# CMPT 409/981: Optimization for Machine Learning

Lecture 20

Sharan Vaswani

November 19, 2024

- Adam:  $w_{k+1} = \Pi_C^k[w_k \eta_k A_k^{-1} m_k]$  where  $A_k = G_k^{\frac{1}{2}}$ ,  $G_0 = 0$  and  $G_k = \beta_2 G_{k-1} + (1 \beta_2) \nabla f_k(w_k) \nabla f_k(w_k)^{\mathsf{T}}$ ,  $m_k = \beta_1 m_{k-1} + (1 \beta_1) \nabla f_k(w_k)$ , for  $\beta_1, \beta_2 \in (0, 1)$ .
- Scalar Adam:  $v_k = \Pi_{\mathcal{C}} \left[ w_k \frac{\eta_k m_k}{\sqrt{\beta_2 G_{k-1} + (1-\beta_2) \|\nabla f_k(w_k)\|^2}} \right]$ ,  $w_{k+1} = \Pi_{\mathcal{C}}[v_k]$ , where  $G_0 = 0$  and  $m_k = \beta_1 m_{k-1} + (1-\beta_1) \nabla f_k(w_k)$ .
- For C>2, run scalar Adam with  $\beta_1=0$  (no momentum),  $\beta_2=\frac{1}{1+C^2}$  and  $\eta_k=\frac{\eta}{\sqrt{k}}$  such that  $\eta<\sqrt{1-\beta_2}$  on the following problem:
- ullet Consider  $\mathcal{C}=[-1,1]$  and the following sequence of linear functions.

$$f_k(w) = \begin{cases} C & w \text{ for } k \text{ mod } 3 = 1 \\ -w & \text{otherwise} \end{cases}$$

We will prove that Adam results in linear regret for the above example.

1

• **Update**:  $w_1 = 1$  and for  $k \ge 1$ ,

$$v_{k+1} := w_k - \frac{\eta_k}{\sqrt{\beta_2 \ G_{k-1} + (1-\beta_2) \left\|\nabla f_k(w_k)\right\|^2}} \nabla f_k(w_k) \text{ and } w_{k+1} = \Pi_{[-1,1]}[v_{k+1}]$$

• We will compare Adam to the "best" fixed decision ( $w^*$ ) that minimizes the regret. To compute  $w^*$ , consider the sequence of 3 functions from iteration 3k to 3k + 2 for  $k \ge 0$ . In this case,

$$w^* := \underset{[-1,1]}{\operatorname{arg\,min}} [f_{3k}(w) + f_{3k+1}(w) + f_{3k+2}(w)] = \underset{[-1,1]}{\operatorname{arg\,min}} [(C-2)w] = -1$$
 (Since  $C > 2$ )

**Claim**: For Adam's iterates, for  $k \ge 0$ , for all  $i \le [3k+1]$ ,  $w_i > 0$  and  $w_{3k+1} = 1$ .

**Proof**: Let us prove the statement by induction. Base case: For k = 0,  $w_{3k+1} = w_1 = 1$ .

**Inductive hypothesis**: Assume that for  $i \leq [3k+1]$ ,  $w_i > 0$  and  $w_{3k+1} = 1$ . We need to prove that (a)  $w_{3k+2} > 0$ , (b)  $w_{3k+3} > 0$  and (c)  $w_{3k+4} = 1$ .

In order to show this, note that  $\nabla f_i(w) = C$  for i mod 3 = 1 and  $\nabla f_i(w) = -1$  otherwise.

Consider the update at iteration (3k+1). By the induction hypothesis, we know that  $w_{3k+1}=1$ .

$$\begin{aligned} v_{3k+2} &= w_{3k+1} - \left[ \frac{\eta_{3k+1}}{\sqrt{\beta_2 \, G_{3k} + (1 - \beta_2) \, \|\nabla f_{3k+1}(w_{3k+1})\|^2}} \, \nabla f_{3k+1}(w_{3k+1}) \right] \\ &= 1 - \left[ \frac{C\eta}{\sqrt{(3k+1) \, (\beta_2 \, G_{3k} + (1 - \beta_2) \, C^2)}} \right] \qquad \text{(Using the value of } \eta_{3k+1}) \\ &\geq 1 - \left[ \frac{C\eta}{\sqrt{(3k+1) \, (1 - \beta_2) \, C^2}} \right] = 1 - \left[ \frac{\eta}{\sqrt{(3k+1) \, (1 - \beta_2)}} \right] \quad \text{(Since } G_{3k} \geq 0) \\ \implies v_{3k+2} > 1 - \frac{1}{\sqrt{3k+1}} > 0 \qquad \qquad \text{(Since } \eta < \sqrt{1 - \beta_2} \text{ and } k \geq 0) \end{aligned}$$

Since 
$$\left[\frac{C\eta}{\sqrt{(3k+1)(\beta_2 \ G_{3k}+(1-\beta_2)C^2)}}\right] > 0$$
,  $v_{3k+2} < 1$ . Since  $v_{3k+2} \in (0,1)$ ,  $w_{3k+2} = v_{3k+2} < 1$  which proves (a).

• For the update at iteration (3k+2), since  $\nabla f_{3k+2}(w) = -1$  for all w,

$$v_{3k+3} = w_{3k+2} + \left[ \frac{\eta}{\sqrt{(3k+2)(\beta_2 G_{3k+1} + (1-\beta_2))}} \right]$$

Since  $w_{3k+2} \in (0,1)$  and  $\frac{\eta}{\sqrt{(3k+2)(\beta_2 G_{3k+1}+(1-\beta_2))}} > 0$ ,  $v_{3k+3} > 0$  and hence  $w_{3k+3} > 0$  which proves (b).

• In order to prove (c), consider iteration 3k + 3. Since  $\nabla f_{3k+3}(w) = -1$  for all w,

$$v_{3k+4} = w_{3k+3} + \left[ \frac{\eta}{\sqrt{(3k+3)(\beta_2 G_{3k+2} + (1-\beta_2))}} \right]$$

From the above update, we can conclude that  $v_{3k+4} > w_{3k+3}$ .

To prove (c), we will show that  $v_{3k+4} \ge 1$  and hence  $w_{3k+4} = \Pi_{[-1,1]}v_{3k+4} = 1$ . For this, we consider two cases – when  $v_{3k+3} \ge 1$  or when  $v_{3k+3} < 1$ .

Case 1: When  $v_{3k+3} \ge 1 \implies w_{3k+3} = 1 \implies v_{3k+4} \ge 1 \implies w_{3k+4} = 1$ .

Case 2: When  $v_{3k+3} < 1 \implies w_{3k+3} = v_{3k+3} < 1$ . Combining iterations (3k+4) and (3k+3),

$$v_{3k+4} = v_{3k+3} + \left[ \frac{\eta}{\sqrt{(3k+3)(\beta_2 G_{3k+2} + (1-\beta_2))}} \right]$$

$$= w_{3k+2} + \left[ \frac{\eta}{\sqrt{(3k+2)(\beta_2 G_{3k+1} + (1-\beta_2))}} \right] + \left[ \frac{\eta}{\sqrt{(3k+3)(\beta_2 G_{3k+2} + (1-\beta_2))}} \right]$$

$$= 1 - \left[ \frac{C\eta}{\sqrt{(3k+1)(\beta_2 G_{3k} + (1-\beta_2)C^2)}} \right] \qquad \text{(Since } v_{3k+2} = w_{3k+2} \text{ and } w_{3k+1} = 1)$$

$$= 1 - \left[ \frac{\eta}{\sqrt{(3k+2)(\beta_2 G_{3k+1} + (1-\beta_2))}} \right] + \left[ \frac{\eta}{\sqrt{(3k+3)(\beta_2 G_{3k+2} + (1-\beta_2))}} \right]$$

In order to show that  $v_{3k+4} \ge 1$ , it is sufficient to show that  $T_1 \le T_2$ .

Recall from Slide 15,  $T_1 \leq \left[\frac{\eta}{\sqrt{(3k+1)(1-\beta_2)}}\right]$ . Let us lower-bound  $T_2$ .

$$T_{2} := \left[ \frac{\eta}{\sqrt{(3k+2)(\beta_{2} G_{3k+1} + (1-\beta_{2}))}} \right] + \left[ \frac{\eta}{\sqrt{(3k+3)(\beta_{2} G_{3k+2} + (1-\beta_{2}))}} \right]$$

$$\geq \left[ \frac{\eta}{\sqrt{(3k+2)(\beta_{2} C^{2} + (1-\beta_{2}))}} \right] + \left[ \frac{\eta}{\sqrt{(3k+3)(\beta_{2} C^{2} + (1-\beta_{2}))}} \right]$$
(Since  $G_{k} \leq C^{2}$  for all  $k$ )

$$= \frac{\eta}{\sqrt{(\beta_2 C^2 + (1 - \beta_2))}} \left[ \sqrt{\frac{1}{3k + 2}} + \sqrt{\frac{1}{3k + 3}} \right]$$

$$\geq \frac{\eta}{\sqrt{(\beta_2 C^2 + (1 - \beta_2))}} \left[ \sqrt{\frac{1}{2(3k + 1)}} + \sqrt{\frac{1}{2(3k + 1)}} \right] = \frac{\sqrt{2}\eta}{\sqrt{(\beta_2 C^2 + (1 - \beta_2))}} \left[ \frac{1}{\sqrt{3k + 1}} \right]$$

$$\implies T_2 \ge \left\lceil \frac{\eta}{\sqrt{(3k+1)(1-\beta_2)}} \right\rceil \ge T_1 \qquad \text{(Since } \beta_2 = \frac{1}{1+C^2} \implies \frac{\beta_2 C^2 + (1-\beta_2)}{2} = 1 - \beta_2 \text{)}$$

Since we have proved that  $T_2 \ge T_1$ ,  $v_{3k+4} = 1 - T_1 + T_2 \ge 1 \implies w_{3k+4} = 1$ . This completes the induction proof.

Hence, for the Adam iterates, for  $k \ge 0$ , for all  $i \le [3k+1]$ ,  $w_i > 0$  and  $w_{3k+1} = 1$ . Now that we have bounds on the Adam iterates, let us compute its regret  $R_{[3k \to 3k+2]}(w^*)$  w.r.t  $w^* = -1$  for iterations 3k to 3k + 2.

$$\begin{split} R_{[3k \to 3k+2]}(w^*) &= [f_{3k}(w_{3k}) - f_{3k}(-1)] + [f_{3k+1}(w_{3k+1}) - f_{3k+1}(-1)] + [f_{3k+2}(w_{3k+2}) - f_{3k+2}(-1)] \\ &= [-w_{3k} - 1] + [C \ w_{3k+1} + C] + [-w_{3k+2} - 1] > 2C - 4 > 0 \\ &\qquad \qquad \qquad \text{(Since } w_{3k} \text{ and } w_{3k+2} \text{ are in } (0,1), \ w_{3k+1} = 1 \text{ and } C > 2) \end{split}$$

- Hence for every three functions, Adam has a regret > 2C 4 and hence  $R_T(w^*) = O(T)$ .
- Both OGD and AdaGrad achieve sublinear regret when run on this example.

- The example takes advantage of the non-monotonicity in the Adam step-sizes resulting in smaller updates for  $k=1 \mod 3$  (when the gradient is positive and will push the iterates towards -1) and larger updates for the other k (when the gradient is negative and will push the iterates towards 1).
- In the example, as C>2 increases, the regret increases,  $\beta_2=\frac{1}{1+C^2}\to 0$ . [?] show that using a "large"  $\beta_2$  and ensuring that  $\beta_1\leq \sqrt{\beta_2}$  (often the choice in practice) can bypass the lower-bound resulting in convergence for Adam (without modifying the update).