

# Political Science 209 - Fall 2018

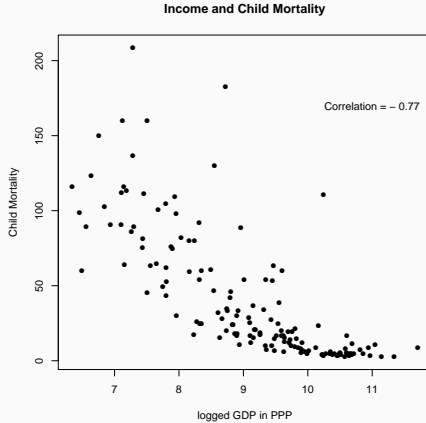
## Linear Regression

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Florian Hollenbach

11th October 2018

# Recall Correlation & Scatterplot



What is the correlation?

## Recall the definition of correlation

$$\text{Correlation (x,y)} = \frac{1}{N} \sum_{i=1}^N \text{z-score of } x_i \times \text{z-score of } y_i$$

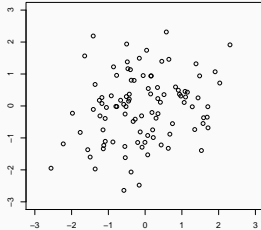
$$\text{Correlation (x,y)} = \frac{1}{N} \sum_{i=1}^N \frac{x_i - \bar{x}}{sd_x} \times \frac{y_i - \bar{y}}{sd_y}$$

# Correlations & Scatterplots/Data points

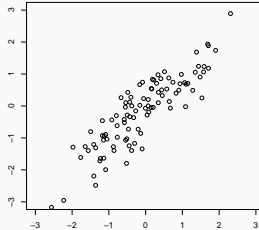
1. positive correlation  $\rightsquigarrow$  upward slope
2. negative correlation  $\rightsquigarrow$  downward slope
3. high correlation  $\rightsquigarrow$  tighter, close to a line
4. correlation **cannot** capture nonlinear relationship

# Correlations & Scatterplots/Data points

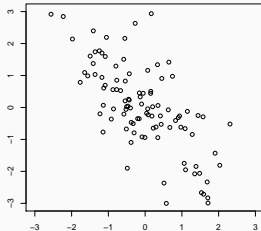
(a) correlation = 0.22



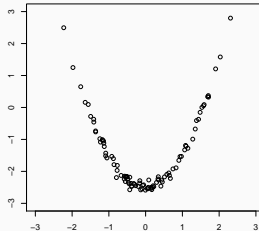
(b) correlation = 0.88



(c) correlation = -0.7



(d) correlation = 0.02



# Moving from Correlation to Linear Regression

Preview:

- linear regression allows us to create predictions
- linear regression specifies direction of relationship
- linear regression allows us to examine more than two variables at the same time (*statistical control*)

# Linear Regression

- regression has one **dependent (y)** and *for now* one **independent (x)** variable
- regression is a statistical method to estimate the linear relationship between variables

# Linear Regression

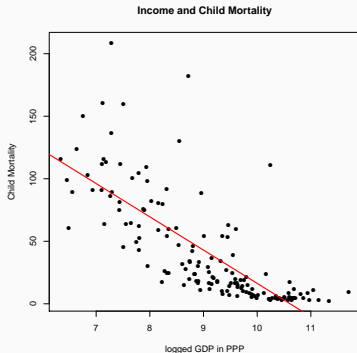
- goal of regression is to approximate the (linear) relationship between  $X$  and  $Y$  as best as possible



# Linear Regression

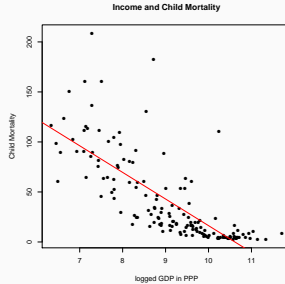
- goal of regression is to approximate the (linear) relationship between  $X$  and  $Y$  as best as possible
- regression is the mathematical model to draw best fitting line through cloud of points

# Linear Regression



Linear regression is the mathematical model to draw best fitting line through cloud of points

# Linear Regression



- regression line is an estimate of the (for now bivariate) relationship between  $x$  and  $y$
- for each  $x$  we have a prediction of  $y$ : what would we expect  $y$  to be given the value of  $x$ ?

