

# **COM738 - Data Science Foundations**

## Week 1

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#### **Contents**

- Overview of Data Science
- Applications of Data Science
- Introduction to Python
- Practical experience-development environment & syntax



## What is Data Science?



#### What is Data Science?

 The process of extracting knowledge or insights from data in various forms, either structured or unstructured

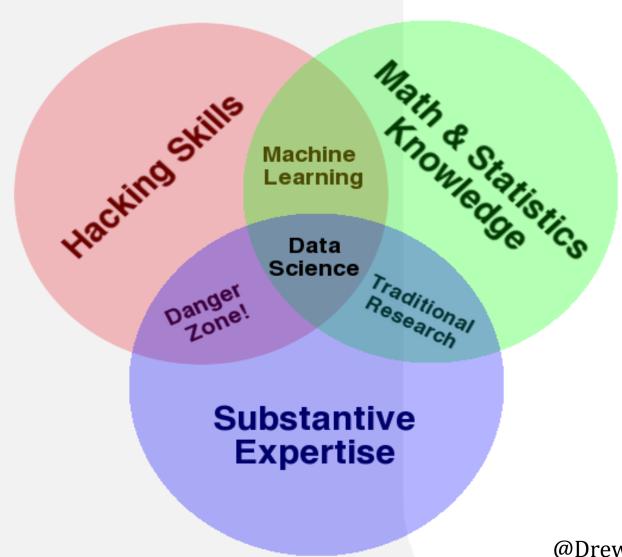
#### Analyse large datasets to:

- Find novel
- Commercially valuable
- Exploitable patterns

Meaningful insights



## **Data Science Venn Diagram**





#### What is a Data Scientist?

Someone with a diverse skill set and capable of:

- Identifying organisational needs
- Collecting, cleaning, and storing diverse datasets
- Performing in-depth analysis on collected data through application of software engineering, statistics and machine learning
- Perform experiments to test hypotheses and identify hidden trends
- Summarise findings and communicate insights to a range of stakeholders to create impact

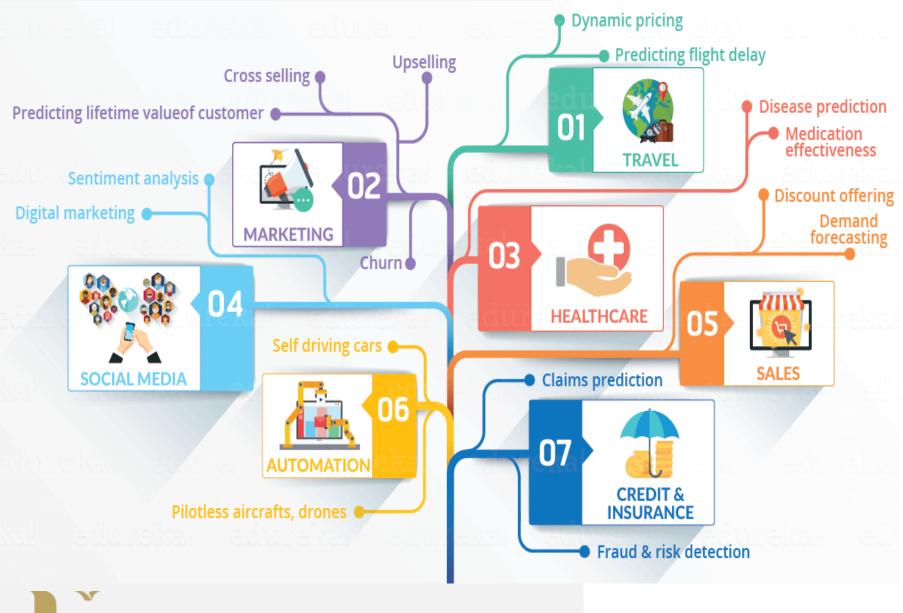


## **Potential of Data Science**



#### What Areas can Benefit from Data Science?







#### **Deliverables**

- Prediction
  - Predict a value based on inputs
- Classification
  - E.g., spam or not spam
- Recommendations
  - E.g., Amazon and Netflix recommendations
- Pattern detection and grouping
  - E.g., classification without known classes



#### **Deliverables**

- Anomaly detection
  - E.g., fraud detection
- Recognition
  - E.g., image, text, audio, video, facial, etc.
- Actionable insights
  - via dashboards, reports, visualizations, ...
- Automated processes and decision-making
  - E.g., credit card approval



#### **Deliverables**

- Scoring and ranking
  - E.g., FICO score
- Segmentation
  - E.g., demographic-based marketing
- Optimization
  - E.g., risk management
- Forecasts
  - E.g., sales and revenue



## Data Science-Meaningful Insight

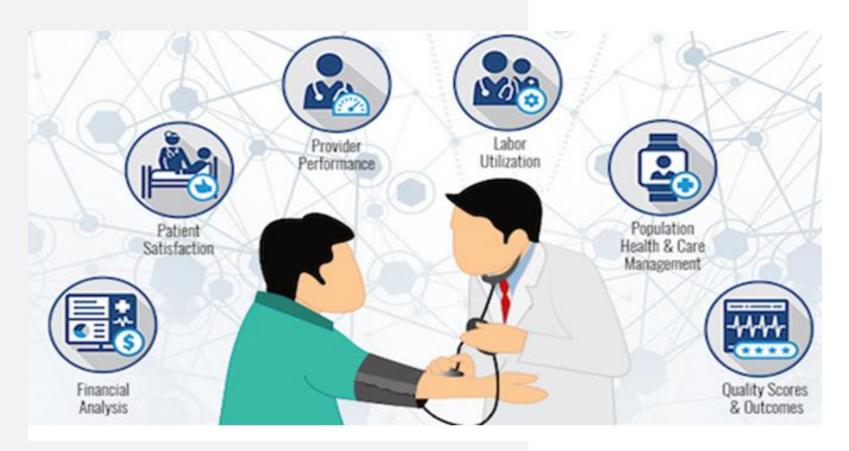
- Predicting from data with certain degree of certainty
- > Generalise what has been observed in data
- ➤ Insight found in data are discoveries
- Discoveries are expressed as models
- Models are used for making prediction

Discoveries → Models → Prediction



## **Example**

#### **Data Science is Transforming Healthcare Systems**





# Data Science - A Brief History

– J.W. Tukey – Published "The Future of Data Analysis", mentioning the relationship between data analysis and statistics.

– P. Naur – Published the "Concise Survey of Computer Methods", mentioning "Data Science".

1996 – "Data Science" used in a conference title

– Introduced as an independent discipline.

– Data Science Journals start publishing



## **Data Science - A Brief History**

#### 2012 - Harvard Business Review



Data Scientist: The Sexiest Job of the 21st Century



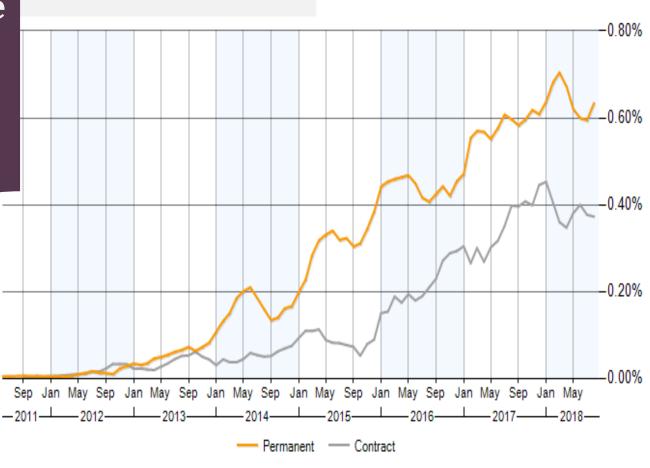


Big Data: The Management Revolution





Data Science
- A Brief
History



Job postings that featured Data Scientist in the job title as a percentage of all IT jobs advertised



## **Big Data**

#### The Rise of Big Data

- **✓ 2000** 25% of data is digital
- ✓ The amount of digital data doubles
- ✓ **2013** Less than 2% of stored data
- ✓ **Datafication** The ability to render into data of the world that have never been quantified before, e.g.:
  - Location (GPS)
  - Friendships (Facebook Likes)
  - Activity levels



Volume 92 • Number :

### The Rise of Big Data

How It's Changing The Way We Think About the World

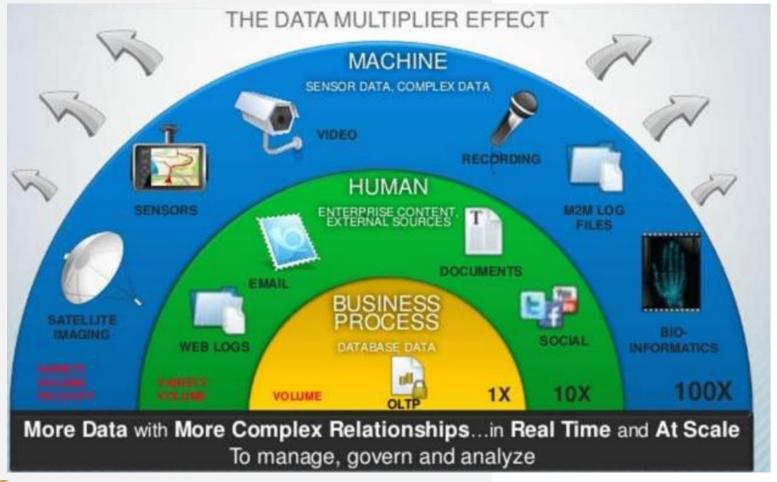
Kenneth Cukier and Viktor Mayer-Schoenberger

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## **Data Types**





## **Big Data**

#### The Rise of Big Data

#### **Profound changes in approaching data:**

- ✓ Massive vs small samples
- ✓ Messiness vs Pristine
- ✓ Correlation vs Causation
  - E.g. Predicting delivery van breakdown



Volume 92 • Number 3

### The Rise of Big Data

How It's Changing The Way We Think About the World

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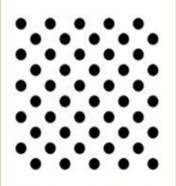
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## **Big Data - The 5 Vs Everyone Must Know**

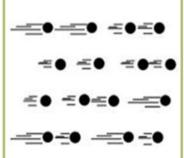
#### Volume



**Data at Rest** 

Terabytes to Exabytes of existing data to process

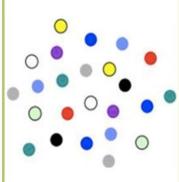
#### Velocity



## Data in Motion

Streaming data, requiring milliseconds to seconds to respond

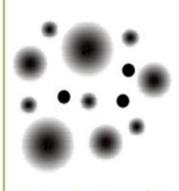
#### Variety



## Data in Many Forms

Structured, unstructured, text, multimedia,...

#### Veracity



#### **Data in Doubt**

Uncertainty due to data inconsistency & incompleteness, ambiguities, latency, deception, model approximations

#### Value



#### Data into Money

Business models can be associated to the data



## **Big Data Variety**

```
SmartCities |
                     Medical Devices
Questionnaires
                    Census Housing
            GoogleThermalCameras News Fith
              Images Security Cameras
                           SmartAppliances
                         GoogleSearches
Wearables
SalesRecords
SmartHomes Weather
  CustomerPurchases
Sensors FacebookMedicalRecords
ConsumerSurveys BusinessApps Public Transport
PaperRecords HealthService
                                       ineLogs
```

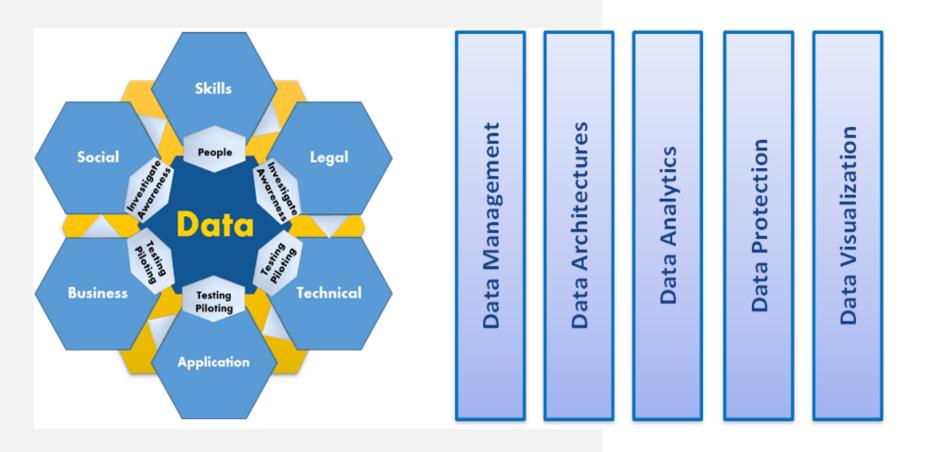






Sizing "Mobile + Social" Big Data Stats

## The Multiple Dimensions of Big Data





## **Big Data vs Data Science**

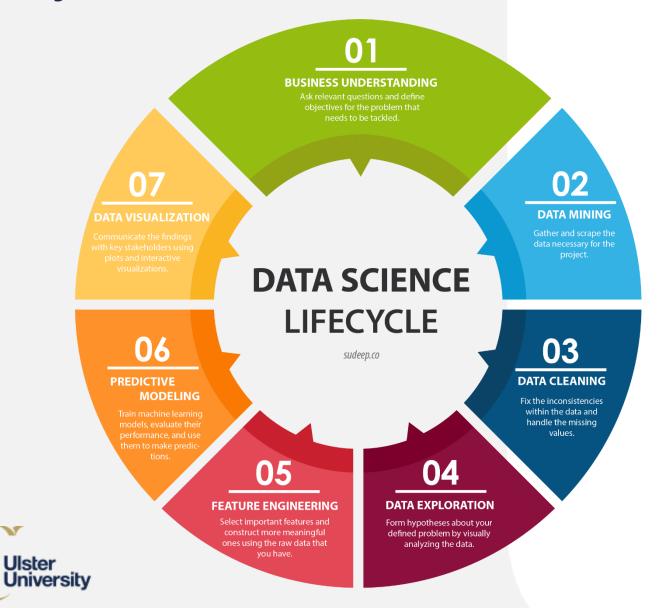
#### Big data

- → Relates to the efficient collection of a large volume of heterogeneous data
- → Data not stored in traditional way can be structured, semistructured or unstructured
- → Not stored in traditional databases

#### **Data Science**

- → Deals with slicing and dicing of the big chunks of data
- → Finding insightful patterns and trends using technology, mathematics and statistical techniques
- → Intelligent analysis of large data





#### Phase 1—Discovery

## (1010) (010101) 01010

#### Before you begin the project:

- Understand various specifications, requirements, priorities and required budget
- Ask the right questions
- Assess if you have the required resources to support the project:
  - People, technology, time and data to support the project
- Frame the business problem and formulate initial hypotheses (IH) to test



#### Phase 2—Data preparation



- Need to explore, preprocess and condition data prior to modeling
- Extract, transform, and load data
- Example:
  - Can use R for data cleaning, transformation, and visualisation
- Will help to spot the outliers and establish a relationship between the variables



#### Phase 3—Model planning



- Get insights into the nature of the data
- Do exploratory data analytics- Descriptive statistics using various statistical formulas and visualization tools
- Determine methods and techniques to draw the relationships between variables → Input features and output
- These relationships will set the base for the algorithms implemented in the next phase



#### Phase 4—Model building

- Build model using training dataset and test the performance of the model using test dataset
- Analyze various learning techniques like classification, association and clustering to build the model



### **Phase 5—Operationalize**



- Final reports, briefings, code and technical documents
- Sometimes a pilot project is also implemented in a real-time production environment
- This will provide a clear picture of the performance and other related constraints on a small scale before full deployment



#### Phase 6—Communicate results

- Evaluate if goal was achieved that was planned in the first phase
- Identify all the key findings
- Communicate to the stakeholders
- Determine if the results of the project are a success or a failure based on the criteria developed in Phase 1



#### **Research Question:**

Could we predict the occurrence of diabetes and take appropriate measures beforehand to prevent it???

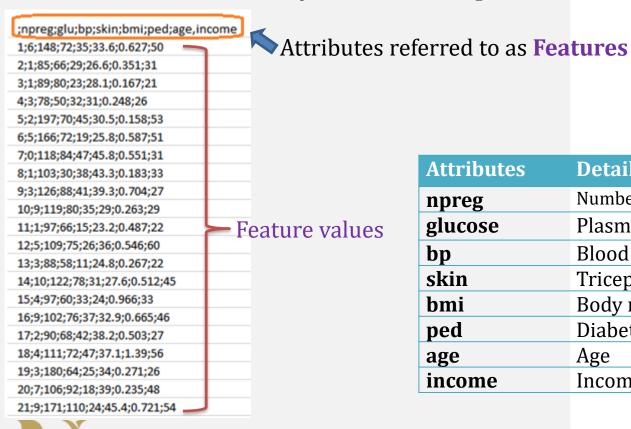




#### Step 1:

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Collect medical history data of the patient



Attributes	Details				
npreg	Number of times pregnant				
glucose	Plasma glucose concentration				
bp	Blood pressure				
skin	Triceps skinfold thickness				
bmi	Body mass index				
ped	Diabetes pedigree function				
age	Age				
income	Income				

#### Step 2:

- Now, once we have the data, clean and prepare the data for data analysis
- This data has a lot of inconsistencies like missing values, blank columns, abrupt values and incorrect data format which need to be cleaned



#### Step 2:

Let's have a look at the sample data below

	npreg	glu	bp	skin	bmi	ped	age	income
1	6	148	72	35	33.6	0.627	50	
2	1	85	66	29	26.6	0.351	31	
3	1	89	6600	23	28.1	0.167	21	
4	3	78	50	32	31	0.248	26	
5	2	197	70	45	30.5	0.158	53	
6	5	166	72	19	25.8	0.587	51	
7	0	118	84	47	45.8	0.551	31	
8	one	103	30	38	43.3	0.183	33	
9	3	126	88	41	39.3	0.704	27	
10	9	119	80	35	29	0.263	29	
11	1	97	66	15	23.2	0.487	22	
12	5	109	75	26	36	0.546	60	
13	3	88	58	11	24.8	0.267	22	
14	10	122	78	31	27.6	0.512	45	
15	4		60	33	24	0.966	33	
16	9	102	76	37	32.9	0.665	46	
17	2	90	68	42	38.2	0.503	27	
18	4	111	72	47	37.1	1.39	56	
19	3	180	64	25	34	0.271	26	
20	7	106	92	18		0.235	48	
21	9	171	110	24	45.4	0.721	54	

- npreg, "one" is written in words whereas it should be in the numeric form like
- *bp* one of the values is 6600 which is impossible
- Income column is blank and also makes no sense in predicting diabetes.
- Therefore, it is redundant to have it here and should be removed from the table



#### Step 2:

- Clean and preprocess this data:
  - Removing outliers
  - Fill up the null values
  - Normalizing the data type

Data preprocessing → Phase 2

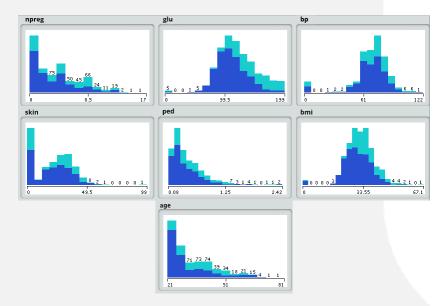
	npreg	glu	bp	skin	bmi	ped	age
1	6	148	72	35	33.6	0.627	50
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19	3	180	64	25	34	0.271	26
20	7	106	92	18	39	0.235	48
21	9	171	110	24	45.4	0.721	54



