Lab2_2

2024-04-12

```
# 5: import code
all.data <- read.csv("cell_infectivity_wide.csv")</pre>
# 6: view the data, also done in console
View(all.data)
# 8,9
f.WT <- all.data$Wild.type.Finch</pre>
f.hLDLR <- all.data[, "hLDLR.Finch"]</pre>
c.WT <- all.data[["Wild.type.Chicken"]]</pre>
c.hLDLR <- all.data[[4]]</pre>
# 10
c.WT[5]
## [1] 1.343797
c.WT
## [1] 4.076517 2.046030 3.952518 3.067056 1.343797 4.398596 4.416107 3.243844
## [9] 5.772721
# second value of f.WT
f.WT[2]
## [1] 0.197452
# Section 3, number 11
mean(f.WT)
## [1] 0.2536965
# mean of f.hLDLR
mean(f.hLDLR)
## [1] 2.143759
```

data.means # print the vector

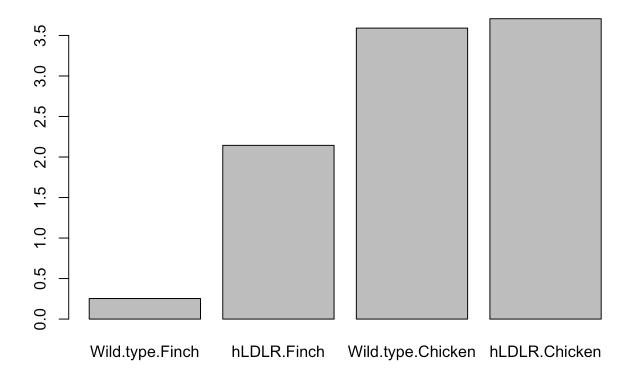
data.means <- colMeans(all.data) # create a vector of group means.

12

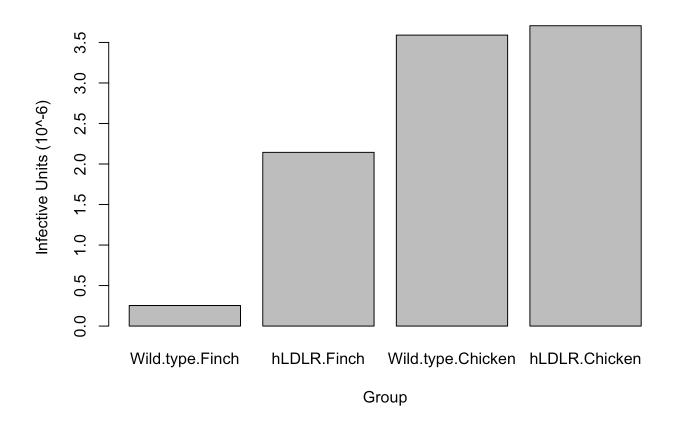
hLDLR.Finch Wild.type.Chicken ## Wild.type.Finch hLDLR.Chicken 0.2536965 ## 2.1437594 3.5907984 3.7064355 # 13, 14 apply(all.data, 2, mean) ## hLDLR.Finch Wild.type.Chicken hLDLR.Chicken Wild.type.Finch ## 0.2536965 2.1437594 3.5907984 3.7064355 # 15 # sd function apply(all.data,2, sd) hLDLR.Chicken ## Wild.type.Finch hLDLR.Finch Wild.type.Chicken ## 0.07997842 1.23324550 1.33656421 0.88193124 # var function apply(all.data,2, var) Wild.type.Finch hLDLR.Finch Wild.type.Chicken hLDLR.Chicken ## ## 0.006396548 1.520894451 1.786403894 0.777802718 # min function apply(all.data,2, min) ## Wild.type.Finch hLDLR.Finch Wild.type.Chicken hLDLR.Chicken 0.1458407 ## 0.4487605 1.3437965 2.4873307 # max function apply(all.data,2, max) ## Wild.type.Finch hLDLR.Finch Wild.type.Chicken hLDLR.Chicken 0.3737137 4.1256479 5.7727210 4.8716014 ## # median function apply(all.data, 2, median) ## Wild.type.Finch hLDLR.Finch Wild.type.Chicken hLDLR.Chicken ## 0.239245 2.067363 3.952518 3.622098 # summary function apply(all.data,2, summary)

```
##
           Wild.type.Finch hLDLR.Finch Wild.type.Chicken hLDLR.Chicken
                 0.1458407
                                                  1.343797
## Min.
                              0.4487605
                                                                2.487331
## 1st Qu.
                 0.1974520
                              1.4996970
                                                  3.067056
                                                                3.136610
## Median
                 0.2392450
                              2.0673629
                                                  3.952518
                                                                3.622098
## Mean
                 0.2536965
                              2.1437594
                                                  3.590798
                                                                3.706436
## 3rd Qu.
                 0.3229060
                              3.2264398
                                                  4.398596
                                                                4.596712
## Max.
                 0.3737137
                              4.1256479
                                                  5.772721
                                                                4.871601
```

```
# 16
barplot(data.means)
```



```
# 17
barplot(data.means, ylab = "Infective Units (10^-6)", xlab = "Group")
```



```
# 18
finch <- data.means[1:2]</pre>
chicken <- data.means[3:4]</pre>
print(finch)
## Wild.type.Finch
                         hLDLR.Finch
          0.2536965
##
                            2.1437594
print(chicken)
## Wild.type.Chicken
                            hLDLR.Chicken
##
             3.590798
                                 3.706436
bar.data <- data.frame(finch, chicken)</pre>
bar.data
```

