431 Part A: Extra Example

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Sleep and Mammals

This example reviews some of the key work we did in Part A of the course. It uses the msleep data set, which is part of the ggplot2 package. Use the help file for that data set or visit http://ggplot2.tidyverse.org/reference/msleep.html to assist in understanding the data.

1. Identify the number of rows and number of variables in the msleep data set.

```
dim(msleep)
```

[1] 83 11

There are 83 rows and 11 variables.

2. Specify the variable with the largest number of missing values in msleep. How many values are missing?

summary(msleep)

| name | genus | vore | |
|------------------|------------------|------------------|--|
| Length:83 | Length:83 | Length:83 | |
| Class :character | Class :character | Class :character | |
| Mode :character | Mode :character | Mode :character | |

| order | conservation | sleep_total | sleep_rem |
|------------------|---------------|-------------------|-----------------|
| Length:83 | Length:83 | Min. : 1.90 |) Min. :0.100 |
| Class :character | Class :charac | ter 1st Qu.: 7.89 | 1st Qu.:0.900 |
| Mode :character | Mode :charac | ter Median :10.10 | Median :1.500 |
| | | Mean :10.43 | Mean :1.875 |
| | | 3rd Qu.:13.79 | 3rd Qu.:2.400 |
| | | Max. :19.90 | • |
| | | | NA's :22 |
| sleep_cycle | awake | brainwt | bodywt |
| Min. :0.1167 | | | Min. : 0.005 |
| 1st Qu.:0.1833 | 1st Qu.:10.25 | 1st Qu.:0.00290 | 1st Qu.: 0.174 |
| Median :0.3333 | Median :13.90 | Median :0.01240 | Median : 1.670 |
| Mean :0.4396 | Mean :13.57 | Mean :0.28158 | Mean : 166.136 |
| 3rd Qu.:0.5792 | 3rd Qu.:16.15 | 3rd Qu.:0.12550 | 3rd Qu.: 41.750 |
| Max. :1.5000 | Max. :22.10 | Max. :5.71200 | Max. :6654.000 |
| NA's :51 | | NA's :27 | |

sleep_cycle is missing for 51 of the 83 mammals.

3. Identify the mammal who remains awake the longest, per day. What is the Z score for this mammal?

Let's calculate the z scores, and then sort by the awake value.

1.47

1.44

```
msleep %>% select(name, awake) %>%
 mutate(z_awake = round((awake - mean(msleep$awake))/sd(msleep$awake),2)) %>%
 arrange(desc(awake))
# A tibble: 83 x 3
              name awake z_awake
             <chr> <dbl> <dbl>
1
           Giraffe 22.10
                            1.92
2
       Pilot whale 21.35
                            1.75
3
                          1.69
             Horse 21.10
 4
          Roe deer 21.00
                            1.67
5
            Donkey 20.90
                            1.65
6 African elephant 20.70
                          1.60
7
      Caspian seal 20.50
                          1.56
8
             Sheep 20.20
                            1.49
```

It's the giraffe, whose Z score is 1.92

... with 73 more rows

Asian elephant 20.10

Cow 20.00

9

10

4. Display your R code to create a data set (called sleep2) from msleep which contains the variables name, order, vore, sleep_cycle and sleep_rem for those animals who have no missing values in any of those four variables, then convert the vore information into a factor.

```
sleep2 <- msleep %>%
  select(name, order, vore, sleep_cycle, sleep_rem) %>%
  filter(complete.cases(name, order, vore, sleep_cycle, sleep_rem)) %>%
  mutate(vore = factor(vore))
sleep2
```

```
# A tibble: 31 x 5
```

