

Now we head for an upper bound " $\Theta(G) \leq \sigma^{-1}$ " for the Shannon capacity of any graph G that has an especially "nice" orthonormal representation. For this let $T = \{v^{(1)}, \dots, v^{(m)}\}$ be an orthonormal representation of G in \mathbb{R}^s , where $v^{(i)}$ corresponds to the vertex. We assume in addition that all the vectors $v^{(i)}$ have the same angle ($\neq 90^\circ$) with the vector $u := \frac{1}{m}(v^{(1)} + \dots + v^{(m)})$, or equivalently that the inner product

$$\langle v^{(i)}, u \rangle = \sigma$$

has the same value $\sigma_T \neq 0$ for all i . Let us call this value σ the constant of the representation T . For the Lovász umbrella that represents C_5 the condition $\langle v^{(i)}, u \rangle = \sigma_T$ certainly holds, for $u = \overrightarrow{OS}$. Now we proceed in the following three steps.

(A) Consider a probability distribution $x = (x_1, \dots, x_m)$ on V and set

$$\mu(x) := |x_1 v^{(1)} + \dots + x_m v^{(m)}|^2$$

and

$$\mu_T(G) := \inf_x \mu(x)$$

Let U be a largest independent set in G with $|U| = \alpha$, and define $x_U = (x_1, \dots, x_m)$ with $x_i = \frac{1}{\alpha}$ if $v_i \in U$ and $x_i = 0$ otherwise. Since all vectors $v^{(i)}$ have unit length and $\langle v^{(i)}, v^{(j)} \rangle = 0$ for any two non-adjacent vertices, we infer

$$\mu_T(G) \leq \mu(x_U) = \left| \sum_{i=1}^m x_i v^{(i)} \right|^2 = \sum_{i=1}^m x_i^2 = \alpha \frac{1}{\alpha^2} = \frac{1}{\alpha}$$

Thus we have $\mu_T(G) \leq \alpha^{-1}$, and therefore

$$\alpha(G) \leq \frac{1}{\mu_T(G)}$$

(B) Next we compute $\mu_T(G)$. We need the Cauchy-Schwarz inequality

$$\langle a, b \rangle^2 \leq |a|^2 |b|^2$$

for vectors $a, b \in \mathbb{R}^s$. Applied to $a = x_1 v^{(1)} + \dots + x_m v^{(m)}$ and $b = u$, the inequality yields

$$\langle x_1 v^{(1)} + \dots + x_m v^{(m)}, u \rangle^2 \leq \mu(x) |u|^2. \quad (1)$$

By our assumption that $\langle v^{(i)}, u \rangle = \sigma_T$ for all i , we have

$$\langle x_1 v^{(1)} + \dots + x_m v^{(m)}, u \rangle = (x_1 + \dots + x_m) \sigma_T = \sigma_T$$

for any distribution x . Thus, in particular, this has to hold for the uniform distribution $(\frac{1}{m}, \dots, \frac{1}{m})$, which implies $|u|^2 = \sigma_T^2$. Hence (5) reduces to

$$\sigma_T^2 \leq \mu(x) \sigma_T \quad \text{or} \quad \mu_T(G) \geq \sigma_T$$

On the other hand, for $x = (\frac{1}{m}, \dots, \frac{1}{m})$ we obtain

$$\mu_T(G) \leq \mu(x) = \left| \frac{1}{m}(v^{(1)} + \dots + v^{(m)}) \right|^2 = |u|^2 = \sigma_T$$

and so we have proved

$$\mu_T(G) = \sigma_T \quad (2)$$

In summary, we have established the inequality

$$\alpha(G) \leq \frac{1}{\sigma_T} \quad (3)$$

for any orthonormal representation T with constant σ_T .

(C) To extend this inequality to $\Theta(G)$, we proceed as before. Consider again the product $G \times H$ of two graphs. Let G and H have orthonormal representations R and S in \mathbb{R}^r and \mathbb{R}^s , respectively, with constants σ_R and σ_S . Let $v = (v_1, \dots, v_r)$ be a vector in R and $w = (w_1, \dots, w_s)$ be a vector in S . To the vertex in $G \times H$ corresponding to the pair (v, w) we associate the vector

$$vw^T := (v_1w_1, \dots, v_1w_s, v_2w_1, \dots, v_2w_s, \dots, v_rw_1, \dots, v_rw_s) \in \mathbb{R}^{rs}$$

It is immediately checked that $R \times S := \{vw^T : v \in R, w \in S\}$ is an orthonormal representation of $G \times H$ with constant $\sigma_R\sigma_S$. Hence by (6) we obtain

$$\mu_{R \times S}(G \times H) = \mu_R(G)\mu_S(H).$$

For $G^n = G \times \dots \times G$ and the representation T with constant σ_T this means

$$\mu_{T^n}(G^n) = \mu_T(G)^n = \sigma_T^n$$

and by (7) we obtain

$$\alpha(G^n) \leq \sigma_T^{-n}, \quad \sqrt[n]{\alpha(G^n)} \leq \sigma_T^{-1}$$

Taking all things together we have thus completed Lovász' argument:

Theorem 1 *whenever $T = \{v^{(1)}, \dots, v^{(m)}\}$ is an orthonormal representation of G with constant σ_T , then*

$$\Theta(G) \leq \frac{1}{\sigma_T} \quad (4)$$

Looking at the Lovász umbrella, we have $u = (0, 0, h = \frac{1}{\sqrt[4]{5}})^T$ and hence $\sigma = \langle v^{(i)}, u \rangle = h^2 = \frac{1}{\sqrt{5}}$, which yields $\Theta(C_5) \leq \sqrt{5}$. Thus Shannon's problem is solved.

Let us carry our discussion a little further. We see from (8) that the larger σ_T is for a representation of G , the better a bound for $\Theta(G)$ we will get. Here is a method that gives us an orthonormal representation for any graph G . To $G = (V, E)$ we associate the adjacency matrix $A = (a_{ij})$, which is defined as follows: Let $V = \{v_1, \dots, v_m\}$, then we set

$$a_{ij} := \begin{cases} 1 & \text{if } v_i v_j \in E \\ 0 & \text{otherwise} \end{cases}$$

A is a real symmetric matrix with 0's in the main diagonal. Now we need two facts from linear algebra. First, as a symmetric matrix, A has m real eigenvalues $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_m$ (some of which may be equal), and the sum of the eigenvalues equals the sum of the diagonal entries of A , that is, 0. Hence the smallest eigenvalue must be negative (except in the trivial case when G has no edges). Let $p = |\lambda_m| = -\lambda_m$ be the absolute value of the smallest eigenvalue, and consider the matrix

$$M := I + \frac{1}{p}A$$

where I denotes the $(m \times m)$ -identity matrix. This M has the eigenvalues $1 + \frac{\lambda_1}{p} \geq 1 + \frac{\lambda_2}{p} \geq \dots \geq 1 + \frac{\lambda_m}{p} = 0$. Now we quote the second result (the principal axis theorem of linear algebra): If $M = (m_{ij})$ is a real symmetric matrix with all eigenvalues ≥ 0 , then there are vectors $v^{(1)}, \dots, v^{(m)} \in \mathbb{R}^s$ for $s = \text{rank}(M)$, such that

$$m_{ij} = \langle v^{(i)}, v^{(j)} \rangle \quad (1 \leq i, j \leq m).$$

In particular, for $M = I + \frac{1}{p}A$ we obtain

$$\langle v^{(i)}, v^{(i)} \rangle = m_{ii} = 1 \quad \text{for all } i$$

and

$$\langle v^{(i)}, v^{(j)} \rangle = \frac{1}{p}a_{ij} \quad \text{for } i \neq j$$

Since $a_{ij} = 0$ whenever $v_i v_j \notin E$, we see that the vectors $v^{(1)}, \dots, v^{(m)}$ form indeed an orthonormal representation of G . Let us, finally, apply this construction to the m -cycles C_m for odd $m > 5$. Here one easily computes $p = |\lambda_{\min}| = 2 \cos \frac{\pi}{m}$ (see the box). Every row of the adjacency matrix contains two 1's, implying that every row of the matrix M sums to $1 + \frac{2}{p}$. For the representation $\{v^{(1)}, \dots, v^{(m)}\}$ this means

$$\langle v^{(i)}, v^{(1)} + \dots + v^{(m)} \rangle = 1 + \frac{2}{p} = 1 + \frac{1}{\cos \frac{\pi}{m}}$$

and hence

$$\langle v^{(i)}, u \rangle = \frac{1}{m}(1 + (\cos \frac{\pi}{m})^{-1}) = \sigma$$

for all i . We can therefore apply our main result (8) and conclude

$$\Theta(C_m) \leq \frac{m}{1 + (\cos \frac{\pi}{m})^{-1}} \quad (5)$$

Notice that because of $\cos \frac{\pi}{m} < 1$ the bound (9) is better than the bound $\Theta(C_m) \leq \frac{m}{2}$ we found before. Note further $\cos \frac{\pi}{5} = \frac{\tau}{2}$, where $\tau = \frac{\sqrt{5}+1}{2}$ is the golden section. Hence for $m = 5$ we again obtain

$$\Theta(C_5) \leq \frac{5}{1 + \frac{4}{\sqrt{5}+1}} = \frac{5(\sqrt{5}+1)}{5 + \sqrt{5}} = \sqrt{5}.$$

The orthonormal representation given by this construction is, of course, precisely the "Lovasz umbrella." And what about C_7, C_9 , and the other odd cycles? By considering $\alpha(C_m^2)$, $\alpha(C_m^3)$ and other small powers the lower bound $\frac{m-1}{2} \leq \Theta(C_m)$ can certainly be increased, but for no odd $m \geq 7$ do the best known lower bounds agree with the upper bound given in (8). So, twenty years after Lovasz' which is $3.2141 \leq \Theta(C_7) \leq 3.3177$. marvelous proof of $\Theta(C_5) = \sqrt{5}$, these problems remain open and are considered very difficult — but after all we had this situation before.

The eigenvalues of C_m

Look at the adjacency matrix A of the cycle C_m . To find the eigenvalues (and eigenvectors) we use the m -th roots of unity. These are given by $1, \zeta, \zeta^2, \dots, \zeta^{m-1}$ for $\zeta = e^{\frac{2\pi i}{m}}$ — see the box on page 25. Let $\lambda = \zeta^k$ be any of these roots, then we claim that $(1, \lambda, \lambda^2, \dots, \lambda^{m-1})^T$ is an eigenvector of A to the eigenvalue $\lambda + \lambda^{-1}$. In fact, by the set-up of A we find

$$A \begin{pmatrix} 1 \\ \lambda \\ \lambda^2 \\ \vdots \\ \lambda^{m-1} \end{pmatrix} = \begin{pmatrix} \lambda + \lambda^{m-1} \\ \lambda^2 + 1 \\ \lambda^3 + \lambda \\ \vdots \\ 1 + \lambda^{m-2} \end{pmatrix} = (\lambda + \lambda^{-1}) \begin{pmatrix} 1 \\ \lambda \\ \lambda^2 \\ \vdots \\ \lambda^{m-1} \end{pmatrix}$$

Since the vectors $(1, \lambda, \dots, \lambda^{m-1})$ are independent (they form a so-called Vandermonde matrix) we conclude that for odd m

$$\begin{aligned} \zeta^k + \zeta^{-k} &= [\cos(2k\pi/m) + i \sin(2k\pi/m)] \\ &\quad + [\cos(2k\pi/m) - i \sin(2k\pi/m)] \\ &= 2 \cos(2k\pi/m) \quad (0 \leq k \leq \frac{m-1}{2}) \end{aligned}$$

are all the eigenvalues of A . Now the cosine is a decreasing function, and So

$$2 \cos\left(\frac{(m-1)\pi}{m}\right) = -2 \cos \frac{\pi}{m}$$

is the smallest eigenvalue of A .

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