Friederike Horn & Bileam Scheuvens Justify all your claims.

Online Gradient Descent

Solution

a) The pythagorean inequality becomes apparent when we expand by \hat{y} :

$$||y - u||^2 = ||y - \hat{y} + \hat{y} - u||^2$$
$$= ||(y - \hat{y}) + (\hat{y} - u)||^2$$

This way we can apply the parallelogram law $||a+b||^2 = ||a||^2 + ||b||^2 + 2\langle a,b\rangle$:

$$= ||y - \hat{y}||^2 + ||\hat{y} - u||^2 + 2\langle y - \hat{y}, \hat{y} - u\rangle$$

Since the $\langle a, b \rangle \geq 0$:

$$\geq ||y - \hat{y}||^2 + ||\hat{y} - u||^2$$

b) Consider the distance between the w_{t+1} and u:

$$||w_{t+1} - u||^2$$

We can expand w_{t+1} with our step rule before applying P_U

$$= ||w_t - \eta \nabla f_t(w_t) - u||^2 - \epsilon$$

Where ϵ describes the nonnegative length lost in the projection. Without ϵ clearly:

$$||w_{t+1} - u||^{2} \le ||w_{t} - \eta \nabla f_{t}(w_{t}) - u||^{2}$$

$$= ||(w_{t} - u) + (-\eta \nabla f_{t}(w_{t}))||^{2}$$

$$= ||(w_{t} - u)||^{2} + ||(-\eta \nabla f_{t}(w_{t}))||^{2} + 2\langle w_{t} - u, -\eta \nabla f_{t}(w_{t})\rangle$$

$$= ||(w_{t} - u)||^{2} + \eta^{2}||\nabla f_{t}(w_{t})||^{2} - 2\eta\langle w_{t} - u, \nabla f_{t}(w_{t})\rangle$$

Remember the term we started with:

$$||w_{t+1} - u||^2 \le ||(w_t - u)||^2 + \eta^2 ||\nabla f_t(w_t)||^2 - 2\eta \langle w_t - u, \nabla f_t(w_t) \rangle$$
$$2\eta \langle w_t - u, \nabla f_t(w_t) \rangle \le ||(w_t - u)||^2 + \eta^2 ||\nabla f_t(w_t)||^2 - ||w_{t+1} - u||^2$$

Divide by 2η :

$$\langle w_t - u, \nabla f_t(w_t) \rangle \le \frac{||(w_t - u)||^2 - ||w_{t+1} - u||^2}{2\eta} + \frac{\eta}{2} ||\nabla f_t(w_t)||^2$$

c) From convexity we know that a (differentiable) function stays above any linear local approximation, that is to say for x, y in the domain of f:

$$f(x) \ge f(y) + \langle \nabla f(y), x - y \rangle$$

$$\Leftrightarrow -\langle \nabla f(y), x - y \rangle \ge f(y) - f(x)$$

$$\Leftrightarrow \langle \nabla f(y), y - x \rangle \ge f(y) - f(x)$$

Translating to our problem, we obtain:

$$f_t(w_t) - f_t(u) \le \langle \nabla f_t(w_t), w_t - u \rangle$$

We can replace the right side with our inequality from b):

$$f_t(w_t) - f_t(u) \le \frac{||(w_t - u)||^2 - ||w_{t+1} - u||^2}{2\eta} + \frac{\eta}{2}||\nabla f_t(w_t)||^2$$

The sum $\sum_{t=1}^{T} f_t(w_t) - f_t(u)$ can then be decomposed to examine both sums on the right hand side individually:

$$\sum_{t=1}^{T} \frac{||(w_t - u)||^2 - ||w_{t+1} - u||^2}{2\eta} = \frac{1}{2\eta} \sum_{t=1}^{T} ||(w_t - u)||^2 - ||w_{t+1} - u||^2$$

It can be seen that this sum is telescopic, that is, it expands to:

$$\frac{1}{2\eta}(||w_1-u||^2-||w_2-2||^2+||w_2-u||^2-||w_3-u||^2+||w_3-u||^2+\dots)$$

leaving only the first and last term.

$$= \frac{||w_1 - u||^2 - ||w_{T+1} - u||^2}{2\eta}$$

Since $w_1 = 0$, $||u||^2 \le D$ we obtain:

$$\leq \frac{D^2 - ||w_{T+1} - u||^2}{2\eta} \\ \leq \frac{D^2}{2\eta}$$

Returning to the original expression and examining the other summand:

$$\frac{\eta}{2} \sum_{t=1}^{T} ||\nabla f_t(w_t)||^2$$

Since that $\nabla |f_t| \leq G$:

$$\leq \frac{\eta}{2} \sum_{t+1}^{T} G^2$$
$$= \frac{\eta}{2} T G^2$$

This concludes the proof that:

$$\max_{u \in U} \sum_{t=1}^{T} f_t(w_t) - f_t(u) \le \frac{D^2}{2\eta} + \frac{\eta}{2} TG^2$$

d) To find the optimal η we investigate the upper bound for critical points.

$$\frac{d}{d\eta} \left(\frac{D^2}{2\eta} + \frac{\eta}{2} T G^2 \right) \stackrel{!}{=} 0$$

$$\frac{-D^2}{2\eta^2} + \frac{T G^2}{2} = 0$$

$$\frac{T G^2}{2} = \frac{D^2}{2\eta^2}$$

$$\eta^2 T G^2 = D^2$$

$$\eta^2 = \frac{D^2}{T G^2}$$

$$\eta = \frac{D}{\sqrt{T} G}$$

This is a minimum, since the second derivative is $\frac{D^2}{6\eta^3}$, which is positive. Inserting this into the original equation yields a worst case regret of:

$$\frac{D^2}{2\left(\frac{D}{\sqrt{T}G}\right)} + \frac{\left(\frac{D}{\sqrt{T}G}\right)}{2}TG^2$$

$$= \frac{D^2\sqrt{T}G}{2D} + \frac{D}{2\sqrt{T}G}TG^2$$

$$= \frac{D\sqrt{T}G}{2} + \frac{D\sqrt{T}G}{2}$$

$$= D\sqrt{T}G$$

Formal proof of the existence of Lagrange Multipliers Solution

- a)
- b)
- c)

Differentiable approximation of \mathcal{L}^1 -approximation Solution

- a)
- b)