

BSA Football initial models

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```
library(janitor)

##
## Attaching package: 'janitor'
## The following objects are masked from 'package:stats':
##
##   chisq.test, fisher.test
pbp_2018 <- clean_names(read_csv("pbp_merged_2018.csv"))

## New names:
## * `` -> `...1`
## Rows: 17783 Columns: 61
## -- Column specification -----
## Delimiter: ","
## chr  (20): game_id, home_team, away_team, season_type, posteam, posteam_type...
## dbl  (36): ...1, Unnamed: 0_pbp, play_id, old_game_id, week, yardline_100, q...
## lgl   (3): run_location, run_gap, was_pressure
## date  (1): game_date
## time  (1): time
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
pbp_2019 <- clean_names(read_csv("pbp_merged_2019.csv"))

## New names:
## Rows: 17730 Columns: 61
## -- Column specification
## ----- Delimiter: "," chr
## (20): game_id, home_team, away_team, season_type, posteam, posteam_type... dbl
## (36): ...1, Unnamed: 0_pbp, play_id, old_game_id, week, yardline_100, q... lgl
## (3): run_location, run_gap, was_pressure date (1): game_date time (1): time
## i Use `spec()` to retrieve the full column specification for this data. i
## Specify the column types or set `show_col_types = FALSE` to quiet this message.
## * `` -> `...1`
pbp_2020 <- clean_names(read_csv("pbp_merged_2020.csv"))

## New names:
## Rows: 18142 Columns: 61
## -- Column specification
## ----- Delimiter: "," chr
## (20): game_id, home_team, away_team, season_type, posteam, posteam_type... dbl
```

```
## (36): ...1, Unnamed: 0_pbp, play_id, old_game_id, week, yardline_100, q... lgl
## (3): run_location, run_gap, was_pressure date (1): game_date time (1): time
## i Use `spec()` to retrieve the full column specification for this data. i
## Specify the column types or set `show_col_types = FALSE` to quiet this message.
## * `` -> `...1`
```

```
pbp_2021 <- clean_names(read_csv("pbp_merged_2021.csv"))
```

```
## New names:
## * `` -> `...1`

## Warning: One or more parsing issues, call `problems()` on your data frame for details,
## e.g.:
##   dat <- vroom(...)
##   problems(dat)
```

