

lab9.R

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```
birds =
read.csv("/Users/stonehuang/Documents/environmental_data/data/bird.sta.csv")
hab =
read.csv("/Users/stonehuang/Documents/environmental_data/data/hab.sta.csv")

birdhab = merge(
  birds,
  hab, by=c("basin", "sub", "sta"))

# Create a contingency table for edge/interior and brown creeper
presence/absence
table(
  birdhab$s.edge,
  birdhab$BRCR > 0)

##
##      FALSE TRUE
##  E    144   29
##  I    559  314

br_creeper_table = table(
  birdhab$s.edge,
  birdhab$BRCR > 0)[, 2:1]

br_creeper_table

##
##      TRUE FALSE
##  E     29   144
##  I    314   559

#Q1. The presence/absence of Brown Creepers does not vary between the
interior and edge of forest stands.
#Q2. Brown Creepers show a significant habitat preference because the p-value
is lower than 0.05.
chisq_BRCR = chisq.test(br_creeper_table)
chisq_BRCR

##
##  Pearson's Chi-squared test with Yates' continuity correction
##
```

```
## data: br_creeper_table
## X-squared = 23.3, df = 1, p-value = 1.386e-06

require(palmerpenguins)

## Loading required package: palmerpenguins

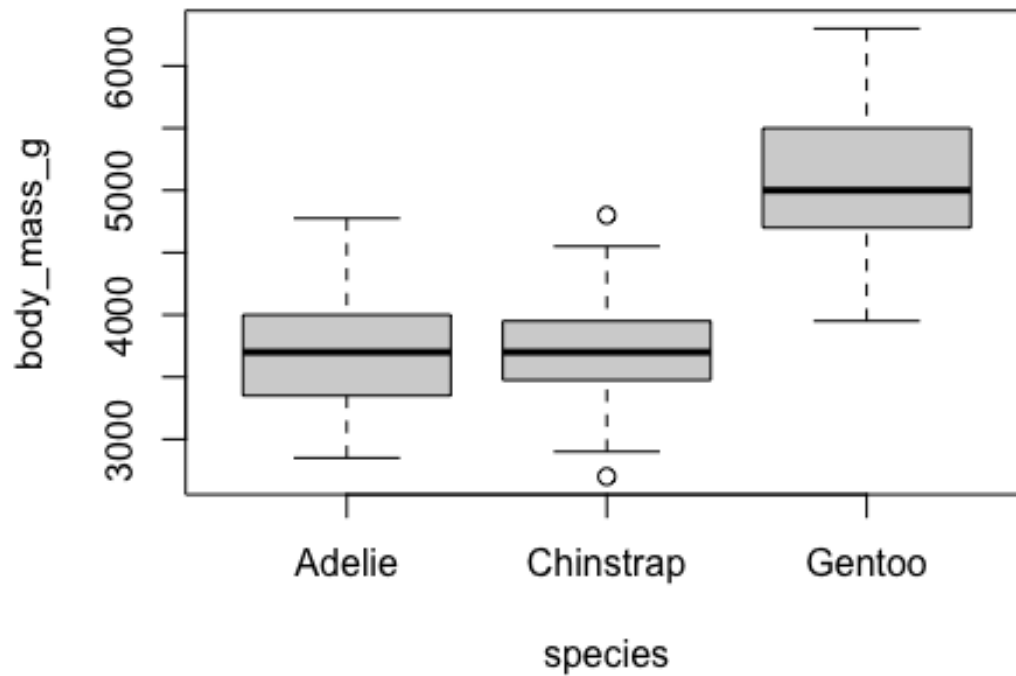
head(penguins)

## # A tibble: 6 × 8
##   species island bill_length_mm bill_depth_mm flipper_length... body_mass_g
##   <fct>   <fct>         <dbl>         <dbl>         <int>         <int>
## 1 Adelie  Torge...         39.1          18.7          181          3750
## 2 Adelie  Torge...         39.5          17.4          186          3800
## 3 Adelie  Torge...         40.3          18           195          3250
## 4 Adelie  Torge...         NA            NA            NA            NA
## 5 Adelie  Torge...         36.7          19.3          193          3450
## 6 Adelie  Torge...         39.3          20.6          190          3650
## # ... with 1 more variable: year <int>

