Quantitative Methods in Political Science Recitation

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Correlation Analysis

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- Scatterplots

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 - Recall that missing data is represented by "." in Stata

Review of regression

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- More work with scatterplots

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 - What does this do again?

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- b is the slope

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- How do we interpret the results? Let's take a look at the output.

. regress gdppc agehinst

Source	SS	df	MS		Number of obs F(1, 152)	
Model Residual	2.7428e+09 2.4133e+09	1 152	2.7428e+09 15876745.7		Prob > F R-squared Adj R-squared	= 0.0000 = 0.5320
Total	5.1561e+09	153	33699753.5		Root MSE	= 3984.6
gdppc	Coef.	Std. I	Err. t	P> t	[95% Conf.	Interval]
agehinst _cons	162.9329 1543.919	12.390 433.1		0.000 0.000	138.4415 688.2255	187.4242 2399.612

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- The slope is given by the coefficient for *agehinst*: 162.9329 or we'll say 163
- The intercept is given by the constant: 1543.919 or we'll say 1544
- We can plug this into our general regression line formula

 $\hat{y} = 1544 + 163x$

Income PerCapita = 1544 + 163 * AgeCurrentRegime

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We can interpret using the templates from class:

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 - When [x] is 0, [y] is expected to equal the [y-intercept].
- R-squared (proportion of variability of dependent variable that is accounted for by the independent variable (or the model):
 - So [XXX] % of the variability in y is explained by x (or by the model).

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- Interpret the R-squared

. regress gdppc investment

Source	SS	df	MS			Number of obs F(1, 129)	
Model Residual Total	1.7454e+09 2.9898e+09 4.7352e+09	1 129 130	2317	54e+09 6820.5 4558.6		Prob > F R-squared Adj R-squared Root MSE	= 0.0000 = 0.3686
gdppc	Coef.	Std.	Err.	t	P> t	[95% Conf.	Interval]
investment _cons	424.7486 -613.219	48.94 841.		8.68 -0.73	0.000 0.467	327.9086 -2278.145	521.5887 1051.707

More on scatterplots

Recall last week we made scatterplots using the *scatter* command.

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- As always you can edit and save graphs in the "Graph Editor" window

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