

Music Recommendation System

Import Libraries

```
In [1]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
from sklearn.preprocessing import StandardScaler
from sklearn.ensemble import RandomForestClassifier
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, confusion_matrix

import warnings
warnings.filterwarnings('ignore')
```

Reading the dataset

```
In [2]: members = pd.read_csv("members.csv")
members
```

```
Out[2]:
```

		msno	city	bd	gender	registered_via	registration_init_time	exp
0	XQxgAYj3klVKjR3oxPPXYFp4soD4TuBghkhMTD4oTw=	1	0	NaN	7	20110820		
1	UizsfmJb9mV54qE9hCYyU07Va97c0ICRLEQX3ae+ztM=	1	0	NaN	7	20150628		
2	D8nEhslOBSoE6VthTaqDX8U6lqjJ7dLdr72mOyLya2A=	1	0	NaN	4	20160411		
3	mCuD+tZ1hERA/o5GPqk38e041J8ZsBaLcu7nGollvhl=	1	0	NaN	9	20150906		
4	q4HRBfVSssAFS9iRfxWrohuk9kCYMKjHOEagUMV6rQ=	1	0	NaN	4	20170126		
...	
34398	Wwd/cudKVuLJ3txRVxlg2Zaeliu+LRUfiBmfrnxhRCY=	1	0	NaN	7	20131111		
34399	g3JGnJX6Hg50IFbrNWfsHwCUmAplkiv2M8sXOaeXolQ=	4	18	male	3	20141024		
34400	IMaPMJuyN+ip9Vqi+z2XuXbFAP2kbHr+EvvCNkFfj+o=	1	0	NaN	7	20130802		
34401	WAnCAJjUty9Stv8yKtV7ZC7PN+ilOy5FX3aUgGPANM=	1	0	NaN	7	20151020		
34402	xH8KpzKGeNNq6dOvy51c/8VzqOiGG+m6vabhsPSDHX4=	1	0	NaN	4	20160815		

34403 rows × 7 columns

```
In [3]: songs=pd.read_csv("songs.csv", nrows=20000)
songs
```

Out[3]:

		song_id	song_length	genre_ids	artist_name	composer	lyricis
0	CXoTN1eb7Al+DntdU1vbcwGRV4SCIDxZu+YD8JP8r4E=		247640	465	張信哲 (Jeff Chang)	董貞	何啟弘
1	o0kFgae9QtnYgRkVPqLJwa05zlhRIUjff7O1tDw0ZDU=		197328	444	BLACKPINK	TEDDY FUTURE BOUNCE Bekuh BOOM	TEDDY
2	DwVvVurfpuZ+XPuFvuccIVQEYpqcPqUkHR0ne1RQzPs0=		231781	465	SUPER JUNIOR	NaN	NaN
3	dKMBWoZyScdxSkihKG+Vf47nc18N9q4m58+b4e7dSSE=		273554	465	S.H.E	湯小康	徐世珍
4	W3bqWd3T+VeHFzHAUfARgW9AvVRaF4N5Yzm4Mr6Eo/o=		140329	726	貴族精選	Traditional	Traditional
...
19995	XTDNdQR/VbqECrUmXlmyeOnhD4dFglDefCw/auQ/mrU=		363946	958	Rachel Podger	Heinrich Ignaz Franz von Biber	NaN
19996	iUWEK/CODxzJtYSPUIp/0SM5yUd8RBrAZeCPwJFu/+c=		319712	958	Various Artists	Johann Sebastian Bach	NaN
19997	ljBHnpgdxRnzxO0IJoivVZdjIDZEUGjOvvVhLKCxwNY=		214274	958	Mozart	NaN	NaN
19998	OOowMAm1BHvDzH0xt33+heZkv2lnWK2sffo9kugb9zU=		223425	465	Jorge Ben Jor	Jorge Ben Jor	NaN
19999	J4QBnnRehImXlQn3wBXPOe91rw5yKabW9Ex0lzB8EL4=		197369	451	Various Artists	Michael Lai	NaN

20000 rows × 7 columns

In [4]:

```
songs_info=pd.read_csv("song_extra_info.csv")
songs_info
```

Out[4]:

	song_id	name	isrc
0	LP7pLJoJFBvyuUwvu+oLzjT+bl+UeBPURCecJsX1jjs=	我們	TWUM71200043
1	ClazTFnk6r0Bnuie44bocdNMM3rdlrq0bCGAsGUWcHE=	Let Me Love You	QMZSY1600015
2	u2ja/bZE3zhCGxvbbOB3zOoUjx27u40cf5g09UXMokQ=	原諒我	TWA530887303
3	92Fqsy0+p6+RHe2EoLKjHahORHR1Kq1TBJoCIW9v+Ts=	Classic	USSM11301446
4	0QFmz/+rJy1Q56C1DuYqT9hKKqi5TUqx0sN0lwvoHrw=	愛投羅網	TWA471306001
...
2295966	hLnetpF6UbPg28sSfXnPE2vsdaGsLvddIXEdJR4VTIA=	Deep Breathing	PLL431720793
2295967	N+6vJ8actKQm0S3Fpf4elipTjoAo9ev28aA5FJN5e40=	In Hiding	US5UL1519827
2295968	pv35uG0ts05mWtirM/AMOWEzbHxIVart5ZzRXqKUY1c=	Il Est Ne Le Divin Enfant	PLL431502294
2295969	QSySnm8jt2Go7byY34/PxsZP6dPCins2j2cyYquNhBo=	The Exodus Song	DEPZ69316095
2295970	DYJKJSgDOKxb19XzOVO81176qTH0OIHCsfzFRm/BG+g=	Like This	US5UL1512426

2295971 rows × 3 columns

In [5]:

```
submission=pd.read_csv("sample_submission.csv", nrows=20000)
submission
```

Out[5]:

	id	target
0	0	0.5
1	1	0.5
2	2	0.5
3	3	0.5
4	4	0.5
...
19995	19995	0.5
19996	19996	0.5
19997	19997	0.5
19998	19998	0.5
19999	19999	0.5

