

# CS70 In Simpler Terms - Note 5

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## 1 Random Variables and Expectations

There are a few main distributions that we will be focusing on in this course, but understanding the probability aspect and how to calculate their Expectations will be key to your success.

**Random Variable:** a function  $X : \omega \Rightarrow R$  that assigns a real number to every outcome  $\omega$  in the probability space. ( $\omega$  is a sample point)

*Example:*  $X(\omega) = \text{number of heads}$ , then  $X(THH) = 2$

### Expectation and Linearity:

- Expectation is the mean/average of the r.v.:  $E(x) = \sum_a xP(x = a)$
- Linearity:
  - For 2 r.v.  $x, y$  on the same probability space:  $E(X + Y) = E(X) + E(Y)$
  - For any constant  $c$ ,  $E(cX) = cE(X)$
  - **Note: Linearity of Expectation holds even when the r.v. is NOT Independent**
- $E(XY) = E(X)E(Y)$  iff  $X, Y$  are independent
- The expected number of fixed points in a random permutation of  $n$  items is always 1 regardless of  $n$  (e.g.  $E(X) = 1$ )
- Indicator Variables general form:
  - Let  $X_i$  be the indicator of whether...
  - $X = X_1 + X_2 + \dots + X_n \Rightarrow E(X) = \sum_{i=1}^n E(X_i)$
  - $E(X^2) = E([X_1 + X_2 + \dots + X_n]^2) = \sum_{i=1}^n E(X_i)^2 + \sum_{i \neq j}^{n(n-1)} E(X_i X_j)$   
**Explanation:** there are  $n(n-1)$  unlike terms in the expansion of  $(X_1 + X_2 + \dots + X_n)(X_1 + X_2 + \dots + X_n)$  (after selecting 1 of the  $n$  terms in the first set, you have  $n-1$  other terms to choose from in the second of the pair)
  - $E(X_i) = E(X_i^2)$  - since you only care about  $Pr(X_i = 1) \Rightarrow E(X_i) = 1 * Pr(X_i = 1) + 0 * Pr(X_i = 0)$

## 2 Variance and Covariance

**Definition:**  $Var(x) = E((x - \mu)^2) = \sigma^2$ , where  $\sigma$  is standard deviation. So  $Var(x) = (\sqrt{\sigma})^2$

- For r.v. with  $E(X) = \mu$ ,  $Var(X) = E(X^2) - \mu^2 = E(X^2) - E(X)^2$ 
  - Proof:  $Var(X) = E((x - \mu)^2) = E(X^2 - 2\mu X + \mu^2)$   
 $= E(X^2) - 2\mu E(x) + \mu^2$  and substituting  $\mu$  for  $E(X)$ , we get  
 $Var(X) = E(X^2) - \mu^2 = E(X^2) - E(X)^2$
- $Var(cX) = c^2 Var(X)$
- Independent r.v.:  $P(X = a, Y = b) = P(X = a)P(Y = b) \forall a, b$
- Covariance
  - Covariance of  $X, Y = E(XY) - E(X)E(Y)$ , and if  $X, Y$  are independent, the covariance is 0
  - $Var(X + Y) = Var(X) + Var(Y) + 2Cov(X, Y)$  and  $Cov(X, Y) = 0$  if  $X, Y$  are independent
  - $Cov(a + bX, c + dY) = bdCov(X, Y)$
  - $Cov(X_1 + X_2, Y) = Cov(X_1, Y) + Cov(X_2, Y)$
- Note that the value of  $Var(X) \geq 0$  so  $E(X^2) \geq E(X)^2$