

# Linear Regression

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# Overview: Part I

# Regression Analysis: What for?

Regression analysis is a statistical technique to:

- (i) **Describe** the relationship between a response variable and a set of predictor variables;
- (ii) **Predict** the value of the response variable based on values of the predictor variables;

## A Bit of History

- ▶ Francis Galton: Study of family resemblances, 1885
- ▶ 928 child-parent pairs: Height of the adult child and the “midparent height” (average height of the father and the mother)
- ▶ “**Regression to mean**”: Children’s heights tend to be more “moderate” than their parents

Child(in) Midparent(in)

1 61.57220 70.07404

2 61.24382 68.22505

3 61.90968 65.12639

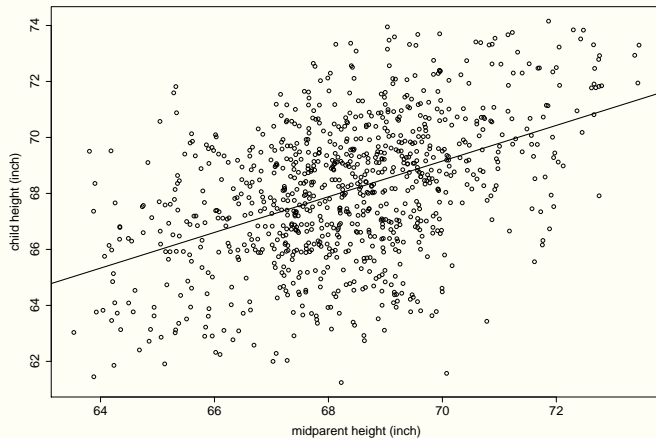
4 61.85769 64.23529

5 61.44986 63.88177

6 62.00005 67.02702

. . . . .

**Figure:** Scatter plot: Child's height versus midparent height



- ▶ Foot-ball shaped data cloud → linear relationship
- ▶ Fitted regression line:

$$Y = 24.54 + 0.637X$$

- ▶ If the midparent's height is 72in, then the child's height is predicted to be:

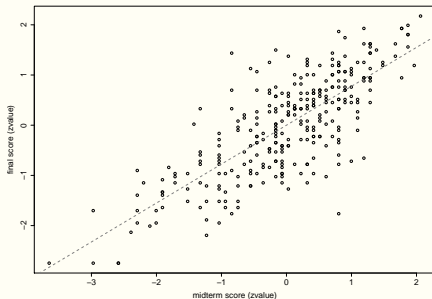
$$24.54 + 0.637 \times 72in = 70.4in.$$

# Overview: Part II



# Exam Scores

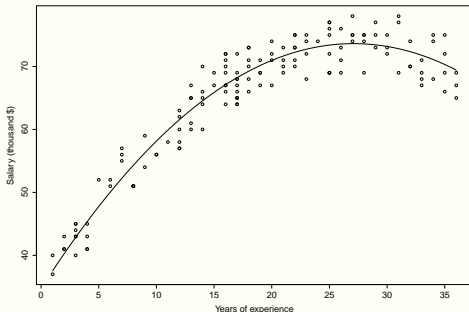
Figure: Standardized final score versus standardized midterm score



*If a student's midterm score is 2 standard deviations above the class mean, then their predicted final score would be ...?*

# Salary

Figure: Salary versus years of experience



*Would a straight line fit the data well?*

# Body Fat

Accurate measure of body fat is costly. It is desirable to use a set of easily obtainable measurements to predict the body fat. E.g., Age (years), Weight (lbs), Height (inches), Neck circumference (cm), Chest (cm), Abdomen 2 (cm), Hip (cm), Thigh (cm), Knee(cm), Ankle (cm), Biceps (cm), Forearm (cm), Wrist (cm).

*Are all these needed for predicting body fat? Are their effects linear?*

# Questions to Be Studied

- ▶ How to estimate the regression relationship?
- ▶ How reliable are the regression estimates?
- ▶ How reliable are the predictions?
- ▶ How to interpret estimated regression coefficients?
- ▶ Does the model fit the data? Do model assumptions hold?
- ▶ How to choose  $X$  variables? How to choose between competing models? How to validate a model?

# Review of Some Basics

# Summation Operator: $\Sigma$

►  $\sum_{i=1}^n Y_i = Y_1 + Y_2 + \cdots + Y_n$

► Variants:

$$\sum_{i=1}^n Y_i, \quad \sum_1^n Y_i, \quad \sum Y_i$$

► Useful facts:

$$\sum_{i=1}^n c = n \cdot c$$

$$\sum_{i=1}^n (Y_i + Z_i) = \sum_{i=1}^n Y_i + \sum_{i=1}^n Z_i$$

$$\sum_{i=1}^n (c \cdot Y_i) = c \cdot \sum_{i=1}^n Y_i$$

**Notes:** in general  $\sum_{i=1}^n (c_i \cdot Y_i) \neq (\sum_{i=1}^n c_i) \cdot (\sum_{i=1}^n Y_i)!$

## Expectation Operator: $E(\cdot)$

- ▶ Discrete random variable:  $E(Y) = \sum_y y \cdot P(Y = y)$
- ▶ Continuous random variable:  $E(Y) = \int y \cdot f(y) dy$
- ▶ Variants: expectation, mean, expected value
- ▶ Useful facts:  $c$  is a (non-random) constant and  $Y, Z$  are random variables

$$E(c) = c$$

$$E(Y + Z) = E(Y) + E(Z)$$

$$E(c + Y) = c + E(Y)$$

$$E(c \cdot Y) = c \cdot E(Y)$$

## Variance Operator: $\text{Var}(\cdot)$

- ▶  $\text{Var}(Y) = E((Y - E(Y))^2)$
- ▶ Variants:  $\sigma^2\{Y\}$
- ▶ Useful facts:  $c$  is a (non-random) constant and  $Y, Z$  are random variables

$$\text{Var}(Y) = E(Y^2) - (E(Y))^2 \geq 0$$

$$\text{Var}(c) = 0$$

$$\text{Var}(Y + Z) = \text{Var}(Y) + \text{Var}(Z), \text{ if } Y \text{ and } Z \text{ are uncorrelated}$$

$$\text{Var}(c + Y) = \text{Var}(Y)$$

$$\text{Var}(c \cdot Y) = c^2 \cdot \text{Var}(Y)$$



## Covariance Operator: $\text{Cov}(\cdot, \cdot)$

- ▶  $\text{Cov}(Y, Z) = E\left(\left(Y - E(Y)\right) \cdot \left(Z - E(Z)\right)\right)$
- ▶ Variants:  $\sigma\{Y, Z\}$
- ▶ Useful facts:  $c_1, c_2$  are (non-random) constants and  $Y, Z$  are random variables

$$\text{Cov}(Y, Z) = E(Y \cdot Z) - E(Y) \cdot E(Z)$$

$$\text{Cov}(Y, Y) = \text{Var}(Y)$$

$$\text{Cov}(c_1 + Y, c_2 + Z) = \text{Cov}(Y, Z)$$

$$\text{Cov}(c_1 \cdot Y, c_2 \cdot Z) = c_1 \cdot c_2 \cdot \text{Cov}(Y, Z)$$

$$\text{Var}(Y + Z) = \text{Var}(Y) + \text{Var}(Z) + 2 \cdot \text{Cov}(Y, Z)$$

# Hypothesis Testing: Components

- ▶ **Null hypothesis**  $H_0$ : “status quo”. E.g.,  $H_0$  : not guilty
  - ▶ Null hypothesis is assumed to be true unless there is strong evidence against it. E.g., A defendant is assumed not guilty unless proved otherwise by evidence that is beyond reasonable doubts.
- ▶ **Alternative hypothesis**  $H_a$ : “new theory”. E.g.,  $H_a$  : guilty
  - ▶ Alternative hypothesis is being accepted **only when one can reject the null hypothesis with strong evidence.**
- ▶ A testing procedure is to ascertain if the evidence (in data) against the null hypothesis is strong enough to reject it.

# Type I and Type II Errors

Decision \ Actual Fact	$H_0$ true	$H_0$ false
	Accept $H_0$	Type II Error
Reject $H_0$	Type I Error	Correct

- ▶ **Significance level  $\alpha$ :** (prespecified) maximum allowable **type I error rate**. E.g.,  $\alpha = 0.05$ ,  $\alpha = 0.1$ ,  $\alpha = 0.01$ . **Type I error rate** is not to exceed the significance level  $\alpha$ .
- ▶ **Type II error rate (=1-power)** is affected by the testing procedure, signal-to-noise ratio, sample size, significance level  $\alpha$ , etc.

# Point Estimator

A procedure to calculate a numerical quantity based on a (random) sample.

- ▶ E.g., Sample mean as an estimator for the population mean; sample proportion as an estimator for the population proportion.
- ▶ An estimator is a random variable.
- ▶ Error in the estimation: a numerical measure of the reliability of the estimation. E.g., standard error (SE) of an estimator.

What is the standard error of the sample mean?

# Confidence Interval

A (random) interval that covers the parameter of interest with a (pre-specified) high probability.

- ▶ Confidence coefficient (level/coverage) : the pre-specified probability of coverage, denoted by  $1 - \alpha$ . E.g.,  $1 - \alpha = 90\%$ ,  $1 - \alpha = 95\%$ ,  $1 - \alpha = 99\%$ .
- ▶ The corresponding interval is called a  $(1 - \alpha)100\%$ -confidence interval.
- ▶ One common form:  $Estimator \pm multiplier_{\alpha} \times SE(Estimator)$

What is a 95% confidence interval for the population mean?