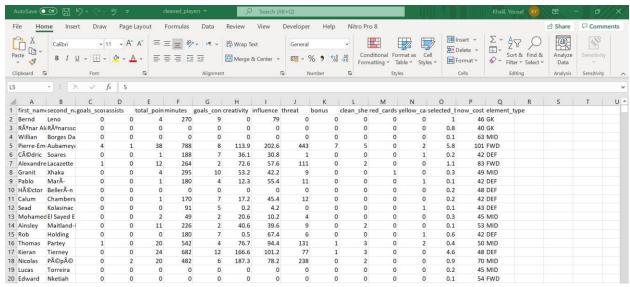
### Introduction

Importing a CSV file into psql is an important procedure that allows the user to extract meaningful insights from spreadsheets. Although excel has a Graphical User Interface (GUI) and is more user-friendly, it fails to handle large CSV files and big data. This is where SQL comes in handy with its ability to handle large files.

This report will explore how to load a CSV file into psql in command line, to perform queries and subqueries on the data, and to export the results of a query (a table) into a new CSV file. The chosen CSV file is a fantasy football player dataset for the 2021/2022 Barclays Premier League season. It contains player statistics such as goals, assists, creativity, and others. The aim of this report will be to extract meaningful insights from the data using queries and subqueries to advise the user on the best players to bug and bargains.

### **Question 1**

We use **psql** -U **postgres** and type in the password to log into the local database.



Fantasy Premier League Players 21/22

We choose **cleaned\_players** (source: <u>github</u>), a database of all players in the 2021/2022 Barclays Premier League season who are available for selection in the Fantasy Premier League game. Fantasy Football is an online game where gamers have a budget to choose a fantasy squad of Barclays Premier League players based on how well they think these players will perform. If the players they pick, perform well the gamers will gain points and move up in the rankings. The csv consists of the following columns:

first\_name: first name of player

**second\_name**: family name

goals\_scored: goals scored so far

assists: goal assists provided

total\_points: number of points gained by player in fantasy across all matches played

minutes: minutes played on the pitch

goals\_conceded: number of goals player's team has conceded

**creativity**: rating of player's ability to create chances

**influence**: rating of player's ability to individually influence the game and its outcome

threat: rating of player's ability to cause trouble for the opposition in offence

bonus: accumulated bonus points

**clean\_sheets**: number of matches player's team has completed without conceding any goals

red\_cards: number of red cards player has received

yellow\_cards: number of yellow cards player has received

**selected\_by\_percent**: percentage of fantasy gamers who have picked the player in their squad

now\_cost: current price in fantasy

element\_type: position of player on the field

## **Question 3**

```
Type "Nelp" for Nelp.

**Command Prompt - psql - U postgres

**Circosoft Windows [Version 18.8, 1943, 1348]
(c) Microsoft Corporation. All rights reserved.

**Ci-\Users\Lenovopsql - U postgres

**Password for user postgres:

**pagl (14.1)

**ARNING: Console code page (437) differs from Windows code page (1252)

**B-bit characters might not work correctly. See psql reference

**page "Notes for Windows users" for details.

**Type "Nelp" for Nelp.

**Ostgress#* CREATE DATABASE bpl_fantasy;

**CREATE DATABASE bpl_fantasy;

**CREATE DATABASE postgress#**

**Password for Users for Windows users for details.

**Password for user postgress for Windows for Users for Windows f
```

We use the **CREATE DATABASE** command followed by the name of choice for the database. In this case, the chosen name of the database is **bpl\_fantasy** for relevance.

# **Question 4**

We use the "\c" command followed by **bpl\_fantasy** in command line to connect to the database.

```
CREATE TABLE BPLFantasy (
1 CREATE TABLE BPLFantasy (
2 first_name text,
 3 second name text,
4 goals scored int,
5 assists int,
6 total_points int,
7 minutes int,
8 goals_conceded int,
9 creativity real,
10 influence real,
11 threat real,
12 bonus int,
13 clean_sheets int,
14 red_cards int,
15 yellow cards int,
16 selected_by_percent real,
   now_cost int,
17
18
   element_type text
19
    );
```

We use **CREATE TABLE** to create the table **BPLFantasy** 

The columns which are of the **integer** datatype are allocated the **int** datatype (an alias for integer). These are take on exact numerical values (no decimals).

The columns which are **floats** are allocated the **real** datatype. Floats may be decimals.

**Strings** are allocated the **text** datatype.

```
bpl_fantasy=#
bpl fantasy=# CREATE TABLE BPLFantasy (
bpl_fantasy(# first_name text,
bpl_fantasy(# second_name text,
bpl_fantasy(# goals_scored int,
bpl_fantasy(# assists int,
bpl fantasy(# total points int,
bpl_fantasy(# minutes int,
bpl fantasy(# goals conceded int,
bpl_fantasy(# creativity real,
bpl fantasy(# influence real,
bpl_fantasy(# threat real,
bpl fantasy(# bonus int,
bpl_fantasy(# clean_sheets int,
bpl fantasy(# red cards int,
bpl_fantasy(# yellow_cards int,
bpl fantasy(# selected by percent real,
bpl_fantasy(# now_cost int,
bpl_fantasy(# element_type text
bpl_fantasy(# );
CREATE TABLE
bpl_fantasy=#
bpl fantasy=# \dt
           List of relations
             Name
                      Type
                                Owner
 public | bplfantasy | table | postgres
1 row)
bpl_fantasy=#
```

We run the **CREATE TABLE** command to create the table and define its columns and their datatypes. Then, we use "\dt" to check that the table was created.

```
cpl_fantasy=#
tpl_fantasy=# \copy BPLFantasy FROM 'C:\Users\Lenovo\Desktop\New folder\Fantasy-Premier-League\data\2021-22\cleaned_players.csv'DELIMITER','CSV HEADER;
COPY 623
bpl_fantasy=#
```

Only "\copy" worked in this case. "\copy" should be followed by the name of the table ("BPLFantasy") that we created in the bpl\_fantasy database. We then include the FROM phrase and specify the directory of the csv file after. We finally use DELIMITER ',' CSV HEADER to designate that the imported file is a csv file.

## **Question 8**



We use **SELECT \* FROM BPLFantasy** to display the entire BPLFantasy table.

### **Question 9**

#### Query 1:

The purpose of this query is to return the number of players in each field position and rank each position by its player count (to find the most abundant field positions).

To accomplish this, we select the **element\_type** (field position) column and **COUNT(\*)** as our aggregate function because the aim is to count the number of players in each element type. We rename the count column as **player\_count**. We also use **GROUP BY** to group the records by element\_type. Finally, to rank the positions by player count, we use **ORDER BY** and limit the result to 20 rows.

```
31 SELECT element_type, COUNT(*)
32 AS player_count
33 FROM BPLFantasy
34 GROUP BY element_type
35 ORDER BY player_count DESC;
```

Query 1

Results Query 1

#### Query 2:

The purpose of this query is to select the top 20 outfield players in the Premier League who have earned the most points so far in the season. If two players have the same total points, their goal record will be compared and the player with the greatest number of goals will be ranked above in the list.

To accomplish this, we use the **SELECT** statement to return the relevant columns. We use **WHERE** to filter the rows to only outfield players. Then, we use **ORDER BY** to rank the rows in descending order by total points and goals scored, and **limit** to 20 records.

```
25 SELECT DISTINCT first_name,
    second_name, total_points,
    goals_scored
26 FROM BPLFantasy
27 WHERE element_type != 'GK'
28 ORDER BY total_points DESC,
    goals_scored DESC
29 LIMIT 20;
```

Query 2

```
bpl_fantasy=# SELECT DISTINCT first_name, second_name, total_points, goals_scored
bpl_fantasy-# FROM BPLFantasy
opl_fantasy-# WHERE element_type != 'GK'
opl fantasy-# ORDER BY total points DESC, goals scored DESC
opl_fantasy-# LIMIT 20;
    first_name | second_name | total_points | goals_scored
Mohamed
                    Salah
                                                117
                    Cancelo
                    Alexander-Arnold
Trent
Conor
                    Gallagher
Michail
                     Vardy
Sadio
Heung-Min
Emile
                     Smith Rowe
                     Benrahma
Jarrod
                     R세diger
Bruno Miguel
                     Borges Fernandes
Andros
Pablo
                     Fornals
                     Trossard
Leandro
                     Tielemans
Youri
                    Dias Belloli
Raphael
Gabriel Fernando
                   de Jesus
(20 rows)
```

Results Query 2

#### **Query 3:**

The purpose of this query is to return the first name, last name, element type, individual influence score, and average score partitioned by **element\_type**, of each player, and rank the players according to their individual influence. This is useful to compare how the individual influence scores of the most influential players compare to the average in their field positions.

