STAT115: Introduction to Biostatistics

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Lecture 29: Tests of Association for Contingency Tables

Outline

- Contingency table
- Looking at the relationship between two categorical variables
- Investigate approaches to test independence of two categorical variables
- Compare observed and expected counts
- Introduce χ^2 distribution

Data: Passengers on the Titanic

• Data from the adult passengers on the titanic. Two variables:

► Class: 1st, 2nd, 3rd or crew

► Survived: yes or no

| | | survived | | |
|-------|-------|----------|-----|-------|
| | | no | yes | Total |
| Class | 1st | 122 | 197 | 319 |
| | 2nd | 167 | 94 | 261 |
| | 3rd | 476 | 151 | 627 |
| | Crew | 673 | 212 | 885 |
| | Total | 1438 | 654 | 2092 |

• Do survival probabilities depend on the class?

Big picture

- We have investigated when both variables have two levels (groups)
- Here one of the variables has four levels
 - ▶ 1st 3rd class, crew
- If the survival probabilities vary by class
 - ► The two variables (class and survival) are related
- If the survival probabilities do not vary by class
 - ► The two variables (class and survival) are independent
 - ▶ Knowing the class of a passenger tells us nothing about their survival probability
 - ▶ Recall: Definition of independence when we looked at probability
- Idea: Compare the observed data to what we would expect if two variables were independent

Expected counts

We can use the margin totals to find the expected counts under independence

$$\mathsf{expected}\;\mathsf{count} = \frac{\mathsf{row}\;\mathsf{total} \times \mathsf{column}\;\mathsf{total}}{\mathsf{table}\;\mathsf{total}}$$

• Work through the Titanic table to understand this

Expected counts: Titanic

expected count =
$$\frac{\text{row total} \times \text{column total}}{\text{table total}} = \frac{319 \times 654}{2092} = 99.73$$

| | | survived | | |
|-------|-------|----------|-------|-------|
| | | no | yes | Total |
| Class | 1st | | 99.73 | 319 |
| | 2nd | | | 261 |
| | 3rd | | | 627 |
| | Crew | | | 885 |
| | Total | 1438 | 654 | 2092 |

Proportion of passengers who are 1st class

- ▶ 15.25% of passengers are 1st class
- · If survival and class are independent
 - Expected number is the total number of passengers who survive × the proportion of passengers who are 1st class

Expected counts: Titanic

expected count =
$$\frac{\text{row total} \times \text{column total}}{\text{table total}} = \frac{627 \times 1438}{2092} = 430.99$$

| | | survived | | |
|-------|-------|----------|-------|-------|
| | | no | yes | Total |
| Class | 1st | | 99.73 | 319 |
| | 2nd | | | 261 |
| | 3rd | 430.99 | | 627 |
| | Crew | | | 885 |
| | Total | 1438 | 654 | 2092 |

• Proportion of passengers who are 3rd class

- ▶ 29.97% of passengers are 3rd class
- · If survival and class are independent
 - Expected number is the total number of passengers who died × the proportion of passengers who are 3rd class

 $\qquad \qquad \mathsf{Or} \ \mathsf{column} \ \mathsf{total} \times \frac{\mathsf{row} \ \mathsf{total}}{\mathsf{table} \ \mathsf{total}}$

Expected counts: Titanic

• Put it all together to give observed (black) and expected (blue)

| | | survived | | |
|-------|-------|--------------|--------------|-------|
| | | no | yes | Total |
| Class | 1st | 122 (219.27) | 197 (99.73) | 319 |
| | 2nd | 167 (179.41) | 94 (81.59) | 261 |
| | 3rd | 476 (430.99) | 151 (196.01) | 627 |
| | Crew | 673 (608.33) | 212 (276.67) | 885 |
| | Total | 1438 | 654 | 2092 |

- The observed and expected counts will vary: there is natural variation in the data
 - ▶ Do they vary more than we would expect if variables are truly independent?

Test for independence/association

- We can look at this with a hypothesis test
 - ▶ H₀ : the two variables are independent
 - ▶ H_A : the two variables are related (associated)
- The test statistic we will use is

$$X^2 = \sum \frac{(\mathsf{observed} - \mathsf{expected})^2}{\mathsf{expected}}$$

 \blacktriangleright For each cell we calculate $\frac{(\text{observed-expected})^2}{\text{expected}}$ and add them up

Test statistic

| | | survived | | |
|-------|-------|--------------|--------------|-------|
| | | no | yes | Total |
| Class | 1st | 122 (219.27) | 197 (99.73) | 319 |
| | 2nd | 167 (179.41) | 94 (81.59) | 261 |
| | 3rd | 476 (430.99) | 151 (196.01) | 627 |
| | Crew | 673 (608.33) | 212 (276.67) | 885 |
| | Total | 1438 | 654 | 2092 |

