

# Lecture 1: Case-Control Association Testing

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## Introduction

- ▶ Association mapping is now routinely being used to identify loci that are involved with complex traits.
- ▶ Technological advances have made it feasible to perform case-control association studies on a genome-wide basis with hundreds of thousands of markers in a single study.
- ▶ We consider testing a genetic marker for association with a disease in a sample of unrelated subjects.
- ▶ Case-control association methods essentially test for independence between trait and allele/genotype.

# Case-Control Association Testing

- ▶ Allelic Association Tests
  - ▶ Allele is treated as the sampling unit
  - ▶ Typically make an assumption of Hardy-Weinberg equilibrium (HWE). Alleles within an individual are conditionally independent, given the trait value.
- ▶ Genotypic Association Tests
  - ▶ Individual is the sampling unit
  - ▶ Does not assume HWE

## Pearson's $\chi^2$ Test for Allelic Association

- ▶ The classical Pearson's  $\chi^2$  test is often used for allelic association testing.
- ▶ This test looks for deviations from independence between the trait and allele.
- ▶ Consider a single marker with 2 allelic types (e.g., a SNP) labeled “1” and “2”
- ▶ Let  $N_{ca}$  be the number of cases and  $N_{co}$  be the number of controls with genotype data at the marker.

## Pearson's $\chi^2$ Test for Allelic Association

- Below is a  $2 \times 2$  contingency table for trait and allelic type

	Cases	Controls	Total
Allele 1	$n_1^{ca}$	$n_1^{co}$	$n_1$
Allele 2	$n_2^{ca}$	$n_2^{co}$	$n_2$
Total	$2N_{ca}$	$2N_{co}$	$T$

- $n_1^{ca}$  is the number of type 1 alleles in the cases and  $n_1^{ca} = 2 \times$  the number of homozygous (1,1) cases + the number of heterozygous (1,2) cases
- $n_2^{co}$  is the number of type 2 alleles in the controls and  $n_2^{co} = 2 \times$  the number of homozygous (2,2) controls + the number of heterozygous (1,2) controls
- Hypotheses
  - $H_0$ : there is *no association* between the row variable and column variable
  - $H_a$ : there *is* an association between the two variables

## Pearson's $\chi^2$ Test for Allelic Association

- ▶ Can use Pearson's  $\chi^2$  test for independence. The statistic is:

$$\chi^2 = \sum_{\text{all cells}} \frac{(\text{Observed cell} - \text{Expected cell})^2}{\text{Expected cell}}$$

- ▶ What is the the expected cell number under  $H_0$ ? For each cell, we have

$$\text{Expected Cell Count} = \frac{\text{row total} \times \text{col total}}{\text{total count}}$$

- ▶ Under  $H_0$ , the  $\chi^2$  test statistic has an approximate  $\chi^2$  distribution with  $(r - 1)(c - 1) = (2 - 1)(2 - 1) = 1$  degree of freedom

## LHON Example: Pearson's $\chi^2$ Test

- ▶ From Phasukijwattana et al. (2010), Leber Hereditary Optic Neuropathy (LHON) disease and genotypes for marker rs6767450:

	CC	CT	TT
Cases	6	8	75
Controls	10	66	163

- ▶ Corresponding  $2 \times 2$  contingency table for allelic type and case-control status

	Cases	Controls	Total
Allele T	158	392	550
Allele C	20	86	106
Total	178	478	656

- ▶ Intuition for the test: Suppose  $H_0$  is true, allelic type and case-control status are independent, then what counts would we expect to observe?

- ▶ Recall that under the independence assumption

## LHON Example: Pearson's $\chi^2$ Test

	Cases	Controls	Total
Allele T	158	392	550
Allele C	20	86	106
Total	178	478	656

- ▶ Let  $n$  be the total number of alleles in the study. Assuming independence, the expected number of case alleles that are of type T is:

$$\begin{aligned}
 & n \times P(\text{Allele is from a Case and Allelic type is T}) \\
 &= nP(\text{Allele is from a Case})P(\text{Allelic type is T}) \\
 &= 656 \left( \frac{178}{656} \right) \left( \frac{550}{656} \right) = \frac{(178)(550)}{656} = 149.2378
 \end{aligned}$$



## LHON Example: Pearson's $\chi^2$ Test

- ▶ Fill in the remaining cells for the expected counts

	Cases	Controls	Total
Allele T	149.2378		
Allele C			
Total			

- ▶ Calculate the  $X^2$  statistic

$$\chi^2 = \frac{(158 - 149.2378)^2}{149.2378} + \dots + \frac{(86 - 77.2378)^2}{77.2378} = 4.369$$

- ▶ What is the  $p$ -value?

$$P(\chi_1^2 \geq 4.369) = .037$$

## LHON Example: Pearson's $\chi^2$ Test

- ▶ Fill in the remaining cells for the expected counts

	Cases	Controls	Total
Allele T	149.2378	400.7622	550
Allele C	28.7622	77.2378	106
Total	178	478	656

- ▶ Calculate the  $\chi^2$  statistic

$$\chi^2 = \frac{(158 - 149.2378)^2}{149.2378} + \dots + \frac{(86 - 77.2378)^2}{77.2378} = 4.369$$

- ▶ What is the  $p$ -value?

$$P(\chi_1^2 \geq 4.369) = .037$$

## Fisher's Exact Test for Allelic Association

- ▶ For contingency tables that have cells with less than 5 observations
- ▶ Consider the table below

	Cases	Controls	Total
Allele T	21	14	35
Allele C	3	10	13
Total	24	24	48

- ▶ The marginal counts of the table are fixed: There are 24 case alleles, 24 control alleles, 35 T alleles, and 13 C alleles
- ▶ Let  $X$  be the number of case alleles that are of type T. A test based on  $X$  can be constructed.
- ▶ Under the null hypothesis,  $X$  will have a hypergeometric distribution where the probability that  $X = x$  is

$$\binom{35}{x} \binom{13}{24-x} / \binom{48}{24}$$

## Fisher's Exact Test for Allelic Association

- ▶ Obtain the probability distribution for  $X$

$x$	11	12	13	14	15	16	17	18	19	20	21	22	23	24
$P(X=x)$														

- ▶  $P(X = 11)$  is

$$\binom{35}{11} \binom{13}{13} / \binom{48}{24} = .00001$$

- ▶  $P(X = 12)$  is

$$\binom{35}{12} \binom{13}{12} / \binom{48}{24} = .00003$$

## Fisher's Exact Test for Allelic Association

- ▶ Obtain the probability distribution for  $X$

$x$	$P(X=x)$
11	.00001
12	.0003
13	.004
14	.021
15	.072
16	.162
17	.241
18	.241
19	.162
20	.072
21	.021
22	.004
23	.0003
24	.00001

- ▶ Construct a rejection region for a two-sided test with  $\alpha = .05$ .
- ▶ Can the null hypothesis be rejected at the .05 level for the observed value  $X = 21$ ?

## Fisher's Exact Test for Allelic Association

- ▶ Obtain the probability distribution for  $X$

$x$	$P(X=x)$
11	.00001
12	.0003
13	.004
14	.021
15	.072
16	.162
17	.241
18	.241
19	.162
20	.072
21	.021
22	.004
23	.0003
24	.00001

- ▶ So, a rejection region for a two-sided test with  $\alpha = .05$  would consist of the following values for  $X$ : 11, 12, 13, 14, 21, 22, 23, and 24.
- ▶ The observed  $X$  value of 21 for the data falls in this region, so the test would reject at the level .05.

## The Armitage Trend Test for Genotypic Association

- ▶ The most common genotypic test for unrelated individuals is the Armitage trend test (Sasieni 1997)
- ▶ Consider a single marker with 2 allelic types (e.g., a SNP) labeled “1” and “2”
- ▶ Let  $Y_i = 2$  if individual  $i$  is homozygous (1,1), 1 if the  $i$  is heterozygous, and 0 if  $i$  is homozygous (2,2)
- ▶ Let  $X_i = 1$  if  $i$  is a case and 0 if  $i$  is a control.
- ▶ A simple linear regression model of

$$Y = \beta_0 + \beta_1 X + \epsilon$$

- ▶  $H_0 : \beta_1 = 0$  vs.  $H_a : \beta_1 \neq 0$

## The Armitage Trend for Genotypic Association

- ▶ To test this hypothesis, the Armitage trend test statistic is

$$A_r = \frac{\hat{\beta}_1^2}{\text{VAR}(\hat{\beta}_1)} = Nr_{xy}^2$$

where  $r_{xy}^2$  is the squared correlation between genotype variable  $Y$  and phenotype variable  $X$ .

- ▶ Note that the variance estimate for  $Y$  that is used in the calculation of the Armitage trend test is the sum of the squared deviations of  $Y$  from the fitted values of  $Y$  for regression with only an intercept term.
- ▶ Under the null hypothesis,  $A_r$  will follow an approximate  $\chi^2$  distribution with 1 degree of freedom.
- ▶ The Armitage trend test can be shown to be valid when HWE does not hold.



