**Netflix!** What started in 1997 as a DVD rental service has since exploded into one of the largest entertainment and media companies.

Given the large number of movies and series available on the platform, it is a perfect opportunity to flex your exploratory data analysis skills and dive into the entertainment industry.

You work for a production company that specializes in nostalgic styles. You want to do some research on movies released in the 1990's. You'll delve into Netflix data and perform exploratory data analysis to better understand this awesome movie decade!

You have been supplied with the dataset netflix\_data.csv, along with the following table detailing the column names and descriptions. Feel free to experiment further after submitting!

## The data

## netflix\_data.csv

Column	Description
show_id	The ID of the show
type	Type of show
title	Title of the show
director	Director of the show
cast	Cast of the show
country	Country of origin
date_added	Date added to Netflix
release_year	Year of Netflix release
duration	Duration of the show in minutes
description	Description of the show
genre	Show genre

Add blockquote

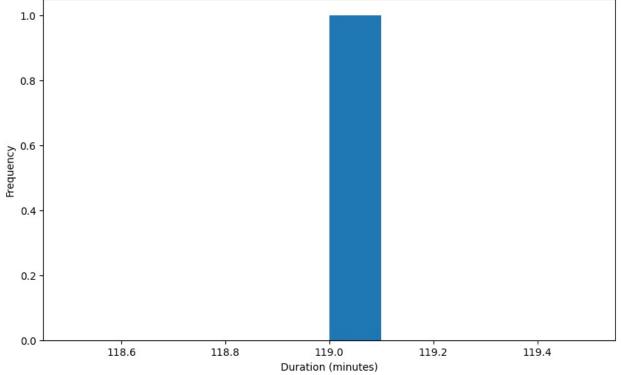


```
# Importing pandas and matplotlib
import pandas as pd
import matplotlib.pyplot as plt
# Read in the Netflix CSV as a DataFrame
netflix df = pd.read csv("netflix.csv")
type(netflix df)
netflix df.head(10)
{"summary":"{\n \"name\": \"netflix df\",\n \"rows\": 10,\n
\"fields\": [\n {\n \"column\": \"show_id\",\n \"properties\": {\n \"dtype\": \"string\",\n
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                                                         \"s3\",\n
\"s10\",\n
                                                                                                                        \"s7\"\n
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                                                                                                        \"description\": \"\"\n
\"dtype\": \"category\",\n \"num_unique_values\": 2,\n
\"samples\": [\n \"TV Show\",\n
                                                                                                                                                       \"Movie\"\
                                                                \"semantic type\": \"\",\n
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                                                                                          \"samples\": [\n
\"num_unique_values\": 10,\n
\"192<del>0</del>\",\n\\"23:59\"\n
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                                                                                                                                                                \"semantic type\":
```

```
\"\",\n \"description\": \"\"\n }\n },\n {\n \"column\": \"director\".\n \"properties\": {\n \"dtype\":
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\"string\",\n \"num_unique_values\": 10,\n
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Rajendranath Zutshi, Vipin Sharma, Amin Hajee, Shri Vallabh
Vyas\\\"\",\n \"\\\"Tedd Chan, Stella Chung, Henley Hii,
Lawrence Koh, Tommy Kuan, Josh Lai, Mark Lee, Susan Leong, Benjamin
\"dtype\": \"string\",\
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\"Mexico\",\n \"Singapore\"\n ],\n
}\
\"\\\"December 20,
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}\n },\n {\n \"column\": \"description\",\n
\"properties\": {\n \"dtype\": \"string\",\n
\"num unique values\": 10,\n \"samples\": [\n
architect and his wife move into a castle that is slated to become a
luxury hotel. But something inside is determined to stop the
renovation.\",\n \"\\\"When an army recruit is found dead,
his fellow soldiers are forced to confront a terrifying secret that's
haunting their jungle island training camp.\\\"\"\n ],\n
\"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"genre\",\n \"properties\": {\
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\"samples\": [\n \"Horror Movies\",\n \"International TV\"\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n ]\
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```

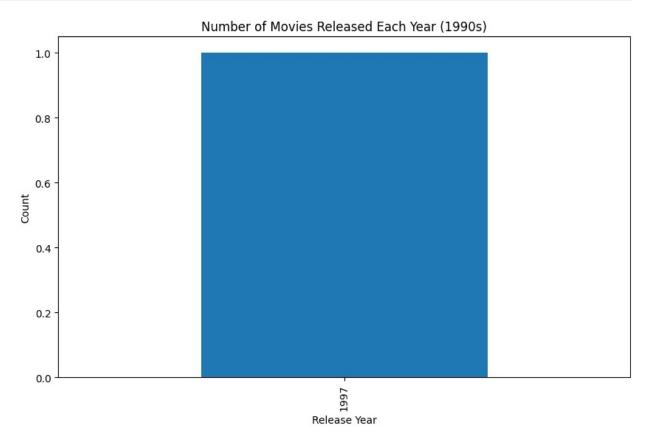
```
df_90s = netflix_df[(netflix_df['type'] == 'Movie') &
(netflix df['release year'] >= 1990) & (netflix df['release year'] <</pre>
2000)]
print(df_90s.describe())
       release year
                      duration
count
                1.0
                           1.0
                         119.0
             1997.0
mean
std
                NaN
                           NaN
             1997.0
                         119.0
min
                         119.0
25%
             1997.0
50%
             1997.0
                         119.0
75%
             1997.0
                         119.0
             1997.0
                         119.0
max
plt.figure(figsize=(10, 6))
plt.hist(df_90s['duration'])
plt.title('Distribution of Movie Durations (1990s)')
plt.xlabel('Duration (minutes)')
plt.ylabel('Frequency')
plt.show()
```

## Distribution of Movie Durations (1990s)



```
plt.figure(figsize=(10, 6))
df_90s['release_year'].value_counts().sort_index().plot(kind='bar')
```

```
plt.title('Number of Movies Released Each Year (1990s)')
plt.xlabel('Release Year')
plt.ylabel('Count')
plt.show()
```



```
f_90s = netflix_df[(netflix_df['type'] == 'Movie') &
  (netflix_df['release_year'] >= 1990) & (netflix_df['release_year'] <
2000)]

# Find the most frequent movie duration
duration = int(df_90s['duration'].mode()[0])

# Print the most frequent movie duration
print(f"The most frequent movie duration in the 1990s is {duration}
minutes.")

The most frequent movie duration in the 1990s is 119 minutes.

df_90s = netflix_df[(netflix_df['type'] == 'Movie') &
  (netflix_df['release_year'] >= 1990) & (netflix_df['release_year'] <
2000)]

# Filter for short action movies
short_action_movies = df_90s[(df_90s['duration'] < 90) &</pre>
```

```
(df_90s['genre'].str.contains('Action', case=False, na=False))]
# Count the number of short action movies
short_movie_count = short_action_movies.shape[0]
# Print the count of short action movies
print(f"The number of short action movies released in the 1990s is
{short_movie_count}.")
The number of short action movies released in the 1990s is 0.
```