

Introduction to Optimization Theory

Homework Assignment 1

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Ex. 2.2

Proof:

For any $\mathbf{x}, \mathbf{y} \in S$, we have $f(\mathbf{x}), f(\mathbf{y}) \leq c$. f is convex so that for any $\lambda \in [0, 1]$, we have

$$f(\lambda \cdot \mathbf{x} + (1 - \lambda) \cdot \mathbf{y}) \leq \lambda f(\mathbf{x}) + (1 - \lambda)f(\mathbf{y}) \leq \lambda \cdot c + (1 - \lambda) \cdot c = c,$$

which means $\lambda \cdot \mathbf{x} + (1 - \lambda) \cdot \mathbf{y} \in S$. Therefore, S is convex.

Ex. 2.4

The argument is wrong in that extreme points are representation-dependent. When we convert an arbitrary LP problem to an equivalence in standard form, we might add new variables and new constraints, which may create extreme points.

For example, polyhedron $P = \{(x, y) \mid 0 \leq x \leq 1\}$ does not have an extreme point, while its standard form $P' = \{(x_0, x_1, y_0, y_1) \mid x_0 - x_1 = 1, (x_0, x_1, y_0, y_1) \geq \mathbf{0}\}$ ($x_0 = x, x_1 = x - 1, y = y_0 - y_1$) has extreme points.

Ex. 2.6

(a)

Proof:

For any $\mathbf{y} \in C$, i.e. $\mathbf{y} = \sum_{i=1}^n \lambda_i \mathbf{A}_i$, $(\lambda_1, \lambda_2, \dots, \lambda_n) \geq \mathbf{0}$, we could define polyhedron

$$P = \left\{ (\lambda_1, \lambda_2, \dots, \lambda_n) \in \mathbb{R}^n \mid \sum_{i=1}^n \lambda_i \mathbf{A}_i = \mathbf{y}, (\lambda_1, \lambda_2, \dots, \lambda_n) \geq \mathbf{0} \right\}.$$

It's obvious that P is non-empty. Notice that P is a polyhedron in standard form, so P does not contain a line, which means P has at least one extreme point (basic feasible solution) $\boldsymbol{\lambda}^* = (\lambda_1^*, \lambda_2^*, \dots, \lambda_n^*)$. For a basic feasible solution, there are at most m non-zero components, which means

$$\mathbf{y} = \sum_{i=1}^n \lambda_i^* \mathbf{A}_i$$

is a representation with at most m non-zero coefficients.

(b)

Proof:

Similar to **(a)**, for any $\mathbf{y} = \sum_{i=1}^n \lambda_i \mathbf{A}_i$, $\sum_{i=1}^n \lambda_i = 1$, $(\lambda_1, \lambda_2, \dots, \lambda_n) \geq \mathbf{0}$, we could define polyhedron

$$\Lambda = \left\{ (\lambda_1, \lambda_2, \dots, \lambda_n) \in \mathbb{R}^n \mid \sum_{i=1}^n \lambda_i \mathbf{A}_i = \mathbf{y}, \sum_{i=1}^n \lambda_i = 1, (\lambda_1, \lambda_2, \dots, \lambda_n) \geq \mathbf{0} \right\}$$

with $m + 1$ equation constraints. Again, we know that Λ has at least one basic feasible solution, with at most $m + 1$ non-zero components, which is a representation of \mathbf{y} with at most $m + 1$ non-zero coefficients.

Ex. 2.7

Proof:

That $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_m$ span \mathbb{R}^n means there exist n linearly independent vectors in $\{\mathbf{a}_i \mid i = 1, 2, \dots, m\}$ (i.e. $\text{rank } \mathbf{A} = \text{rank}[\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_m] = n$), which means the polyhedron does not contain a line. Therefore, there exist n linearly independent vectors that span \mathbb{R}^n in $\{\mathbf{g}_i \mid i = 1, 2, \dots, k\}$, too.

Ex. 2.8

Proof:

\Rightarrow :

If a basis is associated with the basic solution \mathbf{x} , for any i s.t. the i -th column is not in the basis, we set $x_i = 0$. So if $x_i \geq 0$, the i -th column has to be in the basis.

\Leftarrow :

If every column $\mathbf{A}_i, i \in J$ is in the basis, all columns $\{x_i\}$ that are not in the basis have $x_i = 0$. Therefore, the basis is associated with the solution.

Ex. 2.9

(a)

Proof:

If two different bases lead to the same basic solution, more than $n - m$ variables are zero, which means more than n constraints are active at here.

(b)

False.

Degenerate basic solution may have only one basis, e.g. consider polyhedron in standard form $P = \{(x, y) | x = 0, y = 0, (x, y) \geq 0\}$. The polyhedron contains only one point, and obviously there's only one choice of basis.

(c)

False.

The counterexample is the same as (b). There's no adjacent basic solution.