

# Contingency Tables

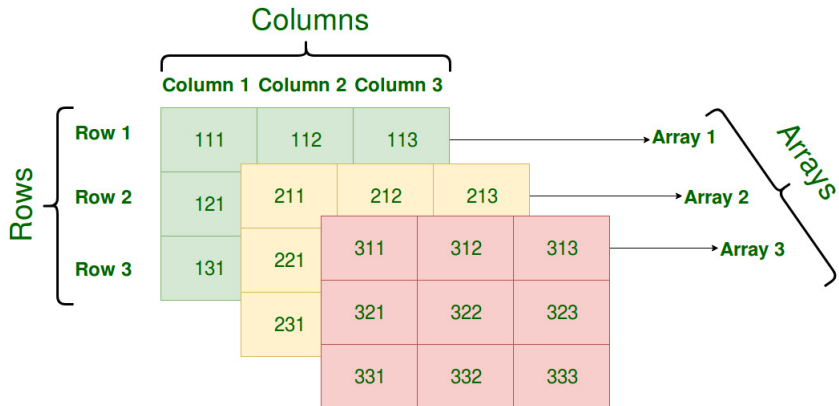
## Three-way

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# Three-way contingency tables

- Three random variables  $X$ ,  $Y$ , and  $Z$
- There are new quantities of interest in this setting
  - Conditional associations
  - Odds ratio between  $X$  and  $Y$  at a particular value of  $Z$
- Conditional associations can be different from marginal associations between  $X$  and  $Y$
- Commonly interest is in the association between  $X$  and  $Y$  after controlling for (conditioning on)  $Z$



## Example

A  $2 \times 2 \times 2$  table from data from Florida 1976-1987

Victim's Race	Defendant's Race	Death Penalty		Percentage Yes
		Yes	No	
White	White	53	414	11.3
	Black	11	37	22.9
Black	White	0	16	0.0
	Black	4	139	2.8
Total	White	53	430	11.0
	Black	15	176	7.9

- $Y$  be the response whether they receive death penalty
- $X$  be the defendant's race
- $Z$  be the victim's race

## Example (continued)

- If we ignore  $Z$  and look at the marginal table we get

$$\hat{\theta}_{XY} = \frac{53 \times 176}{15 \times 430} = 1.45$$

which indicates that whites are more likely to get the death penalty

- Of interest is  $\theta_{XY(1)}$  and  $\theta_{XY(2)}$  the odds ratios between  $X$  and  $Y$  within the two levels of  $Z$ 
  - $Z = \text{white}$ ,  $\hat{\theta}_{XY(1)} = \frac{53 \times 37}{414 \times 11} = 0.43$  (0.42 after adding 0.5 to each cell)
  - $Z = \text{black}$ ,  $\hat{\theta}_{XY(2)} = \frac{0 \times 139}{16 \times 4} = 0$  (0.94 after adding 0.5 to each cell)
- Conditional odds ratios indicate that whites are less likely to get the death penalty

# Three-way contingency tables

## Definition (Simpson's paradox)

When a marginal association can have different direction from the conditional associations this is called *Simpson's paradox*.

## Definition (Conditional Independence)

Variables  $X$  and  $Y$  are conditionally independent given  $Z$  if they are independent in each conditional table.

In a  $2 \times 2 \times K$  table this implies that

$$\theta_{XY(1)} = \cdots = \theta_{XY(k)} = 1$$

## Example

Clinic	Treatment	Response		$\hat{\theta}$
		Success	Failure	
1	A	18	12	1.0
	B	12	8	
2	A	2	8	1.0
	B	8	32	
Total	A	20	20	2.0
	B	20	40	

$X$  and  $Y$  are conditionally independent, but marginally, are dependent

## Definition (Homogeneous Association)

A *homogeneous association* exists if the conditional odds ratios between  $X$  and  $Y$  are identical at all levels of  $Z$ .

# Cochran-Mantel-Haenszel Test

In a  $2 \times 2 \times K$  table we wish to test

$$H_0: \theta_{XY(1)} = \cdots = \theta_{XY(K)} = 1$$

- The null is that  $X$  and  $Y$  are conditionally independent given  $Z$
- The alternative is that at least one of the conditional odds ratios is different from 1



# Cochran-Mantel-Haenszel Test

The test statistic is

$$CMH = \frac{\left[ \sum_{k=1}^K (n_{11k} - E(n_{11k})) \right]^2}{\sum_{k=1}^K V(n_{11k})} \underset{\text{approx.}}{\overset{H_0}{\sim}} \chi_1^2$$

where under independence,

$$E(n_{11k}) = \frac{n_{1+k} n_{+1k}}{n_{++k}}$$
$$V(n_{11k}) = \frac{n_{1+k} n_{2+k} n_{+1k} n_{+2k}}{n_{++k}^2 (n_{++k} - 1)}$$

# CMH test statistic

- Procedure also provides an estimate of the common odds ratio among the  $K$  tables

$$\hat{\theta}_{MH} = \frac{\sum_{k=1}^K (n_{11k} n_{22k} / n_{++k})}{\sum_{k=1}^K (n_{12k} n_{21k} / n_{++k})}$$

- Though this is only useful when all of the individual tables have similar odds ratios

## Remark

The Breslow-Day Test also exists for testing homogeneity of odds ratios, not just for conditional independence.

## Example

Consider a  $2 \times 2 \times 5$  table

```
> apply(MIOC,3,OR)
```

	1	2	3	4	5
	6.465600	8.859104	1.675303	3.786661	3.810890

Five sample odds ratios do not vary “drastically”, proceed with CMH

```
> mantelhaen.test(MIOC)
```

X-squared = 32.793, df = 1, p-value = 1.025e-08

95 percent confidence interval:

2.426983 6.493688

sample estimates:

common odds ratio

3.969895

# We learned

- Conditional odds-ratio
- CHM test for homogeneous association/odds ratio