FEED FORWARD NEURAL NETWORK



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Study map



- 1.Basic programming
 - R-programming
- 2. Perceptron
 - Activity function
- 3. Feed Forward NN
 - Logistic function
- 4. Multi-layer Perceptron
 - XOR gate
- 5. Deep Feed Forward

6. Apply DFF

7. Recurrent Neural Network

8. Apply RNN

• 9. Long / Short Term Memory

10. Apply LSTM

11. Evolutionary Computation

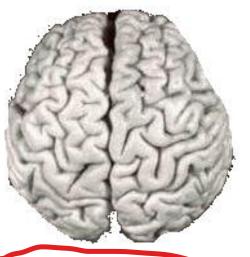
Learning outcome

- To known ANN concepts
- To understand activation functions
- To have experience using neuralnet() implementing AND/OR gate

Topics

- ANN concepts and overview
- Neurons, perceptron, and multilayered neural networks
- Bias, weights, activation functions, and hidden layers
- Forward and backpropagation methods
- neuralnet()

Overview



The brain is importance organ of the human body.

Weighting only 1.5Kilos

86 billion neurons

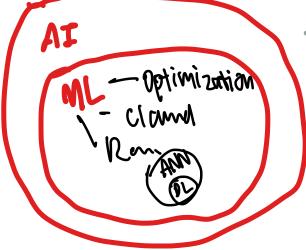


Artificial Neural Network (ANN) is the thought the mimicking the function as the human brain

Artificial Intelligence (AI)

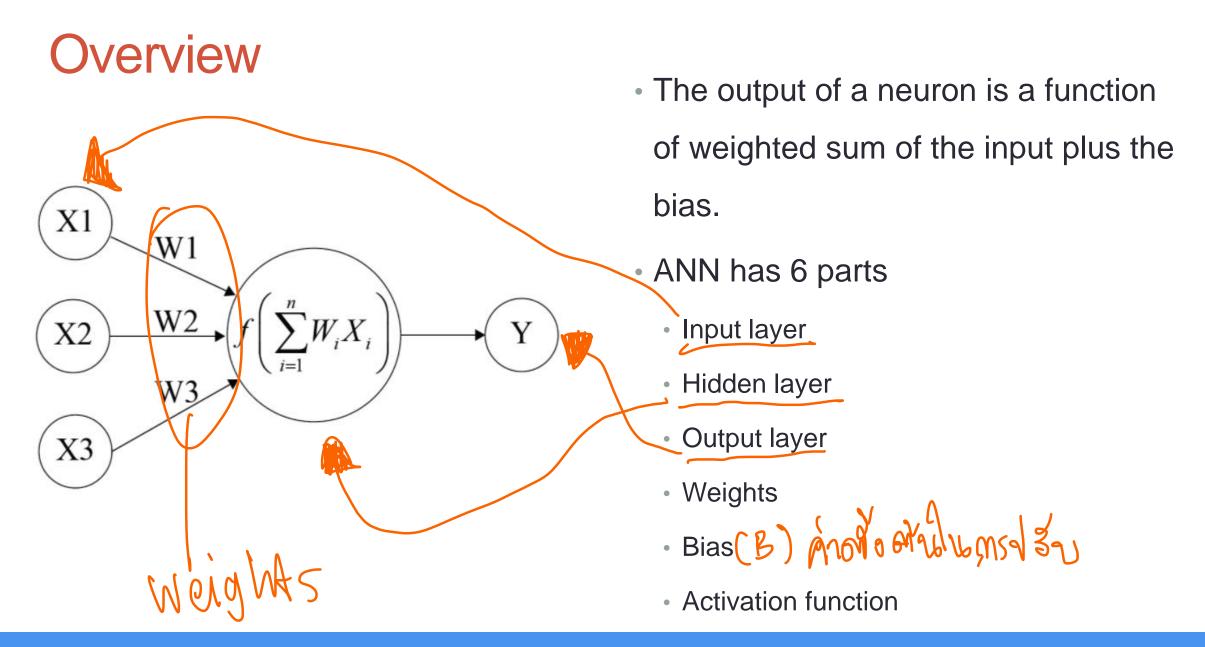
Machine Learning (ML)

Deep learning (DL)



Machine learning (ML) is a field of study giving the computer power to simulate the human intelligent

<u>Yor มานาว</u> <u>Deep Learning (DL)</u> is complex set of neural networks with more layers of processing, trying to work in image recognition, image classification, etc.

