Lecture 7a: Bivariate associations for nominal- and ordinal-level variables Ernesto F. L. Amaral

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Introduction to Sociological Data Analysis (SOCI 600)

Source: Healey, Joseph F. 2015. "Statistics: A Tool for Social Research." Stamford: Cengage Learning. 10th edition. Chapter 12 (pp. 308–341).



Outline

- Use measures of association to describe and analyze the importance (magnitude) vs. statistical significance of a bivariate correlation
- Define association in the context of bivariate tables
- List and explain the three characteristics of a bivariate correlation: (a) does it exist? (b) how strong is it? and (c) what is the pattern or direction of the association?
- Assess the association of variables in a bivariate table by: (a) calculating and interpreting column percentages and (b) computing and interpreting an appropriate measure of association
- Compute and interpret Spearman's rho, a measure of association for "continuous" ordinal-level variables



Basic concepts

- Two variables are said to be associated when they vary together, when one changes as the other changes
- Association can be important evidence for causal relationships, particularly if the association is strong
- If variables are associated, the score on one variable can be predicted from the score of the other variable
- The stronger the association, the more accurate the predictions
- Read the table from column to column, noting the differences across the "within-column" frequency distributions



Bivariate association

- Bivariate association can be investigated by finding answers to three questions
 - 1. Does an association exist?
 - 2. How strong is the association?
 - 3. What is the pattern and/or direction of the association?

Productivity by Job Satisfaction (frequencies)

	Job Satisfaction (X)			
Productivity (Y)	Low	Moderate	High	TOTALS
Low	30	21	7	58
Moderate	20	25	18	63
High	10	15	27	52
TOTALS	60	61	52	173



Bivariate tables

- Most general rules
 - Calculate percentages within the categories of the independent variable
 - Compare percentages across the categories of the independent variable
- When independent variable is the column variable (as is generally the case, but not always)
 - Calculate percentages within the columns (vertically)
 - Compare percentages across the columns (horizontally)
- Briefest version
 - Percentage down
 - Compare across



Percentages

- To detect association within bivariate tables (assuming the column variable is the independent variable)
 - Compute percentages within the columns (vertically)
 - Compare percentages across the columns (horizontally)

Productivity by Job Satisfaction (percentages)

		Job Satisfaction (X))	
Productivity (Y)	Low	Moderate	High	TOTALS
Low	50.0%	34.4%	13.5%	33.5% (58)
Moderate	33.3%	41.0%	34.6%	36.4% (63)
High	16.7%	24.6%	51.9%	30.1% (52)
TOTALS	100.0% (60)	100.0% (61)	100.0%	100.0% (173)



1. Is there an association?

- An association exists if the conditional distributions of one variable change across the values of the other variable
- With bivariate tables, column percentages are the conditional distributions of Y for each value of X
- If the column percentages change, the variables are associated



2. How strong is the association?

- The stronger the correlation, the greater the change in column percentages (or conditional distributions)
- In weak correlations, there is little or no change in column percentages
- In strong correlations, there is marked change in column percentages



3. Pattern of the association

- Which scores of the variables go together?
- To detect, find the cell in each column which has the highest column percentage
- If both variables are ordinal, we can discuss the "direction" as well
 - In positive associations, the variables vary in the same direction
 - As one variable increases, the other variable increases
 - In negative associations, the variables vary in opposite directions
 - As one variable increases, the other variable decreases



Maximum difference

- One way to measure strength is to find the "maximum difference"
 - The biggest difference in column percentages for any row of the table
 - This is a "quick and easy" method: easy to apply but of limited usefulness

The Relationship Between the Maximum Difference and the Strength of the Relationship

Maximum Difference	Strength The strength of the relationship is	
If the maximum difference is		
Between 0 and 10 percentage points	Weak	
Between 10 and 30 percentage points	Moderate	
More than 30 percentage points	Strong	





Measures for nominal variables

- It is always useful to compute column percentages for bivariate tables
- It is also useful to have a summary measure (a single number) to indicate the strength of the association
- For nominal level variables, there are two commonly used measures of association
 - Chi Square based measures
 - Phi (ϕ) or Cramer's V
 - Proportional Reduction in Error (PRE) measure
 - Lambda (λ)



