

# Notes on Descriptive Statistics

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A descriptive statistic is a summary statistic that quantifiably describes or summarises features from a collection of information.

Descriptive statistics is the process of using and analysing those statistics.

## Expectation

The **expected value** of a random variable  $X$ , denoted  $E(X)$  or  $E[X]$  is a weighted average of the possible values that  $X$  can take, with each value being weighted according to the probability of that event occurring.

I.e. intuitively, **expectation is the arithmetic mean of a random variable  $X$ .**

## Definition

### Discrete

If  $F_X(x)$  is the pmf of the discrete random variable  $X$  with  $R_X = \{x_1, x_2, \dots\}$ ,  $E[X]$  is defined as:

$$\begin{aligned} E[X] &= \sum_{x \in R_X} x F_X(x) \\ &= \sum_{x \in R_X} x p_X(x) \\ &= \sum_{x \in R_X} x P(X = x) \end{aligned}$$

### Continuous

If  $F_X(x)$  is the pdf of the continuous random variable  $X$ ,  $E[X]$  is defined as:

$$E[X] = \int_{-\infty}^{\infty} x F_X(x) dx$$

## Example

Here's a very **simple** example: What is the expected value when we roll a fair die?

$$\omega = \{1, 2, 3, 4, 5, 6\}$$

$$P(X = x \mid x \in \omega) = \frac{1}{6}$$

$$E[X] = \sum_{x \in R_X} xP(X = x)$$

$$E[X] = 1 \times P(X = 1) + 2 \times P(X = 2) + 3 \times P(X = 3) + 4 \times P(X = 4) + 5 \times P(X = 5) + 6 \times P(X = 6)$$

$$E[X] = \frac{1}{6} + \frac{2}{6} + \frac{3}{6} + \frac{4}{6} + \frac{5}{6} + \frac{6}{6} = \frac{7}{2}$$

So the expectation is 3.5.

If you think about it, 3.5 is halfway between the possible values the die can take and so this is what you should have expected.

## Variance

The **variance** of a random variable  $X$ , denoted  $Var(X)$  and sometimes  $\sigma^2$ , measures how far a set of numbers is spread out from their average value.

### Definition

The variance of a random variable  $X$  is the expected value of the squared deviation (difference) from the mean of  $X$ . The mean of  $X$  is  $\mu = E[X]$ .

### Discrete

$$\begin{aligned} Var(X) &= E[(X - \mu)^2] \\ &= \sum_{x \in R_X} p_X(x)(x - \mu)^2 \\ &= E[X^2] - E[(X)^2] \\ &= \sum_{x \in R_X} p_X(x)x^2 - \mu^2 \end{aligned}$$

### Continuous

$$\begin{aligned} \text{Var}(X) &= E[(X - \mu)^2] \\ &= \int_{-\infty}^{\infty} p_X(x)(x - \mu)^2 \\ &= E[X^2] - E[(X)^2] \\ &= \int_{-\infty}^{\infty} p_X(x)x^2 - \mu^2 \end{aligned}$$

Where  $p_X(x)$  is shorthand notation for pmf.