

STA 325 Final Project Code

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November 23, 2020

```
library(readr)
library(dplyr)
library(tidyverse)
library(gridExtra)
library(mgcv)
library(patchwork)
library(MASS)
library(EnvStats)
library(tree)
library(randomForest)
library(gbm)
library(bst)
library(plyr)
library(caret)
```

Data Load-In and Initial Cleaning

```
# 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"

# 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>
```

Exploratory Data Analysis

Individual Predictor Variable EDA

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))

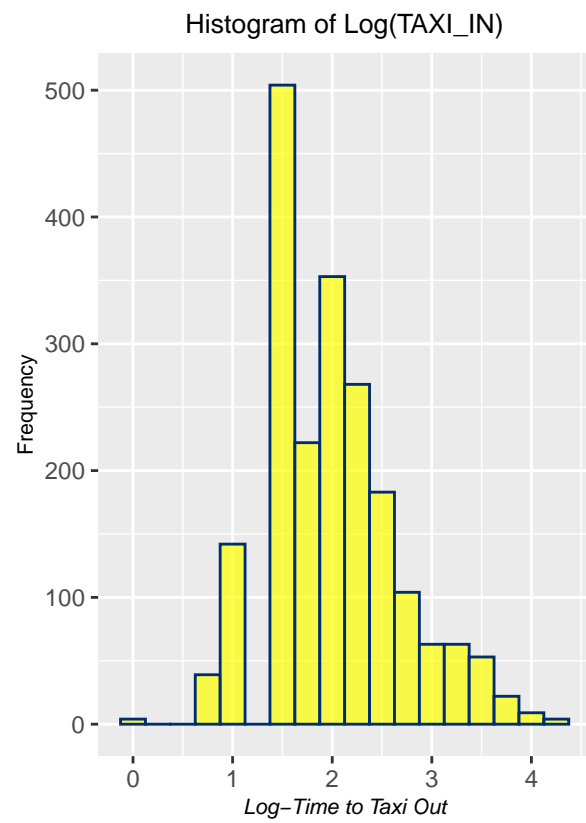
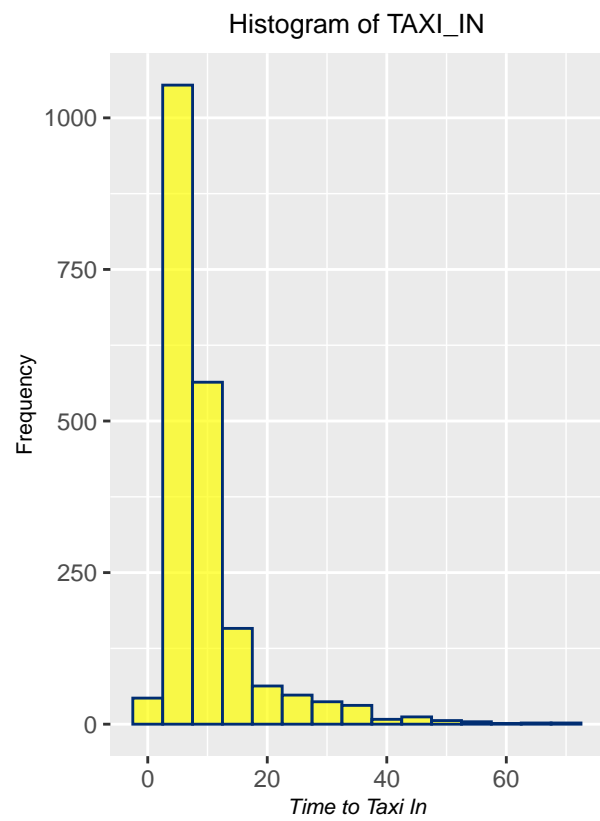
# plot untransformed predictor taxi_out
pTAXI_OUT <- ggplot(data = flights, aes(x = TAXI_OUT)) +
  geom_histogram(binwidth = 5, fill = "#FFFF00", color = "#002D72", alpha = .7) +
  labs(x = "Time to Taxi Out",
       y = "Frequency",
       title = "Histogram of TAXI_OUT") +
  theme(plot.title = element_text(size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

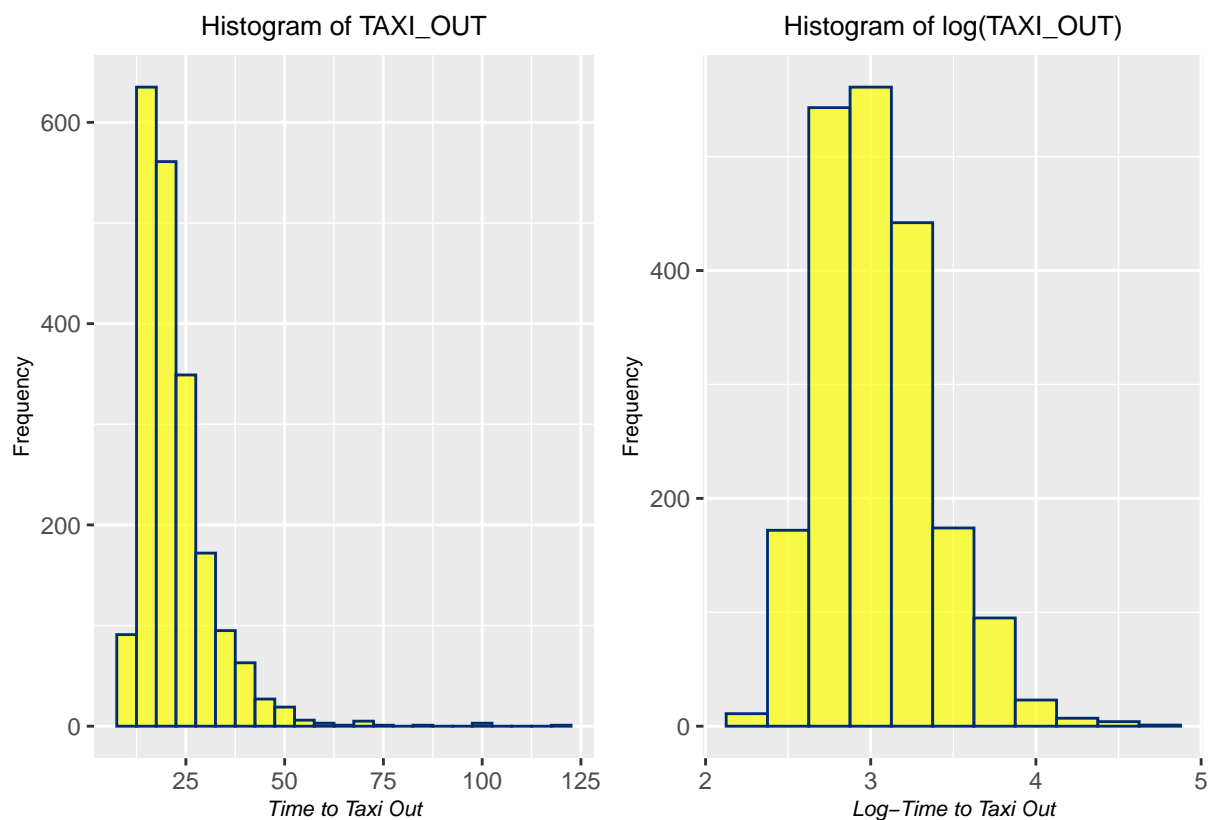
# plot log transformed taxi_out
plog_TAXI_OUT <- ggplot(data = flights, aes(x = log_TAXI_OUT)) +
  geom_histogram(binwidth = .25, fill = "#FFFF00", color = "#002D72", alpha = .7) +
  labs(x = "Log-Time to Taxi Out",
       y = "Frequency",
       title = "Histogram of log(TAXI_OUT)") +
  theme(plot.title = 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(binwidth = .25, fill = "#FFFF00", color = "#002D72", alpha = .7) +
  labs(x = "Log-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



pTAXI_OUT + plog_TAXI_OUT

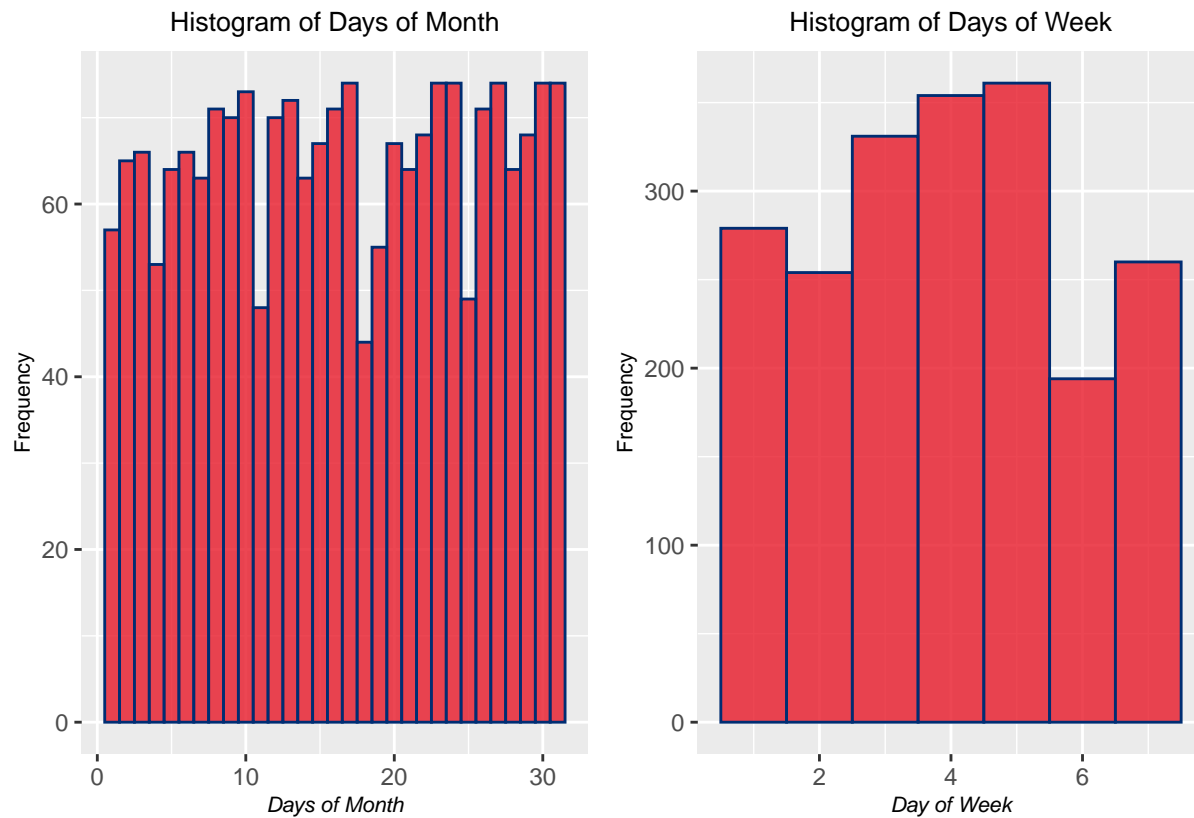


Days of Month and Week

```
# plot predictor DAYS_OF_MONTH
pDOM <- 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
pDOW <- 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))

pDOM + pDOW
```

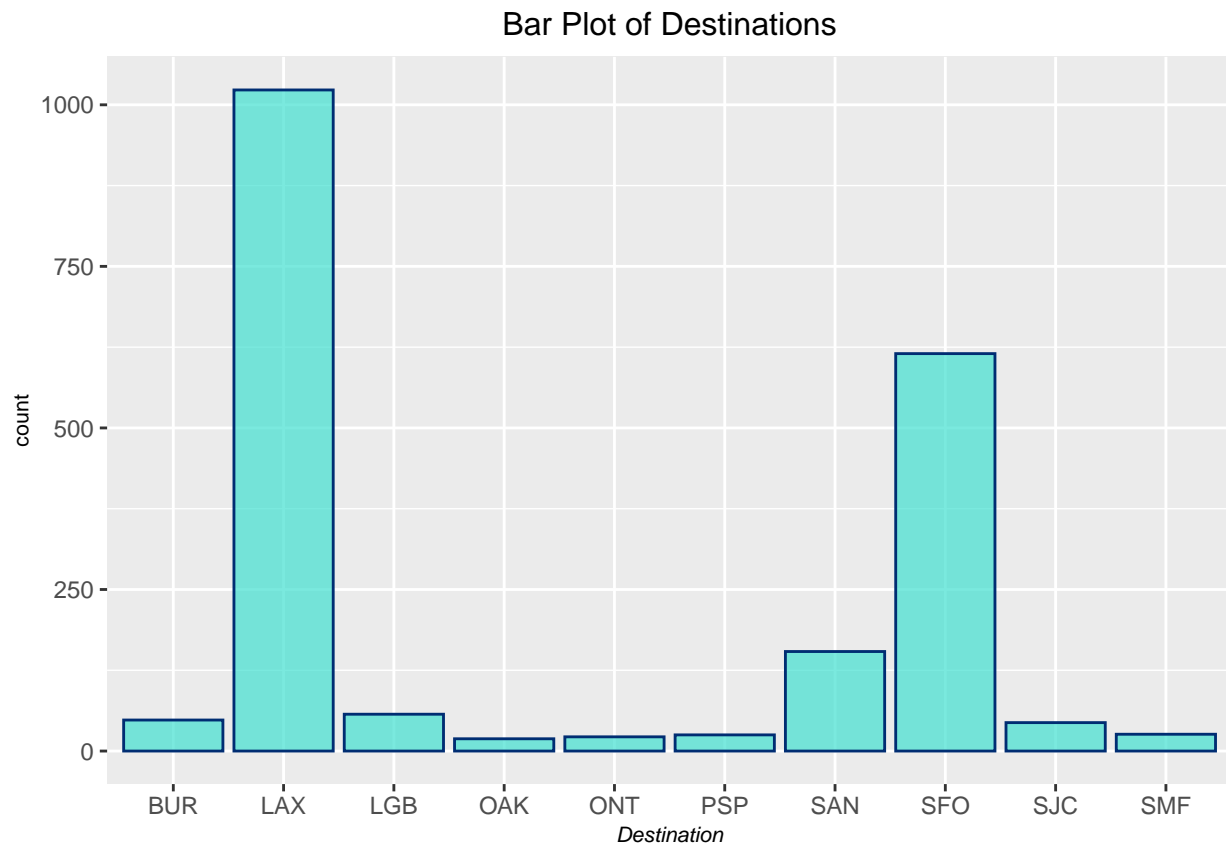


Destination Locations

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

```
# plot destinations in CA
pDEST <- ggplot(data = flights, aes(x = DEST)) +
  geom_bar(fill = "#40E0D0", color = "#002D72", alpha = .7) +
  labs(x = "Destination",
       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))
```

