

DATA STORY TELLING PRESENTATION

- Nivedha A



Table of CONTENTS

01

INSIGHTS: EACH DATASET

- *Bonus Cost Data*
- *Activity Player Data*
- *First Bet Data*
- *Player Details Data*
- *First Deposit Data*

02

KPI CHART USING POWER BI

03

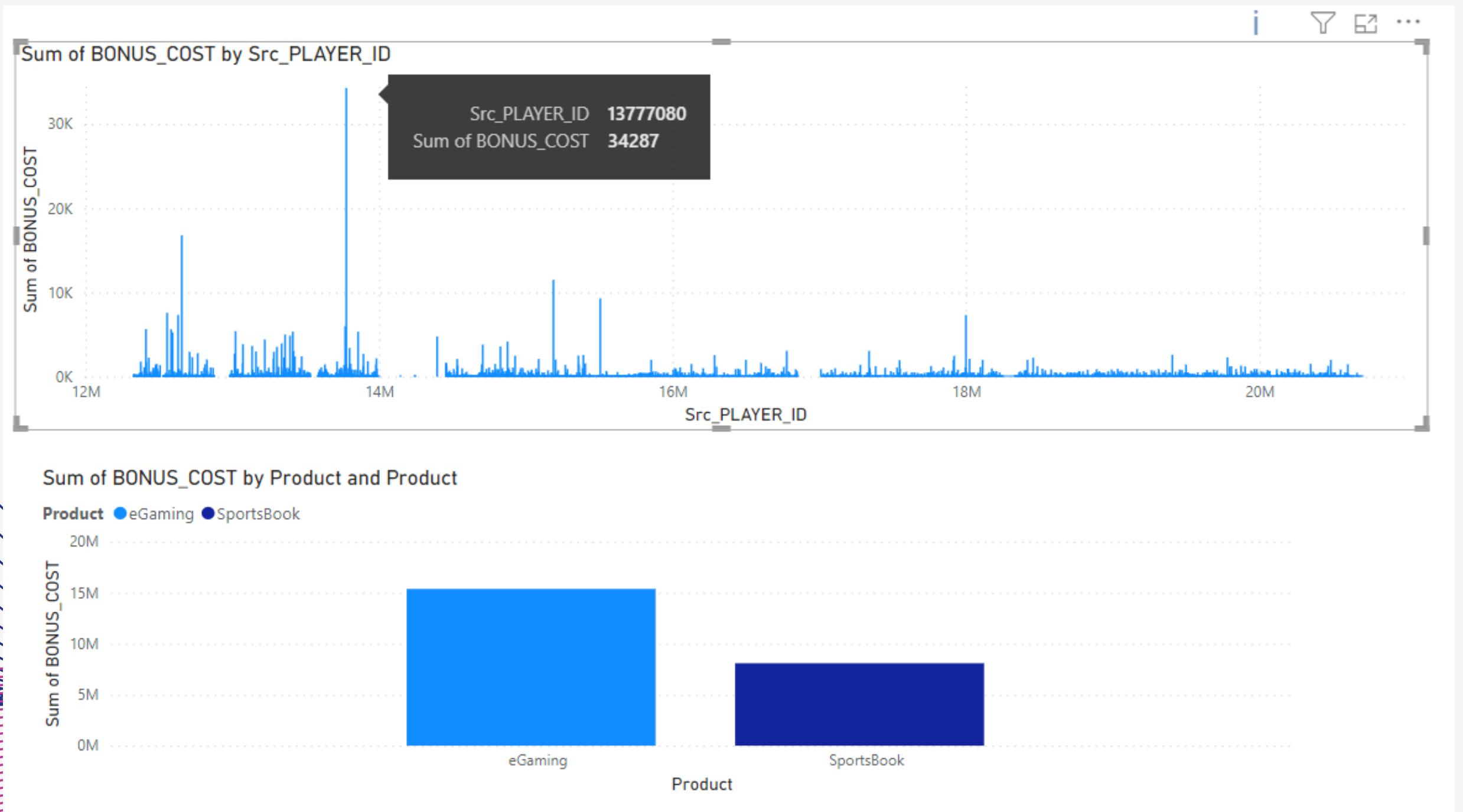
INSIGHTS AND METRICS AFTER
GROUPING THE DATASET



INSIGHTS : BONUS COST DATA

INFERENCE

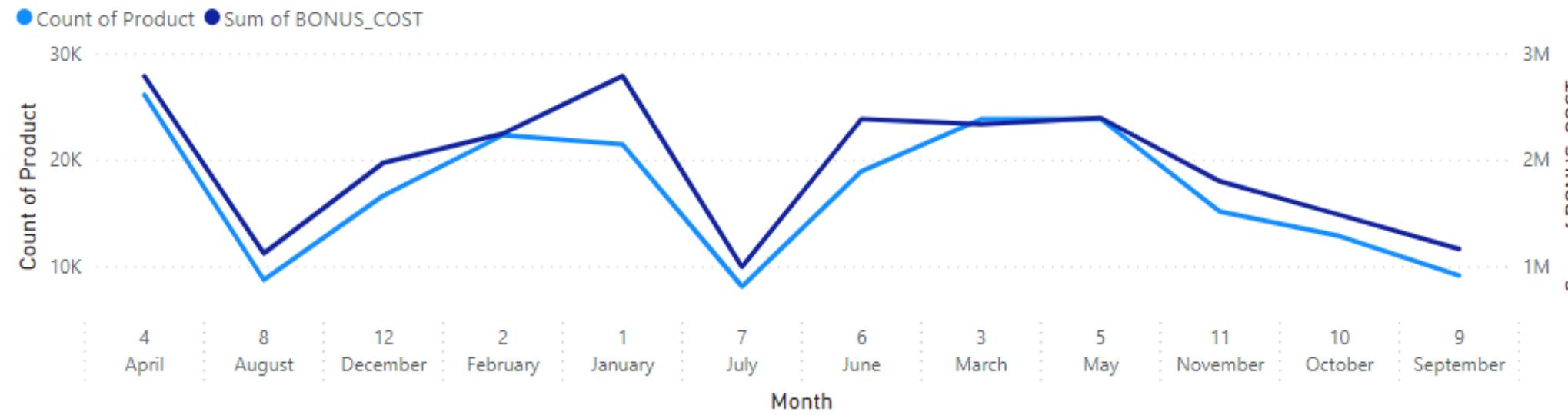
- Player with id **34287** had the maximum **bonus cost**
- **eGaming** produces most **bonus cost** than the **Sports book** in the **product category**



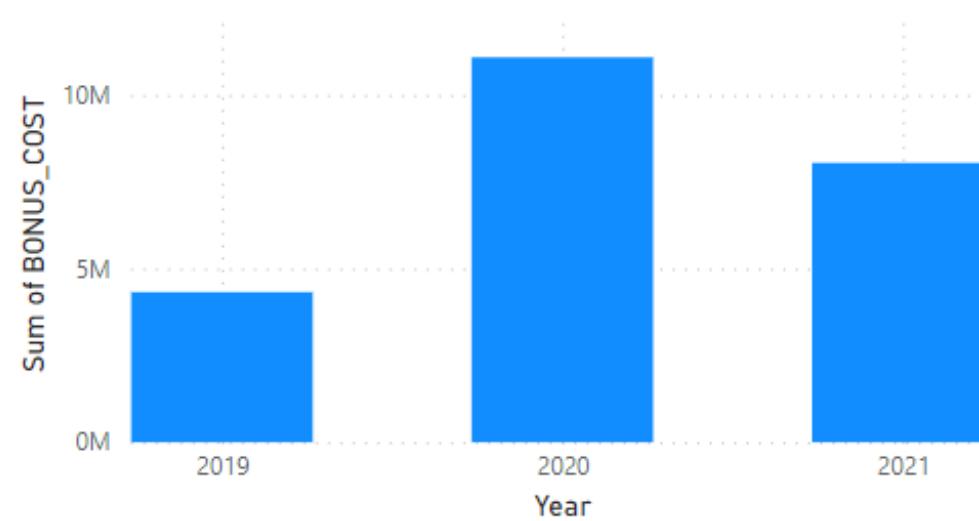
INFERENCE

- **Product count and bonus cost** is low at months August , July and September
- We observe the pattern of raise and fall of **bonus cost** and **product count**
- **Both** are maximum at January which may be probably because it is the start of the year , and at that time bonus and product cost is high
- Comparitively **2020** performed well in obtaining **bonus**
- **eGaming** has lot of **product count** than **Sportsbook**

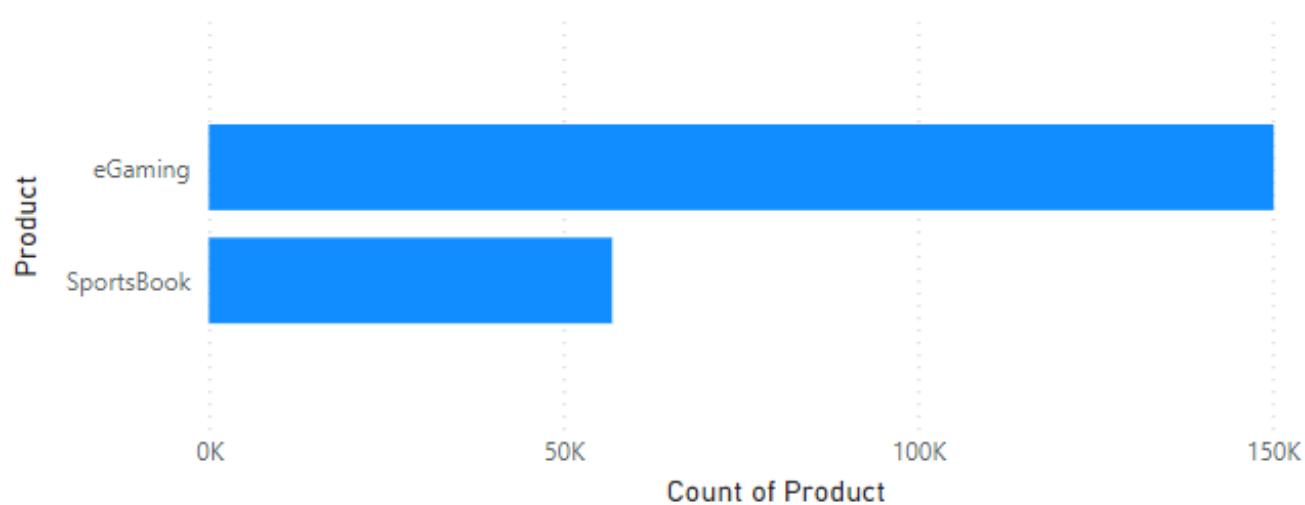
Count of Product and Sum of BONUS_COST by MonthName and Month



