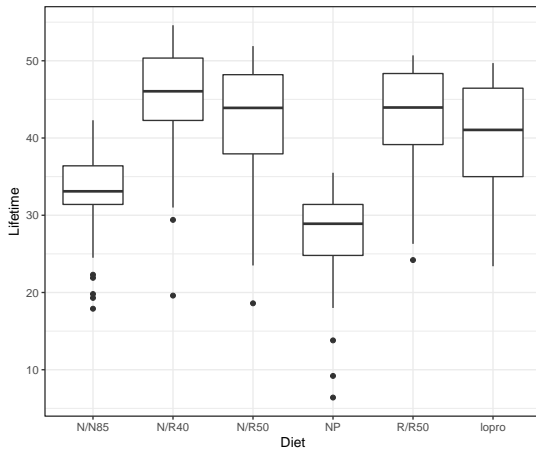


## Set R06 - ANOVA and F-tests

STAT 401 (Engineering) - Iowa State University

April 3, 2017

# Lifetime (months) of mice on different diets



# One-way ANOVA model/assumptions

$$Y_{ij} \overset{ind}{\sim} N(\mu_j, \sigma^2)$$

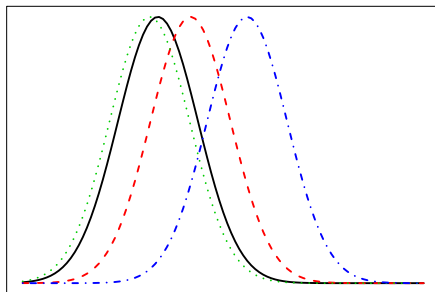
for  $j = 1, \dots, J$  and  $i = 1, \dots, n_j$ .

( $n_j$  means there can be different # of observations in each group)

Assumptions:

- Normality
  - Not skewed
  - Not heavy-tailed
- Common variance for all groups
- Independence
  - No cluster effects
  - No serial effects
  - No spatial effects

# ANOVA assumptions graphically



# What if you want to compare two groups?

We may still be interested in comparing two groups.

Statistical hypothesis: Is there a difference in mean lifetimes between the mice in two groups, e.g. NP and N/N85?

Statistical question: What is the difference in mean lifetimes between the mice in two groups, e.g. NP and N/N85?

## Two-group analysis

Begin with the two group (equal variance) model:

$$Y_{ij} \stackrel{ind}{\sim} N(\mu_j, \sigma^2)$$

with  $j = 1, 2$  and  $i = 1, \dots, n_j$

To perform a hypothesis test or a CI for the difference in means, the relevant quantities are:

- $\bar{Y}_2 - \bar{Y}_1$
- $SE(\bar{Y}_2 - \bar{Y}_1) = s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$
- $t$  distribution with  $n_1 + n_2 - 2$  degrees of freedom

where

$$s_p^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}$$

What if you have more than two groups?

# Multi-group analysis

The multi-group (equal variance) model:

$$Y_{ij} \overset{ind}{\sim} N(\mu_j, \sigma^2)$$

but now  $j = 1, \dots, J$  and  $i = 1, \dots, n_j$

( $n_j$  means there can be different # of observations in each group)

To perform a hypothesis test or a CI for the difference in means, the relevant quantities are:

- $\bar{Y}_2 - \bar{Y}_1$
- $SE(\bar{Y}_2 - \bar{Y}_1) = s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$
- $t$  distribution with  $n_1 + n_2 + \dots + n_J - J$  degrees of freedom

where

$$s_p^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2 + \dots + (n_J - 1)s_J^2}{n_1 + n_2 + \dots + n_J - J}$$

## Hypothesis test for comparison of two means (in multi-group data)

If  $Y_{ij} \stackrel{ind}{\sim} N(\mu_j, \sigma^2)$  for  $j = 1, \dots, J$  and we want to test the hypothesis

- $H_0 : \mu_1 = \mu_2$
- $H_1 : \mu_1 \neq \mu_2$

then we compute:

$$t = \frac{\bar{Y}_1 - \bar{Y}_2}{SE(\bar{Y}_1 - \bar{Y}_2)}$$

where

$$SE(\bar{Y}_1 - \bar{Y}_2) = s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$$

and

$$s_p^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2 + \dots + (n_J - 1)s_J^2}{n_1 + n_2 + \dots + n_J - J}.$$

Then we compare  $t$  to a  $t$  distribution with  $n_1 + n_2 + \dots + n_J - J$  degrees of freedom.