```
## Rows: 18849 Columns: 61
## -- Column specification -----
## Delimiter: ","
## chr  (20): game_id, home_team, away_team, season_type, posteam, posteam_type...
## dbl  (36): ...1, Unnamed: 0_pbp, play_id, old_game_id, week, yardline_100, q...
## lgl   (3): run_location, run_gap, was_pressure
## date  (1): game_date
## time  (1): time
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

```
pbp_2022 <- clean_names(read_csv("pbp_merged_2022.csv"))
```

```
## New names:
## Rows: 18155 Columns: 61
## -- Column specification
## ----- Delimiter: "," chr
## (20): game_id, home_team, away_team, season_type, posteam, posteam_type... dbl
## (36): ...1, Unnamed: 0_pbp, play_id, old_game_id, week, yardline_100, q... lgl
## (3): run_location, run_gap, was_pressure date (1): game_date time (1): time
## i Use `spec()` to retrieve the full column specification for this data. i
## Specify the column types or set `show_col_types = FALSE` to quiet this message.
## * `` -> `...1`
```

```
pbp_2023 <- clean_names(read_csv("pbp_merged_2023.csv"))
```

```
## New names:
## Rows: 16000 Columns: 61
## -- Column specification
## ----- Delimiter: "," chr
## (20): game_id, home_team, away_team, season_type, posteam, posteam_type... dbl
## (36): ...1, Unnamed: 0_pbp, play_id, old_game_id, week, yardline_100, q... lgl
## (3): run_location, run_gap, was_pressure date (1): game_date time (1): time
## i Use `spec()` to retrieve the full column specification for this data. i
## Specify the column types or set `show_col_types = FALSE` to quiet this message.
## * `` -> `...1`
```

```
full <- pbp_2018 %>% full_join(pbp_2019) %>% full_join(pbp_2020) %>% full_join(pbp_2021) %>% full_join(pbp_2022) %>% full_join(pbp_2023)
```

```
## Joining with `by` = join_by(x1, unnamed_0_pbp, play_id, game_id, old_game_id,
## home_team, away_team, season_type, week, posteam, posteam_type, defteam,
```

```

## side_of_field, yardline_100, game_date, quarter_seconds_remaining,
## game_seconds_remaining, qtr, down, time, ydstogo, play_type, yards_gained,
## no_huddle, qb_scramble, pass_length, pass_location, air_yards,
## yards_after_catch, run_location, run_gap, total_home_score, total_away_score,
## posteam_score, defteam_score, score_differential, no_score_prob, ep, epa, wp,
## def_wp, home_wp, away_wp, wpa, season, play_type_nfl, out_of_bounds,
## unnamed_0_pbp_part, nflverse_game_id, possession_team, offense_formation,
## offense_personnel, defenders_in_box, defense_personnel, number_of_pass_rushers,
## ngs_air_yards, time_to_throw, was_pressure, route, defense_man_zone_type,
## defense_coverage_type)`
## Joining with `by = join_by(x1, unnamed_0_pbp, play_id, game_id, old_game_id,
## home_team, away_team, season_type, week, posteam, posteam_type, defteam,
## side_of_field, yardline_100, game_date, quarter_seconds_remaining,
## game_seconds_remaining, qtr, down, time, ydstogo, play_type, yards_gained,
## no_huddle, qb_scramble, pass_length, pass_location, air_yards,
## yards_after_catch, run_location, run_gap, total_home_score, total_away_score,
## posteam_score, defteam_score, score_differential, no_score_prob, ep, epa, wp,
## def_wp, home_wp, away_wp, wpa, season, play_type_nfl, out_of_bounds,
## unnamed_0_pbp_part, nflverse_game_id, possession_team, offense_formation,
## offense_personnel, defenders_in_box, defense_personnel, number_of_pass_rushers,
## ngs_air_yards, time_to_throw, was_pressure, route, defense_man_zone_type,
## defense_coverage_type)`
## Joining with `by = join_by(x1, unnamed_0_pbp, play_id, game_id, old_game_id,
## home_team, away_team, season_type, week, posteam, posteam_type, defteam,
## side_of_field, yardline_100, game_date, quarter_seconds_remaining,
## game_seconds_remaining, qtr, down, time, ydstogo, play_type, yards_gained,
## no_huddle, qb_scramble, pass_length, pass_location, air_yards,
## yards_after_catch, run_location, run_gap, total_home_score, total_away_score,
## posteam_score, defteam_score, score_differential, no_score_prob, ep, epa, wp,
## def_wp, home_wp, away_wp, wpa, season, play_type_nfl, out_of_bounds,
## unnamed_0_pbp_part, nflverse_game_id, possession_team, offense_formation,
## offense_personnel, defenders_in_box, defense_personnel, number_of_pass_rushers,
## ngs_air_yards, time_to_throw, was_pressure, route, defense_man_zone_type,
## defense_coverage_type)`
## Joining with `by = join_by(x1, unnamed_0_pbp, play_id, game_id, old_game_id,
## home_team, away_team, season_type, week, posteam, posteam_type, defteam,
## side_of_field, yardline_100, game_date, quarter_seconds_remaining,
## game_seconds_remaining, qtr, down, time, ydstogo, play_type, yards_gained,
## no_huddle, qb_scramble, pass_length, pass_location, air_yards,
## yards_after_catch, run_location, run_gap, total_home_score, total_away_score,
## posteam_score, defteam_score, score_differential, no_score_prob, ep, epa, wp,
## def_wp, home_wp, away_wp, wpa, season, play_type_nfl, out_of_bounds,
## unnamed_0_pbp_part, nflverse_game_id, possession_team, offense_formation,
## offense_personnel, defenders_in_box, defense_personnel, number_of_pass_rushers,
## ngs_air_yards, time_to_throw, was_pressure, route, defense_man_zone_type,
## defense_coverage_type)`
## Joining with `by = join_by(x1, unnamed_0_pbp, play_id, game_id, old_game_id,
## home_team, away_team, season_type, week, posteam, posteam_type, defteam,
## side_of_field, yardline_100, game_date, quarter_seconds_remaining,
## game_seconds_remaining, qtr, down, time, ydstogo, play_type, yards_gained,
## no_huddle, qb_scramble, pass_length, pass_location, air_yards,
## yards_after_catch, run_location, run_gap, total_home_score, total_away_score,
## posteam_score, defteam_score, score_differential, no_score_prob, ep, epa, wp,
## def_wp, home_wp, away_wp, wpa, season, play_type_nfl, out_of_bounds,