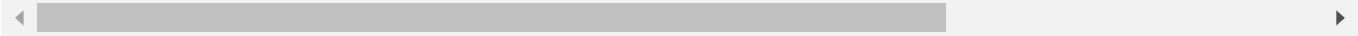
20000 rows × 2 columns

```
In [6]: train_data=pd.read_csv("train.csv", nrows=20000)
train_data
```

Out[6]:

	msno	song_id	source_sy
0	FGtllVqz18RPIwJj/edr2gV78zirAiY/9SmYvia+kCg=	BBzumQNXUHKdEBOB7mAjuzok+IJA1c2Ryg/yzTF6tik=	
1	Xumu+NljS6QYVxDS4/t3SawvJ7viT9hPKXmf0RtLNx8=	bhp/MpSNoqoxOIB+/l8WPQu6jldth4DlpCm3ayXnJqM=	
2	Xumu+NljS6QYVxDS4/t3SawvJ7viT9hPKXmf0RtLNx8=	JNWfrrC7zNN7BdMpslSKa4Mw+xVJYNnxXh3/Epw7QgY=	
3	Xumu+NljS6QYVxDS4/t3SawvJ7viT9hPKXmf0RtLNx8=	2A87tzfnJTSWqD7glZHisolhe4DMdzkbd6LzO1KHjNs=	
4	FGtllVqz18RPIwJj/edr2gV78zirAiY/9SmYvia+kCg=	3qm6XTZ6MOCU11x8FIVbAGH5I5uMkT3/ZalWG1oo2Gc=	
...	
19995	N9u0iiKsqZYNdml12834pcylc7xkifUUHyrb69T0jaU=	NGGXOVTfxaeWP5FCG4FqEXThMN5oArLN3V6gG/XFBnY=	
19996	N9u0iiKsqZYNdml12834pcylc7xkifUUHyrb69T0jaU=	KVcvULyaMxyWdn3ywjZifiGJqkaT6uUKMBLZ+BTsB7Q=	
19997	N9u0iiKsqZYNdml12834pcylc7xkifUUHyrb69T0jaU=	+ns7TUfsDgumML8q2hVjpi+B3dDLB/YlrEDoLuSmlKI=	
19998	N9u0iiKsqZYNdml12834pcylc7xkifUUHyrb69T0jaU=	xf3Py8deCPXun3qc83fyceiXCJ/qZw7pfHxD1x3SvgY=	
19999	N9u0iiKsqZYNdml12834pcylc7xkifUUHyrb69T0jaU=	lxqk50t+WoPjIXelxSVEsKaMJpShQqEq8Sq0cVzVk2A=	

20000 rows × 6 columns



```
In [7]: test_data=pd.read_csv("test.csv", nrows=20000)
test_data
```

Out[7]:	id	msno	song_id
0	0	V8rui7SGk7tDm3zA51DPpn6quutt+vmKMBKa21dp54uM=	WmHKgKMIp1IQMecNdNvDMkvlycZYHnFwDT72I5slssc=
1	1	V8rui7SGk7tDm3zA51DPpn6quutt+vmKMBKa21dp54uM=	y/rsZ9DC7Fwk5F2PK2D5mj+aOBUJAjuu3dZ14NgE0vM=
2	2	/uQAIrAkaczV+nWCd2sPF2ekvXPRipV7q0l+gbLuxjw=	8eZLF0dGVdXBSqoAv5nsLigeH2BvKXzTQYtUM53I0k4=
3	3	1a6oo/iXKatxQx4eS9zTVD+KISVaAFbTlqVvwLC1Y0k=	ztCf8thYsS4YN3GclL/bvoxLm/T5mYBVKOO4C9NiVfQ=
4	4	1a6oo/iXKatxQx4eS9zTVD+KISVaAFbTlqVvwLC1Y0k=	MKVMpsIKcQhMaFEgcEQhEfi5+RZhMYIU3eRDpySrH8Y=
...
19995	19995	g2ZlslHl2Mheh31zqY9cbgx9MKizUzskgEcYUBhuExys=	F/zl2VkhgQs+Lx+XjS74XN1m59vNAVir/SI11wc8Fr4=
19996	19996	g2ZlslHl2Mheh31zqY9cbgx9MKizUzskgEcYUBhuExys=	vM08WBQRO9eZo1K+qTJmjuw2lqbuA3L65ojbGwB4GI0=
19997	19997	g2ZlslHl2Mheh31zqY9cbgx9MKizUzskgEcYUBhuExys=	61cwHmq3kaaSf/yMvcEXUeGmPyG1g8y7am/0fuECBw=
19998	19998	g2ZlslHl2Mheh31zqY9cbgx9MKizUzskgEcYUBhuExys=	icCxTviW2hBsVijNHZnddwcvVi+PE7ywbQEPidLt/4=
19999	19999	g2ZlslHl2Mheh31zqY9cbgx9MKizUzskgEcYUBhuExys=	UHbrHH97KESebiQOL3/2fLOLCX558fZ4BlqmTNYUuqg=

20000 rows × 6 columns

```
In [8]: print(f"The songs_data has {songs.shape[0]} rows and {songs.shape[1]} columns")
print(f"The songs_extra_info_data has {songs_info.shape[0]} rows and {songs_info.shape[1]} columns")
print(f"The members_data has {members.shape[0]} rows and {members.shape[1]} columns")
print(f"The sample_submission_data has {submission.shape[0]} rows and {submission.shape[1]} columns")
print(f"The train_data has {train_data.shape[0]} rows and {train_data.shape[1]} columns")
print(f"The test_data has {test_data.shape[0]} rows and {test_data.shape[1]} columns")
```

The songs_data has 20000 rows and 7 columns
The songs_extra_info_data has 2295971 rows and 3 columns
The members_data has 34403 rows and 7 columns
The sample_submission_data has 20000 rows and 2 columns
The train_data has 20000 rows and 6 columns
The test_data has 20000 rows and 6 columns

```
In [9]: songs.describe()
```

Out[9]:	song_length	language
count	2.000000e+04	20000.000000
mean	2.456958e+05	25.946550
std	1.201716e+05	23.223231
min	4.922000e+03	-1.000000
25%	1.997060e+05	3.000000
50%	2.336850e+05	17.000000
75%	2.731360e+05	52.000000
max	4.025318e+06	59.000000

```
In [10]: print("Columns present in the songs data are:")
for columns in songs.columns:
    print(columns)
```

Columns present in the songs data are:
song_id
song_length
genre_ids
artist_name
composer
lyricist
language

```
In [11]: print(f"Number of records : {songs.shape[0]}")
print(f"Count of distinct song lengths : {len(songs.song_length.unique())}")
```

```

print(f"Count of distinct genre ids : {len(songs.genre_ids.unique())}")
print(f"Count of distinct artist name : {len(songs.artist_name.unique())}")
print(f"Count of distinct composer : {len(songs.composer.unique())}")
print(f"Count of distinct lyricist : {len(songs.lyricist.unique())}")
print(f"Count of distinct language : {len(songs.language.unique())}")

```

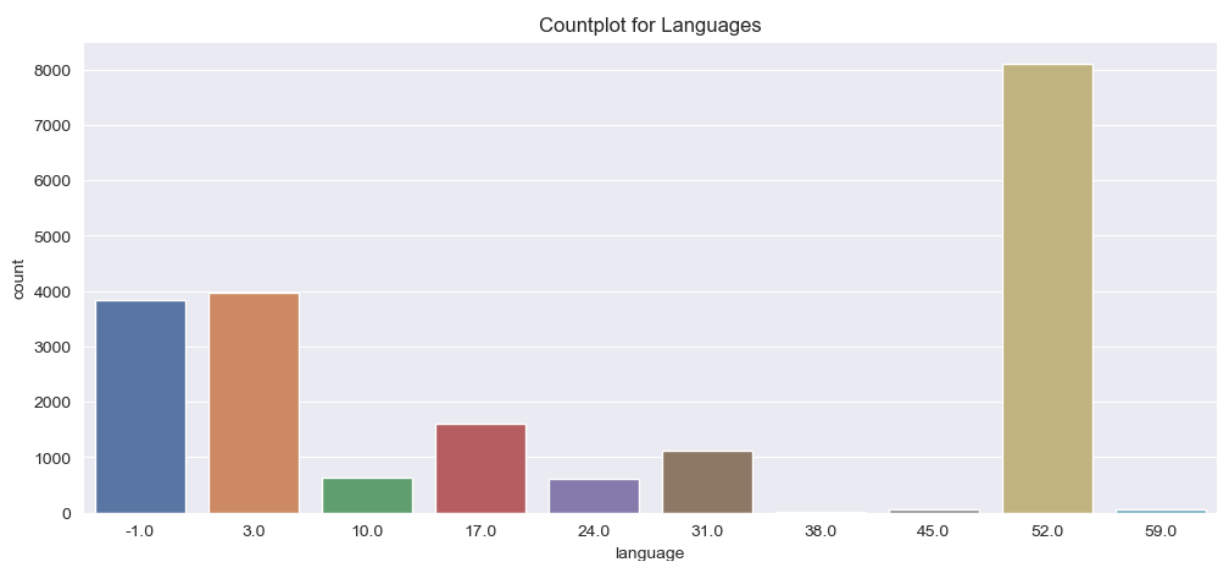
Number of records : 20000
 Count of distinct song lengths : 10734
 Count of distinct genre ids : 275
 Count of distinct artist name : 8378
 Count of distinct composer : 8332
 Count of distinct lyricist : 3977
 Count of distinct language : 10

Data preprocessing

```

In [12]: plt.figure(figsize= (12, 5))
sns.set_style("darkgrid")
ax = sns.countplot(x = songs.language, data = songs, palette="deep")
ax.set_title("Countplot for Languages")
plt.show()

```



```

In [13]: print("Columns present in the Members Data are:")
for columns in members.columns:
    print(columns)

```

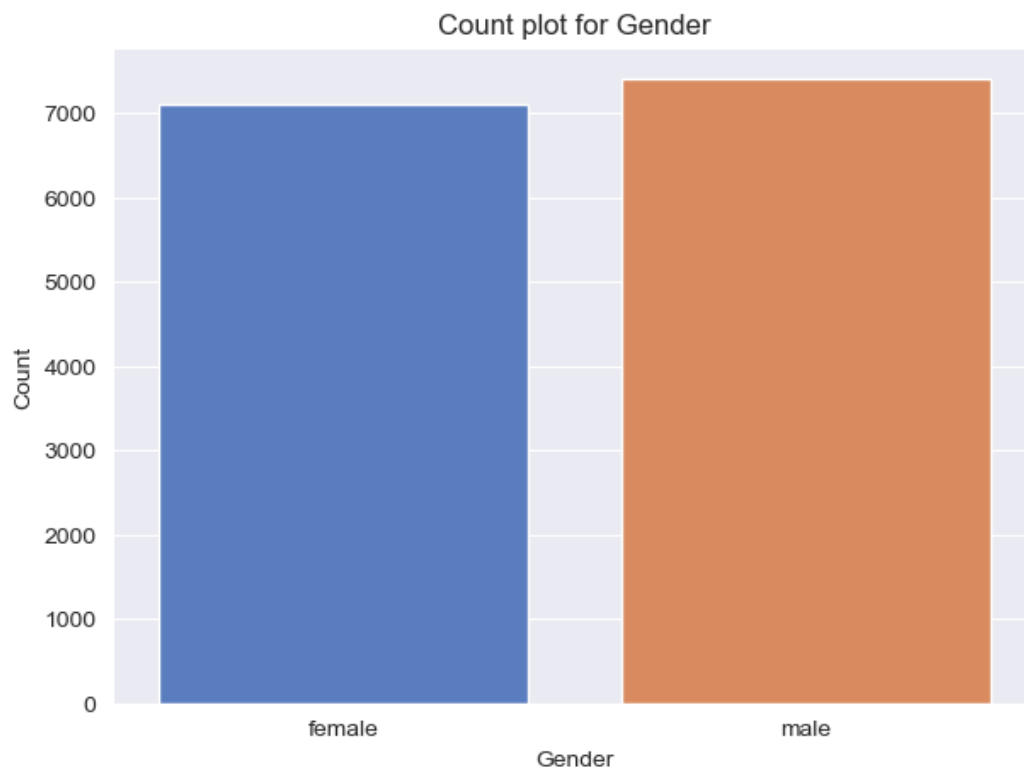
Columns present in the Members Data are:
 msno
 city
 bd
 gender
 registered_via
 registration_init_time
 expiration_date

```

In [14]: plt.figure(figsize= (7, 5))
sns.set_style("darkgrid")
sns.countplot(x='gender', data=members, palette="muted")
plt.xlabel("Gender")
plt.ylabel("Count")
plt.title("Count plot for Gender")

```

Out[14]: Text(0.5, 1.0, 'Count plot for Gender')



```
In [15]: plt.figure(figsize= (7 ,5))
sns.countplot(x="registered_via", data=members, palette="muted")
plt.xlabel("Registration Method")
plt.ylabel("Count")
plt.title("Count plot for Registration Method")
```

Out[15]: Text(0.5, 1.0, 'Count plot for Registration Method')



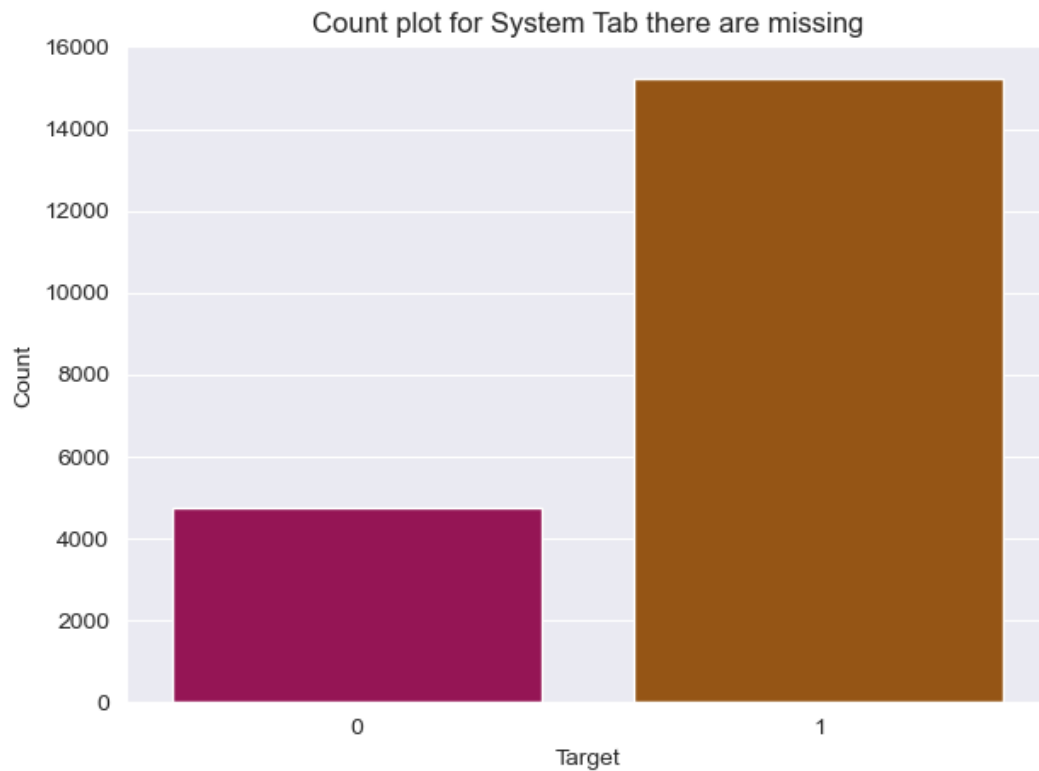
```
In [16]: print(f"Total number of records : {train_data.shape[0]}")
```

Total number of records : 20000

```
In [17]: plt.figure(figsize= (7, 5))
sns.countplot(x='target', data=train_data, palette='brg')
plt.xlabel("Target")
```

```
plt.ylabel("Count")
plt.title("Count plot for System Tab there are missing")
```

Out[17]: Text(0.5, 1.0, 'Count plot for System Tab there are missing')



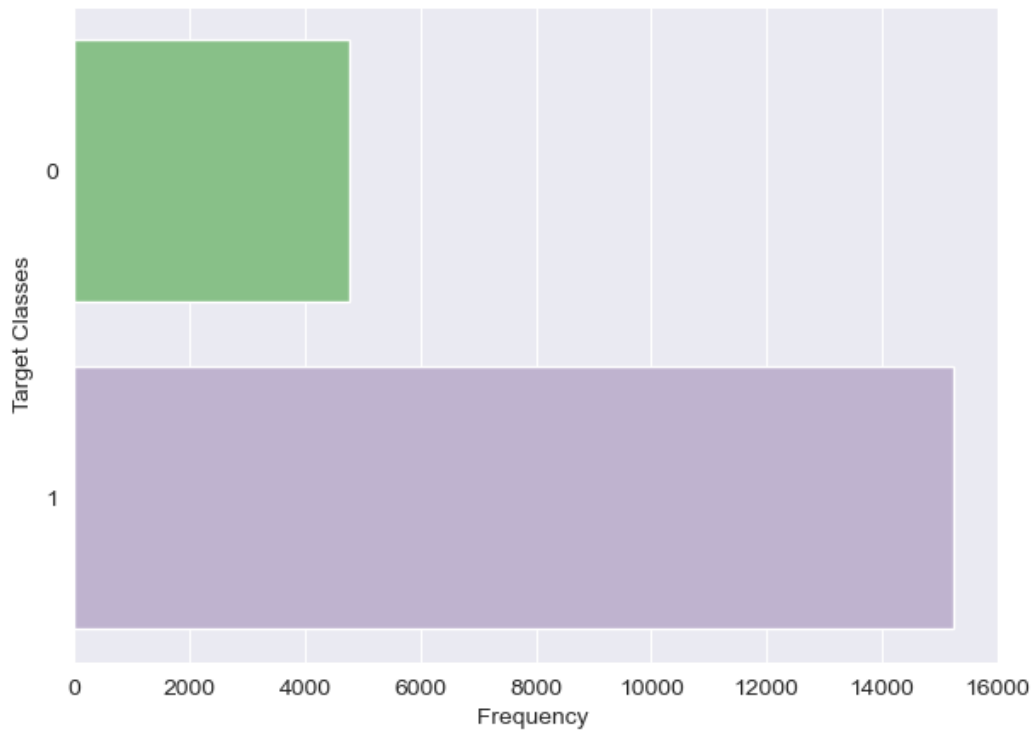
```
In [18]: print("Total percentage for NaN value in target column : ", (train_data["target"].isna().sum()/len(t
```