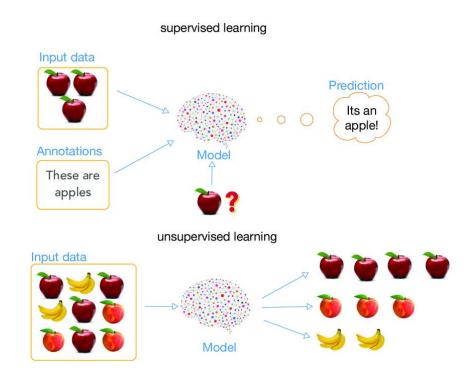


Overview

- Training in Neural Networks
 - Supervised learning is the user that supply
 - Inputs
 - The desired output
 - Unsupervised learning is the user that supply



 The system identifies the pattern and differences in the input without any external assistance.



Activation function

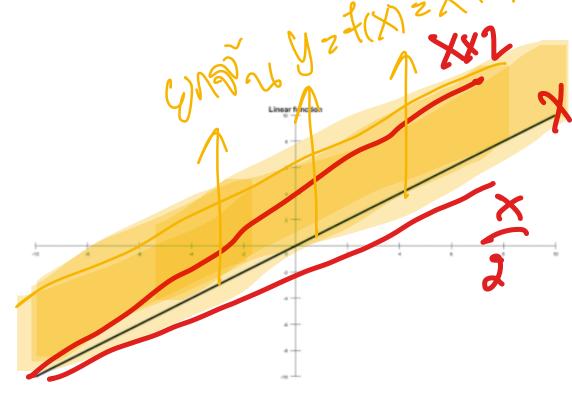
- Activation function is a mathematical function that converts the long-range value of input to the output in a small range.
- Many activation function in NN
 - Linear function
 - Unit step activation function (binary function)
 - Sigmoid function
 - Hyperbolic tangent
 - Rectified Linear Unit

Activation function: Linear function

• Output = Input

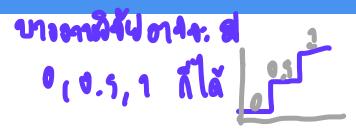
• The output can be gain or lose the value from the input.

- Output = Input / 2 11 1106

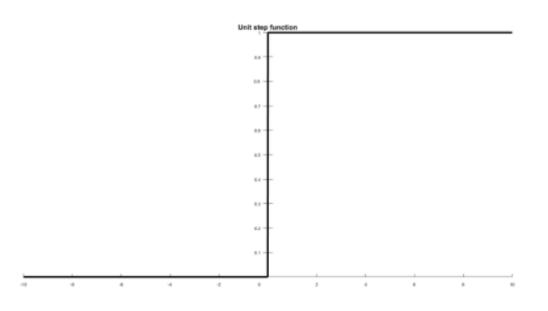


They be gain to the first stope
$$y = f(x) = x$$

Activation function: Unit step function



- The output is divided small levels such as
 - 1 / 0
 - True / False
 - High / Mid / Low

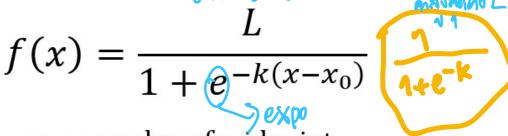


$$f(x) = o \text{ when } x < o,$$

$$1 \text{ when } x >= o$$

Activation function: Sigmoid function symbols & Sigmoid function symbols & Sigmoid function

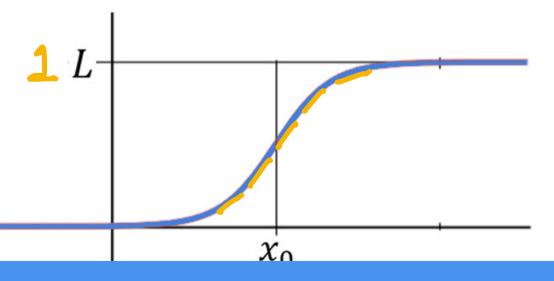
- Characteristic S-shaped curve,
 S-curve
- The class of sigmoid function
 - Logistic function
 - Hyperbolic tangent
 - Softmax function



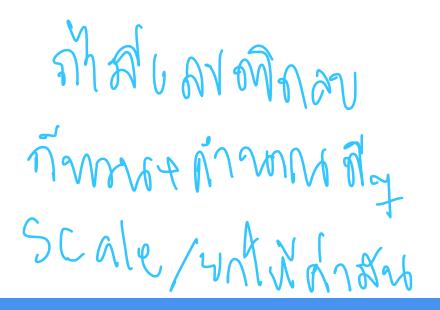
 $x_0 = x$ value of midpoint

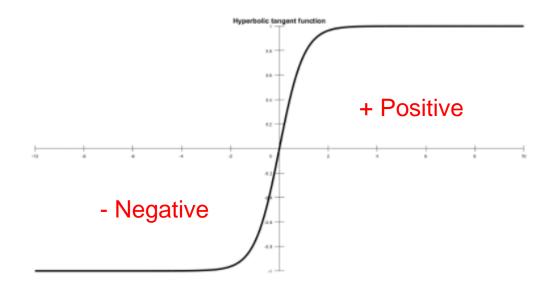
L = maximum value

k = growth rate



 S-curve is in the positive and negative part.

