

# Lecture 19

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## 1 Sampling

We cannot measure the whole population, so we will have to make do with a subset. So we take samples randomly.

**Definition 1.1.** Sample refers to the data  $x_1, \dots, x_n$  collected, with random variables  $X_1, \dots, X_n$ .

Usually, these data are **independent identically distributed** (IID). Suppose each  $X$  has mean  $\mu$  and variance  $\sigma^2$ .

### 1.1 Sample Mean

The empirical mean is

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$

And similarly, the random variable  $\bar{X}$  is defined as

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

and obviously

$$E[X] = \mu$$

## 1.2 Sample Variance

$$S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$$

We want that  $E[S^2] = \sigma^2$ . Note that

$$\begin{aligned} E[S^2] &= E \left[ \frac{1}{n-1} \sum_{i=1}^n (X_i^2 - 2X_i\bar{X} + \bar{X}^2) \right] \\ &= E \left[ \frac{1}{n-1} \left( \sum_{i=1}^n X_i^2 \right) - n\bar{X}^2 \right] \\ &= \frac{1}{n-1} (n\mu^2 + n\sigma^2 - n\mu^2 - n\text{var}(\bar{X})) \\ &= \frac{1}{n-1} \left( n\sigma^2 - n\text{var} \left( \frac{1}{n} \sum_{i=1}^n X_i \right) \right) \\ &= \frac{1}{n-1} \left( n\sigma^2 - \frac{1}{n} \sum_{i=1}^n \text{var}(X_i) \right) \\ &= \frac{1}{n-1} (n\sigma^2 - \sigma^2) \\ &= \sigma^2 \end{aligned}$$