# Diet effect on mice lifetime

```
# A tibble: 6  4
  Diet      n    mean    sd
<fctr> <int>   <dbl>   <dbl>
1 N/N85    57 32.69123 5.125297
2 N/R40    60 45.11667 6.703406
3 N/R50    71 42.29718 7.768195
4      NP    49 27.40204 6.133701
5 R/R50    56 42.88571 6.683152
6 lopro    56 39.68571 6.991695
```

Test for difference in mean lifetime between NP and N/N85, i.e.

$$H_0 : \mu_4 = \mu_1 \text{ vs } H_1 : \mu_4 \neq \mu_1.$$

# Showing work

$$\begin{aligned}
 \bar{Y}_1 - \bar{Y}_4 &= 32.7 - 27.4 = 5.3 \\
 df &= 57 + 60 + 71 + 49 + 56 + 56 - 6 = 343 \\
 s_p^2 &= \frac{(57-1)5.1^2 + (60-1)6.7^2 + (71-1)7.8^2 + (49-1)6.1^2 + (56-1)6.7^2 + (56-1)7.0^2}{57+60+71+49+56+56-6} \\
 &= \frac{15314}{343} = 44.6 \\
 s_p &= \sqrt{s_p^2} = \sqrt{44.6} = 6.7 \\
 SE(\bar{Y}_1 - \bar{Y}_4) &= s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_4}} = 6.7 \sqrt{\frac{1}{57} + \frac{1}{49}} = 1.3 \\
 t &= \frac{\bar{Y}_1 - \bar{Y}_4}{SE(\bar{Y}_1 - \bar{Y}_4)} = \frac{5.3}{1.2} = 4.1 \\
 p &= 2P(t_{343} < -|t|) = 2P(t_{343} < -4.1) = 0.000052
 \end{aligned}$$

So we reject the null hypothesis that there is no difference between mean lifetime of mice on the NP and N/N85 diets.

## Confidence interval for the difference of two means (in multi-group data)

If  $Y_{ij} \stackrel{ind}{\sim} N(\mu_j, \sigma^2)$  for  $j = 1, \dots, J$ , a  $100(1 - \alpha)\%$  confidence interval for  $\mu_1 - \mu_2$  is

$$\bar{Y}_1 - \bar{Y}_2 \pm t_{n_1+n_2+\dots+n_J-J}(1 - \alpha/2)SE(\bar{Y}_1 - \bar{Y}_2)$$

where the  $t$  critical value,  $t_{n_1+n_2+\dots+n_J-J}(1 - \alpha/2)$ , needs to be calculated using a statistical software.

A 95% confidence interval for the difference in mean lifetime for N/N85 minus NP ( $\mu_1 - \mu_4$ ) is

$$5.3 \pm 1.96 \times 1.3 = (2.8, 7.8).$$

The statistical conclusion would be

*In this study, mice on the N/N85 diet lived an average of 5.3 months longer than mice on the NP diet (95% CI (2.8, 7.8)).*

# One-way ANOVA F-test

Are any of the means different?

Hypotheses in English:

$H_0$ : all the means are the same

$H_1$ : at least one of the means is different

Statistical hypotheses:

$$H_0 : \mu_j = \mu \text{ for all } j$$

$$Y_{ij} \stackrel{iid}{\sim} N(\mu, \sigma^2)$$

$$H_1 : \mu_j \neq \mu_{j'} \text{ for some } j \text{ and } j'$$

$$Y_{ij} \stackrel{ind}{\sim} N(\mu_j, \sigma^2)$$

An ANOVA table organizes the relevant quantities for this test and computes the pvalue.

