



# Python Cheat Sheet



**Pandas | Numpy | Sklearn  
Matplotlib | Seaborn  
BS4 | Selenium | Scrapy**

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# Python Basics Cheat Sheet

Here you will find all the Python core concepts you need to know before learning any third-party library.

## Data Types

Integers (int): 1

Float (float): 1.2

String (str): "Hello World"

Boolean: True/False

List: [value1, value2]

Dictionary: {key1:value1, key2:value2, ...}

### Numeric Operators

+	Addition
-	Subtraction
*	Multiplication
/	Division
**	Exponent
%	Modulus
//	Floor division

### Comparison Operators

==	Equal to
!=	Different
>	Greater than
<	Less than
>=	Greater than or equal to
<=	Less than or equal to

## String methods

string.upper(): converts to uppercase

string.lower(): converts to lowercase

string.title(): converts to title case

string.count('l'): counts how many times "l" appears

string.find('h'): position of the "h" first occurrence

string.replace('o', 'u'): replaces "o" with "u"

## Variables

Variable assignment:

```
message_1 = "I'm learning Python"
message_2 = "and it's fun!"
```

String concatenation (+ operator):

```
message_1 + ' ' + message_2
```

String concatenation (f-string):

```
f'{message_1} {message_2}'
```

## List

Creating a list:

```
countries = ['United States', 'India', 'China', 'Brazil']
```

Create an empty list:

```
my_list = []
```

Indexing:

```
>>> countries[0]
United States
```

```
>>> countries[3]
Brazil
```

```
>>> countries[-1]
Brazil
```

Slicing:

```
>>> countries[0:3]
['United States', 'India', 'China']
```

```
>>> countries[1:]
['India', 'China', 'Brazil']
```

```
>>> countries[:2]
['United States', 'India']
```

Adding elements to a list:

```
countries.append('Canada')
countries.insert(0, 'Canada')
```

Nested list:

```
nested_list = [countries, countries_2]
```

Remove element:

```
countries.remove('United States')
countries.pop(0) # removes and returns value
del countries[0]
```

Creating a new list:

```
numbers = [4, 3, 10, 7, 1, 2]
```

Sorting a list:

```
>>> numbers.sort()
[1, 2, 3, 4, 7, 10]
```

```
>>> numbers.sort(reverse=True)
[10, 7, 4, 3, 2, 1]
```

Update value on a list:

```
>>> numbers[0] = 1000
```

```
>>> numbers
[1000, 7, 4, 3, 2, 1]
```

Copying a list:

```
new_list = countries[:]
new_list_2 = countries.copy()
```

## Built-in Functions

Print an object:

```
print("Hello World")
```

Return the length of x:

```
len(x)
```

Return the minimum value:

```
min(x)
```

Return the maximum value:

```
max(x)
```

Returns a sequence of numbers:

```
range(x1,x2,n) # from x1 to x2
(increments by n)
```

Convert x to a string:

```
str(x)
```

Convert x to an integer/float:

```
int(x)
float(x)
```

Convert x to a list:

```
list(x)
```



## Dictionary

Creating a dictionary:

```
my_data = {'name': 'Frank', 'age': 26}
```

Create an empty dictionary:

```
my_dict = {}
```

Get value of key "name":

```
>>> my_data["name"]  
'Frank'
```

Get the keys:

```
>>> my_data.keys()  
dict_keys(['name', 'age'])
```

Get the values:

```
>>> my_data.values()  
dict_values(['Frank', 26])
```

Get the pair key-value:

```
>>> my_data.items()  
dict_items([('name', 'Frank'), ('age', 26)])
```

Adding/updating items in a dictionary:

```
my_data['height'] = 1.7  
my_data.update({'height': 1.8,  
               'languages': ['English', 'Spanish']})  
  
>>> my_data  
{'name': 'Frank',  
 'age': 26,  
 'height': 1.8,  
 'languages': ['English', 'Spanish']}
```

Remove an item:

```
my_data.pop('height')  
del my_data['languages']  
my_data.clear()
```

Copying a dictionary:

```
new_dict = my_data.copy()
```

## If Statement

Conditional test:

```
if <condition>:  
    <code>  
elif <condition>:  
    <code>  
...  
else:  
    <code>
```

Example:

```
if age >= 18:  
    print("You're an adult!")
```

Conditional test with list:

```
if <value> in <list>:  
    <code>
```

## Loops

For loop:

```
for <variable> in <list>:  
    <code>
```

For loop and enumerate list elements:

```
for i, element in enumerate(<list>):  
    <code>
```

For loop and obtain dictionary elements:

```
for key, value in my_dict.items():  
    <code>
```

While loop:

```
while <condition>:  
    <code>
```

## Data Validation

Try-except:

```
try:  
    <code>  
except <error>:  
    <code>
```

Loop control statement:

```
break: stops loop execution  
continue: jumps to next iteration  
pass: does nothing
```

## Functions

Create a function:

```
def function(<params>):  
    <code>  
    return <data>
```

## Modules

Import module:

```
import module  
module.method()
```

OS module:

```
import os  
os.getcwd()  
os.listdir()  
os.makedirs(<path>)
```

## Special Characters

#	Comment
\n	New Line

## Boolean Operators

and	logical AND
or	logical OR
not	logical NOT

## Boolean Operators (Pandas)

&	logical AND
	logical OR
~	logical NOT

Below there are my guides, tutorials and complete Data Science course:

- [Medium Guides](#)
- [YouTube Tutorials](#)
- [Data Science Course](#) (Udemy)

Made by Frank Andrade [frank-andrade.medium.com](https://frank-andrade.medium.com)

# Pandas Cheat Sheet

Pandas provides data analysis tools for Python. All of the following code examples refer to the dataframe below.

df =

	col1	col2
A	1	4
B	2	5
C	3	6

← axis 1

← axis 0

## Getting Started

Import pandas:

```
import pandas as pd
```

Create a series:

```
s = pd.Series([1, 2, 3],
               index=['A', 'B', 'C'],
               name='col1')
```

Create a dataframe:

```
data = [[1, 4], [2, 5], [3, 6]]
index = ['A', 'B', 'C']
df = pd.DataFrame(data, index=index,
                  columns=['col1', 'col2'])
```

Read a csv file with pandas:

```
df = pd.read_csv('filename.csv')
```

Advanced parameters:

```
df = pd.read_csv('filename.csv', sep=',',
                 names=['col1', 'col2'],
                 index_col=0,
                 encoding='utf-8',
                 nrows=3)
```

## Selecting rows and columns

Select single column:

```
df['col1']
```

Select multiple columns:

```
df[['col1', 'col2']]
```

Show first n rows:

```
df.head(2)
```

Show last n rows:

```
df.tail(2)
```

Select rows by index values:

```
df.loc['A'] df.loc[['A', 'B']]
```

Select rows by position:

```
df.loc[1] df.loc[1:]
```

## Data wrangling

Filter by value:

```
df[df['col1'] > 1]
```

Sort by one column:

```
df.sort_values('col1')
```

Sort by columns:

```
df.sort_values(['col1', 'col2'],
               ascending=[False, True])
```

Identify duplicate rows:

```
df.duplicated()
```

Identify unique rows:

```
df['col1'].unique()
```

Swap rows and columns:

```
df = df.transpose()
df = df.T
```

Drop a column:

```
df = df.drop('col1', axis=1)
```

Clone a data frame:

```
clone = df.copy()
```

Connect multiple data frames vertically:

```
df2 = df + 5 #new dataframe
pd.concat([df, df2])
```

Merge multiple data frames horizontally:

```
df3 = pd.DataFrame([[1, 7], [8, 9]],
                   index=['B', 'D'],
                   columns=['col1', 'col3'])
```