Phi (ϕ)

- Phi (φ) is the square root of chi square divided by the sample size (n)
- For 2 x 2 tables
- Ranges from 0.0 to 1.0

$$\phi = \sqrt{\frac{\chi^2}{n}}$$



Cramer's V

- Cramer's V
- For tables larger than 2 x 2
- Ranges from 0.0 to 1.0

$$V = \sqrt{\frac{\chi^2}{n(\min r - 1, c - 1)}}$$

- 1. Find the number of rows (r) and the number of columns (c) in the table. Subtract 1 from the lesser of these two numbers to find (min r 1, c 1)
- 2. Multiply the value you found in step 1 by the sample size (n)
- 3. Divide the value of chi square by the value you found in step 2
- 4. Take the square root of the quantity you found in step 3

Limitations

- Limitations of Chi Square based measures
- Phi and Cramer's V measure only the strength of the association
 - They do not identify the pattern/direction
- To assess pattern/direction, interpret the column percentages in the bivariate table
- Phi and V do not provide a true statistical interpretation
 - All we can say is whether the association is weak, moderate, or strong based on the value



Interpretation of strength

 To interpret the strength of an association using Phi or Cramer's V (Chi Square based measures), follow these guidelines

Guidelines for Interpreting the Strength of the Relationship for Nominal-Level Measures of Association

Measure of Association	Strength The strength of the relationship is	
If the value is		
Between 0.00 and 0.10	Weak	
Between 0.11 and 0.30	Moderate	
Greater than 0.30	Strong	



PRE measures

- The logic of Proportional Reduction in Error (PRE) measures is based on two predictions
 - First prediction, E_1 : How many errors in predicting the value of the dependent variable (Y) do we make if we **ignore** information about the independent variable (X)
 - Second prediction, E_2 : How many errors in predicting the value of the dependent variable (Y) do we make if we take the independent variable (X) into account
- If the variables are associated, we should make fewer errors of the second kind (E₂) than we make of the first kind (E₁)

Lambda (λ)

- Like Phi and Cramer's V
 - Lambda (λ) is used to measure the <u>strength</u> of the association between nominal variables in bivariate tables
- Unlike Phi and Cramer's V
 - Lambda is a PRE measure and its value has a more direct interpretation
 - Phi and Cramer's V are only indexes of strength
 - Lambda tells us the improvement in predicting Y while taking X into account



Calculate Lambda (1)

- To compute Lambda, find E_1 and E_2
- $E_1 = N (largest row total)$
- E_2 = for each column, subtract the largest cell frequency from the column total, then sum

$$\lambda = \frac{(E_1 - E_2)}{E_1}$$



Characteristics of Lambda (1)

- Lambda is asymmetric
 - The value will vary depending on which variable is independent
- When row totals are very unequal, Lambda can be zero even when there is an association between the variables
 - For very unequal row marginals, it's better to use a chi square based measure of association



Limitations of Lambda (λ)

- Lambda gives an indication of the strength of the association only
- It does not give information about pattern
- To analyze the pattern of the association, use column percentages in the bivariate table



Example of Phi, *V*, *λ*

- Various supervisors in the city government of Shinbone, Kansas, have been rated on the extent to which they practice authoritarian styles of leadership and decision making
- Efficiency of each department has also been rated

Efficiency	Authorit	Total		
Efficiency	Low	High	Total	
Low	10	12	22	
High	17	5	22	
Total	27	17	44	



1. Is there an association?

- Calculate the column percentages taking each cell frequency, dividing by the column total, and multiplying by 100
- The column percentages show the efficiency of workers
 (Y) by the authoritarianism of supervisor (X)
- The column percentages change (differ across columns), so these variables are associated

Efficiency	Authorit	Total	
Efficiency	Low	High	Total
Low	10 (37.04%)	12 (70.59%)	22
High	17 (62.96%)	5 (29.41%)	22
Total	27 (100.00%)	17 (100.00%)	44



2. How strong is the association?

- The "maximum difference" is 33.55% (70.59%–37.04%)
- This indicates a "strong" association

