

Sta 325 Final Project

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```
library(readr)
library(dplyr)
library(tidyverse)
library(gridExtra)
library(mgcv)
library(patchwork)

# read data
flights <- read_csv("data/flights.csv")

# find unique airlines, destinations, and types of delays
unique(flights$OP_CARRIER)

## [1] "AA" "DL" "B6" "AS"

unique(flights$DEST)

## [1] "LAX" "SFO" "SJC" "SAN" "PSP" "SMF" "OAK" "LGB" "ONT" "BUR"

class(flights$CARRIER_DELAY)

## [1] "numeric"

# mutate delays and filter out NA arrival delays
flights <- flights %>%
  mutate(CARRIER_DELAY = case_when(CARRIER_DELAY > 0 ~ 1,
                                     TRUE ~ 0),
         WEATHER_DELAY = case_when(WEATHER_DELAY > 0 ~ 1,
                                     TRUE ~ 0),
         NAS_DELAY = case_when(NAS_DELAY > 0 ~ 1,
                                TRUE ~ 0),
         SECURITY_DELAY = case_when(SECURITY_DELAY > 0 ~ 1,
                                     TRUE ~ 0),
         LATE_AIRCRAFT_DELAY = case_when(LATE_AIRCRAFT_DELAY > 0 ~ 1,
                                           TRUE ~ 0)) %>%
  filter(!is.na(ARR_DELAY))

# glimpse data
flights

## # A tibble: 2,033 x 34
##   YEAR MONTH DAY_OF_MONTH DAY_OF_WEEK FL_DATE   OP_CARRIER TAIL_NUM
##   <dbl> <dbl>         <dbl>         <dbl> <date>     <chr>      <chr>
## 1  2020     1             1             3 2020-01-01 AA        N110AN
## 2  2020     1             2             4 2020-01-02 AA        N111ZM
```

```
## 3 2020 1 3 5 2020-01-03 AA N108NN
## 4 2020 1 4 6 2020-01-04 AA N102NN
## 5 2020 1 5 7 2020-01-05 AA N113AN
## 6 2020 1 6 1 2020-01-06 AA N103NN
## 7 2020 1 7 2 2020-01-07 AA N113AN
## 8 2020 1 8 3 2020-01-08 AA N106NN
## 9 2020 1 9 4 2020-01-09 AA N102NN
## 10 2020 1 10 5 2020-01-10 AA N117AN
## # ... with 2,023 more rows, and 27 more variables: OP_CARRIER_FL_NUM <dbl>,
## # ORIGIN <chr>, ORIGIN_CITY_NAME <chr>, DEST <chr>, DEST_CITY_NAME <chr>,
## # CRS_DEP_TIME <dbl>, DEP_TIME <dbl>, DEP_DELAY <dbl>, TAXI_OUT <dbl>,
## # WHEELS_OFF <dbl>, WHEELS_ON <dbl>, TAXI_IN <dbl>, CRS_ARR_TIME <dbl>,
## # ARR_TIME <dbl>, ARR_DELAY <dbl>, CANCELLED <dbl>, CANCELLATION_CODE <lgl>,
## # DIVERTED <dbl>, CRS_ELAPSED_TIME <dbl>, ACTUAL_ELAPSED_TIME <dbl>,
## # AIR_TIME <dbl>, DISTANCE <dbl>, CARRIER_DELAY <dbl>, WEATHER_DELAY <dbl>,
## # NAS_DELAY <dbl>, SECURITY_DELAY <dbl>, LATE_AIRCRAFT_DELAY <dbl>
```

INDIVIDUAL PREDICTORS

Taxi Histograms

```
# plot untransformed predictor taxi_in
pTAXI_IN <- ggplot(data = flights, aes(x = TAXI_IN)) +
  geom_histogram(binwidth = 5, fill = "#FFFF00", color = "#002D72", alpha = .7) +
  labs(x = "Time to Taxi In",
       y = "Frequency",
       title = "Histogram of TAXI_IN") +
  theme(plot.title = element_text(size = 10, hjust = 0.5),
        plot.subtitle = element_text(hjust = 0.5),
        axis.title.x.bottom = element_text(size = 8, face = "italic"),
        axis.title.y.left = element_text(size = 8))

# plot untransformed predictor taxi_out
pTAXI_OUT <- ggplot(data = flights, aes(x = TAXI_OUT)) +
  geom_histogram(binwidth = 5, fill = "#FFFF00", color = "#002D72", alpha = .7) +
  labs(x = "Time to Taxi Out",
       y = "Frequency",
       title = "Histogram of TAXI_OUT") +
  theme(plot.title = element_text(size = 10, hjust = 0.5),
        plot.subtitle = element_text(hjust = 0.5),
        axis.title.x.bottom = element_text(size = 8, face = "italic"),
        axis.title.y.left = element_text(size = 8))

# log transform taxi_in and taxi_out
flights$log_TAXI_OUT <- log(flights$TAXI_OUT)
flights$log_TAXI_IN <- log(flights$TAXI_IN)

# plot log transformed taxi_out
plog_TAXI_OUT <- ggplot(data = flights, aes(x = log_TAXI_OUT)) +
  geom_histogram(fill = "#FFFF00", color = "#002D72", alpha = .7) +
  labs(x = "Time to Taxi Out",
       y = "Frequency",
       title = "Histogram of log(TAXI_OUT)") +
```

```

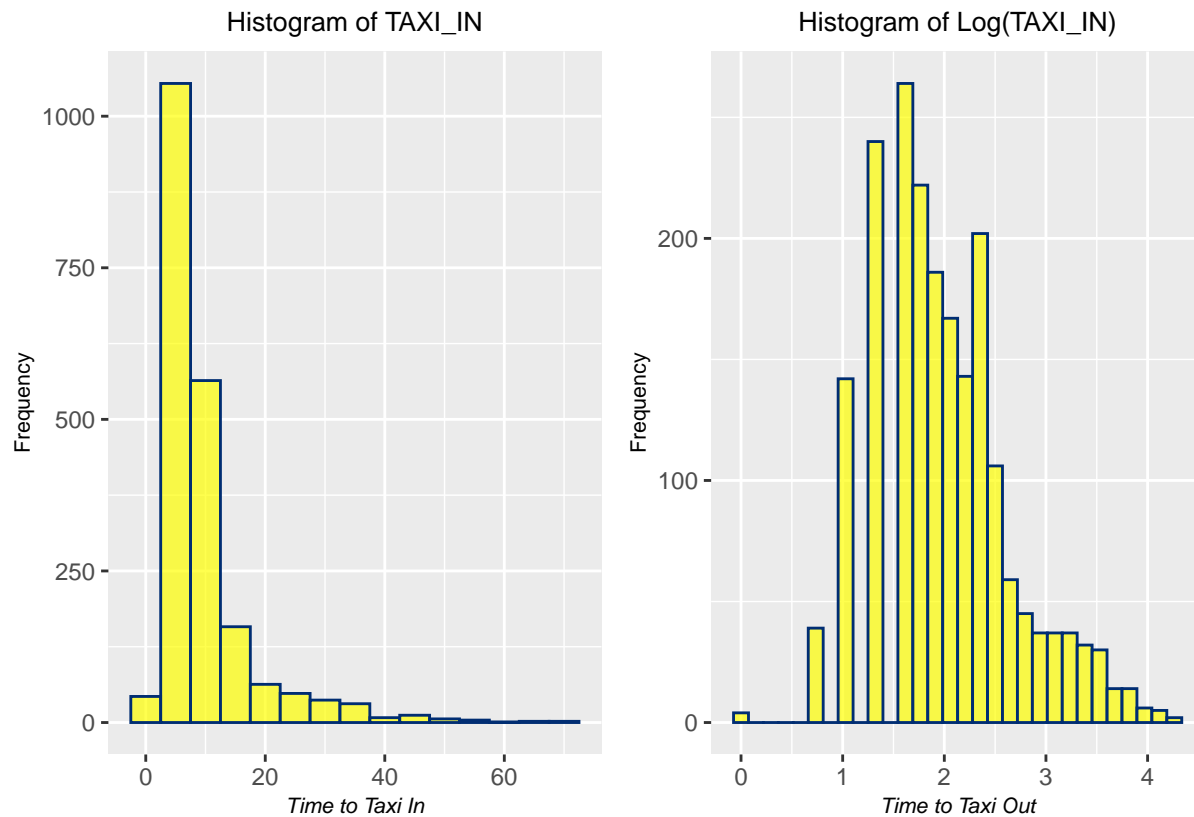
theme(plot.title = element_text(size = 10,hjust = 0.5),
      plot.subtitle = element_text(hjust = 0.5),
      axis.title.x.bottom = element_text(size = 8, face = "italic"),
      axis.title.y.left = element_text(size = 8))

# plot log transform taxi_in
plog_TAXI_IN <- ggplot(data = flights, aes(x = log_TAXI_IN)) +
  geom_histogram(fill = "#FFFF00", color = "#002D72", alpha = .7) +
  labs(x = "Time to Taxi Out",
       y = "Frequency",
       title = "Histogram of Log(TAXI_IN)") +
  theme(plot.title = element_text(size = 10,hjust = 0.5),
        plot.subtitle = element_text(hjust = 0.5),
        axis.title.x.bottom = element_text(size = 8, face = "italic"),
        axis.title.y.left = element_text(size = 8))

pTAXI_IN + plog_TAXI_IN

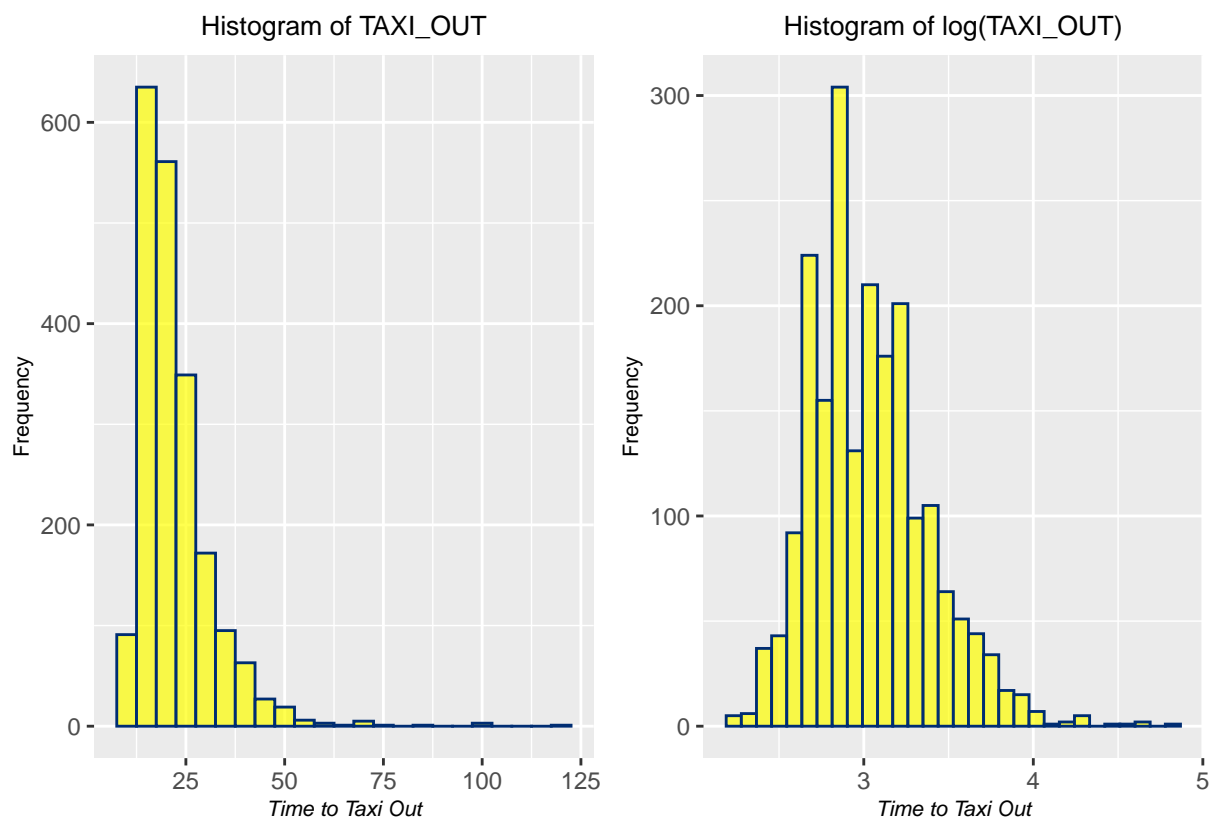
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.



```
pTAXI_OUT + plog_TAXI_OUT
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

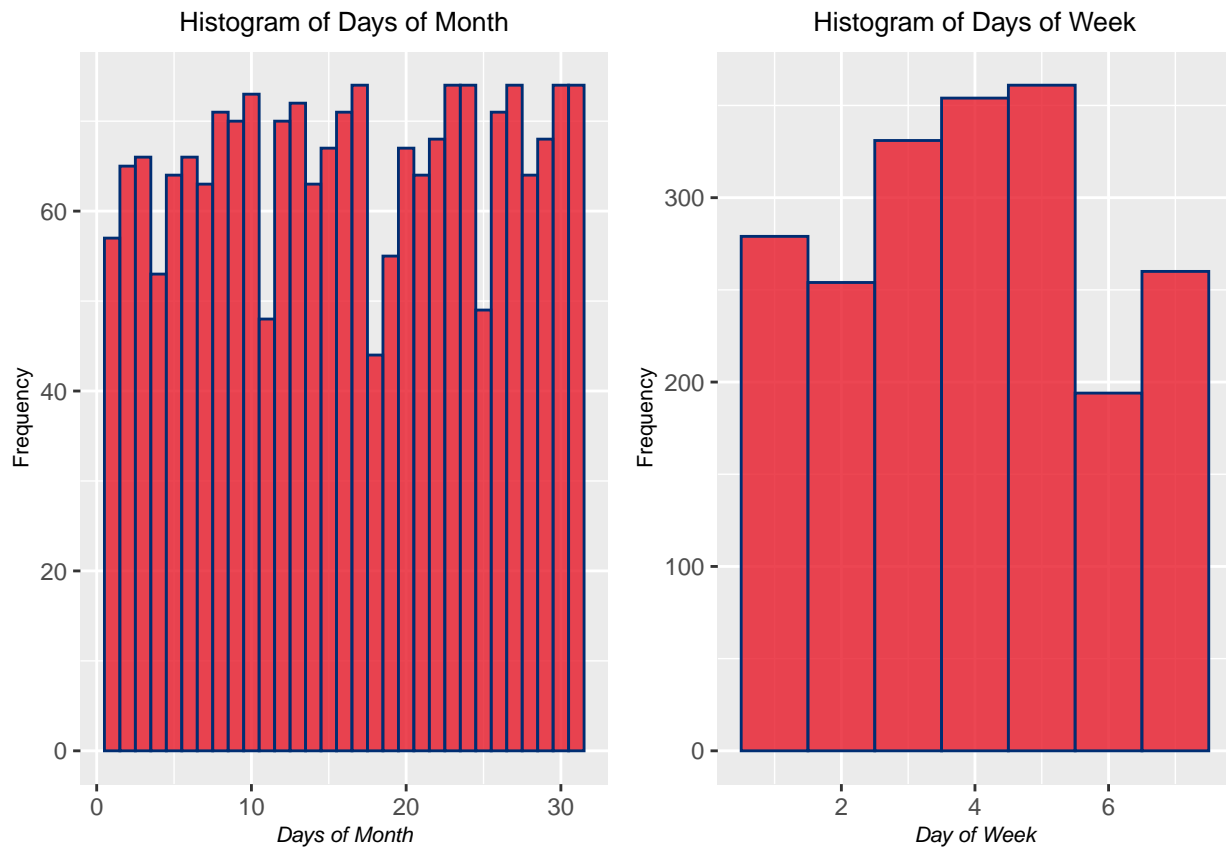


Days of Month and Week

```
# plot predictor DAYS_OF_MONTH
p02 <- ggplot(data = flights, aes(x = DAY_OF_MONTH)) +
  geom_histogram(binwidth = 1, fill = "#E81828", color = "#002D72", alpha = .8) +
  labs(x = "Days of Month",
       y = "Frequency",
       title = "Histogram of Days of Month") +
  theme(plot.title = element_text(size = 10, hjust = 0.5),
        plot.subtitle = element_text(hjust = 0.5),
        axis.title.x.bottom = element_text(size = 8, face = "italic"),
        axis.title.y.left = element_text(size = 8))

# plot predictor DAY_OF_WEEK
p03 <- ggplot(data = flights, aes(x = DAY_OF_WEEK)) +
  geom_histogram(binwidth = 1, fill = "#E81828", color = "#002D72", alpha = .8) +
  labs(x = "Day of Week",
       y = "Frequency",
       title = "Histogram of Days of Week") +
  theme(plot.title = element_text(size = 10, hjust = 0.5),
        plot.subtitle = element_text(hjust = 0.5),
        axis.title.x.bottom = element_text(size = 8, face = "italic"),
        axis.title.y.left = element_text(size = 8))

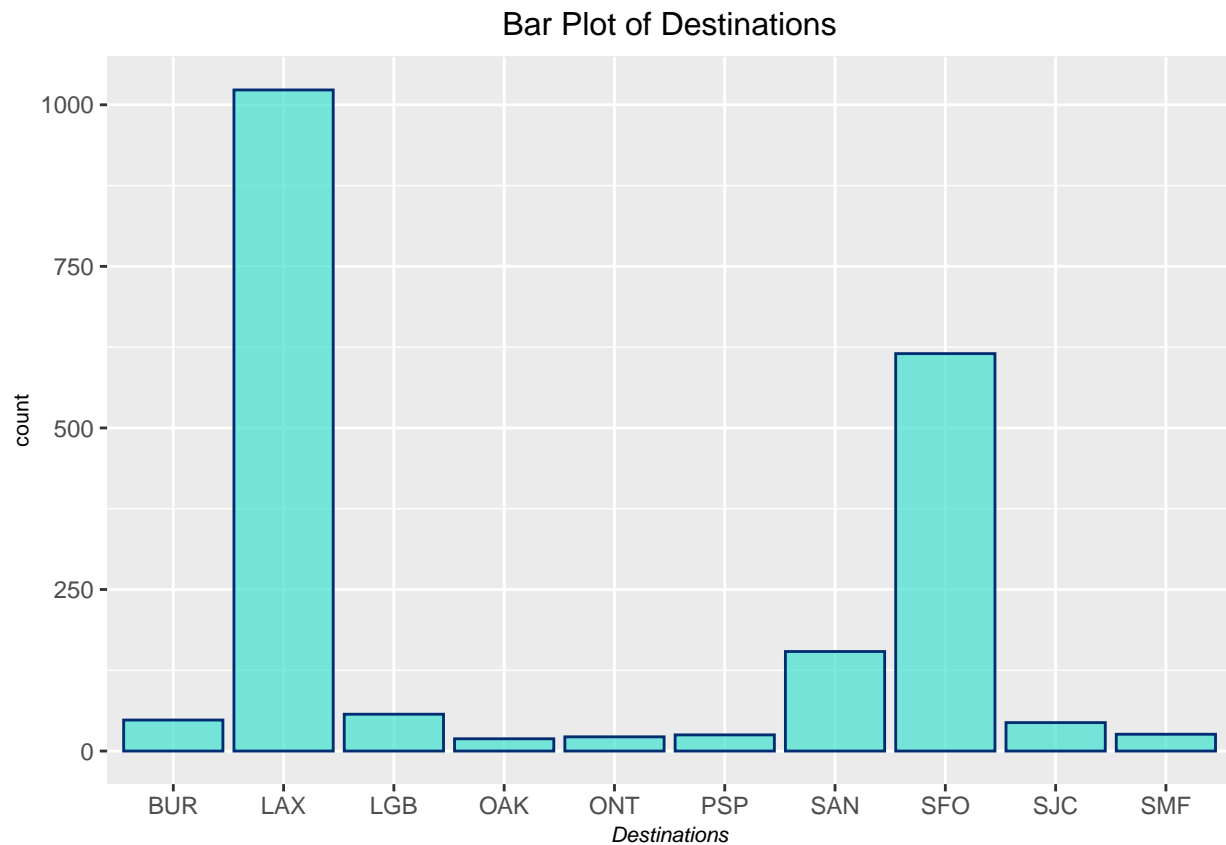
grid.arrange(p02, p03, nrow = 1)
```



Destination Locations

Origin is all JFK, but we could consider the different destination locations.

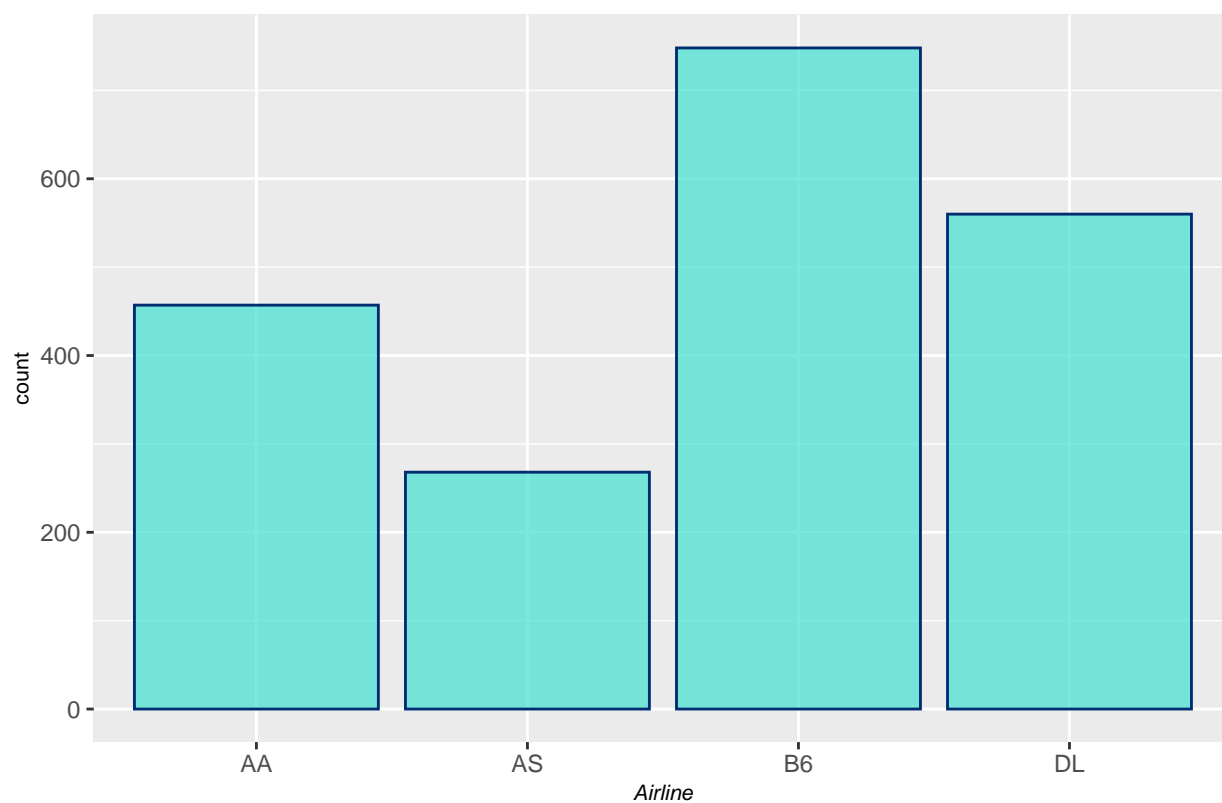
```
# plot destinations in CA
ggplot(data = flights, aes(x = DEST)) +
  geom_bar(fill = "#40E0D0", color = "#002D72", alpha = .7) +
  labs(x = "Destinations",
       title = "Bar Plot of Destinations") +
  theme(plot.title = element_text(size = 12, hjust = 0.5),
        plot.subtitle = element_text(hjust = 0.5),
        axis.title.x.bottom = element_text(size = 8, face = "italic"),
        axis.title.y.left = element_text(size = 8))
```



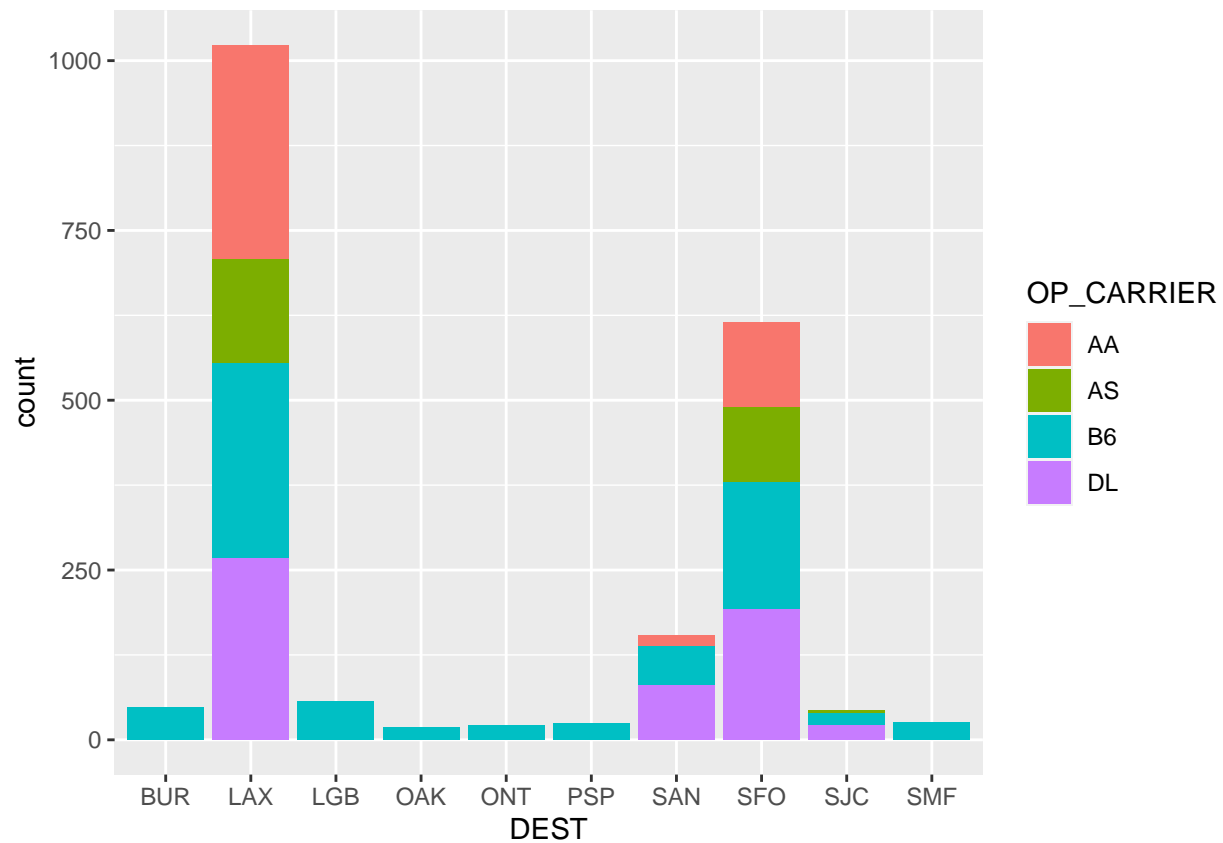
Airlines

