

Computational Thinking 2 - Conditionals

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```
library(tidyverse)
```

```
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr      1.1.4      v readr      2.1.5
v forcats    1.0.0      v stringr    1.5.1
v ggplot2    3.5.2      v tibble     3.2.1
v lubridate  1.9.4      v tidyr      1.3.1
v purrr      1.0.4
-- 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
```

```
library(here)
```

here() starts at C:/Users/conno/OneDrive/Documents/UCSC Undergrad/DataScience4EEB/Computational

1

1.1

```
x <- 5

# Check if the value of x is greater than 10
if(x > 10)
```

```

{
  # Paste takes the value stored in x and combines that with a character string
  print(paste(x, "is greater than 10"))
} else
{
  print(paste(x, "is less than or equal to 10"))
}

```

```
[1] "5 is less than or equal to 10"
```

Q1.1

```

x <- 11

# Check if the value of x is greater than 10
if(x > 10)
{
  # Paste takes the value stored in x and combines that with a character string
  print(paste(x, "is greater than 10"))
} else
{
  print(paste(x, "is less than or equal to 10"))
}

```

```
[1] "11 is greater than 10"
```

output: "11 is greater than 10"

```

x <- "five"

# Check if the value of x is greater than 10
if(x > 10)
{
  # If x is > 10, multiple x by 2

```

```
    print(x*2)
  } else
  {
    # If x is not > 10, divide x by 2
    print(x/2)
  }
  traceback()
```

```
# define a variable
x <- 11

# check the value of x using nested if-else statements
if (x < 10) {
  # if x is less than 10
  print("x is less than 10")
} else {
  # if x is exactly equal to 10
  if (x == 10) {
    print("x is 10!!!")
  } else {
    # if x is greater than 10
    print("x is greater than 10")
  }
}
```

```
[1] "x is greater than 10"
```

```
vec <- c(9, 10, 11, 12)
```

```
# For 1 through the length of the vector "vec"
for (i in 1:length(vec)) {

  # check the value of using nested if-else statements
  if (vec[i] < 10) {
    # if the element is less than 10
    print("value is less than 10")
  }
}
```

```

} else {
  # if the element is exactly equal to 10
  if (vec[i] == 10) {
    # if the element equals 10
    print("value is 10!!!")
  } else {
    # if the element is greater than 10
    print("value is greater than 10")
  }
}
}
}

```

```

[1] "value is less than 10"
[1] "value is 10!!!"
[1] "value is greater than 10"
[1] "value is greater than 10"

```

Q1.2

```

y <- c(-2, 42, 0, 10)

# For 1 through the length of the vector "vec"
for (i in 1:length(y)) {

  # check the value of using nested if-else statements
  if (y[i] < 0) {
    # if the element is negative
    print("value is negative")
  } else {
    # if the element is exactly equal to 0
    if (y[i] == 0) {
      # if the element equals 0
      print("value is 0!!!")
    } else {
      # if the element is positive
      print("value is positive")
    }
  }
}

```

```
}  
}  
}
```

```
[1] "value is negative"  
[1] "value is positive"  
[1] "value is 0!!!"  
[1] "value is positive"
```

1.2

