

Mechanistic Indicators of Childhood Asthma (MICA) Study

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11th SAC Seminar: NEW TRENDS IN CHEMICAL TOXICOLOGY

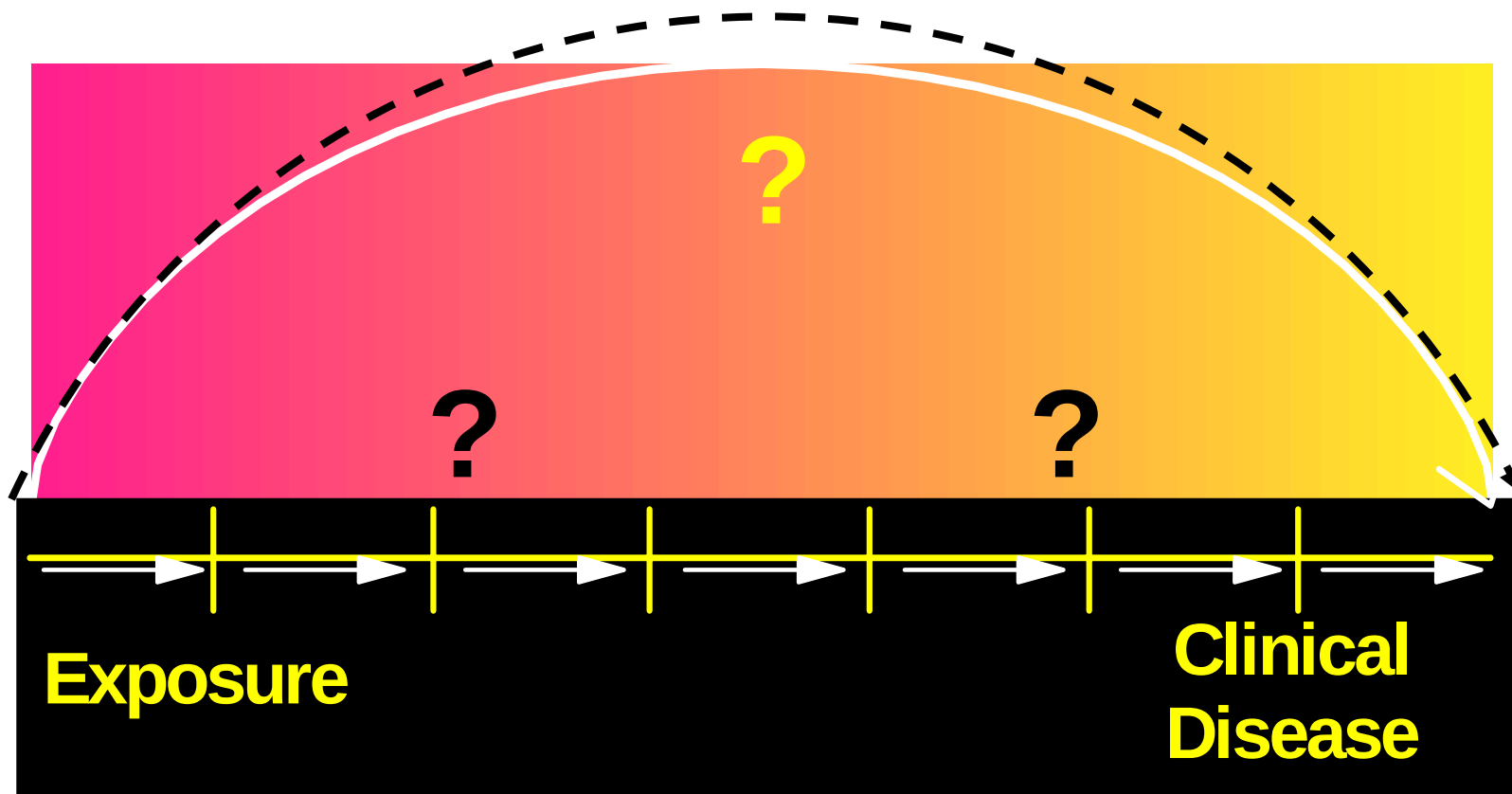
22 – 25 September 2008

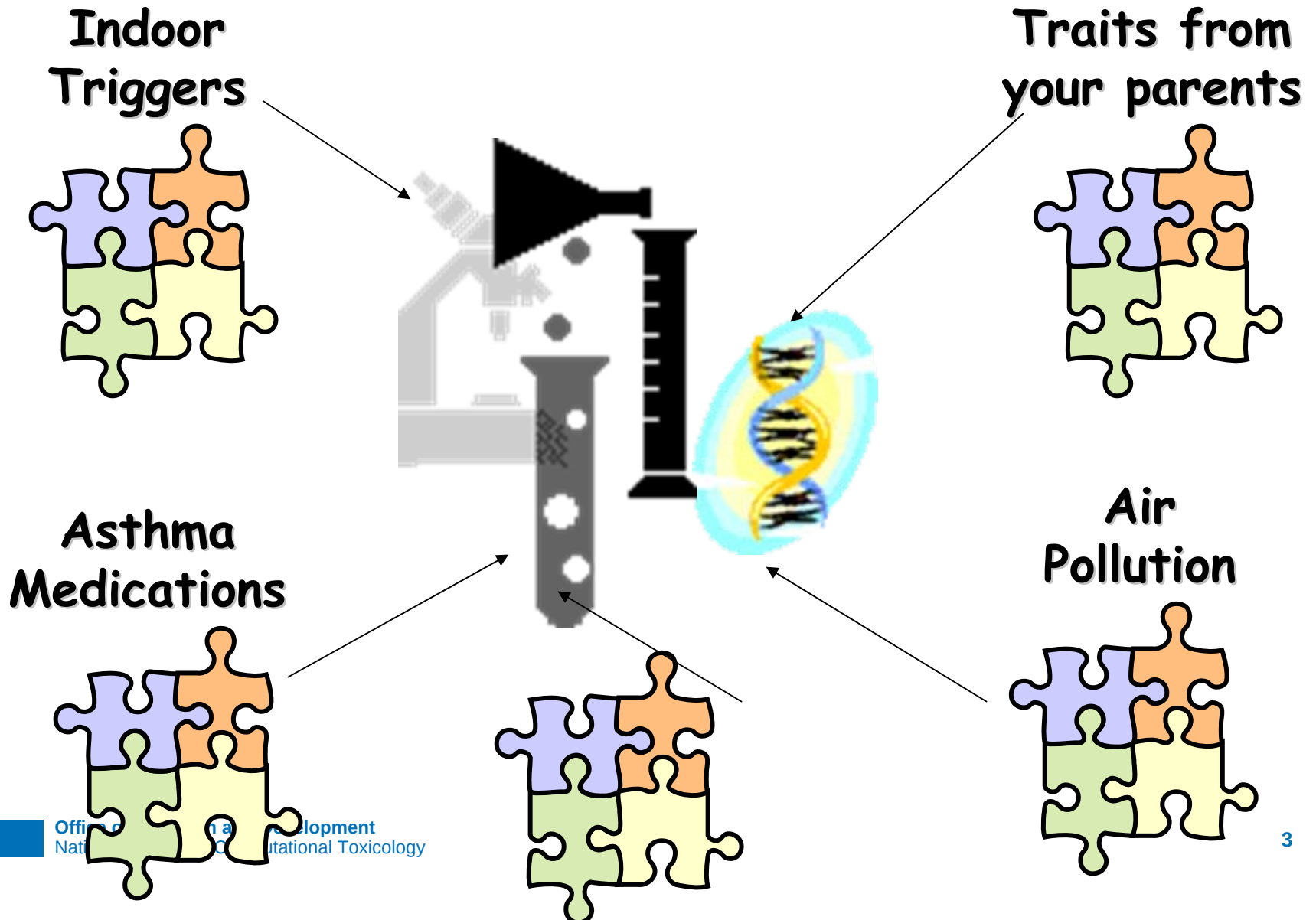
Danilovskaya Hotel, Moscow, Russian Federation

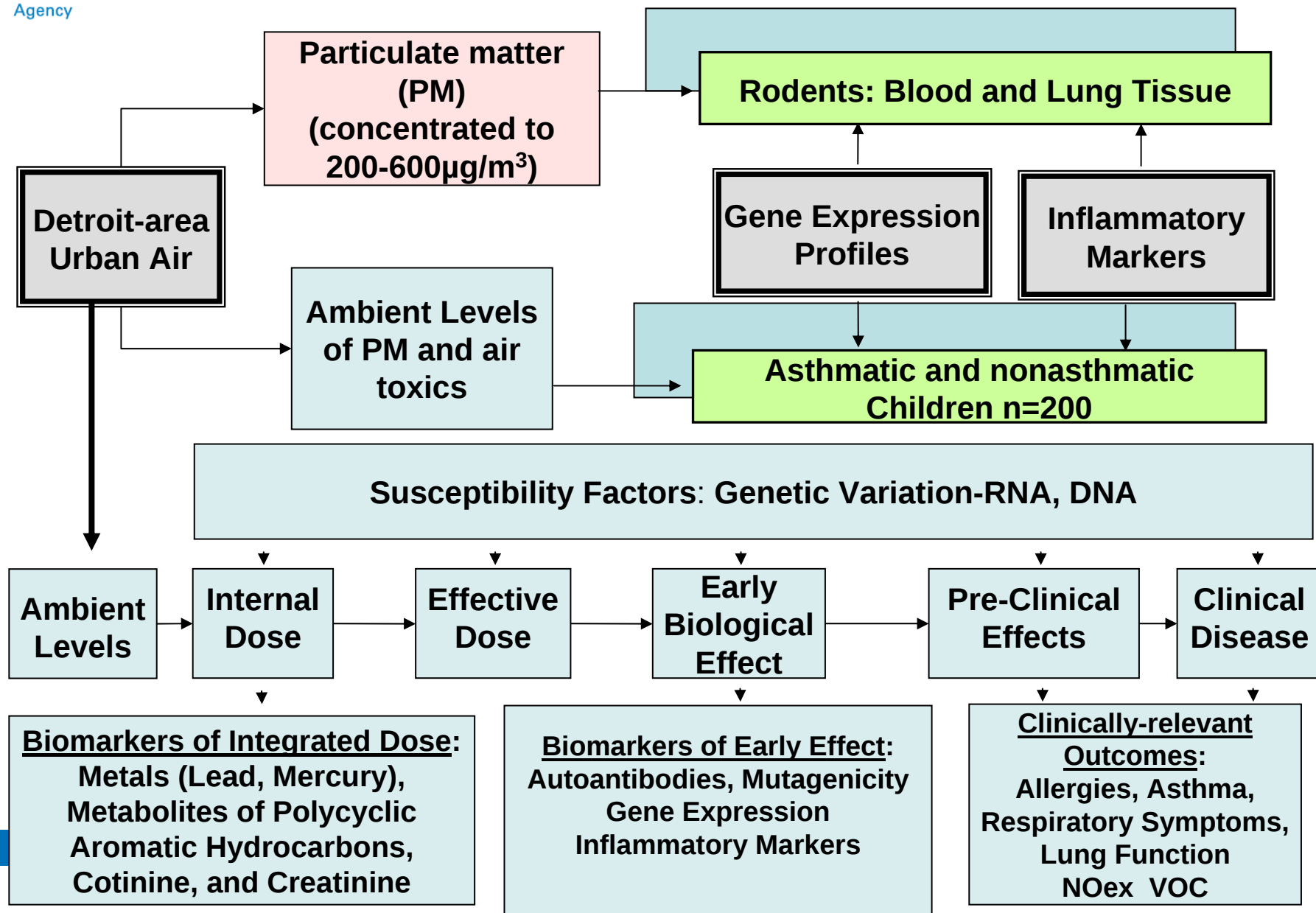


Mechanistic Indicators of Childhood Asthma (MICA)

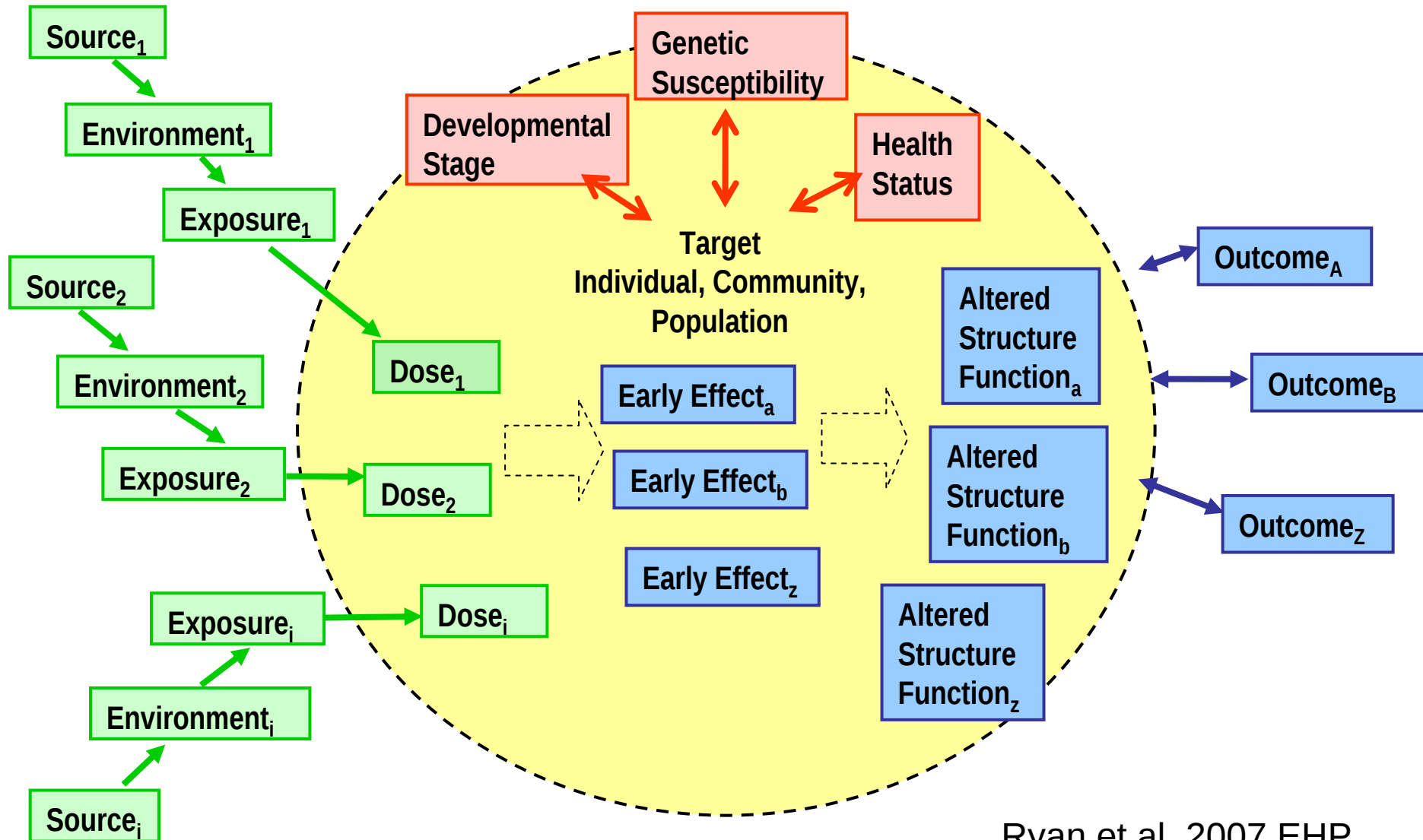
- Apply state-of-the-art technologies to examine the interplay between environmental and genetic factors affecting asthma
- 100 asthmatic and 100 non-asthmatic children, ages 9-12 years (subset of Detroit Children's Health Study cohort)
- Collected multiple types of clinical, demographic, exposure, and gene expression data
- Consider markers of susceptibility, exposure, and effects to analyze and characterize combined risk factors that relate to asthma severity
- <http://www.epa.gov/dears/studies.htm>

