# 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?

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

**Correct**

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

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

**Correct**

- 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).

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 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



- [ ] 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**

- [x] 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**

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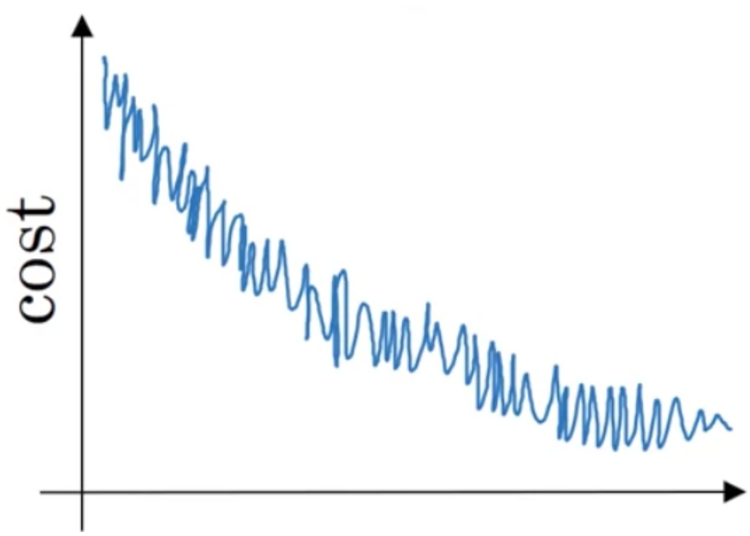
1 / 1
points

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?

- ( ) 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.
- (•) 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.

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1 / 1
points

5.
Suppose the temperature in Casablanca over the first three days of January are the same:

Jan 1st: $\theta_1 = 10^o C$

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**

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

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1 / 1
points

6.
Which of these is NOT a good learning rate decay scheme? Here, t is the epoch number.

- $\alpha = \frac{1}{1+2*t}\alpha_0$
- $\alpha = \frac{1}{\sqrt{t}}\alpha_0$
- $\alpha = 0.95^t \alpha_0$
- $\alpha = e^t \alpha_0$

**Correct**

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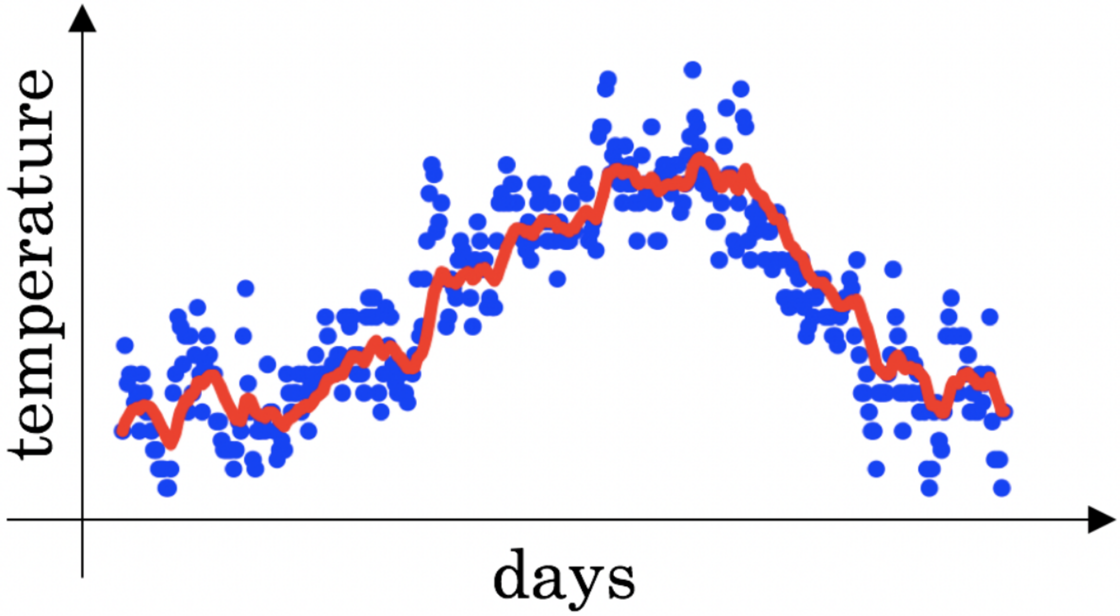
1 / 1
points

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.

**Un-selected is correct**

- [x] 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.

- [x] 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**

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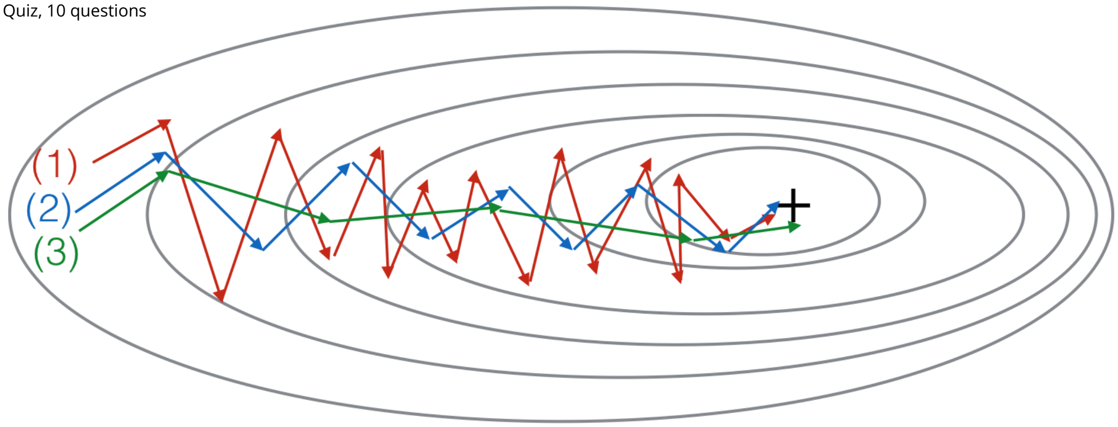
1 / 1
points

8.
Consider this figure:




These plots were generated with gradient descent; with gradient descent with momentum ($\beta$ *[Math Processing Error]* = 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 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 (small $\beta$ ). (3) is gradient descent with momentum (large $\beta$ )

**Correct**

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

- Try mini-batch gradient descent

**Correct**

- Try tuning the learning rate $\alpha$

**Correct**

- Try better random initialization for the weights

Try initializing all the weights to zero

**Un-selected is correct**

Try using Adam

**Correct**

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1 / 1
points

10.
Which of the following statements about Adam is False?

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

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

**Correct**

The learning rate hyperparameter $\alpha$ in Adam usually needs to be tuned.

Adam combines the advantages of RMSProp and momentum