#df3: new dataframe

Only merge complete rows (INNER JOIN):

```
df.merge(df3)
```

Left column stays complete (LEFT OUTER JOIN):

```
df.merge(df3, how='left')
```

Right column stays complete (RIGHT OUTER JOIN):

```
df.merge(df3, how='right')
```

Preserve all values (OUTER JOIN):

```
df.merge(df3, how='outer')
```

Merge rows by index:

```
df.merge(df3, left_index=True,
         right_index=True)
```

Fill NaN values:

```
df.fillna(0)
```

Apply your own function:

```
def func(x):
    return 2**x
df.apply(func)
```

## Arithmetics and statistics

Add to all values:

```
df + 10
```

Sum over columns:

```
df.sum()
```

Cumulative sum over columns:

```
df.cumsum()
```

Mean over columns:

```
df.mean()
```

Standard deviation over columns:

```
df.std()
```

Count unique values:

```
df['col1'].value_counts()
```

Summarize descriptive statistics:

```
df.describe()
```



# Hierarchical indexing

Create hierarchical index:  
`df.stack()`

Dissolve hierarchical index:  
`df.unstack()`

## Aggregation

Create group object:  
`g = df.groupby('col1')`

Iterate over groups:  
`for i, group in g:  
 print(i, group)`

Aggregate groups:  
`g.sum()  
g.prod()  
g.mean()  
g.std()  
g.describe()`

Select columns from groups:  
`g['col2'].sum()  
g[['col2', 'col3']].sum()`

Transform values:  
`import math  
g.transform(math.log)`

Apply a list function on each group:  
`def strsum(group):  
 return ''.join([str(x) for x in group.value])`

`g['col2'].apply(strsum)`

Below there are my guides, tutorials and complete Pandas course:

- [Medium Guides](#)
- [YouTube Tutorials](#)
- [Pandas Course \(Udemy\)](#)

# Data export

Data as NumPy array:  
`df.values`

Save data as CSV file:  
`df.to_csv('output.csv', sep=',')`

Format a dataframe as tabular string:  
`df.to_string()`

Convert a dataframe to a dictionary:  
`df.to_dict()`

Save a dataframe as an Excel table:  
`df.to_excel('output.xlsx')`

## Pivot and Pivot Table

Read csv file 1:  
`df_gdp = pd.read_csv('gdp.csv')`

The pivot() method:  
`df_gdp.pivot(index="year",  
 columns="country",  
 values="gdppc")`

Read csv file 2:  
`df_sales=pd.read_excel(  
 'supermarket_sales.xlsx')`

Make pivot table:  
`df_sales.pivot_table(index='Gender',  
 aggfunc='sum')`

Make a pivot tables that says how much male and female spend in each category:

```
df_sales.pivot_table(index='Gender',  
                      columns='Product line',  
                      values='Total',  
                      aggfunc='sum')
```

# Visualization

The plots below are made with a dataframe with the shape of df\_gdp (pivot() method)

Import matplotlib:  
`import matplotlib.pyplot as plt`

Start a new diagram:  
`plt.figure()`

Scatter plot:  
`df.plot(kind='scatter')`

Bar plot:  
`df.plot(kind='bar',  
 xlabel='data1',  
 ylabel='data2')`

Lineplot:  
`df.plot(kind='line',  
 figsize=(8,4))`

Boxplot:  
`df['col1'].plot(kind='box')`

Histogram over one column:  
`df['col1'].plot(kind='hist',  
 bins=3)`

Piechart:  
`df.plot(kind='pie',  
 y='col1',  
 title='Population')`

Set tick marks:  
`labels = ['A', 'B', 'C', 'D']  
positions = [1, 2, 3, 4]  
plt.xticks(positions, labels)  
plt.yticks(positions, labels)`

Label diagram and axes:  
`plt.title('Correlation')  
plt.xlabel('Nunstück')  
plt.ylabel('Slotermeyer')`