```

```

## unnamed_0_pbp_part, nflverse_game_id, possession_team, offense_formation,
## offense_personnel, defenders_in_box, defense_personnel, number_of_pass_rushers,
## ngs_air_yards, time_to_throw, was_pressure, route, defense_man_zone_type,
## defense_coverage_type)`

trainIndex <- createDataPartition(full$route, p = 0.7, list = FALSE)

train <- full[trainIndex, ]
test <- full[-trainIndex, ]

train <- subset(train, select = -c(run_gap, run_location, yards_after_catch, x1, unnamed_0_pbp,play_id, o
train <- train %>%
  mutate_if(is.character, as.factor) %>%
  dplyr::select(where(~ !any(is.na(.))))

test <- subset(test, select = -c(run_gap, run_location, yards_after_catch, x1, unnamed_0_pbp,play_id, o
test <- test %>%
  mutate_if(is.character, as.factor) %>%
  dplyr::select(where(~ !any(is.na(.))))

nb <- naiveBayes(route ~ ., data = train)

y_pred <- predict(nb, newdata = test)

confusionMatrix(y_pred, test$route)

## Confusion Matrix and Statistics
##
##              Reference
## Prediction ANGLE CORNER CROSS FLAT  GO HITCH  IN  OUT POST SCREEN SLANT WHEEL
## ANGLE      586      0   332  476      2   270   70  371   0    29   44    5
## CORNER       0     36    37    1   35     2    8   18   27     0    1    1
## CROSS       60    171   928  384  234   477  304  572  162   107   88   38
## FLAT      267      0   527 1885     4   169   15  311   0   545   71    4
## GO          3    384   240   12 1735     74  100  121  373     2    7   36
## HITCH     133     68   525  249  280  2962  778 1593  356    23  821   29
## IN         16     64   130   54  118   245  262  195  108     6   11   16
## OUT        49     28   346   58   79   285  193  422   95     8   89    7
## POST        0    190   121    3  516    38   84   68  375     0    6   18
## SCREEN      31      0   321 1000     3    13    0   13    0  2177   24    0
## SLANT       47     14   300  292  122   625  232  811   38    16  999   20
## WHEEL       0      2     3    0    5     0    0    1    0     0    0    2
##
## Overall Statistics
##
##              Accuracy : 0.3866
##              95% CI : (0.3813, 0.392)
##      No Information Rate : 0.1613
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.3104
##
##      McNemar's Test P-Value : NA
##
## Statistics by Class:

```

```
##
##          Class: ANGLE Class: CORNER Class: CROSS Class: FLAT
## Sensitivity          0.49161      0.037618      0.24357      0.42705
## Specificity          0.94808      0.995811      0.90785      0.93063
## Pos Pred Value       0.26819      0.216867      0.26326      0.49631
## Neg Pred Value       0.97967      0.971061      0.89876      0.91030
## Prevalence           0.03726      0.029914      0.11909      0.13797
## Detection Rate       0.01832      0.001125      0.02901      0.05892
## Detection Prevalence 0.06830      0.005189      0.11018      0.11872
## Balanced Accuracy     0.71985      0.516714      0.57571      0.67884
##          Class: GO Class: HITCH Class: IN Class: OUT Class: POST
## Sensitivity          0.55378      0.57403      0.12805      0.09386      0.24446
## Specificity          0.95315      0.81906      0.96784      0.95501      0.96572
## Pos Pred Value       0.56203      0.37892      0.21388      0.25437      0.26427
## Neg Pred Value       0.95163      0.90908      0.94202      0.86569      0.96209
## Prevalence           0.09793      0.16129      0.06395      0.14054      0.04795
## Detection Rate       0.05423      0.09259      0.00819      0.01319      0.01172
## Detection Prevalence 0.09649      0.24434      0.03829      0.05186      0.04435
## Balanced Accuracy     0.75347      0.69655      0.54795      0.52444      0.60509
##          Class: SCREEN Class: SLANT Class: WHEEL
## Sensitivity          0.74734      0.46229      1.136e-02
## Specificity          0.95168      0.91562      9.997e-01
## Pos Pred Value       0.60776      0.28413      1.538e-01
## Neg Pred Value       0.97409      0.95919      9.946e-01
## Prevalence           0.09105      0.06755      5.501e-03
## Detection Rate       0.06805      0.03123      6.252e-05
## Detection Prevalence 0.11197      0.10990      4.064e-04
## Balanced Accuracy     0.84951      0.68896      5.055e-01
```

```
library(h2o)
```

```
## Warning: package 'h2o' was built under R version 4.4.2

##
## -----
##
## Your next step is to start H2O:
##   > h2o.init()
##
## For H2O package documentation, ask for help:
##   > ??h2o
##
## After starting H2O, you can use the Web UI at http://localhost:54321
## For more information visit https://docs.h2o.ai
##
## -----
##
## Attaching package: 'h2o'

## The following objects are masked from 'package:data.table':
##
##   hour, month, week, year

## The following objects are masked from 'package:lubridate':
##
##   day, hour, month, week, year
```

```

## The following objects are masked from 'package:stats':
##
##   cor, sd, var
## The following objects are masked from 'package:base':
##
##   %*%, %in%, &&, ||, apply, as.factor, as.numeric, colnames,
##   colnames<-, ifelse, is.character, is.factor, is.numeric, log,
##   log10, log1p, log2, round, signif, trunc
# Initialize the H2O cluster
h2o.init()

## Connection successful!
##
## R is connected to the H2O cluster:
##   H2O cluster uptime:      10 minutes 50 seconds
##   H2O cluster timezone:    America/Los_Angeles
##   H2O data parsing timezone: UTC
##   H2O cluster version:     3.44.0.3
##   H2O cluster version age:  10 months and 28 days
##   H2O cluster name:        H2O_started_from_R_naren_mlb655
##   H2O cluster total nodes: 1
##   H2O cluster total memory: 6.75 GB
##   H2O cluster total cores: 16
##   H2O cluster allowed cores: 16
##   H2O cluster healthy:     TRUE
##   H2O Connection ip:       localhost
##   H2O Connection port:     54321
##   H2O Connection proxy:    NA
##   H2O Internal Security:    FALSE
##   R Version:                R version 4.4.1 (2024-06-14 ucrt)

## Warning in h2o.clusterInfo():
## Your H2O cluster version is (10 months and 28 days) old. There may be a newer version available.
## Please download and install the latest version from: https://h2o-release.s3.amazonaws.com/h2o/latest.
# Convert the training dataset to an H2O frame
imputed_h2 <- as.h2o(train)

## |

# Define the AutoML settings, limiting to tree-based models
automl <- h2o.automl(
  y = "route", # Response variable
  training_frame = imputed_h2, # Training dataset
  max_runtime_secs = 60, # Maximum runtime (in seconds)
  exclude_algos = c("DeepLearning", "GLM", "StackedEnsemble") # Exclude non-tree-based models
)

## |
## 00:46:32.686: AutoML: XGBoost is not available; skipping it. |

# Extract the best model (one of the tree-based models)
final_model <- automl@leader

# Convert the test dataset to an H2O frame
testing <- as.h2o(test)

```

```
## |
# Make predictions on the test data
predictions <- predict(final_model, testing)
```

```
## |

## Warning in doTryCatch(return(expr), name, parentenv, handler): Test/Validation
## dataset column 'offense_personnel' has levels not trained on: ["0 RB, 4 TE, 1
## WR", "1 RB, 0 TE, 3 WR,1 DB", "1 RB, 0 TE, 3 WR,1 DL", "1 RB, 3 TE, 0 WR,1 LB",
## "2 QB, 2 RB, 2 TE, 0 WR", "2 QB, 6 OL, 1 RB, 2 TE, 0 WR", "4 RB, 1 TE, 0 WR",
## "6 OL, 0 RB, 2 TE, 2 WR"]

## Warning in doTryCatch(return(expr), name, parentenv, handler): Test/Validation
## dataset column 'defense_personnel' has levels not trained on: ["2 DL, 2 LB, 6
## DB, 1 WR", "3 DL, 2 LB, 5 DB, 1 WR", "3 DL, 5 LB, 2 DB, 1 OL", "6 DL, 5 LB, 0
## DB"]

# Display predictions
head(predictions)
```

```
## predict      ANGLE      CORNER      CROSS      FLAT      GO
## 1  HITCH 0.002257819 0.001333023 0.010658690 0.004138312 0.006233827
## 2    OUT 0.060271915 0.003943180 0.041703969 0.131594046 0.006109005
## 3    GO 0.001559473 0.020262228 0.008127529 0.002042808 0.926205195
## 4 CORNER 0.006874723 0.444505678 0.309404234 0.011244697 0.109657234
## 5  HITCH 0.005991998 0.005758866 0.153819179 0.015749394 0.009893108
## 6  POST 0.003736748 0.006241197 0.021848325 0.005808893 0.082809289
##      HITCH      IN      OUT      POST      SCREEN      SLANT
## 1 0.659019970 0.163444720 0.091427255 0.003006899 0.003577947 0.053938749
## 2 0.089692754 0.096556431 0.404995945 0.006285171 0.003276688 0.150604265
## 3 0.003450666 0.004824502 0.008503945 0.012740040 0.001817957 0.002217387
## 4 0.010944960 0.011757950 0.030068656 0.045618328 0.003156106 0.006037435
## 5 0.391989114 0.067650707 0.314920062 0.005248242 0.005118472 0.018230493
## 6 0.053097538 0.102872938 0.261832011 0.433579706 0.006513476 0.017524285
##      WHEEL
## 1 0.0009627888
## 2 0.0049666314
## 3 0.0082482700
## 4 0.0107300000
## 5 0.0056303664
## 6 0.0041355931
```

```
summary(final_model)
```

```
## Model Details:
## =====
##
## H2OMultinomialModel: gbm
## Model Key:  GBM_1_AutoML_4_20241118_04632
## Model Summary:
##  number_of_trees number_of_internal_trees model_size_in_bytes min_depth
## 1           42                504          2892587           15
##  max_depth mean_depth min_leaves max_leaves mean_leaves
## 1          15   15.00000        121         449   364.16864
##
## H2OMultinomialMetrics: gbm
```

```

## ** Reported on training data. **
##
## Training Set Metrics:
## =====
##
## Extract training frame with `h2o.getFrame("AutoML_4_20241118_04632_training_train_sid_a440_1")`
## MSE: (Extract with `h2o.mse`) 0.3421219
## RMSE: (Extract with `h2o.rmse`) 0.5849119
## Logloss: (Extract with `h2o.logloss`) 0.9227366
## Mean Per-Class Error: 0.3655024
## AUC: (Extract with `h2o.auc`) NaN
## AUCPR: (Extract with `h2o.aucpr`) NaN
## R^2: (Extract with `h2o.r2`) 0.9543798
## Confusion Matrix: Extract with `h2o.confusionMatrix(<model>,train = TRUE)`
## =====
## Confusion Matrix: Row labels: Actual class; Column labels: Predicted class
##


|        | ANGLE | CORNER | CROSS | FLAT  | GO   | HITCH | IN   | OUT  | POST | SCREEN | SLANT | WHEEL |
|--------|-------|--------|-------|-------|------|-------|------|------|------|--------|-------|-------|
| ANGLE  | 1656  | 2      | 200   | 470   | 0    | 245   | 9    | 158  | 0    | 36     | 7     | 0     |
| CORNER | 0     | 1154   | 140   | 1     | 567  | 75    | 73   | 79   | 137  | 0      | 7     | 1     |
| CROSS  | 190   | 75     | 5702  | 800   | 306  | 790   | 153  | 465  | 132  | 224    | 53    | 2     |
| FLAT   | 202   | 4      | 363   | 8421  | 14   | 317   | 25   | 235  | 9    | 657    | 53    | 0     |
| GO     | 1     | 81     | 154   | 4     | 6292 | 199   | 101  | 156  | 237  | 0      | 86    | 0     |
| HITCH  | 113   | 14     | 560   | 228   | 106  | 9230  | 214  | 905  | 105  | 16     | 549   | 2     |
| IN     | 41    | 36     | 395   | 44    | 80   | 1241  | 2090 | 510  | 153  | 2      | 183   | 1     |
| OUT    | 253   | 55     | 824   | 586   | 133  | 2381  | 218  | 5384 | 156  | 23     | 479   | 0     |
| POST   | 0     | 67     | 132   | 1     | 674  | 162   | 122  | 181  | 2219 | 0      | 21    | 2     |
| SCREEN | 8     | 0      | 64    | 824   | 0    | 32    | 2    | 19   | 0    | 5846   | 5     | 0     |
| SLANT  | 30    | 3      | 90    | 144   | 28   | 978   | 50   | 408  | 28   | 24     | 3261  | 0     |
| WHEEL  | 0     | 14     | 33    | 34    | 55   | 52    | 11   | 41   | 17   | 1      | 14    | 140   |
| Totals | 2494  | 1505   | 8657  | 11557 | 8255 | 15702 | 3068 | 8541 | 3193 | 6829   | 4718  | 148   |


##


|        | Error  | Rate              |
|--------|--------|-------------------|
| ANGLE  | 0.4050 | = 1,127 / 2,783   |
| CORNER | 0.4834 | = 1,080 / 2,234   |
| CROSS  | 0.3587 | = 3,190 / 8,892   |
| FLAT   | 0.1824 | = 1,879 / 10,300  |
| GO     | 0.1394 | = 1,019 / 7,311   |
| HITCH  | 0.2335 | = 2,812 / 12,042  |
| IN     | 0.5624 | = 2,686 / 4,776   |
| OUT    | 0.4868 | = 5,108 / 10,492  |
| POST   | 0.3803 | = 1,362 / 3,581   |
| SCREEN | 0.1403 | = 954 / 6,800     |
| SLANT  | 0.3535 | = 1,783 / 5,044   |
| WHEEL  | 0.6602 | = 272 / 412       |
| Totals | 0.3117 | = 23,272 / 74,667 |


##
## Hit Ratio Table: Extract with `h2o.hit_ratio_table(<model>,train = TRUE)`
## =====
## Top-10 Hit Ratios:


| k | hit_ratio |
|---|-----------|
| 1 | 0.688323  |
| 2 | 0.883684  |
| 3 | 0.956340  |
| 4 | 0.985643  |
| 5 | 0.996196  |


