```
In [1]:
          import pandas as pd
          import numpy as np
In [3]: data=pd.read_csv("/home/placement/Downloads/movies.csv")
In [4]: data.describe()
Out[4]:
                         srno
                                                  rating
                                                                time
                                       year
           count 49590.000000
                               49590.000000
                                           10814.000000
                                                         45836.000000
           mean 24795.500000
                                2002.303428
                                                          2628.445436
                                                3.451248
             std 14315.544261
                                  12.534555
                                                0.495601
                                                          1604.646265
             min
                      1.000000
                               1913.000000
                                                1.400000
                                                            52.000000
            25%
                 12398.250000
                               1999.000000
                                                3.100000
                                                          1356.000000
                  24795.500000
                                2007.000000
                                                3.500000
                                                          2563.000000
                  37192.750000
                                                3.800000
                                2010.000000
                                                          2877.000000
                               2014.000000
                  49590.000000
                                                4.500000
                                                         28813.000000
In [5]:
          data.head(5)
Out[5]:
                                movie year rating
                                                     time
              srno
                 1 The Nightmare Before 1993
                                                   4568.0
           0
                                               3.9
                 2
                           The Mummy 1932
                                                   4388.0
                                               3.5
           2
                    Orphans of the Storm 1921
                                               3.2 9062.0
                    The Object of Beauty 1991
           3
                                               2.8 6150.0
                 5
                             Night Tide 1963
                                               2.8 5126.0
          data1=data.loc[(data.time<5000)]</pre>
In [7]:
```

In [8]: data1

Out[8]:

	srno	movie	year	rating	time
0	1	The Nightmare Before	1993	3.9	4568.0
1	2	The Mummy	1932	3.5	4388.0
16	17	The Bride of Frankenstein	1935	3.7	4485.0
23	24	Beavis and Butt-head Do America	1996	3.4	4852.0
50	51	Elvis '56	1987	3.8	3518.0
49580	49581	Curious George: A Very Monkey Christmas	2009	3.8	3438.0
49581	49582	Mumfie's White Christmas	1996	2.4	1350.0
49582	49583	Lady Gaga & The Muppets' Holiday Spectacular	2013	3.1	3496.0
49585	49586	Winter Wonderland	2013	2.8	1812.0
49587	49588	Fireplace For Your Home: Crackling Fireplace w	2010	NaN	3610.0

Out[12]:

	srno	movie	year	rating	time
409	410	Believe	2000	3.3	5767.0
416	417	The Prophecy 3: The Ascent	2000	3.4	5048.0
430	431	Scream 3	2000	3.2	7013.0
432	433	Holy Smoke	2000	3.0	6855.0
437	438	Requiem for a Dream	2000	3.9	6087.0
•••					
32557	32558	Shaded Places	2000	2.9	5350.0
36229	36230	The Three Stooges	2000	3.7	5256.0
37333	37334	Les Miserables: Pt. 2	2000	NaN	5170.0
37336	37337	Les Miserables: Pt. 1	2000	NaN	5194.0
39493	39494	The Prophet's Game	2000	3.2	6486.0

In [13]: data.head(100)

Out[13]:

	srno	movie	year	rating	time
0	1	The Nightmare Before	1993	3.9	4568.0
1	2	The Mummy	1932	3.5	4388.0
2	3	Orphans of the Storm	1921	3.2	9062.0
3	4	The Object of Beauty	1991	2.8	6150.0
4	5	Night Tide	1963	2.8	5126.0
95	96	The Hunted	1995	3.4	6605.0
96	97	The Great Waldo Pepper	1975	3.5	6467.0
97	98	Godzilla: King of the Monsters	1956	3.5	4828.0
98	99	Highlander 2: Renegade Version	1991	3.1	6585.0
99	100	High Noon	1952	3.9	5087.0

In [14]: data.head(50)

Out[14]:

	srno	movie	year	rating	time
	0 1	The Nightmare Before	1993	3.9	4568.0
	1 2	The Mummy	1932	3.5	4388.0
	2 3	Orphans of the Storm	1921	3.2	9062.0
	3 4	The Object of Beauty	1991	2.8	6150.0
	4 5	Night Tide	1963	2.8	5126.0
	5 6	One Magic Christmas	1985	3.8	5333.0
	6 7	Muriel's Wedding	1994	3.5	6323.0
	7 8	Mother's Boys	1994	3.4	5733.0
	8 9	Nosferatu: Original Version	1929	3.5	5651.0
	9 10	Nick of Time	1995	3.4	5333.0
1	.0 11	Broken Blossoms	1919	3.3	5367.0
1	l 1 12	Big Night	1996	3.6	6561.0
1	.2 13	The Birth of a Nation	1915	2.9	12118.0
1	.3 14	The Boys from Brazil	1978	3.6	7417.0
1	.4 15	Big Doll House	1971	2.9	5696.0
1	. 5 16	The Breakfast Club	1985	4.0	5823.0
1	.6 17	The Bride of Frankenstein	1935	3.7	4485.0
1	.7 18	Beautiful Girls	1996	3.5	6755.0
1	.8 19	Bustin' Loose	1981	3.7	5598.0
1	.9 20	The Beguiled	1971	3.4	6307.0
2	20 21	Born on the Fourth of July	1989	3.4	8646.0
2	21 22	Broadcast News	1987	3.4	7940.0
2	2 2 23	Swimming with Sharks	1994	3.3	5586.0
2	23 24	Beavis and Butt-head Do America	1996	3.4	4852.0

