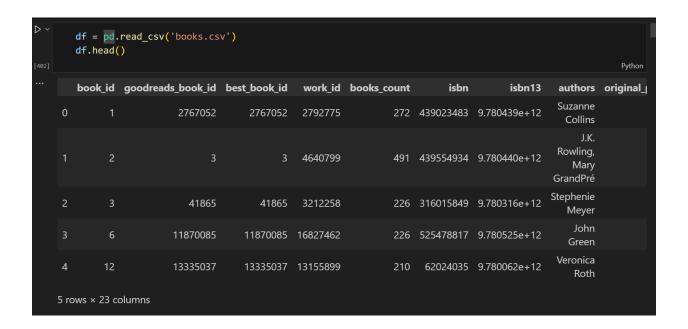
# **Cloud Report**

```
import pandas as pd import matplotlib.pyplot as plt import seaborn as sns from sklearn.impute import SimpleImputer
```

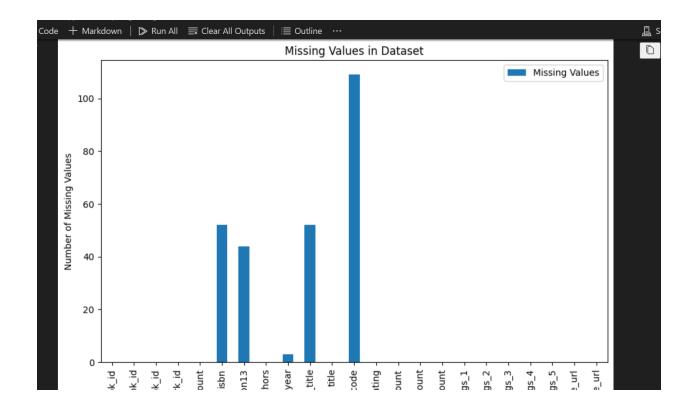
## **Importing libraries**



## Reading our dataset

```
missing_values = pd.DataFrame(df.isnull().sum(), columns=['Missing Values'])
  missing_values
                       Missing Values
               book_id
    goodreads_book_id
          best_book_id
                                    0
               work_id
                                    0
          books_count
                                    0
                  isbn
                                   52
                isbn13
                                   44
               authors
original_publication_year
           original_title
                  title
        language_code
                                  109
```

Here we see the empty values in our dataset



### Plotting missing values in our dataset

```
num_cols = df.select_dtypes(include=['int64', 'float64']).columns
categorical_cols = df.select_dtypes(include=['object']).columns

imputer_numeric = SimpleImputer(strategy='mean')
df[num_cols] = imputer_numeric.fit_transform(df[num_cols])

imputer_categorical = SimpleImputer(strategy='most_frequent')
df[categorical_cols] = imputer_categorical.fit_transform(df[categorical_cols])

print("Missing Values After Imputation:")
print(df.isnull().sum().to_string())
Python
```

Here we impute our empty values in the numeric columns with strategy mean

And we impute our empty values in the categorical columns with strategy mean

# **Output:**

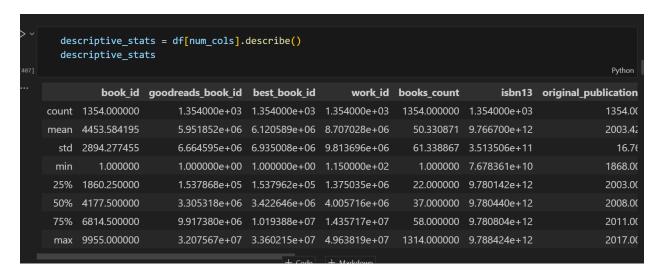
Code 🕂 Iviarkdown   📂 kun Aii	=x Clear All Outputs   := Outline ····
Missing Values After I	mputation:
book_id	0
goodreads_book_id	0
best_book_id	0
work_id	0
books_count	0
isbn	0
isbn13	0
authors	0
original_publication_y	ear 0
original_title	0
title	0
language_code	0
average_rating	0
ratings_count	0
work_ratings_count	0
work_text_reviews_coun	t 0
ratings_1	0
ratings_2	0
ratings_3	0
ratings_4	0
ratings_5	0
image_url	0
small image url	а

```
duplicates = df.duplicated().sum()
duplicates

e6]

Python
```

### Check if there is duplicates in our dataset



### Show our descriptive statistics of our numeric columns

```
df_cleaned = pd.DataFrame()
    df_cleaned = df[df[categorical_cols].apply(lambda x: x.str.contains('Harry Potter', na=False)).any
    (axis=1)]
    harry_potter_books = df_cleaned.iloc[:-1]
    harry_potter_books.head()
Python
```

Here we filter our data to contain harry potter books only

#### **Output:**

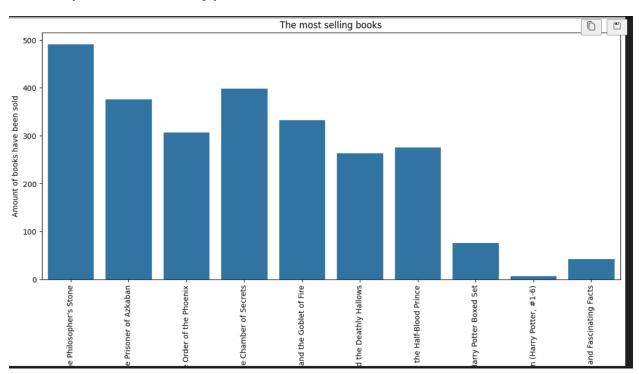
	book_id	goodreads_book_id	best_book_id	work_id	books_count	isbn	isbn13	authors	origin
1	2.0	3.0	3.0	4640799.0	491.0	439554934	9.780440e+12	J.K. Rowling, Mary GrandPré	
6	18.0	5.0	5.0	2402163.0	376.0	043965548X	9.780440e+12	J.K. Rowling, Mary GrandPré, Rufus Beck	
8	21.0	2.0	2.0	2809203.0	307.0	439358078	9.780439e+12	J.K. Rowling, Mary GrandPré	
9	23.0	15881.0	15881.0	6231171.0	398.0	439064864	9.780439e+12	J.K. Rowling, Mary GrandPré	
10	24.0	6.0	6.0	3046572.0	332.0	439139600	9.780439e+12	J.K. Rowling, Mary	

	ha	rry_potter_books['original_title']
109]		
••	1	Harry Potter and the Philosopher's Stone
	6	Harry Potter and the Prisoner of Azkaban
	8	Harry Potter and the Order of the Phoenix
	9	Harry Potter and the Chamber of Secrets
	10	Harry Potter and the Goblet of Fire
	11	Harry Potter and the Deathly Hallows
	12	Harry Potter and the Half-Blood Prince
	96	Complete Harry Potter Boxed Set
	613	Harry Potter Collection (Harry Potter, #1-6)
	1036	The Magical Worlds of Harry Potter: A Treasury
	Name:	original_title, dtype: object

Harry potter books title

```
plt.figure(figsize=(14, 6))
    sns.barplot(x=harry_potter_books['original_title'], y=harry_potter_books['books_count'])
    plt.title("The most selling books")
    plt.xlabel("Columns")
    plt.ylabel("Amount of books have been sold")
    plt.xticks(rotation=90)
    plt.show()
```

# Here we plot the most harry potter selled books



```
ratings_columns = ['ratings_1', 'ratings_2', 'ratings_3', 'ratings_4', 'ratings_5']
   harry_potter_books['Average_Rating'] = harry_potter_books[ratings_columns].mean(axis=1)
   harry potter books['Average Rating'].head()
                                                                                                                        Python
\underline{\text{C:} \text{Users} \text{$$\lambda$ppData$Local$Temp\ipykernel\_17140$\1978471561.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: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.htm">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.htm</a>
  harry_potter_books['Average_Rating'] = harry_potter_books[ratings_columns].mean(axis=1)
       960013.0
       393875.0
6
       368109.6
8
9
       381239.8
10
       373728.4
```

Create new column named Average rating to make analysis on it

```
plt.figure(figsize=(14, 6))
    sns.barplot(x=harry_potter_books['Average_Rating'], y=harry_potter_books['books_count'])
    plt.title("Average rating of most selled books")
    plt.xlabel("Average rating")
    plt.ylabel("Sold books")
    plt.xticks(rotation=90)
    plt.show()
```

Ploting The Average rating of our books with the most sold books

```
correlation_matrix = pd.DataFrame(harry_potter_books[ha_num_cols].corr())

correlation_with_books_count = pd.DataFrame(correlation_matrix['books_count'].sort_values
    (ascending=False))

print("Correlation with books_count:")
print(correlation_with_books_count)
print(50*'*')
Python
```

Create correlation matrix with most books sold if number bigger than 0.5 means strong relationship if less than 0.5 means weak relationship.

```
Correlation with books_count:
                          books count
books_count
                             1.000000
ratings_4
                             0.936038
ratings_3
                             0.908513
Average Rating
                             0.906421
work_ratings_count
                             0.906421
ratings_count
                             0.900125
ratings_5
                             0.891211
work text reviews count
                             0.876589
ratings 2
                             0.807796
ratings 1
                             0.623421
average rating
                            -0.037882
work id
                            -0.138068
isbn13
                            -0.231655
original_publication_year
                            -0.374821
goodreads_book_id
                            -0.596628
best_book_id
                            -0.596628
book id
                            -0.721804
*********************
```

Plotting heat map to see our correlation matrix

