

Distributions

Distribution	Probability Density Function	$\mathbb{E}(X)$	$\text{Var}(X)$
$X \sim \text{Bi}(n, p)$	$p_X(x) = \binom{n}{x} p^x (1-p)^{n-x}$	np	$np(1-p)$
$N \sim \text{G}(p)$	$p_N(n) = (1-p)^n p$	$\frac{1-p}{p}$	$\frac{1-p}{p^2}$
$X \sim \text{Pn}(\alpha t)$	$p_X(x) = \frac{e^{-\alpha} (\alpha)^x}{x!}$	α	α
$X \sim \text{U}(m, n)$	$p_X(x) = \frac{1}{n-m+1}$	$\frac{m+n}{2}$	$\frac{1}{12}((n-m+1)^2 - 1)$
$X \sim \text{R}(a, b)$	$f_X(x) = \frac{1}{b-a}$	$\frac{a+b}{2}$	$\frac{1}{12}(b-a)^2$
$T \sim \exp(\alpha)$	$f_T(t) = \alpha e^{-\alpha t}$, for $t \geq 0$	$\frac{1}{\alpha}$	$\frac{1}{\alpha^2}$
$T \sim \gamma(r, \alpha)$	$f_T(t) = \frac{\alpha^r t^{r-1} e^{-\alpha t}}{\Gamma(r)}$, $t > 0$	$\frac{r}{\alpha}$	$\frac{r}{\alpha^2}$
$X \sim \text{Pareto}(\alpha, \gamma)$	$f_X(x) = \frac{\gamma \alpha^\gamma}{x^{\gamma+1}}$, for $\alpha \leq x < \infty$	$\frac{\gamma \alpha}{\gamma-1}$, for $\gamma > 1$	$\frac{\gamma \alpha^2}{(\gamma-1)^2(\gamma-2)}$, for $\gamma > 2$
$X \sim \text{N}(\mu, \sigma^2)$	$f_X(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2\sigma^2}(x-\mu)^2}$, for $x \in \mathbb{R}$	μ	σ^2
$X \sim \text{C}(m, a)$	$f_X(x) = \frac{1}{\pi} \frac{a}{a^2 + (x-m)^2}$, for $-\infty < x < \infty$	Undefined	Undefined
$Y \sim \text{LN}(\mu, \sigma^2)$	$f_Y(y) = \frac{1}{\sqrt{2\pi\sigma y}} e^{-\frac{(\ln y - \mu)^2}{2\sigma^2}}$, for $y > 0$	$e^{r\mu + \frac{1}{2}r^2\sigma^2}$, $r \geq 0$	$e^{2\mu + \sigma^2}(e^{\sigma^2} - 1)$

Confidence Intervals

Estimating Means

For large r , $t_r \rightarrow \text{N}(0, 1)$ is a good approximation.

Normal, Single Mean, Known σ

$(\bar{x} \pm c \frac{\sigma}{\sqrt{n}})$; with $c = F^{-1}(1 - \frac{\alpha}{2})$ from pivot $\text{N}(0, 1)$.

Normal, Single Mean, Unknown σ

$(\bar{x} \pm c \frac{s}{\sqrt{n}})$; with $c = F^{-1}(1 - \frac{\alpha}{2})$ from pivot t_{n-1} .

Normal, Two Means, Two Known σ 's

$(\bar{x} - \bar{y}) \pm c \sqrt{\frac{\sigma_X^2}{n} + \frac{\sigma_Y^2}{m}}$; with c from pivot $\text{N}(0, 1)$.

Normal, Two Means, Unknown σ 's, Many Samples

$(\bar{x} - \bar{y}) \pm c \sqrt{\frac{s_X^2}{n} + \frac{s_Y^2}{m}}$; with c from pivot $\text{N}(0, 1)$.

Normal, Two Means, Unknown σ 's, Common Variance

$(\bar{x} - \bar{y}) \pm c \cdot s_P \sqrt{\frac{1}{n} + \frac{1}{m}}$; with c from pivot t_{n+m-2} .

$$s_P = \sqrt{\frac{(n-1)s_X^2 + (m-1)s_Y^2}{n+m-2}}$$

Normal, Two Means, Unknown σ 's, \neq Variances

$(\bar{x} - \bar{y}) \pm c \sqrt{\frac{s_X^2}{n} + \frac{s_Y^2}{m}}$; with c from pivot t_r .

$$r = \left(\frac{s_X^2}{n} + \frac{s_Y^2}{m} \right)^2 / \left(\frac{s_X^4}{n^2(n-1)} + \frac{s_Y^4}{m^2(m-1)} \right)$$

Normal, Paired Samples

For pairs (X_i, Y_i) , let $D_i = X_i - Y_i$. $D_i \sim \text{N}(\mu_D, \sigma_D^2)$.

$(\bar{d} \pm c \frac{s_d}{\sqrt{n}})$; with c from pivot t_{n-1} .

Estimating Variance

Normal, Single Variance - estimate of σ^2 not σ !

$\left(\frac{(n-1)s^2}{b}, \frac{(n-1)s^2}{a} \right)$; with a, b from pivot χ_{n-1}^2 .

Normal, Two Variances

A confidence interval for the ratio of the variances σ_X^2/σ_Y^2 is

$\left(a \cdot \frac{s_X^2}{s_Y^2}, b \cdot \frac{s_X^2}{s_Y^2} \right)$; with a, b from pivot $F_{m-1, n-1}$.

Estimating Proportions

Single Proportion

$\approx \left(\hat{p} \pm c \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} \right)$; with c from pivot $\text{N}(0, 1)$.

Two Proportions

$\approx \left(\hat{p}_1 - \hat{p}_2 \pm c \sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n} + \frac{\hat{p}_2(1-\hat{p}_2)}{n}} \right)$; c from pivot $\text{N}(0, 1)$.

Prediction Intervals

Let X^* be a future realisation of X .

If $X \sim \text{N}(\mu, \sigma^2)$, $\bar{X} \sim \text{N}(\mu, \frac{\sigma^2}{n})$, and $(\bar{X} - X^*) \sim \text{N}(\mu, \sigma^2 + \frac{\sigma^2}{n})$.

$\left(\bar{x} \pm c \sqrt{s^2 + \frac{s^2}{n}} \right)$ is a prediction interval with c from t_{n-1} .

If σ is known, use pivot $\text{N}(0, 1)$. If μ is known use χ_{n-1}^2 .

Other Formulae

Sample Variance

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2 = \frac{1}{n-1} \left(\sum_{i=1}^n x_i^2 - n\bar{x}^2 \right)$$