### Optimization algorithms

Quiz, 10 questions

10/10 points (100%)

### **✓** Congratulations! You passed!

Next Item



1/1 points

1.

Which notation would you use to denote the 3rd layer's activations when the input is the 7th example from the 8th minibatch?



 $a^{[8]\{3\}(7)}$ 



 $a^{[3]\{7\}(8)}$ 



 $a^{[8]\{7\}(3)}$ 



 $a^{[3]\{8\}(7)}$ 



Correct



1/1 points

2

Which of these statements about mini-batch gradient descent do you agree with?



Training one epoch (one pass through the training set) using minibatch gradient descent is faster than training one epoch using batch gradient descent.



One iteration of mini-batch gradient descent (computing on a single mini-batch) is faster than one iteration of batch gradient descent.



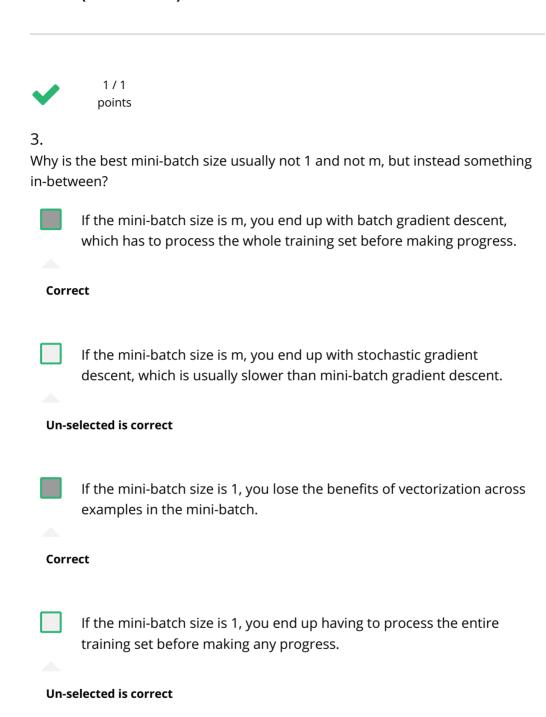
Correct



# Quiz, 10 questions

You should implement mini-batch gradient descent without an Optimization algorithm oop over different mini-batches, so that the algorithm processes all mini-batches at the same time (vectorization).

10/10 points (100%)





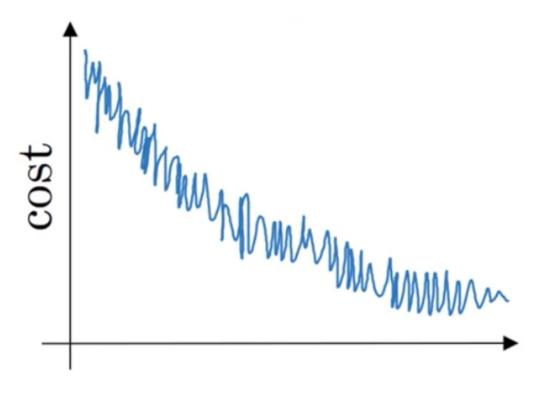
1/1 points

Suppose your learning algorithm's cost J, plotted as a function of the number of

## Optimization algorithms his:

10/10 points (100%)

Quiz, 10 questions



Which of the following do you agree with?

- Whether you're using batch gradient descent or mini-batch gradient descent, this looks acceptable.
- If you're using mini-batch gradient descent, this looks acceptable. But if you're using batch gradient descent, something is wrong.

#### Correct

- Whether you're using batch gradient descent or mini-batch gradient descent, something is wrong.
- If you're using mini-batch gradient descent, something is wrong. But if you're using batch gradient descent, this looks acceptable.



1/1 points

5.

Suppose the temperature in Casablanca over the first three days of January are

## Optimizations algorithms

Quiz, 10 questions

10/10 points (100%)

Jan 1st: 
$$\theta_1 = 10^{\circ} C$$

Jan 2nd:  $\theta_2 10^{\circ} C$ 

(We used Fahrenheit in lecture, so will use Celsius here in honor of the metric world.)

