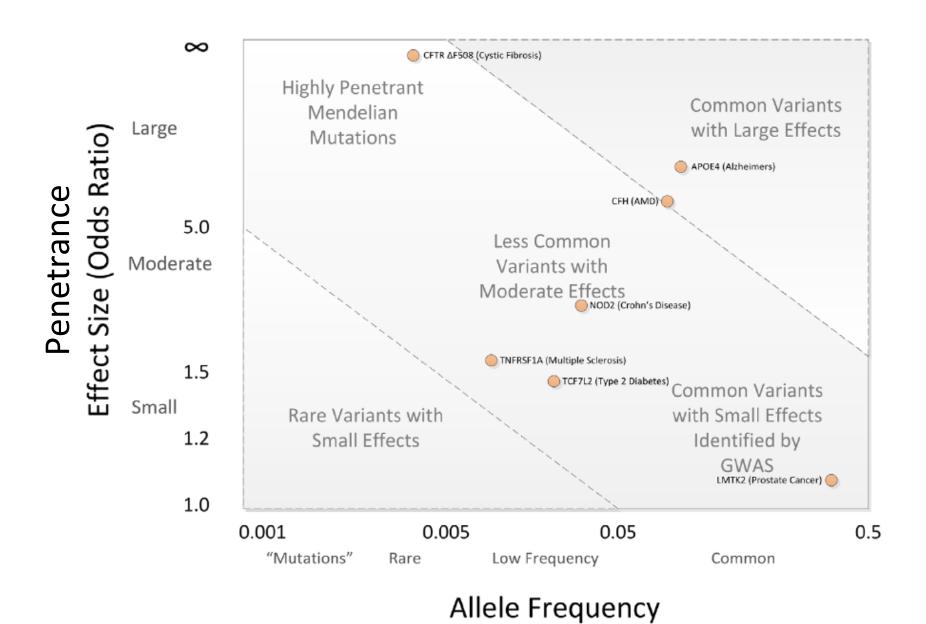
# Genome Wide Association Studies

# What are we trying to do?

Correlate changes in genome with phenotype of interest.

# **GWAS!**



# How tell if a variant is correlated to a phenotype?

- Correlation
- Probability
- Odds ratio/Relative Risk

# Different kinds of Phenotypes

- Categorical
  - Have diabetes or don't
  - Have phenylketonuria or don't
- Quantitative
  - Gradient of disease or biomarker of interest
    - Age of onset of Alzheimer's
    - Blood Pressure
    - HDL concentrations

## Covariance and Correlation

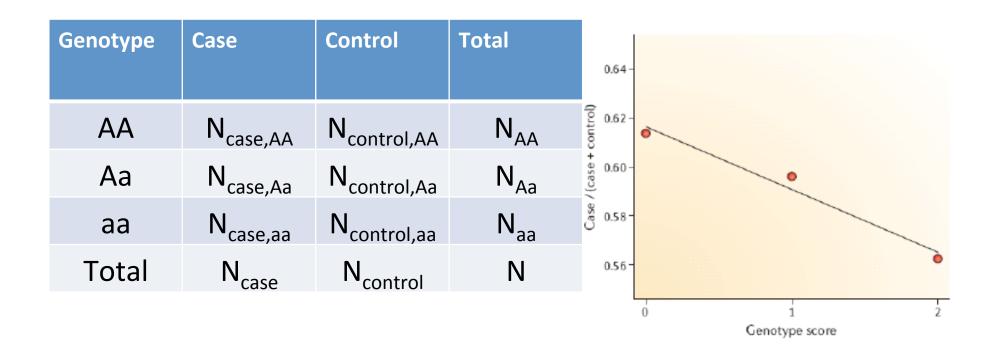
#### Want:

- Degree of association between 2 variables, X &Y
- Given observations  $x_1,...,x_n$  and  $y_1,...,y_n$ 
  - Covariance (sum of O-E divided by number -1)

$$\frac{\sum_{i=1}^{n} (x_i - \overline{x})(y_i - \overline{y})}{n-1}$$

- Correlation coefficient 
$$r = \frac{\sum_{i=1}^{n} (x_i - \overline{x})(y_i - \overline{y})}{\sqrt{\sum_{i=1}^{n} (x_i - \overline{x})^2 \sum_{i=1}^{n} (y_i - \overline{y})^2}}$$

