

# **DEDP PRESENTATION**

A

PROJECT ON

Top\_200\_Comman\_password  
\_by\_Different\_Country\_2022

BY

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## ***Certificate***

*This is to certify that Akshata Dattatray Shirwale has presented a mini project titled Top 200 common passwords by different country in partial fulfillment of the requirements of M.Sc.*

*(Computer Science) Part I (semester I) course.*

*Date: 14/01/2023*

Teacher In-Charge

Head,

Department of Computer Science

# **ACKNOWLEDGEMENT**

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# 1) **Description of the data**

- We take data from kagal.com website (Top 200 common passwords by country 2022).
- We have all country data from A-Z.
- In data we have column name such as- country\_code, country/territory name, rank, password, user count, time to crack password, global rank, time to crack password (in seconds).

## **2) Data observation techniques**

- **Bar graph**
- **Histogram**
- **Heat map**
- **Word cloud**
- **Box plot**
- **Scatter plot**
- **Waterfall chart**
- **Pictogram chart**

### **3) Techniques used** **for Preprocessing Data**

A real-world data generally contains noises, missing values, and maybe in an unusable format which cannot be directly used for processing. Data preprocessing is a required task for cleaning the data and making it suitable for processing which also increases the accuracy and efficiency of the model.

#### **1) Importing Libraries :-**

*In order to perform data preprocessing using Python, we need to import some predefined Python libraries. These libraries are used to perform some specific jobs. There are three specific libraries that we will use for data preprocessing, which are:*

- **Numpy:-**

*Numpy Python library is used for including any type of mathematical operation in the code. It is the fundamental package for scientific calculation in Python. It also supports adding large, multidimensional arrays and matrices.*

- **Matplotlib:-**

*The second library is matplotlib, which is a Python 2D plotting library, and with this library, we need to import a sub-library pyplot. This library is used to plot any type of charts in Python for the code.*

- **Pandas:-**

*The last library is the Pandas library, which is one of the most famous Python libraries and used for importing and managing the datasets. It is an open-source data manipulation and analysis library.*

## **2) Handling Missing data :-**

*The next step of data preprocessing is to handle missing data in the datasets. If our dataset contains some missing data, then it may create a huge problem for our machine learning model. Hence it is necessary to handle missing values present in the dataset.*

### **Calculating the mean :-**

*In this way, we will calculate the mean of that column or row which contains any missing value and will put it in the place of missing value. This strategy is useful for the features which have numeric data.*



## 4) Python Code

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import plotly.express as px
from collections import Counter
from wordcloud import WordCloud
from PIL import Image
from wordcloud import WordCloud, STOPWORDS, ImageColorGenerator
```

```
[6] df=pd.read_csv('/content/sample_data/Top_200_Comman_password_by_Different_Country_2022.csv')
print(df)
print("size of data is:",df.shape)
```

	country_code	country	Rank	Password	User_count	Time_to_crack \
0	au	Australia	1	123456	308483	< 1 second
1	au	Australia	2	password	191880	< 1 second
2	au	Australia	3	lizottes	98220	3 Hours
3	au	Australia	4	password1	86884	< 1 second
4	au	Australia	5	123456789	75856	< 1 second
...	...	...	...	...	...	...
9795	vn	Vietnam	196	hongngoc	2660	3 Hours
9796	vn	Vietnam	197	anhtien	2628	17 Minutes
9797	vn	Vietnam	198	lanhuong	2620	3 Hours
9798	vn	Vietnam	199	congacon	2584	2 Hours
9799	vn	Vietnam	200	conmemay	2532	3 Hours

	Global_rank	Time_to_crack_in_seconds
0	1.0	0
1	5.0	0
2	NaN	10800
3	16.0	0
4	2.0	0
...	...	...
9795	NaN	10800
9796	NaN	1020
9797	NaN	10800
9798	NaN	7200
9799	NaN	10800

```
[9800 rows x 8 columns]
size of data is: (9800, 8)
```

---

```
missing_values=['N/a',"na",np.nan]
df=pd.read_csv('/content/sample_data/Top_200_Comman_password_by_Different_Country_2022.csv',na_values=missing_values)
print(df)
print("size of data is:",df.shape)
print("\n",df.isnull().sum())
```

```
country_code  country  Rank  Password  User_count  Time_to_crack \
0            au  Australia    1    123456    308483    < 1 second
1            au  Australia    2  password    191880    < 1 second
2            au  Australia    3  lizottes    98220     3 Hours
3            au  Australia    4  password1    86884    < 1 second
4            au  Australia    5  123456789    75856    < 1 second
...          ...      ...      ...      ...      ...
9795         vn  Vietnam    196  hongngoc    2660     3 Hours
9796         vn  Vietnam    197  anh tien    2628    17 Minutes
9797         vn  Vietnam    198  lanhuong    2620     3 Hours
9798         vn  Vietnam    199  congacon    2584     2 Hours
9799         vn  Vietnam    200  conmemay    2532     3 Hours
```

```
Global_rank  Time_to_crack_in_seconds
0            1.0                    0
1            5.0                    0
2           NaN                10800
3           16.0                    0
4            2.0                    0
...          ...                  ...
9795         NaN                10800
9796         NaN                1020
9797         NaN                10800
9798         NaN                7200
9799         NaN                10800
```

✓ Done completed at 11:14 PM

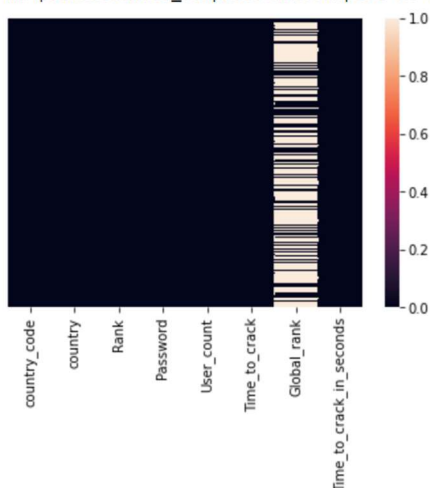
```
Global_rank  Time_to_crack_in_seconds
0            1.0                    0
1            5.0                    0
2           NaN                10800
3           16.0                    0
4            2.0                    0
...          ...                  ...
9795         NaN                10800
9796         NaN                1020
9797         NaN                10800
9798         NaN                7200
9799         NaN                10800
```

```
[9800 rows x 8 columns]
size of data is: (9800, 8)
```

```
country_code    0
country         0
Rank            0
Password        0
User_count      0
Time_to_crack   0
Global_rank     6628
Time_to_crack_in_seconds  0
dtype: int64
```

```
sns.heatmap(df.isnull(),yticklabels=False)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7ff293c06fa0>
```



```

countryc=df.country_code
print(countryc)
country=df.country
print(country)
rank=df.Rank
print(rank)
password=df.Password
print(password)
usercount=df.User_count
print(usercount)
timetocrack=df.Time_to_crack
print(timetocrack)
globalrank=df.Global_rank
print(globalrank)
timetocracksec=df.Time_to_crack_in_seconds
print(timetocracksec)

```

```
print("mean of global rank is",globalrank.mean())
```

```
mean of global rank is 65.33701134930644
```

```

▶ fill_mean=df.fillna(globalrank.mean())
print(fill_mean)

```

```

↳
   country_code  country  Rank  Password  User_count  Time_to_crack \
0             au  Australia    1    123456    308483    < 1 second
1             au  Australia    2  password    191880    < 1 second
2             au  Australia    3  lizottes     98220      3 Hours
3             au  Australia    4  password1     86884    < 1 second
4             au  Australia    5  123456789     75856    < 1 second
...           ...      ...      ...      ...      ...      ...
9795          vn  Vietnam    196  hongngoc     2660      3 Hours
9796          vn  Vietnam    197   anhtien     2628    17 Minutes
9797          vn  Vietnam    198  lanhuong     2620      3 Hours
9798          vn  Vietnam    199  congacon     2584      2 Hours
9799          vn  Vietnam    200  conmemay     2532      3 Hours

```

```

   Global_rank  Time_to_crack_in_seconds
0          1.000000                0
1          5.000000                0
2        65.337011            10800
3        16.000000                0

```

```

   Global_rank  Time_to_crack_in_seconds
0          1.000000                0
1          5.000000                0
2        65.337011            10800
3        16.000000                0
4          2.000000                0
...           ...      ...
9795        65.337011            10800
9796        65.337011            1020
9797        65.337011            10800
9798        65.337011             7200
9799        65.337011            10800

```

```
[9800 rows x 8 columns]
```

```
print(fill_mean.isnull().sum())
```

```
country_code      0
country           0
Rank              0
Password          0
User_count        0
Time_to_crack     0
Global_rank       0
Time_to_crack_in_seconds  0
dtype: int64
```

```
#countries with longest time to crack the password
```

```
cp=fill_mean[fill_mean.Time_to_crack_in_seconds>100000000][['country','Time_to_crack_in_seconds']]
print(cp)
```

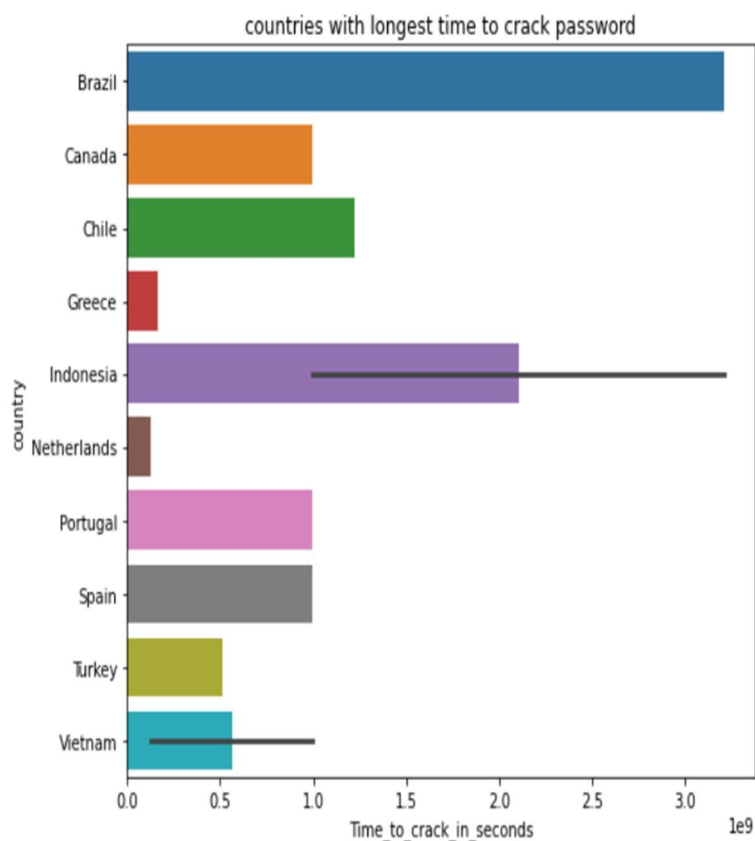
	country	Time_to_crack_in_seconds
793	Brazil	3214080000
999	Canada	996364800
1058	Chile	1221350400
2828	Greece	160704000
3493	Indonesia	3214080000
3495	Indonesia	3214080000
3528	Indonesia	996364800
3554	Indonesia	996364800
5557	Netherlands	128563200
6796	Portugal	996364800
7988	Spain	996364800
8633	Turkey	514252800
9685	Vietnam	128563200
9757	Vietnam	996364800

```
plt.figure(figsize=(8,7))
```

```
plots = sns.barplot(y=cp.country,x=cp.Time_to_crack_in_seconds,data=cp)
```

```
plt.title("countries with longest time to crack password")
```

```
plt.show()
```



```
#Passwords took longest time to crack
pt=fill_mean.nlargest(10,columns='Time_to_crack_in_seconds')[['Password','Time_to_crack_in_seconds']]
print(pt)
```

	Password	Time_to_crack_in_seconds
793	estantevirtual	3214080000
3493	omarbelmestour	3214080000
3495	kallynlavallee	3214080000
1058	paralelepipedo	1221350400
999	ihatethisgame	996364800
3528	pabloparraito	996364800
3554	clayburnclark	996364800
6796	clayburnclark	996364800
7988	clayburnclark	996364800
9757	dothingocthuy	996364800

```
#Above passwords are only alpha passwords it doesn't have numeric passwords
```

