

New Eigenvalue Bound for the Fractional Chromatic Number

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Joint with Krystal Guo

Eigenvalues and Chromatic Numbers

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Theorem (Hoffman 1970)

$$\chi(G) \geq 1 + \frac{\lambda_1}{|\lambda_n|}.$$

Eigenvalues and Chromatic Numbers

Elphick and Wocjan defined

$$s^+ = s^+(G) := \sum_{\lambda_i > 0} \lambda_i^2, \quad s^- = s^-(G) := \sum_{\lambda_i < 0} \lambda_i^2.$$

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Sharp for bipartite graphs, cliques K_n , Payley graph on 9 vertices,...

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These bounds are incomparable with each other. In particular, of the 11,855 graphs on 5,6,7,8 vertices which are connected and non-bipartite, Ando-Lin does better than Hoffman for 11,014 of them.

Eigenvalues and Chromatic Numbers

Conjecture (Elphick-Wocjan)

If G is a connected n -vertex graph, then

$$\min\{s^+, s^-\} \geq n - 1.$$

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So $s^- \geq n$, and a similar argument works for s^+ . □

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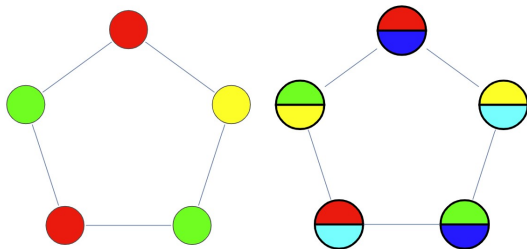
$$\chi(G) \geq 1 + \max \left\{ \frac{s^+}{s^-}, \frac{s^-}{s^+} \right\}.$$

Theorem (Guo-S. 2022)

$$\chi_f(G) \geq 1 + \max \left\{ \frac{s^+}{s^-}, \frac{s^-}{s^+} \right\}.$$

Fractional Chromatic Number

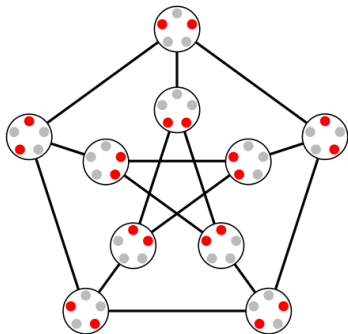
Fractional Chromatic Number



Picture from Wolfram Alpha.

Fractional Chromatic Number

The *Kneser graph* $K_{v;n}$ is the graph whose vertex set consists of n -element subsets of $[v]$ where two sets are adjacent if they are disjoint.



Picture from Wikipedia.

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$$\chi_f(G) = \min_{(v,n): K_{v,n} \in \Phi} \frac{v}{n}.$$

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Theorem

$$\chi_f(G) = \min_{(v,n): K_{v,n} \in \Phi} \frac{v}{n}.$$

This is analogous to

$$\chi(G) = \min_{K_r \in \Phi} r,$$

since an r -coloring of G is equivalent to a homomorphism $\phi : V(G) \rightarrow V(K_r)$.

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$$\frac{\lambda_{\max}(H)}{|\lambda_{\min}(H)|} \geq \max \left\{ \frac{s^+}{s^-}, \frac{s^-}{s^+} \right\}.$$

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This implies the previous result since

$$\chi_f(G) = \min_{(v,n): K_{v;n} \in \Phi} \frac{v}{n} = 1 + \min_{(v,n): K_{v;n} \in \Phi} \frac{\lambda_{\max}(K_{v;n})}{|\lambda_{\min}(K_{v;n})|}.$$

Applications

Theorem (Guo-S. 2022)

$$\chi_f(G) \geq 1 + \max \left\{ \frac{s^+}{s^-}, \frac{s^-}{s^+} \right\} = \frac{2e(G)}{\min\{s^+, s^-\}}.$$

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Corollary (Abiad-De Lima-Desai-Guo-Hogben-Madrid)

The graph $G = C_n$ satisfies $\min\{s^+, s^-\} \geq n - 1$.

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If n is even then $s^+ = s^- = n$, otherwise

$$\frac{2n}{s^\pm} \leq \chi_f(G) = \frac{2n}{n-1}.$$

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Corollary (ADDGHM)

If G is a connected unicyclic graph with cycle length $m \gg \sqrt{n}$, then $\min\{s^+, s^-\} \geq n - 1$.

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One can extend this result somewhat, but it does not hold with “edge-transitive” replaced by “vertex-transitive” (e.g. it fails for $G = K_3$ and $H = \overline{C_6}$).

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Throughout we use the Frobenius norm

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Observe that if $A = \sum \lambda_i u_i u_i^T$, then we can write $A = X - Y$ with

$$X = \sum_{i:\lambda_i>0} \lambda_i u_i u_i^T, \quad Y = - \sum_{i:\lambda_i<0} \lambda_i u_i u_i^T.$$

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With this X, Y are PSD and $\|X\|^2 = s^+$, $\|Y\|^2 = s^-$. Thus to prove $s^+/s^- \leq r$, it suffices to prove a bound of the form

$$\|X\|^2 \leq r \|Y\|^2.$$

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Lemma (Ando-Lin)

Let $\phi : V(G) \rightarrow V(H)$ be a homomorphism, let X, Y be matrices indexed by $V(G)$ such that $XY = 0$ and $X_{[u,v]} = Y_{[u,v]}$ whenever $\{u, v\} \notin E(H)$. If

$$\|X\|^2 \leq (r+1) \sum_{(u,v):\{u,v\} \notin E(H)} \|X_{[u,v]}\|^2,$$

then

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If $\{u, v\} \notin E(H)$, then $A_{[u,v]} = 0$, so $A = X - Y$ implies $X_{[u,v]} = Y_{[u,v]}$, and the condition $XY = 0$ holds if

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The result above holds if $H = K_{r+1}$.

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Hence

$$\|X\|^2 = \sum_u \|X_{[u,u]}\|^2 + \sum_{u \neq v} \|X_{[u,v]}\|^2 \leq (r+1) \sum_u \|X_{[u,u]}\|^2.$$

Proof

Lemma (Main)

If X is PSD and H is vertex and edge-transitive, then

$$\|X\|^2 \leq \left(\frac{\lambda_{\max}(H)}{|\lambda_{\min}(H)|} + 1 \right) \sum_{(u,v):\{u,v\} \notin E(H)} \|X_{[u,v]}\|^2.$$

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To make things simpler, consider the matrix Z indexed by $V(H)$ with $Z_{u,v} := \|X_{[u,v]}\|^2$.

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Claim

It suffices to show that for non-negative PSD matrices Z , we have

$$\sum Z_{u,v} \leq \left(\frac{\lambda_{\max}(H)}{|\lambda_{\min}(H)|} + 1 \right) \sum_{(u,v):\{u,v\} \notin E(H)} Z_{u,v}.$$

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Lemma

If H is edge-transitive, then there exists a set of non-empty graphs $\mathcal{H} = \{H_1, \dots, H_n\}$ on $V(H)$ satisfying the following properties:

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- 4 For every pair of edges $uv, xy \in E(H_i)$, there exists $\pi \in \text{Aut}(H)$ with $\pi(uv) = xy$.

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- 4 For every pair of edges $uv, xy \in E(H_i)$, there exists $\pi \in \text{Aut}(H)$ with $\pi(uv) = xy$.

E.g. if $H = K_{v;n}$ then this holds with H_i the graph on $\binom{[v]}{n}$ where $S \sim T$ if $|S \cap T| = n - i$.

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This matrix is PSD and non-negative with $\sum \bar{Z}_{u,v} = \sum Z_{u,v}$,
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and manipulations together with the previous inequality gives the result.

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- 4 Finally, we proved this using Rayleigh quotient arguments.

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Conjecture (Elphick-Wocjan)

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If G is a connected n -vertex graph, then

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Given a family of graphs \mathcal{H} , define

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