

In [22]:

```
#libraries

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
```

In [23]:

```
dpc_weu_avg_df = pd.read_csv("dpc_weu_2021_2022_tour_3_division_1.csv")
dpc_weu_avg_df.head()
```

Out[23]:

	Player	Position	Total Count	Wins	Losses	Winrate	As Radiant	As Dire	Kills	Deaths	...	Avg. KAL	G
0	Nisha	2	20	11	9	55.00%	10	10	9.30	2.80	...	7.07	
1	Puppey	5	20	11	9	55.00%	10	10	1.80	7.15	...	2.16	
2	Crystallis	1	20	11	9	55.00%	10	10	7.05	3.90	...	4.33	
3	Iceiceice	3	20	11	9	55.00%	10	10	5.25	5.70	...	2.98	
4	Tobi	3	19	11	8	57.89%	5	14	3.53	3.89	...	4.54	

5 rows × 22 columns

Terms:

Dire / Radiant: the two sides of Dota2 teams. Dire, usually represented in color red, owns the top right half of the map, and radiant, usually represented in color blue, owns the bottom left half of the map.

Position 1 - 5:

- 1 = safe-lane carry (the carry player that plays the top lane of dire, or bottom lane of radiant. takes the most resources of the team and deals the most damage in most cases)
- 2 = mid-lane carry (the carry player that plays the midlane of both dire and radiant. takes about the same resources of position 1, but starts involving in teamfights earlier than position 1)
- 3 = off-lane carry (the carry player that plays the bottom lane of dire, or top lane of radiant. takes the least resources of carry positions, usually do stun, tank damage from opponent, and suppress opponent position 1 from farming (getting resources))
- 4 = soft-support (supports position 3. takes a little resources occasionally. Do stun, teamfight initiation, lane helping etc)
- 5 = hard-support (supports position 1. some position 5 heroes can heal. takes the least resources in most cases)

GPM = Golds Per Minute

XPM = Experience Per Minute

Denies = Last hit on their own creeps

HD = Hero Damage

TD = Tower Damage
HH = Hero Heal
GS = Gold Sum

BASIC INFO

All data is gathered from <https://www.datdota.com/> (<https://www.datdota.com/>).

In [24]:

```
#basic info of the df
dpc_weu_avg_df.info
```

38	11	5.86	6.14	...	1.72	493	571	275	6	21
39	8	1.90	6.00	...	1.33	370	458	184	9	19
40	1	2.00	6.00	...	2.08	321	418	140	2	16
41	1	1.00	7.50	...	1.47	328	428	114	6	18
42	2	2.00	10.00	...	0.50	447	428	198	8	18

	HD	TD	HH	GS
0	39091	3843	323	25470
1	12655	582	4012	10392
2	26695	9438	3	23111
3	18919	2946	2440	19175
4	22615	4561	910	16371
5	17455	385	2235	12710
6	11409	251	1939	9079
7	31480	3356	447	25576
8	28023	5254	27	21787
9	27182	2318	375	19408
10	10963	905	5289	9834
11	11515	277	1717	10733

In [25]:

```
#df description
dpc_weu_avg_df.describe()
```

Out[25]:

	Position	Total Count	Wins	Losses	As Radiant	As Dire	Kills	Deaths
count	43.000000	43.000000	43.000000	43.000000	43.000000	43.000000	43.000000	43.000000
mean	3.000000	16.279070	8.139535	8.139535	8.139535	8.139535	4.633023	5.175581
std	1.380131	4.404141	4.683209	3.700588	3.839527	3.522590	2.489144	2.028285
min	1.000000	2.000000	0.000000	2.000000	0.000000	1.000000	1.000000	2.410000
25%	2.000000	16.000000	5.000000	5.000000	5.000000	6.000000	2.245000	3.460000
50%	3.000000	18.000000	10.000000	8.000000	9.000000	7.000000	4.370000	4.890000
75%	4.000000	18.000000	11.000000	11.000000	11.000000	11.000000	6.790000	6.555000
max	5.000000	20.000000	13.000000	14.000000	13.000000	14.000000	9.580000	10.000000

In [26]:

```
#check null values
dpc_weu_avg_df.isna().sum()
```

Out[26]:

```
Player      0
Position    0
Total Count  0
Wins        0
Losses      0
Winrate     0
As Radiant  0
As Dire     0
Kills       0
Deaths      0
Assists     0
KDA         0
Avg. KAL    0
GPM         0
XPM         0
Last Hits   0
Denies      0
LVL         0
HD          0
TD          0
HH          0
GS          0
dtype: int64
```

In [27]:

```
#HD, TD, HH, AND GS are vague abbreviations. Changing names of certain columns
dpc_weu_avg_df.rename(columns = {'HD': 'Hero Damage',
                                'TD': 'Tower Damage',
                                'HH': 'Hero Heal',
                                'GS': 'Gold Sum'}, inplace = True)
print(dpc_weu_avg_df.columns)
```

```
Index(['Player', 'Position', 'Total Count', 'Wins', 'Losses', 'Winrate',
       'As Radiant', 'As Dire', 'Kills', 'Deaths', 'Assists', 'KDA',
       'Avg. KAL', 'GPM', 'XPM', 'Last Hits', 'Denies', 'LVL', 'Hero Damage',
       'Tower Damage', 'Hero Heal', 'Gold Sum'],
      dtype='object')
```

In [28]:

```
#Unique Positions
dpc_weu_avg_df['Position'].unique()
```

Out[28]:

```
array([2, 5, 1, 3, 4], dtype=int64)
```

In [30]:

```
#position 1 to 5 may be vague and unclear to non-dota players. Here I renamed each rows:
#position 1: safelane carry; position 2: midlane carry; position 3: offlane carry; position 4: soft-
dpc_weu_avg_df['Position'].replace({5: 'hard-support(safelane_support)', 4: 'soft-support(offlane-su
dpc_weu_avg_df.head()
```

Out[30]:

	Player	Position	Total Count	Wins	Losses	Winrate	As Radiant	As Dire	Kills	Deaths
0	Nisha	midlane carry	20	11	9	55.00%	10	10	9.30	2.8
1	Puppey	hard- support(safelane_support)	20	11	9	55.00%	10	10	1.80	7.1
2	Crystallis	safelane carry	20	11	9	55.00%	10	10	7.05	3.9
3	Iceiceice	offlane carry	20	11	9	55.00%	10	10	5.25	5.1
4	Tobi	offlane carry	19	11	8	57.89%	5	14	3.53	3.8

5 rows × 22 columns

In [31]:

```
#players count in each position
count_df = dpc_weu_avg_df.groupby(by = 'Position').count()
count_df.drop(count_df.columns[1:], axis = 1, inplace = True)
count_df
```

Out[31]:

	Player
Position	
hard-support(safelane_support)	8
midlane carry	8
offlane carry	11
safelane carry	8
soft-support(offlane-support)	8

General Analysis of All DPC(WEU) Players in the 2021/2022 Tour 3 Division I

In [33]:

```

KDA(kill/death/assist) index of players in each position
re(figsize = (12, 6))
plot(x = 'Position', y = 'KDA', data = dpc_weu_avg_df, order = ['hard-support(safelane_support)',
                                                                'soft-support(offlane-support)',
                                                                'offlane carry',
                                                                'midlane carry',
                                                                'safelane carry']).set(title = 'Average
ks(rotation = 90)

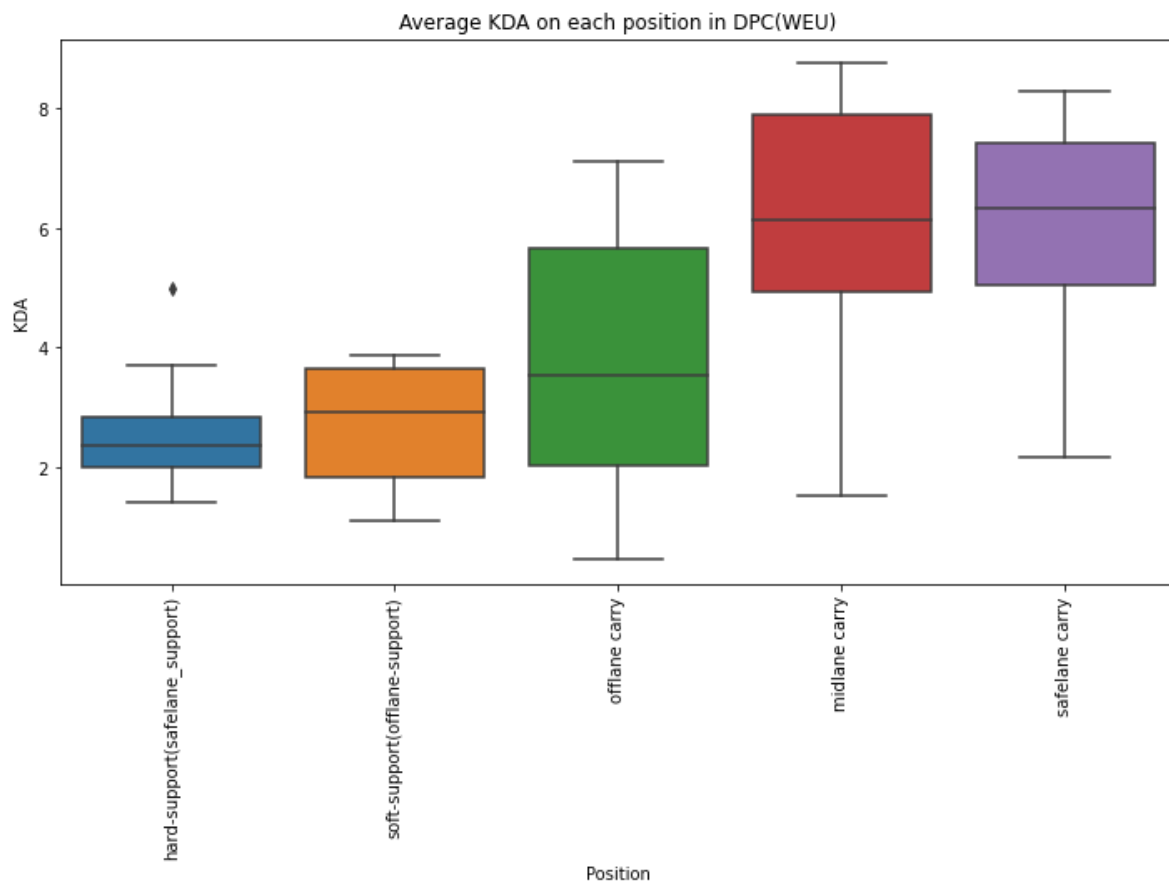
```

Out[33]:

```

(array([0, 1, 2, 3, 4]),
 [Text(0, 0, 'hard-support(safelane_support)'),
  Text(1, 0, 'soft-support(offlane-support)'),
  Text(2, 0, 'offlane carry'),
  Text(3, 0, 'midlane carry'),
  Text(4, 0, 'safelane carry')])

```



In DPC West Europe2021/2022 Tour 3 Division I, midlane carry players (position 2) have the highest KDA (Kill/Death/Assist) ratio. The KDA ranges the most in offlane carry players, ranging from <2 to >8.

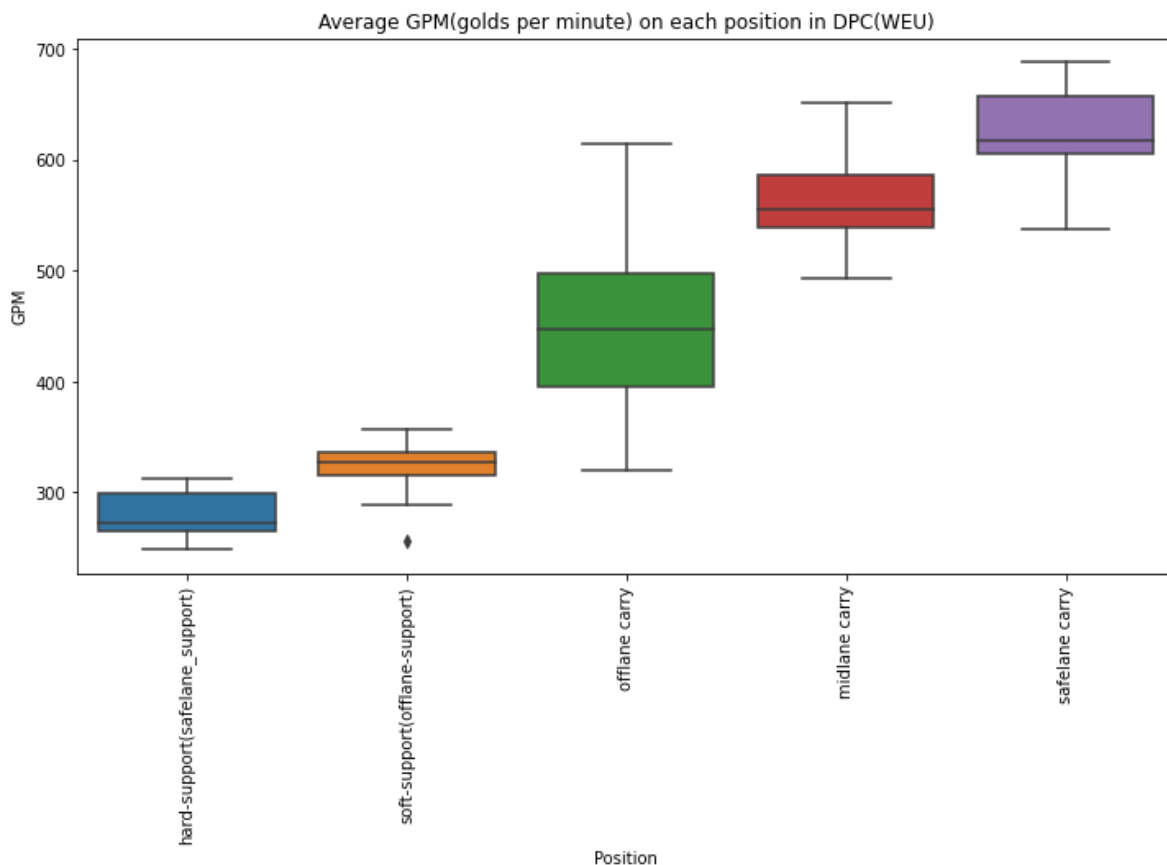
In [38]:

```
#average GPM index of players in each position
plt.figure(figsize = (12, 6))
sns.boxplot(x = 'Position', y = 'GPM', data = dpc_weu_avg_df, order = ['hard-support(safelane_support)',
                                                                    'soft-support(offlane-support)',
                                                                    'offlane carry',
                                                                    'midlane carry',
                                                                    'safelane carry']).set(title = 'Average GPM(golds per minute) on each position in DPC(WEU)')

plt.xticks(rotation = 90)
```

Out[38]:

```
(array([0, 1, 2, 3, 4]),
 [Text(0, 0, 'hard-support(safelane_support)'),
  Text(1, 0, 'soft-support(offlane-support)'),
  Text(2, 0, 'offlane carry'),
  Text(3, 0, 'midlane carry'),
  Text(4, 0, 'safelane carry')])
```



As for GPM (gold per minute, the index that reflects the how much resources on the map a player takes), safelane carry players take the most resources, and offlane carry players take the least among the carry positions. Position 4 and position 5 (supports) take resources much less, with a highest average GPM player at about 360 and 310.

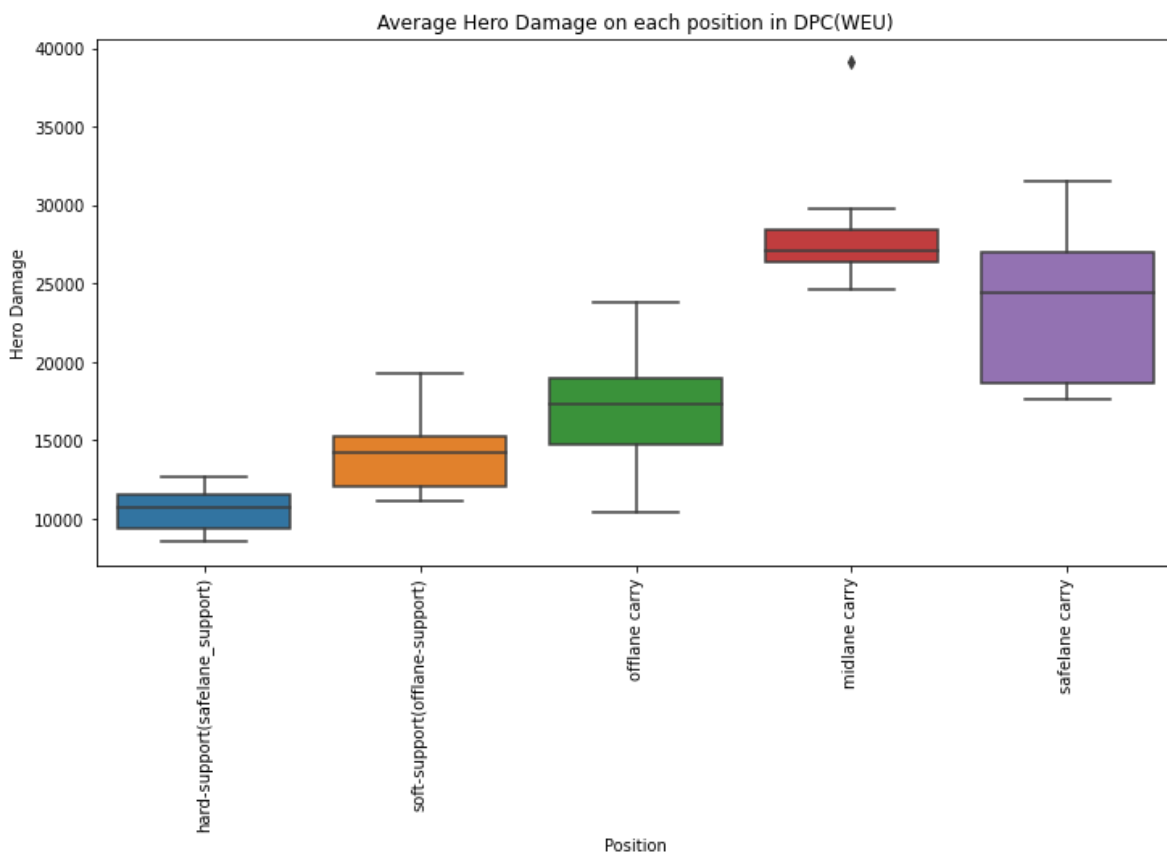
In [89]:

```
#average Hero Damage index of players in each position
plt.figure(figsize = (12, 6))
sns.boxplot(x = 'Position', y = 'Hero Damage', data = dpc_weu_avg_df, order = ['hard-support(safelane support)',
'soft-support(offlane-support)',
'offlane carry',
'midlane carry',
'safelane carry']).set(title = 'Average Hero Damage on each position in DPC(WEU)')

plt.xticks(rotation = 90)
```

Out[89]:

```
(array([0, 1, 2, 3, 4]),
 [Text(0, 0, 'hard-support(safelane support)'),
  Text(1, 0, 'soft-support(offlane-support)'),
  Text(2, 0, 'offlane carry'),
  Text(3, 0, 'midlane carry'),
  Text(4, 0, 'safelane carry')])
```



In DPC(WEU) 2021/2022 Tour 3 Division I, midlane carry players generally deal the most damage. The highest average damage is done by a nisha, a mid-lane carry player, who has an extreme value of almost 40k. Safelane carry players' damage vary hugely ranging from 18k to almost 33k. As for offlane carry players, much less damage is done. This can be possibly be explained with the hero picks in west EU teams. Compared to

previously analyzed DPC(CN) tour, west European teams tend to pick mid and safe-lane players with high damage heroes, instead of those heroes with more heavy focus on teamfight tactics and pushing abilities in China.

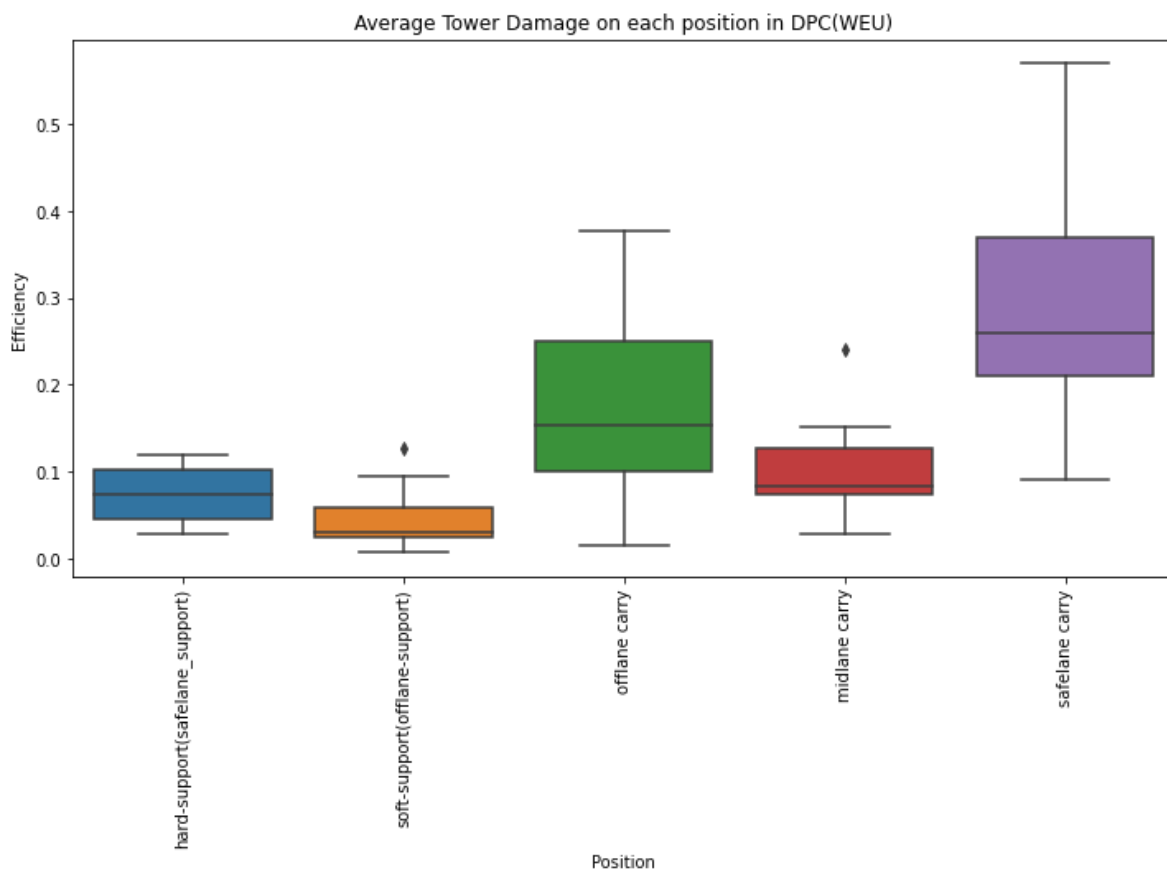
In [90]:

```
g the average gold sum to the tower damage, which states how much gold it takes for a player to deal
'] / dpc_weu_avg_df['Gold Sum']).astype(float)

_avg_df, order = ['hard-support(safelane_support)',
                  'soft-support(offlane-support)',
                  'offlane carry',
                  'midlane carry',
                  'safelane carry']).set(title = 'Average Tower Damage on each position in DPC(WEU)')
```

Out[90]:

```
(array([0, 1, 2, 3, 4]),
 [Text(0, 0, 'hard-support(safelane_support)'),
  Text(1, 0, 'soft-support(offlane-support)'),
  Text(2, 0, 'offlane carry'),
  Text(3, 0, 'midlane carry'),
  Text(4, 0, 'safelane carry')])
```



It can be seen from this chart that most tower damage and pushing is done by position 1, instead of both position 1 and 3 in Chinese teams. It appears that in west European teams, offlane carry players are more dedicated in teamfight, and safelane carry players have hero picks that are both strong in teamfights and pushing. As compared to Chinese region, where pushing ability is done by both offlane and safelane carries, WEU tend to rely mostly on safelane carries. Both region have low midlane carry tower damage efficiency.

In [40]:

```
#Average minutes per game of each player (by dividing GPM from gold sum).
#This would reflect the efficiency of each player and the optimum length of game for them to perform

dpc_weu_avg_df['Average Game Length (min)'] = dpc_weu_avg_df['Gold Sum'] / dpc_weu_avg_df['GPM']
```

In [41]:

```
#Players that requires the largest game length:

max_len = dpc_weu_avg_df[['Player', 'Average Game Length (min)', 'Position', 'KDA']].nlargest(n = 5,
print(max_len)
```

	Player	Average Game Length (min)	Position	KDA
0	Nisha	39.124424	midlane carry	8.76
3	Iceiceice	38.273453	offlane carry	3.54
21	Boxi	37.628125	offlane carry	4.05
16	Zayac	37.176471	soft-support (offlane-support)	2.95
7	Pure	37.174419	safelane carry	8.29

In [42]:

```
#Players that requires the least game length:

min_len = dpc_weu_avg_df[['Player', 'Average Game Length (min)', 'Position', 'KDA']].nsmallest(n = 5,
print(min_len)
```

	Player	Average Game Length (min)	Position	KDA
37	Aramis	29.632812	soft-support (offlane-support)	1.11
29	Saksa	29.906061	soft-support (offlane-support)	2.90
30	Taiga	30.101852	soft-support (offlane-support)	3.57
32	ATF	30.695740	offlane carry	4.91
31	Yuragi	30.700658	safelane carry	7.06

Player **Nisha**, **Zayac**, and **Iceiceice**, all from Team Secret, have the longest average game length, as well as **Boxi** from Team Liquid and **Pure** from Entity. This not only reflects the fact that these three teams plays a longer game than other teams, but also indicates that these players require longer game length and more farming time than others.

On the other hand, **Taiga**, **ATF**, and **Yuragi** from OG all have the least game length on average. It partially reflects that PSG.LGD plays a faster pace game than other teams.

