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```
In [1]: 1 import pandas as pd
        2 import numpy as np
        3 import matplotlib.pyplot as plt
        4 %matplotlib inline
```

```
In [2]: 1 songs = pd.read_csv(r"D:\Data Science\Data Science Career Readiness Program\Datasets\Datasets\Machine_Learning\rolling_stone
        2 songs.head()
```

```
Out[2]:
```

	Unnamed: 0	name	album	release_date	track_number	id	uri	acousticness	danceability	energy
0	0	Concert Intro Music - Live	Licked Live In NYC	2022-06-10	1	2IEkywLJ4ykbhi1yRQvmsT	spotify:track:2IEkywLJ4ykbhi1yRQvmsT	0.0824	0.463	0.98
1	1	Street Fighting Man - Live	Licked Live In NYC	2022-06-10	2	6GVgVJBKkGJoRfarYRvGTU	spotify:track:6GVgVJBKkGJoRfarYRvGTU	0.4370	0.326	0.98
2	2	Start Me Up - Live	Licked Live In NYC	2022-06-10	3	1Lu761pZ0dBTGpzaQoZNW	spotify:track:1Lu761pZ0dBTGpzaQoZNW	0.4160	0.386	0.98
3	3	If You Can't Rock Me - Live	Licked Live In NYC	2022-06-10	4	1agTQzOTUnGNggycEqDH	spotify:track:1agTQzOTUnGNggycEqDH	0.5670	0.369	0.98
4	4	Don't Stop - Live	Licked Live In NYC	2022-06-10	5	7piGJR8YndQBQWVxv6KtQw	spotify:track:7piGJR8YndQBQWVxv6KtQw	0.4000	0.303	0.98

<

```
In [3]: 1 songs.shape
```

```
Out[3]: (1610, 18)
```

```
In [4]: 1 songs.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1610 entries, 0 to 1609
Data columns (total 18 columns):
#   Column              Non-Null Count  Dtype
---  -
0   Unnamed: 0          1610 non-null   int64
1   name                 1610 non-null   object
2   album                1610 non-null   object
3   release_date         1610 non-null   object
4   track_number         1610 non-null   int64
5   id                   1610 non-null   object
6   uri                  1610 non-null   object
7   acousticness         1610 non-null   float64
8   danceability         1610 non-null   float64
9   energy               1610 non-null   float64
10  instrumentalness     1610 non-null   float64
11  liveness             1610 non-null   float64
12  loudness             1610 non-null   float64
13  speechiness         1610 non-null   float64
14  tempo                1610 non-null   float64
15  valence              1610 non-null   float64
16  popularity           1610 non-null   int64
17  duration_ms          1610 non-null   int64
dtypes: float64(9), int64(4), object(5)
memory usage: 226.5+ KB
```

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Trusted

Python 3 (ipykernel)

Run

```
14 tempo          1610 non-null float64
15 valence         1610 non-null float64
16 popularity      1610 non-null int64
17 duration_ms     1610 non-null int64
dtypes: float64(9), int64(4), object(5)
memory usage: 226.5+ KB
```

In [5]: 1 songs.describe()

Out[5]:

	Unnamed: 0	track_number	acousticness	danceability	energy	instrumentalness	liveness	loudness	speechiness	tempo	valence
count	1610.000000	1610.000000	1610.000000	1610.000000	1610.000000	1610.000000	1610.000000	1610.000000	1610.000000	1610.000000	1610.000000
mean	804.500000	8.813885	0.250475	0.468880	0.792352	0.164170	0.49173	-8.971615	0.089512	126.082033	0.582165
std	464.911282	6.560220	0.227397	0.141775	0.179886	0.276249	0.34910	2.994003	0.051631	29.233483	0.231253
min	0.000000	1.000000	0.000009	0.104000	0.141000	0.000000	0.02190	-24.408000	0.023200	46.525000	0.000000
25%	402.250000	4.000000	0.058350	0.362250	0.674000	0.000219	0.15300	-8.982500	0.036500	107.390750	0.404250
50%	804.500000	7.000000	0.183000	0.458000	0.848500	0.013750	0.37950	-8.523000	0.051200	124.404500	0.583000
75%	1206.750000	11.000000	0.403750	0.578000	0.945000	0.179000	0.89375	-4.608750	0.086600	142.356750	0.778000
max	1609.000000	47.000000	0.994000	0.887000	0.999000	0.996000	0.99800	-1.014000	0.624000	216.304000	0.974000

In [6]: 1 songs.columns

Out[6]: Index(['Unnamed: 0', 'name', 'album', 'release\_date', 'track\_number', 'id', 'uri', 'acousticness', 'danceability', 'energy', 'instrumentalness', 'liveness', 'loudness', 'speechiness', 'tempo', 'valence', 'popularity', 'duration\_ms'], dtype='object')

