

What is Pandas

Pandas is a fast, powerful, flexible and easy to use open source data analysis and manipulation tool,built on top of the Python programming language.

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Pandas Series

A Pandas Series is like a column in a table. It is a 1-D array holding data of any type.

Importing Pandas

```
In [4]: import numpy as np  
import pandas as pd
```

Series using String

```
In [6]: # string  
country = ['India', 'Pakistan', 'USA', 'Nepal', 'Srilanka']  
pd.Series(country)
```

```
Out[6]: 0      India  
1    Pakistan  
2        USA  
3      Nepal  
4    Srilanka  
dtype: object
```

```
In [7]: # integers
marks= [13,24,56,78,100]
pd.Series(marks)
```

```
Out[7]: 0      13
1      24
2      56
3      78
4     100
dtype: int64
```

```
In [8]: # custom index
marks = [67,57,89,100]
subjects = ['maths','english','science','hindi']

pd.Series(marks,index=subjects)
```

```
Out[8]: maths      67
english    57
science    89
hindi      100
dtype: int64
```

```
In [10]: # setting a name
marks = pd.Series(marks , index=subjects , name="Jack Marks")
marks
```

```
Out[10]: maths      67
english    57
science    89
hindi      100
Name: Jack Marks, dtype: int64
```

Series from dictionary

When a Pandas Series is converted to a dictionary using the `to_dict()` method, the resulting dictionary has the same keys and values as the Series. The index values of the Series become the keys in the dictionary, and the corresponding values become the values in the dictionary.

```
In [11]: marks = {
    'maths':67,
    'english':57,
    'science':89,
    'hindi':100
}
marks_series = pd.Series(marks,name="jack Marks")
```

```
In [12]: marks_series
```

```
Out[12]: maths      67  
english     57  
science     89  
hindi       100  
Name: jack Marks, dtype: int64
```

Series Attributes

size: Returns the number of elements in the Series.

```
In [13]: # size  
marks_series.size
```

```
Out[13]: 4
```

dtype: Returns the data type of the values in the Series.

```
In [14]: # dtype  
marks_series.dtype
```

```
Out[14]: dtype('int64')
```

name: Returns the name of the Series.

```
In [15]: # name  
marks_series.name
```

```
Out[15]: 'jack Marks'
```

unique is an attribute of a Pandas Series that returns an array of the unique values in the Series.

```
In [16]: # is_unique  
marks_series.is_unique
```

```
Out[16]: True
```

```
In [17]: pd.Series([1,1,2,3,4,44,2]).is_unique #It gives false because of repetition
```

```
Out[17]: False
```

index: Returns the index labels of the Series.

```
In [18]: # index
marks_series.index
```

```
Out[18]: Index(['maths', 'english', 'science', 'hindi'], dtype='object')
```

values: Returns the data contained in the Series as a NumPy array.

```
In [19]: # values
marks_series.values
```

```
Out[19]: array([ 67,  57,  89, 100], dtype=int64)
```

```
In [20]: type(marks_series.values)
```

```
Out[20]: numpy.ndarray
```

Series using read_csv

```
In [21]: # with one col
sub = pd.read_csv("D:\\\\datascience\\\\Nitish isr\\\\Pandas\\\\subs.csv")
```

Pandas.read_csv

Automatically converts everything into data frames not in series.

```
In [23]: type(sub)
```

```
Out[23]: pandas.core.frame.DataFrame
```

```
In [30]: sub.head(5)
```

```
Out[30]: Subscribers gained
```

0	48
1	57
2	40
3	43
4	44

To convert data into series,

we have to apply a parameter called as "Squeeze" is equals to True.

```
In [31]: sub = pd.read_csv("subs.csv",squeeze=True)
```

```
In [32]: type(sub)
```

```
Out[32]: pandas.core.series.Series
```

```
In [33]: sub
```

```
Out[33]: 0      48
1      57
2      40
3      43
4      44
...
360    231
361    226
362    155
363    144
364    172
Name: Subscribers gained, Length: 365, dtype: int64
```

```
In [56]: #With 2 col
k1=pd.read_csv("kohli_ipl.csv",index_col="match_no",squeeze=True)
```

```
In [57]: k1
```

```
Out[57]: match_no
1      1
2      23
3      13
4      12
5      1
...
211    0
212    20
213    73
214    25
215    7
Name: runs, Length: 215, dtype: int64
```

```
In [37]: movies=pd.read_csv( "bollywood.csv", index_col="movie",squeeze=True)
```

In [38]: movies

Out[38]: movie

Uri: The Surgical Strike	Vicky Kaushal
Battalion 609	Vicky Ahuja
The Accidental Prime Minister (film)	Anupam Kher
Why Cheat India	Emraan Hashmi
Evening Shadows	Mona Ambegaonkar
...	
Hum Tumhare Hain Sanam	Shah Rukh Khan
Aankhen (2002 film)	Amitabh Bachchan
Saathiya (film)	Vivek Oberoi
Company (film)	Ajay Devgn
Awara Paagal Deewana	Akshay Kumar

Name: lead, Length: 1500, dtype: object

Series Methods

head(n): Returns the first n elements of the Series.

In [40]: # Head
sub.head()

Out[40]: 0 48
1 57
2 40
3 43
4 44
Name: Subscribers gained, dtype: int64

tail(n): Returns the last n elements of the Series.

In [41]: # tail
kl.tail()

Out[41]: match_no
211 0
212 20
213 73
214 25
215 7
Name: runs, dtype: int64

In [43]: # sample - Gives random data
movies.sample()

Out[43]: movie
Enemmy Sunil Shetty
Name: lead, dtype: object

value_counts(): Returns a Series containing the counts of unique values in the Series.

In [44]: # Value Counts
movies.value_counts()

Out[44]: Akshay Kumar 48
Amitabh Bachchan 45
Ajay Devgn 38
Salman Khan 31
Sanjay Dutt 26
..
Diganth 1
Parveen Kaur 1
Seema Azmi 1
Akanksha Puri 1
Edwin Fernandes 1
Name: lead, Length: 566, dtype: int64

In [45]: #sort_values - temporary changes ##### sort_values(): Returns a sorted Series by kl.sort_values()

Out[45]: match_no
87 0
211 0
207 0
206 0
91 0
...
164 100
120 100
123 108
126 109
128 113
Name: runs, Length: 215, dtype: int64

In [50]: # method chaining
kl.sort_values(ascending=False).head(1).values[0]

Out[50]: 113

```
In [55]: # For permanent Changes use Inplace
kl.sort_values(inplace=True)
kl
```

```
Out[55]: match_no
87      0
211     0
207     0
206     0
91      0
...
164    100
120    100
123    108
126    109
128    113
Name: runs, Length: 215, dtype: int64
```

```
In [60]: # sort_index -> inplace -> movies

movies.sort_index()
```

```
Out[60]: movie
1920 (film)           Rajniesh Duggall
1920: London          Sharman Joshi
1920: The Evil Returns Vicky Ahuja
1971 (2007 film)      Manoj Bajpayee
2 States (2014 film) Arjun Kapoor
...
Zindagi 50-50          Veena Malik
Zindagi Na Milegi Dobara Hrithik Roshan
Zindagi Tere Naam      Mithun Chakraborty
Zokkomon               Darsheel Safary
Zor Lagaa Ke...Haiya!   Meghan Jadhav
Name: lead, Length: 1500, dtype: object
```

```
In [61]: movies.sort_index(ascending=False)
```

```
Out[61]: movie
Zor Lagaa Ke...Haiya!   Meghan Jadhav
Zokkomon               Darsheel Safary
Zindagi Tere Naam      Mithun Chakraborty
Zindagi Na Milegi Dobara Hrithik Roshan
Zindagi 50-50          Veena Malik
...
2 States (2014 film)  Arjun Kapoor
1971 (2007 film)      Manoj Bajpayee
1920: The Evil Returns Vicky Ahuja
1920: London          Sharman Joshi
1920 (film)           Rajniesh Duggall
Name: lead, Length: 1500, dtype: object
```

Series Maths Methods

Difference between Count And Size

Count gives the total number of items present in the series. But only NON missing values but, if we have missing values ,it doesnt count them . But, size gives the total item including missing values

```
In [62]: # count  
k1.count()
```

```
Out[62]: 215
```

sum(): Returns the sum of the values in the Series.

```
In [66]: # sum -> Product  
sub.sum()
```

```
Out[66]: 49510
```

```
In [67]: sub.product() # Multiply the items
```

```
Out[67]: 0
```

Statistical Methods

mean(): Returns the mean value of the Series.

```
In [68]: # mean
```

```
sub.mean()
```

```
Out[68]: 135.64383561643837
```

median(): Returns the median value of the Series.

```
In [72]: # median  
k1.median()
```

```
Out[72]: 24.0
```

mode(): The mode is the value that appears most frequently in the Series.

```
In [74]: # mode  
print(movies.mode())
```

```
0    Akshay Kumar  
dtype: object
```

std(): Returns the standard deviation of the values in the Series.

```
In [71]: # std -> Standard deviation  
sub.std()
```

```
Out[71]: 62.67502303725269
```

```
In [75]: # var -> variance  
sub.var()
```

```
Out[75]: 3928.1585127201556
```

min(): Returns the minimum value of the Series.

```
In [76]: # min  
sub.min()
```

```
Out[76]: 33
```

max(): Returns the maximum value of the Series.

```
In [77]: # max  
sub.max()
```

```
Out[77]: 396
```

describe(): Generates descriptive statistics of the Series.

```
In [79]: # describe  
movies.describe()
```

```
Out[79]: count          1500  
unique         566  
top    Akshay Kumar  
freq            48  
Name: lead, dtype: object
```

```
In [80]: kl.describe()
```

```
Out[80]: count    215.000000
mean      30.855814
std       26.229801
min       0.000000
25%      9.000000
50%     24.000000
75%     48.000000
max     113.000000
Name: runs, dtype: float64
```

```
In [81]: sub.describe()
```

```
Out[81]: count    365.000000
mean      135.643836
std       62.675023
min      33.000000
25%     88.000000
50%    123.000000
75%    177.000000
max     396.000000
Name: Subscribers gained, dtype: float64
```

Series Indexing

```
In [83]: # integer indexing
x = pd.Series([12,13,14,35,46,57,58,79,9])
x[1]
```

```
Out[83]: 13
```

```
In [85]: # negative indexing
movies[-1]
```

```
Out[85]: 'Akshay Kumar'
```

```
In [86]: movies[0]
```

```
Out[86]: 'Vicky Kaushal'
```

```
In [87]: sub[0]
```

```
Out[87]: 48
```

In [90]: # slicing
k1[4:10]

Out[90]: match_no
5 1
6 9
7 34
8 0
9 21
10 3
Name: runs, dtype: int64

In [95]: #Negative slicing

sub[-5:]

Out[95]: 360 231
361 226
362 155
363 144
364 172
Name: Subscribers gained, dtype: int64

In [96]: movies[-5:]

Out[96]: movie
Hum Tumhare Hain Sanam Shah Rukh Khan
Aankhen (2002 film) Amitabh Bachchan
Saathiya (film) Vivek Oberoi
Company (film) Ajay Devgn
Awara Paagal Deewana Akshay Kumar
Name: lead, dtype: object

In [97]: movies[::-2]

Out[97]: movie
Uri: The Surgical Strike Vicky Kaushal
The Accidental Prime Minister (film) Anupam Kher
Evening Shadows Mona Ambegaonkar
Fraud Saiyaan Arshad Warsi
Manikarnika: The Queen of Jhansi Kangana Ranaut
...
Raaz (2002 film) Dino Morea
Waisa Bhi Hota Hai Part II Arshad Warsi
Kaante Amitabh Bachchan
Aankhen (2002 film) Amitabh Bachchan
Company (film) Ajay Devgn
Name: lead, Length: 750, dtype: object

```
In [98]: # Fancy indexing
k1[[1,8,22,11,2]]
```

```
Out[98]: match_no
1      1
8      0
22     38
11     10
2      23
Name: runs, dtype: int64
```

```
In [99]: # Fancy indexing -> indexing with Labels
movies
```

```
Out[99]: movie
Uri: The Surgical Strike           Vicky Kaushal
Battalion 609                      Vicky Ahuja
The Accidental Prime Minister (film) Anupam Kher
Why Cheat India                    Emraan Hashmi
Evening Shadows                     Mona Ambegaonkar
                                         ...
Hum Tumhare Hain Sanam            Shah Rukh Khan
Aankhen (2002 film)               Amitabh Bachchan
Saathiya (film)                  Vivek Oberoi
Company (film)                   Ajay Devgn
Awara Paagal Deewana             Akshay Kumar
Name: lead, Length: 1500, dtype: object
```

```
In [100]: movies['Evening Shadows']
```

```
Out[100]: 'Mona Ambegaonkar'
```

Editing the series

```
In [101]: # using the index number
marks_series
```

```
Out[101]: maths      67
english    57
science    89
hindi      100
Name: jack Marks, dtype: int64
```

```
In [102]: marks_series[1]=88
marks_series
```

```
Out[102]: maths      67
english    88
science    89
hindi      100
Name: jack Marks, dtype: int64
```

```
In [103]: # we can add data , if it doesnt exist
marks_series['social']=90
marks_series
```

```
Out[103]: maths      67
english     88
science     89
hindi       100
social      90
Name: jack Marks, dtype: int64
```

```
In [111]: # using index Label
movies
```

```
Out[111]: movie
Uri: The Surgical Strike           Vicky Kaushal
Battalion 609                      Vicky Ahuja
The Accidental Prime Minister (film) Anupam Kher
Why Cheat India                     Emraan Hashmi
Evening Shadows                     Mona Ambegaonkar
...
Hum Tumhare Hain Sanam            Shah Rukh Khan
Aankhen (2002 film)                Amitabh Bachchan
Saathiya (film)                   Vivek Oberoi
Company (film)                    Ajay Devgn
Awara Paagal Deewana              Akshay Kumar
Name: lead, Length: 1500, dtype: object
```

```
In [114]: movies['Hum Tumhare Hain Sanam'] = 'Jack'
```

```
In [115]: movies
```

```
Out[115]: movie
Uri: The Surgical Strike           Vicky Kaushal
Battalion 609                      Vicky Ahuja
The Accidental Prime Minister (film) Anupam Kher
Why Cheat India                     Emraan Hashmi
Evening Shadows                     Mona Ambegaonkar
...
Hum Tumhare Hain Sanam            Jack
Aankhen (2002 film)                Amitabh Bachchan
Saathiya (film)                   Vivek Oberoi
Company (film)                    Ajay Devgn
Awara Paagal Deewana              Akshay Kumar
Name: lead, Length: 1500, dtype: object
```

Series with Python Functionalities

```
In [117]: # Len/type/dir/sorted/max/min
print(len(sub))
print(type(sub))
```

```
365
<class 'pandas.core.series.Series'>
```

```
In [122]: print(dir(sub))
print(sorted(sub))
```

```
[ 'T', '_AXIS_LEN', '_AXIS_ORDERS', '_AXIS_REVERSED', '_AXIS_TO_AXIS_NUMBER',
'_HANDLED_TYPES', '__abs__', '__add__', '__and__', '__annotations__', '__array__',
 '__array_priority__', '__array_ufunc__', '__array_wrap__', '__bool__',
 '__class__', '__contains__', '__copy__', '__deepcopy__', '__delattr__', '__de-
litem__', '__dict__', '__dir__', '__divmod__', '__doc__', '__eq__', '__finali-
ze__', '__float__', '__floordiv__', '__format__', '__ge__', '__getattr__', '__
getattribute__', '__getitem__', '__getstate__', '__gt__', '__hash__', '__iad-
d__', '__iand__', '__ifloordiv__', '__imod__', '__imul__', '__init__', '__ini-
t_subclass__', '__int__', '__invert__', '__ior__', '__ipow__', '__isub__', '__
iter__', '__itruediv__', '__ixor__', '__le__', '__len__', '__long__', '__lt__',
 '__matmul__', '__mod__', '__module__', '__mul__', '__ne__', '__neg__', '__
new__', '__nonzero__', '__or__', '__pos__', '__pow__', '__radd__', '__rand__',
 '__rdivmod__', '__reduce__', '__reduce_ex__', '__repr__', '__rfloordiv__',
 '__rmatmul__', '__rmod__', '__rmul__', '__ror__', '__round__', '__rpow__',
 '__rsub__', '__rtruediv__', '__rxor__', '__setattr__', '__setitem__', '__
setstate__', '__sizeof__', '__str__', '__sub__', '__subclasshook__', '__true-
div__', '__weakref__', '__xor__', '__accessors__', '__accum_func', '__add_nume-
ric_o-
perations', '__agg_by_level', '__agg_examples_doc', '__agg_see_also_doc', '__alig-
n_fra-
me', '__align_series', '__arith_method', '__as_manager', '__attrs', '__bino-
p', '__can_hold_na', '__check_inplace_and_allows_duplicate_labels', '__check_inp-
lace_setting', '__check_is_chained_assignment_possible', '__check_label_or_leve-
l_ambiguity', '__check_setitem_copy', '__clear_item_cache', '__clip_with_one_bou-
nd', '__clip_with_scalar', '__cmp_method', '__consolidate', '__consolidate_inpla-
ce', '__construct_axes_dict', '__construct_axes_from_arguments', '__construct_res-
ult', '__constructor', '__constructor_expanddim', '__convert', '__convert_dtype',
 '__data', '__dir_additions', '__dir_deletions', '__drop_axis', '__drop_labels_
or_levels', '__duplicated', '__find_valid_index', '__flags', '__from_mgr', '__get-
axis', '__get_axis_name', '__get_axis_number', '__get_axis_resolvers', '__get_blo-
ck_manager_axis', '__get_bool_data', '__get_cacher', '__get_cleaned_column_resol-
vers', '__get_index_resolvers', '__get_label_or_level_values', '__get_numeric_da-
ta', '__get_value', '__get_values', '__get_values_tuple', '__get_with', '__gotite-
m', '__hidden_attrs', '__index', '__indexed_same', '__info_axis', '__info_axis_nam-
e', '__info_axis_number', '__init_dict', '__init_mngr', '__inplace_method', '__inte-
rnal_names', '__internal_names_set', '__is_cached', '__is_copy', '__is_label_or_l-
evel_re-
ference', '__is_label_reference', '__is_level_reference', '__is_mixed_type',
 '__is_view', '__item_cache', '__ixs', '__logical_func', '__logical_method', '__
map_values', '__maybe_update_cacher', '__memory_usage', '__metadata', '__mgr', '__
min_count_stat_function', '__name', '__needs_reindex_multi', '__protect_consolid-
ate', '__reduce', '__reindex_axes', '__reindex_indexer', '__reindex_multi', '__rei-
n-
dex_with_indexers', '__replace_single', '__repr_data_resource__', '__repr_latex__',
 '__reset_cache', '__reset_cacher', '__set_as_cached', '__set_axis', '__set_axi-
s_name', '__set_axis_nocheck', '__set_is_copy', '__set_labels', '__set_name', '__s-
et_value', '__set_values', '__set_with', '__set_with_engine', '__slice', '__stat_a-
xis', '__stat_axis_name', '__stat_axis_number', '__stat_function', '__stat_functi-
on_ddof', '__take_with_is_copy', '__typ', '__update_inplace', '__validate_dtype',
 '__values', '__where', 'abs', 'add', 'add_prefix', 'add_suffix', 'agg', 'aggreg-
ate', 'align', 'all', 'any', 'append', 'apply', 'argmax', 'argmin', 'argsort',
 'array', 'asfreq', 'asof', 'astype', 'at', 'at_time', 'attrs', 'autocor-
r', 'axes', 'backfill', 'between', 'between_time', 'bfill', 'bool', 'clip',
 'combine', 'combine_first', 'compare', 'convert_dtypes', 'copy', 'corr', 'cou-
nt', 'cov', 'cummax', 'cummin', 'cumprod', 'cumsum', 'describe', 'diff', 'di-
v', 'divide', 'divmod', 'dot', 'drop', 'drop_duplicates', 'droplevel', 'dropn-
a', 'dtype', 'dtypes', 'duplicated', 'empty', 'eq', 'equals', 'ewm', 'expandi-
ng', 'explode', 'factorize', 'ffill', 'fillna', 'filter', 'first', 'first_val-
id_index', 'flags', 'floordiv', 'ge', 'get', 'groupby', 'gt', 'hasnans', 'hea-
d', 'hist', 'iat', 'idxmax', 'idxmin', 'iloc', 'index', 'infer_objects', 'int-
erpolate', 'is_monotonic', 'is_monotonic_decreasing', 'is_monotonic_increas-
in
```

```

g', 'is_unique', 'isin', 'isna', 'isnull', 'item', 'items', 'iteritems', 'key
s', 'kurt', 'kurtosis', 'last', 'last_valid_index', 'le', 'loc', 'lt', 'mad',
'map', 'mask', 'max', 'mean', 'median', 'memory_usage', 'min', 'mod', 'mode',
'mul', 'multiply', 'name', ' nbytes', 'ndim', 'ne', 'nlargest', 'notna', 'notn
ull', 'nsmallest', 'nunique', 'pad', 'pct_change', 'pipe', 'plot', 'pop', 'po
w', 'prod', 'product', 'quantile', 'radd', 'rank', 'ravel', 'rdiv', 'rdivmo
d', 'reindex', 'reindex_like', 'rename', 'rename_axis', 'reorder_levels', 're
peat', 'replace', 'resample', 'reset_index', 'rfloordiv', 'rmod', 'rmul', 'ro
lling', 'round', 'rpow', 'rsub', 'rtruediv', 'sample', 'searchsorted', 'sem',
'set_axis', 'set_flags', 'shape', 'shift', 'size', 'skew', 'slice_shift', 'so
rt_index', 'sort_values', 'squeeze', 'std', 'sub', 'subtract', 'sum', 'swapax
es', 'swaplevel', 'tail', 'take', 'to_clipboard', 'to_csv', 'to_dict', 'to_ex
cel', 'to_frame', 'to_hdf', 'to_json', 'to_latex', 'to_list', 'to_markdown',
'to_numpy', 'to_period', 'to_pickle', 'to_sql', 'to_string', 'to_timestamp',
'to_xarray', 'transform', 'transpose', 'truediv', 'truncate', 'tz_convert',
'tz_localize', 'unique', 'unstack', 'update', 'value_counts', 'values', 'va
r', 'view', 'where', 'xs']
[33, 33, 35, 37, 39, 40, 40, 40, 42, 42, 43, 44, 44, 44, 45, 46, 46, 48,
49, 49, 49, 50, 50, 50, 51, 54, 56, 56, 56, 56, 57, 61, 62, 64, 65, 65,
66, 66, 66, 67, 68, 70, 70, 70, 71, 71, 72, 72, 72, 72, 73, 74, 74,
75, 76, 76, 76, 77, 77, 78, 78, 78, 79, 79, 80, 80, 80, 81, 81, 82, 82,
83, 83, 84, 84, 84, 85, 86, 86, 86, 87, 87, 87, 87, 88, 88, 88, 88, 88,
89, 89, 89, 90, 90, 90, 90, 91, 92, 92, 92, 93, 93, 93, 93, 95, 95, 96,
96, 96, 97, 97, 98, 98, 98, 99, 99, 100, 100, 100, 100, 101, 101, 101,
101, 102, 102, 102, 103, 103, 104, 104, 104, 105, 105, 105, 105, 105,
105, 105, 105, 105, 105, 105, 105, 105, 105, 105, 105, 105, 105, 108,
108, 108, 108, 109, 109, 110, 110, 110, 111, 111, 112, 113, 113, 113,
114, 114, 114, 115, 115, 115, 115, 117, 117, 117, 117, 118, 118, 118,
119, 119, 119, 119, 119, 120, 122, 123, 123, 123, 123, 123, 124, 125,
125, 126, 126, 127, 127, 128, 128, 128, 129, 130, 131, 131, 132, 132,
134, 134, 134, 135, 135, 135, 136, 136, 136, 136, 137, 138, 138,
138, 139, 140, 144, 145, 146, 146, 146, 146, 147, 147, 149, 150, 150,
150, 150, 150, 151, 152, 152, 152, 153, 153, 153, 154, 154, 154, 154,
155, 155, 156, 156, 156, 156, 156, 157, 157, 157, 158, 158, 158,
159, 159, 160, 160, 160, 160, 160, 162, 164, 166, 167, 167, 168,
170, 170, 170, 170, 171, 172, 172, 173, 173, 173, 174, 174, 174,
175, 175, 176, 176, 177, 178, 178, 179, 179, 180, 180, 180, 180, 182,
182, 183, 183, 183, 183, 184, 184, 184, 185, 185, 185, 185, 186, 186,
186, 186, 186, 188, 188, 189, 190, 190, 190, 192, 192, 192, 192,
196, 196, 196, 197, 197, 202, 202, 202, 203, 204, 204, 206, 206, 207,
207, 209, 210, 210, 210, 211, 212, 213, 214, 216, 216, 219, 220, 221,
221, 222, 222, 224, 225, 225, 225, 226, 227, 227, 228, 229, 230,
231, 233, 236, 236, 236, 237, 241, 243, 244, 244, 245, 245, 247, 247,
249, 249, 254, 254, 258, 259, 259, 261, 261, 265, 267, 268, 269, 276,
276, 276, 290, 290, 295, 301, 306, 312, 396]

```

In [123]: `print(min(sub))
print(max(sub))`

33
396

In [125]: `# type conversion
list(marks_series)`

Out[125]: [67, 88, 89, 100, 90]

```
In [126]: dict(marks_series)
```

```
Out[126]: {'maths': 67, 'english': 88, 'science': 89, 'hindi': 100, 'social': 90}
```

```
In [129]: # membership operator  
'Hum Tumhare Hain Sanam' in movies # In oprator only searches in index values
```

```
Out[129]: True
```

```
In [133]: "Jack" in movies.values
```

```
Out[133]: True
```

```
In [138]: # Looping  
for i in movies:  
    print(i)
```

```
Vicky Kaushal  
Vicky Ahuja  
Anupam Kher  
Emraan Hashmi  
Mona Ambegaonkar  
Geetika Vidya Ohlyan  
Arshad Warsi  
Radhika Apte  
Kangana Ranaut  
Nawazuddin Siddiqui  
Ali Asgar  
Ranveer Singh  
Prit Kamani  
Ajay Devgn  
Sushant Singh Rajput  
Amitabh Bachchan  
Abhimanyu Dasani  
Talha Arshad Reshi  
Nawazuddin Siddiqui
```

```
In [139]: for i in movies.index:  
    print(i)
```

```
Uri: The Surgical Strike  
Battalion 609  
The Accidental Prime Minister (film)  
Why Cheat India  
Evening Shadows  
Soni (film)  
Fraud Saiyaan  
Bombairiya  
Manikarnika: The Queen of Jhansi  
Thackeray (film)  
Amavas  
Gully Boy  
Hum Chaar  
Total Dhamaal  
Sonchiriya  
Badla (2019 film)  
Mard Ko Dard Nahi Hota  
Hamid (film)  
Photograph (film)  
..'
```

```
In [140]: # Arthematic Operators (Broadcasting)  
100+marks_series
```

```
Out[140]: maths      33  
english     12  
science     11  
hindi        0  
social      10  
Name: jack Marks, dtype: int64
```

```
In [141]: 100+marks_series
```

```
Out[141]: maths      167  
english     188  
science     189  
hindi       200  
social      190  
Name: jack Marks, dtype: int64
```

In [143]: # Relational operators
k1>=50

Out[143]: match_no
1 False
2 False
3 False
4 False
5 False
...
211 False
212 False
213 True
214 False
215 False
Name: runs, Length: 215, dtype: bool

Boolean Indexing on Series

In [146]: # Find no of 50's and 100's scored by kohli
k1[k1>=50].size

Out[146]: 50

In [148]: # find number of ducks
k1[k1 == 0].size

Out[148]: 9

In [149]: # Count number of day when I had more than 200 subs a day
sub[sub>=200].size

Out[149]: 59

In [159]: # find actors who have done more than 20 movies
num_mov=movies.value_counts()
num_mov[num_mov>=20]

Out[159]: Akshay Kumar 48
Amitabh Bachchan 45
Ajay Devgn 38
Salman Khan 31
Sanjay Dutt 26
Shah Rukh Khan 21
Emraan Hashmi 21
Name: lead, dtype: int64

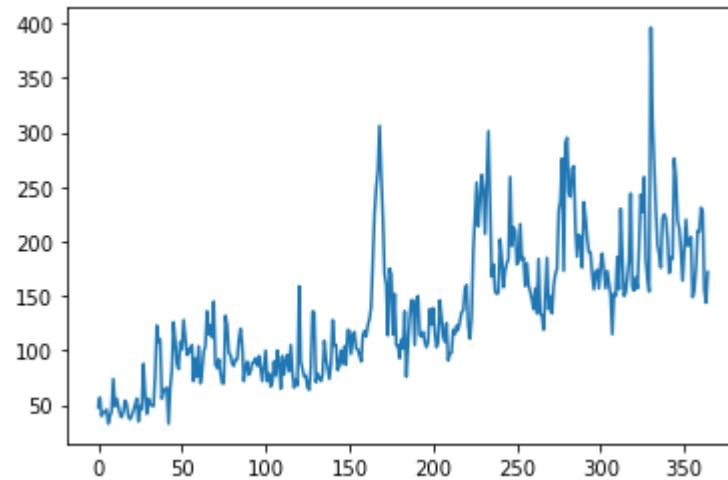
In [160]: num_mov[num_mov>=20].size

Out[160]: 7

Plotting Graphs on Series

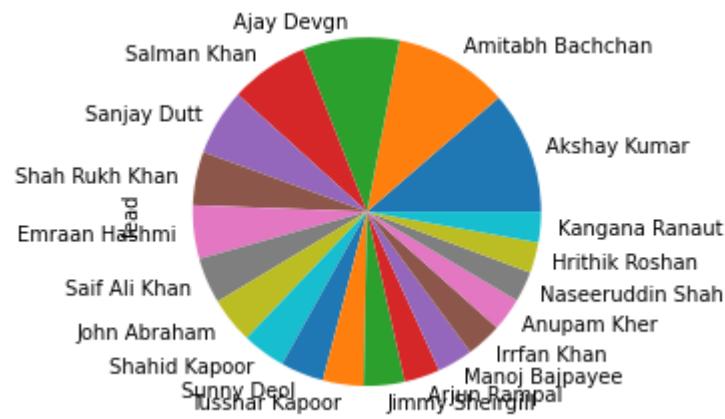
In [162]: `sub.plot()`

Out[162]: <AxesSubplot:>



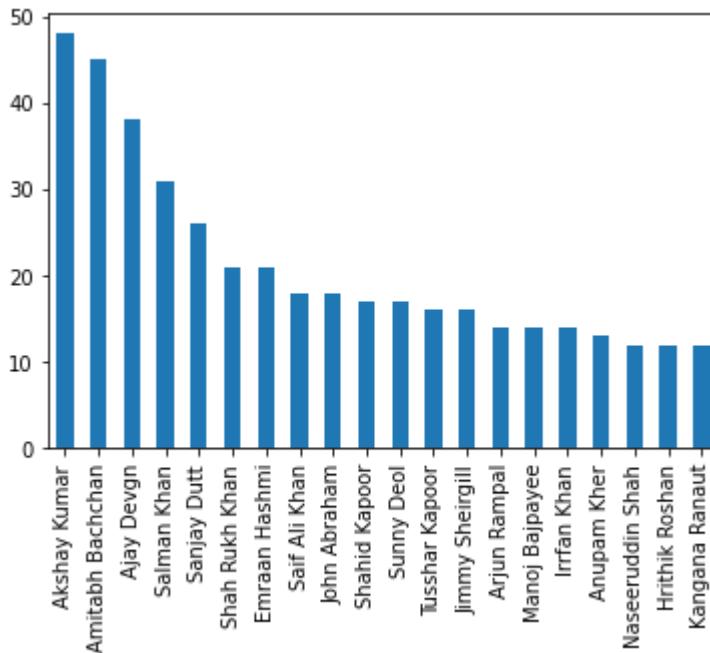
In [164]: `movies.value_counts().head(20).plot(kind="pie")`

Out[164]: <AxesSubplot:ylabel='lead'>



In [165]: `movies.value_counts().head(20).plot(kind="bar")`

Out[165]: <AxesSubplot:>



Some Important Series Methods

In [166]: `# astype
between
clip
drop_duplicates
isnull
dropna
fillna
isin
apply
copy`

In [175]: `# astype
import sys
sys.getsizeof(k1)`

Out[175]: 11752

In [176]: k1

Out[176]:

```
match_no
1      1
2     23
3     13
4     12
5      1
...
211    0
212   20
213   73
214   25
215    7
Name: runs, Length: 215, dtype: int64
```

In [177]: (k1.astype("int16"))

Out[177]:

```
match_no
1      1
2     23
3     13
4     12
5      1
...
211    0
212   20
213   73
214   25
215    7
Name: runs, Length: 215, dtype: int16
```

In [178]: sys.getsizeof(k1.astype("int16"))

Out[178]: 10462

In [181]: # between
k1[k1.between(50,60)]

Out[181]: match_no
15 50
34 58
44 56
57 57
71 51
73 58
80 57
85 56
103 51
122 52
129 54
131 54
137 55
141 58
144 57
182 50
197 51
198 53
209 58
Name: runs, dtype: int64

In [182]: k1[k1.between(50,60)].size

Out[182]: 19

In [183]: # clip
sub.clip(100,200)

Out[183]: 0 100
1 100
2 100
3 100
4 100
...
360 200
361 200
362 155
363 144
364 172
Name: Subscribers gained, Length: 365, dtype: int64

In [186]: *# drop duplicates ##### drop_duplicates(): Returns a Series with duplicates removed*

```
dele = pd.Series([1,2,33,3,3,3,1,23,33,22,33,11])
dele
```

Out[186]:

0	1
1	2
2	33
3	3
4	3
5	3
6	1
7	23
8	33
9	22
10	33
11	11

dtype: int64

In [188]: `dele.drop_duplicates()`

Out[188]:

0	1
1	2
2	33
3	3
7	23
9	22
11	11

dtype: int64

In [189]: `dele.drop_duplicates(keep='last')`

Out[189]:

1	2
5	3
6	1
7	23
9	22
10	33
11	11

dtype: int64

In [190]: `movies.drop_duplicates()`

Out[190]: movie

Uri: The Surgical Strike	Vicky Kaushal
Battalion 609	Vicky Ahuja
The Accidental Prime Minister (film)	Anupam Kher
Why Cheat India	Emraan Hashmi
Evening Shadows	Mona Ambegaonkar
...	
Rules: Pyaar Ka Superhit Formula	Tanuja
Right Here Right Now (film)	Ankit
Talaash: The Hunt Begins...	Rakhee Gulzar
The Pink Mirror	Edwin Fernandes
Hum Tumhare Hain Sanam	Jack
Name: lead, Length: 567, dtype: object	

In [191]: `dele.duplicated().sum()`

Out[191]: 5

In [193]: `kl.duplicated().sum()`

Out[193]: 137

In [194]: `dele.count()`

Out[194]: 12

isin(values): Returns a boolean Series indicating whether each element in the Series is in the provided values

In [198]: `# isnull`

`kl.isnull().sum()`

Out[198]: 0

In [199]: `dele.isnull().sum()`

Out[199]: 0

In [200]: # dropna

```
dele.dropna()
```

Out[200]: 0 1

1 2

2 33

3 3

4 3

5 3

6 1

7 23

8 33

9 22

10 33

11 11

dtype: int64

In [202]: # fillna

```
dele.fillna(0)
```

```
dele.fillna(dele.mean())
```

Out[202]: 0 1

1 2

2 33

3 3

4 3

5 3

6 1

7 23

8 33

9 22

10 33

11 11

dtype: int64

In [205]: # isin

```
kl
```

Out[205]: match_no

1 1

2 23

3 13

4 12

5 1

..

211 0

212 20

213 73

214 25

215 7

Name: runs, Length: 215, dtype: int64

```
In [207]: k1[(k1==49) | (k1==99)]
```

```
Out[207]: match_no
82    99
86    49
Name: runs, dtype: int64
```

```
In [209]: k1[k1.isin([49,99])]
```

```
Out[209]: match_no
82    99
86    49
Name: runs, dtype: int64
```

```
In [210]: # apply
```

```
movies
```

```
Out[210]: movie
Uri: The Surgical Strike           Vicky Kaushal
Battalion 609                      Vicky Ahuja
The Accidental Prime Minister (film) Anupam Kher
Why Cheat India                     Emraan Hashmi
Evening Shadows                     Mona Ambegaonkar
                                         ...
                                         Jack
Hum Tumhare Hain Sanam            Amitabh Bachchan
Aankhen (2002 film)               Vivek Oberoi
Saathiya (film)                  Ajay Devgn
Company (film)                   Akshay Kumar
Awara Paagal Deewana
Name: lead, Length: 1500, dtype: object
```

```
In [212]: movies.apply(lambda x:x.split()) # split name in to two using Lambda function
```

```
Out[212]: movie
Uri: The Surgical Strike           [Vicky, Kaushal]
Battalion 609                      [Vicky, Ahuja]
The Accidental Prime Minister (film) [Anupam, Kher]
Why Cheat India                     [Emraan, Hashmi]
Evening Shadows                     [Mona, Ambegaonkar]
                                         ...
                                         [Jack]
Hum Tumhare Hain Sanam            [Amitabh, Bachchan]
Aankhen (2002 film)               [Vivek, Oberoi]
Saathiya (film)                  [Ajay, Devgn]
Company (film)                   [Akshay, Kumar]
Awara Paagal Deewana
Name: lead, Length: 1500, dtype: object
```

In [213]: `movies.apply(lambda x:x.split()[0]) # select first word`

Out[213]: movie

Uri: The Surgical Strike	Vicky
Battalion 609	Vicky
The Accidental Prime Minister (film)	Anupam
Why Cheat India	Emraan
Evening Shadows	Mona
...	
Hum Tumhare Hain Sanam	Jack
Aankhen (2002 film)	Amitabh
Saathiya (film)	Vivek
Company (film)	Ajay
Awara Paagal Deewana	Akshay

Name: lead, Length: 1500, dtype: object

In [214]: `movies.apply(lambda x:x.split()[0].upper()) # Upper case`

Out[214]: movie

Uri: The Surgical Strike	VICKY
Battalion 609	VICKY
The Accidental Prime Minister (film)	ANUPAM
Why Cheat India	EMRAAN
Evening Shadows	MONA
...	
Hum Tumhare Hain Sanam	JACK
Aankhen (2002 film)	AMITABH
Saathiya (film)	VIVEK
Company (film)	AJAY
Awara Paagal Deewana	AKSHAY

Name: lead, Length: 1500, dtype: object

In [215]: `sub`

Out[215]:

0	48
1	57
2	40
3	43
4	44
...	
360	231
361	226
362	155
363	144
364	172

Name: Subscribers gained, Length: 365, dtype: int64

In [216]: `sub.mean()`

Out[216]: 135.64383561643837

```
In [217]: sub.apply(lambda x: 'good day' if x > sub.mean() else 'bad day')
```

```
Out[217]: 0      bad day
1      bad day
2      bad day
3      bad day
4      bad day
...
360    good day
361    good day
362    good day
363    good day
364    good day
Name: Subscribers gained, Length: 365, dtype: object
```

```
In [229]: # Copy
```

```
k1
```

```
Out[229]: match_no
1      1
2      23
3      13
4      12
5      1
..
211    0
212    20
213    73
214    25
215    7
Name: runs, Length: 215, dtype: int64
```

```
In [230]: new = k1.head()
```

```
In [231]: new[1]=100
```

```
In [232]: new
```

```
Out[232]: match_no
1      100
2      23
3      13
4      12
5      1
Name: runs, dtype: int64
```

In [233]: k1

Out[233]:

```
match_no
1      100
2      23
3      13
4      12
5      1
...
211     0
212    20
213    73
214    25
215     7
Name: runs, Length: 215, dtype: int64
```

In [240]: new = k1.head(5).copy()

In [241]: new[1]=20

In [242]: new

Out[242]:

```
match_no
1      20
2      23
3      13
4      12
5      1
Name: runs, dtype: int64
```

In [250]: k1

Out[250]:

```
match_no
1      100
2      23
3      13
4      12
5      1
...
211     0
212    20
213    73
214    25
215     7
Name: runs, Length: 215, dtype: int64
```

In []:

In []:


```
In [1]: import numpy as np
import pandas as pd
```

Creating DataFrame

```
In [2]: # Using the Lists
student_data = [
    [100,80,10],
    [90,70,7],
    [120,100,14],
    [80,50,2]
]

pd.DataFrame(student_data,columns=['iq','marks','package'])
```

```
Out[2]:
   iq  marks  package
0  100      80       10
1    90      70        7
2   120     100       14
3    80      50        2
```

```
In [3]: # using dicts

student_dict = {
    'name':['peter','saint','noeum','parle','samme','dave'],
    'iq':[100,90,120,80,13,90],
    'marks':[80,70,100,50,11,80],
    'package':[10,7,14,2,15,100]
}
students=pd.DataFrame(student_dict)
students
```

```
Out[3]:
   name    iq  marks  package
0  peter   100      80       10
1   saint    90      70        7
2   noeum   120     100       14
3   parle    80      50        2
4   samme    13      11       15
5    dave    90      80      100
```

```
In [4]: students.set_index('name',inplace=True)
students
```

```
Out[4]:
          iq  marks  package
name
peter    100      80       10
saint     90      70        7
noeum    120     100       14
parle     80      50        2
samme     13      11       15
dave     90      80      100
```

```
In [5]: # Read csv
movies = pd.read_csv("movies.csv")
movies.head()
```

Out[5]:

	title_x	imdb_id	poster_path	wiki_link	title_y	original_title	is_adult	year_
0	Uri: The Surgical Strike	tt8291224	https://upload.wikimedia.org/wikipedia/en/thumb/.../https://en.wikipedia.org/wiki/Uri:_The_Surgical_...	https://en.wikipedia.org/wiki/Uri:_The_Surgical_...	Uri: The Surgical Strike	Uri: The Surgical Strike	0	
1	Battalion 609	tt9472208		Nan	https://en.wikipedia.org/wiki/Battalion_609	Battalion 609	Battalion 609	0
2	The Accidental Prime Minister (film)	tt6986710	https://upload.wikimedia.org/wikipedia/en/thumb/.../https://en.wikipedia.org/wiki/The_Accidental_P...	https://en.wikipedia.org/wiki/The_Accidental_P...	The Accidental Prime Minister	The Accidental Prime Minister	0	
3	Why Cheat India	tt8108208	https://upload.wikimedia.org/wikipedia/en/thumb/.../https://en.wikipedia.org/wiki/Why_Cheat_India	https://en.wikipedia.org/wiki/Why_Cheat_India	Why Cheat India	Why Cheat India	0	
4	Evening Shadows	tt6028796		Nan	https://en.wikipedia.org/wiki/Evening_Shadows	Evening Shadows	Evening Shadows	0

```
In [6]: ipl = pd.read_csv("ipl-matches.csv")
ipl.head()
```

Out[6]:

	ID	City	Date	Season	MatchNumber	Team1	Team2	Venue	TossWinner	TossDecision	SuperOver	WinningTeam	W
0	1312200	Ahmedabad	2022-05-29	2022	Final	Rajasthan Royals	Gujarat Titans	Narendra Modi Stadium, Ahmedabad	Rajasthan Royals	bat	N	Gujarat Titans	Wi
1	1312199	Ahmedabad	2022-05-27	2022	Qualifier 2	Royal Challengers Bangalore	Rajasthan Royals	Narendra Modi Stadium, Ahmedabad	Rajasthan Royals	field	N	Rajasthan Royals	Wi
2	1312198	Kolkata	2022-05-25	2022	Eliminator	Royal Challengers Bangalore	Lucknow Super Giants	Eden Gardens, Kolkata	Lucknow Super Giants	field	N	Royal Challengers Bangalore	Wi
3	1312197	Kolkata	2022-05-24	2022	Qualifier 1	Rajasthan Royals	Gujarat Titans	Eden Gardens, Kolkata	Gujarat Titans	field	N	Gujarat Titans	Wi

DataFrame Attributes and Methods

```
In [7]: # shape
movies.shape
```

Out[7]: (1629, 18)

```
In [8]: ipl.shape
```

Out[8]: (950, 20)

```
In [9]: # dtype
movies.dtypes
```

```
Out[9]: title_x          object
imdb_id           object
poster_path        object
wiki_link          object
title_y            object
original_title     object
is_adult           int64
year_of_release    int64
runtime             object
genres              object
imdb_rating         float64
imdb_votes          int64
story               object
summary              object
tagline              object
actors              object
wins_nominations   object
release_date        object
dtype: object
```

```
In [10]: ipl.dtypes
```

```
Out[10]: ID            int64
City           object
Date            object
Season          object
MatchNumber     object
Team1           object
Team2           object
Venue            object
TossWinner      object
TossDecision    object
SuperOver       object
WinningTeam     object
WonBy            object
Margin           float64
method           object
Player_of_Match object
Team1Players    object
Team2Players    object
Umpire1          object
Umpire2          object
dtype: object
```

```
In [11]: # index
movies.index
```

```
Out[11]: RangeIndex(start=0, stop=1629, step=1)
```

```
In [12]: ipl.index
```

```
Out[12]: RangeIndex(start=0, stop=950, step=1)
```

```
In [13]: # Columns
movies.columns
```

```
Out[13]: Index(['title_x', 'imdb_id', 'poster_path', 'wiki_link', 'title_y',
       'original_title', 'is_adult', 'year_of_release', 'runtime', 'genres',
       'imdb_rating', 'imdb_votes', 'story', 'summary', 'tagline', 'actors',
       'wins_nominations', 'release_date'],
      dtype='object')
```

```
In [14]: ipl.columns
```

```
Out[14]: Index(['ID', 'City', 'Date', 'Season', 'MatchNumber', 'Team1', 'Team2',
       'Venue', 'TossWinner', 'TossDecision', 'SuperOver', 'WinningTeam',
       'WonBy', 'Margin', 'method', 'Player_of_Match', 'Team1Players',
       'Team2Players', 'Umpire1', 'Umpire2'],
      dtype='object')
```

```
In [15]: # Values
students.values
```

```
Out[15]: array([[100,  80,  10],
   [ 90,  70,   7],
   [120, 100,  14],
   [ 80,  50,   2],
   [ 13,  11,  15],
   [ 90,  80, 100]], dtype=int64)
```

```
In [16]: ipl.values
```

```
Out[16]: array([[1312200, 'Ahmedabad', '2022-05-29', ...,
   ['WP Saha', 'Shubman Gill', 'MS Wade', 'HH Pandya', 'DA Miller', 'R Tewatia', 'Rashid Khan', 'R Sai Kishore',
   'LH Ferguson', 'Yash Dayal', 'Mohammed Shami'],
   'CB Gaffaney', 'Nitin Menon'],
   [1312199, 'Ahmedabad', '2022-05-27', ...,
   ['YBK Jaiswal', 'JC Buttler', 'SV Samson', 'D Padikkal', 'SO Hetmyer', 'R Parag', 'R Ashwin', 'TA Boult', 'YS C
hahal', 'M Prasidh Krishna', 'OC McCoy'],
   'CB Gaffaney', 'Nitin Menon'],
   [1312198, 'Kolkata', '2022-05-25', ...,
   ['Q de Kock', 'KL Rahul', 'M Vohra', 'DJ Hooda', 'MP Stoinis', 'E Lewis', 'KH Pandya', 'PVD Chameera', 'Mohsin
Khan', 'Avesh Khan', 'Ravi Bishnoi'],
   'J Madanagopal', 'MA Gough'],
   ...,
   [335984, 'Delhi', '2008-04-19', ...,
   ['T Kohli', 'YK Pathan', 'SR Watson', 'M Kaif', 'DS Lehmann', 'RA Jadeja', 'M Rawat', 'D Salunkhe', 'SK Warne',
   'SK Trivedi', 'MM Patel'],
   'Aleem Dar', 'GA Pratapkumar'],
   [335983, 'Chandigarh', '2008-04-19', ...,
   ['PA Patel', 'ML Hayden', 'MEK Hussey', 'MS Dhoni', 'SK Raina', 'JDP Oram', 'S Badrinath', 'Joginder Sharma',
   'P Amarnath', 'MS Gony', 'M Muralitharan'],
   'MR Benson', 'SL Shastri'],
   [335982, 'Bangalore', '2008-04-18', ...,
   ['SC Ganguly', 'BB McCullum', 'RT Ponting', 'DJ Hussey', 'Mohammad Hafeez', 'LR Shukla', 'WP Saha', 'AB Agarka
r', 'AB Dinda', 'M Kartik', 'I Sharma'],
   'Asad Rauf', 'RE Koertzen']], dtype=object)
```

```
In [17]: # Sample -> to reduce bias
ipl.sample(2)
```

ID	City	Date	Season	MatchNumber	Team1	Team2	Venue	TossWinner	TossDecision	SuperOver	WinningTeam	WonBy	Ma
226	1178402	Mumbai	2019-04-13	2019	27	Mumbai Indians	Rajasthan Royals	Wankhede Stadium	Rajasthan Royals	field	N	Rajasthan Royals	Wickets
63	1304057	Mumbai	2022-04-03	2022	11	Punjab Kings	Chennai Super Kings	Brabourne Stadium, Mumbai	Chennai Super Kings	field	N	Punjab Kings	Runs

In [18]: # info
movies.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1629 entries, 0 to 1628
Data columns (total 18 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   title_x          1629 non-null   object  
 1   imdb_id          1629 non-null   object  
 2   poster_path      1526 non-null   object  
 3   wiki_link        1629 non-null   object  
 4   title_y          1629 non-null   object  
 5   original_title   1629 non-null   object  
 6   is_adult         1629 non-null   int64  
 7   year_of_release  1629 non-null   int64  
 8   runtime          1629 non-null   object  
 9   genres           1629 non-null   object  
 10  imdb_rating     1629 non-null   float64 
 11  imdb_votes       1629 non-null   int64  
 12  story            1609 non-null   object  
 13  summary          1629 non-null   object  
 14  tagline          557 non-null    object  
 15  actors           1624 non-null   object  
 16  wins_nominations 707 non-null   object  
 17  release_date     1522 non-null   object  
dtypes: float64(1), int64(3), object(14)
memory usage: 229.2+ KB
```

In [19]: ipl.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 950 entries, 0 to 949
Data columns (total 20 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   ID               950 non-null   int64  
 1   City             899 non-null   object  
 2   Date             950 non-null   object  
 3   Season           950 non-null   object  
 4   MatchNumber      950 non-null   object  
 5   Team1            950 non-null   object  
 6   Team2            950 non-null   object  
 7   Venue            950 non-null   object  
 8   TossWinner       950 non-null   object  
 9   TossDecision     950 non-null   object  
 10  SuperOver        946 non-null   object  
 11  WinningTeam      946 non-null   object  
 12  WonBy            950 non-null   object  
 13  Margin           932 non-null   float64 
 14  method           19 non-null    object  
 15  Player_of_Match 946 non-null   object  
 16  Team1Players    950 non-null   object  
 17  Team2Players    950 non-null   object  
 18  Umpire1          950 non-null   object  
 19  Umpire2          950 non-null   object  
dtypes: float64(1), int64(1), object(18)
memory usage: 148.6+ KB
```

In [20]: # describe
movies.describe()

Out[20]:

	is_adult	year_of_release	imdb_rating	imdb_votes
count	1629.0	1629.000000	1629.000000	1629.000000
mean	0.0	2010.263966	5.557459	5384.263352
std	0.0	5.381542	1.567609	14552.103231
min	0.0	2001.000000	0.000000	0.000000
25%	0.0	2005.000000	4.400000	233.000000
50%	0.0	2011.000000	5.600000	1000.000000
75%	0.0	2015.000000	6.800000	4287.000000
max	0.0	2019.000000	9.400000	310481.000000

In [21]: `ipl.describe()`

Out[21]:

	ID	Margin
count	9.500000e+02	932.000000
mean	8.304852e+05	17.056867
std	3.375678e+05	21.633109
min	3.359820e+05	1.000000
25%	5.012612e+05	6.000000
50%	8.297380e+05	8.000000
75%	1.175372e+06	19.000000
max	1.312200e+06	146.000000

In [22]: `# isnull
movies.isnull()`

Out[22]:

	title_x	imdb_id	poster_path	wiki_link	title_y	original_title	is_adult	year_of_release	runtime	genres	imdb_rating	imdb_votes	story	summary
0	False	False	False	False	False	False	False	False	False	False	False	False	False	Fal
1	False	False	True	False	False	False	False	False	False	False	False	False	False	Fal
2	False	False	False	False	False	False	False	False	False	False	False	False	False	Fal
3	False	False	False	False	False	False	False	False	False	False	False	False	False	Fal
4	False	False	True	False	False	False	False	False	False	False	False	False	False	Fal
...
1624	False	False	False	False	False	False	False	False	False	False	False	False	False	Fal
1625	False	False	False	False	False	False	False	False	False	False	False	False	False	Fal
1626	False	False	True	False	False	False	False	False	False	False	False	False	False	Fal
1627	False	False	False	False	False	False	False	False	False	False	False	False	False	Fal
1628	False	False	False	False	False	False	False	False	False	False	False	False	False	Fal

1629 rows × 18 columns

In [23]: `movies.isnull().sum()`

Out[23]:

title_x	0
imdb_id	0
poster_path	103
wiki_link	0
title_y	0
original_title	0
is_adult	0
year_of_release	0
runtime	0
genres	0
imdb_rating	0
imdb_votes	0
story	20
summary	0
tagline	1072
actors	5
wins_nominations	922
release_date	107
dtype: int64	

In [24]: `# duplicated
movies.duplicated().sum()`

Out[24]: 0

```
In [25]: # rename
students
```

```
Out[25]:
      iq  marks  package
name
peter   100     80      10
saint    90     70       7
noeum   120    100      14
parle    80     50       2
samme   13      11      15
dave    90     80     100
```

```
In [26]: students.rename(columns={'marks': 'percent', 'package': 'lpa'}, inplace=True)
```

```
In [ ]: students.drop(columns='name', inplace=True)
```

Maths Method

```
In [28]: # sum -> Axis Argument
students.sum(axis=1)
```

```
Out[28]: name
peter    190
saint    167
noeum    234
parle    132
samme     39
dave     270
dtype: int64
```

```
In [29]: # mean
students.mean()
```

```
Out[29]: iq        82.166667
percent   65.166667
lpa       24.666667
dtype: float64
```

```
In [30]: students.min(axis=1)
```

```
Out[30]: name
peter    10
saint     7
noeum    14
parle     2
samme    11
dave     80
dtype: int64
```

```
In [31]: students.var()
```

```
Out[31]: iq        1332.166667
percent   968.166667
lpa       1384.666667
dtype: float64
```

Selecting cols from a DataFrame

```
In [32]: # single cols
movies['title_x']
```

```
Out[32]: 0           Uri: The Surgical Strike
1           Battalion 609
2       The Accidental Prime Minister (film)
3           Why Cheat India
4           Evening Shadows
...
1624      Tera Mera Saath Rahen
1625      Yeh Zindagi Ka Safar
1626      Sabse Bada Sukh
1627          Daaka
1628      Humsafar
Name: title_x, Length: 1629, dtype: object
```

```
In [33]: type(movies['title_x'])
```

```
Out[33]: pandas.core.series.Series
```

```
In [137]: # multiple columns
movies[['year_of_release', 'actors', 'title_x']].head(2)
```

```
In [35]: type(movies[['year_of_release', 'actors', 'title_x']].head(2))
```

```
Out[35]: pandas.core.frame.DataFrame
```

```
In [36]: ipl[['City', 'Team1', 'Team2']]
```

```
Out[36]:
```

	City	Team1	Team2
0	Ahmedabad	Rajasthan Royals	Gujarat Titans
1	Ahmedabad	Royal Challengers Bangalore	Rajasthan Royals
2	Kolkata	Royal Challengers Bangalore	Lucknow Super Giants
3	Kolkata	Rajasthan Royals	Gujarat Titans
4	Mumbai	Sunrisers Hyderabad	Punjab Kings
...
945	Kolkata	Kolkata Knight Riders	Deccan Chargers
946	Mumbai	Mumbai Indians	Royal Challengers Bangalore
947	Delhi	Delhi Daredevils	Rajasthan Royals
948	Chandigarh	Kings XI Punjab	Chennai Super Kings
949	Bangalore	Royal Challengers Bangalore	Kolkata Knight Riders

950 rows × 3 columns

```
In [37]: student_dict = {
    'name': ['peter', 'saint', 'noeum', 'parle', 'samme', 'dave'],
    'iq': [100, 90, 120, 80, 13, 90],
    'marks': [80, 70, 100, 50, 11, 80],
    'package': [10, 7, 14, 2, 15, 100]
}
students = pd.DataFrame(student_dict)
students.set_index('name', inplace=True)
```

```
In [38]: students
```

```
Out[38]:
```

	iq	marks	package
name			
peter	100	80	10
saint	90	70	7
noeum	120	100	14
parle	80	50	2
samme	13	11	15
dave	90	80	100

Selecting rows from a DataFrame

- `iloc` - searches using index positions
- `loc` - searches using index labels

```
In [39]: # single_row
movies.iloc[1]
```

```
Out[39]: title_x          Battalion 609
imdb_id           tt9472208
poster_path        NaN
wiki_link   https://en.wikipedia.org/wiki/Battalion_609 (https://en.wikipedia.org/wiki/Battalion_609)
title_y          Battalion 609
original_title    Battalion 609
is_adult            0
year_of_release      2019
runtime              131
genres                War
imdb_rating          4.1
imdb_votes             73
story     The story revolves around a cricket match betw...
summary   The story of Battalion 609 revolves around a c ...
tagline        NaN
actors    Vicky Ahuja|Shoaib Ibrahim|Shrikant Kamat|Elen...
wins_nominations      NaN
release_date    11 January 2019 (India)
Name: 1, dtype: object
```

```
In [40]: # Multiple rows
movies.iloc[5:10]
```

	title_x	imdb_id	poster_path	wiki_link	title_y	original_title	is_adult	y
5	Soni (film)	tt6078866	https://upload.wikimedia.org/wikipedia/en/thum...	https://en.wikipedia.org/wiki/Soni_(film)	Soni	Soni	0	
6	Fraud Saiyaan	tt5013008	https://upload.wikimedia.org/wikipedia/en/thum...	https://en.wikipedia.org/wiki/Fraud_Saiyaan	Fraud Saiyaan	Fraud Saiyyan	0	
7	Bombairiya	tt4971258	https://upload.wikimedia.org/wikipedia/en/thum...	https://en.wikipedia.org/wiki/Bombairiya	Bombairiya	Bombairiya	0	
8	Manikarnika: The Queen of Jhansi	tt6903440	https://upload.wikimedia.org/wikipedia/en/thum...	https://en.wikipedia.org/wiki/Manikarnika:_The...	Manikarnika: The Queen of Jhansi	Manikarnika: The Queen of Jhansi	0	
9	Thackeray (film)	tt7777196	https://upload.wikimedia.org/wikipedia/en/thum...	https://en.wikipedia.org/wiki/Thackeray_(film)	Thackeray	Thackeray	0	

In [41]: `movies.iloc[5:12:2]`

Out[41]:

	title_x	imdb_id	poster_path	wiki_link	title_y	original_title	is_adult	year_
5	Soni (film)	tt6078866	https://upload.wikimedia.org/wikipedia/en/thumb.../250px-Soni_(film)_Poster.jpg	https://en.wikipedia.org/wiki/Soni_(film)	Soni	Soni	Soni	0
7	Bombairiya	tt4971258	https://upload.wikimedia.org/wikipedia/en/thumb.../250px-Bombairiya_Poster.jpg	https://en.wikipedia.org/wiki/Bombairiya	Bombairiya	Bombairiya	Bombairiya	0
9	Thackeray (film)	tt7777196	https://upload.wikimedia.org/wikipedia/en/thumb.../250px-Thackeray_(film)_Poster.jpg	https://en.wikipedia.org/wiki/Thackeray_(film)	Thackeray	Thackeray	Thackeray	0
11	Gully Boy	tt2395469	https://upload.wikimedia.org/wikipedia/en/thumb.../250px-Gully_Boy_Poster.jpg	https://en.wikipedia.org/wiki/Gully_Boy	Gully Boy	Gully Boy	Gully Boy	0

In [42]: `# Fancy indexing
ipl.iloc[[0, 4, 5]]`

Out[42]:

ID	City	Date	Season	MatchNumber	Team1	Team2	Venue	TossWinner	TossDecision	SuperOver	WinningTeam	WonBy
0	Ahmedabad	2022-05-29	2022	Final	Rajasthan Royals	Gujarat Titans	Narendra Modi Stadium, Ahmedabad	Rajasthan Royals	bat	N	Gujarat Titans	Wickets
4	Mumbai	2022-05-22	2022	70	Sunrisers Hyderabad	Punjab Kings	Wankhede Stadium, Mumbai	Sunrisers Hyderabad	bat	N	Punjab Kings	Wickets
5	Mumbai	2022-05-21	2022	69	Delhi Capitals	Mumbai Indians	Wankhede Stadium, Mumbai	Mumbai Indians	field	N	Mumbai Indians	Wickets

In [43]: `# loc (Location)
students`

Out[43]:

	name	iq	marks	package
peter	100	80	10	
saint	90	70	7	
noeum	120	100	14	
parle	80	50	2	
samme	13	11	15	
dave	90	80	100	

In [44]: `students.loc['parle']`

Out[44]:

```
iq      80
marks   50
package 2
Name: parle, dtype: int64
```

```
In [45]: students.loc['saint':'samme':2]
```

```
Out[45]:
      iq  marks  package
name
saint  90     70      7
parle  80     50      2
```

```
In [46]: # Fancy indexing
students.loc[['saint', 'dave']]
```

```
Out[46]:
      iq  marks  package
name
saint  90     70      7
dave   90     80     100
```

```
In [47]: students.iloc[[0,4,3]]
```

```
Out[47]:
      iq  marks  package
name
peter  100    80     10
samme  13     11     15
parle  80     50      2
```

Selecting both rows and cols

```
In [48]: movies.iloc[0:3,0:3]
```

```
Out[48]:
      title_x  imdb_id  poster_path
0       Uri: The Surgical Strike  tt8291224  https://upload.wikimedia.org/wikipedia/en/thum...
1           Battalion 609  tt9472208          NaN
2  The Accidental Prime Minister (film)  tt6986710  https://upload.wikimedia.org/wikipedia/en/thum...
```

```
In [49]: movies.loc[0:2,'title_x':'poster_path']
```

```
Out[49]:
      title_x  imdb_id  poster_path
0       Uri: The Surgical Strike  tt8291224  https://upload.wikimedia.org/wikipedia/en/thum...
1           Battalion 609  tt9472208          NaN
2  The Accidental Prime Minister (film)  tt6986710  https://upload.wikimedia.org/wikipedia/en/thum...
```

Filtering a DataFrame

```
In [50]: ipl.head(2)
```

```
Out[50]:
   ID   City  Date  Season MatchNumber  Team1  Team2  Venue  TossWinner  TossDecision  SuperOver  WinningTeam  WonBy
0  1312200  Ahmedabad  2022-05-29  2022        Final  Rajasthan Royals  Gujarat Titans  Narendra Modi Stadium, Ahmedabad  Rajasthan Royals  bat  N  Gujarat Titans  Wicke...
1  1312199  Ahmedabad  2022-05-27  2022  Qualifier 2  Royal Challengers Bangalore  Rajasthan Royals  Narendra Modi Stadium, Ahmedabad  Rajasthan Royals  field  N  Rajasthan Royals  Wicke...
```

In [51]: # find all the final winners

```
mask=ipl['MatchNumber'] == 'Final'
new_df= ipl[mask]
new_df[['Season', 'WinningTeam']]
```

Out[51]:

	Season	WinningTeam
0	2022	Gujarat Titans
74	2021	Chennai Super Kings
134	2020/21	Mumbai Indians
194	2019	Mumbai Indians
254	2018	Chennai Super Kings
314	2017	Mumbai Indians
373	2016	Sunrisers Hyderabad
433	2015	Mumbai Indians
492	2014	Kolkata Knight Riders
552	2013	Mumbai Indians
628	2012	Kolkata Knight Riders
702	2011	Chennai Super Kings
775	2009/10	Chennai Super Kings
835	2009	Deccan Chargers
892	2007/08	Rajasthan Royals

In [52]: # In one Line

```
ipl[ipl['MatchNumber']=='Final'][['Season', 'WinningTeam']]
```

Out[52]:

	Season	WinningTeam
0	2022	Gujarat Titans
74	2021	Chennai Super Kings
134	2020/21	Mumbai Indians
194	2019	Mumbai Indians
254	2018	Chennai Super Kings
314	2017	Mumbai Indians
373	2016	Sunrisers Hyderabad
433	2015	Mumbai Indians
492	2014	Kolkata Knight Riders
552	2013	Mumbai Indians
628	2012	Kolkata Knight Riders
702	2011	Chennai Super Kings
775	2009/10	Chennai Super Kings
835	2009	Deccan Chargers
892	2007/08	Rajasthan Royals

In [53]: # how many super over finishes have occurred

```
ipl.head(2)
```

Out[53]:

ID	City	Date	Season	MatchNumber	Team1	Team2	Venue	TossWinner	TossDecision	SuperOver	WinningTeam	WonB
0	Ahmedabad	2022-05-29	2022	Final	Rajasthan Royals	Gujarat Titans	Narendra Modi Stadium, Ahmedabad	Rajasthan Royals	bat	N	Gujarat Titans	Wicke
1	Ahmedabad	2022-05-27	2022	Qualifier 2	Royal Challengers Bangalore	Rajasthan Royals	Narendra Modi Stadium, Ahmedabad	Rajasthan Royals	field	N	Rajasthan Royals	Wicke

```
In [54]: ipl[ipl['SuperOver']=='Y'].shape[0]
```

```
Out[54]: 14
```

```
In [55]: # how many matches has csk won in kolkata
ipl.sample(2)
```

```
Out[55]:
```

	ID	City	Date	Season	MatchNumber	Team1	Team2	Venue	TossWinner	TossDecision	SuperOver	WinningTeam	Wk
364	1082599	Pune	2017-04-11	2017	9	Rising Pune Supergiant	Delhi Daredevils	Maharashtra Cricket Association Stadium	Rising Pune Supergiant	field	N	Delhi Daredevils	1
376	981013	Bangalore	2016-05-24	2016	Qualifier 1	Gujarat Lions	Royal Challengers Bangalore	M Chinnaswamy Stadium	Royal Challengers Bangalore	field	N	Royal Challengers Bangalore	Win



```
In [56]:
```

```
ipl[(ipl['City'] == 'Kolkata') & (ipl['WinningTeam'] == 'Chennai Super Kings')].shape[0]
```

```
Out[56]: 5
```

```
In [57]: # toss winner is match winner in percentage
ipl.sample(2)
```

```
Out[57]:
```

	ID	City	Date	Season	MatchNumber	Team1	Team2	Venue	TossWinner	TossDecision	SuperOver	WinningTeam	WonBy	Ma
168	1216529	Abu Dhabi	2020-10-11	2020/21	27	Delhi Capitals	Mumbai Indians	Sheikh Zayed Stadium	Delhi Capitals	bat	N	Mumbai Indians	Wickets	
127	1254064	Mumbai	2021-04-15	2021	7	Delhi Capitals	Rajasthan Royals	Wankhede Stadium, Mumbai	Rajasthan Royals	field	N	Rajasthan Royals	Wickets	



```
In [58]: (ipl[(ipl['TossWinner']== ipl['WinningTeam'])].shape[0]/ipl.shape[0])*100
```

```
Out[58]: 51.473684210526315
```

```
In [59]: # movies with rating higher than 8 and votes>10000
movies.sample(2)
```

```
Out[59]:
```

	title_x	imdb_id	poster_path	wiki_link	title_y	original_title	is_adult	year_of
24	Junglee (2019 film)	tt7463730	https://upload.wikimedia.org/wikipedia/en/e/e2... https://en.wikipedia.org/wiki/Junglee_(2019_film)	Junglee	Junglee	Junglee	0	
390	Hey Bro	tt4512230	https://upload.wikimedia.org/wikipedia/en/thumb/0/0d/H... https://en.wikipedia.org/wiki/Hey_Bro	Hey Bro	Hey Bro	Hey Bro	0	



```
In [60]: movies[(movies['imdb_rating'] > 8) & (movies['imdb_votes'] > 10000)].shape[0]
```

```
Out[60]: 43
```

```
In [61]: # Action movies with rating higher than 7.5
#mask1=movies['genres'].str.split('/').apply(lambda x:'Action' in x)
mask1=movies['genres'].str.contains('Action')
mask2=movies['imdb_rating']>7.5
movies[mask1 & mask2]
```

