





Genomic approaches to unravel the pathogenesis of Chagas disease

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Desiré Casares Marfil, 2021

Programa de Doctorado en Biomedicina

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Genomic approaches to unravel the pathogenesis of Chagas disease

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Programa de Doctorado en Biomedicina.

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A José, A mis padres y mi hermano, A mis amigos y compañeros.

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ABBREVIATIONS

- 1KGP: The 1000 Genomes Project
- AFR: African population
- AMR: Native American population
- CCR5: Chemokine receptor 5
- CI: Confidence interval
- CNV: Copy number variation
- DC: Dendritic cell
- DMP: Differentially methylated position
- DMR: Differentially methylated region
- DTG: Disease tolerance genes
- DTUs : Discrete typing units
- ELISA: Enzyme-linked immunosorbent assay
- eQTL: Expression-quantitative trait loci
- EUR: European population
- FDR: False discovery rate
- GRS: Genomic risk score
- GWAS: Genome-wide association studies

HLA: Human leukocyte antigen

HP: Haptoglobin

HWE: Hardy-Weinberg equilibrium

IBD: Identity by descent

IFN: interferon

IL: Interleukins

LD: Linkage disequilibrium

MAF: Minor allele frequency

MHC: Major histocompatibility complex

MIF: Migration inhibitory factor

mQTL: Methylation quantitative trait loci

NFAT: Transcriptional nuclear factor of activated T cells

NK: Natural killer cells

NO : Nitric oxide

OR: Odds ratios

PCA: Principal component analysis

PCR: Polymerase chain reaction

PEL: Peruvian from Lima population

PheWAS: Phenome-Wide Association Studies

PI_HAT: IBD proportion

PRG: Pathogen resistance genes

QC: Quality control

RNA-seq: RNA-sequencing

SNPs: Single nucleotide polymorphism

TF: Transcription factor

TIRAP: Toll-interleukin-1 receptor domain containing adapter protein

TLRs: Toll-like receptors

TRIM: Tripartite Motif family

TSS: Transcription start site

WHO : World Health Organization

GENES

APOL1: Apolipoprotein L1

ATP2B4: ATPase Plasma Membrane Ca2+ Transporting 4

BAFT2: Basic Leucine Zipper ATF-Like Transcription Factor 2

CCDC88B: coiled-coil domain containing 88B

CCL2: C-C Motif Chemokine Ligand 2

CCL3: C-C Motif Chemokine Ligand 3

CCL4: C-C Motif Chemokine Ligand 4

CCL5: C-C Motif Chemokine Ligand 5

CCR2: C-C Motif Chemokine Receptor 2

CCR5: C-C Motif Chemokine Receptor 5

CD247: Cluster of Differentiation 247

CD40L: CD40 Ligand

CR1: Complement C3b/C4b Receptor 1

CXCL10: C-X-C Motif Chemokine Ligand 10

CXCL9: C-X-C Motif Chemokine Ligand 9

DUSP22: Dual Specificity Phosphatase 22

EBF2: EBF Transcription Factor 2

FERMT3: FERM Domain Containing Kindlin 3

FL1: Follicular Lymphoma, Susceptibility To, 1

HLA: Human leukocyte antigen

IFN-γ: Interferon gamma

IL-10: Interleukin 10

IL-12: Interleukin 12

IL-15: Interleukin 15

IL-17: Interleukin 17

IL-17A: Interleukin 17A

IL-18: Interleukin 18

IL-2: Interleukin 2

IL-35: Interleukin 35

IL-6: Interleukin 6

IL-7: Interleukin 7

KIR3DS1: Killer Cell Immunoglobulin Like Receptor, Three Ig Domains And Short Cytoplasmic Tail 1

KLF4: Kruppel Like Factor 4

KLF4: Kruppel-like factor 4

LAMP3: Lysosomal Associated Membrane Protein 3

LTA: Lymphotoxin Alpha

MBL2: Mannose-binding lectin 2

NAALADL1: N-Acetylated Alpha-Linked Acidic Dipeptidase Like 1

NLRP3: NLR pyrin domain-containing 3

PLA2G16: Alias for PLAAT3 gene

PLAAT3: Phospholipase A and acyltransferase 3

PLCB3: Phospholipase C Beta 3

POLA2: DNA polymerase alpha 2, accessory subunit

RPP21: Ribonuclease P/MRP Subunit P21

SAC3D1: SAC3 domain-containing protein 1

SHD1: Alias for SAC3D1 gene

SIK1: Salt Inducible Kinase 1

SLC11A1: Solute Carrier Family 11 Member 1

SNX15: Sorting Nexin 15

STAT5: Signal transducer and activator of transcription 5

STX7: Syntaxin 7

TNF- α : Tumor necrosis factor

TRIM10: Tripartite Motif Containing 10

TRIM26: Tripartite Motif Containing 26

TRIM31: Tripartite Motif Containing 31

TRIM39: Tripartite Motif Containing 39

TRIM40: Tripartite Motif Containing 40

VTRNA2-1: Vault RNA 2-1

RESUMEN

La enfermedad de Chagas es una enfermedad infecciosa causada por el parásito *Trypanosoma cruzi*. Esta enfermedad es endémica de Latinoamerica donde el principal vector de transmisión son insectos hematófagos. Se estima que afecta a unas 6-7 millones de personas en todo el mundo ya que, debido a procesos como la migración y globalización, se ha extendido a zonas no endémicas. La enfermedad cursa con una fase aguda seguida de una fase indeterminada en la que la mayoría de individuos permanecen asintomáticos de por vida. Sin embargo, algunos pacientes pueden desarrollar la forma crónica de la enfermedad años después de la infección, manifestándose con sintomatología digestiva y/o cardíaca, siendo esta última la más frecuente y grave conocida como cardiomiopatía chagásica crónica.

Dada las diferencias en el desarrollo de la infección en zonas endémicas así como la evolución diferencial a la forma crónica cardíaca, se ha sugerido la influencia de la variación genética del hospedador en la patogénesis de la enfermedad. La presente tesis doctoral se centra en el estudio del componente genético del hospedador mediante diferentes estrategias de análisis genético, para dilucidar su papel tanto en la infección como en el desarrollo de la cardiomiopatía chagásica crónica.

Para ello y con el objetivo de confirmar asociaciones previamente descritas, se realizaron estudios de genes candidatos para evaluar la relación de variantes genéticas localizadas en los genes *IL6*, *IL17A* e *IL18*, que codifican citoquinas implicadas en la respuesta inmune contra *Trypanosoma cruzi*. Dichas asociaciones se confirmaron para los genes *IL17A* e *IL18*, mientras que la relación de la variante del gen *IL6* fue descartada.

Sin embargo, la evaluación de variantes genéticas a lo largo de todo el genoma mejora la capacidad de identificación de variación asociada con la enfermedad de una manera no sesgada, para lo cual un estudio de asociación del genoma completo (GWAS, del inglés "genome-wide association studies") supuso la herramienta de elección. Así, se realizó un GWAS en individuos procedentes de Colombia, Bolivia y Argentina que, junto con la información de población brasileña previamente publicada, se meta-analizó tanto para la susceptibilidad a la infección como para el desarrollo de la cardiomiopatía chagásica crónica. Este análisis nos permitió identificar una asociación a nivel de significación genómica con la forma cardíaca cercana al gen *SAC3D1*, siendo la primera asociación genómica identificada en esta enfermedad.

Por otro lado, dada la mezcla de las poblaciones ancestrales Nativo Americana, Europea y Africana presente en las poblaciones latinas, se evaluó la relación de estos bloques de ascendencia con la susceptibilidad a la enfermedad de Chagas. Para esto se realizó un análisis de mapeo por mezcla en la cohorte colombiana. Se identificó una asociación de protección de la ascendencia Nativo Americana con la susceptibilidad a la infección en la región del complejo mayor de histocompatibilidad, en contraposición de la ascendencia Europea que resultó de riesgo, poniendo de manifiesto el componente evolutivo en la relación parásitohospedador de estas poblaciones. En esta región genética se localizan numerosos genes implicados en la respuesta inmunológica, entre los que cabría destacar los genes *HLA*, que tienen una gran transcendencia en el reconocimiento antigénico, y algunos miembros de la familia de los *TRIM*, conocidos por su papel en la respuesta inmune frente a patógenos.

Finalmente, se llevó a cabo un análisis de rasgos cuantitativos de metilación (mQTLs, del inglés "methylation quantitative trait *loci*") en la región genómica identificada en el GWAS. Con esta estrategia se analizó

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el efecto de la variación genética en los patrones de metilación de un subconjunto de pacientes procedentes de Bolivia. Entre los resultados más importantes destacamos aquellos mQTLs en genes que se han relacionado previamente con caracteres hematológicos V cardiovasculares. Además, identificamos mQTLs del gen CCDC88B, que tiene un papel relevante en el proceso inflamatorio. Este gen había sido previamente identificado como diferencialmente metilado y expresado en tejido cardiaco de pacientes con cardiomiopatía chagásica crónica, lo cual pone de manifiesto su importancia en la patogénesis de la enfermedad. Nuestros resultados confirmaron la influencia de los genes de esta región, destacando la funcionalidad de los mismos más allá de la variación en la secuencia del ADN. Adicionalmente, la identificación de genes previamente identificados en tejido cardiaco, pone de manifiesto la importancia del uso de tejidos más accesibles en la identificación de biomarcadores.

Los resultados presentados en esta tesis doctoral destacaron el papel de la genética del hospedador en el desarrollo de la enfermedad de Chagas. Esto, junto con los análisis de metilación y expresión global que se están realizando en estas muestras, representan un avance en el conocimiento de la patogénesis de esta enfermedad tropical desatendida. Genomic approaches to unravel the pathogenesis of Chagas disease

ABSTRACT

Chagas disease is an infection caused by the parasite *Trypanosoma cruzi*. This disease is endemic form Latin America, where the main transmission vectors are hematophagous bugs. It is estimated that the infection affects 6-7 million people around the world, where it has been extended to non-endemic areas given migratory movements and globalization. Chagas disease has an acute phase followed by an indeterminate phase, and the majority of individuals remain asymptomatic for life. Nevertheless, some patients could develop the chronic form of the disease even years after the infection, which manifests with digestive and/or cardiac symptomatology. The latter the most frequent and severe form of the disease is known as chronic Chagas cardiomyopathy.

Given the inter-individual differences in the infection in endemic regions as well as the differential development of the chronic cardiac form, it has been suggested the influence of the host genetic component in the pathogenesis of the disease. The present doctoral thesis focused in the analysis of the host genetic component through different genetic analysis approaches, in order to elucidate its role in the infection and the progress to the chronic Chagas cardiomyopathy.

In order to confirm previously described associations, candidate gene studies were performed to assess genetic variants located in the *IL6*, *IL17A* and *IL18* genes, which encode cytokines with relevance in the immune response against *Trypanosoma cruzi*. The associations were confirmed for the *IL17A* and *IL18* variants, while it was discarded for the *IL6* gene variant.

The assessment of genetic variation throughout the entire genome improves the capacity of identifying disease associations in an unbiased fashion; therefore a genome-wide association study (GWAS) was the strategy of choice. For this, Colombian, Bolivian and Argentinian individuals' genomic data was meta-analyzed with previously published data from the Brazilian population. This analysis was carried out for the susceptibility to the infection and the development of chronic Chagas cardiomyopathy. Using this strategy, it was possible to identify an association at genomic significance level with the chronic cardiac form in the vicinity of the *SAC3D1* gene, being this the first genomic association identified for the disease.

On the other hand, given the admixture of Native American, European and African ancestries of the Latin American populations, the relation among ancestry genetic blocks and the susceptibility to Chagas disease was assessed. Thus, an admixture mapping analysis in the Colombian population was performed. This approach revealed a protective association of the Native American ancestry with the susceptibility to the infection located in the major histocompatibility complex region. In contrast, the European ancestry showed a risk association with the disease, highlighting the evolutionary component in the parasite-host relationship of these populations. The HLA genes and the TRIM gene family were highlighted in this locus, which are of great importance for antigenic recognition and their contribution in the immune response against pathogens.

Finally, a methylation quantitative trait *loci* (mQTLs) analysis was performed in the significant genomic region from the GWAS. This strategy allowed us to evaluate the effect of the genetic variation in the methylation patterns in a subset of individuals from Bolivia. Significant

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mQTLs were identified, including the *CCDC88B*, which is an important player in the inflammatory process. Interestingly, a previous analysis in chronic Chagas cardiomyopathy patients revealed this gene as differentially methylated and expressed. Our results highlighted the functionality of the genes within this region beyond the variation in the DNA sequence. In addition, the exposure of genes previously reported in cardiac samples, highlighted the importance of using more accessible tissues in the identification of disease biomarkers.

The results presented in this doctoral thesis emphasized the role of the host genetics in the development of Chagas disease. This, together with the global methylation and expression analyses being carried out, represents an advance in the knowledge of the pathogenesis of this neglected tropical disease. Genomic approaches to unravel the pathogenesis of Chagas disease

INTRODUCTION

INTRODUCTION

Genomic approaches to unravel the pathogenesis of Chagas disease

1. Chagas disease: a neglected tropical disease

Chagas disease, also known as American trypanosomiasis, is an infection caused by the parasitic protozoan *Trypanosoma cruzi*, whose main transmission vector are bugs from the Reduviidae family (1). The disease is endemic in 21 Latin American countries and, according to the last World Health Organization (WHO) report, it is estimated that affects 6-7 million people around the world (https://www.who.int/news-room/fact-sheets/detail/chagas-disease-(american-trypanosomiasis)). Although the infection was restricted to rural areas from Latin America at the beginning, the migratory movements have originated the globalization of the disease, extending it to urban areas and other regions such as USA or Europe (Figure 1) (2).



Figure 1. Regions with cases of Chagas disease in endemic and non-endemic areas (Modified from Sangenito et al, Trop Med Infect Dis. 2020).

The infection has been dated in the pre-Columbian period thanks to the discovery of *T. cruzi* DNA in mummies from Chile/Peru and Brazil as far back as 7,500 BC (3). However, it was not until 1909 that the disease, the ethological agent and the bug vector that transmit it were described for the first time by the Brazilian physician Carlos Ribeiro Justiniano Chagas (1879–1934) in a 2 years-old patient with presence of the parasite in blood (4). Thus, it is believed that the great variability of these insect vectors, as well as the animal reservoirs, have contributed to the persistence of the infection for almost 9,000 years (5).

Over time, other transmission ways of the parasite have diversified, being even more common than the Triatomine bugs in certain endemic regions (6). The transmission by blood transfusion started to be less common in endemic areas after its detection, however it might suppose a challenge in non-endemic regions where blood donations are not screen tested for the presence of the parasite (7). Oral transmission has also been described in Chagas disease, mainly through wild animal meat and the consumption of food contaminated with the feces of the insect vector (6, 8). Moreover, this transmission mechanism has been described to have higher mortality rate in comparison to the vector transmission (8).

After the entry of the parasite, Chagas disease starts with an acute phase, which can be followed by an indeterminate phase that in most cases uses to be asymptomatic (9). However, some Chagas patients develop chronic symptoms that compromise internal organs such as heart, esophagus and colon, comprising this symptomatology the chronic cardiac and digestive forms of the disease, respectively (https://www.cdc.gov/parasites/chagas/disease.html). In spite of the variety of symptoms that Chagas patients can manifest, a high number of them remain undiagnosed and, in combination with the high transmission

rate, it is estimated that 75 million people are in risk of infection (https://www.who.int/health-topics/chagas-disease#tab=tab_1). This produces patient's disability and long-term treatments in the case of the chronic forms, which make Chagas disease one of the most costly neglected tropical diseases (1). In addition, the high mobility emphasizes the economic burden of Chagas disease in non-endemic regions where it is estimated that a 10% of its total burden affects these regions (1, 9, 10).

2. Trypanosoma cruzi life cycle

The life cycle of *T. cruzi* takes place in the main transmission vector, the Triatomine bugs, and in the animal host (11). During this complex cycle, the parasite goes through four different stages: blood-form trypomastigotes, intracellular amastigotes, epimastigotes and metacyclic trypomastigotes (Figure 2) (11). The cycle starts when the insect vector ingests the bloodform trypomastigotes from an infected host through its bite. Once in the mammalian host, the parasite starts division cycles by binary fission in the vector's gut, being this the epimastigote parasitic form (12). Next, the parasite is differentiated into metacyclic trypomastigotes, which migrates to the hindgut of the vector around three weeks later. Normally, the transmission to the mammalian host occurs when the Triatomine vector bites and deposits its feces in the host's skin, which favors the entry of the parasite through the wound (12). Metacyclic trypomastigotes will penetrate into the host' cells to turn into the amastigotes stage, which uses binary fission to multiply and infect the tissue (11). At this point, the host will suffer the acute phase of the disease while the number of amastigotes increases and infect other tissues, turning into the trypomastigotes stage (9, 12). The last ones will reside in blood where they can be transmitted to another bug, closing the life cycle of the parasite (11).

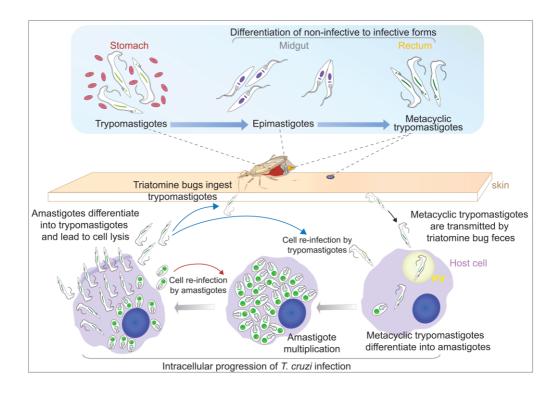


Figure 2. *Trypanosoma cruzi* double-host cycle (Modified from Moretti et al, Trends Parasitol. 2020).

Six phylogenetic lineages or discrete typing units (DTUs) have been described for *T. cruzi* named TcI to TcVI and classified according its genetics, geographical distribution, pathogenesis and response to treatment (9, 13). In addition, different DTUs have been associated with different symptomatology of the disease, for instance the TcII has been related with cardiac complications of the chronic phase while the TcI has been mostly identified in individuals with acute manifestations of the disease (14).

During this cycle, there are several metabolic processes that favor the parasite replication and might affect the pathogenesis of the disease (12).

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One of these processes is the variation of ionic concentrations that affect to the different parasitic stages (12). The calcium (Ca²⁺) and potassium (K⁺) have been described to play an important role in the proliferation and the tissue invasion of the parasite in the human host through the invagination of the cellular membrane (11, 12, 15, 16). Thus, the regulation of these molecules by ionic channels allows to the parasite the adaptation to changing conditions in the vector and the mammalian host (17). In fact, parasite's calcium homeostasis has been proposed as a possible therapeutic target for possible treatments of the disease (18). Another molecule is the nitric oxide (NO), which is involve in the control of parasitic levels through toxic levels and the reduction of growth factors (19).

3. Pathogenesis of Chagas disease

During the acute phase, the first contact with the parasite triggers host immune response that might develop nonspecific clinical symptoms such as fever and inflammation (20). Nevertheless, some of these symptoms can help to the detection of the infection in endemic regions, for example when the parasite entry through the conjunctiva and causes eye inflammation, which is known as the Romaña sign (21). This phase uses to appear a few weeks after the infection with the parasite and last for 4-8 weeks (9). About 1-5% of patients may develop a severe acute phase characterized by myocarditis and/or meningoencephalitis, which can lead to death (4, 9). During the acute phase the screening of seropositive patients is very important for an early administration of treatment in order to palliate symptoms and prevent the chronic phase (22). Thus, detection methods are based on the identification of the parasite and change according to the stage of the disease (23). Therefore, during the acute phase, methods based on concentration or PCR (polymerase chain reaction) techniques predominate, while in the chronic phase serological methods such as ELISA (Enzymelinked immunosorbent assay) are used for screening (23). This supposes key in the control of the disease as most patients remain asymptomatic during the acute phase and for the rest of their lives, being this known as the indeterminate phase (9). Nevertheless, around 30-40% of infected patients develop a chronic phase with organ involvement that can appear even years after contact with the parasite (4). This chronic phase can produce cardiomyopathy, also known as chronic Chagas cardiomyopathy, or involve the digestive system and produce megaoesophagus and megacolon (4, 24).

The chronic Chagas cardiomyopathy is the most severe form of the disease, being the sudden death the main cause of death at this stage of the disease (4, 24-26). This chronic phase affects mainly the myocardium and the conduction system, although it can also affect the circulatory system causing other symptoms such as thrombus formation (27). According to the severity of symptoms, there are different classifications for the chronic cardiac form. These are based on the presence of electrocardiogram abnormalities and its severity (28). Some of these classifications are the Kuschnir and that according to the American Heart Association (28-30). In general, these classify patients into 4 groups based on the results of the echocardiogram and electrocardiogram, as well as physiological abnormalities in the tissue or organ (28, 31).

The chronic digestive form of Chagas disease is less frequent among chronic patients although it can also appear in conjunction with the chronic cardiac form (32). Symptoms of this condition range from dysphagia and motility problems of the digestive track to more severe forms such as megacolon and megaesophagus (33). The digestive affection is mainly reported in regions from Brazil being also related with specific parasitic DTUs, although it has become more common in non-endemic regions (32).

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In this sense, several studies have related the geographical distribution of *T. cruzi* DTUs with the different clinical symptoms prevalence in different regions (34). Thus, the TcI and TcIV strains have been mainly detected among acute cases of Chagas disease while the TcII presented highly prevalence in different cardiac and digestive affections in patients from Brazil (13).

The variety of parasitic strains has also been shown to affect the efficacy of the main treatments currently used, benznidazole and nifurtimox, due to their different resistance (35). Benznidazole and nifurtimox are antiparasitic drugs with better results observed in young patients and during the early stages of the disease, since in the advanced chronic phase and patients of advanced age the side effects are increased (9, 35). For this reason, and especially for the acute phase of the disease, the development of new, less invasive and individualized treatments is necessary. In this sense, precision medicine allows the identification of new markers and therapeutic targets of disease progression useful for the development of new drugs (36).

3.1. Host immune response

The parasitic control and disease progression during the *T. cruzi* infection depend on the innate and adaptive immune responses (Figure 3), as well as of several external factors , such as genetic parasitic variation, the transmission way, neuroendocrine factors or the host's genetic background (37, 38). In the early infection the high parasitic levels requires a strong inflammatory response in order to control the parasite distribution to other tissues (9). Thus, in the acute phase mainly highlight the role of macrophages, natural killer cells (NK), lymphocytes and cytokines production (38).

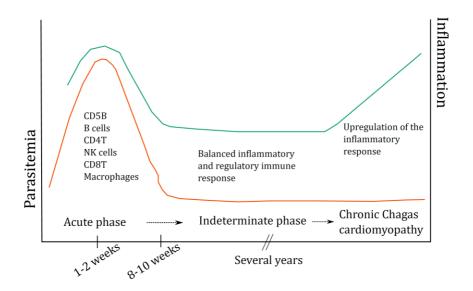


Figure 3. During Chagas disease stages there are different balances between the regulatory and inflammatory immune response as parasite levels in blood or tissues change (parasitemia). Thus, after the entry of the parasite, immune cells are produced in order to control the infection. This originated a balance in the immune system that keeps most individuals asymptomatic for life. However, in 30% of

patients there is an imbalance in the inflammatory state, characterized by the production of proinflammatory cytokines, which affects the cardiomyocytes causing the chronic Chagas cardiomyopathy (Modified from Machado et al, Semin Immunopathol. 2012).

In the innate immunity, the first line of defense is composed of monocytes/macrophages, neutrophils and dendritic cells due to their disposition in the epithelial region (39). In the acute phase, monocytes and neutrophils express a series of receptors specialized in antigen recognition, the most prominent of which are the Toll-like receptors (TLRs) (39). Several T. cruzi molecules act as ligand to these receptors, thus, it has been reported that *T. cruzi* DNA and RNA act as activators for the TLR9 and 7, respectively (40). In addition, during the innate immune response, the release of cytokines seems to be important both in the initial infection and in the progression of the disease (41). Cytokines such as interleukins 6 and 12 (IL-6 and IL-12), interferon gamma (IFN- γ) and the tumor necrosis factor (TNF- α), among others, are the first to be released and responsible for the activation of other cells of the inflammatory response (36, 42). One of these cells activated by cytokines are the NK cells that have a crucial role during intracellular infections by controlling parasitic levels during early stages of Chagas disease (39). Although these cells exert their main function by releasing IFN- γ , they can also fight infection by eliminating extracellular parasites in an IL-12 dependent mechanism, or by influencing the maturation of other cells such as dendritic cells (DCs) (39). DCs are also part of the first line of defense against the parasite playing an important role during the acute phase (43). In this sense, several studies propose these cells as a treatment therapy due to their ability to recognize and internalize antigens during infection, specifically tolerogenic dendritic cells, which have an inhibitory effect in the progression of chronic Chagas cardiomyopathy (43, 44).

During the adaptive immune response, B and T cells play a fundamental role, since antigen recognition and antibody production are crucial processes for the control of the parasite in the organism and to maintain the survival rate (45). The B cells are the main producers of antibodies against *T*. cruzi, providing a bridge between innate and adaptive immune responses (39, 46). Numerous studies have shown that the antibody production is crucial in the course of Chagas disease during the acute and chronic phases (45, 47). Nevertheless, the antibodies produced are not fully efficient eliminating the parasite (39). In this sense, a continued patients' exposure to the parasite, although in low levels, might produces an inflammatory response during the chronic phase that promotes cardiac damage (24). In addition to the production of antibodies, different types of B cells produce a variety of cytokines that stimulates the production of T cells. For example, regulatory B cells produce IL-10 and IL-35, which stimulate the formation of T cell clones as well as promote regulatory T cells (48). Thus, the T cells play an important role in the adaptive immune response against *T. cruzi*, as they are the main responsible for the antigen recognition from the antigen presenting cells during the acute phase (39). In the case of the chronic phase, these cells also produce different types of cytokines that helps to control parasitic levels and the damage level (49).

The immune response during the first stage of the disease is crucial to prevent parasite proliferation in tissues and the development of the chronic phase (4). Thus, it has been shown that chronic Chagas patients have higher levels of T cells with immunoregulatory activity (49). However, the expression of TNF- α and IFN- γ cytokines by these cells has been associated with the development of a more severe chronic cardiac form probably as a consequence of a continuous exposure to the parasite, which has been proposed as a promoter of cardiac damage in Chagas patients (24, 49). This is the premise of the autoimmunity in Chagas disease hypothesis, which is

proposed due to the presence of autoantibodies in the course of the disease (50). Different autoantibodies, such as targets of β 1-adrenergic receptors, have been related with the chronic cardiac form of Chagas disease supporting the idea of anti-self-response that has been also related with parasitic persistence (51). Thus, the fact that better explain this hypothesis is the existence of a molecular mimicry between parasitic and host molecules (50, 52). Nevertheless, even though the existence of autoantibodies has been demonstrated in Chagas disease and patients with the chronic cardiac form, their implication in the development of the disease needs to be clarified (51, 52).

4. The study of the genetic component of diseases.

Genetic studies are based on the identification of association among a specific trait or disease with genetic variation. This variation in the DNA sequence goes from copy number variations (CNVs) or microsatellites to single nucleotide polymorphisms (SNPs), being the last one the most common markers under study (53).

The SNPs are variations in a single position of the DNA sequence with a minor allele frequency (MAF) of at least 1% in the population under study. It is estimated that there are around 38 million SNPs across the genome and, although the majority are in non-coding regions, they compound around 90% of the human genetic variation (54). Genetic analyses allow to evaluate the relation of this genetic variation with a trait or disease given us an idea of the genetic predisposition of a group of individuals to develop it (https://www.nature.com/scitable/definition/snp-295/). In this sense, allelic frequencies of one or more SNPs are compared among cases (individuals affected with the disease) and controls (unaffected individuals)

using different statistical methods. Thus, if they are statistically significant under different significance thresholds, they are considered in association with the disease (55).

The first significant event in genomics research was the sequencing of the human genome, driven by the Human Genome Project in 1990 (56). This announcement would accelerate biomedical research, allowing the determination of the genetic contribution to human traits (57). From this, candidate gene studies emerged as a rapid and cheap first approach by which the genetic relatedness of genes to a trait or disease is assessed by selecting SNPs located in their gene region (58). These types of studies are performed under a previous association hypothesis mainly based on the functional implication of the gene (59).

The determination of common patterns of the DNA sequence variation in different populations by the International HapMap Project would contribute to the genomic development with the determination of the linkage disequilibrium (LD) structure, haplotypes and recombination hotspots in several populations from the different ancestral geographic locations (60). With this knowledge, the first genotyping arrays with considerable coverage of common human genetic variation were designed. Thus, from 2005 these arrays began to be used in genome-wide association studies (GWAS), whose aim is the understanding of the genetic component of a disease or trait, as well as candidate gene studies (61). Nevertheless, GWAS allow the analysis of millions of variants distributed throughout the genome being this its main advantage (62). In addition, they are hypothesis-free studies, which allow working without any prior assumptions (63). The methodology commonly applied in GWAS is based on the genotyping of genomic common SNPs in collected individuals with phenotypic data and clinical information for a given trait or disease; after the corresponding quality controls, the genotypic

data can be phased by the imputation process, assessing a larger number of SNPs and increasing the coverage; later, the association test is assess and the associated signals will be those that pass the genome-wide significance threshold, set at *P-value*< $5x10^{-08}$ that corresponds to the Bonferroni correction for 1,000,000 independent SNPs (Figure 4) (63, 64). The results obtained are usually tested in independent cohorts in a replication process, as well as *in silico* analyses are carried out on available sources to check the relevance of the results (63) (Figure 4).

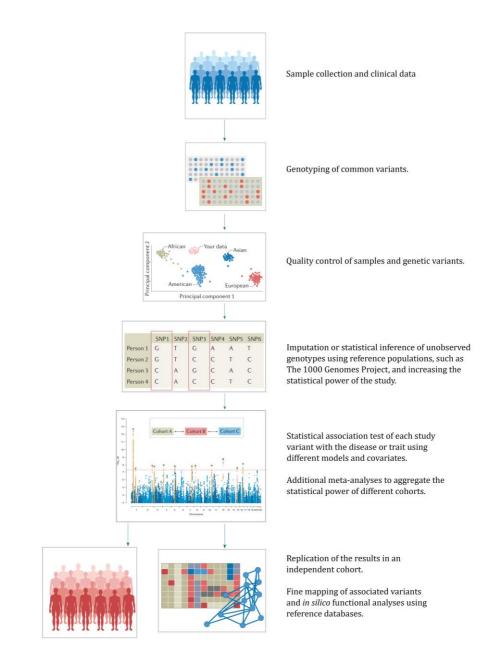


Figure 4. Main steps for genome-wide association (Modified from Uffelmann E, Nat Rev Methods Primers. 2021).

The launch of the 1000 Genomes Project (1KGP) aimed to characterize the human genome and cover low-frequent variants as well as variation with different frequency among populations. This project included 14 representative populations from Europe, East Asia, sub-Saharan Africa and the Americas, supposing the last one an advance in the representation of Latin American populations in reference data (54, 65). In this sense, the inclusion of populations in the genetic studies with mixed ancestry supposed a challenge, as is the case of Latin American populations that have a recently admixture with Native American, European and African continental sources (66). In addition, individuals from these populations show inter-individual variation in ancestry proportions, which has been correlated with the risk of other infectious diseases such as malaria (67, 68). This situation allows the application of admixture mapping studies with the aim to assess disease associations with the varying local ancestries in admixed populations, as a GWAS complementary methodology (69, 70).

Nevertheless, the DNA sequence variations are not the only mechanism that can be associated with a specific phenotype. The epigenetic comprises the set of modifications in gene expression without changing the DNA sequence. These changes can take place under the influence of environmental factors or the genetic variation itself (Figure 5) (71). The most studied epigenetic mechanism is the DNA methylation that consists in a heritable modification of the cytosine DNA by the addition of a methyl group, which in most cases uses to inhibit gene expression (72). Given that the majority of genetic variants are located in non-coding regions, functional studies are required in order to link those associations with specific genes or pathways. Thus, the combination of epigenomic and genomic data allows the assessment of biological and functional significance of the variants associated with a disease. In turn, methylation patterns can be influenced by variations in the DNA sequence, which are known as methylation quantitative trait *loci* (mQTL) (72). The analysis of these interactions allows us to give functional significance to the associated SNPs, which is essential for the genetic characterization of the disease. Therefore, genetic and epigenetic variations and their interaction can be translated into expression changes. Transcriptomic analyses are a good strategy to evaluate the consequences of genetic variations and their interactions (73). These analyses together with the online tools and databases available for different populations make it possible to evaluate the genetic component of different type of diseases, as well as to functionally evaluate disease-associated variants for unraveling the biological mechanisms underlying associations.

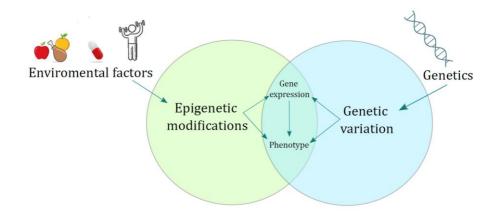


Figure 5. Factors influencing genetic and epigenetic variability and their relationship to gene expression. The dynamic among these modifications can be evaluated in order to assess their relationship with the development or the susceptibility to the disease. (Modified from Arif M, Hypertens Res. 2019).

5. Genetic analysis of infectious diseases

Regarding the host genetic burden in infectious diseases, it is hypothesized that it contributes to the heterogeneity in the response and the

severity of the infection (74). This hypothesis arises from the idea of a continuous exposure of our genomes to the infectious agents for several generations and that, through the action of co-evolution and natural selection, produces this variability in the susceptibility to these diseases (75). In this sense, the evidence of this contribution has been assessed in several infectious diseases, such as malaria, HIV-1 or tuberculosis, mainly through GWAS (74).

One of the most studied parasitic diseases is Malaria. This infections is caused by different *Plasmodium* species and several candidate gene studies and GWAS have reported genetic associations with the development of the disease and its severe forms in several African populations (76-78). Therefore, strong associations have been described for the ABO, CD40L and ATP2B4 genes with the disease (79). In the same way, the Leishmaniasis is a parasitic infection caused by species from the genus Lesinmania where the large number of molecular studies carried out in mouse models has allowed the identification of important genes and pathways in the disease (79). Based on the last point, most of the genetic studies carried out in this disease have been candidate gene studies, highlighting the SLC11A1 and FL1 genes and their association with different forms of the disease (80-82). Recently, a GWAS in cutaneous leishmaniasis has been published in which some suggestive associations in the LAMP3 and STX7 genes were identified, being these genes previously related to the intracellular localization of the parasite (83). Regarding other trypanosomiasis, the African trypanosomiasis is caused from another member of the T. cruzi family, T. brucei. Thus, candidate gene studies highlighted genes related with the immune response, such as *IL6*, or genes related with the identification of the parasite by the host, such as APOL1 (84).

In this way, the large amount of information available today that confirms the host differential genetic susceptibility to contract different infectious diseases highlights the need to deepen this type of genetic studies in order to translate it to clinical and preventive benefit.

5.1. Previous genetic studies in Chagas disease

Genetic studies are great tools to improve the knowledge about the molecular basis of Chagas disease, opening new avenues for the development of biomarkers of disease progression, new therapeutic strategies to suit the requirements of individual patients, and contributing to the control of one of the infections with the greatest socio-economic impact in the Americas. Thus, several studies have assessed the influence of genetic variation in the susceptibility to Chagas disease and the development of the chronic forms, especially in those genes with implication in the immune response.

The major histocompatibility complex (MHC) region is located on chromosome 6 and encode different sets of proteins that play key roles in the immune system and that are responsible for mediating host-pathogen interactions (85). The most relevant molecules in this context are the human leukocyte antigen (HLA), which are cell-surface glycoproteins that bind peptide fragments that either have been synthesized within the cell (HLA class Ia; A, B, and C) or that have been ingested and processed by the cell (HLAclass II; DR, DQ, and DP). Several HLA alleles and haplotypes have been reported to be associated with the susceptibility to *T. cruzi* infection and to the chronic manifestation of Chagas disease (86, 87) (Table 1).

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GeneAlleles/GenotypingPopulationAssociationHLA class IITNFa -238ABrazilInfection(Pissetti etTNFa -1031C and -308AColombiaCCC(Criado et aTNFa -1031C and -308ABrazilInfection(Pissetti etTNFa -1031C and -308ABrazilInfection and All chronic(Criado et al.TNFa microsatellite and -308BrazilInfection and All chronic(Campelo et al.TNFa wicrosatellite and -308BrazilInfection and All chronic(Campelo et al.TNFa wicrosatellite and -308BrazilInfection and All chronic(Campelo et al.TNFa wicrosatellite and -308BrazilNo association(Drigo et al.TNFa -308BrazilNo association(Drigo et al.TNFa -308BrazilNo association(Alvarado-Lata)TNFa -308BrazilNo association(Drima et al.TNFa -308BrazilNo association(Drima et al.TNFa -308BrazilNo association(Alvarado-Lata)TNFa -308BrazilNo association(Alvarado-Lata)TNFa -308BrazilNo association(Alvarado-Lata)TNFa -308BrazilNo association(Alvarado-Lata)TNFa -208BrazilNo association(Alvarado-Lata)LTALTA +252BrazilNo association(Alvarado-Lata)BAT1BAT1 - 22C/GBrazilCCC(Alvarado-Lata)BAT1BAT1 - 22C/GBrazilCCC(Alvarado-Lata)	Table 1. HLA genes	les			
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BAT1-22C/G Brazil CCC (14Z V281L CCC 0		BAT1 – 22C/G and –348C/T	Brazil	CCC	(Ramasawmy et al., 2006a, 2006b)
V281L Bolivia CCC		BAT1 – 22C/G	Brazil	CCC	(Alvarado-Arnez et al., 2018)
		V281L	Bolivia	CCC	(del Puerto et al., 2013)
Brazil CCC (IKBL-62A/T and -262A/G	Brazil	CCC	(Ramasawmy et al., 2008)

The most studied genes are HLA-B and HLA-DRB1. A study conducted in Peruvian population found that DRB1*14-DQB1*0301 haplotype was associated to protection against *T. cruzi* infection in a highly endemic area (88). In the same way, another study in the Mexican population found that the frequencies of HLA-DR4 and HLAB39 were significantly increased in seropositive individuals when compared to healthy controls, while the frequency of the HLA-B35 increased in patients with cardiomyopathy when compared to healthy controls (89). A study performed in Bolivian population found that the frequencies of HLA-B*14:02 and HLA-DRB1*01 were significantly lower in patients suffering from megacolon as well as in those with electrocardiogram alteration and/or megacolon compared with a group of patients with indeterminate symptoms (90). Furthermore, the HLA-B*3505 allele was associated with moderate to severe cutaneous reaction in response to Benznidazole (91). Another less polymorphic HLA-G, MICA and MICB have also been studied (90, 92, 93), where the HLA-G gene exhibited distinct patterns of associations with cardiac and digestive chronic Chagas forms in Brazilian population (93, 94). The variability of HLA alleles associated with Chagas disease in the mentioned studies may reflect the genetic heterogeneity presented among the different population in Latin America.

Within the MHC class III region, the *TNF* and *LTA* genes are the most widely investigated regarding the susceptibility to the infection and the chronic cardiac form. Thus, some studies demonstrated that the *TNF* gene polymorphisms could modify the genetic risk of developing heart disease (95, 96). Nevertheless, although genetic associations were reported in Colombian, Mexican and Brazilian populations (97-100), other authors did not find a significant association between the *TNF* gene variants with the disease (101-103).

Cytokines are mediators of the immune response and help to modulate the progression of the disease by inhibiting parasitic replication in different cell types (101). Most candidate gene studies carried out in Chagas disease have been focused on these molecules and were replicated in different Latin American populations (Table 2). *IL-17*, *IL-18* and *IFN-y* are critical molecules for host defense against a variety of intracellular pathogens. Several studies shown that polymorphisms located in the genes encoding for these cytokines were associated with the susceptibility to *T. cruzi* infection and the chronic Chagas cardiomyopathy, mainly in Colombian and Brazilian populations (101, 104-107). In the case of *IL-18* and *IFNG-*, polymorphisms in these genes have been associated with the risk of infection but not with the development of the chronic cardiac form in Colombian patients (104, 105, 108), while in the case of the Brazilian population associations with the chronic form were reported (101, 109). The authors explained that these discrepancies may be the result of a difference in genotypic distributions between Brazilian and Colombian populations.

Cytokines, chemokines and their receptors. Gene Chr Gene name Gytokines and their receptors Gytokines and their receptors L1A 2 Interleukin 1 alpha Colo L-1B 2 Interleukin 1 beta Mexi L-1B 2 Interleukin 1 receptor antagonist Colo L-1B 2 Interleukin 1 receptor antagonist Colo L-1R 5 Interleukin 4 Colo L-4 5 Interleukin 4 Colo L-4R 16 Interleukin 4 Colo L-4R 16 Interleukin 10 Braz L-4R 1 Interleukin 10 Braz L-4R 5 Interleukin 10 Braz L-10 1 Interleukin 12B Colo L-12B 5 Interleukin 12B Colo L-13B 11 Interleukin 12B Colo L-18 11 Interleukin 12B Colo L-18 11 Interleukin 12B Colo L-18 11 Interleukin 17A Colo L-18 11						
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 ines and their receptors 2 Interleukin 1 alpha 2 Interleukin 1 beta 2 Interleukin 1 beta 5 Interleukin 4 16 Interleukin 4 receptor 7 Interleukin 6 1 Interleukin 10 8 5 Interleukin 10 8 5 Interleukin 11 8 6 Interleukin 128 4 6 Interleukin 18 11 Interleukin 18 12 Interfeukin 18 		Chr	Gene name	Population	Association	Reference
 2 Interleukin 1 alpha 2 Interleukin 1 beta 2 Interleukin 1 beta 5 Interleukin 1 receptor antagonist 5 Interleukin 4 receptor 7 Interleukin 6 1 Interleukin 10 8 5 Interleukin 10 8 6 Interleukin 128 4 6 Interleukin 13 11 Interleukin 18 12 Interfeukin 18 12 Interfeukin 18 	cines and their n	eceptors				
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N 2 Interleukin 1 receptor antagonist 5 Interleukin 1 receptor 16 Interleukin 4 receptor 7 Interleukin 6 1 Interleukin 10 8 5 5 Interleukin 10 11 Interleukin 12B 12 Interleukin 18 12 Interleukin 18	د د	Interlaid	in 1 hota	Colombia	CCC	(Florez et al., 2006)
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7 Interleukin 6 1 Interleukin 10 8 5 Interleukin 12B 4 6 Interleukin 17A 11 Interleukin 18 12 Interferon gamma			n 4 receptor	Colombia	CCC	(Florez et al., 2011)
1 Interleukin 10 E B 5 Interleukin 12B E 4 6 Interleukin 17A C C 11 Interleukin 18 12 E E 12 Interleukin 18 E E E	7	Interleuki		Colombia/Peru	No association	(Torres et al., 2010a)
1 Interleukin 10 0 B 5 Interleukin 12B E 4 6 Interleukin 17A 0 11 Interleukin 18 12 12 Interferon gamma 1				Brazil	CCC	(Costa et al., 2009)
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A 6 Interleukin 17A 0 11 Interleukin 18 1 12 Interferon gamma 1	3B 5	Interleuk	in 12B	Colombia	CCC	(Zafra et al., 2007)
11 Interleukin 18 E 12 Interferon gamma	7 A 6	Interleuk	in 17A	Colombia	Infection and CCC	(Leon Rodriguez et al., 2015)
12 Interferon gamma	~	1 լոետեսի	10	Colombia	Infection	(Leon Rodriguez et al., 2016)
12 Interferon gamma				Brazil	CCC	(Nogueira et al., 2015)
12 IIICIICI 011 Ballilla				Colombia	Infection	(Torres et al., 2010b)
				Brazil	CCC	(Alvarado-Arnez et al., 2018)
TCER 10 Transforming growth factor hata Colo			ming growth factor hata	Colombia/Peru	Infection	(Calzada et al., 2009)
				Brazil	No association	(Alvarado-Arnez et al., 2018)

Cytokines, chemokines and their receptors. G Ghemokines and their receptors Chemokines and their receptors G Chemokines and their receptors G Chemokines and their receptors G Chemokines and their receptors C Chemokines G Chemokines G CCR5 3 CCR5 3 C-C motif chemoki	i and their receptors. r Gene name eceptors	Population		
Gene Ch Chemokines and their r C Cores 3	ceptors	Population		
Chemokines and their r CCR5 3	sceptors		Association	Reference
		Peru	CCC	(Calzada et al., 2001a)
		Colombia	CCC	(Florez et al, 2012; Machuca et al, 2014)
		Venezuela	CCC	(Fernandez-Mestre et al, 2004)
	C C motif chomolaino acconton E	Argentina	CCC	(Juiz et al., 2019)
		Brazil	CCC	(Nogueira et al., 2012)
		Brazil	CCC	(Frade et al., 2013)
		Brazil	All chronic forms	(de Oliveira et al, 2015)
		Brazil	No association	(Lima et al., 2018)
		Colombia	CCC	(Florez et al., 2012; Machuca et al., 2014)
CCR2 3	C-C motif chemokine receptor 2	Argentina	CCC	(Juiz et al., 2019)
17	C-C motif chemokine ligand 2	Brazil	CCC	(Frade et al., 2013; Ramasawmy et al., 2009)
4	C-X-C motif chemokine ligand 9	Brazil	CCC	(Nogueira et al., 2012)
CXCL10 4	C-X-C motif chemokine ligand 10	Brazil	CCC	(Nogueira et al., 2012)
22	Macrophage migration inhibitory factor	actor Colombia/Peru	Infection	(Torres et al., 2009)
CCL5 17	C-C motif chemokine ligand 5	Brazil	No association	(Nogueira et al., 2012)
CCL17 16	C-C motif chemokine ligand 17	Brazil	No association	(Nogueira et al., 2012)
CCL19 9	C-C motif chemokine ligand 179	Brazil	No association	(Nogueira et al., 2012)

Chemokines have been also directly related to cardiac effects, involving processes such as heart tissue repair, arrhythmia and cardiac insufficiency (87). The chemokine receptor 5 (CCR5) is a receptor for the chemokines CCL2, CCL3, CCL4 and CCL5 and is one of the most replicated gene. Interestingly, patients with cardiomyopathy exhibited higher expression of this gene, which resulted in increased inflammation (110). In addition, CCR2 belongs to the same chemokine receptor family and has also been associated with chronic Chagas cardiomyopathy after *T. cruzi* infection. Thus, *CCR5*, CCR2 and their haplotypes have been associated with the development of this chronic cardiomyopathy (111, 112). On its part, chronic Chagas cardiomyopathy patients with ventricular dysfunction displayed reduced genotypic frequencies of variants in CXCL9, CXCL10, and increased in CCR5 as compared to those without the dysfunction (113). Additionally, the chemokine macrophage migration inhibitory factor (MIF) is an integral component of the host antimicrobial alarm system whose coding gene has been also associated with the infection by T. cruzi in Colombian and Peruvian populations (114).

Several other genes, selected for their previous association with other infectious diseases, have been studied for Chagas disease, mainly those related with the immune response. These include genes coding for TLRs and related molecules such as Toll-interleukin-1 receptor domain containing adapter protein (TIRAP). This molecule encodes an adapter protein associated with TLRs, which recognizes microbial pathogens. In the context of Chagas disease two studies have reported an association with the susceptibility of developing chronic Chagas cardiomyopathy (Table 3) (115, 116). Others studies have addressed the association between haptoglobin (HP) polymorphisms with *T. cruzi* infection and its chronic forms, including cardiac and digestive forms in Brazilian and Venezuelan population (117,

118). HP is an acute-phase protein synthesized mainly by the liver during inflammatory processes and also possesses anti-inflammatory and antioxidant properties (119). On the other hand, two studies in the mannose-binding lectin 2 (MBL2) showed association with an increased risk of severe cardiac Chagas disease, related with high MBL serum levels in Brazilian population (120), and a moderate association with the risk of infection in Chilean population (121).

Table 3 Othersgenes. Othersgenes. Conc Chronic Concentration Gene Anne Population Association MB2 It Population association MB2 It Population association MB2 It Brazil Infection and CCC MASP2 It Brazil Infection and CCC MASP2 Forum binding lectin 1 Colspa="2">Colspa="2">Population and CCC MASP2 CCC<						
Chr Cene name Population 16 Haptoglobin Venezuela Infecti 10 Mannose binding lectin 2 Venezuela Infecti 11 Toll-interleukin receptor domain containing adaptor protein Brazil CCC 12 Vitamin D3 receptor C Chile Infecti 13 Toll-interleukin receptor domain containing adaptor protein Brazil CCC 14 Toll-interleukin receptor C Brazil CCC 15 Actin apha cardiac muscle 1 Brazil CCC 16 Ficolin 2 Brazil CCC 17 Yotoxic T-lymphocyte associated protein 4 Venezuela No ass 16 Toll like receptor 1 / Toll like receptor 2 Brazil No ass 17 Toll like receptor 5 Toll like receptor 6 Colombia No ass 17 Toll like receptor 6 Colombia No ass Toll like receptor 6 Colombia No ass 13 Toll like receptor 6 Brazil No ass Colombia No ass 14 <th>Table 3</th> <th></th> <th></th> <th></th> <th></th> <th></th>	Table 3					
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16 Haptoglobin Brazil 10 Mannose binding lectin 2 Viramin binding lectin 2 11 Toll-interleukin receptor domain containing adaptor protein Brazil 12 Viramin Di Sreeptor 3 Brazil 13 Actin alpha cardiar muscle 1 Brazil 14 Mannan binding lectin serine peptidase 2 Brazil 15 Ficolin 2 Brazil 16 Ficolin 2 Brazil 17 Cytotoxic T-Jymphocyte associated protein 4 Venezuela 17 Toll like receptor 1 / Toll like receptor 2 Brazil 17 Toll like receptor 4 Brazil 13 Toll like receptor 9 Brazil 14 Toll like receptor 9 Colombia 17 Toll like receptor 9 Brazil 17 Toll like receptor 9 Colombia 13 Toll like receptor 1 Peru 13 Toll like receptor 1 Peru 13 Toll like receptor 1 Peru 14 Toll like receptor 1	Gene	Chr	Gene name	Population	Association	Reference
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11 Toll-interleukin receptor domain containing adaptor protein Brazil 12 Vitamin D3 receptor Oombia 13 Vitamin D3 receptor Brazil 14 Mannan binding lectin serine peptidase 2 Brazil 15 Actin apha cardiac muscle 1 Brazil 16 Froin 2 Brazil 17 Cytotic T-lymphocyte associated protein 4 Venezuela 18 Toll like receptor 1 Toll like receptor 2 17 Toll like receptor 2 Colombia 17 Toll like receptor 3 Brazil 17 Toll like receptor 4 Brazil 17 Toll like receptor 4 Brazil 17 Toll like receptor 7 Colombia 17 Toll like receptor 4 Brazil 17 Toll like receptor 7 Colombia 17 Toll like receptor 7 Colombia 18 Toll like receptor 7 Colombia 17 Toll like receptor 7 Colombia 18 Toll like receptor 6 Toll like receptor 7 17 Toll like receptor 7 Colombia 18 Toll like receptor 7 Colombia 17 Toll like receptor 7 Colombia 18 Toll like receptor 7 </th <th>MBL2</th> <th>10</th> <th>Mannose binding lectin 2</th> <th>Brazil</th> <th>Infection and CCC</th> <th>(Luz et al., 2016)</th>	MBL2	10	Mannose binding lectin 2	Brazil	Infection and CCC	(Luz et al., 2016)
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12 Vitamin D3 receptor C olombia 1 Amman binding lectin serine peptidase 2 Brazil 1 Atin abha cardiar uuscle 1 Brazil 1 Ficolin 2 Brazil 2 Cytotoxic T-lymphocyte associated protein 4 Brazil 4 Toll like receptor 1 / Toll like receptor 2 Brazil 9 Toll like receptor 2 Colombia 9 Toll like receptor 2 Colombia 1/3 Toll like receptor 4 Brazil 1/3 Toll like receptor 7 Colombia 9 Toll like receptor 4 Brazil 1/3 Toll like receptor 7 Colombia 1 Tyrosine-protein phosphatase non-receptor type 22 Colombia 1 Tyrosine-protein phosphatase non-receptor type 22 Colombia 1 Tyrosine-protein phosphatase non-receptor type 22 Colombia 2 Fordhado ab protein 1 Peru	TIRAP	11	Toll-interleukin receptor domain containing adaptor protein	Brazil	CCC	(Frade et al., 2013; Ramasawmy et al., 2009)
1 Mannan binding lectin serine peptidase 2 Brazil 15 Actin alpha cardiac muscle 1 Brazil 16 Ficoln 2 Brazil 2 Cytotoxic T-lymphocyte associated protein 4 Brazil 4 Toll like receptor 1 / Toll like receptor 2 Brazil 9 Toll like receptor 4 Brazil 1 Toll like receptor 4 Brazil 1 Toll like receptor 4 Brazil 1 Toll like receptor 5 Colombia 1 Toll like receptor 6 Colombia 1 Toll like receptor 7 Colombia 1 Toll like receptor 9 Brazil 1 Toll like receptor 6 Colombia 1 Toll like receptor 7 Colombia 2 Hunan natural resistance-associated macrophage protein 1 Peru 1 Troid des printase Colombia 2 Hunan natural resistance-associated macrophage protein 1 Peru 5 Fordhadox protein 1 Peru	VDR	12	Vitamin D3 receptor C	olombia	CCC	(Leon Rodriguez et al., 2016)
15 Actin alpha cardiac muscle 1 Brazil 9 Ficolin 2 Brazil 2 Cytotoxic T-lymphocyte associated protein 4 Brazil 4 Toll like receptor 1 / Toll like receptor 2 Brazil 9 Toll like receptor 2 Brazil 13 Toll like receptor 4 Colombia 13 Toll like receptor 5 Colombia 13 Toll like receptor 9 Brazil 13 Toll like receptor 5 Colombia 13 Toll like receptor 5 Toll like receptor 9 13 Toll like receptor 5 Toll like receptor 9 13 Toll like receptor 5 Toll like receptor 9 14 Toll like receptor 5 Toll like receptor 9 17 Tritic oxide synthase Toll like receptor 1 17 Nitric oxide synthase Colombia 18 Fordward bac protein 1 Peru	MASP2	1	Mannan binding lectin serine peptidase 2	Brazil	CCC	(Boldt et al., 2011)
9 Ficolin 2 Brazil 2 Cytotoxic T-lymphocyte associated protein 4 Brazil 4 Toll like receptor 1 / Toll like receptor 2 Brazil 4 Toll like receptor 2 Colombia 9 Toll like receptor 2 Colombia 1/3 Toll like receptor 4 Brazil 1/3 Toll like receptor 9 Brazil 1/3 Toll like receptor 9 Brazil 1/3 Toll like receptor 5 Colombia 1/3 Toll like receptor 5 Colombia 1/3 Toll like receptor 5 Colombia 1 Toll like receptor 5 Colombia 1 Toll like receptor 5 Toll like receptor 1 1 Toll like receptor 5 Toll like receptor 9 2 Human natural resistance-associated macrophage protein 1 Peru 1 Nitric oxide synthase Colombia 6 Fordwade Apo protein 3 Colombia	ACTC1	15	Actin alpha cardiac muscle 1	Brazil	000	(Frade et al., 2013)
2 Cytotoxic T-lymphocyte associated protein 4 Brazil 0 4 Toll like receptor 1 / Toll like receptor 2 Brazil 0 4 Toll like receptor 2 Colombia P 9 Toll like receptor 4 Brazil 1 13 Toll like receptor 5 Colombia P 13 Toll like receptor 5 Toll like receptor 9 Colombia P 13 Toll like receptor 5 Toll like receptor 9 Colombia P 13 Toll like receptor 5 Toll like receptor 9 Colombia P 13 Toll like receptor 6 Colombia P P 1 Troll like receptor 7 Colombia P 2 Hunan natural resistance-associated macrophage protein 1 Peru 1 Nitric oxide synthase Colombia P 5 Fordwald ab protein 3 Colombia P	FCN2	6	Ficolin 2	Brazil	All chronic forms	(Luz et al., 2013)
 Grouting of the receptor 2 Toll like receptor 1 / Toll like receptor 2 Toll like receptor 3 Toll like receptor 4 Brazil Toll like receptor 5 / Toll like receptor 9 Toll like receptor 6 Toll like receptor 6 Toll like receptor 6 Toll like receptor 6 Toll like receptor 7 Toll like receptor 7 Toll like receptor 6 Toll like receptor 6 Toll like receptor 6 Toll like receptor 6 Toll like receptor 7 Toll like receptor 6 Toll like receptor 6 Toll like receptor 7 Toll like receptor 6 Toll like receptor 7 Toll like receptor 6 Toll like receptor 6 Toll like receptor 7 Toll like receptor 6 Toll like receptor 6 Toll like receptor 7 Toll like receptor 7<th>CTI AA</th><th>ç</th><th>Citto toxic T. Jiminhocuto a concisted nuclein A</th><th>Brazil</th><th>000</th><th>(Dias et al., 2013)</th>	CTI AA	ç	Citto toxic T. Jiminhocuto a concisted nuclein A	Brazil	000	(Dias et al., 2013)
4 Toll like receptor 1 / Toll like receptor 2 Chile N 4 Toll like receptor 3 Brazil N 9 Toll like receptor 5 Colombia N 1/3 Toll like receptor 5 Colombia N 1/3 Toll like receptor 5 Toll like receptor 5 Colombia N 1/3 Toll like receptor 5 Toll like receptor 5 Colombia N 1/3 Toll like receptor 5 Toll like receptor 5 Colombia N 1/3 Toll like receptor 5 Toll like receptor 6 N N N 1/3 Toll like receptor 5 Toll like receptor 6 N <th>10710</th> <th>4</th> <th>cytotoxic 1-1911.pitocyte associated protein 7</th> <th>Venezuela</th> <th>No association</th> <th>(Fernandez-Mestre et al., 2009)</th>	10710	4	cytotoxic 1-1911.pitocyte associated protein 7	Venezuela	No association	(Fernandez-Mestre et al., 2009)
4 Toll like receptor 1 Durinke receptor 2 Colombia 9 Toll like receptor 4 Brazil 1/3 Toll like receptor 5 Colombia 1/3 Toll like receptor 5 Brazil 1/3 Toll like receptor 5 Brazil 1/3 Toll like receptor 5 Brazil 1 Tyrosine-protein phosphatase non-receptor type 22 Colombia/Peru 1 Tyrosine-protein phosphatase non-receptor type 22 Colombia/Peru 1 Nitric oxide synthase Fortua 6 Forthaad bac protein Peru 1 Peru Colombia	TI D1 / TI D2	~	Toll libr recentor 1 / Toll libr recentor 2	Chile	No association	(Weitzel et al., 2012)
4 Toll like receptor 2 Colombia N 9 Toll like receptor 4 Brazil N 71.89 1/3 Toll like receptor 5/Toll like receptor 9 Brazil N 21 1 Toll like receptor 5/Toll like receptor 9 Brazil N 22 1 Tyrosike-protein phosphatase non-receptor type 22 Colombia/Peru 21 1 Tyrosike-protein phosphatase non-receptor type 22 Colombia/Peru 23 1 Mitric oxide synthase Peru 3 6 Forder and cox protein 3 Colombia	7V71 /TV71	۲		Brazil	No association	(Ramasawmy et al., 2009)
9 Toll like receptor 4 Colombia 7LR9 1/3 Toll like receptor 5 / Toll like receptor 9 Brazil 72 1 Tyrosine-protein phosphatase non-receptor type 22 Colombia/Peru 22 1 Tyrosine-protein phosphatase non-receptor type 22 Colombia/Peru 21 2 Human natural resistance-associated macrophage protein 1 Peru 3 6 Fortheradox protein Colombia	TLR2	4	Toll like receptor 2	Colombia	No association	(Zafra et al, 2008)
71.R9 1/3 Toll like receptor 5 / Toll like receptor 9 Brazil Practice 71.R9 1/3 Toll like receptor 5 / Toll like receptor 9 Brazil Practice 22 1 Tyrosine-protein phosphatase non-receptor type 22 Colombia/Peru Pri 2 Human natural resistance-associated macrophage protein 1 Peru 3 6 Fordadox protein 3 Colombia	TIPA	σ	Toll libe recentor 4	Colombia	No association	(Zafra et al., 2008)
1/3 Toll like receptor 5 / Toll like receptor 6 Brazil h 4 Toll like receptor 6 Gilombia/Peru 1 Tyrosine protein flow sphatase non-receptor type 22 Colombia/Peru 2 Human natural resistance-associated macrophage protein 1 Peru 1 Nitric oxide synthase Colombia/Peru 6 Forkhead has protein 3 Colombia				Brazil	No association	(Ramasawmy et al., 2009)
4 Toll like receptor 6 1 Tyrosine-protein phosphatase non-receptor type 22 Colombia/Peru 2 Human natural resistance-associated macrophage protein 1 Peru 17 Nitric oxide synthase Peru 6 Forkhead box protein 3 Colombia	TLR5/TLR9	1/3	Toll like receptor 5 / Toll like receptor 9	Brazil	No association	(Ramasawmy et al., 2009)
1 Tyrosine-protein phosphatase non-receptor type 22 Colombia/Peru 2 Human natural resistance-associated macrophage protein 1 Peru 17 Nitric oxide synthase Peru 6 Porkhead bay protein 3 Colombia	TLR6	4	Toll like receptor 6	Chile	No association	(Weitzel et al., 2012)
2 Human natural resistance-associated macrophage protein 1 Peru 17 Nitric oxide synthase 6 Forkhead box protein 3 Colombia	PTPN22	1	Tyrosine-protein phosphatase non-receptor type 22	Colombia/Peru	No association	(Robledo et al., 2007)
17 Nitric oxide synthase Peru 6 Forkhead box protein 3 Colombia	NRAMP1	2	Human natural resistance-associated macrophage protein 1	Peru	No association	(Calzada et al., 2001a)
6 Forkhead box protein 3 Colombia	NOS2	17	Nitric oxide synthase	Peru	No association	(Calzada et al., 2002)
	FOX03	9	Forkhead box protein 3	Colombia	No association	(Leon Rodriguez et al., 2016)
TYK2 19 Non-receptor tyrosine-protein kinase Colombia No association	TYK2	19	Non-receptor tyrosine-protein kinase	Colombia	No association	(Leon Rodriguez et al., 2018)

Genomic approaches to unravel the pathogenesis of Chagas disease

OBJECTIVES

Despite previous evidences of the role of the host genetic component in Chagas disease and its chronic cardiac form, the knowledge of the genetic architecture and its functional involvement in the disease is limited. Therefore, the present thesis has the overall objective of determining the genetic basis and molecular mechanisms underlying Chagas disease and its most severe chronic form, the chronic Chagas cardiomyopathy.

The specific objectives were:

1. To assess previously reported genetic associations with Chagas disease using the candidate genes strategy in different Latin American populations.

2. To perform a GWAS in Chagas disease to identify new *loci* associated with susceptibility to infection and the development of chronic Chagas cardiomyopathy.

3. To explore the influence of genetic ancestry in relation to *T.cruzi* infection using an admixture mapping approach.

4. To functionally characterize the GWAS-*loci* associated with chronic Chagas cardiomyopathy by integrating genomic data with DNA methylation data.

Genomic approaches to unravel the pathogenesis of Chagas disease

MATERIAL AND METHODS, RESULTS AND DISCUSSION

Genomic approaches to unravel the pathogenesis of Chagas disease

Chapter 1: Genetic polymorphisms of *IL6, IL17A* and *IL18* genes and their associations with Chagas disease.

As mentioned in the previous chapter, genetic factors and the immune response have been suggested to be determinant in the susceptibility against the infection and the outcome of Chagas disease. For this, the analysis of genetic variants localized in the vicinity of genes involved in the immune response, such as the interleukin encoding genes, might give novel insights in the pathogenesis of the infection and its chronic cardiac form. In this chapter we analyzed *IL6*, *IL17A* and *IL18* genetic variants regarding the predisposition to *T. cruzi* infection and the development of chronic Chagas cardiomyopathy in different Latin American populations. Individuals were classified as seronegative and seropositive for *T. cruzi* antigens, and the last group was divided into asymptomatic and chronic Chagas cardiomyopathy patients. The results showed association of the *IL17A* and *IL18* genetic variants with the differential susceptibility to the infection, while only the *IL18* polymorphisms showed association with the chronic form of the disease, confirming previously published results.

1.1. Materials and methods

1.1.1. Ethics statement

The study was accepted by the Ethics Committees from the Industrial University of Santander and Cardiovascular Foundation, Colombia (Act No. 15/2005), the Vall d'Hebron University Hospital, Barcelona, Spain (PR (AMI) 297/2016), and the National Hospital of Clinics, National University of Cordoba, Argentina (CIEIS HNC 118/2012 and 2/16/2017). Written informed consent was obtained from all subjects prior to participation. The research protocols followed the principles of the Declaration of Helsinki, and informed consent was obtained from all individual participants included in the study.

1.1.2. Study design and patient populations

Candidate-gene case-control studies were performed in Colombian, Bolivian and Argentinian cohorts in order to replicate previous findings and test the association of *IL6*, *IL17A* and *IL18* polymorphisms. Additionally, a meta-analysis was performed combining the tested cohorts for each analyzed gene. Sample sized of each cohort for the *IL6*, *IL17A* and *IL18* genes are summarized in Table 1.1.

	1.1. Sample size from the differ nfection and the chronic Chagas			candidate gene	e study for <i>T.</i>
Gene		Colombia	Bolivia	Argentina	Brasil
	Sex (%female)	55%	69%	71%	-
	Age (mean±SD)	52.12±17.65	48.83±9.45	53.82±13.52	-
IL6	n (seropositive/seronegative)	744/647	-	272/78	-
	n (CCC/asymptomatic)	468/276	100/530	182/90	-
	Sex (%female)	58%	69%	71%	51.71%
IL17A	Age (mean±SD)	45.55±17.19	48.83±9.45	53.82±13.52	62.60±13.70
IL17A	n (seropositive/seronegative)	937/640	-	272/78	260/150
	n (CCC/asymptomatic)	576/361	100/530	182/90	212/48
	Sex (%female)	58%	69%	71%	52%
11.40	Age (mean±SD)	45.55±17.19	48.83±9.45	53.82±13.52	-
IL18	n (seropositive/seronegative)	937/640	-	272/78	-
	n (CCC/asymptomatic)	576/361	100/530	182/90	849/202

Total of individuals included in each candidate gene study from the different populations. CCC chronic Chagas cardiomyopathy.

Colombian cohort: Colombian samples were recruited in an endemic region for Chagas disease (Guanentina and Comunera provinces, Santander department) by the health care team from the Industrial University of Santander and Cardiovascular Foundation from Colombia. Samples were classified as seronegative or seropositive according to the serological tests of recombinant antigen enzyme-linked immunosorbent Assay (ELISA) and commercial indirect hemagglutination. Additionally, basing on complementary tests and clinical findings, seropositive patients were classified as chronic Chagas cardiomyopathy patients or asymptomatic individuals. Thus, in the case of the study of the IL6 polymorphisms, a total of 1,391 Colombian individuals were assessed for the susceptibility to the infection and the development of the chronic cardiac form of the disease. The genetic data for the selected SNPs for this gene of these individuals were additionally meta-analyzed with previously published data from Colombia (Colombia I) and Peru (107). The mean age of the participants was 52.12±17.65 years. Regarding the study of IL17A and IL18 polymorphisms, a total of 406 Colombian

MATERIAL AND METHODS

individuals from the same population as the study by Leon Rodriguez et al. (2016) were taking into account for the analyses (122). In order to increase the sample size, these individuals were included with the previously published Colombian cohort, making a total of 1,577 individuals to assess the susceptibility to both infection and chronic form.

Bolivian cohort: A total of 630 Bolivian individuals residents in Barcelona, Spain were recruited from the Infectious Diseases Department of the Vall d'Hebron University Hospital. This cohort is composed only by seropositive individuals, which were subjected to electrocardiograms, echocardiograms, and chest radiography in order to identify cardiac involvement. The mean age of the chronic group was 50.71±9.41, while for the asymptomatic individuals was 46.93±9.49.

Argentinian cohort: A total of 350 Argentinian individuals from an endemic region for Chagas disease (Cordoba province) were included in this study. The samples were recruited from the National Hospital of Clinics and Sucre Clinic, Cordoba city. As for the Colombian cohort, all participants underwent a serological diagnosis for *T. cruzi* infection through ELISA assays in combination with a commercial indirect hemagglutination test. Seropositive individuals were classified as classified as chronic Chagas cardiomyopathy patients and asymptomatic according to the results of complementary tests and clinical findings, seropositive individuals were. The mean age of participants was 53.82±16.53 years for seronegative individuals, 60.14±10.16 for chronic patients and 49.30±13.65 for the asymptomatic group.

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Brazilian cohort: The Brazilian population was included in the study of *IL17A* and *IL18* polymorphisms based on previous studies. Regarding the study of *IL17A* polymorphisms, data from 410 Brazilian individuals drawn from Reis et al. 2017, originally from South and Southeast regions of Brazil were included in the meta-analysis for the susceptibility to the infection and the chronic Chagas cardiomyopathy (123). In the case of *IL18*, a total of 1,051 Brazilian seropositive patients for antibodies against *T. cruzi* were included in the meta-analysis.

1.1.3. Selected polymorphisms and genotyping.

For each gene under study, selected polymorphisms corresponded to: rs1800795, as a promoter variant of the IL6 gene; rs4711998, rs8193036 and rs2275913 for the IL17A gene, previously assessed in Chagas disease in a Colombian cohort and in a Brazilian cohort (123); and the rs2043055, rs1946518 and rs360719 SNPs in representation of the IL18 gene. LD (R² and D') was estimated using an expectationmaximization algorithm implemented in Haploview V4.2 (124) for the studied cohorts and from the American sub-populations genotype data from the 1KGP phase III (54) in the case of the *IL18* polymorphisms in conjunction with the LDlink website tool (https://ldlink.nci.nih.gov/?tab=ldmatrix).

Genomic DNA from blood samples was isolated following standard procedures, and the genotyping was performed using TaqMan assays (Applied Biosystems, Foster City, California, USA) on a real-time PCR system (7900HT Fast Real-Time PCR System), SNPs were determined by TaqMan 5' allelic discrimination assay method performed by Applied Biosystems.

1.1.4. Statistical analyses

For the candidate gene study, the statistical analyses were performed with the software Plink V1.9 (125). Deviance from Hardy-Weinberg equilibrium (HWE) was determined at the 1% significance level in all groups of individuals. Individuals that did not achieved an SNP completion rate of 95% were filtered out. To test for possible allelic association, logistic regression model and Fisher's exact test were assessed in seropositive vs. seronegative individuals and chronic Chagas cardiomyopathy individuals vs. asymptomatic, using age and sex as covariates. Odds ratios (OR) and 95% confidence intervals (CI) were calculated and in those cases that control for multiple testing were necessary the Benjamini & Hochberg step-up false discovery rate (FDR) correction was used. *P-values* lower than 0.05 were considered as statistically significant.

To assess the consistency of effects across the cohorts, a metaanalysis was performed with METASOFT (126) based on inversevariance-weighted effect size. Heterogeneity across studies was assessed using the Cochran's Q statistic (Q test *P-value*<0.05) and I² heterogeneity index (126). A fixed-effects model was applied for those SNPs without evidence of heterogeneity (Cochran's Q test P-value>0.05), and a random-effects model was applied for SNPs displaying heterogeneity of effects between studies (Cochran's Q test *P-value*≤0.05). The statistical power of the studies was estimated with the Power Calculator for Genetic Studies 2006 (CaTS) software (http://www.sph.umich.edu/csg/abecasis/CaTS) (127) and the allele frequency used for statistical power calculation was the one described for the Americans sub-populations of the 1KGP phase III (54). Evaluation of functionality of the SNPs was performed with the online software

HaploReg v4.1 (128) based on empirical data from the ENCODE project (http://www.genome.gov/encode) and focused specifically on experiments performed on blood and T cells lines in the American population. For regulatory features, Ensembl Browser (54) and ReMap 2018 v1.2 (129) were used.

RESULTS

1.2. Results

The variants under study were in HWE in all the analyzed cohorts (*P-value*>0.01) and the genotyping success rate was >90% for all of them. In addition, allele frequencies in all cases were similar to those described for the Americans sub-populations of the 1KGP phase III (54). Demographic information and sample sizes for the included cohorts are summarized in Table 1.

1.2.1. IL6 results

Regarding the association of *IL6* with *T. cruzi* infection susceptibility, allelic and genotypic frequencies of seronegative and seropositive individuals from the Colombian and Argentinian cohorts were compared for the variant rs1800795 (Table 1.2). In both cohorts, the frequency of the minor allele C, was increased in seronegative compared with seropositive individuals, but no significant association between *IL16* rs1800795 and susceptibility to *T. cruzi* infection was found after applying logistic regression adjusted by sex and age. In the meta-analysis, when combining the newly recruited cohorts (Colombia and Argentina) and the previously published data (Colombian I and Peruvian), the effect of the C allele was consistent among the populations and indicated protection against *T. cruzi* infection. However, no statistically significant differences were observed under a fixed-effects meta-analysis (*P-value*= 0.433, OR=0.94, 95%CI= 0.81-1.10) (Table 1.3).

	enotype and all omatic individua								
				Genotype. N (%)				
Cohorts	Group	Sample size	CC	CG	GG	MAF	OR	(L95-U95)	P-value
			Serc	positive vs. Se	ronegative				
Colombia	Seronegative	639	41 (6.36)	198 (30.7)	406 (62.95)	21.71	0.91	(0.75-1.12)	0.371
COIOIIIDIa	Seropositive	744	26 (3.51)	247 (33.21)	470 (63.18)	20.1	0.91	(0.75-1.12)	0.571
Augontino	Seronegative	76	6 (7.79)	22 (28.57)	49 (63.64)	22.08	0.06	(0.56-1.32)	0.479
Argentina	Seropositive	272	10 (3.7)	85 (31.48)	175 (64.81)	19.44	0.86	(0.56-1.52)	0.479
			CCC	patients vs. Asy	mptomatic				
Colombia	Asymptomatic	276	6 (2.17)	95 (34.42)	175 (63.41)	19.38	1 1 2	(0.81-1.54)	0.502
COIOIIIDIa	CCC	468	21 (4.49)	153 (32.69)	294 (62.82)	20.83	1.12	(0.81-1.54)	0.502
Bolivia	Asymptomatic	529	5 (0.95)	80 (15.12)	444 (83.93)	8.51	1 50	(0.00.2.(1)	0.119
Bolivia	CCC	100	2 (2.00)	18 (18.00)	80 (80.00)	11	1.53	(0.89-2.61)	0.119
A	Asymptomatic	89	1 (1.12)	32 (35.96)	56 (62.92)	19.1	1.05	(0 (2 1 75)	0.070
Argentina	CCC	182	9 (4.97)	53 (29.28)	119 (6575)	19.61	1.05	(0.62-1.75)	0.868

odds ratio, L95-U95 confidence intervals of 95% L lower limit, U upper limit.

