CS229 Section: Python & Numpy

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Content adapted from past CS229 iterations

How is Python related to with others?

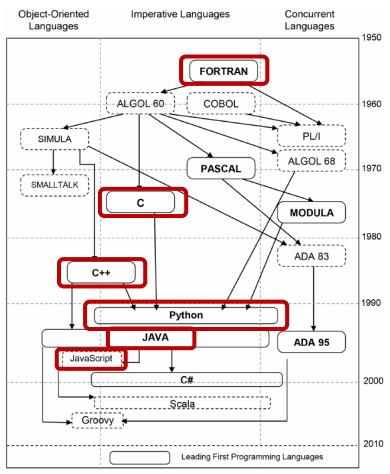
Python 2.0 released in 2000

(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



https://www.researchgate.net/figure/Genealogy-of-Programming-Languages-doi101371-journalpone0088941g001_fig1_260447599

Before you start

Use Anaconda

Create a new environment (full Conda)

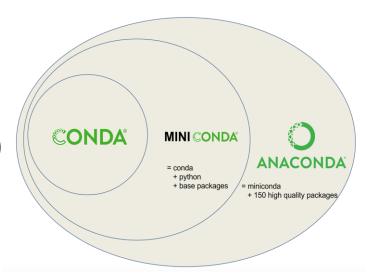
conda create -n cs229

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)

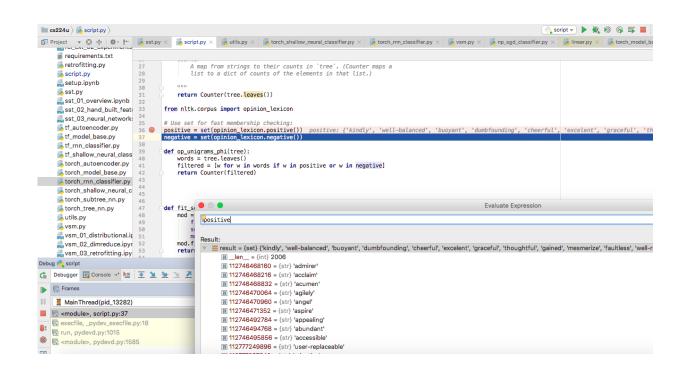




To make you more prepared

PyCharm

- Great debugger
- Proper project management



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

```
File Edit Selection View Go Run Terminal Help
                                                         class_example.py - python_tutorial - Visual Studio Code
                               python_tutorial > • class_example.py > * Vehicle > * compute_price
 V OPEN EDITORS
 X • class example.pv python tutorial
                                    1 class Vehicle:
                                             def __init__(self, make, name, year,
                                                             is_electric=False, price=100):
                                                  self.name = name
                                                  self.make = make
  ■ Numpy tutorial.ipynb
                                                 self.year = year
  F Quick Jupyter Demo.ipynb
                                                 self.is_electric = is_electric
  random_classifier.py
                                                 self.price = price

■ test_data.txt
                                                 self.odometer = 1
  F train_data.txt
                                            def drive(self, distance):
                                                  self.odometer += distance
  F cs229 python friday.pptx
                                            def compute price(self):
                                                 if self.is electric:
                                                      price = self.price / (self.odometer * 0.8)
                                                       price = self.price / self.odometer
                                                 return price
                                  24 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())
> OUTLINE
   020* → Python 3.6.5 64-bit ('base': conda) ⊗ 0 🛦 0
                                                                                                                Ln 18, Col 37 Spaces: 4 UTF-8 CRLF Python R C
```

Basic Python

Strings, Lists, Dictionaries

String manipulation

```
Formatting
```

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 1 = ['one', 'two', 'three']
List creation
                             list 1.append(4)
                             list 1.insert(0, 'ZERO')
                             list_2 = [1, 2, 3]
Insertion/extension
                             list 1.extend(list 2)
                             long_list = [i for i in range(9)]
                             long_long_list = [(i, j) for i in range(3)
List comprehension
                                                                for j in range(5)]
                             long list_list = [[i for i in range(3)]
                                                                for in range(5)]
                             sorted(random list)
Sorting
                             random_list_2 = [(3, 'z'), (12, 'r'), (6, 'e'),
                                                       (8, 'c'), (2, 'g')]
                             sorted(random_list_2, key=lambda x: x[1])
```

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 (https://docs.python.org/3/library/pdb.html)
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

Matrix multiplication like Ax

2D matrix multiplication

Element-wise multiplication

```
array 1 @ array 2
array_1.dot(array_2)
np.dot(array 1, array 2)
weight_matrix = np.array([1, 2, 3, 4]).reshape(2, 2)
sample = np.array([[50, 60]]).T
np.matmul(weight_matrix, sample)
mat1 = np.array([[1, 2], [3, 4]])
mat2 = np.array([[5, 6], [7, 8]])
np.matmul(mat1, mat2)
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])
                                                       array([[1, 3, 5],
                                                                [4, 6, 8],
                                                                 [ 7, 9, 11]])
# Notice that the results here are DIFFERENT!
pp.pprint(op1 + op2)
                                                       array([[1, 2, 3],
pp.pprint(op1 + op2.T)
                                                                [5, 6, 7],
                                                                 [ 9, 10, 11]])
                                                       array([[1, 3, 5],
# Notice that the results here are THE SAME!
                                                                [4, 6, 8],
                                                                 [ 7, 9, 11]])
pp.pprint(op1 + op3)
pp.pprint(op1 + op3.T)
                                                       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)
```

```
Both achieve the effect of \begin{bmatrix} \overrightarrow{\alpha} \\ \overrightarrow{b} \end{bmatrix} \Rightarrow \begin{bmatrix} \overrightarrow{\alpha} \\ \overrightarrow{b} \end{bmatrix} - \begin{bmatrix} \overrightarrow{\alpha} \\ \overrightarrow{b} \\ \overrightarrow{b} \end{bmatrix}
```

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

Why should I vectorize my code?

Shorter code, faster execution

Wall time: 345ms

Wall time: 2.9ms

An example with pairwise distance

Speed up depends on setup and nature of computation

(imagine without Numpy norm)

```
samples = np.random.random((100, 5))
                With loop
                                                  Numpy with broadcasting
total dist = []
                                         diff = samples[: ,np.newaxis, :]
for s1 in samples:
                                                             samples[np.newaxis, :, :]
                                         distances = np.linalg.norm(diff, axis=-1)
    for s2 in samples:
        d = np.linalg.norm(s1 - s2)
                                         avg dist = np.mean(distances)
        total dist.append(d)
avg_dist = np.mean(total_dist)
   Wall time: 162ms
                                                       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/)

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







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

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

Save/show

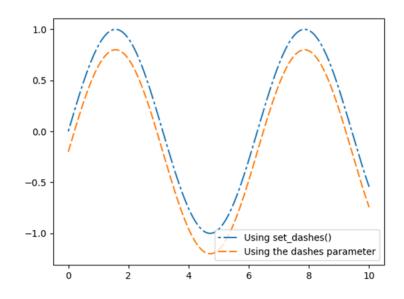
```
fig.savefig("test.png")
plt.show()
```

About as simple as it gets, folks 2.00 1.75 1.50 oltage 1.00 voltage 1.00 0.75 0.50 0.25 0.00 0.00 0.25 0.50 0.75 1.00 1.25 1.50 1.75 2.00

time (s)

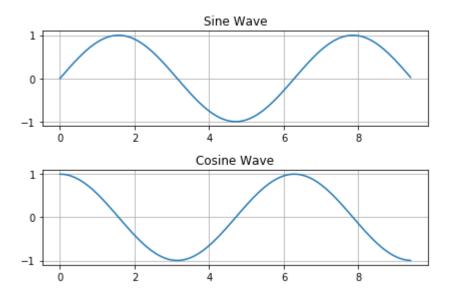
Plot with dash lines and legend

```
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()
```



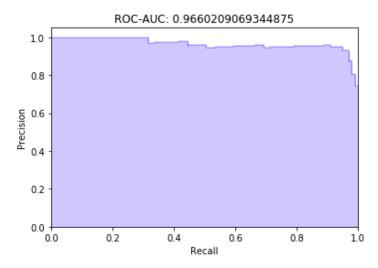
Using subplot

```
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

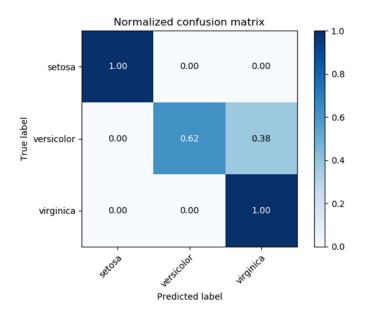
```
def prec rec curve(model, X, Y true, title="", verbose=False):
    probas pred = model.predict_proba(X)[:, 1]
    pos label = 1.0
    precision, recall, thresholds = precision recall curve(Y true,
                                                           probas pred,
                                                           pos label=pos label)
    step kwargs = ({'step': 'post'}
               if 'step' in signature(plt.fill between).parameters
    plt.step(recall, precision, color='b', alpha=0.2,
         where='post')
    plt.fill between(recall, precision, alpha=0.2, color='b', **step kwargs)
    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

```
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,
        vlabel='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?

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)
```

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

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

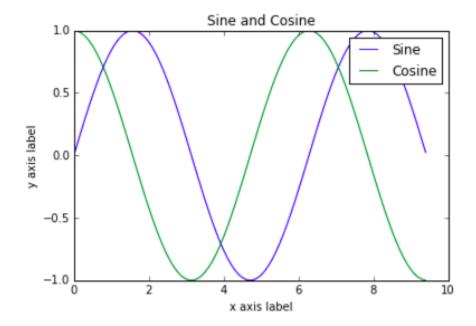
```
my_dict = {i: i ** 2 for i in range(10)}
my_set = {i ** 2 for i in range(10)}
```

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 po
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()
```

