# 105demography

B-MAT-100

## Relationships between variables

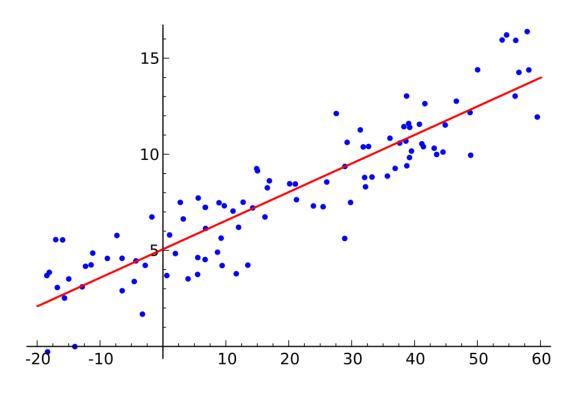
- There often is relationships between different variables
  - Example: radius and area of a circle, gas pressure and temperature...
- This relationships are expressed with an equation:

$$Y = f(X)$$

- X is the independent variable and Y is the dependent variable
- Y is explained by X
- What if your data come from observations and no exact relationship exists?

#### Regression analysis

- Estimation of the relationships between variables
- Given a set of N data points  $(X_i, Y_i)$ , the goal is to find a function f such as  $Y_i \approx f(X_i)$
- This function is called a regression, or a fit



## Correlation theory

- Correlation indicates how closely the data fits the regression
- If the fit is exact, there is a perfect correlation
  - Example: radius and area of a circle
- There is no correlation when the variables are independent
  - Example: two dice rolls
- If the variable are somewhat related, there is some correlation
  - Example: size and weight of an individual

#### Linear regression

- Simple model with a single independent variable
- The regression is a line:

$$f(X) = aX + b$$

• The differences between the prediction of the fit  $f(X_i) = \widehat{Y}_i$  and the actual observation  $Y_i$  are called the residuals  $\varepsilon_i$ 

$$Y_i = \widehat{Y}_i + \varepsilon_i = aX_i + b + \varepsilon_i$$

a and b are obtained by minimizing the sum of squared residuals.
 This is the method of least squares

## Least squares (1/2)

 The goal is to find a and b to minimize

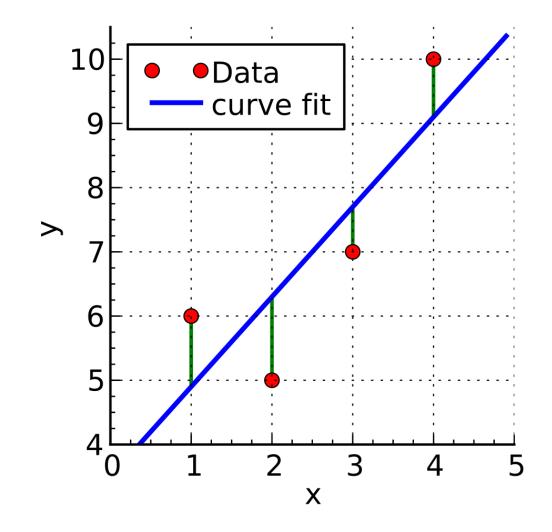
$$S = \sum_{i=1}^{N} \varepsilon_i^2$$

• Since  $Y_i = aX_i + b + \varepsilon_i$  we can write

$$S = \sum_{i=1}^{N} (Y_i - aX_i - b)^2$$

• The minimum is found by setting the gradient to 0:

$$\frac{\partial S}{\partial a} = \frac{\partial S}{\partial b} = 0$$



# Least squares (2/2)

By expanding the derivates we get the following equations

$$\begin{cases}
\sum_{i=1}^{N} Y_i = a \sum_{i=1}^{N} X_i + bN \\
\sum_{i=1}^{N} X_i Y_i = a \sum_{i=1}^{N} X_i^2 + b \sum_{i=1}^{N} X_i
\end{cases}$$

And then

$$a = \frac{N(\sum XY) - (\sum X)(\sum Y)}{N(\sum X^2) - (\sum X)^2}$$

$$b = \frac{(\sum Y)(\sum X^2) - (\sum X)(\sum XY)}{N(\sum X^2) - (\sum X)^2}$$

#### Root-mean-squared deviation

• Quantifies the amount of dispersion of the data around the fit

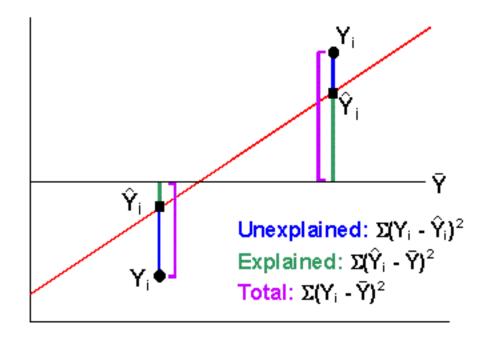
$$s_{Y,X} = \sqrt{\frac{\sum_{i=1}^{N} \varepsilon_i^2}{N}} = \sqrt{\frac{\sum_{i=1}^{N} (Y_i - \widehat{Y}_i)^2}{N}}$$

• Similar properties as the standard deviation: if N is large enough, 68% of the data points are at a distance less than  $s_{Y,X}$ , 95% at less than  $2s_{Y,X}$ , and 99.7% at less than  $3s_{Y,X}$ .

#### Explained and unexplained variances

$$\sum_{i=1}^{N} (Y_i - \bar{Y})^2 = \sum_{i=1}^{N} (Y_i - \hat{Y}_i)^2 + \sum_{i=1}^{N} (\hat{Y}_i - \bar{Y})^2$$

- $\sum (Y \overline{Y})^2$  is the total variance of Y (TSS: Total sum of squares)
- $\sum (Y \widehat{Y}_i)^2$  is the unexplained variance (RSS: Residual sum of squares)
- $\sum (\widehat{Y}_i \overline{Y})^2$  is the explained variance (ESS: Explained sum of squares)



#### Coefficient of correlation

 The correlation coefficient is the proportion of expected variance in the total variance:

$$r = \sqrt{\frac{ESS}{TSS}} = \sqrt{\frac{\sum (\widehat{Y}_i - \overline{Y})^2}{\sum (Y_i - \overline{Y})^2}}$$

• If  $\sigma_Y$  is the standard deviation of Y, r can also be written

$$r = \sqrt{1 - \frac{RSS}{TSS}} = \sqrt{1 - \frac{\sum (Y_i - \widehat{Y}_i)^2}{\sum (Y_i - \overline{Y})^2}} = \sqrt{1 - \frac{s_{Y,X}^2}{\sigma_Y^2}}$$

• r measures the quality of the fit

#### Covariance

• If we suppose there is a linear relationship between two variables X and Y with standard deviations  $\sigma_X$  and  $\sigma_Y$ , we can define the covariance as:

$$cov(X,Y) = \frac{E[(X - E[X])(Y - E[Y])]}{\sum (X_i - \overline{X}) \sum (Y_i - \overline{Y})}$$
$$= \frac{\sum (X_i - \overline{X}) \sum (Y_i - \overline{Y})}{N}$$

• In this case, the coefficient of correlation can now be expressed as:

$$r = \frac{cov(X,Y)}{\sigma_X \sigma_Y}$$

#### 105demography

- Goal: Use the linear least square regression to predict a country's population
- 105demography\_data.csv: data file containing every country's population from 1960 to 2017
- Inputs: one or several country codes

## 105demography

- *Y* is the population
- *X* is the year
- Outputs
  - Coefficients of the fit  $Y = a_X X + b_X$
  - Root-mean-square deviation of the fit
  - Population prediction in 2050 according to this fit
  - Coefficients of the fit  $X = a_Y Y + b_Y$
  - Root-mean-square deviation of the fit
  - Population prediction in 2050 according to this fit
  - Coefficient of correlation

#### Exercise: Sums

- Given a data set  $(X_i, Y_i)$ , compute the following sums:
  - $\sum X$
  - $\sum Y$
  - $\sum X^2$
  - $\sum Y^2$
  - $\sum X.Y$
- It should then be easy to compute the coefficients of a linear fit

#### Exercise: root-mean-square deviation

• Given a data set  $(X_i, Y_i)$  and a linear fit, compute the root-mean-square deviation of the fit.

$$s_{Y,X} = \sqrt{\frac{\sum_{i=1}^{N} (Y_i - \widehat{Y}_i)^2}{N}}$$

## Exercise: data parsing

 Given a country code, parse the data file 105demography\_data.csv and store the corresponding population data for every year from 1960 to 2017.