Python

Python is a high-level, dynamically typed multiparadigm programming language

Python versions

python 2 version no longer available currently we are using python 3

- Python is very simple to pick up;
- Packages useful for ML are available in Python;
- Jupyter Notebooks for interactive programming;
- Extensively used in the industry;
- Python is much more general purpose programming language;

Keywords and identifiers

- Keywords are the reserved words in python;
- We can't use a keyword as variable name, function name or any other identifier; Keywords are case sensitive:
- [Import keyword]
 - Example: False, None, True, class, if, else, return, def, try, while, for, etc Total number of keywords: 36
- Identifiers:
- Name given to entities like class, functions and variables
- Can be a combination of letters, digits and underscores, cannot start with a digit Keywords cannot be used as identifiers, special characters cannot be used
- Python has straight forward Error indications;

```
In [4]: import keyword
In [6]: print(len(keyword.kwlist))
36
```

- Name given to entities like class, functions and variables
- Can be a combination of letters, digits and underscores, cannot start with a digit
- Keywords cannot be used as identifiers, special characters cannot be used
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Comments, indentations and statements

- Start a line with a # or use triple quotes, "" "
- Indentations are used (4 spaces preferred) to make blocks of code, a for loop
- Rather than writing code in a single line try to write in multiple lines (can use) to make code readable
- The written instructions are called statements

Variables and data types in python

- Variable is a location in memory used to store some data;
- Variable declaration is not needed a, b = 10, 'Hi'
- id(a) prints location of a Data types:
- Everything in python is an object; Number: Integers, float and complex
- Boolean: True and False
- Strings: Sequence of Unicode characters, defined with quotes, indexable, slicable
- List: An ordered sequence of items, like an array, can have multiple data type elements, defined with square brackets; Lists are mutable
- Tuple: Defined with parenthesis, can have multiple data type elements, tuple is immutable, can be indexable
- Set: Defined with Curly braces, Set is an unordered collection of unique items; behaves as a set in mathematics; does not support indexing
- Dictionary: an unordered collection of key-value pairs, defined with curly braces and a colon, value accessible with key
- Data types can be converted provided the value is valid in both data types; List('Hello') =
 ['H','e','l','l','o']

```
<class 'int'>
3
4
2
6
9
4
8
<class 'float'>
2.5 3.5 5.0 6.25
```

Booleans:

Python implements all of the usual operators for Boolean logic, but uses English words rather than symbols (&&, ||, etc.):

```
In [8]: t = True
    f = False
    print(type(t))
    print(t and f)
    print(t or f)
    print(not t)
    print(t != f)

<class 'bool'>
    False
    True
    False
    True
    False
    True
```

Strings: Python has great support for strings:

- Sequence of characters: (Unicode (default) or ASCII) S = "kl" or = str(1)
- Access characters of a string as a list;
- Strings are immutable;
- Operations: str1+str2; for i in string: __; lower(), upper(), join(), split(), find(), replace("Bad", "Good")

```
In [9]: hello = 'hello'
world = "world"
print(hello)
print(len(hello))
hw = hello + ' ' + world # String concatenation
print(hw) # prints "hello world"
hw12 = '%s %s %d' % (hello, world, 12) # sprintf style string form
atting
print(hw12) # prints "hello world 12"
hello
5
hello world
hello world
hello world 12
```

Containers

• Python includes several built-in container types: lists, dictionaries, sets, and tuples.

Lists

• A list is the Python equivalent of an array, but is resizeable and can contain elements of different types:

- Data Structures: collection of data elements
- List: Sequence data structures, these are indexable, mutable, defined by square brackets and elements are comma separated **Operations on list:**

```
len(list), append(element), insert(index, element), remove(element)
  (removes first occurrence only), list.append(element), list.extend(list),
  pop(index)
  ['one', 'two'].append(['one', 'two']) = ['one', 'two', ['one', 'two']]
  ['one', 'two'].extend(['one', 'two']) = ['one', 'two', 'one', 'two']
  del lst[1]
  list reverse: list.reverse()
  sorted(list), list.sort()
  lst = [1, 2, 3, 4, 5]; abc = lst; abc.append(6); print(lst)[[1, 2, 3, 4, 5]]
  5, 61
  1st and abc are pointers;
  string.split(' ')
  lst[index]
  lst[slice index start: slice index end]
  lst1 + lst2
  lst.count()
  for ele in lst: print(ele)
  List comprehensions: [i**2 \text{ for } i \text{ in range}(10) \text{ if } i*2 ==0]
  [[row[i] for row in matrix] for i in range(4)]
In [11]: xs = [3, 1, 2]
         print(xs, xs[2])
         print(xs[-1])
         xs[2] = 'foo'
         print(xs)
         xs.append('bar')
         print(xs)
         x = xs.pop()
         print(x, xs)
         [3, 1, 2] 2
         2
         [3, 1, 'foo']
         [3, 1, 'foo', 'bar']
         bar [3, 1, 'foo']
```

