

# CS229

# Python & Numpy

Saahil Jain, Xinkun Nie (adapted from Jingbo Yang, Zhihan Xiong)

# How is python related to with others?

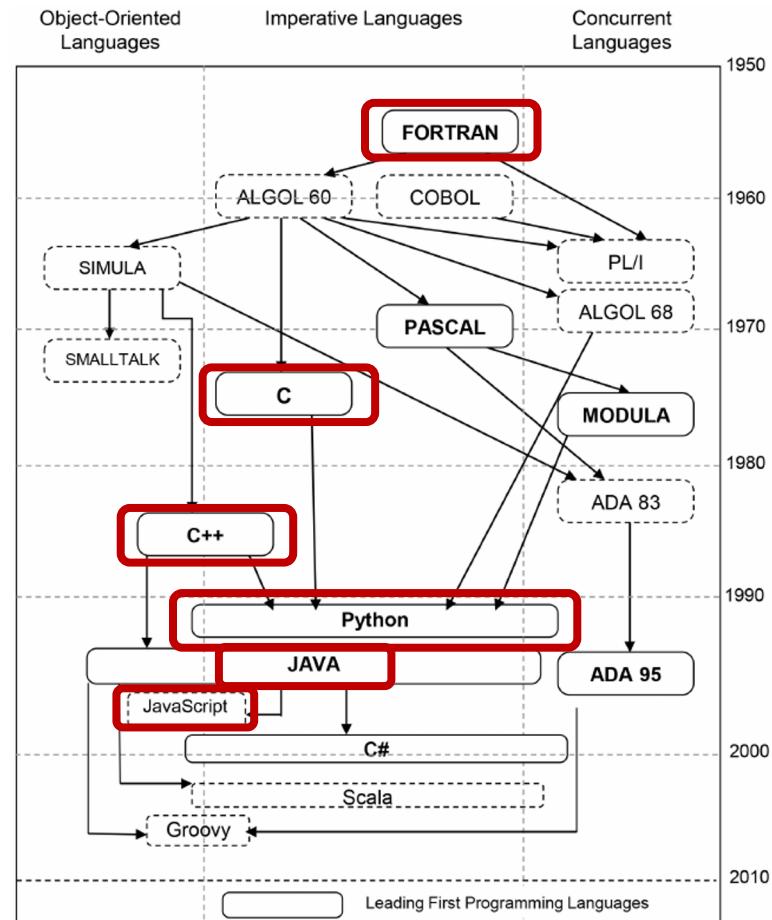
Python 2.0 released in 2000

(Python 2.7 “end-of-life” in 2020)

Python 3.0 released in 2008

(Python 3.6+ for CS 229)

Can run interpreted, like MATLAB



# Before you start

## Use Anaconda

Create a new environment (full Conda)

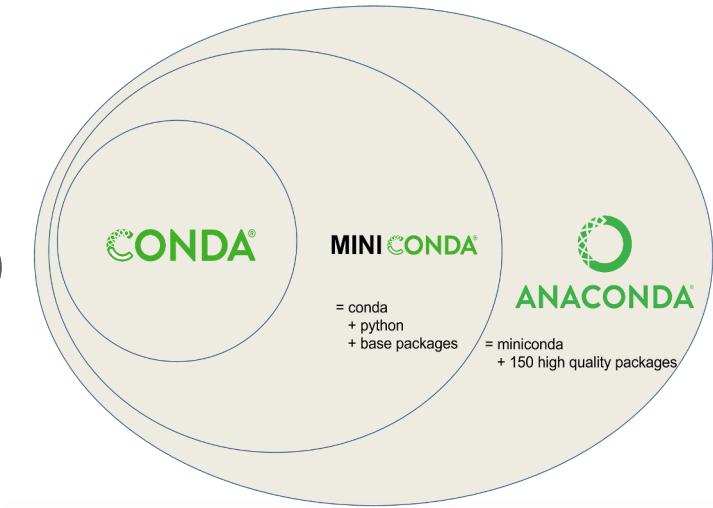
```
conda create -n cs229
```

Create an environment (Miniconda)

```
conda env create -f environment.yml
```

Activate an environment after creation

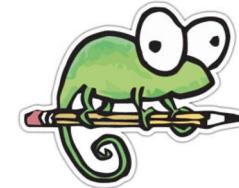
```
conda activate cs229
```



# Notepad is not your friend ...

Get a text editor/IDE

- PyCharm (IDE)
- Visual Studio Code (IDE??)
- Sublime Text (IDE??)
- Notepad ++/gedit
- Vim (for Linux)