Efficiency	Authorit	Total		
Efficiency	Low High		iotai	
Low	10 (37.04%)	12 (70.59%)	22	
High	17 (62.96%)	5 (29.41%)	22	
Total	27 (100.00%)	17 (100.00%)	44	



Phi

$$\phi = \sqrt{\frac{\chi^2}{n}} = \sqrt{\frac{4.70}{44}} = 0.33$$

- Phi = 0.33
- This indicates a "strong" association



Cramer's V

$$V = \sqrt{\frac{\chi^2}{n(\min r - 1, c - 1)}} = \sqrt{\frac{4.70}{44(2 - 1)}} = 0.33$$

- Cramer's V = 0.33
- This indicates a "strong" association



Lambda

- $E_1 = n \text{largest row total} = 44 22 = 22$
- E_2 = for each column, subtract largest cell frequency from the column total = (27 17) + (17 12) = 15

$$\lambda = \frac{(E_1 - E_2)}{E_1} = \frac{22 - 15}{22} = 0.32$$

- Lambda = 0.32
- We reduce our error in predicting the dependent variable by 32% when we take the independent variable into account

3. Pattern of the association

- Low on authoritarianism goes with high on efficiency
- High on authoritarianism goes with low on efficiency
- Therefore, the association is negative: as authoritarianism increases, efficiency decreases

Efficiency	Authorit	Total	
Efficiency	Low	_ow High	
Low	10 (37.04%)	12 (70.59%)	22
High	17 (62.96%)	5 (29.41%)	22
Total	27 (100.00%)	17 (100.00%)	44





Measures for ordinal variables

- Collapsed ordinal variables
 - Have just a few values or scores
 - Use Gamma (G)
 - e.g., social class measured as lower, middle, upper
- Continuous ordinal variables
 - Have many possible scores
 - Resemble interval-ratio level variables
 - Use Spearman's Rho (r_s)
 - e.g., scale measuring attitudes toward handgun control with scores ranging from 0 to 20



Gamma

- Gamma is used to measure the <u>strength</u> and <u>direction</u> of the association
 - Between two ordinal level variables that have been arrayed in a bivariate table
 - Gamma is based on pairs of cases
- Gamma (like Lambda)
 - Tells us the extent to which knowledge of one variable improves our ability to predict the other variable
 - Gamma predicts the order of pairs of cases
 - If two variables are related, the order of pairs on the dependent variable (Y) is predictable from their order on the independent variable (X)
- Before computing and interpreting Gamma, it will always be useful to find and interpret the column percentages

Calculate Gamma

- To compute Gamma, two quantities must be found
 - $-n_s$ is the number of pairs of cases ranked in the <u>same</u> <u>order</u> on both variables
 - n_d is the number of pairs of cases ranked in <u>different</u> order on the variables
 - Always make sure the "low-low" cell is the "top-left" cell in your table before calculation

$$G = \frac{n_S - n_d}{n_S + n_d}$$



Interpretation of Gamma

- The PRE interpretation refers
 - To the percentage of fewer errors made in predicting the order of pairs on the dependent variable (Y) from the order of pairs on the independent variable (X)
 - Compared to the number of errors made in predicting the order of pairs on the dependent variable (Y) while <u>ignoring</u> the independent variable (X)

Guidelines for Interpreting the Strength of the Relationship for Ordinal-Level Measures of Association

Measure of Association	Strength The strength of the relationship is	
If the value is		
Between 0.00 and 0.30	Weak	
Between 0.31 and 0.60	Moderate	
Greater than 0.60	Strong	



Gamma: Strength and direction

- In addition to strength, gamma also identifies the direction of the association
- In a negative association, the variables change in different directions
 - e.g., as age increases, income decreases (or, as age decreases, income increases)
- In a positive association, the variables change in the same direction
 - e.g., as education increases, income increases (or, as education decreases, income decreases)

Example of Gamma: n_s

Efficiency	Authoritarianism Low High	
Effici		
Low	10	12
High	17	5

- To compute n_s,
 multiply each cell
 frequency by all cell
 frequencies below
 and to the right
- $n_s = 10 \times 5 = 50$
- Regardless of how many cells a table has, this procedure is the same



