

# EE578B - Convex Optimization - Winter 2021

## Homework 8

**Due Date:** Sunday, Mar 14<sup>th</sup>, 2021 at 11:59 pm

### 1. Newton's Method

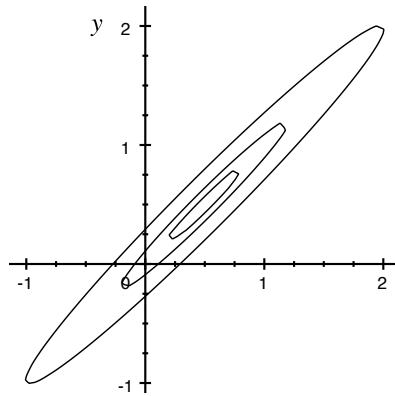
Consider the following unconstrained quadratic program.

$$\min_x f(x) = \frac{1}{2}x^T Qx + c^T x$$

for  $x \in \mathbb{R}^2$

$$Q = \begin{bmatrix} 101 & -99 \\ -99 & 101 \end{bmatrix}, \quad c = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

Note that the level sets look something like



- **(PTS:0-2)** Perform regular first order gradient descent using the update equation

$$x^+ = x - \gamma \frac{\partial f}{\partial x}^T$$

starting from the initial condition  $x = \begin{bmatrix} 10 & 0 \end{bmatrix}^T$  with a fixed step size  $\gamma$ . (Pick the step size.)  
Note: if you want you can use a more sophisticated step-size method.

- **(PTS:0-2)** Plot the trajectory of  $x$  and describe the behavior intuitively.
- **(PTS:0-2)** Perform Newton's Method starting from the same initial condition with the same step size.

$$x^+ = x - \gamma \left( \frac{\partial^2 f}{\partial x^2} \right)^{-1} \frac{\partial f}{\partial x}^T$$

- **(PTS:0-2)** Plot the trajectory of  $x$  and compare the qualitative performance to the first-order gradient descent method.

Now consider the following constrained convex program

$$\begin{aligned} \min_x \quad & 10x_1^4 + 2x_2^4 + 2x_3^4 + 2x_4^4 \\ \text{s.t.} \quad & Ax = b \end{aligned}$$

for  $x \in \mathbb{R}^4$  and

$$A = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix}, \quad b = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

- **(PTS:0-2)** Use Newton's method to perform gradient descent on this constrained optimization problem to solve for the optimal  $x \in \mathbb{R}^n$  and the optimal dual variable  $v \in \mathbb{R}^2$
- **(PTS:0-2)** Compare  $\frac{\partial f}{\partial x}$  and  $v^T A$  at optimum. How do they relate?

## 2. Interior Point Method

Consider the constrained optimization problem

$$\begin{aligned} \min_x \quad & f(x) = 10x_1^4 + x_2^4 \\ \text{s.t.} \quad & h(x) \leq 0 \end{aligned}$$

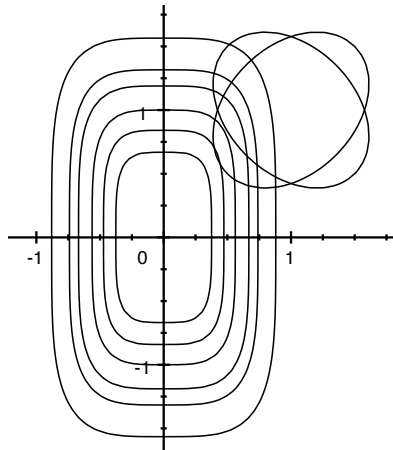
for  $x \in \mathbb{R}^2$

$$h(x) = \begin{bmatrix} (x - \mathbf{1})^T Q_1 (x - \mathbf{1}) - 1 \\ (x - \mathbf{1})^T Q_2 (x - \mathbf{1}) - 1 \end{bmatrix}$$

where  $\mathbf{1}^T = [1 \ 1]$

$$Q_1 = \begin{bmatrix} 3 & 1 \\ 1 & 3 \end{bmatrix}, \quad Q_2 = \begin{bmatrix} 3 & -1 \\ -1 & 3 \end{bmatrix}$$

The constraint level sets and objective level sets are shown in the following figure.



- **(PTS:0-2)** Replace the objective function  $f(x)$  with  $tf(x)$  for some  $t > 0$ . Replace the inequality constraints with equality constraints of the form  $h_i(x) = s_i$  and add barrier function terms of the form  $\mu \ln(s_i)$  to the objective for some  $\mu > 1$ .

- **(PTS:0-2)** Write the Lagrangian for this new optimization problem (with barrier functions) with dual variables  $v \in \mathbb{R}^2$  for the equality constraints.
- **(PTS:0-4)** Write code to perform Newton's method for gradient descent to solve for the optimal  $x \in \mathbb{R}^2$ ,  $s \in \mathbb{R}^2$ , and  $v \in \mathbb{R}^2$  for a given value of  $t$ .
- **(PTS:0-4)** For each value of  $t$ , run Newton's method till  $|x - x^+| < \delta$  for some tolerance  $\delta > 0$ . After  $x$  converges, update  $t$  as  $t^+ = \mu t$  and repeat solving for  $x$ . (Note, it's important that  $\mu > 1$  so that  $t$  will grow, ie. the objective gets more weight as you approach the boundary.) Iterate on this process till the optimal  $x$  is found. How does this method perform for different values of  $\mu$ ?

### 3. Simplex Method - Row Geometry

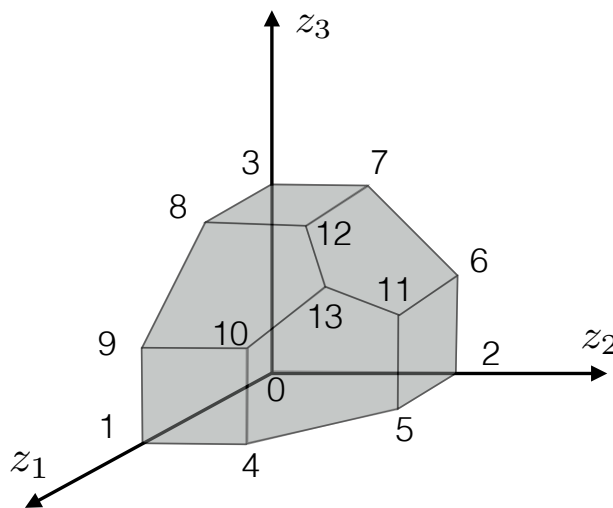
Consider the following linear program for  $z \in \mathbb{R}^3$

$$\begin{aligned} \max_x \quad & c^T z \\ \text{s.t.} \quad & Cz \leq d, \quad x \geq 0 \end{aligned}$$

where

$$c = \begin{bmatrix} 1 & 1 & 1 \end{bmatrix}, \quad C = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix}, \quad d = \begin{bmatrix} 2 \\ 2 \\ 2 \\ 3 \\ 3 \\ 3 \end{bmatrix}$$

Note that this set is the inside of the following geometric shape (with the vertices numbered).



- **(PTS:0-2)** Use a slack variable  $s \in \mathbb{R}^6$  to rewrite the LP in standard form for the simplex method

$$\begin{aligned} \max_x \quad & r^T x \\ \text{s.t.} \quad & Ax = b, x \geq 0 \end{aligned}$$

What is  $x$ ?  $A$ ?  $b$ ? What feasible  $x$  corresponds to  $z = 0$ ?

- **(PTS:0-2)** Write a tableau for the linear program in the form

$$\begin{bmatrix} 1 & -r^T & 0 \\ \mathbf{0} & A & b \end{bmatrix}$$

- **(PTS:0-4)** Starting at the initial solution  $z = 0$  (vertex 0), perform pivot steps to find the optimal solution to the linear program. What is the optimal  $x$ ? What is the corresponding optimal  $z$ ? Which rows of the constraint  $Cz \leq d$  are satisfied with equality?
- **(PTS:0-2)** What route did you follow through the polytope? You can list the route referencing the vertex numbers.

