

Math 430 Assignment 3

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Exercise 2.1

Solution:

(a) It is not a polyhedron. Let us call this set S , then

$$S = \{\mathbf{x} \in \mathbb{R}^2 | \mathbf{Ax} \leq \mathbf{b} \text{ for } \forall \theta \in [0, \frac{\pi}{2}] \wedge \mathbf{x} \geq 0\}$$

where $\mathbf{A} = [\cos \theta \sin \theta]$ and $\mathbf{b} = 1$. Since $\theta \in [0, \frac{\pi}{2}]$, then $\mathbf{A} \geq 0$. We can write

$$[\cos \theta \sin \theta] \begin{bmatrix} x \\ y \end{bmatrix} \leq 1$$

It could be intersection of infinite many halfspaces. By definition of polyhedron, it is intersection of finite many halfspaces. Hence this is not a polyhedron.

(b) It is a polyhedron. Let us call this set S , then

$$S = \{x \in \mathbb{R} | x^2 - 8x + 15 \leq 0\}$$

$$\Rightarrow S = \{x \in \mathbb{R} | (x - 3)(x - 5) \leq 0\}$$

$$\Rightarrow S = \{x \in [3, 5]\} \Rightarrow S = [3, 5]$$

In \mathbb{R} , S is an intersection of two halfspaces. S is a polyhedron.

(c) It is a trivial polyhedron. Since it is empty set, we can choose any \mathbf{A} , \mathbf{b} to satisfy the polyhedron definition. For example,

$$P = \{\mathbf{x} \in \mathbb{R} | \mathbf{Ax} \geq \mathbf{b}\}$$

Let $\mathbf{A} = (-2, 2)^T \in \mathbb{R}^2$ and $\mathbf{b} = (3, -3)^T \in \mathbb{R}^2$, then $P = \emptyset$. P follows the definition of polyhedron, therefore P is polyhedron.

Exercise 2.2

Solution: Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be a convex function, then for every $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ and every $\lambda \in [0, 1]$,

$$f(\lambda \mathbf{x} + (1 - \lambda) \mathbf{y}) \leq \lambda f(\mathbf{x}) + (1 - \lambda) f(\mathbf{y})$$

Let $\mathbf{z}_1, \mathbf{z}_2$ be any two points in S , want to show $\lambda \mathbf{z}_1 + (1 - \lambda) \mathbf{z}_2 \in S$.

$$\begin{aligned} f(\lambda \mathbf{z}_1 + (1 - \lambda) \mathbf{z}_2) &\leq \lambda f(\mathbf{z}_1) + (1 - \lambda) f(\mathbf{z}_2) \\ &\leq \lambda c + (1 - \lambda) c = c \end{aligned}$$

That is

$$f(\lambda \mathbf{z}_1 + (1 - \lambda) \mathbf{z}_2) \leq c \Rightarrow \lambda \mathbf{z}_1 + (1 - \lambda) \mathbf{z}_2 \in S$$

Therefore, S is convex by definition.

Exercise 2.5

Solution:

(a) Proof by contrapositive. Let \mathbf{x} be a non-extreme point of P , then we can find two vector $\mathbf{s}, \mathbf{t} \in P$ and a scalar $\lambda \in [0, 1]$, s.t.

$$\mathbf{x} = \lambda \mathbf{s} + (1 - \lambda) \mathbf{t}$$

Want to show $f(\mathbf{x})$ is not an extreme point of Q . Since both function f, g is affine, then $f(\mathbf{x}) = \mathbf{A}\mathbf{x} + \mathbf{b}$, $g(\mathbf{y}) = \mathbf{B}\mathbf{y} + \mathbf{c}$ where $\mathbf{x} \in P, \mathbf{y} \in Q$, \mathbf{A}, \mathbf{B} are matrix, \mathbf{b}, \mathbf{c} are vector. Therefore,

$$\begin{aligned} f(\mathbf{x}) &= f(\lambda \mathbf{s} + (1 - \lambda) \mathbf{t}) = \mathbf{A}(\lambda \mathbf{s} + (1 - \lambda) \mathbf{t}) + \mathbf{b} \\ &= \lambda \mathbf{A}\mathbf{s} + (1 - \lambda) \mathbf{A}\mathbf{t} + \mathbf{b} = \lambda \mathbf{A}\mathbf{s} + (1 - \lambda) \mathbf{A}\mathbf{t} + \lambda \mathbf{b} + (1 - \lambda) \mathbf{b} \\ &= \lambda (\mathbf{A}\mathbf{s} + \mathbf{b}) + (1 - \lambda) (\mathbf{A}\mathbf{t} + \mathbf{b}) = \lambda f(\mathbf{s}) + (1 - \lambda) f(\mathbf{t}) \end{aligned}$$

That is $f(\mathbf{x})$ is not an extreme point of Q . Therefore, if \mathbf{x} is not an extreme point of P , then $f(\mathbf{x})$ is not an extreme point of Q . That is if $f(\mathbf{x})$ is an extreme point of Q , then \mathbf{x} is an extreme point of P .

