

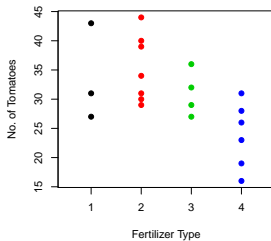
## Lecture 19: ANOVA Part I

### Chapter 5.5

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### Analysis of Variance (ANOVA)

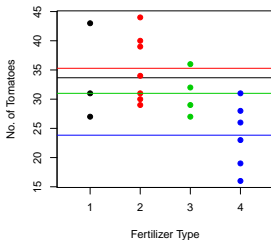
A farmer has the choice of four tomato fertilizers and wants to compare their performance in terms of crop yield.



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## Analysis of Variance (ANOVA)

A farmer has the choice of four tomato fertilizers and wants to compare their performance in terms of crop yield.



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## Analysis of Variance (ANOVA)

We have  $k = 4$  groups AKA **levels of a factor**: the 4 types of fertilizer.

- ▶  $n_i$  plants assigned to each of the  $k = 4$  fertilizers:

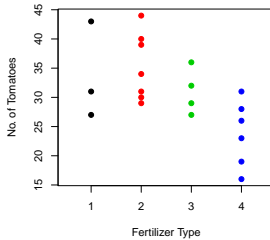
$n_1$	$n_2$	$n_3$	$n_4$	total $n$
3	7	4	6	20

- ▶ Count the number of tomatoes on each plant

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## Tomato Fertilizer

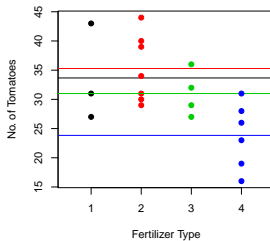
We observe the following, where each point is one tomato plant.



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## Tomato Fertilizer

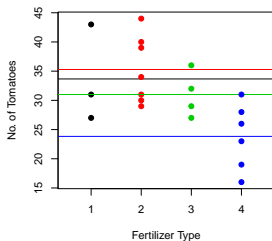
We observe the following, where each point is one tomato plant.  
Plot the sample mean of each level.



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## Tomato Fertilizer

We observe the following, where each point is one tomato plant.  
Plot the sample mean of each level. **Question:** are the mean tomato yields different?



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## Analysis of Variance

Say we have  $k$  groups and want to compare the  $k$  means:

$$\mu_1, \mu_2, \dots, \mu_k$$

We could do  $\binom{k}{2}$  individual two-sample tests.

Ex. for groups 1 & 2:

$$\begin{aligned} H_0 : & \mu_1 = \mu_2 \\ \text{vs. } H_a : & \mu_1 \neq \mu_2 \end{aligned}$$

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## Analysis of Variance

Or we do a single overall test via Analysis of Variance ANOVA:

The hypothesis test is:

$$\begin{array}{ll} H_0 : & \mu_1 = \mu_2 = \dots = \mu_k \\ \text{vs. } H_a : & \text{at least one of the } \mu_i \text{'s are different} \end{array}$$

## Analysis of Variance

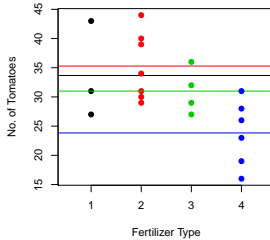
ANOVA asks: where is the overall variability of the observations originate from?

The test statistic used to compute a  $p$ -value is now the **F-statistic**:

$$F = \frac{\text{measure of between-group variability}}{\text{measure of within-group variability}}$$

## Tomato Fertilizer Example

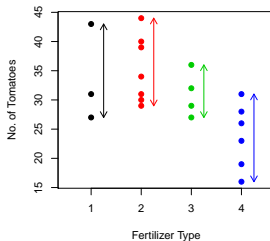
Numerator: the **between-group variation** refers to the variability **between** the levels (the 4 horizontal lines):



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## Tomato Fertilizer Example

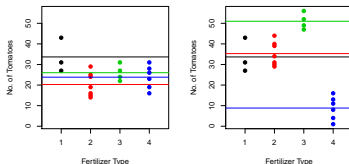
Denominator: the **within-group variation** refers to the variability **within** each level (the 4 vertical arrows):



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## Tomato Fertilizer Example

Now compare the following two plots. Which has “more different” means?



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## Tomato Fertilizer Example

- ▶ They have the **same within-group** variability. Call this value  $W$
- ▶ The right plot has **higher between group** variability b/c the 4 means are more different. Call these values  $B_{left}$  and  $B_{right}$  with  $B_{left} < B_{right}$
- ▶ Recall  $F = \frac{\text{measure of between-group variability}}{\text{measure of within-group variability}}$
- ▶ Since  $\frac{B_{left}}{W} < \frac{B_{right}}{W}$ , thus  $F_{left} < F_{right}$  The right plot as a larger  $F$ -statistic

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## F Distributions

Assuming  $H_0$  is true (that  $\mu_1 = \mu_2 = \dots = \mu_k$ ), the  $F$ -statistic

$$F = \frac{\text{measure of between-group variability}}{\text{measure of within-group variability}}$$

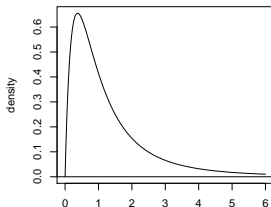
follows the  $F$  distribution with  $df_1 = k - 1$  and  $df_2 = n - k$  degrees of freedom where

- ▶  $n$  = total number of observations
- ▶  $k$  = number of groups

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## F Distributions

For  $df_1 = 4$  and  $df_2 = 6$ , the  $F$  distribution looks like:

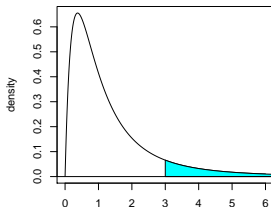


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## F Distributions

$p$ -values are computed where "more extreme" means **larger**. Say the  $F = 3$ , the  $p$ -value is the **area to the right of 3** and is computed in R: `pf(3,df1=4,df2=6,lower.tail=FALSE)`



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## Conducting An F-Test

The results are typically summarized in an **ANOVA table**:

Source of Variation	$df$	$SS$	$MS$	$F$	$p$ -value
Between groups	$k - 1$	$SSTr$	$MSTr = \frac{SSTr}{k-1}$	$\frac{MSTr}{MSE}$	$p$
Within groups	$n - k$	$SSE$	$MSE = \frac{SSE}{n-k}$		
Total	$n - 1$	$SST$			

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## Conditions

1. The observations have to be **independent**. 10% rule.
2. Trade off of  $n$  and **normality** of observations **within each group**.
3. Each of the groups has **constant variance**  $\sigma_1^2 = \dots = \sigma_k^2 = \sigma^2$ .  
Check via:
  - ▶ boxplots
  - ▶ comparing the sample standard deviations  $s_1, \dots, s_k$