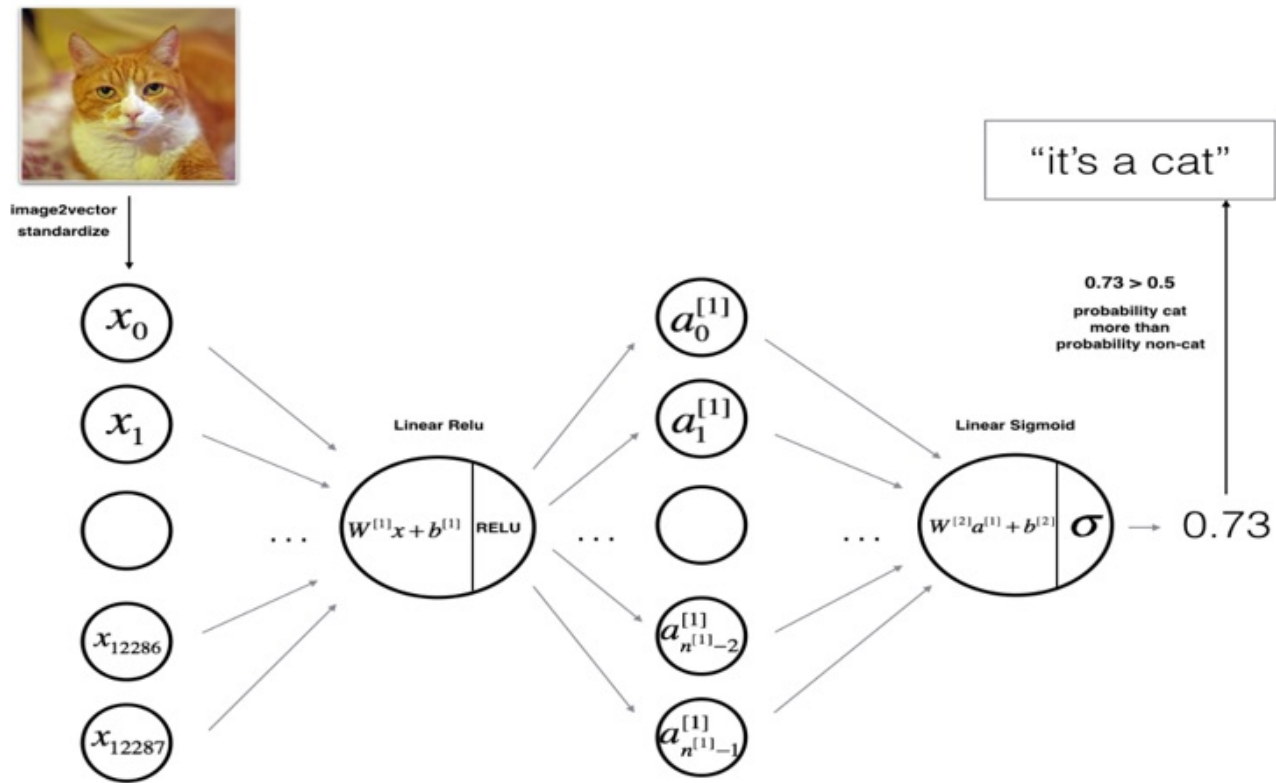


Deep Neural Network for Image Classification

1. 2-Layer Neural Network



- The input is a (64,64,3) image which is flattened to a vector of size (12288, 1).
- The corresponding vector: $[x_0, x_1, \dots, x_{12287}]^T$ is then multiplied by the weight matrix $W^{[1]}$ of size $(n^{[1]}, 12288)$.
- You then add a bias term and take its relu to get the following vector: $[a_0^{[1]}, a_1^{[1]}, \dots, a_{n^{[1]}-1}^{[1]}]^T$.
- You then repeat the same process.
- You multiply the resulting vector by $W^{[2]}$ and add your intercept (bias).
- Finally, you take the sigmoid of the result. If it is greater than 0.5, you classify it to be a cat.

(1) Initialize parameters

parameters = initialize_parameters(n_x, n_h, n_y)

$w_1 / b_1 / w_2 / b_2$

layers - dims

(2) Loop (gradient descent)

for i in range(0, num_iterations):

1. forward propagation

A1, cache = linear_activation_forward(X, w1, b1, activation="relu")

A2, cache = linear_activation_forward(A1, w2, b2, activation="sigmoid")

LINEAR \rightarrow RELU \rightarrow LINEAR \rightarrow SIGMOID

ii. Compute cost

$$\text{cost} = \text{compute_cost}(A2, Y)$$

iii. Backward propagation

$$dA1, dw2, db2 = \text{linear_activation_backward}(dA2, \text{cache2}, \text{activation} = "symmid")$$

$$dA0, dW1, dB1 = \text{linear_activation_backward}(dA1, \text{cache1}, \text{activation} = "relu")$$

$$\text{grads}['dW1'] = dW1$$

...

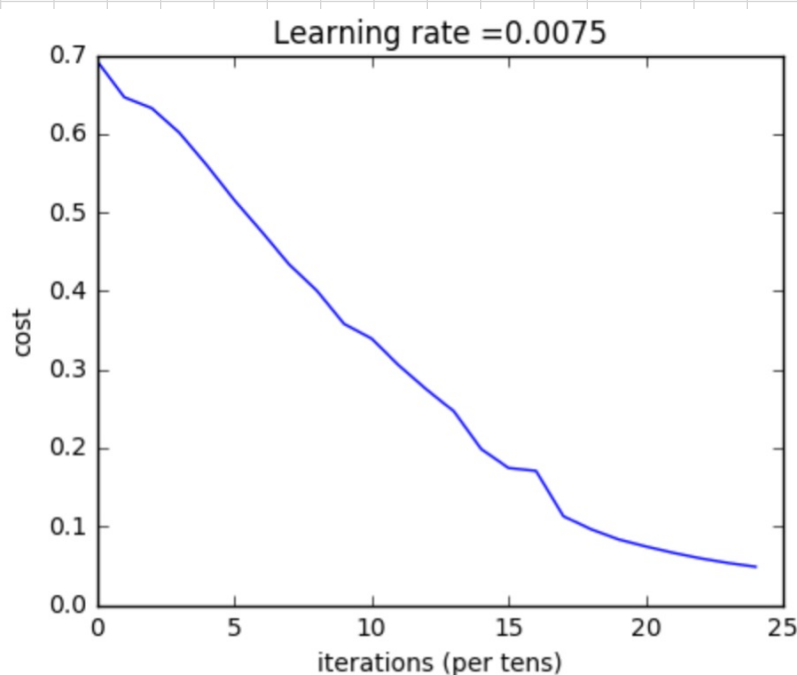
iv. update parameters

$$\text{parameters} = \text{update_parameters}(\text{parameters}, \text{grads}, \text{learning_rate} = \text{learning_rate})$$

$$W1 = \text{parameters}['W1']$$

...

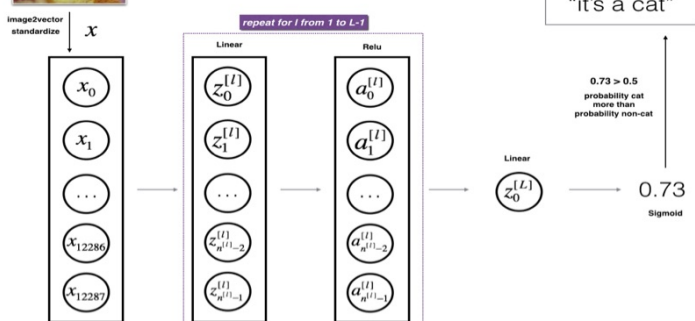
(3) output



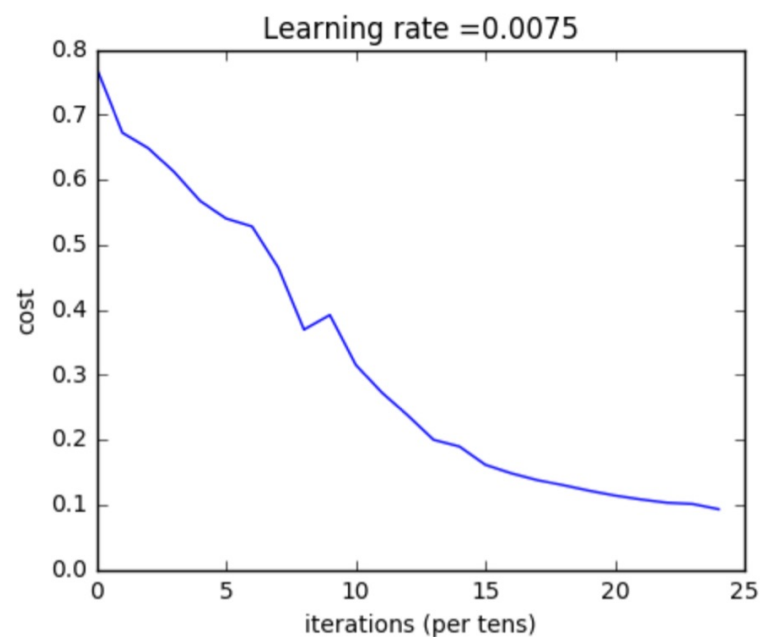
train: 100%

test: 92%

2. L-layer Neural Network



- The input is a (64,64,3) image which is flattened to a vector of size (12288,1).
- The corresponding vector: $[x_0, x_1, \dots, x_{12287}]^T$ is then multiplied by the weight matrix $W^{[1]}$ and then you add the intercept $b^{[1]}$. The result is called the linear unit.
- Next, you take the relu of the linear unit. This process could be repeated several times for each $(W^{[l]}, b^{[l]})$ depending on the model architecture.
- Finally, you take the sigmoid of the final linear unit. If it is greater than 0.5, you classify it to be a cat.



the process of 1-layer Neural Network nearly to 2-layer, and the accuracy of train set is 98%, while the test set is 80% thus, the 5-layer neural network has better performance than the 2-layer neural network on the same test set.