

## Surface and Time

```
library(TennisBT)

##
## Attaching package: 'TennisBT'

## The following object is masked from 'package:stats':
##
##     predict

library(tidyverse)

## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr     1.1.3     v readr     2.1.4
## v forcats   1.0.0     v stringr   1.5.0
## v ggplot2   3.4.4     v tibble    3.2.1
## v lubridate 1.9.3     v tidyrr    1.3.0
## v purrr    1.0.2

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()   masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors

library(BradleyTerry2)

data <- TennisTidyr(2013, 2022, 2023, 20)
train <- data$train
test <- data$test
players <- data$main_names
```

Set up the model as with in the surface experiment.

```
final_grass_clay <- 0.4
final_hard_clay <- 0.8
final_grass_hard <- 0.5

predict_with_surface <- function(time_weights) {

  weighting <- as.data.frame(diag(3))
  rownames(weighting) <- c("Grass", "Hard", "Clay")
  colnames(weighting) <- c("Grass", "Hard", "Clay")
  weighting['Grass', 'Hard'] <- final_grass_hard
  weighting['Hard', 'Grass'] <- final_grass_hard
  weighting['Clay', 'Hard'] <- final_hard_clay
  weighting['Hard', 'Clay'] <- final_hard_clay
  weighting['Clay', 'Grass'] <- final_grass_clay
  weighting['Grass', 'Clay'] <- final_grass_clay
```

```

for (i in 1:nrow(train)) {
  train[i, 'grass_weight'] <- weighting[train[i, 'surface'], 'Grass']
  train[i, 'clay_weight'] <- weighting[train[i, 'surface'], 'Clay']
  train[i, 'hard_weight'] <- weighting[train[i, 'surface'], 'Hard']
}

btm_grass <- BTm(outcome = 1, factor(winner_name, levels=players), factor(loser_name, levels=players))
btm_clay <- BTm(outcome = 1, factor(winner_name, levels=players), factor(loser_name, levels=players),
btm_hard <- BTm(outcome = 1, factor(winner_name, levels=players), factor(loser_name, levels=players),

# Create a dataframe with the Bradley-Terry estimates for each surface
btm_grass_df <- as.data.frame(BTabilities(btm_grass))
btm_clay_df <- as.data.frame(BTabilities(btm_clay))
btm_hard_df <- as.data.frame(BTabilities(btm_hard))

coeff_frames <- list('Grass' = btm_grass_df, 'Clay' = btm_clay_df,
                      'Hard' = btm_hard_df)

draws <- test[, c("winner_name", "loser_name", "surface")]
draws$pred <- NA

# predict the test set using a different model based on the surface
for (i in 1:nrow(draws)) {
  player1 <- draws$winner_name[i]
  player2 <- draws$loser_name[i]
  draw_surface <- draws$surface[i]

  pred <- predict(player1, player2, as.data.frame(coeff_frames[draw_surface]))
  draws$pred[i] <- pred[2]
}

return(draws)
}

```

Test this surface model with all of the time weighting parameters.

```

scores <- data.frame(matrix(0, ncol=5, nrow=10), row.names = seq(0.1,1,0.1))
colnames(scores) <- c("Accuracy", "Avg Prob", "Avg Prob SE", "Avg Log Prob", "Avg Log Prob SE")

row_count = 1
for (recency_weighting in seq(0.1,1,0.1)) {
  cat("w=",recency_weighting,"\\n")
  weights <- get_recency_weights(train, recency_weighting)

  predictions <- predict_with_surface(weights)
  score <- score_predictions(predictions)

  scores[row_count, ]["Accuracy"] <- score$accuracy
  scores[row_count, ]["Avg Prob"] <- score$avg_probability
  scores[row_count, ]["Avg Prob SE"] <- score$avg_prob_se
  scores[row_count, ]["Avg Log Prob"] <- score$avg_log_probability
  scores[row_count, ]["Avg Log Prob SE"] <- score$avg_log_prob_se

```

```

    row_count <- row_count + 1
}

## w= 0.1

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.2

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.3

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.4

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.5

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.6

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
```

```

## w= 0.7

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.8

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.9

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 1

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

```

scores

	Accuracy	Avg	Prob	Avg	Prob	SE	Avg	Log	Prob	Avg	Log	Prob	SE
## 0.1	0.6435331	0.7091272	0.004202919	-0.3615849	0.005925624								
## 0.2	0.6460568	0.6998755	0.004051684	-0.3739154	0.005776137								
## 0.3	0.6429022	0.6947137	0.003921399	-0.3803626	0.005622974								
## 0.4	0.6384858	0.6907279	0.003803034	-0.3851870	0.005473270								
## 0.5	0.6365931	0.6863991	0.003706887	-0.3908332	0.005362498								
## 0.6	0.6365931	0.6822523	0.003619995	-0.3963448	0.005256236								
## 0.7	0.6359621	0.6790322	0.003555305	-0.4006469	0.005172882								
## 0.8	0.6416404	0.6749851	0.003531601	-0.4067163	0.005166466								
## 0.9	0.6365931	0.6752287	0.003514150	-0.4061030	0.005138621								
## 1	0.6403785	0.6742218	0.003528770	-0.4078452	0.005172765								

Then we have that the best time weighting is between 0.1 and 0.3, so we can split it down further.

```

scores <- data.frame(matrix(0, ncol=5, nrow=21), row.names = seq(0.1,0.3,0.01))
colnames(scores) <- c("Accuracy", "Avg Prob", "Avg Prob SE", "Avg Log Prob", "Avg Log Prob SE")

row_count = 1
for (recency_weighting in seq(0.1,0.3,0.01)) {
  cat("w=",recency_weighting,"\\n")
  weights <- get_recency_weights(train, recency_weighting)

  predictions <- predict_with_surface(weights)
  score <- score_predictions(predictions)

  scores[row_count, ]["Accuracy"] <- score$accuracy
  scores[row_count, ]["Avg Prob"] <- score$avg_probability
  scores[row_count, ]["Avg Prob SE"] <- score$avg_prob_se
  scores[row_count, ]["Avg Log Prob"] <- score$avg_log_probability
  scores[row_count, ]["Avg Log Prob SE"] <- score$abg_log_prob_se

  row_count <- row_count + 1
}

## w= 0.1

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.11

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.12

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.13

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

```

```

## w= 0.14

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.15

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.16

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.17

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.18

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.19

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.2

```



```

## w= 0.27

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.28

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.29

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

## w= 0.3

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!

```

scores

	Accuracy	Avg	Prob	Avg	Prob	SE	Avg	Log	Prob	Avg	Log	Prob	SE
## 0.1	0.6435331	0.7091272	0.004202919	-0.3615849	0.005925624								
## 0.11	0.6441640	0.7078832	0.004185573	-0.3632684	0.005910398								
## 0.12	0.6441640	0.7068907	0.004168515	-0.3645707	0.005893191								
## 0.13	0.6454259	0.7055720	0.004152658	-0.3663960	0.005879804								
## 0.14	0.6466877	0.7042927	0.004137977	-0.3681786	0.005868295								
## 0.15	0.6460568	0.7036523	0.004121617	-0.3689582	0.005848987								
## 0.16	0.6454259	0.7030486	0.004105596	-0.3696876	0.005829749								
## 0.17	0.6460568	0.7020724	0.004091921	-0.3710224	0.005817424								
## 0.18	0.6466877	0.7011207	0.004078866	-0.3723289	0.005805984								
## 0.19	0.6447950	0.7009720	0.004063478	-0.3723647	0.005784452								
## 0.2	0.6460568	0.6998755	0.004051684	-0.3739154	0.005776137								
## 0.21	0.6422713	0.7003684	0.004033749	-0.3729319	0.005744984								
## 0.22	0.6416404	0.6999032	0.004019722	-0.3734814	0.005727781								
## 0.23	0.6416404	0.6992562	0.004006816	-0.3743270	0.005713892								
## 0.24	0.6410095	0.6988240	0.003993009	-0.3748311	0.005696636								
## 0.25	0.6422713	0.6978149	0.003982148	-0.3762646	0.005688916								
## 0.26	0.6435331	0.6968232	0.003971297	-0.3776739	0.005680973								

```
## 0.27 0.6435331 0.6962330 0.003958853 -0.3784427 0.005667116
## 0.28 0.6441640 0.6954648 0.003947170 -0.3794990 0.005655822
## 0.29 0.6435331 0.6950860 0.003934199 -0.3799346 0.005639307
## 0.3 0.6429022 0.6947137 0.003921399 -0.3803626 0.005622974
```

Therefore the best model is with surface weighting as defined above, and time weighting of  $w = 0.14$ .