

# ST2334 Probability and Statistics

## Midterm Examination Helpsheet

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### 1 Basic Concepts of Probability

Terminology	Definition	Example
Statistical experiment	Procedure that produces data or observations	Rolling a dice
Sample space $S$	Set of all possible outcomes of a statistical experiment	{1, 2, 3, 4, 5, 6}
Sample point	An outcome in the sample space	1
Event	A subset of the sample space	An odd number facing up

- The sample space itself is an event and called a *sure event*.
- An event that contains no element is called a *null event*  $\emptyset$ .

**Event:** union  $A \cup B$ , intersection  $A \cap B$ , complement  $A'$ .

- Mutually exclusive/disjoint:  $A \cap B = \emptyset$ .  
 $\triangleright P(A \cap B) = 0$  does not mean mutually exclusive (e.g., continuous).
- Contained:  $A \subset B$ .
- Equivalent:  $A \subset B$  and  $B \subset A \Leftrightarrow A = B$ .

$A \cap A' = \emptyset$	$A \cap \emptyset = \emptyset$	$A \cup A' = S$	$(A')' = A$
$A \cup (B \cap C) = (A \cup B) \cap (A \cup C)$	$A \cap (B \cup C) = (A \cap B) \cup (A \cap C)$	$A \cup B = A \cup (B \cap A')$	$A = (A \cap B) \cup (A \cap B')$

- De Morgan's law:  $(A_1 \cup A_2 \cup \dots \cup A_n)' = A_1' \cap A_2' \cap \dots \cap A_n'$ ;  
 $(A_1 \cap A_2 \cap \dots \cap A_n)' = A_1' \cup A_2' \cup \dots \cup A_n'$ .

**Counting:** Multiplication principle + Addition principle.

- Permutation:  $P_r^n = \frac{n!}{r!(n-r)!}$ .
- Combination:  $C_r^n = \binom{n}{r} = \frac{n!}{r!(n-r)!}$ .

**Probability:** How likely event  $A$  occurs,  $P(A)$ .

- Relative frequency:  $f_A = \frac{n_A}{n} \rightarrow P(A)$  as  $n \rightarrow \infty$ .  
 $\triangleright 0 \leq f_A \leq 1$ ;  
 $\triangleright f_A = 1$  if  $A$  occurs in every repetition;  
 $\triangleright$  If  $A$  and  $B$  are mutually exclusive, then  $f_{A \cup B} = f_A + f_B$ .
- Axioms of probability:
  - For any event  $A$ ,  $0 \leq P(A) \leq 1$ ;
  - $P(S) = 1$ ;
  - $A \cap B = \emptyset \Rightarrow P(A \cup B) = P(A) + P(B)$ .  
 $\triangleright P(\emptyset) = 0$ ;  
 $\triangleright$  If  $A_1, A_2, \dots, A_n$  are mutually exclusive events, then  $P_{A_1 \cup A_2 \cup \dots \cup A_n} = P(A_1) + P(A_2) + \dots + P(A_n)$ .  
 $\triangleright$  For any event  $A$ ,  $P(A') = 1 - P(A)$ .  
 $\triangleright$  For any events  $A, B$ ,  $P(A) = P(A \cap B) + P(A \cap B')$ .  
 $\triangleright$  For any events  $A, B$ ,  $P(A \cup B) = P(A) + P(B) - P(A \cap B)$ .  
 $\triangleright$  If  $A \subset B$ , then  $P(A) \leq P(B)$ .
- Conditional probability:  $P(B|A) = \frac{P(A \cap B)}{P(A)}$ .  
 $\triangleright P(A \cap B) = P(A)P(B|A) = P(B)P(A|B)$  if  $P(A), P(B) \neq 0$ .  
 $\triangleright P(A|B) = \frac{P(A)P(B|A)}{P(B)}$ .
- Independence:  $A$  and  $B$  are *independent* if and only if  $P(A \cap B) = P(A)P(B)$ . We denote this by  $A \perp B$ .  
 $\triangleright$  If  $P(A) \neq 0$ ,  $A \perp B$  if and only if  $P(B|A) = P(B)$ .
- Law of total probability: Suppose  $A_1, A_2, \dots, A_n$  is a partition of  $S$ . Then,  $P(B) = \sum_{i=1}^n P(B \cap A_i) = \sum_{i=1}^n P(A_i)P(B|A_i)$ .
- Bayes' theorem: Suppose  $A_1, A_2, \dots, A_n$  is a partition of  $S$ . Then,  $P(A_k|B) = \frac{P(A_k)P(B|A_k)}{\sum_{i=1}^n P(A_i)P(B|A_i)}$ .

### 2 Random Variables

**Random variable:** A random variable  $X : S \rightarrow \mathbb{R}$  assigns a real number to every  $s \in S$ .

- Range space:  $R_X = \{x \mid x = X(s), s \in S\}$ . Either finite or countable.
- Discrete r.v.:  $R_X = \{x_1, x_2, x_3, \dots\}$ .  
 $\triangleright$  Probability (mass) function:  $f(x) = \begin{cases} P(X=x) & \text{for } x \in R_X; \\ 0 & \text{for } x \notin R_X. \end{cases}$   
 $\triangleright$  Probability distribution: Collection of pairs  $(x_i, f(x_i))$ .  
 $\triangleright$  Properties of p.m.f.:
  - $f(x_i) \geq 0$  for all  $x_i \in R_X$ ;
  - $f(x_i) = 0$  for all  $x_i \notin R_X$ ;
  - $\sum_{x_i \in R_X} f(x_i) = 1$ .
  - For any  $B \subset \mathbb{R}$ ,  $P(X \in B) = \sum_{x_i \in B \cap R_X} f(x_i)$ .
- Continuous r.v.:  
 $\triangleright$  Probability density function:
  - $f(x) \geq 0$  for all  $x \in R_X$ ; and  $f(x) = 0$  for all  $x \notin R_X$ ;
  - $\int_{R_X} f(x) dx = 1$ ;
  - $P(a < X \leq b) = \int_a^b f(x) dx$ .
- Cumulative distribution function:  $F(x) = P(X \leq x)$ .  
 $\triangleright P(a \leq X \leq b) = P(X \leq b) - P(X < a) = F(b) - F(a)$ .  
 $\triangleright$  Continuous:  $P(a \leq X \leq b) = P(a < X < b) = F(b) - F(a)$ .  
 $\triangleright$  Right continuous:  $F(a) = \lim_{x \rightarrow a^+} F(x)$ .  
 $\triangleright$  Convergence to 0 and 1 in the limits:  $\lim_{x \rightarrow -\infty} F(x) = 0$ ;  
 $\lim_{x \rightarrow +\infty} F(x) = 1$ .

**Expectation:**

- Discrete:  $\mathbb{E}[X] = \mu_X = \sum_{x_i \in R_X} x_i f(x_i)$ .
- Continuous:  $\mathbb{E}[X] = \mu_X = \int_{-\infty}^{\infty} x f(x) dx = \int_{x \in R_X} x f(x) dx$ .
- $\mathbb{E}[aX + b] = a\mathbb{E}[X] + b$ .
- $\mathbb{E}[X + Y] = \mathbb{E}[X] + \mathbb{E}[Y]$ .
- Discrete:  $\mathbb{E}[g(X)] = \sum_{x \in R_X} g(x)f(x)$ ; continuous:  $\mathbb{E}[g(X)] = \int_{R_X} g(x)f(x) dx$ .

**Variance:**  $\sigma_X^2 = V(X) = \mathbb{E}[(X - \mu_X)^2]$ .

- Discrete:  $V(X) = \sum_{x \in R_X} (x - \mu_X)^2 f(x)$ ;
- Continuous:  $V(X) = \int_{R_X} (x - \mu_X)^2 f(x) dx$ .
- $V(aX + b) = a^2 V(X)$ .
- $V(X) = \mathbb{E}[X^2] - \mathbb{E}[X]^2$ .
- Standard deviation:  $\sigma_X = \sqrt{V(X)}$ .

### 3 Special Probability Distributions

#### Discrete Distributions

**Discrete uniform distribution**

P.m.f.:  $f_X(x) = \begin{cases} \frac{1}{k}, & x = x_1, x_2, \dots, x_k; \\ 0, & \text{otherwise.} \end{cases}$   
 $\mathbb{E}[X] = \frac{1}{k} \sum_{i=1}^k x_i$      $V[X] = \sigma_X^2 = \frac{1}{k} \sum_{i=1}^k x_i^2 - \mu_X^2$ .

**Bernoulli Distribution:**  $X \sim \text{Bern}(p), 0 \leq p \leq 1$ .

P.m.f.:  $f_X(x) = P(X=x) = \begin{cases} p, & x=1 \\ 1-p, & x=0 \end{cases} = p^x(1-p)^{1-x}$   
 $\mu_X = \mathbb{E}[X] = p$   
 $\sigma_X^2 = V(X) = p(1-p) = pq$

**Binomial Distribution:** No. of successes in  $n$  trials of Bernoulli processes;  $X \sim \text{Bin}(n, p)$ .

$f_X(x) = P(X=x) = \binom{n}{x} p^x (1-p)^{n-x}$  for  $x = 0, 1, \dots, n$ .  
 $\mu_X = \mathbb{E}[X] = np$ ;  $\sigma_X^2 = V(X) = np(1-p)$ .

**Negative binomial distribution:** No. of i.i.d. Bernoulli trials until first  $k$  successes.

$f_X(x) = P(X=x) = \binom{x-1}{k-1} p^k (1-p)^{x-k}$  for  $x = k, k+1, \dots$ .  
 $\mathbb{E}[X] = \frac{k}{p}$      $V(X) = \frac{(1-p)k}{p^2}$ .

**Geometric distribution:** No. of i.i.d. Bernoulli trials until first success.

$f_X(x) = P(X=x) = p(1-p)^{x-1}$ .  
 $\mathbb{E}[X] = \frac{1}{p}$      $V(X) = \frac{1-p}{p^2}$ .

**Poisson distribution:** No. of events occurring in a fixed period of time/fixed region;  $X \sim \text{Poisson}(\lambda)$ .

$\mathbb{E}[X] = \lambda$      $V(X) = \lambda$ .

Poisson process: continuous-time process with Poisson( $\lambda$ ) rate  $\lambda$ .

Poisson approximation to Binomial: Let  $X \sim (n, p)$ . Suppose  $n \rightarrow \infty, p \rightarrow 0$  s.t.  $np$  is constant,  $\lim_{p \rightarrow 0, n \rightarrow \infty} P(X=x) = \frac{e^{-np}(np)^x}{x!}$ .

- Good when  $n \geq 20; p \leq 0.05$  or  $n \geq 100; np \leq 10$ .

#### Continuous Distributions

**Continuous uniform distribution:**  $X \sim U(a, b)$ .

$f_X(x) = \begin{cases} \frac{1}{b-a}, & a \leq x \leq b \\ 0, & \text{otherwise} \end{cases}$   
 $\mathbb{E}[X] = \frac{a+b}{2}$      $V(X) = \frac{(b-a)^2}{12}$ .  
 c.d.f.:  $F_X(x) = \begin{cases} 0, & x < a \\ \frac{x-a}{b-a}, & a \leq x \leq b \\ 1, & x > b. \end{cases}$