```
In [12]: 1 sizes = [len(part_1['popularity']), len(part_2['popularity']), len(part_3['popularity']), len(part_4['popularity'])]

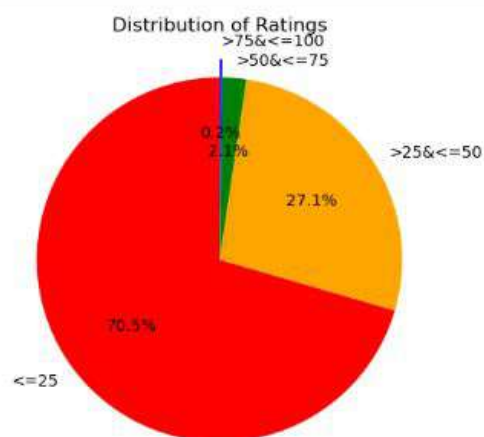
In [13]: 1 colors = ['red', 'orange', 'green', 'blue']

In [14]: 1 labels = ['<=25', '>25&=<=50', '>50&=<=75', '>75&=<=100']

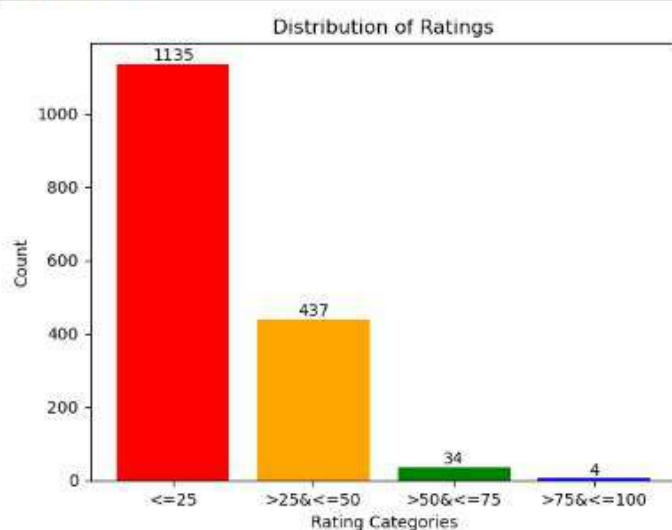
In [15]: 1 x_pos = np.arange(len(labels))

In [16]: 1 explode = (0, 0, 0, 0.1)

In [17]: 1 plt.pie(sizes, explode=explode, labels=labels, colors=colors, autopct='%1.1f%%', startangle=90)
2 plt.axis('equal')
3 plt.title('Distribution of Ratings')
4 plt.show()
```



```
In [18]: 1 plt.bar(x_pos, sizes, color=colors)
2 plt.xlabel('Rating Categories')
3 plt.ylabel('Count')
4 plt.title('Distribution of Ratings')
5
6 plt.xticks(x_pos, labels)
7
8 for i, size in enumerate(sizes):
9     plt.text(i, size, str(size), ha='center', va='bottom')
10
11 plt.show()
```



Like this, we can divide the songs based on the various attributes that have been given.

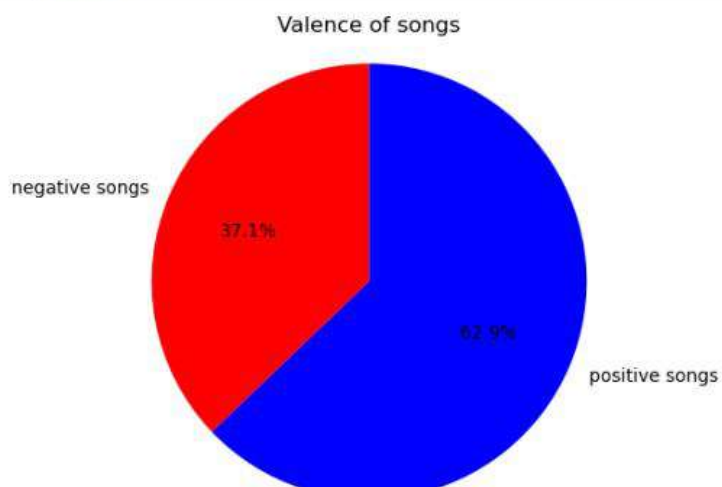
and we can create number of clusters of songs depending on the features.

In [ ]: 1

Out[22]: (1012, 18)

