

Entertainer data analysis

April 29, 2024

1 Entertainer Data Analysis

```
[25]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[7]: df1 = pd.read_csv('Entertainer - Basic Info.csv')
df2 = pd.read_csv('Entertainer - Breakthrough Info.csv')
df3 = pd.read_csv('Entertainer - Last work Info.csv')
```

```
[8]: mer_df = pd.merge(df1, df2, how='left')
```

```
[9]: mer_data = pd.merge(mer_df, df3, how = 'left')
mer_data.head(20)
```

```
[9]:      Entertainer Gender (traditional) Birth Year \
0      Adele F 1988
1  Angelina Jolie F 1975
2  Aretha Franklin F 1942
3      Bette Davis F 1908
4      Betty White F 1922
5      Bing Crosby M 1903
6      Bob Hope M 1903
7      Carol Burnett F 1933
8      Carole Lombard F 1908
9      Carrie Fisher F 1956
10     Cary Grant M 1904
11  Charlie Chaplin M 1889
12     Clara Bow F 1905
13     Clark Gable M 1901
14  David Letterman M 1947
15  Debbie Reynolds F 1932
16  Denzel Washington M 1954
17     Dick Van Dyke M 1925
18  Donald Sutherland M 1935
19   Dustin Hoffman M 1937
```

	Year of Breakthrough/#1 Hit/Award Nomination \
0	2008
1	1999
2	1967
3	1934
4	1952
5	1931
6	1938
7	1959
8	1934
9	1977
10	1933
11	1915
12	1926
13	1934
14	1982
15	1952
16	1989
17	1961
18	1967
19	1967

	Breakthrough Name	Year of First Oscar/Grammy/Emmy \
0	19	2009.0
1	Girl, Interrupted	1999.0
2	I Never Loved a Man (The Way I Love You)	1968.0
3	Of Human Bondage	1935.0
4	Life with Elilzabeth	1976.0
5	Several Songs	1962.0
6	The Big Broadcast of 1938	1940.0
7	The Garry Moore Show	1962.0
8	Twentieth Century	NaN
9	Star Wars	NaN
10	She Done Him Wrong, I'm No Angel	1970.0
11	The Tramp	1929.0
12	Mantrap	NaN
13	It Happened One Night	1934.0
14	Late Night with David Letterman	1981.0
15	Singin' in the Rain	NaN
16	Glory	1989.0
17	Bye Bye Birdie, The Dick Van Dyke Show	1964.0
18	The Dirty Dozen	1995.0
19	The Graduate	1980.0

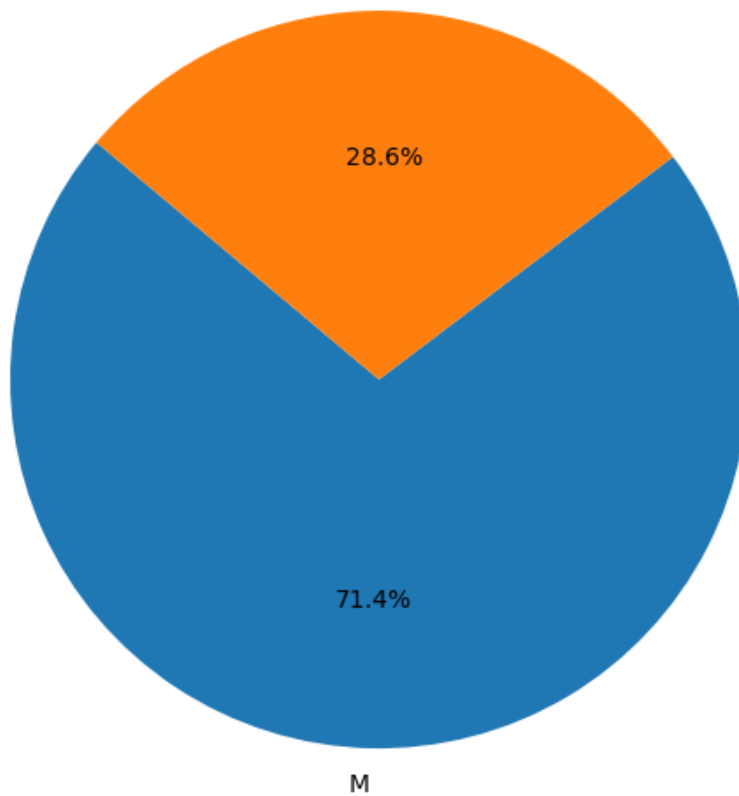
	Year of Last Major Work (arguable)	Year of Death
0	2016	NaN
1	2016	NaN

2	2014	NaN
3	1989	1989.0
4	2016	NaN
5	1974	1977.0
6	1972	2003.0
7	2016	NaN
8	1942	1942.0
9	2016	2016.0
10	1966	1986.0
11	1967	1977.0
12	1933	1965.0
13	1960	1960.0
14	2015	NaN
15	2006	2016.0
16	2016	NaN
17	2015	NaN
18	2016	NaN
19	2016	NaN

2 Explanatory Data Analysis

```
[56]: # Vizualization of number of males and females in the field of entertainment
gender_counts = mer_data['Gender (traditional)'].value_counts()
plt.figure(figsize=(8, 6))
plt.pie(gender_counts, labels=gender_counts.index, autopct='%1.1f%%',
        ↪startangle=140)
plt.title('Distribution of Males and Females')
plt.axis('equal')
plt.show()
```

Distribution of Males and Females



```
[81]: numerical_columns = mer_data.select_dtypes(include=['number'])
      numerical_columns.corr()
```

```
[81]:
```

	Birth Year \
Birth Year	1.000000
Year of Breakthrough/#1 Hit/Award Nomination	0.941740
Year of First Oscar/Grammy/Emmy	0.772863
Year of Last Major Work (arguable)	0.723677
Year of Death	0.561124

	Year of Breakthrough/#1 Hit/Award
Nomination \	
Birth Year	0.941740
Year of Breakthrough/#1 Hit/Award Nomination	1.000000
Year of First Oscar/Grammy/Emmy	0.842745
Year of Last Major Work (arguable)	

0.748675
Year of Death
0.511714

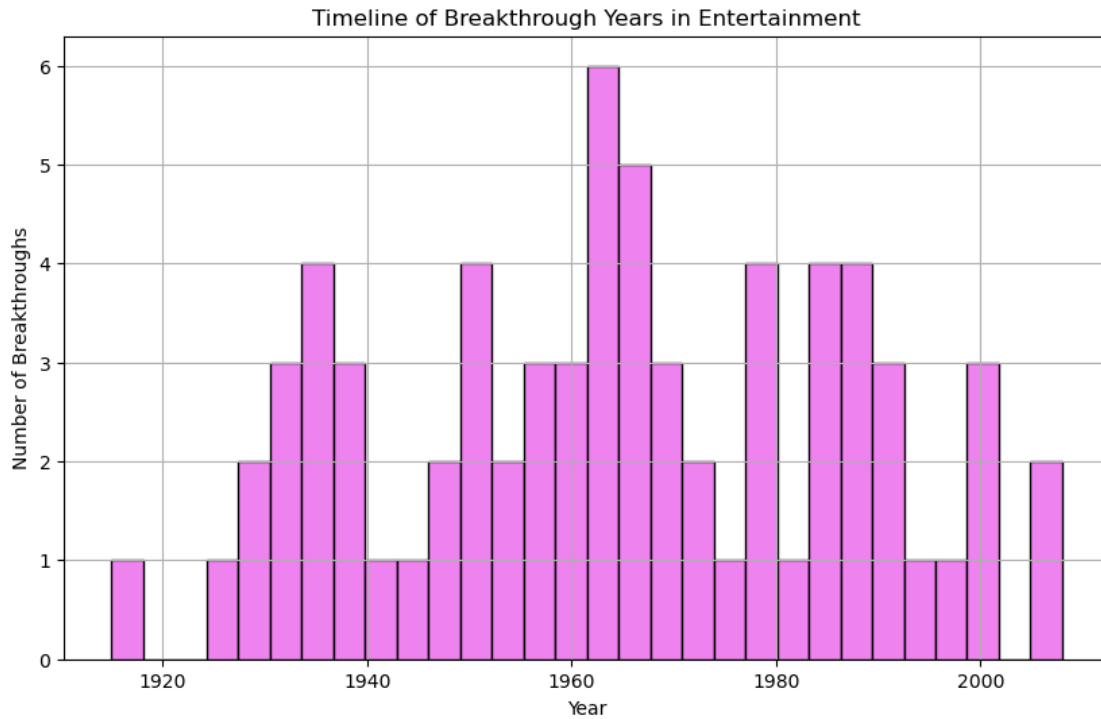
	Year of First Oscar/Grammy/Emmy \
Birth Year	0.772863
Year of Breakthrough/#1 Hit/Award Nomination	0.842745
Year of First Oscar/Grammy/Emmy	1.000000
Year of Last Major Work (arguable)	0.608658
Year of Death	0.360906

	Year of Last Major Work (arguable)
\	
Birth Year	0.723677
Year of Breakthrough/#1 Hit/Award Nomination	0.748675
Year of First Oscar/Grammy/Emmy	0.608658
Year of Last Major Work (arguable)	1.000000
Year of Death	0.848694

	Year of Death
Birth Year	0.561124
Year of Breakthrough/#1 Hit/Award Nomination	0.511714
Year of First Oscar/Grammy/Emmy	0.360906
Year of Last Major Work (arguable)	0.848694
Year of Death	1.000000

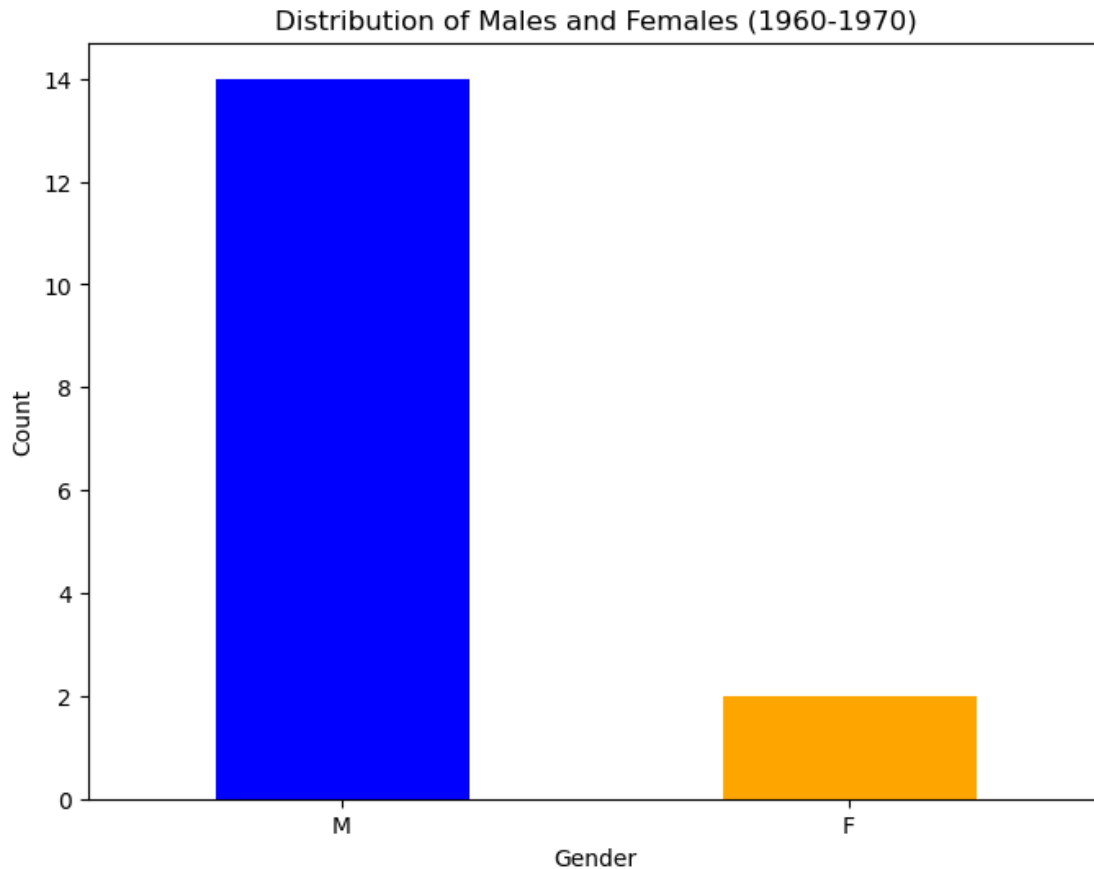
```
[33]: # visualizing the trending period of entertainment from this dataset
plt.figure(figsize=(10, 6))
plt.hist(mer_data['Year of Breakthrough/#1 Hit/Award Nomination'], bins=30,
        color='violet', edgecolor='black')
plt.title('Timeline of Breakthrough Years in Entertainment')
plt.xlabel('Year')
plt.ylabel('Number of Breakthroughs')
plt.grid(True)
plt.show()

# during the period of 1960-70, entertainment was the trending the most
```



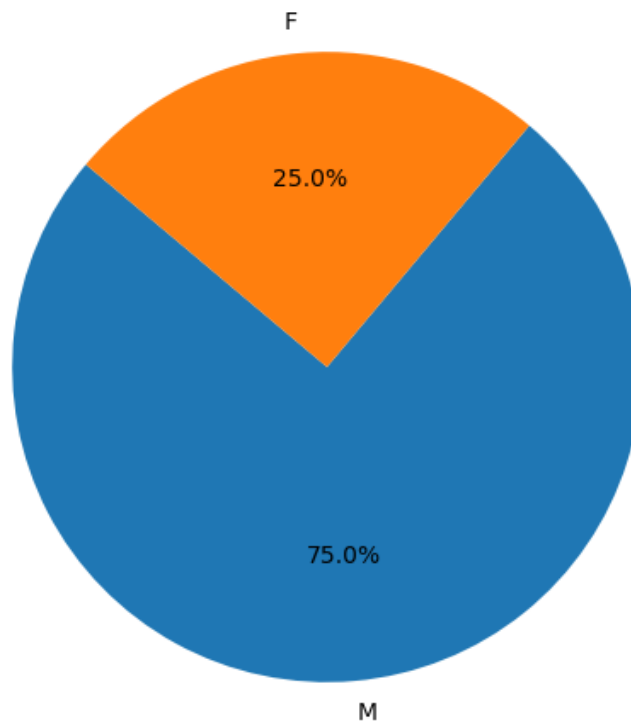
```
[39]: filtered_df = mer_data[(mer_data['Year of Breakthrough/#1 Hit/Award_
↳Nomination'] >= 1960) & (mer_data['Year of Breakthrough/#1 Hit/Award_
↳Nomination'] <= 1970)]
gender_counts = filtered_df['Gender (traditional)'].value_counts()
plt.figure(figsize=(8, 6))
gender_counts.plot(kind='bar', color=['blue', 'orange'])
plt.title('Distribution of Males and Females (1960-1970)')
plt.xlabel('Gender')
plt.ylabel('Count')
plt.xticks(rotation=0)
plt.show()
```

During this trending period, the number of male entertainer is high



```
[58]: modified_df = mer_data.copy()
modified_df['Year of First Oscar/Grammy/Emmy'] = modified_df['Year of First
↳Oscar/Grammy/Emmy'].notna().astype(int)
winners_df = modified_df[modified_df['Year of First Oscar/Grammy/Emmy'] == 1]
gender_counts = winners_df['Gender (traditional)'].value_counts()
plt.figure(figsize=(6, 6))
plt.pie(gender_counts, labels=gender_counts.index, autopct='%1.1f%%',
↳startangle=140)
plt.title('Distribution of Males and Females Who Won an Oscar/Grammy/Emmy
↳Award')
plt.show()
```

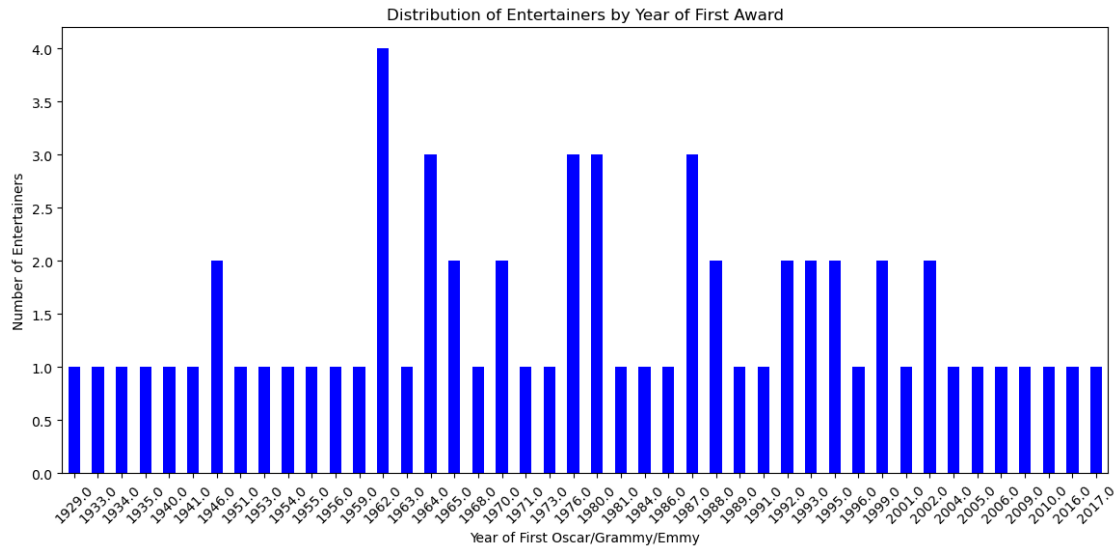
Distribution of Males and Females Who Won an Oscar/Grammy/Emmy Award



```
[85]: # Visualize the year in which maximum number of Oscar was won by Entertainers
award_year_counts = mer_data['Year of First Oscar/Grammy/Emmy'].value_counts().
    ↪sort_index()

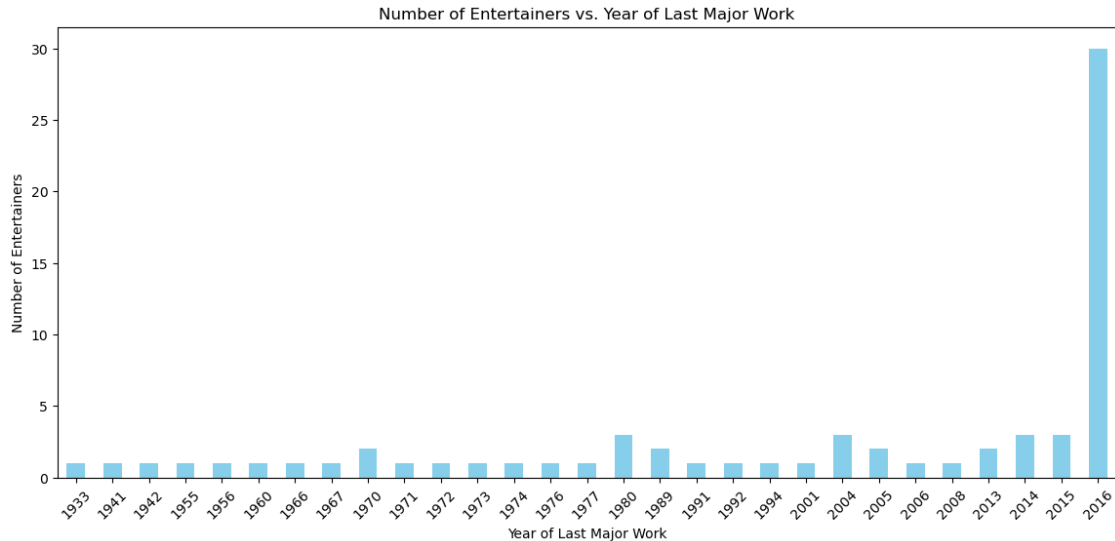
# Plot the bar graph
plt.figure(figsize=(14, 6))
award_year_counts.plot(kind='bar', color='blue')
plt.title('Distribution of Entertainers by Year of First Award')
plt.xlabel('Year of First Oscar/Grammy/Emmy')
plt.ylabel('Number of Entertainers')
plt.xticks(rotation=45) # Rotate x-axis labels if necessary
plt.show()

# In 1962 the Oscar award were awarded to maximum number of entertainers
```

```
[72]: # visualizing the relationship between entertainers and their Year of Last
      ↪ Major Work (arguable).
year_counts = mer_data['Year of Last Major Work (arguable)'].value_counts().
      ↪ sort_index()
plt.figure(figsize=(14, 6))
year_counts.plot(kind='bar', color='skyblue')
plt.title('Number of Entertainers vs. Year of Last Major Work')
plt.xlabel('Year of Last Major Work')
plt.ylabel('Number of Entertainers')
plt.xticks(rotation=45) # Rotate x-axis labels if necessary
plt.show()

# Most of the entertainer's last work is in the year 2016
```



```
[79]: alive_df = mer_data.copy()
alive_df['Year of Death'].fillna('Alive', inplace=True)
num_alive = (alive_df['Year of Death'] == 'Alive').sum()
print("Number of entertainers who are still alive:", num_alive)
```

Number of entertainers who are still alive: 40

/tmp/ipykernel_3979/2298961232.py:2: FutureWarning: Setting an item of incompatible dtype is deprecated and will raise in a future error of pandas. Value 'Alive' has dtype incompatible with float64, please explicitly cast to a compatible dtype first.

```
alive_df['Year of Death'].fillna('Alive', inplace=True)
```

```
[80]: # visualizing the graph of entertainers who are alive

alive_df = mer_data.copy() # Make a copy to avoid modifying the original DataFrame
alive_df['Vital Status'] = alive_df['Year of Death'].apply(lambda x: 'Alive' if pd.isnull(x) else 'Not Alive')

# Count the number of entertainers in each vital status category
vital_status_counts = alive_df['Vital Status'].value_counts()

# Plot the pie chart
plt.figure(figsize=(6, 6))
plt.pie(vital_status_counts, labels=vital_status_counts.index, autopct='%1.1f%%', startangle=140)
plt.title('Distribution of Entertainers by Vital Status')
plt.axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle
```

```
plt.show()
```

Distribution of Entertainers by Vital Status

