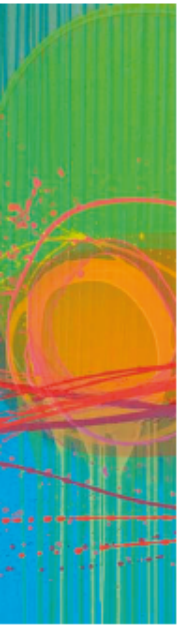
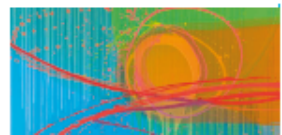


Multiple Regression

Lecture 07



Aims

- Understand When To Use Multiple Regression.
- Understand the multiple regression equation and what the betas represent.
- Understand Different Methods of Regression
 - Hierarchical
 - Stepwise
 - Forced Entry
- Understand How to do a Multiple Regression on PASW/SPSS
- Understand how to Interpret multiple regression.
- Understand the Assumptions of Multiple Regression and how to test them

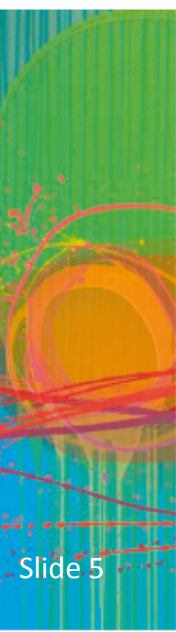
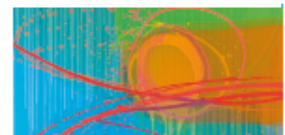
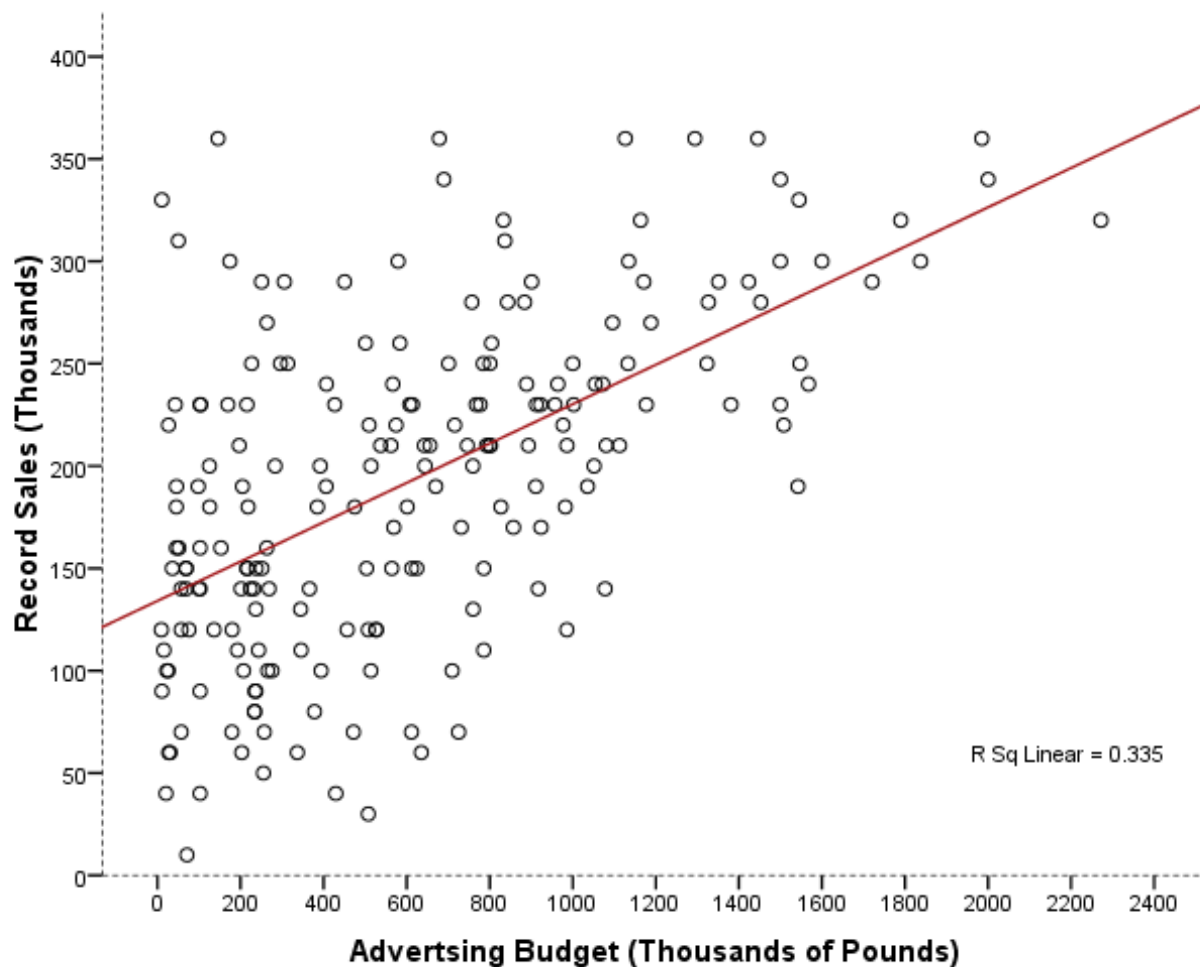
What is Multiple Regression?

- Linear Regression is a model to predict the value of one variable from another.
- Multiple Regression is a natural extension of this model:
 - We use it to predict values of an outcome from *several* predictors.
 - It is a hypothetical model of the relationship between several variables.

Regression: An Example

- A record company boss was interested in predicting record sales from advertising.
- Data
 - 200 different album releases
- Outcome variable:
 - Sales (CDs and Downloads) in the week after release
- Predictor variables
 - The amount (in £s) spent promoting the record before release (see last lecture)
 - Number of plays on the radio (new variable)

The Model with One Predictor



Multiple Regression as an Equation

- With multiple regression the relationship is described using a variation of the equation of a straight line.

$$y = b_0 + b_1X_1 + b_2X_2 + \dots + b_nX_n + \varepsilon_i$$

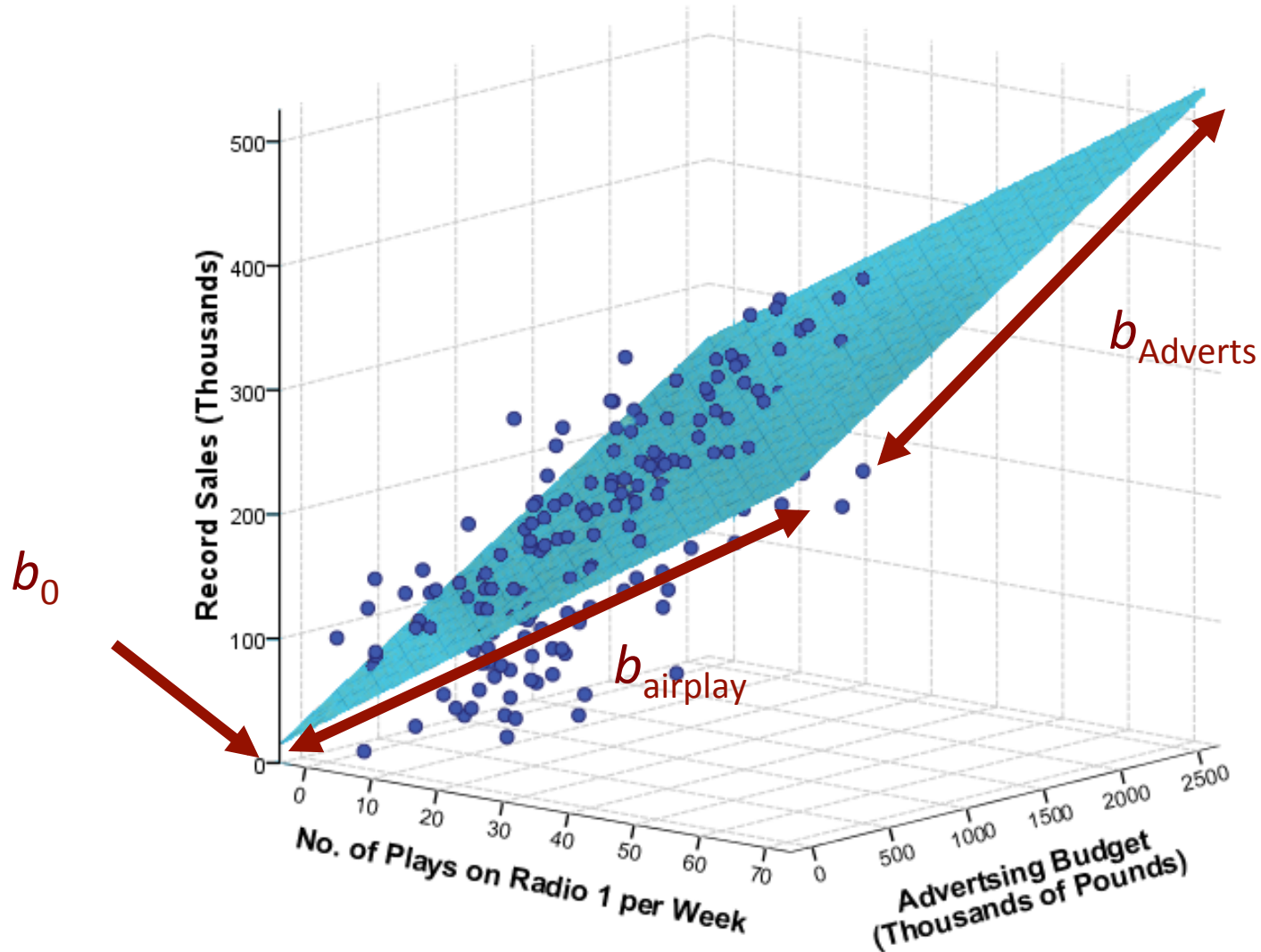
$$b_0$$

- b_0 is the intercept.
- The intercept is the value of the Y variable when all Xs = 0.
- This is the point at which the regression plane crosses the Y-axis (vertical).

Beta Values

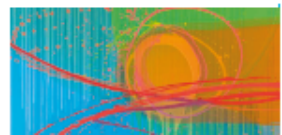
- b_1 is the regression coefficient for variable 1.
- b_2 is the regression coefficient for variable 2.
- b_n is the regression coefficient for n^{th} variable.

The Model with Two Predictors

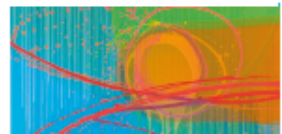


Methods of Regression

- **Hierarchical:**
 - Experimenter decides the order in which variables are entered into the model.
- **Forced Entry:**
 - All predictors are entered simultaneously.
- **Stepwise:**
 - Predictors are selected using their semi-partial correlation with the outcome.

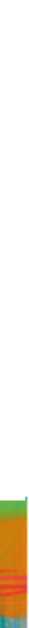
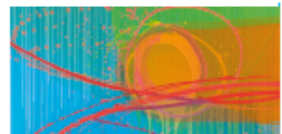


Which method of
regression should I use?



Hierarchical Regression

- Known predictors (based on past research) are entered into the regression model first.
- New predictors are then entered in a separate step/block.
- Experimenter makes the decisions.

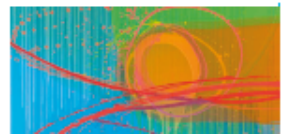


Hierarchical Regression

- **It is the best method:**
 - Based on theory testing.
 - You can see the unique predictive influence of a new variable on the outcome because known predictors are held constant in the model.
- **Bad Point:**
 - Relies on the experimenter knowing what they're doing!

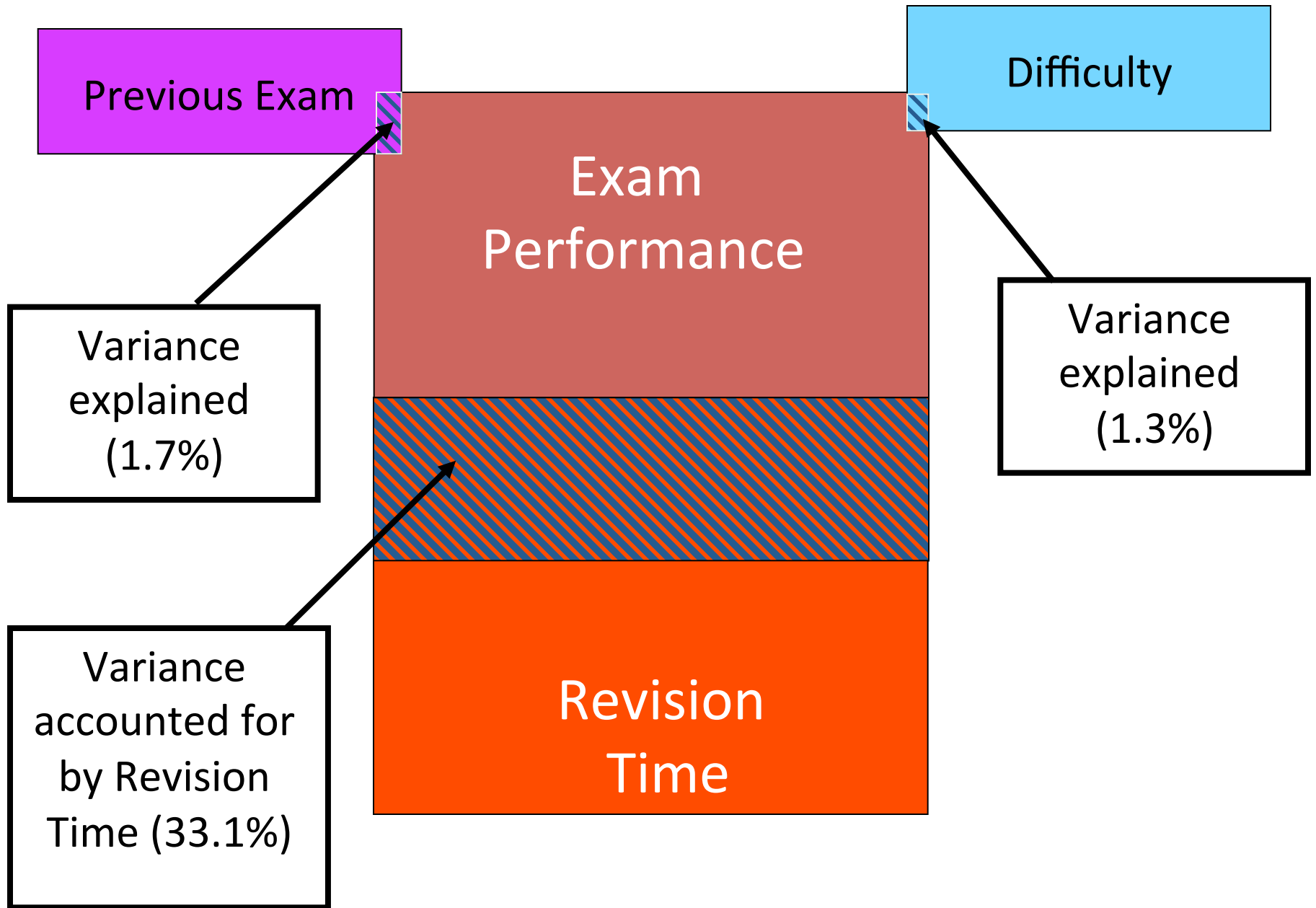
Forced Entry Regression

- All variables are entered into the model simultaneously.
- The results obtained depend on the variables entered into the model.
 - It is important, therefore, to have good theoretical reasons for including a particular variable.



Stepwise Regression I

- Variables are entered into the model based on mathematical criteria.
- Computer selects variables in steps.
- Step 1
 - SPSS looks for the predictor that can explain the most variance in the outcome variable.



Correlations

		Exam Mark (%)	First Year Exam Grade	Difficulty of Question	Revision Time
Exam Mark (%)	Pearson Correlation	1.000	.130**	-.113*	.575**
	Sig. (2-tailed)	.	.005	.014	.000
	N	474	474	474	474
First Year Exam Grade	Pearson Correlation	.130**	1.000	-.012	.047
	Sig. (2-tailed)	.005	.	.959	.303
	N	474	474	474	474
Difficulty of Question	Pearson Correlation	-.113*	-.012	1.000	-.236**
	Sig. (2-tailed)	.014	.959	.	.000
	N	474	474	474	474
Revision Time	Pearson Correlation	.575**	.047	-.236**	1.000
	Sig. (2-tailed)	.000	.303	.000	.
	N	474	474	474	474

** . Correlation is significant at the 0.01 level (2-tailed).

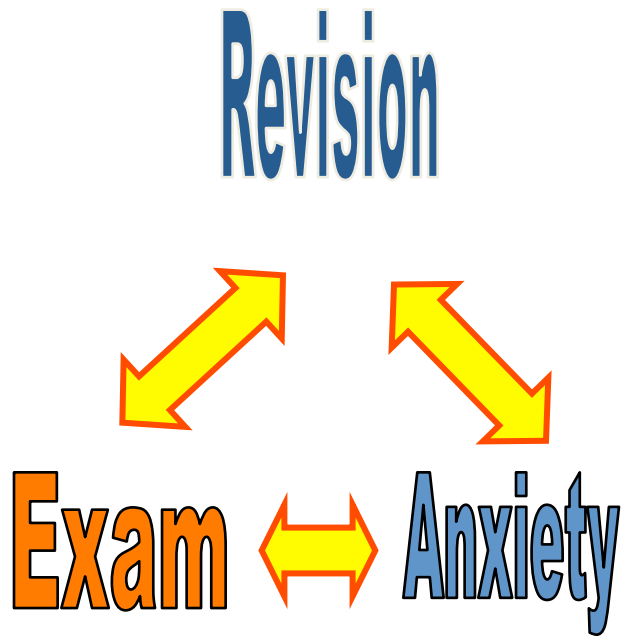
* . Correlation is significant at the 0.05 level (2-tailed).

Stepwise Regression II

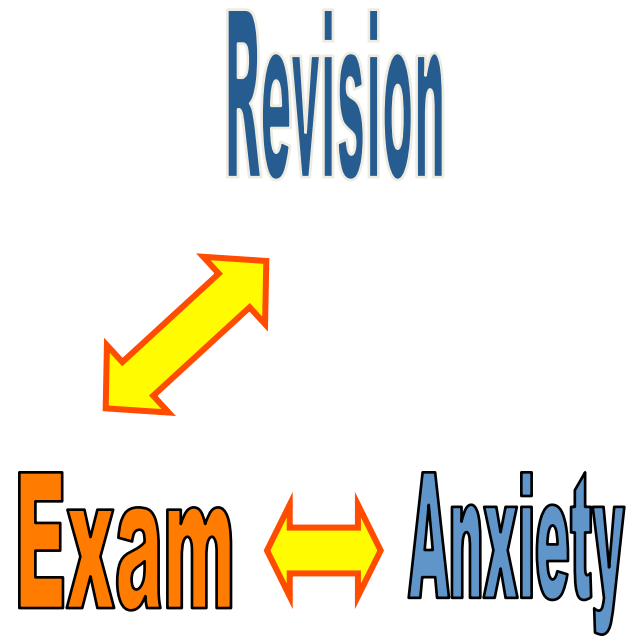
- Step 2:
 - Having selected the 1st predictor, a second one is chosen from the remaining predictors.
 - The *semi-partial correlation* is used as a criterion for selection.

Semi-Partial Correlation

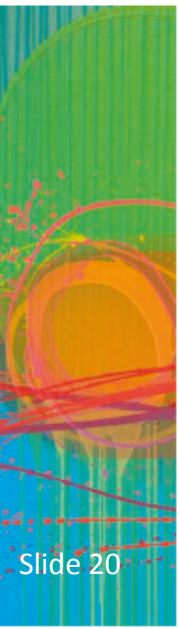
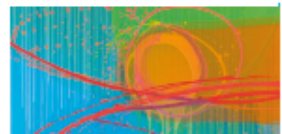
- **Partial correlation:**
 - measures the relationship between two variables, controlling for the effect that a third variable has on them both.
- **A semi-partial correlation:**
 - Measures the relationship between two variables controlling for the effect that a third variable has on only one of the others.



Partial Correlation

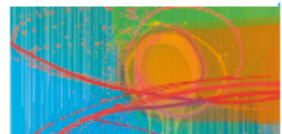
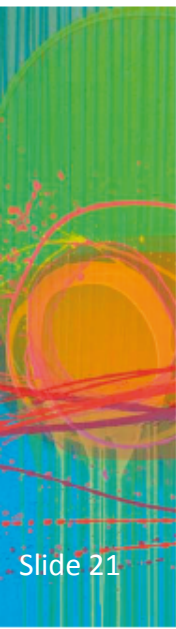


Semi-Partial
Correlation



Semi-Partial Correlation in Regression

- The semi-partial correlation
 - Measures the relationship between a predictor and the outcome, controlling for the relationship between that predictor and any others already in the model.
 - It measures the *unique contribution* of a predictor to explaining the variance of the outcome.



Coefficients^a

Model		Unstandardized Coefficients		Standardized Coefficients	t	Sig.
		B	Std. Error	Beta		
1	(Constant)	-10.863	3.069		-3.539	.000
	Revision Time	3.400	.222	.575	15.282	.000
2	(Constant)	-24.651	5.858		-4.208	.000
	Revision Time	3.371	.221	.570	15.241	.000
	First Year Exam Grade	.175	.063	.103	2.756	.006

a. Dependent Variable: Exam Mark (%)

Excluded Variables^c

Model		Beta In	t	Sig.	Partial	Collinearity
					Correlation	Statistics
		Tolerance				
1	First Year Exam Grade	.103 ^a	2.756	.006	.126	.998
	Difficulty of Question	.024 ^a	.624	.533	.029	.944
2	Difficulty of Question	.024 ^b	.631	.528	.029	.944

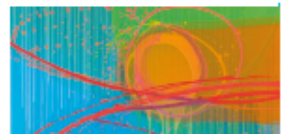
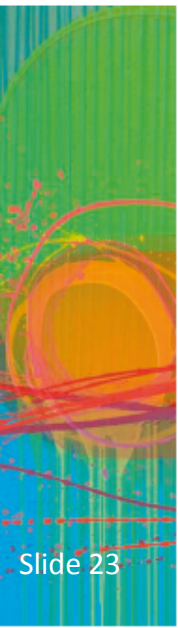
a. Predictors in the Model: (Constant), Revision Time

b. Predictors in the Model: (Constant), Revision Time, First Year Exam Grade

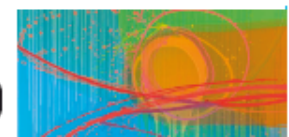
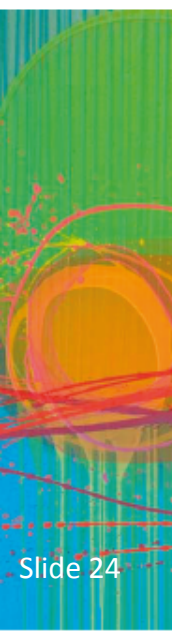
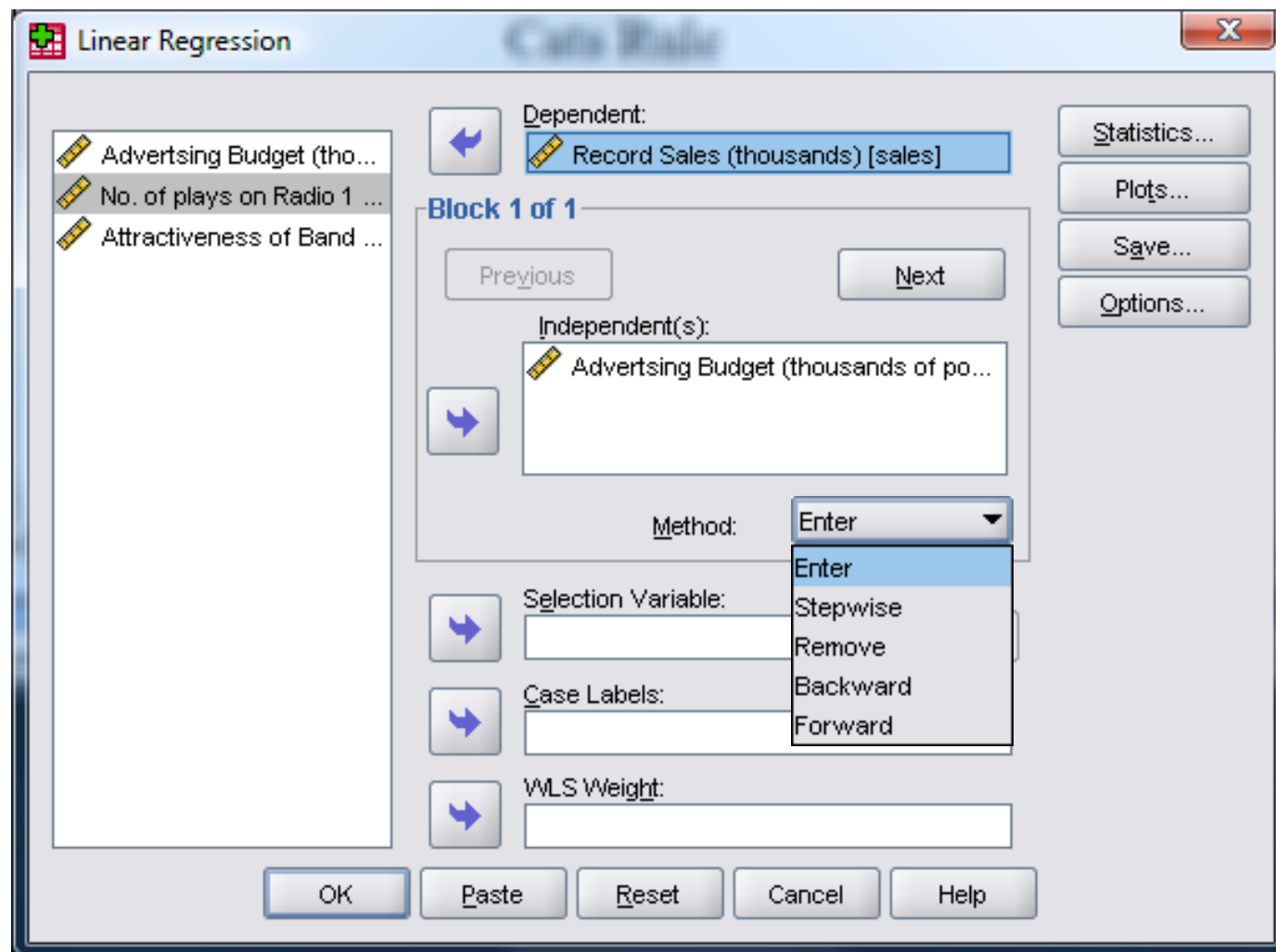
c. Dependent Variable: Exam Mark (%)

Problems with Stepwise Methods

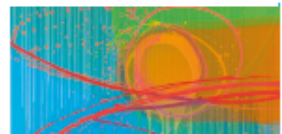
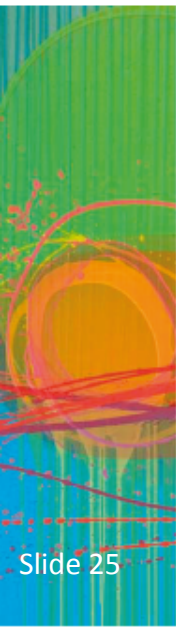
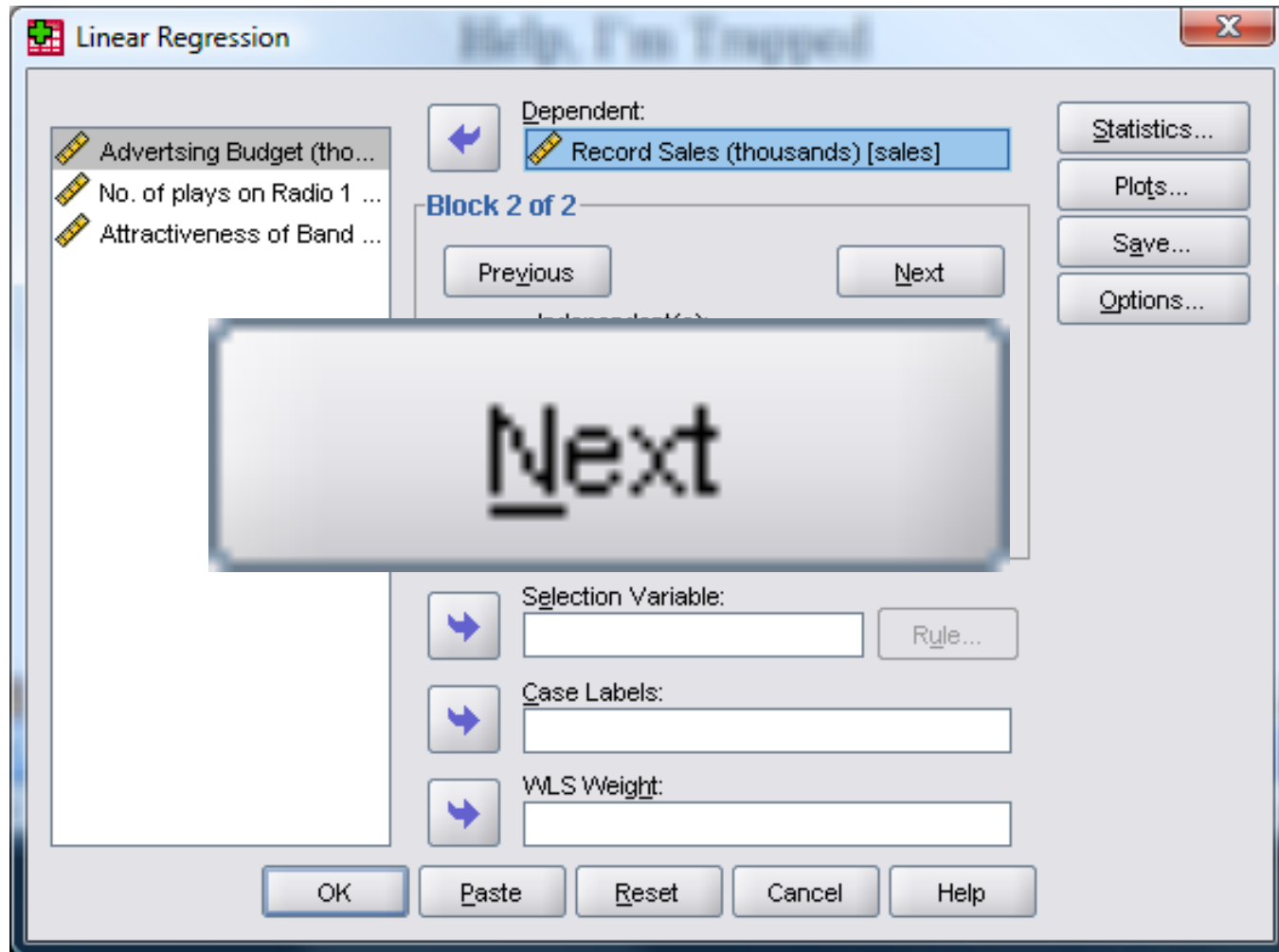
- **Rely on a mathematical criterion.**
 - Variable selection may depend upon only slight differences in the Semi-partial correlation.
 - These slight numerical differences can lead to major theoretical differences.
- **Should be used only for exploration**



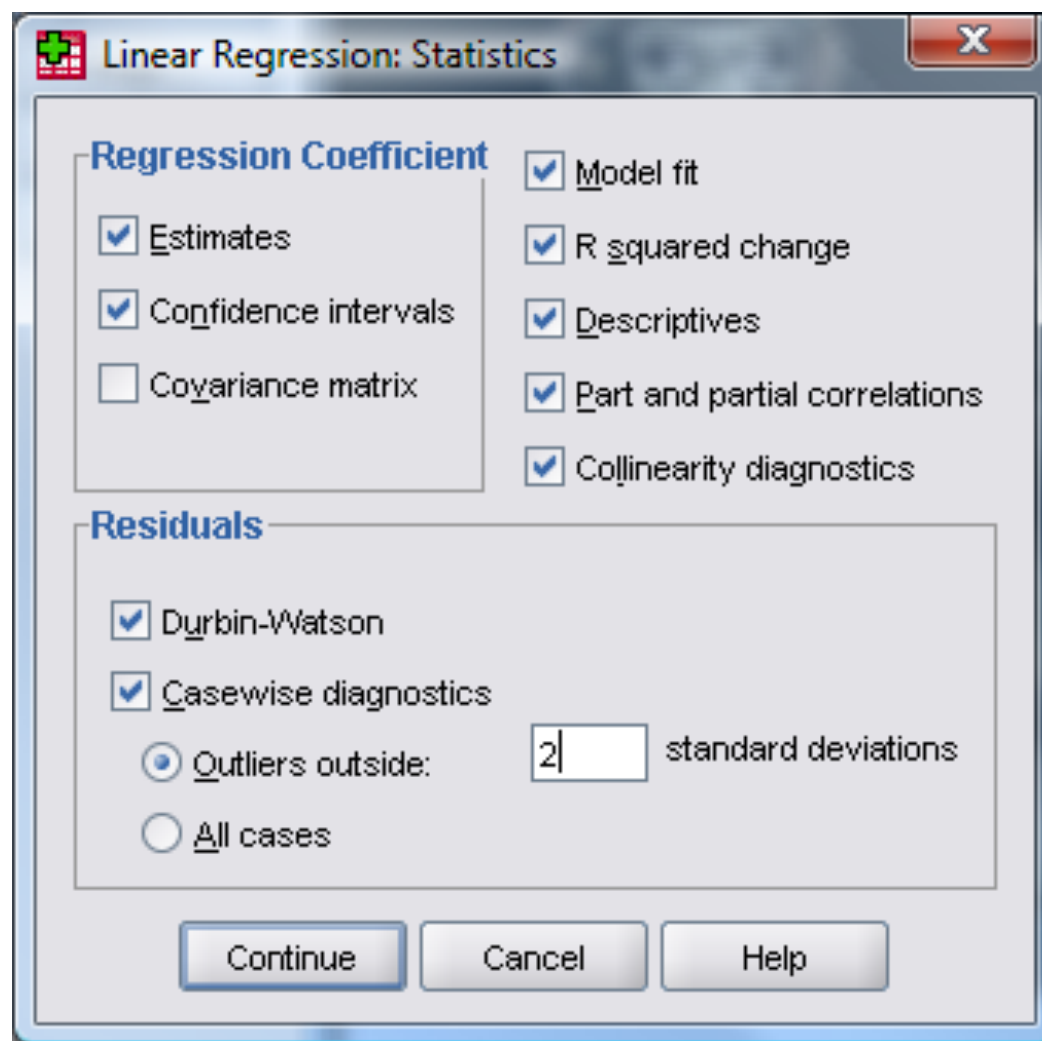
Doing Multiple Regression



Doing Multiple Regression



Regression Statistics



Linear Regression: Save

Predicted Values

- ☒ Unstandardized
- ☒ Standardized
- ☒ Adjusted
- ☐ S.E. of mean predictions

Residuals

- ☐ Unstandardized
- ☒ Standardized
- ☐ Studentized
- ☒ Deleted
- ☒ Studentized deleted

Distances

- ☒ Mahalanobis
- ☒ Cook's
- ☒ Leverage values

Influence Statistics

- ☐ DfBeta(s)
- ☒ Standardized DfBeta(s)
- ☐ DfFit
- ☒ Standardized DfFit
- ☐ Covariance ratio

Prediction Intervals

- ☐ Mean
- ☐ Individual

Confidence Interval: 95 %

Coefficient statistics

- ☐ Create coefficient statistics
- ☒ Create a new dataset
 - Dataset name: RecordDiagnostics
- ☐ Write a new data file
 - File...

Export model information to XML file

Browse...

- ☒ Include the covariance matrix

Continue Cancel Help

Regression Diagnostics

Output: Model Summary

Model Summary^c

Model	R	R Square	Adjusted R Square	Std. Error of the Estimate	Change Statistics					Durbin-Watson
					R Square Change	F Change	df1	df2	Sig. F Change	
1	.578 ^a	.335	.331	65.9914	.335	99.587	1	198	.000	1.950
2	.815 ^b	.665	.660	47.0873	.330	96.447	2	196	.000	

a. Predictors: (Constant), Advertising Budget (thousands of pounds)

b. Predictors: (Constant), Advertising Budget (thousands of pounds), Attractiveness of Band, No. of plays on Radio 1 per week

c. Dependent Variable: Record Sales (thousands)

R and R^2

- R
 - The correlation between the observed values of the outcome, and the values predicted by the model.
- R^2
 - The proportion of variance accounted for by the model.
- Adj. R^2
 - An estimate of R^2 in the population (*shrinkage*).

Output: ANOVA

ANOVA ^c						
Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	433687.833	1	433687.833	99.587	.000 ^a
	Residual	362264.167	198	4354.870		
	Total	1295952.0	199			
2	Regression	861377.418	3	287125.806	129.498	.000 ^b
	Residual	434574.582	196	2217.217		
	Total	1295952.0	199			

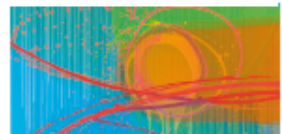
a. Predictors: (Constant), Advertising Budget (thousands of pounds)

b. Predictors: (Constant), Advertising Budget (thousands of pounds), Attractiveness of Band, No. of Plays on Radio 1 per Week

c. Dependent Variable: Record Sales (thousands)

Analysis of Variance: ANOVA

- **The F-test**
 - looks at whether the variance explained by the model (SS_M) is significantly greater than the error within the model (SS_R).
 - It tells us whether using the regression model is significantly better at predicting values of the outcome than using the mean.



Output: betas

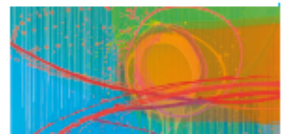
Coefficients^a

Model	Unstandardized Coefficients		Standardized Coefficients	t	Sig.	95% Confidence Interval for B	
	B	Std. Error	Beta			Lower Bound	Upper Bound
1 (Constant)	134.140	7.537		17.799	.000	119.278	149.002
Advertising Budget (thousands of pounds)	.096	.010	.578	9.979	.000	.077	.115
2 (Constant)	-26.613	17.350		-1.534	.127	-60.830	7.604
Advertising Budget (thousands of pounds)	.085	.007	.511	12.261	.000	.071	.099
No. of plays on Radio 1 per week	3.367	.278	.512	12.123	.000	2.820	3.915
Attractiveness of Band	11.086	2.438	.192	4.548	.000	6.279	15.894

a. Dependent Variable: Record Sales (thousands)

How to Interpret Beta Values

- **Beta values:**
 - the change in the outcome associated with a unit change in the predictor.
- **Standardised beta values:**
 - tell us the same but expressed as standard deviations.



Beta Values

- $b_1 = 0.087$.
 - So, as advertising increases by £1, record sales increase by 0.087 units.
- $b_2 = 3589$.
 - So, each time (per week) a song is played on radio 1 its sales increase by 3589 units.

Constructing a Model

$$y = b_0 + b_1X_1 + b_2X_2$$

$$\text{Sales} = 41\,124 + 0.087\text{Adverts} + 3589\text{plays}$$

£1 Million Advertising, 15 plays

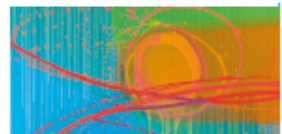
$$\begin{aligned}\text{Sales} &= 41\,124 + (0.087 \times 1,000,000) + (3589 \times 15) \\ &= 41\,124 + 87\,000 + 53\,835 \\ &= 181\,959\end{aligned}$$

Standardised Beta Values

- $\beta_1 = 0.523$
 - As advertising increases by 1 standard deviation, record sales increase by 0.523 of a standard deviation.
- $\beta_2 = 0.546$
 - When the number of plays on radio increases by 1 s.d. its sales increase by 0.546 standard deviations.

Interpreting Standardised Betas

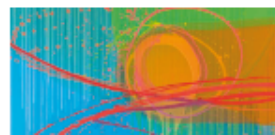
- As advertising increases by £485,655, record sales increase by $0.523 \times 80,699 = 42,206$.
- If the number of plays on radio 1 per week increases by 12, record sales increase by $0.546 \times 80,699 = 44,062$.



Reporting the Model

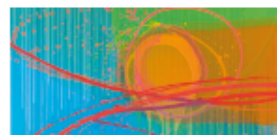
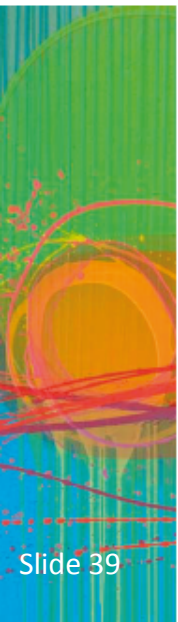
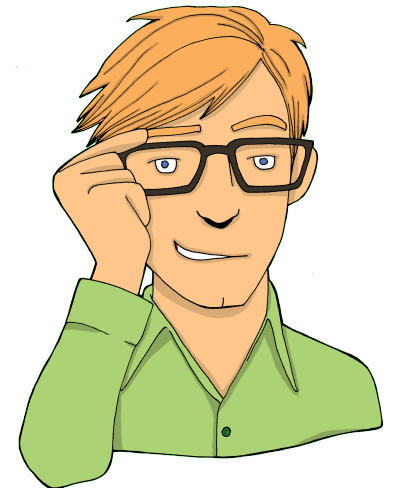
TABLE 7.2 How to report multiple regression

	<i>B</i>	<i>SE B</i>	β
Step 1			
Constant	134.14	7.54	
Advertising Budget	0.10	0.01	.58*
Step 2			
Constant	-26.61	17.35	
Advertising Budget	0.09	0.01	.51*
Plays on BBC Radio 1	3.37	0.28	.51*
Attractiveness	11.09	2.44	.19*



How well does the Model fit the data?

- There are two ways to assess the accuracy of the model in the sample:
- Residual Statistics
 - Standardized Residuals
- Influential cases
 - Cook's distance

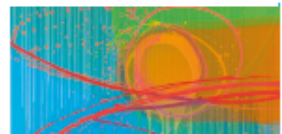
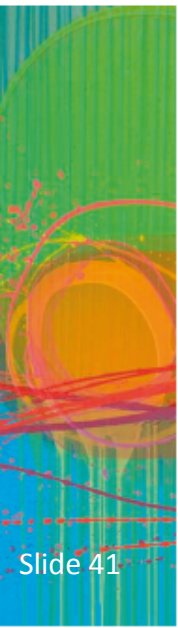


Standardized Residuals

- In an average sample, 95% of standardized residuals should lie between ± 2 .
- 99% of standardized residuals should lie between ± 2.5 .
- Outliers
 - Any case for which the absolute value of the standardized residual is 3 or more, is likely to be an outlier.

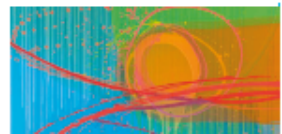
Cook's Distance

- Measures the influence of a single case on the model as a whole.
- Weisberg (1982):
 - Absolute values greater than 1 may be cause for concern.



Generalization

- When we run regression, we hope to be able to generalize the sample model to the entire population.
- To do this, several assumptions must be met.
- Violating these assumptions stops us generalizing conclusions to our target population.



Straightforward Assumptions

- **Variable Type:**
 - Outcome must be continuous
 - Predictors can be continuous or dichotomous.
- **Non-Zero Variance:**
 - Predictors must not have zero variance.
- **Linearity:**
 - The relationship we model is, in reality, linear.
- **Independence:**
 - All values of the outcome should come from a different person.

The More Tricky Assumptions

- **No Multicollinearity:**
 - Predictors must not be highly correlated.
- **Homoscedasticity:**
 - For each value of the predictors the variance of the error term should be constant.
- **Independent Errors:**
 - For any pair of observations, the error terms should be uncorrelated.
- **Normally-distributed Errors**

Multicollinearity

- Multicollinearity exists if predictors are highly correlated.
- This assumption can be checked with collinearity diagnostics.

Coefficients ^a						
Model		Correlations			Collinearity Statistics	
		Zero-order	Partial	Part	Tolerance	VIF
1	Advertising Budget (thousands of pounds)	.578	.578	.578	1.000	1.000
2	Advertising Budget (thousands of pounds)	.578	.659	.507	.986	1.015
	No. of plays on Radio 1 per week	.599	.655	.501	.959	1.043
	Attractiveness of Band	.326	.309	.188	.963	1.038

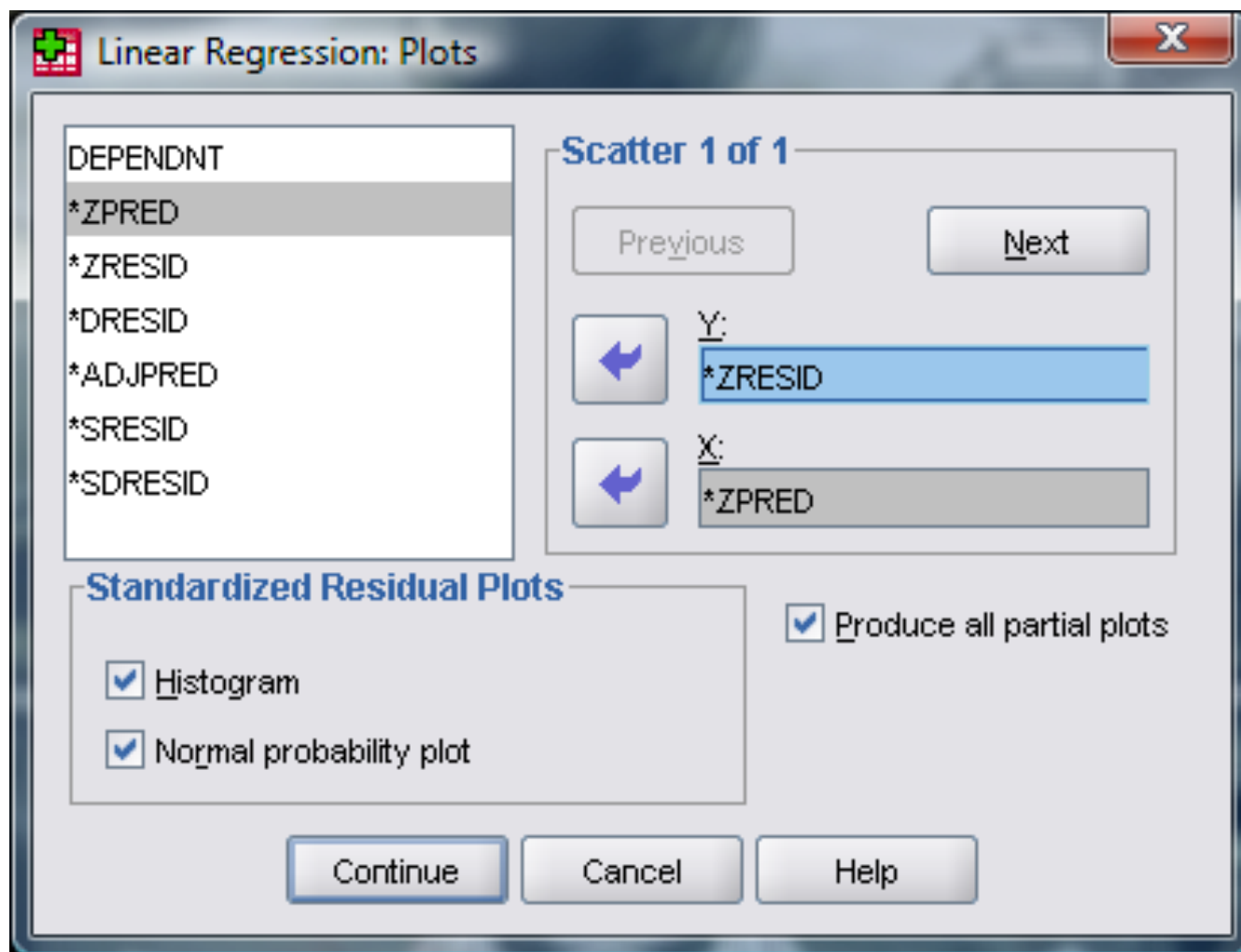
a. Dependent Variable: Record Sales (thousands)

- Tolerance should be more than 0.2 (Menard, 1995)
- VIF should be less than 10 (Myers, 1990)

Checking Assumptions about Errors

- **Homoscedacity/Independence of Errors:**
 - Plot ZRESID against ZPRED.
- **Normality of Errors:**
 - Normal probability plot.

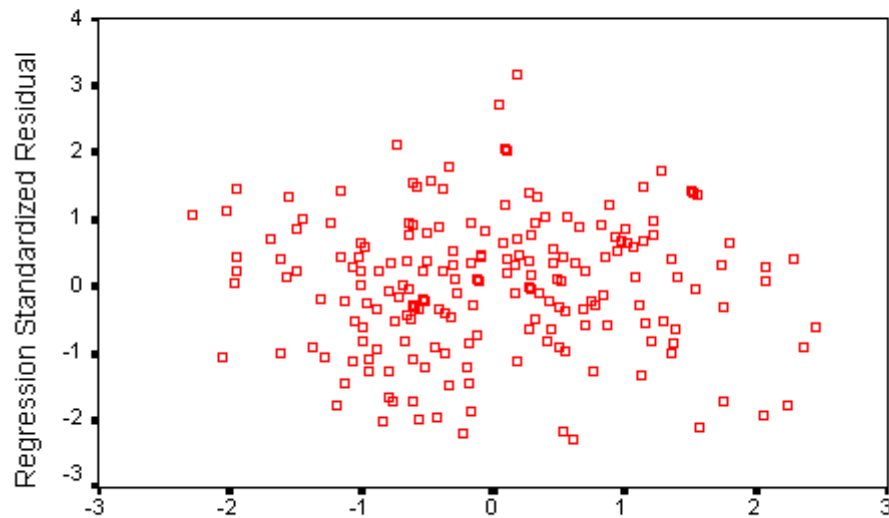
Regression Plots



Homoscedasticity: ZRESID vs. ZPRED

Scatterplot

Dependent Variable: Record Sales

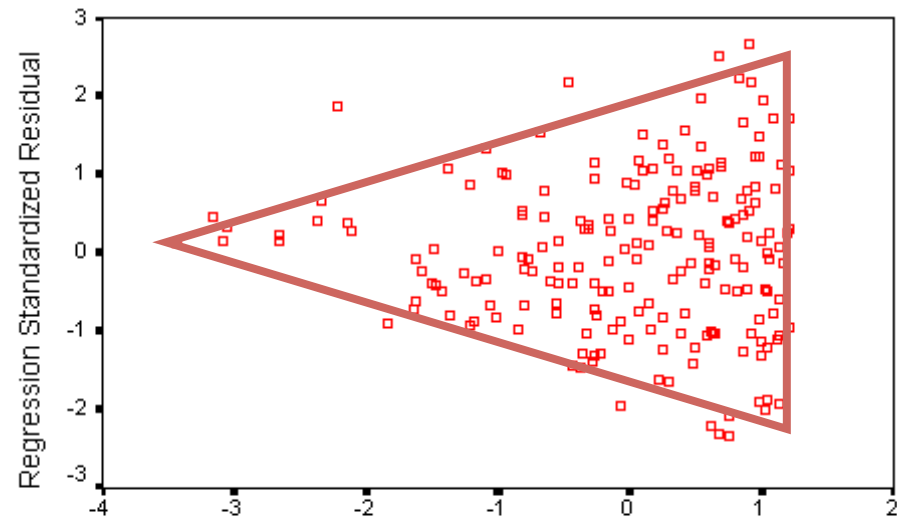


Regression Standardized Predicted Value

Good

Scatterplot

Dependent Variable: OUTCOME



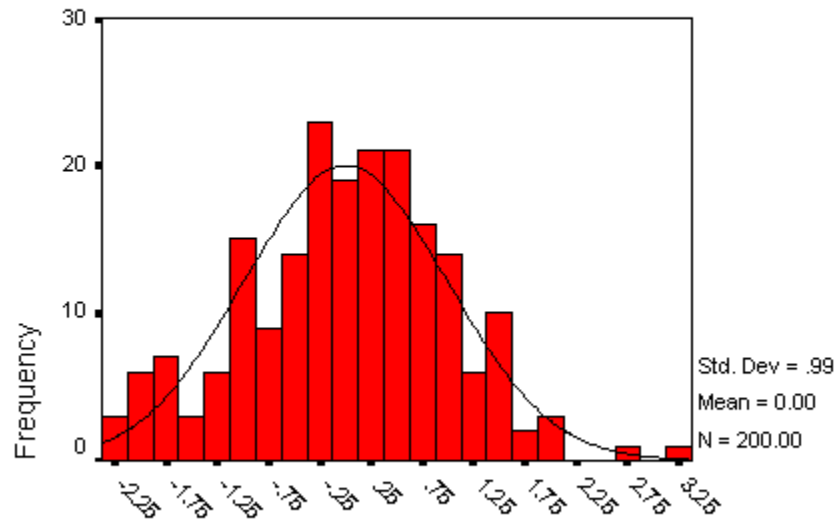
Regression Standardized Predicted Value

Bad

Normality of Errors: Histograms

Histogram

Dependent Variable: Record Sales

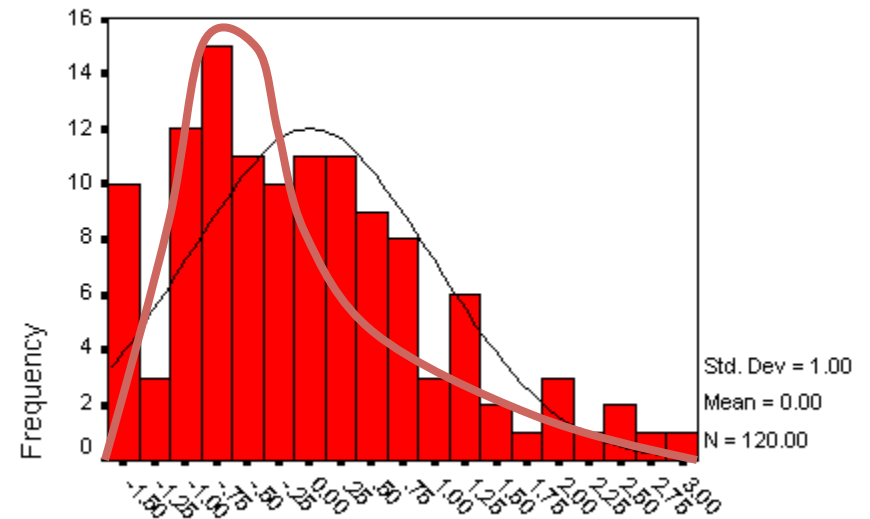


Regression Standardized Residual

Good

Histogram

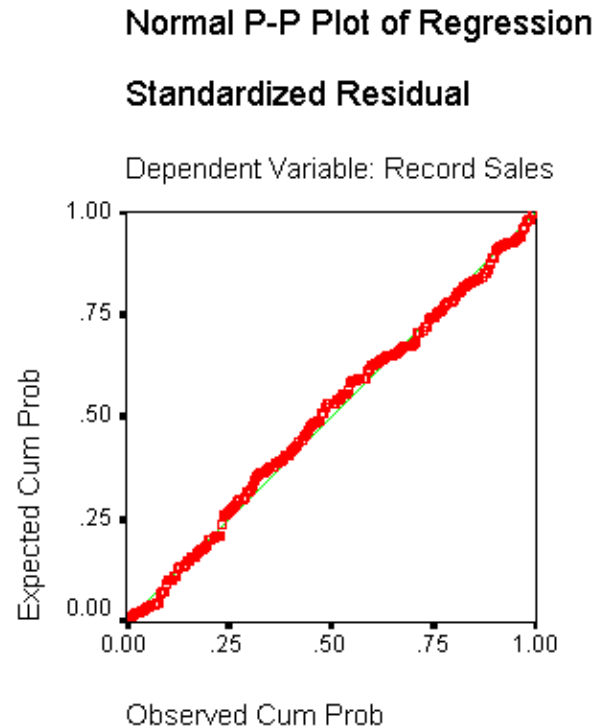
Dependent Variable: OUTCOME



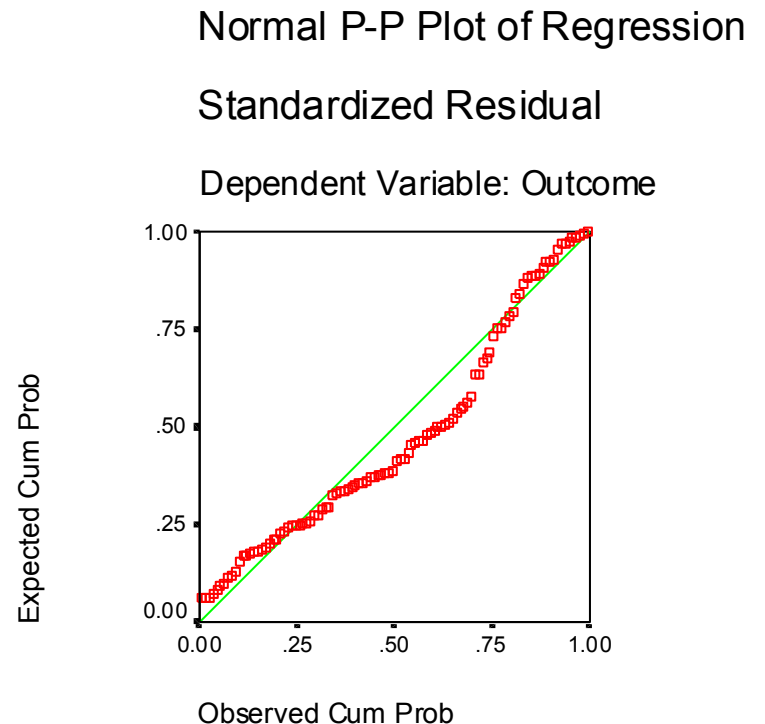
Regression Standardized Residual

Bad

Normality of Errors: Normal Probability Plot



Good



Bad