

Music Data Analysis

Albums Sales & Popularity Data Analysis

James Bond 12/18/2019

In [39]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures
from sklearn.model_selection import train_test_split
```

In [131]:

```
albums = pd.read_csv("./albums.csv")
print(f>Data Shape: {week.shape}<div data-bbox="64 185 306 201" data-label="Text">


Data Shape: (300600, 8)


```

	date	Rank	Title	Artist
0	1955-01-01	1	Poor Little Fool	Ricky Nelson
1	1955-01-01	2	Patricia	Perez Prado And His Orchestra
2	1955-01-01	3	Splish Splash	Bobby Darin
3	1955-01-01	4	Hard Headed Woman	Elvis Presley With The Jordanaires
4	1955-01-01	5	When	Kalin Twins

	Weeks on chart	year	Spotify_Popularity	Artist_Popularity
0	1.0	1955	-1	56
1	NaN	1955	-1	12
2	NaN	1955	-1	62
3	NaN	1955	-1	-1
4	NaN	1955	-1	24

Data Shape: (5300, 6)

	year	Rank	Artist	Title
0	1960	1	Percy Faith	Theme From "A Summer Place"
1	1960	2	Jim Reeves	He'll Have To Go
2	1960	3	Everly Brothers	Cathy's Clown
3	1960	4	Johnny Preston	Running Bear
4	1960	5	Mark Dinning	Teen Angel

	Spotify_Popularity	Artist_Popularity
0	52	52
1	-1	60
2	-1	62
3	-1	34
4	-1	22

First five rows of weekly popularity data set

In [133]:

albums.head(5)

Out[133]:

	id	artist_id	album_title	genre	year_of_pub	num_of_tracks	num_of_sales	rolling_stone_criti
0	1	1767	Call me Cat Moneyless That Doggies	Folk	2006	11	905193	4.
1	2	23548	Down Mare	Metal	2014	7	969122	3.
2	3	17822	Embarrassed Hungry	Latino	2000	11	522095	2.
3	4	19565	Standard Immediate Engineer Slovakia	Pop	2017	4	610116	1.
4	5	24941	Decent Distance Georgian	Black Metal	2010	8	151111	4.

First five rows of yearly data set

In [43]:

year.head(5)

Out[43]:

	year	Rank	Artist	Title	Spotify_Popularity	Artist_Popularity
0	1960	1	Percy Faith	Theme From "A Summer Place"	52	52
1	1960	2	Jim Reeves	He'll Have To Go	-1	60
2	1960	3	Everly Brothers	Cathy's Clown	-1	62
3	1960	4	Johnny Preston	Running Bear	-1	34
4	1960	5	Mark Dinning	Teen Angel	-1	22

In [134]:

```
print(albums.describe, year.describe)
```

```

<bound method NDFrame.describe of          id  artist_id
album_title      genre  \
0          1      1767      Call me Cat Moneyless That Doggies
Folk
1          2      23548                                Down Mare      M
etal
2          3      17822                                Embarrassed Hungry      La
tino
3          4      19565      Standard Immediate Engineer Slovakia
Pop
4          5      24941                                Decent Distance Georgian      Black M
etal
...          ...          ...          ...
...
99995      99996      44624      Mike Pies Malay Albanian Terrible      Pop-
Rock
99996      99997      16345                                Global      R
etro
99997      99998      32674                                MINI      I
ndie
99998      99999      10134                                Marketing Belligerent Toe
Pop
99999      100000      41286                                Lover Barbie Of Rock
Rock

```

```

          year_of_pub  num_of_tracks  num_of_sales  rolling_stone_critic
\
0          2006          11          905193          4.0
1          2014          7          969122          3.0
2          2000          11          522095          2.5
3          2017          4          610116          1.5
4          2010          8          151111          4.5
...          ...          ...          ...          ...
99995      2016          3          871655          2.5
99996      2013          14          146202          5.0
99997      2018          4          620452          2.0
99998      2007          7          643276          4.0
99999      2014          7          466962          3.5

```

```

          mtv_critic  music_maniac_critic
0          1.5          3.0
1          4.0          5.0
2          1.0          2.0
3          2.0          4.0
4          2.5          1.0
...          ...          ...
99995      1.5          1.0
99996      2.5          1.0
99997      4.0          5.0
99998      1.5          4.0
99999      4.5          2.5

```

```

[100000 rows x 10 columns]> <bound method NDFrame.describe of          yea
r  Rank          Artist          Title  \
0      1960      1      Percy Faith      Theme From "A Summer Place"
1      1960      2          Jim Reeves      He'll Have To Go
2      1960      3      Everly Brothers      Cathy's Clown

```

3	1960	4	Johnny Preston	Running Bear
4	1960	5	Mark Dinning	Teen Angel
...
5295	2012	96	Kip Moore	Somethin' 'Bout A Truck
5296	2012	97	Miguel	Adorn
5297	2012	98	Jason Aldean	Fly Over States
5298	2012	99	Eli Young Band	Even If It Breaks Your Heart
5299	2012	100	Linkin Park	Burn It Down

	Spotify_Popularity	Artist_Popularity
0	52	52
1	-1	60
2	-1	62
3	-1	34
4	-1	22
...
5295	65	68
5296	64	84
5297	61	78
5298	66	65
5299	71	86

[5300 rows x 6 columns]>

In [136]:

```
print(albums.info, year.info)
```

```

<bound method DataFrame.info of                                id  artist_id
album_title      genre  \
0              1      1767      Call me Cat Moneyless That Doggies
Folk
1              2      23548                                Down Mare      M
etal
2              3      17822                                Embarrassed Hungry      La
tino
3              4      19565      Standard Immediate Engineer Slovakia
Pop
4              5      24941                                Decent Distance Georgian      Black M
etal
...          ...          ...          ...
...
99995      99996      44624      Mike Pies Malay Albanian Terrible      Pop-
Rock
99996      99997      16345                                Global      R
etro
99997      99998      32674                                MINI      I
ndie
99998      99999      10134                                Marketing Belligerent Toe
Pop
99999      100000      41286                                Lover Barbie Of Rock
Rock

```

```

      year_of_pub  num_of_tracks  num_of_sales  rolling_stone_critic
\
0              2006              11          905193              4.0
1              2014              7          969122              3.0
2              2000              11          522095              2.5
3              2017              4          610116              1.5
4              2010              8          151111              4.5
...          ...          ...          ...          ...
99995      2016              3          871655              2.5
99996      2013              14          146202              5.0
99997      2018              4          620452              2.0
99998      2007              7          643276              4.0
99999      2014              7          466962              3.5

```

```

      mtv_critic  music_maniac_critic
0              1.5              3.0
1              4.0              5.0
2              1.0              2.0
3              2.0              4.0
4              2.5              1.0
...          ...          ...
99995      1.5              1.0
99996      2.5              1.0
99997      4.0              5.0
99998      1.5              4.0
99999      4.5              2.5

```

```

[100000 rows x 10 columns]> <bound method DataFrame.info of                                year
Rank      Artist      Title  \
0      1960      1      Percy Faith      Theme From "A Summer Place"
1      1960      2      Jim Reeves      He'll Have To Go
2      1960      3      Everly Brothers      Cathy's Clown

```


3	1960	4	Johnny Preston	Running Bear
4	1960	5	Mark Dinning	Teen Angel
...
5295	2012	96	Kip Moore	Somethin' 'Bout A Truck
5296	2012	97	Miguel	Adorn
5297	2012	98	Jason Aldean	Fly Over States
5298	2012	99	Eli Young Band	Even If It Breaks Your Heart
5299	2012	100	Linkin Park	Burn It Down

	Spotify_Popularity	Artist_Popularity
0	52	52
1	-1	60
2	-1	62
3	-1	34
4	-1	22
...
5295	65	68
5296	64	84
5297	61	78
5298	66	65
5299	71	86

```
[5300 rows x 6 columns]>
```

```
In [137]:
```

```
X = year['Spotify_Popularity'].values
y = year['Artist_Popularity'].values
```

```
In [140]:
```

```
X
```

```
Out[140]:
```

```
array([[52],
       [-1],
       [-1],
       ...,
       [61],
       [66],
       [71]])
```

```
In [141]:
```

```
X=X.reshape(-1,1)
X
```

```
Out[141]:
```

```
array([[52],
       [-1],
       [-1],
       ...,
       [61],
       [66],
       [71]])
```

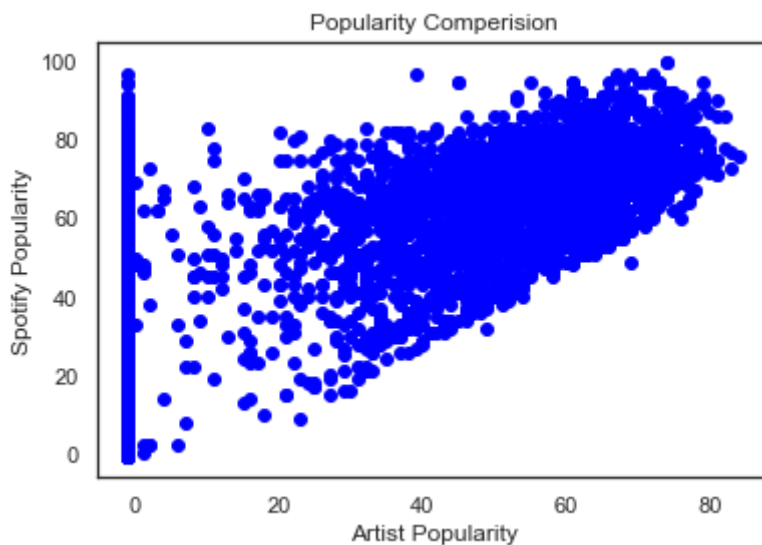
In [142]:

```
x_train, x_test, y_train, y_test = train_test_split(X,y,train_size=0.8,test_size=0.2,random_state=100)
print(f"X_train shape {x_train.shape}")
print(f"y_train shape {y_train.shape}")
print(f"X_test shape {x_test.shape}")
print(f"y_test shape {y_test.shape}")
print(y_test)
```

```
X_train shape (4240, 1)
y_train shape (4240,)
X_test shape (1060, 1)
y_test shape (1060,)
[ 88  61  -1 ...  79 100  49]
```

In [144]:

```
%matplotlib inline
plt.scatter(x_train,y_train,color='blue')
plt.xlabel('Artist Popularity' )
plt.ylabel('Spotify Popularity')
plt.title('Popularity Comperision')
plt.show()
```



In [145]:

```
lm = LinearRegression()
lm.fit(x_train,y_train)
y_predict = lm.predict(x_test)
print(f"Train accuracy {round(lm.score(x_train,y_train)*100,2)} %")
print(f"Test accuracy {round(lm.score(x_test,y_test)*100,2)} %")
```

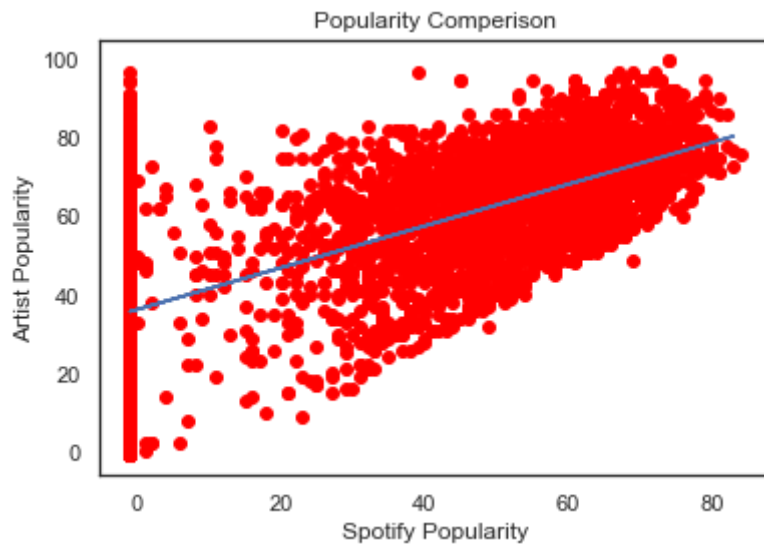
```
Train accuracy 33.88 %
Test accuracy 32.56 %
```

In [146]:

```
plt.scatter(x_train,y_train,color='red')
plt.plot(x_test,y_predict)
plt.xlabel("Spotify Popularity")
plt.ylabel("Artist Popularity")
plt.title("Popularity Comperison")
plt.plot
```

Out[146]:

```
<function matplotlib.pyplot.plot(*args, scalex=True, scaley=True, data=
None, **kwargs)>
```



In [147]:

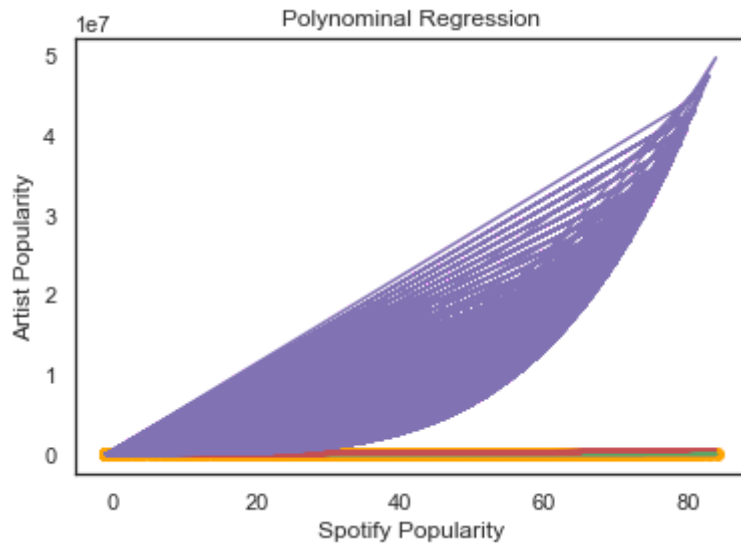
```
poly = PolynomialFeatures(degree = 4)
X_poly = poly.fit_transform(x_train)
poly.fit(X_poly,y_train)
```

Out[147]:

```
PolynomialFeatures(degree=4, include_bias=True, interaction_only=False,
                    order='C')
```

In [148]:

```
%matplotlib inline
plt.scatter(x_train,y_train,color='orange')
plt.plot(x_train, poly.fit_transform(x_train))
plt.xlabel('Spotify Popularity')
plt.ylabel('Artist Popularity')
plt.title('Polynomial Regression')
plt.show()
```



In [149]:

```
uniform_data = np.random.rand(10,12)
data = pd.DataFrame({'x':np.arange(1,10001),'y':np.random.normal(0,4,10000)})
year = pd.read_csv("./year.csv")
print(f>Data Shape: {year.shape}")

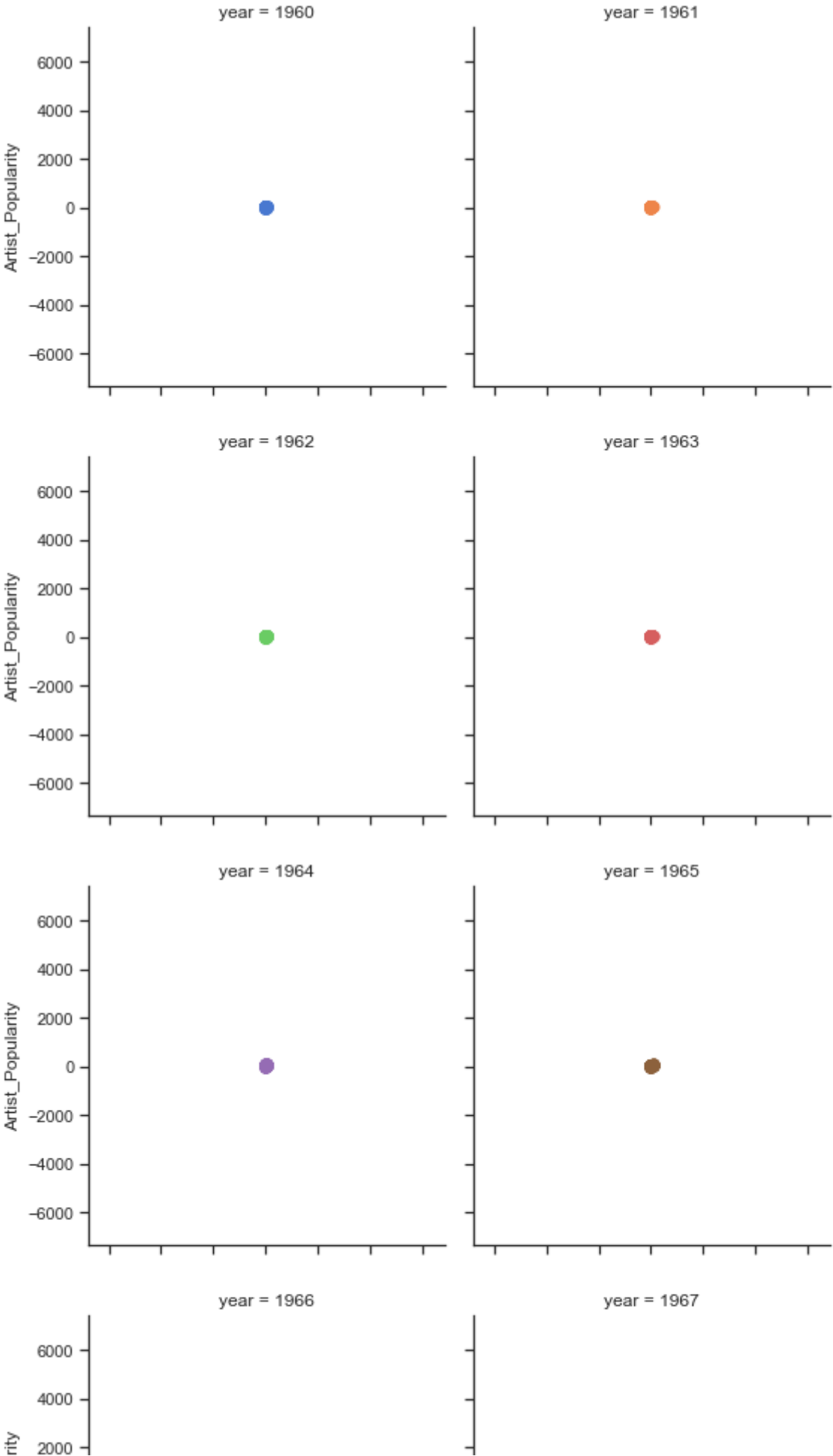
sns.set(style="ticks")