Wild.type.Finch 0.2536965 3.590798

hLDLR.Finch

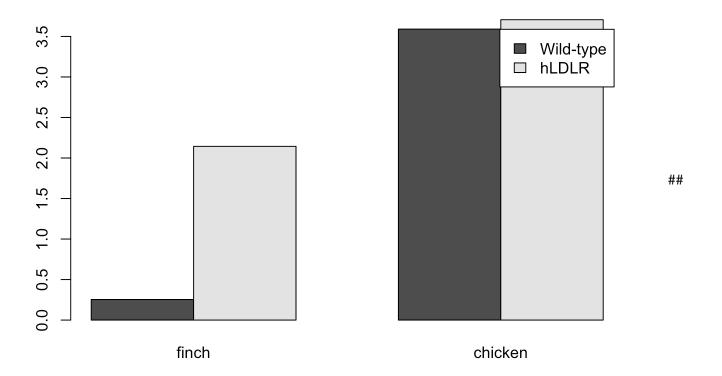
finch chicken

2.1437594 3.706436

```
# 20
rownames(bar.data) <- c("Wild-type", "hLDLR")
bar.data</pre>
```

```
## finch chicken
## Wild-type 0.2536965 3.590798
## hLDLR 2.1437594 3.706436
```

```
# 21
barplot( as.matrix(bar.data), legend= rownames(bar.data), beside = TRUE)
```



Section 5: Assessment

Q1:

- 1. all.data\$Wild.type.Finch
- 2. all.data[, "Wild.type.Finch"] or all.data[, 1]
- 3. all.data[["Wild.type.Finch"]] or all.data[[1]]
- 4. all.data["Wild.type.Finch"] or all.data[1]

Try out by code:

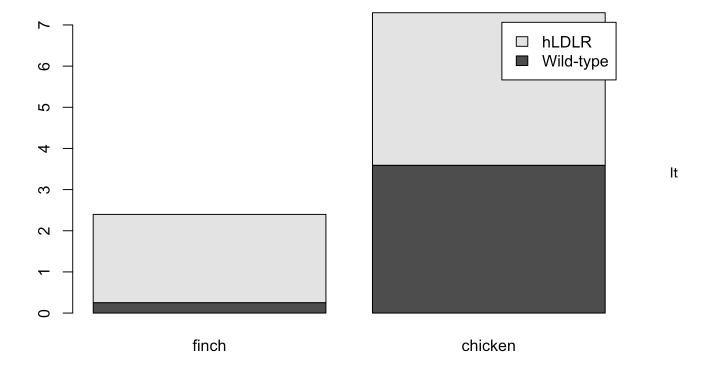
```
a <- all.data$Wild.type.Finch</pre>
b1 <- all.data[, "Wild.type.Finch"]</pre>
b2 <- all.data[, 1]
c1 <- all.data[["Wild.type.Finch"]]</pre>
c2 <- all.data[[1]]</pre>
d1 <- all.data["Wild.type.Finch"]</pre>
d2 <- all.data[1]</pre>
a == b1
a == b2
a == c1
a == c2
a == d1
##
      Wild.type.Finch
##
  [1,]
              TRUE
  [2,]
##
              TRUE
##
  [3,]
              TRUE
  [4,]
##
              TRUE
  [5,]
##
              TRUE
  [6,]
##
              TRUE
  [7,]
##
              TRUE
  [8,]
##
              TRUE
  [9,]
##
              TRUE
a == d2
```

```
Wild.type.Finch
##
##
    [1,]
                      TRUE
##
    [2,]
                      TRUE
    [3,]
                      TRUE
##
##
    [4,]
                      TRUE
    [5,]
##
                      TRUE
##
    [6,]
                      TRUE
##
    [7,]
                      TRUE
##
    [8,]
                      TRUE
    [9,]
##
                      TRUE
```

note only d1 and d2 are in the dataframe format

Q2:

```
barplot( as.matrix(bar.data), legend= rownames(bar.data))
```



seems like instead of having 4 column, we now get 2 columns in total but each column contain 2 different color represent the wild type or hLDLR. Without this code, there seems to have a sub-bars. Thus, the code beside = TRUE will put the value in each column juxtapose instead of stack together.

Q3:

Null hypothesis[H0]: There is no difference in infectivity (number of infective units) between wildtype and hLDLR zebra finch cells.

Alternative hypothesis[H1]: There is a difference in infectivity (number of infective units) between wildtype and hLDLR zebra finch cells.

Q4:

1. mean for both finch group

finch

```
## Wild.type.Finch hLDLR.Finch
## 0.2536965 2.1437594
```

It is clear that wild type finch has a much smaller mean value compare to the hLDLR finch. Thus, this support that idea that cell type has an effect on infective in zebra finch cells.

2. standard deviation for both finch group

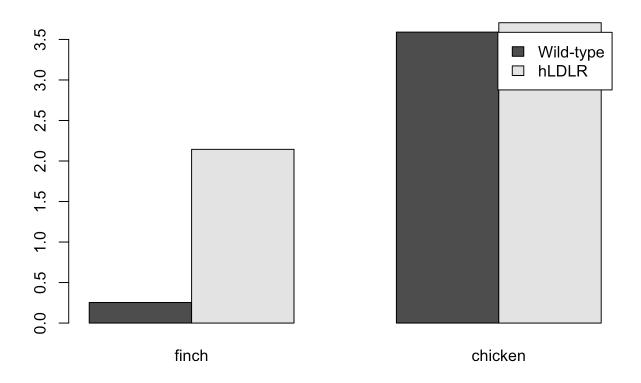
```
apply(all.data,2, sd)
```

##	Wild.type.Finch	hLDLR.Finch Wil	d.type.Chicken	hLDLR.Chicken
##	0.07997842	1.23324550	1.33656421	0.88193124

From here, we get to see that even though the standard deviation for hLDLR group is a little bit higher than the wild type finch, the standard deviation is still very small. Hence, it is a support evident that to reject the null hypothesis.

3. By looking at this barplot:

```
barplot( as.matrix(bar.data), legend= rownames(bar.data), beside = TRUE)
```



We can see that the infective units for hLDLR finch is much higher than the wild type finch.

Therefore, the null hypothesis from Q3 will be rejected since there is a clear connection between cell type and the infective unit in zebra finch.

Q5:

Null hypothesis[H0]: There is no difference in infectivity (number of infective units) between wildtype and hLDLR chicken cell .

Alternative hypothesis[H1]: There is a difference in infectivity (number of infective units) between wildtype and hLDLR chicken cell.

Q6:

The null hypothesis from Q5 will not be rejected. 1. mean for both chicken group:

```
chicken

## Wild.type.Chicken
## 3.590798 3.706436
```

We can see that both group has a very similar mean value. It is a good support that there is no connection between cell type and the infectivity in the chicken cell.

By using the same code from Q4, for the standard derivation, as both group has a similar number of standard derivation and the value is relative small. We will say the data is highly conserves and convince.

From the barplot we used for Q4, since the level of the infecivity unit for both wildtype and hLDLR chicken cell are very close and there is no big difference, we can conclude that there is no clear connection between the cell type and infecitviy in chicken cell. In other word, wide-type or hLDLR cell type does not affected the infectivity units in chicken cells.

Therefore, we will not rejected the null hypothesis from Q5.