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## Chapter 1

# Statistics for the Working Economist

### § 1.1. Measurements

**Definition 1.1.1.**

- (1) The large body of data that is the target of our interest is called the *population*.
- (2) A subset selected from a given population is called a *sample*.

**Definition 1.1.2.** The *mean* of a sample of  $n$  measured responses  $y_1, \dots, y_n$  is given by:

$$\bar{y} = \frac{1}{n} \sum_{i=1}^n y_i.$$

The corresponding *population mean* is denoted  $\mu$ .

**Remark.** We usually cannot measure the value of the population mean,  $\mu$ ; rather,  $\mu$  is an unknown constant that we may want to estimate using sample information.

**Definition 1.1.3.** The *variance* of a sample of measurements  $y_1, \dots, y_n$  is given by:

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})^2.$$

The corresponding *population variance* is denoted by  $\sigma^2$ . The larger the variance of a set of measurements, the greater will be the amount of variation within the set

**Definition 1.1.4.** The *standard deviation* of a sample of measurements is given by:

$$s = \sqrt{s^2}.$$

The corresponding *population standard deviation* is denoted by  $\sigma = \sqrt{\sigma^2}$

**Definition 1.1.5.** An *estimator* is a formula that tells how to calculate the value of an estimate based on the measurements contained in a sample.

**Example 1.1.1.** The sample mean  $\bar{y}$  is one possible point estimator of a population mean  $\mu$ .

## § 1.2. Linear Models

In this chapter, we undertake a study of inferential procedures that can be used when a random variable  $Y$ , called the *dependent variable*, has a mean that is a function of one or more non-random variables  $x_1, \dots, x_k$  called *independent variables*.

**Definition 1.2.1.** A *linear statistical model* relating a random response  $Y$  to a set of independent variables  $x_1, \dots, x_k$  is of the form:

$$Y = \beta_0 + \beta_1 x_1 + \dots + \beta_k x_k + \epsilon,$$

where  $\beta_0, \dots, \beta_k$  are unknown parameters, epsilon is a random variable (typically an error of some sort), and the variables  $x_1, \dots, x_k$  are known values.

**Remark.** Although unreasonable, we will assume that  $E(\epsilon) = 0$ , hence that:

$$E(Y) = \beta_0 + \beta_1 x_1 + \dots + \beta_k x_k.$$