

# Auckland Cycle analysis

Lawrence May

Total number of cyclists counted for each day, with rainfall.

```
library(tidyverse)
```

```
## -- Attaching packages -----
```

```
## v ggplot2 3.3.2    v purrr  0.3.4
## v tibble  3.0.3    v dplyr  1.0.0
## v tidyr   1.1.0    v stringr 1.4.0
## v readr   1.3.1    v forcats 0.5.0
```

```
## -- Conflicts -----
```

```
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

```
#Read in bike data
```

```
bike_files<- list.files("/Users/lawrence/Google Drive/UNI/current/Stats 369/a1/",pattern=".csv",full=TRUE)
bike_data<-map(bike_files,read_csv)
```

```
## Parsed with column specification:
## cols(
##   .default = col_double(),
##   Date = col_character()
## )
```

```
## See spec(...) for full column specifications.
```

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```

```

#Join into one dataframe
bike_data <- dplyr::bind_rows(bike_data[[1]],bike_data[[2]],bike_data[[3]])

bike <- subset(bike_data, select = -c(Date))
bike[is.na(bike)] <- 0

#Take sum of all cyclists on each day
bike<-rowSums(bike)
bike_data <- data.frame(bike_data$Date,bike)
bike_data %>% rename(Date = bike_data.Date, Count = bike) -> bikes

#Read in rain data
rain_files<- list.files("/Users/lawrence/Google Drive/UNI/current/Stats 369/a1/",pattern=".txt",full=TRUE)
rain_data<-map(rain_files,read_csv,skip=9)

```

```

## Parsed with column specification:
## cols(
##   Station = col_double(),
##   'Date(NZST)' = col_double(),
##   'Time(NZST)' = col_character(),
##   'Amount(mm)' = col_double(),
##   'Period(Hrs)' = col_double(),
##   Freq = col_character()
## )

## Warning: 12 parsing failures.
##   row    col expected          actual
## 35091 Station a double  UserName is = tslumley      '/Users/lawrence/Google D
## 35091 NA        6 columns 1 columns      '/Users/lawrence/Google D
## 35092 Station a double  Total number of rows output = 35090      '/Users/lawrence/Google D
## 35092 NA        6 columns 1 columns      '/Users/lawrence/Google D
## 35093 Station a double  Number of rows remaining in subscription = 1793302 '/Users/lawrence/Google D
## .....
## See problems(...) for more details.

```

```

## Parsed with column specification:
## cols(
##   Station = col_double(),
##   'Date(NZST)' = col_double(),
##   'Time(NZST)' = col_character(),
##   'Amount(mm)' = col_double(),
##   'Period(Hrs)' = col_double(),
##   Freq = col_character()
## )

## Warning: 12 parsing failures.
##   row    col expected          actual
## 17305 Station a double  UserName is = tslumley      '/Users/lawrence/Google D
## 17305 NA        6 columns 1 columns      '/Users/lawrence/Google D
## 17306 Station a double  Total number of rows output = 17304      '/Users/lawrence/Google D
## 17306 NA        6 columns 1 columns      '/Users/lawrence/Google D
## 17307 Station a double  Number of rows remaining in subscription = 1775998 '/Users/lawrence/Google D

```

```
## .....
## See problems(...) for more details.
```

```
#Join into one dataframe
rain <- dplyr::bind_rows(rain_data[[1]],rain_data[[2]])
rain %>% filter(Station == 22719) -> rain1
rain %>% filter(Station == 37852) -> rain2

#Aggregate daily rainfall for each locationn
aggregate(rain1$'Amount(mm)' ~ rain1$'Date(NZST)', data=rain1, FUN=sum) -> rain1
aggregate(rain2$'Amount(mm)' ~ rain2$'Date(NZST)', data=rain2, FUN=sum) -> rain2

colnames(rain1) <- c("Date", "Rain")
colnames(rain2) <- c("Date", "Rain")
rain <- rbind(rain1,rain2)

#Take average of daily rain in Mangere and Albany to get approximation for average rainfall across Auckland
aggregate(Rain ~ Date, data=rain, FUN=mean) -> rain

bikes %>% separate(col=Date, into=c("dow","day","month","year")) %>% unite(Date,c(day,month,year), sep = "-")

#Change date to same format and join on date
bikes$Date <- as.Date(bikes$Date, format='%d %b %Y')
rain$Date <- as.Date(as.character(rain$Date), format='%Y%m%d')
combined_df <- inner_join(bikes,rain,by = 'Date')
head(combined_df)
```

```
##   dow      Date month year Count  Rain
## 1 Fri 2016-01-01   Jan 2016  1299 20.25
## 2 Sat 2016-01-02   Jan 2016  1030 19.15
## 3 Sun 2016-01-03   Jan 2016  7423  6.80
## 4 Mon 2016-01-04   Jan 2016 11956  0.05
## 5 Tue 2016-01-05   Jan 2016 10167  0.00
## 6 Wed 2016-01-06   Jan 2016 10387  0.00
```

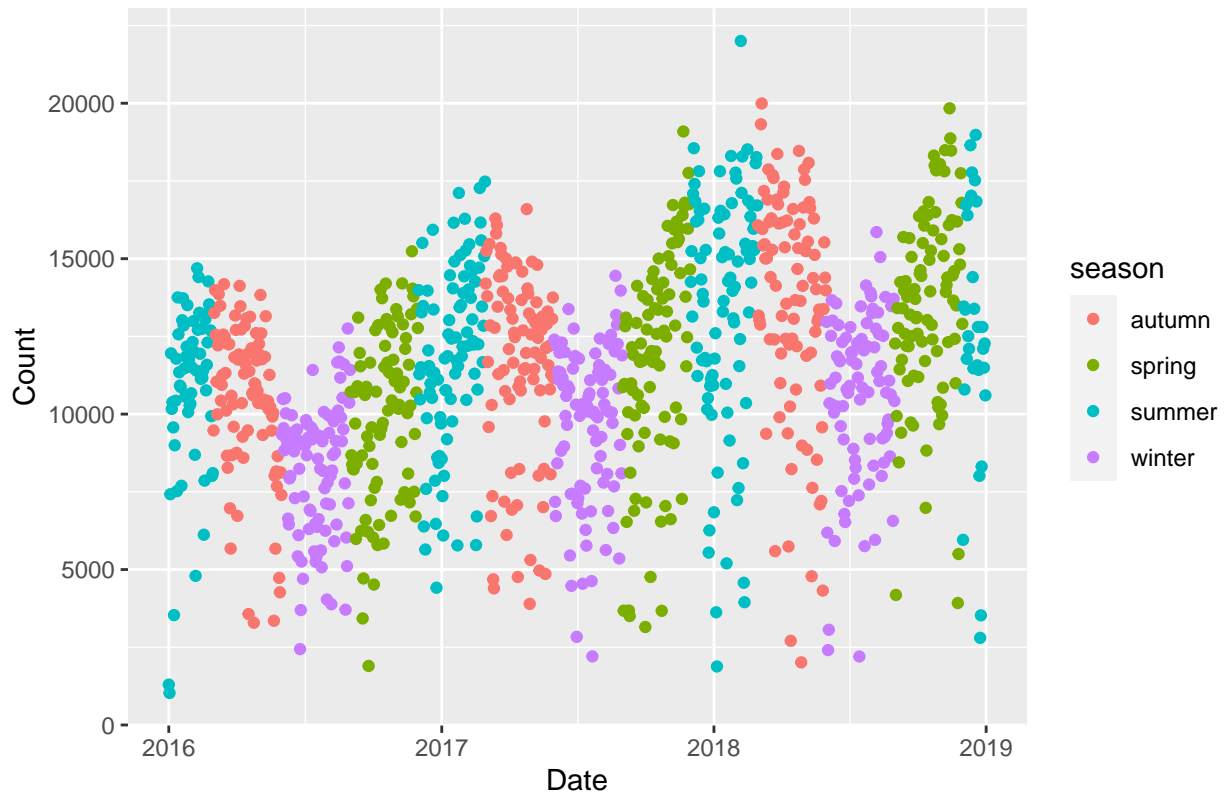
Some graphical analysis

```
#Create season factor variable
spring <- c("Sep", "Oct", "Nov")
summer <- c("Dec", "Jan", "Feb")
autumn <- c("Mar", "Apr", "May")
winter <- c("Jun", "Jul", "Aug")

#Insert into dataframe
combined_df <- mutate(combined_df, season = factor(case_when(month %in% spring ~ "spring",
                                                             month %in% summer ~ "summer", month %in% autumn ~ "autumn",
                                                             month %in% winter ~ "winter",
                                                             TRUE ~ NA_character_)))

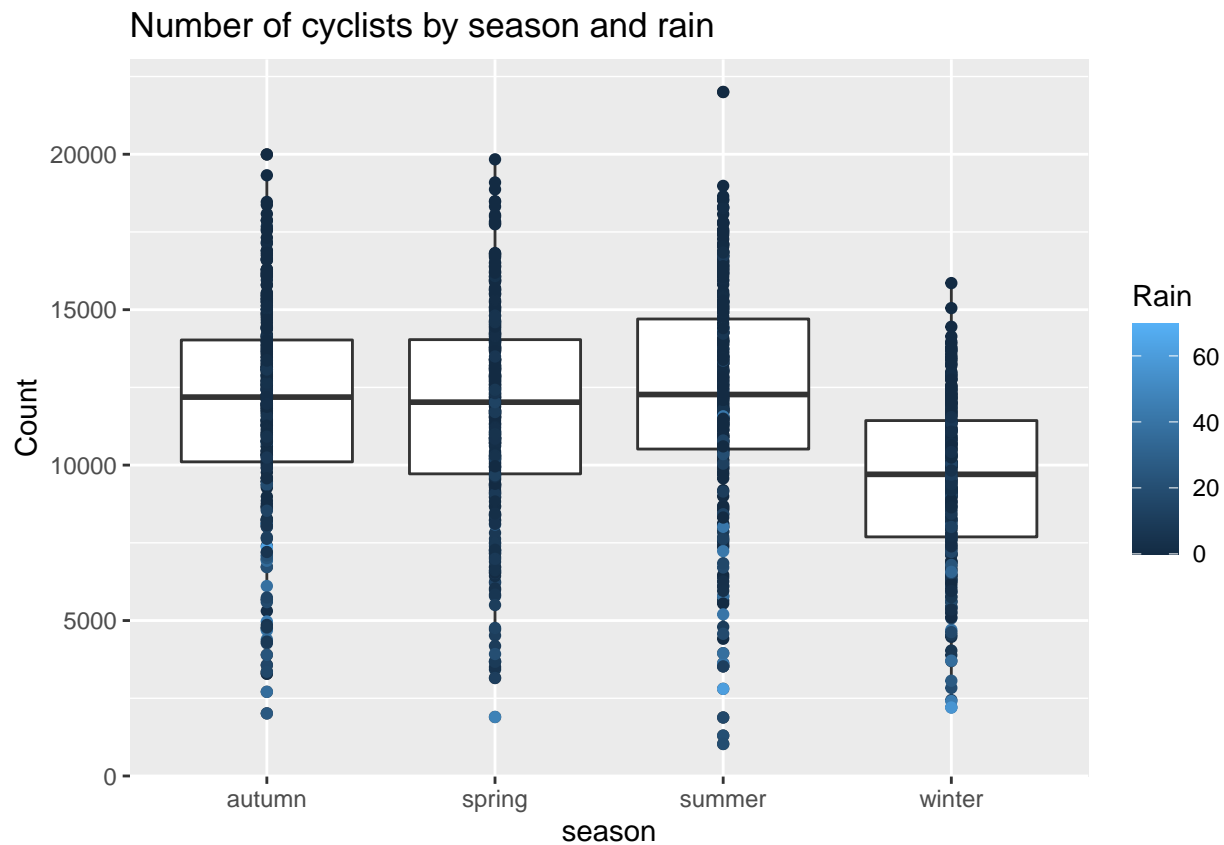
ggplot(combined_df, aes(Date, Count, color = season)) + geom_point()+ggtitle("Number of cyclists over time")
```

Number of cyclists over time and by season



There appears to be an increasing trend of the number of cyclists over time, with a strong seasonal component (higher number of cyclist in spring and summer).

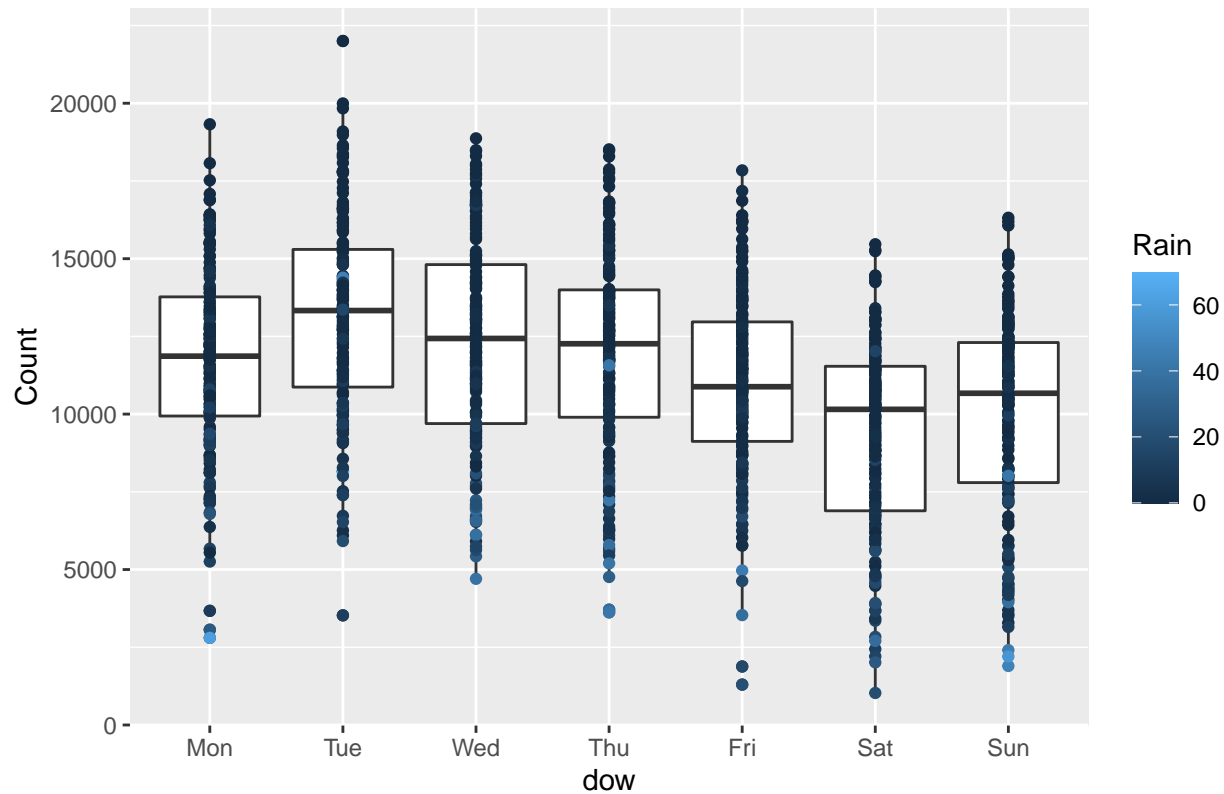
```
ggplot(combined_df, aes(season, Count, color = Rain)) + geom_boxplot()+ggtitle("Number of cyclists by s
```



There is a strong seasonal pattern with most cyclists in the summer, spring and autumn months and far less in the winter months. Rain does appear to have a negative effect as well, and seems to be able to explain many of the negative outliers (light blue = heavy rain).

```
combined_df$dow<-factor(combined_df$dow , levels=c('Mon','Tue','Wed','Thu','Fri','Sat','Sun'))
ggplot(combined_df, aes(dow, Count, color = Rain)) + geom_boxplot()+ggtitle("Number of cyclists by day of week")
```

Number of cyclists by day of the week and rain

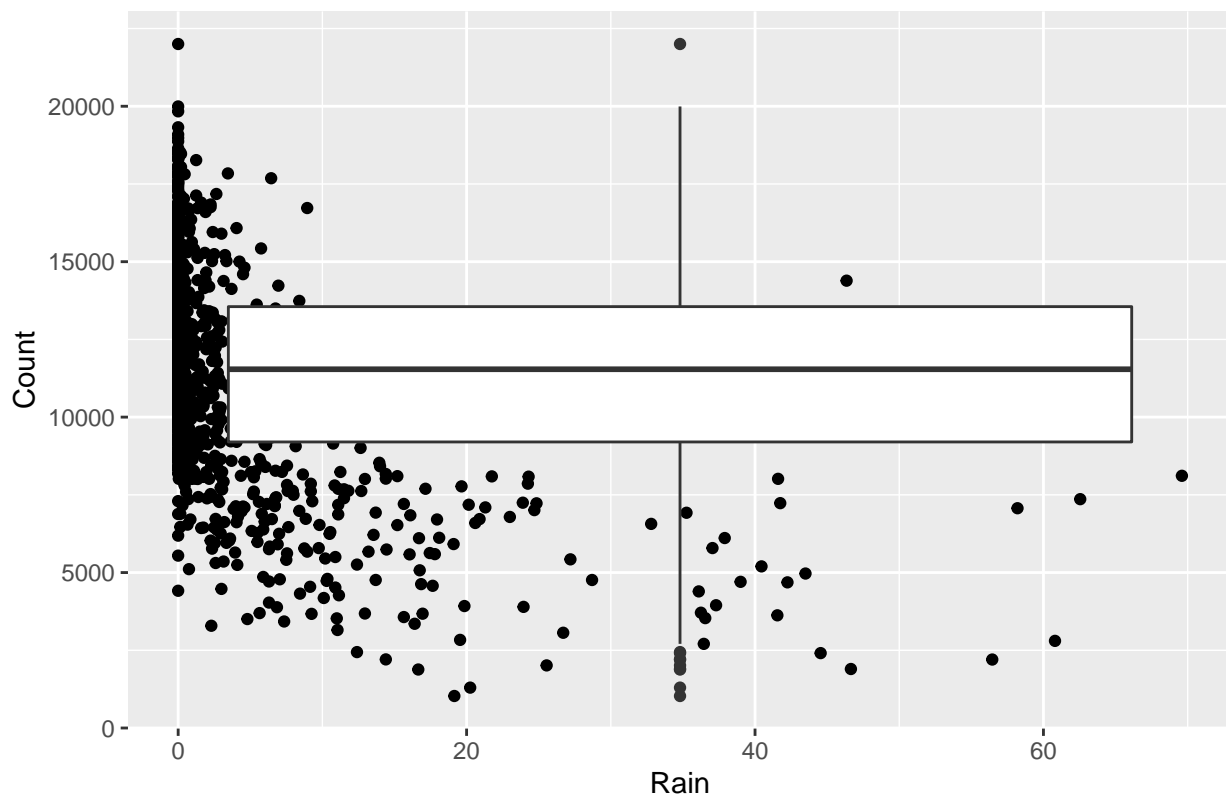


There also appears to be a Day of the week effect, with most cyclists riding on Tuesdays and the least amount of Saturdays.

