

1b. ATP Initial Exploration Part 2

This notebook is a continuation of "1a. ATP Initial Exploration Part 1". The first notebook focused on cleaning and wrangling steps. This second notebook will create subsets to use for analysis.

Table of Contents

1. Importing Libraries and PKL File

2. Creating Subsets

2.1 Creating New Main Dataframe: df_matchstats

2.2 Best of 3 or 5

2.3 Court Surface

3. Exporting Subsets as PKL Files

List of All New Subsets

- New Main Dataframe: "df_matchstats"
- By Number of Sets: "df_BO3" and "df_BO5"
- Hard Courts (3 total): "df_hard", "df_hard_BO3" and "df_hard_BO5"
- Clay Courts (3 total): "df_clay", "df_clay_BO3" and "df_clay_BO5"
- Grass Courts (3 total): "df_grass", "df_grass_BO3" and "df_grass_BO5"

1. Importing Libraries and PKL File

#Import Libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import os
import scipy
import matplotlib
```

#Set Path

```
path = r'/Users/tristansavella/Desktop/Important Things/Data Analytics/CareerFoundry/Data Immersion/Achievement 6/Master Folder ATP/02 Data'
```

```
#Import df_post2000
df_post2000 = pd.read_pickle(os.path.join(path, 'Prepared
Data','df_post2000.pkl'))
```

```
#Show all columns
pd.set_option('display.max_columns', None)
```

```
#Show all rows
pd.set_option('display.max_rows', None)
```

```
#Check Head
df_post2000.head()
```

	tourney_id	tourney_name	surface	draw_size	tourney_level
tourney_date \					
119317	2000-301	Auckland	Hard	32	A
20000110					
119318	2000-301	Auckland	Hard	32	A
20000110					
119319	2000-301	Auckland	Hard	32	A
20000110					
119320	2000-301	Auckland	Hard	32	A
20000110					
119321	2000-301	Auckland	Hard	32	A
20000110					

	match_num	winner_id	winner_seed	winner_entry	
winner_name \					
119317	1	103163	1.0	NaN	Tommy
Haas					
119318	2	102607	NaN	Q	Juan
Balcells					
119319	3	103252	NaN	NaN	Alberto
Martin					
119320	4	103507	7.0	NaN	Juan Carlos
Ferrero					
119321	5	102103	NaN	Q	Michael
Sell					

	winner_hand	winner_ht	winner_ioc	winner_age	loser_id
loser_seed \					
119317	R	188.0	GER	21.7	101543
NaN					
119318	R	190.0	ESP	24.5	102644
NaN					
119319	R	175.0	ESP	21.3	102238
NaN					
119320	R	183.0	ESP	19.9	103819
NaN					
119321	R	180.0	USA	27.3	102765

4.0

	loser_entry	loser_name	loser_hand	loser_ht	loser_ioc
\					
119317	NaN	Jeff Tarango	L	180.0	USA
119318	NaN	Franco Squillari	L	183.0	ARG
119319	NaN	Alberto Berasategui	R	173.0	ESP
119320	NaN	Roger Federer	R	185.0	SUI
119321	NaN	Nicolas Escude	R	185.0	FRA

	loser_age	score	best_of	round	minutes	w_#aces
w_#dfs	\					
119317	31.1	7-5 4-6 7-5	3	R32	108.0	18.0
4.0						
119318	24.3	7-5 7-5	3	R32	85.0	5.0
3.0						
119319	26.5	6-3 6-1	3	R32	56.0	0.0
0.0						
119320	18.4	6-4 6-4	3	R32	68.0	5.0
1.0						
119321	23.7	0-6 7-6(7) 6-1	3	R32	115.0	1.0
2.0						

	w_#ServePoints	w_#1stServesIn	w_#1stWon	w_#2ndWon
w_#ServeGames	\			
119317	96.0	49.0	39.0	28.0
17.0				
119318	76.0	52.0	39.0	13.0
12.0				
119319	55.0	35.0	25.0	12.0
8.0				
119320	53.0	28.0	26.0	15.0
10.0				
119321	98.0	66.0	39.0	14.0
13.0				

	w_bpSaved	w_#bpFaced	l_#aces	l_#dfs	l_#ServePoints	\
119317	3.0	5.0	7.0	8.0	106.0	
119318	5.0	6.0	5.0	10.0	74.0	
119319	1.0	1.0	0.0	6.0	56.0	
119320	0.0	0.0	11.0	2.0	70.0	
119321	6.0	11.0	8.0	8.0	92.0	

	l_#1stServesIn	l_#1stWon	l_#2ndWon	l_#ServeGames	l_bpSaved
\					

119317	55.0	39.0	29.0	17.0	4.0
119318	32.0	25.0	18.0	12.0	3.0
119319	33.0	20.0	7.0	8.0	7.0
119320	43.0	29.0	14.0	10.0	6.0
119321	46.0	34.0	18.0	12.0	5.0
	l_#bpFaced	winner_rank	winner_rank_points	loser_rank	\
119317	7.0	11.0	1612.0	63.0	
119318	6.0	211.0	157.0	49.0	
119319	11.0	48.0	726.0	59.0	
119320	8.0	45.0	768.0	61.0	
119321	9.0	167.0	219.0	34.0	
	loser_rank_points	w_#2ndServePoints	w_%1stServesIn	w_	
%1stWon	\				
119317	595.0	47.0	51		
79					
119318	723.0	24.0	68		
75					
119319	649.0	20.0	63		
71					
119320	616.0	25.0	52		
92					
119321	873.0	32.0	67		
59					
	w_%2ndWon	l_#2ndServePoints	l_%1stServesIn	l_%1stWon	l_
%2ndWon	\				
119317	59	51.0	51	70	
56					
119318	54	42.0	43	78	
42					
119319	60	23.0	58	60	
30					
119320	60	27.0	61	67	
51					
119321	43	46.0	50	73	
39					
	Year				
119317	2000				
119318	2000				
119319	2000				
119320	2000				
119321	2000				

2. Creating Subsets

2.1 Create main dataframe, 'df_matchstats', consisting only of match statistics

2.2 Divide 'df_matchstats' into two groups; best of 3 and best of 5

2.3 Create Subsets for Matches Played By Surface (Hard, Clay and Grass)

2.4 Create Subsets for Matches Played by Big 3 (Participate, Win and Lose)

Final List of Subsets

- Main Dataframe: "df_matchstats"
- By Number of Sets: "df_BO3" and "df_BO5"
- Hard Courts (3 total): "df_hard", "df_hard_BO3" and "df_hard_BO5"
- Clay Courts (3 total): "df_clay", "df_clay_BO3" and "df_clay_BO5"
- Grass Courts (3 total): "df_grass", "df_grass_BO3" and "df_grass_BO5"
- Big 3 Wins (3 total): "df_big3_win", "df_big3_win_BO3" and "df_big3_win_BO5"
- Big 3 Loses (3 total): "df_big3_lose", "df_big3_lose_BO3" and "df_big3_lose_BO5"

#drop irrelevant columns

```
df_matchstats = df_post2000.drop(columns = ['tourney_date',
                                             'draw_size',
                                             'match_num',
                                             'score',
                                             'winner_entry',
                                             'winner_hand',
                                             'loser_entry',
                                             'loser_hand',
                                             'winner_rank_points',
                                             'loser_rank_points',
                                             'winner_seed',
                                             'loser_seed',
                                             ])
```

#Reorder Columns

```
df_matchstats = df_matchstats[['tourney_id',
                                'Year',
                                'tourney_name',
                                'surface',
                                'tourney_level',
                                'winner_id',
                                'winner_ioc',
                                'winner_name',
                                'winner_age',
                                'winner_rank',
                                'winner_ht',
```

```

        'loser_id',
'loser_ioc',
        'loser_name',
        'loser_rank',
'loser_ht',
        'loser_age',
        'best_of',
        'round',
        'minutes',
        'w_#ServeGames',
        'w_#aces',
        'w_#dfs',
        'w_#ServePoints',
        'w_#1stServesIn',
        'w_#2ndServePoints',
        'w_%1stServesIn',
        'w_#1stWon',
        'w_%1stWon',
        'w_#2ndWon',
        'w_%2ndWon',
        'w_bpSaved',
        'w_#bpFaced',
        'l_#ServeGames',
        'l_#aces',
        'l_#dfs',
        'l_#ServePoints',
        'l_#1stServesIn',
        'l_#2ndServePoints',
        'l_%1stServesIn',
        'l_#1stWon',
        'l_%1stWon',
        'l_#2ndWon',
        'l_%2ndWon',
        'l_bpSaved',
        'l_#bpFaced']]

```

#check head

```
df_matchstats.head()
```

	tourney_id	Year	tourney_name	surface	tourney_level	
winner_id \						
119317	2000-301	2000	Auckland	Hard	A	103163
119318	2000-301	2000	Auckland	Hard	A	102607
119319	2000-301	2000	Auckland	Hard	A	103252
119320	2000-301	2000	Auckland	Hard	A	103507
119321	2000-301	2000	Auckland	Hard	A	102103

winner_id	winner_ioc	winner_name	winner_age	winner_rank
119317	GER	Tommy Haas	21.7	11.0
119318	ESP	Juan Balcells	24.5	211.0
119319	ESP	Alberto Martin	21.3	48.0
119320	ESP	Juan Carlos Ferrero	19.9	45.0
119321	USA	Michael Sell	27.3	167.0

loser_id	loser_ioc	loser_name	loser_rank
101543	USA	Jeff Tarango	63.0
102644	ARG	Franco Squillari	49.0
102238	ESP	Alberto Berasategui	59.0
103819	SUI	Roger Federer	61.0
102765	FRA	Nicolas Escude	34.0

loser_age	best_of	round	minutes	w_#ServeGames	w_#aces
31.1	3	R32	108.0	17.0	18.0
24.3	3	R32	85.0	12.0	5.0
26.5	3	R32	56.0	8.0	0.0
18.4	3	R32	68.0	10.0	5.0
23.7	3	R32	115.0	13.0	1.0

w_#ServePoints	w_#1stServesIn	w_#2ndServePoints	w_#1stServesIn
96.0	49.0	47.0	51
76.0	52.0	24.0	68
55.0	35.0	20.0	63
53.0	28.0	25.0	

52

119321	98.0	66.0	32.0
--------	------	------	------

67

	w_#1stWon	w_%1stWon	w_#2ndWon	w_%2ndWon	w_bpSaved
--	-----------	-----------	-----------	-----------	-----------

w_#bpFaced \

119317	39.0	79	28.0	59	3.0
--------	------	----	------	----	-----

5.0

119318	39.0	75	13.0	54	5.0
--------	------	----	------	----	-----

6.0

119319	25.0	71	12.0	60	1.0
--------	------	----	------	----	-----

1.0

119320	26.0	92	15.0	60	0.0
--------	------	----	------	----	-----

0.0

119321	39.0	59	14.0	43	6.0
--------	------	----	------	----	-----

11.0

	l_#ServeGames	l_#aces	l_#dfs	l_#ServePoints	l_#1stServesIn
--	---------------	---------	--------	----------------	----------------

\

119317	17.0	7.0	8.0	106.0	55.0
--------	------	-----	-----	-------	------

119318

	12.0	5.0	10.0	74.0	32.0
--	------	-----	------	------	------

119319

	8.0	0.0	6.0	56.0	33.0
--	-----	-----	-----	------	------

119320

	10.0	11.0	2.0	70.0	43.0
--	------	------	-----	------	------

119321

	12.0	8.0	8.0	92.0	46.0
--	------	-----	-----	------	------

	l_#2ndServePoints	l_%1stServesIn	l_#1stWon	l_%1stWon
--	-------------------	----------------	-----------	-----------

l_#2ndWon \

119317	51.0	51	39.0	70
--------	------	----	------	----

29.0

119318	42.0	43	25.0	78
--------	------	----	------	----

18.0

119319	23.0	58	20.0	60
--------	------	----	------	----

7.0

119320	27.0	61	29.0	67
--------	------	----	------	----

14.0

119321	46.0	50	34.0	73
--------	------	----	------	----

18.0

	l_%2ndWon	l_bpSaved	l_#bpFaced
--	-----------	-----------	------------

119317	56	4.0	7.0
--------	----	-----	-----

119318	42	3.0	6.0
--------	----	-----	-----

119319	30	7.0	11.0
--------	----	-----	------

119320	51	6.0	8.0
--------	----	-----	-----

119321	39	5.0	9.0
--------	----	-----	-----


```
#describe
```

```
df_matchstats.describe()
```

	winner_age	winner_rank	winner_ht	loser_rank
loser_ht \				
count	62530.000000	62441.000000	62332.000000	62234.000000
61801.000000				
mean	26.407667	62.159607	186.182603	95.891731
185.611317				
std	3.928157	86.099346	6.826430	132.046004
6.776734				
min	14.900000	1.000000	163.000000	1.000000
163.000000				
25%	23.500000	17.000000	183.000000	35.000000
181.000000				
50%	26.200000	42.000000	185.000000	65.000000
185.000000				
75%	29.100000	79.000000	190.000000	106.000000
190.000000				
max	42.300000	1890.000000	211.000000	2159.000000
211.000000				

	loser_age	minutes	w_#ServeGames	w_#aces
w_#dfs \				
count	62528.000000	61074.000000	62530.000000	62530.000000
62530.000000				
mean	26.540982	106.384141	12.507484	6.915832
2.652167				
std	4.002836	40.944761	4.221022	5.547041
2.291411				
min	14.500000	3.000000	0.000000	0.000000
0.000000				
25%	23.600000	76.000000	9.000000	3.000000
1.000000				
50%	26.400000	99.000000	11.000000	6.000000
2.000000				
75%	29.300000	129.000000	15.000000	9.000000
4.000000				
max	46.000000	1146.000000	90.000000	113.000000
26.000000				

	w_#ServePoints	w_#1stServesIn	w_#2ndServePoints	w_
%1stServesIn \				
count	62530.000000	62530.000000	62530.000000	
62530.000000				
mean	77.922469	47.881449	30.04102	
61.061427				
std	29.176144	18.902521	13.02635	
8.180105				
min	3.000000	1.000000	1.000000	

12.000000			
25%	56.000000	34.000000	21.000000
56.000000			
50%	73.000000	45.000000	28.000000
61.000000			
75%	94.000000	58.000000	37.000000
67.000000			
max	491.000000	361.000000	130.000000
98.000000			

	w_#1stWon	w_%1stWon	w_#2ndWon	w_%2ndWon
w_bpSaved \				
count	62530.000000	62530.000000	62530.000000	62530.000000
62530.000000				
mean	36.233664	76.289637	16.654630	56.223413
3.464753				
std	13.541931	8.141773	6.983291	10.496364
3.078031				
min	0.000000	0.000000	0.000000	0.000000
0.000000				
25%	27.000000	71.000000	12.000000	50.000000
1.000000				
50%	34.000000	76.000000	16.000000	56.000000
3.000000				
75%	43.000000	82.000000	20.000000	63.000000
5.000000				
max	292.000000	100.000000	82.000000	129.000000
24.000000				

	w_#bpFaced	l_#ServeGames	l_#aces	l_#dfs \
count	62530.000000	62530.000000	62530.000000	62530.000000
mean	5.038893	12.300112	5.108460	3.378682
std	4.034177	4.222271	4.882427	2.530127
min	0.000000	0.000000	0.000000	0.000000
25%	2.000000	9.000000	2.000000	2.000000
50%	4.000000	11.000000	4.000000	3.000000
75%	7.000000	15.000000	7.000000	5.000000
max	30.000000	91.000000	103.000000	26.000000

	l_#ServePoints	l_#1stServesIn	l_#2ndServePoints	l_#1stServesIn \
count	62530.000000	62530.000000	62530.000000	62530.000000
62530.000000				
mean	80.973885	48.472013	32.501871	
59.158452				
std	29.154055	19.195435	12.949551	
8.272118				
min	3.000000	1.000000	1.000000	
12.000000				
25%	60.000000	35.000000	23.000000	

54.000000				
50%	76.000000	45.000000	30.000000	
59.000000				
75%	97.000000	59.000000	40.000000	
65.000000				
max	489.000000	328.000000	161.000000	
97.000000				

	l_#1stWon	l_%1stWon	l_#2ndWon	l_%2ndWon
l_bpSaved \				
count	62530.000000	62530.000000	62530.000000	62530.000000
62530.000000				
mean	32.332768	65.442875	14.972477	44.859332
4.778330				
std	14.349978	9.668002	7.210501	10.276116
3.270382				
min	0.000000	0.000000	0.000000	0.000000
0.000000				
25%	22.000000	60.000000	10.000000	38.000000
2.000000				
50%	30.000000	66.000000	14.000000	45.000000
4.000000				
75%	40.000000	72.000000	19.000000	52.000000
7.000000				
max	284.000000	100.000000	101.000000	100.000000
27.000000				

	l_#bpFaced
count	62530.000000
mean	8.630977
std	4.142344
min	0.000000
25%	6.000000
50%	8.000000
75%	11.000000
max	38.000000

2.2 Best of 3 or 5

```
#Checking for matches played best of 3 and matches played best of 5
df_matchstats['best_of'].value_counts(dropna = False)

best_of
3    50549
5    11981
Name: count, dtype: int64

#Subset: Matches Played Best of 3 Sets
df_B03 = df_matchstats[df_matchstats['best_of']==3]
```

```

#checking shape
df_B03.shape
#50549 was the number I was looking for

(50549, 46)

#Subset: Matches Played Best of 5 Sets
df_B05 = df_matchstats[df_matchstats['best_of']==5]

#checking shape
df_B05.shape
#11981 was the number I was looking for

(11981, 46)

```

2.3: Create Subsets based on Surface:

```

#checking for matches by surface
df_matchstats['surface'].value_counts(dropna = False)

surface
Hard      34182
Clay      20073
Grass      6836
Carpet     1439
Name: count, dtype: int64

```

Hard Courts

- "df_hard"
- "df_hard_B03"
- "df_hard_B05"

```

#creating main df for hard courts
df_hard = df_matchstats[df_matchstats['surface']=='Hard']

#checking shape
df_hard.shape
#correct number of matches

(34182, 46)

#creating 'df_hard_B03'
df_hard_B03 = df_B03[df_B03['surface']=='Hard']

#checking head
df_hard_B03.head()

```

	tourney_id	Year	tourney_name	surface	tourney_level	
winner_id \						
119317	2000-301	2000	Auckland	Hard	A	103163

119318	2000-301	2000	Auckland	Hard	A	102607
119319	2000-301	2000	Auckland	Hard	A	103252
119320	2000-301	2000	Auckland	Hard	A	103507
119321	2000-301	2000	Auckland	Hard	A	102103

	winner_ioc	winner_name	winner_age	winner_rank
winner_ht \				
119317	GER	Tommy Haas	21.7	11.0
188.0				
119318	ESP	Juan Balcells	24.5	211.0
190.0				
119319	ESP	Alberto Martin	21.3	48.0
175.0				
119320	ESP	Juan Carlos Ferrero	19.9	45.0
183.0				
119321	USA	Michael Sell	27.3	167.0
180.0				

	loser_id	loser_ioc	loser_name	loser_rank
loser_ht \				
119317	101543	USA	Jeff Tarango	63.0
				180.0
119318	102644	ARG	Franco Squillari	49.0
				183.0
119319	102238	ESP	Alberto Berasategui	59.0
				173.0
119320	103819	SUI	Roger Federer	61.0
				185.0
119321	102765	FRA	Nicolas Escude	34.0
				185.0

	loser_age	best_of	round	minutes	w_#ServeGames	w_#aces
w_#dfs \						
119317	31.1	3	R32	108.0	17.0	18.0
4.0						
119318	24.3	3	R32	85.0	12.0	5.0
3.0						
119319	26.5	3	R32	56.0	8.0	0.0
0.0						
119320	18.4	3	R32	68.0	10.0	5.0
1.0						
119321	23.7	3	R32	115.0	13.0	1.0
2.0						

w_#ServePoints	w_#1stServesIn	w_#2ndServePoints	w_#1stServesIn
\			

119317	96.0	49.0	47.0
51			
119318	76.0	52.0	24.0
68			
119319	55.0	35.0	20.0
63			
119320	53.0	28.0	25.0
52			
119321	98.0	66.0	32.0
67			

	w_#1stWon	w_%1stWon	w_#2ndWon	w_%2ndWon	w_bpSaved
w_#bpFaced \					
119317	39.0	79	28.0	59	3.0
5.0					
119318	39.0	75	13.0	54	5.0
6.0					
119319	25.0	71	12.0	60	1.0
1.0					
119320	26.0	92	15.0	60	0.0
0.0					
119321	39.0	59	14.0	43	6.0
11.0					

	l_#ServeGames	l_#aces	l_#dfs	l_#ServePoints	l_#1stServesIn
\					
119317	17.0	7.0	8.0	106.0	55.0
119318	12.0	5.0	10.0	74.0	32.0
119319	8.0	0.0	6.0	56.0	33.0
119320	10.0	11.0	2.0	70.0	43.0
119321	12.0	8.0	8.0	92.0	46.0

	l_#2ndServePoints	l_%1stServesIn	l_#1stWon	l_%1stWon
l_#2ndWon \				
119317	51.0	51	39.0	70
29.0				
119318	42.0	43	25.0	78
18.0				
119319	23.0	58	20.0	60
7.0				
119320	27.0	61	29.0	67
14.0				
119321	46.0	50	34.0	73
18.0				

	l_%2ndWon	l_bpSaved	l_#bpFaced
119317	56	4.0	7.0
119318	42	3.0	6.0
119319	30	7.0	11.0
119320	51	6.0	8.0
119321	39	5.0	9.0

#checking shape

df_hard_B03.shape

(28096, 46)

#creating 'df_hard_B05'

df_hard_B05 = df_B05[df_B05['surface']=='Hard']

#checking head

df_hard_B05.head()

	tourney_id	Year	tourney_name	surface	tourney_level
winner_id \					
119876	2000-337	2000	Vienna	Hard	A
102450					
120032	2000-357	2000	Stuttgart Masters	Hard	M
101965					
120267	2000-403	2000	Miami Masters	Hard	M
101948					
120330	2000-404	2000	Indian Wells Masters	Hard	M
102374					
120996	2000-429	2000	Stockholm	Hard	A
102563					

	winner_ioc	winner_name	winner_age	winner_rank
winner_ht \				
119876	GBR	Tim Henman	26.0	10.0
185.0				
120032	RSA	Wayne Ferreira	29.1	19.0
185.0				
120267	USA	Pete Sampras	28.6	2.0
185.0				
120330	ESP	Alex Corretja	25.9	26.0
180.0				
120996	SWE	Thomas Johansson	25.6	57.0
180.0				

	loser_id	loser_ioc	loser_name	loser_rank	loser_ht \
119876	103163	GER	Tommy Haas	28.0	188.0
120032	103720	AUS	Lleyton Hewitt	8.0	180.0
120267	102856	BRA	Gustavo Kuerten	6.0	190.0
120330	102358	SWE	Thomas Enqvist	10.0	190.0
120996	102338	RUS	Yevgeny Kafelnikov	5.0	190.0

w_#dfs	loser_age	best_of	round	minutes	w_#ServeGames	w_#aces
119876	22.5	5	F	124.0	15.0	10.0
1.0						
120032	19.6	5	F	251.0	27.0	18.0
5.0						
120267	23.5	5	F	198.0	22.0	20.0
9.0						
120330	26.0	5	F	120.0	14.0	10.0
1.0						
120996	26.7	5	F	95.0	14.0	7.0
4.0						

w_#ServePoints	w_#1stServesIn	w_#2ndServePoints	w_#1stServesIn
119876	84.0	56.0	28.0
66			
120032	194.0	100.0	94.0
51			
120267	148.0	94.0	54.0
63			
120330	77.0	38.0	39.0
49			
120996	76.0	36.0	40.0
47			

	w_#1stWon	w_%1stWon	w_#2ndWon	w_%2ndWon	w_bpSaved
w_#bpFaced \					
119876	46.0	82	18.0	64	3.0
3.0					
120032	69.0	69	52.0	55	14.0
20.0					
120267	75.0	79	30.0	55	5.0
6.0					
120330	30.0	78	28.0	71	6.0
7.0					
120996	29.0	80	21.0	52	2.0
5.0					

\	l_#ServeGames	l_#aces	l_#dfs	l_#ServePoints	l_#1stServesIn
119876	15.0	5.0	3.0	99.0	51.0
120032	26.0	8.0	8.0	170.0	84.0
120267	21.0	16.0	4.0	159.0	90.0
120330	15.0	8.0	7.0	100.0	54.0
120996	14.0	3.0	10.0	77.0	36.0

	l_#2ndServePoints	l_%1stServesIn	l_#1stWon	l_%1stWon
l_#2ndWon \				
119876	48.0	51	38.0	74
25.0				
120032	86.0	49	64.0	76
40.0				
120267	69.0	56	61.0	67
43.0				
120330	46.0	54	34.0	62
25.0				
120996	41.0	46	25.0	69
13.0				

	l_%2ndWon	l_bpSaved	l_#bpFaced
119876	52	6.0	9.0
120032	46	8.0	14.0
120267	62	11.0	14.0
120330	54	8.0	13.0
120996	31	2.0	9.0

```
#checking shape
df_hard_B05.shape
#6,086 + 28096 = 34182 --> checks out
(6086, 46)
```

Clay Courts

- "df_clay"
- "df_clay_B03"
- "df_clay_B05"

```
#creating main df for clay courts
df_clay = df_matchstats[df_matchstats['surface']=='Clay']
```

```
#checking shape
df_clay.shape
#correct number of matches
```

```
(20073, 46)
```

```
#creating 'df_clay_B03'
df_clay_B03 = df_B03[df_B03['surface']=='Clay']
```

```
#checking head
df_clay_B03.head()
```

	tourney_id	Year	tourney_name	surface	tourney_level
winner_id \					

119348	2000-306	2000	St. Poelten	Clay	A	102247
119349	2000-306	2000	St. Poelten	Clay	A	102287
119350	2000-306	2000	St. Poelten	Clay	A	102869
119351	2000-306	2000	St. Poelten	Clay	A	103082
119352	2000-306	2000	St. Poelten	Clay	A	102446
	winner_ioc		winner_name	winner_age	winner_rank	winner_ht
\						
119348	ITA		Andrea Gaudenzi	26.8	74.0	183.0
119349	ESP		Albert Portas	26.5	71.0	188.0
119350	ESP		Galo Blanco	23.6	70.0	173.0
119351	GER		Markus Hantschk	22.5	94.0	188.0
119352	UKR		Andrei Medvedev	25.7	21.0	193.0
	loser_id	loser_ioc	loser_name	loser_rank	loser_ht	
loser_age						
\						
119348	103017	GER	Nicolas Kiefer	8.0	183.0	22.8
119349	103242	ESP	Juan Giner	399.0	178.0	21.8
119350	102987	BRA	Andre Sa	86.0	185.0	23.0
119351	103819	SUI	Roger Federer	54.0	185.0	18.7
119352	102795	USA	Scott Humphries	720.0	185.0	23.9
	best_of	round	minutes	w_#ServeGames	w_#aces	w_#dfs
w_#ServePoints						
\						
119348	3	R32	76.0	8.0	1.0	3.0
50.0						
119349	3	R32	87.0	13.0	4.0	6.0
74.0						
119350	3	R32	36.0	5.0	1.0	0.0
21.0						
119351	3	R32	54.0	7.0	0.0	0.0
36.0						
119352	3	R32	122.0	15.0	5.0	5.0
108.0						
	w_#1stServesIn		w_#2ndServePoints		w_%1stServesIn	

w_#1stWon \				
119348	35.0	15.0	70	19.0
119349	35.0	39.0	47	26.0
119350	12.0	9.0	57	7.0
119351	17.0	19.0	47	14.0
119352	69.0	39.0	63	46.0

	w_%1stWon	w_#2ndWon	w_%2ndWon	w_bpSaved	w_#bpFaced
l_#ServeGames \					
119348	54	8.0	53	4.0	8.0
119349	74	22.0	56	2.0	5.0
119350	58	7.0	77	0.0	1.0
119351	82	15.0	78	0.0	0.0
119352	66	18.0	46	7.0	12.0

	l_#aces	l_#dfs	l_#ServePoints	l_#1stServesIn
l_#2ndServePoints \				
119348	0.0	4.0	53.0	24.0
119349	0.0	2.0	71.0	46.0
119350	0.0	3.0	25.0	11.0
119351	2.0	2.0	64.0	36.0
119352	4.0	4.0	104.0	67.0

	l_%1stServesIn	l_#1stWon	l_%1stWon	l_#2ndWon	l_%2ndWon
l_bpSaved \					
119348	45	11.0	45	12.0	41
119349	64	22.0	47	13.0	52
119350	44	4.0	36	4.0	28
119351	56	19.0	52	10.0	35
119352	64	45.0	67	12.0	32

```

l_#bpFaced
119348      10.0
119349       9.0
119350       4.0
119351      11.0
119352      10.0

```

```
#checking shape
```

```
df_clay_B03.shape
```

```
(16996, 46)
```

```
#creating 'df_clay_B05'
```

```
df_clay_B05 = df_B05[df_B05['surface']=='Clay']
```

```
#checking head
```

```
df_clay_B05.head()
```

	tourney_id	Year	tourney_name	surface	tourney_level
winner_id \					
119588	2000-317	2000	Amsterdam	Clay	A
101320					
119666	2000-321	2000	Stuttgart Outdoor	Clay	A
102644					
120486	2000-410	2000	Monte Carlo Masters	Clay	M
101611					
120549	2000-414	2000	Hamburg Masters	Clay	M
102856					
120612	2000-416	2000	Rome Masters	Clay	M
102796					

	winner_ioc	winner_name	winner_age	winner_rank
winner_ht \				
119588	SWE	Magnus Gustafsson	33.5	63.0
185.0				
119666	ARG	Franco Squillari	24.9	20.0
183.0				
120486	FRA	Cedric Pioline	30.8	12.0
188.0				
120549	BRA	Gustavo Kuerten	23.6	7.0
190.0				
120612	SWE	Magnus Norman	23.9	4.0
188.0				

	loser_id	loser_ioc	loser_name	loser_rank	loser_ht
loser_age \					
119588	103171	NED	Raemon Sluiter	142.0	185.0
22.2					
119666	103292	ARG	Gaston Gaudio	41.0	175.0
21.6					

120486	103103	SVK	Dominik Hrbaty	24.0	183.0
22.2					
120549	103498	RUS	Marat Safin	14.0	193.0
20.2					
120612	102856	BRA	Gustavo Kuerten	6.0	190.0
23.6					

	best_of	round	minutes	w_#ServeGames	w_#aces	w_#dfs
w_#ServePoints \						
119588	5	F	155.0	21.0	6.0	4.0
118.0						
119666	5	F	228.0	22.0	8.0	2.0
130.0						
120486	5	F	160.0	17.0	12.0	5.0
127.0						
120549	5	F	232.0	28.0	7.0	1.0
165.0						
120612	5	F	185.0	20.0	14.0	1.0
121.0						

	w_#1stServesIn	w_#2ndServePoints	w_%1stServesIn	
w_#1stWon \				
119588	82.0	36.0	69	71.0
119666	63.0	67.0	48	45.0
120486	61.0	66.0	48	48.0
120549	86.0	79.0	52	64.0
120612	81.0	40.0	66	57.0

	w_%1stWon	w_#2ndWon	w_%2ndWon	w_bpSaved	w_#bpFaced
l_#ServeGames \					
119588	86	19.0	52	0.0	0.0
19.0					
119666	71	38.0	56	4.0	9.0
23.0					
120486	78	33.0	50	7.0	10.0
17.0					
120549	74	41.0	51	4.0	11.0
28.0					
120612	70	21.0	52	5.0	9.0
19.0					

	l_#aces	l_#dfs	l_#ServePoints	l_#1stServesIn
l_#2ndServePoints \				
119588	4.0	2.0	121.0	70.0
51.0				

119666	2.0	5.0	168.0	110.0
58.0				
120486	7.0	7.0	113.0	57.0
56.0				
120549	13.0	2.0	177.0	94.0
83.0				
120612	15.0	5.0	142.0	68.0
74.0				

	l_%1stServesIn	l_#1stWon	l_%1stWon	l_#2ndWon	l_%2ndWon
l_bpSaved \					
119588	57	46.0	65	33.0	64
3.0					
119666	65	63.0	57	32.0	55
10.0					
120486	50	42.0	73	22.0	39
4.0					
120549	53	66.0	70	38.0	45
4.0					
120612	47	48.0	70	34.0	45
15.0					

	l_#bpFaced
119588	6.0
119666	18.0
120486	8.0
120549	11.0
120612	21.0


```

#checking shape
df_clay_BO5.shape
#16996 + 3077 = 20073 --> checks out

(3077, 46)

```

Grass Courts

- "df_grass"
- "df_grass_BO3"
- "df_grass_BO5"

```

#creating main df for grass courts
df_grass = df_matchstats[df_matchstats['surface']=='Grass']

#checking shape
df_grass.shape
#correct number of matches

(6836, 46)

```

```
#creating 'df_grass_B03'
df_grass_B03 = df_B03[df_B03['surface']=='Grass']

#checking head
df_grass_B03.head()
```

	tourney_id	Year	tourney_name	surface	tourney_level	winner_id
\						
119410	2000-311	2000	Queen's Club	Grass	A	102179
119411	2000-311	2000	Queen's Club	Grass	A	101150
119412	2000-311	2000	Queen's Club	Grass	A	101086
119413	2000-311	2000	Queen's Club	Grass	A	102257
119414	2000-311	2000	Queen's Club	Grass	A	102533

	winner_ioc	winner_name	winner_age	winner_rank
winner_ht \				
119410	FRA	Antony Dupuis	27.2	105.0
185.0				
119411	ITA	Gianluca Pozzi	34.9	76.0
180.0				
119412	USA	Ronald Agenor	35.5	97.0
180.0				
119413	GBR	Greg Rusedski	26.7	20.0
193.0				
119414	GER	Jens Knippschild	25.3	114.0
190.0				

	loser_id	loser_ioc	loser_name	loser_rank	loser_ht
loser_age \					
119410	101320	SWE	Magnus Gustafsson	77.0	185.0
33.4					
119411	101733	NED	Jan Siemerink	125.0	183.0
30.1					
119412	101820	SUI	Marc Rosset	34.0	201.0
29.5					
119413	101965	RSA	Wayne Ferreira	42.0	185.0
28.7					
119414	102755	USA	Alex Witt	498.0	NaN
24.2					

	best_of	round	minutes	w_#ServeGames	w_#aces	w_#dfs
w_#ServePoints \						
119410	3	R64	113.0	15.0	5.0	0.0
102.0						
119411	3	R64	112.0	13.0	4.0	5.0

92.0						
119412	3	R64	42.0	6.0	2.0	0.0
31.0						
119413	3	R64	67.0	11.0	14.0	5.0
61.0						
119414	3	R64	96.0	14.0	7.0	3.0
70.0						

	w_#1stServesIn	w_#2ndServePoints	w_%1stServesIn	
w_#1stWon \				
119410	63.0	39.0	61	45.0
119411	47.0	45.0	51	35.0
119412	22.0	9.0	70	20.0
119413	34.0	27.0	55	28.0
119414	34.0	36.0	48	29.0

	w_%1stWon	w_#2ndWon	w_%2ndWon	w_bpSaved	w_#bpFaced
l_#ServeGames \					
119410	71	22.0	56	10.0	12.0
14.0					
119411	74	25.0	55	9.0	11.0
13.0					
119412	90	4.0	44	0.0	0.0
5.0					
119413	82	15.0	55	3.0	4.0
11.0					
119414	85	22.0	61	2.0	4.0
14.0					

	l_#aces	l_#dfs	l_#ServePoints	l_#1stServesIn
l_#2ndServePoints \				
119410	6.0	7.0	78.0	44.0
34.0				
119411	6.0	7.0	97.0	54.0
43.0				
119412	3.0	5.0	36.0	16.0
20.0				
119413	7.0	7.0	68.0	33.0
35.0				
119414	8.0	4.0	97.0	50.0
47.0				

	l_%1stServesIn	l_#1stWon	l_%1stWon	l_#2ndWon	l_%2ndWon
l_bpSaved \					
119410	56	34.0	77	17.0	50


```

4.0
119411          55          33.0          61          20.0          46
11.0
119412          44          12.0          75          7.0          35
2.0
119413          48          22.0          66          19.0          54
5.0
119414          51          32.0          64          26.0          55
5.0

```

```

      l_#bpFaced
119410          7.0
119411         16.0
119412          4.0
119413          8.0
119414          8.0

```

#checking shape

```
df_grass_B03.shape
```

```
(4044, 46)
```

#creating 'df_grass_B05'

```
df_grass_B05 = df_B05[df_B05['surface']=='Grass']
```

#checking head

```
df_grass_B05.head()
```

	tourney_id	Year	tourney_name	surface	tourney_level	
winner_id \						
121574	2000-540	2000	Wimbledon	Grass	G	101948
121575	2000-540	2000	Wimbledon	Grass	G	102344
121576	2000-540	2000	Wimbledon	Grass	G	103566
121577	2000-540	2000	Wimbledon	Grass	G	102925
121578	2000-540	2000	Wimbledon	Grass	G	103252

	winner_ioc	winner_name	winner_age	winner_rank
winner_ht \				
121574	USA	Pete Sampras	28.8	3.0
185.0				
121575	SVK	Karol Kucera	26.3	44.0
188.0				
121576	FRA	Michael Llodra	20.1	158.0
190.0				
121577	USA	Justin Gimelstob	23.4	99.0
196.0				

121578	ESP	Alberto Martin	21.8	67.0
--------	-----	----------------	------	------

175.0

	loser_id	loser_ioc	loser_name	loser_rank	loser_ht
--	----------	-----------	------------	------------	----------

loser_age \

121574	103181	CZE	Jiri Vanek	80.0	185.0
--------	--------	-----	------------	------	-------

22.1

121575	102286	ZIM	Wayne Black	166.0	170.0
--------	--------	-----	-------------	-------	-------

26.6

121576	102223	MAR	Karim Alami	29.0	185.0
--------	--------	-----	-------------	------	-------

27.0

121577	102443	GBR	Barry Cowan	172.0	188.0
--------	--------	-----	-------------	-------	-------

25.8

121578	102381	AUT	Werner Eschauer	176.0	188.0
--------	--------	-----	-----------------	-------	-------

26.1

	best_of	round	minutes	w_#ServeGames	w_#aces	w_#dfs
--	---------	-------	---------	---------------	---------	--------

w_#ServePoints \

121574	5	R128	83.0	14.0	10.0	4.0
--------	---	------	------	------	------	-----

73.0

121575	5	R128	88.0	13.0	13.0	9.0
--------	---	------	------	------	------	-----

68.0

121576	5	R128	78.0	13.0	4.0	4.0
--------	---	------	------	------	-----	-----

72.0

121577	5	R128	159.0	21.0	28.0	7.0
--------	---	------	-------	------	------	-----

131.0

121578	5	R128	128.0	18.0	3.0	1.0
--------	---	------	-------	------	-----	-----

91.0

	w_#1stServesIn	w_#2ndServePoints	w_%1stServesIn
--	----------------	-------------------	----------------

w_#1stWon \

121574	48.0	25.0	65	43.0
--------	------	------	----	------

121575	36.0	32.0	52	33.0
--------	------	------	----	------

121576	44.0	28.0	61	33.0
--------	------	------	----	------

121577	81.0	50.0	61	64.0
--------	------	------	----	------

121578	74.0	17.0	81	57.0
--------	------	------	----	------

	w_%1stWon	w_#2ndWon	w_%2ndWon	w_bpSaved	w_#bpFaced
--	-----------	-----------	-----------	-----------	------------

l_#ServeGames \

121574	89	15.0	60	2.0	2.0
--------	----	------	----	-----	-----

14.0

121575	91	18.0	56	2.0	3.0
--------	----	------	----	-----	-----

13.0

121576	75	17.0	60	2.0	4.0
--------	----	------	----	-----	-----

12.0

```

121577      79      27.0      54      4.0      6.0
19.0
121578      77      11.0      64      5.0      7.0
19.0

      l_#aces l_#dfs l_#ServePoints l_#1stServesIn
l_#2ndServePoints \
121574      3.0      3.0      94.0      59.0
35.0
121575      3.0      5.0      87.0      43.0
44.0
121576      3.0      7.0      80.0      50.0
30.0
121577     10.0      7.0     141.0      94.0
47.0
121578      2.0      4.0     124.0      91.0
33.0

      l_%1stServesIn l_#1stWon l_%1stWon l_#2ndWon l_%2ndWon
l_bpSaved \
121574      62      40.0      67      16.0      45
2.0
121575      49      22.0      51      25.0      56
6.0
121576      62      30.0      60      9.0      30
8.0
121577      66      67.0      71      17.0      36
12.0
121578      73      55.0      60      18.0      54
8.0

      l_#bpFaced
121574      6.0
121575     12.0
121576     15.0
121577     17.0
121578     13.0

#checking shape
df_grass_B05.shape
#4,044 + 2,792 = 6,836 --> Checks Out

(2792, 46)

```

3. Exporting Subsets

List of New Main Dataframe and Subsets

- Main Dataframe: "df_matchstats"
- By Number of Sets: "df_BO3" and "df_BO5"

- Hard Courts (3 total): "df_hard", "df_hard_BO3" and "df_hard_BO5"
- Clay Courts (3 total): "df_clay", "df_clay_BO3" and "df_clay_BO5"
- Grass Courts (3 total): "df_grass", "df_grass_BO3" and "df_grass_BO5"

```
#df_matchstats (main dataframe)
df_matchstats.to_pickle(os.path.join(path, 'Prepared Data',
'df_matchstats.pkl'))

#save df_matchstats as CSV
df_matchstats.to_csv(os.path.join(path, 'Prepared
Data', 'df_matchstats.csv'))

#df_B03
df_B03.to_pickle(os.path.join(path, 'Prepared Data',
'Subsets', 'df_B03.pkl'))

#df_B05
df_B05.to_pickle(os.path.join(path, 'Prepared Data',
'Subsets', 'df_B05.pkl'))

#df_hard
df_hard.to_pickle(os.path.join(path, 'Prepared Data',
'Subsets', 'df_hard.pkl'))

#df_hard_B03
df_hard_B03.to_pickle(os.path.join(path, 'Prepared Data',
'Subsets', 'df_hard_B03.pkl'))

#df_hard_B05
df_hard_B05.to_pickle(os.path.join(path, 'Prepared Data',
'Subsets', 'df_hard_B05.pkl'))

#df_clay
df_clay.to_pickle(os.path.join(path, 'Prepared Data',
'Subsets', 'df_clay.pkl'))

#df_clay_B03
df_clay_B03.to_pickle(os.path.join(path, 'Prepared Data',
'Subsets', 'df_clay_B03.pkl'))

#df_clay_B05
df_clay_B05.to_pickle(os.path.join(path, 'Prepared Data',
'Subsets', 'df_clay_B05.pkl'))

#df_grass
df_grass.to_pickle(os.path.join(path, 'Prepared Data',
'Subsets', 'df_grass.pkl'))

#df_grass_B05
df_grass_B05.to_pickle(os.path.join(path, 'Prepared Data',
'Subsets', 'df_grass_B05.pkl'))
```

```
#df_grass_B03  
df_grass_B03.to_pickle(os.path.join(path, 'Prepared Data',  
'Subsets', 'df_grass_B03.pkl'))
```