

MATH 323: Probability

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Contents

1 Formulas

Mean & Variance

$$\text{General} \left| \begin{array}{l} \text{Mean } (\mu) \\ \bar{y} = \frac{1}{n} \sum_{i=1}^n y_i \end{array} \right.$$

$$\left| \begin{array}{l} \text{Variance } (\sigma^2) \\ \frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})^2 \end{array} \right.$$

Expected Value & Variance

General

Probability

$$E(Y)$$

$$V(Y)$$

$$M_x(t)$$

$$p(y)$$

$$\mu = \sum_y yp(y)$$

$$\sigma^2 = E[(Y - \mu)^2] = E(Y^2) - \mu^2$$

$$E(e^{tx})$$

Bernoulli

Realization

$$P_x(x)$$

$$M_x(t)$$

$$E(X)$$

$$V(X)$$

$$x = 0, 1$$

$$p^x(1-p)^{1-x} = p^x q^{1-x}$$

$$(1-p) + pe^t$$

$$\sum_{x=0}^1 xp^x q^{1-x} = p$$

$$E(X^2) - E(X)^2 = p(1-p)$$

Binomial

Realization

$$P_x(x)$$

$$M_x(t)$$

$$E(X)$$

$$V(X)$$

$$x = 0, 1, \dots, n$$

$$C_x^n p^x (1-p)^{n-x}$$

$$(pe^t + 1 - p)^n$$

$$np$$

$$np(1-p)$$

Geometric

Realization

$$P_x(x)$$

$$M_x(t)$$

$$E(X)$$

$$V(X)$$

$$x = 1, 2, \dots$$

$$p(1-p)^{x-1}$$

$$\frac{pe^t}{1-(1-p)e^t}$$

$$\frac{1}{p}$$

$$\frac{1-p}{p^2}$$

Negative Binomial

Realization

$$P_x(x)$$

$$M_x(t)$$

$$E(X)$$

$$V(X)$$

$$x = r, r+1, \dots$$

$$C_{r-1}^{x-1} p^r q^{x-r}$$

$$\left[\frac{pe^t}{1-(1-p)e^t} \right]^r$$

$$r \cdot \frac{1}{p}$$

$$r \cdot \frac{1-p}{p^2}$$

Poisson

Realization

$$x = 0, 1, 2, \dots$$

$$P_x(x)$$

$$\frac{e^{-\lambda} \cdot \lambda^x}{x!}$$

$$M_x(t)$$

$$e^{\lambda(e^t - 1)}$$

$$E(X)$$

$$\lambda$$

$$V(X)$$

$$\lambda$$

Exponential Distribution

Realization

$$X \sim \text{Exp}(\lambda), x > 0$$

$$P_x(x)$$

$$\lambda e^{-\lambda x}$$

$$M_x(t)$$

$$\frac{\lambda}{\lambda - t}$$

$$E(X)$$

$$\frac{1}{\lambda}$$

$$V(X)$$

$$\frac{1}{\lambda^2}$$

Uniform Distribution

Realization

$$a < x < b$$

$$P_x(x)$$

$$\frac{1}{b-a} = c$$

$$M_x(t)$$

$$\frac{e^{tb} - e^{ta}}{t(b-a)}$$

$$E(X)$$

$$\frac{a+b}{2}$$

$$V(X)$$

$$\frac{(b-a)^2}{12}$$

Gamma

Realization

$$x > 0$$

$$P_x(x)$$

$$\frac{\lambda^\alpha x^{\alpha-1} e^{-\lambda x}}{\Gamma(\alpha)}$$

$$M_x(t)$$

$$\frac{\Gamma(\alpha)}{\beta^\alpha} = \int_0^\infty x^{\alpha-1} e^{-\beta x} dx, \quad \Gamma(\alpha) = (\alpha - 1)!$$

$$E(X)$$

$$\left(\frac{\lambda}{\lambda - t}\right)^\alpha$$

$$V(X)$$

$$\alpha\beta$$

$$\alpha\beta^2$$

Beta

Realization

$$0 < x < 1$$

$$P_x(x)$$

$$\frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1}$$

$$\alpha, \beta > 0 \quad B(\alpha, \beta) = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)}$$

