```
In [1]: import pandas as pd
        import numpy as np
In [2]: movies df = pd.read csv('movies.csv')
In [3]: movies df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 9742 entries, 0 to 9741
        Data columns (total 3 columns):
         # Column Non-Null Count Dtype
        --- ----- -----
           movieId 9742 non-null int64
         \cap
            title 9742 non-null object
         2 genres 9742 non-null object
        dtypes: int64(1), object(2)
        memory usage: 228.5+ KB
In [4]: action df = movies df[movies df['genres'] == 'Action'].reset index() #selecting only acti
In [5]: action df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 60 entries, 0 to 59
        Data columns (total 4 columns):
         # Column Non-Null Count Dtype
        --- ----- ------ ----
           index 60 non-null int64
         0
         1 movieId 60 non-null
                                    int64
           title 60 non-null object genres 60 non-null object
           title 60 non-null
        dtypes: int64(2), object(2)
        memory usage: 2.0+ KB
In [6]: action df.head(5)
Out[6]:
          index movield
                                                      title
                                                           genres
        0
                     9
                                          Sudden Death (1995)
                                                            Action
        1
             63
                     71
                                             Fair Game (1995)
                                                            Action
        2
            172
                   204
                               Under Siege 2: Dark Territory (1995)
                                                            Action
        3
            215
                    251
                                            Hunted, The (1995)
                                                            Action
        4
            555
                   667 Bloodsport 2 (a.k.a. Bloodsport II: The Next K...
                                                            Action
In [7]:
        action df.drop('genres', axis= 1, inplace= True)
In [8]: action df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 60 entries, 0 to 59
        Data columns (total 3 columns):
         # Column Non-Null Count Dtype
        --- ----- -----
           index 60 non-null
                                    int64
         0
           movieId 60 non-null
                                    int64
            title 60 non-null object
         2
        dtypes: int64(2), object(1)
        memory usage: 1.5+ KB
```

```
In [9]: ratings df = pd.read csv('ratings.csv')
         ratings df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 100836 entries, 0 to 100835
         Data columns (total 4 columns):
          # Column Non-Null Count Dtype
                        -----
                      100836 non-null int64
          0 userId
          1 movieId 100836 non-null int64
            rating 100836 non-null float64
          3 timestamp 100836 non-null int64
         dtypes: float64(1), int64(3)
         memory usage: 3.1 MB
In [10]: ratings df.drop('timestamp', axis= 1, inplace= True)
In [11]: movie df = ratings df.pivot(index= 'movieId', columns= 'userId',
                                         values= 'rating').reset index(drop= True)
In [12]: movie df.fillna(0, inplace= True)
In [13]: movie df.index = ratings df.movieId.unique()
In [14]: movie_df.iloc[0:5, 0:5]
                               5
Out[14]: userId
                1
                    2
                        3
             1 4.0 0.0 0.0 0.0 4.0
             3 0.0 0.0 0.0 0.0 0.0
             6 4.0 0.0 0.0 0.0 0.0
            47 0.0 0.0 0.0 0.0 0.0
            50 0.0 0.0 0.0 0.0 0.0
In [15]: movie df.fillna(0, inplace= True)
In [16]: from sklearn.metrics import pairwise distances
         from scipy.spatial.distance import cosine, correlation
In [17]: movie_sim = 1 - pairwise_distances(movie_df.values, metric= 'correlation')
In [18]: movie sim df = pd.DataFrame(movie sim)
In [19]: movie sim df.iloc[0:5, 0:5]
Out[19]:
                  0
                                   2
                                            3
                                                    4
         0 1.000000 0.231327 0.173213 -0.028917 0.192474
            0.231327 1.000000 0.191945
                                    0.071269 0.200526
            0.067143 0.370171
         3 -0.028917 0.071269 0.067143
                                    1.000000 0.167910
         4 0.192474 0.200526 0.370171 0.167910 1.000000
In [20]:
         def get similar movies(movieid, topn = 5):
            movieidx = action df[action df.movieId == movieid].index[0]
```

```
top n = action df.sort values(['similarity'], ascending = False)[0:topn]
               return top n
In [21]: get similar movies(71, topn= 5)
Out[21]:
              index movield
                                              title
                                                   similarity
                 63
                          71
                                    Fair Game (1995)
                                                    1.000000
           18
              3345
                        4531
                                    Red Heat (1988)
                                                    0.412808
               7208
                       72874
                             Peacekeeper, The (1997)
                                                   0.333349
          32
               4379
                        6417
                                    Live Wire (1992)
                                                    0.332326
               1910
                       2534
                                    Avalanche (1978) 0.285086
           9
In [22]:
          animation children df = movies df[(movies df['genres'] == 'Animation') |
                                               (movies df['genres'] == 'Children')]
          animation children df.head(5)
Out[22]:
                 movield
                                                  title
                                                         genres
            301
                    343
                             Baby-Sitters Club, The (1995)
                                                        Children
           1097
                   1426
                                Zeus and Roxanne (1997)
                                                        Children
          6973
                  66335 Afro Samurai: Resurrection (2009)
                                                       Animation
                  69469
          7059
                               Garfield's Pet Force (2009)
                                                       Animation
          7195
                  72603
                                Merry Madagascar (2009) Animation
In [23]:
          def get similar movies (movieid, topn= 5):
               movieidx = animation children df[animation children df.movieId == movieid].index[0]
               animation children df['similarity'] = movie sim df.iloc[movieidx]
               top n = animation children df.sort values(['similarity'], ascending= False)[0:topn]
               return top n
In [24]: get similar movies(66335)
          /var/folders/3q/2bh9znv97b109382p2y xj9m0000gn/T/ipykernel 3544/2739994426.py:3: Setting
          WithCopyWarning:
          A value is trying to be set on a copy of a slice from a DataFrame.
          Try using .loc[row indexer, col indexer] = value instead
          See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user
          guide/indexing.html#returning-a-view-versus-a-copy
            animation children df['similarity'] = movie sim df.iloc[movieidx]
Out[24]:
                                                             title
                 movield
                                                                    genres similarity
          6973
                  66335
                                                                             1.000000
                                     Afro Samurai: Resurrection (2009)
                                                                  Animation
          8593
                  117545 Asterix: The Land of the Gods (Astérix: Le dom...
                                                                  Animation
                                                                             1.000000
          9539
                 172587
                                    Vacations in Prostokvashino (1980)
                                                                  Animation
                                                                            0.490280
           9601
                  176051
                            LEGO DC Super Hero Girls: Brain Drain (2017)
                                                                  Animation
                                                                             0.418148
          7279
                  74791 Town Called Panic, A (Panique au village) (2009)
                                                                  Animation
                                                                            0.223780
          action df merged = ratings df.merge(action df, on= 'movieId', how= 'inner')
In [25]:
          action df merged.info()
```

<class 'pandas.core.frame.DataFrame'>

action_df['similarity'] = movie_sim_df.iloc[movieidx]

Int64Index: 186 entries, 0 to 185 Data columns (total 6 columns): # Column Non-Null Count Dtype --------186 non-null int64 0 userId 1 movieId 186 non-null int64 186 non-null float64 2 rating int64 3 index 186 non-null 186 non-null object 4 title 5 similarity 186 non-null float64 dtypes: float64(2), int64(3), object(1) memory usage: 10.2+ KB

In [24]: #animation_children_df = ratings_df.merge(animation_df, on= 'movieId', how= 'inner')
#animation_df.head(5)

Out[24]:		userId	movield	rating	title	genres
	0	50	66335	2.5	Afro Samurai: Resurrection (2009)	Animation
	1	50	81018	3.0	Illusionist, The (L'illusionniste) (2010)	Animation
	2	298	81018	2.0	Illusionist, The (L'illusionniste) (2010)	Animation
	3	318	81018	4.0	Illusionist, The (L'illusionniste) (2010)	Animation
	4	89	69469	5.0	Garfield's Pet Force (2009)	Animation