

# **INF 551- FOUNDATIONS OF DATA MANGEMENT**

## **FINAL REPORT**

### ***JEUGE PARAISO***

#### **Group Members:**

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#### **Project Idea:**

- Gamers often don't know which games to buy, since there is a large variety to pick from.
- This variety arises due to different consoles, gaming genres, game producer, popularity, etc. We wanted to make an application which gives a one-stop solution to the above problems.
- Our web application will give a gamer insight into the best-selling games of the last 40 years, based on different criteria such as sales, genre, publisher, console availability, game versions, etc. This will help him/her to buy a game after a quick one-stop analysis.

#### **Technology Stack Used:**

- Database - Firebase
- Server-side scripting - Flask
- Front-end - HTML, CSS, JavaScript

#### **Dataset:** “Video Game Sales”

<https://www.kaggle.com/gregorut/videogamesales>

Shape of the Data Set: 16,600 rows x 11 columns

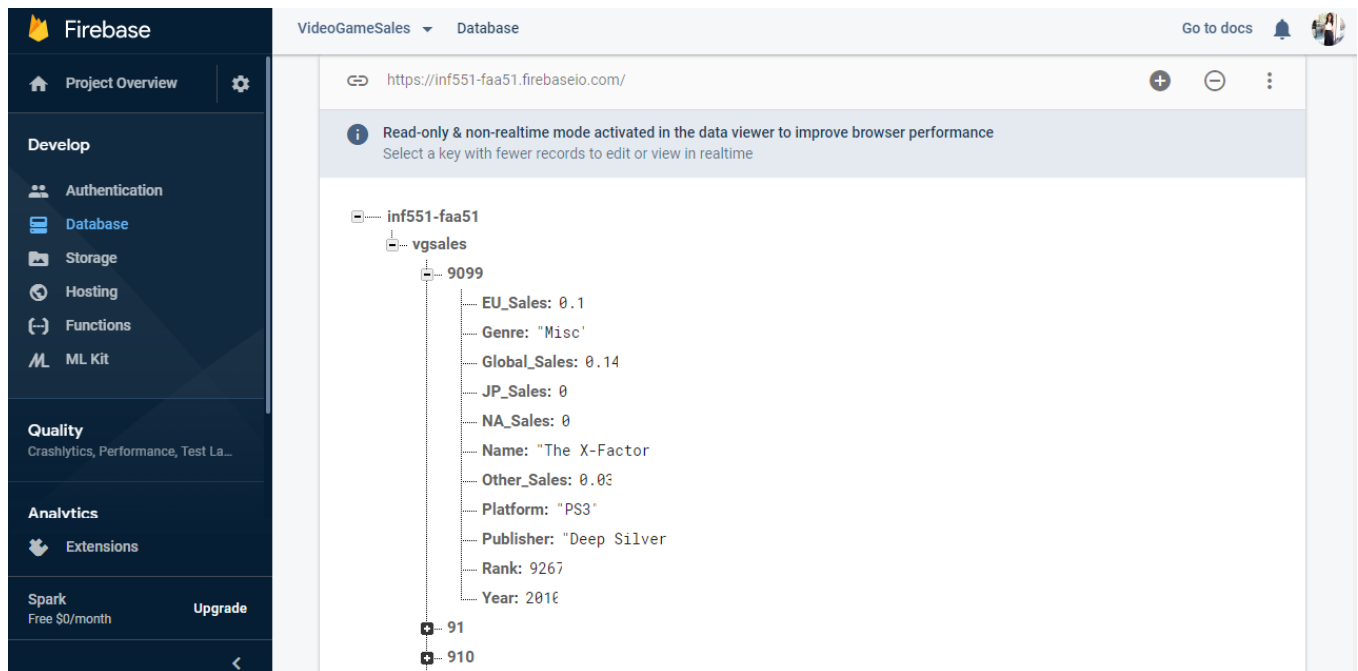
#### **Example Data:**

Rank	Name	Platform	Year	Genre	Publisher	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales
1	Wii Sports	Wii	2006	Sports	Nintendo	41.49	29.02	3.77	8.46	82.74
2	Super Mario	NES	1985	Platform	Nintendo	29.08	3.58	6.81	0.77	40.24
3	Mario Kart W	Wii	2008	Racing	Nintendo	15.85	12.88	3.79	3.31	35.82
4	Wii Sports R	Wii	2009	Sports	Nintendo	15.75	11.01	3.28	2.96	33
5	Pokemon Re	GB	1996	Role-Playing	Nintendo	11.27	8.89	10.22	1	31.37
6	Tetris	GB	1989	Puzzle	Nintendo	23.2	2.26	4.22	0.58	30.26
7	New Super M	DS	2006	Platform	Nintendo	11.38	9.23	6.5	2.9	30.01
8	Wii Play	Wii	2006	Misc	Nintendo	14.03	9.2	2.93	2.85	29.02
9	New Super N	Wii	2009	Platform	Nintendo	14.59	7.06	4.7	2.26	28.62
10	Duck Hunt	NES	1984	Shooter	Nintendo	26.93	0.63	0.28	0.47	28.31
11	Nintendogs	DS	2005	Simulation	Nintendo	9.07	11	1.93	2.75	24.76

## Reason for choosing the Dataset:

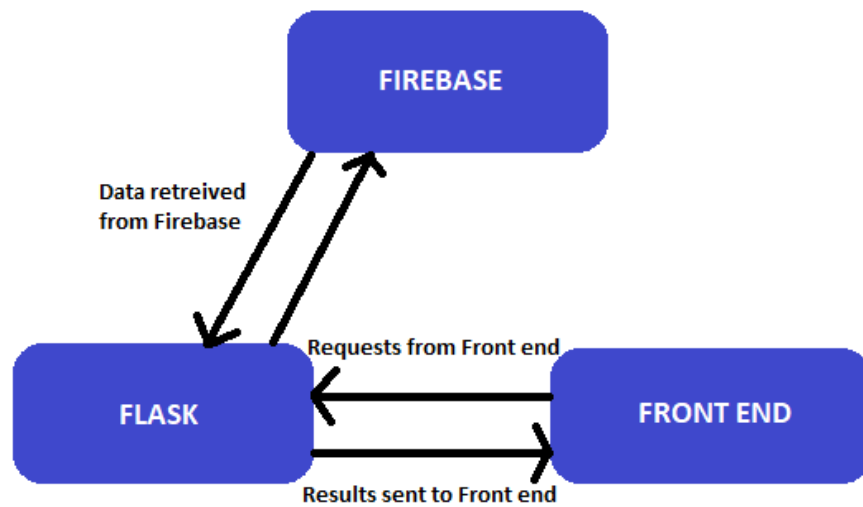
- The dataset has both numerical and non-numerical attributes.
- Size of the dataset makes it ideal for this application as it is more than 1k and less than 100k.
- It is an interesting dataset as it encompasses all those attributes needed for a gamer to buy or analyze a game.

## Firestore Database:



## Data Flow:

- On button clicks on the front end, the request is sent to the server side handler (Flask). Based on these requests, it extracts live data from Firestore every time a gamer chooses a particular sorting or filtering option.
- We perform manipulations on this retrieved data using Python on the back-end.
- The required dataset is sent back to the front end from Flask.



### Working Components:

It is a cloud-based Web application with the following features and implementation.

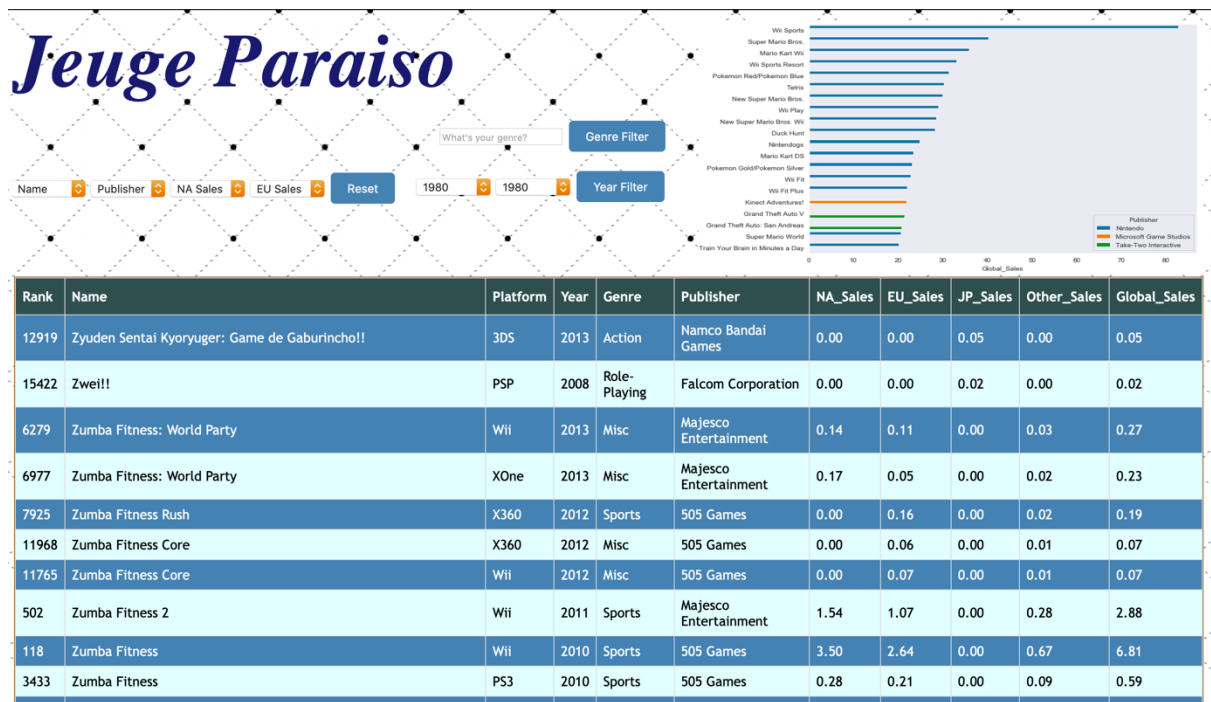
The screenshot shows the 'Jeuge Paraiso' landing page. At the top, there's a title 'Jeuge Paraiso' in a stylized font. Below it, there are search filters: 'What's your genre?' with a 'Genre Filter' button, and '1980' with a 'Year Filter' button. Below the filters is a table with 11 columns: Rank, Name, Platform, Year, Genre, Publisher, NA\_Sales, EU\_Sales, JP\_Sales, Other\_Sales, and Global\_Sales. The table lists 14 games, with 'Wii Sports' at the top and 'Wii Fit' at the bottom.

Rank	Name	Platform	Year	Genre	Publisher	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales
1	Wii Sports	Wii	2006	Sports	Nintendo	41.49	29.02	3.77	8.46	82.74
2	Super Mario Bros.	NES	1985	Platform	Nintendo	29.08	3.58	6.81	0.77	40.24
3	Mario Kart Wii	Wii	2008	Racing	Nintendo	15.85	12.88	3.79	3.31	35.82
4	Wii Sports Resort	Wii	2009	Sports	Nintendo	15.75	11.01	3.28	2.96	33.00
5	Pokemon Red/Pokemon Blue	GB	1996	Role-Playing	Nintendo	11.27	8.89	10.22	1.00	31.37
6	Tetris	GB	1989	Puzzle	Nintendo	23.20	2.26	4.22	0.58	30.26
7	New Super Mario Bros.	DS	2006	Platform	Nintendo	11.38	9.23	6.50	2.90	30.01
8	Wii Play	Wii	2006	Misc	Nintendo	14.03	9.20	2.93	2.85	29.02
9	New Super Mario Bros. Wii	Wii	2009	Platform	Nintendo	14.59	7.06	4.70	2.26	28.62
10	Duck Hunt	NES	1984	Shooter	Nintendo	26.93	0.63	0.28	0.47	28.31
11	Nintendogs	DS	2005	Simulation	Nintendo	9.07	11.00	1.93	2.75	24.76
12	Mario Kart DS	DS	2005	Racing	Nintendo	9.81	7.57	4.13	1.92	23.42
13	Pokemon Gold/Pokemon Silver	GB	1999	Role-Playing	Nintendo	9.00	6.18	7.20	0.71	23.10
14	Wii Fit	Wii	2007	Sports	Nintendo	8.94	8.03	3.60	2.15	22.72