In the case of chronic Chagas cardiomyopathy, the rs1800795*C allelic frequency in Colombian, Bolivian and Argentinian cohorts was increased patients compared with asymptomatic individuals, suggesting a risk for Chagas cardiomyopathy susceptibility in these populations, but in none of these cohorts the results were statistically significant after performing logistic regression adjusted by sex and age (Table 1.2). Moreover, although the allele effect was consistent in all the Latin American cohorts analyzed no significant association was detected for *IL6* rs1800795 and chronic Chagas cardiomyopathy when the fixed-effect meta-analysis was performed (Table 1.3) (*P-value*=0.257, OR=1.11, CI=0.93-1.33).

I ante T.J. Meta	sictinita		111 T. CI UZ	Indepension anarchui	מוווה מווח ב			معمد سالماله معامرة المسالم	רווא זוו דימרו		ö
				Se	ropositive	Seropositive vs. seronegative					
Colombia (N=1,383)	1,383)	Colombia I (N=554)	=554)	Argentina (N=348)	=348)	Peru (N=162)	52)			Meta-analysis (N=2,447)	I=2,447)
OR (L95-U95)	P-value	OR (L95-U95) P-value OR (L95-U95) P-value OR (L95-U95) P-value OR (L95-U95) P-value	P-value	OR (L95-U95)	P-value	OR (L95-U95)	P-value			OR (L95-U95) P-value	P-value
0.91 (0.74-1.12)	0.371	0.91 (0.74-1.12) 0.371 1.11 (0.84-1.48) 0.466 0.86 (0.56-1.32) 0.479 0.55 (0.23-1.31) 0.183	0.466	0.86 (0.56-1.32)	0.479	0.55 (0.23-1.31)	0.183			0.94 (0.81-1.10) 0.433	0.433
				CC	C patients v	CCC patients vs. Asymptomatic					
Colombia (N=744)	=744)	Colombia I (N=239)	=239)	Argentina (N=271)	=271)	Peru (N=78)	8)	Bolivia (N=629)	29)	Meta-analysis (N=1,961)	l=1,961)
OR (L95-U95)	P-value	OR (L95-U95) P-value	P-value	OR (L95-U95)	P-value	0R (L95-U95)	P-value	OR (L95-U95)	P-value	OR (L95-U95)	P-value
1.12 (0.81-1.58)	0.502	1.12 (0.81-1.58) 0.502 1.12 (0.72-1.70) 0.651 1.05 (0.62-1.75) 0.868 3.17 (0.8-12.63) 0.107 1.53 (0.89-2.61) 0.119 1.11 (0.93-1.33) 0.257	0.651	1.05 (0.62-1.75)	0.868	3.17 (0.8-12.63)	0.107	1.53 (0.89-2.61)	0.119	1.11(0.93-1.33)	0.257
CCC chamic Charac	cardiomic	CCC chamic Charace and invention N total individuale included from anch achaet. OD adde meia. 106,1106 annflama	inde included	from orch cohort OD	odde ratio I	05-1105 confidence int	ormule of OE	% I louer limit: II unn	or limit		

CCC chomic Chagas cardiomyopathy, N total individuals included from each cohort, OR odds ratio, L95-U95 confidence intervals of 95% L lower limit, U upper limit.

1.2.2. IL17A results

In the susceptibility to *T. cruzi* infection analysis, only two *IL17A* genetic variants were significant when comparing seropositive and seronegative individuals. Therefore, those variants significantly different between cases and controls after adjust by sex and age, were rs8193036*C and rs2275913*A for the Colombian and Brazilian cohorts, respectively (Table 1.4). In the case of the Argentinian cohort no associations of the variants tested was found. Furthermore, the meta-analysis combining each individual cohort (Colombian, Argentinian and Brazilian) for *IL17A* was performed (Table 1.4). The rs2275913*A allele effect was consistent in the three cohorts and the association improved after the meta-analysis, showing statistically significant results (*P-value*=0.016, OR=1.21, 95% CI=1.06–1.41, under a fixed-effects meta-analysis) after Bonferroni correction, however no association was observed for the *IL17A* rs4711998 and rs8193036 variants in the meta-analysis.

				Sero	Seropositive vs. Seronegative	ative				
	Colombia	ia	Argentina	E	Brazil			Meta-á	Meta-analysis	
SNP	OR (L95-U95)	P-value	0R (L95-U95)	P-value	0R (L95-U95) P-value	P-value	OR (L95-U95)	95)	P-value	
rs4711998*A	rs4711998*A 0.94 (0.78-1.14)	0.582	1.38 (0.90-2.12)	0.143			0.99 (0.84–1.17)	1.17)	0.946	
rs8193036*C	rs8193036*C 0.83 (0.70-0.99)	0.043	1.34 (0.89-20.2)	0.164			0.89 (0.76-1.05)	1.05)	0.169	
rs2275913*A	rs2275913*A 1.16(0.95-1.4)	0.136	1.07(0.67 - 1.69)	0.793	1.46(1.05 - 2.05)	0.032	1.21 (1.06-1.41)	1.41)	0.016	
				CCC p	CCC patients vs. Asymptomatic	matic				
	Colombia	ia	Argentina	в	Bolivia		Brazil		Meta-analysis	sis
SNP	OR (L95-U95)	P-value	OR (L95-U95)	P-value	OR (L95-U95) P-value	P-value	OR (L95-U95)	P-value	0R (L95-U95)	P-value
rs4711998*A	rs4711998*A 0.86 (0.67-1.11)	0.259	1.08(0.69-1.68)	0.751	0.96(0.65 - 1.41)	0.831			0.92 (0.76-1.11)	0.396
rs8193036*C	rs8193036*C 0.92 (0.72-1.18)	0.526	0.74(0.49 - 1.29)	0.164	1.18(0.85 - 1.62)	0.319			0.95(0.80 - 1.14)	0.616
rs2275913*A	0.8 (0.62-1.02)	0.081	0.72(0.43 - 1.21)	0.217	1.14(0.75 - 1.71)	0.543	1.21(0.74 - 1.99)	0.463	0.89 (0.74-1.07)	0.232
				CCC	CCC patients vs. Seronegative	ative				
	Colombia	ia	Argentina	e	Brazil			Meta-a	Meta-analysis	
SNP	OR (L95-U95)	P-value	OR (L95-U95)	P-value	OR (L95-U95) P-value	P-value	OR (L95-U95)	95)	P-value	
rs4711998*A	rs4711998*A 0.93 (0.75-1.16)	0.541	1.47 (0.92-2.37)	0.109			1.02 (0.82-1.23)	1.23)	0.927	
rs8193036*C	rs8193036*C 0.84 (0.67-1.05)	0.133	1.21 (0.78-1.88)	0.389			0.91(0.74 - 1.10)	1.10)	0.323	
rs2275913*A	rs2275913*A 1.14 (0.90-1.44)	0.298	0.99(0.60-1.61)	0.955	1.52(1.08 - 2.15)	0.021	1.20 (1.01-1.45)	1.45)	0.040	

In the case of the analysis of IL17A variants in chronic Chagas cardiomyopathy patients, no significant associations were detected in the Colombian, Bolivian and Argentinian cohorts after the logistic regression adjusted by sex and age (Table 1.4). These results were consistent with previous findings in the Brazilian cohort (123). As consequence, any significant association was detected for the available SNPs when the meta-analysis was performed combining these cohorts (Table 1.4). In addition, in order to evaluate the possible association between *IL17A* genetic variants and chronic Chagas cardiomyopathy, CCC patients were compared with seronegative individuals, as previously performed (106, 123). While in the Colombian and Argentinian cohorts no associations were found after applying logistic regression adjusted by sex and age, in the rs2275913*A allele was nominally significant in the Brazilian cohort, as previously reported (123). The effect of this variant is consistent in the Colombian, Argentinian and Brazilian cohorts, and the association with chronic cardiomyopathy susceptibility improved after the meta-analysis showing nominally statistical differences (P-value=0.040, OR=1.20, 95% CI=1.01-1.45, under a fixed-effects meta-analysis, Table 1.4).

Given its statistical association with the infection and the chronic phase, an *in silico* functional analysis of the *IL17A* variant rs2275913 and the ones in high LD ($R^2 \ge 0.8$) was performed on peripheral mononuclear blood in American population from the 1KGP (Table 5). The annotation indicates that these SNPs map in enhancer regions and marks of histone modifications (H3K4me1, H3K4me3, H3K27ac and chromatin marks), potentially modulating gene expression.

Table 1.5. Functional annot Functional annotation from	ional annotation tation from mon	ı. Kegula Ionuclea	Table 1.5. Functional annotation. Regulatory chromatin states and instone modifications for 1L17A rs2275913 and SNPs in high LD (K 20.8). Functional annotation from mononuclear peripheral blood specifically primary T helper 17 cells.	es and hist specifically	tone modifi 7 primary T	cations for IL helper 17 ce	17A rs227593 lls.	13 and SNPs	in high LD (R	²≥0.8).
Position ^a	SNPs	\mathbb{R}^2	Functionality	MAF (AMR)	eQTL	Chromatin states ^b	Chromatin states ^c	H3K4me1	H3K4me1 H3K4me3	H3K27ac
chr6:52051033	rs2275913	-	Intergenic variant	25%	I	Flank	Promoter	Enhancer	Promoter	Enhancer
chr6:52087034 rs11966760	rs11966760	0.82	Intergenic variant	24%	PAQR8	Enhancer	Promoter	Enhancer	Promoter	Enhancer
chr6:52056386	chr6:52056386 rs16882180	0.8	0.8 Intergenic variant 25%	25%	PAQR8	Ι		Enhancer	Enhancer Promoter	ı
^a According to National Center for imputed marks. H3K4me1 Histon Allele Frequency, AMR American.	onal Center for Biot K4me1 Histone H3 MR American.	echnolog lysine 4 ı	^a According to National Center for Biotechnology Genome Reference Consortium NCBI build GRCh37. ^b Core 15-state model. ^c Chromatin states: 25-state model using 12 imputed marks. H3K4me1 Histone H3 lysine 4 mono-methylation, H3K4me3 Histone H3 lysine 4 tri-methylation, H3K27acHistone H3 lysine 27 acetylation, MAF Minor Allele Frequency, AMR American.	nsortium NC 4me3 Histoı	.BI build GRC ne H3 lysine	h37. ^b Core 15-s 4 tri-methylatio	tate model. ^c Chı n, H3K27acHistu	omatin states: one H3 lysine 2	25-state mode 7 acetylation, N	using 12 1AF Minor

RESULTS

1.2.3. IL18 results

In order to assess the association of *IL18* genetic variants with the differential susceptibility to *T. cruzi* infection, the allelic and genotypic frequencies of seronegative and seropositive individuals from Colombia were compared (Table 1.6). The allelic frequencies of the three SNPs were statistically significant even after multiple testing corrections. The frequency of the minor allele, G, in rs2043055 was significantly reduced in the seronegative compared to seropositive individuals suggesting an association with higher infection risk, while the frequencies of rs1946518*T and rs360719*G alleles were significantly increased in seronegative compared to the seropositive individuals, suggesting an association with the protection against the infection by T. cruzi. Regarding the Argentinian cohort, no associations between *IL18* genetic variants were found (Table 1.6). However, the rs2043055 remained borderline significant for protection against infection by T. cruzi (Pvalue=0.061, OR=0.71, 95% CI=0.49-1.02). In addition, a meta-analysis combining data from Colombian and Argentinean cohorts was performed (Table 1.7). The IL18 rs360719 showed consistent effects among the two meta-analyzed populations with a statistically significant association (P-value= 0.001, 95% CI=0.66-0.89, under a fixed-effects meta-analysis) with an OR for the G allele of 0.76. For this comparison, the sample size attained a statistical power of over 80% for this OR. In both cohorts, the allele effects size were in concordance and this result indicates an association to the protection against *T. cruzi* infection in these cohorts.

Table 1.6. Ge Colombian co	enotype and all ohort.	ele distri	bution for IL1	8 variants in se	eronegative an	d seropos	sitive ind	lividuals from	the
CND		41140	C	Genotype N (%)		MAE		Allele test	
SNP		A1 A2	1 1	1 2	2 2	MAF	OR	(L95-U95)	P-value
				Colombia					
rs2043055	Seronegative	G A	82(13.00)	300(47.54)	249(39.46)	36.77	1.3	(1.10-1.53)	0.002
152043033	Seropositive	ЧA	164(17.69)	450(48.54)	313(33.76)	41.96	1.5	(1.10-1.55)	0.002
rs1946518	Seronegative	T G	163(25.83)	334(52.93)	134(21.24)	52.3	0.79	(0.67-0.92)	0.003
181940510	Seropositive	110	214(23.09)	448(48.33)	265(28.59)	47.25	0.79	(0.67-0.92)	0.005
rs360719	Seronegative	CIA	103(16.32)	299(47.39)	229(36.29)	40.02	0.75	(0.63-0.89)	0.001
18300719	Seropositive	G A	107(11.54)	426(45.95)	394(42.50)	34.52	0.75	(0.63-0.69)	0.001
				Argentina					
rs2043055	Seronegative	CLA	20(25.97)	29(37.66)	28(36.36)	44.81	0.71	(0.49-1.02)	0.061
182043035	Seropositive	G A	33(12.23)	126 (46.67)	111(41.11)	35.56	0.71	(0.49-1.02)	0.001
rs1946518	Seronegative	TIC	19(24.67)	35(45.46)	23(29.87)	47.4	1.02	(0.71 1.40)	0.883
151940518	Seropositive	T G	54(20.00)	151(55.92)	65(24.08)	47.96	1.03	(0.71–1.49)	0.083
rs360719	Seronegative	CIA	11(14.28)	33(42.85)	33(42.85)	35.71	0.87	(0 (0 1 21)	0.552
15500/19	Seropositive	G A	25(9.26)	128(47.40)	117(43.33)	32.96	0.87	(0.60–1.31)	0.552

1 minor allele | 2 major allele; alleles are showed in forward strand. MAP minor allele frequency, OR odds ratio, L95-U95 confidence intervals of 95% L lower limit, U upper limit. Values adjusted by sex and age. Individuals included in Colombia: 927 seropositive and 631 seronegative, and in Argentina: 270 seropositive and 77 seronegative.

Argentinian cohort Bolivian Brazilian Meta-analysis P-value OR (195-U95) P-value OR (195-U95) P-value OR (195-U95) Image (195-U95) Meta-analysis P-value OR (195-U95) P-value OR (105-U156)									D	•					
P-value OR (L95-U95) P-value OR (L95-U95) P-value OR (195-U95) P-value OR $T.$ cruzi infection susceptibility $T.$ cruzi infection susceptibility $T.$ cruzi infection susceptibility 0.02 $0.71 - 0.02$ $0.04 - 1.02$) 0.061 $ -$		Colombian coho	prt	Arg	entinian cohort			Bolivian			Brazilian			Meta-analysis	
T. cruzi infection susceptibility 0.002 0.71 (0.49-1.02) 0.061 - - - 1.17 (1.01-1.36) 0.003 1.03 (0.71-1.49) 0.883 - - - - 0.82 (0.71-0.94) 0.001 0.87 (0.60-1.31) 0.552 - - - 0.76 0.82 (0.71-0.94) 0.001 0.87 (0.60-1.31) 0.552 - - - 0.76 0.82 (0.71-0.94) 0.001 0.87 (0.60-1.31) 0.552 - - - 0.76 0.82 0.76 0.001 0.87 (0.60-1.31) 0.552 - - - 0.76 0.82 0.76 0.037 1.26 (0.82-1.30) 0.088 1.06 (0.85-1.32) 0.598 1.07 0.994 0.97 1.07 0.994 0.91 0.934 - - - 0.95 0.976 0.95 0.95 0.95 0.95 <th>OR</th> <th>(L95-U95)</th> <th></th> <th>OR</th> <th>(L95-U95)</th> <th>P-value</th> <th>OR</th> <th>(L95-U95)</th> <th>P-value</th> <th>OR</th> <th>(L95-U95)</th> <th>P-value</th> <th>OR</th> <th>(T95-U95)</th> <th>P-value</th>	OR	(L95-U95)		OR	(L95-U95)	P-value	OR	(L95-U95)	P-value	OR	(L95-U95)	P-value	OR	(T95-U95)	P-value
0.002 0.71 (0.49-1.02) 0.061 - - - - - 1.17 (1.01-1.36) 0.003 103 (0.71-1.49) 0.883 - - - - 0.82 (0.71-0.94) 0.001 0.87 (0.60-1.31) 0.552 - - - - 0.76 (0.71-0.94) 0.001 0.87 (0.60-1.31) 0.552 - - - 0.76 (0.70-0.89) 0.001 0.87 (0.60-1.31) 0.552 - - - 0.76 (0.71-0.94) 0.001 0.87 (0.60-1.31) 0.552 - - - 0.76 (0.71-0.94) 0.037 1.26 (0.82-1.36) 0.988 1.06 (0.85-1.32) 0.598 1.05 (0.82-1.35) 0.225 0.67 (0.44-1.04) 0.784 1.98 0.066 - - 1.07 (0.90-1.26) 0.234 0.84 0.84 0.84 0.934 -						T. cruzi	infectior	n susceptibility	v						
0.003 1.03 (0.71-1.49) 0.883 - - - - 0.82 (0.71-0.94) 0.001 0.87 (0.60-1.31) 0.552 - - - 0.76 (0.66-0.89) Chronic Chagas cardiomyopathy susceptibility 0.037 1.26 (0.82-1.95) 0.291 1.39 (0.95-2.02) 0.088 1.06 (0.85-1.32) 0.598 1.05 (0.82-1.35) 0.037 1.26 (0.84-1.04) 0.078 1.24 (0.85-1.30) 0.298 1.06 (0.82-1.35) 0.90-1.26) 0.225 0.67 (0.44-1.04) 0.078 1.24 (0.85-1.80) 0.26 - - 1.07 (0.90-1.26) 0.994 0.81 (0.52-1.27) 0.834 - - - 0.95 (0.70-1.15)	1.3	(1.10 - 1.53)	0.002	0.71	(0.49 - 1.02)								1.17	(1.01 - 1.36)	0.035
0.001 0.87 (0.60-1.31) 0.552 - - - - 0.76 (0.66-0.89) Chronic Chagas cardiomyopathy susceptibility 0.037 1.26 (0.82-1.95) 0.291 1.39 (0.95-2.02) 0.088 1.06 (0.85-1.32) 0.598 1.05 (0.82-1.35) 0.225 0.67 (0.44-1.04) 0.078 1.24 (0.85-1.80) 0.26 - 1.07 (0.90-1.26) 0.934 0.81 (0.52-1.27) 0.364 0.98 (0.66-1.45) 0.934 - - 0.95 (0.79-1.15)	0.79		0	1.03	(0.71 - 1.49)								0.82	(0.71 - 0.94)	0.006
Chronic Chagas cardiomyopathy susceptibility 0.037 1.26 (0.82-1.95) 0.291 1.39 (0.95-2.02) 0.088 1.06 (0.85-1.32) 0.598 1.05 (0.82-1.35) 0.025 0.67 (0.44-1.04) 0.078 1.24 (0.85-1.80) 0.26 - 1.07 (0.90-1.26) 0.994 0.81 (0.52-1.27) 0.364 0.98 (0.66-1.45) 0.934 - - 0.95 (0.79-1.15)	0.75		0.001	0.87	(0.60 - 1.31)	0.552	•			•			0.76	(0.66-0.89)	0.001
0.037 1.26 (0.82-1.95) 0.291 1.39 (0.95-2.02) 0.088 1.06 (0.85-1.32) 0.598 1.05 (0.82-1.35) 0.225 0.67 (0.44-1.04) 0.078 1.24 (0.85-1.80) 0.26 - - 1.07 (0.90-1.26) 0.994 0.81 (0.52-1.27) 0.364 0.98 (0.66-1.45) 0.934 - - 0.95 (0.79-1.15)					Chrc	onic Chagas	cardiom	nyopathy susce	epti bili ty						
0.67 (0.44-1.04) 0.078 1.24 (0.85-1.80) 0.26 - - 1.07 (0.90-1.26) 0.81 (0.52-1.27) 0.364 0.98 (0.66-1.45) 0.934 - 0.95 (0.79-1.15)	0.79	(0.64 - 0.99)	-	1.26	(0.82 - 1.95)	0.291	1.39	(0.95 - 2.02)	0.088	1.06	(0.85 - 1.32)		1.05	(0.82 - 1.35)	0.259
0.994 0.81 (0.52-1.27) 0.364 0.98 (0.66-1.45) 0.934 -	1.14	(0.92 - 1.41)	-	0.67	(0.44 - 1.04)	0.078	1.24	(0.85 - 1.80)	0.26				1.07	(0.90 - 1.26)	0.426
	0.99	(0.79 - 1.26)		0.81	(0.52 - 1.27)	0.364	0.98	(0.66 - 1.45)	0.934				0.95	(0.79 - 1.15)	0.629

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RESULTS

The allelic genotypic frequencies of chronic and Chagas cardiomyopathy and asymptomatic patients from Colombia were compared for the *IL18* genetic variant rs2043055, which was statistically significant even after multiple testing correction (*P-value*=0.037, OR=0.79, 95% CI=0.64-0.99, Table 1.8). The frequency of the rs2043055*G allele was significantly incremented in asymptomatic patients, suggesting an association with the protection against the development of Chagas cardiomyopathy. However, no significant differences in allelic frequencies were observed for rs1946518 and The rs2043055 was additionally studied in 1,051 rs360719. seropositive Brazilian patients (Table 1.1) (109). Nevertheless, no significant differences were found in the Brazilian, Bolivian and Argentinian cohorts, only a trend of association can be observed for the SNP rs2043055 in the Bolivian cohort (Table 1.8). On the other hand, the rs1946518 showed an increase of the T allele frequency in asymptomatic individuals from the Argentinian cohort remaining borderline significant (Table 1.8). Further, a meta-analysis combining these results were performed, but the results of the available SNPs showed no significant associations (Table 1.7).

	enotype and all duals from the (c and chro	nic Cha	gas cardiomy	opathy
SNP		41142		Genotype N (%)	MAF		Allele test	
SNP		A1 A2	1 1	1 2	2 2	MAF	OR	(L95-U95)	P-value
				Colombia					
rs2043055	Asymptomatic	G A	83(23.18)	159(44.41)	116(32.40)	45.39	0.79	(0.64-0.99)	0.037
182043035	CCC	GIA	81(14.24)	291(51.14)	197(34.62)	39.81	0.79	(0.04-0.99)	0.037
rs1946518	Asymptomatic	T G	82(22.91)	160(44.69)	116(32.40)	45.25	1.14	(0.92-1.41)	0.225
151940310	CCC	ЦG	132(23.20)	288(50.62)	149(26.19)	48.51	1.14	(0.92-1.41)	0.225
rs360719	Asymptomatic	G A	45(12.57)	155(43.30)	158(44.13)	34.22	0.99	(0.79-1.26)	0.994
15500/19	CCC	GIA	62(10.90)	271(47.63)	236(41.48)	34.71	0.99	(0.79-1.20)	0.994
				Bolivia					
rs2043055	Asymptomatic	G A	72(13.64)	260(49.24)	196(37.12)	38.26	1.39	(0.95-2.02)	0.088
152043033	CCC	uл	16(16.00)	50(50.00)	34(34.00)	41	1.39	(0.93=2.02)	0.000
rs1946518	Asymptomatic	G T	101(19.13)	268(50.76)	159(30.11)	44.51	1.24	(0.85-1.80)	0.26
151940510	CCC	GII	19(19.00)	55(55.00)	26(26.00)	46.5	1.24	(0.85-1.80)	0.20
rs360719	Asymptomatic	G A	54(10.23)	237(44.89)	237(44.89)	32.67	0.98	(0.66-1.45)	0.934
15300719	CCC	ЧА	10(10.00)	50(50.00)	40(40.00)	35	0.90	(0.00=1.43)	0.934
				Argentina					
rs2043055	Asymptomatic	G A	9(10.11)	40(44.94)	40(44.94)	32.58	1.26	(0.82-1.95)	0.291
132043033	CCC	uл	24(13.26)	86(47.51)	71(39.22)	37.02	1.20	(0.02-1.93)	0.271
rs1946518	Asymptomatic	T G	21(23.59)	51(57.30)	17(19.10)	52.25	0.67	(0.44-1.04)	0.078
131740310	CCC	ца	33(18.23)	100(55.25)	48(26.51)	45.86	0.07	(0.44-1.04)	0.070
rs360719	Asymptomatic	G A	8(8.99)	47(52.81)	34(38.20)	35.39	0.81	(0.52-1.27)	0.364
15300/19	CCC	uА	17(9.32)	81(44.75)	83(45.86)	31.77	0.01	(0.32-1.27)	0.304

Total number of individuals: 569 CCC and 358 asymptomatic in the Colombian cohort; 100 CC and 528 asymptomatic in the Bolivian cohort; and 181 CC and 89 asymptomatic in the Argentinian cohort. 1 minor allele | 2 major allele, alleles are showed in forward strand. MAF minor allele frequency, OR odds ratio, L95-U95 confidence intervals of 95% L lower limit, U upper limit. Values adjusted by sex and age.

