

## Lecture 08:

### Scripts, Modules and Packages



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Python for Molecular Sciences

MSSE 272, 3 Units



When you delete a block of code  
that you thought was useless



### Outline

- Overall Structure
- **Classes**
- Let's build a Package!



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### Outline

- **Overall Structure**
- Classes
- Let's build a Package!





many different steps for a simulation or in a data analysis pipeline

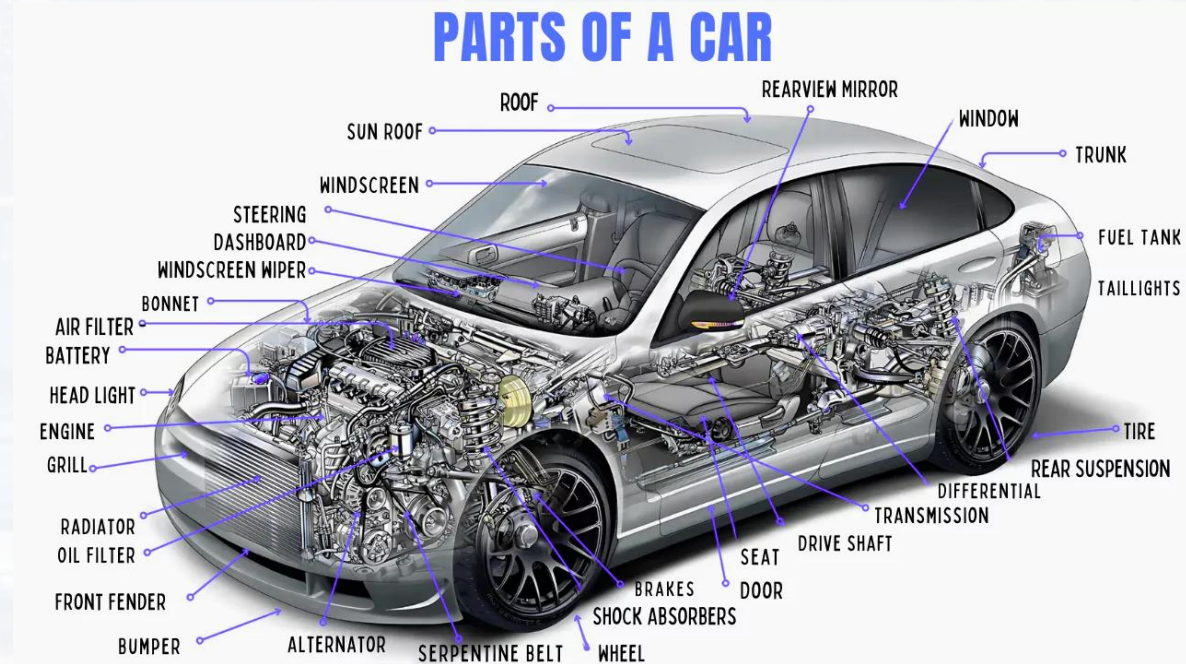
- loading the data file
- extracting data
- analysis part
- visualization
- saving results

separate entities according to their functions

**modularization** or **refactoring**

advantages:

- removing redundancies
- maintainable
- readable
- scalable
- faster



credit: mendmotor

many different steps for a simulation or in a data analysis pipeline

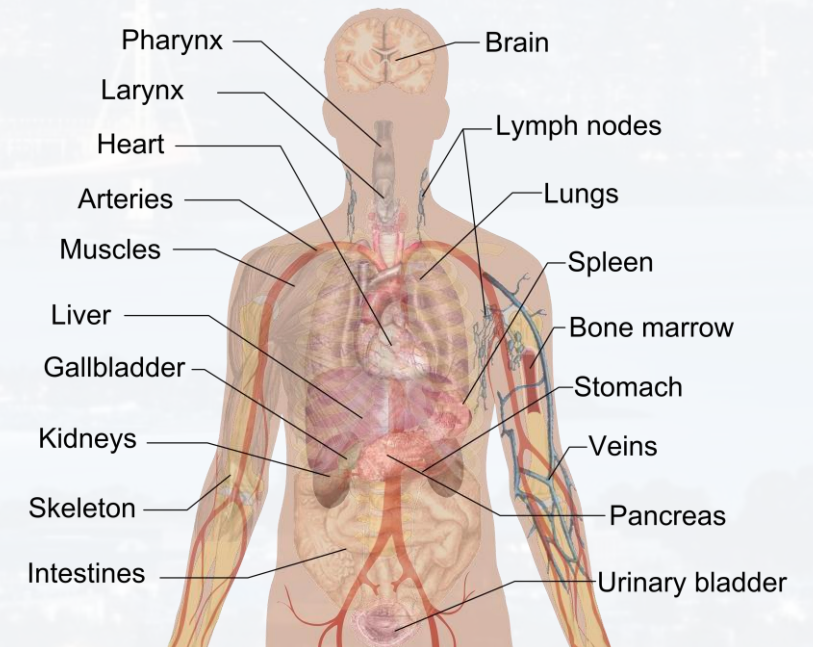
- loading the data file
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separate entities according to their functions

**modularization** or **refactoring**

- advantages:
- removing redundancies
  - maintainable
  - readable
  - scalable
  - faster

### Internal organs



credit: Wikipedia





many different steps for a simulation or in a data analysis pipeline

for example: We need to call the same set of modules for many different tasks

```
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
```

idea: load libraries in the header  
of our package only once

We need to call the same plotting routine at different parts of the code

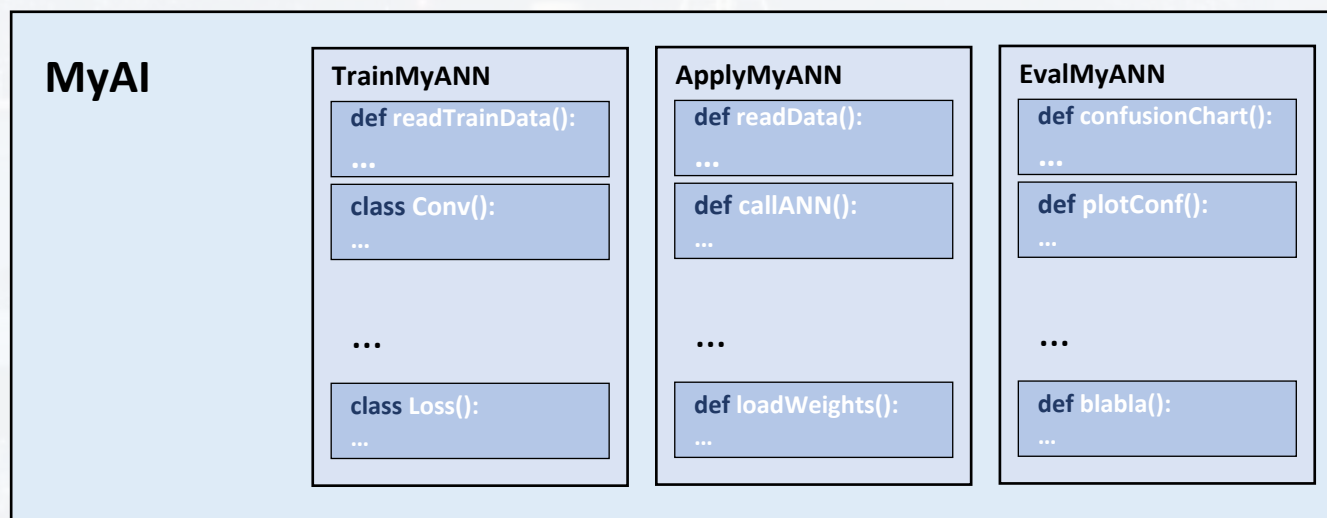
```
plt.scatter(x, y, size, marker = marker, color = color,\
            edgecolor = None, alpha = 0.2)
plt.xlim([-Lim*1.2, Lim*1.2])
plt.ylim([-Lim*1.2, Lim*1.2])
plt.title(title)
plt.show()
```

idea: write a separate plot routine  
→ call it whenever needed



many different steps for a simulation or in a data analysis pipeline

use **functions** and **classes** (see later)





When you delete a block of code  
that you thought was useless



### Outline

- Overall Structure
- **Classes**
- Let's build a Package!





Task: writing a small package that one-hot encodes DNA and plots the result

recap: encoding a sequence with a dictionary

Before we introduce classes: let's do it the old fashion way first!





