Supplementary Methods

Bayesian models of IFN-I resistance of HIV-1 plasma isolates from longitudinally sampled participants

To create a simple model of the temporal dynamics of type I interferon (IFN-I) resistance, IFN $\alpha 2$ and IFN β IC $_{50}$ values were each modeled using a Bayesian change point hierarchical model. The model is based on a segmented regression of the log IC $_{50}$ making the following simplifying assumptions:

- Each participant has a level of resistance at the acute infection stage drawn from separate population-level distributions for typical, non- or fast progressors.
- Each participant has a drop (or rise) in IFN-I resistance from acute levels drawn from separate population-level distributions for typical, non- or fast progressors.
- Each participant has a time to nadir drawn from separate population-level distributions for typical, non- or fast progressors.
- Resistance changes linearly from onset of symptoms to time of nadir.
- The nadir of IFN-I resistance represents a changepoint in the data. Following this
 point, changes in IFN-I resistance are modeled as a linear function of CD4+ T cell
 count changes away from the count found at nadir.
- Where data is not present, CD4+ T cell counts are assumed to be linearly interpolated between adjacent observations.

The log IC₅₀ observation from each viral isolate i was modeled as a normal distribution IC50 $_i \sim \text{Normal}(\mu_i, \sigma)$ with mean μ_i where:

$$\mu_{\mathrm{I}} = \begin{cases} \alpha_{\mathsf{person}_i} + \delta_{\mathsf{person}_i} \frac{\mathsf{time}_i}{s_{\mathsf{person}_i}} & \mathsf{if time}_i < s_{\mathsf{person}_i} \\ \alpha_{\mathsf{person}_i} + \delta_{\mathsf{person}_i} + \beta_{\mathsf{person}_i} (\mathsf{CD4}_{\mathsf{person}_i,\mathsf{time}_i} - \mathsf{CD4}_{\mathsf{person}_i,s_{\mathsf{person}_i}}) & \mathsf{if time}_i \geq s_{\mathsf{person}_i} \end{cases}$$

where the parameters α_j represent the level of IFN-I resistance at symptom onset, δ_j represents the change from symptom onset to nadir and s_j represents the time of nadir in person j. Study participant data is represented by time, corresponding to the time since onset of symptoms and person, recording the participant from which isolate i was collected, CD4 $_{j,k}$ containing the estimated CD4+ T cell count for person j at time k and progression, is the disease progression type (fast/non/typical) for participant j. The hierarchical probabilities

for these parameters were:

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\sigma \sim \mathsf{Gamma}(1,0.1)
\alpha_j \sim \begin{cases} \mathsf{Normal}(\theta_{\alpha,\mathsf{typical}},\tau_\alpha) & \text{if progression}_j = \mathsf{typical} \\ \mathsf{Normal}(\theta_{\alpha,\mathsf{typical}} + \theta_{\alpha,\mathsf{fast}},\tau_\alpha) & \text{if progression}_j = \mathsf{fast} \\ \mathsf{Normal}(\theta_{\alpha,\mathsf{typical}} + \theta_{\alpha,\mathsf{non}},\tau_\alpha) & \text{if progression}_j = \mathsf{non} \end{cases}
\delta_j \sim \begin{cases} \mathsf{Normal}(\theta_{\delta,\mathsf{typical}},\tau_\delta) & \text{if progression}_j = \mathsf{typical} \\ \mathsf{Normal}(\theta_{\delta,\mathsf{typical}} + \theta_{\delta,\mathsf{non}},\tau_\delta) & \text{if progression}_j = \mathsf{fast} \\ \mathsf{Normal}(\theta_{\delta,\mathsf{typical}} + \theta_{\delta,\mathsf{non}},\tau_\delta) & \text{if progression}_j = \mathsf{typical} \end{cases}
s_j \sim \begin{cases} \mathsf{NegativeBinomial}(\theta_{s,\mathsf{typical}},\tau_s) & \text{if progression}_j = \mathsf{typical} \\ \mathsf{NegativeBinomial}(\theta_{s,\mathsf{typical}} \exp(\theta_{s,\mathsf{fast}}),\tau_s) & \text{if progression}_j = \mathsf{fast} \\ \mathsf{NegativeBinomial}(\theta_{s,\mathsf{typical}} \exp(\theta_{s,\mathsf{non}}),\tau_s) & \text{if progression}_j = \mathsf{non} \end{cases}
\beta_j \sim \begin{cases} \mathsf{Normal}(\theta_{\beta,\mathsf{typical}},\tau_\beta) & \text{if progression}_j = \mathsf{typical} \\ \mathsf{Normal}(\theta_{\beta,\mathsf{fast}},\tau_\beta) & \text{if progression}_j = \mathsf{fast} \\ \mathsf{Normal}(\theta_{\beta,\mathsf{non}},\tau_\beta) & \text{if progression}_j = \mathsf{fast} \\ \mathsf{Normal}(\theta_{\beta,\mathsf{non}},\tau_\beta) & \text{if progression}_j = \mathsf{non} \end{cases}
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where j indicates each participant and NegativeBinomial(x,y) represents a negative binomial distribution parameterized such that the expected value is x and the variance is $x+\frac{x^2}{y}$. All hyperparameters were given prior probabilities of $\theta_x \sim \text{Normal}(0,10)$ for parameters representing the means of a distribution and $\tau_x \sim \text{Gamma}(1,0.1)$ for parameters representing standard deviations other than $\theta_{\alpha,\text{typical}}$ and $\theta_{s,\text{typical}}$ which were given a flat prior and $\tau_s \sim \text{Cauchy}(0,10)$.

For computational efficiency, the nadir time parameter s was discretized to weekly intervals, assumed to fall within 1 to 150 weeks after symptom onset and marginalized out of the joint probability:

$$p(\mathsf{IC50},...) = p(...) \prod_{i=1}^n \sum_{s=1}^{150} \mathsf{Normal}(\mathsf{IC50}_i | \mu_{i,s}, \sigma) \mathsf{NegativeBinomial}(s | \theta_s, \tau_s)$$

where ... represents all parameters other than s and $\mu_{i,s}$ is defined the same as μ_i :

$$\mu_{i,s} = \begin{cases} \alpha_{\mathsf{person}_i} + \frac{\mathsf{time}_i}{s} \delta_{\mathsf{person}_i} & \text{if time}_i < s \\ \alpha_{\mathsf{person}_i} + \delta_{\mathsf{person}_i} + \beta_{\mathsf{person}_i} (\mathsf{CD4}_{\mathsf{person}_i,\mathsf{time}_i} - \mathsf{CD4}_{\mathsf{person}_i,s}) & \text{if time}_i \geq s \end{cases}$$

Posterior probabilities were estimated with 50 Markov chain Monte Carlo chains of 5000 iterations each using Stan (91).

Bayesian models of IFN-I resistance of outgrowth and rebound HIV-1 isolates

To compare the IFN-I resistance of viral isolates derived from plasma samples collected during acute, chronic and rebound infections, as well as from viably frozen PBMCs collected during ART suppression (QVOA), IFN $\alpha 2$ and IFN β IC $_{50}$ values were modeled using a Bayesian hierarchical model. The model is based on the assumptions that:

- Isolates found at acute infection form a base level of IFN-I resistance for a given person. Resistances in virus isolated from chronic, ART suppressed and rebound infection for this person are modelled as changes from this initial level.
- The mean IC₅₀ level within each person for acute isolates and the change from acute levels for chronic, QVOA and rebound isolates are drawn from a population-level distribution for that type.
- QVOA isolates are separated into to two populations; a "pre" group composed of QVOA viruses isolated from study participants prior to or in the absence of treatment interruption (ATI) and a "post" group of QVOA viruses isolated from participants following ATI and reinitiation of ART.
- In both QVOA populations, the viruses can include some proportion of rebound-like isolates. This mixture is modeled in both pre- and post-treatment so that differences in mixture proportion between the two populations can be assessed.
- Variation in the potency of INF-I used to experimentally determine IC_{50} values may shift the inferred resistance for isolates tested in other studies. This effect is modeled as a multiplicative shift in IC_{50} for all isolates measured outside this study (acute recipient and chronic donor isolates from ref. 50).
- Isolates from participants who received exogenous IFN $\alpha 2$ during treatment interruption may display altered interferon resistance. This effect is modeled as a multiplicative shift in IC $_{50}$ for all rebound isolates from such participants (participants 004, 030, and 044 from ref. 38).

The log IC $_{50}$ observation from each viral isolate i from acute, chronic and rebound isolates was modeled as a normal distribution:

$$\mathsf{IC50}_i \sim \mathsf{Normal}(\mu_{\mathsf{type}_i,\mathsf{person}_i},\sigma_{\mathsf{type}_i})$$

with the mean resistance for isolate type j from person k:

$$\mu_{j,k} \sim \begin{cases} \mathsf{Normal}(\alpha_k + \beta_{\mathsf{batch}} \mathsf{batch}_k, \psi_j) & \text{if } j = \mathsf{acute} \\ \mathsf{Normal}(\alpha_k + \beta_{j,k} + \beta_{\mathsf{batch}} \mathsf{batch}_k + \beta_{\mathsf{ifn}} \mathsf{IFN}_k, \psi_j) & \text{if } j = \mathsf{rebound} \\ \mathsf{Normal}(\alpha_k + \beta_{j,k} + \beta_{\mathsf{batch}} \mathsf{batch}_k, \psi_j) & \text{otherwise} \end{cases}$$

where type_i indicates whether isolate i was isolated during acute, chronic, QVOA or rebound infection from participant person_i , batch_k indicates when isolates from person_k were tested in another study and IFN_k indicates when person_k was treated with exogenous $\operatorname{IFN}_{\alpha 2}$ prior to and during treatment interruption. Parameters are included for the mean resistance level during acute infection for each person_k , standard deviation of isolates of type i0 within a person i1, standard deviation of mean resistance for type i2 isolates among people i2, change from acute levels in isolates of type i3 in a given participant i3, the effects of exogenous IFN treatment i3, and batch to batch variation in IFN in isolates assayed in previous studies i3, batch.

For QVOA isolates, the IC₅₀ was modeled as a mixture of two populations such that:

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\begin{split} p(\mathsf{IC50}_i | \mu_{\mathsf{qvoa},\mathsf{person}_i}, \sigma_{\mathsf{qvoa}}, \mu_{\mathsf{rebound},\mathsf{person}_i}, \sigma_{\mathsf{rebound}}, \phi_{\mathsf{prePost}_i}) = \\ \phi_{\mathsf{prePost}_i} \mathsf{Normal}(\mathsf{IC50}_i | \mu_{\mathsf{rebound},\mathsf{person}_i}, \sigma_{\mathsf{rebound}}) \\ + (1 - \phi_{\mathsf{prePost}_i}) \mathsf{Normal}(\mathsf{IC50}_i | \mu_{\mathsf{qvoa},\mathsf{person}_i}, \sigma_{\mathsf{qvoa}}) \end{split}
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where prePost $_i$ indicates whether isolate i was isolated pre- or post-ATI and and ϕ_{pre} and ϕ_{post} represent the proportion of rebound-like virus present in pre- and post-ATI QVOA isolates.

The hierarchical parameter priors were modeled as:

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\begin{split} \sigma_j &\sim \mathsf{Gamma}(1,0.1) \\ \psi_j &\sim \mathsf{Gamma}(1,0.1) \\ \phi_{\mathsf{pre}} &\sim \mathsf{Uniform}(0,1) \\ \phi_{\mathsf{post}} &\sim \mathsf{Uniform}(0,1) \end{split}
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where j indicates the isolate type (acute, chronic, QVOA, rebound). All α_k , $\beta_{j,k}$ were given flat priors and $\beta_{\text{ifn}} \sim \text{Normal}(0,10)$ and $\beta_{\text{batch}} \sim \text{Normal}(0,10)$.

Posterior probabilities were estimated with 50 Markov chain Monte Carlo chains of 5,000 iterations each using Stan (91).