```
# plot airline carriers
ggplot(data = flights, aes(x = OP_CARRIER)) +
  geom_bar(fill = "#40E0D0", color = "#002D72", alpha = .7) +
  labs(x = "Airline",
       title = "Bar Plot of Airlines") +
  theme(plot.title = element_text(size = 12, hjust = 0.5),
        plot.subtitle = element_text(hjust = 0.5),
        axis.title.x.bottom = element_text(size = 8, face = "italic"),
        axis.title.y.left = element_text(size = 8))
```

Bar Plot of Airlines

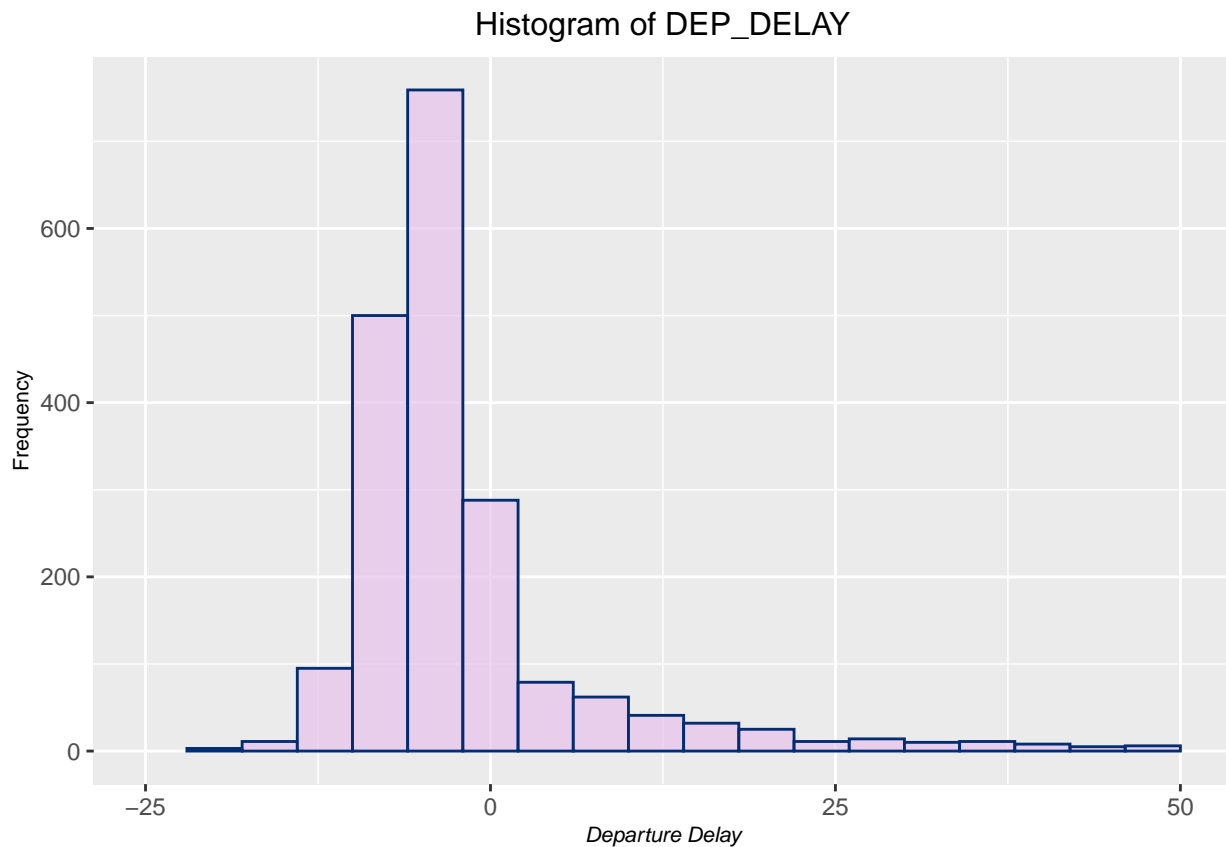


```
# plot airlines by destination  
ggplot(data = flights, aes(x = DEST, fill = OP_CARRIER)) +  
  geom_bar()
```



Depart Delay Histogram

```
# plot DEP_DELAY
ggplot(data = flights, aes(x = DEP_DELAY)) +
  geom_histogram(binwidth = 4, fill = "#e9c2ed", color = "#002D72", alpha = 0.7) +
  xlim(-25, 50) +
  labs(x = "Departure Delay",
       y = "Frequency",
       title = "Histogram of DEP_DELAY") +
  theme(plot.title = element_text(size = 12, hjust = 0.5),
        plot.subtitle = element_text(hjust = 0.5),
        axis.title.x.bottom = element_text(size = 8, face = "italic"),
        axis.title.y.left = element_text(size = 8))
```

```
# plot types of delays
p1 <- ggplot(data = flights, aes(x = CARRIER_DELAY)) +
  geom_bar(fill = "#E81828", color = "#002D72") +
  labs(title = "Carrier Delay")

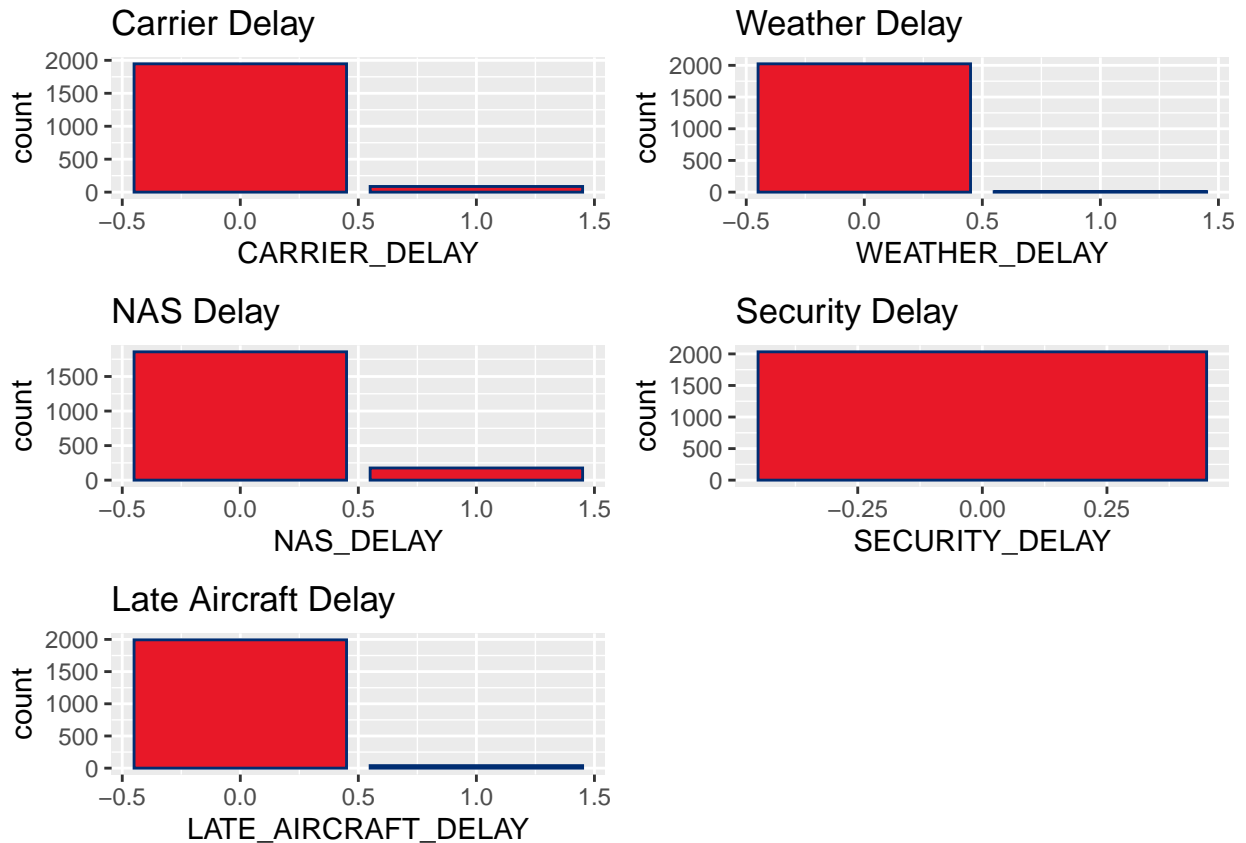
p2 <- ggplot(data = flights, aes(x = WEATHER_DELAY)) +
  geom_bar(fill = "#E81828", color = "#002D72") +
  labs(title = "Weather Delay")

p3 <- ggplot(data = flights, aes(x = NAS_DELAY)) +
  geom_bar(fill = "#E81828", color = "#002D72") +
  labs(title = "NAS Delay")

p4 <- ggplot(data = flights, aes(x = SECURITY_DELAY)) +
  geom_bar(fill = "#E81828", color = "#002D72") +
  labs(title = "Security Delay")

p5 <- ggplot(data = flights, aes(x = LATE_AIRCRAFT_DELAY)) +
  geom_bar(fill = "#E81828", color = "#002D72") +
  labs(title = "Late Aircraft Delay")

grid.arrange(p1,p2,p3,p4,p5, nrow = 3)
```



From this EDA of the categorical variables, we probably should not perform analysis with `SECURITY_DELAY` since all of them are classified as 0.

```
flights %>%
  count(WEATHER_DELAY)
```

```
## # A tibble: 2 x 2
##   WEATHER_DELAY    n
##         <dbl> <int>
## 1             0  2024
## 2             1     9
```

Furthermore, only 9 flights are classified with a weather delay, so it may not be good for our model to include this as a variable for right now.

Overall, the categorical delay predictors I would think we could use are: Carrier Delay, NAS Delay, and Late Aircraft Delay

RESPONSE VARIABLE: ARRIVAL DELAY TIME

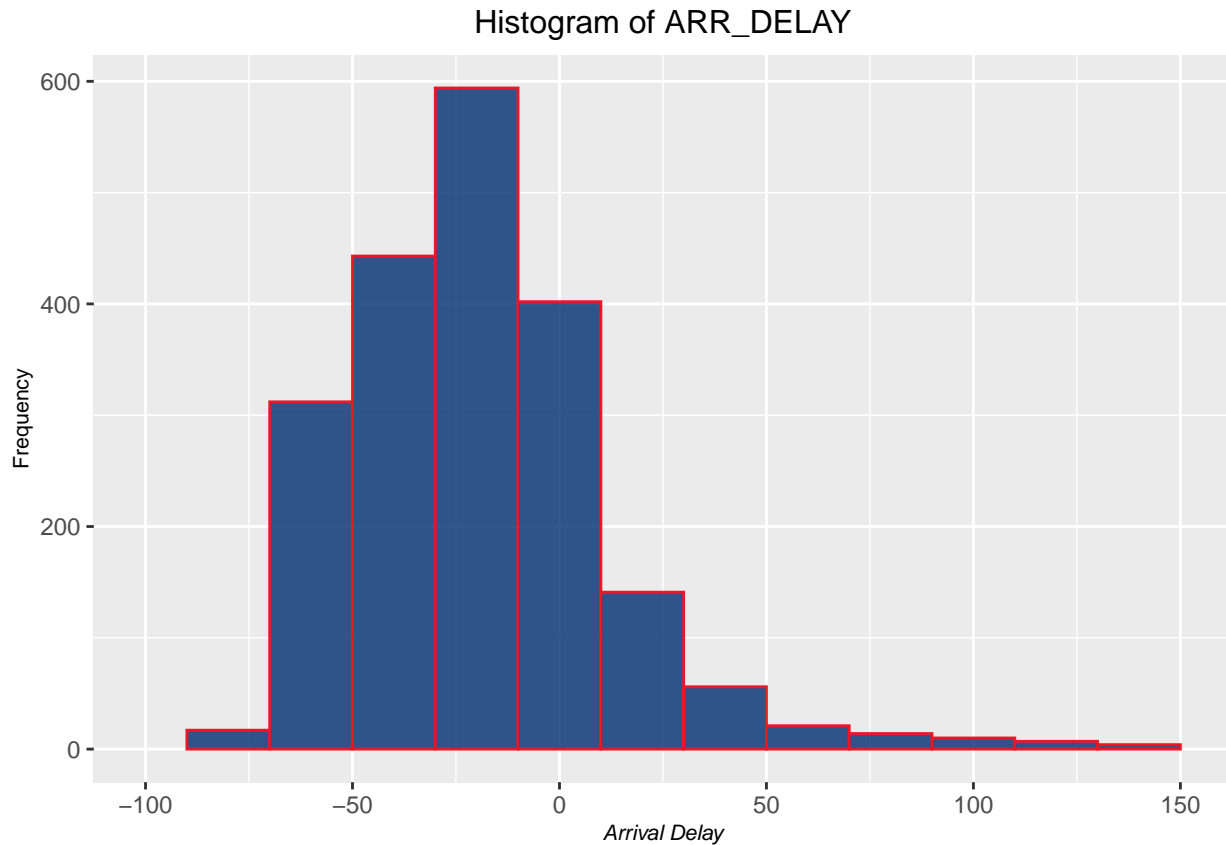
I just made it a different color so that when I scroll up to look at distributions I can easily tell the response from predictors (definitely can change at the end).

```
# plot ARR_DELAY
ggplot(data = flights, aes(x = ARR_DELAY)) +
  geom_histogram(binwidth = 20, fill = "#002D72", color = "#E81828", alpha = 0.8) +
  xlim(-100, 150) +
  labs(x = "Arrival Delay",
       y = "Frequency",
```

```

title = "Histogram of ARR_DELAY") +
theme(plot.title = element_text(size = 12,hjust = 0.5),
      plot.subtitle = element_text(hjust = 0.5),
      axis.title.x.bottom = element_text(size = 8, face = "italic"),
      axis.title.y.left = element_text(size = 8))

```



```

# 2-parameter BC transformation
## can apply to GAM

```

PREDICTORS VS RESPONSE

ARR_DELAY and TAXI_IN / TAXI_OUT

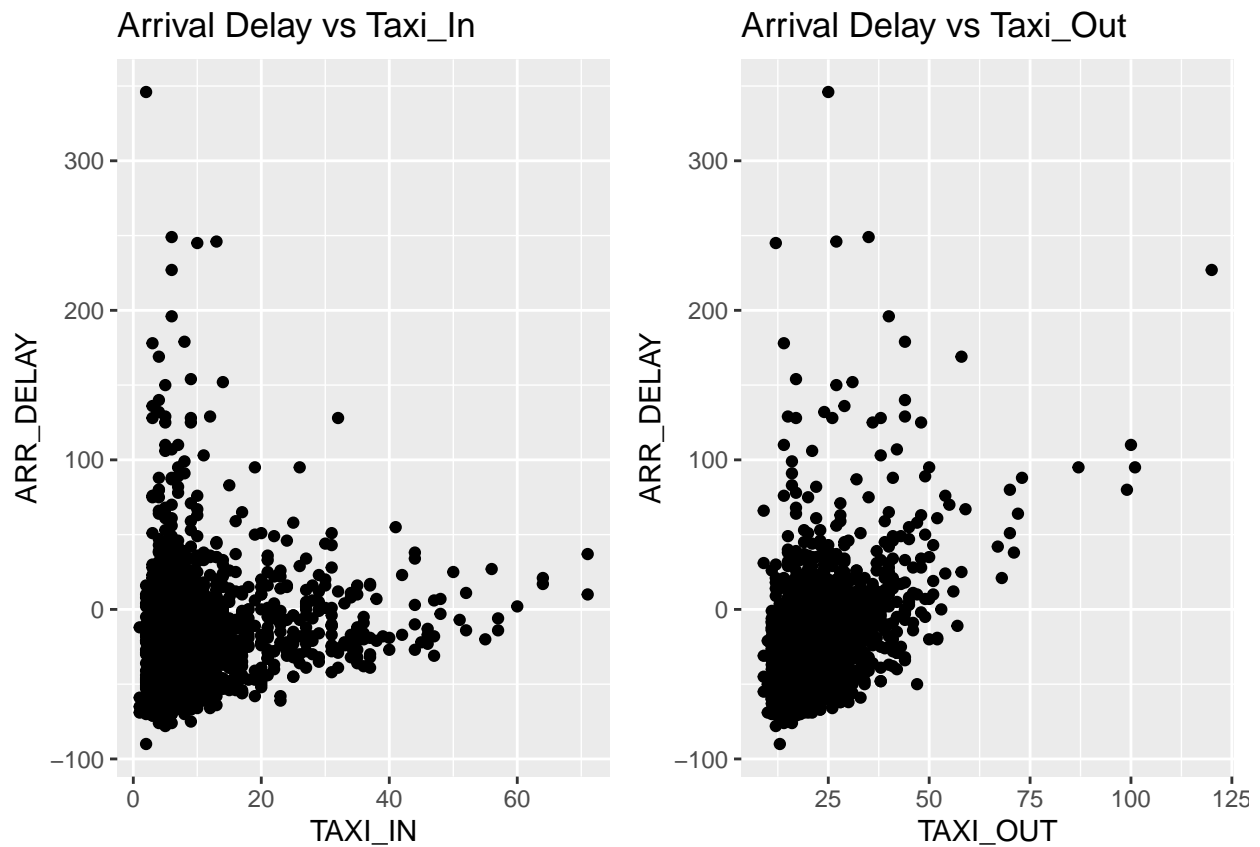
```

p6 <- ggplot(data = flights, aes(y = ARR_DELAY, x = TAXI_IN)) +
  geom_point() +
  labs(title = "Arrival Delay vs Taxi_In")

p7 <- ggplot(data = flights, aes(y = ARR_DELAY, x = TAXI_OUT)) +
  geom_point() +
  labs(title = "Arrival Delay vs Taxi_Out")

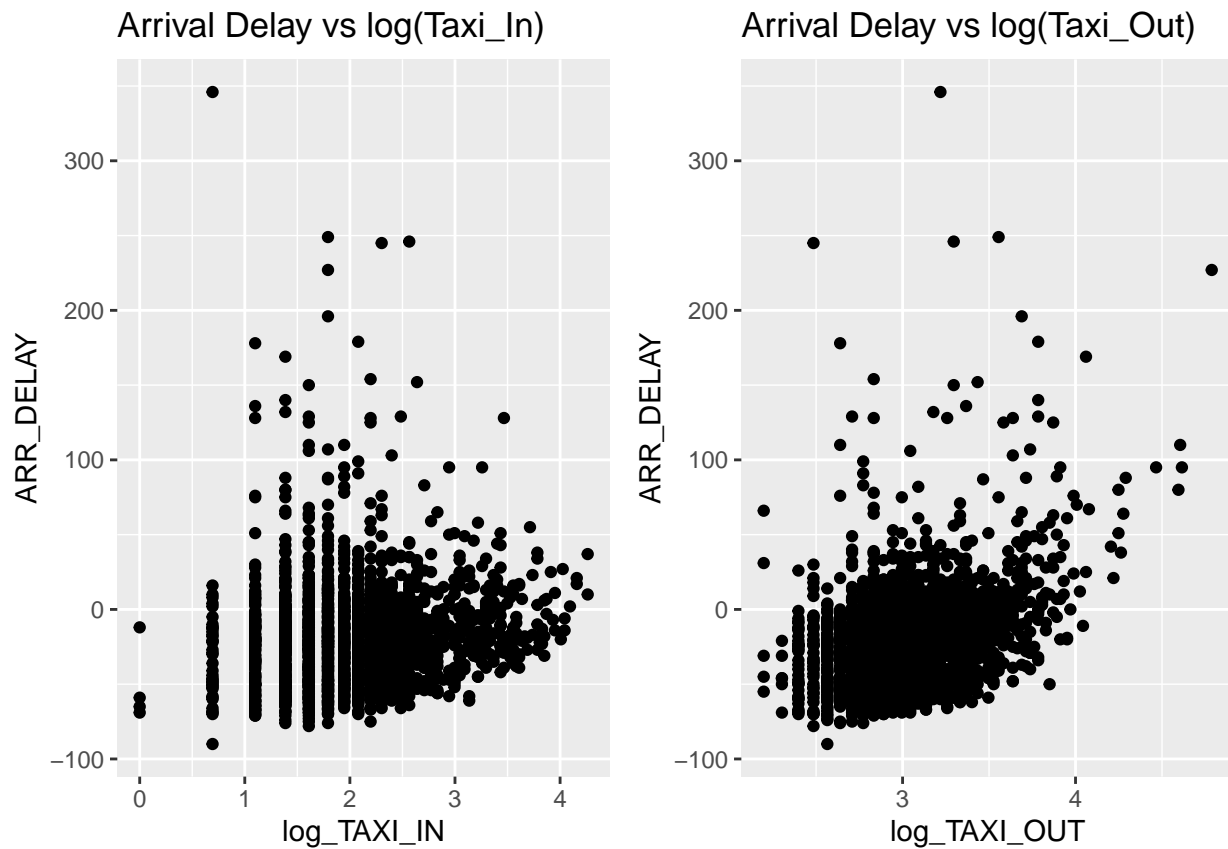
grid.arrange(p6,p7, nrow = 1)

```



```
plog6 <- ggplot(data = flights, aes(y = ARR_DELAY, x = log_TAXI_IN)) +
  geom_point() +
  labs(title = "Arrival Delay vs log(Taxi_In)")

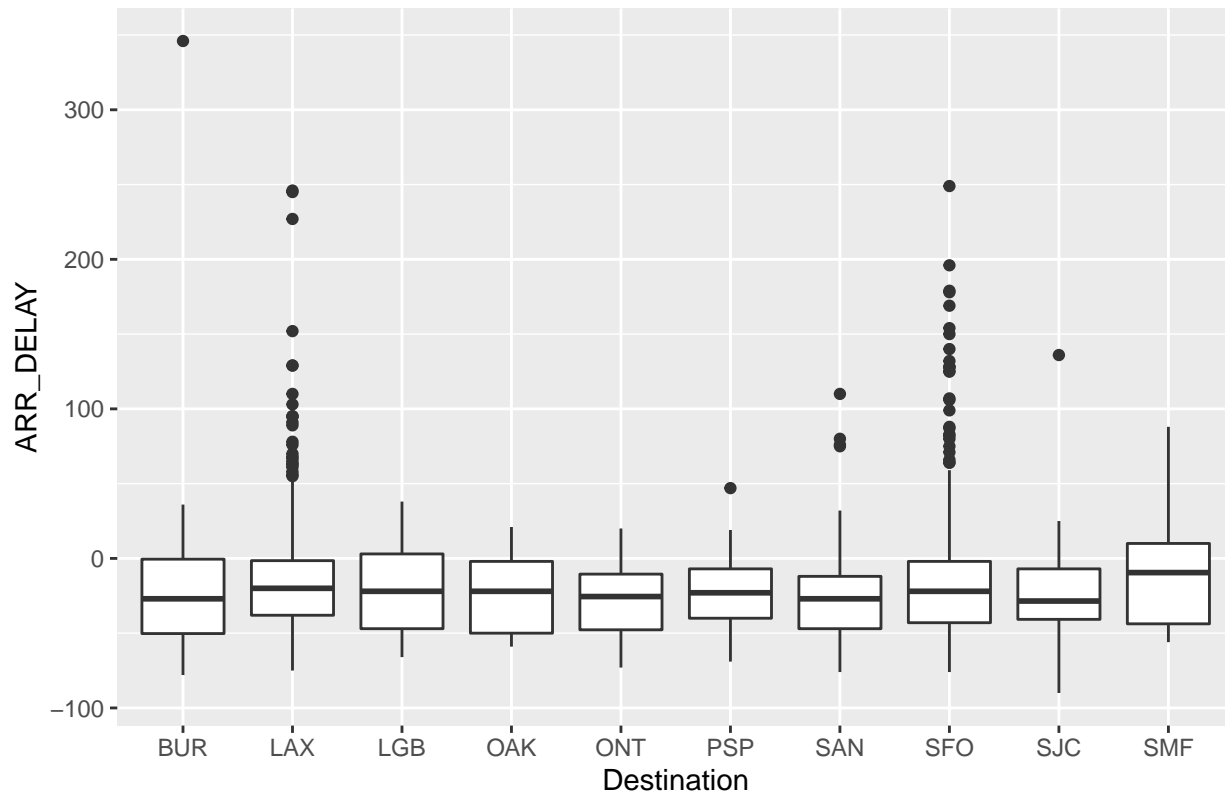
plog7 <- ggplot(data = flights, aes(y = ARR_DELAY, x = log_TAXI_OUT)) +
  geom_point() +
  labs(title = "Arrival Delay vs log(Taxi_Out)")
grid.arrange(plog6, plog7, nrow = 1)
```



These plots above suggest that we may want to transform the variables at some point.

```
ggplot(data = flights, aes(y = ARR_DELAY, x = DEST)) +  
  geom_boxplot() +  
  labs(x = "Destination",  
       title = "Arrival Delay vs Destination")
```

Arrival Delay vs Destination

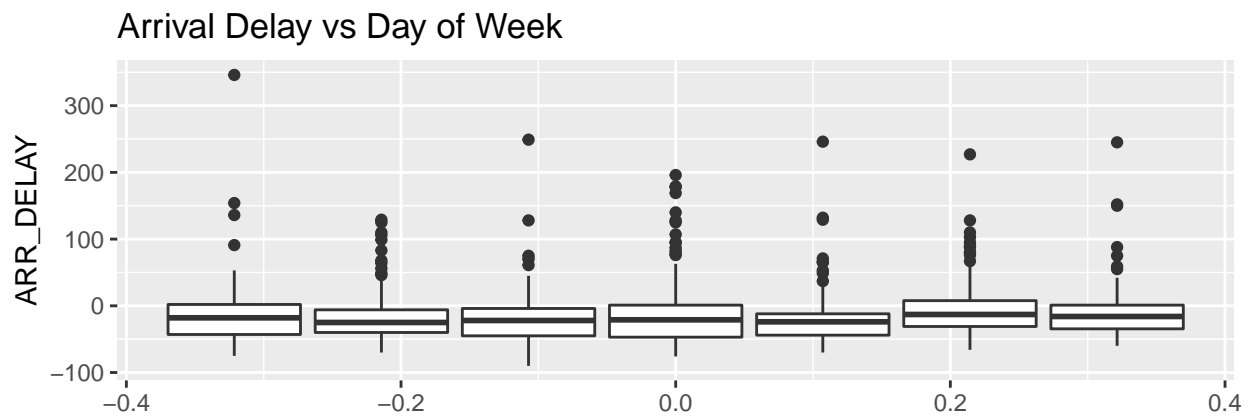
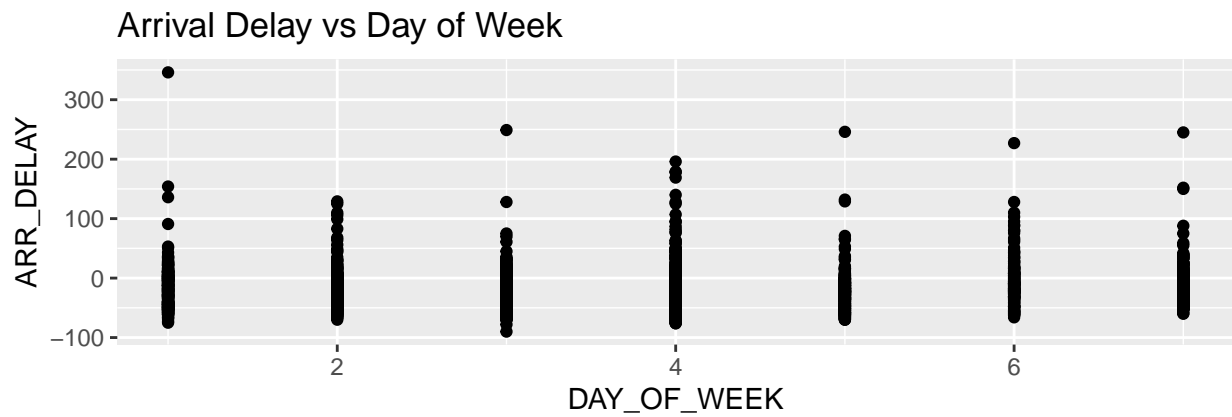


ARR_DELAY and DAY_OF_WEEK

```
p8 <- ggplot(data = flights, aes(y = ARR_DELAY, x = DAY_OF_WEEK)) +
  geom_point() +
  labs(title = "Arrival Delay vs Day of Week")

p9 <- ggplot(data = flights, aes(y = ARR_DELAY, group = DAY_OF_WEEK)) +
  geom_boxplot() +
  labs(title = "Arrival Delay vs Day of Week")

grid.arrange(p8,p9, nrow = 2)
```

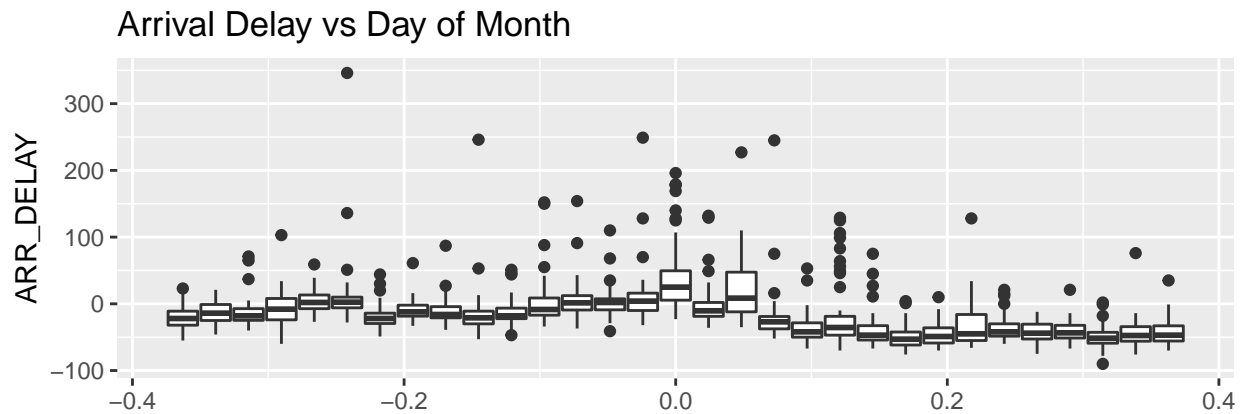
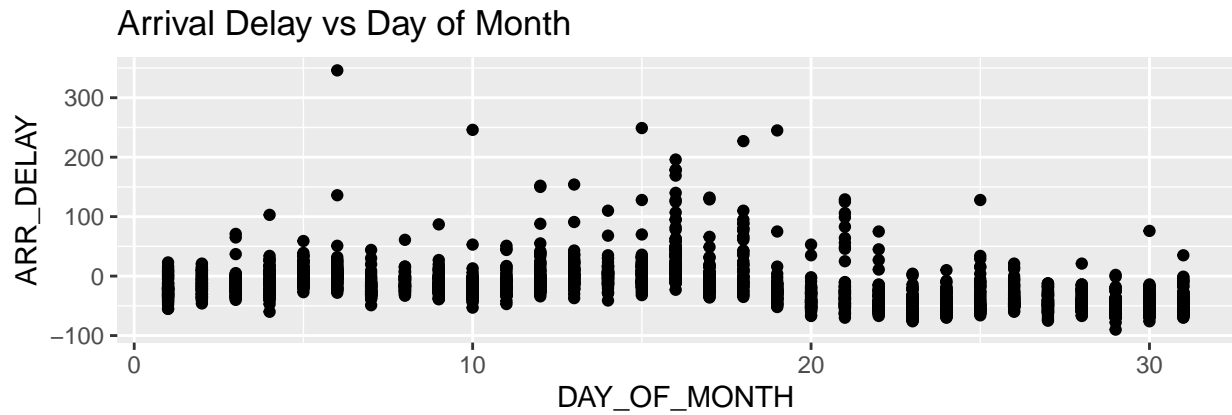


ARR_DELAY and DAY_OF_MONTH

```
p10 <- ggplot(data = flights, aes(y = ARR_DELAY, x = DAY_OF_MONTH)) +
  geom_point() +
  labs(title = "Arrival Delay vs Day of Month")

p11 <- ggplot(data = flights, aes(y = ARR_DELAY, group = DAY_OF_MONTH)) +
  geom_boxplot() +
  labs(title = "Arrival Delay vs Day of Month")

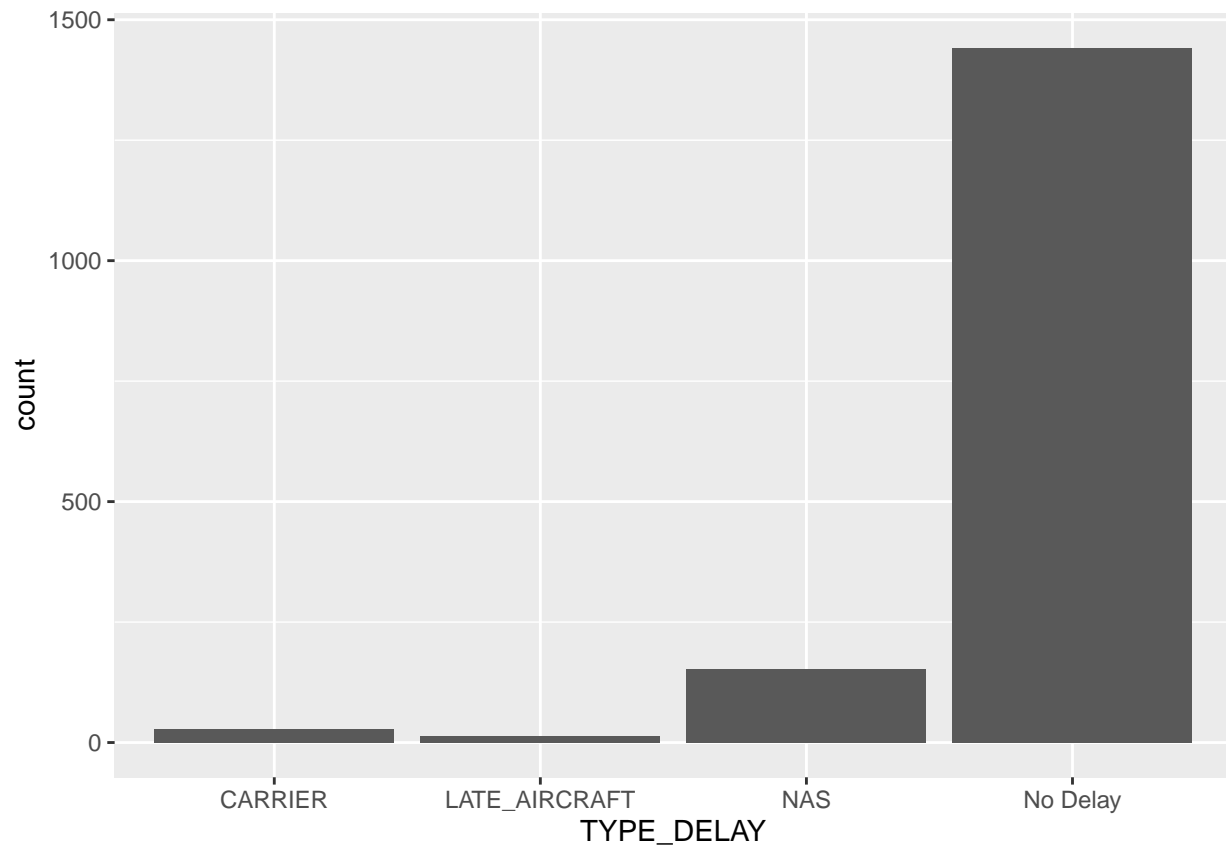
grid.arrange(p10, p11, nrow = 2)
```



Further Data Cleaning

```
# take only SFO/LAX since all 4 carriers fly there
flights <- flights %>%
  filter(DEST == "SFO" | DEST == "LAX") %>%
  mutate(TYPE_DELAY = case_when(
    NAS_DELAY == 1 ~ "NAS",
    CARRIER_DELAY == 1 ~ "CARRIER",
    LATE_AIRCRAFT_DELAY == 1 ~ "LATE_AIRCRAFT",
    TRUE ~ "No Delay"))

ggplot(data = flights, aes(x = TYPE_DELAY)) +
  geom_bar()
```

```
unique(flights$TYPE_DELAY)
```

```
## [1] "No Delay"      "NAS"           "LATE_AIRCRAFT" "CARRIER"
```

SPLITTING DATA

```
set.seed(1234)
flights <- flights %>%
  mutate(id = row_number())
train <- flights %>%
  sample_frac(0.8)
test <- anti_join(flights, train, by = "id")
```

LINEAR MODELS

Variables that I think we could explore: department delay time, days of month, days of week, taxi-in, taxi-out, destination, Carrier Delay, NAS Delay, and Late Aircraft Delay.

Full Log-Transformed Model

```
lm.01 <- lm(ARR_DELAY ~ DEP_DELAY + DAY_OF_WEEK + OP_CARRIER + DEST + CRS_DEP_TIME + CRS_ARR_TIME + log(
#plot(lm.01)
#summary(lm.01)
```

```

library(MASS)

##
## Attaching package: 'MASS'

## The following object is masked from 'package:patchwork':
##
##     area

## The following object is masked from 'package:dplyr':
##
##     select

step_model <- stepAIC(lm.01, direction = "backward", trace = FALSE)
#summary(step_model)

lm.02 <- lm(ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_OUT + log_TAXI_IN + TY
#summary(lm.02)
#anova(step_model, lm.02)

lm.03 <- lm(ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_OUT + log_TAXI_IN + TY
#anova(lm.02, lm.03)

log_linear_model <- lm(ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_OUT + log_T
anova(lm.03, log_linear_model)

## Analysis of Variance Table
##
## Model 1: ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_OUT +
##     log_TAXI_IN + TYPE_DELAY + OP_CARRIER:DEST + DEST:log_TAXI_IN
## Model 2: ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_OUT +
##     log_TAXI_IN + TYPE_DELAY + OP_CARRIER:DEST + DEST:log_TAXI_IN +
##     log_TAXI_OUT:DEP_DELAY
##   Res.Df    RSS Df Sum of Sq    F Pr(>F)
## 1     1294 427667
## 2     1293 425449   1      2218 6.7408 0.00953 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

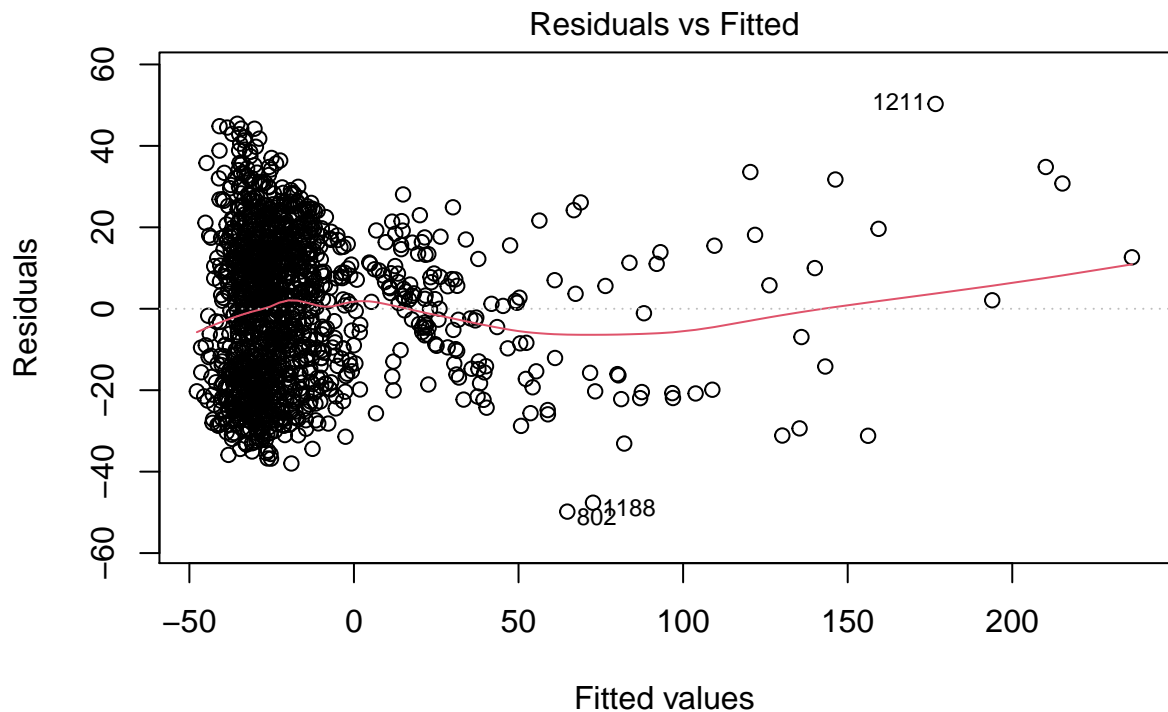
summary(log_linear_model)

##
## Call:
## lm(formula = ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME +
##     log_TAXI_OUT + log_TAXI_IN + TYPE_DELAY + OP_CARRIER:DEST +
##     DEST:log_TAXI_IN + log_TAXI_OUT:DEP_DELAY, data = train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -49.817 -15.330   1.198  13.897  50.301
##
## Coefficients:

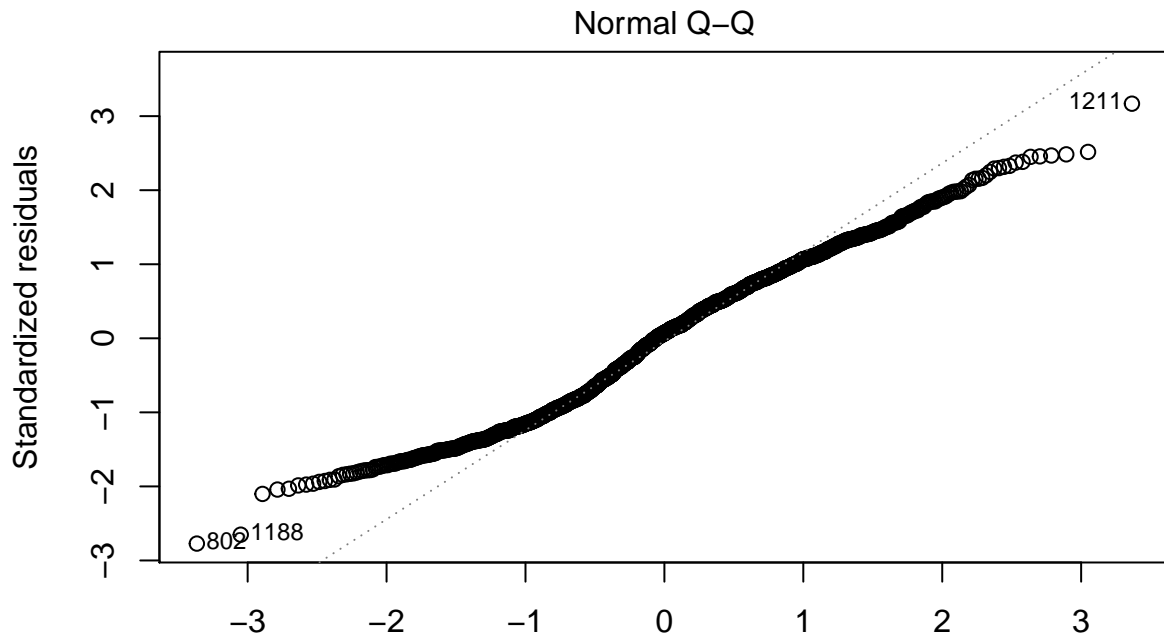
```

```
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -81.901399   7.262197 -11.278 < 2e-16 ***
## DEP_DELAY       0.524207   0.141256   3.711 0.000215 ***
## OP_CARRIERAS  -4.458249   2.087733  -2.135 0.032912 *
## OP_CARRIERB6   5.045463   1.669832   3.022 0.002564 **
## OP_CARRIERDL  -1.493672   1.717998  -0.869 0.384775
## DESTSFO         9.893184   4.366882   2.266 0.023647 *
## CRS_DEP_TIME   -0.004364   0.001070  -4.081 4.77e-05 ***
## log_TAXI_OUT    20.610508   1.617078  12.746 < 2e-16 ***
## log_TAXI_IN      8.433233   1.057392   7.976 3.32e-15 ***
## TYPE_DELAYLATE_AIRCRAFT -3.973566   6.537317  -0.608 0.543408
## TYPE_DELAYNAS   24.019795   4.598524   5.223 2.05e-07 ***
## TYPE_DELAYNo Delay -15.676745   4.540377  -3.453 0.000573 ***
## OP_CARRIERAS:DESTSFO  6.630276   3.374581   1.965 0.049655 *
## OP_CARRIERB6:DESTSFO -4.199151   2.858830  -1.469 0.142121
## OP_CARRIERDL:DESTSFO -1.424895   2.900122  -0.491 0.623282
## DESTSFO:log_TAXI_IN  -5.261163   1.951509  -2.696 0.007110 **
## DEP_DELAY:log_TAXI_OUT  0.113332   0.043651   2.596 0.009530 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 18.14 on 1293 degrees of freedom
## Multiple R-squared:  0.7376, Adjusted R-squared:  0.7344
## F-statistic: 227.2 on 16 and 1293 DF,  p-value: < 2.2e-16
```

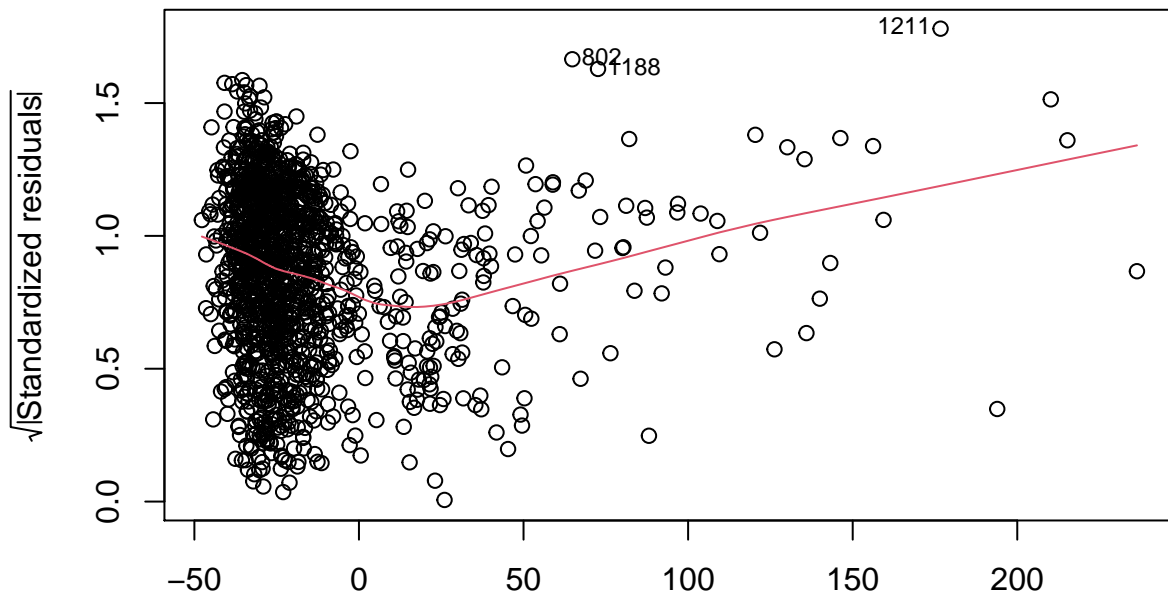
```
plot(log_linear_model)
```



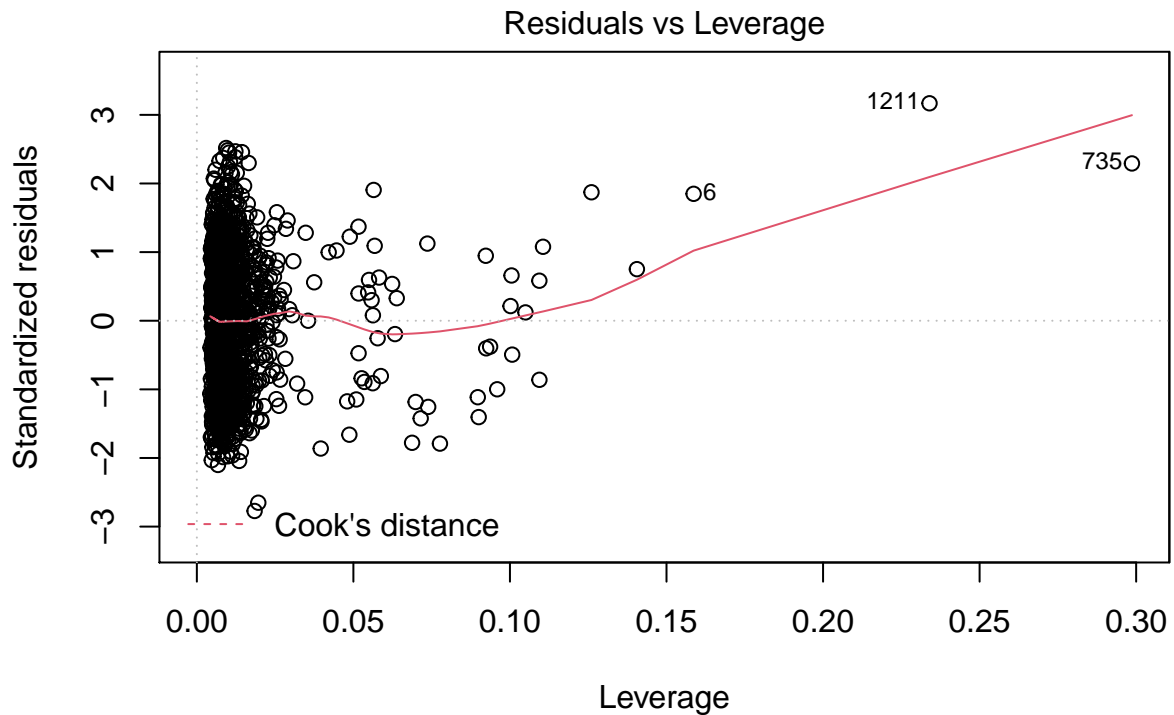
(ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_



(ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_
Scale-Location



(ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_



(ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_

```
## SIGNIFICANT INTERACTIONS
#OP_CARRIER:DEST
#DEST:log_TAXI_IN
#CRS_DEP_TIME:DEST (***** makes zero intuitive sense - might not wanna do this)
#CRS_ARR_TIME:log_TAXI_IN
#log_TAXI_OUT:DEP_DELAY

#log_TAXI_OUT:CRS_DEP_TIME (verrrrrrry close to 0.05)
# library(broom)
# log_linear_preds <- predict(log_linear_model, test)
# log_linear_MSE <- sum((log_linear_preds-test$ARR_DELAY)^2, na.rm=T)/328
# log_linear_MSE
```

Plain Linear Model

```
full_model <- lm(ARR_DELAY ~
  DEP_DELAY +
  DAY_OF_WEEK +
  OP_CARRIER +
  DEST +
  CRS_DEP_TIME +
  CRS_ARR_TIME +
  TAXI_OUT +
  TAXI_IN +
  TYPE_DELAY, train)

# summary(full_model)
# full_model_preds <- predict(full_model, test)
# linear_MSE <- sum((full_model_preds-test$ARR_DELAY)^2, na.rm=T)/328
```

```

# linear_MSE

library(MASS)
step_model <- stepAIC(full_model, trace = FALSE)
#summary(step_model)
plain_linear_model <- step_model

# interaction1 <- lm(ARR_DELAY ~
#                   DEP_DELAY +
#                   OP_CARRIER +
#                   DEST +
#                   CRS_DEP_TIME +
#                   CRS_ARR_TIME +
#                   TAXI_OUT +
#                   TAXI_IN +
#                   TYPE_DELAY,
#                   data = train)

## op_carrier and crs_dep_time almost significant

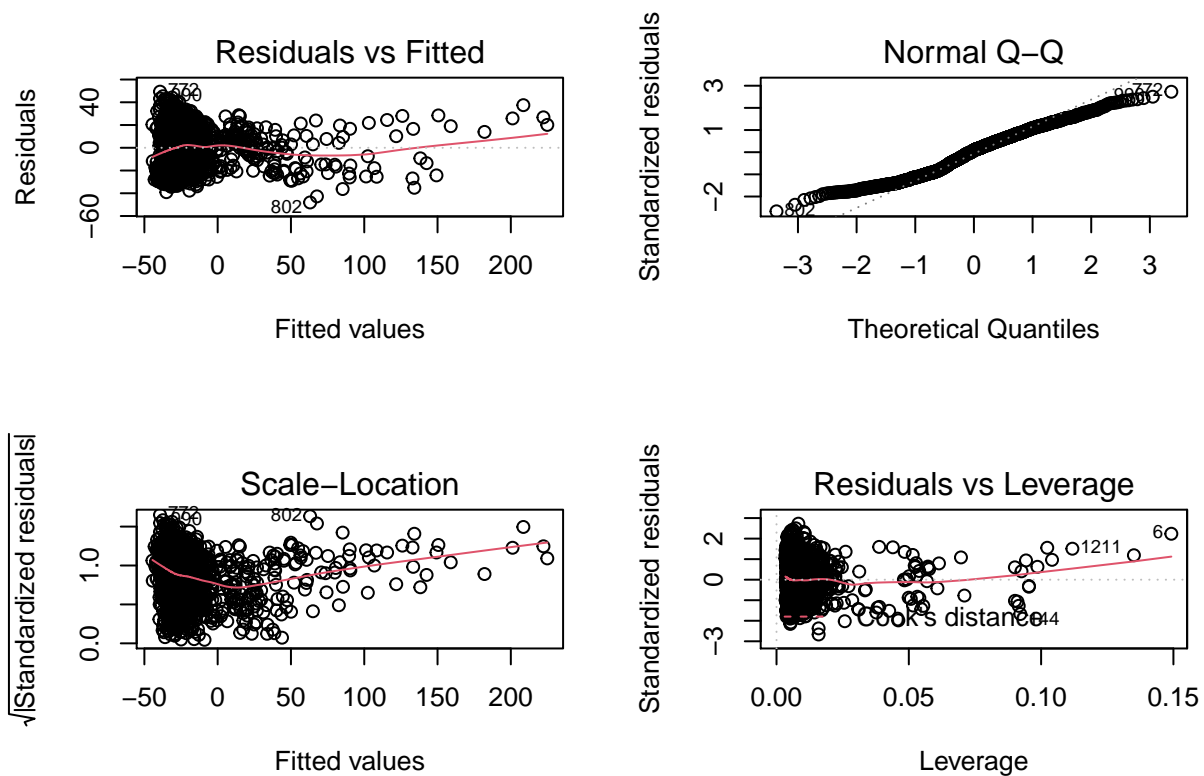
#anova(step_model, interaction1)

plain_linear_model

##
## Call:
## lm(formula = ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME +
##    CRS_ARR_TIME + TAXI_OUT + TAXI_IN + TYPE_DELAY, data = train)
##
## Coefficients:
##          (Intercept)          DEP_DELAY          OP_CARRIERAS
##          -24.104903           0.873248          -1.569862
##          OP_CARRIERB6          OP_CARRIERDL          DESTSF0
##           1.918141          -2.303676          -1.832585
##          CRS_DEP_TIME          CRS_ARR_TIME          TAXI_OUT
##          -0.004231          -0.001525           0.866771
##           TAXI_IN  TYPE_DELAYLATE_AIRCRAFT  TYPE_DELAYNAS
##           0.469992          -2.223475           25.087061
##          TYPE_DELAYNo Delay
##          -13.604813

par(mfrow = c(2,2))
plot(plain_linear_model)

```



Adjusted Model No Log Transforms

```
train$adj_ARR_DELAY = train$ARR_DELAY + 75
adj_linear_model <- lm(adj_ARR_DELAY ~
  DEP_DELAY +
  OP_CARRIER +
  DEST +
  CRS_DEP_TIME +
  CRS_ARR_TIME +
  TAXI_OUT +
  TAXI_IN +
  TYPE_DELAY,
  data = train)

## op_carrier and crs_dep_time almost significant interaction
```

Box-Cox

```
library(EnvStats)

##
## Attaching package: 'EnvStats'
## The following object is masked from 'package:MASS':
##
##   boxcox
## The following objects are masked from 'package:stats':
##
```

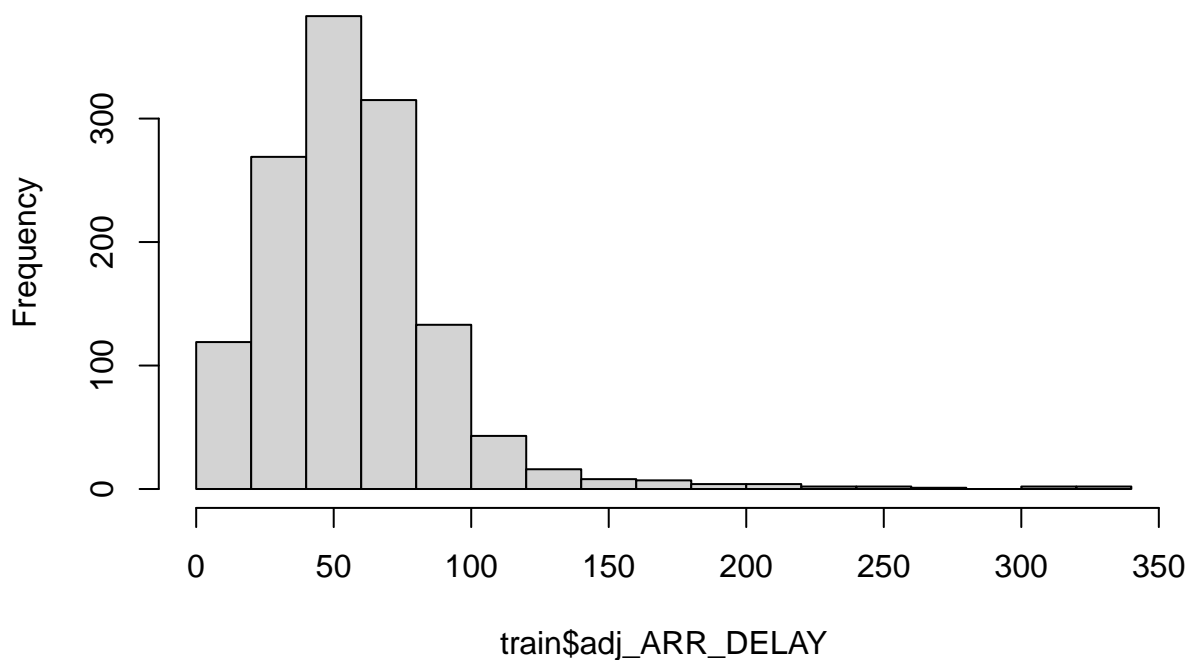
```
##      predict, predict.lm
## The following object is masked from 'package:base':
##
##      print.default
bc_model <- boxcox(adj_linear_model, optimize = TRUE)
bc_lambda <- bc_model$lambda
bc_lambda

## [1] 0.6981479
#plot(bc_model)

library(dplyr)
train <- train %>%
  mutate(bc_adj_ARR_DELAY = ((adj_ARR_DELAY^bc_lambda) - 1)/bc_lambda)

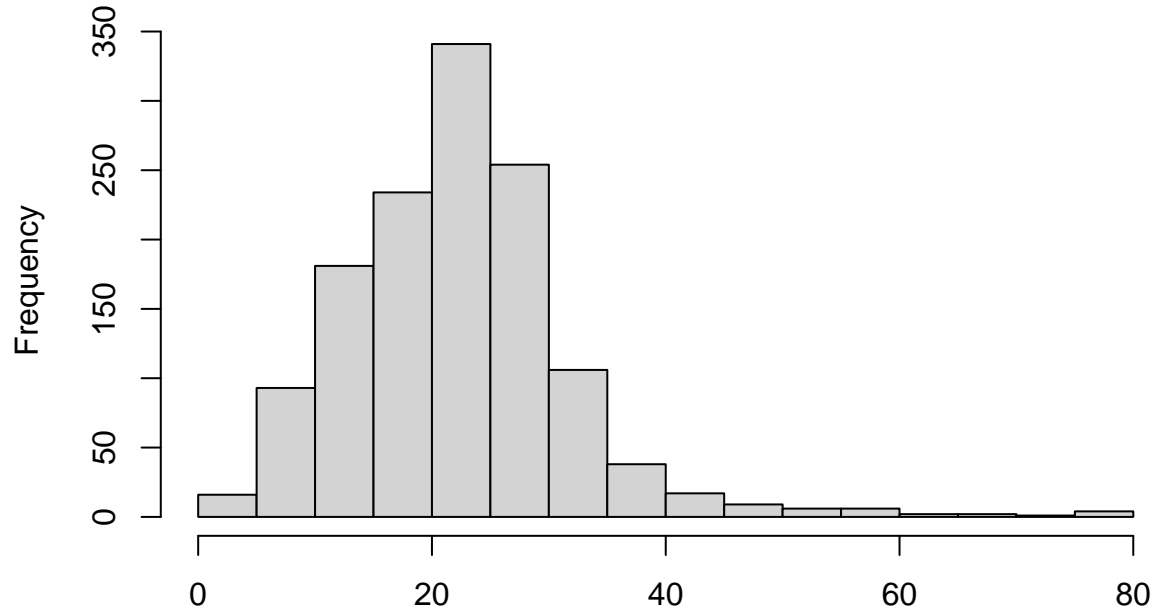
hist(train$adj_ARR_DELAY)
```

Histogram of train\$adj_ARR_DELAY



```
hist(train$bc_adj_ARR_DELAY)
```


Histogram of train\$bc_adj_ARR_DELAY

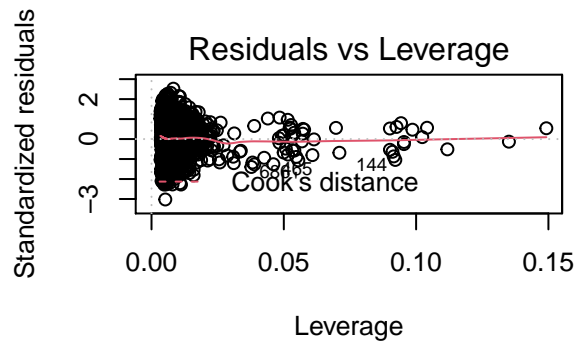
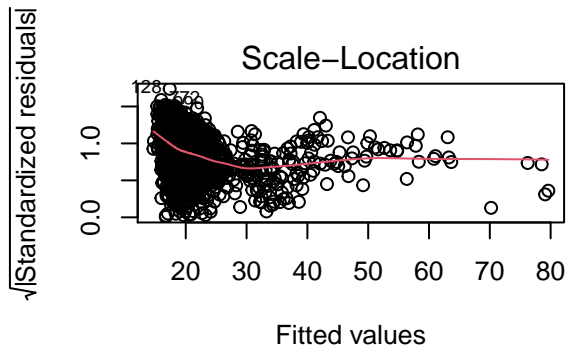
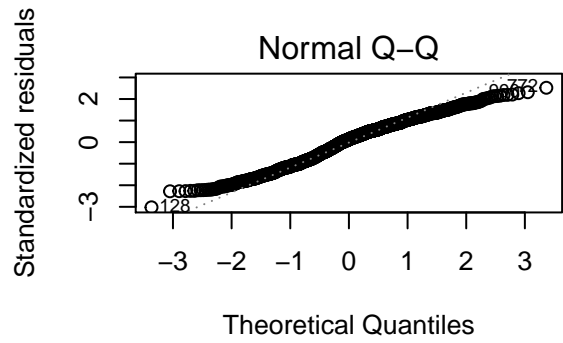
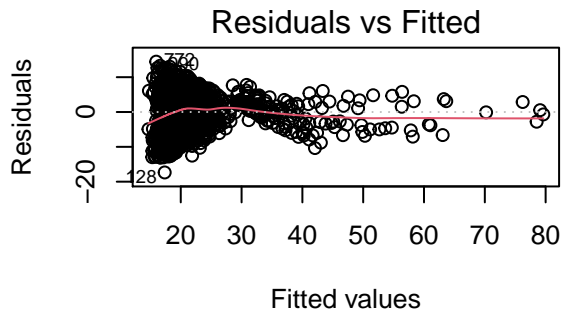


train\$bc_adj_ARR_DELAY

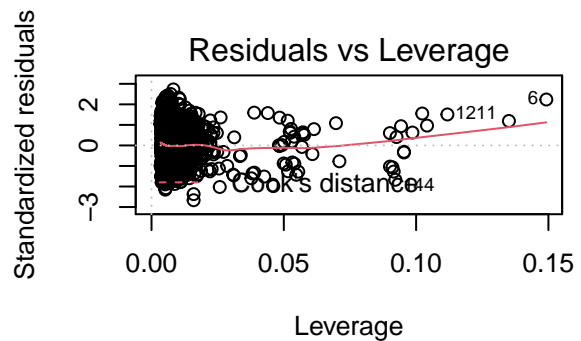
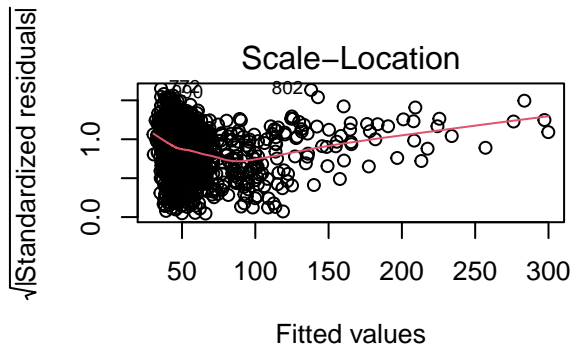
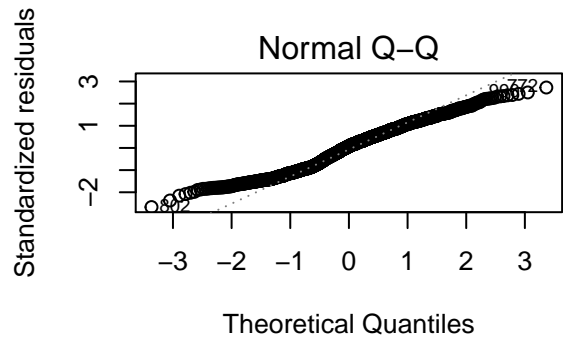
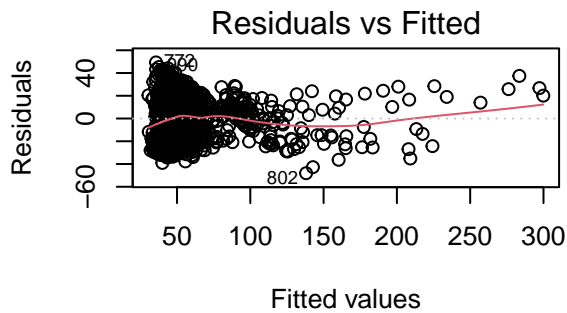
Box-

Cox Transformed Linear Model (No Transformed Predictors)

```
bc_adj_linear_model <- lm(bc_adj_ARR_DELAY ~
  DEP_DELAY +
  OP_CARRIER +
  DEST +
  CRS_DEP_TIME +
  CRS_ARR_TIME +
  TAXI_OUT +
  TAXI_IN +
  TYPE_DELAY,
  data = train)
par(mfrow = c(2,2))
plot(bc_adj_linear_model)
```



```
par(mfrow = c(2,2))
plot(adj_linear_model)
```

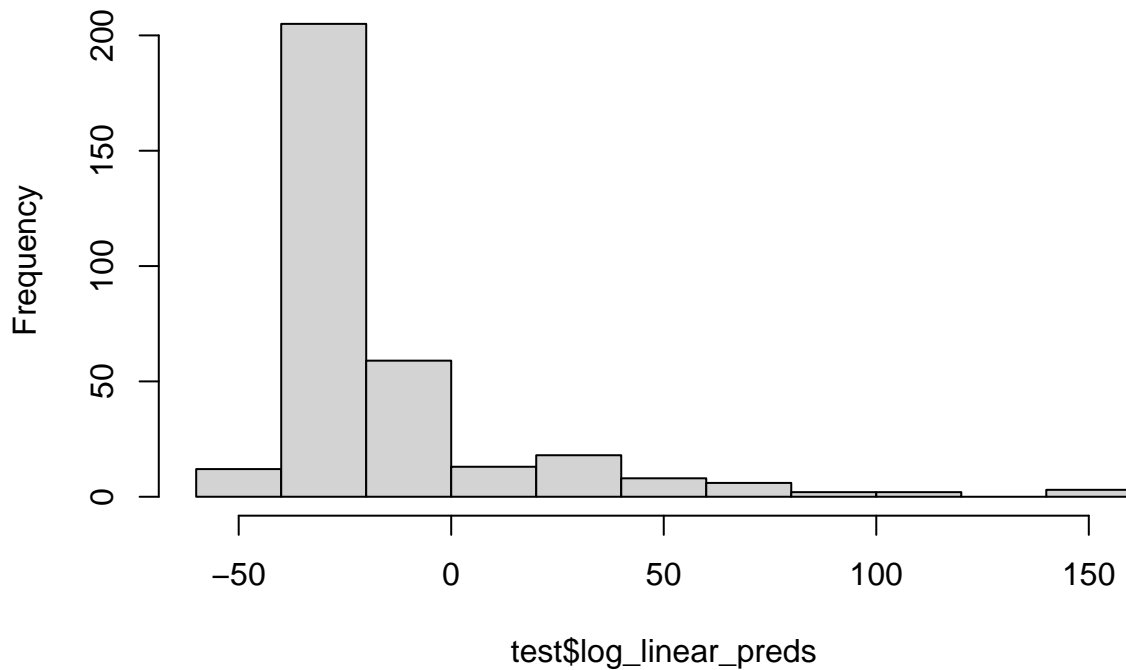


Test Error

LOOKING AT: - log transformed predictors (taxi_in and taxi_out), interactions, no boxcox transformed response - model without any interactions or transformations: ENDED UP PERFORMING THE BEST!!! - box-cox transformed response, no interactions or predicted transformations

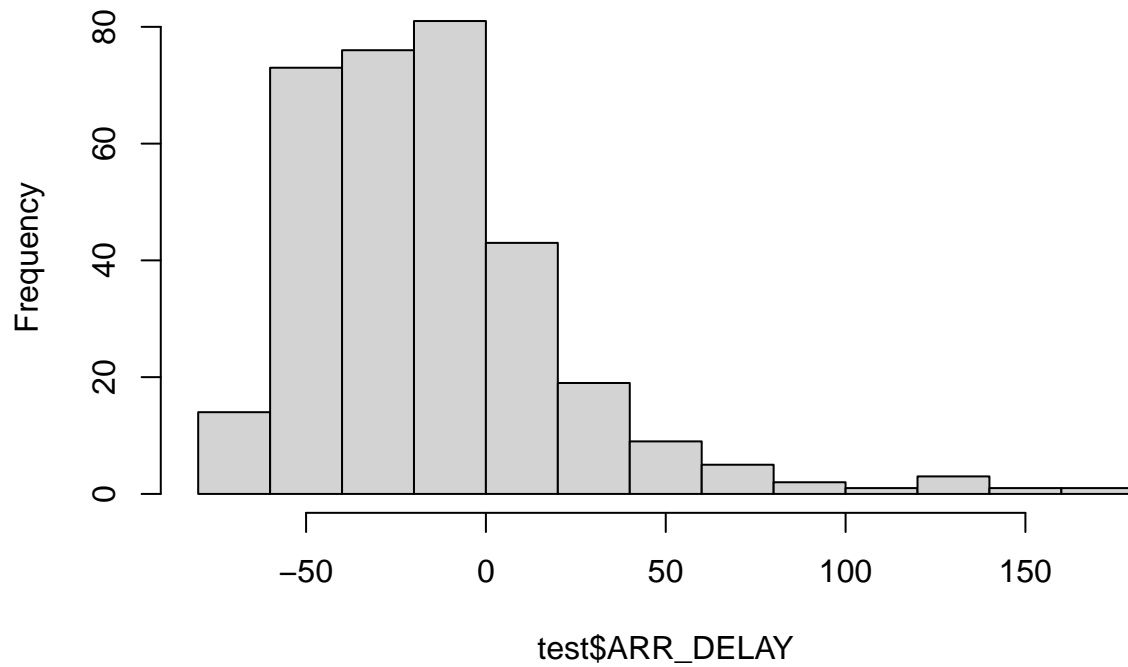
```
#min(test$ARR_DELAY)
test$log_linear_preds <- predict(log_linear_model, test)
hist(test$log_linear_preds)
```

Histogram of test\$log_linear_preds



```
hist(test$ARR_DELAY)
```

Histogram of test\$ARR_DELAY



```
log_linear_MSE <- sum((test$log_linear_preds-test$ARR_DELAY)^2, na.rm=T)/328
log_linear_MSE
```

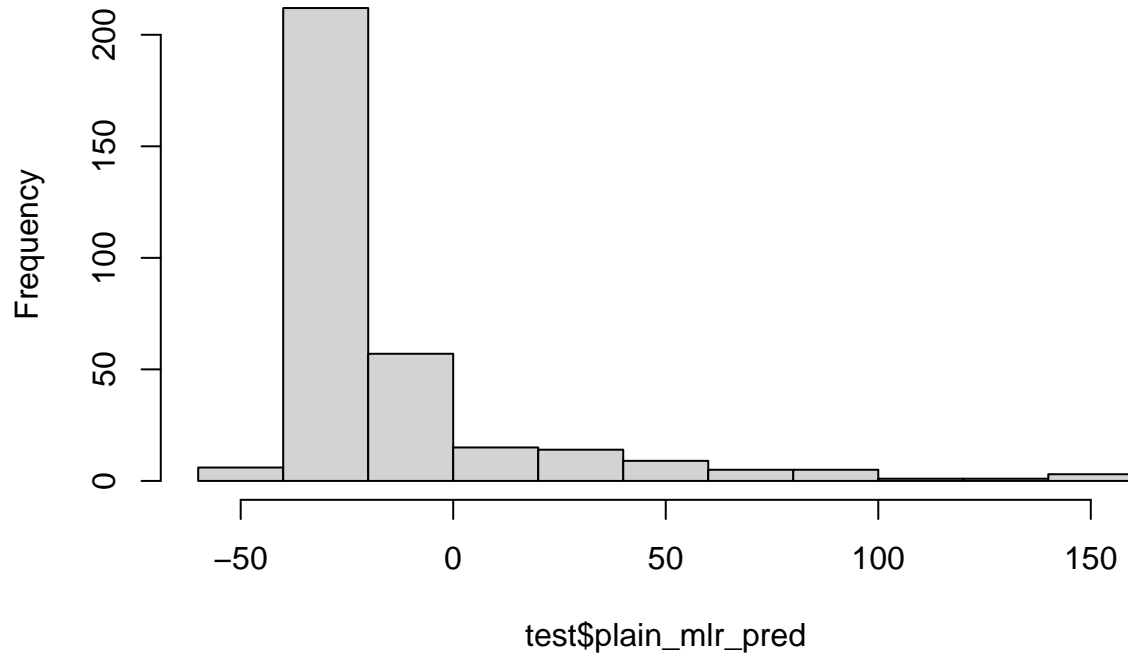
```
## [1] 333.8962
```

```
#min(test$ARR_DELAY)
```

```
test$plain_mlr_pred <- predict(plain_linear_model, test)
```

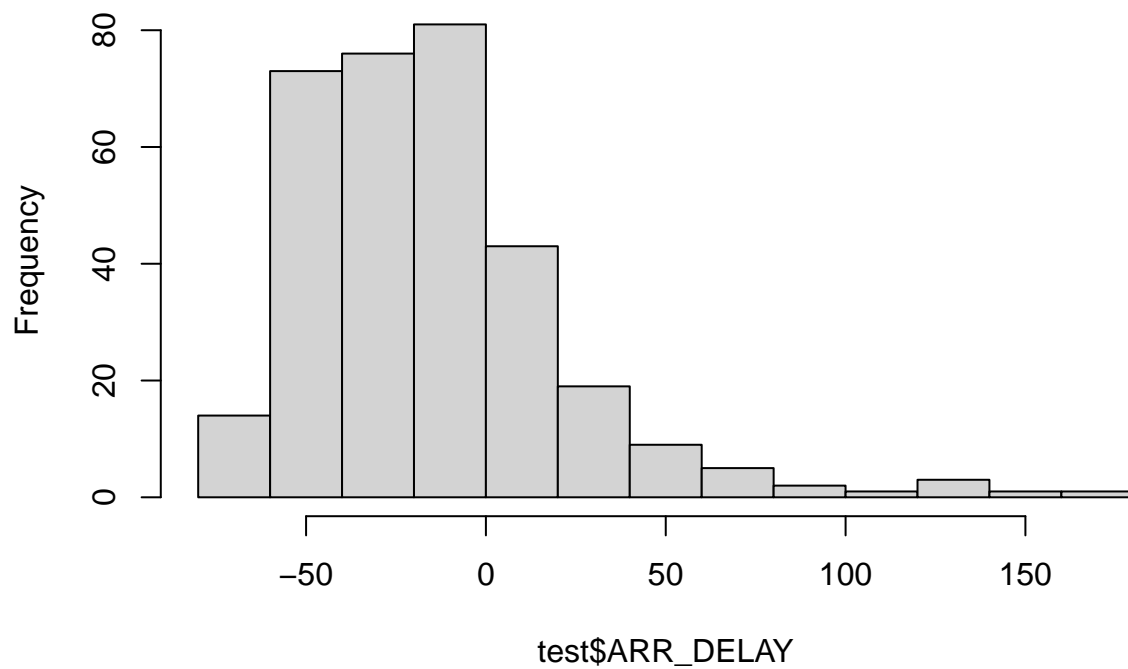
```
hist(test$plain_mlr_pred)
```

Histogram of test\$plain_mlr_pred



```
hist(test$ARR_DELAY)
```

Histogram of test\$ARR_DELAY



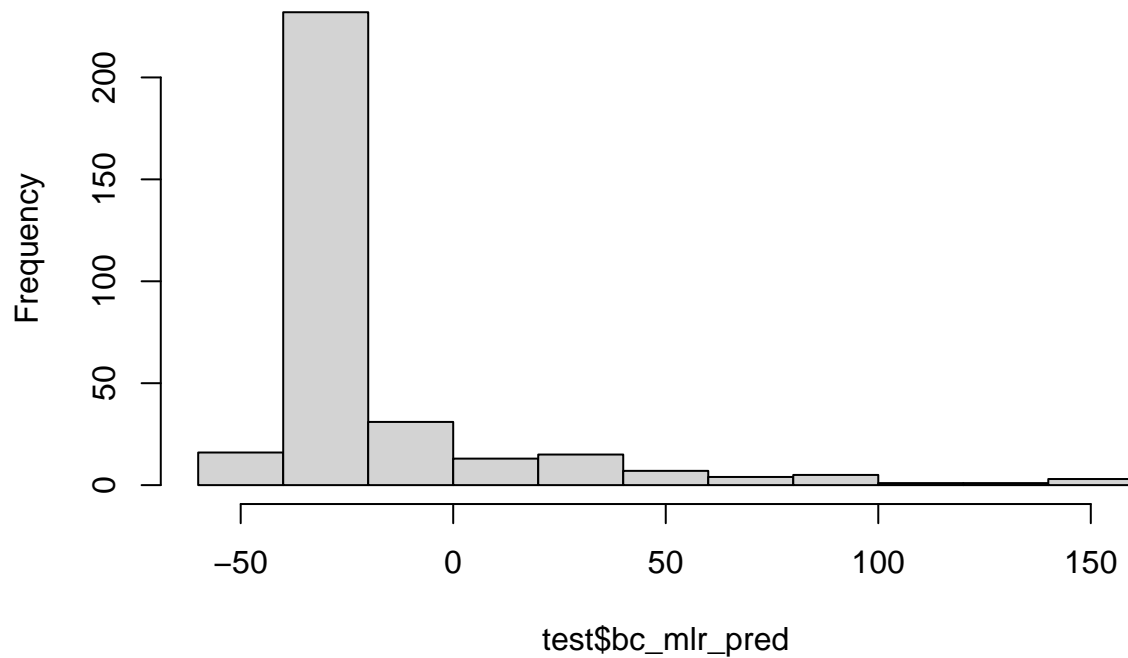
```
plain_linear_model_MSE <- sum((test$ARR_DELAY - test$plain_mlr_pred)^2, na.rm=T)/328  
plain_linear_model_MSE
```

```
## [1] 322.4588
#min(test$ARR_DELAY)
test$adj_ARR_DELAY = test$ARR_DELAY + 77
test$bc_adj_linear_preds <- predict(bc_adj_linear_model, test)
#hist(test$bc_adj_linear_preds)

test <- test %>%
  mutate(adj_linear_preds = ((bc_adj_linear_preds*(bc_lambda) + 1)^(1/bc_lambda)))

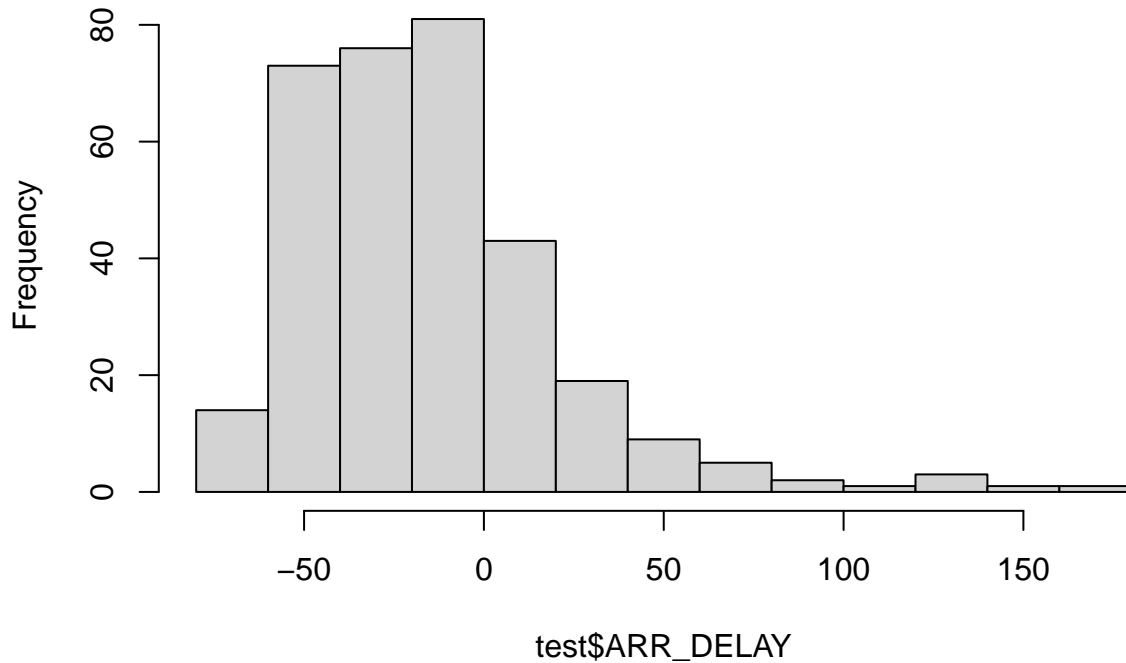
test$bc_mlr_pred = test$adj_linear_preds - 77
hist(test$bc_mlr_pred)
```

Histogram of test\$bc_mlr_pred



```
hist(test$ARR_DELAY)
```

Histogram of test\$ARR_DELAY



```
bc_adj_linear_model_MSE <- sum((test$ARR_DELAY - test$bc_mlr_pred)^2, na.rm=T)/328
bc_adj_linear_model_MSE
```

```
## [1] 334.9217
```

GAM MODEL

Initial Model

fit a gam model with numerical variables on a smoothing spline

```
gam00 <- gam(ARR_DELAY ~ DAY_OF_WEEK +
              OP_CARRIER +
              s(TAXI_IN) +
              s(TAXI_OUT) +
              DEST +
              s(DEP_DELAY) +
              s(CRS_DEP_TIME) +
              s(CRS_ARR_TIME) +
              TYPE_DELAY, data = train)
```

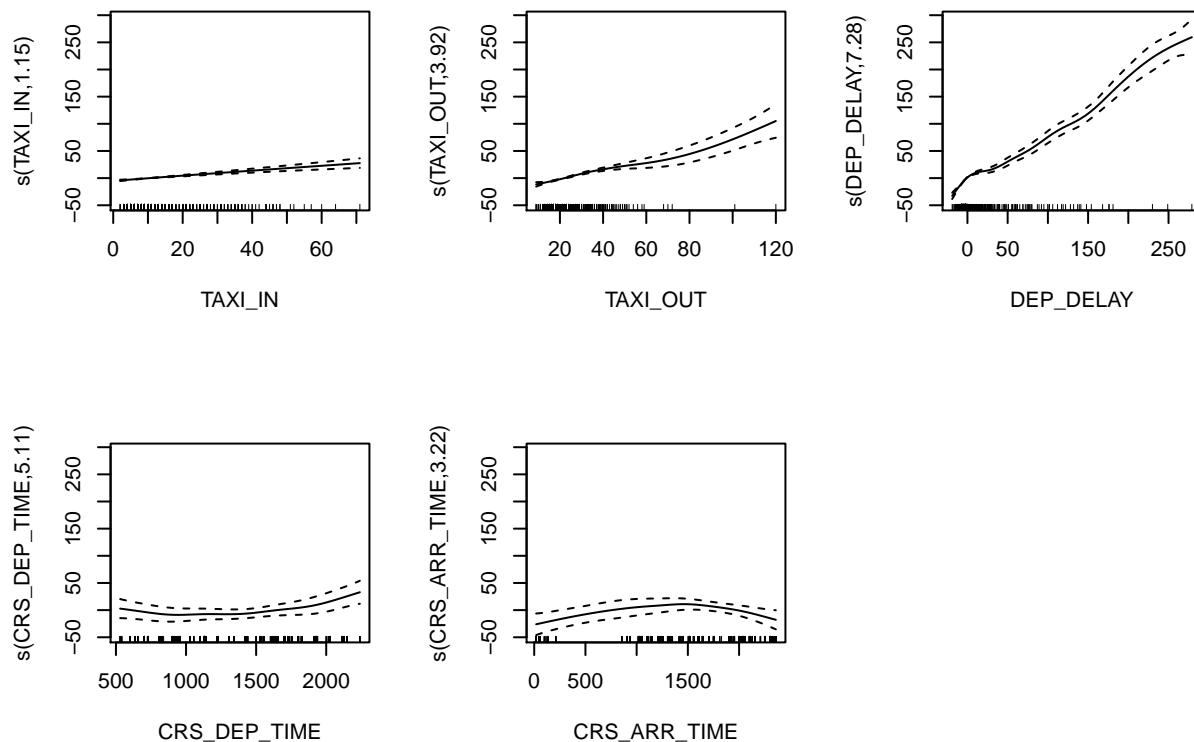
```
summary(gam00)
```

```
##
## Family: gaussian
## Link function: identity
##
## Formula:
## ARR_DELAY ~ DAY_OF_WEEK + OP_CARRIER + s(TAXI_IN) + s(TAXI_OUT) +
##       DEST + s(DEP_DELAY) + s(CRS_DEP_TIME) + s(CRS_ARR_TIME) +
```

```

##      TYPE_DELAY
##
## Parametric coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.6248    4.6488   0.134  0.8931
## DAY_OF_WEEK       0.2465    0.2625   0.939  0.3479
## OP_CARRIERAS     -1.4083    1.6722  -0.842  0.3999
## OP_CARRIERB6      2.8700    1.3616   2.108  0.0352 *
## OP_CARRIERDL     -2.7519    1.3905  -1.979  0.0480 *
## DESTSFO           -0.5607    1.1267  -0.498  0.6188
## TYPE_DELAYLATE_AIRCRAFT -3.3786    6.5848  -0.513  0.6080
## TYPE_DELAYNAS      18.9194    4.5466   4.161 3.38e-05 ***
## TYPE_DELAYNo Delay -22.2130    4.5742  -4.856 1.34e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Approximate significance of smooth terms:
##              edf Ref.df      F  p-value
## s(TAXI_IN)      1.148  1.283 44.974 6.24e-13 ***
## s(TAXI_OUT)      3.922  4.851 46.982 < 2e-16 ***
## s(DEP_DELAY)     7.279  8.272 136.783 < 2e-16 ***
## s(CRS_DEP_TIME)  5.114  6.050   2.121  0.0415 *
## s(CRS_ARR_TIME)  3.221  3.897   1.845  0.1277
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## R-sq.(adj) =  0.749   Deviance explained = 75.5%
## GCV = 317.86   Scale est. = 310.66      n = 1310
par(mfrow = c(2,3))
plot.gam(gam00, se=TRUE)

```

Checking Linearity

TAXI_IN may be linear

