

Step 1: Loading in the Data

```
# Packages used in tutorials
library(MASS)      # boxcox
library(car)       # qqPlot

## Loading required package: carData
library(randtests) # runs.test

## Warning: package 'randtests' was built under R version 4.3.3
# library(forecast) # OPTIONAL if you want auto.arima, not required

bike <- read.csv("trips_per_day.csv")
bike$trip_date <- as.Date(bike$trip_date)

str(bike)

## 'data.frame': 2969 obs. of 2 variables:
## $ trip_date: Date, format: "2016-01-10" "2016-01-11" ...
## $ n_trips : int 2273 3623 2535 2966 2970 2636 4122 3104 1642 4834 ...

head(bike)

##   trip_date n_trips
## 1 2016-01-10   2273
## 2 2016-01-11   3623
## 3 2016-01-12   2535
## 4 2016-02-10   2966
## 5 2016-02-11   2970
## 6 2016-02-12   2636

range(bike$trip_date)

## [1] "2016-01-10" "2024-09-30"
```

Initial Plotting for Time Series

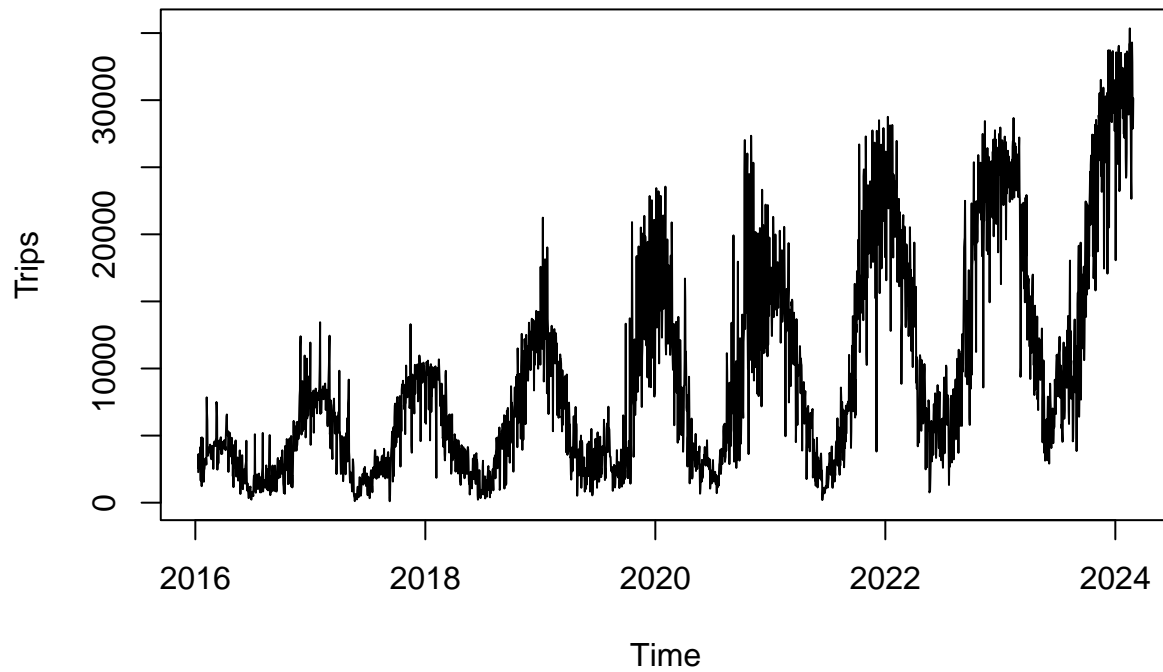
```
# Sort just in case
bike <- bike[order(bike$trip_date), ]

# Extract response as a vector
y <- bike$n_trips

# Daily frequency with yearly seasonality (approx 365)
bike_ts <- ts(
  y,
  start = c(as.numeric(format(min(bike$trip_date), "%Y")),
            as.numeric(format(min(bike$trip_date), "%j"))),
  frequency = 365
)

plot(bike_ts, main = "Daily BikeShare Trips in Toronto", ylab = "Trips")
```

