

**Mathematics 333 : Probability and Statistics
Formula Sheet**

1. Sample variance $s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$
2. $\sum_{i=1}^n (x_i - \bar{x})^2 = \sum_{i=1}^n x_i^2 - \frac{1}{n} (\sum_{i=1}^n x_i)^2$
3. (*Addition Rule for Probabilities.*) For events A and B ,

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

4. For positive integers n and k , with $k \leq n$,

i) Permutations: $P_k^n = n(n-1) \cdots (n-k+1) = \frac{n!}{(n-k)!}$

ii) Combinations: $\binom{n}{k} = \frac{n!}{k!(n-k)!}$

5. Conditional probability formula: $P(A | B) = \frac{P(A \cap B)}{P(B)}$
6. (*Independence*) Events A and B are independent if and only if:

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

7. (*Bayes' formula*) If $\{A_1, A_2, \dots, A_n\}$ are mutually exclusive and exhaustive events, and B is any event, then

$$P(A_k | B) = \frac{P(A_k)P(B | A_k)}{\sum_{i=1}^n P(A_i)P(B | A_i)}, \quad k = 1, 2, \dots, n$$

8. The c.d.f. $F(x)$ of a random variable X :

$$F(x) = P(X \leq x) = \begin{cases} \sum_{y: y \leq x} f(y), & \text{for } X \text{ discrete} \\ \int_{-\infty}^x f(y) dy, & \text{for } X \text{ continuous} \end{cases}$$

9. For any real valued function $h(x)$,

$$E\{h(X)\} = \begin{cases} \sum_x h(x)f(x), & \text{for } X \text{ discrete} \\ \int_{-\infty}^{\infty} h(x)f(x) dx, & \text{for } X \text{ continuous} \end{cases}$$

10. binomial($x; n, p$) = $\binom{n}{x} p^x (1-p)^{n-x}$, $x = 0, 1, 2, \dots, n$;
 $E(X) = np$, $V(X) = np(1-p)$.

11. Poisson($x; \lambda$) = $e^{-\lambda} \left(\frac{\lambda^x}{x!}\right)$, $x = 0, 1, 2, \dots$; $E(X) = V(X) = \lambda$.

12. Exponential density: $f(x) = \lambda e^{-\lambda x}$, $x > 0$; $EX = \lambda^{-1}$, $V(X) = \lambda^{-2}$.

13. Z-score of the sample mean \bar{X} for a sample of size n , is $Z = \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}}$.

14. (i) A *large sample* $100(1 - \alpha)\%$ confidence interval (CI) for the population mean μ , is:

$$\bar{X} \pm z_{\frac{\alpha}{2}} \left(\frac{\sigma}{\sqrt{n}} \right)$$

(ii) A *small sample* $100(1 - \alpha)\%$ confidence interval (CI) for a *normal* population mean μ , is:

$$\bar{X} \pm t_{\frac{\alpha}{2}, n-1} \left(\frac{s}{\sqrt{n}} \right)$$

(iii) A *large sample* $100(1 - \alpha)\%$ confidence interval (CI) for the population proportion p is:

$$\hat{p} \pm z_{\frac{\alpha}{2}} \left(\sqrt{\frac{\hat{p}\hat{q}}{n}} \right)$$

15. A $100(1 - \alpha)\%$ CI for the population variance σ^2 in a Normal distribution is :

$$\left(\frac{(n-1)s^2}{\chi_{\frac{\alpha}{2}, n-1}^2}, \frac{(n-1)s^2}{\chi_{1-\frac{\alpha}{2}, n-1}^2} \right)$$

16. An *exact* small sample $100(1 - \alpha)\%$ CI for the difference ($\mu_1 - \mu_2$) of two normal means, with equal variances (unknown) is:

$$\bar{X} - \bar{Y} \pm t_{\frac{\alpha}{2}, n+m-2} \left(s_p \sqrt{\frac{1}{m} + \frac{1}{n}} \right),$$

where $s_p^2 = \frac{(m-1)s_1^2 + (n-1)s_2^2}{m+n-2}$, and (\bar{X}, s_1^2) , (\bar{Y}, s_2^2) are the sample means and variances based on m and n random sample observations drawn from independent Normal populations with a common unknown variance.

17. A small sample *approximate* $100(1 - \alpha)\%$ CI for the difference ($\mu_1 - \mu_2$) of two Normal means and possibly unequal variances, is:

$$(\bar{X} - \bar{Y}) \pm t_{\frac{\alpha}{2}, \nu} \sqrt{\frac{s_1^2}{m} + \frac{s_2^2}{n}}$$

where, $\nu = \frac{(\{s_1^2/m\} + \{s_2^2/n\})^2}{\frac{(s_1^2/m)^2}{m-1} + \frac{(s_2^2/n)^2}{n-1}}$ is the estimated degrees of freedom rounded down to the nearest integer.