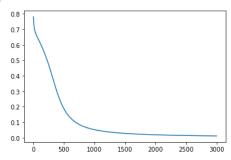
```
#Form a 3 Layer neural network (one input, one hidden, and one output) to Learn the XOR function.
                      #The input layer should contain 2 binary inputs.
                      #Second Layer (first hidden Layer) should contain 3 neurons.
                      #Output layer contains 1 neuron which will produce the output of the XOR function
  In [8]: | import numpy as np
                      # Activation Functions
                      def tanh(x):
                               return np.tanh(x)
                      \begin{tabular}{ll} \beg
                              return 1 - np.square(np.tanh(x))
                      def sigmoid(x):
                              return 1/(1 + np.exp(-x))
                      def d_sigmoid(x):
                              return (1 - sigmoid(x)) * sigmoid(x)
                      # Loss Functions
                      def logloss(y, a):
                               return -(y*np.log(a) + (1-y)*np.log(1-a))
                      def d_logloss(y, a):
                              return (a - y)/(a*(1 - a))
  In [9]: # The layer class
                      class Layer:
                                activationFunctions = {
                                        'tanh': (tanh, d_tanh),
'sigmoid': (sigmoid, d_sigmoid)
                                learning_rate = 0.1
                               self.W = np.random.randn(neurons, inputs)
                                       self.b = np.zeros((neurons, 1))
self.act, self.d_act = self.activationFunctions.get(activation)
                               def feedforward(self, A_prev):
                                       self.A_prev = A_prev
self.Z = np.dot(self.W, self.A_prev) + self.b
self.A = self.act(self.Z)
                                        return self.A
                               def backprop(self, dA):
    dZ = np.multiply(self.d_act(self.Z), dA)
                                       dA_prev = np.dot(self.W.T, dZ)
                                       dW = 1/dZ.shape[1] * np.dot(dZ, self.A_prev.T)
db = 1/dZ.shape[1] * np.sum(dZ, axis=1, keepdims=True)
                                        self.W = self.W - self.learning_rate * dW
                                        self.b = self.b - self.learning_rate * db
                                        return dA_prev
In [10]:
                     x_train = np.array([[0, 0, 1, 1], [0, 1, 0, 1]]) # dim \times m y_train = np.array([[0, 1, 1, 0]]) # 1 \times m
                      #dnn = DeepNeuralNetwork(sizes=[784, 128, 64, 10])
                      m = 4
                      epochs = 3000
                      layers = [Layer(2, 3, 'tanh'), Layer(3, 1, 'sigmoid')]
                      costs = [] # to plot graph
                       for epoch in range(epochs):
                               A = x_{train}
                               for layer in layers:
    A = layer.feedforward(A)
                               cost = 1/m * np.sum(logloss(y_train, A))
                               costs.append(cost)
                               dA = d_logloss(y_train, A)
                               for layer in reversed(layers):
                                      dA = layer.backprop(dA)
In [11]: \mid # predicting
                      A = np.array([[0], [0]]) # dim(=2) x m(=1)
                      for layer in layers:
    A = layer.feedforward(A)
                      print(A)
                    [[0.00805333]]
```

import matplotlib.pyplot as plt
plt.plot(range(epochs), costs)

Out[12]: [<matplotlib.lines.Line2D at 0x1b785b95b50>]



In []: