

**Indian Statistical Institute, Bangalore Centre**  
**B.Math.(Hons.)II Year - 2016-17, First Semester**  
**Optimization**

Back Paper Exam  
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30 December 2016, 10 am -1 pm  
Max.Marks: 100

**NOTE:**

- (i) Answer *any* 7 in Questions 1 – 9 and answer Question 10. WRITE NEATLY.  
(ii)  $M_n(\mathbb{R})$  denotes the set of real  $n \times n$  matrices.

1. Let  $\mathbf{e}_j$ ,  $1 \leq j \leq n$ , denote the standard unit vectors in  $\mathbb{R}^n$ . Put  $\mathbf{E}_{ij} = \mathbf{e}_i \mathbf{e}_j^t$  for  $1 \leq i, j \leq n$ . Define

$$\mathbf{A} = \sum_{i=1}^{n-1} \alpha_i \mathbf{E}_{i,i+1},$$

where  $\alpha_i$  for  $i = 1, \dots, n-1$  are real numbers. Show that  $\mathbf{I} - \mathbf{A}$  is invertible and obtain an expression for  $(\mathbf{I} - \mathbf{A})^{-1}$  in terms of powers of  $\mathbf{A}$ . (12)

2. (a) Find all the  $2 \times 2$  projection matrices. (4)

(b) Suppose  $\mathbf{A}$  is a real  $m \times n$  matrix of rank  $n$  and define  $\mathbf{P} = \mathbf{A}(\mathbf{A}^t \mathbf{A})^{-1} \mathbf{A}^t$ .

i. Verify that the definition of  $\mathbf{P}$  makes sense and show that  $\text{im } \mathbf{P} = \text{im } \mathbf{A}$  and  $\ker \mathbf{P} = \ker \mathbf{A}^t$ . (1+4)

ii. Show that  $\mathbf{P}$  is an orthogonal projection. (3)

3. (a) Let  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  be a set of orthonormal (column) vectors in  $\mathbb{R}^n$  and  $\mathbf{x} \in \mathbb{R}^n$ . Show that

$$\|\mathbf{x} - \sum_{i=1}^k \lambda_i \mathbf{u}_i\| \geq \|\mathbf{x} - \sum_{i=1}^k (\mathbf{x}, \mathbf{u}_i) \mathbf{u}_i\|,$$

for all real numbers  $\lambda_1, \dots, \lambda_k$ , with equality holding if and only if  $\lambda_i = (\mathbf{x}, \mathbf{u}_i)$  for  $i = 1, \dots, k$ . (6)

(b) Let  $\mathbf{A}$  be an  $m \times n$  real matrix and  $\mathbf{b}$  is a given column vectors in  $\mathbb{R}^m$ . If  $\mathbf{P}$  is the orthogonal projection onto  $\text{im } \mathbf{A}$ , show that

$$\|\mathbf{A}\mathbf{x} - \mathbf{b}\| \geq \|\mathbf{b} - \mathbf{P}\mathbf{b}\|,$$

for all  $\mathbf{x} \in \mathbb{R}^n$ , with equality holding if and only if  $\mathbf{A}\mathbf{x} = \mathbf{P}\mathbf{b}$  for some  $\mathbf{x} \in \mathbb{R}^n$ . (6)

4. (a) Let  $\mathbf{A} = ((a_{ij})) \in M_n(\mathbb{R})$  be a positive matrix, that is  $a_{ij} > 0$  for all  $i, j$ . Suppose there is a  $\lambda > 0$  and a vector  $\mathbf{x} \geq \mathbf{0}$ ,  $\mathbf{x} \neq \mathbf{0}$  such that  $\mathbf{A}\mathbf{x} > \lambda\mathbf{x}$ . Show that there is a  $\delta > 0$  such that  $\mathbf{A}\mathbf{x} \geq (\lambda + \delta)\mathbf{x}$ . (4)

- (b) Suppose  $\mathbf{A} \in M_n(\mathbb{R})$ . Show that  $\mathbf{A}$  is a positive matrix if and only if  $\mathbf{Ax} > \mathbf{0}$  for all  $\mathbf{x} \geq \mathbf{0}$ ,  $\mathbf{x} \neq \mathbf{0}$ . (4)
- (c) Suppose  $\mathbf{A} = ((a_{ij})) \in M_n(\mathbb{R})$  is a non-negative matrix, that is  $a_{ij} \geq 0$  for all  $i, j$  and  $\mathbf{x} \geq \mathbf{0}$ ,  $\mathbf{x} \neq \mathbf{0}$ . Define

$$r_{\mathbf{x}} = \min_{1 \leq i \leq n} \left\{ \frac{(\mathbf{Ax})_i}{(\mathbf{x})_i} : (\mathbf{x})_i > 0 \right\},$$

$$R_{\mathbf{x}} = \max \{ \lambda \geq 0 : \mathbf{Ax} \geq \lambda \mathbf{x} \}.$$

Show that  $r_{\mathbf{x}} = R_{\mathbf{x}}$ . (4)

5. (a) Show that 1 is the dominant eigenvalue of the doubly stochastic matrix  $\mathbf{A} = \begin{pmatrix} 1/2 & 0 & 1/3 & 1/6 \\ 1/6 & 1/2 & 1/3 & 0 \\ 0 & 1/6 & 1/2 & 1/3 \\ 1/3 & 0 & 1/6 & 1/2 \end{pmatrix}$  and hence find the limit  $\lim_{k \rightarrow \infty} \mathbf{A}^k$ . (6)

- (b) Suppose  $\mathbf{A} \in M_n(\mathbb{R})$  is a non-negative, irreducible matrix such that its spectral radius equals 1. Prove the following:

i. There are positive vectors  $\mathbf{u}, \mathbf{v} \in \mathbb{R}^n$  such that  $\mathbf{Au} = \mathbf{u}$  and  $\mathbf{A}^t \mathbf{v} = \mathbf{v}$ , with  $\mathbf{v}^t \mathbf{u} = 1$ . (3)

ii. Define  $\mathbf{B} = \mathbf{A} - \mathbf{u}\mathbf{v}^t$  and show that each non-zero eigenvalue of  $\mathbf{B}$  is also an eigenvalue of  $\mathbf{A}$ , but 1 is *not* an eigenvalue of  $\mathbf{B}$ . (3)

6. (a) Solve the following using simplex method:

$$\begin{aligned} & \text{maximize } 3x_1 + x_2 + 3x_3 \\ & \text{subject to } 2x_1 + x_2 + x_3 \leq 2 \\ & \quad \quad \quad x_1 + 2x_2 + 3x_3 \leq 5 \\ & \quad \quad \quad 2x_1 + 2x_2 + x_3 \leq 6 \\ & \quad \quad \quad x_i \geq 0, i = 1, 2, 3. \end{aligned}$$

(6)

- (b) Let  $\mathbf{A}$  be an  $m \times n$  real matrix and  $\mathbf{b}, \mathbf{c}$  are given column vectors in  $\mathbb{R}^m$  and  $\mathbb{R}^n$  respectively. Consider the following linear programming:

$$\text{minimize } \mathbf{c}^t \mathbf{x}, \text{ subject to } \mathbf{Ax} = \mathbf{b}, \mathbf{x} \geq \mathbf{a},$$

where  $\mathbf{a} \geq \mathbf{0}$  is a given vector in  $\mathbb{R}^n$ . Find the dual problem. (6)

7. (a) Suppose  $C$  is a convex set in  $\mathbb{R}^n$  and  $k \geq 2$ . Let  $\mathbf{x}_1, \dots, \mathbf{x}_k$  are in  $C$  and  $t_1, \dots, t_k$  are non-negative real numbers such that  $t_1 + \dots + t_k = 1$ . Show that  $t_1 \mathbf{x}_1 + \dots + t_k \mathbf{x}_k$  is also in  $C$ . (3)

- (b) State the Farkas-Minkowski lemma and prove it using the duality theorem of the linear programming. (6)

(c) Using the theorem of the alternative, show that the following system

$$\begin{pmatrix} 1 & 3 & -5 \\ 1 & -4 & -7 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 2 \\ 3 \end{pmatrix}$$

does not have a non-negative solution. (3)

8. Consider the following LPP:

$$\begin{aligned} & \text{minimize} && -x_1 - 4x_2 - 3x_3 \\ & \text{subject to} && 2x_1 + 2x_2 + x_3 \leq 4 \\ & && x_1 + 2x_2 + 2x_3 \leq 6 \\ & && x_i \geq 0, i = 1, 2, 3. \end{aligned}$$

It is given that an optimal solution of the above is  $(0, 1, 2)$ .

(a) Write down the corresponding dual problem. (4)

(b) Solve the dual problem by the simplex method. (6)

(c) Verify the duality theorem for these duality relations. (2)

9. Let  $\mathbf{A}$  be a real  $m \times n$  matrix. Show that the system  $\mathbf{Ax} > \mathbf{0}$  has a solution if and only if the following statement holds true:

$$\boldsymbol{\lambda} \in \mathbb{R}^m, \boldsymbol{\lambda} \geq \mathbf{0} \text{ and } \boldsymbol{\lambda}^t \mathbf{A} = \mathbf{0} \Rightarrow \boldsymbol{\lambda} = \mathbf{0}.$$

(12)

10. (a) Let  $\mathbf{A}$  be an  $m \times n$  real matrix and  $\mathbf{c}$  is a given column vector  $\mathbb{R}^n$ . Assume that the following statement holds true:

$$\mathbf{x} \in \mathbb{R}^n, \mathbf{Ax} \leq \mathbf{0} \Rightarrow \mathbf{c}^t \mathbf{x} \leq 0.$$

Show that there is a  $\boldsymbol{\lambda} \in \mathbb{R}^m, \boldsymbol{\lambda} \geq \mathbf{0}$  such that  $\mathbf{c}^t = \boldsymbol{\lambda}^t \mathbf{A}$ . (8)

(b) Let  $\mathbf{A}$  be an  $m \times n$  real matrix and  $\mathbf{b}, \mathbf{c}$  are given column vectors in  $\mathbb{R}^m$  and  $\mathbb{R}^n$  respectively. Put  $k = m + n$  and define the  $k \times k$  matrix  $\mathbf{M}$  by  $\mathbf{M} = \begin{pmatrix} \mathbf{O}_{n \times n} & -\mathbf{A}^t \\ \mathbf{A} & \mathbf{O}_{m \times m} \end{pmatrix}$  and  $\mathbf{q} = \begin{pmatrix} \mathbf{c} \\ -\mathbf{b} \end{pmatrix}$ . Consider the linear programme  $(P)$  given by

$$\begin{aligned} & \text{minimize} && \mathbf{q}^t \mathbf{z} \\ & \text{subject to} && \mathbf{Mz} \geq -\mathbf{q} \\ & && \mathbf{z} \geq \mathbf{0}. \end{aligned}$$

Show that the problem  $(P)$  and its dual are the same. Further, show that any feasible solution of  $(P)$  is also optimal. (8)