### **Spotify Data Analysis**

In [1]: import pandas as pd import numpy as np import matplotlib.pyplot as plt import seaborn as sns import warnings warnings.filterwarnings('ignore')

#### Out[2]:

	id	name	popularity	duration_ms	explicit	artists	
0	35iwgR4jXetl318WEWsa1Q	Carve	6	126903	0	['Uli']	[
1	021ht4sdgPcrDgSk7JTbKY	Capítulo 2.16 - Banquero Anarquista	0	98200	0	['Fernando Pessoa']	['14jtF
2	07A5yehtSnoedViJAZkNnc	Vivo para Quererte - Remasterizado	0	181640	0	['Ignacio Corsini']	['5LiO <sub>'</sub>
3	08FmqUhxtyLTn6pAh6bk45	El Prisionero - Remasterizado	0	176907	0	['Ignacio Corsini']	['5LiO
4	08y9GfoqCWfOGsKdwojr5e	Lady of the Evening	0	163080	0	['Dick Haymes']	['3E
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# **#Finding Null values**

## In [3]: pd.isnull(df\_tracks).sum()

Out[3]:	id	0
	name	71
	popularity	0
	duration_ms	0
	explicit	0
	artists	0
	id_artists	0
	release_date	0
	danceability	0
	energy	0
	key	0
	loudness	0
	mode	0
	speechiness	0
	acousticness	0
	instrumentalness	0
	liveness	0
	valence	0
	tempo	0
	time_signature	0
	dtvpe: int64	

# In [4]: df\_tracks.info();

<class 'pandas.core.frame.DataFrame'> RangeIndex: 586672 entries, 0 to 586671 Data columns (total 20 columns):

#	Column	Non-Null Count	Dtype
0	id	586672 non-null	object
1	name	586601 non-null	object
2	popularity	586672 non-null	int64
3	duration_ms	586672 non-null	int64
4	explicit	586672 non-null	int64
5	artists	586672 non-null	object
6	id_artists	586672 non-null	object
7	release_date	586672 non-null	object
8	danceability	586672 non-null	float64
9	energy	586672 non-null	float64
10	key	586672 non-null	int64
11	loudness	586672 non-null	float64
12	mode	586672 non-null	int64
13	speechiness	586672 non-null	float64
14	acousticness	586672 non-null	float64
15	instrumentalness	586672 non-null	float64
16	liveness	586672 non-null	float64
17	valence	586672 non-null	float64
18	tempo	586672 non-null	float64
19	time_signature	586672 non-null	int64
dtyp	es: float64(9), in	t64(6), object(5	)

memory usage: 89.5+ MB

In [5]: sorted\_df = df\_tracks.sort\_values('popularity',ascending=**True**).head(10) sorted\_df

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	id	name	popularity	duration_ms	explicit	artists	
546130	181rTRhCcggZPwP2TUcVqm	Newspaper Reports On Abner, 20 February 1935	0	896575	0	['Norris Goff', 'Chester Lauck', 'Carlton Bric	['3'
546222	0yOCz3V5KMm8l1T8EFc60i	恋は <b>水</b> の 上で	0	188440	0	['Hibari Misora']	[
546221	0y48Hhwe52099UqYjegRCO	私の誕生 日	0	173467	0	['Hibari Misora']	[
546220	0xCmgtf9ka07hkZg3D6PaV	エル・チョクロ (EL CHOCLO)	0	205280	0	['Hibari Misora']	[
546219	0tBXS3VuCPX7KWUFH2nros	<b>恋は不思</b> 議なもの	0	185733	0	['Hibari Misora']	[
546218	0qrKnQtYDVJhKFAXTHYVS9	ゆうべは どうした の (WHATSA MALLA U)	0	183427	0	['Hibari Misora']	[
546217	0nqsDxOeKSwEzp3AUQAAqS	Screen Director's Playhouse, Music For Million	0	1767071	0	['Wilms Herbert', 'June Allyson', 'Joseph Kear	['2r
546216	0kGEdsxVLYjCdfxM9tbezd	ブルーマン ボ	0	162147	0	['Hibari Misora']	[
546215	0bc3PUZurUUXrY7yqoOxjq	Screen Director's Playhouse, Trade Winds direc	0	1776652	0	['Wally Maher', 'Tay Garnett', 'Lurene Tuttle'	
546214	0Wwm0ruSjYMliWG0nyAl1F	Screen Director's Playhouse, It's A Wonderful 	0	1767576	0	['Joseph Granby', 'Jimmy Stewart', 'Irene Tedr	]

In [6]: df\_tracks.describe().transpose()

# Out[6]:

	count	mean	std	min	25%	50%
popularity	586672.0	27.570053	18.370642	0.0	13.0000	27.000000
duration_ms	586672.0	230051.167286	126526.087418	3344.0	175093.0000	214893.000000
explicit	586672.0	0.044086	0.205286	0.0	0.0000	0.000000
danceability	586672.0	0.563594	0.166103	0.0	0.4530	0.577000
energy	586672.0	0.542036	0.251923	0.0	0.3430	0.549000
key	586672.0	5.221603	3.519423	0.0	2.0000	5.000000
loudness	586672.0	-10.206067	5.089328	-60.0	-12.8910	-9.243000
mode	586672.0	0.658797	0.474114	0.0	0.0000	1.000000
speechiness	586672.0	0.104864	0.179893	0.0	0.0340	0.044300
acousticness	586672.0	0.449863	0.348837	0.0	0.0969	0.422000
instrumentalness	586672.0	0.113451	0.266868	0.0	0.0000	0.000024
liveness	586672.0	0.213935	0.184326	0.0	0.0983	0.139000
valence	586672.0	0.552292	0.257671	0.0	0.3460	0.564000
tempo	586672.0	118.464857	29.764108	0.0	95.6000	117.384000
time_signature	586672.0	3.873382	0.473162	0.0	4.0000	4.000000
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In [7]: most\_popular=df\_tracks.query('popularity>90',inplace=**False**).sort\_values('popularity',ascendir most\_popular[:10]

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	id	name	popularity	duration_ms	explicit	artists	
93802	4iJyoBOLtHqaGxP12qzhQI	Peaches (feat. Daniel Caesar & Giveon)	100	198082	1	['Justin Bieber', 'Daniel Caesar', 'Giveon']	[
93803	7IPN2DXiMsVn7XUKtOW1CS	drivers license	99	242014	1	['Olivia Rodrigo']	['
93804	3Ofmpyhv5UAQ70mENzB277	Astronaut In The Ocean	98	132780	0	['Masked Wolf']	['1
92810	5QO79kh1waicV47BqGRL3g	Save Your Tears	97	215627	1	['The Weeknd']	['1
92811	6tDDoYlxWvMLTdKpjFkc1B	telepatía	97	160191	0	[ˈKali Uchisˈ]	[
92813	0VjljW4GlUZAMYd2vXMi3b	Blinding Lights	96	200040	0	['The Weeknd']	['1
93805	7MAibcTli4lisCtbHKrGMh	Leave The Door Open	96	242096	0	['Bruno Mars', 'Anderson .Paak', 'Silk Sonic']	['
92814	6f3Slt0GbA2bPZlz0alFXN	The Business	95	164000	0	['Tiësto']	[
91866	60ynsPSSKe6O3sfwRnIBRf	Streets	94	226987	1	[ˈDoja Catˈ]	
92816	3FAJ6O0NOHQV8Mc5Ri6ENp	AJ6O0NOHQV8Mc5Ri6ENp Heartbreak 94 198371 0 ['Give		['Giveon']	[		
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In [8]: df\_tracks.set\_index("release\_date",inplace=**True**) df\_tracks.index=pd.to\_datetime(df\_tracks.index,format='mixed',dayfirst=**True**) df\_tracks.head()

# Out[8]:

	i	id	name	popularity	duration_ms	explicit	art
release_date							
1922-02-22	35iwgR4jXetI318WEWsa1	Q	Carve	6	126903	0	['1
1922-06-01	021ht4sdgPcrDgSk7JTbK	<Υ	Capítulo 2.16 - Banquero Anarquista	0	98200	0	['Fernaı Pess
1922-03-21	07A5yehtSnoedViJAZkNr	nc	Vivo para Quererte - Remasterizado	0	181640	0	[ˈlgna Cors
1922-03-21	08FmqUhxtyLTn6pAh6bk4	45	El Prisionero - Remasterizado	0	176907	0	[ˈlgna Cors
1922-01-01	08y9GfoqCWfOGsKdwojr5	5e	Lady of the Evening	0	163080	0	[ˈ[ Haym
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In [9]: df\_tracks['duration']=df\_tracks['duration\_ms'].apply(lambda x:round(x/1000)) df\_tracks.drop('duration\_ms',inplace=True,axis=1)

In [10]: corr\_df=df\_tracks.drop(['key','mode','explicit'],axis=1).corr(method='pearson',numeric\_only=**Ti** plt.figure(figsize=(14,6)) heatmap=sns.heatmap(corr\_df,annot=**True**,fmt='.1g',vmin=-1,vmax=1,center=0,cmap='inferr heatmap.set\_title('Correlation Heatmap Between Variable') heatmap.set\_xticklabels(heatmap.get\_xticklabels(),rotation=90);

Correlation Heatmap Between Variable													
popularity -	1	0.2	0.3	0.3	-0.05	-0.4	-0.2	-0.05	0.005	0.07	0.09	0.03	1.00
danceability -	0.2	1	0.2	0.3	0.2	-0.2	-0.2	-0.1	0.5	-0.04	0.1	-0.1	- 0.75
energy -	0.3	0.2	1	0.8	-0.05	-0.7	-0.2	0.1	0.4	0.2	0.2	0.02	0.50
loudness -	0.3	0.3	0.8	1	-0.2	-0.5	-0.3	0.03	0.3	0.2	0.2	0.0003	- 0.50
speechiness -	-0.05	0.2	-0.05	-0.2	1	0.07	-0.1	0.2	0.05	-0.09	-0.1	-0.1	- 0.25
acousticness -	-0.4	-0.2	-0.7	-0.5	0.07	1	0.2	-0.005	-0.2	-0.2	-0.2	-0.06	
instrumentalness -	-0.2	-0.2	-0.2	-0.3	-0.1	0.2	1	-0.04	-0.2	-0.06	-0.04	0.07	- 0.00
liveness -	-0.05	-0.1	0.1	0.03	0.2	-0.005	-0.04	1	-3e-05	-0.01	-0.02	0.002	0.25
valence -	0.005	0.5	0.4	0.3	0.05	-0.2	-0.2	-3e-05	1	0.1	0.1	-0.2	0.50
tempo -	0.07	-0.04	0.2	0.2	-0.09	-0.2	-0.06	-0.01	0.1	1	0.03	-0.001	0.50
time_signature -	0.09	0.1	0.2	0.2	-0.1	-0.2	-0.04	-0.02	0.1	0.03	1	0.04	0.75
duration -	0.03	-0.1	0.02	0.0003	-0.1	-0.06	0.07	0.002	-0.2	-0.001	0.04	1	
	popularity -	danceability -	energy -	loudness -	speechiness -	acousticness -	instrumentalness -	liveness -	valence -	tempo -	time_signature -	duration -	

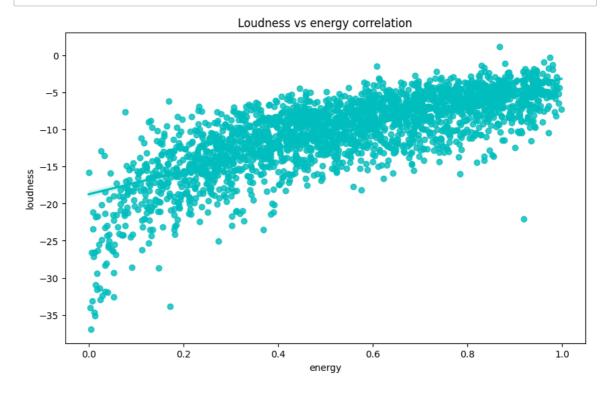
#### **#Sampling data**

In [11]: sample\_df = df\_tracks.sample(int(0.004\*len(df\_tracks)))

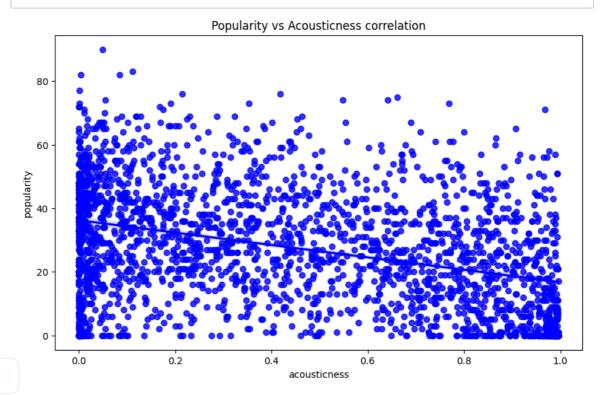
In [12]: print(len(sample\_df))

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In [13]: plt.figure(figsize=(10,6)) sns.regplot(data=sample\_df,y='loudness',x='energy',color='c').set(title="Loudness vs energy c

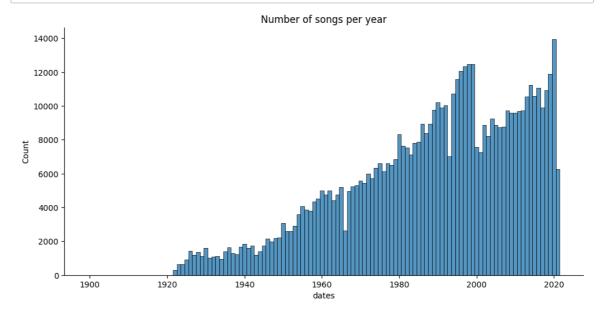


In [14]: plt.figure(figsize=(10,6)) sns.regplot(data=sample\_df,y='popularity',x='acousticness',color='b').set(title="Popularity vs A

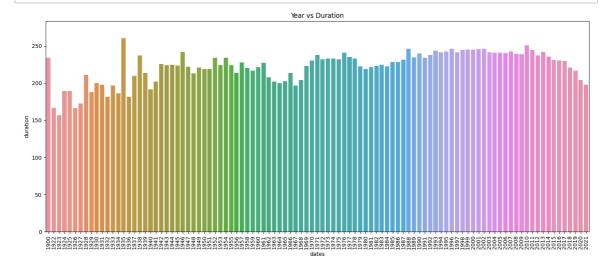


In [15]: df\_tracks['dates']=df\_tracks.index.get\_level\_values('release\_date') df\_tracks.dates=pd.to\_datetime(df\_tracks.dates) years=df\_tracks.dates.dt.year

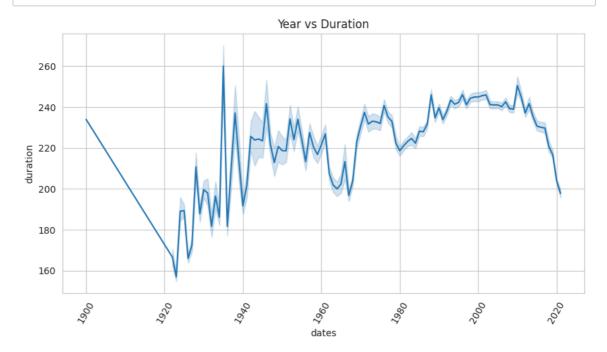
In [16]: sns.displot(years,discrete=**True**,aspect=2,height=5,kind='hist').set(title='Number of songs per



