

# General

Pandas [API Reference](#)

Pandas [User Guide](#)

## Creating DataFrames

	a	b	c
1	4	7	10
2	5	8	11
3	6	9	12

[IO-Tools](#) from files

```
pd.read_csv("filepath")
```

```
df = pd.DataFrame(  
    {"a" : [4 ,5, 6],  
     "b" : [7, 8, 9],  
     "c" : [10, 11, 12]},  
    index = [1, 2, 3])  
Specify values for each column.
```

```
df = pd.DataFrame(  
    [[4, 7, 10],  
     [5, 8, 11],  
     [6, 9, 12]],  
    index=[1, 2, 3],  
    columns=['a', 'b', 'c'])  
Specify values for each row.
```

		a	b	c
n	v			
d	1	4	7	10
	2	5	8	11
e	2	6	9	12

```
df = pd.DataFrame(  
    {"a" : [4 ,5, 6],  
     "b" : [7, 8, 9],  
     "c" : [10, 11, 12]},  
    index = pd.MultiIndex.from_tuples(  
        [('d',1),('d',2),('e',2)],  
        names=['n','v']))  
Create DataFrame with a MultiIndex
```

## Method Chaining

Most pandas methods return a DataFrame so that another pandas method can be applied to the result. This improves readability of code.

```
df = (pd.melt(df)  
     .rename(columns={  
         'variable' : 'var',  
         'value' : 'val'})  
     .query('val >= 200'))
```

# Display & Visualize data

[Display options](#) for DataFrames:

```
pd.set_option('display.max_rows', 4)  
pd.set_option('display.max_columns', 400)  
pd.reset_option('display.max_rows')  
with pd.option_context(...):  
    only set options in „with“ codeblock
```

[Visualize Data](#) in plots

[Style options](#) of DataFrames

Cell highlight, heatmaps, ...

[seaborn.pairplot\(df, hue='column'\)](#)

Matrix of pairwise relationship  
(for [classification](#))

`df.plot.hist()`

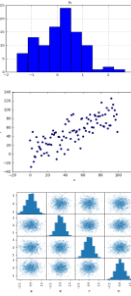
Histogram for each column

`df.plot.scatter(x='w',y='h')`

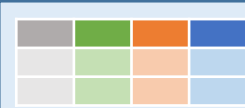
Scatter chart using pairs of points

`df.plotting.scatter_matrix(...)`

Matrix of scatter plots and histograms  
(for [regression](#))

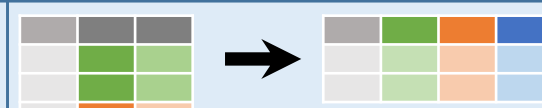


## Reshaping Data – Change layout, sorting, reindexing, renaming



`pd.melt(df)`

Gather columns into rows.



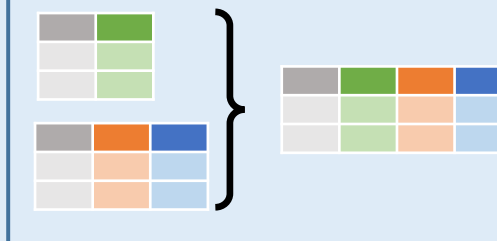
`df.pivot(columns='var', values='val')`

Spread rows into columns.



`pd.concat([df1,df2])`

Append rows of DataFrames



`pd.concat([df1,df2], axis=1)`

Append columns of DataFrames

`df.sort_values('mpg')`

Order rows by values of a column (low to high).

`df.sort_values('mpg',ascending=False)`

Order rows by values of a column (high to low).

`df.rename(columns = {'y':'year'})`

Rename the columns of a DataFrame

`df.sort_index()`

Sort the index of a DataFrame

`df.reset_index()`

Reset index of DataFrame to row numbers, moving  
index to columns.

`df.drop(columns=['Length','Height'])`

Drop columns from DataFrame

## Subset Observations – Selecting Data: rows or columns

`df[df.Length > 7]`

Extract rows that meet logical  
criteria.

`df.drop_duplicates()`

Remove duplicate rows (only  
considers columns).

`df.head(n)`

Select first n rows.

`df.tail(n)`

Select last n rows.

`df.sample(frac=0.5)`

Randomly select fraction of rows.

`df.sample(n=10)`

Randomly select n rows.

`df.iloc[row_ind, column_ind]`

`df.iloc[10:20]`

Select rows by position.

`df.iloc[:, [1,2,5]]`

Select columns in positions 1, 2 and 5 (first  
column is 0).

`df.loc[row_ind, column_ind]`

`df.loc[:, 'x2': 'x4']`

Select all columns between x2 and x4 (inclusive).

`df.loc[df['a'] > 10, ['a','c']]`

Select rows meeting logical condition, and only  
the specific columns.

`df[['width','length','species']]`

Select multiple columns with specific names.

`df['width']` or `df.width`

Select single column with specific name.

`df.filter(regex='regex')`

Select columns whose name matches regular expression *regex*.

`df.nlargest(n, 'value')`

Select and order top n entries.

`df.nsmallest(n, 'value')`

Select and order bottom n entries.

### Logic in Python (and pandas)

<	Less than	!=	Not equal to
>	Greater than	<code>df.column.isin(values)</code>	Group membership
==	Equals	<code>pd.isnull(obj)</code>	Is NaN
<=	Less than or equals	<code>pd.notnull(obj)</code>	Is not NaN
>=	Greater than or equals	<code>&amp;,  , ~, ^, df.any(), df.all()</code>	Logical and, or, not, xor, any, all

### regex (Regular Expressions) Examples

<code>'\.'</code>	Matches strings containing a period '.'
<code>'Length\$'</code>	Matches strings ending with word 'Length'
<code>'^Sepal'</code>	Matches strings beginning with the word 'Sepal'
<code>'^x[1-5]\$'</code>	Matches strings beginning with 'x' and ending with 1,2,3,4,5
<code>'^(?!Species\$).*'</code>	Matches strings except the string 'Species'

## Summarize Data

`df['w'].value_counts()`

Count number of rows with each unique value of variable

`len(df)`

# of rows in DataFrame.

`df['w'].nunique()`

# of distinct values in a column.

`df.describe()`

Basic descriptive statistics for each column (or GroupBy)

`df.shape`

Length and width of dataset



pandas provides a large set of **summary functions** that operate on different kinds of pandas objects (DataFrame columns, Series, GroupBy, Expanding and Rolling (see below)) and produce single values for each of the groups. When applied to a DataFrame, the result is returned as a pandas Series for each column. Examples:

`sum()`

Sum values of each object.

`count()`

Count non-NA/null values of each object.

`median()`

Median value of each object.

`quantile([0.25,0.75])`

Quantiles of each object.

`apply(function)`

Apply function to each object.

`min()`

Minimum value in each object.

`max()`

Maximum value in each object.

`mean()`

Mean value of each object.

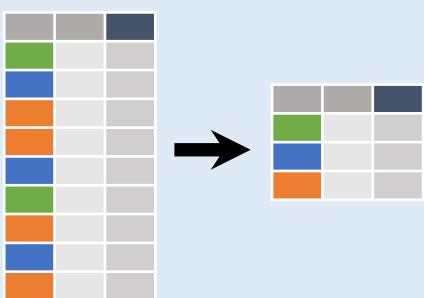
`var()`

Variance of each object.

`std()`

Standard deviation of each object.

## Group Data



`df.groupby(by="col")`

Return a GroupBy object, grouped by values in column named "col".

`df.groupby(level="ind")`

Return a GroupBy object, grouped by values in index level named "ind".

Call `.reset_index()` after grouping!

All of the summary functions listed above can be applied to a group. Additional GroupBy functions:

`size()`

Size of each group.

`agg(function)`

Aggregate group using function.

## Windows

`df.expanding()`

Return an Expanding object allowing summary functions to be applied cumulatively.

`df.rolling(n)`

Return a Rolling object allowing summary functions to be applied to windows of length n.

## Handling Missing Data

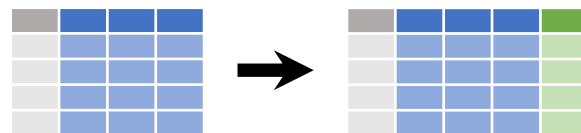
`df.dropna()`

Drop rows with any column having NA/null data.

`df.fillna(value)`

Replace all NA/null data with value.

## Make New Columns



`df.assign(Area=lambda df: df.Length*df.Height)`

Compute and append one or more new columns.

`df['Volume'] = df.Length*df.Height*df.Depth`

Add single column.

`pd.qcut(df.col, n, labels=False)`

Bin column into n buckets.



pandas provides a large set of **vector functions** that operate on all columns of a DataFrame or a single selected column (a pandas Series). These functions produce vectors of values for each of the columns, or a single Series for the individual Series. Examples:

`max(axis=1)`

Element-wise max.

`min(axis=1)`

Element-wise min.

`clip(lower=-10,upper=10)`

Trim values at input thresholds

`abs()`

Absolute value.

The examples below can also be applied to groups. In this case, the function is applied on a per-group basis, and the returned vectors are of the length of the original DataFrame.

`shift(1)`

Copy with values shifted by 1.

`rank(method='dense')`

Ranks with no gaps.

`rank(method='min')`

Ranks. Ties get min rank.

`rank(pct=True)`

Ranks rescaled to interval [0, 1].

`rank(method='first')`

Ranks. Ties go to first value.

`shift(-1)`

Copy with values lagged by 1.

`cumsum()`

Cumulative sum.

`cummax()`

Cumulative max.

`cummin()`

Cumulative min.

`cumprod()`

Cumulative product.

## Apply Functions

`df.pipe()` when chaining function

`df.apply()` to apply function on a axis

`df.agg()` to express multiple agg. operations

`df.transform()` to return same size DataFrame

## Combine Data Sets

adf

x1	x2
A	1
B	2
C	3

bdf

x1	x3
A	T
B	F
D	T



### Standard Joins

x1	x2	x3
A	1	T
B	2	F
C	3	NaN

`pd.merge(adf, bdf, how='left', on='x1')`  
Join matching rows from bdf to adf.

x1	x2	x3
A	1.0	T
B	2.0	F
D	NaN	T

`pd.merge(adf, bdf, how='right', on='x1')`  
Join matching rows from adf to bdf.

x1	x2	x3
A	1	T
B	2	F

`pd.merge(adf, bdf, how='inner', on='x1')`  
Join data. Retain only rows in both sets.

x1	x2	x3
A	1	T
B	2	F
C	3	NaN
D	NaN	T

`pd.merge(adf, bdf, how='outer', on='x1')`  
Join data. Retain all values, all rows.

### Filtering Joins

x1	x2
A	1
B	2

`adf[adf.x1.isin(bdf.x1)]`  
All rows in adf that have a match in bdf.

x1	x2
C	3

`adf[~adf.x1.isin(bdf.x1)]`  
All rows in adf that do not have a match in bdf.

ydf

x1	x2
A	1
B	2
C	3

zdf

x1	x2
B	2
C	3
D	4



### Set-like Operations

x1	x2
B	2
C	3

`pd.merge(ydf, zdf)`  
Rows that appear in both ydf and zdf (Intersection).

x1	x2
A	1
B	2
C	3
D	4

`pd.merge(ydf, zdf, how='outer')`  
Rows that appear in either or both ydf and zdf (Union).

x1	x2
A	1

`pd.merge(ydf, zdf, how='outer', indicator=True)`  
`.query('_merge == "left_only"')`  
`.drop(columns=['_merge'])`  
Rows that appear in ydf but not zdf (Setdiff).