

Chapter 2 Summary

Definition 5: Odds

The odds in favour of an event A is

$$\frac{P(A)}{1-P(A)}$$

The odds against the event A is

$$\frac{1-P(A)}{P(A)}$$

Chapter 3 Summary

Properties of $\binom{n}{k}$

$$1. n^{(k)} = \frac{n!}{(n-k)!} = n(n-1)^{k-1} \text{ for } k \geq 1.$$

$$2. \binom{n}{k} = \frac{n!}{k!(n-k)!} = \frac{n^{(k)}}{k!}$$

$$3. \binom{n}{k} = \binom{n}{n-k} \text{ for all } k = 0, 1, \dots, n.$$

$$4. \binom{n}{0} = \binom{n}{n} = 1$$

$$5. \binom{n}{k} = \binom{n-1}{k-1} + \binom{n-1}{k}$$

$$6. \text{Binomial Theorem: } (1+x)^n = \sum_{k=0}^n \binom{n}{k} x^k$$

3.5 Notes

$$1. n^{(k)} = n(n-1) \cdots (n-k+1)$$

$$2. n^{(0)} = \frac{n!}{(n-0)!} = 1$$

$$0! = 1.$$

$$3. \binom{n}{k} = \frac{n^{(k)}}{k!} = \frac{n(n-1) \cdots (n-k+1)}{k!}$$

$$4. \text{When } n \text{ and } k \text{ are non-negative integers and } k > n \text{ then } \binom{n}{k} = \frac{n^{(k)}}{k!} = \frac{n(n-1) \cdots (1)(0) \cdots (n-k+1)}{k!} = 0$$

Geometric Series

$$\sum_{i=0}^{n-1} t^i = 1 + t + t^2 + \dots + t^{n-1} = \frac{1-t^n}{1-t} \quad \text{for } t \neq 1$$

$$\sum_{k=0}^{\infty} t^k = 1 + t + t^2 + \dots = \frac{1}{1-t} \quad \text{for } |t| < 1$$

Binomial Theorem

$$(1+t)^n = \sum_{k=0}^{\infty} \binom{n}{k} t^k \quad \text{if } |t| < 1$$

Multinomial Theorem

A generalization of the Binomial Theorem is

$$(t_1 + t_2 + \dots + t_k)^n = \sum \frac{n!}{\alpha_1! \alpha_2! \dots \alpha_k!} t_1^{\alpha_1} t_2^{\alpha_2} \dots t_k^{\alpha_k}$$

Hypergeometric Identity

$$\sum_{k=0}^{\infty} \binom{a}{k} \binom{b}{n-k} = \binom{a+b}{n}$$

Exponential Series

$$e^t = 1 + t + \frac{t^2}{2!} + \frac{t^3}{3!} + \dots = \sum_{n=0}^{\infty} \frac{t^n}{n!} \quad \text{for all } t \in \mathbb{R}$$

$$e^t = \lim_{n \rightarrow \infty} \left(1 + \frac{t}{n}\right)^n \quad \text{for all } t \in \mathbb{R}$$

Chapter 4 Summary

De Morgan's Laws

$$\overline{A \cup B} = \bar{A} \cap \bar{B}$$
$$\overline{A \cap B} = \bar{A} \cup \bar{B}$$

Rule 4a: Addition Law of Probability or The Sum Rule

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

Rule 4b: Probability of the union of 3 events

$$P(A \cup B \cup C) = P(A) + P(B) + P(C) - P(AB) - P(AC) - P(BC) + P(ABC)$$

Definition 6: Mutually Exclusive

Events A and B are mutually exclusive if $A \cap B = \emptyset$ and $P(A \cap B) = P(\emptyset) = 0$.

Rule 5a:

Let A and B be mutually exclusive events. Then

$$P(A \cup B) = P(A) + P(B)$$

Rule 5b

Let $A_1, A_2, \dots, A_n$ be mutually exclusive events. Then

$$P(A_1 \cup A_2 \cup \dots \cup A_n) = \sum_{i=1}^n P(A_i)$$

Rule 6:

$$P(A) = 1 - P(\bar{A})$$

Definition 7: Independent Events

Events A and B are independent events if and only if

$$P(A \cap B) = P(A)P(B)$$

Definition 8: Independent Events

Events $A_1, A_2, \dots, A_n$ are independent events if and only if

$$P(A_{i_1} \cap A_{i_2} \cap \dots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2}) \dots P(A_{i_k})$$

for all sets $(i_1, i_2, \dots, i_k)$ of distinct subscripts chosen from $(1, 2, \dots, n)$

Definition 9: Conditional Probability

The conditional probability of event A, given event B, is

$$P(A|B) = \frac{P(A \cap B)}{P(B)}, \text{ provided } P(B) > 0.$$

Rule 7: Product Rules

$$\begin{aligned} P(AB) &= P(A)P(B|A) \\ P(ABC) &= P(A)P(B|A)P(C|AB) \\ P(ABCD) &= P(A)P(B|A)P(C|AB)P(D|ABC) \end{aligned}$$

Rule 8: Law of Total Probability

Suppose that

$$A_1 \cup A_2 \cup \dots \cup A_k = S \text{ and } A_i \cap A_j = \emptyset \text{ if } i \neq j.$$

Let B be an arbitrary event in S, then

$$\begin{aligned} P(B) &= P(BA_1) + P(BA_2) + \dots + P(BA_k) \\ &= \sum_{i=1}^k P(A_i)P(B|A_i) \end{aligned}$$

Chapter 5 Summary

Hypergeometric Distribution

- A collection of N objects with 2 distinct types: success(S) and failure(F)
- There are r successes and $N-r$ failures
- Pick n objects without replacement
- Let X be the number of successes obtained. Then
- X has a Hypergeometric distribution, with parameters N, r, n
- PF:

$$f(x) = P(X=x) = \frac{\binom{r}{x} \binom{N-r}{n-x}}{\binom{N}{n}}$$

Note: $x \leq \min(r, n)$

Binomial Distribution

- An "experiment" with 2 distinct outcomes, success(S) and failure(F)
- $P(S)=p$ and $P(F)=1-p$
- Repeat the experiment n
- Let X be the number of successes obtained. Then,
- $X \sim \text{Binomial}(n, p)$
- PF:

$$f(x) = P(X=x) = \binom{n}{x} p^x (1-p)^{n-x}$$

Negative Binomial Distribution

- Two distinct types of outcome: S and F
- $P(S)=p$ on each trial (independent)
- Continue doing the experiment until k successes have occurred
- Let X be the number of failures before the 1st success. Then,
- $X \sim \text{Negative Binomial}(k, p)$
- PF:

$$f(x) = P(X=x) = \binom{x+k-1}{x} p^k (1-p)^x$$

for $x=0, 1, \dots$, $0 < p < 1$

Geometric Distribution

- Two distinct outcome: S and F
- $P(S)=p$ on each trial (independent)
- Let X be the number of failures before the k th success. Then,
- $X \sim \text{Geometric}(p)$
- PF:

$$f(x) = P(X=x) = (1-p)^x p$$

for $x=0, 1, \dots$ and $0 < p < 1$

Poisson Distribution from Binomial

- Approximate Binomial Distribution when n is large and p is small.
- PF:

$$f(x) = P(X=x) = \frac{n^x e^{-n}}{x!},$$

for $x=0, 1, \dots$

Poisson Distribution from Poisson Process

- Independence.
- Individuality: $P(2 \text{ or more events in } (t, t+\Delta t)) = o(\Delta t)$ as $\Delta t \rightarrow 0$.
- Homogeneity or Uniformity: $P(\text{one event in } (t, t+\Delta t)) = \lambda \Delta t + o(\Delta t)$
- PF:

$$f(x) = P(X=x) = \frac{(\lambda t)^x e^{-\lambda t}}{x!},$$

for $x=0, 1, \dots$

Chapter 7 Summary

Definition 16: Expected Value

Let X be a discrete random variable with $\text{range}(X) = A$ and probability function $f(x)$.

The expected value (also called the mean or the expectation) of X is

$$E(X) = \sum_{x \in A} x f(x)$$

Theorem 17:

Let X be a discrete random variable with $\text{range}(X) = A$ and probability function $f(x)$.

The expected value of some function $g(x)$ of X is given by

$$E(g(X)) = \sum_{x \in A} g(x) f(x)$$

Linearity of Expectation

For $a, b \in \mathbb{R}$ and functions g and g_2

$$E(ag_1(X) + bg_2(X)) = aE(g_1(X)) + bE(g_2(X))$$

Expected Value of Poisson Distribution

If $X \sim \text{Poisson}(\mu)$, then

$$E(X) = \mu.$$

Expected Value of Binomial Distribution

If $X \sim \text{Binomial}(n, p)$, then

$$E(X) = np.$$

Expected Value of Geometric Distribution

If $X \sim \text{Geometric}(p)$, then

$$E(X) = \frac{1-p}{p}$$