Association between ten-eleven methylcytosine dioxygenase 2 genetic variation and viral load in people with HIV

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Introduction: Identifying genetic factors that influence HIV-pathogenesis is critical for understanding disease pathways. Previous studies have suggested a role for the human gene ten-eleven methylcytosine dioxygenase 2 (TET2) in modulating HIV-pathogenesis.

Methods: We assessed whether genetic variation in TET2 was associated with markers of HIV-pathogenesis using both gene level and single nucleotide polymorphism (SNP) level association in 8512 HIV-positive persons across five clinical trial cohorts.

Results: Variation at both the gene and SNP-level of TET2 was found to be associated with levels of HIV viral load (HIV-VL) consistently in the two cohorts that recruited antiretroviral-naïve participants. The SNPs occurred in two clusters of high linkage disequilibrium (LD), one associated with high HIV-VL and the other low HIV-VL, and were predominantly found in Black participants.

Conclusion: Genetic variation in TET2 was associated with HIV-VL in two large antiretroviral therapy (ART)-naive clinical trial cohorts. The role of TET2 in HIV-pathogenesis warrants further investigation.

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AIDS 2023, 37:379-387

Keywords: HIV, ten-eleven methylcytosine dioxygenase 2, HIV viral load, genetics, transcriptional regulation

Introduction

HIV-1 pathogenesis is a multifaceted interaction between host and virus that results in varying degrees of immunedeficiency, high levels of HIV viral load (HIV-VL), chronic inflammation, and coagulopathy. In previous studies, variations in all aspects of these responses have been linked with increased risk of adverse clinical outcomes in people with HIV (PWH) [1–11]. A number of recent studies have linked variation in the host genome

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Received: 4 June 2022; revised: 29 September 2022; accepted: 2 November 2022.

DOI:10.1097/QAD.00000000003427

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with a variety of markers of HIV progression and disease [12-18]. These genetic association studies present a unique opportunity to explore potential contributors to HIV progression and disease by studying the impact of genetic variation in known or suspected host HIV enhancing or resistance factors on relevant biomarkers.

In this study, we focus on Ten-eleven methylcytosine dioxygenase 2 (TET2). TET2 encodes a protein that is involved in locus-specific DNA de-methylation of previously methylated and silenced genes through the oxidation of 5-methylcytosine (5mC) [19,20]. Loss of TET2 function is implicated in the development of both haematological malignancies and solid tumours in people without HIV (reviewed in [21]). However, there is a developing literature associating this gene with viral replication, and in particular retroviral replication, as well as viral associated malignancy. TET2 function has been linked to both Epstein-Barr virus (EBV) and Human Tcell leukaemia virus type-1 (HTLV-1) induced malignancy [22-28], while TET2 also appears to be critical for the regulation of endogenous retroviral elements [29-31]. In the context of HIV, recent work has suggested that the HIV-protein VPR selectively degrades TET2, thereby enhancing IL-6 and IFITM3 and promoting viral replication in monocyte-macrophage lineage cells [32,33], while mutations in TET2 are relatively common in HIV positive persons with clonal haematopoiesis [34]. Additionally, a key upstream regulator of TET2 (IDH1) has previously been implicated in HIV acquisition in molecular work and clinical cohorts [35]. Finally, from our own recent study, we observed associations between singal nucleotide polymorphisms (SNPs) in the TET2 gene with HIV-VL (via GWAS). Although the SNPs of interest were below the 5% minor allele frequency (MAF) cut-off used in that study [36], this signal, combined with the literature associating TET2 function with retroviral replication and viral malignancy, encouraged us to investigate this gene and its upstream regulators further.

