Correlation & Linear Regression

Slides adopted from the Internet

Roadmap

Linear Correlation

Spearman's rho correlation

Kendall's tau correlation

Linear regression

Linear correlation

Recall: Covariance

$$cov(x,y) = \frac{\sum_{i=1}^{n} (x_i - \overline{X})(y_i - \overline{Y})}{n-1}$$

Interpreting Covariance

 $cov(X,Y) > 0 \longrightarrow X$ and Y are positively correlated

 $cov(X,Y) < 0 \longrightarrow X$ and Y are inversely correlated

 $cov(X,Y) = 0 \longrightarrow X$ and Y are independent

Correlation coefficient

Pearson's Correlation Coefficient is standardized covariance (unitless):

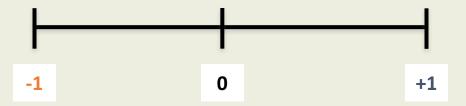
$$r = \frac{\text{cov} \, ariance(x, y)}{\sqrt{\text{var} \, x} \sqrt{\text{var} \, y}}$$

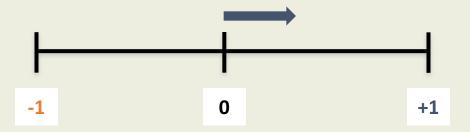
Calculating by hand...

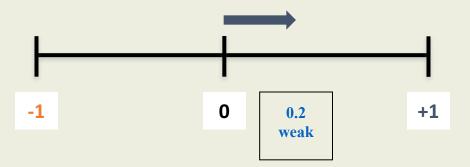
$$\hat{r} = \frac{\sum_{i=1}^{n} (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\text{var } x} \sqrt{\text{var } y}} = \frac{\sum_{i=1}^{n} (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_{i=1}^{n} (x_i - \bar{x})^2} \sqrt{\sum_{i=1}^{n} (y_i - \bar{y})^2}} \sqrt{\frac{\sum_{i=1}^{n} (y_i - \bar{y})^2}{n-1}}$$

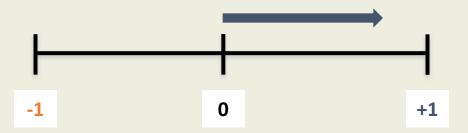
Correlation

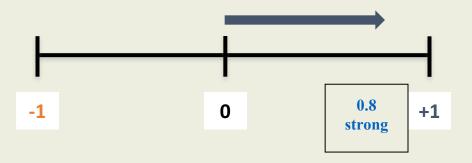
- Measures the relative strength of the *linear* relationship between two variables
- Unit-less
- Ranges between –1 and 1
- The closer to −1, the stronger the negative linear relationship
- The closer to 1, the stronger the positive linear relationship
- The closer to 0, the weaker any positive linear relationship

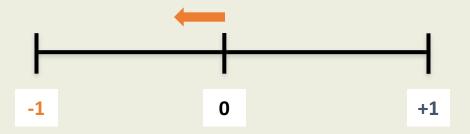


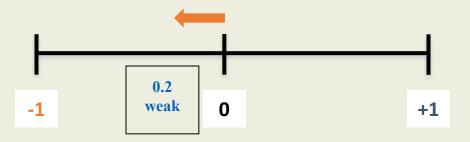


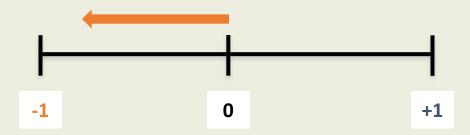


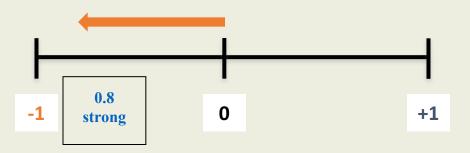


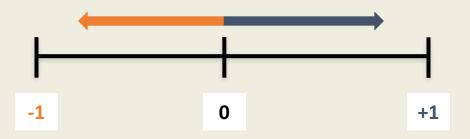




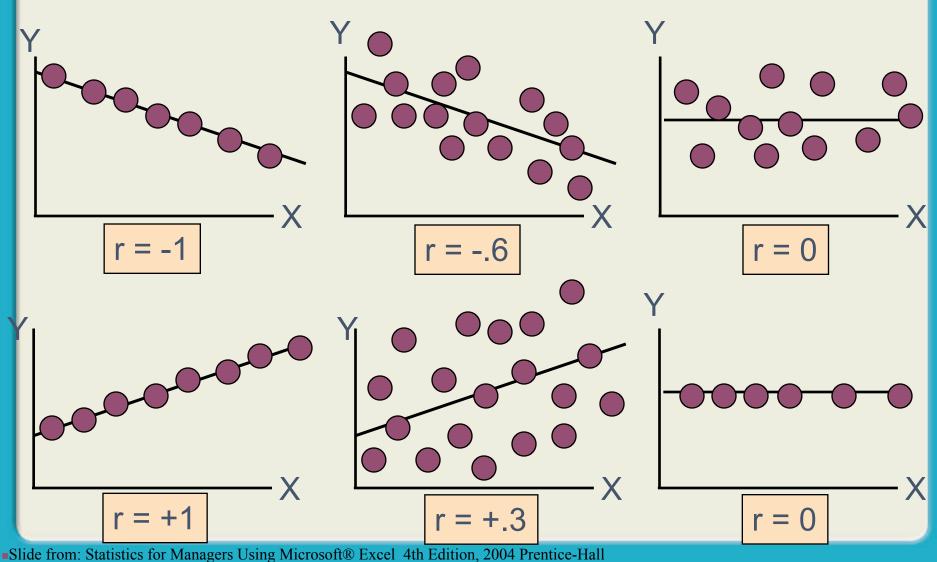






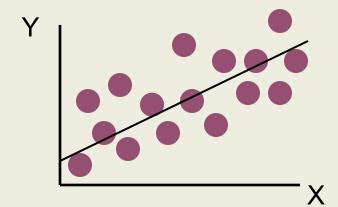


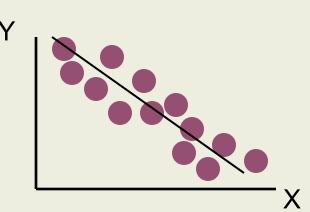
Scatter Plots of Data with Various Correlation Coefficients



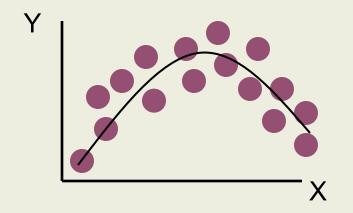
Linear Correlation

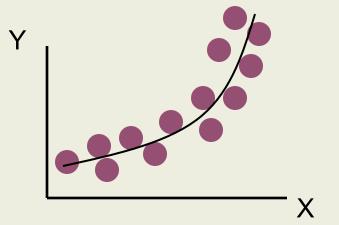
Linear relationships



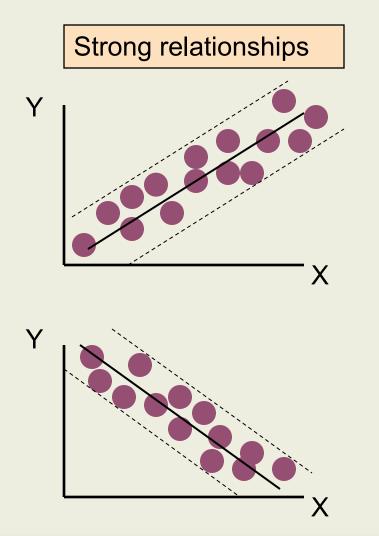


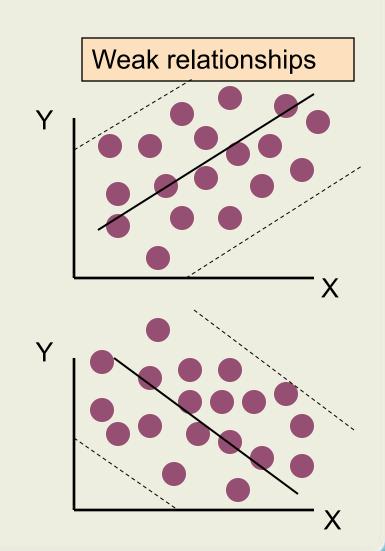
Curvilinear relationships



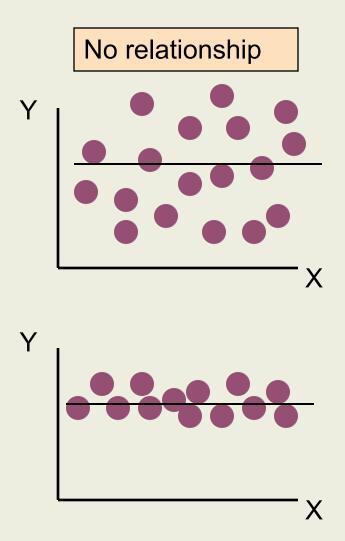


Linear Correlation

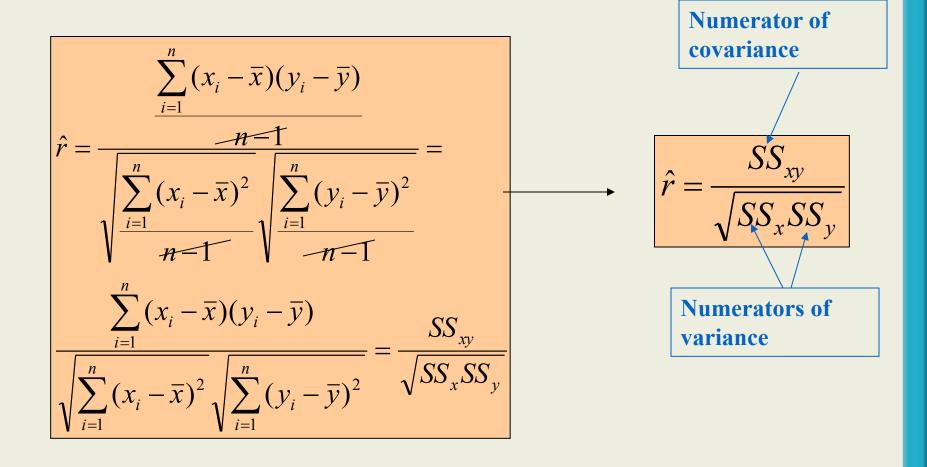




Linear Correlation



Simpler calculation formula...



Pearson r correlation assumptions

Both variables should be normally distributed

 A straight-line (linear) relationship between two variables

Data are normally distributed around the regression line

Roadmap

Linear Correlation

Spearman's rho correlation

Kendall's tau correlation

Linear regression

Spearman's Rank-Order Correlation

For Independence Questions

Welcome to the Spearman's Rho Test of Independence Learning Module

(i.e., does not assume data distribution)

• Spearman's "Rho' is a non-parametric analogue to the Pearson Product Moment Correlation.

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- Spearman's Rho is designed to estimate the coherence or lack of coherence of two variables (as in the Pearson Product Moment Correlation).
- It is calculated based on the rank-ordered (ordinal) data rather than the means and standard deviation used in the Pearson Product Moment Correlation.

 Here is an illustration of the difference between a Pearson Correlation and a Spearman's Rho

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- Are race times of athletes who participated in both biking and running competitions independent of one another? (This is a Pearson Correlation question because we are dealing with continuous variables)

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Individuals	Biking Event race times	Running Event race times
Bob	4.5 hours	4.0 hours
Conrad	7.0 hours	2.5 hours
Dallen	5.2 hours	2.8 hours
Ernie	6.0 hours	2.9 hours
Fen	6.3 hours	3.3 hours
Gaston	5.1 hours	2.3 hours

- Here is an illustration of the difference between a Pearson Correlation and a Spearman's Rho
- Are race times of athletes who participated in both biking and running competitions independent of one another? (This is a Pearson Correlation question because we are dealing with continuous variables)
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	<u> </u>	
Individuals	Biking Event	Running Event
	race times	race times
Bob	1 st	6 th
Conrad	6 th	2 nd
Dallen	3 rd	3 rd
Ernie	4 th	4 th
Fen	5 th	5 th
Gaston	2 nd	1 st

• In summary, if at least one of two variables to be correlated are based on an underlying ordinal measurement, the Spearman's Rho is an appropriate estimate.

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For example -

Interval or continuous Data

Ordinal or rankordered Data

Individuals	Biking Event race times in minutes	Running Event placement
Bob	55	6 th
Conrad	25	2 nd
Dallen	29	3 rd
Ernie	33	4 th
Fen	39	5 th
Gaston	23	1 st

• For example – Interval or continuous Data

Ordinal or rankordered Data

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Bob	55	6 th
Conrad	25	2 nd
Dallen	29	3 rd
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Fen	39	5 th
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Because this
data is
ordinal or
rank ordered
we will use
Spearman's
Rho

• For example – Interval or continuous Data Ordinal or rank-ordered Data

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Bob	55	6 th
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• or

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Ordinal or rankordered Data

Ordinal or rank- Interval or ordered Data continuous Data

	Oracica Data	COTTENTIA DA LA DA LA
Because this	s Biking Event placement	Running Event race times
data is ordinal or	1 st	4.0 hours
rank ordered	6 th	2.5 hours
we will use	3 rd	2.8 hours
Spearman's	4 th	2.9 hours
Rho	5 th	3.3 hours
Gaston	2 nd	2.3 hours

• If both variables are on an interval scale, but one or both are significantly skewed, then Spearman's Rho is an appropriate estimate that compensates for distortion of the mean.

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• For example: Interval – heavily normally skewed data distributed Data

Individuals	Biking Event race times	Running Event race times
Bob	4.5 hours	4.0 hours
Conrad	4.6 hours	2.5 hours
Dallen	4.7 hours	2.8 hours
Ernie	5.0 hours	2.9 hours
Fen	20.0 hours	3.3 hours
Gaston	28.0 hours	2.3 hours

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How to calculate Rho?

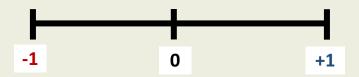
$$\rho = 1 - \frac{6\sum d_i^2}{n(n^2 - 1)}$$

Where:

P= Spearman rank correlation

di= the difference between the ranks of corresponding values Xi and Yi

n= number of value in each data set



• Spearman's Rho renders a result that is similar to the Pearson Correlation



 Therefore it shares the same properties as these other methods:



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A result like this would be evidence of independence

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- For example:

Individuals	Rank order for	Rank order for
	Biking Event	Running Event
Bob	1 st	1 st
Conrad	2 nd	1 st
Dallen	2 nd	2 nd
Ernie	3 rd	3 rd
Fen	4 th	4 th
Gaston	5 th	4 th

^{*}use Kendall's Tau when there are rank ordered ties.

Spearman's Rho Assumptions

- non-parametric: it does not assume any assumptions about the distribution of the data
- Is the appropriate correlation analysis when the variables are measured on a scale that is at least ordinal.
- Scores on one variable must be monotonically related to the other variable.
- Cannot deal with ties

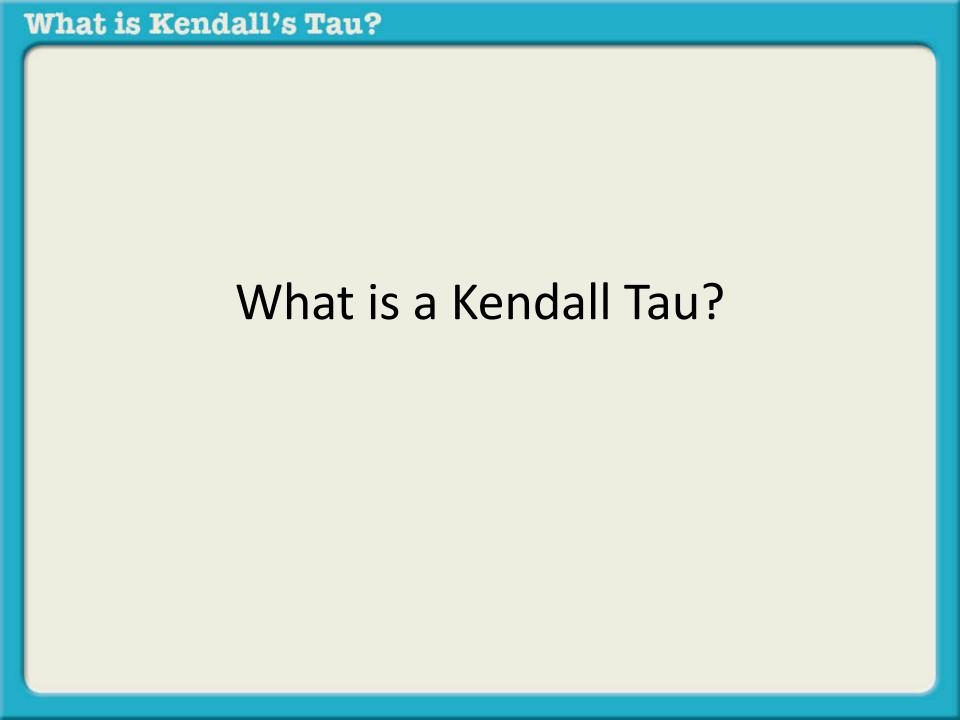
Roadmap

Linear Correlation

Spearman's rho correlation

Kendall's tau correlation

Linear regression



Kendall's Tau is a nonparametric analogue to the Pearson Product Moment Correlation.

Similar to Spearman's Rho, Kendall's Tau operates on rank-ordered (ordinal) data but is particularly useful when there are tied ranks.

Let's consider an investigation that would lend itself to being analyzed by Kendall's Tau:

An iron man competition consists of three consecutive events:

An iron man competition consists of three consecutive events: Biking 110 miles,



An iron man competition consists of three consecutive events: Biking 110 miles, Swimming 2.5 miles



An iron man competition consists of three consecutive events: Biking 110 miles, Swimming 2.5 miles and Running 26.2 miles



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An iron man competition consists of three consecutive events: Biking 110 miles, Swimming 2.5 miles and Running 26.2 miles. Researchers are interested in the relationship between the rank ordered results from the biking and the running events. Here is the data for 6 individuals who competed:

Individuals	Rank order for Biking Event	Rank order for Running Event
Bob		
Conrad		
Dallen		
Ernie		
Fen		
Gaston		

Individuals	Rank order for Biking Event	Rank order for Running Event
Bob	1 st	, and the second
Conrad	2 nd	
Dallen	2 nd	
Ernie	3 rd	
Fen	4 th	
Gaston	5 th	

Individuals	Rank order for Biking Event	Rank order for Running Event
Bob	1 st	1 st
Conrad	2 nd	1 st
Dallen	2 nd	2 nd
Ernie	3 rd	3 rd
Fen	4 th	4 th
Gaston	5 th	4 th

Because both variables are expressed as rank ordered data, we will use either a Kendall's Tau or a Spearman's Rho.

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Note – even if only one variable were ordinal and the other were scaled or nominal, you would still use Kendall's Tau or a Spearman's Rho by virtue of having *one ordinal variable*.

Because there are ties in the data, we will use Kendall's Tau *instead* of the Spearman's Rho.

How to calculate Tau?

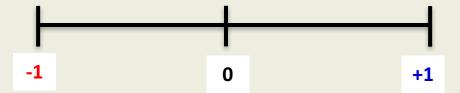
Let $(x_1, y_1), (x_2, y_2), ..., (x_n, y_n)$ be a set of observations of the joint random variables X and Y respectively, such that all the values of (x_i) and (y_i) are unique. Any pair of observations (x_i, y_i) and (x_j, y_j) are said to be *concordant* if the ranks for both elements agree: that is, if both $x_i > x_j$ and $y_i > y_j$ or if both $x_i < x_j$ and $y_i < y_j$. They are said to be *discordant*, if $x_i > x_j$ and $y_i < y_j$ or if $x_i < x_j$ and $y_i > y_j$. If $x_i = x_j$ or $y_i = y_j$, the pair is neither concordant nor discordant.

The Kendall τ coefficient is defined as:

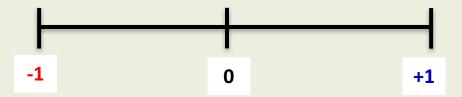
$$\tau = \frac{(\text{number of concordant pairs}) - (\text{number of discordant pairs})}{\frac{1}{2}n(n-1)}.^{[3]}$$

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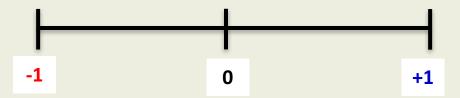
Individuals	Rank order for Biking Event	Rank order for Running Event
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Dallen	2 nd	2 nd
Ernie	3 rd	3 rd
Fen	4 th	4 th
Gaston	5 th	4 th



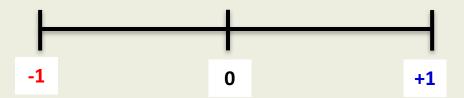
Kendall's Tau renders a result that is similar to Spearman's Rho and the Pearson Correlation



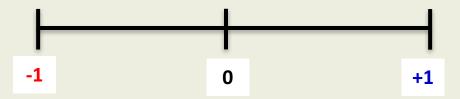
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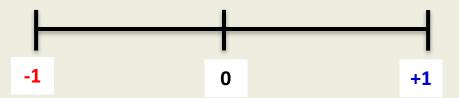
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Kendall's Tau Assumptions

 non-parametric: does not assume any assumptions about the distribution of the data

 Is the appropriate correlation analysis when the variables are measured on a scale that is at least ordinal.

Can deal with ties

Binned Kenall Correlation

Use this "binned" Kendall correlation under two scenarios:

- Skewed data distribution
 - To this end, we look at the average value for each bin and compute the correlation on the binned data.
- Amount of the data so large that rank correlation is computationally expensive
 - The binned correlation retains the qualitative properties that we want to highlight with lower compute cost.

Roadmap

Linear Correlation

Spearman's rho correlation

Kendall's tau correlation

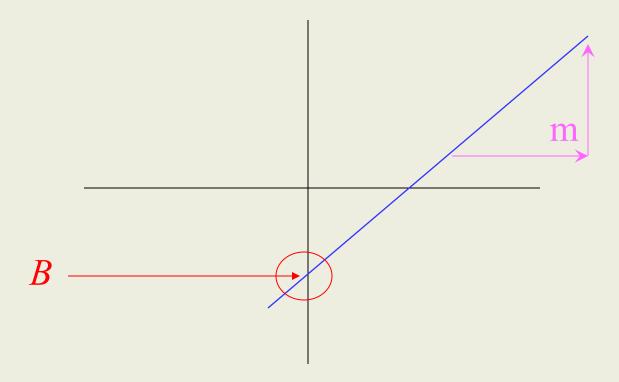
Linear regression

Linear regression

In correlation, the two variables are treated as equals. In regression, one variable is considered independent (=predictor) variable (X) and the other the dependent (=outcome) variable Y.

What is "Linear"?

- Remember this:
- \blacksquare Y=mX+B?



What's Slope?

A slope of 2 means that every 1-unit change in X yields a 2-unit change in Y.

Prediction

If you know something about X, this knowledge helps you predict something about Y. (Sound familiar?...sound like conditional probabilities?)

Regression equation...

Expected value of y at a given level of x=

$$E(y_i/x_i) = \alpha + \beta x_i$$

Predicted value for an individual...

$$y_i = \alpha + \beta * x_i + \text{random error}_i$$

Fixed – exactly on the line

Assumption: Follows a normal distribution

Estimating the intercept and slope: least squares estimation

** Least Squares Estimation

A little calculus....

What are we trying to estimate? β , the slope, from

What's the constraint? We are trying to minimize the squared distance (hence the "least squares") between the observations themselves and the predicted values, or (also called the "residuals", or left-over unexplained variability)

Difference_i =
$$y_i - (\beta x + \alpha)$$
 Difference_i² = $(y_i - (\beta x + \alpha))^2$

Find the β that gives the minimum sum of the squared differences. How do you maximize a function? Take the derivative; set it equal to zero; and solve. Typical max/min problem from calculus....

$$\frac{d}{d\beta} \sum_{i=1}^{n} (y_i - (\beta x_i + \alpha))^2 = 2(\sum_{i=1}^{n} (y_i - \beta x_i - \alpha)(-x_i))$$
$$2(\sum_{i=1}^{n} (-y_i x_i + \beta x_i^2 + \alpha x_i)) = 0...$$

From here takes a little math trickery to solve for β ...

Resulting formulas...

Slope (beta coefficient) =
$$\hat{\beta} = \frac{Cov(x,y)}{Var(x)}$$

Intercept= Calculate:
$$\hat{\alpha} = \overline{y} - \hat{\beta}\overline{x}$$

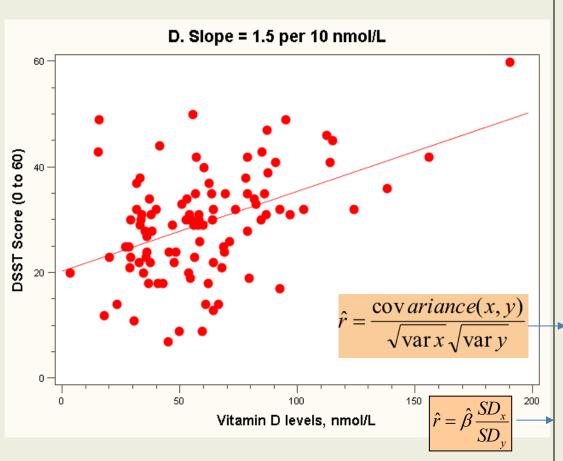
 $(\overline{x},\overline{y})$ Regression line always goes through the point:

Relationship with correlation

$$\hat{r} = \hat{\beta} \frac{SD_x}{SD_y}$$

In correlation, the two variables are treated as equals. In regression, one variable is considered independent (=predictor) variable (X) and the other the dependent (=outcome) variable (X).

Example:



$$SDx = 33 \text{ nmol/L}$$

$$Cov(X,Y) = 163$$

points*nmol/L

Beta =
$$163/33^2 = 0.15$$

points per nmol/L

$$r = 163/(10*33) = 0.49$$

$$r = 0.15 * (33/10) = 0.49$$

Pearson r correlation assumptions

Both variables should be normally distributed

 A straight-line (linear) relationship between two variables

Data are normally distributed around the regression line

Summary

Assumptions	Pearson r /linear regression	Spearman's Rho	Kendall's Tau
distributions of two variables	both are normally distributed	no assumption	no assumption
variable property	both are numbers	at least ordinal	at least ordinal
relationship between two variables	linear	scores on one variable must be monotonically related to the other variable	does not assume monotonic relationship
misc.	data are normally distributed around the regression line	cannot deal with tie	can deal with ties