Programming Principles in Python (CSCI 503/490)

Arrays

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Modules and Packages

- Python allows you to import code from other files, even your own
- A **module** is a collection of definitions
- A **package** is an organized collection of modules
- Modules can be
 - a separate python file
 - a separate C library that is written to be used with Python
 - a built-in module contained in the interpreter
 - a module installed by the user (via conda or pip)
- All types use the same import syntax









What is the purpose of having modules or packages?

- Code reuse: makes life easier because others have written solutions to various problems
- Generally forces an organization of code that works together • Standardizes interfaces; easier maintenance
- Encourages robustness, testing code
- This does take time so don't always create a module or package - If you're going to use a method once, it's not worth putting it in a module - If you're using the same methods over and over in (especially in different projects), a module or package makes sense









Importing modules

- import <module>
- import <module> as <another-identifier>
- from <module> import <identifer-list>
- from <module> import <identifer> as <another-identifier>, ...
- import imports from the top, from ... import imports "inner" names
- as clause renames the imported name

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• Need to use the qualified names when using import (foo.bar.mymethod)





Namespaces

- Namespace is basically a dictionary with names and their values
- Accessing namespaces

builtins , globals(), locals()

- Examine contents of a namespace: dir(<namespace>)
- Python checks for a name in the sequence: local, enclosing, global, builtins
- Each import <module> creates a namespace; access it through dot-syntax:
 - Examples: math.pi, collections.Counter











Using an imported module

- - import math math.log10(100) math.sqrt(196)
- Import module into current namespace and use unqualified name
 - from math import log10, sqrt log10(100)sqrt (196)

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• Import module, and call functions with **fully qualified** name (its namespace)









Reloading a Module?

- If you re-import a module, what happens?
 - import my module my module.SECRET NUMBER # 42
 - Change the definition of SECRET NUMBER to 14
 - import my module my module.SECRET NUMBER # Still 42!
- Modules are cached so they are not reloaded on each import call
- Can reload a module via importlib.reload (<module>)
- Be careful because **dependencies** will persist! (Order matters)





Python Packages

- A package is basically a collection of modules in a directory subtree
- Structures a module namespace by allowing dotted names
- Example:

 For packages that are to be execut added

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of modules in a directory subtree / allowing dotted names

For packages that are to be executed as scripts, __main__.py can also be





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Example







Finding Packages

- Python Package Index (PyPI) is the standard repository (<u>https://pypi.org</u>) and pip (pip installs packages) is the official python package installer
 - Types of distribution: source (sdist) and wheels (binaries)
 - Each package can specify dependencies
 - Creating a PyPI package requires adding some metadata
- <u>Anaconda</u> is a package index, conda is a package manager
 - conda is language-agnostic (not only Python)
 - solves dependencies
 - conda deals with non-Python dependencies
 - has different channels: default, conda-forge (community-led)







Installing Packages

- pip install <package-name>
- conda install <package-name>
- In Jupyter use:
 - %pip install <package-name>
 - %conda install <package-name>
- Arguments can be multiple packages
- (e.g. <u>Alex Birsan</u>)

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• Be careful! Security exploits using package installation and dependencies





Environments

- Both pip and conda support environments
 - venv
 - conda env
- Idea is that you can create different environments for different work
 - environment for cs503
 - environment for research
 - environment for each project





<u>Assignment 4</u>

- Due Today
- UDSA Food Price Data
- Reading & Writing Files
- Iterators
- Numeric Aggregation
- String Formatting
- CSCI 503 students compute and output two additional fields





Assignment 5

- Scripts, modules, packages
- Soon







What is the difference between an array and a list (or a tuple)?





Arrays

- Usually a fixed size—lists are meant to change size
- Are mutable—tuples are not
- Store only one type of data—lists and tuples can store any combination • Are faster to access and manipulate than lists or tuples
- Can be multidimensional:

 - Can have list of lists or tuple of tuples but no guarantee on shape - Multidimensional arrays are rectangles, cubes, etc.





Why NumPy?

- Fast vectorized array operations for data munging and cleaning, subsetting and filtering, transformation, and any other kinds of computations
- Common array algorithms like sorting, unique, and set operations
- Efficient descriptive statistics and aggregating/summarizing data
- Data alignment and relational data manipulations for merging and joining together heterogeneous data sets
- Expressing conditional logic as array expressions instead of loops with ifelif-else branches
- Group-wise data manipulations (aggregation, transformation, function) application).

