```
In [1]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

In [2]: # importing data to use
 df= pd.read_csv('tmdb.movies.csv')
 df

Out[2]:

	Unnamed: 0	genre_ids	id	original_language	original_title	popularity	release_date	
0	0	[12, 14, 10751]	12444	en	Harry Potter and the Deathly Hallows: Part 1	33.533	2010-11-19	Harry P the Hallov
1	1	[14, 12, 16, 10751]	10191	en	How to Train Your Dragon	28.734	2010-03-26	How to T
2	2	[12, 28, 878]	10138	en	Iron Man 2	28.515	2010-05-07	Iro
3	3	[16, 35, 10751]	862	en	Toy Story	28.005	1995-11-22	
4	4	[28, 878, 12]	27205	en	Inception	27.920	2010-07-16	
26512	26512	[27, 18]	488143	en	Laboratory Conditions	0.600	2018-10-13	L; C
26513	26513	[18, 53]	485975	en	_EXHIBIT_84xxx_	0.600	2018-05-01	_EXHIBI7
26514	26514	[14, 28, 12]	381231	en	The Last One	0.600	2018-10-01	The
26515	26515	[10751, 12, 28]	366854	en	Trailer Made	0.600	2018-06-22	Tra
26516	26516	[53, 27]	309885	en	The Church	0.600	2018-10-05	Th

26517 rows × 10 columns

```
In [3]: #printing out first 5 rows
print(df.head())
```

```
Unnamed: 0
                         genre_ids
                                       id original_language \
0
                   [12, 14, 10751] 12444
            0
                                                         en
            1 [14, 12, 16, 10751] 10191
1
                                                         en
2
                     [12, 28, 878]
                                   10138
            2
                                                        en
                   [16, 35, 10751]
3
            3
                                      862
                                                        en
4
            4
                     [28, 878, 12] 27205
                                                         en
```

original_title	popularity	release_date	\
Harry Potter and the Deathly Hallows: Part 1	33.533	2010-11-19	
How to Train Your Dragon	28.734	2010-03-26	
Iron Man 2	28.515	2010-05-07	
Toy Story	28.005	1995-11-22	
Inception	27.920	2010-07-16	
	Harry Potter and the Deathly Hallows: Part 1 How to Train Your Dragon Iron Man 2 Toy Story	Harry Potter and the Deathly Hallows: Part 1 33.533 How to Train Your Dragon 28.734 Iron Man 2 28.515 Toy Story 28.005	How to Train Your Dragon 28.734 2010-03-26 Iron Man 2 28.515 2010-05-07 Toy Story 28.005 1995-11-22

	title	vote_average	vote_count
0	Harry Potter and the Deathly Hallows: Part 1	7.7	10788
1	How to Train Your Dragon	7.7	7610
2	Iron Man 2	6.8	12368
3	Toy Story	7.9	10174
4	Inception	8.3	22186

```
In [4]:
        #printing out last 5 rows
        print(df.tail())
               Unnamed: 0
                                                 id original_language
                                  genre_ids
        26512
                    26512
                                   [27, 18]
                                             488143
                                   [18, 53]
        26513
                    26513
                                             485975
                                                                   en
        26514
                    26514
                               [14, 28, 12]
                                             381231
                                                                   en
        26515
                     26515
                           [10751, 12, 28]
                                             366854
                                                                   en
        26516
                     26516
                                   [53, 27]
                                             309885
                                                                   en
                      original_title popularity release_date
                                                                                 title \
        26512
               Laboratory Conditions
                                              0.6
                                                    2018-10-13 Laboratory Conditions
                     _EXHIBIT_84xxx_
                                              0.6
                                                    2018-05-01
                                                                      _EXHIBIT_84xxx_
        26513
        26514
                        The Last One
                                              0.6
                                                    2018-10-01
                                                                         The Last One
        26515
                        Trailer Made
                                              0.6
                                                    2018-06-22
                                                                         Trailer Made
                           The Church
                                              0.6
                                                    2018-10-05
                                                                            The Church
        26516
               vote_average
                             vote_count
        26512
                        0.0
                                       1
                        0.0
        26513
                                       1
                        0.0
        26514
                                       1
        26515
                        0.0
                                       1
        26516
                        0.0
                                       1
In [5]:
        #getting statistical summary of the data
        print(df.describe())
                Unnamed: 0
                                        id
                                              popularity vote_average
                                                                          vote_count
               26517.00000
                             26517.000000
                                            26517.000000
                                                          26517.000000
                                                                        26517.000000
        count
        mean
               13258.00000
                            295050.153260
                                                3.130912
                                                              5.991281
                                                                          194.224837
        std
                7654.94288
                            153661.615648
                                                4.355229
                                                              1.852946
                                                                           960.961095
                   0.00000
                                 27.000000
                                                0.600000
                                                              0.000000
                                                                             1.000000
        min
        25%
                6629.00000
                            157851.000000
                                                0.600000
                                                              5.000000
                                                                             2.000000
        50%
               13258.00000 309581.000000
                                                1.374000
                                                              6.000000
                                                                             5.000000
        75%
               19887.00000 419542.000000
                                                3.694000
                                                              7.000000
                                                                            28.000000
        max
               26516.00000 608444.000000
                                               80.773000
                                                             10.000000
                                                                        22186.000000
In [6]:
        # printing out how many rows and columns are there
        print(df.shape)
        (26517, 10)
In [7]: #getting general information on the columns
        print(df.info())
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 26517 entries, 0 to 26516
        Data columns (total 10 columns):
                                Non-Null Count Dtype
         #
             Column
                                 -----
         0
             Unnamed: 0
                                 26517 non-null int64
                                 26517 non-null object
         1
             genre_ids
         2
                                 26517 non-null int64
         3
             original_language 26517 non-null
                                                 object
         4
             original title
                                 26517 non-null
                                                object
         5
             popularity
                                 26517 non-null float64
             release_date
         6
                                 26517 non-null object
         7
             title
                                 26517 non-null object
         8
             vote_average
                                 26517 non-null float64
         9
             vote count
                                 26517 non-null int64
        dtypes: float64(2), int64(3), object(5)
        memory usage: 2.0+ MB
        None
```

```
In [8]:
         #renaming first column
         df.rename(columns= {df.columns[0]:'Number'}, inplace= True)
 In [9]: #printing out first 5 rows again to confirm column name change
         print(df.head())
            Number
                               genre ids
                                             id original language
         0
                         [12, 14, 10751]
                                          12444
         1
                  1
                     [14, 12, 16, 10751]
                                          10191
                                                                en
                           [12, 28, 878]
         2
                  2
                                          10138
                                                                en
         3
                  3
                         [16, 35, 10751]
                                            862
                                                                en
         4
                  4
                           [28, 878, 12]
                                          27205
                                                                en
                                           original_title popularity release_date
         0
            Harry Potter and the Deathly Hallows: Part 1
                                                                33.533
                                                                         2010-11-19
         1
                                 How to Train Your Dragon
                                                                28.734
                                                                         2010-03-26
         2
                                               Iron Man 2
                                                                28.515
                                                                         2010-05-07
         3
                                                Tov Story
                                                                28.005
                                                                         1995-11-22
         4
                                                Inception
                                                                27.920
                                                                         2010-07-16
                                                     title vote_average vote_count
            Harry Potter and the Deathly Hallows: Part 1
                                                                    7.7
                                                                               10788
         1
                                 How to Train Your Dragon
                                                                     7.7
                                                                                7610
         2
                                                Iron Man 2
                                                                     6.8
                                                                               12368
         3
                                                Toy Story
                                                                     7.9
                                                                               10174
         4
                                                Inception
                                                                     8.3
                                                                                22186
         #checking for duplicates
In [10]:
         duplicates= df[df.duplicated()]
         print(len(duplicates))
         0
In [11]: # deleting the genre_ids column
         df.drop('genre_ids', axis=1, inplace=True)
In [12]: #printing first 5 rows to confirm column is deleted
         print(df.head())
            Number
                        id original_language
         0
                 0 12444
                                          en
         1
                 1
                    10191
                                          en
         2
                 2 10138
                                          en
         3
                  3
                      862
                                          en
         4
                  4
                    27205
                                          en
                                           original title popularity release date
            Harry Potter and the Deathly Hallows: Part 1
                                                                33.533
                                                                         2010-11-19
         1
                                 How to Train Your Dragon
                                                                28.734
                                                                         2010-03-26
         2
                                                Iron Man 2
                                                                28.515
                                                                         2010-05-07
         3
                                                Toy Story
                                                                28.005
                                                                         1995-11-22
         4
                                                Inception
                                                                27.920
                                                                         2010-07-16
                                                     title vote_average vote_count
            Harry Potter and the Deathly Hallows: Part 1
                                                                                10788
                                                                     7.7
         1
                                 How to Train Your Dragon
                                                                     7.7
                                                                                7610
         2
                                                Iron Man 2
                                                                     6.8
                                                                               12368
         3
                                                Toy Story
                                                                     7.9
                                                                               10174
         4
                                                Inception
                                                                     8.3
                                                                               22186
In [13]: #checking current number of columns
         print(df.shape)
         (26517, 9)
```

```
In [14]: #checking for extraeous values
for col in df.columns:
    print(col, '\n', df[col].value_counts(normalize=True).head(), '\n\n')
```

```
Number
Number
         0.000038
0
17675
         0.000038
         0.000038
17685
17684
         0.000038
         0.000038
17683
Name: proportion, dtype: float64
id
 id
380718
          0.000113
292086
          0.000113
402448
          0.000113
192137
          0.000113
514791
          0.000113
Name: proportion, dtype: float64
original_language
original_language
      0.878342
en
fr
      0.019120
      0.017159
es
      0.011238
ru
jа
      0.009994
Name: proportion, dtype: float64
original_title
original_title
Eden
                 0.000264
Home
                 0.000226
Legend
                 0.000189
Aftermath
                 0.000189
Truth or Dare
                 0.000189
Name: proportion, dtype: float64
popularity
popularity
0.600
         0.265377
1.400
         0.024475
0.840
         0.022137
0.624
         0.003922
0.625
         0.003469
Name: proportion, dtype: float64
release_date
 release_date
2010-01-01
              0.010144
2011-01-01
             0.007542
2014-01-01
              0.005845
2012-01-01
              0.005845
2013-01-01
              0.005468
Name: proportion, dtype: float64
title
title
             0.000264
Eden
Home
             0.000264
Lucky
             0.000189
Legend
             0.000189
```

Aftermath

0.000189

Name: proportion, dtype: float64

```
vote_average

vote_average

6.0 0.073161

7.0 0.058830

5.0 0.056040

10.0 0.047215

8.0 0.046423
```

Name: proportion, dtype: float64

vote_count vote_count 1 0.246672 2 0.114794 3 0.066259 4 0.050798 5 0.036543

Name: proportion, dtype: float64

```
In [15]: df.dtypes
```

```
Out[15]: Number
                                 int64
                                 int64
         original_language
                                object
         original_title
                                object
         popularity
                               float64
                                object
         release_date
         title
                                object
         vote_average
                               float64
                                 int64
         vote_count
         dtype: object
```

In [16]: df.isna().sum() #Checking for null values

```
Out[16]: Number
                               0
         id
                               0
         original_language
                               0
         original title
                               0
         popularity
                               0
         release_date
                               0
                               0
         title
                               0
         vote_average
         vote_count
                               0
         dtype: int64
```

```
In [17]: df.iloc[1000:1050,2] # sourcing rows from 1000 to 1500 for the original_language column
Out[17]: 1000
                  en
         1001
                  en
         1002
                  en
         1003
                  en
         1004
                  ru
         1005
                  en
         1006
                  nl
         1007
                  en
         1008
                  en
         1009
                  en
         1010
                  en
         1011
                  no
         1012
                  en
         1013
                  en
         1014
                  en
         1015
                  en
         1016
                  en
         1017
                  en
         1018
                  en
         1019
                  de
         1020
                  en
         1021
                  en
         1022
                  en
         1023
                  et
         1024
                  en
         1025
                  en
         1026
                  en
         1027
                  it
         1028
                  da
         1029
                  en
         1030
                  en
         1031
                  en
         1032
                  no
         1033
                  de
         1034
                  en
         1035
                  en
         1036
                  nl
         1037
                  tl
         1038
                  es
         1039
                  en
         1040
                  en
         1041
                  en
         1042
                  fr
         1043
                  en
         1044
                  en
         1045
                  lt
         1046
                  en
         1047
                  fr
         1048
                  en
         1049
         Name: original_language, dtype: object
```

```
In [18]: #how many times each Language appears in the data set
    df['original_language'].value_counts()

Out[18]: original_language
    en 23291
        fr 507
        es 455
        ru 298
```

ja 265
bo 1 si 1 sl 1 hz 1

Name: count, Length: 76, dtype: int64

In [19]: ne_languagecount = (df['original_language'] == 'ne').sum() # how many times ne Language
ne_languagecount

Out[19]: 2

In [20]: sampled_df = df.sample(frac=0.2) #creating a sample data frame from the original
 sampled_df

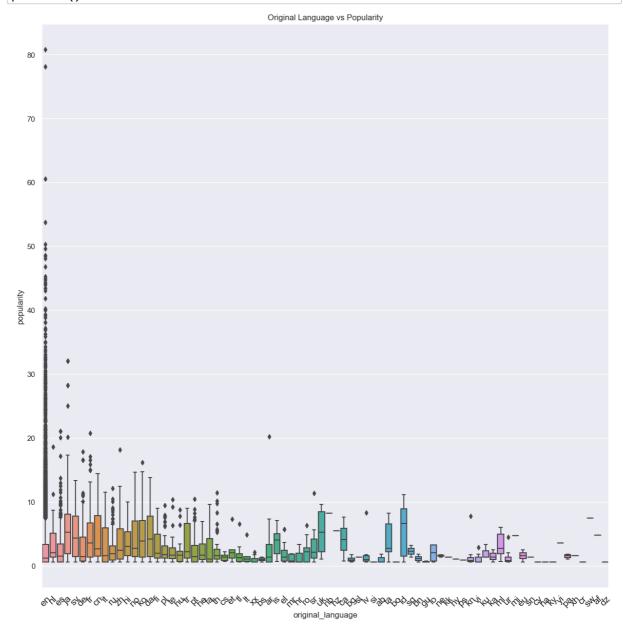
Out[20]:

	Number	id	original_language	original_title	popularity	release_date	title	vote_average	
20107	20107	521317	en	Норе	0.600	2016-03-07	Норе	7.0	
10347	10347	245230	en	l Am Britney Jean	0.600	2013-12-22	l Am Britney Jean	6.8	
26082	26082	534169	en	Penny Palabras	0.748	2018-05-19	Penny Palabras	8.0	
20277	20277	423342	en	Feeding Time	0.600	2016-10-21	Feeding Time	6.0	
11217	11217	243935	en	Rob the Mob	9.175	2014-03-21	Rob the Mob	6.3	
								•••	
22308	22308	456101	en	Aaron's Blood	1.265	2017-06-02	Aaron's Blood	4.8	
10264	10264	439755	en	Wake-up Juice	0.600	2013-02-05	Wake-up Juice	7.0	
11011	11011	186242	en	Re- Emerging: The Jews of Nigeria	0.600	2013-05-17	Re- Emerging: The Jews of Nigeria	0.5	
18467	18467	325186	en	Smothered	2.198	2016-03-29	Smothered	3.3	
7379	7379	135847	en	The Transmission	0.600	2012-09-21	The Transmission	6.5	
5303 rows × 9 columns									

5303 rows × 9 columns

localhost:8888/notebooks/Data Science Project.ipynb

```
In [25]: #creating box plot for language and popularity
    plt.figure(figsize=(15, 15)) #setting desired figure size
    sns.boxplot(x='original_language', y='popularity', data=df)
    plt.title('Original Language vs Popularity')
    plt.xticks(rotation=45, fontsize= 14) # rotate x values at an angle for them to stop be
    plt.show()
```



In conclussion, the movies created in English are most popular followed by japanese. Spanish and french movies follow in close range.

```
In [51]: origlanguage_frequency = df['original_language'].value_counts()
    df['origlanguage_frequency'] = df['original_language'].map(origlanguage_frequency)
    df.head()
```

Out[51]:

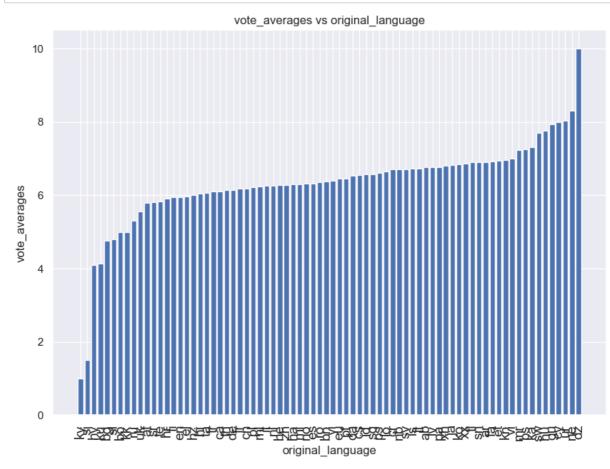
	Number	id	original_language	original_title	popularity	release_date	title	vote_average	vote_cc
0	0	12444	en	Harry Potter and the Deathly Hallows: Part 1	33.533	2010-11-19	Harry Potter and the Deathly Hallows: Part 1	7.7	10
1	1	10191	en	How to Train Your Dragon	28.734	2010-03-26	How to Train Your Dragon	7.7	7
2	2	10138	en	Iron Man 2	28.515	2010-05-07	Iron Man 2	6.8	12
3	3	862	en	Toy Story	28.005	1995-11-22	Toy Story	7.9	10
4	4	27205	en	Inception	27.920	2010-07-16	Inception	8.3	22
4									•

Movies made in English are by far more frequently made than movies in any other language.

```
In [28]: plt.style.available
Out[28]: ['Solarize_Light2',
           '_classic_test_patch',
'_mpl-gallery',
           '_mpl-gallery-nogrid',
           'bmh',
           'classic',
           'dark_background',
           'fast',
           'fivethirtyeight',
           'ggplot',
           'grayscale',
           'seaborn-v0_8',
           'seaborn-v0_8-bright',
           'seaborn-v0 8-colorblind',
           'seaborn-v0_8-dark',
           'seaborn-v0_8-dark-palette',
           'seaborn-v0_8-darkgrid',
           'seaborn-v0_8-deep',
           'seaborn-v0_8-muted',
           'seaborn-v0_8-notebook',
           'seaborn-v0 8-paper',
           'seaborn-v0_8-pastel',
           'seaborn-v0_8-poster',
           'seaborn-v0_8-talk',
           'seaborn-v0_8-ticks',
           'seaborn-v0_8-white',
           'seaborn-v0_8-whitegrid',
           'tableau-colorblind10']
```

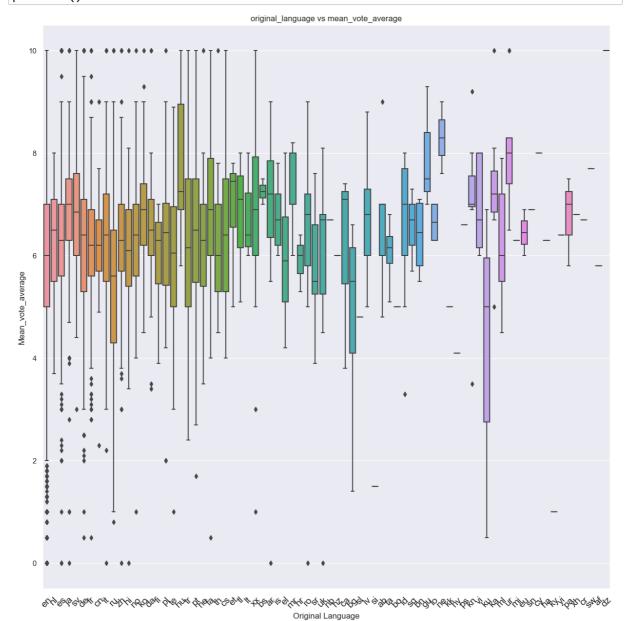
```
In [40]: #plotting according the the mean of average votes for original_languages
   vote_averages = df.groupby('original_language')['vote_average'].mean().reset_index()
   vote_averages = vote_averages.sort_values(by='vote_average')
   plt.figure(figsize=(10, 7))

plt.bar(vote_averages['original_language'], vote_averages['vote_average'])
   plt.xlabel('original_language')
   plt.ylabel('vote_averages')
   plt.title('vote_averages vs original_language')
   plt.xticks(rotation=90, fontsize=14)
   plt.show()
```

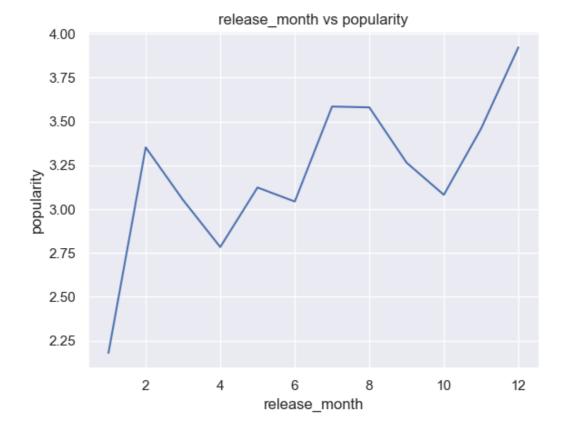


However, this may not be the most accurate way foward as the mean vote_averages tends to be bigger for languages that appear few times and smaller for languages that appear the more often.

```
In [31]: # Plot for vote_average mean and Language
   plt.figure(figsize=(15,15))
    sns.boxplot(x='original_language',y='vote_average', data=df)
   plt.xlabel('Original Language')
   plt.ylabel('Mean_vote_average')
   plt.title('original_language vs mean_vote_average')
   plt.xticks(rotation=45, fontsize=14)
   plt.show()
```

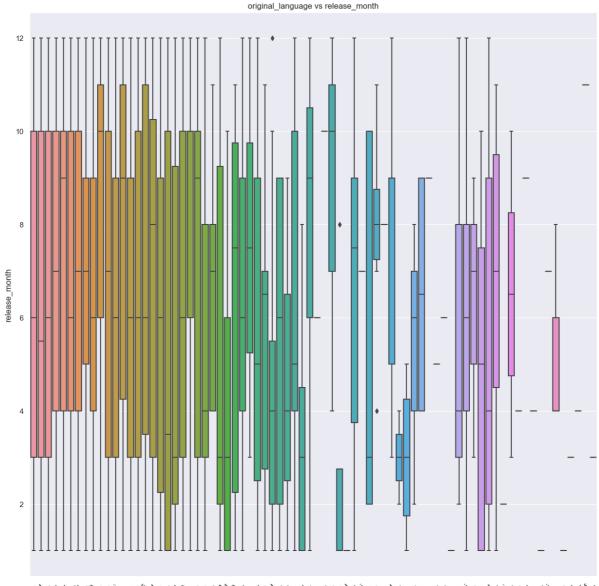


```
In [32]: df['release_date'] = pd.to_datetime(df['release_date']) # ensuring release_date is in def['release_month'] = df['release_date'].dt.month # grouping the release dates by month release_month_groups = df.groupby('release_month')
monthly_popularity = release_month_groups['popularity'].mean() # finding monthly mean for the finding to see trend
plt.plot(monthly_popularity.index, monthly_popularity.values)
plt.xlabel('release_month')
plt.ylabel('popularity')
plt.title('release_month vs popularity')
plt.show()
```



```
In [35]: #making sure the release date is in the datetime format
df['release_date'] = pd.to_datetime(df['release_date'])

#changing release_date to monthly format.
df['release_month'] = df['release_date'].dt.month
plt.figure(figsize=(15,15))
sns.boxplot(x='original_language',y='release_month', data=df)
plt.xlabel('Original_Language')
plt.ylabel('release_month')
plt.title('original_language vs release_month')
plt.xticks(rotation=45, fontsize=14)
plt.show()
```



Wiking Taking Chard Char

For most languages, they peak on the 12th month, December, making it the best time of the year to release movies.