# To make you more prepared

## PyCharm

- Great debugger
- Proper project management

The screenshot shows the PyCharm IDE interface with the following details:

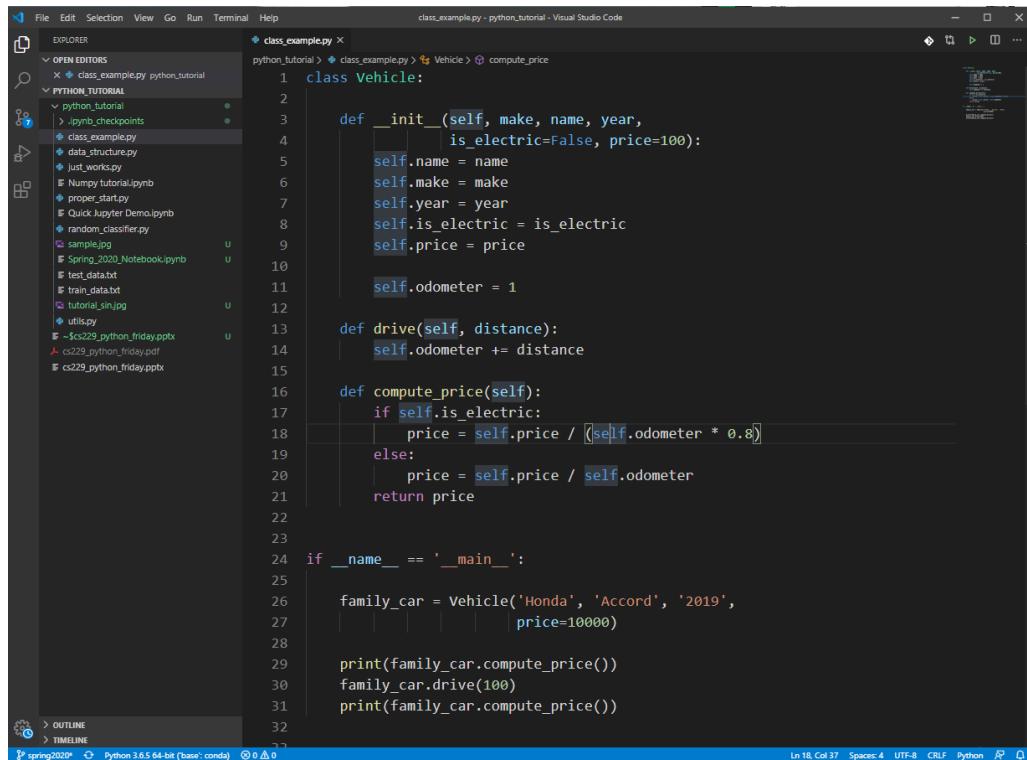
- Project Structure:** The left sidebar shows a project named "cs224u" containing various Python files such as requirements.txt, sst.py, script.ipynb, sst\_01\_overview.ipynb, sst\_02\_hand\_built\_feat.ipynb, sst\_03\_neural\_network.ipynb, tf\_autoencoder.py, tf\_model\_base.py, tf\_rnn\_classifier.py, tf\_shallow\_neural\_classifier.py, torch\_autoencoder.py, torch\_model\_base.py, torch\_rnn\_classifier.py, torch\_shallow\_neural\_classifier.py, torch\_subtree\_nn.py, torch\_tree\_nn.py, utils.py, vsm.py, vsm\_01\_distributional.ipynb, vsm\_02\_dimreduce.ipynb, and vsm\_03\_retrofitting.ipynb.
- Code Editor:** The main editor window displays a Python script with code related to opinion lexicons and word counts. A red circle highlights a breakpoint at line 36. The code includes imports for Counter and nltk.corpus, and defines functions like op\_ungigrams\_phi and fit\_S.
- Breakpoint Dialog:** A small window titled "Breakpoint" is open, showing the variable "positive" with its value set to "{'kindly', 'well-balanced', 'buoyant', 'dumbfounding', 'cheerful', 'excellent', 'graceful', 'thoughtful', 'gained', 'mesmerize', 'faultless', 'well-rounded'}".
- Results Window:** The bottom right shows the "Result" window with the same list of words: "{'kindly', 'well-balanced', 'buoyant', 'dumbfounding', 'cheerful', 'excellent', 'graceful', 'thoughtful', 'gained', 'mesmerize', 'faultless', 'well-rounded'}".
- Toolbars and Menus:** Standard PyCharm toolbars and menus are visible along the top and right edges of the interface.