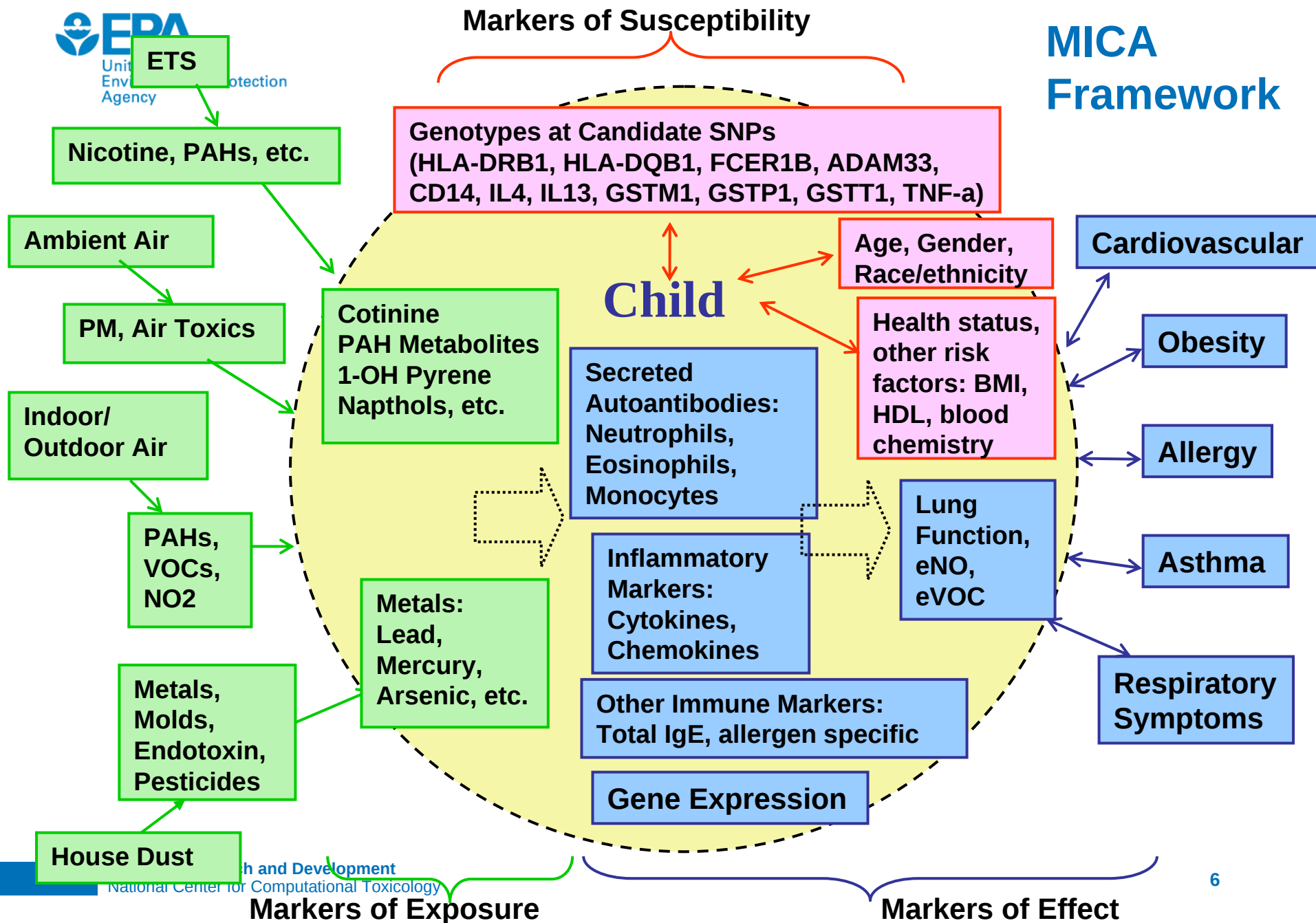




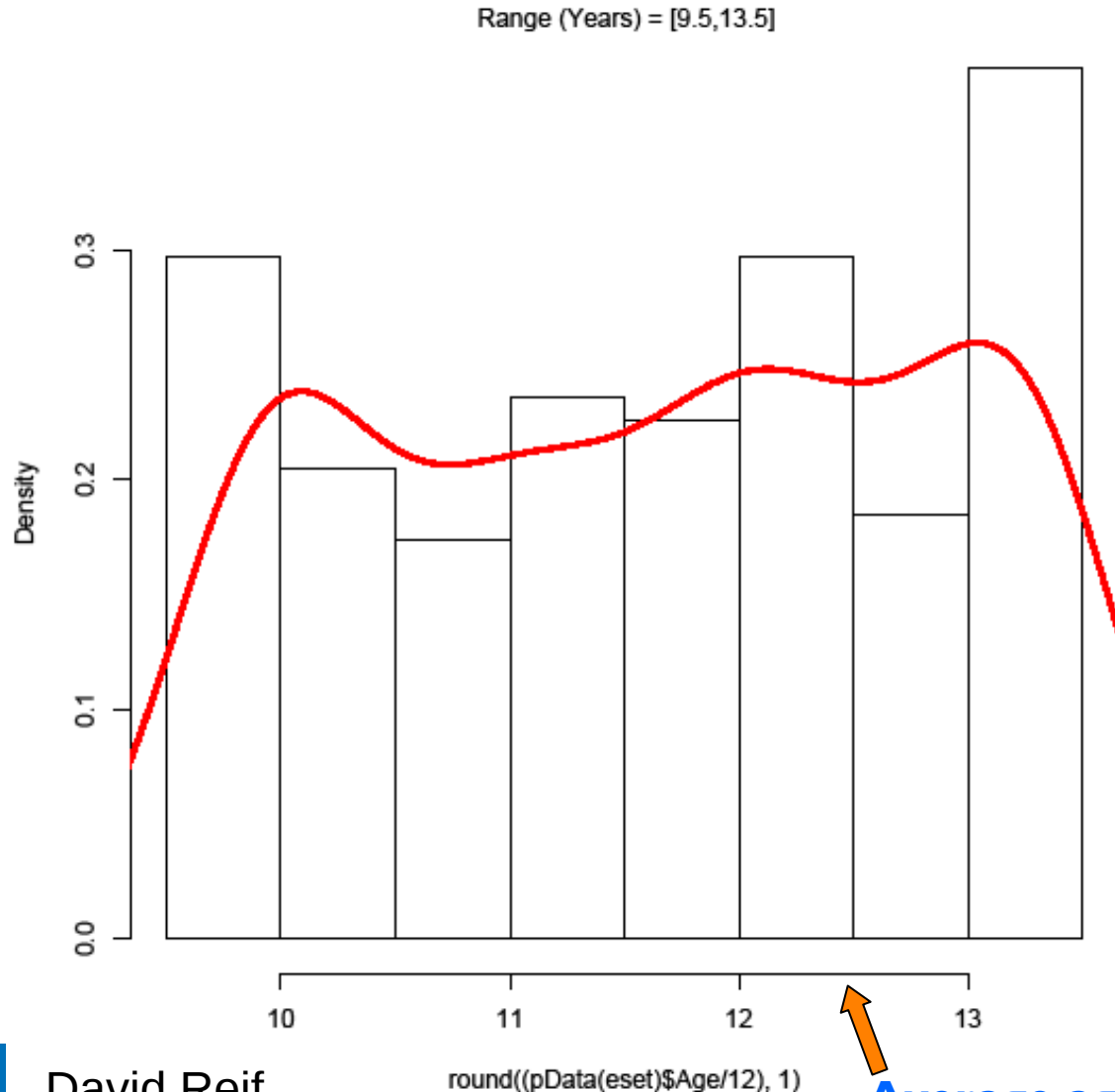


Biomarker Framework





Characteristics of our MICA Study Sample



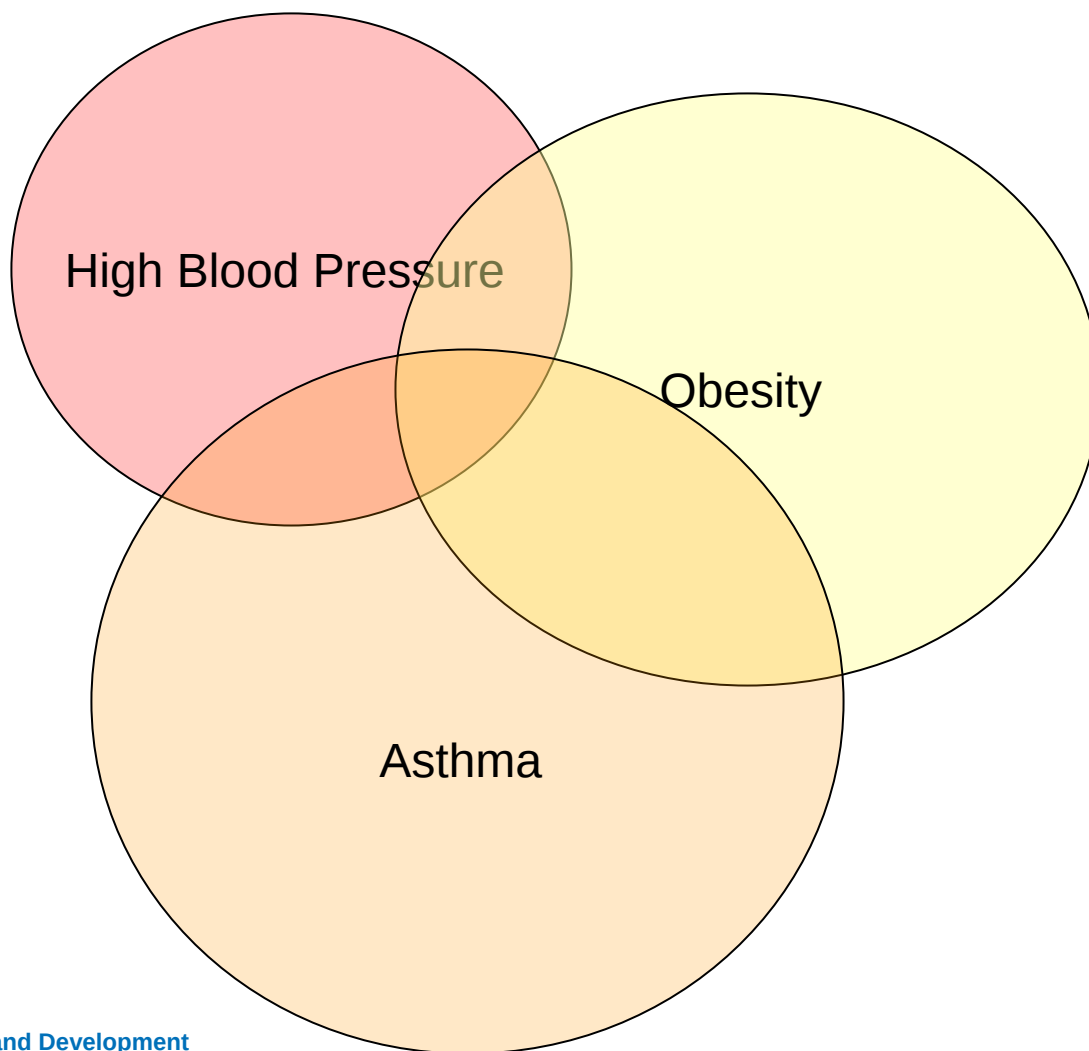
Study sample includes both boys and girls

Study sample is predominantly African-American (>80%)

