

# Lecture 4

## Segment 2

Calculating correlations

# Correlations 2

- Important topics
  - Pearson product-moment correlation coefficient ( $r$ )
  - Covariance

# Correlations 2

- Correlation coefficient ( $r$ )
  - Pearson product-moment correlation coefficient
- $r$  = the degree to which X and Y vary together, relative to the degree to which X and Y vary independently
- $r = (\text{covariance of X \& Y}) / (\text{variance of X \& Y})$

# Correlations 2

- Here comes the math!
- Formulae for  $r$ 
  - Raw score formula
  - z-score formula

# Correlations 2

- Let's quickly review calculations from Lecture 2 on descriptive statistics

# Linsanity!



# Jeremy Lin (10 games)

Points per game	(X-M)	(X-M) <sup>2</sup>
28	5.3	28.09
26	3.3	10.89
10	-12.7	161.29
27	4.3	18.49
20	-2.7	7.29
38	15.3	234.09
23	0.3	0.09
28	5.3	28.09
25	2.3	5.29
2	-20.7	428.49
$M = 227/10 = 22.7$	$M = 0/10 = 0$	$M = 922.1/10 = 92.21$

# Results

- $M = \text{mean} = 22.7$
- $SD = \text{standard deviation} = 9.6$
- $SD^2 = \text{variance} = 92.21$



# Notation

- $M$  = mean
- $SD$  = standard deviation
- $SD^2$  = variance (also known as  $MS$ )
  - $MS$  stands for Mean Squares
  - $SS$  stands for Sum of Squares

# Just one new concept ☺

- $SP = \text{Sum of cross Products}$

# Just one new concept ☺

- Review: To calculate SS
  - For each subject in the sample, calculate their deviation score
    - $(X - M_x)$
  - Square the deviation scores
    - $(X - M_x)^2$
  - Sum the squared deviation scores
    - $SS_x = \Sigma[(X - M_x)^2] = \Sigma[(X - M_x) \times (X - M_x)]$

# Just one new concept ☺

- To calculate SP
  - For each subject in the sample, calculate their deviation scores on both  $X$  and  $Y$ 
    - $(X - M_x)$
    - $(Y - M_y)$
  - Then, for each subject, multiply the deviation score on  $X$  by the deviation score on  $Y$ 
    - $(X - M_x) \times (Y - M_y)$
  - Then sum the “cross products”
    - $SP = \Sigma[(X - M_x) \times (Y - M_y)]$

# Formulae to calculate $r$

Raw score formula:

$$r = SP_{xy} / \sqrt{SS_x SS_y}$$

$$SS_x = \Sigma(X - M_x)^2 = \Sigma[(X - M_x)(X - M_x)]$$

$$SS_y = \Sigma(Y - M_y)^2 = \Sigma[(Y - M_y)(Y - M_y)]$$

$$SP_{xy} = \Sigma[(X - M_x)(Y - M_y)]$$

# Formulae to calculate $r$

Raw score formula continued:

$$r = SP_{xy} / \sqrt{SS_x SS_y}$$

$$r = \frac{\Sigma[(X - M_x)(Y - M_y)]}{\sqrt{\Sigma(X - M_x)^2 \Sigma(Y - M_y)^2}}$$

# Formulae to calculate $r$

z-score formula:

$$r = \Sigma(z_x z_y) / N$$

$$z_x = (X - M_x) / SD_x$$

$$z_y = (Y - M_y) / SD_y$$

$$SD_x = \sqrt{\Sigma(X - M_x)^2 / N}$$

$$SD_y = \sqrt{\Sigma(Y - M_y)^2 / N}$$

# Formulae to calculate $r$

z-score formula continued:

$$z_x = (X - M_x) / SD_x$$

$$z_y = (Y - M_y) / SD_y$$

$$SD_x = \sqrt{\Sigma(X - M_x)^2 / N}$$

$$SD_y = \sqrt{\Sigma(Y - M_y)^2 / N}$$



# Formulae to calculate $r$

Proof of equivalence:

$$z_x = (X - M_x) / \sqrt{\Sigma(X - M_x)^2 / N}$$

$$z_y = (Y - M_y) / \sqrt{\Sigma(Y - M_y)^2 / N}$$

$$r = \Sigma \{ [(X - M_x) / \sqrt{\Sigma(X - M_x)^2 / N}] \\ [(Y - M_y) / \sqrt{\Sigma(Y - M_y)^2 / N}] \} / N$$

# Formulae to calculate $r$

Proof of equivalence continued:

$$r = \Sigma \{ [(X - M_x) / \sqrt{\Sigma(X - M_x)^2 / N}] [(Y - M_y) / \sqrt{\Sigma(Y - M_y)^2 / N}] \} / N$$

$$r = \Sigma [(X - M_x)(Y - M_y)] / \sqrt{\Sigma(X - M_x)^2 \Sigma(Y - M_y)^2}$$

$$r = SP_{xy} / \sqrt{SS_x SS_y}$$

# Variance and covariance

- Variance =  $MS = SS / N$
- Covariance =  $COV = SP / N$
- Correlation is standardized COV
  - Standardized so the value is in the range -1 to 1

# Note on the denominators

- Correlation for descriptive purposes
  - Divide by  $N$
- Correlation for inferential purposes
  - Divide by  $N-1$

# Correlations 2: Review

- Important topics
  - Pearson product-moment correlation coefficient ( $r$ )
  - Covariance

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