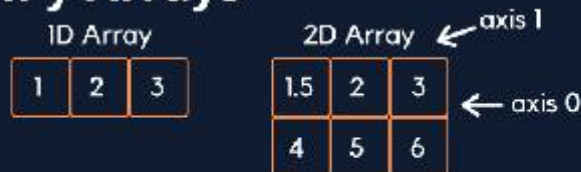
Save most recent diagram:  
`plt.savefig('plot.png')  
plt.savefig('plot.png', dpi=300)  
plt.savefig('plot.svg')`



# NumPy Cheat Sheet

NumPy provides tools for working with arrays. All of the following code examples refer to the arrays below.

## NumPy Arrays



## Getting Started

Import numpy:

```
import numpy as np
```

Create arrays:

```
a = np.array([1,2,3])
b = np.array([(1.5,2,3), (4,5,6)], dtype=float)
c = np.array([[(1.5,2,3), (4,5,6)],
               [(3,2,1), (4,5,6)]],
              dtype = float)
```

Initial placeholders:

```
np.zeros((3,4)) #Create an array of zeros
np.ones((2,3,4),dtype=np.int16)
d = np.arange(10,25,5)
np.linspace( 0,2, 9)
e = np.full((2,2), 7)
f = np.eye(2)
np.random.random((2,2))
np.empty((3,2))
```

Saving & Loading On Disk:

```
np.save('my_array', a)
np.savez('array.npz', a, b)
np.load('my_array.npy')
```

Saving & Loading Text Files

```
np.loadtxt('my_file.txt')
np.genfromtxt('my_file.csv',
              delimiter=',')
np.savetxt('myarray.txt', a,
           delimiter=' ')
```

Inspecting Your Array

```
a.shape
len(a)
b.ndim
e.size
b.dtype #data type
b.dtype.name
b.astype(int) #change data type
```

Data Types

```
np.int64
np.float32
np.complex
np.bool
np.object
np.string_
np.unicode_
```

## Array Mathematics

Arithmetic Operations

```
>>> g = a-b
array([[ -0.5,  0. ,  0. ],
       [-3. ,  3. ,  3. ]])
>>> np.subtract(a,b)

>>> b+a
array([[2.5,  4. ,  6. ],
       [ 5. ,  7. ,  9. ]])
>>> np.add(b,a)

>>> a/b
array([[ 0.66666667,  1. ,  1. ],
       [ 0.25 ,  0.4 ,  0.5 ]])
>>> np.divide(a,b)

>>> a*b
array([[ 1.5,  4. ,  9. ],
       [ 4. , 10. , 18. ]])
>>> np.multiply(a,b)

>>> np.exp(b)
>>> np.sqrt(b)
>>> np.sin(a)
>>> np.log(a)
>>> e.dot(f)
```

Aggregate functions:

```
a.sum()
a.min()
b.max(axis= 0)
b.cumsum(axis= 1) #Cumulative sum
a.mean()
b.median()
a.corrcoef() #Correlation coefficient
np.std(b) #Standard deviation
```

Copying arrays:

```
h = a.view() #Create a view
np.copy(a)
h = a.copy() #Create a deep copy
```

Sorting arrays:

```
a.sort() #Sort an array
c.sort(axis=0)
```

## Array Manipulation

Transposing Array:

```
i = np.transpose(b)
i.T
```

Changing Array Shape:

```
b.ravel()
g.reshape(3,-2)
```

Adding/removing elements:

```
h.resize((2,6))
np.append(h,g)
np.insert(a, 1, 5)
np.delete(a,[1])
```

Combining arrays:

```
np.concatenate((a,d),axis=0)
np.vstack((a,b)) #stack vertically
np.hstack((e,f)) #stack horizontally
```

Splitting arrays:

```
np.hsplit(a,3) #Split horizontally
np.vsplit(c,2) #Split vertically
```

Subsetting

```
b[1,2]
```

1.5	2	3
4	5	6

Slicing:

```
a[0:2]
```

1	2	3
---	---	---

Boolean Indexing:

```
a[a<2]
```

1	2	3
---	---	---



# Scikit-Learn Cheat Sheet



Sklearn is a free machine learning library for Python. It features various classification, regression and clustering algorithms.

## Getting Started

The code below demonstrates the basic steps of using sklearn to create and run a model on a set of data.

The steps in the code include loading the data, splitting into train and test sets, scaling the sets, creating the model, fitting the model on the data using the trained model to make predictions on the test set, and finally evaluating the performance of the model.

```
from sklearn import neighbors, datasets, preprocessing
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
iris = datasets.load_iris()
X, y = iris.data[:, :2], iris.target
X_train, X_test, y_train, y_test = train_test_split(X, y)
scaler = preprocessing.StandardScaler().fit(X_train)
X_train = scaler.transform(X_train)
X_test = scaler.transform(X_test)
knn = neighbors.KNeighborsClassifier(n_neighbors = 5)
knn.fit(X_train, y_train)
y_pred = knn.predict(X_test)
accuracy_score(y_test, y_pred)
```

## Loading the Data

The data needs to be numeric and stored as NumPy arrays or SciPy sparse matrix (numeric arrays, such as Pandas DataFrame's are also ok)

```
>>> import numpy as np
>>> X = np.random.random((10, 5))
array([[0.21, 0.33],
       [0.23, 0.60],
       [0.48, 0.62]])
>>> y = np.array(['A', 'B', 'A'])
array(['A', 'B', 'A'])
```

## Training and Test Data

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y,
                                                    random_state = 0) # Splits data into training and test sets
```

## Preprocessing The Data

### Standardization

Standardizes the features by removing the mean and scaling to unit variance.

```
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler().fit(X_train)
standardized_X = scaler.transform(X_train)
standardized_X_test = scaler.transform(X_test)
```

### Normalization

Each sample (row of the data matrix) with at least one non-zero component is rescaled independently of other samples so that its norm equals one.

```
from sklearn.preprocessing import Normalizer
scaler = Normalizer().fit(X_train)
normalized_X = scaler.transform(X_train)
normalized_X_test = scaler.transform(X_test)
```

### Binarization

Binarize data (set feature values to 0 or 1) according to a threshold.

```
from sklearn.preprocessing import Binarizer
binarizer = Binarizer(threshold = 0.0).fit(X)
binary_X = binarizer.transform(X_test)
```

### Encoding Categorical Features

Imputation transformer for completing missing values.

```
from sklearn import preprocessing
le = preprocessing.LabelEncoder()
le.fit_transform(X_train)
```

### Imputing Missing Values

```
from sklearn.impute import SimpleImputer
imp = SimpleImputer(missing_values=0, strategy='mean')
imp.fit_transform(X_train)
```

### Generating Polynomial Features

```
from sklearn.preprocessing import PolynomialFeatures
poly = PolynomialFeatures(5)
poly.fit_transform(X)
```



## Create Your Model

### Supervised Learning Models

#### Linear Regression

```
from sklearn.linear_model import LinearRegression
lr = LinearRegression(normalize = True)
```

#### Support Vector Machines (SVM)

```
from sklearn.svm import SVC
svc = SVC(kernel = 'linear')
```

#### Naive Bayes

```
from sklearn.naive_bayes import GaussianNB
gnb = GaussianNB()
```

#### KNN

```
from sklearn import neighbors
knn = neighbors.KNeighborsClassifier(n_neighbors = 5)
```

### Unsupervised Learning Models

#### Principal Component Analysis (PCA)

```
from sklearn.decomposition import PCA
pca = PCA(n_components = 0.95)
```

#### K means

```
from sklearn.cluster import KMeans
k_means = KMeans(n_clusters = 3, random_state = 0)
```

## Model Fitting

Fitting supervised and unsupervised learning models onto data.

#### Supervised Learning

```
lr.fit(X, y) #Fit the model to the data
knn.fit(X_train,y_train)
svc.fit(X_train,y_train)
```

#### Unsupervised Learning

```
k_means.fit(X_train) #Fit the model to the data
pca_model = pca.fit_transform(X_train)#Fit to data,then transform
```

## Prediction

#### Predict Labels

```
y_pred = lr.predict(X_test) #Supervised Estimators
y_pred = k_means.predict(X_test) #Unsupervised Estimators
```

#### Estimate probability of a label

```
y_pred = knn.predict_proba(X_test)
```

## Evaluate Your Model's Performance

### Classification Metrics

#### Accuracy Score

```
knn.score(X_test,y_test)
from sklearn.metrics import accuracy_score
accuracy_score(y_test,y_pred)
```

#### Classification Report

```
from sklearn.metrics import classification_report
print(classification_report(y_test,y_pred))
```

#### Confusion Matrix

```
from sklearn .metrics import confusion_matrix
print(confusion_matrix(y_test,y_pred))
```

### Regression Metrics

#### Mean Absolute Error

```
from sklearn.metrics import mean_absolute_error
mean_absolute_error(y_test,y_pred)
```

#### Mean Squared Error

```
from sklearn.metrics import mean_squared_error
mean_squared_error(y_test,y_pred)
```

#### R<sup>2</sup> Score

```
from sklearn.metrics import r2_score
r2_score(y_test, y_pred)
```

### Clustering Metrics

#### Adjusted Rand Index

```
from sklearn.metrics import adjusted_rand_score
adjusted_rand_score(y_test,y_pred)
```

#### Homogeneity

```
from sklearn.metrics import homogeneity_score
homogeneity_score(y_test,y_pred)
```

#### V-measure

```
from sklearn.metrics import v_measure_score
v_measure_score(y_test,y_pred)
```

## Tune Your Model

### Grid Search

```
from sklearn.model_selection import GridSearchCV
params = {'n_neighbors':np.arange(1,3),
          'metric':['euclidean','cityblock']}
grid = GridSearchCV(estimator = knn, param_grid = params)
grid.fit(X_train, y_train)
print(grid.best_score_)
print(grid.best_estimator_)
```



# Data Viz Cheat Sheet



Matplotlib is a Python 2D plotting library that produces figures in a variety of formats.



## Matplotlib

### Workflow

The basic steps to creating plots with matplotlib are Prepare Data, Plot, Customize Plot, Save Plot and Show Plot.

```
import matplotlib.pyplot as plt
```

### Example with lineplot

Prepare data

```
x = [2017, 2018, 2019, 2020, 2021]
y = [43, 45, 47, 48, 50]
```

Plot & Customize Plot

```
plt.plot(x,y,marker='o',linestyle='--',
color='g', label='USA')
plt.xlabel('Years')
plt.ylabel('Population (M)')
plt.title('Years vs Population')
plt.legend(loc='lower right')
plt.yticks([41, 45, 48, 51])
```

Save Plot

```
plt.savefig('example.png')
```

Show Plot

```
plt.show()
```

Markers: 'l', 'o', 'v', '<', '>'

Line Styles: '-', '--', '...', '.'

Colors: 'b', 'g', 'r', 'y' #blue, green, red, yellow

### Barplot

```
x = ['USA', 'UK', 'Australia']
y = [40, 50, 33]
plt.bar(x, y)
plt.show()
```

### Piechart

```
plt.pie(y, labels=x, autopct='%0f %%')
plt.show()
```

### Histogram

```
ages = [15, 16, 17, 30, 31, 32, 35]
bins = [15, 20, 25, 30, 35]
plt.hist(ages, bins, edgecolor='black')
plt.show()
```

### Boxplots

```
ages = [15, 16, 17, 30, 31, 32, 35]
plt.boxplot(ages)
plt.show()
```

### Scatterplot

```
a = [1, 2, 3, 4, 5, 4, 3, 2, 5, 6, 7]
b = [7, 2, 3, 5, 5, 7, 3, 2, 6, 3, 2]
plt.scatter(a, b)
plt.show()
```

## Subplots

Add the code below to make multiple plots with 'n' number of rows and columns.

```
fig, ax = plt.subplots(nrows=1,
ncols=2,
sharey=True,
figsize=(12, 4))
```

Plot & Customize Each Graph

```
ax[0].plot(x, y, color='g')
ax[0].legend()
ax[1].plot(a, b, color='r')
ax[1].legend()
plt.show()
```

## Seaborn

### Workflow

```
import seaborn as sns
import matplotlib.pyplot as plt
import pandas as pd

Lineplot
plt.figure(figsize=(10, 5))
flights = sns.load_dataset("flights")
may_flights=flights.query("month=='May'")
ax = sns.lineplot(data=may_flights,
x="year",
y="passengers")
ax.set(xlabel='x', ylabel='y',
title='my_title', xticks=[1,2,3])
ax.legend(title='my legend',
title_fontsize=13)
plt.show()
```

### Barplot

```
tips = sns.load_dataset("tips")
ax = sns.barplot(x="day",
y="total_bill",
data=tips)
```

### Histogram

```
penguins = sns.load_dataset("penguins")
sns.histplot(data=penguins,
x="flipper_length_mm")
```

### Boxplot

```
tips = sns.load_dataset("tips")
ax = sns.boxplot(x=tips["total_bill"])
```

### Scatterplot

```
tips = sns.load_dataset("tips")
sns.scatterplot(data=tips,
x="total_bill",
y="tip")
```

### Figure aesthetics

```
sns.set_style('darkgrid') #stlyes
sns.set_palette('husl', 3) #palettes
sns.color_palette('husl') #colors
```

Fontsize of the axes title, x and y labels, tick labels and legend:

```
plt.rc('axes', titlesize=18)
plt.rc('axes', labelszsize=14)
plt.rc('xtick', labelszsize=13)
plt.rc('ytick', labelszsize=13)
plt.rc('legend', fontsize=13)
plt.rc('font', size=13)
```



# Web Scraping Cheat Sheet

Web Scraping is the process of extracting data from a website. Before studying BeautifulSoup and Selenium, it's good to review some HTML basics first.

## HTML for Web Scraping

Let's take a look at the HTML element syntax.



This is a single HTML element, but the HTML code behind a website has hundreds of them.

### HTML code example

```
<article class="main-article">
  <h1> Titanic (1997) </h1>
  <p class="plot"> 84 years later ... </p>
  <div class="full-script"> 13 meters. You ... </div>
</article>
```

The HTML code is structured with "nodes". Each rectangle below represents a node (element, attribute and text nodes)



- "Siblings" are nodes with the same parent.
- A node's children and its children's children are called its "descendants". Similarly, a node's parent and its parent's parent are called its "ancestors".
- it's recommended to find element in this order.
  - a. ID
  - b. Class name
  - c. Tag name
  - d. Xpath

## Beautiful Soup

### Workflow

Importing the libraries

```
from bs4 import BeautifulSoup
import requests
```

Fetch the pages

```
result=requests.get("www.google.com")
result.status_code #get status code
result.headers #get the headers
```

Page content

```
content = result.text
```

Create soup

```
soup = BeautifulSoup(content,"lxml")
```

HTML in a readable format

```
print(soup.prettify())
```

Find an element

```
soup.find(id="specific_id")
```

Find elements

```
soup.find_all("a")
soup.find_all("a","css_class")
soup.find_all("a",class="my_class")
soup.find_all("a",attrs={"class":
                        "my_class"})
```

Get inner text

```
sample = element.get_text()
sample = element.get_text(strip=True,
                          separator=' ')
```

Get specific attributes

```
sample = element.get('href')
```

## XPath

We need to learn XPath to scrape with Selenium or Scrapy.

### XPath Syntax

An XPath usually contains a tag name, attribute name, and attribute value.

```
//tagName[@AttributeName="Value"]
```

Let's check some examples to locate the article, title, and transcript elements of the HTML code we used before.

```
//article[@class="main-article"]
//h1
//div[@class="full-script"]
```

### XPath Functions and Operators

XPath functions

```
//tag[contains(@AttributeName, "Value")]
```

XPath Operators: and, or

```
//tag[(expression 1) and (expression 2)]
```

### XPath Special Characters

<code>/</code>	Selects the children from the node set on the left side of this character
<code>//</code>	Specifies that the matching node set should be located at any level within the document
<code>.</code>	Specifies the current context should be used (refers to present node)
<code>..</code>	Refers to a parent node
<code>*</code>	A wildcard character that selects all elements or attributes regardless of names
<code>@</code>	Select an attribute
<code>()</code>	Grouping an XPath expression
<code>[n]</code>	Indicates that a node with index "n" should be selected



# Selenium



## Workflow

```
from selenium import webdriver
web="www.google.com"
path='introduce chromedriver path'
driver = webdriver.Chrome(path)
driver.get(web)
```

### Find an element

```
driver.find_element_by_id('name')
```

### Find elements

```
driver.find_elements_by_class_name()
driver.find_elements_by_css_selector
driver.find_elements_by_xpath()
driver.find_elements_by_tag_name()
driver.find_elements_by_name()
```

### Quit driver

```
driver.quit()
```

### Getting the text

```
data = element.text
```

### Implicit Waits

```
import time
time.sleep(2)
```

### Explicit Waits

```
from selenium.webdriver.common.by import By
from selenium.webdriver.support.ui import WebDriverWait
from selenium.webdriver.support import expected_conditions as EC
```

```
WebDriverWait(driver, 5).until(EC.element_to_be_clickable((By.ID,
'id_name')))) #Wait 5 seconds until an element is clickable
```

### Options: Headless mode, change window size

```
from selenium.webdriver.chrome.options import Options
options = Options()
options.headless = True
options.add_argument('window-size=1920x1080')
driver=webdriver.Chrome(path,options=options)
```

Below there are my guides, tutorials  
and complete web scraping course:

- [Medium Guides](#)
- [YouTube Tutorials](#)
- [Web Scraping Course \(Udemy\)](#)

Made by Frank Andrade [frank-andrade.medium.com](https://frank-andrade.medium.com)

# Scrapy



Scrapy is the most powerful web scraping framework in Python, but it's a bit complicated to set up, so check my [guide](#) or its documentation to set it up.

## Creating a Project and Spider

To create a new project, run the following command in the terminal.

```
scrapy startproject my_first_spider
```

To create a new spider, first change the directory.

```
cd my_first_spider
```

Create an spider

```
scrapy genspider example example.com
```

## The Basic Template

When you create a spider, you obtain a template with the following content.

```
import scrapy
class ExampleSpider(scrapy.Spider):
    name = 'example'
    allowed_domains = ['example.com']
    start_urls = ['http://example.com/']

    def parse(self, response):
        pass
```

Class

Parse method

The class is built with the data we introduced in the previous command, but the parse method needs to be built by us. To build it, use the functions below.

## Finding elements

To find elements in Scrapy, use the response argument from the parse method

```
response.xpath('//tag[@AttributeName="Value"]')
```

## Getting the text

To obtain the text element we use `text()` and either `.get()` or `.getall()`. For example:

```
response.xpath('//h1/text()').get()
response.xpath('//tag[@Attribute="Value"]/text()').getall()
```

## Return data extracted

To see the data extracted we have to use the yield keyword

```
def parse(self, response):
    title = response.xpath('//h1/text()').get()

    # Return data extracted
    yield {'titles': title}
```

## Run the spider and export data to CSV or JSON

```
scrapy crawl example
```

```
scrapy crawl example -o name_of_file.csv
```

```
scrapy crawl example -o name_of_file.json
```