

A Stronger Innate Immune Response During Hyperacute Human Immunodeficiency Virus Type 1 (HIV-1) Infection Is Associated With Acute Retroviral Syndrome

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Background. Acute retroviral syndrome (ARS) is associated with human immunodeficiency virus type 1 (HIV-1) subtype and disease progression, but the underlying immunopathological pathways are poorly understood. We aimed to elucidate associations between innate immune responses during hyperacute HIV-1 infection (hAHI) and ARS.

Methods. Plasma samples obtained from volunteers (\geq 18.0 years) before and during hAHI, defined as HIV-1 antibody negative and RNA or p24 antigen positive, from Kenya, Rwanda, Uganda, Zambia, and Sweden were analyzed. Forty soluble innate immune markers were measured using multiplexed assays. Immune responses were differentiated into volunteers with stronger and comparatively weaker responses using principal component analysis. Presence or absence of ARS was defined based on 11 symptoms using latent class analysis. Logistic regression was used to determine associations between immune responses and ARS.

Results. Of 55 volunteers, 31 (56%) had ARS. Volunteers with stronger immune responses (n = 36 [65%]) had increased odds of ARS which was independent of HIV-1 subtype, age, and risk group (adjusted odds ratio, 7.1 [95% confidence interval {CI}: 1.7–28.8], P = .003). Interferon gamma-induced protein (IP)-10 was 14-fold higher during hAHI, elevated in 7 of the 11 symptoms and independently associated with ARS. IP-10 threshold >466.0 pg/mL differentiated stronger immune responses with a sensitivity of 84.2% (95% CI: 60.4–96.6) and specificity of 100.0% (95% CI]: 90.3–100.0).

Conclusions. A stronger innate immune response during hAHI was associated with ARS. Plasma IP-10 may be a candidate biomarker of stronger innate immunity. Our findings provide further insights on innate immune responses in regulating ARS and may inform the design of vaccine candidates harnessing innate immunity.

Keywords. HIV-1; acute infection; immune responses; acute retroviral syndrome; IP-10.

Virologic and immunologic events during acute human immunodeficiency virus type 1 (HIV-1) infection (AHI) are associated with disease pathogenesis. During the first 10 days of HIV-1 infection (eclipse phase), the virus replicates at the

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infection site and in local lymphoid tissues before undergoing systemic dissemination. Subsequent stages of AHI can be defined by detection of virus RNA only (Fiebig stage I), virus RNA and p24 antigen (Fiebig stage II), and development of HIV-1 specific antibodies (Fiebig stages III-V) [1]. In 40–90% of cases, acute retroviral syndrome (ARS) presents before or during peak plasma viremia 2–3 weeks after HIV-1 infection [2–5]. The number and severity of symptoms have been correlated with rapid disease progression [6–8]. AHI symptoms are also more common in HIV-1 subtype A1 compared to subtype C and D infections [4].

AHI is further characterized by an induction of innate immune responses, including activation of dendritic cells, macrophages, and natural killer (NK) cells [9–11]. The earliest systemic innate immune perturbations occur during the eclipse phase of AHI. As HIV-1 viremia increases, pro-inflammatory

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cytokines, and chemokines are induced with elevations in multiple analytes creating a cytokine storm [12–14]. This systemic activation has been shown to have antiviral activity, enhance other innate immune responses, and shape adaptive immunity to control virus replication [15–19]. However, the exaggerated response also has immunopathological consequences by promoting provirus transcription, enhancing HIV-1 replication and driving CD4+ T-cell depletion [20–22]. Little is known about involvement of innate immune responses in the regulation of AHI symptoms. We hypothesized that activation of strong innate immune responses are associated with AHI symptoms. We sought to test this hypothesis in a hyperacute HIV-1 infection (hAHI) study involving volunteers enrolled into cohorts from 4 African and 1 European countries.

METHODS

Study Design and Ethical Considerations

Data and samples were obtained from African and Swedish adult (≥ 18 years old) volunteers enrolled in acute and early HIV-1 infection studies. African volunteers were enrolled 2006-2011 at sites in Kenya, Rwanda, Uganda, and Zambia under IAVI's protocol C [23]. Volunteers diagnosed with AHI before 2011 and identified from patient medical records at Skåne University Hospital in Lund/Malmö, Sweden, were included. Volunteers with hAHI diagnosis were eligible (defined as HIV-1 antibody negative and RNA [Fiebig stage I] or p24 antigen [Fiebig stage II] positive). Available matched preinfection samples from volunteers were also included in the analysis. All protocol C volunteers provided written informed consent for the use of their samples for biomedical research. Protocol C sites received approvals from respective country-specific ethics review boards [23]. Ethical approval for the Swedish cohort was obtained from the Lund University Ethical Review Board, Sweden (Diarienummer [Dnr] 2013/772).

Characteristics of Volunteers

Protocol C data were obtained from a centralized and curated repository. De-identified data from the Swedish volunteers were obtained from an electronic patient register (InfCareHIV) and medical records. Baseline variables included date of birth, sex, ethnicity, HIV-1 RNA, HIV-1 p24 antigen, and antibody test results, date of HIV-1 diagnosis, transmission risk group (serodiscordant, men who have sex with men [MSM], and heterosexual relationships), and AHI symptoms.

HIV-1 Subtyping

HIV-1 *env* sequences (*V1-V3*, approximately 940 base pairs) were generated, and HIV-1 subtype were phylogenetically determined as previously described [24]. In brief, HIV-1 RNA was extracted, reverse transcribed, polymerase chain reaction (PCR) amplified and sequenced using established protocols. Generated sequences were aligned with HIV-1 subtype

reference sequences from the Los Alamos HIV-1 sequence database (https://www.hiv.lanl.gov/content/index). The general time reversible (GTR) model of nucleotide substitution with gamma distributed rate heterogeneity was used to infer maximum likelihood trees. Branch support was assessed using the approximate likelihood ratio test based on the Shimodaira-Hasegawa (aLRT-SH) method, with branch support of \geq .90 considered significant [25].

Acute Retroviral Syndrome

For protocol C volunteers, AHI symptoms were ascertained 2–6 weeks following the estimated date of infection using a standardized questionnaire [4]. For Swedish volunteers, medical records were reviewed using the protocol C questionnaire as a template; documented AHI symptoms were extracted separately by 2 investigators. AHI symptoms reported at the incident visit or in follow-up visits (but within 6 weeks of incident visit) were extracted. Case files without a reported symptom were considered negative for that symptom. Inconsistencies were discussed and a consensus documented.

Acute retroviral syndrome (ARS) was defined based on 11 AHI symptoms. Previous studies have used various definitions for ARS including volunteers reporting any symptom, ≥ 2 symptoms, ≥ 3 symptoms, or a combination of fever with other symptom(s) [2, 26, 27]. A potential limitation of discrete classification methods is the inability to account for unobserved linkages between symptoms. Latent class analysis (LCA), a structural equation modelling approach, was therefore applied to group volunteers based on the number of AHI symptoms and other unobserved linkages, defined as those with or without ARS.

Innate Immune Responses

Cryopreserved plasma samples collected from whole blood treated with ethylenediaminetetraacetic acid (EDTA) were used. Multiplexed analysis was done in duplicates and measured by electrochemiluminescence using the MesoScaleDiscovery (MSD) multispot 40-plex assay (MesoScaleDiagnostics, LLC) according to manufacturer's instructions [28]. All samples were processed from the IAVI Human Immunology Laboratory, London, United Kingdom. Analyte concentration read-outs, along with respective lower (LLOQ) and upper (ULOQ) limits of quantification, were inferred from standard curves. Analyte concentration, whereas those above the ULOQ were assigned the ULOQ value. The mean of the duplicate observations were analyzed.

Data Analysis

First, ARS was defined based on the number of AHI symptoms and other unobserved linkages using LCA, and volunteers disentangled into those with and without ARS. Second, associations between \log_{10} analyte concentrations with AHI symptoms and with ARS were assessed using Wilcoxon rank sum tests. Analytes suggestive of an association with any of the 11 AHI symptoms and with ARS were further explored using principal component analysis (PCA). The PCA differentiated volunteers into those with stronger and comparatively weaker immune responses. Third, logistic regression was applied to determine associations between immune response and ARS. HIV-1 subtype, risk group, ethnicity, geographic region, and HIV-1 RNA were dichotomized to maximize precision. A threshold of P < .10 was used to select variables from univariable to multivariable models. Finally, individual analytes were assessed for prognostic performance in differentiating stronger immune responses using receiver operating characteristic area under the curve (ROC-AUC).

All analyses were done in Stata/IC version 15.1 (StataCorp LP, California), and outputs generated using GraphPad Prism version 8.4.2 (GraphPad Software, California). A 2-sided type 1 error of 5% was considered statistically significant. Due to sample size limitations and consistent with previous

publications, no formal corrections for multiple comparisons were done [14, 26].

RESULTS

Characteristics of Study Volunteers

Fifty-five volunteers were eligible. The majority of volunteers were diagnosed with HIV-1 by p24 (Fiebig II, n = 42 [76%]) and had a matched preinfection sample (n = 31 [56%]), collected at median 47 (interquartile range [IQR], 22–120) days before HIV-1 infection. A complete set of baseline variables were available for all volunteers, except for HIV-1 RNA that was missing from 62% of volunteers (Table 1). No significant differences in baseline characteristics of volunteers with missing versus with HIV-1 RNA data were found (Supplementary Table 1), and HIV-1 RNA was therefore excluded from further analysis. Most volunteer HIV-1 genetic sequences clustered with subtype A1 (n = 33 [60%], Figure 1). The majority of HIV-1 subtype A1 sequences (n = 26) were from Kenya, whereas all HIV-1 subtype B sequences (n = 8) were from Sweden.

Table 1. Characteristics of Volunteers Diagnosed With Hyperacute HIV-1 Infection From Africa (Rwanda, Uganda, Zambia, and Kenya) and Sweden by Fiebig Staging (N = 55)

Baseline Variables		Fiebig I (N = 13)	Fiebig II (N = 42)	Overall (N = 55)
Age (y)	Median (IQR)	22.6 (20.9–27.1)	29.3 (25.1–34.1)	28.1 (23.0–34.1)
Age group (y)	18.0-24.9	7 (54)	10 (24)	17 (31)
	25.0-34.9	4 (31)	22 (52)	26 (47)
	35.0-44.9	1 (8)	5 (12)	6 (11)
	45.0+	1 (8)	5 (12)	6 (11)
Gender	Female	O (O)	4 (10)	4 (7)
	Male	13 (100)	38 (90)	51 (93)
Year of infection	<2009	4 (31)	15 (36)	19 (35)
	2009–2010	7 (54)	18 (43)	25 (45)
	2011+	2 (15)	9 (21)	11 (20)
Ethnicity	African	13 (100)	29 (69)	42 (76)
	Arab/Asian	O (O)	1 (2)	1 (2)
	Caucasian	O (O)	12 (29)	12 (22)
Risk group ^a	DC	2 (15)	12 (29)	14 (25)
	HET	1 (8)	3 (7)	4 (7)
	MSM	10 (77)	24 (57)	34 (62)
	UNK	O (O)	3 (7)	3 (5)
Country	Sweden	O (O)	13 (31)	13 (24)
	African ^b	13 (100)	29 (69)	42 (76)
HIV-1 RNA (log ₁₀ cpm)	Median (IQR)	4.0 (3.8–4.6)	6.1 (5.9–7.0)	5.0 (4.0-6.0)
HIV-1 RNA group (log ₁₀ cpm)	≤5.0	11 (85)	O (O)	11 (20)
	>5.0	2 (15)	8 (19)	10 (18)
	Missing	O (O)	34 (79)	34 (62)
Matched preinfection sample ^c	No	2 (15)	22 (52)	24 (44)
	Yes	11 (85)	20 (48)	31 (56)
Matched preinfection period ^d	Median (IQR)	85 (22–158)	41 (22–81)	47 (22–120)

Abbreviations: cpm, copies per milliliter; HIV-1, human immunodeficiency virus type 1; IQR, interquartile range.

^aRisk group data (DC [serodiscordant couples], HET [heterosexual], MSM (men who have sex with men] and UNK [unknown]).

^bAfrican countries (protocol C sites from Rwanda [n = 4], Uganda [n = 5], Zambia [n = 4] and Kenya [n = 29]).

^cMatched preinfection samples from Kenya (n = 29), Rwanda (n = 1) and Zambia (n = 1).

^dAvailability of matched preinfection samples by days from sampling to the estimated date of infection (EDI).

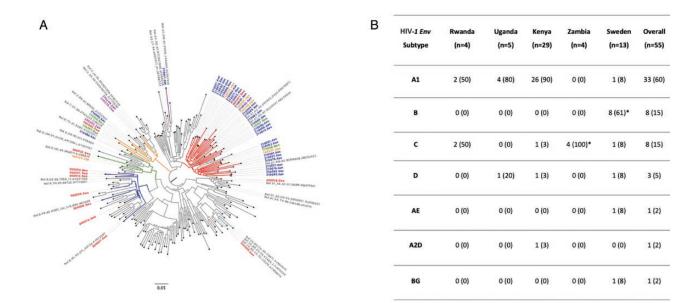


Figure 1. (*A*) Phylogenetic tree showing relatedness of HIV-1 partial *envelope* sequences (*V1-V3* region) from individuals with hyperacute HIV-1 infection from Africa and Sweden (N = 55), with HIV-1 subtype reference sequences from the Los Alamos database (N = 158); and (*B*) summary table showing the distribution of HIV-1 partial *env* (*V1-V3* region) subtypes of individuals with hyperacute infection from Africa and Sweden (N = 55). Branches are colored according to HIV-1 *env* subtype as follows: gray (references), red (subtype A1), dark blue (subtype B), orange (subtype C), green (subtype D), light blue (recombinant BG), purple (recombinant A2D), and brown (recombinant AE). Tip labels are colored according to the country of origin as follows: gray (references; only those clustering with volunteer sequences shown), green (Rwanda), orange (Uganda), blue (Kenya), purple (Zambia), and red (Sweden). *Unable to amplify HIV-1 *env* region from two volunteers, though previous *gag/pol/nef* sequence data suggests subtype C (Zambia, n = 1) and B (Sweden, n = 1) infections and included in the analysis as such. Abbreviation: HIV-1, human immunodeficiency virus type 1.

Acute Retroviral Syndrome

Overall, 43 (78%) volunteers had at least one AHI symptom. The median number of symptoms was 4 (IQR, 1–7). The most common symptom was fever (n = 37 [67%]; Figure 2A). Using LCA, volunteers were grouped into those with ARS (n = 31 [56%]; median symptoms per volunteer, 6 [IQR, 5–8]) and without ARS (n = 24 [44%]; 0.5 [IQR,

0–2.5], Figure 2B). Nine of the 11 AHI symptoms were significantly higher in volunteers with ARS. Myalgia demonstrated the strongest, whereas skin rash and pharyngitis demonstrated no discriminatory potential for ARS. Notably, fever was common in both groups (with and without ARS), suggesting a weak discriminatory ability for ARS (Figure 2C).

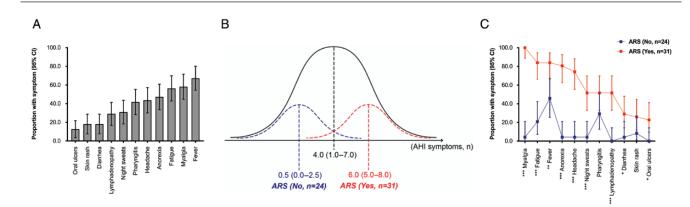


Figure 2. (a) Distribution of volunteers diagnosed with hyperacute HIV-1 infection by AHI symptoms; (b) a schematic illustration of the distribution of AHI symptoms for all volunteers (black curve) and by grouping of volunteers resulting from latent class analysis (LCA, for those without acute retroviral syndrome, [ARS, blue] and for those with ARS [red]). The numbers denote median (interquartile ranges) symptoms per volunteer for all volunteers (in black), for those without ARS (in blue) and for those with ARS (in red); and (c) Graph comparing the distribution of AHI symptoms between volunteers that were defined to be with and without ARS (Fisher exact test P < .05 [*], P < .01 [**], and P < .001 [***], N = 55). ARS was defined based on the 11 AHI symptoms and other unobserved linkages between symptoms using LCA. Incremental latent group models were assessed to predict the goodness of fit. The model with 2 latent groups was the best fit as it had the lowest BIC value (660.5) compared to 3 (678.6), 4 (699.2), or 5 (714.7) groups. Volunteers were grouped based on their predicted posterior probabilities into those with ARS (n = 31 [56%]) and those without ARS (n = 24 [44%]). Abbreviations: AHI, acute human immunodeficiency virus type 1; ARS, acute retroviral syndrome; CI, confidence interval; HIV-1, human immunodeficiency virus type 1; LCA, latent class analysis.

Innate Immune Responses During Hyperacute HIV-1 Infection

SAA, CRP, VCAM-1, and ICAM-1 concentrations were higher than the ULOQ in >90% of volunteers and therefore excluded from further analysis. Compared to matched preinfection responses, 16 analytes were significantly elevated during hAHI (Figure 3*A*; Supplementary Table 2). A 14-fold higher median level of Interferon gamma-induced protein (IP)-10 was observed during hAHI compared to HIV-1 preinfection (2400 [IQR, 251–2400] vs 168 [IQR, 86–204] pg/mL). Substantial between-individual variation in analyte responses was observed (Figure 3*B*). Macrophage inflammatory protein (MIP)-1β, Eotaxin-3, MIP-1α, interleukin (IL)-8, IL-1β, IL-2, IL-4, IL-6, IL-8, IL-12p70, IL-13, and tumor necrosis factor (TNF)-α were positively correlated with each other (Supplementary Figure 1).

Associations Between Innate Immune Responses and Acute Retroviral Syndrome

Overall, 18 analytes were higher in volunteers with any AHI symptoms compared to those without symptoms (Supplementary Figure 2). IP-10 activation stood out, with plasma concentrations being significantly higher across 7 of the 11 AHI symptoms (Figure 4A). Further, there were significantly higher IP-10 (P < .001), interferon (IFN)- γ (P = .020), granulocyte-macrophage colony-stimulating factor (GM-CSF) (P = .033), IL-12 (P = .043), IL-15 (P = .037), Eotaxin-3 (P = .002), Flt-1 (P = .025), IL-10 (P = .047), IL12p70 (P = .027) and IL-13 (P = .047) plasma concentrations amongst

volunteers with versus without ARS (Figure 4*B*; Supplementary Table 3). After adjusting for HIV-1 subtype, age, and risk group, elevated IP-10 was independently associated with ARS (adjusted coefficient, 4.7 [95% confidence interval {CI}: 0.8-8.6], P = .002, Table 2).

Significantly elevated analytes in volunteers with any of the 11 symptoms and in those with ARS were analyzed by PCA. Two principal components (PCs) were determined: PC1 (IP-10, IFN- γ , IL-12, IL-15, and Flt-1); and PC2 (Eotaxin-3, IL-13, and IL-12p70, Supplementary Figure 3). Both PCs were significantly associated with ARS, suggesting a synergistic effect of multiple analytes on ARS (more so for Eotaxin-3, IL-13 and IL-12p70, which were not significantly associated with ARS when assessed independently, Supplementary Table 4).

Furthermore, volunteers were grouped into those with higher PC1 loading and those with higher PC2 loading. Volunteers with higher PC1 loading had significantly higher median plasma concentration for the majority of the 40 analytes compared to those with higher PC2 loading (Figure 5*A*, Supplementary Table 5). Hence, PC1 analytes differentiated volunteers into those with stronger immune responses (n = 36 [65%]) and comparatively weaker immune responses (n = 19 [35%], Figure 5*B*). After controlling for HIV-1 subtype, age and risk group, volunteers with stronger innate immune responses had increased odds of ARS (adjusted odds ratio, 7.1 [95% CI: 1.7–28.8], *P* = .003, Table 3).

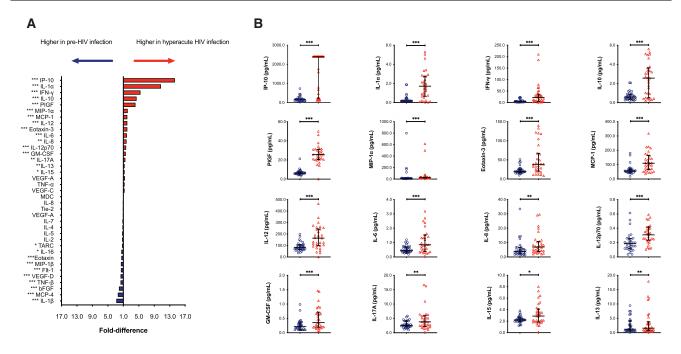
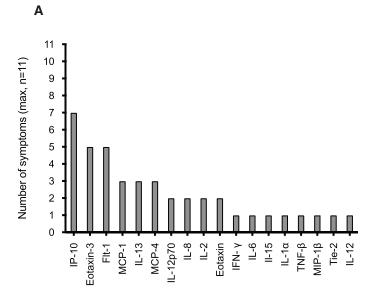


Figure 3. (*A*) Plot illustrating fold-differences in median plasma biomarker concentration from matched pre-infection and hyperacute HIV-1 infection samples. Red bars demonstrate fold difference in analytes that had higher plasma concentrations during hyperacute HIV-1 infection compared to the pre-infection samples. Blue bars denote fold-difference in analytes that had higher concentrations in the pre-infection samples compared to hyperacute HIV-1 infection samples. (*B*) Graphs illustrating analytes that were significantly higher during hyperacute HIV-1 infection compared to preinfection (Wilcoxon paired signed-rank test P < .05 [*], P < .01 [**], and P < .001 [***], N = 31). Abbreviations: bFGF, basic fibroblast growth factor; FIt, FMS-like tyrosine kinase; GM-CSF, granulocyte-macrophage colony-stimulating factor; HIV-1, human immunodeficiency virus type 1; IP, interferon gamma-induced protein; IFN, interferon; IL, interleukin; MCP, monocyte chemoattractant protein; MDC, macrophage-derived chemokine; MIP, macrophage inflammatory protein; PIGF, placental growth factor; TARC, thymus and activation-related chemokine; TNF, tumor necrosis factor; VEGF, vascular endothelial growth factor.



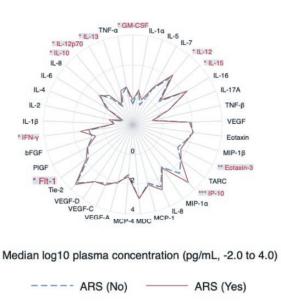


Figure 4. (*A*) Graph summarizing associations between \log_{10} median plasma analyte concentration when comparing volunteers with and those without AHI symptoms (Supplementary Figure 2), and (*B*) plot illustrating differences in \log_{10} median plasma biomarker concentration between hyperacute HIV-1 infected volunteers with ARS compared to mild ARS (Wilcoxon rank sum test P < .05 [*], P < .01 [**], and P < .001 [***], N = 55). Analytes suggestive of an association with any of the 11 AHI symptoms and with ARS (IL-12, IL-15, Eotaxin-3, IP-10, FIt-1, IFN- γ , IL-12p70 and IL-13) were further explored using principal component analysis (PCA). Abbreviations: AHI, acute human immunodeficiency virus type 1; ARS, acute retroviral syndrome; bFGF, basic fibroblast growth factor; FIt, FMS-like tyrosine kinase; GM-CSF, granulocyte-macrophage colony-stimulating factor; HIV-1, human immunodeficiency virus type 1; IFN, interferon; IP, interferon gamma-induced protein; IL, interleukin; MCP, monocyte chemoattractant protein; MDC, macrophage-derived chemokine; MIP, macrophage inflammatory protein; PIGF, placental growth factor; TARC, thymus and activation-related chemokine; TNF, tumor necrosis factor; VEGF, vascular endothelial growth factor.

Next, analytes with higher PC1 loading (IP-10, IFN- γ , IL-12, IL-15, and Flt-1) were analyzed independently to assess their prognostic performance in differentiating a stronger innate immune response. All analytes demonstrated high prognostic value (ROC range, 0.73–0.90), with no significant differences in the AUCs (Supplementary Figure 4). The best performing analyte and threshold to differentiate

Table 2. Logistic Regression Analysis Demonstrating the Association Between Specific Analytes and Acute Retroviral Syndrome (ARS) Among Volunteers With Hyperacute HIV-1 Infection From Africa (Rwanda, Uganda, Zambia, and Kenya) and Sweden (N = 55)

Analyte	Crude Coef. (95% Cl)	<i>P</i> -value	Adjusted Coef. (95% Cl) ^a	<i>P</i> -value
IL-12	.24 (89 to 1.37)	.672		
IL-15	.83 (98 to 2.64)	.355		
Eotaxin-3	3.20 (1.06–3.81)	<.001	3.57 (98 to 8.11)	.078
IP-10	2.44 (1.07–3.81)	<.001	4.71 (.84 – 8.57)	.002
Flt-1	07 (-1.12 to .98)	.891		
IFN-γ	1.13 (.15 – 2.11)	.016	0.58 (-2.10 – 3.26)	.671
IL-12p70	1.10 (44 to 2.64)	.161		
IL-13	1.39 (.14–2.63)	.020	1.37 (-1.77 to 4.51)	.379

Abbreviations: CI, confidence interval; Coef, coefficient; Flt, FMS-like tyrosine kinase; HIV-1, human immunodeficiency virus type 1; IFN, interferon; IL, interleukin; MSM, men who have sex with men.

 $^{a}\text{Adjusted}$ for age group (18.0–24.9 vs >25.0 years), HIV-1 env subtypes (Non-A1 vs A1) and risk group (Non-MSM vs MSM).

a stronger innate immune response was IP-10 (466 pg/mL; AUC: 0.92 [95% CI: 0.84–1.00]; sensitivity: 84.2% [95% CI: 60.4–96.6] and specificity: 100.0% [95% CI: 90.3–100.0], Figure 6; Supplementary Table 6).

Fiebig II Subpopulation Analysis

Due to the dynamic nature of innate immune responses during hAHI, we performed a sensitivity analysis to address the potential confounding effect of combining Fiebig I and II samples. The analysis was restricted to Fiebig II samples (n = 42)and followed the analytical pipeline that was used for the full data set. Overall, the results were consistent with the analysis of the full data set. Briefly, the most common symptom was fever. Volunteers were grouped by LCA into those with (n = 26)[62%]) and without ARS (n = 16 [38%], Supplementary Figure 5). Fifteen analytes were significantly elevated during hAHI, with IP-10 having 9-fold higher levels compared to preinfection levels (Supplementary Figure 6). IP-10, Eotaxin-3, and Flt-1 were elevated in volunteers with ARS (Supplementary Figure 7). IP-10 was independently associated with ARS after controlling for HIV-1 subtype, age, and risk group (Supplementary Table 7). Analytes significantly altered during hAHI and with any symptom were analyzed by PCA. These clustered into 3 PCs indicative of stronger (PC1 and PC2, n = 28 [67%]) and

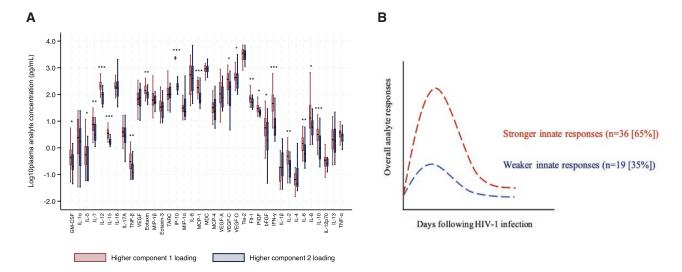


Figure 5. (*A*) Graph demonstrating the differentiation of analyte responses by principal component analysis loadings. Volunteers with higher principal component (PC) 1 loadings had significantly higher plasma concentration in a majority of the analytes compared to those with higher PC2 loadings (Wilcoxon rank sum test P < .05 [*], P < .01 [**], and P < .001 [***], N = 55); (*B*) A schematic illustration of the differentiation of analyte responses into volunteers with and without a cytokine storm; and Wilcoxon signed rank test (P < .05 [*], P < .01 [**], P < .01 [**], P < .05 [**], P < .005 [***], and P < .0005 [***]; paired/matched signed rank test in green, N = 31). Abbreviations: hbFGF, basic fibroblast growth factor; FIt, FMS-like tyrosine kinase; GM-CSF, granulocyte-macrophage colony-stimulating factor; hAHI, hyperacute HIV-1 infection; HIV-1, human immunodeficiency virus type 1; IFN, interferon; IL, interleukin; IP, interferon gamma-induced protein; MCP, monocyte chemoattractant protein; MDC, macrophage-derived chemokine; MIP, macrophage inflammatory protein; PIGF, placental growth factor; TARC, thymus and activation-related chemokine; TNF, tumor necrosis factor; VEGF, vascular endothelial growth factor.

comparatively weaker innate immune responses (PC3, n = 14 [33%], Supplementary Figures 8 and 9; Supplementary Table 8). Volunteers with stronger innate immune responses had

increased odds of ARS (adjusted odds ratio [aOR] 11.9 [95% CI: 1.1–130.4], P = .019, Supplementary Table 9). Finally, IP-10 was the best prognostic performing analyte (threshold: 1752

Table 3. Logistic Regression Analysis Demonstrating the Association Between Innate Immune Responses and Acute Retroviral Syndrome (ARS) Among
Volunteers With Hyperacute HIV-1 Infection From Africa and Sweden (N = 55)

Particulars		ARS, n (%)	Crude OR (95% CI)	P-value	Adjusted OR (95% CI)	<i>P</i> -value
Innate immune responses	Weaker responses	6/19 (32)	Ref	.007	Ref	.003
	Stronger responses	25/36 (69)	4.9 (1.5 – 16.3)		7.1 (1.7–28.8)	
Fiebig staging	Fiebig I	5/13 (38)	Ref	.137		
	Fiebig II	26/42 (62)	2.6 (.7 – 9.3)			
Gender	Male	28/51 (55)	Ref	.422		
	Female	3/4 (75)	2.5 (.2 – 25.3)			
Age group (y)	18.0 - 24.9	13/17 (76)	3.6 (1.0 – 13.1)	.040	2.4 (.5-12.1)	.264
	25.0 +	18/38 (47)	Ref		Ref	
Year of infection	<2009	11/19 (58)	Ref	.712		
	2009 - 2010	15/25 (60)	1.1 (.3 – 3.7)			
	2011+	5/11 (45)	.6 (.1 – 2.7)			
HIV-1 e <i>nv</i> subtype	Non-A1 ^a	9/22 (41)	Ref	.059	Ref	.176
	A1	22/33 (67)	2.9 (1.0 – 8.8)		2.5 (.6 – 9.3)	
Ethnicity	Non-African ^b	7/13 (54)	Ref	.834		
	African	24/42 (57)	1.1 (.3 – 4.0)			
Risk group	Non-MSM ^c	8/21 (38)	Ref	.031	Ref	.084
	MSM	23/34 (68)	3.4 (1.1 – 10.6)		3.3 (.8–12.8)	
Country	Sweden	7/13 (54)	Ref	.834		
	African ^d	24/42 (57)	1.1 (.3–3.9)			

Abbreviations: CI, confidence interval; HIV-1, human immunodeficiency virus type 1; MSM, men who have sex with men; OR, odds ratio; Ref, reference.

 $^{\rm a}{\rm HIV}\textsc{-1}$ env subtypes B (n = 8), C (n = 8), D (n = 3), BG (n = 1), A2D (n = 1), and AE (n = 1).

^bCaucasian (n = 12), Asian/Arabian (n = 1).

 $^{c}\textsc{Discordant}$ couples (n = 14), heterosexual (n = 4), and unknown (n = 3).

 d Rwanda (n = 4), Uganda (n = 5), Zambia (n = 4), and Kenya (n = 29).

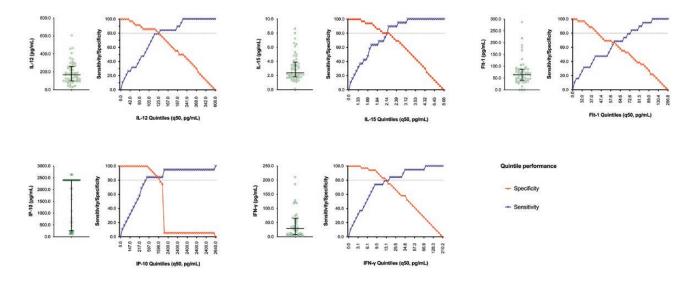


Figure 6. Graphs demonstrating distribution, performance (sensitivity and specificity) and threshold of analytes (IL-12, IL-15, Flt-1, IP-10, and IFN-γ) indicative of a stronger innate immune response. Analytes and threshold with both sensitivity and specificity of more than 80% were further explored. Abbreviations: Flt, FMS-like tyrosine kinase; IFN, interferon; IL, interleukin; IP, interferon gamma-induced protein.

pg/mL; AUC of 0.82 [95% CI: 0.69–0.95]; sensitivity: 71.4% [95% CI: 41.9–91.6] and specificity: 92.9% [95% CI: 76.5–99.1], Supplementary Figures 10 and 11; Supplementary Table 10).

DISCUSSION

Our findings are consistent with previous studies reporting striking elevation of innate immune markers resulting in the well-established cytokine storm during AHI [12-14]. In the absence of a gold standard, we applied a holistic systems approach to differentiate innate immune activation, and observed that a stronger immune response during hAHI was independently associated with ARS. To our knowledge, only one other study has reported associations between innate immune responses during AHI and ARS [26]. Crowell et al included 430 volunteers with AHI, a majority (62%) of whom were enrolled at Fiebig stage III or later. Five of the 10 analyzed analytes were included in our analysis (TNF-a, IL-6, IL-7, IP-10, and MCP-1). In contrast to our findings, TNF- α was elevated in ARS, whereas no association with IP-10 was reported. Several factors may explain these differences. First, because our volunteers were enrolled at Fiebig stages I/II, we were able to observe the earliest innate immune perturbations, including the cytokine storm, which may have been missed by a majority of the volunteers in the study by Crowell et al. Second, Crowell et al studied participants enrolled in Bangkok, Thailand (an HIV-1 CRF01_AE predominant region), whereas we studied participants from Africa and Europe infected by various strains of HIV-1 (subtypes A, B, C, and D, representing >70% of HIV-1 infections globally) [29]. Indeed, varying rates of disease progression have been suggested for different subtypes and circulating recombinant forms (CRFs) [30]. Third, given the different study settings, it is possible that differences in host genetic factors impacted innate immune responses differently [31]. Finally, Crowell et al defined ARS as \geq 3 symptoms, whereas we considered both the number of symptoms and unobserved relatedness between symptoms. This may have resulted in a more sensitive differentiation of ARS.

We identified 16 analytes that had significantly higher plasma concentrations during hAHI compared to preinfection values. Systemic upregulation of innate immune analytes during AHI is consistent with literature, and the association with ARS may be explained by multiple but related events [12-14]. In brief, and during AHI, the virus targets innate immune cells including dendritic cells, macrophages, and NK cells [9–11]. The result is significant destruction of these cells, widespread systemic virus dissemination, and a subsequent immunological milieu resulting in activation of innate and adaptive immune responses. Further, debris released from infected cells and other microbial products are translocated across mucosal barriers causing additional systemic inflammation and cytokine production [9, 13]. Altogether, these processes result in profound damage to cells, blood capillaries and mucosal barriers which may explain subsequent systemic (eg, fever), lymphatic (eg, lymphadenopathy), gastrointestinal (eg, diarrhea), musculoskeletal (eg, myalgia), and neurologic (eg, headache) symptoms. Notably, cytokine responses are activated and peak within 2 weeks of HIV-1 infection [13], whereas symptoms start manifesting 2-3 weeks after infection [3, 32], suggesting that systemic innate immune activation is a precursor for symptomatic manifestations.

The observed IP-10 elevation among volunteers with ARS is consistent with literature [14, 33]. However, its role in the induction of AHI symptoms is less understood. IP-10 is induced during the eclipse phase of AHI by innate immune cells in response to IFN- γ stimulation [34]. It is therefore not surprising that IP-10 clustered closely with IFN-y during PCA analysis. IP-10 induction increases exponentially and in parallel with HIV-1 RNA but prior to symptom manifestation [12]. In turn, IP-10 induces chemotaxis, apoptosis, cell growth inhibition, and angiostasis of different immune cells contributing to the damaging immunological milieu as described above. The role of IP-10 in disease pathogenesis is not unique to HIV-1; elevated IP-10 has been documented in a diverse range of virus infections, such as hepatitis C virus (HCV) infection, respiratory infections, and the more recent severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) infection [35-37]. The utility of IP-10 as a diagnostic marker for infectious diseases has also been explored. For instance, and in a meta-analysis, IP-10 has been suggested as a diagnostic marker for differentiating active from latent tuberculosis [38]. IP-10 has also been suggested as a marker for detection of AHI and for virus load monitoring [33, 39]. Yet the role of IP-10 as a biomarker in the differentiation of stronger immune responses remains undocumented. We propose consideration of IP-10 as a candidate marker in the differentiation of stronger innate immune responses at a threshold of >466 pg/mL.

Our study is not without limitations. First, the majority of our volunteers were missing HIV-1 RNA data. HIV-1 RNA increases and peaks 2-3 weeks after infection [2-5]. As HIV-1 viremia increases, pro-inflammatory cytokines and chemokines are induced, suggesting a possible correlation, as has been demonstrated between HIV-1 RNA and IP-10 [12]. HIV-1 RNA is also associated with ARS [26]. Further studies to disentangle the relationship between HIV-1 RNA, innate immune perturbations, and ARS are therefore warranted. Second, although our strict Fiebig I/II inclusion criteria enabled us to elucidate very early immune responses, it is possible that peak activation of analytes with delayed response may have been missed [12, 13]. Third, it is possible that volunteers seeking hospital care are more likely to present with ARS compared to those from a routine longitudinal cohort. However, there was no significant difference in the number of symptoms reported by the hospital volunteers compared to routine cohort volunteers (median, 5 [IQR, 3-6] vs 4 [IQR, 0–7] symptoms per volunteer respectively, P = .610). Finally, our small sample size and potential bias toward male volunteers (93% of the study population) warrants validation of our findings in larger and more sex-balanced AHI studies.

In conclusion, we applied holistic systems approaches to elucidate ARS from AHI symptoms and to disentangle the myriad of complex but related pathways of soluble innate immune markers. We found compelling evidence of an association between stronger innate immune responses and ARS. Specifically, plasma IP-10 was profoundly activated during AHI and associated with ARS. Furthermore, and in the absence of a gold standard, we also identified plasma IP-10 as a candidate biomarker in the differentiation of a stronger innate immune response. Our findings provide further insights into early immune responses during hAHI, their role in the regulation of ARS, and may have implications for the design of vaccine candidates that harness innate immunity for virus neutralization or replication control.

Supplementary Data

Supplementary materials are available at Clinical Infectious Diseases online. Consisting of data provided by the authors to benefit the reader, the posted materials are not copyedited and are the sole responsibility of the authors, so questions or comments should be addressed to the corresponding author.

Notes

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