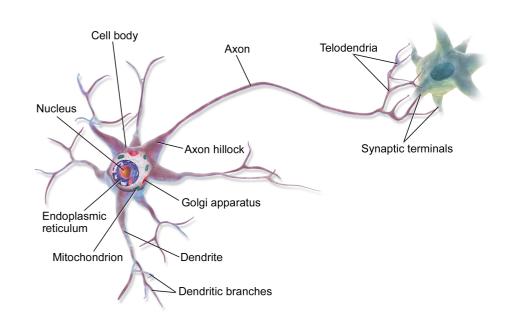
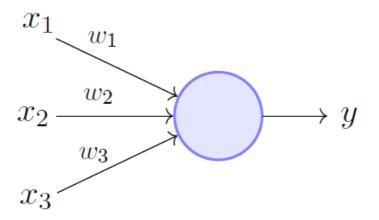
Preview: Neural Networking

Thomas Schwarz, SJ

- Neural networks are biology inspired
 - Instead of using Good Old Al (GOAl):
 - Try to build an artificial brain
- Brains are made up of dendrites / neurons
 - Have inputs that are activated by electricity
 - Have outputs that activate electricity

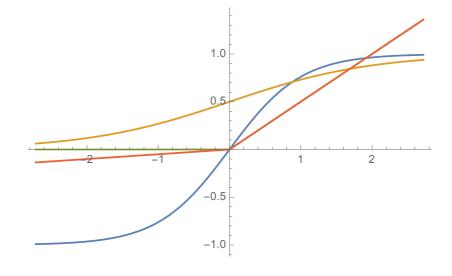


- Make a very simple model:
 - A perceptron is a unit that takes a number of inputs
 - Takes the <u>weighted sum</u> of the inputs
 - Subjects it to an activation function
 - Outputs the result $y = f(\sum_{i=1}^{n} w_i x_i)$

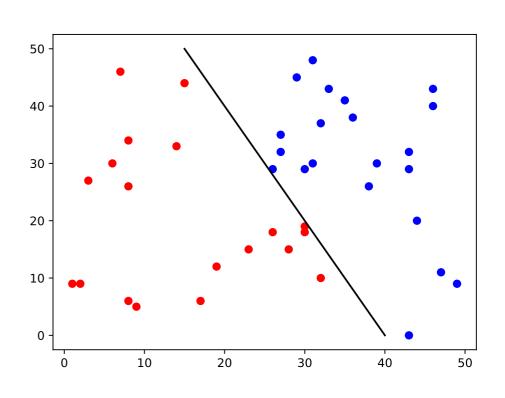


Perceptron Model (Minsky-Papert in 1969)

 Activation functions are various: sigmoid, tanh, step functions, ...



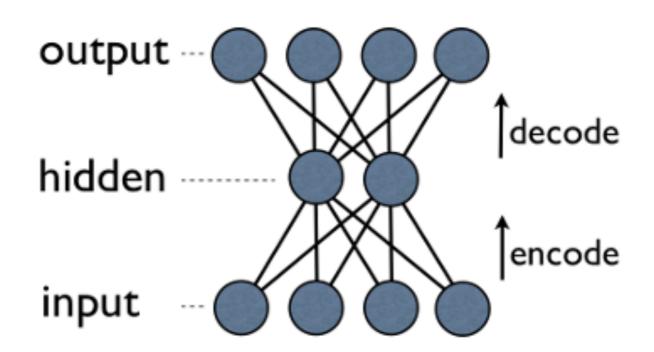
 A single perceptron can be made to classify data points along a hyper-plane



- Perceptrons:
 - 1943: McCulloch and Pitts model of a neuron, very much a perceptron with step-wise activation and equal weight
 - 1949: Hebb: Weights are different in nature
 - 1958: Rosenblatt: Perceptrons work as a generic tool
 - 1969: Minsky and Seymore: Perceptrons are really too limited

Neural Networks

- Create network of perceptrons = neural networks
 - 1989: Neural networks shown to be "universal approximators"
 - 1986: Hinton: Backpropagation learning algorithm



Neural Networks

- 1991: Hochreiter: Deep neural networks (with many different hidden layers) are difficult to train with backpropagation
 - Vanishing or exploding gradients
- 2000 2020: Lots of training data, Use of GPU (70 times faster), New types of networks, Better learning algorithms

Working with Neural Networks

- Several Python based packages for deep-learning
 - Keras
 - PyTorch
- Run best on GPU

- Pima Indian Data Set
 - Preparing the data:
 - Need to load data into a tensor
 - Need to separate training from testing

Imports first

```
from keras.models import Sequential, load_model
from keras.layers import Dense

import numpy as np
import os

np.random.seed(8)
```

Need to load data:

```
Pregnancies, Glucose, BloodPressure, SkinThickness, Insulin, BMI, DiabetesPedigree Function, Age, Outcome 6,148,72,35,0,33.6,0.627,50,1 1,85,66,29,0,26.6,0.351,31,0 8,183,64,0,0,23.3,0.672,32,1 1,89,66,23,94,28.1,0.167,21,0
```

Use Numpy's loadtxt

```
dataset = np.loadtxt("diabetes.csv", delimiter = ',',
    skiprows=1)
```

- Separate output from input parameters
 - Output is the last column
 - This is where slicing comes in

```
X = dataset[:,:-1]
Y = dataset[:,-1]
```

```
>>> X
             , 148.   ,   72.   ,   ...,   33.6   ,     0.627,   50.
array([[ 6.
       [ 1. , 85. , 66. , ..., 26.6 , 0.351, 31.
         8. , 183. , 64. , ..., 23.3 , 0.672,
                                                              ],
             , 121.     ,    72.     ,    ...,    26.2    ,     0.245,     30.
       [ 1. , 126. , 60. , ..., 30.1 , 0.349, 47.
                                                              ],
              , 93. , 70.
                             , ..., 30.4 , 0.315,
                                                        23.
                                                              ]])
>>> Y
array([1., 0., 1., 0., 1., 0., 1., 0., 1., 1., 0., 1., 0., 1., 1., 1., 1., 1.,
      1., 0., 1., 0., 0., 1., 1., 1., 1., 1., 0., 0., 0., 0., 1., 0., 0.,
      0., 0., 0., 1., 1., 1., 0., 0., 0., 1., 0., 1., 0., 0., 1., 0.,
      0., 0., 1., 0., 0., 1., 0., 0., 0., 0., 1., 0., 0., 1., 0., 1., 0.,
```

- Now separate into training and testing data, using slices
 - There are 768 records, we use the last 68 as test:

```
Xtrain = X[:700,:]
Ytrain = Y[:700]
Xtest = X[700:,:]
Ytest = Y[700:]
```

- Now use Keras magic:
 - Define a neural network with
 - an initial layer of 8 neurons,
 - a first layer with 16 neurons
 - a second layer with 8 neurons
 - an output layer with 1 neuron

Sequential models are easiest

```
model=Sequential()
model.add(Dense(16, input_dim=8, activation = 'relu'))
model.add(Dense(8, activation='relu'))
model.add(Dense(1, activation='sigmoid'))
```

Now train

```
model.compile(loss='binary_crossentropy',
  optimizer='adam', metrics=['accuracy'])

model.fit(Xtrain, Ytrain, epochs = 150, batch_size=10,
  verbose=0)
```

• Get accuracy: (Usually around 70% on test data)

```
score=model.evaluate(Xtest, Ytest, verbose=0)
print(f'{model.metrics_names[1]} {score[1]}')
```