

# DATA ANALYSIS

Week 11: F-tests/ANOVAs for nominal data

# lunch with Psychology faculty!



## Lunch with Psychology Faculty

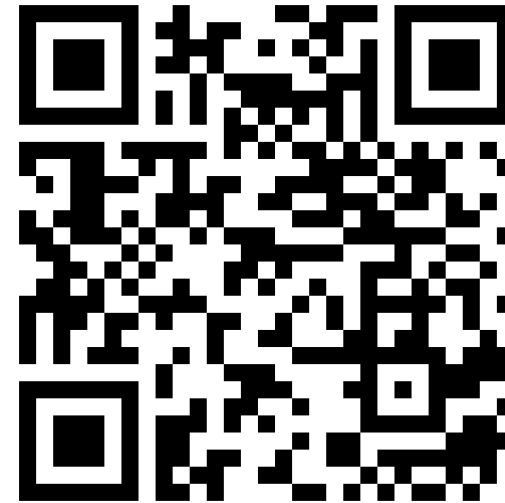
The Psychology Department is hosting lunches with faculty and students this semester.

All lunches will be in **Thorne Dining!** Please meet us at the check-in station at the times mentioned for the specific dates.

The lunches are on the following dates/times:

- Wednesday, February 21 2024 (**12 pm**): Prof. Erika Nyhus and Prof. Hannah Reese
- Tuesday, March 5 2024 (**12 pm**): Prof. Kacie Armstrong, Prof. Suzanne Lovett, and Prof. Thomas Small
- Friday, April 12 2024 (**1.10 pm**): Prof. Abhilasha Kumar and Prof. Samuel Putnam

We look forward to seeing you!



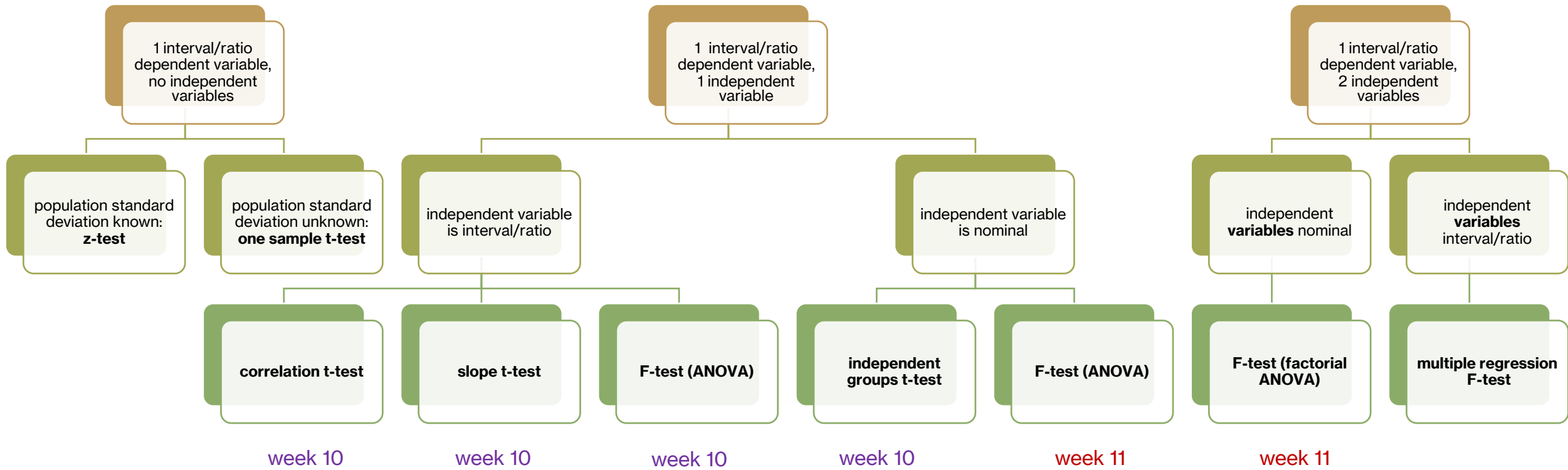
# logistics

- no week 11 or week 12 quiz!
- office hours only 2-4 today
- problem set #4 second attempt due tonight
- problem set #6 (due Apr 9)
- class survey (due Apr 9, Canvas)
- midterm 2 review materials (on Canvas by Friday):
  - midterm 2 practice quiz
  - midterm 2 computational questions
- Apr 10 is a review class for midterm 2

11	W: April 3, 2024	<a href="#">W11: Modeling Relationships II</a>
11	F: April 5, 2024	W11 continued...
12	T: April 9, 2024	<b>Problem Set 6 due</b>
12	W: April 10, 2024	<a href="#">W12: Loose Ends / Exam 2 review</a>
12	F: April 12, 2024	<b>Exam (Midterm) 2</b>
13	W: April 17, 2024	<a href="#">W13: Non-Independent Data</a>
13	F: April 19, 2024	W13 continued...
14	T: April 23, 2024	<b>Problem Set Opt-out Deadline 3</b>
14	W: April 24, 2024	<a href="#">W14: Miscellaneous Data</a>
14	F: April 26, 2024	W14 continued...
15	T: April 30, 2024	<b>Problem Set 7 due</b>
15	W: May 1, 2024	<a href="#">W15: Odds and Ends</a>
15	F: May 3, 2024	<b>Final Exam</b>
16	W: May 8, 2024	<b>Wrapping Up!</b>

# the chart of doom (so far)

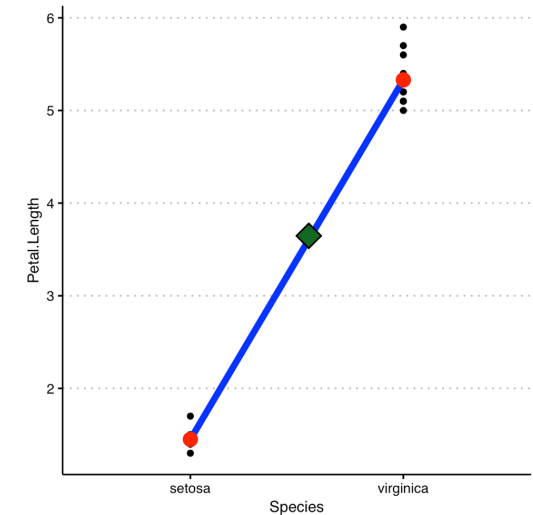
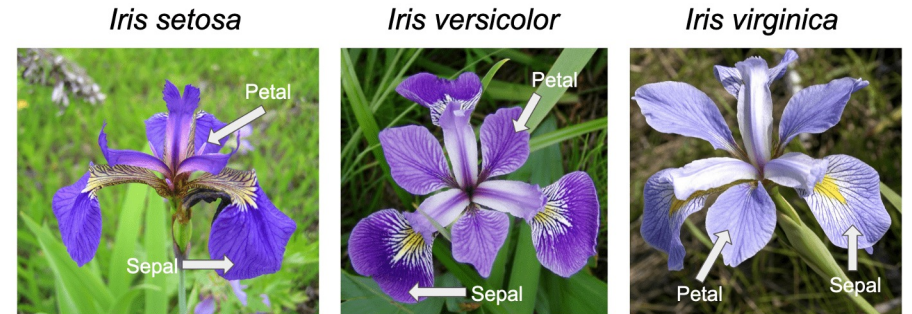
week 7



only for two groups!

