## STAT 22000 Lecture Slides Exploring Categorical Data

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## Outline

This set of slides cover Section 1.7 in the text.

- Ways to summarize of a single categorical variable
- Frequency tables
- Barplots, pie charts
- Ways to summarize of relationships between two categorical variables
- two-way contingency tables
- segmented barplots, standardized segmented barplots, mosaic plot


## Bar Graphs and Pie Charts

## Graphs for Categorical Variables

A categorical variable is summarized by a table showing the count or the percentage of cases in each category, and is often displayed by a bar plot or a pie chart.

Ex: Passengers on Titanic

| Class | Freq | Percent |
| :---: | :---: | :---: |
| 1st | 325 | $14.8 \%$ |
| 2nd | 285 | $12.9 \%$ |
| 3rd | 706 | $32.1 \%$ |
| Crew | 885 | $40.2 \%$ |
| Total | 2201 | $100 \%$ |




## Bar plots

A bar plot is a common way to display a single categorical variable. A bar plot where proportions instead of frequencies are shown is called a relative frequency bar plot.



## How are Bar Plots Different From Histograms?

- Bar plots are used for displaying distributions of categorical variables, while histograms are used for numerical variables.
- The horizontal axis in a histogram is a number line, hence the order of the bars cannot be changed, while in a bar plot the categories can be listed in any order (though some orderings make more sense than others, especially for ordinal variables.)


## Why We Recommend Bar Plots Over Pie Charts?

In a pie chart, the areas of slices represents the percentages of categories. However, it is generally more difficult to compare group sizes in a pie chart than in a bar plot, especially when categories have nearly identical counts or proportions


Without looking at the counts, can you tell which class have fewest people from the pie?

## Why We Recommend Bar Plots Over Pie Charts?

It's much easier to make a wrong pie chart than a wrong bar plot. In a pie chart, the categories must make up a whole. There is no such restriction for a bar plot.


## Another Wrong Pie Chart


http://www.youtube.com/watch?v=-rbyhj8uTT8

## Two-Way Contingency Tables

## Two-Way Contingency Tables

A table that summarizes data for two categorical variables is called a contingency table.
E.g., breakdown of people on Titanic by class and survival status

|  |  | Died | Survived | Total |
| :---: | :---: | :---: | :---: | :---: |
| Class | 1st | 122 | 203 | 325 |
| and | 167 | 118 | 285 |  |
|  | Ord | 528 | 178 | 706 |
|  | Crew | 673 | 212 | 885 |
|  | Sum | 1490 | 711 | 2201 |

The marginal totals give the distributions of the two variables, e.g.,

- overall, 1490 died and 711 survived
- there were 325, 285, and 706 passengers in the 1st, 2nd and 3rd classes, and 885 crew members


## Overall Proportions

Dividing the cell counts in a contingency table by the overall total, we get the proportions of observations in the combinations of the two variables.

Survived

| Class |  | No | Yes | Total |
| :---: | :---: | :---: | :---: | :---: |
|  | 1st | $122 / 2201 \approx 0.06$ | $203 / 2201 \approx 0.09$ | $325 / 2201 \approx 0.15$ |
|  | 2nd | $167 / 2201 \approx 0.08$ | $118 / 2201 \approx 0.05$ | $285 / 2201 \approx 0.13$ |
|  | 3rd | $528 / 2201 \approx 0.24$ | $178 / 2201 \approx 0.08$ | $706 / 2201 \approx 0.32$ |
|  | Crew | $673 / 2201 \approx 0.31$ | $212 / 2201 \approx 0.10$ | $885 / 2201 \approx 0.40$ |
|  | Sum | $1490 / 2201 \approx 0.68$ | $711 / 2201 \approx 0.32$ | 1 |

e.g., of people on Titanic

- $122 / 2201 \approx 6 \%$ were in the 1 st class and died in the disaster
- $212 / 2201 \approx 10 \%$ were survived crew members

Note the marginal totals give the distributions of the two variables, e.g.,

- Overall, $711 / 2201 \approx 32 \%$ of the people survived


## Row Proportions

The row proportions (cell counts divided by the corresponding row totals) give the proportion of people survived in the four classes.

|  |  | Survived |  | Total |
| :---: | :---: | :---: | :---: | :---: |
|  |  | No | Yes |  |
|  | 1st | 122/325 $\approx 0.38$ | 203/325 $\approx 0.62$ | 1 |
| Class | 2nd | $167 / 285 \approx 0.59$ | 118/285 $\approx 0.41$ | 1 |
|  | 3rd | $528 / 706 \approx 0.75$ | 178/706 $\approx 0.25$ | 1 |
|  | Crew | $673 / 885 \approx 0.76$ | 212/885 $\approx 0.24$ | 1 |

e.g.,

- $203 / 325 \approx 62 \%$ of people in the 1 st class survived.
- $178 / 706 \approx 25 \%$ of people in the 3rd class survived.


## Column Proportions

The column proportions (dividing cell counts by the corresponding column totals) give the proportion of people survived in each of the four classes.

|  | Survived |  |
| :---: | :---: | :---: |
| Class | No | Yes |
|  | 1st | $122 / 1490 \approx 0.08$ |
| 2nd | $167 / 1490 \approx 0.11$ | $118 / 711 \approx 0.29$ |
|  | 3rd | $528 / 1490 \approx 0.35$ |
|  | $178 / 711 \approx 0.17$ |  |
| Crew | $673 / 1490 \approx 0.45$ | $212 / 711 \approx 0.30$ |
|  | Sum | 1 |

- Among those who survived, $203 / 711 \approx 29 \%$ were in the 1 st class.
- Among those who died, $673 / 1490 \approx 45 \%$ were crew members


## Independence of Two Categorical Variables

If the row proportions do not change from row to row, we say the two categorical variables are independent. Otherwise, we say they are associated.
E.g., if the survival rates do not change from class to class, we say 'survival' is independent of 'class'. In the Titanic data, the survival of passengers is associated with the class they were in because the survival rates differ substantially from class to class.

We can also define two categorical variables to be independent if the column proportions do not vary from column to column since the two conditions are equivalent (why?)

## Exercise

The table below shows the breakdown of cases of injuries in the U.S in a certain year. by circumstance and gender ${ }^{1}$. Counts are in millions.

|  | Circumstance |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
| Gender | Work | Home | Other | Total |
| Male | 8.0 | 9.8 | 17.8 | 35.6 |
| Female | 1.3 | 11.6 | 12.9 | 25.8 |
| Total | 9.3 | 21.4 | 30.7 | 61.4 |

- What proportion of injury cases occurred at work? 9.3/61.4 $\approx 0.15$
- What proportion of injury cases occurred at work and on women? 1.3/61.4 $\approx 0.02$

[^0]
## Practise (Cont'd)

Circumstance

| Gender | Work | Home | Other | Total |
| :---: | :---: | :---: | :---: | :---: |
| Male | 8.0 | 9.8 | 17.8 | 35.6 |
| Female | 1.3 | 11.6 | 12.9 | 25.8 |
| Total | 9.3 | 21.4 | 30.7 | 61.4 |

- Among all injury cases occurred on women, what proportion occurred at work? $1.3 / 25.8 \approx 0.05$
- Among all injury cases occurred at work, what proportion occurred on women? 1.3/9.3 $\approx 0.14$
- Is the circumstance of injury cases independent of the gender of the victims? No, only $5 \%$ of injury cases on women occurred at work, compared with $8.0 / 36.5 \approx 22 \%$ of cases on men occurred at work.


## Segmented Bar and Mosaic Plots

## Segmented Bar Plots

|  | Survived |  |  |
| :--- | :---: | ---: | ---: |
| Class | No | Yes | Total |
| 1st | 122 | 203 | 325 |
| 2nd | 167 | 118 | 285 |
| 3rd | 528 | 178 | 706 |
| Crew | 673 | 212 | 885 |
| Sum | 1490 | 711 | 2201 |



## Standardized Segmented Bar Plots



Standardized segmented bar plots are convenient for comparing row proportions, and determining whether the two variables are independent.

However, the information of row totals is lost after standardization.

## Mosaic Plots

- bar widths = row totals
- segment lengths within a bar = row proportions



## Class

$$
\begin{aligned}
\text { segment area } & =(\text { barwidth }) \times(\text { segment length }) \\
& =\text { row total } \times(\text { row proportion }) \\
& =\text { row total } \times \frac{\text { cell count }}{\text { row total }}=\text { cell count }
\end{aligned}
$$

## Exercise 1.68 Raise Taxes on the Rich or the Poor

The mosaic plot below shows the relationship between political party affiliation and views on whether it's better to raise taxes on the rich or on the poor for a random sample of registered voters taken nationally in 2015.

Raise taxes on the rich

Raise taxes on the poor Not sure


Raise taxes on the rich

Raise taxes on the poor Not sure


Which political party identification is least common in the sample, Democrats, Republicans, or Indep/Other?

Ans: Indep/Other.

Raise taxes on the rich

Raise taxes on the poor Not sure


Based on this sample, which political party identification had the highest percentage supported raising taxes on the rich? Which had the lowest?

Ans: Democrats the highest, Republicans the lowest.


What percentage of Democrats (in this sample) supported raising taxes on the rich?
(a) below $25 \%$
(b) between $25 \%$ and $50 \%$
(c) between $50 \%$ and $75 \%$
(d) over $75 \%$

Raise taxes on the rich

Raise taxes on the poor Not sure


In this sample, which of the following groups contains the greatest number of subjects?
(a) Democrats who supported raising taxes on the rich.
(b) Democrats who supported raising taxes on the poor.
(c) Republicans who supported raising taxes on the rich.
(d) Republicans who supported raising taxes on the poor.


Based on the mosaic plot, do views on raising taxes and political affiliation appear to be independent?

Instead of looking at survival rates in the four classes, we can also look at the breakdown of the four classes among those who survived and among those who died.



Class
1st 2nd 3 3rd Crew



Survived

## Ways to Inspect Relationships Between Variables

- numerical v.s. numerical
- scatterplots
- categorical v.s. categorical
- contingency tables
- segmented barplots, standardized segmented barplots, mosaic plot
- categorical v.s. numerical
- side-by-side boxplots
- histograms by group on the same horizontal axis


## Example (Diamonds)

Mosaic plot: Carat Weight v.s. Quality of Cut


## Example (Diamonds)



## Example (Diamonds)



## Example (Diamonds)



## Example (Diamonds)

From the mosaic plots, we can see the proportion of low-quality cut diamonds increases substantially whenever the carat weight of diamonds reaches those benchmarks (0.5, $0.7,0.9,1,1.2,1.5$, $2, \ldots$ ). Diamonds with carat weights right above those benchmarks generally have better quality of cut then those just at those benchmarks.

Possible reasons:
Diamond cutters would want to get the heaviest diamond out of a rough stone whenever possible. They might increase the depth of diamonds to increase the carat weight, but result in a loss of brilliance due to light leakage.


[^0]:    ${ }^{1}$ Source: Vital and Health Statistics published by the National Center for Health Statistics

