# Variance of Different Probability Distributions'Formula Sheet

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$$P(A \cup B) = P(A) + P(B) - P(AB) \tag{1}$$

$$P(A - B) = P(A) - P(AB)$$
(2)

$$P(B|A) = \frac{P(AB)}{P(A)} \tag{3}$$

$$P(AB) = P(A|B) \times P(B) \tag{4}$$

$$P(A) = \sum_{i=1}^{n} [P(A|B_i) \times P(B_i)]$$

$$(5)$$

$$P(B_i|A) = \frac{P(B_i) \times P(A|B_i)}{\sum_{j=1}^n P(B_j) \times P(A|B_j)}$$
(6)

$$F(x) = P\{X \leqslant x\}, (-\infty < x < +\infty) \tag{7}$$

$$\lim_{x \to -\infty} F(x) = 1 \tag{8}$$

$$\lim_{x \to +\infty} F(x) = 0 \tag{9}$$

$$P\{X = k\} = C_k^n \times p^k \times (1 - p)^{n - k} \quad X \sim B(n, p)$$
(10)

$$P\{X = k\} = \frac{\lambda^k}{k!} \times e^{-\lambda} \quad X \sim \pi(n, p)$$
(11)

$$F(x) = \int_{-\infty}^{x} f(t) dt$$
 (12)

$$f(x) = \frac{1}{2} \times e^{-|x|} \tag{13}$$

$$f(x) = \frac{1}{b-a}, \quad (a < x < b) \quad X \sim U(a,b)$$
 (14)

$$f(x) = \lambda \times e^{x\lambda} \quad X \sim E(\lambda) \tag{15}$$

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} \times e^{-\frac{(x-\mu)^2}{2\sigma^2}} \quad X \sim N(\mu, \sigma^2)$$
 (16)

$$F_Y(y) = P\{Y \le y\} = P\{g(x) \le y\} = \int_{g(x) \le y} f_X(t) dt$$
 (17)

$$f_Y(y) = \frac{dF_Y(y)}{dy} \tag{18}$$

$$f_Y(y) = f_X[h(y)] \times |h'(y)| \tag{19}$$

$$F(x,y) = \int_{-\infty}^{y} \int_{-\infty}^{x} f(u,v) \, du dv \tag{20}$$

$$f(x,y) = \frac{\partial^2 F(x,y)}{\partial x \partial y} \tag{21}$$

$$f_X(x) = \int_{-\infty}^{+\infty} f(x, y) \, dy \tag{22}$$

$$f_Y(y) = \int_{-\infty}^{+\infty} f(x, y) dx \tag{23}$$

$$f(x,y) = \frac{1}{S}, \quad (x,y) \in G \quad (X,Y) \sim U(G)$$
 (24)

$$F_{X|Y}(x|y) = \int_{-\infty}^{x} \frac{f(u,y)}{f_Y(y)} \, du$$
 (25)

$$f_{X|Y}(x|y) = \frac{f(x,y)}{f_Y(y)}$$
 (26)

$$F(x,y) = F_X(x) \times F_Y(y) \tag{27}$$

$$f(x,y) = f_X(x) \times f_Y(y) \tag{28}$$

### 3.1 Z=X+Y

$$F_Z(z) = P\{X + Y \leqslant Z\} = \iint_{x+y \leqslant z} f(x,y) \, dx dy = \int_{-\infty}^{+\infty} \int_{-\infty}^{z} f(u-y,y) \, du dy$$
 (29)

$$f_Z(z) = \int_{-\infty}^{+\infty} f(z - y, y) \, dy \tag{30}$$

$$f_Z(z) = \int_{-\infty}^{+\infty} f(x, z - x) dx \tag{31}$$

$$f_X * f_Y = \int_{-\infty}^{+\infty} f_X(x) f_Y(z - x) \, dx = \int_{-\infty}^{+\infty} f_X(z - y) f_Y(y) \, dy \tag{32}$$

