

semesterProject

December 3, 2025

1 NBA retirement prediction

```
[572]: # Install dependencies as needed:  
# pip install kagglehub  
import kagglehub  
from kagglehub import KaggleDatasetAdapter  
import pandas as pd  
import matplotlib.pyplot as plt  
import numpy as np  
from sklearn.linear_model import LinearRegression  
from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score  
from sklearn.model_selection import train_test_split
```

2 DATA:

3 Player Totals

```
[573]: # Set the path to the file you'd like to load  
file_path = "Player Totals.csv"  
  
# Load the latest version  
playerTotalsDf = kagglehub.dataset_load(  
    KaggleDatasetAdapter.PANDAS,  
    "sumitrodatta/nba-aba-baa-stats",  
    file_path,  
    # Provide any additional arguments like  
    # sql_query or pandas_kwargs. See the  
    # documentation for more information:  
    # https://github.com/Kaggle/kagglehub/blob/main/README.  
    #md#kaggledatasetadapterpandas  
)  
  
print(playerTotalsDf.head())
```

	season	lg	player	player_id	age	team	pos	g	gs	mp	\
0	2026	NBA	Precious Achiuwa	achiupr01	26.0	SAC	C	14	6.0	293.0	
1	2026	NBA	Steven Adams	adamsst01	32.0	HOU	C	15	4.0	324.0	

```

2    2026 NBA      Bam Adebayo adebab01 28.0 MIA C 14 14.0 425.0
3    2026 NBA      Ochai Agbaji agbajoc01 25.0 TOR SG 15 1.0 204.0
4    2026 NBA      Santi Aldama aldamsa01 25.0 MEM PF 21 2.0 548.0

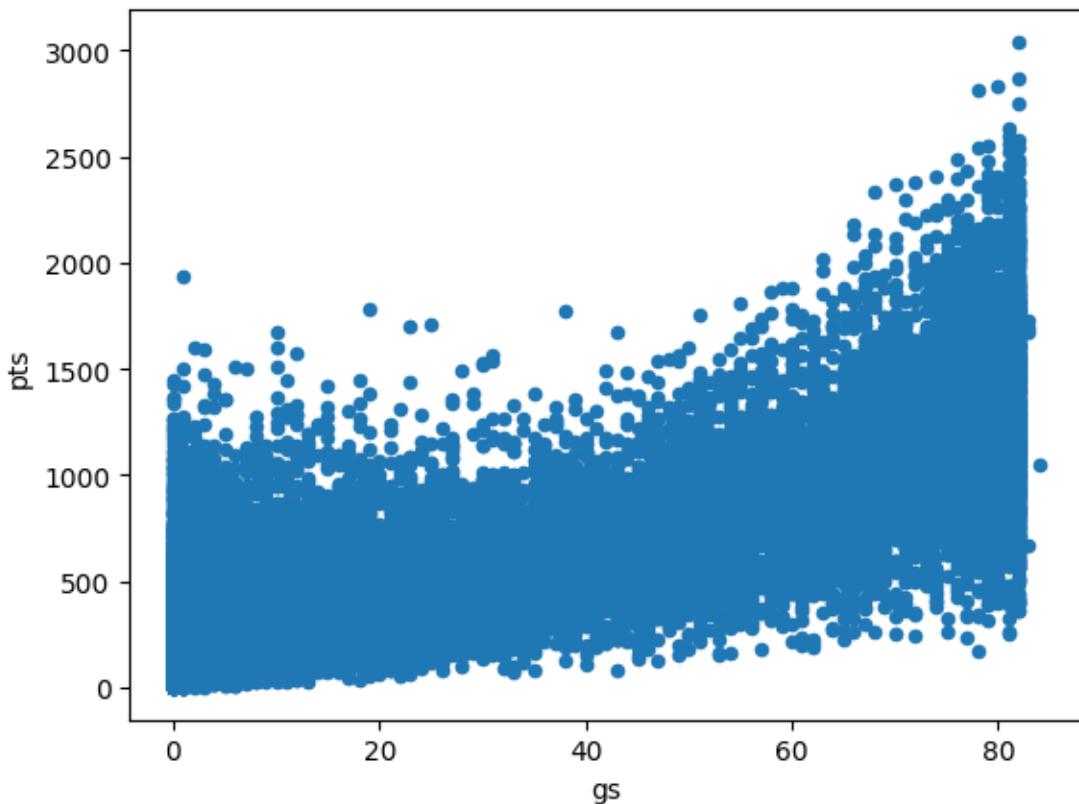
...   orb    drb    trb   ast   stl   blk   tov   pf   pts   trp_dbl
0 ... 25.0  48.0  73.0  14   9.0   6.0   7.0  24.0 102   0.0
1 ... 74.0  67.0 141.0  23  10.0  10.0  15.0 29.0  97   0.0
2 ... 18.0 101.0 119.0  36  13.0  11.0  26.0 20.0 264   0.0
3 ...  5.0  23.0  28.0  10   6.0   2.0   5.0  26.0  46   0.0
4 ... 31.0 108.0 139.0  63  20.0  16.0  23.0 26.0 282   0.0

```

[5 rows x 33 columns]

```
[574]: playerTotalsDf.plot.scatter(x='gs', y='pts')
```

```
[574]: <Axes: xlabel='gs', ylabel='pts'>
```



4 Player Per Game

```
[575]: # Set the path to the file you'd like to load
file_path = "Player Per Game.csv"

# Load the latest version
playerPerGameDf = kagglehub.dataset_load(
    KaggleDatasetAdapter.PANDAS,
    "sumitrodatta/nba-aba-baa-stats",
    file_path,
)

print(playerPerGameDf.head())
```

```
   season   lg      player  player_id   age  team pos   g   gs \
0   2026  NBA  Precious Achiuwa  achiupr01  26.0  SAC  C  14  6.0
1   2026  NBA    Steven Adams  adamsst01  32.0  HOU  C  15  4.0
2   2026  NBA     Bam Adebayo  adebaba01  28.0  MIA  C  14 14.0
3   2026  NBA    Ochai Agbaji  agbajoc01  25.0  TOR  SG  15  1.0
4   2026  NBA    Santi Aldama  aldamsa01  25.0  MEM  PF  21  2.0

   mp_per_game   ...  ft_percent  orb_per_game  drb_per_game  trb_per_game \
0        20.9   ...       0.458          1.8            3.4           5.2
1        21.6   ...       0.775          4.9            4.5           9.4
2        30.4   ...       0.806          1.3            7.2           8.5
3        13.6   ...       0.700          0.3            1.5           1.9
4        26.1   ...       0.673          1.5            5.1           6.6

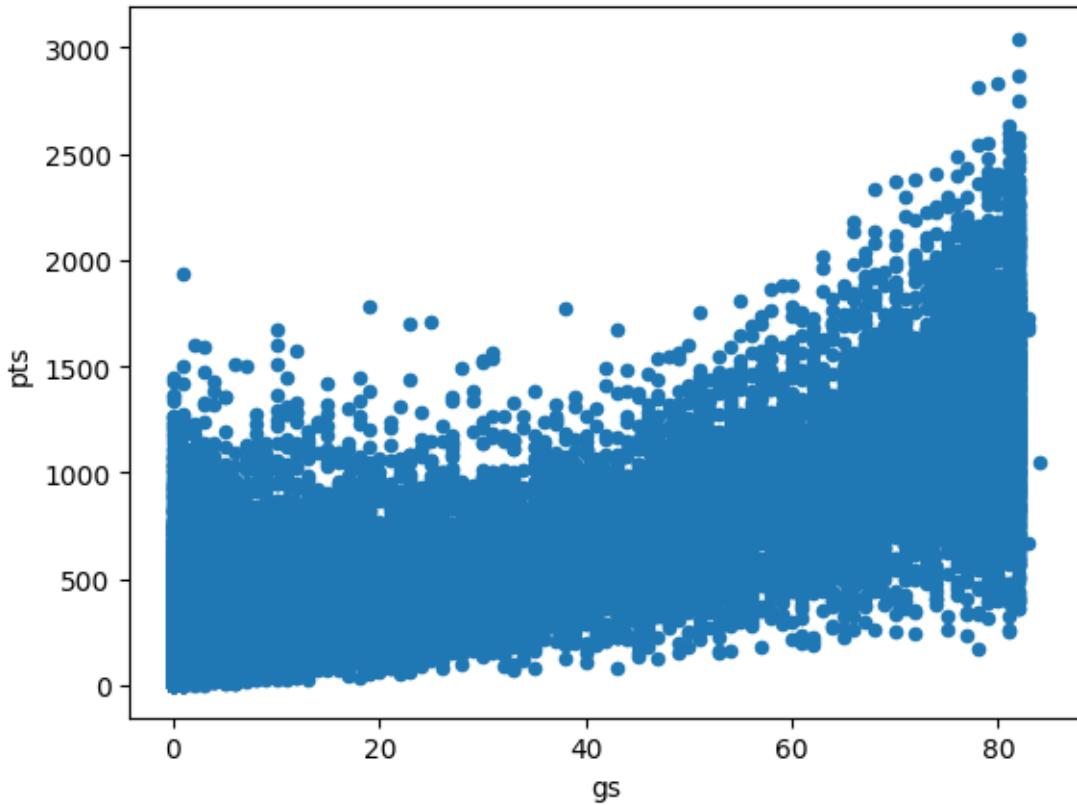
   ast_per_game  stl_per_game  blk_per_game  tov_per_game  pf_per_game \
0         1.0          0.6          0.4          0.5           1.7
1         1.5          0.7          0.7          1.0           1.9
2         2.6          0.9          0.8          1.9           1.4
3         0.7          0.4          0.1          0.3           1.7
4         3.0          1.0          0.8          1.1           1.2

   pts_per_game
0        7.3
1        6.5
2       18.9
3        3.1
4       13.4
```

