



Genetic Variants Linked to HIV-1 Exposure in Cameroonian Children and Adolescents and Their Association to Non-Communicable Diseases

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Abstract

Background: Children with HIV continue to be at a higher risk of non-communicable diseases (NCDs), even with improvements in the treatment of prenatal exposure to HIV-1 infection. This study aimed to identify genetic variations associated with HIV-1 and their potential influence on the risk of NCDs in offspring exposed during pregnancy.

Methods: We conduct a genome-wide association study (GWAS) on a cohort of 689 participants from Cameroon, employing a case-control design and a genomic epidemiology approach. Of these, 421 were HIV-infected and 268 were not. Single nucleotide polymorphisms (SNPs) were performed using Illumina BeadChips.

Results: Twenty SNPs in all were found, and the majority of the alterations were silent or synonymous. Those linked to HIV rapid progression include rs1422884 (GRIA1, Chr5) and exm665143 (TRPV6, Chr7) 7.9×10^{-6} , with $OR = 0.01$, rs4915847 (C1orf87, Chr1), rs7578597, and exm189601 (THADA, Chr 2). The NCD-related SNPs rs1693687 (Chr10) 8.4×10^{-7} with $OD=0.04$ and rs7175885 (Chr15) showed significant associations with slow progression in HIV-infected individuals: THADA (rs7578597, exm189601, Chr2) was linked to type 2 diabetes and metabolic disorders. Both TENM2 (rs10057680, Chr5) and RBFOX1 (rs9302839, Chr16) have been associated with neurocognitive impairment. rs1693687 (Chr10) and rs10231785 (Chr7) were linked to cardiovascular diseases (CVDs). PHLPP1 (rs692916, Chr18) and ZNF208 (rs10426971, Chr19) have been connected to cancer risk and immune response.

Conclusion: The genetic variations linked to protective effects against HIV-1's rapid development as well as other variants with possible genetic resistance mechanisms are highlighted in this work. Additionally, several SNPs overlap with risk factors for NCDs, indicating that there may be common genetic processes. These results highlight the value of precision medicine strategies in HIV care.

Keywords: HIV-1, GWAS, SNPs, metabolic disorders, cardiovascular diseases, non-communicable diseases (NCDs)

Introduction

Complex host-parasite interactions are influenced by both extrinsic and intrinsic factors that affect host adaptation and susceptibility. HIV-1 infection remains a significant global health burden despite global efforts to eradicate it, especially in sub-Saharan Africa, where environmental factors and genetic

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diversity contribute to the variability in treatment outcomes and disease progression [1]. Although antiretroviral therapy (ART) has greatly raised survival rates, persons living with HIV (PLWHIV) are still particularly concerned about long-term consequences including non-communicable diseases (NCDs) [2, 3]. Managing HIV/AIDS in low- and middle-income countries (LMICs) such as Cameroon is more difficult. According to a study, some regions/areas of the country have observed stable HIV prevalence over time at around 5% despite widespread ARV use [4]. While it is difficult to determine why the prevalence stays high within communities, several factors could be speculated, including host genetic variables known to be important in determining HIV susceptibility, disease progression, and ART response [5-7]. However, despite Africa's enormous genetic variety, the majority of genomic research has been done on people in North America and Europe, with very few studies being carried out in African locales [8,9]. The phenotypic heterogeneity seen in exposed people is one of the main obstacles to comprehending HIV-1 infection. Following infection, some people develop quickly, slowly, or do not progress for a long time [6], while others resist HIV-1 infection even after several exposures [10]. A complex interaction between viral, environmental, and host genetic variables influences this diversity in virus acquisition, disease development, and responsiveness to treatment [11]. Through candidate gene techniques and genome-wide association studies (GWAS), host genetic research has played a significant role in identifying genetic variations linked to HIV susceptibility and disease progression over the years [12]. The majority of this research, however, has concentrated on people with European ancestry, which limits their relevance to communities in Africa, which have the highest genetic variety in the world [8, 13, 14]. Due to population-specific genetic variants and structural variations, genome-wide association studies (GWAS) have found multiple SNPs (single nucleotide polymorphisms) linked to HIV-1 progression; however, these findings frequently do not replicate in African populations [13, 15]. Furthermore, while GWAS has mostly focused on HIV-related immune response genes, accumulating data suggests that many of these variations also overlap with genes connected to NCDs, including cancer, metabolic diseases, and neurocognitive deficits [16]. Given the increasing burden of NCDs among PLWHIV, it is crucial to examine genetic risk factors that may contribute to both HIV progression and NCD susceptibility.

Only candidate gene studies, which concentrate on genes linked to HIV progression, resistance, and mother-to-child transmission, have been conducted in Cameroon on children and adolescents exposed to HIV-1 [17-19]. Expanding research to find genetic variants linked to HIV infection as well as metabolic, cardiovascular, and neurodevelopmental disorders that may develop as a result of prolonged exposure to

antiretroviral therapy (ART) and host genetic predisposition is imperative given the rising burden of NCDs among PLWHIV [2, 19]. Although important single nucleotide polymorphisms (SNPs) linked to diabetes, cancer, and other NCDs have been identified by studies conducted in developed countries, little is known about these findings in African contexts [16,20]. Non-communicable diseases (NCDs) are recognized as one of the primary causes of morbidity and mortality worldwide, particularly for people living with HIV (PLWHIV) [3]. By performing a genome-wide association study (GWAS) on Black African children and adolescents who were exposed to HIV-1 at birth and those who were not, this study seeks to close this gap by examining the genetic variability of the participants and finding correlations with variants linked to both HIV-1 susceptibility and NCD risk. This study is an important step in understanding host genetic factors that contribute to HIV-related health disparities and the increasing burden of non-communicable diseases in PLWHIV, especially considering the high genetic diversity of African populations and the limited funding available for large-scale genomic research in LMICs. In order to contribute to precision medicine techniques that are specifically designed for African populations, this research aims to offer new insights into the genetic drivers of disease susceptibility and progression by utilizing genome-wide methodologies [15, 21].

Materials and Methods

Study design, setting and population

This study was done at the Chantal Biya International Reference Center and the BiOmics unit of the Pasteur Institute of Paris. A total of 689 participants were recruited after obtaining a proxy consent form, signed by parents before enrollment. The classification was done according to their profile at the time of the inclusion, focusing on their records files and their biological parameters such as their CD4⁺ T cells count and viral load (VL) values, together with clinical parameters and their status regarding the treatment as described by Dambaya and colleagues (2019). Among the participants, 268 were HIV-1 negative and 421 were HIV-1 positive. We obtained ethical clearance from the Cameroon National Ethics Committee for research of Human health (CNERSH) under N°2013/11/374/L/CNERSH/SP/ and an administrative authorization from the Ministry of Public Health of Cameroon under the N0631-05.14/AAR/MINSANTE/SG/DROS/CRC/ZAMC. As well, this study was conducted following the Declaration of Helsinki.

Data sources, Analytical processes and measurement

Blood collection: five millilitres of blood were collected from each participant included in the study at each appointment. Part of whole blood was used for CD4 count and the rest was centrifuged at 1,000 g for 10 minutes, plasma was aliquoted and stored at -20°C for further analyses.

Determination of CD4 counts: CD4 cell counts were quantified using a FACScalibur flow cytometer (Becton-Dickinson, Immunocytometry System, San Jose, CA, USA).

Determination of viral load: HIV-1 viral load was determined from plasma by Abbott m2000rt Real-Time HIV-1 assay (Abbott Molecular Diagnostics, Wiesbaden, Germany), using the 200 and 600 µl of plasma protocol with a detection limit of 160 copies/ml (1.8log) and 40 copies/ml (1.6log) respectively.

DNA extraction: this was done according to the QIamp DNA Mini and the Blood Mini Handbook, from 200µl of whole human blood or from dried blood spot (DBS), three puncher-out circles from a DBS were used.

Quality control and quantification of DNA samples: we used microarray-based genome-wide association studies (GWAS) to identify disease associations. The experiment was done at the time of sample collection and was undertaken at the Genotype Eukaryotes unit of the Pasteur Institute of Paris (France) actually named Biomic's unit. This platform at the time of this analysis used SNP genotyping on Illumina beadchips. The genotyping chip technology was a standardized and automated (using kits and robots) protocol. From the 285 included subjects, we selected a total of 120 extracted DNA of good quality from the different involved groups of participants. Genotyping clustering was performed using the Illumina Genome studio program.

Quantification was done on the extracted DNA from the whole blood or from the DBS. The PicoGreen assay was useful for the quantification of the amount of double-stranded DNA (dsDNA) that was present in a given sample in an ultra-sensitive fluorescent nucleic acid stain because a specific amount of DNA (at least 75 ng/ul) was required for SNP identification. The measured amount of dsDNA was used as a marker of the success of the genotyping process. The samples were then separated on 1% agarose gel, for checking the DNAs' integrity (no smearing) and to verify the requested amount of DNA needed. Quality control of the raw data was done using the software PLINK which requires a computer server Illumina BeadStudio v3.1 according to several parameters. First of all the genotypes were assigned using a classification provided by Illumina, generated on an African population. We performed a common measure of relatedness (or duplication) between pairs of samples, based on identity by descent (IBD). When a link between samples was observed, one of the two samples was excluded for further analysis. In IBD estimation coefficient greater than 0.20 may suggest relatedness, duplicates, or sample mixture. In the same way, gender identity was also checked at this stage to confirm that self-reported gender is consistent with the observed X and Y chromosomes. After classification, individuals with a call rate (percentage of genotyped SNPs per individual) of less

than 97% were deleted. On the other hand, SNPs with a "call frequency" (percentage of individuals genotyped by SNP) of less than 99% were re-classified. Finally, SNPs with a call frequency of less than 98% (i.e. a missing data rate >2%) were excluded. All of these steps ensure reliable genotyping data with little missing data. Finally, only 96 DNA were genotyped using Illumina Human core exome analysis beadchips. The deviation of the Hardy-Weinberg equilibrium was evaluated for each SNP in each group by comparing the distribution of observed genotype frequencies and theoretical genotypic frequencies using an exact statistical test [21]. The realization of multiple tests has been taken into account by applying the corrections of Bonferroni, one of the simplest approaches to correct for multiple testing.

