

# Recap: Python For AI + Numpy

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STUDY LOCALLY. LIVE GLOBALLY.

# Containers - lists, dictionaries, sets, and tuples.

A list is the Python equivalent of an array, but is resizable and can contain elements of different types.

```
xs = [3, 1, 2] # Create a list
```

```
xs.append('bar') # Add a new element to the end of the list
```

```
nums = list(range(5)) # [0,1,2,3,4]
```

```
print(nums[2:4]) # Get a slice from index 2 to 4 (exclusive); prints "[2, 3]"
```

```
print(nums[2:]) # Get a slice from index 2 to the end; prints "[2, 3, 4]"
```

```
print(nums[:2]) # Get a slice from the start to index 2 (exclusive); prints "[0, 1]"
```

```
print(nums[:]) # Get a slice of the whole list; prints "[0, 1, 2, 3, 4]"
```

```
print(nums[:-1]) # Slice indices can be negative; prints "[0, 1, 2, 3]"
```

# Containers - lists

Iterating a list:

```
animals = ['cat', 'dog', 'monkey']  
for animal in animals:  
    print(animal)
```

```
animals = ['cat', 'dog', 'monkey']  
for idx, animal in enumerate(animals):  
    print('#{}: {}'.format(idx + 1, animal))
```

# Containers - lists

Iterating a list:

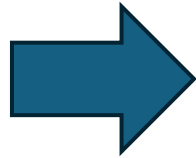
```
animals = ['cat', 'dog', 'monkey']  
for animal in animals:  
    print(animal)
```

```
animals = ['cat', 'dog', 'monkey']  
for idx, animal in enumerate(animals):  
    print('#{}: {}'.format(idx + 1, animal))
```

# Containers - lists

List comprehension:

```
nums = [0, 1, 2, 3, 4]
squares = []
for x in nums:
    squares.append(x ** 2)
```



```
nums = [0, 1, 2, 3, 4]
squares = [x ** 2 for x in nums]

even_squares = [x ** 2 for x in nums if x % 2 == 0]
```

# Containers - Dictionaries

A dictionary stores (key, value) pairs, similar to a `Map` in Java or an object in Javascript.

```
d = {'cat': 'cute', 'dog': 'furry'} # Create a new dictionary with some data  
print(d['cat'])    # Get an entry from a dictionary; prints "cute"  
print('cat' in d)  # Check if a dictionary has a given key; prints "True"
```

```
d['fish'] = 'wet' # Set an entry in a dictionary  
print(d['fish'])  # Prints "wet"
```

```
print(d['monkey']) # KeyError: 'monkey' not a key of d
```

```
print(d.get('monkey', 'N/A')) # Get an element with a default; prints "N/A"  
print(d.get('fish', 'N/A'))  # Get an element with a default; prints "wet"
```

# Containers - Dictionaries

Iterating a dictionary:

```
d = {'person': 2, 'cat': 4, 'spider': 8}
for animal, legs in d.items():
    print('A {} has {} legs'.format(animal, legs))
```

```
nums = [0, 1, 2, 3, 4]
even_num_to_square = {x: x ** 2 for x in nums if x % 2 == 0}
print(even_num_to_square)
```

# Containers - Sets

A set is an unordered collection of distinct elements.

```
animals = {'cat', 'dog'}  
print('cat' in animals) # Check if an element is in a set; prints "True"  
print('fish' in animals) # prints "False"
```

```
animals.add('fish')    # Add an element to a set  
print('fish' in animals)  
print(len(animals))
```

```
animals.add('cat')     # Adding an element that is already in the set does nothing  
print(len(animals))  
animals.remove('cat')  # Remove an element from a set  
print(len(animals))
```



# Containers - Sets

Iterating a set:

```
animals = {'cat', 'dog', 'fish'}  
for idx, animal in enumerate(animals):  
    print('#{}: {}'.format(idx + 1, animal))  
  
print({int(sqrt(x)) for x in range(30)})
```

# Containers - Tuples

A tuple is an (immutable) ordered list of values.

```
t = (5, 6)    # Create a tuple  
print(type(t))
```

```
t[0] = 1 #TypeError: 'tuple' object does not support item assignment
```

# Containers - Tuples

A tuple is an (immutable) ordered list of values.

```
t = (5, 6)    # Create a tuple  
print(type(t))
```

```
t[0] = 1 #TypeError: 'tuple' object does not support item assignment
```

# Class

Classes are a way by which, in object-oriented programming, we can create our own type of data and give them names.

```
class BacGrade:
    def __init__(self):
        self.grades = ["F","A","C","B","D","E"]

    def top(self):
        grades_sorted = sorted(self.grades)
        return grades_sorted[0]

    @staticmethod
    def format(grade='f'):

        return f'My BacII grade is {grade.upper()}'
```

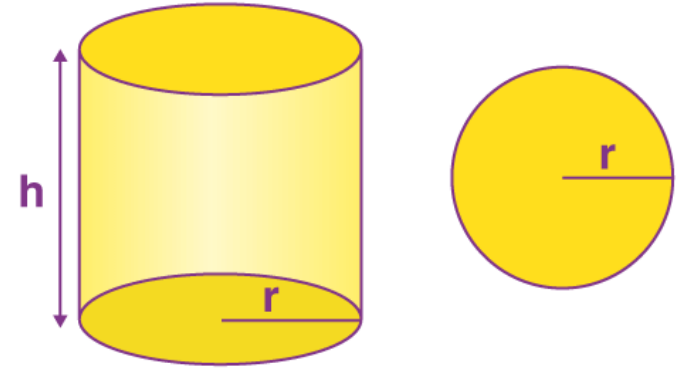
```
bac = BacGrade()
```

```
bac.top() # class method
```

```
BacGrade.format('a') #static  
method
```

# Class - Inheritance

Inheritance is, perhaps, the most powerful feature of object-oriented programming. It just so happens that you can create a class that “inherits” methods, variables, and attributes from another class.



```
class Circle:
    def __init__(self, radius=1.0):
        self.radius = radius

    def get_area_or_volume(self):
        return self.radius**2*math.pi
```

```
c = Circle()
print(c.get_area_or_volume())
```

```
class Cylinder(Circle):
    def __init__(self, radius=1.0, height=2.0):
        super().__init__(radius)
        self.height = height

    def get_area_or_volume(self):
        return self.radius**2*math.pi*self.height
```

```
cy = Cylinder()
print(cy.get_area_or_volume())
```

# File/I/O

File I/O is the ability of a program to take a file as input or create a file as output.

*open("names.txt", "w") # open or create a names.txt file and overwrite new content.*

*open("names.txt", "a") # open a names.txt file and add new content.*

*open("names.txt", "r") # open a names.txt file and read content.*

```
file = open("names.txt", "a")  
file.write(name)  
file.close()
```

*Or*

```
with open("names.txt", "a") as file:  
    file.write(f"{name}\n")
```

# File/I/O

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*open("names.txt", "w") # open or create a names.txt file and overwrite new content.*

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```
file = open("names.txt", "a")  
file.write(name)  
file.close()
```

*Or*

```
with open("names.txt", "a") as file:  
    file.write(f"{name}\n")
```

*For an image file,*

```
from PIL import Image  
image = Image.open(path)
```

# ChatGPT

Prompt: Write a python script to open a RGB image file and convert it to grayscale

```
from PIL import Image  
# Open the RGB image file  
image = Image.open("input_image.jpg")  
# Convert the image to grayscale  
grayscale_image = image.convert("L")  
# Save the grayscale image  
grayscale_image.save("output_grayscale_image.jpg")  
# Show the grayscale image  
grayscale_image.show()
```



# Remarks

- Put in 10,000 hours of deliberate effort and work to become a successful machine learning (or python) practitioner.



[Andrej Karpathy](#)

# 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.

```
import numpy as np  
print(np.__version__)
```

# Numpy

Creating a numpy array:

*`a = np.array([1, 2, 3])` #1D - i.e., rank 1*

*`b = np.array([[1,2,3],[4,5,6]])` #2D - i.e., rank 2*

Other ways:

*`np.zeros (shape)` # e.g. `np.zeros((2,3))`*

*`np.ones, np.eye, np.full` etc.*

# Numpy - Array Indexing

You can index a numpy array in the same way you can index a list.

```
a = np.array([[1,2,3,4], [5,6,7,8], [9,10,11,12]])
```

```
b = a[:2, 1:3] # a subview of a
```

```
a[0, 1] # an item of a
```

```
row_r1 = a[1, :] # 1D view of the second row of a
```

```
row_r2 = a[1:2, :] # 2D view of the second row of a
```

```
row_r3 = a[[1], :] # 2D view of the second row of a
```

# Numpy -Array Indexing

Integer array indexing:

```
a = np.array([[1,2], [3, 4], [5, 6]])
```

```
a[[0, 1, 2], [0, 1, 0]] # 1D view
```

```
np.array([a[0, 0], a[1, 1], a[2, 0]]) # 1D view
```

Boolean array indexing:

```
a = np.array([[1,2], [3, 4], [5, 6]])
```

```
bool_idx = (a > 2)
```

```
a[bool_idx] # OR
```

```
a[a>2]
```

# Datatypes & Operations

## Datatypes:

*`x = np.array([1, 2])` # Let numpy choose the datatype*

*`y = np.array([1.0, 2.0])` # Let numpy choose the datatype*

*`z = np.array([1, 2], dtype=np.int64)` # Force a particular datatype*

## Operations:

*`np.add(x,y)` #  $x + y$*

*`np.subtract(x,y)` #  $x - y$*

*`np.multiply(x,y)` #  $x * y$*

*`np.divide(x,y)` #  $x / y$*

## Matrix Multiplication:

*`x = np.array([[1,2],[3,4]])` # (2,2)*

*`y = np.array([[5,6],[7,8]])` # (2,2)*

*`x.dot(y)`*

*`np.dot(x, y)`*

*`x @ y`*

# Datatypes & Operations

Matrix-Vector Multiplication:

```
x = np.array([[1,2],[3,4]]) # (2,2)
```

```
v = np.array([9,10]) # (2,)
```

```
print(x.dot(v)) # (2,)
```

```
print(np.dot(x, v)) # (2,)
```

```
print(x @ v) # (2,)
```

# Datatypes & Operations

Dot product:

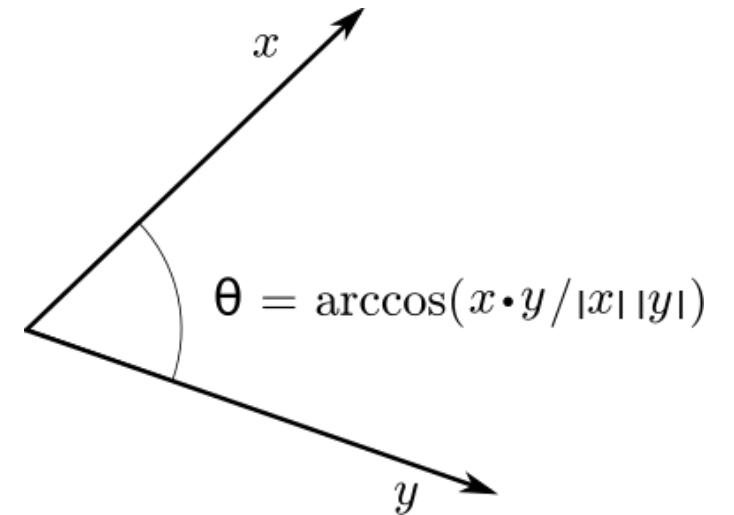
```
v = np.array([9, 10])
```

```
w = np.array([11, 12])
```

```
# Inner product of vectors; both produce 219
```

```
print(v.dot(w))
```

```
print(np.dot(v, w))
```





# Numpy Array Reshaping & Ravelling

```
a1 = np.arange(1, 13)
```

1	2	3	4	5	6	7	8	9	10	11	12
---	---	---	---	---	---	---	---	---	----	----	----

→

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

```
a1.reshape(3, 4)
a1.reshape(-1, 4)
a1.reshape(3, -1)
.ravel() # back to 1D
```

## 3D array from 2D arrays

```
a1 = np.arange(1, 13).reshape(3, 4)
a2 = np.arange(13, 25).reshape(3, -1)
```


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

```
# stack along axis 2
a3_2 = np.stack((a1, a2), axis=2)
a3_2.shape: (3, 4, 2)
```

```
# retrieve a1
a3_2[:, :, 0]
```

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

# Broadcasting

X			y		
1	0	1	1	0	1
1	2	3			
1	2	3			
1	2	2	Y_stracked		
1	2	1	1	0	1
2	2	1	1	0	1
1	1	1	1	0	1
			1	0	1
			1	0	1
			1	0	1
			1	0	1
			1	0	1
			1	0	1

How to compute  $\mathbf{X} + \mathbf{y}$ :

Option 1:

```
for i in range(7):  
     $X[i, :] + y$ 
```

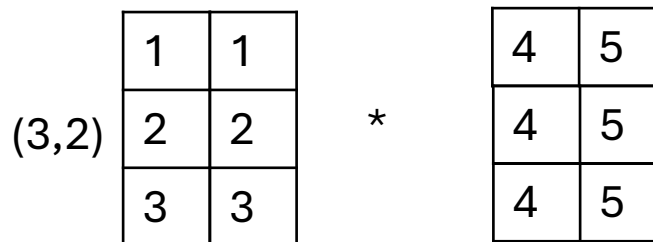
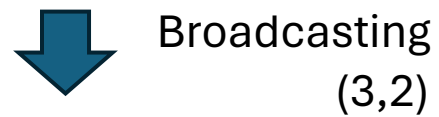
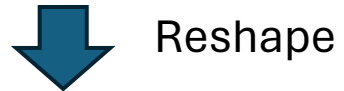
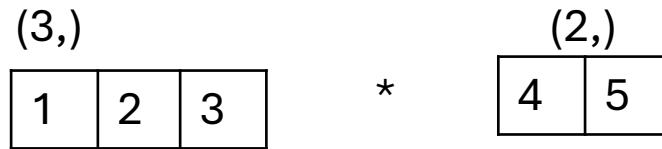
Option 2:

```
 $X + Y\_stracked$ 
```

Broadcasting:

```
 $X + y \# (7,3)$ 
```

# Broadcasting – Cases



# Broadcasting – Cases

(2,3)

1	2	3
4	5	6

+

(3,)

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



Broadcasting

(2,3)

1	2	3
4	5	6

+

(2,3)

1	2	3
1	2	3

# Matplotlib

Import:

```
import matplotlib.pyplot as plt  
%matplotlib inline
```

# Matplotlib

plt.plot -> Line plot :

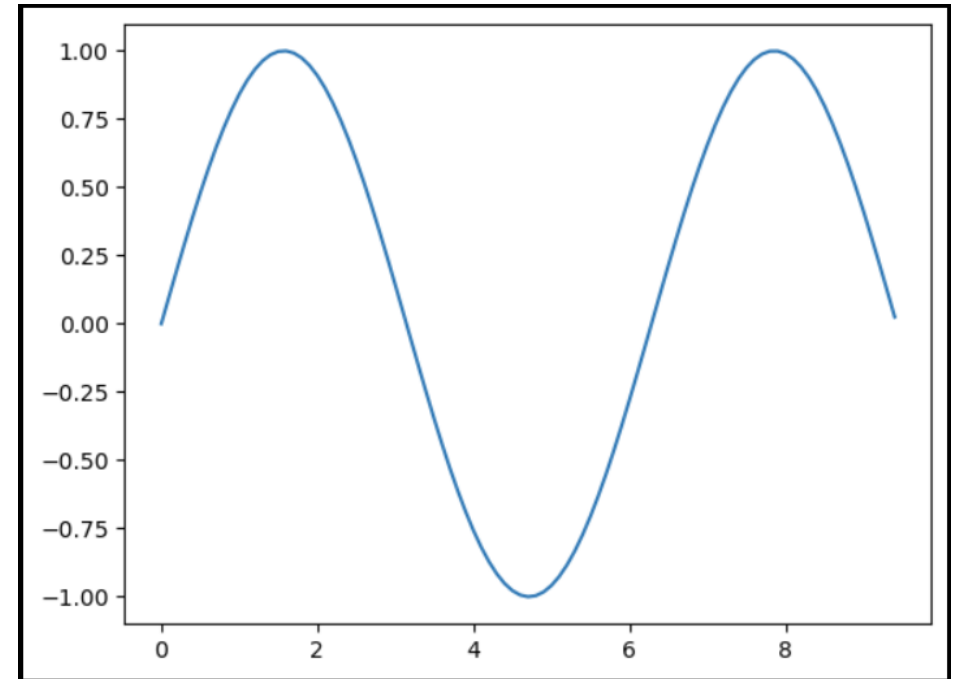
*# Compute the x and y coordinates for points on a sine curve*

*$x = np.arange(0, 3 * np.pi, 0.1)$*

*$y = np.sin(x)$*

*# Plot the points using matplotlib*

*$plt.plot(x, y)$*



# Matplotlib

plt.plot -> Line plot :

*y\_sin = np.sin(x)*