[5 rows x 32 columns]

```
[576]: playerTotalsDf.plot.scatter(x='gs', y='pts')
```

```
[576]: <Axes: xlabel='gs', ylabel='pts'>
```



5 Player Per 36 Mins

```
[577]: # Set the path to the file you'd like to load
file_path = "Per 36 Minutes.csv"

# Load the latest version
per36MinsDf = kagglehub.dataset_load(
    KaggleDatasetAdapter.PANDAS,
    "sumitrodatta/nba-aba-baa-stats",
    file_path,
)

print(per36MinsDf.head())
```

	season	lg	player	player_id	age	team	pos	g	gs	mp	\
0	2026	NBA	Precious Achiuwa	achiupr01	26.0	SAC	C	14	6.0	293	
1	2026	NBA	Steven Adams	adamsst01	32.0	HOU	C	15	4.0	324	
2	2026	NBA	Bam Adebayo	adebab01	28.0	MIA	C	14	14.0	425	
3	2026	NBA	Ochai Agbaji	agbajoc01	25.0	TOR	SG	15	1.0	204	
4	2026	NBA	Santi Aldama	aldamsa01	25.0	MEM	PF	21	2.0	548	

```

... ft_percent orb_per_36_min drb_per_36_min trb_per_36_min \
0 ... 0.458 3.1 5.9 9.0
1 ... 0.775 8.2 7.4 15.7
2 ... 0.806 1.5 8.6 10.1
3 ... 0.700 0.9 4.1 4.9
4 ... 0.673 2.0 7.1 9.1

ast_per_36_min stl_per_36_min blk_per_36_min tov_per_36_min \
0 1.7 1.1 0.7 0.9
1 2.6 1.1 1.1 1.7
2 3.0 1.1 0.9 2.2
3 1.8 1.1 0.4 0.9
4 4.1 1.3 1.1 1.5

pf_per_36_min pts_per_36_min
0 2.9 12.5
1 3.2 10.8
2 1.7 22.4
3 4.6 8.1
4 1.7 18.5

[5 rows x 32 columns]

```

6 Player Per 100 Possessions

```
[578]: # Set the path to the file you'd like to load
file_path = "Per 100 Poss.csv"

# Load the latest version
per100PossDf = kagglehub.dataset_load(
    KaggleDatasetAdapter.PANDAS,
    "sumitrodatta/nba-aba-baa-stats",
    file_path,
)

print(per100PossDf.head())

```

	season	lg	player	player_id	age	team	pos	g	gs	mp	...	\
0	2026	NBA	Precious Achiuwa	achiupr01	26	SAC	C	14	6.0	293	...	
1	2026	NBA	Steven Adams	adamsst01	32	HOU	C	15	4.0	324	...	
2	2026	NBA	Bam Adebayo	adebab01	28	MIA	C	14	14.0	425	...	
3	2026	NBA	Ochai Agbaji	agbajoc01	25	TOR	SG	15	1.0	204	...	
4	2026	NBA	Santi Aldama	aldamsa01	25	MEM	PF	21	2.0	548	...	

	drb_per_100_poss	trb_per_100_poss	ast_per_100_poss	stl_per_100_poss	\
0	7.8	11.8	2.3	1.5	
1	10.3	21.7	3.5	1.5	

```

2          10.8          12.7          3.9          1.4
3          5.4           6.6          2.4          1.4
4          9.4          12.1          5.5          1.7

blk_per_100_poss  tov_per_100_poss  pf_per_100_poss  pts_per_100_poss  \
0              1.0            1.1            3.9          16.5
1              1.5            2.3            4.5          14.9
2              1.2            2.8            2.1          28.3
3              0.5            1.2            6.1          10.8
4              1.4            2.0            2.3          24.6

o_rtg  d_rtg
0  117.0  120.0
1  138.0  110.0
2  112.0  109.0
3   94.0  114.0
4  116.0  113.0

[5 rows x 34 columns]

```

7 Player Advanced Stats

```
[579]: # Set the path to the file you'd like to load
file_path = "Advanced.csv"

# Load the latest version
advancedStatsDf = kagglehub.dataset_load(
    KaggleDatasetAdapter.PANDAS,
    "sumitrodatta/nba-aba-baa-stats",
    file_path,
)

print(advancedStatsDf.head())
```

	season	lg	player	player_id	age	team	pos	g	gs	mp	\	
0	2026	NBA	Precious Achiuwa	achiupr01	26.0	SAC	C	14	6.0	293.0		
1	2026	NBA	Steven Adams	adamsst01	32.0	HOU	C	15	4.0	324.0		
2	2026	NBA	Bam Adebayo	adebab01	28.0	MIA	C	14	14.0	425.0		
3	2026	NBA	Ochai Agbaji	agbajoc01	25.0	TOR	SG	15	1.0	204.0		
4	2026	NBA	Santi Aldama	aldamsa01	25.0	MEM	PF	21	2.0	548.0		
	...		tov_percent	usg_percent	ows	dws	ws	ws_48	obpm	dbpm	bpm	vorp
0	...		7.1	14.3	0.3	0.2	0.4	0.068	-1.4	-0.8	-2.2	0.0
1	...		15.4	12.4	0.9	0.5	1.4	0.208	2.3	-0.3	1.9	0.3
2	...		10.1	24.5	0.3	0.8	1.1	0.119	0.9	1.1	2.0	0.4
3	...		8.6	12.1	-0.2	0.2	0.1	0.017	-6.6	-0.3	-6.9	-0.3
4	...		8.3	20.9	0.6	0.8	1.3	0.118	1.9	1.2	3.1	0.7

```
[5 rows x 30 columns]
```

8 Player play by play stats

```
[580]: # Set the path to the file you'd like to load
file_path = "Player Play By Play.csv"

# Load the latest version
playByPlayDf = kagglehub.dataset_load(
    KaggleDatasetAdapter.PANDAS,
    "sumitrodratta/nba-aba-baa-stats",
    file_path,
)

print(playByPlayDf.head())
```

	season	lg	player	player_id	age	team	pos	g	gs	mp	...	\
0	2026	NBA	Precious Achiuwa	achiupr01	26	SAC	C	14	6	293	...	
1	2026	NBA	Steven Adams	adamsst01	32	HOU	C	15	4	324	...	
2	2026	NBA	Bam Adebayo	adebab01	28	MIA	C	14	14	425	...	
3	2026	NBA	Ochai Agbaji	agbajoc01	25	TOR	SG	15	1	204	...	
4	2026	NBA	Santi Aldama	aldamsa01	25	MEM	PF	21	2	548	...	

	net_plus_minus_per_100_poss	bad_pass_turnover	lost_ball_turnover	\
0	6.4	2	1	
1	9.8	5	3	
2	8.5	10	11	
3	1.3	2	2	
4	-3.5	11	7	

	shooting_foul_committed	offensive_foul_committed	shooting_foul_drawn	\
0	12	3	11	
1	9	5	14	
2	6	2	33	
3	13	1	4	
4	15	4	21	

	offensive_foul_drawn	points_generated_by_assists	and1	fga_blocked
0	2.0	31	6	4
1	0.0	52	1	2
2	2.0	83	3	7
3	2.0	23	0	5
4	3.0	160	8	7

```
[5 rows x 26 columns]
```

9 Height and Weight Data

```
[581]: # Set the path to the file you'd like to load
file_path = "Players.csv"

# Load the latest version
HeightAndWeightDf = kagglehub.dataset_load(
    KaggleDatasetAdapter.PANDAS,
    "drgilermo/nba-players-stats",
    file_path,
)

HeightAndWeightDf = HeightAndWeightDf.rename(columns={'Player': 'player'})
print(HeightAndWeightDf.head())
```

```
      Unnamed: 0      player   height  weight \
0          0  Curly Armstrong    180.0    77.0
1          1     Cliff Barker    188.0    83.0
2          2     Leo Barnhorst    193.0    86.0
3          3      Ed Bartels    196.0    88.0
4          4     Ralph Beard    178.0    79.0

           collage      born  birth_city birth_state
0  Indiana University  1918.0        NaN       NaN
1  University of Kentucky  1921.0  Yorktown       Indiana
2  University of Notre Dame  1924.0        NaN       NaN
3  North Carolina State University  1925.0        NaN       NaN
4  University of Kentucky  1927.0  Hardinsburg    Kentucky
```

10 Injury Data

```
[582]: # Set the path to the file you'd like to load
file_path = "injuries_2010-2020.csv"

# Load the latest version
injuryDF = kagglehub.dataset_load(
    KaggleDatasetAdapter.PANDAS,
    "ghopkins/nba-injuries-2010-2018",
    file_path,
)
injuryDF = injuryDF.rename(columns={'Relinquished': 'player'})

injury_counts = injuryDF['player'].value_counts().reset_index()
injury_counts.columns = ['player', 'num_injuries']

print(injuryDF.head())
```

	Date	Team	Acquired	player	\
0	2010-10-03	Bulls	NaN	Carlos Boozer	
1	2010-10-06	Pistons	NaN	Jonas Jerebko	
2	2010-10-06	Pistons	NaN	Terrico White	
3	2010-10-08	Blazers	NaN	Jeff Ayres	
4	2010-10-08	Nets	NaN	Troy Murphy	

	Notes
0	fractured bone in right pinky finger (out indefinitely)
1	torn right Achilles tendon (out indefinitely)
2	broken fifth metatarsal in right foot (out indefinitely)
3	torn ACL in right knee (out indefinitely)
4	strained lower back (out indefinitely)

11 Merging data into 1 dataset

```
[583]: #Merge data into one big dataset-Amelia
mergedDF = playerTotalsDf
mergedDF = pd.merge(mergedDF, playerPerGameDf, on=["player_id", "season"], how="outer", suffixes=("","_per_game"))
mergedDF = pd.merge(mergedDF, per36MinsDf, on=["player_id", "season"], how="outer", suffixes=("","_per_36_mins"))
mergedDF = pd.merge(mergedDF, per100PossDf, on=["player_id", "season"], how="outer", suffixes=("","_per_100_poss"))
mergedDF = pd.merge(mergedDF, advancedStatsDf, on=["player_id", "season"], how="outer", suffixes=("","_advanced"))
finalDF = mergedDF.drop_duplicates(subset=['season', 'player'])

print(finalDF[26700:26730])
```

	season	lg	player	player_id	age	team	pos	g	gs	\
803813	1960	NBA	George Yardley	yardlge01	31.0	SYR	SF	73	NaN	
803814	1972	NBA	Barry Yates	yatesba01	26.0	PHI	PF	24	1.0	
803815	1962	NBA	Wayne Yates	yateswa01	24.0	LAL	C	37	NaN	
803816	1972	NBA	Charlie Yelverton	yelvech01	23.0	POR	SG	69	NaN	
803817	1982	NBA	Rich Yonakor	yonakri01	23.0	SAS	PF	10	0.0	
803818	2022	NBA	Gabe York	yorkga01	28.0	IND	SG	2	0.0	
803819	2023	NBA	Gabe York	yorkga01	29.0	IND	SG	3	0.0	
803820	2026	NBA	Chris Youngblood	youngch01	23.0	OKC	SG	16	0.0	
803821	1985	NBA	Danny Young	youngda01	22.0	SEA	PG	3	0.0	
803822	1986	NBA	Danny Young	youngda01	23.0	SEA	PG	82	29.0	
803823	1987	NBA	Danny Young	youngda01	24.0	SEA	PG	73	26.0	
803824	1988	NBA	Danny Young	youngda01	25.0	SEA	PG	77	0.0	
803825	1989	NBA	Danny Young	youngda01	26.0	POR	PG	48	2.0	
803826	1990	NBA	Danny Young	youngda01	27.0	POR	PG	82	8.0	
803827	1991	NBA	Danny Young	youngda01	28.0	POR	PG	75	1.0	

803828	1992	NBA	Danny Young	youngda01	29.0	2TM	PG	62	5.0
804071	1993	NBA	Danny Young	youngda01	30.0	DET	PG	65	2.0
804072	1995	NBA	Danny Young	youngda01	32.0	MIL	PG	7	0.0
804073	2015	NBA	James Young	youngja01	19.0	BOS	SG	31	0.0
804074	2016	NBA	James Young	youngja01	20.0	BOS	SG	29	0.0
804075	2017	NBA	James Young	youngja01	21.0	BOS	SG	29	0.0
804076	2018	NBA	James Young	youngja01	22.0	PHI	SG	6	0.0
804077	2025	NBA	Jahmir Young	youngja05	24.0	CHI	PG	6	0.0
804078	2026	NBA	Jahmir Young	youngja05	25.0	MIA	PG	4	0.0
804079	2016	NBA	Joe Young	youngjo01	23.0	IND	PG	41	0.0
804080	2017	NBA	Joe Young	youngjo01	24.0	IND	PG	33	0.0
804081	2018	NBA	Joe Young	youngjo01	25.0	IND	PG	53	1.0
804082	1999	NBA	Korleone Young	youngko01	20.0	DET	SF	3	0.0
804083	1985	NBA	Michael Young	youngmi01	24.0	PHO	SF	2	0.0
804084	1986	NBA	Michael Young	youngmi01	25.0	PHI	SF	2	0.0

	mp	...	tov_percent	usg_percent	ows	dws	ws	ws_48	obpm	\	
803813	2402.0	...		NaN		NaN	6.9	2.1	9.0	0.179	NaN
803814	144.0	...		NaN		NaN	-0.4	0.1	-0.2	-0.075	NaN
803815	263.0	...		NaN		NaN	-0.8	0.3	-0.5	-0.085	NaN
803816	1227.0	...		NaN		NaN	-0.7	0.3	-0.5	-0.018	NaN
803817	70.0	...	6.4		17.7	0.2	0.1	0.3	0.201	-1.2	
803818	21.0	...	9.8		20.8	0.0	0.0	0.0	0.059	-3.7	
803819	56.0	...	0.0		16.4	0.1	0.0	0.1	0.091	-1.7	
803820	80.0	...	3.9		13.8	0.1	0.1	0.2	0.114	-4.6	
803821	26.0	...	16.7		19.3	-0.2	0.0	-0.1	-0.188	-9.9	
803822	1901.0	...	15.7		12.9	2.7	2.1	4.8	0.121	0.2	
803823	1482.0	...	21.0		10.7	2.0	0.9	2.9	0.095	0.8	
803824	949.0	...	13.3		11.8	1.6	0.8	2.4	0.122	0.4	
803825	952.0	...	13.9		13.1	1.0	0.8	1.8	0.091	-1.1	
803826	1393.0	...	17.5		12.9	1.0	1.9	3.0	0.103	-1.2	
803827	897.0	...	14.7		15.4	0.6	1.2	1.8	0.096	-0.6	
803828	1023.0	...	14.2		13.7	0.8	1.1	1.9	0.089	-1.4	
804071	836.0	...	14.2		10.8	0.9	0.5	1.4	0.081	-1.5	
804072	77.0	...	18.7		12.3	0.2	0.0	0.2	0.141	0.9	
804073	332.0	...	4.2		15.8	0.1	0.3	0.3	0.047	-2.5	
804074	199.0	...	11.7		9.1	-0.2	0.2	0.1	0.013	-5.3	
804075	220.0	...	6.1		13.3	0.1	0.2	0.3	0.076	-1.7	
804076	61.0	...	5.7		12.3	0.0	0.0	0.1	0.041	-4.5	
804077	30.0	...	14.5		9.6	0.1	0.0	0.1	0.220	1.2	
804078	17.0	...	9.1		26.2	0.0	0.0	0.0	-0.001	-4.3	
804079	384.0	...	15.5		24.4	-0.5	0.5	-0.1	-0.009	-3.9	
804080	135.0	...	6.0		27.8	-0.2	0.1	-0.1	-0.022	-3.1	
804081	558.0	...	11.6		18.0	0.1	0.4	0.5	0.043	-2.3	
804082	15.0	...	8.4		37.6	0.1	0.0	0.1	0.280	9.9	
804083	11.0	...	0.0		22.2	0.0	0.0	0.0	-0.072	-2.5	
804084	2.0	...	0.0		40.3	0.0	0.0	0.0	-0.818	-28.9	

	dbpm	bpm	vorp
803813	NaN	NaN	NaN
803814	NaN	NaN	NaN
803815	NaN	NaN	NaN
803816	NaN	NaN	NaN
803817	-0.6	-1.8	0.1
803818	2.2	-1.5	0.0
803819	-1.8	-3.5	0.0
803820	-1.0	-5.6	-0.1
803821	3.8	-6.1	0.0
803822	1.7	1.9	1.9
803823	0.3	1.1	1.2
803824	0.7	1.1	0.8
803825	1.2	0.0	0.5
803826	2.1	0.9	1.0
803827	1.1	0.5	0.6
803828	0.6	-0.8	0.3
804071	1.0	-0.5	0.3
804072	2.4	3.3	0.1
804073	-0.9	-3.4	-0.1
804074	0.8	-4.5	-0.1
804075	0.4	-1.3	0.0
804076	-2.4	-6.9	-0.1
804077	-0.8	0.5	0.0
804078	-2.6	-7.0	0.0
804079	-0.8	-4.7	-0.3
804080	-2.1	-5.2	-0.1
804081	-1.2	-3.5	-0.2
804082	-0.5	9.4	0.0
804083	-5.6	-8.0	0.0
804084	-19.1	-48.0	0.0