Best Players / Stand-out Players

In [91]:

```

age kills of players in Dota Pro Circuit West Europe 2021-2022 Summer Division 1 is 4.333902, with a
kill + std_kill

distinct value for selecting distinct players, we list the relative important data of these players
['Kills'].values >= distinct_kill
c_weu_avg_df.loc[mask]
(distinct_players.columns[2:5], axis = 1, inplace = True)
s)

```

	Player	Position	Winrate	As Radiant	As Dire	Kills	Deaths	\
0	Nisha	midlane carry	55.00%	10	10	9.30	2.80	
2	Crystallis	safelane carry	55.00%	10	10	7.05	3.90	
7	Pure	safelane carry	57.89%	5	14	9.58	3.42	
8	Stormstormer	midlane carry	57.89%	5	14	8.37	3.00	
9	MiCKe	midlane carry	72.22%	12	6	7.33	3.17	
15	DyrachYO	safelane carry	38.89%	11	7	6.56	4.89	
19	MATUMBAMAN	safelane carry	72.22%	12	6	9.06	2.50	
20	Zai	offlane carry	72.22%	12	6	6.83	3.28	
22	Limp	midlane carry	27.78%	7	11	6.44	3.56	
23	BOOM	midlane carry	38.89%	11	7	7.28	3.56	
27	Nine	midlane carry	76.47%	13	4	6.76	3.00	
28	skiter	safelane carry	76.47%	13	4	6.82	2.41	
31	Yuragi	safelane carry	62.50%	9	7	7.44	2.69	
32	ATF	offlane carry	62.50%	9	7	6.69	4.38	
34	bzm	midlane carry	62.50%	9	7	7.38	3.31	

	Assists	KDA	Avg. KAL	...	XPM	Last Hits	Denies	LVL	Hero Damage	\
0	10.50	8.76	7.07	...	759	405	10	26	39091	
2	9.85	5.45	4.33	...	688	390	10	25	26695	
7	10.21	8.29	5.78	...	696	396	9	25	31480	
8	12.58	8.55	6.98	...	736	373	10	25	28023	
9	12.06	6.75	6.12	...	679	303	11	23	27182	
15	9.11	5.62	3.20	...	718	366	13	23	24453	
19	9.33	8.06	7.36	...	717	362	12	24	27982	
20	13.72	6.42	6.27	...	657	281	9	23	23821	
22	8.67	4.94	4.25	...	605	290	6	22	29786	
23	8.39	5.54	4.41	...	678	292	9	23	24587	
27	11.06	7.70	5.94	...	681	271	8	22	27013	
28	8.24	7.23	6.24	...	668	330	19	22	18914	
31	8.44	7.06	5.91	...	688	281	10	22	17907	
32	9.81	4.91	3.77	...	641	222	10	21	18101	
34	9.38	4.93	5.06	...	696	292	8	22	26958	

	Tower Damage	Hero Heal	Gold Sum	Efficiency	Average Game Length (min)
0	3843	323	25470	0.150883	39.124424
2	9438	3	23111	0.408377	36.918530
7	3356	447	25576	0.131217	37.174419
8	5254	27	21787	0.241153	35.541599
9	2318	375	19408	0.119435	34.906475
15	5292	1764	22362	0.236651	32.933726
19	13205	181	23176	0.569770	35.655385
20	2054	2693	19526	0.105193	35.959484
22	1219	36	18677	0.065267	36.266019
23	1364	52	17970	0.075904	32.851920
27	1404	1196	17353	0.080908	31.323105
28	5108	196	18966	0.269324	31.297030

31	6637	705	18666	0.355566	30.700658
32	1568	0	15133	0.103615	30.695740
34	1621	886	19214	0.084366	33.242215

[15 rows x 21 columns]

C:\Users\Dominic\AppData\Local\Temp\ipykernel_27988\4205091641.py:9: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

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

distinct_players.drop(distinct_players.columns[2:5], axis = 1, inplace = True)

In [51]:

#top 10 players with respect to KDA index:

```
top_5_kda = dpc_weu_avg_df[['Player', 'Position', 'KDA', 'Winrate']].nlargest(n = 5, columns = 'KDA')
top_5_kda
```

Out[51]:

	Player	Position	KDA	Winrate
0	Nisha	midlane carry	8.76	55.00%
8	Stormstormer	midlane carry	8.55	57.89%
7	Pure	safelane carry	8.29	57.89%
19	MATUMBAMAN	safelane carry	8.06	72.22%
27	Nine	midlane carry	7.70	76.47%

The player that performs the best (based on KDA data) is **Nisha**, midlane carry player from Team Secret, with a winrate of 55.00%. The second best midlane carry player in DPC West EU is **Stormstormer**, from Entity, with a winrate of 57.89.

As for safelane carry players, **Pure** from Entity and **MATUMBAMAN** from Team Liquid stands out from all players with KDA of 8.29 and 8.06. Compared to DPC in China, WEU players with the best performances are mostly midlane or safelane carries. It can be reflected that position 3 (offlane carries) are less focused and are less performing well among all players compared to DPC China, where multiple position 3 players have a high KDA.

In [53]:

```
top_5_kda['Winrate'] = top_5_kda['Winrate'].str.rstrip('%').astype('float') / 100.0
```

In [54]:

```
#top 10 players' winrate:
top_5_kda_wr = top_5_kda[['Player', 'Position', 'KDA', 'Winrate']].nlargest(n = 5, columns = 'Winrate')
top_5_kda_wr
```

Out[54]:

	Player	Position	KDA	Winrate
27	Nine	midlane carry	7.70	0.7647
19	MATUMBAMAN	safelane carry	8.06	0.7222
8	Stormstormer	midlane carry	8.55	0.5789
7	Pure	safelane carry	8.29	0.5789
0	Nisha	midlane carry	8.76	0.5500

This part is only sorting top 5 KDA players with their winrate. We can see that even though Nisha has such a extraordinary performance with an extreme value of average damage and highest KDA of 8.76, his team's winrate is only 0.55. It can draw two possible conclusions, without further data: either Nisha is not a great team player that his performance does not contribute to the winning of the team, or his teammate are not performing good enough to match his extraordinary performance in mid-lane.

In [92]:

```
top_5_rd = dpc_weu_avg_df.nlargest(n = 5, columns = 'KDA')
top_5_rd.drop(['Player', 'Position', 'Total Count', 'Wins', 'Losses', 'Winrate', 'As Radiant', 'As Dire'])
top_5_rd.reset_index()
top_5_rd
```

Out[92]:

	Player	Position	Total Count	Wins	Losses	Winrate	As Radiant	As Dire	Kills	Deaths	...
0	Nisha	midlane carry	20	11	9	55.00%	10	10	9.30	2.80	...
8	Stormstormer	midlane carry	19	11	8	57.89%	5	14	8.37	3.00	...
7	Pure	safelane carry	19	11	8	57.89%	5	14	9.58	3.42	...
19	MATUMBAMAN	safelane carry	18	13	5	72.22%	12	6	9.06	2.50	...
27	Nine	midlane carry	17	13	4	76.47%	13	4	6.76	3.00	...

5 rows × 24 columns

In [93]:

```
top_5_rd['Laning Ability'] = (top_5_rd['Last Hits'] / top_5_rd['Denies']).astype(float)
top_5_rd_data = top_5_rd.drop(['Player', 'Position', 'Total Count', 'Wins', 'Losses', 'Winrate', 'As
top_5_rd_data
```

Out[93]:

	KDA	GPM	XPM	Hero Damage	Tower Damage	Gold Sum	Efficiency	Average Game Length (min)	Laning Ability
0	8.76	651	759	39091	3843	25470	0.150883	39.124424	40.500000
8	8.55	613	736	28023	5254	21787	0.241153	35.541599	37.300000
7	8.29	688	696	31480	3356	25576	0.131217	37.174419	44.000000
19	8.06	650	717	27982	13205	23176	0.569770	35.655385	30.166667
27	7.70	554	681	27013	1404	17353	0.080908	31.323105	33.875000

In [94]:

```
top_5_rd_name = top_5_rd.iloc[:, [0]].copy()
top_5_rd_name
```

Out[94]:

	Player
0	Nisha
8	Stormstormer
7	Pure
19	MATUMBAMAN
27	Nine

In [95]:

```
#Normalilzing data

top_5_rd_data_max_scaled = top_5_rd_data.copy()

for columns in top_5_rd_data_max_scaled.columns:
    top_5_rd_data_max_scaled[columns] = top_5_rd_data_max_scaled[columns] / top_5_rd_data_max_scaled[columns].max()

top_5_rd_data_max_scaled = top_5_rd_data_max_scaled.multiply(10)
top_5_rd_data_max_scaled.iloc[0]
```

Out[95]:

```
KDA                10.000000
GPM                9.462209
XPM                10.000000
Hero Damage        10.000000
Tower Damage       2.910261
Gold Sum           9.958555
Efficiency          2.648144
Average Game Length (min) 10.000000
Laning Ability     9.204545
Name: 0, dtype: float64
```

In [96]:

```
nisha = top_5_rd_data_max_scaled.iloc[0].tolist()
nisha = np.concatenate((nisha, [nisha[0]]))

stormstormer = top_5_rd_data_max_scaled.iloc[1].tolist()
stormstormer = np.concatenate((stormstormer, [stormstormer[0]]))

pure = top_5_rd_data_max_scaled.iloc[2].tolist()
pure = np.concatenate((pure, [pure[0]]))

matumbaman = top_5_rd_data_max_scaled.iloc[3].tolist()
matumbaman = np.concatenate((matumbaman, [matumbaman[0]]))

nine = top_5_rd_data_max_scaled.iloc[4].tolist()
nine = np.concatenate((nine, [nine[0]]))

nisha
```

Out[96]:

```
array([10.          ,  9.4622093 , 10.          , 10.          ,  2.91026126,
        9.9585549 ,  2.6481435 , 10.          ,  9.20454545, 10.          ])
```

In [97]:

```
categories = top_5_rd_data_max_scaled.columns.tolist()
categories = np.concatenate((categories, [categories[0]]))
len(categories)
```

Out[97]:

10

In [98]:

```
label_placement = np.linspace(start = 0, stop = 2*np.pi, num = len(nisha))
```

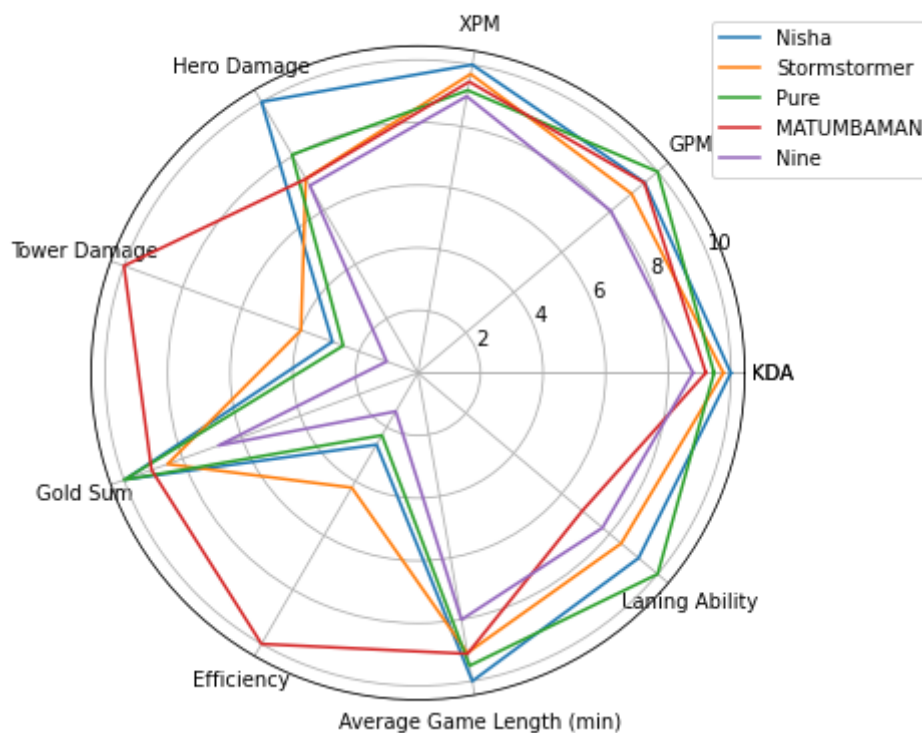
In [99]:

```
plt.figure(figsize = (6, 6))
plt.subplot(polar = True)
plt.plot(label_placement, nisha)
plt.plot(label_placement, stormstormer)
plt.plot(label_placement, pure)
plt.plot(label_placement, matumbaman)
plt.plot(label_placement, nine)
lines, labels = plt.thetagrids(np.degrees(label_placement), labels = categories)
plt.title('Best Players Data Comparison', y = 1.1, fontdict = {'fontsize': 18})
plt.legend(labels = top_5_rd_name['Player'].tolist(), loc = (0.95, 0.8))
```

Out[99]:

<matplotlib.legend.Legend at 0x1ae46fae6d0>

Best Players Data Comparison



In [100]:

```
#isolate 2 best midlane player comparison in the top 5 KDA chart

best_mid = dpc_weu_avg_df[(dpc_cn_avg_df['Player'] == 'Nisha') | (dpc_weu_avg_df['Player'] == 'Nine')
best_mid
```

Out[100]:

	index	Player	Position	Total Count	Wins	Losses	Winrate	As Radiant	As Dire	Kills	...	XPM	Last Hits
0	0	Nisha	midlane carry	20	11	9	55.00%	10	10	9.30	...	759	405
1	27	Nine	midlane carry	17	13	4	76.47%	13	4	6.76	...	681	271

2 rows × 25 columns

In [101]:

```
best_mid['Laning Ability'] = (best_mid['Last Hits'] / best_mid['Denies']).astype(float)
best_mid_data = best_mid.drop(['index', 'Player', 'Position', 'Total Count', 'Wins', 'Losses', 'Winrate'])
best_mid_data
```

Out[101]:

	KDA	GPM	XPM	Hero Damage	Tower Damage	Gold Sum	Efficiency	Average Game Length (min)	Laning Ability
0	8.76	651	759	39091	3843	25470	0.150883	39.124424	40.500
1	7.70	554	681	27013	1404	17353	0.080908	31.323105	33.875

In [102]:

```
best_mid_name = best_mid.iloc[:, [1]].copy()
best_mid_name
```

Out[102]:

	Player
0	Nisha
1	Nine

In [103]:

```
best_mid_data
```

Out[103]:

	KDA	GPM	XPM	Hero Damage	Tower Damage	Gold Sum	Efficiency	Average Game Length (min)	Laning Ability
0	8.76	651	759	39091	3843	25470	0.150883	39.124424	40.500
1	7.70	554	681	27013	1404	17353	0.080908	31.323105	33.875

In [104]:

```
from sklearn.preprocessing import MinMaxScaler
```

In [105]:

```
test = best_mid_data.copy().astype(float)
test
```

Out[105]:

	KDA	GPM	XPM	Hero Damage	Tower Damage	Gold Sum	Efficiency	Average Game Length (min)	Laning Ability
0	8.76	651.0	759.0	39091.0	3843.0	25470.0	0.150883	39.124424	40.500
1	7.70	554.0	681.0	27013.0	1404.0	17353.0	0.080908	31.323105	33.875

In [106]:

```
#Normalilzing data

best_mid_data_max_scaled = best_mid_data.copy()

for columns in best_mid_data_max_scaled.columns:
    best_mid_data_max_scaled[columns] = best_mid_data_max_scaled[columns] / best_mid_data_max_scaled.max()

best_mid_data_max_scaled = best_mid_data_max_scaled.multiply(10)
best_mid_data_max_scaled
```

Out[106]:

	KDA	GPM	XPM	Hero Damage	Tower Damage	Gold Sum	Efficiency	Average Game Length (min)	Laning Ability
0	10.000000	10.000000	10.000000	10.000000	10.000000	10.000000	10.0000	10.000000	10.000000
1	8.789954	8.509985	8.972332	6.910286	3.653396	6.813113	5.3623	8.006023	8.375000

In [107]:

```
nisha2 = best_mid_data_max_scaled.loc[0].tolist()
nisha2 = np.concatenate((nisha2, [nisha2[0]]))

nine2 = best_mid_data_max_scaled.loc[1].tolist()
nine2 = np.concatenate((nine2, [nine2[0]]))

nisha2
```

Out[107]:

```
array([10., 10., 10., 10., 10., 10., 10., 10., 10., 10.])
```

In [108]:

```
categories = best_mid_data_max_scaled.columns.tolist()
categories = np.concatenate((categories, [categories[0]]))
len(categories)
```

Out[108]:

```
10
```

In [109]:

```
label_placement = np.linspace(start = 0, stop = 2*np.pi, num = len(nts))
```

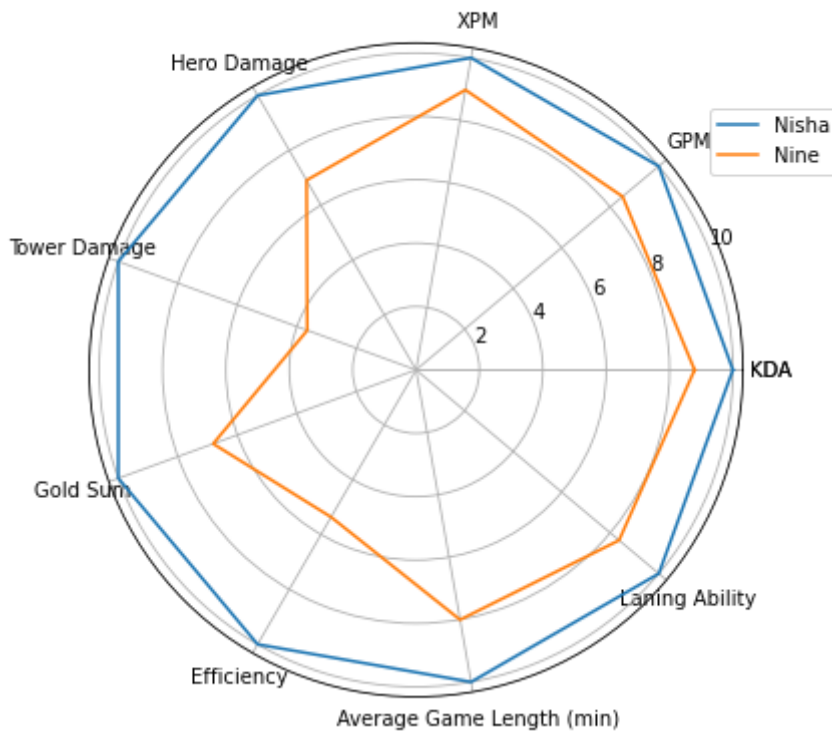
In [110]:

```
plt.figure(figsize = (6, 6))
plt.subplot(polar = True)
plt.plot(label_placement, nisha2)
plt.plot(label_placement, nine2)
lines, labels = plt.thetagrids(np.degrees(label_placement), labels = categories)
plt.title('Best Mid Players Data Comparison', y = 1.1, fontdict = {'fontsize': 18})
plt.legend(labels = best_mid_name['Player'].tolist(), loc = (0.95, 0.8))
```

Out[110]:

```
<matplotlib.legend.Legend at 0x1ae480013d0>
```

Best Mid Players Data Comparison



In [111]:

```
#Player efficiency based on GPM with KDA

a = plt.figure()
a.set_figwidth(20)
a.set_figheight(18)

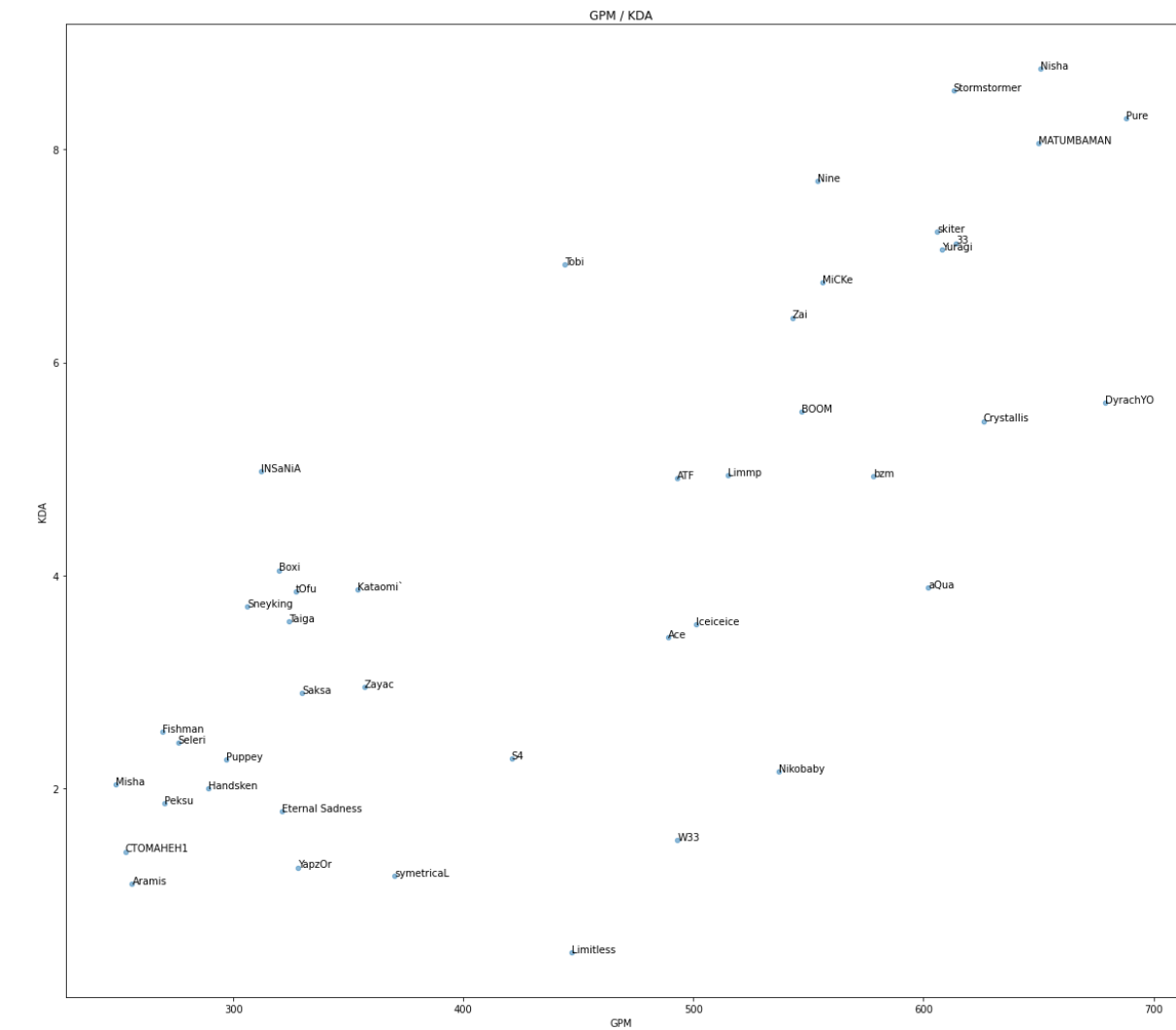
plt.scatter(x = dpc_cn_avg_df['GPM'],
            y = dpc_cn_avg_df['KDA'],
            s = 20,
            alpha = 0.5)

plt.title('GPM / KDA')
plt.xlabel('GPM')
plt.ylabel('KDA')

x, y = dpc_weu_avg_df['GPM'], dpc_weu_avg_df['KDA']
for i, txt in enumerate(dpc_weu_avg_df['Player']):
    plt.annotate(txt, (x[i], y[i]))
    print(i, txt, x[i], y[i], dpc_weu_avg_df['Player'][i])
```

```
0 Nisha 651 8.76 Nisha
1 Puppey 297 2.27 Puppey
2 Crystallis 626 5.45 Crystallis
3 Iceiceice 501 3.54 Iceiceice
4 Tobi 444 6.92 Tobi
5 Kataomi` 354 3.87 Kataomi`
6 Fishman 269 2.53 Fishman
7 Pure 688 8.29 Pure
8 Stormstormer 613 8.55 Stormstormer
9 MiCKe 556 6.75 MiCKe
10 Peksu 270 1.86 Peksu
11 tOfu 327 3.85 tOfu
12 Ace 489 3.42 Ace
13 Handsken 289 2.0 Handsken
14 S4 421 2.28 S4
15 DyrachY0 679 5.62 DyrachY0
16 Zayac 357 2.95 Zayac
17 aQua 602 3.89 aQua
18 INSaNiA 312 4.98 INSaNiA
19 MATUMBAMAN 650 8.06 MATUMBAMAN
20 Zai 543 6.42 Zai
21 Boxi 320 4.05 Boxi
22 Limpin 515 4.94 Limpin
23 BOOM 547 5.54 BOOM
24 Seleri 276 2.43 Seleri
25 33 614 7.12 33
26 Sneyking 306 3.71 Sneyking
27 Nine 554 7.7 Nine
28 skiter 606 7.23 skiter
29 Saksa 330 2.9 Saksa
30 Taiga 324 3.57 Taiga
31 Yuragi 608 7.06 Yuragi
32 ATF 493 4.91 ATF
33 Misha 249 2.04 Misha
34 bzm 578 4.93 bzm
35 CTOMAHEH1 253 1.41 CTOMAHEH1
36 Nikobaby 537 2.16 Nikobaby
37 Aramis 256 1.11 Aramis
38 W33 493 1.52 W33
```

39 symetrical 370 1.18 symetrical
40 Eternal Sadness 321 1.79 Eternal Sadness
41 YapzOr 328 1.26 YapzOr
42 Limitless 447 0.46 Limitless



This chart shows the amount of resource a player takes and the amount of impact (here, only indicated through KDA) they created in the game.

It can be seen that **Nisha** from Team Secret takes about 640 GPM and has the highest KDA of all players. As for the player that has both a high GPM and a high KDA, **Pure** from Entity, **MATUMBAMAN** from Team Liquid, and **Stormstormer** from Entity all stand on the top left corner.

Players that has a high GPM but is only mediocre in KDA includes **DyrachYO** from Gaimin Gladiator, **Crystallis** from Team Secret, **aQua** from goonsquad, **bzm** from OG, and **Nikobaby** from Alliance.

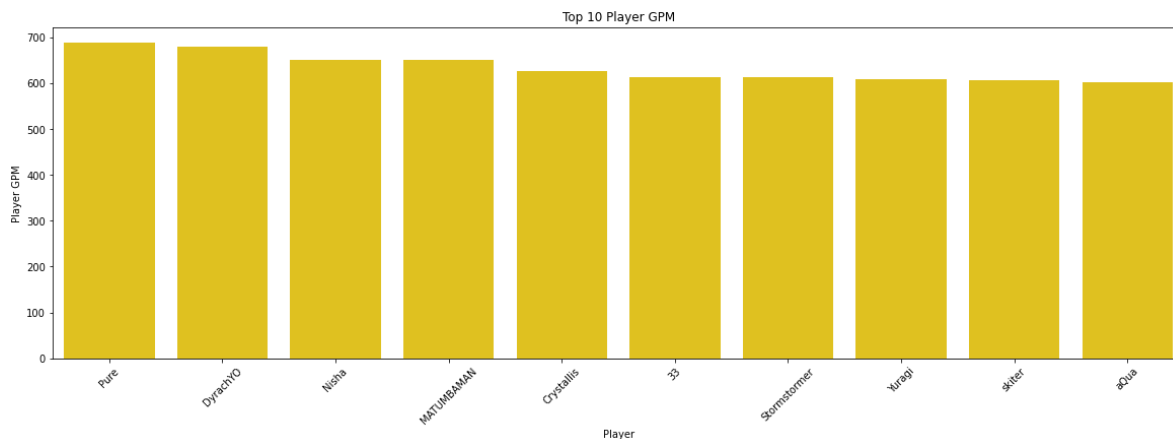
In [112]:

```
# top 10 Players with the highest GPM
```

```
dpc_weu_gpm = dpc_weu_avg_df.sort_values(by = 'GPM', ascending = False)[:10]
plt.figure(figsize = (20, 6))
plt.title("Top 10 Player GPM")
gpm_chart = sns.barplot(x = dpc_weu_gpm['Player'], y = dpc_weu_gpm['GPM'], label = 'Players', color
plt.ylabel('Player GPM')
gpm_chart.set_xticklabels(gpm_chart.get_xticklabels(), rotation = 45)
gpm_chart
```

Out[112]:

```
<AxesSubplot:title={'center':'Top 10 Player GPM'}, xlabel='Player', ylabel='Player GPM'>
```



From the chart, most position 1 players have the highest GPM. **Pure** from Entity has the highest average gpm of all West EU DPC players. **Nisha** from Team Secret has the highest GPM of all position 2 players.

