cob-phase-2

February 12, 2024

0.0.1 1. Analyze the dataset and create graphs using seaborn and matplotlib.Dataset:

Import Necessary Libraries:

[1]:
```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

[2]:
```
# Load the dataset
df = pd.read_csv('/content/dataset - netflix1.csv')
df.head()
```

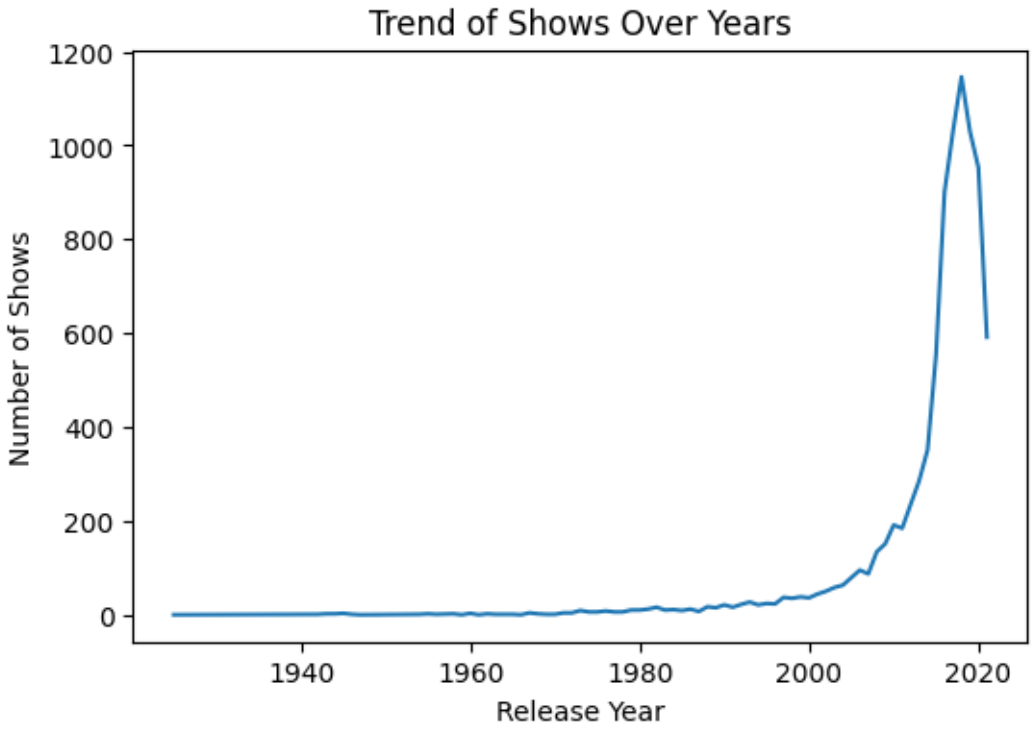
[2]:

	show_id	type	title	director	country	date_added	release_year	rating	duration	listed_in
0	s1	Movie	Dick Johnson Is Dead	Kirsten Johnson	United States	9/25/2021	2020	PG-13	90 min	Documentaries
1	s3	TV Show	Ganglands	Julien Leclercq	France	9/24/2021	2021	TV-MA	1 Season	Crime TV Shows, International TV Shows, TV Act…
2	s6	TV Show	Midnight Mass	Mike Flanagan	United States	9/24/2021	2021	TV-MA	1 Season	TV Dramas, TV Horror, TV Mysteries
3	s14	Movie	Confessions of an Invisible Girl	Bruno Garotti	Brazil	9/22/2021	2021	TV-PG	91 min	Children & Family Movies, Comedies
