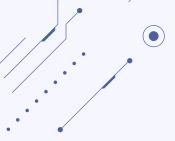
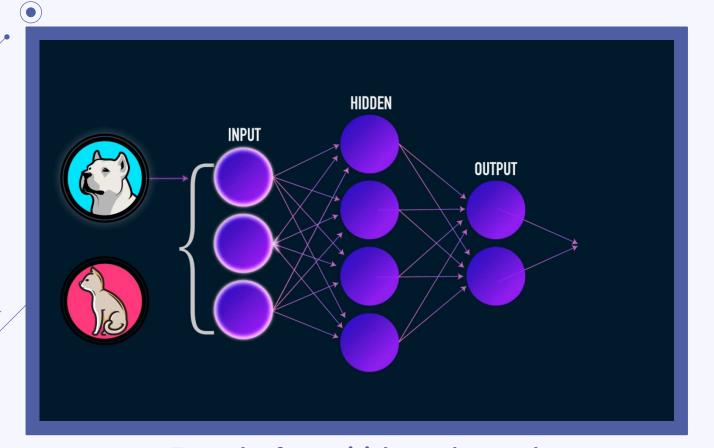
# Multi-layer Perceptron (MLP)

Feedforward Artificial Neural Network (ANN)



# **Summary**

- 1. Introduction
- 2. MLP implementation
  - 3. Results
  - 4. Conclusion



Example of an articial neural network

# **Multi-layer Perceptron**

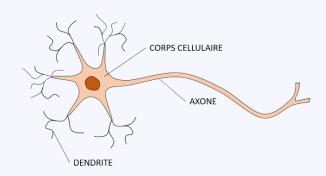


Image 1: Representation of a biological neurone

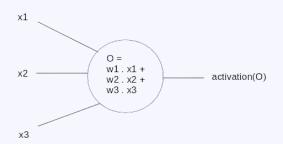


Image 2: Representation of an artificial neurone

An MLP is an Artificial Neural Networks (ANN)

Inspired by biological neural networks

## **Applications:**

Speech and image recognition, text processing...

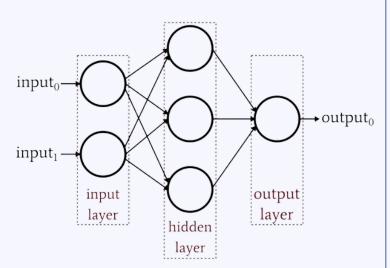
At least **3 layers** of nodes: an **input** layer, a **hidden** layer and a **output** layer.

Learns by **training** on a dataset

Image 1: <a href="https://svtdiderot.fr/cordewener/4eme-cordewener/le-systeme-nerveux/">https://svtdiderot.fr/cordewener/4eme-cordewener/le-systeme-nerveux/</a>

Image 2: https://cdancette.fr/assets/neuron.png

## **Multi-layer Perceptron**



An MLP is an Artificial Neural Networks (ANN)

Inspired by biological neural networks

**Applications:** 

Speech and image recognition, text processing...

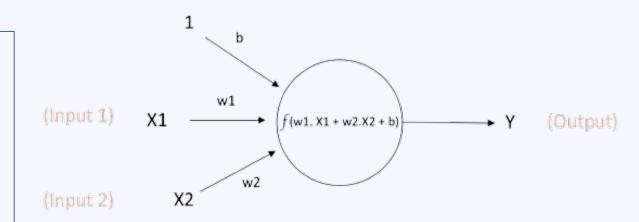
At least **3 layers of nodes**: an **input** layer, a **hidden** layer and a **output** layer.

Learns by **training** on a dataset

## A single neurone

#### Also called **node** or **unit**

- **Inputs**: X1, X2, Bias 1
- Weights: w1, w2, b
- Output: Y
- Activation function:
   non-linear function



Output of neuron = 
$$Y = f(w1. X1 + w2. X2 + b)$$

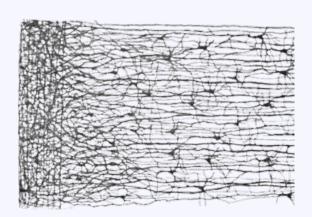


Image 4: Multiple layers in a biological neural network

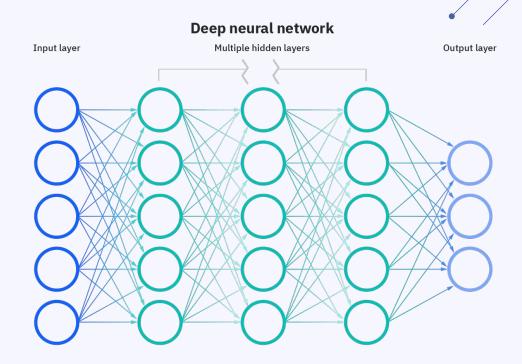
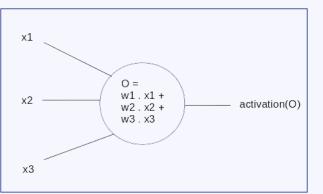
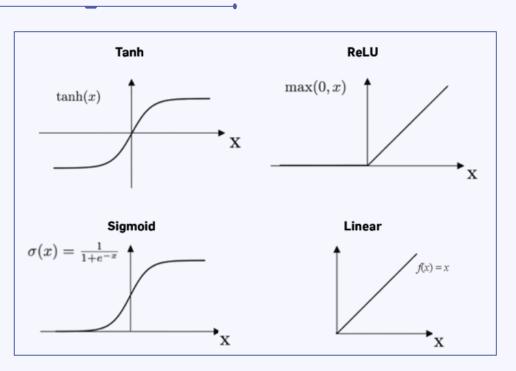


Image 5: Multiple layers in an artificial neural network

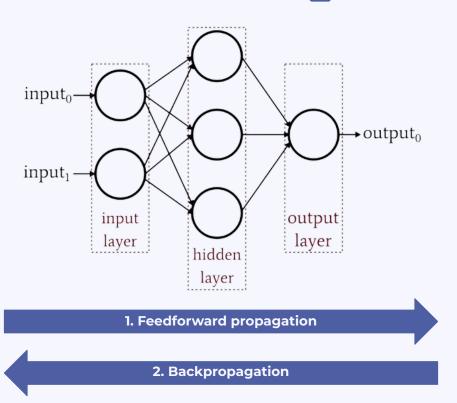








# **Feedforward and Backpropagation**



# Requirements



**Pam50 dataset**: It contains the expression of 50 known human genes with patterns according to a subtype of breast cancer.

# **Data importation**

Data Pam = pd.read csv("data/Data Pam50.csv")

	Unnamed: 0	subtype	ACTR3B	ANLN	BAG1	BCL2	BIRC5	BLVRA
0	Normal.Breast.10	Normal	-1.151	-3.736	0.260	1.300	-2.860	-0.569
1	Normal.Breast.2	Normal	-0.485	-3.739	0.591	1.580	-3.250	-0.533
2	Normal.Breast.3	Normal	0.298	-2.848	0.359	1.292	-2.493	-0.687
3	Normal.Breast.4.Custom	Normal	1.153	-4.717	0.098	1.954	-3.237	-0.535
4	Normal.Breast.7	Normal	-0.287	-3.681	0.441	1.911	-2.156	-0.965
	(***)		***	•••	***			
67	H1AUNC.1319.C	LumB	-0.267	-0.803	-0.409	0.329	0.179	1.542
68	H1AUNC.1323.C	LumB	-0.938	-1.999	-0.910	0.472	-1.564	0.352
69	H1AUNC.1462.C	LumB	-0.950	-1.757	0.250	0.099	-1.289	1.219
70	H1AUNC.1471.C	LumB	0.039	-1.937	0.095	0.137	-0.948	2.833
71	H1AUNC.1474.C	LumB	-1.441	-2.706	-1.142	0.393	-1.330	0.333

Slicing

0 Normal
1 Normal
2 Normal
3 Normal
4 Normal
67 LumB
68 LumB
69 LumB
70 LumB
71 LumB

Raw Labels

label\_raw = Data\_Pam.iloc[:,1]
data raw = Data\_Pam.iloc[:,2:]

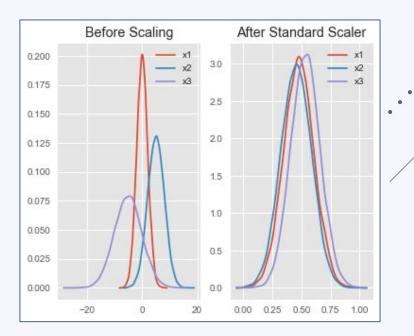
	ACTR3B	ANLN	BAG1	BCL2	BIRC5	BLVRA	CCNB1	CCI
0	-1.151	-3.736	0.260	1.300	-2.860	-0.569	-2.981	-1.
1	-0.485	-3.739	0.591	1.580	-3.250	-0.533	-2.935	-1.
2	0.298	-2.848	0.359	1.292	-2.493	-0.687	-2.810	-1.
3	1.153	-4.717	0.098	1.954	-3.237	-0.535	-3.558	-2.
4	-0.287	-3.681	0.441	1.911	-2.156	-0.965	-2.869	-1.
		2.2						
67	-0.267	-0.803	-0.409	0.329	0.179	1.542	-0.854	-1.
68	-0.938	-1.999	-0.910	0.472	-1.564	0.352	-1.842	-1.
69	-0.950	-1.757	0.250	0.099	-1.289	1.219	-1.135	-1.
70	0.039	-1.937	0.095	0.137	-0.948	2.833	-1.669	-0.
71	-1.441	-2.706	-1.142	0.393	-1.330	0.333	-1.415	-1.

Raw Data

Portion of the data frame

## **Data Scaling**

```
from sklearn.preprocessing import StandardScaler
sc_X = StandardScaler()
X_trainscaled=sc_X.fit_transform(X_train)
X_testscaled=sc_X.transform(X_test)
```



Example of scaling with StandardScaler

## **Cross-validation**

```
loo = LeaveOneOut()
for train, test in loo.split(data scaled):
    clf = MLPClassifier(hidden layer sizes=layer,
         random state=1, max iter=epochs, tol = tols).fit(data scaled[train],
                                                                   label target[train])
                                              total samples
                     iteration 1/N:
                                                                           test set
                                                                           train set
                     iteration 2/N:
                     iteration 3/N:
                     iteration N/N:
```

## Parameters and GridSearchCV

```
layer = [(100,1),(200,1),(100,10),(200,10),(100,20),(200,20),(50,10)]
tols = [10e-4, 10e-5, 10e-3, 10e-2, 10e-6, 10e-7]
epochs = [100,50,20,200,300]
```

```
clf = GridSearchCV(
    MLPClassifier(),
    param_grid= {
        'hidden_layer_sizes': layer,
        'tol': tols,
        'max_iter': epochs
        },
    refit='True',
    cv=2,
    n_jobs=-1,
)
```

## **MLPClassifier parameters**:

- hidden\_layer\_sizes
- max\_iter
- tol (tolerance)



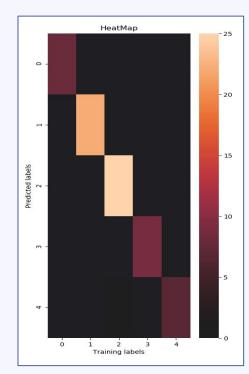
## **Heat Map**

## Repartition of the labels

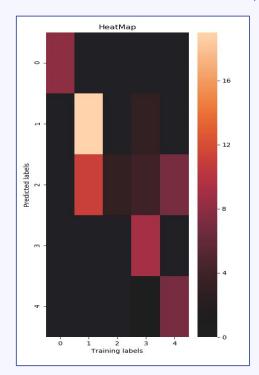
→ Each number is a category for a label

A Heat Map with high accuracy:

 First predicted labels should match with first true labels, etc.

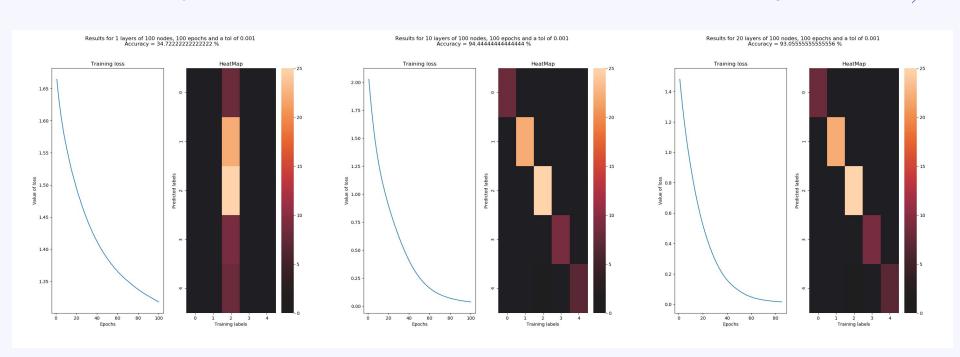


Heat map with high accuracy

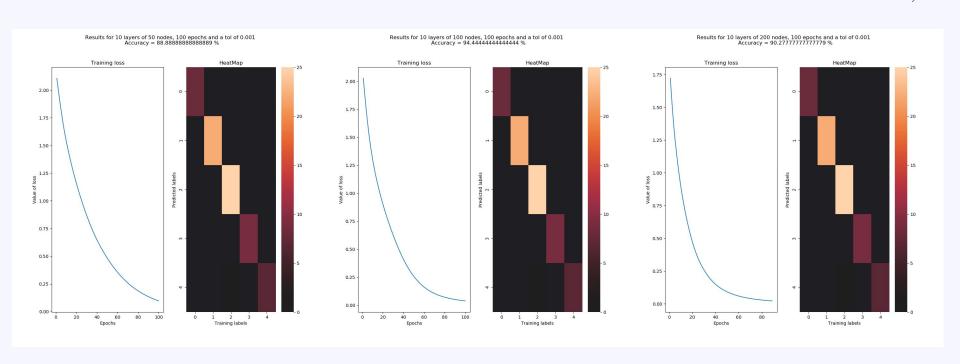


Heat map with low accuracy

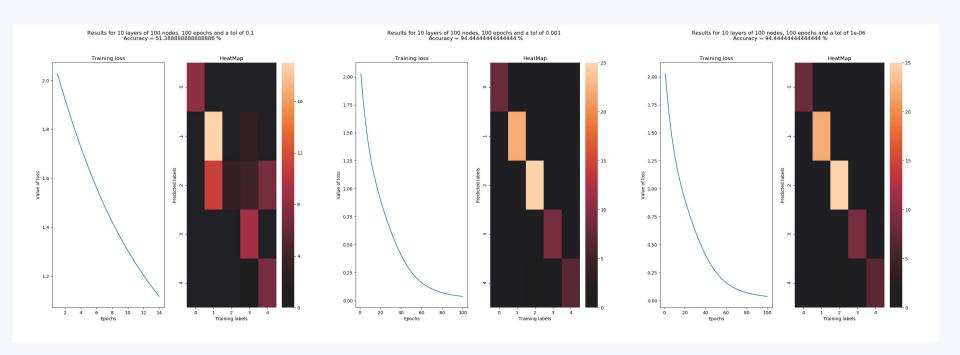
# 1<sup>st</sup> parameter : size of network



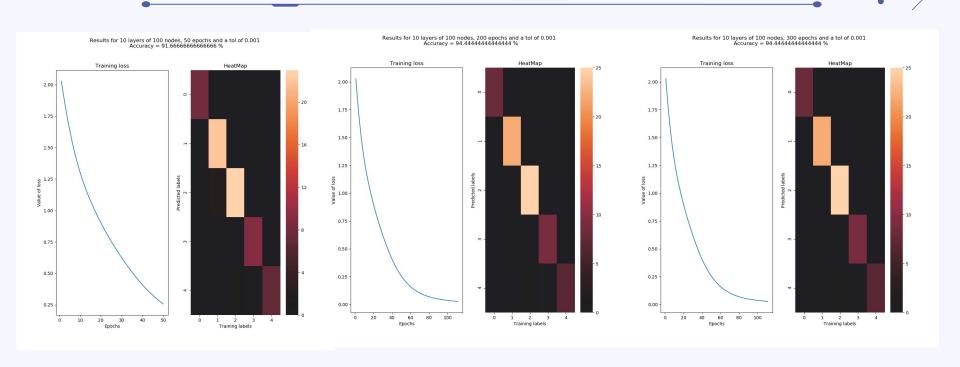
# 1<sup>st</sup> parameter: size of network



# 2<sup>nd</sup> parameter : tolerance

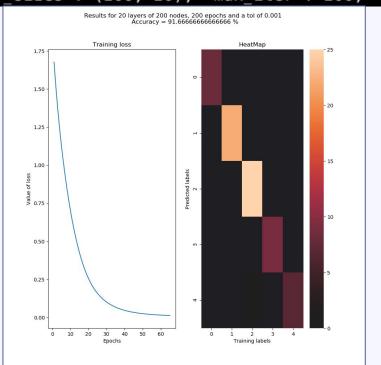


# 3<sup>rd</sup> parameter: number of epochs



## **GridSearchCV**







Scikit-learn documentation:

https://scikit-learn.org/stable/modules/neural\_networks\_supervised.html

https://scikit-learn.org/stable/modules/generated/sklearn.neural\_network.MLPClassifier.html

#### Other:

https://ujjwalkarn.me/2016/08/09/quick-intro-neural-networks/https://www.ibm.com/cloud/learn/neural-networks/https://en.wikipedia.org/wiki/Multilayer\_perceptron

## Git:

https://github.com/TexierLouis/Multilayer-perceptron