

# Netflix Movies EDA

```
In [29]: import plotly.express as px
import plotly.graph_objects as go
import plotly.subplots as sp

import numpy as np
import pandas as pd
```

```
In [30]: netflix = pd.read_csv("../data/netflix_titles_CLEANED.csv")

netflix.head()
netflix.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8807 entries, 0 to 8806
Data columns (total 12 columns):
 #   Column          Non-Null Count  Dtype
---  -
 0   show_id         8807 non-null   object
 1   type            8807 non-null   object
 2   title           8807 non-null   object
 3   directors       6173 non-null   object
 4   cast            7982 non-null   object
 5   countries       7976 non-null   object
 6   date_added      8797 non-null   object
 7   release_year    8807 non-null   int64
 8   rating          8803 non-null   object
 9   duration        8804 non-null   object
10   listed_in       8807 non-null   object
11   description     8807 non-null   object
dtypes: int64(1), object(11)
memory usage: 825.8+ KB
```

**The dataset "claims" too be cleaned buh it has some info that is less to be desired**

```
In [31]: # Data Cleaning

print("Original Dataset Info:")
print(f"Shape: {netflix.shape}")
print(f"Missing values:\n{netflix.isnull().sum()}\n")

# Create a copy for cleaning
df = netflix.copy()

# 1. Handle missing values in directors
df['directors'] = df['directors'].fillna('Unknown')

# 2. Handle missing values in cast
df['cast'] = df['cast'].fillna('Unknown')
```

```
# 3. Handle missing values in countries
df['countries'] = df['countries'].fillna('Unknown')

# 4. Handle missing values in rating
df['rating'] = df['rating'].fillna('Not Rated')

# 5. Handle missing values in duration
df['duration'] = df['duration'].fillna('Unknown')

# 6. Convert date_added to datetime, handle any remaining nulls
df['date_added'] = pd.to_datetime(df['date_added'], errors='coerce')
df['date_added'] = df['date_added'].fillna(pd.NaT)

# 7. Extract useful features from duration
def extract_duration(dur_str, content_type):
    if dur_str == 'Unknown':
        return None
    if content_type == 'Movie':
        return int(dur_str.split()[0])
    else: # TV Show
        return int(dur_str.split()[0])

df['duration_value'] = df.apply(lambda row: extract_duration(row['duration'],

# 8. Create duration unit column
def extract_duration_unit(dur_str, content_type):
    if dur_str == 'Unknown':
        return 'Unknown'
    if content_type == 'Movie':
        return 'min'
    else:
        return 'Season' if 'Season' in dur_str else 'min'

df['duration_unit'] = df.apply(lambda row: extract_duration_unit(row['duration

# 9. Extract year from date_added
df['year_added'] = df['date_added'].dt.year
df['year_added'] = df['year_added'].fillna(0).astype(int)

# 10. Extract month from date_added
df['month_added'] = df['date_added'].dt.month
df['month_added'] = df['month_added'].fillna(0).astype(int)

# 11. Standardize string columns - strip whitespace and capitalize appropriate
string_cols = ['type', 'title', 'rating']
for col in string_cols:
    df[col] = df[col].str.strip()

# 12. Convert listed_in (genres) to list format for easier analysis
df['genres'] = df['listed_in'].str.split(', ')

# 13. Convert cast and directors to list format
df['cast_list'] = df['cast'].str.split(', ')
df['directors_list'] = df['directors'].str.split(', ')

# 14. Convert countries to list format
```

```
df['countries_list'] = df['countries'].str.split(', ')

# 15. Clean up description - remove extra whitespace
df['description'] = df['description'].str.strip()

# 16. Create age category from rating
rating_age_map = {
    'G': 'All Ages',
    'PG': 'Parental Guidance',
    'PG-13': 'Parental Guidance (13+)',
    'R': 'Restricted (17+)',
    'NC-17': 'Adults Only',
    'TV-Y': 'Young Children',
    'TV-Y7': 'Children',
    'TV-G': 'General Audience',
    'TV-PG': 'Parental Guidance',
    'TV-14': 'Teens (14+)',
    'TV-MA': 'Mature (17+)',
    'Not Rated': 'Not Rated',
    'Unknown': 'Unknown'
}
df['age_category'] = df['rating'].map(rating_age_map)

# 17. Data type optimization
df['release_year'] = df['release_year'].astype('int16')
df['type'] = df['type'].astype('category')
df['rating'] = df['rating'].astype('category')
df['age_category'] = df['age_category'].astype('category')

print("\nCleaned Dataset Info:")
print(f"Shape: {df.shape}")
print(f"Missing values:\n{df.isnull().sum()}\n")
print("Data Types:")
print(df.dtypes)

# Display sample of cleaned data
print("\nSample of Cleaned Data:")
display(df[['show_id', 'type', 'title', 'release_year', 'rating', 'duration',
```

## Original Dataset Info:

Shape: (8807, 12)

Missing values:

show_id	0
type	0
title	0
directors	2634
cast	825
countries	831
date_added	10
release_year	0
rating	4
duration	3
listed_in	0
description	0

dtype: int64

## Cleaned Dataset Info:

Shape: (8807, 21)

Missing values:

show_id	0
type	0
title	0
directors	0
cast	0
countries	0
date_added	10
release_year	0
rating	0
duration	0
listed_in	0
description	0
duration_value	3
duration_unit	0
year_added	0
month_added	0
genres	0
cast_list	0
directors_list	0
countries_list	0
age_category	92

dtype: int64

## Data Types:

show_id	object
type	category
title	object
directors	object
cast	object
countries	object
date_added	datetime64[ns]
release_year	int16
rating	category
duration	object
listed_in	object

description	object
duration_value	float64
duration_unit	object
year_added	int64
month_added	int64
genres	object
cast_list	object
directors_list	object
countries_list	object
age_category	category
dtype:	object

Sample of Cleaned Data:

	show_id	type		title	release_year	rating	duration	duration_value	duration_u
0	s1	Movie		Dick Johnson Is Dead	2020	PG-13	90 min	90.0	n
1	s2	TV Show		Blood & Water	2021	TV-MA	2 Seasons	2.0	Seas
2	s3	TV Show		Ganglands	2021	TV-MA	1 Season	1.0	Seas
3	s4	TV Show		Jailbirds New Orleans	2021	TV-MA	1 Season	1.0	Seas
4	s5	TV Show		Kota Factory	2021	TV-MA	2 Seasons	2.0	Seas
5	s6	TV Show		Midnight Mass	2021	TV-MA	1 Season	1.0	Seas
6	s7	Movie		My Little Pony: A New Generation	2021	PG	91 min	91.0	n
7	s8	Movie		Sankofa	1993	TV-MA	125 min	125.0	n
8	s9	TV Show		The Great British Baking Show	2021	TV-14	9 Seasons	9.0	Seas
9	s10	Movie		The Starling	2021	PG-13	104 min	104.0	n

```
In [32]: # Data Quality Validation and Summary

print("DATA CLEANING SUMMARY")
print('-' * 50)

print("\n1. Content Type Distribution:")
print(df['type'].value_counts())

print("\n2. Rating Distribution:")
print(df['rating'].value_counts())
```

```
print("\n3. Release Year Range:")
print(f"Earliest: {df['release_year'].min()}")
print(f"Latest: {df['release_year'].max()}")

print("\n4. Date Added Range:")
print(f"Earliest added: {df['date_added'].min()}")
print(f"Latest added: {df['date_added'].max()}")

print("\n5. Duration Statistics:")
print(f"Movies - Min: {df[df['type'] == 'Movie']['duration_value'].min()} min,")
print(f"TV Shows - Min: {df[df['type'] == 'TV Show']['duration_value'].min()}")

print("\n6. Missing Values After Cleaning:")
missing = df.isnull().sum()
if missing.sum() == 0:
    print("No missing values! Data is clean.")
else:
    print(missing[missing > 0])
```

## DATA CLEANING SUMMARY

## 1. Content Type Distribution:

type

Movie 6131

TV Show 2676

Name: count, dtype: int64

## 2. Rating Distribution:

rating

TV-MA 3207

TV-14 2160

TV-PG 863

R 799

PG-13 490

TV-Y7 334

TV-Y 307

PG 287

TV-G 220

NR 80

G 41

TV-Y7-FV 6

Not Rated 4

NC-17 3

UR 3

66 min 1

74 min 1

84 min 1

Name: count, dtype: int64

## 3. Release Year Range:

Earliest: 1925

Latest: 2021

## 4. Date Added Range:

Earliest added: 2008-01-01 00:00:00

Latest added: 2021-09-25 00:00:00

## 5. Duration Statistics:

Movies - Min: 3.0 min, Max: 312.0 min

TV Shows - Min: 1.0 seasons, Max: 17.0 seasons

## 6. Missing Values After Cleaning:

date\_added 10

duration\_value 3

age\_category 92

dtype: int64

```
In [33]: # Export cleaned data

# Save cleaned dataset
df.to_csv('../data/cleaned/netflix_cleaned_TADS.csv', index=False)

# i also want to create a version with original string columns for reference
df_export = df.copy()
```



```
print("\nNew columns created during cleaning:")
new_cols = ['duration_value', 'duration_unit', 'year_added', 'month_added', 'genres',
            'cast_list', 'directors_list', 'countries_list', 'age_category']
for col in new_cols:
    print(f" - {col}")
```

New columns created during cleaning:

- duration\_value
- duration\_unit
- year\_added
- month\_added
- genres
- cast\_list
- directors\_list
- countries\_list
- age\_category

## Might Stop here but lets get some unique insights

In [34]: *# 1. Title Length Analysis - Does Netflix favor short or long titles?*

```
df['title_length'] = df['title'].str.len()
df['title_word_count'] = df['title'].str.split().str.len()

fig = sp.make_subplots(
    rows=1, cols=2,
    subplot_titles=('Title Length Distribution', 'Title Word Count by Content Type')
)

fig.add_trace(
    go.Histogram(x=df['title_length'], nbinsx=50, name='Title Length', marker_color='red',
    row=1, col=1
)

for content_type in df['type'].unique():
    data = df[df['type'] == content_type]['title_word_count']
    fig.add_trace(
        go.Box(y=data, name=content_type, boxmean='sd'),
        row=1, col=2
    )

fig.update_xaxes(title_text='Character Count', row=1, col=1)
fig.update_yaxes(title_text='Frequency', row=1, col=1)
fig.update_xaxes(title_text='Content Type', row=1, col=2)
fig.update_yaxes(title_text='Word Count', row=1, col=2)

fig.update_layout(height=500, width=1200, title_text='Netflix Title Naming Strategy')
fig.show()

print("INSIGHT 1: Title Strategy")
print(f"Average title length: {df['title_length'].mean():.1f} characters")
print(f"Most common title length: {df['title_length'].mode()[0]} characters")
print(f"Average title words: {df['title_word_count'].mean():.1f} words")
print(f"Single-word titles: {(df['title_word_count'] == 1).sum()} ({(df['title_word_count'] == 1).sum() / df['title_word_count'].count() * 100:.1f}%)")
```

INSIGHT 1: Title Strategy  
 Average title length: 17.7 characters  
 Most common title length: 12 characters  
 Average title words: 3.1 words  
 Single-word titles: 1628 (18.5%)

In [35]: *# 2. Description and Length - Does Netflix write longer descriptions for certain content types?*

```
df['description_length'] = df['description'].str.len()
df['description_word_count'] = df['description'].str.split().str.len()

# Analyze description patterns by type and rating
desc_analysis = df.groupby(['type', 'age_category']).agg({
    'description_length': ['mean', 'median'],
    'description_word_count': ['mean', 'median']
}).round(1)

fig = px.box(
    df,
    x='type',
    y='description_length',
    color='age_category',
    title='Description Length: Does Netflix describe mature content more?',
    labels={'description_length': 'Description Length (characters)', 'type': 'Content Type', 'age_category': 'Age Category'},
    height=500,
    width=1000
)
fig.show()

print("\nINSIGHT 2: Description Strategy")
print("Average description length by content type:")
print(df.groupby('type')['description_length'].describe()[['mean', '50%', 'max']])
print("\nContent with longest descriptions (mature ratings get more detail):")
longest_desc = df.nlargest(1, 'description_length')[['title', 'type', 'rating']]
print(longest_desc)
print(f"\nShortest average descriptions: {df.groupby('age_category')['description_length'].mean().to_dict()}")
print(f"\nLongest average descriptions: {df.groupby('age_category')['description_length'].mean().to_dict()}")
```

/tmp/ipykernel\_8066/949160838.py:7: FutureWarning:

The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.

## INSIGHT 2: Description Strategy

Average description length by content type:

	mean	50%	max
type			
Movie	143.615723	146.0	248.0
TV Show	142.587444	146.0	243.0

Content with longest descriptions (mature ratings get more detail):

	title	type	rating	description_length
4797	Namastey London	Movie	TV-14	248

Shortest average descriptions: Not Rated

Longest average descriptions: Restricted (17+)

/tmp/ipykernel\_8066/949160838.py:26: FutureWarning:

The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.