# What is the equation of a line?

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Equation of a line?  $y = mx + b$

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

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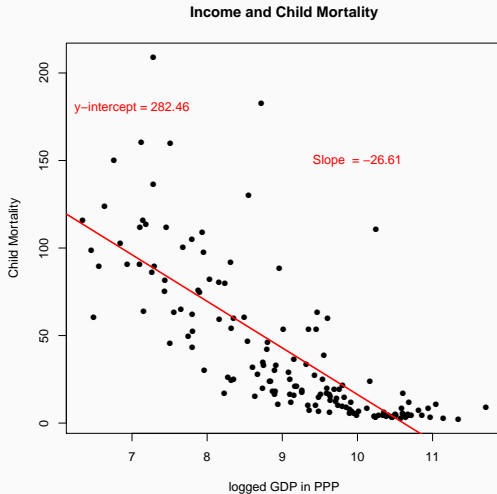
$\alpha \rightarrow$  y-intercept

$\beta \rightarrow$  slope

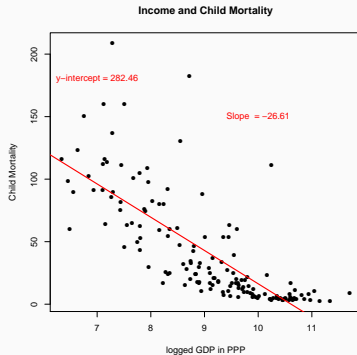
$\epsilon \rightarrow$  error



# Regression equation



# Regression equation



$$Y = 282.46 + -26.61X + \epsilon$$

# Regression equation

Model:

$$Y = \underbrace{\alpha}_{\text{intercept}} + \underbrace{\beta}_{\text{slope}} X + \underbrace{\epsilon}_{\text{error term}}$$

- $Y$ : dependent/outcome/response variable
- $X$ : independent/explanatory variable, predictor
- $(\alpha, \beta)$ : coefficients (parameters of the model)
- $\epsilon$ : unobserved error/disturbance term (mean zero)

# Regression: Interpretation of the Parameters:

$$Y = \underbrace{\alpha}_{\text{intercept}} + \underbrace{\beta}_{\text{slope}} X + \underbrace{\epsilon}_{\text{error term}}$$

- $\alpha + \beta X$ : average of  $Y$  at the given the value of  $X$
- $\alpha$ : the value of  $Y$  when  $X$  is zero
- $\beta$ : increase in  $Y$  associated with one unit increase in  $X$

# Regression equation

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# Regression equation

- but, we don't know the equation that generates the data
- our regression line is an estimate, based on the collected data
- estimates are denoted with little hats:  $\hat{\beta}$ ,  $\hat{\alpha}$
- $(\hat{\alpha}, \hat{\beta})$ : estimated coefficients
- we can use  $(\hat{\alpha}, \hat{\beta}, X)$  to create *predicted values* of  $y$
- $\hat{Y} = \hat{\alpha} + \hat{\beta}x$ : predicted/fitted value

How far off is our line? How do we know?



# Regression equation

How far off is our line? How do we know?

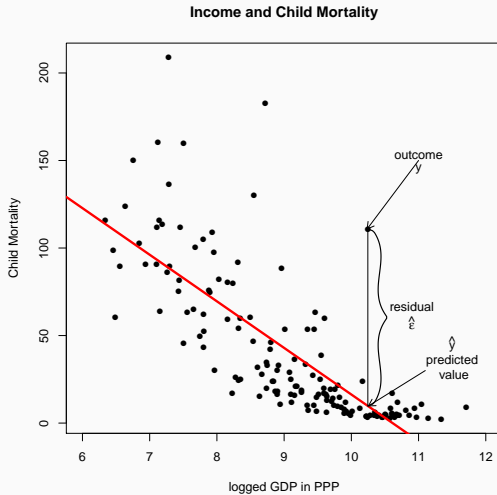
# Regression equation

How far off is our line? How do we know?

$\hat{\epsilon} = \text{true } Y - \hat{Y}$ : residuals/error

$\hat{\epsilon}$ 's are an estimate of how good/bad our line approximates the relationship

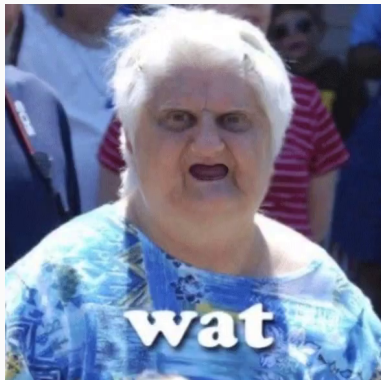
# Regression



- $(\alpha, \beta)$  are estimated from the data
- How do we find  $\alpha, \beta$ ?

