please get a handout from the back!

Stats 101/101G/108 Workshop

Regression and Correlation [RC]

2020

We'll start a (0.05 am...

by Leila Boyle



Stats 101/101G/108 Workshops

The Statistics Department offers workshops and one-to-one/small group assistance for Stats 101/101G/108 students wanting to improve their statistics skills and understanding of core concepts and topics.

Leila's website for Stats 101/101G/108 workshop hand-outs and information is here: www.tinyURL.com/stats-10x

Resources for this workshop, including pdfs of this hand-out and Leila's scanned slides showing her working for each problem are available here: www.tinvURL.com/stats-RC

Want to get in touch with Leila?

Leila Boyle

Undergraduate Statistics Assistance, Department of Statistics Room 303S.288 (second floor of the Science Centre, Building 303S) l.boyle@auckland.ac.nz; (09) 923-9045; 021 447-018

Want help with Stats?

Stats 101/101G/108 appointments

Book your preferred time with Leila here: www.tinyurl.com/appt-stats, or contact her directly (see above for her contact details).

Stats 101/101G/108 Workshops

Workshops are run in a relaxed environment, and allow plenty of time for questions. In fact, this is encouraged ©

Please make sure you bring your calculator with you to all of these workshops!

Preparation at the beginning of the semester:

Multiple identical sessions of a preparation workshop are run at the beginning of the semester to get students off to a good start – come along to whichever one suits your schedule!

o Basic Maths and Calculator skills for Statistics

www.tinyURL.com/stats-BM

First half of the semester

Five theory workshops are held during the first half of the semester:

Exploratory Data Analysis

www.tinyURL.com/stats-EDA

- Proportions and Proportional Reasoning <u>www.tinyURL.com/stats-PPR</u>
- Observational Studies, Experiments, Polls and Surveys

www.tinyURL.com/stats-OSE

Confidence Intervals: Means
www.tinyURL.com/stats-CIM

Confidence Intervals: Proportions

www.tinyURL.com/stats-CIP

Second half of the semester

Five theory workshops and one computing workshop are held during the second half of the semester:

Statistics Theory Workshops

Hypothesis Tests: Means (part 1) www.tinyURL.com/stats-HTM

Hypothesis Tests: Means (part 2005) www.tinyURL.com/stats-HTM

Chi-Square Tests Change Tests C

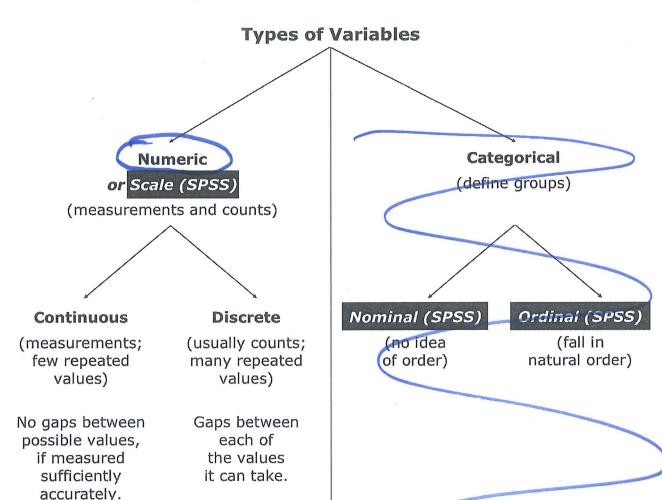
Computer Workshop: Hypothesis Tests in SPSS

www.tinyURL.com/stats-HTS

Useful Computer Resource:

If you haven't used SPSS before, try working your way through this self-paced tutorial: www.tinyURL.com/stats-IS

Regression and Correlation



Examples

Weight; Height; Length; Temperature; Total course mark (out of 100); Age [large range]

Examples

Number of siblings; Number of cars per household; Assignment mark (out of 10); Age [small range]

Examples

Gender; Ethnicity; Degree (BA, BCom, BSc, etc.); Major (Chemistry, Statistics, etc.)

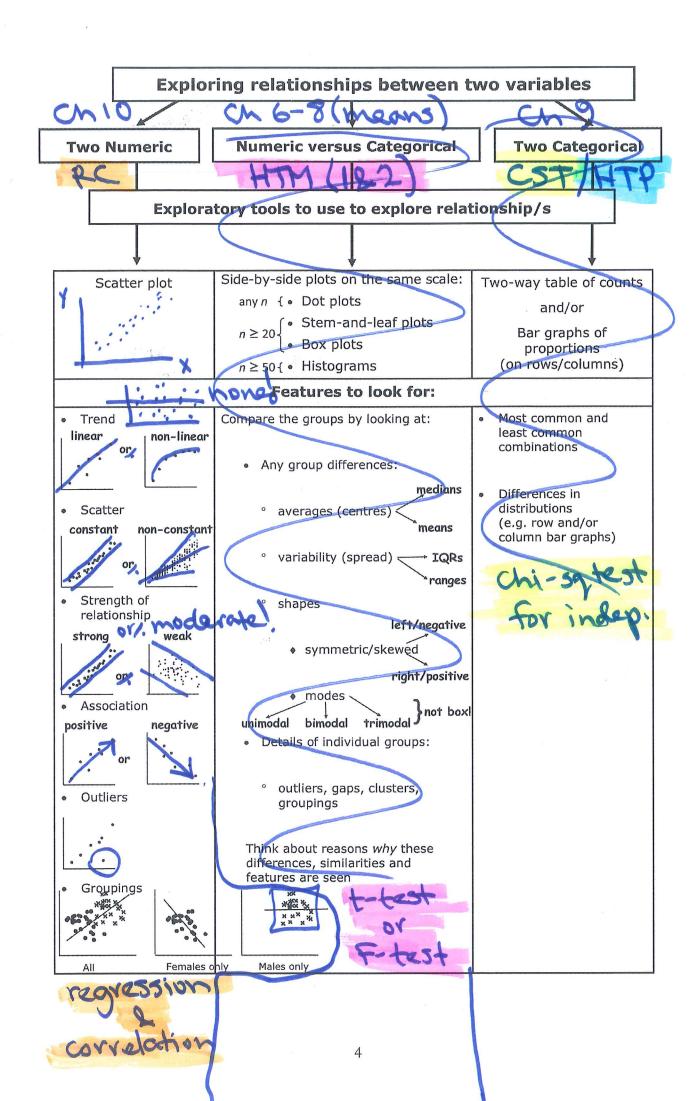
Examples

Size (small, medium, large); Income bracket; Grade (A+, A, ...); Age group (<25, 25-39, 40+)

■Useful reference: Chance Encounters, pages 40 - 42

The main tool for comparing two numeric variables is the scatter plot. What to look for in a scatter plot:

- o Trend (pattern)
- Scatter
- Strength of the relationship
- Association
- Outliers
- o Groupings



Regression

y=mx+c -> y=c+mx

Regression looks at the relationship between two numeric variables where the two variables take on special roles:

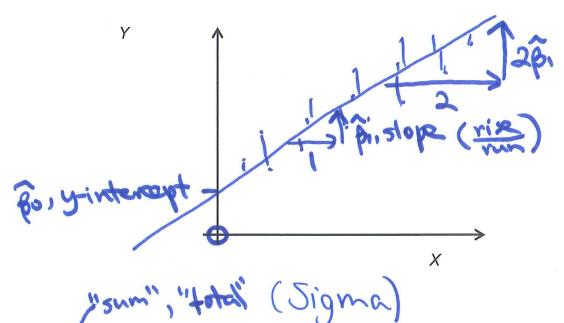
- X is used to **explain** or **predict** the behaviour of Y
- X is the explanatory or independent variable
- Y is the dependent or response variable

The two main components of the regression model are:

- trend and
- o scatter.

see back page for Formulae Sheet

We use a **least squares regression line** $\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x$ fitted by the computer / calculator to estimate the unknown population parameters β_0 and β_1 .



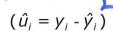
The single **least squares regression line** for each linear regression model:

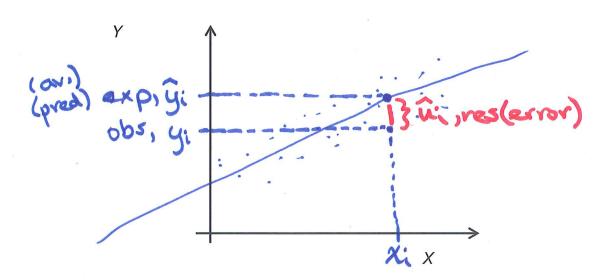
- minimises the sum of the squared residuals/prediction errors
- has \sum esiduals = 0 (but so do many other lines)
- has $(\overline{x}, \overline{y})$ lying on it

Residuals

- Errors, residuals or prediction errors are all terms for the same thing.
- A residual is the (vertical) distance between the actual observed value y_i and the expected estimated value $\hat{y_i}$, i.e.

Errors = observed - expected





Hypotheses

Recall the *t*-test?

 $t_0 = \frac{estimate - hypothesised value}{std error}$

see back page for Formulae Sheet

where in this case the degrees of freedom are:

df = n - 2

 $H_0: \beta_1 = 0$ 5 lope

(there <u>is no</u> linear relationship)

 H_1 : $\beta_1 \neq 0$

(there <u>is a</u> linear relationship)

ory the

	•							
predic	Regression	ays co	omes ou	t in	the same	e way:	7	Af=10-
	Coefficients(a)		NY NY	Z	160	A CONTRACTOR OF THE PARTY OF TH		
	Model	X	Unstandar B		Coefficients Std. Error	Standardized Coefficients Beta	t	Sig.
1-interce	(Constant)		$\hat{\beta}_{o}$		$se(\hat{eta}_o)$		t ₀	p-value
Slope, B	X-axis_Varia	ble	$\hat{\beta}_1$		$se(\hat{eta}_{\scriptscriptstyle 1})$	r	t_0	p-value
	a Dependent Variable	axis_	Variabl	e		4/		
		· +	to: B	= (0 +	- /	Ho =	et-hu
			40:β1 H1:β		A 1			349 0
		12	HI. b	17	0 4	-0%	4 :	= 3 - 1

Correlations

	*	<i>X</i> -axis_	_Variable	Y-axis	_Variable
<i>X</i> -axis_Variable	Pearson Correlation		1		r
	Sig. (2-tailed)				p-value
,	N		n		n
<i>Y</i> -axis_Variable	Pearson Correlation		r	y	1
	Sig. (2-tailed)		p-value		
	N		n		, n

Recall: The P-value:

- In regression, we are carrying out t-tests, just like in Chapter 7 and 8!
- Therefore, the P-value:
 - is the <u>conditional</u> probability of observing a test statistic as extreme as that observed or more so, <u>given that</u> the null hypothesis, H_0 , is true.
 - is the probability that sampling variation would produce an estimate that is at least as far from the hypothesised value than the estimate we obtained from our data, assuming that the null hypothesis, H_0 , is true.
 - \checkmark measures the strength of evidence **against** H_0 .

memorise o

• We interpret the P-value as a <u>description</u> of the strength of evidence against the null hypothesis, H_0 . The smaller the P-value, the stronger the evidence against H_0 :

and evidence against			
	P-value	Evidence against <i>H</i> ₀	
	> 0.10	None	
			5
	≈ 0.07	Weak	Walley 18
	≈ 0.05	Some	- Mark
	≈ 0.01	Strong	
	≤ 0.001	Very Strong	

• An alternative approach often found in research articles and news items is to describe the test result as (statistically) significant or not significant. A test result is said to be significant when the *P-value* is "small enough"; usually people say a *P-value* is "small enough" if it is less than 0.05 (5%):

Testing at a 5% level of significance:

<i>P</i> -value	Test result	Action	_ / _
< 0.05	Significant	Reject H_0 in favour of H_1	V.049
> 0.05	Nonsignificant	Do not reject H ₀	X.051

Testing can be done at any level of significance; 1% is common but 5% is what most researchers use.

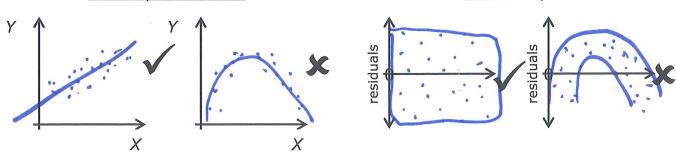
The level of significance can be thought of as a false alarm error rate, i.e. it is the proportion of times that the null hypothesis will be rejected when it is actually true (which can result in action being taken when really no action should be taken).

Thus, a statistically significant result means that a study has produced a "small" P-value (usually < 5%).

- **Assumptions** of simple linear regression are:
 - 1. There is a **linear** relationship between *X* and *Y*.
 - 2. Errors are all independent. Can't be checked
 - 3. Errors are **Normal**ly distributed (with $\mu = 0$).
 - 4. Errors all have the **same std deviation**, σ , regardless of the value of x.
- Assumption checking using plots of the data and residual plots
- 1. There is a **linear** relationship between *X* and *Y*.

Scatterplot of data:

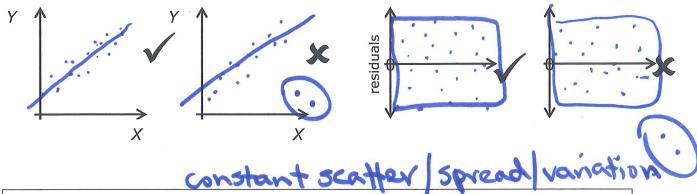
Residual plot:



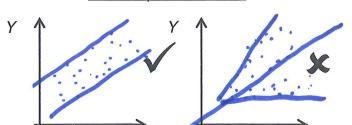
3. Errors are **Normally** distributed (with $\mu = 0$).

Scatterplot of data:

Residual plot:

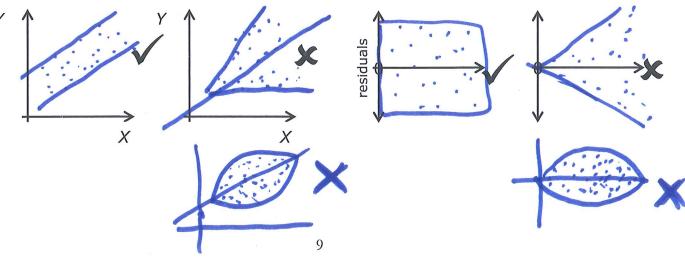


4. Errors all have the **same std deviation**, σ , regardless of the value of x.



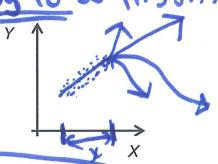
Scatterplot of data:

Residual plot:



0-11.35 am have a break restort on pg

- Estimating / Predicting
 - Within the range of our observed X-values this can be done with confidence. Predicting outside the range of our observed X-values is dangerous. A relationship that fits the data well may not extend outside that range.



Confidence Interval (for the mean)

This estimates the mean Yvalue at a specified value of x. The width of the interval allows for:

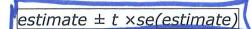
> o uncertainty about the values of β_0 and β_1 .

Prediction Interval

This predicts the Y-value for an individual with a specified value of x.

The width of the interval $\ref{1}$ allows for:

- o uncertainty about the values of β_0 and β_1 and
- uncertainty due to the random scatter about the line.



✓ For a given value of x, the **95%** prediction interval is always wider than the 95% confidence interval for the mean.

\checkmark The Sample Correlation Coefficient, r

 \checkmark r has a value between -1 and +1:













- r = -1
- r = -0.7
- r = -0.4
- r = 0
- r = 0.3
- r = 0.8
- r = -1, then X and Y have a **perfect negative linear relationship**
- r = 0, then X and Y have **no linear** relationship but they may have some other <u>non-linear</u> relationship 💈
- r = 1, then X and Y have a perfect positive linear relationship
- \checkmark r measures the **strength** and **direction** of the **linear** association between two numeric variables
- ✓ r measures how close the points come to lying on a straight line
- \checkmark The value of r is the same if the axes are swapped around it doesn't matter which variable is X and which one is Y
- r has no units \rightarrow a computer / calculator can give you the value of r

Correlation DOES NOT imply causation