$\begin{array}{c} {\bf Measure \ Aspects \ of} \\ {\bf Cut \ Polyhedra:} \\ \ell_1 \text{-embeddability and \ Probability} \end{array}$

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Measure aspects of cut polyhedra: ℓ_1 -embeddability and probability

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1 Preliminaries

We recall in this Section all the definitions that we need for this Chapter and, in particular, the definitions about distance spaces, isometric embeddings, measure spaces, and our main host spaces, namely, the Banach ℓ_{p} - and L_{p} -spaces for $1 \leq p \leq \infty$.

1.0.1 Distance spaces and ℓ_p -spaces

Let X be a set. A function $d: X \times X \to \mathbb{R}_+$ is called a **distance** on X if d is **symmetric**, i.e., satisfies d(i, j) = d(j, i) for all $i, j \in X$, and if d(i, i) = 0 holds for all $i \in X$. Then, (X, d) is called a **distance space**. If d satisfies, in addition, the following inequalities

(1.1)
$$d(i,j) \le d(i,k) + d(j,k) \text{ for all } i,j,k \in X,$$

called **triangle inequalities**, then d is called a **semimetric** on X. Moreover, if d(i, j) = 0 holds only for i = j, then d is a **metric** on X.

Suppose d is a distance on the set $V_n = \{1, \ldots, n\}$. Set $E_n = \{ij : i, j \in V_n, i \neq j\}$, where the symbol ij denotes the unordered pair of the integers i, j, i.e., ij and ji are considered identical. Because of symmetry and since d(i, i) = 0 for $i \in V_n$, we can view the distance d as a vector $d = (d_{ij})_{1 \leq i < j \leq n} \in \mathbb{R}^{E_n}$ and, vice versa, each vector $d \in \mathbb{R}^{E_n}$ yields a symmetric function that is zero on the main diagonal. We will use both representations as a function on $V_n \times V_n$ or as a vector of \mathbb{R}^{E_n} for a distance on V_n .

Given a normed space $(E, \| . \|)$, a metric $d_{\|.\|}$ is defined on E, called **norm** or **Minkowski metric**, by setting

$$d_{\parallel,\parallel}(x,y) = \parallel x - y \parallel$$

for all $x, y \in E$.

We will consider, in particular, the norm metric, denoted by d_{ℓ_p} and called the ℓ_p -**metric**, of the Banach ℓ_p -space $(\mathbb{R}^m, \| \cdot \|_p)$ for $p \ge 1$. Recall that

$$\| x \|_{p} = \left(\sum_{1 \le k \le m} |x_{k}|^{p} \right)^{\frac{1}{p}}$$

for all $x \in \mathbb{R}^m$. The metric space $(\mathbb{R}^m, d_{\ell_p})$ is denoted by ℓ_p^m . Similarly, ℓ_{∞}^m denotes the metric space $(\mathbb{R}^m, d_{\ell_{\infty}})$, where $d_{\ell_{\infty}}$ denotes the norm metric associated with the norm $\|\cdot\|_{\infty}$ which is defined by

$$|| x ||_{\infty} = \max(|x_k| : 1 \le k \le m),$$

for all $x \in \mathbb{R}^m$.

For $1 \leq p < \infty$, the metric space ℓ_p^{∞} consists of the set of infinite sequences $x = (x_i)_{i\geq 0} \in \mathbb{R}^{\mathbb{N}}$ for which the sum $\sum_{i\geq 0} |x_i|^p$ is finite, endowed with the distance $d(x, y) = \left(\sum_{i\geq 0} |x_i - y_i|^p\right)^{\frac{1}{p}}$. In the same way ℓ_{∞}^{∞} is the set of bounded infinite sequences $x \in \mathbb{R}^{\mathbb{N}}$, endowed with the distance $d(x, y) = \max(|x_i - y_i| : i \geq 0)$.

If (X, d) and (X', d') are two distance spaces, (X, d) is said to be **isometrically embeddable** into (X', d') if there exists a map ϕ (the **embedding**) from X to X' such that $d(x, y) = d'(\phi(x), \phi(y))$ for all $x, y \in X$. One says also that (X, d) is an **isometric subspace** of (X', d'). All the embeddings considered here are isometric, so we sometimes omit the adjective "isometric".

A distance space (X, d) is said to be ℓ_p -embeddable if (X, d) is isometrically embeddable into the space ℓ_p^m for some integer $m \ge 1$. The smallest such integer m is called the ℓ_p -dimension of (X, d) and is denoted by $m_p(X, d)$. Then, we denote by

(1.2)
$$m_p(n) = \max(m_p(X, d) : |X| = n \text{ and } (X, d) \text{ is } \ell_p \text{-embeddable})$$

the minimum dimension m such that each ℓ_p -embeddable distance on n points can be embedded in ℓ_p^m . It is known that $m_p(n)$ is finite; in fact, $m_p(n) \leq \binom{n}{2}$ for all n and p (see Section 7.1).

(X, d) is said to be ℓ_p^{∞} -embeddable if it is an isometric subspace of ℓ_p^{∞} .

We are interested here in the study of the distances spaces which can be isometrically embedded in one of the following host spaces: ℓ_p^m , ℓ_p^∞ , or $L_p(\Omega, \mathcal{A}, \mu)$ (see the definition below) for $p \ge 1$ and we are mainly concerned with the cases p = 1, 2.

The case p = 1 is directly relevant to the central topic of this book. Indeed, the distances on n points that are ℓ_1 -embeddable are precisely the members of the cut cone CUT_n ; see Theorem 2.11.

The case p = 2 is also closely related to our topic; see Section 4.3.

On the other hand, it is well known that the distances on n points that are ℓ_{∞} embeddable are precisely the semimetrics on n points, i.e., the members of the semimetric
cone MET_n. To see it, note that if d is a semimetric on the set $V_n = \{1, \ldots, n\}$, then the
mapping $i \in V_n \mapsto (d(1, i), d(2, i), \ldots, d(n - 1, i)) \in \mathbb{R}^{n-1}$ is an isometric embedding of (V_n, d) into ℓ_{∞}^{n-1} . This shows that

$$(1.3) m_{\infty}(n) \le n-1.$$

We also consider isometric embeddings into the hypercube metric space $(\{0, 1\}^m, d_{\ell_1})$, which is a subspace of ℓ_1^m . Note that the hypercube metric space can also be defined

as the graphic metric space $(V(H_m), d_{H_m})$, where H_m denotes the 1-skeleton of the *m*dimensional hypercube and d_{H_m} is its path metric defined on the nodes of H_m . A distance space (X, d) is said to be **hypercube embeddable** if it can be isometrically embedded in some hypercube metric space. Hence, each hypercube embeddable distance space is ℓ_1 -embeddable and, in fact, if *d* is rational valued, then the space (X, d) is hypercube embeddable if and only if $(X, \lambda d)$ is ℓ_1 -embeddable for some scalar λ ; see Proposition 2.8.

1.0.2 Measure spaces and L_p-spaces

For defining the distance space $L_p(\Omega, \mathcal{A}, \mu)$, we need to recall some definitions on measure spaces. Let Ω be a set and let \mathcal{A} be a σ -algebra of subsets of Ω , i.e., \mathcal{A} satisfies the following properties

$$\begin{array}{l} \Omega \in \mathcal{A}, \\ \text{if } A \in \mathcal{A} \text{ then } \Omega \setminus A \in \mathcal{A}, \\ \text{if } A = \bigcup_{1 \leq k \leq \infty} \text{ with } A_k \in \mathcal{A} \text{ for all } k, \text{ then } A \in \mathcal{A}. \end{array}$$

A function $\mu : \mathcal{A} \longrightarrow \mathbb{R}_+$ is a **measure** on \mathcal{A} if it is additive, i.e., $\mu(\bigcup_{k\geq 1} A_k) = \sum_{k\geq 1} \mu(A_k)$ for all pairwise disjoint sets $A_k \in \mathcal{A}$, and satisfies $\mu(\emptyset) = 0$. Note that measures are always assumed to be nonnegative. A **measure space** is a triple $(\Omega, \mathcal{A}, \mu)$ consisting of a set Ω , a σ -algebra \mathcal{A} of subsets of Ω , and a measure μ on \mathcal{A} . A **probability space** is a measure space with total measure $\mu(\Omega) = 1$.

Given a function $f: \Omega \longrightarrow \mathbb{R}$, its L_p -norm is defined by

$$\parallel f \parallel_p = \left(\int_{\Omega} |f(\omega)|^p \mu(d\omega) \right)^{\frac{1}{p}}$$

Then, $L_p(\Omega, \mathcal{A}, \mu)$ denotes the set of functions $f : \Omega \longrightarrow \mathbb{R}$ which satisfy $|| f ||_p < \infty$. The L_p -norm defines a metric structure on $L_p(\Omega, \mathcal{A}, \mu)$, namely, by taking $|| f - g ||_p$ as distance between two functions $f, g \in L_p(\Omega, \mathcal{A}, \mu)$.

A distance space (X, d) is said to be L_p -embeddable if it is a subspace of $L_p(\Omega, \mathcal{A}, \mu)$ for some measure space $(\Omega, \mathcal{A}, \mu)$.

The most classical example of an L_p -space is the space $L_p(\Omega, \mathcal{A}, \mu)$, where Ω is the open interval (0, 1), \mathcal{A} is the family of Borel subsets of (0, 1), and μ is the Lebesgue measure; it is simply denoted by $L_p(0, 1)$.

We now make precise the connections existing between L_p -spaces and ℓ_p -spaces.

If $\Omega = \mathbb{N}$, $\mathcal{A} = 2^{\Omega}$ is the collection of all subsets of Ω , and μ is the cardinality measure, i.e., $\mu(A) = |A|$ if A is a finite subset of Ω and $\mu(A) = \infty$ otherwise, then $L_p(\mathbb{N}, 2^{\mathbb{N}}, |.|)$ coincides with the space ℓ_p^{∞} . If $\Omega = V_m$ is a set of cardinality m, $\mathcal{A} = 2^{\Omega}$, and μ is the cardinality measure, then $L_p(V_m, 2^{V_m}, |.|)$ coincides with ℓ_p^m .

In other words, ℓ_p^m is an isometric subspace of ℓ_p^∞ which, in turn, is L_p -embeddable.

Finally, we introduce one more semimetric space. Let $(\Omega, \mathcal{A}, \mu)$ be a measure space. Set $\mathcal{A}_{\mu} = \{A \in \mathcal{A} : \mu(A) < \infty\}$. One can define a distance d_{μ} on \mathcal{A}_{μ} by setting

$$d_{\mu}(A,B) = \mu(A \triangle B)$$

for all $A, B \in \mathcal{A}_{\mu}$. Then, d_{μ} is a semimetric on \mathcal{A}_{μ} . We call it a **measure semimetric**, and the space $(\mathcal{A}_{\mu}, d_{\mu})$ is called a **measure semimetric space**. The semimetric d_{μ} is also called a Fréchet-Nikodym-Aronszajn distance in the literature. We will consider in Section 6.3 the related Steinhaus distance, which is defined by

$$\frac{\mu(A \triangle B)}{\mu(A \cap B)}$$

for $A, B \in \mathcal{A}_{\mu}$. Note that the measure semimetric space $(\mathcal{A}_{\mu}, d_{\mu})$ is the subspace of $L_1(\Omega, \mathcal{A}, \mu)$ consisting of its 0-1 valued functions. Moreover, if $\Omega = V_m$ is a finite set of cardinality $m, \mathcal{A} = 2^{\Omega}$, and μ is the cardinality measure, then the space $(\mathcal{A}_{\mu}, d_{\mu})$ coincides with the hypercube metric space $(\{0, 1\}^m, d_{\ell_1})$.

1.0.3 Finitude result for L_p -embeddability

Though we are mainly concerned with finite distance spaces, i.e., distance spaces (X, d) with X finite, we also present some results involving infinite distance spaces. For instance, we study in Section 5.1 the normed spaces whose norm metric is L_1 -embeddable. However, the following fundamental result shows that the study of L_p -embeddable spaces can be reduced to the finite case.

THEOREM 1.4 [BCK66] Let $p \ge 1$ and let (X, d) be a distance space. Then, (X, d) is L_p -embeddable if and only if every finite subspace of (X, d) is L_p -embeddable.

We consider in detail L_1 -embeddable distance spaces in Section 2. In particular, we show that, for a finite distance space, the properties of being ℓ_1 -, ℓ_1^{∞} , or L_1 -embeddable are all equivalent to the property of belonging to the cut cone (see Theorem 2.11).

Similar results are known for the case p = 2 (see, e.g., [WW75]). Namely, for a finite distance space (X,d), the properties of being ℓ_2 -, ℓ_2^{∞} -, or L_2 -embeddable are all equivalent (and, then, (X,d) embeds in $\ell_2^{|X|-1}$; see relation (4.18)). Moreover, if X is countable, then (X,d) is ℓ_2^{∞} -embeddable (or, equivalently, L_2 -embeddable) if and only if

each subspace of (X, d) on n + 1 points embeds in ℓ_2^n . A crucial result in the case p = 2 is the well known result by Schoenberg [Sch38b], which shows that L_2 -embeddable spaces can be characterized by the negative type inequalities; see Theorem 4.16 in Section 4. This contrasts with the case p = 1 where no complete characterization by inequalities is known for L_1 -embeddable spaces.

Isometric embeddings among the L_p -spaces.

There is a vast literature on the topic of isometric embeddings among the various L_p -spaces; see, e.g., [WW75, Dor76, Bal87, LV93]. We summarize here some of the main results.

THEOREM 1.5 [Dor76] Let $1 \le p < \infty$, $1 \le r \le \infty$ and $m \in \mathbb{N}, m \ge 2$. Then, ℓ_r^m is an isometric subspace of $L_p(0, 1)$ if and only if one of the following assertions holds. (i) $p \le r < 2$. (ii) r = 2. (iii) m = 2 and p = 1.

Hence, for instance, ℓ_r^3 does not embed isometrically in $L_p(0,1)$ if r > 2. It was already shown in [BCK66] that $L_p(0,1)$ embeds isometrically in $L_1(0,1)$ for all $1 \le p \le 2$.

As a reformulation of the above Theorem, we have the following implications for a distance space (X, d).

• If (X, d) is ℓ_p^2 -embeddable for some $1 \le p \le \infty$, then (X, d) is L_1 -embeddable.

• If (X, d) is ℓ_p -embeddable for some $1 \le p \le 2$, then (X, d) is L_1 -embeddable.

• If (X, d) is ℓ_2 -embeddable, then (X, d) is L_p -embeddable for all $1 \le p \le \infty$.

Let $r \neq p$ such that $1 \leq r, p < \infty$ and let $m \geq 1$ be an integer. Then, ℓ_r^m embeds isometrically in ℓ_p^n for some integer $n \geq 1$ if and only if r = 2 and p is an even integer ([LV93]). Given an even integer p, we can define N(m, p) as the smallest integer $n \geq 1$ for which ℓ_2^m embeds isometrically into ℓ_p^n . It is shown in [LV93] that $N(2,p) = \frac{p}{2} + 1$ and that, for any $p \geq 2$ and $m \geq 1$, $\max(N(m-1,p), N(m,p-2)) \leq N(m,p) \leq \binom{m+p-1}{m-1}$. An exact evaluation of N(m,p) is known for small values of p, m; for instance, N(3,4) = 6, N(3,6) = 11, N(3,8) =16, N(7,4) = 28, N(8,6) = 120, N(23,4) = 276, N(23,6) = 2300, N(24,10) = 98280.

Therefore, given r and $m \in \mathbb{N}$ such that $1 < r \leq 2 < m$, we have that ℓ_r^m does not embed isometrically into ℓ_1^n (n positive integer), but ℓ_r^m embeds into $L_1(0, 1)$ and, moreover, every finite subspace of ℓ_r^m on s points embeds into $\ell_1^{\binom{s}{2}}$.

2 The cut cone and ℓ_1 -metrics

In this Section, we show how the members of the cut cone can be interpreted in terms of metrics and measure spaces. We essentially follow [Ass80b] and [AD82].

In order to make the Chapter self-contained, we recall the definition of the cut cone.

Given a subset S of $V_n = \{1, \ldots, n\}, \delta(S)$ denotes the **cut semimetric** defined by

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(2.1)
$$\delta(S)_{ij} = 1 \text{ if } |S \cap \{i, j\}| = 1, \text{ and } \delta(S)_{ij} = 0 \text{ otherwise},$$

for all $1 \leq i < j \leq n$. Then, CUT_n denotes the cone in \mathbb{R}^{E_n} generated by the cut semimetrics $\delta(S)$ for all subsets $S \subseteq V_n$, CUT_n is called the **cut cone** on n points, and the **cut polytope** CUT_n^{\square} denotes the polytope in \mathbb{R}^{E_n} whose vertices are the cut semimetrics $\delta(S)$ for all subsets S of V_n . So,

(2.2)
$$\operatorname{CUT}_{n} = \{ \sum_{S \subseteq V_{n}} \lambda_{S} \delta(S) : \lambda_{S} \ge 0 \text{ for all } S \subseteq V_{n} \},$$

(2.3)
$$\operatorname{CUT}_{n}^{\Box} = \{ \sum_{S \subseteq V_{n}} \lambda_{S} \delta(S) : \sum_{S \subseteq V_{n}} \lambda_{S} = 1 \text{ and } \lambda_{S} \ge 0 \text{ for all } S \subseteq V_{n} \}.$$

If one considers an arbitrary finite set X instead of V_n , then one defines similarly the cut cone CUT(X) and the cut polytope $\text{CUT}^{\square}(X)$ on X. So, $\text{CUT}(V_n) = \text{CUT}_n$ and $\text{CUT}^{\square}(V_n) = \text{CUT}_n^{\square}$.

2.1 ℓ_1 -spaces (finite case)

Clearly, every member d of the cut cone CUT_n defines a semimetric on n points. Hence arises the question of characterizing the class of semimetrics that belong to the cut cone. Several equivalent characterizations are stated in Theorem 2.11. We now present several intermediate results.

PROPOSITION 2.4 Let $d = (d_{ij})_{1 \leq i < j \leq n} \in \mathbb{R}^{E_n}$. The following assertions are equivalent. (i) $d \in CUT_n$ (resp. $d \in CUT_n^{\Box}$). (ii) There exist a measure space (resp. a probability space) $(\Omega, \mathcal{A}, \mu)$ and $A_1, \ldots, A_n \in \mathcal{A}$ such that $d_{ij} = \mu(A_i \triangle A_i)$ for all $1 \leq i < j \leq n$.

PROOF. Assume $d \in \text{CUT}_n$. Then, $d = \sum_{S \subseteq \{1,...,n\}} \lambda_S \delta(S)$ for some $\lambda_S \ge 0$. We define a measure space $(\Omega, \mathcal{A}, \mu)$ as follows. Let Ω denote the family of subsets of $\{1, \ldots, n\}$, let \mathcal{A} denote the family of subsets of Ω and let μ denote the measure on \mathcal{A} defined by $\mu(A) = \sum_{S \in A} \lambda_S$ for each $A \in \mathcal{A}$ (i.e., A is a collection of subsets of $\{1, \ldots, n\}$). Define $A_i = \{S \in \Omega : i \in S\}$. Then, $\mu(A_i \triangle A_j) = \mu(\{S \in \Omega : |S \cap \{i, j\}| = 1\})$ $= \sum_{S \in \Omega: |S \cap \{i, j\}| = 1} \lambda_S = d_{ij}$ holds, for all $1 \le i < j \le n$. Moreover, if $d \in \text{CUT}_n^{\Box}$, then we have $\sum_S \lambda_S = 1$, i.e. $\mu(\Omega) = 1$, that is $(\Omega, \mathcal{A}, \mu)$ is a probability space.

Conversely, assume $d_{ij} = \mu(A_i \triangle A_j)$ for $1 \le i < j \le n$, where $(\Omega, \mathcal{A}, \mu)$ is a measure space and $A_1, \ldots, A_n \in \mathcal{A}$. Set $A^S = \bigcap_{i \in S} A_i \cap \bigcap_{i \notin S} (\Omega \setminus A_i)$ for each $S \subseteq \{1, \ldots, n\}$. Then, $A_i = \bigcup_{S:i \in S} A^S$, $A_i \triangle A_j = \bigcup_{S:|S \cap \{i,j\}|=1} A_S$ and $\Omega = \bigcup_S A^S$. Therefore, $d = \sum_{S \subseteq \{1,\ldots,n\}} \mu(A^S) \delta(S)$, showing that d belongs to the cut cone CUT_n . Moreover, if $(\Omega, \mathcal{A}, \mu)$ is a probability space, i.e., $\mu(\Omega) = 1$, then $\sum_{S} \mu(A^{S}) = 1$, implying that d belongs to the cut polytope $\operatorname{CUT}_{n}^{\Box}$.

PROPOSITION 2.5 Let $d \in \mathbb{R}^{E_n}$ and (V_n, d) the associated distance space. The following assertions are equivalent.

(i) $d \in CUT_n$. (ii) (V_n, d) is ℓ_1 -embeddable, i.e., there exist n vectors $u_1, \ldots, u_n \in \mathbb{R}^m$ for some m such that $d_{ij} = \| u_i - u_j \|_1$ for all $1 \le i < j \le n$.

PROOF. $(i) \Rightarrow (ii)$. Suppose that $d \in \text{CUT}_n$. Then, $d = \sum_{1 \le k \le m} \lambda_k \delta(S_k)$ with $\lambda_1, \ldots, \lambda_m \ge 0$. For $1 \le i \le n$, define the vector $u_i \in \mathbb{R}^m$ with components $(u_i)_k = \lambda_k$ if $i \in S_k$ and $(u_i)_k = 0$ otherwise, for $1 \le k \le m$. Then $d_{ij} = || u_i - u_j ||_1$ holds, showing that (V_n, d) is ℓ_1 -embeddable.

