

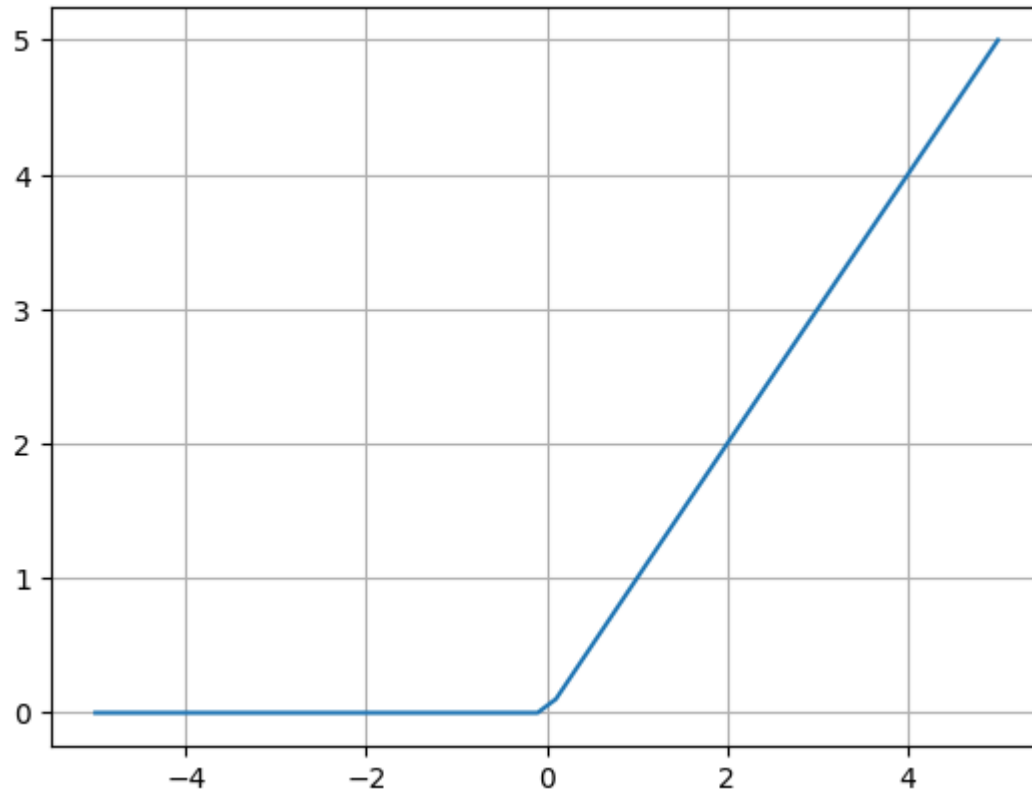
tutorial 5

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In [ ]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

```
In [ ]: def relu(z):
    return np.maximum(0, z)
```

```
In [ ]: x=np.linspace(-5,5,50)
z=relu(x)
```

```
In [ ]: plt.plot(x,z)
plt.grid()
plt.show()
```



```
In [ ]: def initialize_param(layer_sizes):
        params={}
        for i in range (1,len(layer_sizes)):
            params['W'+str(i)]=np.random.randn(layer_sizes[i],layer_sizes[i-1])
            params['b'+str(i)]=np.random.randn((layer_sizes[i],1))*0.01
        return params
```

```
In [ ]: def forward_propagation(X_train,params):
        layers = len(params)//2
        values={}
        for i in range(1,layers+1):
            if i==1:
                values['z'+str(i)]=np.dot(params['w'+str(i)],X_train) +params['b'+str(i)]
                values['A'+str(i)]=relu(values['z'+str(i)])
            else:
```

```

        values['z'+str(i)]=np.dot(params['w'+str(i)],values['A'+str(i-1)])+params['b'+str(i)]
    if i==layers:
        values['A'+str(i)]=values['z'+str(i)]
    else:
        values['A'+str(i)]=relu(values['z'+str(i)])
return values

```

Cost Function

$$J = \frac{1}{2m} \sum (J_{\text{true}} - J_{\text{pred}})^2$$

```

In [ ]: def compute_cost(values,Y_train):
        layers=len(values)//2
        Y_pred=values['A'+str(layers)]
        cost=1/(2*len(Y_train))*np.sum(np.square(Y_pred-Y_train))
        return cost

```

```

In [ ]: def train(inputs, expected_outputs, weights_input_hidden, bias_hidden, weights_hidden_output, bias_output, epochs, learning_ra
        for epoch in range(epochs):
            total_error = 0
            for i in range(len(inputs)):
                input_data = inputs[i]
                expected_output = expected_outputs[i]

                # Forward propagation
                predicted_output, hidden_output = forward_propagation(input_data, weights_input_hidden, bias_hidden, weights_hidde

                # Backward propagation
                weights_input_hidden, bias_hidden, weights_hidden_output, bias_output = backward_propagation(input_data, expected_

                # Calculate error
                total_error += mean_squared_error(expected_output, predicted_output)

            print(f'Epoch {epoch}, Error: {total_error}')

```

```
return weights_input_hidden, bias_hidden, weights_hidden_output, bias_output
```

```
In [ ]: def train(self, X, y, epochs, learning_rate):
        for epoch in range(epochs):
            total_error = 0
            for i in range(len(X)):
                input_data = X[i:i+1]
                expected_output = y[i:i+1]

                # Forward propagation
                predicted_output, hidden_output = self.forward_propagation(input_data)

                # Backward propagation
                self.backward_propagation(input_data, expected_output, predicted_output, hidden_output, learning_rate)

                # Calculate error
                total_error += self.mean_squared_error(expected_output, predicted_output)

            # Print error
            print(f'Epoch {epoch}, Error: {total_error}')
```