4	s8	Movie	Sankofa	Haile Gerima	United States	9/24/2021	1993	TV-MA	125 min	Dramas, Independent Movies, International Movies

Trend of Shows Over Years:

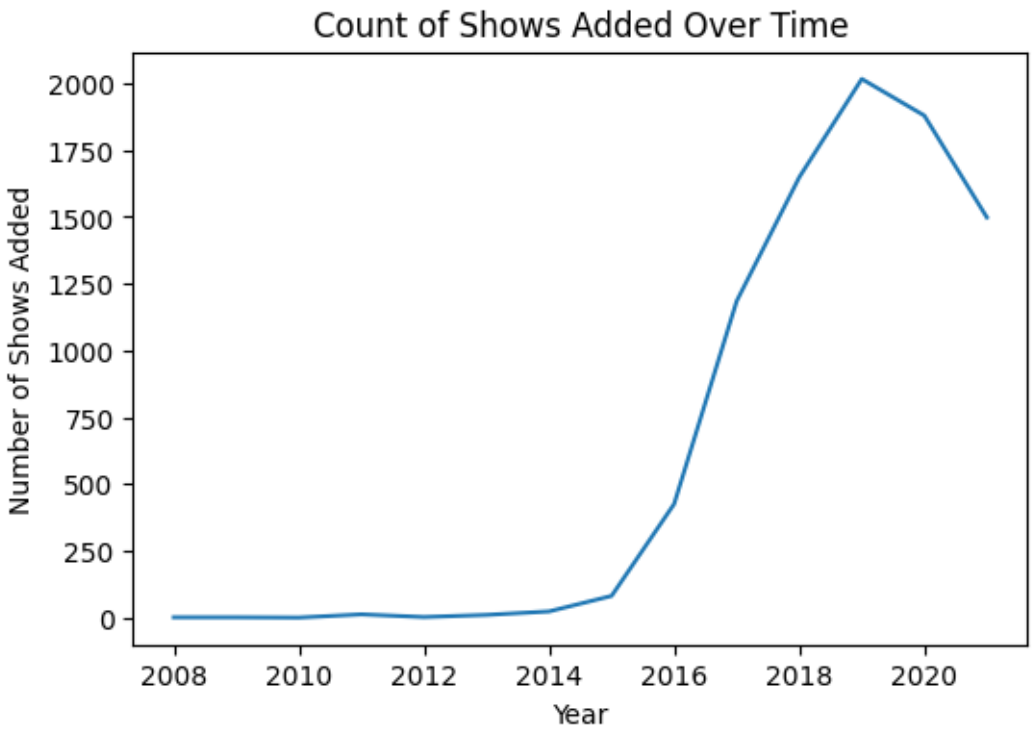
[3]:
```
plt.figure(figsize=(6, 4))
sns.lineplot(data=df.groupby('release_year').size())
plt.title('Trend of Shows Over Years')
plt.xlabel('Release Year')
```

```
plt.ylabel('Number of Shows')
plt.show()
```



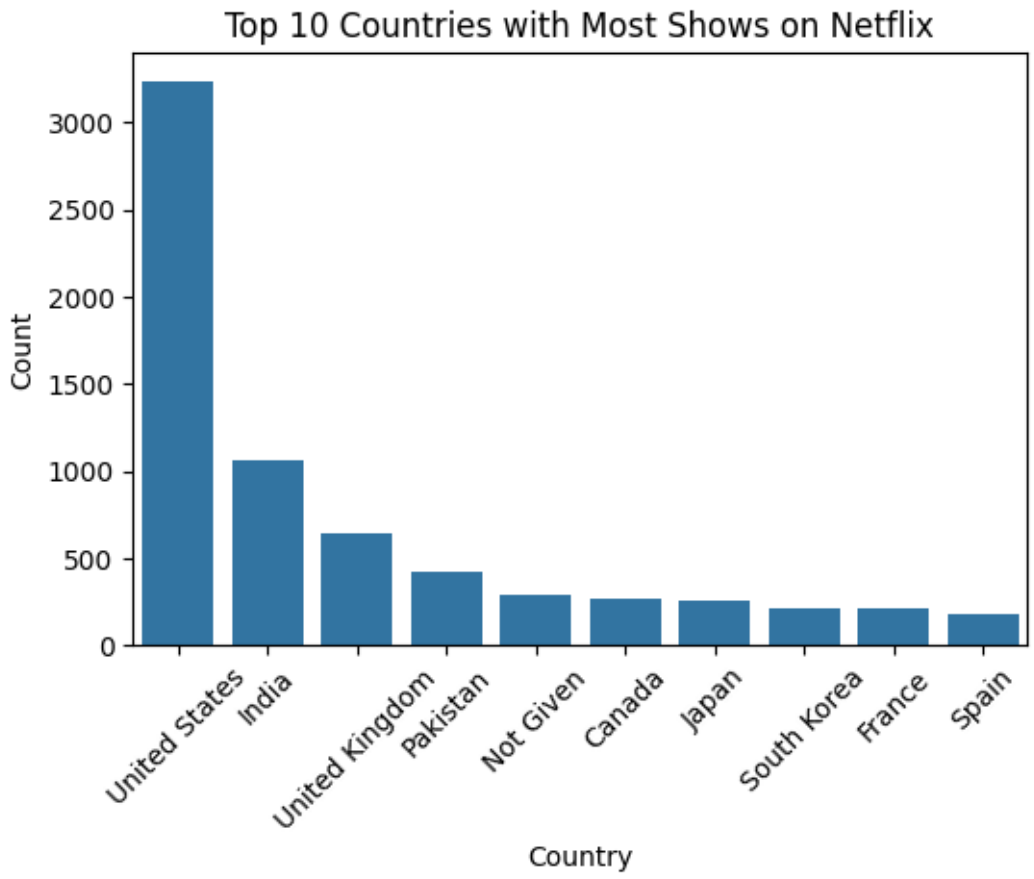
Count of Shows Added Over Time:

[4]:
```
df['date_added'] = pd.to_datetime(df['date_added'])
plt.figure(figsize=(6, 4))
sns.lineplot(data=df.groupby(df['date_added'].dt.year).size())
plt.title('Count of Shows Added Over Time')
plt.xlabel('Year')
plt.ylabel('Number of Shows Added')
plt.show()
```



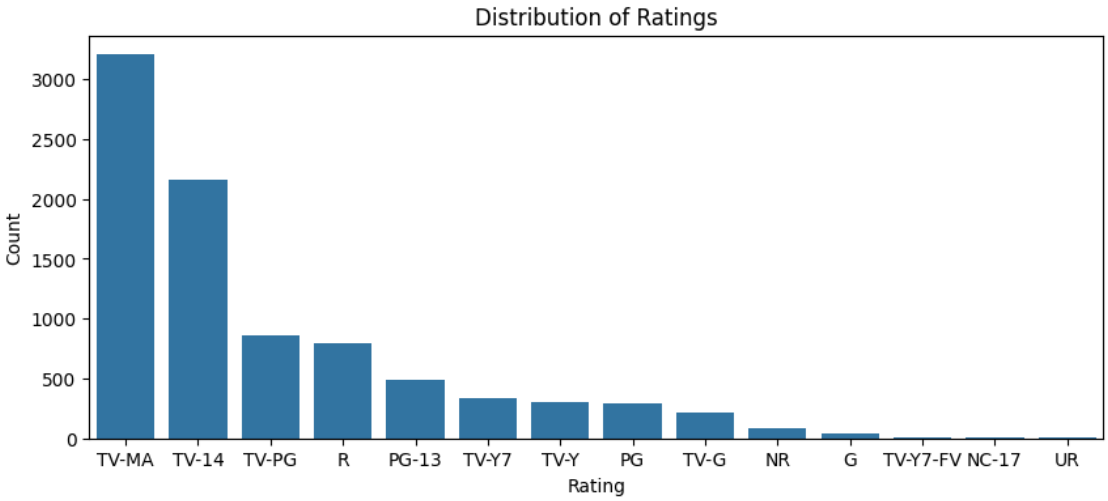
Count of Shows by Country:

[5]:
```
plt.figure(figsize=(6, 4))
sns.countplot(data=df, x='country', order=df['country'].value_counts().index[:
↪10])
plt.xticks(rotation=45)
plt.title('Top 10 Countries with Most Shows on Netflix')
plt.xlabel('Country')
plt.ylabel('Count')
plt.show()
```



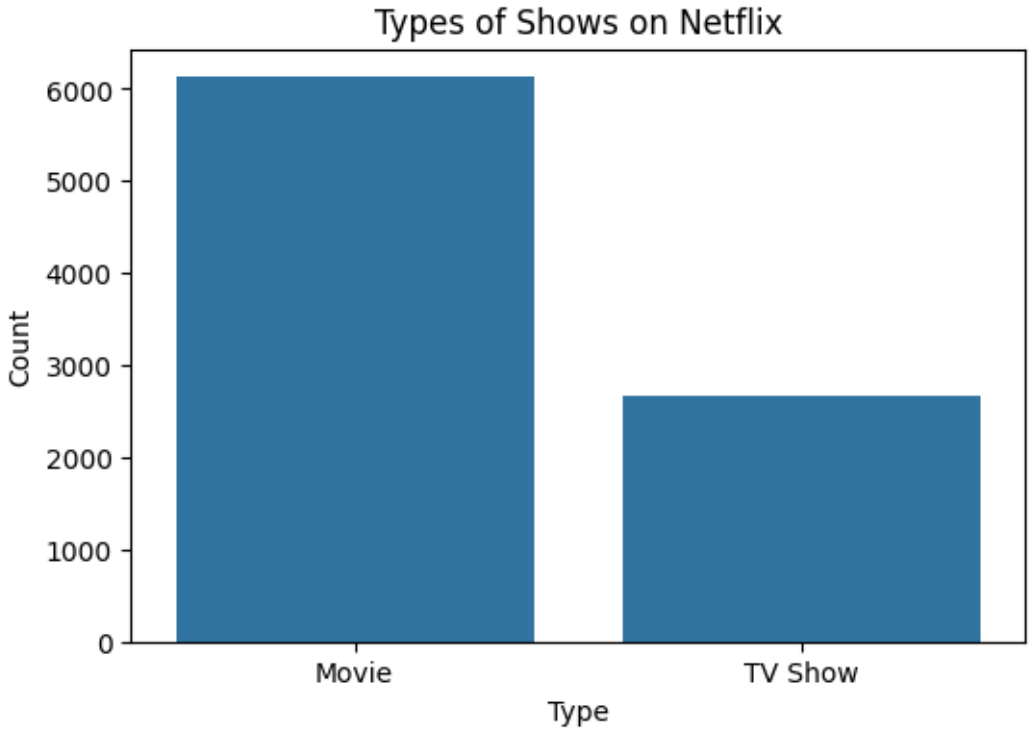
Distribution of Ratings:

[6]:
```
plt.figure(figsize=(10, 4))
sns.countplot(data=df, x='rating', order=df['rating'].value_counts().index)
plt.title('Distribution of Ratings')
plt.xlabel('Rating')
plt.ylabel('Count')
plt.show()
```



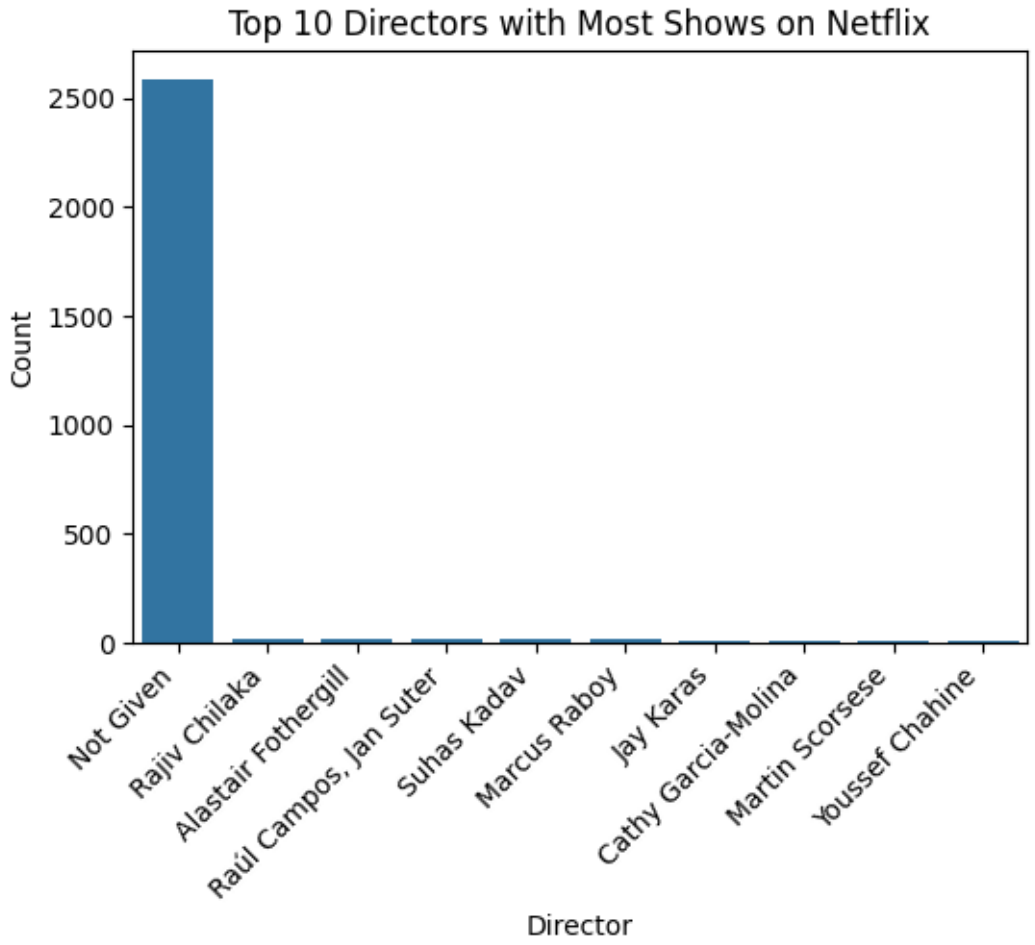
Types of Shows:

[7]:
```
plt.figure(figsize=(6, 4))
sns.countplot(data=df, x='type')
plt.title('Types of Shows on Netflix')
plt.xlabel('Type')
plt.ylabel('Count')
plt.show()
```



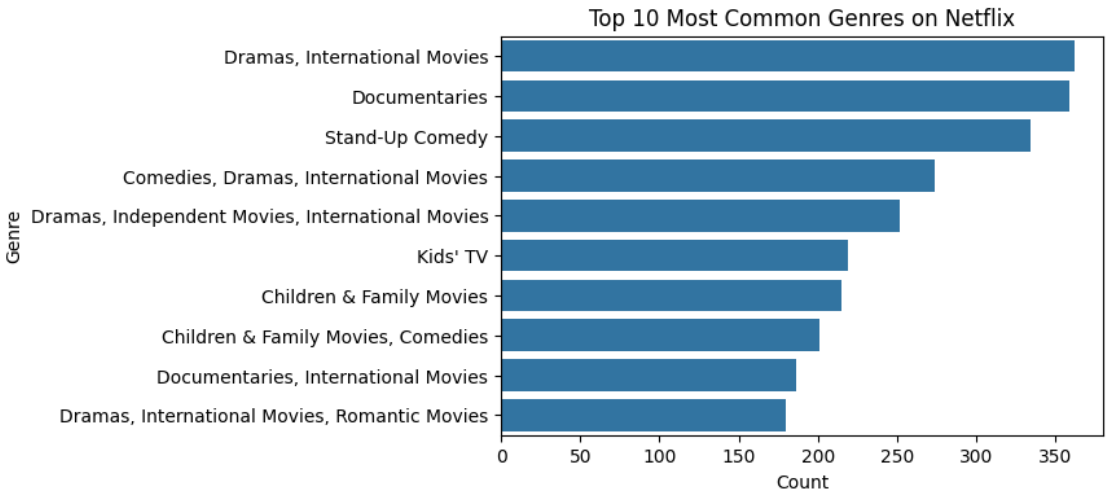
Top Directors with Most Shows:

[8]:
```
plt.figure(figsize=(6, 4))