FYI, professional version free for students: <https://www.jetbrains.com/student/>

# To make you more prepared

## Visual Studio Code

- Light weight
- Wide variety of plugins to enable support for all languages
- Better UI



A screenshot of the Visual Studio Code interface. The left sidebar shows a file tree with several Python files and other files like images and PDFs. The main editor area displays a Python script named `class_example.py`. The script defines a `Vehicle` class with methods for initialization, driving, and computing price based on whether it's electric or not. It also includes a check for the `__name__` variable and creates a `family_car` instance. The status bar at the bottom shows the file name, Python 3.6.5 64-bit (base) conda, and other system information.

```
class Example:
    def __init__(self, make, name, year,
                 is_electric=False, price=100):
        self.name = name
        self.make = make
        self.year = year
        self.is_electric = is_electric
        self.price = price

    def drive(self, distance):
        self.odometer += distance

    def compute_price(self):
        if self.is_electric:
            price = self.price / (self.odometer * 0.8)
        else:
            price = self.price / self.odometer
        return price

if __name__ == '__main__':
    family_car = Vehicle('Honda', 'Accord', '2019',
                         price=10000)
    print(family_car.compute_price())
    family_car.drive(100)
    print(family_car.compute_price())
```

# Basic Python

# String manipulation

## Formatting

```
print('I love CS229. (upper)'.upper())
print('I love CS229. (rjust 20)'.rjust(20))
print('we love CS229. (capitalize)'.capitalize())
print('        I love CS229. (strip)        '.strip())
```

## Concatenation

```
print('I like ' + str(cs_class_code) + ' a lot!')
print(f'{print} (print a function)')
print(f'{type(229)} (print a type)')
```

## Formatting

```
print('Old school formatting: {:.2F}'.format(1.358))
```

# List

## List creation

```
list_1 = ['one', 'two', 'three']
```

```
list_1.append(4)  
list_1.insert(0, 'ZERO')
```

```
list_2 = [1, 2, 3]  
list_1.extend(list_2)
```

```
long_list = [i for i in range(9)]  
long_long_list = [(i, j) for i in range(3)  
                  for j in range(5)]  
long_list_list = [[i for i in range(3)]  
                  for _ in range(5)]
```

```
sorted(random_list)
```

```
random_list_2 = [(3, 'z'), (12, 'r'), (6, 'e'),  
                  (8, 'c'), (2, 'g')]  
sorted(random_list_2, key=lambda x: x[1])
```

## Sorting

# Dictionary and Set

Set

(unordered, unique)

```
my_set = {i ** 2 for i in range(10)}  
{0, 1, 64, 4, 36, 9, 16, 49, 81, 25}
```

Dictionary  
(mapping)

```
my_dict = {(5 - i): i ** 2 for i in range(10)}  
{5: 0, 4: 1, 3: 4, 2: 9, 1: 16, 0: 25, -1: 36,  
-2: 49, -3: 64, -4: 81}  
dict_keys([5, 4, 3, 2, 1, 0, -1, -2, -3, -4])
```

Dictionary update

```
second_dict = {'a': 10, 'b': 11}  
my_dict.update(second_dict)
```

Iterate through items

```
for k, it in my_dict.items():  
    print(k, it)
```

# Numpy

# What is Numpy and why?

Numpy – package for vector and matrix manipulation  
Broadcasting and vectorization saves time and amount of code

FYI, if you are interested in how/why vectorization is faster, checkout the following topics (completely optional, definitely not within scope)

AVX instruction set (SIMD) and structure of x86 and RISC

OpenMP and CUDA for multiprocessing

Assembly-level optimization, memory stride, caching, etc.

Or even about memory management, virtualization

More bare metal → FPGA, TPU

# Convenient math functions, read before use!

Python Command	Description
np.linalg.inv	Inverse of matrix (numpy as equivalent)
np.linalg.eig	Get eigen value (Read documentation on eigh and numpy equivalent)
np.matmul	Matrix multiply
np.zeros	Create a matrix filled with zeros (Read on np.ones)
np.arange	Start, stop, step size (Read on np.linspace)
np.identity	Create an identity matrix
np.vstack	Vertically stack 2 arrays (Read on np.hstack)

# Your friend for debugging

Python Command	Description
array.shape	Get shape of numpy array
array.dtype	Check data type of array (for precision, for weird behavior)
type(stuff)	Get type of a variable
import pdb; pdb.set_trace()	Set a breakpoint ( <a href="https://docs.python.org/3/library/pdb.html">https://docs.python.org/3/library/pdb.html</a> )
print(f'My name is {name}')	Easy way to construct a message

# Basic Numpy usage

Initialization from Python lists

Lists with different types  
(Numpy auto-casts to higher precision, but it should be reasonably consistent)

Numpy supports many types of algebra on an entire array

```
array_1d = np.array([1, 2, 3, 4])
array_1by4 = np.array([[1, 2, 3, 4]])

large_array = np.array([i for i in range(400)])
large_array = large_array.reshape((20, 20))

from_list = np.array([1, 2, 3])
from_list_2d = np.array([[1, 2, 3.0], [4, 5, 6]])
from_list_bad_type = np.array([1, 2, 3, 'a'])

print(f'Data type of integer is {from_list.dtype}')
print(f'Data type of float is {from_list_2d.dtype}'')
```

```
array_1 + 5
array_1 * 5
np.sqrt(array_1)
np.power(array_1, 2)
np.exp(array_1)
np.log(array_1)
```

# Dot product and matrix multiplication

A few ways to write dot product

```
array_1 @ array_2  
array_1.dot(array_2)  
np.dot(array_1, array_2)
```

Matrix multiplication like Ax

```
weight_matrix = np.array([1, 2, 3, 4]).reshape(2, 2)  
sample = np.array([[50, 60]]).T  
np.matmul(weight_matrix, sample)
```

2D matrix multiplication

```
mat1 = np.array([[1, 2], [3, 4]])  
mat2 = np.array([[5, 6], [7, 8]])  
np.matmul(mat1, mat2)
```

Element-wise multiplication

```
a = np.array([i for i in range(10)]).reshape(2, 5)  
  
a * a  
np.multiply(a, a)  
np.multiply(a, 10)
```

# Broadcasting

Numpy compares dimensions of operands, then infers missing/mismatched dimensions so the operation is still valid. Be careful with *DIMENSIONS*

```
op1 = np.array([i for i in range(9)]).reshape(3, 3)
op2 = np.array([[1, 2, 3]])
op3 = np.array([1, 2, 3])
# Notice that the results here are DIFFERENT!
pp pprint(op1 + op2)
pp pprint(op1 + op2.T)
# Notice that the results here are THE SAME!
pp pprint(op1 + op3)
pp pprint(op1 + op3.T)
```

array([[ 1, 3, 5],  
 [ 4, 6, 8],  
 [ 7, 9, 11]])

array([[ 1, 2, 3],  
 [ 5, 6, 7],  
 [ 9, 10, 11]])

array([[ 1, 3, 5],  
 [ 4, 6, 8],  
 [ 7, 9, 11]])

array([[ 1, 3, 5],  
 [ 4, 6, 8],  
 [ 7, 9, 11]])

# Broadcasting for pairwise distance

```
samples = np.random.random((15, 5))

# Without broadcasting
expanded1 = np.expand_dims(samples, axis=1)
tile1 = np.tile(expanded1, (1, samples.shape[0], 1))
expanded2 = np.expand_dims(samples, axis=0)
tile2 = np.tile(expanded2, (samples.shape[0], 1, 1))
diff = tile2 - tile1
distances = np.linalg.norm(diff, axis=-1)
```

```
# With broadcasting
diff = samples[:, np.newaxis, :]
          - samples[np.newaxis, :, :]
distances = np.linalg.norm(diff, axis=-1)
```

```
# With scipy (another math toolbox)
import scipy
distances = scipy.spatial.distance.cdist(samples, samples)
```

Both achieve the effect of

$$\begin{bmatrix} \vec{a} \\ \vec{b} \end{bmatrix} \Rightarrow \begin{bmatrix} \vec{a} & \vec{a} & \vec{a} \\ \vec{b} & \vec{b} & \vec{b} \end{bmatrix} - \begin{bmatrix} \vec{a} & \vec{b} & \vec{a} \\ \vec{b} & \vec{b} & \vec{b} \end{bmatrix}$$

# Why should I vectorize my code?

Shorter code, faster execution

```
a = np.random.random(500000)  
b = np.random.random(500000)
```

With loop

```
dot = 0.0  
for i in range(len(a)):  
    dot += a[i] * b[i]  
  
print(dot)
```

Wall time: 345ms

Numpy dot product

```
print(np.array(a).dot(np.array(b)))
```

Wall time: **2.9ms**

# An example with pairwise distance

Speed up depends on setup and nature of computation

```
samples = np.random.random((100, 5))  
  
With loop  
total_dist = []  
for s1 in samples:  
    for s2 in samples:  
        d = np.linalg.norm(s1 - s2)  
        total_dist.append(d)  
  
avg_dist = np.mean(total_dist)  
  
Numpy with broadcasting  
diff = samples[:, np.newaxis, :] -  
       samples[np.newaxis, :, :]  
distances = np.linalg.norm(diff, axis=-1)  
avg_dist = np.mean(distances)
```

Wall time: 162ms

(imagine without Numpy norm)

Wall time: **3.5ms**

# Plotting

# Other Python packages/tools

## Jupyter Notebook

- Interactive, re-execution, result storage



## Matplotlib

- Visualization (line, scatter, bar, images and even interactive 3D)

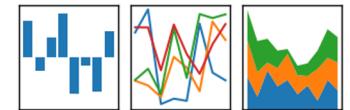


## Pandas (<https://pandas.pydata.org/>)

- Dataframe (database/Excel-like)
- Easy filtering, aggregation (also plotting, but few people uses Pandas for plotting)

## pandas

$$y_{it} = \beta' x_{it} + \mu_i + \epsilon_{it}$$



# Example plots

<https://matplotlib.org/3.1.1/gallery/index.html>

Import

```
import matplotlib
import matplotlib.pyplot as plt
import numpy as np
```

Create data

```
# Data for plotting
t = np.arange(0.0, 2.0, 0.01)
s = 1 + np.sin(2 * np.pi * t)
```

Plotting

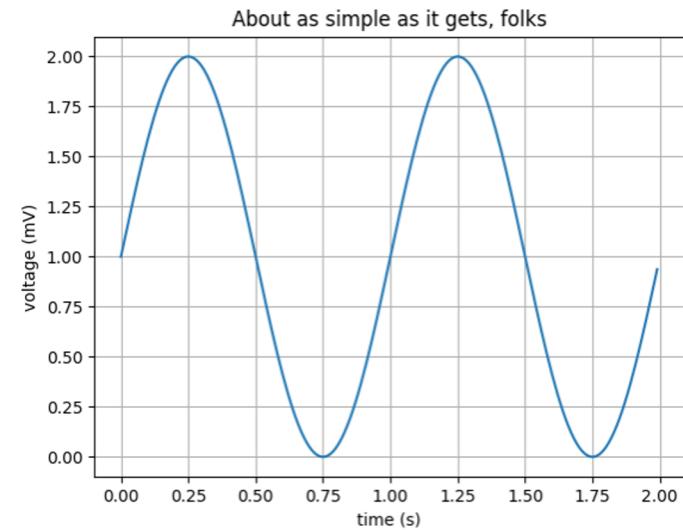
```
fig, ax = plt.subplots()
ax.plot(t, s)
```

Format plot

```
ax.set(xlabel='time (s)', ylabel='voltage (mV)',
       title='About as simple as it gets, folks')
ax.grid()
```

Save/show

```
fig.savefig("test.png")
plt.show()
```



# Plot with dash lines and legend

```
import numpy as np
import matplotlib.pyplot as plt

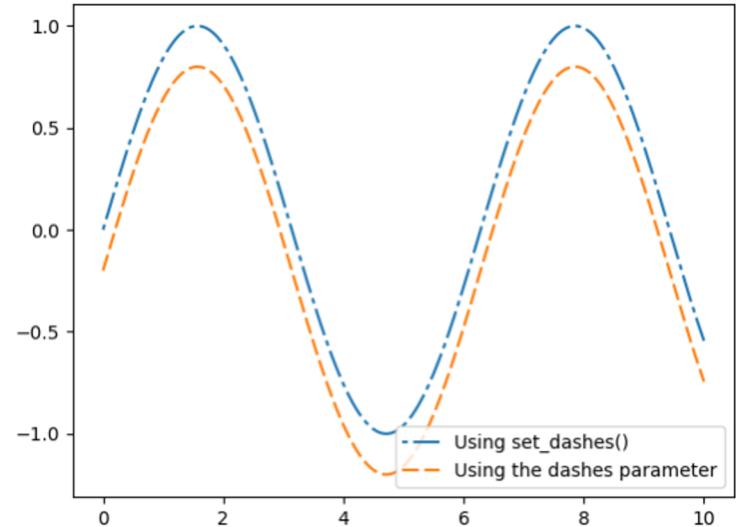
x = np.linspace(0, 10, 500)
y = np.sin(x)

fig, ax = plt.subplots()

line1, = ax.plot(x, y, label='Using set_dashes()')
# 2pt line, 2pt break, 10pt line, 2pt break
line1.set_dashes([2, 2, 10, 2])

line2, = ax.plot(x, y - 0.2, dashes=[6, 2],
                  label='Using the dashes parameter')

ax.legend()
plt.show()
```



# Using subplot

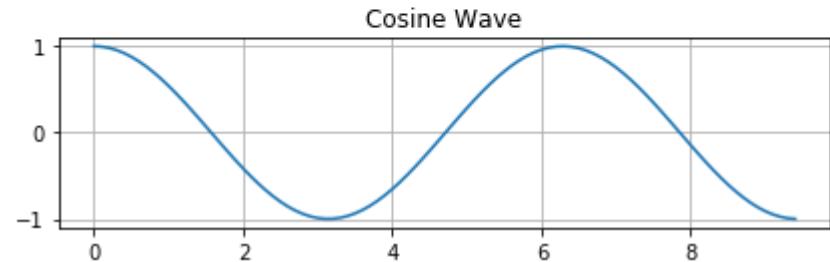
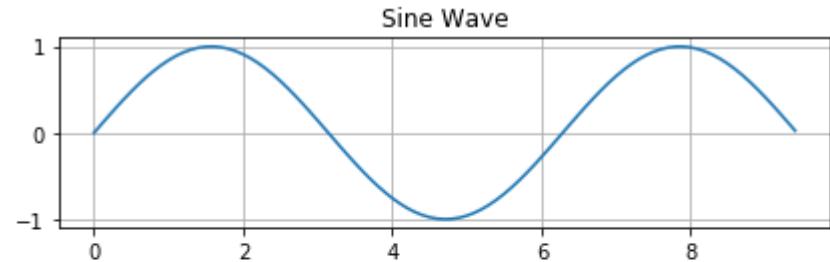
```
x = np.arange(0, 3 * np.pi, 0.1)
y_sin = np.sin(x)
y_cos = np.cos(x)

# Setup grid with height 2 and col 1.
# Plot the 1st subplot
plt.subplot(2, 1, 1)

plt.grid()
plt.plot(x, y_sin)
plt.title('Sine Wave')

# Now plot on the 2nd subplot
plt.subplot(2, 1, 2)
plt.plot(x, y_cos)
plt.title('Cosine Wave')

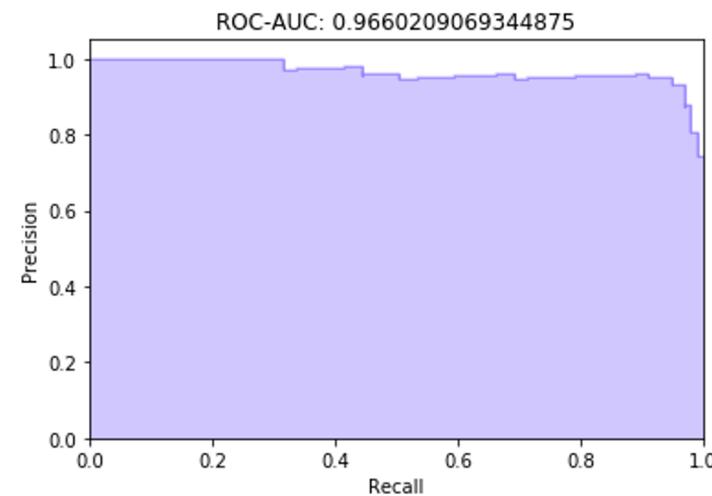
plt.grid()
plt.tight_layout()
```