The converse is trivial. That is \mathbf{x} is an extreme point of P if and only if $f(\mathbf{x})$

is an extreme point of Q .

(b) Let f be a map, and g be a map such that

$$f : \mathbf{x} \mapsto (\mathbf{x}, \mathbf{z}) \text{ for } \forall \mathbf{x} \in P$$

Let $f(\mathbf{x}) = \mathbf{A}\mathbf{x} + \mathbf{c}$ where \mathbf{A} is $(n+k) \times n$ matrix and \mathbf{c} is a vector in \mathbb{R}^{n+k}

$$f(\mathbf{x}) = \begin{bmatrix} \mathbf{I}_{n \times n} \\ - - - \\ \mathbf{B}_{k \times n} \end{bmatrix} \mathbf{x}_{n \times 1} + \begin{bmatrix} \mathbf{0}_{n \times 1} \\ - - - \\ -\mathbf{b}_{k \times 1} \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{n \times 1} \\ - - - \\ \mathbf{z}_{k \times 1} \end{bmatrix}$$

where $\mathbf{I}_{n \times n}$ is identity matrix $n \times n$, $\mathbf{0}_{n \times 1}$ is zero vector in \mathbb{R}^n ,

$\mathbf{B}_{k \times n}$ is a matrix and \mathbf{b} is a vector in \mathbb{R}^k

then $\mathbf{B}_{k \times n}\mathbf{x} - \mathbf{b} = \mathbf{z} \Rightarrow f(\mathbf{x}) \in Q$ for $\forall \mathbf{x} \in P$, f is both affine and injective map.

$$g : (\mathbf{x}, \mathbf{z}) \mapsto \mathbf{x} \text{ for } \forall (\mathbf{x}, \mathbf{z}) \in Q$$

By assumption there are some matrix \mathbf{C} and vector \mathbf{d} , s.t. $\mathbf{C}\mathbf{x} - \mathbf{z} = \mathbf{d}$

Let $g((\mathbf{x}, \mathbf{z})) = \mathbf{D}(\mathbf{x}, \mathbf{z}) + \mathbf{e}$, where \mathbf{D} is $n \times (n+k)$ matrix and \mathbf{e} is a vector in \mathbb{R}^n

$$g((\mathbf{x}, \mathbf{z})) = \begin{bmatrix} \mathbf{I}_{n \times n} & | & \mathbf{0}_{n \times k} \end{bmatrix} \begin{bmatrix} \mathbf{x}_{n \times 1} \\ - - - \\ \mathbf{z}_{k \times 1} \end{bmatrix} + \mathbf{0}_{n \times 1} = \mathbf{x}_{n \times 1}$$

then $g((\mathbf{x}, \mathbf{z})) \in P$ for $\forall (\mathbf{x}, \mathbf{z}) \in Q$. i.e. g is affine and injective map. By construction, both f, g are affine. Moreover, $g(f(\mathbf{x})) = \mathbf{x}$ for $\forall \mathbf{x} \in P$, $f(g(\mathbf{y})) = \mathbf{y}$ for $\forall \mathbf{y} \in Q$. These imply f, g is inverse map and P and Q are isomorphic.

Exercise 2.6

Solution:

(a) Case 1: When $n \leq m$, it is trivial. It is at most m of the coefficients λ_i being nonzero.

Case 2: When $n > m$, assume there are k linear independent vectors in \mathbf{R}^m . WLOG, denoted them as $\mathbf{A}_1, \dots, \mathbf{A}_k$, where $k \leq m < n$. Let $\mathbf{A} \in C$, then

$$\mathbf{A} = \sum_{i=1}^n \lambda_i \mathbf{A}_i, \text{ where } \lambda_i \geq 0 \text{ for } 1 \leq i \leq n$$

Consider the polyhedron,

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

By definition of basic solution and basic feasible solution, a basic solution $x \in \Lambda$ must satisfy all equality constraints and active for at least n linear independent constraints of Λ , but $\sum_{i=1}^n \lambda_i \mathbf{A}_i = \mathbf{y}$ are only k linear independent constraints since \mathbf{A}_i is in \mathbb{R}^m . Therefore, we need $(n - k)$ more linear independent constraints to consider a basic solution is active. By linear algebra, in order to find a basic solution, we need to reduce the matrix to $k \times k$ square matrix. That is the other $(n - k)$ linear independent constraints must be zeros. Hence there are only k linear independent constraints are nonzero. Since $k \leq m$, then at most m of λ_i being nonzero.

(b) Consider a polyhedron,

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

same argument as part a, both $\sum_{i=1}^n \lambda_i \mathbf{A}_i = \mathbf{y}$, $\sum_{i=1}^n \lambda_i = 1$ will provide only $k + 1$ equality. In order to get a basic solution, there must be $n - k - 1$ linear independent constraints are zeros. Therefore, matrix can be reduced to $(k + 1) \times (k + 1)$ square matrix. Since $k \leq m$, when $k = m$ there are $m + 1$ linear independent constraints are nonzero. That is at most $m + 1$ of the coefficients λ_i being nonzero.