# PSY 201: Statistics in Psychology <br> Lecture 10 <br> Correlation 

How changes in one variable correspond to change in another variable.

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## CORRELATION

- two variables may be related
- SAT scores, GPA
- hours in therapy, self-esteem
- grade on homeworks, grade on exams
- number of risk factors, probability of getting AIDS
- height, points in basketball
- ...
- how do we show the relationship?
- scattergrams


## SCATTERGRAMS

- plot value of one variable against the value of the other variable



## RELATIONSHIPS

- Identifying these types of relationships is one of the key issues in statistical analysis
- Consider a 1999 study that reported a relationship between the use of nightlights in a child's room and the tendency of the child to need glasses

- My daughter slept with a nightlight. Was I a bad father?


## COMPLICATIONS

- Clearly there is a relationship between using a nightlight and needing glasses
- However, it's not clear what the nature of the relationship involves
- It could be that the extra light somehow influences the child's eyes and causes the need for glasses
- Or it could be that needing glasses will somehow co-occur with the use of a nightlight (e.g., children who need glasses will want a night light, or their parents will want a nightlight)
- Finding a relationship is necessary for establishing causation, but it is not enough


## SPURIOUS CORRELATION

- Since so many variables get measured, it is easy to identify spurious correlations
- Sometimes there is an explanation for the relationship:

- (increased use of technology)


## SPURIOUS CORRELATION

- Since so many variables get measured, it is easy to identify spurious correlations
- Sometimes there is no explanation for the relationship:



## POSITIVE CORRELATION

- First, we need to understand how to quantify the existence of a relationship.
- Increases in the value of one variable tend to occur with increases in the value of the other variable
- SAT scores and exam scores



## NEGATIVE CORRELATION

- Increases in the value of one variable tend to occur with decreases in the value of the other variable
- temperature and number of people with frostbite



## PERFECT CORRELATIONS

- perfect positive correlation

- perfect negative correlation



## NO CORRELATION

- no correlation
- balance of larger and smaller values




## CORRELATION COEFFICIENT

- quantitative measure of correlation
- bounded between

$$
-1.0 \&+1.0
$$

- correlation coefficient of -1.0 indicates perfect negative correlation
- correlation coefficient of +1.0 indicates perfect positive correlation
- correlation coefficient of 0.0 indicates no correlation
- values in between give ordinal measures of relationship


## PEARSON $r$

- Pearson product-moment correlation coefficient
- one correlation coefficient for quantitative data (the most important one)

$$
r=\frac{\text { degree to which } X \text { and } Y \text { vary together }}{\text { degree to which } X \text { and } Y \text { vary separately }}
$$

- several formulas
- z-scores
- Deviation scores
- Raw scores
- Covariance
- all give the same result!


## z SCORES

- Two steps
- Convert raw scores into z scores
- Find the mean of cross-products

$$
r_{x y}=\frac{\sum z_{x} z_{y}}{n-1}
$$

## z SCORES

- what does this calculation do?
- suppose you have two distributions that have a positive correlation
- then a large value of $X$ will be above $\bar{X}$ and have a positive $z_{X}$ score
- and a corresponding $Y$ will be above $\bar{Y}$ and have a positive $z_{y}$ score
- Thus the cross-product

$$
z_{x} z_{y}
$$

- will be positive


## PEARSON $r$

- also a small value of $X$ will be below $\bar{X}$ and have a negative $z_{X}$ score
- and the corresponding $Y$ will be below $\bar{Y}$ and have a negative $z_{y}$ score
- Thus

$$
z_{x} z_{y}
$$

- will again be positive
- to find the average, sum all the products (positive numbers) we divide by $n-1$

$$
r_{x y}=\frac{\sum z_{x} z_{y}}{n-1}
$$

- still a positive number!


## PEARSON $r$

- exactly the opposite is true for negatively correlated distributions
- then a large value of $X$ will be above $\bar{X}$ and have a positive $z_{X}$ score
- and a corresponding $Y$ will be below $\bar{Y}$ and have a negative $z_{y}$ score
- Thus

$$
z_{x} z_{y}
$$

- will be negative


## PEARSON $r$

- while a small value of $X$ will be below $\bar{X}$ and have a negative $z_{x}$ score
- and the corresponding $Y$ will be above $\bar{Y}$ and have a positive $z_{y}$ score
- Thus

$$
z_{x} z_{y}
$$

- will again be negative
- to find the average, sum all the products (negative numbers) we divide by $n-1$

$$
r_{x y}=\frac{\sum z_{x} z_{y}}{n-1}
$$

- still a negative number!


## DEVIATION FORMULA

- it is awkward to convert to $z$ scores
- we can get the same number with deviation scores

$$
\begin{aligned}
& x=X-\bar{X} \\
& y=Y-\bar{Y}
\end{aligned}
$$

- deviation score formula

$$
r_{x y}=\frac{\Sigma x y}{\sqrt{\sum x^{2} \Sigma y^{2}}}
$$

## RAW SCORE FORMULA

- it is awkward to calculate deviation scores
- raw score formula

$$
r_{x y}=\frac{n \Sigma X Y-\Sigma X \Sigma Y}{\sqrt{\left[n \Sigma X^{2}-(\Sigma X)^{2}\right]\left[n \Sigma Y^{2}-(\Sigma Y)^{2}\right]}}
$$