# ANOVA table

A start of an ANOVA table:

Source of variation	Sum of squares	d.f.	Mean square
Factor A (Between groups)	$SSA = \sum_{j=1}^J n_j (\bar{Y}_j - \bar{Y})^2$	$J - 1$	$\frac{SSA}{J-1}$
Error (Within groups)	$SSE = \sum_{j=1}^J \sum_{i=1}^{n_j} (Y_{ij} - \bar{Y}_j)^2$	$n - J$	$\frac{SSE}{n-J} (= s_p^2)$
Total	$SST = \sum_{j=1}^J \sum_{i=1}^{n_j} (Y_{ij} - \bar{Y})^2$	$n - 1$	

where

- $J$  is the number of groups,
- $n_j$  is the number of observations in group  $j$ ,
- $n = \sum_{j=1}^J n_j$  (total observations),
- $\bar{Y}_j = \frac{1}{n_j} \sum_{i=1}^{n_j} Y_{ij}$  (average in group  $j$ ),
- and  $\bar{Y} = \frac{1}{n} \sum_{j=1}^J \sum_{i=1}^{n_j} Y_{ij}$  (overall average).

# ANOVA table

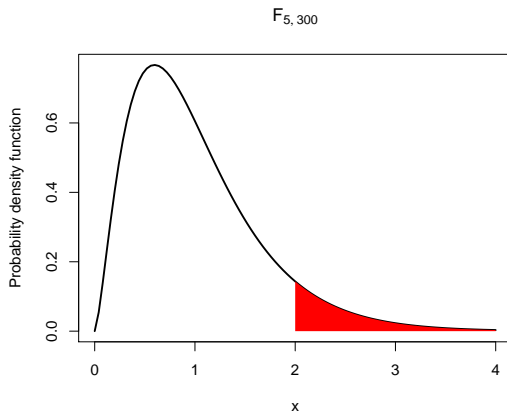
An easier to remember ANOVA table:

Source of variation	Sum of squares	df	Mean square	F-statistic	p-value
Factor A (between groups)	SSA	$J - 1$	$MSA = SSA/J - 1$	$MSA/MSE$	(see below)
Error (within groups)	SSE	$n - J$	$MSE = SSE/n - J$		
Total	$SST=SSA+SSE$	$n - 1$			

Under  $H_0$ ,

- the quantity  $MSA/MSE$  has an F-distribution with  $J - 1$  numerator and  $n - J$  denominator degrees of freedom,
- larger values of  $MSA/MSE$  indicate evidence against  $H_0$ , and
- the p-value is determined by  $P(F_{J-1, n-J} > MSA/MSE)$ .

# F-distribution



# One-way ANOVA F-test (by hand)

```
# A tibble: 6  4
  Diet      n    mean    sd
<fctr> <int>  <dbl>  <dbl>
1  N/N85    57 32.69123 5.125297
2  N/R40    60 45.11667 6.703406
3  N/R50    71 42.29718 7.768195
4    NP     49 27.40204 6.133701
5  R/R50    56 42.88571 6.683152
6  loopro   56 39.68571 6.991695

      n    mean
1 349 38.79713
```

