

ECE 30200 - Probabilistic Methods in Electrical and Computer Engineering

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Background

The following formulas will be instrumental and may be familiar.

Series

$$\sum_{k=0}^n r^k = \frac{1 - r^{n+1}}{1 - r} \quad (1)$$

$$\sum_{n=1}^{\infty} \frac{1}{n^2} = \frac{\pi^2}{6} \quad (2)$$

$$\sum_{k=1}^{\infty} k r^{k-1} = \frac{1}{(1 - r)^2} \quad (3)$$

Combinatorics

$$\binom{n}{k} = \frac{n!}{k!(n - k)!} \quad (4)$$

$$(a + b)^n = \sum_{k=0}^n \binom{n}{k} a^{n-k} b^k \quad (5)$$

$$\binom{n}{k} + \binom{n}{k-1} = \binom{n+1}{k} \quad (6)$$

$$P(n, k) = \frac{n!}{(n - k)!} \quad (7)$$

where $P(n, k)$ is the number of ways to arrange k objects out of n (permutations).

$$C(n, k) = \binom{n}{k} = \frac{n!}{k!(n - k)!} \quad (8)$$

where $C(n, k)$ is the number of ways to choose k objects out of n (combinations).

Approximations

$$f(x) = f(a) + f'(a)(x - a) + \frac{f''(a)}{2!}(x - a)^2 + \dots \quad (9)$$

$$= \sum_{n=0}^{\infty} \frac{f^{(n)}(a)}{n!} (x - a)^n \quad (10)$$

$$1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \dots = \sum_{k=0}^{\infty} \frac{x^k}{k!} \quad (11)$$

$$= e^x \quad (12)$$

$$\sin(x) = x - \frac{x^3}{3!} + \frac{x^5}{5!} - \frac{x^7}{7!} + \dots \quad (13)$$

$$= \sum_{n=0}^{\infty} (-1)^n \frac{x^{2n+1}}{(2n+1)!} \quad (14)$$

$$\cos(x) = 1 - \frac{x^2}{2!} + \frac{x^4}{4!} - \frac{x^6}{6!} + \dots \quad (15)$$

$$= \sum_{n=0}^{\infty} (-1)^n \frac{x^{2n}}{(2n)!} \quad (16)$$

$$\ln(1+x) = x - \frac{x^2}{2} + \frac{x^3}{3} - \frac{x^4}{4} + \dots \quad (17)$$

$$= \sum_{n=1}^{\infty} (-1)^{n+1} \frac{x^n}{n} \quad (18)$$

Calculus

$$\frac{d}{dx} \int_a^x f(t) dt = f(x) \quad (19)$$

$$\int_a^b f'(x) dx = f(b) - f(a) \quad (20)$$

$$\int f(g(x))g'(x) dx = \int f(u) du \quad (21)$$

$$\int u dv = uv - \int v du \quad (22)$$

$$\int \frac{1}{(x-a)(x-b)} dx = \frac{1}{b-a} \ln \left| \frac{x-a}{x-b} \right| + C \quad (23)$$

Linear Algebra

$$\vec{y} = \beta_1 \vec{x}_1 + \beta_2 \vec{x}_2 + \dots + \beta_N \vec{x}_N \quad (24)$$

$$\langle \vec{a}, \vec{b} \rangle = \vec{a} \vec{b}^T \quad (25)$$

$$= \sum_{i=1}^n a_i b_i \quad (26)$$

where $\langle \vec{a}, \vec{b} \rangle$ denotes the inner product of vectors \vec{a} and \vec{b} .

$$\|\vec{x}\|_p = \left(\sum_{i=1}^n |x_i|^p \right)^{1/p} \quad (27)$$

where $\|\vec{x}\|_p$ is the p -norm (or ℓ_p -norm) of vector \vec{x} .

$$\cos(\theta) = \frac{\langle \vec{a}, \vec{b} \rangle}{\|\vec{a}\|_2 \|\vec{b}\|_2} \quad (28)$$

where θ is the angle between vectors \vec{a} and \vec{b} .

$$\hat{\beta} = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \vec{y} \quad (29)$$

where $\hat{\beta}$ is the vector of least squares coefficients, \mathbf{X} is the data matrix, and \vec{y} is the target vector

Probability Laws

A probability law must satisfy three axioms:

1. Non-negativity: $P(A) \geq 0 \forall A \in F$
2. Normalization: $P(\Omega) = 1$
3. Additivity: For any disjoint subsets $\{A_1, A_2, \dots\}$, it holds that

$$P \left[\bigcup_{n=1}^{\infty} A_n \right] = \sum_{n=1}^{\infty} P[A_n]$$

Probability Properties

$$P[A \cup B] = P[A] + P[B] - P[A \cap B] \quad (30)$$

$$P[A \cup B] \leq P[A] + P[B] \quad (31)$$

$$A \subseteq B \implies P[A] \leq P[B] \quad (32)$$

Formal Definitions

Outcomes

An *outcome* is the result of some *experiment*. If that experiment is flipping a coin, the outcome is either heads or tails. We could express the outcome of heads as H , and the outcome of tails as T . The set of all possible outcomes for an experiment is known as a sample space and is denoted by Ω . In this case $\Omega = \{H, T\}$.

Events

An *event* F is a subset of the sample space Ω . The formal definitions of probability are expressed with set notation. So the event where we have neither heads nor tails is written as $\{\}$. The event of heads could be expressed as $\{H\}$, and the event of tails could be expressed as $\{T\}$. The event of either heads or tails is $\{H, T\}$.

Probability Laws

A *probability law* is a function P that maps an event A to a real number in $[0, 1]$. For the coin example, the probability law might be $P(\{\}) = 0$, $P(\{H\}) = 0.5$, $P(\{T\}) = 0.5$, and $P(\{\Omega\}) = 1$. A probability law must satisfy three axioms:

1. Non-negativity: $P(A) \geq 0 \forall A \in F$
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$$P\left[\bigcup_{n=1}^{\infty} A_n\right] = \sum_{n=1}^{\infty} P[A_n]$$

Probability Space

A probability space is a triplet Ω, F, P .

Probability Properties

$$P[A \cup B] = P[A] + P[B] - P[A \cap B] \quad (33)$$

$$P[A \cup B] \leq P[A] + P[B] \quad (34)$$

$$A \subseteq B \implies P[A] \leq P[B] \quad (35)$$

$$P[A|B] = \frac{P[A \cap B]}{P[B]} \quad (36)$$

Outcomes are statistically *independent* if $P(A|B) = P(A)$ (assuming $P(B) > 0$), or equivalently $P(A \cap B) = P(A)P(B)$.

A *random variable* X is a function $X : \Omega \implies \Re$ that maps an outcome $\epsilon \in \Omega$ to a number $X(\epsilon)$ on the real line.

Bayes Theorem states that for any two events A and B such that $P[A] > 0$ and $P[B] > 0$,

$$P[A|B] = \frac{P[B|A]P[A]}{P[B]} \quad (37)$$

The *Law of Total Probability* states that if $\{A_1, A_2, \dots, A_n\}$ is a partition of Ω , then for any $B \subseteq \Omega$,

$$P[B] = \sum_{i=1}^n P[B|A_i]P[A_i] \quad (38)$$

*Reference**Series*

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