## COVARIANCE FORMULA

$$
\text { covariance }=s_{x y}=\frac{\Sigma(X-\bar{X})(Y-\bar{Y})}{n-1}
$$

- average cross-product of deviation scores (similar to variance)
- Pearson $r$ turns out to be:

$$
r_{x y}=\frac{s_{x y}}{s_{x} s_{y}}
$$

- where $s_{x}$ and $s_{y}$ are the standard deviations of their respective distributions


## EXAMPLE

| $X$ | $Y$ |
| :--- | :---: |
| 595 | 68 |
| 520 | 55 |
| 715 | 65 |
| 405 | 42 |
| 680 | 64 |
| 490 | 45 |
| 565 | 56 |
| 580 | 59 |
| 615 | 56 |
| 435 | 42 |
| 440 | 38 |
| 515 | 50 |
| 380 | 37 |
| 510 | 42 |
| 565 | 53 |



## EXAMPLE

- standard score formula

$$
r_{x y}=\frac{\sum z_{x} z_{y}}{n-1}=\frac{12.67}{14}=0.905
$$

| $X$ | $Y$ | $z_{x}$ | $z_{y}$ | $z_{x} z_{y}$ |
| ---: | ---: | ---: | ---: | ---: |
| 595 | 68 | 0.63 | 1.64 | 1.03 |
| 520 | 55 | -0.15 | 0.35 | -0.52 |
| 715 | 65 | 1.88 | 1.34 | 1.26 |
| 405 | 42 | -1.34 | -0.94 | 1.87 |
| 680 | 64 | 1.51 | 1.24 | 0.29 |
| 490 | 45 | -0.46 | -0.64 | 0.14 |
| 565 | 56 | 0.32 | 0.45 | 0.36 |
| 580 | 59 | 0.48 | 0.74 | 0.38 |
| 615 | 56 | 0.84 | 0.45 | 0.97 |
| 435 | 42 | -1.03 | -0.94 | 1.29 |
| 440 | 38 | -0.97 | -1.33 | 0.03 |
| 515 | 50 | -0.20 | -0.15 | 2.29 |
| 380 | 37 | -1.60 | -1.43 | 0.24 |
| 510 | 42 | -0.25 | -0.94 | 0.05 |
| 565 | 53 | 0.32 | 0.15 |  |
| $\sum X=8010$ | $\sum Y=772$ | $\sum z_{x}=0.0$ | $\sum z_{y}=0.0$ | $\sum z_{x} z_{y}=12.67$ |

## EXAMPLE

- deviation score formula

$$
r_{x y}=\frac{\Sigma x y}{\sqrt{\sum x^{2} \Sigma y^{2}}}=\frac{12332.00}{\sqrt{(130460.0)(1429.72)}}=0.903
$$

| $X$ | $Y$ | $x$ | $y$ | $x y$ |
| ---: | ---: | ---: | ---: | ---: |
| 595 | 68 | 61.0 | 16.53 | 1008.33 |
| 520 | 55 | -14.0 | 3.53 | -49.42 |
| 715 | 65 | 181.0 | 13.53 | 2448.93 |
| 405 | 42 | -129.0 | -9.47 | 1221.63 |
| 680 | 64 | 146.0 | 12.53 | 1829.38 |
| 490 | 45 | -44.0 | -6.47 | 284.68 |
| 565 | 56 | 31.0 | 4.53 | 140.43 |
| 580 | 59 | 46.0 | 7.53 | 346.38 |
| 615 | 56 | 81.0 | 4.53 | 366.93 |
| 435 | 42 | -99.0 | -9.47 | 937.53 |
| 440 | 38 | -94.0 | -13.47 | 1266.18 |
| 515 | 50 | -19.0 | -1.47 | 27.93 |
| 380 | 37 | -154.0 | -14.47 | 2228.38 |
| 510 | 42 | -24.0 | -9.47 | 227.28 |
| 565 | 53 | 31.0 | 1.53 | 47.43 |

- $\Sigma x^{2}=130460.0$ and $\Sigma y^{2}=1429.72$


## EXAMPLE

- raw score formula

$$
\begin{aligned}
& n \Sigma X Y-\Sigma X \Sigma Y \\
& r_{x y}=\frac{}{\sqrt{\left[n \Sigma X^{2}-(\Sigma X)^{2}\right]\left[n \Sigma Y^{2}-(\Sigma Y)^{2}\right]}} \\
& (15)(424580)-(8010)(772) \\
& =0.903
\end{aligned}
$$

- $\Sigma X^{2}=4407800$ and $\Sigma Y^{2}=41162$


## EXAMPLE

- covariance formula

$$
r_{x y}=\frac{s_{x y}}{s_{x} s_{y}}=\frac{880.86}{(96.53)(10.11)}=0.903
$$

- where,

$$
\begin{gathered}
s_{x y}=\frac{\Sigma x y}{n-1}=\frac{12332}{14}=880.86 \\
s_{x}=\sqrt{\frac{\Sigma x^{2}}{n-1}}=\sqrt{\frac{130460}{14}}=96.53 \\
s_{y}=\sqrt{\frac{\Sigma y^{2}}{n-1}}=\sqrt{\frac{1429.72}{14}}=10.11
\end{gathered}
$$

## CORRELATION

- $r$ measures correlation between two variables
- not just any two variables
- The two variables must be paired observations.
- Variables must be quantitative (interval or ratio scale).


## CONCLUSIONS

- correlation
- scattergrams
- Pearson $r$
- formulas


## NEXT TIME

- factors affecting $r$
- interpreting $r$

Is there a link between IQ and problem solving ability?

