

# One-way ANOVA Homework

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```
# remove # from line below if you need to install UNKstats package
# remotes::install_github("CalebRother/UNKstats")

library(tidyverse)

## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr     1.1.4     v readr     2.1.5
## v forcats   1.0.1     v stringr   1.6.0
## v ggplot2   4.0.0     v tibble    3.3.0
## v lubridate 1.9.4     v tidyverse 1.3.1
## v purrr    1.2.0

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()   masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors

library(UNKstats)
```

## Question 1:

Researchers want to test a new drug when given at different dosages. They are also including a placebo. They are measuring the score on a psychological examination, and are hoping to find the most effective dose to reduce score. Run the proper tests to determine if each group is unique and what effects they have.

```
drug <- tibble(
  Dose = rep(c("Placebo", "Low", "Med", "High"), each = 12),
  Score = c(
    # Placebo
    0.90, 0.82, 1.00, 1.12, 0.74, 0.93, 1.02, 1.21, 0.84, 0.91, 1.08, 0.86,
    # Low
    0.10, 0.00, -0.10, 0.20, 0.05, -0.05, 0.15, 0.00, -0.10, 0.12, 0.06, 0.01,
    # Med
    -0.30, -0.42, -0.22, -0.50, -0.35, -0.27, -0.31, -0.46, -0.38, -0.20, -0.36, -0.29,
    # High
    0.40, 0.32, 0.50, 0.36, 0.25, 0.46, 0.41, 0.52, 0.35, 0.30, 0.44, 0.39
  )
)
```

1. State the null and alternative hypotheses.
2. Are these data normal? How can you tell?
3. Which dosage/dosages are most effective at reducing the score?

```
## # A tibble: 2 x 6
##   term      df    sumsq   meansq statistic   p.value
```

```

##   <chr>     <int>  <dbl>  <dbl>      <dbl>      <dbl>
## 1 Dose        3 10.9   3.62      339.  2.10e-30
## 2 Residuals   44  0.470  0.0107      NA    NA

```

- What might the biological implications of this test be?

### Question 2:

Professors at UNK are debating if labs should be kept in Biology 105. They want to see if exam scores are higher in classes taught with only lecture, only lab, or both. Run the proper tests to determine if each group is unique and what effects they have.

```

teach <- tibble(
  Method = rep(c("Lab", "Both", "Lecture"), each = 15),
  Exam = c(
    # Lab
    76, 79, 82, 77, 75, 80, 81, 78, 76, 79, 77, 83, 74, 80, 78,
    # Both
    84, 85, 82, 88, 83, 86, 84, 87, 81, 86, 85, 83, 84, 89, 82,
    # Lecture
    80, 82, 79, 83, 81, 80, 82, 81, 84, 78, 83, 80, 81, 79, 82
  )
)

```

- State the null and alternative hypotheses.
- Are these data normal?
- Which teaching method is most effective? How do you know this?

```

## # A tibble: 2 x 6
##   term      df sumsq meansq statistic p.value
##   <chr>     <int> <dbl>  <dbl>      <dbl>      <dbl>
## 1 Method      2 297. 148.       30.1  7.72e-9
## 2 Residuals   42 207.   4.93      NA    NA

```

- What might the teaching implications of this test be?

### Question 3:

Researchers are testing weight gain drugs in comparison to each other. They are also comparing the effects of these drugs to a placebo. Run the proper tests to determine if each group is unique and what effects they have.

```

diet <- tibble(
  Diet = rep(c("A", "B", "C", "Placebo"), each = 14),
  WeightGain = c(
    # A
    1.1, 0.9, 1.5, 1.2, 0.8, 1.0, 1.4, 0.9, 1.0, 1.2, 1.5, 0.9, 1.3, 1.4,
    # B
    1.3, 1.5, 1.2, 0.8, 1.6, 1.5, 1.0, 2.0, 1.5, 1.4, 1.6, 1.2, 1.5, 1.4,
    # C
    2.1, 2.0, 1.6, 2.3, 2.1, 2.0, 1.6, 2.1, 2.5, 1.7, 2.2, 2.1, 2.3, 2.7,
    # Placebo
    0.2, 0.2, -0.2, 0.0, -0.8, -0.4, 0.4, 0.2, 1.0, 0.0, -0.1, -0.2, -0.7, 0.6
  )
)

```

- State the null and alternative hypotheses.

- Are these data normal?

```
diet |>
  group_by(Diet) |>
  summarise(p_value = shapiro.test(WeightGain)$p.value)

## # A tibble: 4 x 2
##   Diet      p_value
##   <chr>     <dbl>
## 1 A         0.228
## 2 B         0.527
## 3 C         0.457
## 4 Placebo   0.952

## # A tibble: 2 x 6
##   term        df    sumsq   meansq statistic   p.value
##   <chr>     <int> <dbl>    <dbl>     <dbl>      <dbl>
## 1 Diet         3  31.3   10.4      88.3  2.11e-20
## 2 Residuals   52  6.15   0.118     NA       NA
```

- Interpret the meaning of the assigned letters. How were they found, and what do they mean?

- Which diet plan was most effective for gaining weight?

#### Question 4:

A field study is measuring time-to-first-foraging (minutes) for four bird colonies. Run the proper tests to determine if each group is unique.

```
set.seed(4517)

n_per <- 18
Colonies_data <- data.frame(
  Colony = rep(c("North", "East", "South", "West"), each = n_per),
  Minutes = c(
    round(rexp(n_per, rate = 1.0), 1),
    round(rexp(n_per, rate = 1.3), 1),
    round(rexp(n_per, rate = 0.7), 1),
    round(rexp(n_per, rate = 1.0), 1)
  )
)
```

- State the null and alternative hypotheses.

- Are these data normal?

```
## # A tibble: 2 x 6
##   term        df    sumsq   meansq statistic   p.value
##   <chr>     <int> <dbl>    <dbl>     <dbl>      <dbl>
## 1 Colony      3   4.35    1.45      5.33  0.00234
## 2 Residuals   68  18.5    0.272     NA       NA
```

- Interpret the meaning of the assigned letters. How were they found, and what do they mean?

- Which colony/colonies were the fastest foragers?

#### Question 5:

A team of behavioral ecologists recorded the number of aggressive interactions among four populations of prairie voles raised in different social environments. They want to know if social upbringing influences aggression levels in male voles.

```

set.seed(9913)

n_per <- 16
vole <- tibble(
  Environment = rep(c("Isolated", "Paired", "Group", "Crowded"), each = n_per),
  Aggression = c(
    round(rnorm(n_per, mean = 8, sd = 3)),
    round(rnorm(n_per, mean = 6, sd = 1.5)),
    round(rnorm(n_per, mean = 4, sd = 2)),
    round(rnorm(n_per, mean = 7, sd = 4))
  )
)

```

1. State the null and alternative hypotheses.

2. Are these data normal?

```

## # A tibble: 2 x 6
##   term        df  sumsq meansq statistic p.value
##   <chr>     <int> <dbl>  <dbl>    <dbl>
## 1 Environment     3  24.4   8.12    0.776   0.512
## 2 Residuals      60 629.   10.5     NA       NA

```

3. Interpret the meaning of the assigned letters. How were they found, and what do they mean?

4. What else is unique about this data and might explain the results? *Hint:* Look at the final plot