

```
import pandas as pd
import numpy as np
```

```
movies = pd.read_csv('imdb-top-1000.csv')
movies.head(2)
```

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating
0	The Shawshank Redemption	1994	142	Drama	9.3
1	The Godfather	1972	175	Crime	9.2

	Director	Star1	No_of_Votes	Gross
0	Frank Darabont	Tim Robbins	2343110	28341469.0
1	Francis Ford Coppola	Marlon Brando	1620367	134966411.0

Groupby - is basically the study of groups
group by apply hota hai categorical column ke upar (jiska ki value humesa change karte rahata hai for ex : genre in our data set)

```
movies.shape
```

```
(1000, 10)
```

genre ke basis pe movie ko categorise karna hai suppose sare Drama type ke movie ko group karna hai then sare crime type ko alag

```
movies.groupby('Genre') # abhi kuch output nahi dikhega bass pata chalega ki hum jo operation kye hain usse DataFrameGroupBy object create ho gya hai
```

```
<pandas.core.groupby.generic.DataFrameGroupBy object at 0x00000290C6ACB8C0>
```

```
genres = movies.groupby('Genre')
```

```
# Applying builtin aggregation functions on groupby objects
genres.sum()
```

yaha pe jo output aa raha hai usme har row me alag alag genre hai means ki ye table genres ke basis pe groupby ho gya hai and sum function ko call
karne se particular column ke sare data add ho gya hai

	Series_Title
Genre	
Action	The Dark KnightThe Lord of the Rings: The Retu...

Adventure	Interstellar	Back to the Future	Inglourious Bast...
Animation	Sen to Chihiro no kamikakushi	The Lion King	Hota...
Biography	Schindler's List	Goodfellas	Hamilton
Comedy	Gisaengchung	La vita è bella	Modern Times
Crime	The Godfather	The Godfather: Part II	12 Angry Me...
Drama	The Shawshank Redemption	Fight Club	Forrest Gump...
Family	E.T. the Extra-Terrestrial	Willy Wonka & the Ch...	
Fantasy		Das Cabinet des Dr. Caligari	Nosferatu
Film-Noir	The Third Man	The Maltese Falcon	Shadow of a Doubt
Horror	Psycho	Alien	The Thing
Mystery	Memento	Rear Window	Vertigo
Thriller			Shutter Island
Western	Il buono, il brutto, il cattivo	Once Upon a Tim...	

		Released_Year	Runtime
\ Genre			
Action	2008200320102001200219991980197719621954200019...		22196
Adventure	2014198520091981196819621959201319751963194819...		9656
Animation	2001199419882016201820172008199719952019200920...		8166
Biography	1993199020202011200220171995198420182013201320...		11970
Comedy	2019199719361931200919641940200120001973196019...		17380
Crime	1972197419571994200219991995199120192006199519...		13524
Drama	1994199919941975202019981946201420061998198819...		36049
Family		19821971	215
Fantasy		19201922	170
Film-Noir		194919411943	312
Horror	19601979198219731968196120171978193320042001		1123
Mystery	200019541958201020121995197219381988201219981997		1429
Thriller		1967	108
Western		1966196819651976	593

	IMDB_Rating
Director \	
Genre	

Action	1367.3	Christopher Nolan	Peter Jackson	Christopher Nola...
Adventure	571.5	Christopher Nolan	Robert Zemeckis	Quentin Tarant...
Animation	650.3	Hayao Miyazaki	Roger Allers	Isao Takahata
Biography	698.6	Steven Spielberg	Martin Scorsese	Thomas Kail
Comedy	1224.7	Bong Joon Ho	Roberto Benigni	Charles Chaplin
Crime	857.8	Francis Ford Coppola	Francis Ford	Coppola
Drama	2299.7	Frank Darabont	David Fincher	Robert Zemeckis
Family	15.6		Steven Spielberg	Mel Stuart
Fantasy	16.0		Robert Wiene	F.W. Murnau
Film-Noir	23.9		Carol Reed	John Huston
Horror	87.0	Alfred Hitchcock	Ridley Scott	John Carpenter
Mystery	95.7	Christopher Nolan	Alfred Hitchcock	Alfred Hitchc...
Thriller	7.8			Terence Young
Western	33.4	Sergio Leone	Sergio Leone	Sergio Leone
East...				Clint East...

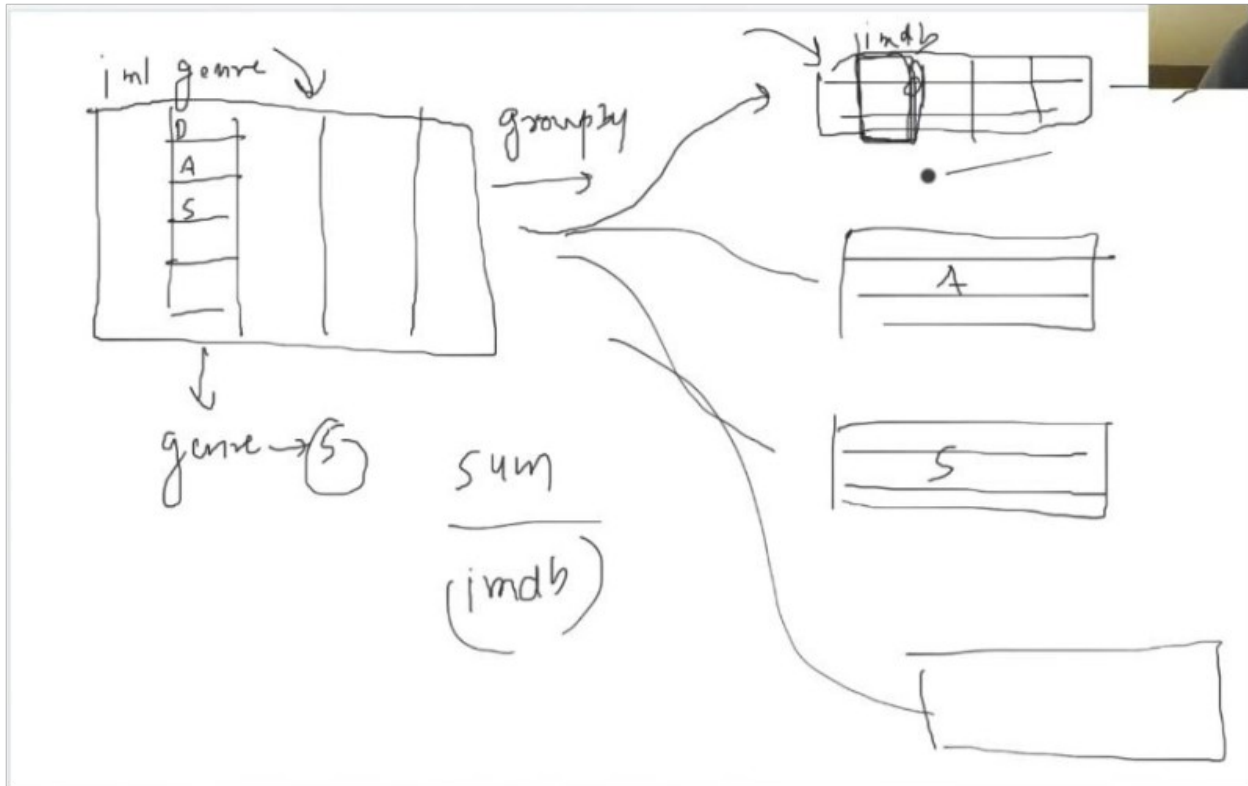
Star1

No_of_Votes \
Genre

Action	Christian Bale	Elijah Wood	Leonardo DiCaprio	Elij...
Adventure	Matthew McConaughey	Michael J. Fox	Brad Pitt	Jürg...
Animation	Daveigh Chase	Rob Minkoff	Tsutomu Tatsumi	Ryûnosu...
Biography	Liam Neeson	Robert De Niro	Lin-Manuel Miranda	Éri...
Comedy	Kang-ho Song	Roberto Benigni	Charles Chaplin	Char...
Crime	Marlon Brando	Al Pacino	Henry Fonda	John Travolta...
Drama	Tim Robbins	Brad Pitt	Tom Hanks	Jack Nicholson
Family		Henry Thomas	Gene Wilder	

Fantasy	Werner Krauss	Max Schreck
146222		
Film-Noir	Orson Welles	Humphrey Bogart
367215		Teresa Wright
Horror	Anthony Perkins	Sigourney Weaver
3742556		Kurt Russell
Mystery	Guy Pearce	James Stewart
4203004		James Stewart
Thriller		Leonardo D...
27733		Audrey Hepburn
Western	Clint Eastwood	Henry Fonda
1289665		Clint Eastwood

	Gross	Metascore
Genre		
Action	3.263226e+10	10499.0
Adventure	9.496922e+09	5020.0
Animation	1.463147e+10	6082.0
Biography	8.276358e+09	6023.0
Comedy	1.566387e+10	9840.0
Crime	8.452632e+09	6706.0
Drama	3.540997e+10	19208.0
Family	4.391106e+08	158.0
Fantasy	7.827267e+08	0.0
Film-Noir	1.259105e+08	287.0
Horror	1.034649e+09	880.0
Mystery	1.256417e+09	633.0
Thriller	1.755074e+07	81.0
Western	5.822151e+07	313.0



`genres.min()` # minimum value nikal ke de dega and kuch chize observe kar ssakte hain jaise ki Action me jo sabse purani movie hai wo 1924 me aayi
 # thi , action movie ka minimum length 45 hai , adventure movie me sabse kam rating 7.6 hai and so on.....

	Series_Title	Released_Year	Runtime	\
Genre				
Action	300	1924	45	
Adventure	2001: A Space Odyssey	1925	88	
Animation	Akira	1940	71	
Biography	12 Years a Slave	1928	93	
Comedy	(500) Days of Summer	1921	68	
Crime	12 Angry Men	1931	80	
Drama	1917	1925	64	
Family	E.T. the Extra-Terrestrial	1971	100	
Fantasy	Das Cabinet des Dr. Caligari	1920	76	
Film-Noir	Shadow of a Doubt	1941	100	
Horror	Alien	1933	71	
Mystery	Dark City	1938	96	
Thriller	Wait Until Dark	1967	108	
Western	Il buono, il brutto, il cattivo	1965	132	
	IMDB_Rating	Director		
Star1	\			
Genre				

Action Khan	7.6	Abhishek Chaubey	Aamir
Adventure Khan	7.6	Akira Kurosawa	Aamir
Animation Molina	7.6	Adam Elliot	Adrian
Biography Brody	7.6	Adam McKay	Adrien
Comedy Khan	7.6	Alejandro G. Iñárritu	Aamir
Crime Devgn	7.6	Akira Kurosawa	Ajay
Drama Deol	7.6	Aamir Khan	Abhay
Family Wilder	7.8	Mel Stuart	Gene
Fantasy Schreck	7.9	F.W. Murnau	Max
Film-Noir Bogart	7.8	Alfred Hitchcock	Humphrey
Horror Perkins	7.6	Alejandro Amenábar	Anthony
Mystery Donnadieu	7.6	Alex Proyas	Bernard-Pierre
Thriller Hepburn	7.8	Terence Young	Audrey
Western Eastwood	7.8	Clint Eastwood	Clint

	No_of_Votes	Gross	Metascore
Genre			
Action	25312	3296.0	33.0
Adventure	29999	61001.0	41.0
Animation	25229	128985.0	61.0
Biography	27254	21877.0	48.0
Comedy	26337	1305.0	45.0
Crime	27712	6013.0	47.0
Drama	25088	3600.0	28.0
Family	178731	4000000.0	67.0
Fantasy	57428	337574718.0	NaN
Film-Noir	59556	449191.0	94.0
Horror	27007	89029.0	46.0
Mystery	33982	1035953.0	52.0
Thriller	27733	17550741.0	81.0
Western	65659	5321508.0	69.0

genres.max()

\ Genre	Series_Title Released_Year		Runtime
Action	Yôjinbô	2019	321
Adventure	Zombieland	PG	228
Animation	Ôkami kodomo no Ame to Yuki	2020	137
Biography	Zerkalo	2020	209
Comedy	Zindagi Na Milegi Dobara	2020	188
Crime	À bout de souffle	2019	229
Drama	Zwartboek	2020	242
Family	Willy Wonka & the Chocolate Factory	1982	115
Fantasy	Nosferatu	1922	94
Film-Noir	The Third Man	1949	108
Horror	The Thing	2017	122
Mystery	Vertigo	2012	138
Thriller	Wait Until Dark	1967	108
Western	The Outlaw Josey Wales	1976	165
\ Genre	IMDB_Rating	Director	Star1
	No_of_Votes		
Action	9.0	Zack Snyder	Yun-Fat Chow
2303232			
Adventure	8.6	Ömer Faruk Sorak	Yves Montand
1512360			
Animation	8.6	Yoshifumi Kondô	Yôji Matsuda
999790			
Biography	8.9	Tom McCarthy	Éric Toledano
1213505			
Comedy	8.6	Zoya Akhtar	Ömer Faruk Sorak
939631			
Crime	9.2	Yavuz Turgul	Vincent Cassel
1826188			
Drama	9.3	Çagan Irmak	Çetin Tekindor
2343110			

Family 372490	7.8	Steven Spielberg	Henry Thomas
Fantasy 88794	8.1	Robert Wiene	Werner Krauss
Film-Noir 158731	8.1	John Huston	Teresa Wright
Horror 787806	8.5	William Friedkin	Sigourney Weaver
Mystery 1129894	8.4	Terry Gilliam	Vidya Balan
Thriller 27733	7.8	Terence Young	Audrey Hepburn
Western 688390	8.8	Sergio Leone	Henry Fonda

	Gross	Metascore
Genre		
Action	936662225.0	98.0
Adventure	874211619.0	100.0
Animation	873839108.0	96.0
Biography	753585104.0	97.0
Comedy	886752933.0	99.0
Crime	790482117.0	100.0
Drama	924558264.0	100.0
Family	435110554.0	91.0
Fantasy	445151978.0	NaN
Film-Noir	123353292.0	97.0
Horror	298791505.0	97.0
Mystery	474203697.0	100.0
Thriller	17550741.0	81.0
Western	31800000.0	90.0

`genres.mean()` # error dega qki bale column ka mean kaise find karega for that data should be integer

```
-----
-----
TypeError                                Traceback (most recent call
last)
File ~\anaconda3\Lib\site-packages\pandas\core\groupby\
groupby.py:1942, in GroupBy._agg_py_fallback(self, how, values, ndim,
alt)
    1941 try:
-> 1942     res_values = self._grouper.agg_series(ser, alt,
preserve_dtype=True)
    1943 except Exception as err:
```



```
File ~\anaconda3\Lib\site-packages\pandas\core\groupby\ops.py:864, in
BaseGrouper.agg_series(self, obj, func, preserve_dtype)
    862     preserve_dtype = True
--> 864 result = self._aggregate_series_pure_python(obj, func)
    866 npvalues = lib.maybe_convert_objects(result, try_float=False)
```

```
File ~\anaconda3\Lib\site-packages\pandas\core\groupby\ops.py:885, in
BaseGrouper._aggregate_series_pure_python(self, obj, func)
    884 for i, group in enumerate(splitter):
--> 885     res = func(group)
    886     res = extract_result(res)
```

```
File ~\anaconda3\Lib\site-packages\pandas\core\groupby\
groupby.py:2454, in GroupBy.mean.<locals>.<lambda>(x)
    2451 else:
    2452     result = self._cython_agg_general(
    2453         "mean",
-> 2454         alt=lambda x: Series(x,
copy=False).mean(numeric_only=numeric_only),
    2455         numeric_only=numeric_only,
    2456     )
    2457     return result.__finalize__(self.obj, method="groupby")
```

```
File ~\anaconda3\Lib\site-packages\pandas\core\series.py:6549, in
Series.mean(self, axis, skipna, numeric_only, **kwargs)
    6541 @doc(make_doc("mean", ndim=1))
    6542 def mean(
    6543     self,
    (...)
    6547     **kwargs,
    6548 ):
-> 6549     return NDFrame.mean(self, axis, skipna, numeric_only,
**kwargs)
```

```
File ~\anaconda3\Lib\site-packages\pandas\core\generic.py:12420, in
NDFrame.mean(self, axis, skipna, numeric_only, **kwargs)
    12413 def mean(
    12414     self,
    12415     axis: Axis | None = 0,
    (...)
    12418     **kwargs,
    12419 ) -> Series | float:
> 12420     return self._stat_function(
    12421         "mean", nanops.nanmean, axis, skipna, numeric_only,
**kwargs
    12422     )
```

```
File ~\anaconda3\Lib\site-packages\pandas\core\generic.py:12377, in
NDFrame._stat_function(self, name, func, axis, skipna, numeric_only,
**kwargs)
```

```

12375 validate_bool_kwarg(skipna, "skipna", none_allowed=False)
> 12377 return self._reduce(
12378     func, name=name, axis=axis, skipna=skipna,
numeric_only=numeric_only
12379 )

```

File ~\anaconda3\Lib\site-packages\pandas\core\series.py:6457, in Series._reduce(self, op, name, axis, skipna, numeric_only, filter_type, **kwds)

```

6453     raise TypeError(
6454         f"Series.{name} does not allow
{kwd_name}={numeric_only} "
6455         "with non-numeric dtypes."
6456     )
-> 6457 return op(delegate, skipna=skipna, **kwds)

```

File ~\anaconda3\Lib\site-packages\pandas\core\nanops.py:147, in bottleneck_switch.__call__.<locals>.f(values, axis, skipna, **kwds)

```

146 else:
--> 147     result = alt(values, axis=axis, skipna=skipna, **kwds)
149 return result

```

File ~\anaconda3\Lib\site-packages\pandas\core\nanops.py:404, in _datetimelike_compat.<locals>.new_func(values, axis, skipna, mask, **kwargs)

```

402     mask = isna(values)
--> 404 result = func(values, axis=axis, skipna=skipna, mask=mask,
**kwargs)
406 if datetimelike:

```

File ~\anaconda3\Lib\site-packages\pandas\core\nanops.py:720, in nanmean(values, axis, skipna, mask)

```

719 the_sum = values.sum(axis, dtype=dtype_sum)
--> 720 the_sum = _ensure_numeric(the_sum)
722 if axis is not None and getattr(the_sum, "ndim", False):

```

File ~\anaconda3\Lib\site-packages\pandas\core\nanops.py:1701, in _ensure_numeric(x)

```

1699 if isinstance(x, str):
1700     # GH#44008, GH#36703 avoid casting e.g. strings to numeric
-> 1701     raise TypeError(f"Could not convert string '{x}' to
numeric")
1702 try:

```

TypeError: Could not convert string 'The Dark KnightThe Lord of the Rings: The Return of the KingInceptionThe Lord of the Rings: The Fellowship of the RingThe Lord of the Rings: The Two TowersThe MatrixStar Wars: Episode V - The Empire Strikes BackStar WarsSeppukuShichinin no samuraiGladiatorLéonTerminator 2: Judgment DayVikram VedhaDangalAvengers: EndgameAvengers: Infinity WarThe Dark

Knight RisesOldeuboiRaiders of the Lost ArkAliensStar Wars: Episode VI
 - Return of the JediUri: The Surgical StrikeK.G.F: Chapter 1Baahubali
 2: The ConclusionGangs of WasseyapurPaan Singh TomarWarriorV for
 VendettaBatman BeginsHera PheriLock, Stock and Two Smoking
 BarrelsAndaz Apna ApnaIndiana Jones and the Last CrusadeDie
 HardRanSholayYôjinbôSherlock Jr.HaiderLoganRushFord v FerrariMad Max:
 Fury RoadA WednesdayTaegukgi hwinalrimyeoKill Bill: Vol. 1Jurassic
 ParkBlade RunnerSanjuroWhite HeatThe GeneralBajrangi
 BhaijaanBabyBãhubali: The BeginningSerbuan maut 2: BerandalGuardians
 of the GalaxyBlade Runner 2049The RevenantTropa de Elite 2: O Inimigo
 Agora é OutroDeadpoolYip ManNefes: Vatan SagolsunTropa de EliteThe
 AvengersThe Bourne UltimatumCasino RoyaleKill Bill: Vol. 2Mou gaan
 douPirates of the Caribbean: The Curse of the Black PearlKnockin' on
 Heaven's DoorThe TerminatorPer un pugno di dollariRio BravoThor:
 RagnarokStar Wars: Episode VII - The Force AwakensX-Men: Days of
 Future PastEdge of TomorrowDistrict 9Star TrekLetters from Iwo
 JimaIron ManYing xiongThe Bourne IdentityThe Blues BrothersDawn of the
 DeadAguirre, der Zorn GottesThe Wild BunchThe Adventures of Robin
 HoodThe GentlemenRaaziDunkirkUdta PunjabRogue OneCaptain America:
 Civil WarAng-ma-reul bo-at-daAjeossiChugyeokjaTakenAvatarApocalyptoHot
 FuzzSerenityGongdong gyeongbi guyeok JSAThe Count of Monte CristoWo hu
 cang longThe Boondock SaintsThe GameTombstoneThe FugitiveLat sau san
 taamDip huet seung hungPredatorEvil Dead IIGhostbustersBadlandsBonnie
 and ClydeThe Longest DayRed RiverKey LargoScarface: The Shame of the
 NationDeadpool 2Mission: Impossible - FalloutKingsman: The Secret
 ServiceCaptain America: The Winter SoldierStar Trek Into DarknessX:
 First ClassSkyfallLucky Number Slevin3:10 to YumaKung fuThe Bourne
 SupremacyMan on FireThe Last SamuraiThe Fifth ElementThe Last of the
 MohicansEmpire of the SunStar Trek II: The Wrath of KhanFirst BloodThe
 Taking of Pelham One Two ThreeEnter the DragonThe French
 ConnectionDirty HarryWhere Eagles DareThe Dirty DozenGoldfingerThe
 Magnificent SevenGuardians of the Galaxy Vol. 2Baby DriverOnly the
 BraveSicarioHell or High WaterDawn of the Planet of the ApesSerbuan
 mautEnd of WatchKick-AssCelda 211Sherlock HolmesEastern PromisesHuo
 Yuan JiaHarry Potter and the Half-Blood Prince300WatchmenLord of
 WarBatoru rowaiaruMinority ReportDie Hard: With a VengeanceFalling
 DownLethal WeaponMad Max 2The WarriorsEscape from Alcatraz' to numeric

The above exception was the direct cause of the following exception:

```

TypeError                                Traceback (most recent call
last)

```

```

Cell In[54], line 1

```

```

----> 1 genres.mean()['Runtime']

```

```

File ~\anaconda3\Lib\site-packages\pandas\core\groupby\
groupby.py:2452, in GroupBy.mean(self, numeric_only, engine,
engine_kwargs)

```

```

    2445     return self._numba_agg_general(

```

```

    2446         grouped_mean,

```

```

2447         executor.float_dtype_mapping,
2448         engine_kwargs,
2449         min_periods=0,
2450     )
2451 else:
-> 2452     result = self._cython_agg_general(
2453         "mean",
2454         alt=lambda x: Series(x,
copy=False).mean(numeric_only=numeric_only),
2455         numeric_only=numeric_only,
2456     )
2457     return result.__finalize__(self.obj, method="groupby")

```

File ~\anaconda3\Lib\site-packages\pandas\core\groupby\groupby.py:1998, in GroupBy._cython_agg_general(self, how, alt, numeric_only, min_count, **kwargs)

```

1995     result = self._agg_py_fallback(how, values,
ndim=data.ndim, alt=alt)
1996     return result
-> 1998 new_mgr = data.grouped_reduce(array_func)
1999 res = self._wrap_agged_manager(new_mgr)
2000 if how in ["idxmin", "idxmax"]:

```

File ~\anaconda3\Lib\site-packages\pandas\core\internals\managers.py:1469, in BlockManager.grouped_reduce(self, func)

```

1465 if blk.is_object:
1466     # split on object-dtype blocks bc some columns may raise
1467     # while others do not.
1468     for sb in blk._split():
-> 1469         applied = sb.apply(func)
1470         result_blocks = extend_blocks(applied, result_blocks)
1471 else:

```

File ~\anaconda3\Lib\site-packages\pandas\core\internals\blocks.py:393, in Block.apply(self, func, **kwargs)

```

387 @final
388 def apply(self, func, **kwargs) -> list[Block]:
389     """
390     apply the function to my values; return a block if we are
not
391     one
392     """
--> 393     result = func(self.values, **kwargs)
395     result = maybe_coerce_values(result)
396     return self._split_op_result(result)

```

File ~\anaconda3\Lib\site-packages\pandas\core\groupby\groupby.py:1995, in GroupBy._cython_agg_general.<locals>.array_func(values)

```

1992     return result

```

```

1994 assert alt is not None
-> 1995 result = self._agg_py_fallback(how, values, ndim=data.ndim,
alt=alt)
1996 return result

```

File ~\anaconda3\Lib\site-packages\pandas\core\groupby\groupby.py:1946, in GroupBy._agg_py_fallback(self, how, values, ndim, alt)

```

1944     msg = f"agg function failed [how->{how},dtype->{ser.dtype}]"
1945     # preserve the kind of exception that raised
-> 1946     raise type(err)(msg) from err
1948 if ser.dtype == object:
1949     res_values = res_values.astype(object, copy=False)

```

TypeError: agg function failed [how->mean,dtype->object]

genres.mean(numeric_only=True) *# this is the correct way in new version qki humare pass string data bhi hai*

	Runtime	IMDB_Rating	No_of_Votes	Gross
Metascore				
Genre				
Action	129.046512	7.949419	420246.581395	1.897224e+08
73.419580				
Adventure	134.111111	7.937500	313557.819444	1.319017e+08
78.437500				
Animation	99.585366	7.930488	268032.073171	1.784326e+08
81.093333				
Biography	136.022727	7.938636	272805.045455	9.404952e+07
76.240506				
Comedy	112.129032	7.901290	178195.658065	1.010572e+08
78.720000				
Crime	126.392523	8.016822	313398.271028	7.899656e+07
77.080460				
Drama	124.737024	7.957439	212343.612457	1.225259e+08
79.701245				
Family	107.500000	7.800000	275610.500000	2.195553e+08
79.000000				
Fantasy	85.000000	8.000000	73111.000000	3.913633e+08
NaN				
Film-Noir	104.000000	7.966667	122405.000000	4.197018e+07
95.666667				
Horror	102.090909	7.909091	340232.363636	9.405902e+07
80.000000				
Mystery	119.083333	7.975000	350250.333333	1.047014e+08
79.125000				
Thriller	108.000000	7.800000	27733.000000	1.755074e+07
81.000000				

```
Western      148.250000      8.350000  322416.250000  1.455538e+07
78.250000
```

```
genres.std()    # standard deviation -> same reason
```

```
-----
-----
ValueError                                Traceback (most recent call
last)
Cell In[56], line 1
----> 1 genres.std()['']
```

```
File ~\anaconda3\Lib\site-packages\pandas\core\groupby\
groupby.py:2641, in GroupBy.std(self, ddof, engine, engine_kwargs,
numeric_only)
    2631     return np.sqrt(
    2632         self._numba_agg_general(
    2633             grouped_var,
    2634             (...)
    2635         )
    2636     )
    2637 else:
-> 2641     return self._cython_agg_general(
    2642         "std",
    2643         alt=lambda x: Series(x, copy=False).std(ddof=ddof),
    2644         numeric_only=numeric_only,
    2645         ddof=ddof,
    2646     )
```

```
File ~\anaconda3\Lib\site-packages\pandas\core\groupby\
groupby.py:1998, in GroupBy._cython_agg_general(self, how, alt,
numeric_only, min_count, **kwargs)
    1995     result = self._agg_py_fallback(how, values,
ndim=data.ndim, alt=alt)
    1996     return result
-> 1998 new_mgr = data.grouped_reduce(array_func)
    1999 res = self._wrap_agged_manager(new_mgr)
    2000 if how in ["idxmin", "idxmax"]:
```

```
File ~\anaconda3\Lib\site-packages\pandas\core\internals\
managers.py:1469, in BlockManager.grouped_reduce(self, func)
    1465 if blk.is_object:
    1466     # split on object-dtype blocks bc some columns may raise
    1467     # while others do not.
    1468     for sb in blk._split():
-> 1469         applied = sb.apply(func)
    1470         result_blocks = extend_blocks(applied, result_blocks)
    1471 else:
```

```
File ~\anaconda3\Lib\site-packages\pandas\core\internals\
```

```

blocks.py:393, in Block.apply(self, func, **kwargs)
    387 @final
    388 def apply(self, func, **kwargs) -> list[Block]:
    389     """
    390     apply the function to my values; return a block if we are
not
    391     one
    392     """
--> 393     result = func(self.values, **kwargs)
    395     result = maybe_coerce_values(result)
    396     return self._split_op_result(result)

```

File ~\anaconda3\Lib\site-packages\pandas\core\groupby\groupby.py:1973, in

GroupBy._cython_agg_general.<locals>.array_func(values)

```

    1971 def array_func(values: ArrayLike) -> ArrayLike:
    1972     try:
-> 1973         result = self._grouper._cython_operation(
    1974             "aggregate",
    1975             values,
    1976             how,
    1977             axis=data.ndim - 1,
    1978             min_count=min_count,
    1979             **kwargs,
    1980         )
    1981     except NotImplementedError:
    1982         # generally if we have numeric_only=False
    1983         # and non-applicable functions
    1984         # try to python agg
    1985         # TODO: shouldn't min_count matter?
    1986         # TODO: avoid special casing SparseArray here
    1987         if how in ["any", "all"] and isinstance(values,
SparseArray):

```

File ~\anaconda3\Lib\site-packages\pandas\core\groupby\ops.py:831, in BaseGrouper._cython_operation(self, kind, values, how, axis, min_count, **kwargs)

```

    829 ids, _, _ = self.group_info
    830 ngroups = self.ngroups
--> 831 return cy_op.cython_operation(
    832     values=values,
    833     axis=axis,
    834     min_count=min_count,
    835     comp_ids=ids,
    836     ngroups=ngroups,
    837     **kwargs,
    838 )

```

File ~\anaconda3\Lib\site-packages\pandas\core\groupby\ops.py:550, in WrappedCythonOp.cython_operation(self, values, axis, min_count,

```

comp_ids, ngroups, **kwargs)
539 if not isinstance(values, np.ndarray):
540     # i.e. ExtensionArray
541     return values._groupby_op(
542         how=self.how,
543         has_dropped_na=self.has_dropped_na,
544         (...)
545         **kwargs,
546     )
--> 550 return self._cython_op_ndim_compat(
551     values,
552     min_count=min_count,
553     ngroups=ngroups,
554     comp_ids=comp_ids,
555     mask=None,
556     **kwargs,
557 )

```

File ~\anaconda3\Lib\site-packages\pandas\core\groupby\ops.py:344, in WrappedCythonOp._cython_op_ndim_compat(self, values, min_count, ngroups, comp_ids, mask, result_mask, **kwargs)

```

341     # otherwise we have OHLC
342     return res.T
--> 344 return self._call_cython_op(
345     values,
346     min_count=min_count,
347     ngroups=ngroups,
348     comp_ids=comp_ids,
349     mask=mask,
350     result_mask=result_mask,
351     **kwargs,
352 )

```

File ~\anaconda3\Lib\site-packages\pandas\core\groupby\ops.py:402, in WrappedCythonOp._call_cython_op(self, values, min_count, ngroups, comp_ids, mask, result_mask, **kwargs)

```

400 out_shape = self._get_output_shape(ngroups, values)
401 func = self._get_cython_function(self.kind, self.how,
values.dtype, is_numeric)
--> 402 values = self._get_cython_vals(values)
403 out_dtype = self._get_out_dtype(values.dtype)
405 result = maybe_fill(np.empty(out_shape, dtype=out_dtype))

```

File ~\anaconda3\Lib\site-packages\pandas\core\groupby\ops.py:230, in WrappedCythonOp._get_cython_vals(self, values)

```

224 how = self.how
226 if how in ["median", "std", "sem", "skew"]:
227     # median only has a float64 implementation
228     # We should only get here with is_numeric, as non-numeric
cases

```



```

229     # should raise in _get_cython_function
--> 230     values = ensure_float64(values)
232 elif values.dtype.kind in "iu":
233     if how in ["var", "mean"] or (
234         self.kind == "transform" and self.has_dropped_na
235     ):
236         # has_dropped_na check need for
test_null_group_str_transformer
237         # result may still include NaN, so we have to cast

```

File pandas_libs\\algos_common_helper.pxi:42, in
pandas._libs.algos.ensure_float64()

ValueError: could not convert string to float: 'The Shawshank
Redemption'

genres.std(numeric_only=True) # similar as last

	Runtime	IMDB_Rating	No_of_Votes	Gross
Metascore				
Genre				
Action	28.500706	0.304258	432946.814748	2.256724e+08
12.421252				
Adventure	33.317320	0.229781	301188.347642	1.697543e+08
12.345393				
Animation	14.530471	0.253221	262173.231571	2.091840e+08
8.813646				
Biography	25.514466	0.267140	271284.191372	1.363251e+08
11.028187				
Comedy	22.946213	0.228771	188653.570564	1.946513e+08
11.829160				
Crime	27.689231	0.335477	373999.730656	1.571191e+08
13.099102				
Drama	27.740490	0.267229	305554.162841	2.201164e+08
12.744687				
Family	10.606602	0.000000	137008.302816	3.048412e+08
16.970563				
Fantasy	12.727922	0.141421	22179.111299	7.606861e+07
NaN				
Film-Noir	4.000000	0.152753	54649.083277	7.048472e+07
1.527525				
Horror	13.604812	0.311302	234883.508691	9.965017e+07
15.362291				
Mystery	14.475423	0.310791	404621.915297	1.567524e+08
18.604435				
Thriller	NaN	NaN	NaN	NaN
NaN				
Western	17.153717	0.420317	263489.554280	1.230626e+07
9.032349				

```
# find the top 3 genres by total earning
movies.groupby('Genre').sum()
```

	Series_Title \
Genre	
Action	The Dark KnightThe Lord of the Rings: The Retu...
Adventure	InterstellarBack to the FutureInglourious Bast...
Animation	Sen to Chihiro no kamikakushiThe Lion KingHota...
Biography	Schindler's ListGoodfellasHamiltonThe Intoucha...
Comedy	GisaengchungLa vita è bellaModern TimesCity Li...
Crime	The GodfatherThe Godfather: Part II12 Angry Me...
Drama	The Shawshank RedemptionFight ClubForrest Gump...
Family	E.T. the Extra-TerrestrialWilly Wonka & the Ch...
Fantasy	Das Cabinet des Dr. CaligariNosferatu
Film-Noir	The Third ManThe Maltese FalconShadow of a Doubt
Horror	PsychoAlienThe ThingThe ExorcistNight of the L...
Mystery	MementoRear WindowVertigoShutter IslandKahaani...
Thriller	Wait Until Dark
Western	Il buono, il brutto, il cattivoOnce Upon a Tim...

	Released_Year	Runtime
\ Genre		
Action	2008200320102001200219991980197719621954200019...	22196
Adventure	2014198520091981196819621959201319751963194819...	9656
Animation	2001199419882016201820172008199719952019200920...	8166
Biography	1993199020202011200220171995198420182013201320...	11970
Comedy	2019199719361931200919641940200120001973196019...	17380
Crime	1972197419571994200219991995199120192006199519...	13524
Drama	1994199919941975202019981946201420061998198819...	36049
Family	19821971	215
Fantasy	19201922	170
Film-Noir	194919411943	312
Horror	19601979198219731968196120171978193320042001	1123
Mystery	200019541958201020121995197219381988201219981997	1429
Thriller	1967	108
Western	1966196819651976	593

Director \ Genre	IMDB_Rating	
Action	1367.3	Christopher NolanPeter JacksonChristopher Nola...
Adventure	571.5	Christopher NolanRobert ZemeckisQuentin Tarant...
Animation	650.3	Hayao MiyazakiRoger AllersIsao TakahataMakoto ...
Biography	698.6	Steven SpielbergMartin ScorseseThomas KailOliv...
Comedy	1224.7	Bong Joon HoRoberto BenigniCharles ChaplinChar...
Crime	857.8	Francis Ford CoppolaFrancis Ford CoppolaSidney...
Drama	2299.7	Frank DarabontDavid FincherRobert ZemeckisMilo...
Family	15.6	Steven SpielbergMel Stuart
Fantasy	16.0	Robert WieneF.W. Murnau
Film-Noir	23.9	Carol ReedJohn HustonAlfred Hitchcock
Horror	87.0	Alfred HitchcockRidley ScottJohn CarpenterWill...
Mystery	95.7	Christopher NolanAlfred HitchcockAlfred Hitchc...
Thriller	7.8	Terence Young
Western	33.4	Sergio LeoneSergio LeoneSergio LeoneClint East...

No_of_Votes \ Genre	Star1
Action	Christian BaleElijah WoodLeonardo DiCaprioElij... 72282412
Adventure	Matthew McConaugheyMichael J. FoxBrad PittJürg... 22576163
Animation	Daveigh ChaseRob MinkoffTsutomu TatsumiRyûnosu... 21978630
Biography	Liam NeesonRobert De NiroLin-Manuel MirandaÉri... 24006844
Comedy	Kang-ho SongRoberto BenigniCharles ChaplinChar... 27620327
Crime	Marlon BrandoAl PacinoHenry FondaJohn Travolta...

```

33533615
Drama      Tim RobbinsBrad PittTom HanksJack NicholsonSur...
61367304
Family      Henry ThomasGene Wilder
551221
Fantasy      Werner KraussMax Schreck
146222
Film-Noir    Orson WellesHumphrey BogartTeresa Wright
367215
Horror      Anthony PerkinsSigourney WeaverKurt RussellEll...
3742556
Mystery      Guy PearceJames StewartJames StewartLeonardo D...
4203004
Thriller      Audrey Hepburn
27733
Western      Clint EastwoodHenry FondaClint EastwoodClint E...
1289665

```

	Gross	Metascore
Genre		
Action	3.263226e+10	10499.0
Adventure	9.496922e+09	5020.0
Animation	1.463147e+10	6082.0
Biography	8.276358e+09	6023.0
Comedy	1.566387e+10	9840.0
Crime	8.452632e+09	6706.0
Drama	3.540997e+10	19208.0
Family	4.391106e+08	158.0
Fantasy	7.827267e+08	0.0
Film-Noir	1.259105e+08	287.0
Horror	1.034649e+09	880.0
Mystery	1.256417e+09	633.0
Thriller	1.755074e+07	81.0
Western	5.822151e+07	313.0

iss dataframe me sare details aa gya but mujhe bass need earning bale se hai

```
movies.groupby('Genre').sum()['Gross']
```

Genre	
Action	3.263226e+10
Adventure	9.496922e+09
Animation	1.463147e+10
Biography	8.276358e+09
Comedy	1.566387e+10
Crime	8.452632e+09
Drama	3.540997e+10
Family	4.391106e+08
Fantasy	7.827267e+08
Film-Noir	1.259105e+08

```
Horror      1.034649e+09
Mystery     1.256417e+09
Thriller    1.755074e+07
Western     5.822151e+07
Name: Gross, dtype: float64
```

```
# ye abhi index ke basis pe sort hai to isse value ke basis pe sort
kar dete hain taki top earning movies dikh jaye
movies.groupby('Genre').sum()['Gross'].sort_values(ascending=False)
```

```
Genre
Drama      3.540997e+10
Action     3.263226e+10
Comedy     1.566387e+10
Animation  1.463147e+10
Adventure  9.496922e+09
Crime      8.452632e+09
Biography  8.276358e+09
Mystery    1.256417e+09
Horror     1.034649e+09
Fantasy    7.827267e+08
Family     4.391106e+08
Film-Noir  1.259105e+08
Western    5.822151e+07
Thriller   1.755074e+07
Name: Gross, dtype: float64
```

```
movies.groupby('Genre').sum()
['Gross'].sort_values(ascending=False).head(3) # top 3 movie jo ki
maximum revenue generate karta hai box office me
```

```
Genre
Drama      3.540997e+10
Action     3.263226e+10
Comedy     1.566387e+10
Name: Gross, dtype: float64
```

```
# m2
movies.groupby('Genre')
['Gross'].sum().sort_values(ascending=False).head(3) # yaha pahle
column nikale phie sum apply kar dye last method me just opposite
# kye the 2nd bala method jyada best hai qki time kam lagega hum
kisi specific column pe hi sum apply kar rahe hai but 1st method me
sare column pe sum
#apply karne ke baad required column ko select kar rahe the
```

```
Genre
Drama      3.540997e+10
Action     3.263226e+10
Comedy     1.566387e+10
Name: Gross, dtype: float64
```

```
# find the genre with heighest avg IMDB rating
```

```
movies.groupby('Genre')['IMDB_Rating'].mean()
```

```
Genre
Action      7.949419
Adventure    7.937500
Animation    7.930488
Biography    7.938636
Comedy       7.901290
Crime        8.016822
Drama        7.957439
Family       7.800000
Fantasy      8.000000
Film-Noir    7.966667
Horror       7.909091
Mystery      7.975000
Thriller     7.800000
Western      8.350000
Name: IMDB_Rating, dtype: float64
```

```
movies.groupby('Genre')
['IMDB_Rating'].mean().sort_values(ascending=False).head(1)
```

```
Genre
Western      8.35
Name: IMDB_Rating, dtype: float64
```

```
#find director with most popularity
```

```
movies.groupby('Director') # director ke basis pe group kar lete
hain and then no_of_votes jisko jyada hoga wo popular hoga
```

```
<pandas.core.groupby.generic.DataFrameGroupBy object at
0x00000290C55C7A40>
```

```
movies.groupby('Director')['No_of_Votes'].sum()
```

```
Director
Aamir Khan      168895
Aaron Sorkin     89896
Abdellatif Kechiche 138741
Abhishek Chaubey  27175
Abhishek Kapoor  32628
...
Zack Snyder     1233675
Zaza Urushadze   40382
Zoya Akhtar      99813
Çagan Irmak     78925
Ömer Faruk Sorak 56960
Name: No_of_Votes, Length: 548, dtype: int64
```

```
movies.groupby('Director')  
["No_of_Votes"].sum().sort_values(ascending=False).head(1)
```

```
Director  
Christopher Nolan    11578345  
Name: No_of_Votes, dtype: int64
```

```
#movies.groupby('Genre')['IMDB_Rating'].max()
```

```
Genre  
Action      9.0  
Adventure   8.6  
Animation   8.6  
Biography   8.9  
Comedy      8.6  
Crime       9.2  
Drama       9.3  
Family      7.8  
Fantasy     8.1  
Film-Noir   8.1  
Horror      8.5  
Mystery     8.4  
Thriller    7.8  
Western     8.8  
Name: IMDB_Rating, dtype: float64
```

```
# find number of movies done by each actor  
movies['Star1'].value_counts()
```

```
Star1  
Tom Hanks      12  
Robert De Niro 11  
Al Pacino      10  
Clint Eastwood 10  
Humphrey Bogart 9  
..            ..  
Preity Zinta   1  
Javier Bardem  1  
Ki-duk Kim     1  
Vladimir Garin 1  
Robert Donat   1  
Name: count, Length: 660, dtype: int64
```

```
# but mujhe groupby ke help se karna hai  
movies.groupby('Star1')['Series_Title'].count() # koi bhi column ko  
pakad sakte the bass no_of_row hi to count karana tha group karane ke  
baad
```

```
Star1  
Aamir Khan      7  
Aaron Taylor-Johnson 1
```

```

Abhay Deol      1
Abraham Attah   1
Adam Driver     1
..
Zbigniew Zamachowski  1
Zooey Deschanel  1
Çetin Tekindor  1
Eric Toledano    1
Ömer Faruk Sorak  1
Name: Series_Title, Length: 660, dtype: int64

```

last bale tarike se jo answer aaya tha wo sort tha on the basis of index and groupby bale method se jo aaya hai wo sorted hai on the basis of index

```

movies.groupby('Star1')
['Series_Title'].count().sort_values(ascending=False)

```

```

Star1
Tom Hanks      12
Robert De Niro 11
Clint Eastwood 10
Al Pacino      10
Leonardo DiCaprio  9
..
Glen Hansard    1
Giuseppe Battiston 1
Giulietta Masina 1
Gerardo Taracena 1
Ömer Faruk Sorak 1
Name: Series_Title, Length: 660, dtype: int64

```

*#har question me kuch steps follw kar rahe hain
step-1 : pahle groupby kar rahe hain kisi specific column pe jaise ki required hai
step-2 : and then required column ko nikal rahe hain
step-3 : phir koi nn koi aggregate function(min,max,sum etc.) use kar rahe hain*

```

# GroupBy Attributes and Methods
# find total number of groups -> len
# find items in each group -> size
# first()/last() -> nth item
# get_group -> vs filtering
# groups
# describe
# sample
# nunique

```

```

# find total number of groups -> len
movies.groupby('Genre')

```



```
<pandas.core.groupby.generic.DataFrameGroupBy object at 0x00000207A42C7890>
```

```
len(movies.groupby('Genre'))
```

```
14
```

```
#m-2
```

```
movies['Genre'].nunique()
```

```
14
```

```
# find items in each group -> size
```

```
movies.groupby('Genre').size()    # tells ki har group me kitne row hai (index ke basis pe sort hoga)
```

```
Genre
Action      172
Adventure    72
Animation    82
Biography    88
Comedy       155
Crime        107
Drama        289
Family        2
Fantasy       2
Film-Noir     3
Horror        11
Mystery       12
Thriller       1
Western        4
dtype: int64
```

```
#m2
```

```
movies['Genre'].value_counts()     #(value ke basis pe sort hoga)
```

```
Genre
Drama      289
Action     172
Comedy     155
Crime      107
Biography   88
Animation   82
Adventure   72
Mystery     12
Horror      11
Western      4
Film-Noir    3
Fantasy      2
Family        2
```

```
Thriller      1
Name: count, dtype: int64
```

```
# first()/last() -> nth item
genres = movies.groupby('Genre')
genres
```

```
<pandas.core.groupby.generic.DataFrameGroupBy object at
0x0000023A99EA51F0>
```

```
genres.first()    # har genres ki first movies aa jayegi
```

	Series_Title	Released_Year	Runtime	\
Genre				
Action	The Dark Knight	2008	152	
Adventure	Interstellar	2014	169	
Animation	Sen to Chihiro no kamikakushi	2001	125	
Biography	Schindler's List	1993	195	
Comedy	Gisaengchung	2019	132	
Crime	The Godfather	1972	175	
Drama	The Shawshank Redemption	1994	142	
Family	E.T. the Extra-Terrestrial	1982	115	
Fantasy	Das Cabinet des Dr. Caligari	1920	76	
Film-Noir	The Third Man	1949	104	
Horror	Psycho	1960	109	
Mystery	Memento	2000	113	
Thriller	Wait Until Dark	1967	108	
Western	Il buono, il brutto, il cattivo	1966	161	

	IMDB_Rating	Director	Star1	\
Genre				
Action	9.0	Christopher Nolan	Christian Bale	
Adventure	8.6	Christopher Nolan	Matthew McConaughey	
Animation	8.6	Hayao Miyazaki	Daveigh Chase	
Biography	8.9	Steven Spielberg	Liam Neeson	
Comedy	8.6	Bong Joon Ho	Kang-ho Song	
Crime	9.2	Francis Ford Coppola	Marlon Brando	
Drama	9.3	Frank Darabont	Tim Robbins	
Family	7.8	Steven Spielberg	Henry Thomas	
Fantasy	8.1	Robert Wiene	Werner Krauss	
Film-Noir	8.1	Carol Reed	Orson Welles	
Horror	8.5	Alfred Hitchcock	Anthony Perkins	
Mystery	8.4	Christopher Nolan	Guy Pearce	
Thriller	7.8	Terence Young	Audrey Hepburn	
Western	8.8	Sergio Leone	Clint Eastwood	

	No_of_Votes	Gross	Metascore
Genre			
Action	2303232	534858444.0	84.0

Adventure	1512360	188020017.0	74.0
Animation	651376	10055859.0	96.0
Biography	1213505	96898818.0	94.0
Comedy	552778	53367844.0	96.0
Crime	1620367	134966411.0	100.0
Drama	2343110	28341469.0	80.0
Family	372490	435110554.0	91.0
Fantasy	57428	337574718.0	NaN
Film-Noir	158731	449191.0	97.0
Horror	604211	32000000.0	97.0
Mystery	1125712	25544867.0	80.0
Thriller	27733	17550741.0	81.0
Western	688390	6100000.0	90.0

genres.last() *# har group ka last movies aa jayega*

Series_Title Released_Year Runtime			
\ Genre			
Action	Escape from Alcatraz	1979	112
Adventure	Kelly's Heroes	1970	144
Animation	The Jungle Book	1967	78
Biography	Midnight Express	1978	121
Comedy	Breakfast at Tiffany's	1961	115
Crime	The 39 Steps	1935	86
Drama	Lifeboat	1944	97
Family	Willy Wonka & the Chocolate Factory	1971	100
Fantasy	Nosferatu	1922	94
Film-Noir	Shadow of a Doubt	1943	108
Horror	The Others	2001	101
Mystery	Lost Highway	1997	134
Thriller	Wait Until Dark	1967	108
Western	The Outlaw Josey Wales	1976	135
IMDB_Rating Director Star1			
No_of_Votes \			

Genre			
Action	7.6	Don Siegel	Clint Eastwood
121731			
Adventure	7.6	Brian G. Hutton	Clint Eastwood
45338			
Animation	7.6	Wolfgang Reitherman	Phil Harris
166409			
Biography	7.6	Alan Parker	Brad Davis
73662			
Comedy	7.6	Blake Edwards	Audrey Hepburn
166544			
Crime	7.6	Alfred Hitchcock	Robert Donat
51853			
Drama	7.6	Alfred Hitchcock	Tallulah Bankhead
26471			
Family	7.8	Mel Stuart	Gene Wilder
178731			
Fantasy	7.9	F.W. Murnau	Max Schreck
88794			
Film-Noir	7.8	Alfred Hitchcock	Teresa Wright
59556			
Horror	7.6	Alejandro Amenábar	Nicole Kidman
337651			
Mystery	7.6	David Lynch	Bill Pullman
131101			
Thriller	7.8	Terence Young	Audrey Hepburn
27733			
Western	7.8	Clint Eastwood	Clint Eastwood
65659			

	Gross	Metascore
Genre		
Action	430000000.0	76.0
Adventure	1378435.0	50.0
Animation	141843612.0	65.0
Biography	35000000.0	59.0
Comedy	679874270.0	76.0
Crime	302787539.0	93.0
Drama	852142728.0	78.0
Family	4000000.0	67.0
Fantasy	445151978.0	NaN
Film-Noir	123353292.0	94.0
Horror	96522687.0	74.0
Mystery	3796699.0	52.0
Thriller	17550741.0	81.0
Western	31800000.0	69.0

genres.nth(6) # har group ka 7th movies (qki indexing 0 se) nikal
ke de dega aap observ karenge ki sare group available nahi hai qki jo

group

#available nahi hai usme 7 movie hai hi nahi

	Series_Title	Released_Year
Runtime \		
16	Star Wars: Episode V - The Empire Strikes Back	1980
124		
27	Se7en	1995
127		
32	It's a Wonderful Life	1946
130		
66	WALL·E	2008
98		
83	The Great Dictator	1940
125		
102	Braveheart	1995
178		
118	North by Northwest	1959
136		
420	Sleuth	1972
138		
724	Get Out	2017
104		

	Genre	IMDB_Rating	Director	Star1 \
16	Action	8.7	Irvin Kershner	Mark Hamill
27	Crime	8.6	David Fincher	Morgan Freeman
32	Drama	8.6	Frank Capra	James Stewart
66	Animation	8.4	Andrew Stanton	Ben Burtt
83	Comedy	8.4	Charles Chaplin	Charles Chaplin
102	Biography	8.3	Mel Gibson	Mel Gibson
118	Adventure	8.3	Alfred Hitchcock	Cary Grant
420	Mystery	8.0	Joseph L. Mankiewicz	Laurence Olivier
724	Horror	7.7	Jordan Peele	Daniel Kaluuya

	No_of_Votes	Gross	Metascore
16	1159315	290475067.0	82.0
27	1445096	100125643.0	65.0
32	405801	82385199.0	89.0
66	999790	223808164.0	95.0
83	203150	288475.0	NaN
102	959181	75600000.0	68.0
118	299198	13275000.0	98.0
420	44748	4081254.0	NaN
724	492851	176040665.0	85.0

get_group -> vs filtering => iske help se kisi particular group ko dekh sakte hain

genres.get_group('Horror')

IMDB_Rating \	Series_Title	Released_Year	Runtime	Genre
49	Psycho	1960	109	Horror
8.5				
75	Alien	1979	117	Horror
8.4				
271	The Thing	1982	109	Horror
8.1				
419	The Exorcist	1973	122	Horror
8.0				
544	Night of the Living Dead	1968	96	Horror
7.9				
707	The Innocents	1961	100	Horror
7.8				
724	Get Out	2017	104	Horror
7.7				
844	Halloween	1978	91	Horror
7.7				
876	The Invisible Man	1933	71	Horror
7.7				
932	Saw	2004	103	Horror
7.6				
948	The Others	2001	101	Horror
7.6				
Metascore	Director	Star1	No_of_Votes	Gross
49	Alfred Hitchcock	Anthony Perkins	604211	32000000.0
97.0				
75	Ridley Scott	Sigourney Weaver	787806	78900000.0
89.0				
271	John Carpenter	Kurt Russell	371271	13782838.0
57.0				
419	William Friedkin	Ellen Burstyn	362393	232906145.0
81.0				
544	George A. Romero	Duane Jones	116557	89029.0
89.0				
707	Jack Clayton	Deborah Kerr	27007	2616000.0
88.0				
724	Jordan Peele	Daniel Kaluuya	492851	176040665.0
85.0				
844	John Carpenter	Donald Pleasence	233106	47000000.0
87.0				
876	James Whale	Claude Rains	30683	298791505.0
87.0				
932	James Wan	Cary Elwes	379020	56000369.0
46.0				
948	Alejandro Amenábar	Nicole Kidman	337651	96522687.0
74.0				

```
genres.get_group('Fantasy')
```

	Series_Title	Released_Year	Runtime	Genre
321	Das Cabinet des Dr. Caligari	1920	76	Fantasy
568	Nosferatu	1922	94	Fantasy

	IMDB_Rating	Director	Star1	No_of_Votes
Gross				
321	8.1	Robert Wiene	Werner Krauss	57428
337574718.0				
568	7.9	F.W. Murnau	Max Schreck	88794
445151978.0				

	Metascore
321	NaN
568	NaN

```
#m2
```

```
movies[movies['Genre']=='Fantasy']
```

	Series_Title	Released_Year	Runtime	Genre
321	Das Cabinet des Dr. Caligari	1920	76	Fantasy
568	Nosferatu	1922	94	Fantasy

	IMDB_Rating	Director	Star1	No_of_Votes
Gross				
321	8.1	Robert Wiene	Werner Krauss	57428
337574718.0				
568	7.9	F.W. Murnau	Max Schreck	88794
445151978.0				

	Metascore
321	NaN
568	NaN

```
# groups
```

```
genres.groups    # ye return karega index(genres) name ke sath movies  
ke sare index   jaise Action categories me kon kon se index ke movies  
hain
```

```
{'Action': [2, 5, 8, 10, 13, 14, 16, 29, 30, 31, 39, 42, 44, 55, 57,  
59, 60, 63, 68, 72, 106, 109, 129, 130, 134, 140, 142, 144, 152, 155,  
160, 161, 166, 168, 171, 172, 177, 181, 194, 201, 202, 216, 217, 223,  
224, 236, 241, 262, 275, 294, 308, 320, 325, 326, 331, 337, 339, 340,  
343, 345, 348, 351, 353, 356, 357, 362, 368, 369, 375, 376, 390, 410,  
431, 436, 473, 477, 479, 482, 488, 493, 496, 502, 507, 511, 532, 535,  
540, 543, 564, 569, 570, 573, 577, 582, 583, 602, 605, 608, 615,  
623, ...], 'Adventure': [21, 47, 93, 110, 114, 116, 118, 137, 178,  
179, 191, 193, 209, 226, 231, 247, 267, 273, 281, 300, 301, 304, 306,  
323, 329, 361, 366, 377, 402, 406, 415, 426, 458, 470, 497, 498, 506,  
513, 514, 537, 549, 552, 553, 566, 576, 604, 609, 618, 638, 647, 675,
```

```

681, 686, 692, 711, 713, 739, 755, 781, 797, 798, 851, 873, 884, 912,
919, 947, 957, 964, 966, 984, 991], 'Animation': [23, 43, 46, 56, 58,
61, 66, 70, 101, 135, 146, 151, 158, 170, 197, 205, 211, 213, 219,
229, 230, 242, 245, 246, 270, 330, 332, 358, 367, 378, 386, 389, 394,
395, 399, 401, 405, 409, 469, 499, 510, 516, 518, 522, 578, 586, 592,
595, 596, 599, 633, 640, 643, 651, 665, 672, 694, 728, 740, 741, 744,
756, 758, 761, 771, 783, 796, 799, 822, 828, 843, 875, 891, 892, 902,
906, 920, 956, 971, 976, 986, 992], 'Biography': [7, 15, 18, 35, 38,
54, 102, 107, 131, 139, 147, 157, 159, 173, 176, 212, 215, 218, 228,
235, 243, 263, 276, 282, 290, 298, 317, 328, 338, 342, 346, 359, 360,
365, 372, 373, 385, 411, 416, 418, 424, 429, 484, 525, 536, 542, 545,
575, 579, 587, 600, 606, 614, 622, 632, 635, 644, 649, 650, 657, 671,
673, 684, 729, 748, 753, 757, 759, 766, 770, 779, 809, 810, 815, 820,
831, 849, 858, 877, 882, 897, 910, 915, 923, 940, 949, 952, 987],
'Comedy': [19, 26, 51, 52, 64, 78, 83, 95, 96, 112, 117, 120, 127,
128, 132, 153, 169, 183, 192, 204, 207, 208, 214, 221, 233, 238, 240,
250, 251, 252, 256, 261, 266, 277, 284, 311, 313, 316, 318, 322, 327,
374, 379, 381, 392, 396, 403, 413, 414, 417, 427, 435, 445, 446, 449,
455, 459, 460, 463, 464, 466, 471, 472, 475, 481, 490, 494, 500, 503,
509, 526, 528, 530, 531, 533, 538, 539, 541, 547, 557, 558, 562, 563,
565, 574, 591, 593, 594, 598, 613, 626, 630, 660, 662, 667, 679, 680,
683, 687, 701, ...], 'Crime': [1, 3, 4, 6, 22, 25, 27, 28, 33, 37, 41,
71, 77, 79, 86, 87, 103, 108, 111, 113, 123, 125, 133, 136, 162, 163,
164, 165, 180, 186, 187, 189, 198, 222, 232, 239, 255, 257, 287, 288,
299, 305, 335, 363, 364, 380, 384, 397, 437, 438, 441, 442, 444, 450,
451, 465, 474, 480, 485, 487, 505, 512, 519, 520, 523, 527, 546, 556,
560, 584, 597, 603, 607, 611, 621, 639, 653, 664, 669, 676, 695, 708,
723, 762, 763, 767, 775, 791, 795, 802, 811, 823, 827, 833, 885, 895,
921, 922, 926, 938, ...], 'Drama': [0, 9, 11, 17, 20, 24, 32, 34, 36,
40, 45, 50, 53, 62, 65, 67, 73, 74, 76, 80, 82, 84, 85, 88, 89, 90,
91, 92, 94, 97, 98, 99, 100, 104, 105, 121, 122, 124, 126, 138, 141,
143, 148, 149, 150, 154, 156, 167, 174, 175, 182, 184, 185, 188, 190,
195, 196, 199, 200, 203, 206, 210, 225, 227, 234, 237, 244, 248, 249,
253, 254, 258, 259, 260, 264, 265, 268, 269, 272, 274, 278, 279, 280,
283, 285, 286, 289, 291, 292, 293, 295, 296, 297, 302, 303, 307, 310,
312, 314, 315, ...], 'Family': [688, 698], 'Fantasy': [321, 568],
'Film-Noir': [309, 456, 712], 'Horror': [49, 75, 271, 419, 544, 707,
724, 844, 876, 932, 948], 'Mystery': [69, 81, 119, 145, 220, 393, 420,
714, 829, 899, 959, 961], 'Thriller': [700], 'Western': [12, 48, 115,
691]]}

```

```
# describe
```

```
genres.describe()
```

	Runtime						
\	count	mean	std	min	25%	50%	75%
max							
Genre							

Action	172.0	129.046512	28.500706	45.0	110.75	127.5	143.25
321.0							
Adventure	72.0	134.111111	33.317320	88.0	109.00	127.0	149.00
228.0							
Animation	82.0	99.585366	14.530471	71.0	90.00	99.5	106.75
137.0							
Biography	88.0	136.022727	25.514466	93.0	120.00	129.0	146.25
209.0							
Comedy	155.0	112.129032	22.946213	68.0	96.00	106.0	124.50
188.0							
Crime	107.0	126.392523	27.689231	80.0	106.50	122.0	141.50
229.0							
Drama	289.0	124.737024	27.740490	64.0	105.00	121.0	137.00
242.0							
Family	2.0	107.500000	10.606602	100.0	103.75	107.5	111.25
115.0							
Fantasy	2.0	85.000000	12.727922	76.0	80.50	85.0	89.50
94.0							
Film-Noir	3.0	104.000000	4.000000	100.0	102.00	104.0	106.00
108.0							
Horror	11.0	102.090909	13.604812	71.0	98.00	103.0	109.00
122.0							
Mystery	12.0	119.083333	14.475423	96.0	110.75	117.5	130.25
138.0							
Thriller	1.0	108.000000	NaN	108.0	108.00	108.0	108.00
108.0							
Western	4.0	148.250000	17.153717	132.0	134.25	148.0	162.00
165.0							

Metascore	IMDB_Rating			Gross		
	count	mean	...	75%	max	
count						
Genre			...			
Action	172.0	7.949419	...	2.674437e+08	936662225.0	143.0
Adventure	72.0	7.937500	...	1.998070e+08	874211619.0	64.0
Animation	82.0	7.930488	...	2.520612e+08	873839108.0	75.0
Biography	88.0	7.938636	...	9.829924e+07	753585104.0	79.0
Comedy	155.0	7.901290	...	8.107809e+07	886752933.0	125.0
Crime	107.0	8.016822	...	7.102163e+07	790482117.0	87.0
Drama	289.0	7.957439	...	1.164461e+08	924558264.0	241.0

Family	2.0	7.800000	...	3.273329e+08	435110554.0
2.0					
Fantasy	2.0	8.000000	...	4.182577e+08	445151978.0
0.0					
Film-Noir	3.0	7.966667	...	6.273068e+07	123353292.0
3.0					
Horror	11.0	7.909091	...	1.362817e+08	298791505.0
11.0					
Mystery	12.0	7.975000	...	1.310949e+08	474203697.0
8.0					
Thriller	1.0	7.800000	...	1.755074e+07	17550741.0
1.0					
Western	4.0	8.350000	...	1.920000e+07	31800000.0
4.0					

	mean	std	min	25%	50%	75%	max
Genre							
Action	73.419580	12.421252	33.0	65.00	74.0	82.00	98.0
Adventure	78.437500	12.345393	41.0	69.75	80.5	87.25	100.0
Animation	81.093333	8.813646	61.0	75.00	82.0	87.50	96.0
Biography	76.240506	11.028187	48.0	70.50	76.0	84.50	97.0
Comedy	78.720000	11.829160	45.0	72.00	79.0	88.00	99.0
Crime	77.080460	13.099102	47.0	69.50	77.0	87.00	100.0
Drama	79.701245	12.744687	28.0	72.00	82.0	89.00	100.0
Family	79.000000	16.970563	67.0	73.00	79.0	85.00	91.0
Fantasy	NaN	NaN	NaN	NaN	NaN	NaN	NaN
Film-Noir	95.666667	1.527525	94.0	95.00	96.0	96.50	97.0
Horror	80.000000	15.362291	46.0	77.50	87.0	88.50	97.0
Mystery	79.125000	18.604435	52.0	65.25	77.0	98.50	100.0
Thriller	81.000000	NaN	81.0	81.00	81.0	81.00	81.0
Western	78.250000	9.032349	69.0	72.75	77.0	82.50	90.0

[14 rows x 40 columns]

sample

genres.sample() # har genres ka koi ek random movies dikha dega

	Series_Title	Released_Year	Runtime
Genre \			
702	Bonnie and Clyde	1967	111
Action			
458	The Wizard of Oz	1939	102
Adventure			
358	Persepolis	2007	96
Animation			
622	American Gangster	2007	157
Biography			
481	The Artist	2011	100
Comedy			

222		Prisoners	2013	153	
Crime					
988		Close Encounters of the Third Kind	1977	138	
Drama					
688		E.T. the Extra-Terrestrial	1982	115	
Family					
568		Nosferatu	1922	94	
Fantasy					
456		The Maltese Falcon	1941	100	Film-
Noir					
844		Halloween	1978	91	
Horror					
69		Memento	2000	113	
Mystery					
700		Wait Until Dark	1967	108	
Thriller					
12		Il buono, il brutto, il cattivo	1966	161	
Western					
	IMDB_Rating	Director	Star1	No_of_Votes	
\					
702	7.8	Arthur Penn	Warren Beatty	102415	
458	8.0	Victor Fleming	George Cukor	371379	
358	8.0	Vincent Paronnaud	Marjane Satrapi	88656	
622	7.8	Ridley Scott	Denzel Washington	392449	
481	7.9	Michel Hazanavicius	Jean Dujardin	230624	
222	8.1	Denis Villeneuve	Hugh Jackman	601149	
988	7.6	Steven Spielberg	Richard Dreyfuss	184966	
688	7.8	Steven Spielberg	Henry Thomas	372490	
568	7.9	F.W. Murnau	Max Schreck	88794	
456	8.0	John Huston	Humphrey Bogart	148928	
844	7.7	John Carpenter	Donald Pleasence	233106	
69	8.4	Christopher Nolan	Guy Pearce	1125712	
700	7.8	Terence Young	Audrey Hepburn	27733	
12	8.8	Sergio Leone	Clint Eastwood	688390	
	Gross	Metascore			

702	909838190.0	86.0
458	2076020.0	92.0
358	4445756.0	90.0
622	130164645.0	76.0
481	44671682.0	89.0
222	61002302.0	70.0
988	132088635.0	90.0
688	435110554.0	91.0
568	445151978.0	NaN
456	2108060.0	96.0
844	47000000.0	87.0
69	25544867.0	80.0
700	17550741.0	81.0
12	6100000.0	90.0

genres.sample(2) # abb do do random movies dikhega abhi error islye aa raha hai qki

```
-----
-----
ValueError                                Traceback (most recent call
last)
```

```
Cell In[48], line 1
----> 1 genres.sample(2)
```

```
File ~\anaconda3\Lib\site-packages\pandas\core\groupby\
groupby.py:5780, in GroupBy.sample(self, n, frac, replace, weights,
random_state)
```

```
    5777         assert frac is not None
    5778         sample_size = round(frac * group_size)
-> 5780         grp_sample = sample.sample(
    5781             group_size,
    5782             size=sample_size,
    5783             replace=replace,
    5784             weights=None if weights is None else
weights_arr[grp_indices],
    5785             random_state=random_state,
    5786         )
    5787         sampled_indices.append(grp_indices[grp_sample])
    5789 sampled_indices = np.concatenate(sampled_indices)
```

```
File ~\anaconda3\Lib\site-packages\pandas\core\sample.py:152, in
sample(obj_len, size, replace, weights, random_state)
```

```
    149         else:
    150             raise ValueError("Invalid weights: weights sum to
zero")
--> 152 return random_state.choice(obj_len, size=size,
replace=replace, p=weights).astype(
    153     np.intp, copy=False
    154 )
```

```
File numpy\\random\\mtrand.pyx:1001, in
numpy.random.mtrand.RandomState.choice()
```

```
ValueError: Cannot take a larger sample than population when
'replace=False'
```

```
genres.sample(2,replace=True) # abb jisme ek hi row hai wo repeate ho
jayega
```

	Series_Title	Released_Year	Runtime	
Genre \				
774	3:10 to Yuma	2007	122	
Action				
144	Warrior	2011	140	
Action				
884	The Peanut Butter Falcon	2019	97	
Adventure				
329	The Martian	2015	144	
Adventure				
43	The Lion King	1994	88	
Animation				
740	Wreck-It Ralph	2012	101	
Animation				
657	Ed Wood	1994	127	
Biography				
373	Cinderella Man	2005	144	
Biography				
463	Knives Out	2019	130	
Comedy				
835	The Purple Rose of Cairo	1985	82	
Comedy				
669	Boyz n the Hood	1991	112	
Crime				
28	The Silence of the Lambs	1991	118	
Crime				
443	La Strada	1954	108	
Drama				
73	The Shining	1980	146	
Drama				
698	Willy Wonka & the Chocolate Factory	1971	100	
Family				
688	E.T. the Extra-Terrestrial	1982	115	
Family				
568	Nosferatu	1922	94	
Fantasy				
568	Nosferatu	1922	94	
Fantasy				
309	The Third Man	1949	104	Film-
Noir				

456	The Maltese Falcon	1941	100	Film-Noir
844	Halloween	1978	91	Horror
544	Night of the Living Dead	1968	96	Horror
145	Shutter Island	2010	138	Mystery
81	Rear Window	1954	112	Mystery
700	Wait Until Dark	1967	108	Thriller
700	Wait Until Dark	1967	108	Thriller
115	Per qualche dollaro in più	1965	132	Western
691	The Outlaw Josey Wales	1976	135	Western

	IMDB_Rating	Director	Star1	No_of_Votes \
774	7.7	James Mangold	Russell Crowe	288797
144	8.2	Gavin O'Connor	Tom Hardy	435950
884	7.6	Tyler Nilson	Michael Schwartz	66346
329	8.0	Ridley Scott	Matt Damon	760094
43	8.5	Roger Allers	Rob Minkoff	942045
740	7.7	Rich Moore	John C. Reilly	380195
657	7.8	Tim Burton	Johnny Depp	164937
373	8.0	Ron Howard	Russell Crowe	176151
463	7.9	Rian Johnson	Daniel Craig	454203
835	7.7	Woody Allen	Mia Farrow	47102
669	7.8	John Singleton	Cuba Gooding Jr.	126082
28	8.6	Jonathan Demme	Jodie Foster	1270197
443	8.0	Federico Fellini	Anthony Quinn	58314
73	8.4	Stanley Kubrick	Jack Nicholson	898237
698	7.8	Mel Stuart	Gene Wilder	178731
688	7.8	Steven Spielberg	Henry Thomas	372490
568	7.9	F.W. Murnau	Max Schreck	88794
568	7.9	F.W. Murnau	Max Schreck	88794
309	8.1	Carol Reed	Orson Welles	158731
456	8.0	John Huston	Humphrey Bogart	148928
844	7.7	John Carpenter	Donald Pleasence	233106
544	7.9	George A. Romero	Duane Jones	116557
145	8.2	Martin Scorsese	Leonardo DiCaprio	1129894
81	8.4	Alfred Hitchcock	James Stewart	444074
700	7.8	Terence Young	Audrey Hepburn	27733
700	7.8	Terence Young	Audrey Hepburn	27733
115	8.3	Sergio Leone	Clint Eastwood	232772
691	7.8	Clint Eastwood	Clint Eastwood	65659

	Gross	Metascore
774	53606916.0	76.0
144	13657115.0	71.0
884	13122642.0	70.0
329	228433663.0	80.0
43	422783777.0	88.0
740	189422889.0	72.0
657	5887457.0	70.0
373	61649911.0	69.0
463	165359751.0	82.0
835	10631333.0	75.0
669	57504069.0	76.0
28	130742922.0	85.0
443	101832153.0	NaN
73	44017374.0	66.0
698	4000000.0	67.0
688	435110554.0	91.0
568	445151978.0	NaN
568	445151978.0	NaN
309	449191.0	97.0
456	2108060.0	96.0
844	47000000.0	87.0
544	89029.0	89.0
145	128012934.0	63.0
81	36764313.0	100.0
700	17550741.0	81.0
700	17550741.0	81.0
115	15000000.0	74.0
691	31800000.0	69.0

genres.nunique() # har genres me kitna unique data hai action bale
 movies me released_year me 61 unique value hai matlab ki bahut ssare
 movies

#aaise honge jo ki same year me release
 hua hoga

Star1 \ Genre	Series_Title	Released_Year	Runtime	IMDB_Rating	Director	
Action 121		172	61	78	15	123
Adventure 59		72	49	58	10	59
Animation 77		82	35	41	11	51
Biography 72		88	44	56	13	76
Comedy 133		155	72	70	11	113

Crime	106	56	65	14	86
85					
Drama	289	83	95	14	211
250					
Family	2	2	2	1	2
2					
Fantasy	2	2	2	2	2
2					
Film-Noir	3	3	3	3	3
3					
Horror	11	11	10	8	10
11					
Mystery	12	11	10	8	10
11					
Thriller	1	1	1	1	1
1					
Western	4	4	4	4	2
2					

	No_of_Votes	Gross	Metascore
Genre			
Action	172	172	50
Adventure	72	72	33
Animation	82	82	29
Biography	88	88	40
Comedy	155	155	44
Crime	107	107	39
Drama	288	287	52
Family	2	2	2
Fantasy	2	2	0
Film-Noir	3	3	3
Horror	11	11	9
Mystery	12	12	7
Thriller	1	1	1
Western	4	4	4


```

# agg method
# passing dict
genres.sum()

```


	Series_Title \
Genre	
Action	The Dark KnightThe Lord of the Rings: The Retu...
Adventure	InterstellarBack to the FutureInglourious Bast...
Animation	Sen to Chihiro no kamikakushiThe Lion KingHota...
Biography	Schindler's ListGoodfellasHamiltonThe Intoucha...
Comedy	GisaengchungLa vita è bellaModern TimesCity Li...
Crime	The GodfatherThe Godfather: Part III12 Angry Me...
Drama	The Shawshank RedemptionFight ClubForrest Gump...
Family	E.T. the Extra-TerrestrialWilly Wonka & the Ch...

Fantasy		Das Cabinet des Dr. Caligari	Nosferatu
Film-Noir	The Third Man	The Maltese Falcon	Shadow of a Doubt
Horror	Psycho	Alien	The Thing
Mystery	Memento	Rear Window	Vertigo
Thriller			Shutter Island
Western	Il buono, il brutto, il cattivo	Once Upon a Tim...	

		Released_Year	Runtime
\ Genre			
Action	2008200320102001200219991980197719621954200019...		22196
Adventure	2014198520091981196819621959201319751963194819...		9656
Animation	2001199419882016201820172008199719952019200920...		8166
Biography	1993199020202011200220171995198420182013201320...		11970
Comedy	2019199719361931200919641940200120001973196019...		17380
Crime	1972197419571994200219991995199120192006199519...		13524
Drama	1994199919941975202019981946201420061998198819...		36049
Family		19821971	215
Fantasy		19201922	170
Film-Noir		194919411943	312
Horror	19601979198219731968196120171978193320042001		1123
Mystery	200019541958201020121995197219381988201219981997		1429
Thriller		1967	108
Western		1966196819651976	593

		IMDB_Rating
Director \ Genre		
Action	1367.3	Christopher NolanPeter JacksonChristopher Nola...
Adventure	571.5	Christopher NolanRobert ZemeckisQuentin Tarant...
Animation	650.3	Hayao MiyazakiRoger AllersIsao TakahataMakoto ...
Biography	698.6	Steven SpielbergMartin ScorseseThomas

Kail0liv...					
Comedy	1224.7	Bong Joon Ho	Roberto Benigni	Charles	
ChaplinChar...					
Crime	857.8	Francis Ford Coppola	Francis Ford		
CoppolaSidney...					
Drama	2299.7	Frank Darabont	David Fincher	Robert	
ZemeckisMilo...					
Family	15.6			Steven Spielberg	Mel
Stuart					
Fantasy	16.0			Robert Wiene	F.W.
Murnau					
Film-Noir	23.9		Carol Reed	John Huston	Alfred
Hitchcock					
Horror	87.0	Alfred Hitchcock	Ridley Scott	John	
CarpenterWill...					
Mystery	95.7	Christopher Nolan	Alfred Hitchcock	Alfred	
Hitchc...					
Thriller	7.8				Terence
Young					
Western	33.4	Sergio Leone	Sergio Leone	Sergio Leone	Clint
East...					

Star1

No_of_Votes \
Genre

Action	Christian Bale	Elijah Wood	Leonardo DiCaprio	Elij...
72282412				
Adventure	Matthew McConaughey	Michael J. Fox	Brad Pitt	Jürg...
22576163				
Animation	Daveigh Chase	Rob Minkoff	Tsutomu Tatsumi	Ryûnosu...
21978630				
Biography	Liam Neeson	Robert De Niro	Lin-Manuel Miranda	Éri...
24006844				
Comedy	Kang-ho Song	Roberto Benigni	Charles Chaplin	Char...
27620327				
Crime	Marlon Brando	Al Pacino	Henry Fonda	John Travolta...
33533615				
Drama	Tim Robbins	Brad Pitt	Tom Hanks	Jack Nicholson
61367304				Sur...
Family			Henry Thomas	Gene Wilder
551221				
Fantasy			Werner Krauss	Max Schreck
146222				
Film-Noir	Orson Welles	Humphrey Bogart	Teresa Wright	
367215				
Horror	Anthony Perkins	Sigourney Weaver	Kurt Russell	Ell...
3742556				
Mystery	Guy Pearce	James Stewart	James Stewart	Leonardo D...

```
4203004
Thriller                                Audrey Hepburn
27733
Western    Clint EastwoodHenry FondaClint EastwoodClint E...
1289665
```

	Gross	Metascore
Genre		
Action	3.263226e+10	10499.0
Adventure	9.496922e+09	5020.0
Animation	1.463147e+10	6082.0
Biography	8.276358e+09	6023.0
Comedy	1.566387e+10	9840.0
Crime	8.452632e+09	6706.0
Drama	3.540997e+10	19208.0
Family	4.391106e+08	158.0
Fantasy	7.827267e+08	0.0
Film-Noir	1.259105e+08	287.0
Horror	1.034649e+09	880.0
Mystery	1.256417e+09	633.0
Thriller	1.755074e+07	81.0
Western	5.822151e+07	313.0

```
# agg method
# passing dict
# yadi aap particular column ke upar koi particular aggregate function
apply karna chahate ho to usse ek dictionary me pass kar do
genres.agg(
    {
        'Runtime': 'mean',
        'IMDB_Rating': 'mean',
        'No_of_Votes': 'sum',
        'Gross': 'sum',
        'Metascore': 'min'
    }
)
```

	Runtime	IMDB_Rating	No_of_Votes	Gross
Metascore				
Genre				
Action	129.046512	7.949419	72282412	3.263226e+10
33.0				
Adventure	134.111111	7.937500	22576163	9.496922e+09
41.0				
Animation	99.585366	7.930488	21978630	1.463147e+10
61.0				
Biography	136.022727	7.938636	24006844	8.276358e+09
48.0				
Comedy	112.129032	7.901290	27620327	1.566387e+10

45.0				
Crime	126.392523	8.016822	33533615	8.452632e+09
47.0				
Drama	124.737024	7.957439	61367304	3.540997e+10
28.0				
Family	107.500000	7.800000	551221	4.391106e+08
67.0				
Fantasy	85.000000	8.000000	146222	7.827267e+08
NaN				
Film-Noir	104.000000	7.966667	367215	1.259105e+08
94.0				
Horror	102.090909	7.909091	3742556	1.034649e+09
46.0				
Mystery	119.083333	7.975000	4203004	1.256417e+09
52.0				
Thriller	108.000000	7.800000	27733	1.755074e+07
81.0				
Western	148.250000	8.350000	1289665	5.822151e+07
69.0				

passing list

suppose har numerical column ke lye [min,max & mean] nikalna chahate hain

genres[['Runtime','IMDB_Rating','No_of_Votes','Gross','Metascore']].agg(['min','max','mean']) # jis bhi column pe multiple functions apply karna hai usse ek list me pass kar do

#and jo functions apply karna hai usse bhi ek list me pass kar do

	Runtime			IMDB_Rating		
No_of_Votes \	min	max	mean	min	max	mean
min						
Genre						
Action	45	321	129.046512	7.6	9.0	7.949419
25312						
Adventure	88	228	134.111111	7.6	8.6	7.937500
29999						
Animation	71	137	99.585366	7.6	8.6	7.930488
25229						
Biography	93	209	136.022727	7.6	8.9	7.938636
27254						
Comedy	68	188	112.129032	7.6	8.6	7.901290
26337						
Crime	80	229	126.392523	7.6	9.2	8.016822
27712						
Drama	64	242	124.737024	7.6	9.3	7.957439
25088						
Family	100	115	107.500000	7.8	7.8	7.800000

178731							
Fantasy	76	94	85.000000	7.9	8.1	8.000000	
57428							
Film-Noir	100	108	104.000000	7.8	8.1	7.966667	
59556							
Horror	71	122	102.090909	7.6	8.5	7.909091	
27007							
Mystery	96	138	119.083333	7.6	8.4	7.975000	
33982							
Thriller	108	108	108.000000	7.8	7.8	7.800000	
27733							
Western	132	165	148.250000	7.8	8.8	8.350000	
65659							

Gross				
\	max	mean	min	max
mean				
Genre				
Action	2303232	420246.581395	3296.0	936662225.0
1.897224e+08				
Adventure	1512360	313557.819444	61001.0	874211619.0
1.319017e+08				
Animation	999790	268032.073171	128985.0	873839108.0
1.784326e+08				
Biography	1213505	272805.045455	21877.0	753585104.0
9.404952e+07				
Comedy	939631	178195.658065	1305.0	886752933.0
1.010572e+08				
Crime	1826188	313398.271028	6013.0	790482117.0
7.899656e+07				
Drama	2343110	212343.612457	3600.0	924558264.0
1.225259e+08				
Family	372490	275610.500000	4000000.0	435110554.0
2.195553e+08				
Fantasy	88794	73111.000000	337574718.0	445151978.0
3.913633e+08				
Film-Noir	158731	122405.000000	449191.0	123353292.0
4.197018e+07				
Horror	787806	340232.363636	89029.0	298791505.0
9.405902e+07				
Mystery	1129894	350250.333333	1035953.0	474203697.0
1.047014e+08				
Thriller	27733	27733.000000	17550741.0	17550741.0
1.755074e+07				
Western	688390	322416.250000	5321508.0	31800000.0
1.455538e+07				

Metascore

	min	max	mean
Genre			
Action	33.0	98.0	73.419580
Adventure	41.0	100.0	78.437500
Animation	61.0	96.0	81.093333
Biography	48.0	97.0	76.240506
Comedy	45.0	99.0	78.720000
Crime	47.0	100.0	77.080460
Drama	28.0	100.0	79.701245
Family	67.0	91.0	79.000000
Fantasy	NaN	NaN	NaN
Film-Noir	94.0	97.0	95.666667
Horror	46.0	97.0	80.000000
Mystery	52.0	100.0	79.125000
Thriller	81.0	81.0	81.000000
Western	69.0	90.0	78.250000

adding both syntax

```
genres.agg(
    {
        'Runtime': ['mean', 'min'],
        'IMDB_Rating': 'mean',
        'No_of_Votes': ['sum', 'max'],
        'Gross': 'sum',
        'Metascore': 'min'
    }
)
```

abb sirf runtime me min and min value and no_of_votes me sum and max value dikhega

	Runtime		IMDB_Rating	No_of_Votes	
Gross \	mean	min	mean	sum	max
sum					
Genre					
Action	129.046512	45	7.949419	72282412	2303232
3.263226e+10					
Adventure	134.111111	88	7.937500	22576163	1512360
9.496922e+09					
Animation	99.585366	71	7.930488	21978630	999790
1.463147e+10					
Biography	136.022727	93	7.938636	24006844	1213505
8.276358e+09					
Comedy	112.129032	68	7.901290	27620327	939631
1.566387e+10					
Crime	126.392523	80	8.016822	33533615	1826188
8.452632e+09					
Drama	124.737024	64	7.957439	61367304	2343110
3.540997e+10					

Family	107.500000	100	7.800000	551221	372490
4.391106e+08					
Fantasy	85.000000	76	8.000000	146222	88794
7.827267e+08					
Film-Noir	104.000000	100	7.966667	367215	158731
1.259105e+08					
Horror	102.090909	71	7.909091	3742556	787806
1.034649e+09					
Mystery	119.083333	96	7.975000	4203004	1129894
1.256417e+09					
Thriller	108.000000	108	7.800000	27733	27733
1.755074e+07					
Western	148.250000	132	8.350000	1289665	688390
5.822151e+07					

	Metascore min
Genre	
Action	33.0
Adventure	41.0
Animation	61.0
Biography	48.0
Comedy	45.0
Crime	47.0
Drama	28.0
Family	67.0
Fantasy	NaN
Film-Noir	94.0
Horror	46.0
Mystery	52.0
Thriller	81.0
Western	69.0

```
# looping on groups
genres
```

```
<pandas.core.groupby.generic.DataFrameGroupBy object at 0x000001E5E9F2CA10>
```

```
for group,data in genres:    # group basically sare group ko specify
                             # karega and data uss group ke sare data ko
    print(type(group),type(data))
```

[illegible]

```

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

```

```

for group,data in genres:
    print(data)      # har group ke data ko print kar dega jab hum data
ko print karayenge

```

	Series_Title	Released_Year
Runtime \		
2	The Dark Knight	2008
152		
5	The Lord of the Rings: The Return of the King	2003
201		
8	Inception	2010
148		
10	The Lord of the Rings: The Fellowship of the Ring	2001
178		
13	The Lord of the Rings: The Two Towers	2002
179		
..
...		
968	Falling Down	1993
113		
979	Lethal Weapon	1987
109		
982	Mad Max 2	1981
96		
983	The Warriors	1979
92		
985	Escape from Alcatraz	1979
112		

	Genre	IMDB_Rating	Director	Star1
No_of_Votes \				
2	Action	9.0	Christopher Nolan	Christian Bale
2303232				
5	Action	8.9	Peter Jackson	Elijah Wood
1642758				
8	Action	8.8	Christopher Nolan	Leonardo DiCaprio
2067042				
10	Action	8.8	Peter Jackson	Elijah Wood
1661481				
13	Action	8.7	Peter Jackson	Elijah Wood
1485555				
..


```

...
968 Action          7.6    Joel Schumacher    Michael Douglas
171640
979 Action          7.6    Richard Donner      Mel Gibson
236894
982 Action          7.6    George Miller       Mel Gibson
166588
983 Action          7.6    Walter Hill         Michael Beck
93878
985 Action          7.6    Don Siegel          Clint Eastwood
121731

```

```

      Gross  Metascore
2    534858444.0      84.0
5    377845905.0      94.0
8    292576195.0      74.0
10   315544750.0      92.0
13   342551365.0      87.0
..
968   40903593.0      56.0
979   65207127.0      68.0
982   12465371.0      77.0
983   22490039.0      65.0
985   43000000.0      76.0

```

```
[172 rows x 10 columns]
```

```

      Series_Title  Released_Year  Runtime    Genre
\
21      Interstellar          2014      169  Adventure
47      Back to the Future      1985      116  Adventure
93      Inglourious Basterds      2009      153  Adventure
110     Das Boot              1981      149  Adventure
114     2001: A Space Odyssey      1968      149  Adventure
..
957  Fear and Loathing in Las Vegas      1998      118  Adventure
964      Dead Man              1995      121  Adventure
966      Apollo 13              PG       140  Adventure
984      The Muppet Movie      1979       95  Adventure
991      Kelly's Heroes        1970      144  Adventure

```

	IMDB_Rating	Director		Star1	No_of_Votes
\					
21	8.6	Christopher Nolan	Matthew McConaughey		1512360
47	8.5	Robert Zemeckis	Michael J. Fox		1058081
93	8.3	Quentin Tarantino	Brad Pitt		1267869
110	8.3	Wolfgang Petersen	Jürgen Prochnow		231855
114	8.3	Stanley Kubrick	Keir Dullea		603517
..
957	7.6	Terry Gilliam	Johnny Depp		259753
964	7.6	Jim Jarmusch	Johnny Depp		90442
966	7.6	Ron Howard	Tom Hanks		269197
984	7.6	James Frawley	Jim Henson		32802
991	7.6	Brian G. Hutton	Clint Eastwood		45338
	Gross	Metascore			
21	188020017.0	74.0			
47	210609762.0	87.0			
93	120540719.0	69.0			
110	11487676.0	86.0			
114	56954992.0	84.0			
..			
957	10680275.0	41.0			
964	1037847.0	62.0			
966	173837933.0	77.0			
984	76657000.0	74.0			
991	1378435.0	50.0			
[72 rows x 10 columns]					
		Series_Title	Released_Year	Runtime	
Genre \					
23	Sen to Chihiro no kamikakushi		2001	125	
Animation					
43	The Lion King		1994	88	
Animation					
46	Hotaru no haka		1988	89	
Animation					
56	Kimi no na wa.		2016	106	
Animation					
58	Spider-Man: Into the Spider-Verse		2018	117	

Animation			
..
..			
956	Mulan	1998	88
Animation			
971	Omohide poro poro	1991	118
Animation			
976	The Little Mermaid	1989	83
Animation			
986	Watership Down	1978	91
Animation			
992	The Jungle Book	1967	78
Animation			

	IMDB_Rating	Director	Star1	
No_of_Votes \				
23	8.6	Hayao Miyazaki	Daveigh Chase	651376
43	8.5	Roger Allers	Rob Minkoff	942045
46	8.5	Isao Takahata	Tsutomu Tatsumi	235231
56	8.4	Makoto Shinkai	Ryûnosuke Kamiki	194838
58	8.4	Bob Persichetti	Peter Ramsey	375110
..
956	7.6	Tony Bancroft	Barry Cook	256906
971	7.6	Isao Takahata	Miki Imai	27071
976	7.6	Ron Clements	John Musker	237696
986	7.6	Martin Rosen	John Hubley	33656
992	7.6	Wolfgang Reitherman	Phil Harris	166409