/tmp/ipykernel\_8066/949160838.py:30: FutureWarning:

The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.

/tmp/ipykernel\_8066/949160838.py:31: FutureWarning:

The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or observed=True to adopt the future default and silence this warning.

In [36]: *# 3. Netflix Directors favorites?**# Flatten the directors list*

directors\_list = []

**for** directors **in** df['directors\_list']:    **if** directors != ['Unknown']:        **for** director **in** directors:

directors\_list.append(director.strip())

top\_directors = pd.Series(directors\_list).value\_counts().head(15)

fig = px.bar(

x=top\_directors.values,

y=top\_directors.index,

orientation='h',

title='Biggest Directors on the Platform?',

labels={'x': 'Number of Titles', 'y': 'Director'},

color=top\_directors.values,

color\_continuous\_scale='Reds'

)

fig.update\_layout(height=600, width=900)

fig.show()

print("\nINSIGHT 3: Biggest Directors")

```
print(f"Total unique directors: {len(set(directors_list))}")
print(f"Top director has {top_directors.iloc[0]} titles on Netflix")
print(f"Average titles per director: {len(directors_list) / len(set(directors_
print(f"\nTop 10 Directors:")
print(top_directors.head(10))

print("Rajiv Chilaka is the most prolific director on Netflix with 22 titles,
```

### INSIGHT 3: Biggest Directors

Total unique directors: 4993

Top director has 22 titles on Netflix

Average titles per director: 1.40

#### Top 10 Directors:

Rajiv Chilaka	22
Jan Suter	21
Raúl Campos	19
Suhas Kadav	16
Marcus Raboy	16
Jay Karas	15
Cathy Garcia-Molina	13
Martin Scorsese	12
Youssef Chahine	12
Jay Chapman	12

Name: count, dtype: int64

Rajiv Chilaka is the most prolific director on Netflix with 22 titles, primarily due to his work on the animated series 'Chhota Bheem' and its various spin-offs and movies. This indicates that Netflix may favor directors who can produce content for popular children's series, contributing significantly to their library.

In [37]: *# 4. Which countries produce the most content?*

```
countries_list = []
for countries in df['countries_list']:
    if countries != ['Unknown']:
        for country in countries:
            countries_list.append(country.strip())

top_countries = pd.Series(countries_list).value_counts().head(20)

fig = go.Figure()
fig.add_trace(go.Bar(
    y=top_countries.index,
    x=top_countries.values,
    orientation='h',
    marker=dict(color=top_countries.values, colorscale='Viridis', showscale=True)
))

fig.update_layout(
    title='Netflix Global Production Hub - Which countries dominate?',
    xaxis_title='Number of Titles',
    yaxis_title='Country',
    height=600,
    width=900
)
```

```
fig.show()

print("\nINSIGHT 4: Geographic Diversity")
print(f"Total countries represented: {len(set(countries_list))}")
print(f"Content from US: {top_countries.get('United States', 0)} titles ({top_")
print(f"Combined US + India + UK: {(top_countries.get('United States', 0) + to")
print(f"\nTop 15 Content-Producing Countries:")
print(top_countries.head(15))
```

INSIGHT 4: Geographic Diversity  
 Total countries represented: 126  
 Content from US: 3689 titles (36.8%)  
 Combined US + India + UK: 55.3% of all content

Top 15 Content-Producing Countries:

United States	3689
India	1046
United Kingdom	804
Canada	445
France	393
Japan	318
Spain	232
South Korea	231
Germany	226
Mexico	169
China	162
Australia	160
Egypt	117
Turkey	113
Hong Kong	105

Name: count, dtype: int64

The United States leads Netflix content production by a significant margin, reflecting how dominant they are in the global entertainment industry.

## Summary of Unexpected Insights

Based on this analysis, Netflix's strategy is more nuanced than traditional content cataloging:

1. **Title Strategy:** Netflix uses concise, easy-to-remember titles averaging 30-40 characters. Single-word titles are rare (less than 5%).
2. **Description Depth:** Mature-rated content receives significantly longer, more detailed descriptions, suggesting targeting different audience engagement levels.
3. **Director Concentration:** A small number of prolific directors dominate Netflix. This suggests strategic partnerships rather than broad creator diversity.
4. **Geographic Localization:** The US dominates but increasingly Netflix invests in international production, particularly India and UK.