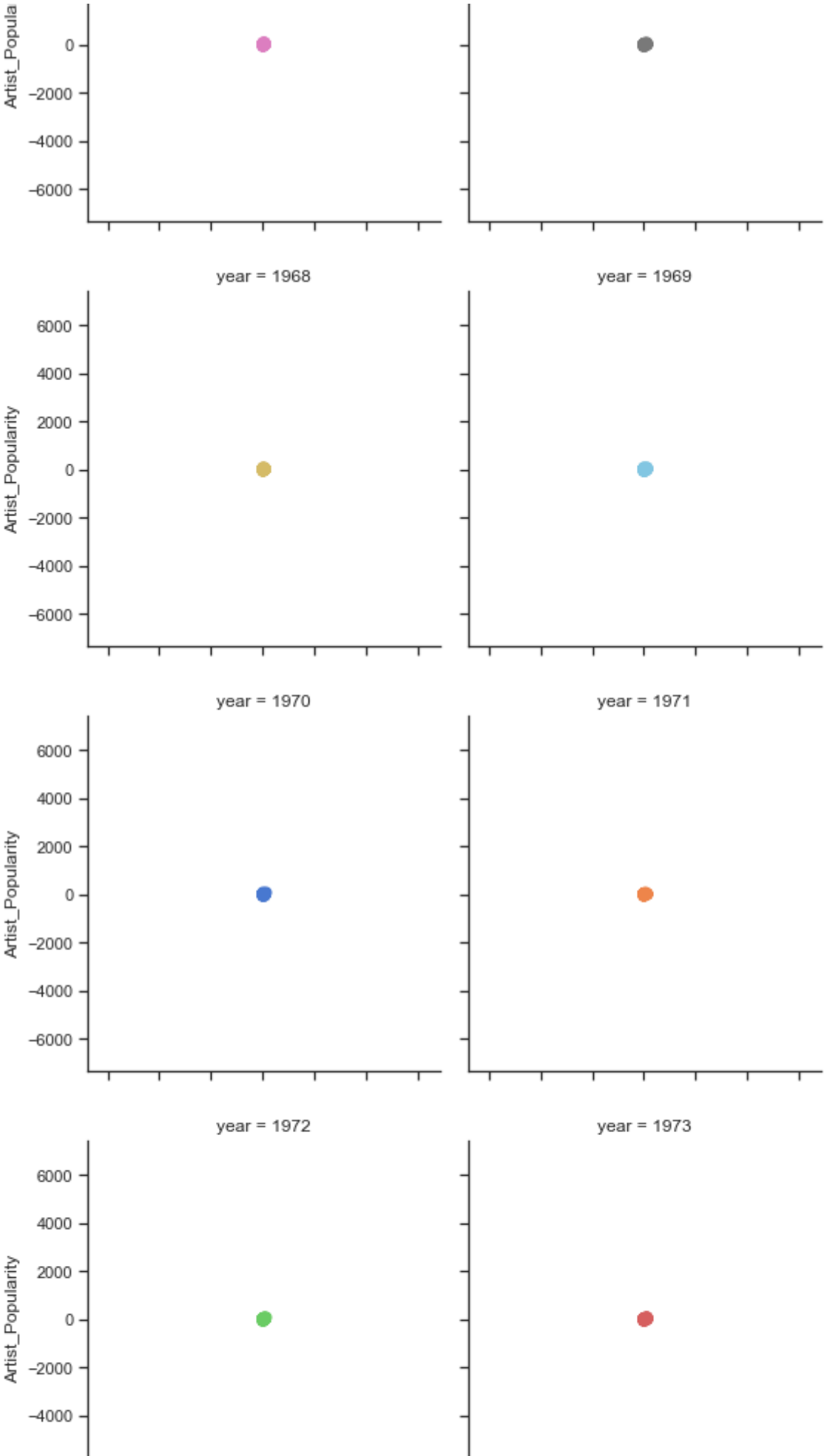
sns.lmplot(x="Spotify_Popularity", y="Artist_Popularity", col="year", hue="year", d
ata=year,
          col_wrap=2, ci=None, palette="muted", height=4,
          scatter_kws={"s": 50, "alpha": 1})
```

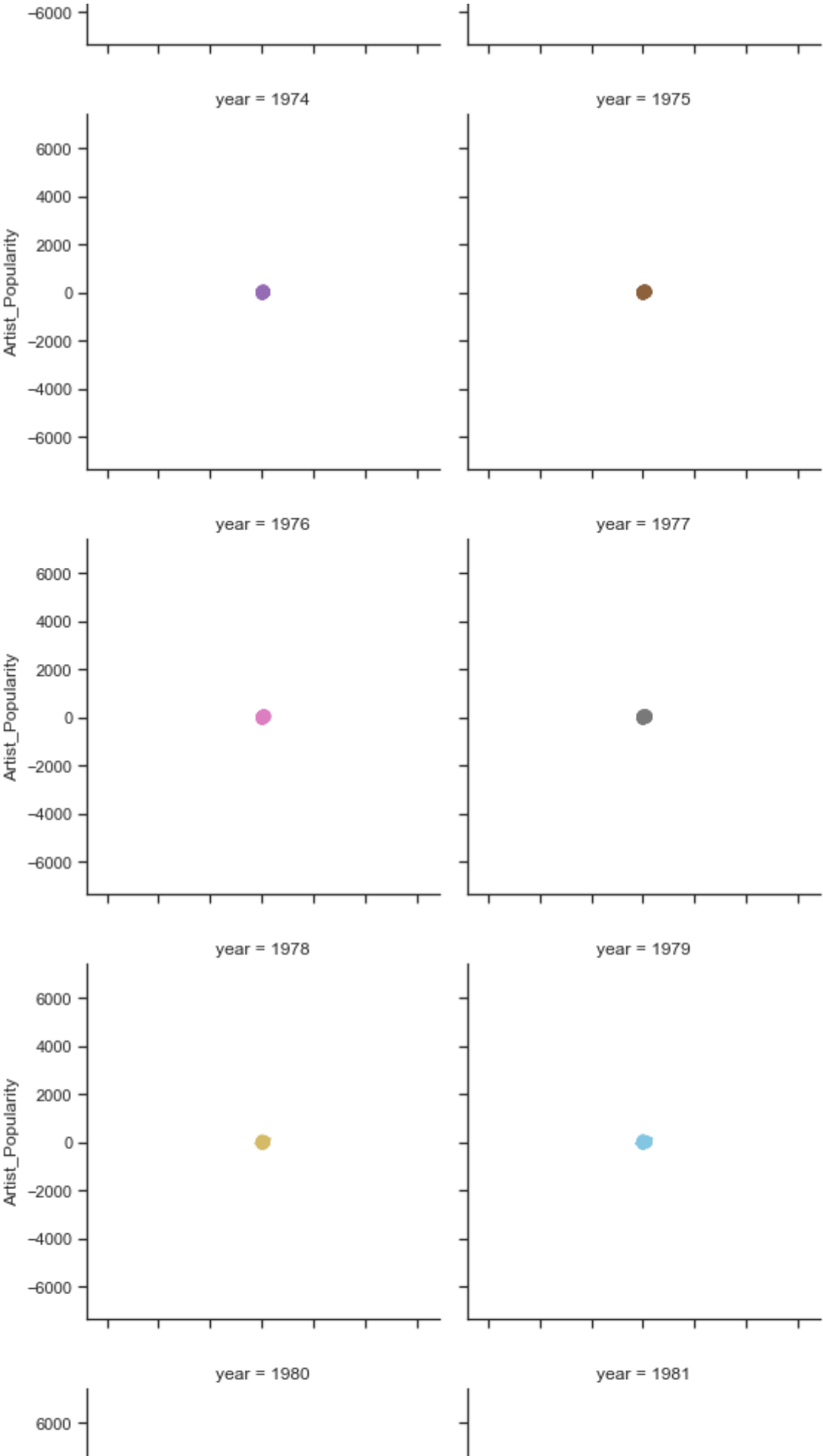
```
Data Shape: (5300, 6)
```

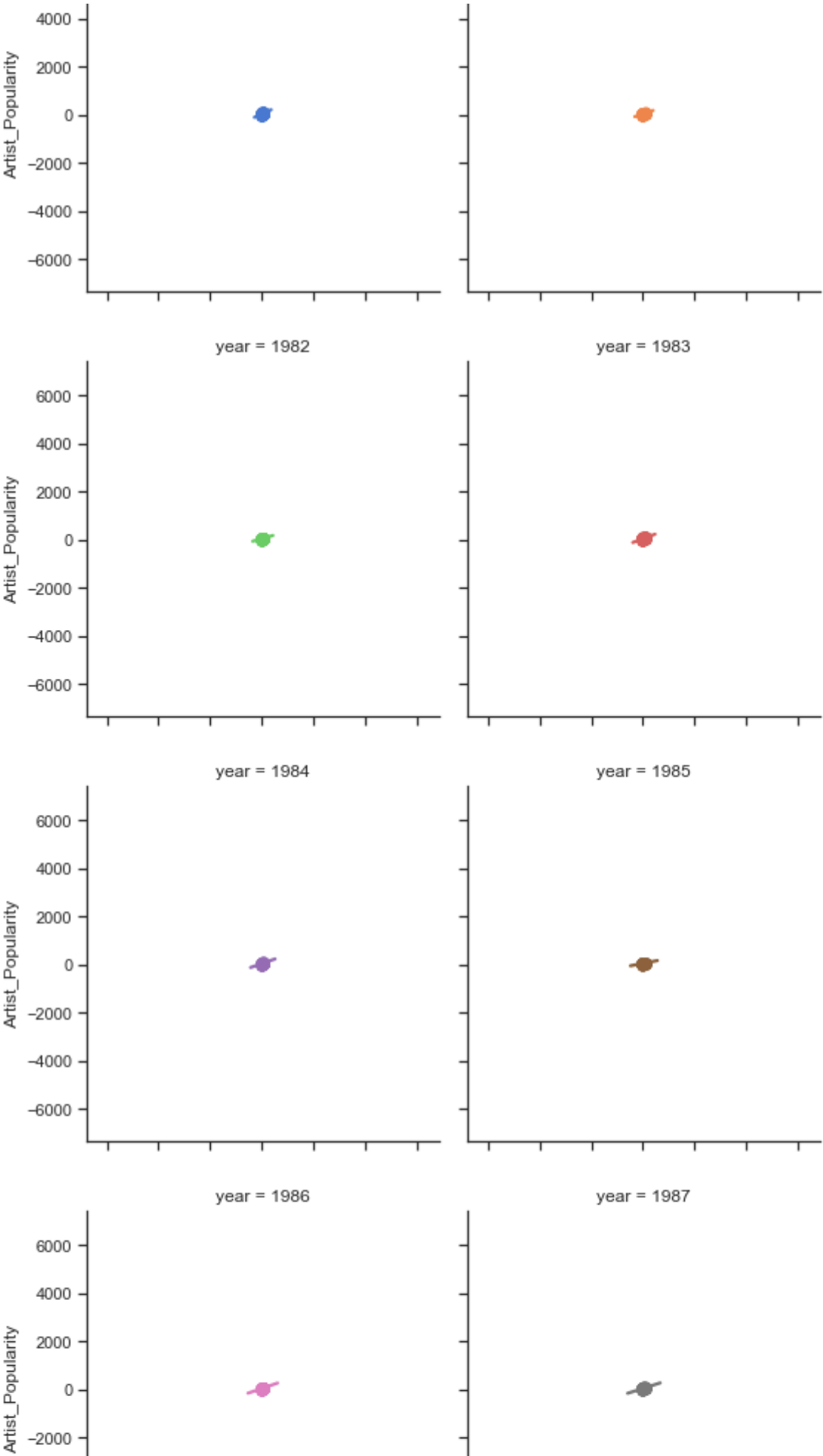
```
Out[149]:
```

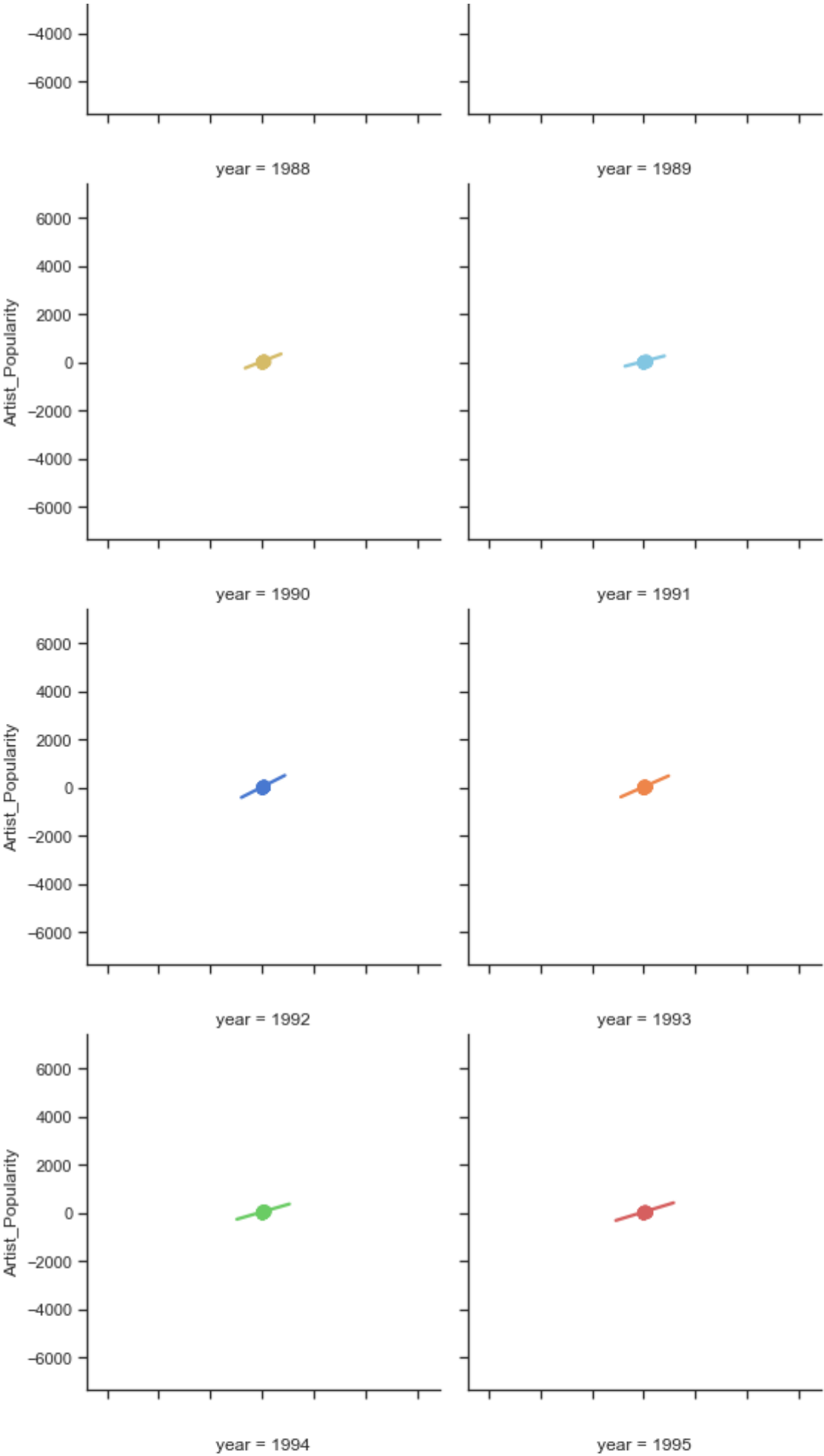
```
<seaborn.axisgrid.FacetGrid at 0x1451d8890>
```

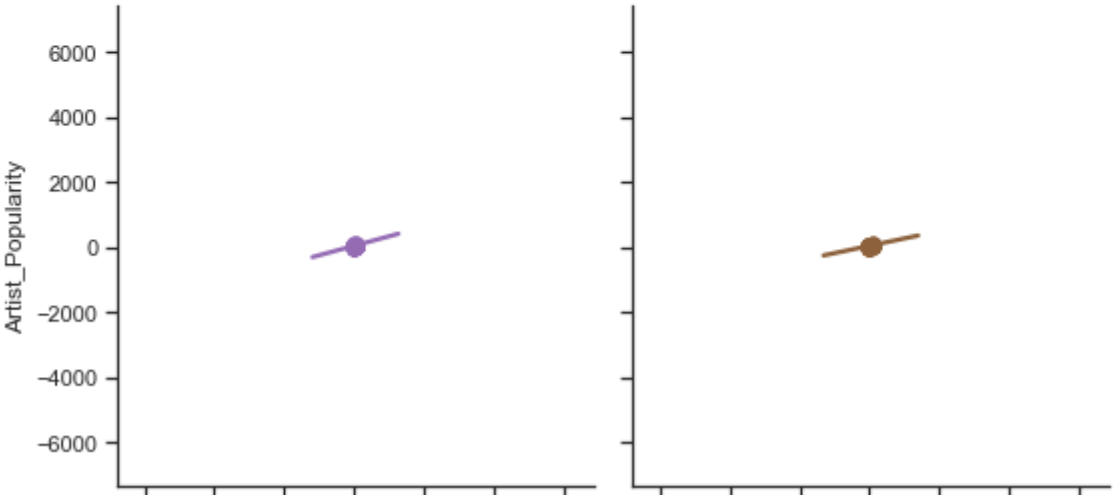






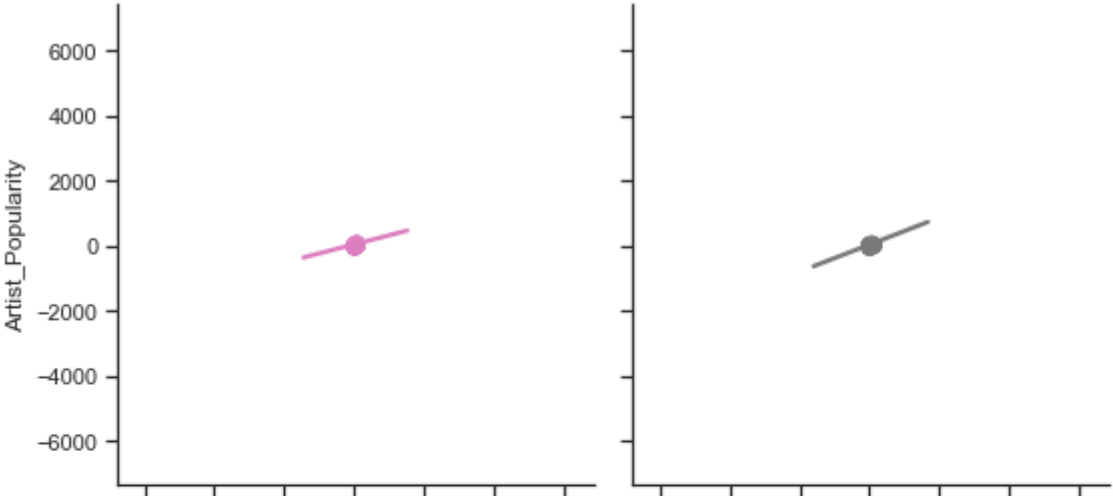






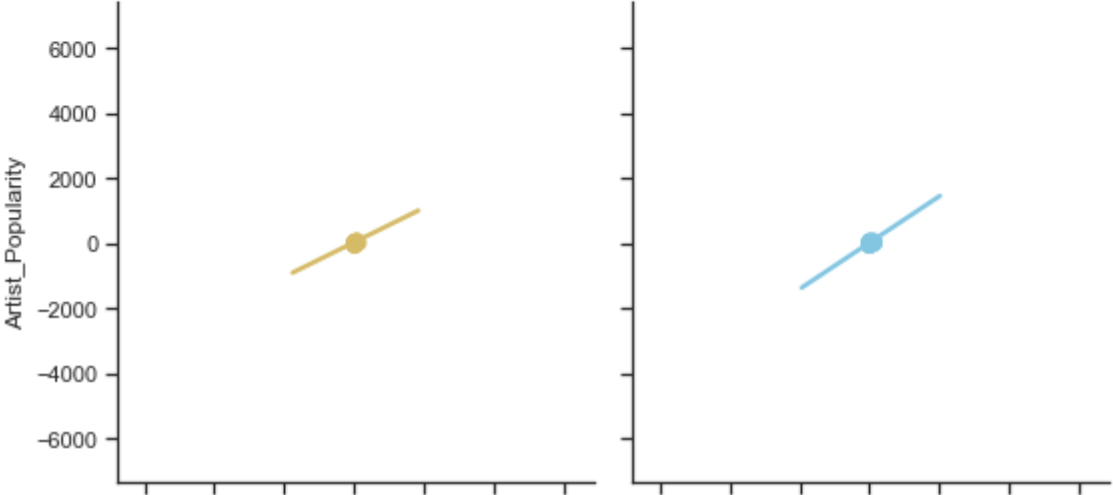
year = 1996

year = 1997



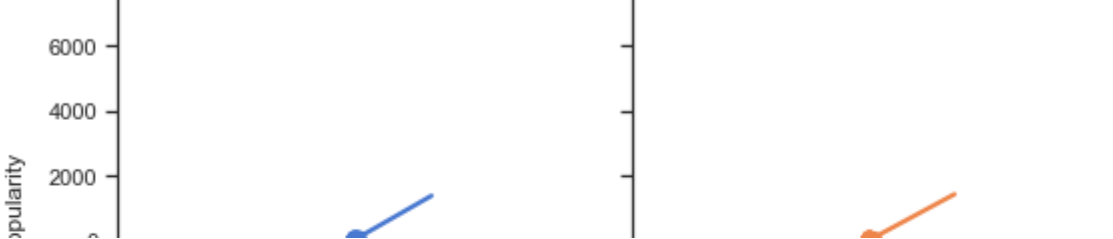
year = 1998

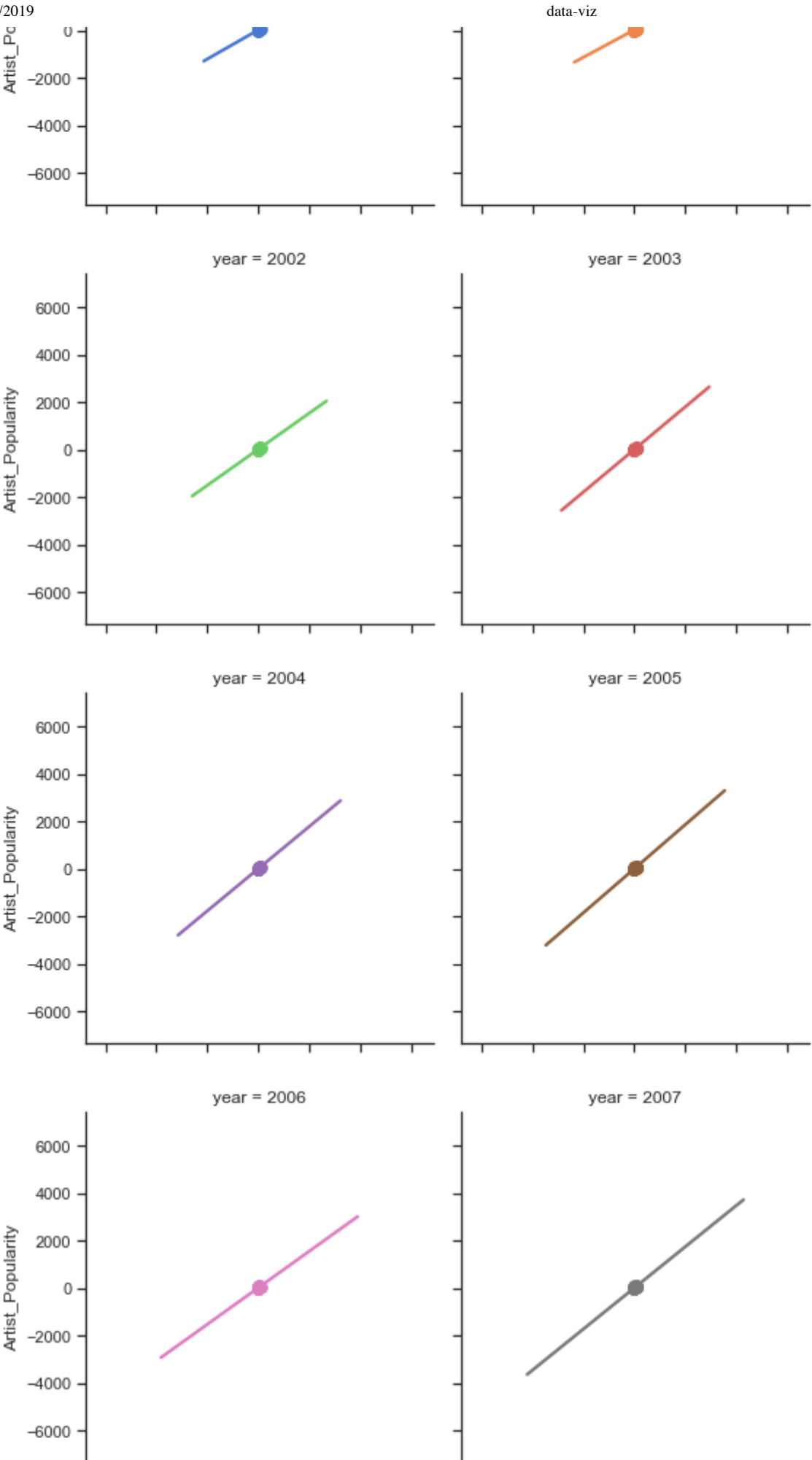
year = 1999

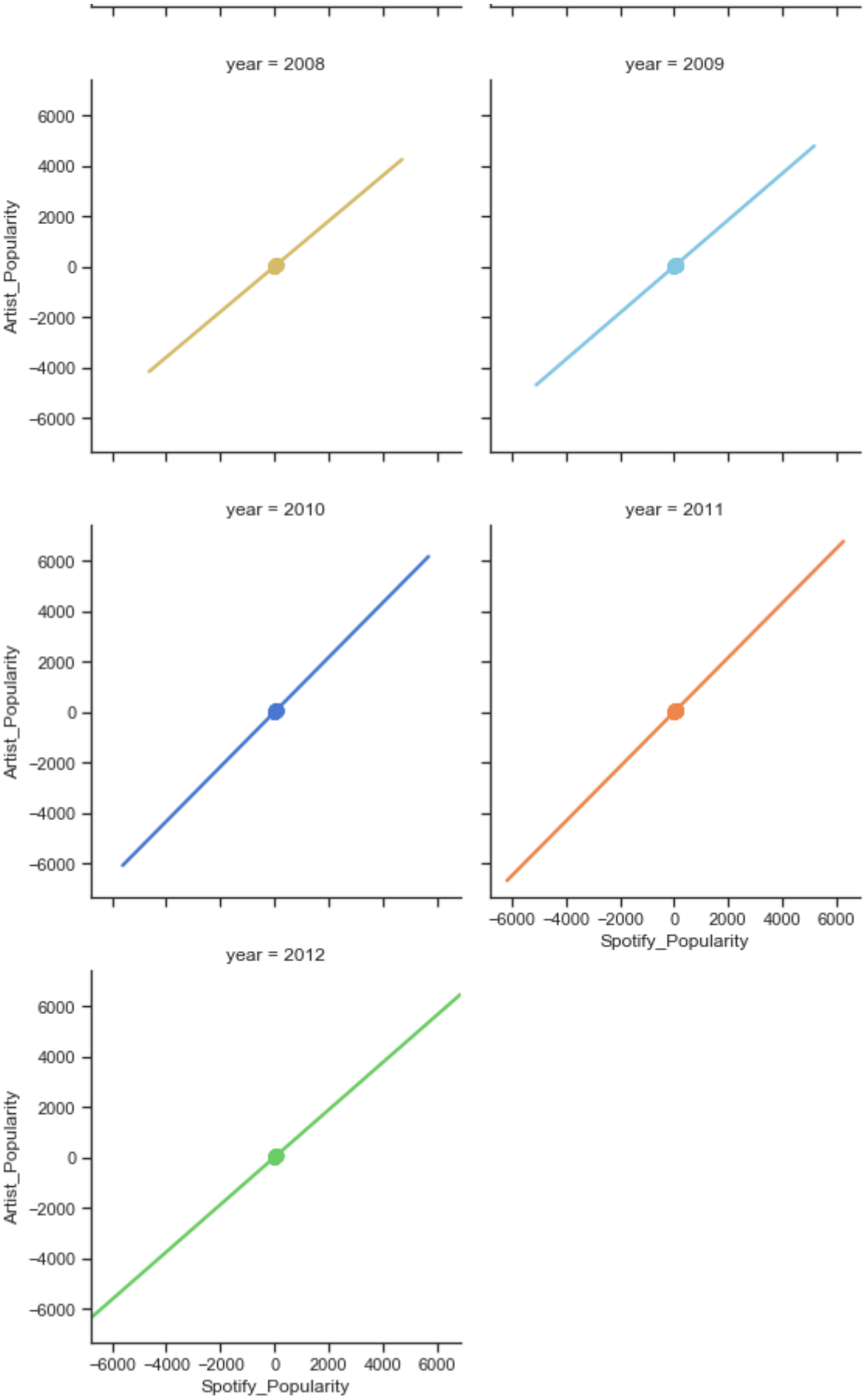


year = 2000

year = 2001







In my data analysis using mathlibpot and seaborn spotify popularity and artist popularity is starting show colleration after 2005. Spotify is not old enough for 2003 however Spotify users do not show popularity to songs earlier than 2003. We have tested it 100 , 500, 1000 and 10000 random variables really breaking point year is 2005.

In [150]:

```
sns.set(style="dark")
year = np.random.RandomState(5000)

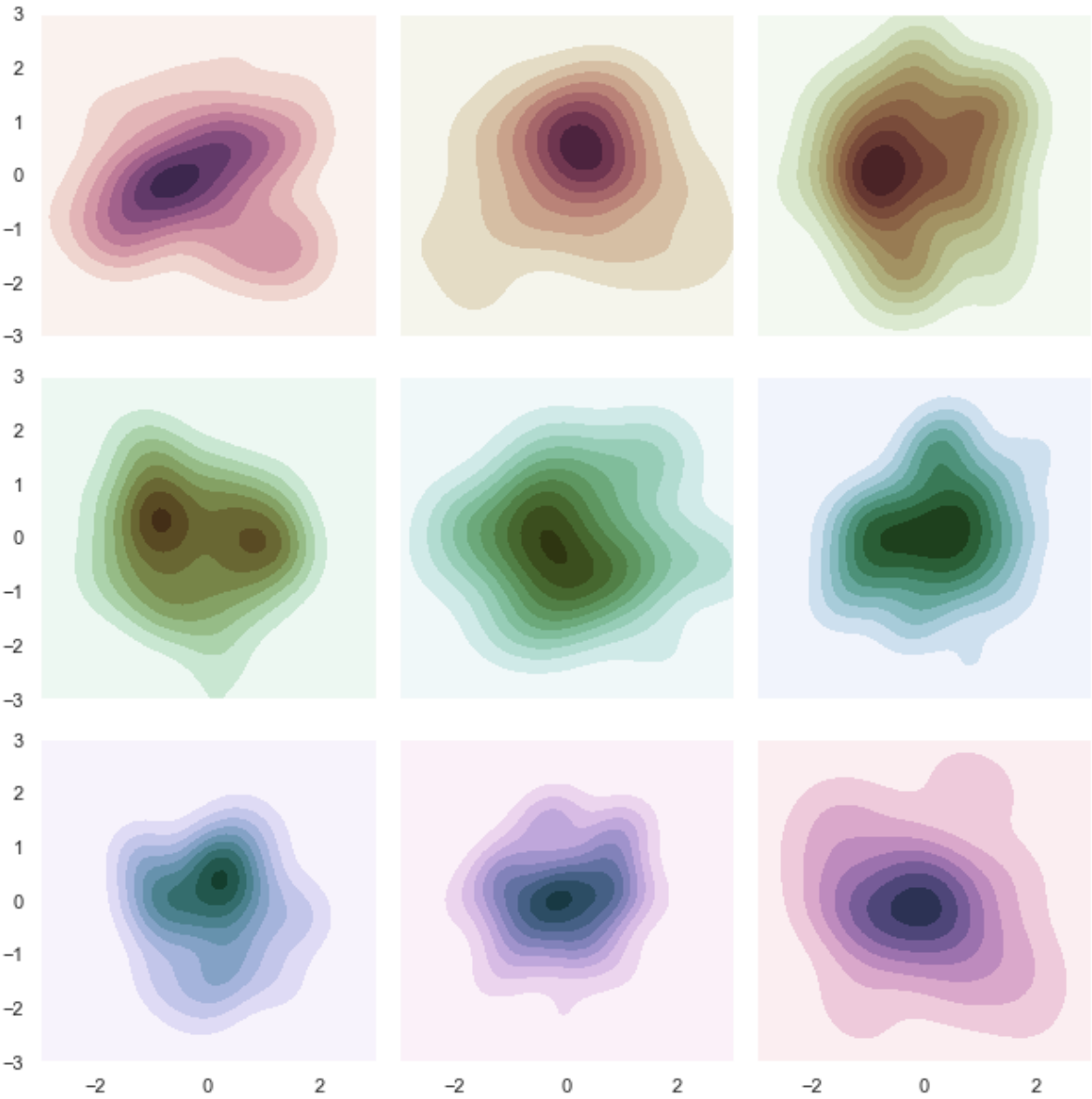
# Set up the matplotlib figure
f, axes = plt.subplots(3, 3, figsize=(9, 9), sharex=True, sharey=True)

# Rotate the starting point around the cubehelix hue circle
for ax, s in zip(axes.flat, np.linspace(0, 3, 10)):

    # Create a cubehelix colormap to use with kdeplot
    cmap = sns.cubehelix_palette(start=s, light=1, as_cmap=True)

    # Generate and plot a random bivariate dataset
    x, y = year.randn(2, 50)
    sns.kdeplot(x, y, cmap=cmap, shade=True, cut=5, ax=ax)
    ax.set(xlim=(-3, 3), ylim=(-3, 3))

f.tight_layout()
```

Also above analysis shows that majority of spotify popularity is not cover up artist popularity

In [151]:

```
albums['num_of_sales'] = albums['num_of_sales']/1000
```

In [152]:

```
albums.head(1)
```

Out[152]:

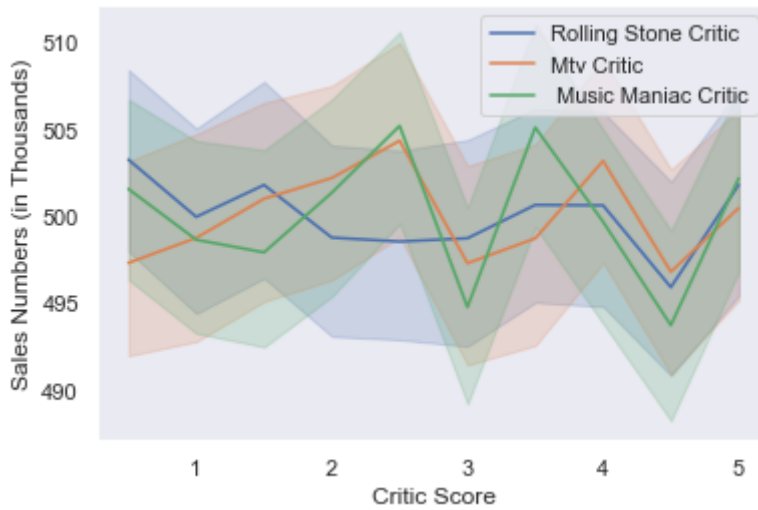
	id	artist_id	album_title	genre	year_of_pub	num_of_tracks	num_of_sales	rolling_stone_critic
0	1	1767	Call me Cat Moneyless That Doggies	Folk	2006	11	905.193	4.0

In [160]:

```
sns.lineplot(x="rolling_stone_critic", y="num_of_sales", data=albums, label="Rolling Stone Critic")
sns.lineplot(x="mtv_critic", y="num_of_sales", data=albums, label="Mtv Critic")
sns.lineplot(x="music_maniac_critic", y="num_of_sales", data=albums, label="Music Maniac Critic")
plt.xlabel('Critic Score')
plt.ylabel('Sales Numbers (in Thousands)')
```

Out[160]:

Text(0, 0.5, 'Sales Numbers (in Thousands)')



Here is my another analysis about 3 Music Critic authorities, Rolling Stone, MTV and Music Maniac. From this analysis 3 Authorities mostly same opinion and good catch giving top critic for high sale albums. Rolling Stone seems more accurate about it. Music Maniac critics the most independent than sales numbers, MTV critics between 1 and 3 is more accurate about sales slope. Rolling Stone behaves giving bad critics by not caring of sales numbers or their album critic opinions are the farrest from the reality of music market. Music Maniacs are behaving in similar line, MTV has good catch about giving low critics to lower sales number albums. This graphic tells us that the critics of these 3 authorities are matching the sales trends of the albums well.

In []: