

## Week 2: Optimization Algorithms

1. Which notation would you use to denote the 3rd layer activations when the input is the 7th example from the 8th mini-batch?

**Ans:**  $a^{[3]8(7)}$ .  $[i]j(k)$  superscript implies i-th layer, j-th minibatch, k-th example

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

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

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

**Ans:**

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

4. Suppose your cost J is plotted as a function of the number of iterations

**Ans:** If you are using mini-batch gradient descent, oscillations are acceptable. With batch gradient descent, something could be wrong

5. Suppose the temperature in Casablanca over the first three days of January are the same: Jan 1st:  $10^\circ\text{C}$ , Jan 2nd:  $10^\circ\text{C}$  Say you use an exponentially weighted average with  $\beta = 0.5$  to track the temperature:

$$v_0 = 0$$
$$v_t = \beta v_{t-1} + (1 - \beta)\theta_t$$

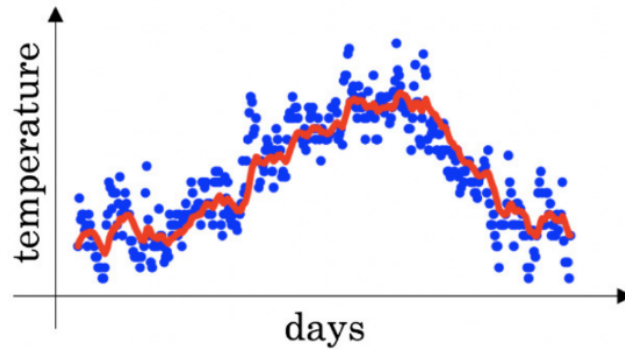
If  $v_2$  is the value computed after day-2 without bias correction, and  $v_{corrected,2}$  is the value you compute with bias correction. What are these values?

**Ans:**  $v_2 = 7.5^\circ\text{C}$ ,  $v_{corrected,2} = 10^\circ\text{C}$

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

**Ans:**  $\alpha = e^t \cdot \alpha_0$ . This will explode the rate instead of decaying.

7. You use an exponentially weighted average on the London temperature dataset. You use the following to track the temperature with  $\beta = 0.9$

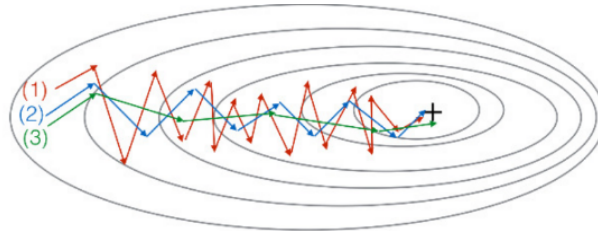


$$v_t = \beta v_{t-1} + (1 - \beta) \theta_t$$

What would happen to your moving average curve as you vary  $\beta$ ? **Ans:**

- Increasing  $\beta$  will shift the average line slightly to the right.
- Decreasing  $\beta$  will create more oscillation within the average line.

8. Which curves correspond to which algorithm? **Ans:** Gradient descent has



highest oscillation, gradient descent with small  $\beta$  for momentum will have some oscillations by few averaging, whereas, the least oscillation will come from gradient descent with large  $\beta$ . Hence,

- 1: GD,
- 2: GD + Momentum (small  $\beta$ ),
- 3: GD + Momentum (large  $\beta$ )

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

Which of the following techniques could help find parameter values that attain a small value for  $J$ ? (Check all that apply)

**Ans:**

- Try ADAM
- Try better random initialization for the weights
- Try tuning the learning rate  $\alpha$
- Try mini-batch gradient descent

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

**Ans:** ADAM should be used with batch gradient computations, not with mini-batches.