Task: writing a small package that one-hot encodes DNA and plots the result

recap: encoding a sequence with a dictionary

```
NT = ['A', 'C', 'G', 'T']
```

```
Code = [[1,0,0,0], [0,1,0,0], [0,0,1,0], [0,0,0,1]]
```

```
Dict = {key: value for key, value in zip(NT, Code)}
```

```
Encoder = lambda Sequence: [Dict[s] for s in Sequence]
```

```
In [2]: Encoder('AACCTCGA')
```

```
Out[2]:
```

```
[[1, 0, 0, 0],  
 [1, 0, 0, 0],  
 [0, 1, 0, 0],  
 [0, 1, 0, 0],  
 [0, 0, 0, 1],  
 [0, 1, 0, 0],  
 [0, 0, 1, 0],  
 [1, 0, 0, 0]]
```



Task: writing a small package that one-hot encodes DNA and plots the result

next step: plotting the one-hot encoded sequence

Seq = 'AACTCGTCTGACCTGTCGTATTGGCA'

```
import numpy as np
import seaborn as sns
```

```
E = np.array(Encoder(Seq)).transpose()
```

- 1) calling the Encoder
- 2) for plotting:
  - turning output into an array
  - transposing it

```
In [2]: Encoder('AACCTCGA')
Out[2]:
[[1, 0, 0, 0],
 [1, 0, 0, 0],
 [0, 1, 0, 0],
 [0, 1, 0, 0],
 [0, 0, 0, 1],
 [0, 1, 0, 0],
 [0, 0, 1, 0],
 [1, 0, 0, 0]]
```

```
sns.heatmap(E, cmap = "Blues", xticklabels = list(Seq), yticklabels = [None]*4)
```



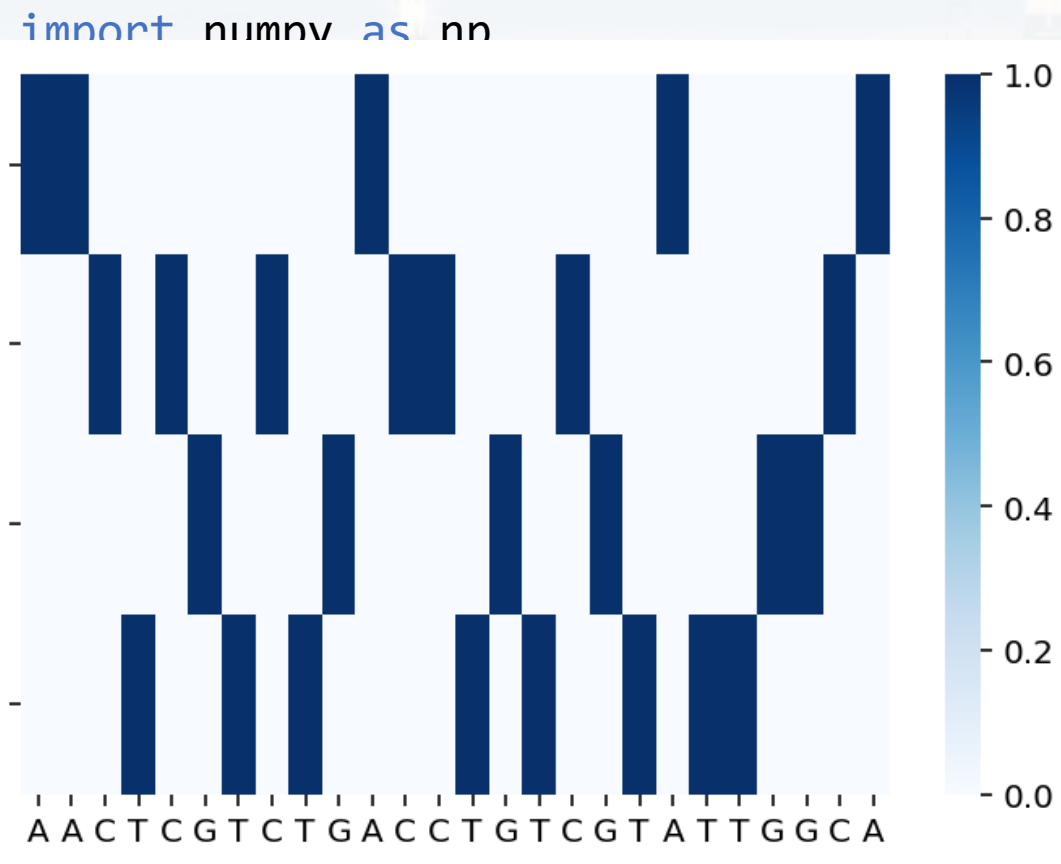


Task: writing a small package that one-hot encodes DNA and plots the result

next step: plotting the one-hot encoded sequence

Seq = 'AACTCGTCTGACCTGTCGTATTGGCA'

```
In [2]: Encoder('AACCTCGA')  
Out[2]:  
[[1, 0, 0, 0],  
 [1, 0, 0, 0],  
 [0, 1, 0, 0],  
 [0, 1, 0, 0],  
 [0, 0, 0, 1],  
 [0, 1, 0, 0],  
 [0, 0, 1, 0],  
 [1, 0, 0, 0]]
```



- 1) calling the Encoder
- 2) for plotting
  - turning output into an array
  - transposing it

```
ls = list(Seq), yticklabels = [None]*4)
```



Task: writing a small package that one-hot encodes DNA and plots the result

If we ran this on a Jupyter Notebook:

```
import numpy as np
import seaborn as sns
```

```
NT      = ['A', 'C', 'G', 'T']
Code    = [[1,0,0,0], [0,1,0,0], [0,0,1,0], [0,0,0,1]]
Dict    = {key: value for key, value in zip(NT,Code)}
Encoder = lambda Sequence: [Dict[s] for s in Sequence]
```

```
Seq = 'AACTCGTCTGACCTGTCGTATTGGCA'
```

```
E = np.array(Encoder(Seq)).transpose()
```

```
sns.heatmap(E, cmap = "Blues", xticklabels = list(Seq), yticklabels=[None]*4)
```

EncodeSequential.ipynb



Task: writing a small package that one-hot encodes DNA and plots the result

Now using `classes`:

- 1) Open a Python Script
- 2) Save it under the name: `OneHotEncoder.py`
- 3) refactoring: we need to call `numpy` and `seaborn` only once!  
we need to define our `dict` and the `Encoder` only once!

```
import numpy as np
import seaborn as sns

NT = ['A', 'C', 'G', 'T']
Code = [[1, 0, 0, 0], [0, 1, 0, 0], [0, 0, 1, 0], [0, 0, 0, 1]]

Dict = {key: value for key, value in zip(NT, Code)}

Encoder = lambda Sequence: [Dict[s] for s in Sequence]
```





Task: writing a small package that one-hot encodes DNA and plots the result

Now using `classes`:

- 1) Open a Python Script
- 2) Save it under the name: `OneHotEncoder.py`
- 3) refactoring: we need to call `numpy` and `seaborn` only once!  
we need to define our dict and the Encoder only once!  
defining the plotting routine as independent function using `def`

```
import numpy as np
import seaborn as sns

NT = ['A', 'C', 'G', 'T']
Code = [[1,0,0,0], [0,1,0,0], [0,0,1,0], [0,0,0,1]]

Dict = {key: value for key, value in zip(NT, Code)}

Encoder = lambda Sequence: [Dict[s] for s in Sequence]
```

```
def MyPlotRoutine(E, Seq):
    sns.heatmap(E, cmap="Blues", xticklabels=list(Seq),\
    .....yticklabels=[None]*4)
```

don't forget the input arguments!



Task: writing a small package that one-hot encodes DNA and plots the result

Now using **classes**: 4) writing a **class**

```
class EncodeMySeq():
```

reading and converting the sequence  
when we call the class

creating the plot by calling the  
plot routine



Task: writing a small package that one-hot encodes DNA and plots the result

Now using `classes`: 4) writing a `class`

```
class EncodeMySeq():
```

```
    def __init__(    , Seq):
```

reading and converting the sequence  
when we call the class

```
        E = np.array(Encoder(Seq)).transpose()
```

```
    def PlotMySeq(    ):
```

```
        MyPlotRoutine(E, Seq)
```

creating the plot by calling the  
plot routine

don't forget the input arguments!

How do we handover these arguments from  
`__init__` to `PlotMySeq`?





Task: writing a small package that one-hot encodes DNA and plots the result

Now using `classes`: 4) writing a `class`

```
class EncodeMySeq():
```

```
    def __init__(self, Seq):
```

reading and converting the sequence  
when we call the class

```
        self.E = np.array(Encoder(Seq)).transpose()
```

```
        self.Seq = Seq
```

```
    def PlotMySeq(self):
```

creating the plot by calling the  
plot routine

```
        MyPlotRoutine(self.E, self.Seq)
```

don't forget the input arguments!

How do we handover these arguments from  
`__init__` to `PlotMySeq`?

→ using `self`



our code now:

```
import numpy as np
import seaborn as sns

NT = ['A', 'C', 'G', 'T']
Code = [[1,0,0,0], [0,1,0,0], [0,0,1,0], [0,0,0,1]]

Dict = {key: value for key, value in zip(NT, Code)}

Encoder = lambda Sequence: [Dict[s] for s in Sequence]

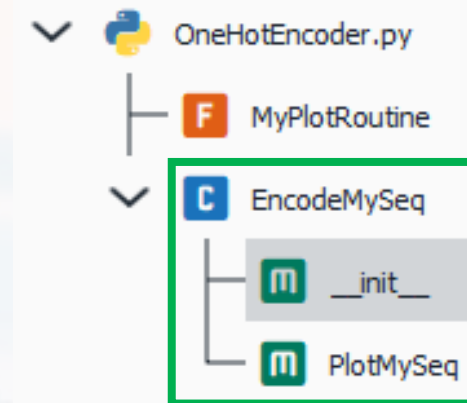
def MyPlotRoutine(E, Seq):
    sns.heatmap(E, cmap = "Blues", xticklabels = list(Seq),\
                yticklabels = [None]*4)

class EncodeMySeq():

    def __init__(self, Seq):

        self.E = np.array(Encoder(Seq)).transpose()
        self.Seq = Seq

    def PlotMySeq(self):
        MyPlotRoutine(self.E, self.Seq)
```



our class has  
two methods



our code now:

```
import numpy as np
import seaborn as sns
```

```
NT = ['A', 'C', 'G', 'T']
Code = [[1,0,0,0], [0,1,0,0], [0,0,1,0], [0,0,0,1]]

Dict = {key: value for key, value in zip(NT, Code)}

Encoder = lambda Sequence: [Dict[s] for s in Sequence]

def MyPlotRoutine(E, Seq):
    sns.heatmap(E, cmap = "Blues", xticklabels = list(Seq),\
                yticklabels = [None]*4)

class EncodeMySeq():

    def __init__(self, Seq):

        self.E = np.array(Encoder(Seq)).transpose()
        self.Seq = Seq

    def PlotMySeq(self):
        MyPlotRoutine(self.E, self.Seq)
```

a final adjustment!





