

MSE201 Statistics Formula Sheet

1. For the **sample space** S , the empty set ϕ , and events A and B :

$$\begin{aligned}P(S) &= 1, & P(\phi) &= 0, & P(A') &= 1 - P(A). \\P(A \cup B) &= P(A) + P(B) - P(A \cap B). \\(A \cup B)' &= A' \cap B', & (A \cap B)' &= A' \cup B'.$$

Conditional probability: $P(A | B) = \frac{P(A \cap B)}{P(B)}$ provided that $P(B) > 0$.

Multiplication rule: $P(A \cap B) = P(A | B)P(B) = P(B | A)P(A)$.

Independence: Events A and B are **independent** if $P(B | A) = P(B)$.

Bayes' rule: $P(A | B) = \frac{P(B | A)P(A)}{P(B)} = \frac{P(B | A)P(A)}{P(B | A)P(A) + P(B | A')P(A')}$.

2. A **discrete random variable** X has the **probability mass function** $\{p_x\} = \{P(X = x)\}$.

The **expectation:** $E(X) = \mu = \sum_x xp_x$.

From random sample x_1, \dots, x_n , the **sample mean** $\bar{x} = \frac{1}{n} \sum_k x_k$ estimates μ .

The **variance** $\text{var}(X) = \sigma^2 = E((X - \mu)^2) = E(X^2) - \mu^2$, where $E(X^2) = \sum_x x^2 p_x$.

The **sample variance:** $s^2 = \frac{1}{n-1} \left\{ \sum_k x_k^2 - \frac{1}{n} (\sum_k x_k)^2 \right\}$ estimates σ^2 .

The **standard deviation:** $\text{sd}(X) = \sigma = \sqrt{\text{var}(X)}$.

3. **Binomial distribution:** X is *Binomial*(n, p).

$$p_x = \binom{n}{x} p^x (1-p)^{n-x} \quad (x = 0, 1, 2, \dots, n); \quad \mu = np, \quad \sigma^2 = np(1-p).$$

Poisson distribution: X is *Poisson*(μ).

$$p_x = \frac{\mu^x e^{-\mu}}{x!} \quad (x = 0, 1, 2, \dots) \quad (\text{with } \mu > 0); \quad \mu = \sigma^2 = \mu.$$

4. For **continuous** random variables:

The **cumulative distribution function** (cdf)

$$F(x) = P(X \leq x) = \int_{x_0=-\infty}^x f(x_0) dx_0$$

The **probability density function** (pdf) $f(x) = \frac{dF(x)}{dx}$.

$$E(X) = \mu = \int_{-\infty}^{\infty} xf(x) dx, \quad E(X^2) = \int_{-\infty}^{\infty} x^2 f(x) dx, \quad \sigma^2 = E(X^2) - \mu^2.$$

5. **Uniform distribution** $Uniform(\alpha, \beta)$:

$$f(x) = \begin{cases} \frac{1}{\beta - \alpha} & (\alpha \leq x \leq \beta), & \mu = (\alpha + \beta)/2, \\ 0 & (\text{otherwise}). & \sigma^2 = (\beta - \alpha)^2/12. \end{cases}$$

Exponential distribution $Exponential(\lambda)$:

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & (x \geq 0), & \mu = 1/\lambda \\ 0 & (\text{otherwise}). & \sigma^2 = 1/\lambda^2. \end{cases}$$

Normal distribution $N(\mu, \sigma^2)$:

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x - \mu)^2}{2\sigma^2}\right) \quad (-\infty < x < \infty); \quad E(X) = \mu, \quad \text{var}(X) = \sigma^2.$$

Standard normal distribution $N(0, 1)$: If $X \sim N(\mu, \sigma^2)$, then $Z = \frac{X - \mu}{\sigma}$ is $N(0, 1)$.

For Z we write $\phi(z)$ for the pdf $f(z)$ and $\Phi(z)$ for the cdf $F(z)$.

6. For a system of k devices, which operate independently:

The **system reliability**, R is the probability of a path of operating devices.

Let $R_i = P(D_i) = P(\text{"device } i \text{ operates"})$.

A system of devices in **series** fails if any device fails.

$$R = P(D_1 \cap D_2 \cap \dots \cap D_k) = R_1 R_2 \dots R_k.$$

A system of devices in **parallel** operates if any device operates.

$$R = P(D_1 \cup D_2 \cup \dots \cup D_k) = 1 - (1 - R_1)(1 - R_2) \dots (1 - R_k).$$

7. If x_1, \dots, x_n are a random sample from $N(\mu, \sigma^2)$ and σ^2 is known, then

the $100(1 - \alpha)\%$ CI for μ is $(\bar{x} - z_{\alpha/2}\sigma/\sqrt{n}, \bar{x} + z_{\alpha/2}\sigma/\sqrt{n})$,

where $z_{\alpha/2}$ is that value such that $1 - \alpha = P(-z_{\alpha/2} < Z < z_{\alpha/2})$, where $Z \sim N(0, 1)$.

If σ^2 is estimated, then the $100(1 - \alpha)\%$ CI for μ is $(\bar{x} - t_{\alpha/2}^{n-1}s/\sqrt{n}, \bar{x} + t_{\alpha/2}^{n-1}s/\sqrt{n})$

where $t_{\alpha/2}^{n-1}$ is that value such that $1 - \alpha = P(-t_{\alpha/2}^{n-1} < T < t_{\alpha/2}^{n-1})$,

where T has a t-distribution with $n - 1$ degrees of freedom.

A **significance test** of H_0 rejects H_0 , if, assuming that H_0 is true, a test statistic is in a rejection region of its sampling distribution.