```
import pandas as pd
 In [1]:
          movies = pd.read_csv("movies.csv")
          movies.head()
 In [3]:
             movield
 Out[3]:
                                                                             genres
          0
                  1
                                 Toy Story (1995)
                                               Adventure|Animation|Children|Comedy|Fantasy
          1
                  2
                                                              Adventure|Children|Fantasy
                                  Jumanji (1995)
                  3
          2
                          Grumpier Old Men (1995)
                                                                     Comedy|Romance
          3
                  4
                           Waiting to Exhale (1995)
                                                               Comedy|Drama|Romance
          4
                  5 Father of the Bride Part II (1995)
                                                                            Comedy
 In [4]:
         import re
          def clean_title(title):
               title = re.sub("[^a-zA-Z0-9]", "", title)
               return title
          movies["clean_title"] = movies["title"].apply(clean_title)
          movies
                 movield
                                              title
Out[4]:
                                                                                 genres
                                                                                                       clean_title
              0
                                     Toy Story (1995)
                                                  Adventure|Animation|Children|Comedy|Fantasy
                                                                                                     Toy Story 1995
                      1
                      2
              1
                                      Jumanji (1995)
                                                                  Adventure|Children|Fantasy
                                                                                                      Jumanji 1995
              2
                      3
                              Grumpier Old Men (1995)
                                                                         Comedy|Romance
                                                                                              Grumpier Old Men 1995
                                                                   Comedy|Drama|Romance
              3
                               Waiting to Exhale (1995)
                                                                                              Waiting to Exhale 1995
                      5 Father of the Bride Part II (1995)
              4
                                                                                Comedy
                                                                                        Father of the Bride Part II 1995
                 209157
          62418
                                         We (2018)
                                                                                 Drama
                                                                                                         We 2018
                             Window of the Soul (2001)
                                                                            Documentary
                                                                                             Window of the Soul 2001
          62419
                 209159
          62420
                 209163
                                   Bad Poems (2018)
                                                                           Comedy|Drama
                                                                                                   Bad Poems 2018
          62421
                 209169
                                   A Girl Thing (2001)
                                                                          (no genres listed)
                                                                                                   A Girl Thing 2001
          62422
                 209171
                         Women of Devil's Island (1962)
                                                                    Action|Adventure|Drama
                                                                                          Women of Devils Island 1962
         62423 rows × 4 columns
          from sklearn.feature_extraction.text import TfidfVectorizer
          vectorizer = TfidfVectorizer(ngram_range=(1,2))
          tfidf = vectorizer.fit_transform(movies["clean_title"])
          from sklearn.metrics.pairwise import cosine_similarity
 In [6]:
          import numpy as np
          def search(title):
               title = clean_title(title)
               query_vec = vectorizer.transform([title])
               similarity = cosine_similarity(query_vec, tfidf).flatten()
               indices = np.argpartition(similarity, -5)[-5:]
               results = movies.iloc[indices].iloc[::-1]
               return results
         # pip install ipywidgets
          #jupyter labextension install @jupyter-widgets/jupyterlab-manager
          import ipywidgets as widgets
          from IPython.display import display
          movie_input = widgets.Text(
              value='Toy Story',
              description='Movie Title:',
              disabled=False
          movie_list = widgets.Output()
          def on_type(data):
              with movie_list:
                   movie_list.clear_output()
                   title = data["new"]
                   if len(title) > 5:
                       display(search(title))
          movie_input.observe(on_type, names='value')
          display(movie_input, movie_list)
          Text(value='Toy Story', description='Movie Title:')
          Output()
 In [8]: movie_id = 89745
          #def find_similar_movies(movie_id):
          movie = movies[movies["movieId"] == movie_id]
          ratings = pd.read_csv("ratings.csv")
          ratings.dtypes
          userId
                          int64
Out[9]:
          movieId
                          int64
          rating
                        float64
          timestamp
                          int64
          dtype: object
In [10]: similar_users = ratings[(ratings["movieId"] == movie_id) & (ratings["rating"] > 4)]["userId"].unique()
          similar_user_recs = ratings[(ratings["userId"].isin(similar_users)) & (ratings["rating"] > 4)]["movieId"]
          similar_user_recs = similar_user_recs.value_counts() / len(similar_users)
          similar_user_recs = similar_user_recs[similar_user_recs > .10]
          all_users = ratings[(ratings["movieId"].isin(similar_user_recs.index)) & (ratings["rating"] > 4)]
          all_user_recs = all_users["movieId"].value_counts() / len(all_users["userId"].unique())
          rec_percentages = pd.concat([similar_user_recs, all_user_recs], axis=1)
          rec_percentages.columns = ["similar", "all"]
          rec_percentages
Out[10]:
                   similar
                               all
               1 0.236083 0.126250
              32 0.103877 0.101516
              47 0.203115 0.146232
              50 0.211067 0.202959
             110 0.182240 0.162835
          134853 0.198641 0.036444
          152081 0.133532 0.020652
          164179 0.128728 0.029124
          166528 0.124751 0.014411
          168252 0.132538 0.016469
         193 rows × 2 columns
          rec_percentages["score"] = rec_percentages["similar"] / rec_percentages["all"]
          rec_percentages = rec_percentages.sort_values("score", ascending=False)
          rec_percentages.head(10).merge(movies, left_index=True, right_on="movieId")
                                     score movield
          17067 1.000000 0.040459 24.716368
                                                                     Avengers, The (2012)
                                                                                                                                      Avengers The 2012
                                             89745
                                                                                              Action|Adventure|Sci-Fi|IMAX
                                                                                            Action|Adventure|Fantasy|IMAX
          20513 0.103711 0.005289 19.610199
                                            106072
                                                               Thor: The Dark World (2013)
                                                                                                                                 Thor The Dark World 2013
                                                             Avengers: Age of Ultron (2015)
          25058 0.241054 0.012367 19.491770
                                            122892
                                                                                                                               Avengers Age of Ultron 2015
                                                                                                  Action|Adventure|Sci-Fi
          19678 0.216534 0.012119 17.867419
                                                                                                Action|Sci-Fi|Thriller|IMAX
                                            102125
                                                                       Iron Man 3 (2013)
                                                                                                                                         Iron Man 3 2013
          16725 0.215043 0.012052 17.843074
                                                     Captain America: The First Avenger (2011)
                                              88140
                                                                                         Action|Adventure|Sci-Fi|Thriller|War
                                                                                                                      Captain America The First Avenger 2011
                                                                            Thor (2011) Action|Adventure|Drama|Fantasy|IMAX
          16312 0.175447 0.010142 17.299824
                                              86332
                                                                                                                                             Thor 2011
          21348 0.287608 0.016737 17.183667
                                            110102 Captain America: The Winter Soldier (2014)
                                                                                              Action|Adventure|Sci-Fi|IMAX Captain America The Winter Soldier 2014
                                                                                                     Action|Sci-Fi|Thriller
          25071 0.214049 0.012856 16.649399
                                            122920
                                                            Captain America: Civil War (2016)
                                                                                                                             Captain America Civil War 2016
          25061 0.136017 0.008573 15.865628
                                                                         Ant-Man (2015)
                                                                                                                                           AntMan 2015
                                            122900
                                                                                                  Action|Adventure|Sci-Fi
          14628 0.242876 0.015517 15.651921
                                             77561
                                                                       Iron Man 2 (2010)
                                                                                        Action|Adventure|Sci-Fi|Thriller|IMAX
                                                                                                                                         Iron Man 2 2010
          def find_similar_movies(movie_id):
In [12]:
               similar_users = ratings[(ratings["movieId"] == movie_id) & (ratings["rating"] > 4)]["userId"].unique()
               similar_user_recs = ratings[(ratings["userId"].isin(similar_users)) & (ratings["rating"] > 4)]["movieId"]
               similar_user_recs = similar_user_recs.value_counts() / len(similar_users)
               similar_user_recs = similar_user_recs[similar_user_recs > .10]
               all_users = ratings[(ratings["movieId"].isin(similar_user_recs.index)) & (ratings["rating"] > 4)]
               all_user_recs = all_users["movieId"].value_counts() / len(all_users["userId"].unique())
               rec_percentages = pd.concat([similar_user_recs, all_user_recs], axis=1)
               rec_percentages.columns = ["similar", "all"]
               rec_percentages["score"] = rec_percentages["similar"] / rec_percentages["all"]
               rec_percentages = rec_percentages.sort_values("score", ascending=False)
               return rec_percentages.head(10).merge(movies, left_index=True, right_on="movieId")[["score", "title", "genres"]]
          import ipywidgets as widgets
In [13]:
          from IPython.display import display
          movie_name_input = widgets.Text(
               value='Toy Story',
               description='Movie Title:',
              disabled=False
          recommendation_list = widgets.Output()
          def on_type(data):
               with recommendation_list:
                   recommendation_list.clear_output()
                   title = data["new"]
                   if len(title) > 5:
                       results = search(title)
                       movie_id = results.iloc[0]["movieId"]
                       display(find_similar_movies(movie_id))
          movie_name_input.observe(on_type, names='value')
          display(movie_name_input, recommendation_list)
          Text(value='Toy Story', description='Movie Title:')
          Output()
```