APPM2720 Assessing accuracy of the least squares model

It is usually straightforward to judge the strength of a linear relationship between two variables by examining a scatterplot. To set the stage for more complicated data situations there is a useful statistic to measure this related to how large the residuals are compared to the original observations.

the standard deviation

By way of background the definition of the standard deiviation (the R function ${\bf sd}$) is $sd=\sqrt{\frac{1}{N-1}\sum_i(X_i-\bar{X})^2}$]

N the total number of observations, and \bar{X} the average.

This is a useful statistic to describe the spread in data especically when it is close to a Gaussian (aka Normal) distribution.

The "minus one" in the denominator is an adjustment because we are subtracting the average from all the data values.

Residuals

Given N pairs of numbers $(x_1, y_1), (x_2, y_2), \dots, x_N, y_N)$. A useful model is to predict the Ys by a straight line in X. Recall the residuals are the difference between the predicted values and the observations.

$$e_i = Y_i - (\beta_0 + \beta_1 X_i)$$

Residual standard error

Residual Sums of Squares = $RSS = \sum_{i} e_i^2$

Residual standard error =
$$RSE = \sqrt{\frac{1}{N-2}RSS}$$

The "minus 2" comes from subtracting the line (slope and intercept).

- The residual standard error summarizes the spread in the residuals and is useful if they are approximately normal (e.g. follows rules for fraction being within certain number of standard deviations of zero.)
- RSE is needed to find confidence intervals for the estimated slope and intercept
- The Residual standard error is also reported by the summary function applied to the Im output. The degrees of freedom reported after this statistic is the number of observations minus the number of parameters being fit by least squares. (2 in the case of a single variable with intercept.)

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Accuracy of the linear fit

Total sum of squares =
$$TSS = \sum_{i} Y_{i} - \bar{Y}^{2}$$

This is the RSS if we just predicted Y with a constant value!

(R squared) =
$$R^2 = 1 - RSS/TSS$$

R squared close to 1 means the line predicts teh data well. close 0 to zero means the line does not explain much of the variability in Y.

Relationship to R squared

I like to compare the standard deviation of the Y values to the standard deviation of the residuals

```
fit<- lm( Y~X)
sd( fit$residuals ) / sd( Y)
```

Want this to be small -- if zero means a perfect fit.

Or in the form of R^2

```
1- (sd( fit$residuals ) / sd( Y))^2
```

Not exactly equal to R squared but easier to think about.