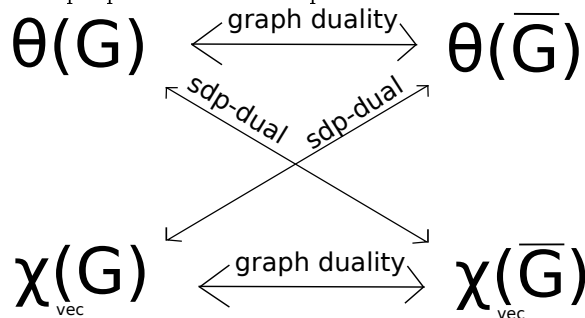


# NOTES ON LOVASZ THETA

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The purpose of this example is to illustrate some of the structure to  $\vartheta(G)$ .



Though the quantities above are scalars, they arise from arrangements of vectors. I will write down the vectors associated with each quantity above when  $G$  is the Petersen Graph.

## DEFINING $\vartheta$

First, we need a series of definitions to define  $\vartheta$ :

**Definition 1.** Given a graph  $G$ , an orthonormal representation of  $G$  is a mapping  $r : V(G) \rightarrow \mathbb{S}^n \subset \mathbb{R}^{n+1}$  (for some  $n \in \mathbb{N}$ ) such that if  $i \neq j \in V(G)$ , with  $i \not\sim j$ , then  $r(i) \perp r(j)$ . (Be careful: a vertex is not adjacent to itself).

Note that each graph has at least one representation, where  $v$  maps the vertices each to its own orthonormal vector.

**Definition 2.** A valuation of an orthonormal representation  $val(r)$  is

$$\min_{\psi} \max_{v \in V(G)} \frac{1}{(\psi^T r(v))^2}$$

where  $\psi$  ranges over all unit vectors (of the target space of  $r$ ).

Given an orthonormal representation, its valuation is how tightly it can be embedded into a cone around some vector ( $\psi$ ).

**Definition 3.** Define  $\vartheta(G)$  to be the minimum valuation over all orthonormal representations of  $G$ .

We can show that this minimum is actually attained. We will use Bolzano-Weistrass. To see this, fix  $n$ , and consider the orthonormal representations of the form:  $v : V(G) \rightarrow \mathbb{S}^{n-1} \subset \mathbb{R}^n$ . Observe that  $\vartheta(G)$  remains unchanged if we require that  $\psi = (1, 0, 0, \dots)$ : These valuations are defined by an inner product, which will not change if we apply a fixed unitary  $U$  to every vector. Choose  $U$  to send  $\psi \mapsto (1, 0, 0, \dots)$ .

Fixing  $\psi$ , take a sequence of orthonormal representations whose values converge to  $\vartheta(G)$ . Observe that these orthonormal representations themselves can be considered as bounded vectors of dimension  $n \cdot V(G)$ , by concatenating all  $V(G)$  vectors of dimension  $n$ . By the Bolzano-Weirstrass theorem, these have a convergent subsequence, so there is an accumulation point,  $r_\infty$ , which we must show is an orthonormal representation.

Our convergent subsequence of orthonormal representations gives rise to  $V(G)$  convergent sequences of vectors. We must show that each sequence of vectors goes to a unit vector, and that when  $i \not\sim j$ , with  $i \neq j$ , we have  $r_\infty(i)^T r_\infty(j) = 0$ . Both of these are consequences of the fact that dot products are continuous:  $0 = \lim_{n \rightarrow \infty} r_n(i)^T r_n(j) = (\lim_{n \rightarrow \infty} r_n(i))^T (\lim_{n \rightarrow \infty} r_n(j))$ .

There is no claim that such optimal representations are unique.

### GRAPHS

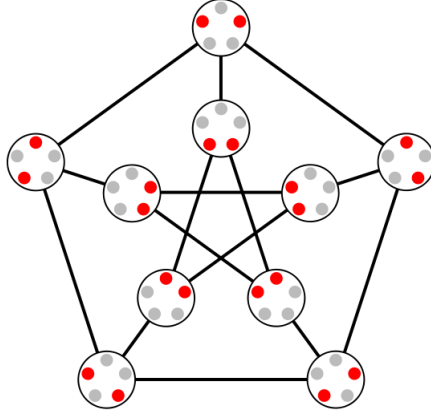
**Definition 4.** Define the Kneser Graph  $k(n, r)$  to have  $\binom{n}{r}$  vertices labeled by  $r$ -element subsets from a universe of size  $n$ . Two vertices are adjacent if their corresponding sets are disjoint. We assume that  $n \geq 2r$

Kneser graphs are vertex and edge transitive. Given any two vertices (edges), there is an automorphism which sends one to the other.

**Theorem 5.** If  $G$  is vertex and edge transitive, then  $\vartheta(G) \vartheta(\overline{G}) = n$ , and  $\vartheta(G) = \frac{-n\lambda_n}{\lambda_1 - \lambda_n}$

This powerful theorem was originally used to find  $\vartheta(k(n, r))$ . The proof of the theorem builds on the relations in the diagram.

**Definition 6.** The Petersen Graph,  $P$ , is the Kneser graph,  $k(5, 2)$ .



We start with some graph properties.

*Claim 7.* The clique number of the kneser graph  $\omega(k(n, r)) = \lfloor \frac{n}{r} \rfloor$ , so  $\omega(P) = 2$

A clique corresponds to a collection of disjoint sets.

*Claim 8.* The coloring number  $\chi(k(n, r)) = n - 2r + 2$ , so  $\chi(P) = 3$

This was a big open problem for many years. The optimal coloring is the following: Order the elements of the universe  $u_1, \dots, u_n$ , and divide them into 3 pieces

with sizes  $n - 2r$ ,  $r$  and  $r$ . Let  $x$  be an  $r$ -set. If it intersects the first piece, color  $x$  with the color  $i$ , where  $i = \min \{i \mid u_i \in x\}$ . Otherwise,  $x$  is contained entirely in the last two pieces. These remaining vertices form a subgraph, where each vertex has a unique neighbor, and these can be colored with two colors.

*Claim 9.* The independence number of the kneser graph is  $\alpha(k(n, r)) = \binom{n-1}{r-1}$ , so  $\alpha(P) = 4$

The collection of  $r$ -subsets which each contain  $u_1$  is a set of this size. It isn't hard to show this is optimal.

*Claim 10.* The clique covering number is  $q(k(n, r)) = \left\lceil \frac{\binom{n}{k}}{\lfloor \frac{n}{k} \rfloor} \right\rceil$ ,  $q(P) = 5$

(!) (Claim found on Wolfram Mathworld.)

*Claim 11.*  $\vartheta(k(n, r)) = \binom{n-1}{r-1}$ , and  $\vartheta(k(n, r)) = \frac{n}{r}$ , so  $\vartheta(P) = 4$ ,  $\vartheta(\overline{P}) = \frac{5}{2}$

This is proven by Theorem 5 and some tricky algebra, but this avoids (or at least obscures) creating explicit orthonormal representations, which is the point of this example.

#### RELATIONS BETWEEN GRAPH CONSTANTS

**Theorem 12.**  $\alpha(G) \chi(G) \geq |V(G)|$

Each color is an independent set, and a proper coloring colors every vertex.

**Theorem 13.** For any graph  $G$ ,  $\alpha(G) \leq \vartheta(G) \leq \chi(\overline{G}) = q(G)$

In an orthonormal representation, an independent set,  $\alpha$ , of  $G$  must be sent to a collection of pairwise independent vectors. For such vectors, it is easy to see that  $\max_{v_i \in \alpha} \frac{1}{(\psi^T r(v_i))^2}$  is minimized when  $\psi = \frac{\sum_{v_i \in \alpha} r(v_i)}{\sqrt{|V(G)|}}$  (when  $\psi$  is between all the vectors.) In this case,  $\frac{1}{(\psi^T r(v_i))^2} = |\alpha|$ , and this lower bound holds for all orthonormal representations. This shows  $\alpha(G) \leq \vartheta(G)$ .

Suppose we have clique cover of size  $q(G)$ . Define an orthonormal representation by choosing  $q(G)$  pairwise orthonormal vectors. Send each clique to one of these vectors. This provides an explicit orthonormal representation with valuation  $q(G)$ . The minimum over all orthonormal representations may be less.

#### ORTHONORMAL REPRESENTATIONS

We start with the graph  $k(n, r)$  and construct an optimal orthonormal representation in dimension  $n$ , with orthonormal basis  $e_1, \dots, e_n$ . The choice is obvious: disjoint sets need to go to orthonormal vectors. Set  $e_i^T r(v_j) = \frac{1}{\sqrt{r}}$  if  $u_i \in v_j$ , and 0 otherwise. Set  $\psi = \frac{1}{\sqrt{n}}(1, 1, \dots, 1)$ . It is immediate that this is an orthonormal representation with valuation  $\frac{1}{(\psi^T r(v_i))^2} = \frac{r \cdot n}{r^2} = \frac{n}{r}$ .

**Definition 14.** Given an orthonormal representation, we can define the cost of a vertex to be  $c(v) = (\psi_1^T(r_1(v_i)))^2$ . This corresponds to the quantum-mechanical probability of measuring  $r_1(v_1)$  when measuring from state  $\psi$ .

**Theorem 15.** (Certification of Orthonormal Representations): if we have two orthonormal representations  $r_1, r_2$  of  $G$  and  $\overline{G}$  and for all  $i \in V(G)$  we have  $c_1(v_i) = \frac{1}{\vartheta}$ , and we also have  $\sum_i c_1(v_i) c_2(v_i) = 1$ , then  $\vartheta = \vartheta(G)$

*Proof.* We have the explicit orthonormal representation  $r_1$ , so  $\vartheta(G) \leq \vartheta$ . For the other direction, we use an alternate definition  $\vartheta$ ,  $\vartheta(G) = \max_{Rep(\overline{G})} \sum_i c(v_i)$ .

$$\vartheta = \sum_i \vartheta c_1(v_i) c_2(v_i) = \sum_i c_2 v_i \leq \max_{r \in Rep(\overline{G})} \sum_i c(v_i) = \vartheta(G)$$

□

The argument above also shows that certificates always exist.

**Definition 16.** Given a non-empty closed convex set  $P \subset \mathbb{R}_+^n$  with the property that  $x \in P$  and  $0 \leq x' \leq x$  then  $x' \in P$ , the antiblocker of  $P$  is

$$AB(P) = \{x \in \mathbb{R}_+^n : y^T x \leq 1 \text{ for all } y \in P\}$$

**Definition 17.**  $TH(G) = \{(c(v_i), v_i \in V(G)) \in \mathbb{R}_+^{V(G)}\}$ . These are assignable probabilities.

**Theorem 18.**  $AB(TH(G)) = TH(\overline{G})$

(!) These concepts provide a geometric description of the  $\vartheta$ .  $\vartheta$  is the maximal of a linear functional over  $TH(G)$ . This linear functional has hyperplanes as its level sets, and the optimal value corresponds to a level set which lies tangent to  $TH(G)$ . Thus,  $\vartheta(G)$  is the smallest simplex  $S$  such that  $TH(G) \subset S$ . If we take the antiblocker of this picture, we seek the largest cube  $C$  such that  $C \subset AB(TH(G)) = TH(\overline{G})$ . This explains the formula  $\vartheta(G) = \min_{O.R.} \max_i \frac{1}{c(v_i)} = \max_{O.R.} \min_i c(v_i)$ .

Next, we give an orthonormal representation of  $\vartheta(P)$ , which will certify the optimality of the orthonormal representation given at the beginning of this section.

Assume a basis of size 10,  $\{e_{s_1}, e_{s_2}, \dots, e_{s_{10}}\}$  labeled by the  $\binom{5}{2}$  subsets of the graph. Let  $\psi = \frac{1}{\sqrt{10}}(1, 1, \dots, 1)$ . Finally, assume that we will have  $e_{s_i}^T r(v_j) = x_{|s_i \cap v_j|}$ . This is a plausible assumption, because it will result in vectors whose orthogonality relations are invariant with respect to the automorphism group of  $P$ .

The fact that intersection sets must be sent to orthonormal vectors translates into the constraint

$$x_0^1 + 3x_1^2 + 4x_0x_1 + 2x_1x_2 = 0$$

At the same time, we would like to minimize  $\frac{10 \cdot (x_2^2 + 6x_1^2 + 3x_0^2)}{(x_2 + 6x_1 + 3x_0)^2}$ . According to Wolfram Alpha the minimum is 4, when  $(x_0, x_1, x_2) = (1, -4 - \sqrt{15}, 6 + \sqrt{15})$ , or when  $(x_0, x_1, x_2) = (1, \sqrt{15} - 4, 6 - \sqrt{15})$ .

## VECTOR COLORINGS OF GRAPHS

**Definition 19.** Given a graph  $G$ , we assign a unit vector to each vertex. This time, we would like adjacent vertices to be sent to vectors whose dot product is as negative as possible. If  $\chi(G) = k$ , then we can associate each color with a vector in the regular  $k$ -simplex in  $\mathbb{R}^{k+1}$ . Such vectors have inner product  $\frac{-1}{k-1}$ . In light of this, we define  $\chi_{vec}(G) = \min \left\{ k \mid v_i^T v_j \leq \frac{-1}{k-1} \right\}$

**Theorem 20.**  $\chi_{vec}(G) = \vartheta(\overline{G})$

This is proven by the fact that the two problems can be expressed as semidefinite-programming duals of one another. Alternatively, there is a concrete way to move between optimal representations of the coloring problem and optimal representations for  $\vartheta$ . Namely, if  $x_i$  is an optimal vector-coloring for  $\vartheta(\overline{G})$  and  $\psi$  is some unit vector orthogonal to each  $x_i$ , then  $u_i = \frac{1}{\sqrt{\vartheta}}(\psi + \sqrt{\vartheta - 1}x_i)$  (!)

Next, we will provide an optimal vector coloring of  $P$ . Assume a basis of size 5, and that we will map  $\star\star\circ\circ\circ \mapsto (a, a, b, b, b)$ , and extend this map by permutations of  $S_5$ . If  $x, y$  are two vector representations of intersecting sets, we would like to minimize

$$\min_{a,b} \frac{x^T y}{\|x\| \|y\|} = \frac{4ab + b^2}{2a^2 + 3b^2}$$

The minimum occurs at  $a = -3, b = 2$ , and gives

$$\frac{x^T y}{\|x\| \|y\|} = \frac{-24 + 4}{18 + 12} = -\frac{2}{3} = \frac{-1}{\frac{5}{2} - 1}$$

Finally, an optimal vector coloring of  $\overline{P}$  can be found by assuming a basis of size 10 (the same basis we used for  $\vartheta(P)$ ) and three variables,  $x_0, x_1, x_2$ . This gives us the optimization problem:

$$\min_{x_0, x_1, x_2} \frac{x^T y}{\|x\| \|y\|} = \frac{x_0^2 + 3x_1^2 + 4x_0x_1 + 2x_1x_2}{3x_0^2 + 6x_1^2 + x_2^2}$$

The minimum (according to Wolfram) is found at  $\left(-\frac{1}{\sqrt{18}}, \frac{1}{\sqrt{18}}, -\frac{1}{\sqrt{2}}\right)$  and gives

$$\frac{x^T y}{\|x\| \|y\|} = \frac{\frac{1}{18} + \frac{3}{18} - 4\frac{1}{18} - 2\frac{1}{6}}{3\frac{1}{18} + 6\frac{1}{18} + \frac{1}{2}} = \frac{-6}{18} = \frac{-1}{4 - 1}$$