# untitled6-1

#### August 5, 2024

<ipython-input-37-b1a88c6dadca>:1: ParserWarning: Falling back to the 'python'
engine because the 'c' engine does not support regex separators (separators > 1
char and different from '\s+' are interpreted as regex); you can avoid this
warning by specifying engine='python'.

df = pd.read\_csv("/content/train\_data.txt",sep=":::",names=['NO','MOVIE\_NAME',
'GENRE','DESCRIPTION'])

### []: print(df)

	NO	MOV	IE_NAME	GENRE	\
0	1	Oscar et la dame rose	(2009)	drama	
1	2	Cupid	(1997)	thriller	
2	3	Young, Wild and Wonderful	(1980)	adult	
3	4	The Secret Sin	(1915)	drama	
4	5	The Unrecovered	(2007)	drama	
•••	•••		•••	•••	
54209	54210	"Bonino"	(1953)	comedy	
54210	54211	Dead Girls Don't Cry	(????)	horror	
54211	54212	Ronald Goedemondt: Ze bestaan echt	(2008)	documentary	
54212	54213	Make Your Own Bed	(1944)	comedy	
54213	54214	Nature's Fury: Storm of the Century	(2006)	history	

#### DESCRIPTION

```
0
        Listening in to a conversation between his do...
1
        A brother and sister with a past incestuous r...
2
        As the bus empties the students for their fie...
3
        To help their unemployed father make ends mee...
4
        The film's title refers not only to the un-re...
54209
        This short-lived NBC live sitcom centered on ...
        The NEXT Generation of EXPLOITATION. The sist...
54210
        Ze bestaan echt, is a stand-up comedy about g...
54211
54212
        Walter and Vivian live in the country and hav...
        On Labor Day Weekend, 1935, the most intense ...
54213
```

#### [54214 rows x 4 columns]

<ipython-input-22-083b3032d583>:1: ParserWarning: Falling back to the 'python'
engine because the 'c' engine does not support regex separators (separators > 1
char and different from '\s+' are interpreted as regex); you can avoid this
warning by specifying engine='python'.

test\_data=pd.read\_csv("/content/test\_data.txt",sep=":::",names=['NO','MOVIE\_NA
ME','DESCRIPTION'])

<ipython-input-39-439b5a6c1fa0>:1: ParserWarning: Falling back to the 'python'
engine because the 'c' engine does not support regex separators (separators > 1
char and different from '\s+' are interpreted as regex); you can avoid this
warning by specifying engine='python'.

test\_data\_solution=pd.read\_csv("/content/test\_data\_solution.txt",sep=":::",nam
es=['NO','MOVIE\_NAME','GENRE','DESCRIPTION'])

#### []: print(test\_data)

	NO	MOM	/IE_NAME \
0	1	Edgar's Lunch	(1998)
1	2	La guerra de papá	(1977)
2	3	Off the Beaten Track	(2010)
3	4	Meu Amigo Hindu	(2015)
4	5	Er nu zhai	(1955)
•••	•••		•••
541	95 54196	"Tales of Light & Dark"	(2013)
541	96 54197	Der letzte Mohikaner	(1965)
541	97 54198	Oliver Twink	(2007)

54198	54199 Slipstream (1973)
54199	54200 Curitiba Zero Grau (2010)
	DESCRIPTION
0	L.R. Brane loves his life - his car, his apar
1	Spain, March 1964: Quico is a very naughty ch
2	One year in the life of Albin and his family
3	His father has died, he hasn't spoken with hi
4	Before he was known internationally as a mart
•••	
54195	Covering multiple genres, Tales of Light & Da
54196	As Alice and Cora Munro attempt to find their
54197	A movie 169 years in the making. Oliver Twist
54198	Popular, but mysterious rock D.J Mike Mallard
54199	Curitiba is a city in movement, with rhythms

## [54200 rows x 3 columns]

# []: print(test\_data\_solution)

	NO	/OM	/IE_NAME	GENRE	\
0	1	Edgar's Lunch	(1998)	thriller	
1	2	La guerra de papá	(1977)	comedy	
2	3	Off the Beaten Track	(2010)	documentary	
3	4	Meu Amigo Hindu	(2015)	drama	
4	5	Er nu zhai	(1955)	drama	
•••			•••	•••	
40123	40124	'Doc'	(1971)	western	
40124	40125	"AN.X.O"	(2015)	action	
40125	40126	Bachke Rehna Re Baba	(2005)	comedy	
40126	40127	Sukeban gerira	(1972)	action	
40127	40128	Na dne	(2014)	drama	

## DESCRIPTION

- 0 L.R. Brane loves his life his car, his apar...
- 1 Spain, March 1964: Quico is a very naughty ch...
- 2 One year in the life of Albin and his family ...
- 3 His father has died, he hasn't spoken with hi...
- 4 Before he was known internationally as a mart...

40123 One night of 1881, Doc Holliday, a famous pok...

- 40124 AN.X.O is a web series with 8 minute episodes...
- 40125 Rukmini is well into her early 40s and is sti...
- 40126 Miko Sugimoto is the leader of the Red Helmet...
- 40127 More than a century ago a play about "former ...

### [40128 rows x 4 columns]

```
[]:
    TRAINING AND TESTING OF DATA
[]: df['DESCRIPTION'].fillna("",inplace=True)
[]: print(df['DESCRIPTION'])
    0
              Listening in to a conversation between his do...
    1
              A brother and sister with a past incestuous r...
    2
              As the bus empties the students for their fie...
    3
              To help their unemployed father make ends mee...
              The film's title refers not only to the un-re...
              This short-lived NBC live sitcom centered on ...
    54209
    54210
              The NEXT Generation of EXPLOITATION. The sist...
    54211
              Ze bestaan echt, is a stand-up comedy about g...
    54212
              Walter and Vivian live in the country and hav ...
              On Labor Day Weekend, 1935, the most intense ...
    54213
    Name: DESCRIPTION, Length: 54214, dtype: object
[]: test_data['DESCRIPTION'].fillna("",inplace=True)
[]: print(test_data["DESCRIPTION"])
    0
              L.R. Brane loves his life - his car, his apar...
              Spain, March 1964: Quico is a very naughty ch...
    1
              One year in the life of Albin and his family ...
    3
              His father has died, he hasn't spoken with hi...
    4
              Before he was known internationally as a mart...
    54195
              Covering multiple genres, Tales of Light & Da...
    54196
              As Alice and Cora Munro attempt to find their...
    54197
              A movie 169 years in the making. Oliver Twist...
    54198
              Popular, but mysterious rock D.J Mike Mallard...
    54199
              Curitiba is a city in movement, with rhythms ...
    Name: DESCRIPTION, Length: 54200, dtype: object
[]: vectorizer=TfidfVectorizer(stop_words='english',max_features=1000)
[]: X_train=vectorizer.fit_transform(df["DESCRIPTION"])
     X_test=vectorizer.fit_transform(test_data["DESCRIPTION"])
[]: le=LabelEncoder()
     Y train=le.fit transform(df["GENRE"])
     Y_test=le.transform(test_data_solution["GENRE"])
```

```
[]: x_train,x_test,y_train,y_test=train_test_split(X_train,Y_train,test_size=0.

→2,random_state=42)
[]: v=LinearSVC()
     v.fit(x_train,y_train)
     prediction=v.predict(x_test)
     print("accuracy:",accuracy_score(y_test,prediction))
     print("classification report:",classification_report(y_test,prediction))
    /usr/local/lib/python3.10/dist-packages/sklearn/svm/_classes.py:32:
    FutureWarning: The default value of `dual` will change from `True` to `'auto'`
    in 1.5. Set the value of `dual` explicitly to suppress the warning.
      warnings.warn(
    accuracy: 0.5391496818223739
    classification report:
                                           precision
                                                         recall f1-score
                                                                             support
                0
                        0.35
                                   0.18
                                             0.24
                                                         263
                1
                        0.42
                                   0.27
                                             0.33
                                                         112
                2
                        0.27
                                   0.09
                                             0.13
                                                         139
                3
                        0.25
                                   0.11
                                             0.15
                                                         104
                4
                        0.00
                                   0.00
                                             0.00
                                                          61
                5
                        0.46
                                   0.47
                                             0.46
                                                        1443
                6
                        0.29
                                   0.06
                                             0.09
                                                         107
                7
                        0.65
                                   0.82
                                             0.72
                                                        2659
                8
                        0.53
                                   0.73
                                             0.61
                                                        2697
                9
                        0.42
                                   0.12
                                             0.19
                                                         150
               10
                        0.10
                                   0.01
                                             0.02
                                                          74
               11
                        0.56
                                   0.50
                                             0.53
                                                          40
               12
                        0.00
                                   0.00
                                             0.00
                                                          45
               13
                        0.49
                                   0.55
                                             0.52
                                                         431
               14
                        0.50
                                   0.45
                                             0.47
                                                         144
               15
                        0.40
                                   0.08
                                                          50
                                             0.13
               16
                        0.15
                                   0.04
                                             0.06
                                                          56
               17
                        0.43
                                   0.09
                                             0.15
                                                          34
               18
                        0.46
                                   0.23
                                             0.31
                                                         192
               19
                        0.22
                                   0.01
                                             0.03
                                                         151
               20
                        0.36
                                   0.22
                                             0.28
                                                         143
               21
                        0.43
                                   0.24
                                             0.31
                                                        1045
               22
                        0.40
                                   0.27
                                             0.32
                                                          93
               23
                                   0.20
                                             0.26
                        0.39
                                                          81
                                                         309
               24
                        0.29
                                   0.07
                                             0.11
               25
                        0.33
                                   0.10
                                             0.15
                                                          20
               26
                        0.68
                                   0.73
                                             0.70
                                                         200
                                             0.54
                                                       10843
        accuracy
                        0.36
                                   0.25
                                             0.27
                                                       10843
       macro avg
```

0.50

10843

0.54

0.50

weighted avg

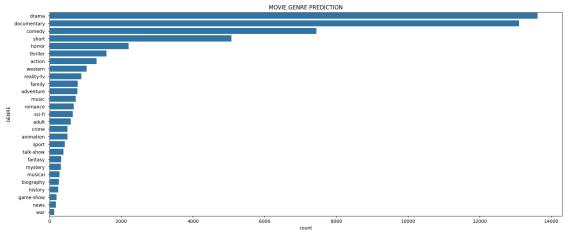
```
[]: naive=MultinomialNB()
    naive.fit(X_train,Y_train)

[]: MultinomialNB()

[]: naive.predict(X_test)

[]: array([8, 8, 7, ..., 8, 8, 7])

[]: plt.figure(figsize=(20,8))
    sns.countplot(y=df["GENRE"],order=df["GENRE"].value_counts().index)
    plt.title("MOVIE GENRE PREDICTION")
    plt.xlabel=("x-axis")
    plt.ylabel=("y-axis")
```



comedy