## LHON Example: Armitage Trend Test

- ▶ Leber Hereditary Optic Neuropathy (LHON) disease and genotypes for marker rs6767450:

	CC	CT	TT
Cases	6	8	75
Controls	10	66	163

- ▶ The Armitage test statistic for this data is

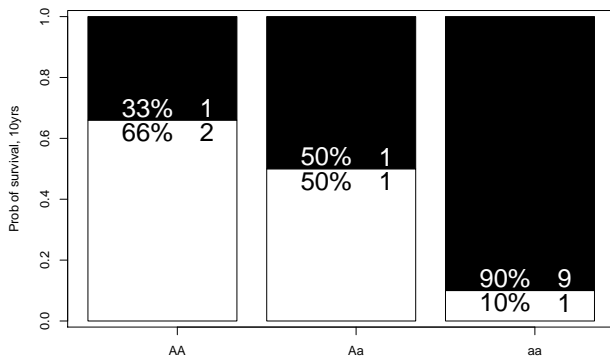
$$A_r = Nr_{xy}^2 = 328(.0114) = 3.74$$

- ▶ The  $p$ -value is

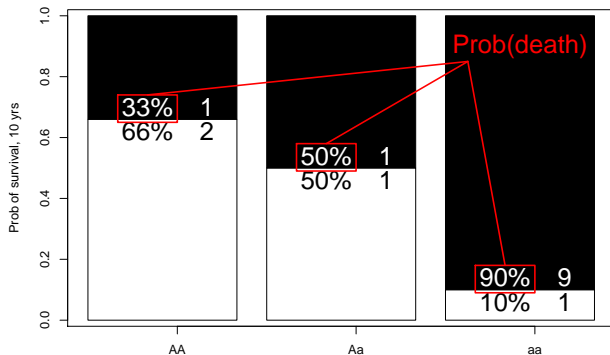
$$P(\chi_1^2 \geq 3.743) = .053$$

## Odds Ratios (ORs) for Genotypes

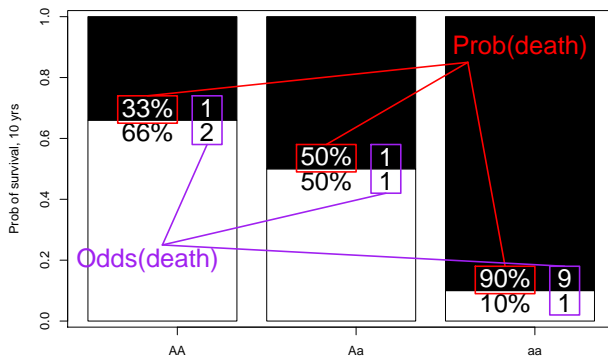
- ▶ What are **odds**? Really just **probability**...
- ▶ Odds are a [gambling-friendly] measure of chance;



# Odds Ratios (ORs) for Genotypes



## Odds Ratios (ORs) for Genotypes



– so what are **odds ratios**?

## Odds Ratios (ORs) for Genotypes

	Cases	Controls
TT	$A$	$B$
CT	$A'$	$B'$
CC	$C$	$D$

- Typically choose a reference genotype.

$$OR_{TT} = \frac{\text{odds of disease in an individual with the TT genotype}}{\text{odds of disease in an individual with the CC genotype}}$$

$$OR_{CT} = \frac{\text{odds of disease in an individual with the CT genotype}}{\text{odds of disease in an individual with the CC genotype}}$$

- $OR_{TT} = 1$  implies no association with and disease. Similarly for  $OR_{CT}$ .
- $OR_{TT} > 1$  or  $OR_{TT} < 1$  implies association with the disease.

## Odds Ratios (ORs) for Genotypes

- ▶ Logistic regression is generally used to get odds ratios and confidence intervals for genotypes.
- ▶ Let  $\pi_i$  be the probability that individual  $i$  is affected with the disease and let  $G_i$  be the genotype for individual  $i$  at the SNP:

$$\log(\text{odds of disease for individual } i | G_i)$$

$$= \log \left( \frac{\pi_i}{1 - \pi_i} \middle| G_i \right)$$

$$= \beta_0 + \beta_{CT} I\{G_i = CT\} + \beta_{TT} I\{G_i = TT\}$$

where  $I\{G_i = CT\}$  is 1 if  $G_i = CT$  and 0 otherwise, and similarly for  $I\{G_i = TT\}$ .

## Odds Ratios (ORs) for Genotypes

- ▶ The coefficient estimates for  $\hat{\beta}_{CT}$  and  $\hat{\beta}_{TT}$  can be used to calculate odds ratios:

$$OR_{CT} = \exp(\hat{\beta}_{CT})$$

$$OR_{TT} = \exp(\hat{\beta}_{TT})$$

- ▶ 95% CI for  $OR_{CT}$  is

$$\exp(\hat{\beta}_{CT} \pm 1.96 \times \text{s.e.}(\hat{\beta}_{CT}))$$

## Odds Ratios for LHON Example

- ▶ Leber Hereditary Optic Neuropathy (LHON) disease and genotypes for marker rs6767450:

	CC	CT	TT
Cases	6	8	75
Controls	10	66	163

- ▶ We will use the R software package to obtain odds ratios and confidence intervals for this data set (as well as Pearson's  $\chi^2$  test and Armitage Trend tests).
- ▶ Exercises and some commands for analyzing the LHON data with R can be found on the following webpage:

[http://faculty.washington.edu/tathornt/SISG\\_MODULE8.html](http://faculty.washington.edu/tathornt/SISG_MODULE8.html)



## Odds Ratios (ORs) based on Allele Counting

- ▶ We can also obtain allelic odds ratios
- ▶ Odds ratios based on an allele counting model essentially assumes an additive model
- ▶ Genotype  $TT$  has twice the risk (or protection) of heterozygous genotype  $CT$ .
- ▶ Same risk (or protection) for the comparison of heterozygous  $CT$  genotype and homozygous  $CC$  genotype.

	Cases	Controls
T	$n_A$	$n_B$
C	$n_C$	$n_D$

$$\begin{aligned}
 OR_T &= \frac{\text{odds of disease with T allele}}{\text{odds of disease with C allele}} \\
 &= \frac{(n_A/n_B)}{(n_C/n_D)} = \frac{n_A \times n_D}{n_B \times n_C}
 \end{aligned}$$

## Odds Ratios (ORs) Allele Counting

	Cases	Controls
T	$n_A$	$n_B$
C	$n_C$	$n_D$

- ▶  $OR_T = 1$  implies no association with and disease
- ▶  $OR_T > 1$  or  $OR_T < 1$  implies association with the disease

## Confidence Intervals for Odds Ratios (ORs)

	Cases	Controls
T	$n_A$	$n_B$
C	$n_C$	$n_D$

$$OR = \frac{n_A \times n_D}{n_B \times n_C}$$

$$s.e.(\log(OR)) = \sqrt{\frac{1}{n_A} + \frac{1}{n_B} + \frac{1}{n_C} + \frac{1}{n_D}}$$

- ▶ Lower limit of 95% CI

$$= \exp(\log(OR) - 1.96 \times s.e.(\log(OR)))$$

- ▶ Upper limit of 95% CI

$$= \exp(\log(OR) + 1.96 \times s.e.(\log(OR)))$$

## Confidence Intervals for Odds Ratios (ORs)

rs6767450	Cases	Controls
T	158	392
C	20	86

$$OR = \frac{n_A \times n_D}{n_B \times n_C}$$

$$s.e.(\log(OR)) = \sqrt{\frac{1}{n_A} + \frac{1}{n_B} + \frac{1}{n_C} + \frac{1}{n_D}}$$

- ▶ Lower limit of 95% CI

$$= \exp(\log(OR) - 1.96 \times s.e.(\log(OR)))$$

- ▶ Upper limit of 95% CI

$$= \exp(\log(OR) + 1.96 \times s.e.(\log(OR)))$$

## LHON Example: Confidence Intervals for Odds Ratios (ORs)

rs6767450	Cases	Controls
T	158	392
C	20	86

$$OR = \frac{158 \times 86}{392 \times 20} = 1.7332$$

$$s.e.(log(OR)) = \sqrt{\frac{1}{158} + \frac{1}{392} + \frac{1}{20} + \frac{1}{86}}$$

- ▶ Lower limit of 95% CI

$$= \exp(\log(OR) - 1.96 \times s.e.(log(OR)))$$

$$= \exp(\log(1.7332) - 1.96 \times 0.2665) = 1.03$$

- ▶ Upper limit of 95% CI = 2.92

## References

- ▶ Phasukijwattana N, Kunhapan B, Stankovich J, Chuenkongkaew WL, Thomson R, Thornton T, Bahlo M, Mushiroda T, Nakamura Y, Mahasirimongkol S, et al. (2010). Genome-wide linkage scan and association study of PARL to the expression of LHON families in Thailand. *Hum. Genet.* **128**, 39-49.
- ▶ Sasieni P (1997). From genotypes to genes: doubling the sample size. *Biometrics* **5**, 1254-1261.