Out[61]:

	title_x	imdb_id	poster_path	wiki_link	title_y	original_title
0	Uri: The Surgical Strike	tt8291224	https://upload.wikimedia.org/wikipedia/en/thumb/0/0e/Uri_The_Surgical_Strike.jpg/220px-Uri_The_Surgical_Strike.jpg	https://en.wikipedia.org/wiki/Uri:_The_Surgical_Strike	Uri: The Surgical Strike	Uri: The Surgical Strike
41	Family of Thakurganj	tt8897986	https://upload.wikimedia.org/wikipedia/en/9/99/Family_of_Thakurganj.jpg	https://en.wikipedia.org/wiki/Family_of_Thakurganj	Family of Thakurganj	Family of Thakurganj
84	Mukkabaaz	tt7180544	https://upload.wikimedia.org/wikipedia/en/thumb/2/2d/Mukkabaaz.jpg/220px-Mukkabaaz.jpg	https://en.wikipedia.org/wiki/Mukkabaaz	The Brawler	Mukkabaaz
106	Raazi	tt7098658	https://upload.wikimedia.org/wikipedia/en/thumb/2/2f/Raazi.jpg/220px-Raazi.jpg	https://en.wikipedia.org/wiki/Raazi	Raazi	Raazi

Adding new cols

```
In [62]: movies['country']='India'
movies.sample(2)
```

Out[62]:

	title_x	imdb_id	poster_path	wiki_link	title_y	original_title	is_adult	year_of_release
915	Wanted (2009 film)	tt1084972	https://upload.wikimedia.org/wikipedia/en/thumb/0/0f/Wanted_(2009_film).jpg/220px-Wanted_(2009_film).jpg	https://en.wikipedia.org/wiki/Wanted_(2009_film)	Wanted	Wanted	0	0
241	Shaadi Mein Zaroor Aana	tt7469726	https://upload.wikimedia.org/wikipedia/en/thumb/9/9a/Shaadi_Mein_Zaroor_Aana.jpg/220px-Shaadi_Mein_Zaroor_Aana.jpg	https://en.wikipedia.org/wiki/Shaadi_Mein_Zaroor_Aana	Shaadi Mein Zaroor Aana	Shaadi Mein Zaroor Aana	0	0

In [63]: movies.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1629 entries, 0 to 1628
Data columns (total 19 columns):
 #   Column           Non-Null Count  Dtype  
 --- 
 0   title_x          1629 non-null   object  
 1   imdb_id          1629 non-null   object  
 2   poster_path       1526 non-null   object  
 3   wiki_link         1629 non-null   object  
 4   title_y          1629 non-null   object  
 5   original_title    1629 non-null   object  
 6   is_adult          1629 non-null   int64  
 7   year_of_release   1629 non-null   int64  
 8   runtime           1629 non-null   object  
 9   genres             1629 non-null   object  
 10  imdb_rating       1629 non-null   float64 
 11  imdb_votes        1629 non-null   int64  
 12  story              1609 non-null   object  
 13  summary            1629 non-null   object  
 14  tagline            557 non-null    object  
 15  actors             1624 non-null   object  
 16  wins_nominations  707 non-null   object  
 17  release_date       1522 non-null   object  
 18  country            1629 non-null   object  
dtypes: float64(1), int64(3), object(15)
memory usage: 241.9+ KB
```

```
In [138]: # From Existing ones
movies['actors'].str.split('|').apply(lambda x:x[0])
```

```
In [139]: # From Existing ones
movies['lead actor']= movies['actors'].str.split('|').apply(lambda x:x[0])
movies
```

Important DataFrame Functions

```
In [68]: ipl.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 950 entries, 0 to 949
Data columns (total 20 columns):
 #   Column            Non-Null Count  Dtype  
--- 
 0   ID                950 non-null    int64  
 1   City               899 non-null    object  
 2   Date               950 non-null    object  
 3   Season              950 non-null    object  
 4   MatchNumber         950 non-null    object  
 5   Team1              950 non-null    object  
 6   Team2              950 non-null    object  
 7   Venue               950 non-null    object  
 8   TossWinner          950 non-null    object  
 9   TossDecision        950 non-null    object  
 10  SuperOver           946 non-null    object  
 11  WinningTeam         946 non-null    object  
 12  WonBy               950 non-null    object  
 13  Margin              932 non-null    float64 
 14  method              19 non-null     object  
 15  Player_of_Match    946 non-null    object  
 16  Team1Players        950 non-null    object  
 17  Team2Players        950 non-null    object  
 18  Umpire1             950 non-null    object  
 19  Umpire2             950 non-null    object  
dtypes: float64(1), int64(1), object(18)
memory usage: 148.6+ KB
```

```
In [69]: ipl['ID']=ipl['ID'].astype('int32')
```

```
In [70]: ipl.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 950 entries, 0 to 949
Data columns (total 20 columns):
 #   Column            Non-Null Count  Dtype  
--- 
 0   ID                950 non-null    int32  
 1   City               899 non-null    object  
 2   Date               950 non-null    object  
 3   Season              950 non-null    object  
 4   MatchNumber         950 non-null    object  
 5   Team1              950 non-null    object  
 6   Team2              950 non-null    object  
 7   Venue               950 non-null    object  
 8   TossWinner          950 non-null    object  
 9   TossDecision        950 non-null    object  
 10  SuperOver           946 non-null    object  
 11  WinningTeam         946 non-null    object  
 12  WonBy               950 non-null    object  
 13  Margin              932 non-null    float64 
 14  method              19 non-null     object  
 15  Player_of_Match    946 non-null    object  
 16  Team1Players        950 non-null    object  
 17  Team2Players        950 non-null    object  
 18  Umpire1             950 non-null    object  
 19  Umpire2             950 non-null    object  
dtypes: float64(1), int32(1), object(18)
memory usage: 144.9+ KB
```

```
In [71]: ipl['Season'] = ipl['Season'].astype('category')
```

In [72]: `ipl.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 950 entries, 0 to 949
Data columns (total 20 columns):
 #   Column      Non-Null Count  Dtype  
 --- 
 0   ID          950 non-null    int32  
 1   City         899 non-null    object  
 2   Date         950 non-null    object  
 3   Season       950 non-null    category
 4   MatchNumber  950 non-null    object  
 5   Team1        950 non-null    object  
 6   Team2        950 non-null    object  
 7   Venue         950 non-null    object  
 8   TossWinner   950 non-null    object  
 9   TossDecision 950 non-null    object  
 10  SuperOver   946 non-null    object  
 11  WinningTeam 946 non-null    object  
 12  WonBy        950 non-null    object  
 13  Margin        932 non-null    float64 
 14  method        19 non-null    object  
 15  Player_of_Match 946 non-null  object  
 16  Team1Players 950 non-null    object  
 17  Team2Players 950 non-null    object  
 18  Umpire1      950 non-null    object  
 19  Umpire2      950 non-null    object  
dtypes: category(1), float64(1), int32(1), object(17)
memory usage: 139.0+ KB
```

In [73]: `ipl['Team1'] = ipl['Team1'].astype('category')`
`ipl['Team2'] = ipl['Team2'].astype('category')`

In [74]: `ipl.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 950 entries, 0 to 949
Data columns (total 20 columns):
 #   Column      Non-Null Count  Dtype  
 --- 
 0   ID          950 non-null    int32  
 1   City         899 non-null    object  
 2   Date         950 non-null    object  
 3   Season       950 non-null    category
 4   MatchNumber  950 non-null    object  
 5   Team1        950 non-null    category
 6   Team2        950 non-null    category
 7   Venue         950 non-null    object  
 8   TossWinner   950 non-null    object  
 9   TossDecision 950 non-null    object  
 10  SuperOver   946 non-null    object  
 11  WinningTeam 946 non-null    object  
 12  WonBy        950 non-null    object  
 13  Margin        932 non-null    float64 
 14  method        19 non-null    object  
 15  Player_of_Match 946 non-null  object  
 16  Team1Players 950 non-null    object  
 17  Team2Players 950 non-null    object  
 18  Umpire1      950 non-null    object  
 19  Umpire2      950 non-null    object  
dtypes: category(3), float64(1), int32(1), object(15)
memory usage: 127.4+ KB
```

More Important Functions

Value Counts

In [143]: # value_counts(series and dataframe)

```
marks = pd.DataFrame([
    [100,80,10],
    [90,70,7],
    [120,100,14],
    [80,70,14],
    [80,70,14]
],columns=['iq','marks','package'])

marks
```

Out[143]:

	iq	marks	package
0	100	80	10
1	90	70	7
2	120	100	14
3	80	70	14
4	80	70	14

In [76]: marks.value_counts()

```
Out[76]: iq    marks    package
80     70      14        2
90     70      7         1
100    80      10        1
120    100     14        1
dtype: int64
```

In [77]: # find which player has won most potm -> in finals and qualifiers
ipl.sample(2)

Out[77]:

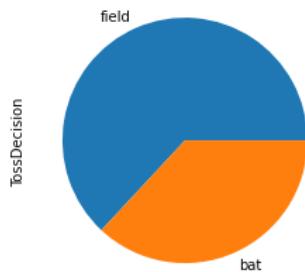
ID	City	Date	Season	MatchNumber	Team1	Team2	Venue	TossWinner	TossDecision	SuperOver	WinningTeam	WonBy	
304	1136570	Kolkata	2018-04-14	2018	10	Kolkata Knight Riders	Sunrisers Hyderabad	Eden Gardens	Sunrisers Hyderabad	field	N	Sunrisers Hyderabad	Wickets
561	598028	Dharamsala	2013-05-16	2013	67	Kings XI Punjab	Delhi Daredevils	Himachal Pradesh Cricket Association Stadium	Delhi Daredevils	field	N	Kings XI Punjab	Runs

```
In [78]: ipl[~ipl['MatchNumber'].str.isdigit()][['Player_of_Match']].value_counts() # To reverse the contet use tilt ~
```

```
Out[78]: KA Pollard      3  
F du Plessis      3  
SK Raina         3  
A Kumble         2  
MK Pandey        2  
YK Pathan        2  
M Vijay          2  
JJ Bumrah        2  
AB de Villiers    2  
SR Watson        2  
HH Pandya        1  
Harbhajan Singh   1  
A Nehra          1  
V Sehwag          1  
UT Yadav          1  
MS Bisla          1  
BJ Hodge          1  
MEK Hussey        1  
MS Dhoni          1  
CH Gayle          1  
MM Patel          1  
DE Bollinger      1  
AC Gilchrist      1  
RG Sharma          1  
DA Warner          1  
MC Henriques      1  
JC Buttler         1  
RM Patidar         1  
DA Miller          1  
VR Iyer            1  
SP Narine          1  
RD Gaikwad         1  
TA Boult           1  
MP Stoinis         1  
KS Williamson      1  
RR Pant            1  
SA Yadav           1  
Rashid Khan        1  
AD Russell          1  
KH Pandya          1  
KV Sharma          1  
NM Coulter-Nile     1  
Washington Sundar    1  
BCJ Cutting         1  
M Ntini             1  
Name: Player_of_Match, dtype: int64
```

```
In [79]: # Toss decision plot  
ipl['TossDecision'].value_counts().plot(kind='pie')
```

```
Out[79]: <AxesSubplot:ylabel='TossDecision'>
```



```
In [80]: # No.of matches each team has played
(ipl['Team1'].value_counts() + ipl['Team2'].value_counts()).sort_values(ascending=False)
```

```
Out[80]: Mumbai Indians      231
Royal Challengers Bangalore 226
Kolkata Knight Riders       223
Chennai Super Kings        208
Rajasthan Royals           192
Kings XI Punjab            190
Delhi Daredevils           161
Sunrisers Hyderabad         152
Deccan Chargers             75
Delhi Capitals              63
Pune Warriors               46
Gujarat Lions               30
Punjab Kings                28
Gujarat Titans              16
Rising Pune Supergiant      16
Lucknow Super Giants        15
Kochi Tuskers Kerala        14
Rising Pune Supergiants     14
dtype: int64
```

Sort values

```
In [81]: x = pd.Series([12,14,1,56,89])
x
```

```
Out[81]: 0    12
1    14
2     1
3    56
4    89
dtype: int64
```

```
In [82]: x.sort_values(ascending=True)
```

```
Out[82]: 2    1
0    12
1    14
3    56
4    89
dtype: int64
```

```
In [83]: movies.sample(2)
```

		title_x	imdb_id	poster_path	wiki_link	title_y	original_title	is_adult	year_of_i
107		Hope Aur Hum	tt8324474	https://upload.wikimedia.org/wikipedia/en/thumb/...	https://en.wikipedia.org/wiki/Hope_Aur_Hum	Hope Aur Hum	Hope Aur Hum	0	
666		Tere Naal Love Ho Gaya	tt2130242	https://upload.wikimedia.org/wikipedia/en/thumb/...	https://en.wikipedia.org/wiki/Tere_Naal_Love_H...	Tere Naal Love Ho Gaya	Tere Naal Love Ho Gaya	0	

```
In [84]: movies.sort_values('title_x', ascending=False)
```

Out[84]:

	title_x	imdb_id	poster_path	wiki_link	title_y	original_title	is_adult
1623	Zubeidaa	tt0255713	https://upload.wikimedia.org/wikipedia/en/thumb/.../Zubeidaa.jpg	https://en.wikipedia.org/wiki/Zubeidaa	Zubeidaa	Zubeidaa	0
939	Zor Lagaa Ke...Haiya!	tt1479857	https://upload.wikimedia.org/wikipedia/en/thumb/.../Zor_Lagaa_Ke...Haiya!.jpg	https://en.wikipedia.org/wiki/Zor_Lagaa_Ke...Haiya!	Zor Lagaa Ke...Haiya!	Zor Lagaa Ke...Haiya!	0
756	Zokkomon	tt1605790	https://upload.wikimedia.org/wikipedia/en/thumb/.../Zokkomon.jpg	https://en.wikipedia.org/wiki/Zokkomon	Zokkomon	Zokkomon	0
670	Zindagi Tere Naam	tt2164702	https://upload.wikimedia.org/wikipedia/en/thumb/.../Zindagi_Tere_Naam.jpg	https://en.wikipedia.org/wiki/Zindagi_Tere_Naam	Zindagi Tere Naam	Zindagi Tere Naam	0
778	Zindagi Na Milegi Dobara	tt1562872	https://upload.wikimedia.org/wikipedia/en/thumb/.../Zindagi_Na_Milegi_Dobara.jpg	https://en.wikipedia.org/wiki/Zindagi_Na_Milegi_Dobara	Zindagi Na Milegi Dobara	Zindagi Na Milegi Dobara	0
...
1039	1971 (2007 film)	tt0983990	https://upload.wikimedia.org/wikipedia/en/thumb/.../1971_(2007_film).jpg	https://en.wikipedia.org/wiki/1971_(2007_film)	1971	1971	0
723	1920: The Evil Returns	tt2222550	https://upload.wikimedia.org/wikipedia/en/e/e7/.../1920_The_Evil_Returns.jpg	https://en.wikipedia.org/wiki/1920:_The_Evil_R...	1920: Evil Returns	1920: Evil Returns	0
287	1920: London	tt5638500	https://upload.wikimedia.org/wikipedia/en/thumb/.../1920_London.jpg	https://en.wikipedia.org/wiki/1920_London	1920 London	1920 London	0
1021	1920 (film)	tt1301698	https://upload.wikimedia.org/wikipedia/en/thumb/.../1920_(film).jpg	https://en.wikipedia.org/wiki/1920_(film)	1920	1920	0
1498	16 December (film)	tt0313844	https://upload.wikimedia.org/wikipedia/en/thumb/.../16_December_(film).jpg	https://en.wikipedia.org/wiki/16_December_(film)	16-Dec	16-Dec	0

1629 rows × 19 columns

```
In [85]: students = pd.DataFrame(
    {
        'name': ['nitish', 'ankit', 'rupesh', np.nan, 'mrityunjay', np.nan, 'rishabh', np.nan, 'aditya', np.nan],
        'college': ['bit', 'iit', 'vit', np.nan, np.nan, 'vlsi', np.nan, np.nan, 'git'],
        'branch': ['eee', 'it', 'cse', np.nan, 'me', 'ce', 'civ', 'cse', 'bio', np.nan],
        'cgpa': [6.66, 8.25, 6.41, np.nan, 5.6, 9.0, 7.4, 10, 7.4, np.nan],
        'package': [4, 5, 6, np.nan, 6, 7, 8, 9, np.nan, np.nan]
    }
)
students
```

Out[85]:

	name	college	branch	cgpa	package
0	nitish	bit	eee	6.66	4.0
1	ankit	iit	it	8.25	5.0
2	rupesh	vit	cse	6.41	6.0
3	NaN	NaN	NaN	NaN	NaN
4	mrityunjay	NaN	me	5.60	6.0
5	NaN	vlsi	ce	9.00	7.0
6	rishabh	ssit	civ	7.40	8.0
7	NaN	NaN	cse	10.00	9.0
8	aditya	NaN	bio	7.40	NaN
9	NaN	git	NaN	NaN	NaN

In [86]: students.sort_values('name', ascending=False, na_position='first') # inplace=True (for Permanent Changes)

Out[86]:

	name	college	branch	cgpa	package
3	NaN	NaN	NaN	NaN	NaN
5	NaN	vlsi	ce	9.00	7.0
7	NaN	NaN	cse	10.00	9.0
9	NaN	git	NaN	NaN	NaN
2	rupesh	vit	cse	6.41	6.0
6	rishabh	ssit	civ	7.40	8.0
0	nitish	bit	eee	6.66	4.0
4	mrityunjay	NaN	me	5.60	6.0
1	ankit	iit	it	8.25	5.0
8	aditya	NaN	bio	7.40	NaN

In [87]: students

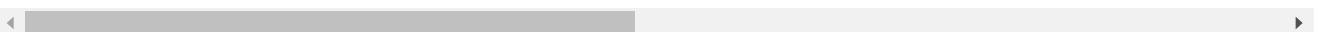
Out[87]:

	name	college	branch	cgpa	package
0	nitish	bit	eee	6.66	4.0
1	ankit	iit	it	8.25	5.0
2	rupesh	vit	cse	6.41	6.0
3	NaN	NaN	NaN	NaN	NaN
4	mrityunjay	NaN	me	5.60	6.0
5	NaN	vlsi	ce	9.00	7.0
6	rishabh	ssit	civ	7.40	8.0
7	NaN	NaN	cse	10.00	9.0
8	aditya	NaN	bio	7.40	NaN
9	NaN	git	NaN	NaN	NaN

In [88]: `movies.sort_values(['year_of_release', 'title_x'], ascending=[True, False]).head(2)`

Out[88]:

	title_x	imdb_id	poster_path	wiki_link	title_y	original_title	is_adult	ye
1623	Zubeidaa	tt0255713	https://upload.wikimedia.org/wikipedia/en/thum...	https://en.wikipedia.org/wiki/Zubeidaa	Zubeidaa	Zubeidaa	0	
1625	Yeh Zindagi Ka Safar	tt0298607	https://upload.wikimedia.org/wikipedia/en/thum...	https://en.wikipedia.org/wiki/Yeh_Zindagi_Ka_S...	Zindagi Ka Safar	Yeh Zindagi Ka Safar	0	



rank(Series)

In [89]: `batsman=pd.read_csv("batsman_runs_ipl.csv")`

In [90]: `batsman.head(2)`

Out[90]:

	batter	batsman_run
0	AAshish Reddy	280
1	A Badoni	161

In [91]: `batsman['batsman_run'].rank(ascending=False)`

Out[91]:

0	166.5
1	226.0
2	535.0
3	329.0
4	402.5
	...
600	594.0
601	343.0
602	547.5
603	27.0
604	256.0

Name: batsman_run, Length: 605, dtype: float64

In [92]: `batsman['batsman_rank'] = batsman['batsman_run'].rank(ascending=False)`
`batsman.sort_values('batsman_rank')`

Out[92]:

	batter	batsman_run	batsman_rank
569	V Kohli	6634	1.0
462	S Dhawan	6244	2.0
130	DA Warner	5883	3.0
430	RG Sharma	5881	4.0
493	SK Raina	5536	5.0

512	SS Cottrell	0	594.0
466	S Kaushik	0	594.0
203	IC Pandey	0	594.0
467	S Ladda	0	594.0
468	S Lamichhane	0	594.0

605 rows × 3 columns

sort_index (Series and dataframe)

```
In [93]: marks = {  
    'maths':67,  
    'english':57,  
    'science':89,  
    'hindu':100  
}  
  
marks_series = pd.Series(marks)  
marks_series
```

```
Out[93]: maths      67  
english     57  
science     89  
hindu      100  
dtype: int64
```

```
In [94]: marks_series.sort_index()
```

```
Out[94]: english     57  
hindu      100  
maths      67  
science     89  
dtype: int64
```

```
In [95]: movies.sort_index(ascending=False)
```

Out[95]:

	title_x	imdb_id	poster_path	wiki_link	title_y	original_title	is_adult
1628	Humsafar	tt2403201	https://upload.wikimedia.org/wikipedia/en/thumb/0/0d/Humsafar.jpg/220px-Humsafar.jpg	https://en.wikipedia.org/wiki/Humsafar	Humsafar	Humsafar	0
1627	Daaka	tt10833860	https://upload.wikimedia.org/wikipedia/en/thumb/0/0e/Daaka.jpg/220px-Daaka.jpg	https://en.wikipedia.org/wiki/Daaka	Daaka	Daaka	0
1626	Sabse Bada Sukh	tt0069204		Nan	https://en.wikipedia.org/wiki/Sabse_Bada_Sukh	Sabse Bada Sukh	Sabse Bada Sukh
1625	Yeh Zindagi Ka Safar	tt0298607	https://upload.wikimedia.org/wikipedia/en/thumb/0/0f/Yeh_Zindagi_Ka_Safar.jpg/220px-Yeh_Zindagi_Ka_Safar.jpg	https://en.wikipedia.org/wiki/Yeh_Zindagi_Ka_Safar	Yeh Zindagi Ka Safar	Yeh Zindagi Ka Safar	0
1624	Tera Mera Saath Rahen	tt0301250	https://upload.wikimedia.org/wikipedia/en/2/2b/Tera_Mera_Saath_Rahen.jpg	https://en.wikipedia.org/wiki/Tera_Mera_Saath_Rahen	Tera Mera Saath Rahen	Tera Mera Saath Rahen	0
...
4	Evening Shadows	tt6028796		Nan	https://en.wikipedia.org/wiki/Evening_Shadows	Evening Shadows	Evening Shadows
3	Why Cheat India	tt8108208	https://upload.wikimedia.org/wikipedia/en/thumb/0/0a/Why_Cheat_India.jpg/220px-Why_Cheat_India.jpg	https://en.wikipedia.org/wiki/Why_Cheat_India	Why Cheat India	Why Cheat India	0
2	The Accidental Prime Minister (film)	tt6986710	https://upload.wikimedia.org/wikipedia/en/thumb/0/0a/The_Accidental_Prime_Minister_film.jpg/220px-The_Accidental_Prime_Minister_film.jpg	https://en.wikipedia.org/wiki/The_Accidental_Prime_Minister	The Accidental Prime Minister	The Accidental Prime Minister	0
1	Battalion 609	tt9472208		Nan	https://en.wikipedia.org/wiki/Battalion_609	Battalion 609	Battalion 609
0	Uri: The Surgical Strike	tt8291224	https://upload.wikimedia.org/wikipedia/en/thumb/0/0a/Uri:_The_Surgical_Strike.jpg/220px-Uri:_The_Surgical_Strike.jpg	https://en.wikipedia.org/wiki/Uri:_The_Surgical_Strike	Uri: The Surgical Strike	Uri: The Surgical Strike	0

1629 rows × 19 columns



```
In [96]: # set_index(dataframe) -> inplace
batsman.set_index('batter', inplace=True)
```

In [99]: how to delete columns

```
File "C:\Users\user\AppData\Local\Temp\ipykernel_17496\2400687195.py", line 1
    how to delete columns
      ^
SyntaxError: invalid syntax
```

```
In [140]: # This code drops two columns from the 'batsman' dataframe: 'Level_0' and 'index'.
# The 'inplace=True' parameter ensures that the original dataframe is modified instead of creating a new one.
```

```
batsman.drop(['index'], axis=1, inplace=True)
```

```
In [101]: # reset_index(series + dataframe) -> drop parameter
batsman.reset_index(inplace=True)
```

```
In [102]: batsman
```

Out[102]:

	batter	batsman_run	batsman_rank
0	A Ashish Reddy	280	166.5
1	A Badoni	161	226.0
2	A Chandila	4	535.0
3	A Chopra	53	329.0
4	A Choudhary	25	402.5
...
600	Yash Dayal	0	594.0
601	Yashpal Singh	47	343.0
602	Younis Khan	3	547.5
603	Yuvraj Singh	2754	27.0
604	Z Khan	117	256.0

605 rows × 3 columns

```
In [103]: # how to replace existing index without Loosing
batsman.reset_index().set_index('batsman_rank')
```

Out[103]:

	index	batter	batsman_run
	batsman_rank		
166.5	0	A Ashish Reddy	280
226.0	1	A Badoni	161
535.0	2	A Chandila	4
329.0	3	A Chopra	53
402.5	4	A Choudhary	25
...
594.0	600	Yash Dayal	0
343.0	601	Yashpal Singh	47
547.5	602	Younis Khan	3
27.0	603	Yuvraj Singh	2754
256.0	604	Z Khan	117

605 rows × 3 columns

```
In [104]: batsman
```

Out[104]:

	batter	batsman_run	batsman_rank
0	A Ashish Reddy	280	166.5
1	A Badoni	161	226.0
2	A Chandila	4	535.0
3	A Chopra	53	329.0
4	A Choudhary	25	402.5
...
600	Yash Dayal	0	594.0
601	Yashpal Singh	47	343.0
602	Younis Khan	3	547.5
603	Yuvraj Singh	2754	27.0
604	Z Khan	117	256.0

605 rows × 3 columns

```
In [105]: # series to dataframe using reset_index  
marks_series.reset_index()
```

```
Out[105]:  
    index    0  
0   maths   67  
1   english  57  
2   science  89  
3   hindi   100
```

```
In [106]: type(marks_series.reset_index())
```

```
Out[106]: pandas.core.frame.DataFrame
```

```
rename(dataframe) -> index
```

```
In [107]: movies.set_index('title_x', inplace=True)
```

In [108]: movies

Out[108]:

	imdb_id	poster_path	wiki_link	title_y	original_title	is_adult	year_
	title_x						
Uri: The Surgical Strike	tt8291224	https://upload.wikimedia.org/wikipedia/en/thumb/.../https://en.wikipedia.org/wiki/Uri:_The_Surgical_...	Uri: The Surgical Strike	Uri: The Surgical Strike	Uri: The Surgical Strike	0	
Battalion 609	tt9472208	NaN	https://en.wikipedia.org/wiki/Battalion_609	Battalion 609	Battalion 609	0	
The Accidental Prime Minister (film)	tt6986710	https://upload.wikimedia.org/wikipedia/en/thumb/.../https://en.wikipedia.org/wiki/The_Accidental_P...	The Accidental Prime Minister	The Accidental Prime Minister	The Accidental Prime Minister	0	
Why Cheat India	tt8108208	https://upload.wikimedia.org/wikipedia/en/thumb/.../https://en.wikipedia.org/wiki/Why_Cheat_India	Why Cheat India	Why Cheat India	Why Cheat India	0	
Evening Shadows	tt6028796	NaN	https://en.wikipedia.org/wiki/Evening_Shadows	Evening Shadows	Evening Shadows	0	
...
Tera Mera Saath Rahen	tt0301250	https://upload.wikimedia.org/wikipedia/en/2/2b.../https://en.wikipedia.org/wiki/Tera_Mera_Saath_...	Tera Mera Saath Rahen	Tera Mera Saath Rahen	Tera Mera Saath Rahen	0	
Yeh Zindagi Ka Safar	tt0298607	https://upload.wikimedia.org/wikipedia/en/thumb/.../https://en.wikipedia.org/wiki/Yeh_Zindagi_Ka_S...	Yeh Zindagi Ka Safar	Yeh Zindagi Ka Safar	Yeh Zindagi Ka Safar	0	
Sabse Bada Sukh	tt0069204	NaN	https://en.wikipedia.org/wiki/Sabse_Bada_Sukh	Sabse Bada Sukh	Sabse Bada Sukh	0	
Daaka	tt10833860	https://upload.wikimedia.org/wikipedia/en/thumb/.../https://en.wikipedia.org/wiki/Daaka	Daaka	Daaka	Daaka	0	
Humsafar	tt2403201	https://upload.wikimedia.org/wikipedia/en/thumb/.../https://en.wikipedia.org/wiki/Humsafar	Humsafar	Humsafar	Humsafar	0	

1629 rows × 18 columns

In [109]: #Rename the columns
movies.rename(columns={'imdb_id' : 'imdb', 'poster_path':'link'},inplace=True)

In [110]: movies

Out[110]:

	imdb	link	wiki_link	title_y	original_title	is_adult	year_
	title_x						
Uri: The Surgical Strike	tt8291224	https://upload.wikimedia.org/wikipedia/en/thumb.../	https://en.wikipedia.org/wiki/Uri:_The_Surgical...	Uri: The Surgical Strike	Uri: The Surgical Strike	0	
Battalion 609	tt9472208		NaN	https://en.wikipedia.org/wiki/Battalion_609	Battalion 609	Battalion 609	0
The Accidental Prime Minister (film)	tt6986710	https://upload.wikimedia.org/wikipedia/en/thumb.../	https://en.wikipedia.org/wiki/The_Accidental_P...	The Accidental Prime Minister	The Accidental Prime Minister	0	
Why Cheat India	tt8108208	https://upload.wikimedia.org/wikipedia/en/thumb.../	https://en.wikipedia.org/wiki/Why_Cheat_India	Why Cheat India	Why Cheat India	0	
Evening Shadows	tt6028796		NaN	https://en.wikipedia.org/wiki/Evening_Shadows	Evening Shadows	Evening Shadows	0
...
Tera Mera Saath Rahen	tt0301250	https://upload.wikimedia.org/wikipedia/en/2/2b.../	https://en.wikipedia.org/wiki/Tera_Mera_Saath_...	Tera Mera Saath Rahen	Tera Mera Saath Rahen	0	
Yeh Zindagi Ka Safar	tt0298607	https://upload.wikimedia.org/wikipedia/en/thumb.../	https://en.wikipedia.org/wiki/Yeh_Zindagi_Ka_S...	Yeh Zindagi Ka Safar	Yeh Zindagi Ka Safar	0	
Sabse Bada Sukh	tt0069204		NaN	https://en.wikipedia.org/wiki/Sabse_Bada_Sukh	Sabse Bada Sukh	Sabse Bada Sukh	0
Daaka	tt10833860	https://upload.wikimedia.org/wikipedia/en/thumb.../		https://en.wikipedia.org/wiki/Daaka	Daaka	Daaka	0
Humsafar	tt2403201	https://upload.wikimedia.org/wikipedia/en/thumb.../		https://en.wikipedia.org/wiki/Humsafar	Humsafar	Humsafar	0

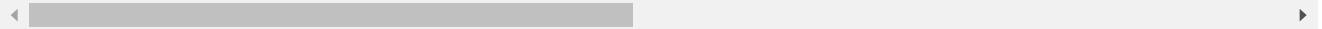
1629 rows × 18 columns

```
In [111]: # Rename the index
movies.rename(index={'Uri: The Surgical Strike':'uri','Humsafar':'Hum'})
```

Out[111]:

	imdb	link	wiki_link	title_y	original_title	is_adult	year_
	title_x						
uri	tt8291224	https://upload.wikimedia.org/wikipedia/en/thumb/.../The_Surgical_Strike.jpg	https://en.wikipedia.org/wiki/Uri:_The_Surgical_Strike	Uri: The Surgical Strike	Uri: The Surgical Strike	0	
Battalion 609	tt9472208		https://en.wikipedia.org/wiki/Battalion_609	Battalion 609	Battalion 609	0	
The Accidental Prime Minister (film)	tt6986710	https://upload.wikimedia.org/wikipedia/en/thumb/.../The_Accidental_Prime_Minister.jpg	https://en.wikipedia.org/wiki/The_Accidental_Prime_Minister	The Accidental Prime Minister	The Accidental Prime Minister	0	
Why Cheat India	tt8108208	https://upload.wikimedia.org/wikipedia/en/thumb/.../Why_Cheat_India.jpg	https://en.wikipedia.org/wiki/Why_Cheat_India	Why Cheat India	Why Cheat India	0	
Evening Shadows	tt6028796		https://en.wikipedia.org/wiki/Evening_Shadows	Evening Shadows	Evening Shadows	0	
...	
Tera Mera Saath Rahen	tt0301250	https://upload.wikimedia.org/wikipedia/en/2/2b/.../Tera_Mera_Saath_Rahen.jpg	https://en.wikipedia.org/wiki/Tera_Mera_Saath_Rahen	Tera Mera Saath Rahen	Tera Mera Saath Rahen	0	
Yeh Zindagi Ka Safar	tt0298607	https://upload.wikimedia.org/wikipedia/en/thumb/.../Yeh_Zindagi_Ka_Safar.jpg	https://en.wikipedia.org/wiki/Yeh_Zindagi_Ka_Safar	Yeh Zindagi Ka Safar	Yeh Zindagi Ka Safar	0	
Sabse Bada Sukh	tt0069204		https://en.wikipedia.org/wiki/Sabse_Bada_Sukh	Sabse Bada Sukh	Sabse Bada Sukh	0	
Daaka	tt10833860	https://upload.wikimedia.org/wikipedia/en/thumb/.../Daaka.jpg	https://en.wikipedia.org/wiki/Daaka	Daaka	Daaka	0	
Hum	tt2403201	https://upload.wikimedia.org/wikipedia/en/thumb/.../Humsafar.jpg	https://en.wikipedia.org/wiki/Humsafar	Humsafar	Humsafar	0	

1629 rows × 18 columns



unique

```
In [112]: # unique(series)
temp = pd.Series([1,1,2,2,3,3,4,4,5,5,np.nan,np.nan])
print(temp)
temp.unique()
```

```
0    1.0
1    1.0
2    2.0
3    2.0
4    3.0
5    3.0
6    4.0
7    4.0
8    5.0
9    5.0
10   NaN
11   NaN
dtype: float64
```

```
Out[112]: array([ 1.,  2.,  3.,  4.,  5., nan])
```

```
In [113]: ipl['Season'].unique()
```

```
Out[113]: ['2022', '2021', '2020/21', '2019', '2018', ..., '2012', '2011', '2009/10', '2009', '2007/08']
Length: 15
Categories (15, object): ['2007/08', '2009', '2009/10', '2011', ..., '2019', '2020/21', '2021', '2022']
```

- `nunique` : returns the number of unique elements in a pandas Series or DataFrame. It doesn't count the missing values (NaNs) by default. If you want to count the missing values, you can set the argument "dropna" to False.
- `unique`: returns the unique elements in a pandas Series or DataFrame. It counts the missing values (NaNs) by default. If you don't want to count the missing values, you can use the "dropna" argument and set it to True.

```
In [114]: len(ipl['Season'].unique())
```

```
Out[114]: 15
```

```
In [115]: # nunique(series + dataframe) -> does not count nan -> dropna parameter
ipl['Season'].nunique()
```

```
Out[115]: 15
```

`isnull(series + dataframe)`

```
In [116]: students
```

```
Out[116]:
```

	name	college	branch	cgpa	package
0	nitish	bit	eee	6.66	4.0
1	ankit	iit	it	8.25	5.0
2	rupesh	vit	cse	6.41	6.0
3	NaN	NaN	NaN	NaN	NaN
4	mrityunjay	NaN	me	5.60	6.0
5	NaN	vlsi	ce	9.00	7.0
6	rishabh	ssit	civ	7.40	8.0
7	NaN	NaN	cse	10.00	9.0
8	aditya	NaN	bio	7.40	NaN
9	NaN	git	NaN	NaN	NaN

```
In [117]: students['name'].isnull()
```

```
Out[117]: 0    False
1    False
2    False
3     True
4    False
5     True
6    False
7     True
8    False
9     True
Name: name, dtype: bool
```

```
In [118]: # notnull(series + dataframe)
students['name'].notnull()
```

```
Out[118]: 0    True
1    True
2    True
3   False
4    True
5   False
6    True
7   False
8    True
9   False
Name: name, dtype: bool
```

```
In [119]: students['name'][students['name'].notnull()]
```

```
Out[119]: 0      nitish
1      ankit
2      rupesh
3  mrityunjay
4      rishabh
5      aditya
Name: name, dtype: object
```

```
In [120]: # hasnans(series)
students['college'].hasnans
```

```
Out[120]: True
```

```
In [121]: students.isnull()
```

```
Out[121]:
   name  college  branch  cgpa  package
0  False   False  False  False  False
1  False   False  False  False  False
2  False   False  False  False  False
3   True    True   True   True   True
4  False    True  False  False  False
5   True   False  False  False  False
6  False   False  False  False  False
7   True    True  False  False  False
8  False    True  False  False   True
9   True   False   True   True   True
```

In [122]: `students.notnull()`

Out[122]:

	name	college	branch	cgpa	package
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	False	False	False	False	False
4	True	False	True	True	True
5	False	True	True	True	True
6	True	True	True	True	True
7	False	False	True	True	True
8	True	False	True	True	False
9	False	True	False	False	False

dropna -> for Missing Values

In [123]: `# dropna(series + dataframe) -> how parameter -> works like or
students['name'].dropna()`

Out[123]:

0	nitish
1	ankit
2	rupesh
4	mrityunjay
6	rishabh
8	aditya

Name: name, dtype: object

how : {'any', 'all'}, default 'any'

Determine if row or column is removed from DataFrame, when we have at least one NA or all NA.

- * 'any' : If any NA values are present, drop that row or column.
- * 'all' : If all values are NA, drop that row or column.

In [124]: `students.dropna(how='any')`

Out[124]:

	name	college	branch	cgpa	package
0	nitish	bit	eee	6.66	4.0
1	ankit	iit	it	8.25	5.0
2	rupesh	vit	cse	6.41	6.0
6	rishabh	ssit	civ	7.40	8.0

In [125]: `students.dropna(how='all')`

Out[125]:

	name	college	branch	cgpa	package
0	nitish	bit	eee	6.66	4.0
1	ankit	iit	it	8.25	5.0
2	rupesh	vit	cse	6.41	6.0
4	mrityunjay	NaN	me	5.60	6.0
5	NaN	vlsi	ce	9.00	7.0
6	rishabh	ssit	civ	7.40	8.0
7	NaN	NaN	cse	10.00	9.0
8	aditya	NaN	bio	7.40	NaN
9	NaN	git	NaN	NaN	NaN

subset : array-like, optional

Labels along other axis to consider, e.g. if you are dropping rows these would be a list of columns to include.

```
In [126]: students.dropna(subset=['name'])
```

Out[126]:

	name	college	branch	cgpa	package
0	nitish	bit	eee	6.66	4.0
1	ankit	iit	it	8.25	5.0
2	rupesh	vit	cse	6.41	6.0
4	mrityunjay	NaN	me	5.60	6.0
6	rishabh	ssit	civ	7.40	8.0
8	aditya	NaN	bio	7.40	NaN

```
In [127]: students.dropna(subset=['name','college'])
```

Out[127]:

	name	college	branch	cgpa	package
0	nitish	bit	eee	6.66	4.0
1	ankit	iit	it	8.25	5.0
2	rupesh	vit	cse	6.41	6.0
6	rishabh	ssit	civ	7.40	8.0

fill na (series + dataframe)

```
In [128]: students
```

Out[128]:

	name	college	branch	cgpa	package
0	nitish	bit	eee	6.66	4.0
1	ankit	iit	it	8.25	5.0
2	rupesh	vit	cse	6.41	6.0
3	NaN	NaN	NaN	NaN	NaN
4	mrityunjay	NaN	me	5.60	6.0
5	NaN	vlsi	ce	9.00	7.0
6	rishabh	ssit	civ	7.40	8.0
7	NaN	NaN	cse	10.00	9.0
8	aditya	NaN	bio	7.40	NaN
9	NaN	git	NaN	NaN	NaN

```
In [129]: students['name'].fillna('unknown')
```

Out[129]:

0	nitish
1	ankit
2	rupesh
3	unknown
4	mrityunjay
5	unknown
6	rishabh
7	unknown
8	aditya
9	unknown

Name: name, dtype: object

In [130]: `students.fillna('0')`

Out[130]:

	name	college	branch	cgpa	package
0	nitish	bit	eee	6.66	4.0
1	ankit	iit	it	8.25	5.0
2	rupesh	vit	cse	6.41	6.0
3	0	0	0	0	0
4	mrityunjay	0	me	5.6	6.0
5	0	vlsi	ce	9.0	7.0
6	rishabh	ssit	civ	7.4	8.0
7	0	0	cse	10.0	9.0
8	aditya	0	bio	7.4	0
9	0	git	0	0	0

In [131]:

`students['package'].fillna(students['package'].mean())`

Out[131]: 0 4.000000

1 5.000000

2 6.000000

3 6.428571

4 6.000000

5 7.000000

6 8.000000

7 9.000000

8 6.428571

9 6.428571

Name: package, dtype: float64

In [132]: `students['name'].fillna(method='ffill') #forward fill method`

Out[132]: 0 nitish

1 ankit

2 rupesh

3 rupesh

4 mrityunjay

5 mrityunjay

6 rishabh

7 rishabh

8 aditya

9 aditya

Name: name, dtype: object

In [133]: `students['name'].fillna(method='bfill') # backward fill method`

Out[133]: 0 nitish

1 ankit

2 rupesh

3 mrityunjay

4 mrityunjay

5 rishabh

6 rishabh

7 aditya

8 aditya

9 NaN

Name: name, dtype: object

drop_duplicates

In [134]: `# drop_duplicates(series + dataframe) -> works Like and -> duplicated()`

In [135]: marks

Out[135]: {'maths': 67, 'english': 57, 'science': 89, 'hindi': 100}

In [144]: marks

Out[144]:

	iq	marks	package
0	100	80	10
1	90	70	7
2	120	100	14
3	80	70	14
4	80	70	14

In [145]: marks.drop_duplicates()

Out[145]:

	iq	marks	package
0	100	80	10
1	90	70	7
2	120	100	14
3	80	70	14

In [146]: marks.drop_duplicates(keep='last')

Out[146]:

	iq	marks	package
0	100	80	10
1	90	70	7
2	120	100	14
3	80	70	14

In [148]: # find the last match played by virat kohli in Delhi
ipl.sample(2)

Out[148]:

	ID	City	Date	Season	MatchNumber	Team1	Team2	Venue	TossWinner	TossDecision	SuperOver	WinningTeam	WonBy	M
6	1304114	Mumbai	2022-05-20	2022	68	Chennai Super Kings	Rajasthan Royals	Brabourne Stadium, Mumbai	Chennai Super Kings	bat	N	Rajasthan Royals	Wickets	
276	1136598	Indore	2018-05-06	2018	38	Rajasthan Royals	Kings XI Punjab	Holkar Cricket Stadium	Kings XI Punjab	field	N	Kings XI Punjab	Wickets	

In [149]: ipl['all_Players'] = ipl['Team1Players'] + ipl['Team2Players']
ipl.sample(2)

Out[149]:

	Player_of_Match	Team1Players	Team2Players	Umpire1	Umpire2						
MA iram ium, yauk	Rajasthan Royals	bat ...	Chennai Super Kings	Wickets	8.0	NaN	MEK Hussey	['MEK Hussey', 'M Vijay', 'SK Raina', 'JA Mork...']	['SR Watson', 'R Dravid', 'AL Menaria', 'J Bot...']	SS Hazare	RB Tiffin
iede ium, nbai	Delhi Capitals	field ...	Royal Challengers Bangalore	Runs	16.0	NaN	KD Karthik	['F du Plessis', 'Anuj Rawat', 'V Kohli', 'GJ ...']	['PP Shaw', 'DA Warner', 'MR Marsh', 'RR Pant...']	Chirra	Madanagopal J

```
In [154]: def did_kohli_play(players_list):
    return 'V Kohli' in players_list
```

```
In [160]: ipl['did_kohli_play'] = ipl['all Players'].apply(did_kohli_play)
```

```
In [165]: ipl.sample(2)
```

```
Out[165]:
```

ID	City	Date	Season	MatchNumber	Team1	Team2	Venue	TossWinner	TossDecision	Margin	method	Player_of_Match
940	335991	Chandigarh	2008-04-25	2007/08	10	Kings XI Punjab	Mumbai Indians	Punjab Cricket Association Stadium, Mohali	Mumbai Indians	field	...	66.0
826	419114	Delhi	2010-03-17	2009/10	9	Delhi Daredevils	Mumbai Indians	Feroz Shah Kotla	Delhi Daredevils	field	...	98.0

2 rows × 23 columns

```
In [166]: ipl[(ipl['City'] == 'Delhi') & (ipl['did_kohli_play'] == True)]
```

```
Out[166]:
```

ID	City	Date	Season	MatchNumber	Team1	Team2	Venue	TossWinner	TossDecision	Margin	method	Player_of_Match
208	1178421	Delhi	2019-04-28	2019	46	Delhi Capitals	Royal Challengers Bangalore	Arun Jaitley Stadium	Delhi Capitals	bat	...	16.0
269	1136605	Delhi	2018-05-12	2018	45	Delhi Daredevils	Royal Challengers Bangalore	Arun Jaitley Stadium	Royal Challengers Bangalore	field	...	5.0
318	1082646	Delhi	2017-05-14	2017	56	Delhi Daredevils	Royal Challengers Bangalore	Feroz Shah Kotla	Royal Challengers Bangalore	bat	...	10.0
467	829757	Delhi	2015-04-26	2015	26	Delhi Daredevils	Royal Challengers Bangalore	Feroz Shah Kotla	Royal Challengers Bangalore	field	...	10.0
571	598054	Delhi	2013-05-10	2013	57	Delhi Daredevils	Royal Challengers Bangalore	Feroz Shah Kotla	Delhi Daredevils	field	...	4.0
638	548372	Delhi	2012-05-17	2012	67	Delhi Daredevils	Royal Challengers Bangalore	Feroz Shah Kotla	Delhi Daredevils	field	...	21.0
746	501227	Delhi	2011-04-26	2011	30	Delhi Daredevils	Royal Challengers Bangalore	Feroz Shah Kotla	Royal Challengers Bangalore	field	...	3.0
801	419140	Delhi	2010-04-04	2009/10	35	Delhi Daredevils	Royal Challengers Bangalore	Feroz Shah Kotla	Delhi Daredevils	bat	...	37.0
933	335998	Delhi	2008-04-30	2007/08	17	Delhi Daredevils	Royal Challengers Bangalore	Feroz Shah Kotla	Royal Challengers Bangalore	field	...	10.0

9 rows × 23 columns

```
In [168]: ipl[( ipl['City'] == 'Delhi') & (ipl['did_kohli_play'] == True)].drop_duplicates(subset=['City','did_kohli_play'],keep='last')
```

Out[168]:

	ID	City	Date	Season	MatchNumber	Team1	Team2	Venue	TossWinner	TossDecision	Margin	method	Player_of_Match	Te
208	1178421	Delhi	2019-04-28	2019	46	Delhi Capitals	Royal Challengers Bangalore	Arun Jaitley Stadium	Delhi Capitals	bat ...	16.0	NaN	S Dhawan	[F D]

1 rows × 23 columns

drop(series + dataframe)

```
In [169]: # Series
temp = pd.Series([10,2,3,16,45,78,10])
temp
```

```
Out[169]: 0    10
1     2
2     3
3    16
4    45
5    78
6    10
dtype: int64
```

```
In [170]: temp.drop(index=[0,6])
```

```
Out[170]: 1    2
2     3
3    16
4    45
5    78
dtype: int64
```

```
In [171]: students
```

Out[171]:

	name	college	branch	cgpa	package
0	nitish	bit	eee	6.66	4.0
1	ankit	iit	it	8.25	5.0
2	rupesh	vit	cse	6.41	6.0
3	NaN	NaN	NaN	NaN	NaN
4	mrityunjay	NaN	me	5.60	6.0
5	NaN	vlsi	ce	9.00	7.0
6	rishabh	ssit	civ	7.40	8.0
7	NaN	NaN	cse	10.00	9.0
8	aditya	NaN	bio	7.40	NaN
9	NaN	git	NaN	NaN	NaN

```
In [172]: students.drop(columns=['branch', 'cgpa']) # To delete Columns
```

Out[172]:

	name	college	package
0	nitish	bit	4.0
1	ankit	iit	5.0
2	rupesh	vit	6.0
3	NaN	NaN	NaN
4	mrityunjay	NaN	6.0
5	NaN	vlsi	7.0
6	rishabh	ssit	8.0
7	NaN	NaN	9.0
8	aditya	NaN	NaN
9	NaN	git	NaN

```
In [173]: students.drop(index=[0,8]) # to delete rows
```

Out[173]:

	name	college	branch	cgpa	package
1	ankit	iit	it	8.25	5.0
2	rupesh	vit	cse	6.41	6.0
3	NaN	NaN	NaN	NaN	NaN
4	mrityunjay	NaN	me	5.60	6.0
5	NaN	vlsi	ce	9.00	7.0
6	rishabh	ssit	civ	7.40	8.0
7	NaN	NaN	cse	10.00	9.0
9	NaN	git	NaN	NaN	NaN

```
In [174]: students.set_index('name')
```

Out[174]:

name	college	branch	cgpa	package
nitish	bit	eee	6.66	4.0
ankit	iit	it	8.25	5.0
rupesh	vit	cse	6.41	6.0
NaN	NaN	NaN	NaN	NaN
mrityunjay	NaN	me	5.60	6.0
NaN	vlsi	ce	9.00	7.0
rishabh	ssit	civ	7.40	8.0
NaN	NaN	cse	10.00	9.0
aditya	NaN	bio	7.40	NaN
NaN	git	NaN	NaN	NaN

```
In [175]: students.set_index('name').drop(index=['nitish', 'aditya']) # delete by names
```

Out[175]:

name	college	branch	cgpa	package
ankit	iit	it	8.25	5.0
rupesh	vit	cse	6.41	6.0
NaN	NaN	NaN	NaN	NaN
mrityunjay	NaN	me	5.60	6.0
NaN	vlsi	ce	9.00	7.0
rishabh	ssit	civ	7.40	8.0
NaN	NaN	cse	10.00	9.0
NaN	git	NaN	NaN	NaN

apply(series + dataframe)

```
In [176]: # series
temp = pd.Series([10,20,30,40,50])
temp
```

```
Out[176]: 0    10
1    20
2    30
3    40
4    50
dtype: int64
```

```
In [177]: def sigmoid(value):
    return 1/(1+np.exp(-value))
```

```
In [178]: temp.apply(sigmoid)
```

```
Out[178]: 0    1.000045
1    1.000000
2    1.000000
3    1.000000
4    1.000000
dtype: float64
```

```
In [179]: # On data Frame
points = pd.DataFrame(
{
    '1st point':[(3,4),(-6,5),(0,0),(-10,1),(4,5)],
    '2nd point':[(-3,4),(0,0),(2,2),(10,10),(1,1)]
})
points
```

```
Out[179]:
      1st point  2nd point
0     (3, 4)    (-3, 4)
1     (-6, 5)   (0, 0)
2     (0, 0)     (2, 2)
3     (-10, 1)  (10, 10)
4     (4, 5)     (1, 1)
```

```
In [180]: def euclidean(row):
    point_A = row['1st point']
    point_B = row['2nd point']
    return ((point_A[0] - point_B[0])**2 + (point_A[1] - point_B[1])**2)**0.5
```

```
In [182]: points.apply(euclidean, axis=1) # mention axis=1 because its row wise
```

```
Out[182]: 0    6.000000
1    7.810250
2    2.828427
3    21.931712
4    5.000000
dtype: float64
```

```
In [184]: points['distance'] = points.apply(euclidean, axis=1)
points
```

```
Out[184]:
      1st point  2nd point  distance
0     (3, 4)    (-3, 4)  6.000000
1     (-6, 5)   (0, 0)   7.810250
2     (0, 0)     (2, 2)   2.828427
3     (-10, 1)  (10, 10) 21.931712
4     (4, 5)     (1, 1)   5.000000
```

Groupby (Applies on Categorical data)

```
In [187]: movies = pd.read_csv("imdb-top-1000.csv")
```

```
In [189]: movies.head(1)
```

```
Out[189]:
```

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	Director	Star1	No_of_Votes	Gross	Metascore
0	The Shawshank Redemption	1994	142	Drama	9.3	Frank Darabont	Tim Robbins	2343110	28341469.0	80.0

```
In [190]: movies.groupby('Genre')
```

```
Out[190]: <pandas.core.groupby.generic.DataFrameGroupBy object at 0x000001EFE1E1F670>
```

```
In [195]: genres = movies.groupby('Genre')
```

```
In [196]: # Applying builtin aggregation fuctions on groupby objects
genres.sum()
```

```
Out[196]:
```

Genre	Runtime	IMDB_Rating	No_of_Votes	Gross	Metascore
Action	22196	1367.3	72282412	3.263226e+10	10499.0
Adventure	9656	571.5	22576163	9.496922e+09	5020.0
Animation	8166	650.3	21978630	1.463147e+10	6082.0
Biography	11970	698.6	24006844	8.276358e+09	6023.0
Comedy	17380	1224.7	27620327	1.566387e+10	9840.0
Crime	13524	857.8	33533615	8.452632e+09	6706.0
Drama	36049	2299.7	61367304	3.540997e+10	19208.0
Family	215	15.6	551221	4.391106e+08	158.0
Fantasy	170	16.0	146222	7.827267e+08	0.0
Film-Noir	312	23.9	367215	1.259105e+08	287.0
Horror	1123	87.0	3742556	1.034649e+09	880.0
Mystery	1429	95.7	4203004	1.256417e+09	633.0
Thriller	108	7.8	27733	1.755074e+07	81.0
Western	593	33.4	1289665	5.822151e+07	313.0

```
In [197]: genres.mean()
```

```
Out[197]:
```

Genre	Runtime	IMDB_Rating	No_of_Votes	Gross	Metascore
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

In [198]: `genres.min()`

Out[198]:

Genre	Series_Title	Released_Year	Runtime	IMDB_Rating	Director	Star1	No_of_Votes	Gross	Metascore
Action	300	1924	45	7.6	Abhishek Chaubey	Aamir Khan	25312	3296.0	33.0
Adventure	2001: A Space Odyssey	1925	88	7.6	Akira Kurosawa	Aamir Khan	29999	61001.0	41.0
Animation	Akira	1940	71	7.6	Adam Elliot	Adrian Molina	25229	128985.0	61.0
Biography	12 Years a Slave	1928	93	7.6	Adam McKay	Adrien Brody	27254	21877.0	48.0
Comedy	(500) Days of Summer	1921	68	7.6	Alejandro G. Iñárritu	Aamir Khan	26337	1305.0	45.0
Crime	12 Angry Men	1931	80	7.6	Akira Kurosawa	Ajay Devgn	27712	6013.0	47.0
Drama	1917	1925	64	7.6	Aamir Khan	Abhay Deol	25088	3600.0	28.0
Family	E.T. the Extra-Terrestrial	1971	100	7.8	Mel Stuart	Gene Wilder	178731	4000000.0	67.0
Fantasy	Das Cabinet des Dr. Caligari	1920	76	7.9	F.W. Murnau	Max Schreck	57428	337574718.0	NaN
Film-Noir	Shadow of a Doubt	1941	100	7.8	Alfred Hitchcock	Humphrey Bogart	59556	449191.0	94.0
Horror	Alien	1933	71	7.6	Alejandro Amenábar	Anthony Perkins	27007	89029.0	46.0
Mystery	Dark City	1938	96	7.6	Alex Proyas	Bernard-Pierre Donnadieu	33982	1035953.0	52.0
Thriller	Wait Until Dark	1967	108	7.8	Terence Young	Audrey Hepburn	27733	17550741.0	81.0
Western	Il buono, il brutto, il cattivo	1965	132	7.8	Clint Eastwood	Clint Eastwood	65659	5321508.0	69.0

In [206]: `# find the top 3 genres by total earning`

```
movies.groupby('Genre').sum()['Gross'].sort_values(ascending=False).head(3)
```

Out[206]:

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

In [212]: `#Second Approach`

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

Out[212]:

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

In [214]: `movies.groupby('Genre')['Gross'].sum().sort_values(ascending=False).head(3)`

Out[214]:

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

In [217]: `# find the genre with highest avg IMDB rating`

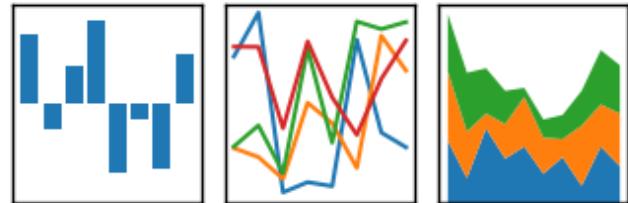
```
movies.groupby('Genre')['IMDB_Rating'].mean().sort_values(ascending=False).head(1)
```

Out[217]:

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

Pandas groupby

groupby in pandas is a function that lets you group data in a DataFrame based on specific criteria, and then apply aggregate functions to each group. It's a powerful tool for data analysis that allows you to quickly and easily calculate summary statistics for your data.



```
In [1]: import numpy as np
import pandas as pd
```

Groupby (Applies on Categorical data)

```
In [2]: movies = pd.read_csv("imdb-top-1000.csv")
movies.head(3)
```

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	Director	Star1	No_of_Votes
0	The Shawshank Redemption	1994	142	Drama	9.3	Frank Darabont	Tim Robbins	2343110
1	The Godfather	1972	175	Crime	9.2	Francis Ford Coppola	Marlon Brando	1620367
2	The Dark Knight	2008	152	Action	9.0	Christopher Nolan	Christian Bale	2303232

```
In [3]: movies.groupby('Genre')
```

```
Out[3]: <pandas.core.groupby.generic.DataFrameGroupBy object at 0x0000024A502D5550>
```

In [4]: `genres = movies.groupby('Genre')`

In [5]: `# Applying builtin aggregation functions on groupby objects`
`genres.sum().head(2)`

Out[5]:

Genre	Runtime	IMDB_Rating	No_of_Votes	Gross	Metascore
Action	22196	1367.3	72282412	3.263226e+10	10499.0
Adventure	9656	571.5	22576163	9.496922e+09	5020.0

In [6]: `#Second Approach`
`movies.groupby('Genre')['Gross'].sum().sort_values()`

Out[6]:

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

In [7]: `# find the genre with highest avg IMDB rating`
`movies.groupby('Genre')['IMDB_Rating'].mean().sort_values(ascending=False).head(1)`

Out[7]:

Genre	IMDB_Rating
Western	8.35
Name: IMDB_Rating, dtype: float64	

In [8]: `# find director with most popularity`
`movies.groupby('Director')['No_of_Votes'].sum().sort_values(ascending=False).head(1)`

Out[8]:

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

In [9]: `# find the highest rated movie of each genre
movies.head(1)`

Out[9]:

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	Director	Star1	No_of_Votes
0	The Shawshank Redemption	1994	142	Drama	9.3	Frank Darabont	Tim Robbins	2343110

In [10]: `# find number of movies done by each actor
#movies['Star1'].value_counts() # first method
movies.groupby('Star1')['Series_Title'].count().sort_values(ascending=False)`

Out[10]: 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

GroupBy Attributes and Methods

In [12]: `# find total number of groups -> len
len(movies.groupby('Genre')) # 14 different groups`

Out[12]: 14

In [13]: `movies['Genre'].nunique() # second method`

Out[13]: 14

```
In [14]: # find items in each group -> size  
movies.groupby('Genre').size() # index based
```

```
Out[14]: 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
```

```
In [15]: movies['Genre'].value_counts() # second method
```

```
Out[15]: 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: Genre, dtype: int64
```

```
In [18]: # first()/Last() -> nth item
#movies.groupby('Genre').first()
#movies.groupby('Genre').last()
movies.groupby('Genre').nth(6) ---> #gives 7th movie
```

Out[18]:

	Series_Title	Released_Year	Runtime	IMDB_Rating	Director	Star1	No_of_Votes
Genre							
Action	Star Wars: Episode V - The Empire Strikes Back	1980	124	8.7	Irvin Kershner	Mark Hamill	115931
Adventure	North by Northwest	1959	136	8.3	Alfred Hitchcock	Cary Grant	29919
Animation	WALL·E	2008	98	8.4	Andrew Stanton	Ben Burtt	99979
Biography	Braveheart	1995	178	8.3	Mel Gibson	Mel Gibson	95918
Comedy	The Great Dictator	1940	125	8.4	Charles Chaplin	Charles Chaplin	20315
Crime	Se7en	1995	127	8.6	David Fincher	Morgan Freeman	144509
Drama	It's a Wonderful Life	1946	130	8.6	Frank Capra	James Stewart	40580
Horror	Get Out	2017	104	7.7	Jordan Peele	Daniel Kaluuya	49285
Mystery	Sleuth	1972	138	8.0	Joseph L. Mankiewicz	Laurence Olivier	4474



In [19]: # get_group -> vs filtering

movies.groupby('Genre').get_group('Horror')

Out[19]:

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	Director	Star1	No_of_Votes
49	Psycho	1960	109	Horror	8.5	Alfred Hitchcock	Anthony Perkins	6042
75	Alien	1979	117	Horror	8.4	Ridley Scott	Sigourney Weaver	7878
271	The Thing	1982	109	Horror	8.1	John Carpenter	Kurt Russell	3712
419	The Exorcist	1973	122	Horror	8.0	William Friedkin	Ellen Burstyn	3623
544	Night of the Living Dead	1968	96	Horror	7.9	George A. Romero	Duane Jones	1165
707	The Innocents	1961	100	Horror	7.8	Jack Clayton	Deborah Kerr	270
724	Get Out	2017	104	Horror	7.7	Jordan Peele	Daniel Kaluuya	4928
844	Halloween	1978	91	Horror	7.7	John Carpenter	Donald Pleasence	2331
876	The Invisible Man	1933	71	Horror	7.7	James Whale	Claude Rains	306
932	Saw	2004	103	Horror	7.6	James Wan	Cary Elwes	3790
948	The Others	2001	101	Horror	7.6	Alejandro Amenábar	Nicole Kidman	3376

< >

In [20]: movies.groupby('Genre').get_group('Fantasy')

Out[20]:

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	Director	Star1	No_of_Votes
321	Das Cabinet des Dr. Caligari	1920	76	Fantasy	8.1	Robert Wiene	Werner Krauss	57428
568	Nosferatu	1922	94	Fantasy	7.9	F.W. Murnau	Max Schreck	88794

< >

In [21]: movies[movies['Genre']=='Fantasy'] #Second Method

Out[21]:

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	Director	Star1	No_of_Votes
321	Das Cabinet des Dr. Caligari	1920	76	Fantasy	8.1	Robert Wiene	Werner Krauss	57428
568	Nosferatu	1922	94	Fantasy	7.9	F.W. Murnau	Max Schreck	88794

< >

In [24]: # groups

```
movies.groupby('Genre').groups ---> #Dictionary= gives index position
```

Out[24]: {'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]}

In [25]:

```
# describe
movies.groupby('Genre').describe()
```

Out[25]:

										Runtime	IMDB_Rating	.
	Genre	count	mean	std	min	25%	50%	75%	max	count	mean	.
Action	Action	172.0	129.046512	28.500706	45.0	110.75	127.5	143.25	321.0	172.0	7.949419	.
Adventure	Adventure	72.0	134.111111	33.317320	88.0	109.00	127.0	149.00	228.0	72.0	7.937500	.
Animation	Animation	82.0	99.585366	14.530471	71.0	90.00	99.5	106.75	137.0	82.0	7.930488	.
Biography	Biography	88.0	136.022727	25.514466	93.0	120.00	129.0	146.25	209.0	88.0	7.938636	.
Comedy	Comedy	155.0	112.129032	22.946213	68.0	96.00	106.0	124.50	188.0	155.0	7.901290	.
Crime	Crime	107.0	126.392523	27.689231	80.0	106.50	122.0	141.50	229.0	107.0	8.016822	.
Drama	Drama	289.0	124.737024	27.740490	64.0	105.00	121.0	137.00	242.0	289.0	7.957439	.
Family	Family	2.0	107.500000	10.606602	100.0	103.75	107.5	111.25	115.0	2.0	7.800000	.
Fantasy	Fantasy	2.0	85.000000	12.727922	76.0	80.50	85.0	89.50	94.0	2.0	8.000000	.
Film-Noir	Film-Noir	3.0	104.000000	4.000000	100.0	102.00	104.0	106.00	108.0	3.0	7.966667	.
Horror	Horror	11.0	102.090909	13.604812	71.0	98.00	103.0	109.00	122.0	11.0	7.909091	.
Mystery	Mystery	12.0	119.083333	14.475423	96.0	110.75	117.5	130.25	138.0	12.0	7.975000	.
Thriller	Thriller	1.0	108.000000	NaN	108.0	108.00	108.0	108.00	108.0	1.0	7.800000	.
Western	Western	4.0	148.250000	17.153717	132.0	134.25	148.0	162.00	165.0	4.0	8.350000	.

14 rows × 40 columns



In [30]:

```
# sample
movies.groupby('Genre').sample(2,replace=True)
```

Out[30]:

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	Director	Star1	No_1
308	White Heat	1949	114	Action	8.1	Raoul Walsh	James Cagney	
171	Die Hard	1988	132	Action	8.2	John McTiernan	Bruce Willis	
300	Ben-Hur	1959	212	Adventure	8.1	William Wyler	Charlton Heston	
47	Back to the Future	1985	116	Adventure	8.5	Robert Zemeckis	Michael J. Fox	
640	Les triplettes de Belleville	2003	80	Animation	7.8	Sylvain Chomet	Michèle Caucheteux	
799	South Park: Bigger, Longer & Uncut	1999	81	Animation	7.7	Trey Parker	Trey Parker	
632	The World's Fastest Indian	2005	127	Biography	7.8	Roger Donaldson	Anthony Hopkins	
372	Mar adentro	2014	126	Biography	8.0	Alejandro Amenábar	Javier Bardem	
846	Love and Death	1975	85	Comedy	7.7	Woody Allen	Woody Allen	
660	The Sandlot	1993	101	Comedy	7.8	David Mickey Evans	Tom Guiry	
397	Bound by Honor	1993	180	Crime	8.0	Taylor Hackford	Damian Chapa	
108	Scarface	1983	170	Crime	8.3	Brian De Palma	Al Pacino	
53	Capharnaüm	2018	126	Drama	8.4	Nadine Labaki	Zain Al Rafeea	
894	Creed	2015	133	Drama	7.6	Ryan Coogler	Michael B. Jordan	
688	E.T. the Extra-Terrestrial	1982	115	Family	7.8	Steven Spielberg	Henry Thomas	
688	E.T. the Extra-Terrestrial	1982	115	Family	7.8	Steven Spielberg	Henry Thomas	
568	Nosferatu	1922	94	Fantasy	7.9	F.W. Murnau	Max Schreck	
321	Das Cabinet des Dr. Caligari	1920	76	Fantasy	8.1	Robert Wiene	Werner Krauss	
309	The Third Man	1949	104	Film-Noir	8.1	Carol Reed	Orson Welles	
712	Shadow of a Doubt	1943	108	Film-Noir	7.8	Alfred Hitchcock	Teresa Wright	
49	Psycho	1960	109	Horror	8.5	Alfred Hitchcock	Anthony Perkins	

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	Director	Star1	No_of_Votes
948	The Others	2001	101	Horror	7.6	Alejandro Amenábar	Nicole Kidman	10
81	Rear Window	1954	112	Mystery	8.4	Alfred Hitchcock	James Stewart	10
959	Dark City	1998	100	Mystery	7.6	Alex Proyas	Rufus Sewell	10
700	Wait Until Dark	1967	108	Thriller	7.8	Terence Young	Audrey Hepburn	10
700	Wait Until Dark	1967	108	Thriller	7.8	Terence Young	Audrey Hepburn	10
691	The Outlaw Josey Wales	1976	135	Western	7.8	Clint Eastwood	Clint Eastwood	10
48	Once Upon a Time in the West	1968	165	Western	8.5	Sergio Leone	Henry Fonda	10

In [31]:

```
# nunique
movies.groupby('Genre').nunique() # unique --> unique items , nunique--> gives
```

Out[31]:

Genre	Series_Title	Released_Year	Runtime	IMDB_Rating	Director	Star1	No_of_Votes	Gr
Action	172	61	78	15	123	121	172	172
Adventure	72	49	58	10	59	59	72	72
Animation	82	35	41	11	51	77	82	82
Biography	88	44	56	13	76	72	88	88
Comedy	155	72	70	11	113	133	155	155
Crime	106	56	65	14	86	85	107	107
Drama	289	83	95	14	211	250	288	288
Family	2	2	2	1	2	2	2	2
Fantasy	2	2	2	2	2	2	2	2
Film-Noir	3	3	3	3	3	3	3	3
Horror	11	11	10	8	10	11	11	11
Mystery	12	11	10	8	10	11	12	12
Thriller	1	1	1	1	1	1	1	1
Western	4	4	4	4	2	2	4	4

aggregate method

In [33]:

```
# passing dict
movies.groupby('Genre').agg(
{
    'Runtime':'mean',
    'IMDB_Rating':'mean',
    'No_of_Votes':'sum',
    'Gross':'sum',
    'Metascore':'min'
}
)
```

Out[33]:

Genre	Runtime	IMDB_Rating	No_of_Votes	Gross	Metascore
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

In [37]: #Passsing List

```
movies.groupby('Genre').agg(['min', 'max', 'mean', 'sum', 'median'])
```

Out[37]:

Genre	Runtime						IMDB_Rating			...		
	min	max	mean	sum	median	min	max	mean	sum	median	...	
Action	45	321	129.046512	22196	127.5	7.6	9.0	7.949419	1367.3	7.9	...	
Adventure	88	228	134.111111	9656	127.0	7.6	8.6	7.937500	571.5	7.9	...	
Animation	71	137	99.585366	8166	99.5	7.6	8.6	7.930488	650.3	7.9	...	
Biography	93	209	136.022727	11970	129.0	7.6	8.9	7.938636	698.6	7.9	...	
Comedy	68	188	112.129032	17380	106.0	7.6	8.6	7.901290	1224.7	7.9	...	
Crime	80	229	126.392523	13524	122.0	7.6	9.2	8.016822	857.8	8.0	...	
Drama	64	242	124.737024	36049	121.0	7.6	9.3	7.957439	2299.7	8.0	...	
Family	100	115	107.500000	215	107.5	7.8	7.8	7.800000	15.6	7.8	...	40
Fantasy	76	94	85.000000	170	85.0	7.9	8.1	8.000000	16.0	8.0	...	337
Film-Noir	100	108	104.000000	312	104.0	7.8	8.1	7.966667	23.9	8.0	...	1
Horror	71	122	102.090909	1123	103.0	7.6	8.5	7.909091	87.0	7.8	...	
Mystery	96	138	119.083333	1429	117.5	7.6	8.4	7.975000	95.7	8.0	...	10
Thriller	108	108	108.000000	108	108.0	7.8	7.8	7.800000	7.8	7.8	...	17
Western	132	165	148.250000	593	148.0	7.8	8.8	8.350000	33.4	8.4	...	5

14 rows × 25 columns



In [39]: # Adding both the syntax

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

Out[39]:

Genre	Runtime		IMDB_Rating		No_of_Votes		Gross		Metascore	
	min	mean	mean	sum	max	sum	sum	min		
Action	45	129.046512	7.949419	72282412	2303232	3.263226e+10		33.0		
Adventure	88	134.111111	7.937500	22576163	1512360	9.496922e+09		41.0		
Animation	71	99.585366	7.930488	21978630	999790	1.463147e+10		61.0		
Biography	93	136.022727	7.938636	24006844	1213505	8.276358e+09		48.0		
Comedy	68	112.129032	7.901290	27620327	939631	1.566387e+10		45.0		
Crime	80	126.392523	8.016822	33533615	1826188	8.452632e+09		47.0		
Drama	64	124.737024	7.957439	61367304	2343110	3.540997e+10		28.0		
Family	100	107.500000	7.800000	551221	372490	4.391106e+08		67.0		
Fantasy	76	85.000000	8.000000	146222	88794	7.827267e+08		NaN		
Film-Noir	100	104.000000	7.966667	367215	158731	1.259105e+08		94.0		
Horror	71	102.090909	7.909091	3742556	787806	1.034649e+09		46.0		
Mystery	96	119.083333	7.975000	4203004	1129894	1.256417e+09		52.0		
Thriller	108	108.000000	7.800000	27733	27733	1.755074e+07		81.0		
Western	132	148.250000	8.350000	1289665	688390	5.822151e+07		69.0		

In [40]: # Looping on groups

```
for group , data in movies.groupby('Genre'):  
    print(data)
```

ime \	Series_Title	Released_Year	Runt
2	The Dark Knight	2008	
152	The Lord of the Rings: The Return of the King	2003	
201	Inception	2010	
8			
148	The Lord of the Rings: The Fellowship of the Ring	2001	
178	The Lord of the Rings: The Two Towers	2002	
13			
179			
..
...			
968	Falling Down	1993	
113	Lethal Weapon	1987	
979			
109			
982	Mad Max 2	1981	
..			

In [41]: # find the highest rated movie of each genre

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

for group, data in movies.groupby('Genre'):
    df.append(data[data['IMDB_Rating'] == data['IMDB_Rating'].max()])
df
```

Out[41]:

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	Director	Star1
2	The Dark Knight	2008	152	Action	9.0	Christopher Nolan	Christian Bale
21	Interstellar	2014	169	Adventure	8.6	Christopher Nolan	Matthew McConaughey
23	Sen to Chihiro no kamikakushi	2001	125	Animation	8.6	Hayao Miyazaki	Daveigh Chase
7	Schindler's List	1993	195	Biography	8.9	Steven Spielberg	Liam Neeson
19	Gisaengchung	2019	132	Comedy	8.6	Bong Joon Ho	Kang-ho Song
26	La vita è bella	1997	116	Comedy	8.6	Roberto Benigni	Roberto Benigni
1	The Godfather	1972	175	Crime	9.2	Francis Ford Coppola	Marlon Brando
0	The Shawshank Redemption	1994	142	Drama	9.3	Frank Darabont	Tim Robbins
688	E.T. the Extra-Terrestrial	1982	115	Family	7.8	Steven Spielberg	Henry Thomas
698	Willy Wonka & the Chocolate Factory	1971	100	Family	7.8	Mel Stuart	Gene Wilder
321	Das Cabinet des Dr. Caligari	1920	76	Fantasy	8.1	Robert Wiene	Werner Krauss
309	The Third Man	1949	104	Film-Noir	8.1	Carol Reed	Orson Welles
49	Psycho	1960	109	Horror	8.5	Alfred Hitchcock	Anthony Perkins
69	Memento	2000	113	Mystery	8.4	Christopher Nolan	Guy Pearce
81	Rear Window	1954	112	Mystery	8.4	Alfred Hitchcock	James Stewart
700	Wait Until Dark	1967	108	Thriller	7.8	Terence Young	Audrey Hepburn
12	Il buono, il brutto, il cattivo	1966	161	Western	8.8	Sergio Leone	Clint Eastwood



split (apply) combine

In [42]: `# apply -> builtin function
movies.groupby('Genre').apply(min)`

Out[42]:

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	Director	Star1
Genre							
Action	300	1924	45	Action	7.6	Abhishek Chabey	Aamir Khan
Adventure	2001: A Space Odyssey	1925	88	Adventure	7.6	Akira Kurosawa	Aamir Khan
Animation	Akira	1940	71	Animation	7.6	Adam Elliot	Adrian Molina
Biography	12 Years a Slave	1928	93	Biography	7.6	Adam McKay	Adrien Brody
Comedy	(500) Days of Summer	1921	68	Comedy	7.6	Alejandro G. Iñárritu	Aamir Khan
Crime	12 Angry Men	1931	80	Crime	7.6	Akira Kurosawa	Ajay Devgn
Drama	1917	1925	64	Drama	7.6	Aamir Khan	Abhay Deol
Family	E.T. the Extra-Terrestrial	1971	100	Family	7.8	Mel Stuart	Gene Wilder
Fantasy	Das Cabinet des Dr. Caligari	1920	76	Fantasy	7.9	F.W. Murnau	Max Schreck
Film-Noir	Shadow of a Doubt	1941	100	Film-Noir	7.8	Alfred Hitchcock	Humphrey Bogart
Horror	Alien	1933	71	Horror	7.6	Alejandro Amenábar	Anthony Perkins
Mystery	Dark City	1938	96	Mystery	7.6	Alex Proyas	Bernard-Pierre Donnadieu
Thriller	Wait Until Dark	1967	108	Thriller	7.8	Terence Young	Audrey Hepburn
Western	Il buono, il brutto, il cattivo	1965	132	Western	7.8	Clint Eastwood	Clint Eastwood

```
In [43]: # find number of movies starting with A for each group
def foo(group):
    print(group) # type = Dataframe
    return group
```

```
In [44]: movies.groupby('Genre').apply(foo)
```

		Series_Title	Released_Year	Runt
ime \				
2		The Dark Knight	2008	
152		The Lord of the Rings: The Return of the King	2003	
201		Inception	2010	
8				
148		The Lord of the Rings: The Fellowship of the Ring	2001	
178		The Lord of the Rings: The Two Towers	2002	
13				
179				
..		
...				
968		Falling Down	1993	
113				
979		Lethal Weapon	1987	
109				

```
In [49]: #Custom Logic
def foo(group):
    return group['Series_Title'].str.startswith('A').sum()
```

```
In [50]: movies.groupby('Genre').apply(foo)
```

```
Out[50]: 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
```

```
In [51]: # 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
```

```
In [52]: movies.groupby('Genre').apply(rank_movie)
```

Out[52]:

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	Director	Star1	No_of_
0	The Shawshank Redemption	1994	142	Drama	9.3	Frank Darabont	Tim Robbins	23
1	The Godfather	1972	175	Crime	9.2	Francis Ford Coppola	Marlon Brando	16
2	The Dark Knight	2008	152	Action	9.0	Christopher Nolan	Christian Bale	23
3	The Godfather: Part II	1974	202	Crime	9.0	Francis Ford Coppola	Al Pacino	11
4	12 Angry Men	1957	96	Crime	9.0	Sidney Lumet	Henry Fonda	6
...
995	Breakfast at Tiffany's	1961	115	Comedy	7.6	Blake Edwards	Audrey Hepburn	16
996	Giant	1956	201	Drama	7.6	George Stevens	Elizabeth Taylor	1
997	From Here to Eternity	1953	118	Drama	7.6	Fred Zinnemann	Burt Lancaster	1
998	Lifeboat	1944	97	Drama	7.6	Alfred Hitchcock	Tallulah Bankhead	1
999	The 39 Steps	1935	86	Crime	7.6	Alfred Hitchcock	Robert Donat	1

1000 rows × 11 columns



```
In [55]: # find normalized IMDB rating group wise
#x_normalized = (x - x minimum) / (x maximum - x minimum)
def normal(group):
    group['normal_rating'] = (group['IMDB_Rating'] - group['IMDB_Rating'].min())
    return group

movies.groupby('Genre').apply(normal)
```

Out[55]:

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	Director	Star1	No_of_
0	The Shawshank Redemption	1994	142	Drama	9.3	Frank Darabont	Tim Robbins	234
1	The Godfather	1972	175	Crime	9.2	Francis Ford Coppola	Marlon Brando	162
2	The Dark Knight	2008	152	Action	9.0	Christopher Nolan	Christian Bale	230
3	The Godfather: Part II	1974	202	Crime	9.0	Francis Ford Coppola	Al Pacino	112
4	12 Angry Men	1957	96	Crime	9.0	Sidney Lumet	Henry Fonda	68
...
995	Breakfast at Tiffany's	1961	115	Comedy	7.6	Blake Edwards	Audrey Hepburn	16
996	Giant	1956	201	Drama	7.6	George Stevens	Elizabeth Taylor	15
997	From Here to Eternity	1953	118	Drama	7.6	Fred Zinnemann	Burt Lancaster	14
998	Lifeboat	1944	97	Drama	7.6	Alfred Hitchcock	Tallulah Bankhead	12
999	The 39 Steps	1935	86	Crime	7.6	Alfred Hitchcock	Robert Donat	11

1000 rows × 11 columns

groupby on multiple cols

```
In [57]: combo = movies.groupby(['Director', 'Star1'])
#size
combo.size()
```

```
Out[57]: 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
```

```
In [59]: #if we want particular combo---> get_group
combo.get_group(('Aamir Khan', 'Amole Gupte'))
```

	Series_Title	Released_Year	Runtime	Genre	IMDB_Rating	Director	Star1	No_of_Votes
65	Taare Zameen Par	2007	165	Drama	8.4	Aamir Khan	Amole Gupte	168895

```
In [62]: # find the most earning actor-> director combo
combo['Gross'].sum().sort_values(ascending=False).head(2)
```

```
Out[62]: Director      Star1
Akira Kurosawa Toshirô Mifune    2.999877e+09
Anthony Russo   Joe Russo        2.205039e+09
Name: Gross, dtype: float64
```

```
In [75]: # find the best(in-terms of metascore(avg)) actor->genre combo
movies.groupby(['Star1', 'Genre'])['Metascore'].mean().reset_index().sort_value
```

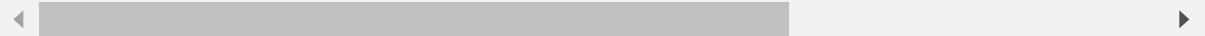
	Star1	Genre	Metascore
230	Ellar Coltrane	Drama	100.0

In [77]: # agg on multiple groupby
combo.agg(['min', 'max', 'mean'])

Out[77]:

Director	Star1	Runtime			IMDB_Rating			No_of_Votes			n
		min	max	mean	min	max	mean	min	max	mean	
Aamir Khan	Amole Gupte	165	165	165.0	8.4	8.4	8.4	168895	168895	168895.0	122386!
Aaron Sorkin	Eddie Redmayne	129	129	129.0	7.8	7.8	7.8	89896	89896	89896.0	85309041!
Abdellatif Kechiche	Léa Seydoux	180	180	180.0	7.7	7.7	7.7	138741	138741	138741.0	219967!
Abhishek Chaubey	Shahid Kapoor	148	148	148.0	7.8	7.8	7.8	27175	27175	27175.0	21842830!
Abhishek Kapoor	Amit Sadh	130	130	130.0	7.7	7.7	7.7	32628	32628	32628.0	112252!
...
Zaza Urushadze	Lembit Ulfsak	87	87	87.0	8.2	8.2	8.2	40382	40382	40382.0	14450!
Zoya Akhtar	Hrithik Roshan	155	155	155.0	8.1	8.1	8.1	67927	67927	67927.0	310848!
	Vijay Varma	154	154	154.0	8.0	8.0	8.0	31886	31886	31886.0	556653!
Çagan Irmak	Çetin Tekindor	112	112	112.0	8.3	8.3	8.3	78925	78925	78925.0	46185536!
Ömer Faruk Sorak	Cem Yilmaz	127	127	127.0	8.0	8.0	8.0	56960	56960	56960.0	19620607!

898 rows × 15 columns



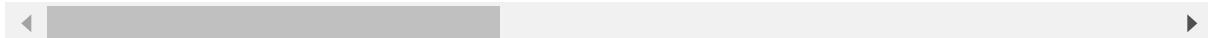
Excercise

```
In [78]: ipl = pd.read_csv("deliveries.csv")
ipl.head(2)
```

Out[78]:

	match_id	inning	batting_team	bowling_team	over	ball	batsman	non_striker	bowler	is_s
0	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	1	DA Warner	S Dhawan	TS Mills	
1	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	2	DA Warner	S Dhawan	TS Mills	

2 rows × 21 columns



```
In [80]: ipl.shape # ball by ball dataset
```

Out[80]: (179078, 21)

```
In [87]: # find the top 10 batsman in terms of runs
ipl.groupby('batsman')['batsman_runs'].sum().sort_values(ascending=False).rese
```

Out[87]:

	batsman	batsman_runs
0	V Kohli	5434
1	SK Raina	5415
2	RG Sharma	4914
3	DA Warner	4741
4	S Dhawan	4632
5	CH Gayle	4560
6	MS Dhoni	4477
7	RV Uthappa	4446
8	AB de Villiers	4428
9	G Gambhir	4223

```
In [94]: # find the batsman with max no of sixes
six = ipl[ipl['batsman_runs']==6]

six.groupby('batsman')['batsman'].count().sort_values(ascending=False).head(2)
```

Out[94]:

batsman	
CH Gayle	327
AB de Villiers	214
Name: batsman, dtype: int64	

In [105]: # find batsman with most number of 4's and 6's in last 5 overs

```
temp = ipl[ipl['over'] > 15]
temp = temp[(temp['batsman_runs'] == 4) | (temp['batsman_runs'] == 6)]

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

Out[105]: 'MS Dhoni'

In [110]: # find V Kohli's record against all teams

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

temp.groupby('bowling_team')['batsman_runs'].sum().sort_values(ascending=False)
```

Out[110]:

	bowling_team	batsman_runs
0	Delhi Daredevils	763
1	Chennai Super Kings	749

In [118]: # Create a function that can return the highest score of any batsman

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

temp.groupby('match_id')['batsman_runs'].sum().sort_values(ascending=False).head(1)
```

Out[118]: 113

In [122]: def highest(batsman):

```
    temp = ipl[ipl['batsman'] == batsman]
    return temp.groupby('match_id')['batsman_runs'].sum().sort_values(ascending=False).head(1)
```

In [123]: highest('DA Warner')

Out[123]: 126

In []:

```
In [1]: import pandas as pd
import numpy as np
```

```
In [2]: courses = pd.read_csv("courses.csv")
students = pd.read_csv("students.csv")
may = pd.read_csv("reg-month1.csv")
june = pd.read_csv("reg-month2.csv")
matches = pd.read_csv("matches.csv")
deliveries = pd.read_csv("deliveries.csv")
```

```
In [3]: courses.head(2)
```

Out[3]:

	course_id	course_name	price
0	1	python	2499
1	2	sql	3499

```
In [4]: students.head(2)
```

Out[4]:

	student_id	name	partner
0	1	Kailash Harjo	23
1	2	Esha Butala	1

```
In [5]: may.head(2)
```

Out[5]:

	student_id	course_id
0	23	1
1	15	5

```
In [6]: june.head(2)
```

Out[6]:

	student_id	course_id
0	3	5
1	16	7

```
In [7]: matches.head(2)
```

Out[7]:

	id	season	city	date	team1	team2	toss_winner	toss_decision	result	dl_applied	winner	win_by_runs	win_by_wickets	player_of_match
0	1	2017	Hyderabad	2017-04-05	Sunrisers Hyderabad	Royal Challengers Bangalore	Royal Challengers Bangalore	field	normal	0	Sunrisers Hyderabad	35	0	Yuvraj Singh
1	2	2017	Pune	2017-04-06	Mumbai Indians	Rising Pune Supergiant	Rising Pune Supergiant	field	normal	0	Rising Pune Supergiant	0	7	SPD Singh

Concat

It is a powerful function that allows you to concatenate two or more DataFrames along a particular axis (row-wise or column-wise). You can control how the data is concatenated by specifying several parameters, such as axis, join, ignore_index, and keys.

```
In [8]: regs = pd.concat([may,june],ignore_index=True) # Vertically merged  
regs
```

Out[8]:

	student_id	course_id
0	23	1
1	15	5
2	18	6
3	23	4
4	16	9
5	18	1
6	1	1
7	7	8
8	22	3
9	15	1
10	19	4
11	1	6
12	7	10
13	11	7
14	13	3
15	24	4
16	21	1
17	16	5
18	23	3
19	17	7
20	23	6
21	25	1
22	19	2
23	25	10
24	3	3
25	3	5
26	16	7
27	12	10
28	12	1
29	14	9
30	7	7
31	7	2
32	16	3
33	17	10
34	11	8
35	14	6
36	12	5
37	12	7
38	18	8
39	1	10
40	1	9
41	2	5
42	7	6
43	22	5
44	22	6
45	23	9
46	23	5
47	14	4
48	14	1
49	11	10
50	42	9
51	50	8
52	38	1

In [9]: # Multi_index DataFrame

```
multi = pd.concat([may,june],keys=['may','june'])  
multi
```

Out[9]:

		student_id	course_id
may	0	23	1
	1	15	5
	2	18	6
	3	23	4
	4	16	9
	5	18	1
	6	1	1
	7	7	8
	8	22	3
	9	15	1
	10	19	4
	11	1	6
	12	7	10
	13	11	7
	14	13	3
	15	24	4
	16	21	1
	17	16	5
	18	23	3
	19	17	7
	20	23	6
	21	25	1
	22	19	2
	23	25	10
	24	3	3
june	0	3	5
	1	16	7
	2	12	10
	3	12	1
	4	14	9
	5	7	7
	6	7	2
	7	16	3
	8	17	10
	9	11	8
	10	14	6
	11	12	5
	12	12	7
	13	18	8
	14	1	10
	15	1	9
	16	2	5
	17	7	6
	18	22	5
	19	22	6
	20	23	9
	21	23	5
	22	14	4
	23	14	1
	24	11	10
	25	42	9
	26	50	8
	27	38	1

```
In [10]: multi.loc['may']
```

Out[10]:

	student_id	course_id
0	23	1
1	15	5
2	18	6
3	23	4
4	16	9
5	18	1
6	1	1
7	7	8
8	22	3
9	15	1
10	19	4
11	1	6
12	7	10
13	11	7
14	13	3
15	24	4
16	21	1
17	16	5
18	23	3
19	17	7
20	23	6
21	25	1
22	19	2
23	25	10
24	3	3

```
In [11]: multi.loc['june', 0]
```

Out[11]: student_id 3

course_id 5

Name: (june, 0), dtype: int64

In [12]: # Horizontally placed

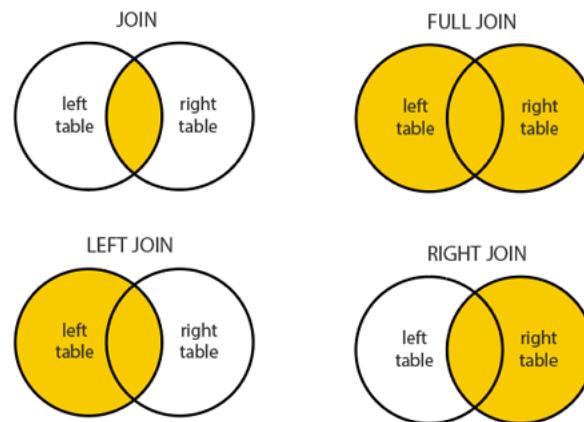
pd.concat([may,june],axis=1)

Out[12]:

	student_id	course_id	student_id	course_id
0	23.0	1.0	3	5
1	15.0	5.0	16	7
2	18.0	6.0	12	10
3	23.0	4.0	12	1
4	16.0	9.0	14	9
5	18.0	1.0	7	7
6	1.0	1.0	7	2
7	7.0	8.0	16	3
8	22.0	3.0	17	10
9	15.0	1.0	11	8
10	19.0	4.0	14	6
11	1.0	6.0	12	5
12	7.0	10.0	12	7
13	11.0	7.0	18	8
14	13.0	3.0	1	10
15	24.0	4.0	1	9
16	21.0	1.0	2	5
17	16.0	5.0	7	6
18	23.0	3.0	22	5
19	17.0	7.0	22	6
20	23.0	6.0	23	9
21	25.0	1.0	23	5
22	19.0	2.0	14	4
23	25.0	10.0	14	1
24	3.0	3.0	11	10
25	NaN	NaN	42	9
26	NaN	NaN	50	8
27	NaN	NaN	38	1

Merge

On Joins



Inner Join

For joining any data ,

In each set of data, there should be a "common" column. Students[student_id] and regs[student_id] are listed here. We combine based on the student_id, however the inner join only displays the data that is "Common" across the two dataframes.

In [13]: `students.merge(regs, how= 'inner' , on = 'student_id').tail()`

Out[13]:

	student_id	name	partner	course_id
45	23	Chhavi Lachman	18	9
46	23	Chhavi Lachman	18	5
47	24	Radhika Suri	17	4
48	25	Shashank D'Alia	2	1
49	25	Shashank D'Alia	2	10

Left Join

Here we have same column --- > course_id

on basis on this we can merge using left join.

Regardless of whether or not the right side data leaves, it prints all of the left side data. so , we can see left data (Numpy , c++) but we cannot see any right side data which is student_id here, courses reflect = Left and regs reflect = right

In [14]:

`courses.merge(regs,how='left',on='course_id').tail(5)`

Out[14]:

	course_id	course_name	price	student_id
50	10	pyspark	2499	17.0
51	10	pyspark	2499	1.0
52	10	pyspark	2499	11.0
53	11	Numpy	699	NaN
54	12	C++	1299	NaN

Right join

In [15]:

```
temp_df = pd.DataFrame({
    'student_id':[26,27,28],
    'name':['Nitish','Ankit','Rahul'],
    'partner':[28,26,17]
})
students = pd.concat([students,temp_df],ignore_index=True)
```

In [16]:

`students.tail()`

Out[16]:

	student_id	name	partner
23	24	Radhika Suri	17
24	25	Shashank D'Alia	2
25	26	Nitish	28
26	27	Ankit	26
27	28	Rahul	17

Regs data(50,51,52) in the current case does not contain students data, however even this, data is printed since the join was done right.

why?

because when using a right join, all right side data is printed regardless of whether the left side data exists or not.

here right reflects = regs , Left reflects = students

In [17]: `students.merge(regs, how='right', on='student_id').tail(5)`

Out[17]:

	student_id	name	partner	course_id
48	14	Pranab Natarajan	22.0	1
49	11	David Mukhopadhyay	20.0	10
50	42		Nan	9
51	50		Nan	8
52	38		Nan	1

Since there is no course_id in the student data in the current case, "Nan" data is displayed.

Why was a left join performed using the student_id? Regardless of whether or not the right side data leaves, it prints all of the left side data.

here Left reflects = students , right reflects = regs

In [18]: `students.merge(regs, how='left', on='student_id').tail(5)`

Out[18]:

	student_id	name	partner	course_id
55	25	Shashank D'Alia	2	1.0
56	25	Shashank D'Alia	2	10.0
57	26	Nitish	28	Nan
58	27	Ankit	26	Nan
59	28	Rahul	17	Nan

Outer join

Initially the left join data is clearly apparent with (Nitish, Ankit, Rahul) data written,

but the right side data (course id) is blank. like which,

Right join shows Nan even though we don't have any data for (42, 50, 38), but we can see the course's id column because it's a right join.

Finally, we may view both data sets, both common and individual, regardless of whether they have ever been. As seen in the outer join

In [19]: `students.merge(regs, how='outer', on='student_id').tail(10)`

Out[19]:

	student_id	name	partner	course_id
53	23	Chhavi Lachman	18.0	5.0
54	24	Radhika Suri	17.0	4.0
55	25	Shashank D'Alia	2.0	1.0
56	25	Shashank D'Alia	2.0	10.0
57	26	Nitish	28.0	Nan
58	27	Ankit	26.0	Nan
59	28	Rahul	17.0	Nan
60	42		Nan	9.0
61	50		Nan	8.0
62	38		Nan	1.0

In [20]: `# 1. find total revenue generated
regs.merge(courses, how='inner', on='course_id')['price'].sum()`

Out[20]: 154247

In [27]: `# 2. find month by month revenue
temp = pd.concat([may,june], keys=['may','june']).reset_index()
temp.merge(courses, on='course_id').groupby('level_0')['price'].sum()`

Out[27]:

level_0	price
june	65072
may	89175

Name: price, dtype: int64

```
In [32]: # 3. Print the registration table
# cols -> name -> course -> price

regs.merge(students, on = 'student_id').merge(courses , on='course_id')
```

Out[32]:

	student_id	course_id	name	partner	course_name	price
0	23	1	Chhavi Lachman	18	python	2499
1	15	1	Preet Sha	16	python	2499
2	18	1	Fardeen Mahabir	13	python	2499
3	1	1	Kailash Harjo	23	python	2499
4	21	1	Seema Kota	15	python	2499
5	25	1	Shashank D'Alia	2	python	2499
6	12	1	Radha Dutt	19	python	2499
7	14	1	Pranab Natarajan	22	python	2499
8	23	4	Chhavi Lachman	18	machine learning	9999
9	19	4	Qabeel Raman	12	machine learning	9999
10	24	4	Radhika Suri	17	machine learning	9999
11	14	4	Pranab Natarajan	22	machine learning	9999
12	23	3	Chhavi Lachman	18	data analysis	4999
13	16	3	Elias Dodiya	25	data analysis	4999
14	22	3	Yash Sethi	21	data analysis	4999
15	13	3	Munni Varghese	24	data analysis	4999
16	3	3	Parveen Bhalla	3	data analysis	4999
17	23	6	Chhavi Lachman	18	power bi	1899
18	18	6	Fardeen Mahabir	13	power bi	1899
19	1	6	Kailash Harjo	23	power bi	1899
20	7	6	Tarun Thaker	9	power bi	1899
21	22	6	Yash Sethi	21	power bi	1899
22	14	6	Pranab Natarajan	22	power bi	1899
23	23	9	Chhavi Lachman	18	plotly	699
24	16	9	Elias Dodiya	25	plotly	699
25	1	9	Kailash Harjo	23	plotly	699
26	14	9	Pranab Natarajan	22	plotly	699
27	23	5	Chhavi Lachman	18	tableau	2499
28	15	5	Preet Sha	16	tableau	2499
29	16	5	Elias Dodiya	25	tableau	2499
30	22	5	Yash Sethi	21	tableau	2499
31	3	5	Parveen Bhalla	3	tableau	2499
32	12	5	Radha Dutt	19	tableau	2499
33	2	5	Esha Butala	1	tableau	2499
34	18	8	Fardeen Mahabir	13	pandas	1099
35	7	8	Tarun Thaker	9	pandas	1099
36	11	8	David Mukhopadhyay	20	pandas	1099
37	16	7	Elias Dodiya	25	ms excel	1599
38	7	7	Tarun Thaker	9	ms excel	1599
39	11	7	David Mukhopadhyay	20	ms excel	1599
40	17	7	Yasmin Palan	7	ms excel	1599
41	12	7	Radha Dutt	19	ms excel	1599
42	1	10	Kailash Harjo	23	pyspark	2499
43	7	10	Tarun Thaker	9	pyspark	2499
44	11	10	David Mukhopadhyay	20	pyspark	2499
45	17	10	Yasmin Palan	7	pyspark	2499
46	25	10	Shashank D'Alia	2	pyspark	2499
47	12	10	Radha Dutt	19	pyspark	2499
48	7	2	Tarun Thaker	9	sql	3499
49	19	2	Qabeel Raman	12	sql	3499

```
In [33]: regs.merge(students, on = 'student_id').merge(courses , on='course_id')[['name','course_name','price']]
```

Out[33]:

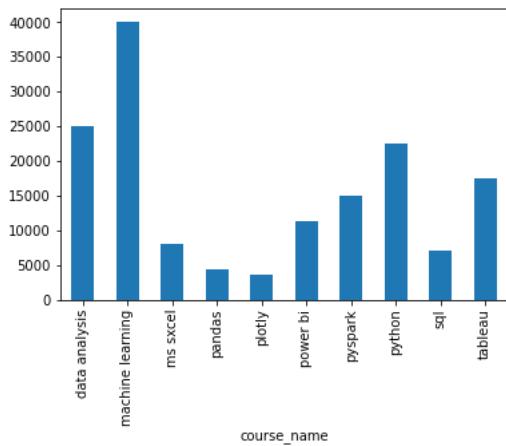
	name	course_name	price
0	Chhavi Lachman	python	2499
1	Preet Sha	python	2499
2	Fardeen Mahabir	python	2499
3	Kailash Harjo	python	2499
4	Seema Kota	python	2499
5	Shashank D'Alia	python	2499
6	Radha Dutt	python	2499
7	Pranab Natarajan	python	2499
8	Chhavi Lachman	machine learning	9999
9	Qabeel Raman	machine learning	9999
10	Radhika Suri	machine learning	9999
11	Pranab Natarajan	machine learning	9999
12	Chhavi Lachman	data analysis	4999
13	Elias Dodiya	data analysis	4999
14	Yash Sethi	data analysis	4999
15	Munni Varghese	data analysis	4999
16	Parveen Bhalla	data analysis	4999
17	Chhavi Lachman	power bi	1899
18	Fardeen Mahabir	power bi	1899
19	Kailash Harjo	power bi	1899
20	Tarun Thaker	power bi	1899
21	Yash Sethi	power bi	1899
22	Pranab Natarajan	power bi	1899
23	Chhavi Lachman	plotly	699
24	Elias Dodiya	plotly	699
25	Kailash Harjo	plotly	699
26	Pranab Natarajan	plotly	699
27	Chhavi Lachman	tableau	2499
28	Preet Sha	tableau	2499
29	Elias Dodiya	tableau	2499
30	Yash Sethi	tableau	2499
31	Parveen Bhalla	tableau	2499
32	Radha Dutt	tableau	2499
33	Esha Butala	tableau	2499
34	Fardeen Mahabir	pandas	1099
35	Tarun Thaker	pandas	1099
36	David Mukhopadhyay	pandas	1099
37	Elias Dodiya	ms excel	1599
38	Tarun Thaker	ms excel	1599
39	David Mukhopadhyay	ms excel	1599
40	Yasmin Palan	ms excel	1599
41	Radha Dutt	ms excel	1599
42	Kailash Harjo	pyspark	2499
43	Tarun Thaker	pyspark	2499
44	David Mukhopadhyay	pyspark	2499
45	Yasmin Palan	pyspark	2499
46	Shashank D'Alia	pyspark	2499
47	Radha Dutt	pyspark	2499
48	Tarun Thaker	sql	3499
49	Qabeel Raman	sql	3499

```
In [38]: # 4. Plot bar chart for revenue/course
regs.merge(courses, on = 'course_id').groupby('course_name')['price'].sum()
```

```
Out[38]: course_name
          data analysis      24995
          machine learning    39996
          ms excel           7995
          pandas              4396
          plotly               3495
          power bi             11394
          pyspark              14994
          python               22491
          sql                  6998
          tableau              17493
Name: price, dtype: int64
```

```
In [41]: regs.merge(courses, on = 'course_id').groupby('course_name')['price'].sum().plot(kind='bar')
```

```
Out[41]: <AxesSubplot:xlabel='course_name'>
```



intersect1d

Find the intersection of two arrays. Return the sorted, unique values that are in both of the input arrays.

```
In [45]: # 5. find students who enrolled in both the months
common_students_id = np.intersect1d(may['student_id'], june['student_id'])
common_students_id
```

```
Out[45]: array([ 1,  3,  7, 11, 16, 17, 18, 22, 23], dtype=int64)
```

```
In [47]: students[students['student_id'].isin(common_students_id)]
```

```
Out[47]:
```

student_id	name	partner
0	Kailash Harjo	23
2	Parveen Bhalla	3
6	Tarun Thaker	9
10	David Mukhopadhyay	20
15	Elias Dodiya	25
16	Yasmin Palan	7
17	Fardeen Mahabir	13
21	Yash Sethi	21
22	Chhavi Lachman	18

numpy.setdiff1d()

function find the set difference of two arrays and return the unique values in arr1 that are not in arr2.

```
In [52]: # 6. find course that got no enrollment
# courses['course_id']
# regs['course_id']

course_id_list = np.setdiff1d(courses['course_id'], regs['course_id'])
courses[courses['course_id'].isin(course_id_list)]
```

Out[52]:

	course_id	course_name	price
10	11	Numpy	699
11	12	C++	1299

```
In [53]: # 7. find students who did not enroll into any courses

student_id_list = np.setdiff1d(students['student_id'], regs['student_id'])
students[students['student_id'].isin(student_id_list)]
```

Out[53]:

	student_id	name	partner
3	4	Mario Dugal	14
4	5	Kusum Bahri	6
5	6	Lakshmi Contractor	10
7	8	Radheshyam Dey	5
8	9	Nitika Chatterjee	4
9	10	Aayushman Sant	8
19	20	Hanuman Hegde	11
25	26	Nitish	28
26	27	Ankit	26
27	28	Rahul	17

```
In [55]: students[students['student_id'].isin(student_id_list)].shape[0]
```

Out[55]: 10

```
In [56]: # Percentage of students Enrolled

(10/28)*100
```

Out[56]: 35.714285714285715

Self Join

A self join is a regular join, but the table is joined with itself.

here, left_on = partner from outside students on left , right_on =student_id from inside students on right .

```
In [60]: # 8. Print student name -> partner name for all enrolled students
# self join
students.merge(students, how = 'inner', left_on = 'partner', right_on= 'student_id')[['name_x','name_y']]
```

Out[60]:

	name_x	name_y
0	Kailash Harjo	Chhavi Lachman
1	Esha Butala	Kailash Harjo
2	Parveen Bhalla	Parveen Bhalla
3	Marlo Dugal	Pranab Natarajan
4	Kusum Bahri	Lakshmi Contractor
5	Lakshmi Contractor	Aayushman Sant
6	Tarun Thaker	Nitika Chatterjee
7	Radheshyam Dey	Kusum Bahri
8	Nitika Chatterjee	Marlo Dugal
9	Aayushman Sant	Radheshyam Dey
10	David Mukhopadhyay	Hanuman Hegde
11	Radha Dutt	Qabeel Raman
12	Munni Varghese	Radhika Suri
13	Pranab Natarajan	Yash Sethi
14	Preet Sha	Elias Dodiya
15	Elias Dodiya	Shashank D'Alia
16	Yasmin Palan	Tarun Thaker
17	Fardeen Mahabir	Munni Varghese
18	Qabeel Raman	Radha Dutt
19	Hanuman Hegde	David Mukhopadhyay
20	Seema Kota	Preet Sha
21	Yash Sethi	Seema Kota
22	Chhavi Lachman	Fardeen Mahabir
23	Radhika Suri	Yasmin Palan
24	Rahul	Yasmin Palan
25	Shashank D'Alia	Esha Butala
26	Nitish	Rahul
27	Ankit	Nitish

```
In [70]: # 9. find top 3 students who did most number enrollments
regs.merge(students, on='student_id').groupby(['student_id', 'name'])['name'].count().sort_values(ascending=False).head(3)
```

```
Out[70]: student_id    name
23          Chhavi Lachman    6
7           Tarun Thaker    5
1            Kailash Harjo    4
Name: name, dtype: int64
```

```
In [81]: # 10. find top 5 students who spent most amount of money on courses
regs.merge(students , on = 'student_id').merge(courses, on= 'course_id').groupby(['student_id', 'name'])['price'].sum().sort_values()
```

```
Out[81]: student_id    name
23          Chhavi Lachman  22594
14          Pranab Natarajan 15096
19          Qabeel Raman    13498
7           Tarun Thaker    10595
24          Radhika Suri     9999
Name: price, dtype: int64
```

```
In [82]: # Alternate syntax for merge
# students.merge(regs)

pd.merge(students,regs , how='inner', on= 'student_id')
```

Out[82]:

	student_id	name	partner	course_id
0	1	Kailash Harjo	23	1
1	1	Kailash Harjo	23	6
2	1	Kailash Harjo	23	10
3	1	Kailash Harjo	23	9
4	2	Esha Butala	1	5
5	3	Parveen Bhalla	3	3
6	3	Parveen Bhalla	3	5
7	7	Tarun Thaker	9	8
8	7	Tarun Thaker	9	10
9	7	Tarun Thaker	9	7
10	7	Tarun Thaker	9	2
11	7	Tarun Thaker	9	6
12	11	David Mukhopadhyay	20	7
13	11	David Mukhopadhyay	20	8
14	11	David Mukhopadhyay	20	10
15	12	Radha Dutt	19	10
16	12	Radha Dutt	19	1
17	12	Radha Dutt	19	5
18	12	Radha Dutt	19	7
19	13	Munni Varghese	24	3
20	14	Pranab Natarajan	22	9
21	14	Pranab Natarajan	22	6
22	14	Pranab Natarajan	22	4
23	14	Pranab Natarajan	22	1
24	15	Preet Sha	16	5
25	15	Preet Sha	16	1
26	16	Elias Dodiya	25	9
27	16	Elias Dodiya	25	5
28	16	Elias Dodiya	25	7
29	16	Elias Dodiya	25	3
30	17	Yasmin Palan	7	7
31	17	Yasmin Palan	7	10
32	18	Fardeen Mahabir	13	6
33	18	Fardeen Mahabir	13	1
34	18	Fardeen Mahabir	13	8
35	19	Qabeel Raman	12	4
36	19	Qabeel Raman	12	2
37	21	Seema Kota	15	1
38	22	Yash Sethi	21	3
39	22	Yash Sethi	21	5
40	22	Yash Sethi	21	6
41	23	Chhavi Lachman	18	1
42	23	Chhavi Lachman	18	4
43	23	Chhavi Lachman	18	3
44	23	Chhavi Lachman	18	6
45	23	Chhavi Lachman	18	9
46	23	Chhavi Lachman	18	5
47	24	Radhika Suri	17	4
48	25	Shashank D'Alia	2	1
49	25	Shashank D'Alia	2	10

In [87]: # IPL Problems

```
# find top 3 stadiums with highest sixes/match ratio
matches
```

Out[87]:

	id	season	city	date	team1	team2	toss_winner	toss_decision	result	dl_applied	winner	win_by_runs	win_by_wickets	player_
0	1	2017	Hyderabad	2017-04-05	Sunrisers Hyderabad	Royal Challengers Bangalore	Royal Challengers Bangalore	field	normal	0	Sunrisers Hyderabad	35	0	Yuvraj Singh
1	2	2017	Pune	2017-04-06	Mumbai Indians	Rising Pune Supergiant	Rising Pune Supergiant	field	normal	0	Rising Pune Supergiant	0	7	Sachin Tendulkar
2	3	2017	Rajkot	2017-04-07	Gujarat Lions	Kolkata Knight Riders	Kolkata Knight Riders	field	normal	0	Kolkata Knight Riders	0	10	MS Dhoni
3	4	2017	Indore	2017-04-08	Rising Pune Supergiant	Kings XI Punjab	Kings XI Punjab	field	normal	0	Kings XI Punjab	0	6	Gautam Gambhir
4	5	2017	Bangalore	2017-04-08	Royal Challengers Bangalore	Delhi Daredevils	Royal Challengers Bangalore	bat	normal	0	Royal Challengers Bangalore	15	0	Kumar Sangakkara
...
631	632	2016	Raipur	2016-05-22	Delhi Daredevils	Royal Challengers Bangalore	Royal Challengers Bangalore	field	normal	0	Royal Challengers Bangalore	0	6	AB De Villiers
632	633	2016	Bangalore	2016-05-24	Gujarat Lions	Royal Challengers Bangalore	Royal Challengers Bangalore	field	normal	0	Royal Challengers Bangalore	0	4	AB de Villiers
633	634	2016	Delhi	2016-05-25	Sunrisers Hyderabad	Kolkata Knight Riders	Kolkata Knight Riders	field	normal	0	Sunrisers Hyderabad	22	0	MC Marylebone
634	635	2016	Delhi	2016-05-27	Gujarat Lions	Sunrisers Hyderabad	Sunrisers Hyderabad	field	normal	0	Sunrisers Hyderabad	0	4	Catriona Curran
635	636	2016	Bangalore	2016-05-29	Sunrisers Hyderabad	Royal Challengers Bangalore	Sunrisers Hyderabad	bat	normal	0	Sunrisers Hyderabad	8	0	Brett D'Oliveira

636 rows × 18 columns



In [89]: deliveries

Out[89]:

	match_id	inning	batting_team	bowling_team	over	ball	batsman	non_striker	bowler	is_super_over	...	bye_runs	legbye_runs	noball_runs	penalt
0	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	1	DA Warner	S Dhawan	TS Mills	0	...	0	0	0	0
1	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	2	DA Warner	S Dhawan	TS Mills	0	...	0	0	0	0
2	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	3	DA Warner	S Dhawan	TS Mills	0	...	0	0	0	0
3	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	4	DA Warner	S Dhawan	TS Mills	0	...	0	0	0	0
4	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	5	DA Warner	S Dhawan	TS Mills	0	...	0	0	0	0
...
179073	11415	2	Chennai Super Kings	Mumbai Indians	20	2	RA Jadeja	SR Watson	SL Malinga	0	...	0	0	0	0
179074	11415	2	Chennai Super Kings	Mumbai Indians	20	3	SR Watson	RA Jadeja	SL Malinga	0	...	0	0	0	0
179075	11415	2	Chennai Super Kings	Mumbai Indians	20	4	SR Watson	RA Jadeja	SL Malinga	0	...	0	0	0	0
179076	11415	2	Chennai Super Kings	Mumbai Indians	20	5	SN Thakur	RA Jadeja	SL Malinga	0	...	0	0	0	0
179077	11415	2	Chennai Super Kings	Mumbai Indians	20	6	SN Thakur	RA Jadeja	SL Malinga	0	...	0	0	0	0

179078 rows × 21 columns

```
In [94]: temp = pd.merge(deliveries,matches ,how ='inner',left_on='match_id',right_on='id')
temp.head(2)
```

Out[94]:

	match_id	inning	batting_team	bowling_team	over	ball	batsman	non_striker	bowler	is_super_over	...	result	dl_applied	winner	win_by_runs	win
0	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	1	DA Warner	S Dhawan	TS Mills	0	...	normal	0	Sunrisers Hyderabad	35	
1	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	2	DA Warner	S Dhawan	TS Mills	0	...	normal	0	Sunrisers Hyderabad	35	

2 rows × 39 columns

```
In [101]: six_df=temp[temp['batsman_runs']==6]
six_df.head(2)
```

Out[101]:

	match_id	inning	batting_team	bowling_team	over	ball	batsman	non_striker	bowler	is_super_over	...	result	dl_applied	winner	win_by_runs
10	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	2	4	DA Warner	S Dhawan	A Choudhary	0	...	normal	0	Sunrisers Hyderabad	35
47	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	8	4	MC Heniques	S Dhawan	TM Head	0	...	normal	0	Sunrisers Hyderabad	35

2 rows × 39 columns

```
In [105]: #stadium --> sixes
number_six = six_df.groupby('venue')['venue'].count()
number_six.head()
```

```
Out[105]: venue
Barabati Stadium          68
Brabourne Stadium         114
Buffalo Park              27
De Beers Diamond Oval    34
Dr DY Patil Sports Academy 173
Name: venue, dtype: int64
```

```
In [108]: # Number of matches
number_matches = matches['venue'].value_counts()
number_matches.head()
```

```
Out[108]: M Chinnaswamy Stadium          66
Eden Gardens                 61
Feroz Shah Kotla            60
Wankhede Stadium             57
Rajiv Gandhi International Stadium, Uppal 49
Name: venue, dtype: int64
```

```
In [112]: (number_six/number_matches).sort_values(ascending=False).head()
```

```
Out[112]: Holkar Cricket Stadium           17.600000
M Chinnaswamy Stadium                13.227273
Sharjah Cricket Stadium               12.666667
Himachal Pradesh Cricket Association Stadium 12.000000
Dr. Y.S. Rajasekhar Reddy ACA-VDCA Cricket Stadium 11.727273
Name: venue, dtype: float64
```

```
In [113]: # find orange cap holder of all the seasons
```

```
Out[113]:
```

	match_id	inning	batting_team	bowling_team	over	ball	batsman	non_striker	bowler	is_super_over	...	result	dl_applied	winner	win_by_runs
0	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	1	DA Warner	S Dhawan	TS Mills	0	...	normal	0	Sunrisers Hyderabad	35
1	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	2	DA Warner	S Dhawan	TS Mills	0	...	normal	0	Sunrisers Hyderabad	35
2	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	3	DA Warner	S Dhawan	TS Mills	0	...	normal	0	Sunrisers Hyderabad	35
3	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	4	DA Warner	S Dhawan	TS Mills	0	...	normal	0	Sunrisers Hyderabad	35
4	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	5	DA Warner	S Dhawan	TS Mills	0	...	normal	0	Sunrisers Hyderabad	35
...	
150455	636	2	Royal Challengers Bangalore	Sunrisers Hyderabad	20	2	Sachin Baby	CJ Jordan	B Kumar	0	...	normal	0	Sunrisers Hyderabad	8
150456	636	2	Royal Challengers Bangalore	Sunrisers Hyderabad	20	3	Sachin Baby	CJ Jordan	B Kumar	0	...	normal	0	Sunrisers Hyderabad	8
150457	636	2	Royal Challengers Bangalore	Sunrisers Hyderabad	20	4	Iqbal Abdulla	Sachin Baby	B Kumar	0	...	normal	0	Sunrisers Hyderabad	8
150458	636	2	Royal Challengers Bangalore	Sunrisers Hyderabad	20	5	Sachin Baby	Iqbal Abdulla	B Kumar	0	...	normal	0	Sunrisers Hyderabad	8
150459	636	2	Royal Challengers Bangalore	Sunrisers Hyderabad	20	6	Iqbal Abdulla	Sachin Baby	B Kumar	0	...	normal	0	Sunrisers Hyderabad	8

150460 rows × 39 columns

```
In [114]: df = pd.merge(deliveries,matches ,how ='inner',left_on='match_id',right_on='id')

df.head(2)
```

```
Out[114]:
   match_id  inning  batting_team  bowling_team  over  ball  batsman  non_striker  bowler  is_super_over  ...  result  dl_applied  winner  win_by_runs  win_
0          1       1  Sunrisers Hyderabad  Royal Challengers Bangalore  1     1    DA Warner      S Dhawan    TS Mills        0  ...  normal  0  Sunrisers Hyderabad  35
1          1       1  Sunrisers Hyderabad  Royal Challengers Bangalore  1     2    DA Warner      S Dhawan    TS Mills        0  ...  normal  0  Sunrisers Hyderabad  35
```

2 rows × 39 columns

```
In [117]: df.groupby(['season','batsman'])['batsman_runs'].sum()
```

```
Out[117]:
   season  batsman
2008      A Chopra      42
             A Kumble      13
             A Mishra       37
             A Mukund        0
             A Nehra         3
             ...
2017      Washington Sundar     9
             YK Pathan     143
             YS Chahal      13
             Yuvraj Singh   252
             Z Khan         4
Name: batsman_runs, Length: 1531, dtype: int64
```

```
In [120]: df.groupby(['season','batsman'])['batsman_runs'].sum().reset_index().sort_values('batsman_runs',ascending=False)
```

```
Out[120]:
   season  batsman  batsman_runs
1383    2016      V Kohli      973
1278    2016      DA Warner    848
910     2013      MEK Hussey    733
684     2012      CH Gayle     733
852     2013      CH Gayle     720
...
1467    2017      MM Patel      0
658     2012      AC Blizzard    0
475     2011      AB Dinda      0
1394    2017      AD Nath      0
58      2008      L Balaji      0
```

1531 rows × 3 columns

```
In [123]: '[']['batsman_runs'].sum().reset_index().sort_values('batsman_runs',ascending=False).drop_duplicates(subset='season',keep='first')
```

```
Out[123]:
   season  batsman  batsman_runs
1383    2016      V Kohli      973
910     2013      MEK Hussey    733
684     2012      CH Gayle     733
1088    2014      RV Uthappa    660
1422    2017      DA Warner    641
446     2010      SR Tendulkar   618
115     2008      SE Marsh     616
502     2011      CH Gayle     608
229     2009      ML Hayden     572
1148    2015      DA Warner    562
```

```
In [124]: bupby(['season','batsman'])['batsman_runs'].sum().reset_index().sort_values('batsman_runs',ascending=False).sort_values('season')
```

Out[124]:

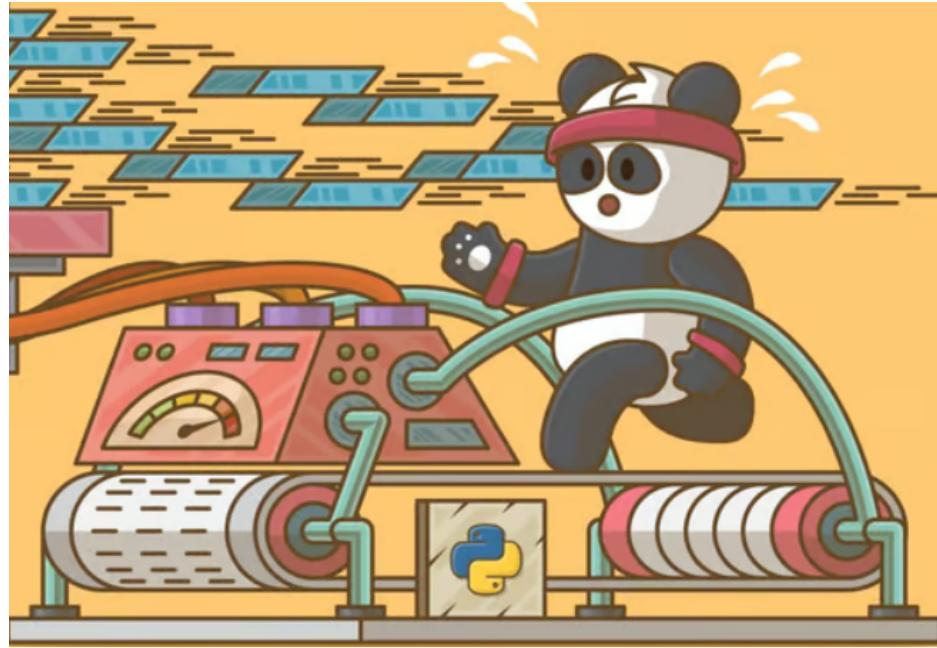
	season	batsman	batsman_runs
58	2008	L Balaji	0
45	2008	I Sharma	11
12	2008	AM Nayar	206
31	2008	DNT Zoysa	11
67	2008	M Ntini	11
...
1424	2017	DL Chahar	14
1515	2017	Swapnil Singh	12
1516	2017	TA Boult	5
1470	2017	MP Stoinis	17
1400	2017	AR Bawne	12

1531 rows × 3 columns

In []:

What is MultiIndex in Pandas?

In pandas, a multi-index, also known as a **hierarchical index**, is a way to represent two or more dimensions of data in a single index. This is useful when you have data that can be grouped or categorized by more than one variable.



```
In [1]: import numpy as np
import pandas as pd
```

Series is 1D and DataFrames are 2D objects

- But why?
- And what exactly is index?

```
In [2]: # can we have multiple index? Let's try
index_val = [('cse', 2019), ('cse', 2020), ('cse', 2021), ('cse', 2022), ('ece', 2019), ('ece', 2020), ('ece', 2021), ('ece', 2022)]
data = pd.Series([1,2,3,4,5,6,7,8], index=index_val)
data
```

```
Out[2]: (cse, 2019)    1
(cse, 2020)    2
(cse, 2021)    3
(cse, 2022)    4
(ece, 2019)    5
(ece, 2020)    6
(ece, 2021)    7
(ece, 2022)    8
dtype: int64
```

```
In [3]: # The problem?
#data['cse']
```

The solution : multiindex series-2D(also known as Hierarchical Indexing)

multiple index levels within a single index

```
In [4]: # how to create multiindex object
# 1. pd.MultiIndex.from_tuples()
index_val = [('cse', 2019), ('cse', 2020), ('cse', 2021), ('cse', 2022), ('ece', 2019), ('ece', 2020), ('ece', 2021), ('ece', 2022)]
multiindex = pd.MultiIndex.from_tuples(index_val)
multiindex.levels[0]
```

```
Out[4]: Index(['cse', 'ece'], dtype='object')
```

```
In [5]: # 2. pd.MultiIndex.from_product()
pd.MultiIndex.from_product([['cse', 'ece'], [2019, 2020, 2021, 2022]])
```

```
Out[5]: MultiIndex([('cse', 2019),
 ('cse', 2020),
 ('cse', 2021),
 ('cse', 2022),
 ('ece', 2019),
 ('ece', 2020),
 ('ece', 2021),
 ('ece', 2022)],
```

```
In [6]: # creating a series with multiindex object
sample = pd.Series([1,2,3,4,5,6,7,8],index=multiindex)
sample
```

```
Out[6]: cse 2019    1
        2020    2
        2021    3
        2022    4
      ece 2019    5
        2020    6
        2021    7
        2022    8
dtype: int64
```

```
In [7]: # how to fetch items from such a series
sample[['cse',2022]]
```

```
Out[7]: 4
```

```
In [8]: sample['cse']
```

```
Out[8]: 2019    1
        2020    2
        2021    3
        2022    4
dtype: int64
```

unstack

reshape the given Pandas DataFrame by transposing specified row level to column level

```
In [9]: temp = sample.unstack()
temp
```

```
Out[9]:
2019 2020 2021 2022
cse   1    2    3    4
ece   5    6    7    8
```

stack

reshapes the given DataFrame by converting the column label to a row index.

```
In [10]: temp.stack()
```

```
Out[10]: cse 2019    1
        2020    2
        2021    3
        2022    4
      ece 2019    5
        2020    6
        2021    7
        2022    8
dtype: int64
```

so why we should study Multi Index

because we can convert any dataframe dimension, including 3D, 4D, 10D, and 20D, to 1Dimension (Series) and 2Dimension (Dataframes).

```
In [11]: # multi index dataframes
branch_df1 = pd.DataFrame(
    [
        [1,2],
        [3,4],
        [5,6],
        [7,8],
        [9,10],
        [11,12],
        [13,14],
        [15,16],
    ],
    index = multiindex,
    columns = ['avg_package', 'students']
)
branch_df1
```

Out[11]:

		avg_package	students
cse	2019	1	2
	2020	3	4
	2021	5	6
	2022	7	8
ece	2019	9	10
	2020	11	12
	2021	13	14
	2022	15	16

In [12]: branch_df1.loc['cse']

Out[12]:

	avg_package	students
2019	1	2
2020	3	4
2021	5	6
2022	7	8

In [13]: branch_df1['avg_package']

```
Out[13]: cse  2019    1
          2020    3
          2021    5
          2022    7
         ece  2019    9
          2020   11
          2021   13
          2022   15
Name: avg_package, dtype: int64
```

In [14]: branch_df1['students']

```
Out[14]: cse  2019    2
          2020    4
          2021    6
          2022    8
         ece  2019   10
          2020   12
          2021   14
          2022   16
Name: students, dtype: int64
```

In [15]: `branch_df1.loc['ece']`

Out[15]:

	avg_package	students
2019	9	10
2020	11	12
2021	13	14
2022	15	16

multiindex df from columns perspective

In [16]: `branch_df2 = pd.DataFrame([[1,2,0,0], [3,4,0,0], [5,6,0,0], [7,8,0,0]], index = [2019,2020,2021,2022], columns = pd.MultiIndex.from_product([[['delhi','mumbai'],['avg_package','students']]])`

branch_df2

Out[16]:

	delhi		mumbai	
	avg_package	students	avg_package	students
2019	1	2	0	0
2020	3	4	0	0
2021	5	6	0	0
2022	7	8	0	0

In [17]: `branch_df2['delhi']`

Out[17]:

	avg_package	students
2019	1	2
2020	3	4
2021	5	6
2022	7	8

In [18]: `branch_df2.loc[2019]`

Out[18]:

delhi	avg_package	1
	students	2
mumbai	avg_package	0
	students	0
Name:	2019, dtype:	int64

In [19]: `branch_df2.iloc[1]`

Out[19]:

delhi	avg_package	3
	students	4
mumbai	avg_package	0
	students	0
Name:	2020, dtype:	int64

Multindex df in terms of both cols and index

```
In [20]: branch_df3 = pd.DataFrame(
    [
        [1,2,0,0],
        [3,4,0,0],
        [5,6,0,0],
        [7,8,0,0],
        [9,10,0,0],
        [11,12,0,0],
        [13,14,0,0],
        [15,16,0,0],
    ],
    index = multiindex,
    columns = pd.MultiIndex.from_product([['delhi','mumbai'],['avg_package','students']])
)

branch_df3

#here index= multiindex is a name , already we have stored data of ece and cse in above
```

Out[20]:

		delhi		mumbai	
		avg_package	students	avg_package	students
cse	2019	1	2	0	0
	2020	3	4	0	0
	2021	5	6	0	0
	2022	7	8	0	0
ece	2019	9	10	0	0
	2020	11	12	0	0
	2021	13	14	0	0
	2022	15	16	0	0

Stacking and Unstacking

In [21]: branch_df1

Out[21]:

		avg_package	students
cse	2019	1	2
	2020	3	4
	2021	5	6
	2022	7	8
ece	2019	9	10
	2020	11	12
	2021	13	14
	2022	15	16

```
In [22]: # After applying Unstack
branch_df1.unstack()
```

Out[22]:

	avg_package				students			
	2019	2020	2021	2022	2019	2020	2021	2022
cse	1	3	5	7	2	4	6	8
ece	9	11	13	15	10	12	14	16

```
In [23]: branch_df1.unstack().unstack()
```

```
Out[23]: avg_package  2019  cse    1
              ece    9
            2020  cse    3
              ece   11
            2021  cse    5
              ece   13
            2022  cse    7
              ece   15
      students  2019  cse    2
              ece   10
            2020  cse    4
              ece   12
            2021  cse    6
              ece   14
            2022  cse    8
              ece   16
dtype: int64
```

The stack() method

It can be used to move the columns to the index. This means that the columns will become the rows, and the rows will become the columns.
The stack method can be used to move the columns to the index

```
In [24]: # After applying Unstack + stack
branch_df1.unstack().stack()
```

```
Out[24]:      avg_package  students
cse  2019        1        2
        2020        3        4
        2021        5        6
        2022        7        8
ece  2019        9       10
        2020       11       12
        2021       13       14
        2022       15       16
```

```
In [25]: # applying multiple stack
branch_df1.unstack().stack().stack()
```

```
Out[25]: cse  2019  avg_package    1
              students    2
            2020  avg_package    3
              students    4
            2021  avg_package    5
              students    6
            2022  avg_package    7
              students    8
      ece  2019  avg_package    9
              students   10
            2020  avg_package   11
              students   12
            2021  avg_package   13
              students   14
            2022  avg_package   15
              students   16
dtype: int64
```

```
In [26]: # Example : 2
branch_df2
```

Out[26]:

	delhi		mumbai	
	avg_package	students	avg_package	students
2019	1	2	0	0
2020	3	4	0	0
2021	5	6	0	0
2022	7	8	0	0

The Unstack()

It is method can be used to move the index to the columns. This means that the index will become the rows, and the rows will become the columns.

The unstack method can be used to move the index to the columns

```
In [27]: branch_df2.unstack()
```

```
Out[27]: delhi    avg_package  2019    1
              2020    3
              2021    5
              2022    7
            students    2019    2
              2020    4
              2021    6
              2022    8
      mumbai    avg_package  2019    0
              2020    0
              2021    0
              2022    0
            students    2019    0
              2020    0
              2021    0
              2022    0
dtype: int64
```

```
In [28]: branch_df2.stack()
```

Out[28]:

	delhi	mumbai
2019	avg_package	1
	students	0
2020	avg_package	3
	students	0
2021	avg_package	5
	students	0
2022	avg_package	7
	students	0

In [29]: `branch_df2.stack().stack()`

```
Out[29]: 2019  avg_package  delhi      1
          mumbai      0
          students    delhi      2
                      mumbai      0
2020  avg_package  delhi      3
          mumbai      0
          students    delhi      4
                      mumbai      0
2021  avg_package  delhi      5
          mumbai      0
          students    delhi      6
                      mumbai      0
2022  avg_package  delhi      7
          mumbai      0
          students    delhi      8
                      mumbai      0
dtype: int64
```

In [30]: `# Working on 4D data
branch_df3`

Out[30]:

		delhi		mumbai	
		avg_package	students	avg_package	students
cse	2019	1	2	0	0
	2020	3	4	0	0
	2021	5	6	0	0
	2022	7	8	0	0
ece	2019	9	10	0	0
	2020	11	12	0	0
	2021	13	14	0	0
	2022	15	16	0	0

In [31]: `branch_df3.stack()`

Out[31]:

		delhi	mumbai	
cse	2019	avg_package	1	0
		students	2	0
2020	avg_package	3	0	
		students	4	0
2021	avg_package	5	0	
		students	6	0
2022	avg_package	7	0	
		students	8	0
ece	2019	avg_package	9	0
		students	10	0
2020	avg_package	11	0	
		students	12	0
2021	avg_package	13	0	
		students	14	0
2022	avg_package	15	0	
		students	16	0

In [32]: `branch_df3.stack().stack()`

```
Out[32]: cse  2019  avg_package  delhi      1
              avg_package  mumbai     0
              students    delhi      2
              students    mumbai     0
            2020  avg_package  delhi      3
              avg_package  mumbai     0
              students    delhi      4
              students    mumbai     0
            2021  avg_package  delhi      5
              avg_package  mumbai     0
              students    delhi      6
              students    mumbai     0
            2022  avg_package  delhi      7
              avg_package  mumbai     0
              students    delhi      8
              students    mumbai     0
        ece  2019  avg_package  delhi      9
              avg_package  mumbai     0
              students    delhi     10
              students    mumbai     0
            2020  avg_package  delhi     11
              avg_package  mumbai     0
              students    delhi     12
              students    mumbai     0
            2021  avg_package  delhi     13
              avg_package  mumbai     0
              students    delhi     14
              students    mumbai     0
            2022  avg_package  delhi     15
              avg_package  mumbai     0
              students    delhi     16
              students    mumbai     0
dtype: int64
```

In [33]: `# Unstacking on 4D data
branch_df3.unstack()`

Out[33]:

	delhi						mumbai									
	avg_package			students			avg_package			students						
	2019	2020	2021	2022	2019	2020	2021	2022	2019	2020	2021	2022	2019	2020	2021	2022
cse	1	3	5	7	2	4	6	8	0	0	0	0	0	0	0	0
ece	9	11	13	15	10	12	14	16	0	0	0	0	0	0	0	0

```
In [34]: branch_df3.unstack().unstack()
```

```
Out[34]: delhi    avg_package  2019  cse    1
              ece    9
              2020  cse    3
              ece   11
              2021  cse    5
              ece   13
              2022  cse    7
              ece   15
      students    2019  cse    2
              ece   10
              2020  cse    4
              ece   12
              2021  cse    6
              ece   14
              2022  cse    8
              ece   16
mumbai    avg_package  2019  cse    0
              ece    0
              2020  cse    0
              ece    0
              2021  cse    0
              ece    0
              2022  cse    0
              ece    0
      students    2019  cse    0
              ece    0
              2020  cse    0
              ece    0
              2021  cse    0
              ece    0
              2022  cse    0
              ece    0
dtype: int64
```

Working with multiindex dataframes

```
In [35]: # Multi index dataframes works same as normal dataframes
branch_df3
```

```
Out[35]:
```

	delhi		mumbai	
	avg_package	students	avg_package	students
cse 2019	1	2	0	0
2020	3	4	0	0
2021	5	6	0	0
2022	7	8	0	0
ece 2019	9	10	0	0
2020	11	12	0	0
2021	13	14	0	0
2022	15	16	0	0

```
In [36]: # head and tail
branch_df3.head()
```

```
Out[36]:
```

	delhi		mumbai	
	avg_package	students	avg_package	students
cse 2019	1	2	0	0
2020	3	4	0	0
2021	5	6	0	0
2022	7	8	0	0
ece 2019	9	10	0	0

```
In [37]: # shape
branch_df3.shape
```

```
Out[37]: (8, 4)
```

```
In [38]: # info
branch_df3.info()
```

```
<class 'pandas.core.frame.DataFrame'>
MultiIndex: 8 entries, ('cse', 2019) to ('ece', 2022)
Data columns (total 4 columns):
 #   Column           Non-Null Count  Dtype  
---  --  
 0   (delhi, avg_package)    8 non-null   int64  
 1   (delhi, students)      8 non-null   int64  
 2   (mumbai, avg_package)  8 non-null   int64  
 3   (mumbai, students)    8 non-null   int64  
dtypes: int64(4)
memory usage: 932.0+ bytes
```

```
In [39]: # duplicated -> isnull
branch_df3.duplicated()
```

```
Out[39]: cse  2019    False
          2020    False
          2021    False
          2022    False
         ece  2019    False
          2020    False
          2021    False
          2022    False
dtype: bool
```

```
In [40]: branch_df3.isnull()
```

```
Out[40]:
```

	delhi		mumbai	
	avg_package	students	avg_package	students
cse 2019	False	False	False	False
2020	False	False	False	False
2021	False	False	False	False
2022	False	False	False	False
ece 2019	False	False	False	False
2020	False	False	False	False
2021	False	False	False	False
2022	False	False	False	False

```
In [41]: # Extracting rows single
```

```
branch_df3.loc[('cse', 2022)]
```

```
Out[41]: delhi    avg_package    7
          students      8
         mumbai    avg_package    0
          students      0
Name: (cse, 2022), dtype: int64
```

```
In [42]: # Extracting multiple rows
```

```
branch_df3.loc[('cse', 2019):('ece', 2020):2]
```

```
Out[42]:
```

	delhi		mumbai	
	avg_package	students	avg_package	students
cse 2019	1	2	0	0
2021	5	6	0	0
ece 2019	9	10	0	0

```
In [43]: # Using iloc
branch_df3.iloc[0:5:2]
```

Out[43]:

		delhi		mumbai	
		avg_package	students	avg_package	students
cse	2019	1	2	0	0
	2021	5	6	0	0
ece	2019	9	10	0	0

```
In [44]: # Extracting single columns
branch_df3['delhi']['students']
```

```
Out[44]: cse 2019    2
          2020    4
          2021    6
          2022    8
         ece 2019    10
          2020   12
          2021   14
          2022   16
Name: students, dtype: int64
```

```
In [45]: # we want to extract delhi - students , mumbai - avg_package
branch_df3
```

Out[45]:

		delhi		mumbai	
		avg_package	students	avg_package	students
cse	2019	1	2	0	0
	2020	3	4	0	0
	2021	5	6	0	0
	2022	7	8	0	0
ece	2019	9	10	0	0
	2020	11	12	0	0
	2021	13	14	0	0
	2022	15	16	0	0

```
In [46]: #here [:] all rows ,
#columns : delhi=avg_package[0],students[1],mumbai=avg_package[2],students[3]
branch_df3.iloc[:,1:3]
```

Out[46]:

		delhi	mumbai
		students	avg_package
cse	2019	2	0
	2020	4	0
	2021	6	0
	2022	8	0
ece	2019	10	0
	2020	12	0
	2021	14	0
	2022	16	0

```
In [47]: # Extracting both rows and columns
branch_df3.iloc[[0,4],[1,2]]
```

Out[47]:

		delhi	mumbai
		students	avg_package
cse	2019	2	0
ece	2019	10	0

```
In [48]: # sort index
# both -> descending -> diff order
# based on one level
```

```
branch_df3
```

Out[48]:

		delhi		mumbai	
		avg_package	students	avg_package	students
cse	2019	1	2	0	0
	2020	3	4	0	0
	2021	5	6	0	0
	2022	7	8	0	0
ece	2019	9	10	0	0
	2020	11	12	0	0
	2021	13	14	0	0
	2022	15	16	0	0

```
In [49]: branch_df3.sort_index(ascending=False)
```

Out[49]:

		delhi		mumbai	
		avg_package	students	avg_package	students
ece	2022	15	16	0	0
	2021	13	14	0	0
	2020	11	12	0	0
	2019	9	10	0	0
cse	2022	7	8	0	0
	2021	5	6	0	0
	2020	3	4	0	0
	2019	1	2	0	0

```
In [50]: # if we want year in descending order
branch_df3.sort_index(ascending=[False ,True])
```

Out[50]:

		delhi		mumbai	
		avg_package	students	avg_package	students
ece	2019	9	10	0	0
	2020	11	12	0	0
	2021	13	14	0	0
	2022	15	16	0	0
cse	2019	1	2	0	0
	2020	3	4	0	0
	2021	5	6	0	0
	2022	7	8	0	0

```
In [51]: # multiindex dataframe(col) -> transpose
branch_df3.transpose()
```

Out[51]:

		cse				ece			
		2019	2020	2021	2022	2019	2020	2021	2022
delhi	avg_package	1	3	5	7	9	11	13	15
	students	2	4	6	8	10	12	14	16
mumbai	avg_package	0	0	0	0	0	0	0	0
	students	0	0	0	0	0	0	0	0

In [52]: # swapLevel
branch_df3

Out[52]:

		delhi		mumbai	
		avg_package	students	avg_package	students
cse	2019	1	2	0	0
	2020	3	4	0	0
	2021	5	6	0	0
	2022	7	8	0	0
ece	2019	9	10	0	0
	2020	11	12	0	0
	2021	13	14	0	0
	2022	15	16	0	0

In [53]: # On rows
branch_df3.swaplevel()

Out[53]:

		delhi		mumbai	
		avg_package	students	avg_package	students
2019	cse	1	2	0	0
2020	cse	3	4	0	0
2021	cse	5	6	0	0
2022	cse	7	8	0	0
2019	ece	9	10	0	0
2020	ece	11	12	0	0
2021	ece	13	14	0	0
2022	ece	15	16	0	0

In [54]: # on columns
branch_df3.swaplevel(axis=1)

Out[54]:

		avg_package		students	
		delhi	delhi	mumbai	mumbai
cse	2019	1	2	0	0
	2020	3	4	0	0
	2021	5	6	0	0
	2022	7	8	0	0
ece	2019	9	10	0	0
	2020	11	12	0	0
	2021	13	14	0	0
	2022	15	16	0	0

Long(Tall) Vs Wide data

“Long” format

country	year	metric
x	1960	10
x	1970	13
x	2010	15
y	1960	20
y	1970	23
y	2010	25
z	1960	30
z	1970	33
z	2010	35

“Wide” format

country	yr1960	yr1970	yr2010
x	10	13	15
y	20	23	25
z	30	33	35

Wide format is where we have a single row for every data point with multiple columns to hold the values of various attributes.

Long format is where, for each data point we have as many rows as the number of attributes and each row contains the value of a particular attribute for a given data point.

Melt -- Converting wide data to long Data.

```
In [55]: # melt -> simple example branch
# wide to Long
pd.DataFrame({'cse':[120]})
```

```
Out[55]:
cse
0    120
```

```
In [56]: pd.DataFrame({'cse':[120]}).melt()
```

```
Out[56]:
variable  value
0        cse    120
```

```
In [57]: # melt -> branch with year
pd.DataFrame({'cse':[120], 'ece':[100], 'mech':[50]}).melt()
```

```
Out[57]:
variable  value
0        cse    120
1        ece    100
2      mech     50
```

```
In [58]: # we can name the variable and value
pd.DataFrame({'cse':[120], 'ece':[100], 'mech':[50]}).melt(var_name='branch',value_name='num_students')
```

Out[58]:

	branch	num_students
0	cse	120
1	ece	100
2	mech	50

```
In [59]: pd.DataFrame(
{
    'branch':['cse', 'ece', 'mech'],
    '2020':[100,150,60],
    '2021':[120,130,80],
    '2022':[150,140,70]
})
```

Out[59]:

	branch	2020	2021	2022
0	cse	100	120	150
1	ece	150	130	140
2	mech	60	80	70

```
In [60]: pd.DataFrame(
{
    'branch':['cse', 'ece', 'mech'],
    '2020':[100,150,60],
    '2021':[120,130,80],
    '2022':[150,140,70]
}).melt()
```

Out[60]:

	variable	value
0	branch	cse
1	branch	ece
2	branch	mech
3	2020	100
4	2020	150
5	2020	60
6	2021	120
7	2021	130
8	2021	80
9	2022	150
10	2022	140
11	2022	70

```
In [61]: # dont include 'branch' to rows
pd.DataFrame(
    {
        'branch': ['cse', 'ece', 'mech'],
        '2020': [100, 150, 60],
        '2021': [120, 130, 80],
        '2022': [150, 140, 70]
    }
).melt(id_vars=['branch'])
```

Out[61]:

	branch	variable	value
0	cse	2020	100
1	ece	2020	150
2	mech	2020	60
3	cse	2021	120
4	ece	2021	130
5	mech	2021	80
6	cse	2022	150
7	ece	2022	140
8	mech	2022	70

the `melt()` method is used to reshape a DataFrame from wide to long format. This means that the columns of the DataFrame are converted into rows, and the values in the columns are converted into columns.

```
In [62]: # adding variable and value names.
pd.DataFrame(
    {
        'branch': ['cse', 'ece', 'mech'],
        '2020': [100, 150, 60],
        '2021': [120, 130, 80],
        '2022': [150, 140, 70]
    }
).melt(id_vars=['branch'], var_name='year', value_name='students')
```

Out[62]:

	branch	year	students
0	cse	2020	100
1	ece	2020	150
2	mech	2020	60
3	cse	2021	120
4	ece	2021	130
5	mech	2021	80
6	cse	2022	150
7	ece	2022	140
8	mech	2022	70

```
In [63]: # melt --> Real world examples.
```

```
deaths = pd.read_csv("time_series_covid19_deaths_global.csv")
confirm = pd.read_csv("time_series_covid19_confirmed_global.csv")
```

```
In [64]: deaths.head(2)
```

Out[64]:

	Province/State	Country/Region	Lat	Long	1/22/20	1/23/20	1/24/20	1/25/20	1/26/20	1/27/20	...	12/24/22	12/25/22	12/26/22	12/27/22
0	NaN	Afghanistan	33.93911	67.709953	0	0	0	0	0	0	...	7845	7846	7846	7846
1	NaN	Albania	41.15330	20.168300	0	0	0	0	0	0	...	3595	3595	3595	3595

2 rows × 1081 columns

In [65]: `deaths.shape`

Out[65]: (289, 1081)

In [66]: `deaths = deaths.melt(id_vars=['Province/State', 'Country/Region', 'Lat', 'Long'], var_name='date', value_name='no. of deaths')`

In [67]: `# After converting columns into rows,
which is converting wide format to long format using 'melt'`
`deaths.shape`

Out[67]: (311253, 6)

In [75]: `deaths.head()`

Out[75]:

	Province/State	Country/Region	Lat	Long	date	no. of deaths
0	NaN	Afghanistan	33.93911	67.709953	1/22/20	0
1	NaN	Albania	41.15330	20.168300	1/22/20	0
2	NaN	Algeria	28.03390	1.659600	1/22/20	0
3	NaN	Andorra	42.50630	1.521800	1/22/20	0
4	NaN	Angola	-11.20270	17.873900	1/22/20	0

In [68]: `confirm.head(2)`

Out[68]:

	Province/State	Country/Region	Lat	Long	1/22/20	1/23/20	1/24/20	1/25/20	1/26/20	1/27/20	...	12/24/22	12/25/22	12/26/22	12/27/22
0	NaN	Afghanistan	33.93911	67.709953	0	0	0	0	0	0	0	207310	207399	207438	207460
1	NaN	Albania	41.15330	20.168300	0	0	0	0	0	0	0	333749	333749	333751	333751

2 rows × 1081 columns

In [69]: `confirm.shape`

Out[69]: (289, 1081)

In [73]: `confirm = confirm.melt(id_vars=['Province/State', 'Country/Region', 'Lat', 'Long'], var_name='date', value_name='no. of confirmed')`

In [76]: `confirm.head()`

Out[76]:

	Province/State	Country/Region	Lat	Long	date	no. of confirmed
0	NaN	Afghanistan	33.93911	67.709953	1/22/20	0
1	NaN	Albania	41.15330	20.168300	1/22/20	0
2	NaN	Algeria	28.03390	1.659600	1/22/20	0
3	NaN	Andorra	42.50630	1.521800	1/22/20	0
4	NaN	Angola	-11.20270	17.873900	1/22/20	0

In [74]: `# After converting columns into rows,
which is converting wide format to long format using 'melt'`
`confirm.shape`

Out[74]: (311253, 6)

```
In [77]: # Now merge both data frames as per desire
confirm.merge(deaths, on =['Province/State', 'Country/Region', 'Lat', 'Long', 'date'])
```

Out[77]:

	Province/State	Country/Region	Lat	Long	date	no. of confirmed	no. of deaths
0	NaN	Afghanistan	33.939110	67.709953	1/22/20	0	0
1	NaN	Albania	41.153300	20.168300	1/22/20	0	0
2	NaN	Algeria	28.033900	1.659600	1/22/20	0	0
3	NaN	Andorra	42.506300	1.521800	1/22/20	0	0
4	NaN	Angola	-11.202700	17.873900	1/22/20	0	0
...
311248	NaN	West Bank and Gaza	31.952200	35.233200	1/2/23	703228	5708
311249	NaN	Winter Olympics 2022	39.904200	116.407400	1/2/23	535	0
311250	NaN	Yemen	15.552727	48.516388	1/2/23	11945	2159
311251	NaN	Zambia	-13.133897	27.849332	1/2/23	334661	4024
311252	NaN	Zimbabwe	-19.015438	29.154857	1/2/23	259981	5637

311253 rows × 7 columns

```
In [80]: # desired columns
deaths, on =['Province/State', 'Country/Region', 'Lat', 'Long', 'date'])[[['Country/Region', 'date', 'no. of confirmed', 'no. of deaths']]
```

Out[80]:

	Country/Region	date	no. of confirmed	no. of deaths
0	Afghanistan	1/22/20	0	0
1	Albania	1/22/20	0	0
2	Algeria	1/22/20	0	0
3	Andorra	1/22/20	0	0
4	Angola	1/22/20	0	0
...
311248	West Bank and Gaza	1/2/23	703228	5708
311249	Winter Olympics 2022	1/2/23	535	0
311250	Yemen	1/2/23	11945	2159
311251	Zambia	1/2/23	334661	4024
311252	Zimbabwe	1/2/23	259981	5637

311253 rows × 4 columns

Pivot table -- Converting Long data to wide data.

the **Pivot table** takes simple column wise data as input, and groups as the entire Into 2 dimensional table that provides a multi dimensional summarization of the data.

Pivot table generally used on categorical data

```
In [81]: import seaborn as sns
```

```
In [83]: df = sns.load_dataset('tips')
df.head()
```

Out[83]:

	total_bill	tip	sex	smoker	day	time	size
0	16.99	1.01	Female	No	Sun	Dinner	2
1	10.34	1.66	Male	No	Sun	Dinner	3
2	21.01	3.50	Male	No	Sun	Dinner	3
3	23.68	3.31	Male	No	Sun	Dinner	2
4	24.59	3.61	Female	No	Sun	Dinner	4

```
In [85]: # On gender basis average total bill
df.groupby('sex')[ 'total_bill'].mean()
```

```
Out[85]: sex
Male      20.744076
Female    18.056897
Name: total_bill, dtype: float64
```

```
In [88]: # On gender basis. Who smokes more? On average.
df.groupby(['sex', 'smoker'])['total_bill'].mean().unstack()
```

```
Out[88]: smoker Yes      No
          sex
          _____
          Male 22.284500 19.791237
          Female 17.977879 18.105185
```

```
In [89]: # Using Pivot table method
```

```
df.pivot_table(index ='sex',columns='smoker',values = 'total_bill')
```

```
Out[89]: smoker Yes      No
          sex
          _____
          Male 22.284500 19.791237
          Female 17.977879 18.105185
```

```
In [90]: # Aggregate function.
```

```
# Print, the total amount smokers of the bill, Not mean Or average,
df.pivot_table(index ='sex',columns='smoker',values = 'total_bill',aggfunc='sum')
```

```
Out[90]: smoker Yes      No
          sex
          _____
          Male 1337.07 1919.75
          Female 593.27 977.68
```

```
In [91]: # count of people
```

```
df.pivot_table(index ='sex',columns='smoker',values = 'total_bill',aggfunc='count')
```

```
Out[91]: smoker Yes  No
          sex
          _____
          Male 60 97
          Female 33 54
```

```
In [92]: # standard deviation
```

```
df.pivot_table(index ='sex',columns='smoker',values = 'total_bill',aggfunc='std')
```

```
Out[92]: smoker Yes      No
          sex
          _____
          Male 9.911845 8.726566
          Female 9.189751 7.286455
```

In [93]: # All columns together --- gives average

```
df.pivot_table(index='sex',columns='smoker')
```

Out[93]:

	size		tip		total_bill	
smoker	Yes	No	Yes	No	Yes	No
sex						
Male	2.500000	2.711340	3.051167	3.113402	22.284500	19.791237
Female	2.242424	2.592593	2.931515	2.773519	17.977879	18.105185

In [95]: # single column

```
df.pivot_table(index='sex',columns='smoker')['tip']
```

Out[95]:

smoker	Yes	No
sex		
Male	3.051167	3.113402
Female	2.931515	2.773519

In [96]: df.pivot_table(index='sex',columns='smoker')['size']

Out[96]:

smoker	Yes	No
sex		
Male	2.500000	2.711340
Female	2.242424	2.592593

In [98]: # Multi dimensional -5D
df.head(2)

Out[98]:

	total_bill	tip	sex	smoker	day	time	size
0	16.99	1.01	Female	No	Sun	Dinner	2
1	10.34	1.66	Male	No	Sun	Dinner	3

In [100]: # 5D - 5 Dimensional data

```
df.pivot_table(index=['sex','smoker'],columns=['day','time'],values='total_bill')
```

Out[100]:

	day	Thur	Fri	Sat	Sun
time	Lunch	Dinner	Lunch	Dinner	Dinner
sex	smoker				
Male	Yes	19.171000	NaN	11.386667	25.892
	No	18.486500	NaN	17.475	19.929063
Female	Yes	19.218571	NaN	13.260000	20.266667
	No	15.899167	18.78	15.980000	16.540000
				22.750	19.003846
					20.824286

In [102]: df.pivot_table(index=['sex','smoker'],columns=['day','time'])

Out[102]:

	tip								total_bill								Fri		Sat		
	Fri		Sat		Sun		Thur		Fri		Sat		Sun		Thur		Fri		Sat		
	Dinner	Lunch	Dinner	Lunch	Dinner	Lunch	Dinner	Dinner	Lunch	Dinner	Lunch	Dinner	Dinner	Lunch	Dinner	Lunch	Dinner	Lunch	Dinner	Lunch	
	NaN	1.666667	2.4	2.629630	2.600000	3.058000	NaN	1.90	3.246	2.879259	3.521333	19.171000	NaN	11.386667	25.892	21.837778					
	NaN	NaN	2.0	2.656250	2.883721	2.941500	NaN	NaN	2.500	3.256563	3.115349	18.486500	NaN	NaN	17.475	19.929063					
	NaN	2.000000	2.0	2.200000	2.500000	2.990000	NaN	2.66	2.700	2.868667	3.500000	19.218571	NaN	13.260000	12.200	20.266667					
		2.0	3.000000	2.0	2.307692	3.071429	2.437083	3.0	3.00	3.250	2.724615	3.329286	15.899167	18.78	15.980000	22.750	19.003846				

In [103]: `df.pivot_table(index=['sex','smoker'],columns=['day','time'],aggfunc={'size':'mean','tip':'max','total_bill':'sum'})`

Out[103]:

		size				tip				total_bill							
day	Thur	Fri		Sat		Sun		Thur		Fri		Sat		Sun		Thur	
		time	Lunch	Dinner	Lunch	Dinner	Dinner	Lunch	Dinner	Lunch	Dinner	Dinner	Lunch	Dinner	Lunch	Dinner	Lunch
sex	smoker																
Male	Yes	2.300000	NaN	1.666667	2.4	2.629630	2.600000	5.00	NaN	2.20	4.73	10.00	6.5	191.71	0.00	34.16	12
	No	2.500000	NaN	NaN	2.0	2.656250	2.883721	6.70	NaN	NaN	3.50	9.00	6.0	369.73	0.00	0.00	3
Female	Yes	2.428571	NaN	2.000000	2.0	2.200000	2.500000	5.00	NaN	3.48	4.30	6.50	4.0	134.53	0.00	39.78	4
	No	2.500000	2.0	3.000000	2.0	2.307692	3.071429	5.17	3.0	3.00	3.25	4.67	5.2	381.58	18.78	15.98	2

In [106]: `# Margins.`

`df.pivot_table(index='sex',columns= 'smoker',values = 'total_bill',aggfunc='sum',margins=True)`

Out[106]:

smoker	Yes	No	All
sex			
Male	1337.07	1919.75	3256.82
Female	593.27	977.68	1570.95
All	1930.34	2897.43	4827.77

In [108]: `# Plotting Graphs.`

`expense= pd.read_csv("expense_data.csv")`

In [110]: `expense.head(2)`

Out[110]:

	Date	Account	Category	Subcategory	Note	INR	Income/Expense	Note.1	Amount	Currency	Account.1	
0	3/2/2022 10:11	CUB - online payment	Food		NaN	Brownie	50.0	Expense	NaN	50.0	INR	50.0
1	3/2/2022 10:11	CUB - online payment	Other		NaN	To lended people	300.0	Expense	NaN	300.0	INR	300.0

In [115]: `# Categories`

`expense['Category'].value_counts()`

Out[115]:

Food	156
Other	60
Transportation	31
Apparel	7
Household	6
Allowance	6
Social Life	5
Education	1
Salary	1
Self-development	1
Beauty	1
Gift	1
Petty cash	1

Name: Category, dtype: int64

In [117]: `expense.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 277 entries, 0 to 276
Data columns (total 11 columns):
 #   Column      Non-Null Count  Dtype  
--- 
 0   Date        277 non-null    object  
 1   Account     277 non-null    object  
 2   Category    277 non-null    object  
 3   Subcategory 0 non-null     float64 
 4   Note         273 non-null    object  
 5   INR          277 non-null    float64 
 6   Income/Expense 277 non-null  object  
 7   Note.1       0 non-null     float64 
 8   Amount       277 non-null    float64 
 9   Currency     277 non-null    object  
 10  Account.1   277 non-null    float64 
dtypes: float64(5), object(6)
memory usage: 23.9+ KB
```

In [119]: `# Converting integer to daytime format`
`expense['Date'] = pd.to_datetime(expense['Date'])`

In [121]: `expense.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 277 entries, 0 to 276
Data columns (total 11 columns):
 #   Column      Non-Null Count  Dtype    
--- 
 0   Date        277 non-null    datetime64[ns]
 1   Account     277 non-null    object  
 2   Category    277 non-null    object  
 3   Subcategory 0 non-null     float64  
 4   Note         273 non-null    object  
 5   INR          277 non-null    float64  
 6   Income/Expense 277 non-null  object  
 7   Note.1       0 non-null     float64  
 8   Amount       277 non-null    float64  
 9   Currency     277 non-null    object  
 10  Account.1   277 non-null    float64 
dtypes: datetime64[ns](1), float64(5), object(5)
memory usage: 23.9+ KB
```

In [123]: `# Extracting month from the date column`
`expense['Date'].dt.month_name()`

Out[123]:

0	March
1	March
2	March
3	March
4	March
...	
272	November
273	November
274	November
275	November
276	November

Name: Date, Length: 277, dtype: object

In [124]: `expense['month']= expense['Date'].dt.month_name()`

In [125]: `expense.head(2)`

Out[125]:

	Date	Account	Category	Subcategory	Note	INR	Income/Expense	Note.1	Amount	Currency	Account.1	month
0	2022-03-02 10:11:00	CUB - online payment	Food	NaN	Brownie	50.0	Expense	NaN	50.0	INR	50.0	March
1	2022-03-02 10:11:00	CUB - online payment	Other	NaN	To lended people	300.0	Expense	NaN	300.0	INR	300.0	March

In [126]: # Using pivot table

expense.pivot_table(index ='month', columns='Category', values ='INR', aggfunc='sum')

Out[126]:

Category	Allowance	Apparel	Beauty	Education	Food	Gift	Household	Other	Petty cash	Salary	Self-development	Social Life	Transportation
month													
December	11000.0	2590.0	196.0		6440.72	Nan	4800.0	1790.0	Nan	Nan	400.0	513.72	914.0
February	Nan	798.0	Nan		5579.85	Nan	2808.0	20000.0	Nan	Nan	Nan	1800.00	5078.8
January	1000.0	Nan	Nan	1400.0	9112.51	Nan	4580.0	13178.0	Nan	8000.0	Nan	200.00	2850.0
March	Nan	Nan	Nan		195.00	Nan	Nan	900.0	Nan	Nan	Nan	Nan	30.0
November	2000.0	Nan	Nan		3174.40	115.0	Nan	2000.0	3.0	Nan	Nan	Nan	331.0

In [128]: # fill values of NAN

expense.pivot_table(index ='month', columns='Category', values ='INR', aggfunc='sum',fill_value =0)

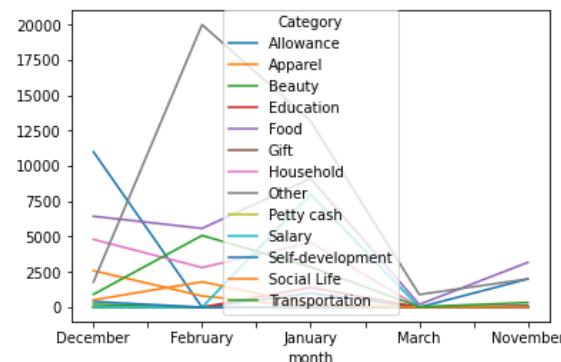
Out[128]:

Category	Allowance	Apparel	Beauty	Education	Food	Gift	Household	Other	Petty cash	Salary	Self-development	Social Life	Transportation
month													
December	11000	2590	196	0	6440.72	0	4800	1790	0	0	400	513.72	914.0
February	0	798	0	0	5579.85	0	2808	20000	0	0	0	1800.00	5078.8
January	1000	0	0	1400	9112.51	0	4580	13178	0	8000	0	200.00	2850.0
March	0	0	0	0	195.00	0	0	900	0	0	0	0.00	30.0
November	2000	0	0	0	3174.40	115	0	2000	3	0	0	0.00	331.0

In [131]: # plot

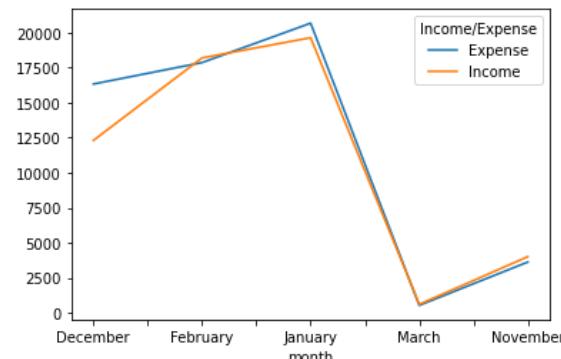
expense.pivot_table(index ='month', columns='Category', values ='INR', aggfunc='sum',fill_value =0).plot()

Out[131]: <AxesSubplot:xlabel='month'>



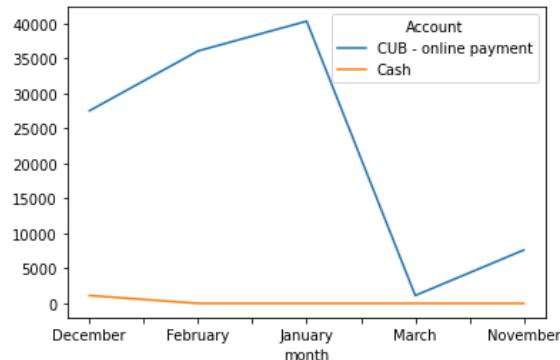
In [132]: expense.pivot_table(index ='month', columns='Income/Expense', values ='INR', aggfunc='sum',fill_value =0).plot()

Out[132]: <AxesSubplot:xlabel='month'>



```
In [133]: expense.pivot_table(index ='month', columns='Account', values ='INR', aggfunc='sum',fill_value =0).plot()
```

```
Out[133]: <AxesSubplot:xlabel='month'>
```



```
In [ ]:
```

```
In [1]: import pandas as pd
import numpy as np
```

What are vectorized operations

vectorized operations are a way to perform operations on entire arrays of data at once, which is faster than doing them one at a time.

```
In [2]: a = np.array([1,2,3,4])
a * 4
```

```
Out[2]: array([ 4,  8, 12, 16])
```

```
In [3]: # problem in vectorized operations in vanilla python
s = ['cat','mat',None,'rat']
[i.startswith('c') for i in s]
## Throws an error , because Startswith only works on strings
```

```
In [4]: # How pandas solves this issue?
```

```
s = pd.Series(['cat','mat',None,'rat'])
s
```

```
Out[4]: 0      cat
        1      mat
        2    None
        3      rat
       dtype: object
```

here , **str = string accessor**

```
In [5]: s.str.startswith('c') # Fast and optimized for larger datasets.
```

```
Out[5]: 0    True
        1   False
        2    None
        3   False
       dtype: object
```

```
In [6]: # import titanic dataset
df =pd.read_csv("titanic.csv")
```

In [7]: `df.head(1)`

Out[7]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0		1	0	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.25	NaN	

In [8]: `df['Name']`

Out[8]:

0	Braund, Mr. Owen Harris
1	Cumings, Mrs. John Bradley (Florence Briggs Th... Heikkinen, Miss. Laina
2	Futrelle, Mrs. Jacques Heath (Lily May Peel)
3	Allen, Mr. William Henry
4	...
886	Montvila, Rev. Juozas
887	Graham, Miss. Margaret Edith
888	Johnston, Miss. Catherine Helen "Carrie"
889	Behr, Mr. Karl Howell
890	Dooley, Mr. Patrick

Name: Name, Length: 891, dtype: object

Common Functions

`lower/upper/capitalize/title`

In [9]:

```
# Upper
df['Name'].str.upper() # converts into Capital Words
```

Out[9]:

0	BRAUND, MR. OWEN HARRIS
1	CUMINGS, MRS. JOHN BRADLEY (FLORENCE BRIGGS TH... HEIKKINEN, MISS. LAINA
2	FUTRELLE, MRS. JACQUES HEATH (LILY MAY PEEL)
3	ALLEN, MR. WILLIAM HENRY
4	...
886	MONTVILA, REV. JUOZAS
887	GRAHAM, MISS. MARGARET EDITH
888	JOHNSTON, MISS. CATHERINE HELEN "CARRIE"
889	BEHR, MR. KARL HOWELL
890	DOOLEY, MR. PATRICK

Name: Name, Length: 891, dtype: object

In [10]: # Lower
df['Name'].str.lower() # converts into small words

Out[10]: 0 braund, mr. owen harris
1 cumings, mrs. john bradley (florence briggs th...
2 heikkinen, miss. laina
3 futrelle, mrs. jacques heath (lily may peel)
4 allen, mr. william henry
...
886 montvila, rev. juozas
887 graham, miss. margaret edith
888 johnston, miss. catherine helen "carrie"
889 behr, mr. karl howell
890 dooley, mr. patrick
Name: Name, Length: 891, dtype: object

In [11]: # title
df['Name'].str.title() # converts into starting letter of Word to Capital

Out[11]: 0 Braund, Mr. Owen Harris
1 Cumings, Mrs. John Bradley (Florence Briggs Th...
2 Heikkinen, Miss. Laina
3 Futrelle, Mrs. Jacques Heath (Lily May Peel)
4 Allen, Mr. William Henry
...
886 Montvila, Rev. Juozas
887 Graham, Miss. Margaret Edith
888 Johnston, Miss. Catherine Helen "Carrie"
889 Behr, Mr. Karl Howell
890 Dooley, Mr. Patrick
Name: Name, Length: 891, dtype: object

In [12]: # Lets try to find the longest name In the passengers
df['Name'].str.len().max()

Out[12]: 82

In [13]: df['Name'][df['Name'].str.len()== 82]

Out[13]: 307 Penasco y Castellana, Mrs. Victor de Satode (M...
Name: Name, dtype: object

In [14]: df['Name'][df['Name'].str.len()== 82].values[0]

Out[14]: 'Penasco y Castellana, Mrs. Victor de Satode (Maria Josefa Perez de Soto y Vallejo)'

strip

In [16]: 'jack'.strip()

Out[16]: 'jack'

In [17]: df['Name'].str.strip() # removes spaces

Out[17]: 0 Braund, Mr. Owen Harris
 1 Cummings, Mrs. John Bradley (Florence Briggs Th...
 2 Heikkinen, Miss. Laina
 3 Futrelle, Mrs. Jacques Heath (Lily May Peel)
 4 Allen, Mr. William Henry
 ...
 886 Montvila, Rev. Juozas
 887 Graham, Miss. Margaret Edith
 888 Johnston, Miss. Catherine Helen "Carrie"
 889 Behr, Mr. Karl Howell
 890 Dooley, Mr. Patrick
 Name: Name, Length: 891, dtype: object

split

In [19]: # split
 df['Name'].str.split(',')

Out[19]: 0 [Braund, Mr. Owen Harris]
 1 [Cummings, Mrs. John Bradley (Florence Briggs ...
 2 [Heikkinen, Miss. Laina]
 3 [Futrelle, Mrs. Jacques Heath (Lily May Peel)]
 4 [Allen, Mr. William Henry]
 ...
 886 [Montvila, Rev. Juozas]
 887 [Graham, Miss. Margaret Edith]
 888 [Johnston, Miss. Catherine Helen "Carrie"]
 889 [Behr, Mr. Karl Howell]
 890 [Dooley, Mr. Patrick]
 Name: Name, Length: 891, dtype: object

In [21]: # Split -> get
 df['Name'].str.split(',').str.get(0)

Out[21]: 0 Braund
 1 Cummings
 2 Heikkinen
 3 Futrelle
 4 Allen
 ...
 886 Montvila
 887 Graham
 888 Johnston
 889 Behr
 890 Dooley
 Name: Name, Length: 891, dtype: object

In [22]: `df['Name'].str.split(',') .str.get(1)`

Out[22]:

0	Mr. Owen Harris
1	Mrs. John Bradley (Florence Briggs Thayer)
2	Miss. Laina
3	Mrs. Jacques Heath (Lily May Peel)
4	Mr. William Henry
...	
886	Rev. Juozas
887	Miss. Margaret Edith
888	Miss. Catherine Helen "Carrie"
889	Mr. Karl Howell
890	Mr. Patrick

Name: Name, Length: 891, dtype: object

In [23]: `df['last_name'] = df['Name'].str.split(',') .str.get(0)`

In [25]: `df.head(1)`

Out[25]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.25	NaN	

In [29]: *# it is used to split the Name column of the DataFrame df into two columns # FirstName and LastName.*

```
df['Name'].str.split(',') .str.get(1).str.strip().str.split(' ',n=1, expand=True)
```

Out[29]:

	0	1
0	Mr.	Owen Harris
1	Mrs. John Bradley (Florence Briggs Thayer)	
2	Miss.	Laina
3	Mrs. Jacques Heath (Lily May Peel)	
4	Mr.	William Henry
...
886	Rev.	Juozas
887	Miss.	Margaret Edith
888	Miss.	Catherine Helen "Carrie"
889	Mr.	Karl Howell
890	Mr.	Patrick

891 rows × 2 columns

In [30]: `df[['title', 'firstname']] = df['Name'].str.split(',').str.get(1).str.strip().st`

In [31]: `df.head(1)`

Out[31]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Emba
0		1	0	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.25	NaN	

◀ ▶

In [33]: `# Number of titles`

`df['title'].value_counts()`

Out[33]:

Mr.	517
Miss.	182
Mrs.	125
Master.	40
Dr.	7
Rev.	6
Mlle.	2
Major.	2
Col.	2
the	1
Capt.	1
Ms.	1
Sir.	1
Lady.	1
Mme.	1
Don.	1
Jonkheer.	1
Name: title, dtype: int64	

replace

In [36]: `df['title'] = df['title'].str.replace('Ms.', 'Miss.')
df['title'] = df['title'].str.replace('Mlle.', 'Miss.')`

C:\Users\user\AppData\Local\Temp\ipykernel_15952\1805277261.py:1: FutureWarning: The default value of regex will change from True to False in a future version.

`df['title'] = df['title'].str.replace('Ms.', 'Miss.')`

C:\Users\user\AppData\Local\Temp\ipykernel_15952\1805277261.py:2: FutureWarning: The default value of regex will change from True to False in a future version.

`df['title'] = df['title'].str.replace('Mlle.', 'Miss.')`

```
In [37]: df['title'].value_counts()
```

```
Out[37]: Mr.          517
Miss.        185
Mrs.         125
Master.       40
Dr.           7
Rev.          6
Major.         2
Col.           2
Don.           1
Mme.           1
Lady.          1
Sir.           1
Capt.          1
the            1
Jonkheer.      1
Name: title, dtype: int64
```

Filtering

In [38]: *# startswith/endswith*

df[df['firstname'].str.startswith('A')]

Out[38]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
13	14	0	3	Andersson, Mr. Anders Johan	male	39.0	1	5	347082	31.2750
22	23	1	3	McGowan, Miss. Anna "Annie"	female	15.0	0	0	330923	8.0292
35	36	0	1	Holverson, Mr. Alexander Oskar	male	42.0	1	0	113789	52.0000
38	39	0	3	Vander Planke, Miss. Augusta Maria	female	18.0	2	0	345764	18.0000
61	62	1	1	Icard, Miss. Amelie	female	38.0	0	0	113572	80.0000
...
842	843	1	1	Serepeca, Miss. Augusta	female	30.0	0	0	113798	31.0000
845	846	0	3	Abbing, Mr. Anthony	male	42.0	0	0	C.A. 5547	7.5500
866	867	1	2	Duran y More, Miss. Asuncion	female	27.0	1	0	SC/PARIS 2149	13.8583
875	876	1	3	Najib, Miss. Adele Kiamie "Jane"	female	15.0	0	0	2667	7.2250
876	877	0	3	Gustafsson, Mr. Alfred Ossian	male	20.0	0	0	7534	9.8458

95 rows × 15 columns



In [40]: # ends with

```
df[df['firstname'].str.endswith('z')]
```

Out[40]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin
69	70	0	3	Kink, Mr. Vincenz	male	26.0	2	0	315151	8.6625	Na
679	680	1	1	Cardeza, Mr. Thomas Drake Martinez	male	36.0	0	1	PC 17755	512.3292	B5
721	722	0	3	Jensen, Mr. Svend Lauritz	male	17.0	1	0	350048	7.0542	Na

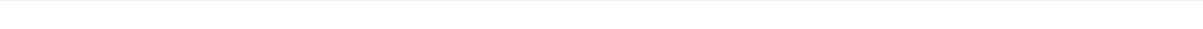


In [41]: # isdigit/isalpha...