#Q3
fit_species = lm(formula = body_mass_g ~ species, data = penguins)
#Q4
fit_sex = lm(formula = body_mass_g ~ sex, data = penguins)
#Q5
fit_both = lm(formula = body_mass_g ~ sex*species, data = penguins)

#Q6
boxplot(body_mass_g ~ species, data = penguins, main = "Boxplot for the
fit_species model")
```

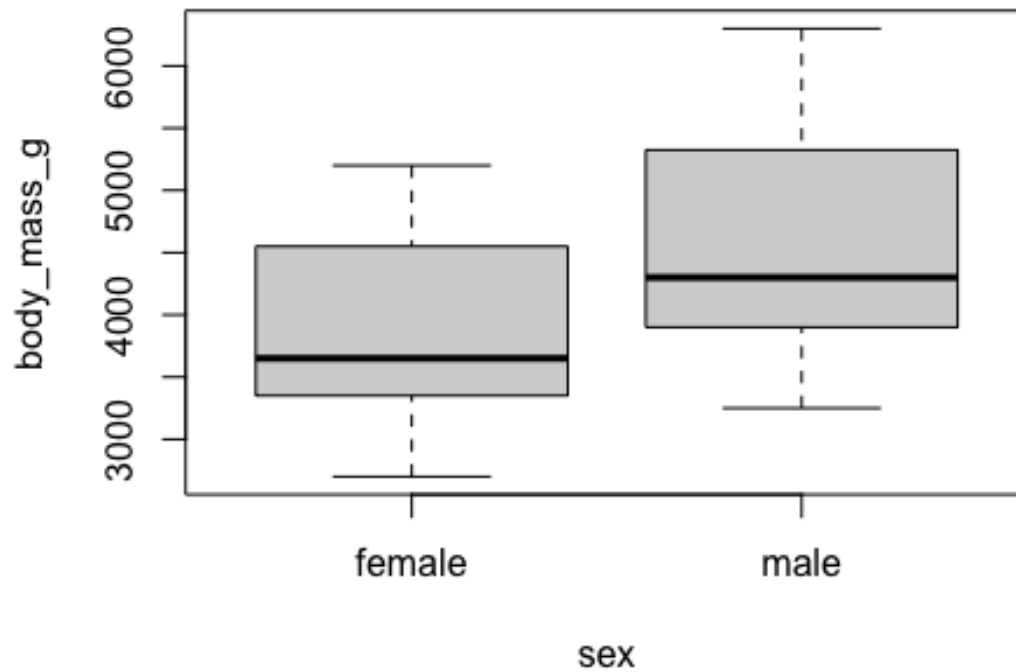
Boxplot for the fit_species model



#Q7

```
boxplot(body_mass_g ~ sex, data = penguins, main = "Boxplot for the fit_sex  
model")
```

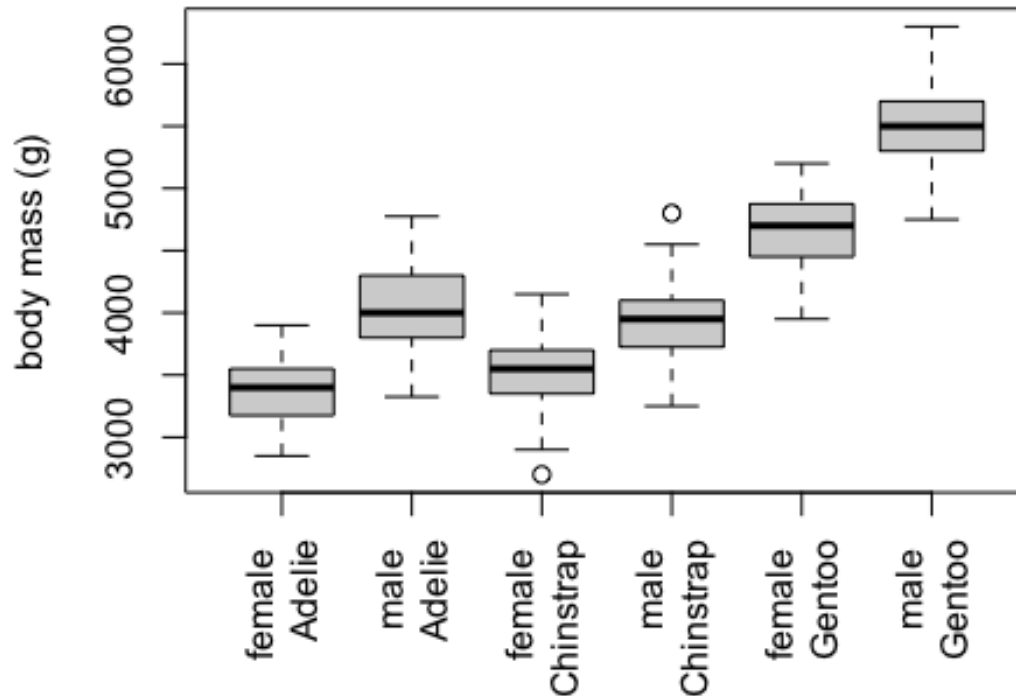
Boxplot for the fit_sex model



#Q8

