

Implementation of Neural Network

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We'll try to implement a L-layer neural network. The implementation could be divided into 5 steps:

o. First things first

Layers_dims

[12288, 20, 7, 5, 1]

Use a dictionary to record parameters($W[i]$, $b[i]$) of DNN

parameters = {"W1": W1, "b1": b1, "W2": W2, "b2": b2}

Shape of parameter matrixes: ($Z[i] = W[i] * X[i] + b[i]$)

i	$W[i]$	$X[i] = Z[i-1]$	$b[i]$	$Z[i]$	
1	20, 12888	12888, 209	20, 1	20, 209	
2	7, 20	20, 209	7, 1	7, 209	
3	5, 7	7, 209	5, 1	5, 209	
4	1, 5	5, 209	1, 1	1, 209	

Functions	Usage	Input	Out
initialize_parameters_deep	Initialize parameter dictionary of DNN (Refer to Initialization in course2)	Num of Layers	parameters{ $W[i]$, $b[i]$ }
L_model_forward	Forward Propagation For each layer, $Z[i] = W[i] \cdot A[i] + b[i]$	X, parameters{ $W[i]$, $b[i]$ }	AL (The output layer) cache(A_{prev} , W, b)
compute_cost	Compute cost	AL, Y	cost
L_model_backward	Backward Propagation For each layer(in reversed order), compute gradient dA, dW, db	AL, Y, cache(A_{prev} , W, b)	Grads{ $dA[i]$, $dW[i]$, $db[i]$ }
update_parameters	For each layer, parameters[$W[i]$] = parameters[$W[i]$] - learning_rate * grads[$dW[i]$] parameters[$b[i]$] = parameters[$b[i]$] - learning_rate * grads[$db[i]$]	parameters{ $W[i]$, $b[i]$ }, Grads{ $dA[i]$, $dW[i]$, $db[i]$ }, learning_rate	parameters{ $W[i]$, $b[i]$ } (new value)
linear_activation_forward	Z, linear_cache = linear_forward(A_{prev} , W, b) A , activation_cache = Activation_function(Z)	A_{prev} , W, b, activation	A, cache(linear_cache, activation_cache)
linear_forward	$Z = W \cdot A + b$	A, W, b	Z, cache(A,W,b)
linear_activation_backward	Calculate dW, db for current layer	dA, cache(linear_cache, activation_cache), activation	dA_{prev} , dW, db
linear_backward	Calculate dW, db	dZ, cache(A_{prev} , W, b)	dA_{prev} , dW, db

W, b indicates the $parameters['WL']$, $parameters['bL']$ for a specific layer L

1. Load dataset

	Shape	comment
Train_x_org	(209, 64, 64, 3)	209 original 64x64-pixel colored pictures
Train_x	(12288, 209)	Train_x_org is reversed and flattened (12288 = 64x64x3). The values are standardized to range [0,1]
Train_y	(1, 209)	Train_y is reversed

2. Initialize parameters

Here we will use Random Initialization.

3. Train NN

For each iteration:

L_model_forward

Compute_cost

L_model_backward

Update_parameters

4. Predict