```
df[df['firstname'].str.isdigit()]
```

Out[41]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarke



regex

In [42]: # applying regex

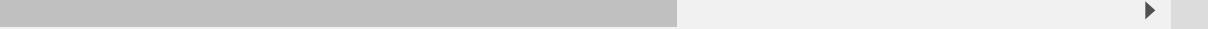
contains

search john -> both case

```
df[df['firstname'].str.contains('john', case=False)]
```

Out[42]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th... Turpin, Mrs. William John Robert (Dorothy Ann ... Rogers, Mr. William John	female	38.0 27.0 NaN	1 1 0	0 0 0	PC 17599 11668 S.C./A.4. 23567
41	42	0	2	Doling, Mrs. John T (Ada Julia Bone)	female	34.0	0	1	231919
45	46	0	3						
98	99	1	2						



In [44]: # find Lastnames with start and end char vowel (aeiou)
df[df['last_name'].str.contains('^[^aeiouAEIOU].+[^aeiouAEIOU]\$')]]

Out[44]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...)	female	38.0	1	0	PC 17599	71.2833
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250
5	6	0	3	Moran, Mr. James	male	NaN	0	0	330877	8.4583
6	7	0	1	McCarthy, Mr. Timothy J	male	54.0	0	0	17463	51.8625
...
884	885	0	3	Suttehall, Mr. Henry Jr	male	25.0	0	0	SOTON/OQ 392076	7.0500
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.0000
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.4500
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.0000
890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.7500

671 rows × 15 columns



slicing

In [45]: `df['Name'].str[:4] # first 4 characters`

Out[45]:

0	Brau
1	Cumi
2	Heik
3	Futr
4	Alle
	...
886	Mont
887	Grah
888	John
889	Behr
890	Dool

Name: Name, Length: 891, dtype: object

In [46]: `df['Name'].str[::-2] # alternate characters`

Out[46]:

0	Ban,M.Oe ars
1	Cmns r.Jh rde Foec rgsTae)
2	Hiknn is an
3	Ftel,Ms aqe et Ll a el
4	Aln r ila er
	...
886	Mnvl,Rv uzs
887	Gaa,Ms.Mrae dt
888	Jhso,Ms.CteieHln"are
889	Bh,M.Kr oel
890	Doe,M.Ptik

Name: Name, Length: 891, dtype: object

In [47]: `df['Name'].str[::-1] # reverse`

Out[47]:

0	sirraH newO .rM ,dnuarB
1)reyahT sggirB ecnerolF(yeldarB nhoJ .srM ,sg...
2	aniaL .ssiM ,nenikkieH
3)leeP yaM yliL(htaeH seuqcaJ .srM ,ellertuF
4	yrneH mailliW .rM ,nellA
	...
886	sazouJ .veR ,alivtnoM
887	htidE teragraM .ssiM ,maharG
888	"eirraC" neleH enirehtaC .ssiM ,notsnhoJ
889	llewoH lraK .rM ,rheB
890	kcirtaP .rM ,yelooD

Name: Name, Length: 891, dtype: object

In []:

```
In [1]: import numpy as np
import pandas as pd
```

Timestamp Object

Time stamps reference particular moments in time (e.g., Oct 24th, 2022 at 7:00pm)

Vectorized date and time operations are a powerful tool for working with date and time data. They can be used to quickly and easily perform a wide variety of operations on date and time data.

Creating Timestamp objects

```
In [2]: # creating a timestamp
pd.Timestamp('2023/05/12')
# This time stamp contains year-month- Day ,hour-minute-Second
```

```
Out[2]: Timestamp('2023-05-12 00:00:00')
```

```
In [3]: # type
type(pd.Timestamp('2023/05/12'))
```

```
Out[3]: pandas._libs.tslibs.timestamps.Timestamp
```

```
In [4]: # Variations
pd.Timestamp('2023-05-12')
```

```
Out[4]: Timestamp('2023-05-12 00:00:00')
```

```
In [5]: pd.Timestamp('2023,05,12')
```

```
Out[5]: Timestamp('2023-12-01 00:00:00')
```

```
In [6]: pd.Timestamp('2023.05.12')
```

```
Out[6]: Timestamp('2023-05-12 00:00:00')
```

```
In [7]: # only year
pd.Timestamp('2023') # It Automatically assigns First day of the year.
```

```
Out[7]: Timestamp('2023-01-01 00:00:00')
```

```
In [8]: # Using text
pd.Timestamp('12th May 2023')
```

```
Out[8]: Timestamp('2023-05-12 00:00:00')
```

```
In [9]: # Provide time also
pd.Timestamp('12th May 2023 4:40PM')
```

```
Out[9]: Timestamp('2023-05-12 16:40:00')
```

```
In [10]: # using Python's datetime object
import datetime as dt

x = pd.Timestamp(dt.datetime(2023,5,12,4,42,56))
x
```

Out[10]: Timestamp('2023-05-12 04:42:56')

```
In [12]: # Fetching attributes
x.year
```

Out[12]: 2023

```
In [15]: x.day
```

Out[15]: 12

```
In [17]: x.time()
```

Out[17]: datetime.time(4, 42, 56)

```
In [18]: x.month
```

Out[18]: 5

why separate objects to handle data and time when python already has datetime functionality?

- syntax wise datetime is very convenient
- But the performance takes a hit while working with huge data. List vs Numpy Array
- The weaknesses of Python's datetime format inspired the NumPy team to add a set of native time series data type to NumPy.
- The datetime64 dtype encodes dates as 64-bit integers, and thus allows arrays of dates to be represented very compactly.

```
In [19]: import numpy as np
date = np.array('2023-05-12', dtype=np.datetime64)
date
```

Out[19]: array('2023-05-12', dtype='datetime64[D]')

```
In [21]: # We can operate vector operations on Date.
date + np.arange(20)
```

```
Out[21]: array(['2023-05-12', '2023-05-13', '2023-05-14', '2023-05-15',
               '2023-05-16', '2023-05-17', '2023-05-18', '2023-05-19',
               '2023-05-20', '2023-05-21', '2023-05-22', '2023-05-23',
               '2023-05-24', '2023-05-25', '2023-05-26', '2023-05-27',
               '2023-05-28', '2023-05-29', '2023-05-30', '2023-05-31'],
              dtype='datetime64[D]')
```

- Because of the uniform type in NumPy datetime64 arrays, this type of operation can be accomplished much more quickly than if we were working directly with Python's datetime objects, especially as arrays get large
- Pandas **Timestamp** object combines the ease-of-use of python datetime with the efficient storage and vectorized interface of numpy.datetime64
- From a group of these Timestamp objects, Pandas can construct a DatetimeIndex that can be used to index data in a Series or DataFrame

DatetimeIndex Object

A collection of pandas timestamp

```
In [22]: # using strings
pd.DatetimeIndex(['2023/05/12','2023/01/01','2025/01/22'])

Out[22]: DatetimeIndex(['2023-05-12', '2023-01-01', '2025-01-22'], dtype='datetime64[ns]', freq=None)

In [23]: pd.DatetimeIndex(['2023/05/12','2023/01/01','2025/01/22'])[0]

Out[23]: Timestamp('2023-05-12 00:00:00')

In [25]: # type
type(pd.DatetimeIndex(['2023/05/12','2023/01/01','2025/01/22']))

Out[25]: pandas.core.indexes.datetimes.DatetimeIndex
```

To store a **single date**, we use `Timestamp`.

And to store **multiple date** and time we use date time index.

```
In [26]: # using python datetime object
pd.DatetimeIndex([dt.datetime(2023,5,12),dt.datetime(2023,1,1),dt.datetime(2025,1,1)])

Out[26]: DatetimeIndex(['2023-05-12', '2023-01-01', '2025-01-01'], dtype='datetime64[ns]', freq=None)

In [27]: # using pd.timestamps
dt_index = pd.DatetimeIndex([pd.Timestamp(2023,1,1),pd.Timestamp(2022,1,1),pd.Timestamp(2021,1,1)])

In [28]: dt_index

Out[28]: DatetimeIndex(['2023-01-01', '2022-01-01', '2021-01-01'], dtype='datetime64[ns]', freq=None)

In [29]: # using datatimeindex as series index

pd.Series([1,2,3],index=dt_index)

Out[29]: 2023-01-01    1
2022-01-01    2
2021-01-01    3
dtype: int64
```

date_range function

```
In [33]: # generate daily dates in a given range
pd.date_range(start='2023/5/12',end='2023/6/12',freq='D')

Out[33]: DatetimeIndex(['2023-05-12', '2023-05-13', '2023-05-14', '2023-05-15',
'2023-05-16', '2023-05-17', '2023-05-18', '2023-05-19',
'2023-05-20', '2023-05-21', '2023-05-22', '2023-05-23',
'2023-05-24', '2023-05-25', '2023-05-26', '2023-05-27',
'2023-05-28', '2023-05-29', '2023-05-30', '2023-05-31',
'2023-06-01', '2023-06-02', '2023-06-03', '2023-06-04',
'2023-06-05', '2023-06-06', '2023-06-07', '2023-06-08',
'2023-06-09', '2023-06-10', '2023-06-11', '2023-06-12'],
dtype='datetime64[ns]', freq='D')
```

```
In [34]: # Alternate days
pd.date_range(start='2023/5/12', end='2023/6/12', freq='2D')
```

```
Out[34]: DatetimeIndex(['2023-05-12', '2023-05-14', '2023-05-16', '2023-05-18',
 '2023-05-20', '2023-05-22', '2023-05-24', '2023-05-26',
 '2023-05-28', '2023-05-30', '2023-06-01', '2023-06-03',
 '2023-06-05', '2023-06-07', '2023-06-09', '2023-06-11'],
 dtype='datetime64[ns]', freq='2D')
```

```
In [35]: # 2 days gap
pd.date_range(start='2023/5/12', end='2023/6/12', freq='3D')
```

```
Out[35]: DatetimeIndex(['2023-05-12', '2023-05-15', '2023-05-18', '2023-05-21',
 '2023-05-24', '2023-05-27', '2023-05-30', '2023-06-02',
 '2023-06-05', '2023-06-08', '2023-06-11'],
 dtype='datetime64[ns]', freq='3D')
```

```
In [36]: # B -> business days (MON-FRI)
pd.date_range(start='2023/5/12', end='2023/6/12', freq='B')
```

```
Out[36]: DatetimeIndex(['2023-05-12', '2023-05-15', '2023-05-16', '2023-05-17',
 '2023-05-18', '2023-05-19', '2023-05-22', '2023-05-23',
 '2023-05-24', '2023-05-25', '2023-05-26', '2023-05-29',
 '2023-05-30', '2023-05-31', '2023-06-01', '2023-06-02',
 '2023-06-05', '2023-06-06', '2023-06-07', '2023-06-08',
 '2023-06-09', '2023-06-12'],
 dtype='datetime64[ns]', freq='B')
```

```
In [37]: # W -> one week per day (SUN)
pd.date_range(start='2023/5/12', end='2023/6/12', freq='W')
```

```
Out[37]: DatetimeIndex(['2023-05-14', '2023-05-21', '2023-05-28', '2023-06-04',
 '2023-06-11'],
 dtype='datetime64[ns]', freq='W-SUN')
```

```
In [38]: # if you want specific Day (THU)
pd.date_range(start='2023/5/12', end='2023/6/12', freq='w-THU')
```

```
Out[38]: DatetimeIndex(['2023-05-18', '2023-05-25', '2023-06-01', '2023-06-08'],
 dtype='datetime64[ns]', freq='W-THU')
```

```
In [39]: # H -> Hourly data(factor)
pd.date_range(start='2023/5/12', end='2023/6/12', freq='H')
```

```
Out[39]: DatetimeIndex(['2023-05-12 00:00:00', '2023-05-12 01:00:00',
 '2023-05-12 02:00:00', '2023-05-12 03:00:00',
 '2023-05-12 04:00:00', '2023-05-12 05:00:00',
 '2023-05-12 06:00:00', '2023-05-12 07:00:00',
 '2023-05-12 08:00:00', '2023-05-12 09:00:00',
 ...
 '2023-06-11 15:00:00', '2023-06-11 16:00:00',
 '2023-06-11 17:00:00', '2023-06-11 18:00:00',
 '2023-06-11 19:00:00', '2023-06-11 20:00:00',
 '2023-06-11 21:00:00', '2023-06-11 22:00:00',
 '2023-06-11 23:00:00', '2023-06-12 00:00:00'],
 dtype='datetime64[ns]', length=745, freq='H')
```

```
In [41]: # For every six hours
pd.date_range(start='2023/5/12', end='2023/6/12', freq='6H')
```

```
Out[41]: DatetimeIndex(['2023-05-12 00:00:00', '2023-05-12 06:00:00',
 '2023-05-12 12:00:00', '2023-05-12 18:00:00',
 '2023-05-13 00:00:00', '2023-05-13 06:00:00',
 '2023-05-13 12:00:00', '2023-05-13 18:00:00',
 '2023-05-14 00:00:00', '2023-05-14 06:00:00',
 ...
 '2023-06-09 18:00:00', '2023-06-10 00:00:00',
 '2023-06-10 06:00:00', '2023-06-10 12:00:00',
 '2023-06-10 18:00:00', '2023-06-11 00:00:00',
 '2023-06-11 06:00:00', '2023-06-11 12:00:00',
 '2023-06-11 18:00:00', '2023-06-12 00:00:00'],
 dtype='datetime64[ns]', length=125, freq='6H')
```

```
In [42]: # M -> Month end
pd.date_range(start='2023/5/12', end='2023/6/12', freq='M')
```

```
Out[42]: DatetimeIndex(['2023-05-31'], dtype='datetime64[ns]', freq='M')
```

```
In [47]: # MS -> Month start
pd.date_range(start='2023/5/12', end='2028/6/12', freq='MS')
```

```
Out[47]: DatetimeIndex(['2023-06-01', '2023-07-01', '2023-08-01', '2023-09-01',
 '2023-10-01', '2023-11-01', '2023-12-01', '2024-01-01',
 '2024-02-01', '2024-03-01', '2024-04-01', '2024-05-01',
 '2024-06-01', '2024-07-01', '2024-08-01', '2024-09-01',
 '2024-10-01', '2024-11-01', '2024-12-01', '2025-01-01',
 '2025-02-01', '2025-03-01', '2025-04-01', '2025-05-01',
 '2025-06-01', '2025-07-01', '2025-08-01', '2025-09-01',
 '2025-10-01', '2025-11-01', '2025-12-01', '2026-01-01',
 '2026-02-01', '2026-03-01', '2026-04-01', '2026-05-01',
 '2026-06-01', '2026-07-01', '2026-08-01', '2026-09-01',
 '2026-10-01', '2026-11-01', '2026-12-01', '2027-01-01',
 '2027-02-01', '2027-03-01', '2027-04-01', '2027-05-01',
 '2027-06-01', '2027-07-01', '2027-08-01', '2027-09-01',
 '2027-10-01', '2027-11-01', '2027-12-01', '2028-01-01',
 '2028-02-01', '2028-03-01', '2028-04-01', '2028-05-01',
 '2028-06-01'],
 dtype='datetime64[ns]', freq='MS')
```

```
In [46]: # A -> Year end
pd.date_range(start='2023/5/12', end='2030/6/12', freq='A')
```

```
Out[46]: DatetimeIndex(['2023-12-31', '2024-12-31', '2025-12-31', '2026-12-31',
 '2027-12-31', '2028-12-31', '2029-12-31'],
 dtype='datetime64[ns]', freq='A-DEC')
```

```
In [49]: # using periods(number of results)
pd.date_range(start='2023/5/12', periods = 30, freq='D')
```

```
Out[49]: DatetimeIndex(['2023-05-12', '2023-05-13', '2023-05-14', '2023-05-15',
 '2023-05-16', '2023-05-17', '2023-05-18', '2023-05-19',
 '2023-05-20', '2023-05-21', '2023-05-22', '2023-05-23',
 '2023-05-24', '2023-05-25', '2023-05-26', '2023-05-27',
 '2023-05-28', '2023-05-29', '2023-05-30', '2023-05-31',
 '2023-06-01', '2023-06-02', '2023-06-03', '2023-06-04',
 '2023-06-05', '2023-06-06', '2023-06-07', '2023-06-08',
 '2023-06-09', '2023-06-10'],
 dtype='datetime64[ns]', freq='D')
```

```
In [50]: # Hour (using periods)
pd.date_range(start='2023/5/12', periods =30,freq='H')
```

```
Out[50]: DatetimeIndex(['2023-05-12 00:00:00', '2023-05-12 01:00:00',
 '2023-05-12 02:00:00', '2023-05-12 03:00:00',
 '2023-05-12 04:00:00', '2023-05-12 05:00:00',
 '2023-05-12 06:00:00', '2023-05-12 07:00:00',
 '2023-05-12 08:00:00', '2023-05-12 09:00:00',
 '2023-05-12 10:00:00', '2023-05-12 11:00:00',
 '2023-05-12 12:00:00', '2023-05-12 13:00:00',
 '2023-05-12 14:00:00', '2023-05-12 15:00:00',
 '2023-05-12 16:00:00', '2023-05-12 17:00:00',
 '2023-05-12 18:00:00', '2023-05-12 19:00:00',
 '2023-05-12 20:00:00', '2023-05-12 21:00:00',
 '2023-05-12 22:00:00', '2023-05-12 23:00:00',
 '2023-05-13 00:00:00', '2023-05-13 01:00:00',
 '2023-05-13 02:00:00', '2023-05-13 03:00:00',
 '2023-05-13 04:00:00', '2023-05-13 05:00:00'],
 dtype='datetime64[ns]', freq='H')
```

```
In [51]: # 6 Hours (using periods)
pd.date_range(start='2023/5/12', periods =30,freq='6H')
```

```
Out[51]: DatetimeIndex(['2023-05-12 00:00:00', '2023-05-12 06:00:00',
 '2023-05-12 12:00:00', '2023-05-12 18:00:00',
 '2023-05-13 00:00:00', '2023-05-13 06:00:00',
 '2023-05-13 12:00:00', '2023-05-13 18:00:00',
 '2023-05-14 00:00:00', '2023-05-14 06:00:00',
 '2023-05-14 12:00:00', '2023-05-14 18:00:00',
 '2023-05-15 00:00:00', '2023-05-15 06:00:00',
 '2023-05-15 12:00:00', '2023-05-15 18:00:00',
 '2023-05-16 00:00:00', '2023-05-16 06:00:00',
 '2023-05-16 12:00:00', '2023-05-16 18:00:00',
 '2023-05-17 00:00:00', '2023-05-17 06:00:00',
 '2023-05-17 12:00:00', '2023-05-17 18:00:00',
 '2023-05-18 00:00:00', '2023-05-18 06:00:00',
 '2023-05-18 12:00:00', '2023-05-18 18:00:00',
 '2023-05-19 00:00:00', '2023-05-19 06:00:00'],
 dtype='datetime64[ns]', freq='6H')
```

```
In [52]: # Month (using periods)
pd.date_range(start='2023/5/12', periods =30,freq='M')
```

```
Out[52]: DatetimeIndex(['2023-05-31', '2023-06-30', '2023-07-31', '2023-08-31',
 '2023-09-30', '2023-10-31', '2023-11-30', '2023-12-31',
 '2024-01-31', '2024-02-29', '2024-03-31', '2024-04-30',
 '2024-05-31', '2024-06-30', '2024-07-31', '2024-08-31',
 '2024-09-30', '2024-10-31', '2024-11-30', '2024-12-31',
 '2025-01-31', '2025-02-28', '2025-03-31', '2025-04-30',
 '2025-05-31', '2025-06-30', '2025-07-31', '2025-08-31',
 '2025-09-30', '2025-10-31'],
 dtype='datetime64[ns]', freq='M')
```

to_datetime function

converts an existing objects to pandas timestamp/datetimeindex object

```
In [59]: # simple series example
s = pd.Series(['2023/5/12', '2022/1/1', '2021/2/1'])
pd.to_datetime(s).dt.year # converting string to datetime
```

```
Out[59]: 0    2023
1    2022
2    2021
dtype: int64
```

```
In [60]: pd.to_datetime(s).dt.day
```

```
Out[60]: 0    12
1     1
2     1
dtype: int64
```

```
In [61]: pd.to_datetime(s).dt.day_name()
```

```
Out[61]: 0      Friday
1    Saturday
2    Monday
dtype: object
```

```
In [62]: pd.to_datetime(s).dt.month_name()
```

```
Out[62]: 0      May
1  January
2  February
dtype: object
```

```
In [63]: # with errors -> coerce
```

```
s = pd.Series(['2023/1/1', '2022/1/1', '2021/130/1'])
pd.to_datetime(s, errors='coerce') #NaT = Not a Time
```

```
Out[63]: 0    2023-01-01
1    2022-01-01
2        NaT
dtype: datetime64[ns]
```

```
In [64]: pd.to_datetime(s, errors='coerce').dt.year
```

```
Out[64]: 0    2023.0
1    2022.0
2        NaN
dtype: float64
```

```
In [65]: pd.to_datetime(s, errors='coerce').dt.month_name()
```

```
Out[65]: 0    January
1    January
2        NaN
dtype: object
```

Real World example

```
In [66]: df = pd.read_csv("expense_data.csv")
```

In [69]: `df.head()`

Out[69]:

	Date	Account	Category	Subcategory	Note	INR	Income/Expense	Note.1	Amount	Currency	Account.1
0	3/2/2022 10:11	CUB - online payment	Food		NaN Brownie	50.0	Expense	NaN	50.0	INR	50.0
1	3/2/2022 10:11	CUB - online payment	Other		NaN To lended people	300.0	Expense	NaN	300.0	INR	300.0
2	3/1/2022 19:50	CUB - online payment	Food		NaN Dinner	78.0	Expense	NaN	78.0	INR	78.0
3	3/1/2022 18:56	CUB - online payment	Transportation		NaN Metro	30.0	Expense	NaN	30.0	INR	30.0
4	3/1/2022 18:22	CUB - online payment	Food		NaN Snacks	67.0	Expense	NaN	67.0	INR	67.0

In [70]: `df.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 277 entries, 0 to 276
Data columns (total 11 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Date             277 non-null    object 
 1   Account          277 non-null    object 
 2   Category         277 non-null    object 
 3   Subcategory      0 non-null     float64
 4   Note             273 non-null    object 
 5   INR              277 non-null    float64
 6   Income/Expense   277 non-null    object 
 7   Note.1           0 non-null     float64
 8   Amount            277 non-null    float64
 9   Currency          277 non-null    object 
 10  Account.1        277 non-null    float64
dtypes: float64(5), object(6)
memory usage: 23.9+ KB
```

In [72]: `# converting object to date time type`
`df['Date'] = pd.to_datetime(df['Date'])`

In [73]: `df.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 277 entries, 0 to 276
Data columns (total 11 columns):
 #   Column      Non-Null Count  Dtype  
--- 
 0   Date        277 non-null    datetime64[ns]
 1   Account     277 non-null    object  
 2   Category    277 non-null    object  
 3   Subcategory 0 non-null     float64 
 4   Note         273 non-null    object  
 5   INR          277 non-null    float64 
 6   Income/Expense 277 non-null  object  
 7   Note.1       0 non-null     float64 
 8   Amount       277 non-null    float64 
 9   Currency    277 non-null    object  
 10  Account.1   277 non-null    float64 
dtypes: datetime64[ns](1), float64(5), object(5)
memory usage: 23.9+ KB
```

dt accessor

Accessor object for datetimelike properties of the Series values.

In [75]: `df['Date'].dt.year`

```
Out[75]: 0      2022
1      2022
2      2022
3      2022
4      2022
...
272    2021
273    2021
274    2021
275    2021
276    2021
Name: Date, Length: 277, dtype: int64
```

In [76]: `df['Date'].dt.month`

```
Out[76]: 0      3
1      3
2      3
3      3
4      3
..
272    11
273    11
274    11
275    11
276    11
Name: Date, Length: 277, dtype: int64
```

```
In [77]: df['Date'].dt.month_name()
```

```
Out[77]: 0      March
1      March
2      March
3      March
4      March
...
272    November
273    November
274    November
275    November
276    November
Name: Date, Length: 277, dtype: object
```

```
In [80]: df['Date'].dt.day_name()
```

```
Out[80]: 0      Wednesday
1      Wednesday
2      Tuesday
3      Tuesday
4      Tuesday
...
272    Monday
273    Monday
274    Sunday
275    Sunday
276    Sunday
Name: Date, Length: 277, dtype: object
```

```
In [86]: df['Date'].dt.is_month_end
```

```
Out[86]: 0      False
1      False
2      False
3      False
4      False
...
272    False
273    False
274    False
275    False
276    False
Name: Date, Length: 277, dtype: bool
```

```
In [87]: df['Date'].dt.is_year_end
```

```
Out[87]: 0      False
1      False
2      False
3      False
4      False
...
272    False
273    False
274    False
275    False
276    False
Name: Date, Length: 277, dtype: bool
```

```
In [90]: df['Date'].dt.is_quarter_end
```

```
Out[90]: 0      False
1      False
2      False
3      False
4      False
...
272    False
273    False
274    False
275    False
276    False
Name: Date, Length: 277, dtype: bool
```

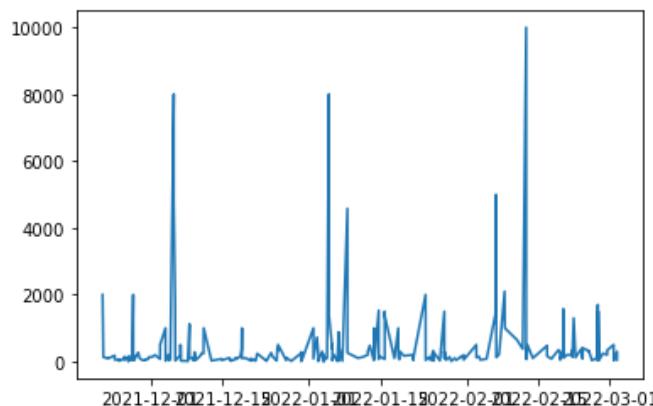
```
In [91]: df['Date'].dt.is_quarter_start
```

```
Out[91]: 0      False
1      False
2      False
3      False
4      False
...
272    False
273    False
274    False
275    False
276    False
Name: Date, Length: 277, dtype: bool
```

```
In [94]: ## Plot Graph
import matplotlib.pyplot as plt
```

```
In [95]: plt.plot(df['Date'],df['INR'])
```

```
Out[95]: []
```



```
In [96]: # Money spent day name wise (bar chart)
df['day_name'] = df['Date'].dt.day_name()
```

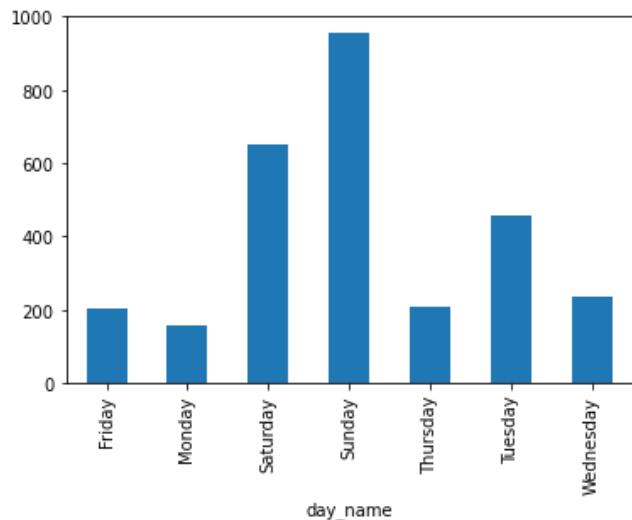
In [97]: `df.head()`

Out[97]:

	Date	Account	Category	Subcategory	Note	INR	Income/Expense	Note.1	Amount	Currency	Account.1
0	2022-03-02 10:11:00	CUB - online payment	Food		NaN Brownie	50.0	Expense	NaN	50.0	INR	50.0
1	2022-03-02 10:11:00	CUB - online payment	Other		NaN To lended people	300.0	Expense	NaN	300.0	INR	300.0
2	2022-03-01 19:50:00	CUB - online payment	Food		NaN Dinner	78.0	Expense	NaN	78.0	INR	78.0
3	2022-03-01 18:56:00	CUB - online payment	Transportation		NaN Metro	30.0	Expense	NaN	30.0	INR	30.0
4	2022-03-01 18:22:00	CUB - online payment	Food		NaN Snacks	67.0	Expense	NaN	67.0	INR	67.0

In [99]: `df.groupby('day_name')['INR'].mean().plot(kind='bar')`

Out[99]: <AxesSubplot:xlabel='day_name'>



In [100]: `# Money spent month name wise (pie chart)
df['month_name'] = df['Date'].dt.month_name()`

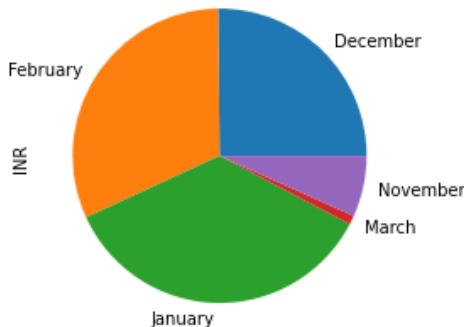
In [101]: `df.head()`

Out[101]:

	Date	Account	Category	Subcategory	Note	INR	Income/Expense	Note.1	Amount	Currency	Account.1
0	2022-03-02 10:11:00	CUB - online payment	Food		NaN Brownie	50.0	Expense	NaN	50.0	INR	50.0
1	2022-03-02 10:11:00	CUB - online payment	Other		NaN To lended people	300.0	Expense	NaN	300.0	INR	300.0
2	2022-03-01 19:50:00	CUB - online payment	Food		NaN Dinner	78.0	Expense	NaN	78.0	INR	78.0
3	2022-03-01 18:56:00	CUB - online payment	Transportation		NaN Metro	30.0	Expense	NaN	30.0	INR	30.0
4	2022-03-01 18:22:00	CUB - online payment	Food		NaN Snacks	67.0	Expense	NaN	67.0	INR	67.0

In [102]: `df.groupby('month_name')['INR'].sum().plot(kind = 'pie')`

Out[102]: <AxesSubplot:ylabel='INR'>



In [109]: `# Average
df.groupby('month_name')['INR'].mean().plot(kind = 'bar')`

Out[109]: <AxesSubplot:xlabel='month_name'>

