

Problem Sheet 3

Problems in Part A will be discussed in class. Problems in Part B come with solutions and should be tried at home.

Part A

(3.1) Recall that a norm was a function $\|\cdot\|$ on \mathbb{R}^n such that

- $\|\mathbf{x}\| \geq 0$ and $\|\mathbf{x}\| = 0 \Leftrightarrow \mathbf{x} = \mathbf{0}$;
- $\|\lambda\mathbf{x}\| = |\lambda|\|\mathbf{x}\|$ for $\lambda \in \mathbb{R}$;
- $\|\mathbf{x} + \mathbf{y}\| \leq \|\mathbf{x}\| + \|\mathbf{y}\|$.

The most prominent examples are

$$\|\mathbf{x}\|_2 = \sqrt{\sum_{i=1}^n x_i^2}, \quad \|\mathbf{x}\|_1 = \sum_{i=1}^n |x_i|, \quad \|\mathbf{x}\|_\infty = \max_{1 \leq i \leq n} |x_i|.$$

(a) Show that for any norm, the *norm balls*

$$B(\mathbf{p}, r) := \{\mathbf{x} : \|\mathbf{x} - \mathbf{p}\| \leq r\}$$

are convex sets.

(b) To a convex set C with $\mathbf{0} \in \text{int}C$ we can associate the *polar set*

$$C^* := \{\mathbf{y} \in \mathbb{R}^n : \forall \mathbf{x} \in C, \langle \mathbf{x}, \mathbf{y} \rangle \leq 1\}.$$

Show that the polar of a convex set is again convex.

(c) Describe the polar sets of the norm balls $B(\mathbf{0}, 1)$ for the 1, 2, and ∞ norms.

(3.2) Let $C, D \subseteq \mathbb{R}^n$ be disjoint, non-empty, closed, bounded convex sets. Show that there exists an affine hyperplane H strictly separating C and D (i.e., $C \subset \text{int}H_-$ and $D \subset \text{int}H_+$). Hint: consider the set $C - D = \{\mathbf{x} - \mathbf{y} : \mathbf{x} \in C, \mathbf{y} \in D\}$ and use the separation theorem for a convex set and a point from the lecture. Give an example of closed convex sets C and D that cannot be strictly separated by a hyperplane.

(3.3) (Facility location.) Facility location problems concern the location of points (facilities or devices on a circuit) in a region, some of which are required to be connected by links (streets or wires). The objective is to locate some of the points that are not fixed in a way that minimizes the transport cost along the links. The distance can be measured with respect to the usual Euclidean norm or the 1-norm.

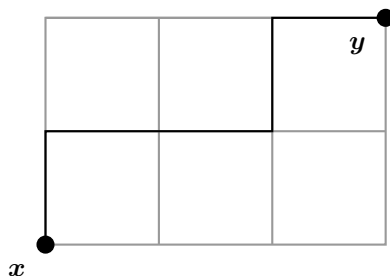


Figure 1: Manhattan distance

- (a) Given a rectangular grid, show that the length of a path from coordinate x to y is given by

$$d(x, y) := \|x - y\|_1.$$

This is an example of the so-called *Manhattan distance*.

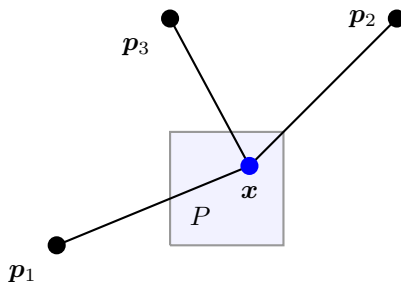
- (b) Suppose now that we are given N fixed points p_1, \dots, p_N in on a fine grid in \mathbb{R}^2 and would like to find a point x that minimizes the total 1-norm distance

$$\text{minimize} \quad \sum_{i=1}^N \|p_i - x\|_1 = \sum_{i=1}^N |p_{i,1} - x_1| + |p_{i,2} - x_2|.$$

Assume in addition that the point x is constrained to be in a polyhedral region

$$P = \{x : Ax \leq b\}.$$

Formulate this optimal placement problem as a linear programming problem.

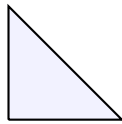


Part B

- (3.4) A *polyhedron* P is a convex set defined by a system of linear inequalities

$$P = \{x \in \mathbb{R}^n : Ax \leq b\},$$

where $\mathbf{A} \in \mathbb{R}^{m \times n}$, $\mathbf{x} \in \mathbb{R}^d$ and $\mathbf{b} \in \mathbb{R}^m$. A vertex \mathbf{z} of P is a point of P that can not be expressed as a convex combination of distinct points in P . For $\mathbf{z} \in P$, let $\mathbf{A}_{\mathbf{z}}$ denote the matrix consisting of those rows \mathbf{a}_i^\top of \mathbf{A} where $\mathbf{a}_i^\top \mathbf{z} = b_i$. For example, in Figure 2 the matrix corresponding to the vertex $\mathbf{z} = (1, 0)^\top$ is given by



$$\mathbf{A} = \begin{pmatrix} 1 & 1 \\ -1 & 0 \\ 0 & -1 \end{pmatrix}, \quad \mathbf{b} = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \quad \mathbf{A}_{\mathbf{z}} = \begin{pmatrix} 1 & 1 \\ 0 & -1 \end{pmatrix}.$$

Figure 2: A triangle

Show that the vertices of P are precisely the points \mathbf{z} for which $\mathbf{A}_{\mathbf{z}}$ has rank n . In particular, every polyhedron has only finitely many vertices.

(3.5) (Text classification) Consider the vocabulary consisting of the words