In [17]: total\_dr=df\_tracks.duration fig\_dims=(18,7) fig,ax=plt.subplots(figsize=fig\_dims) fig=sns.barplot(x=years,y=total\_dr,ax=ax,errwidth=**False**).set(title='Year vs Duration') plt.xticks(rotation=90);



In [18]: sns.set\_style(style='whitegrid') fig\_dims=(10,5) fig,ax=plt.subplots(figsize=fig\_dims) fig=sns.lineplot(x=years,y=total\_dr,ax=ax).set(title="Year vs Duration") plt.xticks(rotation=60);



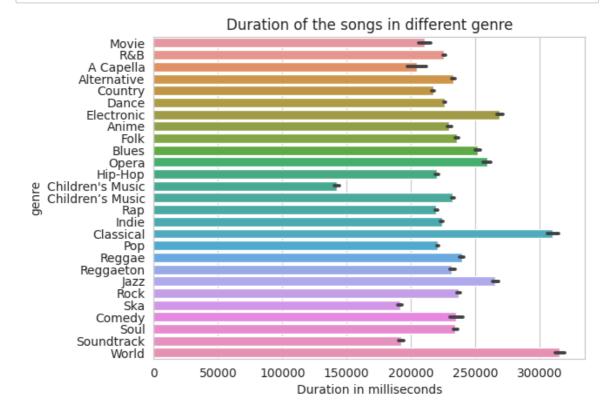
### **#Analysing genres**

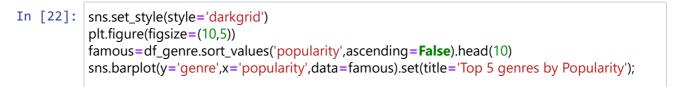
In [19]: df\_genre=pd.read\_csv("/kaggle/input/ultimate-spotify-tracks-db/SpotifyFeatures.csv")

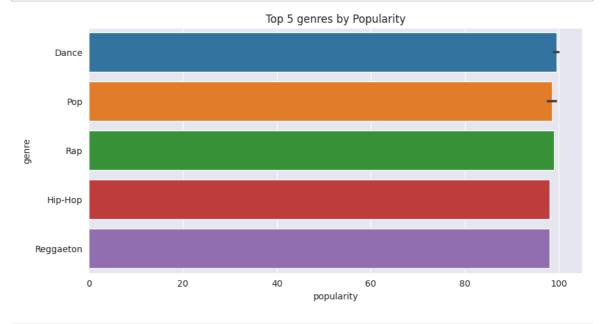
In [20]: df\_genre.head()

	genre	artist_name	track_name	track_id	popularity	acousticness	dancea
0	Movie	Henri Salvador	C'est beau de faire un Show	0BRjO6ga9RKCKjfDqeFgWV	0	0.611	
1	Movie	Martin & les fées	Perdu d'avance (par Gad Elmaleh)	0BjC1NfoEOOusryehmNudP	1	0.246	
2	Movie	Joseph Williams	Don't Let Me Be Lonely Tonight	0CoSDzoNIKCRs124s9uTVy	3	0.952	
3	Movie	Henri Salvador	Dis-moi Monsieur Gordon Cooper	0Gc6TVm52BwZD07Ki6tlvf	0	0.703	
4	Movie	Fabien Nataf	Ouverture	0luslXpMROHdEPvSl1fTQK	4	0.950	

In [21]: plt.title("Duration of the songs in different genre")
sns.color\_palette("rocket",as\_cmap=**True**)
sns.barplot(y='genre',x='duration\_ms',data=df\_genre)
plt.xlabel("Duration in milliseconds")
plt.ylabel('genre');







In [ ]: