

# DataSet Kaggle con i Modelli

May 17, 2024

## 1 DataSet Kaggle

### 1.1 Caricamento e Visualizzazione dei Dati dei Titoli Netflix

```
[1]: # Importa le librerie, carica il file csv nel dataframe e lo stampa
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
percorso_file_csv = "C:\\Users\\zetam\\Desktop\\2_
↳Superiore\\Robotica\\netflix_titles.csv"
df = pd.read_csv(percorso_file_csv)
print(df)
```

	show_id	type	title	director \
0	s1	Movie	Dick Johnson Is Dead	Kirsten Johnson
1	s2	TV Show	Blood & Water	NaN
2	s3	TV Show	Ganglands	Julien Leclercq
3	s4	TV Show	Jailbirds New Orleans	NaN
4	s5	TV Show	Kota Factory	NaN
...	...	...	...	...
8802	s8803	Movie	Zodiac	David Fincher
8803	s8804	TV Show	Zombie Dumb	NaN
8804	s8805	Movie	Zombieland	Ruben Fleischer
8805	s8806	Movie	Zoom	Peter Hewitt
8806	s8807	Movie	Zubaan	Mozez Singh

	cast	country \
0	NaN	United States
1	Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban...	South Africa
2	Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi...	NaN
3	NaN	NaN
4	Mayur More, Jitendra Kumar, Ranjan Raj, Alam K...	India
...	...	...
8802	Mark Ruffalo, Jake Gyllenhaal, Robert Downey J...	United States
8803	NaN	NaN
8804	Jesse Eisenberg, Woody Harrelson, Emma Stone, ...	United States
8805	Tim Allen, Courteney Cox, Chevy Chase, Kate Ma...	United States

8806 Vicky Kaushal, Sarah-Jane Dias, Raaghav Chanan... India

	date_added	release_year	rating	duration	\
0	September 25, 2021	2020	PG-13	90 min	
1	September 24, 2021	2021	TV-MA	2 Seasons	
2	September 24, 2021	2021	TV-MA	1 Season	
3	September 24, 2021	2021	TV-MA	1 Season	
4	September 24, 2021	2021	TV-MA	2 Seasons	
...	...	...	...	...	
8802	November 20, 2019	2007	R	158 min	
8803	July 1, 2019	2018	TV-Y7	2 Seasons	
8804	November 1, 2019	2009	R	88 min	
8805	January 11, 2020	2006	PG	88 min	
8806	March 2, 2019	2015	TV-14	111 min	

	listed_in	\
0	Documentaries	
1	International TV Shows, TV Dramas, TV Mysteries	
2	Crime TV Shows, International TV Shows, TV Act...	
3	Docuseries, Reality TV	
4	International TV Shows, Romantic TV Shows, TV ...	
...	...	
8802	Cult Movies, Dramas, Thrillers	
8803	Kids' TV, Korean TV Shows, TV Comedies	
8804	Comedies, Horror Movies	
8805	Children & Family Movies, Comedies	
8806	Dramas, International Movies, Music & Musicals	

	description
0	As her father nears the end of his life, filmm...
1	After crossing paths at a party, a Cape Town t...
2	To protect his family from a powerful drug lor...
3	Feuds, flirtations and toilet talk go down amo...
4	In a city of coaching centers known to train I...
...	...
8802	A political cartoonist, a crime reporter and a...
8803	While living alone in a spooky town, a young g...
8804	Looking to survive in a world taken over by zo...
8805	Dragged from civilian life, a former superhero...
8806	A scrappy but poor boy worms his way into a ty...

[8807 rows x 12 columns]

## 1.2 Identificazione del Tipo di Programma più Frequente nei Titoli Netflix

```
[12]: # Conta quante volte compare ogni tipo di programma e stampa quello con il
      ↪ numero maggiore
      # utilizzando il metodo idxmax
      import pandas as pd
      percorso_file_csv = "C:\\Users\\zetam\\Desktop\\2_
      ↪ Superiore\\Robotica\\netflix_titles.csv"
      df = pd.read_csv(percorso_file_csv)
      tipo_programma = df['type'].value_counts().idxmax()
      print(tipo_programma)
```

Movie

## 1.3 Conteggio dei Programmi Netflix per Anno di Rilascio

```
[13]: # Conta quanti programmi ci sono per ogni anno e stampa i numeri
      import pandas as pd

      anno_programma = df['release_year'].value_counts()
      print(anno_programma)
```

```
release_year
2018      1147
2017      1032
2019      1030
2020       953
2016       902
...
1959         1
1925         1
1961         1
1947         1
1966         1
Name: count, Length: 74, dtype: int64
```

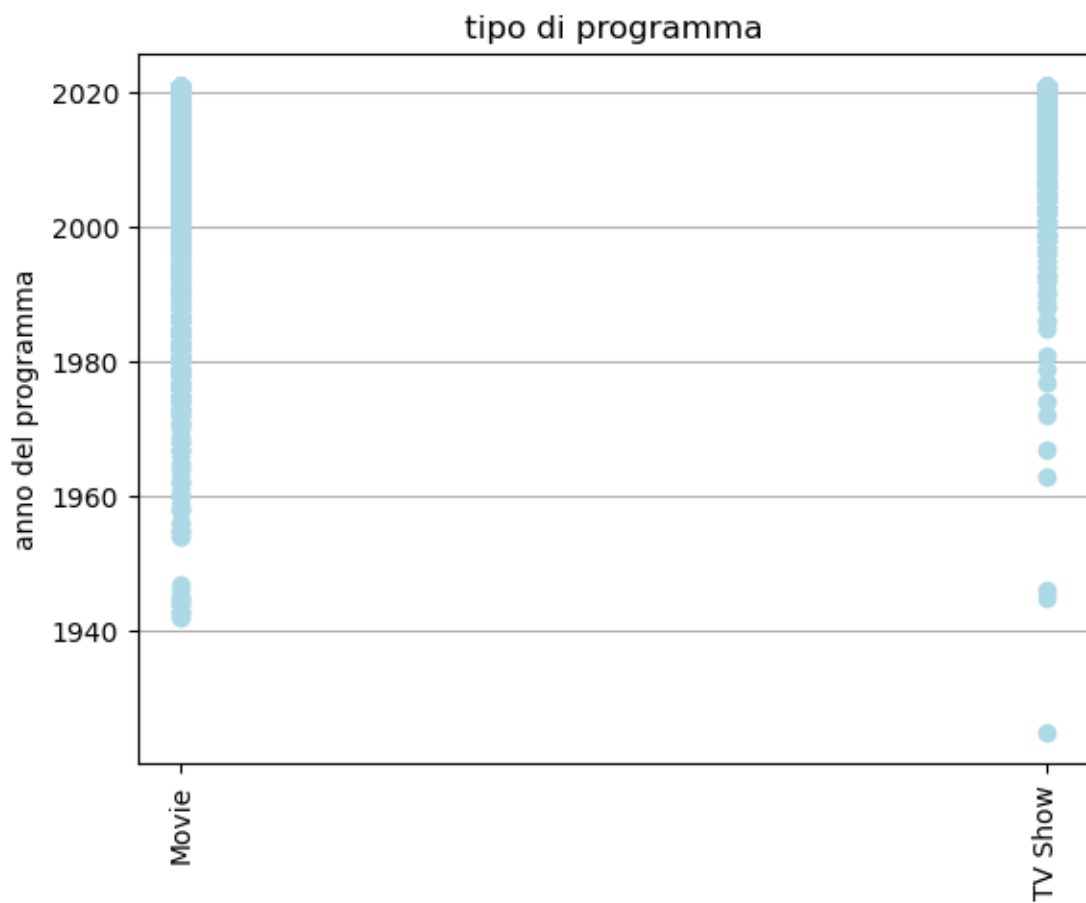
## 1.4 Identificazione dell'Anno con il Maggior Numero di Programmi Netflix

```
[14]: # Conta quanti programmi ci sono per ogni anno e stampa l'anno che ne ha di più
      import pandas as pd
      anno_prog = df['release_year'].value_counts().idxmax()
      print(anno_prog)
```

2018

## 1.5 Visualizzazione della Distribuzione dei Tipi di Programmi Netflix nel Corso degli Anni

```
[15]: # Grafico a dispersione che mostra la distribuzione dei tipi di programmi negli anni
      ↪anni
import matplotlib.pyplot as plt
plt.plot(df['type'],df['release_year'], marker='o', linestyle='',
      ↪color='lightblue')
plt.title('tipo di programma')
plt.ylabel('anno del programma')
plt.xticks(rotation=90)
plt.grid(True, axis="y")
plt.show()
```



## 1.6 Identificazione e Stampa delle Righe con Valori Mancanti nel DataFrame

```
[16]: # identifica le righe con valori mancanti e lo stampa
righe_con_dati_mancanti = df[df.isnull().any(axis=1)]
righe_con_dati_mancanti
```

```
[16]:
```

	show_id	type	title	director	\
0	s1	Movie	Dick Johnson Is Dead	Kirsten Johnson	
1	s2	TV Show	Blood & Water	NaN	
2	s3	TV Show	Ganglands	Julien Leclercq	
3	s4	TV Show	Jailbirds New Orleans	NaN	
4	s5	TV Show	Kota Factory	NaN	
...	...	...	...	...	
8795	s8796	TV Show	Yu-Gi-Oh! Arc-V	NaN	
8796	s8797	TV Show	Yunus Emre	NaN	
8797	s8798	TV Show	Zak Storm	NaN	
8800	s8801	TV Show	Zindagi Gulzar Hai	NaN	
8803	s8804	TV Show	Zombie Dumb	NaN	

	cast	\
0	NaN	
1	Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban...	
2	Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi...	
3	NaN	
4	Mayur More, Jitendra Kumar, Ranjan Raj, Alam K...	
...	...	
8795	Mike Liscio, Emily Bauer, Billy Bob Thompson, ...	
8796	Gökhan Atalay, Payidar Tüfekçioğlu, Baran Akbu...	
8797	Michael Johnston, Jessica Gee-George, Christin...	
8800	Sanam Saeed, Fawad Khan, Ayesha Omer, Mehreen ...	
8803	NaN	

	country	date_added	\
0	United States	September 25, 2021	
1	South Africa	September 24, 2021	
2	NaN	September 24, 2021	
3	NaN	September 24, 2021	
4	India	September 24, 2021	
...	...	...	
8795	Japan, Canada	May 1, 2018	
8796	Turkey	January 17, 2017	
8797	United States, France, South Korea, Indonesia	September 13, 2018	
8800	Pakistan	December 15, 2016	
8803	NaN	July 1, 2019	

	release_year	rating	duration	\
0	2020	PG-13	90 min	

1	2021	TV-MA	2 Seasons
2	2021	TV-MA	1 Season
3	2021	TV-MA	1 Season
4	2021	TV-MA	2 Seasons
...	...	...	...
8795	2015	TV-Y7	2 Seasons
8796	2016	TV-PG	2 Seasons
8797	2016	TV-Y7	3 Seasons
8800	2012	TV-PG	1 Season
8803	2018	TV-Y7	2 Seasons

	listed_in \
0	Documentaries
1	International TV Shows, TV Dramas, TV Mysteries
2	Crime TV Shows, International TV Shows, TV Act...
3	Docuseries, Reality TV
4	International TV Shows, Romantic TV Shows, TV ...
...	...
8795	Anime Series, Kids' TV
8796	International TV Shows, TV Dramas
8797	Kids' TV
8800	International TV Shows, Romantic TV Shows, TV ...
8803	Kids' TV, Korean TV Shows, TV Comedies

	description
0	As her father nears the end of his life, filmm...
1	After crossing paths at a party, a Cape Town t...
2	To protect his family from a powerful drug lor...
3	Feuds, flirtations and toilet talk go down amo...
4	In a city of coaching centers known to train I...
...	...
8795	Now that he's discovered the Pendulum Summonin...
8796	During the Mongol invasions, Yunus Emre leaves...
8797	Teen surfer Zak Storm is mysteriously transpor...
8800	Strong-willed, middle-class Kashaf and carefre...
8803	While living alone in a spooky town, a young g...

[3475 rows x 12 columns]

## 1.7 Calcolo e Stampa del Numero Totale di Righe con Dati Mancanti

```
[18]: # calcola il numero totale di righe con dati mancanti e lo assegna alla
      ↪variabile tot_dati_mancanti e poi la stampa
      tot_dati_mancanti = righe_con_dati_mancanti.shape[0]
      tot_dati_mancanti
```

[18]: 3475

## 1.8 Identificazione e Rimozione delle Righe con Valori Mancanti dal DataFrame

```
[2]: # identifica le righe con valori mancanti e le rimuove dal dataframe df1, poi
      ↳ lo stampa
      df1=df.dropna(inplace=False)
      df1
```

```
[2]:
```

	show_id	type	title	director \
7	s8	Movie	Sankofa	Haile Gerima
8	s9	TV Show	The Great British Baking Show	Andy Devonshire
9	s10	Movie	The Starling	Theodore Melfi
12	s13	Movie	Je Suis Karl	Christian Schwochow
24	s25	Movie	Jeans	S. Shankar
...	...	...	...	...
8801	s8802	Movie	Zinzana	Majid Al Ansari
8802	s8803	Movie	Zodiac	David Fincher
8804	s8805	Movie	Zombieland	Ruben Fleischer
8805	s8806	Movie	Zoom	Peter Hewitt
8806	s8807	Movie	Zubaan	Mozez Singh

```
cast \
```

7	Kofi Ghanaba, Oyafunmike Ogunlano, Alexandra D...
8	Mel Giedroyc, Sue Perkins, Mary Berry, Paul Ho...
9	Melissa McCarthy, Chris O'Dowd, Kevin Kline, T...
12	Luna Wedler, Jannis Niewöhner, Milan Peschel, ...
24	Prashanth, Aishwarya Rai Bachchan, Sri Lakshmi...
...	...
8801	Ali Suliman, Saleh Bakri, Yasa, Ali Al-Jabri, ...
8802	Mark Ruffalo, Jake Gyllenhaal, Robert Downey J...
8804	Jesse Eisenberg, Woody Harrelson, Emma Stone, ...
8805	Tim Allen, Courteney Cox, Chevy Chase, Kate Ma...
8806	Vicky Kaushal, Sarah-Jane Dias, Raaghav Chanan...

```
country date_added \
```

7	United States, Ghana, Burkina Faso, United Kin...	September 24, 2021
8	United Kingdom	September 24, 2021
9	United States	September 24, 2021
12	Germany, Czech Republic	September 23, 2021
24	India	September 21, 2021
...	...	...
8801	United Arab Emirates, Jordan	March 9, 2016
8802	United States	November 20, 2019
8804	United States	November 1, 2019
8805	United States	January 11, 2020
8806	India	March 2, 2019

```
release_year rating duration \
```

7	1993	TV-MA	125 min
8	2021	TV-14	9 Seasons
9	2021	PG-13	104 min
12	2021	TV-MA	127 min
24	1998	TV-14	166 min
...	...	...	...
8801	2015	TV-MA	96 min
8802	2007	R	158 min
8804	2009	R	88 min
8805	2006	PG	88 min
8806	2015	TV-14	111 min

	listed_in \
7	Dramas, Independent Movies, International Movies
8	British TV Shows, Reality TV
9	Comedies, Dramas
12	Dramas, International Movies
24	Comedies, International Movies, Romantic Movies
...	...
8801	Dramas, International Movies, Thrillers
8802	Cult Movies, Dramas, Thrillers
8804	Comedies, Horror Movies
8805	Children & Family Movies, Comedies
8806	Dramas, International Movies, Music & Musicals

	description
7	On a photo shoot in Ghana, an American model s...
8	A talented batch of amateur bakers face off in...
9	A woman adjusting to life after a loss contend...
12	After most of her family is murdered in a terr...
24	When the father of the man she loves insists t...
...	...
8801	Recovering alcoholic Talal wakes up inside a s...
8802	A political cartoonist, a crime reporter and a...
8804	Looking to survive in a world taken over by zo...
8805	Dragged from civilian life, a former superhero...
8806	A scrappy but poor boy worms his way into a ty...

[5332 rows x 12 columns]

## 1.9 Creazione di una Matrice Booleana per Indicare Valori Mancanti nel DataFrame

```
[3]: # Utilizza il metodo isnull() sul DataFrame df per creare una matrice booleana
      ↪ (valori True o False)
      # missing_matrix che indica se c'è un valore mancante (NaN) in ciascuna
      ↪ posizione del DataFrame.
```



```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
missing_matrix = df.isnull()
missing_matrix
```

```
[3]:
```

	show_id	type	title	director	cast	country	date_added	\
0	False	False	False	False	True	False	False	
1	False	False	False	True	False	False	False	
2	False	False	False	False	False	True	False	
3	False	False	False	True	True	True	False	
4	False	False	False	True	False	False	False	
...	...	...	...	...	...	...	...	
8802	False	False	False	False	False	False	False	
8803	False	False	False	True	True	True	False	
8804	False	False	False	False	False	False	False	
8805	False	False	False	False	False	False	False	
8806	False	False	False	False	False	False	False	

	release_year	rating	duration	listed_in	description
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
...	...	...	...	...	...
8802	False	False	False	False	False
8803	False	False	False	False	False
8804	False	False	False	False	False
8805	False	False	False	False	False
8806	False	False	False	False	False

[8807 rows x 12 columns]

## 1.10 Selezione e Stampa dei Nomi delle Colonne Numeriche del DataFrame

```
[4]: # seleziona le colonne del Df che contengono dati numerici e le mette nella
      ↪ variabile numeric_cols, e poi stampa il nome delle colonne
numeric_cols = df.select_dtypes(include=['number'])
numeric_cols.columns
```

```
[4]: Index(['release_year'], dtype='object')
```

### 1.11 Calcolo del Numero di Valori Mancanti per Ogni Colonna in un DataFrame

```
[5]: # calcola il numero di valori mancanti per ogni colonna
df.isnull().sum()
```

```
[5]: show_id          0
      type            0
      title           0
      director       2634
      cast           825
      country        831
      date_added      10
      release_year    0
      rating          4
      duration        3
      listed_in       0
      description     0
      dtype: int64
```

### 1.12 Calcolo della Percentuale di Valori Mancanti per Ogni Colonna in un DataFrame

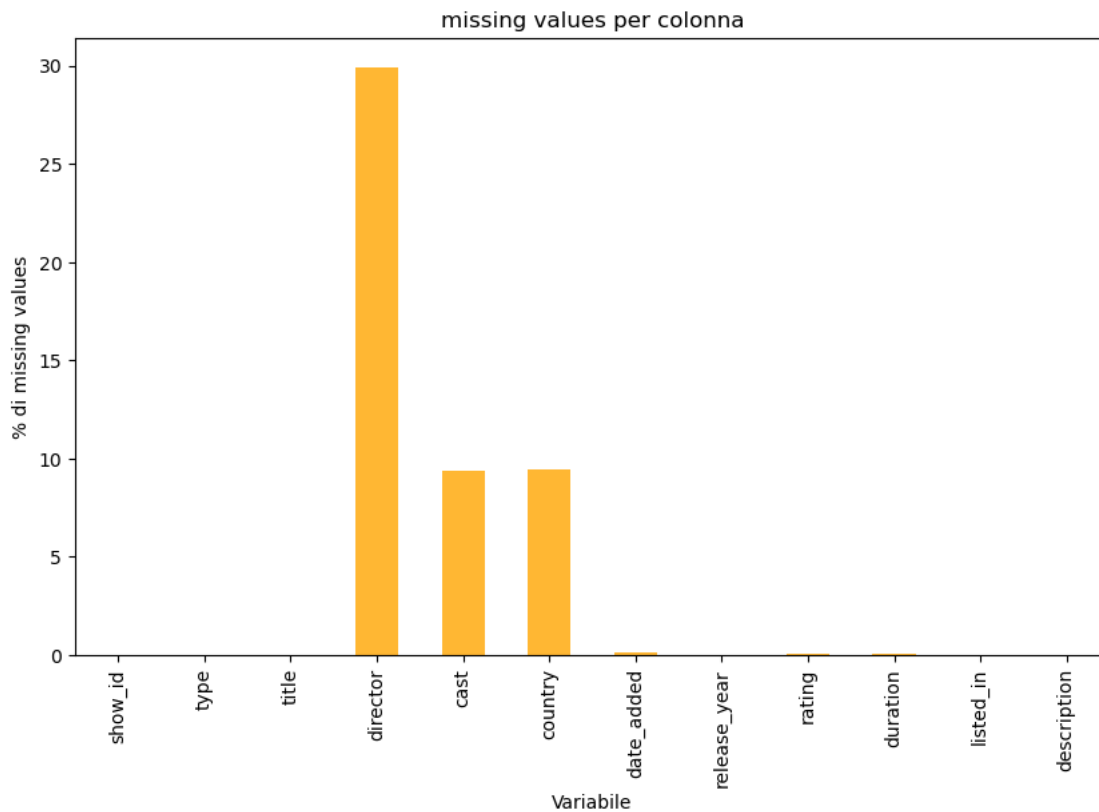
```
[6]: # calcola per ogni colonna la percentuale di valori mancanti su tutte le righe
      ↳ del dataframe
missing_percent = df.isnull().sum() / len(df) * 100
missing_percent
```

```
[6]: show_id          0.000000
      type            0.000000
      title           0.000000
      director       29.908028
      cast           9.367549
      country        9.435676
      date_added      0.113546
      release_year    0.000000
      rating          0.045418
      duration        0.034064
      listed_in       0.000000
      description     0.000000
      dtype: float64
```

### 1.13 Calcolo della Percentuale di Valori Mancanti per Ogni Colonna in un DataFrame e Creazione del Grafico a Barre Corrispondente

```
[7]: # calcola per ogni colonna la percentuale di valori mancanti su tutte le righe
      ↳ del dataframe e poi crea il grafico a barre
missing_percent= (df.isnull().sum()) / len(df) * 100
```

```
plt.figure(figsize=(10,6))
missing_percent.plot(kind='bar', color='orange', alpha=0.8)
plt.xlabel('Variabile')
plt.ylabel('% di missing values')
plt.title('missing values per colonna')
plt.xticks(rotation=90)
plt.show()
```



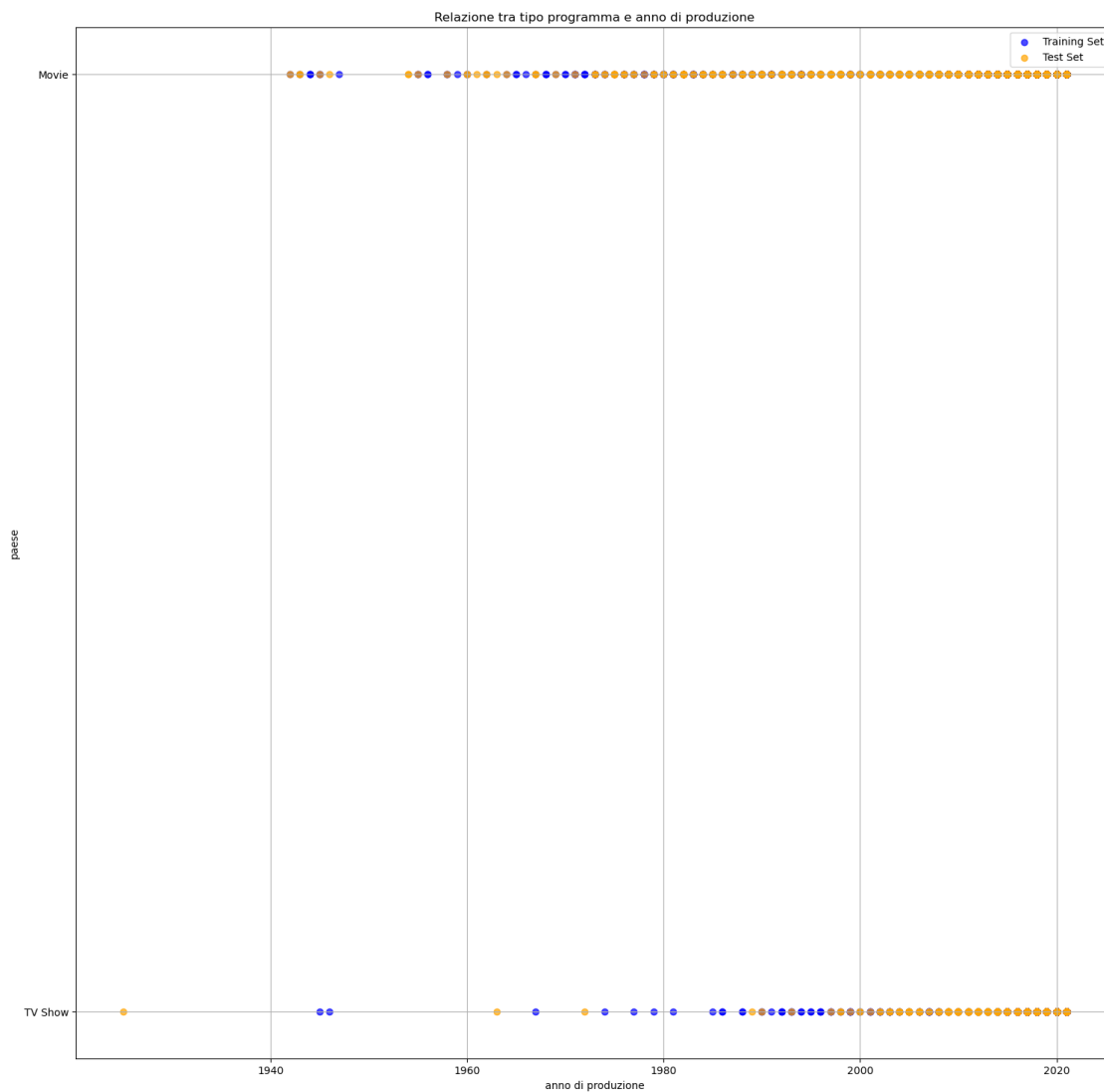
### 1.14 Suddivisione del Dataset in Training e Test Set, Creazione di un Grafico a Dispersione e Stampa delle Dimensioni dei Set

```
[9]: import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
# Suddivisione del dataset in training set (70%) e test set (30%)
X_train, X_test, y_train, y_test = \
    train_test_split(df['release_year'], df['type'], test_size=0.3,
    random_state=42)
# Creazione di un grafico a dispersione
plt.figure(figsize=(2^16, 2^16))
plt.scatter(X_train, y_train, label='Training Set', color='blue', alpha=0.7)
```

```

plt.scatter(X_test, y_test, label='Test Set', color='orange', alpha=0.7)
plt.xlabel('anno di produzione')
plt.ylabel('paese')
plt.title('Relazione tra tipo programma e anno di produzione')
plt.legend()
plt.grid(True)
plt.show()
# Stampare le dimensioni dei training set e test set
print("Dimensioni del Training Set (tipo programma e anno di produzione):
      ↪",X_train.shape, y_train.shape)
print("Dimensioni del Test Set (tipo programma e anno di produzione):",X_test.
      ↪shape, y_test.shape)

```



Dimensioni del Training Set (tipo programma e anno di produzione): (6164,)

(6164,)

Dimensioni del Test Set (tipo programma e anno di produzione): (2643,) (2643,)

### 1.15 Creazione di Tre Subset Casuali da un DataFrame

```
[10]: # Creare tre subset di dimensioni simili
# primo subset: campione casuale di 1/3 delle righe del df di partenza
subset1 = df.sample(frac=1/3)
# stampa il numero di righe del subset1
l1=len(subset1)
print(l1)
df = df.drop(subset1.index)
# secondo subset: campione casuale con metà delle righe rimanenti (la metà dei
↳ 2/3 rimanenti)
subset2 = df.sample(frac=1/2)
# stampa il numero di righe del subset2
l2=len(subset2)
print(l2)
df = df.drop(subset2.index)
# terzo subset: le righe restanti
subset3 = df
# stampa il numero di righe del subset3
l3=len(subset3)
print(l3)
```

2936

2936

2935

### 1.16 Calcolo delle Percentuali dei Valori Unici per il Paese nel Subset1

```
[22]: percentuali_subset1 = subset1['country'].value_counts(normalize=True)
percentuali_subset1
```

```
[22]: country
United States    0.329981
India            0.124767
United Kingdom   0.057728
Japan            0.029795
South Korea      0.024209
...
United Kingdom, Hungary, Australia    0.000372
United States, China, Hong Kong       0.000372
United States, Morocco                0.000372
India, Germany                        0.000372
United States, Hungary, Ireland, Canada 0.000372
Name: proportion, Length: 347, dtype: float64
```

## 1.17 Identificazione degli Outliers nell'Anno di Rilascio

```
[10]: import pandas as pd
import matplotlib.pyplot as plt
# Lista con outliers da entrambi i lati
# Calcola la media e la deviazione standard
mean_value = df['release_year'].mean()
print('media anno:')
print(mean_value)
std_dev = df['release_year'].std()
print('deviazione standard:')
print(std_dev)
# Identifica gli outliers considerando  $\pm 3 * dev\_std$  dalla media
outliers = df[(df['release_year'] > mean_value + 3 * std_dev) |
               (df['release_year'] < mean_value - 3 * std_dev)]
outliers
```

```
media anno:
2014.1801975701146
deviazione standard:
8.819312130834057
```

```
[10]:
```

	show_id	type	title \
41	s42	Movie	Jaws
42	s43	Movie	Jaws 2
43	s44	Movie	Jaws 3
44	s45	Movie	Jaws: The Revenge
131	s132	Movie	Blade Runner: The Final Cut
...	...	...	...
8739	s8740	Movie	Why We Fight: The Battle of Russia
8745	s8746	Movie	Willy Wonka & the Chocolate Factory
8748	s8749	Movie	Winter of Our Dreams
8763	s8764	Movie	WWII: Report from the Aleutians
8792	s8793	Movie	Young Tiger

	director \
41	Steven Spielberg
42	Jeannot Szwarc
43	Joe Alves
44	Joseph Sargent
131	Ridley Scott
...	...
8739	Frank Capra, Anatole Litvak
8745	Mel Stuart
8748	John Duigan
8763	John Huston
8792	Mu Chu

	cast \
41	Roy Scheider, Robert Shaw, Richard Dreyfuss, L...
42	Roy Scheider, Lorraine Gary, Murray Hamilton, ...
43	Dennis Quaid, Bess Armstrong, Simon MacCorkind...
44	Lorraine Gary, Lance Guest, Mario Van Peebles,...
131	Harrison Ford, Rutger Hauer, Sean Young, Edwar...
...	...
8739	NaN
8745	Gene Wilder, Jack Albertson, Peter Ostrum, Roy...
8748	Judy Davis, Bryan Brown, Cathy Downes, Baz Luh...
8763	NaN
8792	Qiu Yuen, Charlie Chin, Jackie Chan, Hu Chin, ...

	country	date_added \
41	United States	September 16, 2021
42	United States	September 16, 2021
43	United States	September 16, 2021
44	United States	September 16, 2021
131	United States	September 1, 2021
...	...	...
8739	United States	March 31, 2017
8745	United States, East Germany, West Germany	January 1, 2020
8748	Australia	November 1, 2016
8763	United States	March 31, 2017
8792	Hong Kong	November 1, 2016

	release_year	rating	duration \
41	1975	PG	124 min
42	1978	PG	116 min
43	1983	PG	98 min
44	1987	PG-13	91 min
131	1982	R	117 min
...	...	...	...
8739	1943	TV-PG	82 min
8745	1971	G	100 min
8748	1981	NR	86 min
8763	1943	TV-PG	45 min
8792	1973	NR	81 min

	listed_in \
41	Action & Adventure, Classic Movies, Dramas
42	Dramas, Horror Movies, Thrillers
43	Action & Adventure, Horror Movies, Thrillers
44	Action & Adventure, Horror Movies, Thrillers
131	Action & Adventure, Classic Movies, Cult Movies
...	...
8739	Documentaries

```

8745 Children & Family Movies, Classic Movies, Come...
8748 Classic Movies, Dramas
8763 Documentaries
8792 Action & Adventure, International Movies

```

```

description
41 When an insatiable great white shark terrorize...
42 Four years after the last deadly shark attacks...
43 After the staff of a marine theme park try to ...
44 After another deadly shark attack, Ellen Brody...
131 In a smog-choked dystopian Los Angeles, blade ...
...
8739 This installment of Frank Capra's acclaimed do...
8745 Zany Willy Wonka causes a stir when he announc...
8748 After the death of a long-ago lover, married p...
8763 Filmmaker John Huston narrates this Oscar-nomi...
8792 Aided only by a tough female police officer, a...

```

[217 rows x 12 columns]

## 2 USO DEI MODELLI

### 3 Utilizzo della Random Forest Classifier

```

[11]: import category_encoders as ce
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import classification_report

# 'release_year' è la variabile target
X = df.drop('release_year', axis=1)
y = df['release_year']

# splitto i dati in train e test set
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
↳random_state=42)

# creo un codificatore di tutte le variabili categoriali
encoder = ce.OrdinalEncoder(cols=['show_id', 'type', 'title', 'director',
↳'cast', 'country', 'date_added', 'rating', 'duration', 'listed_in',
↳'description'])

# adatto e trasformo con il codificatore i train data
X_train = encoder.fit_transform(X_train)

# adatto e trasformo con il codificatore i test data

```



```

X_test = encoder.transform(X_test)

# inizializzo il classificatore RandomForest con 100 stime
rfc = RandomForestClassifier(random_state=42)

# addestrqa il modello al train set
rfc.fit(X_train, y_train)

# effettua le predizioni sul test set
y_pred = rfc.predict(X_test)

accuracy = accuracy_score(y_test, y_pred)
cl_rep = classification_report(y_test, y_pred)

# visualizza l'accuratezza e il report di classificazione
print("Accuratezza del modello", accuracy)
print("\nReport di classificazione:\n", cl_rep)

```

Accuratezza del modello 0.1424517593643587

Report di classificazione:

	precision	recall	f1-score	support
1925	0.00	0.00	0.00	1
1942	0.00	0.00	0.00	1
1943	0.00	0.00	0.00	2
1954	0.00	0.00	0.00	2
1960	0.00	0.00	0.00	2
1961	0.00	0.00	0.00	1
1962	0.00	0.00	0.00	1
1963	0.00	0.00	0.00	2
1967	0.00	0.00	0.00	3
1969	0.00	0.00	0.00	1
1971	0.00	0.00	0.00	1
1973	0.00	0.00	0.00	2
1974	0.00	0.00	0.00	2
1975	0.00	0.00	0.00	4
1976	0.00	0.00	0.00	3
1977	0.00	0.00	0.00	2
1978	0.00	0.00	0.00	1
1979	0.00	0.00	0.00	2
1980	0.00	0.00	0.00	1
1981	0.00	0.00	0.00	2
1982	0.00	0.00	0.00	6
1983	0.00	0.00	0.00	1
1984	0.00	0.00	0.00	3
1985	0.00	0.00	0.00	3
1986	0.00	0.00	0.00	3

1987	0.00	0.00	0.00	1
1988	0.00	0.00	0.00	2
1989	0.00	0.00	0.00	5
1990	0.00	0.00	0.00	3
1991	0.00	0.00	0.00	3
1992	0.00	0.00	0.00	6
1993	0.00	0.00	0.00	4
1994	0.00	0.00	0.00	3
1995	0.00	0.00	0.00	5
1996	0.00	0.00	0.00	9
1997	0.00	0.00	0.00	9
1998	0.00	0.00	0.00	7
1999	0.00	0.00	0.00	4
2000	0.17	0.17	0.17	6
2001	0.00	0.00	0.00	5
2002	0.00	0.00	0.00	5
2003	0.00	0.00	0.00	16
2004	0.00	0.00	0.00	17
2005	0.00	0.00	0.00	14
2006	0.00	0.00	0.00	26
2007	0.00	0.00	0.00	17
2008	0.02	0.08	0.03	24
2009	0.00	0.00	0.00	33
2010	0.20	0.02	0.04	42
2011	0.00	0.00	0.00	39
2012	0.14	0.02	0.03	51
2013	0.00	0.00	0.00	59
2014	0.00	0.00	0.00	69
2015	0.03	0.01	0.01	101
2016	0.16	0.09	0.11	194
2017	0.13	0.28	0.18	202
2018	0.16	0.30	0.21	211
2019	0.14	0.17	0.15	189
2020	0.15	0.26	0.19	193
2021	0.30	0.18	0.23	136
accuracy			0.14	1762
macro avg	0.03	0.03	0.02	1762
weighted avg	0.12	0.14	0.12	1762

```

C:\Users\zetam\anaconda3\Lib\site-
packages\sklearn\metrics\_classification.py:1469: UndefinedMetricWarning:
Precision and F-score are ill-defined and being set to 0.0 in labels with no
predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))
C:\Users\zetam\anaconda3\Lib\site-
packages\sklearn\metrics\_classification.py:1469: UndefinedMetricWarning:

```

Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control this behavior.

```
_warn_prf(average, modifier, msg_start, len(result))
```

C:\Users\zetam\anaconda3\Lib\site-

packages\sklearn\metrics\\_classification.py:1469: UndefinedMetricWarning:

Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control this behavior.

```
_warn_prf(average, modifier, msg_start, len(result))
```

### 3.1 Uso della Logistic Regression

```
[4]: import category_encoders as ce
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, classification_report

# 'release_year' è la variabile target
X = df.drop('release_year', axis=1)
y = df['release_year']

# splitto i dati in train e test set
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
    random_state=42)
# creo un codificatore di tutte le variabili categoriali
encoder = ce.OrdinalEncoder(cols=['show_id', 'type', 'title', 'director',
    'cast', 'country', 'date_added', 'rating', 'duration', 'listed_in',
    'description'])

# adatto e trasformo con il codificatore i train data
X_train = encoder.fit_transform(X_train)

# adatto e trasformo con il codificatore i test data
X_test = encoder.transform(X_test)

# inizializzazione del classificatore di regressioni logistica
log_reg_classifier = LogisticRegression(random_state=42)

# addestra il modello sul train set
log_reg_classifier.fit(X_train, y_train)

# effettua le predizioni sul test set
y_pred = log_reg_classifier.predict(X_test)

# valutazione prestazioni modello
accuracy = accuracy_score(y_test, y_pred)
cl_rep = classification_report(y_test, y_pred)
```

```
# visualizza l'accuratezza e il report di classificazione
print("Accuratezza del modello", accuracy)
print("\nReport di classificazione:\n", cl_rep)
```

Accuratezza del modello 0.13904653802497163

Report di classificazione:

	precision	recall	f1-score	support
1925	0.00	0.00	0.00	1
1942	0.00	0.00	0.00	1
1943	0.00	0.00	0.00	2
1954	0.00	0.00	0.00	2
1960	0.00	0.00	0.00	2
1961	0.00	0.00	0.00	1
1962	0.00	0.00	0.00	1
1963	0.00	0.00	0.00	2
1967	0.00	0.00	0.00	3
1969	0.00	0.00	0.00	1
1971	0.00	0.00	0.00	1
1973	0.00	0.00	0.00	2
1974	0.00	0.00	0.00	2
1975	0.00	0.00	0.00	4
1976	0.00	0.00	0.00	3
1977	0.00	0.00	0.00	2
1978	0.00	0.00	0.00	1
1979	0.00	0.00	0.00	2
1980	0.00	0.00	0.00	1
1981	0.00	0.00	0.00	2
1982	0.00	0.00	0.00	6
1983	0.00	0.00	0.00	1
1984	0.00	0.00	0.00	3
1985	0.00	0.00	0.00	3
1986	0.00	0.00	0.00	3
1987	0.00	0.00	0.00	1
1988	0.00	0.00	0.00	2
1989	0.00	0.00	0.00	5
1990	0.00	0.00	0.00	3
1991	0.00	0.00	0.00	3
1992	0.00	0.00	0.00	6
1993	0.00	0.00	0.00	4
1994	0.00	0.00	0.00	3
1995	0.00	0.00	0.00	5
1996	0.00	0.00	0.00	9
1997	0.00	0.00	0.00	9
1998	0.00	0.00	0.00	7
1999	0.00	0.00	0.00	4
2000	0.00	0.00	0.00	6

2001	0.00	0.00	0.00	5
2002	0.00	0.00	0.00	5
2003	0.00	0.00	0.00	16
2004	0.00	0.00	0.00	17
2005	0.00	0.00	0.00	14
2006	0.00	0.00	0.00	26
2007	0.00	0.00	0.00	17
2008	0.00	0.00	0.00	24
2009	0.00	0.00	0.00	33
2010	0.00	0.00	0.00	42
2011	0.00	0.00	0.00	39
2012	0.00	0.00	0.00	51
2013	0.00	0.00	0.00	59
2014	0.00	0.00	0.00	69
2015	0.00	0.00	0.00	101
2016	0.12	0.29	0.18	194
2017	0.16	0.11	0.13	202
2018	0.10	0.09	0.10	211
2019	0.00	0.00	0.00	189
2020	0.13	0.47	0.20	193
2021	0.21	0.40	0.28	136
accuracy				0.14 1762
macro avg		0.01	0.02	0.01 1762
weighted avg		0.07	0.14	0.09 1762

```
C:\Users\zetam\anaconda3\Lib\site-
packages\sklearn\linear_model\_logistic.py:460: ConvergenceWarning: lbfgs failed
to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
```

Increase the number of iterations (max\_iter) or scale the data as shown in:

<https://scikit-learn.org/stable/modules/preprocessing.html>

Please also refer to the documentation for alternative solver options:

[https://scikit-learn.org/stable/modules/linear\\_model.html#logistic-regression](https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression)

```
n_iter_i = _check_optimize_result(
```

```
C:\Users\zetam\anaconda3\Lib\site-
packages\sklearn\metrics\_classification.py:1469: UndefinedMetricWarning:
Precision and F-score are ill-defined and being set to 0.0 in labels with no
predicted samples. Use `zero_division` parameter to control this behavior.
```

```
_warn_prf(average, modifier, msg_start, len(result))
```

```
C:\Users\zetam\anaconda3\Lib\site-
packages\sklearn\metrics\_classification.py:1469: UndefinedMetricWarning:
Precision and F-score are ill-defined and being set to 0.0 in labels with no
predicted samples. Use `zero_division` parameter to control this behavior.
```

```
_warn_prf(average, modifier, msg_start, len(result))
```

```
C:\Users\zetam\anaconda3\Lib\site-  
packages\sklearn\metrics\_classification.py:1469: UndefinedMetricWarning:  
Precision and F-score are ill-defined and being set to 0.0 in labels with no  
predicted samples. Use `zero_division` parameter to control this behavior.  
_warn_prf(average, modifier, msg_start, len(result))
```

### 3.2 Uso della Support Vector Classifier

```
[25]: import category_encoders as ce  
from sklearn.model_selection import train_test_split  
from sklearn.svm import SVC  
from sklearn.metrics import accuracy_score, classification_report  
  
# 'release_year' è la variabile target  
X = df.drop('release_year', axis=1)  
y = df['release_year']  
  
# splitto i dati in train e test set  
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,  
                                                    random_state=42)  
# creo un codificatore di tutte le variabili categoriali  
encoder = ce.OrdinalEncoder(cols=['show_id', 'type', 'title', 'director',  
                                'cast', 'country', 'date_added', 'rating', 'duration', 'listed_in',  
                                'description'])  
  
# adatto e trasformo con il codificatore i train data  
X_train = encoder.fit_transform(X_train)  
  
# adatto e trasformo con il codificatore i test data  
X_test = encoder.transform(X_test)  
# inizializzazione del classificatore SVC  
svm_cl = SVC(random_state=42)  
  
# addestra il modello sul train set  
svm_cl.fit(X_train, y_train)  
  
# effettua le predizioni sul test set  
y_pred = svm_cl.predict(X_test)  
  
# valutazione prestazioni modello  
accuracy = accuracy_score(y_test, y_pred)  
cl_rep = classification_report(y_test, y_pred)  
  
# visualizza l'accuratezza e il report di classificazione  
print("Accuratezza del modello", accuracy)  
print("\nReport di classificazione:\n", cl_rep)
```

Accuratezza del modello 0.12776831345826234

Report di classificazione:

	precision	recall	f1-score	support
1944	0.00	0.00	0.00	1
1963	0.00	0.00	0.00	1
1967	0.00	0.00	0.00	1
1968	0.00	0.00	0.00	1
1971	0.00	0.00	0.00	1
1974	0.00	0.00	0.00	2
1978	0.00	0.00	0.00	1
1980	0.00	0.00	0.00	1
1981	0.00	0.00	0.00	1
1982	0.00	0.00	0.00	5
1983	0.00	0.00	0.00	1
1986	0.00	0.00	0.00	1
1989	0.00	0.00	0.00	1
1990	0.00	0.00	0.00	1
1992	0.00	0.00	0.00	1
1993	0.00	0.00	0.00	3
1994	0.00	0.00	0.00	2
1995	0.00	0.00	0.00	2
1996	0.00	0.00	0.00	1
1997	0.00	0.00	0.00	4
1998	0.00	0.00	0.00	2
1999	0.00	0.00	0.00	5
2000	0.00	0.00	0.00	5
2001	0.00	0.00	0.00	5
2002	0.00	0.00	0.00	4
2003	0.00	0.00	0.00	4
2004	0.00	0.00	0.00	7
2005	0.00	0.00	0.00	4
2006	0.00	0.00	0.00	6
2007	0.00	0.00	0.00	6
2008	0.00	0.00	0.00	10
2009	0.00	0.00	0.00	9
2010	0.00	0.00	0.00	8
2011	0.00	0.00	0.00	16
2012	0.00	0.00	0.00	17
2013	0.00	0.00	0.00	25
2014	0.00	0.00	0.00	22
2015	0.00	0.00	0.00	36
2016	0.00	0.00	0.00	46
2017	0.26	0.13	0.17	78
2018	0.12	0.87	0.21	70
2019	0.00	0.00	0.00	66
2020	0.17	0.07	0.10	61
2021	0.00	0.00	0.00	43

accuracy			0.13	587
macro avg	0.01	0.02	0.01	587
weighted avg	0.07	0.13	0.06	587

```
C:\Users\zetam\anaconda3\Lib\site-
packages\sklearn\metrics\_classification.py:1469: UndefinedMetricWarning:
Precision and F-score are ill-defined and being set to 0.0 in labels with no
predicted samples. Use `zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
C:\Users\zetam\anaconda3\Lib\site-
packages\sklearn\metrics\_classification.py:1469: UndefinedMetricWarning:
Precision and F-score are ill-defined and being set to 0.0 in labels with no
predicted samples. Use `zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
C:\Users\zetam\anaconda3\Lib\site-
packages\sklearn\metrics\_classification.py:1469: UndefinedMetricWarning:
Precision and F-score are ill-defined and being set to 0.0 in labels with no
predicted samples. Use `zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
```