# Case/Control Study (Categorical)



X<sup>2</sup> test with 1 or 2 df (1 if single allele, 2 if genotype)

$$\chi^{2} = \sum_{i=1}^{n} \frac{(O_{i} - E_{i})^{2}}{E_{i}}$$

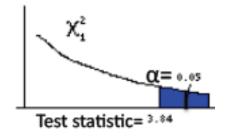
 $O_i$  = observed frequency for  $i^{th}$  outcome (the value can be read off of the contingency table)

 $E_i$  = expected frequency for  $i^{th}$  outcome

(the value can be obtained as described in the previous slides)

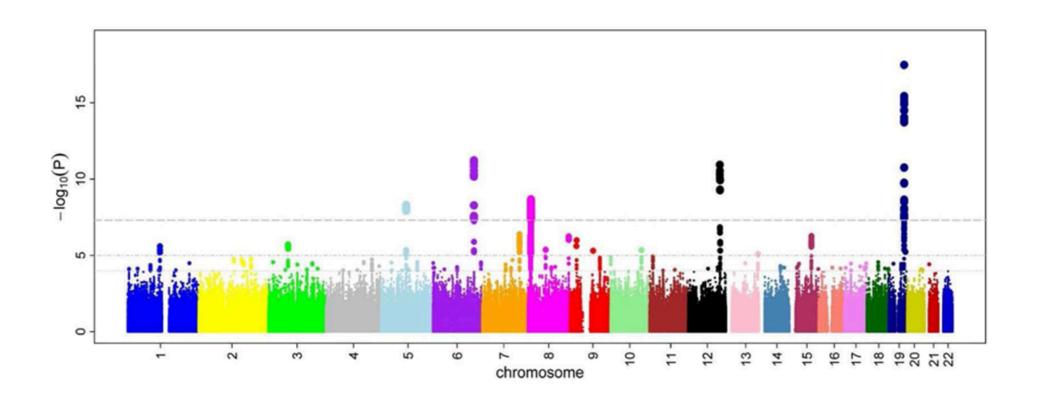
n = total number of outcomes

The probability distribution of this statistic is given by the chi-square distribution.



Using chi-square test, we can test how well observed values fit expected values computed under the independence hypothesis

#### Manhattan Plot



**GWAS** for microcirculation

### Let's do a GWAS

Allele	Short Hair	Long Hair	Totals	Expected
Т	3	12	15	7.5
G	9	0	9	4.5
Totals	12	12	24	

Allele	Short Hair	Long Hair	Expected
Т	3	12	7.5
G	9	0	4.5

$$\chi^{2} = \sum_{i=0}^{\infty} \frac{(O_{i} - E_{i})^{2}}{E_{i}}$$

$$\chi^{2} = \left(\frac{(3 - 7.5)^{2}}{7.5}\right) + \left(\frac{(9 - 4.5)^{2}}{4.5}\right) + \left(\frac{(12 - 7.5)^{2}}{7.5}\right) + \left(\frac{(0 - 4.5)^{2}}{4.5}\right)$$

$$\chi^{2} = 14.4$$

# Significance

	P										
DF	0.995	0.975	0.20	0.10	0.05	0.025	0.02	0.01	0.005	0.002	0.001
1	0.0000393	0.000982	1.642	2.706	3.841	5.024	5.412	6.635	7.879	9.550	10.828
2	0.0100	0.0506	3.219	4.605	5.991	7.378	7.824	9.210	10.597	12.429	13.816
3	0.0717	0.216	4.642	6.251	7.815	9.348	9.837	11.345	12.838	14.796	16.266
4	0.207	0.484	5.989	7.779	9.488	11.143	11.668	13.277	14.860	16.924	18.467
5	0.412	0.831	7.289	9.236	11.070	12.833	13.388	15.086	16.750	18.907	20.515

By alleles – 1 df

 $X^2 = 14.4$ 

Therefore p<0.001

# What we actually care about

- How much does this allele/genotype increase your risk of a disease/phenotype
- Odds ratio (OR) or Risk ratio (RR)
  - Risk ratio depends on penetrance

