# **BST 270: Final Project**

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# Table of contents

Set-Up	2
Preface	2
"Why Some Tennis Matches Take Forever"	3
Table 1: Events	3
Original	3
Re-creation	4
Table 2: Players	5
Original	5
Re-creation	6
Summary/Analysis	7
"We Watched 906 Foul Balls To Find Out Where The Most Dangerous Ones Land"	8
Figure 2	8
Original	8
Re-creation	9
Summary/Analysis	
Appendix	11
A: Session Information	
B: Codo	19

## Set-Up

## **Preface**

Tables were generated using the kable and kableExtra packages and data visualizations using ggplot2 from the Tidyverse. This document was generated using Quarto in RStudio. Additional details are available in the Appendix (e.g. package versions, R/RStudio versions, etc.)

The here package has been used to load files, which *should* be robust to different operating systems, users, etc. as long as the file structure of this project (as described in the README) is followed. If issues persist, all specified file paths can be easily found searching for here ( within this document and replacing with the user's preferred method of file path specification.

Each article has a brief section that displays the original table(s) or figure from the table, my recreation of the same results, and a brief summary section commenting on the reproducibility

# "Why Some Tennis Matches Take Forever"

From Carl Bialik's article "Why Some Tennis Matches Take Forever", we will recreate his two tables (included below).

### Table 1: Events

## Original

We will first look at the original table from Bialik's article:

Surface Speeds Average time added per point in men's tennis						
Fastest tournaments						
	TOURNAMENT	SURFACE	YEARS RUNNING	TIME ADDED		
1	Wimbledon	Grass	1991-2014	-2.98s		
2	London Olympics	Grass	2012	-2.62		
3	Manchester	Grass	1991-94	-2.33		
4	Eastbourne	Grass	2009-14	-1.78		
5	Birmingham	Carpet	1991	-1.63		
6	Queen's Club	Grass	1991-2014	-1.50		
7	Lyon	Hard	2009	-1.35		
8	Las Vegas	Hard	2006-08	-1.32		
9	Stockholm Masters	Carpet	1991-94	-0.90		
10	Nottingham	Grass	1995-2008	-0.82		
Slowest tournaments						
	TOURNAMENT	SURFACE	YEARS RUNNING	TIME ADDED		
196	Birmingham	Clay	1994	+4.498		
197	Oporto	Clay	1995-96	+4.69		
198	Genova	Clay	1991-93	+4.71		
199	Bologna	Clay	1991-98	+4.74		
200	Merano	Clay	1999	+4.93		
201	Viña del Mar	Clay	2009	+4.96		
202	Florence	Clay	1991-94	+5.08		
203	Costa do Sauipe	Clay	2004-11	+5.19		
204	Maceio	Clay	1992	+5.28		
205	Rio Open	Clay	2015	+5.38		
Oth	er notable tournam	ents				
	TOURNAMENT	SURFACE	YEARS RUNNING	TIME ADDE		
21	U.S. Open	Hard	1991-2014	-0.17		
24	Australian Open	Hard	1991-2015	-0.11		
54	Roland Garros	Clay	1991-2014	+0.79		
	THIRTYEIGHT			FROM JEFF SACKMAI		

Re-creation

The data used to recreate these tables is located in Bialik's provided events.csv file.

Fas	Fastest tournaments					
	Tournament	Surface	Years Running	Time Added		
1	Wimbledon	Grass	1991-2014	-2.98s		
2	London Olympics	Grass	2012	-2.62		
3	Manchester	Grass	1991-1994	-2.33		
4	Eastbourne	Grass	2009-2014	-1.78		
5	Birmingham	Carpet	1991	-1.63		
6	Queen's Club	Grass	1991-2014	-1.50		
7	Lyon	Hard	2009	-1.35		
8	Las Vegas	Hard	2006-2008	-1.32		
9	Stockholm Masters	Carpet	1991-1994	-0.90		
10	Nottingham	Grass	1995-2008	-0.82		

Slowest tournaments					
	Tournament	Surface	Years Running	Time Added	
196	Birmingham	Clay	1994	+4.49s	
197	Oporto	Clay	1995-1996	+4.69	
198	Genova	Clay	1991-1993	+4.71	
199	Bologna	Clay	1991-1998	+4.74	
200	Merano	Clay	1999	+4.93	
201	Viña del Mar	Clay	2009	+4.96	
202	Florence	Clay	1991-1994	+5.08	
203	Costa Do Sauipe	Clay	2004-2011	+5.19	
204	Maceio	Clay	1992	+5.28	
205	Rio Open	Clay	2015	+5.38	

Other notable tournaments					
	Tournament	Surface	Years Running	Time Added	
21	US Open	Hard	1991-2014	-0.17s	
24	Australian Open	Hard	1991-2015	-0.11	
54	Roland Garros	Clay	1991-2014	+0.79	

# Table 2: Players

# Original

Again, we first we can review the original table(s) presented in the article:

Fas	test players		Slov	vest players	
1	Dustin Brown	-6.37s	209	Michael Chang	+4.94s
2	Rohan Bopanna	-4.95	210	Joao Cunha Silva	+5.10
3	Chris Guccione	-4.63	211	Julian Knowle	+5.15
4	Benoit Paire	-4.56	212	John McEnroe	+5.22
5	Lukas Dlouhy	-4.35	213	Lucas Arnold Ker	+5.35
6	Brendan Evans	-4.25	214	T.J. Middleton	+5.60
7	Igor Sijsling	-4.19	215	Martin Stringari	+5.68
8	Lukas Rosol	-4.13	216	Rafael Nadal	+5.92
9	Alexander Kudryavtsev	-4.05	217	Nicolas Massu	+6.21
10	Sam Querrey	-3.99	218	Emanuel Couto	+6.35
Oth	er notable players				
22	Goran Ivanisevic	-3.15s			
36	Roger Federer	-2.43			
124	Novak Djokovic	+2.21			
141	Andy Murray	+2.53			
191	Pat Cash	+3.73			
202	Ivan LendI	+4.35			
	line Oeronien	+4.51			
203	Jim Courier	T4.JI			

#### Re-creation

Now we can begin attempting to recreate this table, doing our best to mirror the format of the 538 results shown above in structure/formatting.

We generate generate each of the three tables included in the image above, using the players\_time.csv table. Player rank (the integer column) was generated as noted below, player name taken from the player column, and the added time from the seconds\_added\_per\_point column:

Fastest players			Slowest players		
1	Dustin Brown	-6.37s	209	Michael Chang	4.94s
2	Rohan Bopanna	-4.95	210	Joao Cunha Silva	+5.10
3	Chris Guccione	-4.63	211	Julian Knowle	+5.15
4	Benoit Paire	-4.56	212	John McEnroe	+5.22
5	Lukas Dlouhy	-4.35	213	Lucas Arnold Ker	+5.35
6	Brendan Evans	-4.25	214	T.J. Middleton	+5.60
7	Igor Sijsling	-4.19	215	Martin Stringari	+5.68
8	Lukas Rosol	-4.13	216	Rafael Nadal	+5.92
9	Alexander Kudryavtsev	-4.05	217	Nicolas Massu	+6.21
10	Sam Querrey	-3.99	218	Emanuel Couto	+6.35

Other notable players					
22	Goran Ivanisevic	-3.15s			
36	Roger Federer	-2.43			
124	Novak Djokovic	+2.21			
141	Andy Murray	+2.53			
191	Pat Cash	+3.73			
202	Ivan Lendl	+4.35			
203	Jim Courier	+4.51			
207	Jimmy Connors	+4.90			

We see our results mirror those in Bialik's original article exactly, and we were able to easily identify and match the data to recreate these tables from the provided GitHub repository (even without any code given by the author).

## Summary/Analysis

For Bialik's article on the pace of play in modern tennis, the data for the presented tables was immediately available and well-formatted to corroborate these results and easily output tables that are similar to those in Bialik's original article. The data used was sparse (with no additional information), which necessitated little documentation.

However considering the larger article, the two figures are presented without sufficient data to re-created. The provided data does not include any temporal data on player tendencies for the second figure, and the events data has been aggregated so that annual data is not available to create the first figure. The article also discusses regression modelling for pace of play but with no presentation of results or sharing of code to use in evaluating the model.

As an entertainment article, we obviously would not expect the same level of rigor for reproducibility as we might expect for a peer-reviewed publication or scientific article. The visualization and modelling are presented with no available data or thorough discussion of the methods to generate the results, while the tables in Bialik's article are easily reproduced from the shared materials.

# "We Watched 906 Foul Balls To Find Out Where The Most Dangerous Ones Land"

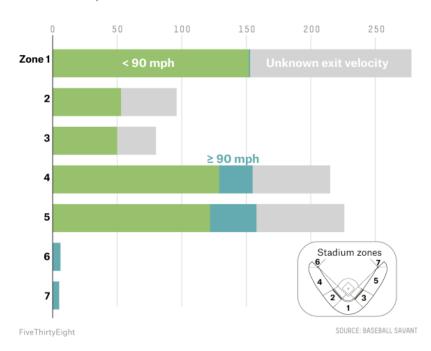
Data and analysis presented from Annette Choi's article We Watched 906 Fould Balls to Find Out Where the most Dangerous Ones Land.

Figure 2

**Original** 

#### The hardest-hit fouls seem to land in unprotected areas

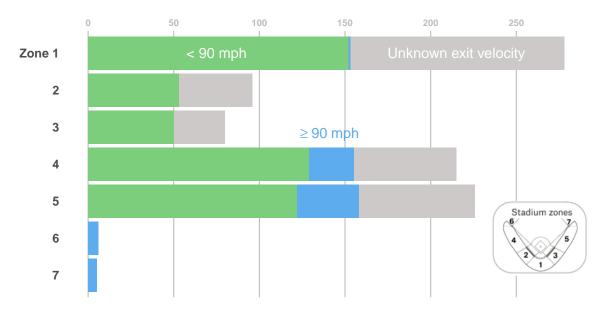
Foul balls by the stadium zone they landed in and their exit velocity, among 906 fouls hit this season in the most foul-heavy day at the 10 MLB stadiums that produced the most fouls as of June 5



### Re-creation

## The hardest hit foul-balls seem to land in unprotected areas

Foul balls by the stadium zone they landed in and their exit velocity, among 906 foul balls hit this season in the most foul–heavy day at the 10 MLB stadiums that produced the most fouls as of June 5



## Summary/Analysis

For Annette Choi's article We Watched 906 Fould Balls to Find Out Where the most Dangerous Ones Land, I recreated only her final figure, which summarizes the landing spot of foul balls into Choi's defined "zones", stratified by the exit velocity of the hit.

The data was immediately available and easy to use in generating this figure (i.e. only little wrangling necessary). The data contained three variables regarding the final "zones", with no indication in the article which was used to generate the figure (or in the analyses throughout the article). I settled on the used\_zone variable, which uses the observed zone (camera\_zone) of the foul ball when available and otherwise imputed a predicted\_zone if the camera did not capture the final zone (and camera\_zone was then missing). This imputation was discussed briefly in the article, but its use as the primary outcome variable was not mentioned explicitly in either the article or the figure.

Although I did not present the table and other figure, the table could be reproduced with the exception of the stadium column (although this could be inferred and created using the matchup's home team). The ballpark figure could be recreated using the available data, although without shared code the method of creation of the ballpark figure is unknown and could only be approximated from the provided information.

## **Appendix**

#### A: Session Information

```
R version 4.2.1 (2022-06-23 ucrt)
Platform: x86_64-w64-mingw32/x64 (64-bit)
Running under: Windows 10 x64 (build 22621)
Matrix products: default
locale:
[1] LC_COLLATE=English_United States.utf8
[2] LC_CTYPE=English_United States.utf8
[3] LC_MONETARY=English_United States.utf8
[4] LC_NUMERIC=C
[5] LC_TIME=English_United States.utf8
attached base packages:
[1] grid
              stats
                        graphics grDevices utils
                                                      datasets methods
[8] base
other attached packages:
[1] png_0.1-7
                     kableExtra_1.3.4 stringr_1.4.1
                                                       dplyr_1.0.10
[5] ggplot2_3.4.0
                     here_1.0.1
                                      knitr_1.40
loaded via a namespace (and not attached):
 [1] compiler_4.2.1
                       pillar_1.8.1
                                         tools_4.2.1
                                                           digest_0.6.29
 [5] viridisLite_0.4.1 jsonlite_1.8.2
                                         evaluate_0.17
                                                           lifecycle_1.0.3
 [9] tibble_3.1.8
                       gtable_0.3.1
                                         pkgconfig_2.0.3
                                                           rlang_1.0.6
[13] cli_3.4.1
                       DBI_1.1.3
                                         rstudioapi_0.14
                                                           yam1_2.3.5
[17] xfun_0.33
                       fastmap_1.1.0
                                         xml2 1.3.3
                                                           httr 1.4.4
[21] withr_2.5.0
                       systemfonts_1.0.4 generics_0.1.3
                                                           vctrs_0.5.1
[25] webshot_0.5.4
                       rprojroot_2.0.3
                                         tidyselect_1.2.0 svglite_2.1.0
[29] glue_1.6.2
                       R6_2.5.1
                                         fansi_1.0.3
                                                           rmarkdown_2.17
[33] farver_2.1.1
                       magrittr_2.0.3
                                         scales_1.2.1
                                                           htmltools_0.5.3
[37] rvest_1.0.3
                       assertthat_0.2.1 colorspace_2.0-3 utf8_1.2.2
[41] stringi_1.7.8
                       munsell_0.5.0
```

