

Read data from xlrd



- `myWorkBook.sheets()` – Returns the object of individual sheet to work on.
- `myWorkBook.sheet_names()` – Returns the list of all sheets.
- `myWorkBook.sheet_by_index()` – Returns the object of individual sheet based on index value.
- `myWorkBook.nsheets` – An integer containing the number of sheets in a workbook.
- `Sheet.name` – Returns the sheet name.
- `sheet.nrows` and `sheet.ncols` – Contain the number of rows and columns in a worksheet.
- `cellname` and `colname` – Defined functions in xlrd that provide the cell name and colname.

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Xlwt – write XL files



```
import xlwt
book = xlwt.Workbook()
sheet1 = book.add_sheet('names')

# Add data using row and col
sheet1.write(0,0,'A1')
sheet1.write(0,1,'B1')

# Add data using row object
row1 = sheet1.row(1)
row1.write(0,'A2')
row1.write(1,'B2')

# Add data using row ID
sheet1.row(2).write(0,'A3')
sheet1.row(2).write(1,'B3')
sheet1.flush_row_data()

book.save('writeToXL.xls')
```

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Xlutils – Append Data to XL file

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```
from xlutils.copy import copy
import xlrd

book = xlrd.open_workbook("Sample.xls")
copy_book = copy(book)           #creates a writeable copy
sheet1 = copy_book.get_sheet(0)  # get a first sheet

col = 3
for row in range(1):
    # Write the data to row, column
    sheet1.write(row,col, 10)
    sheet1.write(row,col + 1, 10)

copy_book.save("Sample.xls")
```

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Web scraping Example using urllib

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```
import re, urllib

def get_quote(symbol):
    base_url = 'http://finance.google.com/finance?q='
    content = urllib.urlopen(base_url + symbol).read()
    find_q = re.search(r'<span\sid="ref_\d+.*">(.+)<', content)
    if find_q:
        quote = find_q.group(1)
    else:
        quote = 'no quote available for: %s' % symbol
    return quote

Companys = ['google', 'ibm', 'microsoft', 'apple', 'nvidia']
for i,value in enumerate(Companys):
    print '%s --> %s' % (Companys[i],get_quote(value))
```

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Web scraping Example using bs4



```
from bs4 import BeautifulSoup
import requests, re

myurl = "http://finance.yahoo.com/q/op?s=" + symbol + "+Options"
html = requests.get(myurl).content
soup = BeautifulSoup(html, "html.parser")
print soup.find('span', attrs={'class':'time_rtq_ticker'}).text
```

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Some Common important modules



1. Requests. The Requests module is a an elegant and simple HTTP library for Python It's a must have for every python developer.
2. Scrapy. If you are involved in webscraping then this is a must have library for you.
3. wxPython is a wrapper for the cross-platform GUI API (often referred to as a 'toolkit) for the Python programming language
4. Pillow. A friendly fork of PIL (Python Imaging Library). It is more user friendly than PIL and is a must have for anyone who works with images.
5. SQLAlchemy. A database library. Many love it and many hate it. It includes dialects for SQLite, Postgresql, MySQL, Oracle, MS-SQL, Firebird, Sybase and others,
6. BeautifulSoup. This parse xml and html library very useful for beginners.
7. Twisted. The most important tool for any network application developer. It has a very beautiful api and is used by a lot of famous python developers.
8. NumPy. How can we leave this very important library ? It provides some advance math functionalities to python.
9. SciPy. When we talk about NumPy then we have to talk about scipy. It is a library of algorithms and mathematical tools for python and has caused many scientists to switch from ruby to python.

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Some Common important modules

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1. matplotlib. A numerical plotting library. It is very useful for any data scientist or any data analyzer.
2. Pygame. Which developer does not like to play games and develop them ? This library will help you achieve your goal of 2d game development.
3. Pyglet. A 3d animation and game creation engine. This is the engine in which the famous python port of minecraft was made
4. pyQT. A GUI toolkit for python. It is my second choice after wxpython for developing GUI's for my python scripts.
5. nltk. Natural Language Toolkit – I realize most people won't be using this one, but it's generic enough. It is a very useful library if you want to manipulate strings. But it's capacity is beyond that. Do check it out.
6. nose. A testing framework for python. It is used by millions of python developers. It is a must have if you do test driven development.

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Objective – Module 11

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Introduction to Machine Learning

- What is Machine Learning?
- Areas of Implementation of Machine Learning,
- Why companies prefer ML
- Major Classes of Learning Algorithms
- Supervised vs Unsupervised Learning, Learning
- Why Numpy?
- Learning Numpy and Scipy,
- Basic plotting using Matplotlib
- Algorithms using Skikit learn

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What is Machine Learning?



Machine learning was defined in 1959 by Arthur Samuel as the "field of study that gives computers the ability to learn without being explicitly programmed." This means imbuing knowledge to machines without hard-coding it. – via Wikipedia

Machine learning is a method of data analysis that automates analytical model building. Using algorithms that iteratively learn from data, machine learning allows computers to find hidden insights without being explicitly programmed where to look. – via SAS

Method of teaching computers or programs/scripts to make and improve predictions or behaviours based on some training data. Machine learning is a huge field, with hundreds of different algorithms for solving different stats problems – via Ethans

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Fields of study

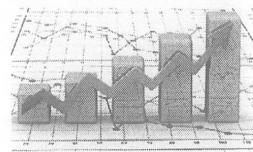


Gaming and AI

Samuel created a computer checkers in 1959, and seminal research on machine learning beginning in 1949. He thought that teaching computers to play games was very fruitful for developing tactics appropriate to general problems.



Robotics and self driven cars



Forecasting & Bioinformatics



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Why companies using Machine learning?



- Online recommendation engines like those from Amazon, snap deal and flip kart. Machine learning applications help them to sell products.
- Got to know the product sentiments from the market from Twitter and Facebook, Machine learning applications help them to analyse the market response.
- Remember face recognition and thumb impression software's for laptop and mobile, Machine learning applications help them to build cool products.
- Fraud detection on time, Machine learning applications help them to build secure products.
- What will be the weather forecast for tomorrow? , Machine learning applications help them to build forecasting products.
- Looking for the automated engine for routing the traffic in the cluster nodes? , Machine learning applications help them to build robotics products.

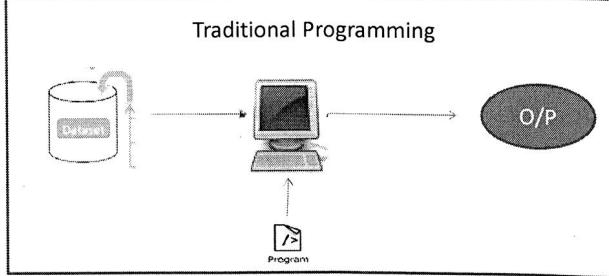
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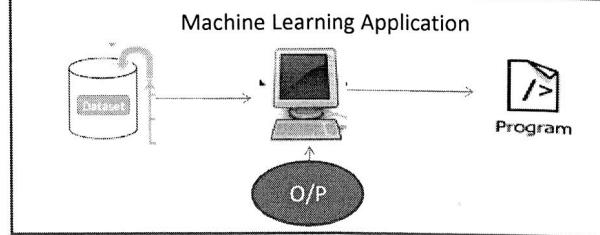
Comparison of two programming's



Traditional Programming



Machine Learning Application



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How is machine learning used today?

Ever wonder how an online retailer provides nearly instantaneous offers for other products that may interest you? Or how lenders can provide near-real-time answers to your loan requests? Many of our day-to-day activities are powered by machine learning algorithms, including:

- Fraud detection.
- Web search results.
- Real-time ads on web pages and mobile devices.
- Text-based sentiment analysis.
- Credit scoring and next-best offers.
- Prediction of equipment failures.
- New pricing models.
- Network intrusion detection.
- Pattern and image recognition.
- Email spam filtering.

Via: SAS website www.ethans.co.in

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Type of learning in ML

- Supervised learning
- Unsupervised learning
- Semi-supervised
- Reinforcement learning

Two of the most widely adopted machine learning methods are **supervised learning** and **unsupervised learning**. Most machine learning – about 70 percent – is supervised learning. Unsupervised learning accounts for 10 to 20 percent. Semi-supervised and reinforcement learning are two other technologies that are sometimes used. Via SAS

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Supervised learning

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Supervised learning algorithms are trained using labeled training dataset. The training data consist of a set of training examples.

In supervised learning, each example is a pair consisting of an input object (typically a vector) and a desired output value (also called the supervisory signal).

A supervised learning algorithm analyzes the training data and produces an inferred function, which can be used for mapping new examples

You Viewed
Product A
 add to cart

Customers who viewed this also viewed:
Product B
 add to cart

Product C
 add to cart

Product D
 add to cart

Recommendation Engine www.ethans.co.in

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Examples of Supervised Learning

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Employee Salary (\$1,000s)	Employee Satisfaction Rating
35	40
40	35
45	45
50	40
55	55
60	50
65	60
70	65
75	70
80	75
85	80

Linear Regression - Problem

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Regression



Regression example

Predict House Pricing:

Based on: House size, floor Size, area, features, Builder etc

Predict Body BMI:

Based on: Sex, height, weight etc

Predict crop yields:

Based on : Soil conditions, rainfall in an areas, Fertilizers used, water level etc

Predict Life expectancy:

Based on: Eating Pattern, exercises pattern, smoker, drinker , disease status, forefather history etc.

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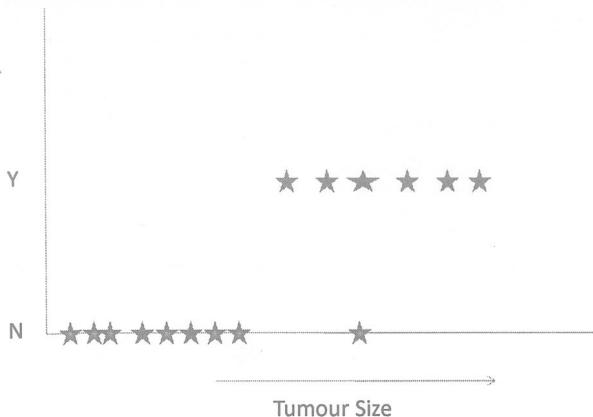
Examples of Supervised Learning



Malign
ant or
not?

Y

N



Classification Problem

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Examples of Supervised Learning

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SVM Algorithm

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Classification

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- text categorization** - (e.g., spam filtering)
- fraud detection** – Finance, anti virus etc
- optical character recognition** – Handwriting recognition
- machine vision** - (e.g., palm detections, thumb impression face detections)
- natural-language processing** - (e.g., spoken language understanding)
- market segmentation** - (e.g. predict if customer will respond to promotion)
- Bioinformatics** - (e.g., classify proteins according to their function)

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Building accurate - Classification



- Get enough training set examples
- It should good in performance on training set
- Classifier should not be not too "complex" to predict
- Classifiers should be "as simple as possible, but no simpler"
- "simplicity" closely related to prior expectations
- In general cases get two label of the classifiers
- Test it many times to get the expected result before finalizing.

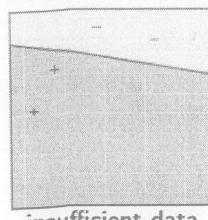
Measure "complexity" by:

- number bits needed to write down
- number of parameters

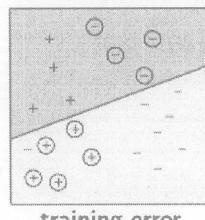
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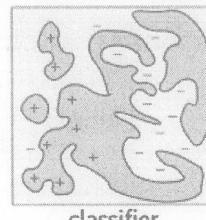
Bad Classifiers



insufficient data



training error
too high



classifier
too complex

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Unsupervised Learning

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Input data is not labelled and does not have a known result. Data is organised to form cluster of data.

A model is prepared by creating structures present in the input data.

This may be to extract general rules. It may through a mathematical process to systematically reduce redundancy, or it may be to organize data by similarity.

Example algorithms include: the k-Medians, Expectation Maximisation (EM), Hierarchical Clustering and k-Means.

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Unsupervised example:

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 **IPL 9, Qualifier 1: AB de Villiers' 'Superman' knock helps RCB storm ...**
Zee News - 8 hours ago
With the spectacular win, RCB are into their third IPL final and one step away from their maiden title. Lions on other hand will have another...

IPL: AB de Villiers pulls off a heist, RCB reach third final
Times of India - 9 hours ago

IPL 9 RCB vs GL: AB De Villiers stands tall amidst ruins, Bangalore ...
Deccan Chronicle - 7 hours ago

IPL 2016 Highlights: RCB vs GL - De Villiers, Abdulla Take Royal ...
NDTVSports.com - 10 hours ago

IPL: De Villiers guides RCB to IPL 2016 final, beat Gujarat by 4 wickets
Hindustan Times - 13 hours ago

Daniel Vettori and Bangalore into IPL final thanks to stars de Villiers ...
Opinion - Stuff.co.nz - 5 hours ago



Explore in depth (675 more articles)

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Unsupervised example:

Raw Data Clustered Data

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NumPy

NumPy is an extension to the Python programming language, adding support for large, multi-dimensional arrays and matrices, along with a large library of high-level mathematical functions to operate on these arrays

To install NumPy run:
`python setup.py install`

To perform an in-place build that can be run from the source folder run:
`python setup.py build_ext --inplace`

The NumPy build system uses distutils and numpy.distutils. setuptools is only used when building via pip or with python setupegg.py.

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NumPy performance Test

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```
author_ = 'Ethans'

import timeit

# Sum of 10000 number from Python
py_sec = timeit.timeit
('sum(x*x for x in xrange(10000))', number=10000)

# Sum of 10000 number using numpy
np_sec = timeit.timeit
('na.dot(na)', setup="import numpy as np; na=np.arange(10000)", number=10000)

print("Normal Python: %f sec" %py_sec)
print("Through NumPy: %f sec" %np_sec)
```

```
>>>
Normal Python: 22.872751 sec
Through NumPy: 0.176976 sec
```

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Getting Started with Numpy

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```
>>> # Importing Numpy module
>>> import numpy
>>> import numpy as np
```

- Arrays are the central feature of NumPy.
- Arrays in Python are similar to lists in Python, the only difference being the array elements should be of the same type

```
import numpy as np

a = np.array([1,2,3], float)      # Accept two arguments list and type
print a                          # Return array([ 1.,  2.,  3.])
print type(a)                    # Return <type 'numpy.ndarray'>
print a[2]                        # 3.0
print a[1:]                       # array([ 2.,  3.])
print a[:2]                       # array([ 1.,  2.])
```

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Two Dimensional Array



```
import numpy as np
a = np.array([[1,2,3], [4,5,6]], int) # Accept two arguments list and type
>>> a[0][0] # Return 1
>>> a[0,0] # Return 1
>>> a[1,2] # Return 6
>>> a[1:] # Return array([[4, 5, 6]])
>>> a[1,:]
>>> a[:2] # Return array([[1, 2, 3],[4, 5, 6]])
>>> a[:,2] # Return array([3, 6])
>>> a[:,1] # Return array([2, 5])
>>> a[:,0] # Return array([1, 4])
```

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Two Dimensional Array



```
import numpy as np
a = np.array([[1,2,3], [4,5,6]], int) # Accept two arguments list and type
>>> a.shape # Return the dimension of a i.e. (2,3)
>>> a.dtype # Return the type of array i.e. dtype('int32')
>>> b = a.reshape(3,2) # Return the new shape of array
>>> b.shape # Return (3,2)
>>> len(a) # Return length of a
>>> 2 in a # Check if 2 is available in a
>>> b.tolist() # Return [[1, 2], [3, 4], [5, 6]]
>>> list(b) # Return [array([1, 2]), array([3, 4]), array([5, 6])]
>>> c = b
>>> b[0][0] = 10
>>> c # Return array([[10, 2], [3, 4], [5, 6]])
```

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Two Dimensional Array



```
>>> b                                # Return array([[10, 2],[3, 4],[5, 6]])
# Convert the multidimensional array to single dimension
>>> b.flatten()                      # Return array([10, 2, 3, 4, 5, 6])

# Concatenating two arrays
>>> a = np.array([1,2,3])
>>> b = np.array([4,5,6])
>>> np.concatenate((a,b))            # Returns array([1, 2, 3, 4, 5, 6])

# arange function and ones function
>>> a = np.arange(5, dtype=float)      # Return array([ 0., 1., 2., 3., 4.])
>>> b = np.ones((3,2), dtype=float)    # Return array([[ 1., 1.], [ 1., 1.], [ 1., 1.]])
```

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Update Array



```
# Assign the new values
-----
>>> c = np.array([[1, 2, 3], [4, 5, 6], [7, 8, 9]])
>>> c[0,0] = 10
>>> c
array([[10, 2, 3],
       [4, 5, 6],
       [7, 8, 9]])
>>> c[:,2] = [10, 10, 10]
>>> c
array([[10, 2, 10],
       [4, 5, 10],
       [7, 8, 10]])
>>> c[0, :] = [10, 10, 10]
>>> c
array([[10, 10, 10],
       [4, 5, 10],
       [7, 8, 10]])
>>> c.shape
```

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Standard Maths functions



```
# 'max', 'mean', 'min', sum, std functions
>>> np.arange(100).max()
99
>>> np.arange(100).mean()
49.5
>>> np.arange(100).min()
0
>>> np.arange(100).sum()
4950
>>> a.std()
2.8722813232690143
>>> a.all()
False
>>> a.any()
True
>>> a.nonzero()
(array([1, 2, 3, 4, 5, 6, 7, 8, 9]),)
(array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
```

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Copy of arrays



```
# Referential operation also valid here
>>> a = np.array([1,2,3])
>>> a
array([1, 2, 3])
>>> b = a
>>> b[0] = 10
>>> a[2] = 20
>>> a
array([10, 2, 20])
>>> b
array([10, 2, 20])
```

```
# Base copy
>>> a
array([10, 2, 20])
>>> d = a.view()
>>> d
array([10, 2, 20])
>>> a[0] = 20
>>> d[1] = 30
>>> a
array([20, 30, 20])
>>> d
array([20, 30, 40])
>>> a is d
False
>>> d.base is a
True
```

```
# Shallow copy
>>> a
array([10, 2, 20])
>>> c = a.copy()
>>> c
array([10, 2, 20])
>>> a
array([10, 2, 20])
>>> c[0] = 20
>>> c
array([20, 2, 20])
>>> a
array([10, 2, 20])
```

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List vs Array

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```
# List operations
>>> list1 = [1,2,3]
>>> list2 = [4,5,6]

>>> list1 + list2
[1, 2, 3, 4, 5, 6]

>>> list2 * 2
[4, 5, 6, 4, 5, 6]
```

```
>>> import numpy as np
>>> a1 = np.array([1,2,3])
>>> a2 = np.array([4,5,6])
>>> a1 + a2
array([5, 7, 9])
>>> a1 * a2
array([ 4, 10, 18])
>>> a2 / a1
array([4, 2, 2])
>>> a1 ** a2
array([ 1, 32, 729])
```

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Boolean Opr

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```
>>> a = np.arange(10)
>>> a
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
>>>
>>> div = a % 2 == 0
>>> div
array([ True, False,  True, False,  True, False,  True, False,  True, False],
>>> a[div]
array([0, 2, 4, 6, 8])
```

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Maths Functions



```
# Mathematical functions
>>> c
array([[1, 2, 4],
       [5, 7, 8]])
>>> c.transpose()
array([[1, 5],
       [2, 7],
       [4, 8]])
>>> np.invert(c)
array([[-2, -3, -5],
       [-6, -8, -9]])
>>> np.eye(3)
array([[ 1.,  0.,  0.],
       [ 0.,  1.,  0.],
       [ 0.,  0.,  1.]])
>>> c = np.array([[1,2], [2,3]])
>>> np.dot(c, c) # Returns dot product array([[ 5,  8],
>>> a = np.array([1,2,3,4,5,6,7,8,9])
>>> np.median(a) # Return 5.0
>>> np.corrcoef(a) # Return 1.0
```

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Help Page



```
# Documentation of any function
>>> np.info(np.sum)
sum(a, axis=None, dtype=None, out=None, keepdims=False)

Sum of array elements over a given axis.
```

Parameters

a : array_like
Elements to sum.

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Sort function



```
# Sort function
>>> a
array([ 2,  4,  5,  1,  2,  3, 20, 10, 15])
>>> a.sort()
>>> a
array([ 1,  2,  2,  3,  4,  5, 10, 15, 20])
>>> a = np.array([2,4,5,1,2,3,20,10,15])
>>> np.sort(a)
array([ 1,  2,  2,  3,  4,  5, 10, 15, 20])
>>> a.argsort()
array([3, 0, 4, 5, 1, 2, 7, 8, 6])
>>> a[a.argsort()]
array([ 1,  2,  2,  3,  4,  5, 10, 15, 20])
>>> a = np.array([10,2,1])
>>> a.argsort()
array([2, 1, 0])
>>> a[a.argsort()]
array([ 1,  2, 10])
```

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Tile function



```
>>> b = np.array([[1, 2], [3, 4]])
>>> np.tile(b, 2)
array([[1, 2, 1, 2],
       [3, 4, 3, 4]])
>>> np.tile(b, (2, 1))
array([[1, 2],
       [3, 4],
       [1, 2],
       [3, 4]])

>>> c = np.array([1, 2, 3, 4])
>>> np.tile(c, (4, 1))
array([[1, 2, 3, 4],
       [1, 2, 3, 4],
       [1, 2, 3, 4],
       [1, 2, 3, 4]])
```

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