Let's try to understand how a `class` works!

1) Compiling our package `OneHotEncoder.py`

2) Creating three sequences

```
S1 = 'AAACCTCTGGTAATT'  
S2 = 'GGTTTGACACATGTCCGT'  
S3 = 'TTAGTCTTGT'
```

3) **initializing** an instance (object) of the class `EncodeMySeq`

```
S1 = 'AAACCTCTGGTAATT'  
S2 = 'GGTTTGACACATGTCCGT'  
S3 = 'TTAGTCTTGT'
```

```
En1 = EncodeMySeq(|
```

```
En1__EncodeMySeq(Seq)
```

```
No documentation available
```

```
En1 = EncodeMySeq(S1)
```

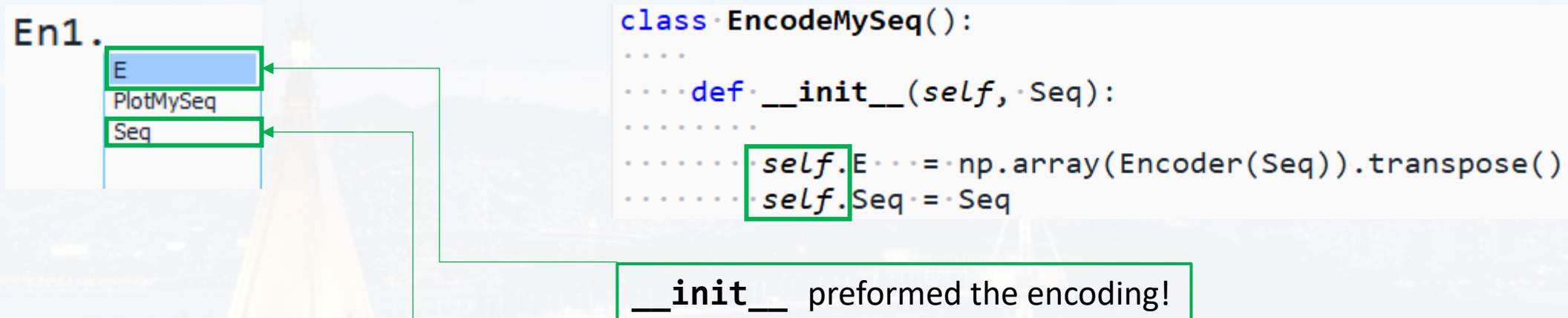
creating an instance (En1) of `EncodeMySeq`  
→ runs the `__init__` method



Let's try to understand how a `class` works!

```
En1 = EncodeMySeq(S1)
```

creating an instance (En1) of EncodeMySeq  
→ runs the `__init__` method



E, PlotMySeq and Seq are so called **attributes** of the `class` EncodeMySeq

```
In [38]: print(En1.E)  
[[1 1 1 0 0 0 0 0 0 0 0 1 1 0 0]  
 [0 0 0 1 1 0 1 0 0 0 0 0 0 0 0]  
 [0 0 0 0 0 0 0 0 0 1 1 0 0 0 0]  
 [0 0 0 0 0 1 0 1 0 0 1 0 0 1 1]]
```



E, PlotMySeq and Seq are so called **attributes** of the **class** EncodeMySeq

```
In [38]: print(En1.E)
[[1 1 1 0 0 0 0 0 0 0 0 1 1 0 0]
 [0 0 0 1 1 0 1 0 0 0 0 0 0 0 0]
 [0 0 0 0 0 0 0 0 1 1 0 0 0 0 0]
 [0 0 0 0 0 1 0 1 0 0 1 0 0 1 1]]
```

**attributes** can be functions or variables (and many other types too)

```
In [39]: dir(En1)
```

Out[39]:

```
['E',  
 'PlotMySeq',  
 'Seq',
```

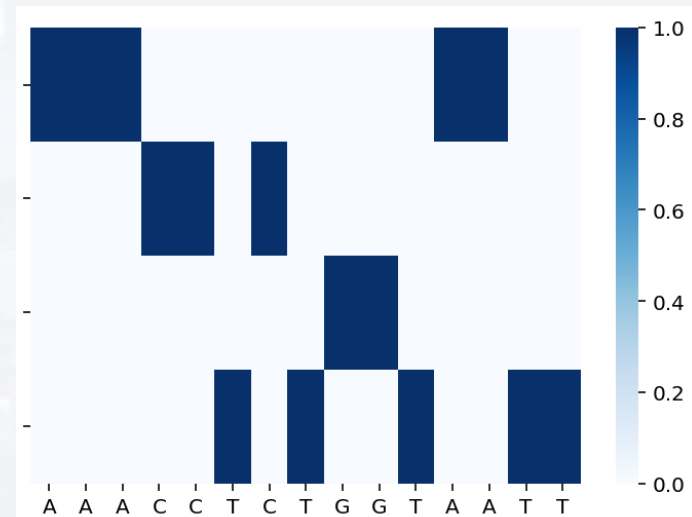
```
type(En1.Seq)
str
```

```
type(En1.E)
numpy.ndarray
```

```
type(En1.PlotMySeq)
method
```

## running the plotting routine

## En1.PlotMySeq()





We can create as many **instance** of a **class** as we want!

```
En1 = EncodeMySeq(S1)
```

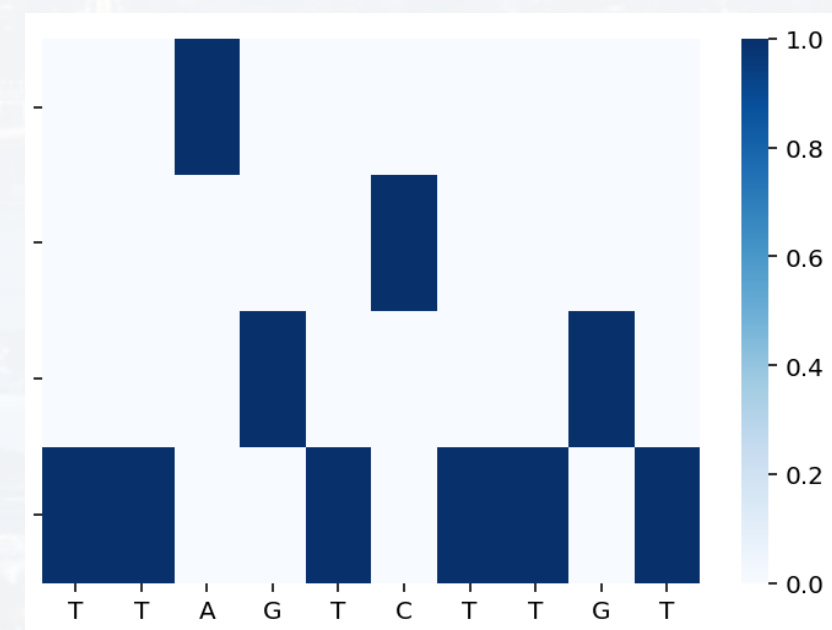
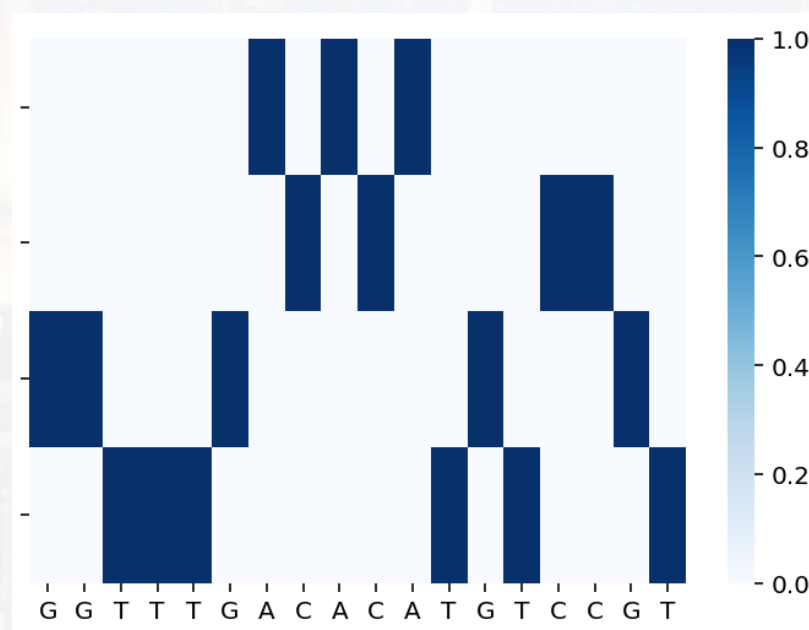
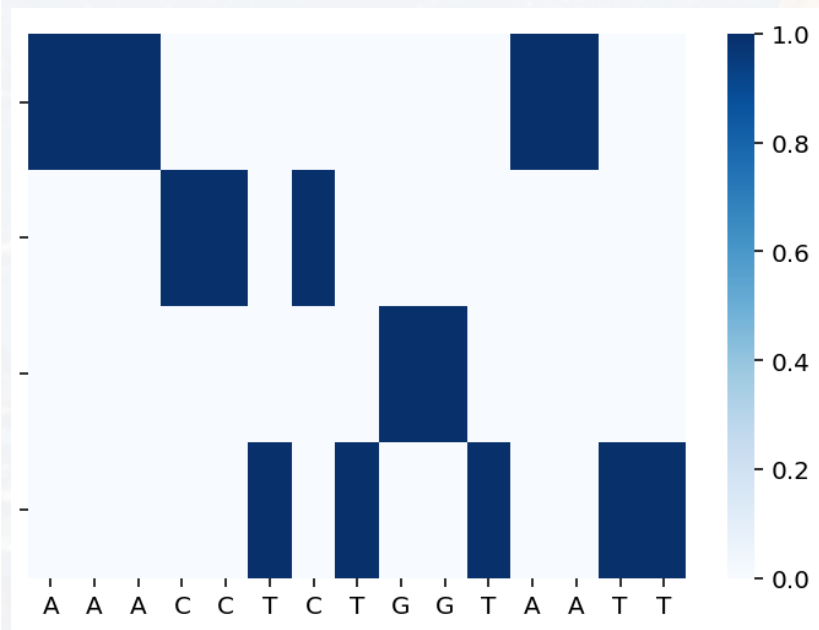
Makes the code scalable! → parallelization!

```
En2 = EncodeMySeq(S2)
```

```
En3 = EncodeMySeq(S3)
```

```
S = [S1, S2, S3]
```

```
[EncodeMySeq(s).PlotMySeq() for s in S]
```

