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```
import pandas as pd
In [1]:
         data=pd.read_csv("C:\\Users\\Dell\\OneDrive\\Desktop\\excel books\\musicdata.csv")
In [2]:
         print(data.head())
In [3]:
            Unnamed: 0
                                                          Track Name
        0
                     0
                                                       Bijlee Bijlee
        1
                                                         Expert Jatt
        2
                     2
                        Kaun Nachdi (From "Sonu Ke Titu Ki Sweety")
                     3
        3
                                                         Na Na Na Na
        4
                     4
                                                         Patiala Peg
                                                                Album Name \
                               Artists
        0
                         Harrdy Sandhu
                                                             Bijlee Bijlee
                                 Nawab
                                                               Expert Jatt
        1
        2
           Guru Randhawa, Neeti Mohan
                                         High Rated Gabru - Guru Randhawa
        3
                                J Star
                                                               Na Na Na Na
        4
                        Diljit Dosanjh Do Gabru - Diljit Dosanjh & Akhil
                          Album ID
                                                   Track ID Popularity Release Date \
           3tG0IGB24sRhGFLs5F1Km8 1iZLpuGMr4tn1F5bZu32Kb
                                                                          2021-10-30
                                                                     70
        1
           2gibg5SCTep0wsIMefGzkd
                                    7rr6n1NFIcQXCsi43P0YNl
                                                                     65
                                                                           2018-01-18
                                    3s7m0jmCXGcM8tmlvjCvAa
                                                                           2019-03-02
           6EDbwGsQNQRLf73c7QwZ2f
                                                                     64
           4xBqgoiRSOMU1VlKuntVQW
                                    5GjxbFTZAMhrVfVrNrrwrG
                                                                     52
        3
                                                                                 2015
           1uxDllRe9CPhdr8rhz2QCZ 6TikcWOLRsPq66GBx2jk67
                                                                     46
                                                                           2018-07-10
                                     ... Energy Key Loudness Mode Speechiness
           Duration (ms) Explicit
        0
                   168450
                              False
                                          0.670
                                                    1
                                                         -5.313
                                                                    0
                                                                            0.1430
        1
                   199535
                              False
                                          0.948
                                                    6
                                                         -2.816
                                                                    0
                                                                            0.1990
        2
                   183373
                              False
                                          0.830
                                                    4
                                                         -3.981
                                                                    0
                                                                            0.0455
        3
                                                    3
                                                                    1
                                                                            0.0413
                   209730
                              False ...
                                          0.863
                                                         -3.760
        4
                   188314
                              False
                                          0.811
                                                    5
                                                         -3.253
                                                                    0
                                                                            0.1840
                                    . . .
           Acousticness Instrumentalness Liveness
                                                      Valence
                                                                  Tempo
        0
                  0.2690
                                  0.000000
                                              0.0733
                                                         0.643
                                                                100.004
        1
                  0.2980
                                  0.000000
                                              0.0784
                                                         0.647
                                                                172.038
        2
                                              0.0419
                                                         0.753
                                                                127.999