Example of Gamma: n_d

Efficiency	Authoritarianism		
Effici	Low	High	
Low	10	12	
High	17	5	

- To compute n_d,
 multiply each cell
 frequency by all cell
 frequencies below
 and to the left
- $n_d = 12 \times 17 = 204$
- This procedure is the same for any size table



Calculate Gamma

$$G = \frac{n_s - n_d}{n_s + n_d} = \frac{50 - 204}{50 + 204} = -0.61$$

Efficiency	Authorit	Authoritarianism					
Effici	Low	High					
Low	10	12					
High	17	5					



Interpretation of direction

- Gamma = -0.61
- Gamma is negative, so the association between authoritarianism and efficiency is negative
- As one variable decreases the other variable increases



Interpretation of strength

- Gamma = -0.61
- The absolute value of Gamma is 0.61
 - According to the guideline table this indicates a strong association
- PRE interpretation
 - We would make 61% fewer errors if we predicted the order of pairs on efficiency (Y) from the order of pairs on authoritarianism (X)
 - Compared to predicting the order of pairs on efficiency (Y) while <u>ignoring</u> authoritarianism (X)



Spearman's Rho (r_s)

- Measure of association for ordinal-level variables with a broad range of different scores and few ties between cases on either variable
- Computing Spearman's Rho
 - 1. Rank cases from high to low on each variable
 - 2. Use ranks, not the scores, to calculate Rho

$$r_{S} = 1 - \frac{6\sum D^{2}}{n(n^{2} - 1)}$$

where $\sum D^2$ is the sum of the squared differences in ranks



Interpreting Spearman's Rho

- Spearman's Rho is positive
 - As the rank of one variable increases, the rank of the other variable also increases

- Spearman's Rho is negative
 - As the rank of one variable increases, the rank of the other variable decreases



Example of Spearman's Rho (r_s)

Scores on Involvement in Jogging and Self-Esteem

Jogger	Involvement in Jogging (X)	Self-Esteem (Y)
Wendy	18	15
Debbie	17	18
Phyllis	15	12
Stacey	12	16
Evelyn	10	6
Tricia	9	10
Christy	8	8
Patsy	8	7
Marsha	5	5
Lynn	1	2



Computing Spearman's Rho (r_s)

Computing Spearman's Rho

	Involvement (X)	Rank	Self-Image (Y)	Rank	D	D^2
Wendy	18	1	15	3	-2	4
Debbie	17	2	18	1	1	1
Phyllis	15	3	12	4	-1	1
Stacey	12	4	16	2	2	4
Evelyn	10	5	6	8	-3	9
Tricia	9	6	10	5	1	1
Christy	8	7.5	8	6	1.5	2.25
Patsy	8	7.5	7	7	0.5	0.25
Marsha	5	9	5	9	0	0
Lynn	1	10	2	10	0	0
				1 - 113 .	$\Sigma D = 0$	$\Sigma D^2 = 22.5$



Result of Spearman's Rho (r_s)

- In the column headed D^2 , each difference is squared to eliminate negative signs
- The sum of this column is $\sum D^2$, and this quantity is entered directly into the formula

$$r_{S} = 1 - \frac{6\sum D^{2}}{n(n^{2} - 1)} = 1 - \frac{6(22.5)}{10(100 - 1)} = 0.86$$



Interpreting Spearman's Rho (r_s)

- Rho is positive, therefore jogging and self-image share a positive association
 - As jogging rank increases, self-image rank also increases
- On its own, Rho does not have a good strength interpretation
 - But Rho² is a PRE measure
 - For this example, $Rho^2 = (0.86)^2 = 0.74$
 - We would make 74% fewer errors if we used the rank of jogging (X) to predict the rank on self-image (Y) compared to if we ignored the rank on jogging

GSS example

Is opinion about immigration different by sex?