```
In [23]: 1 sizes = [len(valence1['valence']), len(valence2['valence'])]
2 colors = ['red', 'blue']
3 labels = ['negative songs', 'positive songs']
4 explode = (0, 0)
5 x_pos = np.arange(len(labels))
```

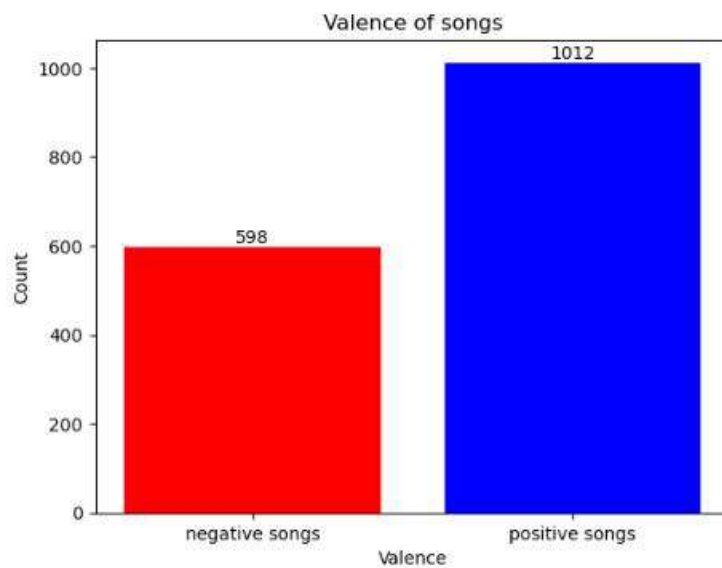
```
In [24]: 1 plt.pie(sizes, explode=explode, labels=labels, colors=colors, autopct='%1.1f%%', startangle=90)
2 plt.axis('equal')
3 plt.title('Valence of songs')
4 plt.show()
```





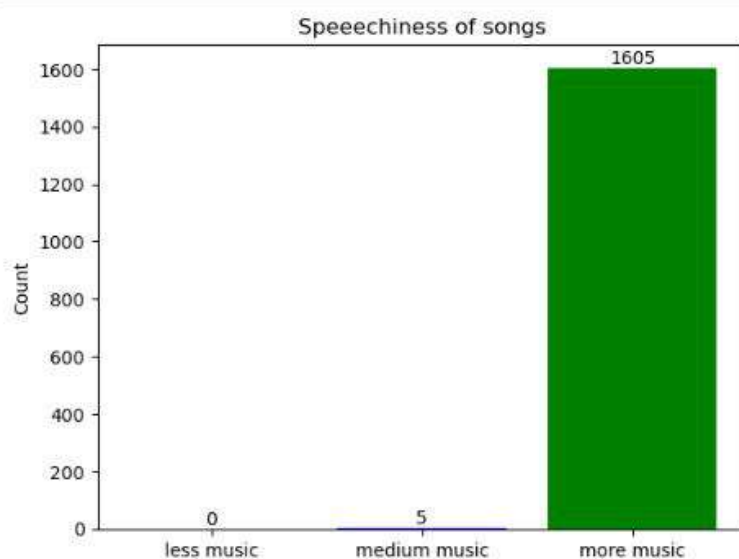


```
In [25]: 1 plt.bar(x_pos, sizes, color=colors)
2         plt.xlabel('Valence')
3         plt.ylabel('Count')
4         plt.title('Valence of songs')
5
6         plt.xticks(x_pos, labels)
7
8         for i, size in enumerate(sizes):
9             plt.text(i, size, str(size), ha='center', va='bottom')
10
11         plt.show()
```