$$M_x(t)$$

$$E(X)$$

$$\frac{\alpha}{\alpha+\beta}$$

$$V(X)$$

$$\frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$$

Normal

Realization

 $P_x(x)$ $E(X)$ $V(X)$ Standard z

$$-\infty < x < \infty$$

$$\frac{1}{\sigma\sqrt{2\pi}} e^{-(x-\mu)^2/(2\sigma^2)}$$

$$\mu$$

$$\sigma^2$$

$$\frac{x-\mu}{\sigma}$$

Chi-square is a beta distribution where $\alpha = \frac{v}{2}$ for some v , and $\beta = 2$ ($\lambda = \frac{1}{2}$). This makes $\mu = v$ and $\sigma^2 = 2v$

Combination & Permutation

- $P_r^n = \frac{n!}{(n-r)!}$
- $\binom{n}{r} = C_r^n = \frac{n!}{r!(n-r)!}$
- Permutation of n objects of k kinds, where n_i is the number of times type k occurs, is $\frac{n!}{n_1!n_2!\dots n_k!}$
- Circular permutations (round table) with n items taken r at a time = $\frac{P_r^n}{r}$

PMF & CDF

PMF: Probability Mass Function

Discrete r.v.: $P(x) = P(X = x)$

- $0 \leq P(x) \leq 1, \forall x$
- $\sum_x P(x) = 1$

CDF: Cumulative distribution function: $F_x(x) = P(X \leq x), x \in \mathbb{R}$

- $0 \leq F(x) \leq 1$
- $\lim_{x \rightarrow -\infty} F(x) = 0, \lim_{x \rightarrow \infty} F(x) = 1$
- Non-decreasing: $x_1 \leq x_2 \Rightarrow F(x_1) \leq F(x_2)$
- Right continuous

Moment Generating Function

Given $m(t) = E(e^{tY})$, $E(Y^k) = m^{(k)}(0) = \mu'_k$, $\mu'_1 = \mu$.

In finding derivatives, we can also find it against $\log(m_x(t))$, as $m(0) = 1$

For simplification, note that $\sum_0^\infty \frac{f^t}{y!} = e^f$

Properties

- $M_{x+a}(t) = E[e^{(x+a)t}] = e^{at}M_x(t)$
- $M_{bx}(t) = E[e^{(bx)t}] = M_x(bt)$
- $M_{\frac{x+a}{b}}(t) = e^{\frac{a}{b}t}M_x(\frac{t}{b})$

Inequalities

- Markov: $E(x) < \infty \Rightarrow$
 $P(x \geq c) \leq \frac{E(x)}{c}, \forall c > 0$
- Tchebysheff: $E(x), Var(x) < \infty \Rightarrow$
 $P(|x - \mu| \geq k\sigma) \leq \frac{1}{k^2}$ or
 $P(|x - \mu| < k\sigma) \geq 1 - \frac{1}{k^2}$
- Chernoff: $M_x(t)$ is a mgf of r.v. x , and $c > 0 \Rightarrow$
 $P(x \geq c) \leq e^{-tc} M_x(t) \forall t > 0$
 $P(x \leq c) \leq e^{-tc} M_x(t) \forall t < 0$

Misc

- $P(A | B) = \frac{A \cap B}{P(B)}$ if $P(B) > 0$
- Skewness

$$\frac{\mu^3}{\sigma^3} \begin{cases} > 0 & \text{right} \\ = 0 & \text{symmetric} \\ < 0 & \text{left} \end{cases} \quad (1)$$

- Kurtosis

$$\frac{\mu^4}{\sigma^4} - 3 \begin{cases} > 0 & \text{peaked} \\ = 0 & \text{normal} \\ < 0 & \text{flat} \end{cases} \quad (2)$$

- Jensen: $f(x)$ is a convex function if:
 $a \cdot f(x_1) + (1 - a) \cdot f(x_2) \geq f(ax_1 + (1 - a)x_2)$, where $0 \leq a \leq 1$
This also means that $E(f(x)) \geq f(E(x))$

2 Theories & Definitions

Kolmogorov Axioms

$\forall A \in \Sigma:$

m!