Slicing:

In addition to accessing list elements one at a time, Python provides concise syntax to access sublists; this is known as slicing:

```
In [12]: nums = list(range(5))
         print(nums)
         print(nums[2:4])
         print(nums[2:])
         print(nums[:2])
         print(nums[:])
         print(nums[:-1])
         nums[2:4] = [8, 9]
         print(nums)
         [0, 1, 2, 3, 4]
         [2, 3]
         [2, 3, 4]
         [0, 1]
         [0, 1, 2, 3, 4]
         [0, 1, 2, 3]
         [0, 1, 8, 9, 4]
```

Loops:

You can loop over the elements of a list like this:

```
Control flow: while loop
While loop: block of code runs until a test expression is true; lst = [10, 20,
30, 40, 50]
index = 0
while index < len(lst):
product * = lst[index]
index += 1 # increment statement is important 0: 1*10, 1: 10*20, 2: 200*30,...
We can use an else block when the test condition fails;
Control flow: for loop
Used to iterate over a sequence; for element in sequence:
statement(s) for ele in 1st:
product *=ele
range(): range(10) will generate a list of 10 numbers from 0 to 9;
 In [13]: animals = ['cat', 'dog', 'monkey']
          for animal in animals:
              print(animal)
          cat
          dog
          monkey
```

```
In [14]: animals = ['cat', 'dog', 'monkey']
    for idx, animal in enumerate(animals):
        print('#%d: %s' % (idx + 1, animal))

#1: cat
#2: dog
#3: monkey
```

List comprehensions:

When programming, frequently we want to transform one type of data into another. As a simple example, consider the following code that computes square numbers:

We can simplified by above code

Dictionary

An unordered collection of key value pairs; O(1) for time complexity for search tasks; Dictionary is mutable; Operations: dict.pop(key), dict.clear(), dict.fromkeys(list, values), dict.items(), .keys(), .values(), .copy()
 Dictionary Comprehension:
 for pair in d.items(): print(pair)
 {k:v for k,v in d.items() if v>2}
 {k:v for k+'c',v*2 in d.items() if v>2}

```
In [17]: d = {'cat': 'cute', 'dog': 'furry'} # Create a new dictionary with
         some data
         print(d['cat'])
         print('cat' in d)
         d['fish'] = 'wet'
         print(d['fish'])
         # print(d['monkey'])
         print(d.get('monkey', 'N/A'))
         print(d.get('fish', 'N/A'))
         del d['fish']
         print(d.get('fish', 'N/A'))
         cute
         True
         wet
         N/A
         wet
         N/A
In [19]: | d = {'person': 2, 'cat': 4, 'spider': 8}
         for animal, legs in d.items():
             print('A {0} has {1} legs'.format(animal, legs))
         # Prints "A person has 2 legs", "A cat has 4 legs", "A spider has 8
         legs"
         A person has 2 legs
         A cat has 4 legs
         A spider has 8 legs
In [20]: nums = [0, 1, 2, 3, 4]
         even num to square = \{x: x ** 2 \text{ for } x \text{ in nums if } x % 2 == 0\}
         print(even num to square) # Prints "{0: 0, 2: 4, 4: 16}"
         \{0: 0, 2: 4, 4: 16\}
```

