
REFERENCE PAGES

STATISTICS

DESCRIPTIVE STATISTICS

1. Sample mean: $\bar{X} = \frac{1}{n} \left(\sum_{i=1}^n X_i \right)$
2. Sample variance: $s_n^2 = \frac{1}{n-1} \left(\sum_{i=1}^n (X_i - \bar{X})^2 \right)$
3. Skewness (asymmetry): $\gamma_1 = \frac{\mu^3}{\sigma^3}$
4. Kurtosis (peakedness): $\gamma_2 = \frac{\mu^4}{\sigma^4}$

RULES OF THUMB

$$P(|X - \mu| < 1\sigma) = 0.6826$$

5. $P(|X - \mu| < 2\sigma) = 0.9544$

$$P(|X - \mu| < 3\sigma) = 0.9973, \text{ "3}\sigma \text{ rule"}$$

CONFIDENCE INTERVAL ESTIMATION

6. Confidence interval of μ when σ^2 is known, using normal distribution:

a) 95% confidence interval: $\left[\bar{X} - 1.96 \frac{\sigma}{\sqrt{n}}, \bar{X} + 1.96 \frac{\sigma}{\sqrt{n}} \right],$

b) 90% confidence interval: $\left[\bar{X} - 1.645 \frac{\sigma}{\sqrt{n}}, \bar{X} + 1.645 \frac{\sigma}{\sqrt{n}} \right].$

7. Confidence interval of μ when σ^2 is unknown, using t-distribution:

$$T = \frac{\bar{X} - \mu}{\sqrt{s_n^2/n}} \sim t(n-1).$$

HYPOTHESIS TESTING

8. One-tail Procedure Example:

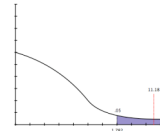
1) $H_0 : \mu = 0, H_1 : \mu > 0$ [one tail]

2) $\bar{X} = 0.031, s_n^2 = 0.0001, n = 13$ [b/c σ^2 is unknown, use t-distribution]

$$t = \frac{\bar{X} - \mu}{\sqrt{s_n^2/n}} = \frac{(0.031 - 0)\sqrt{13}}{.01} = 11.18 \sim t(12)$$

3) $\alpha = 0.05, t_\alpha = 1.782 < 11.18 = t$

4) We reject H_0 .



9. Two-tail Procedure Example:

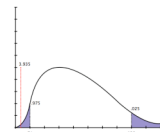
1) $H_0 : \sigma^2 = 0.4, H_1 : \sigma^2 \neq 0.4$ [two tail]

2) $s_n^2 = 0.0159, n = 100$ [b/c testing for population variance, use χ^2 -distribution]

$$\chi^2 = \frac{(n-1)s_n^2}{\sigma^2} \sim \chi^2(n-1) = \frac{(99)(0.0159)}{0.4} = 3.935 \sim \chi^2(99)$$

3) $\alpha = 0.05, 3.935 < 74$

4) We reject H_0 .



REFERENCE PAGES

STATISTICS

NORMAL DISTRIBUTION OR $N > 30$ AND σ_i KNOWN

10. One-sample z-test: $z = \frac{\bar{X} - \mu_0}{(\sigma/\sqrt{n})}$, distribution of sample mean.

11. Two-sample z-test: $z = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$, distribution of independent sample means.

t DISTRIBUTION OR $N < 30$ AND σ_i UNKNOWN, BUT EQUAL

12. One-sample t-test: $T = \frac{\bar{X} - \mu_0}{(s_n/\sqrt{n})}$, $df = n - 1$, distribution of sample mean.

13. If $X \sim N(0,1)$ and $Y \sim \chi^2(n)$, then $T = \frac{X}{\sqrt{Y/n}}$, $df = n$, distribution of X related to sample mean and Y related to sample variance, then their ratio has to do with t-distribution.

14. Two-sample t-test: $T = \frac{\bar{X} - \bar{Y}}{\sqrt{\frac{1}{n_x} + \frac{1}{n_y}} \sqrt{\frac{(n_x-1)s_x^2 + (n_y-1)s_y^2}{n_x+n_y-2}}} \sim t(n_x + n_y - 2)$, assuming $X \sim N(\mu_1, \sigma^2)$,

$Y \sim N(\mu_2, \sigma^2)$, and $H_0 : \mu_1 = \mu_2$.

15. As $n \rightarrow 30$, t-distribution converges to normal distribution.

χ^2 DISTRIBUTION

16. One-sample χ^2 test: $\chi^2 = \frac{(n-1)s_n^2}{\sigma^2} \sim \chi^2(n-1)$, distribution of sample variance.

F-DISTRIBUTION

17. If $X \sim \chi^2(m)$ and $Y \sim \chi^2(n)$, then $F = \frac{(X/m)}{(Y/n)} \sim F(m, n)$, ratios of two sets of sample

variances have to do with F-distribution.

18. ANOVA uses F-distribution.