```
3 labels = ['less music', 'medium music', 'more music']
4 explode = (0.1, 0.1, 0)
5 x_pos = np.arange(len(labels))
```

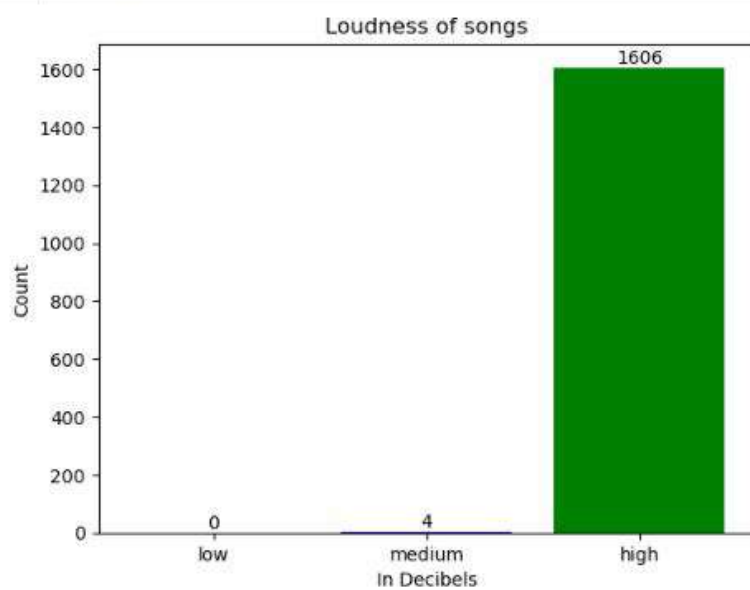
```
In [33]: 1 plt.bar(x_pos, sizes, color=colors)
2 plt.xlabel('Speechiness')
3 plt.ylabel('Count')
4 plt.title('Speechiness of songs')
5
6 plt.xticks(x_pos, labels)
7
8 for i, size in enumerate(sizes):
9     plt.text(i, size, str(size), ha='center', va='bottom')
10
11 plt.show()
```

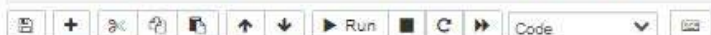






```
In [40]: 1 plt.bar(x_pos, sizes, color=colors)
2         plt.xlabel('In Decibels')
3         plt.ylabel('Count')
4         plt.title('Loudness of songs')
5
6         plt.xticks(x_pos, labels)
7
8         for i, size in enumerate(sizes):
9             plt.text(i, size, str(size), ha='center', va='bottom')
10
11         plt.show()
```



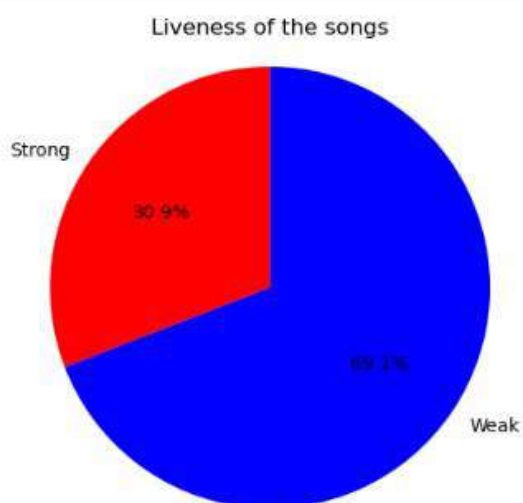


In [44]: weak.shape

Out[44]: (1113, 18)

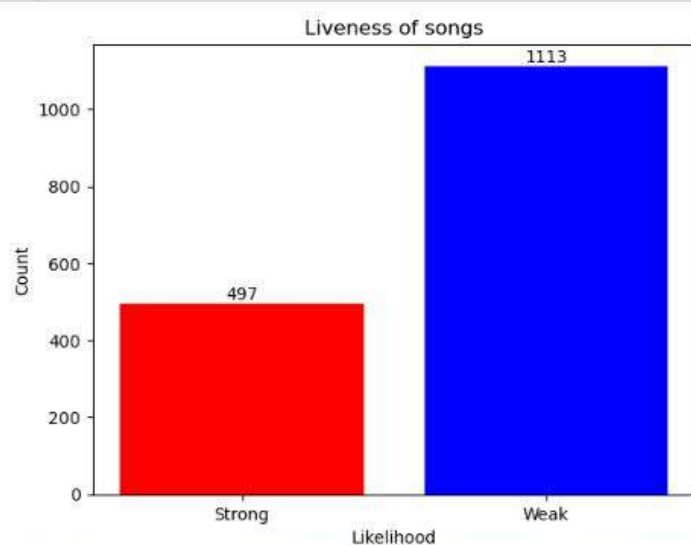
```
In [45]: 1 sizes = [len(strong['liveness']), len(weak['liveness'])]
2 colors = ['red', 'blue']
3 labels = ['Strong', 'Weak']
4 explode = (0, 0)
5 x_pos = np.arange(len(labels))
```