pDEST

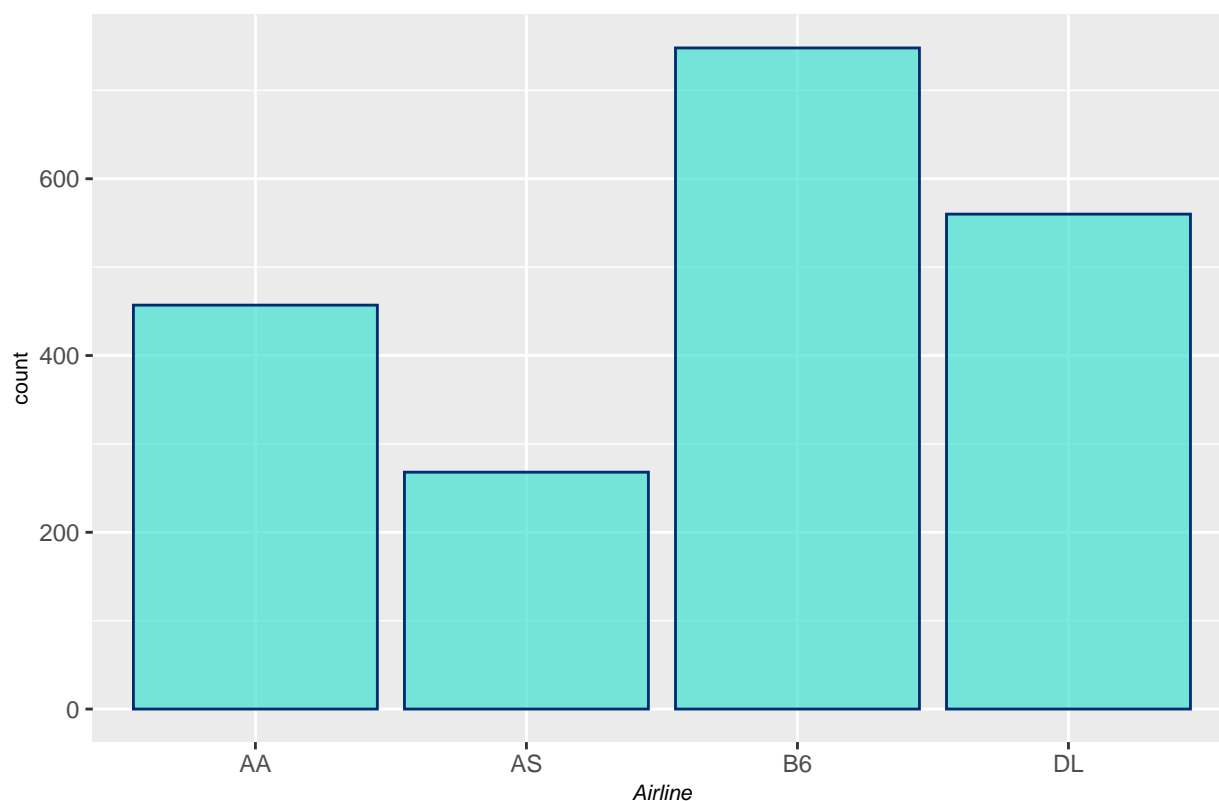


Airlines

```
# plot airline carriers
pLINE <- 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))
```

pLINE

Bar Plot of Airlines



```
# plot airlines by destination
pLINEDEST <- ggplot(data = flights, aes(x = DEST, fill = OP_CARRIER)) +
  geom_bar() +
  labs(x = "Destination",
       title = "Bar Plot of Airlines by Destination",
       fill = "Airline")
  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))
```

```
## List of 4
## $ axis.title.x.bottom:List of 11
## ..$ family      : NULL
## ..$ face        : chr "italic"
## ..$ colour      : NULL
## ..$ size        : num 8
## ..$ hjust       : NULL
## ..$ vjust       : NULL
## ..$ angle       : NULL
## ..$ lineheight   : NULL
## ..$ margin      : NULL
## ..$ debug       : NULL
## ..$ inherit.blank: logi FALSE
## .. attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.title.y.left :List of 11
## ..$ family      : NULL
```



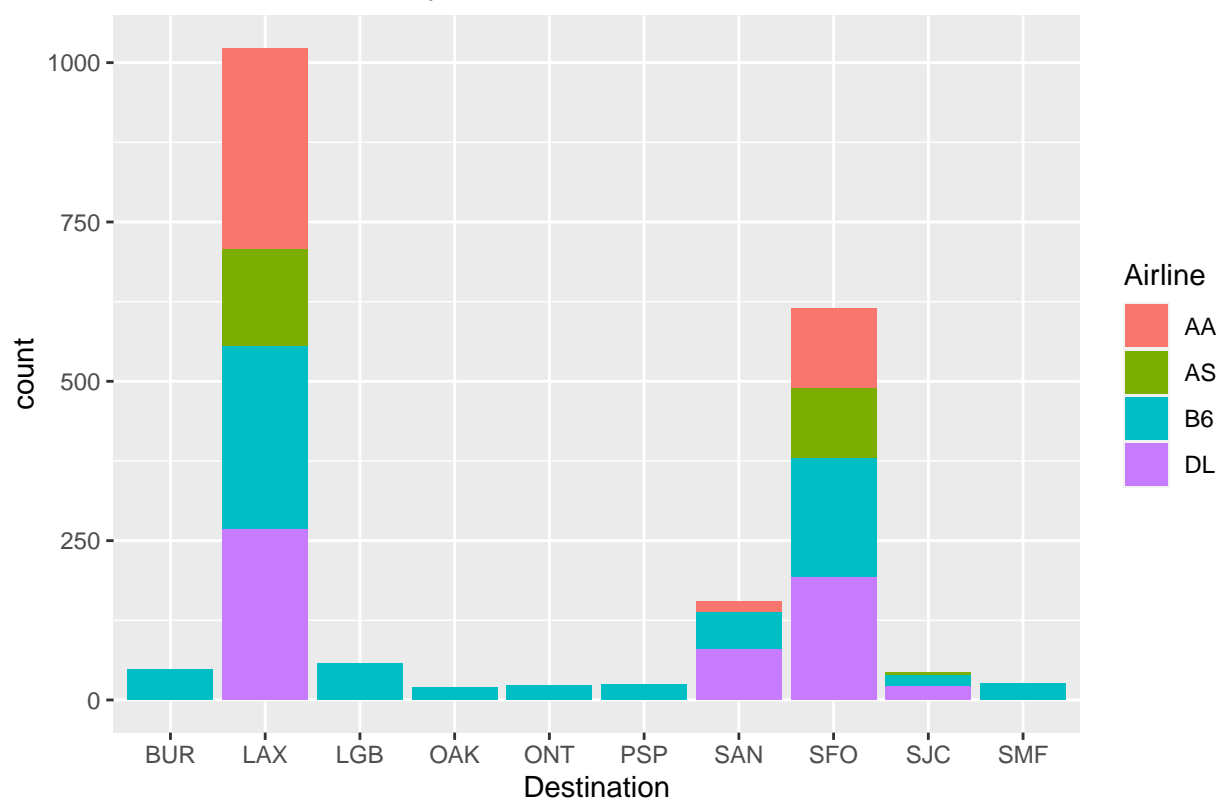
```

## ..$ face          : NULL
## ..$ colour        : NULL
## ..$ size          : num 8
## ..$ hjust         : NULL
## ..$ vjust         : NULL
## ..$ angle         : NULL
## ..$ lineheight    : NULL
## ..$ margin        : NULL
## ..$ debug         : NULL
## ..$ inherit.blank: logi FALSE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ plot.title      :List of 11
## ..$ family        : NULL
## ..$ face          : NULL
## ..$ colour        : NULL
## ..$ size          : num 12
## ..$ hjust         : num 0.5
## ..$ vjust         : NULL
## ..$ angle         : NULL
## ..$ lineheight    : NULL
## ..$ margin        : NULL
## ..$ debug         : NULL
## ..$ inherit.blank: logi FALSE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ plot.subtitle   :List of 11
## ..$ family        : NULL
## ..$ face          : NULL
## ..$ colour        : NULL
## ..$ size          : NULL
## ..$ hjust         : num 0.5
## ..$ vjust         : NULL
## ..$ angle         : NULL
## ..$ lineheight    : NULL
## ..$ margin        : NULL
## ..$ debug         : NULL
## ..$ inherit.blank: logi FALSE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## - attr(*, "class")= chr [1:2] "theme" "gg"
## - attr(*, "complete")= logi FALSE
## - attr(*, "validate")= logi TRUE

```

pLINEDEST

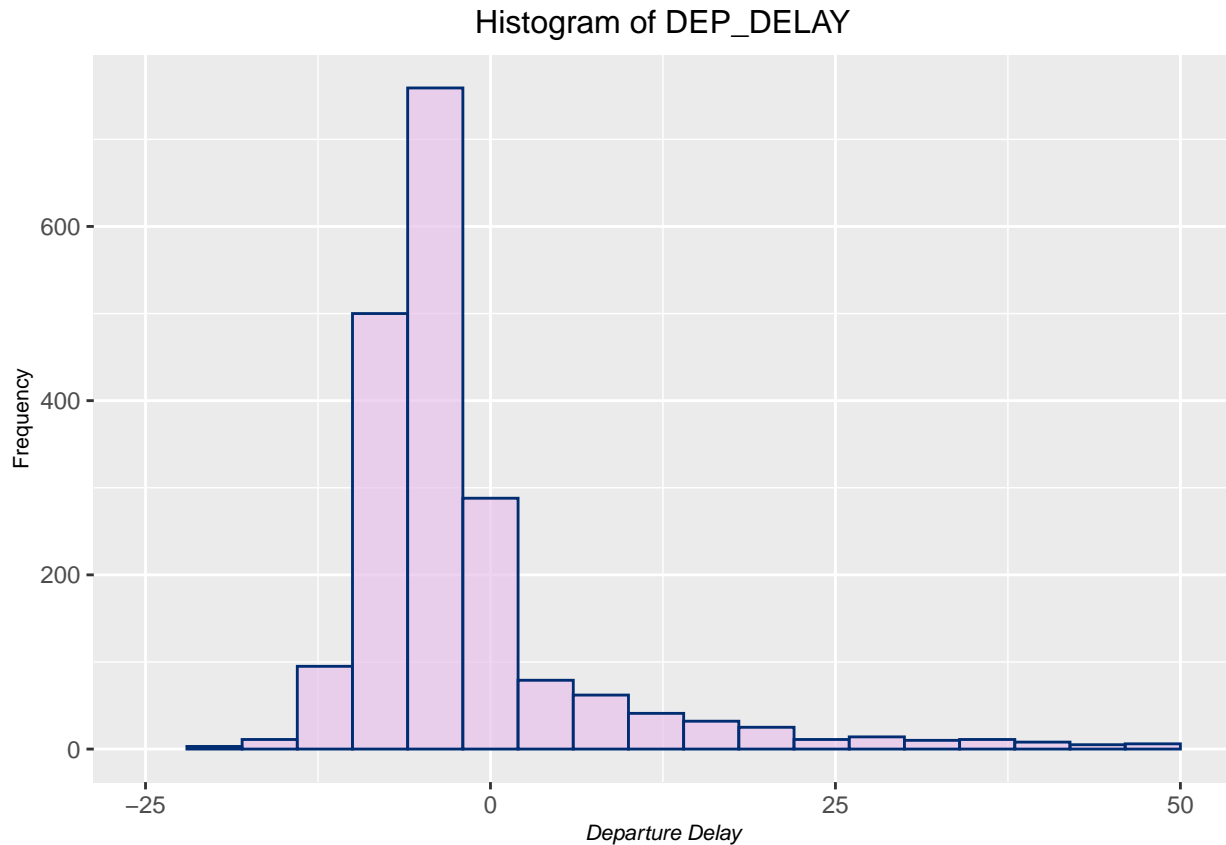
Bar Plot of Airlines by Destination



Depart Delay Histogram

```
# plot DEP_DELAY
pDEPDELAY <- 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))

pDEPDELAY
```



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

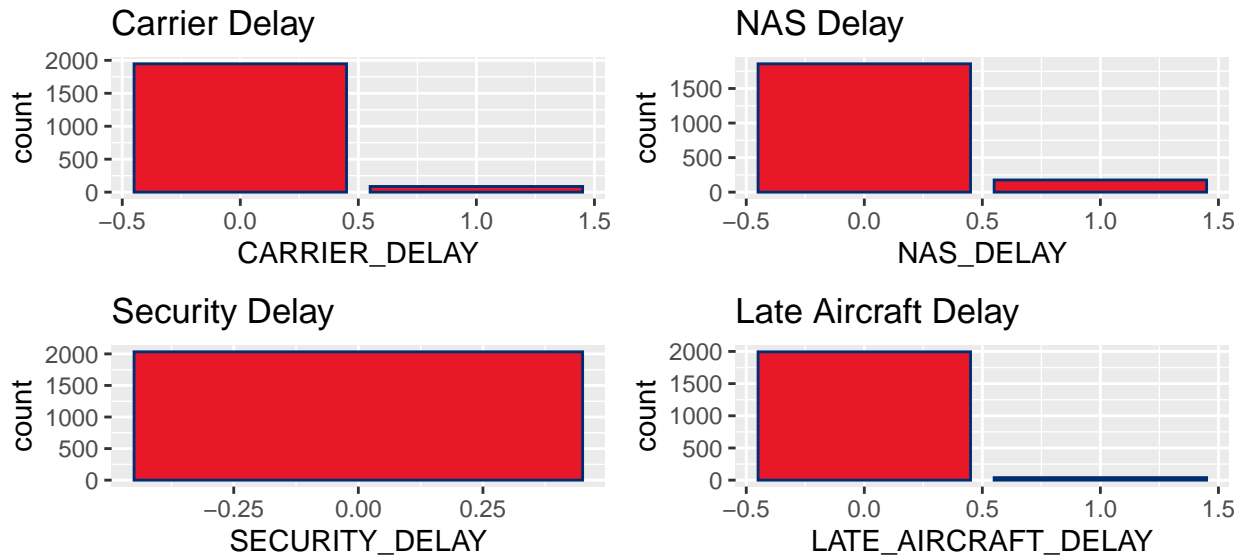
#p2 <- ggplot(data = flights, aes(x = WEATHER_DELAY)) +
#geom_bar(fill = "#E81828", color = "#002D72") +
#labs(title = "Weather Delay")

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,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.

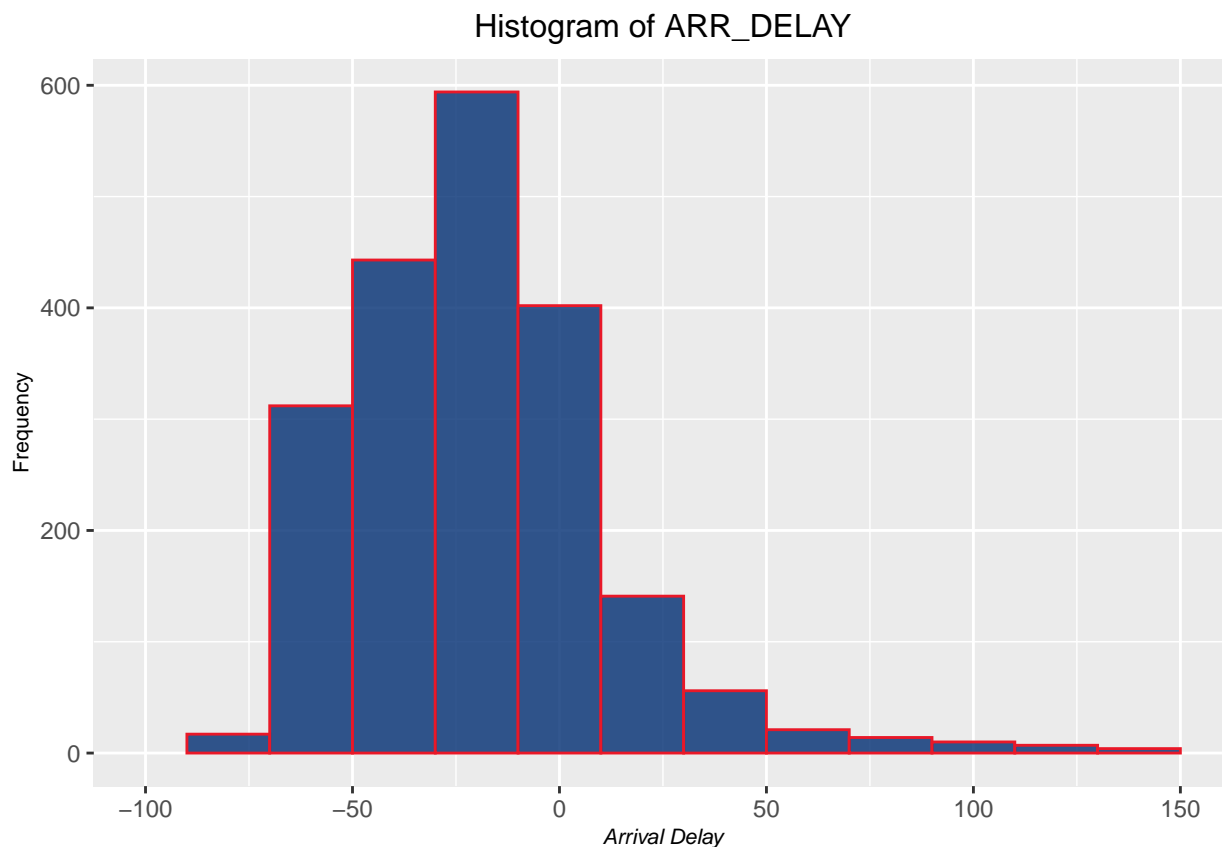
Furthermore, only 9 flights are classified with a weather delay, so it may not be good for our model to include this as a variable for right now.

