Congratulations! You passed!

TO PASS 80% or higher

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Optimization algorithms

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1. Which notation would you use to denote the 3rd layer's activations when the input is the 7th example from the 8th minibatch?

1 / 1 point

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

Correct

2. Which of these statements about mini-batch gradient descent do you agree with?

1 / 1 point

- ( ) 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.
- ( ) 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).

Correct

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

1 / 1 point

- [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.

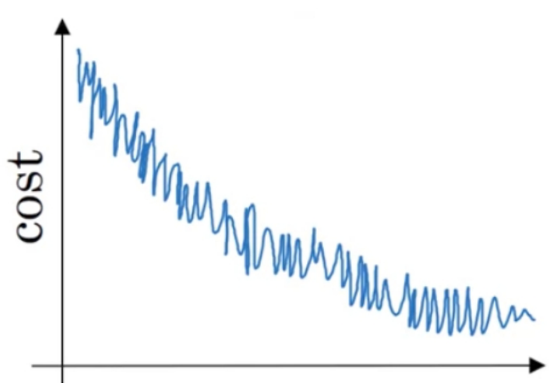
Correct

- [ ] If the mini-batch size is 1, you end up having to process the entire training set before making any progress.
- [ ] If the mini-batch size is m, you end up with stochastic gradient descent, which is usually slower than mini-batch gradient descent.
- [x] If the mini-batch size is 1, you lose the benefits of vectorization across examples in the mini-batch.

Correct

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

1 / 1 point



I

Which of the following do you agree with?

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

✓ **Correct**

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

1 / 1 point

Jan 1st: $\theta_1 = 10^oC$

Jan 2nd: $\theta_2 10^oC$

(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.)

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

✓ **Correct**

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

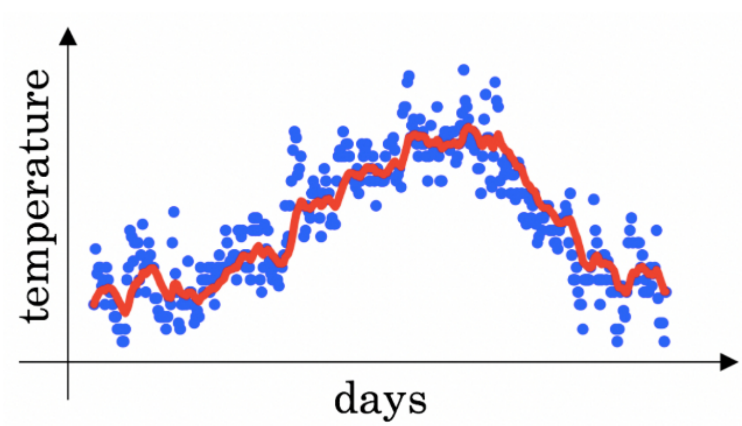
1 / 1 point

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

✓ **Correct**

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 β? (Check the two that apply)

1 / 1 point



- ☐ Decreasing β will shift the red line slightly to the right.

- [x] Increasing β will shift the red line slightly to the right.

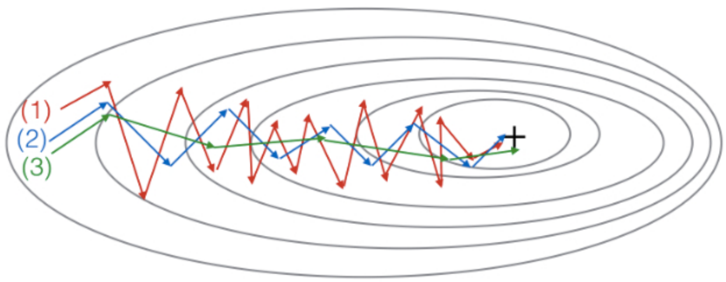
> **Correct**
> True, remember that the red line corresponds to $\beta = 0.9$. In lecture we had a green line $$\beta = 0.98) that is slightly shifted to the right.

- [x] Decreasing β 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 β will create more oscillations within the red line.

8. Consider this figure: **1 / 1 point**



These plots were generated with gradient descent; with gradient descent with momentum (β = 0.5) and gradient descent with momentum (β = 0.9). Which curve corresponds to which algorithm?

- ( ) (1) is gradient descent. (2) is gradient descent with momentum (large β) . (3) is gradient descent with momentum (small β)
- (•) (1) is gradient descent. (2) is gradient descent with momentum (small β). (3) is gradient descent with momentum (large β)
- ( ) (1) is gradient descent with momentum (small β). (2) is gradient descent. (3) is gradient descent with momentum (large β)
- ( ) (1) is gradient descent with momentum (small β), (2) is gradient descent with momentum (small β), (3) is gradient descent

> **Correct**

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) **1 / 1 point**

- [x] Try using Adam

> **Correct**

- [x] Try better random initialization for the weights

> **Correct**

- [x] Try tuning the learning rate α

> **Correct**

- [ ] Try initializing all the weights to zero
- [x] Try mini-batch gradient descent

> **Correct**

10. Which of the following statements about Adam is False? **1 / 1 point**

- ( ) We usually use "default" values for the hyperparameters β_1, β_2 and ε in Adam ($\beta_1 = 0.9$, $\beta_2 = 0.999$, $\varepsilon = 10^{-8}$)

- ◯ The learning rate hyperparameter α in Adam usually needs to be tuned.
- ◉ Adam should be used with batch gradient computations, not with mini-batches.
- ◯ Adam combines the advantages of RMSProp and momentum

✓ **Correct**