Data Science CS301 An Overview of Anova Models

Week 7-8
Fall 2024
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Are you here today?!



https://forms.gle/iaY7zBmxj8KvsDMa8



A New Test ...

- A statistical test to compare the means of multiple groups
- Allows us to determine the existence of significant differences between them.
- Commonly used in research studies to analyze the effects of different variables on an outcome (i.e., useful in scientific research fields.



Analysis of Variance (Anova)

In its simplest form, ANOVA provides a statistical test of whether two (or more) population means are equal, and therefore generalizes the t-test beyond two means.

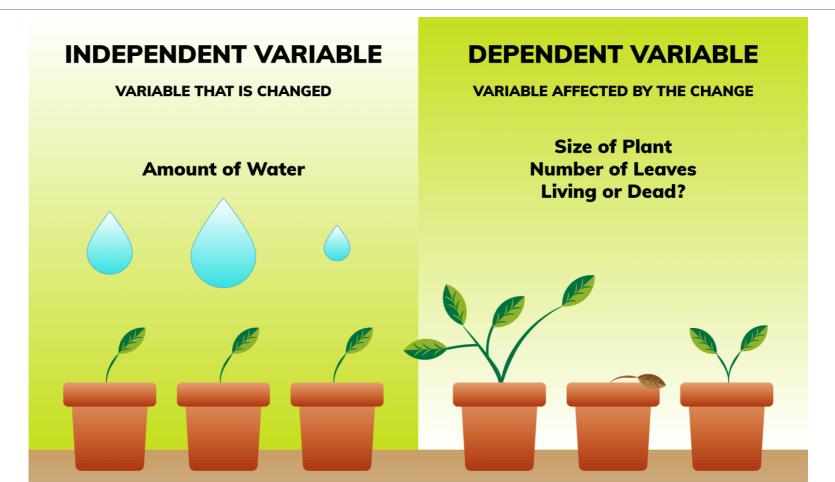
- Quick example:
 - Independent variable is social media use, and you assign groups to low, medium, and high levels of social media use.
 - Find out if there is a difference in hours of sleep per night
 - Find out more! CLICK HERE

URL



Independent, Dependent Variables?

- Independent is seemingly random or unpredictable
- Dependent is not random; behavior depends on independent variable.

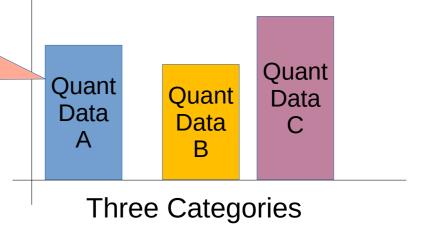




When to Use ANOVA?

- The independent variable should have at least three levels (i.e. at least three different groups or categories)
- ANOVA tells you if the *dependent* variable changes according to the level of the *independent* variable

Quantitative data is data that can be counted or measured in *numerical* values. The two main types of quantitative data are *discrete* data and *continuous* data.





ANOVA: Different from t-Test

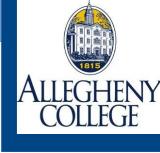
- The Student's t-test is used to compare the means between two groups (levels)
- ANOVA is used to compare the means among three or more groups (levels)
- One way ANOVA: is a hypothesis test in which only one categorical variable or single factor is considered
- Makes comparisons between of means of three or more samples
- Each test studies differences in means and the spread of distributions (i.e., variance) across groups
- The statistical mechanisms of each test calculate statistical significance differently



Avoiding Type-1 Errors

- When there are more than two means, then we could use the t-test to compare the means
- But! When conducting multiple t-tests, we run into more type-1 errors from the test
- Using ANOVA reduces type I error rates
- A Type I error is to reject the null hypothesis when we should have accepted it
- Erroneous conclusions that results are statistically significant when, actually, they are not significant





Did You Say, Errors?

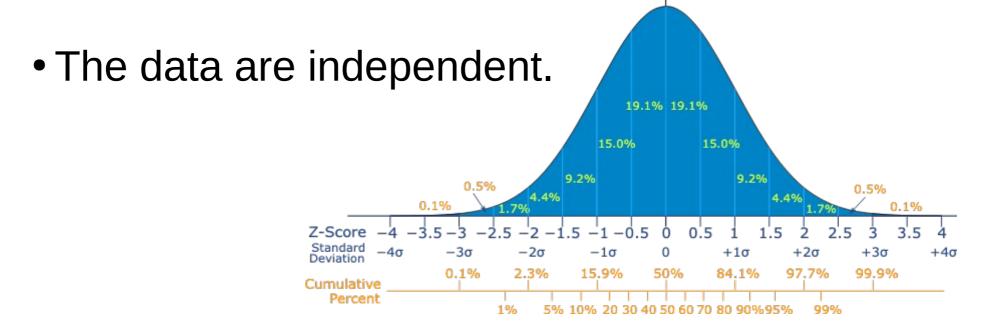
Type I and Type II Error

Null hypothesis is	True	False
Rejected	Type I error False positive Probability = α	Correct decision True positive Probability = 1 - β
Not rejected	Correct decision True negative Probability = 1 - α	Type II error False negative Probability = β



Assumptions of ANOVA

- The responses for each factor level have a normal population distribution
- These distributions have the same variance



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Testing Hypotheses in Anova

- Null hypothesis (H0) of the ANOVA: there is no difference between means
 - Written as, H0: $\mu 1 == \mu 2 == \mu 3$
- Alternative hypothesis (Ha): the mean of at least one group is different from the others
 - Written as, Ha: μ1 != μ2 != μ3
- For,
 - HO = the null hypothesis,
 - Ha = the alternative hypothesis,
 - 11 = the mean of population 1
 - \square = the mean of population 2
 - \square = the mean of population 3



Groups of Dogs

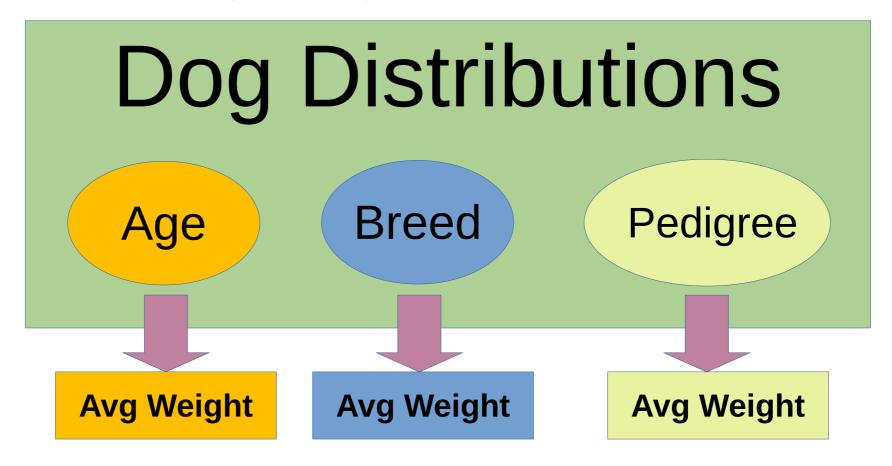


The Dog Show



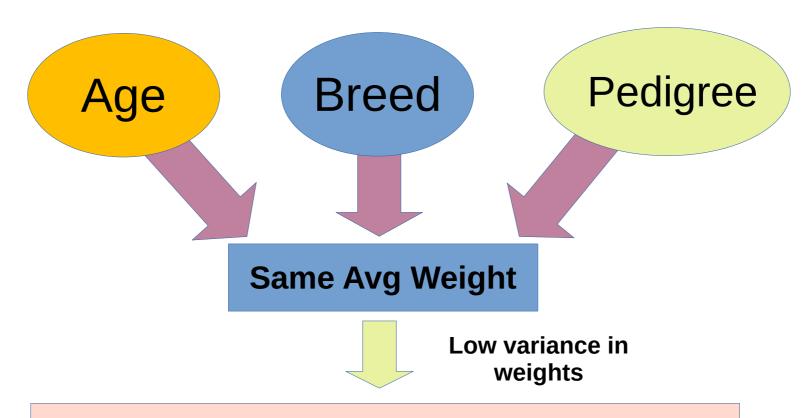
Prediction by Group Type

 We want to predict the weight of a dog based on a certain set of characteristics for each dog (age, breed, and pedigree)









If groups have the same mean, then it isn't reasonable to conclude that the groups are, in fact, separate in any meaningful way

Similar Means From Low Variances



- Separate dogs into distributions having low variance of dog weights (a *heterogeneous* group) ...
- The weight-means between groups should be distinct
- If weight-means were similar between groups, then the groups would be similar.

We create two groups: are-cute and have floppy ears.

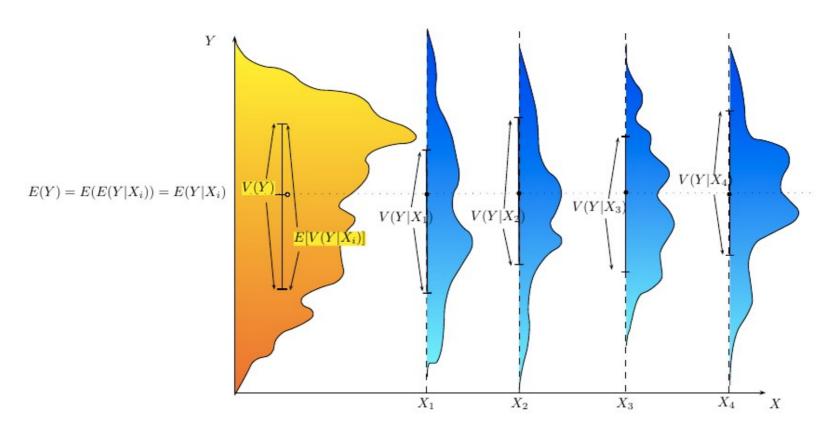
The mean-weights would be similar from each group.

All these Beagles are (both) **cute** and have **floppy ears**!





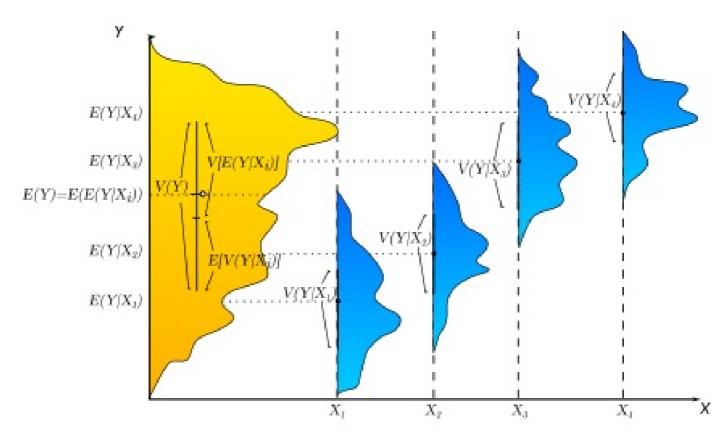




- Orange: mean-weights of all dogs
- Blues: All mean-weights by types of groups
- Groups do not explain variation in distributions



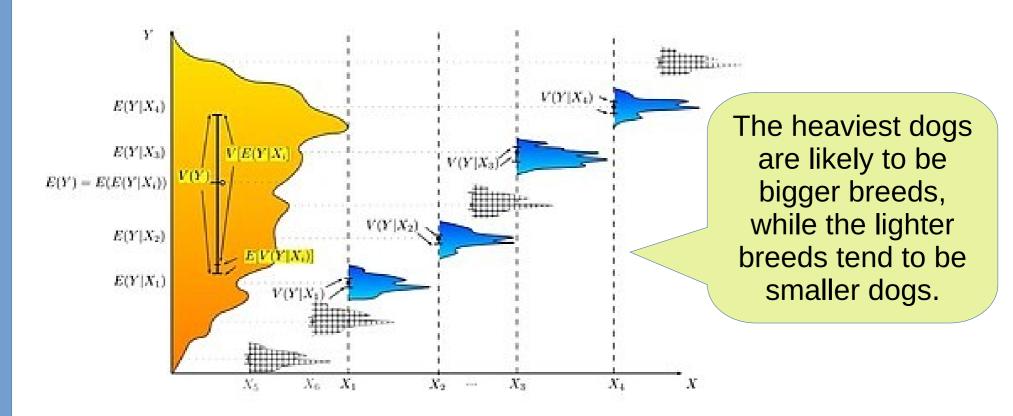




- Orange: mean-weights of all dogs
- Blues: All mean-weights by types of groups
- Groups are becoming distinguishable by meanweights
 https://en.wikipedia.org/wiki/Analysis_of variance







- Orange: mean-weights of all dogs
- Blues: All mean-weights by types of groups
- Groups are distinguishable by mean-weights







Dog groups to similar types of weight







Dog groups to similar types of weight



So What Then?

ANOVA provides a statistical test of equality of weight-means across the groups of dogs.

An ANOVA generalizes the t-test beyond two means.









Follow
the
code
from the
Tutorial
and
address
some
questions.
URL on
next slide.



Let's Look at Some Code!

```
# clear out the variables
 rm(list = ls())
# clear out all plots from previous
work.
 graphics.off()
# clear the console
 cat("\014")
# install libraries
if(!require('tidyverse')) {
 install.packages('tidyverse')
 library('tidyverse')
```

Clean that work space!





Create Random Data to Test

```
# Set seed for reproducibility
set.seed(123)

# Generate synthetic data
group1 <- rnorm(30, mean = 5, sd = 1)
group2 <- rnorm(30, mean = 7, sd = 1)
group3 <- rnorm(30, mean = 6, sd = 1)
```

Randomly generated and normal distribution

```
> group1
[1] 4.439524 4.769823 6.558708 5.070508 5.129288 6.715065 5.460916 3.734939 4.313147
[10] 4.554338 6.224082 5.359814 5.400771 5.110683 4.444159 6.786913 5.497850 3.033383
[19] 5.701356 4.527209 3.932176 4.782025 3.973996 4.271109 4.374961 3.313307 5.837787
[28] 5.153373 3.861863 6.253815
```



Groups Into Data Frame

```
# Set seed for reproducibility
set.seed(123)
# Generate synthetic data
group 1 < -rnorm(30, mean = 5, sd = 1)
group2 <- rnorm(30, mean = 7, sd = 1)
group3 < - \text{rnorm}(30, \text{mean} = 6, \text{sd} = 1)
# Place into a data frame
data <- data.frame(
 value = c(group1, group2, group3),
 group = factor(rep(c("Group 1", "Group 2", "Group 3"),
each = 30)
```



Run the Test, Visualize to Understand Results

```
# Perform ANOVA
anova_result <- aov(value ~ group, data = data)
# Display the summary of the ANOVA
summary(anova result)
# Visualize means by boxplot
ggplot(data, aes(x = group, y = value)) +
 geom boxplot(fill = c("lightblue", "lightgreen",
"lightcoral")) +
 theme minimal() +
 labs(title = "Boxplot of Values by Group",
    x = "Group",
    y = "Value")
```

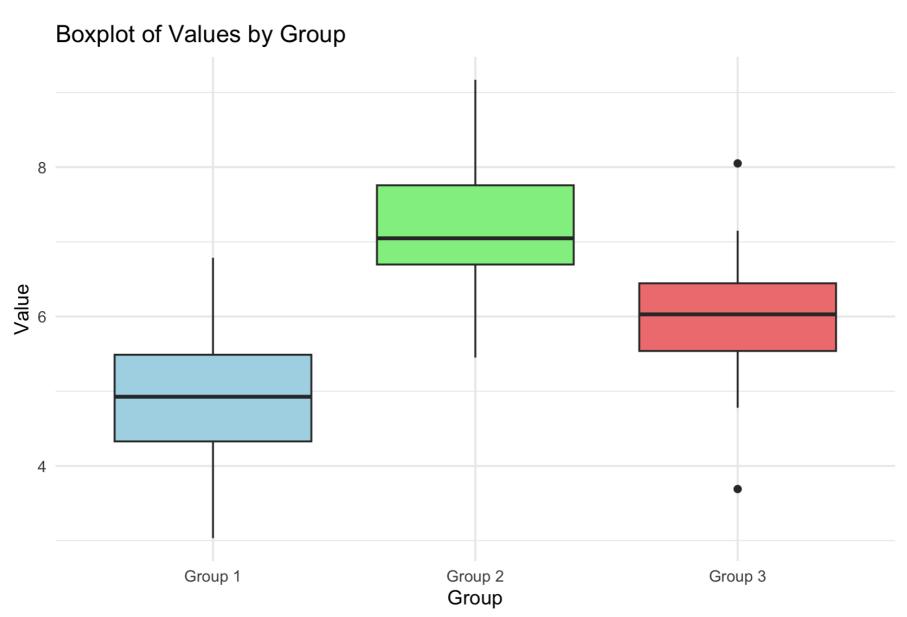


Check the Significance

The *p*-value is significant



Visualize Difference Between Group Means





Anyone For Completing a Tutorial?

ANOVA in R | A Complete Step-by-Step Guide with Examples

Published on March 6, 2020 by Rebecca Bevans. Revised on November 17, 2022.

ANOVA is a statistical test for estimating how a quantitative dependent variable changes according to the levels of one or more categorical independent variables. ANOVA tests whether there is a difference in means of the groups at each level of the independent variable.

https://www.scribbr.com/statistics/anova-in-r/