Correlation and LSR

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We can calculate (usually using a computer) a value r which we call *correlation* on a set of data points.

Correlation facts:

- Correlations can range from -1 to 1, inclusive
- For correlations of 1 or -1, all the points lie on a line
- A correlation of 0 means no linear association of the variables
- We usually compare data against a line unless we say otherwise
- Low correlations don't necessarily mean no association

Warning:

If two variables are correlated, this does not imply that one causes the other. They may be influenced by a third *lurking* variable, or the correlation may also be *spurious*. (There are enough data sets floating around that some will appear to have an uncanny similarity for no real reason.)

Spurious correlations



Lurking variables can cause two data sets to really be connected, and we may think they are spurious. For example, white cars get more speeding tickets than other colors (according to cnet.com), followed by red cars. However, the people who choose those colors may be more likely to drive fast. Or, perhaps police really are more likely to stop white cars. Who knows!? But something like this is more likely to be caused by a lurking variable of some sort.

Examples of zero correlation



Examples of zero correlation





This technically has a correlation of r = -.28 but what does this mean? The correct model would fit perfectly. If you extended this spiral for several more turns, the linear correlation would drop even more. Here, low correlation means you have the wrong model. There is a relationship between these variables! But a line does not capture the relationship!

- For small data sets: BE CAREFUL
- The more significant the correlation, the farther from zero r is
- Positive correlation gives best-fit lines with positive slope
- Negative correlation gives best-fit lines with negative slope









Best-fit lines or LSR (Least Squares Regression) means that we find a line that decreases the squares of the residuals. The arithmetic for calculating residuals and squaring them is very much the same process we used to calculate variance and standard deviations.

Residuals



You can often find a reasonable best-fit line by eyeballing it, and to be honest, your eyeball-fit line may be every bit as good at estimating the relationship as your super-duper computer-generated one. However, this is such a common task, we need to be able to pass it off to a computer, and to get repeatable results.



Where do you think the line should go?

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Estimations based on models come in two broad categories: Interpolation: Estimates within the domain of the given data Extrapolation: Estimates outside the domain of the given data If you have good solid data, interpolations are often relatively safe. However, the farther away from the data you get, often the less accurate your predictions will become. Think of the weather predictions. There are, of course, amazing exceptions. We know with great precision when Halley's comet will pass by the earth, based on computations taken a long time past! It passes by only about once every 75 years! But you should understand that data from the past typically do not accurately predict the future!

Influential outliers



By county in Florida: Total votes vs. Buchanan votes

In the presidential election of 2000, a confusing ballot caused a third party candidate to receive many more votes than he should have. This graphic demonstrates evidence for this claim. Those votes lost due to this error were part of what caused one party to lose the presidency that year.

If you drew an LSR line on that scatterplot with and without the point from Palm Beach County, you would get a different line!

This is a real-world example of an influential outlier!

If you sketch a line by hand, you will want to estimate the slope and intercept.

To estimate the intercept, draw the line until it hits the y-axis.

To estimate the slope, pick two convenient points on your sketch:

$$slope = \frac{y_2 - y_1}{x_2 - x_1}$$

 $y = intercept + slope \cdot x$

For whatever reason, statisticians insist on writing lines this way:

$responseVariable = intercept + slope \cdot explanatoryVariable$

This is simplified as y = a + bx and often confuses students, because suddenly the letter *b* is being used as the slope rather than the intercept, which they are used to. So avoid using the letter *b* altogether and just "B" careful! If you have a line $y = P + Q \cdot x$ and you have an x value you wish to find a y for, just push the x through the formula as usual.

Example: y = 1 + 3x could be your LSR line and you wish to predict what y will go with x = 4.

 $y = 1 + 3 \cdot 4 = 13$

If you have a line $y = P + Q \cdot x$ and you have an y value you wish to find an x for, you need to solve for that x value.

Example: y = 1 + 3x could be your LSR line and you wish to predict what x will go with y = 7.

7 = 1 + 3x6 = 3x2 = x

This fits together with correlation very nicely. When you create an LSR line on StatCrunch, it gives you all the information together: A LSR line, correlation (which it calls R instead of r), and R^2 , as well as p-values.

We will look at the p-values for the Intercept and Slope to see how significant these are.

LSR Significant Intercept



Your Intercept may be very significant.

The p-value is very low for the Intercept but not the slope.

LSR Significant Slope



Your Slope may be very significant.

The p-value is very low for the Slope but not the intercept. Note: This is a common issue when not recentering data!

LSR Significant Slope and Intercept



Your Slope and Intercept may be very significant. The p-value is very low for the Slope and the Intercept.



Perhaps neither the Intercept nor Slope is significant.

R^2

The square of the correlation, or R^2 , tells us how much of the variation in the data can be attributed to the model.

In the above slides, where we had high correlation, the points were mostly along their best-fit line, and R^2 was around 0.999 or very close to 1 or 100%. This means that most of the variation was accounted for by the line model. For the low-correlation slides, R^2 was near zero, meaning that none of the variation was captured by the model.

If the correlation for a LSR model is r = .5, then only .25 or 25% of the variation is explained by the model.

However, if the correlation is r = .9 then 81% of the variation is explained by the model!

We often plot residuals against the line y = 0 to emphasize the variation which remains after the model (in this case just a line) has been removed.

Original LSR analysis



If we had used the correct model...



We do not do polynomial interpolation in STAT202. But, it's just a click away! In real life, don't hesitate to use it if you want to!

LSR with line model and Residuals Plot



- Will not reveal anything the scatterplot and model won't reveal
- Does often make it easier to see
- Average value must be zero

If the residuals cross over the line y = 0 a few times in quite distinct places, you probably need a new model.

When your residual plot hugs the line y = 0 more closely for part of the data, you can see where your model does a better job.



If you switch your two variables (x and y) for all your data and calculate an LSR line, this is not the same as just flipping the line. The distances (residuals) are computed on the independent variable only, so if you switch the roles of your variables, you'll change the model entierly.

However, if your data do follow a best fit line fairly well, and you flip everything, you will probably end up with two very good LSR fits! They should be in roughly the same place.

Remember that p-values greater than 0.05 are generally "boring".

MEMORY QUESTIONS 7 today!



























