

# Optimization algorithms

**10/10 points (100%)**

Quiz, 10 questions

**✓ 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?

 $a^{[3]} \{8\} (7)$ **Correct** $a^{[8]} \{7\} (3)$  $a^{[3]} \{7\} (8)$  $a^{[8]} \{3\} (7)$ 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 mini-batch gradient descent is faster than training one epoch using batch gradient descent.



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

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

Correct



1 / 1  
points

3.

Why is the best mini-batch size usually not 1 and not  $m$ , but instead something in-between?



If the mini-batch size is  $m$ , you end up with batch gradient descent, which has to process the whole training set before making progress.



Correct



If the mini-batch size is 1, you end up having to process the entire training set before making any progress.



Un-selected is correct



If the mini-batch size is  $m$ , you end up with stochastic gradient descent, which is usually slower than mini-batch gradient descent.



Un-selected is correct



If the mini-batch size is 1, you lose the benefits of vectorization across examples in the mini-batch.



Correct



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points

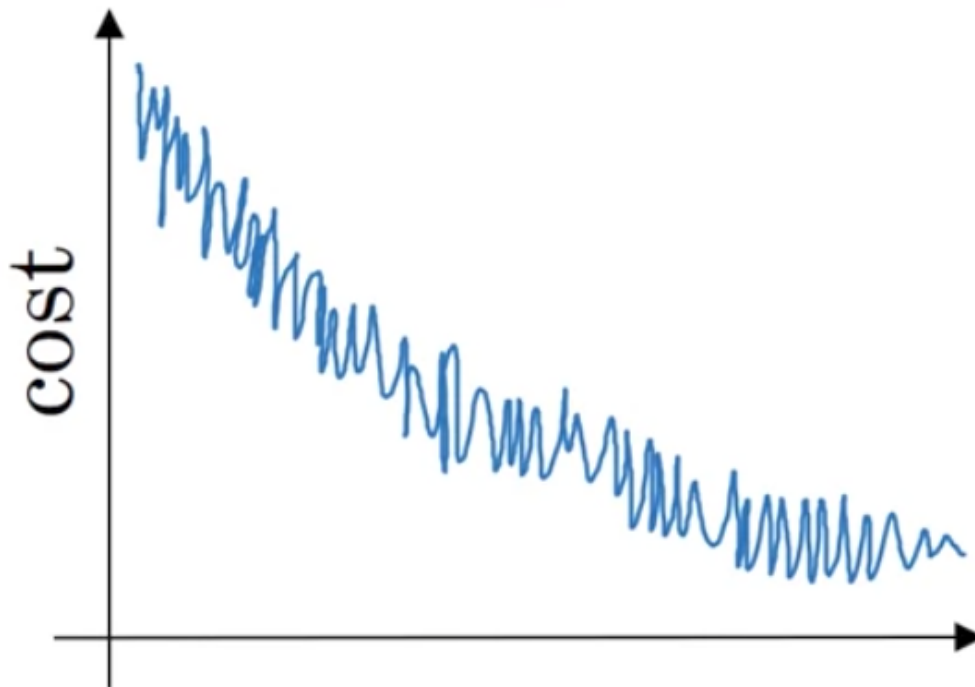
4.

## Optimization algorithms

Suppose your learning algorithm's cost  $J$ , plotted as a function of the number of iterations, looks like this:

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Which of the following do you agree with?

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



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points

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5.

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

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.)

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



Correct

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

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

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



1 / 1  
points

6.

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



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

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

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



Correct



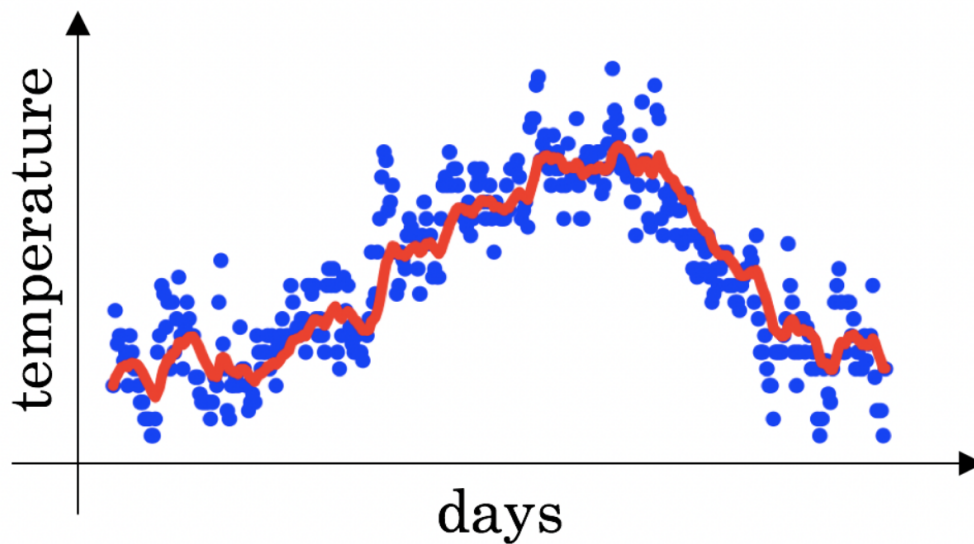
$$\alpha = \frac{1}{\sqrt{t}} \alpha_0$$



1 / 1  
points

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)



Decreasing  $\beta$   
will shift the red line slightly to the right.

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Increasing  $\beta$   
will shift the red line slightly to the right.

**Correct**

True, remember that the red line corresponds to  $\beta = 0.9$   
. In lecture we had a green line ( $\beta = 0.98$ ) that is slightly shifted to the right.



Decreasing  $\beta$   
will create more oscillation within the red line.

**Correct**

True, remember that the red line corresponds to  $\beta = 0.9$   
. In lecture we had a yellow line ( $\beta = 0.98$ ) that had a lot of oscillations.



Increasing  $\beta$   
will create more oscillations within the red line.

**Un-selected is correct**

1 / 1  
points

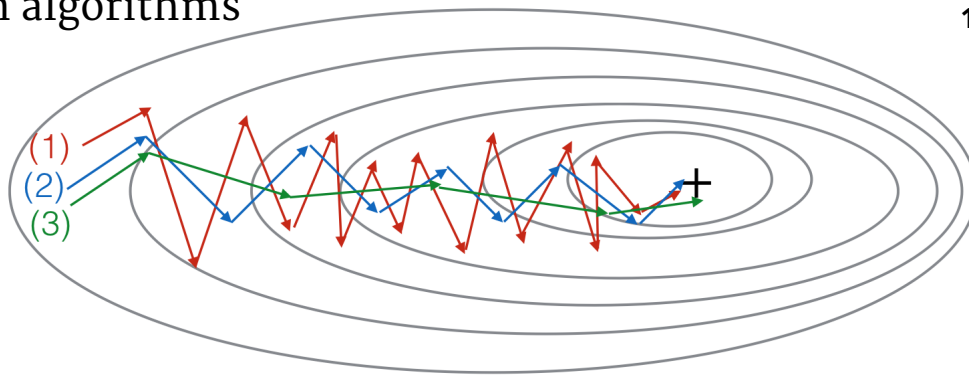
8.

Consider this figure:

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



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points

9.

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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  $J(W^{[1]}, b^{[1]}, \dots, W^{[L]}, b^{[L]})$ .

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Which of the following techniques could help find parameter values that attain a small value for  $J$ ? (Check all that apply)

☐

Try mini-batch gradient descent



Correct

☐

Try better random initialization for the weights



Correct

☐

Try initializing all the weights to zero



Un-selected is correct

☐

Try using Adam



Correct

☐

Try tuning the learning rate  $\alpha$



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.