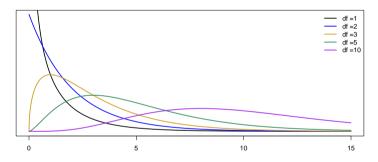
$$X^{2} = \frac{(122 - 219.27)^{2}}{219.27} + \frac{(197 - 99.73)^{2}}{99.73} + \dots + \frac{(212 - 276.67)^{2}}{276.67}$$
$$= 177.8$$

Test statistic

- If the null hypothesis is true
 - ▶ The test statistic, X^2 , will be a realisation from a χ^2 -distribution with $(R-1)\times (C-1)$ degrees of freedom
 - $-\ R$ is the number of rows; C is the number of columns
- Titanic data: R=4, C=2
 - $df = (4-1) \times (2-1) = 3$

Detour: χ^2 -distribution

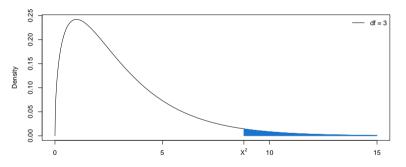
• The χ^2 -distribution is a distribution for positive random variables



- ► It is asymmetric (positively skewed)
- ▶ It has one parameters: degrees of freedom

Finding a p-value

- ullet An extreme X^2 -value is one that is as large, or larger, than that observed
 - ▶ Indicative of increased divergence between observed and expected counts



- The p-value (blue area) is given by 1-pchisq(X2, df)
 - ightharpoonup pchisq(X2, df) gives probability of a value less than X^2

- Data: each row is an observation
 - ► Titanic data: each row is a passenger
- Import into R

```
titanic = read.csv('titanic.csv')
head(titanic)
```

```
##
     Class Survived
      Crew
                Yes
      Crew
                Yes
## 3
       2nd
                 No
      1st
                Yes
## 5
      Crew
                Yes
## 6
       3rd
                  No
```

We use the table function to obtain contingency table

```
titan = table(titanic$Class, titanic$Survived)
```

- ► First argument: variable 1 (class of passenger)
- Second argument: variable 2 (survived: yes / no)

```
titan
##
            No Yes
##
           122 197
##
     1st
##
     2nd
           167
               94
     3rd
          476 151
##
     Crew 673 212
##
```

```
addmargins(titan)
##
##
                Yes
                      Sum
             No
##
            122
                 197
                      319
     1st
##
     2nd
            167
                      261
##
     3rd
            476
                 151
                      627
            673
     Crew
                 212
                      885
           1438
     Sum
                 654 2092
```

• The function addmargins includes the margins on the table

• The R function chisq.test evaluates the test

```
out1 = chisq.test(titan)
out1
##
## Pearson's Chi-squared test
##
## data: titan
## X-squared = 177.8, df = 3, p-value <2e-16</pre>
```

- The p-value $< \alpha = 0.05$. Observing a test statistic as large as we did is unusual if the two variables were independent
 - ▶ Evidence in support of H_A: that the variables are not independent

• The chisq.test function can return the expected counts

- Still important to know:
 - ► How to calculate them
 - ▶ What they represent (expected counts if variables are independent)

 χ^2 -test

- If R=2 and C=2: we have a 2×2 contingency table, e.g. smallpox in Boston
 - ightharpoonup The χ^2 test is identical to test for difference in proportions
 - $ightharpoonup H_0: p_1 p_2 = 0 \text{ and } H_A: p_1 p_2 \neq 0$
 - ► E.g. or smallpox data, following two concepts are the same:
 - Probability of death differs between those innoculated and those note;
 - There is an association between innoculation status (yes/no) and mortality (died/survived)
- The χ^2 test can also be used if both R>2 and C>2
- The χ^2 test is unreliable if any of the expected counts < 5
 - ▶ Options for resolving this problem are beyond the scope of course

Testing for Independence for Smallpox Data (in R)

```
SmallpoxTable = rbind(x,n-x)
SmallpoxTable # rbind combines rows to make a matrix (tabular array)
     [,1] [,2]
          844
        6
      238 5136
out1 = chisq.test(SmallpoxTable)
011t.1
##
    Pearson's Chi-squared test with Yates' continuity correction
##
## data: SmallpoxTable
## X-squared = 26, df = 1, p-value = 3e-07
out1$expected
      [,1] [,2]
     33.3 817
     210.7 5163
```

Testing for Independence for Smallpox Data

Results

- P-value is tiny: 3.37×10^{-7}
- Data highly inconsistent with H_0 (i.e. assumption of independence / non-association between variables)
- Conclude there is (very) strong evidence of an associated between mortality and innoculation status
- Test p-value and conclusion mirrors exactly that from last lecture, where we tested equality of mortality probabilities.

Summary

- χ^2 test for independence of contingency table
- Idea: compare observed counts with those expected under independence
- Assess evidence using χ^2 -distribution
- Analysis 2×2 contingency table equivalent to comparing proportions