```



```

## 6 6 0.999156
## 7 7 0.999839
## 8 8 0.999973
## 9 9 1.000000
## 10 10 1.000000
##
##
##
##
##
## H2OMultinomialMetrics: gbm
## ** Reported on cross-validation data. **
## ** 5-fold cross-validation on training data (Metrics computed for combined holdout predictions) **
##
## Cross-Validation Set Metrics:
## =====
##
## Extract cross-validation frame with `h2o.getFrame("AutoML_4_20241118_04632_training_train_sid_a440_1
## MSE: (Extract with `h2o.mse`) 0.472678
## RMSE: (Extract with `h2o.rmse`) 0.6875158
## Logloss: (Extract with `h2o.logloss`) 1.394882
## Mean Per-Class Error: 0.6159273
## AUC: (Extract with `h2o.auc`) NaN
## AUCPR: (Extract with `h2o.aucpr`) NaN
## R^2: (Extract with `h2o.r2`) 0.9369708
## Hit Ratio Table: Extract with `h2o.hit_ratio_table(<model>,xval = TRUE)`
## =====
## Top-10 Hit Ratios:
##      k hit_ratio
## 1 1 0.467743
## 2 2 0.691323
## 3 3 0.823483
## 4 4 0.901375
## 5 5 0.944326
## 6 6 0.971098
## 7 7 0.987156
## 8 8 0.993839
## 9 9 0.997254
## 10 10 0.998621
##
##
##
##
## Cross-Validation Metrics Summary:
##
##              mean      sd  cv_1_valid  cv_2_valid
## accuracy      0.465734 0.001663    0.466988    0.466318
## auc            NA    0.000000         NA         NA
## err            0.534265 0.001663    0.533012    0.533682
## err_count     7978.400000 24.684004 7960.000000 7970.000000
## logloss        1.402171 0.005930    1.405412    1.395664
## max_per_class_error 0.990018 0.005940    0.989796    0.986667
## mean_per_class_accuracy 0.382614 0.002988    0.386587    0.384509
## mean_per_class_error 0.617386 0.002988    0.613413    0.615491
## mse            0.470717 0.002075    0.471573    0.469794

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## pr_auc          NA 0.000000          NA          NA
## r2              0.937224 0.000707    0.936762    0.937761
## rmse            0.686087 0.001513    0.686712    0.685415
##               cv_3_valid cv_4_valid cv_5_valid
## accuracy        0.463336 0.464743 0.467287
## auc              NA      NA      NA
## err              0.536664 0.535257 0.532713
## err_count        8014.000000 7993.000000 7955.000000
## logloss          1.410315 1.401923 1.397542
## max_per_class_error 0.989011 0.984615 1.000000
## mean_per_class_accuracy 0.381188 0.378905 0.381880
## mean_per_class_error 0.618812 0.621095 0.618120
## mse              0.473011 0.471590 0.467616
## pr_auc          NA      NA      NA
## r2              0.936914 0.936514 0.938170
## rmse            0.687758 0.686724 0.683824
##
## Scoring History:
##      timestamp    duration number_of_trees training_rmse
## 1  2024-11-18 00:47:24 52.197 sec           0      0.91667
## 2  2024-11-18 00:47:25 52.710 sec           5      0.81341
## 3  2024-11-18 00:47:25 53.242 sec          10      0.75006
## 4  2024-11-18 00:47:26 53.768 sec          15      0.69621
## 5  2024-11-18 00:47:26 54.266 sec          20      0.66711
## 6  2024-11-18 00:47:27 54.777 sec          25      0.64051
## 7  2024-11-18 00:47:27 55.276 sec          30      0.62080
## 8  2024-11-18 00:47:28 55.761 sec          35      0.60474
## 9  2024-11-18 00:47:28 56.250 sec          40      0.58918
## 10 2024-11-18 00:47:29 56.444 sec          42      0.58491
##      training_logloss training_classification_error training_auc training_pr_auc
## 1          2.48491          0.88837          NA          NA
## 2          1.76351          0.46277          NA          NA
## 3          1.48965          0.43606          NA          NA
## 4          1.29128          0.41143          NA          NA
## 5          1.18913          0.38803          NA          NA
## 6          1.09971          0.36538          NA          NA
## 7          1.03552          0.34722          NA          NA
## 8          0.98442          0.33157          NA          NA
## 9          0.93571          0.31560          NA          NA
## 10         0.92274          0.31168          NA          NA
##
## Variable Importances: (Extract with `h2o.varimp`)
## =====
##
## Variable Importances:
##      variable relative_importance scaled_importance percentage
## 1      ngs_air_yards      49820.445312      1.000000    0.467103
## 2      time_to_throw      19372.673828      0.388850    0.181633
## 3      possession_team      15778.593750      0.316709    0.147936
## 4      defense_personnel      7446.482910      0.149466    0.069816
## 5      defense_coverage_type      4339.363770      0.087100    0.040685
## 6      offense_formation      2779.407959      0.055789    0.026059
## 7      offense_personnel      2279.188721      0.045748    0.021369
## 8      defenders_in_box      1580.870361      0.031731    0.014822

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## 9	defense_man_zone_type	1359.008423	0.027278	0.012742
## 10	number_of_pass_rushers	1289.365967	0.025880	0.012089
## 11	was_pressure	612.861511	0.012301	0.005746