                  0.0357
                                  0.000000
                                              0.0916
                                                                 95.000
        3
                  0.3760
                                  0.000014
                                                         0.807
        4
                  0.0259
                                  0.000000
                                                                175.910
                                              0.3110
                                                         0.835
        [5 rows x 22 columns]
```

```
<class 'pandas.core.frame.DataFrame'>
        RangeIndex: 100 entries, 0 to 99
        Data columns (total 22 columns):
         #
             Column
                                Non-Null Count
                                                Dtype
         ---
             -----
                                _____
         0
              Unnamed: 0
                                100 non-null
                                                int64
             Track Name
                                94 non-null
                                                object
         1
         2
             Artists
                                94 non-null
                                                object
         3
             Album Name
                                94 non-null
                                                object
         4
             Album ID
                                100 non-null
                                                object
         5
             Track ID
                                100 non-null
                                                object
         6
             Popularity
                                100 non-null
                                                int64
         7
             Release Date
                                100 non-null
                                                object
         8
             Duration (ms)
                                100 non-null
                                                int64
         9
              Explicit
                                100 non-null
                                                bool
         10 External URLs
                                100 non-null
                                                object
             Danceability
                                                float64
         11
                                100 non-null
         12
                                100 non-null
                                                float64
             Energy
         13 Key
                                100 non-null
                                                int64
         14 Loudness
                                100 non-null
                                                float64
         15 Mode
                                100 non-null
                                                int64
         16 Speechiness
                                                float64
                                100 non-null
         17 Acousticness
                                100 non-null
                                                float64
         18 Instrumentalness
                                100 non-null
                                                float64
         19 Liveness
                                                float64
                                100 non-null
         20 Valence
                                                float64
                                100 non-null
         21 Tempo
                                100 non-null
                                                float64
         dtypes: bool(1), float64(9), int64(5), object(7)
        memory usage: 16.6+ KB
         data.index
In [5]:
        RangeIndex(start=0, stop=100, step=1)
Out[5]:
In [6]:
         data.isnull().sum()
        Unnamed: 0
                             0
Out[6]:
        Track Name
                             6
        Artists
                             6
                             6
        Album Name
        Album ID
                             0
        Track ID
                             0
                             0
        Popularity
        Release Date
                             0
        Duration (ms)
                             0
                             0
        Explicit
        External URLs
        Danceability
                             0
                             0
         Energy
                             0
        Key
                             0
        Loudness
        Mode
                             0
        Speechiness
                             0
        Acousticness
                             0
         Instrumentalness
                             0
        Liveness
                             0
        Valence
                             0
        Tempo
                             0
        dtype: int64
         data.describe()
In [7]:
```

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Out[7]:		Unnamed: 0	Popularity	Duration (ms)	Danceability	Energy	Key	Loudness	
	count	100.000000	100.000000	100.000000	100.000000	100.00000	100.00000	100.000000	100.
	mean	49.500000	50.950000	210543.180000	0.767210	0.79763	4.54000	-4.399930	0.
	std	29.011492	16.496326	37961.050214	0.085302	0.11572	3.64434	1.612703	0.
	min	0.000000	0.000000	141862.000000	0.501000	0.47700	0.00000	-8.272000	0.
	25%	24.750000	46.000000	186098.500000	0.714750	0.71125	1.00000	-5.465250	0.
	50%	49.500000	56.500000	205076.000000	0.772000	0.81700	4.00000	-4.252500	0.
	75%	74.250000	62.000000	226079.000000	0.826500	0.88125	7.25000	-3.163250	1.
	max	99.000000	72.000000	367818.000000	0.959000	0.98800	11.00000	-0.223000	1.