Sets

```
• Sets are unordered collection of unique items; Mutable, non-indexable; S = {1, 2, 3} Sets does not allow duplicate numbers; set([1,2,3,1]) = (1, 2, 3) Operations: set.update(elements or sets), set.discard(element), set.remove(element), set.pop(), s.clear() Set1 | Set2: Union; Set1.union(Set2) Set1 & Set2, Set1.intersection(Set2) Set1 - Set2, Set1.difference(Set2) Set1 - Set2, Set1.symmetric_difference(Set2)): Union - Intersection Frozenset: immutable sets: Set1 = frozenset([1,2,3,4])
```

```
In [21]: animals = {'cat', 'dog'}
          print('cat' in animals)
          print('fish' in animals)
          animals.add('fish')
          print('fish' in animals)
          print(len(animals))
          animals.add('cat')
          print(len(animals))
          animals.remove('cat')
          print(len(animals))
          True
         False
         True
          3
          3
          2
In [22]: | animals = {'cat', 'dog', 'fish'}
          for idx, animal in enumerate(animals):
              print('#%d: %s' % (idx + 1, animal))
          # Prints "#1: fish", "#2: dog", "#3: cat"
         #1: fish
         #2: dog
         #3: cat
In [23]: from math import sqrt
          nums = \{int(sqrt(x)) \text{ for } x \text{ in } range(30)\}
          print(nums) # Prints "{0, 1, 2, 3, 4, 5}"
          \{0, 1, 2, 3, 4, 5\}
```

Tuple

- A tuple is similar to list; Tuple is immutable, its elements cannot be altered;
- T = "abcd", # comma is important to create a tuple
- Tuple access: T[1]
- Changing a tuple: a list in a tuple is mutable;
- Concat tuples using +
- Deletion: whole tuple will be deleted
- Tuple.count(), tuple.index(element), element in tuple, element not in tuple, len(tuple), sorted(tuple), min(tuple), max(tuple), sum(tuple)

Functions

- Functions: a group of related statements that perform a specific task;
- Converts a program into smaller chunk which makes management easy

def function():

- Doc string "statements return
- Doc strings is written to explain the working of the function (function.doc)
- Scope and Life Time of Variables: Portion of the code where the variable is recognized and Lifetime is the period throughout
- which the variable exists in memory
- Variable inside a function are local variables which are destroyed once the function finishes execution; - - - Global variables are not destroyed unless deleted;
- Program to print highest common factor (HCF):
- def computeHCF(a, b): """
- Computing HCF of two numbers """
- Smaller =b if a>b else a
- hcf = 1

```
`for i in range(1, smaller + 1):
```

```
if (a\%i==0) and (b\%i==0): hcf = i return hcf
```

```
In [25]: def sign(x):
             if x > 0:
                  return 'positive'
             elif x < 0:
                  return 'negative'
             else:
                  return 'zero'
         for x in [-1, 0, 1]:
             print(sign(x))
         # Prints "negative", "zero", "positive"
         negative
         zero
         positive
```

Classes

• The syntax for defining classes in Python is straightforward:

```
In [26]: class Greeter(object):
             # Constructor
             def __init__(self, name):
                 self.name = name # Create an instance variable
             # Instance method
             def greet(self, loud=False):
                 if loud:
                     print('HELLO, %s!' % self.name.upper())
                     print('Hello, %s' % self.name)
         g = Greeter('Fred')
         g.greet()
         g.greet(loud=True)
         Hello, Fred
```

HELLO, FRED!

Numpy

Numpy is the core library for scientific computing in Python. It provides a high-performance multidimensional array object, and tools for working with these arrays.

Arrays

A numpy array is a grid of values, all of the same type, and is indexed by a tuple of nonnegative integers. The number of dimensions is the rank of the array; the shape of an array is a tuple of integers giving the size of the array along each dimension.

```
In [27]: import numpy as np

a = np.array([1, 2, 3])
print(type(a))
print(a.shape)
print(a[0], a[1], a[2])
a[0] = 5
print(a)

b = np.array([[1,2,3],[4,5,6]])
print(b.shape)
print(b[0, 0], b[0, 1], b[1, 0])

<class 'numpy.ndarray'>
(2.)
```

```
<class 'numpy.ndarray'>
(3,)
1 2 3
[5 2 3]
(2, 3)
1 2 4
```

```
In [28]: import numpy as np
         a = np.zeros((2,2))
         print(a)
         b = np.ones((1,2))
         print(b)
         c = np.full((2,2), 7)
         print(c)
         d = np.eye(2)
         print(d)
         e = np.random.random((2,2))
         print(e)
         [[0. 0.]
          [0. 0.]]
         [[1. 1.]]
         [[7 7]
          [7 7]]
         [[1. 0.]
          [0. 1.]]
         [[0.89032768 0.43775097]
          [0.65058856 0.74750872]]
```