#### 4. Simplex Method - Column Geometry

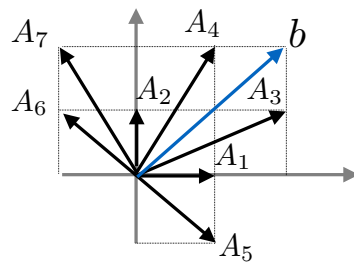
Consider the following linear program for  $x \in \mathbb{R}^7$

$$\begin{aligned} \max_x \quad & r^T x \\ \text{s.t.} \quad & Ax = b, x \geq 0 \end{aligned}$$

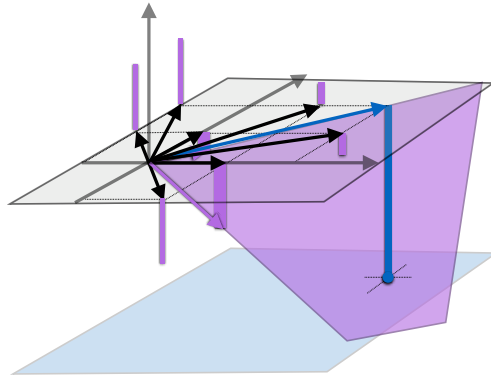
where

$$A = [A_1 \quad \cdots \quad A_7] = \begin{bmatrix} 1 & 0 & 2 & 1 & 1 & -1 & -1 \\ 0 & 1 & 1 & 2 & -1 & 1 & 2 \end{bmatrix}$$

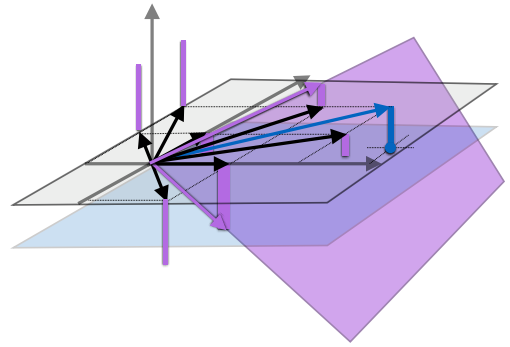
- **(PTS:0-2)** Draw the columns of  $A$  as vectors in  $\mathbb{R}^2$ .
- **(PTS:0-2)** Suppose  $b = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$ . Find all possible pairs of basis vectors ( $A_i$  and  $A_{i'}$ ) such that  $\begin{bmatrix} A_i & A_{i'} \end{bmatrix} \begin{bmatrix} x_i \\ x_{i'} \end{bmatrix} = b$ , for  $x \geq 0$ . (Hint: there are 4 pairs. Drawing  $b$  with the columns of  $A$  may help.)
- **(PTS:0-2)** Suppose  $b = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$ . Find all possible pairs of basis vectors ( $A_i$  and  $A_{i'}$ ) such that  $\begin{bmatrix} A_i & A_{i'} \end{bmatrix} \begin{bmatrix} x_i \\ x_{i'} \end{bmatrix} = b$ , for  $x \geq 0$ . (Hint: there are 7 pairs. Drawing  $b$  with the columns of  $A$  may help.)
- **(PTS:0-4)** Now consider the reward vector  $r^T = [-3 \quad -1 \quad -1 \quad 1 \quad -3 \quad 3 \quad 3]$  for  $b^T = [2 \quad 2]^T$ . Write the tableau for the linear program to maximize  $r^T x$ . Perform the pivot steps shown in the following illustrations.



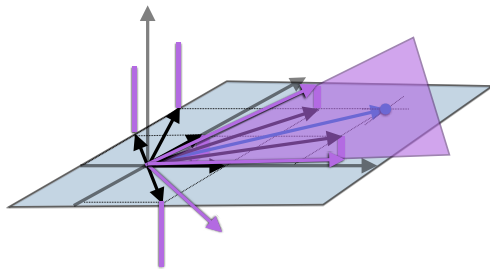
**Start:**



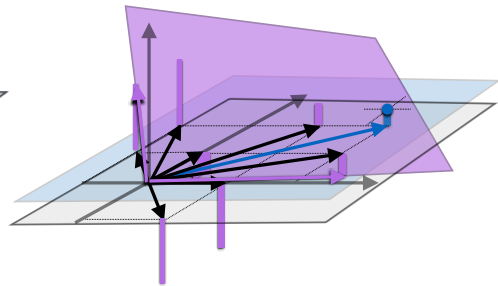
**Step 1:**



**Step 2:**

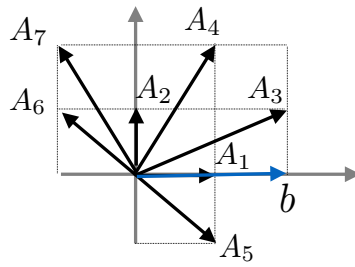


**Step 3:**

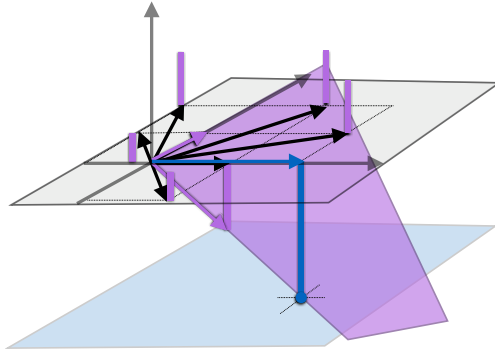


What is the optimal  $x$  and  $r^T x$ ?

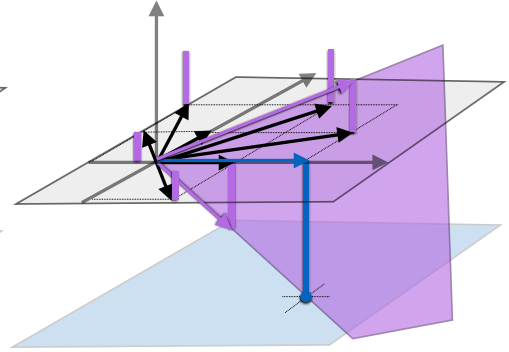
- **(PTS:0-4)** Now consider the reward vector  $r^T = [-3 \ 0 \ 1 \ 2 \ 1 \ -1 \ 2]$  for  $b^T = [2 \ 0]^T$ . Write the tableau for the linear program to maximize  $r^T x$ . Perform the pivot steps shown in the following illustrations.



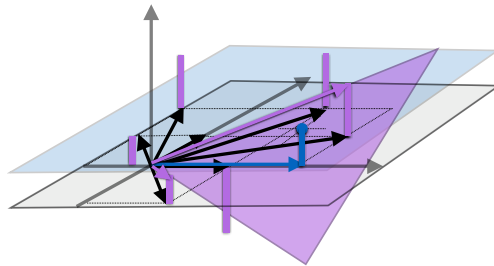
**Start:**



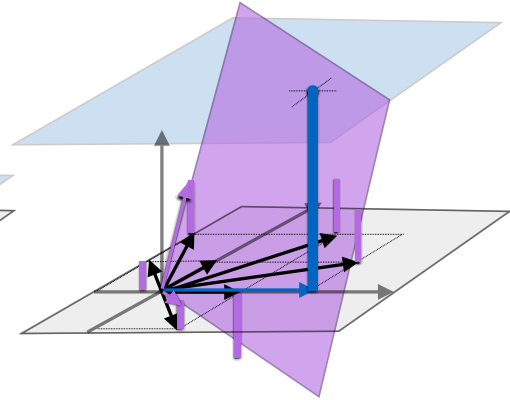
**Step 1:**



**Step 2:**



**Step 3:**



What is the optimal  $x$  and  $r^T x$ ?

- (PTS:0-2) Which individual  $x_i$ 's could correspond to the positive and negative part of a single unconstrained variable?