

## LECTURE 5 AND 6

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### 5 RANDOM VARIABLES

#### 5.1 Random Variables.

**Definition 1. Random Variable( $X$ ):** A variable whose value is the numerical output of a random event. There are many forms of Randvars but each is the output of some probability function.

#### 5.2 Discrete Random Variables.

**Definition 2. Discrete Random Variable:** A random variable  $X$  is said to be discrete if it takes either finitely many or a countable many values.

**Remark.** Examples:

- (1) Experiment: Toss a fair coin  $n$  times and let  $X$  be the number of Heads.
- (2) Experiment: The number  $X$  of accidents on I-95 during next week.
- (3) Experiment: Let  $X$  be the number of coin-tosses until the first Head comes up

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**Definition 3. Continous Random Variable:** A random variable  $X$  is said to be discrete if it can take any number within a variable.

**Remark.** Examples include the exact time of an event happening or measurements.

## 5.2 Probability Mass Function.

## 5.3 The Bernoulli Distributions.

**Definition 4. Bernoulli Trial:** A defined single trial that is classified as:

- (1) Binary Outcomes, ie Success or Failure
- (2) Independent observations
- (3) Fixed probability

**Remark.** Some times Bernoulli Randvars are called **Indicator Variables**

**Definition 5. Probability Mass Function:** An equation which defines the probability the  $X$  can take on a value, used for Bernoulli Trials:

$$P(X = x) = p^x(1 - p)^{1-x}$$

where  $X$  is the random variable and  $x$  is the specific value(1,0)

**Definition 6. Bernoulli Distribution:** A probabilistic deescription of the successes or failures of a single Bernoulli experiment/trial. Probability of Success(1) in a Bernouli trial is seen as  $p$  whereas probability of failure(0) is seen as  $1 - p$

**Definition 7. Mean and Variance of a Bernouli Distribution:** Because the values a Bernoulli trial can have(1,0), we can see that the mean  $\mu$  can be described as

$$E(X) = p$$

and the variance as

$$Var[X] = pq = p(1 - p)$$

## 5.4 Binomial Distributions.

**Definition 8. Binomial Distribution:** Distribution of the number of successes of a fixed amount( $n$ ) Bernoulli trials. For a binomial distribution the probability of individual events can be calculated using the formula:

$$P(X = x) = {}_nC_x p^x q^{n-x}$$

**Definition 9. Mean and Variance of a Binomial Distribution:** Because a Binomial distribution is multiple Bernoulli trials the mean can be found by

$$\mu = np$$

and the variance can be found by

$$\sigma^2 = np(1 - p)$$

### **Cumulative Distribution Functions(CDF).**

**Definition 10. Cumulative Distribution Functions(CDF):** Used to calculate the accumulated probability left of an area. Essentially the integral for values before X, made for Continuous distributions. The PDF for a continuous distribution is found by

$$f(x) = \frac{1}{b - a}$$

whereas the CDF is found by

### **Expectations of Discrete Randvars.**

**Definition 11. Expected Value of Binomial Distribution:** For a Binomial Distribution, the mean ends up being the expected value in that

$$E(X) = \mu = n \cdot p$$

across the different values resulting in:

$$E(x) = \sum x \cdot p(x)$$

### **Variance of Expectation.**

**Definition 12. Variance of Expected Value:** The variance of the expected value depends on the variance of the distribution, which means that we get the relation

$$Var(X) = E[(X - \mu)^2] = E(X^2) - (E(X))^2 = E(X^2) - \mu^2$$

which results in

$$\sigma^2 = \sum (x - \mu)^2 \cdot p(x)$$