```
In [46]: 1 plt.pie(sizes, explode=explode, labels=labels, colors=colors, autopct='%1.1f%%', startangle=90)
2 plt.axis('equal')
3 plt.title('Liveness of the songs')
4 plt.show()
```





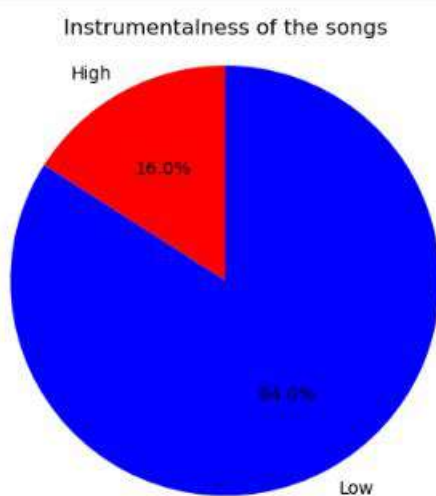
```
In [47]: 1 plt.bar(x_pos, sizes, color=colors)
2 plt.xlabel('Likelihood')
3 plt.ylabel('Count')
4 plt.title('Liveness of songs')
5
6 plt.xticks(x_pos, labels)
7
8 for i, size in enumerate(sizes):
9     plt.text(i, size, str(size), ha='center', va='bottom')
10
11 plt.show()
```





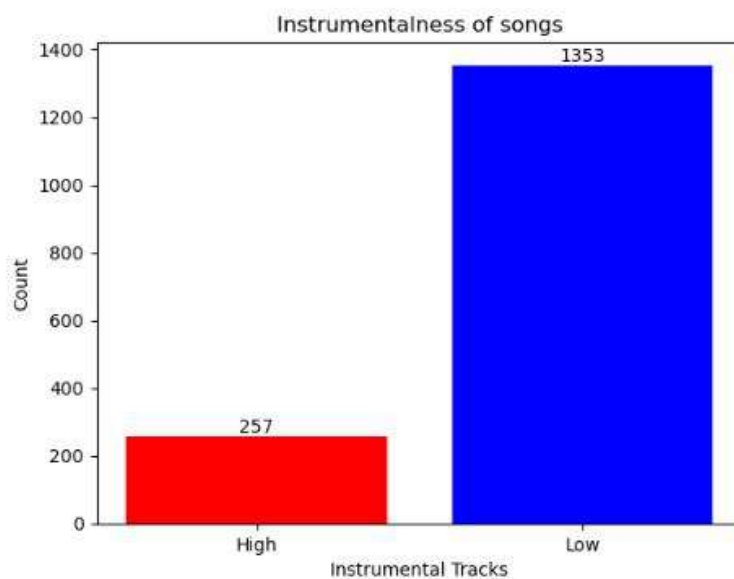
```
In [52]: 1 sizes = [len(high['instrumentalness']), len(low['instrumentalness'])]
2 colors = ['red', 'blue']
3 labels = ['High', 'Low']
4 explode = (0, 0)
5 x_pos = np.arange(len(labels))
```

```
In [53]: 1 plt.pie(sizes, explode=explode, labels=labels, colors=colors, autopct='%1.1f%%', startangle=90)
2 plt.axis('equal')
3 plt.title('Instrumentalness of the songs')
4 plt.show()
```



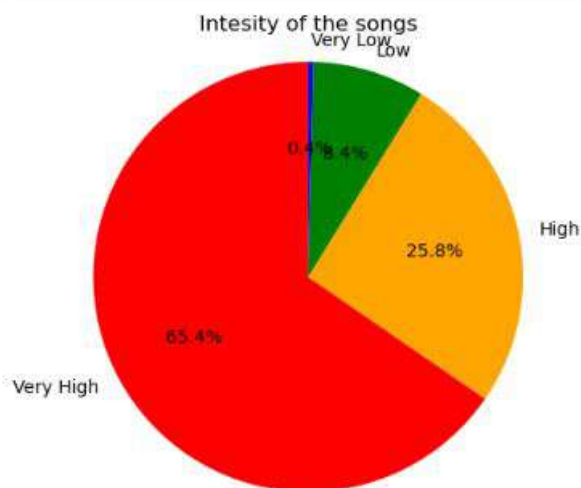


```
In [54]: 1 plt.bar(x_pos, sizes, color=colors)
2 plt.xlabel('Instrumental Tracks')
3 plt.ylabel('Count')
4 plt.title('Instrumentalness of songs')
5
6 plt.xticks(x_pos, labels)
7
8 for i, size in enumerate(sizes):
9     plt.text(i, size, str(size), ha='center', va='bottom')
10
11 plt.show()
```



```
In [59]: 1 sizes = [len(very_high['energy']), len(high['energy']), len(low['energy']), len(very_low['energy'])]
2 colors = ['red', 'orange', 'green', 'blue']
3 labels = ['Very High', 'High', 'Low', 'Very Low']
4 explode = (0, 0, 0, 0)
5 x_pos = np.arange(len(labels))
```

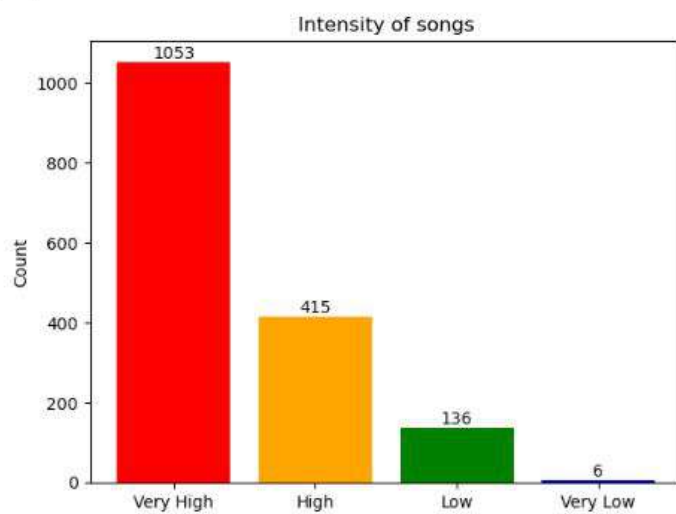
```
In [60]: 1 plt.pie(sizes, explode=explode, labels=labels, colors=colors, autopct='%1.1f%%', startangle=90)
2 plt.axis('equal')
3 plt.title('Intesity of the songs')
4 plt.show()
```



```
In [61]: 1 plt.bar(x_pos, sizes, color=colors)
2 plt.xlabel('Energy')
3 plt.ylabel('Count')
4 plt.title('Intensity of songs')
5
```