 $(ii) \Rightarrow (i)$. Assume that (V_n, d) is ℓ_1 -embeddable, i.e., there exist n vectors $u_1, \ldots, u_n \in \mathbb{R}^m$ for some $m \ge 1$ such that $d_{ij} = || u_i - u_j ||_1$, for $1 \le i < j \le n$. We show that $d \in \text{CUT}_n$. It suffices to show the result for the case m = 1 by additivity of the ℓ_1 -norm. Hence, $d_{ij} = |u_i - u_j|$ where $u_1, \ldots, u_n \in \mathbb{R}$. Without loss of generality, we can suppose that $0 = u_1 \le u_2 \le \ldots \le u_n$. Then, $d = \sum_{1 \le k \le n-1} (u_{k+1} - u_k) \delta(\{1, 2, \ldots, k - 1, k\})$ holds, showing that $d \in \text{CUT}_n$.

REMARK 2.6 The proof of Proposition 2.5 shows that, if d is a distance on V_n , then d is ℓ_1^m -embeddable whenever d can be decomposed as a nonnegative combination of m cut semimetrics.

There is a characterization for hypercube embeddable semimetrics analogue to that of Proposition 2.5.

PROPOSITION 2.7 Let $d \in \mathbb{R}^{E_n}$ and (V_n, d) be the associated distance space. The following assertions are equivalent.

(i) $d = \sum_{S} \lambda_{S} \delta(S)$ for some nonnegative integer scalars λ_{S} .

(ii) (V_n, d) is hypercube embeddable, i.e., there exist n vectors $u_1, \ldots, u_n \in \{0, 1\}^m$ for some m such that $d_{ij} = || u_i - u_j ||_1$ for all $1 \le i < j \le n$.

(iii) There exist a finite set Ω and n subsets A_1, \ldots, A_n of Ω such that $d_{ij} = |A_i \triangle A_j|$ for all $1 \le i < j \le n$.

(iv) (V_n, d) is an isometric subspace of $(\mathbb{Z}^m, d_{\ell_1})$ for some integer $m \geq 1$.

PROOF. The proof of $(i) \iff (ii)$ is analogous to that of Proposition 2.5. Namely, for $(i) \Longrightarrow (ii)$, assume $d = \sum_{1 \le k \le m} \delta(S_k)$ (allowing repetitions). Consider the binary $n \times m$ matrix M whose columns are the incidence vectors of the sets S_1, \ldots, S_m . If u_1, \ldots, u_n denote the rows of M, then $d_{ij} = || u_i - u_j ||_1$ holds, providing an embedding of (V_n, d) in the hypercube of dimension m. Conversely, for $(ii) \Longrightarrow (i)$, consider the matrix M whose rows are the n given vectors u_1, \ldots, u_n . Let S_1, \ldots, S_m be the subsets of $\{1, \ldots, n\}$ whose incidence vectors are the columns of M. Then, $d = \sum_{1 \le k \le m} \delta(S_k)$ holds, giving a decomposition of d as a nonnegative integer combination of cuts.

(iii) is a reformulation of (ii), $(iii) \Longrightarrow (iv)$ is obvious, and $(iv) \Longrightarrow (i)$ follows from the proof of the implication $(ii) \Longrightarrow (i)$ of Proposition 2.5.

The next result follows immediately from Propositions 2.5 and 2.7.

PROPOSITION 2.8 Let (V_n, d) be a distance space where d is rational valued. Then, (V_n, d) is ℓ_1 -embeddable if and only if $(V_n, \lambda d)$ is hypercube embeddable for some scalar λ .

The equivalence $(i) \iff (ii)$ from Proposition 2.7 can be generalized in the context of Hamming spaces and multicuts.

Recall that, given $x, y \in \mathbb{R}^k$, their **Hamming distance** $d_H(x, y)$ is defined as the number of positions where the coordinates of x and y differ. Hence, when considered on binary vectors, the Hamming distance coincides with the ℓ_1 -distance.

Let $q \geq 2$ be an integer and let S_1, \ldots, S_q be q subsets of V_n that partition V_n . Then, the **multicut vector** $\delta(S_1, \ldots, S_q)$ is the vector of \mathbb{R}^{E_n} defined by

 $\begin{aligned} \delta(S_1, \dots, S_q)_{ij} &= 0 & \text{if } i, j \in S_h \text{ for some } h, \ 1 \leq h \leq p, \\ \delta(S_1, \dots, S_q)_{ij} &= 1 & \text{otherwise} \\ \text{for } 1 \leq i < j \leq n. \end{aligned}$

PROPOSITION 2.9 Let $d \in \mathbb{R}^{E_n}$ and (V_n, d) be the associated distance space. The following assertions are equivalent.

(i) $d = \sum_{(S_1,\ldots,S_q)} partition \text{ of } _{V_n} \lambda_{S_1\ldots,S_q} \delta(S_1,\ldots,S_q)$ for some nonnegative integers $\lambda_{S_1\ldots,S_q}$.

(ii) (V_n, d) is an isometric subspace of the Hamming space $(\{0, 1, \ldots, q-1\}^m, d_H)$ for some integer $m \geq 1$.

The following result will permit us to link ℓ_1 - and L_1 -embeddability.

LEMMA 2.10 Let (X, d) be a distance space. The following assertions are equivalent. (i) (X, d) is L_1 -embeddable.

(ii) (X, d) is a subspace of a measure semimetric space $(\mathcal{A}_{\mu}, d_{\mu})$ for some measure space $(\Omega, \mathcal{A}, \mu)$.

PROOF. The implication $(ii) \Rightarrow (i)$ is clear, since $(\mathcal{A}_{\mu}, d_{\mu})$ is a subspace of $L_1(\Omega, \mathcal{A}, \mu)$. We check $(i) \Rightarrow (ii)$. It suffices to show that each space $L_1(\Omega, \mathcal{A}, \mu)$ is a subspace of $(\mathcal{B}_{\nu}, d_{\nu})$ for some measure space (T, \mathcal{B}, ν) . Set $T = \Omega \times \mathbb{R}$, $\mathcal{B} = \mathcal{A} \times \mathcal{R}$ where \mathcal{R} is the family of Borel subsets of \mathbb{R} , and $\nu = \mu \otimes \lambda$ where λ is the Lebesgue measure on \mathbb{R} . For $f \in L_1(\Omega, \mathcal{A}, \mu)$, let $E(f) = \{(\omega, s) \in \Omega \times \mathbb{R} : s > f(\omega)\}$ denote its epigraph. Then, the map $f \mapsto E(f) \triangle E(0)$ provides an isometric embedding from $L_1(\Omega, \mathcal{A}, \mu)$ to $(\mathcal{B}_{\nu}, d_{\nu})$, since $\parallel f - g \parallel_1 = \nu(E(f) \triangle E(g))$ holds.

We summarize in the next Theorem the equivalent characterizations that we have obtained for members of the cut cone CUT_n .

THEOREM 2.11 Let $d \in \mathbb{R}^{E_n}$ and (V_n, d) be the associated distance space. The following assertions are equivalent.

(i) $d \in CUT_n$.

(*ii*) (V_n, d) is ℓ_1 -embeddable.

(iii) (V_n, d) is L_1 -embeddable.

(iv) There exist a measure space $(\Omega, \mathcal{A}, \mu)$ and $A_1, \ldots, A_n \in \mathcal{A}$ such that $d_{ij} = \mu(A_i \triangle A_i)$ for all $1 \le i < j \le n$.

(v) (V_n, d) is an isometric subspace of ℓ_1^{∞} .

2.2 L_1 -spaces (infinite case)

Theorem 2.11 remains partially valid for the case of a distance space (X,d) where the set X is infinite. Indeed, the equivalence $(iii) \iff (iv)$ holds by Lemma 2.10 and the implication $(ii) \implies (iii)$ holds trivially. In fact, there is an infinite analogue of the cut cone, as we now see.

For each subset Y of X, let δ_Y denote the cut function induced by Y defined by $\delta_Y(x,y) = 1$ if $|Y \cap \{x,y\}| = 1$, $\delta_Y(x,y) = 0$ otherwise, for $x, y \in X$; so δ_Y is just the symmetric function corresponding to the cut semimetric $\delta(Y)$. Let $\mathcal{D}(X)$ denote the set of all cut functions δ_Y for $Y \subseteq X$.

Let $C_1(X)$ denote the set of all semimetrics d on X for which (X, d) is L_1 -embeddable.

THEOREM 2.12 Let (X, d) be a distance space. The following assertions are equivalent. (i) (X, d) is L_1 -embeddable.

(ii) There exists a measure ν on $\mathcal{D}(X)$ such that $d(x,y) = \int_{\mathcal{D}(X)} \delta(x,y)\nu(d\delta)$ for $x, y \in X$.

PROOF. $(i) \Rightarrow (ii)$. Assume (X, d) is L_1 -embeddable. Then, by Lemma 2.10, there exist a measure space $(\Omega, \mathcal{A}, \mu)$ and a map $x \mapsto A_x$ from X to \mathcal{A}_μ such that $d(x, y) = \mu(A_x \triangle A_y)$

for $x, y \in X$. For $\omega \in \Omega$, set $A^{\omega} = \{x \in X : \omega \in A_x\}$. We define a measure ν on $\mathcal{D}(X)$ additively by setting $\nu(\{\delta_Y\}) = \mu(\{\omega \in \Omega : A^{\omega} = Y\})$ for each $Y \subseteq X$. Note that $\omega \in A_x$ if and only if $x \in A^{\omega}$ and $\omega \in A_x \triangle A_y$ if and only if $|A^{\omega} \cap \{x, y\}| = 1$. Therefore,

$$d(x,y) = \mu(A_x \triangle A_y) = \mu(\{\omega \in \Omega : |A^{\omega} \cap \{x,y\}| = 1\})$$

= $\mu(\{\omega \in \Omega : \delta_{A^{\omega}}(x,y) = 1\})$
= $\mu(\bigcup_{Y \subseteq X: \delta_Y(x,y) = 1} \{\omega \in \Omega : A^{\omega} = Y\})$
= $\int_{\mathcal{D}(X)} \delta(x,y)\nu(d\delta).$

 $(ii) \Rightarrow (i)$. Conversely, assume that $d = \int_{\mathcal{D}(X)} \delta\nu(d\delta)$ for some non negative measure on $\mathcal{D}(X)$. Fix $s \in X$ and set $A_x = \{\delta \in \mathcal{D}(X) : \delta(s, x) = 1\}$ for each $x \in X$. Then, $d(x, y) = \nu(A_x \triangle A_y)$ holds, since $\delta(x, y) = 0$ if $\delta \notin A_x \triangle A_y$ and $\delta(x, y) = 1$ if $\delta \in A_x \triangle A_y$. This shows, using Lemma 2.10, that (X, d) is L_1 -embeddable.

THEOREM 2.13 (i) $C_1(X)$ is a convex cone.

(ii) The extremal rays of $C_1(X)$ are the rays generated by the nonzero cut functions δ_Y for $Y \subseteq X$, $\emptyset \neq Y \neq X$.

PROOF. (i) follows from Corollary 4.25 (i).

We check (*ii*). It is easy to see that each cut function lies on an extreme ray of $C_1(X)$ (it lies, in fact, on an extreme ray of the semimetric cone). Consider now $d \in C_1(X)$ which is not a cut function. We can suppose that $d(x_1, x_2) = 1$, $d(x_1, x_3) = \alpha > 0$ and $d(x_2, x_3) = \beta > 0$ for some $x_1, x_2, x_3 \in X$ with $\alpha \ge \beta$. Set $d_1 = \int_{\mathcal{D}(X)} \delta(x_1, x_2) \delta(x_1, x_3) \delta \nu(d\delta)$ and $d_2 = d - d_1$. Then, $d_1, d_2 \in C_1(X)$ by Theorem 2.12. But $d_1(x_1, x_2) = \frac{1+\alpha-\beta}{2} > 0$, since $2\delta(x_1, x_2)\delta(x_1, x_3) = \delta(x_1, x_2) + \delta(x_1, x_3) - \delta(x_2, x_3)$ for each cut function δ . Also, $d_1(x_2, x_3) = 0$ and $d_2(x_2, x_3) = \beta$. Therefore d does not lie on an extreme ray of $C_1(X)$ since $d = d_1 + d_2$ where d_1 and d_2 are not proportional to d.

By Theorem 1.4 applied in the case p = 1, we have that a distance space is L_1 -embeddable if and only if every finite subspace of it is L_1 -embeddable. Therefore, L_1 -embeddability can be characterized by a family of linear inequalities, each involving only a finite number of variables. In other words, characterizing L_1 -embeddability amounts to finding the facet defining inequalities of the cut cone CUT_n for all $n \geq 2$.

3 The correlation cone and $\{0, 1\}$ -covariances

As before, we set $V_n = \{1, ..., n\}$ and $E_n = \{ij : i, j \in V_n, i \neq j\}$ denotes the set of unordered pairs of elements of V_n . Given a subset S of V_n , let $\pi(S) = (\pi(S)_{ij})_{1 \leq i \leq j \leq n} \in$ $\mathbb{R}^{V_n \cup E_n}$ (identifying the diagonal pair *ii* with the element $i \in V_n$) be defined by

(3.1)
$$\pi(S)_{ij} = 1 \text{ if } i, j \in S \text{ and } \pi(S)_{ij} = 0 \text{ otherwise}$$

for all $i, j \in V_n$; $\pi(S)$ is called a **correlation vector**. The **correlation cone** COR_n is the cone generated by all correlation vectors $\pi(S)$ for $S \subseteq V_n$ and the **correlation polytope** $\operatorname{COR}_n^{\square}$ is the convex hull of the correlation vectors $\pi(S)$ for $S \subseteq V_n$. So,

(3.2)
$$\operatorname{COR}_{n} = \{ \sum_{S \subseteq V_{n}} \lambda_{S} \pi(S) : \lambda_{S} \ge 0 \text{ for all } S \subseteq V_{n} \},$$

(3.3)
$$\operatorname{COR}_{n}^{\Box} = \{ \sum_{S \subseteq V_{n}} \lambda_{S} \pi(S) : \sum_{S \subseteq V_{n}} \lambda_{S} = 1 \text{ and } \lambda_{S} \ge 0 \text{ for all } S \subseteq V_{n} \}.$$

The correlation cone and/or polytope were considered in many papers, among them, [MD72, Dez73, Erd87, Isa89, Pad89, Sim90, BH91, Pit86, Pit91]. The polytope COR_n^{\Box} is also known under the name of **boolean quadric polytope** ([Pad89]). The terminology "correlation polytope" was introduced by Pitowsky ([Pit91]). It is motivated by the fact that COR_n^{\Box} arises naturally in the context of probability; see Proposition 3.13. Actually, this interpretation was already used in [MD72] in the context of quantum mechanics for describing the pair distributions of particles in lattice sites.

It is sometimes convenient to consider an arbitrary finite subset X instead of V_n . Then, the correlation cone is denoted by COR(X) and the correlation polytope by $COR^{\square}(X)$.

3.1 The covariance mapping

A simple but fundamental property is that the cut cone CUT_{n+1} and the correlation cone COR_n (resp. the cut polytope $\text{CUT}_{n+1}^{\square}$ and the correlation polytope COR_n^{\square}) are in one-to-one correspondance via the following covariance mapping.

The **covariance mapping** ξ is the mapping from the space $\mathbb{R}^{E_{n+1}}$ (indexed by the $\binom{n+1}{2}$ pairs of elements of V_{n+1}) to the space $\mathbb{R}^{V_n \cup E_n}$ (indexed by the *n* elements of V_n and the $\binom{n}{2}$ pairs of elements of V_n) defined by $p = \xi(d)$ for $d = (d_{ij})_{1 \le i < j \le n+1}$ and $p = (p_{ij})_{1 \le i \le j \le n}$ with

(3.4)
$$p_{ij} = \frac{1}{2}(d_{i,n+1} + d_{j,n+1} - d_{ij}) \text{ for all } 1 \le i \le j \le n$$

or, equivalently,

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(3.5)
$$\begin{cases} d_{ij} = p_{ii} + p_{jj} - 2p_{ij} & \text{for } 1 \le i < j \le n \\ d_{i,n+1} = p_{ii} & \text{for } 1 \le i \le n. \end{cases}$$

The covariance mapping ξ is a linear bijection from $\mathbb{R}^{E_{n+1}}$ to $\mathbb{R}^{V_n \cup E_n}$. One can easily check that, for any subset S of V_n , $\xi(\delta(S)) = \pi(S)$ holds. Therefore,

(3.6)
$$\xi(\operatorname{CUT}_{n+1}) = \operatorname{COR}_n \text{ and } \xi(\operatorname{CUT}_{n+1}^{\Box}) = \operatorname{COR}_{n+1}^{\Box}.$$

In the same way, given a finite subset X and an element $x_0 \in X$, the cut cone $\operatorname{CUT}(X)$ and the correlation cone $\operatorname{COR}(X \setminus \{x_0\})$ (resp. the cut polytope $\operatorname{CUT}^{\square}(X)$ and the correlation polytope $\operatorname{COR}^{\square}(X \setminus \{x_0\})$) are in one-to-one linear correspondance, via the covariance mapping ξ , also denoted as ξ_{x_0} if one wants to stress the choice of the point x_0 . For the sake of clarity, we rewrite the definition.

Let X be a set (not necessarly finite), $x_0 \in X$, let d be a distance on X and let p be a symmetric function on $X \setminus \{x_0\}$. Then, $p = \xi(d) = \xi_{x_0}(d)$ if

(3.7)
$$p(x,y) = \frac{1}{2}(d(x,x_0) + d(y,x_0) - d(x,y)) \text{ for all } x, y \in X \setminus \{x_0\}$$

or, equivalently,

(3.8)
$$\begin{cases} d(x, x_0) = p(x, x) & \text{for all } x \in X \setminus \{x_0\}, \\ d(x, y) = p(x, x) + p(y, y) - 2p(x, y) & \text{for all } x, y \in X \setminus \{x_0\}. \end{cases}$$

Therefore, for X finite,

$$\xi_{x_0}(\operatorname{CUT}(X)) = \operatorname{COR}(X \setminus \{x_0\}) \text{ and } \xi_{x_0}(\operatorname{CUT}^{\square}(X)) = \operatorname{COR}^{\square}(X \setminus \{x_0\}).$$

Note that, if one uses relation (3.7) for computing $p(x, x_0)$, then one obtains that $p(x, x_0) = 0$ for all $x \in X$. This explains why we consider p as being defined only on the pairs of elements from $X \setminus \{x_0\}$.

The covariance mapping appeared in many different areas of mathematics. See, for instance, [Cri88], [CP93] (where, for a metric space (X, d) and its image $p = \xi(d)$, the quantity p(x, y) is known as the **Gromov product** of $x, y \in X \setminus \{x_0\}$), [Fic87] (where it is called a **linear generalized similarity function**).

The connection between cut and correlation polyhedra, which is formulated in (3.6), was rediscovered independently by several authors (e.g., in [Ham65, Dez73, Sim90]).

3.2 Covariances

We now introduce the notion of *M*-covariance. This notion is studied in [Ass79, Ass80b] for *M* being a subset of a Hilbert space. We consider here only the case when $M = \mathbb{R}$ or $M = \{0, 1\}$.

DEFINITION 3.9 Let M be a subset of \mathbb{R} . A symmetric function $p: X \times X \longrightarrow \mathbb{R}$ is called an M-covariance if there exist a measure space $(\Omega, \mathcal{A}, \mu)$ and functions $f_x \in L_2(\Omega, \mathcal{A}, \mu)$ taking values in M, for all $x \in X$, such that

$$p(x,y) = \int_{\Omega} f_x(\omega) f_y(\omega) \mu(d\omega) \text{ for all } x, y \in X.$$

In particular, p is a $\{0,1\}$ -covariance if and only if there exist a measure space $(\Omega, \mathcal{A}, \mu)$ and sets $A_x \in \mathcal{A}_{\mu}$, for all $x \in X$, such that

$$p(x,y) = \mu(A_x \cap A_y)$$
 for all $x, y \in X$.

The next two Lemmas show how \mathbb{R} -covariances and $\{0, 1\}$ -covariances are related to L_2 - and L_1 -embeddable distance spaces, respectively, via the covariance mapping. These facts will be extensively used in the sequel.

LEMMA 3.10 Let X be a set and $x_0 \in X$. Let d be a distance on X and let $p = \xi_{x_0}(d)$ be the corresponding symmetric function on $X \setminus \{x_0\}$. Then, (X, \sqrt{d}) is L_2 -embeddable if and only if p is an \mathbb{R} -covariance on $X \setminus \{x_0\}$.

PROOF. It is an immediate verification.

LEMMA 3.11 [Dez73] Let X be a set and $x_0 \in X$. Let d be a distance on X and let $p = \xi_{x_0}(d)$ be the corresponding symmetric function on $X \setminus \{x_0\}$. Then, (X,d) is L_1 -embeddable if and only if p is a $\{0,1\}$ -covariance on $X \setminus \{x_0\}$.

PROOF. By Lemma 2.10, (X, d) is L_1 -embeddable if and only if there exist a measure space $(\Omega, \mathcal{A}, \mu)$ and sets $A_x \in \mathcal{A}_{\mu}$ for $x \in X$ such that $d(x, y) = \mu(A_x \triangle A_y)$ for all $x, y \in X$. Without loss of generality, we can suppose that $A_{x_0} = \emptyset$. Then, it follows from relation (3.7) that $p(x, y) = \frac{1}{2}(\mu(A_x) + \mu(A_y) - \mu(A_x \triangle A_y)) = \mu(A_x \cap A_y)$ for all $x, y \in X \setminus \{x_0\}$.

The following finitude result is a consequence of Lemma 3.11 and Theorem 1.4.

PROPOSITION 3.12 Let p be a symmetric function on X. Then, p is a $\{0,1\}$ -covariance on X if and only if, for each finite subset Y of X, the restriction of p to Y is a $\{0,1\}$ covariance on Y.

We now give an interpretation of the members of the correlation cone and polytope which is an analogue of Proposition 2.4 (via the covariance mapping). It was rediscovered in [Pit86].

PROPOSITION 3.13 Let $p = (p_{ij})_{1 \le i \le j \le n} \in \mathbb{R}^{V_n \cup E_n}$. The following assertions are equivalent. (i) $p \in COR_n$ (resp. $p \in COR_n^{\Box}$).

(ii) There exist a measure space (resp. a probability space) $(\Omega, \mathcal{A}, \mu)$ and $A_1, \ldots, A_n \in \mathcal{A}$ such that $p_{ij} = \mu(A_i \cap A_j)$ for all $1 \leq i \leq j \leq n$.

Therefore, for X finite, p is a $\{0,1\}$ -covariance on X if and only if p belongs to the correlation cone COR(X).

For the members of the correlation cone which can be written as a nonnegative *integer* combination of correlation vectors, we can assume that the measure space in Proposition 3.13 (*ii*) is endowed with the cardinality measure. Namely, we have the following result, which is an analogue of Proposition 2.7 (*i*) \iff (*iii*) (via the covariance mapping).

PROPOSITION 3.14 Let $p = (p_{ij})_{1 \le i \le j \le n} \in \mathbb{R}^{V_n \cup E_n}$. The following assertions are equivalent.

(i) $p = \sum_{S \subseteq V_n} \lambda_S \pi(S)$ for some nonnegative integers λ_S . (ii) There exist a finite set Ω and n subsets A_1, \ldots, A_n of Ω such that $p_{ij} = |A_i \cap A_j|$ for all $1 \le i \le j \le n$.

A vector p satisfying the conditions of Proposition 3.14 is sometimes called an **inter**section pattern in the literature (see, e.g., [DR84]). Testing whether a given vector p is an intersection pattern is an NP-complete problem ([Chv80]). However, this problem is polynomial when restricted to some classes of vectors; for instance, it is polynomial when retricted to the class of the vectors p such that $p_{ii} = 2$ for all $i \in V_n$. We refer to Chapter ??? (on hypercube embedding) for results related to these questions.

3.3 The Boole problem

We now describe an application of the interpretation of the correlation cone and polytope given in Proposition 3.13 for the following **Boole problem**.

Let $(\Omega, \mathcal{A}, \mu)$ be a probability space and let A_1, \ldots, A_n be *n* events of \mathcal{A} . A classical question, which goes back to Boole [Boo54], is the following:

Suppose we are given the values $p_i = \mu(A_i)$ for $1 \le i \le n$, what is the best estimation of $\mu(A_1 \cup \ldots A_n)$?

It is easy to see that the answer is $\max(p_1, \ldots, p_n) \le \mu(A_1 \cup \ldots A_n) \le \min(1, \sum_{1 \le i \le n} p_i)$. The next natural question is the following:

Suppose we are given the values $p_i = \mu(A_i)$ for $1 \le i \le n$ and the values of the joint probabilities $p_{ij} = \mu(A_i \cap A_j)$ for $1 \le i < j \le n$. What is the best estimation of $\mu(A_1 \cup \ldots \cup A_n)$ in terms of the p_i 's and the p_{ij} 's ?

In fact, an answer to this problem is given by the facet defining inequalities for the correlation polytope COR_n^{\Box} . Namely,

$$\mu(A_1 \cup \ldots \cup A_n) \ge \max(w^T p : w^T z \le 1 \text{ is facet defining for } \operatorname{COR}_n^{\Box})$$

(see Proposition 3.17 and relation (3.20)).

The approach described below for obtaining estimations of $\mu(A_1 \cup \ldots \cup A_n)$ uses linear programming; it was considered in [KM76], [Pit91]. Given $p \in \text{COR}_n$, consider the following two linear programming problems.

(3.15)
$$z_{min} := \min_{\substack{\text{subject to}\\ S \subseteq V_n}} \sum_{\substack{\emptyset \neq S \subseteq V_n\\ \emptyset \neq S \subseteq V_n}} \lambda_S \sum_{\substack{\emptyset \neq S \subseteq V_n\\ \lambda_S \geq 0 \text{ for } \emptyset \neq S \subseteq V_n}} \lambda_S = p$$

(3.16)
$$z_{max} := \max_{\substack{\text{subject to}\\ \lambda_S \ge 0 \text{ for } \emptyset \neq S \subseteq V_n}} \sum_{\substack{\emptyset \neq S \subseteq V_n \\ \lambda_S \neq S \subseteq V_n \\ \lambda_S \ge 0 \text{ for } \emptyset \neq S \subseteq V_n}} \sum_{\substack{\emptyset \neq S \subseteq V_n \\ \lambda_S \ge 0 \text{ for } \emptyset \neq S \subseteq V_n}} z_{max}$$

PROPOSITION 3.17 Let z_{min} and z_{max} be defined by the relations (3.15) and (3.16). Then, $z_{min} \leq \mu(A_1 \cup \ldots \cup A_n) \leq z_{max}$.

PROOF. Let $p \in COR_n$ be defined by $p_{ij} = \mu(A_i \cap A_j)$ for all $1 \le i \le j \le n$ (setting $p_{ii} = p_i$). For $S \subseteq V_n$, set $A^S = \bigcap_{i \in S} A_i \cap \bigcap_{i \notin S} (\Omega \setminus A_i)$. Then, $A_i \cap A_j = \bigcup_{S \subseteq V_n: i, j \in S} A^S$, $\Omega = \bigcup_{S \subseteq V_n} A^S$ and $A_1 \cup \ldots \cup A_n = \bigcup_{S \subseteq V_n: S \neq \emptyset} A^S$. Therefore, $p = \sum_{S \subseteq V_n: S \neq \emptyset} \mu(A^S)\pi(S)$ holds, with $\mu(A^S) \ge 0$ for all S. Hence $(\mu(A^S) : \emptyset \ne S \subseteq V_n)$ is a feasible solution to the programs (3.15) and (3.16), with objective value $\mu(A_1 \cup \ldots \cup A_n)$. This shows the result.

The dual programs to (3.15) and (3.16) are the following programs (3.18) and (3.19), respectively.

(3.18)
$$\begin{array}{c} \max & w^T p \\ \text{subject to} & w^T \pi(S) \leq 1 \quad \text{for } \emptyset \neq S \subseteq V_n \end{array}$$

(3.19)
$$\begin{array}{ccc} \min & w^T p \\ \text{subject to} & w^T \pi(S) \ge 1 \quad \text{for } \emptyset \neq S \subseteq V_n \end{array}$$

By linear programming duality, we obtain that

(3.20)
$$z_{min} = \max(w^T p : w^T z \le 1 \text{ is a valid inequality for } \operatorname{COR}_n^{\Box})$$

and one can easily verify that, in relation (3.20), it is sufficient to consider facet defining inequalities. Similarly,

$$z_{max} = \min(w^T p : w^T z \ge 1)$$
 is facet defining for the polytope

$$\operatorname{Conv}(\{\pi(S): \emptyset \neq S \subseteq V_n\})$$

(The latter polytope is distinct from COR_n^{\Box} since it does not contain the origin.)

Therefore, by (3.20), every valid inequality for $\operatorname{COR}_n^{\Box}$ yields a lower bound for $\mu(A_1 \cup \ldots \cup A_n)$ in terms of the joint probabilities $p_{ij} = \mu(A_i \cap A_j)$ for $1 \leq i \leq j \leq n$. We now give some examples of such lower bounds.

Suppose $p = \sum_{S} \lambda_{S} \pi(S)$ with $\lambda_{S} \geq 0$ for all S. Let $u \in \mathbb{R}^{V_{n} \cup E_{n}}$ be defined by $u_{i} = n$ for $i \in V_{n}$ and $u_{ij} = -2$ for $ij \in E_{n}$. By taking the scalar product of both sides of $p = \sum_{S} \lambda_{S} \pi(S)$ with u, we obtain that $n \sum_{1 \leq i \leq n} p_{i} - 2 \sum_{1 \leq i < j \leq n} p_{ij} = \sum_{S} \lambda_{S} |S|(n+1-|S|)$, where $n \leq |S|(n+1-|S|) \leq \lfloor \frac{n+1}{2} \rfloor \lceil \frac{n+1}{2} \rceil$ if $S \neq \emptyset$. Hence, we deduce that

$$\frac{n\sum_{1\leq i\leq n}p_i - 2\sum_{1\leq i< j\leq n}p_{ij}}{\lfloor\frac{n+1}{2}\rfloor\lceil\frac{n+1}{2}\rceil} \leq \sum_{\emptyset\neq S\subseteq V_n}\lambda_S \leq \frac{n\sum_{1\leq i\leq n}p_i - 2\sum_{1\leq i< j\leq n}p_{ij}}{n}$$

and, therefore, from the definition of z_{min} , z_{max} and from Proposition 3.17,

(3.21)
$$\frac{n\sum_{1\leq i\leq n}p_i-2\sum_{1\leq i< j\leq n}p_{ij}}{\lfloor\frac{n+1}{2}\rfloor\lceil\frac{n+1}{2}\rceil} \leq \mu(A_1\cup\ldots\cup A_n) \leq \frac{n\sum_{1\leq i\leq n}p_i-2\sum_{1\leq i< j\leq n}p_{ij}}{n}.$$

The inequality

(3.22)
$$2k \sum_{1 \le i \le n} p_i - 2 \sum_{1 \le i < j \le n} p_{ij} \le k(k+1)$$

is valid for the correlation polytope COR_n^{\square} , for all $1 \leq k \leq n-1$; it is facet defining if $1 \leq k \leq n-2$ and $n \geq 4$. Setting $b_{n+1} = 2k+1-n$ and $b_1 = \ldots = b_n = 1$, the inequality (3.22) corresponds (via the covariance map) to the inequality

(3.23)
$$\sum_{1 \le i < j \le n+1} b_i b_j x_{ij} \le k(k+1)$$

which is valid for the cut polytope $\text{CUT}_{n+1}^{\square}$. Note that the inequality (3.23) is a switching of the hypermetric inequality $Hyp_{n+1}(1, \ldots, 1, -1, \ldots, -1, 2k + 1 - n)$ (with n - k coefficients +1 and k coefficients -1) (see Section 4.1 for definitions). Therefore, we have the following lower bound for $\mu(A_1 \cup \ldots \cup A_n)$:

(3.24)
$$\frac{2}{k+1} \sum_{1 \le i \le n} p_i - \frac{2}{k(k+1)} \sum_{1 \le i < j \le n} p_{ij} \le \mu(A_1 \cup \ldots \cup A_n)$$

for each $k, 1 \le k \le n-1$. The bound (3.24) was found independently by several authors, including [Chu41, DS67, Gal77]. Note that (3.24) coincides with the lower bound of (3.21) in the case n = 2k. The case k = 1 of (3.24) gives the bound

$$\sum_{1 \le i \le n} p_i - \sum_{1 \le i < j \le n} p_{ij} \le \mu(A_1 \cup \ldots \cup A_n)$$

which is a special case of the Bonferroni bound (3.29) mentioned below.

More generally, given integers b_1, \ldots, b_n and $k \ge 0$, the inequality

(3.25)
$$\sum_{1 \le i \le n} b_i (2k+1-b_i) p_i - 2 \sum_{1 \le i < j \le n} b_i b_j p_{ij} \le k(k+1)$$

is valid for $\operatorname{COR}_n^{\Box}$. This yields the bound

$$\frac{1}{k(k+1)} \left(\sum_{1 \le i \le n} p_i b_i (2k+1-b_i) - 2 \sum_{1 \le i < j \le n} b_i b_j p_{ij} \right) \le \mu(A_1 \cup \ldots \cup A_n).$$

The inequality (3.25) can alternatively be written as

(3.26)
$$(\sum_{1 \le i \le n} b_i p_i - k) (\sum_{1 \le i \le n} b_i p_i - k - 1) \ge 0$$

with the convention that, when developing the product, the expression $p_i p_j$ is replaced by the variable p_{ij} (setting $p_{ii} = p_i$). The inequality (3.25) (or (3.26)) (or special cases of it) was considered by many authors (e.g., [Yos70, MD72, KM76, Erd87, Mes87, Pit91]).

(Note that, if we set $b_{n+1} = 2k + 1 - \sum_{1 \le i \le n} b_i$, the inequality (3.25) corresponds (via the covariance mapping) to the inequality

$$\sum_{1 \le i < j \le n+1} b_i b_j x_{ij} \le k(k+1),$$

which is valid for the cut polytope $\text{CUT}_{n+1}^{\square}$.)

3.4 Generalization to higher order correlations

Clearly, much of the treatment of Section 3.3 can be generalized to higher order correlations. Namely, let \mathcal{I} be a family of subsets of V_n . Given a subset S of V_n , its \mathcal{I} -correlation vector $\pi^{\mathcal{I}}(S) \in \mathbb{R}^{\mathcal{I}}$ is defined by

 $\begin{cases} \pi^{\mathcal{I}}(S)_I = 1 & \text{if } I \subseteq S, \\ \pi^{\mathcal{I}}(S)_I = 0 & \text{otherwise,} \end{cases}$

for all $I \in \mathcal{I}$. Then, the cone $\operatorname{COR}_n(\mathcal{I})$ (resp. the polytope $\operatorname{COR}_n^{\square}(\mathcal{I})$) is defined as the conic hull (resp. the convex hull) of all \mathcal{I} -correlation vectors $\pi^{\mathcal{I}}(S)$ for $S \subseteq V_n$.

Given an integer $1 \leq m \leq n$, let $\mathcal{I}_{\leq m}$ denote the collection of all subsets of V_n of cardinality less or equal to m. Hence, $\mathcal{I}_{\leq 2}$ consists of all singletons and pairs of elements of V_n and $\operatorname{COR}_n(\mathcal{I}_{\leq 2})$, $\operatorname{COR}_n^{\Box}(\mathcal{I}_{\leq 2})$ coincide with COR_n , $\operatorname{COR}_n^{\Box}$, respectively.

For $\mathcal{I} = 2^{V_n}$, which consists of all subsets of V_n , $\operatorname{COR}_n^{\square}(2^{V_n})$ is a simplex of dimension $2^n - 1$ and $\operatorname{COR}_n(2^{V_n})$ is a simplicial cone of dimension $2^n - 1$. This implies, in particular, that every correlation polytope $\operatorname{COR}_n^{\square}(\mathcal{I})$ arises as a projection of the simplex $\operatorname{COR}_n^{\square}(2^{V_n})$ (namely, on the subspace $\mathbb{R}^{2^{V_n}\setminus\mathcal{I}}$).

Proposition 3.13 remains valid for the case of arbitrary \mathcal{I} -correlations.

PROPOSITION 3.27 Let \mathcal{I} be a nonempty collection of subsets of $\{1, \ldots, n\}$ and let $p = (p_I)_{I \in \mathcal{I}} \in \mathbb{R}^{\mathcal{I}}$. The following assertions are equivalent. (i) $p \in COR_n(\mathcal{I})$ (resp. $p \in COR_n^{\square}(\mathcal{I})$).

(ii) There exist a measure space (resp. a probability space) $(\Omega, \mathcal{A}, \mu)$ and $A_1, \ldots, A_n \in \mathcal{A}$ such that $p_I = \mu(\cap_{i \in I} A_i)$ for all $I \in \mathcal{I}$.

We have the following general formulation of the Boole problem.

Suppose we are given the values of the joint probabilities $p_I = \mu(\cap_{i \in I} A_i)$, for all $I \in \mathcal{I}$. What is the best estimation of $\mu(A_1 \cup \ldots \cup A_n)$ in terms of the p_I 's?

The same reasoning as in the case of the usual pairwise correlations permits to show the following generalization of Proposition 3.17. PROPOSITION 3.28 Let $p_I = \mu(\bigcap_{I \in \mathcal{I}} A_i)$ for $I \in \mathcal{I}$. Then, $z_{min} \leq \mu(A_1 \cup \ldots \cup A_n) \leq z_{max}$, where $z_{min} = \min(\sum_{\emptyset \neq S \subseteq V_n} \lambda_S : p = \sum_{\emptyset \neq S \subseteq V_n} \lambda_S \pi^{\mathcal{I}}(S)$ with $\lambda_S \geq 0$ for all S) $= \max(w^T p : w^T z \leq 1$ is facet defining for $COR_n^{\Box}(\mathcal{I})$) and $z_{max} = \max(\sum_{\emptyset \neq S \subseteq V_n} \lambda_S : p = \sum_{\emptyset \neq S \subseteq V_n} \lambda_S \pi^{\mathcal{I}}(S)$ with $\lambda_S \geq 0$ for all S) $= \min(w^T p : w^T z \leq 1$ is facet defining for $Conv(\{\pi^{\mathcal{I}}(S) : \emptyset \neq S \subseteq V_n\}))$.

Boros and Prekopa [BP89] consider in detail the case $\mathcal{I} = \mathcal{I}_{\leq m}$; they describe a method for finding bounds on $\mu(A_1 \cup \ldots \cup A_n)$ in terms of the quantities $S_k = \sum_{1 \leq i_1 < \ldots < i_k \leq n} \mu(A_{i_1} \cap \ldots \cap A_{i_k})$ for $1 \leq k \leq n$. For instance, the following bounds hold

(3.29)
$$\begin{cases} \mu(A_1 \cup \ldots \cup A_n) \ge \sum_{1 \le i \le m} (-1)^{i-1} S_i & \text{for } m \text{ even} ,\\ \mu(A_1 \cup \ldots \cup A_n) \le \sum_{1 \le i \le m} (-1)^{i-1} S_i & \text{for } m \text{ odd} . \end{cases}$$

They were first discovered by Bonferroni [Bon36]. Several improvements of them have been proposed, most recently in [Gra93].

The inequality (3.26) can be easily generalized to the case of the polytope $\operatorname{COR}_n^{\square}(\mathcal{I}_{\leq 2m})$ for any $m \geq 1$. Given integers b_1, \ldots, b_n and $k_1, \ldots, k_m \geq 0$, the inequality

$$\prod_{1 \le l \le m} (\sum_{1 \le i \le n} b_i p_i - k_l) (\sum_{1 \le i \le n} b_i p_i - k_l - 1) \ge 0$$

is clearly valid for the polytope $\operatorname{COR}_n^{\square}(\mathcal{I}_{\leq 2m})$. Thus arises the question of determining the parameters $b_1, \ldots, b_n, k_1, \ldots, k_m$ for which it is facet defining. This problem is, however, already difficult for the case m = 1 of the correlation polytope $\operatorname{COR}_n^{\square}$.

4 Conditions for L₁-embeddability

We present in Section 4.1 some of the most important known necessary conditions for L_1 -embeddability, namely, the hypermetric and the negative type conditions. There are many other known necessary conditions for L_1 -embeddability, arising from known valid inequalities for the cut cone; they are described in Chapter ????(on facets). We focus here on the hypermetric and negative type conditions since they will be used repeatedly in this Chapter and throughout the book. In Section 4.3, we show the implications existing between the properties of being L_1 -, L_2 -embeddable, of negative type, or hypermetric, for a distance space. In Section 4.4, we present two operations, the direct sum and the tensor product, which preserve, respectively, L_1 -embeddable distance spaces and $\{0, 1\}$ -covariances.

4.1 Hypermetric and negative type conditions

4.1.1 Hypermetric and negative type inequalities

Let $n \geq 2, b_1, \ldots, b_n$ be integers such that $\epsilon := \sum_{1 \leq i \leq n} b_i \in \{0, 1\}$. We consider the inequality

(4.1)
$$\sum_{1 \le i < j \le n} b_i b_j d_{ij} \le 0.$$

If $\sum_{1 \leq i \leq n} b_i = 1$, then the inequality (4.1) is called a **hypermetric inequality** and is denoted by $Hyp_n(b_1, \ldots, b_n)$. If $\sum_{1 \leq i \leq n} b_i = 0$, then the inequality (4.1) is called an **inequality of negative type** and is denoted by $Neg_n(b_1, \ldots, b_n)$. The inequality (4.1) is said to be **pure** if $|b_i| = 0, 1$ for all $i \in V_n$. We can suppose that at least two of the b_i 's are nonzero (else, the inequality (4.1) is void). Hence, $\sum_{1 \leq i \leq n} |b_i| = 2k + \epsilon$ for some integer $k \geq 1$. The inequality (4.1) is then said to be $(2k + \epsilon)$ -gonal.

In particular, the 2-gonal inequality is the inequality of negative type $Neg_n(b_1, \ldots, b_n)$, where $b_i = 1$, $b_j = -1$ and $b_h = 0$ for $h \in V_n \setminus \{i, j\}$, for some distinct $i, j \in V_n$; it is nothing but the nonnegativity constraint

$$d_{ij} \geq 0.$$

The pure 3-gonal inequality is the hypermetric inequality $Hyp_n(b_1, \ldots, b_n)$, where $b_i = b_j = 1$, $b_k = -1$ and $b_h = 0$ for $h \in V_n \setminus \{i, j, k\}$, for some distinct $i, j, k \in V_n$; it is nothing but the triangle inequality (1.1).

For $\epsilon = 0, 1$, the pure $(2k + \epsilon)$ -gonal inequality reads

$$\sum_{1 \le r < s \le k+\epsilon} d_{i_r i_s} + \sum_{1 \le r < s \le k} d_{j_r j_s} - \sum_{\substack{1 \le r \le k+\epsilon \\ 1 < s < k}} d_{i_r j_s} \le 0,$$

where $i_1, \ldots, i_k, i_{k+\epsilon}, j_1, \ldots, j_k$ are distinct indices of V_n .

