
Uni-variate (1D) inference about mean of one-sample normal population

Model X_1, X_2, \dots, X_n iid $X_i \sim N(\mu, \sigma^2)$, μ, σ^2 unknown

Hypothesis $H_0: \mu = \mu_0$ against $H_1: \mu \neq \mu_0$

Point-estimators $\hat{\mu} = \bar{x} = \frac{1}{n} \sum_{i=1}^n x_i \sim N\left(\mu, \frac{\sigma^2}{n}\right)$
(ML, unbiased, consistent estimator)

$$\hat{\sigma}^2 = S^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2 \sim \frac{\sigma^2}{n-1} X_{n-1}^2$$

(not ML, but unbiased, consistent)

Test-statistic $H_0 \Rightarrow T = \frac{\bar{x} - \mu_0}{S/\sqrt{n}} \sim t_{n-1}$

Test of hypothesis calculate value of test-statistic $T = t$

approach I:

Select level of significance, α (level of confidence $1 - \alpha$)

Reject H_0 if $|t| > t_{n-1, \alpha/2}$

approach II:

calculate p-value: $p = 2P(t_{n-1} > |t|)$

Reject H_0 for all $\alpha > p$

Alternative test-statistic $H_0 \Rightarrow T^2 = \frac{(\bar{x} - \mu_0)^2}{S^2/n} = n(\bar{x} - \mu_0)(S^2)^{-1}(\bar{x} - \mu_0)$
 $T^2 \sim "(t_{n-1})^2" = F_{1, n-1}$
 T^2 could be used for testing H_0 , using the $F_{1, n-1}$ distribution

Confidence interval for μ $P\left(-t_{n-1, \alpha/2} < \frac{\bar{x} - \mu}{S/\sqrt{n}} < t_{n-1, \alpha/2}\right) = 1 - \alpha \Rightarrow$
100(1 - α)% confidence interval for μ :
 $\left[\bar{x} - \frac{S}{\sqrt{n}} t_{n-1, \alpha/2}, \bar{x} + \frac{S}{\sqrt{n}} t_{n-1, \alpha/2}\right]$
