

# Big Data and Security

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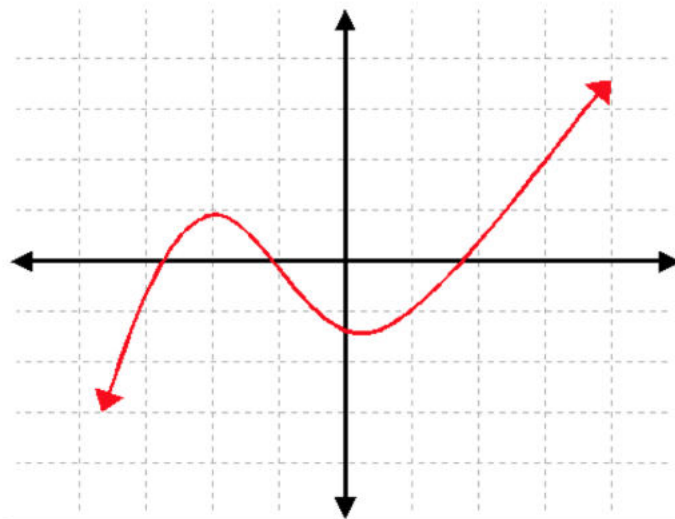
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Linear Regression

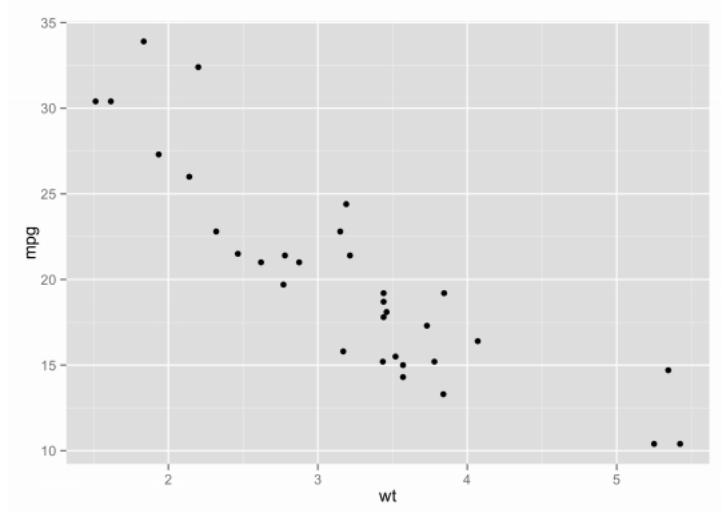
# Regressions

- How do we determine the relationship between variables?
  - We will think about it like a function:  $X$  goes in, and  $Y$  comes out



