

FREEFELLOW

FORMULA SHEET

EXAM P

SOA / CAS · Probability

39

FORMULAS

7

TOPICS

[freefellow.org/soa-p/formulas](https://freefellow.org/soa-p/formulas)

PROBABILITY FUNDAMENTALS

5 items

Combinations (unordered)

$$\binom{n}{r} = \frac{n!}{r!(n-r)!}$$

$n$  objects taken  $r$  at a time, order does not matter

Addition rule (three events)

$$P(A \cup B \cup C) = P(A) + P(B) + P(C) - P(A \cap B) - P(A \cap C) - P(B \cap C) + P(A \cap B \cap C)$$

Addition rule (two events)

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

Permutations (ordered)

$${}_n P_r = \frac{n!}{(n-r)!}$$

$n$  objects taken  $r$  at a time, order matters

Set complement

$$P(A^c) = 1 - P(A)$$

CONDITIONAL PROBABILITY & BAYES

5 items

Independence of two events

$A$  and  $B$  are independent iff  $P(A \cap B) = P(A)P(B)$

(equivalently:  $P(A | B) = P(A)$ )

Conditional probability

$$P(A | B) = \frac{P(A \cap B)}{P(B)}, \quad P(B) > 0$$

Law of total probability

$$P(A) = \sum_i P(A | B_i) P(B_i)$$

where  $\{B_i\}$  is a partition of the sample space

Bayes' theorem

$$P(B_i | A) = \frac{P(A | B_i) P(B_i)}{\sum_j P(A | B_j) P(B_j)}$$

Multiplication rule

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

DISCRETE DISTRIBUTIONS

6 items

Poisson distribution — PMF, mean, variance

$$P(X = k) = \frac{e^{-\lambda} \lambda^k}{k!}$$

$$E[X] = \lambda, \quad \text{Var}(X) = \lambda$$

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

Negative Binomial distribution — PMF, mean, variance

$$P(X = k) = \binom{k-1}{r-1} p^r (1-p)^{k-r} \text{ (trials for } r\text{th success)}$$

$$E[X] = \frac{r}{p}, \quad \text{Var}(X) = \frac{r(1-p)}{p^2}$$

Binomial distribution — PMF, mean, variance

$$P(X = k) = \binom{n}{k} p^k (1-p)^{n-k}$$

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

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

Bernoulli distribution — PMF, mean, variance

$$P(X = 1) = p, \quad P(X = 0) = 1 - p$$

$$E[X] = p, \quad \text{Var}(X) = p(1-p)$$

Hypergeometric distribution — PMF, mean, variance

$$P(X = k) = \frac{\binom{K}{k} \binom{N-K}{n-k}}{\binom{N}{n}}$$

$$E[X] = \frac{nK}{N}, \quad \text{Var}(X) = \frac{nK(N-K)(N-n)}{N^2(N-1)}$$

Geometric distribution — PMF, mean, variance

$$P(X = k) = (1-p)^{k-1} p \text{ (number of trials to first success)}$$

$$E[X] = \frac{1}{p}, \quad \text{Var}(X) = \frac{1-p}{p^2}$$

$k = 1, 2, \dots$

**Normal distribution – PDF**

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$$

$$E[X] = \mu, \quad \text{Var}(X) = \sigma^2$$

**Memoryless property (Exponential / Geometric)**

$$P(X > s+t \mid X > s) = P(X > t)$$

Only the Exponential (continuous) and Geometric (discrete) satisfy this.

**Exponential distribution – PDF, CDF, mean, variance**

$$f(x) = \lambda e^{-\lambda x}, \quad F(x) = 1 - e^{-\lambda x}, \quad x > 0$$

$$E[X] = \frac{1}{\lambda}, \quad \text{Var}(X) = \frac{1}{\lambda^2}$$

**Uniform distribution – PDF, mean, variance**

$$f(x) = \frac{1}{b-a}, \quad a < x < b$$

$$E[X] = \frac{a+b}{2}, \quad \text{Var}(X) = \frac{(b-a)^2}{12}$$

**Weibull distribution – PDF, mean**

$$f(x) = \frac{\tau}{\theta} \left(\frac{x}{\theta}\right)^{\tau-1} e^{-(x/\theta)^\tau}, \quad x > 0$$

$$E[X] = \theta \Gamma\left(1 + \frac{1}{\tau}\right)$$

**Gamma distribution – PDF, mean, variance**

$$f(x) = \frac{x^{\alpha-1} e^{-x/\theta}}{\Gamma(\alpha) \theta^\alpha}, \quad x > 0$$

$$E[X] = \alpha\theta, \quad \text{Var}(X) = \alpha\theta^2$$

$\alpha$ =shape,  $\theta$ =scale

**Lognormal distribution – PDF, mean, variance**

$$f(x) = \frac{1}{x\sigma\sqrt{2\pi}} \exp\left(-\frac{(\ln x - \mu)^2}{2\sigma^2}\right), \quad x > 0$$

$$E[X] = e^{\mu+\sigma^2/2}, \quad \text{Var}(X) = e^{2\mu+2\sigma^2}(e^{\sigma^2} - 1)$$

**Beta distribution – PDF, mean**

$$f(x) = \frac{x^{\alpha-1}(1-x)^{\beta-1}}{B(\alpha, \beta)}, \quad 0 < x < 1$$

$$E[X] = \frac{\alpha}{\alpha + \beta}, \quad \text{Var}(X) = \frac{\alpha\beta}{(\alpha + \beta)^2(\alpha + \beta + 1)}$$

## JOINT &amp; MARGINAL DISTRIBUTIONS

**Conditional PDF**

$$f_{Y|X}(y | x) = \frac{f_{X,Y}(x, y)}{f_X(x)}$$

**Joint PDF – marginal densities**

$$f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x, y) dy$$

$$f_Y(y) = \int_{-\infty}^{\infty} f_{X,Y}(x, y) dx$$

**Independence of continuous RVs**

$X, Y$  independent iff  $f_{X,Y}(x, y) = f_X(x) f_Y(y)$  for all  $x, y$

## EXPECTATION &amp; VARIANCE

**Covariance**

$$\text{Cov}(X, Y) = E[XY] - E[X] E[Y]$$

If  $X, Y$  independent:  $\text{Cov}(X, Y) = 0$

**MGF definition and moment extraction**

$$M_X(t) = E[e^{tX}]$$

$$E[X^n] = M_X^{(n)}(0) \text{ (} n\text{th derivative at } t = 0\text{)}$$

**Correlation coefficient**

$$\rho_{XY} = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y}, \quad -1 \leq \rho \leq 1$$

**Variance of a linear combination**

$$\text{Var}(aX + bY) = a^2 \text{Var}(X) + b^2 \text{Var}(Y) + 2ab \text{Cov}(X, Y)$$

**Variance shortcut**

$$\text{Var}(X) = E[X^2] - (E[X])^2$$

**Law of total expectation**

$$E[X] = E[E[X | Y]]$$

**Law of total variance**

$$\text{Var}(X) = E[\text{Var}(X | Y)] + \text{Var}(E[X | Y])$$

**Ordinary deductible — payment per loss**

$$Y^L = \max(X - d, 0)$$

$$E[Y^L] = E[X] - E[X \wedge d]$$

where  $d$  = deductible,  $X$  = ground-up loss

**Stop-loss (aggregate) premium**

$$E[(S - d)_+] = E[S] - E[S \wedge d]$$

$S$  = aggregate loss,  $d$  = retention (aggregate deductible)

**Limited expected value (LEV)**

$$E[X \wedge u] = \int_0^u S(x) dx = \int_0^u [1 - F(x)] dx$$

for  $X \geq 0$

**Payment per payment (excess loss)**

$$e(d) = E[X - d \mid X > d] = \frac{E[X] - E[X \wedge d]}{1 - F(d)}$$

$d$  = deductible

**Policy limit — payment per loss**

$$Y^L = \min(X, u) = X \wedge u$$

$$E[Y^L] = E[X \wedge u]$$

$u$  = policy limit