The functional annotations of the three *IL18* variants were also explored using HaploReg v4.1, which indicates a location of these variants in a regulatory region of the genome. The annotation based on the epigenomic information of rs2043055 indicates that this SNP maps in an enhancer region, which is correlated with active gene expression in primary mononuclear cells and in T cells from peripheral blood. The rs1946518 and rs360719 variants mapped in a region enriched in histone marks: H3K4me3, H3K9ac, a hallmark of active promoter region, and H3K27ac in enhancer region in mononuclear cells and T cells. Furthermore, according to ReMap 2018 v1.2 these three SNPs mapped in regulatory regions of the human genome and it has been described as transcription factors.

1.3. Discussion

Association studies offer a potentially powerful approach to identify genetic variations that are involved in the immunopathogenesis of Chagas disease (36, 130). However, individual genetic association studies frequently have limitations and the results may be specific to the population of the study. The meta-analysis approach has been proposed to resolve these limitations, to increase the power of statistical analyses (131, 132) and to reach to more conclusive results in order to improve our understanding of the genetic basis underlying Chagas disease.

1.3.1. IL6

Several studies have found associations between *IL6* polymorphisms and infectious diseases susceptibility (133-136). In this sense, the IL6 polymorphism, rs1800795, has been associated with other affections, such as *Toxoplasma gondii* (134), *Chlamydia trachomatis*, and *Chlamidia* pneumoniae infections (136). In this sense, certain functional polymorphisms in genes encoding cytokines may determine the pathogenesis of Chagas disease, as IL-6 is involved in in the acute and chronic phases of Chagas disease (36, 87, 137-141). For this, several polymorphisms within the *IL6* gene region were assessed to confirm its association with Chagas disease. One of these polymorphisms was the variant rs1800795, which is located in the promoter region of the *IL6* gene so it can modulate its expression (142-144). Although previous studies determined differences in IL-6 expression among T. cruzi infected and non-infected patients in previous studies (145, 146), the association analysis of this variant did not show significant associations with the infection or the chronic phase in the five independent Latin American populations. The observed differences in IL-6 expression among T. cruzi infected and non-infected patients in previous studies

(145, 146), could be due to wide and complex mechanisms, since its regulation involves interactions between transcription factors, other promoter polymorphisms and DNA methylation (147-151). In spite of this, the lack of association of *IL6* promoter variant rs1800795 in this analysis does not exclude the contribution of other genetic variants within the gene region to be involved in the human susceptibility to Chagas disease.

This is the largest candidate gene study carried out in *IL6*, with a total of 3,087 individuals from different countries included. The statistical power in the meta-analysis was 96% to detect associations with OR=1.3 at the 5% significance level. This, together with the fact that the minor allele frequencies of the variant analyzed presented variation among cohorts, reflect the complex genetic architecture of Latin Americans, a recently admixed population (66, 152). Taking this into consideration, these results do not provide an evidence of significant association between *IL6* rs1800795 with *T. cruzi* infection or chronic Chagas cardiomyopathy in the populations under study. Thus, further studies in other genetic variants of this gene in well-powered cohorts are needed to definitively discard *IL6* gene variants as susceptibility markers for Chagas disease.

1.3.1. IL17A

In the early stages of the infection, the IL-17A is a crucial cytokine secreted by a wide range cell types such as Th17, B cells, innate lymphoid cells, CD4+, CD8+, gamma-delta T and invariant NKT (153-157). Thus, given the importance of this cytokine in the immune response, three *IL17A* genetic variants were assessed in four Latin American populations, Colombia, Argentina, Bolivia and Brazil. The

analyses showed evidence of the implication of rs2275913 in the T. cruzi infection and the development of chronic Chagas cardiomyopathy. The rs2275913 is a functional polymorphism that modifies the binding of the transcriptional nuclear factor of activated T cells (NFAT) in the IL-17A promoter and, as showed in the in silico analysis, is also located in promoter histone marks, which potentially modulates gene expression. Thus, the association observed in the analysis might be suggesting a variation in gene expression that ends in a lower IL-17A production. In this sense, the reduction in the levels of this cytokine has been described to impede a rapid proinflammatory activation of chemokines and cytokines for the resolution of *T. cruzi* infection (155, 156). On their side, the substitution of the G by an A allele in the associated variant has been demonstrated to be significantly associated with autoimmune diseases and cancer (158-161). However, results reported in serum showed that the A allele was associated with a higher (161-163), lower (164, 165) or no significant (166) levels of transcription and synthesis of the protein. As the results show, the A allele of the rs2275913 shows a risk effect in terms of *T. cruzi* infection. This genetic variation might produce changes in IL17A gene expression modifying IL-17A production, which would impede a rapid proinflammatory activation of chemokines and cytokines for the resolution of *T. cruzi* infection (155, 156). To confirm this, functional studies of this variant are required the complexity of the functional effect of this polymorphism.

Several studies showed the immunomodulatory role of IL-17A in the chronic phase of Chagas disease (167-170). *IL17* expression by Th17 cells and B cells were found in patients with cardiac involvement with higher frequency when compared with asymptomatic patients, correlating with worse cardiac function (168). Furthermore, an

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production of IL-17A produce a proinflammatory exacerbated environment in Chagas severe heart disease (171, 172). Nevertheless, the present association test showed no significant differences when *IL17A* genetic variants were compared among chronic cardiac patients and asymptomatic individuals from Colombia, Argentina, Bolivia and Brazil were compared. The lack of association observed of *IL17A* genetic variants in the presented analyses might be consequence of an insufficient statistical power or genetic heterogeneity among the populations under study. This lack of replication may occur if the studied polymorphism is not the causal variant but is rather in LD with it, as LD patterns depend on the genetic background of the founder population and population history. In this sense, Latin American populations suppose always a challenge due to their genetic diversity and recent admixture (65, 152, 173). Interestingly, at the present time, IL-17A have become a relevant drug target in various forms of autoimmune and inflammatory diseases, mainly as negative modulators of the secreted protein (156, 174). Thus, two antibodies are currently in Phase IV of drug development for the treatment of immune system diseases, namely, Secukinumab and Ixekizumab (Anatomical Therapeutic Chemical [ATC] code L04AC10 and L04AC13, respectively). Given the role of IL-17A as a key cytokine in the pathogenesis of Chagas disease, the opportunities for drug repurposing becomes very important, as there are only two treatments available, Benznidazole and Nifurtimox, with high rates of adverse effects and treatment withdrawal (91, 175).

1.3.1. IL18

IL-18 is produced by a wide variety of cells, including dendritic cells, macrophages, keratinocytes, intestinal epithelial cells, and osteoblasts, with a key pathophysiological role in health and disease (176) This

interleukin also plays an important role in the regulation of IFN- γ production and development of Th1 response. Given the numerous roles of IL-18 in the immune response, several studies have highlighted its implication in the acute and chronic phase of Chagas disease (177-180). Considering that infectious diseases exert significant selective genetic pressure, it has been proposed two mechanisms to explain the implication of the genetic background in the pathogenesis of Chagas disease (137). First, pathogen resistance genes (PRG) would be involved in inhibits infection by directly reducing pathogen burden and secondly, disease tolerance genes (DTG) will operate to minimize tissue damaging effects of the pathogen (181-183). Consequently, polymorphisms in PRG and DTG will be associated with differential disease progression. One of the most relevant disease tolerance genes identified was related to directly or indirectly inhibit IFN- γ production or Th1 differentiation (137), and therefore, IL-18 could be implicated in this regulation.

Taking this into consideration, three *IL18* genetic variants were assessed in four Latin American populations, being the largest candidate gene study conducted in this gene and Chagas disease to date. These results evidenced the implication of the *IL18* rs360719 polymorphism in the susceptibility to the infection although, when comparing cardiomyopathy and asymptomatic patients, no significant associations were detected. The previous study in a Colombian cohort was the first to report an association for rs2043055, rs1946518 and rs360719 with *T. cruzi* infection, suggesting that this association was mainly driven by the polymorphism rs360719 (122). After the enlargement of this cohort, the association of these variants was performed in an Argentinian cohort and only rs2043055 showed a borderline genetic association but

in the opposite direction compared with the Colombian cohort. These differences could be due to the complex genetic structure of Latin American individuals, reflected by the recent admixture among Native American, European, and African source populations (152). Also, as might occur with IL17A polymorphisms, the lack of replication may occur if the assessed polymorphism is not the causal variant but is rather in LD. Although the rs360719 in the Argentinian cohort showed no association with *T. cruzi* infection, probably as consequence of an insufficient statistical power in this case, the meta-analysis showed that this variant presented a consistent effect among the two cohorts, indicating protection against *T. cruzi* infection. This variant is located in the promoter region of the *IL18* gene and its functional annotation revealed its location in histone marks in primary mononuclear cells and T helper naive cells from peripheral blood, also being described as transcription factor. In addition, it has been described that IL18 rs360719 polymorphism leads to loss of the octamer 1 (OCT-1) transcription factor binding site. OCT-1 is known to be a ubiquitously expressed factor involved in the regulation of certain cytokines, like IL-18 (184).

Local expression of *IL18* in chronic Chagas cardiomyopathy heart tissue has been described and seems to be associated with mononuclear inflammatory infiltrates, cardiomyocyte destruction and fibrosis (171). Previous analyses of the *IL18* genetic variant rs2043055 in the Brazilian population showed nominal significant differences in the genotypic frequencies among moderate and severe chronic cardiac patients (109). In the present analyses, when compare chronic Chagas cardiomyopathy with asymptomatic patients the same variant showed significant association in the Colombian cohort. However, these results were not

validated in the Bolivian and Argentinian cohorts. These discrepancies in results could be due to the genetic heterogeneity among the study cohorts (152), as the impact of European colonization and slave trade from western Africa has been demonstrated to altered the genomes of Native Americans in multiple and dynamic ways (66, 185). Thus, previous results have showed that the prevalence of *T. cruzi* infection is strongly and independently associated with higher levels of African and Native American ancestry in a Brazilian population (186). This heterogeneity in ancestry proportions across geographic regions, and also within countries themselves, are challenging in association studies in order to find generalizable results across populations (187-189). All this, suggests that a fine-scale genomics perspective might represent a powerful tool to understand the role of genetics in this neglected disease diagnosis and prognosis.

In summary, these results validated the previous work on *IL18*, highlighting the role of rs360719 in the susceptibility to *T. cruzi* infection (122). Although, no evidence of association was found between the *IL18* genetic variants and chronic Chagas cardiomyopathy in the populations under study, the meta-analyses supposes a powerful approach to identify genetic variants that influence susceptibility of common diseases (190, 191). Thus, in the context of Chagas disease is necessary to contemplate the challenges of studying such an heterogeneous populations like Latin Americans with recent admixture, where finescale genomic assessments may be necessary (152). Therefore, further studies are needed to reach more conclusive results concerning the genetic variability in the IL18 genomic region.

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Chapter 2: Identification of novel susceptibility *loci* in chronic Chagas cardiomyopathy throughout a genomewide association study

Complex diseases are influenced by the added contribution of several common genetic variants. In addition, the genetic architecture of these traits makes a challenge the assessment of its influence using candidate gene approaches. Thus, GWAS represent a great approach in the study of the genetic component of infectious diseases. For this, in order to better elucidate the genetic basis of Chagas disease and chronic Chagas cardiomyopathy, this chapter shows the results of the largest GWAS and meta-analysis of different Latin American populations carried out to the date in this disease. The meta-analysis revealed a novel genome-wide statistically significant association with chronic Chagas cardiomyopathy development nearby the *SAC3D1* gene that, in combination with the *in silico* functional relationships between the associated variant and the *SNX15*, *BAFT2*, and *FERMT3* genes, provides further insights into the pathogenesis of this neglected disease.

2.1. Materials and methods

2.1.1. Study populations, sample collection and ethical considerations

Samples from 3 different Latin American countries: Colombia, Bolivia, and Argentina were included in this study and meta-analyzed with data from the previous GWAS in a Brazilian population (192), comprising a total of 3,699 genomic DNA recruited samples. The description of sample origin is available elsewhere (193). Chagas disease cases were defined by their serological status using recombinant antigen and a commercial indirect hemagglutination test for ELISA assays. Participants were classified as seropositive and seronegative for *T. cruzi* antigens and comparisons were performed among them in order to assess susceptibility to the infection in the Colombian and Argentinian cohort. The Bolivian and Brazilian cohorts were composed only of seropositive individuals. In addition, all the seropositive patients recruited in this study were subjected to electrocardiograms, echocardiograms, and chest radiography to identify cardiac involvement. For the assessment of the chronic Chagas cardiomyopathy susceptibility, those patients with cardiac abnormalities were compared to asymptomatic individuals. Data included from the Brazilian cohort were described elsewhere (192). Sample size and demographic information are summarized in Table 2.1. The study was approved by the Ethics Committees from the Industrial University of Santander and Cardiovascular Foundation, Colombia (Act No. 15/2005), the Vall D'Hebron University Hospital, Barcelona, Spain (PR (AMI) 297/2016), and the National Hospital of Clinics, National University of Cordoba, Argentina (CIEIS HNC 118/2012 and 2/16/2017). The Ethics Committees for the Brazilian cohort was described elsewhere (192). Protocols used in

the study followed the principles of the Declaration of Helsinki, and all individuals included in the study signed written informed consents.

Table 2.1. Demog	graphical cl	aracteristi	cs and clas	sification	of data coll	ections.			
				Seropo	sitive				Seronegative
	Colo	mbia	Bol	ivia	Arge	ntina	Bra	azil [*]	Colombia
	CCC	ASY	CCC	ASY	CCC	ASY	ССС	ASY	
Subjects, N	577	442	97	518	100	62	207	306	1,104
Sex (% females)	280 (48 %)	282 (64%)	57 (59%)	364 (70%)	65 (65%)	40 (65%)	81 (39%)	154 (53%)	672 (61%)
Age (Mean±SD)	67.4±14.3	57.2±15.6	50.7±9.4	46.4±9.5	58.2±9.4	47.5±14.0			49.5±16.7

The sample sizes refer to data passing the genotyping quality controls.

Abbreviations: CCC, chronic chagas Cardiomyopathy; ASY, asymptomatic.

* Data set from Deng et al, 2013

2.1.2. Genotyping, quality control and imputation

Genomic DNA from blood samples from the Colombia, Bolivian, and Argentinian cohorts were isolated using the *QIAamp Midi*DNA Kit following manufacturer's recommendations. The samples were genotyped using the Global Screening Array Platform (Illumina Inc., San Diego, CA, USA). Quality controls (QCs) of genotyped data and consistency of sex assignment were performed using Plink v.1.9 (125). Individuals and variants with missing genotype rate <5% were excluded. SNPs with different call rates between cases and controls (*P-value*<0.05), or deviating from HWE with a *P-value*≤1×10⁻⁰⁶ in the control group were removed.

A principal component analysis (PCA) was performed using Plink1.9 to compare our data with the data from 1KGP. The genomic inflation factor (λ) was calculated to control the type I error rate with the package "gap" for R version 3.6.1. After QCs, genotyped data were imputed with the Michigan

Imputation Server using the admixed American population from 1KGP phase 3 as reference panel (54). After imputation, variants were filtered out by MAF<1% and the imputation quality metric Rsq<0.3.

2.1.3. Statistical analysis

For each case-control collection, association testing for imputed allele dosages was performed using a mixed model with EMMAX (194) implemented in EPACTS (https://genome.sph.umich.edu/wiki/EPACTS). Given the common environmental exposure to triatomine vectors of individuals living together in endemic areas, related individuals were maintained in the analysis. In this sense, genotyped data were used to calculate the pairwise identity by descent (IBD) proportion (PI HAT). duplicated individuals were removed (PI_HAT>0.8) and related individuals were considered with a PI_HAT ≥ 0.4. EMMAX calculates the kinship matrix to include it as a covariate in the statistical analysis, as well as sex and age. The summary level statistics of each cohort, including the Brazilian cohort, were meta-analyzed using METASOFT (126) under a fixed and random-effect models based on variants' heterogeneity (Cochran's Q test *P-value*). As most of imputed variants (>95%) did not show heterogeneity (Cochran's Q test Pvalue>0.05), those analyzed under a random-effect model were omitted from further analysis. Conservative suggestive associations were established at a *P-value*<1×10⁻⁰⁶ and genome-wide significance at *P-value*<5×10⁻⁰⁸ as usually employed for GWAS. The SNPnexus software (https://www.snpnexus.org/v4/) was used to annotate the suggestive and significant signals, and LocusZoom (http://locuszoom.org/) for the regional association plots.

2.1.4. Functional annotation of associated variants

An *in silico* approach was used to investigate the potential biological consequences of the associated variant. The LDlink tool (195) was used to calculate proxies in high LD with the top signal (r²>0.8) and different databases were queried. The Open Targets Genetic was used to assess their functional implication (https://genetics.opentargets. org/), and complemented with PheWeb version 1.1.17 (http://pheweb.sph.umich.edu/SAIGE-UKB/) to assess these variants in previous Phenome-Wide Association Studies (PheWAS). We used HaploReg v4.1 (196) to evaluate regulatory genomic regions and Capture HiC Plotter (197) to assess long-distance physical interactions. RegulomeDB (https://regulomedb.org/regulomesearch/) and GTEx (https://gtexportal.org/home/) were used to evaluate expressionquantitative trait *loci* (eQTLs) in relevant tissues.

RESULTS

2.2. Results

A total of 3,413 postQCs samples from 4 independent Latin American cohorts from Colombia, Bolivia, Argentina, and Brazil were included in this study. After QCs and imputation, 7,846,902 SNPs for the Colombian and 8,408,292 SNPs for the Argentinian cohorts were analyzed correcting by age, sex, and kinship in the case of susceptibility to the infection by *T. cruzi* analysis, while in the case of chronic Chagas cardiomyopathy development, a total of 8,218,190 SNPs were tested in the meta-analysis including the Colombia, Bolivian, Argentinian, and Brazilian populations.

2.2.1. Population structure

Given the ethnic admixture of Latin American populations, the first 2 principal components were used to evaluate the genetic heterogeneity existing in the samples included in this study and in comparison with the admixed American subpopulation from 1KGP. The results of these analyses showed that the heterogeneity present in our samples is consistent with the observed in the admixed American subpopulation of 1KGP. Despite this, the Argentinian and Bolivian sample collections seemed to be more homogenous, while a considerable genetic variability with the Brazilian cohort was detected, reinforcing the high degree of genetic admixture of this population (Figure 2.1). In the association analyses within each cohort, the effect of population structure was taken into account by using a mixed-model of association (194).

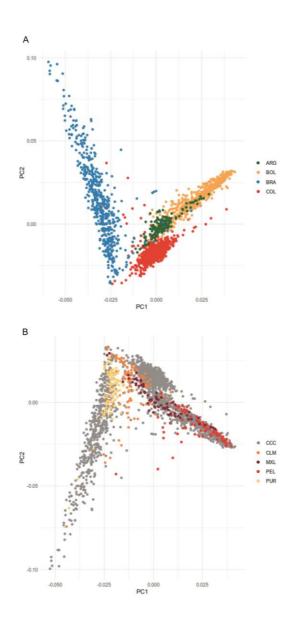


Figure 2.1. Principal component analysis of (A) study populations in comparison with (B) the Admixed American populations from the 1000 Genomes Project. Abbreviations: ARG Argentina; BOL Bolivia; BRAZ Brazil; COL Colombia. For 1000 Genomes Project populations: CLM Colombian; MXL Mexican; PEL Peruvian; PUR Puerto Rican.

2.2.2. Differential Susceptibility to T. cruzi infection

Individuals from the Colombian and Argentinian cohorts were classified and compared according to their serological status in seropositive and seronegative for *T. cruzi* infection. No statistically relevant results were identified in the association test in the Argentinian population, mainly due to the reduced sample size of this cohort. However, in the case of the Colombian cohort, we detected 7 suggestive associations (*P-value*<1×10⁻⁰⁶) contained in 4 different *loci* due to the LD (Table 2.2). The strongest association is an intronic variant of the *EBF2* gene (rs147475322; *P-value*=2.15×10⁻⁰⁷, OR=1.20, 95% CI 0.76–1.90). The other 3 SNPs are located in an intronic region of *CD247* gene (rs554994388) and map downstream and upstream of *SIK1* (rs229347) and *IL18* (rs4937075) genes, respectively (Table 2.2). Remarkably, a different SNP within the *IL18* gene was previously associated in a candidate gene assessment (193); however, this variant was nominally associated in our study and did not reach the suggestive level of significance.

Table	2.2. Sug	gestive* signa	ls associated v	vith suscept	ibility	to Trypa	nosoma	<i>cruzi</i> infecti	ion in the Colo	mbian cohort
Chr	Locus	Вр	SNP	SNP status	Rsq	Effect allele	Allele Freq	p-value	OR (95% CI)	Function
1	CD247	167431884	rs554994388	Imputed	0.5	G	0.03	5.35E-07	1.33 (1.19-1.49)	Intronic
8	EBF2	25979367	rs147475322	Imputed	0.86	А	0.03	2.15E-07	1.2 (0.76-1.90)	Intronic
11	IL18	112127040	rs4937075	Imputed	0.77	С	0.77	8.34E-07	1.02 (0.86-1.21)	Intergenic
21	SIK1	43428190	rs229347	Imputed	0.88	А	0.6	8.93E-07	1.07 (1.04-1.10)	Intergenic

*Suggestive associations were considered with a *p*-value ≤1E-06

Associated loci were determined in order to the LD of contained variants and the most significant SNP of each loci were reported in the table.

The r-squared metric (Rsq) indicates the correlation between input genotypes and imputed dosages.

Abbreviations: Allele Freq allele frequency; bp base pair; chr chromosome; CI confidence interval; OR odds ratio; SNP info SNP information.

2.2.3. Differential susceptibility to the development of chronic Chagas cardiomyopathy

In order to identify novel genetic variants associated with differential chronic Chagas cardiomyopathy development, seropositive samples were stratified according to their cardiological status, and patients with cardiac abnormalities were compared to asymptomatic individuals in the association test. After QCs, genotype imputation, and filtering, a case-control association analysis from GWAS data was performed in the Colombian, Bolivian and Argentinian samples comprising a total of 2,309 individuals (981 chronic cardiac patients and 1,328 asymptomatic individuals). After correcting by age, sex, and kinship, deviations in genomic inflation factors were not observed. Next, summary statistics from every cohort, including the Brazilian data, were meta-analyzed by an inverse variance-weighted effect size and those shared by at least 2 data sets were considered. Additionally, we did not observe inflation in the meta-analysis (λ =1.03). As result, one statistically significant associated signal was identified at GWASlevel (*P-value* $<5 \times 10^{-08}$) and 22 suggestive associations (*P-value* $<1 \times 10^{-06}$) in 3 different autosomal *loci* (Table 2.3; Figure 2.2). The strongest association (rs2458298; *P-value*=3.27×10⁻⁰⁸, OR=0.90, 95% CI 0.87–0.94) is located in chromosome 11 in an intronic region of the NAALADL1 gene. This signal is followed by several proxy variants in high or moderate LD (r²>0.4) located nearby NAALADL1, SAC3D1, and SNX15 genes (Figure 2.3). Regarding the rest of the suggestive signals, they are located in intergenic regions close to CDH8 and KLF4 genes (Table 2.3).

			e significant pment in th			00	ive* sigi	nals asso	ciated wit	h chronic Cha	gas
Chr	Locus	Вр	SNP	SNP status	Rsq	Effect allele	Allele Freq	p-value Q	p-value	OR (95% CI)	Function
5	CDH8	18718887	rs10472156	Imputed	0.81	А	0.73	0.72	8.89E-07	1.1 (1.06-1.14)	Intergenic
9	KLF4	107479876	rs10759240	Genotyped	-	С	0.54	0.98	4.68E-07	0.92 (0.89-0.95)	Intergenic
11	SAC3D1	65047341	rs2458298	Imputed	0.95	G	0.26	0.98	3.27E-08	0.9 (0.87-0.94)	Intronic

*Suggestive associations were considered with a *p*-value \leq 1E-06

Associated *loci* were determined in order to the LD of contained variants and the most significant SNP of each *loci* were reported in the table.