sns.countplot(data=df, x='director', order=df['director'].value_counts().index[:
↪10])
plt.xticks(rotation=45)
plt.title('Top 10 Directors with Most Shows on Netflix')
plt.xlabel('Director')
plt.ylabel('Count')
plt.xticks(rotation=45, ha = 'right')
plt.show()
```



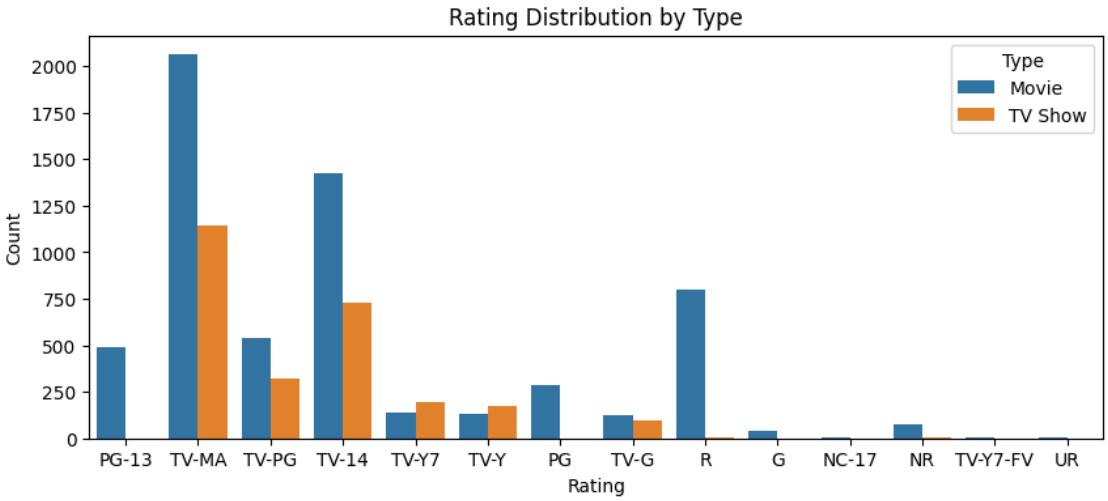
Most Common Genres:

[9]:
```
plt.figure(figsize=(6, 4))
sns.countplot(data=df, y='listed_in', order=df['listed_in'].value_counts().
↪index[:10])
plt.title('Top 10 Most Common Genres on Netflix')
plt.xlabel('Count')
plt.ylabel('Genre')
plt.show()
```



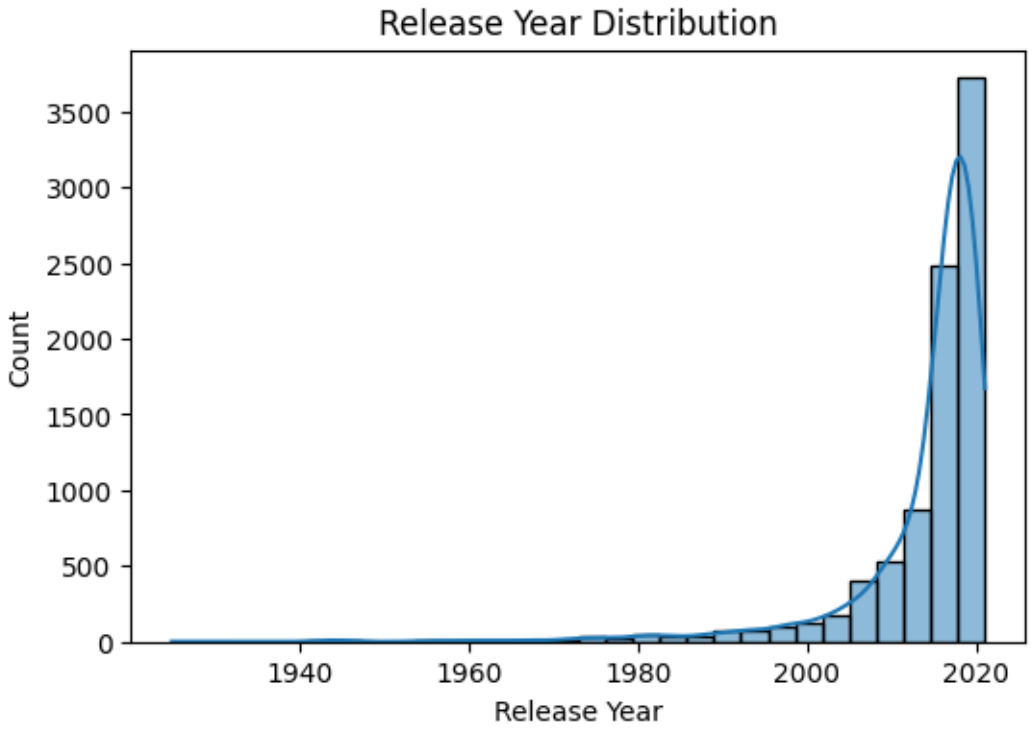
Rating Distribution by Type:

[10]:
```
plt.figure(figsize=(10, 4))
sns.countplot(data=df, x='rating', hue='type')
plt.title('Rating Distribution by Type')
plt.xlabel('Rating')
plt.ylabel('Count')
plt.legend(title='Type')
plt.show()
```



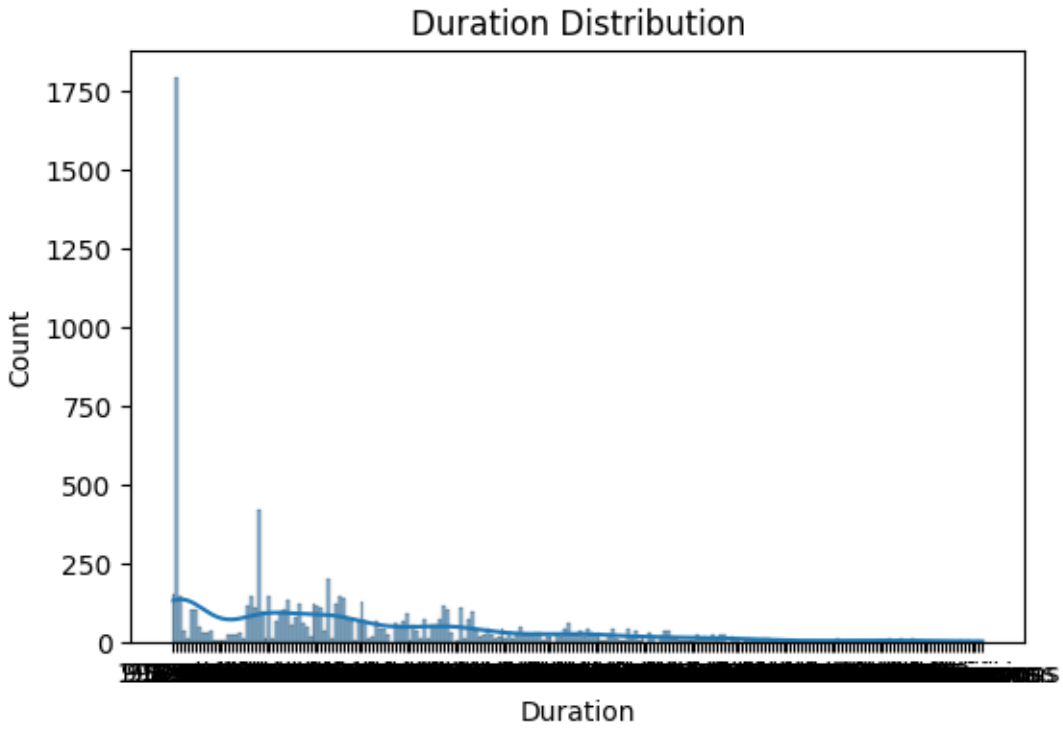
Release Year Distribution:

[11]:
```
plt.figure(figsize=(6, 4))
sns.histplot(data=df, x='release_year', bins=30, kde=True)
plt.title('Release Year Distribution')
plt.xlabel('Release Year')
plt.ylabel('Count')
plt.show()
```



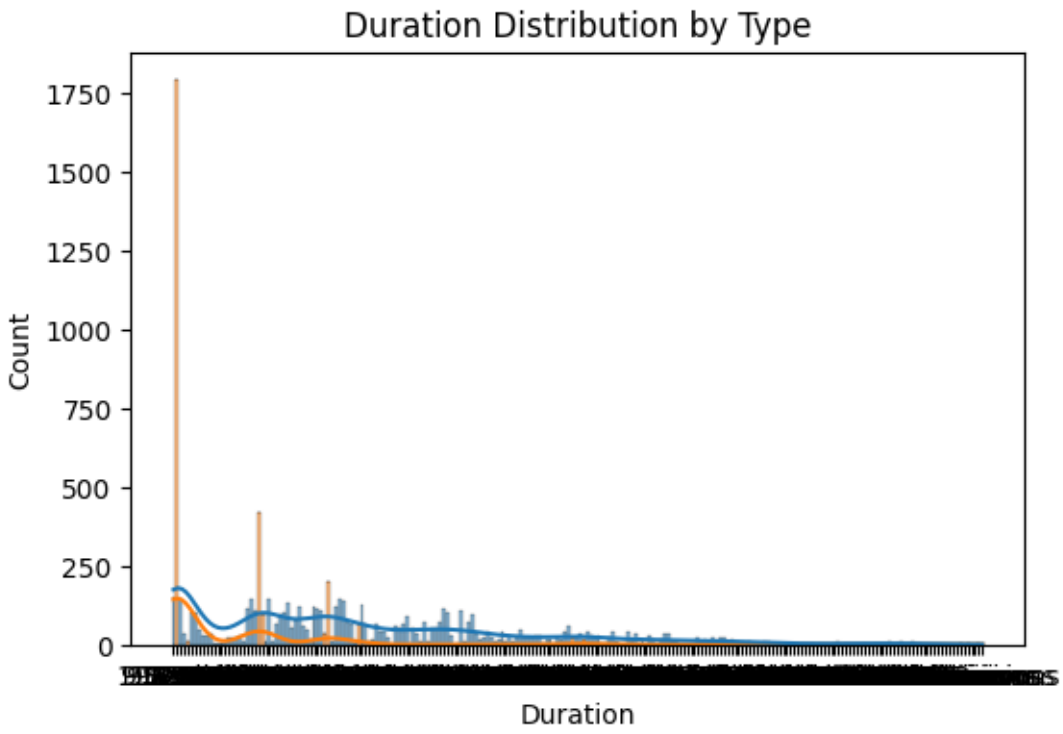
Duration Distribution:

[12]:
```
plt.figure(figsize=(6, 4))
sns.histplot(data=df, x='duration', bins=30, kde=True)
plt.title('Duration Distribution')
plt.xlabel('Duration')
plt.ylabel('Count')
plt.show()
```



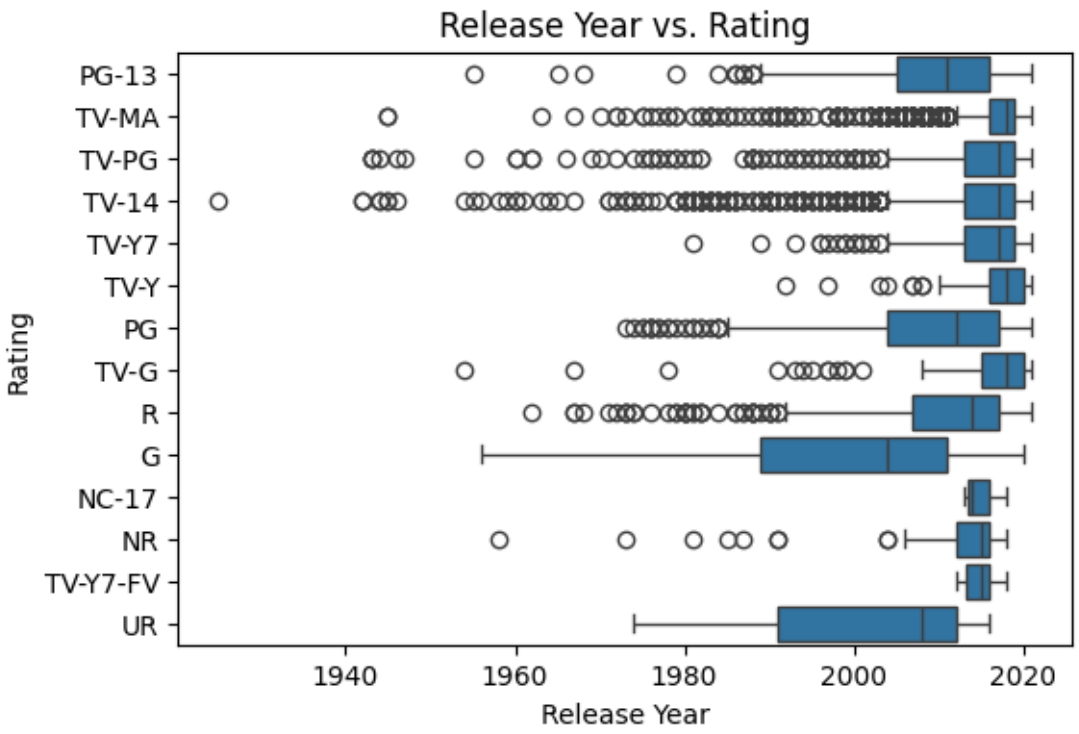
Duration Distribution by Type:

[13]:
```
plt.figure(figsize=(6, 4))
sns.histplot(data=df, x='duration', bins=30, kde=True, hue='type',␣
 ↪multiple='stack', legend=False)
plt.title('Duration Distribution by Type')
plt.xlabel('Duration')
plt.ylabel('Count')
plt.show()
```



Release Year vs. Rating:

[14]:
```
plt.figure(figsize=(6, 4))
sns.boxplot(data=df, x='release_year', y='rating')
plt.title('Release Year vs. Rating')
plt.xlabel('Release Year')
plt.ylabel('Rating')
plt.show()
```



0.0.2 2. Train a simple linear regressing model on a dataset and predict the output:

Import Necessary Libraries:

[15]:
```
import pandas as pd
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error
import matplotlib.pyplot as plt
```

[16]:
```
train_df = pd.read_csv('/content/train dataset - train.csv')
test_df = pd.read_csv('/content/test dataset - test.csv')
```

[17]:
```
print(train_df.head())
```

```
      x          y
0  24.0  21.549452
1  50.0  47.464463
2  15.0  17.218656
3  38.0  36.586398
4  87.0  87.288984
```

[18]:
```
print(test_df.head())
```

```
    x          y
```

```
0  77  79.775152
1  21  23.177279
2  22  25.609262
3  20  17.857388
4  36  41.849864
```

Inspect the dataset:

[19]:
```
print("\033[1mTrain datset:\033[0m")
train_df.dtypes
```

Train datset:

[19]:
```
x    float64
y    float64
dtype: object
```

[20]:
```
print("\033[1mTest datset:\033[0m")
test_df.dtypes
```

Test datset:

[20]:
```
x      int64
y    float64
dtype: object
```

[21]:
```
print(train_df.isnull().sum())
print(test_df.isnull().sum())
```

```
x    0
y    1
dtype: int64
x    0
y    0
dtype: int64
```

[22]:
```
# Fill null values in the dataset with the mean of the respective columns
train_df.fillna(train_df.mean(), inplace=True)
test_df.fillna(test_df.mean(), inplace=True)
```

Split the Train and Test Datasets into Features (X) and Target (y):

[23]:
```
X_train = train_df[['x']]  # Features in the training dataset
y_train = train_df['y']    # Target in the training dataset

X_test = test_df[['x']]   # Features in the test dataset
y_test = test_df['y']     # Target in the test dataset
```

Train the Linear Regression Model:

[24]:
```
# Create a Linear Regression model
model = LinearRegression()

# Train the model on the training data
model.fit(X_train, y_train)
```

[24]:
```
LinearRegression()
```

Make Predictions on the Test Set:

[25]:
```
# Predict the output for the test set
y_pred = model.predict(X_test)
```

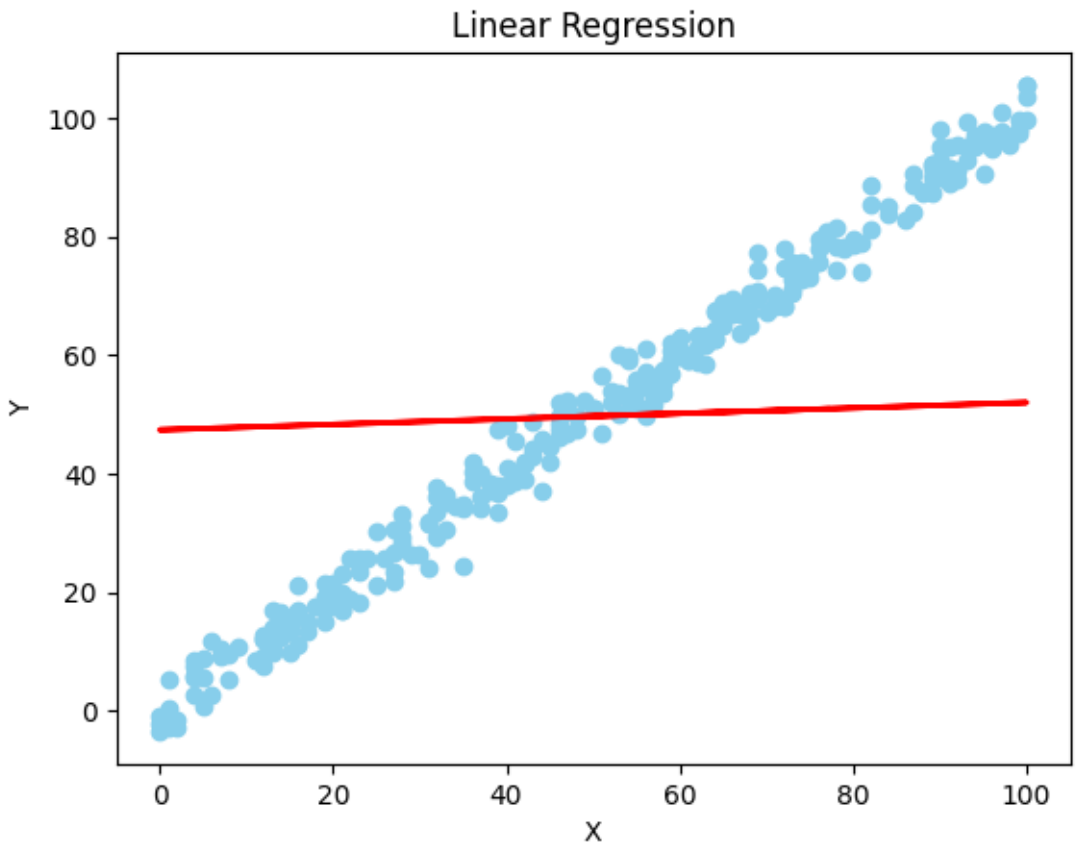
Evaluate the Model:

[26]:
```
# Calculate the Mean Squared Error
mse = mean_squared_error(y_test, y_pred)
print(f"Mean Squared Error: {mse}")
```

```
Mean Squared Error: 770.3012816202481
```

Visualize the Regression Line:

[27]:
```
plt.scatter(X_test, y_test, color='skyblue')  # Scatter plot of the test data
plt.plot(X_test, y_pred, color='red', linewidth=2)  # Regression line
plt.title('Linear Regression')
plt.xlabel('X')
plt.ylabel('Y')
plt.show()
```


Linear Regression
Y
X
0
20
40
60
80
100