## Regression: How do we find $\alpha, \beta$ ?

We minimize the sum of the squared residuals



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We minimize the *sum of the squared residuals (SSR)*

$$\text{SSR} = \sum_{i=1}^n \hat{\epsilon}_i^2 = \sum_{i=1}^n (Y_i - \hat{Y}_i)^2 = \sum_{i=1}^n (Y_i - \hat{\alpha} - \hat{\beta}X_i)^2$$

## Regression: How do we find $\alpha, \beta$ ?

We minimize the *sum of the squared residuals (SSR)*

$$\text{SSR} = \sum_{i=1}^n \hat{\epsilon}_i^2 = \sum_{i=1}^n (Y_i - \hat{Y}_i)^2 = \sum_{i=1}^n (Y_i - \hat{\alpha} - \hat{\beta}X_i)^2$$

This also minimizes the root mean squared error:  $\text{RMSE} = \sqrt{\frac{1}{n}\text{SSR}}$

$$\begin{aligned}\hat{\alpha} &= \bar{Y} - \hat{\beta}\bar{X} \\ \hat{\beta} &= \frac{\sum_{i=1}^n (Y_i - \bar{Y})(X_i - \bar{X})}{\sum_{i=1}^n (X_i - \bar{X})^2}\end{aligned}$$

OR:



$$\begin{aligned}\hat{\alpha} &= \bar{Y} - \hat{\beta}\bar{X} \\ \hat{\beta} &= \frac{\sum_{i=1}^n (Y_i - \bar{Y})(X_i - \bar{X})}{\sum_{i=1}^n (X_i - \bar{X})^2}\end{aligned}$$

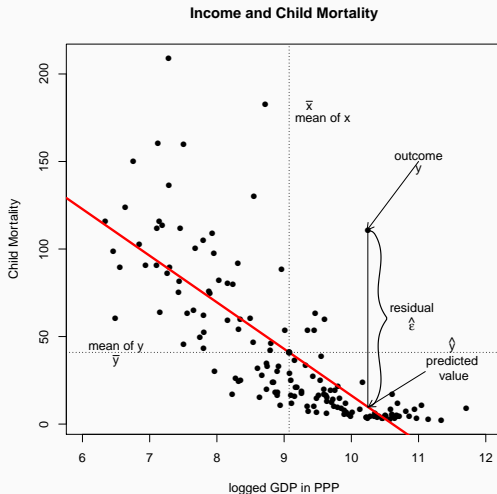
OR:

$$\hat{\beta} = \text{correlation of } X \text{ and } Y \times \frac{\text{standard deviation of } Y}{\text{standard deviation of } X}$$

Regression line always goes through the point of averages  $(\hat{X}, \hat{Y})$

$$\hat{Y} = (\bar{Y} - \hat{\beta}\bar{X}) + \hat{\beta}\bar{X} = \bar{Y}$$

# Regression always goes through point of averages



# Regression NOT by Hand

Enough math!

Fitting/estimating a regression in *R*:

```
lm(dependent ~ independent, data = data_object)
```

# Regression NOT by Hand

Fitting/estimating a regression in *R*:

```
data <- read.csv("bivariate_data.csv")  
data <- subset(data, Year ==2010)  
result <- lm(Child.Mortality ~ log(GDP) , data = data)  
summary(result)
```

# Regression NOT by Hand

```
result <- lm(Child.Mortality ~ log(GDP) , data = data)
coef(result) ### coefficients
```

| (Intercept) | log(GDP)  |
|-------------|-----------|
| 282.45870   | -26.61347 |

*R*-output:

(Intercept):  $\alpha$

$\log(GDP)$ :  $\beta$

How well does our regression line fit the data?

How well does the model predict the outcome?

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How well does the model predict the outcome?

$R^2$  or *coefficient of determination*:

$$R^2 = 1 - \frac{\text{SSR}}{\text{Total sum of squares (TSS)}} = 1 - \frac{\sum_{i=1}^n \hat{\epsilon}_i^2}{\sum_{i=1}^n (Y_i - \bar{Y})^2}$$



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$R^2$  is also defined as the *explained variance* in  $Y$

How much of the deviation of  $Y$  from the average is explained by  $X$ ?

# Model Fit

```
result <- lm(Child.Mortality ~ log(GDP) , data = data)
summary(result)
```

Call:

```
lm(formula = Child.Mortality ~ log(GDP), data = data)
```

Residuals:

| Min     | 1Q      | Median | 3Q     | Max     |
|---------|---------|--------|--------|---------|
| -49.455 | -15.418 | -4.161 | 10.847 | 132.136 |

Coefficients:

|             | Estimate | Std. Error | t value | Pr(> t )   |
|-------------|----------|------------|---------|------------|
| (Intercept) | 282.459  | 16.569     | 17.05   | <2e-16 *** |
| log(GDP)    | -26.613  | 1.809      | -14.71  | <2e-16 *** |

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codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 27.57 on 150 degrees of freedom

Multiple R-squared: 0.5906, Adjusted R-squared: 0.5878

F-statistic: 216.4 on 1 and 150 DF, p-value: < 2.2e-16