Figure 1 shows the pure 4-gonal and 5-gonal inequalities or, rather, their left hand sides. It should be understood as follows: a plain edge between two nodes i and j indicates a coefficient +1 for the variable d_{ij} and a dotted edge indicates a coefficient -1.

Figure 1

The negative type inequalities are classical inequalities in analysis; they were used, in particular, by Schoenberg [Sch37, Sch38a, Sch38b]. The hypermetric inequalities were introduced by Deza [Dez61, Dez62] and later, independently, by Kelly [Kel70].

4.1.2 Hypermetric and negative type distance spaces

We now turn to the definition of a hypermetric distance space, or of a negative type distance space. Basically, a distance space (X,d) is said to be hypermetric (resp. of negative type) if d satisfies all hypermetric inequalities (resp. all inequalities of negative type). More precisely, we have the following definitions.

Let (X, d) be a distance space. Then, (X, d) is said to be of **negative type** if, for all $n \ge 2, x_1, \ldots, x_n \in X, b_1, \ldots, b_n \in \mathbb{Z}$ with $\sum_{1 \le i \le n} b_i = 0$, the following inequality holds

(4.2)
$$\sum_{i,j=1}^{n} b_i b_j d(x_i, x_j) \le 0$$

(X,d) is said to be **hypermetric** if, for all $n \ge 2, x_1, \ldots, x_n \in X, b_1, \ldots, b_n \in \mathbb{Z}$ with $\sum_{1 \le i \le n} b_i = 1$, the inequality (4.2) holds. For $\epsilon = 0, 1$ and $k \in \mathbb{Z}, k \ge 1, (X,d)$ is said to be $(2k + \epsilon)$ -gonal if, for all $n \ge 2, x_1, \ldots, x_n \in X, b_1, \ldots, b_n \in \mathbb{Z}$ with $\sum_{1 \le i \le n} b_i = \epsilon$ and $\sum_{1 \le i \le n} |b_i| = 2k + \epsilon$, the inequality (4.2) holds.

Note that, in the above definitions, we do not require that the points x_1, \ldots, x_n be distinct. For instance, suppose that $x_1 = x_2$. Then, $d(x_1, x_2) = 0$ and $d(x_1, x_i) = d(x_2, x_i)$ for all *i* and, therefore, the inequality (4.2) reads

$$\sum_{2 \le i < j \le n} b'_i b'_j d(x_i, x_j) \le 0$$

after setting $b'_2 = b_1 + b_2, b'_3 = b_3, \ldots, b'_n = b_n$. In other words, we could have assumed in the above definitions that the inequality (4.2) is pure, i.e., that $|b_i| = 0, 1$ for all *i*. For

instance, the distance space (X, d) is $(2k + \epsilon)$ -gonal if and only if, for all (not necessarily distinct) $x_1, \ldots, x_k, x_{k+\epsilon}, y_1, \ldots, y_k \in X$, the following inequality holds

(4.3)
$$\sum_{1 \le i < j \le k+\epsilon} d(x_i, x_j) + \sum_{1 \le i < j \le k} d(y_i, y_j) - \sum_{\substack{1 \le i \le k+\epsilon \\ 1 \le j \le k}} d(x_i, y_j) \le 0.$$

In particular, (X, d) is 5-gonal if and only if, for all $x_1, x_2, x_3, y_1, y_2 \in X$, the following inequality holds

(4.4)
$$\sum_{1 \le i < j \le 3} d(x_i, x_j) + d(y_1, y_2) - \sum_{\substack{1 \le i \le 3\\ j=1,2}} d(x_i, y_j) \le 0.$$

Clearly, if (X, d) is of negative type, then the inequality (4.2) holds for all $b_1, \ldots, b_n \in \mathbb{R}$ with $\sum_{1 \le i \le n} b_i = 0$. Some more implications among the k-gonal conditions are summarized in the next result.

LEMMA 4.5 Let (X, d) be a distance space. (i) If (X, d) is (2k + 1)-gonal, then (X, d) is (2k + 2)-gonal, for any integer $k \ge 1$. (ii) If (X, d) is (k + 2)-gonal, then (X, d) is k-gonal, for any integer $k \ge 2$.

PROOF. (i) Let $x_1, \ldots, x_{k+1}, y_1, \ldots, y_{k+1}$ be 2k + 2 points of X. By assumption, (X, d) satisfies each of the k + 1 inequalities (4.3) obtained by considering all y_i 's except one. Similarly, (X, d) satisfies each of the k + 1 inequalities (4.3) obtained by considering all x_i 's except one (and exchanging the role of the x_i 's and y_i 's). If we sum up these 2k + 2 inequalities, we deduce that (X, d) satisfies the (2k + 2)-gonal inequality (4.3) relative to the points $x_1, \ldots, x_{k+1}, y_1, \ldots, y_{k+1}$.

(*ii*) We check that (X, d) is (2k - 1)-gonal whenever it is (2k + 1)-gonal; the other case is similar. Let $x_1, \ldots, x_k, y_1, \ldots, y_{k-1}$ be 2k - 1 points of X. Let $x \in X$ and set $x_{k+1} = y_k = x$. By assumption, (X, d) satisfies the (2k + 1)-gonal inequality (4.3) relative to the points $x_1, \ldots, x_{k+1}, y_1, \ldots, y_k$. But, the latter inequality, after some cancellations, is nothing but the (2k - 1)-gonal inequality (4.3) relative to the points $x_1, \ldots, x_k, y_1, \ldots, y_{k-1}$.

REMARK 4.6 The proof of Lemma 4.5 (i) shows, in fact, that the pure (2k + 2)-gonal inequality follows from the pure (2k + 1)-gonal inequalities. However, for $k \ge 2$ integer, the k-gonal inequalities do **not** follow from the (k + 2)-gonal inequalities (the proof of Lemma 4.5 (ii) works indeed at the level of distance spaces since we make the assumption that the two points x_{k+1} and y_k of X coincide).

Equality case in the hypermetric and negative type inequalities. The following question is considered in [Kel70, Ass84, Bal90]. What are the distance spaces, within a given class, that satisfy a given hypermetric or negative type inequality at equality? For instance, Kelly [Kel70] characterizes the finite subspaces of (\mathbb{R}, d_{ℓ_1}) that satisfy the

(2k+1)-gonal inequality at equality. Namely, given $x_1, \ldots, x_{k+1}, y_1, \ldots, y_k \in \mathbb{R}$, the equality

$$\sum_{1 \le i < j \le k+1} |x_i - x_j| + \sum_{1 \le i < j \le k} |y_i - y_j| - \sum_{1 \le i \le k+1 \ 1 \le j \le k} |x_i - y_j| = 0$$

holds if and only if y_1, \ldots, y_k separate x_1, \ldots, x_{k+1} , i.e., there exist a permutation α of $\{1, \ldots, k+1\}$ and a permutation β of $\{1, \ldots, k\}$ such that

$$x_{\alpha(1)} \leq y_{\beta(1)} \leq x_{\alpha(2)} \leq y_{\beta(2)} \leq \ldots \leq y_{\beta(k)} \leq x_{\alpha(k+1)}.$$

Generalizations and related results can be found in [Kel70, Ass84].

Ball (Lemma 4,[Bal90]) characterizes the reals x_1, \ldots, x_n for which the distance space $(\{x_1, \ldots, x_n\}, d_{\ell_1})$ satisfies the negative type inequality $Neg_n(-(n-4), 1, \ldots, 1, -2)$ at equality. This result is used for Proposition 7.4 (i) in Section 7.1, for deriving a lower bound on the minimum ℓ_1 -dimension of a distance space.

The hypermetric cone $\operatorname{HYP}(X)$ (resp. the negative type cone $\operatorname{NEG}(X)$) is defined as the set of all the distances d on X that are hypermetric (resp. of negative type). In the finite case, $X = V_n$, the cone $\operatorname{HYP}(V_n)$ is simply denoted by HYP_n and the cone $\operatorname{NEG}(V_n)$ by NEG_n ; both are assumed to be cones in \mathbb{R}^{E_n} . Hence, HYP_n (resp. NEG_n) is the cone in \mathbb{R}^{E_n} defined by the inequalities (4.1) for all integers b_1, \ldots, b_n with $\sum_{1 \le i \le n} b_i = 1$ (resp. $\sum_{1 \le i \le n} b_i = 0$).

4.1.3 Analogues of the hypermetric and negative type conditions for covariances

We now introduce the notion of a function of positive type, which will turn out to be closely related to that of a distance of negative type.

DEFINITION 4.7 A symmetric function $p: X \times X \longrightarrow \mathbb{R}$ is said to be of **positive type** on X if, for all $n \ge 2, x_1, \ldots, x_n \in X$, the matrix $(p(x_i, x_j))_{1 \le i,j \le n}$ is positive semidefinite, *i.e.*, the inequality $\sum_{1 \le i,j \le n} b_i b_j p(x_i, x_j) \ge 0$ holds for all $b_1, \ldots, b_n \in \mathbb{R}$ (or, equivalently, for all $b_1, \ldots, b_n \in \mathbb{Z}$).

The next Lemma 4.8 shows that the notions of functions of positive type and distances of negative type are, in fact, equivalent (via the covariance mapping). Then, Lemma 4.9 is an analogue of Lemma 4.8 for hypermetric inequalities. Both results will be used very often in the book and, in particular, in this Chapter and in Chapter ??? (on hypermetrics).

LEMMA 4.8 Let X be a set and $x_0 \in X$. Let d be a distance on X and $p = \xi_{x_0}(d)$ be the corresponding symmetric function on $X \setminus \{x_0\}$. The following assertions are equivalent. (i) (X,d) is of negative type, i.e., for all $n \geq 2, x_1, \ldots, x_n \in X, b_1, \ldots, b_n \in \mathbb{Z}$ with $\sum_{\substack{1 \leq i \leq n \\ (ii) \\ p \ is \ a \ function \ of \ positive \ type \ on \ X \setminus \{x_0\}, \ i.e., \ for \ all \ n \geq 2, \ x_1, \dots, x_n \in X \setminus \{x_0\},$

 $b_1,\ldots,b_n\in\mathbb{Z},\ \sum_{1\leq i,j\leq n}b_ib_jp(x_i,x_j)\geq 0.$

PROOF. The proof is based on the following observation. Given $x_1, \ldots, x_n \in X \setminus \{x_0\}$, $b_1, \ldots, b_n \in \mathbb{Z}$, we have that

$$\begin{split} \sum_{1 \le i,j \le n} b_i b_j p(x_i, x_j) &= \sum_{1 \le i,j \le n} \frac{1}{2} (d(x_i, x_0) + d(x_j, x_0) - d(x_i, x_j)) \\ &= \sum_{1 \le i \le n} b_i (\sum_{1 \le j \le n} b_j) d(x_0, x_i) - \sum_{1 \le i,j \le n} b_i b_j d(x_i, x_j) \\ &= - \left(\sum_{0 \le i,j \le n} b_i b_j d(x_i, x_j) \right), \end{split}$$

after setting $b_0 = -\sum_{1 \le j \le n} o_j$.

LEMMA 4.9 Let X be a set and $x_0 \in X$. Let d be a distance on X and $p = \xi_{x_0}(d)$ be the corresponding symmetric function on $X \setminus \{x_0\}$. The following assertions are equivalent. (i) (X, d) is hypermetric, i.e., for all $n \ge 2, x_1, \ldots, x_n \in X$, $b_1, \ldots, b_n \in \mathbb{Z}$ with $\sum_{1 \le i \le n} b_i = 0$ 1, $\sum_{1 \leq i,j \leq n} b_i b_j d(x_i, x_j) \leq 0$ holds. (ii) For all $n \geq 2$, $x_1, \ldots, x_n \in X \setminus \{x_0\}, b_1, \ldots, b_n \in \mathbb{Z}$, p satisfies the inequality $\sum_{1 < i,j < n} b_i b_j p(x_i, x_j) - \sum_{1 < i < n} b_i p(x_i, x_i) \ge 0.$

PROOF. The proof is based on the following observation. Given $x_1, \ldots, x_n \in X \setminus \{x_0\}$, $b_1,\ldots,b_n\in\mathbb{Z}$, then

$$\sum_{\substack{1 \le i,j \le n \ b i \ b j \ p}(x_i, x_j) - \sum_{\substack{1 \le i \le n \ b i \ p}(x_i, x_i) \\ = \sum_{\substack{1 \le i \le n \ b i} (\sum_{\substack{1 \le j \le n \ b j \ -1}) d(x_0, x_i) - \sum_{\substack{1 \le i,j \le n \ b i \ b j \ d}(x_i, x_j)) \\ = -(\sum_{\substack{0 \le i,j \le n \ b i \ b j \ d}(x_i, x_j)),$$

setting $b_0 = 1 - \sum_{\substack{1 \le i \le n \ b j}} b_j.$

after s $\log v_0 = 1 \angle 1 \leq j \leq n$

4.2Characterization of L_2 -embeddability

We present in this Section several equivalent characterizations for L_2 -embeddable distance spaces.

We start with some preliminary results. Given a set of vectors $v_1, \ldots, v_n \in \mathbb{R}^k$ $(k \ge 1)$, their **Gram matrix** is the $n \times n$ matrix whose (i, j)-th entry is $v_i^T v_j$; its rank is equal to the rank of the system (v_1, \ldots, v_n) . The next Lemma 4.10, which states the connection existing between Gram matrices and positive semidefinite matrices, is well known; we give the proof for completeness.

LEMMA 4.10 Let $A = (a_{ij})_{1 \leq i,j \leq n}$ be a symmetric matrix which is positive semidefinite and let $k \leq n$ be its rank. Then, A is a Gram matrix, i.e., there exist vectors $v_1, \ldots, v_n \in \mathbb{R}^k$ such that $a_{ij} = v_i^T v_j$ for $1 \leq i,j \leq n$. Moreover, if v'_1, \ldots, v'_n are other vectors of \mathbb{R}^k such that $a_{ij} = v_i'^T v'_j$ for $1 \leq i,j \leq n$, then $v'_i = T(v_i)$, $1 \leq i \leq n$, for some orthogonal transformation T of \mathbb{R}^k . The system (v_1, \ldots, v_n) has rank k.

PROOF. By assumption, A has k non zero eigenvalues which are positive. Hence, there exists an $n \times n$ matrix Q_0 such that $A = Q_0 D Q_0^T$, where D is an $n \times n$ matrix whose entries are all zero except k diagonal entries, say with indices $(1, 1), \ldots, (k, k)$, equal to 1. Denote by Q the $n \times k$ submatrix of Q_0 consisting of its first k columns. Then, $A = QQ^T$ holds, i.e., $a_{ij} = v_i^T v_j$ for $1 \le i, j \le n$, where v_1, \ldots, v_n denote the rows of Q. It is easy to see that (v_1, \ldots, v_n) has the same rank k as A.

Let Q' be another $n \times k$ matrix such that $A = Q'Q'^T$. Both matrices Q, Q' have rank k, hence there exists a $k \times k$ non singular matrix B such that Q' = QB. Let Q_1 be a non singular $k \times k$ submatrix of Q formed, say, by its first k rows, and let Q'_1 denote the $k \times k$ submatrix of Q' formed by its first k rows. Then, $Q'_1 = Q_1B$. From the equality $Q_1Q_1^T = Q'_1(Q'_1)^T$, we obtain that BB^T is the identity matrix, i.e., B is an orthogonal transformation of \mathbb{R}^k .

Let M be a symmetric $n \times n$ matrix. The **inertia** In(M) of M is defined as the triple (p,q,s), where p (resp. q,s) denotes the number of positive (resp. negative, zero) eigenvalues of M; hence, n = p + q + s. If P is a nonsingular matrix, then it is well known that the two matrices M and PMP^T have the same inertia (this result is known as **Sylvester's law of inertia**).

LEMMA 4.11 Let
$$M$$
 be a symmetric matrix with the following block decomposition

$$M = \begin{pmatrix} A & B \\ \hline B^T & C \end{pmatrix}, \text{ where } C \text{ is nonsingular. Then, } In(M) = In(C) + In(A - BC^{-1}B^T).$$

(The matrix $A - BC^{-1}B^T$ is also known as the Schur complement of C in M.)

PROOF. One verifies easily the following identity:

$$\begin{pmatrix} I & BC^{-1} \\ 0^T & I \end{pmatrix} \begin{pmatrix} A - BC^{-1}B^T & 0 \\ 0^T & C \end{pmatrix} \begin{pmatrix} I & 0 \\ C^{-1}B^T & I \end{pmatrix} = \begin{pmatrix} A & B \\ B^T & C \end{pmatrix} = M.$$
By Sylvester's law of inertia, we obtain that the matrices M and $\begin{pmatrix} A - BC^{-1}B^T & 0 \\ 0^T & C \end{pmatrix}$

have the same inertia. Therefore, $In(M) = In(C) + In(A - BC^{-1}B^{T})$.

LEMMA 4.12 Let M be a symmetric $n \times n$ matrix and let U be a subspace of \mathbb{R}^n such that $x^T M x \leq 0$ holds for all $x \in U$. If U has dimension n-1, then M has at most one positive eigenvalue.

PROOF. Suppose, for contradiction, that M has two positive eigenvalues λ_1 and λ_2 . Let u_1 and u_2 be eigenvectors for λ_1 and λ_2 , respectively, with $u_1^T u_2 = 0$ and $|| u_1 ||_2 = || u_2 ||_2 = 1$. Let V denote the subspace of \mathbb{R}^n spanned by u_1 and u_2 . Then, $x^T M x > 0$ holds for all $x \in V, x \neq 0$; indeed, if $x = a_1 u_1 + a_2 u_2$, then $x^T M x = a_1^2 \lambda_1 + a_2^2 \lambda_2 > 0$ if $(a_1, a_2) \neq (0, 0)$. As U and V have respective dimensions n - 1 and 2, there exists $x \in U \cap V$ with $x \neq 0$. Then, $x^T M x \leq 0$ since $x \in U$ and $x^T M x > 0$ since $x \in V$, yielding a contradiction.

Let (X, d) be a distance space with $X = \{1, \ldots, n\}$. Let D denote the corresponding distance matrix; it is the $n \times n$ symmetric matrix whose (i, j)-th entry is d(i, j) (with diagonal terms all equal to 0). We also consider the symmetric $(n + 1) \times (n + 1)$ matrix M(X, d) defined by

(4.13)
$$M(X,d) = \begin{pmatrix} D & -e \\ \hline \\ -e^T & 0 \end{pmatrix}$$

where $e = (1, ..., 1) \in \mathbb{R}^n$. (The bordered matrix M(X, d) is closely related to the Cayley-Menger matrix of the distance space (X, d); indeed, the latter is defined in the same way but it has the borders e, e^T instead of $-e, -e^T$.)

Let $x_0 \in X$ and let $p = \xi_{x_0}(d)$ denote the image of d under the covariance mapping ξ_{x_0} (recall relation (3.8)). Let P(X, d) denote the $(n-1) \times (n-1)$ matrix whose (i, j)-th entry is p(i, j) for $i, j \in X \setminus \{x_0\}$. The next Lemma 4.14 establishes a relation between the ranks of the matrices M(X, d), P(X, d), and $(I - \frac{1}{n}J)D(I - \frac{1}{n}J)$, where I denotes the identity matrix and J the all ones matrix.

LEMMA 4.14 $rank(P(X,d)) = rank(M(X,d)) - 2 = rank((I - \frac{1}{n}J)D(I - \frac{1}{n}J)).$

PROOF. The equalities $\operatorname{rank}(M(X,d)) = \operatorname{rank}(P(X,d)) + 2$ and $\operatorname{rank}((I - \frac{1}{n}J)D(I - \frac{1}{n}J)) = \operatorname{rank}(P(X,d))$ can be checked by doing some row/column manipulations on M(X,d) and $(I - \frac{1}{n}J)D(I - \frac{1}{n}J)$, and applying relation (3.8).

We now present a classical result, due to Schoenberg [Sch35, Sch38b], on the characterization of L_2 -embeddable distance spaces. Theorem 4.16 gives a characterization of the distance spaces that are isometrically L_2 -embeddable in terms of the negative type inequalities and Theorem 4.15 is an equivalent formulation in the context of covariances.

THEOREM 4.15 Let p be a symmetric function on X. Then, p is of positive type on X if and only if p is an \mathbb{R} -covariance on X.

PROOF. Suppose first that p is an \mathbb{R} -covariance on X. Then, $p(x, y) = \int_{\Omega} f_x(\omega) f_y(\omega) \mu(d\omega)$ for all $x, y \in X$, where f_x are real valued functions of $L_2(\Omega, \mathcal{A}, \mu)$. Let $b \in \mathbb{Z}^X$. Then, $\sum_{x,y \in X} b_x b_y p(x,y) = \int_{\Omega} |f_x(\omega)|^2 \mu(d\omega) \ge 0$. This shows that p is of positive type on X. Conversely, suppose that p is of positive type on X. We show that p is an \mathbb{R} -covariance on X. From the finitude result from Theorem 1.4 (for p = 2) and Lemma 3.10, we can suppose that X is finite. By assumption, the matrix $(p(i,j))_{i,j\in X}$ is positive semidefinite and, thus, is a Gram matrix, by Lemma 4.10. This shows that p is an \mathbb{R} -covariance on X.

THEOREM 4.16 Let (X,d) be a distance space. Then, (X,d) is of negative type if and only if (X,\sqrt{d}) is L_2 -embeddable.

PROOF. It follows from Theorem 4.15, after applying Lemmas 3.10 and 4.8.

When the distance space (X, \sqrt{d}) is ℓ_2 -embeddable with X finite, the associated distance matrix D is also known as a **Euclidian distance matrix** in the literature (see, e.g., [Gow85, HWLT91]).