Original landing page

#### (i) **Sorting:**

- for non-numerical data on attributes of 'Name' and 'Publisher'.
- for numerical data on attribute of 'NA\_Sales' and 'Global\_Sales'.



Name Descending Sorting



Publisher Descending Sorting



NA Sales Descending Sorting



EU Sales Descending Sorting

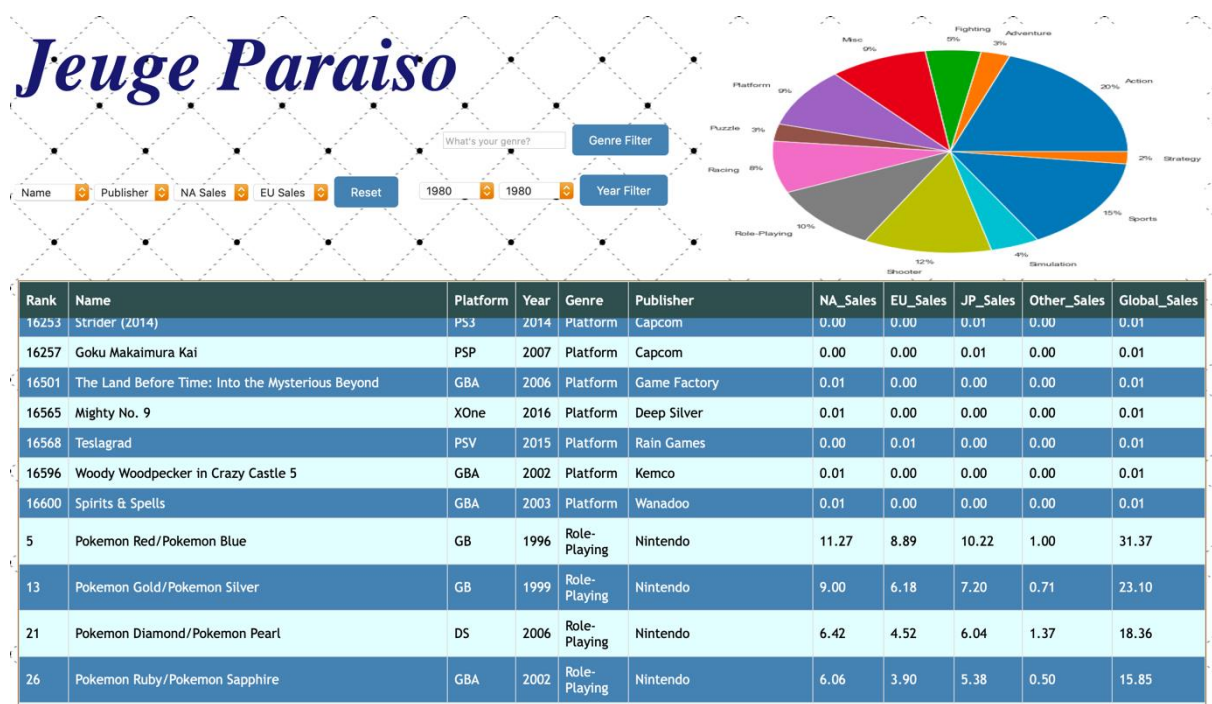
## Feature Implementation:

- We've created four sorting buttons, each with the ascending and descending options.
- As explained before, flask gets the data from firebase after the request is sent from the front end.

- We convert this data into a dataframe, and then sort it based on the request received from the gamer.
- The sorted dataframe is displayed as a HTML table on the front end.
- Since we're using Flask, we've used 'render template()' to return four different templates on the basis of the sorting option selected. This is because each option is associated with a different visualization as well. (discussed later in this document).

## (ii) Filter function

- for non-numerical data on the attribute 'Genre'.
- for numerical data on the attribute 'Year'.



Genre Filter on syllable 'pl'





Year Filter on 1984 - 1993

## Feature Implementation:

### GENRE:

- The gamer can choose to enter any syllable of the gaming genre he/she is interested in. This can be entered in any case as well.
- On the back-end, our code contains a list of all the possible genres present in the database.
- The entered syllable is converted into lower case and compared with the list of genres(all in lower case). If the syllable matches any of the words in the list, those matched words are appended to another list.
- The above case insensitive comparison logic helps the gamer to input any kind of text in any case.
- Now, on the data retrieved from firebase, we pass items in the newly appended list to 'Genre'.
- The results of the above operations are appended to an empty dataframe and the final dataframe result is displayed as an HTML table on the front end.
- Since we're using Flask, we've used 'render template()' to return a different template when genre filtering is selected. This is because each option is associated with a different visualization as well. (discussed later in this document).
- **Error Handling** : If the user enters a genre or syllable that does belong to any genre in the database, it is taken to a page (with the help of render\_template( ))where it is prompted to enter the correct genre.

**Oops! Looks like you took a wrong turn ! Please try some other genre.**

[GO BACK!](#)

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Error Page for 'Genre'

### **YEAR:**

- The starting and end years are taken from a POST form method of two separate drop-down boxes.
- These two years along with the years between them are then stored in a list.
- Now, on the data retrieved from firebase, we pass the items in the newly appended list to 'Year'.
- The results of the above operations are appended to an empty dataframe and the final dataframe result is displayed as an HTML table on the front end.
- Since we're using Flask, we've used 'render template()' to return a different template when genre filtering is selected . This is because each option is associated with a different visualization as well. (discussed later in this document).
- **Error Handling** : If the user enters a starting year which is greater than the ending year, it is taken to a page (with the help of render\_template() )where it is prompted to select the correct range of years.

**Oops! Looks like you took a wrong turn ! Please select a valid range.**

[GO BACK!](#)

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Error Page for 'Year'



### (iii) **Visual Representation:**

- Non-dynamic graphs and charts for sorting and filtering functions: (the above attached screenshots have the respective visualizations)
- On clicking of ASC/DES of **Name** button : Bar Chart of the top 20 selling games of the last 40 years, along with their respective publishers.
- On clicking of ASC/DES of **Publisher** button : Pie-chart showing publishers' contribution to the global sales of games in the last 40 years.
- On clicking of ASC/DES of **NA Sales** button : Double bar chart of the top 10 games with the highest sales in North America, and comparison of the NA Sale figures to the global sales of these games.
- On clicking of ASC/DES of **EU Sales** button : Double bar chart of the top 10 games with the highest sales in Europe, and comparison of the EU Sale figures to the global sales of these games.
- On clicking of **Year Filter** button : Line graph of global sales of all the top games over the entire year range from 1980 - 2020
- On clicking of **Genre Filter** button : Pie-chart showing genres' contribution to the global sales of games in the last 40 years.

### **Group Responsibility:**

Pratishtha Singh:

- Loading of the data on Firebase
- Implementing filtering function on the dataset

Sharad Narayan Sharma:

- Implementation of Sorting function
- Creation of Visualization graphs.

NOTE : We've re-submitted the final presentation PPT as we made some changes to our final project.