Position Analysis - Position 5 (Hard Support)

In [113]:

#Best Player at each position:

#isolating position 5 (hard support) players

mask_5 = dpc_weu_avg_df['Position'].values == 'hard-support(safelane_support)'

pos_5_players = dpc_weu_avg_df.loc[mask_5]

print(pos_5_players)

	Player	Position	Total Count	Wins	Losses	\
1	Puppey	hard-support(safelane_support)	20	11	9	
6	Fishman	hard-support(safelane_support)	19	11	8	
10	Peksu	hard-support(safelane_support)	18	5	13	
18	INSAiA	hard-support(safelane_support)	18	13	5	
24	Seleri	hard-support(safelane_support)	18	7	11	
26	Sneyking	hard-support(safelane_support)	17	13	4	
33	Misha	hard-support(safelane_support)	16	10	6	
35	CTOMAHEH1	hard-support(safelane_support)	14	0	14	

	Winrate	As Radiant	As Dire	Kills	Deaths	...	XPM	Last Hits	Denies	\
1	55.00%	10	10	1.80	7.15	...	420	74	6	
6	57.89%	5	14	2.37	9.21	...	379	36	8	
10	27.78%	7	11	1.89	7.94	...	346	55	3	
18	72.22%	12	6	2.89	4.28	...	411	63	2	
24	38.89%	11	7	1.89	6.89	...	378	57	3	
26	76.47%	13	4	2.12	4.41	...	460	63	5	
33	62.50%	9	7	1.69	7.81	...	381	32	2	
35	0.00%	3	11	1.79	7.57	...	303	60	3	

	LVL	Hero	Damage	Tower Damage	Hero Heal	Gold Sum	Efficiency	\
1	19		12655	582	4012	10392	0.056005	
6	18		11409	251	1939	9079	0.027646	
10	17		10963	905	5289	9834	0.092028	
18	18		12100	1061	3850	10806	0.098186	
24	16		8582	1057	5801	8865	0.119233	
26	18		9398	398	7691	9920	0.040121	
33	16		10406	904	1359	7779	0.116210	
35	14		9429	371	1128	7989	0.046439	

	Average Game Length (min)
1	34.989899
6	33.750929
10	36.422222
18	34.634615
24	32.119565
26	32.418301
33	31.240964
35	31.577075

[8 rows x 24 columns]

In [114]:

```
#For position 5, the best players should be analyzed based on their general impact.
#By dividing gold sum from hero damage, we can see the ratio of gold they consumes (resources of a t
#I will call this index 'resource/impact ratio'
pos_5_players['resource/impact ratio'] = ((pos_5_players['Hero Damage'] + pos_5_players['Hero Heal']
top_pos_5_players = pos_5_players[['Player', 'KDA', 'Winrate', 'GPM', 'XPM', 'Hero Damage', 'Hero He
print(top_pos_5_players)
```

	Player	KDA	Winrate	GPM	XPM	Hero Damage	Hero Heal	\
26	Sneyking	3.71	76.47%	306	460	9398	7691	
10	Peksu	1.86	27.78%	270	346	10963	5289	
24	Seleri	2.43	38.89%	276	378	8582	5801	

	resource/impact ratio
26	1.722681
10	1.652634
24	1.622448

C:\Users\Dominic\AppData\Local\Temp\ipykernel_27988\1032153592.py:4: 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 (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
pos_5_players['resource/impact ratio'] = ((pos_5_players['Hero Damage'] + pos_5_players['Hero Heal']) / pos_5_players['Gold Sum']).astype(float)
```

Position Analysis - Position 4 (Soft Support)

In [81]:

#Best Player at each position:

#isolating position 4 (Soft support) players

mask_4 = dpc_weu_avg_df['Position'].values == 'soft-support(offlane-support)'

pos_4_players = dpc_weu_avg_df.loc[mask_4]

print(pos_4_players)

	Player	Position	Total Count	Wins	Losses	\
5	Kataomi`	soft-support(offlane-support)	19	11	8	
11	tOfu	soft-support(offlane-support)	18	7	11	
13	Handsken	soft-support(offlane-support)	18	5	13	
16	Zayac	soft-support(offlane-support)	18	11	7	
29	Saksa	soft-support(offlane-support)	17	13	4	
30	Taiga	soft-support(offlane-support)	16	10	6	
37	Aramis	soft-support(offlane-support)	14	0	14	
41	YapzOr	soft-support(offlane-support)	2	0	2	

	Winrate	As Radiant	As Dire	Kills	Deaths	...	XPM	Last Hits	Denies	\
5	57.89%	5	14	4.37	5.05	...	476	104	2	
11	38.89%	11	7	2.89	5.72	...	441	101	5	
13	27.78%	7	11	3.61	7.72	...	409	73	3	
16	61.11%	9	9	2.67	7.56	...	477	116	5	
29	76.47%	13	4	2.94	5.88	...	438	91	4	
30	62.50%	9	7	2.38	5.56	...	428	84	9	
37	0.00%	3	11	1.79	8.64	...	323	57	2	
41	0.00%	1	1	1.00	7.50	...	428	114	6	

	LVL	Hero	Damage	Tower	Damage	Hero	Heal	Gold	Sum	Efficiency	\
5	20		17455		385		2235		12710	0.030291	
11	18		11515		277		1717		10733	0.025808	
13	18		14184		309		3291		10215	0.030250	
16	20		19223		635		2224		13272	0.047845	
29	18		12312		1257		1299		9869	0.127369	
30	17		14513		919		986		9753	0.094227	
37	15		11149		60		387		7586	0.007909	
41	18		14286		284		396		11610	0.024462	

	Average Game Length (min)
5	35.903955
11	32.822630
13	35.346021
16	37.176471
29	29.906061
30	30.101852
37	29.632812
41	35.396341

[8 rows x 24 columns]

In [82]:

```
#For position 4, the best players should be analyzed based on their assist and their general impact.
#By dividing gold sum from hero damage, we can see the ratio of gold they consumes (resources of a t
#I will call this index 'resource/impact ratio'
pos_4_players['resource/impact ratio'] = ((pos_4_players['Hero Damage'] + pos_4_players['Hero Heal']
top_pos_4_players = pos_4_players[['Player', 'KDA', 'Winrate', 'GPM', 'XPM', 'Hero Damage', 'Hero He
print(top_pos_4_players)
```

	Player	KDA	Winrate	GPM	XPM	Hero Damage	Hero Heal	\
13	Handsken	2.00	27.78%	289	409	14184	3291	
16	Zayac	2.95	61.11%	357	477	19223	2224	
30	Taiga	3.57	62.50%	324	428	14513	986	

	resource/impact ratio
13	1.710720
16	1.615958
30	1.589152

C:\Users\Dominic\AppData\Local\Temp\ipykernel_27988\3353686895.py:4: 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 (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
pos_4_players['resource/impact ratio'] = ((pos_4_players['Hero Damage'] + pos_4_players['Hero Heal']) / pos_4_players['Gold Sum']).astype(float)
```

Position Analysis - Position 3 (Off-lane Carry)

In [83]:

```
#Best Player at each position:

#isolating position 3 (offlane carry) players
mask_3 = dpc_weu_avg_df['Position'].values == 'offlane carry'
pos_3_players = dpc_weu_avg_df.loc[mask_3]
print(pos_3_players)
```

	Player	Position	Total Count	Wins	Losses	Winrate	\
3	Iceiceice	offlane carry	20	11	9	55.00%	
4	Tobi	offlane carry	19	11	8	57.89%	
12	Ace	offlane carry	18	7	11	38.89%	
14	S4	offlane carry	18	5	13	27.78%	
20	Zai	offlane carry	18	13	5	72.22%	
21	Boxi	offlane carry	18	13	5	72.22%	
25	33	offlane carry	17	13	4	76.47%	
32	ATF	offlane carry	16	10	6	62.50%	
39	symetrical	offlane carry	10	0	10	0.00%	
40	Eternal Sadness	offlane carry	2	0	2	0.00%	
42	Limitless	offlane carry	2	0	2	0.00%	

	As Radiant	As Dire	Kills	Deaths	...	XPM	Last Hits	Denies	LVL	\
3	10	10	5.25	5.70	...	631	280	11	23	
4	5	14	3.53	3.89	...	545	233	22	22	
12	11	7	4.11	5.06	...	582	259	15	21	
14	7	11	3.56	6.22	...	513	200	11	20	
20	12	6	6.83	3.28	...	657	281	9	23	
21	12	6	4.44	4.89	...	487	75	3	20	
25	13	4	5.94	2.82	...	668	287	7	22	
32	9	7	6.69	4.38	...	641	222	10	21	
39	2	8	1.90	6.00	...	458	184	9	19	
40	1	1	2.00	6.00	...	418	140	2	16	
42	0	2	2.00	10.00	...	428	198	8	18	

	Hero Damage	Tower Damage	Hero Heal	Gold Sum	Efficiency	\
3	18919	2946	2440	19175	0.153638	
4	22615	4561	910	16371	0.278602	
12	16890	4162	1288	16369	0.254261	
14	17111	3430	1272	14010	0.244825	
20	23821	2054	2693	19526	0.105193	
21	12567	182	1167	12041	0.015115	
25	17324	7616	1420	20184	0.377329	
32	18101	1568	0	15133	0.103615	
39	12613	1130	480	12292	0.091930	
40	10386	2205	1838	11558	0.190777	
42	19048	1354	1022	13820	0.097974	

	Average Game Length (min)
3	38.273453
4	36.871622
12	33.474438
14	33.277910
20	35.959484
21	37.628125
25	32.872964
32	30.695740
39	33.221622
40	36.006231
42	30.917226

[11 rows x 24 columns]

In [84]:

```
#For position 3, the best players should be analyzed based on their KDA, hero damage, tower damage,
#The index would be calculated by (hero damage + tower damage / gold sum) * KDA
pos_3_players['carry index'] = ((pos_3_players['Hero Damage'] + pos_3_players['Tower Damage']) / pos
top_pos_3_players = pos_3_players[['Player', 'KDA', 'Winrate', 'Hero Damage', 'Tower Damage', 'Gold
print(top_pos_3_players)
```

	Player	KDA	Winrate	Hero Damage	Tower Damage	Gold Sum	carry index
4	Tobi	6.92	57.89%	22615	4561	16371	1.660009
42	Limitless	0.46	0.00%	19048	1354	13820	1.476266
14	S4	2.28	27.78%	17111	3430	14010	1.466167

C:\Users\Dominic\AppData\Local\Temp\ipykernel_27988\3398803657.py:3: 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

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```
pos_3_players['carry index'] = ((pos_3_players['Hero Damage'] + pos_3_players['Tower Damage']) / pos_3_players['Gold Sum']).astype(float)
```

Position Analysis - Position 2 (Mid-lane Carry)

In [85]:

#Best Player at each position:

#isolating position 2 (midlane carry) players

mask_2 = dpc_weu_avg_df['Position'].values == 'midlane carry'

pos_2_players = dpc_weu_avg_df.loc[mask_2]

print(pos_2_players)

	Player	Position	Total Count	Wins	Losses	Winrate	\
0	Nisha	midlane carry	20	11	9	55.00%	
8	Stormstormer	midlane carry	19	11	8	57.89%	
9	MiCkE	midlane carry	18	13	5	72.22%	
22	Limp	midlane carry	18	5	13	27.78%	
23	BOOM	midlane carry	18	7	11	38.89%	
27	Nine	midlane carry	17	13	4	76.47%	
34	bzm	midlane carry	16	10	6	62.50%	
38	W33	midlane carry	14	0	14	0.00%	

	As Radiant	As Dire	Kills	Deaths	...	XPM	Last Hits	Denies	LVL	\
0	10	10	9.30	2.80	...	759	405	10	26	
8	5	14	8.37	3.00	...	736	373	10	25	
9	12	6	7.33	3.17	...	679	303	11	23	
22	7	11	6.44	3.56	...	605	290	6	22	
23	11	7	7.28	3.56	...	678	292	9	23	
27	13	4	6.76	3.00	...	681	271	8	22	
34	9	7	7.38	3.31	...	696	292	8	22	
38	3	11	5.86	6.14	...	571	275	6	21	

	Hero Damage	Tower Damage	Hero Heal	Gold Sum	Efficiency	\
0	39091	3843	323	25470	0.150883	
8	28023	5254	27	21787	0.241153	
9	27182	2318	375	19408	0.119435	
22	29786	1219	36	18677	0.065267	
23	24587	1364	52	17970	0.075904	
27	27013	1404	1196	17353	0.080908	
34	26958	1621	886	19214	0.084366	
38	24698	463	1318	16140	0.028686	

	Average Game Length (min)
0	39.124424
8	35.541599
9	34.906475
22	36.266019
23	32.851920
27	31.323105
34	33.242215
38	32.738337

[8 rows x 24 columns]

In [86]:

```
#For position 2, the best players should be analyzed based on their KDA, hero damage, tower damage,
#The index would be calculated by (hero damage + tower damage / gold sum) * KDA
pos_2_players['carry index'] = ((pos_2_players['Hero Damage'] + pos_2_players['Tower Damage']) / pos
top_pos_2_players = pos_2_players[['Player', 'KDA', 'Winrate', 'Hero Damage', 'Tower Damage', 'Gold
print(top_pos_2_players)
```

	Player	KDA	Winrate	Hero Damage	Tower Damage	Gold Sum	carry index
0	Nisha	8.76	55.00%	39091	3843	25470	1.685669
22	Limm	4.94	27.78%	29786	1219	18677	1.660063
27	Nine	7.70	76.47%	27013	1404	17353	1.637584

C:\Users\Dominic\AppData\Local\Temp\ipykernel_27988\1646841894.py:3: SettingWithCopyWarning:

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```
pos_2_players['carry index'] = ((pos_2_players['Hero Damage'] + pos_2_players['Tower Damage']) / pos_2_players['Gold Sum']).astype(float)
```

Position Analysis - Position 1 (Safe-lane Carry)

In [87]:

```
#Best Player at each position:

#isolating position 1 (safelane carry) players
mask_1 = dpc_weu_avg_df['Position'].values == 'safelane carry'
pos_1_players = dpc_weu_avg_df.loc[mask_1]
print(pos_1_players)
```

	Player	Position	Total Count	Wins	Losses	Winrate	As Radiant	\
2	Crystallis	safelane carry	20	11	9	55.00%	10	
7	Pure	safelane carry	19	11	8	57.89%	5	
15	DyrachYO	safelane carry	18	7	11	38.89%	11	
17	aQua	safelane carry	18	5	13	27.78%	7	
19	MATUMBAMAN	safelane carry	18	13	5	72.22%	12	
28	skiter	safelane carry	17	13	4	76.47%	13	
31	Yuragi	safelane carry	16	10	6	62.50%	9	
36	Nikobaby	safelane carry	14	0	14	0.00%	3	

	As Dire	Kills	Deaths	...	XPM	Last Hits	Denies	LVL	Hero Damage	\
2	10	7.05	3.90	...	688	390	10	25	26695	
7	14	9.58	3.42	...	696	396	9	25	31480	
15	7	6.56	4.89	...	718	366	13	23	24453	
17	11	5.22	3.50	...	643	367	9	23	24424	
19	6	9.06	2.50	...	717	362	12	24	27982	
28	4	6.82	2.41	...	668	330	19	22	18914	
31	7	7.44	2.69	...	688	281	10	22	17907	
36	11	4.43	3.57	...	630	302	5	22	17625	

	Tower Damage	Hero Heal	Gold Sum	Efficiency	Average Game Length (min)
2	9438	3	23111	0.408377	36.918530
7	3356	447	25576	0.131217	37.174419
15	5292	1764	22362	0.236651	32.933726
17	5413	685	21570	0.250950	35.830565
19	13205	181	23176	0.569770	35.655385
28	5108	196	18966	0.269324	31.297030
31	6637	705	18666	0.355566	30.700658
36	1631	514	17788	0.091691	33.124767

[8 rows x 24 columns]

In [88]:

```
#For position 1, the best players should be analyzed based on their KDA, hero damage, tower damage,
#The index would be calculated by (hero damage + tower damage / gold sum) * KDA
pos_1_players['carry index'] = ((pos_1_players['Hero Damage'] + pos_1_players['Tower Damage']) / pos
top_pos_1_players = pos_1_players[['Player', 'KDA', 'Winrate', 'Hero Damage', 'Tower Damage', 'Gold
print(top_pos_1_players)
```

	Player	KDA	Winrate	Hero Damage	Tower Damage	Gold Sum	carry index
19	MATUMBAMAN	8.06	72.22%	27982	13205	23176	1.777140
2	Crystallis	5.45	55.00%	26695	9438	23111	1.563455
17	aQua	3.89	27.78%	24424	5413	21570	1.383264

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```
pos_1_players['carry index'] = ((pos_1_players['Hero Damage'] + pos_1_players['Tower Damage']) / pos_1_players['Gold Sum']).astype(float)
```