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

Response Variable EDA

```
# plot ARR_DELAY
pARRDELAY <- 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))
```

pARRDELAY



Predictors vs. Response EDA

TAXI_IN / TAXI_OUT vs. ARR_DELAY

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

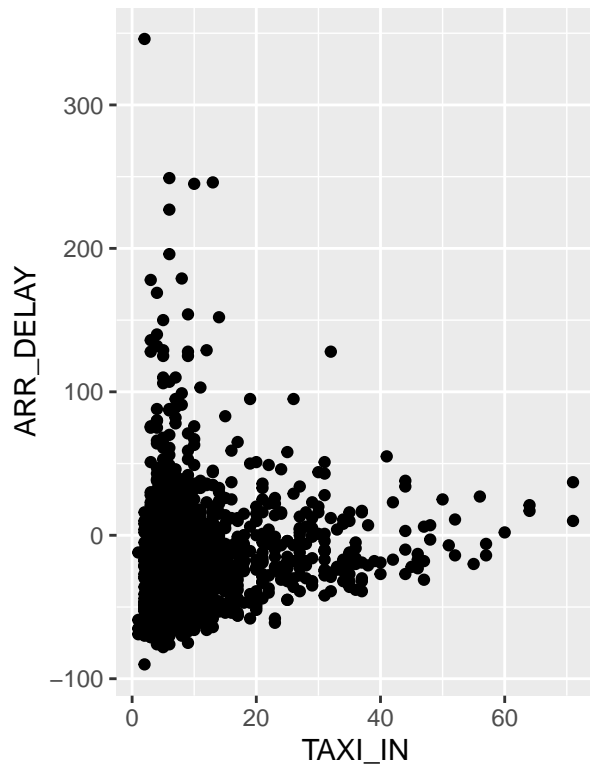
p7 <- ggplot(data = flights, aes(y = ARR_DELAY, x = TAXI_OUT)) +
  geom_point() +
  labs(title = "Arrival Delay vs Taxi_Out")

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

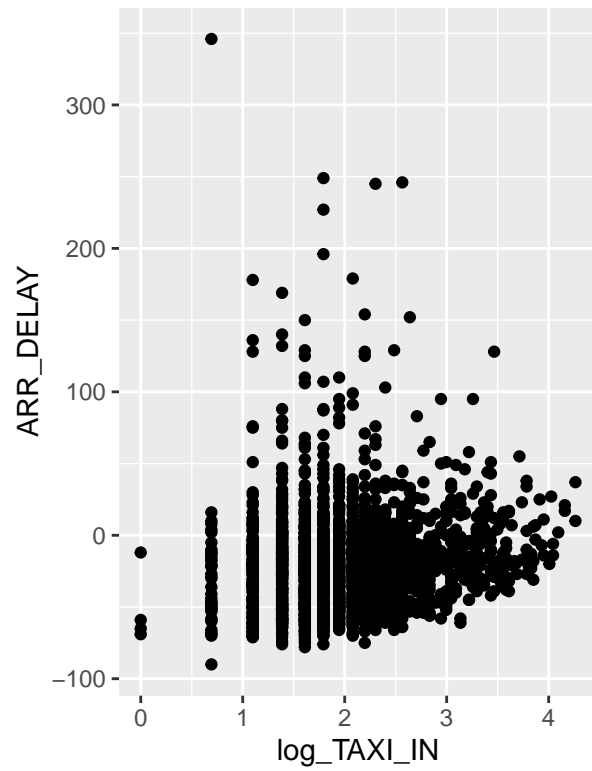
plog7 <- ggplot(data = flights, aes(y = ARR_DELAY, x = log_TAXI_OUT)) +
  geom_point() +
  labs(title = "Arrival Delay vs log(Taxi_Out)")

p6 + plog6
```

Arrival Delay vs Taxi_In

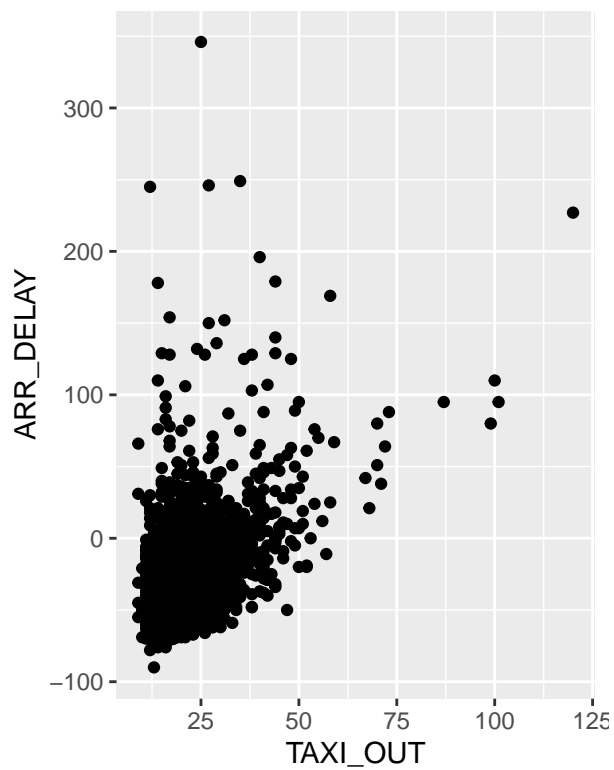


Arrival Delay vs log(Taxi_In)

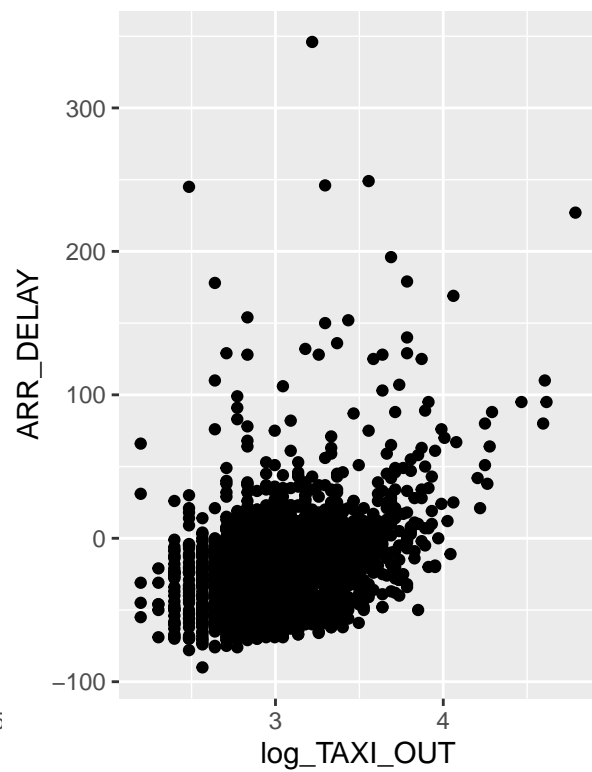


p7 + plog7

Arrival Delay vs Taxi_Out



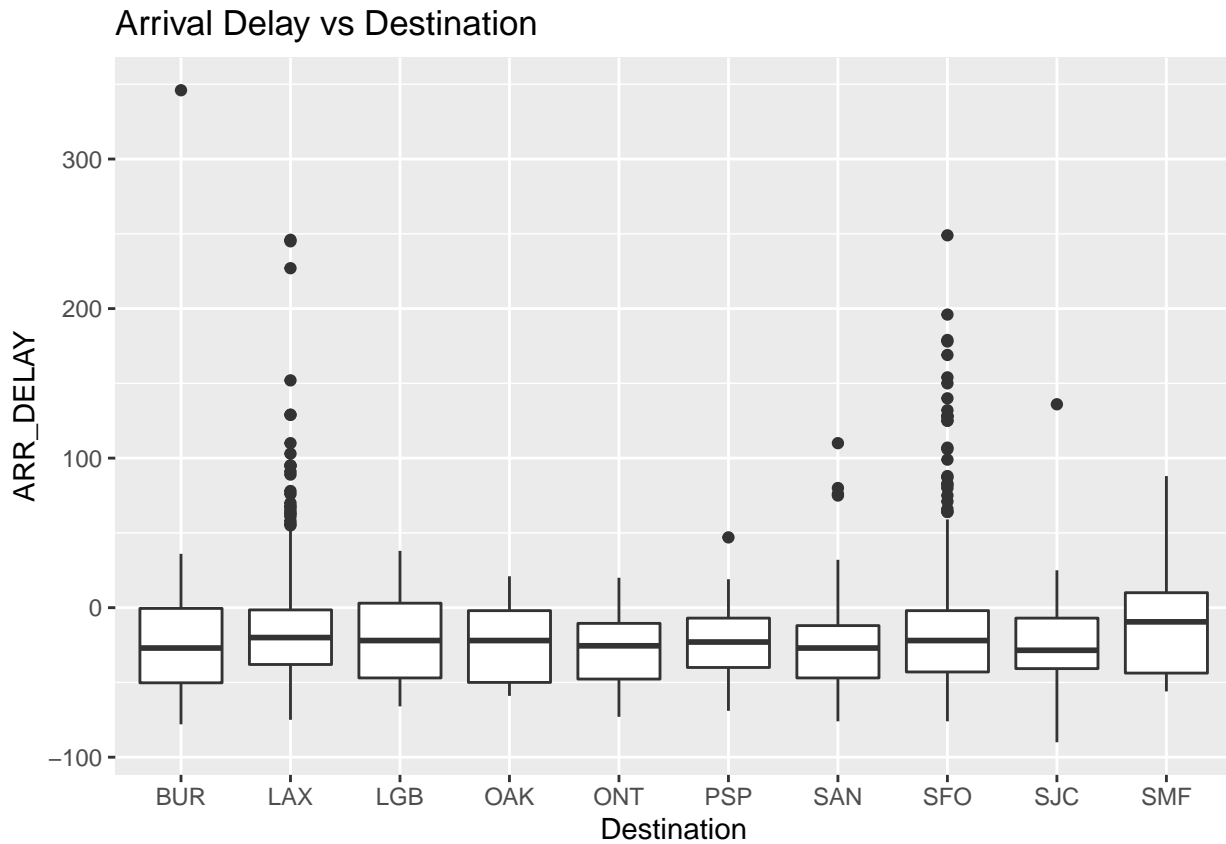
Arrival Delay vs log(Taxi_Out)



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

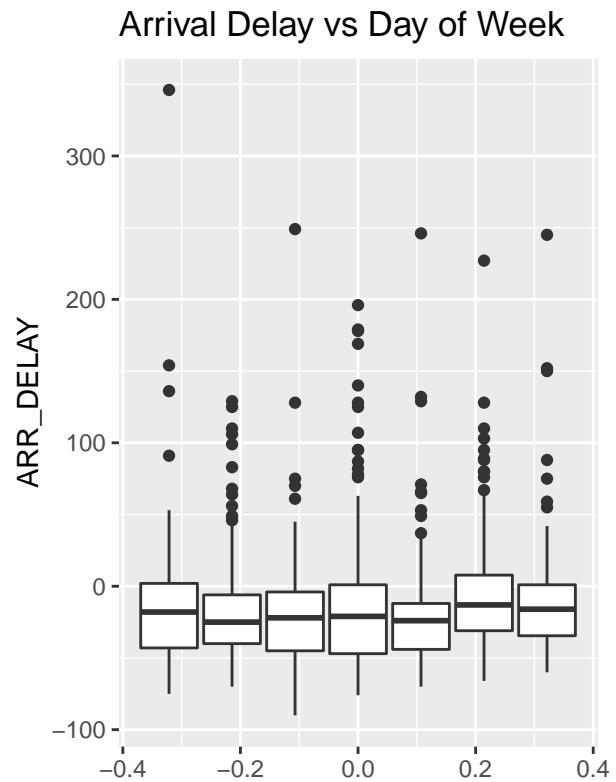
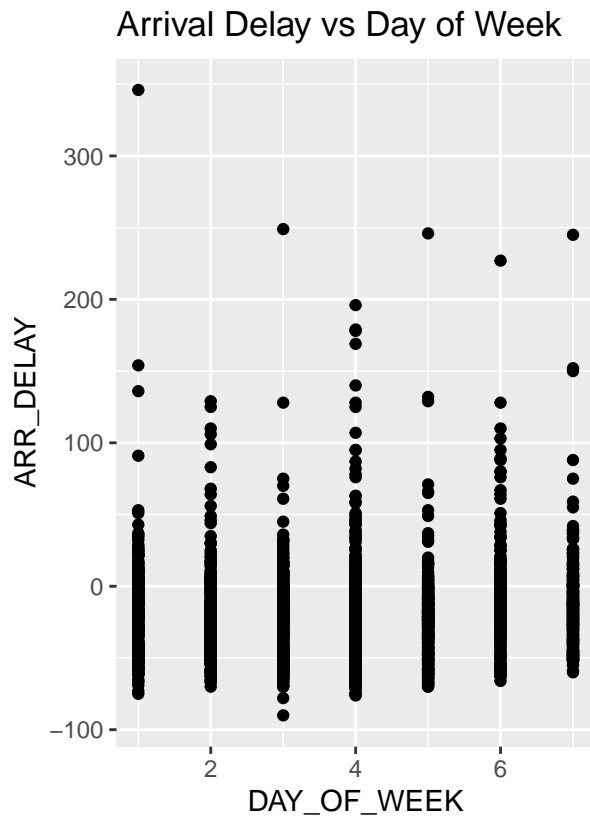
DEST vs. ARR_DELAY

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



DAY_OF_WEEK vs. ARR_DELAY

```
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")  
  
p8 + p9
```

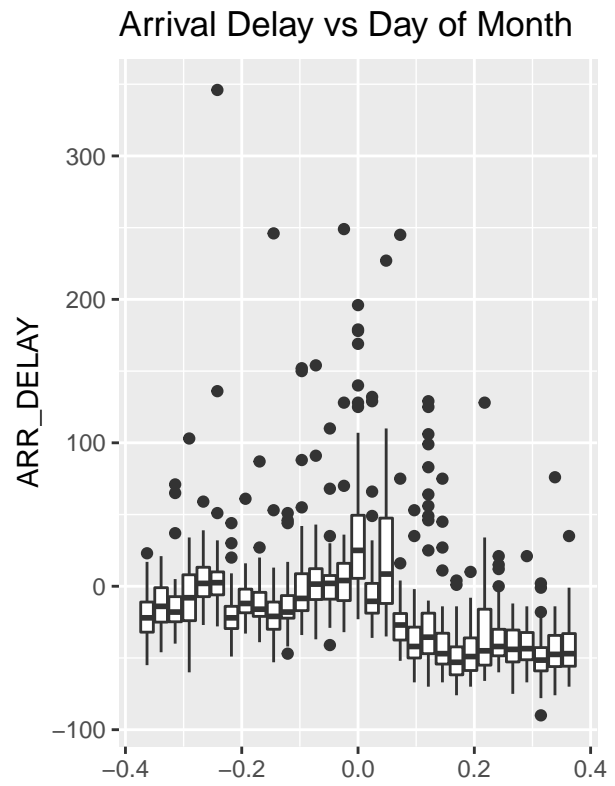
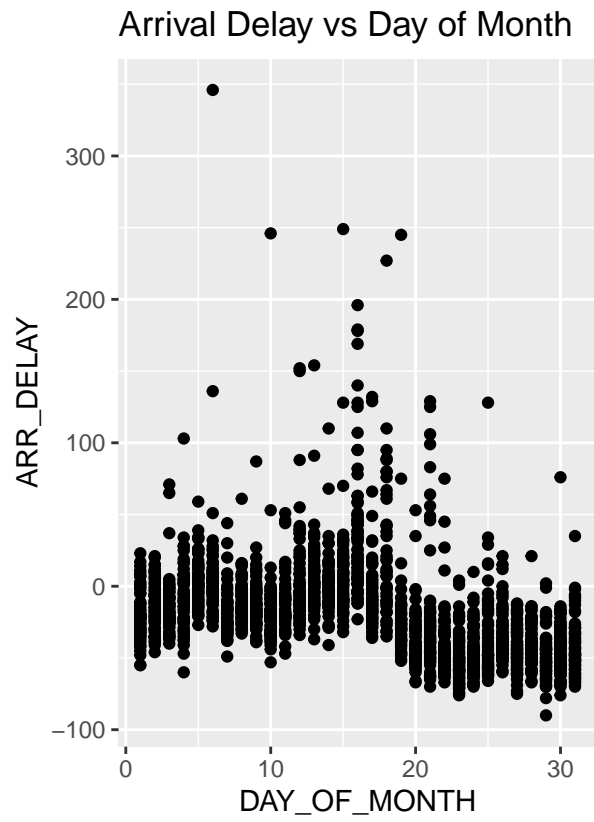


ARR_DELAY and DAY_OF_MONTH

