

```
In [629]: import pandas as pd
import numpy as np
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.metrics.pairwise import linear_kernel
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.metrics.pairwise import cosine_similarity
```

```
In [630]: #Initialize a TFIDF Vectorsizer
tfidf = TfidfVectorizer(stop_words='english')
```

```
In [631]: movies=pd.read_csv("IMDBdata_MainData.csv")
```

```
In [632]: movies.columns
```

```
Out[632]: Index(['Title', 'Year', 'Rated', 'Released', 'Runtime', 'Genre', 'Director',
                'Writer', 'Actors', 'Plot', 'Language', 'Country', 'Awards',
                'Poster',
                'Ratings.Source', 'Ratings.Value', 'Metascore', 'imdbRating',
                'imdbVotes', 'imdbID', 'Type', 'DVD', 'BoxOffice', 'Production',
                'Website', 'Response', 'tomatoURL'],
                dtype='object')
```

```
In [633]: movies=movies[['Title', 'Plot', 'Director', 'Genre']]
```

```
In [634]: movies.head()
```

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Out[634]:
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	Title	Plot	Director	Genre
0	Code Name: K.O.Z.	A look at the 17-25 December 2013 corruption s...	Celal Çimen	Crime, Mystery
1	Saving Christmas	Kirk is enjoying the annual Christmas party ex...	Darren Doane	Comedy, Family
2	Superbabies: Baby Geniuses 2	A group of smart-talking toddlers find themsel...	Bob Clark	Comedy, Family, Sci-Fi
3	Daniel der Zauberer	Evil assassins want to kill Daniel Kublbock, t...	Ulli Lommel	Comedy, Crime, Fantasy
4	Manos: The Hands of Fate	A family gets lost on the road and stumbles up...	Harold P. Warren	Horror

```
In [635]: movies['Plot'].fillna(" ", inplace=True)
```

```
In [636]: #Create a map of movie tilte to its index
AllIndex = pd.Series(movies.index, index=movies['Title']).drop_duplicates()
```

```
In [637]: #Get the TFIDF matrix fittend on movie's Plot
matrix = tfidf.fit_transform(movies['Plot'])
```

```
In [638]: #Use cosine similarity measure to calculate the similarity between the movies
#Since you have used the TF-IDF vectorizer, calculating the dot product will directly give you
#the cosine similarity score
cosine_similarity = linear_kernel(matrix,matrix)
```

```
In [639]: #Function to give top 10 movie recommendations

def recommendations(title,cs=cosine_similarity):
    """
    :param: title - title of the movie to find similar movies
    :param: cs - Cosine similarty matrix
    """
    index=AllIndex[title]

    #Get all similarity scores for the given movie
    scores = list(enumerate(cs[index]))
    scores = sorted(scores, key=lambda x: x[1], reverse=True)

    # Get the scores of the 10 most similar movies
    scores = scores[1:11]

    # Get the movie indices for top 10
    movie_indices = [i[0] for i in scores]

    return movies['Title'].iloc[movie_indices]
recommendations('The Dark Knight Rises')
```

```
Out[639]: 1960          Batman
103          The Dark Knight
4674          Yes
1218          The Siege
2239          Godsend
2458          Elizabeth
3749  The Wind That Shakes the Barley
1064          Batman Returns
2122          George and the Dragon
3249          Flipped
Name: Title, dtype: object
```

```
In [640]: recommendations('The Godfather: Part II')
```

```
Out[640]: 898          The Aviator
1486      The Godfather: Part III
4377          The Purge
1556          Absolute Power
4348          The Betrayed
2172      Midnight in Paris
4245          Boyhood
1429      The Pursuit of Happyness
2247          Dead Man Down
1013      The Nutcracker in 3D
Name: Title, dtype: object
```