## Upload the Dataset

from google.colab import files
uploaded = files.upload()



Choose Files netflix\_titles.csv

• netflix\_titles.csv(text/csv) - 3399671 bytes, last modified: 5/8/2025 - 100% done

Saving netflix titles.csv to netflix titles.csv

### Load the Dataset

import pandas as pd

df = pd.read\_csv('/content/netflix\_titles.csv')
df.head()

<b>→</b> ▼		show_id	type	title	director	cast	country	date_added	release_year	rating	duration	listed_ir
	0	s1	Movie	Dick Johnson Is Dead	Kirsten Johnson	NaN	United States	September 25, 2021	2020	PG-13	90 min	Documentaries
	1	s2	TV Show	Blood & Water	NaN	Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban	South Africa	September 24, 2021	2021	TV-MA	2 Seasons	Internationa TV Shows, TV Dramas, TV Mysteries
	2	s3	TV Show	Ganglands	Julien Leclercq	Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi	NaN	September 24, 2021	2021	TV-MA	1 Season	Crime T\ Shows Internationa TV Shows, T\ Act
	3	s4	TV Show	Jailbirds New Orleans	NaN	NaN	NaN	September 24, 2021	2021	TV-MA	1 Season	Docuseries Reality T∖
	4	s5	TV Show	Kota Factory	NaN	Mayur More, Jitendra Kumar, Ranjan Raj, Alam K	India	September 24, 2021	2021	TV-MA	2 Seasons	Internationa TV Shows Romantic T\ Shows, TV

Next steps: (

Generate code with df

View recommended plots

New interactive sheet

# Data Exploration

df.info()

df.describe(include='all')

<<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 8807 non-null 1 object type 8807 non-null 2 object title 3 director 6173 non-null object 4 cast 7982 non-null object 5 7976 non-null object country 6 date\_added 8797 non-null object 7 release\_year 8807 non-null int64 8 rating object 8803 non-null 9 8804 non-null duration object 10 listed\_in 8807 non-null object

object

11 description 8807 non-null dtypes: int64(1), object(11) memory usage: 825.8+ KB

	show_id	type	title	director	cast	country	date_added	release_year	rating	duration	listed
count	8807	8807	8807	6173	7982	7976	8797	8807.000000	8803	8804	8
unique	8807	2	8807	4528	7692	748	1767	NaN	17	220	
top	s8807	Movie	Zubaan	Rajiv Chilaka	David Attenborough	United States	January 1, 2020	NaN	TV-MA	1 Season	Dran Internatic Mo
freq	1	6131	1	19	19	2818	109	NaN	3207	1793	
mean	NaN	NaN	NaN	NaN	NaN	NaN	NaN	2014.180198	NaN	NaN	1
std	NaN	NaN	NaN	NaN	NaN	NaN	NaN	8.819312	NaN	NaN	١
min	NaN	NaN	NaN	NaN	NaN	NaN	NaN	1925.000000	NaN	NaN	١
25%	NaN	NaN	NaN	NaN	NaN	NaN	NaN	2013.000000	NaN	NaN	١
50%	NaN	NaN	NaN	NaN	NaN	NaN	NaN	2017.000000	NaN	NaN	١
75%	NaN	NaN	NaN	NaN	NaN	NaN	NaN	2019.000000	NaN	NaN	١
max	NaN	NaN	NaN	NaN	NaN	NaN	NaN	2021.000000	NaN	NaN	١

## Check for Missing Values and Duplicates

print(df.isnull().sum())
print("\nDuplicate Rows:", df.duplicated().sum())

0 show\_id 0 type title 0 2634 director 825 cast country 831 date\_added 10 release\_year 4 rating duration 3 0 listed\_in description dtype: int64