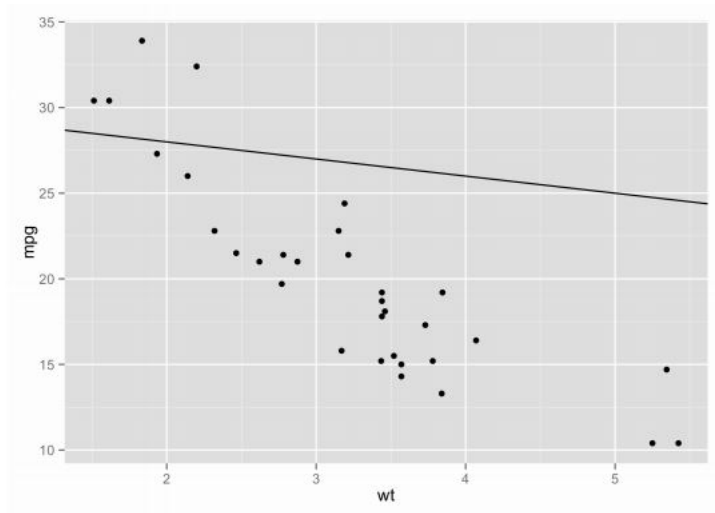
# Regressions

- Instead of a nice functional relationship, we have messy clouds of data



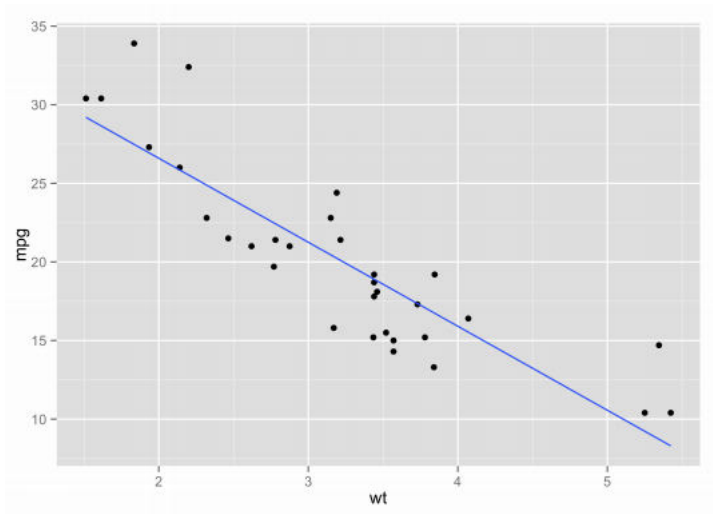
# Regressions

- The goal is to pick a function that fits the data “well”



# Regressions

- What does it mean for this line to fit “better” than the last?



# Regressions

- The main model we use is called “linear regression”
  - We will take a straight line (hence linear)
  - We want it to be near the data.
  - The line which was “bad” was further away than seemed necessary
- We use an equation like:

$$Y = \alpha + \beta X + \varepsilon$$

# The Regression Model

- This is not a math course, but let's look at what goes into this model
  - $X$ ,  $Y$  – random variables. We have draws of this data out of some population.
  - $\alpha$ ,  $\beta$  – parameters. These parameters represent the relationship between  $X$  and  $Y$
  - $\varepsilon$  – This is the error term: the other things which are not in  $X$  but affect  $Y$
- More intuitively
  - If  $Y$  is education and  $X$  is schooling
  - $\beta$  is the amount that an extra unit of schooling is associated with higher wages
  - $\varepsilon$  represents other factors which are not schooling that also affect wages

# Predictions

- How does this relate to prediction?
- We are trying to predict  $y_i$  with  $x_i$
- How do we do this?
  - We know the formula for  $y_i$  from our model

$$y_i = \alpha + \beta x_i + \varepsilon_i$$

- But if we don't know  $y_i$ , we don't know  $\varepsilon_i$  either.
- The best we can do is:

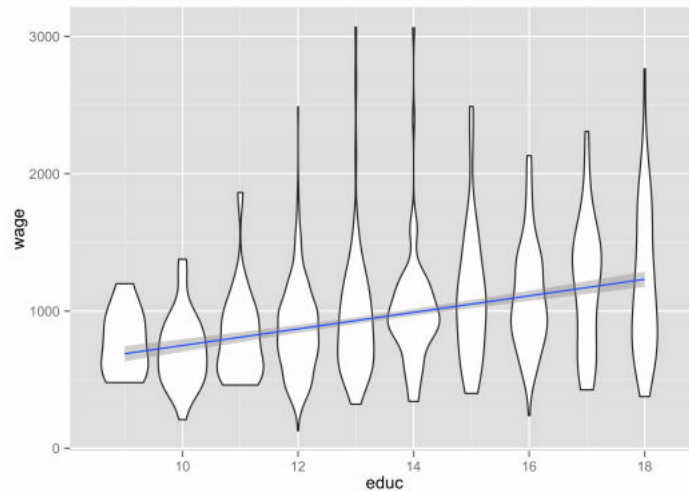
$$\hat{y}^i = \alpha + \beta x_i$$

- $\hat{y}^i$  is called the predicted value of  $y$



# Wages and Education

- X is years of schooling, Y is wages
- Prediction



# What Things Can We Model With A Regression?

- What is  $Y$  ?
- What is  $X$ ?

# Examples

- Crime and policing at the state level
  - What is the sampling frame?
  - What is  $X$ ,  $Y$  ?
- Crime as a function of income from survey data
  - What is the sampling frame?
  - What is  $X$ ,  $Y$  ?

# Where Next?

- First, we'll talk about assumptions
- Linear regression was a very simple model
- But we can discuss a wide variety of extension now that we know about this model

# Lesson Summary

- We try to predict an outcome random variable  $Y$  with an input random variable  $X$
- To do this, we use a sample of pairs of  $(x_i, y_i)$  observation
- We choose the “best” parameters to fit the model
- We get a relationship between  $X$  and  $Y$  that can be used for prediction or analysis