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import numpy as np





Creating arrays

- data1 = [6, 7, 8, 0, 1]arr1 = np.array(data1)
- data2 = [[1.5,2,3,4], [5,6,7,8]]arr2 = np.array(data2)
- data3 = np.array([6, "abc", 3.57]) # !!! check !!!
- Can check the type of an array in dtype property
- Types:
 - arr1.dtype # dtype('int64')
 - arr3.dtype # dtype('<U21'), unicode plus # chars





lypes

- "But I thought Python wasn't stingy about types..."
- numpy aims for speed
- Able to do array arithmetic
- int16, int32, int64, float32, float64, bool, object
- Can specify type explicitly
 - arr1 float = np.array(data1, dtype='float64')
- astype method allows you to convert between different types of arrays:

arr = np.array([1, 2, 3, 4, 5])arr.dtype float arr = arr.astype(np.float64)







numpy data types (dtypes)

Туре	Type code	Descriptio
int8, uint8	i1, u1	Signed an
int16, uint16	i2, u2	Signed an
int32, uint32	i4, u4	Signed an
int64, uint64	i8, u8	Signed an
float16	f2	Half-precis
float32	f4 or f	Standard s
float64	f8 or d	Standard o
		Python f1
float128	f16 or g	Extended-
complex64,	c8, c16,	Complex r
complex128,	c32	
complex256		
bool	?	Boolean ty
object	0	Python ob
string_	S	Fixed-leng
		string dty
unicode_	U	Fixed-leng specificati

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- d unsigned 8-bit (1 byte) integer types
- id unsigned 16-bit integer types
- nd unsigned 32-bit integer types
- id unsigned 64-bit integer types
- ision floating point
- single-precision floating point; compatible with C float
- double-precision floating point; compatible with C double and
- loat object
- -precision floating point
- numbers represented by two 32, 64, or 128 floats, respectively
- ype storing True and False values
- pject type; a value can be any Python object
- gth ASCII string type (1 byte per character); for example, to create a pe with length 10, use 'S10'
- gth Unicode type (number of bytes platform specific); same
- ion semantics as string_(e.g., 'U10')

[W. McKinney, Python for Data Analysis]









Array Shape

- Our normal way of checking the size of a collection is... len
- How does this work for arrays?
- arr1 = np.array([1,2,3,6,9]) len(arr1) # 5
- arr2 = np.array([[1.5,2,3,4],[5,6,7,8]])len(arr2) # 2
- All dimension lengths \rightarrow shape: arr2.shape # (2,4)
- Number of dimensions: arr2.ndim # 2
- Can also reshape an array:
 - arr2.reshape(4,2)
 - arr2.reshape(-1,2) # what happens here?









Speed Benefits

- Compare random number generation in pure Python versus numpy
- Python:
 - import random %timeit rolls list = [random.randrange(1,7)
- With NumPy:
 - %timeit rolls array = np.random.randint(1, 7, 60 000)
- Significant speedup (80x+)

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for i in range(0, 60 000)]









Array Programming

- Lists:
 - C = [] for i in range(len(a)): c.append(a[i] + b[i])
- How to improve this?







Array Programming

- Lists:
 - C = | | for i in range(len(a)): c.append(a[i] + b[i])
 - -c = [aa + bb for aa, bb in zip(a,b)]
- NumPy arrays:
 - -c = a + b
- More functional-style than imperative
- Internal iteration instead of external







Operations

- a = np.array([1, 2, 3])b = np.array([6, 4, 3])
- (Array, Array) Operations (**Element-wise**)
 - Addition, Subtraction, Multiplication
 - -a + b # array([7, 6, 6])
- (Scalar, Array) Operations (**Broadcasting**):
 - Addition, Subtraction, Multiplication, Division, Exponentiation
 - a ** 2 # array([1, 4, 9])
 - -b + 3 # array([9, 7, 6])









More on Array Creation

- Zeros: np.zeros(10)
- Ones: np.ones((4,5)) # shape
- Empty: np.empty((2,2))
- _like versions: pass an existing array and matches shape with specified contents
- Range: np.arange(15) # constructs an array, not iterator!









Indexing

- Same as with lists plus shorthand for 2D+
 - $\operatorname{arr1} = \operatorname{np.array}([6, 7, 8, 0, 1])$
 - arr1[1]
 - arr1[-1]
- What about two dimensions?
 - $\operatorname{arr2} = \operatorname{np.array}([[1.5, 2, 3, 4], [5, 6, 7, 8]])$
 - arr[1][1]
 - arr[1,1] # shorthand









2D Indexing



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axis 1		
0	1	2
, 0	0, 1	0, 2
, 0	1, 1	1, 2
, 0	2, 1	2, 2

[W. McKinney, Python for Data Analysis]











Slicing

- 1D: Similar to lists
 - arr1 = np.array([6, 7, 8, 0, 1])
 - arr1[2:5] # np.array([8,0,1]), sort of
- Can **mutate** original array:
 - arr1[2:5] = 3 # supports assignment
 - arr1 # the original array changed
- Slicing returns views (copy the array if original array shouldn't change)
 - arr1[2:5] # a view
 - arr1[2:5].copy() # a new array







Slicing

- 2D+: comma separated indices as shorthand:
 - $\operatorname{arr2} = \operatorname{np.array}([[1.5, 2, 3, 4], [5, 6, 7, 8]])$
 - a[1:3,1:3]
 - a[1:3,:] # works like in single-dimensional lists
- Can combine index and slice in different dimensions
 - a[1,:] # gives a row
 - a[:,1] # gives a column







2D Array Slicing

How to obtain the blue slice from array arr?

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2D Array Slicing

How to obtain the blue slice from array arr?

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arr[:2,1:]

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