

INTRODUCTION TO PYTHON

ECRF Tutorial

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7th July 2015

LEARNING OUTCOMES

- Understand why Python is becoming a popular language for Scientific Research
- Know how to run basic Python scripts
- Understand basic Python syntax, data structures and control flow
- Be aware of how Python can be used in a similar fashion to Matlab

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Setting up Python

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OVERVIEW

WHY LEARN PYTHON?

- Commonly used in research software

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- Large community of users

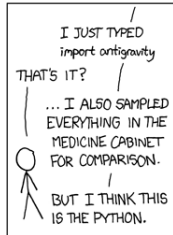
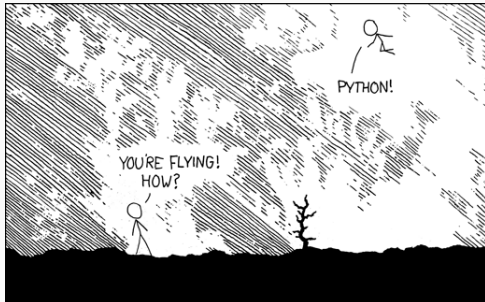


Figure: Taken from <https://xkcd.com/353/>

POPULARITY

Jun 2015	Jun 2014	Change	Programming Language	Ratings	Change
1	2	▲	Java	17.822%	+1.71%
2	1	▼	C	16.788%	+0.60%
3	4	▲	C++	7.756%	+1.33%
4	5	▲	C#	5.056%	+1.11%
5	3	▼	Objective-C	4.339%	-6.60%
6	8	▲	Python	3.999%	+1.29%
7	10	▲	Visual Basic .NET	3.168%	+1.25%
8	7	▼	PHP	2.868%	+0.02%
9	9		JavaScript	2.295%	+0.30%
10	17	▲▲	Delphi/Object Pascal	1.869%	+1.04%

Figure: Taken from
<http://www.tiobe.com/index.php/content/paperinfo/tpci/index.html>

- Advantages:
 - **Dynamic typing** = No types cluttering up screen (`int`, `String` etc...)
 - No semi-colons
 - **Significant whitespace** forces you to write clean code
- Disadvantages:
 - Dynamically typed means you lack compile time checking
 - Also means it can't be compiled ahead of time so it's run in an interpreter = slow

APPLICATION AREAS

- Research software
- Web development
- Rapid prototyping
- Creating GUI front ends
- Glue code
- Utility scripts
- Quick scripts

SETTING UP PYTHON

- Two versions of Python in current use, **Python 2** and **Python 3**
- Python 3 first came out 7 years ago and last major version of Python 2 (2.7) came out 5 years ago
- Yet there's still a considerable userbase still on 2.7
- The **only** reason some people are still on Python 2 is due to package compatibility
- **Use Python 3** (current version is 3.4)

- Installed by default on most Linux operating systems (including Mac), or will be in official repositories
- Will probably have both `Python2` and `Python3`, check which version you have by running `python -V`
- On Windows download from <https://www.python.org/downloads/>

- There is a package manager called **Pip** that installs new packages for you (ships with 3.X, not 2.X)
- `pip install <package>`
- **don't run pip with sudo**
- If using an IDE can install through that as well

- Comes with a **Read-Evaluate-Print-Loop (REPL)** (separate executable on Windows)
- Useful for checking algorithms, or small functions
- Use as a calculator

- Python files (known as **modules**) are saved with **.py** file extension
- No limit to what's saved in a module, can have as many classes and functions as you want
- To run manually just execute `python file.py` from command line
- **NB:** In Windows you'll need to set Path before doing this

- For quick scripts I personally use a text editor and run it manually
- Useful IDEs include:
 - **PyCharm** - My favourite, has full student licence
 - **PyDev** - Eclipse plugin
 - **Spyder** - Matlab style IDE, focuses on scientific computing
- **IDLE** - Comes with Python on Windows. **Rubbish.**

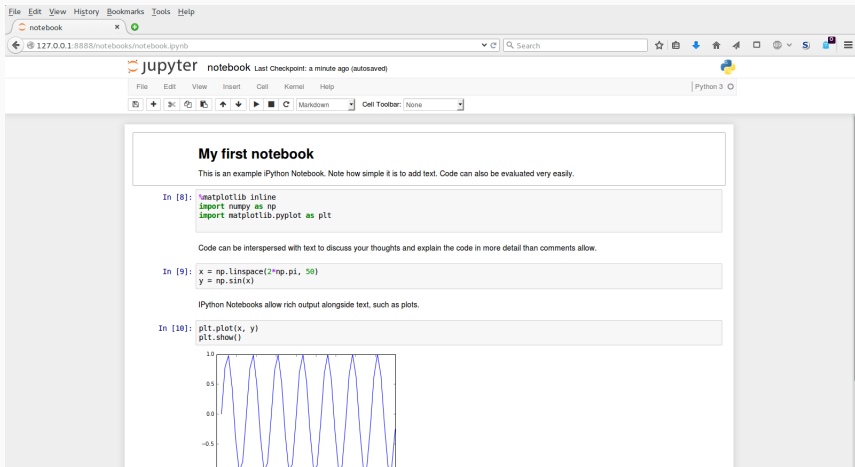
```

File Edit View Navigate Code Refactor Run Tools VCS Window Help
Nostradamus nostradamus ea fitness.py
ea.py population.py fitness.py util.py selector.py breeder.py classifier_breeders.py diversity.py runtimes.cfg main.py
Structure
Debug

497     patterns
498     classes (list): The class labels as integers of the dataset.
499
500     Returns:
501     The AUC of the classifier to the dataset as a floating value.
502     """
503     # Convert to probabilities
504     probabilities = nap(util.scores_to_probabilities, outputs)
505
506     # If have just 2 classes then can save time by not calculating MAUC
507     # and just calculate the single A-value instead.
508     if self.num_classes == 2:
509         raw_fitness = util.a_value(itertools.zip(classes, probabilities))
510         if raw_fitness < 0.5: # Can invert if below diagonal for 2 class
511             raw_fitness = 1 - raw_fitness
512     else:
513         raw_fitness = util.MAUC(zip(classes, probabilities), num_classes=self.num_classes)
514
515     return raw_fitness
516
517 def pairwise_AUC(self, outputs, classes):
518     """
519     Calculates the pairwise AUCs for the input output scores and class labels.
520
521     Returns then in a list of tuples with (class0, class1, AUC)
522     """
523     probabilities = nap(util.scores_to_probabilities, outputs)
524     class_pairs = [x for x in itertools.combinations(xrange(self.num_classes), 2)]
525
526     # Have to take average of A value with both classes acting as label 0 as this
527     # gives different outputs for more than 2 classes
528     pairwise_avals = []
529     for pairing in class_pairs:
530         a_val = (util.a_value(itertools.zip(classes, probabilities), zero_label=pairing[0], one_label=pairing[1]) +
531                 util.a_value(itertools.zip(classes, probabilities), zero_label=pairing[1], one_label=pairing[0])) / 2.0
532         pairwise_avals.append((pairing[0], pairing[1], a_val))
533
534     return pairwise_avals
535
536 def predict_label(self, output, dataset=None):
537     """
538     Predicts a label from the given output to an unknown

```


- Multimedia environments with code, output, text and plots all alongside one another
- Very useful for scientific work
- Also very handy for sharing work with colleagues
- Can expand on ideas and explain implementation details in more depth than comments



The screenshot shows a web browser window displaying a Jupyter Notebook. The browser's address bar shows the URL `127.0.0.1:8888/notebooks/notebook.py?nb`. The Jupyter interface includes a top menu bar with options like File, Edit, View, History, Bookmarks, Tools, and Help. Below this is a toolbar with icons for file operations and a 'Python 3' kernel indicator. The notebook content is displayed in a white box with a title 'My first notebook' and a subtitle 'This is an example iPython Notebook. Note how simple it is to add text. Code can also be evaluated very easily.'

The notebook contains three code cells:

- In [8]:** `!matplotlib inline`
`import numpy as np`
`import matplotlib.pyplot as plt`
- In [9]:** `x = np.linspace(2*np.pi, 50)`
`y = np.sin(x)`
- In [10]:** `plt.plot(x, y)`
`plt.show()`

Below the third code cell, a plot is displayed showing a blue sine wave oscillating between -0.5 and 1.0 on the y-axis. The x-axis represents values from 2π to 50.