PROPOSITION 4.17 Let (X, d) be a finite distance space of negative type. Then, (X, \sqrt{d}) is ℓ_2 -embeddable and its minimum ℓ_2 -dimension $m_2(X, \sqrt{d})$ satisfies: $m_2(X, \sqrt{d}) = \operatorname{rank}(P(X, d)) = \operatorname{rank}(M(X, d)) - 2 = \operatorname{rank}((I - \frac{1}{n}J)D(I - \frac{1}{n}J)).$

PROOF. In view of Lemma 4.14, we have only to check that $m_2(X, \sqrt{d}) = \operatorname{rank}(P(X, d))$. Set $k = m_2(X, \sqrt{d})$ and $r = \operatorname{rank}(P(X, d))$. By Lemma 4.10, P(X, d) is the Gram matrix of a system of vectors $v_i \in \mathbb{R}^r$ for $i \in X \setminus \{x_0\}$. Then, $v_i, i \in X$, provide an ℓ_2 -embedding of (X, \sqrt{d}) , if we set $v_{x_0} = 0$. This implies that $r \ge k$. On the other hand, there exist vectors $u_i \in \mathbb{R}^k$, $i \in X$, such that $d(i, j) = || u_i - u_j ||_2^2$ for all $i, j \in X$. We can suppose without loss of generality that $u_{x_0} = 0$. Then, P(X, d) coincides with the Gram matrix of $u_i, i \in X \setminus \{x_0\}$, which implies that $r \le k$. Hence, r = k holds. Therefore, the parameter $m_2(n)$, which we recall is defined as the minimum ℓ_2 -dimension of an ℓ_2 -embeddable distance space on n points, satisfies

$$(4.18) m_2(n) = n - 1.$$

We now present two additional equivalent characterizations for distance spaces of negative type.

THEOREM 4.19 Let (X, d) be a finite distance space with $X = \{1, ..., n\}$. Let D be the associated $n \times n$ distance matrix and let M(X, d) be the bordered matrix defined by (4.13). Consider the following assertions.

(i) (X, d) is of negative type.

(ii) The matrix $(I - \frac{1}{n}J)(-D)(I - \frac{1}{n}J)$ is positive semidefinite.

(iii) The matrix M(X, d) has exactly one positive eigenvalue.

(iv) The matrix D has exactly one positive eigenvalue.

Then, $(i) \iff (ii)$ [Gow85], $(i) \iff (iii)$ [HW88], and $(i) \implies (iv)$. Moreover, if D has a constant row sum, then $(i) \iff (iv)$ [KS93].

PROOF. (i) \iff (ii) Set $K = I - \frac{1}{n}J$ and A = K(-D)K. Then, for $x \in \mathbb{R}^n$, we have that $x^T A x = y^T(-D)y$, setting y = Kx. One checks easily that the range of K consists of the vectors $y \in \mathbb{R}^n$ such that $\sum_{1 \le i \le n} y_i = 0$. Therefore, we obtain that A is positive semidefinite if and only if $y^T(-D)y \ge 0$ for all $y \in \mathbb{R}^n$ such that $\sum_{1 \le i \le n} y_i = 0$, i.e., (X, d) is of negative type.

(i) \iff (iii) Let Q be an orthogonal $n \times n$ matrix such that Qe is equal to the vector $\begin{pmatrix} Q & 0 \end{pmatrix}$

$$e' = (0, \dots, 0, 1). \text{ Set } D' = QDQ^T, Q' = \left(\begin{array}{c|c} & & \\ \hline & & \\ \hline & & \\ 0^T & 1 \end{array}\right), \text{ and } M' = Q'M(X, d)Q'^T. \text{ Hence,}$$
$$M' = \left(\begin{array}{c|c} D' & -e' \\ \hline & -e'^T & 0 \end{array}\right) = \left(\begin{array}{c|c} D'_0 & b & 0 \\ \hline & b^T & \beta & -1 \\ \hline & 0 & -1 & 0 \end{array}\right), \text{ if we let } D' \text{ be pictured as } \left(\begin{array}{c|c} D'_0 & b \\ \hline & b^T & \beta \end{array}\right),$$

where D'_0 is an $(n-1) \times (n-1)$ matrix and $b \in \mathbb{R}^{n-1}$. As the matrix $\left(\begin{array}{c|c} \beta & -1 \\ \hline -1 & 0 \end{array}\right)$ is nonsingular, we can apply Lemma 4.11 for computing the inertia of M'. We obtain that $\operatorname{In}(M') = \operatorname{In}(D'_0) + \operatorname{In}\left(\begin{array}{c|c} \beta & -1 \\ \hline -1 & 0 \end{array}\right)$. By Sylvester's law of inertia, M(X,d) and M'

have the same inertia and, in particular, both M(X,d) and M' have the same number of positive eigenvalues. One checks easily that the matrix $\begin{pmatrix} \beta & | -1 \\ -1 & 0 \end{pmatrix}$ has exactly one positive eigenvalue. Therefore, M(X,d) has one positive eigenvalue if and only if D'_0 has no positive eigenvalue, i.e., $x^T D'_0 x \leq 0$ for all $x \in \mathbb{R}^{n-1}$. But, $x^T D'_0 x \leq 0$ holds for all $x \in \mathbb{R}^{n-1}$ if and only if $y^T D' y \leq 0$ holds for all $y \in \mathbb{R}^n$ such that $e'^T y = 0$ (because $y^T D' y = x^T D'_0 x + x_n (2b^T x + \beta x_n)$, if $y = (x, x_n)$) or, equivalently, $z^T D z \leq 0$ holds for all $z \in \mathbb{R}^n$ such that $e^T z = 0$, i.e., (X, d) is of negative type.

 $(i) \implies (iv)$ The matrix D has at least one positive eigenvalue since D has its diagonal terms equal to 0. If (X, d) is of negative type then, by Lemma 4.12, D has at most one positive eigenvalue since $x^T D x \leq 0$ holds for all x in an (n-1)-dimensional subspace of \mathbb{R}^n . Therefore, D has one positive eigenvalue.

Finally, suppose that D has a constant row sum, equal to s, and that (iv) holds. Then, s is the positive eigenvalue of D since De = se holds. This implies that the matrix $\frac{s}{n}J - D$ is positive semidefinite. Hence, by Lemma 4.10, there exists an $n \times k$ matrix X such that $\frac{s}{n}J - D = XX^T$. Let v_1, \ldots, v_n denote the row vectors of X. Then, we have that $\frac{s}{n} - d(i,j) = v_i^T v_j$ for all i,j or, equivalently, $d(i,j) = ||v_i - v_j||_2^2$ for all i,j. This shows that (X, \sqrt{d}) is ℓ_2 -embeddable, i.e., by Theorem 4.16, that (X, d) is of negative type.

4.3 A chain of implications

We summarize in this Section the implications existing between the properties of being L_1 -, L_2 -embeddable, of negative type, and hypermetric.

THEOREM 4.20 Let (X, d) be a distance space. Consider the following assertions. (i) (X, d) is L_2 -embeddable. (ii) (X, d) is L_1 -embeddable. (iii) (X, d) is hypermetric. (iv) (X, d) is of negative type. (v) (X, \sqrt{d}) is L_2 -embeddable. We have the chain of implications (i) \Longrightarrow (ii) \Longrightarrow (iv) \Longrightarrow (iv) \iff (v).

PROOF. $(i) \implies (ii)$ is a classical result in analysis. For $(ii) \implies (iii)$, it suffices to check that every finite subspace of (X, d) is hypermetric, i.e., that every member of the cut cone satisfies the hypermetric inequalities or, equivalently, that each cut semimetric satisfies the hypermetric inequalities. Indeed, given a subset S of V_n and $b_1, \ldots, b_n \in \mathbb{Z}$ with $\sum_{1 \le i \le n} b_i = 1$, we have that

with $\sum_{1 \leq i \leq n} b_i = 1$, we have that $\sum_{1 \leq i < j \leq n} \overline{b}_i b_j \delta(S)_{ij} = (\sum_{i \in S} b_i) (\sum_{i \in V_n \setminus S} b_i) = \sum_{i \in S} b_i (1 - \sum_{i \in S} b_i) \leq 0$

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since all b_i 's are integers. For $(iii) \implies (iv)$, we use the equivalent formulations of the negative type and hypermetric conditions given in Lemmas 4.8 and 4.9. Finally, $(iv) \iff (v)$ holds by Theorem 4.16.

EXAMPLE 4.21 Let $d(K_{2,3})$ denote the path metric of the complete bipartite graph $K_{2,3}$ with node set $\{x_1, x_2, x_3\} \cup \{y_1, y_2\}$. Then, $d(K_{2,3})(x_i, x_j) = d(K_{2,3})(y_1, y_2) = 1$ and $d(K_{2,3})(x_i, y_j) = 2$. Hence, $d(K_{2,3})$ violates the pentagonal inequality (4.4). Therefore, $d(K_{2,3})$ is not hypermetric and, thus, not L_1 -embeddable.

From the implication $(ii) \Longrightarrow (iii)$ from Theorem 4.20, we have the inclusion $C_1(X) \subseteq$ HYP(X) (recall that $C_1(X)$ is the cone of all L_1 -embeddable distances on X). This inclusion is, in general, strict. It is strict, in particular, if $7 \le |X| < \infty$ or $X = \mathbb{N}$.

For showing the strict inclusion $\text{CUT}_n \subset \text{HYP}_n$ for $n \ge 7$, it suffices to exhibit an inequality which defines a facet for CUT_n and is not hypermetric. Many such inequalities are described in Chapter ??? (on facets).

For showing the strict inclusion $C_1(\mathbb{N}) \subset \operatorname{HYP}(\mathbb{N})$, consider the distance d on \mathbb{N} obtained by taking the spherical *t*-extension of the path metric of the Schläfli graph G_{27} , i.e., d_{ij} is the shortest length of a path joining i and j in G_{27} if i and j are both nodes of G_{27} and $d_{ij} = t$ otherwise. For $t \geq \frac{4}{3}$, d is hypermetric but d is not L_1 -embeddable (since the path metric of G_{27} lies on an extreme ray of the hypermetric cone on 27 points) ([Gri92]).

However, there are many examples of classes of distance spaces (X, d) for which the properties of being hypermetric and L_1 -embeddable are equivalent. We present such examples with X infinite in Sections 5.1 and 5.2.

We summarize in Remark 4.22 below a list of distance spaces (X, d) for which L_1 embeddability can be characterized by a set \mathcal{I} of inequalities that are all hypermetric or of negative type.

REMARK 4.22 (i) (V_n, d) with $n \leq 6$; \mathcal{I} consists of the hypermetric inequalities, i.e., $CUT_n = HYP_n$ for $n \leq 6$ ([Dez61] for $n \leq 5$ and [AM89] for n = 6).

(*ii*) A normed space $(\mathbb{R}^m, d_{\parallel,\parallel})$; \mathcal{I} consists of the negative type inequalities, a normed space $(\mathbb{R}^m, d_{\parallel,\parallel})$ whose unit ball is a polytope; \mathcal{I} consists of the 7-gonal inequalities (see Theorems 5.1 and 5.2).

(*iii*) (L, d_v) where (L, \preceq) is a lattice with distance $d_v(x, y) = v(v \lor y) - v(x \land y)$ for $x, y \in L; \mathcal{I}$ consists of the 5-gonal inequalities or, equivalently, \mathcal{I} consists of the negative type inequalities (see Theorem 5.6 and Example 5.9).

 $(iv)(\mathcal{A},d)$ where \mathcal{A} is a family of subsets of a set Ω which is stable under the symmetrical difference and $d(A,B) = v(A \triangle B)$ for $A, B \in \mathcal{A}$ with v nonnegative and $v(\emptyset) = 0$; \mathcal{I} consists of the inequalities of negative type (see Example 5.10).

(v) The graphic space (V, d(G)) where G is a connected bipartite graph with node set V; \mathcal{I} consists of the 5-gonal inequalities (see Chapter ????(on graphs)).

(vi) The graphic space (V, d(G)) where G is a connected graph on at least 38 nodes and having a node adjacent to all other nodes; \mathcal{I} consists of the negative type inequalities and the 5-gonal inequalities (see Chapter ????(on graphs)).

4.4 The direct sum and tensor product operations

We present two operations, the direct sum and the tensor product, which preserve, respectively, L_1 -embeddability and $\{0, 1\}$ -covariances.

DEFINITION 4.23 (i) Let (X_1, d_1) and (X_2, d_2) be two distance spaces. Their **direct sum** is the distance space $(X_1 \times X_2, d_1 \oplus d_2)$, where

$$d_1 \oplus d_2((x_1, x_2), (y_1, y_2)) = d_1(x_1, y_1) + d_2(x_2, y_2)$$
 for all $x_1, y_1 \in X_1, x_2, y_2 \in X_2$.

(ii) Let $p_1 : X_1 \times X_1 \longrightarrow \mathbb{R}$ and $p_2 : X_2 \times X_2 \longrightarrow \mathbb{R}$ be two symmetric functions. Their **tensor product** is the symmetric function $p_1 \otimes p_2 : (X_1 \times X_2) \times (X_1 \times X_2) \longrightarrow \mathbb{R}$ defined by

$$p_1 \otimes p_2((x_1, x_2), (y_1, y_2)) = p_1(x_1, y_1)p_2(x_2, y_2)$$
 for all $x_1, y_1 \in X_1, x_2, y_2 \in X_2$.

PROPOSITION 4.24 [Ass79, Ass80b] (i) If (X_1, d_1) and (X_2, d_2) are L_1 -embeddable, then their direct sum $(X_1 \times X_2, d_1 \oplus d_2)$ is L_1 -embeddable.

(i) If p_1 is a $\{0, 1\}$ -covariance on X_1 and p_2 is a $\{0, 1\}$ -covariance on X_2 , then their tensor product $p_1 \otimes p_2$ is a $\{0, 1\}$ -covariance on $X_1 \times X_2$.

PROOF. (i) By assumption, there exist a measure space $(\Omega_i, \mathcal{A}_i, \mu_i)$ and an isometric embedding ϕ_i of (X_i, d_i) into $L_1(\Omega_i, \mathcal{A}_i, \mu_i)$, for i = 1, 2. Let $(\Omega = \Omega_1 \cup \Omega_2, \mathcal{A}, \mu)$ denote the measure space obtained by extending \mathcal{A}_i and μ_i to $\Omega_1 \cup \Omega_2$. We obtain an isometric embedding of $(X_1 \times X_2, d_1 \oplus d_2)$ into $(\Omega, \mathcal{A}, \mu)$ by setting $\phi(x_1, x_2)(\omega) = \phi_i(x_i)(\omega)$ if $\omega \in \Omega_i$, for i = 1, 2. Indeed,

$$\begin{aligned} d_1 \oplus d_2((x_1, x_2), (y_1, y_2)) &= d_1(x_1, y_1) + d_2(x_2, y_2) \\ &= \parallel \phi_1(x_1) - \phi_1(y_1) \parallel + \parallel \phi_2(x_2) - \phi_2(y_2) \parallel \\ &= \parallel \phi(x_1, x_2) - \phi(y_1, y_2) \parallel . \end{aligned}$$

(*ii*) By assumption, there exist a measure space $(\Omega_i, \mathcal{A}_i, \mu_i)$ and a mapping $x \in X_i \mapsto A_x^{(i)} \in (\mathcal{A}_i)_{\mu_i}$ such that $p_i(x, y) = \mu_i(A_x^{(i)} \cap A_y^{(i)})$ for all $x, y \in X_i$, for i = 1, 2. Set

$$\begin{split} \Omega &= \Omega_1 \times \Omega_2, \ \mathcal{A} = \{A_1 \times A_2 : A_1 \in \mathcal{A}_1, A_2 \in \mathcal{A}_2\}, \text{ and take } \mu = \mu_1 \otimes \mu_2. \text{ Then, we have } \\ \text{that } p_1 \otimes p_2((x_1, x_2), (y_1, y_2)) = p_1(x_1, y_1) p_2(x_2, y_2) = \mu_1(A_{x_1}^{(1)} \cap A_{y_1}^{(1)}) \mu_2(A_{x_2}^{(2)} \cap A_{y_2}^{(2)}) \\ &= \mu_1 \otimes \mu_2((A_{x_1}^{(1)} \times A_{x_2}^{(2)}) \cap (A_{y_1}^{(1)} \times A_{y_2}^{(2)})). \text{ This shows that } p_1 \otimes p_2 \text{ is a } \{0, 1\}\text{-covariance on } \\ X_1 \times X_2. \end{split}$$

COROLLARY 4.25 (i) If (X, d_1) and (X, d_2) are L_1 -embeddable, then $(X, d_1 + d_2)$ is L_1 -embeddable.

(ii) If p_1 and p_2 are $\{0,1\}$ -covariances on X, then p_1p_2 is a $\{0,1\}$ -covariance on X.

PROOF. (i) follows from Theorems 1.4 and 2.11. In a more elementary way, (i) follows from Proposition 4.24 (i), since $(X, d_1 + d_2)$ is a subspace from $(X_1 \times X_2, d_1 \oplus d_2)$ (via the embedding $x \mapsto (x, x)$). (ii) follows from Proposition 4.24 (ii), since p_1p_2 identifies with the restriction of $p_1 \otimes p_2$ to the diagonal subset $\{(x, x) : x \in X\}$ of $X \times X$.

5 Two cases of complete characterization of L₁-embeddability

We present in this Section two classes of distance spaces for which L_1 -embeddability can be fully characterized using only hypermetric or negative type inequalities. The first class consists of metric spaces arising from normed spaces and the second one consists of metric spaces arising from metric lattices equipped with a valuation.

5.1 L₁-metrics from normed spaces

Let $(E, \|\cdot\|)$ be a normed space. We consider the associated metric space $(E, d_{\|\cdot\|})$, where $d_{\|\cdot\|}$ is the norm metric defined by

$$d_{\parallel,\parallel}(x,y) = \parallel x - y \parallel$$

for all $x, y \in E$.

In this Section, we give a characterization of the norms on $E = \mathbb{R}^m$ for which the metric space $(\mathbb{R}^m, d_{\parallel,\parallel})$ is L_1 -embeddable.

We first recall some definitions.

Let K be a **convex body** in \mathbb{R}^m , i.e., K is a nonempty convex compact subset of \mathbb{R}^m . We suppose that the origin 0 belongs to the interior of K. K is said to be **centrally symmetric** if $-x \in K$ for all $x \in K$. The **polar** K^* of a convex body K is the convex body defined by

$$K^* = \{ x \in \mathbb{R}^m : x^T y \leq 1 \text{ for all } y \in K \}.$$

The notions of convex bodies and of norms are, in fact, equivalent, in the following sense. First, if $\| \cdot \|$ is a norm on \mathbb{R}^m then its **unit ball**

$$B = \{ x \in \mathbb{R}^m : \parallel x \parallel \le 1 \}$$

is clearly a centrally symmetric convex body. Conversely, let K be a centrally symmetric convex body (containing the origin in its interior). The **support function** $h(K, .) : \mathbb{R}^m \to \mathbb{R}$ of K is defined by

$$h(K, x) = \max(x^T y : y \in K)$$

for $x \in \mathbb{R}^m$. One can easily check that h(K, .) defines a norm on \mathbb{R}^m , whose unit ball is the polar K^* of K. This norm can be alternatively defined by

$$h(K, x) = \min(\lambda > 0 : \frac{x}{\lambda} \in K^*)$$

for all $x \in \mathbb{R}^m$.

A convex polytope is called a **zonotope** if it is the vector sum of some line segments. A convex body which can be approximated by zonotopes with respect to the Blaschke-Hausdorff metric is called a **zonoid**. Zonotopes and zonoids are central objects in convex geometry and they are also relevant to many other fields (see, e.g., [Bol69, SW83] for a survey). They are, in particular, relevant to the topic of L_1 -metrics as we explain below.

We now present several equivalent characterizations for L_1 -embeddability of a normed metric space $(\mathbb{R}^m, d_{\parallel,\parallel})$.

THEOREM 5.1 (see [Bol69, SW83]) Let $\| . \|$ be a norm on \mathbb{R}^m and let B be its unit ball. The following assertions are equivalent.

(i) $d_{\parallel \cdot \parallel}$ is of negative type.

(ii) $d_{\parallel,\parallel}$ is hypermetric.

(*iii*) $(\mathbb{R}^m, d_{\parallel,\parallel})$ is L_1 -embeddable.

(iv) The polar of B is a zonoid.

Precise reference for the equivalence $(i) \iff (ii) \iff (iv)$ can be found in [SW83] and $(iii) \iff (iv)$ is proved in [Bol69]. L_1 -embeddability of norm metrics can be characterized by much simpler inequalities when the unit ball of the normed space is a polytope.

THEOREM 5.2 ([Ass80a, Ass84, Wit78], see [SW83]). Let $\parallel . \parallel$ be a norm on \mathbb{R}^m for which the unit ball B is a polytope. The following assertions are equivalent. $(i) \parallel . \parallel$ satisfies Hlawka's inequality

$$||x|| + ||y|| + ||z|| + ||x + y + z|| \ge ||x + y|| + ||x + z|| + ||y + z||$$

for all $x, y, z \in \mathbb{R}^m$. (ii) $\| . \|$ satisfies the 7-gonal inequality

$$\sum_{1 \le i < j \le 4} \| x_i - x_j \| + \sum_{1 \le h < k \le 3} \| y_h - y_k \| \le \sum_{\substack{1 \le i \le 4 \\ 1 \le k \le 3}} \| x_i - y_k \|$$

for all $x_1, x_2, x_3, x_4, y_1, y_2, y_3 \in \mathbb{R}^m$. (iii) The polar of B is a zonotope. (iv) $(\mathbb{R}^m, d_{\parallel,\parallel})$ is L_1 -embeddable.