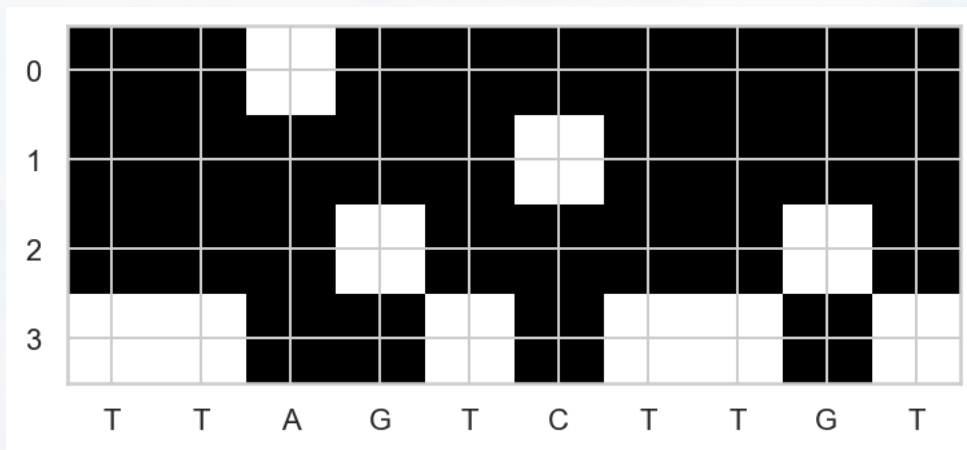






Our package is now more maintainable!

Example: adding a **new plot routine**



1) adding the plot routine to the header

```
def MyPlotRoutine2(E, Seq):  
    plt.imshow(E, cmap = 'gray')  
    plt.xticks(ticks = range(len(Seq)), labels = list(Seq))  
    plt.show()
```



Our package is now more maintainable!

Example: adding a **new plot routine**

renaming the old plotting routine from **MyPlotRoutine** to **MyPlotRoutine1**

```
def MyPlotRoutine1(E, Seq):  
    ...sns.heatmap(E, cmap="Blues", xticklabels=list(Seq),\  
    ...yticklabels=[None]*4)  
    ...plt.show()  
    ...  
    ...  
  
def MyPlotRoutine2(E, Seq):  
    ...plt.imshow(E, cmap='gray')  
    ...plt.xticks(ticks=range(len(Seq)), labels=list(Seq))  
    ...plt.show()  
  
class EncodeMySeq():
```



Our package is now more maintainable!

Example: adding a **new plot routine**

```
class EncodeMySeq():  
    ...  
    def __init__(self, Seq):  
        ...  
        self.E = np.array(Encoder(Seq)).transpose()  
        self.Seq = Seq  
  
    def PlotMySeq1(self):  
        ...  
        MyPlotRoutine1(self.E, self.Seq)  
        ...  
  
    def PlotMySeq2(self):  
        ...  
        MyPlotRoutine2(self.E, self.Seq)
```

don't forget to rename the old plot routine

2) adding the plot routine to the **class**



## Our package is now more maintainable!

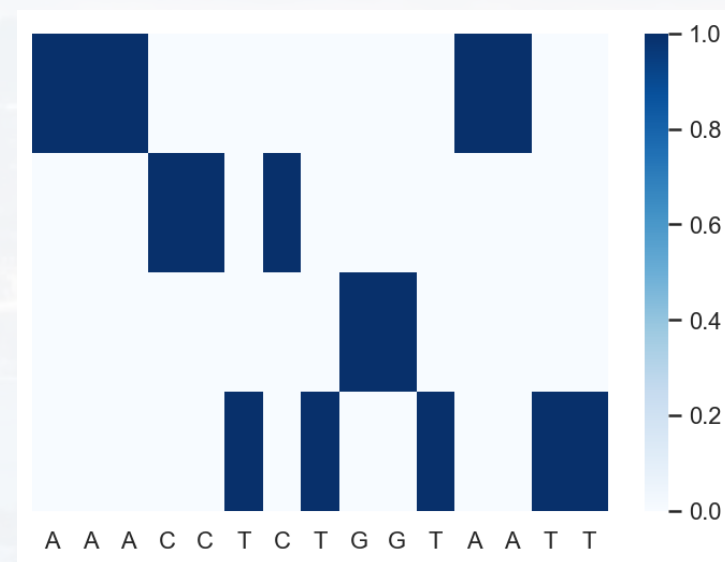
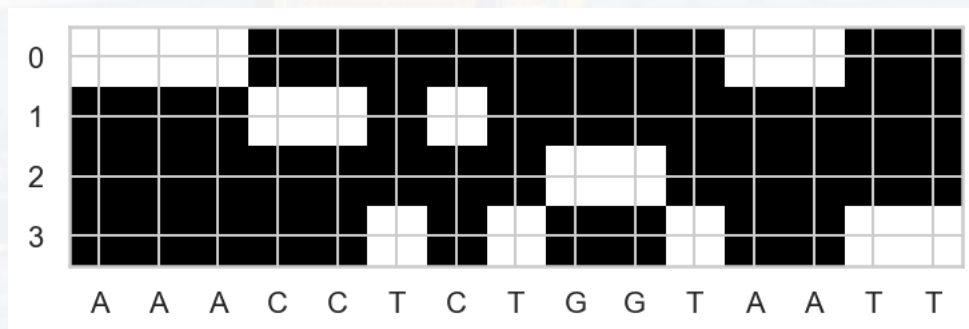
## Example: adding a new plot routine

We can now choose a plot without reloading the data (saves time)!

```
E = EncodeMySeq(S1)
```

## E.PlotMySeq1()

E.PlotMySeq2()





We can also loop over all sequences **and** plot routines via looping over the attributes using **getattr**

Checking out **getattr** first:

```
E = EncodeMySeq(S1)
```

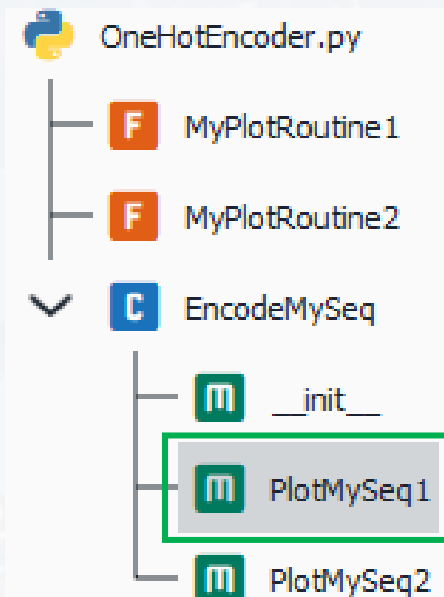
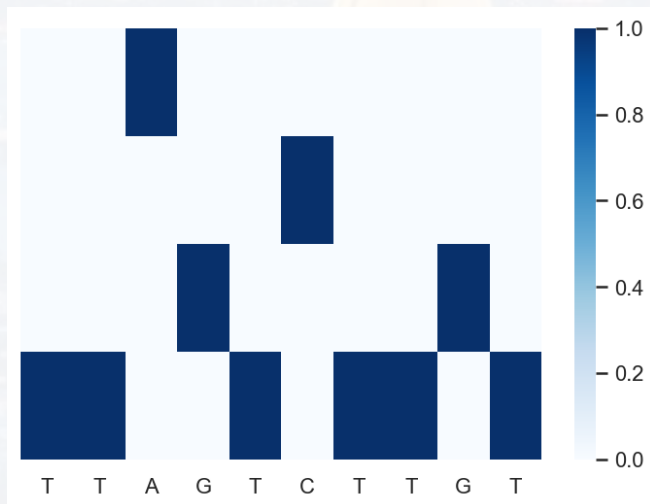
initializing an instance as before

```
M = getattr(E, 'PlotMySeq1')
```

