

# Statistics Formula Sheet

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## Estimation and MLEs

An estimator  $T$  of  $\theta$  is **unbiased** if:

$$\mathbb{E}(T) = \theta,$$

and **consistent** if:

$$\begin{aligned} \text{Var}(T) &\rightarrow 0 \quad \text{and} \\ \mathbb{E}(T) &\rightarrow \theta \\ \text{as } n &\rightarrow \infty \end{aligned}$$

for example,  $\hat{\sigma}^2 = s^2 = \frac{1}{n-1} (\sum_{i=1}^n X_i^2 - n\bar{X}^2)$  is unbiased and consistent.

The **maximum likelihood estimation**  $\hat{\alpha}$  is the solution to the equation

$$\frac{\partial}{\partial \alpha} L(x_1, \dots, x_n; \theta) = \frac{\partial}{\partial \alpha} \prod_{i=1}^n f(x_i; \theta) = 0$$

where  $f(x_i; \theta) = \mathbb{P}(X_i = x_i)$  and

## Confidence Intervals:

These are the  $100(1 - \alpha)$  **confidence intervals** for various stuff. Define  $X_1, \dots, X_n \sim N(0, 1)$ , and remember that you **cannot** reject ' $H_0 : p = p_0$ ' if  $p_0 \in (a, b)$  where  $(a, b)$  is the confidence interval.

$$\begin{aligned} \mu; \sigma \text{ known} &: \bar{x} \pm z_{\alpha/2} \frac{\sigma}{\sqrt{n}} \\ \mu; \sigma \text{ unknown} &: \bar{x} \pm t_{n-1; \alpha/2} \frac{\sigma}{\sqrt{n}} \\ \hat{\beta} &: \bar{\beta} \pm t_{n-2; \alpha/2} \times \text{s.e.}(\hat{\beta}) \\ \hat{\alpha} &: \bar{\alpha} \pm t_{n-2; \alpha/2} \times \text{s.e.}(\hat{\alpha}) \\ \hat{\mathbb{E}}(Y_0) &: (\bar{\alpha} + \bar{\beta}x_0) \pm t_{n-2; \alpha/2} \sqrt{s^2 \left( \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{XX}} \right)} \\ \hat{Y}_0 &: (\bar{\alpha} + \bar{\beta}x_0) \pm t_{n-2; \alpha/2} \sqrt{s^2 \left( 1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{XX}} \right)} \end{aligned}$$

## Critical Regions

Remember that you **reject**  $H_0$  if  $t$ , the observed test statistic, is **in** the critical region, ie  $t = \frac{\bar{x}-65}{1.2}$  (pg.38 of the notes), reject  $H_0$  if  $t \leq -1.96 = -z_{\alpha/2}$  or if  $t \geq 1.96 = z_{\alpha/2}$ .

Type of test	Test Statistic	reject $H_0$ if:	For use when:
<b>z-test</b>	$ T  = \frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}}$	$ t  \geq z_{\alpha/2}$	$\sigma^2$ is known.
<b>one sample t</b>	$ T  = \frac{\bar{X} - \mu_0}{S/\sqrt{n}}$	$ t  \geq t_{n-1; \alpha/2}$	$\sigma^2$ is unknown.
<b>paired t</b>	$ T  = \frac{D}{S/\sqrt{n}}$	$ t  \geq t_{n-1; \alpha/2}$	data is paired as $(X_i, Y_i)$ .
<b>two sample t</b>	$ T  = \frac{\bar{X} - \bar{Y}}{S_p/\sqrt{\frac{1}{m} + \frac{1}{n}}}$	$ t  \geq t_{m+n-2; \alpha/2}$	$\bar{X} \sim N(\mu_X, \frac{\sigma^2}{n})$ and $\bar{Y} \sim N(\mu_Y, \frac{\sigma^2}{m})$
<b>F-test</b>	$ T  = \frac{r_{SS0} - r_{SS1}/m}{r_{SS1}/n}$	$ t  \geq F_{m, n; \alpha/2}$	ANOVA Tests

and p-values are found using:  $p\text{-value} = \mathbb{P}(|T| \geq t)$

## Various Distributional Stuff

### Chi-Squared

The chi-squared  $\chi^2$  distribution is defined by

$$\sum_{i=1}^n Z^2 = \sum_{i=1}^n \left( \frac{Y_i - \mu_i}{\sigma} \right)^2 \sim \chi_n^2,$$

where each  $Y_i \sim N(\mu_i, \sigma)$ , and has the that

$$\frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2.$$

### t-distribution

$$\frac{\bar{X} - \mu}{\sqrt{S^2/n}} = \frac{Z}{\sqrt{Y/(n-1)}} \sim t_{n-1}$$

where  $Z = \frac{\bar{X} - \mu}{\sqrt{\sigma^2/n}} \sim N(0, 1)$  and  $Y = \frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2$  are *independent* variables.

### F-distribution

Let  $U$  and  $V$  be rvs with  $U \sim \chi_{m-1}^2$  and  $V \sim \chi_{n-1}^2$ , then

$$\frac{U/m}{V/n} = \frac{(rss_0 - rss_1)/m}{rss_1/n} \sim F_{m,n}$$

where  $rss_0 = SS_{Tot}$  and  $rss_1 = SS_W$  are *residual sums of squares* (and  $SS_B = rss_0 - rss_1$ ). Also we have that

$$\frac{S_X^2}{S_Y^2} \sim F_{m-1, n-1}$$

And lastly, if  $X \sim t_n$  then  $X^2 \sim F_{1,n}$ .

## Assumptions of linear models

When forming linear models, the assumptions are that **the observations:**

- are Normally distributed
- are Independent
- have constant variance
- are linearly regressed

## ANOVA

When conducting ANOVA tests, the *assumptions* are that **the observations:**

- are Normally distributed
- are Independent
- have constant variance
- have expected values dependent only on their groups

### What ANOVA test do I use? (WANOVATDIU?)

**One-Way** if testing blocks and treatments that have single entries, and

**Two-Way** if you're concerned with seeing if the blocks and treatments have individual effects on a variable.

## LSD

Remember to **ONLY USE LSD's IF YOU REJECT  $H_0$** !

### One-Way:

$$L.S.D. = t_{n-k; \alpha/2} \sqrt{s^2 \left( \frac{1}{n_i} + \frac{1}{n_j} \right)}$$

where  $n_i$  and  $n_j$  are sample sizes for samples  $i$  and  $j$ .

### Two-Way: No Replications

For Blocks:

$$L.S.D. = t_{(b-1)(k-1); \alpha/2} \sqrt{2 \frac{s^2}{k}}$$

For Treatments:

$$L.S.D. = t_{(b-1)(k-1); \alpha/2} \sqrt{2 \frac{s^2}{b}}$$

### Two-Way: With Replications

For Blocks:

$$L.S.D. = t_{rbk-b-k+1; \alpha/2} \sqrt{2 \frac{s^2}{rk}}$$

For Treatments:

$$L.S.D. = t_{rbk-b-k+1; \alpha/2} \sqrt{2 \frac{s^2}{rb}}$$

where  $r$  is the number of replications per group.