

Recitation Note

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Chapter 1: Set Theory, Probability, and Single Experiment

1. From Set Theory to Probability (of the single experiment)

(a)

Set Theory	Probability
Element	Outcome
Subset	Event
Universal Set	Sample Space

(b) $\mathbb{P}(\text{Outcome})$ = the possibility that the outcome appears in the sample space. Outcomes are always **Mutually Exclusive**.

(c) $\mathbb{P}(\text{Event})$ = Event constitutes by different combinations of outcomes, including *null set* and the whole sample space.

(d) Apparently, $\mathbb{P}(\emptyset) = 0$ since there is no element in *null set*, and $\mathbb{P}(\text{Sample Space}) = 1$

2. Probability Properties

(a) $\mathbb{P}(A \cup B) = \mathbb{P}(A) + \mathbb{P}(B) - \mathbb{P}(A \cap B)$

(b) **Mutually Exclusive:** $\mathbb{P}(A \cap B) = 0$ so that $\mathbb{P}(A \cup B) = \mathbb{P}(A) + \mathbb{P}(B)$

(c) **Collectively Exhaustive:** $\mathbb{P}(A \cup B) = 1$

(d) **Partitions:** $\mathbb{P}(\cup_{i=1}^N A_i) = \sum_{i=1}^N \mathbb{P}(A_i) = 1$

3. Conditional Probability

(a) $\mathbb{P}(A|B) = \frac{\mathbb{P}(AB)}{\mathbb{P}(B)}$

(b) **If A_i are Mutually Exclusive:** $\mathbb{P}(A|B) = \sum_{i=1}^n \mathbb{P}(A_i|B)$

(c) **If B_i are Partitions of A:** $\mathbb{P}(A) = \sum_{i=1}^m \mathbb{P}(A \cap B_i) = \sum_{i=1}^m \mathbb{P}(A|B_i)\mathbb{P}(B_i)$ —Law of Total Number

4. Baye's Theorem: $\mathbb{P}(A|B) = \frac{\mathbb{P}(B|A)\mathbb{P}(A)}{\mathbb{P}(B)}$

5. **Independent:** $\mathbb{P}(A \cap B) = \mathbb{P}(A)\mathbb{P}(B)$ so that $\mathbb{P}(A \cup B) = \mathbb{P}(A) + \mathbb{P}(B) - \mathbb{P}(A)\mathbb{P}(B)$

Chapter 2: Sequential Experiments

1. Tree Diagrams

2. Counting Methods

(a) Multiplication: $n \times k_1 \times k_2 \times \dots$

(b) Sampling without Replacement

i. Permutation: $\frac{n!}{(n-k)!}$

ii. Combination: $\binom{n}{k} = \frac{n!}{k!(n-k)!}$

iii. Combination is Permutation without order. Combination is also called n choose k.

(c) Sampling with Replacement: n^k

(d) Multiple Combination: $\binom{n}{k_1, k_2, \dots, k_m} = \frac{n!}{k_1! k_2! \dots k_m!}$ where $n = \sum_{i=1}^m k_i$

(e) Multiple Permutation?

3. Independent Trails

(a) *Theorem 2.8:* The Probability of n_0 failures and n_1 successes in $n = n_0 + n_1$ Independent Trails with success rate p is

$$\mathbb{P}(n_0, n_1) = \binom{n}{n_1} (1-p)^{n_0} p^{n_1} = \binom{n}{n_0} (1-p)^{n_0} p^{n_1}$$

(b) *Theorem 2.9:* $n = n_0 + n_1 + \dots$ and success rates are p_0, p_1, \dots

4. Reliability Analysis

(a) Components in series

(b) Components in parallel

Chapter 3: View Graphs

1. Discrete Random Variables: Assign numerical value to outcomes

2. Probability Mass Function: $\sum_{x \in X} \mathbb{P}_X(x) = 1$