Array indexing

- Numpy offers several ways to index into arrays.
- **Slicing:** Similar to Python lists, numpy arrays can be sliced. Since arrays may be multidimensional, you must specify a slice for each dimension of the array:

```
In [29]: import numpy as np
    a = np.array([[1,2,3,4], [5,6,7,8], [9,10,11,12]])
    b = a[:2, 1:3]
    print(a[0, 1])
    b[0, 0] = 77
    print(a[0, 1])
2
77
```

```
In [30]: import numpy as np
         a = np.array([[1,2,3,4], [5,6,7,8], [9,10,11,12]])
         row r1 = a[1, :]
         row_r2 = a[1:2, :]
         print(row r1, row r1.shape)
         print(row r2, row r2.shape)
         col r1 = a[:, 1]
         col r2 = a[:, 1:2]
         print(col_r1, col_r1.shape)
         print(col r2, col r2.shape)
         [5 6 7 8] (4,)
         [[5 6 7 8]] (1, 4)
         [ 2 6 10] (3,)
         [[2]
          [ 6]
          [10]] (3, 1)
```

• Integer array indexing: When you index into numpy arrays using slicing, the resulting array view will always be a subarray of the original array. In contrast, integer array indexing allows you to construct arbitrary arrays using the data from another array. Here is an example:

```
In [31]: import numpy as np

a = np.array([[1,2], [3, 4], [5, 6]])
    print(a[[0, 1, 2], [0, 1, 0]]) # Prints "[1 4 5]"
    print(np.array([a[0, 0], a[1, 1], a[2, 0]])) # Prints "[1 4 5]"
    print(a[[0, 0], [1, 1]]) # Prints "[2 2]"
    print(np.array([a[0, 1], a[0, 1]])) # Prints "[2 2]"

[1 4 5]
    [1 4 5]
    [2 2]
    [2 2]
```

```
In [32]: import numpy as np
         # Create a new array from which we will select elements
         a = np.array([[1,2,3], [4,5,6], [7,8,9], [10, 11, 12]])
         print(a)
         b = np.array([0, 2, 0, 1])
         print(a[np.arange(4), b]) # Prints "[ 1 6 7 11]"
         a[np.arange(4), b] += 10
         print(a)
         [[ 1 2
                  31
          [ 4
               5
                  6]
          [7 8 9]
          [10 11 12]]
         [1 6 7 11]
         [[11 2 3]
          [ 4 5 16]
          [17 8 9]
          [10 21 12]]
```

Boolean array indexing:

• Boolean array indexing lets you pick out arbitrary elements of an array. Frequently this type of indexing is used to select the elements of an array that satisfy some condition. Here is an example:

```
In [33]: import numpy as np

a = np.array([[1,2], [3, 4], [5, 6]])
bool_idx = (a > 2)
print(bool_idx)
print(a[bool_idx])
print(a[a > 2])  # Prints "[3 4 5 6]"

[[False False]
        [ True True]
        [ True True]]
        [3 4 5 6]
        [3 4 5 6]
```

Datatypes

• Every numpy array is a grid of elements of the same type. Numpy provides a large set of numeric datatypes that you can use to construct arrays. Numpy tries to guess a datatype when you create an array, but functions that construct arrays usually also include an optional argument to explicitly specify the datatype. Here is an example:

numpy math

```
In [35]: import numpy as np
         x = np.array([[1,2],[3,4]], dtype=np.float64)
         y = np.array([[5,6],[7,8]], dtype=np.float64)
         print(x + y)
         print(np.add(x, y))
         print(x - y)
         print(np.subtract(x, y))
         print(x * y)
         print(np.multiply(x, y))
         print(x / y)
         print(np.divide(x, y))
         print(np.sqrt(x))
         [[ 6. 8.]
          [10. 12.]]
         [[ 6. 8.]
          [10. 12.]]
         [-4. -4.]
          [-4. -4.]]
         [-4. -4.]
          [-4. -4.]
         [[ 5. 12.]
          [21. 32.]]
         [[ 5. 12.]
          [21. 32.]]
         [[0.2
                       0.33333333
          [0.42857143 0.5
                       0.333333331
         [[0.2
          [0.42857143 0.5
                       1.414213561
         [[1.
          [1.73205081 2.
                                 11
```

```
In [36]: import numpy as np
         x = np.array([[1,2],[3,4]])
         y = np.array([[5,6],[7,8]])
         v = np.array([9,10])
         w = np.array([11, 12])
         print(v.dot(w))
         print(np.dot(v, w))
         print(x.dot(v))
         print(np.dot(x, v))
         print(x.dot(y))
         print(np.dot(x, y))
         219
         219
         [29 67]
         [29 67]
         [[19 22]
          [43 50]]
         [[19 22]
          [43 50]]
In [37]: import numpy as np
         x = np.array([[1,2],[3,4]])
         print(np.sum(x))
         print(np.sum(x, axis=0))
         print(np.sum(x, axis=1))
         10
         [4 6]
         [3 7]
In [38]: | import numpy as np
         x = np.array([[1,2], [3,4]])
         print(x)
         print(x.T)
         v = np.array([1,2,3])
         print(v)
         print(v.T)
         [[1 2]
          [3 4]]
         [[1 3]
          [2 4]]
         [1 2 3]
         [1 2 3]
```

Broadcasting

• Broadcasting is a powerful mechanism that allows numpy to work with arrays of different shapes when performing arithmetic operations. Frequently we have a smaller array and a larger array, and we want to use the smaller array multiple times to perform some operation on the larger array.

```
In [39]: import numpy as np
         x = np.array([[1,2,3], [4,5,6], [7,8,9], [10, 11, 12]])
         v = np.array([1, 0, 1])
         y = np.empty_like(x)
         for i in range(4):
             y[i, :] = x[i, :] + v
         print(y)
         [[22
                  4]
          [5 5 7]
          [8 8 10]
          [11 11 13]]
In [40]: import numpy as np
         x = np.array([[1,2,3], [4,5,6], [7,8,9], [10, 11, 12]])
         v = np.array([1, 0, 1])
         vv = np.tile(v, (4, 1))
         print(vv)
         y = x + vv
         print(y)
         [[1 0 1]
          [1 0 1]
          [1 0 1]
          [1 0 1]]
         [[2 2 4]
          [5 5 7]
          [8 8 10]
          [11 11 13]]
In [41]: import numpy as np
         x = np.array([[1,2,3], [4,5,6], [7,8,9], [10, 11, 12]])
         v = np.array([1, 0, 1])
         y = x + v
         print(y)
         [[2 2 4]
          [557]
          [8 8 10]
          [11 11 13]]
```

```
In [42]: import numpy as np
         # Compute outer product of vectors
         v = np.array([1,2,3]) \# v \ has \ shape (3,)
         w = np.array([4,5]) # w has shape (2,)
         print(np.reshape(v, (3, 1)) * w)
         x = np.array([[1,2,3], [4,5,6]])
         print(x + v)
         print((x.T + w).T)
         print(x + np.reshape(w, (2, 1)))
         print(x * 2)
         [[4 5]
          [ 8 10]
          [12 15]]
         [[2 4 6]
          [5 7 9]]
         [[ 5 6 7]
          [ 9 10 11]]
         [[ 5 6 7]
          [ 9 10 11]]
         [[2 4 6]
```

SciPy

[8 10 12]]

Numpy provides a high-performance multidimensional array and basic tools to compute with and manipulate these arrays. SciPy builds on this, and provides a large number of functions that operate on numpy arrays and are useful for different types of scientific and engineering applications.

```
In [6]: import numpy as np
    from scipy.misc import imread, imsave, imresize
    import matplotlib.pyplot as plt
    img = imread('cat.jpeg')
    print(img.dtype, img.shape) # Prints "uint8 (400, 248, 3)"
    img_tinted = img * [1, 0.95, 0.9]
    img_tinted = imresize(img_tinted, (300, 300))
    imsave('cat_tinted.jpeg', img_tinted)
    plt.subplot(1, 2, 1)
    plt.imshow(img)
    plt.subplot(1, 2, 2)
    plt.imshow(np.uint8(img_tinted))
    plt.show()
```

uint8 (400, 248, 3)



MATLAB files

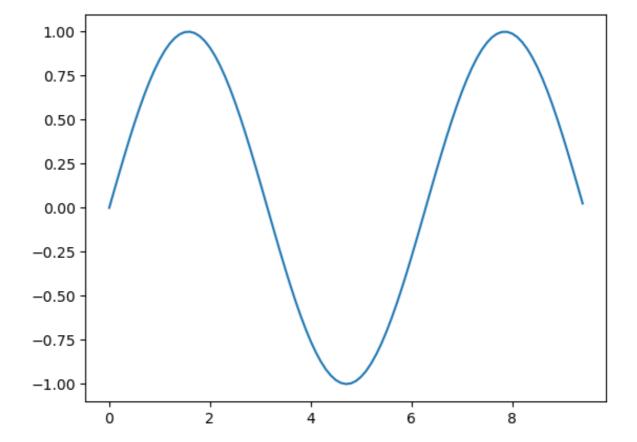
The functions scipy.io.loadmat and scipy.io.savemat allow you to read and write MATLAB files.

```
In [7]: import numpy as np
        from scipy.spatial.distance import pdist, squareform
        x = np.array([[0, 1], [1, 0], [2, 0]])
        print(x)
        d = squareform(pdist(x, 'euclidean'))
        print(d)
        [[0 1]
         [1 0]
         [2 0]]
                      1.41421356 2.23606798]
        [[0.
         [1.41421356 0.
                                 1.
         [2.23606798 1.
                                 0.
                                           ]]
```