	Gross	Metascore
23	10055859.0	96.0
43	422783777.0	88.0
46	150734678.0	94.0
56	5017246.0	79.0
58	190241310.0	87.0
..
956	120620254.0	71.0
971	453243.0	90.0
976	111543479.0	88.0
986	232841485.0	64.0

992 141843612.0 65.0

[82 rows x 10 columns]

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating
\					
7	Schindler's List	1993	195	Biography	8.9
15	Goodfellas	1990	146	Biography	8.7
18	Hamilton	2020	160	Biography	8.6
35	The Intouchables	2011	112	Biography	8.5
38	The Pianist	2002	150	Biography	8.5
..
923	La Vie En Rose	2007	140	Biography	7.6
940	Finding Neverland	2004	106	Biography	7.6
949	Blow	2001	124	Biography	7.6
952	The Hurricane	1999	146	Biography	7.6
987	Midnight Express	1978	121	Biography	7.6

	Director	Star1	No_of_Votes	Gross
Metascore				
7	Steven Spielberg	Liam Neeson	1213505	96898818.0
94.0				
15	Martin Scorsese	Robert De Niro	1020727	46836394.0
90.0				
18	Thomas Kail	Lin-Manuel Miranda	55291	440984783.0
90.0				
35	Olivier Nakache	Éric Toledano	760360	13182281.0
57.0				
38	Roman Polanski	Adrien Brody	729603	32572577.0
85.0				
..
...				
923	Olivier Dahan	Marion Cotillard	82781	10301706.0
66.0				
940	Marc Forster	Johnny Depp	198677	51680613.0
67.0				
949	Ted Demme	Johnny Depp	240714	52990775.0
52.0				
952	Norman Jewison	Denzel Washington	91557	50668906.0
74.0				

987	Alan Parker	Brad Davis	73662	35000000.0
59.0				

[88 rows x 10 columns]

Runtime \	Series_Title	Released_Year
19	Gisaengchung	2019
132		
26	La vita è bella	1997
116		
51	Modern Times	1936
87		
52	City Lights	1931
87		
64	3 Idiots	2009
170		
..
...		
977	The Naked Gun: From the Files of Police Squad!	1988
85		
978	Planes, Trains & Automobiles	1987
93		
989	The Long Goodbye	1973
112		
994	A Hard Day's Night	1964
87		
995	Breakfast at Tiffany's	1961
115		

No_of_Votes	Genre \	IMDB_Rating	Director	Star1
19	Comedy	8.6	Bong Joon Ho	Kang-ho Song
552778				
26	Comedy	8.6	Roberto Benigni	Roberto Benigni
623629				
51	Comedy	8.5	Charles Chaplin	Charles Chaplin
217881				
52	Comedy	8.5	Charles Chaplin	Charles Chaplin
167839				
64	Comedy	8.4	Rajkumar Hirani	Aamir Khan
344445				
..
..				..
977	Comedy	7.6	David Zucker	Leslie Nielsen
152871				
978	Comedy	7.6	John Hughes	Steve Martin
124773				
989	Comedy	7.6	Robert Altman	Elliott Gould
26337				

994	Comedy	7.6	Richard Lester	John Lennon
40351				
995	Comedy	7.6	Blake Edwards	Audrey Hepburn
166544				

	Gross	Metascore
19	53367844.0	96.0
26	57598247.0	59.0
51	163245.0	96.0
52	19181.0	99.0
64	6532908.0	67.0
..
977	78756177.0	76.0
978	49530280.0	72.0
989	959000.0	87.0
994	13780024.0	96.0
995	679874270.0	76.0

[155 rows x 10 columns]

IMDB_Rating \	Series_Title	Released_Year	Runtime	Genre
1	The Godfather	1972	175	Crime
9.2				
3	The Godfather: Part II	1974	202	Crime
9.0				
4	12 Angry Men	1957	96	Crime
9.0				
6	Pulp Fiction	1994	154	Crime
8.9				
22	Cidade de Deus	2002	130	Crime
8.6				
..
.				
958	Funny Games	1997	108	Crime
7.6				
960	Sleepers	1996	147	Crime
7.6				
974	The Godfather: Part III	1990	162	Crime
7.6				
980	Blood Simple	1984	99	Crime
7.6				
999	The 39 Steps	1935	86	Crime
7.6				

	Director	Star1	No_of_Votes	Gross
Metascore				
1	Francis Ford Coppola	Marlon Brando	1620367	134966411.0
100.0				
3	Francis Ford Coppola	Al Pacino	1129952	57300000.0
90.0				

4	Sidney Lumet	Henry Fonda	689845	4360000.0
96.0				
6	Quentin Tarantino	John Travolta	1826188	107928762.0
94.0				
22	Fernando Meirelles	Kátia Lund	699256	7563397.0
79.0				
..
...				
958	Michael Haneke	Susanne Lothar	65058	217345863.0
69.0				
960	Barry Levinson	Robert De Niro	186734	49100000.0
49.0				
974	Francis Ford Coppola	Al Pacino	359809	66666062.0
60.0				
980	Joel Coen	Ethan Coen	87745	2150000.0
82.0				
999	Alfred Hitchcock	Robert Donat	51853	302787539.0
93.0				

[107 rows x 10 columns]

	Series_Title	Released_Year	Runtime	Genre	\
0	The Shawshank Redemption	1994	142	Drama	
9	Fight Club	1999	139	Drama	
11	Forrest Gump	1994	142	Drama	
17	One Flew Over the Cuckoo's Nest	1975	133	Drama	
20	Soorarai Pottru	2020	153	Drama	
..
990	Giù la testa	1971	157	Drama	
993	Blowup	1966	111	Drama	
996	Giant	1956	201	Drama	
997	From Here to Eternity	1953	118	Drama	
998	Lifeboat	1944	97	Drama	

	IMDB_Rating	Director	Star1
No_of_Votes \			
0	9.3	Frank Darabont	Tim Robbins
2343110			
9	8.8	David Fincher	Brad Pitt
1854740			
11	8.8	Robert Zemeckis	Tom Hanks
1809221			
17	8.7	Milos Forman	Jack Nicholson
918088			
20	8.6	Sudha Kongara	Suriya
54995			
..
..			
990	7.6	Sergio Leone	Rod Steiger
30144			

993	7.6	Michelangelo Antonioni	David Hemmings
56513			
996	7.6	George Stevens	Elizabeth Taylor
34075			
997	7.6	Fred Zinnemann	Burt Lancaster
43374			
998	7.6	Alfred Hitchcock	Tallulah Bankhead
26471			

	Gross	Metascore
0	28341469.0	80.0
9	37030102.0	66.0
11	330252182.0	82.0
17	112000000.0	83.0
20	556832648.0	NaN
..
990	696690.0	77.0
993	632532802.0	82.0
996	195217415.0	84.0
997	30500000.0	85.0
998	852142728.0	78.0

[289 rows x 10 columns]

	Series_Title	Released_Year	Runtime
Genre \			
688	E.T. the Extra-Terrestrial	1982	115
Family			
698	Willy Wonka & the Chocolate Factory	1971	100
Family			

	IMDB_Rating	Director	Star1	No_of_Votes
Gross \				
688	7.8	Steven Spielberg	Henry Thomas	372490
435110554.0				
698	7.8	Mel Stuart	Gene Wilder	178731
4000000.0				

	Metascore				
688	91.0				
698	67.0				
	Series_Title	Released_Year	Runtime	Genre \	
321	Das Cabinet des Dr. Caligari	1920	76	Fantasy	
568	Nosferatu	1922	94	Fantasy	

	IMDB_Rating	Director	Star1	No_of_Votes
Gross \				
321	8.1	Robert Wiene	Werner Krauss	57428
337574718.0				
568	7.9	F.W. Murnau	Max Schreck	88794
445151978.0				

321	Metascore					
568	NaN					
	NaN					
	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	
\						
309	The Third Man	1949	104	Film-Noir	8.1	
456	The Maltese Falcon	1941	100	Film-Noir	8.0	
712	Shadow of a Doubt	1943	108	Film-Noir	7.8	
	Director	Star1	No_of_Votes	Gross		
Metascore						
309	Carol Reed	Orson Welles	158731	449191.0		
97.0						
456	John Huston	Humphrey Bogart	148928	2108060.0		
96.0						
712	Alfred Hitchcock	Teresa Wright	59556	123353292.0		
94.0						
	Series_Title	Released_Year	Runtime	Genre		
IMDB_Rating \						
49	Psycho	1960	109	Horror		
8.5						
75	Alien	1979	117	Horror		
8.4						
271	The Thing	1982	109	Horror		
8.1						
419	The Exorcist	1973	122	Horror		
8.0						
544	Night of the Living Dead	1968	96	Horror		
7.9						
707	The Innocents	1961	100	Horror		
7.8						
724	Get Out	2017	104	Horror		
7.7						
844	Halloween	1978	91	Horror		
7.7						
876	The Invisible Man	1933	71	Horror		
7.7						
932	Saw	2004	103	Horror		
7.6						
948	The Others	2001	101	Horror		
7.6						
	Director	Star1	No_of_Votes	Gross		
Metascore						
49	Alfred Hitchcock	Anthony Perkins	604211	32000000.0		
97.0						

75	Ridley Scott	Sigourney Weaver	787806	78900000.0	
89.0					
271	John Carpenter	Kurt Russell	371271	13782838.0	
57.0					
419	William Friedkin	Ellen Burstyn	362393	232906145.0	
81.0					
544	George A. Romero	Duane Jones	116557	89029.0	
89.0					
707	Jack Clayton	Deborah Kerr	27007	2616000.0	
88.0					
724	Jordan Peele	Daniel Kaluuya	492851	176040665.0	
85.0					
844	John Carpenter	Donald Pleasence	233106	47000000.0	
87.0					
876	James Whale	Claude Rains	30683	298791505.0	
87.0					
932	James Wan	Cary Elwes	379020	56000369.0	
46.0					
948	Alejandro Amenábar	Nicole Kidman	337651	96522687.0	
74.0					
	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating \
69	Memento	2000	113	Mystery	8.4
81	Rear Window	1954	112	Mystery	8.4
119	Vertigo	1958	128	Mystery	8.3
145	Shutter Island	2010	138	Mystery	8.2
220	Kahaani	2012	122	Mystery	8.1
393	Twelve Monkeys	1995	129	Mystery	8.0
420	Sleuth	1972	138	Mystery	8.0
714	The Lady Vanishes	1938	96	Mystery	7.8
829	Spoorloos	1988	107	Mystery	7.7
899	El cuerpo	2012	112	Mystery	7.6
959	Dark City	1998	100	Mystery	7.6
961	Lost Highway	1997	134	Mystery	7.6
	Director		Star1	No_of_Votes	
Gross \					
69	Christopher Nolan		Guy Pearce	1125712	
25544867.0					
81	Alfred Hitchcock		James Stewart	444074	
36764313.0					
119	Alfred Hitchcock		James Stewart	364368	
3200000.0					
145	Martin Scorsese		Leonardo DiCaprio	1129894	
128012934.0					
220	Sujoy Ghosh		Vidya Balan	57806	
1035953.0					
393	Terry Gilliam		Bruce Willis	578443	
57141459.0					
420	Joseph L. Mankiewicz		Laurence Olivier	44748	
4081254.0					

714	Alfred Hitchcock	Margaret Lockwood	47400
474203697.0			
829	George Sluizer	Bernard-Pierre Donnadieu	33982
367916835.0			
899	Oriol Paulo	Jose Coronado	57549
140340673.0			
959	Alex Proyas	Rufus Sewell	187927
14378331.0			
961	David Lynch	Bill Pullman	131101
3796699.0			

	Metascore
--	-----------

69	80.0
81	100.0
119	100.0
145	63.0
220	NaN
393	74.0
420	NaN
714	98.0
829	NaN
899	NaN
959	66.0
961	52.0

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating \
700	Wait Until Dark	1967	108	Thriller	7.8

	Director	Star1	No_of_Votes	Gross	Metascore
700	Terence Young	Audrey Hepburn	27733	17550741.0	81.0

	Series_Title	Released_Year	Runtime	Genre \
12	Il buono, il brutto, il cattivo	1966	161	Western
48	Once Upon a Time in the West	1968	165	Western
115	Per qualche dollaro in più	1965	132	Western
691	The Outlaw Josey Wales	1976	135	Western

	IMDB_Rating	Director	Star1	No_of_Votes	Gross \
12	8.8	Sergio Leone	Clint Eastwood	688390	6100000.0
48	8.5	Sergio Leone	Henry Fonda	302844	5321508.0
115	8.3	Sergio Leone	Clint Eastwood	232772	15000000.0

```
691          7.8  Clint Eastwood  Clint Eastwood          65659
31800000.0
```

```
Metascore
12      90.0
48      80.0
115     74.0
691     69.0
```

```
for group,data in genres:
    print(group)                #har group ko print kar dega
```

```
Action
Adventure
Animation
Biography
Comedy
Crime
Drama
Family
Fantasy
Film-Noir
Horror
Mystery
Thriller
Western
```

```
# find the heighest rated movie of each genre
for group,data in genres:
    print(data['IMDB_Rating'].max())    # ye har genre ke maxm
imdb_rating ko nikal ke dega
```

```
9.0
8.6
8.6
8.9
8.6
9.2
9.3
7.8
8.1
8.1
8.5
8.4
7.8
8.8
```

```
for group,data in genres:
    print(data[data['IMDB_Rating'] == data['IMDB_Rating'].max()] )
```

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	\
2	The Dark Knight	2008	152	Action	9.0	
	Director	Star1	No_of_Votes	Gross	Metascore	
2	Christopher Nolan	Christian Bale	2303232	534858444.0	84.0	
	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	\
21	Interstellar	2014	169	Adventure	8.6	
	Director	Star1	No_of_Votes	Gross	\	
21	Christopher Nolan	Matthew McConaughey	1512360	188020017.0		
21	Metascore	74.0				
	Series_Title	Released_Year	Runtime	Genre	\	
23	Sen to Chihiro no kamikakushi	2001	125	Animation		
	IMDB_Rating	Director	Star1	No_of_Votes	Gross	\
23	8.6	Hayao Miyazaki	Daveigh Chase	651376	10055859.0	
23	Metascore	96.0				
	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	\
7	Schindler's List	1993	195	Biography	8.9	
	Director	Star1	No_of_Votes	Gross	Metascore	
7	Steven Spielberg	Liam Neeson	1213505	96898818.0	94.0	
	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	\
19	Gisaengchung	2019	132	Comedy	8.6	
26	La vita è bella	1997	116	Comedy	8.6	
	Director	Star1	No_of_Votes	Gross	Metascore	
19	Bong Joon Ho	Kang-ho Song	552778	53367844.0	96.0	
26	Roberto Benigni	Roberto Benigni	623629	57598247.0	59.0	
	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	\
1	The Godfather	1972	175	Crime	9.2	
	Director	Star1	No_of_Votes	Gross	Metascore	
1	Francis Ford Coppola	Marlon Brando	1620367	134966411.0	100.0	
	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	\

0	The Shawshank Redemption	1994	142	Drama	9.3
	Director	Star1	No_of_Votes	Gross	Metascore
0	Frank Darabont	Tim Robbins	2343110	28341469.0	80.0
	Series_Title	Released_Year	Runtime		
Genre \					
688	E.T. the Extra-Terrestrial	1982	115		
Family					
698	Willy Wonka & the Chocolate Factory	1971	100		
Family					
	IMDB_Rating	Director	Star1	No_of_Votes	
Gross \					
688	7.8	Steven Spielberg	Henry Thomas	372490	
435110554.0					
698	7.8	Mel Stuart	Gene Wilder	178731	
4000000.0					
	Metascore				
688	91.0				
698	67.0				
	Series_Title	Released_Year	Runtime	Genre \	
321	Das Cabinet des Dr. Caligari	1920	76	Fantasy	
	IMDB_Rating	Director	Star1	No_of_Votes	
Gross \					
321	8.1	Robert Wiene	Werner Krauss	57428	
337574718.0					
	Metascore				
321	NaN				
	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating
Director \					
309	The Third Man	1949	104	Film-Noir	8.1
Carol Reed					
	Star1	No_of_Votes	Gross	Metascore	
309	Orson Welles	158731	449191.0	97.0	
	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating
Director \					
49	Psycho	1960	109	Horror	8.5
Alfred Hitchcock					
	Star1	No_of_Votes	Gross	Metascore	
49	Anthony Perkins	604211	32000000.0	97.0	
	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating \
69	Memento	2000	113	Mystery	8.4
81	Rear Window	1954	112	Mystery	8.4
	Director	Star1	No_of_Votes	Gross	

```

Metascore
69 Christopher Nolan      Guy Pearce      1125712  25544867.0
80.0
81 Alfred Hitchcock James Stewart      444074  36764313.0
100.0
      Series_Title Released_Year Runtime      Genre IMDB_Rating \
700 Wait Until Dark      1967      108 Thriller      7.8

      Director      Star1 No_of_Votes      Gross Metascore
700 Terence Young Audrey Hepburn      27733  17550741.0      81.0

      Series_Title Released_Year Runtime      Genre \
12 Il buono, il brutto, il cattivo      1966      161 Western

      IMDB_Rating      Director      Star1 No_of_Votes      Gross
\
12      8.8 Sergio Leone Clint Eastwood      688390  61000000.0

      Metascore
12      90.0

# aache se dekhne ke lye
df = pd.DataFrame(columns=movies.columns)
df # ek empty dataframe bana lye like movies
dataframe

Empty DataFrame
Columns: [Series_Title, Released_Year, Runtime, Genre, IMDB_Rating,
Director, Star1, No_of_Votes, Gross, Metascore]
Index: []

df = pd.DataFrame(columns=movies.columns)

for group,data in genres:

    df= pd.concat([df,data[data['IMDB_Rating'] ==
data['IMDB_Rating'].max()]])

df

C:\Users\jayra\AppData\Local\Temp\ipykernel_23732\1685166070.py:5:
FutureWarning: The behavior of DataFrame concatenation with empty or
all-NA entries is deprecated. In a future version, this will no longer
exclude empty or all-NA columns when determining the result dtypes. To
retain the old behavior, exclude the relevant entries before the
concat operation.
    df= pd.concat([df,data[data['IMDB_Rating'] ==
data['IMDB_Rating'].max()]])

```

		Series_Title	Released_Year	Runtime	
Genre \					
2		The Dark Knight	2008	152	
Action					
21		Interstellar	2014	169	
Adventure					
23	Sen to Chihiro no kamikakushi		2001	125	
Animation					
7		Schindler's List	1993	195	
Biography					
19		Gisaengchung	2019	132	
Comedy					
26		La vita è bella	1997	116	
Comedy					
1		The Godfather	1972	175	
Crime					
0	The Shawshank Redemption		1994	142	
Drama					
688	E.T. the Extra-Terrestrial		1982	115	
Family					
698	Willy Wonka & the Chocolate Factory		1971	100	
Family					
321	Das Cabinet des Dr. Caligari		1920	76	
Fantasy					
309		The Third Man	1949	104	Film-
Noir					
49		Psycho	1960	109	
Horror					
69		Memento	2000	113	
Mystery					
81		Rear Window	1954	112	
Mystery					
700		Wait Until Dark	1967	108	
Thriller					
12	Il buono, il brutto, il cattivo		1966	161	
Western					
IMDB_Rating		Director		Star1	
No_of_Votes \					
2	9.0	Christopher Nolan	Christian Bale		
2303232					
21	8.6	Christopher Nolan	Matthew McConaughey		
1512360					
23	8.6	Hayao Miyazaki	Daveigh Chase		
651376					
7	8.9	Steven Spielberg	Liam Neeson		
1213505					
19	8.6	Bong Joon Ho	Kang-ho Song		
552778					
26	8.6	Roberto Benigni	Roberto Benigni		

623629			
1	9.2	Francis Ford Coppola	Marlon Brando
1620367			
0	9.3	Frank Darabont	Tim Robbins
2343110			
688	7.8	Steven Spielberg	Henry Thomas
372490			
698	7.8	Mel Stuart	Gene Wilder
178731			
321	8.1	Robert Wiene	Werner Krauss
57428			
309	8.1	Carol Reed	Orson Welles
158731			
49	8.5	Alfred Hitchcock	Anthony Perkins
604211			
69	8.4	Christopher Nolan	Guy Pearce
1125712			
81	8.4	Alfred Hitchcock	James Stewart
444074			
700	7.8	Terence Young	Audrey Hepburn
27733			
12	8.8	Sergio Leone	Clint Eastwood
688390			

	Gross	Metascore
2	534858444.0	84.0
21	188020017.0	74.0
23	10055859.0	96.0
7	96898818.0	94.0
19	53367844.0	96.0
26	57598247.0	59.0
1	134966411.0	100.0
0	28341469.0	80.0
688	435110554.0	91.0
698	4000000.0	67.0
321	337574718.0	NaN
309	449191.0	97.0
49	32000000.0	97.0
69	25544867.0	80.0
81	36764313.0	100.0
700	17550741.0	81.0
12	6100000.0	90.0

```
df = pd.DataFrame(columns=movies.columns)
for group, data in genres:
    # Filter for the highest IMDB rating in the current group
    highest_rating_row = data[data['IMDB_Rating'] ==
data['IMDB_Rating'].max()]
    # Use pd.concat to append the highest_rating_row to df
    df = pd.concat([df, highest_rating_row])
```

df

C:\Users\jayra\AppData\Local\Temp\ipykernel_23732\492345485.py:6:
FutureWarning: The behavior of DataFrame concatenation with empty or
all-NA entries is deprecated. In a future version, this will no longer
exclude empty or all-NA columns when determining the result dtypes. To
retain the old behavior, exclude the relevant entries before the
concat operation.

```
df = pd.concat([df, highest_rating_row])
```

		Series_Title	Released_Year	Runtime	
Genre \					
2		The Dark Knight	2008	152	
Action					
21		Interstellar	2014	169	
Adventure					
23	Sen to Chihiro no kamikakushi		2001	125	
Animation					
7		Schindler's List	1993	195	
Biography					
19		Gisaengchung	2019	132	
Comedy					
26		La vita è bella	1997	116	
Comedy					
1		The Godfather	1972	175	
Crime					
0		The Shawshank Redemption	1994	142	
Drama					
688	E.T. the Extra-Terrestrial		1982	115	
Family					
698	Willy Wonka & the Chocolate Factory		1971	100	
Family					
321	Das Cabinet des Dr. Caligari		1920	76	
Fantasy					
309		The Third Man	1949	104	Film-
Noir					
49		Psycho	1960	109	
Horror					
69		Memento	2000	113	
Mystery					
81		Rear Window	1954	112	
Mystery					
700		Wait Until Dark	1967	108	
Thriller					
12	Il buono, il brutto, il cattivo		1966	161	
Western					
IMDB_Rating		Director		Star1	
No_of_Votes \					

