

IMDB Case Study using Pandas

```
In [ ]: import numpy as np
import pandas as pd
movies=pd.read_csv('movies.csv')
directors=pd.read_csv('directors.csv')
```

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

Out[]:	Unnamed: 0	id	budget	popularity	revenue	title	vote_average	vote_count	director_id	year	month	day
0	0	43597	237000000	150	2787965087	Avatar	7.2	11800	4762	2009	Dec	Thursday
1	1	43598	300000000	139	961000000	Pirates of the Caribbean: At World's End	6.9	4500	4763	2007	May	Saturday
2	2	43599	245000000	107	880674609	Spectre	6.3	4466	4764	2015	Oct	Monday
3	3	43600	250000000	112	1084939099	The Dark Knight Rises	7.6	9106	4765	2012	Jul	Monday
4	5	43602	258000000	115	890871626	Spider-Man 3	5.9	3576	4767	2007	May	Tuesday

```
In [ ]: directors.head()
```

Out[]:	Unnamed: 0	director_name	id	gender
0	0	James Cameron	4762	Male
1	1	Gore Verbinski	4763	Male
2	2	Sam Mendes	4764	Male
3	3	Christopher Nolan	4765	Male
4	4	Andrew Stanton	4766	Male

```
In [ ]: df=movies.merge(directors,left_on='director_id',right_on='id',how='inner')
```

```
In [ ]: df.head()
```

Out[]:	Unnamed: 0_x	id_x	budget	popularity	revenue	title	vote_average	vote_count	director_id	year	month	day
0	0	43597	237000000	150	2787965087	Avatar	7.2	11800	4762	2009	Dec	Thursday
1	1	43598	300000000	139	961000000	Pirates of the Caribbean: At World's End	6.9	4500	4763	2007	May	Saturday
2	2	43599	245000000	107	880674609	Spectre	6.3	4466	4764	2015	Oct	Monday
3	3	43600	250000000	112	1084939099	The Dark Knight Rises	7.6	9106	4765	2012	Jul	Monday
4	5	43602	258000000	115	890871626	Spider-Man 3	5.9	3576	4767	2007	May	Tuesday

```
In [ ]: df.drop(['director_id','id_y','id_x'],axis=1,inplace=True)
```

```
In [ ]: df.head()
```

Out[]:

	Unnamed: 0_x	budget	popularity	revenue	title	vote_average	vote_count	year	month	day	Unnamed: 0_y	director
0	0	237000000	150	2787965087	Avatar	7.2	11800	2009	Dec	Thursday	0	
1	1	300000000	139	961000000	Pirates of the Caribbean: At World's End	6.9	4500	2007	May	Saturday	1	Gore Verbinski
2	2	245000000	107	880674609	Spectre	6.3	4466	2015	Oct	Monday	2	Sam Mendes
3	3	250000000	112	1084939099	The Dark Knight Rises	7.6	9106	2012	Jul	Monday	3	Christopher Nolan
4	5	258000000	115	890871626	Spider-Man 3	5.9	3576	2007	May	Tuesday	5	Sam Raimi

In []:

```
df['budget']=df['budget']/10000000
```

In []:

```
df.head()
```

Out[]:

	Unnamed: 0_x	budget	popularity	revenue	title	vote_average	vote_count	year	month	day	Unnamed: 0_y	director_name
0	0	23.7	150	2787965087	Avatar	7.2	11800	2009	Dec	Thursday	0	James Cameron
1	1	30.0	139	961000000	Pirates of the Caribbean: At World's End	6.9	4500	2007	May	Saturday	1	Gore Verbinski
2	2	24.5	107	880674609	Spectre	6.3	4466	2015	Oct	Monday	2	Sam Mendes
3	3	25.0	112	1084939099	The Dark Knight Rises	7.6	9106	2012	Jul	Monday	3	Christopher Nolan
4	5	25.8	115	890871626	Spider-Man 3	5.9	3576	2007	May	Tuesday	5	Sam Raimi

In []:

```
df['revenue']=np.round(df['revenue']/1000000,2)
```

In []:

```
df.head()
```

Out[]:

	Unnamed: 0_x	budget	popularity	revenue	title	vote_average	vote_count	year	month	day	Unnamed: 0_y	director_name
0	0	23.7	150	2787.97	Avatar	7.2	11800	2009	Dec	Thursday	0	James Cameron
1	1	30.0	139	961.00	Pirates of the Caribbean: At World's End	6.9	4500	2007	May	Saturday	1	Gore Verbinski
2	2	24.5	107	880.67	Spectre	6.3	4466	2015	Oct	Monday	2	Sam Mendes
3	3	25.0	112	1084.94	The Dark Knight Rises	7.6	9106	2012	Jul	Monday	3	Christopher Nolan
4	5	25.8	115	890.87	Spider-Man 3	5.9	3576	2007	May	Tuesday	5	Sam Raimi

In []:

```
df.shape
```

Out[]:

```
(1465, 13)
```

In []:

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1465 entries, 0 to 1464
Data columns (total 13 columns):
#   Column              Non-Null Count  Dtype
---  -
0   Unnamed: 0_x         1465 non-null   int64
1   budget               1465 non-null   float64
2   popularity            1465 non-null   int64
3   revenue              1465 non-null   float64
4   title                 1465 non-null   object
5   vote_average          1465 non-null   float64
6   vote_count           1465 non-null   int64
7   year                  1465 non-null   int64
8   month                 1465 non-null   object
9   day                   1465 non-null   object
10  Unnamed: 0_y          1465 non-null   int64
11  director_name         1465 non-null   object
12  gender                1341 non-null   object
dtypes: float64(3), int64(5), object(5)
memory usage: 148.9+ KB
```

```
In [ ]: df.describe(include='number')
```

```
Out[ ]:
```

	Unnamed: 0_x	budget	popularity	revenue	vote_average	vote_count	year	Unnamed: 0_y
count	1465.000000	1465.000000	1465.000000	1465.000000	1465.000000	1465.000000	1465.000000	1465.000000
mean	1627.391809	4.802295	30.855973	143.253952	6.368191	1146.396587	2002.615017	278.192491
std	1187.182894	4.935541	34.845214	206.491831	0.818033	1578.077438	8.680141	258.059631
min	0.000000	0.000000	0.000000	0.000000	3.000000	1.000000	1976.000000	0.000000
25%	639.000000	1.400000	11.000000	17.380000	5.900000	216.000000	1998.000000	83.000000
50%	1425.000000	3.300000	23.000000	75.780000	6.400000	571.000000	2004.000000	202.000000
75%	2393.000000	6.600000	41.000000	179.250000	6.900000	1387.000000	2009.000000	417.000000
max	4768.000000	38.000000	724.000000	2787.970000	8.300000	13752.000000	2016.000000	1442.000000

```
In [ ]: df.describe(include='object')
```

```
Out[ ]:
```

	title	month	day	director_name	gender
count	1465	1465	1465	1465	1341
unique	1465	12	7	199	2
top	El Mariachi	Dec	Friday	Steven Spielberg	Male
freq	1	193	654	26	1309

```
In [ ]: #find the gender wise vote_Average
df.groupby('gender')['vote_average'].mean()
```

```
Out[ ]:
```

	vote_average
gender	
Female	6.262500
Male	6.381742

dtype: float64

```
In [ ]: df.groupby('director_name')['title'].count()
```

title

199 rows × 1 columns

```
In [ ]: df.groupby('director_name')['title'].count().sort_values(ascending=False)
```

title

199 rows × 1 columns

```
In [ ]: df.groupby('director_name').get_group('Steven Spielberg')
```

	Unnamed: 0_x	budget	popularity	revenue	title	vote_average	vote_count	year	month	day	Unnamed: 0_y	director
37	53	18.50	75	786.64	Indiana Jones and the Kingdom of the Crystal Skull	5.7	2495	2008	May	Wednesday	37	Steven Spielberg
105	175	14.00	44	183.35	The BFG	6.0	1000	2016	Jun	Wednesday	37	Steven Spielberg
110	185	13.20	48	591.74	War of the Worlds	6.2	2322	2005	Jun	Tuesday	37	Steven Spielberg
114	190	13.00	89	371.94	The Adventures of Tintin	6.7	2061	2011	Oct	Tuesday	37	Steven Spielberg
166	275	10.20	65	358.37	Minority Report	7.1	2608	2002	Jun	Thursday	37	Steven Spielberg
219	363	10.00	34	235.93	A.I. Artificial Intelligence	6.8	1974	2001	Jun	Friday	37	Steven Spielberg
					The Lost							

296	508	7.30	2	229.07	World: Jurassic Park	6.2	2487	1997	May	Friday	37	Sf
304	521	6.00	57	219.42	The Terminal	7.0	1910	2004	Jun	Thursday	37	Sf
309	528	7.00	29	130.36	Munich	6.9	696	2005	Dec	Thursday	37	Sf
333	572	7.00	33	300.85	Hook	6.6	1532	1991	Dec	Wednesday	37	Sf
342	585	6.60	29	177.58	War Horse	7.0	992	2011	Dec	Sunday	37	Sf
360	628	7.00	76	481.84	Saving Private Ryan	7.9	5048	1998	Jul	Friday	37	Sf
363	633	6.50	36	275.29	Lincoln	6.7	1429	2012	Nov	Friday	37	Sf
390	675	6.30	40	920.10	Jurassic Park	7.6	4856	1993	Jun	Friday	37	Sf
495	883	5.20	73	352.11	Catch Me If You Can	7.7	3795	2002	Dec	Wednesday	37	Sf
552	1006	4.80	80	474.17	Indiana Jones and the Last Crusade	7.6	3152	1989	May	Wednesday	37	Sf
635	1187	4.00	48	165.48	Bridge of Spies	7.2	2583	2015	Oct	Thursday	37	Sf
647	1211	3.60	3	74.00	Amistad	6.8	316	1997	Dec	Wednesday	37	Sf
766	1510	3.50	10	31.76	1941	5.6	143	1979	Dec	Thursday	37	Sf
834	1697	2.80	66	333.00	Indiana Jones and the Temple of Doom	7.1	2781	1984	May	Wednesday	37	Sf
901	1818	2.20	104	321.37	Schindler's List	8.3	4329	1993	Nov	Monday	37	Sf
995	2085	1.80	68	389.93	Raiders of the Lost Ark	7.7	3854	1981	Jun	Friday	37	Sf
997	2087	2.00	52	303.79	Close Encounters of the Third Kind	7.2	1098	1977	Nov	Wednesday	37	Sf
1135	2520	1.50	17	146.29	The Color Purple	7.7	338	1985	Dec	Wednesday	37	Sf
1239	2967	1.05	56	792.91	E.T. the Extra-Terrestrial	7.3	3269	1982	Apr	Saturday	37	Sf
1252	3006	1.00	12	29.45	Twilight Zone: The Movie	6.2	161	1983	Jun	Friday	37	Sf

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In []:

df.groupby('director_name').groups

```
Out[ ]: {'Adam McKay': [176, 323, 366, 505, 839, 916], 'Adam Shankman': [265, 300, 350, 404, 458, 843, 999, 1231], 'Alejandra González Iñárritu': [106, 749, 1015, 1034, 1077, 1405], 'Alex Proyas': [95, 159, 514, 671, 873], 'Alexander Payne': [793, 1006, 1101, 1211, 1281], 'Andrew Adamson': [11, 43, 328, 501, 947], 'Andrew Niccol': [533, 603, 701, 722, 1439], 'Andrzej Bartkowiak': [349, 549, 754, 911, 924], 'Andy Fickman': [517, 681, 909, 926, 973, 1023], 'Andy Tennant': [314, 320, 464, 593, 676, 885], 'Ang Lee': [99, 134, 748, 840, 1089, 1110, 1132, 1184], 'Anne Fletcher': [610, 650, 736, 789, 1206], 'Antoine Fuqua': [310, 338, 424, 467, 576, 808, 818, 1105], 'Atom Egoyan': [946, 1128, 1164, 1194, 1347, 1416], 'Barry Levinson': [313, 319, 471, 594, 878, 898, 1013, 1037, 1082, 1143, 1185, 1345, 1378], 'Barry Sonnenfeld': [13, 48, 90, 205, 591, 778, 783], 'Ben Stiller': [209, 212, 547, 562, 850], 'Bill Condon': [102, 307, 902, 1233, 1381], 'Bobby Farrelly': [352, 356, 481, 498, 624, 630, 654, 806, 928, 972, 1111], 'Brad Anderson': [1163, 1197, 1350, 1419, 1430], 'Brett Ratner': [24, 39, 188, 207, 238, 292, 405, 456, 920], 'Brian De Palma': [228, 255, 318, 439, 747, 905, 919, 1088, 1232, 1261, 1317, 1354], 'Brian Helgeland': [512, 607, 623, 742, 933], 'Brian Levant': [418, 449, 568, 761, 860, 1003], 'Brian Robbins': [416, 441, 669, 962, 988, 1115], 'Bryan Singer': [6, 32, 33, 44, 122, 216, 297, 1326], 'Cameron Crowe': [335, 434, 488, 503, 513, 698], 'Catherine Hardwicke': [602, 695, 724, 937, 1406, 1412], 'Chris Columbus': [117, 167, 204, 218, 229, 509, 656, 897, 996, 1086, 1129], 'Chris Weitz': [17, 500, 794, 869, 1202, 1267], 'Christopher Nolan': [3, 45, 58, 59, 74, 565, 641, 1341], 'Chuck Russell': [177, 410, 657, 1069, 1097, 1339], 'Clint Eastwood': [369, 426, 447, 482, 490, 520, 530, 535, 645, 727, 731, 786, 787, 899, 974, 986, 1167, 1190, 1313], 'Curtis Hanson': [494, 579, 606, 711, 733, 1057, 1310], 'Danny Boyle': [527, 668, 1083, 1085, 1126, 1168, 1287, 1385], 'Darren Aronofsky': [113, 751, 1187, 1328, 1363, 1458], 'Darren Lynn Bousman': [1241, 1243, 1283, 1338, 1440], 'David Ayer': [50, 273, 741, 1024, 1146, 1407], 'David Cronenberg': [541, 767, 994, 1055, 1254, 1268, 1334], 'David Fincher': [62, 213, 253, 383, 398, 478, 522, 555, 618, 785], 'David Gordon Green': [543, 862, 884, 927, 1376, 1418, 1432, 1459], 'David Koepp': [443, 644, 735, 1041, 1209], 'David Lynch': [583, 1161, 1264, 1340, 1456], 'David O. Russell': [422, 556, 609, 896, 982, 989, 1229, 1304], 'David R. Ellis': [582, 634, 756, 888, 934], 'David Zucker': [569, 619, 965, 1052, 1175], 'Dennis Dugan': [217, 260, 267, 293, 303, 718, 780, 977, 1247], 'Donald Petrie': [427, 507, 570, 649, 858, 894, 1106, 1331], 'Doug Liman': [52, 148, 251, 399, 544, 1318, 1451], 'Edward Zwick': [92, 182, 346, 566, 791, 819, 825], 'F. Gary Gray': [308, 402, 491, 523, 697, 833, 1272, 1380], 'Francis Ford Coppola': [487, 559, 622, 646, 772, 1076, 1155, 1253, 1312], 'Francis Lawrence': [63, 72, 109, 120, 679], 'Frank Coraci': [157, 249, 275, 451, 577, 599, 963], 'Frank Oz': [193, 355, 473, 580, 712, 813, 987], 'Garry Marshall': [329, 496, 528, 571, 784, 893, 1029, 1169], 'Gary Fleder': [518, 667, 689, 867, 981, 1165], 'Gary Winick': [258, 797, 798, 804, 1454], 'Gavin O'Connor': [820, 841, 939, 953, 1444], 'George A. Romero': [250, 1066, 1096, 1278, 1367, 1396], 'George Clooney': [343, 450, 831, 966, 1302], 'George Miller': [78, 103, 233, 287, 1250, 1403, 1450], 'Gore Verbinski': [1, 8, 9, 107, 119, 633, 1040], 'Guillermo del Toro': [35, 252, 419, 486, 1118], 'Gus Van Sant': [595, 1018, 1027, 1159, 1240, 1311, 1398], 'Guy Ritchie': [124, 215, 312, 1093, 1225, 1269, 1420], 'Harold Ramis': [425, 431, 558, 586, 788, 1137, 1166, 1325], 'Ivan Reitman': [274, 643, 816, 883, 910, 935, 1134, 1242], 'James Cameron': [0, 19, 170, 173, 344, 1100, 1320], 'James Ivory': [1125, 1152, 1180, 1291, 1293, 1390, 1397], 'James Mangold': [140, 141, 557, 560, 829, 845, 958, 1145], 'James Wan': [30, 617, 1002, 1047, 1337, 1417, 1424], 'Jan de Bont': [155, 224, 231, 270, 781], 'Jason Friedberg': [812, 1010, 1012, 1014, 1036], 'Jason Reitman': [792, 1092, 1213, 1295, 1299], 'Jaume Collet-Serra': [516, 540, 640, 725, 1011, 1189], 'Jay Roach': [195, 359, 389, 397, 461, 703, 859, 1072], 'Jean-Pierre Jeunet': [423, 485, 605, 664, 765], 'Joe Dante': [284, 525, 638, 1226, 1298, 1428], 'Joe Wright': [85, 432, 553, 803, 814, 855], 'Joel Coen': [428, 670, 691, 707, 721, 889, 906, 980, 1157, 1238, 1305], 'Joel Schumacher': [128, 184, 348, 484, 572, 614, 652, 764, 876, 886, 1108, 1230, 1280], 'John Carpenter': [537, 663, 686, 861, 938, 1028, 1080, 1102, 1329, 1371], 'John Glen': [601, 642, 801, 847, 864], 'John Landis': [524, 868, 1276, 1384, 1435], 'John Madden': [457, 882, 1020, 1249, 1257], 'John McTiernan': [127, 214, 244, 351, 534, 563, 648, 782, 838, 1074], 'John Singleton': [294, 489, 732, 796, 1120, 1173, 1316], 'John Whitesell': [499, 632, 763, 1119, 1148], 'John Woo': [131, 142, 264, 371, 420, 675, 1182], 'Jon Favreau': [46, 54, 55, 382, 759, 1346], 'Jon M. Chu': [100, 225, 810, 1099, 1186], 'Jon Turteltaub': [64, 180, 372, 480, 760, 846, 1171], 'Jonathan Demme': [277, 493, 1000, 1123, 1215], 'Jonathan Liebesman': [81, 143, 339, 1117, 1301], 'Judd Apatow': [321, 710, 717, 865, 881], 'Justin Lin': [38, 123, 246, 1437, 1447], 'Kenneth Branagh': [80, 197, 421, 879, 1094, 1277, 1288], 'Kenny Ortega': [412, 852, 1228, 1315, 1365], 'Kevin Reynolds': [53, 502, 639, 1019, 1059], ...}
```

```
In [ ]: df.groupby('director_name').ngroups
```

```
Out[ ]: 199
```

```
In [ ]: #which director is more productive
final_data=df.groupby('director_name').aggregate({'year':['min','max'],'title':['count']})
```

```
In [ ]: final_data
```

Out[]:

		year		title
		min	max	count
director_name				
	Adam McKay	2004	2015	6
	Adam Shankman	2001	2012	8
	Alejandro González Iñárritu	2000	2015	6
	Alex Proyas	1994	2016	5
	Alexander Payne	1999	2013	5

	Wes Craven	1984	2011	10
	Wolfgang Petersen	1981	2006	7
	Woody Allen	1977	2013	18
	Zack Snyder	2004	2016	7
	Zhang Yimou	2002	2014	6

199 rows × 3 columns

In []:

```
final_data.sort_values(by=('title','count'),ascending=False,inplace=True)
```

In []:

```
final_data
```

Out[]:

	year		title
	min	max	count
director_name			
Steven Spielberg	1977	2016	26
Martin Scorsese	1976	2013	19
Clint Eastwood	1982	2014	19
Woody Allen	1977	2013	18
Robert Rodriguez	1992	2014	16
...
Stephen Daldry	2000	2014	5
Tom Tykwer	1998	2012	5
Tim Hill	1999	2011	5
Uwe Boll	2005	2013	5
Wayne Wang	1999	2011	5

199 rows × 3 columns

In []:

```
final_data['year_active']=final_data[('year','max')]-final_data[('year','min')]
final_data
```

Out[]:

	year	title	year_active	
	min	max	count	
director_name				
Steven Spielberg	1977	2016	26	39
Martin Scorsese	1976	2013	19	37
Clint Eastwood	1982	2014	19	32
Woody Allen	1977	2013	18	36
Robert Rodriguez	1992	2014	16	22
...
Stephen Daldry	2000	2014	5	14
Tom Tykwer	1998	2012	5	14
Tim Hill	1999	2011	5	12
Uwe Boll	2005	2013	5	8
Wayne Wang	1999	2011	5	12

199 rows × 4 columns

In []:

```
final_data['productivity']=final_data['title','count']/final_data['year_active']
final_data['productivity']=final_data['productivity']*100
final_data
```

Out[]:

	year	title	year_active	productivity	
	min	max	count		
director_name					
Steven Spielberg	1977	2016	26	39	66.666667
Martin Scorsese	1976	2013	19	37	51.351351
Clint Eastwood	1982	2014	19	32	59.375000
Woody Allen	1977	2013	18	36	50.000000
Robert Rodriguez	1992	2014	16	22	72.727273
...
Stephen Daldry	2000	2014	5	14	35.714286
Tom Tykwer	1998	2012	5	14	35.714286
Tim Hill	1999	2011	5	12	41.666667
Uwe Boll	2005	2013	5	8	62.500000
Wayne Wang	1999	2011	5	12	41.666667

199 rows × 5 columns

In []:

```
final_data.sort_values('productivity',ascending=False,inplace=True)
```

In []:

```
final_data
```


0

```
year    title  year_active  productivity
```

min	max	count
-----	-----	-------

director_name

Tyler Perry	2006	2013	9	7	128.571429
Jason Friedberg	2006	2010	5	4	125.000000
Shawn Levy	2002	2014	11	12	91.666667
Adam Shankman	2001	2012	8	11	72.727273
Robert Rodriguez	1992	2014	16	22	72.727273
...
Lawrence Kasdan	1985	2012	5	27	18.518519
Luc Besson	1985	2014	5	29	17.241379
Michael Apted	1980	2010	5	30	16.666667
Robert Redford	1980	2010	5	30	16.666667
Sidney Lumet	1976	2006	5	30	16.666667

199 rows × 5 columns

In []:

df

Out[]:

Unnamed:
0_x

budget popularity

Materials

title no

2000 104

count

1024 1025

46

Unm

med: 0_y

order to

0	0	23.7000	150	2787.97	Avatar	7.2	11800	2009	Dec	Thursday	0	J. Can
1	1	30.0000	139	961.00	Pirates of the Caribbean: At World's End	6.9	4500	2007	May	Saturday	1	Gore Vert
2	2	24.5000	107	880.67	Spectre	6.3	4466	2015	Oct	Monday	2	Sam Me
3	3	25.0000	112	1084.94	The Dark Knight Rises	7.6	9106	2012	Jul	Monday	3	Christo
4	5	25.8000	115	890.87	Spider-Man 3	5.9	3576	2007	May	Tuesday	5	Sam F
...
1460	4736	0.0000	3	0.32	The Last Waltz	7.9	64	1978	May	Monday	47	M. Scoll
1461	4743	0.0027	19	3.15	Clerks	7.4	755	1994	Sep	Tuesday	607	Kevin S
1462	4748	0.0000	7	0.00	Rampage	6.0	131	2009	Aug	Friday	386	Uwe
1463	4749	0.0000	3	0.00	Slacker	6.4	77	1990	Jul	Friday	773	Rik Linl
1464	4768	0.0220	14	2.04	El Mariachi	6.6	238	1992	Sep	Friday	335	R. Rodr

1465 rows x 13 columns

In []:

```
df.groupby('director name').get_group('Tyler Perry')
```

Out[]:

	Unnamed: 0_x	budget	popularity	revenue	title	vote_average	vote_count	year	month	day	Unnamed: 0_y	director_name
915	1843	0.0	5	0.00	A Madea Christmas	7.0	35	2013	Dec	Friday	792	Tyler Perry
991	2073	0.0	1	37.00	For Colored Girls	7.0	22	2010	Nov	Friday	792	Tyler Perry
1007	2107	2.0	2	60.07	Why Did I Get Married Too?	6.1	29	2010	Apr	Friday	792	Tyler Perry
1009	2110	0.0	7	0.00	Madea's Witness Protection	5.9	52	2012	Jun	Friday	792	Tyler Perry
1062	2287	0.0	2	0.00	I Can Do Bad All By Myself	6.0	40	2009	Sep	Friday	792	Tyler Perry
1098	2393	0.0	3	90.51	Madea Goes to Jail	6.4	52	2009	Feb	Monday	792	Tyler Perry
1140	2531	1.5	2	55.18	Why Did I Get Married?	6.1	33	2007	Oct	Friday	792	Tyler Perry
1172	2677	0.0	4	0.00	Good Deeds	6.2	45	2012	Feb	Thursday	792	Tyler Perry
1244	2984	0.6	5	57.23	Madea's Family Reunion	6.0	77	2006	Feb	Friday	792	Tyler Perry

Insights

- Tyler Perry is the most productive director
- Steven spielberg have director most number of movies

In []:

```
# 1.Which month number of movies have been directed
df.groupby('month')['title'].count()
```

Out[]:

title	
month	
Apr	90
Aug	111
Dec	193
Feb	104
Jan	60
Jul	127
Jun	133
Mar	99
May	116
Nov	117
Oct	149
Sep	166

dtype: int64

Insight 1:

December has the highest number of movies release(193) ,and least number of movies released in January,suggesting it is a peak month for the film industry,likely due to holiday seasons and increased audience engagement.

In []:

```
# 2.Is there any luck day for movie release
df.groupby('day')['title'].count()
```

Out[]:

	title
day	
	Friday 654
	Monday 67
	Saturday 47
	Sunday 46
	Thursday 277
	Tuesday 98
	Wednesday 276

dtype: int64

Insight 2:

Friday is the most popular day for movie release (654), indicating a strategic choice to maximize weekened audience engagement

In []:

```
# 3. Find out top 5 profitable movies
df[['title','revenue']].sort_values(by='revenue',ascending=False).head()
```

Out[]:

	title	revenue
0	Avatar	2787.97
19	Titanic	1845.03
30	Furious 7	1506.25
36	Transformers: Dark of the Moon	1123.75
199	The Lord of the Rings: The Return of the King	1118.89

Insights 3:

The top 5 Movies made a highest revenue top 1 is '**Avatar**' Generated **27billion** Revenue and top 5 is '**The Lord of the rings:the return of the king**' Generated **11billion** revenue

In []:

```
df
```

Out[]:

	Unnamed: 0_x	budget	popularity	revenue	title	vote_average	vote_count	year	month	day	Unnamed: 0_y	director_r
0	0	23.7000	150	2787.97	Avatar	7.2	11800	2009	Dec	Thursday	0	J: Can
1	1	30.0000	139	961.00	Pirates of the Caribbean: At World's End	6.9	4500	2007	May	Saturday	1	Gore Vert
2	2	24.5000	107	880.67	Spectre	6.3	4466	2015	Oct	Monday	2	Sam Me
3	3	25.0000	112	1084.94	The Dark Knight Rises	7.6	9106	2012	Jul	Monday	3	Christo
4	5	25.8000	115	890.87	Spider-Man 3	5.9	3576	2007	May	Tuesday	5	Sam f
...	
1460	4736	0.0000	3	0.32	The Last Waltz	7.9	64	1978	May	Monday	47	M Sco
1461	4743	0.0027	19	3.15	Clerks	7.4	755	1994	Sep	Tuesday	607	Kevin S
1462	4748	0.0000	7	0.00	Rampage	6.0	131	2009	Aug	Friday	386	Uwe
1463	4749	0.0000	3	0.00	Slacker	6.4	77	1990	Jul	Friday	773	Ric Lin
1464	4768	0.0220	14	2.04	El Mariachi	6.6	238	1992	Sep	Friday	335	R Rodr

1465 rows × 13 columns



In []:

```
#4.Director with highest vote_avg and popularity
df.groupby('director_name')[['vote_average','popularity']].max().head()
```

Out[]:

	vote_average	popularity
director_name		
Adam McKay	7.3	57
Adam Shankman	7.5	35
Alejandro González Iñárritu	7.6	100
Alex Proyas	7.3	95
Alexander Payne	7.4	40

Insight 4:

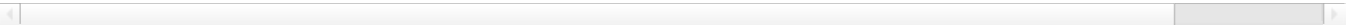
The top5 director with the highest vote average and popularity .Top1 director Adam Shankman vote average is 7.5 and popularity is 100 .

In []:

```
#5.Top3 budget movies
df.sort_values('budget',ascending=False).head(3)
```

Out[]:

	Unnamed: 0_x	budget	popularity	revenue	title	vote_average	vote_count	year	month	day	Unnamed: 0_y	director_n
12	17	38.0	135	1045.71	Pirates of the Caribbean: On Stranger Tides	6.4	4948	2011	May	Saturday	13	Rob Mar
1	1	30.0	139	961.00	Pirates of the Caribbean: At World's End	6.9	4500	2007	May	Saturday	1	Gore Verb
6	10	27.0	57	391.08	Superman Returns	5.4	1400	2006	Jun	Wednesday	10	Bryan Si



Insight 5:

The top 3 budget movies were directed by **Rob Marshall**, **Gore Verbinski** and **Bryan Singer**

```
In [ ]: #6.director who have directed top budget movies
df.sort_values('budget',ascending=False)[['director_name','budget']].head()
```

```
Out[ ]:
```

	director_name	budget
12	Rob Marshall	38.0
1	Gore Verbinski	30.0
6	Bryan Singer	27.0
4	Sam Raimi	25.8
9	Gore Verbinski	25.5

Insight 6:

Rob Marshall Directed a movie with the highest budget of **380 million**, **Gore Verbinski** Directed two movies with budgets over **\$250 million**