Sum of BONUS_COST by Year



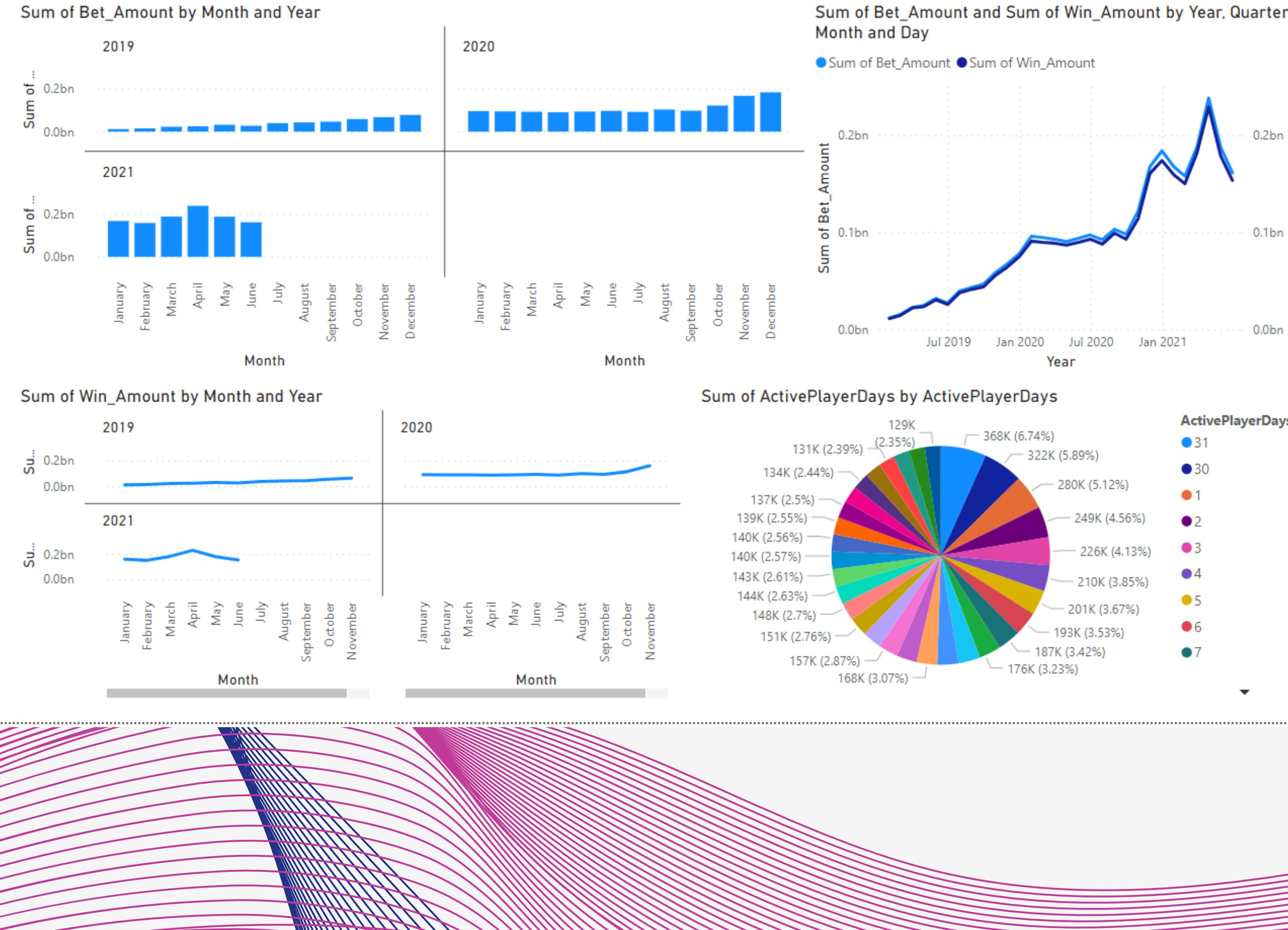
Count of Product by Product





INSIGHTS:PLAYER ACTIVITY DATA

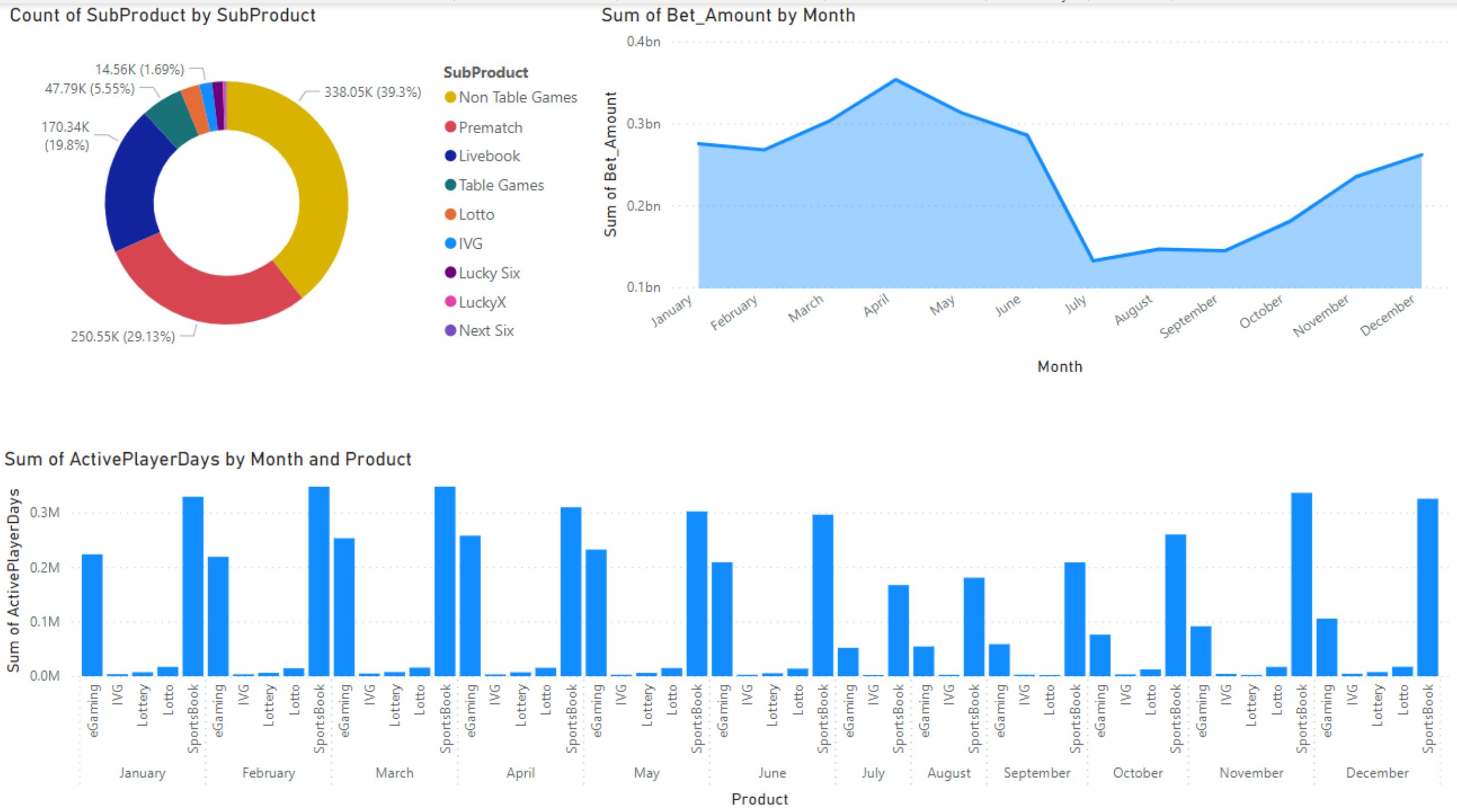
INFERENCE



- It seems that **Bet amount** have been increased through the years
- In **2019 and 2020, bet amount** gradually increased throughout the year
- In **2021 , bet amount** have high and low fluctuation but only to light extent
- **Bet amount and win amount** are almost equal / close to eachother throughout the year
- **Winning amount** is consistent in both 2019 and 2020 and have light fluctuation in 2021
- We observe large number of **Active players** in the end and start of the month

INFERENCE

- **Non Table Games** have maximum count and **Lucky Six** have minimum count
- The **bet amount** is minimum at mid month (July) and maximum at April
- In all the months **SportsBook** have highest **Active player days** and the second highest is **eGaming**
- **IVG** have least Activeplayer days

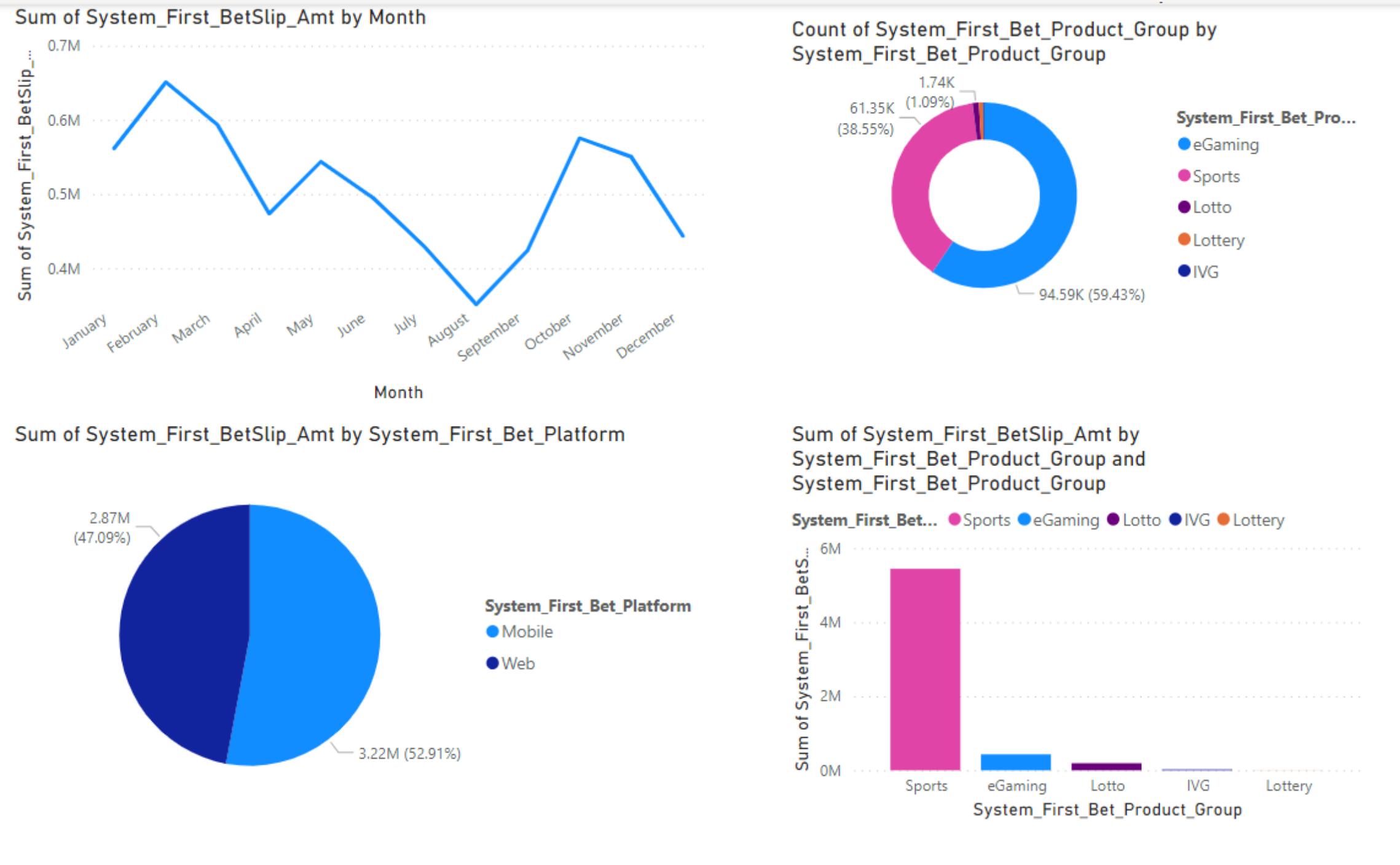




INSIGHTS :FIRST BET DATA

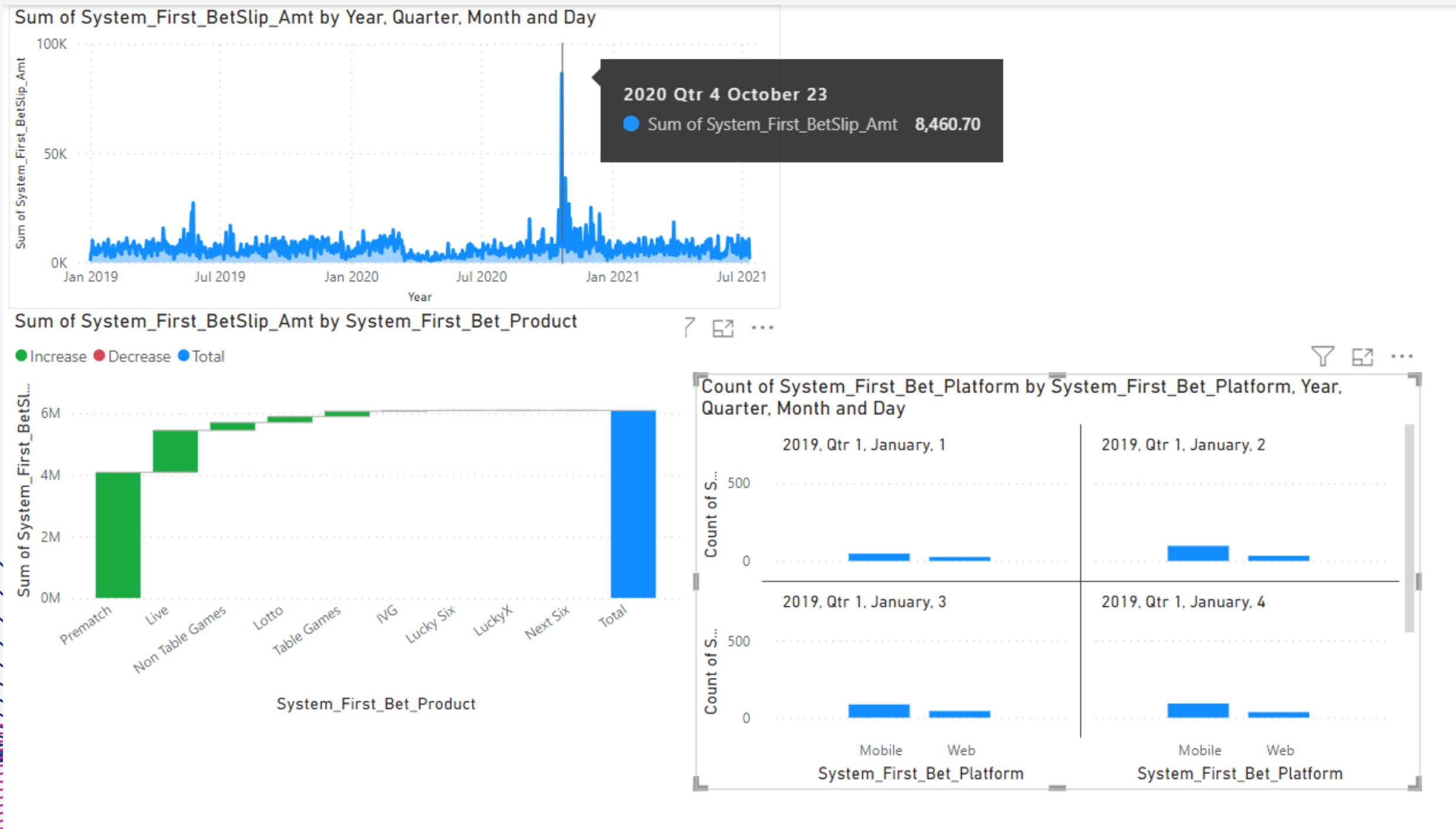
INFERENCE

- By the **bet slip amount** we can infer that the least is observed at August and Maximum at February . Also there is consistent raise and fall of Bet slip amount
- Even though the count of **eGaming** is maximum followed by **sports** and the least is **IVG** (*by doughnut chart*)
- The maximum **bet slip amount** is produced by **Sports** followed by **eGaming** and we can observe that the difference of bet slip amount is huge between Sports and eGaming (*by bar chart*)
- Large number of users use **Mobile** platform than **web**



INFERENCE

- The highest **Bet slip amount** is recorded from October 19 to October 24
- We have observed Increase in first **bet slip amount** in all the first bet product (*by waterfall chart*)
- Comparing all the years , its clear that most people prefer **mobile** than **web**

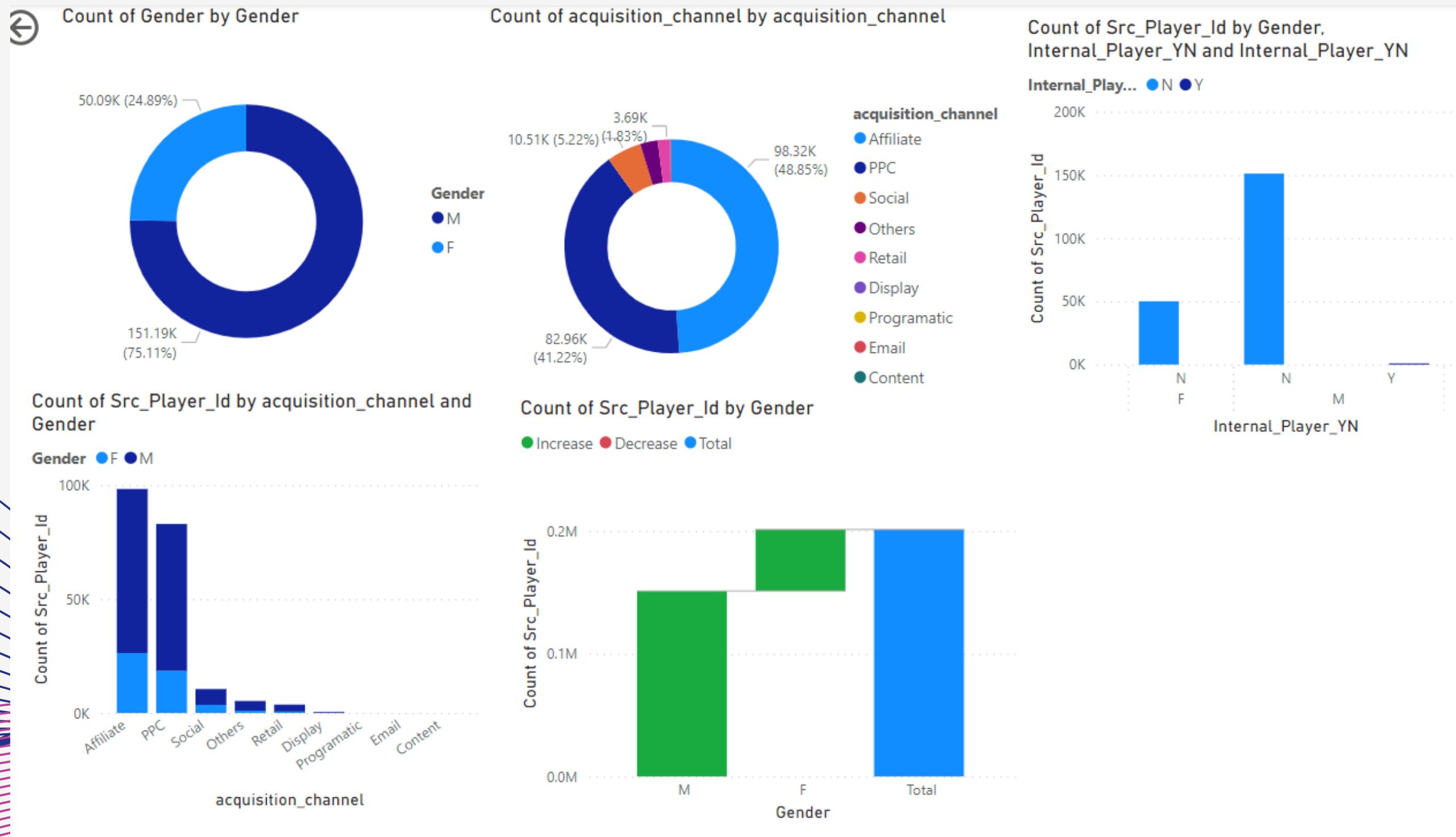




INSIGHTS:PLAYER DETAILS DATA

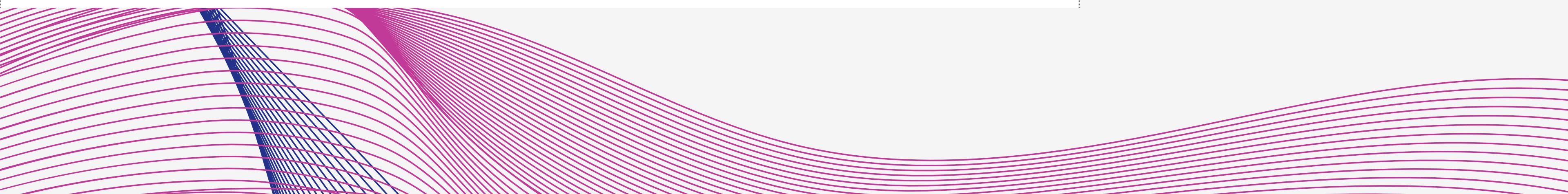
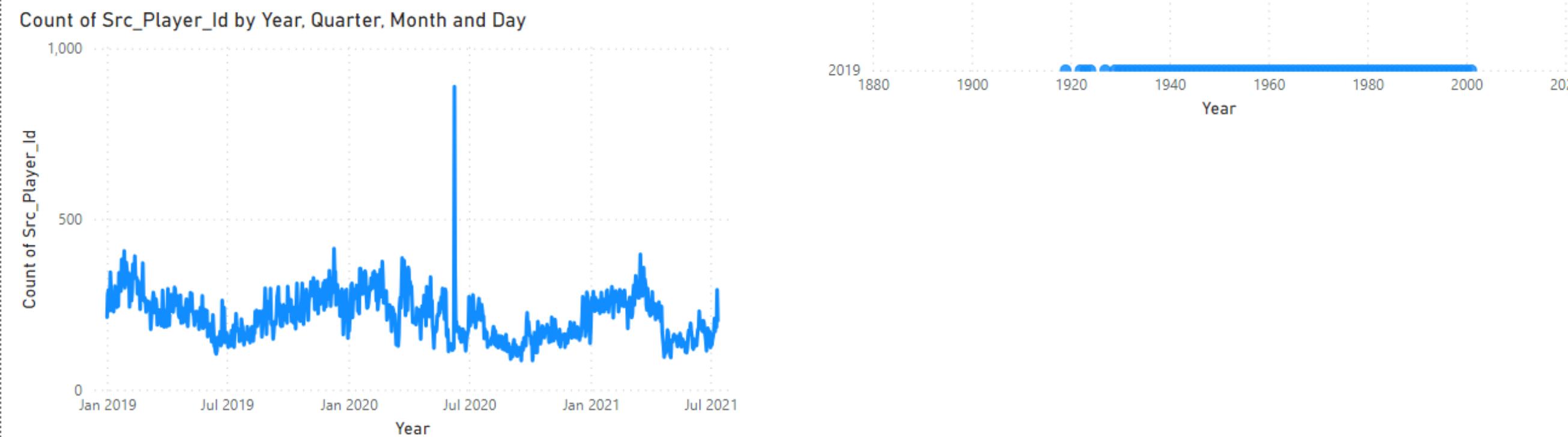
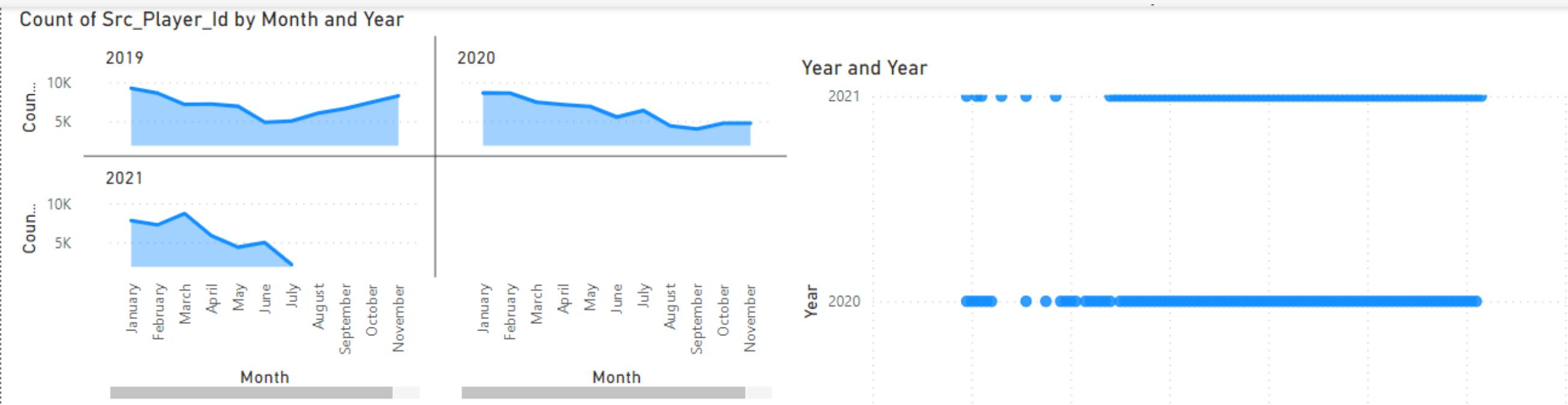
INFERENCE

- More number of **male** players than **female** players
- In **aquisition channel affiliate** is in high count followed by **PPC** and the least is **content**
- **Male** have maximum count in all **aquisition channel**
- Internal player is "**No**" most probably for both **Female** and **Male**
- Count of **Male and Female** have increased throughout the year



INFERENCE

- We can observe that the **players** are high at starting of year and there is gradual decrease and again it increases
- Highest count of **players** are observed in June month
- The scatter plot is to know if there is any relationship between DOB and signup date , we can observe that regardless of when they are born most of the people have been signed up on 2020