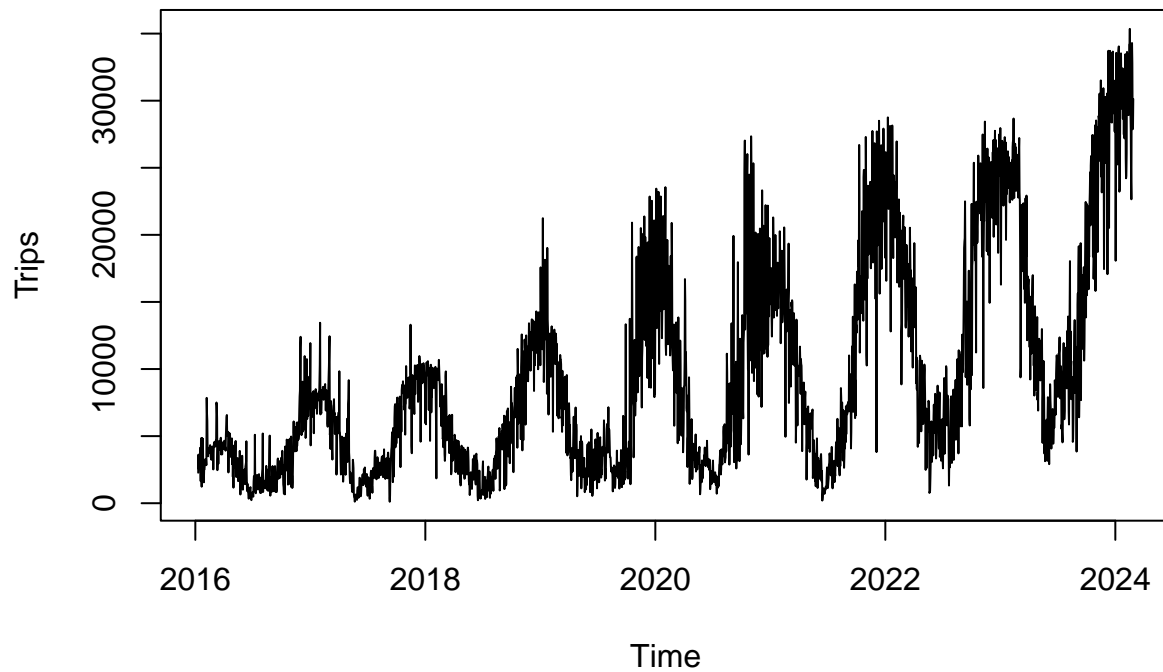
## Daily BikeShare Trips in Toronto



Step 2: EDA

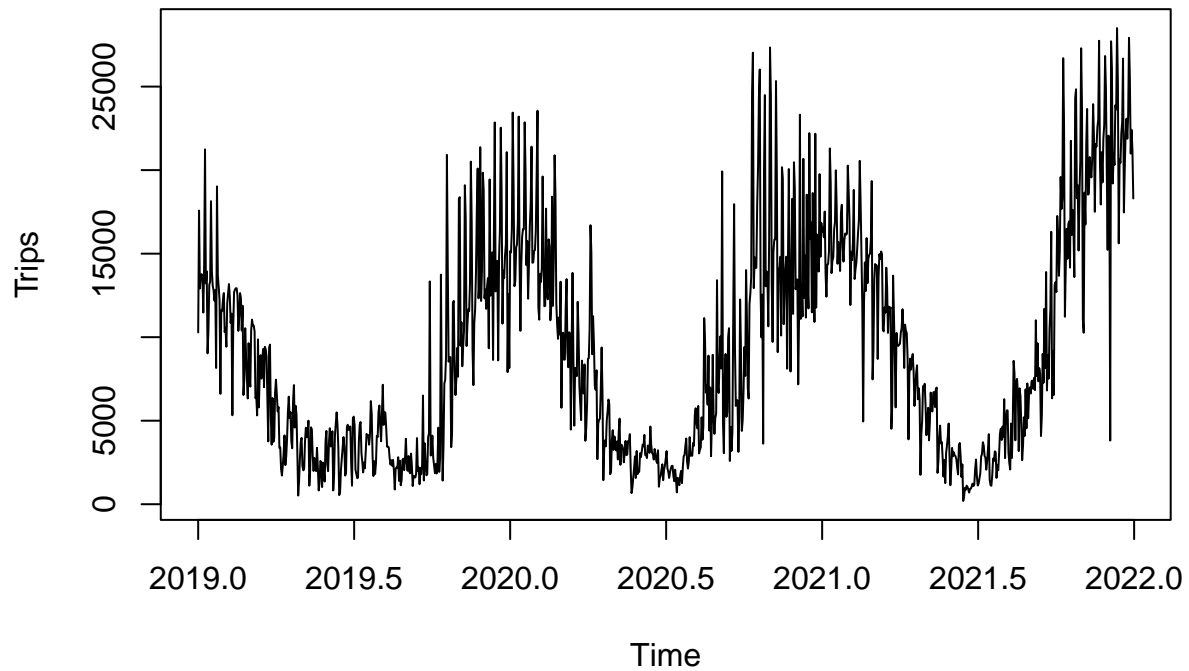
```
par(mfrow = c(1, 1))  
plot(bike_ts, main = "Daily Trips", ylab = "Trips")
```

