

Test $H_0 : \mu_1 = \mu_2$ at the $\alpha=.05$ l.o.s. }
 $H_0 : \mu_1 = \mu_3$ at the $\alpha=.05$ l.o.s. }
 $H_0 : \mu_2 = \mu_3$ at the $\alpha=.05$ l.o.s. }

⇒ $\alpha = \text{P}(\text{type I error for each test}) \quad (= \alpha_{\text{Comparison-Wise}})$
 $\text{P}(\text{at least one type I error}) \gg \alpha \quad (= \alpha_{\text{Experiment-Wise}})$

ANalysis Of VAriance (ANOVA)

$H_0 : \mu_1 = \mu_2 = \mu_3$

H_a : not all means are equal

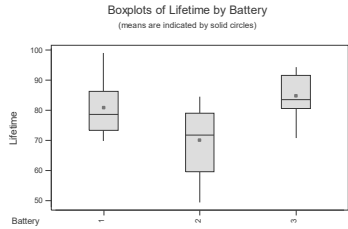
$$F_s = \frac{\text{variation between samples}}{\text{variation within samples (pooled)}}$$

ANalysis Of VAriance

	Source	d.f.	SS	MS	F
(Between Groups)	Treatment	$t-1$	$\sum_{i=1}^t \sum_{j=1}^{n_i} (\bar{Y}_{i\cdot} - \bar{Y}_{\cdot\cdot})^2$	$SST / (t-1)$	MST / MSE
(Within Groups)	Error	$N-t$	$\sum_{i=1}^t \sum_{j=1}^{n_i} (Y_{ij} - \bar{Y}_{i\cdot})^2$	$SSE / (N-t)$	
	Total	$N-1$	$\sum_{i=1}^t \sum_{j=1}^{n_i} (Y_{ij} - \bar{Y}_{\cdot\cdot})^2$		

Statistical Model: $Y_{ij} = \mu + \alpha_i + \varepsilon_{ij} \quad \varepsilon_{ij} \sim N(0, \sigma^2)$

Parameter Estimates:
 $\hat{\mu} = \bar{Y}_{\cdot\cdot}$
 $\hat{\alpha}_i = \bar{Y}_{i\cdot} - \bar{Y}_{\cdot\cdot}$
 $\hat{\varepsilon}_{ij} = Y_{ij} - \bar{Y}_{i\cdot}$
 $\hat{\sigma}^2 = MSE$



$$N = n_1 + n_2 + \dots + n_t$$

$$Y_{ij} = j^{\text{th}} \text{ obs. in } i^{\text{th}} \text{ treatment group}$$

$$\bar{Y}_{..} = \sum_{i=1}^t \sum_{j=1}^{n_i} Y_{ij} / N = \text{Grand Mean}$$

$$\bar{Y}_{i.} = \sum_{j=1}^{n_i} Y_{ij} / n_i = \text{mean of } i^{\text{th}} \text{ trt group}$$

$$SS(\text{Total}) = \sum_{i=1}^t \sum_{j=1}^{n_i} (Y_{ij} - \bar{Y}_{..})^2$$

Using Only Sample 1 and Sample 2

```
ANOVA
> file <- "http://www.uvm.edu/~rsingle/stat211/data/extra/ch6_BatteryLife.txt"
> dat <- read.table(file, header=T, na.strings=c("NA",""))
> dat.1.2 <- dat[dat$battery %in% 1:2,]
> mod2.aov <- aov(lifetime ~ battery, data=dat.1.2)
> summary(mod2.aov)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
battery	1	749.6	749.6	7.312	0.0124 *
Residuals	24	2460.4	102.5		

T-TEST

```
> t.test(lifetime ~ battery, data=dat2, var.equal=TRUE)

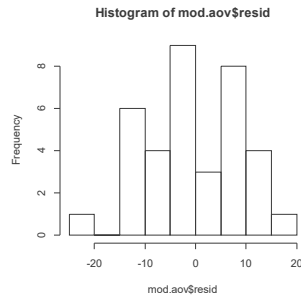
Two Sample t-test
data: lifetime by battery
t = 2.7041, df = 24, p-value = 0.01239 [Note: 2.7041^2 = 7.312]
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval: 2.550145, 18.991760
```

Using All 3 Samples

```
> mod3.aov <- aov(lifetime ~ as.factor(battery), data=dat)
> summary(mod3.aov)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
as.factor(battery)	2	1438	718.8	8.046	0.00142 **
Residuals	33	2948	89.3		

```
n lifetime.Mean lifetime.SD SE* Lower95% Upper95%
1 12 80.896667 9.385130 2.7285 75.345 86.448
2 14 70.125714 10.711163 2.5261 64.986 75.265
3 10 84.857000 7.361748 2.9889 78.776 90.938
*SE uses the pooled estimate of the error variance
SE = sqrt(MSE/n_i)
```



Post-Hoc Tests / Multiple Comparisons

Fisher's Protected T-tests (Least Significant Difference):

* Two stage procedure, with protection via overall ANOVA F-test (STOP if overall F is not significant)

Critical Value (for difference): $t_{\frac{\alpha}{2}, N-t}$ Note: this is the ordinary T critical value

$$\text{Rejection Region: } t_{s(ij)} = \frac{|\bar{Y}_{i.} - \bar{Y}_{j.}|}{\sqrt{MSE(1/n_i + 1/n_j)}} \geq t_{\frac{\alpha}{2}, N-t}$$

Tukey's HSD Procedure (Studentized Range Test):

Critical Value (for difference): $q_{\alpha}(k, df_{Error})$ for k tests using the Studentized Range Distribution

$$\text{Rejection Region: } \frac{|\bar{Y}_{i.} - \bar{Y}_{j.}|}{\sqrt{MSE/n}} \geq q_{\alpha}(k, df_{Error})$$

Fisher's LSD

```
pairwise.t.test(dat$lifetime, dat$battery, p.adj = "none") #No adjustment
p.adj = "bonf"
```

Level	Mean
3 A	84.857000
1 A	80.896667
2 B	70.125714

Levels not connected by same letter are significantly different. (unadjusted for Multiple Tests)

Level	Level	Difference	SE.diff	LowerCL	UpperCL	p.value(unadjusted)
3	2	14.73129	3.913428	6.76936	22.69322	0.0007*
1	2	10.77095	3.718329	3.20596	18.33595	0.0066*
3	1	3.960330	4.047031	-4.27341	12.19408	0.3349

Tukey's HSD

```
> dat$battery <- as.factor(dat$battery)
> mod.aov <- aov(lifetime ~ battery, data=dat)
> TukeyHSD(mod.aov)
> plot(TukeyHSD(mod.aov))
```

Tukey multiple comparisons of means
95% family-wise confidence level

Fit: aov(formula = lifetime ~ battery, data = dat)

\$battery	diff	lwr	upr	p adj
2-1	-10.770952	-19.894968	-1.646937	0.0177794
3-1	3.960333	-5.970249	13.890915	0.5953302
3-2	14.731286	5.128537	24.334035	0.0018431

95% family-wise confidence level

