Extract and Load the Dataset

```
import zipfile
import os
\mbox{\tt\#} Define path and extract the zip file
zip_path = '/content/sample_data/archive (7).zip'
extract_dir = '/content/sample_data/genre_dataset'
with zipfile.ZipFile(zip_path, 'r') as zip_ref:
    zip_ref.extractall(extract_dir)
# View extracted files
os.listdir(extract_dir)
→ ['Genre Classification Dataset']
Double-click (or enter) to edit
import pandas as pd
# Load and parse train_data.txt
train_path = os.path.join(extract_dir, 'Genre Classification Dataset/train_data.txt')
ids, titles, genres, descriptions = [], [], []
with open(train_path, 'r', encoding='utf-8') as f:
    for line in f:
         parts = line.strip().split(" ::: ")
         if len(parts) == 4:
             ids.append(parts[0])
             titles.append(parts[1])
             genres.append(parts[2].lower().split(', ')) # handles multiple genres
             descriptions.append(parts[3])
# Create DataFrame
df = pd.DataFrame({
    'id': ids,
     'title': titles,
     'genres': genres,
     'description': descriptions
df.head()
```





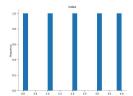
index	id	title	genres	description
0	1	Oscar et la dame rose (2009)	drama	Listening in to a conversation between his doctor and parents, 10-year-old Oscar learns what nobody has the courage to tell him. He only has a few weeks to live. Furious, he refuses to speak to anyone except straight-talking Rose, the lady in pink he meets on the hospital stairs. As Christmas approaches, Rose uses her fantastical experiences as a professional wrestler, her imagination, wit and charm to allow Oscar to live life and love to the full, in the company of his friends Pop Corn, Einstein, Bacon and childhood sweetheart Peggy Blue.
1	2	Cupid (1997)	thriller	A brother and sister with a past incestuous relationship have a current murderous relationship. He murders the women who reject him and she murders the women who get too close to him.
2	3	Young, Wild and Wonderful (1980)	adult	As the bus empties the students for their field trip to the Museum of Natural History, little does the tour guide suspect that the students are there for more than just another tour. First, during the lecture films, the coeds drift into dreams of the most erotic fantasies one can imagine. After the films, they release the emotion of the fantasies in the most erotic and uncommon ways. One slips off to the curator's office for a little "acquisition." Another finds the anthropologist to see what bones can be identified. Even the head teacher isn't immune. Soon the tour is over, but as the bus departs, everyone admits it was quite an education.
3	4	The Secret Sin (1915)	drama	To help their unemployed father make ends meet, Edith and her twin sister Grace work as seamstresses . An invalid, Grace falls prey to the temptations of Chinatown opium and becomes an addict, a condition worsened by a misguided physician who prescribes morphine to ease her pain. When their father strikes oil, the family enjoys a new prosperity and the sisters meet the eligible Jack Herron, a fellow oil prospector. To Grace's shock, Jack falls in love with Edith and in her jealousy, Grace tells Jack that Edith, not she, has a drug problem. Hinting that her sister will soon need more morphine, Grace arranges for a dinner in Chinatown with the couple. While her sister and Jack dance, Grace slips away to an opium den. Edith follows her, but ends up in the wrong den and is arrested in an ensuing drug raid. After he bails her out of jail, Edith takes an angry Jack to search for Grace and stumbles across her half-conscious body lying in the street. The truth about the sisters is revealed, and after sending Grace to a sanitarium in the country, Jack and Edith are married.
4	5	The Unrecovered (2007)	drama	The film's title refers not only to the un-recovered bodies at ground zero, but also to the state of the nation at large. Set in the hallucinatory period of time between September 11 and Halloween of 2001, The Unrecovered examines the effect of terror on the average mind, the way a state of heightened anxiety and/or alertness can cause the average person to make the sort of imaginative connections that are normally made only by artists and conspiracy theorists-both of whom figure prominently in this film. The Unrecovered explores the way in which irony, empathy, and paranoia relate to one another in the wake of 9/11.

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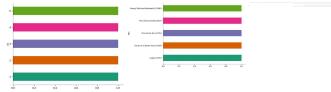


Like what you see? Visit the $\underline{\text{data table notebook}}$ to learn more about interactive tables. WARNING:root:Quickchart encountered unexpected dtypes in columns: "(['genres'],)"

Distributions



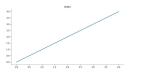
Categorical distributions



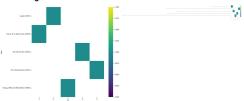
Time series



Values



2-d categorical distributions



Faceted distributions

<string>:5: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `le

<string>:5: FutureWarning:

Next steps: Generate code with df View recommended plots New interactive sheet

Text Cleaning

```
import re
import nltk
from nltk.corpus import stopwords
from nltk.stem import SnowballStemmer
nltk.download('stopwords')
stop_words = set(stopwords.words('english'))
stemmer = SnowballStemmer("english")
def clean_text(text):
    text = text.lower()
    text = re.sub(r'\\\\\\\\', '', text) # remove punctuation
    text = re.sub(r'\\\\\\\\', '', text) # remove digits
    words = text.split()
    words = [stemmer.stem(word) for word in words if word not in stop_words]
    return ' '.join(words)

df['clean_description'] = df['description'].apply(clean_text)
df[['description', 'clean_description']].head()
```

1 to 5 of 5 entries	Filter

Listening in to a conversation between his doctor and parents, 10-year-old Oscar learns listen convers doctor parent year old oscar learn nobodi courag tell what nobody has the courage to tell him. He only has a few weeks to live. Furious, he week live furious refus speak anyon except straight talk rose ladi refuses to speak to anyone except straight-talking Rose, the lady in pink he meets on the pink meet hospit stair christma approach rose use fantast experi 0 hospital stairs. As Christmas approaches, Rose uses her fantastical experiences as a profession wrestler imagin wit charm allow oscar live life love full professional wrestler, her imagination, wit and charm to allow Oscar to live life and love to compani friend pop corn einstein bacon childhood sweetheart the full, in the company of his friends Pop Corn, Einstein, Bacon and childhood sweetheart peggi blue A brother and sister with a past incestuous relationship have a current murderous

relationship. He murders the women who reject him and she murders the women who get too close to him.

description

brother sister past incestu relationship current murder relationship murder women reject murder women get close

clean description

As the bus empties the students for their field trip to the Museum of Natural History, little does the tour guide suspect that the students are there for more than just another tour. First, during the lecture films, the coeds drift into dreams of the most erotic fantasies one can imagine. After the films, they release the emotion of the fantasies in the most erotic and uncommon ways. One slips off to the curator's office for a little "acquisition." Another finds the anthropologist to see what bones can be identified. Even the head teacher isn't immune. Soon the tour is over, but as the bus departs, everyone admits it was quite an education.

bus empti student field trip museum natur histori littl tour guid suspect student anoth tour first lectur film co drift dream erot fantasi one imagin film releas emot fantasi erot uncommon way one slip curat offic littl acquisit anoth find anthropologist see bone identifi even head teacher immun soon tour bus depart everyon admit quit educ

To help their unemployed father make ends meet, Edith and her twin sister Grace work as seamstresses . An invalid, Grace falls prey to the temptations of Chinatown opium and becomes an addict, a condition worsened by a misguided physician who prescribes morphine to ease her pain. When their father strikes oil, the family enjoys a new prosperity and the sisters meet the eligible Jack Herron, a fellow oil prospector. To Grace's shock, Jack falls in love with Edith and in her jealousy, Grace tells Jack that Edith, not she, has a drug

help unemploy father make end meet edith twin sister grace work seamstress invalid grace fall prey temptat chinatown opium becom addict condit worsen misguid physician prescrib morphin eas pain father strike oil famili enjoy new prosper sister meet elig jack herron fellow oil prospector grace shock jack fall love edith jealousi grace tell jack edith drug problem hint sister soon need morphin grace arrang dinner chinatown coupl sister jack danc grace slip away opium den edith follow end wrong den arrest ensu drug raid bail jail edith take angri jack search grace stumbl across half conscious bodi lie street truth sister reveal send grace sanitarium countri jack edith marri

problem. Hinting that her sister will soon need more morphine, Grace arranges for a dinner in Chinatown with the couple. While her sister and Jack dance, Grace slips away to an opium den. Edith follows her, but ends up in the wrong den and is arrested in an ensuing drug raid. After he bails her out of jail, Edith takes an angry Jack to search for Grace and stumbles across her half-conscious body lying in the street. The truth about the sisters is revealed, and after sending Grace to a sanitarium in the country, Jack and Edith are married

> film titl refer un recov bodi ground zero also state nation larg set hallucinatori period time septemb halloween unrecov examin effect terror averag mind way state heighten anxieti alert caus averag person make sort imagin connect normal made artist conspiraci theorist figur promin film unrecov explor way ironi empathi paranoia relat one anoth wake

The film's title refers not only to the un-recovered bodies at ground zero, but also to the state of the nation at large. Set in the hallucinatory period of time between September 11 and Halloween of 2001, The Unrecovered examines the effect of terror on the average mind, the

way a state of heightened anxiety and/or alertness can cause the average person to make the sort of imaginative connections that are normally made only by artists and conspiracy theorists-both of whom figure prominently in this film. The Unrecovered explores the way in which irony, empathy, and paranoia relate to one another in the wake of 9/11.

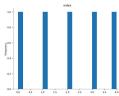
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index

Like what you see? Visit the data table notebook to learn more about interactive tables.

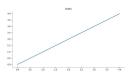
Distributions



Categorical distributions

Time series

Values



2-d categorical distributions

```
from \ sklearn.feature\_extraction.text \ import \ TfidfVectorizer
vectorizer = TfidfVectorizer(max features=5000)
X = vectorizer.fit_transform(df['clean_description'])
Encode Genre Labels with MultiLabelBinarizer
from sklearn.preprocessing import MultiLabelBinarizer
mlb = MultiLabelBinarizer()
y = mlb.fit_transform(df['genres'])
print("Genres:", mlb.classes_)
print("Shape of X:", X.shape)
print("Shape of y:", y.shape)
Genres: ['action' 'adult' 'adventure' 'animation' 'biography' 'comedy' 'crime' 'documentary' 'drama' 'family' 'fantasy' 'game-show' 'history' 'horror'
       'music' 'musical' 'mystery' 'news' 'reality-tv' 'romance' 'sci-fi' 'short' 'sport' 'talk-show' 'thriller' 'war' 'western']
     Shape of X: (54214, 5000)
     Shape of y: (54214, 27)
Train-Test Split
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
Train Classifiers with OneVsRest
from sklearn.linear_model import LogisticRegression
from sklearn.multiclass import OneVsRestClassifier
logreg = OneVsRestClassifier(LogisticRegression(max_iter=1000))
logreg.fit(X_train, y_train)
\rightarrow
             OneVsRestClassifier
                  estimator:
             LogisticRegression
          ▶ LogisticRegression ②
Evaluation
from sklearn.metrics import classification_report, accuracy_score, hamming_loss
y_pred = logreg.predict(X_test)
print("Classification Report:\n", classification_report(y_test, y_pred, target_names=mlb.classes_))
print("Hamming Loss:", hamming_loss(y_test, y_pred))
→ Classification Report:
                     precision
                                   recall f1-score
                                                        support
            action
                          0.83
                                     0.08
                                                0.14
                                                            263
             adult
                          0.73
                                     0.07
                                                0.13
                                                            112
        adventure
                                     0.05
                                                0.09
                          0.44
        animation
                          0.00
                                     0.00
                                                0.00
                                                            104
                          0.00
                                     0.00
                                                0.00
        biography
                                                             61
            comedy
                          0.76
                                     0.30
                                                0.43
                                                           1443
                          0.67
                                     0.02
                                                0.04
                                                            107
            crime
      documentary
                          0.81
                                     0.68
                                                0.74
                                                           2659
             drama
                          0.69
                                     9.46
                                                0.55
                                                           2697
            family
                          1.00
                                     0.03
                                                0.06
                                                            150
           fantasy
                          0.00
                                     0.00
                                                0.00
                                                             74
        game-show
                          1.00
                                     0.23
                                                0.37
                                                             40
                                                             45
           history
                          0.00
                                     0.00
                                                0.00
           horror
                          0.82
                                     0.34
                                                0.48
            music
                          0.64
                                     0.24
                                                0.35
                                                            144
                          0.00
                                     0.00
                                                0.00
           musical
                                                             50
                          0.00
                                     0.00
                                                0.00
                                                             56
          mystery
             news
                          0.00
                                     0.00
                                                0.00
                                                             34
       reality-tv
                          0.75
                                     0.03
                                                0.06
                                                            192
           romance
                          0.00
                                     0.00
                                                0.00
                                                            151
            sci-fi
                          0.67
                                     0.06
                                                0.10
                                                            143
                          0.73
                                     0.11
                                                0.19
                                                           1045
             short
```

sport

war western

talk-show

thriller

0.69

0.75

0.54

0.00

0.98

0.10

0.04

0.04

0.00

0.56

0.17

0.07

0.08

0.00

0.71

93

81

309

20

200

```
micro avg
                        0.76
                                   0.37
                                             0.49
                                                      10843
        macro avg
                        0.50
                                   0.13
                                             0.18
                                                      10843
     weighted avg
                        0.71
                                   0.37
                                             0.45
                                                       10843
                                   0.37
                                             0.36
                                                       10843
      samples avg
                        0.36
     Hamming Loss: 0.027787171105440957
     /usr/local/lib/python3.11/dist-packages/sklearn/metrics/ classification.py:1565: UndefinedMetricWarning: Precision is ill-defined ar
       _warn_prf(average, modifier, f"{metric.capitalize()} is", len(result))
     /usr/local/lib/python3.11/dist-packages/sklearn/metrics/_classification.py:1565: UndefinedMetricWarning: Precision is ill-defined ar
       _warn_prf(average, modifier, f"{metric.capitalize()} is", len(result))
from sklearn.ensemble import RandomForestClassifier
rf = OneVsRestClassifier(RandomForestClassifier(n_estimators=100))
rf.fit(X_train, y_train)
print("RF Accuracy:", rf.score(X_test, y_test))
F Accuracy: 0.20308032832241998
Feature Importance (for Logistic Regression)
import numpy as np
# Feature importance for each genre
for i, genre in enumerate(mlb.classes_):
    top10 = np.argsort(logreg.estimators [i].coef [0])[-10:]
    print(f"\nTop features for genre '{genre}':")
    print([vectorizer.get_feature_names_out()[j] for j in top10])
\overline{2}
     Top features for genre 'action': ['skill', 'hero', 'kill', 'reveng', 'gangster', 'singh', 'fight', 'martial', 'assassin', 'action']
     Top features for genre 'adult':
     ['sexi', 'fantasi', 'seduc', 'orgi', 'bound', 'scene', 'hot', 'sexual', 'gag', 'sex']
     Top features for genre 'adventure':
     ['tie', 'ship', 'bound', 'kidnap', 'rescu', 'buxom', 'jungl', 'gag', 'bondag', 'adventur']
```

B