## Daily Trips



```
# maybe a zoom on a couple of years
plot(window(bike_ts, start = c(2019, 1), end = c(2021, 365)),
     main = "Daily Trips: 2019-2021", ylab = "Trips")
```

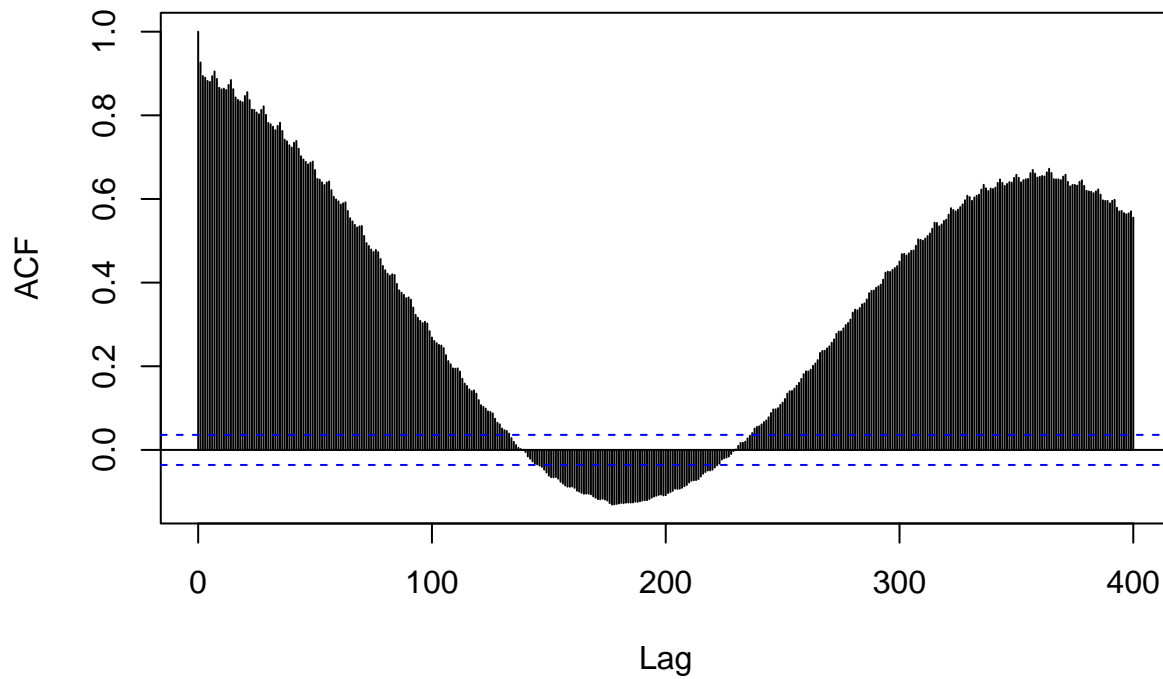
### Daily Trips: 2019–2021



Confirming Seasonality with ACF and Spectrum

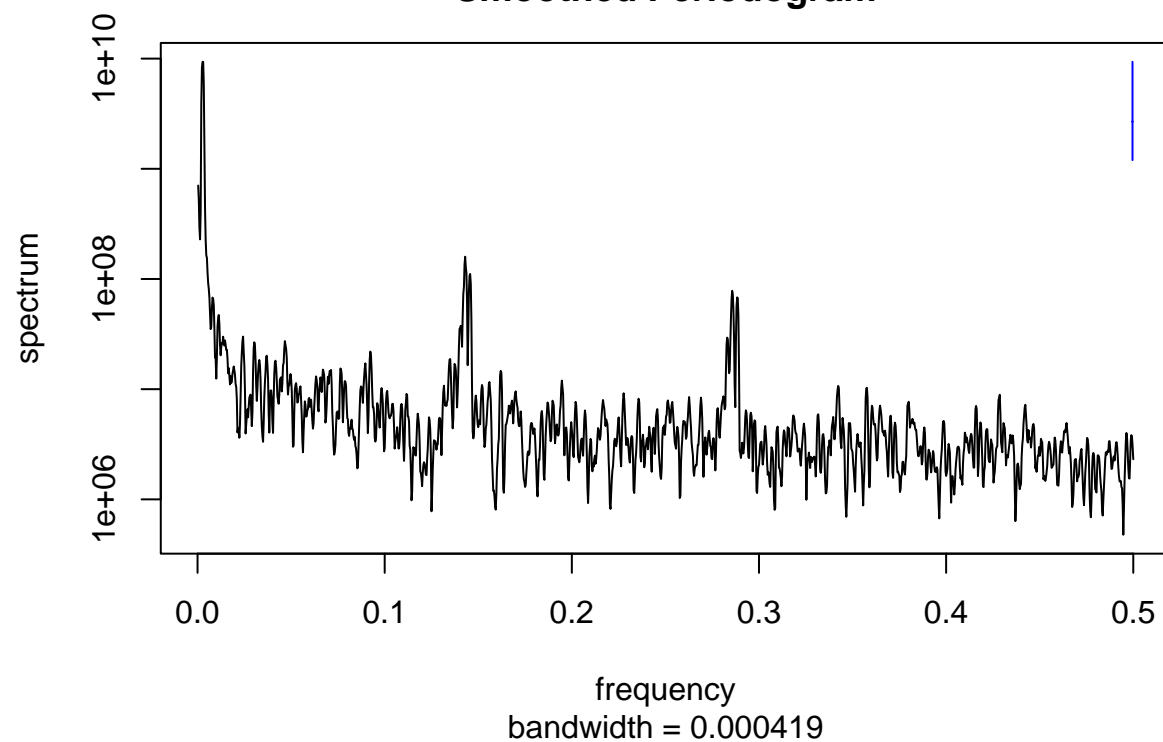
```
acf(as.vector(bike_ts), lag.max = 400,
    main = "ACF of Daily Trips")
```