Duplicate Rows: 0

#### Visualize a Few Features

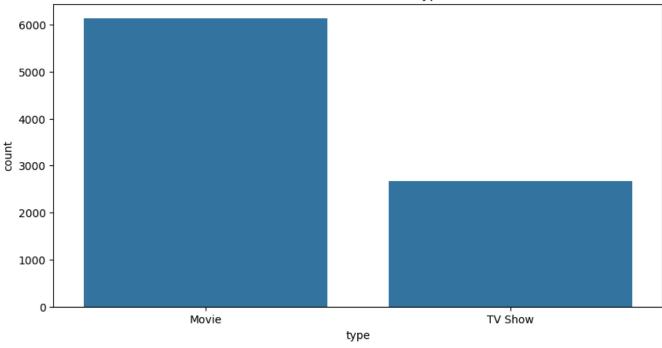
```
import seaborn as sns
import matplotlib.pyplot as plt

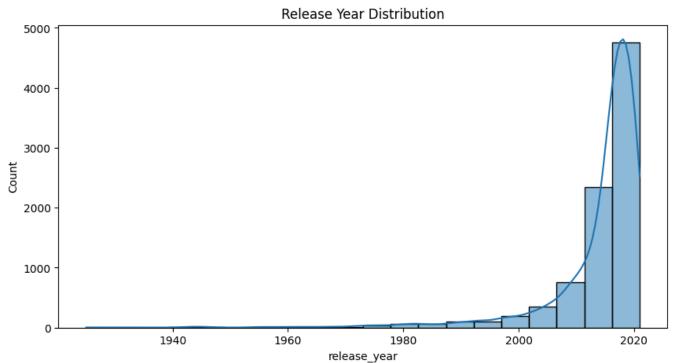
plt.figure(figsize=(10,5))
sns.countplot(data=df, x='type')
plt.title("Distribution of Content Types")
plt.show()

plt.figure(figsize=(10,5))
sns.histplot(df['release_year'], bins=20, kde=True)
plt.title("Release Year Distribution")
plt.show()
```









## **Identify Target and Features**

```
# This dataset doesn't have a traditional 'target' column.
# For example, we could try to predict the 'type' (Movie/TV Show) based on other features.
target = 'type'
features = df.drop(columns=['type', 'show_id', 'title', 'description'])
```

## Convert Categorical Columns to Numerical

```
df['date_added'] = pd.to_datetime(df['date_added'], format='mixed', errors='coerce')
df['year_added'] = df['date_added'].dt.year
df['month_added'] = df['date_added'].dt.month
df = df.drop(columns=['date_added'])
One-Hot Encoding
df_encoded = pd.get_dummies(df, columns=['rating', 'country', 'listed_in'], drop_first=True)
Feature Scaling
print(df_encoded.columns.tolist())
df_encoded.rename(columns={
    'Year Added': 'year_added',
    'Month Added': 'month_added'
}, inplace=True)
missing_cols = [col for col in numerical_cols if col not in df_encoded.columns]
if missing_cols:
    print(f"Missing columns: {missing_cols}")
else:
    df_encoded[numerical_cols] = scaler.fit_transform(df_encoded[numerical_cols])
🚁 ['show_id', 'type', 'title', 'director', 'cast', 'date_added', 'release_year', 'duration', 'description', 'rating_
     Missing columns: ['year_added', 'month_added']
Train-Test Split
from sklearn.model_selection import train_test_split
X = df_encoded.drop(columns=['type'])
y = df['type']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
Model Building
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
# Load your dataset
df = pd.read_csv('/content/netflix_titles.csv') # Adjust path as needed
# Save target separately
target = df['type']
# Drop unnecessary columns
df = df.drop(columns=['show_id', 'title', 'description', 'cast', 'director', 'type'], errors='ignore')
# Convert date_added
df['date_added'] = pd.to_datetime(df['date_added'], format='mixed', errors='coerce')
df['year_added'] = df['date_added'].dt.year
df['month_added'] = df['date_added'].dt.month
df = df.drop(columns=['date_added'])
```

```
# Drop rows with missing values in critical columns
df = df.dropna(subset=['rating', 'country', 'release_year'])

# Encode categorical features
df_encoded = pd.get_dummies(df, drop_first=True)

# Align target with the filtered DataFrame
target = target.loc[df_encoded.index].astype('category').cat.codes

# Train-test split
X_train, X_test, y_train, y_test = train_test_split(df_encoded, target, test_size=0.2, random_state=42)

# Train model
model = RandomForestClassifier()
model.fit(X_train, y_train)

* RandomForestClassifier ③ ?

RandomForestClassifier()
```