Study sample includes children diagnosed as **asthmatic** and **non-asthmatic**

Distribution of age across all categories is fairly uniform

MICA Childhood Study Multiple Risk Factors

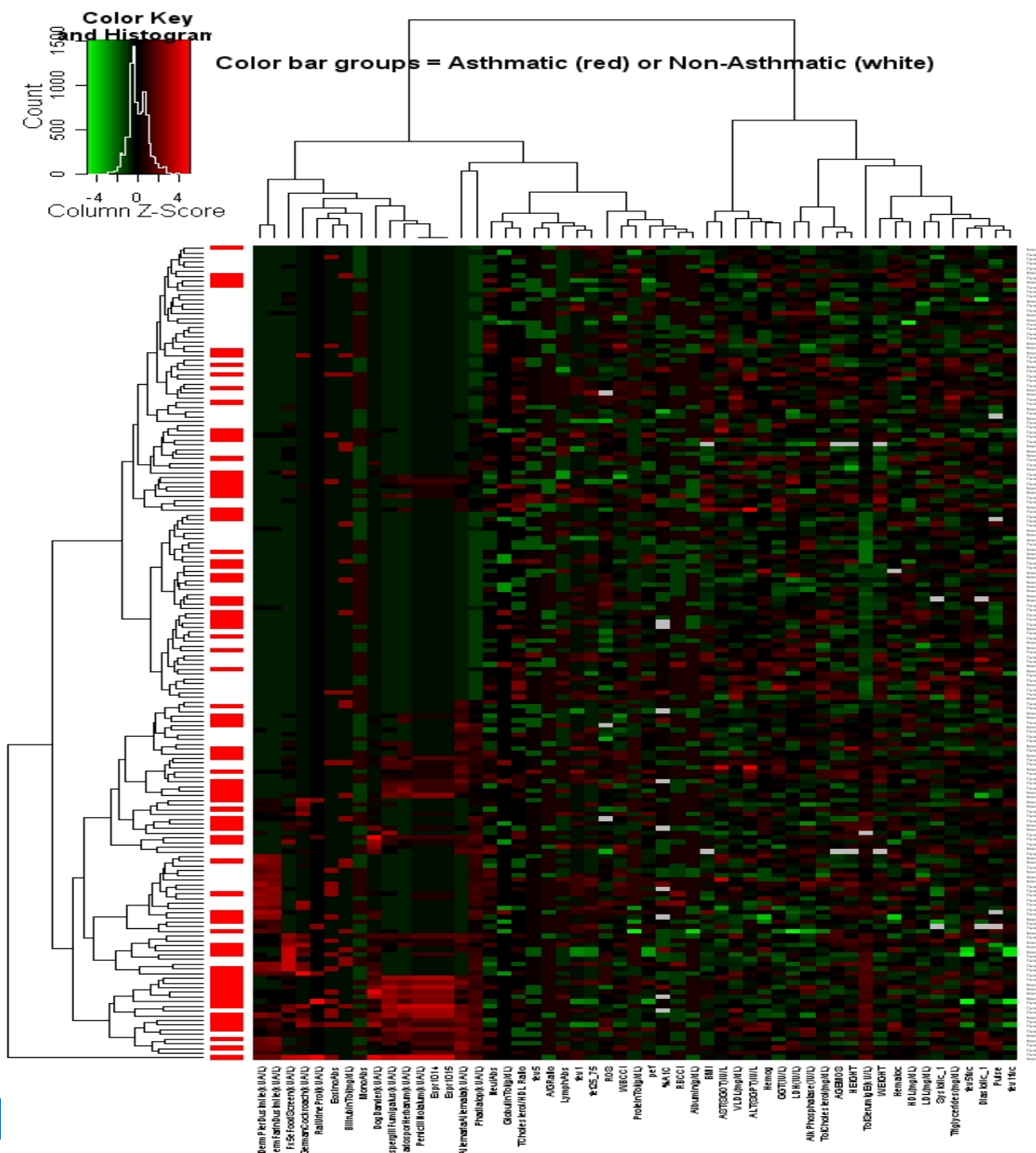


Collaborative NCCT-MICA Goals

- The objectives of the NCCT include advancement of a systems approach to evaluate complex relationships between
 - environmental factors
 - physiological biomarkers
 - health outcomes.
- NCCT collaborating to apply advanced statistical and machine learning methods to evaluate biomarker data collected in MICA
- Contribution of the NCCT component:
 - analyze genetic and gene expression data
 - use a systems biology approach to put data into framework for evaluating ecogenetics

Data Analysis Methods

- Traditional statistics: linear regression, logistic regression, ANOVA, linear discriminant analysis.
- Machine learning: recursive partitioning trees, bootstrap aggregation (bagging) techniques, evolutionary computation-optimized classifiers, multifactor dimensionality reduction, random forests.
- Bioinformatics: protein interaction databases, knowledge (literature) mining tools, biological pathway database and inference software.
- Graphical approaches: cluster diagrams, expression “heat” maps, dendrograms, overlaid scatter plots (both exploratory and summary), distributional “violin” plots, regression plots.