# review: iris dataset

- our goal is to build the **best model for petal lengths**
- petal length ( $Y$ ) =  $a + b$  (species: $X$ )
- we conducted an **independent groups t-test** to evaluate the significance of adding a “species” mean to our model over and above the grand mean
- t-test compared the  $M_{setosa}$  to  $M_{virginica}$





# review: independent groups t-test

step 1:  
state the  
hypotheses

$$H_0: \beta = 0 \text{ or } \mu_2 - \mu_1 = 0$$

$$H_1: \beta \neq 0 \text{ or } \mu_2 - \mu_1 \neq 0$$

step 2:  
set criteria  
for decision

$\alpha = .05$   
find  $t_{critical}$  based on  
one vs. two tailed  
test and degrees of  
freedom  
 $df = n_1 + n_2 - 2$

step 3:  
collect  
data

- (1) compute  $s_{M_2 - M_1} = \sqrt{\frac{s_p^2}{n_1} + \frac{s_p^2}{n_2}}$   
and  $b = M_2 - M_1$
- (2) compute  $t_{observed} = \frac{b - \beta}{s_{M_2 - M_1}}$
- (3) find p-value for t-score

step 4:  
make a  
decision!

check whether  $t_{observed}$  is  
beyond  $t_{critical}$  and  
p-value < .05. if so, reject  
null hypothesis!

# review: independent groups t-test

- **step 1: state the hypotheses**

- $H_0: \mu_{virginica} - \mu_{setosa} = 0$ : mean petal lengths for both species are equal, i.e.,
- $H_1: \mu_{virginica} - \mu_{setosa} \neq 0$ : mean petal lengths for both species are not equal, i.e.,

- **step 2: set criteria for decision**

$$df = n_1 + n_2 - 2 = 10 + 10 - 2 = 18$$

$$t_{18} = t_{critical} = 2.1009$$

- **step 3: collect data**

$$s_p^2 = \frac{SS_1 + SS_2}{df_1 + df_2} = .0525$$

$$s_{M_2 - M_1} = \sqrt{\frac{s_p^2}{n_1} + \frac{s_p^2}{n_2}} = .1025$$

$$b = M_2 - M_1 = 3.88$$

$$t_{observed} = \frac{b - 0}{s_{M_2 - M_1}} = \frac{(3.88) - 0}{.1025} = 37.844 \text{ and p-value} < .0001$$

- **step 4: make a decision!**

# assumptions: t-test

$$s_p^2 = \frac{SS_1 + SS_2}{df_1 + df_2}$$
$$s_{M2-M1} = \sqrt{\frac{s_p^2}{n_1} + \frac{s_p^2}{n_2}}$$

- interval/ratio dependent variable
- independent observations (between-subjects design)
- normality
  - when data are not normal, the t-test is not appropriate
  - BUT: t-tests are fairly robust to minor violations for large n
- homogeneity of variances
  - we assume that the populations from which samples are drawn have equal variances to compute a “pooled” estimate of variance for the independent groups t-test
  - Welch’s test is done for unequal variances





**questions?**

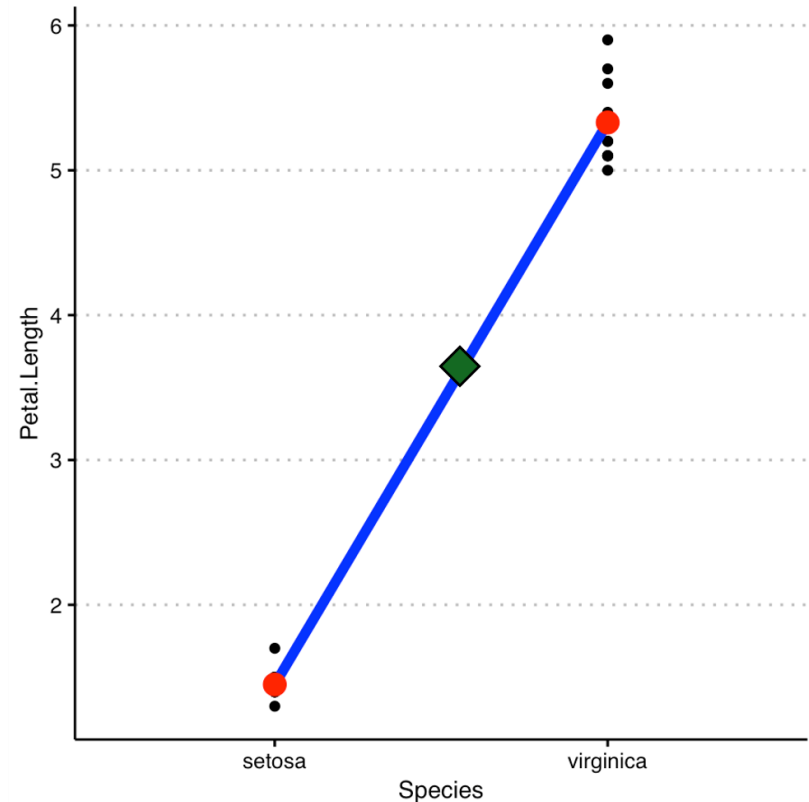
# F-test for two groups

- just as we did an “overall” test for linear regression, we can do the same here for the iris dataset, where we **compare** the grand mean model with the species mean model

- recall that  $F = \frac{MS_{model}}{MS_{error}} = \frac{SS_{model}/df_{model}}{SS_{error}/df_{error}}$

- and  $SS_{total} = SS_{model} + SS_{error}$

- how do we obtain  $SS_{total}$ ,  $SS_{error}$ , and  $SS_{model}$ ?



# F-test for two groups

- $SS_{total}$  represents score deviations from grand mean ( $M_Y$ )

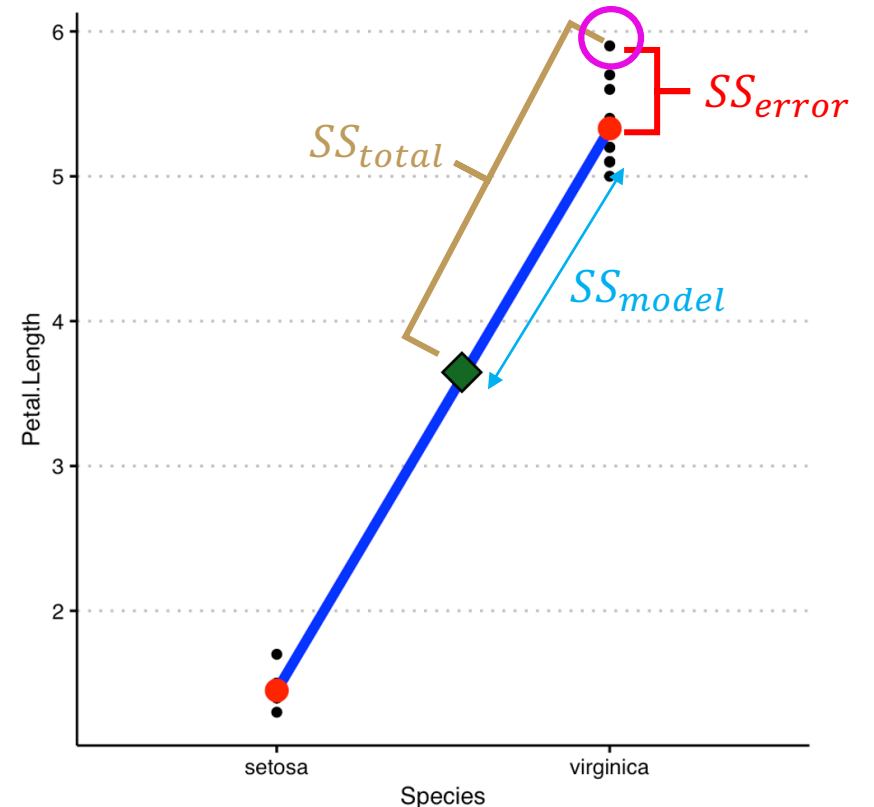
$$SS_{total} = \sum (Y - M_Y)^2$$

- $SS_{error}$  represents the deviations of each score from its group mean

$$SS_{error} = \sum (Y - \hat{Y})^2 = \sum (Y - M_{group})^2$$

- $SS_{model}$  represents the **gains** we get if we substitute each score with the group mean instead of the grand mean

$$SS_{model} = \sum \sum n_i (M_{group} - M_Y)^2 = SS_{total} - SS_{error}$$



# NHST for two independent groups (F-test)

step 1:  
state the  
hypotheses

$$H_0: \mu_2 - \mu_1 = 0$$
$$H_1: \mu_2 - \mu_1 \neq 0$$

step 2:  
set criteria  
for decision

$$\alpha = .05$$

find  $F_{critical}$  based  
on **right** tailed test  
and degrees of  
freedom

