



✓ **Congratulations! You passed!**

TO PASS 80% or higher

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GRADE  
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## Optimization algorithms

LATEST SUBMISSION GRADE

100%

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

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

✓ Correct

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

1 / 1 point

- ☒ 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).
- ☐ Training one epoch (one pass through the training set) using mini-batch gradient descent is faster than training one epoch using batch gradient descent.

✓ Correct

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

1 / 1 point

- ☐ 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.
- ☒ 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.

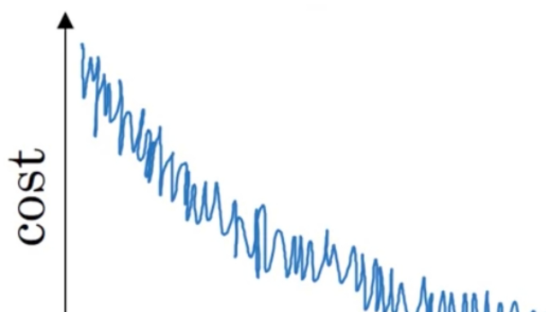
✓ Correct

- ☒ If the mini-batch size is 1, you 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





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

✓ Correct

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

1 / 1 point

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

Jan 2nd:  $\theta_2 = 10^\circ C$

(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.)

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

✓ Correct

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

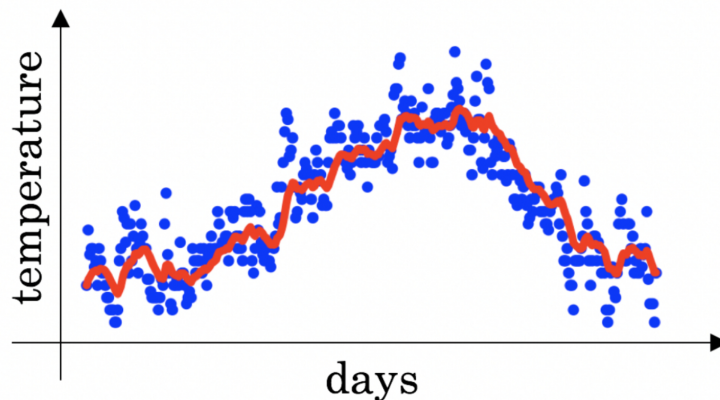
1 / 1 point

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

✓ 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  $\beta$ ? (Check the two that apply)

1 / 1 point



☐ Decreasing  $\beta$  will shift the red line slightly to the right.

☒ 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.

☒ Decreasing  $\beta$  will create more oscillation within the red line.

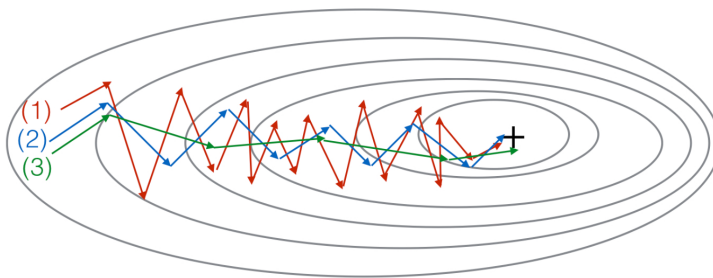
✓ Correct

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.

8. Consider this figure:

1 / 1 point



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

☒ (1) is gradient descent. (2) is gradient descent with momentum (small  $\beta$ ). (3) is gradient descent with momentum (large  $\beta$ )

☐ (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

✓ 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  $\mathcal{J}(W^{[1]}, b^{[1]}, \dots, W^{[L]}, b^{[L]})$ . 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

☒ Try better random initialization for the weights

✓ Correct

☒ Try using Adam

✓ Correct

☐ Try initializing all the weights to zero

☒ Try mini-batch gradient descent

✓ Correct

☒ Try tuning the learning rate  $\alpha$

✓ Correct

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

1 / 1 point

- ☐ We usually use "default" values for the hyperparameters  $\beta_1, \beta_2$  and  $\epsilon$  in Adam ( $\beta_1 = 0.9, \beta_2 = 0.999, \epsilon = 10^{-8}$ )
- ☒ Adam should be used with batch gradient computations, not with mini-batches.
- ☐ Adam combines the advantages of RMSProp and momentum
- ☐ The learning rate hyperparameter  $\alpha$  in Adam usually needs to be tuned.

✓ Correct