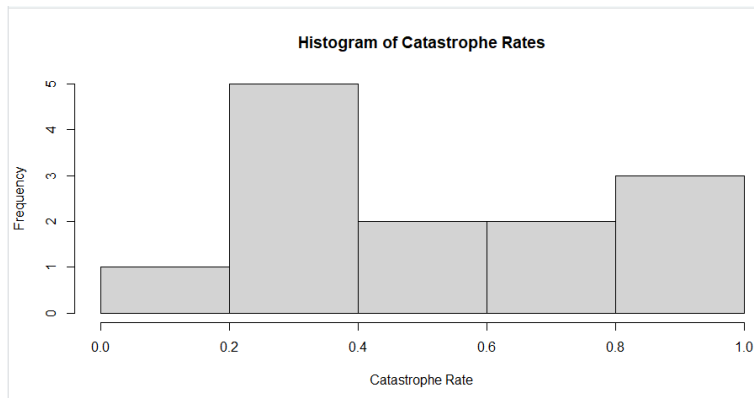


1.



2.

```
> shapiro.test(catrate$cat.rate)

      Shapiro-Wilk normality test

data:  catrate$cat.rate
W = 0.86202, p-value = 0.04097
```

3. The null hypothesis for the Shapiro test is that the data was sampled from a normal distribution.

4. The p-value is less than the standard alpha of 0.05 so there is strong evidence to reject the null, so there is strong evidence that the data came from a non-normal distribution.

5.

```
> t.test(catrate$cat.rate, mu = 2/7)

      One Sample t-test

data:  catrate$cat.rate
t = 2.9595, df = 12, p-value = 0.01193
alternative hypothesis: true mean is not equal to 0.2857143
95 percent confidence interval:
 0.3526250 0.7261295
sample estimates:
mean of x
0.5393773
```

6. The null hypothesis is that there is no difference between the catastrophic rate and the pond late-filling rate.

7. This is a two-tailed t-test because we are looking for a difference, not necessarily a sided (greater/less than) difference.

8. The p-value from the test is .01193. If this was a false positive rate, then this would mean that in one out every 100 experiments (~1%) you would have a false positive value.

9. The confidence interval was a 95% interval from 0.353 to 0.726 the interval does not include 0.

10. From the t-test, there is evidence to reject the null because the p-value is less than the standard of 0.05, but it is not so much less so the evidence is definitely present, but not that strong.

11.

```
> wilcox.test(catrate$cat.rate, mu = 2/7)

      Wilcoxon signed rank test with continuity correction

data:  catrate$cat.rate
V = 85, p-value = 0.006275
alternative hypothesis: true location is not equal to 0.2857143
```

12. P-value for Wilcoxon = 0.006275, P-value for T-test = .01193. Both values are less than alpha (.05), but the value for the Wilcoxon test is an order of magnitude smaller than the T-test value.
13. In this case there is strong evidence to reject the null hypothesis because the p-value is much smaller than the established value of alpha (.05).
14. In both cases, we reject the null because the p-values are smaller than the alpha of .05. For the t-test this means that there is a difference between the observed catastrophe rate and catastrophe rate specifically from late pond-filling. In the Wilcoxon test, this means that there is also a difference between the two, but this time it accounts for the fact that the sample is not normally distributed.
15. I think that the Wilcoxon test is more appropriate for this data set. Both the histogram and the Shapiro test indicate that the data is not normally distributed, so the Wilcoxon test is adjusted for the non-normality.

16.

```
> dat_adelie = subset(penguin_dat, species == "Adelie")
> shapiro.test(dat_adelie$flipper_length_mm)

      Shapiro-Wilk normality test

data:  dat_adelie$flipper_length_mm
W = 0.99339, p-value = 0.72

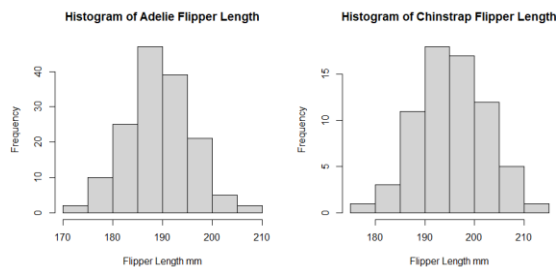
> dat_chinstrap = subset(penguin_dat, species == "Chinstrap")
> shapiro.test(dat_chinstrap$flipper_length_mm)

      Shapiro-Wilk normality test

data:  dat_chinstrap$flipper_length_mm
W = 0.98891, p-value = 0.8106
```

17. The p-values in both cases are greater than the standard alpha of .05 so this means that we do not have sufficient evidence to reject the null. The null hypothesis being that the data came from a normally distributed population.

18.



19. The alternative hypothesis is that there is a difference in flipper length between the two penguin species.

20.

```
> t.test(flipper_length_mm ~ species, data = penguin_dat)

      Welch Two Sample t-test

data:  flipper_length_mm by species
t = -5.7804, df = 119.68, p-value = 6.049e-08
alternative hypothesis: true difference in means between group Adelie and group Chinstrap is not equal to 0
95 percent confidence interval:
 -7.880530 -3.859244
sample estimates:
mean in group Adelie mean in group Chinstrap
      189.9536         195.8235
```