```
library(lterdatasampler)
```

Q1.3

```
?nwt_pikas
```

```
starting httpd help server ... done
```

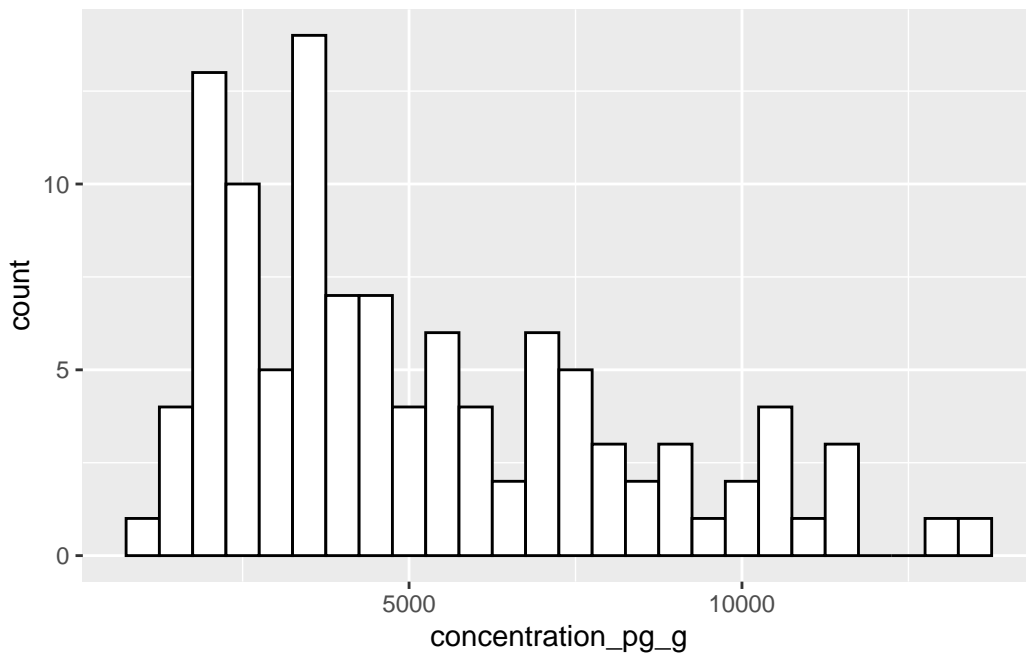
Stress was measured by observing the amount of glucocorticoid metabolite present in pika feces. The variable is called `concentration_pg_m` and the units are in picogram GCM/gram dry pika feces

Q1.4

```
nwt_pikas <- nwt_pikas
```

each row represents an individual pika poop

```
# Make a histogram
nwt_pikas %>%
  ggplot(aes(x = concentration_pg_g)) +
  # Add the histogram geom, which only needs an x-axis
  # Choose a binwidth of 500 picogram GCM/gram
  geom_histogram(binwidth = 500,
                 fill = "white",
                 color = "black")
```

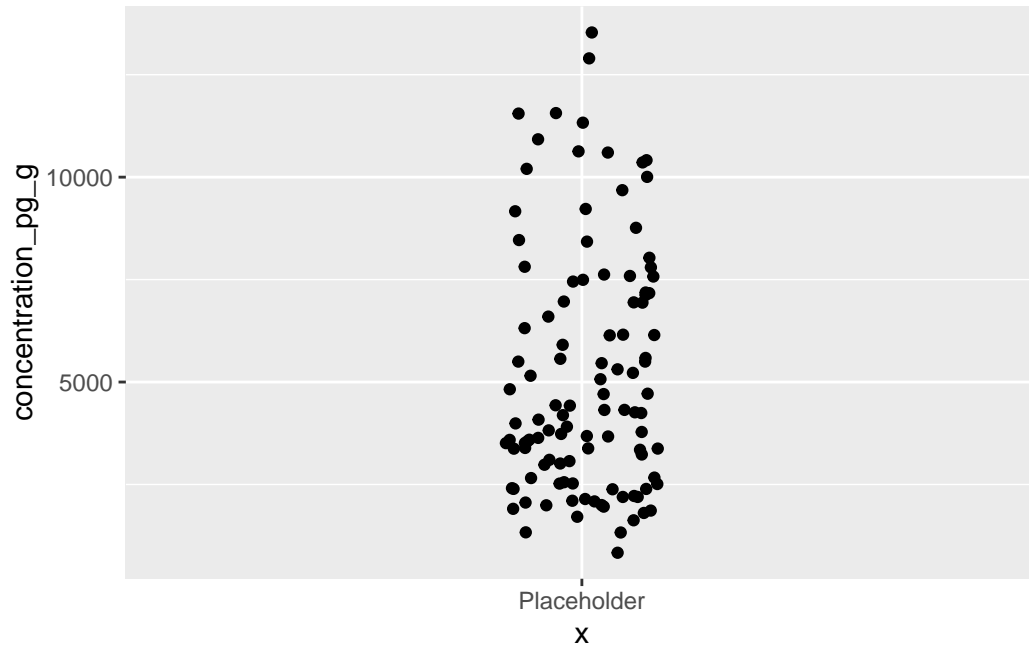


```
# Make a scatterplot with jittered points
nwt_pikas %>%
  # We're adding a little placeholder axis just so we can see the point distribution
  ggplot(aes(x = "Placeholder",
```

```

      y = concentration_pg_g)) +
# Add the geom_jitter geom
geom_jitter(width = 0.1)

```



```

nwt_pikas_categ <- nwt_pikas %>%
# Call the new column stress_category
mutate(stress_category = case_when(
# When the value is > 5000, make the new column's value "Stressed!!!!"
concentration_pg_g > 5000 ~ "Stressed!!!!!!",
# Otherwise, make the new column's value "Chill"
.default = "Chill"
))

# Check out the first 6 rows, but remove the utm columns just for visibility
head(nwt_pikas_categ %>% select(-c(utm_easting, utm_northing)))

```

```

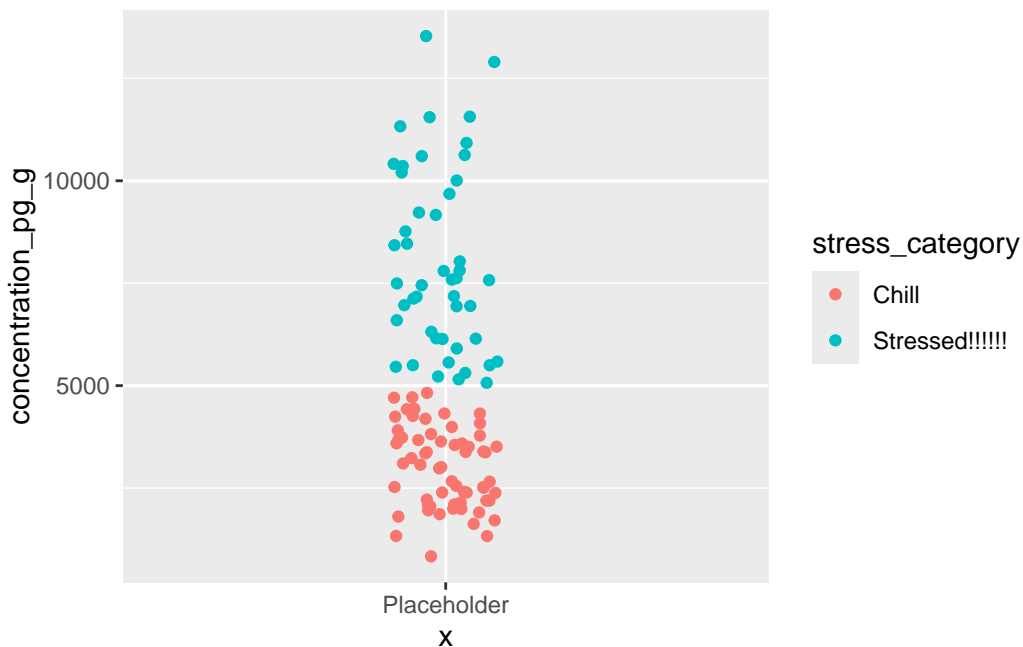
# A tibble: 6 x 7
  date       site      station sex  concentration_pg_g elev_m stress_category
<date>    <fct>    <fct> <fct>      <dbl>    <dbl> <chr>
1 2018-06-08 Cable Gate Cable G~ male      11563.   3343. Stressed!!!!!!
2 2018-06-08 Cable Gate Cable G~ male      10629.   3353. Stressed!!!!!!
3 2018-06-08 Cable Gate Cable G~ male      10924.   3358. Stressed!!!!!!

```

4	2018-06-13	West Knoll	West Kn~	male	10414.	3578.	Stressed!!!!!!
5	2018-06-13	West Knoll	West Kn~	male	13531.	3584.	Stressed!!!!!!
6	2018-06-13	West Knoll	West Kn~	<NA>	7799.	3595.	Stressed!!!!!!

Q1.5

