

# CSIT 5500 Advanced Algorithms

2020 Spring Semester

## Written Assignment 5 solution

1. (10 points)

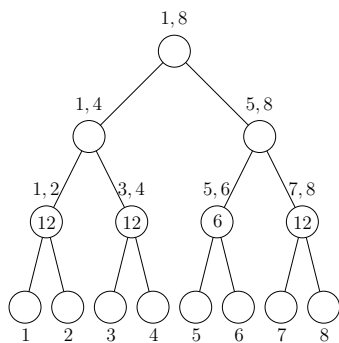
The final array  $C$ :

	0	1	2
0	13	14	15
1	13	15	14

$a$	$\hat{f}_a$
0	14
1	13
2	15
3	14
4	13
5	15
6	14
7	13
8	15
9	14

2. (10 points)

(a)



(b) We stop at entry  $[3, 4]$  because  $12 + 12 > 20$ , and we report 4.

3. (10 points)

- $\nabla f = \begin{bmatrix} 2x \\ 2y \end{bmatrix}$ .

$$\langle \nabla f(x_t), z \rangle = 2x_{t_x}z_x + 2x_{t_y}z_y.$$

- $D$  is the convex hull of  $P$ , so  $\operatorname{argmin}_{z \in D} \langle \nabla f(x_t), z \rangle$  can be obtained by examining all vertices of  $P$ .

•

$$\begin{aligned}
& -\langle \nabla f(x_t), z_t - x_t \rangle \\
&= -2[x_{t_x}(z_{t_x} - x_{t_x}) + x_{t_y}(z_{t_y} - x_{t_y})] \\
&= -2(x_{t_x}z_{t_x} - x_{t_x}^2 + x_{t_y}z_{t_y} - x_{t_y}^2) \\
&= 2(x_{t_x}^2 - x_{t_x}z_{t_x} - x_{t_y}z_{t_y} + x_{t_y}^2).
\end{aligned}$$

Stopping condition is  $\langle \nabla f(x_t), z_t - x_t \rangle \geq 0$  or  $|\langle \nabla f(x_t), z_t - x_t \rangle| \leq \epsilon f(x_t)/(1 + \epsilon)$ . Below is the proof.

Let  $p^*$  denote the point in  $D$  so that  $f(p^*, p_y^*)$  is the minimum.

By convexity of  $f$ , we have  $f(p^*) \geq f(x_t) + \langle \nabla f(x_t), p^* - x_t \rangle$ .

By Lemma 2 in the lecture notes, we have  $f(p^*) \geq f(x_t) + \langle \nabla f(x_t), z_t - x_t \rangle$ .

If  $\langle \nabla f(x_t), z_t - x_t \rangle \geq 0$ , the only possibility is  $\langle \nabla f(x_t), z_t - x_t \rangle = 0$ , and thus  $f(p^*) = f(x_t)$ .

If  $\langle \nabla f(x_t), z_t - x_t \rangle < 0$ , then  $\langle \nabla f(x_t), z_t - x_t \rangle \geq -\epsilon f(x_t)/(1 + \epsilon)$  when the algorithm stops. We have

$$\begin{aligned}
& f(p^*) \geq f(x_t) + \langle \nabla f(x_t), z_t - x_t \rangle \\
& \rightarrow f(p^*) \geq f(x_t) - \epsilon f(x_t)/(1 + \epsilon) \\
& \rightarrow f(p^*) \geq f(x_t)/(1 + \epsilon) \\
& \rightarrow (1 + \epsilon)f(p^*) \geq f(x_t)
\end{aligned}$$

- $f(x_t + \rho(z_t - x_t)) = (x_{t_x} + \rho(z_{t_x} - x_{t_x}))^2 + (x_{t_y} + \rho(z_{t_y} - x_{t_y}))^2$   
Take the derivative of it with respect to  $\rho$ , we have

$$\begin{aligned}
& 2(x_{t_x} + \rho(z_{t_x} - x_{t_x}))(z_{t_x} - x_{t_x}) + 2(x_{t_y} + \rho(z_{t_y} - x_{t_y}))(z_{t_y} - x_{t_y}) \\
&= 2x_{t_x}(z_{t_x} - x_{t_x}) + 2\rho(z_{t_x} - x_{t_x})^2 + 2x_{t_y}(z_{t_y} - x_{t_y}) + 2\rho(z_{t_y} - x_{t_y})^2 \\
&= 2\rho((z_{t_x} - x_{t_x})^2 + (z_{t_y} - x_{t_y})^2) + 2x_{t_x}(z_{t_x} - x_{t_x}) + 2x_{t_y}(z_{t_y} - x_{t_y})
\end{aligned}$$

By setting  $2\rho((z_{t_x} - x_{t_x})^2 + (z_{t_y} - x_{t_y})^2) + 2x_{t_x}(z_{t_x} - x_{t_x}) + 2x_{t_y}(z_{t_y} - x_{t_y})$  to zero, we have

$$\begin{aligned}
& 2\rho((z_{t_x} - x_{t_x})^2 + (z_{t_y} - x_{t_y})^2) + 2x_{t_x}(z_{t_x} - x_{t_x}) + 2x_{t_y}(z_{t_y} - x_{t_y}) = 0 \\
& \rightarrow 2\rho((z_{t_x} - x_{t_x})^2 + (z_{t_y} - x_{t_y})^2) = -2x_{t_x}(z_{t_x} - x_{t_x}) - 2x_{t_y}(z_{t_y} - x_{t_y}) \\
& \rightarrow \rho = -(x_{t_x}(z_{t_x} - x_{t_x}) + x_{t_y}(z_{t_y} - x_{t_y})) / ((z_{t_x} - x_{t_x})^2 + (z_{t_y} - x_{t_y})^2)
\end{aligned}$$

$$\text{Let } \rho^* = -\frac{(x_{t_x}(z_{t_x} - x_{t_x}) + x_{t_y}(z_{t_y} - x_{t_y}))}{((z_{t_x} - x_{t_x})^2 + (z_{t_y} - x_{t_y})^2)}.$$

We have

$$\begin{cases} \operatorname{argmin}_{\rho \in [0,1]} f(x_t + \rho(z_t - x_t)) = \rho^* & \text{if } 0 \leq \rho^* \leq 1. \\ \operatorname{argmin}_{\rho \in [0,1]} f(x_t + \rho(z_t - x_t)) = \operatorname{argmin}_{\rho \in \{0,1\}} f(x_t + \rho(z_t - x_t)) & \text{Otherwise.} \end{cases}$$

- For every pair of points  $p, q \in D$ ,

$$\begin{aligned} & \|\nabla f(p) - \nabla f(q)\| \leq L\|p - q\| \\ \rightarrow & \ 2\|p - q\| \leq L\|p - q\| \\ \rightarrow & \ 2 \leq L. \end{aligned}$$