Total percentage for NaN value in target column : 0.0 %

```
In [19]: duplicate_value1 = len(train_data["song_id"])-train_data["song_id"].nunique()
print("Total number of duplicate song id : ", duplicate_value1)
print("Total percentage of duplicate song id : ", (duplicate_value1/len(train_data["song_id"]))*100,
```

Total number of duplicate song id : 10187
Total percentage of duplicate song id : 50.934999999999995 %

```
In [20]: plt.figure(figsize=(7, 5))
sns.countplot(y=train_data["target"], data=train_data, palette="Accent")
plt.ylabel("Target Classes")
plt.xlabel("Frequency ")
plt.show()
```



```
In [21]: songs_info.head()
```

```
Out[21]:
```

	song_id	name	isrc
0	LP7pLJoJFBvyuUwvu+oLzjT+bl+UeBPURCecJsX1jjs=	我們	TWUM71200043
1	ClazTFnk6r0Bnuie44bocdNMM3rdlrq0bCGAsGUWcHe=	Let Me Love You	QMZSY1600015
2	u2ja/bZE3zhCGxvbbOB3zOoUjx27u40cf5g09UXMoKQ=	原諒我	TWA530887303
3	92Fqsy0+p6+RHe2EoLKjHahORHR1Kq1TBJoCIW9v+Ts=	Classic	USSM11301446
4	0QFmz/+rJy1Q56C1DuYqT9hKKqi5TUqx0sN0lwvoHrw=	愛投羅網	TWA471306001

```
In [22]: songs_info.isnull().sum()
```

```
Out[22]: song_id      0
name          2
isrc        136548
dtype: int64
```

```
In [23]: songs.isnull().sum()
```

```
Out[23]: song_id      0
song_length  0
genre_ids    346
artist_name  0
composer    8382
lyricist    14332
language    0
dtype: int64
```

```
In [24]: songs['genre_ids'].fillna(' ', inplace=True)
```

```
In [25]: songs['composer'].fillna(' ', inplace=True)
songs['lyricist'].fillna(' ', inplace=True)
songs['language'].fillna((52.0), inplace=True)
```

```
In [26]: songs.isnull().sum()
```



```
Out[26]: song_id      0
song_length  0
genre_ids    0
artist_name  0
composer     0
lyricist     0
language     0
dtype: int64
```

```
In [27]: train_data.isnull().sum()
```

```
Out[27]: msno          0
song_id      0
source_system_tab    67
source_screen_name   576
source_type         50
target           0
dtype: int64
```

```
In [28]: train_data = train_data.drop(['source_system_tab', 'source_screen_name', 'source_type'], axis=1)
train_data.head()
```

```
Out[28]:
```

	msno	song_id	target
0	FGtllVqz18RPiwJj/edr2gV78zirAiY/9SmYvia+kCg=	BBzumQNXUHKdEBOB7mAJuzok+IJA1c2Ryg/yzTF6tik=	1
1	Xumu+NljS6QYVxDS4/t3SawvJ7viT9hPKXmf0RtLNx8=	bhp/MpSNoqoxOIB+/l8WPqu6jldth4DlpCm3ayXnJqM=	1
2	Xumu+NljS6QYVxDS4/t3SawvJ7viT9hPKXmf0RtLNx8=	JNWfrrC7zNN7BdMpslSKa4Mw+xVJYNnxXh3/Epw7QgY=	1
3	Xumu+NljS6QYVxDS4/t3SawvJ7viT9hPKXmf0RtLNx8=	2A87tzfnJTSWqD7glZHisolhe4DMdzkdb6LzO1KHjNs=	1
4	FGtllVqz18RPiwJj/edr2gV78zirAiY/9SmYvia+kCg=	3qm6XTZ6MOCU11x8FIVbAGH5I5uMkT3/ZalWG1oo2Gc=	1

```
In [29]: train_data.shape
```

```
Out[29]: (20000, 3)
```

```
In [30]: train_data.rename(columns={'msno': "user_id"}, inplace=True)
train_data.head()
```

```
Out[30]:
```

	user_id	song_id	target
0	FGtllVqz18RPiwJj/edr2gV78zirAiY/9SmYvia+kCg=	BBzumQNXUHKdEBOB7mAJuzok+IJA1c2Ryg/yzTF6tik=	1
1	Xumu+NljS6QYVxDS4/t3SawvJ7viT9hPKXmf0RtLNx8=	bhp/MpSNoqoxOIB+/l8WPqu6jldth4DlpCm3ayXnJqM=	1
2	Xumu+NljS6QYVxDS4/t3SawvJ7viT9hPKXmf0RtLNx8=	JNWfrrC7zNN7BdMpslSKa4Mw+xVJYNnxXh3/Epw7QgY=	1
3	Xumu+NljS6QYVxDS4/t3SawvJ7viT9hPKXmf0RtLNx8=	2A87tzfnJTSWqD7glZHisolhe4DMdzkdb6LzO1KHjNs=	1
4	FGtllVqz18RPiwJj/edr2gV78zirAiY/9SmYvia+kCg=	3qm6XTZ6MOCU11x8FIVbAGH5I5uMkT3/ZalWG1oo2Gc=	1

```
In [31]: songs.head()
```

```
Out[31]:
```

	song_id	song_length	genre_ids	artist_name	composer	lyricist	la
0	CXoTN1eb7Al+DntdU1vbcwGRV4SCIDxZu+YD8JP8r4E=	247640	465	張信哲 (Jeff Chang)	董貞	何啟弘	
1	o0kFgae9QtnYgRkVPqLJwa05zlhRIUjff7O1tDw0ZDU=	197328	444	BLACKPINK	TEDDY FUTURE BOUNCE Bekuh BOOM	TEDDY	
2	DwVvVurfpuZ+XPuFvuclVQEYpqcPukHR0ne1RQzPs0=	231781	465	SUPER JUNIOR			
3	dKMBWoZyScdxSkihKG+Vf47nc18N9q4m58+b4e7dSSE=	273554	465	S.H.E	湯小康	徐世珍	
4	W3bqWd3T+VeHFzHAUfARgW9AvVRaF4N5Yzm4Mr6Eo/o=	140329	726	貴族精選	Traditional	Traditional	

```
In [32]: df = train_data.merge(songs, on="song_id")
df.head()
```

```
Out[32]:
```

	user_id	song_id	target	song_length
0	FGtIlVqz18RPIwJj/edr2gV78zirAiY/9SmYvia+kCg=	3Hg5kugV1S0wzEVLAefqjIV5UHzb7bCrdBRQIGygLvU=	1	24
1	hZyOA+0yqCIPLt6ulEndf8fG8szH/95eKMbaxLE5z30=	3Hg5kugV1S0wzEVLAefqjIV5UHzb7bCrdBRQIGygLvU=	1	24
2	0LhkakIQDn36HZXI6CIQSO7W7jpkZAY+9MvYgPOZGrA=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	1	24
3	MofmAMt7P8LlCF4+LLlcjYlhYUzmv13L/LRwYFxiGYE=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	1	24
4	U9Z+N+szYGJHTPMn/C0V7ylyIC24fDI0RDRWChXATkg=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	0	24

```
In [33]: df = df.drop(['song_length', 'language'], axis=1)
df.head()
```

```
Out[33]:
```

	user_id	song_id	target	genre_id
0	FGtIlVqz18RPIwJj/edr2gV78zirAiY/9SmYvia+kCg=	3Hg5kugV1S0wzEVLAefqjIV5UHzb7bCrdBRQIGygLvU=	1	12
1	hZyOA+0yqCIPLt6ulEndf8fG8szH/95eKMbaxLE5z30=	3Hg5kugV1S0wzEVLAefqjIV5UHzb7bCrdBRQIGygLvU=	1	12
2	0LhkakIQDn36HZXI6CIQSO7W7jpkZAY+9MvYgPOZGrA=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	1	4
3	MofmAMt7P8LlCF4+LLlcjYlhYUzmv13L/LRwYFxiGYE=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	1	4
4	U9Z+N+szYGJHTPMn/C0V7ylyIC24fDI0RDRWChXATkg=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	0	4

```
In [34]: songs_info.head()
```

```
Out[34]:
```

	song_id	name	isrc
0	LP7pLJoJFBvyuUwvu+oLzjT+bl+UeBPURCecJsX1jjs=	我們	TWUM71200043
1	ClazTFnk6r0Bnuie44bocdNMM3rdlrq0bCGAsGUWcHE=	Let Me Love You	QMZSY1600015
2	u2ja/bZE3zhCGxvbbOB3zOoUjx27u40cf5g09UXMoKQ=	原諒我	TWA530887303
3	92Fqsy0+p6+RHe2EoLKjHahORHR1Kq1TBJoCIW9v+Ts=	Classic	USSM11301446
4	0QFmz/+rJy1Q56C1DuYqT9hKKqi5TUqx0sN0lwvoHrw=	愛投羅網	TWA471306001

```
In [35]: df = df.merge(songs_info,on="song_id").drop('isrc',axis=1)
df.head()
```

Out[35]:

		user_id	song_id	target	genre_i
0	FGtIlVqz18RPIwJj/edr2gV78zirAiY/9SmYvia+kCg=	3Hg5kugV1S0wzEVLAEfqiV5UHzb7bCrdBRQIGygLvU=		1	12
1	hZyOA+0yqCIPLt6ulEndf8fG8szH/95eKMbaxLE5z30=	3Hg5kugV1S0wzEVLAEfqiV5UHzb7bCrdBRQIGygLvU=		1	12
2	0LhkakIQDn36HZXI6CIQSO7W7jpkZay+9MvYgPOZGrA=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=		1	4
3	MofmAMt7P8LlcF4+LLlcjylhYUzmv13L/LRwYFxiGYE=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=		1	4
4	U9Z+N+szYGJHTPMn/C0V7ylylC24fDI0RDRWChXATkg=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=		0	4

```
In [36]: df.rename(columns={'name':'song_name'}, inplace=True)
df.head()
```

Out[36]:

		user_id	song_id	target	genre_i
0	FGtIlVqz18RPIwJj/edr2gV78zirAiY/9SmYvia+kCg=	3Hg5kugV1S0wzEVLAEfqiV5UHzb7bCrdBRQIGygLvU=	1	12	
1	hZyOA+0yqCIPLt6ulEndf8fG8szH/95eKMbaxLE5z30=	3Hg5kugV1S0wzEVLAEfqiV5UHzb7bCrdBRQIGygLvU=	1	12	
2	0LhkakIQDn36HZXI6CIQSO7W7jpkZAY+9MvYgPOZGrA=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	1	4	
3	MofmAMt7P8LlcF4+LLlcjYlhYUzmv13L/LRwYFxiGYE=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	1	4	
4	U9Z+N+szYGJHTPMn/C0V7ylylC24fDI0RDRWChXATkg=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	0	4	

Data cleaning

```
In [37]: df['genre_ids'].value_counts()
```

```

Out[37]:
465          710
458          400
444           65
1609          57
921           49
359           32
          26
139           24
2022           21
1259           20
2122           13
139|125|109      7
726              7
451              7
437              7
958              6
786|947           6
465|1259           4
1011              4
786              4
947              4
691              3
921|465           3
430              3
921|458           2
458|1287          2
698              2
444|1259          2
829              2
850              2
1152             1
880|458           1
465|829           1
864|857|850|843   1
465|798           1
474              1
864|850|726|857|843 1
388              1
864|786|850|857|843 1
940              1
1609|465          1
465|2122          1
423              1
726|242           1
437|850           1
Name: genre_ids, dtype: int64

```

```

In [38]: df['genre_ids']=df['genre_ids'].str.replace('|', ' ', regex=True)
df['genre_ids'].value_counts()

```

```

Out[38]: 465          710
         458          400
         444           65
        1609           57
         921           49
        359           32
           26
        139           24
       2022           21
       1259           20
       2122           13
       139 125 109           7
        726           7
        451           7
        437           7
       958           6
       786 947           6
       465 1259           4
       1011           4
        786           4
        947           4
        691           3
       921 465           3
        430           3
       921 458           2
       458 1287           2
        698           2
       444 1259           2
        829           2
       850           2
       1152           1
       880 458           1
       465 829           1
       864 857 850 843           1
       465 798           1
        474           1
       864 850 726 857 843           1
       388           1
       864 786 850 857 843           1
        940           1
       1609 465           1
       465 2122           1
        423           1
       726 242           1
       437 850           1
Name: genre_ids, dtype: int64

```

```

In [39]: df['artist_name']=df['artist_name'].str.replace('|', ' ', regex=True)
         df['composer']=df['composer'].str.replace('/', ' ', regex=True)
         df['lyricist']=df['lyricist'].str.replace('/', ' ', regex=True)
         df['artist_name']=df['artist_name'].str.lower()
         df['composer']=df['composer'].str.lower()
         df['lyricist']=df['lyricist'].str.lower()

```

```

In [42]: df['songs_details']=df['artist_name']+' '+df['composer']+df['lyricist']
         df.head()

```

Out[42]:

		user_id	song_id	target	genre_i
0	FGtIlVqz18RPIwJj/edr2gV78zirAiY/9SmYvia+kCg=	3Hg5kugV1S0wzEVLAefqjIV5UHzb7bCrdBRQIGygLvU=	1	12	
1	hZyOA+0yqCIPLt6ulEndf8fG8szH/95eKMbaxLE5z30=	3Hg5kugV1S0wzEVLAefqjIV5UHzb7bCrdBRQIGygLvU=	1	12	
2	0LhkakIQDn36HZXI6CIQSO7W7jkpZAY+9MvYgPOZGrA=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	1	4	
3	MofmAMt7P8LlcF4+LLlcjYlhYUzmv13L/LRwYFxiGYE=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	1	4	
4	U9Z+N+szYGJHTPMn/C0V7yllylC24fDI0RDRWChXATkg=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	0	4	

In [43]: `df.user_id.value_counts()`

```

Out[43]: EozJegFxTFIwDb9aJ708kSUHAgx4ZIVqf7IuN5Zck50=    19
          V5U4EGk2kaSKaUGSwhU6g3HBefxf1EvAy1vWPu6UBQs=    18
          Bwg9yS76qujJJekSYSzfJrM1kjK5Ui7KFkgUcjuXRCg=    12
          W9NYSCff57nmfyYCiX6IbW0/G3YuwC18h/rld+BGxMY=    11
          Uz1Qoa9tdrcpYdh4wksoh+SpwCFcKvRGPA+xLNqghmo=    11
          ..
          rb7TT328utsdnd8C0yhstig0zciXIURo7M464E60EHg=     1
          hSn7jMfIURFu+1W3PDIDTxbhM5SxRg9VFRoH23Rm2Ic=     1
          yrMfQXudhDaA/bOePZtkKErbjZc5pALG79FHPayEy5U=     1
          iP3eF1In0rH61CfGvmwVYj4CgFcQQ0iVZG7MBA+Plgo=     1
          j2Sx5B7BrjqCiT3ZwWk4AvepwM14QEalhTPi2/sgdG4=     1
          Name: user_id, Length: 975, dtype: int64

```

In [45]: `df.duplicated().sum()`

Out[45]: 0

In [46]: *#Creating a copy file before performing a similarity*
`main_df=df.copy()`
`main_df.head()`

Out[46]:

		user_id	song_id	target	genre_i
0	FGtIlVqz18RPIwJj/edr2gV78zirAiY/9SmYvia+kCg=	3Hg5kugV1S0wzEVLAefqjIV5UHzb7bCrdBRQIGygLvU=	1	12	
1	hZyOA+0yqCIPLt6ulEndf8fG8szH/95eKMbaxLE5z30=	3Hg5kugV1S0wzEVLAefqjIV5UHzb7bCrdBRQIGygLvU=	1	12	
2	0LhkakIQDn36HZXI6CIQSO7W7jkpZAy+9MvYgPOZGrA=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	1	4	
3	MofmAMt7P8LlcF4+LLlcjYlhYUzmv13L/LRwYFxiGYE=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	1	4	
4	U9Z+N+szYGJHTPMn/C0V7yllylC24fDI0RDRWChXATkg=	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	0	4	

In [47]: `main_df.songs_details.duplicated().sum()`

Out[47]: 889

In [49]: `main_df.shape`

Out[49]: (1509, 9)

```
In [50]: main_df.duplicated().sum()
```

Out[50]: 0

```
In [51]: main_df=main_df.drop(['user_id'], axis=1)
```

```
In [52]: main_df
```

Out[52]:

	song_id	target	genre_ids	artist_name	composer	lyricist	song_name
--	---------	--------	-----------	-------------	----------	----------	-----------

0	3Hg5kugV1S0wzEVLAefqjIV5UHzb7bCrdBRQIGygLvU=	1	1259	designer	sidney selby adnan khan		Pan
1	3Hg5kugV1S0wzEVLAefqjIV5UHzb7bCrdBRQIGygLvU=	1	1259	designer	sidney selby adnan khan		Pan
2	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	1	458	莊心妍	鄭建浩	鄭建浩	我過的很
3	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	1	458	莊心妍	鄭建浩	鄭建浩	我過的很
4	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	0	458	莊心妍	鄭建浩	鄭建浩	我過的很
...
1504	gtenKB6Uz9z5MnC8GlvaDSyW+6m6JhmgRBoFc/Jin2U=	1	465	various artists	jung joonil	jung joonil	Fine C
1505	7kGd6s2v5Ywl4fsESa10lIGKkGE+V0QtWGhwiwNTPao=	1	465	郭靜 (claire kuo)	木蘭號aka 陳韋伶	木蘭號 aka 陳韋伶	我不是你那首情
1506	ceQpMUI3zi3wbvUuwa2gcOzzvCv6QoagUpKHU9dwJQU=	1	465	曾沛慈 (pets tseng)	梁正	葛大為 + 梁正	這裡還有
1507	Ny0HzjYum9lyotgPXzdRrcXhx20sFbpdSW68VRvtGfQ=	1	465	郭靜 (claire kuo)	陳小霞	姚若龍	在樹上唱
1508	NGGXOVTfxaeWP5FCG4FqEXThMN5oArLN3V6gG/XFBnY=	0	465	小樂 (吳思賢) (ben wu)	秦洋	姚若龍	最大的缺

1509 rows × 8 columns

```
In [53]: main_df.duplicated().sum()
```

Out[53]: 715

```
In [54]: main_df=main_df.drop_duplicates()
main_df
```

Out[54]:

	song_id	target	genre_ids	artist_name	composer	lyricist	song_name
0	3Hg5kugV1S0wzEVLAefqjIV5UHzb7bCrdBRQIGygLvU=	1	1259	designer	sidney selby adnan khan		Pan
2	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	1	458	莊心妍	鄭建浩	鄭建浩	我過的很
4	skehue/d/R59G71dXYpntDwdjRRPlweN3JE8g40TgZU=	0	458	莊心妍	鄭建浩	鄭建浩	我過的很
21	reXuGcEWDDCnL0K3Th//3DFG4S1ACSpJMzA+CFipo1g=	1	458	周湯豪 (nickthereal)	周湯豪	周湯豪 \崔惟楷	帥到分
73	reXuGcEWDDCnL0K3Th//3DFG4S1ACSpJMzA+CFipo1g=	0	458	周湯豪 (nickthereal)	周湯豪	周湯豪 \崔惟楷	帥到分
...
1504	gtenKB6Uz9z5MnC8GlvaDSyW+6m6JhmgRBoFc/Jin2U=	1	465	various artists	jung joonil	jung joonil	Fine D
1505	7kGd6s2v5Ywl4fsESa10lIGKkGE+V0QtWGhwiwNTPao=	1	465	郭靜 (claire kuo)	木蘭號aka 陳韋伶	木蘭號 aka陳韋伶	我不是你那首情
1506	ceQpMUI3zi3wbvUuwa2gcOzzvCv6QoagUpKHU9dwJQU=	1	465	曾沛慈 (pets tseng)	梁正	葛大為 +梁正	這裡還有
1507	Ny0HzjYum9lyotgPXzdRrcXhx20sFbpdSW68VRvtGfQ=	1	465	郭靜 (claire kuo)	陳小霞	姚若龍	在樹上唱
1508	NGGXOVTfxaeWP5FCG4FqEXThMN5oArLN3V6gG/XFBnY=	0	465	小樂 (吳思賢) (ben wu)	秦洋	姚若龍	最大的缺

794 rows × 8 columns



```
In [55]: main_df.reset_index(inplace=True)
```

```
In [57]: main_df.shape
```

Out[57]: (794, 9)

Mapping frequent words

```
In [59]: from sklearn.feature_extraction.text import TfidfVectorizer
tfidf=TfidfVectorizer(analyzer='word', stop_words='english')
tfidf_matrix=tfidf.fit_transform(main_df['songs_details'])
```

```
In [60]: tfidf_matrix
```

Out[60]: <794x1932 sparse matrix of type '<class 'numpy.float64'>' with 3696 stored elements in Compressed Sparse Row format>

Building Similarity

```
In [61]: from sklearn.metrics.pairwise import cosine_similarity
```



```

In [62]: cosine_similarity = cosine_similarity(tfidf_matrix)

In [63]: cosine_similarity

Out[63]: array([[1., 0., 0., ..., 0., 0., 0.],
 [0., 1., 1., ..., 0., 0., 0.],
 [0., 1., 1., ..., 0., 0., 0.],
 ...,
 [0., 0., 0., ..., 1., 0., 0.],
 [0., 0., 0., ..., 0., 1., 0.],
 [0., 0., 0., ..., 0., 0., 1.]])

In [64]: sorted(list(enumerate(cosine_similarity[0])), reverse=True, key=lambda x:x[1])[1:6]

Out[64]: [(658, 0.6015656934945277),
 (78, 0.0911295308657028),
 (1, 0.0),
 (2, 0.0),
 (3, 0.0)]

In [65]: #In which you can recommend only index
def recommend(song):
    song_index=main_df[main_df['song_name']==song].index[0]
    distances=cosine_similarity[song_index]
    song_list=sorted(list(enumerate(distances)), reverse=True, key=lambda x:x[1])[1:6]
    for i in song_list:
        print(i[0])

```

User based Recommender - Content

```

In [67]: def recommend(song):
    song_index=main_df[main_df['song_name']==song].index[0]
    distances=cosine_similarity[song_index]
    song_list=sorted(list(enumerate(distances)), reverse=True, key=lambda x:x[1])[1:10]
    for i in song_list:
        print(main_df.iloc[i[0]].song_name)

In [68]: recommend('Panda')

Tiimmy Turner
La La La
我過的很好
我過的很好
帥到分手
帥到分手
迷些路 (Lost On The Way)
迷些路 (Lost On The Way)
Bokurano Yume

```

Results:

The results of the Music Recommendation System project are highly dependent on the specific implementation, data preprocessing techniques, and model selection. By employing collaborative filtering or content-based filtering approaches, the recommendation system was able to provide personalized music recommendations to users. The system's recommendations aimed to enhance user engagement, satisfaction, and enjoyment of the music streaming platform.

You can find this project on [GitHub](#).