[30 rows x 153 columns]

12 Creating column for year retired, NaN if still playing

```
[584]: # players not retired yet
activePlayers = finalDF.groupby("player")["season"].max().eq(2026)

retirementYears = finalDF.groupby("player")["season"].max().
    ↪where(~activePlayers)

finalDF['retired'] = finalDF['player'].map(retirementYears)

finalDF[26700:26730]
```

/var/folders/_b/qm30jgm53fqdy5bnwf_lv5x00000gn/T/ipykernel_76804/2776414382.py:6

: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
finalDF['retired'] = finalDF['player'].map(retirementYears)
```

[584]:		season	lg	player	player_id	age	team	pos	g	gs	\
803813	1960	NBA	George Yardley	yardlge01	31.0	SYR	SF	73	NaN		
803814	1972	NBA	Barry Yates	yatesba01	26.0	PHI	PF	24	1.0		
803815	1962	NBA	Wayne Yates	yateswa01	24.0	LAL	C	37	NaN		
803816	1972	NBA	Charlie Yelverton	yelvech01	23.0	POR	SG	69	NaN		
803817	1982	NBA	Rich Yonakor	yonakri01	23.0	SAS	PF	10	0.0		
803818	2022	NBA	Gabe York	yorkgaa01	28.0	IND	SG	2	0.0		
803819	2023	NBA	Gabe York	yorkgaa01	29.0	IND	SG	3	0.0		
803820	2026	NBA	Chris Youngblood	youngch01	23.0	OKC	SG	16	0.0		
803821	1985	NBA	Danny Young	youngda01	22.0	SEA	PG	3	0.0		
803822	1986	NBA	Danny Young	youngda01	23.0	SEA	PG	82	29.0		
803823	1987	NBA	Danny Young	youngda01	24.0	SEA	PG	73	26.0		
803824	1988	NBA	Danny Young	youngda01	25.0	SEA	PG	77	0.0		
803825	1989	NBA	Danny Young	youngda01	26.0	POR	PG	48	2.0		
803826	1990	NBA	Danny Young	youngda01	27.0	POR	PG	82	8.0		
803827	1991	NBA	Danny Young	youngda01	28.0	POR	PG	75	1.0		
803828	1992	NBA	Danny Young	youngda01	29.0	2TM	PG	62	5.0		
804071	1993	NBA	Danny Young	youngda01	30.0	DET	PG	65	2.0		
804072	1995	NBA	Danny Young	youngda01	32.0	MIL	PG	7	0.0		
804073	2015	NBA	James Young	youngja01	19.0	BOS	SG	31	0.0		
804074	2016	NBA	James Young	youngja01	20.0	BOS	SG	29	0.0		
804075	2017	NBA	James Young	youngja01	21.0	BOS	SG	29	0.0		
804076	2018	NBA	James Young	youngja01	22.0	PHI	SG	6	0.0		
804077	2025	NBA	Jahmir Young	youngja05	24.0	CHI	PG	6	0.0		
804078	2026	NBA	Jahmir Young	youngja05	25.0	MIA	PG	4	0.0		
804079	2016	NBA	Joe Young	youngjo01	23.0	IND	PG	41	0.0		
804080	2017	NBA	Joe Young	youngjo01	24.0	IND	PG	33	0.0		
804081	2018	NBA	Joe Young	youngjo01	25.0	IND	PG	53	1.0		
804082	1999	NBA	Korleone Young	youngko01	20.0	DET	SF	3	0.0		
804083	1985	NBA	Michael Young	youngmi01	24.0	PHO	SF	2	0.0		
804084	1986	NBA	Michael Young	youngmi01	25.0	PHI	SF	2	0.0		
	mp	...	usg_percent	ows	dws	ws	ws_48	obpm	dbpm	bpm	\
803813	2402.0	...	NaN	6.9	2.1	9.0	0.179	NaN	NaN	NaN	
803814	144.0	...	NaN	-0.4	0.1	-0.2	-0.075	NaN	NaN	NaN	
803815	263.0	...	NaN	-0.8	0.3	-0.5	-0.085	NaN	NaN	NaN	
803816	1227.0	...	NaN	-0.7	0.3	-0.5	-0.018	NaN	NaN	NaN	
803817	70.0	...	17.7	0.2	0.1	0.3	0.201	-1.2	-0.6	-1.8	
803818	21.0	...	20.8	0.0	0.0	0.0	0.059	-3.7	2.2	-1.5	

803819	56.0	...	16.4	0.1	0.0	0.1	0.091	-1.7	-1.8	-3.5
803820	80.0	...	13.8	0.1	0.1	0.2	0.114	-4.6	-1.0	-5.6
803821	26.0	...	19.3	-0.2	0.0	-0.1	-0.188	-9.9	3.8	-6.1
803822	1901.0	...	12.9	2.7	2.1	4.8	0.121	0.2	1.7	1.9
803823	1482.0	...	10.7	2.0	0.9	2.9	0.095	0.8	0.3	1.1
803824	949.0	...	11.8	1.6	0.8	2.4	0.122	0.4	0.7	1.1
803825	952.0	...	13.1	1.0	0.8	1.8	0.091	-1.1	1.2	0.0
803826	1393.0	...	12.9	1.0	1.9	3.0	0.103	-1.2	2.1	0.9
803827	897.0	...	15.4	0.6	1.2	1.8	0.096	-0.6	1.1	0.5
803828	1023.0	...	13.7	0.8	1.1	1.9	0.089	-1.4	0.6	-0.8
804071	836.0	...	10.8	0.9	0.5	1.4	0.081	-1.5	1.0	-0.5
804072	77.0	...	12.3	0.2	0.0	0.2	0.141	0.9	2.4	3.3
804073	332.0	...	15.8	0.1	0.3	0.3	0.047	-2.5	-0.9	-3.4
804074	199.0	...	9.1	-0.2	0.2	0.1	0.013	-5.3	0.8	-4.5
804075	220.0	...	13.3	0.1	0.2	0.3	0.076	-1.7	0.4	-1.3
804076	61.0	...	12.3	0.0	0.0	0.1	0.041	-4.5	-2.4	-6.9
804077	30.0	...	9.6	0.1	0.0	0.1	0.220	1.2	-0.8	0.5
804078	17.0	...	26.2	0.0	0.0	0.0	-0.001	-4.3	-2.6	-7.0
804079	384.0	...	24.4	-0.5	0.5	-0.1	-0.009	-3.9	-0.8	-4.7
804080	135.0	...	27.8	-0.2	0.1	-0.1	-0.022	-3.1	-2.1	-5.2
804081	558.0	...	18.0	0.1	0.4	0.5	0.043	-2.3	-1.2	-3.5
804082	15.0	...	37.6	0.1	0.0	0.1	0.280	9.9	-0.5	9.4
804083	11.0	...	22.2	0.0	0.0	0.0	-0.072	-2.5	-5.6	-8.0
804084	2.0	...	40.3	0.0	0.0	0.0	-0.818	-28.9	-19.1	-48.0

	vorp	retired
803813	NaN	1960.0
803814	NaN	1972.0
803815	NaN	1962.0
803816	NaN	1972.0
803817	0.1	1982.0
803818	0.0	2023.0
803819	0.0	2023.0
803820	-0.1	NaN
803821	0.0	1995.0
803822	1.9	1995.0
803823	1.2	1995.0
803824	0.8	1995.0
803825	0.5	1995.0
803826	1.0	1995.0
803827	0.6	1995.0
803828	0.3	1995.0
804071	0.3	1995.0
804072	0.1	1995.0
804073	-0.1	2018.0
804074	-0.1	2018.0
804075	0.0	2018.0

```

804076 -0.1    2018.0
804077  0.0     NaN
804078  0.0     NaN
804079 -0.3    2018.0
804080 -0.1    2018.0
804081 -0.2    2018.0
804082  0.0    1999.0
804083  0.0    1990.0
804084  0.0    1990.0

```

[30 rows x 154 columns]

```

[585]: ret_year_mp = finalDF.loc[finalDF['season'] == finalDF['retired'], [
    ['player_id', 'retired', 'mp']].rename(columns={'mp': 'mp_at_retirement'})]
finalDF = finalDF.merge(ret_year_mp, on=['player_id', 'retired'], how='left')
finalDF[26700:26730]

```

	season	lg	player	player_id	age	team	pos	g	gs	\
26700	1960	NBA	George Yardley	yardlge01	31.0	SYR	SF	73	NaN	
26701	1972	NBA	Barry Yates	yatesba01	26.0	PHI	PF	24	1.0	
26702	1962	NBA	Wayne Yates	yateswa01	24.0	LAL	C	37	NaN	
26703	1972	NBA	Charlie Yelverton	yelvech01	23.0	POR	SG	69	NaN	
26704	1982	NBA	Rich Yonakor	yonakri01	23.0	SAS	PF	10	0.0	
26705	2022	NBA	Gabe York	yorkga01	28.0	IND	SG	2	0.0	
26706	2023	NBA	Gabe York	yorkga01	29.0	IND	SG	3	0.0	
26707	2026	NBA	Chris Youngblood	youngch01	23.0	OKC	SG	16	0.0	
26708	1985	NBA	Danny Young	youngda01	22.0	SEA	PG	3	0.0	
26709	1986	NBA	Danny Young	youngda01	23.0	SEA	PG	82	29.0	
26710	1987	NBA	Danny Young	youngda01	24.0	SEA	PG	73	26.0	
26711	1988	NBA	Danny Young	youngda01	25.0	SEA	PG	77	0.0	
26712	1989	NBA	Danny Young	youngda01	26.0	POR	PG	48	2.0	
26713	1990	NBA	Danny Young	youngda01	27.0	POR	PG	82	8.0	
26714	1991	NBA	Danny Young	youngda01	28.0	POR	PG	75	1.0	
26715	1992	NBA	Danny Young	youngda01	29.0	2TM	PG	62	5.0	
26716	1993	NBA	Danny Young	youngda01	30.0	DET	PG	65	2.0	
26717	1995	NBA	Danny Young	youngda01	32.0	MIL	PG	7	0.0	
26718	2015	NBA	James Young	youngja01	19.0	BOS	SG	31	0.0	
26719	2016	NBA	James Young	youngja01	20.0	BOS	SG	29	0.0	
26720	2017	NBA	James Young	youngja01	21.0	BOS	SG	29	0.0	
26721	2018	NBA	James Young	youngja01	22.0	PHI	SG	6	0.0	
26722	2025	NBA	Jahmir Young	youngja05	24.0	CHI	PG	6	0.0	
26723	2026	NBA	Jahmir Young	youngja05	25.0	MIA	PG	4	0.0	
26724	2016	NBA	Joe Young	youngjo01	23.0	IND	PG	41	0.0	
26725	2017	NBA	Joe Young	youngjo01	24.0	IND	PG	33	0.0	
26726	2018	NBA	Joe Young	youngjo01	25.0	IND	PG	53	1.0	
26727	1999	NBA	Korleone Young	youngko01	20.0	DET	SF	3	0.0	
26728	1985	NBA	Michael Young	youngmi01	24.0	PHO	SF	2	0.0	

26729	1986	NBA	Michael Young	youngmi01	25.0	PHI	SF	2	0.0
-------	------	-----	---------------	-----------	------	-----	----	---	-----

	mp	...	ows	dws	ws	ws_48	obpm	dbpm	bpm	vorp	retired	\
26700	2402.0	...	6.9	2.1	9.0	0.179	NaN	NaN	NaN	NaN	1960.0	
26701	144.0	...	-0.4	0.1	-0.2	-0.075	NaN	NaN	NaN	NaN	1972.0	
26702	263.0	...	-0.8	0.3	-0.5	-0.085	NaN	NaN	NaN	NaN	1962.0	
26703	1227.0	...	-0.7	0.3	-0.5	-0.018	NaN	NaN	NaN	NaN	1972.0	
26704	70.0	...	0.2	0.1	0.3	0.201	-1.2	-0.6	-1.8	0.1	1982.0	
26705	21.0	...	0.0	0.0	0.0	0.059	-3.7	2.2	-1.5	0.0	2023.0	
26706	56.0	...	0.1	0.0	0.1	0.091	-1.7	-1.8	-3.5	0.0	2023.0	
26707	80.0	...	0.1	0.1	0.2	0.114	-4.6	-1.0	-5.6	-0.1	NaN	
26708	26.0	...	-0.2	0.0	-0.1	-0.188	-9.9	3.8	-6.1	0.0	1995.0	
26709	1901.0	...	2.7	2.1	4.8	0.121	0.2	1.7	1.9	1.9	1995.0	
26710	1482.0	...	2.0	0.9	2.9	0.095	0.8	0.3	1.1	1.2	1995.0	
26711	949.0	...	1.6	0.8	2.4	0.122	0.4	0.7	1.1	0.8	1995.0	
26712	952.0	...	1.0	0.8	1.8	0.091	-1.1	1.2	0.0	0.5	1995.0	
26713	1393.0	...	1.0	1.9	3.0	0.103	-1.2	2.1	0.9	1.0	1995.0	
26714	897.0	...	0.6	1.2	1.8	0.096	-0.6	1.1	0.5	0.6	1995.0	
26715	1023.0	...	0.8	1.1	1.9	0.089	-1.4	0.6	-0.8	0.3	1995.0	
26716	836.0	...	0.9	0.5	1.4	0.081	-1.5	1.0	-0.5	0.3	1995.0	
26717	77.0	...	0.2	0.0	0.2	0.141	0.9	2.4	3.3	0.1	1995.0	
26718	332.0	...	0.1	0.3	0.3	0.047	-2.5	-0.9	-3.4	-0.1	2018.0	
26719	199.0	...	-0.2	0.2	0.1	0.013	-5.3	0.8	-4.5	-0.1	2018.0	
26720	220.0	...	0.1	0.2	0.3	0.076	-1.7	0.4	-1.3	0.0	2018.0	
26721	61.0	...	0.0	0.0	0.1	0.041	-4.5	-2.4	-6.9	-0.1	2018.0	
26722	30.0	...	0.1	0.0	0.1	0.220	1.2	-0.8	0.5	0.0	NaN	
26723	17.0	...	0.0	0.0	0.0	-0.001	-4.3	-2.6	-7.0	0.0	NaN	
26724	384.0	...	-0.5	0.5	-0.1	-0.009	-3.9	-0.8	-4.7	-0.3	2018.0	
26725	135.0	...	-0.2	0.1	-0.1	-0.022	-3.1	-2.1	-5.2	-0.1	2018.0	
26726	558.0	...	0.1	0.4	0.5	0.043	-2.3	-1.2	-3.5	-0.2	2018.0	
26727	15.0	...	0.1	0.0	0.1	0.280	9.9	-0.5	9.4	0.0	1999.0	
26728	11.0	...	0.0	0.0	0.0	-0.072	-2.5	-5.6	-8.0	0.0	1990.0	
26729	2.0	...	0.0	0.0	0.0	-0.818	-28.9	-19.1	-48.0	0.0	1990.0	

mp_at_retirement

26700	2402.0
26701	144.0
26702	263.0
26703	1227.0
26704	70.0
26705	56.0
26706	56.0
26707	NaN
26708	77.0
26709	77.0
26710	77.0
26711	77.0

```

26712      77.0
26713      77.0
26714      77.0
26715      77.0
26716      77.0
26717      77.0
26718      61.0
26719      61.0
26720      61.0
26721      61.0
26722        NaN
26723        NaN
26724    558.0
26725    558.0
26726    558.0
26727      15.0
26728    459.0
26729    459.0

```

[30 rows x 155 columns]

```

[586]: #Make a variable that holds career length(ex: 10 years)-make a method to
       ↪calculate this for each player(value_count?)-Amelia
careerLen = finalDF.groupby("player_id")["season"].nunique().reset_index()

careerLen.columns = ["player_id", "career_length"]
finalDF = finalDF.merge(careerLen, on="player_id", how="left")
finalDF[["player", "career_length"]][26703:26713]

```

```

[586]:          player  career_length
26703  Charlie Yelverton           1
26704      Rich Yonakor           1
26705      Gabe York            2
26706      Gabe York            2
26707  Chris Youngblood          1
26708      Danny Young           10
26709      Danny Young           10
26710      Danny Young           10
26711      Danny Young           10
26712      Danny Young           10

```

```

[587]: #Create a copy of the DF that removes all the active players so we can gather
       ↪more accurate predictive conclusions about retirement.
retiredDF = finalDF.copy()
retiredDF = retiredDF[retiredDF['retired'].notna()]
retiredDF = retiredDF.reset_index(drop=True)
retiredDF

```