Genome-wide association analysis

We research the association between HIV status (positive infant vs negative participants) and HIV progression and, the association between the different groups of infected individuals with logistic regression, using PLINK.

Statistical methods

In this GWAS study, we used bead chips containing more than 240,000 SNPs, so the statistical significance of the SNP association has been set at 2.08×10^{-7} . In order to identify additional signals while controlling the risk of false discoveries, we also calculated the FDR (False-Discovery Rate) q-value for each p-value: the q-value estimates the proportion of false positives below a threshold of p-value [22]. The local estimation FDR has been applied in this study with a threshold of 25%. For each SNP passing the statistical threshold of significance, the quality control was individually re-verified. Then, for each identified signal, we verified that the allelic frequency in an HIV-positive population was similar to that of the population control, to confirm that the observed association accurately reflects progression to AIDS and not HIV-1 infection. For the identification of population stratification, case and control genotypes were analyzed using the EIGENSTRAT method which allows detecting and correcting the stratification of our study population. This method, based on a principal component analysis, allows to modelling of the ancestral differences according to continuous axes of variation.

Results

Summary Participant enrolment, Classification, and Genomic Sample Processed in the CBIRC HIV Pediatric Cohort (2010–2015) involved in the GWAS analysis

A total of 689 infants were recruited at the Chantal Biya International Reference Center (CBIRC), from October 2010 to April 2015. Among them, 268(38.9%) were HIV-negative

cases, 68 (25.4%) were HIV-exposed uninfected (HEU), and 200 (74.6%) were HIV non-exposed uninfected (HNEU). The group of positive participants was composed of 15 enrolled LTNP (3.6%), 80 SP (19%) and 42 RP (9.9%). The rest, 284 (67.5%) participants were excluded because of not respect any criteria of classification of these three different groups. From the 285 extracted genomic materials to be genotyped

in the genomic laboratory of the Pasteur Institute of Paris, only 120 passed the initial quality control filtering. Five individuals have been removed from the study for gender misclassification. Finally, principal component association (PCA) was performed on 96 samples to assess population stratification. After this correction, these participants were then considered for further analysis (Figure 1).

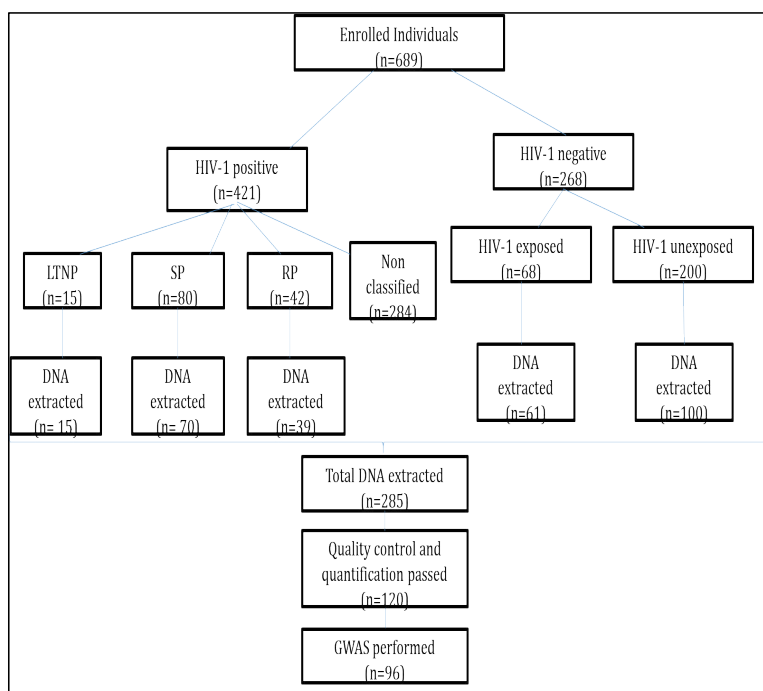


Figure 1: Participant Recruitment, Classification, and Genomic Sample Processing in the CBIRC HIV Pediatric Cohort (2010–2015) involved in the GWAS analysis

Table 1: Socio-demographic and immune-virologic characteristics of participants involved in the GWAS analysis