# Plot area under curve

```
def prec_rec_curve(model, X, Y_true, title="", verbose=False):
    probas_pred = model.predict_proba(X)[:, 1]
    pos_label = 1.0
    precision, recall, thresholds = precision_recall_curve(Y_true,
                                                             probas_pred,
                                                             pos_label=pos_label)
    step_kw_args = ({'step': 'post'}
                   if 'step' in signature(plt.fill_between).parameters
                   else {})
    plt.step(recall, precision, color='b', alpha=0.2,
             where='post')
    plt.fill_between(recall, precision, alpha=0.2, color='b', **step_kw_args)

    plt.xlabel('Recall')
    plt.ylabel('Precision')
    plt.ylim([0.0, 1.05])
    plt.xlim([0.0, 1.0])
    plt.title(title+ "ROC-AUC: {}".format(auc(recall, precision)))
    plt.show()
```



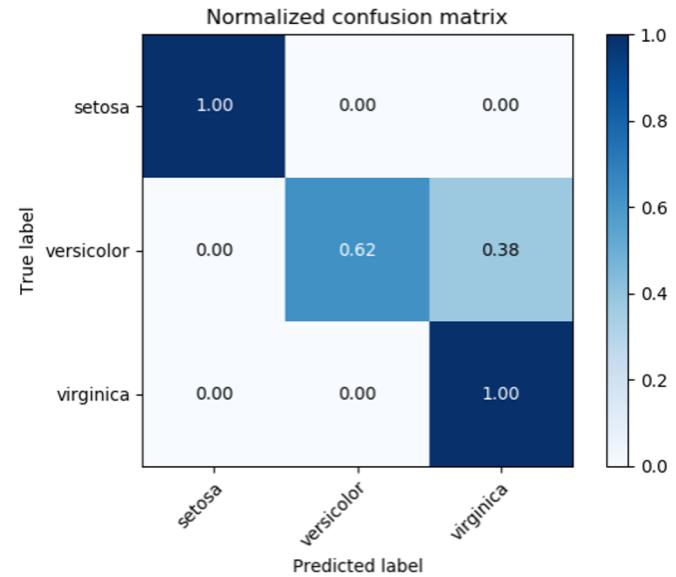
# Confusion matrix

[https://scikit-learn.org/stable/auto\\_examples/model\\_selection/plot\\_confusion\\_matrix.html](https://scikit-learn.org/stable/auto_examples/model_selection/plot_confusion_matrix.html)

```
fig, ax = plt.subplots()
im = ax.imshow(cm, interpolation='nearest', cmap=cmap)
ax.figure.colorbar(im, ax=ax)
# We want to show all ticks...
ax.set(xticks=np.arange(cm.shape[1]),
       yticks=np.arange(cm.shape[0]),
       xticklabels=classes, yticklabels=classes,
       ylabel='True label', xlabel='Predicted label',
       title=title)

# Rotate the tick labels and set their alignment.
plt.setp(ax.get_xticklabels(), rotation=45, ha='right',
         rotation_mode='anchor')

# Loop over data dimensions and create text annotations.
fmt = '.2f' if normalize else 'd'
thresh = cm.max() / 2.
for i in range(cm.shape[0]):
    for j in range(cm.shape[1]):
        ax.text(j, i, format(cm[i, j], fmt),
                ha='center', va='center',
                color="white" if cm[i, j] > thresh else "black")
fig.tight_layout()
```



Good luck on your  
HW/Project!

Questions?

# Links

[CS 231N Python Tutorial](#)

Additional slides in  
case of Q&A

# Where does my program start?

It just works

```
def do_something(number):
    for i in number:
        print(f'Hello {i}')

do_something(5)
```

A function

Properly

```
def do_something(number):
    for i in number:
        print(f'Hello {i}')

if __name__ == '__main__':
    do_something(5)
```

# What is a class?

Initialize the class to get an **instance** using some parameters

**Instance** variable

Does something with the **instance**

```
class Vehicle:  
    def __init__(self, make, name, year,  
                 is_electric=False, price=100):  
        self.name = name  
        self.make = make  
        self.year = year  
        self.is_electric = is_electric  
        self.price = price  
  
        self.odometer = 0  
  
    def drive(self, distance):  
        self.odometer += distance  
  
    def compute_price(self):  
        if self.is_electric:  
            price = self.price / (self.odometer * 0.8)  
        else:  
            price = self.price / self.odometer  
        return price
```

