

Contingency Tables

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- ▶ *Univariate* summaries are the first step in a statistical analysis, but most analyses involve establishing relationships between *multiple variables*
 - ▶ These slides focus on methods for expressing relationships between *two categorical variables*

Two variables, X and Y , are **associated** if the distribution of X depends upon the distribution of Y

- ▶ Usually, we designate an **explanatory variable** (suspected cause) and a **response variable** (suspected outcome)
 - ▶ This is done using prior knowledge (ie: Exam #1 score could cause final grade, but not vice versa)

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Note:

1. Association is general term, we'll soon cover specific types of association (ie: linear, non-linear, etc.)

Association

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Note:

1. Association is general term, we'll soon cover specific types of association (ie: linear, non-linear, etc.)
2. Observing an association between X and Y doesn't mean that X causes Y , or that Y causes X , *causation* is a complex topic that we'll discuss soon

Contingency Tables

- ▶ For *two categorical variables*, we can display frequencies for *each combination* of the variables in a **contingency table** (also called a two-way frequency table)
- ▶ Below is a two-way frequency table describing the historic 2015-16 Golden State Warriors season:

	Win	Loss
Home	39	2
Away	34	7

What do you think the raw data that was used to construct this table looks like? Try writing out a few rows.

	Win	Loss
Home	39	2
Away	34	7

Practice (solution)

Recognize you're only able to discern the last two columns from the contingency table on the prior slide

Date	Opp	Location	Win
10/27/2015	NOP	Home	W
10/30/2015	HOU	Away	W
10/31/2015	NOP	Away	W
11/2/2015	MEM	Home	W
11/4/2015	LAC	Home	W
11/6/2015	DEN	Home	W
11/7/2015	SAC	Away	W
11/9/2015	DET	Home	W
11/11/2015	MEM	Away	W
11/12/2015	MIN	Away	W
11/14/2015	BRK	Home	W
11/17/2015	TOR	Home	W
11/19/2015	LAC	Away	W
11/20/2015	CHI	Home	W
11/22/2015	DEN	Away	W
11/24/2015	LAL	Home	W
11/27/2015	PHO	Away	W
11/28/2015	SAC	Home	W
11/30/2015	UTA	Away	W
12/2/2015	CHO	Away	W
12/5/2015	TOR	Away	W
12/6/2015	BRK	Away	W
12/8/2015	IND	Away	W
12/11/2015	BOS	Away	W
12/12/2015	MIL	Away	L
12/16/2015	PHO	Home	W
12/18/2015	MIL	Home	W

Margins

A useful step when working with contingency tables is to add *table margins*:

	Win	Loss	Row Total
Home	39	2	41
Away	34	7	41
Column Total	73	9	82

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- ▶ Row and column totals are sometimes called **marginal distributions**
 - ▶ The marginal distribution of the “win” variable (win/loss) is characterized by the frequencies $\{73, 9\}$ and the proportions $\{0.89, 0.11\}$
 - ▶ The marginal distribution of the “location” variable (home/away) is characterized by the frequencies $\{41, 41\}$ and the proportions $\{0.5, 0.5\}$

Conditional Proportions

- ▶ From a contingency table, **conditional proportions** allow us to determine whether the two variables displayed are *associated*
- ▶ There are two types of conditional proportions: **row proportions** are calculated using each row's total, the bottom table show how to calculate these

	Win	Loss	Row Total
Home	39	2	41
Away	34	7	41
Column Total	73	9	82

	Win	Loss	Row Total
Home	$39/41 = 0.95$	$2/41 = 0.05$	1
Away	$34/41 = 0.83$	$7/41 = 0.17$	1
Column Total	$73/82 = 0.89$	$9/82 = 0.11$	1

Conditional Proportions

Column proportions are calculated in a similar way:

	Win	Loss	Row Total
Home	39	2	41
Away	34	7	41
Column Total	73	9	82

	Win	Loss	Row Total
Home	$39/73 = 0.53$	$2/9 = 0.22$	$41/82 = 0.5$
Away	$34/73 = 0.47$	$7/41 = 0.78$	$41/82 = 0.5$
Column Total	1	1	1

Conditional Distributions and Association

- ▶ Two variables are **associated** if the distribution of one variable depends upon that of the other variable
- ▶ So, we might compare the distribution of win/loss proportions *conditional upon a game being at home* with the distribution of win/loss proportions *conditional upon a game being away*
 - ▶ If these distributions differ, the variables “location” and “win” are associated

1. Using the row proportions given below, do you think there is an association between whether the Warriors were home/away and winning?
2. How would you explain this association?

	Win	Loss	Row Total
Home	0.95	0.05	1
Away	0.83	0.17	1
Column Total	0.89	0.11	1

Practice (solution)

1. Yes, there is an association between “location” and “win”
2. The warriors look to be *more likely* to win when playing at home. In other words, the distribution of wins/losses for home games differs from the distribution of wins/losses for away games.

- ▶ Recognize that row and column proportions tell you fundamentally different things about your data
 - ▶ In our example, *row proportions* can describe the *proportion of wins conditional on the game being at home*
 - ▶ Contrast that with *column proportions*, which can describe the *proportion of home games conditional on that game being a win*

- ▶ Recognize that row and column proportions tell you fundamentally different things about your data
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 - ▶ Contrast that with *column proportions*, which can describe the *proportion of home games conditional on that game being a win*
- ▶ The row proportions suggest how often home games were won, while the column proportions suggest how often wins were home games
 - ▶ This distinction doesn't seem to matter much here, but let's look at another example

- ▶ Were crew members on the Titanic more likely to survive than 1st class passengers?
 - ▶ Use row or column proportions from the contingency table below to support your answer

	Survived	Died
Crew	212	673
1st Class	203	122

Practice (solution)

- No, using *row proportions* we see that $\frac{212}{623+212} = 0.24$, or 24% of the crew survived; while $\frac{203}{122+203} = 0.62$, or 62% of first class passengers survived

	Survived	Died
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Practice (solution)

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	Survived	Died
Crew	212	673
1st Class	203	122

- ▶ Notice that this particular question *cannot be answered* using column proportions
 - ▶ The proportion of survivors who were crew is $\frac{212}{212+203} = 0.51$, while the proportion of survivors who were first class passengers is $\frac{203}{212+203} = 0.49$

Practice (solution)

- ▶ No, using *row proportions* we see that $\frac{212}{623+212} = 0.24$, or 24% of the crew survived; while $\frac{203}{122+203} = 0.62$, or 62% of first class passengers survived

	Survived	Died
Crew	212	673
1st Class	203	122

- ▶ Notice that this particular question *cannot be answered* using column proportions
 - ▶ The proportion of survivors who were crew is $\frac{212}{212+203} = 0.51$, while the proportion of survivors who were first class passengers is $\frac{203}{212+203} = 0.49$
 - ▶ Conditioning on the column variable is problematic here because the *marginal distribution* of 1st class/crew is *skewed towards crew*
 - ▶ In other words, most of the survivors were crew members because there were so many more crew members, not because individual crew members were more likely to survive

- 1) Contingency tables display the possible combinations of two categorical variables
- 2) Row proportions or column proportions within a contingency are used to find and describe associations
- 3) Just because an association exists does not mean that one variable caused changes in the other