```
[587]:      season    lg        player  player_id   age team  pos   g   gs \
0       1991  NBA  Alaa Abdelnaby  abdelal01  22.0  POR  PF  43  0.0
1       1992  NBA  Alaa Abdelnaby  abdelal01  23.0  POR  PF  71  1.0
2       1993  NBA  Alaa Abdelnaby  abdelal01  24.0  2TM  PF  75  52.0
3       1994  NBA  Alaa Abdelnaby  abdelal01  25.0  BOS  PF  13  0.0
4       1995  NBA  Alaa Abdelnaby  abdelal01  26.0  2TM  PF  54  0.0
...
24079    ...    ...    ...    ...    ...    ...    ...    ...
24079    2019  NBA     Ante Žižić  zizican01  22.0  CLE  C  59  25.0
24080    2020  NBA     Ante Žižić  zizican01  23.0  CLE  C  22  0.0
24081    1983  NBA      Jim Zoet  zoetji01  29.0  DET  C  7  0.0
24082    1971  NBA      Bill Zopf  zopfb01  22.0  MIL  PG  53  NaN
24083    1949  BAA     Matt Zunic  zunicma01  29.0  WSC  NaN  56  NaN

          mp   ...  dws   ws  ws_48  obpm  dbpm  bpm  vorp  retired \
0      290.0  ...  0.5  0.5  0.079  -3.4  -1.2  -4.6  -0.2  1995.0
1      934.0  ...  1.5  2.1  0.110  -2.3  -0.4  -2.6  -0.1  1995.0
2     1311.0  ...  1.3  2.0  0.074  -2.4  -1.5  -3.9  -0.6  1995.0
3      159.0  ...  0.1  -0.1 -0.032  -5.3  -2.2  -7.4  -0.2  1995.0
4      506.0  ...  0.7  0.3  0.027  -4.4  0.1  -4.3  -0.3  1995.0
...
24079  1082.0  ...  0.3  2.0  0.087  -1.1  -2.1  -3.2  -0.3  2020.0
24080  221.0  ...  0.2  0.5  0.106  -1.7  -1.5  -3.2  -0.1  2020.0
24081   30.0  ...  0.0  -0.1 -0.123  -5.6  0.2  -5.4  -0.1  1983.0
24082  398.0  ...  0.4  -0.1 -0.011  NaN  NaN  NaN  NaN  1971.0
24083    NaN  ...  1.8  2.0  NaN  NaN  NaN  NaN  NaN  1949.0

          mp_at_retirement  career_length
0                  506.0           5
1                  506.0           5
2                  506.0           5
3                  506.0           5
4                  506.0           5
...
24079                 ...           ...
24080                 ...           ...
24081                 ...           ...
24082                 ...           ...
24083                 ...           ...

[24084 rows x 156 columns]
```

```
[588]: avgData = retiredDF.copy()
avgData["gs_g_ratio"] = avgData["gs"] / avgData["g"].replace(0, np.nan)

numeric_cols = [
    'career_length', 'mp', 'gs_g_ratio', 'fg', 'fga', 'fg_percent',
    'x3p', 'x3pa', 'x3p_percent',
```

```

'x2p', 'x2pa', 'x2p_percent',
'e_fg_percent', 'ft', 'fta', 'ft_percent'
]
non_numeric_cols = ['player', 'pos']
numeric_means = avgData.groupby('player_id')[numeric_cols].mean().reset_index()
non_numeric_first = avgData.groupby('player_id')[non_numeric_cols].first().
    ↪reset_index()
retirement_age = avgData.groupby('player_id')['age'].max().
    ↪reset_index(name='retirementAge')
avgData = numeric_means.merge(non_numeric_first, on='player_id')
avgData = avgData.merge(retirement_age, on='player_id')

# add heights and weights
avgData = avgData.merge(HeightAndWeightDf[['player', 'height', 'weight']], ▾
    ↪on='player', how='left')
# add injuries
avgData = avgData.merge(injury_counts, on='player', how='left')

avgData

```

[588]:

	player_id	career_length	mp	gs_g_ratio	fg	\	
0	abdelal01	5.0	640.000000	0.141484	124.000000		
1	abdulka01	20.0	2872.300000	1.000000	791.850000		
2	abdulma01	10.0	1920.200000		NaN	359.700000	
3	abdulma02	9.0	1736.444444	0.510361	390.444444		
4	abdulta01	6.0	801.166667	0.517169	120.000000		
...		
4881	zipsepa01	2.0	833.500000	0.315657	84.500000		
4882	zizican01	3.0	505.666667	0.162076	91.000000		
4883	zoetji01	1.0	30.000000	0.000000	1.000000		
4884	zopfbi01	1.0	398.000000		NaN	49.000000	
4885	zunicma01	1.0		NaN	NaN	98.000000	
	fga	fg_percent	x3p	x3pa	x3p_percent	...	\
0	247.200000	0.486400	0.000000	0.600000	0.000000	...	
1	1415.350000	0.558350	0.100000	1.800000	0.033300	...	
2	816.200000	0.432000	NaN	NaN	NaN	...	
3	882.555556	0.439778	52.666667	148.777778	0.316444	...	
4	287.666667	0.414833	3.000000	12.666667	0.254500	...	
...	
4881	227.500000	0.372000	35.000000	104.500000	0.334500	...	
4882	156.666667	0.617667	0.000000	0.000000	NaN	...	
4883	5.000000	0.200000	0.000000	0.000000	NaN	...	
4884	135.000000	0.363000	NaN	NaN	NaN	...	
4885	323.000000	0.303000	NaN	NaN	NaN	...	
	e_fg_percent	ft	fta	ft_percent	player	\	

0	0.486400	45.000000	64.200000	0.658000	Alaa Abdelnaby
1	0.565700	335.600000	465.200000	0.727700	Kareem Abdul-Jabbar
2	NaN	189.300000	250.200000	0.749700	Walt Hazzard
3	0.466889	116.777778	129.000000	0.893111	Mahmoud Abdul-Rauf
4	0.421000	62.000000	88.166667	0.654833	Tariq Abdul-Wahad
...
4881	0.449000	25.000000	32.500000	0.767500	Paul Zipser
4882	0.617667	42.666667	60.000000	0.722000	Ante Žižić
4883	0.200000	0.000000	0.000000	NaN	Jim Zoet
4884	NaN	20.000000	36.000000	0.556000	Bill Zopf
4885	NaN	77.000000	109.000000	0.706000	Matt Zunic

	pos	retirementAge	height	weight	num_injuries
0	PF	26.0	208.0	108.0	NaN
1	C	41.0	NaN	NaN	NaN
2	SG	31.0	201.0	102.0	NaN
3	PG	31.0	188.0	83.0	NaN
4	SG	28.0	198.0	101.0	NaN
...
4881	SF	23.0	203.0	97.0	12.0
4882	C	23.0	NaN	NaN	NaN
4883	C	29.0	216.0	108.0	NaN
4884	PG	22.0	185.0	77.0	NaN
4885	None	29.0	NaN	NaN	NaN

[4886 rows x 23 columns]

```
[589]: predCols = ['fg_percent', 'x3p_percent', 'x2p_percent', 'e_fg_percent', 'ft_percent',
   ↪'mp', 'gs_g_ratio', 'fg', 'fga', 'x3p', 'x3pa', 'x2p', 'x2pa', 'ft', 'fta', 'height',
   ↪'weight', 'num_injuries']

X = avgData[predCols].fillna(avgData[predCols].mean())
y = avgData['career_length'].fillna(avgData['career_length'].mean())
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
   ↪random_state=42)
```

[590]: X_train

679	0.422500	0.346333	0.448500	0.473000	0.788333
4602	0.440200	0.224136	0.440029	0.440403	0.585600
1894	0.466000	0.176000	0.490000	0.473000	0.667000
3299	0.667000	0.224136	0.667000	0.667000	0.690108
1261	0.457750	0.370625	0.526375	0.538875	0.790000
...
4426	0.322600	0.224136	0.440029	0.440403	0.677600

466	0.444727	0.296000	0.459091	0.463545	0.870000		
3092	0.275000	0.224136	0.440029	0.440403	0.667000		
3772	0.230500	0.224136	0.440029	0.440403	0.726000		
860	0.380000	0.100000	0.404500	0.383500	0.893000		
	mp	gs_g_ratio	fg	fga	x3p	x3pa	\
679	750.833333	0.211742	121.666667	292.166667	26.500000	78.333333	
4602	1030.600000	0.000000	200.200000	461.400000	13.886148	41.210164	
1894	467.000000	0.216121	104.000000	223.000000	3.000000	17.000000	
3299	6.000000	0.000000	2.000000	3.000000	0.000000	0.000000	
1261	1002.750000	0.330943	113.375000	248.125000	39.625000	110.000000	
...	
4426	118.000000	0.216121	109.600000	347.800000	13.886148	41.210164	
466	1958.636364	0.703244	355.272727	792.727273	35.818182	100.909091	
3092	803.591551	0.216121	11.000000	40.000000	13.886148	41.210164	
3772	803.591551	0.216121	27.000000	92.500000	13.886148	41.210164	
860	79.000000	0.000000	19.000000	45.500000	0.500000	3.500000	
	x2p	x2pa	ft	fta	height	weight	\
679	95.166667	213.833333	61.500000	76.166667	196.000000	93.000000	
4602	107.157474	226.933226	99.800000	158.500000	211.000000	99.000000	
1894	101.000000	206.000000	20.000000	30.000000	198.393443	94.397055	
3299	2.000000	3.000000	0.000000	0.000000	201.000000	111.000000	
1261	73.750000	138.125000	62.250000	78.125000	198.393443	94.397055	
...	
4426	107.157474	226.933226	69.800000	101.600000	190.000000	88.000000	
466	319.454545	691.818182	162.181818	185.727273	180.000000	81.000000	
3092	107.157474	226.933226	10.000000	15.000000	198.393443	94.397055	
3772	107.157474	226.933226	13.750000	23.250000	178.000000	74.000000	
860	18.500000	42.000000	6.500000	8.000000	198.000000	86.000000	
	num_injuries						
679	15.046458						
4602	15.046458						
1894	15.046458						
3299	15.046458						
1261	15.046458						
...	...						
4426	15.046458						
466	15.046458						
3092	15.046458						
3772	15.046458						
860	15.046458						

[3908 rows x 18 columns]

[591]: X_test

```
[591]:      fg_percent  x3p_percent  x2p_percent  e_fg_percent  ft_percent  \
1149    0.437909    0.135714    0.438091    0.439545    0.694909
393     0.406000    0.139200    0.430900    0.430900    0.560750
1268    0.444000    0.224136    0.444000    0.444000    0.628250
4233    0.500000    0.000000    0.501500    0.500000    0.776000
4181    0.418100    0.284000    0.438100    0.439800    0.816500
...
3112    0.000000    0.224136    0.440029    0.440403    0.500000
84      0.467000    0.224136    0.440029    0.440403    0.851500
2086    0.370000    0.261000    0.407000    0.403000    0.858000
4126    0.250000    0.000000    0.500000    0.250000    0.500000
3614    0.404000    0.334000    0.445000    0.467250    0.798000

      mp  gs_g_ratio      fg      fga      x3p      x3pa  \
1149  1587.454545  0.697676  292.363636  624.545455  0.454545  2.181818
393   1169.800000  0.216285  66.500000  163.500000  4.300000  20.500000
1268   534.500000  0.305125  39.500000  85.250000  0.000000  0.000000
4233   409.000000  0.056818  35.000000  85.000000  0.000000  0.500000
4181  1789.000000  0.432840  290.400000  687.000000  27.600000  92.100000
...
3112   11.000000  0.216121  0.000000  4.000000  13.886148  41.210164
84     533.500000  0.216121  67.500000  143.000000  13.886148  41.210164
2086   2045.500000  0.216121  329.500000  879.500000  61.500000  223.500000
4126   12.000000  0.000000  1.000000  4.000000  0.000000  2.000000
3614   587.500000  0.038377  78.750000  190.000000  24.250000  71.250000

      x2p      x2pa      ft      fta      height      weight  \
1149  291.909091  622.363636  149.818182  203.545455  213.0    124.0
393    62.200000  143.000000  22.600000  40.300000  231.0     90.0
1268   39.500000  85.250000  26.250000  42.250000  211.0    115.0
4233   35.000000  84.500000  19.500000  28.000000  193.0     79.0
4181  262.800000  594.900000  210.100000  255.100000  196.0     95.0
...
3112  107.157474  226.933226  1.000000  2.000000  190.0     83.0
84    107.157474  226.933226  38.500000  45.500000  208.0    104.0
2086  268.000000  656.000000  374.500000  440.000000  193.0     90.0
4126   1.000000  2.000000  1.000000  2.000000  201.0    102.0
3614   54.500000  118.750000  31.000000  38.000000  185.0     88.0

      num_injuries
1149    15.046458
393     15.046458
1268    15.046458
4233    15.046458
4181    43.000000
...
3112    15.046458
```

```
84      15.046458  
2086    15.046458  
4126    15.046458  
3614    15.046458
```

[978 rows x 18 columns]

```
[592]: y_train
```

```
[592]: 679      6.0  
4602     10.0  
1894      1.0  
3299      1.0  
1261      8.0  
...  
4426      5.0  
466       11.0  
3092      1.0  
3772      4.0  
860       2.0  
Name: career_length, Length: 3908, dtype: float64
```

```
[593]: y_test
```

```
[593]: 1149     11.0  
393      10.0  
1268     4.0  
4233     2.0  
4181     10.0  
...  
3112     1.0  
84       2.0  
2086     2.0  
4126     1.0  
3614     4.0  
Name: career_length, Length: 978, dtype: float64
```

```
[594]: avgCareerLength = avgData['career_length'].mean()  
print(avgCareerLength)
```

```
avgRetirementAge = avgData['retirementAge'].mean()  
print(avgRetirementAge)
```

4.929390094146541
27.281930184804928

```
[595]: avgData.loc[avgData['player'] == 'LeBron James']
```

[595]: Empty DataFrame
Columns: [player_id, career_length, mp, gs_g_ratio, fg, fga, fg_percent, x3p, x3pa, x3p_percent, x2p, x2pa, x2p_percent, e_fg_percent, ft, fta, ft_percent, player, pos, retirementAge, height, weight, num_injuries]
Index: []
[0 rows x 23 columns]

13 Predictions

```
[596]: #Function to return predicted retirement values
def calculate_y_hat(X_train, X_test, Y_train, with_intercept=True):
    X = np.array(X_train)
    Y = np.array(Y_train)
    reg = LinearRegression(fit_intercept=with_intercept)
    reg.fit(X_train, Y_train)
    #Determining the yHat value given the fit and linear regression
    yHat = reg.predict(X_test)
    return yHat

# Define the columns to use for prediction
#predCols = ['fg_percent', 'x3p_percent', 'x2p_percent', 'e_fg_percent', 'ft_percent',
#            #'mp', 'gs_g_ratio', 'fg', 'fga', 'x3p', 'x3pa', 'x2p', 'x2pa', 'ft', 'fta',
#            #'height', 'weight', 'num_injuries']

# Fill missing values with column means
#X_train_features = trainData[predCols].fillna(trainData[predCols].mean())
#X_test_features = testData[predCols].fillna(testData[predCols].mean())
#y_train = trainData['career_length'].fillna(trainData['career_length'].mean())
#y_test = testData['career_length'].fillna(testData['career_length'].mean())

# Use the function to get predicted retirement ages
y_pred = calculate_y_hat(X_train, X_test, y_train, with_intercept=True)

# Round predictions for readability
#avgData['predictedCareerLength'] = avgData['predictedCareerLength'].round(0)

# Display results
print(y_test.head(20))
print(y_pred[:20])
```

1149	11.0
393	10.0
1268	4.0
4233	2.0
4181	10.0

```
2680    14.0
3949    16.0
4854     2.0
3757     1.0
3979     6.0
3418     4.0
2509     2.0
1371     1.0
3478     2.0
4748     7.0
2577    16.0
538      7.0
718      1.0
4502    16.0
4473     2.0
Name: career_length, dtype: float64
[ 9.48810784  6.83087212  4.90436618  3.10126108  9.79916924  10.05950646
 9.19021371  0.57583574  3.04748741  8.31209126  4.27512803  4.04146995
 0.59106491  1.79122387  4.79168957 14.49667684  8.37077751  2.63653874
 9.89709037  0.55857179]
```

```
[597]: df = pd.DataFrame()
df['career_length'] = y_test
df['predicted_career_length'] = y_pred
df.head(20)
```

```
[597]:   career_length  predicted_career_length
1149          11.0           9.488108
393           10.0           6.830872
1268           4.0           4.904366
4233           2.0           3.101261
4181          10.0           9.799169
2680          14.0          10.059506
3949          16.0           9.190214
4854           2.0           0.575836
3757           1.0           3.047487
3979           6.0           8.312091
3418           4.0           4.275128
2509           2.0           4.041470
1371           1.0           0.591065
3478           2.0           1.791224
4748           7.0           4.791690
2577          16.0          14.496677
538            7.0           8.370778
718            1.0           2.636539
4502          16.0           9.897090
4473           2.0           0.558572
```

```
[598]: #y = testData['career_length']
#y_hat = testData['predictedCareerLength']

mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
r2 = r2_score(y_test, y_pred)

print(f"RMSE: {rmse:.2f}")
print(f"R2: {r2:.4f}")
```

RMSE: 2.57

R2: 0.6699

```
[599]: print(avgData.loc[avgData['player'] == 'Trae Young'])
```

Empty DataFrame

Columns: [player_id, career_length, mp, gs_g_ratio, fg, fga, fg_percent, x3p, x3pa, x3p_percent, x2p, x2pa, x2p_percent, e_fg_percent, ft, fta, ft_percent, player, pos, retirementAge, height, weight, num_injuries]

Index: []

[0 rows x 23 columns]

```
[600]: #Plot of true retirement age vs predicted retirement age
```

```
plotData = avgData[avgData['career_length'] > 0]

x = plotData['career_length'].values.reshape(-1, 1)
y = plotData['predictedCareerLength'].values
reg = LinearRegression()
reg.fit(x, y)
y_fit = reg.predict(x)

plt.figure(figsize=(8,6))
plt.scatter(x, y, alpha=0.6, color='blue', label='Data points')
plt.plot(x, y_fit, color='green')
plt.xlabel("True career length")
plt.ylabel("Predicted career length")
plt.title("Predicted vs true career length")
plt.show()
```

<pre>KeyError</pre>	<pre>Traceback (most recent call last)</pre>
<pre>File ~/CS577/577venv1/lib/python3.12/site-packages/pandas/core/indexes/base.py:</pre>	<pre></pre>
<pre> ↳3812, in Index.get_loc(self, key)</pre>	<pre></pre>
<pre> 3811 try:</pre>	<pre></pre>
<pre> -> 3812 return self._engine.get_loc(casted_key)</pre>	<pre></pre>
<pre> 3813 except KeyError as err:</pre>	<pre></pre>

```

File pandas/_libs/index.pyx:167, in pandas._libs.index.IndexEngine.get_loc()
File pandas/_libs/index.pyx:196, in pandas._libs.index.IndexEngine.get_loc()
File pandas/_libs/hashtable_class_helper.pxi:7088, in pandas._libs.hashtable.
    ↪PyObjectHashTable.get_item()

File pandas/_libs/hashtable_class_helper.pxi:7096, in pandas._libs.hashtable.
    ↪PyObjectHashTable.get_item()

KeyError: 'predictedCareerLength'

```

The above exception was the direct cause of the following exception:

```

KeyError                                     Traceback (most recent call last)
Cell In[600], line 5
      2 plotData = avgData[avgData['career_length'] > 0]
      3 x = plotData['career_length'].values.reshape(-1, 1)
----> 5 y = plotData[                    ].values
      6 reg = LinearRegression()
      7 reg.fit(x, y)

File ~/CS577/577venv1/lib/python3.12/site-packages/pandas/core/frame.py:4113, in DataFrame.__getitem__(self, key)
    ↪DataFrame._getitem_(self, key)
  4111 if self.columns.nlevels > 1:
  4112     return self._getitem_multilevel(key)
-> 4113 indexer = self.columns.get_loc(key)
  4114 if is_integer(indexer):
  4115     indexer = [indexer]

File ~/CS577/577venv1/lib/python3.12/site-packages/pandas/core/indexes/base.py:
    ↪3819, in Index.get_loc(self, key)
  3814     if isinstance(casted_key, slice) or (
  3815         isinstance(casted_key, abc.Iterable)
  3816         and any(isinstance(x, slice) for x in casted_key)
  3817     ):
  3818         raise InvalidIndexError(key)
-> 3819     raise KeyError(key) from err
  3820 except TypeError:
  3821     # If we have a listlike key, _check_indexing_error will raise
  3822     # InvalidIndexError. Otherwise we fall through and re-raise
  3823     # the TypeError.
  3824     self._check_indexing_error(key)

KeyError: 'predictedCareerLength'

```

```
[ ]: finalDF.plot.scatter(x='mp', y='pts')

[ ]: finalDF = finalDF.dropna()
# Linear regression with mp and pts
MP = np.array(finalDF['mp']).reshape(-1, 1)
PTS = np.array(finalDF['pts']).reshape(-1, 1)
model = LinearRegression()
model.fit(MP, PTS)
prediction = model.predict(MP)

[ ]: plt.scatter(finalDF['mp'], finalDF['pts'])
line = prediction
plt.plot(finalDF['mp'], line, color='red')
plt.xlabel('mp')
plt.ylabel('pts')
plt.title('Simple Linear Regression, predicting mp and pts')
plt.show()

[ ]: #More plotting
finalDF["age"] = pd.to_numeric(finalDF["age"])
ageDF = finalDF[(finalDF["age"])> 0]
aveAge = ageDF.groupby("age")["mp_per_game"].mean()
youngest = aveAge.index.min()

plt.plot(aveAge.index, aveAge.values)
plt.title("average mins per game")
plt.xlabel("age")
plt.ylabel("mins per game")
plt.xlim(youngest)
plt.show()

[ ]: plt.plot(finalDF['career_length'])
```

14 AMELIA CODE

```
[ ]: # Install dependencies as needed:
# pip install kagglehub[pandas-datasets]
import kagglehub
from kagglehub import KaggleDatasetAdapter

# Set the path to the file you'd like to load
file_path = "Player Totals.csv"

# Load the latest version
totals_df = kagglehub.load_dataset(
    KaggleDatasetAdapter.PANDAS,
    "sumitrodatta/nba-aba-baa-stats",
```

```

    file_path,
    # Provide any additional arguments like
    # sql_query or pandas_kwargs. See the
    # documentation for more information:
    # https://github.com/Kaggle/kagglehub/blob/main/README.
    ↵md#kaggledatasetadapterpandas
)

print("First 5 records:", totals_df.head())

# Import kagglehub if not already imported
import kagglehub
from kagglehub import KaggleDatasetAdapter

# Path to the new file
file_path_shooting = "Player Shooting.csv"

# Load the Player Shooting.csv file
df_shooting = kagglehub.load_dataset(
    KaggleDatasetAdapter.PANDAS,
    "sumitrodatta/nba-aba-baa-stats",
    file_path_shooting,
    # You can also provide additional pandas kwargs like index_col,
    ↵parse_dates, etc.
)
print("First 5 records of Player Shooting:", df_shooting.head())


# Install dependencies as needed:
# pip install kagglehub[pandas-datasets]
import kagglehub
from kagglehub import KaggleDatasetAdapter

# Set the path to the file you'd like to load
file_path = "Player Season Info.csv"

# Load the latest version
season_info_df = kagglehub.load_dataset(
    KaggleDatasetAdapter.PANDAS,
    "sumitrodatta/nba-aba-baa-stats",
    file_path,
    # Provide any additional arguments like
    # sql_query or pandas_kwargs. See the
    # documentation for more information:
    # https://github.com/Kaggle/kagglehub/blob/main/README.
    ↵md#kaggledatasetadapterpandas

```

```

)

print("First 5 records:", season_info_df.head())


# Install dependencies as needed:
# pip install kagglehub[pandas-datasets]
import kagglehub
from kagglehub import KaggleDatasetAdapter


# Set the path to the file you'd like to load
file_path = "Player Award Shares.csv"


# Load the latest version
award_shares_df = kagglehub.load_dataset(
    KaggleDatasetAdapter.PANDAS,
    "sumitrodatta/nba-aba-baa-stats",
    file_path,
    # Provide any additional arguments like
    # sql_query or pandas_kwargs. See the
    # documentation for more information:
    # https://github.com/Kaggle/kagglehub/blob/main/README.
    ↵md#kaggledatasetadapterpandas
)
print("First 5 records:", award_shares_df.head())


# Install dependencies as needed:
# pip install kagglehub[pandas-datasets]
import kagglehub
from kagglehub import KaggleDatasetAdapter


# Set the path to the file you'd like to load
file_path = "Player Career Info.csv"


# Load the latest version
career_df = kagglehub.load_dataset(
    KaggleDatasetAdapter.PANDAS,
    "sumitrodatta/nba-aba-baa-stats",
    file_path,
    # Provide any additional arguments like
    # sql_query or pandas_kwargs. See the
    # documentation for more information:
    # https://github.com/Kaggle/kagglehub/blob/main/README.
    ↵md#kaggledatasetadapterpandas
)

```

```

print("First 5 records:", career_df.head())


# Install dependencies as needed:
# pip install kagglehub[pandas-datasets]
import kagglehub
from kagglehub import KaggleDatasetAdapter


# Set the path to the file you'd like to load
file_path = "Player Per Game.csv"


# Load the latest version
per_games_df = kagglehub.load_dataset(
    KaggleDatasetAdapter.PANDAS,
    "sumitrodatta/nba-aba-baa-stats",
    file_path,
    # Provide any additional arguments like
    # sql_query or pandas_kwargs. See the
    # documentation for more information:
    # https://github.com/Kaggle/kagglehub/blob/main/README.
    # ↵md#kaggledatasetadapterpandas
)
print("First 5 records:", per_games_df.head())

```

```

[ ]: mergedDF = totals_df
mergedDF = pd.merge(mergedDF,per_games_df,on=["player_id", "season"], ↵
    ↵how="outer",suffixes=("","_per_game"))
mergedDF = pd.merge(mergedDF,df_shooting,on=["player_id", "season"], ↵
    ↵how="outer",suffixes=("","_shooting"))
mergedDF = pd.merge(mergedDF,season_info_df,on=["player_id", "season"], ↵
    ↵how="outer",suffixes=("","_season_info"))
mergedDF = pd.merge(mergedDF,award_shares_df,on=["player_id", "season"], ↵
    ↵how="outer",suffixes=("","_award"))
finalDF = pd.merge(mergedDF,career_df,on="player_id", how="left",suffixes=( "", ↵
    ↵"_career"))
print(finalDF[223000:223030])

```