. svy: tab letin1 sex if year==2016, col
(running tabulate on estimation sample)

Number of strata = 65Number of PSUs = 130 Number of obs = 1,845 Population size = 1,841.4241 Design df = 65

number of immigrant s to america nowadays respondents sex should be male female Total .056 .0607 .0586 increase increase . 122 .1115 .1163 .4028 remain t .4108 .3961 .2241 .236 reduced .2305 reduced .1871 .1957 .1918 Total 1 1 1

Key: column proportion

***Commands for measures of association:

***Lambda

lambda letin1 sex if year==2016

***Chi square, Cramer's V, Gamma

tab letin1 sex if year==2016, chi V gamma

***Spearman's rank correlation coefficient

spearman letin1 sex if year==2016



Source: 2016 General Social Survey.

Lambda

- . ***Lambda
- . lambda letin1 sex

number of immigrants to america nowadays		ents sex	
should be	male	female	Total
increased a lot	49	59	108
increased a little	104	114	218
remain the same as it	329	413	742
reduced a little	181	238	419
reduced a lot	156	202	358
Total	819	1,026	1,845

Source: 2016 General Social Survey.



Chi square, Cramer's V, Gamma

- . ***Chi square, Cramer's V, Gamma
- . tab letin1 sex, chi V gamma

number of immigrants to america nowadays	respond	ents sex	
should be	male	female	Total
<pre>increased a lot increased a little remain the same as it</pre>	49 104 329	59 114 413	108 218 742
reduced a little	181	238	419
reduced a lot	156	202	358
Total	819	1,026	1,845

```
Pearson chi2(4) = 1.3515 Pr = 0.853
Cramér's V = 0.0271
gamma = 0.0321 ASE = 0.035
```

```
. ***Test statistic for Gamma: Z = gamma / ASE
```

.91714286

```
. di 1-normal(0.91714286) // p-value
.17953389
```

Source: 2016 General Social Survey.



[.] di 0.0321/0.035 // test statistic

Spearman's Rho in Stata

- . ***Spearman's rho (rank correlation coefficient)
- . spearman letin1 sex

```
Number of obs = 1845

Spearman's rho = 0.0212

Test of Ho: letin1 and sex are independent

|Prob > |t| = 0.3637
```

 $Rho^2 = (0.0212)^2 = 0.00045 = 0.045\%$



Edited table

Table 1. Opinion of the U.S. adult population about how should the number of immigrants to the country be nowadays by sex, 2004, 2010, and 2016

				· · · · · · · · · · · · · · · · · · ·	
Opinion About	Male	Female	Total	Measures of association	p-value
Number of Immigrants	(%)	(%)	(%)	Weasures of association	p-value
2004					
Increase a lot	3.19	3.74	3.48	Chi square: 2.3397	0.6740
Increase a little	6.55	6.53	6.54	Cramer's V: 0.0343	
Remain the same	36.25	34.22	35.17	Lambda: 0.0000	
Reduce a little	27.61	28.90	28.30	Gamma: -0.0050	0.4415
Reduce a lot	26.40	26.61	26.51	Spearman's rho: -0.0032	0.8852
Total	100.00	100.00	100.00		
(sample size)	(914)	(1,069)	(1,983)		
2010					
Increase a lot	4.84	3.80	4.26	Chi square: 7.0998	0.1310
Increase a little	7.33	11.10	9.44	Cramer's V: 0.0714	
Remain the same	36.44	35.46	35.89	Lambda: 0.0000	
Reduce a little	25.17	24.01	24.52	Gamma: -0.0472	0.1248
Reduce a lot	26.22	25.62	25.88	Spearman's rho: -0.0310	0.2476
Total	100.00	100.00	100.00		
(sample size)	(595)	(798)	(1,393)		
2016					
Increase a lot	5.60	6.07	5.86	Chi square: 1.3515	0.8530
Increase a little	12.20	11.15	11.63	Cramer's V: 0.0271	
Remain the same	41.08	39.61	40.28	Lambda: 0.0000	
Reduce a little	22.41	23.60	23.05	Gamma: 0.0321	0.1795
Reduce a lot	18.71	19.57	19.18	Spearman's rho: 0.0212	0.3637
Total	100.00	100.00	100.00		
(sample size)	(819)	(1,026)	(1,845)		

Note: Column percentages were estimated taking into account the complex survey design of the General Social Survey. Source: 2004, 2010, 2016 General Social Surveys.

