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## DNA Methylation Signatures Predict HIV Prognosis and Mortality V.1

In 1 collection

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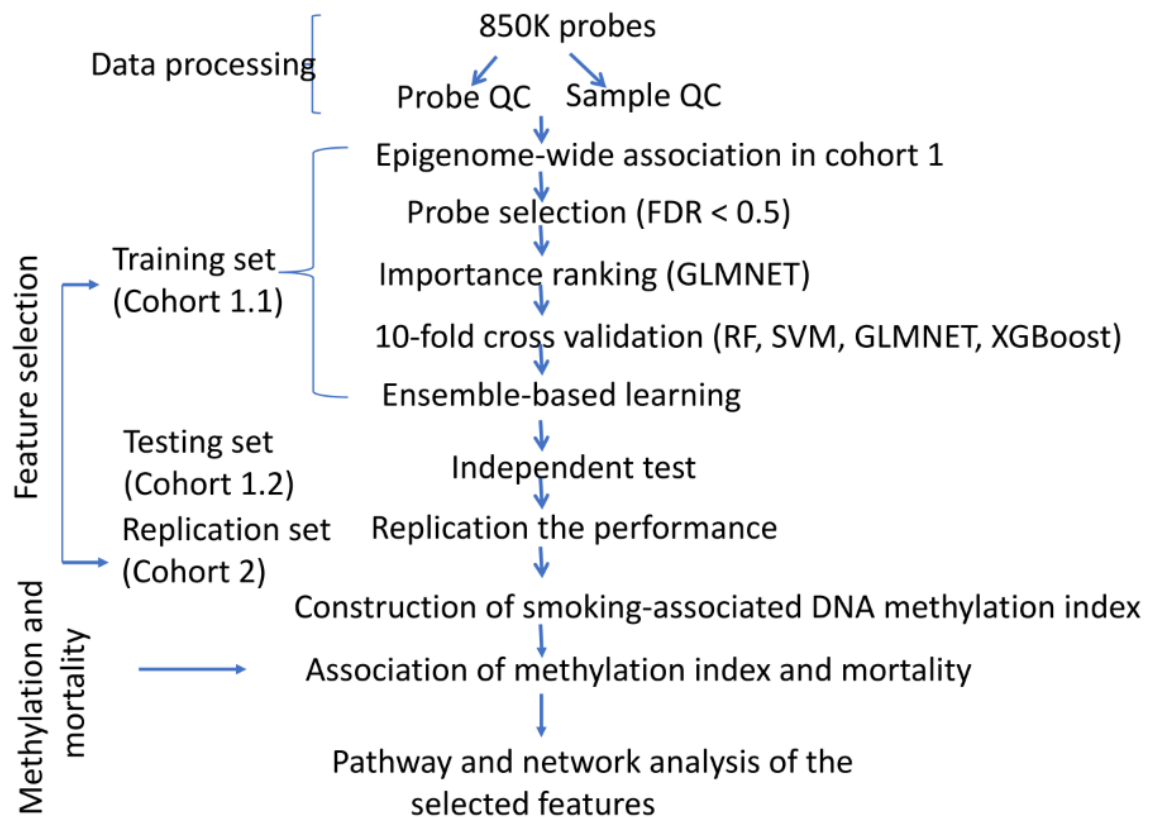
### ABSTRACT

**Background** The effects of tobacco smoking upon epigenome-wide methylation signatures in white blood cells (WBCs) collected from persons living with HIV may have important implications for their immune-related outcomes, including frailty and mortality. The application of a machine learning approach to the analysis of CpG methylation in the epigenome enables the selection of phenotypically relevant features from high-dimensional data. Using this approach, we now report that a set of smoking-associated DNA-methylated CpGs predicts HIV prognosis and mortality in an HIV-positive veteran population.

**Methods and Findings** A total of 1,137 DNA samples were extracted from WBCs in HIV-positive male veterans. Here, HIV prognosis was measured by Veteran Aging Cohort Study Index that was highly correlated with HIV disease frailty. We first conducted two epigenome-wide association studies (EWAS) for tobacco smoking in two HIV-positive sample sets. We then selected smoking-associated DNA methylation features using a machine learning approach to differentiate good and poor HIV prognosis and to predict mortality. The performance of the prediction was validated in an independent sample set. We identified 137 epigenome-wide significant CpGs in meta-EWAS analysis for smoking in HIV-positive samples ( $p < 1.70 \times 10^{-7}$ ). Using an in-house developed bioinformatic pipeline, we selected 698 CpG sites that predicted high HIV frailty [(Area Under Curve (AUC) = 0.73, (95%CI: 0.63 ~ 0.83)] and was replicated in an independent sample [(AUC = 0.78, 95%CI: 0.73 ~ 0.83)]. We further found an association of a DNA methylation index constructed from the 698-CpGs that were associated with a 5-year survival rate [HR = 1.46; 95%CI: 1.06 ~ 2.02,  $p = 0.02$ ]. Interestingly, the 698-CpGs located on 445 genes were enriched on the integrin signaling pathway ( $p = 9.55 \times 10^{-5}$ , False discovery rate = 0.036), which is responsible for regulation of the cell cycle, differentiation, and adhesion.

**Conclusion** We demonstrated that smoking-associated DNA methylation features in white blood cells predict HIV infection prognosis and mortality in a population living with HIV. Feature selection is a powerful approach to identify DNA methylation biomarkers for the prediction of medical outcomes.

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## Epigenome-wide Association Analysis

2 Following protocols of <https://www.nature.com/protocolexchange/protocols/6335>

## Meta-analysis

3 We conducted an EWAS meta-analysis by combining the data from the discovery and replication samples. Effect size and p-values for each probe were obtained from analyses in cohort 1 and cohort 2 samples, respectively. We performed fixed-effects, inverse-variance meta-analysis, with scheme parameters of sample size and standard error by implementing the METAL (ver: 2010-02-08) program, combining summary statistics in two sample sets. We investigated heterogeneity in two sample sets using the I<sup>2</sup>-statistic.


- 4 We developed an in-house ensemble-based feature selection R package, *SmartFeatureSelection* (<https://bitbucket.org/starrcofly/smartfeatureselection.git>) to link smoking-associated CpG sites to HIV outcomes. *SmartFeatureSelection* used a supervised machine learning approach and merges EWAS with step-wise feature selection. The protocol included: 1) filtering of irrelevant probes based on methylation association analysis; 2) consideration of the relative attribution of each probe to rank the importance of the probes; 3) application of multiple machine learning methods to optimize the model; 4) building a model based on greedy ensemble by applying a weighted average of 4 machine learning models; 5) validation of the model in two independent samples.
- Considering the samples were processed at different times and platforms, batch effects were removed using Function `removeBatchEffect` in R `limma` (ver. 3.32.10) library before performing machine learning prediction. To reduce redundant DNA methylation signals and noise for improving the prediction accuracy of HIV frailty, CpG sites with FDR < 0.5 from EWAS in cohort 1 were selected for machine learning. The samples in cohort 1 were randomly divided into a training set and a test set with ratio of 8:2. We first built a model using the training set, in which each sample was labeled poor (VACS index >50) or good prognosis (VACS index ≤50). We then tested the model by performing 10-fold cross validation in the testing set, and the best-performed model was tested in an independent replication set.

#### Association of Smoking-associated DNA Methylation Index and Mortality

- 5 To examine whether the select CpG site methylation was associated with mortality, we constructed a methylation index from the selected smoking-associated CpG sites following the previous formula[34]. A separate index was constructed for hypomethylated and hypermethylated CpG sites, respectively.
- The association of the DNA methylation risk index with all-cause mortality was examined by Kaplan–Meier plots and log-rank tests in all samples. Cox regression model was then used to adjust for age, antiretroviral therapy, HIV-1 loads, and CD4 counts. In the Cox regression model, the DNA methylation index score was a categorical variable (using the highest quartiles as the reference category) or a continuous variable (calculating HR for a decrease in DNA methylation by one standard deviation).  $Index_{hypo}$  and  $index_{hyper}$  were evaluated for the prediction of mortality separately.

#### Gene Enrichment Analysis

- 6 Pathway and network analysis was conducted for the selected CpG sites on the nearest genes by employing Ingenuity Pathway Analysis (IPA). For genes with multiple CpG sites, the lowest p-value at the CpG site within a gene was used to represent the gene level significance. Significant pathways were defined at a FDR < 0.05.

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