```
#Types of passwords
```

```
c=[]
for i in fill_mean.Password:
    if i.isdigit():
        c.append('Numeric')
    elif i.isalpha():
        c.append('alpha')
    else:
        c.append('mixed')
```

```
fill_mean['Type_pass']=c
print(fill_mean)
```

```
country_mean
```

	country_code	country	...	Time_to_crack_in_seconds	Type_pass
0	au	Australia	...	0	Numeric
1	au	Australia	...	0	alpha
2	au	Australia	...	10800	alpha
3	au	Australia	...	0	mixed
4	au	Australia	...	0	Numeric
...	...	...	...	...	...
9795	vn	Vietnam	...	10800	alpha
9796	vn	Vietnam	...	1020	alpha
9797	vn	Vietnam	...	10800	alpha
9798	vn	Vietnam	...	7200	alpha
9799	vn	Vietnam	...	10800	alpha

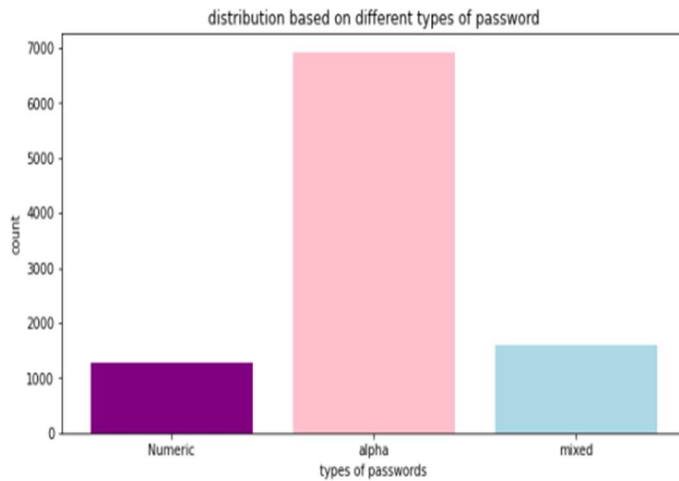
```
[9800 rows x 9 columns]
```

```
count1=Counter(fill_mean.Type_pass)
print(count1)
keys=count1.keys()
```

```
print(keys)
values=count1.values()
print(values)
fig=plt.figure(figsize=(10,5))
c=['purple','pink','lightblue']
plt.bar(keys,values,color=c)
plt.xlabel("types of passwords")
plt.ylabel("count")
plt.title("distribution based on different types of password")
plt.show()
```

```
Counter({'alpha': 6923, 'mixed': 1588, 'Numeric': 1289})
dict_keys(['Numeric', 'alpha', 'mixed'])
dict_values([1289, 6923, 1588])
```



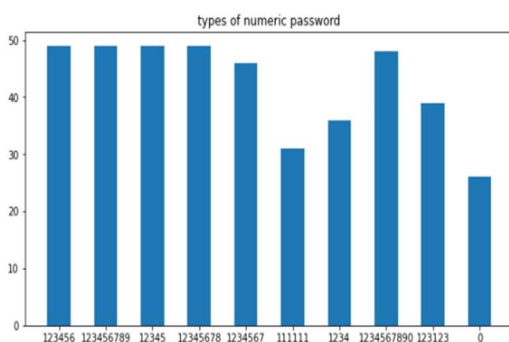


```
#Top used numeric passwords
num=[]
for i in fill_mean.Password:
    if i.isdigit():
        num.append(i)
new_dict=dict.fromkeys(num,0)
print(new_dict,type(new_dict))

for total in new_dict:
    new_dict[total]=num.count(total)
print(new_dict)
dict_items=new_dict.items()
ft=list(dict_items)[:10]
print(ft)
```

```
{'123456': 0, '123456789': 0, '12345': 0, '12345678': 0, '1234567': 0, '111111': 0, '1234': 0, '1234567890': 0, '123123': 0, '0': 0, '654321': 0, '123': 0, '123456': 49, '123456789': 49, '12345': 49, '12345678': 49, '1234567': 46, '111111': 31, '1234': 36, '1234567890': 48, '123123': 39, '0': 26, '654321': 46, [('123456', 49), ('123456789', 49), ('12345', 49), ('12345678', 49), ('1234567', 46), ('111111', 31), ('1234', 36), ('1234567890', 48), ('123123', 39), ('0',
```

```
a=dict(ft)
numeric=a.keys()
count=a.values()
fig=plt.figure(figsize=(10,5))
plt.bar(numeric,count,width=0.5)
plt.xlabel("password")
plt.ylabel("count")
plt.title("types of numeric password")
plt.show()
```



```
#Top used alpha passwords
num1=[]
for i in fill_mean.Password:
    if i.isalpha():
        num1.append(i)
new_dict1=dict.fromkeys(num1,0)
print(new_dict1,type(new_dict1))

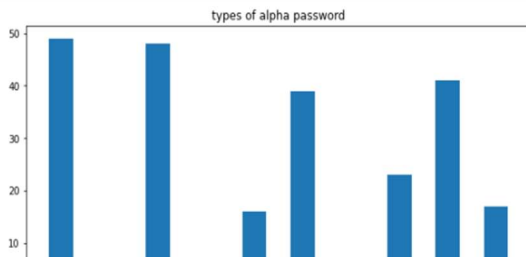
for total in new_dict1:
    new_dict1[total]=num1.count(total)
print(new_dict1)
dict_items1=new_dict1.items()
ft=list(dict_items1)[:10]
print(ft)
```

```
{'password': 0, 'lizottes': 0, 'qwerty': 0, 'holden': 0, 'charlie': 0, 'dragon': 0, 'australia': 0, 'princess': 0, 'iloveyou': 0, 'chocolate': 0, 'soccer': 0, 'monkey': 0, 'tigers': 0, 'michael': 0, 'poke
{'password': 49, 'lizottes': 1, 'qwerty': 48, 'holden': 2, 'charlie': 16, 'dragon': 39, 'australia': 1, 'princess': 23, 'iloveyou': 41, 'chocolate': 17, 'soccer': 11, 'monkey': 28, 'tigers': 1, 'michael':
[('password', 49), ('lizottes', 1), ('qwerty', 48), ('holden', 2), ('charlie', 16), ('dragon', 39), ('australia', 1), ('princess', 23), ('iloveyou', 41), ('chocolate', 17)]
```

```

a=dict(ft)
alpha=a.keys()
count=a.values()
fig=plt.figure(figsize=(10,5))
plt.bar(alpha,count,width=0.5)
plt.xlabel("password")
plt.ylabel("count")
plt.title("types of alpha password")
plt.show()

```



```

#Top used mixed passwords
num2=[]
for i in fill_mean.Password:
    if (i.isalpha()==False and i.isdigit()==False) :
        num2.append(i)
new_dict2=dict.fromkeys(num2,0)
print(new_dict2,type(new_dict2))

for total in new_dict2:
    new_dict2[total]=num2.count(total)
print(new_dict2)
dict_items2=new_dict2.items()
ft=list(dict_items2)[:10]
print(ft)

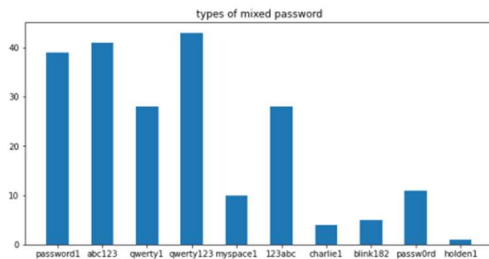
{'password1': 0, 'abc123': 0, 'qwerty1': 0, 'qwerty123': 0, 'myspace1': 0, '123abc': 0, 'charlie1': 0, 'blink182': 0, 'password': 0, 'holden1': 0, 'trustno1': 0, 'hello123': 0, 'hello1': 0, 'australia1': 0, 'password1': 39, 'abc123': 41, 'qwerty1': 28, 'qwerty123': 43, 'myspace1': 10, '123abc': 28, 'charlie1': 4, 'blink182': 5, 'password': 11, 'holden1': 1, 'trustno1': 7, 'hello123': 6, 'hello1': 2, 'australia1': 2, 'password1': 39, ('abc123', 41), ('qwerty1', 28), ('qwerty123', 43), ('myspace1', 10), ('123abc', 28), ('charlie1', 4), ('blink182', 5), ('password', 11), ('holden1', 1)]

```

```

a=dict(ft)
mixed=a.keys()
count=a.values()
fig=plt.figure(figsize=(10,5))
plt.bar(mixed,count,width=0.5)
plt.xlabel("password")
plt.ylabel("count")
plt.title("types of mixed password")
plt.show()

```



```

fill_mean['Length']=fill_mean['Password'].str.len()
print(fill_mean)

```

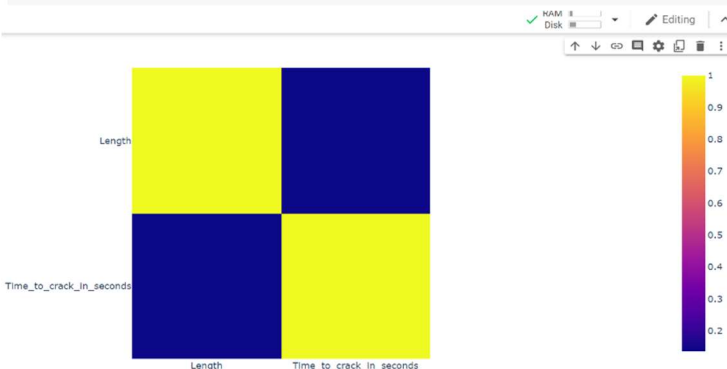
	country_code	country	...	Time_to_crack_in_seconds	Length
0	au	Australia	...	0	6
1	au	Australia	...	0	8
2	au	Australia	...	10800	8
3	au	Australia	...	0	9
4	au	Australia	...	0	9
...	...	...	...	...	...
9795	vn	Vietnam	...	10800	8
9796	vn	Vietnam	...	1020	7
9797	vn	Vietnam	...	10800	8
9798	vn	Vietnam	...	7200	8
9799	vn	Vietnam	...	10800	8

[9800 rows x 9 columns]

```

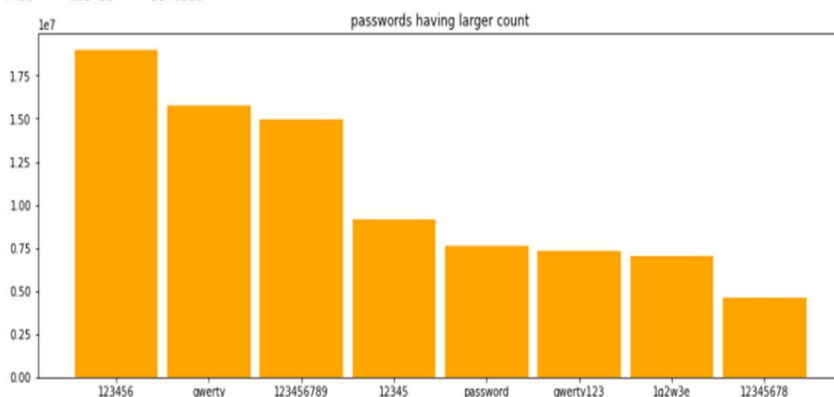
#correlation between password length and time to crack
fill_mean['Length']=fill_mean['Password'].str.len()
dp=fill_mean[['Length','Time_to_crack_in_seconds']].corr()
px.imshow(dp)

```

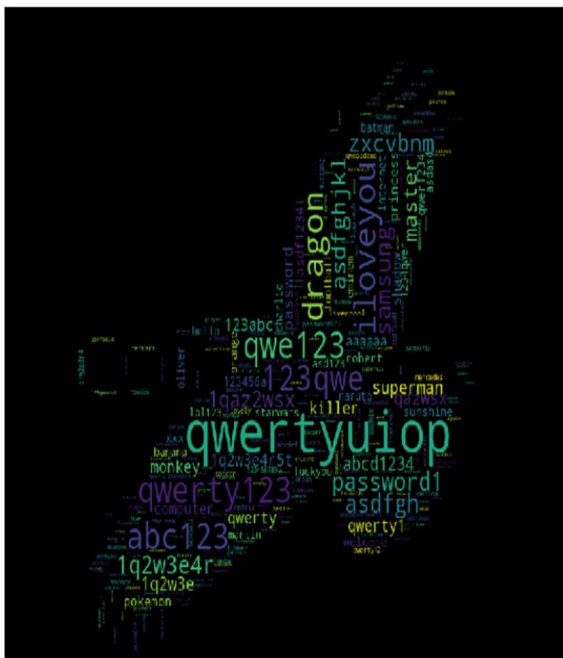


