	TUGAS BESAR A MACHINE LEARNING  • Arjuna Marcelino - 13519021  • Sharon Bernadetha Marbun - 13519092  • Epata Tuah - 13519120  • Giant Andreas Tambunan - 13519127
In [ ]:	<pre>Membaca Data dan Model  # Membaca data import csv  def readData():     data = []</pre>
	<pre>target_class = [] dataFile = open('data.csv', "r") reader = csv.DictReader(dataFile) for row in reader:     data.append([int(row["x1"]), int(row["x2"])])     target_class.append(int(row["f"]))  return data, target_class</pre>
	<pre># Membaca Model # filename ex: modelsigmoid.txt def readModel(filename):     file = open(filename, "r")     lines = file.readlines()  # init array activation = []</pre>
	<pre>weigth = [] bias = []  num_atr = int(lines[0]) i = 1 while i &lt; len(lines):     splitline = lines[i].strip("\n").split(" ")     # baris yang mengandung fungsi aktivasi     if(("sigmoid" in splitline) or ("relu" in splitline) or ("linear" in splitline) or "softmax" in splitline):</pre>
	<pre>activation.append(splitline[1])  i +=1     splitline = lines[i].strip("\n").split(" ")     bias.append(list(map(int, splitline)))  i+=1 else:     w = []</pre>
	<pre>for num in range(num_atr):</pre>
In [ ]:	<pre># Fungsi Aktivasi import math import numpy as np  def linear(x, kwargs=None):     return x  def sigmoid(x):</pre>
	<pre>value = float(1 / (1 + math.exp(x * -1))) threshold = 0.1 if value &lt; threshold:     return 0 else:     return 1  def relu(x):     alpha = 0.0</pre>
	<pre>max_value = 1.0 threshold = 0.0 if x &lt; threshold:     if x &gt; x*alpha:         return x     else:         return x*alpha else:</pre>
	<pre>return min(x, max_value)  def softmax(arr):     arr_exp = np.exp(arr)     return arr_exp / arr_exp.sum()  Neural Network Class and Layer Class</pre>
In [ ]:	<pre>class Layer:     definit(self,bias:list[int], weight:list[int]], activation:str):         self.weight = weight         self.bias = bias         self.activation = activation</pre>
	<pre>def get_sigma(self, input:list[int]) -&gt; list[int]:     if(len(self.weight) == len(bias)):         result = []         for i in range(len(self.bias)):             res = 1*self.bias[i]             for j in range(len(input)):                 res = res + input[j]*self.weight[j][i]                 result.append(res)</pre>
	<pre>return result  def get_result(self, input:list[int]):     sigma = self.get_sigma(input)     result = []     for i in range(len(sigma)):         result.append(self.activate_function(sigma[i]))     return result</pre>
	<pre>def activate_function(self, x):     if(self.activation == "linear"):         return linear(x)     elif(self.activation == "sigmoid"):         return sigmoid(x)     elif(self.activation == "relu"):         return relu(x)     elif(self.activation == "softmax"):</pre>
	<pre>return softmax(x)  def getDiGraph(self, index, max):     # untuk visualisasi Digraph     arr = []  # bias     i = 1     for bs in self.bias:</pre>
	<pre>a = f'b{index}' b = f'h{index+1}_{i}' if(index+1 == max):     b = f'y' arr.append([a, b, bs]) i += 1  # weight i = 1</pre>
	<pre>for x in self.weight:     j = 1     for w in x:         a = f'x{index}_{i}'         if(index&gt;0):             a = f'h{index}_{i}'         b = f'h{index+1}_{j}'         if(index+1 == max):</pre>
	<pre>b = f'y' arr.append([a, b, w])  j += 1 i += 1  return arr</pre>
	<pre>def solve(self, input:list[int]):     print("= = = = = = = ")     print(f"Input \t\t: {input}")     print(f"weight \t\t: {self.weight}")     print(f"Activation\t: {self.activation}")     print(f"Sigma \t\t: {self.get_sigma(input)}")     print()     print(f"Result \t\t: {self.get_result(input=input)}")     print("= = = = = = = = ")</pre>
In [ ]:	Neural Network Class  class FNNN:     definit(self, bias:list[list[int]], weight:list[list[int]], activation:list[str], input:list[list[int]]):         self.layers = []         self.input = input         # layer paling awal index 0
	<pre>i = 0 j = 0 for i in range(len(activation)):     layer = Layer(bias=bias[i], weight=weight[i], activation=activation[i])     self.add_layer(layer)  def add_layer(self, layer:Layer):     self.layers.append(layer)</pre>
	<pre>def resolve(self):     result = []     i = 1     for x in self.input:         print(f"\n### INPUT {i} ####")         input = x         for layer in self.layers:             idx = self.layers.index(layer)             print(f"\n====LAYER {idx}====")             layer.solve(input)</pre>
	<pre>layer.solve(input)   input = layer.get_result(input)  result.append(input[0])   i += 1  print() print(f"RESULT ==&gt; {result}")</pre>
	<pre>def get_results(self):     result = []     for x in self.input:         input = x         for layer in self.layers:             input = layer.get_result(input)          result.append(input[0])  return result</pre>
	return result  Eksekusi  Input Batch Model XOR Relu-Linear  print("# # # # # # # # # # # # # # # # # # ") print("# Feed Forward Neural Network : XOR #") print("# # # # # # # # # # # # # # # # # # #
	<pre>print("# # # # # # # # # # # # # # # # # # #</pre>
	<pre>print() print(f"TARGET CLASS\t\t: {target}") print(f"RESULT CLASS\t\t: {result}") print("==========================") if(result == target):     print("Result: GOOD PREDICT")     print("GOOD") else:     print("Result: BAD PREDICT")</pre>
	<pre>print("NO GOOD")  ### VISUALIZATION from graphviz import Digraph  dGraph = Digraph("FNNN: XOR", filename="model relu-linier.gv")  max = len(fnnn.layers)</pre>
	<pre>for layer in fnnn.layers:     idxL = fnnn.layers.index(layer)     edges = layer.getDiGraph(idxL, max)     for ed in edges:         dGraph.edge(ed[0], ed[1], str(ed[2]))  dGraph.view()  # # # # # # # # # # # # # # # # # # #</pre>
	# Feed Forward Neural Network : XOR # # # # # # # # # # # # # # # # # # #  #### INPUT 1 ####  ====LAYER 0==== = = = = = = = =
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	Method : [[1], [2]] Activation : linear Sigma : [0.0]  Result : [0.0]  = = = = = = = = = = = = = = = = = = =
	<pre>Input : [0, 1] weight : [[1, 1], [1, 1]] Activation : relu Sigma : [1, 0]</pre> Result : [1, 0] ====================================
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	Sigma : [-1.0]  Result : [-1.0]  = = = = = = = = = = = = = = = = = = =
	No GOOD 'model relu-linier.gv.pdf'  Input Batch Model XOR Sigmoid  print("# # # # # # # # # # # # # # # # # # ") print("# Feed Forward Neural Network : XOR #") print("# # # # # # # # # # # # # # # # # # ")
	<pre>bias, weigth , activation = readModel("modelsigmoid.txt") data, target = readData() fnnn = FNNN(bias, weigth, activation, data) result = fnnn.get_results()  # print result etc fnnn.resolve() print()</pre>
	<pre>print(f"TARGET CLASS\t\t: {target}") print(f"RESULT CLASS\t\t: {result}") print("=========================") if(result == target):     print("Result: GOOD PREDICT")     print("GOOD") else:     print("Result: BAD PREDICT")     print("NO GOOD")</pre>
	<pre>### VISUALIZATION from graphviz import Digraph  dGraph = Digraph("FNNN: XOR", filename="model sigmoid.gv")  max = len(fnnn.layers) for layer in fnnn.layers:</pre>
	<pre>idxL = fnnn.layers.index(layer)   edges = layer.getDiGraph(idxL, max)   for ed in edges:</pre>
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	<pre>weight : [[20, -20], [20, -20]] Activation : sigmoid Sigma : [10, 10]  Result : [1, 1] = = = = = = = = = = = = = = = = = = =</pre>
	<pre>weight : [[20], [20]] Activation : sigmoid Sigma : [10]  Result : [1] = = = = = = = = = = = = = = = = = = =</pre>
	<pre>Input : [1, 0] weight : [[20, -20], [20, -20]] Activation : sigmoid Sigma : [10, 10]  Result : [1, 1] = = = = = = = = = = = = = = = = = = =</pre>
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	#### INPUT 4 ####  ====LAYER 0====  = = = = = = = = = = = = = = = =
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	<pre>data = [[1,1]] target = [0] fnnn = FNNN(bias, weigth, activation, data) result = fnnn.get_results()  # print result etc fnnn.resolve() print() print(f"TARGET CLASS\t\t: {target}")</pre>
	<pre>print(f"TARGET CLASS\t\t: {target}") print(f"RESULT CLASS\t\t: {result}") print("====================================</pre>
	<pre>### VISUALIZATION from graphviz import Digraph  dGraph = Digraph("FNNN: XOR", filename="model sigmoid.gv")  max = len(fnnn.layers) for layer in fnnn.layers:    idxL = fnnn.layers.index(layer)    edges = layer extPiCroph(idyL, max)</pre>
	<pre>idxL = fnnn.layers.index(layer)   edges = layer.getDiGraph(idxL, max)   for ed in edges:</pre>
	#### INPUT 1 ####  ====LAYER 0====  = = = = = = =
	Result : [1, 0] = = = = = = = = = = = = = = = = = = =
	Result : [0] = = = = = = = = = = = = = = = = = = =
Out[ ]:	'model sigmoid.gv.pdf'  Input 1 Instance Model Relu-Linear  print("# # # # # # # # # # # # # # # # # # #
	<pre>bias, weigth , activation = readModel("model relu-linier.txt") data = [[1,1]] target = [0] fnnn = FNNN(bias, weigth, activation, data) result = fnnn.get_results()  # print result etc fnnn.resolve() print() print(f"TARGET CLASS\t\t: {target}")</pre>
	<pre>print() print(f"TARGET CLASS\t\t: {target}") print(f"RESULT CLASS\t\t: {result}") print("====================================</pre>
	<pre>### VISUALIZATION from graphviz import Digraph  dGraph = Digraph("FNNN: XOR", filename="model relu-linier.gv")  max = len(fnnn.layers) for layer in fnnn.layers:</pre>
	<pre>for layer in fnnn.layers:     idxL = fnnn.layers.index(layer)     edges = layer.getDiGraph(idxL, max)     for ed in edges:         dGraph.edge(ed[0], ed[1], str(ed[2]))  dGraph.view()  # # # # # # # # # # # # # # # # # # #</pre>
	<pre># # # # # # # # # # # # # # # # # # #### INPUT 1 ####  ====LAYER 0==== = = = = = = = = = = = =</pre>
	Sigma : [2, 1]  Result : [1.0, 1] = = = = = = = = = = = = = = = = = = =
	TARGET CLASS   [0]   RESULT ==> [-1.0]   RESULT CLASS   RESULT CLASS
Out[ ]:	RESULT CLASS : [-1.0]