



Midterm Review



Midterm Exam Rules

- **Zoom meeting: go to your TA section**
- April 29, Thursday 4 pm
- 110 minutes, starts promptly
- Calculators
 - No cell phones, computers, pagers
- Reference sheet:
 - Letter size, 1-side handwritten only
 - no problems or examples
- Turn in both reference sheet & exam: No submission of reference sheet imposes a penalty



Midterm Exam Rules

- Write and sign Honor Code
- Show your desk area using webcam in the beginning of the exam
- No speaking nor leaving the desk before completing exam/submission
- You can ask questions using chat window
- Submit both solutions and reference sheet
- Early submission is fine. Leave the zoom meeting after the submission is complete and TA's confirmation
- Internet disconnection: record locally until rejoining the zoom meeting and submit it with time stamp after the exam*
- Proctoring is recorded incl the submission process



Midterm Review

- Descriptive Statistics
- Probability reasoning questions
 - Unions and Intersections, Complements
 - Independence
 - Law of Total probability
- Conditional Probability and Bayes' Theorem
- Discrete Random Variables and Probability Distribution
 - Binomial Distribution



Measure of Location

Mean

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

Median

- Order the n data points from smallest to largest
 - the number in position $\frac{n+1}{2}$ if n is odd
 - the average of the numbers in positions $\frac{n}{2}$ and $\frac{n}{2} + 1$ if n is even

Mode

- The value that has the highest frequency



Measure of Variability

Sample Variance

$$\begin{aligned}\text{Var}[x] &= s^2 \\ &= \frac{\sum (x_i - \bar{x})^2}{n-1} \\ &= \frac{S_{xx}}{n-1}\end{aligned}$$

Sample Standard Deviation

$$s = \sqrt{s^2}$$



Quartiles and Percentiles

Quartiles

- Lower fourth (lower quartile)
- Upper fourth (upper quartile)
- Fourth spread (fs)

$$f_s = \text{IQR} = \text{upper fourth} - \text{lower fourth}$$

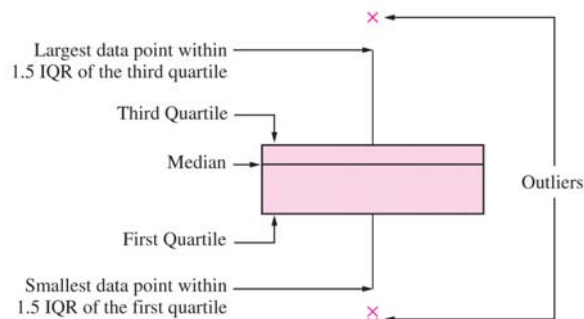
Percentiles

- The 25th percentile = Lower quartile (Q1)
- The 50th percentile = Median (Q2)
- The 75th percentile = Upper quartile (Q3)



Boxplots

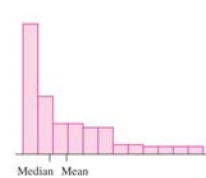
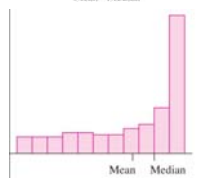
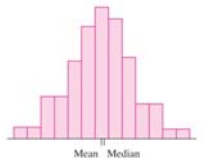
- A boxplot is a graphic that presents the median, the first and third quartiles, and any outliers present in the sample



(Navidi, 2010)



Histogram



(Khan Academy, 2019)

- A histogram is symmetric if its right half is a mirror image of its left half
 - Mean \cong Median
- Histograms that are not symmetric are referred to as skewed
 - skewed to the left, or negatively skewed
 - a histogram with a long right-hand tail
 - the mean $<$ the median
 - skewed to the right, or positively skewed
 - A histogram with a long left-hand tail
 - the mean $>$ the median



Intersection, Union, Complement

- Intersection : $A \cap B$
- Union : $A \cup B$
$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$
- Complement \bar{A}
$$P(\bar{A}) = 1 - P(A)$$
- $(A \cup C) \cap (B \cup C) = (A \cap B) \cup C$
$$\overline{A \cup B} = \bar{A} \cap \bar{B}$$
- Venn Diagram



Axioms of Probability

- For any event A , $0 \leq P(A) \leq 1$
- Axiom 3: $P(S) = 1$
- If A and B are *mutually exclusive* then
$$A \cap B = \phi, \quad P(A \cup B) = P(A) + P(B)$$
- If $A_1, A_2, A_3, \dots, A_k$ is a finite collection of mutually exclusive events,
$$P(A_1 \cup A_2 \cup \dots \cup A_k) = \sum_{i=1}^k P(A_i)$$
- If A_1 and A_2 are *collectively exhaustive*
$$A \cup B = S$$



Permutations, Combinations

- Permutations
 - Selection of choices that can't be repeated
 - Chosen without replacement
 - Ordered matters
$$P_{k,n} = n \times (n-1) \times (n-2) \times \dots \times (n-k+1) = \frac{n!}{(n-k)!}$$
- Combinations
 - Similar to permutations, but order does not matter (dealing cards)
$$C_{k,n} = \binom{n}{k} = \frac{P_{k,n}}{k!} = \frac{n!}{k!(n-k)!}$$
 - $P_{k,n} = k! \times C_{k,n}$



Conditional Probability

- $P(A|B)$ denotes the probability of event A occurring given event B occurs

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

- Tree diagram



Multiplication Rule

- The probability of A and B = $P(A \cap B)$
 - $P(A \cap B) = P(A) P(B|A)$
 - $P(A \cap B) = P(B) P(A|B)$
- Independence
 - A and B are independent if $P(A|B) = P(A)$
 - This is true if and only if $P(B|A) = P(B)$
 - If A and B are independent, then $P(A \cap B) = P(A)P(B)$



Law of Total Probability

- Let A_1, \dots, A_k be mutually exclusive and exhaustive events. Then for any other event B ,

$$P(B) = P(B|A_1)P(A_1) + \dots + P(B|A_k)P(A_k) \\ = \sum_{i=1}^k P(B|A_i)P(A_i)$$

- Or otherwise restated

$$P(B) = \sum_{i=1}^k P(A_i \cap B) \\ = \sum_{i=1}^k P(B|A_i)P(A_i)$$



Bayes' Theorem

$$P(B | A) = \frac{P(A | B)P(B)}{P(A)} \\ = \frac{P(A | B_i)P(B_i)}{P(A | B_1)P(B_1) + P(A | B_2)P(B_2) + \dots P(A | B_N)P(B_N)}$$



Probability Mass Function

- A Probability Mass Function (pmf), also called *probability distribution*, is a function $p(x)$ that assigns to each possible value x that the random variable X can take, its probability

$$\begin{aligned} p(x) &= P(X = x) \\ &= P(\text{all } s \in S : X(s) = x) \end{aligned}$$

- $p(x_i) \geq 0$ for each possible value x_i of X
- $\sum_{\text{all } x_i} p(x_i) = 1$



Cumulative Distribution Function

- The cumulative distribution function (cdf) of a r.v. X is

$$F(x) = P(X \leq x) = \sum_{x_i \leq x} P(X = x_i)$$

which gives the sum of the probabilities up to that value x

- Any probability distribution must follow the axioms of probability
 - $F(-\infty)=0$; $F(\infty) = 1$
 - $F(x) \geq 0$ and is weakly increasing
 - It is continuous in x



Expected Value

- A random variable has probabilities p associated with outcomes x with set of possible values D and pmf $p(x)$

$$E[X] = \mu_x = \sum_{x \in D} x \cdot p(x)$$

$$E[h(X)] = \sum_D h(x) \cdot p(x)$$

- Properties of expected values

$$E(X_1 + X_2) = E[X_1] + E[X_2]$$

$$E(aX + b) = a \times E[X] + b$$



Variance and Standard Deviation

- Variance

$$\begin{aligned} V(X) = \sigma^2 &= E[(X - \mu)^2] = \sum_D (x - \mu)^2 \cdot p(x) \\ &= E[X^2] - (E[X])^2 \end{aligned}$$

- Standard deviation

$$\sigma = \sqrt{\sigma^2}$$

- Properties

$$\sigma^2(aX + b) = a^2 \sigma^2(X)$$

$$\sigma(aX + b) = |a| \sigma(X)$$



Bernoulli Distribution

- Bernoulli Probability Distribution

x	0	1
p(x)	1-p	p

$$\mu = E[X] = p$$

$$\sigma^2 = \text{Var}[X] = p(1-p)$$



Binomial Distributions Theorem

- $X \sim \text{Bin}(n, p)$ $\binom{n}{x} = \frac{n!}{x!(n-x)!}$

- Cumulative Distribution Function

$$\sum_{i=0}^x \binom{n}{i} p^i (1-p)^{n-i}$$

- Expected values and variance

$$E[X] = np$$

$$\text{Var}[X] = np(1-p)$$



Hypergeometric Distribution

- If X is the number of S 's in a completely random sample of size n drawn from a population consisting of M S 's and $(N-M)$ F 's,

$$P(X = x) = h(x; n, M, N) = \frac{\binom{M}{x} \binom{N-M}{n-x}}{\binom{N}{n}}$$

$$E(X) = n \frac{M}{N}$$

$$V(X) = \left(\frac{N-n}{N-1}\right) n \frac{M}{N} \left(1 - \frac{M}{N}\right)$$



Negative Binomial Distribution

- The probability that takes $X=x$ failures to get r successes, with probability of success p is:

$$nb(x; r, p) = \binom{x+r-1}{r-1} p^r (1-p)^x, \quad x = 0, 1, 2, \dots$$

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

$$V(X) = \frac{r(1-p)}{p^2}$$