4									•
In [8]:	clean	ed_data=dat	:a.drop(col	Lumns=['Unname	ed: 0'])				

cleaned_data

Out[8]:

	Track Name	Artists	Album Name	Album ID	Track ID	Pc
0	Bijlee Bijlee	Harrdy Sandhu	Bijlee Bijlee	3tG0IGB24sRhGFLs5F1Km8	1iZLpuGMr4tn1F5bZu32Kb	
1	Expert Jatt	Nawab	Expert Jatt	2gibg5SCTep0wsIMefGzkd	7rr6n1NFlcQXCsi43P0YNl	
2	Kaun Nachdi (From "Sonu Ke Titu Ki Sweety")	Guru Randhawa, Neeti Mohan	High Rated Gabru - Guru Randhawa	6EDbwGsQNQRLf73c7QwZ2f	3s7m0jmCXGcM8tmlvjCvAa	
3	Na Na Na Na	J Star	Na Na Na Na	4xBqgoiRSOMU1VIKuntVQW	5GjxbFTZAMhrVfVrNrrwrG	
4	Patiala Peg	Diljit Dosanjh	Do Gabru - Diljit Dosanjh & Akhil	1uxDllRe9CPhdr8rhz2QCZ	6TikcWOLRsPq66GBx2jk67	
•••						
95	NaN	NaN	NaN	2jw92hf4mnlSbYywvU3Anj	3OZr3vo7SmYpn5XqeQEAOM	
96	Move Your Lakk	Diljit Dosanjh, Badshah, Sonakshi Sinha	Move Your Lakk	0V06TMGQQQkvKxNmFlKyEj	3aYMKdSitJeHUCZO8Wt6fw	
97	Patola (From "Blackmail")	Guru Randhawa, Preet Hundal	Patola (From "Blackmail")	2XAAIDEpPb57NsKgAHLGVQ	17LZzRCY0iFWlDDuAG7BlM	
98	Ban Ja Rani (From "Tumhari Sulu")	Guru Randhawa	High Rated Gabru - Guru Randhawa	6EDbwGsQNQRLf73c7QwZ2f	7cQtGVoPCK9DlspeYjdHOA	
99	Hauli Hauli (From "De De Pyaar De")	Garry Sandhu, Neha Kakkar, Mellow D	Dance Syndrome	6e1XB070vlPaxGDAsi8AF6	4XyKoSEWrkQjl4AekJYWNx	

100 rows × 21 columns

```
In [9]: columns_with_na = ['Track Name','Artists','Album Name']
    cleaned_data[columns_with_na]=cleaned_data[columns_with_na].fillna('unkown')
In [10]: cleaned_data
```

Out[10]:

	Track Name	Artists	Album Name	Album ID	Track ID	Pc
0	Bijlee Bijlee	Harrdy Sandhu	Bijlee Bijlee	3tG0IGB24sRhGFLs5F1Km8	1iZLpuGMr4tn1F5bZu32Kb	
1	Expert Jatt	Nawab	Expert Jatt	2gibg5SCTep0wsIMefGzkd	7rr6n1NFlcQXCsi43P0YNl	
2	Kaun Nachdi (From "Sonu Ke Titu Ki Sweety")	Guru Randhawa, Neeti Mohan	High Rated Gabru - Guru Randhawa	6EDbwGsQNQRLf73c7QwZ2f	3s7m0jmCXGcM8tmlvjCvAa	
3	Na Na Na Na	J Star	Na Na Na Na	4xBqgoiRSOMU1VIKuntVQW	5GjxbFTZAMhrVfVrNrrwrG	
4	Patiala Peg	Diljit Dosanjh	Do Gabru - Diljit Dosanjh & Akhil	1uxDllRe9CPhdr8rhz2QCZ	6TikcWOLRsPq66GBx2jk67	
•••						
95	unkown	unkown	unkown	2jw92hf4mnlSbYywvU3Anj	3OZr3vo7SmYpn5XqeQEAOM	
96	Move Your Lakk	Diljit Dosanjh, Badshah, Sonakshi Sinha	Move Your Lakk	0V06TMGQQQkvKxNmFlKyEj	3aYMKdSitJeHUCZO8Wt6fw	
97	Patola (From "Blackmail")	Guru Randhawa, Preet Hundal	Patola (From "Blackmail")	2XAAIDEpPb57NsKgAHLGVQ	17LZzRCY0iFWlDDuAG7BlM	
98	Ban Ja Rani (From "Tumhari Sulu")	Guru Randhawa	High Rated Gabru - Guru Randhawa	6EDbwGsQNQRLf73c7QwZ2f	7cQtGVoPCK9DlspeYjdHOA	
99	Hauli Hauli (From "De De Pyaar De")	Garry Sandhu, Neha Kakkar, Mellow D	Dance Syndrome	6e1XB070vlPaxGDAsi8AF6	4XyKoSEWrkQjl4AekJYWNx	

100 rows × 21 columns

→

In [11]: cleaned_data.isnull().sum()

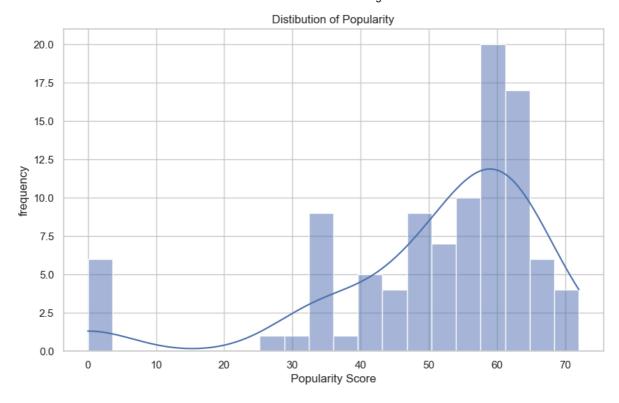
```
Track Name
                               0
Out[11]:
          Artists
                               0
          Album Name
                               0
          Album ID
                               0
          Track ID
                               0
          Popularity
                               0
          Release Date
                               0
                               0
          Duration (ms)
          Explicit
                               0
          External URLs
                               0
          Danceability
                               0
          Energy
                               0
                               0
          Key
                               0
          Loudness
          Mode
                               0
          Speechiness
                               0
          Acousticness
                               0
          Instrumentalness
                               0
          Liveness
                               0
          Valence
                               0
          Tempo
                               0
          dtype: int64
```

In [34]: cleaned_data.describe()

```
Duration (ms) Explicit Danceability
Out[34]:
                   Popularity
                                                                        Energy
                                                                                       Key
                                                                                              Loudness
                                                                                                            Mo
           count 100.000000
                                   100.000000
                                                100.00
                                                          100.000000
                                                                      100.00000 100.00000
                                                                                            100.000000 100.000
                    50.950000 210543.180000
                                                                                              -4.399930
           mean
                                                  0.01
                                                            0.767210
                                                                        0.79763
                                                                                   4.54000
                                                                                                           0.430
                    16.496326
                                37961.050214
                                                  0.10
                                                            0.085302
                                                                        0.11572
                                                                                   3.64434
                                                                                              1.612703
                                                                                                           0.497
              std
                     0.000000 141862.000000
                                                  0.00
                                                            0.501000
                                                                        0.47700
                                                                                   0.00000
                                                                                              -8.272000
                                                                                                           0.000
             min
             25%
                    46.000000
                               186098.500000
                                                  0.00
                                                            0.714750
                                                                        0.71125
                                                                                   1.00000
                                                                                              -5.465250
                                                                                                           0.000
             50%
                    56.500000
                               205076.000000
                                                  0.00
                                                            0.772000
                                                                        0.81700
                                                                                   4.00000
                                                                                              -4.252500
                                                                                                           0.000
             75%
                    62.000000
                               226079.000000
                                                  0.00
                                                            0.826500
                                                                        0.88125
                                                                                   7.25000
                                                                                              -3.163250
                                                                                                           1.000
                                                                        0.98800
                                                                                              -0.223000
             max
                    72.000000 367818.000000
                                                  1.00
                                                            0.959000
                                                                                  11.00000
                                                                                                           1.000
```

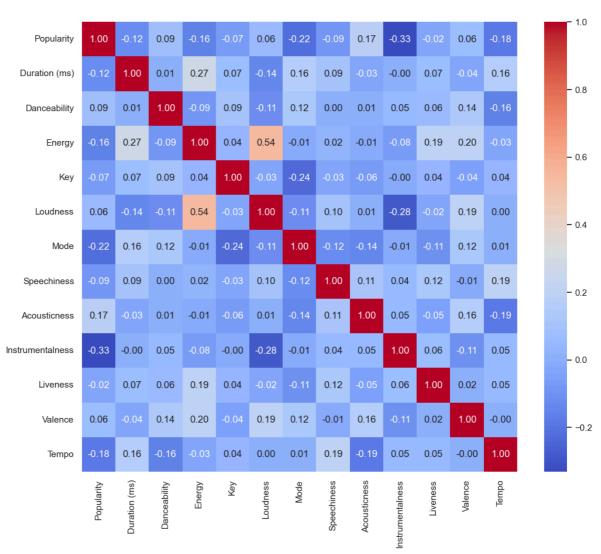
```
In [12]: import seaborn as sns
  import matplotlib.pyplot as plt

In [13]: sns.set(style='whitegrid')
  plt.figure(figsize=(10,6))
  sns.histplot(cleaned_data['Popularity'],bins=20,kde=True)
  plt.title('Distibution of Popularity')
  plt.xlabel('Popularity Score')
  plt.ylabel('frequency')
  plt.show()
```



In [14]: plt.figure(figsize=(12,10))
 corr_matrix=cleaned_data.select_dtypes(include=['float64', 'int64']).corr()
 sns.heatmap(corr_matrix,annot=True,cmap='coolwarm',fmt=".2f")

Out[14]: <AxesSubplot:>



```
features=['Danceability', 'Energy', 'Loudness', 'Acousticness', 'Valence', 'Speechir
In [15]:
             plt.figure(figsize=(15,10))
             for i , feature in enumerate(features,1):
                  plt.subplot(2,3,i)
                  sns.scatterplot(x=cleaned_data[feature],y=cleaned_data['Popularity'],hue=cleaned
                  plt.title(f'{feature} v/s Popularity')
                  plt.xlabel(feature)
                  plt.ylabel('Popularity')
             plt.tight_layout()
             plt.show()
                         Danceability v/s Popularity
                                                                Energy v/s Popularity
                                                                                                     Loudness v/s Popularity
              60
                                                   60
            Popularity
8 8
                                                    40
                                                  g 30
              20
                                                   20
                  Explicit
                                                        Explicit
                                                                                                                     Explicit
              0
                 0.5
                                          0.9
                                                        0.5
                                                             0.6
                       0.6
                                    0.8
                                                                   0.7
                                                                         0.8
                             Danceability
                                                                    Energy
                                                                                                         Loudness
                         Acousticness v/s Popularity
                                                                                                    Speechiness v/s Popularity
                                                                Valence v/s Popularity
              60
                                                   60
              50
            Popularity
8 8
                                                  arity
                                                    40
                                                  Popul
30
                                                                                       Popu
30
              20
                                                   20
                                                                                         20
              10
                                                        Explici
                                                                                                                     Explicit
                                            False
                                                          False
                                                                                                                       False
                 0.0
                     0.1
                          0.2
                               0.3
                                    0.4
                                        0.5
                                                           0.2
                                                                        0.6
                                                                               0.8
                                                                                                  0.10
                                                                                                           0.20
                                                                                                                0.25
                                                                                                                     0.30
In [16]:
             from sklearn.model_selection import train_test_split
             from sklearn.linear model import LinearRegression #y=mx+c
             from sklearn.metrics import mean squared error, r2 score
             from sklearn.preprocessing import StandardScaler #z=x-u/sigma
```

In [17]: cleaned_data.dtypes

```
Track Name
                               object
Out[17]:
         Artists
                               object
         Album Name
                               object
         Album ID
                               object
         Track ID
                               object
         Popularity
                                int64
         Release Date
                               object
                               int64
         Duration (ms)
         Explicit
                                 bool
         External URLs
                               object
         Danceability
                              float64
         Energy
                              float64
                                int64
         Key
         Loudness
                              float64
         Mode
                                int64
                              float64
         Speechiness
         Acousticness
                              float64
         Instrumentalness
                              float64
         Liveness
                              float64
         Valence
                              float64
         Tempo
                              float64
         dtype: object
In [18]:
         cleaned_data['Explicit']=cleaned_data['Explicit'].astype('int')
          cleaned_data['Explicit']
In [19]:
Out[19]:
                0
         2
                0
          3
                0
         4
                0
         95
         96
               0
         97
                0
         98
                0
         99
         Name: Explicit, Length: 100, dtype: int32
In [20]: | features=['Danceability', 'Valence']
          x=cleaned data[features]
          y=cleaned_data['Popularity']
          scaler=StandardScaler()
          x_scaled=scaler.fit_transform(x)
          X_train,X_test,Y_train,Y_test=train_test_split(x_scaled,y,test_size=0.2,random_stat
In [21]:
In [22]:
         model=LinearRegression()
In [23]:
          model.fit(X_train,Y_train)
         LinearRegression()
Out[23]:
          y_pred=model.predict(X_test)
In [24]:
         y_pred
In [25]:
```

```
array([49.35952842, 46.8821292 , 48.73216991, 51.90165825, 52.37423224,
Out[25]:
                 50.3400056 , 47.81632694, 49.97176225, 46.27102562, 49.66480711,
                 53.49744559, 49.71086162, 50.79208592, 50.74596714, 53.31894992,
                 50.98846133, 49.00102128, 43.19213957, 53.51038667, 51.32723902])
In [26]:
          Y_test
          42
                43
Out[26]:
          94
                50
          37
                50
                59
          6
          79
                48
          32
                61
          24
                59
          0
                70
          33
                72
          93
                60
          68
                54
          60
                47
          57
                35
          2
                64
          54
                63
                54
          72
          78
                58
          16
                59
                59
          25
          27
                53
          Name: Popularity, dtype: int64
         mse= mean_squared_error(Y_test,y_pred)
In [27]:
In [28]:
          mse
          121.45929939785506
Out[28]:
In [29]:
          accuracy =r2_score(Y_test,y_pred)
 In [ ]:
          coefficients = pd.Series(model.coef_, index=features)
In [31]:
          coefficients
          Danceability
                          1.121259
Out[31]:
                          1.793277
          Valence
          dtype: float64
In [33]:
          #y=m1x+m2x+c
 In [ ]:
```