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# Penetrance estimates for incidental genomic findings in ACMG-59

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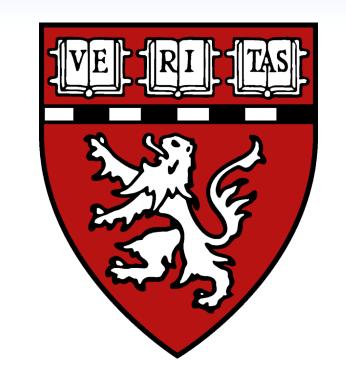
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# Penetrance Estimates for Incidental Genomic Findings

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## INTRODUCTION

(Genetic Testing and Relevant Datasets)

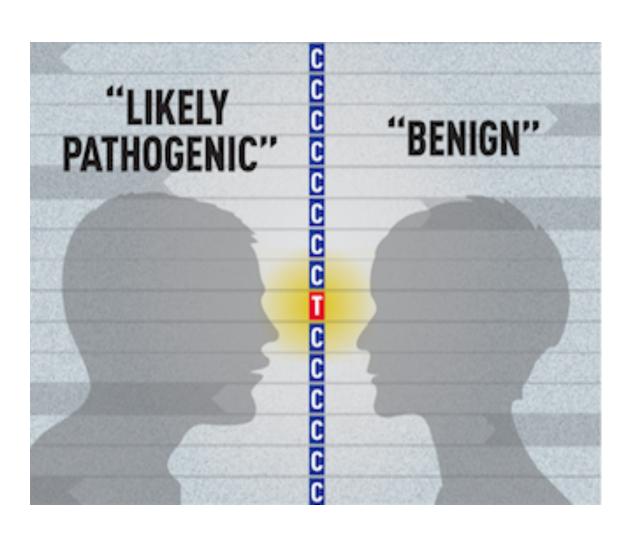
Genetic testing: a difference from the reference genome (variant) may indicate disease.

Incidental finding: variant in gene <u>unrelated</u> to diagnostic indication that prompted sequencing.

-Due to multiple testing and low priors, these typically have <u>high rates of false positives</u>, so we normally don't report them.

ACMG (American College of Medical Genetics & Genomics): recommends an exception for 56 genes thought to be more indicative of disease.

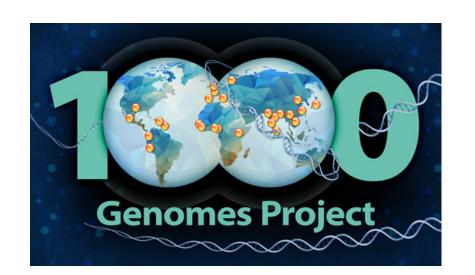




1000 Genomes Project: contains whole-genome sequence data for 2,504 healthy adults from diverse ethnic populations.

**ExAC:** aggregates population-level data from 60,706 diverse human whole-genome sequences.

ClinVar: central repository of interpretations for genetic variants (benign vs. pathogenic).





# **OBJECTIVES**

- 1. Develop an ETL workflow for extraction, transformation, and loading of genomic and interpretation data from relevant sources.
- 2. Evaluate variant distribution across a healthy, diverse cohort (1000 Genomes).
- 3. Estimate plausible penetrance ranges for the ACMG recommendations.

# PENETRANCE MODEL-

 $Penetrance = P(D|V) = \frac{P(D) * P(V|D)}{P(V)} = \frac{(prevalence)(allelic heterogeneity)}{(allele frequency)}$ 

where D = disease, V = any variant

Penetrance: Probability of developing disease, given a positive genetic test result.

Prevalence: Proportion of general population with disease.

Allelic Heterogeneity: Proportion of diseased population with a pathogenic variant.

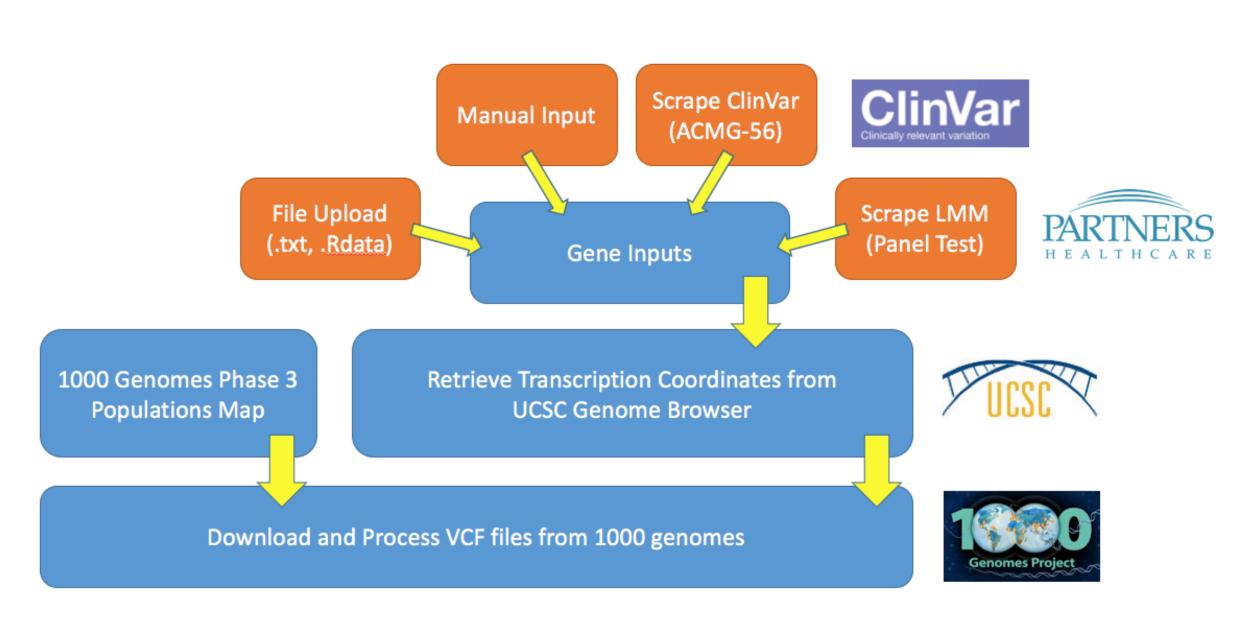
Allele Frequency: Proportion of general population with a pathogenic variant.