	srno	movie	year	rating	time
24	25	Brighton Beach Memoirs	1986	3.4	6564.0
25	26	The Best of Times	1986	3.4	6247.0
26	27	Brassed Off	1996	3.5	6040.0
27	28	Last Tango in Paris	1972	3.1	7732.0
28	29	Leprechaun 2	1994	3.2	5125.0
29	30	Incident at Oglala: The Leonard Peltier Story	1992	3.7	5487.0
30	31	Kalifornia	1993	3.4	7095.0
31	32	The Lady Vanishes	1938	3.7	5762.0
32	33	Jingle All the Way	1996	3.6	5371.0
33	34	Killing Zoe	1993	3.4	5773.0
34	35	King of Beggars	1992	3.6	6025.0
35	36	Into the Woods	1990	4.0	9077.0
36	37	Joe Kidd	1972	3.7	5252.0
37	38	In Too Deep	1999	3.9	5823.0
38	39	King Kong	1976	3.2	8044.0
39	40	Internal Affairs	1990	3.5	6885.0
40	41	Jesus Christ Superstar	1973	3.6	6388.0
41	42	In the Name of the Father	1993	3.9	7972.0
42	43	Easy Money	1987	2.8	5794.0
43	44	Do the Right Thing	1989	3.6	7186.0
44	45	Days of Heaven	1978	3.4	5628.0
45	46	Drop Zone	1994	3.4	6087.0
46	47	Escape from L.A.	1996	3.3	6039.0
47	48	Emma	1996	3.5	7260.0
48	49	Disco Godfather	1979	3.0	5869.0
49	50	The Eiger Sanction	1975	3.5	7726.0

In [15]: data.tail(50)

Out[15]:

	srno	movie	year	rating	time
49540	49541	Life With Boys: Season 1: This Time the Proble	2011	NaN	1368.0
49541	49542	Life With Boys: Season 1: In the Principal's O	2011	NaN	1368.0
49542	49543	El Fuente: 5994 MP	2013	3.5	471.0
49543	49544	Life With Boys: Season 1: Monkey Talk with Boys	2011	NaN	1368.0
49544	49545	El Fuente: 50 MP	2013	3.3	465.0
49545	49546	Bo Burnham: what.	2013	4.1	3614.0
49546	49547	Life With Boys: Season 1: A Perfect Life with	2011	NaN	1368.0
49547	49548	Life With Boys: Season 1: Wrestling with Boys	2011	NaN	1369.0
49548	49549	Life With Boys: Season 1	2011	4.1	NaN
49549	49550	El Fuente: 30 MP	2013	3.3	471.0
49550	49551	El Fuente: 2997 MP	2013	3.2	471.0
49551	49552	El Fuente: 25 MP	2013	3.3	465.0
49552	49553	El Fuente: 24 MP	2013	3.5	484.0
49553	49554	Max Steel	2013	4.1	NaN
49554	49555	Saving Santa	2013	3.8	5038.0
49555	49556	Lilyhammer: Season 1 (Recap)	2013	4.2	194.0
49556	49557	Shinobi Girl	2012	2.0	NaN
49557	49558	GLOW: The Story of the Gorgeous Ladies of Wres	2012	3.7	4593.0
49558	49559	Mitt (Trailer)	2013	3.0	138.0
49559	49560	My Hope America with Billy Graham: The Cross	2013	NaN	1706.0
49560	49561	The Square (Trailer)	2014	3.6	154.0
49561	49562	My Hope America with Billy Graham	2013	3.9	NaN
49562	49563	My Hope America with Billy Graham: Defining Mo	2013	NaN	1791.0
49563	49564	My Hope America with Billy Graham: Lose to Gain	2013	NaN	1400.0

	srno	movie	year	rating	time
49564	49565	American Addict	2013	3.5	5377.0
49565	49566	My Hope America with Billy Graham	2013	3.9	NaN
49566	49567	El Fuente: 2997 MP10	2013	2.7	470.0
49567	49568	El Fuente: 25 MP10	2013	2.9	464.0
49568	49569	El Fuente: 24 MP10	2013	2.8	484.0
49569	49570	El Fuente: 23976 MP10	2013	2.9	484.0
49570	49571	The Short Game (Trailer)	2013	4.1	156.0
49571	49572	El Fuente: 5994 MP10	2013	2.8	471.0
49572	49573	El Fuente: 50 MP10	2013	2.9	464.0
49573	49574	El Fuente: 30 MP10	2013	2.8	470.0
49574	49575	Greg Fitzsimmons: Life on Stage	2013	3.3	3671.0
49575	49576	Dave Foley: Relatively Well	2013	3.2	3446.0
49576	49577	Barbie: Life in the Dreamhouse: Barbie Life in	2013	NaN	1390.0
49577	49578	Barbie: Life in the Dreamhouse: Barbie Life in	2013	NaN	1458.0
49578	49579	Transformers Prime Beast Hunters: Predacons Ri	2013	4.2	3950.0
49579	49580	Underground: The Julian Assange Story	2012	3.7	5665.0
49580	49581	Curious George: A Very Monkey Christmas	2009	3.8	3438.0
49581	49582	Mumfie's White Christmas	1996	2.4	1350.0
49582	49583	Lady Gaga & The Muppets' Holiday Spectacular	2013	3.1	3496.0
49583	49584	Sunset Strip	2012	3.0	5770.0
49584	49585	Silver Bells	2013	3.5	5287.0
49585	49586	Winter Wonderland	2013	2.8	1812.0
49586	49587	Top Gear: Series 19: Africa Special	2013	NaN	6822.0
49587	49588	Fireplace For Your Home: Crackling Fireplace w	2010	NaN	3610.0
49588	49589	Kate Plus Ei8ht	2010	2.7	NaN
49589	49590	Kate Plus Ei8ht: Season 1	2010	2.7	NaN

Out[18]:

	srno	movie	year	rating	time
984	985	Paid in Full	2002	4.1	5883.0
1061	1062	The Pianist	2002	4.1	8945.0
1285	1286	Pirates of the Caribbean: The Curse of the Bla	2003	4.1	8586.0
1305	1306	Ken Burns: Jazz	2001	4.1	NaN
1363	1364	The Life of Mammals	2002	4.2	NaN
•••					
47765	47766	Brew Masters: Season 1	2010	4.1	NaN
48212	48213	Cake Boss: Next Great Baker: Season 1	2010	4.1	NaN
48854	48855	Pit Bulls & Parolees	2009	4.3	NaN
48874	48875	Brew Masters	2010	4.1	NaN
49025	49026	Cake Boss: Next Great Baker	2010	4.1	NaN

Out[19]:

	srno	movie	year	rating	time
484	485	Coming to Light: Edward S. Curtis and the Nort	2000	4.0	5027.0
560	561	The Emperor's New Groove	2000	4.0	4703.0
816	817	Dil Chahta Hai	2001	4.0	11110.0
865	866	Black Hawk Down	2001	4.0	8660.0
920	921	We Were Soldiers	2002	4.0	8309.0
47718	47719	Killer Whales	2010	4.0	2620.0
47793	47794	King Tut Unwrapped: Season 1	2010	4.0	NaN
48066	48067	Police Women of Dallas: Season 1	2010	4.0	NaN
48784	48785	King Tut Unwrapped	2010	4.0	NaN
49028	49029	Police Women of Dallas	2010	4.0	NaN

287 rows × 5 columns

localhost:8888/notebooks/movies.ipynb

In [24]: datat=data.sort_values('time')
 datat

Out[24]:

	srno	movie	year	rating	time
40150	40151	Trailer: Pain	2012	3.6	52.0
41081	41082	Trailer: Get to Work	2012	3.3	55.0
41082	41083	Trailer: Give and Take	2012	3.3	66.0
43166	43167	Trailer: Emperor	2013	3.1	67.0
43330	43331	Trailer: Blood Angel	2013	4.2	69.0
49556	49557	Shinobi Girl	2012	2.0	NaN
49561	49562	My Hope America with Billy Graham	2013	3.9	NaN
49565	49566	My Hope America with Billy Graham	2013	3.9	NaN
49588	49589	Kate Plus Ei8ht	2010	2.7	NaN
49589	49590	Kate Plus Ei8ht: Season 1	2010	2.7	NaN

NaN 6822.0

NaN 3610.0

```
In [26]: datar=data.sort_values('rating')
datar
```

Out[26]:		srno	movie	year	rating	time	
	40934	40935	Lagegi	2007	1.4	NaN	
	42115	42116	Sun Yaar Chill Maar	2007	1.4	NaN	
	40826	40827	Lagegi	2007	1.4	NaN	
	42160	42161	Sun Yaar Chill Maar	2007	1.4	NaN	
	41396	41397	Meri Toh Lag Gayi Naukri	2011	1.5	NaN	
	49563	49564	My Hope America with Billy Graham: Lose to Gain	2013	NaN	1400.0	
	49576	49577	Barbie: Life in the Dreamhouse: Barbie Life in	2013	NaN	1390.0	
	49577	49578	Barbie: Life in the Dreamhouse: Barbie Life in	2013	NaN	1458.0	

Top Gear: Series 19: Africa Special 2013

Fireplace For Your Home: Crackling Fireplace w... 2010

49590 rows × 5 columns

49586 49587

49587 49588

localhost:8888/notebooks/movies.ipynb

```
In [34]: cr=x.cursor()
In [42]: cr.execute('insert into data(id,name,age,dept,total)values(121,"vinay",22,"ece",454)')
Out[42]: 1
In [44]: x.commit()
         x.close()
In [2]: !pip3 install pymysql
         Requirement already satisfied: pymysql in ./anaconda3/lib/python3.10/site-packages (1.0.3)
In [3]: import pymysql
In [17]: x=pymysql.connect(host='localhost',user='root',password='Password@123',database='project1')
In [18]: cr=x.cursor()
```

```
In [26]: cr.execute('insert into state(srno,state,language,city) values(1,"AP","TELUGU","NSP")')
         cr.execute('insert into state(srno,state,language,city) values(2,"TN","TAMIL",CHENNAI")')
         cr.execute('insert into state(srno,state,language,city) values(3,"KERALA","MALAYALAM","TCR")')
         cr.execute('insert into state(srno,state,language,city) values(4, "KARNATAKA", "KANNADA", "BANGLORE")')
         cr.execute('insert into state(srno,state,language,city) values(5, "GOA", "KONGINI", "PANAJI")')
                                                    Traceback (most recent call last)
         ProgrammingError
         Cell In[26], line 2
               1 cr.execute('insert into state(srno, state, language, city) values(1, "AP", "TELUGU", "NSP")')
         ----> 2 cr.execute('insert into state(srno,state,language,city) values(2,"TN","TAMIL",CHENNAI")')
               3 cr.execute('insert into state(srno,state,language,city) values(3,"KERALA","MALAYALAM","TCR")')
               4 cr.execute('insert into state(srno, state, language, city) values(4, "KARNATAKA", "KANNADA", "BANGLORE")'
         File ~/anaconda3/lib/python3.10/site-packages/pymysgl/cursors.py:158, in Cursor.execute(self, guery, args)
             154
                      pass
             156 guery = self.mogrify(guery, args)
         --> 158 result = self. query(query)
             159 self. executed = query
             160 return result
         File ~/anaconda3/lib/python3.10/site-packages/pymysgl/cursors.py:325, in Cursor. guery(self, g)
              323 \text{ conn} = \text{self. get db()}
             324 self. clear result()
         --> 325 conn.query(g)
             326 self. do get result()
             327 return self.rowcount
         File ~/anaconda3/lib/python3.10/site-packages/pymysgl/connections.py:549, in Connection.guery(self, sgl, un
         buffered)
                      sql = sql.encode(self.encoding, "surrogateescape")
              547
             548 self. execute command(COMMAND.COM QUERY, sql)
         --> 549 self._affected_rows = self. read query result(unbuffered=unbuffered)
              550 return self. affected rows
         File ~/anaconda3/lib/python3.10/site-packages/pymysql/connections.py:779, in Connection. read guery result
         (self, unbuffered)
             777 else:
              778
                      result = MySQLResult(self)
```

```
--> 779
            result.read()
    780 self. result = result
    781 if result.server status is not None:
File ~/anaconda3/lib/python3.10/site-packages/pymysgl/connections.py:1157, in MySOLResult.read(self)
   1155 def read(self):
   1156
            try:
                first packet = self.connection. read packet()
-> 1157
                if first packet.is ok packet():
   1159
   1160
                    self. read ok packet(first packet)
File ~/anaconda3/lib/python3.10/site-packages/pymysgl/connections.py:729, in Connection. read packet(self,
 packet type)
            if self. result is not None and self. result.unbuffered active is True:
    727
                self. result.unbuffered active = False
    728
--> 729
            packet.raise for error()
    730 return packet
File ~/anaconda3/lib/python3.10/site-packages/pymysgl/protocol.py:221, in MysglPacket.raise for error(self)
    219 if DEBUG:
            print("errno =", errno)
    220
--> 221 err.raise mysgl exception(self. data)
File ~/anaconda3/lib/python3.10/site-packages/pymysgl/err.py:143, in raise mysgl exception(data)
    141 if errorclass is None:
            errorclass = InternalError if errno < 1000 else OperationalError
    142
--> 143 raise errorclass(errno, errval)
ProgrammingError: (1064, 'You have an error in your SQL syntax; check the manual that corresponds to your M
vSOL server version for the right syntax to use near \'")\' at line 1')
```

In []: