



Graphs Homomorphisms and Spectral Conditions

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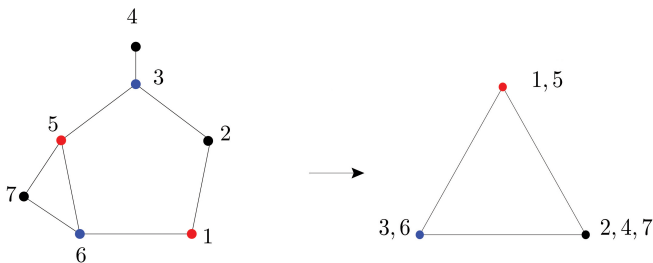


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- We order the eigenvalues of A_G , L_G , and T_G , respectively, as follows

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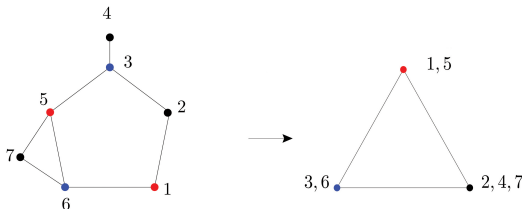
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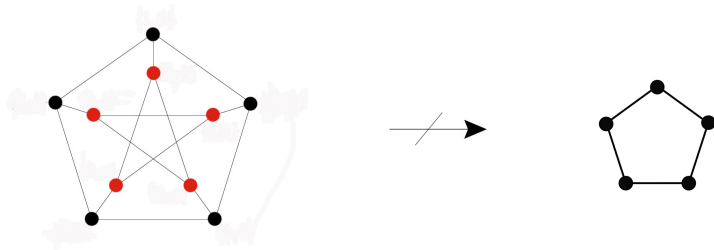
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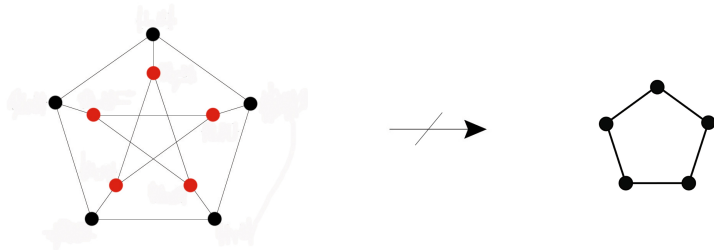
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- ▶ Circular chromatic number of the Petersen graph is 3.
- ▶ Let G be a graph with n vertices. If $\bar{d} = \frac{2|E(G)|}{n}$ then,

$$\chi_f(G) \geq \frac{\lambda_n^G}{\lambda_n^G - \bar{d}}.$$



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- ▶ **A Fisher-type inequality for G -designs:** for a G -design on v points with $|V(G)| = n$ we have,

$$b < v \quad \Rightarrow \quad \max\left(\frac{v}{\lambda_n^G}, \frac{v-2}{\mu_{n-1}^G}\right) \leq \frac{r}{\lambda} \leq \min\left(\frac{v}{\lambda_2^G}, \frac{v-2}{\mu_1^G}\right).$$



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Professor J. Nešetřil & O. Serra
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Thank You!