In boldface the genome-wide significant signal.

The r-squared metric (Rsq) indicates the correlation between input genotypes and imputed dosages.

Abbreviations: Allele Freq allele frequency; bp base pair; chr chromosome; CI confidence interval; OR odds ratio; *p-value* Q Cochran's Q test; rsq r-square; SNP info SNP information.

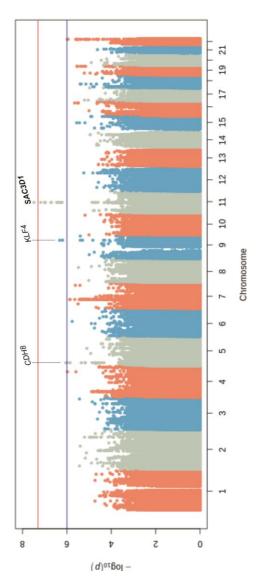


Figure 2.2. Manhattan plot of the association test in chronic Chagas cardiomyopathy meta-analysis results. Y and X axes refer the $-\log_{10}$ transformed P-values and positions in chromosomes, respectively. The red horizontal line refers the genome-wide association threshold, stablished at P-value< 5×10^{-08} , and the blue line refers the suggestive threshold (P-value< 1×10^{-06}).

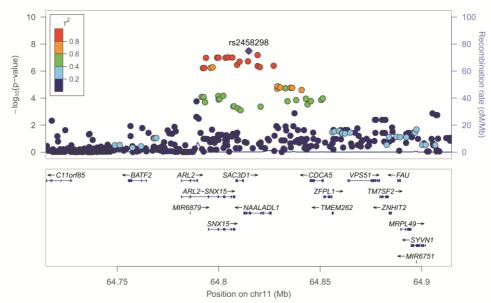


Figure 2.3. Regional association plot for the genome-wide significant signal. The –log10 transformed P-values of associated variants are plotted by position. The genomic position indicated correspond the top variant associated in this region. The rest of the variants are colored in function of their degree of LD with the associated signal. LD is based on pairwise r² values from the American population from the 1KGP. The blue line corresponds to the estimated recombination rates. Nearby genes to the associated and suggestive signals are shown in the bottom of the plot.

2.2.4. Functional analysis

To explore functional features of the genomic association obtained from the meta-analysis in chronic Chagas cardiomyopathy, several databases were queried. Regarding the top associated variant rs2458298, although is located in an intronic region of *NAALADL1*, it is closer to the transcription start site (TSS) of SAC3D1, being this considered as its nearest gene. PheWAS information from Open Targets Genetics revealed a risk association of this SNP with cardiovascular traits such hypertension (Pas valuePheWas=7.4×10⁻⁰⁶) and high blood pressure (P-valuePheWas=5.8×10⁻ ⁰⁶) in the UK Biobank database, as well as with abdominal aortic aneurysm (P-valuePheWas=1.9×10⁻⁰³) in a subset of patients. Significant eQTLs (i.e. loci that explain part of the variance in gene expression in specific tissues) were described identifying rs2458298 as correlated with SNX15 expression in heart atrial appendage (*P*-value = $9.9 \times 10-10$) and heart left ventricle (*P*value=1.3×10⁻¹¹). The same SNP was also correlated with the expression of FERMT3 in a lymphoblastoid cell line. Using the Capture Hi-C Plotter platform, 42 interactions of this SNP with different gene regions across chromosome 11 were identified in the experiments done by Mifsud and colleagues (198) in a lymphoblastoid cell line (GM12878). One of those relevant promoter interactions is with the BATF2 gene, which has been previously implicated in T. cruzi infection (199). All these results are summarized in Table 2.4.

		<i>ico</i> functional a development	nalysis of th	e associated var	iant with chronic Chagas
Chr	SNP	Nearest gene	eQTLs	C-HiC Genes	PheWAS
11 r	s2458298	SAC3D1 NAALADL1	SNX15 ^{°a} FERMT3 ^b	<i>BATF2</i> in GM12878 ^b	Hypertension, high blood pressure and abdominal aortic aneurysm

^aeGENE in heart atrial appendage and heart left ventricle

Queried data bases were Open Targets Genetics, GTEx, RegulomeDB, Capture HiC Plotter, and PheWEB

Abreviations: Chr chromosome; C-HiC capture Hi-C; eQTL expression quantitative trait loci; PheWAS phenome-wide association study.

^blymphoblastoid cell line

2.3. Discussion

Several suggestive *loci* have been identified in association with differential susceptibility to *T. cruzi* infection in our study. Among these variants, highlight those located in intergenic regions near IL18 and CD247 given their implication in the immune response (200, 201). As described in previous chapters, IL18 encodes a proinflammatory cytokine that is involved in the innate and adaptive response, and is critical for T-cell differentiation into IFN-γ producing Th1-type T cells and for IFN-γ production by NK cells (201). IFN- γ is a key player in the acute *T. cruzi* infection and a pathogen resistance gene, aiding the control of *T. cruzi* parasitism by stimulating surrounding cells to produce $TNF-\alpha$ and other inflammatory mediators leading to the generation of peroxynitrite (137). Additionally, IFN- γ gene expression has been reported to increase in chronic cardiac patients in several studies conducted in Latin American populations, indicating an important role of this molecule in the pathogenesis of the chronic cardiac form through the induction of inflammatory damage (202). Here, we confirm the association of this gene with the differential susceptibility to T. cruzi infection; although given the directed nature of candidate gene studies, different SNPs of the same locus were reported (122, 193). Regarding *CD247*, this gene encodes a subunit of the T-cell receptor CD3 complex that plays an important role in antigen recognition (200). In Chagas disease, previous works have confirmed lower expression of CD3 in isolated blood cells from individuals with high exposure to the parasite in Colombian populations (203). Finally, genetic variants in the CD247 were also previously associated with autoimmunity (204-206) pointing to the existence of shared genes in autoimmune and infectious diseases as described elsewhere (207). Remarkably, a statistically significant association at genomic level was identified near the SAC3D1 gene region (rs2458298) with the development of chronic Chagas cardiomyopathy. *SAC3D1*, also known as *SHD1*, has been identified as a transcriptional regulator of *STAT5* (208), also associated with cardioprotection in humans (209). Additionally, the STAT-5 signaling by IL-2, IL-7, and IL-15 receptors has been shown to be perturbed in peripheral and heart-infiltrating T cells in chronic Chagas cardiomyopathy (210). An *in silico* functional assessment of the most associated variant and its proxies was performed in order to acknowledge the interactions and processes where they are involved.

Interestingly, physical contact of chromatin regions in this genetic location revealed interaction with *BATF2*. This gene encodes a transcription factor (TF) that has been related with Chagas disease, regulating the IL-23-Th17 pathway and suggesting an immunoregulatory function during *T. cruzi* infection in a mouse model (199). On the other hand, the top signal has been associated with several cardiovascular traits in PheWas data from UK Biobank. Additionally, this variant is an eOTL of SNX15 in heart atrial appendage and heart left ventricle. This gene encodes a member of the sorting nexin protein family, which has been related with cardiovascular diseases (211). Additionally, the associated variant was also an eQTL for FERMT3 (Fermitin Family Member 3) in lymphoblastoid cell lines. The FERMT3 belongs to the Kindlins family and it encodes a protein that plays an important role in the regulation of thrombosis, as well as in maintaining the cytoskeleton of erythrocytes (212). This gene was previously associated with serum triglyceride levels, considered as a risk factor for coronary heart disease (213). Although the diagnosis of chronic Chagas cardiomyopathy involves the exclusion of other cardiomyopathies, the association with this gene may implicate common pathways among them. Other suggestive signals are those located in chromosome 9 near to *KLF4*, which encodes the Kruppel-like factor 4, a TF related with cardiac mitochondrial homeostasis in a mouse model (214), as well as with a high expression of the inducible

nitric oxide synthase and NO production in different cell types (215). Taking into account that higher production of NO has been reported in chronic Chagas cardiomyopathy severity (216), it would be possible a relationship among *KLF4* and chronic Chagas cardiomyopathy through the NO production.

This study has certain limitations. First, a cross-sectional assessment and the posterior seroconversion or development of cardiac events cannot be ruled out. In this case, the inclusion of cases in the control group will bias the results towards the null hypothesis of no association, weakening the risk of a false positive result. Second, the associated genes have been nominated as related with the disease through a functional assessment in publicly available non-endemic populations. Further functional analyses will guarantee the acceptance or rebuttal of these hypotheses. Finally, the small sample size in the overall study, but especially in the Chagas disease susceptibility assessment, bounds the statistical power for detecting lower frequency variants or with smaller effects. This is the reason why suggestive variants were highlighted, including their potential functional relevance. The main strength of our study is that we reveal a consistent and reproducible association with chronic Chagas cardiomyopathy in 4 Latin American populations. Genetic studies in admixed populations are challenging given their differential ancestry proportions (217), which is further complicated when taking also into account the genetic diversity of the infectious agent (13). This heterogeneity has been previously observed in a related parasitic disease such as Leishmania (83). Further studies evaluating the humanparasite genetic interactions may improve the knowledge of disease pathogenesis. Finally, the inclusion of underrepresented populations in genetic studies is a great opportunity to broaden the knowledge of complex diseases and deepen their molecular mechanisms (218), which is especially important in Chagas disease, due to its significant socio-economic burden in both endemic and non-endemic countries (3).

We identified a novel locus near the gene region of *SAC3D1* associated with the chronic Chagas cardiomyopathy as well as several suggestive associations with the differential susceptibility to the infection by *T. cruzi* in several Latin American populations. These findings highlight the role of the host genetic component in Chagas disease, providing important novel leads to understand the pathogenesis of this neglected disease.

Chapter 3: Differential associations of genetic ancestries with Chagas disease susceptibility in the Colombian population using an admixture mapping approach

The complexity of the host genetic implication in infectious diseases, such as Chagas disease, makes possible the contribution population-specific polymorphisms to the disease risk, especially in recently admixed populations. This is the case of Latin American populations, which are a three-way admixture between Native American, European and African populations, being the Colombian the one with the most disparate admixture proportions. In order to assess genomic susceptibility regions where affected individuals share their genetic ancestry, this chapter shows the results of the first admixture mapping study carried out in Chagas disease for the Colombian population. A two-stage study was conducted and subjects were classified as seropositive and seronegative for T. cruzi, obtaining a protective association of the Native American ancestry with Chagas disease in the Major Histocompatibility Complex region. Additionally, a selective sweep scan in an AMR reference population from 1KGP together with an *in silico* functional analysis highlighted the Tripartite Motif family and the Human Leukocyte Antigen genes, with crucial role in the immune response against pathogens.

3.1. Material and Methods

3.1.1. Ethical considerations

The protocols used in the study followed the Declaration of Helsinki principles and informed consent was obtained from all individuals included in the study design. The Industrial University of Santander and Cardiovascular Foundation (Colombia) Ethics Committee approved this study (Act No. 15/2005).

3.1.2. Study population and genotyping

All donors were recruited by the health care system from the Industrial University of Santander and Cardiovascular Foundation in the provinces of Guanentina, Comunera and Garcia Rovira, which are the provinces with the highest prevalence of Chagas disease in the Santander department in Colombia (219). We used a two-stage case-control design where individuals were classified as seropositive (cases) or seronegative (controls) for T. cruzi antigens according to an indirect hemagglutination commercial test (Chagatest, Wiener, Argentina) and enzymelinked immunosorbent assays (Test ELISA Chagas III, Grupo Bios, Chile; Chagas ELISA IgG + IgM, Vircell, Spain). Samples from stage 1 comprised 1,576 individuals (933 classified as cases and 643 as controls) as described elsewhere (220). Further sample recruitment composed stage 2, including 654 independent samples (122 cases and 532 controls) according to the same classification criteria. Genomic DNA isolation of blood samples was performed using the QIAamp Midi DNA Kit (OIAGEN. Germanv) following manufacturer's recommendations. All samples were genotyped with the Global Screening

Array Platform (Illumina Inc., San Diego, CA, USA) as described elsewhere (221). As part of the QCs of genotyped data, individuals and variants with missing genotype rate >5%, SNPs with different call rates between cases and controls (*P-value*<0.05), and SNPs with large deviations from HWE in the control group (*P-value*≤1×10⁻⁶) was removed. QCs were performed using PLINK v.1.9 (125).

3.1.3. Admixture mapping analysis

Reference data from the 1KGP Phase III (54) was used as representatives of the parental populations in the downstream stage 1 global and local ancestry assessments. Briefly, we used data from Native American (AMR, n=85), European (EUR, n=503) and African (AFR, n=504) reference populations. Regarding the AMR population, only Peruvian from Lima (PEL) were included because this population has been described to have the highest proportion of AMR ancestry after the Maya population (222). All EUR subpopulations were taken into account (Utah residents, Finnish, British, Iberian and Toscani populations). In the case of AFR, both the African Caribbean from Barbados, and individuals with African ancestry from southwest US were excluded as they are populations with recent admixture events (223). Using PLINK, we intersected the autosomal SNPs from stage 1 genotyped data (post QC 493,271 SNPs) with those from the 1KGP dataset, removing from the latter those variants with missing genotype rate >5% or with large deviations from HWE (*P*-value $\leq 1 \times 10^{-6}$) in at least one population. After the intersection of datasets and QCs, a total of 471,342 SNPs were kept for further analysis.

Global ancestry assessment. Individual global ancestries were obtained with ADMIXTURE v1.3.0 (224), which calculates average ancestry

proportions across the genome. We estimated the best number of ancestral populations (K) using a 10-time cross validation with random seeds. The best fitting was obtained for K=3, i.e. assuming 3 ancestral populations distinguishing AMR, EUR and AFR ancestries.

Local ancestry assessment and association testing. Local ancestries were estimated with ELAI v1.0 (225) assuming a three-way admixture and 10 generations since the last admixture event as has been indicated by previous studies (66, 152). As QC steps, a correlation was calculated for individual global and averaged local ancestries across the genome using R v3.6.1. Principal components were calculated using PLINK and plotted along with those from the representative parental populations from 1KGP.

For the admixture mapping, the three local ancestries by separate were tested for case–control association using the EMMAX mixed model (194) implemented in EPACTS (https://genome.sph.umich.edu/wiki/EPACTS). As related individuals were considered, EMMAX calculates the kinship matrix to include it as a covariate in the model, in addition to age and sex. We controlled for the type I error rate by calculating the λ using an in-house script for R v3.6.1. Significance was adjusted by the number of ancestry blocks across the genome and the number of generations since the admixture using the R package STEAM (66). Based on that, the significance threshold for the admixture mapping study was established at *P-value*<3.07×10⁻⁰⁶, similar to estimates that have been declared in independent studies in Latin American populations (66).

3.1.4. Fine mapping

A fine mapping assessment was performed to elucidate the most likely genetic variant(s) underlying the admixture mapping study results combining data from the two stages. Briefly, genotypic data from stage 2 were subjected to the same QCs that were used in stage 1. After this, imputation of both stages was performed with the Michigan Imputation Server using the admixed American population from 1KGP phase 3 as reference panel. Imputed variants were filtered by their MAF and the imputation quality metric Rsq. Those variants satisfying both a MAF>1% and Rsq>0.3 were kept for the study. Association testing of imputed allele dosages was performed in both stages by separate using the EMMAX mixed model implemented in EPACTS, including age, sex and the kinship matrix as covariates. Summary-level statistics of each stage were meta-analyzed using METASOFT v2.0 (126). A random or fixed-effect size meta-analysis was selected for each variant based on the results for the Cochran's Q-test of heterogeneity. Significant and suggestive thresholds for the assessed region established at *P-value*<3.23×10⁻⁰⁴ and *P-value*<6.46×10⁻⁰³, were respectively, according to the estimates of GEC software (226) based on the LD structure from the stage 1 samples. In order to confirm the association of the specific allele with the local AMR ancestry, samples were stratified according to their ancestry punctuations and allele frequencies were recalculated in cases and controls separately. The statistical power of the fine mapping analysis was estimated using the Power Calculator for Two Stage Association Studies software CaTS (127).

3.1.5. Selective sweep analysis

Given the high rate of adaptive signals in the genome triggered by parasites, and that admixture serves as a mechanism driving adaptive evolution in humans (68, 227), we used iSAFE v1.0.4 (228) to provide evidence of a selective sweep embedded in the admixture mapping region and to pinpoint the most likely favored variant. iSAFE exploits the evolutionary contributions hidden in the flanking regions surrounding the region under selection to provide a ranking of variants (iSAFE-score) based on their contribution to the overall signal of selection. For the analysis, we used 1KGP data from unrelated subjects from PEL population (n=77) and from a random selection of 10% Yoruba individuals drawn from the reference (n=91), which represents a non-target or outgroup population. iSAFE was executed enabling the IgnoreGaps flag and the default MaxFreq value (0.95). Ancestral fasta sequences for Homo sapiens (GRCh37) were downloaded ENSEMBL from release 75 (http://ftp.ensembl.org/pub/release-75/fasta/ancestral_alleles/).

3.1.6. In silico functional analysis

In silico functional analyses were performed to assess the biological consequences of the leading associated variants in the fine mapping and of those variants prioritized in the selective sweep analysis. The Open Targets Genetics portal (<u>https://genetics.opentargets.org/</u>) was used for functional annotation, to assess trait associations based on PheWAS and to retrieve the evidence of eQTLs in relevant tissues based on eQTLGen database (<u>https://www.eqtlgen.org/</u>). Additionally, long-distance physical interactions and regulatory genomic regions were considered using Capture HiC Plotter (197) and HaploReg v4.1 (196).

Chapter 3: Admixture mapping in Chagas disease

3.2. Results

This study has a two-stage case-control design and samples were classified as seropositive (cases) and seronegative (controls) for parasite antigens. Demographic characteristics are summarized in Table 3.1. Cases and controls from the stage 1 samples were matched in terms of global ancestries (Figure 3.1). Local ancestry blocks were estimated for the 471,342 positions corresponding to those of genotyped SNPs in stage 1. Local ancestry punctuations per individual were averaged and compared with their global ancestries showing a high correlation among them (r=0.89–0.96) and indicating consistent global and local ancestry proportions in the population under study. Cases and controls were also matched in terms of local ancestries, and supervised and unsupervised analysis provided consistent results, reinforcing the selection of the representatives of the parental populations in the ancestry assessment.

Table 3.1. Demographical characteristics and sample size of the Colombian collections.							
Sta	ge 1	Stage 2					
Seropositive	Seronegative	Seropositive	Seronegative				
998	659	122	532				
913	592	122	512				
503 (55%)	357 (60%)	65 (53%)	315 (62%)				
62.7±16.2	49.0±17.6	64.7±11.2	50.0±15.6				
	ns. Sta Seropositive 998 913 503 (55%)	Stage 1 Seropositive Seronegative 998 659 913 592 503 (55%) 357 (60%)	Stage 1 Stage Seropositive Seropositive 998 659 122 913 592 122 503 (55%) 357 (60%) 65 (53%)				

Genotyping quality controls (QC).

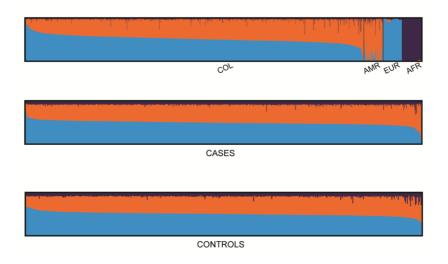


Figure 3.1. Admixture estimations (K=3) for individuals included in Stage 1 (Chagas disease cases and controls) and for AMR, EUR and AFR reference populations from 1KGP.

RESULTS

Admixture mapping results based on local ancestry estimations were not inflated because of the presence of population stratification for any of the ancestries after λ correction (λ_{AMR} =1.00, λ_{EUR} =1.00, λ_{AFR} =1.00). These results revealed genome-wide significant associations of AMR and EUR ancestries with Chagas disease in the positions of the chromosome 6 region 30,079,993-30,332,160 according to build hg19 (Figure 3.2). We observed a lack of association of AFR ancestry with the serological status. These results revealed a differential susceptibility to the infection associated with two of the ancestries in this particular region, where AMR was in higher proportion among seronegative individuals overall, therefore associated with a protective effect. The associated region is located in the MHC locus where the leading signal corresponds to rs115833233 (AMR ancestry OR=0.74, 95% CI=0.66–0.83, *P-value*= 4.53×10^{-08}), which is located in the untranslated region (UTR) of exon 5 in the Tripartite Motif Containing 40 (*TRIM40*) gene. The association of AMR ancestry was unrelated with the genotypes in that position because conditioning by the allele dosage of the rs115833233 variant did not change the results. Thus, the admixture mapping peak was not explained by the genetic variation of that SNP (Table 3.2).

Table 3.2. Joint SNP-ancestry analysis in the discoverystage.						
Factor	OR (95% CI)	p-value				
AMR ancestry	0.74 (0.66-0.83)	4.53E-08				
Allele dosage rs115833233	1.00 (0.89-1.13)	0.99				
AMR ancestry (conditioned on rs115833233)	0.73 (0.65-0.82)	2.15E-08				

Abbreviations: AMR Native American ancestry; CI confidence interval; OR odds ratio

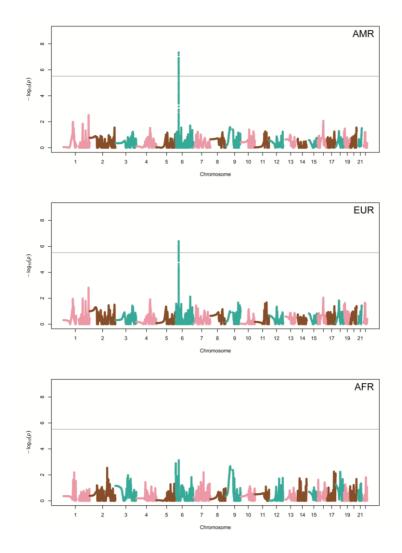


Figure 3.2. Manhattan plots of the admixture mapping results based on local ancestry estimates of Native American (AMR; up), European (EUR; middle) and African (AFR bottom). Y and X axes refer to the –log10 transformed P-values and hg19 positions in chromosomes, respectively. The horizontal line indicates the significance threshold (*P-value*= 3.07×10^{-06}).

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To identify potential variants explaining the result, imputed genotype data from the significant admixture mapping region were assessed in the stage 1 samples and in independent samples from stage 2. This analysis identified 23 variants in high LD (r²>0.8) with nominal significance and consistent direction of effects in the two stages (Table 3.3). The leading variant was rs2032134, located intergenic to the Ribonuclease P/MRP Subunit P21 (RPP21) and TRIM39 genes (OR for the C allele =0.93, 95%) CI=0.90-0.97, *P-value*=3.54×10⁻⁰⁴) (Figure 3.3). We confirmed this association was dependent on the AMR ancestry because adjusting the models by the ancestry score resulted not significant (OR for the C allele=0.99, 95% CI=0.95-1.03, P-value=0.766). In addition, we confirmed that the C allele was associated with local AMR ancestry as stratifying cases and controls by this ancestry proportion, the allele frequency was increased among carriers of local AMR ancestry in cases and controls in comparison with individuals bearing other ancestries in that position. Functional analysis showed that rs2032134 is an eOTL for different HLA members in different tissues, most significantly for the HLA-C in whole blood (P*value*=5.1×10⁻³⁵) according to eQTLGen. Evidence of long-distance interactions between rs2032134 and another member of the Tripartite Motif (TRIM) family, TRIM31, was observed in macrophages and neutrophils (229), as well as in lymphoblastoid cell lines (198). Further functional assessments for rs2032134 and its best LD-proxies are summarized in Table 3.4.

1.09 (1.03-1.15) 3.67E-03 1.07 (1.03-1.11) 3.54E-04 p-value Table 3.3. Association testing results and allele frequencies on imputed data for the admixture mapping associated region. Meta-analysis OR (95%CI) p-value OR (95%CI) cases/controls EA frequency 0.25/0.16Stage 2 1.05 (1.01-1.10) 2.57E-02 p-value OR (95%CI) (cases/controls) EA frequency 0.19/0.14Stage 1 rs2032134*T 6:30360509 CHR:BP SNP_ID(*EA)

Abbreviations: EA effect allele; BP base pair; CHR chromosome; CI confidence interval; OR odds ratio; SNP single nucleotide polymorphism.