retrieving the attribute *'PlotMySeq1'* which is a method

```
M()
```

running the method as before

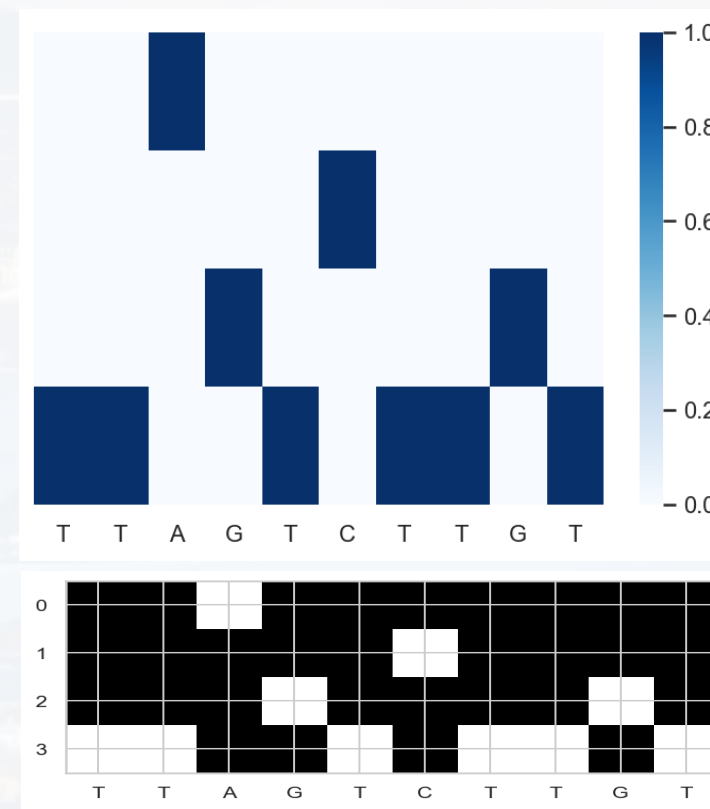
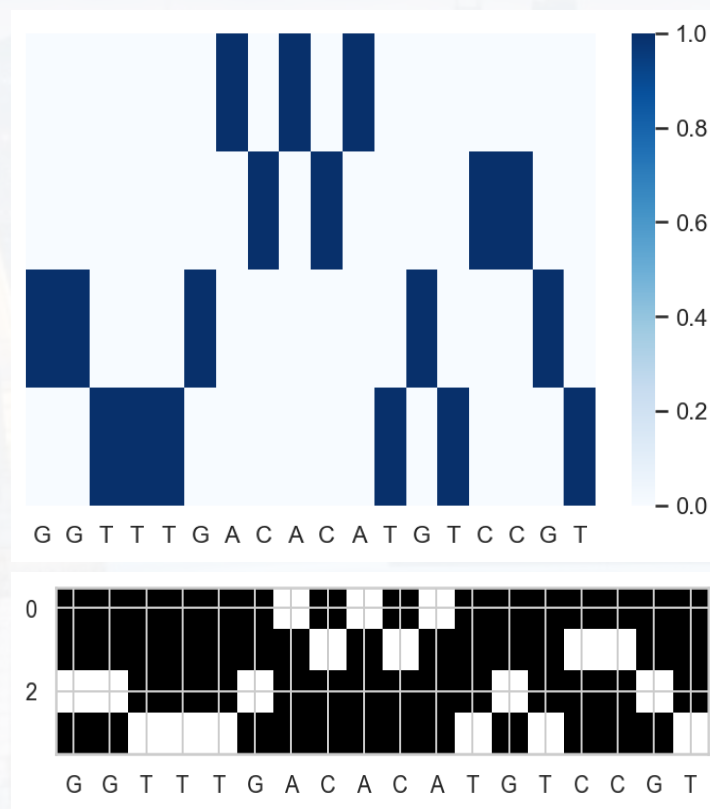
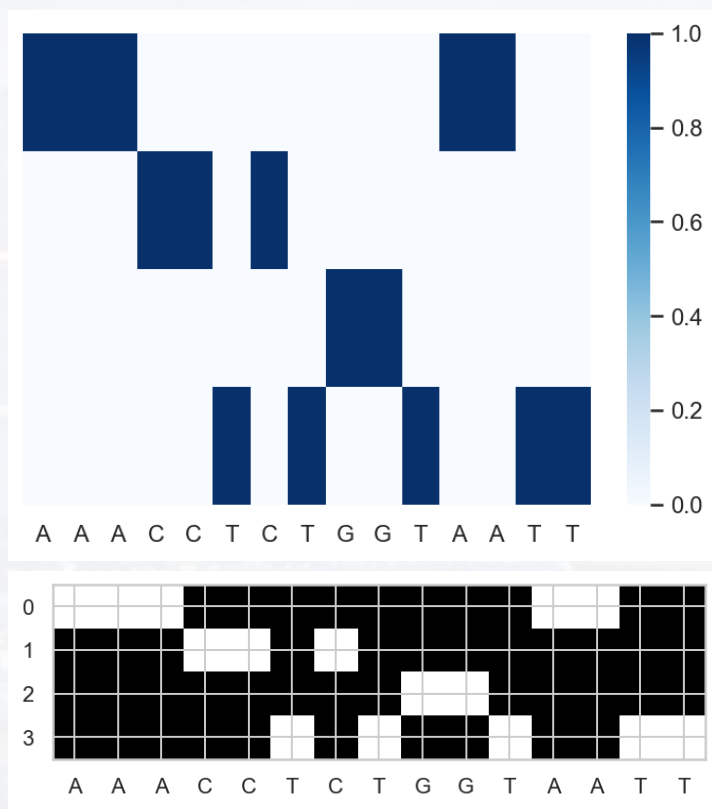




We can also loop over all sequences **and** plot routines via looping over the attributes using **getattr**

```
L = ['PlotMySeq1', 'PlotMySeq2']
```

```
[getattr(EncodeMySeq(s), M)() for s in S for M in L]
```







When you delete a block of code  
that you thought was useless



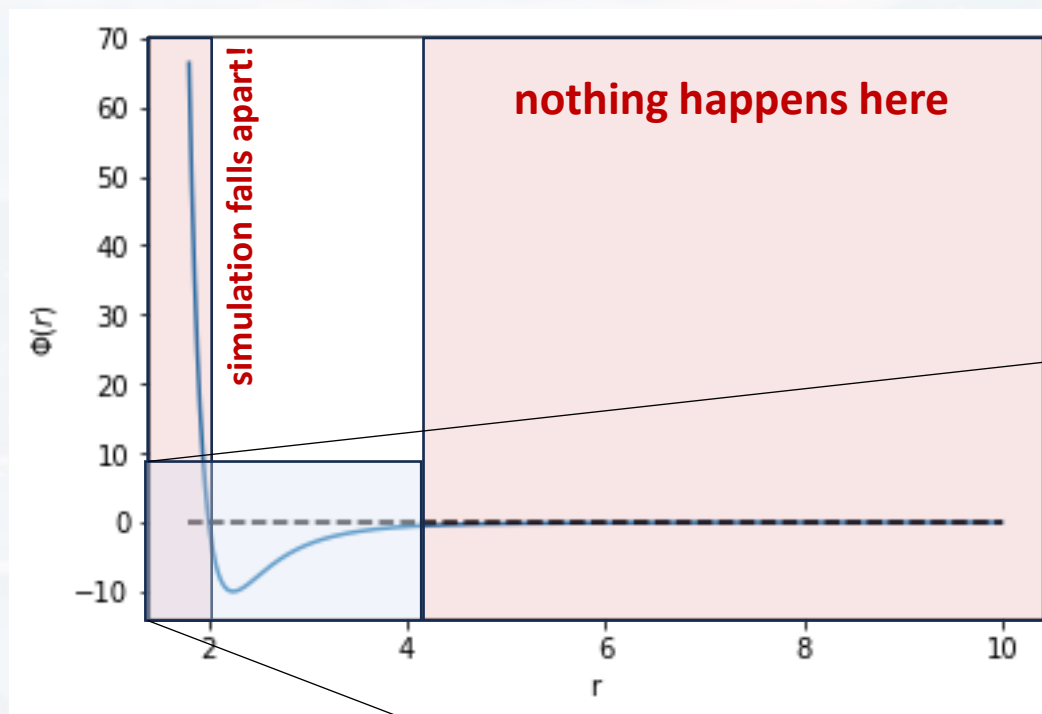
### Outline

- Overall Structure
- Classes
- **Let's build a Package!**

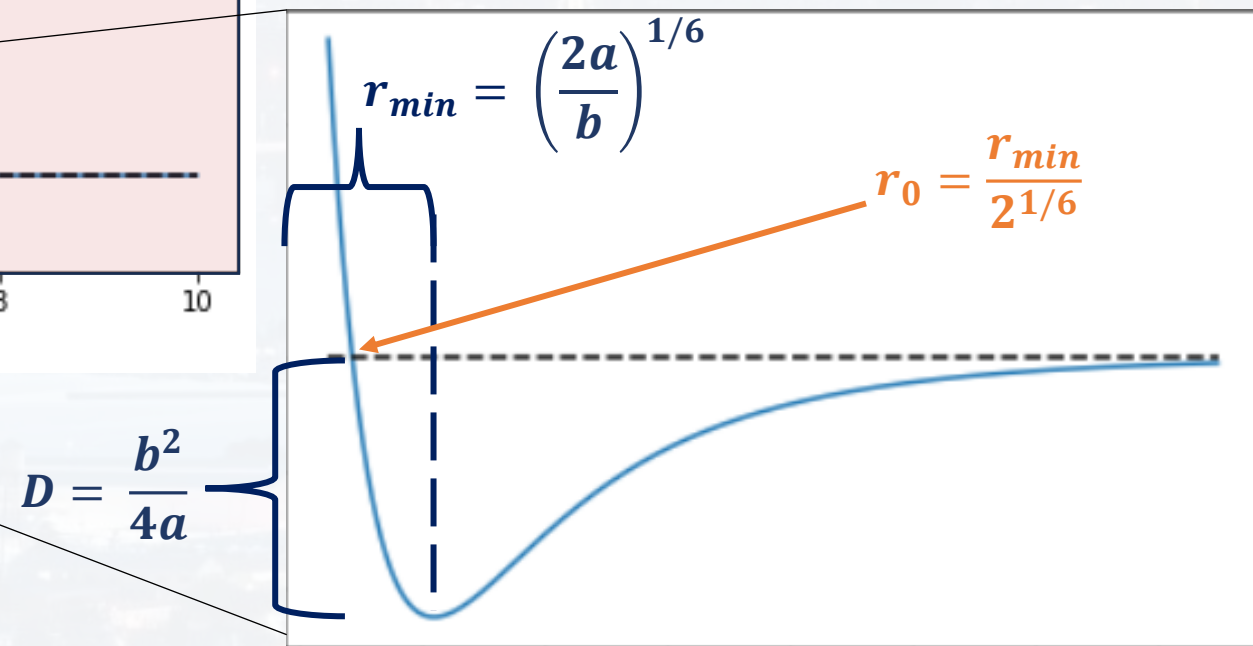


We already built a full package, but now let us repeat the process, but this time working towards the **next project**.

Goal: modelling the **motion of particles** in a **Lennard-Jones** potential  $U_{LJ}$

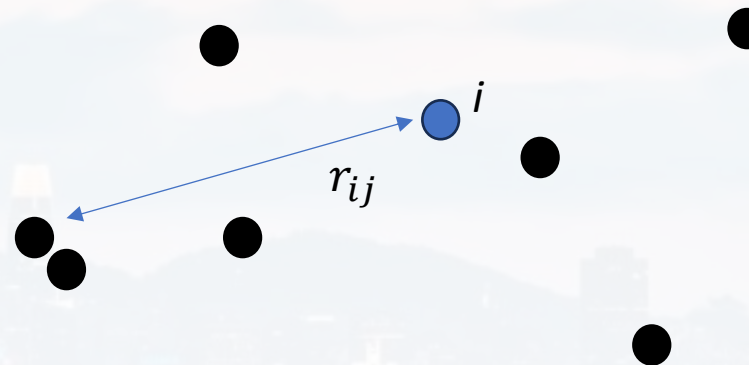
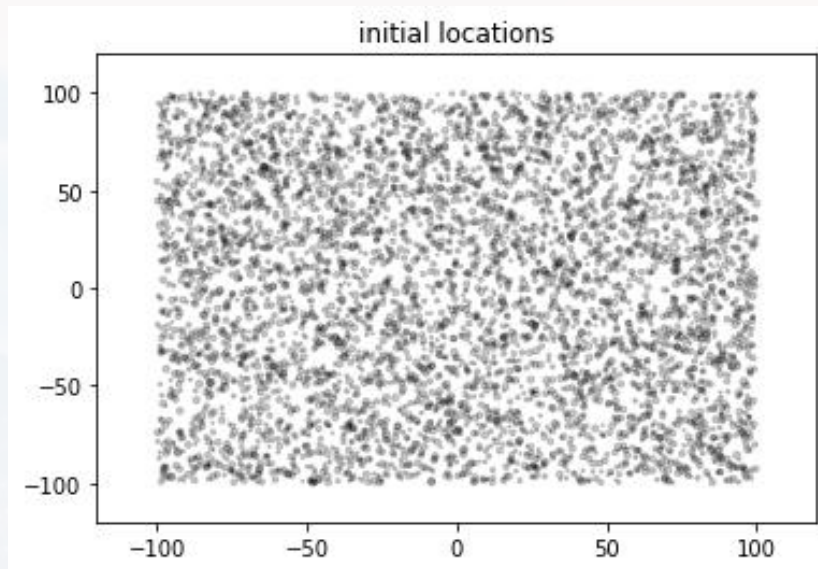


$$U_{LJ} = \frac{a}{r^{12}} - \frac{b}{r^6} \quad a, b = \text{const}$$

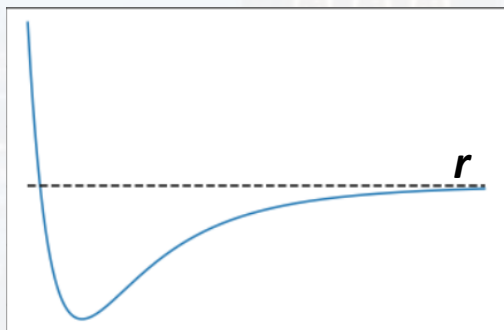




Goal: modelling the motion of particle in a **Lennard-Jones** potential  $U_{LJ}$



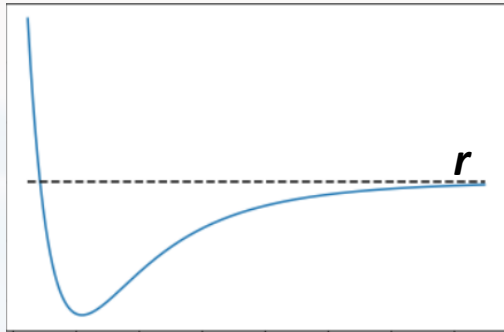
Each particle  $i$  feels the potential  $U_{tot}(r_{ij})$  as sum of all the repulsion/attraction to the other particles



$$U_{tot}(i) = \sum_j \frac{a}{r_{ij}^{12}} - \frac{b}{r_{ij}^6}$$

$$r_{ij} = \sqrt{(x_i - x_j)^2 + (y_i - y_j)^2}$$





$$U_{tot}(i) = \sum_j \frac{a}{r_{ij}^{12}} - \frac{b}{r_{ij}^6}$$

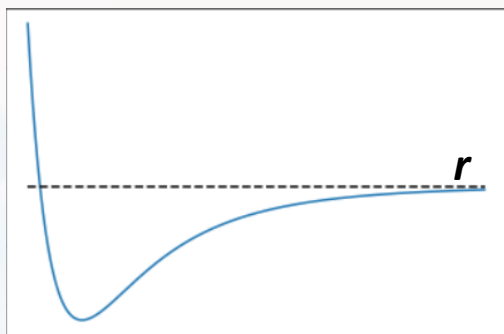
$$r_{ij} = \sqrt{(x_i - x_j)^2 + (y_i - y_j)^2}$$

We need to calculate the **distance** between **each** of the particles

L:        number of particles  
x:        vector with x coordinates  
y:        vector with y coordinates

```
X        = np.tile(x, (L, 1))  
Y        = np.tile(y, (L, 1))
```

```
Dx       = X - X.transpose()  
Dy       = Y - Y.transpose()
```



$$U_{tot}(i) = \sum_j \frac{a}{r_{ij}^{12}} - \frac{b}{r_{ij}^6}$$

from the **distances**, we can  
calculate the **total potential**

$$r_{ij} = \sqrt{(x_i - x_j)^2 + (y_i - y_j)^2}$$

We need to calculate the **distance**  
between **each** of the particles

```
def Potential(Dx, Dy, a, b):  
    r2 = Dx**2 + Dy**2  
  
    return a/(r2**6) - b/(r2**3)
```

This function calculates the potential for all  
distances, but **does not** sum up to  $U_{tot}(i)$  yet.

**your task:**

- 1) modify the function so that it calculates  $U_{tot}$
- 2) avoid  $U = \infty$  for  $r = 0$



```
def Potential(Dx, Dy, a, b):  
    r2 = Dx**2 + Dy**2  
  
    return a/(r2**6) - b/(r2**3)
```

This function calculates the potential for all distances, but **does not** sum up to  $U_{tot}(i)$  yet.