Groups (number; % male)	Mean age	Mean CD4 cells (%)	Mean Viral load copies (Log)
LTNP (9; 11%)	13.44	477 ^a (21)	145097 ^b (4.06)
SP (33; 45%)	11.33	608 ^a (22)	44048 ^b (3.6)
RP (21; 36%)	3.04	1243 ^a (23)	38467 ^b (4.2)
HEU (27; 38%)	5.53	/	/
HNEU (6; 50%)	7.5	/	/

LTNP: long term non progressor; SP: slow progressors; RP: rapid progressors; HEU: HIV exposed uninfected; HNEU: HIV non-exposed uninfected; a:p-value for mean CD4 cells (p=0.02); b:p-value for mean Viral load copies (p =0.02)

Socio-demographic and immune-virologic characteristics of participants involved in the GWAS analysis

Socio-demographic and immune-virologic characteristics of participants involved in the GWAS analysis are resume in table 1. From the 15 enrolled LTNP, 9 past the last quality control analysis with 33 SP and 21 RP. Inside the group

of HIV-1 negative the majority, 82% were HIV-1 exposed uninfected and the rest was non-exposed. Most of the participants were originated from the Center South and West regions of the country.

Sample relatedness and principal component analysis (ancestry plot)

Principle component analysis plot was used to identify axes of genetic variation for each individual. The relatedness of samples is shown in figure 2, each point represents an individual (figure 3), and no patients and no control were too different from others in terms of their genotypes (black crosses and triangles are our HIV positive and controls samples, next to their genotypic ancestry in green in the figure 3). The differences appear clearly when the genetic comparison is between our study population versus Asiatic (Japanese and Chinese) shown in purple and located on the upper left of figure 3, with the European population in red located on the lower left part. All the samples (cases and controls) completely match with African ancestry population in green located on the upper right part of figure 3.

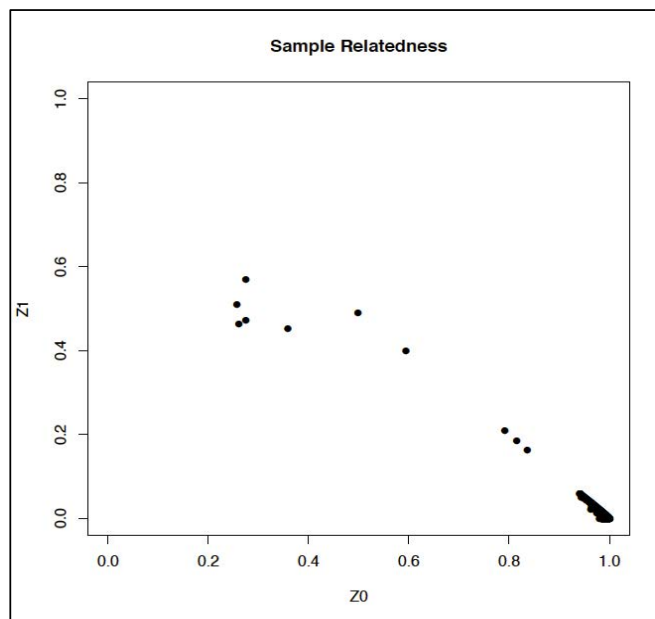


Figure 2: Relatedness in the participant's ancestry sample. A plot of the probability of sharing 1 allele ($z1$) identical by descent against the probability of sharing zero allele ($z0$).

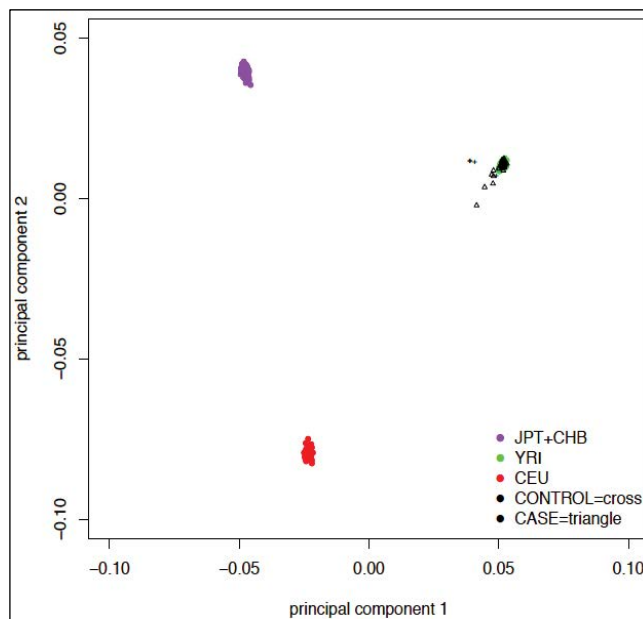


Figure 3: Principal component ancestry of the overall perinatally infected and uninfected children and adolescents compared to Japanese and Chinese (JPT+CHB) European (CEU); and African (YRI)

Table 2: List of the best SNP signal results and genes involved in the overall study population of infected and uninfected participants

SNPs name	CHR	Allele	Gene symbol	Characteristic	Mutation
rs1693687	10	[T/C]	/		/
rs9302839	16	[A/G]	RBFOX1		Silent
rs7175885	15	[T/C]	/		/
rs6140412	20	[T/C]	/		/
rs6136995	20	[T/G]	C20orf26		Silent
rs10057680	5	[A/G]	TENM2		Silent
rs10804512	3	[A/G]	/		/
rs6718327	2	[T/C]	/		/
rs1422884	5	[T/C]	GRIA1		Silent
exm665143	7	[A/G]	TRPV6	Exon variant	Missense_M681T
rs4915847	1	[T/G]	C1orf87		Silent
rs7578597	2	[T/C]	THADA	Exon variant	Missense_T1187A
exm189601	2	[T/C]	THADA	Exon variant	Missense_T1187A
rs12214839	6	[A/G]	FUT9		Silent
rs9270986	6	[A/C]	/		/
rs5926263	X	[A/G]	LOC100873065		Silent
rs10231785	7	[A/G]			
rs40428	12	[T/C]	/		/
rs692916	18	[T/C]	PHLPP1		Silent
rs10426971	19	[A/G]	ZNF208		Silent

Legend: CHR= chromosome number

SNPs with best p values obtained in the overall study population

We identified in overall 20 SNPs in our study population which have the best p values as shown in table 2. Eleven genes have been identified with only three missense mutations in the overall study population. The majority of the SNPs were located on chromosome 2 (with 4 SNPs), followed by 20 and 6 with 2 SNPs respectively. Most of the significant SNPs (45% and 18%) were involved in cancer and type 2 diabetes respectively, the other SNPs were associated with only one disease. Table 3 summarizes the different SNPs identified in NCDs in our study group that were reported in other studies.

Table 3: Identifier SNPs and their localization on genes involved in some NCDs

Gene symbol	SNPs	Role/Disease involved
RBFOX1 (RNA Binding Fox-1 Homolog 1)	rs9302839	Involved in Psychiatric disorders, Autism, Epilepsy and Colorectal Cancer. [23-25]
C20orf26 (chromosome 20 open reading frame 26)	rs6136995	Oxidoreductase activities are involved in innate and adaptative immunity [26]
TENM2 (Teneurin Transmembrane Protein 2)	rs10057680	involved in cancer tissue, and chronic kidney disease [27]
GRIA1 (glutamate ionotropic receptor AMPA type subunit 1)	rs1422884	Intellectual development disorder [28]
TRPV6 (Transient Receptor Potential cation channel subfamily V member 6)	exm665143	Hyperparathyroidism, Breast cancer [29-31]
C1orf87 (chromosome 1 open reading frame 87)	rs4915847	Involved in somatic mutations in cancer, spinal muscular atrophy [32]
THADA (Thyroid Adenoma Associated)	exm189601	Associated with type 2 diabetes and polycystic ovary syndrome [33-35]
FUT9 (Fucosyltransferase 9)	rs12214839	Associated with Susceptibility to placental malaria infection, Intellectual Developmental Disorder, and Colorectal cancer [36,37]
LOC100873065	rs5926263	Cancer, autism and autoimmune disease [38,39]
PHLPP1 (PH domain and Leucine-rich repeat Protein Phosphatase)	rs692916	Involved in multiple types of cancer, obesity and type 2 diabetes [40]
ZNF208 (Zinc finger protein)	rs10426971	Bind DNA and regulate gene transcription [41,42]

SNPs identified in perinatally HIV-1 infected participants: impact on disease progression

Comparison between long-term non-progressors (LTNP) versus Rapid Progressors (RP)

After accessing the genotype magnitude of SNPs between LTNP and RP using a Manhattan plot (Figure 4), showing many SNPs with $\log_{10} p$ values $< 10^{-4}$, only 2 SNPs (rs1422884 and exm665143) emerge with an unadjusted p-value of 7.9×10^{-6} (Table 4). The odd ratio of $r=0.01[0.0001-0.2010]$ and the false discovery rate (FDR_{BH}) was 81%. Adjusted value with Bonferroni correction indicates a p-value of 4.5×10^{-5} (upper horizontal blue line) located in chromosomes 5 and 7 respectively.

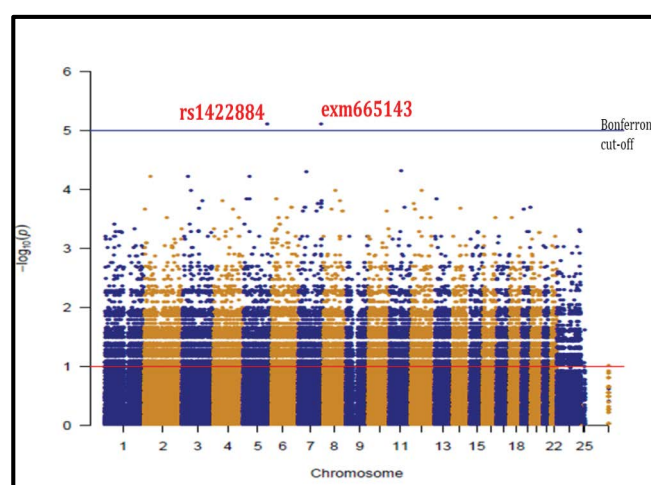


Figure 4: Allelic association plotted as $-\log_{10} p$ -values for a genome-wide association study in LTNP and RP HIV-infected children and adolescents for 22 autosomes and the X chromosome. Each dot corresponds to an SNP, with its chromosomal position (x-axis)

Comparison between long-term non-progressor (LTNP) and Slow Progressor (SP)

When we compared LTNP and SP as shown in red colour in figure 5, 3 SNPs appeared above $p < 10^{-5}$ rs1693687, rs9302839 and rs7175885. The best statistical significances were observed with rs1693687 ($p=8.4 \times 10^{-7}$) and rs9302839 ($p=9.8 \times 10^{-6}$) located respectively on chromosomes 10 and 16. As shown in table 5 the odds ratios were not significant. The FDR-BH of only rs1693687 was below the threshold of 25% and above for the 2 other SNPs (86%).

Comparison between Rapid Progressor (RP) and Slow Progressor (SP)

By comparing RP and SP (Table 6) and considering their unadjusted p values, 8 SNPs showed a signal with a p-value between 0.00024 and 10^{-6} , 3 of them located on chromosomes 1, and 2 and could be involved in rapid disease progression

Table 4: Association analysis results between LTNP and RP

CHR	Gene	SNP	BP	A1	F_A	F_U	A2	CHISQ	p-value	OR
										FDR-BH
5	GRIA 1	rs1422884	153009489	T	0.05	0.727	C	19.95	7.90E-06	0.01
										0.81
7	TPV6	exm665143	142569596	A	0.05	0.727	G	19.95	7.90E-06	0.01
										0.81

CHR: chromosome; SNP: single nucleotide polymorphism; BP: base-pair location; A1 minor allele; A2 major allele; F_A & F_U frequency of the minor allele in LTNP and RP respectively; CHISQ: Chi-square probability value; Nominal P unadjusted asymptotic probability value; OR odds ratio; the level of significance is $10E^{-7}$. FDR_BH = False Discovery Rate Benjamin-Hochberg; G= Guanine; C= Cytosine; A=Adenine; T=Thymine

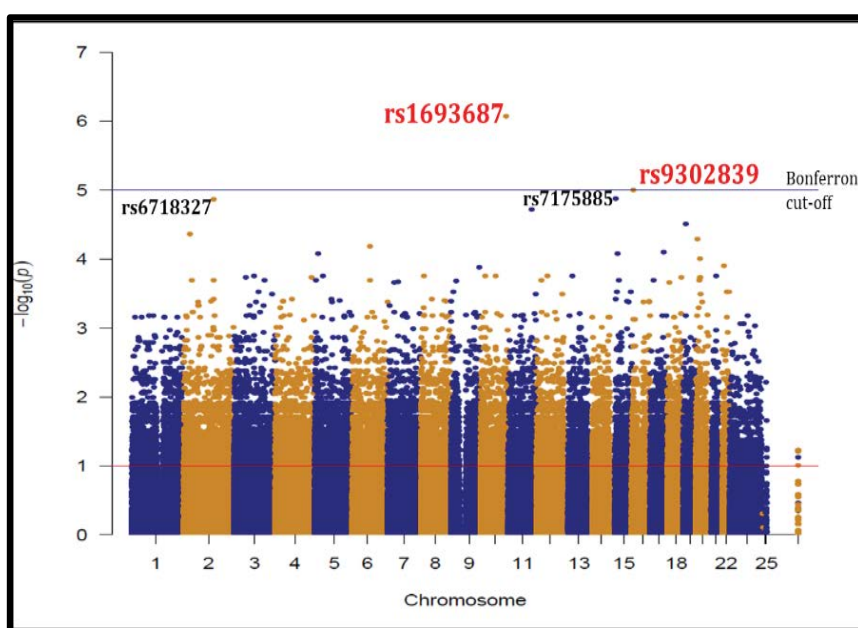


Figure 5: Association plotted as $-\log_{10} p$ -values for a genome-wide association study in LTNP and SP HIV-infected children and adolescents for 22 autosomes and the X chromosome. Each dot corresponds to an SNP, with its chromosomal position (x-axis)

Table 5: Association analysis results for SNPs identified in LTNP and SP

CHR	Gene	SNP	BP	A1	F_A	F_U	A2	CHISQ	p-value	OR
										FDR-BH
10	/	rs1693687	123438113	T	0.105	0.727	C	24.26	8.4×10^{-7}	0.044
										0.24
16	RBFOX1	rs9302839	7187784	A	0.026	0.5	G	19.54	9.8×10^{-6}	0.027
										0.86
15	/	rs7175885	25055302	C	0.05	0.4	T	13.46	1.3×10^{-5}	0.08
										0.86

CHR: chromosome; SNP: single nucleotide polymorphism; BP: base-pair location; A1 minor allele; A2 major allele; F_A & F_U frequency of the minor allele in RP and SP respectively; CHISQ: Chi-square probability value; Nominal P unadjusted asymptotic probability value; OR odds ratio; the level of significance is $10E^{-7}$. FDR_BH = False Discovery Rate Benjamin-Hochberg; G= Guanine; C= Cytosine; T=Thymine

if we consider the odds ratio value as shown in table 6. The most significant SNPs are rs4915847 located on the C1orf87 gene, exm189601 and rs75785997 on the gene THADA. By calculating the FDR-BH of the 3 SNPs, all of them were above the threshold (36%).

SNPs identified in HIV-1 perinatally infected participants: impact on disease acquisition

We identified 7 genome-wide SNPs with unadjusted

p-values not greater than 10^{-6} log when comparing HIV-1 infected participants (LTNP, SP, and RP) versus negative controls (HEU). Only 2 SNPs, rs40428 and rs692916 of the PHLPP1 gene located respectively on chromosomes 12 and 18 have a p-value of 10^{-6} log without significant odd ratio as shown in table 7. The unadjusted p-value gives a value of 6.6×10^{-6} and 9.6×10^{-6} , respectively for rs40428 and rs692916. The FDR-BH was above the threshold for both of them.

Table 6: Association analysis results for SNPs identified in rapid progressors and slow progressors participants

CHR	Gene	SNP	BP	A1	F_A	F_U	A2	CHISQ	p-value	OR FDR-BH
1	C1orf87	rs4915847	60461958	G	0.65	0.078	T	21.39	3.7×10^{-6}	21.67 0.36
2	THADA	exm189601	43732823	C	0.6	0.052	T	21.44	3.7×10^{-6}	27 0.36
2	THADA	rs7578597	43732823	C	0.6	0.052	T	21.44	3.7×10^{-6}	27 0.36

CHR: chromosome; SNP: single nucleotide polymorphism; BP: base-pair location; A1 minor allele; A2 major allele; F_A & F_U frequency of the minor allele in RP and SP respectively; CHISQ: Chi-square probability value; Nominal P unadjusted asymptotic probability value; OR odds ratio; FDR_BH = False Discovery Rate Benjamin-Hochberg; the level of significance is $10E^{-7}$. G= Guanine; C= Cytosine; T=Thymine

Table 7: SNPs potentially associated with HIV-1 virus acquisition

CHR	SNP	A1	Frequency LTNP	A2	Frequency HEU	CHISQ	p-value	OR FDR-BH
12	rs40428	T	0	C	0.6154	20.31	6.6×10^{-6}	0 0.84
			SP		HEU			
18	rs692916	A	0.2105	G	0.7692	19.58	9.6×10^{-6}	0.08 0.86

Legend: CHR=chromosome; OR= odds ratio and frequency is given for the A1 allele where $OR > 1$ indicates a protective effect; FDR_BH = False Discovery Rate Benjamin-Hochberg; CHISQ= Chi-square; G= Guanine; C= Cytosine; A=Adenine; T=Thymine

Discussion

In this study, we identified in overall 20 SNPs with the best signals, with some potentially associated with some profiles and/or clinical status of participants. Limited studies have observed these SNPs in such a pediatric context related to the use of genome-wide association studies (GWAS). Missense or non-synonymous mutations which are characterized by changes in protein sequence, were few represented compared to the majority of the identified SNPs who were synonymous. Those synonymous SNPs also termed silent mutations [43] occur at the non-coding region of the gene [44] and naturally do not affect the gene product. However, they could also be found in the coding region because each amino acid is coded by more than one codon. In most cases, they were previously not considered important. However, recent studies show that synonymous SNPs affect gene function and expression

by changing the expression of neighbouring genes. Many studies have demonstrated silent mutation could also affect the function of the cell by altering the gene's expression and regulation [45]. Some research reveals that they can have disease-driving effects, this impact of silent mutation has been demonstrated in some NCDs such as cancer development [46-48]. Thus, missense or silent SNPs observed in our study population need to be more investigated.

Concerning SNPs potentially involved in disease progression only 2 SNPs with p value of 7.9×10^{-6} , emerged when comparing LTNP to RP. The same comparison done between RP versus SP brings out 3 other SNPs with p values not greater than 10^{-6} with a higher odd ratio. These five SNPs were more represented in the RP participants and therefore need to be more investigate in order to effectively characterize their role in rapid disease progression. To the best of our knowledge, these SNPs with the best signals

have never been identified in GWAS studies regarding HIV disease progression. These new signals such as those found by other researchers will help to better understand host genetic variability in the progression of the disease [49-51]. On the contrary, only 2 SNPs also require further research related to slow disease progression as they have presented the best p values when comparing LTNP to SP. Nevertheless, this finding is controversial since those 2 SNPs were more distributed inside the SP group instead of LTNP. These results can be attributed to our poor sample size especially referring to the LTNP group of participants. The same explanation can be made referring to the poor number of SNPs with a very low odd ratio, observed when assessing the SNPs involved in HIV-1 disease acquisition. Overall, when considering the significant signal ($p=2.08 \times 10^{-7}$), needed to consider the significant impact of an SNP in this study, we observed that only one SNP has a p-value above this threshold. Moreover, by assessing the adjusted p values and using false discovery values, no SNP passed this significant signal. Referring to the fact that a wide significance threshold is usually set in the region between $p < 10^{-7}$ and $p < 10^{-8}$ [52-55], we can realize that our GWAS results globally are not enough statistically significant to make a clear conclusion. This poor genome-wide significance can be justified by many reasons, among them, the low sample size mentioned above. It was difficult for us due to the lack of technical and financial reasons, to ascertain and analyze more samples in both cases and controls in order to obtain adequate power for a typical genome-wide association study. In case samples, the number of LTNP analyzed was not enough compared to the sample size usually involved in genome-wide association studies all over the world. Regarding the number of SNPs which did not pass the quality control, it could be attributed to the genotyping tools especially the beadchips used in our analysis, in addition to the high genetic variability of the African population. African populations are genetically more diverse than European and Asian populations [56-58], the fundamental reason has always been the challenge of how to develop an appropriate methodology for GWAS in Africa in order to search for significant associations for a better exploitation of the results [59,60]. This methodology was economically challenging for us, and our experience just explained one of the reasons why GWAS is very rare in African countries. If some GWAS have been already conducted in some African countries, they are still rare and this has to be improved because GWAS studies might provide important insights into the development of more effective therapeutics public health interventions, and even vaccines. This is by looking for the molecular mechanisms involved in the resistance and susceptibility to disease. To the best of our knowledge, no GWAS study has ever been conducted in Cameroon, especially in a pediatric setting of HIV infection and this explorative GWAS study can serve as an indicator.

Our poor sample size may also have limited our ability to detect statistically high significant associations in some regions of interest, in particular for small effects. However, this study provides some new insights into the genetics of HIV-MTCT and aims to facilitate future genetic studies for this phenotype. Apart from the SNP identified on FUT9 which is associated both with placental malaria infection [35] and colon cancer [36], the observed signals were mostly associated with non-communicable diseases, such as cancer, epilepsy, type 2 diabetes, hyperparathyroidism, intellectual development disorder, and spinal muscular atrophy. These findings corroborate those of Auslander et al. (2017) who predict a variant seen in FUT9, which catalyzes the biosynthesis of Ley glycolipids, as a driver of advanced-stage colon cancer. The variant rs9302839 of RBFOX1 found in this study like those previously reported by other authors, are causal factors of psychiatric disorders, autism, epilepsy and colorectal cancer [22-24]. Overall the majority of variants with the best signals found on TENM2, C1orf87, PHLPP1 and TRPV6 genes were associated with different types of cancer [26,28,29,31,39,]. In HIV-1 infection tat gene have been associated with neurogenerative disorder (HAND) by acting on brain cells and neurons, even in the absence of active HIV-1 replication [61-65]. These results bring out the importance of the exploration of the whole human genome in order to anticipate and ameliorate the follow-up of the patients.

Conclusions

In this study, we focused on children and adolescents who were exposed to HIV-1 by conducting a GWAS in a Cameroonian population. Our findings offer new insights into host genetic determinants impacting HIV disease progression and the potential risk of NCDs in PLWHIV. Twenty SNPs with the strongest associated signals were found by our analysis; many of these had not been previously documented in pediatric HIV investigations. The majority of SNPs were synonymous (silent) mutations. Five SNPs were also found to be strongly linked to the progression of HIV disease; rapid progressors (RP) had the largest proportion of these SNPs when compared to slow progressors (SP) and long-term non-progressors (LTNP). Furthermore, two SNPs were linked to slow illness development. Notably, several SNPs identified in our study, including variants in FUT9, RBFOX1, THADA, TRPV6, and TENM2 have been linked in the past to a number of NCDs, including epilepsy, hyperparathyroidism, type 2 diabetes, intellectual developmental abnormalities, and many cancers. These results highlight the significance of incorporating genomic insights into HIV care regimens by indicating that some genetic variations may have a dual function in HIV progression and NCD vulnerability.

Recommendation

This work emphasizes how urgently larger-scale genetic research in African populations is needed to further precision medicine strategies for HIV treatment. It will be essential to comprehend how genetic variants affect HIV progression and NCD risk in order to create tailored therapies and enhance the long-term health of people living with HIV in sub-Saharan Africa.

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Limitations of the Study

The limited financial facility received did not give rise to analysing more samples as a genetic association study required in order to draw reliable conclusions.

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Conflict of interest

The authors declare no conflict of interest.

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