

AC-3
August 2015 QE
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1. (20 pts) Solve the following linear program,

$$\begin{array}{ll}\text{maximize} & -x_1 - 2x_2 + 4x_3 \\ \text{subject to} & x_1 + 2x_2 - x_3 = 5 \\ & 2x_1 + 3x_2 - x_3 = 6 \\ & x_1 \text{ free}, x_2 \geq 0, x_3 \leq 0\end{array}$$

$$\left(\begin{array}{ccc|c} 1 & 2 & -1 & 5 \\ 2 & 3 & -1 & 6 \end{array}\right) \rightarrow \left(\begin{array}{ccc|c} 1 & 2 & -1 & 5 \\ 0 & 1 & -1 & 4 \end{array}\right) \rightarrow \left(\begin{array}{ccc|c} 1 & 1 & 0 & 1 \\ 0 & 1 & -1 & 4 \end{array}\right)$$

The answer is

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} \alpha \\ 1 - \alpha \\ -3 - \alpha \end{bmatrix}$$

So the function to be maximized is

$$-\alpha - 2(1 - \alpha) + 4(-3 - \alpha) = -14 - 3\alpha$$

Since $x_2 \geq 0$ and $x_3 \leq 0$,

$$\begin{cases} 1 - \alpha \geq 0 \\ -3 - \alpha \leq 0 \end{cases} \rightarrow \begin{cases} \alpha \leq 1 \\ \alpha \geq -3 \end{cases} \rightarrow -3 \leq \alpha \leq 1$$

To maximize the function, *alpha* is chosen to be -3. Therefore,

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -3 \\ 4 \\ 0 \end{bmatrix}.$$

2. (20 pts) Formulate the first-order necessary conditions for the quadratic program,

$$\begin{aligned} & \text{minimize} && \frac{1}{2} \mathbf{x}^T \mathbf{Q} \mathbf{x} - \mathbf{b}^T \mathbf{x} \\ & \text{subject to} && \mathbf{A} \mathbf{x} = \mathbf{c}, \end{aligned}$$

where $\mathbf{x} \in \mathbb{R}^n$, $\mathbf{c} \in \mathbb{R}^m$, $m \leq n$, and $\mathbf{Q} = \mathbf{Q}^T > 0$.

- (a) (15 pts) Represent the obtained conditions as a system of linear equations and **write down the solution to the problem**.
- (b) (5 pts) What is the condition, involving \mathbf{A} and \mathbf{Q} , that must be satisfied for the solution to be unique?

(a)

$$\begin{aligned} \mathcal{L}(x, \lambda) &= \frac{1}{2} \mathbf{x}^T \mathbf{Q} \mathbf{x} - \mathbf{b}^T \mathbf{x} + \lambda (\mathbf{A} \mathbf{x} - \mathbf{c}) \\ \nabla_{\mathbf{x}} \mathcal{L}(x, \lambda) &= \mathbf{x}^{*T} \mathbf{Q} - \mathbf{b}^T + \lambda \mathbf{A} = 0 \\ \mathbf{x}^{*T} &= (\mathbf{b}^T - \lambda \mathbf{A}) \mathbf{Q}^{-1} \\ \mathbf{x}^* &= \mathbf{Q}^{-1} (\mathbf{b} - \lambda \mathbf{A}^T) \\ \mathbf{A} \mathbf{x}^* &= \mathbf{c} \\ \mathbf{A} \mathbf{Q}^{-1} (\mathbf{b} - \lambda^* \mathbf{A}^T) &= \mathbf{c} \\ \mathbf{A} \mathbf{Q}^{-1} \mathbf{b} - \lambda^* \mathbf{A} \mathbf{Q}^{-1} \mathbf{A}^T &= \mathbf{c} \\ \lambda^* &= (\mathbf{A} \mathbf{Q}^{-1} \mathbf{A}^T)^{-1} (\mathbf{A} \mathbf{Q}^{-1} \mathbf{b} - \mathbf{c}) \\ \mathbf{x}^* &= \mathbf{Q}^{-1} \left(\mathbf{b} - (\mathbf{A} \mathbf{Q}^{-1} \mathbf{A}^T)^{-1} (\mathbf{A} \mathbf{Q}^{-1} \mathbf{b} - \mathbf{c}) \mathbf{A}^T \right) \end{aligned}$$

- (b) \mathbf{A} and \mathbf{Q} need to be full rank for the solution to be unique.

3. (20 pts) Consider the optimization problem,

$$\begin{aligned} & \text{maximize} && -x_1^2 + x_1 - x_2 - x_1x_2 \\ & \text{subject to} && x_1 \geq 0, x_2 \geq 0 \end{aligned}$$

(a) (10 pts) Characterize feasible directions at the point

$$\mathbf{x}^* = \begin{bmatrix} \frac{1}{2} \\ 0 \end{bmatrix}.$$

(b) (10 pts) Write down the second-order necessary condition for \mathbf{x}^* . Does the point \mathbf{x}^* satisfy this condition?

Rewrite the problem,

$$\begin{aligned} & \text{minimize} && x_1^2 - x_1 + x_2 + x_1x_2 \\ & \text{subject to} && x_1 \geq 0, x_2 \geq 0 \end{aligned}$$

(a) For \mathbf{d} to be feasible, we need $d_2 \geq 0$ and d_1 can take an arbitrary value in \mathbb{R} .

(b)

$$\begin{aligned} \nabla f(\mathbf{x}) &= \begin{bmatrix} 2x_1 - 1 + x_2 \\ 1 + x_1 \end{bmatrix} \\ \mathbf{F}(\mathbf{x}) &= \begin{bmatrix} 2 & 1 \\ 1 & 0 \end{bmatrix} \\ \mathbf{d}^T \nabla f(\mathbf{x}^*) &= [d_1 \quad d_2] \begin{bmatrix} 0 \\ \frac{3}{2} \end{bmatrix} = \frac{3}{2}d_2 = 0 \\ d_2 &= 0 \\ \mathbf{d}^T \mathbf{F}(\mathbf{x}) \mathbf{d} &= [d_1 \quad d_2] \begin{bmatrix} 2 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} d_1 \\ d_2 \end{bmatrix} \\ &= 2d_1^2 + 2d_1d_2 \\ &= 2d_1^2 \\ &\geq 0 \end{aligned}$$

Therefore, the point \mathbf{x}^* satisfy the SONC.

4. (20 pts) Consider the following primal problem:

$$\begin{aligned} & \text{maximize} && x_1 + 2x_2 \\ & \text{subject to} && -2x_1 + x_2 + x_3 = 2 \\ & && -x_1 + 2x_2 + x_4 = 7 \\ & && x_1 + x_5 = 3 \\ & && x_i \geq 0, \quad i = 1, 2, 3, 4, 5 \end{aligned}$$

- (a) (5 pts) Construct the dual problem corresponding to the above primal function.
 (b) (15 pts) It is known that the solution to the above primal is $\mathbf{x}^* = [3 \ 5 \ 3 \ 0 \ 0]^T$. Find the solution to the dual.
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(a) Rewrite the problem,

$$\begin{aligned} & \text{minimize} && -x_1 - 2x_2 \\ & \text{subject to} && -2x_1 + x_2 + x_3 = 2 \\ & && -x_1 + 2x_2 + x_4 = 7 \\ & && x_1 + x_5 = 3 \\ & && x_i \geq 0, \quad i = 1, 2, 3, 4, 5 \end{aligned}$$

so the dual problem is

$$\begin{aligned} & \text{maximize} && [\lambda_1 \ \lambda_2 \ \lambda_3] \begin{bmatrix} 2 \\ 7 \\ 3 \end{bmatrix} \\ & \text{subject to} && [\lambda_1 \ \lambda_2 \ \lambda_3] \begin{bmatrix} -2 & 1 & 1 & 0 & 0 \\ -1 & 2 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 1 \end{bmatrix} \leq [-1 \ -2 \ 0 \ 0 \ 0] \end{aligned}$$

(b)

$$\begin{aligned} \lambda^T \mathbf{a}_i &= \mathbf{c}_i \\ &\rightarrow \begin{cases} -2\lambda_1 - \lambda_2 + \lambda_3 = -1 \\ \lambda_1 + 2\lambda_2 = -2 \\ \lambda_1 = 0 \end{cases} \\ &\rightarrow \begin{cases} \lambda_1 = 0 \\ \lambda_2 = -1 \\ \lambda_3 = -2 \end{cases} \end{aligned}$$

Therefore, the solution to the dual is

$$\lambda = \begin{bmatrix} 0 \\ -1 \\ -2 \end{bmatrix}$$

5. (20 pts) Find the minimizer of

$$f(x_1, x_2) = \frac{1}{2}x_1^2 + x_2^2 + x_1 + \frac{1}{2}x_2 + 3$$

using the conjugate gradient algorithm. The starting point is $\mathbf{x}^{(0)} = [0 \ 0]^T$.

$$\nabla f(\mathbf{x}) = \begin{bmatrix} x_1 + 1 \\ 2x_2 + \frac{1}{2} \end{bmatrix}$$

$$\mathbf{Q} = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}$$

$$\mathbf{g}^{(0)} = \nabla f(\mathbf{x}^{(0)})$$

$$= \begin{bmatrix} 1 \\ \frac{1}{2} \end{bmatrix}$$

$$\mathbf{d}^{(0)} = -\mathbf{g}^{(0)}$$

$$= \begin{bmatrix} -1 \\ -\frac{1}{2} \end{bmatrix}$$

$$\alpha_0 = - \frac{\begin{bmatrix} 1 & \frac{1}{2} \end{bmatrix} \begin{bmatrix} -1 \\ -\frac{1}{2} \end{bmatrix}}{\begin{bmatrix} -1 & -\frac{1}{2} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} -1 \\ -\frac{1}{2} \end{bmatrix}}$$

$$= \frac{5}{6}$$

$$\mathbf{x}^{(1)} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} + \frac{5}{6} \begin{bmatrix} -1 \\ -\frac{1}{2} \end{bmatrix}$$

$$= \begin{bmatrix} -\frac{5}{6} \\ -\frac{5}{12} \end{bmatrix}$$

$$\mathbf{g}^{(1)} = \nabla f(\mathbf{x}^{(1)})$$

$$= \begin{bmatrix} \frac{1}{6} \\ -\frac{1}{3} \end{bmatrix}$$

$$\beta_0 = \frac{\begin{bmatrix} \frac{1}{6} & -\frac{1}{3} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} -1 \\ -\frac{1}{2} \end{bmatrix}}{\begin{bmatrix} -1 & -\frac{1}{2} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} -1 \\ -\frac{1}{2} \end{bmatrix}}$$

$$= \frac{1}{9}$$

$$\mathbf{d}^{(1)} = -\mathbf{g}^{(1)}$$

$$= \begin{bmatrix} -\frac{1}{6} \\ \frac{1}{3} \end{bmatrix} + \frac{1}{9} \begin{bmatrix} -1 \\ -\frac{1}{2} \end{bmatrix}$$

$$\begin{aligned}
&= \begin{bmatrix} -\frac{5}{18} \\ \frac{5}{18} \end{bmatrix} \\
\alpha_1 &= -\frac{\begin{bmatrix} \frac{1}{6} & -\frac{1}{3} \end{bmatrix} \begin{bmatrix} -\frac{5}{18} \\ \frac{5}{18} \end{bmatrix}}{\begin{bmatrix} -\frac{5}{18} & \frac{5}{18} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} -\frac{5}{18} \\ \frac{5}{18} \end{bmatrix}} \\
&= \frac{3}{5} \\
\mathbf{x}^{(2)} &= \begin{bmatrix} -\frac{5}{6} \\ -\frac{5}{12} \end{bmatrix} + \frac{3}{5} \begin{bmatrix} -\frac{5}{18} \\ \frac{5}{18} \end{bmatrix} \\
&= \begin{bmatrix} -1 \\ -\frac{1}{4} \end{bmatrix} \\
\mathbf{g}^{(2)} &= \nabla f(\mathbf{x}^{(2)}) \\
&= \begin{bmatrix} 0 \\ 0 \end{bmatrix}
\end{aligned}$$

Therefore, $\mathbf{x}^{(2)} = \begin{bmatrix} -1 \\ -\frac{1}{4} \end{bmatrix}$ is the minimizer.