```
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")

p10 + p11
```

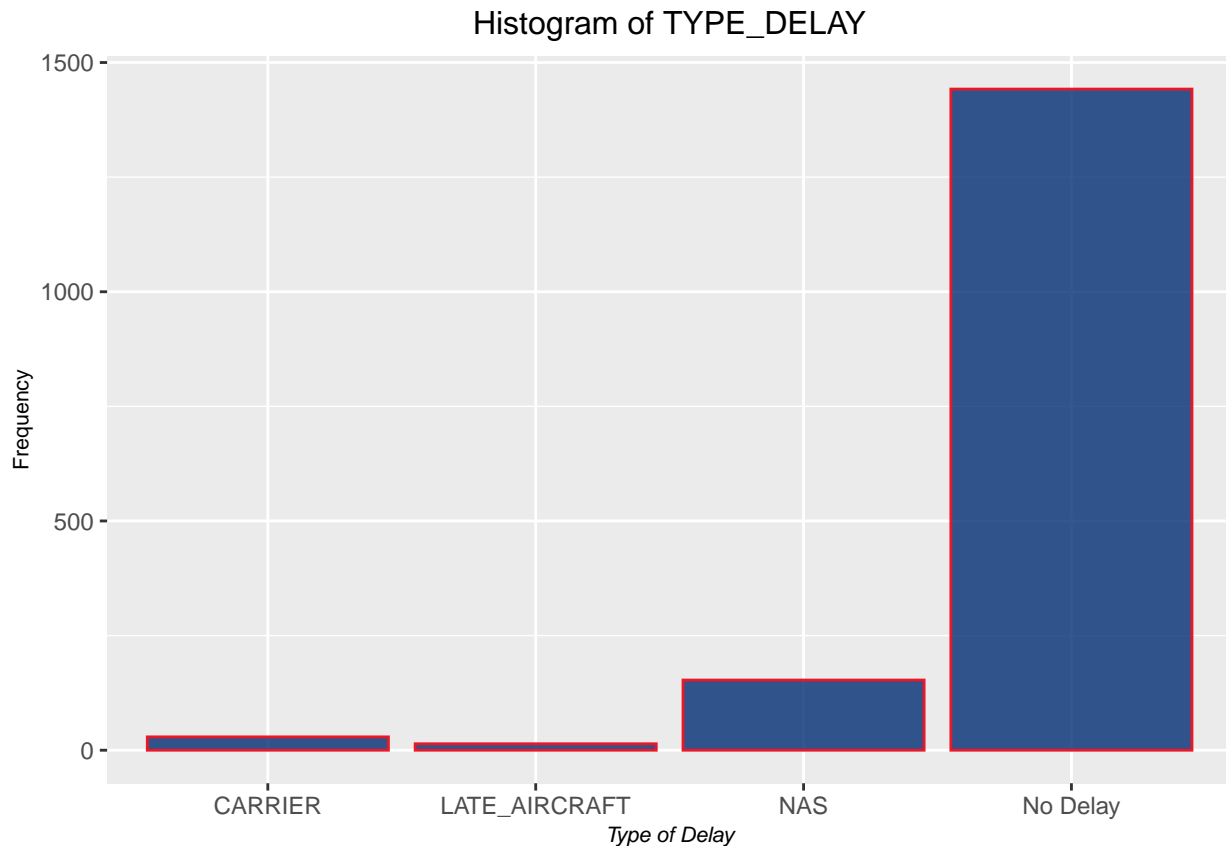
Additional Data Cleaning

New Bounds and Variable

```
# take only SFO/LAX since all 4 carriers fly there
# create TYPE_DELAY as a factor variable for type of delays
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",
                                WEATHER_DELAY == 1 ~ "WEATHER",
                                TRUE ~ "No Delay"))
```

```
pTYPE <- ggplot(data = flights, aes(x = TYPE_DELAY)) +
  geom_bar(fill = "#002D72", color = "#E81828", alpha = 0.8) +
  labs(x = "Type of Delay",
       y = "Frequency",
       title = "Histogram of TYPE_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))
```

pTYPE



Test and Training Set Split

We will use a 80-20 split of training and test sets.

```
set.seed(1234)

flights <- flights %>%
  dplyr::mutate(id = row_number())

train <- flights %>%
  sample_frac(0.8)
test <- anti_join(flights, train, by = "id")
```

Modeling

(1) Multiple Linear Regression

(a) Baesline Linear Model (with AIC Selection)

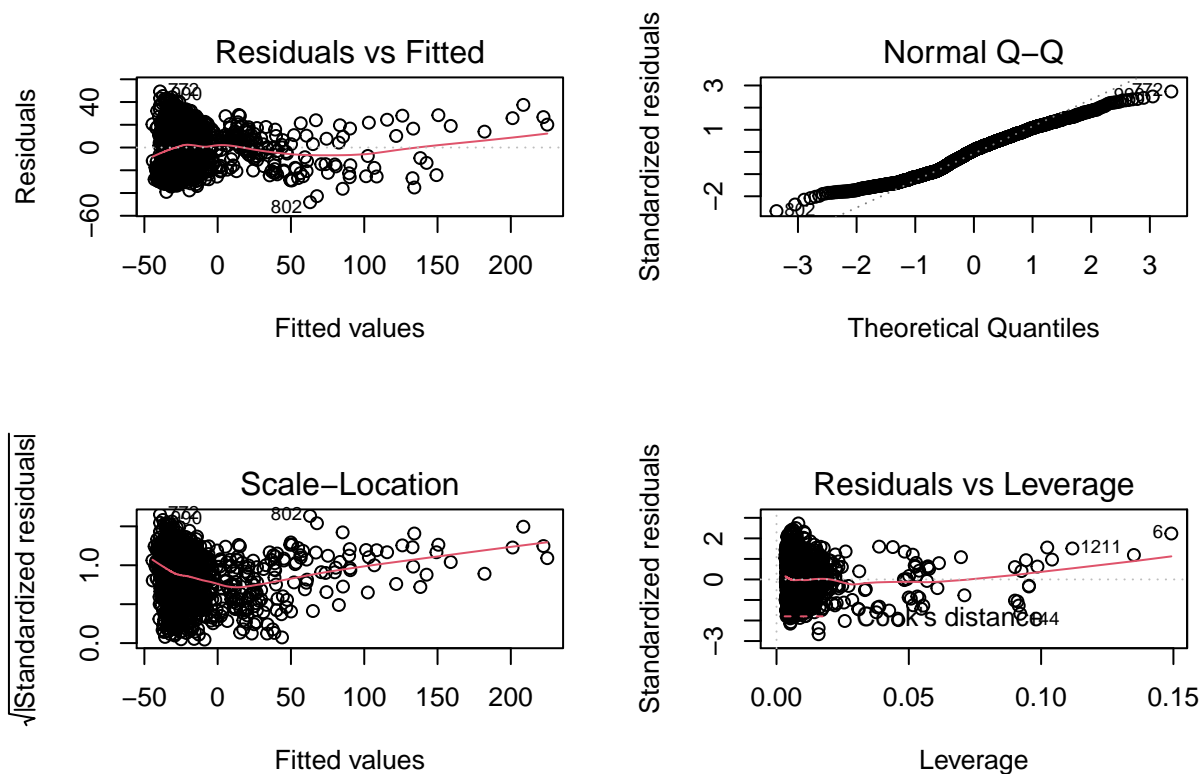
```
# create linear model with all revelant variables
full_model <- lm(ARR_DELAY ~
  DEP_DELAY +
  DAY_OF_WEEK +
  OP_CARRIER +
  DEST +
  CRS_DEP_TIME +
  CRS_ARR_TIME +
  TAXI_OUT +
  TAXI_IN +
  TYPE_DELAY, train)

# summary(full_model)

# use AIC model selection on full model
step_model <- stepAIC(full_model, trace = FALSE)
#summary(step_model)

# update full model
plain_linear_model <- step_model

# plot full model
par(mfrow = c(2,2))
plot(plain_linear_model)
```



(b) Full Log-Transformed Model

```
# fitting models and running ANOVA tests to identify interactions

## first model
lm.01 <- lm(ARR_DELAY ~ DEP_DELAY + DAY_OF_WEEK + OP_CARRIER + DEST + CRS_DEP_TIME + CRS_ARR_TIME + log_TAXI_OUT + log_TAXI_IN, data = train)
#plot(lm.01)
#summary(lm.01)

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

## third model
lm.02 <- lm(ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_OUT + log_TAXI_IN + TYPE_DELAY, data = train)
#summary(lm.02)
#anova(step_model, lm.02)

## fourth model
lm.03 <- lm(ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_OUT + log_TAXI_IN + TYPE_DELAY, data = train)
#anova(lm.02, lm.03)

## final log model
log_linear_model <- lm(ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_OUT + log_TAXI_IN + TYPE_DELAY, data = train)
anova(lm.03, log_linear_model)

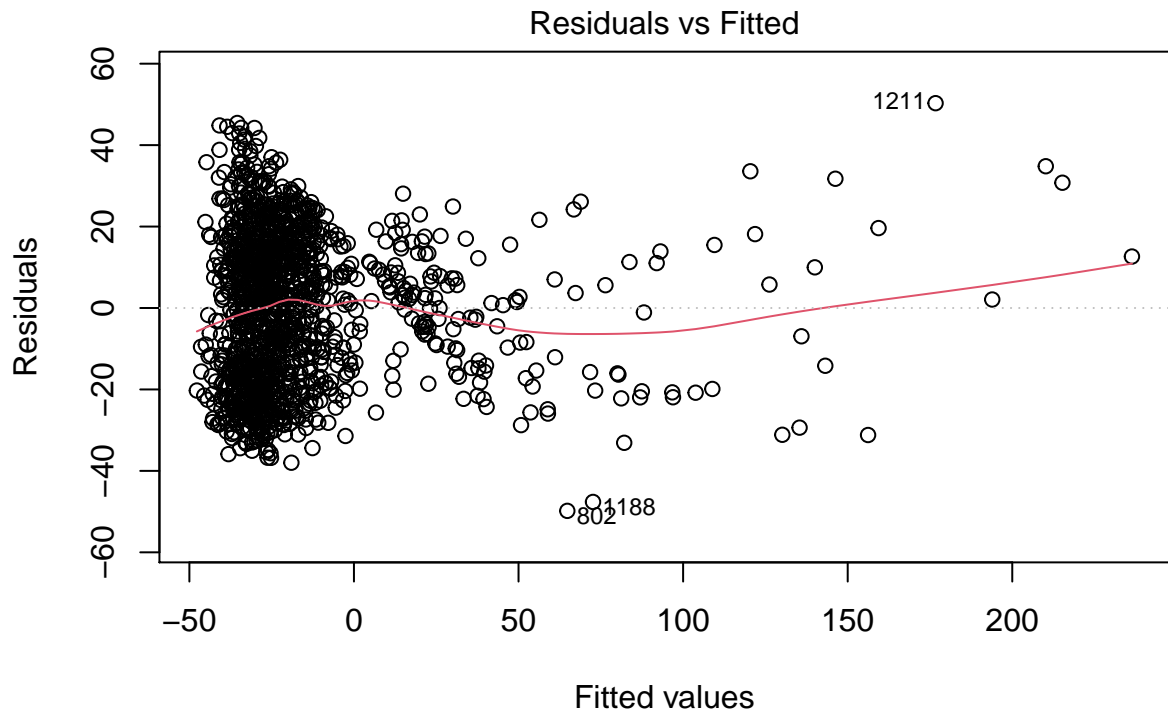
## Analysis of Variance Table
##
## Model 1: ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_OUT +
##   log_TAXI_IN + TYPE_DELAY + OP_CARRIER:DEST + DEST:log_TAXI_IN
## Model 2: ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_OUT +
##   log_TAXI_IN + TYPE_DELAY + 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)

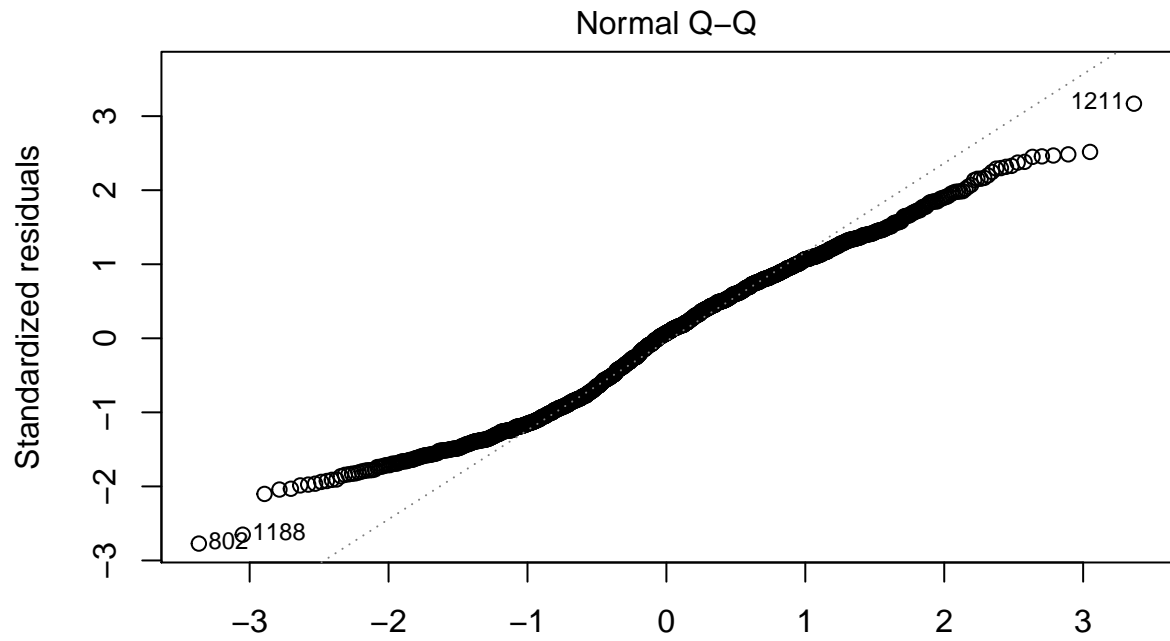
##
## Call:
## lm(formula = ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME +
##   log_TAXI_OUT + log_TAXI_IN + TYPE_DELAY + OP_CARRIER:DEST +
##   DEST:log_TAXI_IN + log_TAXI_OUT:DEP_DELAY, data = train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -49.817 -15.330   1.198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

