## Optimization algorithms

10/10 points (100%)

Item

Quiz, 10 questions

<b>✓</b> Congra	atulations! You passed!
•	1/1 points
1	politics
	notation would you use to denote the 3rd layer's activations when out is the 7th example from the 8th minibatch?
	$a^{[8]\{3\}(7)}$
	$a^{[3]\{7\}(8)}$
0	$a^{[3]\{8\}(7)}$
Corr	ect
	$a^{[8]\{7\}(3)}$
<b>✓</b>	1 / 1 points
2. Which agree	of these statements about mini-batch gradient descent do you with?
0	One iteration of mini-batch gradient descent (computing on a single mini-batch) is faster than one iteration of batch gradient descent.
<u> </u>	
Corr	ect
	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).

Optimizations  Quiz, 10 questions	mini-batch gradient descent is faster than training one epoch using using batch gradient descent.
	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 1, you lose the benefits of vectorization across examples in the mini-batch.
	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 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
	1/1 points

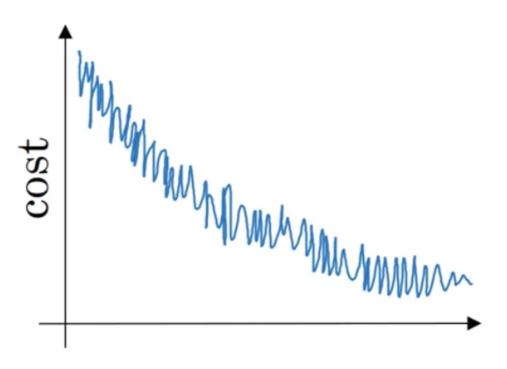
4.

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

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Quiz, 10 questions



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, 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.
- Whether you're using batch gradient descent or mini-batch gradient descent, something is wrong.



1/1 points

5.

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

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Jan 1st: 
$$heta_1=10^oC$$

Jan 2nd:  $\theta_2 10^o C$ 

(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

$$igcup_2=10$$
,  $v_2^{corrected}=10$ 

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

$$igcup_2=10$$
,  $v_2^{corrected}=7.5$ 



1/1 points

6.

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



$$lpha=e^tlpha_0$$

Correct

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

$$\bigcirc \quad \alpha = \frac{1}{\sqrt{t}} \, \alpha_0$$

$$lpha = 0.95^t lpha_0$$

1/1

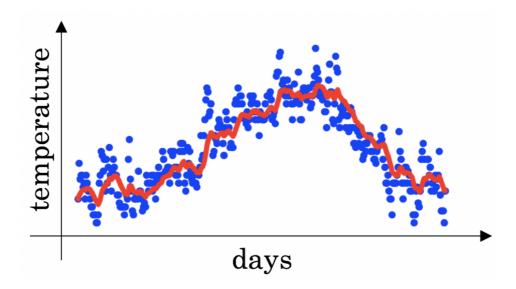
# Optimization algorithms

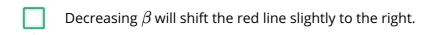
10/10 points (100%)

Quiz, 10 questions

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)



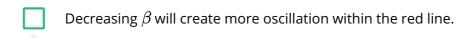


### **Un-selected is correct**

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.



### 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.

# Optimization algorithms Un-selected is correct

Quiz, 10 questions

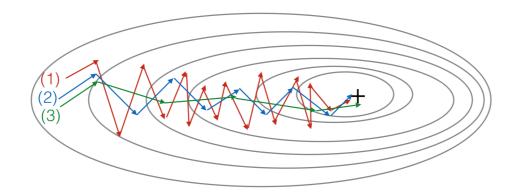
10/10 points (100%)



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. (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
- (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 / 1 points

Optimizations Quiz, 10 questions	9. Suppose batch gradient descent in a deep network is taking excessively large $\mathbf{S}$ and $\mathbf{S}$ of the parameters that achieves a small value for the $0$ 00% 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)
	Try tuning the learning rate $\alpha$
	Try using Adam  Correct
	Try better random initialization for the weights  Correct
	Try mini-batch gradient descent  Correct
	Try initializing all the weights to zero  Un-selected is correct
	1/1 points  10. Which of the following statements about Adam is False?
	We usually use "default" values for the hyperparameters $eta_1,eta_2$ and $arepsilon$ in Adam ( $eta_1=0.9$ , $eta_2=0.999$ , $arepsilon=10^{-8}$ )

be tuned.

The learning rate hyperparameter  $\boldsymbol{\alpha}$  in Adam usually needs to

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Optimization algorithms Id be used with batch gradient computations, not Optimization algorithms (100%) With mini-batches.

Correct

3 R