#### chr32 7420804 ohr32 7472206 ohr32 7473337 ohr32 7479580 chr32 7462967 shr32 7896570 chr32 7492364 Dog Coat Length TC AA GT TT AG TT CG Short Cost AA TT CC GG TC GG GG **Short Cost** CC GA GT CT AG CT Short Coat TC GA GG CT AA CG **Short Coat** CC GA GT CT AG CT GG Short Coat TT TC AA GG GG CC CG **Short Coat** CC AA TT TT GG GG Long Cost TC AA TT TT GG GG TT Long Cost TC AA TT TT GG TT GG Long Cost CC AA TT TT GG CT CG Long Cost TT AA TT TT GG TT GG Long Coat CC AA ш TT GG TT GG Long Coat

#### Chr32 7490570

Allele	Short	Long	Totals	Expected
Τ	6	11	17	8.5
С	6	1	7	3.5
Totals	12	12	24	

$$\chi^2 = \sum_{i=0}^{\infty} \frac{(O_i - E_i)^2}{E_i}$$

$$\chi^{2} = \left(\frac{(6-8.5)^{2}}{8.5}\right) + \left(\frac{(11-8.5)^{2}}{8.5}\right) + \left(\frac{(6-3.5)^{2}}{3.5}\right) + \left(\frac{(1-3.5)^{2}}{3.5}\right)$$

$$\chi^2 = 5.0$$

Still significant 0.05>p>0.025

Allele	Short	Long	Totals
Т	6	11	17
С	6	1	7
Totals	12	12	24

$$P(Short \mid T) = \frac{(Short \& T)}{(T)} = \frac{6}{17} = 0.35$$

$$P(Short \mid C) = \frac{(Short \& C)}{(C)} = \frac{6}{7} = 0.86$$

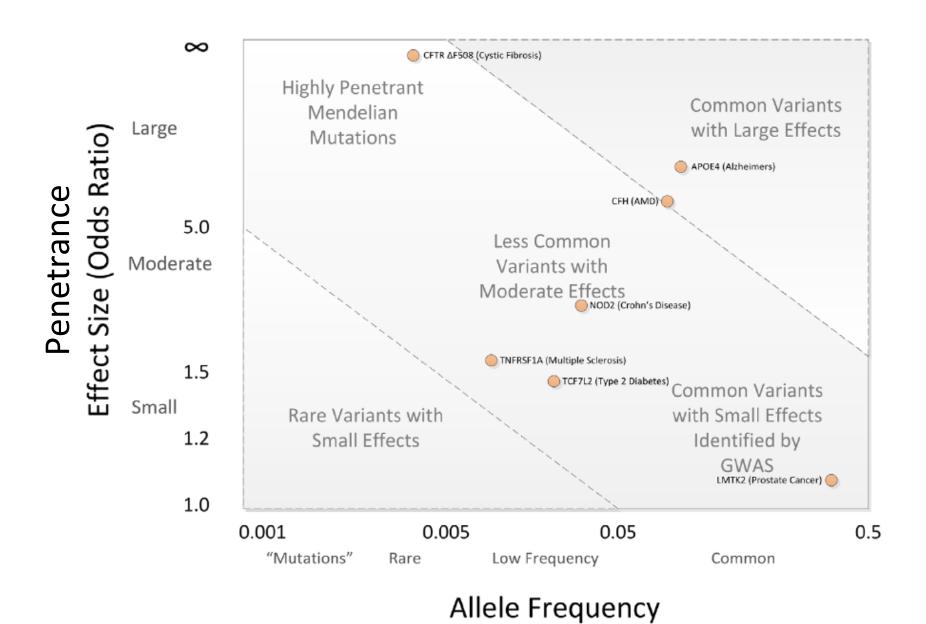
$$Odds(T) = \frac{P(Short \mid T)}{1 - P(Short \mid T)} = 0.55$$

$$Odds(C) = \frac{P(Short \mid C)}{1 - P(Short \mid C)} = 6.0$$

$$OR = \frac{Odds(C)}{Odds(T)} = 11$$

# Quantitative Phenotypes

- Can't use χ<sup>2</sup>
- Need to use a regression on the data
  - Linear is a good first step
- Lots of ongoing research on how to improve this



# Why do GWAS usually find common variants with moderate to large effects?

- Need large numbers of individuals to find rare variants (statistical power)
- Need moderate effect to reach statistical significance
- If very high effect, can still find, like Chr32 7473337

# Some problems with GWAS

- Clinical phenotypes aren't always good enough
  - May be multiple pathophysiologies that have different causes
  - Clinical categories may be too broad to distinguish
- SNPs sometimes not clear how they relate
- Often find nothing
  - Moving toward pathway analysis instead of gene by gene

# Questions?