**your task:**

- 1) modify the function so that it calculates  $U_{tot}$
- 2) avoid  $U = \infty$  for  $r = 0$

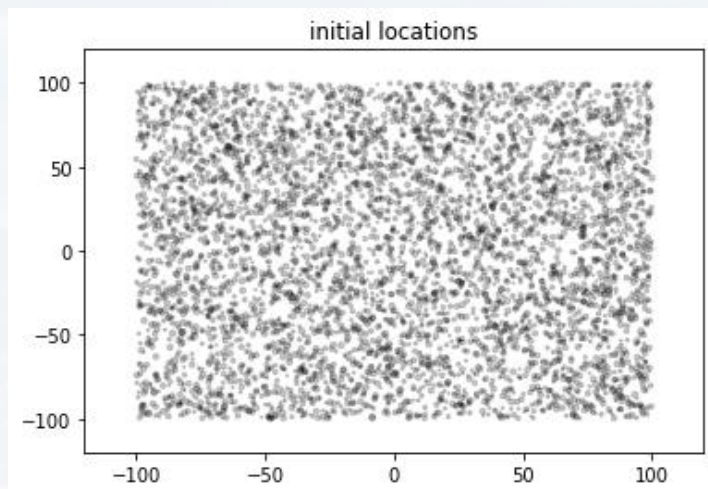
hint: **no loop is necessary** for any of these steps (use algebra with `np.eye`, `np.ones` and `np.dot`),  
but you can run a loop if the matrix multiplications are too tricky!





Goal: modelling the motion of particle in a **Lennard-Jones** potential  $U_{LJ}$

We also want to plot the locations of the particles for each time step → add a plotting routine!



```
def PlotLocations(x, y, size, Lim, title, marker = 'o', color = 'black'):  
    plt.scatter(x, y, size, marker = marker, color = color, edgecolor = None,\  
                alpha = 0.2)  
    plt.xlim([-Lim*1.2, Lim*1.2])  
    plt.ylim([-Lim*1.2, Lim*1.2])  
    plt.title(title)  
    plt.show()
```





How do the particles move?

**Metropolis:**

- 1) suggest a random move  $\Delta x$  and  $\Delta y$  for all particles
- 2) calculate  $\Delta U_{tot}(x, y)_{LJ}$  based on  $\Delta x$  and  $\Delta y$  for **each particle**
- 3) move or not:
  - a) move those particles where  $\Delta U_{tot}(x, y)_{LJ} < 0$
  - b) for those particles where  $\Delta U_{tot}(x, y)_{LJ} > 0$ 
    - draw a **random number**  $\rho$  from a **uniform distribution** in the interval **(0, 1)**
    - move those particles for which  $\rho < \exp \left[ -\frac{\Delta U_{tot}(x, y)_{LJ}}{T} \right]$
- 4) repeat

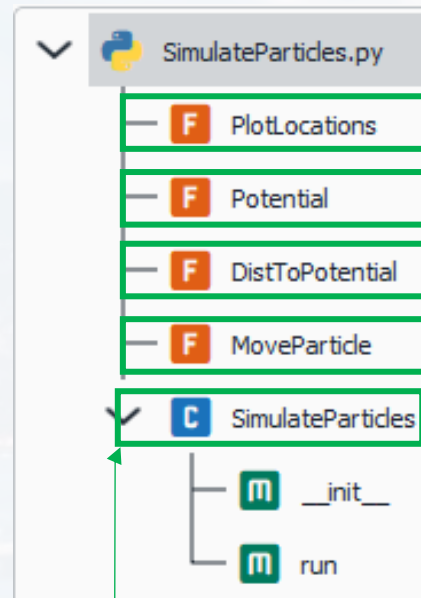
**Note:**

- 1) T is the temperature. Try  $T = 1$  first. Vary T later, once the code works
- 2) Don't worry too much about this algorithm. You will learn **more about it in Chem 273**



Goal: modelling the motion of particle in a **Lennard-Jones** potential  $U_{LJ}$

Your package could have this structure:



routine that plots **current locations** of the particles

Routine that **defines the potential**. Here: Lennard-Jones

routine that **calculates the distances for each particle and the total potential**, calls method Potential

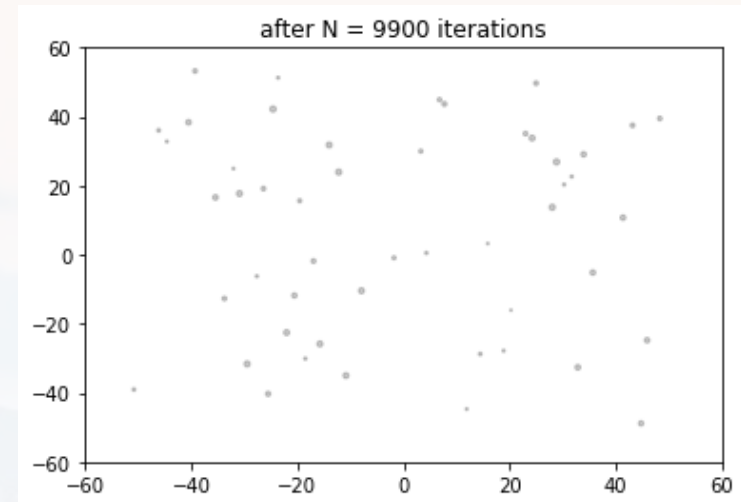
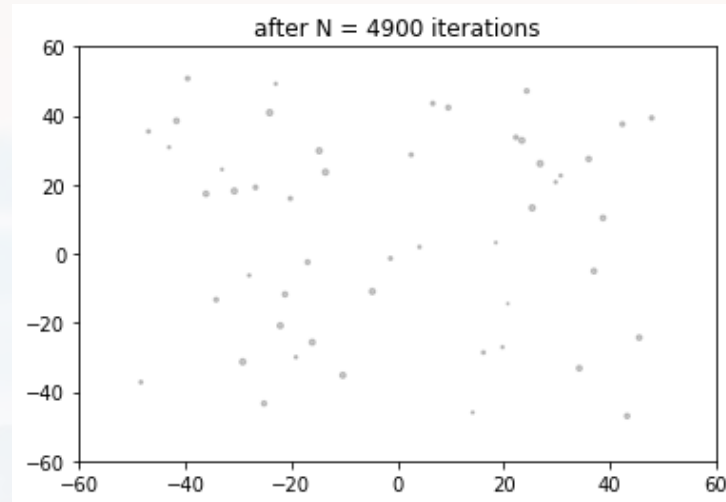
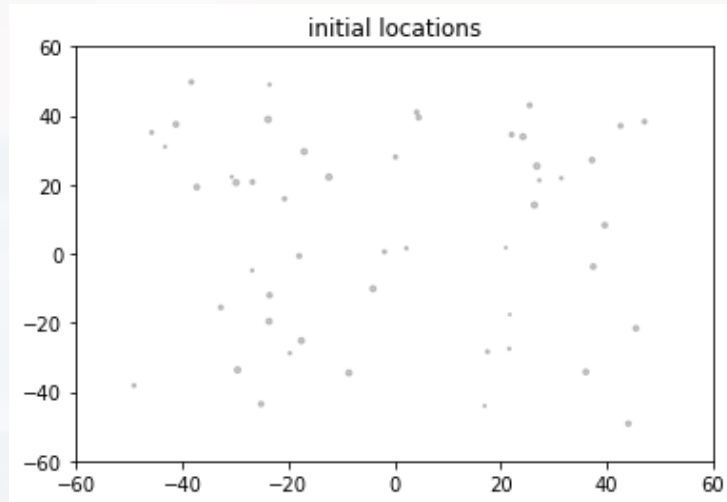
routine that calculates **if a particle is allowed to move** based on  $\rho < \exp \left[ -\frac{\Delta U_{tot}(x,y)_{LJ}}{T} \right]$

**class** that runs the simulation:

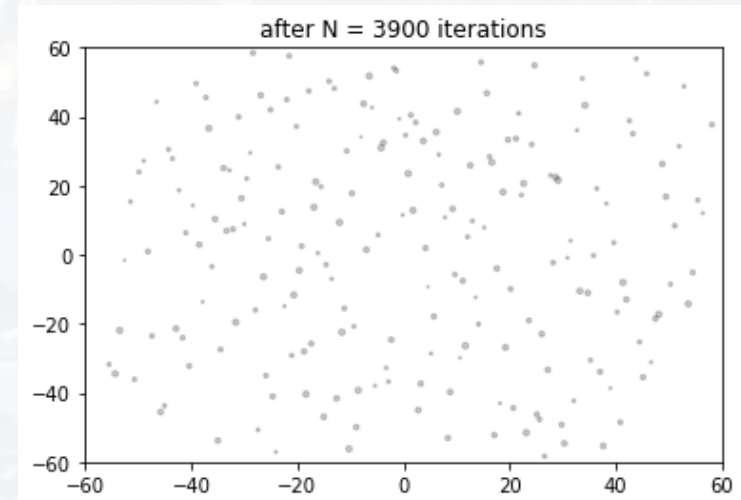
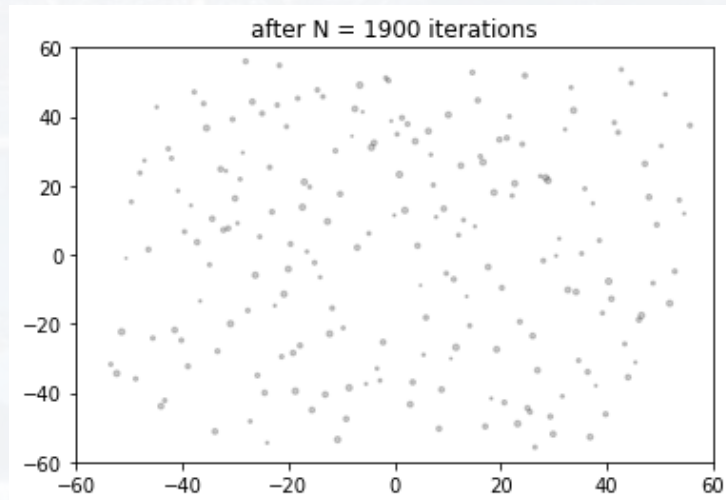
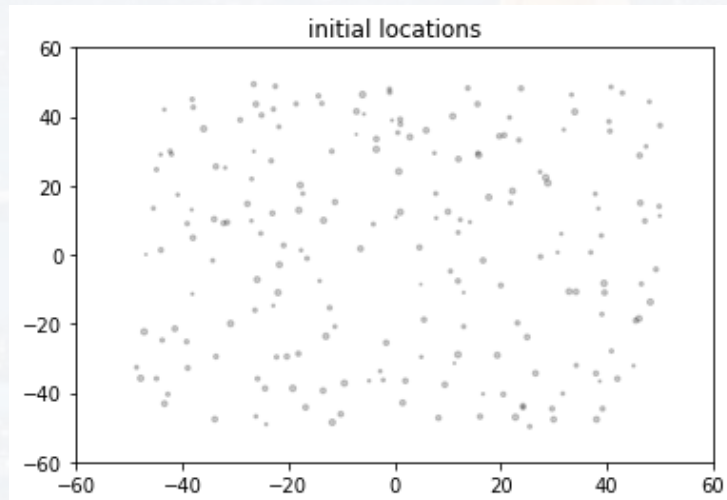
**\_\_init\_\_** takes **number of particles** and **generates initial vectors x and y** with random values  
**run** **runs** the actual **simulation** by calling the routines from above. It takes number of **iterations** (say 1000), values for **a** and **b** and **T** (defaults are 1).



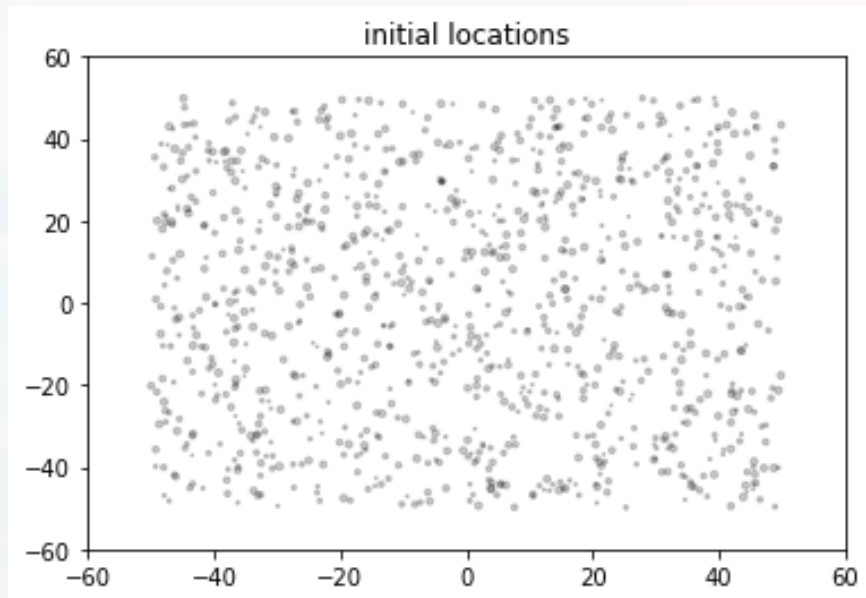
**N = 50**



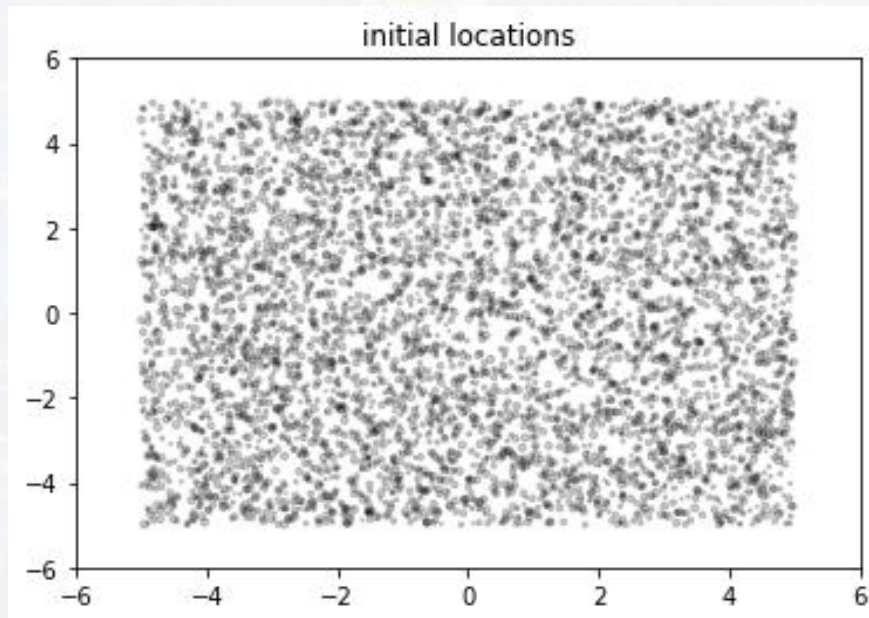
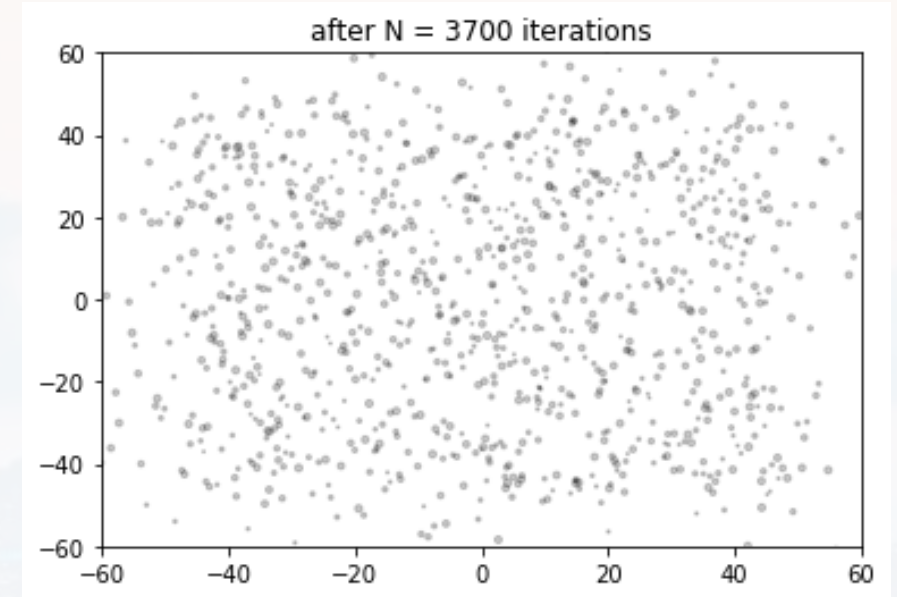
**N = 200**



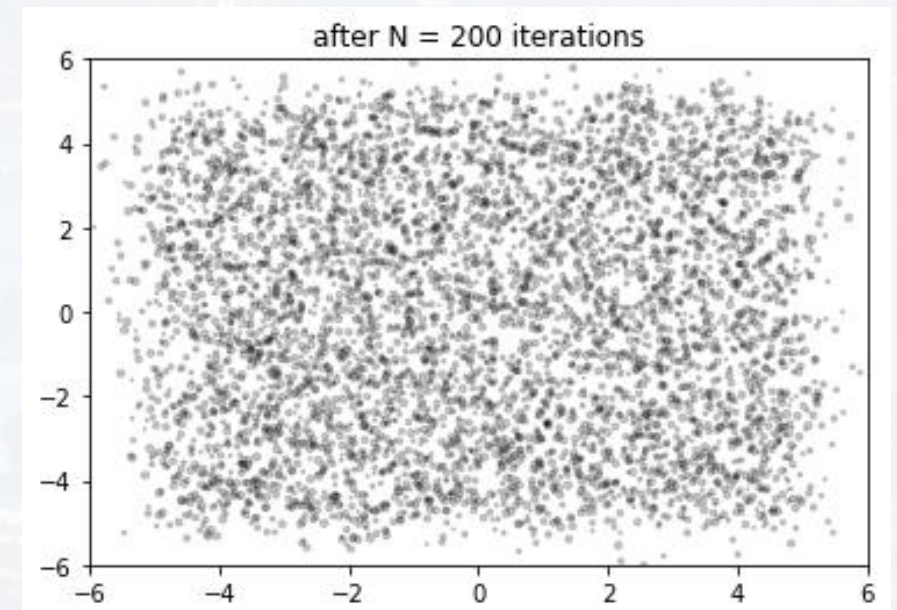




**N = 1000**



**N = 5000**







Thank you very much for your attention!

