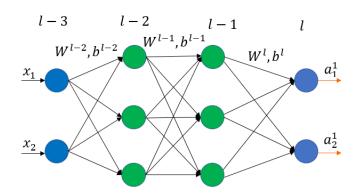
## IMPLEMENT NEURAL NETWORK



## 1. CLASS LAYER

```
from abc import abstractmethod
class Layer:
    def __init__(self):
       self.input = None
        self.output = None
        self.input_shape = None
        self.output_shape = None
        raise NotImplementedError
    @abstractmethod
    def input(self):
        return self.input
    @abstractmethod
    def output(self):
        return self.output
    @abstractmethod
    def input_shape(self):
        return self.input_shape
    @abstractmethod
    def output_shape(self):
        return self.output_shape
    @abstractmethod
    def forward_propagation(self, input):
        raise NotImplementedError
    @abstractmethod
    def backward_propagation(self, output_error, learning_rate):
        raise NotImplementedError
```

#### 2. FULL CONNECTED LAYER

```
# from .layer import Layer
import numpy as np
class FCLayer(Layer):
    def __init__(self, input_shape, output_shape):
        :param input_shape: (1, 3)
        :param output_shape: (1, 4)
        (1X3) (3x4) \Rightarrow (1, 4)
        (3, 1) (1, 4) \Rightarrow (3x4)
        self.input_shape = input_shape
        self.output_shape = output_shape
        self.weights = np.random.rand(input_shape[1], output_shape[1]) - 0.5
        self.bias = np.random.rand(1, output_shape[1]) - 0.5
    def forward_propagation(self, input):
        self.input = input
        self.output = np.dot(self.input, self.weights) + self.bias
        return self.output
    def backward_propagation(self, output_error, learning_rate):
        curent_layer_err = np.dot(output_error, self.weights.T)
        dweight = np.dot(self.input.T, output_error)
        self.weights -= dweight*learning_rate
        self.bias -= learning_rate*output_error
        return curent layer err
```

#### 3. CLASS ACTIVATION LAYER

```
# from .layer import Layer

class ActivationLayer(Layer):
    def __init__(self, input_shape, output_shape, activation,
activation_prime):
        """
            :param input_shape: dau vao input mang (1, 4)
            :param output_shape: mang
            :param activation: ham
            :param activation_prime: ham
            """
```

```
self.input_shape = input_shape
self.output_shape = output_shape
self.activation = activation
self.activation_prime = activation_prime

def forward_propagation(self, input):
    self.input = input
    self.output = self.activation(input)
    return self.output

def backward_propagation(self, output_error, learning_rate):
    return self.activation_prime(self.input)*output_error
```

### 4. CLASS NEURAL NETWORK LAYER

```
class Network:
    def __init__(self):
        self.layers = []
        self.loss = None
        self.loss_prime = None
    def add(self, layer):
        self.layers.append(layer)
    def setup_loss(self, loss, loss_prime):
        self.loss = loss
        self.loss_prime = loss_prime
    def predict(self, input):
        :param input: [[1, 3]] = > 1, [[1, 3], [3, 5], [3, 4]]
        :return: kết quả dự đoán
        ....
        result = []
        n = len(input)
        for i in range(n):
            output = input[i]
            for layer in self.layers:
                output = layer.forward_propagation(output)
            result.append(output)
        return result
    def fit(self, x_train, y_train, learning_rate, epochs):
        n = len(x_train)
        for i in range(epochs):
            err = 0
```

```
for j in range(n):
    #lan truyen tiến
    output = x_train[j]
    for layer in self.layers:
        output = layer.forward_propagation(output)

#tính lỗi của từng
    err += self.loss(y_train[j], output)

#lan truyền ngược
    error = self.loss_prime(y_train[j], output)
    for layer in reversed(self.layers):
        error = layer.backward_propagation(error, learning_rate)

err = err / n

print('epoch : %d/%d err = %f'%(i, epochs, err))
```

# 5. CÁU HÌNH DỰ ĐOÁN DỰA TRÊN NEURAL NETWORK

```
#from network.network import Network
#from layers.FCLayer import FCLayer
#from layers.activation_layer import ActivationLayer
import numpy as np
def relu(z):
    :param z: numpy array
    :return: 0 nếu z <= 0
          z néu z > 0
          [1, -3, 9, -7] \Rightarrow [1, 0, 9, 0]
    return np.maximum(0, z)
def relu_prime(z):
    .....
    :param z: numpy array
    :return: array 1, 0 \ge 0 \Rightarrow 1, z < 0 \Rightarrow 0
    z[z<0]=0
    z[z>0]=1
    return z
def loss(y_true, y_pred):
```

```
return 0.5*(y_pred-y_true)**2

def loss_prime(y_true, y_pred):
    return y_pred-y_true

x_train = np.array([[[0, 0]], [[0, 1]], [[1, 0]], [[1, 1]]])
y_train = np.array([[[0]], [[1]], [[1]], [[0]]])

net = Network()
net.add(FCLayer((1, 2), (1, 3)))
net.add(FCLayer((1, 2), (1, 3)))
net.add(FCLayer((1, 3), (1, 3), relu, relu_prime))
net.add(FCLayer((1, 3), (1, 1)))
net.add(ActivationLayer((1, 1), (1, 1), relu, relu_prime))

net.setup_loss(loss, loss_prime)

net.fit(x_train, y_train, epochs=1000, learning_rate=0.01)

out = net.predict([[0, 1]])

print(out)
```