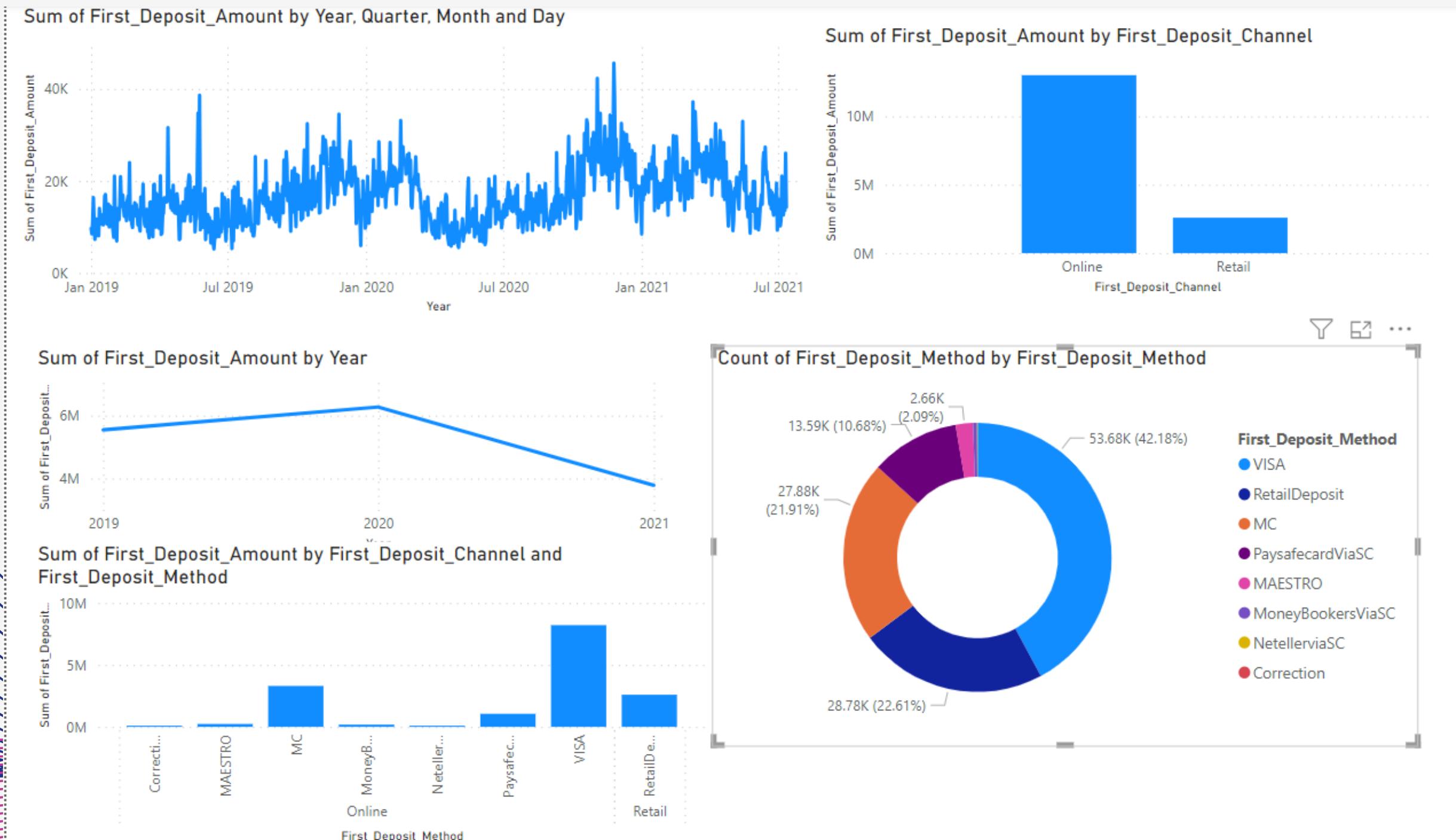




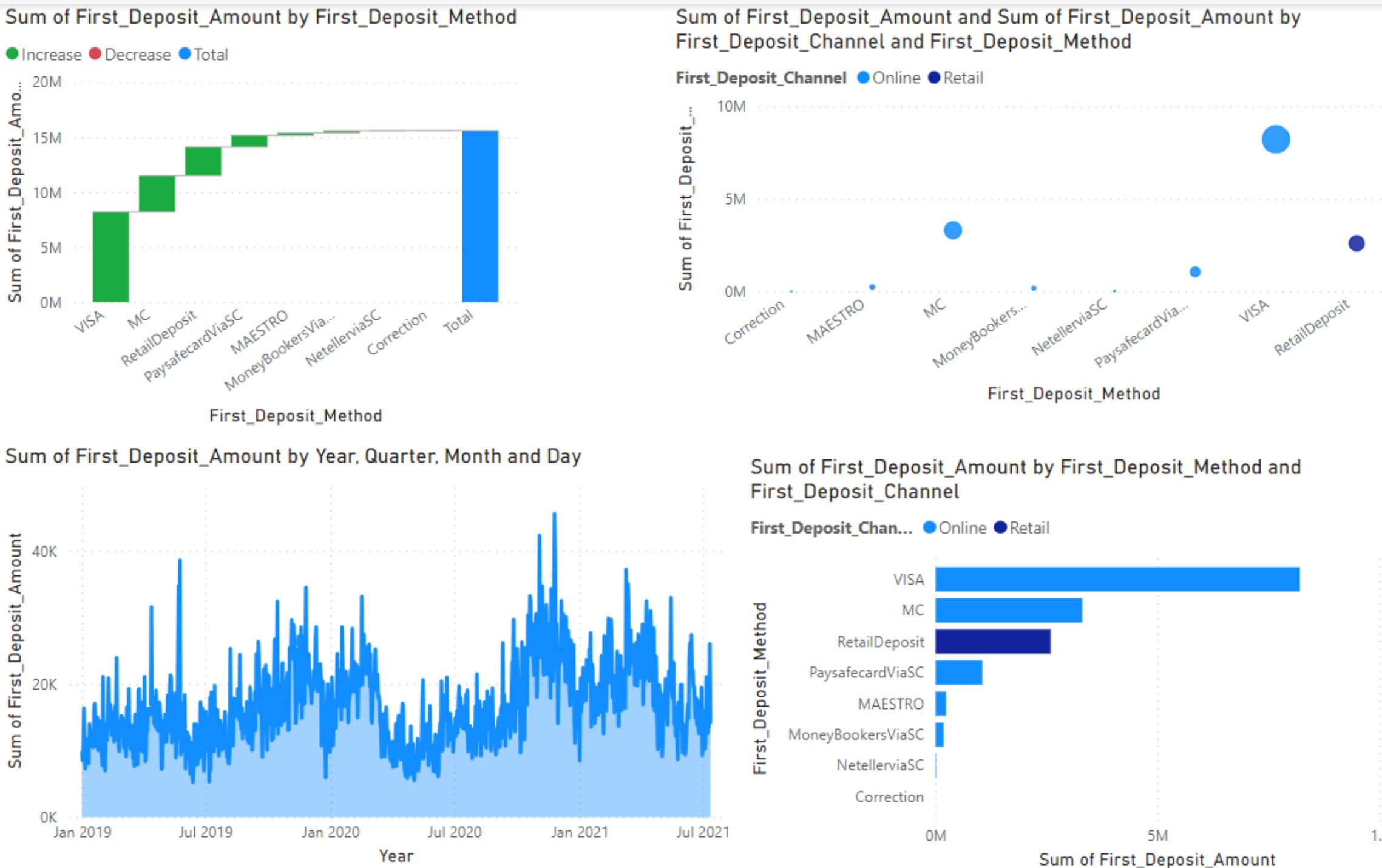
INSIGHTS :FIRST DEPOSIT DATA

INFERENCE

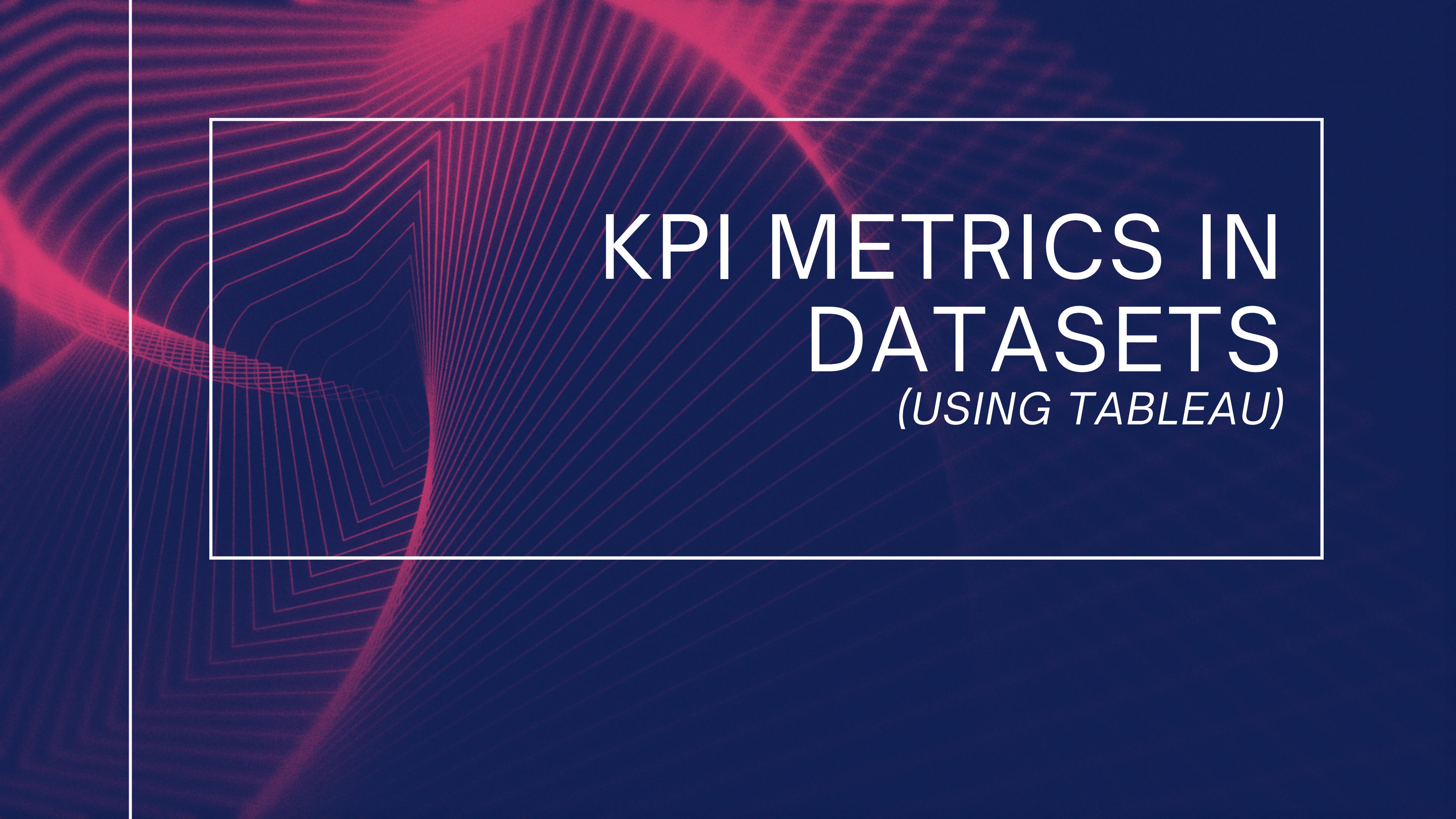
- First **deposit amount** is maximum at November 2020 and minimum at December 2019
- **First deposit amount by Online channel** is greater than **retail channel**
- **Deposit amounts** have been considerably decreased throughout 3 years
- In "Online" deposit channel "**VISA**" method have large value and in **Retail** deposit channel , "**Retail deposit** have large value .
- **VISA** method have highest count followed by **Retail Deposit**(eventhough Retail deposit is used only by the ones who use retail chnnel) and **MC** and the least one is **Correction**



INFERENCE

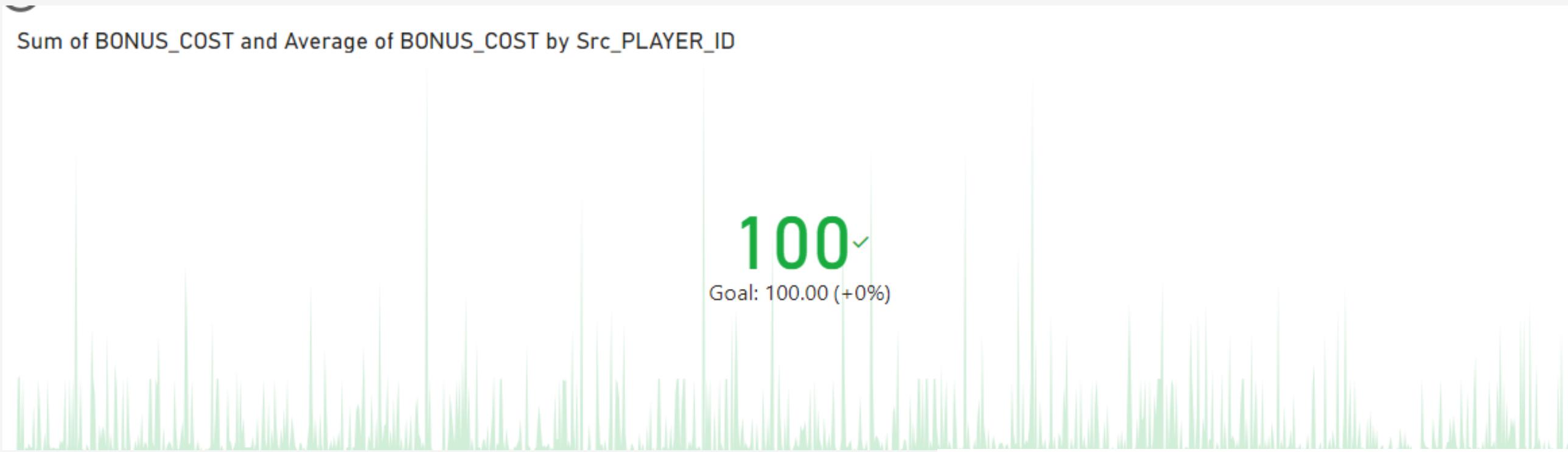


- Deposit methods have seen increase in their use
- The inference we got from the previous slide for first deposit amount in each channel for every method is visualized by *bubble chart*
- The **first deposit amount** is maximum at end of 2020
- **Retail channel** only follows **Retail Deposit method** and rest of the methods are taken by **online channel**
- maximum number of **first deposit amount** are done by **VISA method through online**

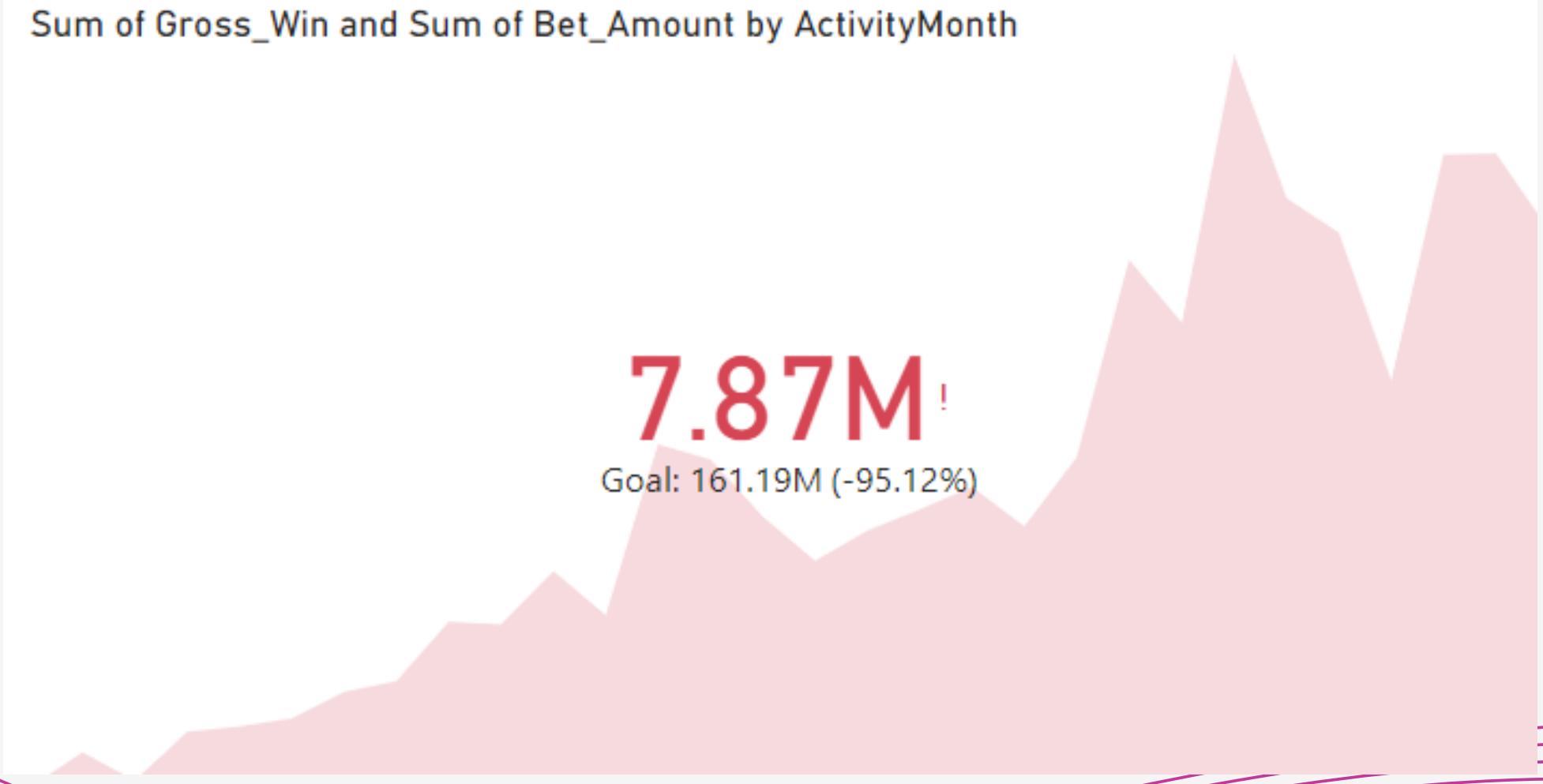


KPI METRICS IN DATASETS

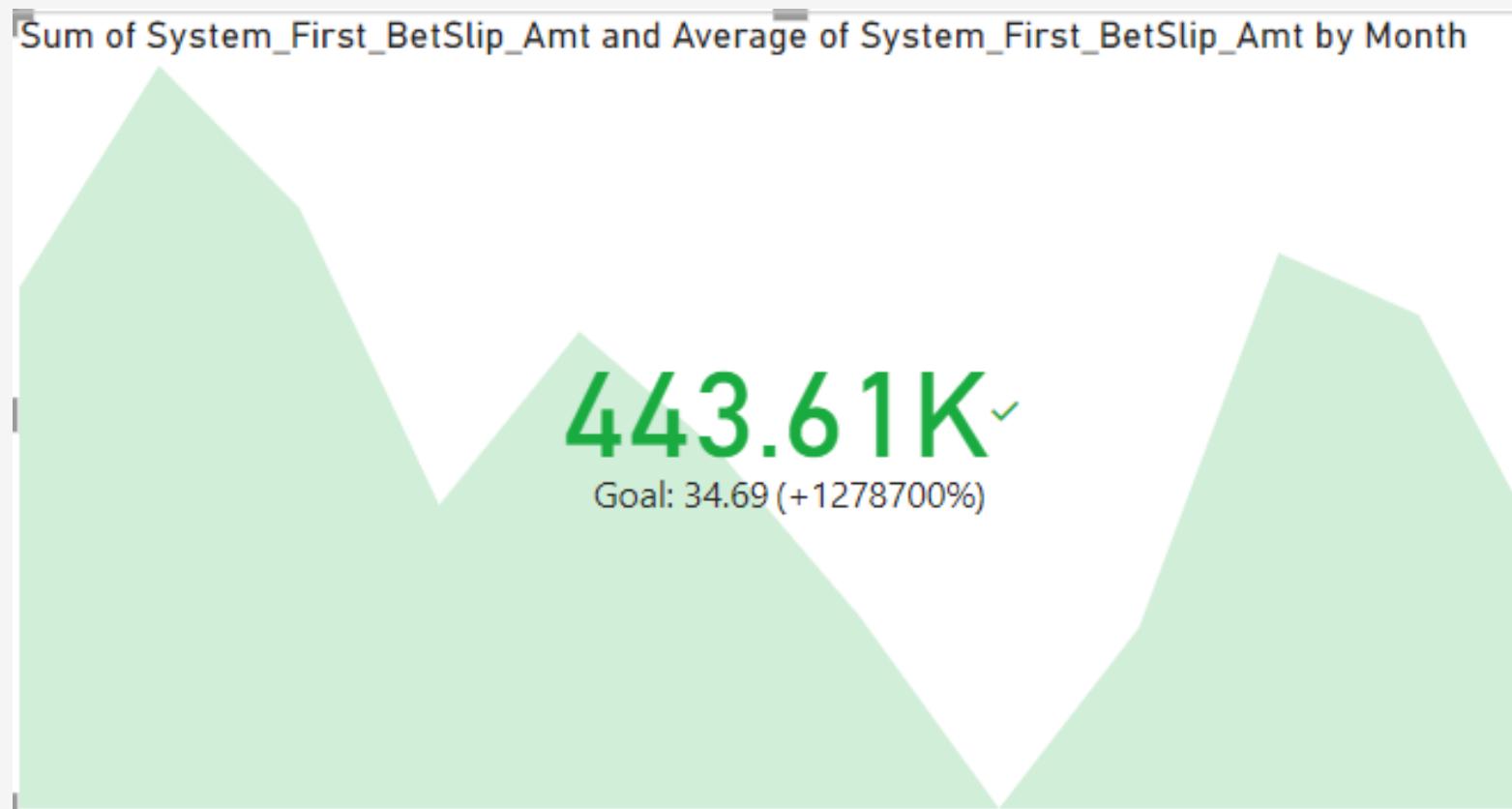
(USING TABLEAU)



- Keeping the target value as average of Bonus cost itself , We can say that we reached the target



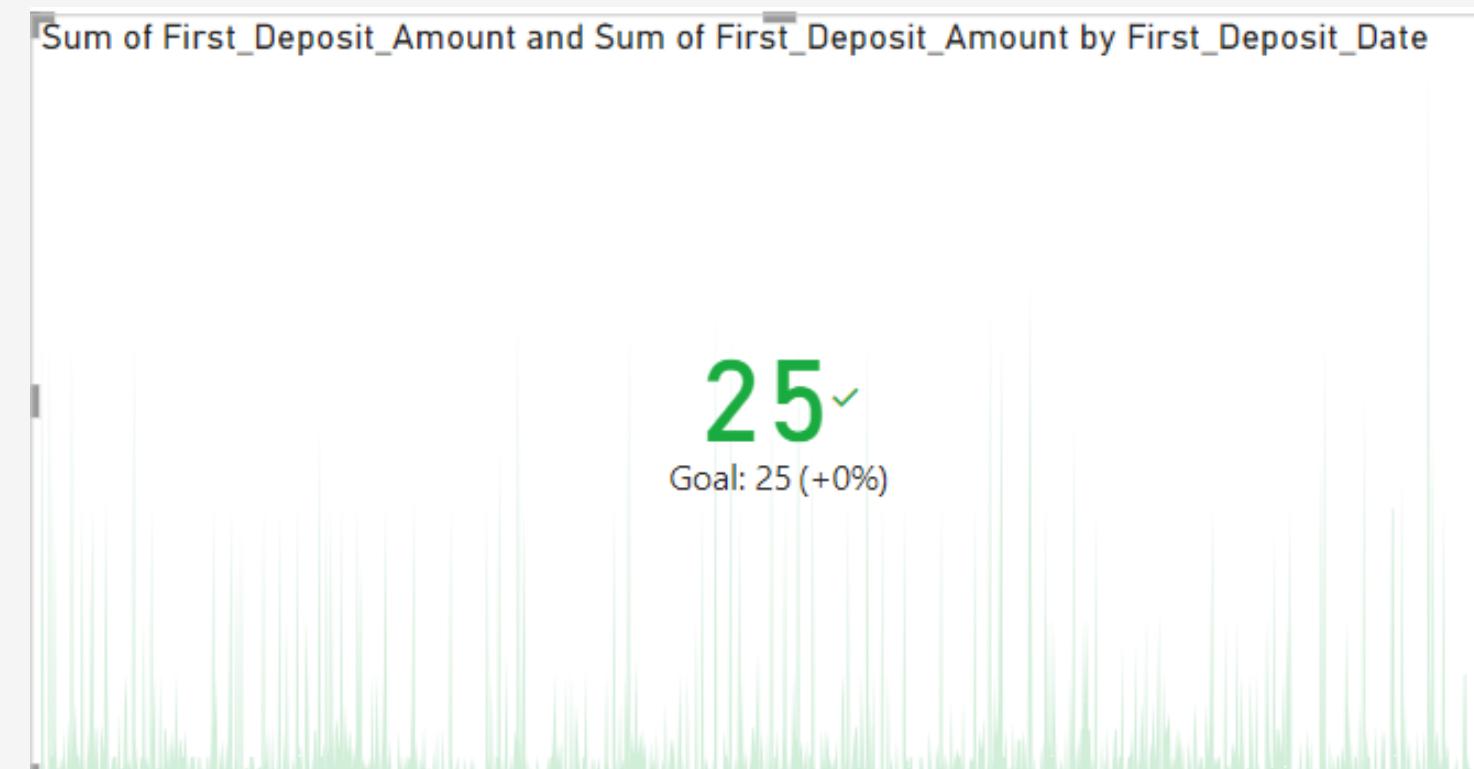
- It is profit when the Win amount is larger than bet amount but we see that the performance is poor



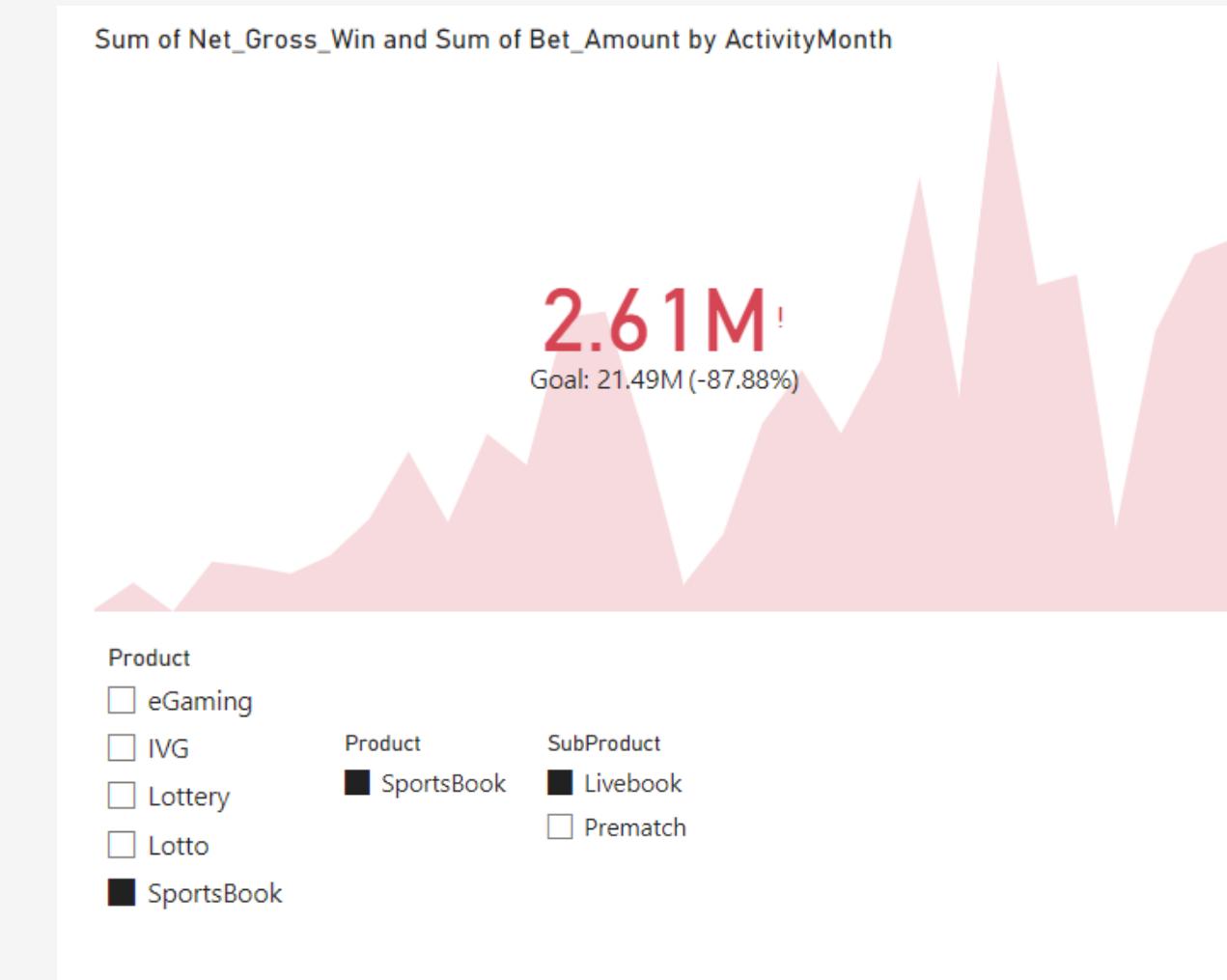
- Keeping the target as average of bet slip amount , we can see that we reached the goal



- We do always observe we have more than average number of players



- The deposit amount is always greater than average amount



- The win amount is always lesser than bet amount in all the product and sub product



INSIGHTS AND METRICS AFTER JOINING AND CORRELATING DATASETS *(USING PYTHON)*

DATA AFTER MERGING AND PREPROCESSING

- Merged all data with the common column "Src_Player_id" using join function to get insights between the datas

	src_player_id	ActivityMonth_x	Product_x	SubProduct	ActivePlayerDays	Bet_Amount	Win_Amount	Gross_Win	Net_Gross_Win	Src_PLAYER_ID	...
2	12845036	2019-05-31	SportsBook	Prematch	28	1229.48	1177.85	41.6300	34.969200	12845036.0	...
6	12580168	2020-12-31	eGaming	Non Table Games	1	74.00	54.60	18.6674	15.680616	12580168.0	...
7	12580168	2020-12-31	eGaming	Non Table Games	1	74.00	54.60	18.6674	15.680616	12580168.0	...
8	12580168	2020-12-31	eGaming	Non Table Games	1	74.00	54.60	18.6674	15.680616	12580168.0	...
9	12580168	2020-12-31	eGaming	Non Table Games	1	74.00	54.60	18.6674	15.680616	12580168.0	...
...
3492207	15816105	2021-05-31	eGaming	Table Games	1	125.00	50.00	75.0000	63.000000	15816105.0	...
3492208	13435803	2020-12-31	eGaming	Non Table Games	1	31.40	11.45	19.9500	16.758000	13435803.0	...
3492209	13435803	2020-12-31	eGaming	Non Table Games	1	31.40	11.45	19.9500	16.758000	13435803.0	...
3492210	13435803	2020-12-31	eGaming	Non Table Games	1	31.40	11.45	19.9500	16.758000	13435803.0	...
3492211	13435803	2020-12-31	eGaming	Non Table Games	1	31.40	11.45	19.9500	16.758000	13435803.0	...

2020634 rows × 35 columns

Note : After merging , the dataset is too large to save as Excel and work in Power Bi because the size was beyond the capacity so here I used python to get insights

INSIGHTS : MERGED DATASET

- Total bonus cost per player
(DESCENDING ORDER)

src_player_id	
12996364	1993563.0
12582603	935906.0
13928829	934550.0
12655429	931676.0
13781394	815512.0
...	
13879352	0.0
16488856	0.0
13059359	0.0
20119385	0.0
14889757	0.0

"Note: While the above data we derived provides valuable insights, I chose not to display the top 5 rows/values in the bar chart intentionally from the above derived data to showcases an alternative approach to obtaining essential information directly as a table itself without visualizing, especially when dealing with large number of values to avoid clutter and enhance clarity.

- Total bet amount per player
(DESCENDING ORDER)

Src_PLAYER_ID	
12346653.0	1.464642e+08
13410752.0	1.329271e+08
12655429.0	1.016396e+08
12996364.0	1.014185e+08
13781394.0	1.007991e+08
...	
12347439.0	0.000000e+00
14870785.0	0.000000e+00
12423049.0	0.000000e+00
16446757.0	0.000000e+00
12368770.0	0.000000e+00

- Total win amount per player
(DESCENDING ORDER)

Src_PLAYER_ID	
12346653.0	1.402861e+08
13410752.0	1.285004e+08
12655429.0	1.138036e+08
12996364.0	9.696702e+07
13781394.0	9.086329e+07
...	
18560156.0	0.000000e+00
18560772.0	0.000000e+00
18563985.0	0.000000e+00
12745145.0	0.000000e+00
20696055.0	0.000000e+00

- Player activity per month

ActivityMonth_x	
2019-01-31	1296
2019-02-28	2149
2019-03-31	2635
2019-04-30	3201
2019-05-31	3261
2019-06-30	2829
2019-07-31	2896
2019-08-31	3403
2019-09-30	3808
2019-10-31	4618
2019-11-30	5346
2019-12-31	6016
2020-01-31	6384
2020-02-29	6792
2020-03-31	6435
2020-04-30	4917
2020-05-31	5373
2020-06-30	6559
2020-07-31	7036
2020-08-31	7042
2020-09-30	7407
2020-10-31	8604
2020-11-30	8997
2020-12-31	9241
2021-01-31	9908
2021-02-28	10199
2021-03-31	11004
2021-04-30	11225
2021-05-31	10607
2021-06-30	9336

- Acquisition channel distribution

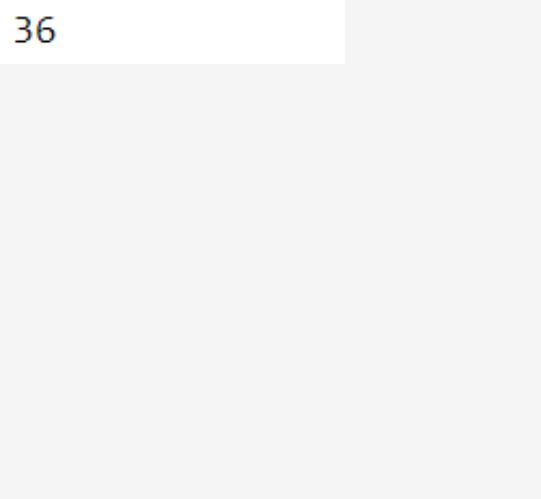
PPC	1124335
Affiliate	581673
Others	175362
Retail	74828
Social	53253
Display	11164
Programmatic	16
Email	3

- Average bet amount by gender

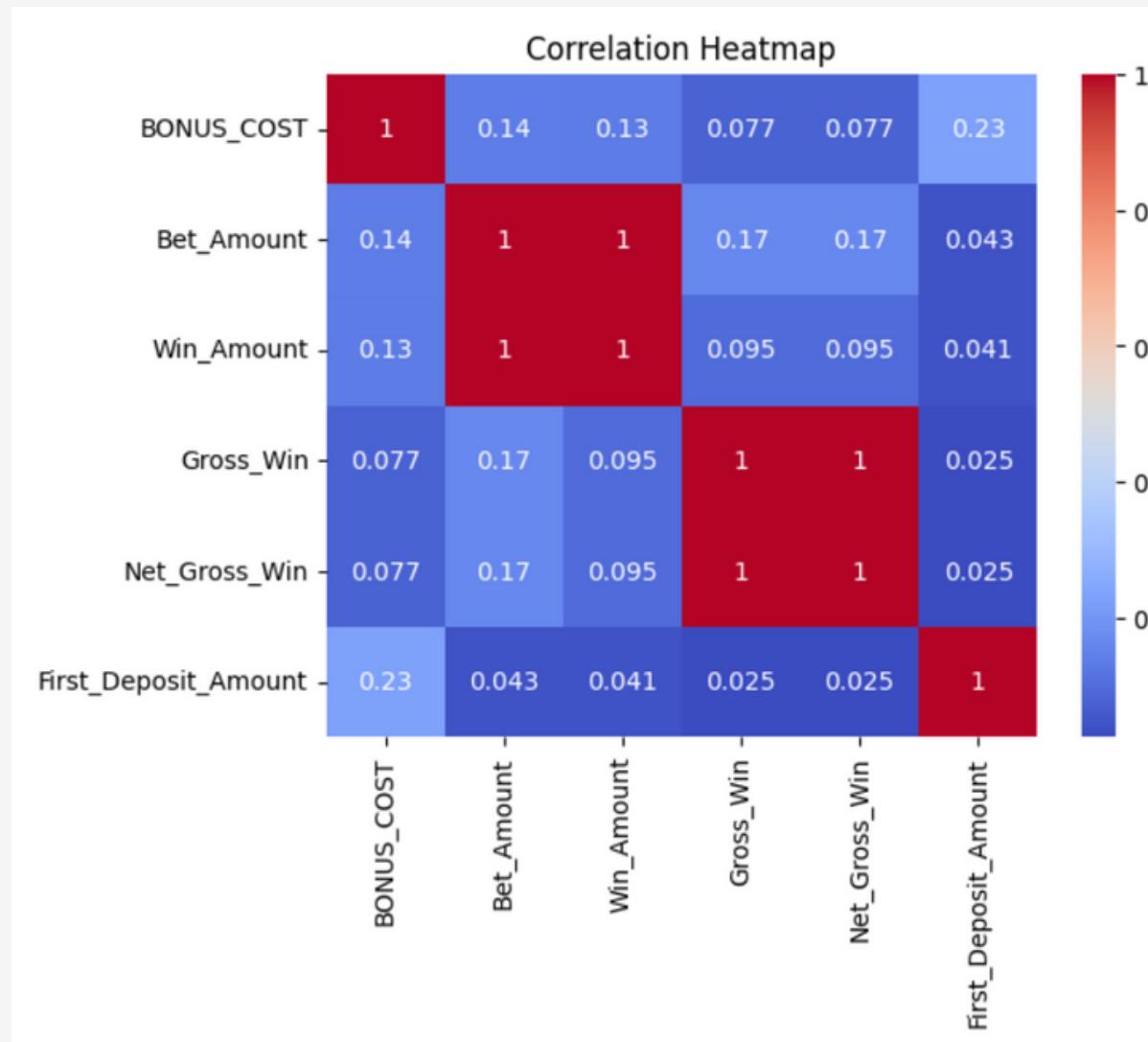
Gender	
F	6603.015914
M	5287.293176
Name:	Bet_Amount, dt

INSIGHTS : MERGED DATASET

- Average age of players



- Correlation matrix (HEAT MAP)

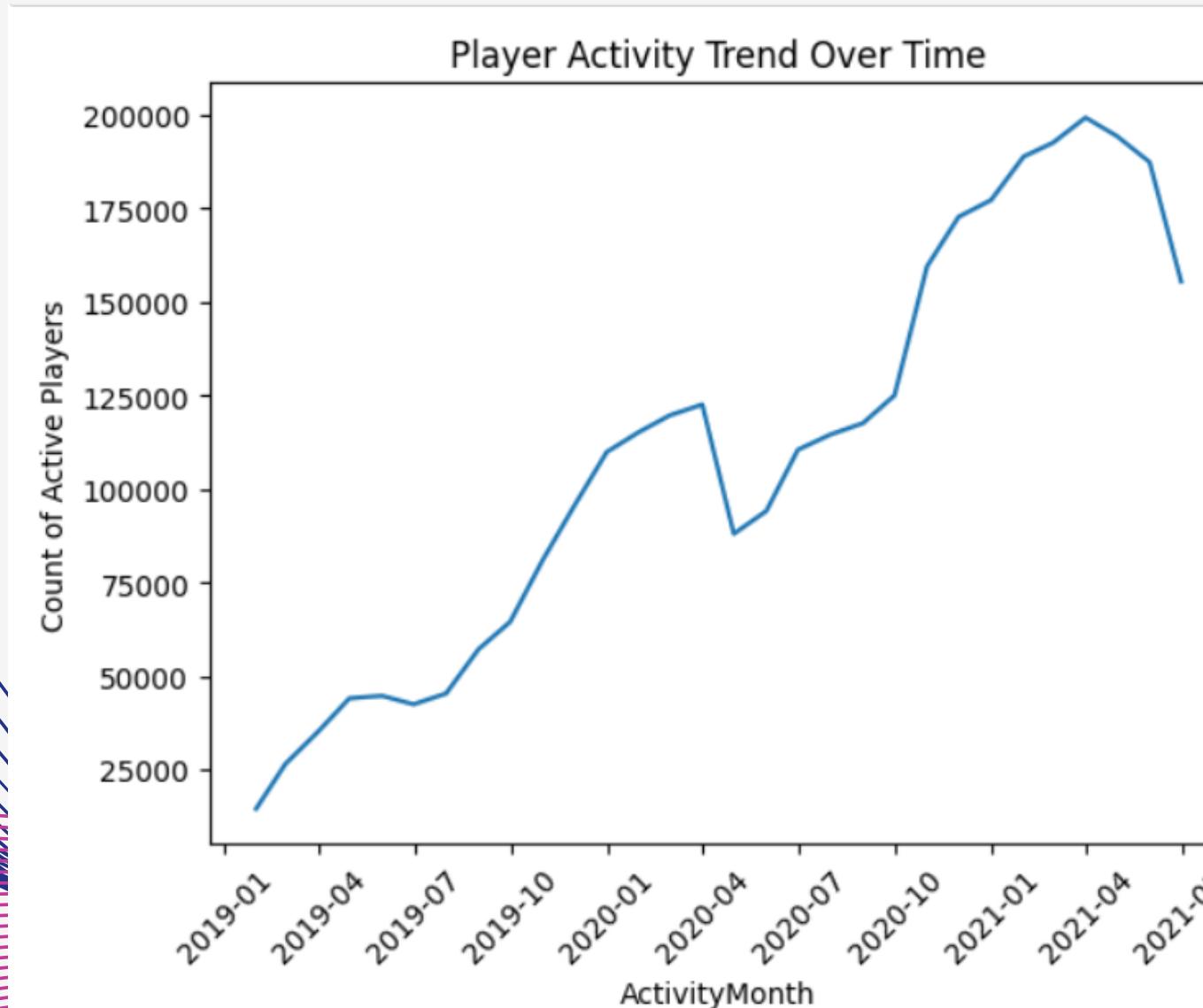


- Top players by net gross win

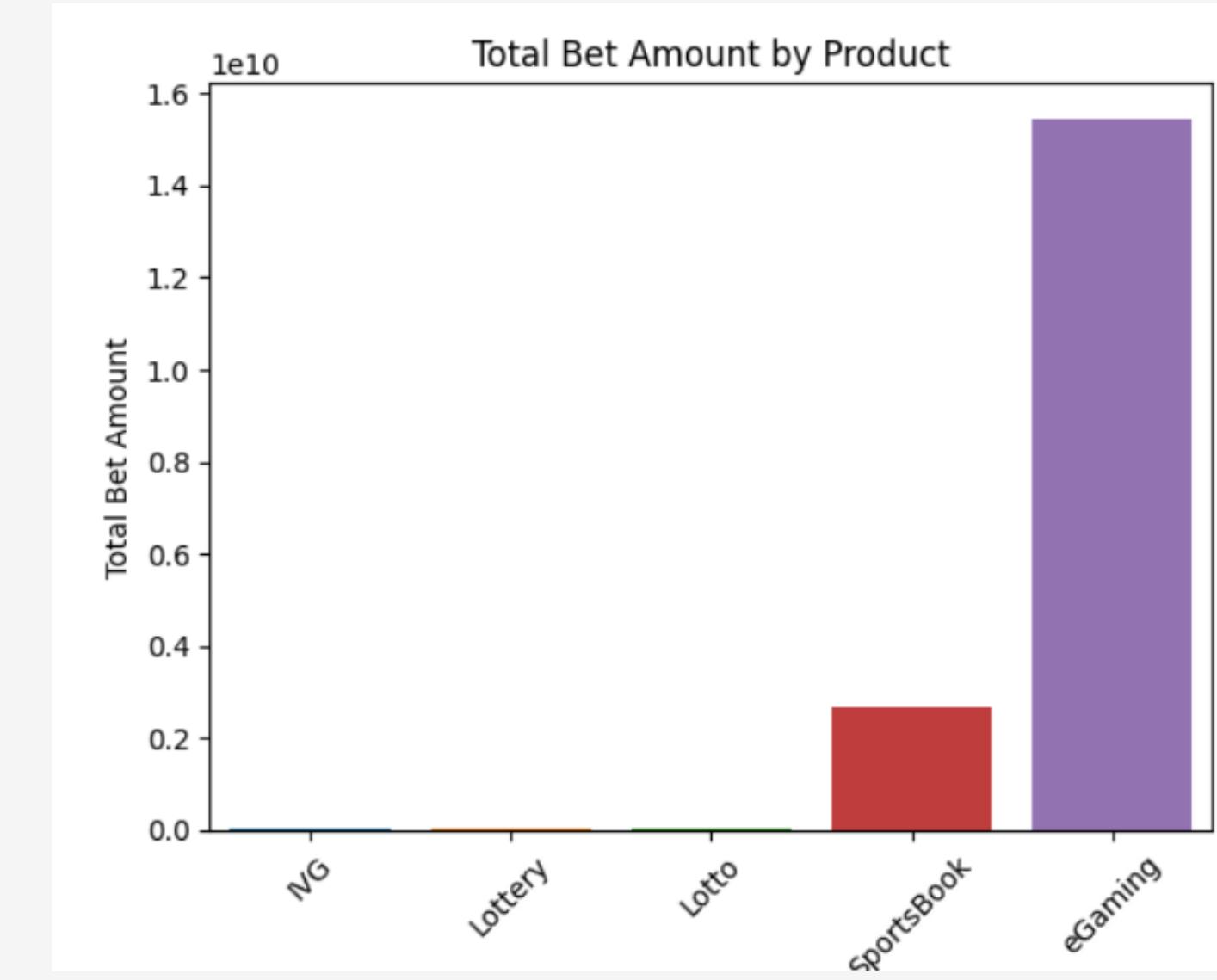
Src_Player_Id_player_details	Net_Gross_Win
2459839	20077064.000000
2459840	20077064.000000
799342	nan
799343	nan
3217697	13781394.000000
3217698	13781394.000000
3217699	13781394.000000
3217700	13781394.000000
3217701	13781394.000000
3217702	13781394.000000

INSIGHTS : MERGED DATASET

- Player activity trend

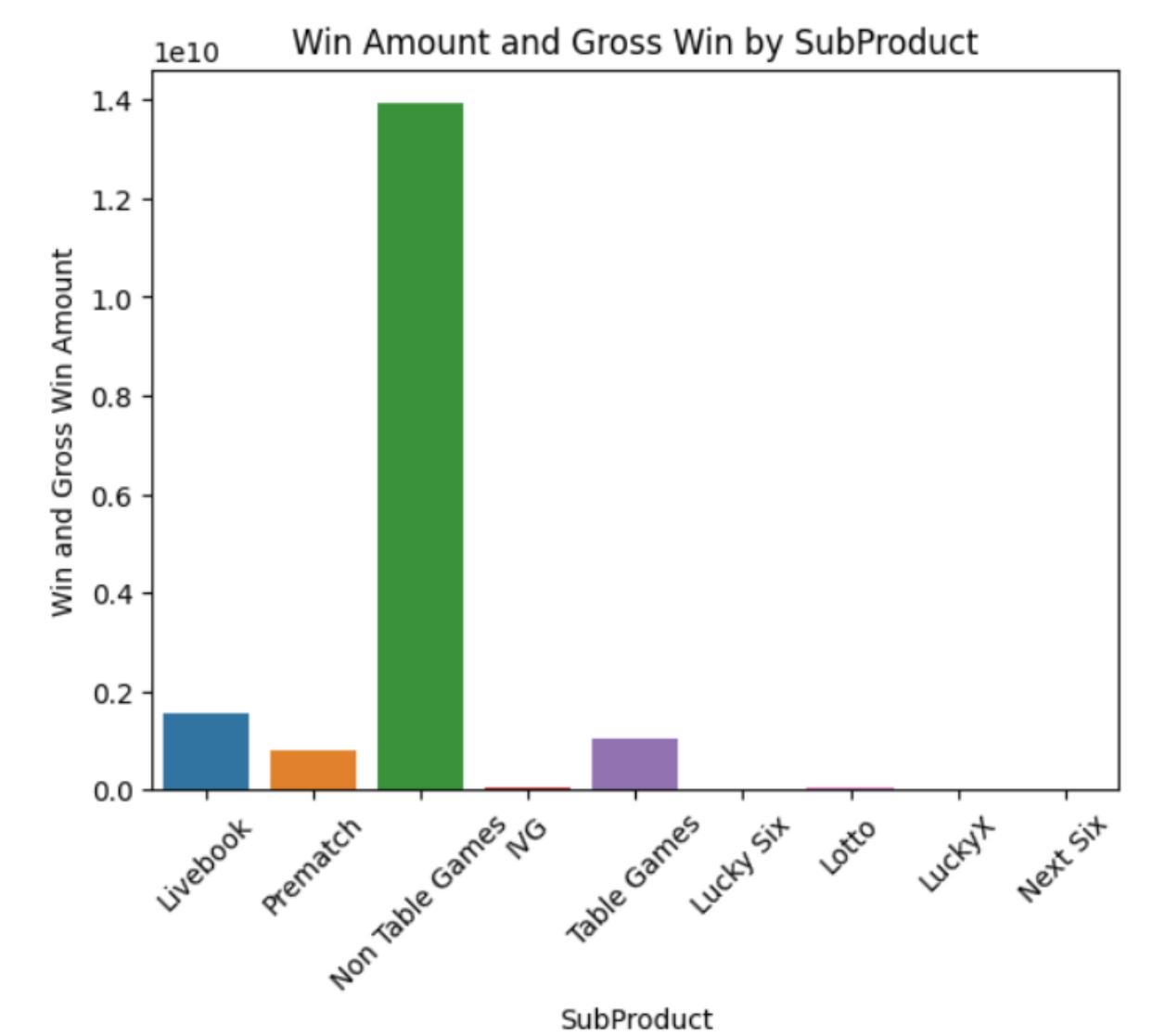


- Total bet amount by product

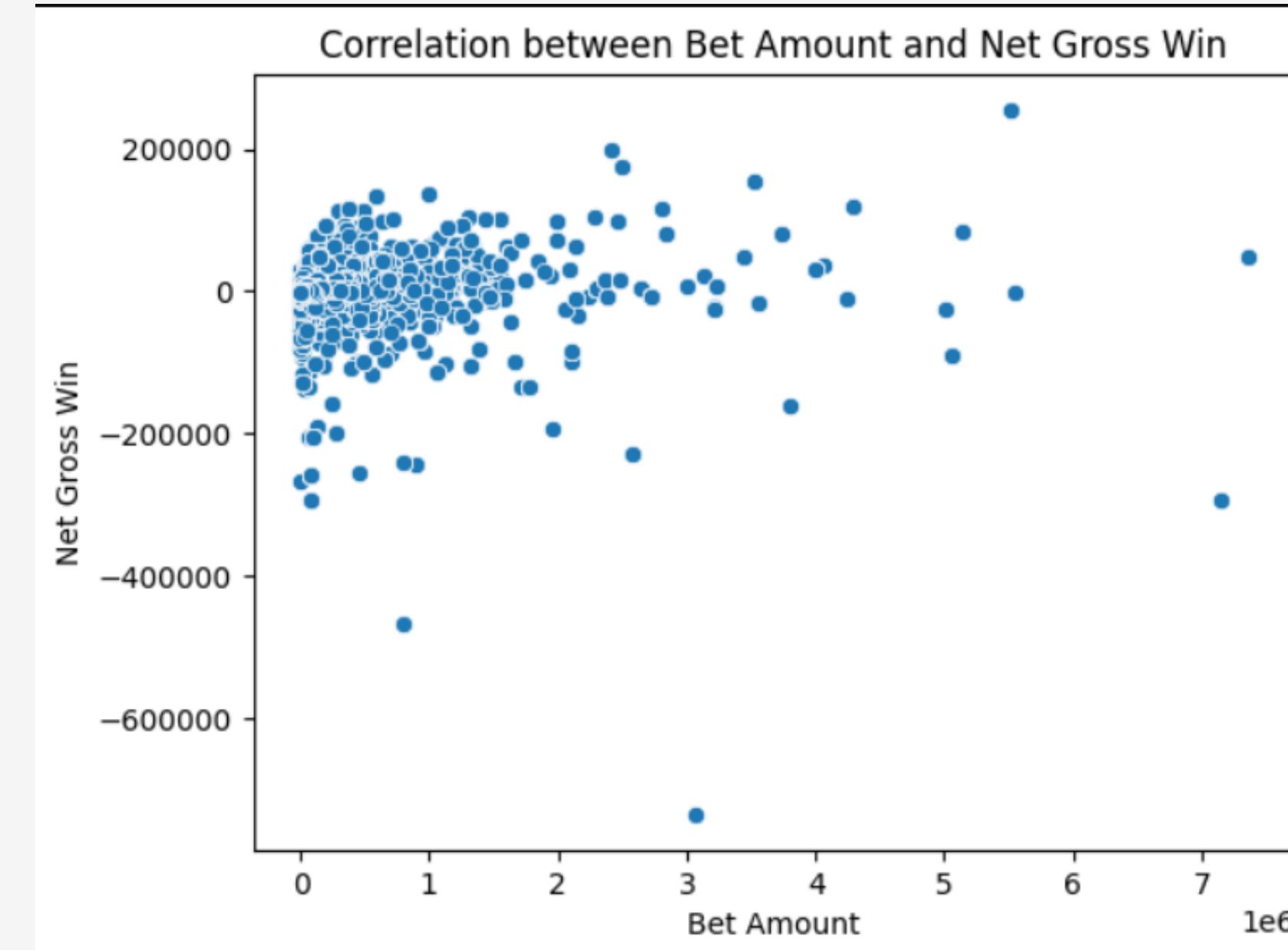


INSIGHTS : MERGED DATASET

- Win amount and gross win amount by sub product

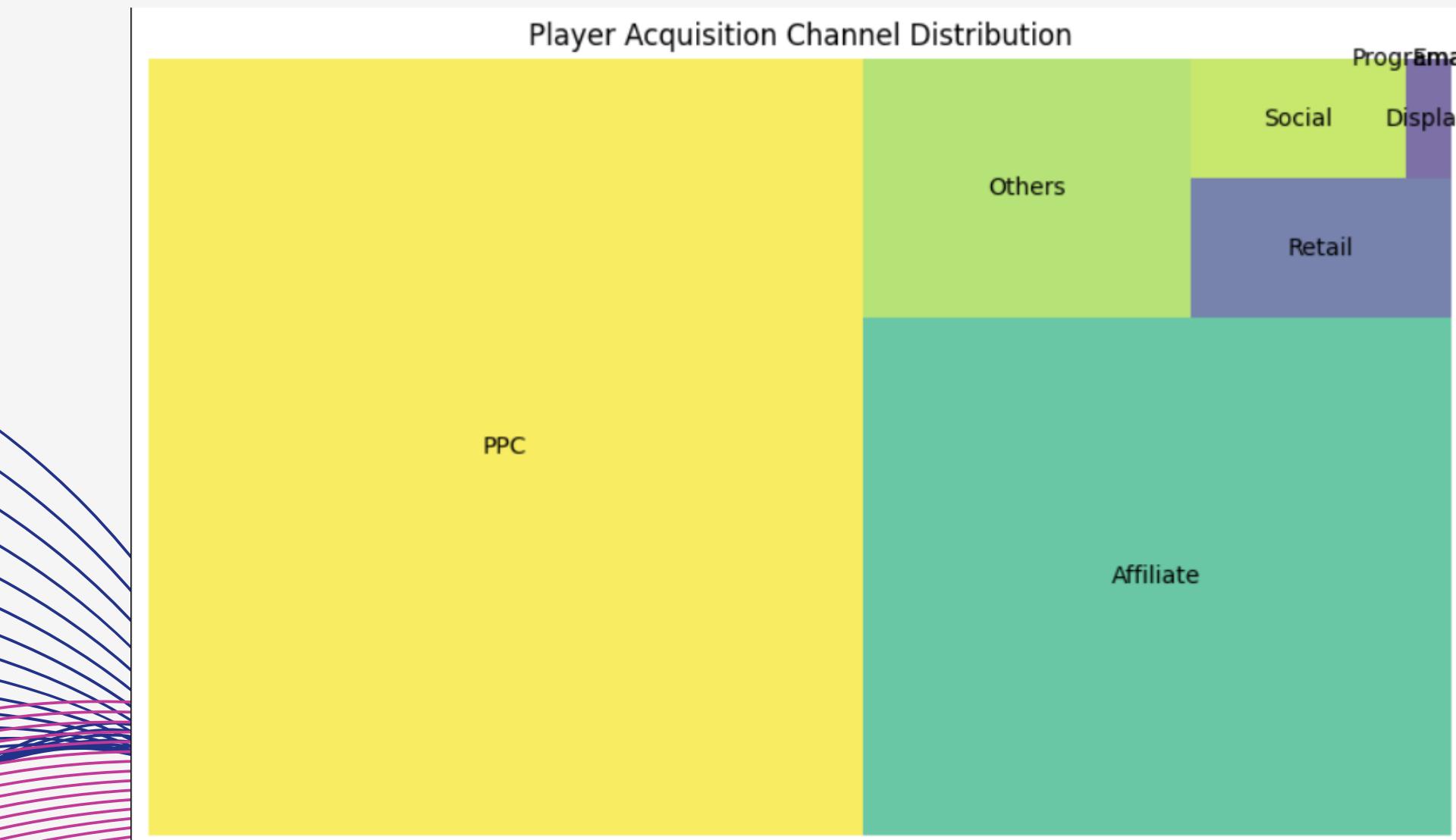


- Correlation between Bet amount and Net gross win

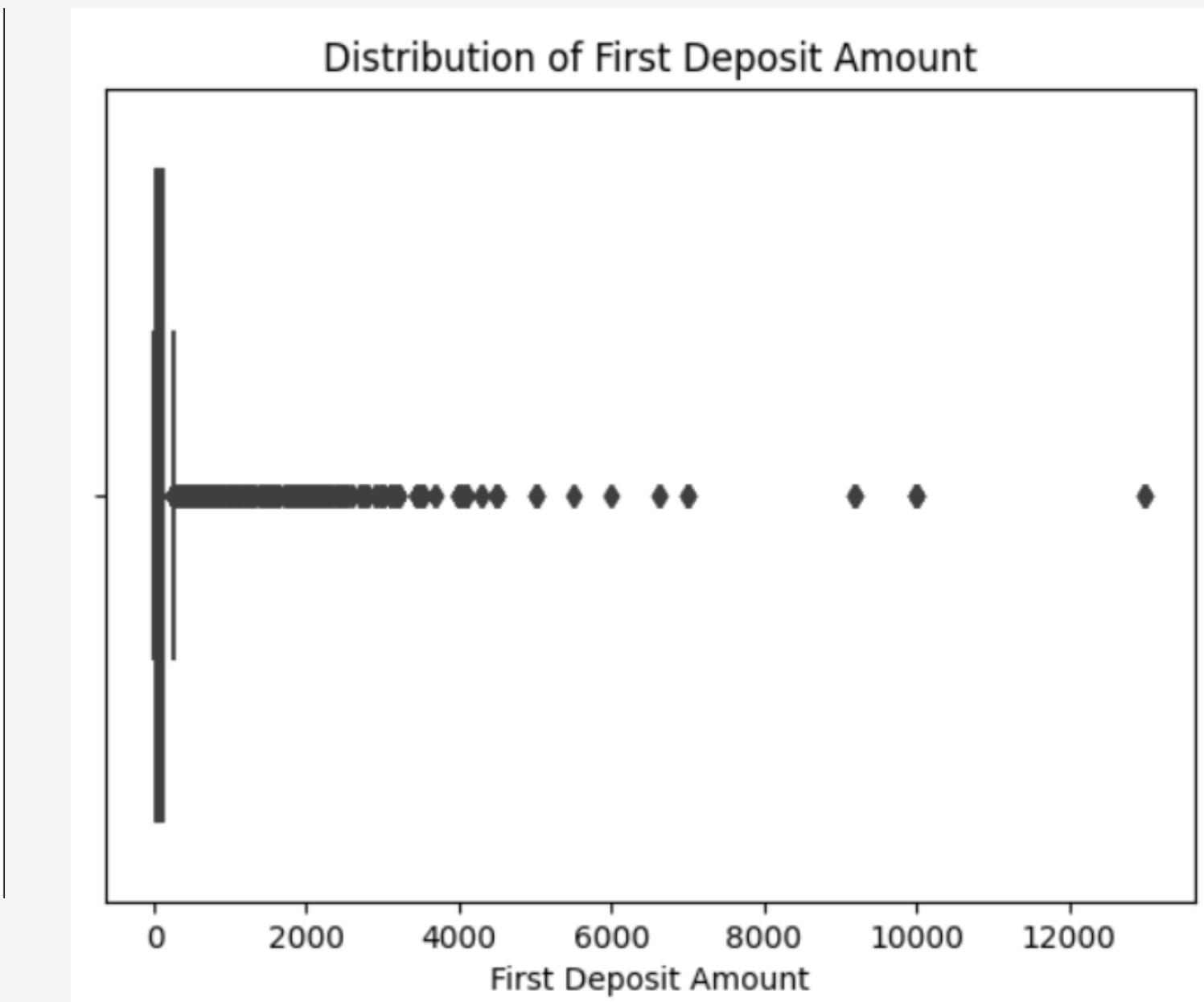


INSIGHTS : MERGED DATASET

- Player Acquisition by tree map

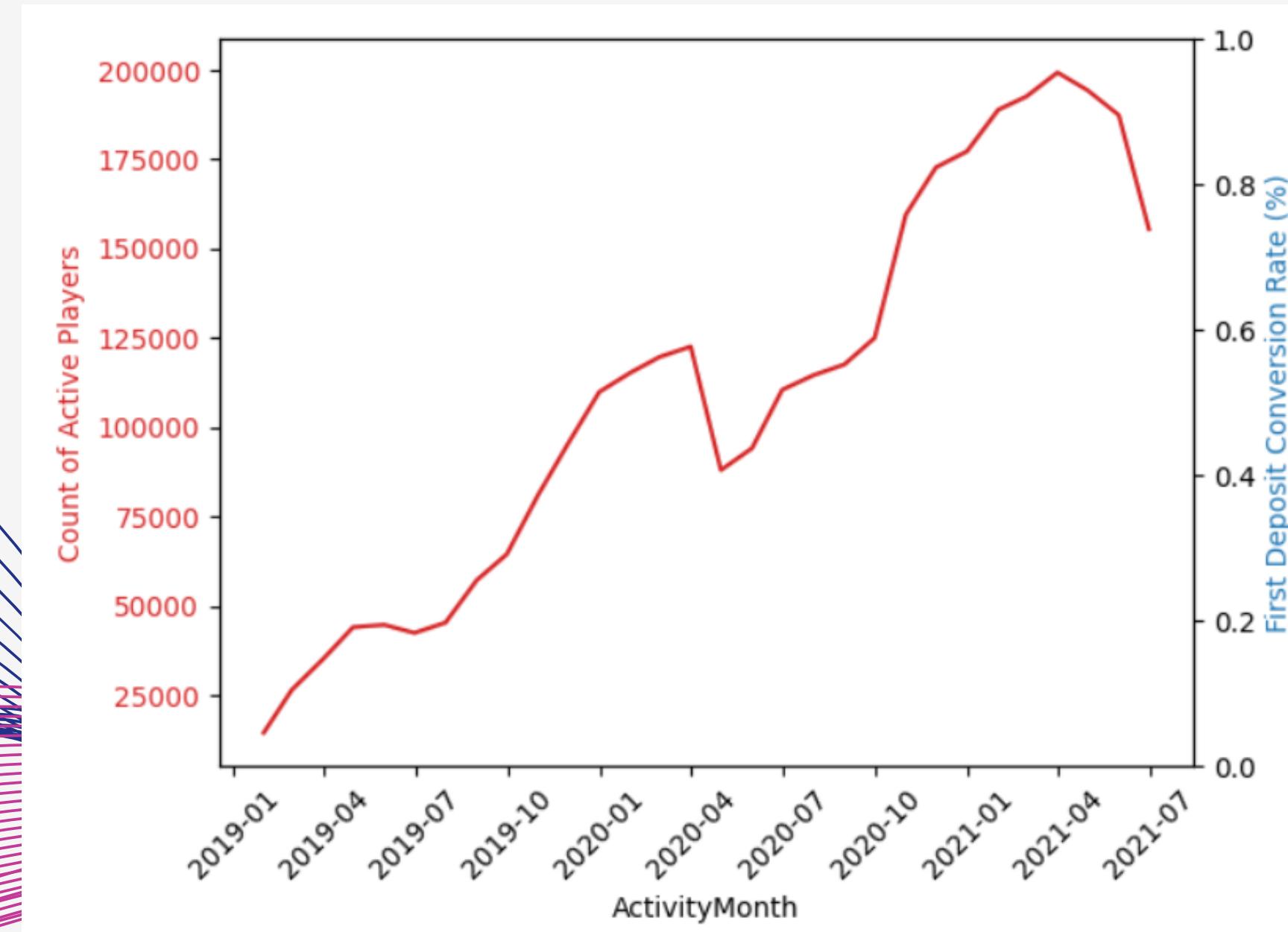


- Distribution of 1st deposit amount

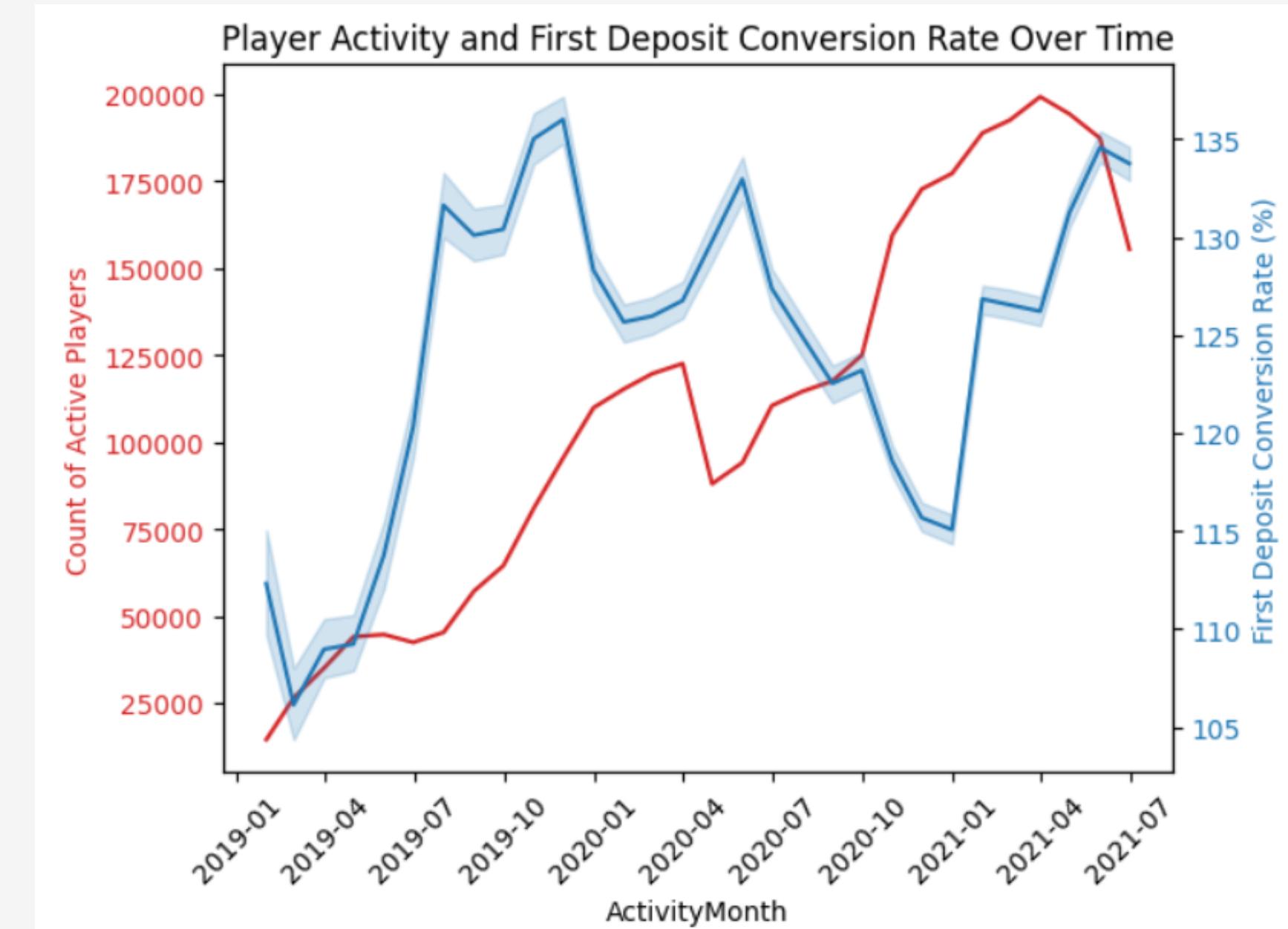


INSIGHTS : MERGED DATASET

- Active players throughout the month



- Player activity and deposit amount over time



Thank You

I hope you found this presentation insightful and valuable.

I am grateful for your time and consideration

Once Again Thank You !

Note :

Tools used : Power Bi and Python

No. of plots : 50