Note: This is the summary note from Udacity Introduction to Deep Learning with PyTorch

Neural Network

- Find the boundary that separate two / more data set
- Complicated data -> need complicate algorithm
- · Classify problem

Classification

- · Classify problem
- How do we find the line / boundary between two data set

Linear Boundary

• Linear boundary -- Linear equation

Equation: $w_1x_1 + w_2x_2 + b = 0$

Vector notation: WX + b = 0

Weight: $W(w_1, w_2)$

Input: $X(x_1, x_2)$

Label: y = 0 or 1

$$prediction: \ \hat{y} = \begin{cases} 1 \ if \ WX + b \ge 0 \\ 0 \ if \ WX + b < 0 \end{cases}$$

 Mission: have Y-hat resemble to Y as close as possible in order to create linear boundary

Higher Dimension

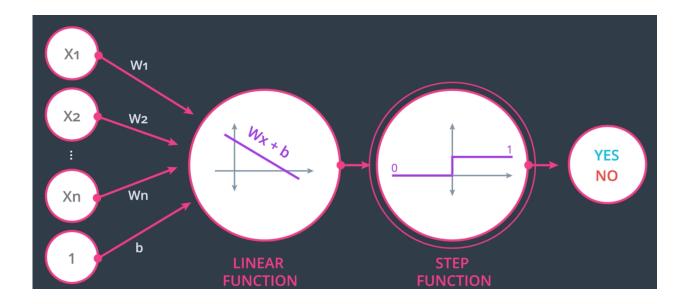
· Focus on what if we have more features / inputs

$$w_1 x_1 + w_2 x_2 + \dots + w_n x_n + b = 0$$

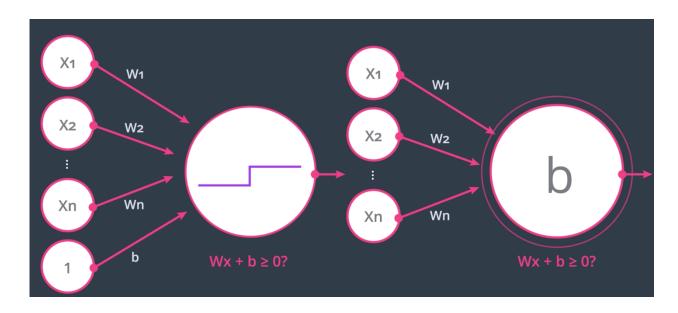
Conclude: still same vector notation and prediction as in linear boundary

Perceptron

- Has features similar to human brain
- Building block of neural network ---> encoding the linear boundary equation into small graph
- Step function (can be discrete or continuous or some other functions) that's why it should live in a separate node



Two ways represent precentron



Perceptron as Logical Operator

- Show application of perceptron as Logical operator
- https://medium.com/@stanleydukor/neural-representation-ofand-or-not-xor-and-xnor-logic-gates-perceptron-algorithmb0275375fea1

Perceptron trick

- How we split the data?
 - Misclassified points wants the line close to them
 - So move the line to the misclassified points
- Point A(x, y), Line: $w_1x_1 + w_2x_2 + b = 0$, Learning Rate = 0.1: this is used so that the line doesn't move drastically over already correct classified points, so move little by little
 - If A in positive area want Line to come closer, we do subtraction
 - $d = w_1 (x * 0.1)$
 - $e = w_2 (y * 0.1)$
 - f = b (1 * 0.1) (1 is added for bias unit)
 - New Line: $dx_1 + ex_2 + f = 0$
 - If A in negative area want Line to come closer, we do addition
 - $d = w_1 + (x * 0.1)$
 - $e = w_2 + (y * 0.1)$
 - f = b + (1 * 0.1) (1 is added for bias unit)
 - New Line: $dx_1 + ex_2 + f = 0$

Perceptron algorithm

Algorithm

For a point with coordinates (p, q), label y, learning rate α and prediction given by the equation $\hat{y} = step(w_1x_1 + w_2x_2 + b)$:

- If the point is correctly classified, do nothing.
- If the point is classified positive, but it has a negative label:
 - $\alpha p w_1$
 - $\alpha q w_2$
 - $(\alpha * 1) b$
- If the point is classified negative, but it has a positive label:
 - $\alpha p + w_1$
 - $\alpha q + w_2$
 - $(\alpha * 1) + b$

Note:

- Step is the step function in this case the discrete step function that return either 1 or 0
- we do this until we have minimum classified points or until a given threshold given

Mini summary:

- We know what is the problem that we try to solve: classification (Linear data)
- We know what the perceptron is = single layer neural network
- Why we use perceptron?
- Solution (The algorithm that create linear boundary between points)
- · We can split two points with Perceptron algorithm