#### Step 3:

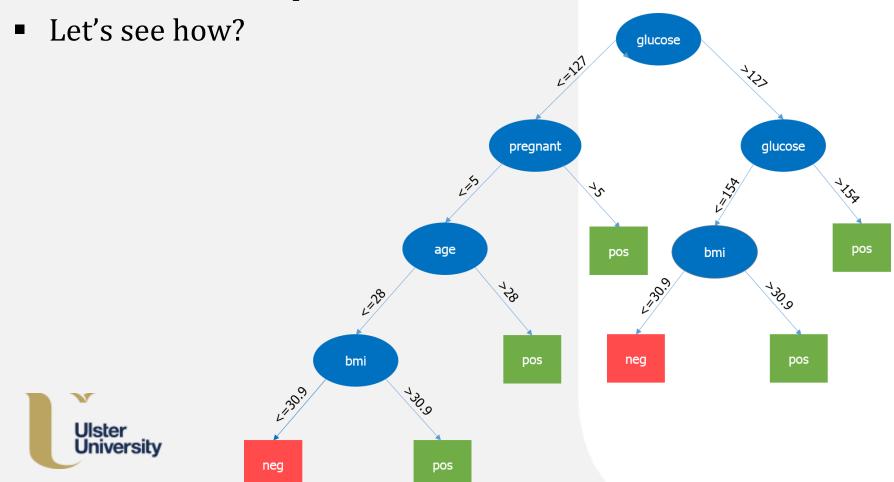
#### **Exploratory data analytics**

- Use functions like describe in R→ Number of missing values and unique values
- Summary function which will give us statistical information like mean, median, range, min and max values
- Use visualization techniques like histograms, line graphs, box plots to get a fair idea of the distribution of data



#### Step 4:

 Based on insights derived from the previous step, the best fit for this kind of problem is the decision tree



#### Step 5:

- Run a small pilot project to check if results are appropriate > Validation
- Also look for performance constraints if any
- If the results are not accurate, replan and rebuild the model.

#### Step 6:

- Once the project executed successfully
- Share the output for full deployment





# Introduction to Python

#### **Python**



- Released in 1989
- Current version: 3.9.0 (January 2021)
- Simplicity -Precise, efficient, readable syntax
- High-Level abstracts details such as memory management
- General purpose AI, statistics, visualisation, etc.
- Comprehensive libraries analysis, machine learning, visualisation
- Interpreted language evaluation of code happens immediately, rather than going through a compile and run cycle
- Recommended for those with a Computer Science / Software Engineering background



#### R

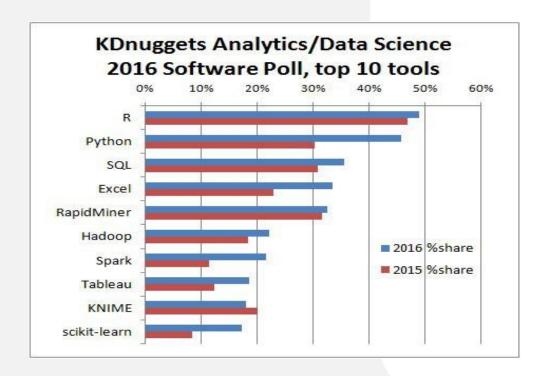
- Released in 1992
- Current version: 3.6.3
- Comprehensive statistics and data mining packages
- Steeper learning curve
- More intuitive for those with a mathematics / statistician background





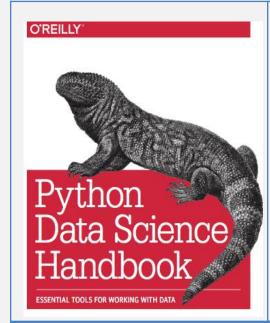
# Python vs R

- ✓ Both excellent languages
- ✓ No "one size fits all"
- ✓ Python serves as a good platform for introduction to data science (less time learning, more time doing)
- ✓ R will be introduced from Week 7 onwards

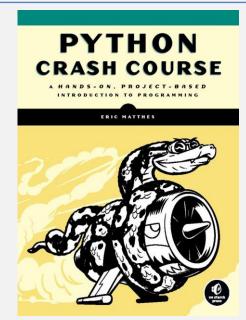




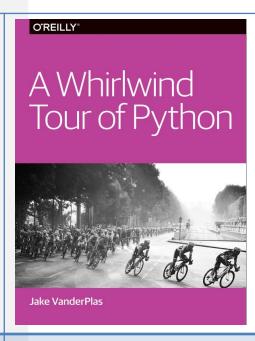
#### **Learning Python-Free Learning Resource**



https://github.com/jakevdp/Py thonDataScienceHandbook



https://github.com/MrAlex6204 /Books/blob/master/pythoncrash-course.pdf



https://www.oreilly.com/progr amming/free/files/awhirlwind-tour-of-python.pdf

http://python.org/

- documentation, tutorials, beginners guide, core distribution, ...



### **Installing Python**

# ANACONDA

#### Easiest method: Use Anaconda

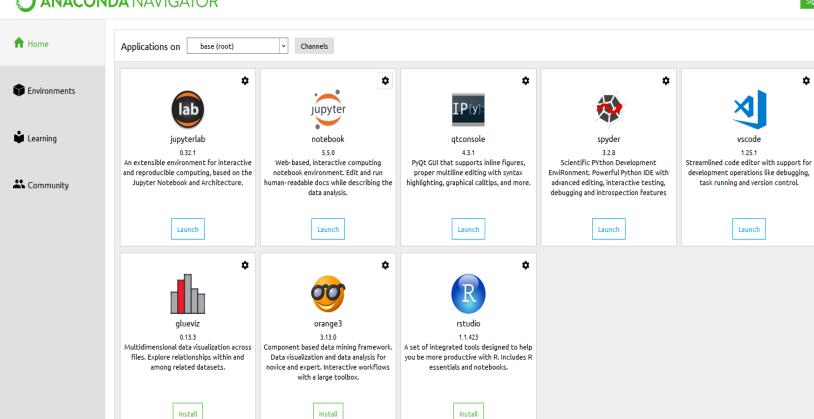
- ✓ A free Python package manager
- ✓ Can be used to install and update Python Packages
- ✓ Including all dependencies
- ✓ Both a Command Line Interface and GUI
- ✓ Already installed in the labs
- ✓ For home use and documentation: <a href="https://docs.continuum.io/">https://docs.continuum.io/</a>





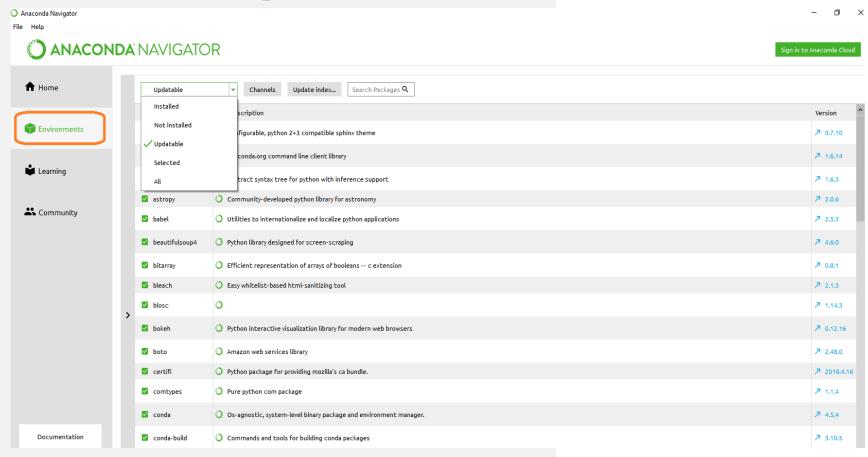






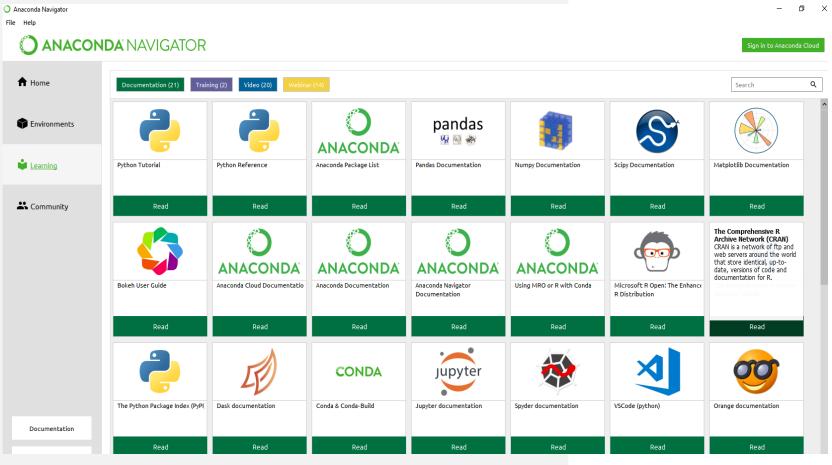
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#### **Installed Packages and Version**



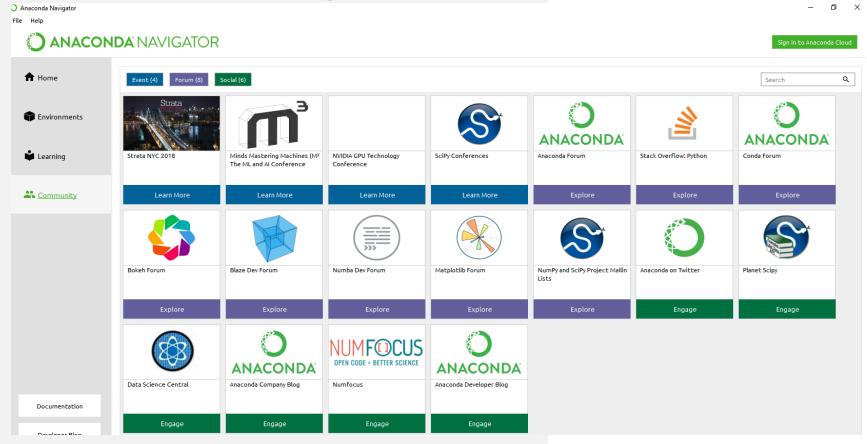


#### **Learning materials**





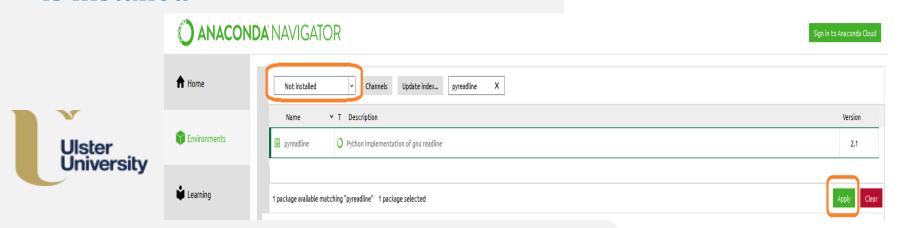
#### Access to community:





# **Jupyter Notebook**

- ✓ A web application to create documents with live code, visualisations, images and text
- ✓ Ideal for beginners and advanced users
- ✓ Jupyter = Julia, Python, and R the first languages supported
- ✓ Main Components:
  - Kernel Runs and inspects user code
  - Dashboard For access to all notebooks created, and for control over the kernel
- ✓ To enable Tab Completion, ensure the "pyreadline" package is installed



# **Jupyter Notebook**

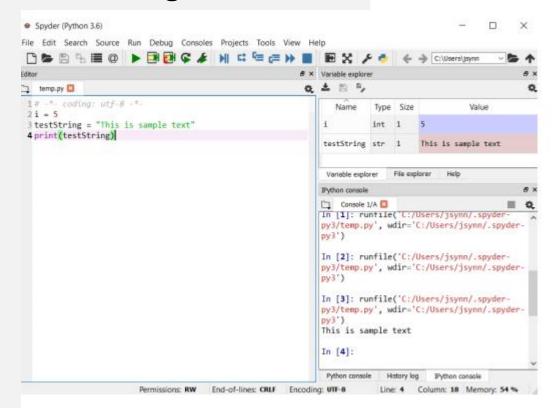
- ✓ Allows segments of code to be re-run quickly
- ✓ Assess impact, experiment, debug
- **✓ View code and output together**
- ✓ Annotate and share code
- ✓ Host remotely, (e.g. high powered server), access anywhere



# Spyder

W.

- ✓ Scientific Python Development IDE
- ✓ Full featured IDE. Ideal for larger codebases
- ✓ Syntax highlighting
- ✓ Exploration and editing of variables from a GUI





### **Jupyter QtConsole**

- ✓ A very lightweight application, acts like an "enhanced terminal"
- ✓ Useful for quickly checking Python object help documents
- ✓ Also supports inline figures, multi-line editing, syntax highlighting etc.

```
File Edit View Kernel Window Help

Jupyter QtConsole 4.3.1
Python 2.7.12 |Anaconda 4.1.1 (64-bit)| (default, Jun 29 2016, 11:87:13) [MSC v. 1580 64 bit (AMD64)]
Type "copyright", "credits" or "license" for more information.

IPython 4.2.6 -- An enhanced Interactive Python.
? -> Introduction and overview of IPython's features.

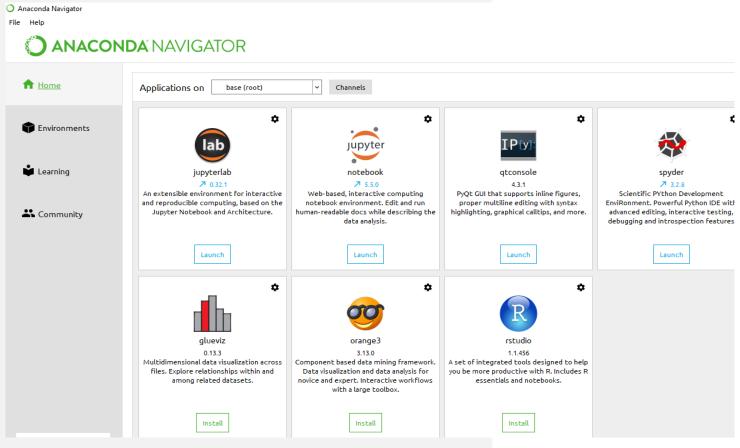
%quickref -> Quick reference.
help -> Python's own help system.
object? -> Details about 'object', use 'object??' for extra details.

In [1]: |
```

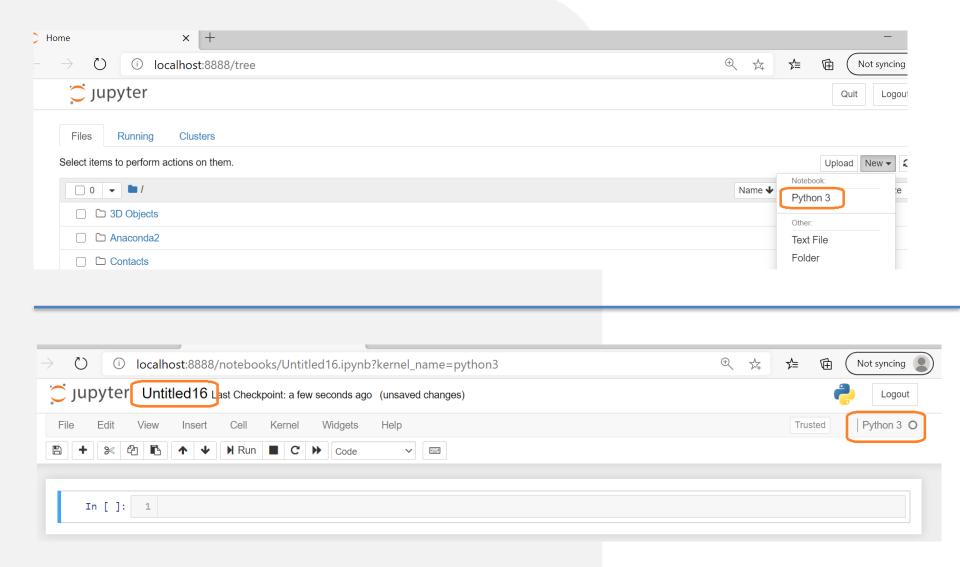


### Working with Jupyter Notebook

Launch the Jupyter Notebook from Anaconda Navigator









- Whitespace Formatting
  - Python uses **indentation** to delimit blocks of code.
  - Other languages use curly brackets {}
  - E.g.

```
for i in [1,2,3,4,5]:
    print(i) #first line in 'for i' block
    for j in [6,7,8,9,10]:
        print(j)
        print(i + j) #Last line in 'for j' block
    print(i) #Last line in 'for i' block
    print("end") #First line after loops.
```

Whitespace within lines has no impact



- Whitespace Formatting
  - Ignored inside:
    - Parentheses (round brackets) ()
    - Curly brackets, aka. Braces { }
    - Brackets, aka. Square Brackets []

\* Both variables are same.

But the second one is more readable presentation

You can also use a backslash to continue a statement on a new line



- Comments: testString = "functional line" #Comment in line #Comment in separate line
- Parenthesis ()
  - Used for grouping mathematical operations:

```
In [18]: 2*(5+2)
Out[18]: 14
```

• And to indicate a function is being called:

```
print('text')
text
```

Note: Must be used during a function call even if no parameters are being supplied:

numbers.sort()
print(numbers)

[1, 2, 3, 4]

```
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```

- Terminology
  - A statement is an instruction that the language implementation understands. In most languages, statements do not return values.
  - Afunction is a subroutine that can be called elsewhere in the program. They often return values.
  - A control structure is a statement that can be used to direct the flow of execution (e.g. an if statement)
  - An expression is a piece of code that evaluates to a value (e.g. 1+3)



- The *print* function
  - Something you will notice when following online tutorials:
    - Python **2.x** *print* acts is a **statement**

```
numbers = [4,3,1,2]
numbers.sort()
print numbers
```

• Python **3.x** – *print* is a **function** 

```
numbers = [4,3,1,2]
numbers.sort()
print(numbers)
```



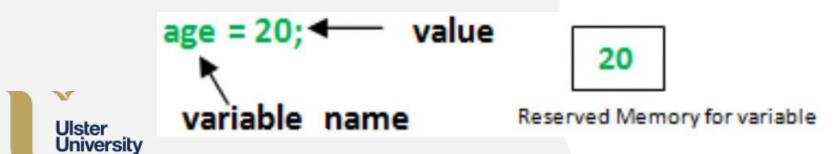
Saving the work in Jupyter notebook



#### 

#### **Variables**

- In Python, these are pointers to a bucket in memory, e.g.:
  - x=2
- This is defining a pointer, called x
- This pointer points to memory bucket, containing the value 2
- Therefore:
  - Python variables just point tovarious objects
  - Avariable <u>can be assigned to different object types</u>
  - There is no need to declare the variable or its type (e.g. int x = 2)
  - Python is dynamically typed to its variable type



#### Python Collections (Arrays)- Built-In Data Structures

There are four collection data types in Python:

- List ordered and changeable. Allows duplicate members
- Tuple ordered and unchangeable. Allows duplicate members
- Set unordered and unindexed. No duplicate members
- Dictionary unordered, changeable and indexed. No duplicate members

LIST	TUPLE	DICTIONARY	SET
Allows duplicate members	Allows duplicate members	No duplicate members	No duplicate members
Changeble Not changeable		Changeable, indexed	Cannot be changed, but can be added, non -indexed
Ordered	Ordered	Unordered	Unordered
Square bracket []	Round brackets ( )	Curly brackets{ }	Curly brackets{ }



Built-In Data Structures- Collection types

- Aka Compound types
- Act as containers for other types
- These are:

Type Name	Description	Example
list	Ordered collection	[1, 2, 3]
tuple	Immutable ordered collection	(1, 2, 3)
dict	Unordered (Key, Value) mapping	{'a':1, 'b':2,' c':3}
set	Unordered collection of <b>unique</b> values	{1, 2, 3}



Built-In Data Structures

#### Lists

- Ordered and Mutable.
- Definition: animals = ["cat", "dog", "mouse"]
- Useful properties and methods:

Property/Method	Description	Example
len()	Return the number of elements in the list	len(animals) returns 3
append()	Add the supplied object to the end of the list	animals.append("elephant")
Concatenation	Combine two lists	animals + ["cow","eel"]
sort()	Sorts in ascending order.	animals.sort()



#### Built-In Data Structures

- Lists
  - Can contain multiple data types simultaneously:

```
mixedItems = ["green", 3, None, [1,2,3], True]
mixedItems
['green', 3, None, [1, 2, 3], True]
```

- Indexing and Slicing allow access to specific elements.
  - Indexing
    - Single elements, using a zero-based index.

```
animals = ["cat", "dog', "chicken"]
animals[1]
'dog'
```

Negative indexes allow us to start from the end of the list:



```
animals = ["cat","dog", chicken"]
animals[-1]
'chicken'
```

Python index starts from 0

#### List

```
In [3]: 1 thislist = ["apple", "banana", "cherry"]
2 print(thislist)
['apple', 'banana', 'cherry']
```

#### **Access Items**

```
In [4]: 1 thislist = ["apple", "banana", "cherry"]
2 print(thislist[1])
banana
```

#### **Change Item Value**

#### Loop Through a List

```
In [6]: 1 thislist = ["apple", "banana", "cherry"]
2 for x in thislist[1]:
    print(x)

b
    a
    n
    a
    n
    a
    n
    a
```

#### **List Length**

```
In [8]: 1 thislist = ["apple", "banana", "cherry"]
2 print(len(thislist))
```

#### **Add Items**

To add an item to the end of the list: use the append() method:

```
In [9]: 1 thislist = ["apple", "banana", "cherry"]
2 thislist.append("orange")
3 print(thislist)
['apple', 'banana', 'cherry', 'orange']
```

#### To add an item at the specified index: use the insert() method:

```
In [10]: 1 thislist = ["apple", "banana", "cherry"]
2 thislist.insert(1, "orange")
3 print(thislist)
['apple', 'orange', 'banana', 'cherry']
```

- Built-In Data Structures
  - Lists
    - Slicing
      - Facilitates accessing multiple values in a list
      - Use the format: **listName[firstIndex:lastIndex]**

```
word="sentence"
word[0:3]
'sen'
```

- Note that first Index is inclusive, last Index is not.
- The following table shows which positive/negative index corresponds to which list item

Index	0	1	2	3	4	5	6	7	8
Listitem	S	e	n	t	e	n	c	e	Ï
Negative Index	-9	-8	-7	-6	-5	-4	-3	-2	-1



- Built-In Data Structures
  - Lists

Slicing

[ start index included : last index excluded]

Index	0	1	2	3	4	5	6	7	8	9
Listitem	1	2	3	4	5	6	7	8	9	10
Negative Index	-10	-9	-8	-7	-6	-5	-4	-3	-2	-1

We can also omit the first, or last index, defaulting to the length of the index,
 e.g.:



Next retrieval is done by adding +1 to the current index, therefore traversal always goes in forward direction for both positive and negative index

- Built-In Data Structures
  - Lists

[ start index included : last index excluded : specify pace ]

Slicing

Index	0	1	2	3	4	5	6	7	8	9
Listitem	1	2	3	4	5	6	7	8	9	10
Negative Index	-10	-9	-8	-7	-6	-5	-4	-3	-2	-1

numbers[::2]
[1, 3, 5, 7, 9]

numbers[2:8:2]

numbers[::-1]

[10, 9, 8, 7, 6, 5, 4, 3, 2, 1]

• Sections of lists can be replaced as follows:



- Built-In DataStructures
  - Tuples
    - Similar to lists, however: Defined with parenthesis or nobrackets.

```
t = ("red", "green", "blue") t = "red", "green", "blue"
```

- Are **Immutable** (They can not be changed once created)
- Slicing and indexing functions as in Lists.

```
In [15]: 1 thistuple = ("apple", "banana", "cherry")
2 print(thistuple[1])
banana
```

TypeError: 'tuple' object does not support item assignment



- Built-In Data Structures
  - Sets
    - Unordered collection of unique items. Will not store duplicates.

primes = 
$$\{2,3,5,7\}$$
  
odds =  $\{1,3,5,7,9\}$ 

- Useful operations including:
  - Union of sets containing items appearing in either

```
primes | odds
primes.union(odds)
{1, 2, 3, 5, 7, 9}
```

• **Intersection** of sets – containing items appearing in both

```
primes & odds
primes.intersection(odds)
{3, 5, 7}
```

Difference in sets – containing items in set 1 but not set 2.

```
primes - odds
primes.difference(odds)
{2}
```

• **Symmetric Difference** in sets – items appearing in only one set.

```
primes ^ odds
primes.symmetric_difference(odds)

{1, 2, 9}
```



- Built-In Data Structures
  - Dictionaries
    - Unordered mappings of keys to values
    - Mutable
    - Created by comma-separated lists of *key: value* pairs

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

- Items are accessed using the list indexing syntax
- However, the index is a key in the dictionary, not an integer:

```
numbers['two']
2
```

New entries can be added via:

```
numbers['four'] = 4
{'four': 4, 'one': 1, 'three': 3, 'two': 2}
```



- Variables and mutable objects
  - Mutable object = object whose value can change
  - Eg list, set, dict
  - Note: Be aware when using two variables to point at the same *mutable* object! e.g.

```
x = [1,2,3]

y = x #x and y now both point to the same object which contains [1,2,3]

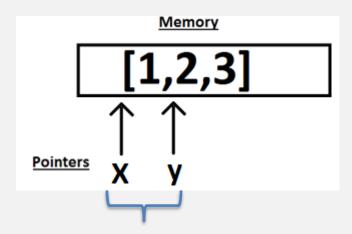
y.append(4) #x and y now both point to the same object which contains [1,2,3,4]

x

[1, 2, 3, 4]
```

 Modifying y has appeared to modify x. This won't make sense unless we remember that python variables are pointers to objects, not the objects themselves





- > x and y are pointing to the same memory location
- > So this memory location can be accessed either using *x* or *y*
- ➤ Therefore, *y.append*(4) is adding 4 to the list





- Variables and immutable objects
  - Immutable object object whose value is unchangeable once it is created
  - Int, float, bool, string, tuple, and other simple types
  - This makes arithmetic operations perform as expected, e.g.

```
x = 5
y = x
x = x * 2
print("x = ",x)
print("y = ",y)

x = 10
y = 5
```

• When we call  $x = x^* 2$ , we are not modifying the value of the 5 object, instead we are changing which objectx points to.



#### **Mutable vs Immutable**

- int, float, bool, str, tuple, unicode are immutable
- list, set, dict are mutable

Check the difference between variable holding mutable object and when variable holding immutable objects what happens?

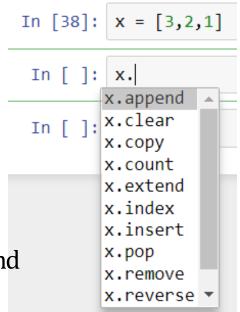


- Objects in Python
  - Everything is an object!
  - Objects have:
    - attributes (metadata)
    - methods (functionality)
  - These are accessed using the dot syntax(.)
  - Tab auto complete helps identify available attributes and methods

#### Note:

- In Jupyter Notebook, you must ensure the variable assignment has been executed prior to using tab autocomplete
  - Otherwise no suggestions, or incorrect suggestions, will be given





- Operators
  - Arithmetic Operators
  - Assignment Operators
  - Comparison Operators
  - Boolean Operators
  - Identity & Membership Operators



#### • Arithmetic

Operator	Name	Description	Example
a+b	Addition	Sum of a and b	2+3=5
a-b	Subtraction	Difference of a and b	2-3=-1
a* b	Multiplication	Product of a and b	2* 3=6
a/b	True Division	Quotient of a and b	2 / 3=0.6666
a// b	Floor Division	Quotient of a and b, without fractional parts	2// 3=0
a%b	Modulus	Remainder after division of a by b	2%3=2
a** b	Exponentiation	araised to the power ofb	2**3 =8
-a	Negation	The negative of a	a=2, -a outputs -2



- Assignment Operators
  - Python contains built-in update operators for all the arithmetic operations.

```
a = a + 2 #instead of this,
a += 2 #we can do this.
```

- Other examples:
  - a-= b
  - a\*= b
  - a/=b
  - etc.



- Comparison Operators
  - To compare different values, returning *True* or *False*.

Operator	Description	Example
a== b	a equal to b	2==2 returns <i>True</i>
a!= b	anot equal to b	2==3 returns <i>True</i>
a <b< td=""><td>aless than b</td><td>1&lt;2 returns <i>True</i></td></b<>	aless than b	1<2 returns <i>True</i>
a>b	a greater than b	2>1 returns <i>True</i>
a<=b	aless than or equal to b	1 <= 2 returns <i>True</i>
a>=b	agreater than or equal to b	2 >= 1 returns <i>True</i>

They can be combined for more complicated comparisons,
 e.g.



**40 < a < 125** #Check if a is between **40** and **125** 

True

- Boolean Operators
  - To compare the values of Boolean statements
  - Useful in control flow
  - *and, or, not*:

```
x=10
(x>1) and (x<9)
False

(x>1) or (x<9)
True

not (x<9)
True</pre>
```



Identity and Membership Operators

Category	Operator	Description	
Idontity	a <b>is</b> b	True if a and b are identical objects	
Identity	a <b>isnot</b> b	True if a and b are not identical objects	
Membership	a <b>in</b> b	True if a is a member of b	
	a <b>not in</b> b	True if a is not a member of b	

Identity and Equality are not the same

```
a = [1,2,3]
b = [1,2,3]

a==b #These objects ARE equal

True

a is b #These objects do not have the same identity / are not identical

False

a=b
a is b #The objects that a and b point to are identical (they are pointing to the same object)

True
```

- Membership Operators
  - Checkfor membership within compound objects

```
'elephant' in ['giraffe','elephant','computer']
True
7 not in [4,3,1]
True
```

• This is distinctly easier to use than other languages which may require the creation of loops and equality checks



#### • Built-In Types – Simple Values

Туре	Example	Description
int	x=1	Integers (whole numbers)
Float	x = 1.0	Floating-point numbers (real numbers)
Complex	x=1+2j	Complex numbers (real + imaginary components)
Bool	x=True	Boolean: True/False values
Str	x='hello'	String: Characters or Text
NoneType	x=None	Special object indicating nulls.

#### Integers

- Any number without a decimal point
- Division returns a floating-point type (Python 3.x)



5/2

2.5

- Built-In Types Simple Values
  - Floating-Point Numbers
    - Store fractional numbers either decimal notation (0.000005) or exponential notation (5e-6) float (5)
    - Cancast from int to floatusing: 5.0
  - Strings
    - Created with 'single' or "double" quotes
    - Many useful string functions are builtin, including:

Function	Description
len(s)	Length of s
s.upper()	Make suppercase
s.lower()	Make s lowercase
s.capitalize()	Capitalise s
S+S	Concatenation
s[0]	Access the first character in s



- Built-In Types Simple Values
  - Boolean Type
    - Two possible values *True* and *False*
    - Returned by comparison operators.
    - **Note:** They are case sensitive! *True* and *False* must be capitalised
    - The *bool()* object constructor can be used to generateBoolean values based on these rules:

Туре	Result	Example
Integer/Float	True if not 0 False if 0	bool(0) returns False
None	False	bool(None)
Strings	True if not empty False if empty	bool("") returns False
Sequences	True if not empty False if empty empty	bool([]) returns false



- Built-In Types Simple Values
  - None Type
    - Aspecial type with only one possible value: *None*
    - Used to refer to nullobjects
    - Often the default return value for functions that do not return any value



#### **Getting Help from Documentation**

- Use Jupyter Notebook get information on the various classes you are using. For example, type:
- <className>? Useful information about a classor collection, e.g. docstrings, function call arguments, constructors, etc.
- *<className>??* Further information, such as source code.
- Note: If using QtConsole, Press'q' to exit the further information page

```
In [107]: list?

Init signature: list(self, /, *args, **kwargs)
Docstring:
list() -> new empty list
list(iterable) -> new list initialized from iterable's items
Type: type
```



# **Getting Help from Documentation**

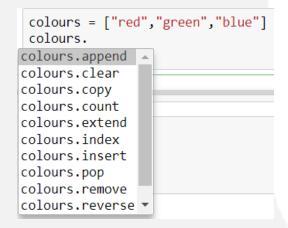
• **Wildcards** can also be used to search for objects in the current

namespace. E.g. *l\*?* 

In [104]: l\*?

len
license
list
locals

 Tab Completion provides a way to explore the structure / attributes of any object you are using. This also works with filenames and directory structures.





# **Getting Help from Documentation**

Information about a specific object can be obtained by:

 Press Shift + Tab in Jupyter Notebook within function parameters to view the function signature:

```
In [103]: colours = ["red", "green", "blue"]
    colours.sort()

Docstring: L.sort(key=None, reverse=False) -> None -- stable sort *IN PLACE*
Type: builtin_function_or_method
```

- Control Flow
  - Conditional statements are used to execute certain blocks of code only when particular conditions are met.
  - If Statement:

```
if x == True:
    print("This will not be printed")
elif x == False:
    print("This will be printed")
else:
    print("this will not be printed")
```



- Loops
  - Repetitions often used to process collections
  - For Loop:

```
for x in range(1,5):
    print(x)

1
2
3
4
```

- Optional Keywords:
  - Continue Skip the remainder of the current loop and continue to the next iteration
  - Break Exit the loop entirely



- Loops
  - While Loop: When the number of iterations is notknown

```
while x < 100:
    x=x*x
    print(x)

4
16
256</pre>
```