```
ggplot(combined_df, aes(Rain, Count)) + ggtitle("Number of cyclists and the effect of rain") + geom_point
```

```
## Warning: Continuous x aesthetic -- did you forget aes(group=...)?
```

## Number of cyclists and the effect of rain



More rain leads to less than average number of cyclists.

Simple regression model to estimate number of cyclists and effect of rain on cycle activity

```
combined_df$year <- as.numeric(combined_df$year)
bike.lm <- lm(Count ~ year + season + dow + Rain, data = combined_df)
summary(bike.lm)
```

```
##
## Call:
## lm(formula = Count ~ year + season + dow + Rain, data = combined_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -10259.7  -1232.2    194.5   1358.7   9367.8
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -3.121e+06  1.623e+05 -19.228  < 2e-16 ***
## year         1.554e+03   8.048e+01  19.308  < 2e-16 ***
## seasonspring -3.924e+02   1.859e+02  -2.111  0.034968 *
## seasonsummer  3.098e+02   1.859e+02   1.667  0.095891 .
## seasonwinter -2.361e+03   1.850e+02 -12.765  < 2e-16 ***
## dowTue        1.403e+03   2.457e+02   5.710  1.46e-08 ***
## dowWed         8.775e+02   2.459e+02   3.569  0.000374 ***
## dowThu         4.781e+02   2.458e+02   1.945  0.051996 .
## dowFri        -7.154e+02   2.453e+02  -2.917  0.003608 **
## dowSat        -2.246e+03   2.453e+02  -9.157  < 2e-16 ***
```

```
## dowSun      -1.574e+03  2.453e+02  -6.417 2.07e-10 ***
## Rain        -2.249e+02  8.229e+00 -27.328 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2173 on 1084 degrees of freedom
## Multiple R-squared:  0.6097, Adjusted R-squared:  0.6057
## F-statistic: 153.9 on 11 and 1084 DF,  p-value: < 2.2e-16
```

There is a highly significant effect of year on cyclists, the number of cyclists increases by 1500 every year, on average. There is also a strong seasonal effect, most notably of winter which decreases the number of cyclists by 2400 compared to the baseline autumn. As expected by the graph, there is also a strong effect of day of the week, most notably Tuesday which, on average, has 1400 cyclists more than a typical Monday.

Based on the model, rain does have a large, and highly significant, effect on people cycling in Auckland: each additional millimeter of rain precipitation per day resulted in 225 people less cycling per day. This can be seen in the graphs as well: On days with higher than usual precipitation (i.e more than 40mm) the count of people cycling is significantly less than what is usually the case in that month.