To accomplish this, we use the **SELECT** statement to select the relevant columns. At the end of the SELECT statement, we add the **avg(influence)** before the **OVER** window function. **avg(influence)** will be calculated for each **element\_type** since we partitioned by **element\_type**. Finally, we rank the rows by the individual influence of each player using **ORDER BY**.

```
38 SELECT DISTINCT first_name,
    second_name, element_type,
    influence,
39 avg(influence) OVER(PARTITION
    BY element_type)
40 FROM BPLFantasy
41 ORDER BY influence DESC
42 LIMIT 20;
43
```

Query 3

```
ppl fantasy=# SELECT DISTINCT first name, second name, element_type, influence,
opl fantasy-# avg(influence) OVER(PARTITION BY element type)
ppl_fantasy-# FROM BPLFantasy
pl fantasy-# LIMIT 20;
                 second name
                                 element type | influence
 first name
Mohamed
                                                               74.87843140620811
Youri
                                                               74.87843140620811
Bruno Miguel
               Borges Fernandes
                                                               74.87843140620811
               Krul
                                                       318.4
                                                               69.55211278082619
Michail
                                   FWD
                                                               66.14651135078003
Declan
                                                       308.6
                                                               74.87843140620811
Illan
                                                       307.8
                                                               69.55211278082619
Michael
                                                       306.4
                                                               84.91658742768222
Conor
               Gallagher
                                                       300.4
                                                               74.87843140620811
                                                       298.4
                                                               69.55211278082619
Trent
               Alexander-Arnold
                                                       294.4
                                                               84.91658742768222
                                                               74.87843140620811
Andros
               Townsend
               Dias Belloli
                                                       281.8
                                                               74.87843140620811
                                                       280.8
                                                               74.87843140620811
               Malheiro de S-í
                                                               69.55211278082619
               Vardy
                                   FWD
                                                               66.14651135078003
               Tarkowski
                                                       276.6
                                                               84.91658742768222
Emiliano
               Mart | nez
                                                         268
                                                               69.55211278082619
Jarrod
                                                         268
                                                               74.87843140620811
Pablo
               Fornals
                                                       267.2
                                                               74.87843140620811
20 rows)
```

Results Query 3

#### Query 4:

The aim of this query is to find the 20 most lethal forwards in the league (highest threat) and compare their threat scores to the average threat score of all forwards in the league.

To accomplish this, we first creat the CTE **avg\_threats** which groups all the records by element\_type (position) and computes the average threat of each position. Using this CTE, we creat another CTE **fwd\_avg\_threats** which selects only the row belonging to the element\_type **FWD.** Then for the main query, we select the first name, second name, individual threat level, and average threat level for forwards from the CTE **fwd\_avg\_threats.** We filter the rows to display only forwards since we only want to compare threat levels between forwards and not other positions. We also filter them to display only records with threat scores above the average for forwards (using the fwd\_avg\_threat CTE). We finally order the rows by descending threat level and limited the rows to 20. This result could not have been accomplished without a subquery.

```
44
   WITH avg_threats AS
45
46
   SELECT element_type, AVG(threat)
47 AS avg threat
48 FROM BPLFantasy
49 GROUP BY element type
   ), fwd_avg_threat AS
51
   (
52
   SELECT avg_threat
   FROM avg threats
   WHERE element_type = 'FWD')
55
56
57 SELECT DISTINCT first name, second name, threat, (SELECT *
58 FROM fwd avg threat)
59 FROM BPLFantasy
60 WHERE element_type = 'FWD'
61 AND threat > (SELECT *
62 FROM fwd_avg_threat)
63 ORDER BY threat DESC
64 LIMIT 20;
```

#### Query 4

```
bpl_fantasy=# WITH avg_threats AS
bpl_fantasy(# SELECT element_type, AVG(threat)
bpl_fantasy(# Avg_threat
bpl_fantasy(# FROM BPLFantasy
bpl_fantasy(# FROM BPLFantasy
bpl_fantasy(# GROUP BY element_type
bpl_fantasy(# ), fwd_avg_threat AS
bpl_fantasy(# ), fwd_avg_threat AS
bpl_fantasy(# SELECT avg_threat
bpl_fantasy(# SELECT avg_threat
bpl_fantasy(# WHERE element_type = 'FWD')
bpl_fantasy(# FROM avg_threats
bpl_fantasy-# FROM fwd_avg_threat)
bpl_fantasy-# SELECT DISTINCT first_name, second_name, threat, (SELECT *
bpl_fantasy-# FROM BPLFantasy
bpl_fantasy-# FROM BPLFantasy
bpl_fantasy-# FROM Fwd_avg_threat)
bpl_fantasy-# FROM fwd_avg_threat
bpl_
```

Results Query 4

#### Query 5:

This query is very useful for any gamer who is looking to add a great midfielder to their squad for a bargain. It returns all midfielders who cost lower than the average cost of a midfielder, but who have at the same time produced more points than the average for a midfielder. It also ranks these players by decreasing total points and decreasing creativity (an important indicator for a midfielder). This is very good example of how a gamer can leverage this data to gain an advantage in fantasy football.

To accomplish this, (just like Query 4) we first create a CTE for the average of the total points (average\_points) grouped by element\_type. Then using this CTE, we create a new CTE mid\_avg\_points to retrieve the average total points for a midfielder. Then, we do the same procedure to find the average price, and the average price for a midfielder. Finally, in the main query, we select the first name, last name, creativity score, price, and total points columns. We filter the rows with the WHERE statement to retrieve only midfielders, who cost less than the average (using mid\_avg\_price CTE), and who produced more points than the average (using mid\_avg\_points CTE). Finally, we use ORDER BY to order in descending order by the number of points then the creativity score.

```
WITH average_points AS
      SELECT element_type, AVG(total_points)
      AS avg_points
      FROM BPLFantasy
      GROUP BY element_type
      HAVING element_type = 'MID'
      ), mid_avg_points AS
      SELECT avg_points
      FROM average_points
      WHERE element_type = 'MID'),
      average price AS
      SELECT element_type, AVG(now_cost)
      AS avg_price
      FROM BPLFantasy
      GROUP BY element_type
      HAVING element_type = 'MID'
      ), mid_avg_price AS
      SELECT avg_price
      FROM average_price
      WHERE element type = 'MID')
      SELECT DISTINCT first name, second name, creativity, now cost, total points
102
      FROM BPLFantasy
      WHERE element_type = 'MID'
      AND now_cost < (SELECT * FROM mid_avg_price)
104
      AND total_points > (SELECT * FROM mid_avg_points)
      ORDER BY total_points DESC, creativity DESC
      LIMIT 20:
```

```
fantasy=# WITH average_points AS
opl_fantasy(# SELECT element_type, AVG(total_points)
opl_fantasy(# AS avg_points
opl_fantasy(# FROM BPLFantasy
pl_fantasy(# ), mid_avg_points AS
opl_fantasy(# SELECT avg_points
ppl_fantasy(# FROM average_points
opl_fantasy(# WHERE element_type = 'MID'),
ppl_fantasy-# average_price AS
ppl_fantasy(# AS avg_price
ppl_fantasy(# FROM BPLFantasy
ppl_fantasy(# GROUP BY element_type
opl_fantasy(# HAVING element_type = 'MID'
ppl_fantasy(# ), mid_avg_price AS
ppl_fantasy(# SELECT avg_price
ppl_fantasy(# FROM average_price
opl_fantasy(# WHERE element_type = 'MID')
opl_fantasy-#
pl_fantasy-# SELECT DISTINCT first_name, second_name, creativity, now_cost, total_points
opl_fantasy-# WHERE element_type = 'MID'
opl_fantasy-# AND now_cost < (SELECT * FROM mid_avg_price)
pl_fantasy-# AND total_points > (SELECT * FROM mid_avg_points)
ppl_fantasy-# ORDER BY total_points DESC, creativity DESC
                                           | creativity | now_cost | total_points
```

#### Results Query 5 P1

Pierre-Emile	H qjbjerg	110.4	49	36
James	McArthur	156.9	45	33
Allan	Marques Loureiro	127.3	46	32
Christian	N≒rgaard	79.2	50	31
Naby	Keita	72.5	50	30
Ashley	Westwood	229.5	53	28
Stuart	Dallas	184.7	51	27
Alexis	Mac Allister	58.3	54	27
R-  ben Diogo	da Silva Neves	196.7	54	25
Douglas Luiz	Soares de Paulo	176.1	46	25
Mateusz	Klich	164.2	54	25
Vitaly	Janelt	115.2	50	25
Jo-úo Filipe Iria	Santos Moutinho	238.8	50	24
Mathias	Normann	93.5	45	24
Isaac	Hayden	32.1	45	24
Adam	Lallana	91.8	54	23
(20 rows)				

Results Query 5 P2

This query returns a table of the top 20 goalkeepers with most clean sheets and goals conceded (descending order), who have played more minutes than the average for the league.

To accomplish this, we first use the **SELECT** statement to filter the columns we want which are first name, last name, goals conceded, and clean sheets. We use the **WHERE** statement to filter the data to only goalkeepers who have played more minutes than the average (used a subquery). Then, we use **ORDER BY** to order the rows by clean sheets and goals conceded and limited the table to 20 rows.

To create the table that stores the result of the above query, we use **CREATE TABLE** followed by the name of the table I chose (**top\_gks**). Then, we copy the query into parentheses after **AS**. This means that the result of the query is stored in the new table.

```
112
113    SELECT DISTINCT first_name, second_name,
        goals_conceded, clean_sheets
114    FROM BPLFantasy
115    WHERE element_type = 'GK' AND minutes > (SELECT AVG(
        minutes) FROM BPLFantasy)
116    ORDER BY clean_sheets DESC, goals_conceded DESC
117    LIMIT 20;
118
```

```
ppl_fantasy=# SELECT DISTINCT first_name, second_name, goals_conceded, clean_sheets
bpl_fantasy+# FROM BPLFantasy
bpl_fantasy+# WHERE element_type = 'GK' AND minutes > (SELECT AVG(minutes) FROM BPLFantasy)
bpl_fantasy-# UMERE Probable P
```

```
pl_fantasy=# CREATE TABLE top_gks
pl_fantasy-# AS (SELECT DISTINCT first_name, second_name, goals_conceded, clean_sheets
ppl_fantasy(# FROM BPLFantasy
opl_fantasy(# WHERE element_type = 'GK' AND minutes > (SELECT AVG(minutes) FROM BPLFantasy)
opl_fantasy(# ORDER BY clean_sheets DESC, goals_conceded ASC
opl fantasy(# LIMIT 20);
SELECT 20
opl_fantasy=# SELECT * FROM top_gks;
first_name | second_name | goals_conceded | clean_sheets
Ederson
Aaron
            Ramses Becker
            McCarthy
Robert
             Lloris
             Raya Martin
Jordan
            Mart | nez
            de Gea
            Krul
Illan
Ben
20 rows)
```

We use **COPY** to copy the table **top\_gks** from Question 10 to the new csv file **top\_gks.csv**. Then, we specify the **delimiter** as a **comma** and indicated that the first row is a header row.

We then open the CSV in Visual Studio Code.

```
bpl_fantasy=#
bpl_fantasy=# \COPY top_gks TO 'C:\Users\Lenovo\Desktop\New folder\Fantasy-Premier-League\data\2021-22\top_gks.csv' DELIMITER ',' CSV HEADER;
COPY 20
bpl_fantasy=#
```

```
top_gks.csv - Visua
Trust this window to enable all features. Manage
                                                                                  Learn More
       Queries HW4
                           ■ Untitled-1
                                           ■ top_gks.csv X
       C: > Users > Lenovo > Desktop > New folder > Fantasy-Premier-League > data > 2021-22 > 📕 top_gks.csv
              first_name, second_name, goals_conceded, clean_sheets
              Edouard, Mendy, 4, 6
              Ederson, Santana de Moraes, 6, 6
              Aaron, Ramsdale, 4,5
              Alisson, Ramses Becker, 11,5
              Alex, McCarthy, 12,5
              Robert, Sánchez, 12,4
              Vicente, Guaita, 14,4
品
              Hugo, Lloris, 16,4
              David, Raya Martin, 9, 3
              Jordan, Pickford, 13,3
              Lukasz, Fabianski, 13,3
              Emiliano, Martínez, 17,3
              José, Malheiro de Sá, 12, 2
              David, de Gea, 17, 2
              Tim, Krul, 26, 2
              Nick, Pope, 17,1
              Illan, Meslier, 18,1
              Kasper, Schmeichel, 18,1
              Daniel, Bachmann, 7,0
              Ben, Foster, 12,0
```

### **Conclusion**

To sum up this report, we first created a new local database **bpl\_fantasy** using the CREATE DATABASE function. To later copy the CSV into the database as a relation, we created a new empty table **BPLFantasy** which contains the columns that have identical names and datatypes to the columns in the CSV file. Then, we used \copy to copy the CSV into our new table.

To understand the data better, we wrote five queries of varying complexity. The first query showed us how many players were in the league from each field position. The second query showed us the names and stats of the players who have generated the most fantasy points and scored the most goals. The third query identified the most influential outfield players and compared their influence to the average influence of all players in their position. The fourth query identified the most lethal forwards using CTEs and compared their threat scores to the average threat score of all forwards. The fifth and final query is the most useful one since it identified the midfielders who produce the most fantasy points and are most creative, while having an above average creativity in their respective positions. Finally, we queried a table of top

performing goalkeepers (least goals conceded and most clean sheets) and who have played more minutes than the average. We then exported the results of this table into a CSV file **top\_gks** and opened the file in Visual Studio Code.

This assignment was a great learning experience for me since I got to practice queries and subqueries. I feel much more confident now about planning and writing queries