```
boxplot(body_mass_g ~ sex*species, data = penguins, las = 3, names =  
c("female \n Adelie", "male \n Adelie", "female \n Chinstrap", "male \n  
Chinstrap", "female \n Gentoo", "male \n Gentoo"), ylab = "body mass (g)",  
xlab = "", main = "Boxplot for the fit_both model")
```

Boxplot for the fit_both model



#Q9

#fit_species model may have problems fulfilling the homogeneity assumption because the boxes have very different width.

#Q10. The null hypothesis of the Bartlett test is that the variances are the same among groups.

#Q11. 0.0501

```
bartlett.test(body_mass_g ~ species, data = penguins)
```

```
##
```

```
## Bartlett test of homogeneity of variances
```

```
##
```

```
## data: body_mass_g by species
```

```
## Bartlett's K-squared = 5.9895, df = 2, p-value = 0.05005
```

#Q12. 0.0319

```
bartlett.test(body_mass_g ~ sex, data = penguins)
```

```
##
```

```
## Bartlett test of homogeneity of variances
```

```
##
```

```
## data: body_mass_g by sex
```

```
## Bartlett's K-squared = 4.6017, df = 1, p-value = 0.03194
```

#Q13. 0.1741

```
dat_groups = aggregate(
  body_mass_g ~ sex*species,
  data = penguins,
  FUN = c)
str(dat_groups)

## 'data.frame':    6 obs. of  3 variables:
## $ sex          : Factor w/ 2 levels "female","male": 1 2 1 2 1 2
## $ species      : Factor w/ 3 levels "Adelie","Chinstrap",...: 1 1 2 2 3 3
## $ body_mass_g:List of 6
## ..$ : int  3800 3250 3450 3625 3200 3700 3450 3325 3400 3800 ...
## ..$ : int  3750 3650 4675 3800 4400 4500 4200 3600 3950 3800 ...
## ..$ : int  3500 3525 3950 3250 4150 3800 3700 3575 3700 3450 ...
## ..$ : int  3900 3650 3725 3750 3700 3775 4050 4050 3300 4400 ...
## ..$ : int  4500 4450 4550 4800 4400 4650 4650 4200 4150 4800 ...
## ..$ : int  5700 5700 5400 5200 5150 5550 5850 5850 6300 5350 ...
```

```
bartlett.test(dat_groups$body_mass_g)

##
## Bartlett test of homogeneity of variances
##
## data:  dat_groups$body_mass_g
## Bartlett's K-squared = 7.6908, df = 5, p-value = 0.1741
```

#Q14

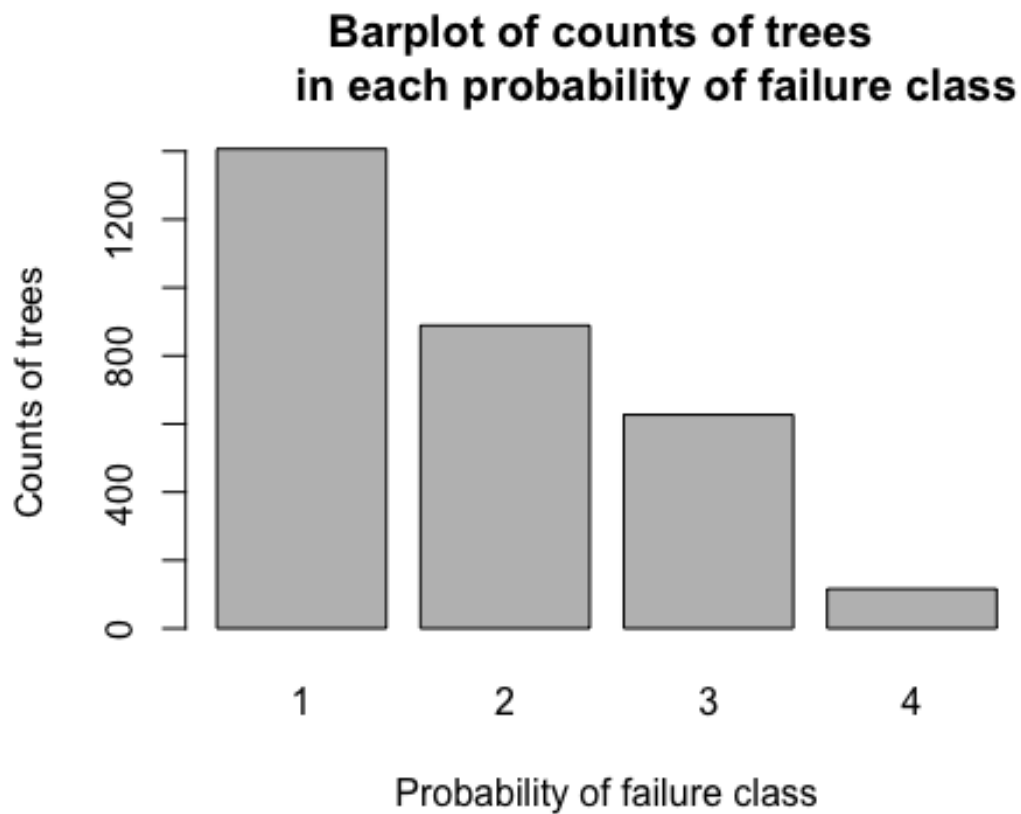
#The fit_sex model has issues with heterogeneity. The p-value is lower than 0.05, so we reject the null hypothesis that the variances are the same among groups.

#Q15

```
dat_fl =
read.csv("/Users/stonehuang/Documents/environmental_data/data/trees_FL.csv")
head(dat_fl)
```

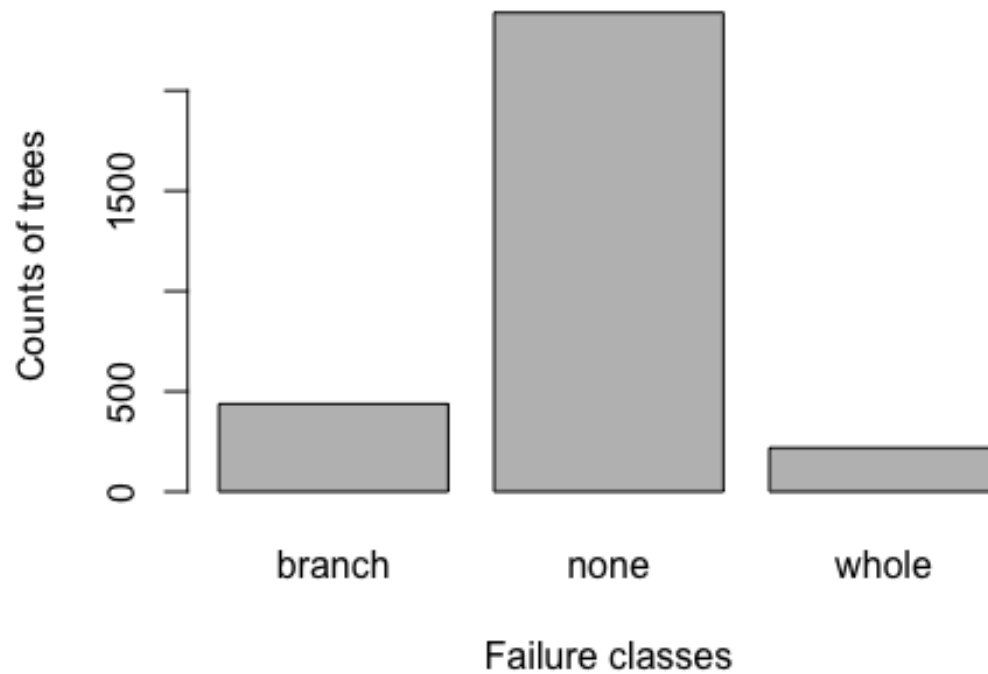
```
##   DBH_in HeighttoTop_ft   Species CanopyWidth_ft MostSignificantDefect
## 1    8.0          12.5 Chinese elm           13                Dead
## 2    8.0          16.0 Chinese elm           20                Decay
## 3    5.5          15.5 Crapemyrtle           15 Poor Tree Architecture
## 4    6.2          14.0 Crapemyrtle           24 Poor Tree Architecture
## 5   16.0          17.0 Crapemyrtle           27 Poor Tree Architecture
## 6   15.2          20.0 Crapemyrtle           19 Poor Tree Architecture
##   Failure_Standardized ProbabilityofFailure
## 1                    whole                4
## 2                    whole                3
## 3                    whole                1
## 4                    whole                1
## 5                     none                2
## 6                     none                2
```