```
In [61]: 1 plt.bar(x_pos, sizes, color=colors)
2         plt.xlabel('Energy')
3         plt.ylabel('Count')
4         plt.title('Intensity of songs')
5
6         plt.xticks(x_pos, labels)
7
8         for i, size in enumerate(sizes):
9             plt.text(i, size, str(size), ha='center', va='bottom')
10
11         plt.show()
```

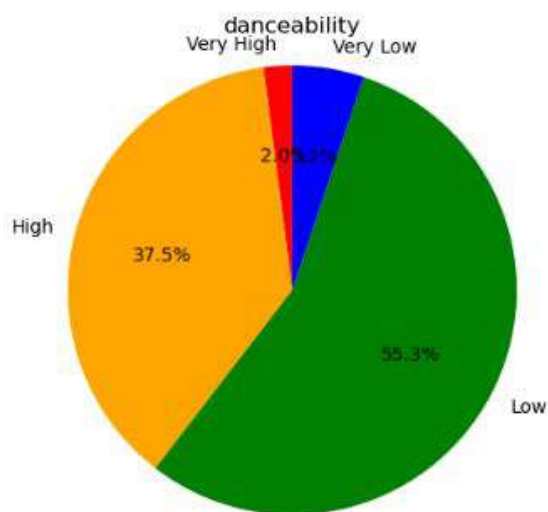






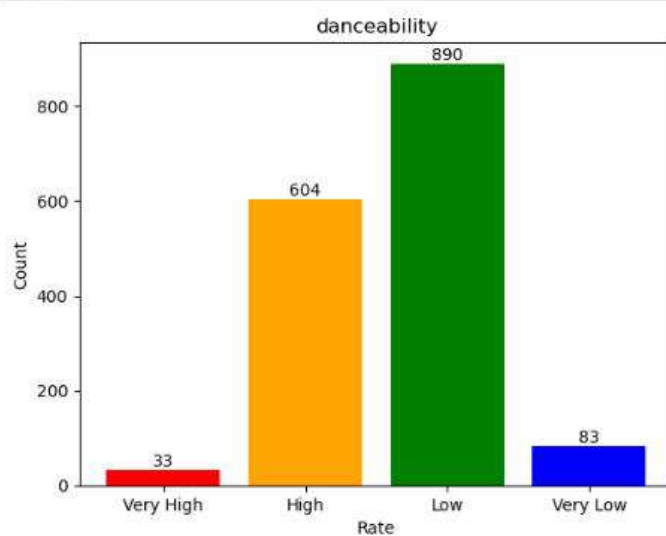
```
2 colors = ['red', 'orange', 'green', 'blue']
3 labels = ['Very High', 'High', 'Low', 'Very Low']
4 explode = (0, 0, 0, 0)
5 x_pos = np.arange(len(labels))
```

```
In [67]: 1 plt.pie(sizes, explode=explode, labels=labels, colors=colors, autopct='%1.1f%%', startangle=90)
2 plt.axis('equal')
3 plt.title('danceability')
4 plt.show()
```



```
In [68]: 1 plt.bar(x_pos, sizes, color=colors)
2 plt.xlabel('Rate')
3 plt.ylabel('Count')
4 plt.title('danceability')
```

```
In [68]: 1 plt.bar(x_pos, sizes, color=colors)
2         plt.xlabel('Rate')
3         plt.ylabel('Count')
4         plt.title('danceability')
5
6         plt.xticks(x_pos, labels)
7
8         for i, size in enumerate(sizes):
9             plt.text(i, size, str(size), ha='center', va='bottom')
10
11         plt.show()
```

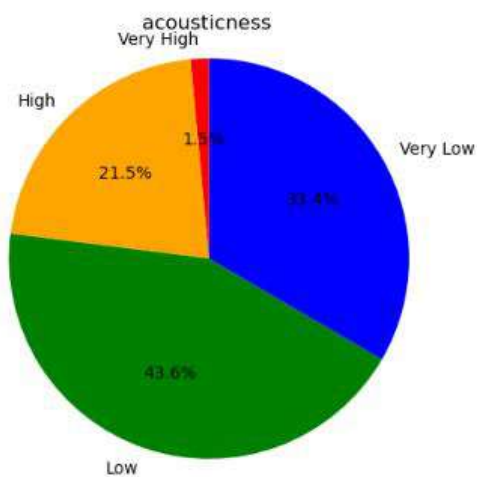


Clustering songs based on acoustictness



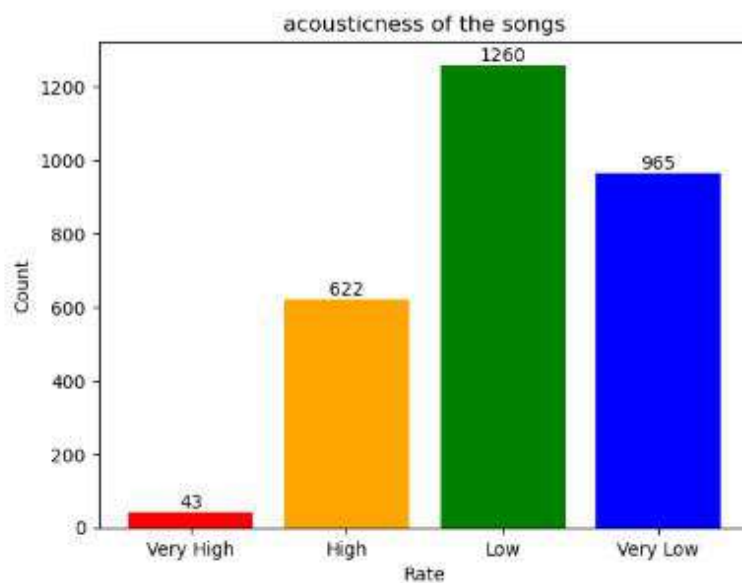
```
2 colors = ['red', 'orange', 'green', 'blue']
3 labels = ['Very High', 'High', 'Low', 'Very Low']
4 explode = (0, 0, 0, 0)
5 x_pos = np.arange(len(labels))
```

```
In [74]: 1 plt.pie(sizes, explode=explode, labels=labels, colors=colors, autopct='%1.1f%%', startangle=90)
2 plt.axis('equal')
3 plt.title('acousticness')
4 plt.show()
```



```
In [75]: 1 plt.bar(x_pos, sizes, color=colors)
2 plt.xlabel('Rate')
3 plt.ylabel('Count')
4 plt.title('acousticness of the songs')
```

```
In [75]: 1 plt.bar(x_pos, sizes, color=colors)
2 plt.xlabel('Rate')
3 plt.ylabel('Count')
4 plt.title('acousticness of the songs')
5
6 plt.xticks(x_pos, labels)
7
8 for i, size in enumerate(sizes):
9     plt.text(i, size, str(size), ha='center', va='bottom')
10
11 plt.show()
```



Like this, we can divide the songs based on the various attributes that have been given.  
and we can create number of clusters of songs depending on the features.



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Python 3 (ipykernel)

Run Code

## CASE STUDY

Let's create a list of songs where the conditions for attributes of the songs are given below

Popularity - Medium

Valence - Positive

Speechiness - Less Speech and more Music

Loudness - High

Liveness - Strong

Instrumentalness - Moderate

Energy - High

Danceability - Medium

Acousticness - Medium

```
In [76]: 1 songs[(songs['popularity'] <= 50) &
2         (songs['valence'] >= 0.4) &
3         (songs['speechiness'] <= 0.33) &
4         (songs['loudness'] >= -38) &
5         (songs['liveness'] >= 0.6) &
6         (songs['instrumentalness'] >= 0.3) & (songs['instrumentalness'] <= 0.7) &
7         (songs['energy'] >= 0.4) &
8         (songs['danceability'] >= 0.3) & (songs['danceability'] <= 0.7) &
9         (songs['acousticness'] >= 0.3) & (songs['acousticness'] <= 0.7)]
```

Out[76]:

Unnamed: 0	name	album	release_date	track_number	id	un	acousticness	den
18	It's Only Rock 'n' Roll - Live	Licked Live In NYC	2022-05-10	10	1Wb0v5UDv3wS34WWJ0AN	spotify:track:1Wb0v5UDv3wS34WWJ0AN	0.675	
89	Start Me Up - Live at Wembley Stadium 1982	Start Me Up (Super Deluxe)	2021-10-22	24	4F6Mct0K0a0h1u0UHLv6A	spotify:track:4F6Mct0K0a0h1u0UHLv6A	0.489	
281	Not Fade Away - Live	Voodoo Lounge Uncut (Live)	2018-11-18	2	1b0D9H50BFD5eH0vabXfC	spotify:track:1b0D9H50BFD5eH0vabXfC	0.409	
389	Not Fade Away - Live	Voodoo Lounge Uncut (Live)	2018-11-18	2	8YCDfK6LCC8pV0K0p0K0wG	spotify:track:8YCDfK6LCC8pV0K0p0K0wG	0.409	
839	Intro (Take The A Train) - Live (Remastered 2009)	Star Line	1950-08-01	1	2cSgynX2H0L1mb1eB0C	spotify:track:2cSgynX2H0L1mb1eB0C	0.483	
980	Around And Around - Live	Love You Live (Remastered)	1977-09-23	4	hwbdumQ03SfWnSP1uKEy	spotify:track:hwbdumQ03SfWnSP1uKEy	0.808	
1523	Oh Baby (We Got A Good Thing Goin') - Remaster...	Now!	1985-02-13	10	2d0nL0m0K0f1u0XfNH1LQ	spotify:track:2d0nL0m0K0f1u0XfNH1LQ	0.494	