## ACF of Daily Trips



```
spec_bike <- spectrum(as.vector(bike_ts), spans = 5)
```

## Series: x Smoothed Periodogram



```
1 / spec_bike$freq[which.max(spec_bike$spec)] # estimated period
```

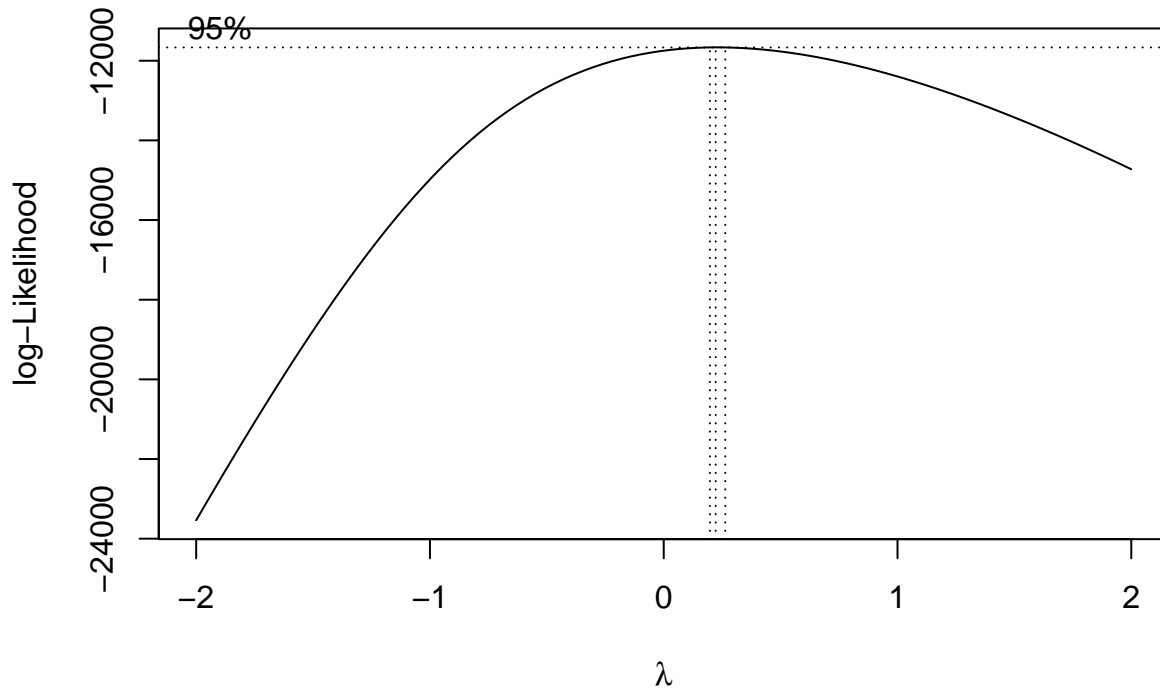
```
## [1] 333.3333
```

Step 3: Box-Cox transformation

```
# Simple intercept-only model (like in tutorial)
```

```
bc_model_raw <- lm(bike_ts ~ 1)
```

```
boxcox_raw <- MASS::boxcox(bc_model_raw, lambda = seq(-2, 2, 0.1))
```



```
(lambda_opt_raw <- boxcox_raw$x[which.max(boxcox_raw$y)])