```
# Make a scatterplot with jittered points
nwt_pikas_categ %>%
  # We're adding a little placeholder axis just so we can see the point distribution
  ggplot(aes(x = "Placeholder",
             y = concentration_pg_g,
             color = stress_category)) +
  # Add the geom_jitter geom
  geom_jitter(width = 0.1)
```



```
nwt_pikas_categ2 <- nwt_pikas_categ %>%
  # Create a new column called month
  # then, extract the month from the date using the month() function
```



```
mutate(month = month(date)) %>%
# Lastly, relocate the month column after the date column so it's more easily visible to u
relocate(month, .after = date)

head(nwt_pikas_categ2)
```

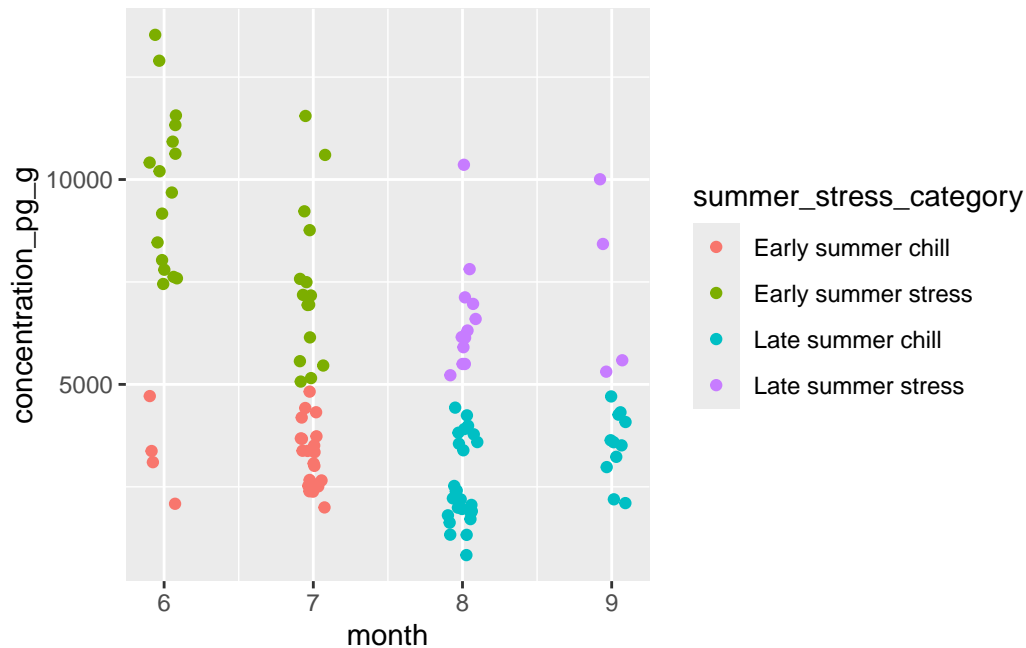
```
# A tibble: 6 x 10
  date      month site      station      utm_easting utm_northing sex
<date>    <dbl> <fct>      <fct>      <dbl>      <dbl> <fct>
1 2018-06-08      6 Cable Gate Cable Gate 1      451373      4432963 male
2 2018-06-08      6 Cable Gate Cable Gate 2      451411      4432985 male
3 2018-06-08      6 Cable Gate Cable Gate 3      451462      4432991 male
4 2018-06-13      6 West Knoll West Knoll 3      449317      4434093 male
5 2018-06-13      6 West Knoll West Knoll 4      449342      4434141 male
6 2018-06-13      6 West Knoll West Knoll 5      449323      4434273 <NA>
# i 3 more variables: concentration_pg_g <dbl>, elev_m <dbl>,
# stress_category <chr>
```

```
nwt_pikas_summerstress <- nwt_pikas_categ2 %>%
  mutate(summer_stress_category = case_when(
    (month == 6 | month == 7) & concentration_pg_g > 5000 ~ "Early summer stress",
    (month == 6 | month == 7) & concentration_pg_g <= 5000 ~ "Early summer chill",
    (month == 8 | month == 9) & concentration_pg_g > 5000 ~ "Late summer stress",
    (month == 8 | month == 9) & concentration_pg_g <= 5000 ~ "Late summer chill",
    .default = "NA"
  ))

head(nwt_pikas_summerstress)
```

```
# A tibble: 6 x 11
  date      month site      station      utm_easting utm_northing sex
<date>    <dbl> <fct>      <fct>      <dbl>      <dbl> <fct>
1 2018-06-08      6 Cable Gate Cable Gate 1      451373      4432963 male
2 2018-06-08      6 Cable Gate Cable Gate 2      451411      4432985 male
3 2018-06-08      6 Cable Gate Cable Gate 3      451462      4432991 male
4 2018-06-13      6 West Knoll West Knoll 3      449317      4434093 male
5 2018-06-13      6 West Knoll West Knoll 4      449342      4434141 male
6 2018-06-13      6 West Knoll West Knoll 5      449323      4434273 <NA>
# i 4 more variables: concentration_pg_g <dbl>, elev_m <dbl>,
# stress_category <chr>, summer_stress_category <chr>
```

```
# Make a scatterplot with jittered points
nwt_pikas_summerstress %>%
  # We're adding a little placeholder axis just so we can see the point distribution
  ggplot(aes(x = month,
             y = concentration_pg_g,
             color = summer_stress_category)) +
  # Add the geom_jitter geom
  geom_jitter(width = 0.1)
```



2

Q2.1

We are using the iris dataset.

```
iris <- iris
```

Q2.2

We are going to try to arbitrarily classify whether a flower is “big” or small” based on its sepal length. We are using if/else statements to iterate through the column.

Q2.3

```
#assign this column to a vector
sepal_length <- iris$Sepal.Length

for (i in 1:20) {
  #determine if flower is big
  if (sepal_length[i] >= 5.5) {
    print("flower is big")
    #determine is flower is small
  } else {
    print("flower is small")
  }
}
```

```
[1] "flower is small"
[1] "flower is small"
[1] "flower is small"
[1] "flower is small"
[1] "flower is small"
[1] "flower is small"
[1] "flower is small"
[1] "flower is small"
[1] "flower is small"
[1] "flower is small"
[1] "flower is small"
[1] "flower is small"
[1] "flower is small"
```

```
[1] "flower is small"
[1] "flower is big"
[1] "flower is big"
[1] "flower is small"
[1] "flower is small"
[1] "flower is big"
[1] "flower is small"
```

```
#Output is REALLY long, so we will not iterate through the whole vector. We will only do the
```

Q2.4

We are going to try to arbitrarily classify whether a flower is “big” or small” based on what species it is using sepal length. Use `case_when()` to iterate through the dataset based on species type and sepal length, taking both into consideration. This will make our previous for loop better by making the output into a column in the dataset.

Q2.5

```
#make new dataset using iris
iris_size <- iris %>%
  #use mutate to make new column for flower size
  #use case_when() to set instructions for new column
  mutate(flower_size = case_when(
    #separate based on species and sepal length
    Species == "setosa" & Sepal.Length >= 5.0 ~ "Big",
    Species == "setosa" & Sepal.Length < 5.0 ~ "Small",
    Species == "versicolor" & Sepal.Length >= 6.5 ~ "Big",
    Species == "versicolor" & Sepal.Length < 6.5 ~ "Small",
    Species == "virginica" & Sepal.Length >= 6.5 ~ "Big",
    Species == "virginica" & Sepal.Length < 6.5 ~ "Small"
  ))

head(iris_size)
```

	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species	flower_size
1	5.1	3.5	1.4	0.2	setosa	Big
2	4.9	3.0	1.4	0.2	setosa	Small
3	4.7	3.2	1.3	0.2	setosa	Small
4	4.6	3.1	1.5	0.2	setosa	Small
5	5.0	3.6	1.4	0.2	setosa	Big
6	5.4	3.9	1.7	0.4	setosa	Big