```
gam01 <- gam(ARR_DELAY ~ DAY_OF_WEEK +
              OP_CARRIER +
              TAXI_IN +
              s(TAXI_OUT) +
              DEST +
              s(DEP_DELAY) +
              s(CRS_DEP_TIME) +
              s(CRS_ARR_TIME) +
              TYPE_DELAY, data = train)

anova(gam00, gam01, test = "F")
```

```
## Analysis of Deviance Table
##
## Model 1: ARR_DELAY ~ DAY_OF_WEEK + OP_CARRIER + s(TAXI_IN) + s(TAXI_OUT) +
##   DEST + s(DEP_DELAY) + s(CRS_DEP_TIME) + s(CRS_ARR_TIME) +
##   TYPE_DELAY
## Model 2: ARR_DELAY ~ DAY_OF_WEEK + OP_CARRIER + TAXI_IN + s(TAXI_OUT) +
##   DEST + s(DEP_DELAY) + s(CRS_DEP_TIME) + s(CRS_ARR_TIME) +
##   TYPE_DELAY
##   Resid. Df Resid. Dev      Df Deviance      F Pr(>F)
## 1    1276.6    397738
## 2    1276.9    397845 -0.29646  -106.39 1.1552 0.1929
```

based on anova test, the model with a smoothing spline on TAXI_IN is a better fit

More Anova

DAY_OF_WEEK, DEST, and CRS_ARR_TIME have very high p-values, so let's try an anova test without including them

```
gam02 <- gam(ARR_DELAY ~ OP_CARRIER +
              s(TAXI_IN) +
              s(TAXI_OUT) +
              s(DEP_DELAY) +
              s(CRS_DEP_TIME) +
              TYPE_DELAY, data = train)

anova(gam00, gam02, test = "F")
```

```
## Analysis of Deviance Table
##
## Model 1: ARR_DELAY ~ DAY_OF_WEEK + OP_CARRIER + s(TAXI_IN) + s(TAXI_OUT) +
##   DEST + s(DEP_DELAY) + s(CRS_DEP_TIME) + s(CRS_ARR_TIME) +
##   TYPE_DELAY
## Model 2: ARR_DELAY ~ OP_CARRIER + s(TAXI_IN) + s(TAXI_OUT) + s(DEP_DELAY) +
##   s(CRS_DEP_TIME) + TYPE_DELAY
##   Resid. Df Resid. Dev      Df Deviance      F Pr(>F)
## 1      1276.6      397738
## 2      1280.0      400550 -3.3459  -2811.2  2.7045 0.03822 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

based on the anova test, the model excluding these variables is a better fit

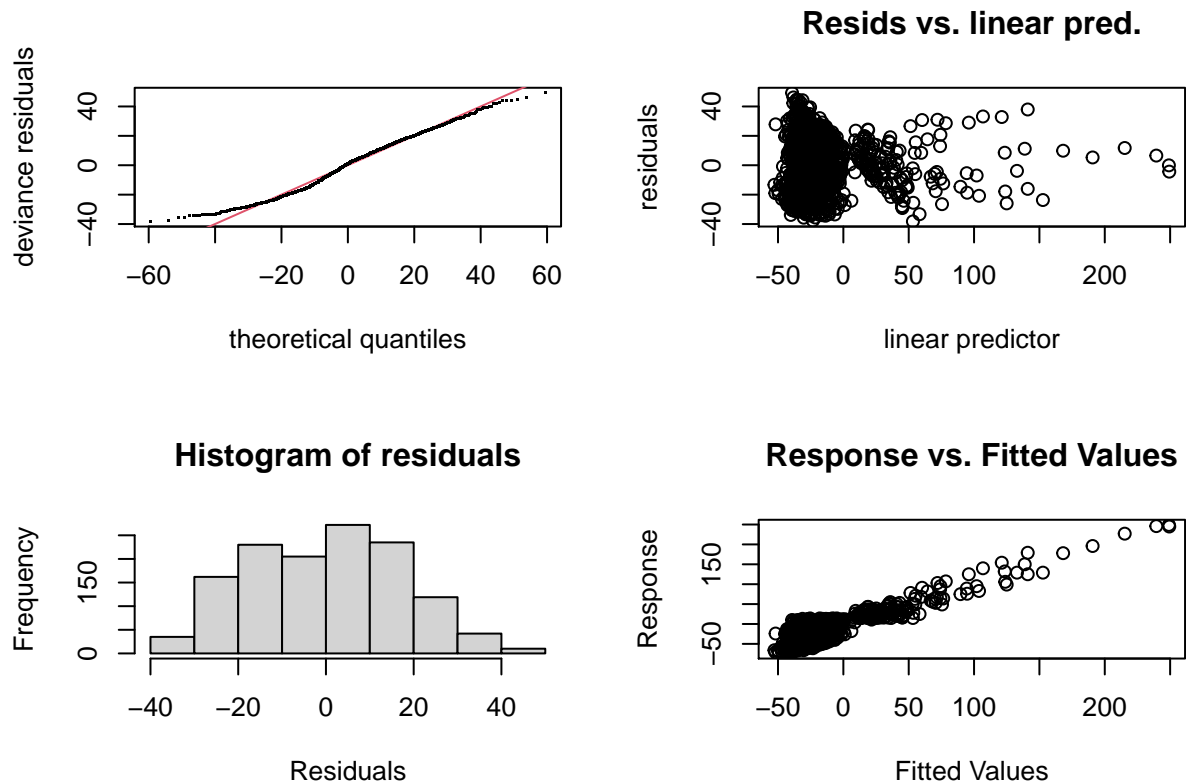
Model Diagnostics

```
summary(gam02)

##
## Family: gaussian
## Link function: identity
##
## Formula:
## ARR_DELAY ~ OP_CARRIER + s(TAXI_IN) + s(TAXI_OUT) + s(DEP_DELAY) +
##   s(CRS_DEP_TIME) + TYPE_DELAY
##
## Parametric coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      1.828      4.530   0.404  0.6865
## OP_CARRIERAS     -1.676      1.664  -1.007  0.3140
## OP_CARRIERB6      2.486      1.354   1.836  0.0666 .
## OP_CARRIERDL     -3.137      1.381  -2.272  0.0233 *
## TYPE_DELAYLATE_AIRCRAFT -3.199      6.601  -0.485  0.6281
## TYPE_DELAYNAS      18.795      4.556   4.126 3.93e-05 ***
## TYPE_DELAYNo Delay -22.413      4.588  -4.885 1.16e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Approximate significance of smooth terms:
##              edf Ref.df      F p-value
## s(TAXI_IN)      1.260  1.478 43.326 7.39e-14 ***
```

```
## s(TAXI_OUT)      4.308  5.298  44.490 < 2e-16 ***
## s(DEP_DELAY)     7.384  8.348 134.478 < 2e-16 ***
## s(CRS_DEP_TIME) 6.781  7.883   5.229 2.36e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## R-sq.(adj) =  0.748   Deviance explained = 75.3%
## GCV = 318.64   Scale est. = 312.13     n = 1310
```

```
par(mfrow = c(2,2))
gam.check(gam02)
```

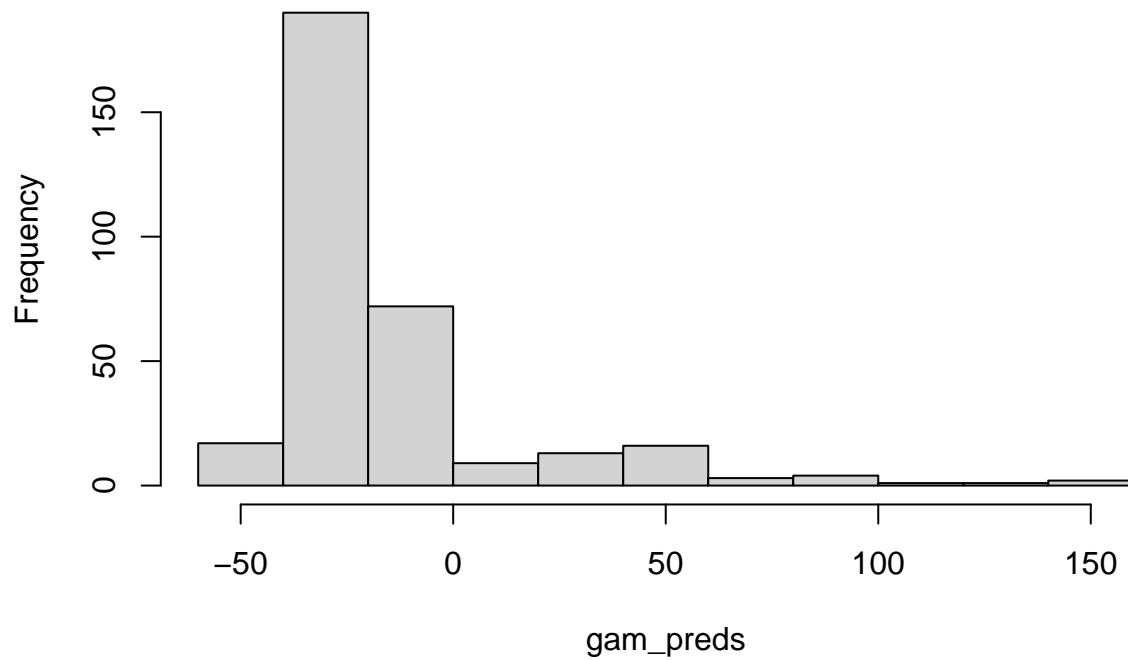


```
##
## Method: GCV   Optimizer: magic
## Smoothing parameter selection converged after 12 iterations.
## The RMS GCV score gradient at convergence was 0.0007156955 .
## The Hessian was positive definite.
## Model rank =  43 / 43
##
## Basis dimension (k) checking results. Low p-value (k-index<1) may
## indicate that k is too low, especially if edf is close to k'.
##
##          k'   edf k-index p-value
## s(TAXI_IN)   9.00 1.26   0.99   0.32
## s(TAXI_OUT)   9.00 4.31   1.07   0.99
## s(DEP_DELAY)   9.00 7.38   0.99   0.31
## s(CRS_DEP_TIME) 9.00 6.78   0.97   0.14
```

Test Error

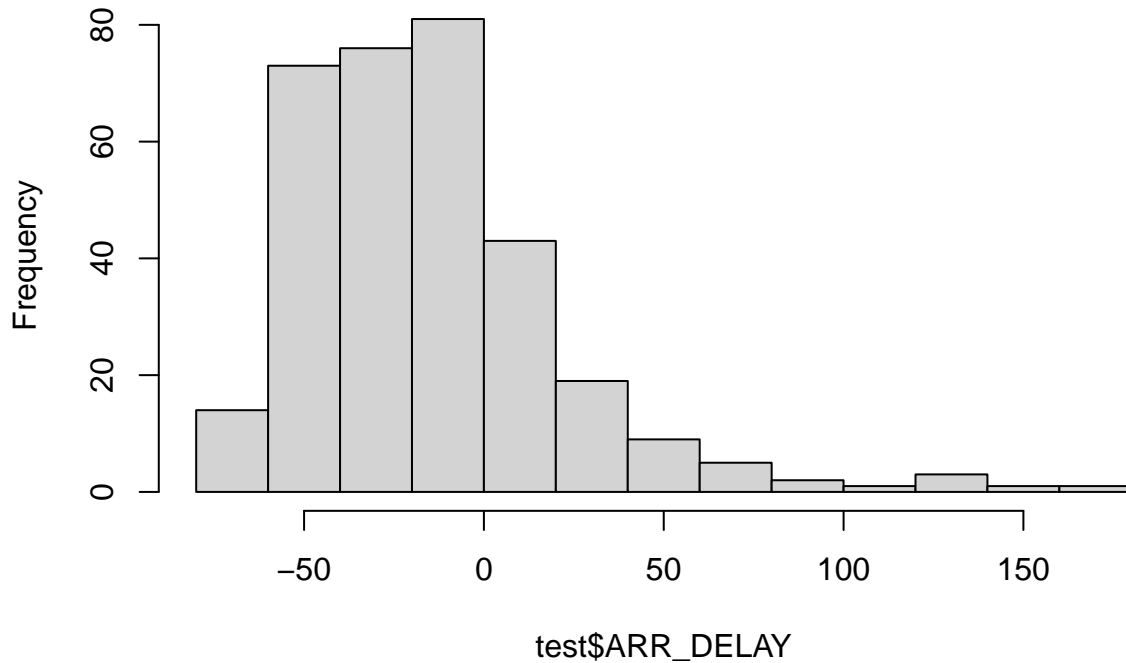
```
gam_preds <- predict.gam(gam02, newdata = test)
hist(gam_preds)
```

Histogram of gam_preds



```
hist(test$ARR_DELAY)
```

Histogram of test\$ARR_DELAY



```
gam_MSE <- sum((test$ARR_DELAY - gam_preds)^2, na.rm=T)/328
gam_MSE
```

```
## [1] 312.2953
```

Boxcox Transformed GAM

```
gambc <- gam(bc_adj_ARR_DELAY ~ OP_CARRIER +
  s(TAXI_IN) +
  s(TAXI_OUT) +
  s(DEP_DELAY) +
  s(CRS_DEP_TIME) +
  TYPE_DELAY, data = train)
```

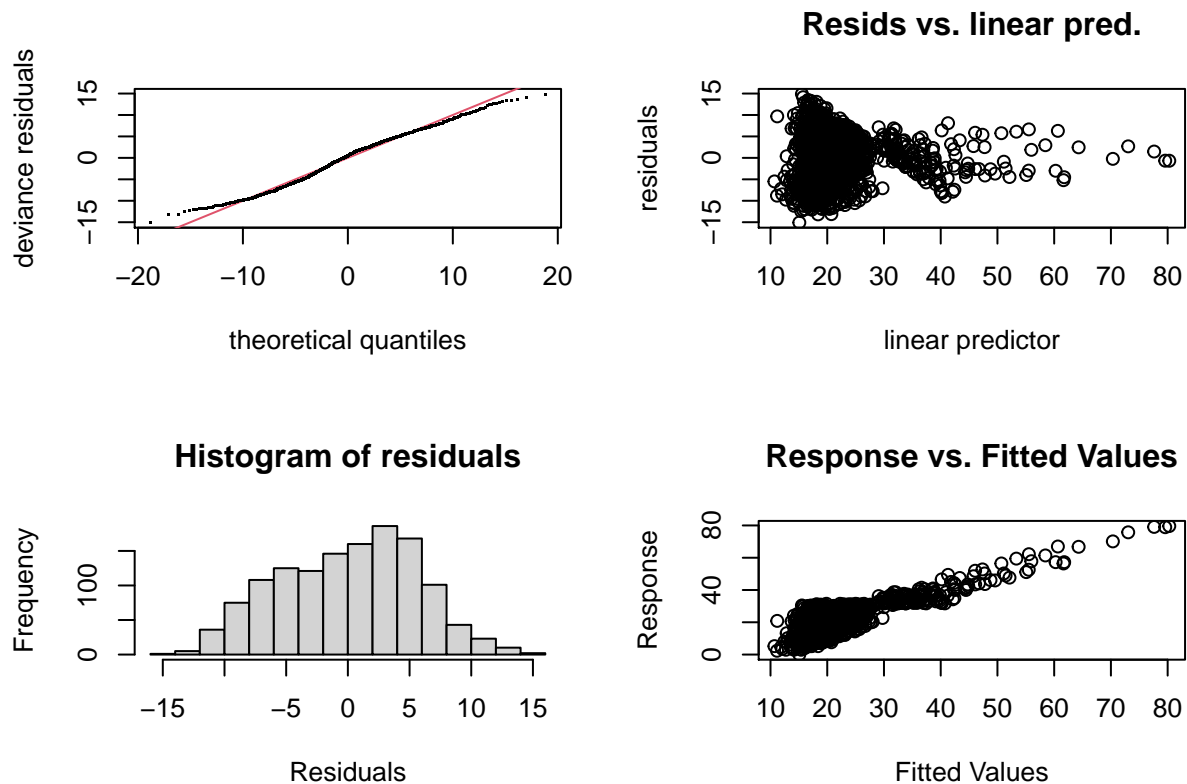
```
summary(gambc)
```

```
##
## Family: gaussian
## Link function: identity
##
## Formula:
## bc_adj_ARR_DELAY ~ OP_CARRIER + s(TAXI_IN) + s(TAXI_OUT) + s(DEP_DELAY) +
##   s(CRS_DEP_TIME) + TYPE_DELAY
##
## Parametric coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    28.2688     1.4316  19.746 < 2e-16 ***
## OP_CARRIERAS   -0.5081     0.5258  -0.966  0.33407
## OP_CARRIERB6    0.7942     0.4275   1.858  0.06340 .
## OP_CARRIERDL   -0.9165     0.4366  -2.099  0.03598 *
```

```
## TYPE_DELAYLATE_AIRCRAFT -1.0709      2.0885  -0.513  0.60822
## TYPE_DELAYNAS           3.9308      1.4403   2.729  0.00644 **
## TYPE_DELAYNo Delay      -7.1102      1.4503  -4.903  1.07e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Approximate significance of smooth terms:
##              edf Ref.df      F  p-value
## s(TAXI_IN)    1.409  1.721 38.501 2.66e-14 ***
## s(TAXI_OUT)    3.241  4.048 46.659 < 2e-16 ***
## s(DEP_DELAY)   7.435  8.383 76.966 < 2e-16 ***
## s(CRS_DEP_TIME) 6.825  7.920  5.521 8.48e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## R-sq.(adj) =  0.662   Deviance explained = 66.8%
## GCV = 31.859   Scale est. = 31.229      n = 1310
```

BC Model Diagnostics

```
par(mfrow = c(2,2))
gam.check(gambc)
```

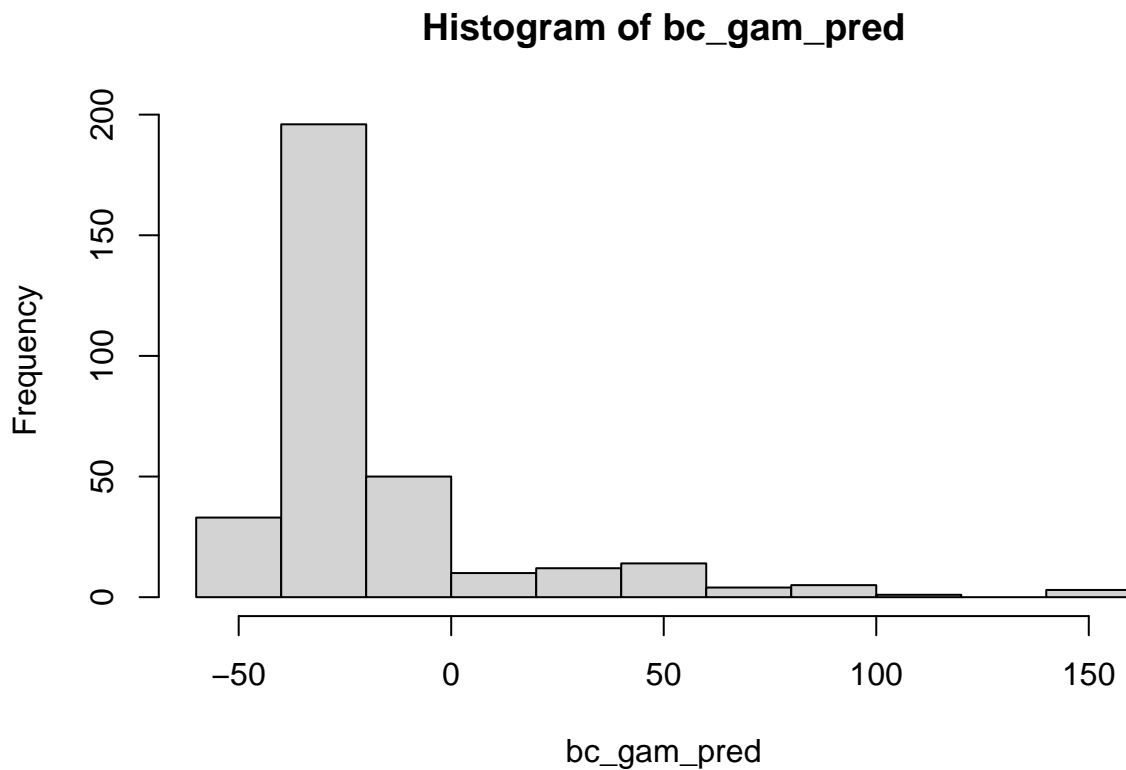


```
##
## Method: GCV   Optimizer: magic
## Smoothing parameter selection converged after 11 iterations.
## The RMS GCV score gradient at convergence was 0.0002369716 .
## The Hessian was positive definite.
## Model rank = 43 / 43
```

```
##
## Basis dimension (k) checking results. Low p-value (k-index<1) may
## indicate that k is too low, especially if edf is close to k'.
##
##           k'   edf k-index p-value
## s(TAXI_IN)   9.00 1.41   0.99   0.36
## s(TAXI_OUT)   9.00 3.24   1.06   0.98
## s(DEP_DELAY)   9.00 7.43   0.98   0.23
## s(CRS_DEP_TIME) 9.00 6.82   0.96   0.03 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

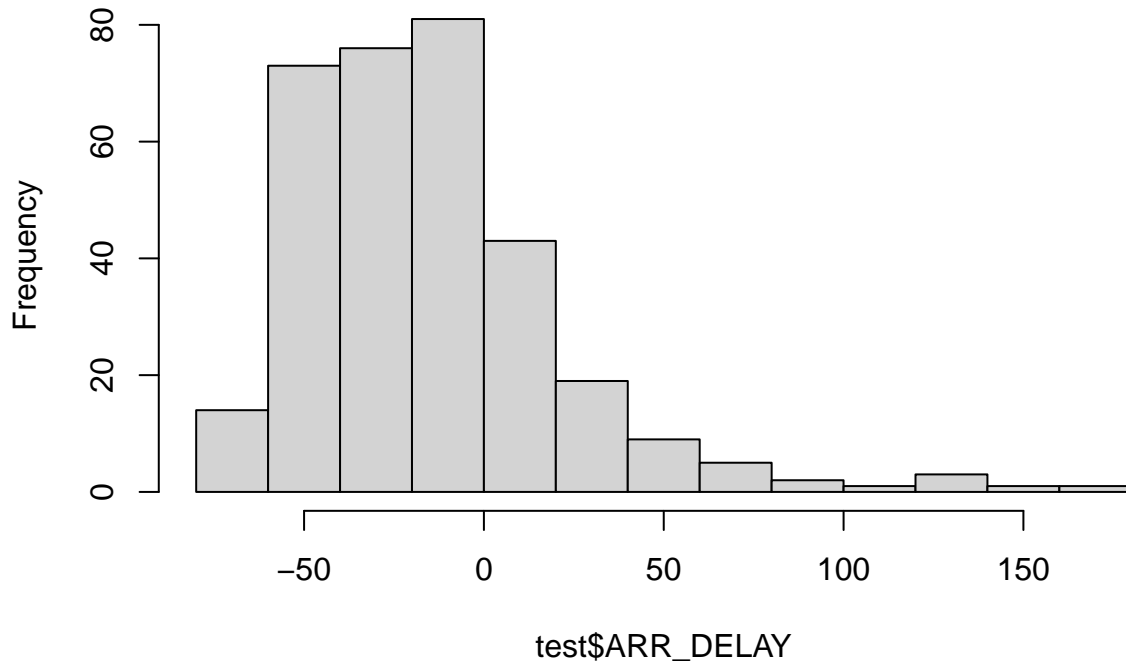
BC Test Error

```
gambc_preds <- predict.gam(gambc, newdata = test)
adjgam_preds <- ((gambc_preds*(bc_lambda) + 1)^(1/bc_lambda))
bc_gam_pred = adjgam_preds - 77
hist(bc_gam_pred)
```



```
hist(test$ARR_DELAY)
```

Histogram of test\$ARR_DELAY



```
gambc_MSE <- sum((test$ARR_DELAY - bc_gam_pred)^2, na.rm=T)/328  
gambc_MSE
```

```
## [1] 317.4533
```

TREES

Random Forests

```
library(tree)  
library(randomForest)
```

By default, `randomForest()` uses $p/3$ variables when building a random forest of regression trees.

```
set.seed(1)  
  
# optimal number of predictors (param = mtry) used = 2 based on CV  
rf.delay <- randomForest(ARR_DELAY ~ DAY_OF_MONTH +  
  TAXI_IN +  
  TAXI_OUT +  
  DEST +  
  DEP_DELAY +  
  CARRIER_DELAY +  
  NAS_DELAY,  
  data = train, na.action = na.omit, importance = TRUE,  
  ntree=10000, mtry=2)  
  
# test error  
yhat.rf <- predict(rf.delay, newdata = test)  
rf.MSE <- sum((test$ARR_DELAY - yhat.rf)^2, na.rm=T)/328
```



```
rf.MSE
```

```
## [1] 155.0148
```

Using the `importance()` function, we can view the importance of each variable.

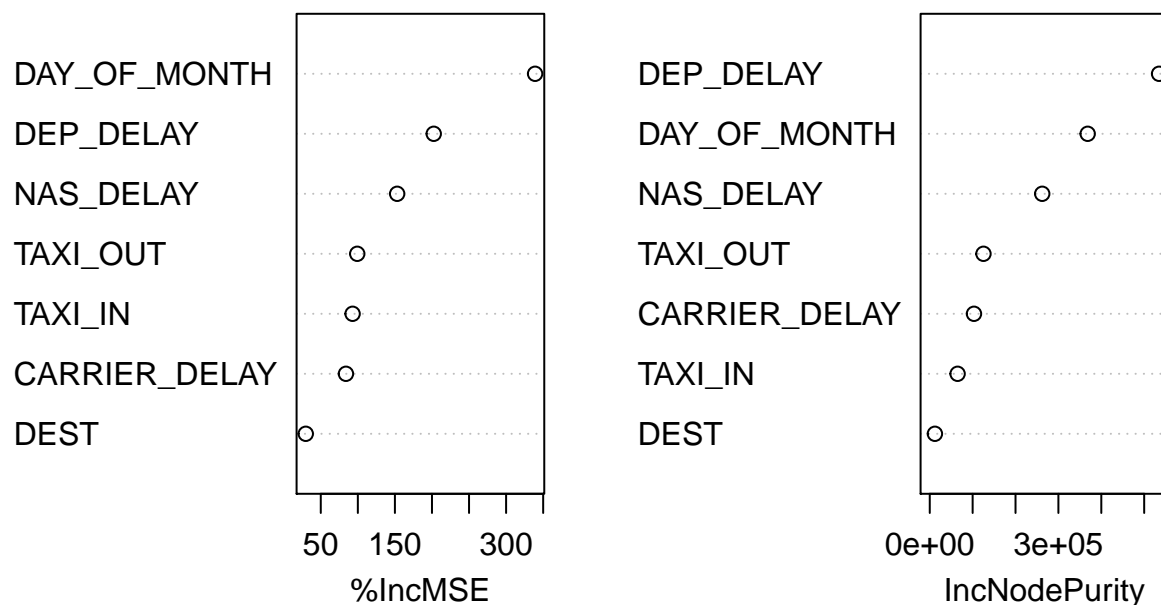
```
importance(rf.delay)
```

```
##           %IncMSE IncNodePurity
## DAY_OF_MONTH 339.08627    368261.46
## TAXI_IN      92.93643     64905.19
## TAXI_OUT     99.21626    125258.87
## DEST         29.76223     12009.51
## DEP_DELAY    202.38571    534776.35
## CARRIER_DELAY 83.95478    103105.51
## NAS_DELAY    153.05251    262232.63
```

Two measures of variable importance are reported. The former is based on the mean decrease in accuracy in predictions on the out of bag samples when a given variable is excluded from the model. The latter is a measure of the total decrease in node impurity that results from splits over that variable, averaged over all trees (this was plotted in Figure 8.9 in the text). In the case of regression trees, the node impurity is measured by the training RSS and for classification trees by the deviance. Plots of these importance measures can be produced using the `varImpPlot()` function.

```
varImpPlot(rf.delay)
```

rf.delay



4. Boosting

Here we use the `gbm()` package, and within it the `gbm()` function, to fit boosted regression trees to the `train` data set. We run `gbm()` with the option `distribution = "gaussian"` since this is a regression problem.

The argument `n.trees = 150` indicates that we want 150 trees, and the option `interaction.depth = 3` limits the depth of each tree.

```
library(gbm)
library(bst)
library(plyr)
library(caret)
set.seed(1)

# find ideal hyper-parameters through CV
gbmFit <- train(ARR_DELAY ~ DAY_OF_MONTH +
                TAXI_IN +
                TAXI_OUT +
                DEP_DELAY +
                CARRIER_DELAY +
                NAS_DELAY +
                LATE_AIRCRAFT_DELAY, data = train,
                method = "gbm")
```

## Iter	TrainDeviance	ValidDeviance	StepSize	Improve
## 1	1150.5648	nan	0.1000	108.0894
## 2	1062.7120	nan	0.1000	98.0251
## 3	999.8411	nan	0.1000	54.3611
## 4	937.5774	nan	0.1000	53.2725
## 5	877.3090	nan	0.1000	60.5384
## 6	834.8623	nan	0.1000	42.6809
## 7	782.6249	nan	0.1000	41.6720
## 8	736.9349	nan	0.1000	47.9831
## 9	695.5806	nan	0.1000	40.5227
## 10	657.6592	nan	0.1000	36.4710
## 20	421.6438	nan	0.1000	15.4628
## 40	244.7935	nan	0.1000	1.2456
## 60	184.0732	nan	0.1000	1.4904
## 80	155.0727	nan	0.1000	0.9043
## 100	140.8694	nan	0.1000	0.3294
## 120	133.2812	nan	0.1000	0.2873
## 140	127.0999	nan	0.1000	0.1086
## 150	124.7120	nan	0.1000	-1.0101
##				
## Iter	TrainDeviance	ValidDeviance	StepSize	Improve
## 1	1091.1076	nan	0.1000	130.0192
## 2	961.6851	nan	0.1000	157.4429
## 3	858.5946	nan	0.1000	101.8085
## 4	790.1614	nan	0.1000	64.1696
## 5	710.8182	nan	0.1000	78.4191
## 6	649.3869	nan	0.1000	63.5996
## 7	598.3666	nan	0.1000	39.1273
## 8	545.5722	nan	0.1000	54.2270
## 9	500.1991	nan	0.1000	34.5864
## 10	467.1205	nan	0.1000	34.3254
## 20	287.3160	nan	0.1000	5.7970
## 40	166.8623	nan	0.1000	3.5853
## 60	132.0353	nan	0.1000	0.5102
## 80	118.0411	nan	0.1000	0.0546
## 100	110.1089	nan	0.1000	0.7461

##	120	103.2643	nan	0.1000	-0.5363
##	140	100.1627	nan	0.1000	-0.0438
##	150	98.7409	nan	0.1000	-0.4050
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1066.5079	nan	0.1000	173.6475
##	2	923.2312	nan	0.1000	140.7612
##	3	801.1209	nan	0.1000	93.9866
##	4	714.0700	nan	0.1000	88.0321
##	5	640.2316	nan	0.1000	68.1676
##	6	569.5613	nan	0.1000	55.4898
##	7	505.9868	nan	0.1000	58.5620
##	8	456.4054	nan	0.1000	49.7937
##	9	419.3526	nan	0.1000	27.1763
##	10	388.8173	nan	0.1000	26.1372
##	20	233.6893	nan	0.1000	7.8144
##	40	135.9065	nan	0.1000	1.1181
##	60	110.4129	nan	0.1000	0.0033
##	80	101.0143	nan	0.1000	-1.0713
##	100	93.8919	nan	0.1000	0.3906
##	120	90.0736	nan	0.1000	-0.2786
##	140	86.1896	nan	0.1000	0.0213
##	150	84.9967	nan	0.1000	-0.3078
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1038.4961	nan	0.1000	79.2910
##	2	964.6743	nan	0.1000	63.7220
##	3	904.4326	nan	0.1000	63.8304
##	4	846.0564	nan	0.1000	35.8399
##	5	796.9160	nan	0.1000	33.8357
##	6	741.3574	nan	0.1000	53.0141
##	7	697.1677	nan	0.1000	42.4855
##	8	658.3158	nan	0.1000	31.6678
##	9	620.2354	nan	0.1000	34.9740
##	10	583.9651	nan	0.1000	37.1194
##	20	365.3850	nan	0.1000	16.3536
##	40	211.8031	nan	0.1000	3.8751
##	60	162.5473	nan	0.1000	1.5584
##	80	139.3494	nan	0.1000	0.7306
##	100	126.2946	nan	0.1000	0.2218
##	120	117.2882	nan	0.1000	0.1613
##	140	110.9744	nan	0.1000	0.3039
##	150	108.7748	nan	0.1000	0.2844
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	977.1061	nan	0.1000	113.4577
##	2	866.4147	nan	0.1000	113.7672
##	3	768.6013	nan	0.1000	79.1478
##	4	692.0857	nan	0.1000	73.6998
##	5	626.7474	nan	0.1000	64.8212
##	6	575.0267	nan	0.1000	52.8815
##	7	521.4352	nan	0.1000	40.5435
##	8	473.4024	nan	0.1000	32.9714
##	9	435.6518	nan	0.1000	33.5803

##	10	404.0142	nan	0.1000	30.8877
##	20	237.3946	nan	0.1000	9.0847
##	40	148.7742	nan	0.1000	0.6326
##	60	116.9303	nan	0.1000	-0.0815
##	80	103.5219	nan	0.1000	0.3205
##	100	97.1446	nan	0.1000	0.0267
##	120	93.7601	nan	0.1000	-0.2853
##	140	90.1817	nan	0.1000	-0.2491
##	150	88.4311	nan	0.1000	0.0747
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	979.2186	nan	0.1000	157.4003
##	2	843.8291	nan	0.1000	140.3897
##	3	728.9358	nan	0.1000	107.9369
##	4	657.4075	nan	0.1000	73.6883
##	5	583.9841	nan	0.1000	74.7565
##	6	525.9856	nan	0.1000	47.9538
##	7	471.0073	nan	0.1000	45.7894
##	8	421.1825	nan	0.1000	45.8401
##	9	381.2552	nan	0.1000	28.7010
##	10	349.2513	nan	0.1000	25.4642
##	20	197.4749	nan	0.1000	7.5401
##	40	118.1161	nan	0.1000	0.2761
##	60	96.2397	nan	0.1000	0.0293
##	80	89.0182	nan	0.1000	0.3721
##	100	83.6028	nan	0.1000	-0.2090
##	120	79.7778	nan	0.1000	-0.2337
##	140	76.0114	nan	0.1000	0.0092
##	150	74.8119	nan	0.1000	-0.2349
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1043.1800	nan	0.1000	65.1818
##	2	959.4253	nan	0.1000	70.1816
##	3	896.6926	nan	0.1000	61.6784
##	4	843.1070	nan	0.1000	50.6219
##	5	796.5353	nan	0.1000	40.8339
##	6	746.4385	nan	0.1000	46.4100
##	7	702.8579	nan	0.1000	41.7416
##	8	663.6213	nan	0.1000	38.0004
##	9	625.0179	nan	0.1000	39.5442
##	10	591.8549	nan	0.1000	27.8403
##	20	378.0878	nan	0.1000	12.1186
##	40	231.7994	nan	0.1000	3.0440
##	60	180.3416	nan	0.1000	1.4298
##	80	156.3828	nan	0.1000	0.6180
##	100	144.4843	nan	0.1000	0.4221
##	120	136.6853	nan	0.1000	-0.0499
##	140	130.7566	nan	0.1000	0.0179
##	150	128.6565	nan	0.1000	-0.3876
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	988.9693	nan	0.1000	135.6243
##	2	878.1054	nan	0.1000	115.2549
##	3	782.7530	nan	0.1000	68.8872

##	4	706.5929	nan	0.1000	65.4505
##	5	647.2079	nan	0.1000	53.1487
##	6	584.1888	nan	0.1000	65.6355
##	7	536.8114	nan	0.1000	43.1166
##	8	496.6526	nan	0.1000	42.0299
##	9	460.4912	nan	0.1000	35.7958
##	10	431.7768	nan	0.1000	27.7186
##	20	263.1444	nan	0.1000	9.9121
##	40	160.8285	nan	0.1000	1.3644
##	60	131.1793	nan	0.1000	0.8514
##	80	117.1487	nan	0.1000	0.2811
##	100	109.9067	nan	0.1000	-0.3520
##	120	103.3028	nan	0.1000	-0.0456
##	140	99.4621	nan	0.1000	-1.4216
##	150	97.8404	nan	0.1000	-0.1779
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	963.7173	nan	0.1000	145.6485
##	2	830.0235	nan	0.1000	140.3019
##	3	729.2431	nan	0.1000	112.9741
##	4	642.8786	nan	0.1000	85.5526
##	5	579.7839	nan	0.1000	67.1794
##	6	525.0721	nan	0.1000	54.9593
##	7	482.4734	nan	0.1000	44.9557
##	8	440.6059	nan	0.1000	36.2972
##	9	405.1308	nan	0.1000	30.6701
##	10	377.1782	nan	0.1000	32.2158
##	20	223.1937	nan	0.1000	8.3633
##	40	137.5210	nan	0.1000	2.7908
##	60	115.6683	nan	0.1000	-0.1178
##	80	103.5987	nan	0.1000	0.1486
##	100	97.4186	nan	0.1000	0.0468
##	120	92.1760	nan	0.1000	-0.6028
##	140	87.8751	nan	0.1000	-0.0790
##	150	86.0851	nan	0.1000	-0.2352
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1114.9069	nan	0.1000	111.0734
##	2	1026.6882	nan	0.1000	60.7724
##	3	960.8236	nan	0.1000	65.4554
##	4	905.6734	nan	0.1000	56.6590
##	5	844.7358	nan	0.1000	56.0554
##	6	790.0563	nan	0.1000	37.1147
##	7	747.9614	nan	0.1000	43.9995
##	8	703.2455	nan	0.1000	37.1534
##	9	660.5436	nan	0.1000	37.6661
##	10	620.8812	nan	0.1000	37.8569
##	20	394.0971	nan	0.1000	13.1611
##	40	227.3770	nan	0.1000	3.0090
##	60	175.9880	nan	0.1000	0.6697
##	80	153.5385	nan	0.1000	0.4214
##	100	141.7942	nan	0.1000	0.0400
##	120	133.9781	nan	0.1000	0.2538
##	140	128.5789	nan	0.1000	-0.6216

```

##      150      127.0912          nan      0.1000      0.2749
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1058.4375          nan      0.1000     179.6786
##      2       934.4213          nan      0.1000     123.0235
##      3       826.0630          nan      0.1000     119.9622
##      4       741.0864          nan      0.1000      76.9391
##      5       673.9855          nan      0.1000      59.9329
##      6       614.3471          nan      0.1000      63.7129
##      7       565.6583          nan      0.1000      51.6028
##      8       515.4718          nan      0.1000      48.5894
##      9       473.3732          nan      0.1000      39.6733
##     10       437.3717          nan      0.1000      25.2694
##     20       255.5560          nan      0.1000       9.7985
##     40       158.6039          nan      0.1000       0.8082
##     60       130.2847          nan      0.1000       1.6124
##     80       119.4850          nan      0.1000       0.1706
##    100       113.3957          nan      0.1000      -0.4501
##    120       107.7894          nan      0.1000       0.0992
##    140       103.8321          nan      0.1000      -0.2636
##    150       102.7621          nan      0.1000      -0.2770
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1050.1920          nan      0.1000     157.4568
##      2       908.9149          nan      0.1000     134.2770
##      3       813.2572          nan      0.1000      88.2433
##      4       713.7046          nan      0.1000      97.3873
##      5       630.2140          nan      0.1000      73.9902
##      6       555.7110          nan      0.1000      72.0744
##      7       505.4664          nan      0.1000      57.3740
##      8       459.7692          nan      0.1000      50.6590
##      9       418.2648          nan      0.1000      38.7313
##     10       381.5394          nan      0.1000      28.0108
##     20       218.5273          nan      0.1000       8.5399
##     40       139.6013          nan      0.1000       1.0679
##     60       115.4817          nan      0.1000       0.4317
##     80       105.5851          nan      0.1000      -0.4976
##    100        98.7866          nan      0.1000      -0.0657
##    120        94.8003          nan      0.1000      -0.2332
##    140        92.0789          nan      0.1000      -0.4928
##    150        89.9621          nan      0.1000      -0.4581
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1363.0377          nan      0.1000     115.9742
##      2      1261.0838          nan      0.1000      95.4869
##      3      1169.9134          nan      0.1000      89.4782
##      4      1098.3658          nan      0.1000      58.2659
##      5      1033.3767          nan      0.1000      65.9647
##      6       975.8048          nan      0.1000      57.3396
##      7       916.9678          nan      0.1000      51.0980
##      8       866.2389          nan      0.1000      48.7142
##      9       822.0977          nan      0.1000      45.7419
##     10       780.3579          nan      0.1000      40.0106
##     20       473.4706          nan      0.1000      21.2260

```

##	40	261.5182	nan	0.1000	4.9859
##	60	188.8009	nan	0.1000	2.0729
##	80	156.3942	nan	0.1000	0.2772
##	100	139.4622	nan	0.1000	0.4014
##	120	130.4820	nan	0.1000	0.2730
##	140	123.5861	nan	0.1000	-0.1129
##	150	120.8073	nan	0.1000	0.0414
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1313.9267	nan	0.1000	178.6950
##	2	1154.3750	nan	0.1000	158.1254
##	3	1021.4058	nan	0.1000	96.9070
##	4	907.3896	nan	0.1000	93.8171
##	5	821.2174	nan	0.1000	79.4315
##	6	735.9728	nan	0.1000	65.2562
##	7	672.5909	nan	0.1000	59.5681
##	8	625.1451	nan	0.1000	44.8287
##	9	569.7428	nan	0.1000	37.7258
##	10	521.7478	nan	0.1000	38.0977
##	20	296.2888	nan	0.1000	12.0787
##	40	167.2275	nan	0.1000	1.9542
##	60	133.7161	nan	0.1000	0.8959
##	80	115.3132	nan	0.1000	-0.3287
##	100	107.6759	nan	0.1000	0.3238
##	120	102.2019	nan	0.1000	-0.2904
##	140	96.2270	nan	0.1000	0.0778
##	150	94.6262	nan	0.1000	-0.3259
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1284.6787	nan	0.1000	216.3611
##	2	1110.8225	nan	0.1000	156.8866
##	3	957.2629	nan	0.1000	157.3410
##	4	837.9827	nan	0.1000	107.2450
##	5	739.2871	nan	0.1000	94.1477
##	6	657.4487	nan	0.1000	70.9209
##	7	592.1601	nan	0.1000	65.7733
##	8	533.5499	nan	0.1000	54.9894
##	9	481.1973	nan	0.1000	54.6397
##	10	450.3253	nan	0.1000	32.3668
##	20	242.1455	nan	0.1000	16.5360
##	40	134.1906	nan	0.1000	1.2931
##	60	107.5346	nan	0.1000	0.6984
##	80	96.4049	nan	0.1000	0.4291
##	100	88.7862	nan	0.1000	0.3195
##	120	84.0447	nan	0.1000	-0.3354
##	140	79.9662	nan	0.1000	-0.0714
##	150	78.3658	nan	0.1000	-0.0550
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1119.6872	nan	0.1000	64.0268
##	2	1028.6291	nan	0.1000	89.2680
##	3	955.2203	nan	0.1000	51.7304
##	4	886.0096	nan	0.1000	76.9672
##	5	833.6934	nan	0.1000	52.1605

##	6	792.9316	nan	0.1000	36.7743
##	7	744.9836	nan	0.1000	38.6176
##	8	697.4934	nan	0.1000	37.4781
##	9	656.1072	nan	0.1000	39.4088
##	10	622.1726	nan	0.1000	33.2767
##	20	396.7593	nan	0.1000	17.3952
##	40	235.1882	nan	0.1000	1.0045
##	60	179.7141	nan	0.1000	1.2229
##	80	154.9271	nan	0.1000	0.6423
##	100	143.1362	nan	0.1000	0.2508
##	120	135.1987	nan	0.1000	0.0048
##	140	129.8053	nan	0.1000	-0.5869
##	150	127.0031	nan	0.1000	0.1124
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1030.7566	nan	0.1000	138.8045
##	2	931.4097	nan	0.1000	108.5589
##	3	823.2101	nan	0.1000	119.9196
##	4	737.1424	nan	0.1000	83.8840
##	5	682.1829	nan	0.1000	52.8919
##	6	617.9732	nan	0.1000	55.2543
##	7	564.1432	nan	0.1000	44.3180
##	8	518.3703	nan	0.1000	41.4856
##	9	485.0086	nan	0.1000	35.0296
##	10	459.1011	nan	0.1000	24.7205
##	20	265.7766	nan	0.1000	6.5308
##	40	163.6818	nan	0.1000	1.6940
##	60	134.7479	nan	0.1000	0.6209
##	80	121.5619	nan	0.1000	0.4184
##	100	111.6734	nan	0.1000	-0.3092
##	120	106.5058	nan	0.1000	-0.0169
##	140	101.4909	nan	0.1000	0.1760
##	150	98.6352	nan	0.1000	0.0204
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1030.4886	nan	0.1000	169.4195
##	2	894.3337	nan	0.1000	147.4280
##	3	781.2207	nan	0.1000	99.2478
##	4	692.5225	nan	0.1000	86.5989
##	5	614.2082	nan	0.1000	74.9347
##	6	556.0481	nan	0.1000	53.3782
##	7	506.4840	nan	0.1000	45.6967
##	8	456.4495	nan	0.1000	44.0783
##	9	425.2643	nan	0.1000	34.2144
##	10	394.9264	nan	0.1000	30.4285
##	20	233.0086	nan	0.1000	7.0539
##	40	142.0020	nan	0.1000	1.1186
##	60	116.1049	nan	0.1000	0.5247
##	80	102.4701	nan	0.1000	-0.5115
##	100	94.1522	nan	0.1000	0.1074
##	120	88.0607	nan	0.1000	0.2569
##	140	84.4240	nan	0.1000	0.1709
##	150	82.9162	nan	0.1000	-0.0851
##					

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1151.1140	nan	0.1000	74.8637
##	2	1073.9338	nan	0.1000	73.2535
##	3	1014.7446	nan	0.1000	60.9543
##	4	947.6675	nan	0.1000	59.7251
##	5	902.7024	nan	0.1000	48.6272
##	6	840.8978	nan	0.1000	61.2264
##	7	795.5811	nan	0.1000	39.0083
##	8	748.1772	nan	0.1000	48.2803
##	9	709.2787	nan	0.1000	36.1646
##	10	675.0291	nan	0.1000	35.6680
##	20	425.3133	nan	0.1000	16.5287
##	40	244.8662	nan	0.1000	1.7448
##	60	187.7142	nan	0.1000	1.3188
##	80	158.9160	nan	0.1000	0.8889
##	100	145.4590	nan	0.1000	0.4266
##	120	136.4194	nan	0.1000	0.2414
##	140	130.0089	nan	0.1000	0.3067
##	150	128.3017	nan	0.1000	-1.5994
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1089.0163	nan	0.1000	147.1273
##	2	969.9221	nan	0.1000	116.2787
##	3	880.5462	nan	0.1000	78.6197
##	4	812.5230	nan	0.1000	59.9873
##	5	726.9211	nan	0.1000	84.5285
##	6	677.4496	nan	0.1000	52.2021
##	7	612.9353	nan	0.1000	50.6274
##	8	561.4688	nan	0.1000	48.4978
##	9	512.3283	nan	0.1000	45.8716
##	10	483.8197	nan	0.1000	29.7787
##	20	277.0257	nan	0.1000	10.2003
##	40	170.0154	nan	0.1000	0.9824
##	60	138.0782	nan	0.1000	0.1468
##	80	121.8979	nan	0.1000	1.0083
##	100	113.0314	nan	0.1000	0.7136
##	120	106.1635	nan	0.1000	0.4506
##	140	101.5379	nan	0.1000	-0.1965
##	150	99.3361	nan	0.1000	0.0973
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1076.3660	nan	0.1000	166.0326
##	2	934.8358	nan	0.1000	127.5496
##	3	821.5449	nan	0.1000	106.6992
##	4	746.0042	nan	0.1000	77.6186
##	5	662.0705	nan	0.1000	88.5528
##	6	587.9474	nan	0.1000	68.8832
##	7	524.7118	nan	0.1000	57.8505
##	8	479.7467	nan	0.1000	47.0668
##	9	436.0507	nan	0.1000	39.3705
##	10	399.8243	nan	0.1000	31.9315
##	20	220.4503	nan	0.1000	6.6917
##	40	137.8457	nan	0.1000	0.0340
##	60	111.5257	nan	0.1000	0.4765

##	80	103.2541	nan	0.1000	0.3290
##	100	95.8835	nan	0.1000	0.0615
##	120	90.3473	nan	0.1000	-1.0049
##	140	86.4337	nan	0.1000	-0.4626
##	150	84.6927	nan	0.1000	-0.2075
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1339.4377	nan	0.1000	113.2613
##	2	1216.2615	nan	0.1000	99.2026
##	3	1126.4295	nan	0.1000	68.4633
##	4	1046.2734	nan	0.1000	82.6301
##	5	983.9641	nan	0.1000	62.1021
##	6	914.1832	nan	0.1000	58.4017
##	7	852.3048	nan	0.1000	55.8210
##	8	808.5896	nan	0.1000	38.0359
##	9	761.6549	nan	0.1000	46.5411
##	10	717.0981	nan	0.1000	45.1370
##	20	436.3161	nan	0.1000	5.0169
##	40	247.8442	nan	0.1000	4.0908
##	60	189.6155	nan	0.1000	1.2188
##	80	163.7345	nan	0.1000	-0.4299
##	100	151.1381	nan	0.1000	0.3984
##	120	143.2722	nan	0.1000	0.3737
##	140	137.0679	nan	0.1000	0.2390
##	150	134.8423	nan	0.1000	-0.2984
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1293.9024	nan	0.1000	152.8322
##	2	1141.1291	nan	0.1000	163.8377
##	3	1004.9357	nan	0.1000	125.6723
##	4	904.6677	nan	0.1000	101.5808
##	5	806.7888	nan	0.1000	95.9764
##	6	727.2010	nan	0.1000	56.4767
##	7	671.4384	nan	0.1000	51.6276
##	8	611.6293	nan	0.1000	47.2025
##	9	561.1445	nan	0.1000	48.9747
##	10	523.9407	nan	0.1000	41.5284
##	20	290.1424	nan	0.1000	8.3678
##	40	169.4780	nan	0.1000	2.2472
##	60	139.4834	nan	0.1000	0.5103
##	80	124.4535	nan	0.1000	0.4551
##	100	116.3915	nan	0.1000	0.0335
##	120	109.1921	nan	0.1000	-0.3405
##	140	105.2343	nan	0.1000	-0.4163
##	150	103.8512	nan	0.1000	-0.3317
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1267.3354	nan	0.1000	185.0019
##	2	1107.6952	nan	0.1000	184.8420
##	3	963.0019	nan	0.1000	143.9640
##	4	842.8009	nan	0.1000	109.3526
##	5	751.5434	nan	0.1000	106.4052
##	6	668.2674	nan	0.1000	99.3912
##	7	596.6896	nan	0.1000	53.4083

##	8	542.7084	nan	0.1000	57.2947
##	9	501.0338	nan	0.1000	42.9532
##	10	455.1269	nan	0.1000	46.6512
##	20	255.6455	nan	0.1000	8.0047
##	40	152.1644	nan	0.1000	0.5435
##	60	122.5953	nan	0.1000	0.5605
##	80	109.4411	nan	0.1000	0.0806
##	100	102.2738	nan	0.1000	0.0991
##	120	96.9834	nan	0.1000	-0.6580
##	140	92.7473	nan	0.1000	-0.6350
##	150	90.8796	nan	0.1000	-0.2192

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1309.6891	nan	0.1000	119.7429
##	2	1192.8718	nan	0.1000	131.6083
##	3	1106.9567	nan	0.1000	79.0799
##	4	1030.6708	nan	0.1000	83.9500
##	5	960.5429	nan	0.1000	60.8600
##	6	892.6716	nan	0.1000	56.2380
##	7	833.7406	nan	0.1000	62.1241
##	8	785.0503	nan	0.1000	50.2280
##	9	731.9059	nan	0.1000	58.2727
##	10	686.1060	nan	0.1000	47.9452
##	20	400.7869	nan	0.1000	13.7796
##	40	224.7410	nan	0.1000	2.2285
##	60	170.6971	nan	0.1000	1.0392
##	80	145.5117	nan	0.1000	0.8516
##	100	132.2849	nan	0.1000	0.1497
##	120	123.5844	nan	0.1000	-0.1524
##	140	117.6299	nan	0.1000	0.0698
##	150	115.2245	nan	0.1000	-0.0096

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1215.0085	nan	0.1000	178.3146
##	2	1083.5683	nan	0.1000	136.7143
##	3	980.2517	nan	0.1000	107.0384
##	4	875.5395	nan	0.1000	91.5845
##	5	777.3021	nan	0.1000	89.5319
##	6	697.5710	nan	0.1000	81.7018
##	7	629.7203	nan	0.1000	48.0097
##	8	574.1415	nan	0.1000	61.9125
##	9	524.1105	nan	0.1000	43.0415
##	10	487.4983	nan	0.1000	38.4445
##	20	262.9601	nan	0.1000	5.6366
##	40	159.6440	nan	0.1000	1.8309
##	60	129.1143	nan	0.1000	0.1300
##	80	114.9550	nan	0.1000	0.0602
##	100	106.5549	nan	0.1000	0.1337
##	120	101.1085	nan	0.1000	-0.0446
##	140	96.6816	nan	0.1000	-0.2191
##	150	94.8257	nan	0.1000	-0.4427

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1202.8157	nan	0.1000	211.6199

##	2	1043.5734	nan	0.1000	158.7406
##	3	894.0216	nan	0.1000	117.4613
##	4	778.1679	nan	0.1000	113.2346
##	5	696.3307	nan	0.1000	97.7456
##	6	617.0662	nan	0.1000	76.7895
##	7	548.1352	nan	0.1000	55.9634
##	8	497.3939	nan	0.1000	48.5479
##	9	452.4059	nan	0.1000	43.9317
##	10	409.5711	nan	0.1000	38.9295
##	20	226.4281	nan	0.1000	9.2007
##	40	131.8420	nan	0.1000	1.4361
##	60	107.8706	nan	0.1000	0.4472
##	80	97.8616	nan	0.1000	0.1277
##	100	90.9571	nan	0.1000	-0.1795
##	120	86.1733	nan	0.1000	-0.4309
##	140	81.2866	nan	0.1000	-0.3627
##	150	79.7317	nan	0.1000	-0.1314

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1197.8704	nan	0.1000	105.9670
##	2	1116.4273	nan	0.1000	79.7892
##	3	1039.6990	nan	0.1000	79.0002
##	4	967.4720	nan	0.1000	69.3039
##	5	914.7650	nan	0.1000	54.4291
##	6	861.5939	nan	0.1000	60.7256
##	7	806.6787	nan	0.1000	52.6622
##	8	764.8375	nan	0.1000	42.8954
##	9	732.0972	nan	0.1000	34.8340
##	10	682.1197	nan	0.1000	22.3085
##	20	422.4240	nan	0.1000	15.8138
##	40	244.9176	nan	0.1000	2.6851
##	60	186.9602	nan	0.1000	1.5573
##	80	160.2296	nan	0.1000	0.5846
##	100	147.3465	nan	0.1000	-0.0290
##	120	140.7966	nan	0.1000	-0.7363
##	140	135.9467	nan	0.1000	0.0404
##	150	133.1098	nan	0.1000	0.0520

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1138.3148	nan	0.1000	161.1864
##	2	1003.7702	nan	0.1000	111.5627
##	3	885.9315	nan	0.1000	88.6904
##	4	787.9267	nan	0.1000	80.9427
##	5	718.3539	nan	0.1000	73.0769
##	6	654.9425	nan	0.1000	66.9388
##	7	593.8499	nan	0.1000	57.3728
##	8	552.5312	nan	0.1000	42.2236
##	9	506.1835	nan	0.1000	32.1548
##	10	470.7422	nan	0.1000	20.3838
##	20	280.8118	nan	0.1000	10.4112
##	40	168.5224	nan	0.1000	1.9261
##	60	138.6952	nan	0.1000	0.5653
##	80	125.8292	nan	0.1000	0.5666
##	100	116.6583	nan	0.1000	-0.0397

##	120	110.5151	nan	0.1000	-0.0843
##	140	104.5755	nan	0.1000	-0.5364
##	150	102.4542	nan	0.1000	-0.0456
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1122.9821	nan	0.1000	178.7064
##	2	977.8391	nan	0.1000	140.7097
##	3	857.8848	nan	0.1000	101.6812
##	4	750.4677	nan	0.1000	97.1855
##	5	662.0855	nan	0.1000	56.7001
##	6	585.8205	nan	0.1000	66.1089
##	7	536.4547	nan	0.1000	49.5552
##	8	486.1739	nan	0.1000	47.3448
##	9	446.9455	nan	0.1000	40.6060
##	10	411.8919	nan	0.1000	34.5107
##	20	234.3615	nan	0.1000	8.2405
##	40	140.8457	nan	0.1000	0.6081
##	60	115.9656	nan	0.1000	-0.3337
##	80	105.0992	nan	0.1000	-0.0829
##	100	99.1904	nan	0.1000	-0.3598
##	120	93.7467	nan	0.1000	-0.6929
##	140	89.7672	nan	0.1000	-0.1951
##	150	88.6174	nan	0.1000	-0.6462
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1023.4582	nan	0.1000	68.3441
##	2	967.9584	nan	0.1000	59.5204
##	3	909.0685	nan	0.1000	63.8819
##	4	854.1288	nan	0.1000	43.4207
##	5	810.2001	nan	0.1000	44.7913
##	6	765.1487	nan	0.1000	46.8162
##	7	725.9762	nan	0.1000	37.0026
##	8	675.6022	nan	0.1000	31.4953
##	9	644.0094	nan	0.1000	28.9816
##	10	616.3479	nan	0.1000	29.9801
##	20	393.2961	nan	0.1000	8.9774
##	40	235.7948	nan	0.1000	3.5970
##	60	179.8041	nan	0.1000	1.5017
##	80	150.9343	nan	0.1000	0.7221
##	100	136.3889	nan	0.1000	-1.4436
##	120	129.0320	nan	0.1000	0.1599
##	140	123.3141	nan	0.1000	0.2362
##	150	121.3434	nan	0.1000	-0.7116
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	982.6173	nan	0.1000	147.4154
##	2	873.1261	nan	0.1000	93.7166
##	3	780.8316	nan	0.1000	83.4734
##	4	705.9900	nan	0.1000	79.7419
##	5	650.7736	nan	0.1000	59.2770
##	6	603.7625	nan	0.1000	46.4562
##	7	557.2159	nan	0.1000	44.0364
##	8	507.2621	nan	0.1000	38.9514
##	9	458.6382	nan	0.1000	36.5356

##	10	425.9856	nan	0.1000	32.0136
##	20	268.7626	nan	0.1000	10.8314
##	40	163.7822	nan	0.1000	1.5170
##	60	129.1078	nan	0.1000	-0.2815
##	80	115.6694	nan	0.1000	-0.5982
##	100	108.3858	nan	0.1000	0.0414
##	120	102.1804	nan	0.1000	0.2199
##	140	97.5815	nan	0.1000	-1.0489
##	150	95.6573	nan	0.1000	-0.4461
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	960.5786	nan	0.1000	154.4063
##	2	843.3030	nan	0.1000	117.0651
##	3	741.9885	nan	0.1000	110.7751
##	4	647.6381	nan	0.1000	82.1691
##	5	577.1404	nan	0.1000	57.7678
##	6	518.5489	nan	0.1000	49.1300
##	7	471.1645	nan	0.1000	41.3249
##	8	436.8741	nan	0.1000	25.9718
##	9	404.6363	nan	0.1000	25.0365
##	10	374.0788	nan	0.1000	25.2631
##	20	221.2741	nan	0.1000	8.9664
##	40	130.5958	nan	0.1000	1.2295
##	60	104.8111	nan	0.1000	-0.9402
##	80	96.0620	nan	0.1000	-0.1090
##	100	89.7521	nan	0.1000	-0.1655
##	120	85.2111	nan	0.1000	-0.5734
##	140	81.8484	nan	0.1000	-0.4011
##	150	79.9122	nan	0.1000	0.0527
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1173.3057	nan	0.1000	53.0337
##	2	1104.1929	nan	0.1000	75.5040
##	3	1034.8149	nan	0.1000	73.8075
##	4	976.3153	nan	0.1000	57.9518
##	5	916.5336	nan	0.1000	55.0927
##	6	871.8483	nan	0.1000	46.3780
##	7	820.0678	nan	0.1000	41.5526
##	8	777.0558	nan	0.1000	52.3215
##	9	733.4139	nan	0.1000	29.6760
##	10	698.0279	nan	0.1000	33.0200
##	20	441.0445	nan	0.1000	16.1338
##	40	264.6720	nan	0.1000	2.5522
##	60	199.0969	nan	0.1000	1.4790
##	80	170.6772	nan	0.1000	0.8804
##	100	156.7461	nan	0.1000	0.2365
##	120	148.7043	nan	0.1000	-0.8564
##	140	142.8518	nan	0.1000	0.1958
##	150	140.7197	nan	0.1000	-0.0893
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1172.7678	nan	0.1000	100.4373
##	2	1038.9700	nan	0.1000	143.6029
##	3	956.5260	nan	0.1000	68.2986

##	4	860.1925	nan	0.1000	92.9443
##	5	776.9312	nan	0.1000	62.7130
##	6	701.8254	nan	0.1000	71.9763
##	7	639.9996	nan	0.1000	59.3219
##	8	586.9620	nan	0.1000	39.7598
##	9	540.6001	nan	0.1000	41.4294
##	10	509.9882	nan	0.1000	31.6544
##	20	310.5632	nan	0.1000	11.7259
##	40	187.8151	nan	0.1000	2.0228
##	60	150.1506	nan	0.1000	-0.0571
##	80	133.2271	nan	0.1000	0.3945
##	100	122.1332	nan	0.1000	0.6666
##	120	117.1130	nan	0.1000	-0.1163
##	140	112.4926	nan	0.1000	-0.3512
##	150	111.4169	nan	0.1000	-0.2453

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1098.0963	nan	0.1000	136.6717
##	2	1001.6572	nan	0.1000	110.3699
##	3	876.4455	nan	0.1000	116.3827
##	4	768.7673	nan	0.1000	89.0480
##	5	694.6607	nan	0.1000	69.1399
##	6	628.2615	nan	0.1000	65.2577
##	7	566.9362	nan	0.1000	58.2404
##	8	523.7986	nan	0.1000	44.3024
##	9	476.6739	nan	0.1000	34.2358
##	10	441.0850	nan	0.1000	36.1024
##	20	252.1497	nan	0.1000	9.0238
##	40	151.3068	nan	0.1000	1.0013
##	60	126.2079	nan	0.1000	0.3360
##	80	115.3667	nan	0.1000	-0.8247
##	100	106.5691	nan	0.1000	0.0017
##	120	101.4904	nan	0.1000	0.1990
##	140	97.1520	nan	0.1000	-0.1910
##	150	95.2360	nan	0.1000	-0.2821

##

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1201.4410	nan	0.1000	106.1190
##	2	1112.0699	nan	0.1000	89.6509
##	3	1043.7033	nan	0.1000	68.3250
##	4	987.0400	nan	0.1000	55.9010
##	5	924.1874	nan	0.1000	53.4938
##	6	877.9404	nan	0.1000	48.3207
##	7	823.8676	nan	0.1000	46.5479
##	8	773.5133	nan	0.1000	50.9513
##	9	726.6435	nan	0.1000	41.1811
##	10	687.2881	nan	0.1000	32.1867
##	20	428.2893	nan	0.1000	16.4960
##	40	247.9174	nan	0.1000	3.8608
##	60	183.2643	nan	0.1000	1.8177
##	80	153.9239	nan	0.1000	0.7093
##	100	139.5839	nan	0.1000	-0.0992
##	120	130.6775	nan	0.1000	-0.0728
##	140	124.8761	nan	0.1000	-0.5797

```

##      150      122.2978          nan      0.1000      0.0473
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1142.7927          nan      0.1000   151.0889
##      2      1014.7561          nan      0.1000   124.4026
##      3       908.0770          nan      0.1000    88.4686
##      4       812.2367          nan      0.1000    92.8222
##      5       754.5227          nan      0.1000    55.9591
##      6       686.4518          nan      0.1000    64.5867
##      7       626.9016          nan      0.1000    60.8306
##      8       568.7257          nan      0.1000    39.6262
##      9       533.7047          nan      0.1000    38.9119
##     10       490.3844          nan      0.1000    25.6961
##     20       280.8341          nan      0.1000    11.5409
##     40       162.4504          nan      0.1000     2.3763
##     60       125.9654          nan      0.1000    -0.4606
##     80       113.8271          nan      0.1000     0.2730
##    100       107.0324          nan      0.1000    -0.2279
##    120       100.7149          nan      0.1000    -0.0411
##    140        97.1166          nan      0.1000    -0.1578
##    150        95.0111          nan      0.1000    -0.1561
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1156.1045          nan      0.1000   168.0392
##      2       991.6610          nan      0.1000   178.6425
##      3       873.8349          nan      0.1000   122.0701
##      4       775.5410          nan      0.1000    89.8189
##      5       685.0005          nan      0.1000    82.8261
##      6       624.5754          nan      0.1000    57.5314
##      7       558.5024          nan      0.1000    43.5408
##      8       503.7201          nan      0.1000    45.5906
##      9       461.8048          nan      0.1000    49.5252
##     10       429.3099          nan      0.1000    32.1101
##     20       226.6858          nan      0.1000    10.5130
##     40       129.7483          nan      0.1000     1.1179
##     60       106.9612          nan      0.1000     0.1356
##     80        96.7065          nan      0.1000     0.0488
##    100        91.2641          nan      0.1000     0.0641
##    120        87.7454          nan      0.1000    -0.2613
##    140        84.6140          nan      0.1000     0.0046
##    150        83.3537          nan      0.1000    -0.1391
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1035.7901          nan      0.1000    58.7296
##      2       965.8432          nan      0.1000    50.9052
##      3       906.2053          nan      0.1000    65.9404
##      4       842.6515          nan      0.1000    62.7205
##      5       795.6941          nan      0.1000    43.9271
##      6       751.0854          nan      0.1000    44.0143
##      7       705.3684          nan      0.1000    43.3203
##      8       662.9367          nan      0.1000    29.9252
##      9       628.5286          nan      0.1000    23.8324
##     10       591.7874          nan      0.1000    36.1055
##     20       380.0726          nan      0.1000    13.3722

```


##	40	215.3355	nan	0.1000	3.7943
##	60	164.8433	nan	0.1000	1.1424
##	80	140.1305	nan	0.1000	0.1823
##	100	127.3285	nan	0.1000	0.3753
##	120	118.5926	nan	0.1000	0.1254
##	140	112.7833	nan	0.1000	0.2753
##	150	110.4796	nan	0.1000	0.1521
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	984.4417	nan	0.1000	104.8746
##	2	876.2773	nan	0.1000	106.6737
##	3	801.2080	nan	0.1000	77.5352
##	4	712.2964	nan	0.1000	79.4696
##	5	643.4845	nan	0.1000	42.7919
##	6	579.3777	nan	0.1000	58.4736
##	7	525.9397	nan	0.1000	40.6531
##	8	478.4460	nan	0.1000	44.2109
##	9	448.4120	nan	0.1000	31.1903
##	10	418.4793	nan	0.1000	30.5114
##	20	248.7991	nan	0.1000	6.4246
##	40	150.1354	nan	0.1000	0.9824
##	60	119.8076	nan	0.1000	0.0967
##	80	107.4000	nan	0.1000	-0.7326
##	100	99.6906	nan	0.1000	-0.2750
##	120	94.1053	nan	0.1000	0.2156
##	140	91.0177	nan	0.1000	-0.1580
##	150	89.8213	nan	0.1000	0.1896
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	951.9572	nan	0.1000	128.8859
##	2	842.2726	nan	0.1000	117.5267
##	3	744.2457	nan	0.1000	88.7939
##	4	659.7421	nan	0.1000	79.6048
##	5	595.3616	nan	0.1000	65.9342
##	6	547.8268	nan	0.1000	49.5625
##	7	501.8491	nan	0.1000	45.1066
##	8	453.5627	nan	0.1000	53.1859
##	9	408.1017	nan	0.1000	38.3713
##	10	376.2601	nan	0.1000	26.7730
##	20	209.4982	nan	0.1000	8.3822
##	40	123.2374	nan	0.1000	2.6032
##	60	97.5851	nan	0.1000	0.9637
##	80	87.1513	nan	0.1000	0.1307
##	100	80.4227	nan	0.1000	-0.0476
##	120	75.7581	nan	0.1000	-0.0517
##	140	72.9355	nan	0.1000	-0.2326
##	150	71.9447	nan	0.1000	-0.1492
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1201.6287	nan	0.1000	90.9115
##	2	1119.7142	nan	0.1000	83.7937
##	3	1040.6954	nan	0.1000	87.8484
##	4	985.5103	nan	0.1000	58.0410
##	5	923.7542	nan	0.1000	56.8231

##	6	870.1986	nan	0.1000	47.7431
##	7	813.9796	nan	0.1000	59.0307
##	8	771.3084	nan	0.1000	45.1861
##	9	734.6957	nan	0.1000	38.0202
##	10	691.2948	nan	0.1000	45.7223
##	20	426.1964	nan	0.1000	16.5375
##	40	247.8649	nan	0.1000	3.8290
##	60	185.7757	nan	0.1000	1.6255
##	80	156.0575	nan	0.1000	0.7319
##	100	142.2254	nan	0.1000	-0.8207
##	120	134.5285	nan	0.1000	0.2423
##	140	128.5601	nan	0.1000	0.1166
##	150	126.1168	nan	0.1000	0.2170
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1157.7912	nan	0.1000	143.5886
##	2	1022.0738	nan	0.1000	138.6245
##	3	905.0947	nan	0.1000	97.2034
##	4	809.8742	nan	0.1000	90.6408
##	5	728.5079	nan	0.1000	73.5228
##	6	655.3187	nan	0.1000	60.1245
##	7	604.2588	nan	0.1000	35.3958
##	8	556.4355	nan	0.1000	45.5072
##	9	511.7440	nan	0.1000	45.0474
##	10	476.4448	nan	0.1000	27.4272
##	20	272.3418	nan	0.1000	10.5469
##	40	165.1427	nan	0.1000	3.5678
##	60	133.0993	nan	0.1000	-0.6801
##	80	121.5918	nan	0.1000	-0.0320
##	100	114.5075	nan	0.1000	0.0147
##	120	109.9636	nan	0.1000	-0.1104
##	140	104.0401	nan	0.1000	0.0509
##	150	102.6555	nan	0.1000	0.0085
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1133.5197	nan	0.1000	172.7048
##	2	987.1972	nan	0.1000	143.5850
##	3	872.7645	nan	0.1000	118.7208
##	4	764.0503	nan	0.1000	112.4616
##	5	665.5178	nan	0.1000	92.4002
##	6	592.2673	nan	0.1000	80.0497
##	7	536.5960	nan	0.1000	64.6191
##	8	485.7798	nan	0.1000	47.8213
##	9	445.0319	nan	0.1000	41.1958
##	10	411.4633	nan	0.1000	31.3130
##	20	232.8683	nan	0.1000	9.9203
##	40	137.9395	nan	0.1000	1.1614
##	60	115.5033	nan	0.1000	-0.2852
##	80	103.5464	nan	0.1000	-0.1662
##	100	96.6148	nan	0.1000	0.3704
##	120	92.0743	nan	0.1000	-0.4650
##	140	87.9470	nan	0.1000	-0.5296
##	150	86.5364	nan	0.1000	0.1443
##					

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1072.0599	nan	0.1000	83.9007
##	2	995.1452	nan	0.1000	75.6876
##	3	930.6467	nan	0.1000	60.1760
##	4	882.8934	nan	0.1000	36.4684
##	5	826.8277	nan	0.1000	57.7570
##	6	777.8295	nan	0.1000	45.0836
##	7	738.5092	nan	0.1000	38.4305
##	8	694.2602	nan	0.1000	35.9653
##	9	657.8653	nan	0.1000	14.7778
##	10	614.2662	nan	0.1000	45.3666
##	20	381.4929	nan	0.1000	13.7471
##	40	225.4717	nan	0.1000	2.9268
##	60	169.7447	nan	0.1000	1.1790
##	80	143.4027	nan	0.1000	0.1550
##	100	130.2860	nan	0.1000	0.3389
##	120	121.7222	nan	0.1000	0.2099
##	140	115.8403	nan	0.1000	-0.0558
##	150	113.2317	nan	0.1000	0.2352
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1019.1468	nan	0.1000	93.3516
##	2	923.9510	nan	0.1000	96.2785
##	3	817.5182	nan	0.1000	95.1558
##	4	746.5360	nan	0.1000	67.5560
##	5	669.3585	nan	0.1000	64.8072
##	6	605.2731	nan	0.1000	61.6085
##	7	552.7118	nan	0.1000	31.8123
##	8	509.3889	nan	0.1000	34.3121
##	9	473.4643	nan	0.1000	37.5760
##	10	434.2131	nan	0.1000	32.7532
##	20	250.3380	nan	0.1000	9.0373
##	40	149.8441	nan	0.1000	2.3425
##	60	121.5830	nan	0.1000	0.4301
##	80	107.6926	nan	0.1000	-0.6998
##	100	100.4119	nan	0.1000	0.0089
##	120	96.2707	nan	0.1000	-0.3374
##	140	90.8735	nan	0.1000	-0.0021
##	150	90.0521	nan	0.1000	-0.2075
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	997.9016	nan	0.1000	156.2244
##	2	874.7602	nan	0.1000	136.0710
##	3	766.1950	nan	0.1000	88.8837
##	4	673.5437	nan	0.1000	86.0353
##	5	608.7514	nan	0.1000	68.2482
##	6	548.6146	nan	0.1000	57.5847
##	7	501.5420	nan	0.1000	49.3042
##	8	456.5943	nan	0.1000	40.2833
##	9	417.2413	nan	0.1000	37.7478
##	10	387.6704	nan	0.1000	27.5050
##	20	216.7646	nan	0.1000	9.6328
##	40	132.6365	nan	0.1000	1.6443
##	60	107.6826	nan	0.1000	0.0679

##	80	94.8867	nan	0.1000	0.1861
##	100	86.7963	nan	0.1000	0.4310
##	120	80.0393	nan	0.1000	0.0909
##	140	76.8072	nan	0.1000	-0.1429
##	150	75.2394	nan	0.1000	-0.0434
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1080.8189	nan	0.1000	76.6571
##	2	1003.9710	nan	0.1000	66.6505
##	3	942.4747	nan	0.1000	58.5403
##	4	886.9356	nan	0.1000	55.1046
##	5	839.6075	nan	0.1000	47.9903
##	6	781.6543	nan	0.1000	58.4151
##	7	735.4703	nan	0.1000	39.6462
##	8	691.4888	nan	0.1000	42.5628
##	9	654.6201	nan	0.1000	36.2675
##	10	622.4645	nan	0.1000	28.5020
##	20	401.3246	nan	0.1000	8.4799
##	40	234.4538	nan	0.1000	3.2475
##	60	175.0800	nan	0.1000	0.7847
##	80	150.1060	nan	0.1000	0.2590
##	100	137.6066	nan	0.1000	-0.6271
##	120	128.8004	nan	0.1000	0.3464
##	140	123.6679	nan	0.1000	-0.3460
##	150	121.3646	nan	0.1000	0.1333
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1032.4696	nan	0.1000	155.1064
##	2	945.9092	nan	0.1000	107.4908
##	3	833.1954	nan	0.1000	98.7479
##	4	748.2650	nan	0.1000	94.3671
##	5	680.7824	nan	0.1000	70.5340
##	6	622.6394	nan	0.1000	63.3733
##	7	568.8359	nan	0.1000	61.2356
##	8	521.6860	nan	0.1000	39.0593
##	9	482.9395	nan	0.1000	44.5384
##	10	444.5591	nan	0.1000	37.3012
##	20	257.2536	nan	0.1000	9.2497
##	40	156.4679	nan	0.1000	1.0572
##	60	124.0190	nan	0.1000	0.7324
##	80	110.4133	nan	0.1000	0.6957
##	100	101.9957	nan	0.1000	0.7617
##	120	95.8166	nan	0.1000	0.0013
##	140	93.3269	nan	0.1000	-0.2322
##	150	91.7920	nan	0.1000	-0.2663
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	996.8675	nan	0.1000	155.6846
##	2	868.4262	nan	0.1000	125.6263
##	3	766.1887	nan	0.1000	104.5108
##	4	675.4711	nan	0.1000	58.7944
##	5	600.3957	nan	0.1000	68.9915
##	6	533.0195	nan	0.1000	66.1509
##	7	477.9536	nan	0.1000	37.5246

##	8	440.1525	nan	0.1000	41.8343
##	9	402.6834	nan	0.1000	37.8598
##	10	374.8409	nan	0.1000	31.5991
##	20	210.2067	nan	0.1000	8.9055
##	40	128.6086	nan	0.1000	-1.2271
##	60	109.0801	nan	0.1000	-0.2201
##	80	98.2993	nan	0.1000	-0.1650
##	100	92.4266	nan	0.1000	0.0777
##	120	86.3903	nan	0.1000	-0.2130
##	140	82.5415	nan	0.1000	-0.2469
##	150	81.0329	nan	0.1000	-0.4306
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1222.9937	nan	0.1000	95.5819
##	2	1145.0232	nan	0.1000	77.1941
##	3	1077.9556	nan	0.1000	67.2316
##	4	1009.5096	nan	0.1000	67.9725
##	5	945.3913	nan	0.1000	58.7099
##	6	894.1958	nan	0.1000	54.7122
##	7	841.2722	nan	0.1000	47.8715
##	8	798.8638	nan	0.1000	30.1145
##	9	756.3627	nan	0.1000	37.6571
##	10	714.7252	nan	0.1000	42.2972
##	20	434.9873	nan	0.1000	17.1626
##	40	248.3850	nan	0.1000	4.5311
##	60	189.8807	nan	0.1000	1.6512
##	80	163.0919	nan	0.1000	0.6497
##	100	150.7166	nan	0.1000	-0.4042
##	120	142.7196	nan	0.1000	0.2574
##	140	136.4626	nan	0.1000	-0.1399
##	150	133.4146	nan	0.1000	-0.1580
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1160.9798	nan	0.1000	137.7709
##	2	1025.0834	nan	0.1000	130.3225
##	3	926.5422	nan	0.1000	101.9054
##	4	835.0766	nan	0.1000	95.1952
##	5	755.7430	nan	0.1000	67.7841
##	6	696.4705	nan	0.1000	59.9653
##	7	639.3003	nan	0.1000	52.7958
##	8	578.5829	nan	0.1000	60.5109
##	9	525.8460	nan	0.1000	45.8725
##	10	483.9510	nan	0.1000	44.4311
##	20	277.9536	nan	0.1000	9.0038
##	40	169.2679	nan	0.1000	0.9201
##	60	139.9209	nan	0.1000	-0.2076
##	80	126.1292	nan	0.1000	-0.7480
##	100	117.6557	nan	0.1000	0.2180
##	120	110.9438	nan	0.1000	-0.6432
##	140	106.7862	nan	0.1000	0.0718
##	150	104.1988	nan	0.1000	-0.0185
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1122.3496	nan	0.1000	214.4010

##	2	977.3170	nan	0.1000	141.5716
##	3	881.4311	nan	0.1000	89.5670
##	4	802.2643	nan	0.1000	73.8740
##	5	706.5898	nan	0.1000	103.1031
##	6	638.2308	nan	0.1000	74.1872
##	7	570.3551	nan	0.1000	51.2075
##	8	521.9023	nan	0.1000	47.8098
##	9	471.4959	nan	0.1000	40.5367
##	10	430.4478	nan	0.1000	38.1211
##	20	234.7109	nan	0.1000	9.8324
##	40	139.1237	nan	0.1000	1.2925
##	60	115.0282	nan	0.1000	0.5032
##	80	103.9629	nan	0.1000	-0.3077
##	100	96.6829	nan	0.1000	-0.2997
##	120	91.8252	nan	0.1000	0.0046
##	140	88.5128	nan	0.1000	-0.2414
##	150	86.9650	nan	0.1000	-0.2051

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1087.8570	nan	0.1000	76.6095
##	2	1013.8603	nan	0.1000	62.7495
##	3	957.5153	nan	0.1000	55.9620
##	4	897.3468	nan	0.1000	61.6687
##	5	836.9038	nan	0.1000	49.7647
##	6	787.3263	nan	0.1000	47.1981
##	7	744.5487	nan	0.1000	36.7272
##	8	699.7224	nan	0.1000	44.5974
##	9	664.5324	nan	0.1000	34.5989
##	10	632.6207	nan	0.1000	32.1634
##	20	404.4873	nan	0.1000	14.4809
##	40	243.6616	nan	0.1000	4.6382
##	60	187.3818	nan	0.1000	1.8495
##	80	161.7907	nan	0.1000	0.4776
##	100	149.1679	nan	0.1000	0.4633
##	120	141.2548	nan	0.1000	-0.4979
##	140	135.7785	nan	0.1000	-0.1640
##	150	133.6173	nan	0.1000	-0.5843

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1050.9876	nan	0.1000	112.8077
##	2	918.7068	nan	0.1000	129.8786
##	3	808.9025	nan	0.1000	98.3689
##	4	728.8537	nan	0.1000	78.2674
##	5	676.8000	nan	0.1000	55.3235
##	6	612.2399	nan	0.1000	54.5006
##	7	559.6181	nan	0.1000	50.6218
##	8	515.0434	nan	0.1000	40.8312
##	9	473.0174	nan	0.1000	36.9899
##	10	433.1366	nan	0.1000	33.8391
##	20	272.3469	nan	0.1000	9.4617
##	40	170.5693	nan	0.1000	2.2894
##	60	143.1123	nan	0.1000	0.3532
##	80	126.9598	nan	0.1000	0.1493
##	100	118.7461	nan	0.1000	-0.8097

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##      120      113.7615      nan      0.1000     -0.1565
##      140      109.6232      nan      0.1000     -0.4357
##      150      107.6110      nan      0.1000      0.4073
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1002.9726      nan      0.1000   130.6249
##      2       874.5304      nan      0.1000   106.8671
##      3       784.4297      nan      0.1000    92.7181
##      4       697.1941      nan      0.1000    88.3228
##      5       626.8668      nan      0.1000    74.1835
##      6       573.3715      nan      0.1000    54.5405
##      7       514.4491      nan      0.1000    67.7620
##      8       464.2703      nan      0.1000    44.0149
##      9       427.8894      nan      0.1000    36.7926
##     10       396.0215      nan      0.1000    30.5870
##     20       228.4352      nan      0.1000     7.3448
##     40       143.6600      nan      0.1000     0.7433
##     60       121.9373      nan      0.1000     0.2120
##     80       111.5679      nan      0.1000    -0.0402
##    100       103.4978      nan      0.1000     0.1086
##    120        97.0031      nan      0.1000    -0.1387
##    140        92.7187      nan      0.1000    -0.0460
##    150        90.3598      nan      0.1000    -0.0037
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1090.7159      nan      0.1000    84.5760
##      2      1007.2289      nan      0.1000    74.1523
##      3       946.8949      nan      0.1000    60.3607
##      4       887.4432      nan      0.1000    59.7682
##      5       819.2801      nan      0.1000    70.1452
##      6       771.3496      nan      0.1000    47.5261
##      7       720.1783      nan      0.1000    53.3296
##      8       681.0785      nan      0.1000    40.7998
##      9       643.3239      nan      0.1000    30.1461
##     10       606.5380      nan      0.1000    36.8915
##     20       377.9292      nan      0.1000    12.3079
##     40       217.9400      nan      0.1000     3.2746
##     60       166.1905      nan      0.1000     1.2989
##     80       141.4818      nan      0.1000     0.7027
##    100       128.4313      nan      0.1000     0.1308
##    120       119.5264      nan      0.1000    -0.1771
##    140       113.3529      nan      0.1000    -0.0105
##    150       110.9493      nan      0.1000     0.0478
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1033.9575      nan      0.1000   132.8711
##      2       923.8711      nan      0.1000    96.0836
##      3       819.2553      nan      0.1000    97.9891
##      4       725.4915      nan      0.1000    69.5400
##      5       664.0824      nan      0.1000    62.4558
##      6       606.7370      nan      0.1000    58.9446
##      7       551.1123      nan      0.1000    55.5651
##      8       507.4330      nan      0.1000    44.5145
##      9       467.8108      nan      0.1000    36.8782

```

##	10	434.9003	nan	0.1000	32.6813
##	20	248.0589	nan	0.1000	12.5799
##	40	146.4711	nan	0.1000	1.8295
##	60	114.5459	nan	0.1000	0.7619
##	80	102.0801	nan	0.1000	0.0312
##	100	94.9411	nan	0.1000	0.5516
##	120	89.8323	nan	0.1000	-0.5575
##	140	86.1568	nan	0.1000	0.0435
##	150	84.8165	nan	0.1000	-0.1532
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1026.8621	nan	0.1000	142.8041
##	2	921.2654	nan	0.1000	108.3263
##	3	795.7200	nan	0.1000	106.8504
##	4	696.4259	nan	0.1000	106.4723
##	5	617.6322	nan	0.1000	72.1743
##	6	552.1506	nan	0.1000	65.2443
##	7	494.2715	nan	0.1000	50.5568
##	8	445.6477	nan	0.1000	46.1631
##	9	408.5658	nan	0.1000	39.8022
##	10	374.8992	nan	0.1000	30.8978
##	20	208.5083	nan	0.1000	6.9347
##	40	122.2220	nan	0.1000	2.0757
##	60	99.9701	nan	0.1000	0.5794
##	80	89.6386	nan	0.1000	0.2813
##	100	83.1020	nan	0.1000	0.1541
##	120	78.2733	nan	0.1000	-0.2972
##	140	75.3554	nan	0.1000	-0.1857
##	150	73.6533	nan	0.1000	-0.1366
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1118.8115	nan	0.1000	73.4980
##	2	1050.4907	nan	0.1000	66.5460
##	3	978.8147	nan	0.1000	67.2210
##	4	917.3993	nan	0.1000	54.8446
##	5	857.4446	nan	0.1000	58.4640
##	6	810.0807	nan	0.1000	42.8145
##	7	764.8915	nan	0.1000	45.2697
##	8	717.6212	nan	0.1000	48.6299
##	9	678.6333	nan	0.1000	38.5026
##	10	642.8427	nan	0.1000	36.2583
##	20	400.0696	nan	0.1000	15.5698
##	40	232.2462	nan	0.1000	3.8881
##	60	174.7613	nan	0.1000	1.5945
##	80	148.1350	nan	0.1000	0.5106
##	100	135.2250	nan	0.1000	-0.9972
##	120	126.5791	nan	0.1000	0.3248
##	140	119.9147	nan	0.1000	0.1349
##	150	117.7438	nan	0.1000	0.0888
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1062.1767	nan	0.1000	153.5348
##	2	936.4652	nan	0.1000	115.1786
##	3	829.3559	nan	0.1000	85.4330