```

```
## [1] 0.2222222
```

```
tim <- time(bike_ts) # continuous time index
```

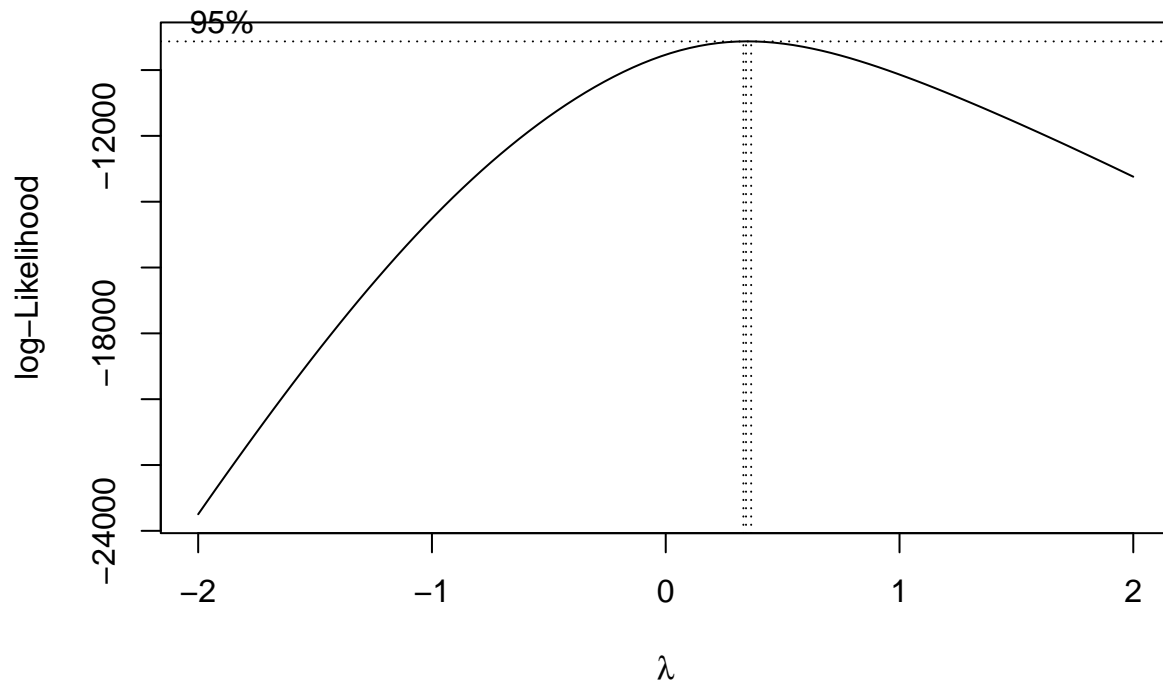
```
# Season: year and day-of-year or month; simplest is month
```

```
# Build a monthly factor from dates (instead of cycle, since this is daily)
```

```
month <- factor(format(bike$trip_date, "%m"))
```

```
reg_for_bc <- lm(bike_ts ~ tim + month)
```

```
boxcox_mod <- MASS::boxcox(reg_for_bc, lambda = seq(-2, 2, 0.1))
```



```
(lambda_opt_mod <- boxcox_mod$x[which.max(boxcox_mod$y)])
```

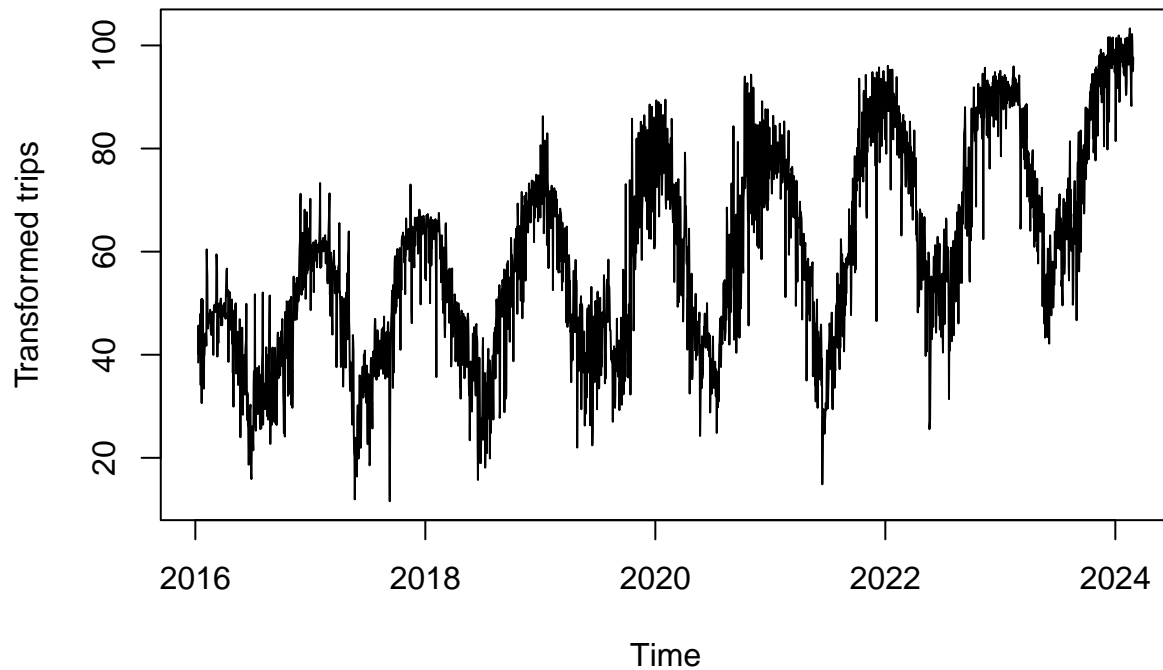
```
## [1] 0.3434343
```

```
lam <- lambda_opt_mod  # keep this for later
```

```
if (lam == 0) {
  y_trans <- log(bike_ts)
} else if (lam > 0) {
  y_trans <- (bike_ts^lam - 1) / lam
} else {
  # negative lambda + use minus sign trick like in lectures
  y_trans <- -(bike_ts^lam)
}
```

```
plot(y_trans, main = "Transformed Daily Trips", ylab = "Transformed trips")
```

## Transformed Daily Trips

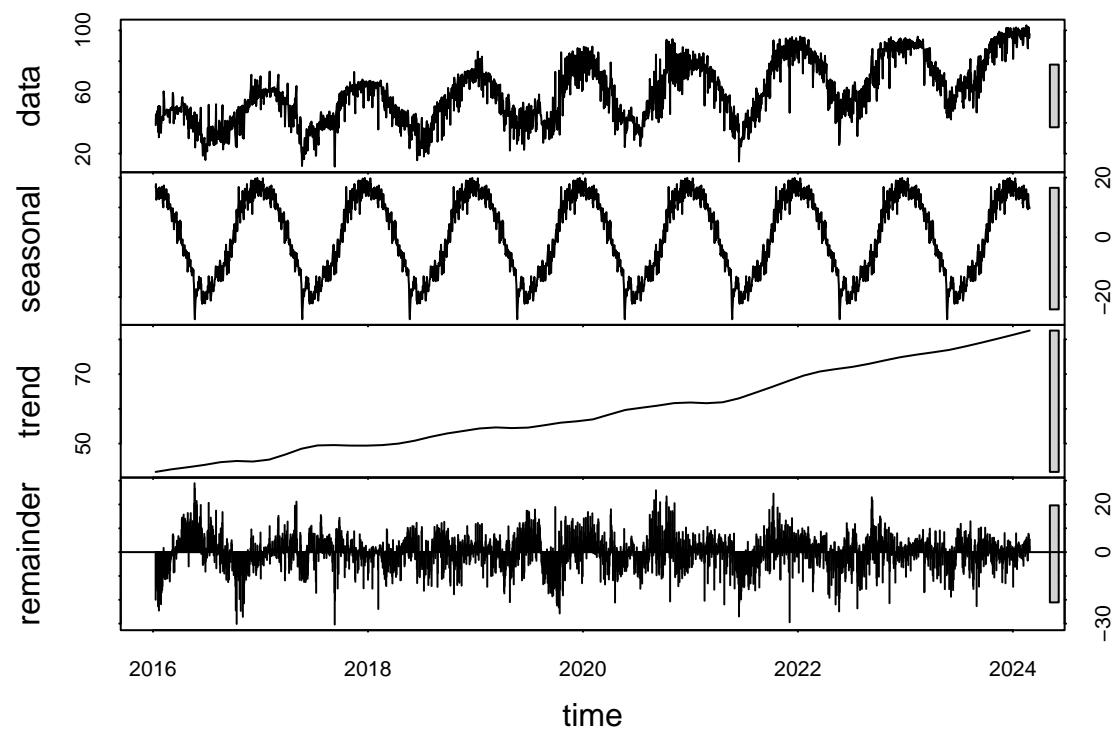


cal Decomposition

```
Decomp_bike <- stl(y_trans, s.window = "periodic")  
plot(Decomp_bike, main = "STL Decomposition of Transformed Daily Trips")
```

Classi-

## STL Decomposition of Transformed Daily Trips



```

# Extract components if needed
bike_seasonal <- Decomp_bike$time.series[, "seasonal"]
bike_trend    <- Decomp_bike$time.series[, "trend"]
bike_remainder <- Decomp_bike$time.series[, "remainder"]

tim    <- time(y_trans)
month  <- factor(format(bike$trip_date, "%m"))
dow    <- factor(weekdays(bike$trip_date)) # optional, but nice

# Simple model: linear trend + monthly seasonality
reg1 <- lm(y_trans ~ tim + month)
summary(reg1)

##
## Call:
## lm(formula = y_trans ~ tim + month)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -36.930  -4.421   0.841   5.135  33.503
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -1.010e+04  1.288e+02 -78.419  < 2e-16 ***
## tim          5.022e+00  6.378e-02  78.738  < 2e-16 ***
## month02      6.976e-01  7.537e-01   0.926  0.354735
## month03      6.327e+00  7.358e-01   8.599  < 2e-16 ***
## month04      1.183e+01  7.421e-01  15.935  < 2e-16 ***
## month05      2.364e+01  7.360e-01  32.113  < 2e-16 ***
## month06      3.001e+01  7.424e-01  40.431  < 2e-16 ***
## month07      3.270e+01  7.130e-01  45.862  < 2e-16 ***
## month08      3.376e+01  7.130e-01  47.344  < 2e-16 ***
## month09      3.207e+01  7.186e-01  44.635  < 2e-16 ***
## month10      2.345e+01  7.404e-01  31.666  < 2e-16 ***
## month11      1.429e+01  7.468e-01  19.132  < 2e-16 ***
## month12      2.715e+00  7.403e-01   3.667  0.000249 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.143 on 2956 degrees of freedom
## Multiple R-squared:  0.8212, Adjusted R-squared:  0.8205
## F-statistic: 1131 on 12 and 2956 DF, p-value: < 2.2e-16

reg2 <- lm(y_trans ~ poly(tim, 2, raw = FALSE) + month)
summary(reg2)

##
## Call:
## lm(formula = y_trans ~ poly(tim, 2, raw = FALSE) + month)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -36.363  -4.468   0.913   5.123  33.023
##
## Coefficients:

```

```
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)      41.6116    0.5175  80.412 < 2e-16 ***
## poly(tim, 2, raw = FALSE)1 642.5618    8.1150  79.182 < 2e-16 ***
## poly(tim, 2, raw = FALSE)2  47.9377    8.1680   5.869 4.87e-09 ***
## month02           0.6949    0.7495   0.927 0.353935
## month03           6.3211    0.7317   8.639 < 2e-16 ***
## month04          11.8129    0.7379  16.008 < 2e-16 ***
## month05          23.6142    0.7319  32.264 < 2e-16 ***
## month06          29.9821    0.7382  40.614 < 2e-16 ***
## month07          32.4508    0.7102  45.690 < 2e-16 ***
## month08          33.5098    0.7102  47.183 < 2e-16 ***
## month09          31.8278    0.7158  44.468 < 2e-16 ***
## month10          23.4850    0.7363  31.897 < 2e-16 ***
## month11          14.3325    0.7427  19.299 < 2e-16 ***
## month12           2.7584    0.7362   3.747 0.000183 ***
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
```

```
## Residual standard error: 8.097 on 2955 degrees of freedom
```

```
## Multiple R-squared:  0.8233, Adjusted R-squared:  0.8225
```

```
## F-statistic: 1059 on 13 and 2955 DF, p-value: < 2.2e-16
```

```
reg3 <- lm(y_trans ~ poly(tim, 2, raw = FALSE) + month + dow)
summary(reg3)
```

```
##
```

```
## Call:
```

```
## lm(formula = y_trans ~ poly(tim, 2, raw = FALSE) + month + dow)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max
## -37.086  -4.311   0.836   4.901  31.512
```

```
##
```

```
## Coefficients:
```

```
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)      42.3325    0.6270  67.516 < 2e-16 ***
## poly(tim, 2, raw = FALSE)1 642.5699    8.0310  80.011 < 2e-16 ***
## poly(tim, 2, raw = FALSE)2  48.1134    8.0835   5.952 2.96e-09 ***
## month02           0.6592    0.7418   0.889 0.374229
## month03           6.2757    0.7242   8.666 < 2e-16 ***
## month04          11.8192    0.7303  16.184 < 2e-16 ***
## month05          23.5691    0.7244  32.538 < 2e-16 ***
## month06          29.9529    0.7307  40.995 < 2e-16 ***
## month07          32.4546    0.7029  46.171 < 2e-16 ***
## month08          33.4604    0.7029  47.603 < 2e-16 ***
## month09          31.8053    0.7084  44.897 < 2e-16 ***
## month10          23.4927    0.7287  32.240 < 2e-16 ***
## month11          14.2872    0.7350  19.438 < 2e-16 ***
## month12           2.7472    0.7287   3.770 0.000166 ***
## dowMonday        -1.9523    0.5495  -3.553 0.000387 ***
## dowSaturday      -1.0741    0.5501  -1.953 0.050951 .
## dowSunday        -2.8167    0.5501  -5.120 3.25e-07 ***
## dowThursday       0.3757    0.5501   0.683 0.494652
## dowTuesday       -0.2132    0.5501  -0.388 0.698380
## dowWednesday      0.8004    0.5508   1.453 0.146250
```

```
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 8.014 on 2949 degrees of freedom  
## Multiple R-squared:  0.8273, Adjusted R-squared:  0.8262  
## F-statistic: 743.3 on 19 and 2949 DF,  p-value: < 2.2e-16
```