Say you use an exponentially weighted average with  $\beta=0.5$  to track the temperature:  $v_0=0$ ,  $v_t=\beta v_{t-1}+(1-\beta)\theta_t$ . If  $v_2$  is the value computed after day 2 without bias correction, and  $v_2^{corrected}$  is the value you compute with bias correction. What are these values? (You might be able to do this without a calculator, but you don't actually need one. Remember what is bias correction doing.)

$$v_2 = 7.5, v_2^{corrected} = 7.5$$

$$v_2 = 10, v_2^{corrected} = 7.5$$

$$v_2 = 10, v_2^{corrected} = 10$$

$$v_2 = 7.5, v_2^{corrected} = 10$$

Correct



1/1 points

6

Which of these is NOT a good learning rate decay scheme? Here, t is the epoch number.

$$\alpha = \frac{1}{1+2*t} \alpha_0$$

$$\alpha = e^t \alpha_0$$

Correct

$$\alpha = 0.95^t \alpha_0$$

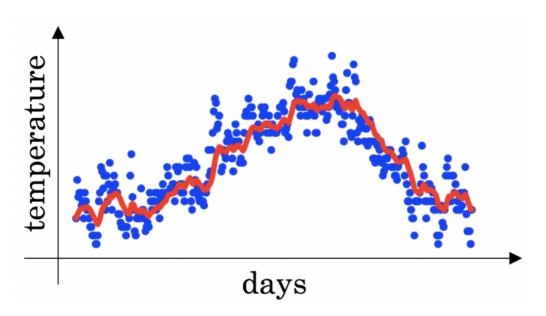
$$\alpha = \frac{1}{\sqrt{t}} \alpha_0$$

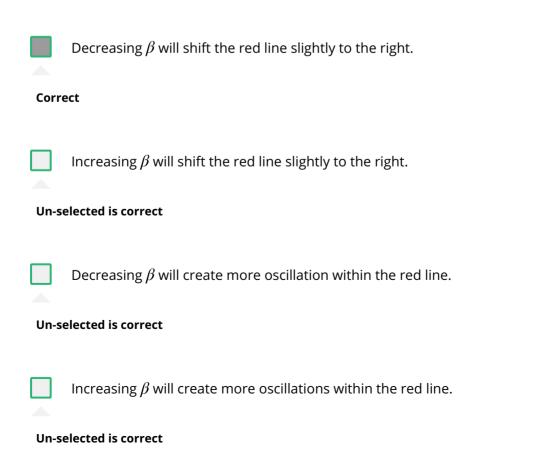
## Optimization algorithms Quiz, 10 questions

10/10 points (100%)

7

You use an exponentially weighted average on the London temperature dataset. You use the following to track the temperature:  $v_t = \beta v_{t-1} + (1-\beta)\theta_t$ . The red line below was computed using  $\beta = 0.9$ . What would happen to your red curve as you vary  $\beta$ ? (Check the two that apply)





#### Optimization algorithms

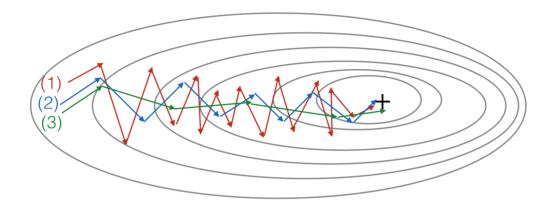
Quiz, 10 questions

10/10 points (100%)



1/1 points

8. Consider this figure:



These plots were generated with gradient descent; with gradient descent with momentum ( $\beta$  = 0.5) and gradient descent with momentum ( $\beta$  = 0.9). Which curve corresponds to which algorithm?

- (1) is gradient descent with momentum (small  $\beta$ ). (2) is gradient descent. (3) is gradient descent with momentum (large  $\beta$ )
- (1) is gradient descent. (2) is gradient descent with momentum (large  $\beta$  ) . (3) is gradient descent with momentum (small  $\beta$ )
- (1) is gradient descent. (2) is gradient descent with momentum (small  $\beta$  ). (3) is gradient descent with momentum (large  $\beta$ )

#### Correct

(1) is gradient descent with momentum (small  $\beta$ ), (2) is gradient descent with momentum (small  $\beta$ ), (3) is gradient descent



points

9.

Suppose batch gradient descent in a deep network is taking excessively long to find a value of the parameters that achieves a small value for the cost function  $\mathcal{J}(W^{[1]},b^{[1]},\ldots,W^{[L]},b^{[L]})$ . Which of the following techniques could help find parameter values that attain a small value for  $\mathcal{J}$ ? (Check all that apply)

# Optimization algorithms Try initializing all the weights to zero

Quiz, 10 questions

**Un-selected** is correct

10/10 points (100%)



Try tuning the learning rate lpha

Correct



Try better random initialization for the weights

Correct



Try using Adam

Correct



Try mini-batch gradient descent

Correct



1/1 points

10.

Which of the following statements about Adam is False?



Adam should be used with batch gradient computations, not with mini-batches.

Correct

We

We usually use "default" values for the hyperparameters  $\beta_1,\beta_2$  and  $\varepsilon$  in Adam ( $\beta_1=0.9$ ,  $\beta_2=0.999$ ,  $\varepsilon=10^{-8}$ )

The learning rate hyperparameter lpha in Adam usually needs to be tuned.

Adam combines the advantages of RMSProp and momentum

## Optimization algorithms

Quiz, 10 questions 10/10 points (100%)