# To use a class

Instantiate a class,

get an **instance**

Call an instance method

```
if __name__ == '__main__':
    family_car = Vehicle('Honda', 'Accord', '2019',
                           price=10000)
    print(family_car.compute_price())
    family_car.drive(100)
    print(family_car.compute_price())
```

# String manipulation

## Formatting

```
stripped = '    I love CS229! '.strip()
```

```
upper_case = 'i love cs 229! '.upper()
```

```
capitalized = 'i love cs 229! '.capitalize()
```

## Concatenation

```
joined = 'string 1' + ' ' + 'string 2'
```

## Formatting

```
formatted = 'Formatted number {:.2F}'.format(1.2345)
```

# Basic data structures

## List

```
example_list = [1, 2, '3', 'four']
```

## Set (unordered, unique)

```
example_set = set([1, 2, '3', 'four'])
```

## Dictionary (mapping)

```
example_dictionary =  
    {  
        '1': 'one',  
        '2': 'two',  
        '3': 'three'  
    }
```

# More on List

2D list

```
list_of_list = [[1,2,3], [4,5,6], [7,8,9]]
```

List comprehension

```
initialize_a_list = [i for i in range(9)]
```

```
initialize_a_list = [i ** 2 for i in range(9)]
```

```
initialize_2d_list = [[i + j for i in range(5)] for j in range(9)]
```

Insert/Pop

```
my_list.insert(0, 'stuff')
```

```
print(my_list.pop(0))
```

# More on List

Sort a list

```
random_list = [3,12,5,6]
```

```
sorted_list = sorted(random_list)
```

```
random_list = [(3, 'A'),(12, 'D'),(5, 'M'),(6, 'B')]
```

```
sorted_list = sorted(random_list, key=lambda x: x[1])
```

# More on Dict/Set

## Comprehension

```
my_dict = {i: i ** 2 for i in range(10)}
```

```
my_set = {i ** 2 for i in range(10)}
```

## Get dictionary keys

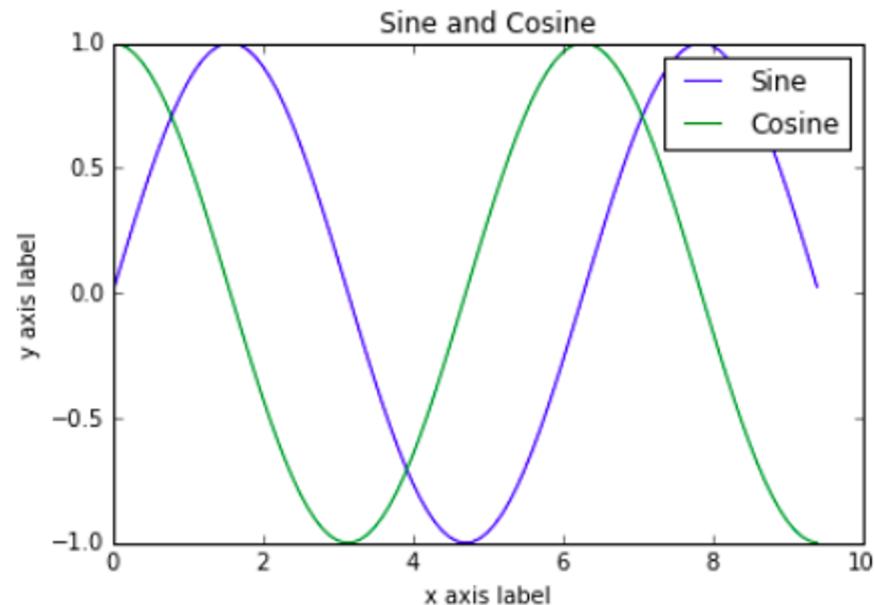
```
my_dict.keys()
```

# Another way for legend

```
import numpy as np
import matplotlib.pyplot as plt

# Compute the x and y coordinates for points on both curves
x = np.arange(0, 3 * np.pi, 0.1)
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'])
plt.show()
```



# Scatter plot

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

girls_grades = [89, 90, 70, 89, 100, 80, 90, 100, 80, 34]
boys_grades = [30, 29, 49, 48, 100, 48, 38, 45, 20, 30]
grades_range = [10, 20, 30, 40, 50, 60, 70, 80, 90, 100]
plt.scatter(grades_range, girls_grades, color='r')
plt.scatter(grades_range, boys_grades, color='g')
plt.xlabel('Grades Range')
plt.ylabel('Grades Scored')
plt.show()
```