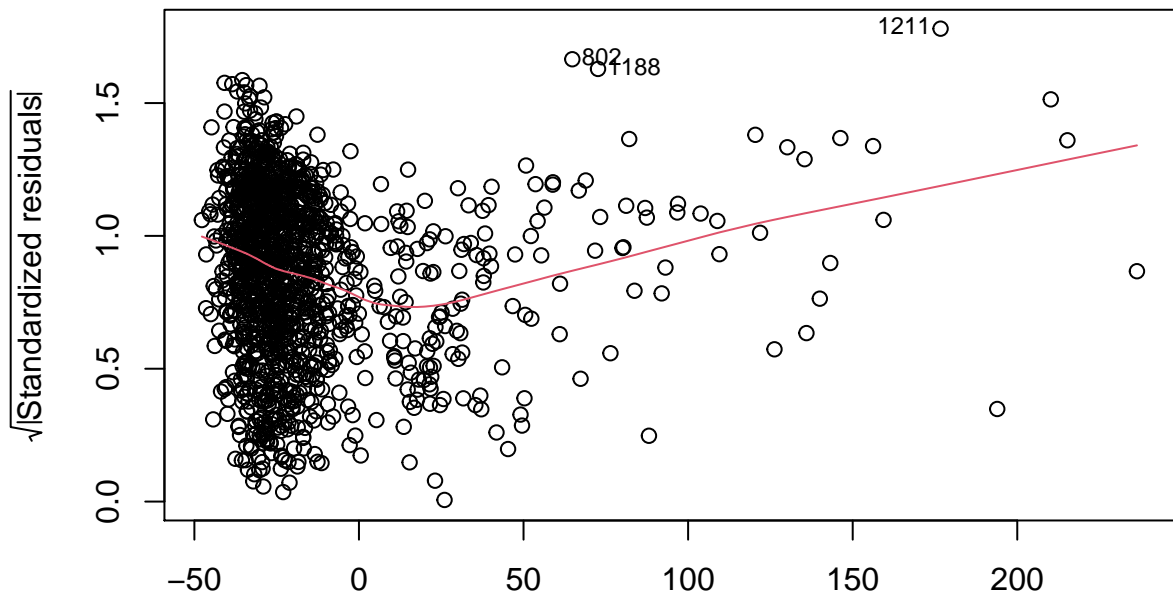
```
plot(log_linear_model)
```



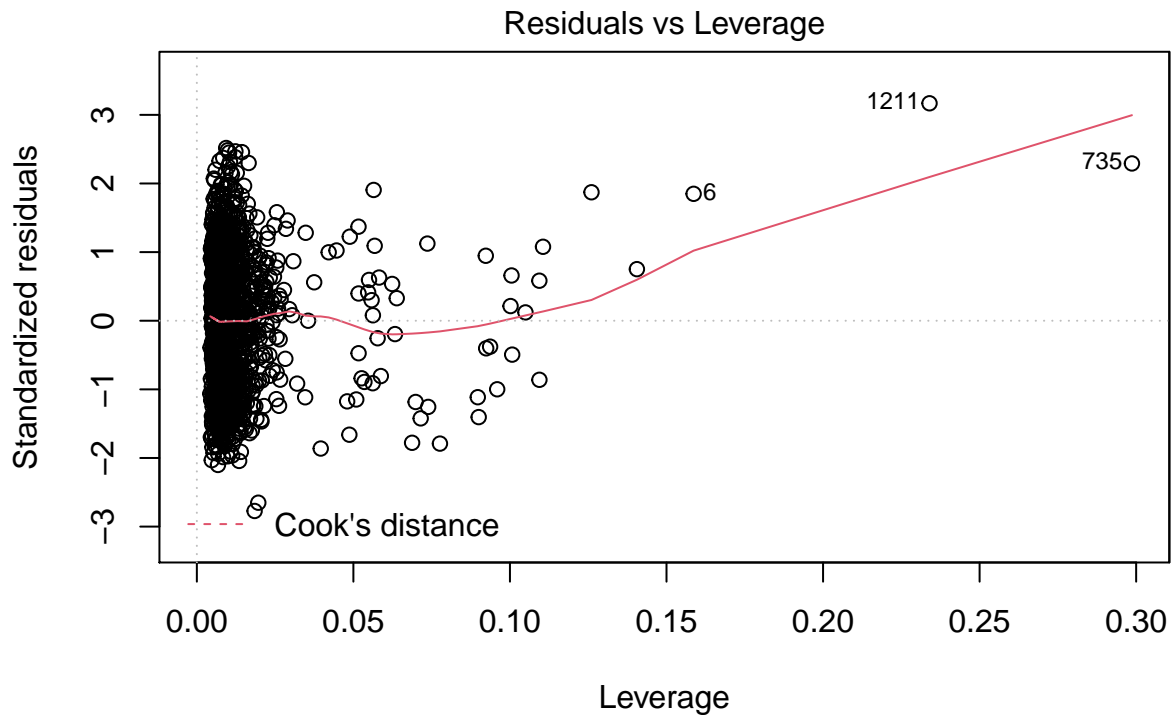
(ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST + CRS_DEP_TIME + log_TAXI_



(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_
Scale-Location



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

(c) Box-Cox-Transformed Response (w/o Log-Transformed Predictors)

```
# adjust response with vertical shift to account for negative responses
## min(train$ARR_DELAY) = -74 --> shift up by constant 75

## transform response
train$adj_ARR_DELAY = train$ARR_DELAY + 75

## fit lm for future Box-Cox use
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)
```

Box-Cox Setup

```
# box cox transform (same as case study)
## call EnvStats library
bc_model <- EnvStats::boxcox(adj_linear_model, optimize = TRUE)
# find optimal lambda parameter
bc_lambda <- bc_model$lambda
bc_lambda
```



```
## [1] 0.6981479
```

```
#plot(bc_model)
```

```
# add Box-Cox response as variable to train set
```

```
train <- train %>%
```

```
  mutate(bc_adj_ARR_DELAY = ((adj_ARR_DELAY^bc_lambda) - 1)/bc_lambda)
```

```
# plot training ARR_DELAY
```

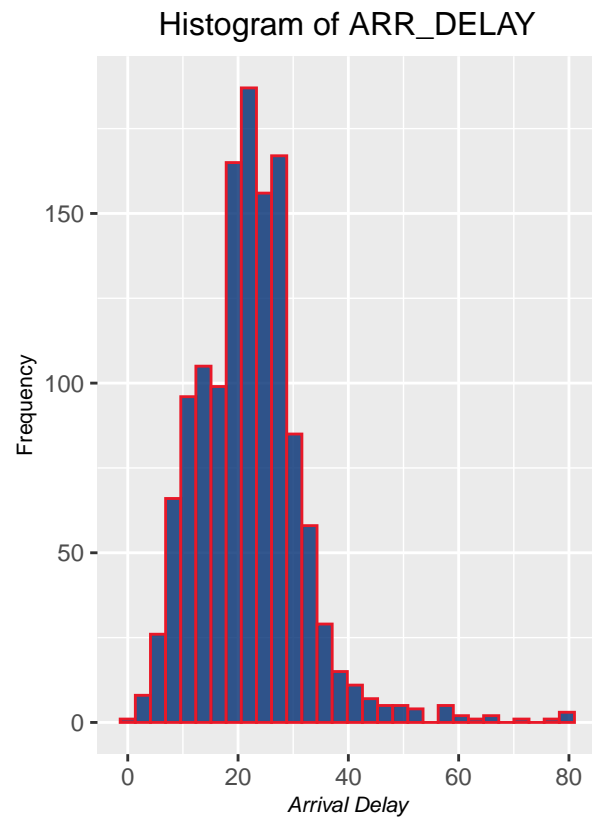
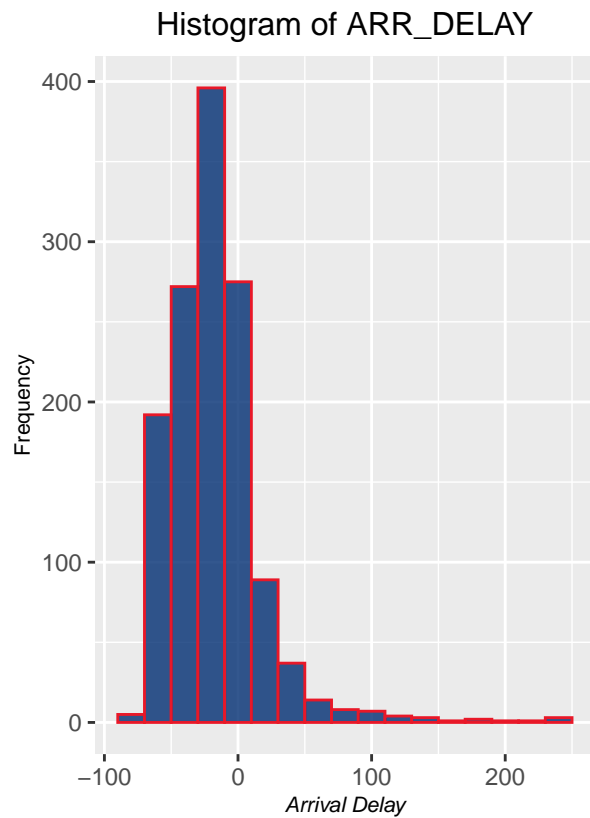
```
ptrain_ARRDELAY <- ggplot(data = train, aes(x = ARR_DELAY)) +  
  geom_histogram(binwidth = 20, fill = "#002D72", color = "#E81828", alpha = 0.8) +  
  labs(x = "Arrival Delay",  
       y = "Frequency",  
       title = "Histogram of ARR_DELAY") +  
  theme(plot.title = element_text(size = 12,hjust = 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))
```

```
# plot Box-Cox training ARR_DELAY
```

```
ptrain_bcARRDELAY <- ggplot(data = train, aes(x = bc_adj_ARR_DELAY)) +  
  geom_histogram(fill = "#002D72", color = "#E81828", alpha = 0.8) +  
  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))
```

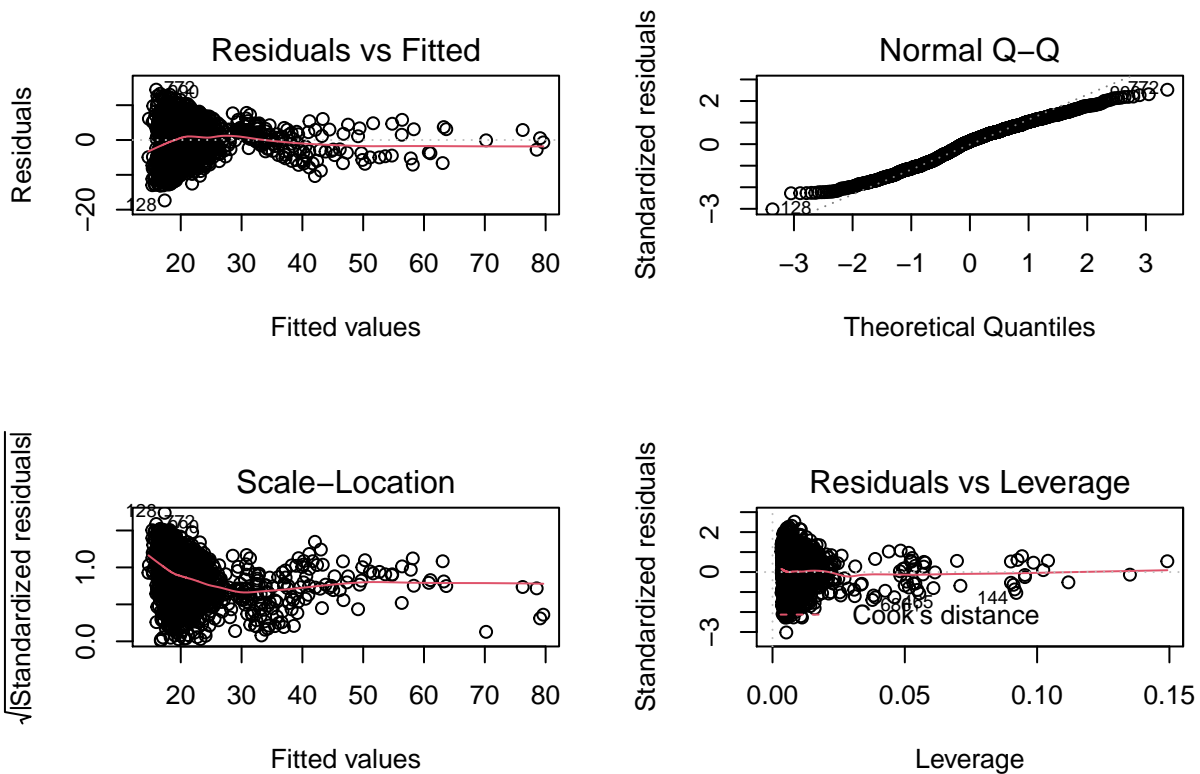
```
ptrain_ARRDELAY + ptrain_bcARRDELAY
```

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



Final Box-Cox Linear Model

```
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)
```



```
summary(bc_adj_linear_model)
```

```
##
## Call:
## lm(formula = bc_adj_ARR_DELAY ~ DEP_DELAY + OP_CARRIER + DEST +
##     CRS_DEP_TIME + CRS_ARR_TIME + TAXI_OUT + TAXI_IN + TYPE_DELAY,
##     data = train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -17.3556  -4.6043   0.7628   4.3534  14.4508
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    21.9481352    1.5512203   14.149 < 2e-16 ***
## DEP_DELAY       0.2019305    0.0085688   23.566 < 2e-16 ***
## OP_CARRIERAS  -0.3865272    0.5255474  -0.735  0.462183
## OP_CARRIERB6   0.4704172    0.4324948   1.088  0.276937
## OP_CARRIERDL  -0.3935191    0.4384069  -0.898  0.369559
## DESTSFO        -0.6404205    0.3427150  -1.869  0.061894 .
## CRS_DEP_TIME    -0.0012308    0.0003474  -3.543  0.000410 ***
## CRS_ARR_TIME    -0.0005200    0.0002776  -1.873  0.061326 .
## TAXI_OUT        0.2468270    0.0192971  12.791 < 2e-16 ***
## TAXI_IN         0.1471062    0.0194314   7.571 7.02e-14 ***
## TYPE_DELAYLATE_AIRCRAFT -0.9730991    2.0595772  -0.472  0.636667
## TYPE_DELAYNAS    5.1599555    1.4212105   3.631  0.000294 ***
## TYPE_DELAYNo Delay -5.5982679    1.4072908  -3.978 7.33e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
## Residual standard error: 5.755 on 1297 degrees of freedom
## Multiple R-squared:  0.6445, Adjusted R-squared:  0.6412
## F-statistic: 195.9 on 12 and 1297 DF,  p-value: < 2.2e-16
```

MLR Test Error Metrics & Predictions

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

```
# predictions
## baseline
test$plain_mlr_pred <- predict(plain_linear_model, test)
## log
test$log_linear_preds <- predict(log_linear_model, test)
## bc
test$adj_ARR_DELAY = test$ARR_DELAY + 77
test$bc_adj_linear_preds <- predict(bc_adj_linear_model, test)
test <- test %>%
  mutate(adj_linear_preds = ((bc_adj_linear_preds*(bc_lambda) + 1)^(1/bc_lambda)))
test$bc_mlr_pred = test$adj_linear_preds - 77

## histogram of predictions vs. actual
# actual ARR_DELAY in test set
#hist(test$ARR_DELAY)
ptest_actual <- ggplot(data = test, aes(x = ARR_DELAY)) +
  geom_histogram(fill = "#E81828", color = "#002D72", alpha = 0.8) +
  labs(x = "Arrival Delay",
       y = "Frequency",
       title = "Actual") +
  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))

# predicted ARR_DELAY in test set -- baseline lm
ptest_baseline_preds <- ggplot(data = test, aes(x = plain_mlr_pred)) +
  geom_histogram(fill = "#E81828", color = "#002D72", alpha = 0.8) +
  labs(x = "Arrival Delay",
       y = "Frequency",
       title = "Baseline Linear Model") +
  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))

# predicted ARR_DELAY in test set -- lm with log-trans
# hist(test$log_linear_preds)
ptest_lm_log_preds <- ggplot(data = test, aes(x = log_linear_preds)) +
  geom_histogram(fill = "#E81828", color = "#002D72", alpha = 0.8) +
  labs(x = "Arrival Delay",
       y = "Frequency",
       title = "Linear Model w/ Log-Transformations") +
  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))

# predicted ARR_DELAY in test set -- lm with Box-Cox model
#hist(test$log_linear_preds)
ptest_bc_preds <- ggplot(data = test, aes(x = bc_mlr_pred)) +
  geom_histogram(fill = "#E81828", color = "#002D72", alpha = 0.8) +
  labs(x = "Arrival Delay",
       y = "Frequency",
       title = "Linear Model w/ Box-Cox Response") +
  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))

mlr_patchwork <- ptest_actual + (ptest_baseline_preds / ptest_bc_preds / ptest_lm_log_preds)

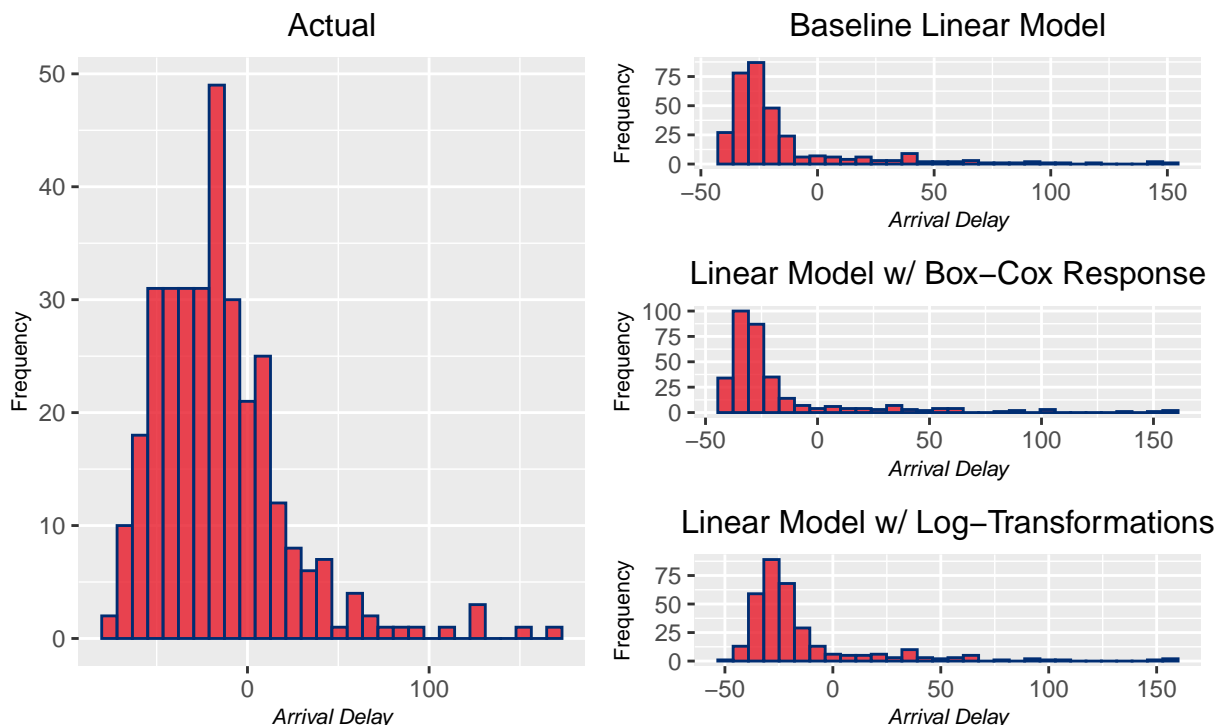
mlr_patchwork + plot_annotation(
  title = 'Comparing Distributions of ARR_DELAY',
  subtitle = 'Histograms of actual test values and MLR-predicted values'
)

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

```

Comparing Distributions of ARR_DELAY

Histograms of actual test values and MLR-predicted values



```

# test MSE calculations
plain_linear_model_MSE <- sum((test$ARR_DELAY - test$plain_mlr_pred)^2, na.rm=T)/length(test$ARR_DELAY)
plain_linear_model_MSE

## [1] 322.4588

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

## [1] 333.8962

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

## [1] 334.9217

```

(2) Generalized Additive Models

(a) Initial GAM: No Box-Cox on Response

```

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)

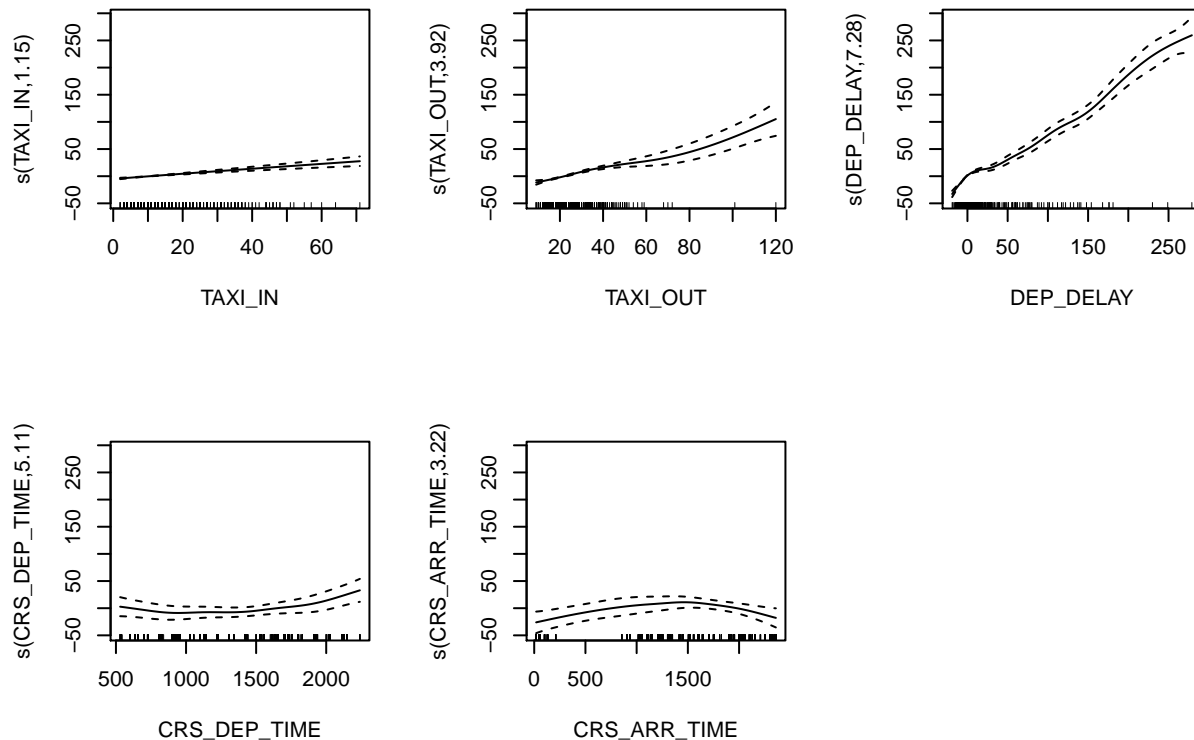
summary(gam00)

##
## 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 Tests 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

Tuned Initial GAM

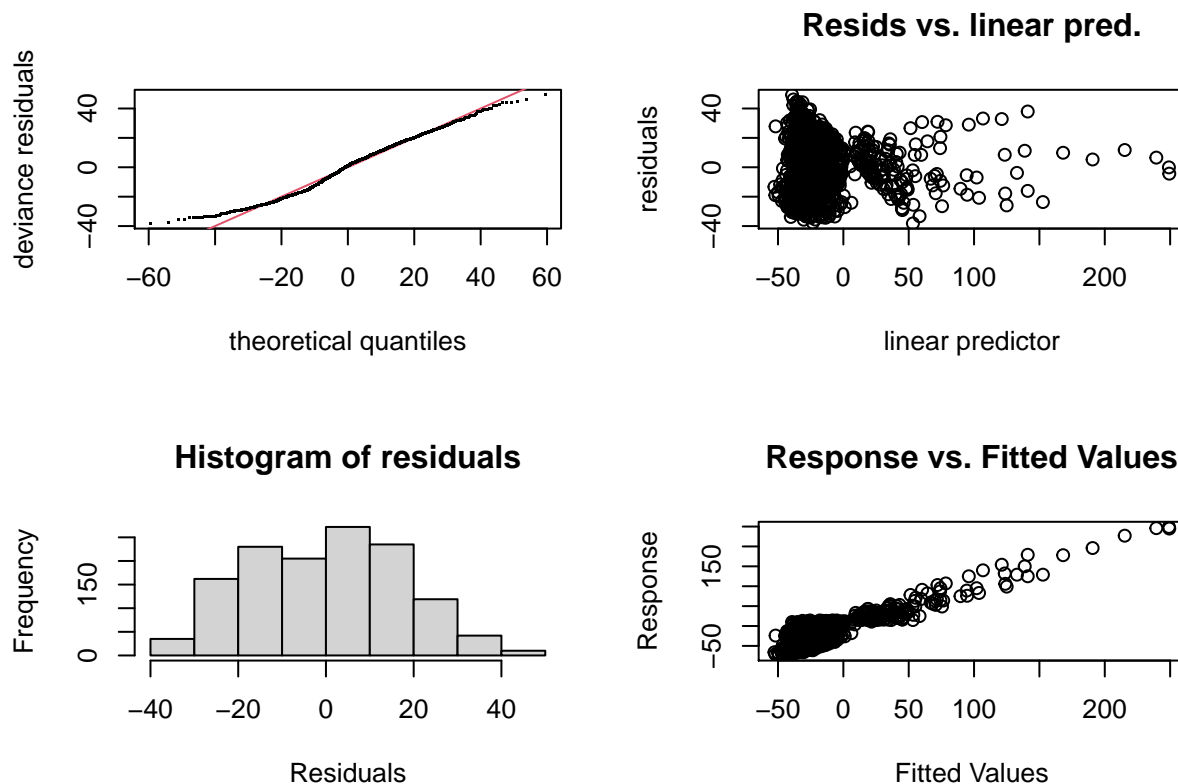
```
# final fit
summary(gam02)

##
## 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
```

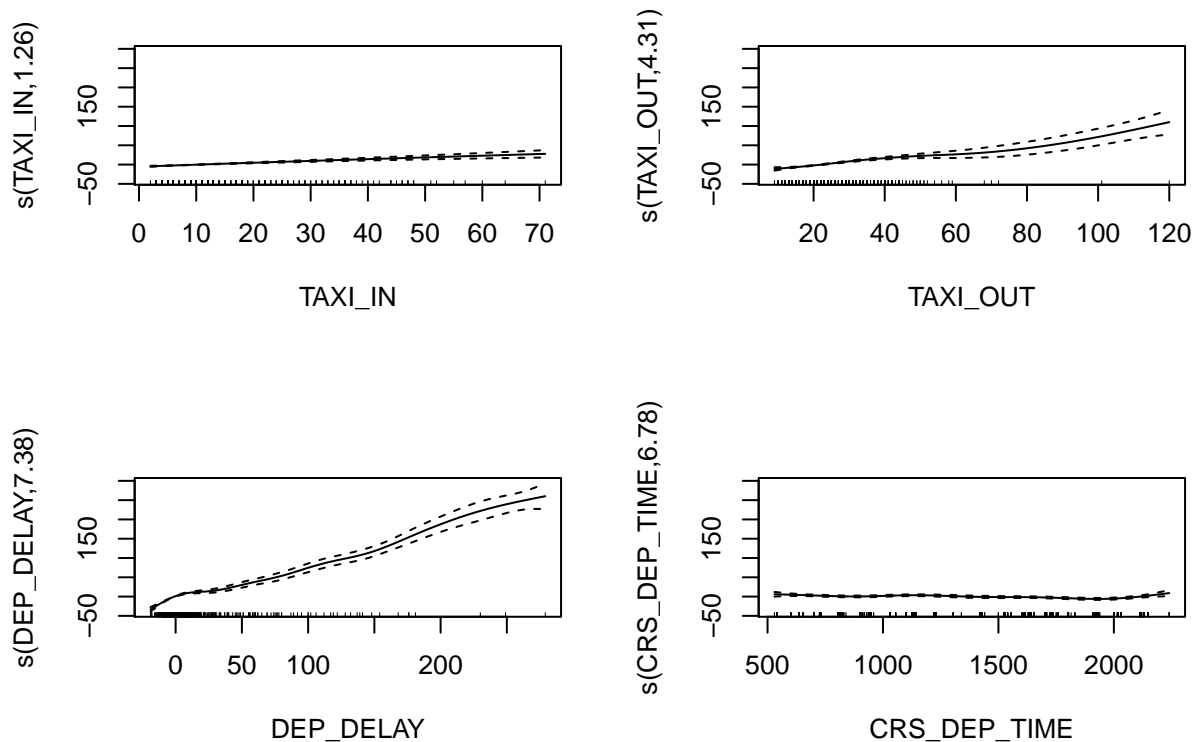
```
# diagnostic plots
par(mfrow = c(2,2))
gam.check(gam02)
```



```
##
## 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

# predictor plots
par(mfrow = c(2,2))
plot(gam02)
```



(b) Secondary GAM: Box-Cox on Response

```
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)

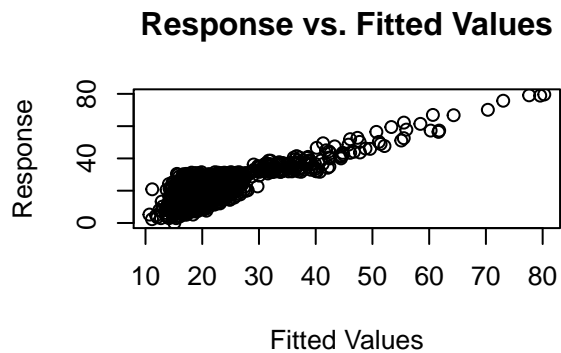
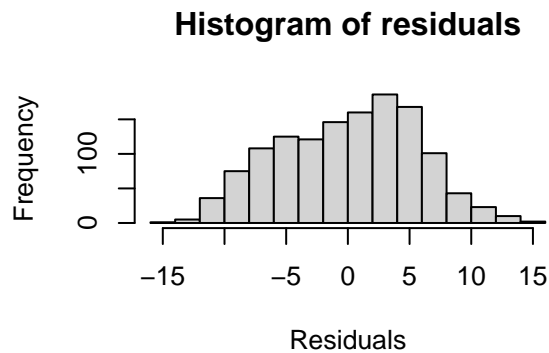
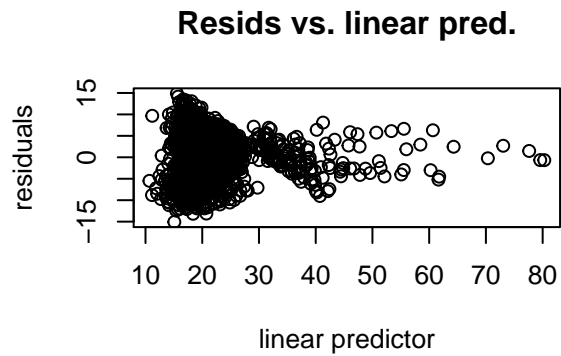
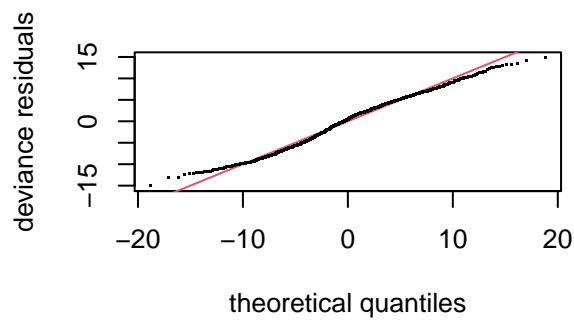
# final BC GAM fit
summary(gambc)
```

```
##
## Family: gaussian
## Link function: identity
##
```

```

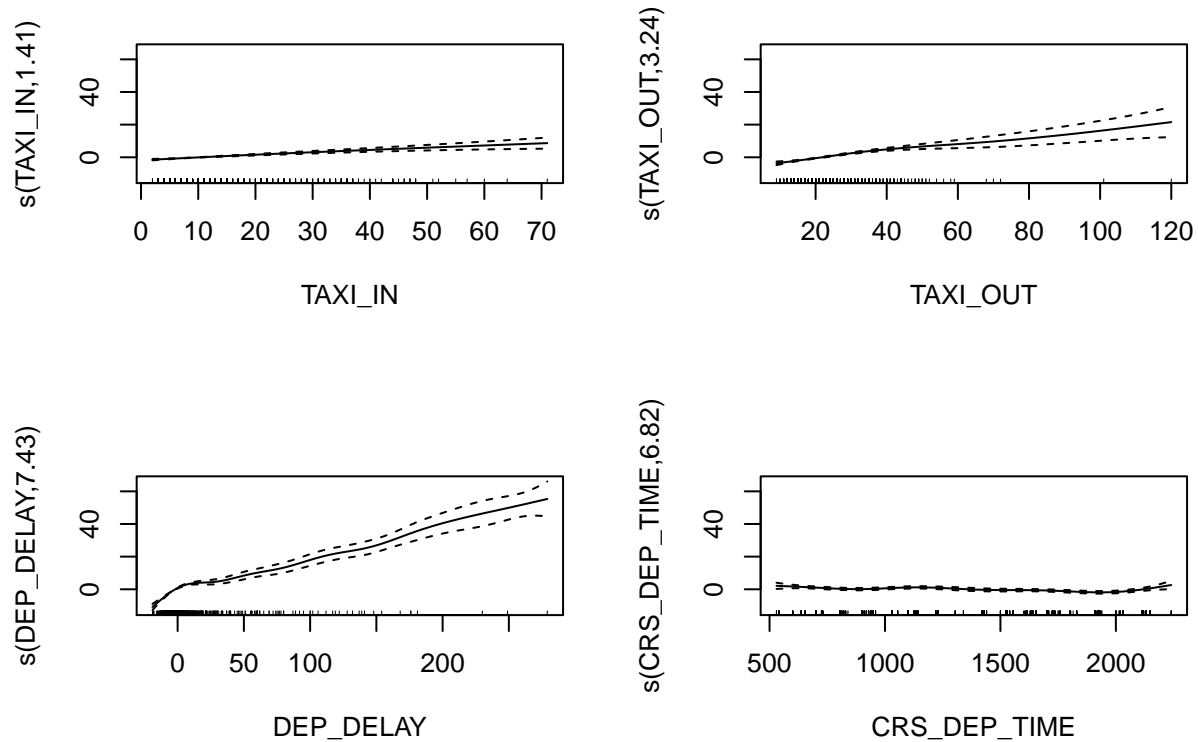
## Formula:
## bc_adj_ARR_DELAY ~ OP_CARRIER + s(TAXI_IN) + s(TAXI_OUT) + s(DEP_DELAY) +
##      s(CRS_DEP_TIME) + TYPE_DELAY
##
## Parametric coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      28.2688      1.4316  19.746 < 2e-16 ***
## OP_CARRIERAS     -0.5081      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
# diagnostic plots
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
```

```
# predictor plots
par(mfrow = c(2,2))
plot(gambc)
```

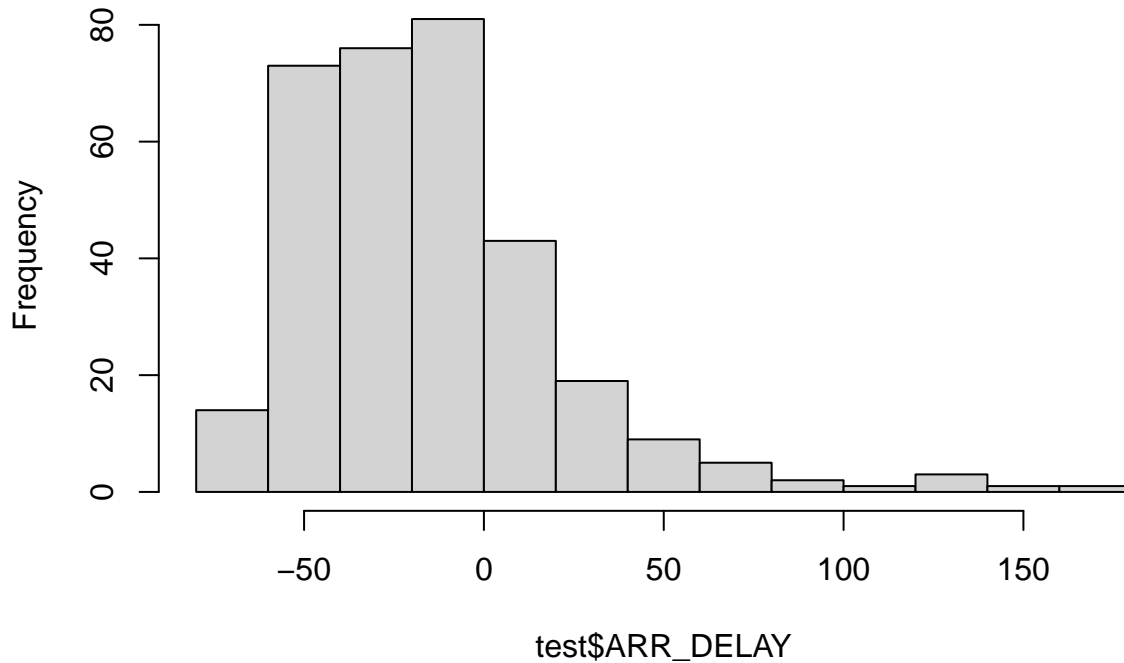


GAM Test Error Metrics & Predictions

```
# GAM predictions
## no BC
gam_preds <- predict.gam(gam02, newdata = test)
## BC
gambc_preds <- predict.gam(gambc, newdata = test)
adjgam_preds <- ((gambc_preds*(bc_lambda) + 1)^(1/bc_lambda))
bc_gam_pred = adjgam_preds - 77

## histogram of predictions vs. actual
# actual ARR_DELAY in test set
hist(test$ARR_DELAY)
```

Histogram of test\$ARR_DELAY



```
pptest_actual <- ggplot(data = test, aes(x = ARR_DELAY)) +
  geom_histogram(fill = "#40E0D0", color = "#002D72", alpha = 0.8) +
  labs(x = "Arrival Delay",
       y = "Frequency",
       title = "Actual") +
  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))

# predicted ARR_DELAY in test set -- baseline lm
pptest_gam <- ggplot(data = test, aes(x = gam_preds)) +
  geom_histogram(fill = "#40E0D0", color = "#002D72", alpha = 0.8) +
  labs(x = "Arrival Delay",
       y = "Frequency",
       title = "GAM") +
  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))

# predicted ARR_DELAY in test set -- lm with log-trans
pptest_gam_bc <- ggplot(data = test, aes(x = bc_gam_pred)) +
  geom_histogram(fill = "#40E0D0", color = "#002D72", alpha = 0.8) +
  labs(x = "Arrival Delay",
       y = "Frequency",
       title = "GAM w/ Box-Cox Response") +
  theme(plot.title = element_text(size = 12, hjust = 0.5),
        plot.subtitle = element_text(hjust = 0.5),
```

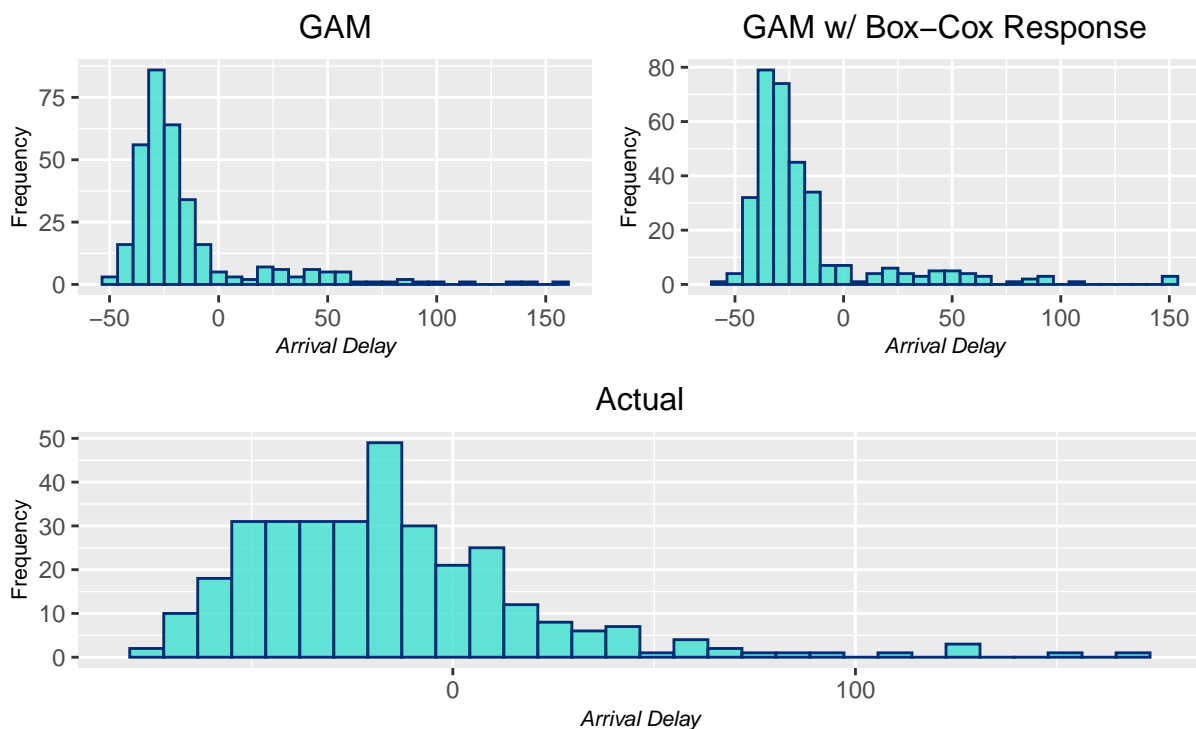
```
axis.title.x.bottom = element_text(size = 8, face = "italic"),
axis.title.y.left = element_text(size = 8))

gam_patchwork <- (ptest_gam + ptest_gam_bc) / ptest_actual
gam_patchwork + plot_annotation(
  title = 'Comparing Distributions of ARR_DELAY',
  subtitle = 'Histograms of actual test values and GAM-predicted values'
)

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

Comparing Distributions of ARR_DELAY

Histograms of actual test values and GAM-predicted values



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

## [1] 312.2953

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

## [1] 317.4533
```

(3) Tree-Based Models

(a) Random Forests

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)

```

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

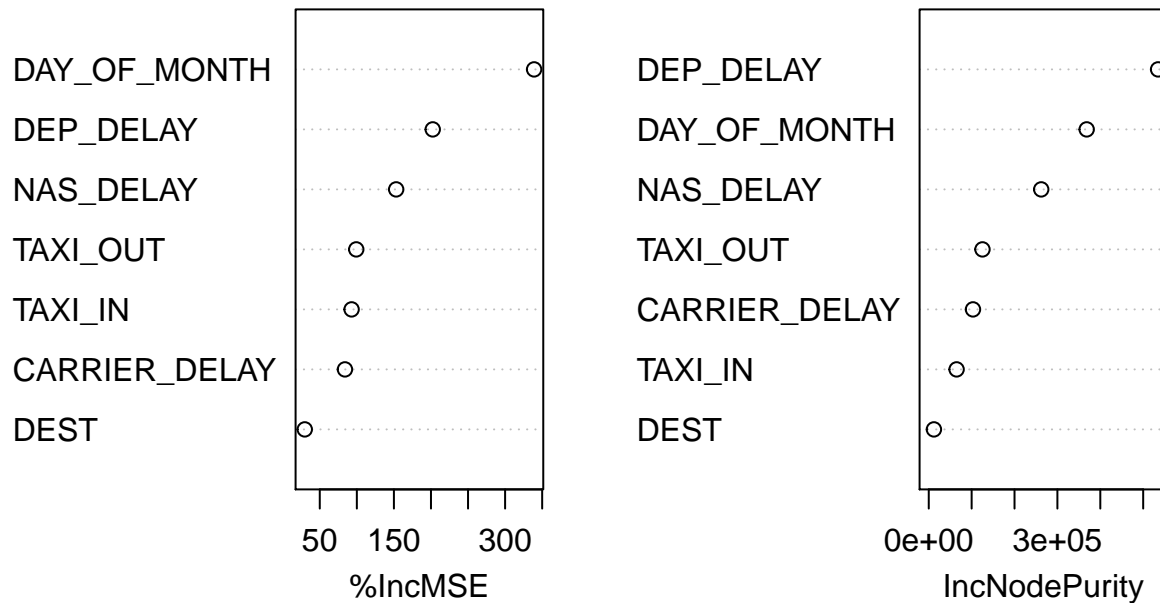
```
importance(rf.delay)
```

##		%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



(b) 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.

##	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

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##	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

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##      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

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##	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

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##      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
##

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##	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
##	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

```

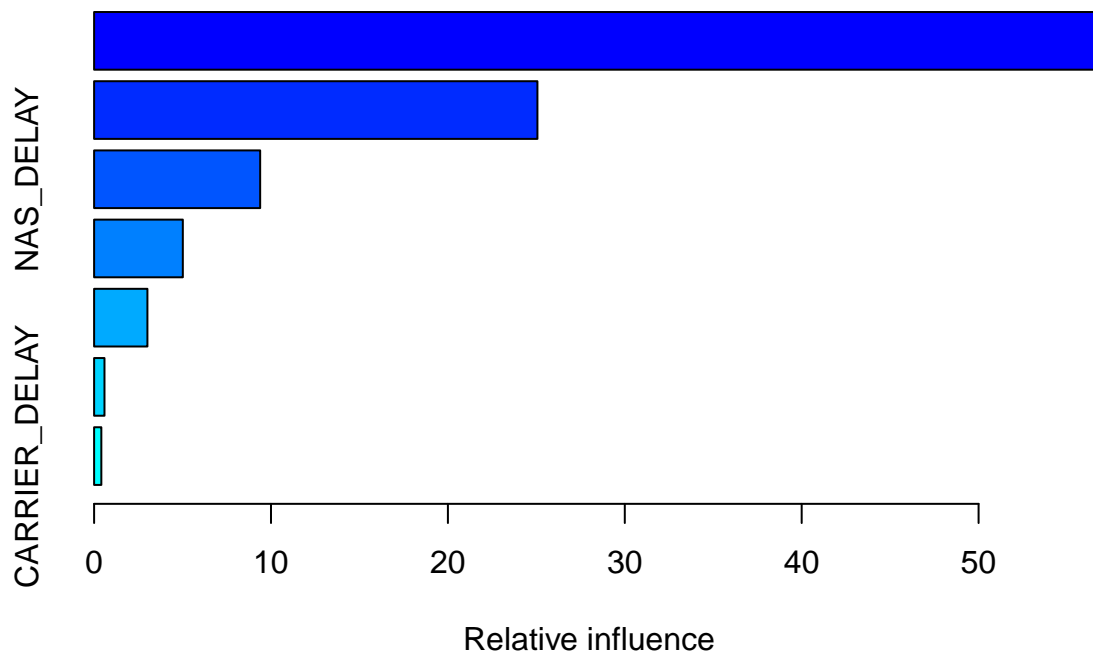
##	40	151.0315	nan	0.1000	0.7579
##	60	127.0778	nan	0.1000	-0.1092
##	80	116.4214	nan	0.1000	-0.0738
##	100	107.3310	nan	0.1000	-0.5775
##	120	102.4719	nan	0.1000	-0.3353
##	140	96.2524	nan	0.1000	-0.0157
##	150	94.1987	nan	0.1000	0.1255
##					
##	Iter	TrainDeviance	ValidDeviance	StepSize	Improve
##	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
```