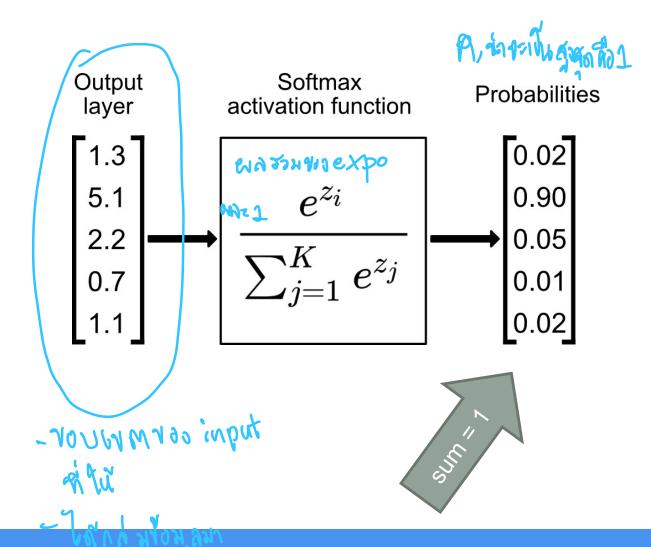




$$f(x) = \tanh(x)$$

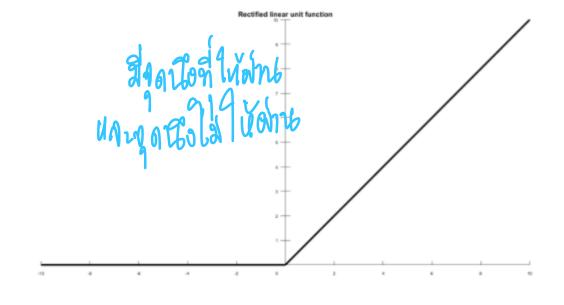
Activation function: Softmax function

 The output has from the normalization of the input data



Activation function: Rectified Linear Unit

- ReLU activation function
- The pass-through input to the output when the input value is in the positive part.
- The output can be gain or lose the value from the input.



$$f(x) = o \text{ when } x < o,$$

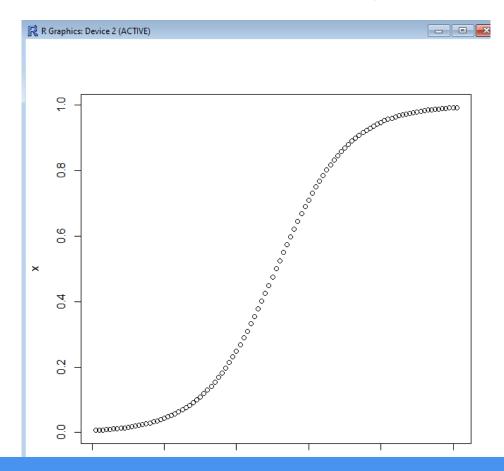
 $x \text{ when } x >= o$



Activity 3.1 Test activation function

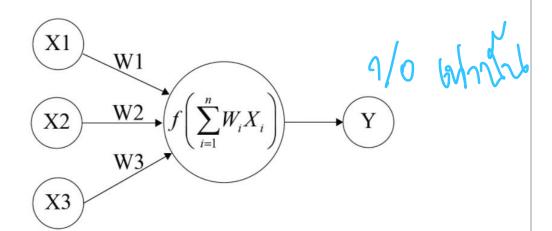
Given student plot output of four activation functions, by running R code below.

```
logistic = function(x)
  1/(1 + \exp(-x))
x = seq(-5,5,0.1)
maxx = length(x)
y = 1:maxx
x = logistic(x)
nlot(v x)
```