*y\_cos = np.cos(x)*

*# Plot the points using matplotlib*

*plt.plot(x, y\_sin)*

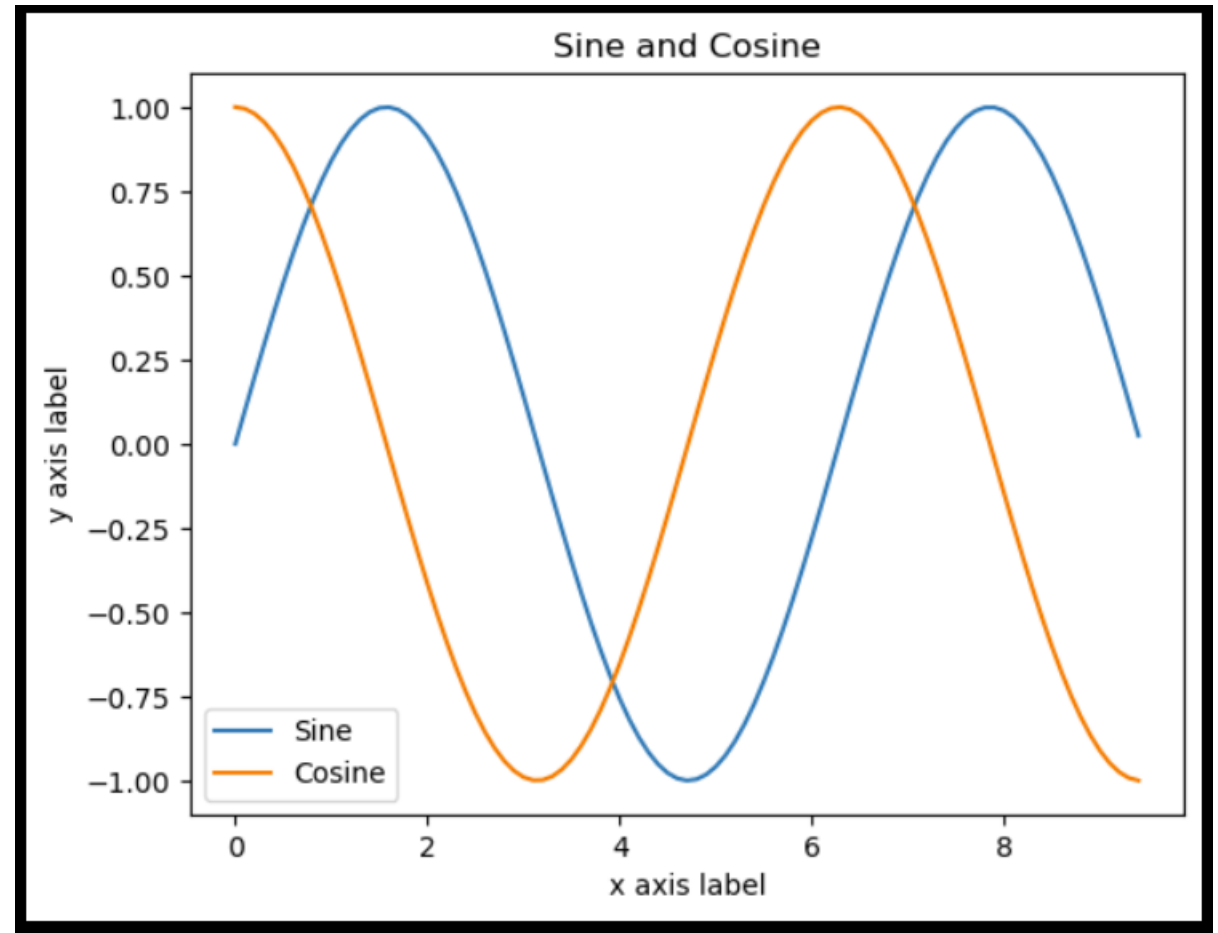
*plt.plot(x, y\_cos)*

*plt.xlabel('x axis label')*

*plt.ylabel('y axis label')*

*plt.title('Sine and Cosine')*

*plt.legend(['Sine', 'Cosine'])*



# ChatGPT

Prompt: Write a python script to generate a line plot of sin and cosine functions with proper axis labels

plt.plot -> Line plot :

*# Generate x values from 0 to  $2\pi$  with a step of 0.1*

*`x = np.arange(0, 2*np.pi, 0.1)`*

*# Calculate sin and cosine values for the x values*

*`y_sin = np.sin(x)`*

*`y_cos = np.cos(x)`*

*# Plot sin and cosine functions*

*`plt.plot(x, y_sin, label='sin(x)')`*

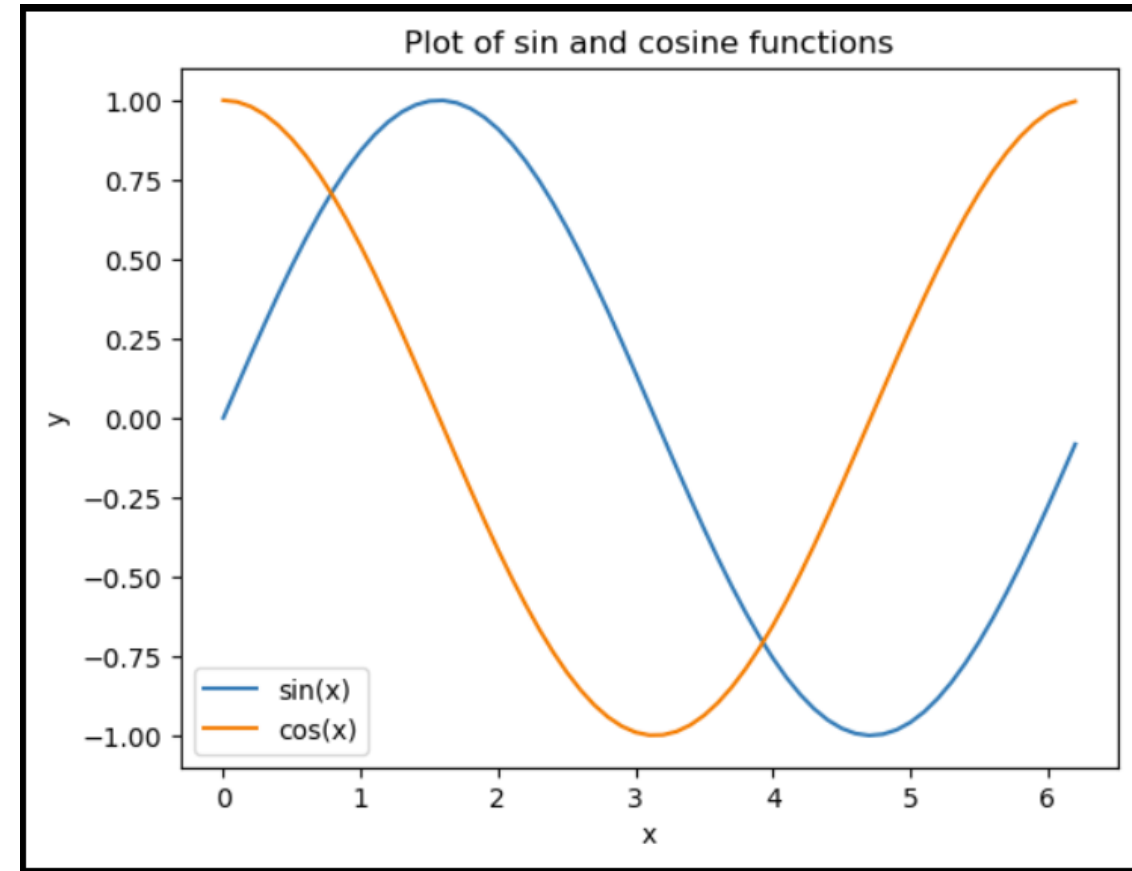
*`plt.plot(x, y_cos, label='cos(x)')`*

*# Add labels and title*

*`plt.xlabel('x')`*

*`plt.ylabel('y')`*

*`plt.title('Plot of sin and cosine functions')`*





# Matplotlib

plt.plot -> Line plot :

*#grid that has height 2 and width 1,  
# and set the first such subplot as active.*

*plt.subplot(2, 1, 1)*

*# Make the first plot*

*plt.plot(x, y\_sin)*

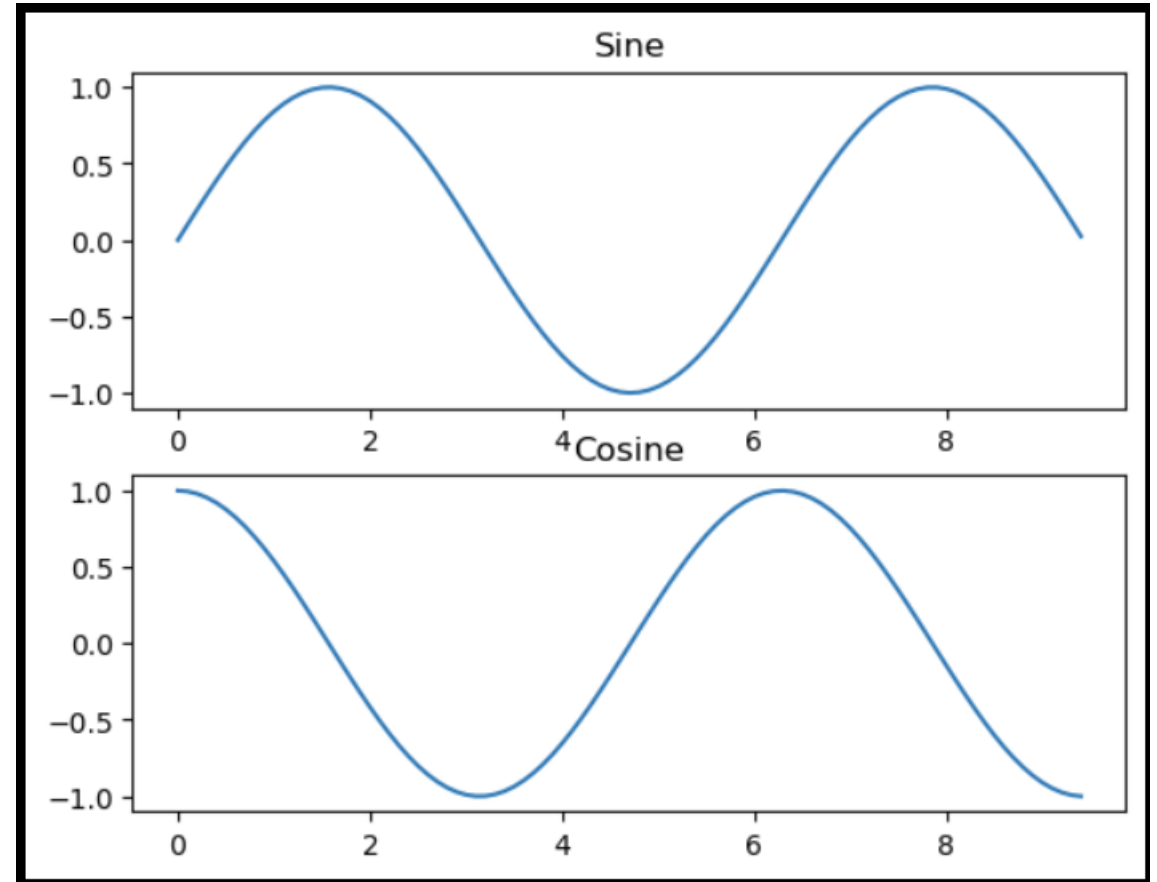
*plt.title('Sine')*

*# Set the second subplot as active, and make the  
second plot.*

*plt.subplot(2, 1, 2)*

*plt.plot(x, y\_cos)*

*plt.title('Cosine')*



# Matplotlib

plt.plot -> Line plot :

*#grid that has height 2 and width 1,  
# and set the first such subplot as active.*

*plt.subplot(2, 1, 1)*

*# Make the first plot*

*plt.plot(x, y\_sin)*

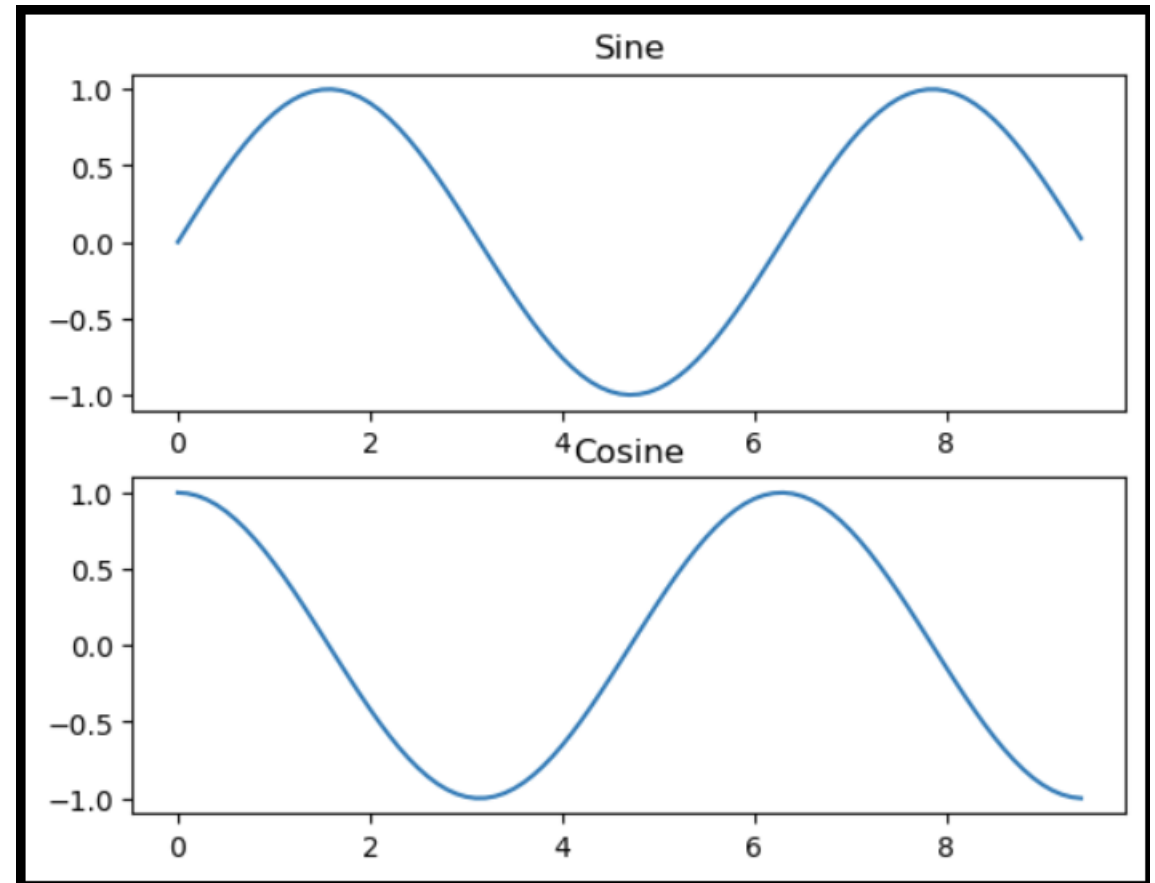
*plt.title('Sine')*

*# Set the second subplot as active, and make the  
second plot.*

*plt.subplot(2, 1, 2)*

*plt.plot(x, y\_cos)*

*plt.title('Cosine')*

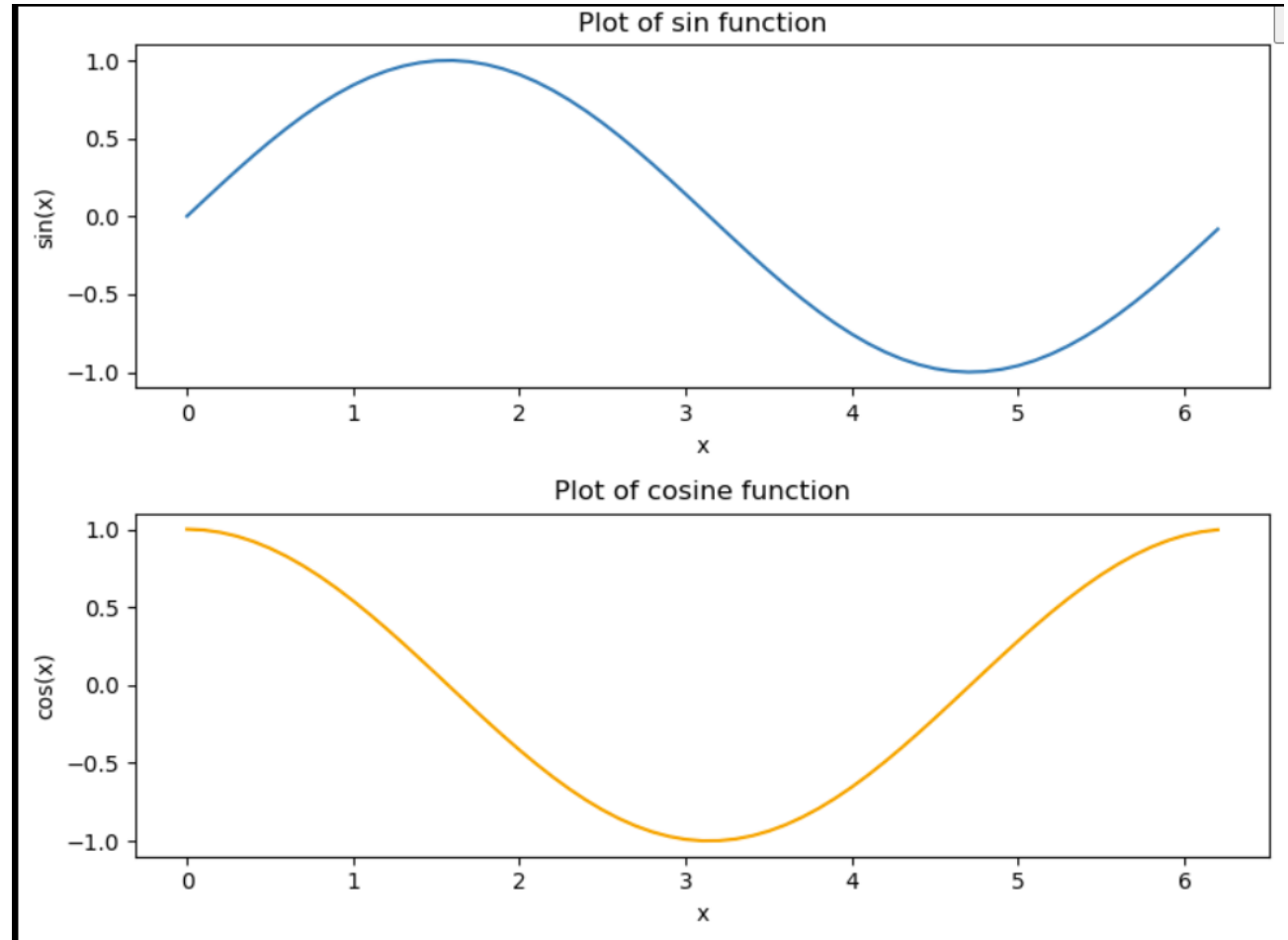


# ChatGPT

Prompt: Write a python script to generate two subplots of sin and cosine functions with proper axis labels

plt.plot -> Line plot :

```
# Create subplots  
fig, (ax1, ax2) = plt.subplots(2, 1, figsize=(8, 6))  
# Plot sin function in the first subplot  
ax1.plot(x, y_sin, label='sin(x)')  
ax1.set_xlabel('x')  
ax1.set_ylabel('sin(x)')  
ax1.set_title('Plot of sin function')  
# Plot cosine function in the second subplot  
ax2.plot(x, y_cos, label='cos(x)', color='orange')  
ax2.set_xlabel('x')  
ax2.set_ylabel('cos(x)')  
ax2.set_title('Plot of cosine function')  
# Adjust layout to prevent overlap  
plt.tight_layout()
```



# Simple Linear Regression

$$\hat{\beta} = (\mathbf{X}^{\top} \mathbf{X})^{-1} \mathbf{X}^{\top} \mathbf{y}$$

$$\hat{\mathbf{y}} = \mathbf{X} \hat{\beta}$$

$\hat{\beta}$  = ordinary least squares estimator

$\mathbf{X}$  = matrix regressor variable  $X$

$\top$  = matrix transpose

$\mathbf{y}$  = vector of the value of the response variable

$\hat{\mathbf{y}}$  = predicted values

$$\hat{\beta} = \begin{bmatrix} \beta_0 \\ \beta_1 \end{bmatrix}$$