### Evaluation

```
from sklearn.metrics import classification_report, accuracy_score

y_pred = model.predict(X_test)
print(classification_report(y_test, y_pred))
print("Accuracy:", accuracy_score(y_test, y_pred))
```

<b>→</b>	precision	recall	f1-score	support
0	0.98	0.99	0.98	1150
1	0.97	0.94	0.96	445
accuracy			0.98	1595
macro avg	0.98	0.97	0.97	1595
weighted avg	0.98	0.98	0.98	1595

Accuracy: 0.9768025078369906

### Make Predictions from New Input

```
sample = X_test.iloc[0:1]
prediction = model.predict(sample)
print("Predicted Type:", prediction[0])
```

→ Predicted Type: 0

## Convert to DataFrame and Encode

```
# Step 1: Fit the scaler on your training data
scaler = StandardScaler()
scaler.fit(df_encoded[numerical_cols]) # df_encoded is your training dataset
# Step 2: Now you can transform new data safely
new_df[numerical_cols] = scaler.transform(new_df[numerical_cols])
import joblib
```

```
joblib.dump(model, 'model.pkl')
joblib.dump(scaler, 'scaler.pkl')
model = joblib.load('model.pkl')
scaler = joblib.load('scaler.pkl')
Predict the Final Grade (assuming 'type')
feature_columns = df_encoded.columns # Save this right after encoding
# Create your new input as a DataFrame
new_data = {
    'country': ['India'],
    'rating': ['TV-MA'],
    'release_year': [2021],
    'duration': ['1 Season'],
    'year_added': [2022],
    'month added': [7],
    'listed_in': ['Dramas, International TV Shows']
new_df = pd.DataFrame(new_data)
# One-hot encode new df using same logic
new_df_encoded = pd.get_dummies(new_df, drop_first=True)
# Align new input with training columns
new_df_encoded = new_df_encoded.reindex(columns=feature_columns, fill_value=0)
final_prediction = model.predict(new_df_encoded)
print("Final Prediction:", "TV Show" if final_prediction[0] == 1 else "Movie")
→ Final Prediction: TV Show
Deployment - Building an Interactive App python Copy Edit
!pip install gradio
import gradio as gr
→ Collecting gradio
       Downloading gradio-5.29.0-py3-none-any.whl.metadata (16 kB)
     Collecting aiofiles<25.0,>=22.0 (from gradio)
       Downloading aiofiles-24.1.0-py3-none-any.whl.metadata (10 kB)
     Requirement already satisfied: anyio<5.0,>=3.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (4.9.0)
     Collecting fastapi<1.0,>=0.115.2 (from gradio)
       Downloading fastapi-0.115.12-py3-none-any.whl.metadata (27 kB)
     Collecting ffmpy (from gradio)
       Downloading ffmpy-0.5.0-py3-none-any.whl.metadata (3.0 kB)
     Collecting gradio-client==1.10.0 (from gradio)
       Downloading gradio_client-1.10.0-py3-none-any.whl.metadata (7.1 kB)
     Collecting groovy~=0.1 (from gradio)
       Downloading groovy-0.1.2-py3-none-any.whl.metadata (6.1 kB)
     Requirement already satisfied: httpx>=0.24.1 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.28.1)
     Requirement already satisfied: huggingface-hub>=0.28.1 in /usr/local/lib/python3.11/dist-packages (from gradio)
     Requirement already satisfied: jinja2<4.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (3.1.6)
     Requirement already satisfied: markupsafe<4.0,>=2.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (3.
     Requirement already satisfied: numpy<3.0,>=1.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (2.0.2)
     Requirement already satisfied: orjson~=3.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (3.10.18)
```

```
Requirement already satisfied: packaging in /usr/local/lib/python3.11/dist-packages (from gradio) (24.2)
Requirement already satisfied: pandas<3.0,>=1.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (2.2.2)
Requirement already satisfied: pillow<12.0,>=8.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (11.2.
Requirement already satisfied: pydantic<2.12,>=2.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (2.1
Collecting pydub (from gradio)
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Collecting python-multipart>=0.0.18 (from gradio)
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Requirement already satisfied: pyyaml<7.0,>=5.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (6.0.2)
Collecting ruff>=0.9.3 (from gradio)
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Collecting safehttpx<0.2.0,>=0.1.6 (from gradio)
 Downloading safehttpx-0.1.6-py3-none-any.whl.metadata (4.2 kB)
Collecting semantic-version~=2.0 (from gradio)
 Downloading semantic_version-2.10.0-py2.py3-none-any.whl.metadata (9.7 kB)
Collecting starlette<1.0,>=0.40.0 (from gradio)
 Downloading starlette-0.46.2-py3-none-any.whl.metadata (6.2 kB)
Collecting tomlkit<0.14.0,>=0.12.0 (from gradio)
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Requirement already satisfied: typing-extensions~=4.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (
Collecting uvicorn>=0.14.0 (from gradio)
 Downloading uvicorn-0.34.2-py3-none-any.whl.metadata (6.5 kB)
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Requirement already satisfied: websockets<16.0,>=10.0 in /usr/local/lib/python3.11/dist-packages (from gradio-cl
Requirement already satisfied: idna>=2.8 in /usr/local/lib/python3.11/dist-packages (from anyio<5.0,>=3.0->gradi
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Requirement already satisfied: h11>=0.16 in /usr/local/lib/python3.11/dist-packages (from httpcore==1.*->httpx>=
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Requirement already satisfied: requests in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.28.1
Requirement already satisfied: tqdm>=4.42.1 in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.
Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.11/dist-packages (from pandas<3.
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.11/dist-packages (from pandas<3.0,>=1.0->g
Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.11/dist-packages (from pandas<3.0,>=1.0-
Requirement already satisfied: annotated-types>=0.6.0 in /usr/local/lib/python3.11/dist-packages (from pydantic
Requirement already satisfied: pydantic-core==2.33.2 in /usr/local/lib/python3.11/dist-packages (from pydantic<2
```

#### Create a Prediction Function

## Create the Gradio Interface

```
import gradio as gr
import pandas as pd

# Save the feature columns after training
feature_columns = df_encoded.columns

# Define the prediction function
def predict_netflix_type(country, rating, release_year, duration, year_added, month_added, listed_in):
    # Create a new DataFrame for input
    new_input = pd.DataFrame({
        'country': [country],
        'rating': [rating],
        'release_year': [release_year],
        'duration': [duration],
        'year_added': [year_added],
        'month added': [month added].
```

```
'listed in': [listed in]
   })
   # Encode the input similar to training
   new_encoded = pd.get_dummies(new_input, drop_first=True)
    # Reindex to match training features
   new_encoded = new_encoded.reindex(columns=feature_columns, fill_value=0)
    # Predict
   prediction = model.predict(new_encoded)[0]
    return "TV Show" if prediction == 1 else "Movie"
# Create the Gradio interface
interface = gr.Interface(
    fn=predict_netflix_type,
    inputs=[
        gr.Textbox(label="Country"),
       gr.Textbox(label="Rating (e.g. TV-MA)"),
       gr.Number(label="Release Year"),
        gr.Textbox(label="Duration (e.g. 1 Season or 90 min)"),
        gr.Number(label="Year Added"),
       gr.Number(label="Month Added"),
       gr.Textbox(label="Genres (Listed In)")
    ],
   outputs="text",
   title="Netflix Content Type Predictor",
   description="Predict whether the content is a Movie or TV Show based on input features."
)
interface.launch()
```

It looks like you are running Gradio on a hosted a Jupyter notebook. For the Gradio app to work, sharing must be e

Colab notebook detected. To show errors in colab notebook, set debug=True in launch()

\* Running on public URL: <a href="https://b4e268096ec7ded9d5.gradio.live">https://b4e268096ec7ded9d5.gradio.live</a>

This share link expires in 1 week. For free permanent hosting and GPU upgrades, run `gradio deploy` from the termi