

#### Lecture 32, November 9

#### **Inference for Regression**

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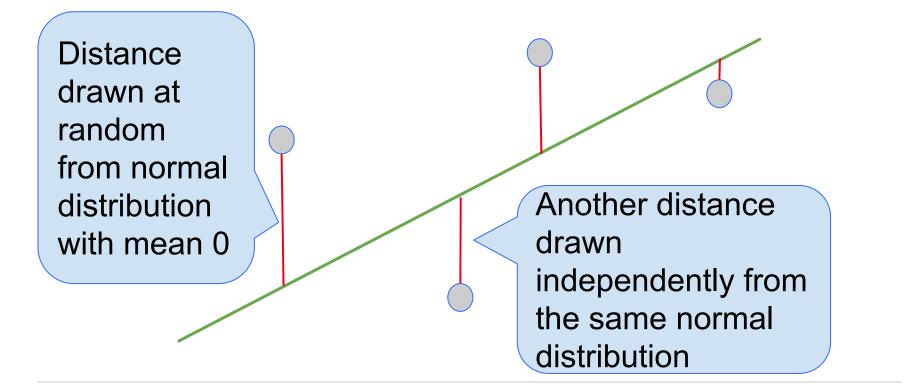
#### Announcements

• Congratulations on submitting the project!

• Homework due Wed/Thurs as usual.

• Friday is an Academic and Administrative Holiday. No lecture and no office hours.

## A "Model": Signal + Noise



#### What We Get to See



# **Confidence Interval for True Slope**

- Bootstrap the scatter plot.
- Find the slope of the regression line through the bootstrapped plot.
- Repeat.
- Draw the empirical histogram of all the generated slopes.
- Get the "middle 95%" interval.
- That's an approximate 95% confidence interval for the slope of the true line.



## **Rain on the Regression Parade**

We observed a slope based on our sample of points.

But what if the sample scatter plot got its slope just by chance?

What if the true line is actually FLAT?

(Demo)

# Slope

- Null hypothesis: The slope of the true line is 0.
- Alternative hypothesis: No, it's not.
- Method:
  - Construct a bootstrap confidence interval for the true slope.
  - If the interval doesn't contain 0, reject the null hypothesis.
  - If the interval does contain 0, there isn't enough evidence to reject the null hypothesis.

# **Regression Prediction**

#### If the model is good,

- Regression line is close to true line
- Given a new value of x, predict y by finding the point on the regression line at that *x*
- Bootstrap the scatter plot
- Get a new prediction using the regression line that goes through the resampled plot (Demo)
- Repeat the two steps above many times
- Get an interval of predictions of y for the given x

### **Predictions at Different Values of** *x*

- Since *y* is correlated with *x*, the predicted values of *y* depend on the value of *x*.
- The width of the prediction interval also depends on *x*.
  Typically, intervals are wider for values of *x* that are further away from the mean of *x*.