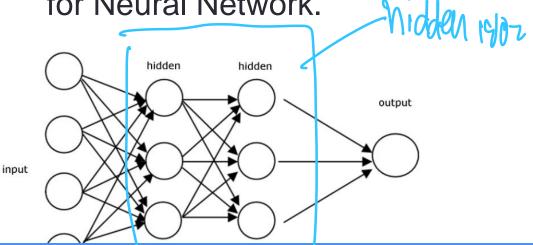


Perceptron vs Multi-Layer Perceptron

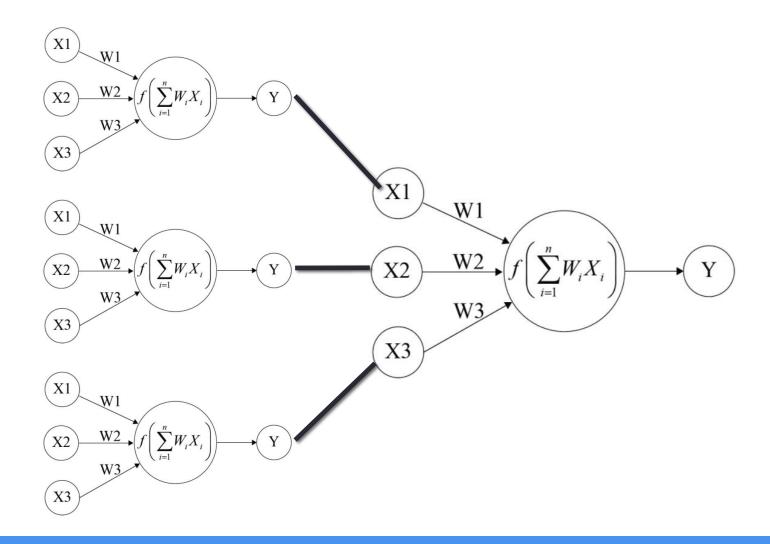
- Perceptron
 - A single neuron
 - Set of input is two categories
 - Usually, it uses step function



- Multi-Layer Perceptron
 - Combined many layer of the perceptron
 - Most of widely used architecture for Neural Network.



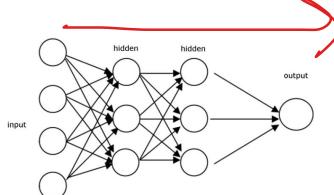
A large model is created from small models



Feed-Forward and Feed-Back networks

Feed-forward neural network has

- One direction
- ชิ ฟพงเลีย
- Input signals are feed to the input layer.
- Most of feed-forward network
 - MLP
 - Radial basic NN

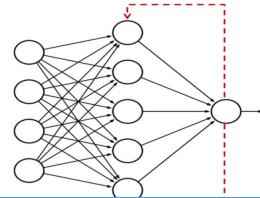




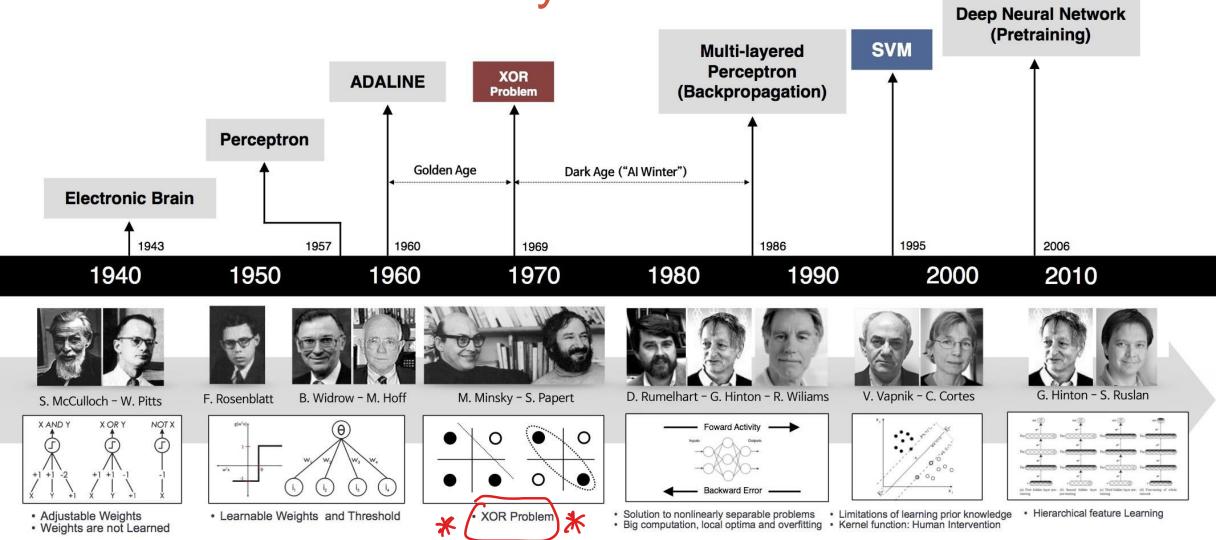
Feedback networks

- A neural or layer has received the processed signal from the next.
- Most of feedback network
 - Elman

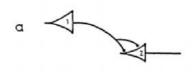
Hopfield

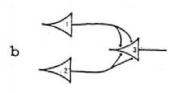


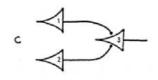
Neural Network History

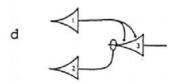


1940, Electronic Brain









EXPRESSION FOR THE FIGURES

- a $N_2(t) = N_1(t-1)$
- b $N_3(t) : \equiv N_1(t-1) \vee N_2(t-1)$
- c $N_3(t) = N_1(t-1) N_2(t-1)$
- d $N_3(t) = .N_1(t-1) \sim N_2(t-1)$

- In 1943, McCulloch and Pitts tried to understand how the brain could produce highly complex patterns by using many basic cells that are connected.
- These basic brain cells are called neurons.
- The McCulloch and Pitts model of a neuron, called an MCP neuron

- Research Questions/Answers
 - Adjustable weights
 - Weights are not learned



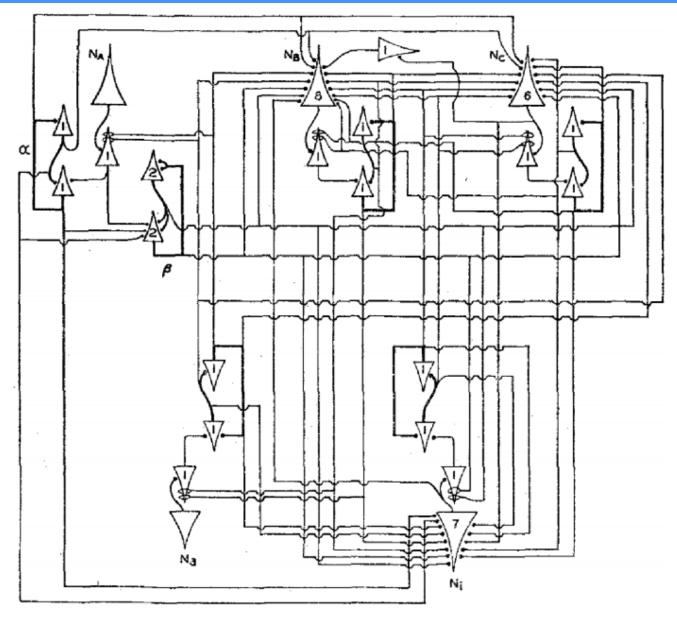
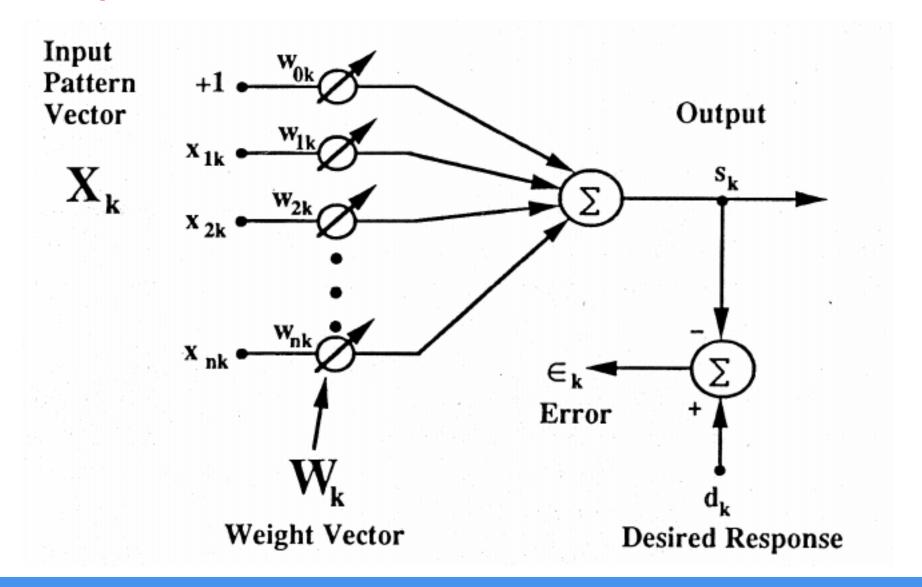


Fig. 7. Nicolas Rashevsky's complex neural network example (from [46, p. 32])

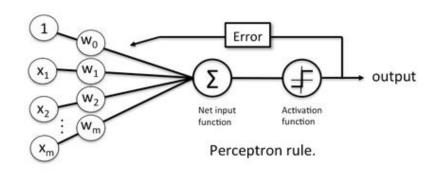
1950, Perceptron

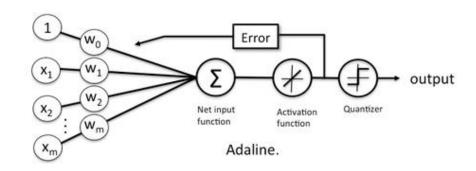


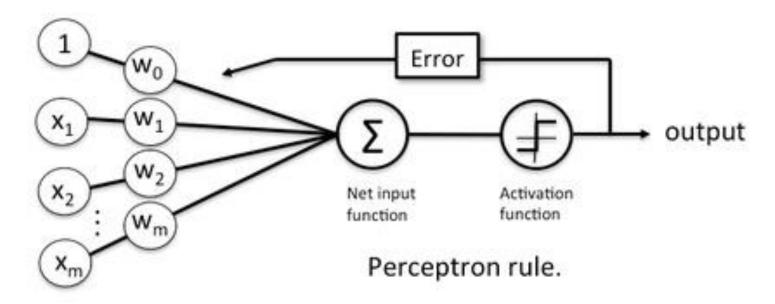
History: 1960s, ADALINE

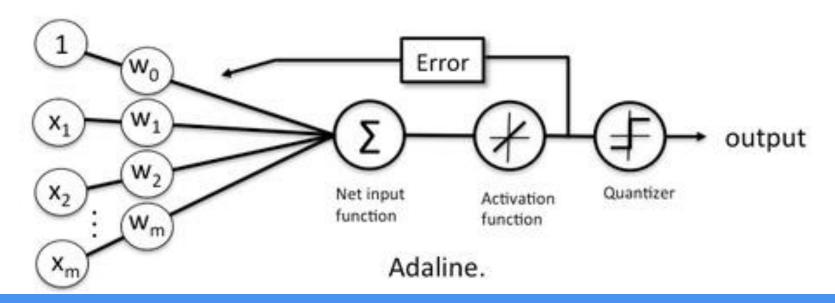
- ADALINE is aberration from ADAptive Linear Neural.
- In the past, the computer is different from today.
- Main problem is how to adjust weight of the perceptron.
- ADALINE uses a feedback after the activation function and before a quantize function comparing for weights adjustment

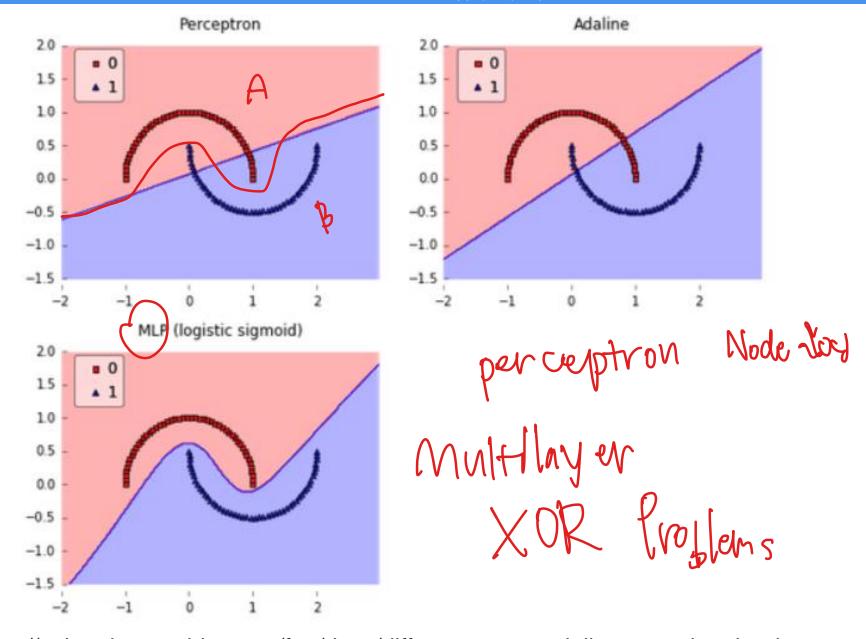












Library nerualnet()

Package	neuralnet
Description	Training of NN using backpropagation Allow user adjust activation function,

Usage

o neuralnet(formula, data, hidden = 1, threshold = 0.01, stepmax = 1e+05, act.fct = "logistic", rep=1 ...)

a ctivation function

Installation neuralnet()

ATUAASOUM ASOURS

There are command steps in R

- 1. install.packages("neuralnet") #Install "nerualnet" in the system.
- 2. library("neuralnet")
- 3. ??neuralnet

#call "neuralnet"

```
> install.packages("neuralnet")
Warning: package 'neuralnet' is in use and will not be installed
> update.packages("neuralnet")
> library("neuralnet")
```

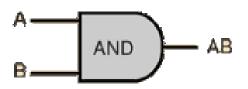
> ??neuralnet

Help pages:

neuralnet::confidence.interval
neuralnet::gwplor
neuralnet::neuralnet-package
neuralnet::predict.nn
fir Mutnimmial Log-linear Models
Evaluates Hessian for a Neural Network
fir Neural Networks
nnet::nnet fir Neural Networks
nnet::predict.nnet fir Neural Networks
nnet::nnet fir Neural Networks
nnet::predict.nnet fir Neural Networks

Activity 3.2 Train AND gate with neuralnet()





Α	В	Out
0	0	0
0	1	0
1	0	0
1	1	1

• AND = c(0,0,0,1)

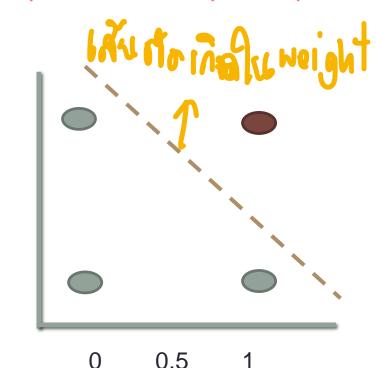
R Console

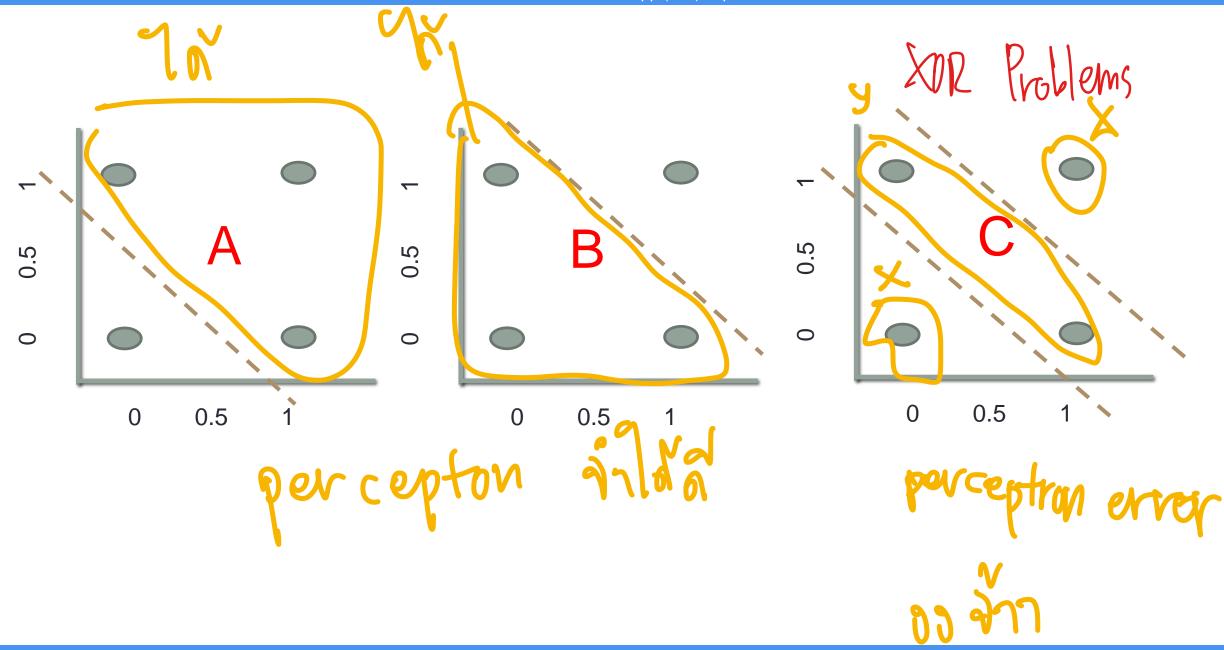
- > AND = c(0,0,0,1)
- > truthtable = expand.grid(c(0,1), c(0,1))

0

- > AND.data <- data.frame(truthtable, AND)
- > print(AND.data)

- truthtable = expand.grid(c(0,1), c(0,1))
- AND.data <- data.frame(truthtable, AND)
- print(AND.data)





Activity 3.2(2) Train AND gate with neuralnet()

net <- neuralnet(AND~Var1+Var2, AND.data, hidden=0, rep=1)

```
print(net)
```

```
> print(net)
neuralnet(formula = AND ~
    rep = 1)
$response
  AND
    0
    0
    0
$covariate
     Var1 Var2
[1,1]
[2,]
[3,]
[4,]
$model.list
$model.list$response
[1] "AND"
```

```
$model.list$variables
[1] "Var1" "Var2"
$err.fct
function (x, y)
   1/2 * (y - x)^2
<bytecode: 0x0000000151ae128>
<environment: 0x000000000ee42798>
attr(,"type")
[1] "sse"
$act.fct
function (x)
    1/(1 + \exp(-x))
<bytecode: 0x0000000151b29b0>
<environment: 0x000000000ee3ee00>
attr(,"type")
```

```
$data
  Var1 Var2 AND
$exclude
NULL
$net.result
$net.result[[1]]
            [,1]
[1,] -0.2408318
      0.2506263
[3,] 0.2510820
      0.7425401
```

Activity 3.2(3) Train AND gate with neuralnet()

```
plot(net)
$startweights
$startweights[[1]]
$startweights[[1]][[1]]
            [,1]
    -0.1119942
     -0.7498539
[3,] -0.7493983
$result.matrix
                                                                           AND
                            [,1]
                    0.125070634
error
                   n 006033500
reached.threshold
                   65.000000000
steps
                   -0.240831818
Intercept.to.AND
                                                    Var2
Var1.to.AND
                    0.491458166
                    0.491913796
Var2.to.AND
attr(,"class")
                                                         Error: 0.125071 Steps: 65
    "nn"
```

Activity 3.2(4) Train AND gate with neuralnet()

```
plot(net)
$startweights
$startweights[[1]]
$startweights[[1]][[1]]
            [,1]
[1, ] -0.1119942
[2,1 -0.7498539]
[3,1 -0.7493983]
                                                    Var1
$result.matrix
                                                                            AND
                            [,1]
                    0.125070634
error
reached.threshold
                    0.006833508
                   65.000000000
steps
                   -0.240831818
Intercept.to.AND
                                                     Var2
Var1.to.AND
                    0.491458166
Var2.to.AND
                    0.491913796
attr(,"class")
                                                          Error: 0.125071 Steps: 65
    "nn"
```

Activity 3.3 Predict the model AND gate

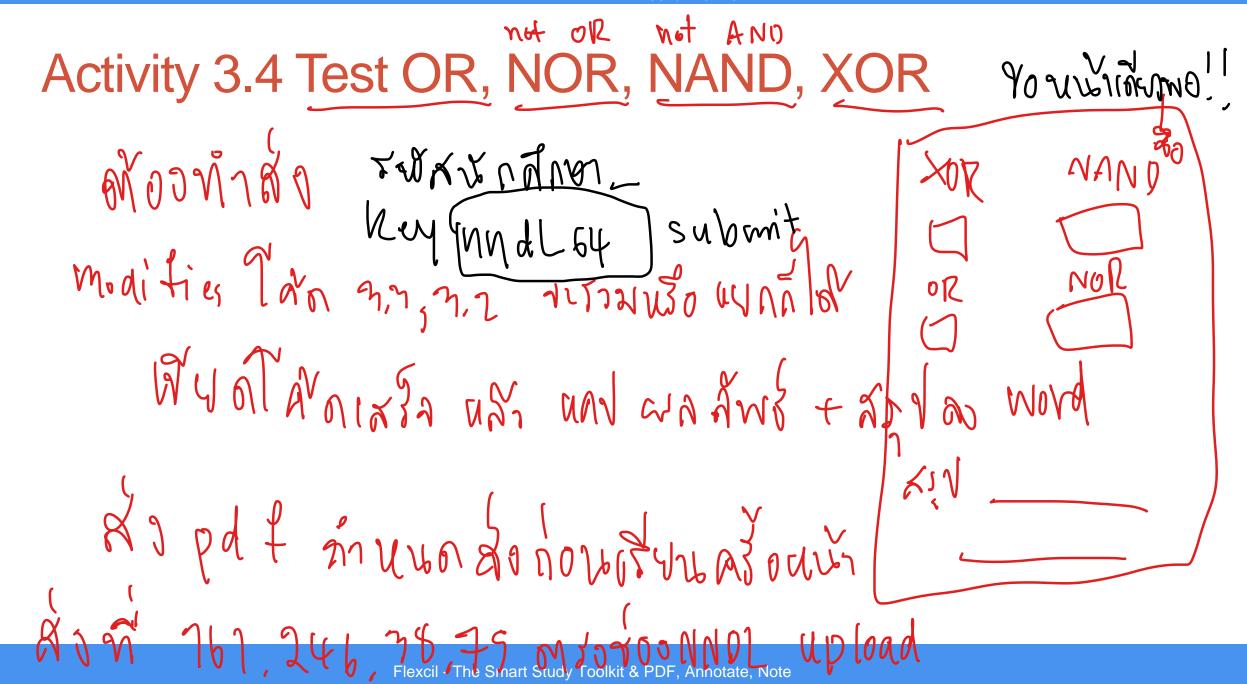
```
$startweights
$startweights[[1]]
$startweights[[1]][[1]]
[1,] -0.1119942
     -0.7498539
     -0.7493983
                                  Error: 0.125071 Steps: 65
Sresult.matrix
                            [,1]
                    0.125070634
error
reached.threshold
                    0.006833508
                   65.000000000
steps
                   -0.240831818
Intercept.to.AND
Var1.to.AND
                    0.491458166
Var2.to.AND
                    0.491913796
attr(,"class")
    "nn"
```

```
var1 = c(0,1,0,1) f(x) f
```

Activity 3.2 – 3.3 ALL CODE

```
library("neuralnet")
AND = c(0,0,0,1)
truthtable = expand.grid(c(0,1), c(0,1))
AND.data <- data.frame(truthtable, AND)
print(AND.data)
net <- neuralnet( AND~Var1+Var2, AND.data,
hidden=0, rep=5)
print(net)
plot(net)
```

```
logistic = function(x)
  1/(1 + \exp(-x))
var1 = c(0,1,0,1)
var2 = c(0,0,1,1)
n = (net result.matrix[4] +
        net$result.matrix[5] * var1 +
        net$result.matrix[6] * var2)
logistic(n)
```



Summary