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

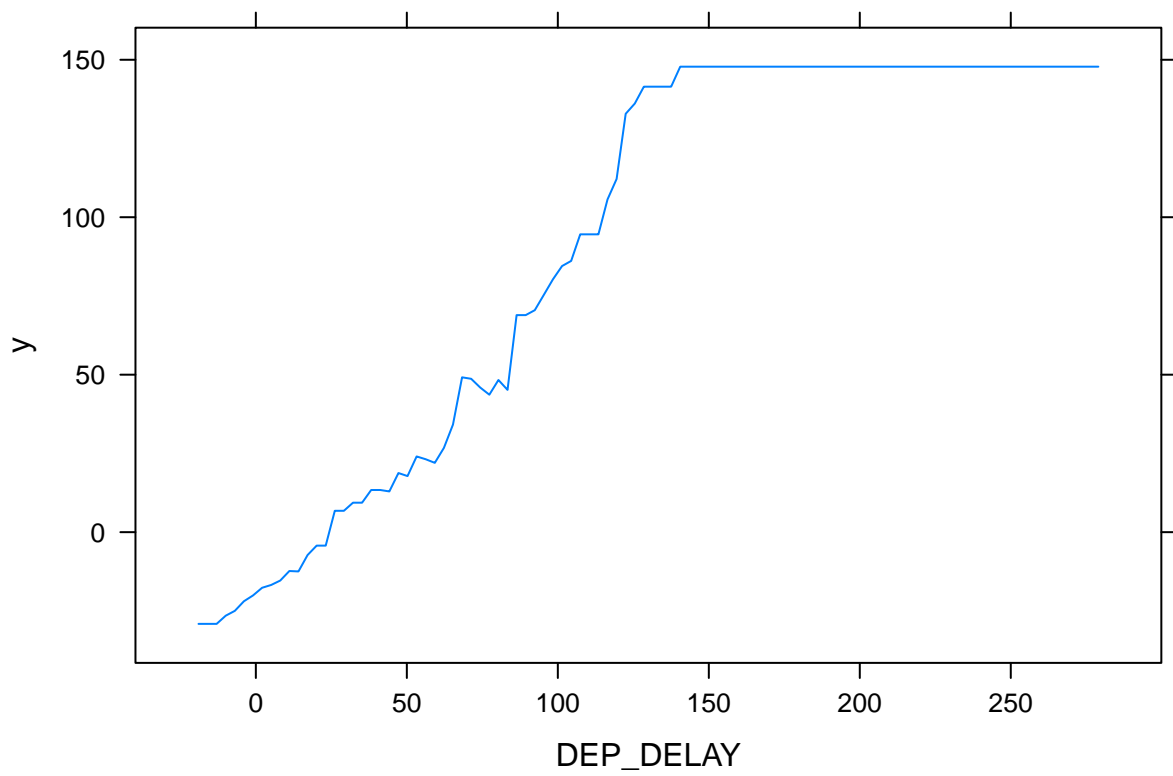
```
summary(boost.delay)
```



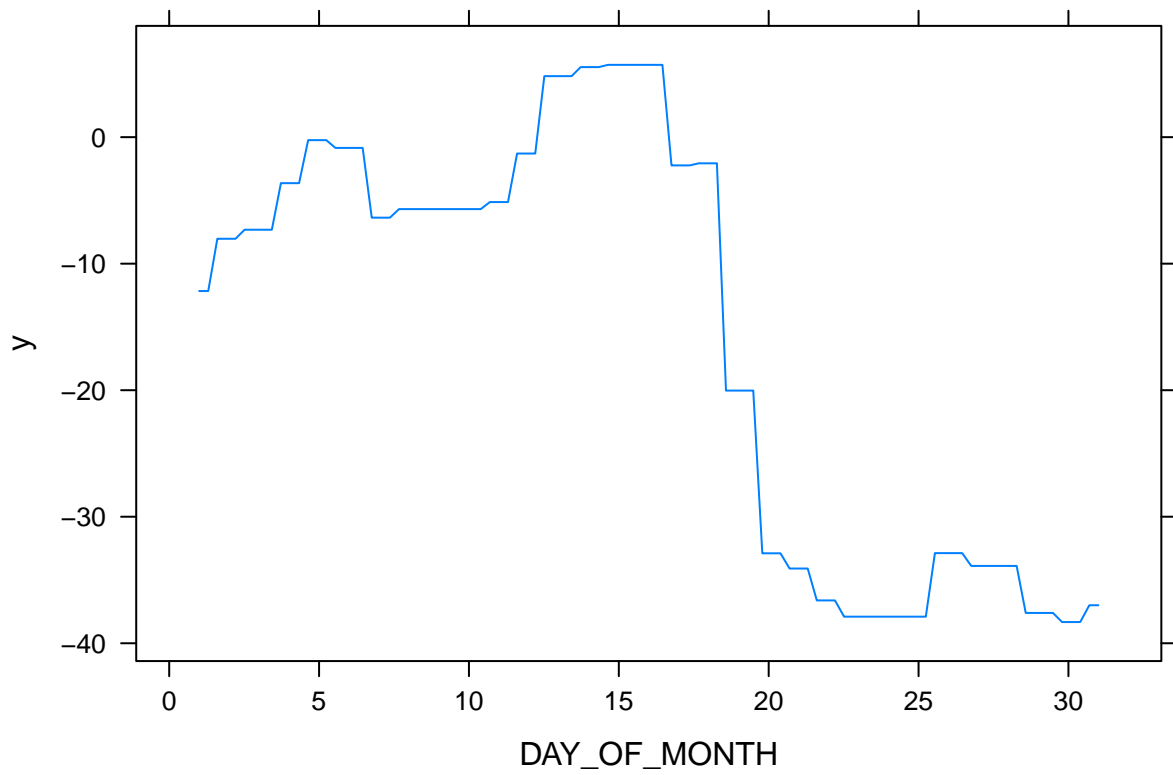
```
##           var      rel.inf
## DEP_DELAY      DEP_DELAY 56.5279213
## DAY_OF_MONTH    DAY_OF_MONTH 25.0608132
## NAS_DELAY       NAS_DELAY  9.3867001
## TAXI_OUT        TAXI_OUT  5.0157480
## TAXI_IN         TAXI_IN   3.0135950
## LATE_AIRCRAFT_DELAY LATE_AIRCRAFT_DELAY 0.5862760
## CARRIER_DELAY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



Trees Test Error Metrics & Predictions

```
# predictions
## random forest
yhat.rf <- predict(rf.delay, newdata = test)
## boosting
yhat.boost <- predict(boost.delay, newdata = test,
                      n.trees = 150)

## histogram of predictions vs. actual
# actual ARR_DELAY in test set
ptest_actual <- ggplot(data = test, aes(x = ARR_DELAY)) +
  geom_histogram(fill = "#FFFF00", color = "#002D72", alpha = 0.8) +
  labs(x = "Arrival Delay",
       y = "Frequency",
       title = "Actual") +
  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))

# predicted ARR_DELAY in test set -- baseline lm
ptest_rf <- ggplot(data = test, aes(x = yhat.rf)) +
  geom_histogram(fill = "#FFFF00", color = "#002D72", alpha = 0.8) +
  labs(x = "Arrival Delay",
       y = "Frequency",
       title = "Random Forest") +
  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))

# predicted ARR_DELAY in test set -- lm with log-trans
ptest_boost <- ggplot(data = test, aes(x = yhat.boost)) +
  geom_histogram(fill = "#FFFF00", color = "#002D72", alpha = 0.8) +
  labs(x = "Arrival Delay",
       y = "Frequency",
       title = "Boosting") +
  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))

# grid.arrange(ptest_actual, ptest_rf, ptest_boost, ncol = 2,
#               width = c(2,1,1),
#               layout_matrix = rbind(c(1,2),
#                                     c(1, 3)))

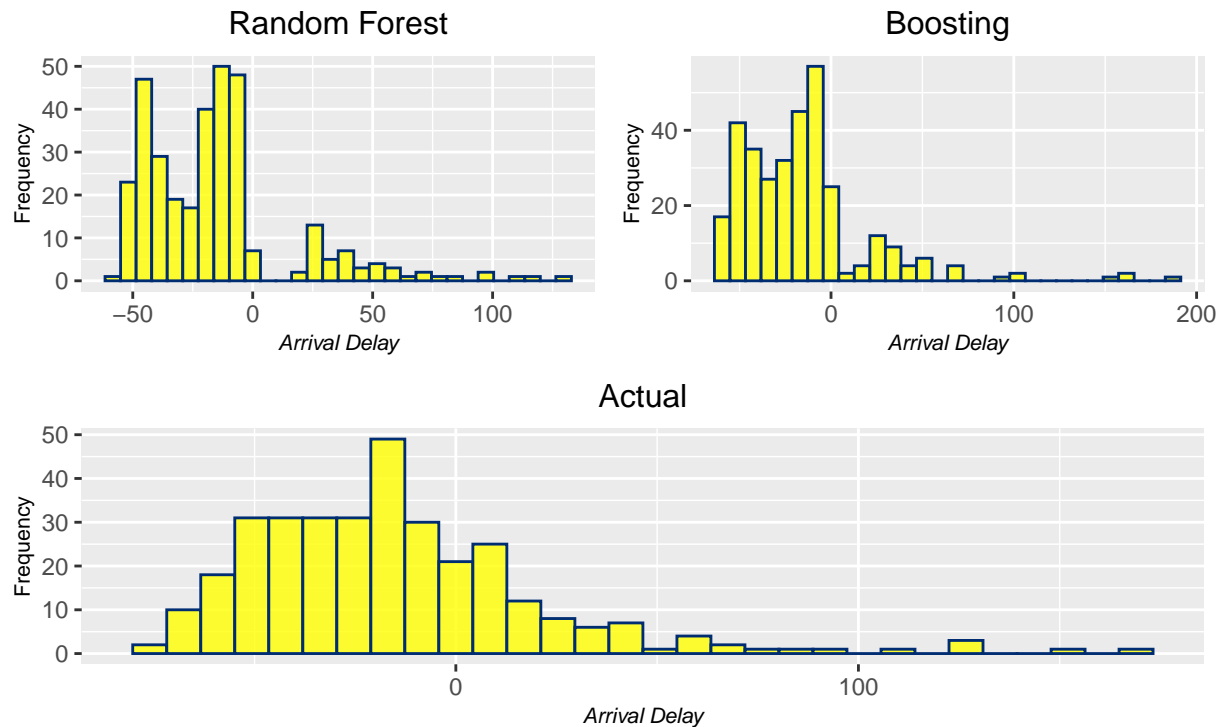
trees_patchwork <- (ptest_rf | ptest_boost) / ptest_actual
trees_patchwork + plot_annotation(
  title = 'Comparing Distributions of ARR_DELAY',
  subtitle = 'Histograms of actual test values and tree-based predicted values'
)
```



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

Comparing Distributions of ARR_DELAY

Histograms of actual test values and tree-based predicted values



```
# test MSE calculations
```

```
rf.MSE <- sum((test$ARR_DELAY - yhat.rf)^2, na.rm=T)/length(test$ARR_DELAY)
rf.MSE
```

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

```
boost_MSE <- sum((test$ARR_DELAY-yhat.boost)^2, na.rm = T)/length(test$ARR_DELAY)
boost_MSE
```

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

Test Error Tables

```
customGreen = "#71CA97"
customRed = "#ff7f7f"

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

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)

pctchange_1 <- round(-((log_linear_MSE - plain_linear_model_MSE)/plain_linear_model_MSE)*100, digits = 4)
pctchange_2 <- round(-((bc_adj_linear_model_MSE - plain_linear_model_MSE)/plain_linear_model_MSE)*100, digits = 4)
pctchange_3 <- round(-((gam_MSE - plain_linear_model_MSE)/plain_linear_model_MSE)*100, digits = 4)
pctchange_4 <- round(-((gambc_MSE - plain_linear_model_MSE)/plain_linear_model_MSE)*100, digits = 4)
pctchange_5 <- round(-((rf.MSE - plain_linear_model_MSE)/plain_linear_model_MSE)*100, digits = 4)
pctchange_6 <- round(-((boost_MSE - plain_linear_model_MSE)/plain_linear_model_MSE)*100, digits = 4)

model.pctchange <- c("---", pctchange_1, pctchange_2, pctchange_3, pctchange_4, pctchange_5, pctchange_6)

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

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

```
#colnames(errors.df4) <- c("Model Name", "Model Type", "Interactions?", "Model MSE", "Model CV MSE", "Model AIC")
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      Tree-Based Regression
## 7                      Boosting      Tree-Based Regression
##  model.mse.char model.pctchange
## 1           322.46           ---
## 2           333.90          -3.5469
## 3           334.92          -3.865
## 4           312.30           3.1519
## 5           317.45           1.5523
## 6           155.01           51.9272
## 7           129.80           59.7479
```

```
mlr.model.names <- c("Baseline Linear", "Selected Linear w/ Log-Transformed Predictors", "Selected Linear w/ Box-Cox")
mlr.mse.char <- c("322.46", "333.90", "334.92")
```

```
mlr.error.df <- data.frame(mlr.model.names, mlr.mse.char)
```

```
formattable(mlr.error.df,
  col.names = c("Model Name", "Model MSE"),
  list(
    mlr.model.names = formatter("span", style = x ~ ifelse(x == "Baseline Linear",
      style(color = "purple", font.weight = "bold"), NA)),
    mlr.mse.char = formatter("span", style = x ~ ifelse(x == "322.46",
      style(color = "purple", font.weight = "bold"), NA))
  ))
```

Model Name

Model MSE

Baseline Linear

322.46

Selected Linear w/ Log-Transformed Predictors

333.90

Selected Linear w/ Box-Cox

334.92

```
gam.model.names <- c("GAM", "GAM w/ Box-Cox")
gam.mse.char <- c("312.30", "317.45")
```

```
gam.error.df <- data.frame(gam.model.names, gam.mse.char)
```

```
formattable(gam.error.df,
  col.names = c("Model Name", "Model MSE"),
```

```

      list(
gam.model.names = formatter("span", style = x ~ ifelse(x == "GAM",
  style(color = "purple", font.weight = "bold"), NA)),
gam.mse.char = formatter("span", style = x ~ ifelse(x == "312.30",
  style(color = "purple", font.weight = "bold"), NA))
))

```

Model Name

Model MSE

GAM

312.30

GAM w/ Box-Cox

317.45

```

tree.model.names <- c("Random Forest", "Boosting")
tree.mse.char <- c("155.01", "129.80")

tree.error.df <- data.frame(tree.model.names, tree.mse.char)

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

```

Model Name

Model MSE

Random Forest

155.01

Boosting

129.80

```

formattable(errors.df,
  col.names = c("Model Name", "Model Type", "Model MSE", "Model Percent Improvement"),
  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 == "Tree-Based Regression ",
  style(color = "purple", font.weight = "bold"), NA)),
model.pctchange = formatter("span",
  style = x ~ style(font.weight = "bold",
    color = ifelse(x == "---", "black",
      ifelse(x > 0, customGreen, ifelse(x < 0,
x ~ icontext(ifelse(x>0, "arrow-up", "arrow-down"), x)

```

))	
Model Name	
Model Type	
Model MSE	
Model Percent Improvement	
Baseline Linear	
Multiple Linear Regression	
322.46	
—	
Selected Linear w/ Log-Transformed Predictors	
Multiple Linear Regression	
333.90	
-3.5469	
Selected Linear w/ Box-Cox	
Multiple Linear Regression	
334.92	
-3.865	
GAM	
Generalized Additive Model	
312.30	
3.1519	
GAM w/ Box-Cox	
Generalized Additive Model	
317.45	
1.5523	
Random Forest	
Tree-Based Regression	
155.01	
51.9272	
Boosting	
Tree-Based Regression	
129.80	
59.7479	