Table 3.4. In silico functional assessment of the fine mapping and selective sweepanalyses top variants.								
Chr	SNP	Function	Nearest gene	eQTLs ^a	C-HiC genes	PheWAS		
6	rs2032134	Intergenic	RPP21 / TRIM39	HLA-C, HCG18, HLA-E, HLA-G, TRIM10	<i>TRIM31</i> in GM12878 ^b , macrophages ^c and neutrophils ^c	-		
6	rs9261440	Intergenic	TRIM31 / TRIM40	HLA-F, HLA-G, HLA-A, HLA-L, TRIM26, TRIM10	<i>TRIM10</i> in GM12878 ^b	Hematological measurement, eosinphil count		

a eGENES in whole blood

b lymphoblastoid cell line from Mifsud et al.

c Macrophages and neutrophil from Javierre et al.

Queried data bases were Open Targets Genetics, eQTLGen, HaploReg and Capture HiC Plotter

Abbreviations: Chr chromosome; C-HiC capture Hi-C; eQTL expression quantitative trait *loci* ; PheWAS phenome-wide association study.

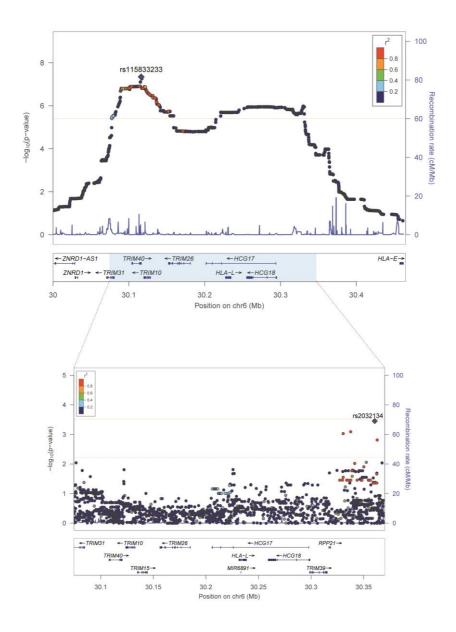


Figure 3.3. Regional plots of the results from admixture (top) and fine mapping (bottom) analyses, showing the $-\log_{10}$ transformed P-value in the y-axis and the hg19 genomic position in the x-axis. Top: Region centered on the significant variants

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for the AMR ancestry association. The horizontal line indicates the significance threshold (*P-value*= 3.07×10^{-6}). Estimated recombination rates (light blue line) are plotted on the right y-axis. Bottom: Meta-analysis results from stages 1 and 2 within the admixture mapping significant region (chr6:30,079,993-30,332,160) and its proxies (r²>0.8) indicating the associated variant with the lowest significance in the region (rs2032134). The results for the remaining single nucleotide polymorphisms (SNPs) are color coded to reflect their degree of LD with the indicated SNP based on pairwise r² values in AMR. The horizontal lines indicate the significant (solid line, *P-value*= 3.23×10^{-04}) and suggestive thresholds (broken line, *P-value*= 6.46×10^{-03}).

In order to assess adaptive signals within the significant admixture mapping region, we performed a selective sweep analysis in this region using a reference population. From this analysis, we found that the iSAFE scores were high; with the top ranking variants corresponding to positions 30,087-30,102 Mb (iSAFE score ≥ 0.331). The top 25 variants ranked by the iSAFE score mapped within (n=5; only rs28400887 was coding but synonymous) or nearby (n=20) the TRIM31 gene, with the furthest SNP located at 21.1 Kb from the gene (Figure 3.4). Given the excellent performance of iSAFE in prioritizing the most likely favored variant in 94% of the times (228), we performed a functional assessment of the variant with top scoring (rs9261440; iSAFE score=0.340) assuming it to be the variant driving the selective sweep. According to Open Targets Genetics portal, rs9261440 is an eQTL for some HLA class I genes in whole blood, where the (*P-value*=5.4×10⁻¹⁴³), most significant were HLA-F HLA-G (Pvalue=2.4×10-280) and HLA-A (P-value=2.4×10-156), and also for the TRIM family member *TRIM26* (*P-value*=7.9×10⁻⁷⁹). Additionally, HaploReg indicates that this variant is an eQTL of HLA-L and TRIM10 in whole blood (230). In agreement with this, PheWAS data from Open Targets Genetics indicates that this variant associates with hematological measurements,

particularly with the eosinophil count in the UK Biobank (*P-value*= 7.6×10^{-43}).

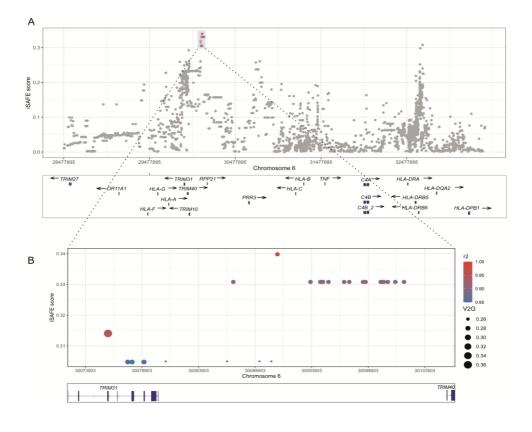


Figure 3.4. Selective sweep scan results for Native Americans (AMR) from the 1000 Genomes Project. Y and X axes represent the iSAFE scores and hg19 genomic positions, respectively. (A) iSAFE scores within the MHC region (chr6:28,477,895-33,389,603). Top ranking variants included in the significant admixture mapping region are highlighted in red (chr6:30,079,993-30,332,160). (B) Zoom in of the admixture mapping significant region, where the top 25 variants according to their iSAFE score are represented. Genetic variants are depicted in colored dots to ref lect its LD with the variant with the highest iSAFE score based on pairwise r² values in AMR population. Their dot size correlates to their composite variant to genes score (V2G) from Open Targets Genetics (<u>https://genetics.opentargets.org/</u>).

3.3. Discussion

In this chapter we describe the admixture mapping strategy to leverage varying local genomic ancestries in Colombians to identify *loci* associated with differential susceptibility to Chagas disease. Significant associations of AMR and EUR ancestries within the MHC locus with the development of the infection were found, pointing out to the role of the immune response on the disease risk. Additionally, fine mapping assessments and a selective sweep scan of this region prioritized variants with potential functional implications in the disease and highlighting how powerful this strategy is for identifying regions that had been previously overlooked in other genomic studies.

Several studies in bioarchaeological material confirmed that the American trypanosomiasis already existed long before European settlement when ancestral populations domesticated plants and animals in the process of sedentarization. This provided the vector with food availability and a more rapid domiciliation (3, 5). Host-parasite co-evolution is considered one of the most important generators of biological diversity in the genome, as is the case of selective sweeps at loci with functional role in their interaction (227). It occurs when parasites trigger host adaptations, which will lead the parasite to adapt again to this new environment in their hosts (227). This coevolution has been described for *T. cruzi* (39, 231) and, based on the observed protective association of the AMR ancestry, one can speculate that the long-term exposure of ancestral populations with larger proportions of AMR ancestry may have provided a more efficient immune response to the parasite, as previously hypothesized, with an adequate defense against pathogens that are endemic to the New World (68). This response could be responsible for parasite clearance, precluding the establishment of a serological response. This ancestry-specific effective

response against parasites and other infectious agents has been previously hypothesized to explain the enrichment of African alleles within the HLA-B locus in a Colombian population (67). Another study correlated the EUR and AFR ancestries with differential immune response to bacterial infections and hypothesized that this could be related to the differential pathogen exposure of each population and different selective pressures after the human migrations out of Africa (232). Local adaptations have shaped human genetic variation together with drift and migrations, and admixed populations are likely to have a larger number of genetic variants that have functional effects (68, 233). Therefore, it is particularly striking that the selective sweep scan within the significant admixture mapping region revealed variants associated with eosinophil counts because eosinophils have cytotoxic functions to fight parasitic infections, and along with macrophages, monocytes and neutrophils, are the innate immune cells responsible for the control of the initial infection by *T. cruzi* (234, 235). Interestingly, blood cell traits have been reported to differ among ancestries and are subjected to different selective pressures (236).

Several genes of the TRIM family mapped within the significant admixture mapping region, and the top variant was located in the *TRIM40* gene (Figure 3.3). More importantly, the variants prioritized both by the fine mapping and the selective sweep scan are also eQTLs of TRIM genes in whole blood. The TRIM family is E3 ubiquitin kinases that play an important role in immune signaling pathways, and the expansion of this multigene family suggests their key regulatory role during the immune response against pathogens (211, 237). When the pathogens are recognized by the immune system through the pattern-recognition receptors, several immune responses are initiated, including the production of interferons (IFNs), leading to the expression of TRIM proteins, among others (237). The upregulation of TRIM genes in response to *IFN-* γ has been reported in

DISCUSSION

human monocytes and macrophages (238). The expression of *IFN-* γ is induced by IL-18, which is a pro-inflammatory cytokine mainly produced by macrophages (39). Remarkably, genetic variants of the gene encoding *IL-18* have been associated with *T. cruzi* infection in previous candidate gene studies and showed suggestive association in a recently published GWAS (122, 193, 221).

The fact that the top variant from the fine mapping analysis mapped in an intergenic region, made us speculate on its potential relation with nearby genes through an in silico functional analysis. This strategy suggested the significant correlation of this variant with the *HLA-C* and *HLA-G* expression in whole blood as eQTLs. HLA genes are well-known for their role in the modulation of the immune response during T. cruzi infection (36). A previous study associated the HLA-C*03 allele with the susceptibility to the chronic cardiac form of the disease in a Venezuelan population (239). Regarding the HLA-G, several alleles from the HLA-G*03 UTR region were tested in a Brazilian population, reporting evidence of association with different clinical forms of the disease (93). Additionally, lower gene expression and plasma concentrations of this gene have been described in the chronic phase of *T. cruzi* and Plasmodium falciparum infections (240). Further long-distance chromosome interaction analysis of this variant revealed a significant interaction with the promoter of TRIM31 in macrophages, neutrophils and the lymphoblastoid cell line GM12878. This gene is a member of the TRIM family that acts as a regulator of the NLR pyrin domain-containing 3 (NLRP3) expression (241). The NLRP3 is an inflammasome component that is well-known to be activated by molecular patterns associated with pathogens. The activation of the inflammasome is crucial for the control of intracellular protozoan parasitic infections, like T. *cruzi* and the production of nitric oxide during its acute infection in mouse models (216, 242). Interestingly, TRIM31 has been also prioritized by the selective sweep scan as a locus under putative positive selection in the AMR population.

There are some major limitations in this study that need to be taken into consideration. One of those limitations relates to the identification of the causal agent of the selective signal. The local adaptive signal maps within the MHC locus, which is a well-recognized target of selection. However, the causal factor underlying the selective sweep has been identified only in few instances in humans, such as skin pigmentation (243), lactase persistence (244) and adaptation to altitude (245). Therefore, we cannot infer nor guarantee that T. cruzi is the causal factor driving this putative adaptive signal. Another limitation is the limited statistical power, especially in the fine mapping, where we only had 80% power to detect variants with an allele frequency >20% and minimum effect allele of 1.4, which is fairly larger than expected for common variants in complex traits (246). Larger sample sizes would improve the statistical power of the study. Another limitation refers to the analyzed genetic variants, as only common SNPs were assessed in the fine mapping, and structural or less frequent genetic variants underlying the admixture mapping signal of Chagas susceptibility remain unexplored. Further insights concerning these types of genetic variants may be assessed by next-generation sequencing approaches. Moreover, despite the advantages of admixture mapping studies for identifying susceptibility loci, we are not in the position to warrantee if the results are generalizable to other populations with AMR ancestry, considering as well the variability of the infectious agent throughout the American continent. Finally, the scarce representation of AMR populations in reference databases reduces the precision to provide the exact functional implication of the associated variants, given the differences in LD structure among populations as well as other potential phenotypic features in different cell types.

DISCUSSION

This chapter encompasses the results of the first admixture mapping analysis carried out in Chagas disease. This assessment allowed us to associate the AMR local ancestry at the MHC locus with protection from Chagas disease, highlighting the role of the immune response during the acute phase of the infection.

Chapter 4: Functional assessment of the GWAS associated *loci* through its influence on methylation levels

GWAS associated and suggestive variants use to map in non-coding regions of the genome, which exert their effect modulating gene expression. This modulation might be driven by the methylation of the DNA through methylation quantitative trait loci (mQTL), which are correlations among SNPs and DNA methylation levels in the context of a specific trait. Thus, in order to functionally characterize GWAS associated variants that map in regulatory regions, these multi-omics integration analyses are performed. The aim of this study was to evaluate the functional mechanism underlying previously reported region associated with chronic the Chagas cardiomyopathy by a mQTL analysis in a subset of individuals from the Bolivian population. As result, we identified mQTLs in the CCDC88B, PLAAT3 and POLA2 genes, which have been associated with cardiovascular diseases in previous studies. Interestingly, the CCDC88B was previously identified as differentially methylated and expressed in heart biopsies of chronic Chagas cardiomyopathy patients. These results, in addition to suggest novel genes that could play a role in the chronic Chagas cardiomyopathy pathogenesis, evidence the functional relevance of the previously associated loci.

4.1. Material and methods

4.1.1. Study samples and ethical considerations

All donors were recruited by the health care system from the Vall D'Hebron University Hospital, Barcelona, Spain. The study population is composed by a subset of patients from Bolivia from the previous GWAS (221). A total of 57 seropositive individuals for *T. cruzi* antigens were classified into chronic Chagas cardiomyopathy patients (cases, n=37) and asymptomatic (controls, n=20) according to the presence of cardiac abnormalities. Patients were subjected to electrocardiogram and echocardiogram tests to determinate cardiac abnormalities, while additional clinical information or complementary tests were retrieved from the medical history. Case-control sample size and demographic information are summarized in Table 4.1.

Table 4.1. Demographical characteristics and sample size.					
	CCC	ASY			
Subjects, N	37	20			
Sex (% females)	18 (48.65%)	16 (80%)			
Age (Mean±SD)	52.8±9.9	45.5±12.9			

The sample sizes refer to data passing the genotyping quality controls. Abbreviations: CCC, chronic chagas Cardiomyopathy; ASY, asymptomatic.

This study was approved by the Ethics Committee from the Vall D'Hebron University Hospital, Barcelona, Spain (PR (AMI) 297/2016). The Ethics Committees for the GWAS data was described elsewhere (221). Protocols used in the study followed the principles of the Declaration of Helsinki and all individuals included in the study signed written informed consents.

4.1.2. Data preparation

SNP genotyping and imputation

As extensively described elsewhere (221), blood DNA was isolated and genotyped using the Global Screening Array Platform (Illumina Inc., San Diego, CA, USA). After passing their corresponding QCs using the software PLINK v.1.9 (125), genotyped data was imputed with the Michigan Imputation Server (247).

After imputation, genomic information from the 2 Mb flanking region centered in the variant associated with the disease (rs2458298, chr11:63,814,813-65,814,813, according to build hg19) was extracted from a subset of individuals from the Bolivian cohort, and transformed to PLINK format. Imputed SNPs were filtered by their imputation quality metric Rsq and MAF, keeping for the analysis those that satisfied both MAF>5% and Rsq>0.3.

DNA methylation data preprocessing

Genomic DNA from blood samples from the 57 Bolivian individuals was isolated using the QIAamp MidiDNA Kit (QIAGEN, Germany) following manufacturer's recommendations. From this, methylation patterns were determined using the Infinium MethylationEPIC Bead Chip array (Illumina, Inc., San Diego, CA, USA). A total of 850,000 DNAm sites were assessed and passed to downstream data processing and normalization using the R/Bioconductor package *minfi* (248). Probes positioned in SNP positions and those with a detection *p-value*<0.01 were removed. The clustering of methylated and unmethylated probes did not show significant differences in the samples. After this, the preprocessQuantile method for normalization was performed (249) and the *M* values matrix was obtained by the log₂-transformation of the DNA methylation ratio. This matrix was utilized to compare sample groups (cardiomyopathic and asymptomatic) by an eBayes-moderated paired t test using the limma package (250). As for the genomic data, methylation data was extracted for the same 2 Mb region centered in the variant associated with the disease and *p*-value of <0.05 was considered statistically significant.

4.1.3. Methylation quantitative trait loci statistical analysis.

In order to link genetic variants to variations in DNAm, preprocessed methylation and imputed genotyped data from the significant region were integrated for the 57 individuals included in this study. The analysis of *cis*-mQTLs for cases and controls was performed using the R/Bioconductor package *MatrixEQTL* version 2.3 (251). For the correlation analyses, sex and age were used as covariates and those SNP-DNAm site pairs with a maximum distance of 1 Mb were tested. In order to assess the functionality of the locus associated with the cardiac form of Chagas disease, the integration analysis was carried out in the region of 2 Mb centered in the associated GWAS variant (rs2458298, chr11:63,814,813-65,814,813 according to build hg19). Those interactions with a FDR of <0.05 were considered as significant.

4.1.4. In silico functional analyses

An *in silico* functional analysis was performed to assess the biological consequences of the genes related with the leading significant interactions. For this, genes within the DNAm sites were annotated using the R/Bioconductor package IlluminaHumanMethylationEPICmanifest (https://bioconductor.org/packages/release/data/annotation/html/Illumi naHumanMethylationEPICmanifest.html).

In order to prioritize *cis*-mQTLs linked to the variant associated with the disease, the LD link tool (195) was used to calculate SNPs in LD (r²>0.4) with it in the admixed American population from The 1KGP. The Open Targets Genetics web tool (<u>https://genetics.opentargets.org/</u>) was used to evaluate the biological and disease implications of the genes associated with the DNAm sites from the prioritized *cis*-mQTLs.

4.2. Results

4.2.1. Identification of cis-mQTLs in the previously associated chronic Chagas cardiomyopathy locus

Genomic and methylation data from a total of 57 Chagas disease patients were integrated to determine *cis*-mQTLs. We limited our analysis to the genomic region centered in the variant previously associated with the chronic cardiac form of the disease (221). A total of 2,611 SNPs were tested against 2,647 DNAm sites in 37 chronic Chagas cardiomyopathy patients and 20 asymptomatic individuals after passing their corresponding QC criteria. We identified 6,958 significant *cis*-mQTLs (FDR<0.05) at maximal 1 Mb distance of the GWAS leading variant (Figure 4.1). These *cis*-mQTLs were composed by 2,143 unique SNPs and 152 unique DNAm sites. On average, the distance between the SNPs and DNAm sites that form the *cis*-mQTLs was ~114 kb.

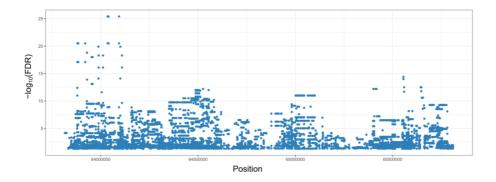


Figure 4.1. Regional scatter plot of the *cis*-mQTL results for the selected genomic region under analysis (chr11:63814813-65814813). Representation of the 6,958 *cis*-mQTLs identified in the genomic region previously associated to the disease. The x-axis corresponds to the genomic position of the SNP from each mQTL interaction; the y-axis refers to the transformed False Discovery Rate (FDR) of each signal.

First, we evaluated if the chronic Chagas cardiomyopathy associated variant and those in LD acted as *cis*-mQTLs. We identified six SNPs with moderate LD (mean r^2 =0.55) with the leading variant as *cis*-mQTLs of a DNAm site located between 200-1,500 bp of distance to the TSS (TSS1500) of the SAC3 Domain Containing 1 (*SAC3D1*) gene (Table 4.2). The risk alleles of these variants were associated with higher methylation levels. However, these changes were not significant when comparing the methylation status in asymptomatic and cardiomyopathic patients. Nevertheless we observe a highly significant positive correlation of GWAS and *cis*-mQTL *p*-values (Figure 4.2) suggesting an association of both methylation and genetic signals. The most significant *cis*-mQTL in the region, which mapped in the gene body of the Phospholipase C Beta 3 (*PLCB3*), also showed no significant correlation of the allele changes with the methylation levels, we cannot reassure that this is directly related to the disease as this alteration in

methylation levels occurs in a similar manner in asymptomatic and CCC patients.

Table 4.2. <i>cis</i> -mQTLs with SNPs in moderate LD (r ² >0.4) with the genetic variant associated to the chronic chagas cardiomyopathy.									ronic
SNP ID	SNP position	DNAm site Id ^a	DNAm position	DNAm site function ^b	DNAm site gene	FDR	Beta	R^{2c}	Distance (bp) ^d
rs11231929	11:64810771		11:64808083	TSS1500	SAC3D1	1.85E-05	0.26	0.49	2,688
rs496427	11:64838173					6.82E-05	0.26	0.46	30,090
rs947799	11:64851305	cg22690720				8.06E-05	0.26	0.45	43,222
rs475089	11:64832939	cg22690720				1.46E-04	0.26	0.44	24,856
rs498636	11:64807541					3.11E-04	0.23	0.42	-542
rs6591869	11:64802267					9.49E-04	0.22	0.39	-5,816

a DNAm site identificator

b Position of the DNAm site respecting its annotated gene. TSS1500 refers a position 1500 bp to the gene promoter.

c mQTL correlation.

d SNP-DNAm site genetic distance.

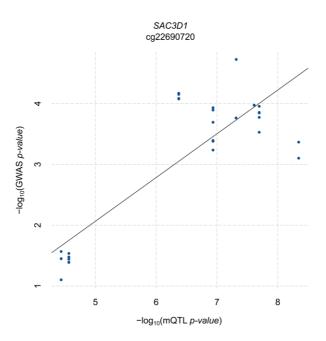


Figure 4.2. Comparison of the transformed significance levels ($-\log_{10} p$ -*values*) of the associated SNPs in the previous GWAS with those corresponding to their *cis*-mQTLs. The transformed *p*-*values* of the mQTL

and GWAS analyses were compared for those SNPs that form mQTL with the DNAm position cg22690720 located in the gene region of the *SAC3D1*.

Second, taking into account the total *cis*-mQTLs identified in the evaluated region, we found 268 interactions with significant methylation changes when comparing asymptomatic individuals and chronic Chagas cardiomyopathy patients. 89 (33%) DNAm sites were located in promotors (distance to TSS < 1500 bp), 175 (65%) in gene-bodies and 4 (1%) in intergenic regions (Table 4.3). The majority of these interactions affected a DNAm site located in the gene body of the DNA polymerase alpha 2, accessory subunit (POLA2) (Figure 4.3A) and also contain the most significant *cis*-mQTL in this locus (11:65063553–cg22229551, FDR=1.04x10-¹¹, R²=0.72). Additionally nine SNPs from these *cis*-mQTLs were nominally associated with the disease in the previous GWAS (221).

Table 4.3. <i>cis</i> -mQTLs interactions that produce differentially methylation among chronic Chagas cardiomyopathy patients and asymptomatic individuals.								
SNP ID	DNAm site ID ^a	Distance (bp) ^b	DNAm site function ^c	DNAm site gene	mQTL p-value	mQTL FDR	Beta	R ^{2d}
rs72922019	cg00022866	-104203	Gene body	CCDC88B	0.0000158	0.0189	-0.37	0.3
rs1111934	cg14293362	450196	TSS200	TM7SF2	0.0000335	0.0338	-0.3	0.28
rs12292693	cg19940438	-146	TSS1500	SPDYC	1.24E-10	0.000000775	0.67	0.55
rs12292693	cg16849481	-260	TSS1500	SPDYC	2.93E-08	0.0000943	0.47	0.45
rs484147	cg22229551	-241058	Gene body	POLA2	0.00000878	0.0118	-0.3	0.32
rs61884708	cg26579892	478276	Gene body	PLA2G16	0.0000048	0.00722	-0.6	0.33
rs507062	cg23228688	69522	Intergenic	-	0.0000201	0.0228	-0.71	0.3

a DNAm site identificator

b SNP-DNAm site genetic distance.

c Position of the DNAm site respecting its annotated gene. TSS1500 and TSS200 refers a position at 1500-200 or less than 200 bp to the gene promoter.

d mQTL correlation.

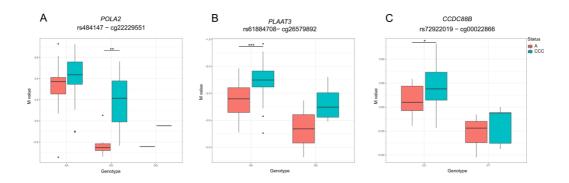


Figure 4.3. Box plots of three of the most interesting genes comparing DNAm levels between chronic Chagas cardiomyopathy patients (CCC) and asymptomatic individuals (A). Comparison of individuals' SNP genotypes and CpGs Mvalues for the most interesting mQTL genes that also showed differences in methylation patterns in cases and controls. Significant differences between genotypes are marked with symbols according to their level of significance calculated with an ANOVA test (* *p*-value≤0.05; ** *p*-value≤0.01; *** *p*-value≤0.001). Genotypes with no significant differences do not have a symbol.

The *cis*-mQTL with the most significant variation in the methylation patterns among cases and controls (*p-value*=8.09x10⁻⁰⁴), corresponded to the DNAm site within in the gene body of the Phospholipase A and acyltransferase 3 (*PLAAT3*) gene (rs61884708–cg26579892, FDR=7.22x10⁻⁰³, beta=0.12, R²=0.33, Figure 4.3B). Among these results it is interesting also to highlight the interactions with the DNAm site located in the coiled-coil domain containing 88B (*CCDC88B*) gene body (rs72922019 – cg00022866, FDR=1.89x10⁻⁰², beta=0.08, R²=0.30, Figure 4.3C). This gene was previously reported to be differentially methylated and differentially expressed when comparing biopsies from heart tissue of chronic Chagas cardiomyopathy

patients and healthy donors (252), and we were able to corroborate these findings in an independent population.

4.2.2. Functional annotation analysis

To explore functional features of the genes related to the identified DNAm sites, several databases and online tools were queried. The majority of the methylated positions in the region were located within gene bodies (Figure 4.4), and those *cis*-mQTLs that interacted with DNAm sites located in intergenic regions were removed from further functional analyses.

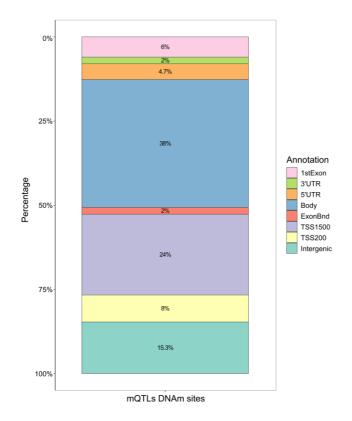


Figure 4.4. DNA methylation (DNAm) site positions respecting their nearest genes. For the 152 unique DNAm sites identified, their positions were compared with the Human Methylation EPIC manifest from Illumina and were expressed in percentages. According to their position, the DNAm sites can be located in the first exon (1st exon), in the 3' or 5' regulatory regions (3'UTR and 5'UTR), in the gene body (body), exons (exonBnd), at 1,500 or 200 to the transcription start site (TSS1500 and TSS200) or intergenic (intergenic).

According to the Open Targets Genetics database there are several association studies that have related the *POLA2* with hematological traits, such as mean corpuscular hemoglobin or red blood cell count (236, 253). One of these studies has also found an association of this locus with

cardiovascular disease (*p*-value= $6x10^{-16}$), highlighting the role of this gene in this trait (253). Additionally, different *loci* in the *CCDC88B* gene have been also associated with cardiovascular disease in the same dataset (*p*-value= $1x10^{-10}$) (253). Taking together, the *in silico* functional analysis of the resulting genes suggests the relationship of the associated region with different cardiovascular traits, supporting its relevance in the pathogenesis of the chronic Chagas cardiomyopathy.

DISCUSSION

4.3. Discussion

In this chapter we show the results of the first *cis*-mQTL analysis in Chagas disease patients from the Bolivian population, leveraging a previously reported genomic association, and its correlation with specific DNAm levels that may impact the regulation of gene expression. This assessment has provided further functional characterization of the associated locus, and has revealed significant interactions with relevant genes that have been previously related with several cardiovascular traits. As mentioned in previous chapters, the chronic cardiac form of the disease shows damage and inflammation of the myocardium as a response to the parasite, but also the microvascular system can be involved (254). Interestingly, these results revealed the regulation of gene expression of an inflammation related gene, validating previous findings, and highlighting the use of whole blood as a surrogate tissue for studying the chronic form of the disease. In addition, we identified regulatory mechanisms potentially involved on disease pathogenesis, exposing the functional implication of the previously associated *loci*.

Considering the sustained host-parasite interaction throughout human history, host genetic variation has been linked to the heterogeneous response observed to infectious agents (74). Nevertheless, functional studies are needed to determine the relevance of this genetic variation on disease pathogenesis (72). In this sense, epigenomic modifications are particularly ligated to pathological processes behind the chronic phases of infections, even when the first stages have been overcome (255, 256). Several of these modifications have been described in infectious diseases like viral (257), bacterial (258) and specifically, DNAm changes has been reported in parasitic diseases, such as Malaria, Leishmaniasis and Chagas disease (252, 259).

These results showed cis-mQTLs composed by SNPs in LD with the leading GWAS variant that were correlated with changes in the DNAm levels in the vicinity of the promoter of SAC3D1. This gene, as mentioned in chapter 2, is a transcriptional regulator of STAT5 (208), and increased phosphorylation levels of this protein have been associated with protection in human hearts (209). Among the significant interactions identified in the region, the top signal corresponded to a DNAm site located in the gene PLCB3, which encodes a member of the phospholipase C family. This protein family has been described to act in the induction of cardiac hypertrophy (260, 261). Moreover, another member of the phospholipase C family, the PLCB2, showed a significant correlation between methylation and gene expression levels in cardiac tissue, compared to healthy donors in a previous study in chronic Chagas cardiomyopathy (252). Although our results do not show significant differences on the methylation levels among asymptomatic and cardiomyopathic patients, we might speculate their potential role on the disease, based on their localization within the GWAS locus and their functional relevance.

From the total number of significant *cis*-mQTLs it is relevant to highlight those that produce substantial changes in methylation levels when comparing patients with chronic Chagas cardiomyopathy and asymptomatic individuals, since these interactions could have a greater implication on the regulation of gene expression in the pathogenesis of the disease. From these, the *CCDC88B* gene is of special interest because of its involvement in inflammatory conditions (262). This gene showed higher methylation levels within the gene body in cardiomyopathy patients, which is related with tissue-dependent regulation of gene expression (263). The *CCDC88B* has been associated with the most severe form of Malaria, cerebral malaria, where patients suffer inflammation of the brain tissue (264). In this study, the authors tested the role of the *CCDC88B* during this affection in a murine

DISCUSSION

model and showed that is a potential regulator of T cells function. This gene plays an important role in the inflammatory and host responses during parasitic infections (264, 265). This is of substantial interest since inflammation of the cardiac tissue also occurs during chronic Chagas cardiomyopathy (254, 266). In addition, CCDC88B has been also described in a murine model as relevant in the immune function of dendritic cells (262). In this sense, one subtype of dendritic cells, the tolerogenic dendritic cells, have been recently proposed to have therapeutic properties in a mouse model of chronic Chagas cardiomyopathy, given their role in the reduction of inflammation and fibrosis that contributes to reduce disease progression (44, 267). Interestingly, the CCDC88B was previously reported to suffer changes on its methylation and expression levels when cardiac tissues from patients with chronic Chagas cardiomyopathy and healthy donors were compared, showing also a correlation among both methylation an expression variations (252). The identification of this gene in whole blood from independent study samples highlights its role in the pathogenesis of the chronic Chagas cardiomyopathy, as well as emphasizes the use of whole blood as a surrogate tissue that allows the detection of disease-related molecules in a less invasive way. Thus, despite the high tissue specificity of the DNAm patterns (268), surrogate tissues as whole blood could be widely used, as in other diseases with high correlation of results (269).

The *PLAAT3* gene, also known as *PLA2G16*, showed the most significant variation in methylation levels, being also the chronic Chagas cardiomyopathy patients' methylation levels higher in this locus in comparison with asymptomatic individuals. This gene encodes an adipose-specific phospholipase whose expression has been reported to be reduced in patients with peripheral artery disease, which is a common affection in patients with coronary artery disease (270, 271). Thus, as mentioned for *PLCB3*, the role of the phospholipases during *T. cruzi* infection has been

widely studied as part of the lipid metabolism (272). Nevertheless, although this phospholipase seems to be related with other cardiovascular traits and exhibit higher expression in heart tissue, its potential relation with Chagas disease remains to be determined.

Finally, the most significant *cis*-mQTLs that produced variation in the methylation levels among cases and controls corresponded to DNAm sites located in the gene body of the *POLA2*. As for the other genes, *POLA2* showed higher methylation levels in chronic Chagas cardiomyopathy patients. Our *in* silico functional analyses related this gene with different hematological measurements in other association studies, in addition to be related with cardiovascular disease (236, 253). One of the hematological traits associated with POLA2 is the mean corpuscular volume, which has been previously related with heart failure (273). In this sense, the relation of this gene with the pathogenesis of the chronic Chagas cardiomyopathy remains unclear, as gene expression information would be necessary to verify the direction of the reported methylation patterns. However, the identification of genes previously related to several cardiovascular traits and cardiac damage reinforces the role of the region under study previously associated with the disease (221). Therefore, our results suggest the DNAm as a possible driver of the variation observed in the differential development of the chronic Chagas cardiomyopathy.

This study has some limitations that are worth mentioning here. First, as the variation explained by *cis*-mQTLs is higher in blood (274), we focused our work in these interactions. Nevertheless, the addition of genetic distal interactions (i.e. *trans*-mQTLs) to the analysis could be also informative about other genes and pathways relevant for the disease. Interestingly, we used whole blood as a surrogate tissue and were able to validate the involvement of a previously reported gene in the cardiomyopathy. Thus,

DISCUSSION

further studies focused on the cardiac tissue would be interesting in order to validate the rest of the results. In order to confirm the methylation patterns associated with the identified interactions, validation assays such as pyrosequencing techniques would be of high interest in further studies focused in methylation patters. Another limitation is the analysis of variants with a MAF higher than 5%, which limits the number of low-frequency and rare variants evaluated in the region that might contribute to the disease risk as occurs in other complex diseases (246). Increasing the sample size may further allow the correlation of low frequency alleles with DNAm levels, and increase the statistical power in the differential methylation assessment. Finally, replication analyses would help to elucidate the generalization of these results to other Latin American populations, where the genetic variability of the parasite may contribute to different epigenetic modifications, as it has been previously reported for other parasitic infections (275).

This chapter describes the results of the first *cis*-mQTLs analysis carried out in chronic Chagas cardiomyopathy in patients from the Bolivian population. Numerous significant interactions between genetic variants and DNAm levels were identified within the only genomic region significantly associated with the disease. Several of these interactions showed also differential methylation in chronic Chagas cardiomyopathy patients, and validated previous findings. These results provide novel functional insights in the genetic background and gene regulation underlying the pathogenesis of the most severe form of Chagas disease.

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GENERAL DISCUSSION

Genomic approaches to unravel the pathogenesis of Chagas disease

In the present thesis, different aspects of the host genetics in *T. cruzi* infection have been addressed given the nature of this complex infectious disease. Several type of genetic studies, candidate and GWAS, were described here highlighting the role of the genetic variation at different levels in Chagas disease. It is known that in addition to genomics, information provided form different types of data, like epigenomics as well as the molecular interaction among them, give a better understanding of the molecular complexity behind a disease (276).

The genetic associations described in our candidate genes studies pointed out, and confirmed, the involvement of the host genetic background in the immune response to the infection by the parasite. In addition, with the purpose of analyzing a great number of the genetic variability of the genome, a genome-wide analysis and meta-analysis of several Latin American populations were carried out, being the largest GWAS in Chagas disease to date. This analysis identified a genomic associated *locus* on chromosome 11, whose functional assessment was related with numerous genes previously associated with Chagas disease and other cardiac symptomatology such as SAC3D1, BATF2 or SNX15. The absence of association at genomic level in one of the regions most related to the immune response, such as the HLA genes, may lead to the exclusion of this region as part of the genes related to the genetic contribution to the disease. Nevertheless, the complementary analysis of the local ancestry proportions in the Colombian population and their relationship to infection revealed the association of European and Native American ancestries with increased risk and protection to Chagas disease respectively. This association was observed in a region of chromosome 6 where the HLA and TRIM family genes are located. In addition, the selective sweep assessment highlighted several adaptive signals in this region functionally related with different HLA genes. In this way, we were able to observe an association in a region related to several

HLA genes, which highlights the relevance of these *loci* in the susceptibility to the infection, which could be driven by the selective pressure on this region. This leads us to believe that further studies focused on the analysis of SNPs located in the HLA region, as well as classical HLA alleles and amino acid positions, will allow us to have better knowledge of the role of this region in Chagas disease, as has been assessed in other infections such as Malaria, Tuberculosis or Leishmania (277-279). Occasionally, the absence of association in genomic regions such as this one may be due to the type of variant evaluated. Thus, structural variations such as CNVs may explain part of the variability in this region for Chagas disease (280). This structural variation has been assessed in other infectious diseases such as Malaria and HIV-1 infection (281). In the last one, it could be observed how the CNVs of the *KIR3DS1* gene, in combination with a specific HLA-B allele, showed a relation with the HIV-1 viral load and leaded to a slower progression to AIDS, being an example of the implication of the HLA in the disease identified by structural variations (281). Another important region where the effects of structural variations are key in the immune response is the complement system. This is an important component of the innate immune response that is also part of the first line of defense in infections by microorganisms (282). CNVs located in this region involve changes in the enzymatic cascade whose relationship with predisposition to autoimmune diseases is known (283). In the case of Chagas disease, it is known that during the acute phase there is an interaction between the parasite and different components of the complement system that leads to their activation in order to internalize the parasite. However, the enzymatic cascade is interrupted allowing the entry of trypanomastigotes into the cells and originating the chronic phase of the disease (282). Translating this interaction into the genetic context, one study evaluated the relationship among the disease with SNPs located in the gene region of one complement elements, the CR1 protein (284). In this study,

GENERAL DISCUSSION

associations with the disease were found in several SNPs located in the gene coding for the CR1 protein, which the authors relate to an amino acid change that may affect the functionality of the protein and leads to the parasite persistence common in chronic Chagas disease patients (284). In this regard, although previous studies have evaluated the interaction of *T. cruzi* with the different complement elements, future analyses of structural variation in these characteristic regions would provide insight into the molecular mechanisms that lead to the activation or inactivation of this system during the acute and chronic phases of the disease.

Despite the efforts to apply genetic studies to underrepresented populations such as the Latin American ones, the limited number of references makes it difficult to assign functional outcomes to these populations, so they must be hypothesized. In addition, standard genomic references may leave out uncommon variants present in other populations. This was demonstrated in a sample of individuals of African descent, where it was observed that a considerable amount of sequences from the African genome were missing from the reference genome (285). These differences in genetic architecture between populations are also evident in the replicability of genetic studies in underrepresented populations.

In this sense, during 2018 the GWAS carried out in Latin American populations represent slightly more than 5% (286). This percentage will be maintained during 2021, being also comparable to the number of individuals included in this type of studies (Figure 6) (287). Thus, these events suggest that more population specific reference genomes are needed for population-based studies of human genetics.

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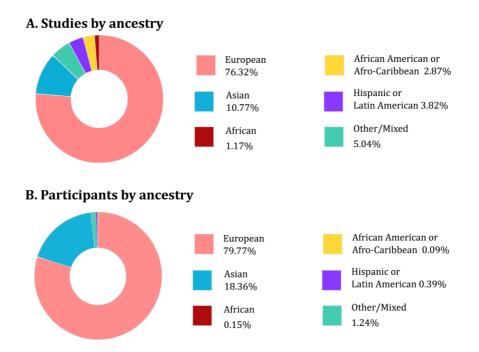


Figure 6. Number of genome-wide association studies and participants included in them in the discovery stages classified by ancestry during 2021 (Modified from https://gwasdiversitymonitor.com/).

The underrepresentation of Latin American populations in the reference panels is also reflected in functional analyses. In the case of epigenomics, as in their integration with other omics sciences, it is frequent also to use public databases in which, in most cases, the reference data used belong to the European population. This was especially relevant when it came to functionally characterizing the mQTLs identified in the associated region of chromosome 11. Regarding this, an important aspect to take into account in methylation studies is the obtaining of relevant tissue for the disease under study due to, like other epigenetic modifications, it is considered tissuespecific (288). This is remarkable in the case of tissues with limited access

such as the cardiac one and, as mentioned in previous chapters, the use of surrogate markers from tissues that are more accessible allow us to assess the epigenetic background behind a disease. Here, whole blood has been demonstrated to be a good candidate reporter of the influence of methylation changes in genomic variation in the context of the chronic Chagas cardiomyopathy. This aspect is also relevant not only considering the access to these tissues but also in terms of sample collection. The access to certain endemic regions and the cultural component in occasions can influence in the representation of ethnic minority groups (289). In this sense, it is important to take into consideration that some endemic areas have a difficult access and the risk that supposes the sample collection of an infectious disease. This limits the obtaining of larger sample sizes required for genome-wide association studies, which affects their statistical power. Therefore, it is noteworthy the observation of an association at the genomic level in the meta-analysis, despite the sample was limited and heterogeneous. Here, another important aspect that was necessary to take into consideration was the genetic heterogeneity present among and within Latin American populations. This variety is driven by the different admixture proportions of the European, Native American and African ancestries along and within the different Latin American countries (290).

The use of different host-based genetic approaches allowed us to better understand the pathogenesis of this disease and, ultimately, can lead to improve treatment, outcomes and the development of new drugs and vaccines. In this regard, a first step lies in the assessment of existing drugs that can be reused to combat the disease, which is known as drug repurposing. This approach allows optimizing the treatments development reducing time and cost of their production for which it is necessary to take into account the patient's genetic background in the context of the disease to be treated (291). In the case of parasitic diseases, these drugs should be

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focused on controlling parasitemia, which is difficult to generalize among different parasites. Despite this, recent studies have evaluated the trypanocidal activity of some drugs such as nimesulide, an anti-inflammatory previously used in cancer and diabetes (292). In this sense, another interesting approach in the case of this disease would be focused on the chronic form and the inflammation that occurs in it, for example through the use of drugs used in other myocarditis. Therefore, the identification of common genetic biomarkers is an excellent tool in the knowledge of individual susceptibility to the disease and to be able to apply preventive measures.

FUTURE DIRECTIONS

Genomic approaches to unravel the pathogenesis of Chagas disease

The results widely discussed here are a novel contribution to the host genetic field in Chagas disease, although as mentioned, the assessment of several types of biomarkers is crucial for the knowledge of the molecular pathways involved. In this sense, our group recently obtained methylation data of individuals with chronic Chagas cardiomyopathy classified according to the severity of the disease using the Kuschnir scale. These methylation patterns were analyzed in 57 individuals belonging to the Bolivian cohort that also were used in the mQTL analysis described in chapter 4. Preliminary analyses of these data showed differences in the methylation patterns at a differentially methylated positions (DMPs) level when comparing the groups according to their severity (Figure 7). In this way, two main clusters can be observed showing a tendency to hypomethylation and hypermethylation, respectively, depending on the severity of the disease. These results are indicative of the existence of genetic variation in chronic Chagas cardiomyopathy patients beyond the variation in the DNA sequence.

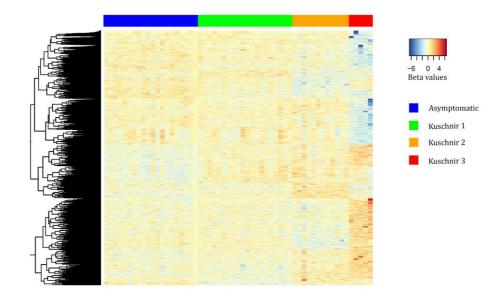


Figure 7. Heatmap of the DMPs (*p-value*<0.05) of 57 seropositive individuals from the bolivian cohort classified according to the Kuschnir scale. Hypermethylated CpGs

have higher beta values (red), while hypomethylated positions display lower beta values (blue). The rows repressent each CpG positon and the columns reffers the individuals.

Additionally, the analysis of differentially methylated regions (DMRs) in these patients is also ongoing. Preliminary results when comparing asymptomatic individuals with severe chronic Chagas cardiomyopathy patients (Kuschnir 2 and 3) pointed out two significant DMRs nearby the genes *DUSP22* and *VTRNA2-1*, which encode a phosphatase and a RNA polymerase III transcript. In this sense, protein phosphorylation is important for cellular signaling and, interestingly, the implication of the deregulation of these proteins in diseases such as cardiomyopathies or inflammatory conditions is well-known (293). Further *in silico* functional analyses of significant DMPs and DMRs will be needed in order to characterize these regions and their relation with the disease. Therefore, the comparison of these results with previous similar analyses carried out in the disease would clarify common molecular pathways when using different tissues.

Taking into account the relationship between DNA methylation and gene expression, the analysis of RNA-sequencing (RNA-seq) data will allows us to identify which genes are being expressed in a pathological process. In the case of Chagas disease, the latest study of the gene expression patterns was carried out in combination with a methylation analysis in a Brazilian population (252). In this sense, differences between Latin American populations can also be extrapolated to other genetic studies and, as a consequence, the expression patterns may also vary between populations (294, 295). Thus, the analysis of gene expression patterns in the Bolivian population would allow us to assess the differences and similarities of the pathways involved in the chronic cardiac form of the disease among different populations. To this end, the gene expression profile of part of the Bolivian cohort included in the previous genetic analyses has been determined using the RNA-seq technology, and the association analysis with the disease and its integration with genetic and methylation data are being carried out. After this, further steps of this work will be focused in the functional characterization of these results and their enrichment in specific molecular pathways that may be related to the pathogenesis of the disease, which is also currently being carried out.

In general, these methodologies provide a wide amount of information about the biology of the disease that will help in the development of predictive models of disease risk, such as genomic risk score (GRS) studies, for the identification of markers of disease prognosis as well as therapeutic targets, especially focused on chronic Chagas cardiomyopathy. With the efforts behind this doctoral thesis we hope to have contributed to the knowledge of this neglected tropical disease. Genomic approaches to unravel the pathogenesis of Chagas disease

CONCLUSIONS

1. The candidate gene studies performed confirmed the role of *IL17A* and *IL18*, but not *IL6* polymorphisms in Chagas disease. These results expose the activation of the host immune response at genetic level during initial stages of Chagas disease.

2. The largest GWAS in Chagas disease, and the meta-analysis of different Latin American populations, revealed a novel genome-wide statistically significant association with the development of chronic Chagas cardiomyopathy. This association was located in an intronic region of the *NAALADL1* gene and the vicinity of the *SAC3D1* and *SNX15* genes.

3. The functional characterization of the GWAS-identified novel locus using reference databases, revealed its relation with hypertension and blood pressure traits in PheWAS data. In addition, the variant was identified as eQTL in cardiac tissue for the *SNX15* gene, and further chromatin structure database showed its interaction with the promoter of *BATF2*.

4. The admixture mapping analysis revealed a protective association of the local Native American ancestry with the susceptibility to *T. cruzi* infection. This signal is located in the major histocompatibility complex region, where the TRIM and HLA gene families stood out for their functionality in the immune response against pathogens.

5. The mQTL analysis within the GWAS associated locus, allowed the identification of genes related to hematological and cardiac traits, as well as previously related to chronic Chagas cardiomyopathy, such as *POLA2*, *PLAAT3* and *CCDC88B*.

Genomic approaches to unravel the pathogenesis of Chagas disease

CONCLUSIONES

1. Los estudios de genes candidatos confirmaron el papel de las variantes de los genes *IL17A* e *IL18* en la enfermedad de Chagas. Sin embargo, la asociación con la enfermedad del gen *IL6* fue descartada. Estos resultados resaltaron la activación de la respuesta inmune del hospedador a nivel genético durante la fase inicial de la enfermedad.

2. El GWAS y meta-análisis de mayor tamaño muestral llevado a cabo hasta la fecha en la enfermedad de Chagas, reveló una asociación estadísticamente significativa a nivel genómico con el desarrollo de la cardiomiopatía Chagásica crónica. Esta asociación está localizada en un intrón del gen *NAALADL1* y en las inmediaciones de los genes *SAC3D1* y *SNX15*.

3. La caracterización funcional de la variante asociada en el GWAS reveló su relación con rasgos cardiovasculares. Además, fue identificada como eQTL en tejido cardíaco para el gen *SNX15* en bases de datos de referencia y que interacciona con el promotor de *BATF2* en datos de estructura de la cromatina.

4. El estudio de mapeo por mezcla reveló una asociación de protección de la ascendencia local Nativo Americana con la susceptibilidad a la infección por *T. cruzi.* Esta señal se localiza en la región del complejo mayor de histocompatibilidad, donde destaca la funcionalidad de las familias de genes TRIM y HLA en la respuesta inmune frente a patógenos.

5. El estudio de mQTLs en el locus asociado en el GWAS permitió la identificación de genes relacionados con caracteres hematológicos, cardíacos y previamente relacionados con la cardiomiopatía chagásica crónica, como el *POLA2, PLAAT3* y *CCDC88B*.

Genomic approaches to unravel the pathogenesis of Chagas disease

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