```
barplot(table(dat_fl$ProbabilityofFailure), ylab = "Counts of trees", xlab =  
"Probability of failure class", main = "Barplot of counts of trees  
in each probability of failure class")
```



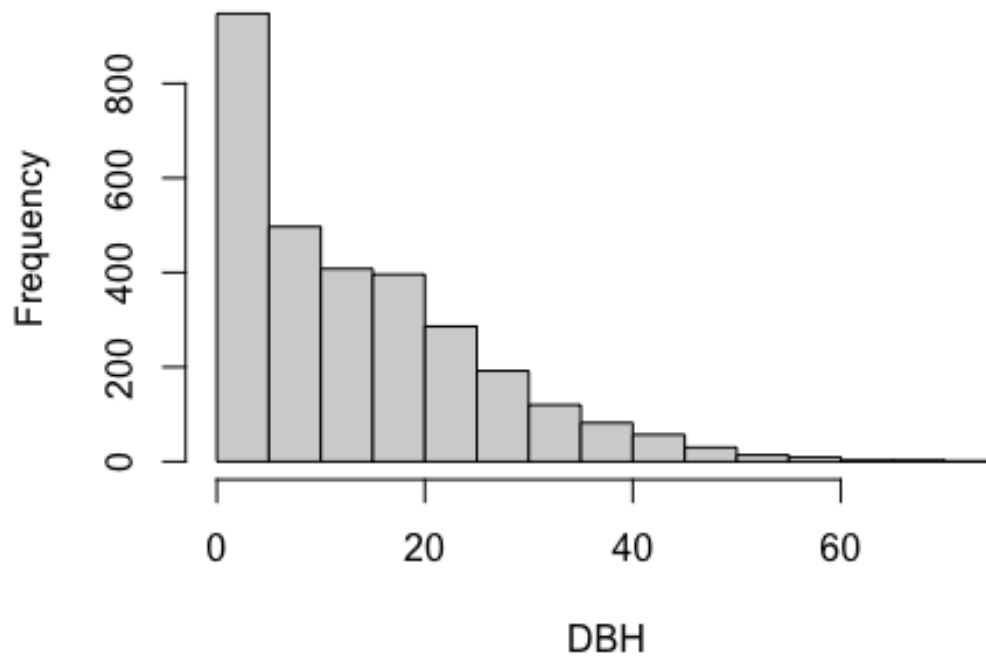
```
barplot(table(dat_fl$Failure_Standardized), ylab = "Counts of trees", xlab =  
"Failure classes", main = "Barplot of the counts of trees  
in each of the failure classes")
```

**Barplot of the counts of trees
in each of the failure classes**



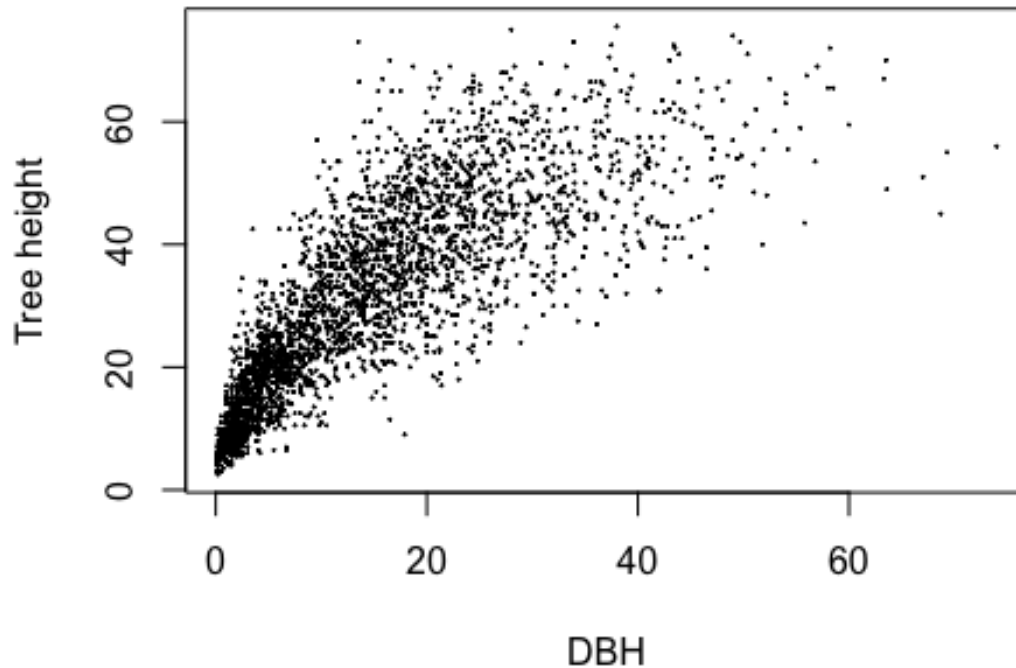
```
hist(dat_fl$DBH_in, xlab = "DBH", main = "Histogram of DBH")
```


Histogram of DBH



```
plot(dat_fl$HeighttoTop_ft ~ dat_fl$DBH_in, ylab = "Tree height", xlab =  
"DBH", main = "Scatterplot of DBH and tree height", cex = 0.1)
```

Scatterplot of DBH and tree height



#Q16. The null hypothesis for the Kolmogorov-Smirnov test is that there is no difference in DBH between whole-tree failures and intact trees.

#Q17. 0.02125. The distribution of DBH is not the same for the two groups. The p-value is lower than 0.05 so we reject the null hypothesis.

```
whole = subset(dat_fl, Failure_Standardized == "whole")
```

```
none = subset(dat_fl, Failure_Standardized == "none")
```

```
ks.test(whole$DBH_in, none$DBH_in)
```

```
## Warning in ks.test(whole$DBH_in, none$DBH_in): p-value will be approximate  
in
```

```
## the presence of ties
```

```
##
```

```
## Two-sample Kolmogorov-Smirnov test
```

```
##
```

```
## data: whole$DBH_in and none$DBH_in
```

```
## D = 0.10643, p-value = 0.02125
```

```
## alternative hypothesis: two-sided
```

#Q18. Tree height increases with increasing DBH. The relationship is curved and monotonic.

#Q19. Spearman

```
cor.test(  
  dat_fl$DBH_in,
```

```

dat_fl$HeighttoTop_ft,
method='spearman',
exact=FALSE)

##
## Spearman's rank correlation rho
##
## data: dat_fl$DBH_in and dat_fl$HeighttoTop_ft
## S = 541454081, p-value < 2.2e-16
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
## rho
## 0.8850462

#Q20.  $p\text{-value} < 2.2e-16$ . I conclude that the two variables are significantly
correlated.

#Q21.  $X\text{-squared} = 202.65$ ,  $p\text{-value} < 2.2e-16$ 
dat_fl$fail = factor(dat_fl$Failure_Standardized != "none")
levels(dat_fl$fail) = c("No Fail", "Fail")
fl_table_2 = table(
  dat_fl$ProbabilityofFailure,
  dat_fl$fail)
fl_table_2

##
##      No Fail Fail
## 1      1239  168
## 2       670  219
## 3       421  206
## 4        53   63

chisq_trees = chisq.test(fl_table_2)
chisq_trees

##
## Pearson's Chi-squared test
##
## data: fl_table_2
## X-squared = 202.65, df = 3, p-value < 2.2e-16

#Q22. -136
round(
  chisq_trees$observed - chisq_trees$expected,
  digits = 0)

##
##      No Fail Fail
## 1      136 -136
## 2      -27  27

```

##	3	-71	71
##	4	-38	38

#Q23. There were fewer tree failures than expected by chance in failure probability category #1.

#Q24. There more tree failures than expected by chance in failure probability category #4.

#Q25. I conclude that the probability of failure rating system is effective! There were more tree failures than expected in higher failure probability categories.