

DSFA Spring 2020

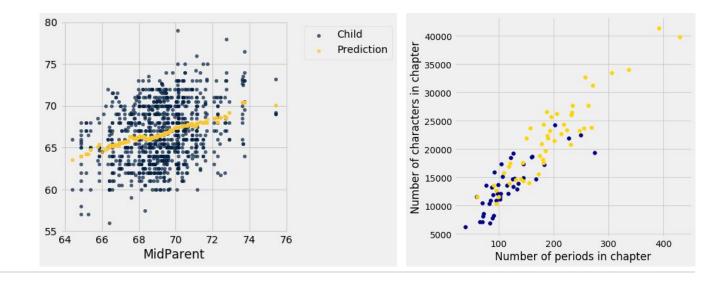
#### Lecture 24

Residuals

## **Prediction**

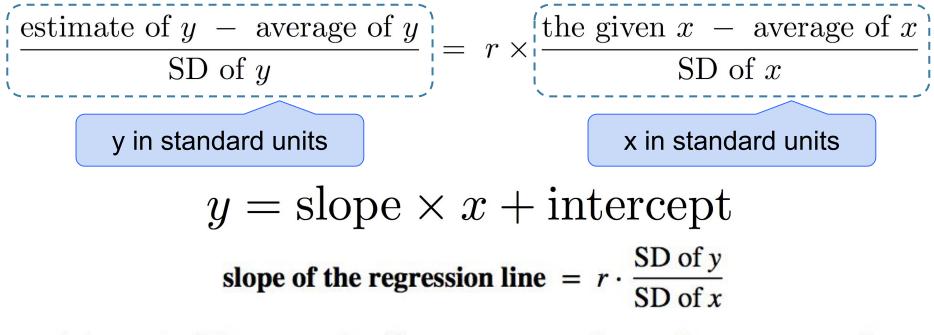
If we have a line describing the relation between two

variables, we can make predictions



## **Regression Line Equation**

In original units, the regression line has this equation:



**intercept of the regression line** = average of y - slope  $\cdot$  average of x

# **Errors and Predictions**

- error = actual value prediction
- RMSE = root mean square error
- Regression line has the minimum RMSE of all lines
- Names:
  - Regression line
  - Least squares line
  - "Best fit" line

## **Non-linear regression**

#### **Residuals**

#### Residuals

- Error in regression prediction
- residual
  - = observed *y* regression prediction of *y*
  - = vertical distance between each point and the best line

# **Residual Plot**

A scatter diagram of residuals

- Should look like an unassociated blob for linear relations
- But still contains patterns for non-linear relations
- Can reveal whether linear regression is appropriate







(Demo)

## **Mean and Stdev of Residuals**

No matter what the scatter looks like...

- mean(residuals) = 0
- SD(residuals) = RMSE = SD(y) \* sqrt(1  $r^2$ )



# **Clustering around line**

- "The correlation measures how clustered the points are about a straight line."
- SD(residuals) = RMSE = SD(y) \* sqrt(1  $r^2$ )
- so, RMSE / SD(y) = sqrt(1  $r^2$ )

#### **Bounds**

#### Rule of thumb:

- About 68% of values within 1 RMSE of prediction
- About 95% of values within 2 RMSE of prediction
- etc.

## What we can learn from *r*

- How clustered points are around a line
- How *y* depends on *x*
- How accurate linear regression predictions will be