We therefore hypothesized that TET2 is a critical regulator of HIV-pathogenesis, and that genetic variation in this gene and upstream regulators will impact gene function and therefore the specific pathogenic process it is involved in. However, as it was not abundantly clear from the molecular data how TET2 was involved in HIVpathogenesis (both inflammatory and viral load mechanisms were proposed [32,33]), we sought to perform an exploratory targeted genetic association study to determine what, if any, role TET2 has in HIV-pathogenesis. Biomarkers tested include markers of disease progression and immune dysfunction (HIV-VL, CD4⁺ T-cell count, CD8⁺ T-cell count and CD4/CD8 ratio) as well as markers of inflammation and coagulopathy (interleukin 6 [IL-6], C-reactive protein [CRP], D-dimer). We conducted this targeted genetic association study on TET2 and two regulators of TET2 function - IDH1 and IDH2 [37]. Associations were assessed at both the gene and SNP-level using five independent trials from the International Network for Strategic Initiatives in Global HIV Trials (INSIGHT) network.

Materials and methods

Study population

Our study population included HIV-positive adults from five clinical trials: START (NCT00867048) [40], (NCT00027352), ESPRIT SMART [41] [42] (NCT00004978), STALWART [43] (NCT00110812) and FIRST [44] (NCT00000922). The trials were approved by the institutional review board or ethics committee at each contributing centre. Participants analysed in this study were those who provided written informed consent for the trials and for genetic analyses. This was only a subset of the main trial and therefore the number of participants analysed in this study is less than presented in the main trial report. Details on the individual trials can be found in the primary papers and supplemental methods, Supplemental Digital Content, http://links.lww.com/QAD/C708.

Genotyping and quality control

Participants were genotyped using a custom content Affymetrix Axiom SNP array, consisting of 770 558 probes, enriched with markers related to immune dysfunction. The Ensembl Gene database, assembly hg19/GRCh37, was used to annotate genes within a 5 kb window of each variant. Genotyping was performed at Advanced BioMedical Laboratories, using standardized quality assurance procedures. As part of SNP-level QC, probes were removed if any of the following applied: probe was duplicate, multiallelic or nonautosomal, missingness >0.03, reproducibility less than 0.90, MAF <1% or deviation from the Hardy-Weinberg equilibrium with $P < 1 \times 10^{-6}$. At sample-level QC, participants were excluded if there was a sex mismatch, duplication or missingness. In order to capture potentially relevant TET2 phenotypes present in related individuals, participants were not excluded due to cryptic relatedness or outlying heterozygosity. After sample and SNP QC SNPs that mapped to TET2 as well as IDH1 and IDH2, $(\pm 5 \text{ kb})$ were selected for further analysis. These genes included TET2 as well as IDH1 and IDH2, two regulators of TET2 function [37].

Statistical analyses

Association of TET2 pathway genetic variation with markers of HIV-pathogenesis

Gene level associations were estimated using the SKAT-O method [45–47]. To control for population stratification, we calculated eigenvectors of the genetic data using EIGENSTRAT [48]. Eigenvectors were calculated independently for each trial cohort. Eigenvectors were calculated using all SNPs (not just TET2 pathway SNPs) that passed QC and were present above 5% MAF. The first

four eigenvectors and sex were included as explanatory variables in the SKAT-O model.

We investigated associations of TET2 SNPs with study entry measurements of HIV-VL, IL-6, hs-CRP, D-dimer, CD4⁺ T-cell count, CD8⁺ T-cell count, CD4/CD8 ratio and CD4 nadir, using separate linear regression models for each biomarker (response) and SNP (predictor) combination. Associations were assessed using an additive genetic model. Associations with HIV-VL were only assessed in ART-naive cohorts (START, FIRST and STALWART). In order to ensure an approximately normal distribution of the biomarkers, HIV-VL underwent log₁₀ transformation, IL-6, hs-CRP and D-dimer underwent log₂-transformation, whereas CD4⁺ T-cell count, CD8⁺ T-cell count, CD4/CD8 ratio and CD4 nadir were not transformed. The first four eigenvectors and sex were included as explanatory variables in the regression models. Associations were estimated separately in each of the five cohorts. To control for inflation of Type I error due to multiple testing, we used the Benjamini-Hochberg FDR method to limit the false discovery rate to at least 5% for each biomarker and cohort [49].

We conducted several sensitivity analyses on the SNPs that were associated with HIV-VL in the START and FIRST studies. These are described in supplemental methods, Supplemental Digital Content, http://links.lww.com/QAD/C708.

To assess whether alternate alleles were more frequent among Black compared with non-Black participants, we used a two-proportions *z*-test. To determine whether SNPs occurred in clusters of high linkage disequilibrium (LD), we ranked the SNPs associated with HIV-VL in both START and FIRST by effect size and performed pairwise LD analyses using LDheatmap [50]. The LDheatmaps were made independently for the START and FIRST cohorts and included all participants from the respective cohorts.

All statistical analyses were performed using R version 3.5.1 [51]. Figures were generated using R or GraphPad Prism.

Results

Characteristics of participants in the INSIGHT genetic cohorts

A total of 8512 participants from the five cohorts were included in this study. Baseline characteristics are summarized in Table 1 and distributions of the biomarkers tested are shown in Figures 1–5, Supplemental Digital Content, http://links.lww.com/QAD/C708. Participant characteristics varied substantially between studies. The START (early HIV infection with preserved immune parameters) and FIRST (advanced

untreated HIV infection) trials were the only studies that enrolled exclusively antiretroviral therapy (ART)-naive participants. Participants were predominantly White, with a large percentage of Black participants in the START, SMART and FIRST studies, 23, 38, and 57%, respectively. SMART and FIRST enrolled predominantly in the United States, whereas the other studies had wider enrolment profiles (including Europe, Asia, Africa, Australia, South America). The SCREE plots for the eigenvectors calculated to control for this population stratification are presented in Figure 6, Supplemental Digital Content, http://links.lww.com/QAD/C708.

SNP quality control and minor allele frequency filtering

Eight hundred and eighty-eight SNPs spanning the TET2 (796), IDH1 (7) and IDH2 (85) genes were present in the GeneCHIP array. After SNP quality control (QC) and cohort specific MAF filtering, a total of 292 SNPs were included in START genetic association analyses, 345 in FIRST, 317 in SMART, 262 in ESPRIT and 276 in STALWART.

Association of genetic variation in the ten-eleven methylcytosine dioxygenase 2 pathway with markers of HIV-pathogenesis

First, the associations between genes of the TET2 pathway and markers of interest were assessed using the SKAT-O method. TET2 was associated with levels of HIV-VL at study entry in the two trials that included only ART-naive participants, FIRST (P=0.0005) and START (P=0.0001), but not any of the other cohorts (Table 2). TET2 was also associated with CD4⁺ T-cell count in the START study (P=0.016) (Table 2). TET2 was not associated with any other marker in any of the five studies. SKAT-O did not reveal any significant associations with either IDH1 or IDH2 (data not shown).

Next, we estimated associations of individual SNPs with each of the biomarkers, separately for each of the five cohorts. In this analysis, 15 SNPs were significantly (q < 0.05) associated with HIV-VL in both the START and FIRST cohorts (Fig. 1d, e; Table 3). Additionally, we found 13 and 8 variants, respectively, that were associated with HIV-VL in either START or FIRST, but not in both studies. The SNPs that were significantly associated with HIV-VL in only one of the two studies showed a similar effect size and direction in the other study. All SNPs associated with HIV-VL were located within the TET2 gene, except for a single SNP located in IDH1 (rs12694101). SNPs that were associated with baseline HIV-VL in the START study were also associated with longitudinal VL levels while participants remained off ART (Figure 7, Supplemental Digital Content, http:// links.lww.com/QAD/C708). The alternate allele of the SNPs associated with HIV-VL were predominantly found in PWH of Black race (Tables 2-5, Supplemental Digital Content, http://links.lww.com/QAD/C708) across a

Table 1.	Study	entry	demographics	for	the t	five	cohorts.
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	START	FIRST	ESPRIT	SMART	STALWART
Participants $(n =)$	2546	544	2891	2283	244
Age (years): median (IQR)	36 (29, 45)	38 (32, 44)	40 (34, 46)	44 (38, 50)	36 (30, 44)
Female (%)	20.1	20.2	18	26	18
Race					
Asian (%)	1	<1	12	1	27
Black (%)	23	57	9	38	6
White/other (%)	76	43	79	60	68
CD4 ⁺ cell count (cells/µl): median (IQR)	651 (585, 759)	181 (43, 345)	451 (368, 582)	572 (455, 773)	419 (358, 516)
HIV RNA level (log ₁₀ copies/ml): median (IOR)	4.17 (3.54, 4.66)	5.09 (4.53, 5.54)	1.70 (1.70, 2.60)	2.60 (1.70, 3.45)	4.38 (3.86, 4.79)
CD8 ⁺ T-cell count (cells/µl): median (IOR)	1062 (790, 1431)	n/a	n/a	n/a	n/a
Nadir CD4 ⁺ T-cell count (cells/µl): median (IOR)	545 (473, 642)	n/a	n/a	n/a	n/a
CD4/CD8 ratio: median (IOR)	0.62 (0.46, 0.84)	n/a	n/a	n/a	n/a
IL-6 (pg/ml): median (IOR)	1.47 (1.02, 2.21)	n/a	1.90 (1.30, 2.80)	1.91 (1.19, 3.23)	n/a
CRP (µg/ml): median (IQR)	1.82 (0.77, 4.15)	n/a	1.49 (0.69, 3.21)	1.79 (0.76, 4.51)	n/a
D-dimer (µg/ml): median (IQR)	0.31 (0.22, 0.47)	n/a	0.26 (0.18, 0.37)	0.22 (0.13, 0.40)	n/a
On ART at study entry (%)	0	0	100	79.5	0
ART-naive (%)	100	100	0	6.0	78
Geographical region ^a					
U.S. (%)	18	100	25	80.4	10
Europe/Australia/Israel (%)	49	0	49	11.2	34
South America/Mexico (%)	20	0	14	5.5	30
Asia (%)	0	0	11	0.4	25
Africa (%)	13	0	1	2.5	<1%

CRP, C-reactive protein; IL-6, interleukin 6; IQR, interquartile range. ^aCountry/region of residence.

variety of geographic regions of enrolment (Tables 2 and 3, Supplemental Digital Content, http://links.lww.com/QAD/C708).

The results of the sensitivity analyses for SNPs significantly associated with HIV-VL in START and FIRST were consistent with the main models (Tables 6 and 7, Supplemental Digital Content, http://links.lww.com/QAD/C708 and Figures 8–10, Supplemental Digital Content, http://links.lww.com/QAD/C708). For the subset analyses, 578 and 309 participants were assessed in the Black only analysis, for START and FIRST, respectively, 2444 participants were assessed in the analysis that excluded participants exhibiting cryptic and outlying heterozygosity, from the START study only. 2365 participants were assessed in the analysis that excluded recently infected participants, from the START study only.

We also identified six SNPs that were associated with IL-6 levels at study entry in the ESPRIT cohort (Fig. 1a) and one in the START cohort (Fig. 1e). However, these SNPs were not the same across the different cohorts (Table 1, Supplemental Digital Content, http://links.lww.com/ QAD/C708). Associations with other markers of interest were not observed in any of the cohorts.

Linkage disequilibrium of SNPs associated with HIV-viral load

Among the 36 SNPs associated with HIV-VL at study entry in the START and FIRST cohorts, 14 were associated with lower viral load, while 22 were associated with a higher HIV-VL. These SNPs form two clusters of high linkage disequilibrium (LD), one containing the SNPs that were associated with high

Outcome	START	FIRST	ESPRIT	SMART	STALWART
CD4 ⁺ T-cell count	0.023	0.868	0.126	0.391	0.604
CD4/CD8 ratio	0.843	N/A	N/A	N/A	N/A
CD4 nadir	0.430	N/A	N/A	N/A	N/A
CD8 ⁺ T-cell count	1.000	N/A	N/A	N/A	N/A
HIV-VL	0.000152	0.00067	Not assessed ^a	Not assessed ^a	0.458
IL-6	0.062	N/A	0.071	0.431	N/A
CRP	0.806	N/A	0.059	0.470	N/A
D-dimer	0.905	N/A	0.244	0.791	N/A

CRP, C-reactive protein; IL-6, interleukin 6; TET2, ten-eleven methylcytosine dioxygenase 2; VL, viral load.

^aAs ESPRIT and SMART did not recruit ART-naive individuals, HIV-VL was not assessed in these cohorts.



Fig. 1. Associations between SNPs and markers of HIV pathogenesis. *Q*-values were $-\log_{10}$ transformed and plotted on the y-axis; higher values denote smaller *q*-values, and stronger evidence for the presence of an association. SNPs are grouped by associations to a specific marker of interest and plotted on the x-axis. Each dot represents a single SNP. Associations are considered significant with *q*-value <0.05.

HIV-VL, the other all SNPs associated with low HIV-VL (Fig. 2).

Discussion

Here we report on the effect of genetic variation in TET2 and related genes on clinical markers of HIV-pathogenesis. Gene level analysis revealed significant associations between TET2 with HIV-VL in two independent ARTnaive cohorts (START and FIRST), and with CD4⁺ T-cell count in one cohort (START). No gene level associations were observed for IDH1 or IDH2. SNP level analyses confirmed an association with HIV-VL, identifying a number of SNPs associated with HIV-VL in the START and FIRST cohorts. The association with CD4⁺ T-cell count was not confirmed at the SNP-level. No associations were observed with any of the other biomarkers.

Our findings at both gene and SNP level suggest that variation in the TET2 gene impacts HIV-VL in PWH not taking ART. This effect on HIV-VL appears to be driven by several SNPs that group into two distinct LD clusters, and are predominantly seen in persons of Black race. In the START cohort, we observed 13 SNPs significantly (*q*-value < 0.05) associated with lower HIV-VL at study entry, whereas 15 SNPs were significantly associated with a higher HIV-VL at study entry. These results were replicated using a linear mixed effects model which included follow-up HIV-VL data. All SNPs in the LD cluster associated with a higher HIV-VL in START were associated with a higher HIV-VL in FIRST; whereas all SNPs in the LD cluster associated with lower HIV-VL in

Tuble 5. Results and description of stars that were observed to be associated with three L in the stratt and thest const	Table 3.	Results and	description	of SNPs that	were observed	to be	associated	with	HIV-VL	in the	START	and FIR	ST coho	rts.
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dbSNP RS ID	Gene (±5 kb)	Chr	Physical position	Strand	Ref allele	Alt allele	START estimate ^a	START <i>P</i> -value	START q-value	START rank	FIRST estimate ^a	FIRST <i>P</i> -value	FIRST q-value	FIRST rank
rs77936891	TET2	4	106122657	+	А	G	-0.264	0.000025	0.0052	1	-0.1869	0.028	0.2587	37
rs75555015	TET2	4	106151781	+	А	G	-0.2839	0.000036	0.0052	2	-0.1838	0.059	0.4421	45
rs115930414	TET2	4	106173177	+	G	Т	-0.3001	0.000059	0.0057	3	-0.191	0.057	0.4421	44
rs78305976	TET2	4	106095219	+	Т	А	-0.2705	0.000082	0.0059	4	-0.1934	0.023	0.2386	33
rs139100435	TET2	4	106142080	+	CTT	_	0.3156	0.00012	0.0073	5	0.5107	0.00031	0.0072	15
rs80014899	TET2	4	106141445	+	А	G	-0.2453	0.00021	0.0092	6	-0.1728	0.040	0.3282	42
rs72955180	TET2	4	106133750	+	G	Т	0.4022	0.00024	0.0092	7	N/A	N/A	N/A	N/A
rs72950501	TET2	4	106197936	+	С	Т	0.3072	0.00028	0.0092	8	0.5511	0.000081	0.0058	4
rs76644731	TET2	4	106132312	+	А	G	-0.2451	0.00028	0.0092	9	-0.1894	0.026	0.2531	35
rs72955179	TET2	4	106133471	+	G	А	0.3022	0.00034	0.0096	10	0.5107	0.00031	0.0072	14
rs72955193	TET2	4	106145468	+	А	G	0.3781	0.00039	0.0096	11	0.5505	0.00017	0.0058	9
rs17035325	TET2	4	106110796	+	Т	С	-0.2394	0.00040	0.0096	12	-0.1761	0.036	0.305	41
rs77738124	TET2	4	106173540	+	Т	С	-0.2367	0.00044	0.0096	13	-0.1561	0.059	0.4421	46
rs59519484	TET2	4	106190732	+	С	А	0.2992	0.00046	0.0096	14	0.5141	0.00036	0.0077	16
rs142786189	TET2	4	106162216	+	Т	_	-0.2393	0.00053	0.0104	15	-0.1902	0.027	0.2587	36
rs78763791	LOC101929491;	4	106065102	+	С	А	-0.2107	0.00073	0.0133	16	-0.2485	0.0025	0.0371	23
	TET2													
rs72963014	TET2	4	106168623	+	С	G	0.348	0.00091	0.0155	17	0.5336	0.00022	0.0062	12
rs72963032	TET2	4	106182446	+	А	G	0.3407	0.0010	0.0164	18	0.5451	0.00011	0.0058	6
rs59479204	TET2	4	106160495	+	А	G	0.3411	0.0011	0.0164	19	0.5458	0.00018	0.0058	11
rs17035308	LOC101929491;	4	106064349	+	G	С	-0.213	0.0012	0.0169	20	-0.2392	0.0038	0.0521	25
	TET2													
rs58322634	TET2	4	106166139	+	А	G	0.2467	0.0013	0.0176	21	0.5561	0.000040	0.0058	1
rs137883243	TET2	4	106184236	+	ΑΤΑΤΑΑΑ	_	-0.2338	0.0013	0.0178	22	-0.2156	0.026	0.2531	34
rs72961199	TET2	4	106159386	+	G	А	0.3389	0.0015	0.0189	23	0.554	0.00015	0.0058	7
rs60382101	TET2	4	106160161	+	А	G	0.3372	0.0016	0.0192	24	0.554	0.00015	0.0058	8
rs72963038	TET2	4	106189156	+	С	А	0.3282	0.0017	0.0203	25	0.5731	0.000053	0.0058	2
rs79305653	TET2	4	106200234	+	Т	С	-0.2254	0.0022	0.0242	26	-0.1809	0.062	0.4476	48
rs72952305	TET2	4	106201653	+	Т	G	0.314	0.0035	0.0377	27	0.544	0.00026	0.0069	13
rs72963036	TET2	4	106188548	+	А	Т	0.3011	0.0038	0.0397	28	0.577	0.00010	0.0058	5
rs72963031	TET2	4	106182132	+	А	Т	0.2422	0.014	0.1276	31	0.4754	0.00068	0.0138	17
rs60786079	TET2	4	106197750	+	G	А	0.1663	0.015	0.1402	32	0.5261	0.000061	0.0058	3
rs72963046	TET2	4	106194419	+	С	А	0.2547	0.019	0.1601	34	0.5501	0.00017	0.0058	10
rs6811468	TET2	4	106082473	+	А	G	0.1042	0.19	0.7998	69	0.3239	0.0024	0.0371	22
rs12694101	IDH1; PIKFYVE	2	209128892	+	G	А	0.0096	0.77	0.9674	231	-0.1913	0.0014	0.0233	20
rs72963007	TET2	4	106164723	+	G	А	N/A	N/A	N/A	N/A	0.5476	0.00093	0.0169	19
rs72961197	TET2	4	106158738	+	G	А	N/A	N/A	N/A	N/A	0.5476	0.00093	0.0169	18
rs72955158	TET2	4	106110032	+	А	G	N/A	N/A	N/A	N/A	0.4082	0.0018	0.0302	21

N/A indicates that a particular SNP did not pass QC or the MAF cut-off for that particular study and was therefore not included in the GLM. MAF, minor allele frequency; QC, quality control; TET2, ten-eleven methylcytosine dioxygenase 2; VL, viral load. *SNPs were ordered by *P*-value (low to high) with rank = 1 for the SNP with the lowest *P*-value.

^aEstimates are presented for the alternate allele.

START were also associated with a lower HIV-VL in the FIRST study. Many of the SNPs (15/36) in the two clusters were significant across both studies, while the SNPs that were significantly associated with HIV-VL in only one of the two studies showed a similar effect size and direction in the other. Since SNP MAF and QC filtering were performed per study, some SNPs were only present in one study or the other. Amongst these discordant SNPs was rs72963007, which was significantly associated $(\beta = 0.55, q = 0.02)$ with HIV-VL in the FIRST cohort but was below the MAF cut-off in the START cohort. Of the TET2 SNPs that were associated with HIV-VL in our study, only rs72963007 has been reported in the literature, where it was associated with an increased risk of adult T-cell leukaemia caused by HTLV-1 in persons of African descent [23]. Like HIV, HTLV-1 is a retrovirus, with many similar genetic elements, and this association suggests a critical role for this SNP (and TET2 more generally) in retroviral pathogenesis.

The exact mechanism through which TET2 affects HIV-VL, and why some SNPs are associated with higher HIV-VL and others with lower HIV-VL, is not clear. Prior molecular work has focused on the relationship between TET2 and HIV-pathogenesis in monocyte/macrophage cells and cell-lines [32,33], with recent work suggesting that TET2 is only downregulated in HIV-infected monocyte-derived-macrophages and not CD4⁺ T cells [38]. If true, this could explain the relatively small effect these SNPs have on HIV-VL, as monocyte derived macrophages represent a smaller subset of the total infected cells than their T-cell counterparts. The SNPs themselves are in high LD and are spread out across the TET2 gene (although those associated with higher HIV-VL tend to congregate in the 5' end of TET2). Given the previous evidence associating TET2 with transcriptional regulation of endogenous retroviral elements [29-31], it may be that TET2 is playing a role in transcriptional regulation of HIV. Alternatively, previously studies have



Fig. 2. Pairwise linkage disequilibrium (LD) analysis of SNPs significantly associated with VL in the START (a) and/or FIRST (b) cohorts. The LD shows two distinct clusters of SNPs in both studies, consisting of SNPs that are associated with lower VL (on the left), and those associated with higher VL (on the right). All TET2 SNPs that associated with higher VL in START are also associated with higher VL in FIRST. Likewise, TET2 SNPs that associated with a lower VL in START are also associated with a lower VL in FIRST. Likewise, TET2 SNPs that associated with a lower VL in START are also associated with a lower VL in FIRST. Likewise, TET2 SNPs that associated with a lower VL in START are also associated with lower VL in FIRST. LD is measured using R^2 of each SNP pair; darker colour denotes stronger associations. (c) TET2 SNPs associated with lower VL are spread out across the TET2 gene. (d) TET2 SNPs associated with higher VL tend to cluster towards the 5' end of TET2. TET2, ten-eleven methylcytosine dioxygenase 2; VL, viral load.

suggested two independent mechanisms for TET2 related enhancement of HIV-VL, mediated through the VPR. The first via modulation of IL-6 and the second via IFITM3 [32,33], and it is possible that these SNPs are either preventing or enhancing one of these mechanisms. Although, in our study we did not observe consistent associations between TET2 and IL-6, and further laboratory work is required to elucidate the relationship between TET2 and HIV-VL.

Our study has several limitations. First, the five cohorts have different populations. This is particularly relevant for participants from the two trials that recruited only ARTnaive participants. Those in FIRST were more progressed than participants of START (as indicated by lower CD4⁺ T-cell count and higher HIV-VL). This difference means the HIV-VL phenotype is not directly comparable between these two cohorts, but as associations were found in both cohorts, it may be that the mechanism through which the TET2 mutations affect HIV-VL is independent of progression status. However, this also means that we may have been able to validate other signals (e.g. the association with CD4⁺ T-cell count and gene-level variation in the START cohort) if we were able to access cohorts with similar characteristics. Another limitation is the large number of associations explored and the less stringent multiple testing correction methodology used (compared to the Bonferroni method preferred by large GWAS'). Instead, we relied on validation of the associations across the

more than one cohort to increase confidence in our observations. We believe this, combined with the totality of molecular evidence suggesting that TET2 is involved in retroviral and HIV-pathogenesis, justifies the use of the less stringent multiple testing threshold utilized. Despite this, the exploratory nature of this study means that results should be interpreted cautiously and further studies (both laboratory based and in a different clinical populations) are required to validate the results observed here. Finally, the associations for both IDH1 and IDH2 were limited by the number of SNPs from these genes in the Affymetrix array we utilized. Only seven and 85 SNPs, compared to 796 for TET2, were included for IDH1 and IDH2, respectively.

In summary, we showed that genetic variations in TET2 were associated with levels of HIV-VL in ART-naive HIV-positive persons. These results support previous work reporting a key role for TET2 in retroviral replication and pathogenesis. This is the first in-human evidence of the importance of this gene in HIVpathogenesis. These observations warrant further investigation in additional cohorts and through molecular work. Of particularly interest is exploring the effects of these SNPs in larger populations of Black persons, since the risk allele of this SNP appears substantially more frequent in this subgroup. This group remains understudied in large genetic association studies, despite Black persons representing the vast majority of PWH globally [39].

Acknowledgements

The authors would like to thank all participants and staff from the trials described in this study.

See *N Engl J Med* 2015;373:795–807 for the complete list of START investigators.

See *Lancet* 2006;368:2125–35 for the complete list of FIRST investigators.

See *N Engl J Med* 2006;355:2283–96 for the complete list of SMART investigators.

See *N Engl J Med* 2009;361:1548–59 for the complete list of ESPRIT investigators.

See *PLoS ONE* 2010; 5(2):e9334. doi:10.1371/journal. pone.0009334 for the complete list of STALWART investigators.

Funding: This study was funded by the Danish National Research Foundation (DNRF126). The clinical trials described in this study were supported by the National Institute of Health (UM1-AI06864, UM1-AI120197, 1U01-AI36780, U01-AI46957, U01-AI042170, U01-AI046362), National Institute of Allergy and Infectious Diseases, National Institutes of Health Clinical Center, National Cancer Institute, National Heart, Lung, and Blood Institute, Eunice Kennedy Shriver National Institute of Child Health and Human Development, National Institute of Mental Health, National Institute of Neurological Disorders and Stroke, National Institute of Arthritis and Musculoskeletal and Skin Diseases, Agence Nationale de Recherches sur le SIDA et les Hépatites Virales (France), National Health and Medical Research Council (Australia), National Research Foundation (Denmark), Bundes ministerium für Bildung und Forschung (Germany), European AIDS Treatment Network, Medical Research Council (United Kingdom), National Institute for Health Research, National Health Service (United Kingdom), and University of Minnesota. Antiretroviral drugs were donated to the central drug repository by AbbVie, Bristol-Myers Squibb, Gilead Sciences, GlaxoSmithKline/ViiV Healthcare, Janssen Scientific Affairs, and Merck.

Conflicts of interest

D.D.M., C.R.M, A.G.Z., J.R., M.H. and J.D.L. were supported by the Danish National Research Foundation (DNRF126). B.G. was supported by the National Institute of Health (UM1 AI120197). A.H.B. was supported by the Lundbeckfonden (grant number R219-2016-762). L.D.D. was supported by the EACS Medical Exchange Programme and the Spanish Society of Infectious Diseases and Clinical Microbiology. M.H. participated in advisory boards for AstraZeneca, Gilead, GSK, MSD, Roche and Sobi and received speakers honoraria from Gilead and GSK. M.P. received Funding from BMS, Celgene, Gilead, Janssen, ViiV Pharmaceuticals all outside submitted work (to institution). Provision of drug and other materials for studies from Astex/ Otsuka, Celgene, CSL, Emergent Biosolutions, Janssen Grifols, Takeda, Verastem, ViiV Pharmaceuticals, all outside submitted work (to institution). Advisory board/ speakers panel for Celgene, Gilead, outside submitted work. For the remaining authors, no conflicts relating to the submitted work were declared.

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