# $\leftarrow \quad \underset{\text{Quiz, 10 questions}}{\text{Optimization algorithms}}$

8/10 points (80.00%)

	<b>~</b>	Congratulations! You passed!	Next Item				
<b>~</b>	1 / 1 point						
1. Which	notation w	ould you use to denote the 3rd layer's activations when the input is the 7th exa	mple from the 8th minibatch?				
	$a^{[3]\{7\}(8)}$						
	$a^{[8]\{7\}(3)}$						
	$a^{[8]\{3\}(7)}$						
0	$a^{[3]\{8\}(7)}$						
Corr	ect						
<ul> <li>2. Which of these statements about mini-batch gradient descent do you agree with?</li> <li>Training one epoch (one pass through the training set) using mini-batch gradient descent is faster than training one epoch using batch gradient descent.</li> <li>One iteration of mini-batch gradient descent (computing on a single mini-batch) is faster than one iteration of batch gradient descent.</li> <li>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).</li> </ul> This should not be selected							
<b>X</b>	0 / 1 point						
	3. Why is the best mini-batch size usually not 1 and not m, but instead something in-between?						
	If the min	i-batch size is 1, you end up having to process the entire training set before ma	king 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 Optimization algorithms Quiz, 10 questions

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### **Un-selected is correct**

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

#### This should be selected

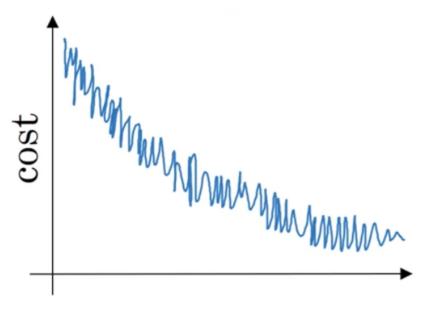
If the mini-batch size is m, you end up with batch gradient descent, which has to process the whole training set before making progress.

#### Correct



point

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



Which of the following do you agree with?

$\bigcirc$	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, this looks acceptable. But if you're using batch gradient descent, something is wrong.

### Correct

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



## Optimization algorithms

Quiz, 10 questions

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1/1 point

5.

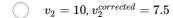
Suppose the temperature in Casablanca over the first three days of January are the same:

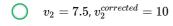
Jan 1st:  $heta_1=10^oC$ 

Jan 2nd:  $heta_2 10^o C$ 

(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.)





Correct

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

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



1/1 point

6.

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



Correct

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

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

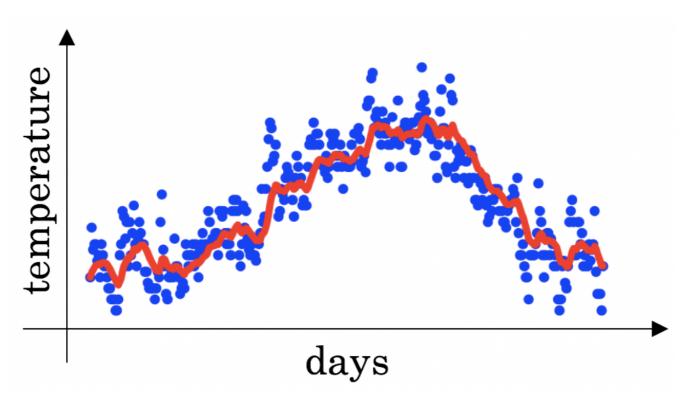
$$\alpha = \frac{1}{\sqrt{t}} \alpha_0$$



1/1 point

7.

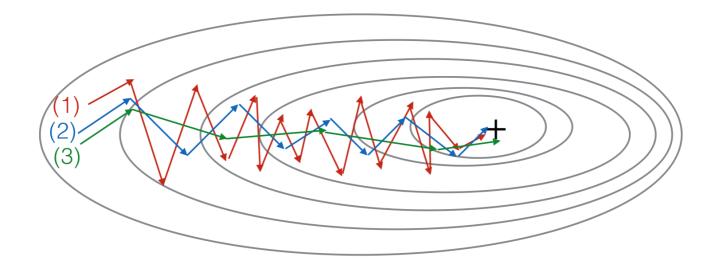
You use an exponentially weighted average on the London temperature dataset. You use the following to track the temperature:  $e^{i\omega_t} = \beta \Omega ptimization in a least ballow was computed using <math>\beta = 0.9$ . What would happen to your red curve as you vary  $\beta = 0.9$ . What would happen to your red curve as you vary  $\beta = 0.9$ . What would happen to your red curve as you vary  $\beta = 0.9$ . What would happen to your red curve as you vary  $\beta = 0.9$ . What would happen to your red curve as you vary  $\beta = 0.9$ .



Decreasing $eta$ will shift the red line slightly to the right.						
Un-selected is correct						
Increasing $eta$ 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 $eta$ will create more oscillation within the red line.						
Correct True, remember that the red line corresponds to $eta=0.9$ . In lecture we had a yellow line $\$$ \beta = 0.98 that had a lot of						
oscillations.						
Increasing $eta$ will create more oscillations within the red line.						
Un-selected is correct						



8/10 points (80.00%)



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 with momentum (small $\beta$ ), (2) is gradient descent with momentum (small $\beta$ ), (3) is gradient descent
	(1) is gradient descent. (2) is gradient descent with momentum (large $\beta$ ) . (3) is gradient descent with momentum (small $\beta$ )
0	(1) is gradient descent. (2) is gradient descent with momentum (small $\beta$ ). (3) is gradient descent with momentum (large $\beta$ )
Corr	rect

•

1/1 point

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  $\sqrt{3}W^{[1]}, b^{[1]}, \dots, W^{[L]}, b^{[L]}$ ). Which of the following techniques could help find parameter values that attain a small value for  $\sqrt{3}$  (Check all that apply)

	Try better random initialization for the weights			
Corr	ect			
	Try initializing all the weights to zero			
Un-selected is correct				
	Try mini-batch gradient descent			

Correct

<del>-</del>	Optimization algorithms ्राष्ट्र भरुंपुद्धक्रीलाः	8/10 points (80.00%)						
Corr	Correct							
Corr	Try tuning the learning rate $lpha$							
<b>~</b>	1/1 point							
10. <b>Which</b>	of the following statements about Adam is False?							
0	Adam should be used with batch gradient computations, not with mini-batches.							
Corr	ect							
	We usually use "default" values for the hyperparameters $eta_1,eta_2$ and $\ \ $ in Adam ( $eta_1=0.9,eta_2=0.9$	$999, = 10^{-8}$						
	The learning rate hyperparameter $lpha$ in Adam usually needs to be tuned.							
	Adam combines the advantages of RMSProp and momentum							



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