| | name | order | vore | sleep_cycle | sleep rem | | |
|---------------------|----------------------------|-------------------------|---------------|-------------|-------------|--|--|
| | <chr></chr> | <chr></chr> | <fctr></fctr> | <dbl></dbl> | <dbl></dbl> | | |
| 1 | Greater short-tailed shrew | Soricomorpha | omni | 0.1333333 | 2.3 | | |
| 2 | Cow | Artiodactyla | herbi | 0.6666667 | 0.7 | | |
| 3 | Three-toed sloth | Pilosa | herbi | 0.7666667 | 2.2 | | |
| 4 | Northern fur seal | Carnivora | carni | 0.3833333 | 1.4 | | |
| 5 | Dog | Carnivora | carni | 0.3333333 | 2.9 | | |
| 6 | Guinea pig | Rodentia | herbi | 0.2166667 | 0.8 | | |
| 7 | Chinchilla | Rodentia | herbi | 0.1166667 | 1.5 | | |
| 8 | Lesser short-tailed shrew | Soricomorpha | omni | 0.1500000 | 1.4 | | |
| 9 | Long-nosed armadillo | Cingulata | carni | 0.3833333 | 3.1 | | |
| 10 | North American Opossum | ${\tt Didelphimorphia}$ | omni | 0.3333333 | 4.9 | | |
| # with 21 more rows | | | | | | | |

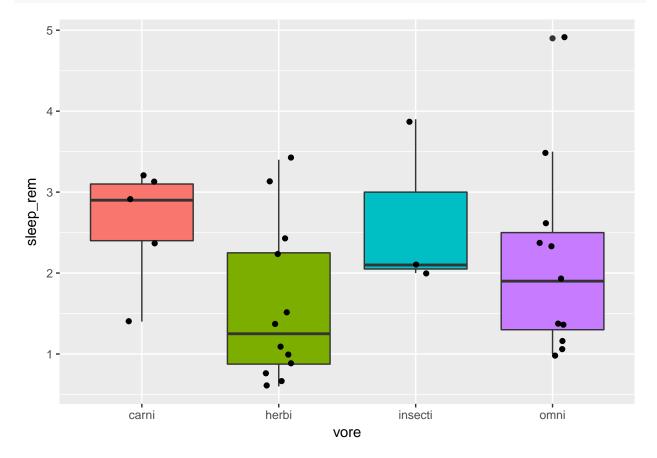
5. According to the order variable, how many Primates (see the order variable) exist in your sleep2 data set?

```
sleep2 %>% filter(order == "Primates")
# A tibble: 5 x 5
                         vore sleep_cycle sleep_rem
        name
                order
                <chr> <fctr>
                                    <dbl>
                                               <dbl>
       <chr>
      Galago Primates
                                0.5500000
                                                 1.1
1
                         omni
2
       Human Primates
                                1.5000000
                                                 1.9
                         omni
3
     Macaque Primates
                         omni
                                0.7500000
                                                1.2
4 Chimpanzee Primates
                         omni
                                1.4166667
                                                 1.4
      Baboon Primates
                         omni
                                0.6666667
                                                1.0
```

There should be five.

6. Draw a plot to compare the sleep_rem levels by vore group using your sleep2 data. What do you conclude?

```
ggplot(sleep2, aes(x = vore, y = sleep_rem, fill = vore)) +
geom_boxplot() +
geom_jitter(width = 0.1) +
guides(fill = FALSE)
```



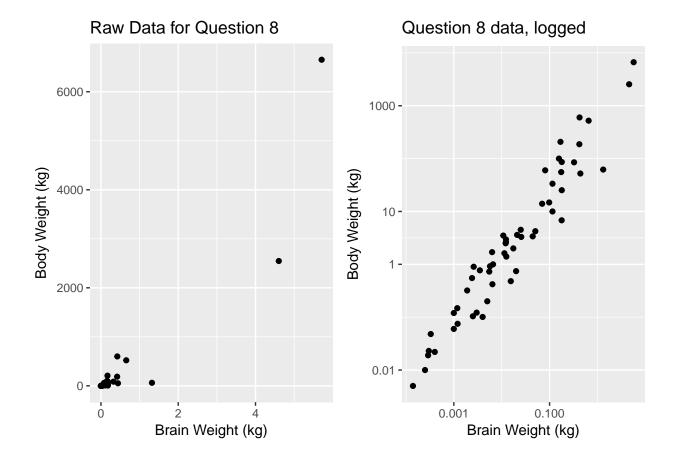
In general, carnivores have the longest sleep_rem while herbivores have the shortest, at least in terms of

medians. There are very few values (looks like just three) in the insectivores, and it looks like just five in the carnivores.

7. Produce R code using the %>% pipe to produce a table which answers these two questions for the sleep2 data: [a] Which vore group has the largest mean sleep_rem level? [b] Which vore group has the largest mean sleep_cycle?

```
sleep2 %>% group_by(vore) %>% summarize(n(), mean(sleep_rem), mean(sleep_cycle))
# A tibble: 4 x 4
    vore `n()` `mean(sleep_rem)` `mean(sleep_cycle)`
   <fctr> <int>
                           <dbl>
                        2.600000
                                          0.3733333
   carni 5
            12
   herbi
                        1.591667
                                          0.4180556
3 insecti
            3
                        2.666667
                                          0.1611111
    omni
            11
                        2.154545
                                          0.5924242
```

8. Now, return to the original msleep data for questions 8-10. Build a scatterplot of brainwt and bodywt, first using the raw data and then using a logarithmic scale for each variable.



9. Fit a linear model to the scatterplot you drew in part 8 for which a linear model seems more appropriate, and specify and describe (in a sentence of two) both the fitted least squares equation *and* the Pearson correlation coefficient.

Clearly the log-log model is the better choice. I'll revert to a natural logarithm for the plot below, and the model.

```
model9 <- lm(log(bodywt) ~ log(brainwt), data = sleep1)
summary(model9)</pre>
```

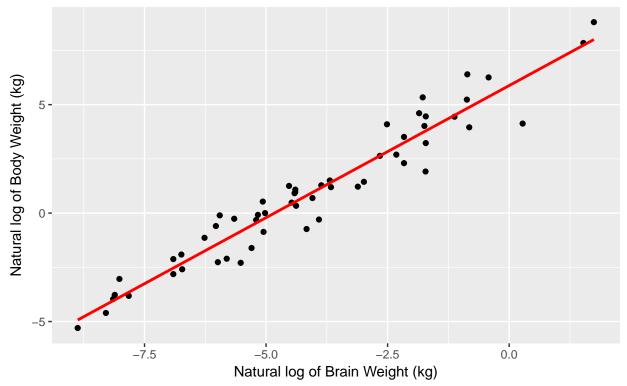
```
lm(formula = log(bodywt) ~ log(brainwt), data = sleep1)
Residuals:
   Min
             1Q Median
                             ЗQ
                                    Max
-2.0902 -0.4282 0.1054 0.5723 1.6211
Coefficients:
             Estimate Std. Error t value Pr(>|t|)
(Intercept)
              5.87919
                         0.21393
                                   27.48
                                           <2e-16 ***
log(brainwt)
              1.21785
                         0.04482
                                   27.17
                                           <2e-16 ***
               0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Signif. codes:
```

```
Residual standard error: 0.8403 on 54 degrees of freedom
Multiple R-squared: 0.9319, Adjusted R-squared: 0.9306
F-statistic: 738.4 on 1 and 54 DF, p-value: < 2.2e-16

ggplot(sleep1, aes(x = log(brainwt), y = log(bodywt))) +
    geom_point() +
    geom_smooth(method = "lm", se = FALSE, col = "red") +
    labs(title = "Question 8 data: Natural Logs, with Linear Model",
        subtitle = "log(Body Weight) = 5.88 + 1.22 log(Brain Weight)",
        x = "Natural log of Brain Weight (kg)",
        y = "Natural log of Body Weight (kg)")
```

Question 8 data: Natural Logs, with Linear Model

log(Body Weight) = 5.88 + 1.22 log(Brain Weight)



10. Identify the mammal in your model (in part 9) with the largest (in absolute value) regression residual.

```
3 Brazilian tapir 1.621108
4 Cow 1.565557
5 Owl monkey 1.538485
6 Thirteen-lined ground squirrel 1.447518
7 Squirrel monkey 1.411995
8 Giant armadillo 1.275983
9 Tenrec 1.264385
10 Galago 1.036076
```

... with 46 more rows

The answer is Human.