2	9.0	Christopher Nolan	Christian Bale
2303232			
21	8.6	Christopher Nolan	Matthew McConaughey
1512360			
23	8.6	Hayao Miyazaki	Daveigh Chase
651376			
7	8.9	Steven Spielberg	Liam Neeson
1213505			
19	8.6	Bong Joon Ho	Kang-ho Song
552778			
26	8.6	Roberto Benigni	Roberto Benigni
623629			
1	9.2	Francis Ford Coppola	Marlon Brando
1620367			
0	9.3	Frank Darabont	Tim Robbins
2343110			
688	7.8	Steven Spielberg	Henry Thomas
372490			
698	7.8	Mel Stuart	Gene Wilder
178731			
321	8.1	Robert Wiene	Werner Krauss
57428			
309	8.1	Carol Reed	Orson Welles
158731			
49	8.5	Alfred Hitchcock	Anthony Perkins
604211			
69	8.4	Christopher Nolan	Guy Pearce
1125712			
81	8.4	Alfred Hitchcock	James Stewart
444074			
700	7.8	Terence Young	Audrey Hepburn
27733			
12	8.8	Sergio Leone	Clint Eastwood
688390			

	Gross	Metascore
2	534858444.0	84.0
21	188020017.0	74.0
23	10055859.0	96.0
7	96898818.0	94.0
19	53367844.0	96.0
26	57598247.0	59.0
1	134966411.0	100.0
0	28341469.0	80.0
688	435110554.0	91.0
698	4000000.0	67.0
321	337574718.0	NaN
309	449191.0	97.0
49	32000000.0	97.0

69	25544867.0	80.0
81	36764313.0	100.0
700	17550741.0	81.0
12	6100000.0	90.0

```
# split (apply) combine    Strategy
# apply -> builtin function
```

```
genres.apply(min)
```

```
C:\Users\jayra\AppData\Local\Temp\ipykernel_23732\1443917609.py:1:
FutureWarning: The provided callable <built-in function min> is
currently using np.minimum.reduce. In a future version of pandas, the
provided callable will be used directly. To keep current behavior pass
the string np.minimum.reduce instead.
```

```
    genres.apply(min)
```

```
C:\Users\jayra\AppData\Local\Temp\ipykernel_23732\1443917609.py:1:
DeprecationWarning: DataFrameGroupBy.apply operated on the grouping
columns. This behavior is deprecated, and in a future version of
pandas the grouping columns will be excluded from the operation.
Either pass `include_groups=False` to exclude the groupings or
explicitly select the grouping columns after groupby to silence this
warning.
```

```
    genres.apply(min)
```

	Series_Title	Released_Year	Runtime
Genre \			
Genre			
Action	300	1924	45
Action			
Adventure	2001: A Space Odyssey	1925	88
Adventure			
Animation	Akira	1940	71
Animation			
Biography	12 Years a Slave	1928	93
Biography			
Comedy	(500) Days of Summer	1921	68
Comedy			
Crime	12 Angry Men	1931	80
Crime			
Drama	1917	1925	64
Drama			
Family	E.T. the Extra-Terrestrial	1971	100
Family			
Fantasy	Das Cabinet des Dr. Caligari	1920	76
Fantasy			

Film-Noir	Shadow of a Doubt	1941	100
Film-Noir			
Horror	Alien	1933	71
Horror			
Mystery	Dark City	1938	96
Mystery			
Thriller	Wait Until Dark	1967	108
Thriller			
Western	Il buono, il brutto, il cattivo	1965	132
Western			
	IMDB_Rating	Director	
Star1 \			
Genre			
Action	7.6	Abhishek Chaubey	Aamir
Khan			
Adventure	7.6	Akira Kurosawa	Aamir
Khan			
Animation	7.6	Adam Elliot	Adrian
Molina			
Biography	7.6	Adam McKay	Adrien
Brody			
Comedy	7.6	Alejandro G. Iñárritu	Aamir
Khan			
Crime	7.6	Akira Kurosawa	Ajay
Devgn			
Drama	7.6	Aamir Khan	Abhay
Deol			
Family	7.8	Mel Stuart	Gene
Wilder			
Fantasy	7.9	F.W. Murnau	Max
Schreck			
Film-Noir	7.8	Alfred Hitchcock	Humphrey
Bogart			
Horror	7.6	Alejandro Amenábar	Anthony
Perkins			
Mystery	7.6	Alex Proyas	Bernard-Pierre
Donnadiou			
Thriller	7.8	Terence Young	Audrey
Hepburn			
Western	7.8	Clint Eastwood	Clint
Eastwood			
	No_of_Votes	Gross	Metascore
Genre			
Action	25312	3296.0	NaN
Adventure	29999	61001.0	NaN
Animation	25229	128985.0	NaN
Biography	27254	21877.0	NaN

Comedy	26337	1305.0	NaN
Crime	27712	6013.0	NaN
Drama	25088	3600.0	NaN
Family	178731	4000000.0	67.0
Fantasy	57428	337574718.0	NaN
Film-Noir	59556	449191.0	94.0
Horror	27007	89029.0	46.0
Mystery	33982	1035953.0	NaN
Thriller	27733	17550741.0	81.0
Western	65659	5321508.0	69.0

find number of movies starting with A for each group

```
def foo(group):
    print(group)
    return group
```

```
genres.apply(foo)
```

	Series_Title	Released_Year
Runtime \		
2	The Dark Knight	2008
152		
5	The Lord of the Rings: The Return of the King	2003
201		
8	Inception	2010
148		
10	The Lord of the Rings: The Fellowship of the Ring	2001
178		
13	The Lord of the Rings: The Two Towers	2002
179		
..
...		
968	Falling Down	1993
113		
979	Lethal Weapon	1987
109		
982	Mad Max 2	1981
96		
983	The Warriors	1979
92		
985	Escape from Alcatraz	1979
112		

	Genre	IMDB_Rating	Director	Star1
No_of_Votes \				
2	Action	9.0	Christopher Nolan	Christian Bale
2303232				
5	Action	8.9	Peter Jackson	Elijah Wood
1642758				

8	Action	8.8	Christopher Nolan	Leonardo DiCaprio
2067042				
10	Action	8.8	Peter Jackson	Elijah Wood
1661481				
13	Action	8.7	Peter Jackson	Elijah Wood
1485555				
..
...				
968	Action	7.6	Joel Schumacher	Michael Douglas
171640				
979	Action	7.6	Richard Donner	Mel Gibson
236894				
982	Action	7.6	George Miller	Mel Gibson
166588				
983	Action	7.6	Walter Hill	Michael Beck
93878				
985	Action	7.6	Don Siegel	Clint Eastwood
121731				

	Gross	Metascore
2	534858444.0	84.0
5	377845905.0	94.0
8	292576195.0	74.0
10	315544750.0	92.0
13	342551365.0	87.0
..
968	40903593.0	56.0
979	65207127.0	68.0
982	12465371.0	77.0
983	22490039.0	65.0
985	43000000.0	76.0

[172 rows x 10 columns]

	Series_Title	Released_Year	Runtime	Genre
\				
21	Interstellar	2014	169	Adventure
47	Back to the Future	1985	116	Adventure
93	Inglourious Basterds	2009	153	Adventure
110	Das Boot	1981	149	Adventure
114	2001: A Space Odyssey	1968	149	Adventure
..
957	Fear and Loathing in Las Vegas	1998	118	Adventure
964	Dead Man	1995	121	Adventure

966	Apollo 13	PG	140	Adventure
984	The Muppet Movie	1979	95	Adventure
991	Kelly's Heroes	1970	144	Adventure

	IMDB_Rating	Director	Star1	No_of_Votes
\				
21	8.6	Christopher Nolan	Matthew McConaughey	1512360
47	8.5	Robert Zemeckis	Michael J. Fox	1058081
93	8.3	Quentin Tarantino	Brad Pitt	1267869
110	8.3	Wolfgang Petersen	Jürgen Prochnow	231855
114	8.3	Stanley Kubrick	Keir Dullea	603517
..
957	7.6	Terry Gilliam	Johnny Depp	259753
964	7.6	Jim Jarmusch	Johnny Depp	90442
966	7.6	Ron Howard	Tom Hanks	269197
984	7.6	James Frawley	Jim Henson	32802
991	7.6	Brian G. Hutton	Clint Eastwood	45338

	Gross	Metascore
21	188020017.0	74.0
47	210609762.0	87.0
93	120540719.0	69.0
110	11487676.0	86.0
114	56954992.0	84.0
..
957	10680275.0	41.0
964	1037847.0	62.0
966	173837933.0	77.0
984	76657000.0	74.0
991	1378435.0	50.0

[72 rows x 10 columns]

	Series_Title	Released_Year	Runtime
Genre \			
23	Sen to Chihiro no kamikakushi	2001	125
Animation			

43		The Lion King	1994	88
Animation				
46		Hotaru no haka	1988	89
Animation				
56		Kimi no na wa.	2016	106
Animation				
58		Spider-Man: Into the Spider-Verse	2018	117
Animation				
..	
..				
956		Mulan	1998	88
Animation				
971		Omohide poro poro	1991	118
Animation				
976		The Little Mermaid	1989	83
Animation				
986		Watership Down	1978	91
Animation				
992		The Jungle Book	1967	78
Animation				
	IMDB_Rating	Director	Star1	
No_of_Votes \				
23	8.6	Hayao Miyazaki	Daveigh Chase	651376
43	8.5	Roger Allers	Rob Minkoff	942045
46	8.5	Isao Takahata	Tsutomu Tatsumi	235231
56	8.4	Makoto Shinkai	Ryûnosuke Kamiki	194838
58	8.4	Bob Persichetti	Peter Ramsey	375110
..
956	7.6	Tony Bancroft	Barry Cook	256906
971	7.6	Isao Takahata	Miki Imai	27071
976	7.6	Ron Clements	John Musker	237696
986	7.6	Martin Rosen	John Hubley	33656
992	7.6	Wolfgang Reitherman	Phil Harris	166409
	Gross	Metascore		
23	10055859.0	96.0		
43	422783777.0	88.0		
46	150734678.0	94.0		

56	5017246.0	79.0
58	190241310.0	87.0
..
956	120620254.0	71.0
971	453243.0	90.0
976	111543479.0	88.0
986	232841485.0	64.0
992	141843612.0	65.0

[82 rows x 10 columns]

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating
\					
7	Schindler's List	1993	195	Biography	8.9
15	Goodfellas	1990	146	Biography	8.7
18	Hamilton	2020	160	Biography	8.6
35	The Intouchables	2011	112	Biography	8.5
38	The Pianist	2002	150	Biography	8.5
..
923	La Vie En Rose	2007	140	Biography	7.6
940	Finding Neverland	2004	106	Biography	7.6
949	Blow	2001	124	Biography	7.6
952	The Hurricane	1999	146	Biography	7.6
987	Midnight Express	1978	121	Biography	7.6

	Director	Star1	No_of_Votes	Gross
Metascore				
7	Steven Spielberg	Liam Neeson	1213505	96898818.0
94.0				
15	Martin Scorsese	Robert De Niro	1020727	46836394.0
90.0				
18	Thomas Kail	Lin-Manuel Miranda	55291	440984783.0
90.0				
35	Olivier Nakache	Éric Toledano	760360	13182281.0
57.0				
38	Roman Polanski	Adrien Brody	729603	32572577.0
85.0				
..
...				
923	Olivier Dahan	Marion Cotillard	82781	10301706.0

66.0					
940	Marc Forster	Johnny Depp	198677	51680613.0	
67.0					
949	Ted Demme	Johnny Depp	240714	52990775.0	
52.0					
952	Norman Jewison	Denzel Washington	91557	50668906.0	
74.0					
987	Alan Parker	Brad Davis	73662	35000000.0	
59.0					

[88 rows x 10 columns]

	Series_Title	Released_Year
Runtime \		
19	Gisaengchung	2019
132		
26	La vita è bella	1997
116		
51	Modern Times	1936
87		
52	City Lights	1931
87		
64	3 Idiots	2009
170		
..
...		
977	The Naked Gun: From the Files of Police Squad!	1988
85		
978	Planes, Trains & Automobiles	1987
93		
989	The Long Goodbye	1973
112		
994	A Hard Day's Night	1964
87		
995	Breakfast at Tiffany's	1961
115		

	Genre	IMDB_Rating	Director	Star1
No_of_Votes \				
19	Comedy	8.6	Bong Joon Ho	Kang-ho Song
552778				
26	Comedy	8.6	Roberto Benigni	Roberto Benigni
623629				
51	Comedy	8.5	Charles Chaplin	Charles Chaplin
217881				
52	Comedy	8.5	Charles Chaplin	Charles Chaplin
167839				
64	Comedy	8.4	Rajkumar Hirani	Aamir Khan
344445				
..
				..

977	Comedy	7.6	David Zucker	Leslie Nielsen
152871				
978	Comedy	7.6	John Hughes	Steve Martin
124773				
989	Comedy	7.6	Robert Altman	Elliott Gould
26337				
994	Comedy	7.6	Richard Lester	John Lennon
40351				
995	Comedy	7.6	Blake Edwards	Audrey Hepburn
166544				

	Gross	Metascore
19	53367844.0	96.0
26	57598247.0	59.0
51	163245.0	96.0
52	19181.0	99.0
64	6532908.0	67.0
...
977	78756177.0	76.0
978	49530280.0	72.0
989	959000.0	87.0
994	13780024.0	96.0
995	679874270.0	76.0

[155 rows x 10 columns]

	Series_Title	Released_Year	Runtime	Genre
IMDB_Rating \				
1	The Godfather	1972	175	Crime
9.2				
3	The Godfather: Part II	1974	202	Crime
9.0				
4	12 Angry Men	1957	96	Crime
9.0				
6	Pulp Fiction	1994	154	Crime
8.9				
22	Cidade de Deus	2002	130	Crime
8.6				
...
.				
958	Funny Games	1997	108	Crime
7.6				
960	Sleepers	1996	147	Crime
7.6				
974	The Godfather: Part III	1990	162	Crime
7.6				
980	Blood Simple	1984	99	Crime
7.6				
999	The 39 Steps	1935	86	Crime

7.6

	Director	Star1	No_of_Votes	Gross
Metascore				
1	Francis Ford Coppola	Marlon Brando	1620367	134966411.0
100.0				
3	Francis Ford Coppola	Al Pacino	1129952	57300000.0
90.0				
4	Sidney Lumet	Henry Fonda	689845	4360000.0
96.0				
6	Quentin Tarantino	John Travolta	1826188	107928762.0
94.0				
22	Fernando Meirelles	Kátia Lund	699256	7563397.0
79.0				
..
...				
958	Michael Haneke	Susanne Lothar	65058	217345863.0
69.0				
960	Barry Levinson	Robert De Niro	186734	49100000.0
49.0				
974	Francis Ford Coppola	Al Pacino	359809	66666062.0
60.0				
980	Joel Coen	Ethan Coen	87745	2150000.0
82.0				
999	Alfred Hitchcock	Robert Donat	51853	302787539.0
93.0				

[107 rows x 10 columns]

	Series_Title	Released_Year	Runtime	Genre	\
0	The Shawshank Redemption	1994	142	Drama	
9	Fight Club	1999	139	Drama	
11	Forrest Gump	1994	142	Drama	
17	One Flew Over the Cuckoo's Nest	1975	133	Drama	
20	Soorarai Pottru	2020	153	Drama	
..
990	Giù la testa	1971	157	Drama	
993	Blowup	1966	111	Drama	
996	Giant	1956	201	Drama	
997	From Here to Eternity	1953	118	Drama	
998	Lifeboat	1944	97	Drama	

IMDB_Rating	Director	Star1
No_of_Votes \		
0	9.3	Frank Darabont
2343110		Tim Robbins
9	8.8	David Fincher
1854740		Brad Pitt
11	8.8	Robert Zemeckis
1809221		Tom Hanks
17	8.7	Milos Forman
		Jack Nicholson

```

918088
20          8.6          Sudha Kongara          Suriya
54995
..          ...          ...          ...
..
990          7.6          Sergio Leone          Rod Steiger
30144
993          7.6  Michelangelo Antonioni  David Hemmings
56513
996          7.6          George Stevens  Elizabeth Taylor
34075
997          7.6          Fred Zinnemann  Burt Lancaster
43374
998          7.6          Alfred Hitchcock  Tallulah Bankhead
26471

```

```

          Gross  Metascore
0      28341469.0      80.0
9      37030102.0      66.0
11     330252182.0      82.0
17     112000000.0      83.0
20     556832648.0      NaN
..          ...          ...
990     696690.0      77.0
993    632532802.0      82.0
996    195217415.0      84.0
997     30500000.0      85.0
998    852142728.0      78.0

```

[289 rows x 10 columns]

```

          Series_Title  Released_Year  Runtime
Genre \
688          E.T. the Extra-Terrestrial      1982      115
Family
698  Willy Wonka & the Chocolate Factory      1971      100
Family

```

```

          IMDB_Rating          Director          Star1  No_of_Votes
Gross \
688          7.8  Steven Spielberg  Henry Thomas      372490
435110554.0
698          7.8          Mel Stuart  Gene Wilder      178731
4000000.0

```

```

          Metascore
688          91.0
698          67.0
          Series_Title  Released_Year  Runtime  Genre \
321  Das Cabinet des Dr. Caligari      1920      76  Fantasy
568          Nosferatu      1922      94  Fantasy

```

	IMDB_Rating	Director	Star1	No_of_Votes
Gross \				
321	8.1	Robert Wiene	Werner Krauss	57428
337574718.0				
568	7.9	F.W. Murnau	Max Schreck	88794
445151978.0				

	Metascore				
321	NaN				
568	NaN				
\	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating
309	The Third Man	1949	104	Film-Noir	8.1
456	The Maltese Falcon	1941	100	Film-Noir	8.0
712	Shadow of a Doubt	1943	108	Film-Noir	7.8

	Director	Star1	No_of_Votes	Gross
Metascore				
309	Carol Reed	Orson Welles	158731	449191.0
97.0				
456	John Huston	Humphrey Bogart	148928	2108060.0
96.0				
712	Alfred Hitchcock	Teresa Wright	59556	123353292.0
94.0				

	Series_Title	Released_Year	Runtime	Genre
IMDB_Rating \				
49	Psycho	1960	109	Horror
8.5				
75	Alien	1979	117	Horror
8.4				
271	The Thing	1982	109	Horror
8.1				
419	The Exorcist	1973	122	Horror
8.0				
544	Night of the Living Dead	1968	96	Horror
7.9				
707	The Innocents	1961	100	Horror
7.8				
724	Get Out	2017	104	Horror
7.7				
844	Halloween	1978	91	Horror
7.7				
876	The Invisible Man	1933	71	Horror
7.7				
932	Saw	2004	103	Horror
7.6				

948	The Others		2001	101	Horror
7.6					
	Director		Star1	No_of_Votes	Gross
Metascore					
49	Alfred Hitchcock	Anthony Perkins		604211	32000000.0
97.0					
75	Ridley Scott	Sigourney Weaver		787806	78900000.0
89.0					
271	John Carpenter	Kurt Russell		371271	13782838.0
57.0					
419	William Friedkin	Ellen Burstyn		362393	232906145.0
81.0					
544	George A. Romero	Duane Jones		116557	89029.0
89.0					
707	Jack Clayton	Deborah Kerr		27007	2616000.0
88.0					
724	Jordan Peele	Daniel Kaluuya		492851	176040665.0
85.0					
844	John Carpenter	Donald Pleasence		233106	47000000.0
87.0					
876	James Whale	Claude Rains		30683	298791505.0
87.0					
932	James Wan	Cary Elwes		379020	56000369.0
46.0					
948	Alejandro Amenábar	Nicole Kidman		337651	96522687.0
74.0					
	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating \
69	Memento	2000	113	Mystery	8.4
81	Rear Window	1954	112	Mystery	8.4
119	Vertigo	1958	128	Mystery	8.3
145	Shutter Island	2010	138	Mystery	8.2
220	Kahaani	2012	122	Mystery	8.1
393	Twelve Monkeys	1995	129	Mystery	8.0
420	Sleuth	1972	138	Mystery	8.0
714	The Lady Vanishes	1938	96	Mystery	7.8
829	Spoorloos	1988	107	Mystery	7.7
899	El cuerpo	2012	112	Mystery	7.6
959	Dark City	1998	100	Mystery	7.6
961	Lost Highway	1997	134	Mystery	7.6
	Director		Star1	No_of_Votes	
Gross \					
69	Christopher Nolan		Guy Pearce	1125712	
25544867.0					
81	Alfred Hitchcock		James Stewart	444074	
36764313.0					
119	Alfred Hitchcock		James Stewart	364368	
3200000.0					
145	Martin Scorsese		Leonardo DiCaprio	1129894	

128012934.0			
220	Sujoy Ghosh	Vidya Balan	57806
1035953.0			
393	Terry Gilliam	Bruce Willis	578443
57141459.0			
420	Joseph L. Mankiewicz	Laurence Olivier	44748
4081254.0			
714	Alfred Hitchcock	Margaret Lockwood	47400
474203697.0			
829	George Sluizer	Bernard-Pierre Donnadieu	33982
367916835.0			
899	Oriol Paulo	Jose Coronado	57549
140340673.0			
959	Alex Proyas	Rufus Sewell	187927
14378331.0			
961	David Lynch	Bill Pullman	131101
3796699.0			

	Metascore
--	-----------

69	80.0
81	100.0
119	100.0
145	63.0
220	NaN
393	74.0
420	NaN
714	98.0
829	NaN
899	NaN
959	66.0
961	52.0

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating \
700	Wait Until Dark	1967	108	Thriller	7.8

	Director	Star1	No_of_Votes	Gross	Metascore
700	Terence Young	Audrey Hepburn	27733	17550741.0	81.0

	Series_Title	Released_Year	Runtime	
Genre \				
12	Il buono, il brutto, il cattivo	1966	161	Western
48	Once Upon a Time in the West	1968	165	Western
115	Per qualche dollaro in più	1965	132	Western
691	The Outlaw Josey Wales	1976	135	Western

IMDB_Rating	Director	Star1	No_of_Votes
-------------	----------	-------	-------------

Gross \				
12	8.8	Sergio Leone	Clint Eastwood	688390
6100000.0				
48	8.5	Sergio Leone	Henry Fonda	302844
5321508.0				
115	8.3	Sergio Leone	Clint Eastwood	232772
15000000.0				
691	7.8	Clint Eastwood	Clint Eastwood	65659
31800000.0				

	Metascore
12	90.0
48	80.0
115	74.0
691	69.0

C:\Users\jayra\AppData\Local\Temp\ipykernel_23732\2904676014.py:1:
 DeprecationWarning: DataFrameGroupBy.apply operated on the grouping
 columns. This behavior is deprecated, and in a future version of
 pandas the grouping columns will be excluded from the operation.
 Either pass `include_groups=False` to exclude the groupings or
 explicitly select the grouping columns after groupby to silence this
 warning.

genres.apply(foo)

		Series_Title
Released_Year \		
Genre		
Action	2	The Dark Knight
2008		
	5	The Lord of the Rings: The Return of the King
2003		
	8	Inception
2010		
	10	The Lord of the Rings: The Fellowship of the Ring
2001		
	13	The Lord of the Rings: The Two Towers
2002		
...		...
...		
Thriller	700	Wait Until Dark
1967		
Western	12	Il buono, il brutto, il cattivo
1966		
	48	Once Upon a Time in the West
1968		
	115	Per qualche dollaro in più
1965		
	691	The Outlaw Josey Wales

1976

		Runtime	Genre	IMDB_Rating	Director \
Genre					
Action	2	152	Action	9.0	Christopher Nolan
	5	201	Action	8.9	Peter Jackson
	8	148	Action	8.8	Christopher Nolan
	10	178	Action	8.8	Peter Jackson
	13	179	Action	8.7	Peter Jackson
...	
Thriller	700	108	Thriller	7.8	Terence Young
Western	12	161	Western	8.8	Sergio Leone
	48	165	Western	8.5	Sergio Leone
	115	132	Western	8.3	Sergio Leone
	691	135	Western	7.8	Clint Eastwood

		Star1	No_of_Votes	Gross	Metascore
Genre					
Action	2	Christian Bale	2303232	534858444.0	84.0
	5	Elijah Wood	1642758	377845905.0	94.0
	8	Leonardo DiCaprio	2067042	292576195.0	74.0
	10	Elijah Wood	1661481	315544750.0	92.0
	13	Elijah Wood	1485555	342551365.0	87.0
...	
Thriller	700	Audrey Hepburn	27733	17550741.0	81.0
Western	12	Clint Eastwood	688390	6100000.0	90.0
	48	Henry Fonda	302844	5321508.0	80.0
	115	Clint Eastwood	232772	15000000.0	74.0
	691	Clint Eastwood	65659	31800000.0	69.0

[1000 rows x 10 columns]

```
def foo(group):  
    print(group['Series_Title'].str.startswith('A'))    # ek boolean  
    series_milega  
    return group
```

```
genres.apply(foo)
```

```
2      False  
5      False  
8      False  
10     False  
13     False  
...  
968    False  
979    False  
982    False  
983    False  
985    False
```

Name: Series_Title, Length: 172, dtype: bool

21 False
47 False
93 False
110 False
114 False

...

957 False
964 False
966 True
984 False
991 False

Name: Series_Title, Length: 72, dtype: bool

23 False
43 False
46 False
56 False
58 False

...

956 False
971 False
976 False
986 False
992 False

Name: Series_Title, Length: 82, dtype: bool

7 False
15 False
18 False
35 False
38 False

...

923 False
940 False
949 False
952 False
987 False

Name: Series_Title, Length: 88, dtype: bool

19 False
26 False
51 False
52 False
64 False

...

977 False
978 False
989 False
994 True
995 False

Name: Series_Title, Length: 155, dtype: bool

```
1      False
3      False
4      False
6      False
22     False
...
958    False
960    False
974    False
980    False
999    False
Name: Series_Title, Length: 107, dtype: bool
0      False
9      False
11     False
17     False
20     False
...
990    False
993    False
996    False
997    False
998    False
Name: Series_Title, Length: 289, dtype: bool
688    False
698    False
Name: Series_Title, dtype: bool
321    False
568    False
Name: Series_Title, dtype: bool
309    False
456    False
712    False
Name: Series_Title, dtype: bool
49     False
75     True
271    False
419    False
544    False
707    False
724    False
844    False
876    False
932    False
948    False
Name: Series_Title, dtype: bool
69     False
81     False
119    False
```

```

145     False
220     False
393     False
420     False
714     False
829     False
899     False
959     False
961     False
Name: Series_Title, dtype: bool
700     False
Name: Series_Title, dtype: bool
12      False
48      False
115     False
691     False
Name: Series_Title, dtype: bool

```

```

C:\Users\jayra\AppData\Local\Temp\ipykernel_23732\2904676014.py:1:
DeprecationWarning: DataFrameGroupBy.apply operated on the grouping
columns. This behavior is deprecated, and in a future version of
pandas the grouping columns will be excluded from the operation.
Either pass `include_groups=False` to exclude the groupings or
explicitly select the grouping columns after groupby to silence this
warning.
  genres.apply(foo)

```

		Series_Title
Released_Year \		
Genre		
Action	2	The Dark Knight
2008		
	5	The Lord of the Rings: The Return of the King
2003		
	8	Inception
2010		
	10	The Lord of the Rings: The Fellowship of the Ring
2001		
	13	The Lord of the Rings: The Two Towers
2002		
...		...
...		
Thriller	700	Wait Until Dark
1967		
Western	12	Il buono, il brutto, il cattivo
1966		
	48	Once Upon a Time in the West
1968		
	115	Per qualche dollaro in più

1965
691 The Outlaw Josey Wales
1976

		Runtime	Genre	IMDB_Rating	Director \
Genre					
Action	2	152	Action	9.0	Christopher Nolan
	5	201	Action	8.9	Peter Jackson
	8	148	Action	8.8	Christopher Nolan
	10	178	Action	8.8	Peter Jackson
	13	179	Action	8.7	Peter Jackson
...	
Thriller	700	108	Thriller	7.8	Terence Young
Western	12	161	Western	8.8	Sergio Leone
	48	165	Western	8.5	Sergio Leone
	115	132	Western	8.3	Sergio Leone
	691	135	Western	7.8	Clint Eastwood

		Star1	No_of_Votes	Gross	Metascore
Genre					
Action	2	Christian Bale	2303232	534858444.0	84.0
	5	Elijah Wood	1642758	377845905.0	94.0
	8	Leonardo DiCaprio	2067042	292576195.0	74.0
	10	Elijah Wood	1661481	315544750.0	92.0
	13	Elijah Wood	1485555	342551365.0	87.0
...	
Thriller	700	Audrey Hepburn	27733	17550741.0	81.0
Western	12	Clint Eastwood	688390	6100000.0	90.0
	48	Henry Fonda	302844	5321508.0	80.0
	115	Clint Eastwood	232772	15000000.0	74.0
	691	Clint Eastwood	65659	31800000.0	69.0

[1000 rows x 10 columns]

```
def foo(group):
    return (group['Series_Title'].str.startswith('A')).sum() # ek
boolean series milega
```

```
genres.apply(foo)
```

C:\Users\jayra\AppData\Local\Temp\ipykernel_23732\2904676014.py:1:
DeprecationWarning: DataFrameGroupBy.apply operated on the grouping
columns. This behavior is deprecated, and in a future version of
pandas the grouping columns will be excluded from the operation.
Either pass `include_groups=False` to exclude the groupings or
explicitly select the grouping columns after groupby to silence this
warning.

```
genres.apply(foo)
```

```
Genre
Action      10
Adventure    2
Animation    2
Biography    9
Comedy       14
Crime        4
Drama       21
Family       0
Fantasy      0
Film-Noir    0
Horror       1
Mystery      0
Thriller     0
Western      0
dtype: int64
```

```
# find ranking of each movie in the group according to IMDB score
def rank_movie(group):
    group['genre_rank'] = group['IMDB_Rating'].rank(ascending=False)
    return group
```

```
genres.apply(rank_movie)
```

```
C:\Users\jayra\AppData\Local\Temp\ipykernel_23732\2409710219.py:1:
DeprecationWarning: DataFrameGroupBy.apply operated on the grouping
columns. This behavior is deprecated, and in a future version of
pandas the grouping columns will be excluded from the operation.
Either pass `include_groups=False` to exclude the groupings or
explicitly select the grouping columns after groupby to silence this
warning.
```

```
genres.apply(rank_movie)
```

		Series_Title
Released_Year \		
Genre		
Action	2	The Dark Knight
2008		
	5	The Lord of the Rings: The Return of the King
2003		
	8	Inception
2010		
	10	The Lord of the Rings: The Fellowship of the Ring
2001		
	13	The Lord of the Rings: The Two Towers
2002		
...		...
...		
Thriller	700	Wait Until Dark

1967			
Western	12		Il buono, il brutto, il cattivo
1966			
	48		Once Upon a Time in the West
1968			
	115		Per qualche dollaro in più
1965			
	691		The Outlaw Josey Wales
1976			

		Runtime	Genre	IMDB_Rating	Director \
Genre					
Action	2	152	Action	9.0	Christopher Nolan
	5	201	Action	8.9	Peter Jackson
	8	148	Action	8.8	Christopher Nolan
	10	178	Action	8.8	Peter Jackson
	13	179	Action	8.7	Peter Jackson
...					
Thriller	700	108	Thriller	7.8	Terence Young
Western	12	161	Western	8.8	Sergio Leone
	48	165	Western	8.5	Sergio Leone
	115	132	Western	8.3	Sergio Leone
	691	135	Western	7.8	Clint Eastwood

		Star1	No_of_Votes	Gross	
Metascore \					
Genre					
Action	2	Christian Bale	2303232	534858444.0	84.0
	5	Elijah Wood	1642758	377845905.0	94.0
	8	Leonardo DiCaprio	2067042	292576195.0	74.0
	10	Elijah Wood	1661481	315544750.0	92.0
	13	Elijah Wood	1485555	342551365.0	87.0
...					
Thriller	700	Audrey Hepburn	27733	17550741.0	81.0
Western	12	Clint Eastwood	688390	6100000.0	90.0
	48	Henry Fonda	302844	5321508.0	80.0
	115	Clint Eastwood	232772	15000000.0	74.0
	691	Clint Eastwood	65659	31800000.0	69.0

		genre_rank
Genre	Action	2
		1.0
		5
		2.0
		8
...		3.5
		10
		3.5
		13
		6.0
Thriller	700	...
		1.0
	Western	12
		1.0
		48
...		2.0
		115
		3.0
		691
		4.0

[1000 rows x 11 columns]

find normalized IMDB rating group wise

Normalization Formula

$$X \text{ normalized} = \frac{(X - X \text{ minimum})}{(X \text{ maximum} - X \text{ minimum})}$$

```
def normal(group):
    group['norm_rating'] =(group['IMDB_Rating'] -
group['IMDB_Rating'].min())/(group['IMDB_Rating'].max() -
group['IMDB_Rating'].min())
    return group
```

```
genres.apply(normal)
```

C:\Users\jayra\AppData\Local\Temp\ipykernel_23732\2582220829.py:5:
DeprecationWarning: DataFrameGroupBy.apply operated on the grouping columns. This behavior is deprecated, and in a future version of pandas the grouping columns will be excluded from the operation. Either pass `include_groups=False` to exclude the groupings or explicitly select the grouping columns after groupby to silence this

```
warning.  
genres.apply(normal)
```

Released_Year \		Series_Title			
Genre					
Action	2	The Dark Knight			
	5	The Lord of the Rings: The Return of the King			
	8	Inception			
	10	The Lord of the Rings: The Fellowship of the Ring			
	13	The Lord of the Rings: The Two Towers			
...		...			
...					
Thriller	700	Wait Until Dark			
1967					
Western	12	Il buono, il brutto, il cattivo			
	48	Once Upon a Time in the West			
1968					
	115	Per qualche dollaro in più			
1965					
	691	The Outlaw Josey Wales			
1976					
		Runtime	Genre	IMDB_Rating	Director \
Genre					
Action	2	152	Action	9.0	Christopher Nolan
	5	201	Action	8.9	Peter Jackson
	8	148	Action	8.8	Christopher Nolan
	10	178	Action	8.8	Peter Jackson
	13	179	Action	8.7	Peter Jackson
...	
Thriller	700	108	Thriller	7.8	Terence Young
Western	12	161	Western	8.8	Sergio Leone
	48	165	Western	8.5	Sergio Leone
	115	132	Western	8.3	Sergio Leone
	691	135	Western	7.8	Clint Eastwood
		Star1	No_of_Votes	Gross	
Metascore \	Genre				
Genre					
Action	2	Christian Bale	2303232	534858444.0	84.0

	5	Elijah Wood	1642758	377845905.0	94.0
	8	Leonardo DiCaprio	2067042	292576195.0	74.0
	10	Elijah Wood	1661481	315544750.0	92.0
	13	Elijah Wood	1485555	342551365.0	87.0
...	
Thriller	700	Audrey Hepburn	27733	17550741.0	81.0
Western	12	Clint Eastwood	688390	6100000.0	90.0
	48	Henry Fonda	302844	5321508.0	80.0
	115	Clint Eastwood	232772	15000000.0	74.0
	691	Clint Eastwood	65659	31800000.0	69.0

		norm_rating
Genre		
Action	2	1.000000
	5	0.928571
	8	0.857143
	10	0.857143
	13	0.785714
...		...
Thriller	700	NaN
Western	12	1.000000
	48	0.700000
	115	0.500000
	691	0.000000

[1000 rows x 11 columns]

groupby on multiple cols

```
duo = movies.groupby(['Director', 'Star1'])
duo
```

<pandas.core.groupby.generic.DataFrameGroupBy object at 0x0000023A9C458410>

#size

```
duo.size()
```

Director	Star1	
Aamir Khan	Amole Gupte	1
Aaron Sorkin	Eddie Redmayne	1
Abdellatif Kechiche	Léa Seydoux	1

```

Abhishek Chaubey      Shahid Kapoor      1
Abhishek Kapoor      Amit Sadh          1
...
Zaza Urushadze      Lembit Ulfsak      1
Zoya Akhtar          Hrithik Roshan     1
                    Vijay Varma        1
Çagan Irmak          Çetin Tekindor     1
Ömer Faruk Sorak     Cem Yilmaz         1
Length: 898, dtype: int64

```

```
#get_group
```

```
duo.get_group(('Aamir Khan','Amole Gupte'))
```

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	
Director \						
65	Taare Zameen Par	2007	165	Drama	8.4	Aamir Khan

	Star1	No_of_Votes	Gross	Metascore
65	Amole Gupte	168895	1223869.0	NaN

```
# # find the most earning actor->director combo
```

```
duo['Gross'].sum() # abhi sare actor director ka earning ek sath aa jayega
```

Director	Star1	
Aamir Khan	Amole Gupte	1223869.0
Aaron Sorkin	Eddie Redmayne	853090410.0
Abdellatif Kechiche	Léa Seydoux	2199675.0
Abhishek Chaubey	Shahid Kapoor	218428303.0
Abhishek Kapoor	Amit Sadh	1122527.0
		...
Zaza Urushadze	Lembit Ulfsak	144501.0
Zoya Akhtar	Hrithik Roshan	3108485.0
	Vijay Varma	5566534.0
Çagan Irmak	Çetin Tekindor	461855363.0
Ömer Faruk Sorak	Cem Yilmaz	196206077.0

Name: Gross, Length: 898, dtype: float64

```
duo['Gross'].sum().sort_values(ascending=True)
```

Director	Star1	
Anders Thomas Jensen	Ulrich Thomsen	1.305000e+03
Thomas Jahn	Til Schweiger	3.296000e+03
Jaco Van Dormael	Jared Leto	3.600000e+03
Shane Meadows	Paddy Considine	6.013000e+03
Jeong-beom Lee	Won Bin	6.460000e+03
		...
Werner Herzog	Klaus Kinski	1.124605e+09
Christopher Nolan	Christian Bale	1.242940e+09
Billy Wilder	William Holden	1.286779e+09

Anthony Russo	Joe Russo	2.205039e+09
Akira Kurosawa	Toshirô Mifune	2.999877e+09

Name: Gross, Length: 898, dtype: float64

```
duo['Gross'].sum().sort_values(ascending=True).head(1)
```

Director	Star1	
Anders Thomas Jensen	Ulrich Thomsen	1305.0

Name: Gross, dtype: float64

```
# find the best(in-terms of metascore(avg)) actor -> genre combo
movies.groupby(['Star1', 'Genre'])['Metascore'].mean()
```

Star1	Genre	
Aamir Khan	Action	NaN
	Adventure	84.0
	Comedy	67.0
Aaron Taylor-Johnson	Action	66.0
Abhay Deol	Drama	NaN
...		
Zbigniew Zamachowski	Comedy	88.0
Zoey Deschanel	Comedy	76.0
Çetin Tekindor	Drama	NaN
Éric Toledano	Biography	57.0
Ömer Faruk Sorak	Comedy	NaN

Name: Metascore, Length: 829, dtype: float64

```
movies.groupby(['Star1', 'Genre'])['Metascore'].mean().reset_index() #
reset index karne se dataframe ban jayega
```

	Star1	Genre	Metascore
0	Aamir Khan	Action	NaN
1	Aamir Khan	Adventure	84.0
2	Aamir Khan	Comedy	67.0
3	Aaron Taylor-Johnson	Action	66.0
4	Abhay Deol	Drama	NaN
...
824	Zbigniew Zamachowski	Comedy	88.0
825	Zoey Deschanel	Comedy	76.0
826	Çetin Tekindor	Drama	NaN
827	Éric Toledano	Biography	57.0
828	Ömer Faruk Sorak	Comedy	NaN

[829 rows x 3 columns]

```
movies.groupby(['Star1', 'Genre'])
['Metascore'].mean().reset_index().sort_values('Metascore', ascending=F
alse).head(1)
```

	Star1	Genre	Metascore
230	Ellar Coltrane	Drama	100.0

```
# agg on multiple groupby
duo[['Runtime','IMDB_Rating','No_of_Votes','Gross','Metascore']].agg(['min','max','mean'])
```

		Runtime			IMDB_Rating
\		min	max	mean	min
max	mean				
Director	Star1				
Aamir Khan	Amole Gupte	165	165	165.0	8.4
8.4	8.4				
Aaron Sorkin	Eddie Redmayne	129	129	129.0	7.8
7.8	7.8				
Abdellatif Kechiche	Léa Seydoux	180	180	180.0	7.7
7.7	7.7				
Abhishek Chaubey	Shahid Kapoor	148	148	148.0	7.8
7.8	7.8				
Abhishek Kapoor	Amit Sadh	130	130	130.0	7.7
7.7	7.7				
...	
...	
Zaza Urushadze	Lembit Ulfsak	87	87	87.0	8.2
8.2	8.2				
Zoya Akhtar	Hrithik Roshan	155	155	155.0	8.1
8.1	8.1				
	Vijay Varma	154	154	154.0	8.0
8.0	8.0				
Çagan Irmak	Çetin Tekindor	112	112	112.0	8.3
8.3	8.3				
Ömer Faruk Sorak	Cem Yilmaz	127	127	127.0	8.0
8.0	8.0				

		No_of_Votes		
Gross	\	min	max	mean
min				
Director	Star1			
Aamir Khan	Amole Gupte	168895	168895	168895.0
1223869.0				
Aaron Sorkin	Eddie Redmayne	89896	89896	89896.0
853090410.0				
Abdellatif Kechiche	Léa Seydoux	138741	138741	138741.0
2199675.0				
Abhishek Chaubey	Shahid Kapoor	27175	27175	27175.0
218428303.0				
Abhishek Kapoor	Amit Sadh	32628	32628	32628.0
1122527.0				
...	

```

...
Zaza Urushadze      Lembit Ulfsak      40382    40382    40382.0
144501.0
Zoya Akhtar         Hrithik Roshan     67927    67927    67927.0
3108485.0
                    Vijay Varma         31886    31886    31886.0
5566534.0
Çagan Irmak         Çetin Tekindor     78925    78925    78925.0
461855363.0
Ömer Faruk Sorak    Cem Yilmaz          56960    56960    56960.0
196206077.0

```

```

\
Metascore
max      mean      min

```

```

Director      Star1
Aamir Khan     Amole Gupte      1223869.0    1223869.0    NaN
NaN
Aaron Sorkin    Eddie Redmayne   853090410.0  853090410.0  77.0
77.0
Abdellatif Kechiche Léa Seydoux      2199675.0    2199675.0    89.0
89.0
Abhishek Chaubey Shahid Kapoor     218428303.0  218428303.0  NaN
NaN
Abhishek Kapoor Amit Sadh          1122527.0    1122527.0    40.0
40.0

```

```

...      ...      ...
...
Zaza Urushadze      Lembit Ulfsak      144501.0    144501.0    73.0
73.0
Zoya Akhtar         Hrithik Roshan     3108485.0    3108485.0    NaN
NaN
                    Vijay Varma         5566534.0    5566534.0    65.0
65.0
Çagan Irmak         Çetin Tekindor     461855363.0  461855363.0  NaN
NaN
Ömer Faruk Sorak    Cem Yilmaz          196206077.0  196206077.0  NaN
NaN

```

```

Director      Star1      mean
Aamir Khan     Amole Gupte      NaN
Aaron Sorkin    Eddie Redmayne   77.0
Abdellatif Kechiche Léa Seydoux      89.0
Abhishek Chaubey Shahid Kapoor     NaN
Abhishek Kapoor Amit Sadh          40.0
...      ...

```


Zaza Urushadze	Lembit Ulfsak	73.0
Zoya Akhtar	Hrithik Roshan	NaN
	Vijay Varma	65.0
Çagan Irmak	Çetin Tekindor	NaN
Ömer Faruk Sorak	Cem Yilmaz	NaN

[898 rows x 15 columns]

EXERCISE

```
ipl = pd.read_csv('deliveries.csv')
ipl
```

	match_id	inning	batting_team	
bowling_team \				
0	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
1	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
2	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
3	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
4	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
...		
...				
179073	11415	2	Chennai Super Kings	Mumbai Indians
179074	11415	2	Chennai Super Kings	Mumbai Indians
179075	11415	2	Chennai Super Kings	Mumbai Indians
179076	11415	2	Chennai Super Kings	Mumbai Indians
179077	11415	2	Chennai Super Kings	Mumbai Indians

	over	ball	batsman	non_striker	bowler
is_super_over ... \					
0	1	1	DA Warner	S Dhawan	TS Mills
0 ...					
1	1	2	DA Warner	S Dhawan	TS Mills
0 ...					
2	1	3	DA Warner	S Dhawan	TS Mills
0 ...					
3	1	4	DA Warner	S Dhawan	TS Mills
0 ...					
4	1	5	DA Warner	S Dhawan	TS Mills

```

0 ...
... ... ... ... ...
...
179073 20 2 RA Jadeja SR Watson SL Malinga
0 ...
179074 20 3 SR Watson RA Jadeja SL Malinga
0 ...
179075 20 4 SR Watson RA Jadeja SL Malinga
0 ...
179076 20 5 SN Thakur RA Jadeja SL Malinga
0 ...
179077 20 6 SN Thakur RA Jadeja SL Malinga
0 ...

```

	bye_runs	legbye_runs	noball_runs	penalty_runs	batsman_runs
\					
0	0	0	0	0	0
1	0	0	0	0	0
2	0	0	0	0	4
3	0	0	0	0	0
4	0	0	0	0	0
...
179073	0	0	0	0	1
179074	0	0	0	0	2
179075	0	0	0	0	1
179076	0	0	0	0	2
179077	0	0	0	0	0

	extra_runs	total_runs	player_dismissed	dismissal_kind
fielder				
0	0	0	NaN	NaN
NaN				
1	0	0	NaN	NaN
NaN				
2	0	4	NaN	NaN
NaN				
3	0	0	NaN	NaN
NaN				
4	2	2	NaN	NaN

```

NaN
...      ...      ...      ...      ...
...
179073      0      1      NaN      NaN
NaN
179074      0      2      NaN      NaN
NaN
179075      0      1      SR Watson      run out  KH
Pandya
179076      0      2      NaN      NaN
NaN
179077      0      0      SN Thakur      lbw
NaN

```

```
[179078 rows x 21 columns]
```

```
ipl.shape
```

```
(179078, 21)
```

```
ipl.head()
```

```

      match_id  inning      batting_team      bowling_team
over \
0      1      1  Sunrisers Hyderabad  Royal Challengers Bangalore
1
1      1      1  Sunrisers Hyderabad  Royal Challengers Bangalore
1
2      1      1  Sunrisers Hyderabad  Royal Challengers Bangalore
1
3      1      1  Sunrisers Hyderabad  Royal Challengers Bangalore
1
4      1      1  Sunrisers Hyderabad  Royal Challengers Bangalore
1

      ball  batsman non_striker  bowler  is_super_over  ...  bye_runs
\
0      1  DA Warner  S Dhawan  TS Mills      0  ...      0
1      2  DA Warner  S Dhawan  TS Mills      0  ...      0
2      3  DA Warner  S Dhawan  TS Mills      0  ...      0
3      4  DA Warner  S Dhawan  TS Mills      0  ...      0
4      5  DA Warner  S Dhawan  TS Mills      0  ...      0

      legbye_runs  noball_runs  penalty_runs  batsman_runs  extra_runs  \
0      0      0      0      0      0
1      0      0      0      0      0

```

2	0	0	0	4	0
3	0	0	0	0	0
4	0	0	0	0	2

	total_runs	player_dismissed	dismissal_kind	fielder
0	0	NaN	NaN	NaN
1	0	NaN	NaN	NaN
2	4	NaN	NaN	NaN
3	0	NaN	NaN	NaN
4	2	NaN	NaN	NaN

[5 rows x 21 columns]

find the top 10 batsman in terms of run scored

har batsman ne jitne ball kheli hai usko grouped kar lo
`ipl.groupby('batsman')`

<pandas.core.groupby.generic.DataFrameGroupBy object at 0x0000023AA3FC7AA0>

`ipl.groupby('batsman')['batsman_runs'].sum()` *# har group ke batsman ka run sum ho jayega abhi index ke according sort hai value ke according kar lo*

```
batsman
A Ashish Reddy      280
A Chandila          4
A Chopra            53
A Choudhary         25
A Dananjaya         4
...
YV Takawale        192
Yashpal Singh       47
Younis Khan         3
Yuvraj Singh       2765
Z Khan             117
Name: batsman_runs, Length: 516, dtype: int64
```

`ipl.groupby('batsman')['batsman_runs'].sum().sort_values(ascending=False)`

```
batsman
V Kohli      5434
SK Raina     5415
RG Sharma    4914
DA Warner    4741
S Dhawan     4632
...
IC Pandey    0
J Denly      0
```

```
P Raj          0
Sunny Gupta    0
L Ablish       0
Name: batsman_runs, Length: 516, dtype: int64
```

```
ipl.groupby('batsman')
['batsman_runs'].sum().sort_values(ascending=False).head(10)
```

```
batsman
V Kohli          5434
SK Raina         5415
RG Sharma        4914
DA Warner        4741
S Dhawan         4632
CH Gayle         4560
MS Dhoni         4477
RV Uthappa       4446
AB de Villiers   4428
G Gambhir        4223
Name: batsman_runs, dtype: int64
```

```
# find the batsman with max no of sixes
ipl['batsman_runs'] == 6 # ek boolean series mil jayega
```

```
0      False
1      False
2      False
3      False
4      False
...
179073  False
179074  False
179075  False
179076  False
179077  False
Name: batsman_runs, Length: 179078, dtype: bool
```

```
six = ipl[ipl['batsman_runs'] == 6]
six                                     # sare ball ka details mil
jayega jaha pe ki six laga hai
```

	match_id	inning	batting_team
bowling_team \			
10	1	1	Sunrisers Hyderabad Royal Challengers Bangalore
47	1	1	Sunrisers Hyderabad Royal Challengers Bangalore
75	1	1	Sunrisers Hyderabad Royal Challengers Bangalore
89	1	1	Sunrisers Hyderabad Royal Challengers Bangalore

91	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
...
...				
178987	11415	2	Chennai Super Kings	Mumbai Indians
179048	11415	2	Chennai Super Kings	Mumbai Indians
179061	11415	2	Chennai Super Kings	Mumbai Indians
179062	11415	2	Chennai Super Kings	Mumbai Indians
179063	11415	2	Chennai Super Kings	Mumbai Indians

	over	ball	batsman	non_striker	bowler
is_super_over	\				
10	2	4	DA Warner	S Dhawan	A Choudhary
0					
47	8	4	MC Henriques	S Dhawan	TM Head
0					
75	13	2	Yuvraj Singh	MC Henriques	A Choudhary
0					
89	15	3	Yuvraj Singh	MC Henriques	S Aravind
0					
91	15	5	MC Henriques	Yuvraj Singh	S Aravind
0					

...
...					
178987	6	4	SR Watson	SK Raina	SL Malinga
0					
179048	16	1	DJ Bravo	SR Watson	SL Malinga
0					
179061	18	2	SR Watson	DJ Bravo	KH Pandya
0					
179062	18	3	SR Watson	DJ Bravo	KH Pandya
0					
179063	18	4	SR Watson	DJ Bravo	KH Pandya
0					

	...	bye_runs	legbye_runs	noball_runs	penalty_runs
batsman_runs	\				
10	...	0	0	0	0
6					
47	...	0	0	0	0
6					
75	...	0	0	0	0
6					
89	...	0	0	0	0

```

6
91      ...      0      0      0      0
6
...      ...      ...      ...      ...      ...
...
178987  ...      0      0      0      0
6
179048  ...      0      0      0      0
6
179061  ...      0      0      0      0
6
179062  ...      0      0      0      0
6
179063  ...      0      0      0      0
6

      extra_runs  total_runs  player_dismissed  dismissal_kind
fielder
10              0           6                NaN                NaN
NaN
47              0           6                NaN                NaN
NaN
75              0           6                NaN                NaN
NaN
89              0           6                NaN                NaN
NaN
91              0           6                NaN                NaN
NaN
...            ...            ...            ...            ...
.
178987          0           6                NaN                NaN
NaN
179048          0           6                NaN                NaN
NaN
179061          0           6                NaN                NaN
NaN
179062          0           6                NaN                NaN
NaN
179063          0           6                NaN                NaN
NaN

[8170 rows x 21 columns]

six.groupby('batsman')# batsman ke basis pe group kar diye and abb
kisi bhi column ko extract karo suppose batsman colum ko hi pakarte
hain

<pandas.core.groupby.generic.DataFrameGroupBy object at
0x0000023AA3B3BBC0>

```

```
six.groupby('batsman')['batsman'] # abb iski count nikal lo
```

```
<pandas.core.groupby.generic.SeriesGroupBy object at  
0x0000023A9CFF9B80>
```

```
six.groupby('batsman')['batsman'].count() # abb sort kar lo
```

```
batsman  
A Ashish Reddy      15  
A Choudhary         1  
A Flintoff          2  
A Hales             6  
A Mishra            5  
...  
Y Venugopal Rao    37  
YK Pathan          161  
YV Takawale         3  
Yuvraj Singh       149  
Z Khan              2  
Name: batsman, Length: 336, dtype: int64
```

```
six.groupby('batsman')['batsman'].count().sort_values(ascending=False)
```

```
batsman  
CH Gayle           327  
AB de Villiers     214  
MS Dhoni           207  
SK Raina           195  
RG Sharma          194  
...  
CK Langeveldt      1  
JDS Neesham        1  
SK Trivedi         1  
D Shorey           1  
MN van Wyk         1  
Name: batsman, Length: 336, dtype: int64
```

```
six.groupby('batsman')  
['batsman'].count().sort_values(ascending=False).head(1)
```

```
batsman  
CH Gayle      327  
Name: batsman, dtype: int64
```

```
# find batsman with most number of 4's and 6's in last 5 overs
```

```
# hume unn sare ball ki need hi nahi hai jo ki 1 se 15 over tak feka  
gya hai  
temp=ipl[ipl['over']>15]  
temp
```


	match_id	inning	batting_team		
bowling_team \					
93 Bangalore	1	1	Sunrisers Hyderabad	Royal Challengers	
94 Bangalore	1	1	Sunrisers Hyderabad	Royal Challengers	
95 Bangalore	1	1	Sunrisers Hyderabad	Royal Challengers	
96 Bangalore	1	1	Sunrisers Hyderabad	Royal Challengers	
97 Bangalore	1	1	Sunrisers Hyderabad	Royal Challengers	
...			
...					
179073 Indians	11415	2	Chennai Super Kings	Mumbai	
179074 Indians	11415	2	Chennai Super Kings	Mumbai	
179075 Indians	11415	2	Chennai Super Kings	Mumbai	
179076 Indians	11415	2	Chennai Super Kings	Mumbai	
179077 Indians	11415	2	Chennai Super Kings	Mumbai	
	over	ball	batsman	non_striker	bowler
is_super_over \					
93 0	16	1	MC Henriques	Yuvraj Singh	YS Chahal
94 0	16	2	MC Henriques	Yuvraj Singh	YS Chahal
95 0	16	3	Yuvraj Singh	DJ Hooda	YS Chahal
96 0	16	4	DJ Hooda	Yuvraj Singh	YS Chahal
97 0	16	5	Yuvraj Singh	DJ Hooda	YS Chahal
...
...					
179073 0	20	2	RA Jadeja	SR Watson	SL Malinga
179074 0	20	3	SR Watson	RA Jadeja	SL Malinga
179075 0	20	4	SR Watson	RA Jadeja	SL Malinga
179076 0	20	5	SN Thakur	RA Jadeja	SL Malinga
179077 0	20	6	SN Thakur	RA Jadeja	SL Malinga

	...	bye_runs	legbye_runs	noball_runs	penalty_runs
batsman_runs \					

93	...	0	0	0	0
0					
94	...	0	0	0	0
0					
95	...	0	0	0	0
1					
96	...	0	0	0	0
1					
97	...	0	0	0	0
1					
...
...					

179073	...	0	0	0	0
1					
179074	...	0	0	0	0
2					
179075	...	0	0	0	0
1					
179076	...	0	0	0	0
2					
179077	...	0	0	0	0
0					

	extra_runs	total_runs	player_dismissed	dismissal_kind
fielder				

93	0	0	NaN	NaN
NaN				
94	0	0	MC Henriques	caught
Sachin Baby				
95	0	1	NaN	NaN
NaN				
96	0	1	NaN	NaN
NaN				
97	0	1	NaN	NaN
NaN				
...
...				

179073	0	1	NaN	NaN
NaN				
179074	0	2	NaN	NaN
NaN				
179075	0	1	SR Watson	run out
Pandya				
179076	0	2	NaN	NaN
NaN				
179077	0	0	SN Thakur	lbw
NaN				

KH

[40400 rows x 21 columns]

```
temp[(temp['batsman_runs'] ==4) | (temp['batsman_runs'] ==6)]
```

	match_id	inning	batting_team	
bowling_team \				
101	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
105	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
109	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
114	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
115	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
...
...				
179052	11415	2	Chennai Super Kings	Mumbai Indians
179061	11415	2	Chennai Super Kings	Mumbai Indians
179062	11415	2	Chennai Super Kings	Mumbai Indians
179063	11415	2	Chennai Super Kings	Mumbai Indians
179071	11415	2	Chennai Super Kings	Mumbai Indians

	over	ball	batsman	non_striker	bowler
is_super_over \					
101	17	3	Yuvraj Singh	DJ Hooda	TS Mills
0					
105	18	1	DJ Hooda	Yuvraj Singh	A Choudhary
0					
109	18	5	Yuvraj Singh	DJ Hooda	A Choudhary
0					
114	19	2	Yuvraj Singh	DJ Hooda	TS Mills
0					
115	19	3	Yuvraj Singh	DJ Hooda	TS Mills
0					
...
...					
179052	16	5	SR Watson	DJ Bravo	SL Malinga
0					
179061	18	2	SR Watson	DJ Bravo	KH Pandya
0					
179062	18	3	SR Watson	DJ Bravo	KH Pandya
0					

179063	18	4	SR Watson	DJ Bravo	KH Pandya
0					
179071	19	6	RA Jadeja	SR Watson	JJ Bumrah
0					
	...	bye_runs	legbye_runs	noball_runs	penalty_runs
batsman_runs	\				
101	...	0	0	0	0
4					
105	...	0	0	0	0
6					
109	...	0	0	0	0
4					
114	...	0	0	0	0
4					
115	...	0	0	0	0
6					
...
...					
179052	...	0	0	0	0
4					
179061	...	0	0	0	0
6					
179062	...	0	0	0	0
6					
179063	...	0	0	0	0
6					
179071	...	4	0	0	0
4					
	extra_runs	total_runs	player_dismissed	dismissal_kind	
fielder					
101	0	4	NaN	NaN	
NaN					
105	0	6	NaN	NaN	
NaN					
109	0	4	NaN	NaN	
NaN					
114	0	4	NaN	NaN	
NaN					
115	0	6	NaN	NaN	
NaN					
...
.					
179052	0	4	NaN	NaN	
NaN					
179061	0	6	NaN	NaN	
NaN					
179062	0	6	NaN	NaN	

NaN				
179063	0	6	NaN	NaN
NaN				
179071	4	8	NaN	NaN
NaN				

[7592 rows x 21 columns]

```
temp.groupby('batsman')
['batsman'].count().sort_values(ascending=False).head(1).index[0]
```

'MS Dhoni'

find V Kohli's record against all teams

```
temp = ipl[ipl['batsman'] == 'V Kohli']
temp
```

	match_id	inning	batting_team		
bowling_team \					
2590	12	1	Royal Challengers Bangalore	Mumbai	Indians
2591	12	1	Royal Challengers Bangalore	Mumbai	Indians
2593	12	1	Royal Challengers Bangalore	Mumbai	Indians
2594	12	1	Royal Challengers Bangalore	Mumbai	Indians
2597	12	1	Royal Challengers Bangalore	Mumbai	Indians
...			
...					
177522	11345	2	Royal Challengers Bangalore	Sunrisers	Hyderabad
177523	11345	2	Royal Challengers Bangalore	Sunrisers	Hyderabad
177524	11345	2	Royal Challengers Bangalore	Sunrisers	Hyderabad
177525	11345	2	Royal Challengers Bangalore	Sunrisers	Hyderabad
177527	11345	2	Royal Challengers Bangalore	Sunrisers	Hyderabad

	over	ball	batsman	non_striker	bowler
is_super_over \					
2590	1	2	V Kohli	CH Gayle	TG Southee
0					
2591	1	3	V Kohli	CH Gayle	TG Southee
0					
2593	1	5	V Kohli	CH Gayle	TG Southee
0					

2594	1	6	V Kohli	CH Gayle	TG Southee	
0						
2597	2	1	V Kohli	CH Gayle	Harbhajan Singh	
0						
...
...						
177522	1	7	V Kohli	AB de Villiers	B Kumar	
0						
177523	2	1	V Kohli	AB de Villiers	K Ahmed	
0						
177524	2	2	V Kohli	AB de Villiers	K Ahmed	
0						
177525	2	3	V Kohli	AB de Villiers	K Ahmed	
0						
177527	2	5	V Kohli	AB de Villiers	K Ahmed	
0						
	...	bye_runs	legbye_runs	noball_runs	penalty_runs	
batsman_runs	\					
2590	...	0	0	0	0	
0						
2591	...	0	0	0	0	
1						
2593	...	0	0	0	0	
0						
2594	...	0	0	0	0	
1						
2597	...	0	0	0	0	
0						
...
...						
177522	...	0	0	0	0	
1						
177523	...	0	0	0	0	
4						
177524	...	0	0	0	0	
6						
177525	...	0	0	0	0	
1						
177527	...	0	0	0	0	
0						
	extra_runs	total_runs	player_dismissed	dismissal_kind		
fielder						
2590	1	1	NaN	NaN		
NaN						
2591	0	1	NaN	NaN		
NaN						
2593	0	0	NaN	NaN		

NaN				
2594	0	1	NaN	NaN
NaN				
2597	0	0	NaN	NaN
NaN				
...
..				
177522	0	1	NaN	NaN
NaN				
177523	0	4	NaN	NaN
NaN				
177524	0	6	NaN	NaN
NaN				
177525	0	1	NaN	NaN
NaN				
177527	0	0	V Kohli	caught WP
Saha				

[4211 rows x 21 columns]

```
temp.groupby('bowling_team')['batsman_runs'].sum()
```

bowling_team	
Chennai Super Kings	749
Deccan Chargers	306
Delhi Capitals	66
Delhi Daredevils	763
Gujarat Lions	283
Kings XI Punjab	636
Kochi Tuskers Kerala	50
Kolkata Knight Riders	675
Mumbai Indians	628
Pune Warriors	128
Rajasthan Royals	370
Rising Pune Supergiant	83
Rising Pune Supergiants	188
Sunrisers Hyderabad	509

Name: batsman_runs, dtype: int64

```
temp.groupby('bowling_team')['batsman_runs'].sum().reset_index()
```

	bowling_team	batsman_runs
0	Chennai Super Kings	749
1	Deccan Chargers	306
2	Delhi Capitals	66
3	Delhi Daredevils	763
4	Gujarat Lions	283
5	Kings XI Punjab	636
6	Kochi Tuskers Kerala	50
7	Kolkata Knight Riders	675

8	Mumbai Indians	628
9	Pune Warriors	128
10	Rajasthan Royals	370
11	Rising Pune Supergiant	83
12	Rising Pune Supergiants	188
13	Sunrisers Hyderabad	509

Create a function that can return the highest score of any batsman in ipl

	match_id	inning	batting_team	
bowling_team \				
0	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
1	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
2	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
3	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
4	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore
...
...				
179073	11415	2	Chennai Super Kings	Mumbai Indians
179074	11415	2	Chennai Super Kings	Mumbai Indians
179075	11415	2	Chennai Super Kings	Mumbai Indians
179076	11415	2	Chennai Super Kings	Mumbai Indians
179077	11415	2	Chennai Super Kings	Mumbai Indians

	over	ball	batsman	non_striker	bowler
is_super_over ... \					
0	1	1	DA Warner	S Dhawan	TS Mills
0 ...					
1	1	2	DA Warner	S Dhawan	TS Mills
0 ...					
2	1	3	DA Warner	S Dhawan	TS Mills
0 ...					
3	1	4	DA Warner	S Dhawan	TS Mills
0 ...					
4	1	5	DA Warner	S Dhawan	TS Mills
0 ...					
...
...					
179073	20	2	RA Jadeja	SR Watson	SL Malinga

0	...					
179074	20	3	SR Watson	RA Jadeja	SL Malinga	
0	...					
179075	20	4	SR Watson	RA Jadeja	SL Malinga	
0	...					
179076	20	5	SN Thakur	RA Jadeja	SL Malinga	
0	...					
179077	20	6	SN Thakur	RA Jadeja	SL Malinga	
0	...					

	bye_runs	legbye_runs	noball_runs	penalty_runs	batsman_runs
\					
0	0	0	0	0	0
1	0	0	0	0	0
2	0	0	0	0	4
3	0	0	0	0	0
4	0	0	0	0	0
...
179073	0	0	0	0	1
179074	0	0	0	0	2
179075	0	0	0	0	1
179076	0	0	0	0	2
179077	0	0	0	0	0

	extra_runs	total_runs	player_dismissed	dismissal_kind
fielder				
0	0	0	NaN	NaN
NaN				
1	0	0	NaN	NaN
NaN				
2	0	4	NaN	NaN
NaN				
3	0	0	NaN	NaN
NaN				
4	2	2	NaN	NaN
NaN				
...
...				
179073	0	1	NaN	NaN

NaN				
179074	0	2	NaN	NaN
NaN				
179075	0	1	SR Watson	run out KH
Pandya				
179076	0	2	NaN	NaN
NaN				
179077	0	0	SN Thakur	lbw
NaN				

```
[179078 rows x 21 columns]

temp= ipl[ipl['batsman'] == 'V Kohli']
temp.groupby('match_id')
['batsman_runs'].sum().sort_values(ascending=False).head(1).values[0]
113

def highest(batsman):
    temp = ipl[ipl['batsman'] == batsman]
    return temp.groupby('match_id')
['batsman_runs'].sum().sort_values(ascending=False).head(1).values[0]

highest('MS Dhoni')
89

highest('DA Warner')
126

import matplotlib.pyplot as plt
```

	Series_Title	Released_Year	Runtime	Genre
IMDB_Rating \				
0	The Shawshank Redemption	1994	142	Drama
9.3				
1	The Godfather	1972	175	Crime
9.2				
2	The Dark Knight	2008	152	Action
9.0				
3	The Godfather: Part II	1974	202	Crime
9.0				
4	12 Angry Men	1957	96	Crime
9.0				
..
...				
995	Breakfast at Tiffany's	1961	115	Comedy
7.6				

996	Giant	1956	201	Drama
7.6				
997	From Here to Eternity	1953	118	Drama
7.6				
998	Lifeboat	1944	97	Drama
7.6				
999	The 39 Steps	1935	86	Crime
7.6				

	Director	Star1	No_of_Votes	Gross
\				
0	Frank Darabont	Tim Robbins	2343110	28341469.0
1	Francis Ford Coppola	Marlon Brando	1620367	134966411.0
2	Christopher Nolan	Christian Bale	2303232	534858444.0
3	Francis Ford Coppola	Al Pacino	1129952	57300000.0
4	Sidney Lumet	Henry Fonda	689845	4360000.0
..
995	Blake Edwards	Audrey Hepburn	166544	679874270.0
996	George Stevens	Elizabeth Taylor	34075	195217415.0
997	Fred Zinnemann	Burt Lancaster	43374	30500000.0
998	Alfred Hitchcock	Tallulah Bankhead	26471	852142728.0
999	Alfred Hitchcock	Robert Donat	51853	302787539.0

	Metascore
0	80.0
1	100.0
2	84.0
3	90.0
4	96.0
..	...
995	76.0
996	84.0
997	85.0
998	78.0
999	93.0

[1000 rows x 10 columns]