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

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

   ✓ Correct

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

   1 / 1 point

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

   - [x] If the mini-batch size is 1, you lose the benefits of vectorization across examples in the mini-batch.

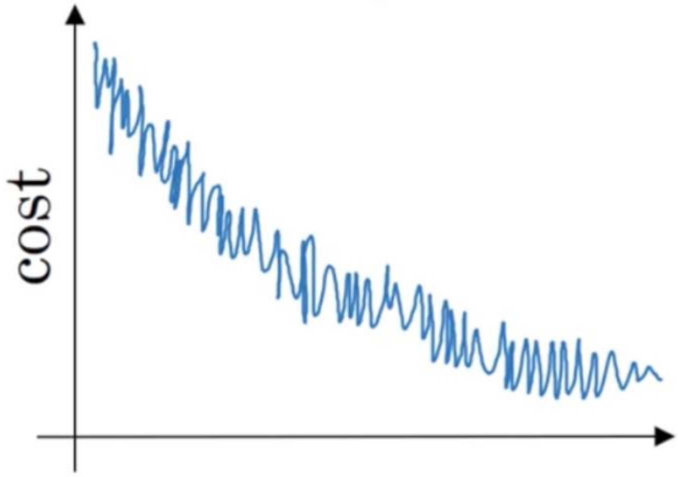
   ✓ Correct

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

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?

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

✓ **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.)

- (selected) $v_2 = 7.5$, $v_2^{corrected} = 10$
- $v_2 = 7.5$, $v_2^{corrected} = 7.5$
- $v_2 = 10$, $v_2^{corrected} = 10$
- $v_2 = 10$, $v_2^{corrected} = 7.5$

✓ **Correct**

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

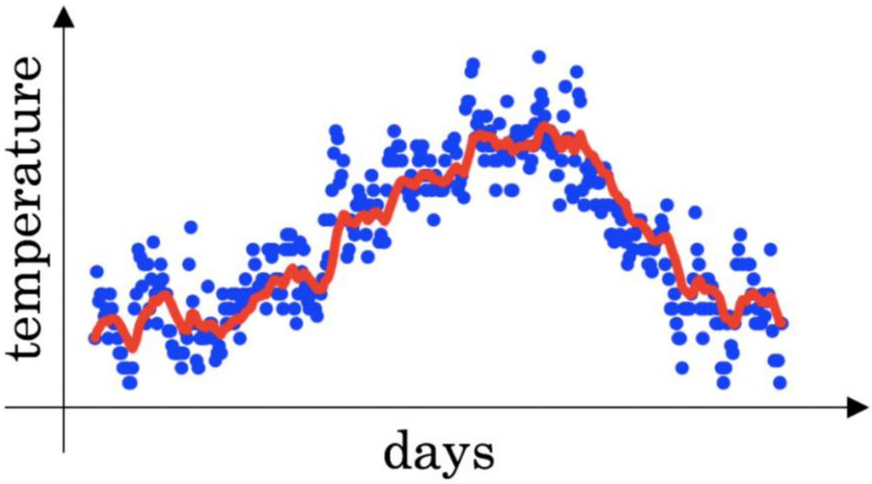
1 / 1 point

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

✓ 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.
- ☑ 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.

- ☑ Decreasing β will create more oscillation within the red line.

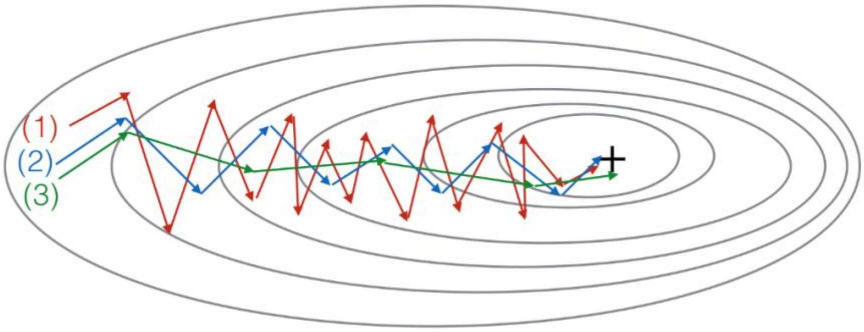
✓ Correct

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

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

1 / 1 point

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

✓ **Correct**

- [x] Try tuning the learning rate α

✓ **Correct**

- [x] Try using Adam

✓ **Correct**

- [x] Try better random initialization for the weights

✓ **Correct**

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

1 / 1 point

- ○ Adam combines the advantages of RMSProp and momentum
- ○ The learning rate hyperparameter α in Adam usually needs to be tuned.
- ○ We usually use "default" values for the hyperparameters β_1, β_2 and ε in Adam ($\beta_1 = 0.9$, $\beta_2 = 0.999$, $\varepsilon = 10^{-8}$)
- ◉ Adam should be used with batch gradient computations, not with mini-batches.

✓ Correct