ACS example

Is educational attainment different by age group?

. tab educgr agegr, col

Key
frequency
column percentage

	agegr									
educgr	0	16	20	25	35	45	55	65	Total	
Less than high school	571,701	89,702	10,262	25,198	30,960	35,040	39,879	74,522	877,264	
	99.97	52.61	5.51	6.49	8.25	8.52	8.44	11.67	27.29	
High school	157	59,928	71,447	119,445	111,837	141,857	184,217	259,161	948,049	
	0.03	35.15	38.39	30.78	29.79	34.50	38.97	40.58	29.49	
Some college	0	20,766	72,420	93,352	85,507	91,946	107,832	123,053	594,876	
	0.00	12.18	38.92	24.05	22.78	22.36	22.81	19.27	18.51	
College	0	105	29,469	102,919	85,850	85,309	84,454	98,425	486,531	
	0.00	0.06	15.84	26.52	22.87	20.75	17.86	15.41	15.14	
Graduate school	0	0	2,495	47,199	61,261	57,053	56,382	83,429	307,819	
	0.00	0.00	1.34	12.16	16.32	13.87	11.93	13.06	9.58	
Total	571,858	170,501	186,093	388,113	375,415	411,205	472,764	638,590	3,214,539	
	100.00	100.00	100.00	100.00	100.00	100.00	100.00	100.00	100.00	

Spearman's Rho in Stata

. spearman educgr agegr

```
Number of obs = 3214539
Spearman's rho = 0.4405
```

Test of Ho: educgr and agegr are independent Prob > |t| = 0.0000

 $Rho^2 = (0.4405)^2 = 0.1940 = 19.40\%$



ACS example: percentages

- Use column percentages from this table
- . tab educgr agegr [fweight=perwt], col

Key
frequency
column percentage

		agegr										
educgr	0	16	20	25	35	45	55	65	Total			
Less than high school	64932988	9592001	1233939	3146621	3999381	4047164	4092972	6713748	97758814			
	99.97	55.79	5.67	6.95	9.59	9.73	9.68	12.81	29.88			
High school	17628	5676286	8516860	14302836	12637092	14222739	16105938	20704168	92183547			
	0.03	33.02	39.11	31.59	30.31	34.20	38.09	39.51	28.18			
Some college	0	1915448	8462363	11380862	9705561	9436932	9710019	10211276	60822461			
	0.00	11.14	38.86	25.14	23.28	22.69	22.96	19.48	18.59			
College	0	8720	3288424	11420420	9104449	8441402	7508620	8093763	47865798			
	0.00	0.05	15.10	25.22	21.84	20.30	17.76	15.44	14.63			
Graduate school	0	0	276404	5026278	6240807	5444101	4864635	6684594	28536819			
	0.00	0.00	1.27	11.10	14.97	13.09	11.51	12.76	8.72			
Total	64950616	17192455	21777990	45277017	41687290	41592338	42282184	52407549	327167439			
	100.00	100.00	100.00	100.00	100.00	100.00	100.00	100.00	100.00			

Edited table

Table 1. Distribution of U.S. population by educational attainment and age group, 2018

Educational	Age group									
attainment	0–15	16–19	20–24	25–34	35–44	45–54	55–64	65+		
Less than high school	99.97	55.79	5.67	6.95	9.59	9.73	9.68	12.81		
High school	0.03	33.02	39.11	31.59	30.31	34.20	38.09	39.51		
Some college	0.00	11.14	38.86	25.14	23.28	22.69	22.96	19.48		
College	0.00	0.05	15.10	25.22	21.84	20.30	17.76	15.44		
Graduate school	0.00	0.00	1.27	11.10	14.97	13.09	11.51	12.76		
Total	100.00	100.00	100.00	100.00	100.00	100.00	100.00	100.00		
Population size (N)	64,950,616	17,192,455	21,777,990	45,277,017	41,687,290	41,592,338	42,282,184	52,407,549		
Sample size (n)	571,858	170,501	186,093	388,113	375,415	411,205	472,764	638,590		
Spearman's Rho	0.4405	p-value	: 0.000							

Source: 2018 American Community Survey.



