XGBoost Comparison

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```
library(tidyverse)
library(xgboost)
library(Metrics)

set.seed(5353)

df = read.table("auto-ins.tsv", header = TRUE)
```

```
df <- df %>%
  mutate(reg.grp = case_when(
      region == "Highlands" ~ "HLPC",
      region == "Lakeview" ~ "HLPC",
      region == "Piety Corner" ~ "HLPC",
      region == "Bleachery" ~ "BSBW",
      region == "Banks Square" ~ "BSBW",
      region == "Warrendale" ~ "BSBW",
      region == "The Lanes" ~ "HLPC",
      region == "The Chemistry" ~ "Chem",
      TRUE ~ "Other"))
df$reg.grp <- factor(df$reg.grp,</pre>
                    levels = c("HLPC", "BSBW", "Chem"))
df <- df %>%
    mutate(use.colps = case_when(
      vehicle.use == "Business" ~ "B&C",
      vehicle.use == "Commute" ~ "B&C",
      vehicle.use == "Private" ~ "Prv",
      TRUE ~ "Other"))
df$use.colps <- factor(df$use.colps,</pre>
                        levels = c("Prv", "B&C"))
df$v.bod <- fct_collapse(df$vehicle.body,</pre>
                            FML = c("Minibus", "Station Wagon"),
                            MED = c("Hatchback", "SUV", "Sedan"),
                            HST = c("Truck", "Panel Van", "Roadster"))
bks \leftarrow c(16, 20, 24, 30, 40, 50, 60, 76)
```

```
lbs <- c("17-20", "21-24", "25-30", "31-40",
         "41-50", "51-60", "61-75")
df$age.cat <- cut(df$age, breaks = bks, labels = lbs)</pre>
df1 <- df %>%
  select(claims,
         gender,
         age.cat,
         use.colps,
         reg.grp,
         v.bod,
         exposure)
head(df1)
##
     claims gender age.cat use.colps reg.grp v.bod exposure
## 1
          1 Female
                     17-20
                                   Prv
                                           HLPC
                                                  MED
                                                           1.00
## 2
          0 Male 61-75
                                           HLPC
                                                           1.00
                                   Prv
                                                  MED
## 3
          0 Female 31-40
                                   B&C
                                           HLPC
                                                  MED
                                                           0.75
## 4
          0 Female 25-30
                                   B&C
                                           HLPC
                                                           0.50
                                                  MED
## 5
          2 Male 21-24
                                   B&C
                                           Chem
                                                  MED
                                                           1.00
## 6
              Male 17-20
                                   B&C
                                           HLPC
                                                  MED
                                                           1.00
Gender <- model.matrix(~gender-1, df1)</pre>
Age <- model.matrix(~age.cat-1, df1)
Use <- model.matrix(~use.colps-1, df1)</pre>
Reg <- model.matrix(~reg.grp-1, df1)</pre>
Body <- model.matrix(~v.bod-1, df1)</pre>
df_num <- cbind(Gender, Age, Use, Reg, Body)</pre>
df_matrix <- data.matrix(df_num)</pre>
df_label <- as_vector(df1$claims)</pre>
df_exposure <- as_vector(df1$exposure)</pre>
train idx <- sample(10000,7000, replace = FALSE)</pre>
train_data <- df_matrix[train_idx,]</pre>
train_label <- df_label[train_idx]</pre>
train_exposure <- df_exposure[train_idx]</pre>
test_data <- df_matrix[-train_idx,]</pre>
test_label <- df_label[-train_idx]</pre>
test_exposure <- df_exposure[-train_idx]</pre>
dtrain <- xgb.DMatrix(data = train_data, label = train_label)</pre>
dtest <- xgb.DMatrix(data = test_data, label = test_label)</pre>
setinfo(dtrain, "base_margin", log(train_exposure)) # For offset
```

```
## [1] TRUE
```

```
setinfo(dtest, "base_margin", log(test_exposure))

## [1] TRUE

fin_mod <- glm(claims~gender + age.cat + use.colps + reg.grp + v.bod + gender:age.cat, data = df1[train

asdf <- fin_mod$coefficients

fdsa <- tibble(Coefficient = names(asdf), Value = asdf)

knitr::kable(asdf)</pre>
```

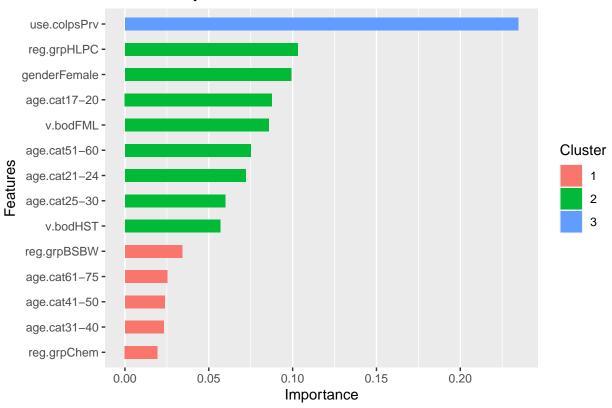
	X
(Intercept)	-2.2058848
genderMale	0.4904278
age.cat21-24	-0.1678105
age.cat25-30	-0.3144790
age.cat31-40	-0.6812966
age.cat41-50	-0.8779139
age.cat 51-60	-1.7584890
age.cat61-75	-0.3582758
use.colpsB&C	0.7022586
reg.grpBSBW	-0.3831956
reg.grpChem	-0.2166617
v.bodFML	-0.4413887
v.bodHST	0.2789719
genderMale:age.cat21-24	-0.0887434
genderMale:age.cat25-30	-0.0514458
${\tt genderMale:} {\tt age.cat 31-40}$	-0.1558785
${\tt genderMale:} {\tt age.cat41-50}$	-0.0157540
${\tt genderMale:} {\tt age.cat51-60}$	0.6719896
$\underline{{\rm gender Male: age. cat 61\text{-}75}}$	-0.8294848

```
## [1] train-poisson-nloglik:0.822982
## Will train until train_poisson_nloglik hasn't improved in 4 rounds.
##
## [11] train-poisson-nloglik:0.401141
## [21] train-poisson-nloglik:0.340956
```

```
## [31] train-poisson-nloglik:0.332176
## [41] train-poisson-nloglik:0.329866
## [51] train-poisson-nloglik:0.329513
## Stopping. Best iteration:
## [51] train-poisson-nloglik:0.329513
model2 <- xgboost(data = dtrain,</pre>
                 objective = "count:poisson",
                 \max.depth = 2,
                 early_stopping_rounds = 4,
                 print_every_n = 10,
                 nrounds = 500,
                 min_child_weight = 1,
                 gamma = 1)
## [1] train-poisson-nloglik:0.823034
## Will train until train_poisson_nloglik hasn't improved in 4 rounds.
## [11] train-poisson-nloglik:0.402943
## [21] train-poisson-nloglik:0.344469
## [31] train-poisson-nloglik:0.336040
## [41] train-poisson-nloglik:0.333399
## [51] train-poisson-nloglik:0.331704
## Stopping. Best iteration:
## [56] train-poisson-nloglik:0.331222
model3 <- xgboost(data = dtrain,</pre>
                 objective = "count:poisson",
                 \max.depth = 2,
                 #early_stopping_rounds = 4,
                 print_every_n = 10,
                 nrounds = 500,
                 min_child_weight = 1,
                 gamma = 1)
## [1] train-poisson-nloglik:0.823034
## [11] train-poisson-nloglik:0.402943
## [21] train-poisson-nloglik:0.344469
## [31] train-poisson-nloglik:0.336040
## [41] train-poisson-nloglik:0.333399
## [51] train-poisson-nloglik:0.331704
## [61] train-poisson-nloglik:0.331222
## [71] train-poisson-nloglik:0.331222
## [81] train-poisson-nloglik:0.331222
## [91] train-poisson-nloglik:0.331222
## [101]
            train-poisson-nloglik:0.331222
## [111]
            train-poisson-nloglik:0.331222
## [121]
         train-poisson-nloglik:0.331222
## [131]
           train-poisson-nloglik:0.331222
## [141]
           train-poisson-nloglik:0.331222
## [151] train-poisson-nloglik:0.331222
## [161]
         train-poisson-nloglik:0.331222
## [171]
          train-poisson-nloglik:0.331222
```

```
## [181]
            train-poisson-nloglik:0.331222
## [191]
            train-poisson-nloglik:0.331222
## [201]
            train-poisson-nloglik:0.331222
## [211]
            train-poisson-nloglik:0.331222
## [221]
            train-poisson-nloglik:0.331222
## [231]
            train-poisson-nloglik:0.331222
## [241]
            train-poisson-nloglik:0.331222
## [251]
            train-poisson-nloglik:0.331222
## [261]
            train-poisson-nloglik:0.331222
            train-poisson-nloglik:0.331222
## [271]
## [281]
            train-poisson-nloglik:0.331222
            train-poisson-nloglik:0.331222
## [291]
## [301]
            train-poisson-nloglik:0.331222
## [311]
            train-poisson-nloglik:0.331222
## [321]
            train-poisson-nloglik:0.331222
## [331]
            train-poisson-nloglik:0.331222
## [341]
            train-poisson-nloglik:0.331222
## [351]
            train-poisson-nloglik:0.331222
## [361]
            train-poisson-nloglik:0.331222
## [371]
            train-poisson-nloglik:0.331222
## [381]
            train-poisson-nloglik:0.331222
## [391]
            train-poisson-nloglik:0.331222
## [401]
            train-poisson-nloglik:0.331222
## [411]
            train-poisson-nloglik:0.331222
## [421]
            train-poisson-nloglik:0.331222
## [431]
            train-poisson-nloglik:0.331222
## [441]
            train-poisson-nloglik:0.331222
## [451]
            train-poisson-nloglik:0.331222
## [461]
            train-poisson-nloglik:0.331222
## [471]
            train-poisson-nloglik:0.331222
## [481]
            train-poisson-nloglik:0.331222
## [491]
            train-poisson-nloglik:0.331222
## [500]
            train-poisson-nloglik:0.331222
mat <- xgb.importance(names(df_matrix),</pre>
                      model = model)
xgb.ggplot.importance(mat)
```

Feature importance



```
xgb_pred <- predict(model, dtest)
xgb2_pred <- predict(model2, dtest)
xgb3_pred <- predict(model3, dtest)
glm_pred <- predict(fin_mod, df1[-train_idx,], type = "response")

xgb2_o_pred <- predict(model2, dtrain)
glm_o_pred <- predict(fin_mod, df1[train_idx,], type = "response")
xgb3_o_pred <- predict(model3, dtrain)

msep_fit <- function(x,y) {
   ans <- mean((y-x)^2)
   return(ans)
}

msep_fit(xgb_pred, test_label)

## [1] 0.1172858

msep_fit(xgb2_pred, test_label)</pre>
```

[1] 0.117169

msep_fit(xgb3_pred, test_label)

[1] 0.1171656

msep_fit(glm_pred, test_label)

[1] 0.1172515

msep_fit(xgb2_o_pred, train_label)

[1] 0.1370088

msep_fit(xgb3_o_pred, train_label)

[1] 0.1370118

msep_fit(glm_o_pred, train_label)

[1] 0.1372614