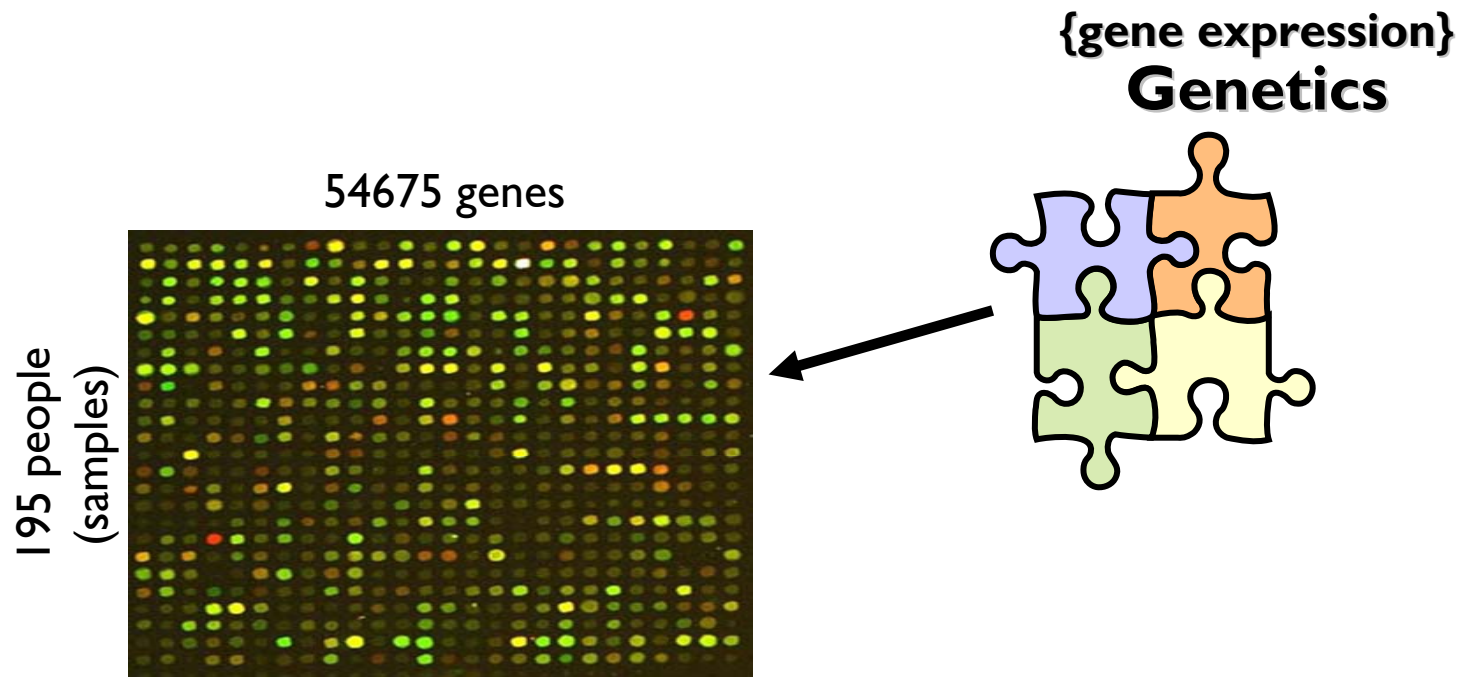


MICA: Physiological Markers/ Health Status

Red – Asthmatic

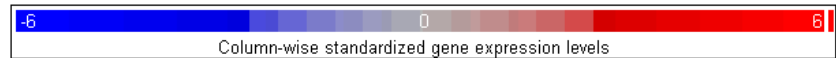
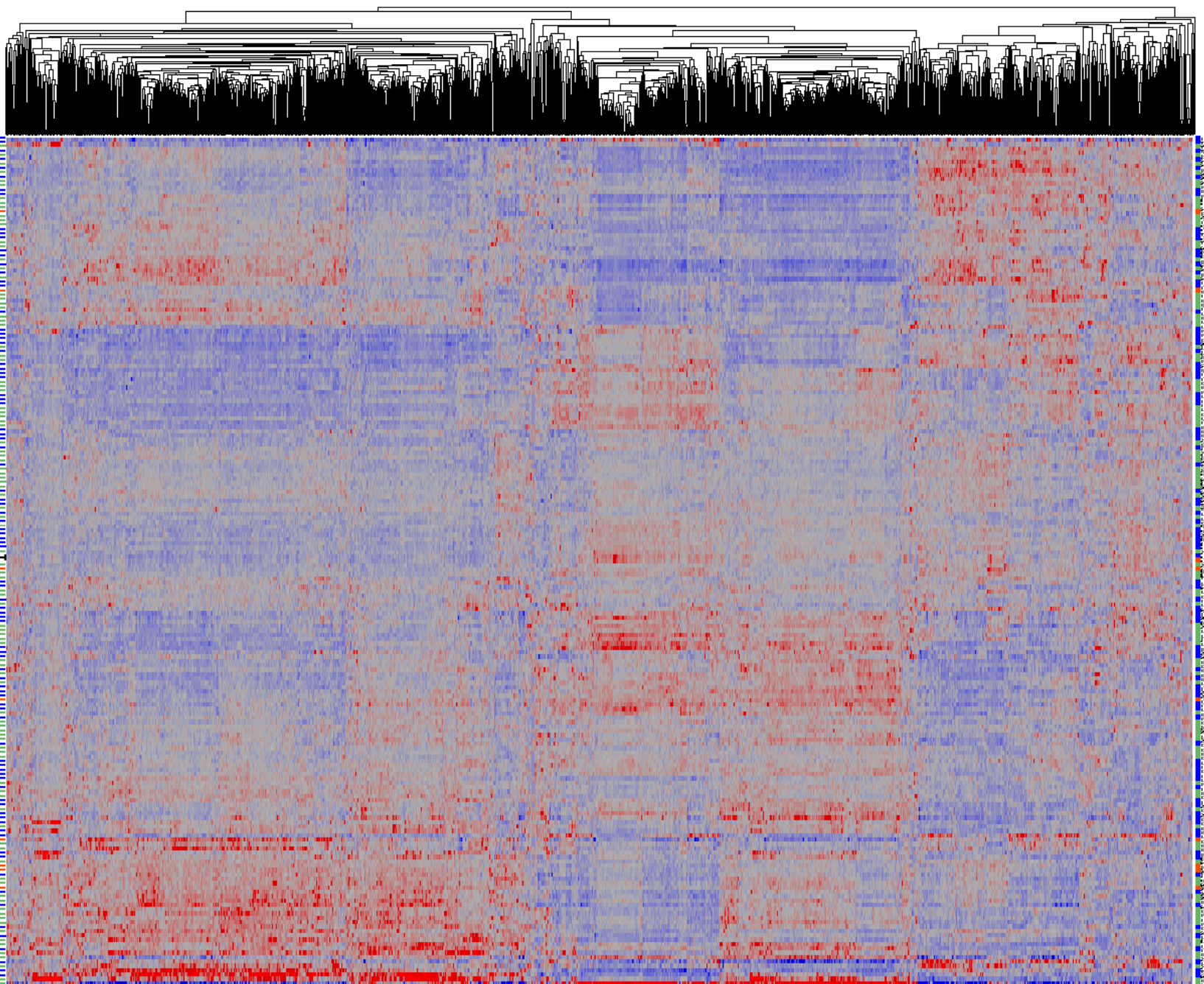
White – Non Asthmatic

Gene expression measured using oligonucleotide microarrays



IMPORTANT: Total RNA was taken from whole blood samples in the absence of any deliberate experimental perturbation.

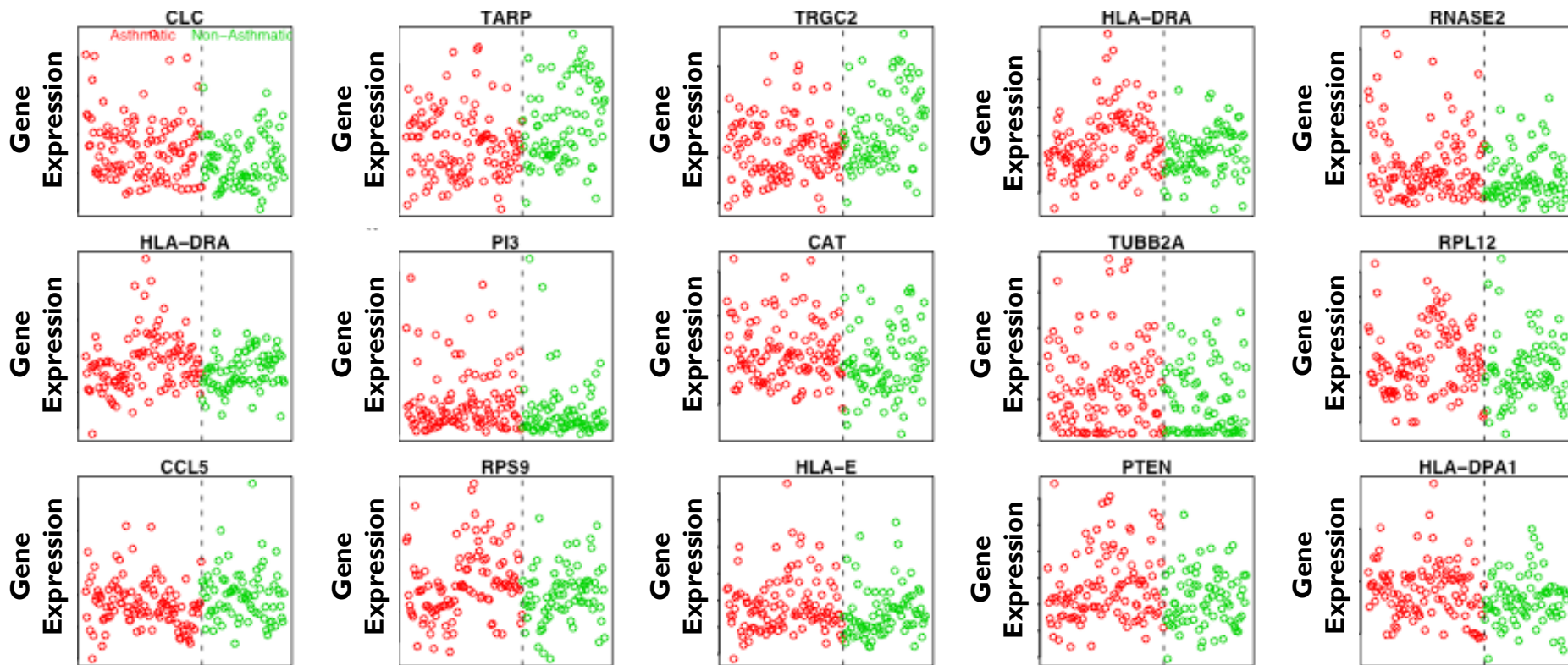
Asthma (D6DrDiagnosis)
■ (Unknown)
■ 0: No
■ 1: Yes



Can Gene Expression Distinguish Subtypes of Asthmatics?

- To identify subtypes, first applying an unbiased (i.e. without knowledge of asthma status) analysis to assess association between gene expression data and information on clinical, demographic, and exposure indicators.
- Next, select only gene expression probe sets that are significantly correlated with at least one of the demographic, clinical, or exposure indicators.
 - This filtering method prevents selecting only genes whose expression is associated with broadly-defined, imperfect asthma diagnoses.
- Examination of the genes differentiating asthma subtypes in this context highlights mechanistic genomic etiologies underlying the disease. These include subtypes of asthma characterized by
 - Patterns of gene expression associated with immune over-stimulation and household allergy exposures
 - Combinations of genomic biomarkers with demographic factors such as gender

Why not just do the usual “here are some main-effect genes that discriminate asthmatics versus non-asthmatics”?

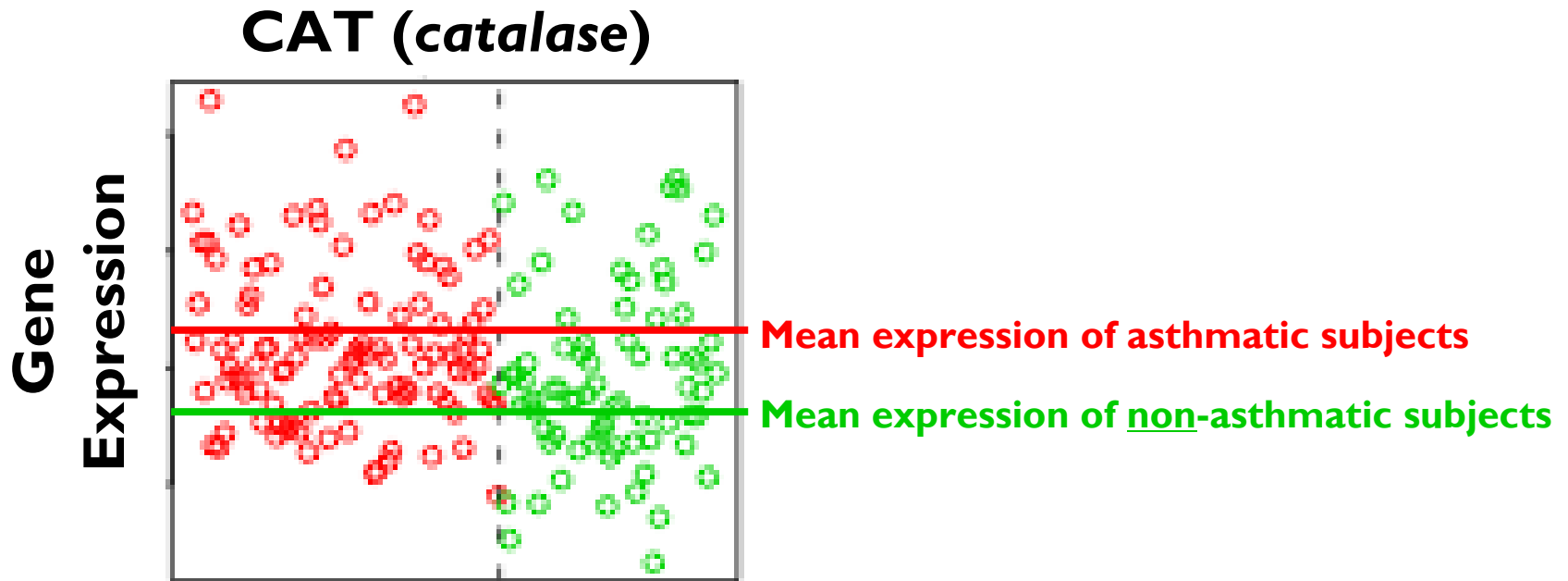


Asthmatic
Non-Asthmatic

Does **not** address gene expression profiles associated with subtypes of asthma

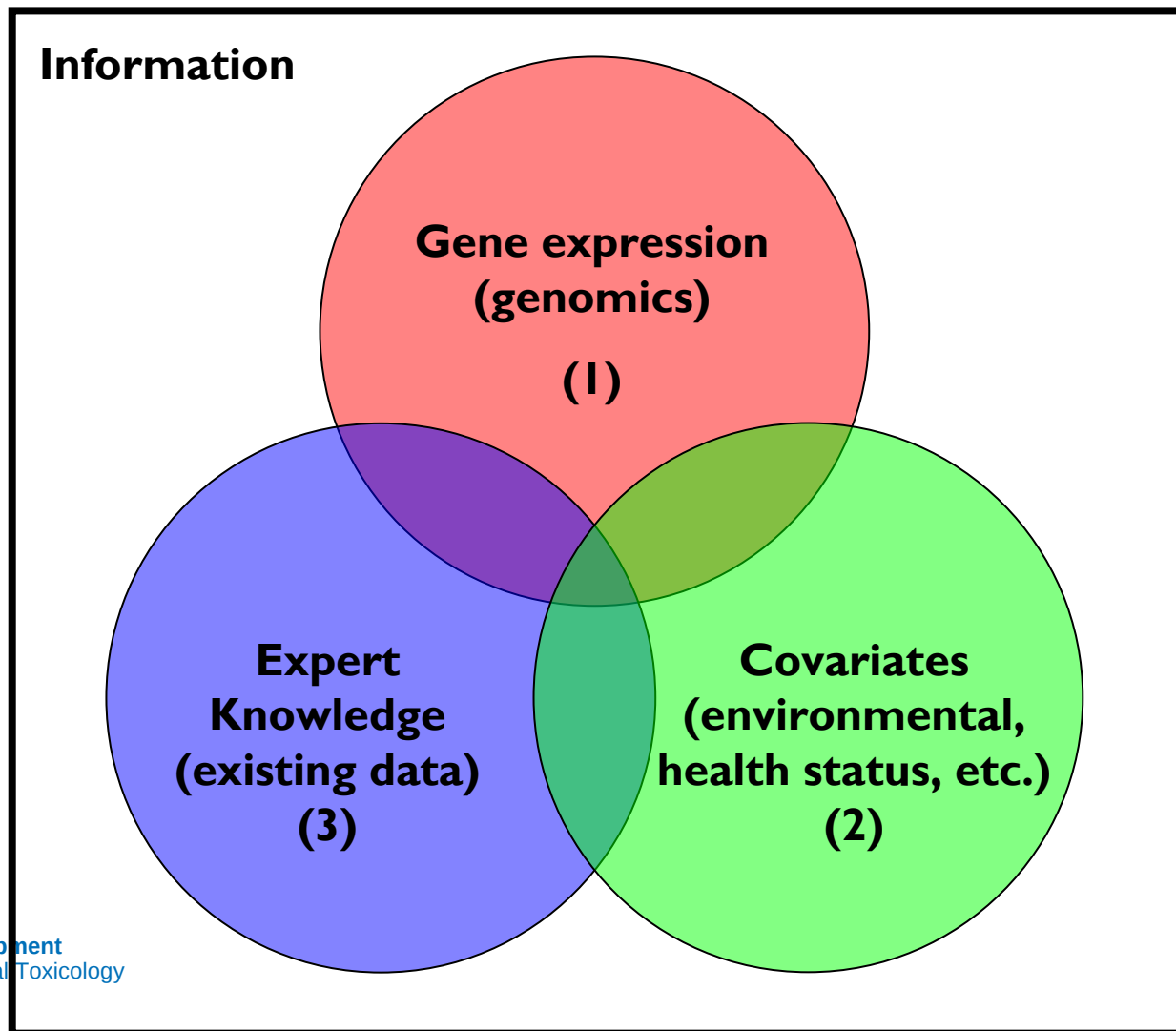
Does **not** link gene expression with other biomarkers or covariates (Where is the context?)

Why not just do the usual “here are some main-effect genes that discriminate asthmatics versus non-asthmatics”?

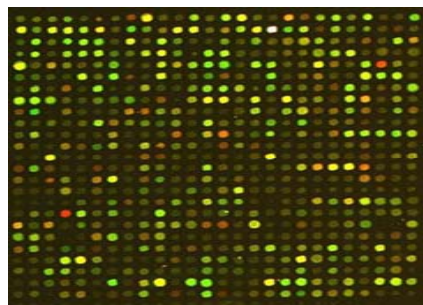


Reliance on group means **ignores** the complexity of asthma etiology

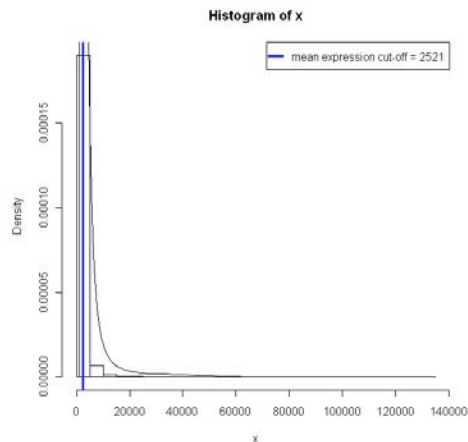
The ultimate goal is to glean mechanistic information regarding asthma subtypes



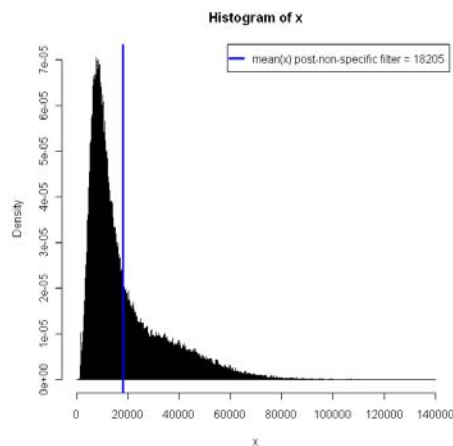
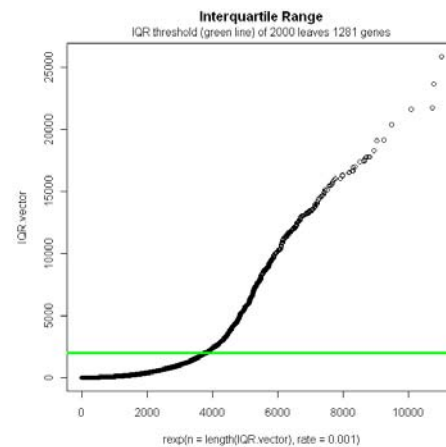
Analysis pipeline for the gene expression data



