Optimization algorithms

Teste, 10 questions

✔ **Parabéns! Você foi aprovado!**

✔ 1/1 pontos

1.

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

- ○ $a^{[3]\{7\}(8)}$

- ○ $a^{[8]\{3\}(7)}$

- ◉ $a^{[3]\{8\}(7)}$

   Correto

- ○ $a^{[8]\{7\}(3)}$

✔ 1/1 pontos

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.

- ○ 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).

- ○

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

Correto

✔ 1/1
pontos

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 end up having to process the entire training set before making any progress.

Não selecionado está correto

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

Correto

☐ 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.

Correto

☐ If the mini-batch size is m, you end up with stochastic gradient descent, which is usually slower than mini-batch gradient descent.

Não selecionado está correto

✔ 1/1
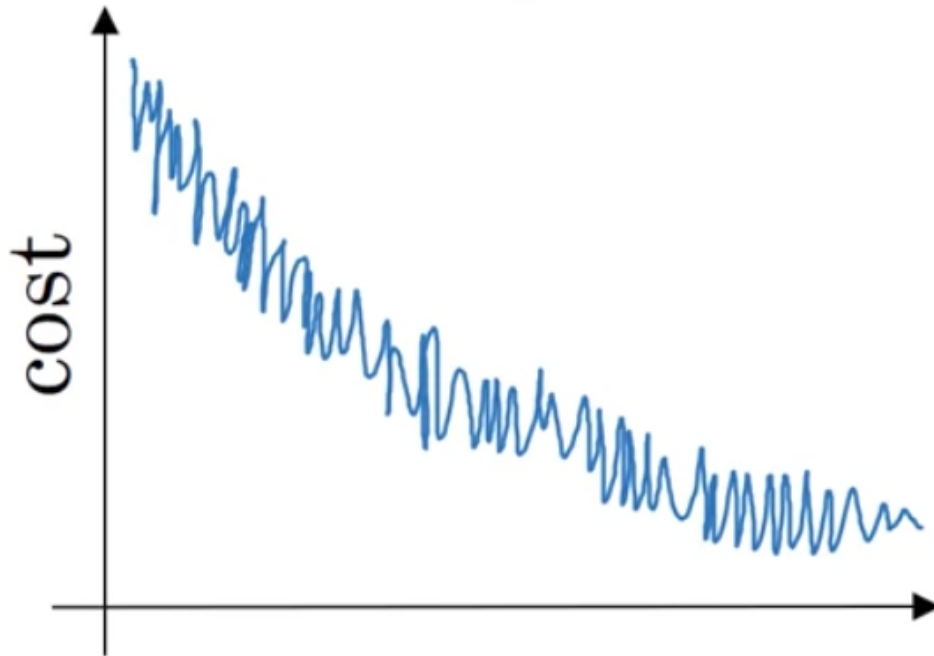pontos

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.

Correto

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

✔ 1/1 pontos

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.)

○ $v_2 = 10$, $v_2^{corrected} = 7.5$

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

○ $v_2 = 10$, $v_2^{corrected} = 10$

◉ $v_2 = 7.5$, $v_2^{corrected} = 10$

Correto

✔ 1/1 pontos

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

○ $\alpha = \frac{1}{1+2*t} \alpha_0$

◉ $\alpha = e^t \alpha_0$

Correto

○ $\alpha = 0.95^t \alpha_0$

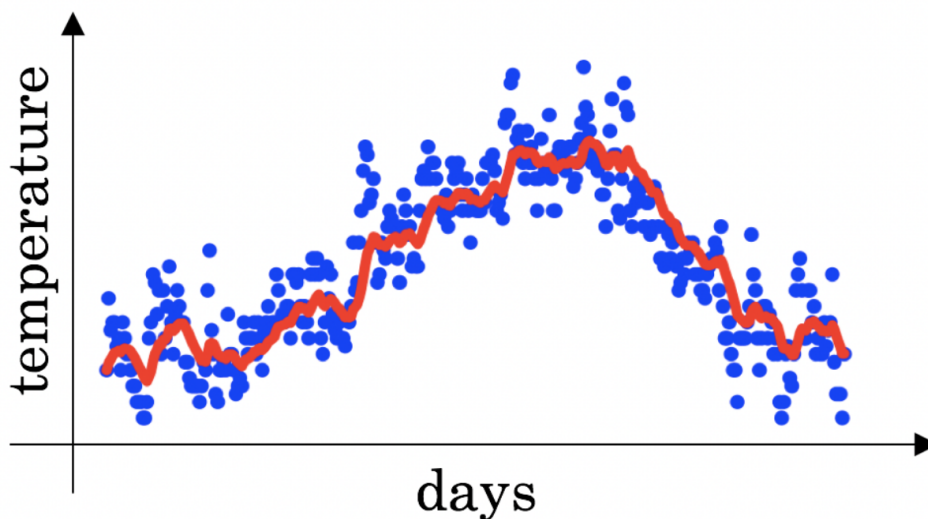○ $\alpha = \frac{1}{\sqrt{t}} \alpha_0$

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10/10 points (100%)

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)

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

Não selecionado está correto

☐ Increasing β will shift the red line slightly to the right.

Correto

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.

Correto

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.

Optimization algorithms

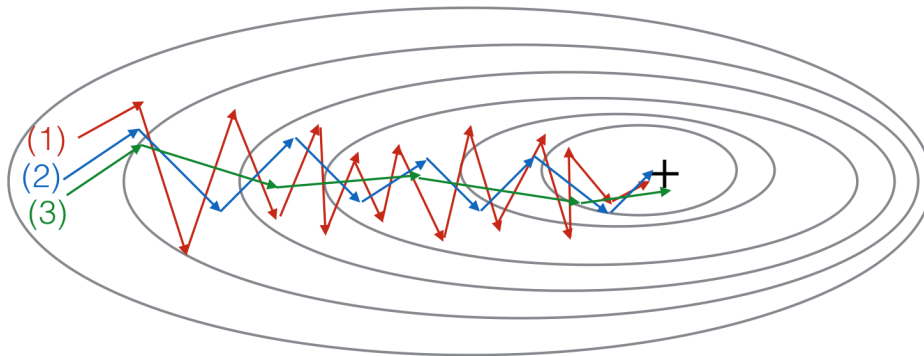**10/10 points (100%)**

Não selecionado está correto

Teste, 10 questions

✔ 1/1
pontos

8.
Consider this figure:

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 (small β). (3) is gradient descent with momentum (large β)

Correto

○ (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

○ (1) is gradient descent. (2) is gradient descent with momentum (large β) . (3) is gradient descent with momentum (small β)

✔ 1/1
pontos

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)

☑ Try mini-batch gradient descent

Correto

☑ Try tuning the learning rate α

Correto

☑ Try better random initialization for the weights

Correto

☑ Try using Adam

Correto

☐ Try initializing all the weights to zero

Não selecionado está correto

✔ 1/1
pontos

10.

Which of the following statements about Adam is False?

○ The learning rate hyperparameter α in Adam usually needs to be tuned.

○ Adam combines the advantages of RMSProp and momentum

○ 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.

Correto

10/10 points (100%)

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