

TEST CATALOG FOR THE BINOMIAL DISTRIBUTION

Statistical model:

- $X \sim \text{binomial}(n, p)$
- Parameter estimate: $\hat{p} = x/n$
- Where the observation is $x = \text{'number of successes out of } n \text{ trials'}$

Hypothesis test (two-tailed):

- $H_0: p = p_0$
- $H_1: p \neq p_0$
- Test size: $z = \frac{x - np_0}{\sqrt{np_0(1-p_0)}} \sim N(0,1)$
- Approximate p-value: $2 \cdot |1 - \Phi(|z|)|$

95% confidence interval:

- $p_- = \frac{1}{n+1.96^2} \left[x + \frac{1.96^2}{2} - 1.96 \sqrt{\frac{x(n-x)}{n} + \frac{1.96^2}{4}} \right]$
- $p_+ = \frac{1}{n+1.96^2} \left[x + \frac{1.96^2}{2} + 1.96 \sqrt{\frac{x(n-x)}{n} + \frac{1.96^2}{4}} \right]$

TEST CATALOG FOR THE POISSON DISTRIBUTION

Statistical model:

- $X \sim poisson(\lambda \cdot t)$
- Parameter estimate: $\hat{\lambda} = x/t$
- Where the observation is $x =$ 'number of arrivals/events observed over a period of time t '

Hypothesis test (two-tailed):

- $H_0: \lambda = \lambda_0$
- $H_1: \lambda \neq \lambda_0$
- Test size: $z = \frac{x - \lambda \cdot t}{\sqrt{\lambda \cdot t}} \sim N(0,1)$
- Approximate p-value: $2 \cdot |1 - \Phi(|z|)|$

95% confidence interval:

- $\lambda_- = \frac{1}{t} \left[x + \frac{1.96^2}{2} - 1.96 \sqrt{x + \frac{1.96^2}{4}} \right]$
- $\lambda_+ = \frac{1}{t} \left[x + \frac{1.96^2}{2} + 1.96 \sqrt{x + \frac{1.96^2}{4}} \right]$

TEST CATALOG FOR THE MEAN (KNOWN VARIANCE)

Statistical model:

- X_1, X_2, \dots, X_n are i.i.d. samples of a random variable X with mean μ and variance σ^2 .
- Parameter estimate:

$$\hat{\mu} = \bar{x} = \frac{1}{n} \sum_{i=1}^n X_i \sim N(\mu, \sigma^2/n)$$

- Where the observation is \bar{x} = 'the average of n samples drawn from X 's distribution'.
- NOTE: The statistical model is only true if n is sufficiently large ($n \geq 30$) or if the samples are drawn from a normal population with mean μ and variance σ^2 .

Hypothesis test (two-tailed):

- $H_0: \mu = \mu_0$
- $H_1: \mu \neq \mu_0$
- Test size: $z = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}} \sim N(0,1)$
- Approximate p-value: $2 \cdot |1 - \Phi(|z|)|$

95% confidence interval:

- $\mu_- = \bar{x} - 1.96 \cdot \sigma/\sqrt{n}$
- $\mu_+ = \bar{x} + 1.96 \cdot \sigma/\sqrt{n}$

TEST CATALOG FOR THE MEAN (UNKNOWN VARIANCE)

Statistical model:

- X_1, X_2, \dots, X_n are i.i.d. samples of a random variable X with mean μ and variance σ^2 .
- Parameter estimates:

$$\hat{\mu} = \bar{x} = \frac{1}{n} \sum_{i=1}^n X_i \sim N(\mu, \sigma^2/n)$$

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

- Where the observation is \bar{x} = 'the average of n samples drawn from X 's distribution'.
- NOTE: The statistical model is only true if n is sufficiently large ($n \geq 30$) or if the samples are drawn from a normal population with mean μ and variance σ^2 .

Hypothesis test (two-tailed):

- $H_0: \mu = \mu_0$
- $H_1: \mu \neq \mu_0$
- Test size: $t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}} \sim t(n-1)$
- Approximate p-value: $2 \cdot (1 - t_{cdf}(|t|))$

95% confidence interval:

- $\mu_- = \bar{x} - t_0 \cdot s/\sqrt{n}$
- $\mu_+ = \bar{x} + t_0 \cdot s/\sqrt{n}$

where $t_0 = \text{tinv}(1-0.05/2, n-1)$

TEST CATALOG FOR COMPARING TWO MEANS (KNOWN VARIANCE)

Statistical model:

- $X_{1i} \sim N(\mu_1, \sigma_1^2), i = 1, 2, \dots, n_1$ and $X_{2i} \sim N(\mu_2, \sigma_2^2) i = 1, 2, \dots, n_2$
- Parameter estimate:

$$\hat{\delta} = \bar{x}_1 - \bar{x}_2 \sim N(\mu_1 - \mu_2, \sigma_1^2/n_1 + \sigma_2^2/n_2)$$

- Where the observation is $\bar{x}_1 - \bar{x}_2$ = 'the difference between two sample means'.

Hypothesis test (two-tailed):

- $H_0: \mu_1 = \mu_2$
- $H_1: \mu_1 \neq \mu_2$
- Test size: $z = \frac{(\bar{x}_1 - \bar{x}_2)}{\sigma \sqrt{1/n_1 + 1/n_2}} \sim N(0,1)$
- Approximate p-value: $2 \cdot |1 - \Phi(|z|)|$

95% confidence interval:

- $\delta_- = (\bar{x}_1 - \bar{x}_2) - 1.96 \cdot \sigma \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$
- $\delta_+ = (\bar{x}_1 - \bar{x}_2) + 1.96 \cdot \sigma \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$

TEST CATALOG FOR COMPARING TWO MEANS (UNKNOWN VARIANCE)

Statistical model:

- $X_{1i} \sim N(\mu_1, \sigma_1^2), i = 1, 2, \dots, n_1$ and $X_{2i} \sim N(\mu_2, \sigma_2^2) i = 1, 2, \dots, n_2$
- Parameter estimate:

$$\hat{\delta} = \bar{x}_1 - \bar{x}_2 \sim N(\mu_1 - \mu_2, \sigma_1^2/n_1 + \sigma_2^2/n_2)$$
$$s^2 = \frac{1}{n_1 + n_2 - 2} \left((n_1 - 1)s_1^2 + (n_2 - 1)s_2^2 \right)$$

- Where the observation is $\bar{x}_1 - \bar{x}_2 =$ 'the difference between two sample means'.

Hypothesis test (two-tailed):

- $H_0: \mu_1 = \mu_2$
- $H_1: \mu_1 \neq \mu_2$
- Test size: $t = \frac{(\bar{x}_1 - \bar{x}_2)}{s\sqrt{1/n_1 + 1/n_2}} = \sim t(n_1 + n_2 - 2)$
- Approximate p-value: $2 \cdot (1 - t_{cdf}(|t|, n_1 + n_2 - 2))$

95% confidence interval:

- $\delta_- = (\bar{x}_1 - \bar{x}_2) - t_0 \cdot s \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$
- $\delta_+ = (\bar{x}_1 - \bar{x}_2) + t_0 \cdot s \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$

where $t_0 = \text{tinv}(1-0.05/2, n_1+n_2-2)$

TEST CATALOG FOR PAIRED DATA

Statistical model:

- $d_i = X_{1i} - X_{2i}$, where $d_i \sim N(\delta, \sigma^2)$, $i = 1, 2, \dots, n$
- Parameter estimate:

$$\hat{\delta} = \bar{d} = \frac{1}{n} \sum_{i=1}^n X_{1i} - X_{2i}$$

$$s_d^2 = \frac{1}{n-1} \sum_{i=1}^n (d_i - \bar{d})^2$$

- Where the observation is \bar{d} = 'the average of the differences between paired samples'.

Hypothesis test (two-tailed):

- $H_0: \delta = \delta_0$
- $H_1: \delta \neq \delta_0$
- Test size: $t = \frac{\bar{d} - \delta_0}{s_d / \sqrt{n}} = \sim t(n-1)$
- Approximate p-value: $2 \cdot (1 - t_{cdf}(|t|, n-1))$

95% confidence interval:

- $\delta_- = \bar{d} - t_0 \cdot \frac{s_d}{\sqrt{n}}$
- $\delta_+ = \bar{d} + t_0 \cdot \frac{s_d}{\sqrt{n}}$

where $t_0 = \text{tinv}(1-0.05/2, n-1)$

TEST CATALOG FOR THE SLOPE IN SIMPLE LINEAR REGRESSION

Statistical model:

- $y_i \sim N(\alpha + \beta x_i, \sigma^2)$ for $i = 1, 2, \dots, n$ are independent samples.
- Parameter estimates:

$$\hat{\beta} = \frac{\sum_{i=1}^n (x_i - \bar{x}) \cdot (y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

$$\hat{\alpha} = \bar{y} - \hat{\beta} \cdot \bar{x}$$

$$s_r^2 = \frac{1}{n-2} \sum_{i=1}^n (y_i - (\hat{\alpha} + \hat{\beta} x_i))^2$$

Hypothesis test (two-tailed):

- $H_0: \beta = \beta_0$
- $H_1: \beta \neq \beta_0$
- Test size: $t = \frac{\hat{\beta} - \beta_0}{s_r \sqrt{1 / \sum_{i=1}^n (x_i - \bar{x})^2}} \sim t(n-2)$
- Approximate p-value: $2 \cdot (1 - t_{cdf}(|t|, n-2))$

95% confidence interval:

- $\beta_- = \hat{\beta} - t_0 \cdot s_r \sqrt{\frac{1}{\sum_{i=1}^n (x_i - \bar{x})^2}}$
- $\beta_+ = \hat{\beta} + t_0 \cdot s_r \sqrt{\frac{1}{\sum_{i=1}^n (x_i - \bar{x})^2}}$

where $t_0 = \text{tinv}(1-0.05/2, n-2)$.

TEST CATALOG FOR THE INTERCEPT IN SIMPLE LINEAR REGRESSION

Statistical model:

- $y_i \sim N(\alpha + \beta x_i, \sigma^2)$ for $i = 1, 2, \dots, n$ are independent samples.
- Parameter estimates:

$$\hat{\beta} = \frac{\sum_{i=1}^n (x_i - \bar{x}) \cdot (y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

$$\hat{\alpha} = \bar{y} - \hat{\beta} \cdot \bar{x}$$

$$s_r^2 = \frac{1}{n-2} \sum_{i=1}^n (y_i - (\hat{\alpha} + \hat{\beta} x_i))^2$$

Hypothesis test (two-tailed):

- $H_0: \alpha = \alpha_0$
- $H_1: \alpha \neq \alpha_0$
- Test size: $t = \frac{\hat{\alpha} - \alpha_0}{s_r \sqrt{\frac{1}{n} + \frac{\bar{x}^2}{\sum_{i=1}^n (x_i - \bar{x})^2}}} \sim t(n-2)$
- Approximate p-value: $2 \cdot (1 - t_{cdf}(|t|, n-2))$

95% confidence interval:

- $\alpha_- = \hat{\alpha} - t_0 \cdot s_r \sqrt{\frac{1}{n} + \frac{\bar{x}^2}{\sum_{i=1}^n (x_i - \bar{x})^2}}$
- $\alpha_+ = \hat{\alpha} + t_0 \cdot s_r \sqrt{\frac{1}{n} + \frac{\bar{x}^2}{\sum_{i=1}^n (x_i - \bar{x})^2}}$

where $t_0 = \text{tinv}(1-0.05/2, n-2)$.