

Key Equations

Matrix Operations

Transpose

$$A_{m \times n}^t = B_{n \times m}, \text{ where } b_{ij} = a_{ji}$$

Scalar Multiplication

$$kA = C, \text{ where } c_{ij} = ka_{ij}$$

Addition

$$A + D = E, \text{ where } e_{ij} = a_{ij} + d_{ij}$$

Multiplication

$$A_{m \times n} F_{n \times p} = G_{m \times p}, \text{ where } g_{ij} = \sum_{k=1}^n a_{ik} f_{kj}$$

Inverse

$$\text{For } H = \begin{bmatrix} a & b \\ c & d \end{bmatrix}, H^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \text{ if } ad \neq bc$$

Statistics of One Variable

Population

$$\text{Mean: } \mu = \frac{\sum x}{N}$$

Sample

$$\bar{x} = \frac{\sum x}{n}$$

Weighted Mean

$$\bar{x}_w = \frac{\sum w_i x_i}{\sum w_i}$$

$$\text{Variance: } \sigma^2 = \frac{\sum(x - \mu)^2}{N}$$

$$s^2 = \frac{\sum(x - \bar{x})^2}{n - 1}$$

$$\text{Standard Deviation: } \sigma = \sqrt{\frac{\sum(x - \mu)^2}{N}}$$

$$s = \sqrt{\frac{\sum(x - \bar{x})^2}{n - 1}} = \sqrt{\frac{n(\sum x^2) - (\sum x)^2}{n(n-1)}} = \sqrt{\frac{n\sum x^2 - n(\sum x)^2}{n-1}}$$

$$\text{Z-score: } z = \frac{x - \mu}{\sigma}$$

$$z = \frac{x - \bar{x}}{s}$$

$$\text{Grouped Data: } \mu \doteq \frac{\sum f_i m_i}{\sum f_i}$$

$$\bar{x} \doteq \frac{\sum f_i m_i}{\sum f_i}, \text{ where } m_i \text{ is midpoint of } i\text{th interval}$$

$$\sigma \doteq \sqrt{\frac{\sum f_i(m_i - \mu)^2}{N}}$$

$$s \doteq \sqrt{\frac{\sum f_i(m_i - \bar{x})^2}{n - 1}}$$

Statistics of Two Variables

Correlation Coefficient

$$r = \frac{s_{XY}}{s_X s_Y}$$

$$= \frac{n \sum xy - \sum x \sum y}{\sqrt{n \sum x^2 - (\sum x)^2} \sqrt{n \sum y^2 - (\sum y)^2}}$$

Least Squares Line of Best Fit

$$y = ax + b, \text{ where } a = \frac{n \sum xy - \sum x \sum y}{n \sum x^2 - (\sum x)^2} \text{ and } b = \bar{y} - a \bar{x}$$

Coefficient of Determination

$$r^2 = \frac{\sum(y_{est} - \bar{y})^2}{\sum(y - \bar{y})^2}$$

Permutations and Organized Counting

Factorial: $n! = n \times (n - 1) \times (n - 2) \times \dots \times 3 \times 2 \times 1$

Permutations

$$r \text{ objects from } n \text{ different objects: } {}_n P_r = \frac{n!}{(n - r)!}$$

$$n \text{ objects with some alike: } \frac{n!}{a!b!c!...}$$

Combinations and the Binomial Theorem

Combinations

r items chosen from n different items: ${}_n C_r = \frac{n!}{(n-r)!r!}$

at least one item chosen from n distinct items: $2^n - 1$

at least one item chosen from several different sets of identical items: $(p+1)(q+1)(r+1)\dots-1$

Pascal's Formula: ${}_n C_r = {}_{n-1} C_{r-1} + {}_{n-1} C_r$

Binomial Theorem: $(a+b)^n = \sum_{r=0}^n {}_n C_r a^{n-r} b^r$

Introduction to Probability

Equally Likely Outcomes: $P(A) = \frac{n(A)}{n(S)}$

Complement of A : $P(A') = 1 - P(A)$

Odds: odds in favour of $A = \frac{P(A)}{P(A')}$

If odds in favour of $A = \frac{h}{k}$, $P(A) = \frac{h}{h+k}$

Conditional Probability: $P(A|B) = \frac{P(A \text{ and } B)}{P(B)}$

Independent Events: $P(A \text{ and } B) = P(A) \times P(B)$

Dependent Events: $P(A \text{ and } B) = P(A) \times P(B|A)$

Mutually Exclusive Events: $P(A \text{ or } B) = P(A) + P(B)$

Non-Mutually Exclusive Events: $P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$

Markov Steady State: $S^{(n)} = S^{(n)}P$

Discrete Probability Distributions

Expectation: $E(x) = \sum_{i=1}^n x_i P(x_i)$

Discrete Uniform Distribution: $P(x) = \frac{1}{n}$

Binomial Distribution: $P(x) = {}_n C_x p^x q^{n-x}$

$E(x) = np$

Geometric Distribution: $P(x) = q^x p$

$E(x) = \frac{q}{p}$

Hypergeometric Distribution:

$$P(x) = \frac{{C}_r x {C}_{n-r}}{C_n}$$

$$E(x) = \frac{r}{n}$$

$$P(x) = \frac{C(k, x) \times C(N-k, n-x)}{C(N, n)}$$

Continuous Probability Distributions

Exponential Distribution: $y = ke^{-kx}$, where $k = \frac{1}{\mu}$

$$E(x) = \mu$$

Normal Distribution: $y = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2}(\frac{x-\mu}{\sigma})^2}$

Normal Approximation to Binomial Distribution: $\mu = np$ and $\sigma = \sqrt{npq}$ if $np > 5$ and $nq > 5$

Distribution of Sample Means: $\mu_{\bar{x}} = \mu$ and $\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$

Confidence Intervals: $\bar{x} - z_{\frac{\alpha}{2}} \frac{\sigma}{\sqrt{n}} < \mu < \bar{x} + z_{\frac{\alpha}{2}} \frac{\sigma}{\sqrt{n}}$

$$\hat{p} - z_{\frac{\alpha}{2}} \frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}} < p < \hat{p} + z_{\frac{\alpha}{2}} \frac{\sqrt{\hat{p}(1-\hat{p})}}{\sqrt{n}}$$