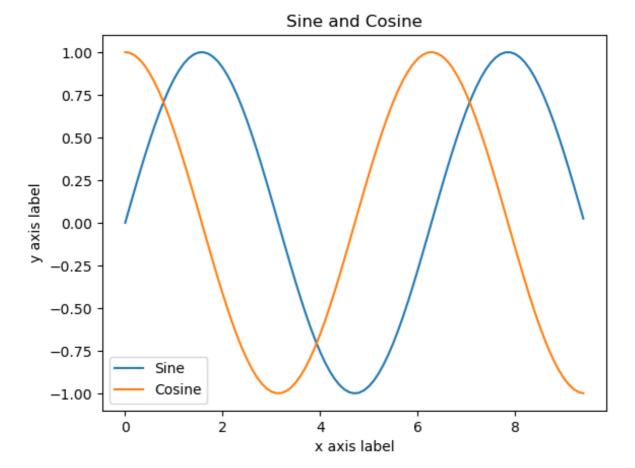
Plotting

• The most important function in matplotlib is plot, which allows you to plot 2D data. Here is a simple example:

```
In [1]: import numpy as np
   import matplotlib.pyplot as plt
   x = np.arange(0, 3 * np.pi, 0.1)
   y = np.sin(x)
   plt.plot(x, y)
   plt.show()
```



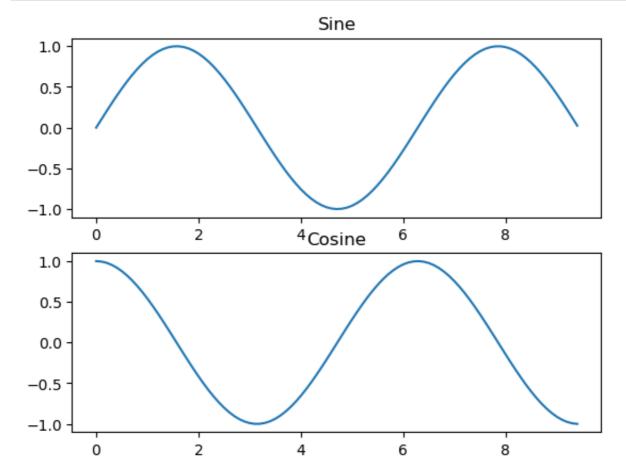
```
In [2]: import numpy as np
    import matplotlib.pyplot as plt
    x = np.arange(0, 3 * np.pi, 0.1)
    y_sin = np.sin(x)
    y_cos = np.cos(x)
    plt.plot(x, y_sin)
    plt.plot(x, y_cos)
    plt.xlabel('x axis label')
    plt.ylabel('y axis label')
    plt.title('Sine and Cosine')
    plt.legend(['Sine', 'Cosine'])
    plt.show()
```



Subplots

• You can plot different things in the same figure using the subplot function. Here is an example:

```
In [3]: import numpy as np
    import matplotlib.pyplot as plt
    x = np.arange(0, 3 * np.pi, 0.1)
    y_sin = np.sin(x)
    y_cos = np.cos(x)
    plt.subplot(2, 1, 1)
    plt.plot(x, y_sin)
    plt.title('Sine')
    plt.subplot(2, 1, 2)
    plt.plot(x, y_cos)
    plt.title('Cosine')
    plt.show()
```



Images

• You can use the imshow function to show images. Here is an example:

```
In [4]: import numpy as np
    from scipy.misc import imread, imresize
    import matplotlib.pyplot as plt
    img = imread('cat.jpeg')
    img_tinted = img * [1, 0.95, 0.9]
    plt.subplot(1, 2, 1)
    plt.imshow(img)
    plt.subplot(1, 2, 2)
    plt.imshow(np.uint8(img_tinted))
    plt.show()
```



In []: