$\leftarrow$	Optimization algorithms	8/10 points (80.00%)
	Quiz, 10 questions	
	Congratulations! You passed!	Next Item
<b>~</b>	1 / 1 points	
1.		
Which	n notation would you use to denote the 3rd layer's activations when the in	input is the 7th example from the 8th minibatch?
	$a^{[3]\{7\}(8)}$	
0	$a^{[3]\{8\}(7)}$	
Corr	rect	
	$a^{[8]\{3\}(7)}$	
	$a^{[8]\{7\}(3)}$	
	1/1	
<b>\</b>	points	
2.	n of these statements about mini-batch gradient descent do you agree wi	uith2
WITICH		
	Training one epoch (one pass through the training set) using mini-batc using batch gradient descent.	ch gradient descent is faster than training one epoch
	You should implement mini-batch gradient descent without an explicit	it for-loop over different mini-batches, so that the
	algorithm processes all mini-batches at the same time (vectorization).	
0	One iteration of mini-batch gradient descent (computing on a single m descent.	mini-batch) is faster than one iteration of batch gradient
	descent.	
Corr	rect	
<b>~</b>	1 / 1 points	
3.		
	s the best mini-batch size usually not 1 and not m, but instead something	ng in-between?
	If the mini-batch size is 1, you lose the benefits of vectorization across	s examples in the mini-batch.
Corr	rect	

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

# descent. Optimization algorithms

8/10 points (80.00%)

### Un-sel@ctied1is questictns

	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 batch gradient descent, which has to process the whole training set before making
progress.

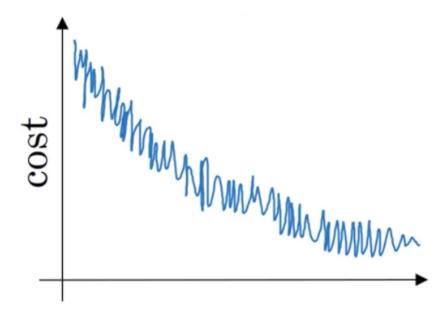
Correct



1/1 points

4.

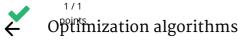
Suppose your learning algorithm's cost J, plotted as a function of the number of iterations, looks like this:



Which of the following do you agree with?

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

Correct



8/10 points (80.00%)

5. Quiz, 10 questions

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

Jan 1st:  $heta_1=10^oC$ 

Jan 2nd:  $heta_2 10^o C$ 

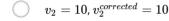
(We used Fahrenheit in lecture, so will use Celsius here in honor of the metric world.)

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.)



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

Correct



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

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



6.

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



Correct

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

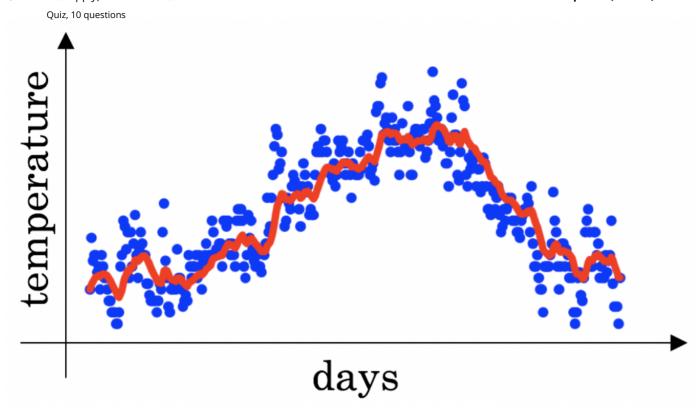
$$lpha = 0.95^t lpha_0$$

$$lpha = rac{1}{1+2*t}lpha_0$$



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 two Details 2 to 10 points (80.00%)



## **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  $\boldsymbol{\beta}$  will create more oscillation within the red line.

# This should be selected

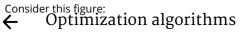
Increasing eta will create more oscillations within the red line.

# This should not be selected

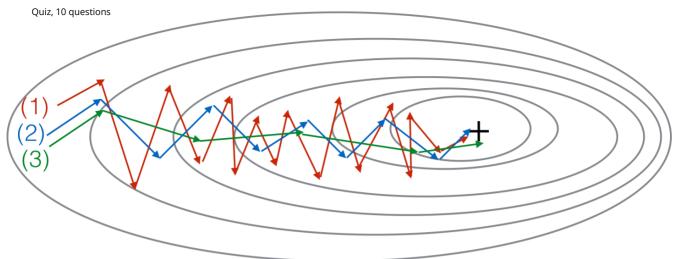
False. Increasing  $\beta$  will cause fewer oscillations



1/1 points 8.



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

(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 eta). (3) is gradient descent with momentum (large eta)

Correct



0 / 1 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  $\mathcal{J}(W^{[1]},b^{[1]},...,W^{[L]},b^{[L]})$ . Which of the following techniques could help find parameter values that attain a small value for  $\mathcal{J}$ ? (Check all that apply)

Try initializing all the weights to zero

Un-selected is correct

Try mini-batch gradient descent

Correct

Try better random initialization for the weights

Correct

Try using Adam

	ect

# ← Optimization algorithms

8/10 points (80.00%)

•	-	-	٦
			ı

Quiz, 10 questions Try tuning the learning rate lpha

This should be selected



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