# METHODS & WORKFLOW

#### ETL for Datasets

Pipeline + UI using R/Shiny/Markdown

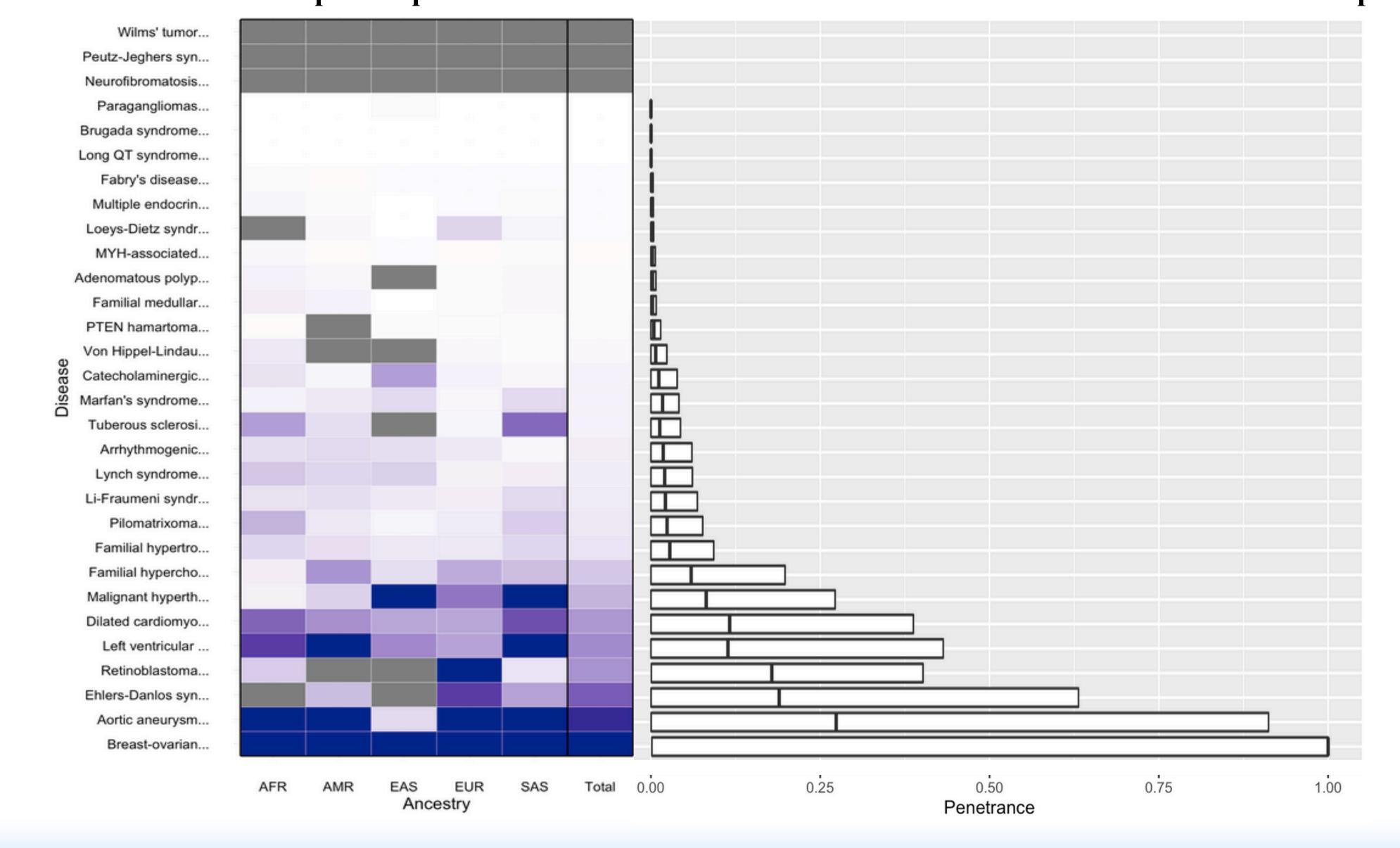
- 1. Extract: query UCSC Genome Browser for gene regions and retrieve corresponding VCF files from 1000 Genomes. Download ExAC manually from gene-level searches.
- 2. Transform: separate variants with multiple alternates; convert genotypes to allele counts, collect missense variants.
- 3. Load: Stage and merge final data objects.



https://github.com/jamesdiao/2016-paper-ACMG-penetrance

# **KEY FIGURES**

Heatmap + Barplot: Penetrance Estimates are Low and Variable between Ancestral Groups



## CONCLUSIONS

- 1. High counts: 40-80% of individuals have an incidental finding under ACMG guidelines, far higher than empirical disease prevalences.
- 2. Clustered distribution: by ethnicity AFR (African) have the most findings, EAS (East Asian) have the fewest.
- 3. High sensitivity: findings dominated by a few high-frequency variants.
- 4. Very low penetrance estimates:

Out of the 30 diseases (22 with data):

- (a) 20 have max theoretical penetrance < 50%
- (b) 12 have max theoretical penetrance < 5%
- 5. High uncertainty around parameters: translates into very large errors bars.
- -This is a preliminary "letter-of-the-law" evaluation and does <u>not</u> yet demonstrate real-world effects on patients.

## **NEXT STEPS**

### 1. Identify questionable variants:

- (a) high-frequency (common findings)
- (b) highly enriched in 1 ethnic population.
- 2. Validation with empirical penetrance values and other sequencing datasets (e.g. gnomAD).
- 3. Model biases in parameter estimates (prevalence, pathogenicity, etc.)
- 4. Confer with clinical collaborators
  to determine alternate protocols at Laboratory of
  Molecular Medicine and Partners HealthCare.

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