

**Section 7.3: Variance and Standard Deviation**

The **Variance** of a random variable  $X$  is the measure of degree of dispersion, or spread, of a probability distribution about its mean (i.e. how much on average each of the values of  $X$  deviates from the mean).

Note: A probability distribution with a small (large) spread about its mean will have a small (large) variance.

**Variance of a Random Variable X**

Suppose a random variable has the probability distribution

$x$	$x_1$	$x_2$	$\dots$	$x_n$
$P(X = x)$	$p_1$	$p_2$	$\dots$	$p_n$

and expected value  $E(X) = \mu$ . Then the variance of the random variable  $X$  is

$$\text{Var}(X) = p_1(x_1 - \mu)^2 + p_2(x_2 - \mu)^2 + \dots + p_n(x_n - \mu)^2$$

Note: We square each since some may be negative.

Standard Deviation measures the same thing as the variance. The standard deviation of a Random Variable  $X$  is

$$\sigma = \sqrt{\text{Var}(X)} = \sqrt{p_1(x_1 - \mu)^2 + p_2(x_2 - \mu)^2 + \dots + p_n(x_n - \mu)^2}$$

**Example 1:** Compute the mean, variance and standard deviation of the random variable  $X$  with probability distribution as follows:

$X$	$P(X=x)$
-3	0.4
2	0.3
5	0.3



**Chebychev's Inequality**

Let  $X$  be a random variable with expected value  $\mu$  and standard deviation  $\sigma$ . Then, the probability that a randomly chosen outcome of the experiment lies between  $\mu - k\sigma$  and  $\mu + k\sigma$  is at least  $1 - \frac{1}{k^2}$ ; that is,

$$P(\mu - k\sigma \leq X \leq \mu + k\sigma) \geq 1 - \frac{1}{k^2}$$

**Example 3:** A probability distribution has a mean 20 and a standard deviation of 3. Use Chebychev's Inequality to estimate the probability that an outcome of the experiment lies between 12 and 28.

**Example 4:** A light bulb has an expected life of 200 hours and a standard deviation of 2 hours. Use Chebychev's Inequality to estimate the probability that one of these light bulbs will last between 190 and 210 hours?