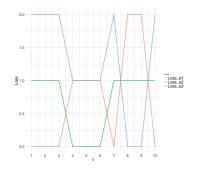
Advanced Machine Learning

Follow the leader for OQO problems





Learning goals

 Prove that FTL works for online quadratic optimization problems

FTL FOR OQO PROBLEMS

- One popular instantiation of the online learning problem is the problem of online quadratic optimization (OQO).
- In its most general form, the loss function is thereby defined as

$$L(a_t, z_t) = \frac{1}{2} ||a_t - z_t||_2^2,$$

where $\mathcal{A}, \mathcal{Z} \subset \mathbb{R}^d$.

ullet Proposition: Using FTL on any online quadratic optimization problem with $\mathcal{A}=\mathbb{R}^d$ and $V=\sup_{z\in\mathcal{Z}}||z||_2$, leads to a regret of

$$R_T^{\text{FTL}} \leq 4V^2 (\log(T) + 1).$$



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 - ullet In the following, we denote $a_1^{\mathrm{FTL}}, a_2^{\mathrm{FTL}}, \dots$ simply by a_1, a_2, \dots



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Reminder (Useful Lemma):

$$R_T^{\text{FTL}} \leq \sum\nolimits_{t = 1}^T {L(a_t^{\text{FTL}}, z_t)} - \sum\nolimits_{t = 1}^T {L(a_{t + 1}^{\text{FTL}}, z_t)}$$



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Using this lemma, we just have to show that

$$\sum_{t=1}^{T} \left(L(a_t, z_t) - L(a_{t+1}, z_t) \right) \le 4L^2 \cdot \left(\log(T) + 1 \right). \tag{1}$$



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 So, we will prove (1). For this purpose, we compute the explicit form of the actions of FTL for this type of online learning problem.



• Claim: It holds that $a_t = \frac{1}{t-1} \cdot \sum_{s=1}^{t-1} z_s$, if $L(a,z) = \frac{1}{2} ||a-z||_2^2$.



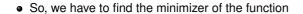
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 - Recall that

$$a_t^{\text{FTL}} = \arg\min_{a \in \mathcal{A}} \sum_{s=1}^{t-1} L(a, z_s) = \arg\min_{a \in \mathcal{A}} \sum_{s=1}^{t-1} \frac{1}{2} ||a - z_s||_2^2.$$



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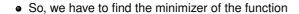


$$f(a) := \sum_{s=1}^{t-1} \frac{1}{2} ||a-z_s||_2^2 = \sum_{s=1}^{t-1} \frac{1}{2} (a-z_s)^{\top} (a-z_s).$$



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• Compute $\nabla f(a) = \sum_{s=1}^{t-1} a - z_s = (t-1)a - \sum_{s=1}^{t-1} z_s$, which we set to zero and solve with respect to a to obtain the claim.

(f is convex, so that this leads indeed to a minimizer.)



• Hence, a_t is the empirical average of z_1, \ldots, z_{t-1} and we can provide the following incremental update formula for its computation

$$a_{t+1} = \frac{1}{t} \cdot \sum_{s=1}^{t} z_s = \frac{1}{t} \left(z_t + \sum_{s=1}^{t-1} z_s \right)$$

= $\frac{1}{t} (z_t + (t-1)a_t) = \frac{1}{t} z_t + (1 - \frac{1}{t}) a_t$.



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• From the last display we derive that

$$a_{t+1} - z_t = (1 - \frac{1}{t}) \cdot a_t + \frac{1}{t}z_t - z_t = (1 - \frac{1}{t}) \cdot (a_t - z_t).$$



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· Claim:

$$L(a_t, z_t) - L(a_{t+1}, z_t) \le \frac{1}{t} \cdot ||a_t - z_t||_2^2.$$
 (2)

Reminder:
$$a_{t+1} - z_t = \left(1 - \frac{1}{t}\right) \cdot \left(a_t - z_t\right)$$
.

Indeed, this can be seen as follows

$$L(a_{t}, z_{t}) - L(a_{t+1}, z_{t}) = \frac{1}{2} ||a_{t} - z_{t}||_{2}^{2} - \frac{1}{2} ||a_{t+1} - z_{t}||_{2}^{2}$$

$$= \frac{1}{2} \left(||a_{t} - z_{t}||_{2}^{2} - ||a_{t+1} - z_{t}||_{2}^{2} \right)$$

$$= \frac{1}{2} \left(||a_{t} - z_{t}||_{2}^{2} - ||(1 - \frac{1}{t}) \cdot (a_{t} - z_{t})||_{2}^{2} \right).$$



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· And from this,

$$L(a_{t}, z_{t}) - L(a_{t+1}, z_{t}) = \frac{1}{2} \left(||a_{t} - z_{t}||_{2}^{2} - \left(1 - \frac{1}{t}\right)^{2} \cdot ||a_{t} - z_{t}||_{2}^{2} \right)$$

$$= \frac{1}{2} \left(1 - \left(1 - \frac{1}{t}\right)^{2} \right) \cdot ||a_{t} - z_{t}||_{2}^{2}$$

$$= \left(\frac{1}{t} - \frac{1}{2t^{2}} \right) \cdot ||a_{t} - z_{t}||_{2}^{2}$$

$$\leq \frac{1}{t} \cdot ||a_{t} - z_{t}||_{2}^{2}.$$





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• Since by assumption $L = \sup_{z \in \mathcal{Z}} ||z||_2$ and a_t is the empirical average of z_1, \ldots, z_{t-1} , we have that $||a_t||_2 \le L$.



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- Now the triangle inequality states that for any two vectors $x,y\in\mathbb{R}^d$ it holds that

$$||x + y||_2 \le ||x||_2 + ||y||_2$$

so that

$$||a_t - z_t||_2 \le ||a_t||_2 + ||z_t||_2 \le 2L.$$
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• Summing over all t in (2) and using (3) we arrive at

$$\sum_{t=1}^{T} (L(a_t, z_t) - L(a_{t+1}, z_t)) \le \sum_{t=1}^{T} \left(\frac{1}{t} \cdot ||a_t - z_t||_2^2 \right) \le \sum_{t=1}^{T} \frac{1}{t} \cdot (2L)^2$$

$$= 4L^2 \cdot \sum_{t=1}^{T} \frac{1}{t}.$$



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• Now, it holds that $\sum_{t=1}^{I} \frac{1}{t} \leq \log(T) + 1$, so that we obtain

$$\sum_{t=1}^{T} \left(L(a_t, z_t) - L(a_{t+1}, z_t) \right) \leq 4L^2 \cdot \sum_{t=1}^{T} \frac{1}{t} \leq 4L^2 \cdot \left(\log(T) + 1 \right),$$

which is what we wanted to prove.