Practical Notebook 1

PLEASE NOTE, the assignments throughout the course are designed to be solved by searching online.

If you have questions about a function, try to Google it or run the command you want more information from with the question mark, ? in front of it

Introduction to Python and Jupyter Notebook

The purpose of this lab is to familiarize oneself with the basics of Python, working with the interactive environment Jupyter Notebooks, as well as some introductory problems with regards to matrix operations, loading data, and plotting. Python is extensively used in machine learning applications.

Jupyter Notebook is an interactive environment for executing Python code (among other languages). These notebooks are executed in sequential order and consist of text and code blocks. Pressing shift + enter or the play arrow to the left will execute one segment at a time and move one down to the next segment. If you are running a notebook with Google Colab, the notebook will be inactive if no work is executed.

Below are presented some examples of Python code some commonly applicable operations.

Print statements

Print statements in pythons can be done in multiple ways. Below are listed some of the common methods for printing variables in Python

```
In [1]: # The various ways for printing values in Python
answer = 42
print("The answer is", answer)
print("The answer is " + str(answer))
print("The answer is %s" % answer)
print(f"The answer is {answer}")

# Printing special character
print("\nItem 1 \nItem 2\n")

The answer is 42
Item 1
Item 2
```

Math functions

```
In [2]: # Math functions
        answer = 42.62
        print("Answer = %.2f with two decimals" % answer)
        print(f"Answer = {answer:.2f} with two decimals\n")
        # Division
        print("5/2 =", 5/2)
        print("5//2 = ", 5//2)
        # Rounding
        print("\nRounding numbers", int(answer))
        print("Rounding numbers", round(answer))
        Answer = 42.62 with two decimals
        Answer = 42.62 with two decimals
        5/2 = 2.5
        5//2 = 2
        Rounding numbers 42
        Rounding numbers 43
```

Lists

One of the most common data types used in Python are lists. Lists can expand or shrink dynamically and contain any data type.

```
In [3]: # Inserting into an empty list
        generic_list = []
        generic_list.append((42, "Answer", True))
        print(generic_list)
        # Generating a list up to a range
         ranged_list = list(range(1, 5))
         print("\nRanged based list:", ranged_list)
        [(42, 'Answer', True)]
        Ranged based list: [1, 2, 3, 4]
        Indexing of a list starts at 0 in python
In [4]: # Retreving elements from a list
        generic_list = [0, 1, 2, 3, 4, 5, 6]
        print("generic_list[0] =", generic_list[0])
        generic_list[0] = 0
        ... and can be done in reverse order
In [5]: print("generic_list[-1] =", generic_list[-1])
        generic_list[-1] = 6
```

```
In [6]: print("Print elements in index 1-5:", generic_list[1:5])
        print("Print all elements up to index 5:", generic_list[:5])
        Print elements in index 1-5: [1, 2, 3, 4]
        Print all elements up to index 5: [0, 1, 2, 3, 4]
In [7]: print("Every other element in a list:", generic_list[::2])
        Every other element in a list: [0, 2, 4, 6]
        Assignment 1a)
In [8]: # ASSIGNMENT
        # print every odd number of the list
        print(generic_list[1::2])
        [1, 3, 5]
        Lists can also be nested in other lists
In [9]: double_list = [[1, 2, 3], ["Bert", "Elmo", "Big Bird"]]
        print(double_list)
        print(double list[1])
        print(double_list[1][1])
        [[1, 2, 3], ['Bert', 'Elmo', 'Big Bird']]
```

Built in list functions

Elmo

['Bert', 'Elmo', 'Big Bird']

min, max, concatinating lists, insertion, removal

```
In [10]: generic_list = [1, 932, 77, 52, 2]
         print("min(generic_list) =", min(generic_list))
         print("max(generic_list) =", max(generic_list))
         print("sum(generic_list) =", sum(generic_list))
         print()
         concaticated_list = ["Bert", "Elmo"] + ["Big Bird"]
         print(concaticated_list)
         # append
         concaticated_list.append("Erni")
         print(concaticated_list)
         # insert at index
         concaticated_list.insert(1, "The Count")
         print(concaticated_list)
         # remove based on index
         concaticated_list.pop(3)
         print(concaticated_list)
```

```
concaticated list.remove("Erni")
          print(concaticated_list)
         # verify if an element is in a list
         Bert_in_list = "Bert" in concaticated_list
          print("\nBert is in the list? ", Bert_in_list)
         min(generic_list) = 1
         max(generic_list) = 932
         sum(generic_list) = 1064
          ['Bert', 'Elmo', 'Big Bird']
          ['Bert', 'Elmo', 'Big Bird', 'Erni']
          ['Bert', 'The Count', 'Elmo', 'Big Bird', 'Erni']
         ['Bert', 'The Count', 'Elmo', 'Erni']
         ['Bert', 'The Count', 'Elmo']
         Bert is in the list? True
         Ranged based loops on lists
In [11]: # Range based for Loop
         length = 5
         for i in range(length):
             print(f''\{i\} \times 2 = \{i*2\}'')
         0 \times 2 = 0
         1 \times 2 = 2
         2 \times 2 = 4
         3 \times 2 = 6
         4 \times 2 = 8
In [12]: # Non pythonic way of printing a list
         name_list = ["Ada Lovelace", "Alan Turing", "Grace Hopper"]
         for i in range(len(name_list)):
             print(name_list[i], "a pioneer in computer science")
         Ada Lovelace a pioneer in computer science
         Alan Turing a pioneer in computer science
         Grace Hopper a pioneer in computer science
In [13]: # Printing a list
         # Assignment: print content of the list with out the `range` function
         name_list = ["Ada Lovelace", "Alan Turing", "Grace Hopper"]
         for i in name list:
             print(i, "a pioneer in computer science")
         Ada Lovelace a pioneer in computer science
         Alan Turing a pioneer in computer science
         Grace Hopper a pioneer in computer science
In [14]: # There are multiple ways to do a ranged based for loop in Python
         name_list = ["Ada Lovelace", "Alan Turing", "Grace Hopper"]
         for i in range(1, (len(name_list)+1)):
             print(i, name_list[i-1])
```

remove value from list

```
print()
name_list = ["Ada Lovelace", "Alan Turing", "Grace Hopper"]
index_list = [1, 2, 3]
for i, name in zip(index_list, name_list):
    print(i, name)
print()
name_list = ["Ada Lovelace", "Alan Turing", "Grace Hopper"]
for i, name in enumerate(name_list):
    print(i+1, name)
print()
1 Ada Lovelace
2 Alan Turing
3 Grace Hopper
1 Ada Lovelace
2 Alan Turing
3 Grace Hopper
1 Ada Lovelace
2 Alan Turing
3 Grace Hopper
```

Assignment 1b)

```
In [15]: # ASSIGNMENT:
    # sort the "generic_list" in ascending order and print it.

generic_list = [1, 932, 77, 52, 2]
# YOUR CODE HERE
print(sorted(generic_list))
# Hint: There is a sorting function in python that will do this. Google it.
```

[1, 2, 52, 77, 932]

Sets

Sets are effective methods for filtering a collection of duplicate values

Assignment 1c)

```
In [16]: lst1 = [0, 1, 3, 5, 6, 7, 5, 8, 9, 10, 12]
lst2 = [1, 2, 2, 1, 2, 4, 5, 8, 2, 13, 9, 11]

# ASSIGNMENT:
# a) print the union of elements between the two lists,
# i.e. all the elements should be included once.
set1 = set(lst1)
set2 = set(lst2)
print(set1.union(set2))

# b) print the intersection of elements between the two lists,
```

```
# i.e. the elements that appear in both lists.
#
print(set1.intersection(set2))
# HINT: Look into `sets` in Python
# YOUR CODE HERE
```

```
{0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13}
{8, 1, 5, 9}
```

functions

Functions in can return multiple objects from a function. The results can either be retrieved as a tuple and specify an index or as individual values

Assignment 1d)

```
In [17]: # ASSIGNMENT:
    # Write a function that will return from a list:
    # 1. The highest
    # 2. The Lowest
    # 3. The first
    # 4. The last elements from a list

def our_custom_function(lst):
    return max(lst), min(lst), lst[0], lst[-1]

lst = [6, 90, 42, -1, 45]

highest, lowest, first, last = our_custom_function(lst)

assert highest == 90
assert lowest == -1
assert first == 6
assert last == 45
```

Assignment 1e)

```
In [18]: # List comprehention
    squared = [x**2 for x in range(10)]
    print("Squared list: \t\t\t ", squared)

# ASSIGNMENT:
# Generate a list of the 20 Fibbonacci numbers,
# BUT exclude all odd numbers Fibbonacci numbers

def fib(n):
    if n <= 1:
        return 1
    else:
        return fib(n-1) + fib(n-2)

# Make sure here you use list comprehension. Find out more about list comprehension</pre>
```

```
# TODO YOUR CODE HERE.

lst = [fib(x) for x in range(20) if fib(x) % 2 == 0]

assert lst == [2, 8, 34, 144, 610, 2584]
```

Squared list: [0, 1, 4, 9, 16, 25, 36, 49, 64, 81]

Python and Libraries

Much like Node, Ruby, and similar languages; Python as language attempts to have a small and extendable library of core functions, where extra functionality can be extended with core libraries or external libraries.

IF are executing a Jupyter Notebook or Python code locally on your computer, we strongly suggest you create and activate a virtual environment for installing your packages such as NumPy, pandas, matplotlib, or similar. This is to ensure that the packages installed to execute this lab do not override previously installed Python libraries. To install packages, we recommend using the Anaconda distribution or Python pip. See this small python guide about how to set up python notebooks locally.

NumPy

NumPy is an external Python library for matrix and vector operations. This means that the library needs to be installed and imported. If you are using a Jupyter Notebook or a plain Python file you may need to install NumPy. In Google Colab, NumPy and other commonly used libraries are pre-installed. Please note that a Python list is not the same as a vector or matrix.

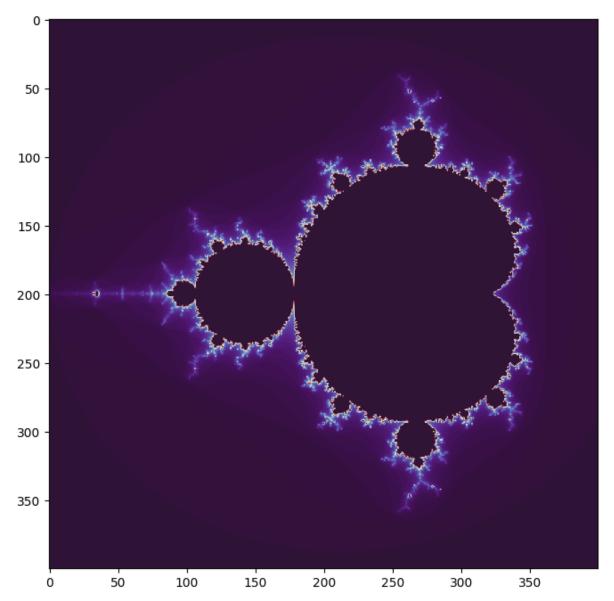
Below is an illustration of how NumPy can be used to illustrate the mathematics of the Mandelbrot set. We use it as a motivating example of what is possible to do with NumPy and other libraries.

```
In [19]: import numpy as np
         import matplotlib.pyplot as plt
         def mandelbrot(h=400, w=400, max_iter=20):
             """Returns an image of the Mandelbrot fractal of size (h,w)."""
             y, x = \text{np.ogrid}[-1.4:1.4:h*1j, -2:0.8:w*1j]
             c = x + y * 1j
             z = c
             divtime = max_iter + np.zeros(z.shape, dtype=int)
             for i in range(max_iter):
                 z = z^{**2} + c
                 diverge = z * np.conj(z) > 2**2
                                                           # who is diverging
                 div_now = diverge & (divtime == max_iter) # who is diverging now
                 divtime[div_now] = i
                                                             # note when
                 z[diverge] = 2
                                                             # avoid diverging too much
```

return divtime

plt.figure(figsize=(8,8)) # change the sizes to view the figure more easily
plt.imshow(mandelbrot(400, 400, 100), cmap='twilight_shifted')

Out[19]: <matplotlib.image.AxesImage at 0x1c16a6a5060>



An installed library can be imported

import numpy.

A library can be imported with an alias

import numpy as np

```
In [20]: import numpy as np
```

The NumPy library is mainly written in C and wrapped in Python for ease of use.

```
In [21]: # Creates a vector of ones
vec = np.array([1, 2, 3])
```

```
# Vectors and lists support mathematical operations differently!
         lst = [1, 2, 3]
         print("2*1st:", 2*1st)
         assert len(2*lst) == 6
         print("2*vec:", 2*vec)
         assert len(2*vec) == 3
         2*lst: [1, 2, 3, 1, 2, 3]
         2*vec: [2 4 6]
In [22]: # One dimentional arrays
         zeros = np.zeros(4)
         print(zeros)
         ranged = np.arange(4)
         print(ranged)
         ranged = np.linspace(0, 1, 4)
         print(ranged)
         custom = np.array([1, 7, 9, 3])
         print(custom)
         # Creating arrays from lists
         custom = np.array([5, 42, 82])
         print(custom)
         [0. 0. 0. 0.]
         [0 1 2 3]
                      0.33333333 0.66666667 1.
         [0.
                                                      ]
         [1 7 9 3]
         [ 5 42 82]
         Making multidimensional arrays
In [23]: # From the start
         multi_arr = np.ones([3, 2])
         print(multi_arr)
         print()
         # Or manually assign the shape
         multi_arr = np.array([[1,1],
                                [1,1],
                                [1,1]]).astype(np.float32)
         print(multi_arr)
         [[1. 1.]
          [1. 1.]
          [1. 1.]]
         [[1. 1.]
          [1. 1.]
          [1. 1.]]
```

```
In [24]: arr = np.array([[1, 2, 3],
                          [4, 5, 6]])
         stacked_horizontal = np.hstack((arr, arr))
         print(stacked_horizontal)
         print()
         stacked_verticaly = np.vstack((arr, arr))
         print(stacked_verticaly)
         [[1 2 3 1 2 3]
          [4 5 6 4 5 6]]
         [[1 2 3]
          [4 5 6]
          [1 2 3]
          [4 5 6]]
         Assignment 1f)
In [25]: import numpy as np
         arr = np.array([1, 2, 3, 4, 5, 6])
         print(f'shape before: {arr.shape}')
         print(f'dimension before: {arr.ndim}')
         # ASSIGNMENT
         # extend or reshape the 1D array into a 2D array - LOOKUP ONLINE
         # i.e. it should have the shape (1,6).
         # YOUR CODE HERE
         arr = np.reshape(arr, (1,6))
         print(f'\nshape after: {arr.shape}')
         print(f'dimension after: {arr.ndim}')
         assert arr.ndim == 2
         assert arr.shape[0] == 1 and arr.shape[1] == 6
         # Feel free to reshape the array to other dimensions as well
         shape before: (6,)
         dimension before: 1
         shape after: (1, 6)
         dimension after: 2
         Transposing an array
In [26]: # Transposing a matrix
         multi_arr = np.array([[1, 2, 3],
                                [4, 5, 6]])
         print("A =\n", multi_arr)
```

```
multi_arr_transp = multi_arr.T
multi_arr_transp = np.transpose(multi_arr)

print("\nA.T =\n", multi_arr_transp)

A =
  [[1 2 3]
  [4 5 6]]

A.T =
  [[1 4]
  [2 5]
  [3 6]]
```

We can also perform the expected mathematical operation on arrays

```
In [27]: arr = np.array([1, 2, 3, 4])
         # Addition
         print("Add one to array", arr + 1)
         # Multiplication
         print("Multiply array by 2", arr*2)
         # Division
         print("Divide array by 2", arr/2)
         # Raised to some power
         print("Log of an array", np.power(arr, 2))
         # Logarithm
         print("Log of an array", np.log(arr))
         print("Log2 of an array", np.log2(arr))
         print("Log2 of an array", np.log(arr)/np.log(2))
         Add one to array [2 3 4 5]
         Multiply array by 2 [2 4 6 8]
         Divide array by 2 [0.5 1. 1.5 2.]
         Log of an array [ 1 4 9 16]
         Log of an array [0. 0.69314718 1.09861229 1.38629436]
         Log2 of an array [0.
                                  1. 1.5849625 2.
                                                                ]
         Log2 of an array [0.
                                             1.5849625 2.
                                                                1
```

As well as commonly used matrix operations

```
arr_mult = A @ B # or
arr_mult = np.matmul(A, B)
print("A*B\n", arr_mult)
# fun fact... The matrix A*B is known as a vampire matrix
A =
[[1 0 0]
[3 1 0]
 [4 0 1]]
b =
[[23 10 12]
[0 0 0]
 [0 0 0]]
A*B
[[23 10 12]
[69 30 36]
[92 40 48]]
```

Assignment 1g)

```
In [29]: M = np.array([[23, 10, 12],
                       [69, 30, 36],
                        [92, 40, 48]])
         # ASSIGNMENT:
         # a) Square an matrix M
         # b) Square the values in a Matrix M
         # Squaring the matrix
         # YOUR CODE HERE
         sqr_arr = np.matmul(M,M)
         assert np.all(sqr_arr == np.array([[2323, 1010, 1212],
                                            [6969, 3030, 3636],
                                            [9292, 4040, 4848]]))
         # Square the values in an array
         # YOUR CODE HERE
         sqr_val = M**2
         assert np.all(sqr_val == np.array([[529, 100, 144],
                                            [4761, 900, 1296],
                                            [8464, 1600, 2304]]))
```

Assignment 1h)

```
In [30]: rand = np.random.RandomState(42)
```

```
In [31]: # Logical operations
    arr = np.array([1,2,3,4,5,6])
    filtered_arr = arr[(arr > 2) & (arr < 6)]
    print(filtered_arr)

filtered_arr = arr[arr%2==0]
    print(filtered_arr)</pre>
[3 4 5]
```

[2 4 6]

Assignment 1i)

```
In [32]: # ASSIGNMENT:
    # Filter out duplicate AND odd numbers

arr = np.array([1, 4, 2, 2, 3, 4, 4, 4, 2, 3, 4, 5, 6])

# YOUR CODE HERE
arr = np.unique(arr[(arr%2==0)])

assert np.all(arr == np.array([2, 4, 6]))
```

Assignment 1i)

What is the difference between the following two statements?

a)

```
{python}
arr1 = np.array([1, 2, 3])
arr2 = arr1
arr2[0] = 100

b)

{python}
arr1 = np.array([1, 2, 3])
arr2 = arr1.copy()
arr2[0] = 100
```

YOUR ANSWER:

```
In [33]:

ASSIGNMENT:

Your answer here...

In a), arr1 and arr2 are referencing to the same vector, meaning that the change in In b), arr2 is a copy of arr1, thus they are referencing equivalent vectors, but no

"""
```

Out[33]: '\nASSIGNMENT:\nYour answer here...\nIn a), arr1 and arr2 are referencing to the s ame vector, meaning that the change in arr2 changes arr1 aswell (since it is the s ame vector).\nIn b), arr2 is a copy of arr1, thus they are referencing equivalent vectors, but not the same vector. The change in arr2 does not affect arr1\n\n'

Random Walk

The task here is to create a random walk function and plot the result.

You will probably not find all the information of how to solve this task in the lab instructons entirely, so you will need to find the answer online.

A random walk is categorized as taking a random step in any number of directions and then accumumalating the distance one has taken.

We want you to make a function which either steps up (+1) or down (-1) in each step (1D random walk) for a 1000 steps and aggrigate the result. If you for example walk [up, up, down, up, up, down, down, down], this corresponds to [1,1,-1,1,-1,-1,-1]. With the starting position 0 this will put you in position [1,2,1,2,3,2,1,0] for the different timesteps.

Assignment 1k)

```
In [34]: # ASSIGNMENT:
    # Complete the random walk function

def random_walk(n=1000):
    steps = np.vectorize(lambda x: -1 if x == 0 else x)(np.random.randint(2, size=n return np.cumsum(steps))

np.random.seed(7)
steps = np.arange(1000)
walk = random_walk(1000)
```

To better illustrate what a random walk does, we can plot the results with a plotting library. For Python, the most commonly used plotting library is Matplotlib. We will discuss Matplotlib in greater detail in a section further on in the lab.

```
In [35]: import matplotlib.pyplot as plt

plt.figure()
 plt.plot(steps, walk)
 plt.title("Random walks")
```

```
plt.ylabel("distance (d)")
plt.xlabel("steps (s)")
```

Out[35]: Text(0.5, 0, 'steps (s)')