$$df_1 = k - 1$$
$$df_2 = n - k$$

step 3:  
collect  
data

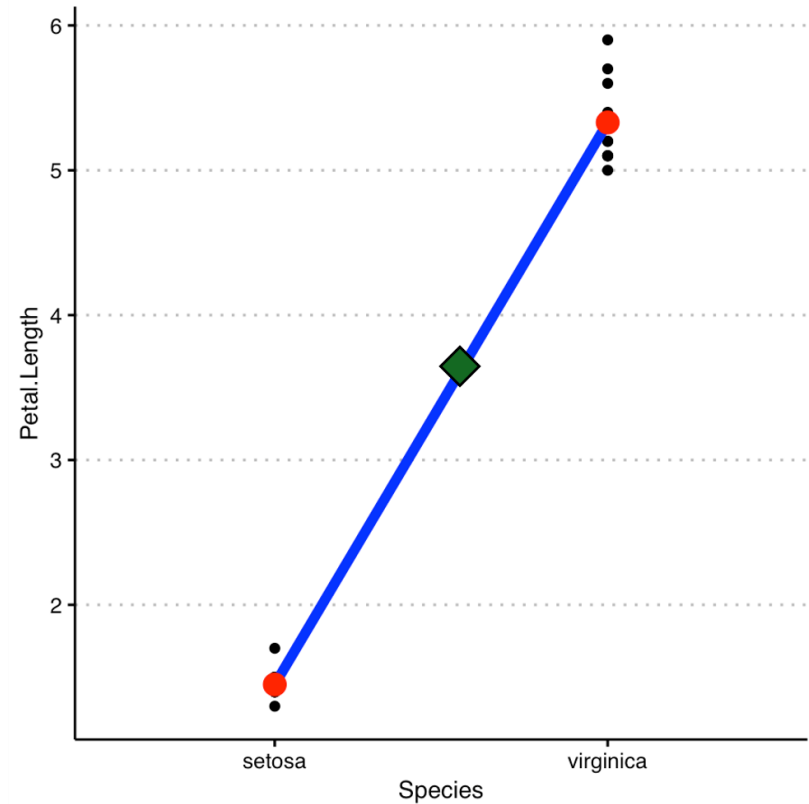
- (1) compute  $SS_{model}$  and  $SS_{error}$
- (2) compute  $F_{observed} = \frac{MS_{model}}{MS_{error}}$
- (3) find p-value for F-score

step 4:  
make a  
decision!

check whether  $F$  is  
beyond  $F_{critical}$  and  
p-value  $< .05$ . if so, reject  
null hypothesis!

# activity: F-test for iris dataset

- conduct the F test for the [iris dataset](#)



# F-test for iris dataset

- **step 1: state the hypotheses**

- $H_0: \mu_{virginica} - \mu_{setosa} = 0$ : petal lengths for both species are equal
- $H_1: \mu_{virginica} - \mu_{setosa} \neq 0$ : petal lengths for species are different

- **step 2: set criteria for decision**

$k = 2$  : number of levels of independent variable OR estimated parameters

$$df_1 = k - 1 = 2 - 1 = 1$$

$$df_2 = n - k = 20 - 2 = 18$$

$$F(df_1, df_2) = F(1, 18) = F_{critical} = 4.414$$

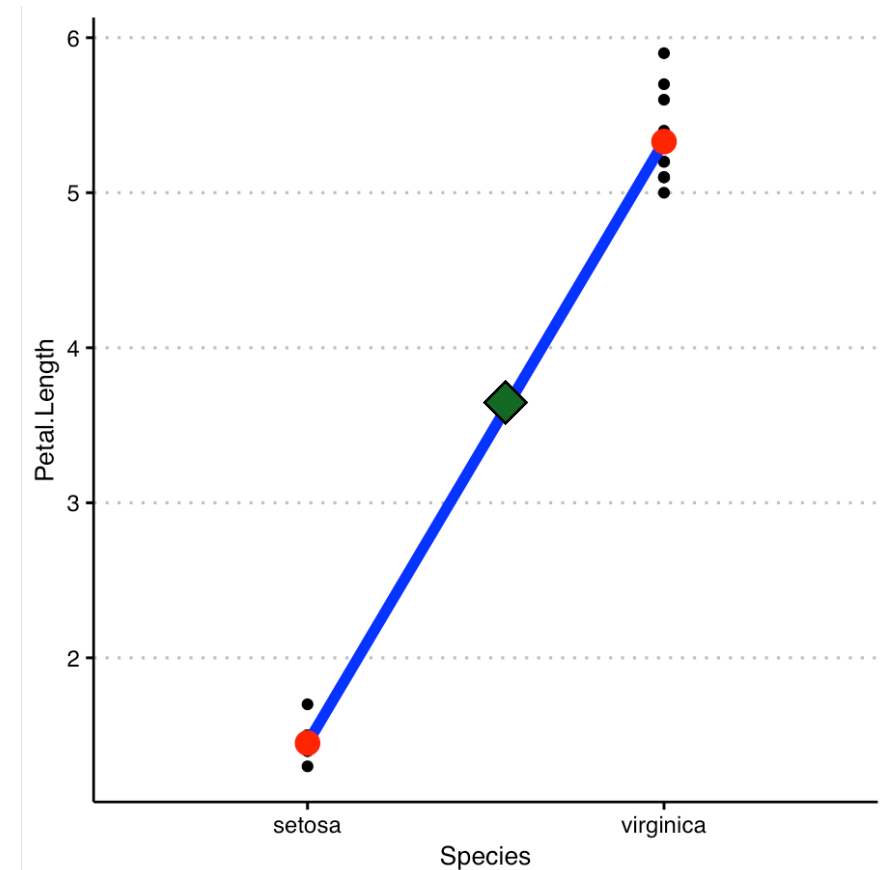


## step 3a: obtaining $SS_{total}$

- what is  $SS_{total}$ ?  $SS_{total}$  is the error left over after the grand mean has been fit to the data

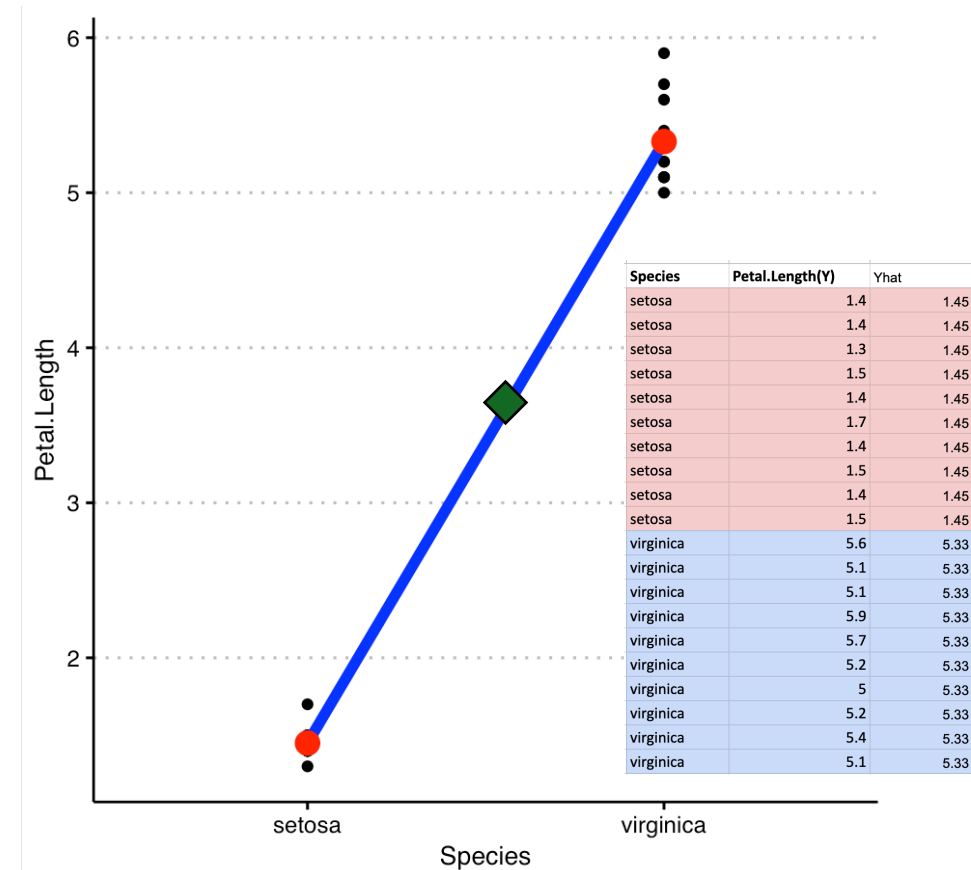
$$SS_{total} = \sum (Y - M_Y)^2$$

- for iris,  $SS_{total} = 76.218$



## step 3b: obtaining $SS_{error}$

- $SS_{error}$  is the error that is left over after our species model has been fit
- our species model substitutes each raw score with the mean of the specific species
- $SS_{error} = \sum(Y - \hat{Y})^2 = \sum(Y - M_{group})^2$
- for iris,  $SS_{error} = .946$



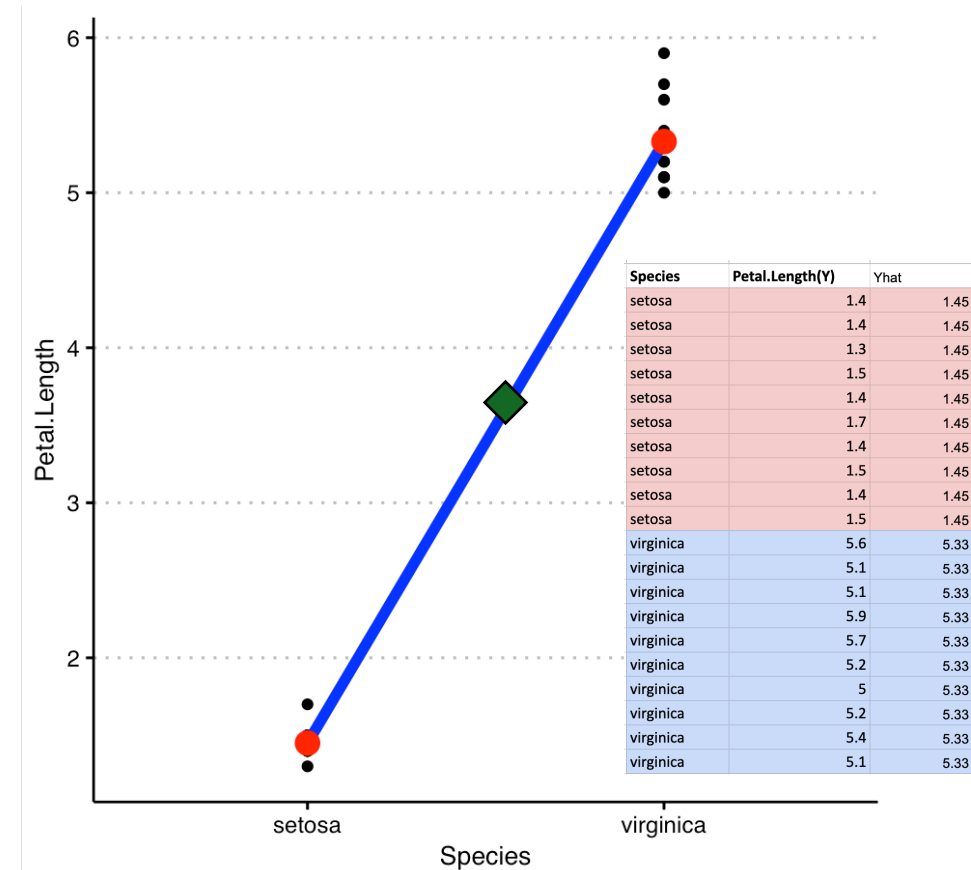
## step 3c: obtaining $SS_{model}$

- how can we obtain  $SS_{model}$ ?

$$SS_{total} = SS_{model} + SS_{error}$$

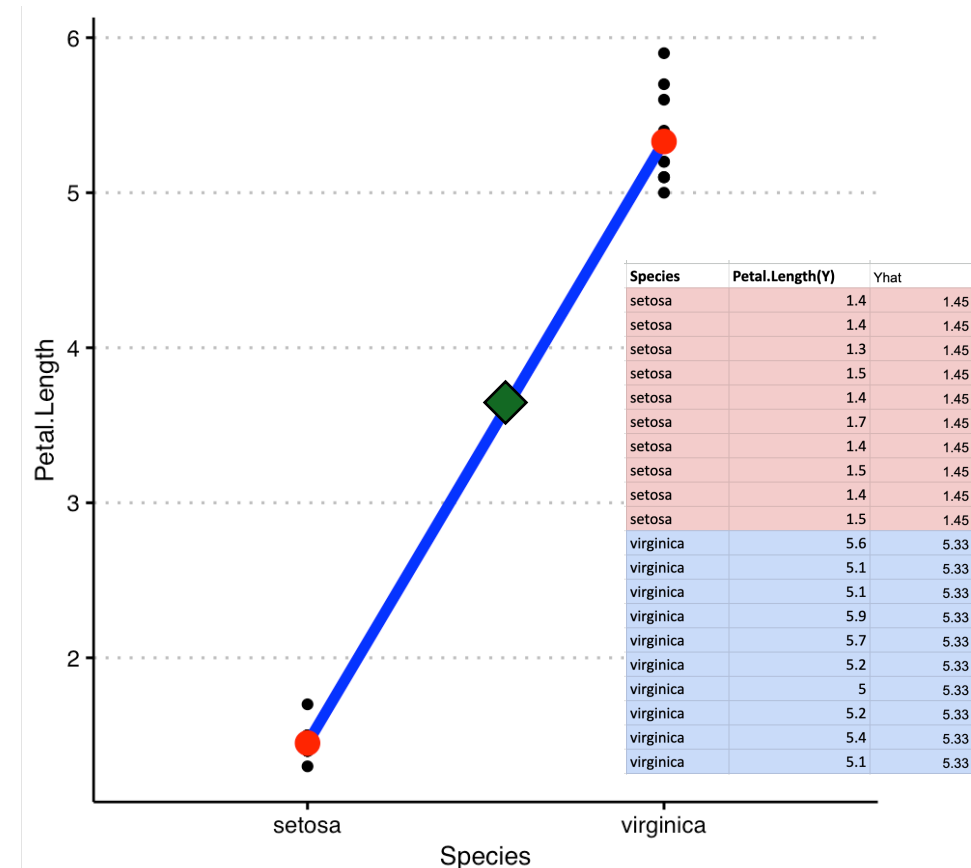
$$\text{thus, } SS_{model} = SS_{total} - SS_{error}$$

- for iris,  $SS_{total} = 76.218$  and  $SS_{error} = .946$
- $SS_{model} = 75.272$



## step 3d: obtaining $F_{observed}$

- $F_{observed} = \frac{MS_{model}}{MS_{error}} = \frac{SS_{model}/df_{model}}{SS_{error}/df_{error}} = 1432.24$
- p-value = <.0001
- $F_{critical} = 4.414$
- thus,  $F(1,18) = 1432.24, p < .0001$ 
  - we can reject the null hypothesis
  - petal lengths of setosa and virginica are significantly different
  - also,  $t^2 = F!!$



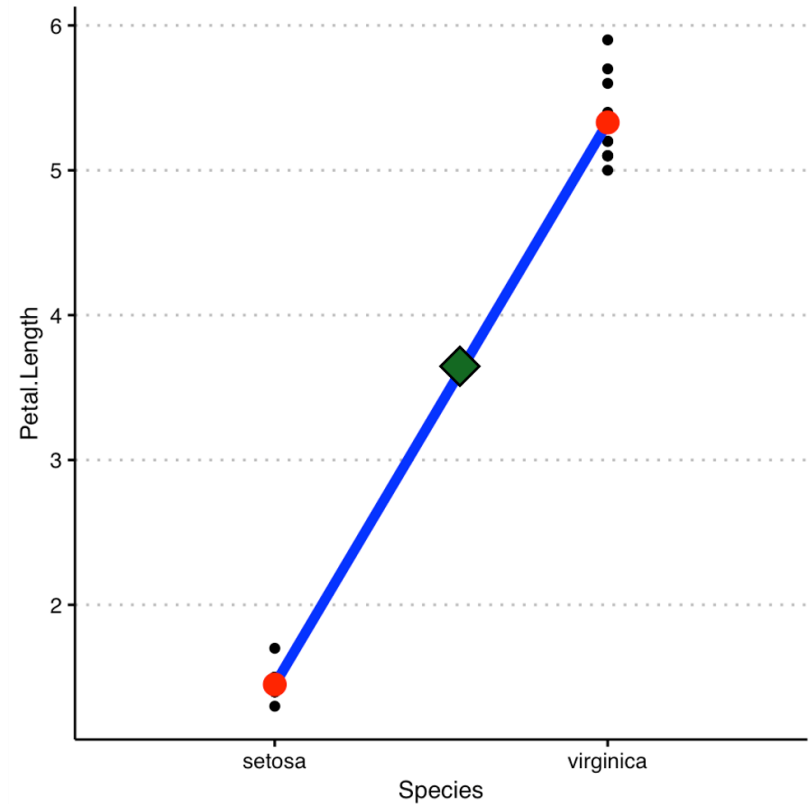
# F-table

		SS	df	MS	F	p-value
$SS_{model}$	species	75.272	1	75.272	1432.24	<.0001
$SS_{error}$	residual	0.946	18	0.0526		

```
Response: Petal.Length
      Sum Sq Df F value    Pr(>F)
Species  75.272  1  1432.2 < 2.2e-16 ***
Residuals  0.946 18
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
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# testing more than two groups

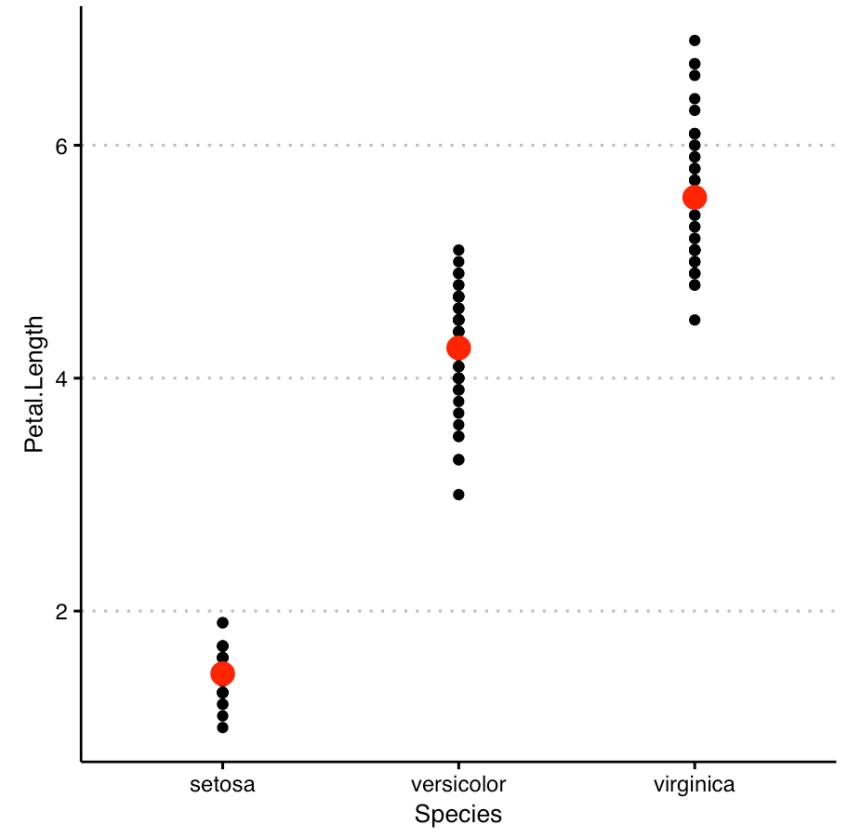
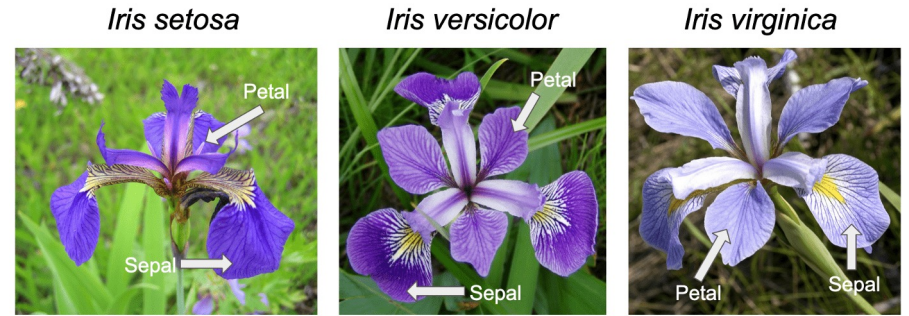
- an independent groups t-test is a **special case of linear models** ( $Y = a + bX$ ) when  $X$  is nominal
- it is *also* a special case of only **comparing two groups**
- example of comparing **more than two groups?**



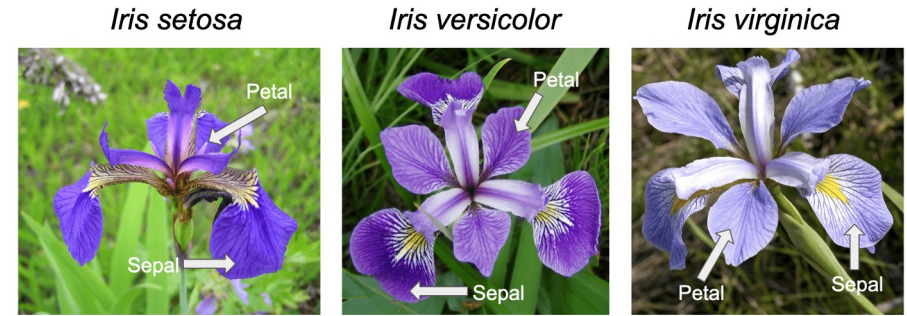


# revisiting iris

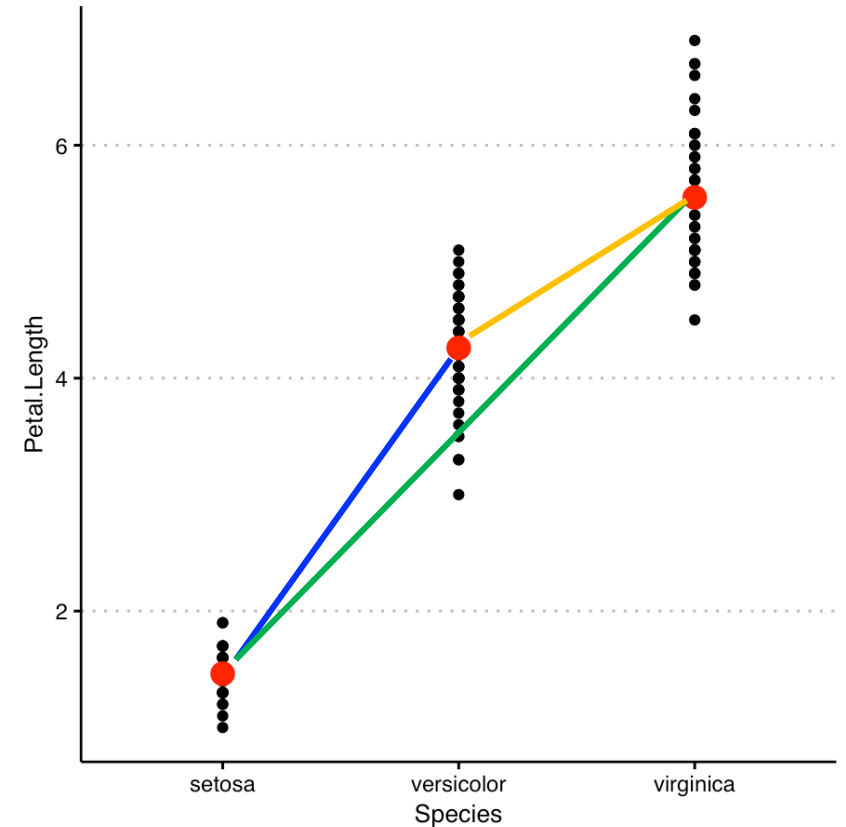
- recall that the iris dataset actually contains information about **three** species (setosa, virginica, and versicolor)



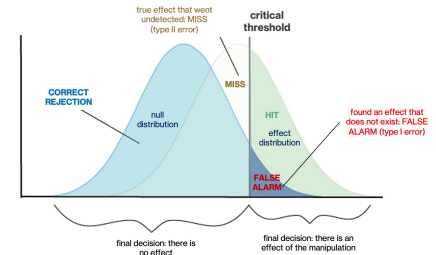
# revisiting iris



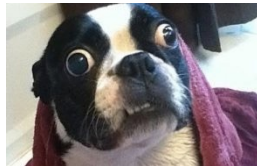
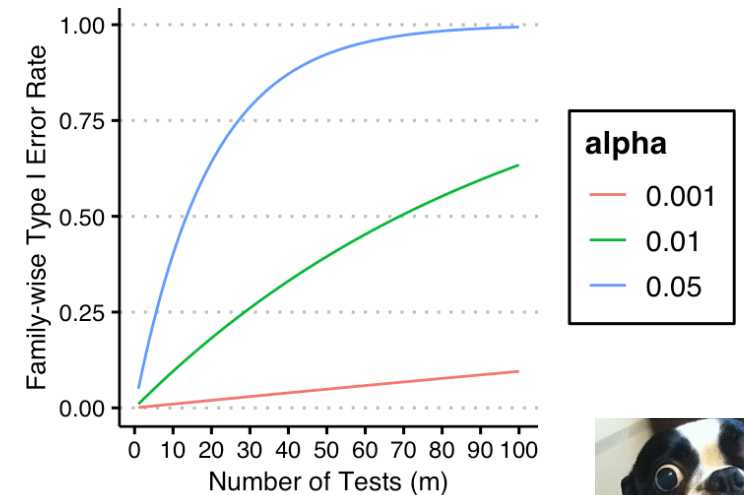
- what if we wanted to look at all three species?
- how many possible comparisons are involved?
  - $M_{virginica} - M_{setosa}$
  - $M_{versicolor} - M_{setosa}$
  - $M_{virginica} - M_{versicolor}$
- we could fit individual linear models for each comparison and conduct the t-test/F-test for each comparison



# multiple tests and type I errors

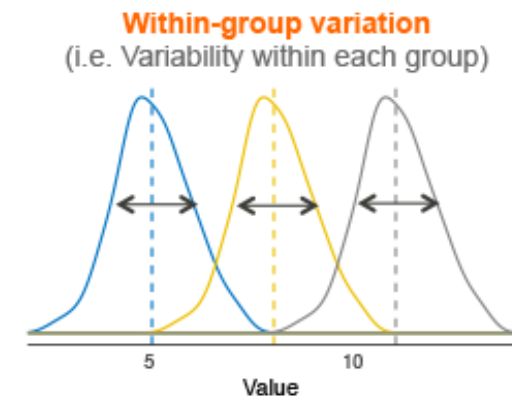
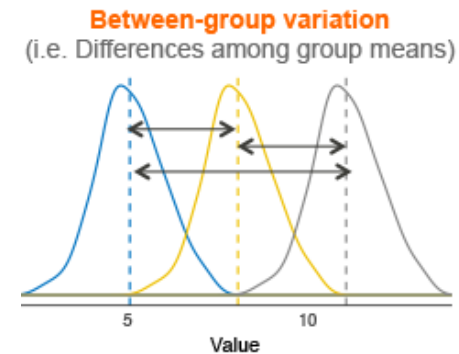


- each time a hypothesis test is conducted with some  $\alpha$ -level, there is  $\alpha$  % probability of making a type I error
- as more tests are conducted, this probability increases
  - P(type I error in one test) =  $\alpha$
  - P (no type I error in one test) =  $1 - \alpha$
  - P (no type I error in  $m$  tests) =  $(1 - \alpha)^m$
  - P (at least one type I error in  $m$  tests) =  $1 - (1 - \alpha)^m$
- two solutions
  - correct for multiple comparisons
  - do an “overall” test before jumping in



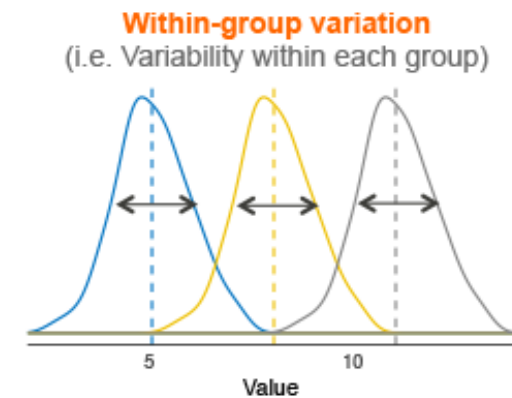
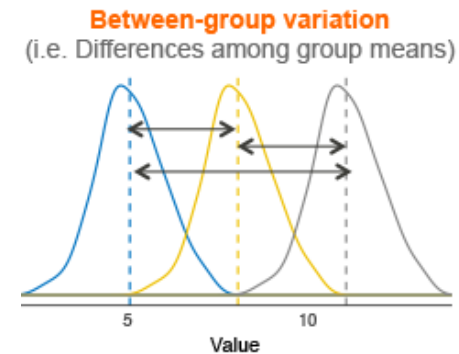
# F test aka ANOVA!

- we already saw how an analysis of variance / F-test can help us assess “overall” fit of the model
- formally, ANOVA is a *generalized* t-test for more than two means/groups!
- we first evaluate whether the overall model explains variance over and above random chance
  - $F = \frac{MS_{model}}{MS_{error}} = \frac{SS_{model}/df_{model}}{SS_{error}/df_{error}}$
  - If  $F > 1$ , the group differences are greater than what would be expected as random variation within groups
- if this test is significant, we then go in to look for pairwise differences between groups



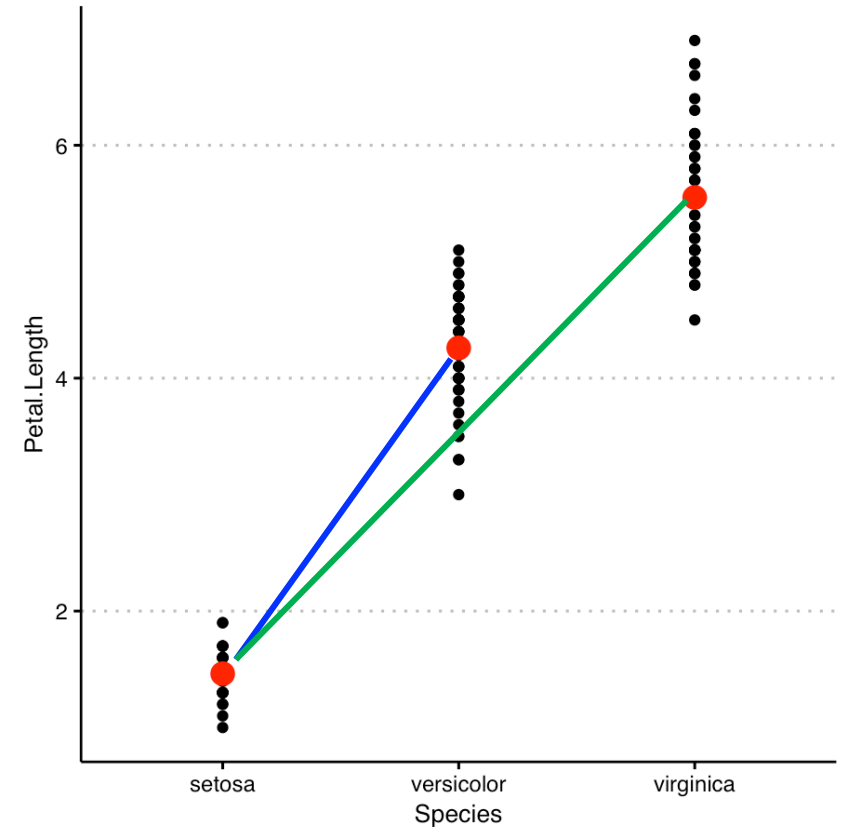
# types of ANOVAs = complex linear models

- $n$  (independent variables)
  - one-way: one independent variable
  - two-way / three-way
  - ( $n > 3$ )-way: crazy land
- within or between subjects
  - between subjects: regular ANOVA
    - independent observations: each raw score comes from *different* individuals!
  - within-subjects: repeated measures ANOVA
    - non-independent observations: multiple raw scores from from the same individuals



# revisiting iris

- when more than two groups are involved, we need to **expand our model** to include multiple groups
- petal length =  $a + b(\text{virginica}) + c(\text{versicolor})$ 
  - $a = M_{\text{setosa}}$
  - $b = M_{\text{virginica}} - M_{\text{setosa}}$
  - $c = M_{\text{versicolor}} - M_{\text{setosa}}$





# NHST for one-way ANOVA

- **step 1: state the hypotheses**

- $H_0$ : no change in mean petal lengths due to species, i.e.,  $\mu_{setosa} = \mu_{virginica} = \mu_{versicolor}$
- $H_1$ : there is **at least one mean difference (no claims about where!)**

- **step 2: set criteria for decision**

$$F(df_1, df_2) = F_{critical}$$

- **step 3: collect data**

- **step 4: make a decision!**

# NHST for one-way ANOVA

step 1:  
state the  
hypotheses

step 2:  
set criteria  
for decision

step 3:  
collect  
data

step 4:  
make a  
decision!

$H_0: \mu_1 = \mu_2 = \dots = \mu_n$   
 $H_1: \text{at least one mean difference}$

$\alpha = .05$   
find  $F_{critical}$  based  
on **right** tailed test  
and degrees of  
freedom  
 $df_1 = k - 1$   
 $df_2 = n - k$

- (1) compute  $SS_{model}$  and  $SS_{error}$
- (2) compute  $F_{observed} = \frac{MS_{model}}{MS_{error}}$
- (3) find p-value for F-score

check whether  $F$  is  
beyond  $F_{critical}$  and  
p-value  $< .05$ . if so, reject  
null hypothesis!

# step 2: set criteria for decision

- we need to find the values of  $df_1$  and  $df_2$  and the corresponding critical F value

$$F(df_1, df_2) = F_{critical}$$

- k denotes the **number of levels of the independent variable** OR estimated parameters

- $df_{model} = k - 1 = 3 - 1 = 2$

- $df_{error} = n - k = 30 - 3 = 27$

- $F_{critical} = F(df_1, df_2)$  at  $\alpha = 0.05$

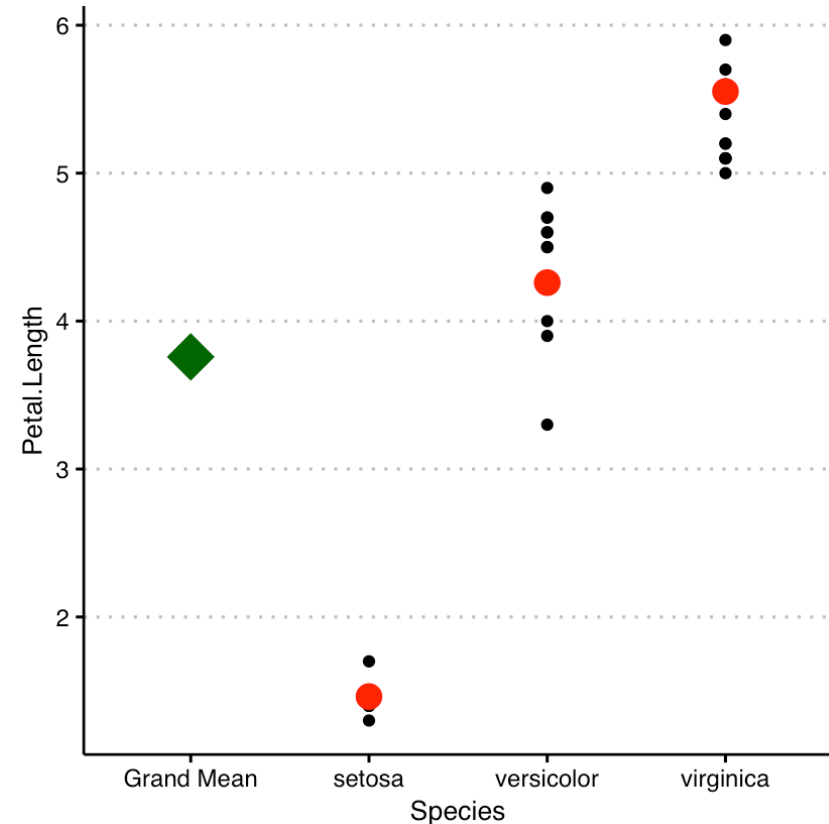
- $F(2,27)$  at  $\alpha = 0.05 = 3.59$

# step 3: obtaining $F_{observed}$

- we need to obtain estimates of

$$F_{observed} = \frac{MS_{model}}{MS_{error}} = \frac{SS_{model}/df_{model}}{SS_{error}/df_{error}}$$

- $SS_{total} = SS_{model} + SS_{error}$ 
  - $SS_{total}$  represents error left over after grand mean
  - $SS_{error}$  represents error left over after species mean model
  - $SS_{model} = SS_{total} - SS_{error}$



# step 3: obtaining $F_{observed}$

- $SS_{total}$  represents deviations from grand mean ( $M_Y$ )

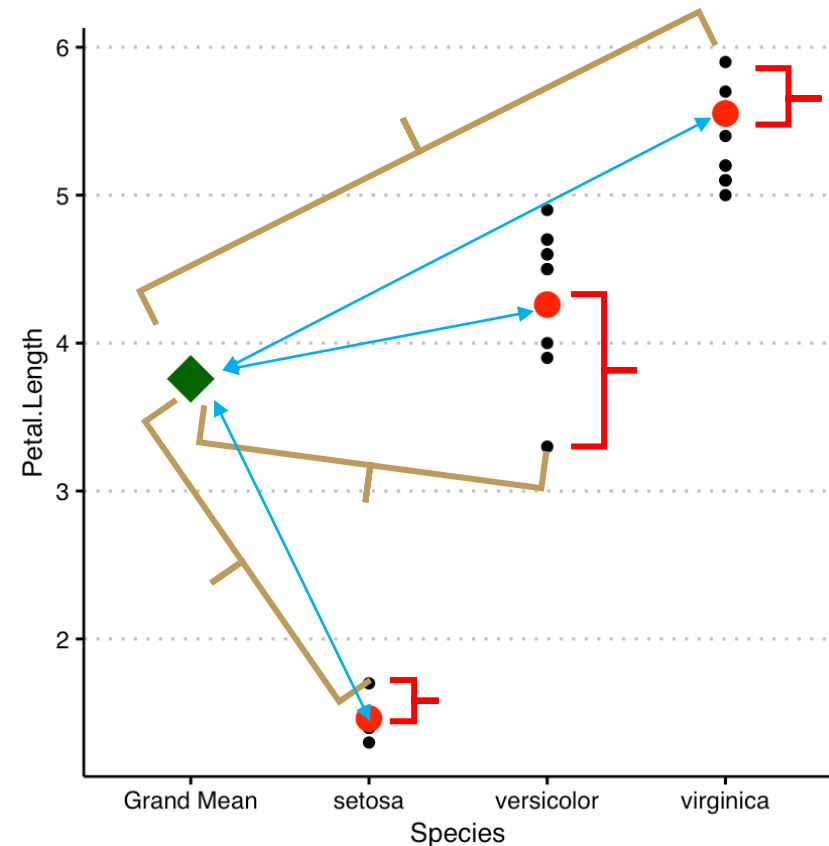
$$SS_{total} = \sum (Y - M_Y)^2$$

- $SS_{error}$  represents the deviations of each score from its group mean

$$SS_{error} = \sum (Y - \hat{Y})^2 = \sum (Y - M_{group})^2$$

- $SS_{model}$  represents the **gains** we get if we substitute each score with the group mean instead of the grand mean

$$SS_{model} = \sum \sum n_i (M_{group} - M_Y)^2 = SS_{total} - SS_{error}$$



# step 3 and 4: obtaining F & decisions

- calculate each of these for iris
  - $SS_{total} = 84.76$
  - $SS_{error} = 3.087$
  - $SS_{model} = 81.67$
- calculate  $F_{observed} = \frac{MS_{model}}{MS_{error}} = \frac{SS_{model}/df_{model}}{SS_{error}/df_{error}} = 357.1778, p < .0001$
- $F_{critical} = 3.59$  at  $\alpha = 0.05$
- therefore, the null hypothesis can be rejected! i.e., species differ in their petal lengths
- $F(2,27) = 357.18, p < .0001$

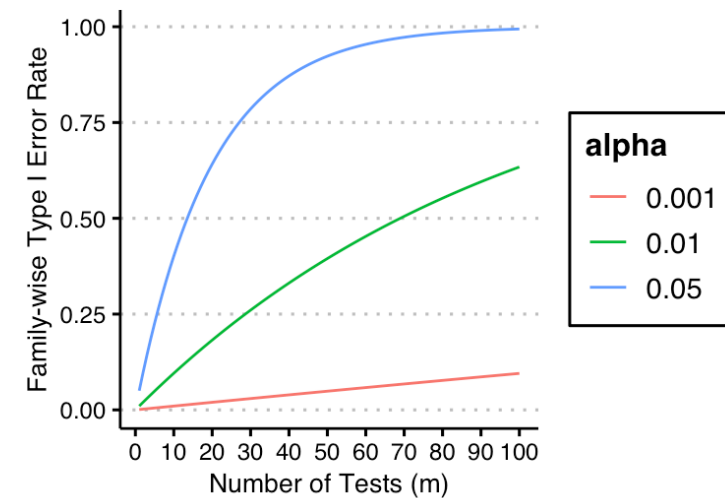
# F-table

		SS	df	MS	F	p-value
$SS_{model}$	species	81.67	2	75.272	357.1778	<.0001
$SS_{error}$	residual	3.087	27	0.0526		

```
Response: Petal.Length
          Sum Sq Df F value    Pr(>F)
Species   81.675  2  357.18 < 2.2e-16 ***
Residuals  3.087 27
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

# what now?

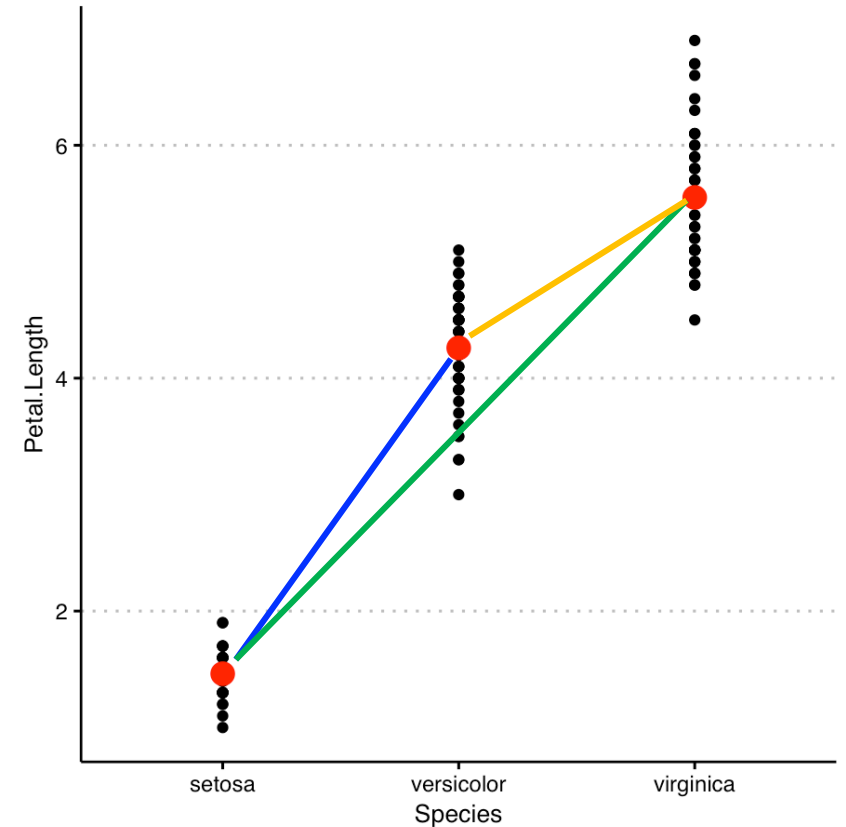
- once we know that the “overall” test has detected something meaningful, we can look for specific differences by conducting **pairwise t-tests**
  - $M_{virginica} - M_{setosa}$
  - $M_{versicolor} - M_{setosa}$
  - $M_{virginica} - M_{versicolor}$
- BUT...what about the **type I error??**
- we correct for multiple comparisons





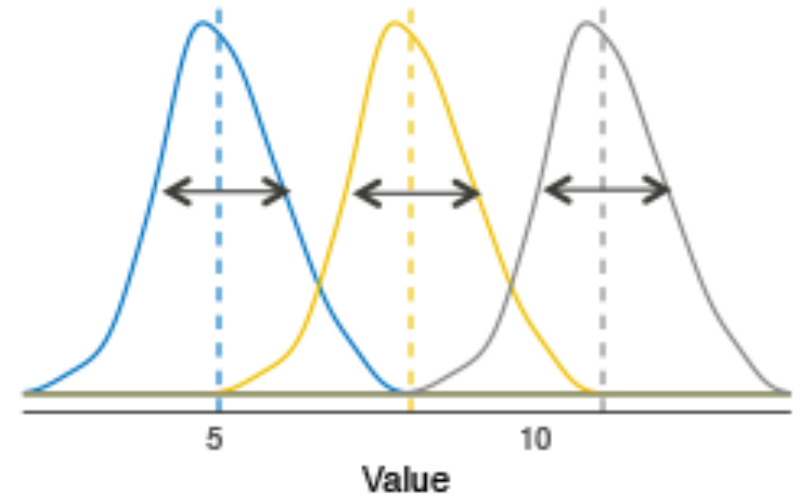
# post-hoc tests

- when pairwise t-tests are conducted after an “overall” / omnibus test (ANOVA), they are called **post-hoc** tests
- several **corrections** exist in the literature
  - Tukey’s Honest Significant Difference Test: moderately conservative
  - Scheffe’s test: very conservative
  - Fisher LSD: very liberal
- most statistical software will allow you to apply a correction, so we will not cover the specifics
- visual inspection is useful in these situations



# one-way ANOVA assumptions

- independent observations within each sample
- normality
- homogeneity of variances



# next time

- **before** class
  - *watch*: [Hypothesis Testing \(two-groups: F-test\)](#) [12 min]
  - *watch*: [Hypothesis Testing \(one-way ANOVA: F-test\)](#) [18 min]
  - *read*: [optional] Read Chapter 14 from the Gravetter & Wallnau (2017) textbook.
  - *start*: Problem Set #6 (Chapter 10 and Chapter 12 problems)
- **during** class
  - two independent variables / two-way ANOVA