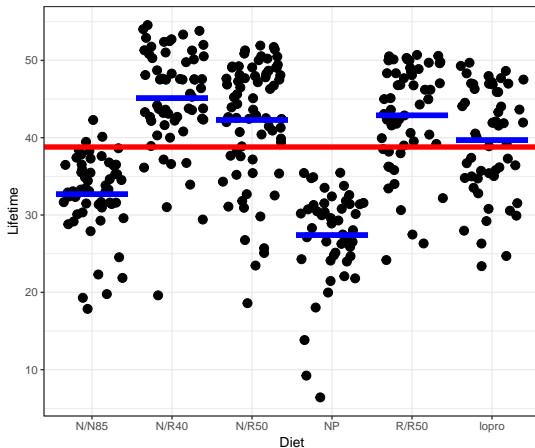
So

$$\begin{aligned}
 SSA &= 57 \times (32.7 - 38.8)^2 + 60 \times (45.1 - 38.8)^2 + 71 \times (42.3 - 38.8)^2 + 49 \times (27.4 - 38.8)^2 \\
 &\quad + 56 \times (42.9 - 38.8)^2 + 56 \times (39.7 - 38.8)^2 = 12734 \\
 SST &= (35.5 - 38.8)^2 + (35.4 - 38.8)^2 + (34.9 - 38.8)^2 + \dots + (19.6 - 38.8)^2 + (47.6 - 38.8)^2 = 28031 \\
 SSE &= SST - SSA = 28031 - 12734 = 15297 \\
 J - 1 &= 5 \\
 n - J &= 349 - 6 = 343 \\
 n - 1 &= 348 \\
 MSA &= SSA / J - 1 = 12734 / 5 = 2547 \\
 MSE &= SSE / n - J = 15297 / 343 = 44.6 = s_p^2 \\
 F &= MSA / MSE = 2547 / 44.6 = 57.1 \\
 p &= P(F_{5,343} > 57.1) < 0.0001
 \end{aligned}$$



# As a picture

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# R code and output for one-way ANOVA

```
m <- lm(Lifetime~Diet, case0501)
anova(m)
```

Analysis of Variance Table

Response: Lifetime

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
Diet	5	12734	2546.8	57.104	< 2.2e-16 ***
Residuals	343	15297	44.6		

---

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# General F-tests

The one-way ANOVA F-test is an example of a general hypothesis testing framework that uses F-tests. This framework can be used to test

- composite alternative hypotheses or, equivalently,
- a full vs a reduced model.

The general idea is to balance the amount of variability remaining when moving from the reduced model to the full model measured using the sums of squared errors (SSEs) relative to the amount of complexity, i.e. parameters, added to the model.

# Simple vs Composite Hypotheses

Suppose

$$Y_{ij} \stackrel{\text{ind}}{\sim} N(\mu_j, \sigma^2)$$

for  $j = 1, \dots, 3$  then a **simple hypothesis** is

- $H_0 : \mu_1 = \mu_2$
- $H_1 : \mu_1 \neq \mu_2$

and a **composite hypothesis** is

- $H_0 : \mu_1 = \mu_2 = \mu_3$
- $H_1 : \mu_j \neq \mu_{j'}$  for some  $j$  and  $j'$

since there are four possibilities under  $H_1$

- $\mu_1 = \mu_2 \neq \mu_3$
- $\mu_2 = \mu_3 \neq \mu_1$
- $\mu_3 = \mu_1 \neq \mu_2$
- none of  $\mu_1, \mu_2, \mu_3$  are equal

# Testing Composite hypotheses

If  $Y_{ij} \stackrel{\text{ind}}{\sim} N(\mu_j, \sigma^2)$  for  $j = 1, \dots, J$  and we want to test the **composite hypothesis**

- $H_0 : \mu_j = \mu$  for all  $j$
- $H_1 : \mu_j \neq \mu_{j'}$  for some  $j$  and  $j'$

think about this as two models:

- $H_0 : Y_{ij} \stackrel{\text{ind}}{\sim} N(\mu, \sigma^2)$  (**reduced**)
- $H_1 : Y_{ij} \stackrel{\text{ind}}{\sim} N(\mu_j, \sigma^2)$  (**full**)

We can use an F-test to calculate a p-value for tests of this type.

## Nested models: full vs reduced

### Definition

Two models are **nested** if the **reduced** model is a special case of the **full** model.

For example, consider the full model

$$Y_{ij} \stackrel{\text{ind}}{\sim} N(\mu_j, \sigma^2).$$

One special case of this model occurs when  $\mu_j = \mu$  and thus

$$Y_{ij} \stackrel{\text{ind}}{\sim} N(\mu, \sigma^2).$$

is a reduced model and these two models are nested.

# Calculating the sum of squared residuals (errors)

Model	<i>Full</i>	<i>Reduced</i>
Assumption	$H_1 : Y_{ij} \stackrel{ind}{\sim} N(\mu_j, \sigma^2)$	$H_0 : Y_{ij} \stackrel{iid}{\sim} N(\mu, \sigma^2)$
Mean	$\hat{\mu}_j = \bar{Y}_j = \frac{1}{n_j} \sum_{i=1}^{n_j} Y_{ij}$	$\hat{\mu} = \bar{Y} = \frac{1}{n} \sum_{j=1}^J \sum_{i=1}^{n_j} Y_{ij}$
Residual	$r_{ij} = Y_{ij} - \hat{\mu}_j = Y_{ij} - \bar{Y}_j$	$r_{ij} = Y_{ij} - \hat{\mu} = Y_{ij} - \bar{Y}$
SSE	$\sum_{j=1}^J \sum_{i=1}^{n_j} r_{ij}^2$	$\sum_{j=1}^J \sum_{i=1}^{n_j} r_{ij}^2$

# General F-tests

Do the following

1. Calculate

Extra sum of squares =

Residual sum of squares (reduced) - Residual sum of squares (full)

2. Calculate

Extra degrees of freedom =

# of mean parameters (full) - # of mean parameters (reduced)

3. Calculate F-statistics

$$F = \frac{\text{Extra sum of squares} / \text{Extra degrees of freedom}}{\hat{\sigma}_{full}^2}$$

4. Pvalue is  $P(F_{ndf,ddf} > F)$

- numerator degrees of freedom (nnd) = Extra degrees of freedom
- denominator degrees of freedom (ddf) =  $n$  - # of mean parameters (full)



## Example

Recall the mice data set.

Consider the hypothesis that all diets have a common mean lifetime except NP.

Let

$$Y_{ij} \stackrel{\text{ind}}{\sim} N(\mu_j, \sigma^2)$$

with  $j = 1$  being the NP group then the hypotheses are

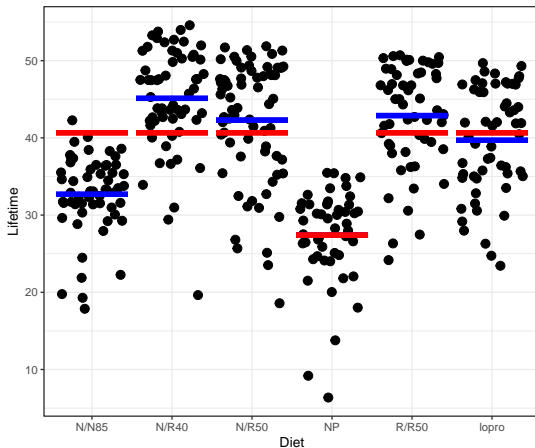
- $H_0 : \mu_j = \mu$  for  $j \neq 1$
- $H_1 : \mu_j \neq \mu_{j'}$  for some  $j, j' = 2, \dots, 6$

As models:

- $H_0 : Y_{i1} \sim N(\mu_1, \sigma^2)$  and  $Y_{ij} \sim N(\mu, \sigma^2)$  for  $j \neq 1$
- $H_1 : Y_{ij} \sim N(\mu_j, \sigma^2)$

# As a picture

```
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```



# Making R do the calculations

```
case0501$NP = factor(case0501$Diet == "NP")

modR = lm(Lifetime~NP, case0501)
modF = lm(Lifetime~Diet, case0501)
anova(modR,modF)
```

Analysis of Variance Table

Model 1: Lifetime ~ NP

Model 2: Lifetime ~ Diet

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	347	20630				
2	343	15297	4	5332.2	29.89	< 2.2e-16 ***

---

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# Are there differences in means amongst low calorie diets?

Let  $Y_{ij}$  be the lifetime in months for mouse  $i$  in group  $j$  where the groups are N/N85 ( $j=1$ ), N/R40 ( $j=2$ ), N/R50 ( $j=3$ ), NP ( $j=4$ ), R/R50 ( $j=5$ ), and lopro ( $j=6$ ). Assume

$$Y_{ij} \stackrel{\text{ind}}{\sim} N(\mu_j, \sigma^2)$$

and test the hypotheses

$$H_0: \mu_2 = \mu_3 = \mu_5 = \mu_6$$

$H_1$ : at least one of  $\mu_2, \mu_3, \mu_5, \mu_6$  is different from the rest

Implicitly, we are allowing  $\mu_1$  and  $\mu_4$  to be different from the others.

# Making R do the calculations

```
case0501$local = ifelse(case0501$Diet=='N/N85', 1, 2) # NP is 2 here
case0501$local[case0501$Diet=='NP'] = 0             # now NP is 1
case0501$local = factor(case0501$local)
mod1 = lm(Lifetime~1, case0501)
modR = lm(Lifetime~local, case0501)
modF = lm(Lifetime~Diet, case0501)
anova(mod1, modR, modF)
```

## Analysis of Variance Table

Model 1: Lifetime ~ 1

Model 2: Lifetime ~ local

Model 3: Lifetime ~ Diet

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	348	28031				
2	346	16163	2	11868.5	133.0585	< 2.2e-16 ***
3	343	15297	3	865.4	6.4682	0.0002863 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

```
anova(modF) # To get the pooled estimate of the variance for the full model
```

## Analysis of Variance Table

Response: Lifetime

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
Diet	5	12734	2546.8	57.104	< 2.2e-16 ***
Residuals	343	15297	44.6		

---

# Lack-of-fit F-test

Let  $Y_{ij}$  be the  $i^{th}$  observation from the  $j^{th}$  group where the group is defined by those observations having the same explanatory variable value ( $X_j$ ).

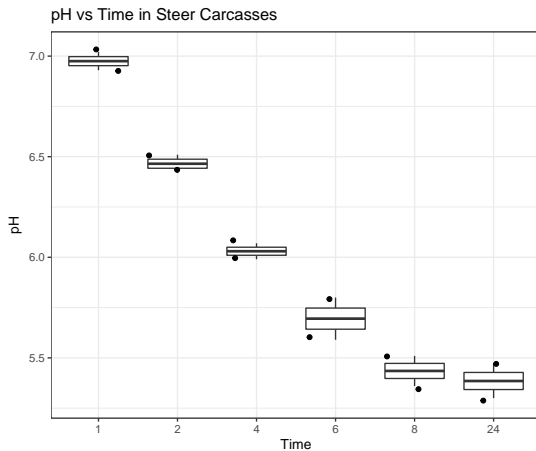
Two models:

ANOVA:  $Y_{ij} \stackrel{ind}{\sim} N(\mu_j, \sigma^2)$  (full)

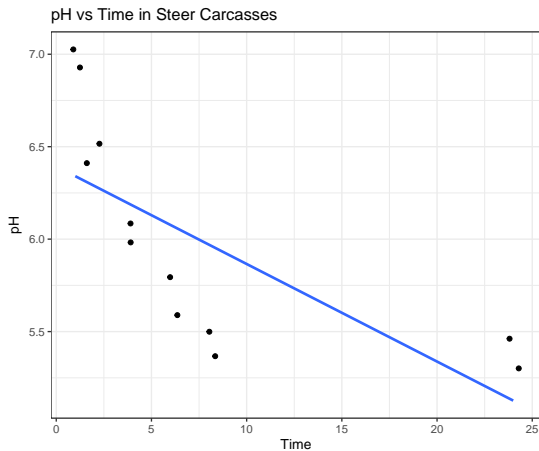
Regression:  $Y_{ij} \stackrel{ind}{\sim} N(\beta_0 + \beta_1 X_j, \sigma^2)$  (reduced)

- Regression model is reduced:
  - ANOVA has  $J$  parameters for the mean
  - Regression has 2 parameters for the mean
  - Set  $\mu_j = \beta_0 + \beta_1 X_j$ .
- Small pvalues indicate a lack-of-fit, i.e. the reduced model is not adequate.
- Lack-of-fit F-test requires multiple observations at a few  $X_j$  values!

# pH vs Time - ANOVA



# pH vs Time - Regression





```
# Use as.factor to turn a continuous variable into a categorical variable
m_anova = lm(pH ~ as.factor(Time), Sleuth3::ex0816)
m_reg   = lm(pH ~ Time, Sleuth3::ex0816)
anova(m_reg, m_anova)
```

#### Analysis of Variance Table

Model 1: pH ~ Time

Model 2: pH ~ as.factor(Time)

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	10	1.97289				
2	6	0.05905	4	1.9138	48.616	0.0001048 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

# Summary

- Use t-tests for simple hypothesis tests and CIs
- Use F-tests for composite hypothesis tests
  - One-way ANOVA F-test
  - General F-tests
  - Lack-of-fit F-tests

Think about F-tests as comparing models.