

Properties of Discrete Random variables

1. Expectation/Mean/Expected Value

$$E(X) = \sum_{i=1}^n x_i p(x_i)$$

- ◆ Weighted average of the values the random variable X can take
- ◆ Weighting by the probability of each value
- ◆ Measure of centrality

2. Expectation of a constant multiple of a random variable

$$E[aX + b] = aE[X] + b$$

$$E[aX + b] = \sum_{i=1}^n (ax_i + b)p(x_i) \quad \text{From definition 1}$$

$$= \sum_{i=1}^n (ax_i)p(x_i) + \sum_{i=1}^n bp(x_i) \quad \text{By multiplying out the brackets}$$

$$= a \sum_{i=1}^n x_i p(x_i) + b \sum_{i=1}^n p(x_i) \quad \text{From the properties of summation}$$

$$= aE[X] + b \sum_{i=1}^n p(x_i) \quad \text{From definition 1}$$

$$= aE[X] + b.1 \quad \text{From axioms of probability}$$

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$$= aE[X] + b$$

3. Expectation of a function of random variable

$$E[g(X)] = \sum_i g(x_i)p(x_i)$$

- ♦ The expectation of a function of a random variable is **not equal** to the function of the expectation $E[g(X)] \neq g[E(X)]$

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4. Variance

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

Let $\mu = E[X]$

$$E[(X - \mu)^2] = \sum_{i=1}^n (x_i - \mu)^2 p(x_i) \quad \text{From definition}$$

$$= \sum_{i=1}^n (x_i^2 - 2\mu x_i + \mu^2) p(x_i) \quad \text{Multiplying out}$$

$$= \sum_{i=1}^n x_i^2 p(x_i) + \sum_{i=1}^n -2\mu x_i p(x_i) + \sum_{i=1}^n \mu^2 p(x_i)$$

$$= E[X^2] + \sum_{i=1}^n -2\mu x_i p(x_i) + \sum_{i=1}^n \mu^2 p(x_i) \quad \text{From definition 3}$$

$$= E[X^2] - 2\mu \sum_{i=1}^n x_i p(x_i) + \mu^2 \sum_{i=1}^n p(x_i) \quad \text{Properties of summations}$$

$$= E[X^2] - 2\mu\mu + \mu^2 \sum_{i=1}^n p(x_i)$$

$$= E[X^2] - 2\mu\mu + \mu^2 \quad \text{Axioms of probability}$$

$$= E[X^2] - \mu^2$$

$$= E[X^2] - (E[X])^2$$

5. Variance of a constant property

$$\text{Var}[a] = 0$$

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6. Variance of a constant multiple

$$\text{Var}[aX] = a\text{Var}[X]$$

$$\text{Var}[aX] = E[(aX - E[aX])^2]$$

$$= E[(aX - aE[X])^2] \quad \text{From definition 2}$$

$$= E[(aX - a\mu)^2] \quad \text{Letting } \mu = E[X]$$

$$= \sum_{i=1}^n (ax_i - a\mu)^2 p(x_i) \quad \text{From definition}$$

$$= \sum_{i=1}^n a^2 (x_i - \mu)^2 p(x_i)$$

$$= a^2 \sum_{i=1}^n (x_i - \mu)^2 p(x_i)$$

$$= a^2 \text{Var}[X] \quad \text{From definition 4}$$

7. Expectation of the sum of two finite countable variables

If X is a random variable with sample space $\{x_1, x_2, \dots, x_m\}$ and Y is an independent random variable with sample space $\{y_1, y_2, \dots, y_n\}$ then the sample space of the joint distribution will be given by a set of pairs

$$\{x_1, y_1\}, \{x_1, y_2\}, \dots, \{x_1, y_n\}$$

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$$\{x_2, y_1\}, \{x_2, y_2\}, \dots, \{x_2, y_n\}$$

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$$\{x_m, y_1\}, \{x_m, y_2\}, \dots, \{x_m, y_n\}$$

The expectation of the sum of the two variables is then given by

$$\sum_{i=1}^m \sum_{j=1}^n (x_i + y_j) p(x_i, y_j)$$

Multiplying out we get

$$\sum_{i=1}^m \sum_{j=1}^n x_i p(x_i, y_j) + \sum_{i=1}^m \sum_{j=1}^n y_j p(x_i, y_j)$$

Noting that $\sum_{j=1}^n p(x_i, y_j) = p(x_i)$ and $\sum_{i=1}^m p(x_i, y_j) = p(y_j)$

$$\sum_{i=1}^m x_i p(x_i) + \sum_{j=1}^n y_j p(y_j)$$

Therefore we can note that

$$E[X + Y] = E[X] + E[Y]$$

8. Expectation of the sum of n identically distributed random variables

We can calculate the expectation of the sum of n identically distributed random variables denoted by X_1, X_2, \dots, X_n as $E[X_1] + E[X_2] + \dots + E[X_n]$ which is equal to

$$n \cdot E[X_n]$$