##	4	757.2288	nan	0.1000	79.2473
##	5	700.1942	nan	0.1000	61.7006
##	6	636.2802	nan	0.1000	66.7213
##	7	580.7349	nan	0.1000	56.5523
##	8	534.5268	nan	0.1000	46.2899
##	9	498.0534	nan	0.1000	35.5052
##	10	458.5808	nan	0.1000	36.8635
##	20	272.1561	nan	0.1000	10.5912
##	40	163.0717	nan	0.1000	1.0363
##	60	126.9545	nan	0.1000	1.4551
##	80	113.7320	nan	0.1000	-0.0505
##	100	106.1635	nan	0.1000	-0.7638
##	120	100.3572	nan	0.1000	0.5649
##	140	95.7085	nan	0.1000	-0.1186
##	150	93.0152	nan	0.1000	0.3598
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1041.6123	nan	0.1000	154.3690
##	2	914.0338	nan	0.1000	142.5481
##	3	801.1364	nan	0.1000	110.1066
##	4	733.4377	nan	0.1000	74.4354
##	5	652.7770	nan	0.1000	91.6788
##	6	578.6338	nan	0.1000	59.0238
##	7	526.6763	nan	0.1000	49.7396
##	8	479.9661	nan	0.1000	42.8199
##	9	438.8906	nan	0.1000	41.4579
##	10	404.4224	nan	0.1000	35.9367
##	20	229.6770	nan	0.1000	9.7288
##	40	135.5758	nan	0.1000	0.8887
##	60	109.8398	nan	0.1000	-0.2751
##	80	97.5181	nan	0.1000	-0.1177
##	100	91.6485	nan	0.1000	-0.3932
##	120	85.1088	nan	0.1000	-0.1882
##	140	80.8995	nan	0.1000	-0.4833
##	150	78.4903	nan	0.1000	-0.2523
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1119.8773	nan	0.1000	94.3423
##	2	1036.5403	nan	0.1000	84.1755
##	3	967.3030	nan	0.1000	68.0679
##	4	900.4328	nan	0.1000	51.4887
##	5	840.7201	nan	0.1000	57.2646
##	6	789.3604	nan	0.1000	53.1029
##	7	737.3305	nan	0.1000	48.8495
##	8	694.6946	nan	0.1000	46.8030
##	9	651.7255	nan	0.1000	37.3301
##	10	615.7655	nan	0.1000	37.5008
##	20	383.5992	nan	0.1000	11.8737
##	40	228.2604	nan	0.1000	2.4331
##	60	178.3682	nan	0.1000	1.1799
##	80	152.9885	nan	0.1000	0.4506
##	100	139.4953	nan	0.1000	0.1046
##	120	131.4240	nan	0.1000	0.2497
##	140	125.9992	nan	0.1000	-0.6988

```

##      150      123.9954          nan      0.1000      0.1049
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1056.6943          nan      0.1000   152.1538
##      2       922.9485          nan      0.1000   127.0554
##      3       812.9224          nan      0.1000   112.0540
##      4       735.7370          nan      0.1000    77.1514
##      5       660.4394          nan      0.1000    77.7973
##      6       594.5295          nan      0.1000    54.8795
##      7       538.1276          nan      0.1000    50.0000
##      8       496.2987          nan      0.1000    41.3740
##      9       463.3502          nan      0.1000    29.0453
##     10       423.1220          nan      0.1000    32.9575
##     20       254.4723          nan      0.1000    10.8771
##     40       157.7539          nan      0.1000     3.1032
##     60       129.3769          nan      0.1000     0.6312
##     80       115.9494          nan      0.1000     0.5027
##    100       109.1067          nan      0.1000    -0.1916
##    120       103.6767          nan      0.1000    -0.1559
##    140       100.4952          nan      0.1000    -0.0890
##    150        98.7492          nan      0.1000    -0.2977
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1048.6867          nan      0.1000   169.7189
##      2       898.1875          nan      0.1000   132.8233
##      3       779.0176          nan      0.1000   115.3169
##      4       688.4945          nan      0.1000    76.3549
##      5       601.5160          nan      0.1000    76.0706
##      6       535.5268          nan      0.1000    49.9857
##      7       491.4052          nan      0.1000    44.9608
##      8       456.8617          nan      0.1000    33.7188
##      9       427.1252          nan      0.1000    27.1048
##     10       396.9068          nan      0.1000    34.3282
##     20       227.6345          nan      0.1000     7.7327
##     40       139.1342          nan      0.1000     2.4652
##     60       116.2420          nan      0.1000     0.1765
##     80       104.1271          nan      0.1000    -0.5082
##    100        96.7440          nan      0.1000    -0.1467
##    120        92.1058          nan      0.1000    -0.1483
##    140        87.7991          nan      0.1000     0.0628
##    150        86.2169          nan      0.1000    -0.0998
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1189.0101          nan      0.1000    86.6900
##      2      1114.6414          nan      0.1000    71.3947
##      3      1047.4181          nan      0.1000    57.8647
##      4       977.1987          nan      0.1000    66.0326
##      5       907.3896          nan      0.1000    67.8968
##      6       851.5658          nan      0.1000    54.0091
##      7       800.4883          nan      0.1000    43.4781
##      8       753.3136          nan      0.1000    40.4555
##      9       709.8473          nan      0.1000    39.5049
##     10       668.1243          nan      0.1000    37.6030
##     20       423.7775          nan      0.1000    18.7096

```

##	40	248.4562	nan	0.1000	3.1495
##	60	192.5803	nan	0.1000	1.5092
##	80	169.4225	nan	0.1000	0.4061
##	100	156.9656	nan	0.1000	0.2653
##	120	149.3569	nan	0.1000	0.0663
##	140	145.4897	nan	0.1000	-0.1711
##	150	143.3884	nan	0.1000	-0.4252
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1129.1701	nan	0.1000	153.7598
##	2	996.7433	nan	0.1000	97.6723
##	3	909.4208	nan	0.1000	88.9440
##	4	821.4093	nan	0.1000	82.8959
##	5	755.1422	nan	0.1000	67.1834
##	6	678.6369	nan	0.1000	67.5101
##	7	618.9197	nan	0.1000	56.0494
##	8	569.8699	nan	0.1000	52.9390
##	9	533.0902	nan	0.1000	36.0775
##	10	490.8521	nan	0.1000	41.1104
##	20	279.8185	nan	0.1000	8.8339
##	40	181.4204	nan	0.1000	1.5549
##	60	150.0737	nan	0.1000	1.6910
##	80	137.2608	nan	0.1000	0.7824
##	100	129.9877	nan	0.1000	-0.4570
##	120	123.8775	nan	0.1000	-0.3129
##	140	118.0144	nan	0.1000	-0.7412
##	150	115.0476	nan	0.1000	-0.3169
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1106.8309	nan	0.1000	137.4255
##	2	983.8373	nan	0.1000	111.8664
##	3	876.5559	nan	0.1000	110.1780
##	4	766.9048	nan	0.1000	109.8097
##	5	682.1685	nan	0.1000	80.4737
##	6	622.5092	nan	0.1000	66.1048
##	7	550.4590	nan	0.1000	62.3545
##	8	498.6682	nan	0.1000	46.0724
##	9	457.5230	nan	0.1000	46.5795
##	10	421.0962	nan	0.1000	37.5097
##	20	236.7550	nan	0.1000	9.0588
##	40	151.0315	nan	0.1000	0.7579
##	60	127.0778	nan	0.1000	-0.1092
##	80	116.4214	nan	0.1000	-0.0738
##	100	107.3310	nan	0.1000	-0.5775
##	120	102.4719	nan	0.1000	-0.3353
##	140	96.2524	nan	0.1000	-0.0157
##	150	94.1987	nan	0.1000	0.1255
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1283.6954	nan	0.1000	73.1715
##	2	1182.6265	nan	0.1000	84.5721
##	3	1092.3751	nan	0.1000	87.0078
##	4	1020.7277	nan	0.1000	66.1521
##	5	954.5623	nan	0.1000	65.1395

##	6	892.8784	nan	0.1000	38.6726
##	7	839.0687	nan	0.1000	52.9767
##	8	785.7557	nan	0.1000	57.1875
##	9	743.5452	nan	0.1000	39.0069
##	10	695.9698	nan	0.1000	48.5670
##	20	422.0979	nan	0.1000	15.6932
##	40	243.9402	nan	0.1000	4.0555
##	60	186.8147	nan	0.1000	0.8337
##	80	161.0412	nan	0.1000	0.2119
##	100	147.5159	nan	0.1000	0.2091
##	120	138.3987	nan	0.1000	0.0832
##	140	131.6827	nan	0.1000	0.1823
##	150	128.7122	nan	0.1000	-0.0164
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1207.7910	nan	0.1000	180.1243
##	2	1063.2926	nan	0.1000	153.4980
##	3	965.0687	nan	0.1000	85.7630
##	4	864.6020	nan	0.1000	91.2689
##	5	773.1352	nan	0.1000	82.4426
##	6	700.6816	nan	0.1000	69.2203
##	7	631.2889	nan	0.1000	68.3590
##	8	573.0636	nan	0.1000	51.0187
##	9	528.0273	nan	0.1000	45.9456
##	10	476.1582	nan	0.1000	44.3333
##	20	278.1279	nan	0.1000	9.1232
##	40	169.9533	nan	0.1000	1.3212
##	60	137.5625	nan	0.1000	0.8528
##	80	123.0038	nan	0.1000	-0.2596
##	100	112.5488	nan	0.1000	0.2349
##	120	105.2482	nan	0.1000	-0.0700
##	140	101.6530	nan	0.1000	-0.3782
##	150	100.3467	nan	0.1000	-0.7668
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1187.7043	nan	0.1000	214.3587
##	2	1020.9794	nan	0.1000	181.5930
##	3	893.9061	nan	0.1000	127.2004
##	4	778.8480	nan	0.1000	120.6036
##	5	695.0268	nan	0.1000	74.5797
##	6	629.3644	nan	0.1000	77.4514
##	7	561.2605	nan	0.1000	67.4979
##	8	510.4674	nan	0.1000	53.8901
##	9	459.3352	nan	0.1000	37.7979
##	10	416.3479	nan	0.1000	41.0710
##	20	227.3097	nan	0.1000	6.8951
##	40	136.2936	nan	0.1000	1.5643
##	60	115.3820	nan	0.1000	0.0036
##	80	103.7335	nan	0.1000	-0.0722
##	100	96.6125	nan	0.1000	0.0211
##	120	92.4467	nan	0.1000	-0.1759
##	140	88.4003	nan	0.1000	-0.2170
##	150	86.1169	nan	0.1000	-0.0322
##					

##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1210.0056	nan	0.1000	77.0098
##	2	1131.9875	nan	0.1000	77.4601
##	3	1050.1754	nan	0.1000	78.9251
##	4	985.7537	nan	0.1000	63.0267
##	5	929.2235	nan	0.1000	51.8651
##	6	868.4312	nan	0.1000	66.4035
##	7	813.4592	nan	0.1000	52.3478
##	8	772.3243	nan	0.1000	43.1009
##	9	738.1129	nan	0.1000	33.0682
##	10	699.9654	nan	0.1000	40.1154
##	20	434.8572	nan	0.1000	14.7241
##	40	252.9659	nan	0.1000	4.0820
##	60	190.3498	nan	0.1000	2.0213
##	80	162.8234	nan	0.1000	0.6628
##	100	150.7828	nan	0.1000	0.5502
##	120	142.9481	nan	0.1000	0.0537
##	140	137.5727	nan	0.1000	-0.5494
##	150	135.5533	nan	0.1000	0.0043
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1144.5054	nan	0.1000	171.3193
##	2	1026.1129	nan	0.1000	122.7750
##	3	929.5790	nan	0.1000	92.9228
##	4	835.4790	nan	0.1000	75.4115
##	5	752.1271	nan	0.1000	78.2724
##	6	683.9247	nan	0.1000	71.9112
##	7	620.0264	nan	0.1000	54.5971
##	8	565.8734	nan	0.1000	40.3096
##	9	516.9392	nan	0.1000	38.0251
##	10	484.2238	nan	0.1000	36.4699
##	20	292.2535	nan	0.1000	10.8410
##	40	165.6017	nan	0.1000	2.9954
##	60	134.2776	nan	0.1000	0.5575
##	80	120.1004	nan	0.1000	0.0713
##	100	111.6263	nan	0.1000	-1.0180
##	120	105.9832	nan	0.1000	-0.2969
##	140	102.5158	nan	0.1000	0.0533
##	150	100.7794	nan	0.1000	-0.2743
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	1	1163.8655	nan	0.1000	156.2903
##	2	1016.5769	nan	0.1000	143.8181
##	3	882.9052	nan	0.1000	133.6532
##	4	772.9999	nan	0.1000	83.3790
##	5	688.3052	nan	0.1000	60.3592
##	6	613.6813	nan	0.1000	68.7803
##	7	550.7211	nan	0.1000	64.4275
##	8	497.8653	nan	0.1000	47.6486
##	9	458.5626	nan	0.1000	40.8449
##	10	420.3436	nan	0.1000	30.5602
##	20	229.2189	nan	0.1000	15.5717
##	40	134.3119	nan	0.1000	0.8417
##	60	112.1139	nan	0.1000	1.0587

```
##      80      99.2406      nan    0.1000    0.1316
##     100     92.5674      nan    0.1000   -0.1656
##     120     87.5723      nan    0.1000    0.1431
##     140     83.9197      nan    0.1000   -0.3731
##     150     81.6334      nan    0.1000    0.2738
##
## Iter   TrainDeviance   ValidDeviance   StepSize   Improve
##      1      1066.3687         nan    0.1000  169.8292
##      2       927.1047         nan    0.1000  114.6841
##      3       807.6555         nan    0.1000  121.2462
##      4       718.3953         nan    0.1000   99.4117
##      5       634.5350         nan    0.1000   79.5987
##      6       573.8356         nan    0.1000   69.3062
##      7       515.3659         nan    0.1000   47.3729
##      8       470.2866         nan    0.1000   36.9176
##      9       429.8505         nan    0.1000   39.8279
##     10       400.3352         nan    0.1000   30.4731
##     20       229.6228         nan    0.1000    8.5911
##     40       144.8718         nan    0.1000    1.4341
##     60       123.6587         nan    0.1000    1.1849
##     80       111.3196         nan    0.1000    0.1940
##    100       105.9478         nan    0.1000   -0.0948
##    120       101.1540         nan    0.1000   -0.2309
##    140        97.2260         nan    0.1000   -0.1688
##    150        95.4646         nan    0.1000   -0.2985
```

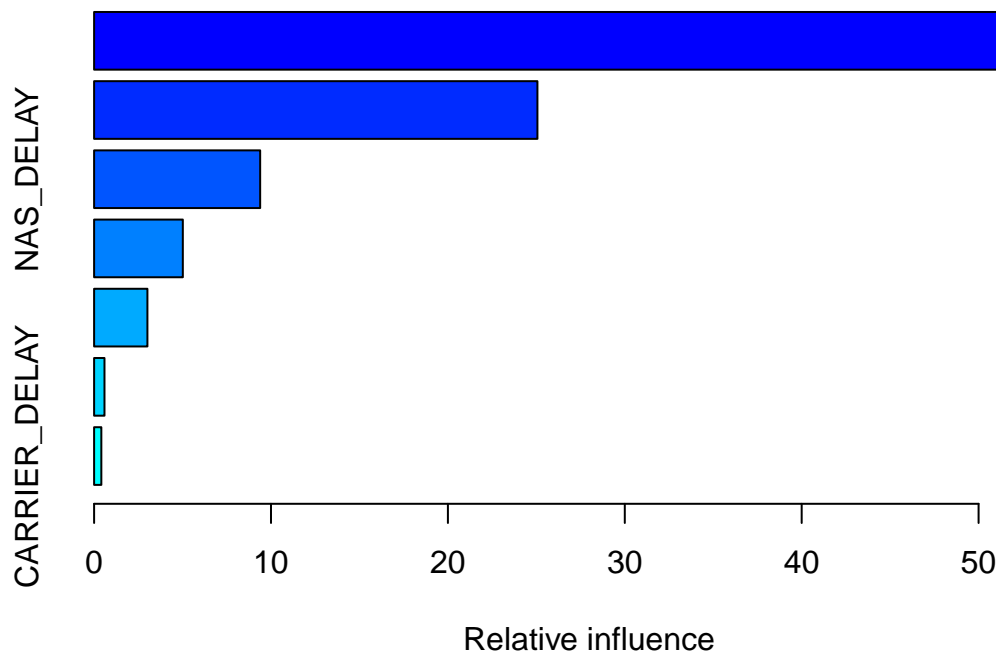
```
gbmFit
```

```
## Stochastic Gradient Boosting
##
## 1310 samples
##    7 predictor
##
## No pre-processing
## Resampling: Bootstrapped (25 reps)
## Summary of sample sizes: 1310, 1310, 1310, 1310, 1310, 1310, ...
## Resampling results across tuning parameters:
##
##  interaction.depth  n.trees  RMSE      Rsquared  MAE
##      1              50    15.43312  0.8413952  11.624536
##      1             100    13.21950  0.8674105   9.881563
##      1             150    12.57194  0.8778294   9.287849
##      2              50    13.58503  0.8617336   9.939209
##      2             100    12.40400  0.8806513   8.916071
##      2             150    12.05968  0.8869600   8.584478
##      3              50    13.00810  0.8714726   9.378071
##      3             100    12.10156  0.8865241   8.592430
##      3             150    11.86209  0.8907303   8.365195
##
## Tuning parameter 'shrinkage' was held constant at a value of 0.1
##
## Tuning parameter 'n.minobsinnode' was held constant at a value of 10
## RMSE was used to select the optimal model using the smallest value.
## The final values used for the model were n.trees = 150, interaction.depth =
##    3, shrinkage = 0.1 and n.minobsinnode = 10.
```

```
# boosted model with cross-validated hyper-parameters
boost.delay <- gbm(Arr_delay ~ DAY_OF_MONTH +
  TAXI_IN +
  TAXI_OUT +
  DEP_DELAY +
  CARRIER_DELAY +
  NAS_DELAY +
  LATE_AIRCRAFT_DELAY,
  data = train, distribution = "gaussian",
  n.trees=150, interaction.depth=3, shrinkage=0.1, cv.folds=10)
```

The `summary()` function also provides a relative influence plot and also outputs the relative influence statistics.

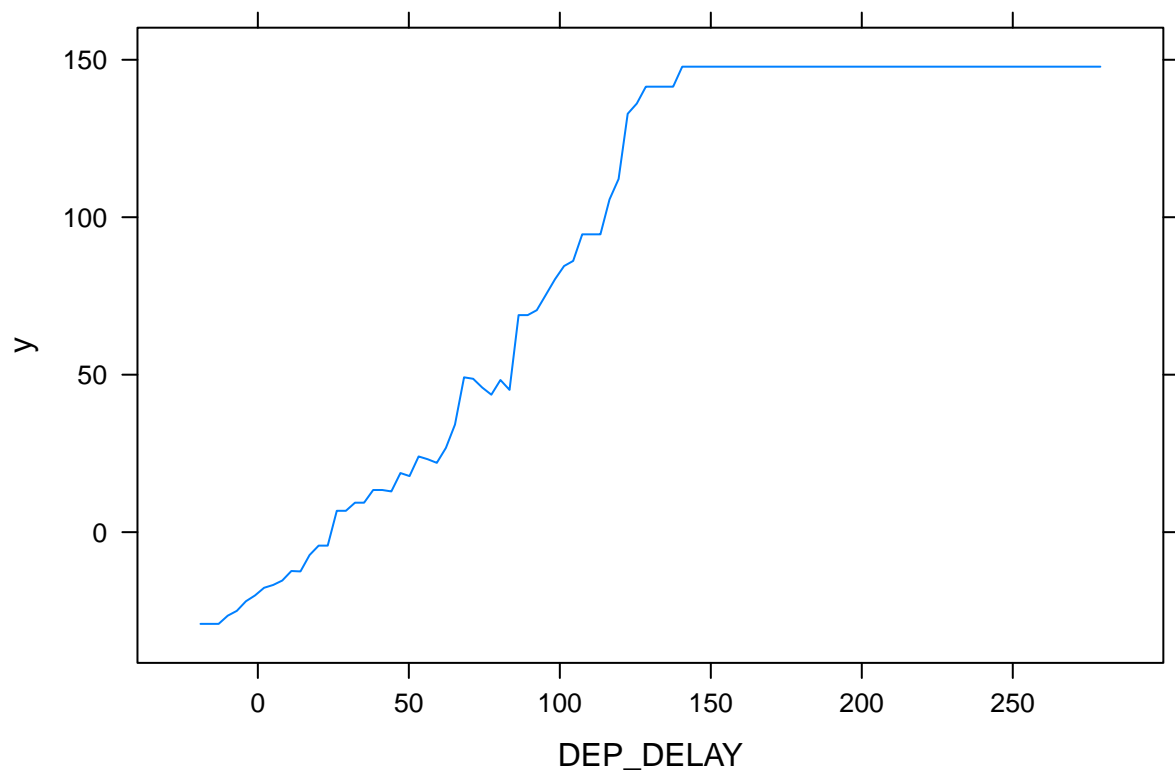
```
summary(boost.delay)
```



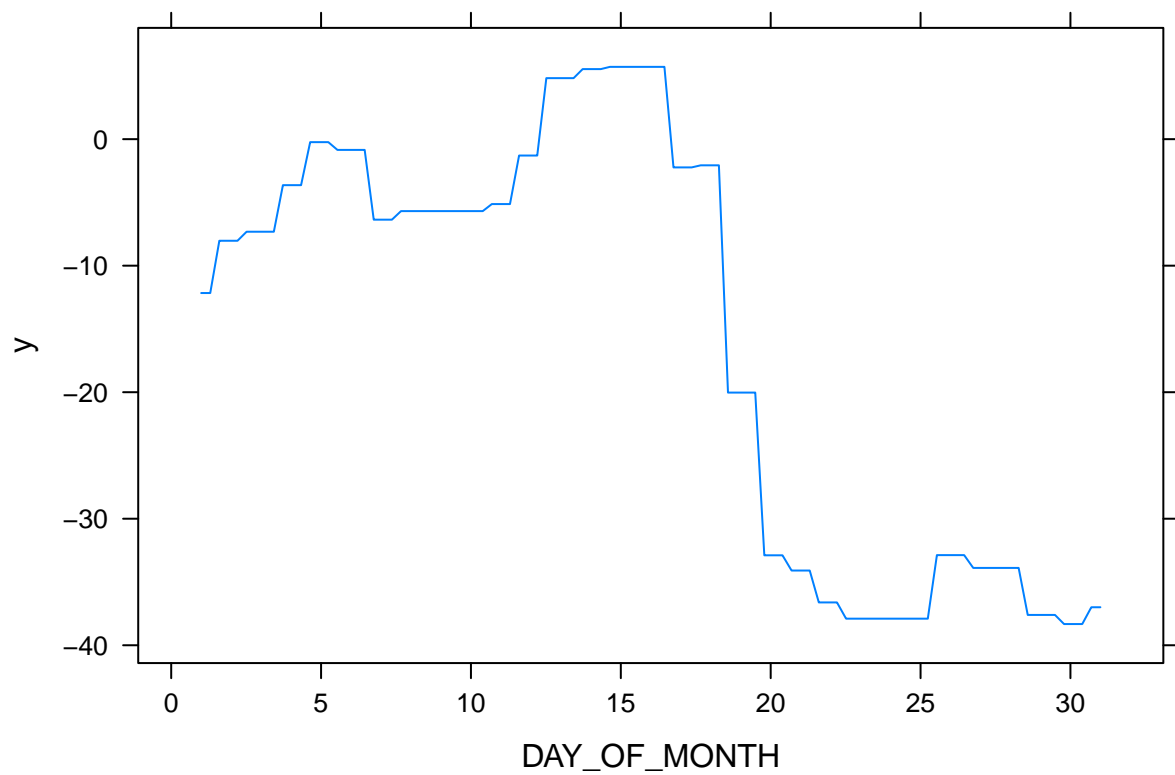
```
##              var    rel.inf
## DEP_DELAY      DEP_DELAY 56.5279213
## DAY_OF_MONTH   DAY_OF_MONTH 25.0608132
## NAS_DELAY      NAS_DELAY  9.3867001
## TAXI_OUT       TAXI_OUT  5.0157480
## TAXI_IN        TAXI_IN  3.0135950
## LATE_AIRCRAFT_DELAY LATE_AIRCRAFT_DELAY 0.5862760
## CARRIER_DELAY CARRIER_DELAY 0.4089464
```

We see that `DEP_DELAY` and `DAY_OF_MONTH` are by far the most important variables. We can also produce *partial dependence plots* for these two variables. These plots illustrate the marginal effect of the selected variables on the response after *integrating* out the other variables.

```
par(mfrow = c(1,2))
plot(boost.delay, i = "DEP_DELAY")
```



```
plot(boost.delay, i = "DAY_OF_MONTH")
```



We now use the boosted model to predict ARR_DELAY on the test set:


```
yhat.boost <- predict(boost.delay, newdata =test,
                      n.trees = 150)
boost_MSE <- sum((test$ARR_DELAY-yhat.boost)^2, na.rm = T)/328
boost_MSE
```

```
## [1] 129.7965
```

Test Errors

```
options(scipen = 5, digits = 4)
model.names <- c("Baseline Linear", "Selected Linear w/ Log-Transformed Predictors", "Selected Linear w/
model.types <- c("Multiple Linear Regression", "Multiple Linear Regression", "Multiple Linear Regression")

#model4.ints <- c("FALSE", "FALSE", "FALSE", "FALSE")

model.mse <- c(plain_linear_model_MSE, log_linear_MSE,
              bc_adj_linear_model_MSE, gam_MSE,
              gambc_MSE, rf.MSE, boost_MSE)
model.mse.char <- c("322.46", "333.90", "334.92", "312.30", "317.45", "155.01", "129.80")

#model4.cumse_var <- c(mlr4_1_cv, mlr4_4_bc_cv, ridge.mom4.cumse, gam4_bc_gcv)
#model4.cumse <- c(2.284e+20, "25.66", 25.62, 25.79)

#pctchange4_3 <- round(-((ridge.mom4.cumse - mlr4_4_bc_cv)/mlr4_4_bc_cv)*100, digits = 4)
#pctchange4_4 <- round(-((gam4_bc_gcv - mlr4_4_bc_cv)/mlr4_4_bc_cv)*100, digits = 4)

#model4.pctchange <- c("---", "---", pctchange4_3, pctchange4_4)

errors.df <- data.frame(model.names,
                        model.types,
                        model.mse.char
                        )

#model4.ints,
#model4.cumse,
#model4.pctchange

#colnames(errors.df4) <- c("Model Name", "Model Type", "Interactions?", "Model MSE", "Model CV MSE", "M
errors.df
```

```
##              model.names              model.types
## 1              Baseline Linear Multiple Linear Regression
## 2 Selected Linear w/ Log-Transformed Predictors Multiple Linear Regression
## 3              Selected Linear w/ Box-Cox Multiple Linear Regression
## 4                                GAM Generalized Additive Model
## 5              GAM w/ Box-Cox Generalized Additive Model
## 6              Random Forest              Random Forest
## 7              Boosting              Boosting
##  model.mse.char
## 1              322.46
## 2              333.90
## 3              334.92
```

```
## 4      312.30
## 5      317.45
## 6      155.01
## 7      129.80
```

```
library(formattable)
```

```
##
## Attaching package: 'formattable'
## The following object is masked from 'package:bst':
##
##      gradient
## The following object is masked from 'package:MASS':
##
##      area
## The following object is masked from 'package:patchwork':
##
##      area
```

```
formattable(errors.df,
  col.names = c("Model Name", "Model Type", "Model MSE"),
  list(
    model.names = formatter("span", style = x ~ ifelse(x == "Boosting",
      style(color = "purple", font.weight = "bold"), NA)),
    model.mse.char = formatter("span", style = x ~ ifelse(x == "129.80",
      style(color = "purple", font.weight = "bold"), NA)),
    model.types = formatter("span", style = x ~ ifelse(x == "Boosting",
      style(color = "purple", font.weight = "bold"), NA))
  ))
```

Model Name

Model Type

Model MSE

Baseline Linear

Multiple Linear Regression

322.46

Selected Linear w/ Log-Transformed Predictors

Multiple Linear Regression

333.90

Selected Linear w/ Box-Cox

Multiple Linear Regression

334.92

GAM

Generalized Additive Model

312.30

GAM w/ Box-Cox

Generalized Additive Model

317.45

Random Forest

Random Forest

155.01

Boosting

Boosting

129.80