# 3.2 $Z = \frac{Y}{X}, Z = XY$

$$U = \frac{Y}{X}, V = XY \tag{33}$$

$$f_U(u) = \int_{-\infty}^{\infty} |x| f(x, xu) dx$$
 (34)

$$f_V(v) = \int_{-\infty}^{\infty} |x| f(x, \frac{v}{x}) dx$$
 (35)

### 3.3 M=max(X,Y), N=min(X,Y)

$$F_M(z) = F_X(z)F_Y(z) \tag{36}$$

$$F_N(z) = 1 - [1 - F_X(z)][1 - F_Y(z)]$$
(37)

Table 1: Mathematical Expectation

Distribution Type	E(X)
(0-1) distribution	p
Binomial distribution	np
Poisson distribution	$\lambda$
Uniform distribution	$\frac{a+b}{2}$
Exponential distribution	$\frac{1}{\lambda}$
normal distribution	$\mu$

Table 2: Variance of Different Probability Distributions

Distribution Type	Variance, D(X)
(0-1) distribution	p(1 - p)
Binomial distribution	np(1-p)
Poisson distribution	$\lambda$
Uniform distribution	$\frac{(b-a)^2}{12}$
Exponential distribution	$\frac{1}{\lambda^2}$
Normal distribution	$\sigma^2$

$$E(X) = \int_{-\infty}^{+\infty} x f(x) \, dx \tag{38}$$

$$E(Y) = E(g(X)) = \int_{-\infty}^{+\infty} g(x)f(x) dx \tag{39}$$

$$E(CX) = CE(X) \tag{40}$$

$$E(X+Y) = E(X) + E(Y) \tag{41}$$

$$E(XY) = E(X)E(Y) \tag{42}$$

$$E(X \pm C) = E(X) \pm C \tag{43}$$

$$D(X) = \sum_{k=1}^{n} [x_k - E(X)]^2 p_k$$
(44)

$$D(X) = \int_{-\infty}^{+\infty} [x - E(X)]^2 f(x) \, dx \tag{45}$$

$$D(X) = E(X^{2}) - [E(X)]^{2}$$
(46)

$$D(X \pm C) = D(X) \tag{47}$$

$$D(CX) = C^2 D(X) \tag{48}$$

$$Cov(X,Y) = E(XY) - E(X) \times E(Y)$$
(49)

$$Cov(X,Y) = Cov(Y,X)$$
(50)

$$Cov(aX, bY) = abCov(X, Y)$$
(51)

$$Cov(X_1 + X_2, Y) = Cov(X_1, Y) + Cov(X_2, Y)$$
 (52)

$$Cov(X,Y) = E(XY) - E(X)E(Y)$$
(53)

$$X^* = \frac{X - E(X)}{\sqrt{D(X)}}\tag{54}$$

$$Y^* = \frac{Y - E(Y)}{\sqrt{D(Y)}}\tag{55}$$

$$\rho_{xy} = Cov(X^*, Y^*) = \frac{Cov(X, Y)}{\sqrt{(D(X))}\sqrt{(D(Y))}}$$
(56)

$$P(|X - \mu| \ge \epsilon) \le \frac{\sigma^2}{\epsilon^2} \tag{57}$$

$$P(|X - \mu| < \epsilon) \ge 1 - \frac{\sigma^2}{\epsilon^2} \tag{58}$$

$$\frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} = N(0, 1) \tag{59}$$

Mean: 
$$\bar{x} = \frac{1}{n} \sum_{i=1}^{n} x_i$$
 (60)

Sample Variance: 
$$s^2 = \frac{1}{n-1} \sum_{i=1}^{n} (x_i - \bar{x})^2$$
 (61)

Sample Standard Deviation: 
$$s = \sqrt{\frac{1}{n-1} \sum_{i=1}^{n} (x_i - \bar{x})^2}$$
 (62)

Population Variance: 
$$\sigma^2 = \frac{1}{N} \sum_{i=1}^{N} (x_i - \mu)^2$$
 (63)