```
pass1=a['Password']
print(a)
user=a['User_count']
fig=plt.figure(figsize=(15,5))
plt.bar(pass1,user,color='orange',width=0.9)
plt.xlabel("Password")
plt.ylabel("User count")
plt.title("passwords having larger count")
plt.show()
```

	Password	User_count
7000	123456	19000630
7001	qwerty	15738011
7002	123456789	14975791
7003	12345	9139679
1200	123456	8159358
7004	password	7593503
7005	qwerty123	7343013
7006	1q2w3e	7051194
7007	12345678	4632962
9400	123456	3572081



```
#weekest password
mask=np.array(Image.open("/content/birdimage.jpg"))
mask
weekpass=fill_mean[fill_mean.Time_to_crack_in_seconds==0]['Password'].to_list()
#print(weekpass)
wc=WordCloud(stopwords=STOPWORDS,mask=mask,background_color="black",
             max_words=5000,max_font_size=1000,random_state=99,
             width=mask.shape[1],height=mask.shape[0])
wc.generate(" ".join(weekpass))
plt.figure(figsize=(10,10),facecolor='k')
plt.imshow(wc,interpolation='None')
plt.axis("off")
plt.show()
```

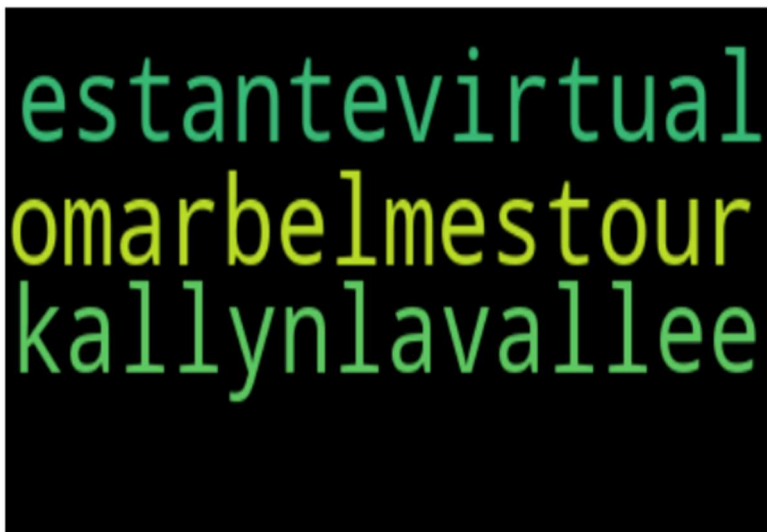




```
#strongest password
strongpass=fill_mean[fill_mean.Time_to_crack=='Centuries']['Password'].to_list()
print(strongpass)
```

```
['estantevirtual', 'omarbelmestour', 'kallynlavallee']
```

```
#create wordcloud image
wordcloud=WordCloud().generate(" ".join(strongpass))
#display image
plt.figure(figsize=(10,10))
plt.imshow(wordcloud,interpolation='bilinear')
plt.axis("off")
plt.show()
```



## 5) Data Visualization & Interpretation

- In countries we see Brazil country take longest time to crack the password.
- We can see the alpha passwords are mostly used in all the countries.

- In numeric password pattern (1234567890) , In alpha pattern (password/qwerty) , In mixed (qwerty123) password are used most rather than remaining patterns.
- Password (123456) has large user\_count in all countries.
- In we have analyze the weakest passwords with the help of word cloud.
- By word cloud image qwertyuiop is bolder so this is the weakest password.
- In this we have analyse the strongest passwords with the help of word cloud.
- In this case kallynlavallee is bolder so this is the strongest password.

## **6) Conclusion**

- ❖ In this project we have analyzed the which countries take longer time to crack the passwords, password which took longer time to crack, types of passwords, strongest and weakest passwords.
- ❖ finally we concluded that we have to use alpha\_password or mixed password because it is difficult to crack that type of password.

- ❖ For data processing we used data cleaning method.
- ❖ For analysis we used bar chart, horizontal bar chart, heat map and word cloud.

## **7) Bibliography**

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