Mean and Standard Deviation for Random Variables

Reading

• Section 2.4.1, 2.4.2

Practice Problems

2.6.4 (Page 123) 2.34, 2.36, 2.38, 2.39

Notes

Mean of a Random Variable

The **mean** or **expected value** of a random variable can be thought of as the "long-term average", meaning the average of the outcomes of an ever increasing number of trials of the experiment.

Denoted E(X) or μ_X .

If the variable X takes the values $x_1, x_2, \dots x_n$ with probabilities p_1, p_2, \dots, p_n respectively, then the mean is defined as:

$$E(X) = \mu_X = p_1 x_1 + p_2 x_2 + \dots + p_n x_n$$

You can think of this as a *weighted average* of the values, with their probabilities as weights. This makes sense: We want to take all values into account, but those values that have a higher probability are meant to appear more often, and so should contribute more. *Each value contributes an amount proportionate to its relative frequency*.

As a simple example, consider the example from the last section, with probability table:

Then for the mean we would have:

$$E(X) = \frac{1}{2} \cdot 0 + \frac{1}{4} \cdot 1 + \frac{1}{4} \cdot 2 = \frac{3}{4} = 0.75$$

We can think of this as saying that if you were to play that game repeatedly, you would be gaining on average \$0.75 per game. You can also think of it as the "fair price to pay to play the game".

We examined a number of games in the previous section. Compute the mean of the random variables in each of those games.

Standard Deviation of a Random Variable

The standard deviation follows a similar formula:

$$\sigma_X^2 = p_1(x_1 - \mu_X)^2 + p_2(x_2 - \mu_X)^2 + \dots + p_n(x_n - \mu_X)^2$$

So we look at how far each value is from the mean, square to remove the signs, average while accounting for the different probabilities, and finally take a square root.

This square of the standard deviation, typically called the *Variance* Var(X), you will often see written as $E((X - \mu_X)^2)$.

Compute the standard deviation for each of the examples discussed so far.