Population Standard Deviation: 
$$\sigma = \sqrt{\frac{1}{N} \sum_{i=1}^{N} (x_i - \mu)^2}$$
 (64)

k-th Raw Moment: 
$$\mu_k = E(X^k) = \int_{-\infty}^{\infty} x^k f(x) dx$$
 (65)

k-th Central Moment: 
$$\mu'_k = E\left[(X - \mu)^k\right] = \int_{-\infty}^{\infty} (x - \mu)^k f(x) dx$$
 (66)

#### **5.1 Distribution**

$$\chi^2 = X_1^2 + X_2^2 + \dots + X_n^2 \quad \chi^2 \sim \chi^2(n)$$
 (67)

$$X \sim N(0,1), \quad Y \sim \chi^2(n)$$
 (68)

$$T = \frac{X}{\sqrt{\frac{Y}{n}}} \quad T \sim t(n) \tag{69}$$

$$U \sim \chi^2(m), \quad V \sim \chi^2(n) \tag{70}$$

$$F = \frac{\frac{U}{m}}{\frac{V}{N}} \quad F \sim F(m, n) \tag{71}$$

$$\frac{1}{F} \sim F(n, m) \tag{72}$$

$$F_{1-\alpha}(n,m) = \frac{1}{F_{\alpha}(m,n)} \tag{73}$$

(74)

#### **Sampling distribution** 5.2

$$\frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \sim N(0, 1) \tag{75}$$

$$\frac{(n-1)S^2}{\sigma^2} \sim \chi^2(n-1)$$
 (76)

$$\frac{\bar{X} - \mu}{\frac{S}{\sqrt{n}}} \sim t(n-1) \tag{77}$$

$$\frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{m} + \frac{\sigma_2^2}{n}}} \sim N(0, 1)$$

$$\frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)}{S_{\omega} \sqrt{\frac{1}{m} + \frac{1}{n}}} \sim t(m + n - 2)$$
(78)

$$\frac{(X-Y) - (\mu_1 - \mu_2)}{S_{\omega} \sqrt{\frac{1}{m} + \frac{1}{n}}} \sim t(m+n-2)$$
(79)

$$S_{\omega} = \sqrt{\frac{(m-1)S_X^2 + (n-1)S_Y^2}{m+n-2}}$$
 (80)

$$\frac{\frac{S_X^2}{S_Y^2}}{\frac{\sigma_1^2}{\sigma_2^2}} \sim F(m-1, n-1)$$
 (81)

$$\bar{X} = E(X) = \int_{-\infty}^{+\infty} x f(x) \, dx \tag{82}$$

$$\bar{X} = E(X) = \int_{-\infty}^{+\infty} x f(x) dx$$

$$L(\theta) = \prod_{i=1}^{n} f(x_i; \theta)$$

$$\frac{dL}{d\theta} = 0$$
(82)
(83)

$$\frac{dL}{d\theta} = 0 \tag{84}$$

Table 3: Test Statistics and Confidence Intervals for Normally Distributed Population Parameters

Parameter	Statistic	Estimator	<b>Confidence Interval</b>
$\mu, \sigma^2$ known	Z	$\frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}}$	$\bar{X} \pm Z_{\frac{\alpha}{2}} \frac{\sigma}{\sqrt{n}}$
	$\chi^2$	$\frac{(n-1)s^2}{\sigma^2}$	$\left[\frac{(n-1)s^2}{\chi_{1-\frac{\alpha}{2},n-1}^2}, \frac{(n-1)s^2}{\chi_{\frac{\alpha}{2},n-1}^2}\right]$
$\mu$ , $\sigma^2$ unknown	Т	$\frac{\bar{X}-\mu}{\frac{s}{\sqrt{n}}}$	$\bar{X} \pm t_{\frac{\alpha}{2}, n-1} \frac{s}{\sqrt{n}}$
	$\chi^2$	$\frac{(n-1)s^2}{\sigma^2}$	$\frac{(n-1)s^2}{\chi^2_{1-\frac{\alpha}{2},n-1}}, \frac{(n-1)s^2}{\chi^2_{\frac{\alpha}{2},n-1}}$