Actually, the implication $(ii) \Longrightarrow (i)$ of Theorem 5.2 remains valid for general norms. Namely, if an arbitrary norm on \mathbb{R}^m satisfies the 7-gonal inequality, then it also satisfies Hlawka's inequality ([Ass84]).

The above results can be partially extended to the more general concept of projective metrics. A continuous metric d on \mathbb{R}^m is called a **projective metric** if it satisfies d(x,z) =d(x, y) + d(y, z) for any collinear points x, y, z lying in that order on a common line. Clearly, every norm metric is projective. The cone of projective metrics is the object considered by the fourth Hilbert problem in \mathbb{R}^m (see [Ale88], [Amb82]).

We have the following characterization of L_1 -embeddability for projective metrics.

THEOREM 5.3 [Ale88] Let d be a projective metric on \mathbb{R}^m . The following assertions are equivalent.

(i) (\mathbb{R}^m, d) is L_1 -embeddable. (ii) d is hypermetric. (iii) There exists a positive Borel measure μ on the hyperplanesets of \mathbb{R}^m satisfying $\left\{ \begin{array}{ll} \mu([[x]]) = 0 & \text{for all } x \in \mathbb{R}^m \\ 0 < \mu([[x,y]]) < \infty & \text{for all } x \neq y \in \mathbb{R}^m \end{array} \right.$

and such that d is defined by the following formula (called Crofton formula):

$$d(x, y) = \mu([[x, y]]) \text{ for } x, y \in \mathbb{R}^m,$$

where [[x, y]] denotes the set of hyperplanes meeting the segment [x, y].

In dimension m = 2, Theorem 5.3 (ii) always holds (see [Ale88]), i.e., every projective metric on \mathbb{R}^2 is L_1 -embeddable. On the other hand, the norm metric $d_{\ell_{\infty}}$ arising from the norm $||x||_{\infty} = \max(|x_1|, |x_2|, |x_3|)$ in \mathbb{R}^3 is not L_1 -embeddable since it is not hypermetric. Indeed, the points $x_1 = (1, 1, 0), x_2 = (1, -1, 0), x_3 = (-1, 1, 0), y_1 = (0, 0, 0)$ and $y_2 = (0, 0, 1)$ violate the 5-gonal inequality (4.4) ([Kel70]).

5.2 L₁-metrics from lattices

In this Section, we consider a class of metric spaces arising from lattices. A good reference on lattices is [Bir67].

Let (L, \preceq) be a **lattice** (possibly infinite), i.e., a partially ordered set in which any two elements $x, y \in L$ have a join $x \lor y$ and a meet $x \land y$. A function $v : L \longrightarrow \mathbb{R}_+$ satisfying

(5.4)
$$v(x \lor y) + v(x \land y) = v(x) + v(y) \text{ for all } x, y \in L.$$

is called a valuation on L. The valuation v is said to be isotone if $v(x) \le v(y)$ whenever $x \le y$ and positive if v(x) < v(y) whenever $x \le y, x \ne y$. Set

(5.5)
$$d_v(x,y) = v(x \lor y) - v(x \land y) \text{ for all } x, y \in L.$$

One can easily check that (L, d_v) is a semimetric space if v is an isotone valuation on Land (L, d_v) is a metric space if v is a positive valuation on L; in the latter case, L is called a **metric lattice** (see [Bir67]). Clearly, every metric lattice is **modular**, i.e., satisfies $x \wedge (y \vee z) = (x \wedge y) \vee z$ for all x, y, z with $z \preceq x$. A lattice is called **distributive** if $x \wedge (y \vee z) = (x \wedge y) \vee (x \wedge z)$ for all x, y, z. The following result gives a characterization of the L_1 -embeddable metric lattices.

THEOREM 5.6 [Kel70] Let L be a metric lattice with positive valuation v. The following assertions are equivalent.

(i) L is a distributive lattice. (ii) (L, d_v) is 5-gonal.

(iii) (L, d_v) is hypermetric.

(iv) (L, d_v) is L_1 -embeddable.

PROOF. It suffices to show the implications $(ii) \Rightarrow (i)$ and $(i) \Rightarrow (iv)$. $(ii) \Rightarrow (i)$. Using the definition of the valuation v and applying the 5-gonal inequality (4.4) to the points $x_1 = x \lor y, x_2 = x \land y, x_3 = z, y_1 = x, y_2 = y$, we obtain the inequality: $2(v(x \lor y \lor z) - v(x \land y \land z)) \le v(x \lor y) + v(x \lor z) + v(y \lor z) - v(x \land y) - v(x \land z) - v(y \land z)$. By applying again the 5-gonal inequality to the points $x_1 = x, x_2 = y, x_3 = z, y_1 = x \lor y$, $y_2 = x \wedge y$, we obtain the reverse inequality. Therefore, the equality holds in the above inequality. In fact, this condition of equality is equivalent to L being distributive (see [Bir67]).

 $(i) \Rightarrow (iv)$. Take a finite subset L_0 of L. We show that (L_0, d_v) is L_1 -embeddable. Let K be the sublattice of L generated by L_0 . Suppose K has length n. Then, K is isomorphic to a ring \mathcal{N} of subsets of a set X, |X| = n ("ring" means closed under \cup and \cap)(see [Bir67] p.58). Via this isomorphism, we have a valuation, again denoted by v, defined on \mathcal{N} . We can assume without loss of generality that $v(\emptyset) = 0$. Then, v can be extended to a valuation v^* on 2^X satisfying $v^*(S) = \sum_{x \in S} v^*(\{x\})$ for $S \subseteq X$. Now, if $x \mapsto S_x$ is the isomorphism from K to \mathcal{N} , then we have the embedding $x \mapsto S_x$ from (L_0, d_v) to $(2^X, v^*)$ which is isometric. Indeed, $d_v(x, y) = v(x \lor y) - v(x \land y) = v(S_x \cup S_y) - v(S_x \cap S_y) = v^*(S_x \cup S_y) - v^*(S_x \cap S_y) = v^*(S_x \triangle S_y)$. This shows that every finite subset of (L, d_v) is L_1 -embeddable, and thus, from Theorem 1.4, (L, d_v) is L_1 -embeddable.

EXAMPLE 5.7 [Ass79] Let (\mathbb{N}^*, \preceq) denote the lattice consisting of the set \mathbb{N}^* of positive integers with order relation $x \preceq y$ if x divides y. Then, for $x, y \in \mathbb{N}^*$, $x \land y$ is the g.c.d. of x and y and $x \lor y$ is their l.c.m. One checks easily that (\mathbb{N}^*, \preceq) is a distributive lattice. Therefore, (\mathbb{N}^*, d_v) is L_1 -embeddable for every positive valuation v on \mathbb{N}^* . For instance, $x \in \mathbb{N}^* \mapsto v(x) := \log x$ is a positive valuation on \mathbb{N}^* ; hence, the metric d_v , defined by $d_v(x, y) = \log(\frac{l.c.m.(x,y)}{g.c.d.(x,y)})$ for all integers $x, y \ge 1$, is L_1 -embeddable.

We now present an analogue of Theorem 5.6 in the context of semigroups. We recall that a commutative semigroup (S, +) consists of a set S equipped with a composition rule + which is commutative and associative. We assume the existence of a neutral element denoted by 0.

THEOREM 5.8 ([BC76], see [Ass79, Ass80b]) Let (S, +) be a commutative semigroup with neutral element 0 and let $v: S \mapsto \mathbb{R}_+$ be a mapping such that v(0) = 0. Set

$$D_v(x,y) = 2v(x+y) - v(2x) - v(2y)$$
 for $x, y \in S$.

Assume that one of the following assertions (i) or (ii) holds. (i) (S, +) is a group. (ii) For each $x \in S$, there exists an integer $n \ge 1$ such that 2nx = x. Then, (S, D_v) is L_1 -embeddable if and only if (S, D_v) is of negative type.

PROOF. We only give a sketch of the proof of the implication: if (S, D_v) is of negative type, then (S, D_v) is L_1 -embeddable.

Let $p_v : S^2 \longrightarrow \mathbb{R}$ denote the symmetric function obtained by applying the covariance transformation ξ to D_v , i.e., $p_v(x, y) = \frac{1}{2}(D_v(x, 0) + D_v(y, 0) - D_v(x, y)) = v(x) + v(y) - v(x + y)$ for $x, y \in S$. By Lemma 3.11, showing that (S, D_v) is L_1 -embeddable amounts to showing that p_v is a $\{0, 1\}$ -covariance on S. By assumption, (S, D_v) is of negative type or, equivalently, by Lemma 4.8, p_v is of positive type on S. Berg and Christensen [BC76] show that, under this assumption, the function v is of the form

$$v(x) = h(x) + \int_{\hat{S} - \{\hat{1}\}} (1 - \rho(x)) \mu(d\rho) \text{ for all } x \in S,$$

where

- the function $h: S \longrightarrow \mathbb{R}_+$ satisfies h(x+y) = h(x) + h(y) for all $x, y \in S$,

- \hat{S} denotes the set of characters on S, i.e., of the functions $\rho : S \longrightarrow [-1, 1]$ satisfying $\rho(x + y) = \rho(x)\rho(y)$ for all $x, y \in S$ and $\rho(0) = 1$, and $\hat{1}$ is the unit character defined by $\hat{1}(x) = 1$ for all $x \in S$,

- μ is a nonnegative Radon measure on $\hat{S} - \{\hat{1}\}$ such that $\int_{\hat{S} - \{\hat{1}\}} (1 - \rho(x)) \mu(d\rho) < \infty$ for all $x \in S$.

Therefore, we have that $p_v(x,y) = \int_{\hat{S}-\{\hat{1}\}} (1-\rho(x))(1-\rho(y))\mu(d\rho)$ for all $x,y \in S$. In case (i), every character on S takes only values ± 1 . Setting $A_x = \{\rho \in \hat{S} : \rho(x) = -1\}$ for $x \in S$, we obtain that $p_v(x,y) = 4\mu(A_x \cap A_y)$ for all $x, y \in S$. In case (ii), every character on S takes only values 0, 1. Setting $A_x = \{\rho \in \hat{S} : \rho(x) = 0\}$ for $x \in S$, we obtain that $p_v(x,y) = \mu(A_x \cap A_y)$ for all $x,y \in S$. Therefore, p_v is a $\{0,1\}$ -covariance, i.e., (S, D_v) is L_1 -embeddable.

EXAMPLE 5.9 Let (L, \leq) be a lattice and let S be a subset of L which is stable under the join operation \lor of L and contains the least element 0 of L. Then, (S, \lor) is a commutative semigroup satisfying Theorem 5.8 (ii). Therefore, given a mapping $v : S \longrightarrow \mathbb{R}_+$ such that v(0) = 0, (S, D_v) is L_1 -embeddable if and only if (S, D_v) is of negative type, with $D_v(x, y) = 2v(x \lor y) - v(x) - v(y)$ for $x, y \in S$. In particular, if v is a valuation on L, i.e., satisfies (5.4), then D_v coincides with d_v (which is defined in (5.5)) and, therefore, we have the following variation of Theorem 5.6: (L, d_v) is L_1 -embeddable if and only if (L, d_v) is of negative type.

EXAMPLE 5.10 Let \mathcal{A} be a family of subsets of a set Ω and suppose that \mathcal{A} is stable under the symmetrical difference. Then, (\mathcal{A}, Δ) is a commutative group. Let $v : \mathcal{A} \mapsto \mathbb{R}_+$ be a mapping such that $v(\emptyset) = 0$ and set $d(\mathcal{A}, \mathcal{B}) = v(\mathcal{A} \Delta \mathcal{B})(=\frac{D_v(\mathcal{A}, \mathcal{B})}{2})$ for $\mathcal{A}, \mathcal{B} \in \mathcal{A}$. Then, by Theorem 5.8, (\mathcal{A}, d) is L_1 -embeddable if and only if (\mathcal{A}, d) is of negative type.

6 Metric transforms preserving L₁-embeddability

Let (X, d) be a distance space and let $F : \mathbb{R}_+ \longrightarrow \mathbb{R}_+$ be a function. We define the distance space (X, F(d)), called **metric transform** of (X, d), by setting $F(d)_{ij} = F(d_{ij})$ for all $i, j \in X$.

A general question is to find nontrivial functions F which preserve certain properties, e.g., metricity, L_1 - or L_2 -embeddability, of the original distance space. We present in this Section several results of this type about metric transforms. We refer to [WW75] for a good exposition of some of these results.

LEMMA 6.1 [Blu70] Suppose that F is nondecreasing, concave and F(0) = 0. If (X, d) is a semimetric space, then (X, F(d)) is also a semimetric space.

PROOF. Let $i, j, k \in X$. We show that $F(d_{ij}) \leq F(d_{ik}) + F(d_{jk})$ holds. But, $F(d_{ij}) \leq F(d_{ik} + d_{jk})$ holds since F is nondecreasing and $F(d_{ik} + d_{jk}) \leq F(d_{ik}) + F(d_{jk})$ holds since F is concave (to see it, verify that the function $t \mapsto F(t + d_{jk}) - F(t) - F(d_{jk})$ is nondecreasing).

A function satisfying the conditions of Lemma 6.1 is called a **scale**. Examples of scales include:

• $F(t) = \frac{t}{1+t}, t \ge 0.$

• $F(t) = 1 - \exp(-\lambda t)$ for $t \ge 0$, where λ is a positive scalar; it is called the **Schoenberg** scale.

• $F(t) = t^{\alpha}$ for $t \ge 0$, where $0 < \alpha \le 1$; it is called the **power scale**.

6.1 Metric transforms of ℓ_2 -spaces

We present two classical results about the functions F for which the metric transform $F(\ell_2^m)$ of the *m*-dimensional Euclidian space $\ell_2^m = (\mathbb{R}^m, d_{\ell_2})$ is L_2 -embeddable or embeddable into some Euclidian space ℓ_2^n .

THEOREM 6.2 (i) [Sch38a] Let $2 \le m \le n$ be integers. The functions $t \in \mathbb{R}_+ \mapsto F(t)$ that are nonnegative, continuous, satisfy F(0) = 0, and for which $F(\ell_2^m)$ is isometrically embeddable in ℓ_2^n are of the form F(t) = ct ($t \ge 0$), where $c \ge 0$.

(ii) [vNS41] Let $m \ge 1$ be an integer. The functions $t \in \mathbb{R}_+ \mapsto F(t)$ that are nonnegative, continuous, satisfy F(0) = 0, and for which $F(\ell_2^m)$ is isometrically L_2 -embeddable are of the form

$$F(t) = \left(\int_0^\infty \frac{1 - \Omega_m(tu)}{u^2} \sigma'(u) du\right)^{1/2} \quad (t \ge 0)$$

where $u \in \mathbb{R}_+ \mapsto \sigma(u)$ is nondecreasing, $\sigma(0) = 0$, and $\int_1^\infty \frac{1}{u^2} \sigma'(u) du < \infty$ (with σ' denoting the first derivative of σ). The function Ω_m is defined by

$$\Omega_m(t) = 1 - \frac{t^2}{2 \cdot m} + \frac{t^4}{2 \cdot 4 \cdot m \cdot (m+2)} - \frac{t^6}{2 \cdot 4 \cdot 6 \cdot m \cdot (m+2) \cdot (m+4)} + \dots$$

For m = 1, we have $\Omega_1(t) = \cos(t)$ and, thus, the functions F are of the form

$$F(t) = \left(\int_0^\infty \frac{\sin^2(tu)}{u^2} \sigma'(u) du\right)^{1/2} \quad (t \in \mathbb{R}_+).$$

PROOF. As an illustration, let us give the proof of the easy implication in (*ii*) for the case m = 1. Let F be defined as in the case m = 1 of Theorem 6.2 (*ii*). By Theorem 4.16, in order to show that $F(\ell_2^1)$ is L_2 -embeddable, it suffices to check that $F^2(\ell_2^1)$ is of negative type. By Lemma 4.8, this is equivalent to checking that its image under the covariance mapping is of positive type. Let $b_1, \ldots, b_k \in \mathbb{R}$ and $x_1, \ldots, x_k \in \mathbb{R}$; we show that the inequality $\sum_{1 \leq i,j \leq k} b_i b_j (F^2(x_i) + F^2(x_j) - F^2(x_i - x_j)) \geq 0$ holds. For this, we use the identity

$$\begin{aligned} \sin^2(x_i u) + \sin^2(x_j u) - \sin^2((x_i - x_j)u) &= 2\sin^2(x_i u)\sin^2(x_j u) + \frac{\sin(2x_i u)\sin(2x_j u)}{2}. \\ \text{Indeed, we deduce from it that} \\ \sum_{1 \le i, j \le k} b_i b_j (F^2(x_i) + F^2(x_j) - F^2(x_i - x_j)) \\ &= \int_0^\infty \left(2\left(\sum_{i=1}^k b_i \sin^2(x_i u)\right)^2 + \frac{1}{2}\left(\sum_{i=1}^k b_i \sin(2x_i u)\right)^2\right) \frac{d\sigma(u)}{u^2} \ge 0. \end{aligned}$$

EXAMPLE 6.3 [Sch37] Consider the function $F(t) = t^{\alpha}$ $(t \ge 0)$ where $0 < \alpha < 1$. Then, $F(\ell_2^1)$ is L_2 -embeddable. Indeed, F satisfies the conditions of Theorem 6.2 (*ii*); this fact relies on the following integral formula

$$t^{2\alpha} = c_{\alpha}^{-1} \int_{0}^{\infty} u^{-1-2\alpha} \sin^{2}(tu) du \ (t \ge 0) \ \text{ where } c_{\alpha} = \int_{0}^{\infty} u^{-1-2\alpha} \sin^{2}(u) du.$$

6.2 The Schoenberg scale

We now consider the **Schoenberg scale**, $F(t) = 1 - \exp(-\lambda t)$ $(t \in \mathbb{R}_+)$, where λ is a positive scalar. We show below that this scale preserves L_1 -embeddability and the negative type property.

THEOREM 6.4 [Sch38b] Let (X, d) be a distance space. The following assertions are equivalent.

(i) (X, d) is of negative type. (ii) The symmetric function $p: X \times X \longrightarrow \mathbb{R}$, defined by $p(x, y) = \exp(-\lambda d(x, y))$ for $x, y \in X$, is of positive type for all $\lambda > 0$. (iii) $(X, 1 - \exp(-\lambda d))$ is of negative type for all $\lambda > 0$.

PROOF. Note that the properties involved in Theorem 6.4 are all of finite type, i.e., they hold if and only if they hold for any finite subset of X. Hence, we can assume that X is finite, say $X = \{1, ..., n\}$.

(i) \Rightarrow (ii) Since (X, d) is of negative type then, by Theorem 4.16, (X, \sqrt{d}) is ℓ_{2} embeddable, i.e., there exist $x^{(1)}, \ldots, x^{(n)} \in \mathbb{R}^m$ $(m \ge 1)$ such that $d_{jk} = (||x^{(j)} - x^{(k)}||_2)^2$ for all $j, k \in X$. Let $b_1, \ldots, b_n \in \mathbb{R}$. We show that $\sum_{1 \le j, k \le n} b_j b_k \exp\left(-\lambda(||x^{(j)} - x^{(k)}||_2)^2\right) \ge 0$. For this, we use the following classical identity

$$\exp(-x^2) = 2^{-1} \pi^{-\frac{1}{2}} \int_{-\infty}^{\infty} \exp(ixu) \exp(-\frac{u^2}{4}) du.$$

(Here, *i* denotes the complex square root of unity.) Indeed, we have that
$$\begin{split} &\sum_{j,k\in X} b_j b_k \exp\left(-\lambda(\parallel x^{(j)} - x^{(k)} \parallel_2)^2\right) \\ &= \sum_{j,k\in X} b_j b_k \prod_{1\leq h\leq m} \exp(-\lambda(x_h^{(j)} - x_h^{(k)})^2) \\ &= \sum_{i,j\in X} b_j b_k 2^{-m} \pi^{-\frac{m}{2}} \prod_{1\leq h\leq m} \int_{-\infty}^{\infty} \exp(i\sqrt{\lambda}(x_h^{(j)} - x_h^{(k)})u_h) \exp(-\frac{u_h^2}{4}) du_h \\ &= \sum_{j,k\in X} b_j b_k 2^{-m} \pi^{-\frac{m}{2}} \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} \exp(i\sqrt{\lambda}(x^{(j)} - x^{(k)})^T u) \exp(-\frac{1}{4}\sum_{1\leq h\leq m} u_h^2) du_1 \dots du_m \\ &= 2^{-m} \pi^{-\frac{m}{2}} \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} \left|\sum_{j\in X} b_j \exp(i\sqrt{\lambda}x^{(j)^T} u)\right|^2 \exp(-\frac{1}{4}(\sum_{1\leq h\leq m} u_h^2)) du_1 \dots du_m \geq 0. \\ (ii) \implies (iii) \text{ Set } d'_{ij} = \exp(-\lambda d_{ii}) + \exp(-\lambda d_{jj}) - 2\exp(-\lambda d_{ij})) = 2(1 - \exp(-\lambda d_{ij})) \\ \text{for } i, j \in X, \text{ i.e., } d' \text{ arises from } p = \exp(-\lambda d) \text{ by applying the inverse of the covariance} \\ \text{mapping (defined in (3.8)). Applying Lemma 4.8, we obtain that <math>(X, d')$$
 is of negative type, i.e., $(X, 1 - \exp(-\lambda d))$ is of negative type. \\ (iii) \implies (i) \text{ Let } b_1, \dots, b_n \in \mathbb{R} \text{ with } \sum_{1\leq i\leq n} b_i = 0. \text{ We show that the inequality} \\ \sum_{1\leq i\leq j\leq n} b_i b_j d_{ij} \leq 0 \text{ holds. By expanding in series the exponential function, we ob-} \end{split}

tain that

$$\sum_{1 \le i < j \le n} b_i b_j (1 - \exp(-\lambda d_{ij})) = \lambda \left(\sum_{1 \le i < j \le n} b_i b_j d_{ij} - \frac{\lambda}{2} \sum_{1 \le i < j \le n} b_i b_j d_{ij}^2 + \frac{\lambda^2}{3!} \sum_{1 \le i < j \le n} b_i b_j d_{ij}^3 - \ldots \right) \le 0$$
for all $\lambda > 0$, since $1 - \exp(-\lambda d)$ is of negative type. By taking the limit when $\lambda \to 0$, we obtain that $\sum_{1 \le i < j \le n} b_i b_j d_{ij} \le 0$.

Remark that Theorem 6.4 remains valid if we assume only that (ii) and (iii) hold for a set of positive λ 's admitting 0 as accumulation point. The same remark also applies to the next Theorem 6.5 (i). THEOREM 6.5 [Ass79, Ass80b] Let (X, d) be a distance space. Then,

(i) (X, d) is L_1 -embeddable if and only if $(X, 1 - \exp(-\lambda d))$ is L_1 -embeddable for all $\lambda > 0$. (ii) Let ν be a positive measure on \mathbb{R} and set $f(t) = \int_0^\infty (1 - \exp(-\lambda t))\nu(d\lambda)$ for $t \ge 0$. If (X, d) is L_1 -embeddable, then (X, f(d)) is L_1 -embeddable.

PROOF. (i) Assume that (X, d) is L_1 -embeddable. Hence, (X, d) is an isometric subspace of a measure semimetric space $(\mathcal{A}_{\mu}, d_{\mu})$ for some measure space $(\Omega, \mathcal{A}, \mu)$. Set v(A) = $1 - \exp(-\lambda\mu(A))$ for $A \in \mathcal{A}_{\mu}$. Then, the distance space $(\mathcal{A}_{\mu}, 1 - \exp(\lambda d_{\mu}))$ coincides with the space $(\mathcal{A}_{\mu}, d_{\nu})$, which is defined as in Example 5.10. Therefore, in order to show that $(\mathcal{A}_{\mu}, 1 - \exp(-\lambda d_{\mu}))$ is L_1 -embeddable, it suffices to show that $(\mathcal{A}_{\mu}, 1 - \exp(-\lambda d_{\mu}))$ is of negative type. But, we know from Theorem 6.4 that $(\mathcal{A}_{\mu}, 1 - \exp(-\lambda d_{\mu}))$ is of negative type, since $(\mathcal{A}_{\mu}, d_{\mu})$ is L_1 -embeddable and, thus, of negative type.

The proof of the converse implication is analogue to that of the implication $(iii) \Longrightarrow (i)$ of Theorem 6.4 (replacing the negative type inequality by an arbitrary inequality valid for the cut cone CUT(Y) where Y is a finite subset of X).

(*ii*) Again we may suppose that (X,d) is an isometric subspace of $(\mathcal{A}_{\mu}, d_{\mu})$ for some measure space $(\Omega, \mathcal{A}, \mu)$. Set $v(A) = f(\mu(A))$ for $A \in \mathcal{A}_{\mu}$. Then, the space $(\mathcal{A}_{\mu}, f(d_{\mu}))$ coincides with the space $(\mathcal{A}_{\mu}, d_{v})$, which is constructed as in Example 5.10. We check that $(\mathcal{A}_{\mu}, d_{v})$ is of negative type, which will imply that $(\mathcal{A}_{\mu}, f(d_{\mu}))$ is L_{1} -embeddable. Indeed, let $A_{i} \in \mathcal{A}_{\mu}, b_{i} \in \mathbb{R}$ for $i \in X$ such that $\sum_{i \in X} b_{i} = 0$. Then,

 $\sum_{i,j\in X} b_i b_j d_\nu (A_i \triangle A_j) = \int_0^\infty \sum_{i,j\in X} b_i b_j (1 - \exp(-\lambda \mu (A_i \triangle A_j))\nu(d\lambda) \ge 0$ because $1 - \exp(-\lambda d)$ is of negative type and ν is a positive measure.

EXAMPLE 6.6 [Ass79, Ass80b] If (X, d) is L_1 -embeddable, then (X, d^{α}) is L_1 -embeddable for all $0 \le \alpha \le 1$.

This is a consequence of Theorem 6.5 (*ii*) and of the following integral formula

$$t^{\alpha} = e_{\alpha}^{-1} \int_0^\infty (1 - \exp(-\lambda^2 t)) \lambda^{-1-2\alpha} d\lambda \ (t \ge 0) \quad \text{where } e_{\alpha} = \int_0^\infty (1 - \exp(-\lambda^2)) \lambda^{-1-2\alpha} d\lambda.$$

6.3 The biotope transform

We mention another transformation which preserves L_1 -embeddability. Let d be a distance on a set X and let s be a point of X. We define a new distance $d^{(s)}$ on X by setting

$$d^{(s)}(i,j) = \frac{d(i,j)}{d(i,s) + d(j,s) + d(i,j)}$$

for all $i, j \in X$. In particular, if $(\Omega, \mathcal{A}, \mu)$ is a measure space and (X, d) is the measure semimetric space $(\mathcal{A}_{\mu}, d_{\mu})$, then its transform $d_{\mu}^{(\emptyset)}$ takes the form

$$d_{\mu}^{(\emptyset)}(A,B) = \frac{\mu(A \triangle B)}{\mu(A) + \mu(B) + \mu(A \triangle B)} = \frac{\mu(A \triangle B)}{2\mu(A \cup B)}$$

for $A, B \in \mathcal{A}_{\mu}$. The distance $(A, B) \in \mathcal{A}_{\mu} \times \mathcal{A}_{\mu} \mapsto \frac{\mu(A \triangle B)}{\mu(A \cup B)}$ is called the **Steinhaus distance**. The distance $(A, B) \mapsto \frac{|A \triangle B|}{|A \cup B|}$, which is obtained in the special case when μ is the cardinality measure, is also called the **biotope distance**. This terminology comes from the fact that this distance is used in some biological problems for the study of biotopes (see [MS58]). As a consequence of the next Proposition 6.7, the Steinhaus and biotope distances are L_1 -embeddable.

PROPOSITION 6.7 (i) [MS58] If d is a semimetric on X, then $d^{(s)}$ is a semimetric on X. (ii) [Ass80b] If (X, d) is L_1 -embeddable, then $(X, d^{(s)})$ is also L_1 -embeddable.

PROOF. (i) follows from (ii) and the fact that a distance space on at most 4 points is L_1 -embeddable if and only if it is a semimetric space (see Remark 4.22 (i)). (ii) We can suppose that (X, d) is an isometric subspace of some measure semimetric space $(\mathcal{A}_{\mu}, d_{\mu})$, i.e., $d(i, j) = \mu(A_i \triangle A_j)$ where $A_i \in \mathcal{A}_{\mu}$ for all $i, j \in X$, and we can suppose without loss of generality that $A_s = \emptyset$. Hence, as was already observed, $d^{(s)}(i, j) = \frac{\mu(A_i \triangle A_j)}{2\mu(A_i \cup A_j)}$ for all $i, j \in X$. By Lemma 3.11, showing that $(X, d^{(s)})$ is L_1 -embeddable amounts to showing that $p = \xi_s(d^{(s)})$ is a $\{0, 1\}$ -covariance. From (3.7), p is defined by $p(i, j) = \frac{1}{2}(d^{(s)}(i, s) + d^{(s)}(j, s) - d^{(s)}(i, j))$ for $i, j \in X \setminus \{s\}$. Hence, $p(i, j) = \frac{1}{4} + \frac{1}{4}\frac{\mu(A_i \cap A_j)}{\mu(A_i \cup A_j)}$ for $i, j \in X \setminus \{s\}$. Therefore, it suffices to show that the symmetric function $(i, j) \in (X \setminus \{s\})^2 \mapsto \frac{\mu(A_i \cap A_j)}{\mu(A_i \cup A_j)}$ is a $\{0, 1\}$ -covariance. For this, we use the identity $\frac{\mu(A \cap B)}{\mu(A \cup B)} = \frac{\mu(A \cap B)}{\mu(\Omega)} \left(\sum_{i \ge 0} \left(\frac{\mu(\overline{A} \cap \overline{B})}{\mu(\Omega)}\right)^i\right)$ (which follows from the identity $\sum_{i \ge 0}(1-u)^i = \frac{1}{u}$ for all $0 < u \le 1$) and the fact that $\{0, 1\}$ -covariances are preserved under taking sum, product and limit ([Ass80b], see Corollary 4.25 (ii)).

6.4 The power scale

We finally consider the **power scale**, $F(t) = t^{\alpha}$ for $t \ge 0$, where $0 < \alpha < 1$. The question is to determine the largest exponent α for which the power scale preserves some metric properties as hypermetricity, ℓ_1 -, or ℓ_2 -embeddability.

We consider the parameters g(n), h(n), t(n), $c_1(n)$, $c_2(n)$ which are defined as follows: q(n) (resp. $h(n), t(n), c_1(n), c_2(n)$) is the maximum exponent $\alpha, 0 \leq \alpha \leq 1$, such that the distance space (V_n, d^{α}) is *i*-gonal for all $i \leq 2\lfloor \frac{n}{2} \rfloor + 1$ (resp. hypermetric, of negative type, ℓ_1 -embeddable, ℓ_2 -embeddable).

From Theorem 4.20 and Theorem 4.16, we have the relations $c_2(n) \leq c_1(n) \leq h(n) \leq \min(t(n), g(n))$ and $t(n) = 2c_2(n)$, respectively.

Set $\gamma(s) = \log_2(1 + \frac{1}{s})$ for s > 0. The following results are given in [DM90]. • $g(2n) = \gamma(n-1)$ for $n \ge 2$ and $g(2n+1) = \gamma(n)$ for $n \ge 1$. Note that, if $d(K_{n,n})$ denotes the path metric of the complete bipartite graph $K_{n,n}$, then $(d(K_{n,n}))^{\alpha}$ violates the 2n-gonal inequality if $\alpha > \gamma(n-1)$. This shows that $g(2n) \leq \alpha$ $\gamma(n-1)$. Similarly, $(d(K_{n,n+1}))^{\alpha}$ violates the 2n+1-gonal inequality if $\alpha > \gamma(n)$, showing that $g(2n+1) \leq \gamma(n)$.

• $h(n) \ge \gamma(n-1)$ for all $n \ge 2$. This implies that $c_2(n) \ge \frac{\gamma(n-1)}{2}$ for all $n \ge 2$.

• Let $d(K_{m,n})$ denote the path metric of the complete bipartite graph $K_{m,n}$. Then, $d(K_{m,n})^{\alpha}$ is ℓ_2 -embeddable if and only if $c \leq \frac{1}{2}\gamma(\frac{2mn}{m+n}-1)$. This implies that

$$\begin{cases} c_2(2n) \le \frac{1}{2}\gamma(n-1) & \text{for all } n \ge 2, \\ c_2(2n+1) \le \frac{1}{2}\gamma(\frac{2n(n+1)}{2n+1} - 1) & \text{for all } n \ge 1. \end{cases}$$

Deza and Maehara [DM90] conjecture that the above inequalities for $c_2(2n)$ and $c_2(2n+1)$ hold at equality. It is known that $c_2(3) = 1 = \frac{1}{2}\gamma(\frac{1}{3})$ (easy), $c_2(4) = \frac{1}{2} = \frac{1}{2}\gamma(1)$ ([Blu70]) and $c_2(6) = \frac{1}{2}\gamma(2)$ ([DM90]), i.e., the conjecture holds for n = 3, 4, 6.

We summarize the known information for n = 3, 4, 5, 6:

- $q(3) = h(3) = c_1(3) = c_2(3) = 1$, and t(3) = 2,
- $t(4) = g(4) = h(4) = c_1(4) = 1$ and $c_2(4) = \frac{1}{2}$,
- $g(5) = h(5) = c_1(5) = \gamma(2) = \log_2(\frac{3}{2})$ and $\frac{1}{2}\gamma(2) \le c_2(5) \le \frac{1}{2}\gamma(\frac{7}{5}) = \frac{1}{2}\log_2(\frac{12}{7})$. $t(6) = g(6) = h(6) = c_1(6) = \gamma(2) = \log_2(\frac{3}{2})$ and $c_2(6) = \frac{1}{2}\gamma(2) = \frac{1}{2}\log_2(\frac{3}{2})$.

Additional questions on ℓ_1 -embeddings 7

In this Section, we address the following two questions.

- Evaluate the minimum ℓ_p -dimension $m_p(n)$ of an ℓ_p -embeddable distance space on n points.

- Determine the smallest integer c(m) such that, for every distance space (X, d), (X, d) is ℓ_1^m -embeddable if and only if every subspace of (X,d) on c(m) points embeds in ℓ_1^m .

7.1 On the minimum ℓ_p -dimension

We consider here the problem of evaluating the minimum ℓ_p -dimension $m_p(n)$ of an ℓ_p embeddable space on n points. We recall the definition of $m_p(n)$ from relation (1.2), i.e., $m_p(n)$ is the smallest integer m such that any ℓ_p -embeddable space on n points can embedded in ℓ_p^m . The results we present come essentially from [Bal90] and can be stated as follows.

As was already observed in relations (1.3) and (4.18), $m_{\infty}(n) \leq n-1$ and $m_2(n) = n-1$ but, for general p, it is not immediate that $m_p(n)$ is finite. Wolfe [Wol67] showed that $m_{\infty}(n) \leq n-2$. Witsenhausen proved that $m_1(n) \leq \binom{n}{2}$ and Ball extended the result for any $p \geq 1$. In other words, every ℓ_p -embeddable distance on n points can embedded in ℓ_p^m , where $m = \binom{n}{2}$. Moreover, for $1 \leq p < 2$, this result is essentially best possible, since $m_p(n) \geq \binom{n-1}{2}$ for $1 , <math>n \geq 3$, and $m_1(n) \geq \binom{n-2}{2}$ for $n \geq 4$. One can also show that $m_1(4) = m_{\infty}(4) = 2$, $m_1(5) = 3$ and $m_1(6) = 6$. It is conjectured in [Bal90] that $m_1(n) = \binom{n-2}{2}$ for all $n \geq 5$.

Ball's proof for the upper bound $m_p(n) \leq {n \choose 2}$ is based on an application of Caratheodory's Theorem to the cut cone (if p = 1) or the cone NOR_n(p) (for $p \geq 1$; see the definition below). We first present the result in the case p = 1.

PROPOSITION 7.1 $m_1(n) \leq {n \choose 2}$.

PROOF. Let d be a distance on n points that is ℓ_1 -embeddable, i.e., d belongs to the cut cone CUT_n . We show that d can be embedded in ℓ_1^m , where $m = \binom{n}{2}$. For this, it suffices to show that d can be decomposed as a nonnegative linear combination of at most $\binom{n}{2}$ distinct cut semimetrics (recall Remark 2.6). Let H denote the hyperplane in \mathbb{R}^{E_n} defined by the equation $\sum_{1 \leq i < j \leq n} x_{ij} = 1$. Then, the section $\operatorname{CUT}_n \cap H$ of the cut cone CUT_n by H is a polytope of dimension $\binom{n}{2} - 1$ whose vertices are the vectors $\frac{\delta(S)}{|\delta(S)|}$ for all subsets S of V_n . Set $a = \sum_{1 \leq i < j \leq n} d_{ij}$. Then, $\frac{d}{a} \in \operatorname{CUT}_n \cap H$ and, thus, by Caratheodory's Theorem, $\frac{d}{a}$ can be written as the convex hull of at most $\binom{n}{2}$ members of $\{\frac{\delta(S)}{|\delta(S)|} : S \subseteq V_n\}$. This shows that d can written as the conic hull of at most $\binom{n}{2}$ cut semimetrics.

The result from Proposition 7.1 can extended for any $p \ge 1$, using the following cone NOR_n(p) instead of the cut cone CUT_n.

Given an integer $p \ge 1$, let $NOR_n(p)$ denote the set of all distances d on V_n for which $d^{\frac{1}{p}}$ is ℓ_p -embeddable, i.e., there exist n vectors $v_1, \ldots, v_n \in \mathbb{R}^m$ $(m \ge 1)$ such that $d_{ij} = || v_i - v_j ||_p^p$ for all $1 \le i < j \le n$.

Note that, if p = 1, then NOR_n(1) coincides with the cut cone CUT_n (by Proposition 2.5). An element $d \in NOR_n(p)$ is said to be **linear** if $d^{\frac{1}{p}}$ is ℓ_n^1 -embeddable, i.e., if

there exist $x_1, \ldots, x_n \in \mathbb{R}$ such that $d_{ij} = |x_i - x_j|^p$ for all $1 \le i < j \le n$. For example, each cut semimetric belongs to NOR_n(p) and is linear, i.e.,

$$\operatorname{CUT}_n \subseteq \operatorname{NOR}_n(p)$$
 for all $1 \leq i < j \leq n$.

We collect in the next result some properties of the set $NOR_n(p)$.

LEMMA 7.2 (i) $NOR_n(p)$ is a cone.

(ii) Let $d \in NOR_n(p)$. Then, $d^{\frac{1}{p}}$ is ℓ_p^m -embeddable if and only if d is the sum of m linear members of $NOR_n(p)$. In particular, if d lies on an extreme ray of $NOR_n(p)$, then d is linear.

PROOF. (i) Let $d, d' \in \text{NOR}_n(p)$. We show that $d + d' \in \text{NOR}_n(p)$. By assumption, there exist some vectors $u_1, \ldots, u_n, v_1, \ldots, v_n \in \mathbb{R}^m$ $(m \ge 1)$ such that $d_{ij} = || u_i - u_j ||_p^p$ and $d'_{ij} = || v_i - v_j ||_p^p$ for all $1 \le i < j \le n$. Set $w_i = (u_i, v_i) \in \mathbb{R}^{2m}$ for all $1 \le i \le n$. Then, $d_{ij} + d'_{ij} = || w_i - w_j ||_p^p$ for all $1 \le i < j \le n$. This shows that $d + d' \in \text{NOR}_n(p)$. Hence, $\text{NOR}_n(p)$ is a cone.

(*ii*) Let $d \in NOR_n(p)$. If $d^{\frac{1}{p}}$ is ℓ_p^m -embeddable, then there exist $u_1, \ldots, u_n \in \mathbb{R}^m$ such that $d_{ij} = || u_i - u_j ||_p^p = \sum_{1 \le h \le m} |(u_i)_h - (u_j)_h|^p$ for $1 \le i < j \le n$. Hence, $d = d^1 + \ldots + d^m$, where d^h denotes the distance on V_n defined by $(d^h)_{ij} = |(u_i)_h - (u_j)_h|$ for $1 \le i < j \le n$. This shows that d is the sum of m linear members of $NOR_n(p)$, since d^1, \ldots, d^m belong to $NOR_n(p)$ and are linear, by construction. The converse implication holds similarly.

PROPOSITION 7.3 $m_p(n) \leq {n \choose 2}$.

PROOF. We sketch the proof. Let H denote again the hyperplane in \mathbb{R}^{E_n} defined by the equation $\sum_{1 \leq i < j \leq n} x_{ij} = 1$. Set $L = \{d \in \operatorname{NOR}_n(p) : d \in H \text{ and } d \text{ is linear}\}$. One can show that L is a compact set and that $\operatorname{NOR}_n(p) \cap H$ is a $\binom{n}{2} - 1$ -dimensional convex set which coincides with the convex hull of L. As in the proof of Proposition 7.1, Caratheodory's Theorem implies that every member of $\operatorname{NOR}_n(p)$ can be written as the sum of $\binom{n}{2}$ linear members of $\operatorname{NOR}_n(p)$. Now, suppose d is an ℓ_p -embeddable distance on n points. Then, $d^p \in \operatorname{NOR}_n(p)$ and, thus, d^p is the sum of $\binom{n}{2}$ members of $\operatorname{NOR}_n(p)$, i.e., d embeds in ℓ_p^m , where $m = \binom{n}{2}$.

PROPOSITION 7.4 (i) $m_1(n) \ge \binom{n-2}{2}$ for $n \ge 4$. (ii) $m_p(n) \ge \binom{n-1}{2}$ for $1 and <math>n \ge 3$. PROOF. (i) Set $m = \binom{n-2}{2}$. We exhibit a semimetric d on V_n which embeds in ℓ_1^m but not in ℓ_1^k if k < m. Set $d = \sum_{2 \le r < s \le n-1} \delta(\{r, s, n\})$, i.e.,

 $\begin{cases} d_{1n} = \binom{n-2}{2} \\ d_{1i} = n-3 & \text{for } 2 \le i \le n-1 \\ d_{ij} = 2(n-4) & \text{for } 2 \le i \le j \le n-1 \\ d_{in} = \binom{n-3}{2} & \text{for } 2 \le i \le n-1. \end{cases}$ By construction, *d* embeds isometrically in ℓ_1^m . We show that *d* cannot be embedded in

By construction, d embeds isometrically in ℓ_1^m . We show that d cannot be embedded in ℓ_1^k if k < m. For this, we consider the following inequality of negative type $Neg_n(-(n-4), 1, \ldots, 1, -2)$, i.e., the inequality

(7.5)
$$2(n-4)x_{1n} - (n-4)\sum_{2 \le i \le n-1} x_{1i} - 2\sum_{2 \le i \le n-1} x_{in} + \sum_{2 \le i < j \le n-1} x_{ij} \le 0.$$

Let F denote the face of the cone $\operatorname{NOR}_n(1)$ (=CUT_n) defined by the inequality (7.5). One can easily check that d satisfies the inequality (7.5) at equality, i.e., d lies on the face F, and that the only cut semimetrics lying on F are the cut semimetrics $\delta(\{r, s, n\})$ for $2 \leq r < s \leq n - 1$. Moreover, these cut semimetrics are linearly independent, i.e., F is a simplicial face of $\operatorname{NOR}_n(1)$. One can also show (Lemma 4, [Bal90]) that

(*) the only linear members of $NOR_n(1)$ lying on F are of the form $\lambda\delta(\{r, s, n\})$ for $2 \le r < s \le n-1$ and $\lambda > 0$.

Let us suppose that d embeds in ℓ_1^k . Then, d is the sum of k linear members of NOR_n(1). From (*) above, we deduce that d can be written as a nonnegative linear combination of k of the cuts $\delta(\{r, s, n\})$ for $2 \leq r < s \leq n - 1$. Since d lies on a simplicial face, the latter decomposition of d must coincide with the initial decomposition $d = \sum_{2 \leq r < s \leq n-1} \delta(\{r, s, n\})$. Therefore, $k = \binom{n-2}{2}$.

(*ii*) We only sketch the proof, which is along the same lines as for (*i*). Set $m = \binom{n-1}{2}$. Consider the vectors $v_1, \ldots, v_n \in \mathbb{R}^m$ defined by

$$(v_i)_{rs} = \begin{cases} 1 & \text{if } r = i \\ -1 & \text{if } s = i \\ 0 & \text{otherwise} \end{cases}$$

for $1 \leq r < s \leq n$. Define the distance d on V_n by setting $d_{ij} = || v_i - v_j ||_p$ for $1 \leq i < j \leq n$. So d embeds in ℓ_p^m by construction. One can show that d does not embed in ℓ_p^k if k < m by using, as in case (i), a special inequality which is valid for the cone NOR_n(p) and is satisfied at equality by d^p . Namely, one uses the inequality

$$\sum_{1 \le i < j \le n} \| u_i - u_j \|_p^p - (n + 2^{p-1} - 2) \| u_i \|_p^p \le 0$$

which holds for any set of n vectors $u_1, \ldots, u_n \in \mathbb{R}^h$ $(h \ge 1)$ if $1 \le p \le 2$ ([Bal87]).

REMARK 7.6 The proof of the lower bound $\binom{n-2}{2}$ from Proposition 7.4 (i) uses essentially the fact that the inequality (7.5) defines a simplicial face F of NOR_n(1) that contains only $\binom{n-2}{2}$ linear members (up to multiple). Observe that the hypermetric inequality $\operatorname{Hyp}_n(-(n-4), 1, \ldots, 1, -1)$, i.e.,

$$(n-4)x_{1n} - (n-4)\sum_{2 \le i \le n-1} x_{1i} - \sum_{2 \le i \le n-1} x_{in} + \sum_{2 \le i < j \le n-1} x_{ij} \le 0$$

defines a simplicial facet G of NOR_n(1) that contains the face F. Hence, G contains $\binom{n}{2} - 1$ cut semimetrics but G may contain additional linear points. For this reason, the lower bound from Proposition 7.4 (i) cannot be improved using the facet G instead of the face F.

Linial, London and Rabinovitch [LLR93] define the **metric dimension** of a graph G as the smallest integer m for which there exists a norm $\| \cdot \|$ on \mathbb{R}^m such that the graphic space (V, d_G) of the graph G can be isometrically embedded into the space $(\mathbb{R}^m, d_{\|\cdot\|})$. The definition extends clearly to an arbitrary semimetric space. Hence, rather than looking only at embeddings in a fixed Banach ℓ_p -space, [LLR93] considers embeddings in an arbitrary normed space.

Actually, the notion of metric dimension is linked with ℓ_{∞} -embeddings in the following way. Let (V_n, d) be a semimetric space. Then, its metric dimension is equal to the minimum rank of a system of vectors $v_1, \ldots, v_n \in \mathbb{R}^k$ $(k \ge 1)$ providing an ℓ_{∞} -embedding of (V_n, d) , i.e., such that $d_{ij} = || v_i - v_j ||_{\infty}$ for all $1 \le i < j \le n$.

The metric dimension of several graphs is computed in [LLR93]. In particular, $dim(K_n) = \lceil \log_2(n) \rceil$, $dim(T) = O(\log_2(n))$ for a treee on *n* nodes (both being realized by an ℓ_{∞} -embedding), $dim(C_{2n}) = n$ for a cycle on 2n nodes (realized by an ℓ_1 -embedding), $dim(K_{n\times 2}) \ge n-1$ for the cocktail party graph (i.e., K_{2n} minus a perfect matching). It is also shown in [LLR93] that, if *G* is a graph on *n* nodes with metric dimension *d*, then each vertex has degree $\le 3^d - 1$, *G* has diameter $\ge \frac{1}{2}(n^{\frac{1}{d}} - 1)$, and there exists a subset *S* of $O(dn^{1-\frac{1}{d}})$ nodes whose deletion disconnects *G* and so that each connected component of $G \setminus S$ has no more than $(1 - \frac{1}{d} + o(1))n$ nodes.

7.2 Compactness results for ℓ_1 -embeddability in the plane

Let $m \ge 1$ be an integer and let $p \ge 1$. Define $c_p(m)$ as the smallest integer such that an arbitrary distance space (X, d) is ℓ_p^m -embeddable if and only if every subspace of (X, d) on $c_p(m)$ points is ℓ_p^m -embeddable. By convention, we set $c_p(m) = \infty$ if $c_p(m)$ does not exist.

The study of the parameter $c_p(m)$ is motivated by the following result of Menger for the case p = 2. Menger [Men28] showed that a distance space (X, d) embeds isometrically in the Euclidian space $(\mathbb{R}^m, d_{\ell_2})$ if and only if each subspace of (X, d) on m + 3 points embeds isometrically in $(\mathbb{R}^m, d_{\ell_2})$. In other words, $c_2(m) = m + 3$ for each $m \ge 1$.

Thus arises naturally the question of looking for analogues of Menger's Theorem for the case of arbitrary ℓ_p -metrics and, in particular, in the case p = 1.

Since the spaces (\mathbb{R}, d_{ℓ_1}) and (\mathbb{R}, d_{ℓ_2}) are identical, we deduce from Menger's Theorem that $c_1(1) = 4$. Malitz and Malitz [MM92] show that $6 \leq c_1(2) \leq 11$ and $c_1(m) \geq 2m + 1$ for all $m \geq 1$. The latter result is improved by J. Schmerl who proves that $c_1(m) \geq 2m + 2$ for all $m \geq 1$.

It is conjectured in [MM92] that $c_1(m)$ exists for all m and that $c_1(m) = 2m + 2$ for all m.

Note that if we know that $c_1(m)$ exists, then this implies the existence of a polynomial time algorithm for checking embeddability of a finite distance space in the space $(\mathbb{R}^m, d_{\ell_1})$, for any given m. We recall that, on the other hand, checking ℓ_1 -embeddability of a finite distance space (i.e., embeddability into some $(\mathbb{R}^m, d_{\ell_1})$ for unrestricted m) is NP-complete ([Kar85]).

8 Examples of the use of the L_1 -metric

8.1 The L₁-metric in probability theory

Let $(\Omega, \mathcal{A}, \mu)$ be a probability space and let $X : \Omega \longrightarrow \mathbb{R}$ be a random variable with finite expected value $E(X) = \int_{\Omega} |X(\omega)| \mu(d\omega) < \infty$, i.e., $X \in L_1(\Omega, \mathcal{A}, \mu)$. Let F_X denote the distribution function of X, i.e., $F_X(x) = \mu(\{\omega \in \Omega : X(\omega) = x\})$ for $x \in \mathbb{R}$; when it exists, its derivative F'_X is called the density of X. A great variety of metrics on random variables are studied in the monography [Rac91]; among them, the following are based on the L_1 -metric.

• The usual L_1 -metric between the random variables:

 $L_1(X,Y) = E(|X-Y|) = \int_{\Omega} |X(\omega) - Y(\omega)| \mu(d\omega).$

• The Monge-Kantorovich-Wasserstein metric (i.e., the L_1 -metric between the distribution functions): $k(X, Y) = \int_{\mathbb{R}} |F_X(x) - F_Y(x)| dx$.

• The total valuation metric (i.e., the L_1 -metric between the densities when they exist): $\sigma(X,Y) = \frac{1}{2} \int_{\mathbb{R}} |F'_X(x) - F'_Y(x)| dx.$

• The engineer metric (i.e., the L_1 -metric between the expected values): EN(X,Y) = |E(X) - E(Y)|.

• The indicator metric: $i(X, Y) = E(1_{X \neq Y}) = \mu(\{\omega \in \Omega : X(\omega) \neq Y(\omega)\}).$

In fact, the L_p -analogues $(1 \le p \le \infty)$ of the above metrics, especially of the first two, are also used in probability theory.

Several results are known, establishing links among the above metrics. One of the

main such results is the Monge-Kantorovich mass-transportation theorem which shows that the second metric k(X, Y) can be viewed as a minimum of the first metric $L_1(X, Y)$ over all joint distributions of X and Y with fixed marginal. A relationship between the $L_1(X, Y)$ and the engineer metric EN(X, Y) is given by [Rac91] as solution of a moment problem. Similarly, a connection between the total valuation metric $\sigma(X, Y)$ and the indicator metric i(X, Y) is given in Dobrushin's theorem on the existence and uniqueness of Gibbs fields in statistical physics. See [Rac91] for a detailed account of the above topics.

We mention another example of use of the L_1 -metric in probability theory, namely for Gaussian random fields. We refer to [Nod87, Nod89] for a detailed account. Let $B = (B(x); x \in M)$ be a centered Gaussian system with parameter space $M, 0 \in M$. The variance of the increment is denoted by

$$d(x, y) := E((B(x) - B(y))^2)$$
 for $x, y \in M$.

When (M, d) is a metric space which is L_1 -embeddable, the Gaussian system is called a Lévy's Brownian motion with parameter space (M, d). The case $M = \mathbb{R}^n$ and $d(x, y) = \|x - y\|_2$ gives the usual Brownian motion with *n*-dimensional parameter. By Lemma 2.10, (M, d) is L_1 -embeddable if and only if there exist a non negative measure space (H, ν) and a map $x \mapsto A_x \subseteq H$ with $\nu(A_x) < \infty$ for $x \in M$, such that $d(x, y) = \nu(A_x \triangle A_y)$ for $x, y \in M$. Hence, a Gaussian system admits a representation called of Chentsov type

$$B(x) = \int_{A_x} W(dh)$$
 for $x \in M$

in terms of a Gaussian random measure based on the measure space (H, ν) with $d(x, y) = \nu(A_x \triangle A_y)$ if and only if d is L_1 -embeddable.

This Chentsov type representation can be compared with the Crofton formula for projective metrics from Theorem 5.3. Actually both come naturally together in [Amb82] (see parts A.8-A.9 of Appendix A there).

8.2 The ℓ_1 -metric in statistical data analysis

A data structure is a pair (I,d), where I is a finite set, called bf population, and $d: I \times I \longrightarrow \mathbb{R}_+$ is a symmetric map with $d_{ii} = 0$ for $i \in I$, called dissimilarity index. The typical problem in statistical data analysis is to choose a "good representation" of a data structure; usually, "good" means a representation allowing to represent the data structure visually by a graphic display. Each sort of visual display corresponds, in fact, to a special choice of the dissimilarity index as a distance and the problem turns out to be the classical isometric embedding problem in special classes of metrics.

For instance, in hierarchical classification, the case when d is ultrametric corresponds to the possibility of a so-called indexed hierarchy (see [Joh67]). A natural extension is the case when d is the path metric of a weighted tree, i.e., d satisfies the four point condition (see Chapter ??? (on graphs)); then the data structure is called an **additive tree**. Also, data structures (I, d) for which d is ℓ_2 -embeddable are considered in factor analysis and multidimensional scaling. These two cases together with cluster analysis are the main three techniques for studying data structures. The case when d is ℓ_1 -embeddable is a natural extension of the ultrametric and ℓ_2 cases.

An ℓ_p -approximation consists of minimizing the estimator $|| e ||_p$, where e is a vector or a random variable (representing an error, deviation, etc.). The following criteria are used in statistical data analysis:

- the ℓ_2 -norm, in the least square method; or its square,
- the ℓ_{∞} -norm, in the minimax or Chebychev method,
- the ℓ_1 -norm, in the least absolute values (LAV) method.

In fact, the ℓ_1 criterion has been increasingly used. Its importance can be seen, for instance, from the volume [Dod87b] of proceedings of a conference entitled "Statistical data analysis based on the L_1 norm and related methods"; we refer, in particular, to [Dod87a], [Fic87], [Cal87], [Vaj87].

8.3 The ℓ_1 -metric in computer vision and pattern recognition

The ℓ_p -metrics are also used in the new area called pattern recognition, or robot vision, or digital topology; see, e.g., [RK86], [Hor86].

A computer picture is a subset of \mathbb{Z}^n (or of a scaling $\frac{1}{m}\mathbb{Z}^n$ of \mathbb{Z}^n) which is called a **digital** *n*-*D*-space (or an *n*-*D m*-quantized space). Usually, pictures are represented in the digital plane \mathbb{Z}^2 or in the digital 3-*D*-space \mathbb{Z}^3 . The points of \mathbb{Z}^n are called the **pixels**.

Given a picture in \mathbb{Z}^n , i.e., a subset A of \mathbb{Z}^n , one way to define its **volume** vol(A) is by vol(A) := |A|, i.e., as the number of pixels contained in A. Then, the distance

$$d(A,B) = vol(A \triangle B)$$

is used in digital topology for evaluating the distance between pictures. It is a digital analogue of the symmetric difference metric used in convex geometry, where the distance between two convex bodies A and B in \mathbb{R}^n is defined as the *n*-dimensional volume of their symmetric difference.

The above metric and other metrics on \mathbb{Z}^n are used for studying analogues of clasical geometric notions as volume, perimeter, shape complexity, etc., for computer pictures.

The metrics on \mathbb{Z}^n that are mainly used are the ℓ_1 -, ℓ_∞ -metrics, as well as the ℓ_2 -metric after rounding to the nearest upper (or lower) integer.

When considered on \mathbb{Z}^n , the ℓ_1 -metric is also called the **grid metric** and the ℓ_{∞} -metric is called the **lattice metric** (or **Chebyshev metric**, or **uniform metric**). More specific names are used in the case n = 2. Then, the ℓ_1 -metric is also called the **city-block metric** (or **Manhattan metric**, or **taxi-cab metric**, or **rectilinear metric**), or the **4-metric** since each point of \mathbb{Z}^2 has exactly 4 closest neighbours in \mathbb{Z}^2 for the ℓ_1 -metric. Similarly, the ℓ_{∞} -metric on \mathbb{Z}^2 is called the **chessboard metric**, or the **8-metric** since each pixel has exactly 8 closest neighbours in \mathbb{Z}^2 . Note indeed that the unit sphere $S^1_{\ell_1}$ (centered at the origin) for the ℓ_1 -norm in \mathbb{R}^2 contains exactly 4 integral points while the unit sphere $S^1_{\ell_{\infty}}$ for the ℓ_{∞} -norm contains 8 integral points.

Observe that the ℓ_1 -metric, when considered on \mathbb{Z}^n , can be seen as the path metric of an (infinite) graph on \mathbb{Z}^n . Namely, consider the graph on \mathbb{Z}^n where two lattice points are adjacent if their ℓ_1 -distance is equal to 1; this graph is nothing but the usual grid. Then, the shortest path distance of two lattice points in the grid is equal to their ℓ_1 -distance. Similarly, the ℓ_{∞} -metric on \mathbb{Z}^n is the path metric of the graph on \mathbb{Z}^n where adjacency is defined by the pairs at ℓ_{∞} -distance one; actually, adjacency corresponds to the king move in chessboard terms.

There are some other useful metrics on \mathbb{Z}^2 which are obtained by combining the ℓ_1 and ℓ_{∞} -metrics. The following two examples, the octogonal and the hexagonal distances, are path metrics; hence, in order to define them, it suffices to describe the pairs of lattice points at distance 1, i.e., to describe their unit balls.

The octogonal distance d_{oct} .

For each $(x, y) \in \mathbb{Z}^2$, its unit sphere $S_{oct}^1(x, y)$, centered at (x, y), is defined by

$$S_{oct}^{1}(x,y) = S_{\ell_{1}}^{3}(x,y) \cap S_{\ell_{\infty}}^{2}(x,y),$$

where $S^3_{\ell_1}(x,y)$ denotes the ℓ_1 -sphere of radius 3 and $S^2_{\ell_{\infty}}(x,y)$ the ℓ_{∞} -sphere of radius 2, centered at (x,y). Hence, $S^1_{oct}(x,y)$ contains exactly 8 integral points; note that moving from (x,y) to its eight neighbours at distance 1 corresponds to the knight move in chessboard terms. Figure 2 shows the spheres $S^3_{\ell_1}$, $S^2_{\ell_{\infty}}$, and S^1_{oct} .

The **hexagonal distance** or **6-metric** d_{hex} . Its unit sphere $S_{hex}^1(x, y)$, centered at $(x, y) \in \mathbb{Z}^2$, is defined by

$$S_{hex}^{1}(x,y) = S_{\ell_{1}}^{1}(x,y) \cup \{(x-1,y-1), (x-1,y+1)\} \text{ for } x \text{ even},$$

$$S_{hex}^{1}(x,y) = S_{\ell_{1}}^{1}(x,y) \cup \{(x+1,y-1), (x+1,y+1)\} \text{ for } x \text{ odd}.$$

The unit sphere $S_{hex}^1(x, y)$ contains exactly 6 integral points. Figure 3 shows the unit spheres $S_{hex}^1(0,0)$ and $S_{hex}^1(1,-3)$.

Several other modifications of the ℓ_1 -metric on the plane have been considered; see, e.g., [Ber91] and references there.

Figure 2

Figure 3

In practice, the subset $(\mathbb{Z}_k)^n := \{0, 1, \ldots, k-1\}^n$ is considered instead of the full space \mathbb{Z}^n . Note that $(\mathbb{Z}_2)^n$ is nothing but the vertex set of the *n*-dimensional hypercube and $((\mathbb{Z}_2)^n, d_{\ell_1})$ is the *n*-dimensional hypercube metric space. Note also that $(\mathbb{Z}_3)^2$ is the unit ball (centered at (1,1)) of the space $(\mathbb{Z}^n, d_{\ell_\infty})$. $(\mathbb{Z}_4)^n$ is known as the **tic-tac-toe board** (or **Rubik's** *n*-**cube**) and $(\mathbb{Z}_k)^2$, $(\mathbb{Z}_k)^3$ are called, respectively, the *k*-grill and the *k*-framework.

Other distances are used on $(\mathbb{Z}_k)^n$, in particular in coding theory, namely, the **Ham**ming distance d_H defined by

$$d_H(x,y) = |\{1 \le i \le n : x_i \ne y_i\}| \text{ for all } x, y \in (\mathbb{Z}_k)^n,$$

and the **Lee distance** d_{Lee} defined by

$$d_{Lee}(x,y) = \sum_{1 \le i \le n} \min(|x_i - y_i|, k - |x_i - y_i|) \text{ for all } x, y \in (\mathbb{Z}_k)^n.$$

The metric space (\mathbb{Z}_k, d_{Lee}) can be seen as a discrete analogue of the elliptic metric space (which consists of the set of all the lines in \mathbb{R}^2 going through the origin and where the distance between two such lines is their angle).

The ℓ_1 -distance and the Hamming distance coincide when restricted to $(\mathbb{Z}_2)^n$, i.e., the spaces $((\mathbb{Z}_2)^n, d_{\ell_1})$ and $((\mathbb{Z}_2)^n, d_H)$ are identical. Also, $(\mathbb{Z}_k, d_{\ell_1})$ coincides with the graphic metric space of the path P_k on k nodes, (\mathbb{Z}_k, d_H) coincides with the graphic space of the complete graph K_k on k nodes, and (\mathbb{Z}_k, d_{Lee}) coincides with the graphic space of the cycle C_k on k nodes. Therefore, the spaces $((\mathbb{Z}_k)^n, d_{\ell_1}), ((\mathbb{Z}_k)^n, d_H)$ and $((\mathbb{Z}_k)^n, d_{Lee})$ coincide with the graphic space of the cartesian product G^n , where G is P_k , K_k and C_k , respectively.

One can easily check that

(i) P_k embeds isometrically in the (k-1)-dimensional hypercube, i.e., $(\mathbb{Z}_k, d_{\ell_1})$ is an isometric subspace of $((\mathbb{Z}_2)^{k-1}, d_{\ell_1})$ (simply, label each $x \in \mathbb{Z}_k$ by the binary string $1 \dots 10 \dots 0$ of lenght k-1 whose first x letters are equal to 1). Hence, $((\mathbb{Z}_k)^n, \Phi d_{\ell_1})$ is an isometric subspace of $((\mathbb{Z}_2)^{n(k-1)}, d_{\ell_1})$.

(*ii*) $((\mathbb{Z}_k)^n, d_H)$ is an isometric subspace of $((\mathbb{Z}_2)^{kn}, \frac{1}{2}d_{\ell_1})$ (label each $x \in \mathbb{Z}_k$ by the binary string of lenght k whose letters are all equal to 0 except the (x + 1)th one equal to 1).

(*iii*) The even cycle C_{2k} embeds isometrically into the k-dimensional hypercube. Therefore, $((\mathbb{Z}_{2k})^n, d_{Lee})$ is an isometric subspace of $((\mathbb{Z}_2)^{nk}, d_{\ell_1})$. Also, $(((\mathbb{Z}_{2k+1})^n, d_{Lee}))$ is an isometric subspace of $((\mathbb{Z}_2)^{(2k+1)n}, \frac{1}{2}d_{\ell_1})$ (since the odd cycle C_{2k+1} embeds isometrically into the (2k+1)-dimensional halfcube).

More details about the ℓ_1 -embeddings of the graphs P_k , C_k and K_k can be found in Chapter 3.

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