1. $P(A) \geq 0$ (non-negative)
2. $P(\Sigma) = 1$ (normed)

$$3. P\left(\bigcup_{j=1}^{\infty} A_j\right) = \sum_{j=1}^{\infty} P(A_j) \text{ when } A_i \cap A_j = \emptyset \quad \forall i \neq j \text{ (linearly-additive)}$$

De-Morgan's Theorem

m!

$$(a) (A \cap B)^C = A^C \cup B^C$$

$$(b) (A \cup B)^C = A^C \cap B^C$$

Partition

Let $\{A_i\}_{i=1}^{\infty(k)}$ be a sequence of sets $A_i \subseteq \Omega, \forall i$. If:

m!

$$i. A_i \cap A_j = \emptyset, \forall i \neq j$$

$$ii. \bigcup_{i=1}^{\infty(k)} A_i = \Sigma$$

then we say $\{A_i\}_{i=1}^{\infty(k)}$ is a partition of sample space Ω : $\Omega = \bigcup_{j=1}^k A_j$

Inequalities (Boole, Bonferroni)

m!

A. Boole's Inequality

Let $A_1, A_2, \dots \in \mathcal{F}$

$$\Rightarrow P\left(\bigcup_{i=1}^{\infty} A_i\right) \leq \sum_{i=1}^{\infty} P(A_i)$$

B. Bonferroni's Inequality

$$P\left(\bigcap_{i=1}^{\infty} A_i\right) \geq 1 - \sum_{i=1}^{\infty} P(A_i^C)$$

Baye's Rule

Assume $\{A_1, A_2, \dots, A_k\}$ is a partition of Ω such that $P(A_i) > 0 \forall i$

For $B \subseteq \Omega$:

C. Law of Total Probability

$$\begin{aligned}
 P(B) &= P\left(\bigcup_{i=1}^k (A_i \cap B)\right) \\
 &= \sum_{i=1}^k P(A_i \cap B) \\
 &= \sum_{i=1}^k P(B | A_i)P(A_i)
 \end{aligned} \tag{3}$$

D.

$$\begin{aligned}
 P(A_i | B) &= \frac{P(A_i \cap B)}{P(B)} \\
 &= \frac{P(B | A_i)P(A_i)}{\sum_{i=1}^k P(B | A_i)P(A_i)}
 \end{aligned} \tag{4}$$

3 Misc

- If $\{A_i\}_{i=1}^{\infty}$ is non-decreasing, then $\lim_{n \rightarrow \infty} A_n = \bigcup_{i=1}^{\infty} A_i$
- If $\{A_i\}_{i=1}^{\infty}$ is non-increasing, then $\lim_{n \rightarrow \infty} A_n = \bigcap_{i=1}^{\infty} A_i$

Independence

A and B are independent ($A \perp\!\!\!\perp B$) if any of the following holds:

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

Note that \emptyset is independent to any event, and that

$$A \perp\!\!\!\perp B \Leftrightarrow A^C \perp\!\!\!\perp B \Leftrightarrow A^C \perp\!\!\!\perp B^C \Leftrightarrow A \perp\!\!\!\perp B^C$$

4 Questions

- E. How many ways can 4 married couples seat themselves around a circular table if no couple can sit next to each other?
- F. What is the probability of getting a full house? (3 cards of some rank and 2 cards of another rank)
- G. Find the support and realization of X , where X is the number of heads, and we are tossing a fair coin three times.

5 Answers

H. $\frac{31}{105}$

Compute probabilities that i couple(s) sit next to each other for i in 1..4 and find complement of sum.

I. 0.00144

Basic combination probability:

$$\frac{(C_1^{13} \cdot C_3^4) \cdot (C_1^{12} \cdot C_2^4)}{C_5^{52}}$$

- J.
- Realization = number of occurrences per sample event. eg, HHH has realization 3, and THT has realization 1
 - Support = range of x . Here, it is 0 to 3
 - Table of x to $P(x)$ showing the probability distribution of X:

x	$P(x)$
0	$\frac{1}{8}$
1	$\frac{3}{8}$
2	$\frac{3}{8}$
3	$\frac{1}{8}$