#### B: Code

```
###############
#### Set-Up ####
###############
library(here) # used for file management
library(ggplot2) # data visualization
library(dplyr) # data wrangling
library(stringr) # data wrangling (string detection, manipulation)
library(kableExtra) # LaTeX/PDF table styling
library(knitr)
###################
### Data Import ###
###################
events <- read.csv(here("Data", "tennis-time", "events_time.csv"), header=T) %>%
 mutate(Rank=row_number())
players <- read.csv(here("Data", "tennis-time", "players_time.csv")) %>%
  mutate(Rank=row_number()) # adding the "Rank" variable (the integers in each table
            # above), descending (1=fastest player, lowest `seconds_added_per_point`)
###############
### Table 1 ###
##############
# Original
knitr::include_graphics(here("Images", "tennis-time", "bialik-tennis-time-table21.png"))
# Re-creation
events_10f <- events %>%
  arrange(seconds_added_per_point) %>%
 head(10) %>%
 mutate(seconds = # reformatting as string to include the "s" for first obs
           case_when(row_number()==1 ~ paste0(sprintf("%0.2f", seconds_added_per_point)
                                               , "s")
                     , seconds_added_per_point<0 ~ sprintf("%0.2f", seconds_added_per_point
                     , T ~ paste0("+",
```

```
sprintf("%0.2f", seconds_added_per_point))
                     )
         ) %>%
 select(Rank, tournament, surface, years, seconds)
events_10s <- events %>%
 mutate(Rank=row number()) %>%
 arrange(desc(seconds_added_per_point)) %>%
 head(10) %>%
 arrange(seconds_added_per_point) %>%
 mutate(seconds = # reformatting as string to include the "s" for the first obs
           case_when(row_number()==1 ~ paste0("+"
                                        , sprintf("%0.2f", seconds_added_per_point)
                     , seconds_added_per_point<0 ~ sprintf("%0.2f", seconds_added_per_point
                     , T ~ paste0("+", sprintf("%0.2f", seconds_added_per_point))
         ) %>%
 select(Rank, tournament, surface, years, seconds)
events_ntbl <- events %>%
 filter(str_detect(tournament, "US Op|Australian Open|Roland Garr")) %>%
 mutate(seconds = # reformatting as string to include the "s" for the first obs
           case_when(row_number()==1 ~ paste0(sprintf("%0.2f", seconds_added_per_point), "
                     , seconds_added_per_point<0 ~ sprintf("%0.2f", seconds_added_per_point
                     , T ~ paste0("+", sprintf("%0.2f", seconds_added_per_point))
         ) %>%
 select(Rank, tournament, surface, years, seconds)
event_cols <- c("", "Tournament", "Surface", "Years Running", "Time Added")
events_10f %>%
 kable(col.names=event_cols
        # , bottomrule=''
        , booktabs=T
        ) %>%
 kable_styling(latex_options = "HOLD_position") %>%
 add_header_above(c("Fastest tournaments"=5), align = "1"
                   , bold = T
                   , underline = F, line = F)
```

```
events_10s %>%
  kable(col.names=event_cols
        # , bottomrule=''
        , booktabs=T
        ) %>%
  kable_styling(latex_options = "HOLD_position") %>%
  add_header_above(c("Slowest tournaments"=5), align = "1"
                   , bold = T
                   , underline = F, line = F)
events_ntbl %>%
  kable(col.names=event_cols
        # , bottomrule=''
        , booktabs=T
        ) %>%
  kable_styling(latex_options = "HOLD_position") %>%
  add_header_above(c("Other notable tournaments"=5), align = "l"
                   , bold = T
                   , underline = F, line = F)
##############
### Table 2 ###
###############
# Original
knitr::include_graphics(here("Images", "tennis-time", "bialik-tennis-time-table1.png"))
# Re-creation
# Top 10 Fastest Players (lower/more negative seconds_added per_point is faster pace)
t10_f <- players %>%
  arrange(seconds_added_per_point) %>% # sort ascending
  head(10) %>% # top 10
  mutate(seconds = # reformatting as above
           case_when(row_number()==1 ~ paste0(sprintf("%0.2f", seconds_added_per_point)
                                               , "s")
                     , T ~ sprintf("%0.2f"
                                    , seconds_added_per_point)
                     )
         ) %>%
```

```
select(Rank, player, seconds)
# "Top" 10 slowest players (largest values)
t10_s <- players %>%
  arrange(desc(seconds_added_per_point)) %>%
  head(10) %>%
  arrange(seconds_added_per_point) %>%
  mutate(seconds =
           case_when(row_number()==1 ~ paste0(sprintf("%0.2f", seconds_added_per_point)
                                               , "s")
                     , seconds_added_per_point<0 ~ sprintf("%0.2f", seconds_added_per_point
                     , T ~ paste0("+"
                              , sprintf("%0.2f", seconds_added_per_point))
         ) %>%
  select(Rank, player, seconds)
# Specific players ("notable") taken from the full data set
tbl_pls <- players %>%
  filter(
    str_detect(player
, "Goran|Roger Fed|Novak Djok|Andy Murray|Pat Cash|Ivan Lend|Jim Courier|Jimmy Connors"))
  arrange(seconds_added_per_point) %>%
  mutate(seconds =
           case_when(row_number()==1 ~ paste0(sprintf("%0.2f", seconds_added_per_point)
                                               , "s")
                     , seconds_added_per_point<0 ~ sprintf("%0.2f", seconds_added_per_point
                     , T ~ paste0("+", sprintf("%0.2f", seconds_added_per_point))
         ) %>%
  select(Rank, player, seconds)
t10_f %>%
  kable(col.names=NULL
        # , bottomrule=''
        , booktabs=T
  kable_styling(latex_options ="HOLD_position") %>%
  add_header_above(c("Fastest players"=3)
                   , bold=T, align="l"
```

```
, underline=F, line=F
t10_s %>%
  kable(col.names=NULL
        # , bottomrule=''
        , booktabs=T
        ) %>%
  kable_styling(latex_options ="HOLD_position") %>%
  add_header_above(c("Slowest players"=3)
                    , bold=T, align="l"
                    , underline=F, line=F
tbl_pls %>%
  kable(col.names=NULL
        # , bottomrule=''
        , booktabs=T
        ) %>%
  kable_styling(latex_options ="HOLD_position") %>%
  add_header_above(c("Other notable players"=3)
                    , bold=T, align="l"
                    , underline=F, line=F
knitr::include_graphics(here("Images", "foul-balls", "choi-foul-0625-2-1.png"))
foul_balls <- read.csv(here("Data", "foul-balls", "foul-balls.csv"))</pre>
library(png); library(grid)
img <- readPNG(here("Images", "foul-balls", "stadium_zones.png"))</pre>
g <- rasterGrob(img, interpolate=TRUE)</pre>
foul_balls %>%
  mutate(Velocity_Cat =
           case_when(exit_velocity<90 ~ 1</pre>
                      , exit_velocity>=90 ~ 2
                      , TRUE ~ 3)
         , Zone_f = case_when(
           used_zone == 1 ~ "Zone 1"
           , TRUE ~ as.character(used_zone)
         )) %>%
  mutate(fvc = factor(Velocity_Cat, levels=c(3, 2, 1))) %>%
  ggplot(aes(x=reorder(as.factor(Zone_f), -used_zone), fill=fvc)) +
```

```
geom_bar() +
 scale_y_continuous(breaks=seq(0, 250, 50)
                     , position="right") +
 scale_fill_manual(values=c("snow3", "steelblue2", "palegreen3")) +
 coord_flip() +
 ylab("") + xlab("") +
 # ggthemes::theme fivethirtyeight() +
 ggtitle(label="The hardest hit foul-balls seem to land in unprotected areas"
          , subtitle = "Foul balls by the stadium zone they landed in and their exit velocity
 theme(legend.position="none", plot.title = element_text(size=16)
        , panel.background = element_rect(fill = 'white', color = 'white')
        , panel.grid.major.x = element_line(colour = "gray", linewidth = 0.5)
        , axis.text.y = element_text(face="bold", size = 12)
        , axis.text.x = element_text(color="gray", face="bold")
        , axis.ticks = element_blank()
 annotation_custom(g, xmin=1, xmax=3, ymin=210, ymax=320) +
 annotate("text", x = 7, y = 75, label="< 90 mph", col="white", size=5) +
 annotate("text", x = 7, y = 215, label="Unknown exit velocity", col="white", size=5) +
 annotate("text", x = 4.8, y = 140, label=("''>= 90 ~ 'mph'"), col="steelblue2", size=5,
sessionInfo()
```