HELLO WORLD

- You can see that unlike C or Java there is no `main` function
- Instead it runs the script entirely
- Useful but also means that when we import the script run into problems (discuss later)

DATA STRUCTURES AND FUNCTIONS

PRIMITIVES

- Basic ints, floats, strings, booleans
- Variables are all **references**
- Variables can be used to reference different types

```
a = 5          # int
a = "foo"     # Previously referenced an int

b = 5.0       # float
c = "hello"   # string
d = True      # boolean
```

- Control flow in Python is denoted by **whitespace**.
- Python replaces {}s from Java with indentation (4 spaces)
- No parentheses around condition
- No **end** statement, just revert back to higher indentation level

WHILE LOOP

```
i = 0
while i < 10:
    print(i)
    i += 1
print("The last value is",i)
```

0
1
2
3
4
5
6
7
8
9

The last value is 10

```
for i in range(10):  
    print(i)
```

0
1
2
3
4
5
6
7
8
9

CONDITIONALS

- `If` and `Else` statements are implemented in a similar fashion
- `Else-if` statements take the syntax `elif`
- Compound statements are formed with `and`, `or`

```
# Obtain a random number somehow
```

```
my_number = random_number()
```

```
if my_number > 1000:
```

```
    print("Number is greater than 1000!")
```

```
elif my_number < 10 and my_number % 2 == 0:
```

```
    print("Number is less than 10 and even")
```

```
else:
```

```
    print("Not interested in this number")
```

- One of the main advantages of Python is the extremely powerful data structures it has
- These come in the default namespace and don't need to be imported
- Three main ones:
 - Lists
 - Tuples
 - Dictionaries

- The most common data structure
- **Heterogeneous dynamic array**
- Heterogeneous = multiple different types
- Dynamic = non-fixed size
- Create lists with []s
- Use []s for indexing as well

```
foo = [1, "hello", 3.4] # Literal instantiation
foo.append(5) # Can add items
foo.remove("hello") # ... And remove them
foo[0] # Evaluates to 1
foo[-1] # Returns the last value in the list, 5
```

```
# Can check for membership with 'in'.
contains_1 = 1 in foo # Evaluates to True
```

```
[1, 3.4, 5]
```

- Any object that overrides the `iterable` property allows for direct iteration over (think for each loops in Java)
- This includes all Python data structures
- Remember the for loop example from before? `range(x)` is a function which returns a list

ITERABLE DATA STRUCTURE

- Any object that overrides the `iterable` property allows for direct iteration over (think for each loops in Java)
- This includes all Python data structures
- Remember the for loop example from before? `range(x)` is a function which returns a list

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

```
# Therefore could have rewritten for loop  
for i in foo:  
    print(i)
```

Docs show full range of functions which work on lists
<https://docs.python.org/3/tutorial/datastructures.html>

- A common operation is to iterate through adding items to a list
- This is so common that Python designers decided to add syntactical sugar to reduce typing - **list comprehension**
- Can be used any time have iteration

LIST COMPREHENSIONS

```
# Creates a list of squared integers up to 10
foo = []
for i in range(10):
    foo.append(i**2)
```

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LIST COMPREHENSIONS

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```
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```

```
# Exact same output as above
```

```
foo = [i**2 for i in range(10)]
```

```
# Can be used for any iterable object
```

```
# Creates a list of lines in a text file
```

```
# Python will automatically detect line breaks
```

```
myfile = open('testfile.txt', 'r')
```

```
lines = [line for line in myfile]
```

- Heterogeneous fixed-size arrays
- Uses less memory and bit quicker to access than lists
- Useful for data structures such as Java classes which just contain fields
- Denoted by parentheses

```
foo = (1, "hello", 3.4)

# Can't add or remove items

# Iterable in the same fashion as lists
for item in foo:
    print(item)
```

- **Hash map** / associative array
- Indexes items in a container by a hash (typically a **string**)
- Create with {}s in the form
`{<key1>: <value1>, <key2>: <value2>... }`
- Iterating returns the **keys**

```
# Dictionary instantiation
foo = {'cat': 'miaow', 'dog': 'woof'}

# Index with squared brackets
cat_noise = foo['cat'] # 'miaow'
dog_noise = foo['dog'] # 'woof'

# Add new values in the same way
foo['pig'] = 'oink'

# Prints dog, cat, pig
for item in foo:
    print(item)
```


COUNTING EXAMPLE

A useful application is counting items

```
text = ("There", "are", "only", "ten", "types", "of" "people",
        "in", "the", "world", ",", "those", "that", "understand",
        "binary" "and" "those" "that" "don't")
counts = {}
for word in text:
    if word not in counts:
        counts[word] = 1
    else:
        counts[word] += 1

{'ten': 1, 'types': 1, 'people': 1, 'in': 1, 'the': 1, 'those': 2,
 'binary': 1, 'world': 1, ',', 'only': 1, 'are': 1, "don't": 1,
 'and': 1, 'understand': 1, 'of': 1, 'that': 2, 'There': 1}
```

- Differences with Java functions:
 - No notion of public/private
 - Don't state argument type
 - Don't state return type
 - Function body defined by **indentation** rather than {}s
 - **def** is the keyword which creates a function

```
def function_name(arg1, arg2):  
    code here  
    ...  
    more code  
    return value
```

```
new_value = function_name(val1, val2)
```

EXAMPLE FUNCTION

- Will calculate `mean` of a list
- Will use the inbuilt functions `sum` and `len`

- Specify arguments by **name** rather than position
- Allows for optional arguments to set default values
- Makes code **self-documenting**

How readable is this?

```
my_plot(x, y, True, False, True, 'Time',  
        'Amplitude')
```

KEYWORD ARGUMENTS

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- Allows for optional arguments to set default values
- Makes code **self-documenting**

How readable is this?

```
my_plot(x, y, True, False, True, 'Time',  
        'Amplitude')
```

Now you can see what each argument does

```
my_plot(x, y, show_gridlines=True, save=False,  
        x_label='Time', y_label='Amplitude')
```

KEYWORD ARGUMENTS

```
# Here's how you define a function with keyword args
def run_ea(population, data, generations=1000):
    for i in range(generations):
        outputs = run_population(population)
        fitnesses = fitness_function(outputs)
        population = breed(population, fitnesses)
    return population.winner
```

```
# Can omit keyword args in the call to use
# the default value
```

```
run_ea(my_pop, some_data)
```

```
# Or pass in a value
```

```
run_ea(my_pop, some_data, generations=5000)
```

IMPORTING MODULES

- Can reuse code we've already in other modules with `import` statements
- Same syntax for using library code
- Have to be careful with top level functions!

```
import foo
```

```
foo.some_function()
```

OR can use aliases for long module names

```
import somereallylongmodulename as m
```

```
m.some_function()
```

- Importing a Python module runs all code at the highest indent level
- Can lead to unexpected behaviour
- Want to be able to control the entry point into the program using a main function like other languages
- In Python this is achieved with the following

```
if __name__ == "__main__":  
    # Code here will only be run when this module  
    # is executed directly with  
    # 'python modulename.py',  
    # and *not* when imported  
    do_stuff()
```


REPLACING MATLAB WITH PYTHON

DIFFERENCES FROM MATLAB

- Matlab is designed for **Scientific Programming**, Python is a **general purpose** programming language
- Matlab comes with an IDE, Python is just the language
- Python is fully cross-platform compatible
- Python has more **sane syntax** choices, such as indexing with square brackets and 0-based indices
- **String manipulation** is far easier in Python than in Matlab
- Python has a much better **modular** system

- The Python language is great for general programming uses but by itself it lacks little support for scientific programming
- It needs:
 - Quick data structures (**static arrays**)
 - **Vectorized** operations
 - Library of Scientific functions

- The Python language is great for general programming uses but by itself it lacks little support for scientific programming
- It needs:
 - Quick data structures (**static arrays**)
 - **Vectorized** operations
 - Library of Scientific functions
- The 3rd party libraries in the **SciPy** stack solve this!
- Three packages which are often included together
 - NumPy
 - SciPy
 - Matplotlib

- Lists are dynamic sized and heterogeneous (memory and slow to access)
- Tuples are heterogeneous (memory)
- NumPy provides **fixed size homogeneous** C-style arrays
- And **vectorized** operations like Matlab (can vectorise your own code)

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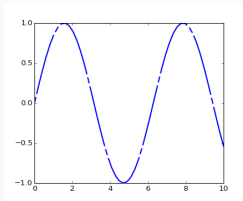
```
import numpy as np
```

```
# Creates an array of 60 values between 0 and 2*PI  
x = np.linspace(0, 2*np.pi, 60)
```

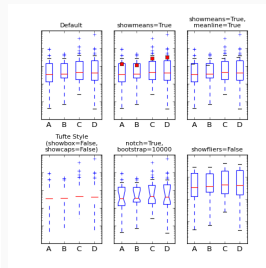
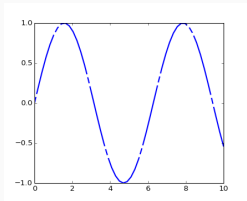
```
# Calculates the corresponding sine values  
y = np.sin(x)
```

- Built on the top of NumPy
- Complex numbers
- Signal processing
- Polynomials
- Integration
- Statistics
- Most functionality from all **Matlab toolboxes** (even paid ones) except Simulink
- Look on website to see the full range of functionality

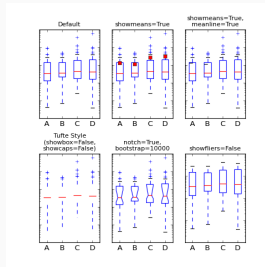
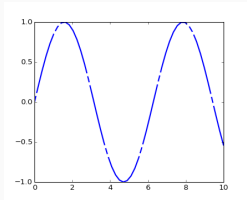
- Easy to use plotting interface
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- Machine learning in Python
- Classification and regression
- Covers feature selection and resampling as well
- Model examples:
 - SVMs
 - Decision trees
 - Neural networks
 - Ensembles

- Data analysis package - very useful
- Introduces the **data frame** data structure from **R**
- Access columns by **name**, summary measures easily available
- Plays well with the rest of the SciPy stack

	mutation_rate	run	accuracy
0	0.04	1	0.93
1	0.04	2	0.92
2	0.04	3	0.91
3	0.08	1	0.94
4	0.08	2	0.93
5	0.08	3	0.92
6	0.12	1	0.90
7	0.12	2	0.89
8	0.12	3	0.88

```
import pandas as pd
import numpy as np
```

```
# Can access columns by name
```

```
data = pd.read_csv('dummy_data.csv')
grouped = data.groupby('mutation_rate')
summaries = grouped['accuracy'].agg([np.mean, np.std])
```

mutation_rate	mean	std
0.04	0.92	0.01
0.08	0.93	0.01
0.12	0.89	0.01

- If Numpy wasn't sufficient or applicable there are a few other options
- **Cython** (www.cython.org)
 - Compiles to C
 - Can run straight away (after compiling) or can get far better speed ups by adding types

SPEEDING UP CODE

- If Numpy wasn't sufficient or applicable there are a few other options
- **Cython** (www.cython.org)
 - Compiles to C
 - Can run straight away (after compiling) or can get far better speed ups by adding types
- **PyPy** (www.pypy.org)
 - JIT compiler for Python
 - My tool of choice for speeding up code
 - Don't need to do anything special, just install it and run `pypy foo.py` rather than `python foo.py`
 - It can run every standard library but not necessarily third party ones, particularly those that have a lot written in C
 - I.e. No PyPy + Numpy combo...**yet**

OBJECT-ORIENTED PROGRAMMING IN PYTHON

- Duck typing
- If it looks like a duck and quacks like a duck, it's probably a duck
- I.e. as it's statically typed we can pass any object into a function, as long as it can run the method require it's fine
- Less verbose code, but not checked at compile time

```
class Foo:
    # Constructor is a special method
    def __init__(self, x):
        self.x = x
        self._y = x * 20 #

class Bar(Foo): # Declare superclass after the name
    pass # Pass is used to provide an empty class or method body

my_foo = Foo(5)
print(my_foo.x) # Attributes are always public

# Nothing is preventing me from accessing this private attribute
# although I should be aware that I'm not using the code as
# intended
print(my_foo._y)
```

- Instance methods must explicitly call `self` as the first argument
- Referring to an instance attribute must be prepended with `self`
- All attributes and methods are `public` and cannot be made private
- The convention is to prepend 'private' attributes and methods with an underscore, so that the programmer is aware that the particular object should only be accessed within that class
- Some built-in methods are pre and post pended with `two` underscores, indicating that you are overriding a **special** method

- Not obliged to use OO
- No interfaces
- Multiple inheritance
- Classes are first order objects, can pass them around

- Extremely similar to Java, but different keywords
- Exceptions get raised up the call stack, don't need to keep explicitly try to catch the error at every level
- `try-and-except` and `throwing` errors

```
try:  
    my_num = int(input("Please enter a number: "))  
# If the user enters a string the cast will fail  
except ValueError:  
    print("That wasn't a valid number!")
```

- `sys`, `os`, and `shutil`: Allow interactions with the file system, the operating system and run external programs in a shell.
- `collections`: Additional useful data structures
- `argparse`: Easily parse arguments to make a fully functional command line program
- `sqlite3`: Access basic SQL databases.
- `urllib`: Access websites from within your programs.

- General Python support:
 - [Official Python Tutorial](#)
 - [Learn Python the Hard Way](#) - Teaches general programming concepts as well as Python
 - [Dive into Python](#) - Written for experienced programmers
- Scientific libraries support:
 - numpy.org/
 - docs.scipy.org/doc/scipy-0.15.1/reference/
 - matplotlib.org
 - scikit-learn.org/stable/
 - pandas.pydata.org/

THANKS FOR LISTENING!