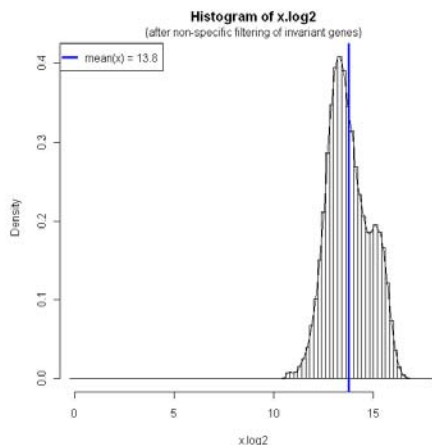
**Gender
adjustment**



**IQR
filtering**

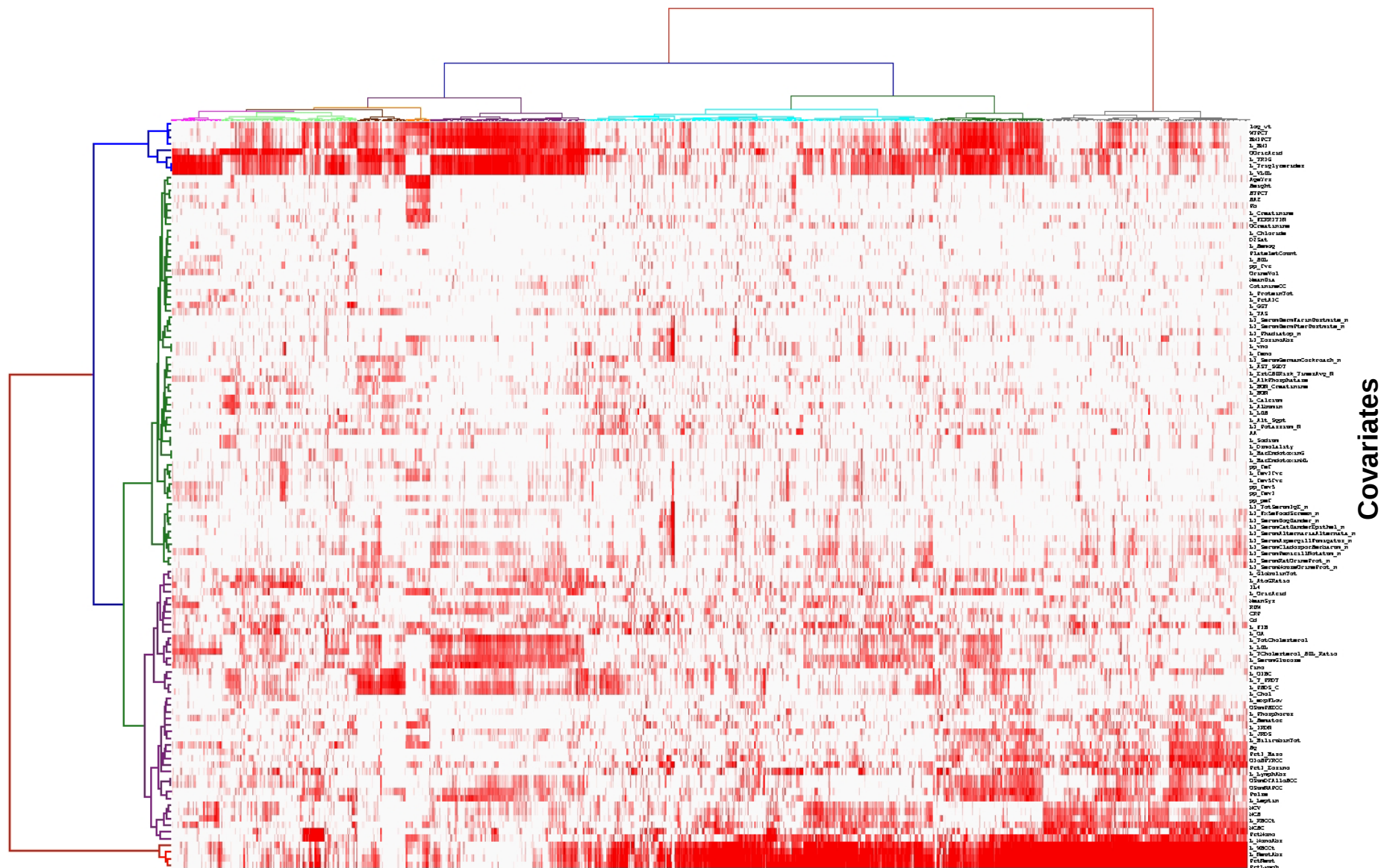


**Log₂
transformation**

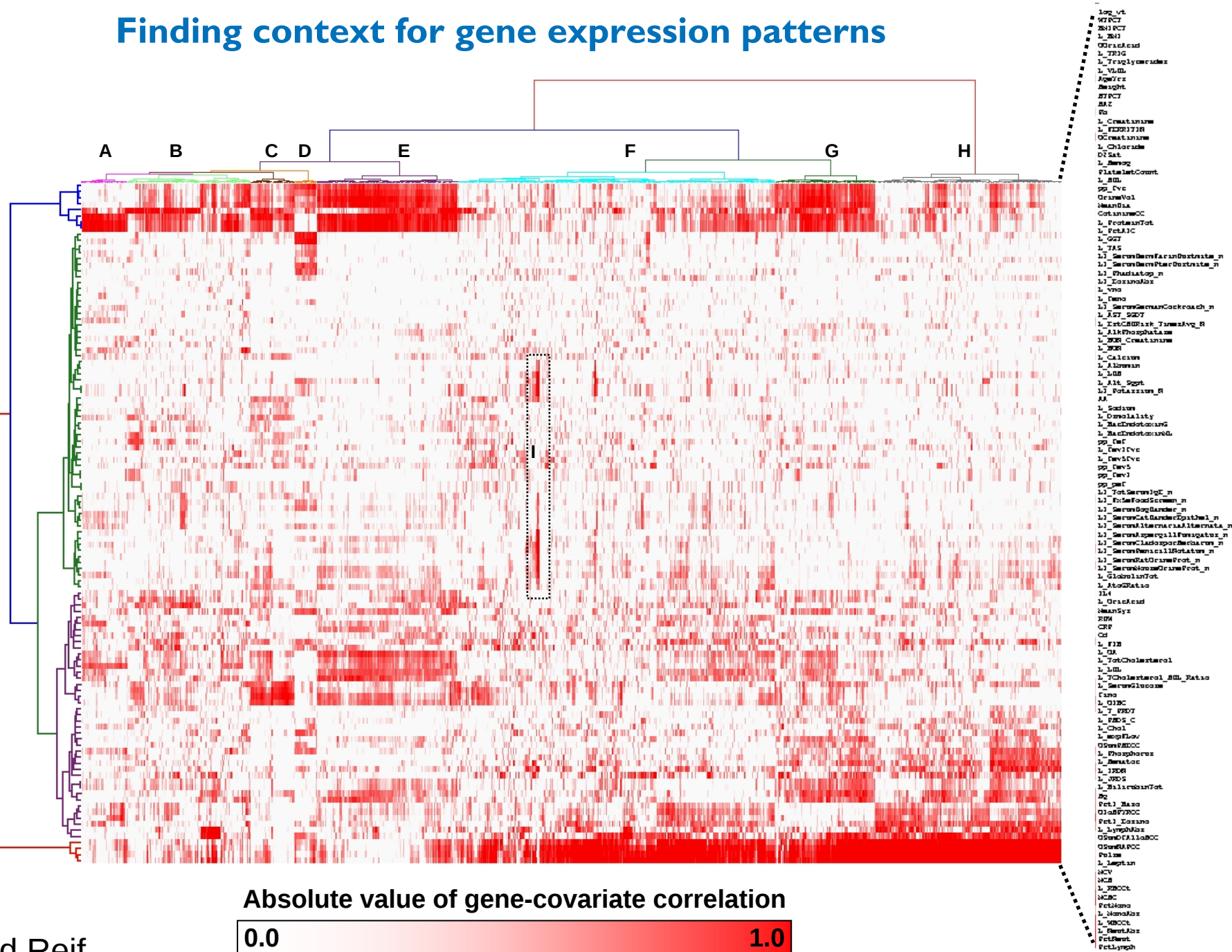




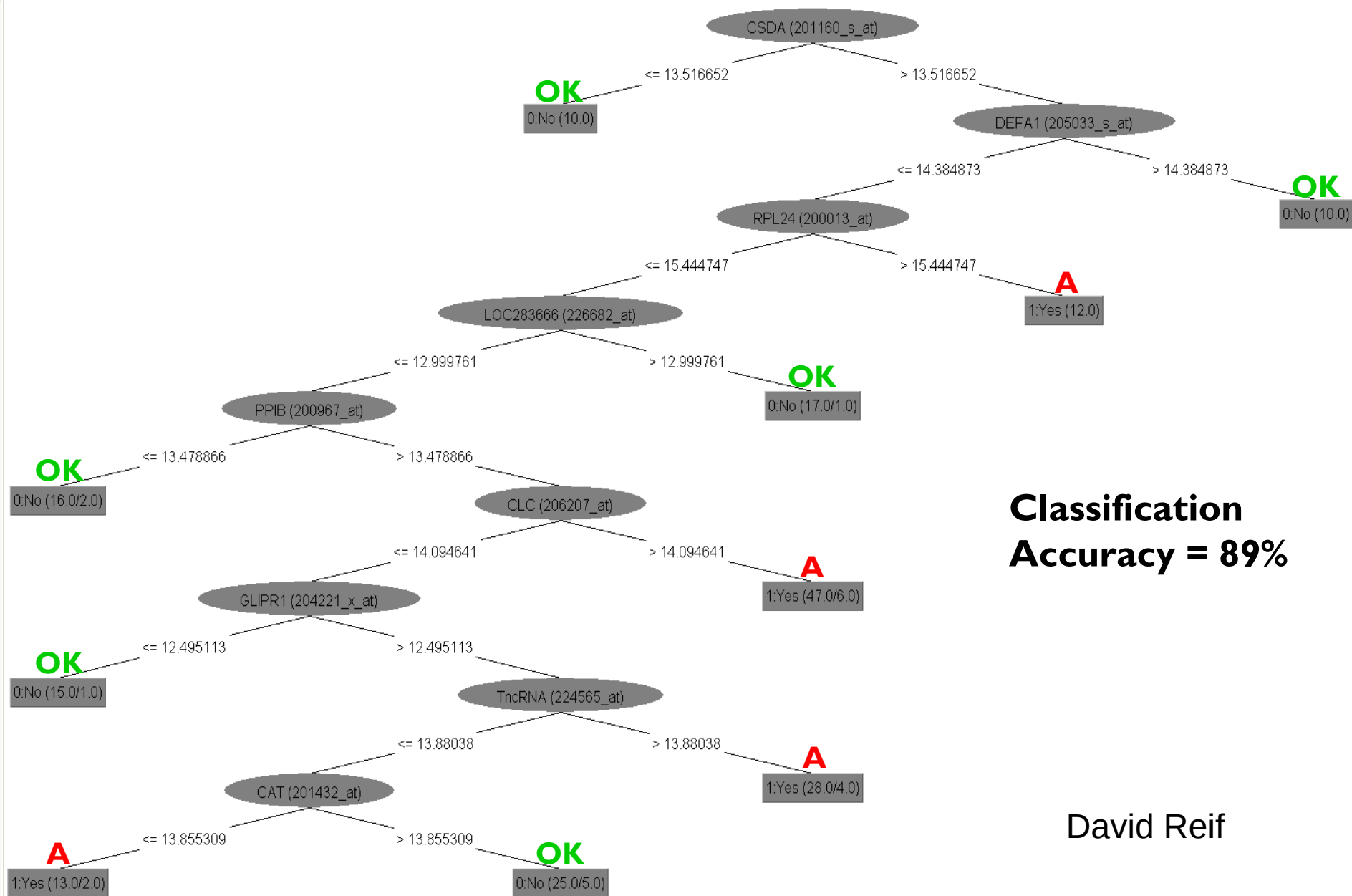
Results in a list of genes having significant correlation with at least one MICA covariate



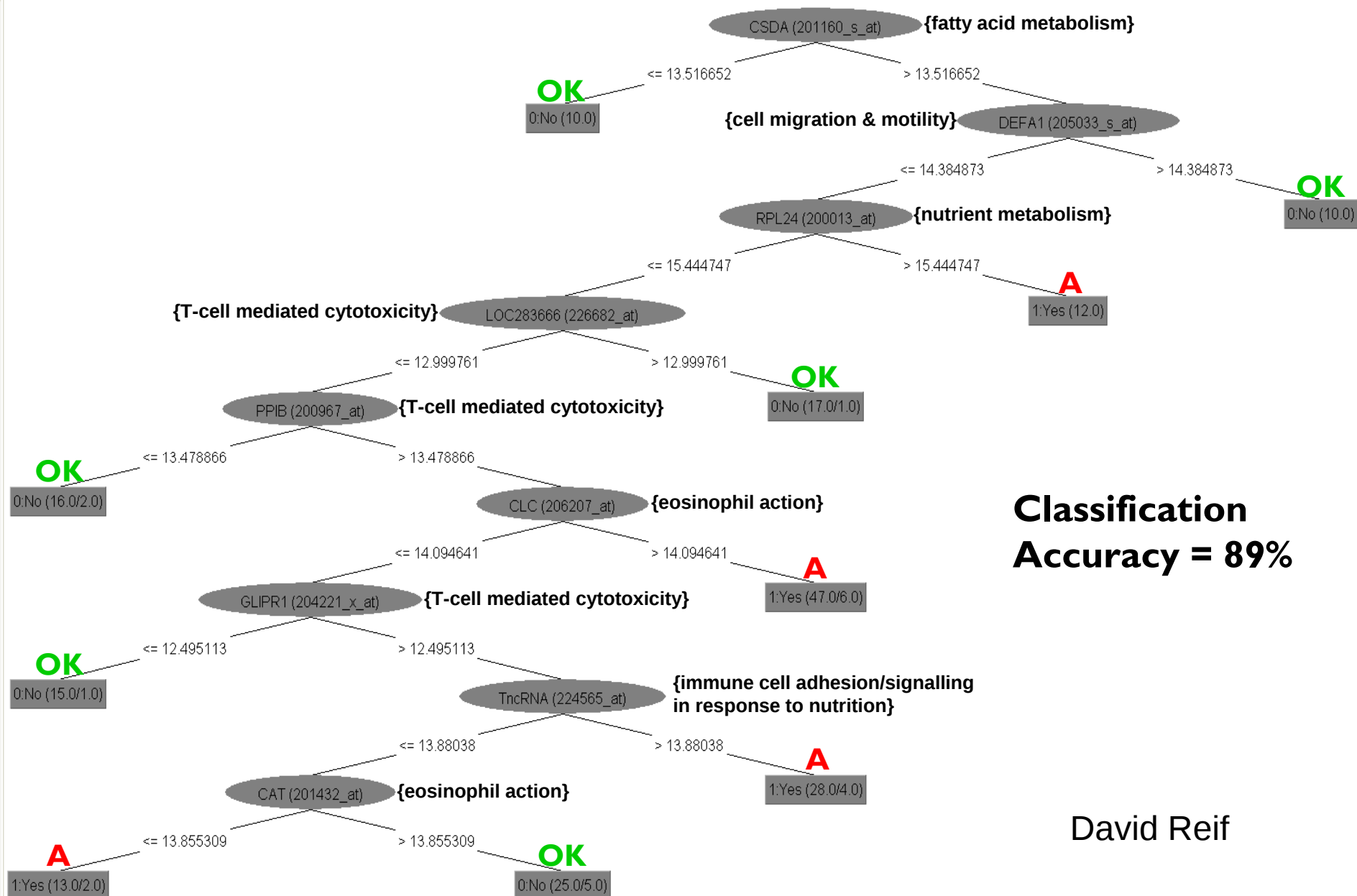
Finding context for gene expression patterns



Can we discriminate between subtypes of asthma?

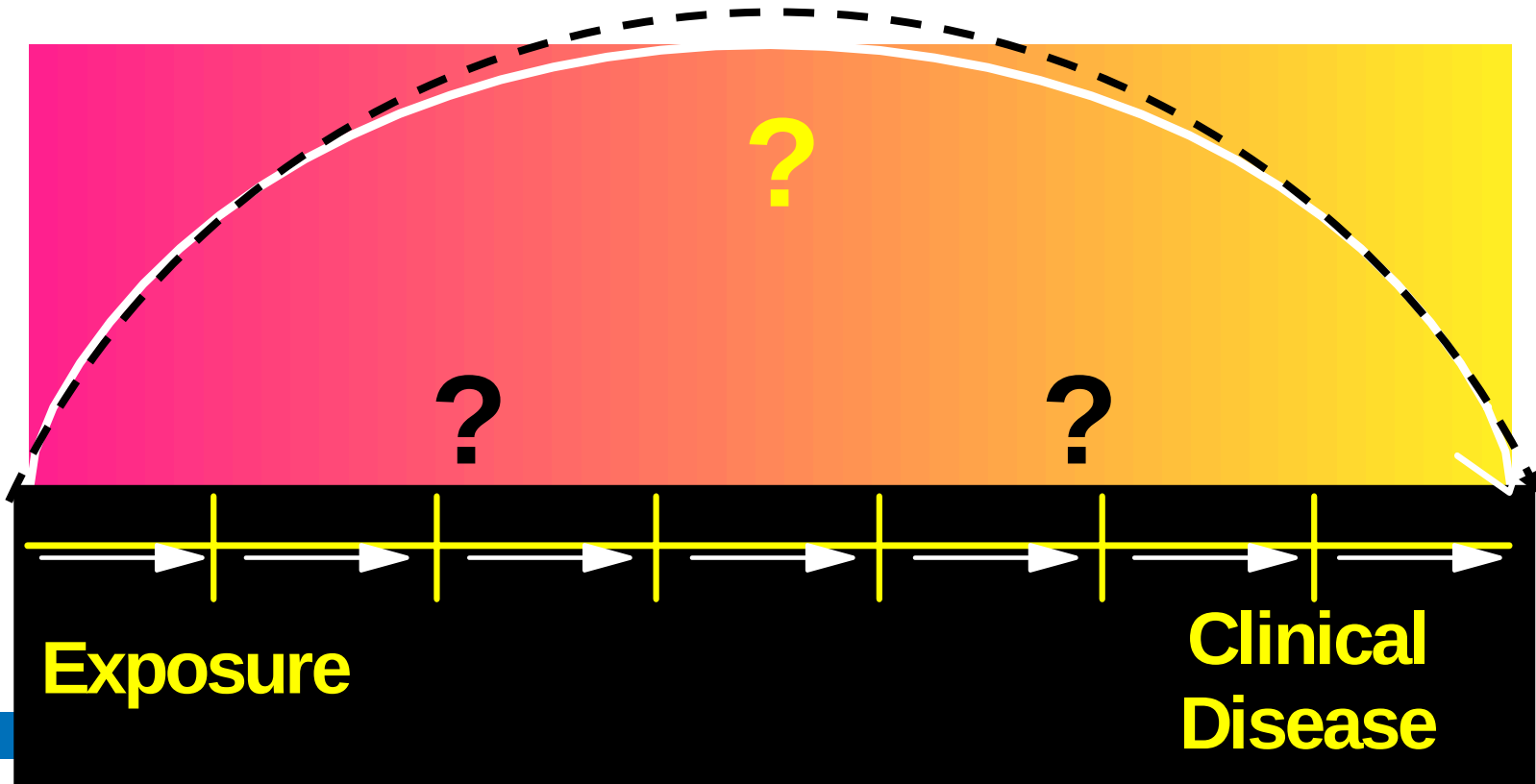


Can we leverage covariate information to put gene expression asthma classifier results in context?



Summary

- Integration of diverse set of exposure, effects and susceptibility measures
- High-data content technologies, elucidating the genetic and environmental basis for toxicity and disease



Acknowledgements

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Disclaimer

Although this work was reviewed by EPA and approved for presentation, it may not necessarily reflect official Agency policy.