

Metrics

A *metric space* $M = (V, d)$ consists of a set of points V and a function $d : V \times V \rightarrow \mathbb{R}_{\geq 0}$ satisfying the properties:

- (Symmetry) $d(x, y) = d(y, x)$ for all $x, y \in V$.
- (Triangle inequality) $d(x, y) + d(y, z) \geq d(x, z)$ for all $x, y, z \in V$.

The definition above is often called a *semi-metric*, and a metric space is also required to satisfy the property that $d(x, y) = 0 \iff x = y$. However, we will blur the distinction between semi-metrics and metrics.

Diameter. The *diameter* of a metric is $\max_{x, y \in V} d(x, y)$.

Ball. The *ball* $\mathbf{B}(x, r) := \{y \in V \mid d(x, y) \leq r\}$. The *open ball* $\mathbf{B}^\circ(x, r) := \{y \in V \mid d(x, y) < r\}$.

r -net. A set of points $N \subseteq V$ which is:

- (r -packing) $d(x, y) \geq r$ for all $x, y \in N$, and
- (r -covering) for $x \in V$, there exists $y \in N$ such that $d(x, y) \leq r$.

One can build an r -net using a simple greedy algorithm.

Distortion

Distortion. Given metrics $M = (V, d)$ and $M' = (V', d')$, and a map $f : M \rightarrow M'$,

- the *expansion* of f is $\max_{x, y \in V} \frac{d'(f(x), f(y))}{d(x, y)}$.
- the *contraction* of f is $\max_{x, y \in V} \frac{d(x, y)}{d'(f(x), f(y))}$.
- the *distortion* of f is

$$\text{expansion} \times \text{contraction} = \max_{x, y \in V} \frac{d'(f(x), f(y))}{d(x, y)} \times \max_{x, y \in V} \frac{d(x, y)}{d'(f(x), f(y))}.$$

If $\text{distortion}(f) \leq D$, we write this as $M \xrightarrow{D} M'$. When we write $M_1 \xrightarrow{\geq D} M_2$, this is a *lower bound* statement: every map $f : V \rightarrow V'$ has distortion at least D .

This naturally extends to the case when \mathcal{G} is a *family of metrics or graphs*, then $M \xrightarrow{D} \mathcal{G}$ implies that *there exists* $M' \in \mathcal{G}$ such that $M \xrightarrow{D} M'$; similarly, $M \xrightarrow{\geq D} \mathcal{G}$ implies that *for all* $M' \in \mathcal{G}$, it holds that $M \xrightarrow{D} M'$. If $M \xrightarrow{1} M'$, then we say that M *isometrically embeds* into M' ; or just that M *embeds into* M' .

Metric Families

ℓ_p Spaces. For $1 \leq p < \infty$, the metric space ℓ_p consists of all infinite sequences $x = (x_i)_{i \leq 0}$ in $\mathbb{R}^{\mathbb{N}}$ for which $\sum_i |x_i|^p$ is finite; the distance is given by $|x - y|_p := (\sum_i |x_i - y_i|^p)^{1/p}$. The space ℓ_∞ is the set of bounded infinite sequences x , with the distance $|x - y|_\infty = \max_i (|x_i - y_i|)$. Often we will deal with ℓ_p^m for some finite m , when these sequences x just represent points in m -dimensional space \mathbb{R}^m ; the distances are defined in the same way as above.

We say a metric M is an ℓ_p -metric (or it belongs to ℓ_p) if there is an isometric embedding of M into ℓ_p .

Tree-Metric. A metric $M = (V, d)$ is a *tree metric* if there exists a tree $T = (V \cup S, E)$ with edge-weights, such that the shortest-path distance in T according to these edge-weights (denoted by d_T) agrees with d on all pairs in $V \times V$ —in other words, $d_T(x, y) = d(x, y)$ for all $x, y \in V$.

Given a class \mathcal{G} of graphs, one can define a \mathcal{G} -metric in the same way as above. E.g., we will often talk about *planar graph metrics*.

k -HST. A *k -Hierarchical well-Separated Tree* is rooted (weighted) tree with the following properties: (a) it is a balanced tree—all the leaves are at the same depth, (b) given any node x in the tree, all the children edges of x have the same length l_x , and the length of the edge from x to its parent node p_x (if any) has length $k \times l_x$. Hence, if the length of the root’s children edges is L , and the height of the tree is h , then the edge lengths on any root-leaf path are $(L, L/k, L/k^2, \dots, L/k^{h-1})$.

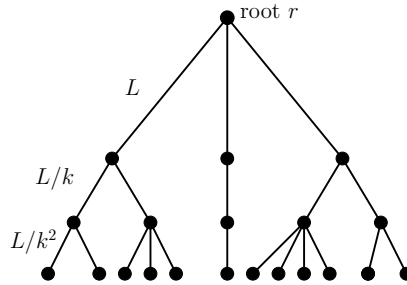


Figure 1: A k -HST with height 3.

Distributions over trees. Given a metric space $M = (V, \delta)$ on $|V| = n$ points, let \mathcal{T} be the set of trees $T = (V, E_T)$ on the with vertex set V with edge lengths $\ell : E_T \rightarrow \mathbb{R}$ such that each edge $e = \{u, v\} \in E_T$ has length $\ell(\{u, v\}) \geq \delta(u, v)$ —i.e., trees whose distances *dominate* those in M .

A probability distribution \mathcal{D} on this set of “dominating” trees \mathcal{T} is said to α -*approximate* the metric M if for every $u, v \in V$,

$$\mathbf{E}_{T \sim \mathcal{D}}[d_T(u, v)] \leq \alpha \cdot \delta(u, v). \quad (1)$$

I.e., for any two points, the expected distance in a random tree (drawn from this distribution) is at most α times what it was in (V, δ) .

Other Useful Definitions

Padded Decomposition. A metric $M = (V, d)$ is said to admit an α -*padded decomposition* if there exists a randomized procedure that takes as input a parameter $\Delta > 0$, and outputs a (random) partition V_1, V_2, \dots, V_k of the set V with the following properties:

- each set V_i has diameter at most Δ ,
- for any $\rho > 0$, the probability $\Pr[\mathbf{B}(x, \rho)$ split by partitioning] $\leq \alpha \cdot \frac{\rho}{\Delta}$.

Note that this probability is taken over the randomness of the padded decomposition procedure. (The ball $\mathbf{B}(x, \rho)$ is *split by* the partitioning if it is not contained within any single “cluster” V_i .)

Tree Cover. Given a metric $M = (V, d)$, an (α, k) -tree cover is a collection of trees $\mathcal{T} = \{T_1, T_2, \dots, T_k\}$ such that for any pair of nodes $x, y \in V$, there exists a tree $T_j \in \mathcal{T}$ with

$$d(x, y) \leq d_{T_j}(x, y) \leq \alpha \cdot d(x, y).$$

Neighborhood Cover. Given a metric $M = (V, d)$, an (α, r, t) -neighborhood cover is a collection $\mathcal{S} = \{S_1, S_2, \dots\}$ of subsets $S_i \subseteq V$ of points such that (a) for each point $x \in V$, there is a subset S_j that contains the r -ball $B(x, r) = \{x' \in V \mid d(x, x') \leq r\}$, (b) each point $x \in V$ is contained in at most t of the subsets in \mathcal{S} , and (c) each subset S_i has diameter at most $O(\alpha r)$.

Graphs

Outerplanar Graphs. These are planar graph such that there exists a face containing all the vertices; often this face is drawn as the outer face, hence the name. Equivalently, these are the graphs that exclude $K_{2,3}$ and K_4 as minors.

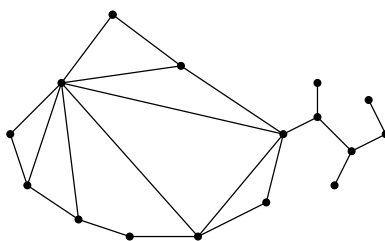


Figure 2: An outerplanar graph.

Series-Parallel Graphs. An (s, t) -series-parallel graph G is either (a) a single edge (s, t) , or (b) the graph obtained by taking an (s_1, t_1) -series-parallel graph and an (s_2, t_2) -series-parallel graph and identifying $s_1 = s_2 = s$ and $t_1 = t_2 = t$ (this is called a *parallel composition*, or (c) the graph obtained by taking an (s_1, t_1) -series-parallel graph and an (s_2, t_2) -series-parallel graph and identifying $t_1 = s_2$ and setting $s_1 = s$ and $t_2 = t$ (this is called a *series composition*). A series-parallel graph G is a graph that contains vertices s and t such that G is an (s, t) -series-parallel graph.

Equivalently, take any planar graph that excludes K_4 as a minor: each 2-node-connected component of this is a series-parallel graph.

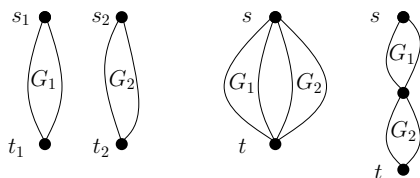


Figure 3: Parallel and Series compositions.

Expander Graphs. A (d, α) -expander graph on n vertices is a d -regular graph $G_n = (V_n, E_n)$ such that for every set $S \subseteq V_n$ with $|S| \leq n/2$, the number of edges in ∂S (i.e., with one endpoint in S and the other in $V \setminus S$) is at least $\alpha|S|$.

We are interested in families of graphs (for infinitely many values of n) where both the degree d and the “expansion parameter” α are constants (independent of the size n). In this case, we just refer to the graphs as constant-degree expander graphs.

One can show (by a probabilistic construction) that there exist constant degree expander graphs; explicit constructions are known as well. For more details, see the survey by Linial, Hoory and Wigderson.

Diamond Graphs. Let the graph G_0 be a single edge, and for each $i \geq 1$, let G_i be obtained by taking G_{i-1} and replacing each edge by an (s, t) -series-parallel graphs consisting of two paths of length 2 (see figure below).

