

Flag Algebras on Limits of Graph Sequences

Today, we finally define some algebras \mathcal{A} and \mathcal{A}^σ . We try to develop an analogue for $P(\cdot, G)$ in the limit. We start by defining \mathcal{A} . We denote the set all graphs up to isomorphism by \mathcal{F} and the set of all graphs on exactly ℓ vertices by \mathcal{F}_ℓ . Let $\mathbb{R}\mathcal{F}$ be the set of all formal linear combinations of graphs in \mathcal{F} . This $\mathbb{R}\mathcal{F}$ will be used to derive \mathcal{A} . In order to do that, we need to somehow define addition and multiplication on $\mathbb{R}\mathcal{F}$.

1: How to *obviously* define the addition of elements in $\mathbb{R}\mathcal{F}$ and the multiplication by a real number?

Solution: For addition of $a, b \in \mathbb{R}\mathcal{F}$, just add the coefficients together.

Recall we used linear combinations of graphs last time, but they had some extra interpretation using $P(\cdot, G)$. Our goal is to eventually have a similar interpretation for $\mathbb{R}\mathcal{F}$ and \mathcal{A} .

2: Let $F_1, F_2 \in \mathcal{F}$. Can you write $F_1 \cdot F_2$ as an element in $\mathbb{R}\mathcal{F}$ inspired by our efforts last time?

Solution:

$$F_1 \cdot F_2 = \sum_{F \in \mathcal{F}_\ell} P(F_1, F_2; F)F,$$

where $v(F_1) + v(F_2) = \ell$. Notice, there is NO $+o(1)$.

3: How to “define” a multiplication of $a, b \in \mathbb{R}\mathcal{F}$?

Solution: One could say that having $a, b \in \mathbb{R}\mathcal{F}$ it would be possible to do entry wise product from the previous exercise. A small trouble is that $F_1 \cdot F_2$ can be written in many ways as a linear combination so this is not a great multiplication.

In addition to the “uniqueness trouble” in multiplication for $\mathbb{R}\mathcal{F}$, we would also like to enforce identities such as

$$\begin{array}{c} \bullet \\ | \\ \bullet \end{array} = \frac{1}{3} \cdot \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \end{array} + \frac{2}{3} \cdot \begin{array}{c} \bullet \text{---} \bullet \\ \diagdown \bullet \end{array} + \begin{array}{c} \bullet \text{---} \bullet \\ \diagup \bullet \end{array} \quad (1)$$

We do this by factorizing $\mathbb{R}\mathcal{F}$ by a suitable subspace \mathcal{K} . Let \mathcal{K} be a linear subspace of $\mathbb{R}\mathcal{F}$ generated by all elements of the form

$$F - \sum_{F' \in \mathcal{F}_\ell} P(F, F')F', \quad (2)$$

where $F \in \mathcal{F}$ and $v(F) \leq \ell$.

4: What is the value of (2) in our previous interpretation using $P(\cdot, G)$?

Solution: 0

Finally, the algebra \mathcal{A} is $\mathbb{R}\mathcal{F}$ factorized by \mathcal{K} . Notice that $a, b \in \mathbb{R}\mathcal{F}$ belong to the same equivalence class iff $a = b + c$ for some $c \in \mathcal{K}$.

This also fixes the uniqueness in multiplication (which we are not proving here). Personal note: I just think of it all as $\mathbb{R}\mathcal{F}$, where some additional equations hold. Not as the formal \mathcal{A} . An analogue is \mathbb{Q} for \mathcal{A} and $\mathbb{Z} \times \mathbb{Z}$ for $\mathbb{R}\mathcal{F}$. We get $\frac{1}{2} = \frac{2}{4} = \frac{4}{8} = \frac{3}{6} = \dots$.

Now we try to make \mathcal{A} somehow useful by providing a similar interpretation as $P(?, G)$ from last time. Notice we do not have $+o(1)$ in multiplication, so we cannot really use $P(?, G)$ but we will use limits of $P(?, G)$.

Definition: A sequence of graphs $(G_n)_{n \in \mathbb{N}}$ is *convergent* if for every finite graph H , $\lim_{n \rightarrow \infty} P(H, G_n)$ exists.

Notice that the *limit of convergent sequence* can be viewed as vector of numbers in $[0, 1]$ indexed by all finite graphs, i.e., a function $\mathcal{F} \rightarrow [0, 1]$.

5: Show that $(K_n)_{n \in \mathbb{N}}$ is convergent, where K_n is the complete graph on n vertices. What is the limit?

Solution: The limit is quite simple. If H is a complete graph, then the limit of $p(H, K_n)$ is 1, otherwise it is 0.

6: Show that $(P_n)_{n \in \mathbb{N}}$ is convergent, where P_n is a path on n vertices. What is the limit?

Solution: The limit is not difficult. For any H that contains an edge, $p(H, P_n)$ is 0 and if H has no edges, then $p(H, P_n) = 1$.

Now we get to the promised replacement of $P(., G)$. Let $Hom(\mathcal{A}, \mathbb{R})$ be the set of all homomorphisms from \mathcal{A} to \mathbb{R} . So for any $\phi \in Hom(\mathcal{A}, \mathbb{R})$ and $a, b \in \mathcal{A}$, we have that $\phi(a + b) = \phi(a) + \phi(b)$ and $\phi(a \cdot b) = \phi(a) \cdot \phi(b)$. Since $p(H, G) \in [0, 1]$, we consider only $Hom^+(\mathcal{A}, \mathbb{R})$, which are homomorphism ϕ such that $\phi(H) \geq 0$ for all $H \in \mathcal{F}$. Notice $\phi(\emptyset) = 1$.

It can be proved, that $Hom^+(\mathcal{A}, \mathbb{R})$ corresponds exactly to the convergent sequences of graphs.

Theorem (special for graphs) Let $a \in \mathbb{R}\mathcal{F}$. $\phi(a) = 0$ for all $\phi \in Hom^+(\mathcal{A}, \mathbb{R})$ iff $a \in \mathcal{K}$.

7: Recall that previously we stated Mantel's theorem as $\left| \begin{smallmatrix} \bullet \\ | \\ \bullet \end{smallmatrix} \right| \leq \frac{1}{2}$ if $\begin{smallmatrix} \bullet & & \bullet \\ & \blacktriangle & \\ \bullet & & \bullet \end{smallmatrix} = 0$. What does it formally translate in our new interpretation using $Hom^+(\mathcal{A}, \mathbb{R})$?

Solution: It is saying that for every $\phi \in Hom^+(\mathcal{A}, \mathbb{R})$ holds that

$$\text{if } \phi \left(\begin{smallmatrix} \bullet & & \bullet \\ & \blacktriangle & \\ \bullet & & \bullet \end{smallmatrix} \right) = 0 \text{ then } \phi \left(\begin{smallmatrix} \bullet \\ | \\ \bullet \end{smallmatrix} \right) \leq \frac{1}{2}.$$

We could also develop the whole flag algebras with \mathcal{F} being *triangle-free* graphs instead of all graphs. Let's call such algebras \mathcal{A}_{∇} . Then the statement would be that for all

$$\phi \in Hom^+(\mathcal{A}_{\nabla}, \mathbb{R}) \text{ holds } \phi \left(\begin{smallmatrix} \bullet \\ | \\ \bullet \end{smallmatrix} \right) \leq \frac{1}{2}.$$

We are interested in finding $a \in \mathcal{A}$ such that for EVERY $\phi \in Hom^+(\mathcal{A}, \mathbb{R})$ we have $\phi(a) \geq 0$. If this happens, we just write $a \geq 0$. Note that Hatami and Norin showed that determining if $a \geq 0$ is not algorithmically decidable.

Example of such a was given last time

$$0 \leq 3 \cdot \begin{smallmatrix} \bullet & & \bullet \\ & \bullet & \\ \bullet & & \bullet \end{smallmatrix} - \begin{smallmatrix} \bullet & & \bullet \\ & \text{---} & \\ \bullet & & \bullet \end{smallmatrix} - \begin{smallmatrix} \bullet & & \bullet \\ & \text{---} & \\ & & \bullet \end{smallmatrix} + 3 \cdot \begin{smallmatrix} \bullet & & \bullet \\ & \blacktriangle & \\ \bullet & & \bullet \end{smallmatrix}. \tag{3}$$

Now we explore a way how to obtain various $a \in \mathcal{A}$ such that $a \geq 0$. Last time we also considered labeled graphs with some fixed embedding of a graph σ . By following the same path as for the unlabeled case, we can define \mathcal{A}^σ . Let $\sigma \in \mathcal{F}$ be graph with vertices labeled by $1, \dots, v(\sigma)$.

- \mathcal{F}^σ is the set of all graphs each containing a fixed induced labeled copy of σ .
- \mathcal{F}_ℓ^σ is the set of elements in \mathcal{F}^σ on ℓ vertices.

- $\mathbb{R}\mathcal{F}^\sigma$ are all formal linear combinations of elements in \mathcal{F}^σ .
- \mathcal{K}^σ is a linear subspace generated by $F - \sum_{F' \in \mathcal{F}_\ell^\sigma} P(F, F')F'$, where $F \in \mathcal{F}^\sigma$ and $v(F) \leq \ell$.
- \mathcal{A}^σ is $\mathbb{R}\mathcal{F}^\sigma$ factorized by \mathcal{K}^σ .
- addition in \mathcal{A}^σ comes from $\mathbb{R}\mathcal{F}^\sigma$
- multiplication in \mathcal{A}^σ defined as an extension of $F_1 \cdot F_2 = \sum_{F \in \mathcal{F}_\ell^\sigma} P(F_1, F_2; F)F$, where $F_1, F_2 \in \mathcal{F}$ and $v(F_1) + v(F_2) - v(\sigma) = \ell$.
- $Hom^+(\mathcal{A}^\sigma, \mathbb{R})$ is the set of homomorphism ϕ^σ , where $\phi^\sigma(F) \geq 0$ for all $F \in \mathcal{F}^\sigma$.
- $F \in \mathcal{F}^\sigma$ is called a σ -flag.

8: What is 1 for multiplication in \mathcal{A}^σ ? Let $a, b \in \mathcal{A}^\sigma$ and $ab = a$. What is b ?

Solution: $b = \sigma$, it also means the equivalence class of σ in \mathcal{A} .

Let σ be fixed. We will describe an interaction between $Hom^+(\mathcal{A}, \mathbb{R})$ and $Hom^+(\mathcal{A}^\sigma, \mathbb{R})$. First, we fix some $\phi \in Hom^+(\mathcal{A}, \mathbb{R})$. This ϕ corresponds to some converging graph sequence $(G_n)_{n \in \mathbb{N}}$. In order to make some connection with \mathcal{A}^σ , we would need a sequence of labeled graphs in \mathcal{F}^σ . However, instead of creating a sequence of labeled graphs, we create a sequence of probability distributions generated by picking different copies of σ . For each n separately, we do the following. Pick uniformly at random a labeled copy of σ in G_n , denote result by G_n^σ . This allows us to evaluate $P(F, G_n^\sigma)$ for all $F \in \mathcal{A}^\sigma$. By the random choice of σ in G_n , we get some probability distribution $\mathbf{P}_{G_n}^\sigma$ on the functions $P(\cdot, G_n^\sigma)$. These $\mathbf{P}_{G_n}^\sigma$ then weakly converge to a (unique) probability distribution \mathbf{P}_ϕ^σ on $\phi^\sigma \in Hom^+(\mathcal{A}^\sigma, \mathbb{R})$. We will get back to \mathbf{P}_ϕ^σ in a bit.

Let σ be fixed. Our next goal is to define an *averaging* operator that can translate expressions from \mathcal{A}^σ to \mathcal{A} . Let $F \in \mathcal{F}^\sigma$. Since F has a labeled copy of σ , we can view F and (G, θ) , where $G \in \mathcal{F}$ is an unlabeled copy of F and θ is a function from $1, \dots, v(\sigma)$ to $V(G)$ that identifies the labeled copy of σ . Let θ' be an injective function from $1, \dots, v(\sigma)$ to $V(G)$ chosen uniformly at random. Define $q_\theta(F)$ to be the probability that (G, θ') is isomorphic to F .

9: Calculate $q_\theta \left(\begin{array}{c} \bullet \quad \bullet \\ \diagdown \quad \diagup \\ 1 \quad \square \end{array} \right) = \frac{1}{3} \qquad q_\theta \left(\begin{array}{c} \bullet \quad \bullet \\ \diagdown \quad \diagup \\ 1 \quad \square \end{array} \right) = \frac{2}{3}$

Finally, for any type σ and $F \in \mathcal{F}^\sigma$, we define $[[\cdot]]_\sigma : \mathcal{F}^\sigma \rightarrow \mathbb{R}\mathcal{F}$ as

$$[[F]]_\sigma = q_\theta(F) \cdot G,$$

where G is an unlabeled F . Its linear extension is then the *averaging operator* $[[\cdot]]_\sigma : \mathbb{R}\mathcal{F}^\sigma \rightarrow \mathbb{R}\mathcal{F}$. It is linear mapping, *not* a homomorphism.

10: Calculate

$$\left[\left[\begin{array}{c} 2 \quad \bullet \quad \bullet \\ \diagdown \quad \diagup \\ \bullet \quad \bullet \\ \diagdown \quad \diagup \\ 1 \quad \square \end{array} \right] \right]_\sigma = \frac{6}{20} \begin{array}{c} \bullet \quad \bullet \\ \diagdown \quad \diagup \\ \bullet \quad \bullet \\ \diagdown \quad \diagup \\ \bullet \quad \bullet \end{array}$$

A *crucial feature* is that if $a \in \mathcal{A}^\sigma$ and $\phi \in Hom^+(\mathcal{A}, \mathbb{R})$, then $\phi([[a]]_\sigma)$ is closely related to the expected value of $\phi^\sigma(a)$, where ϕ^σ is chosen according to \mathbf{P}_ϕ^σ . Specifically,

$$\phi([[\sigma]]_\sigma) \cdot \mathbb{E}_{\mathbf{P}_\phi^\sigma} [\phi^\sigma(a)] = \phi([[a]]_\sigma). \tag{4}$$

This can be viewed as an analogue of $P(B) \cdot P(A|B) = P(A \wedge B)$. Bonus: For a given $\phi \in Hom^+(\mathcal{A}, \mathbb{R})$, it can be directly shown there exists a unique probability distribution \mathbf{P}_ϕ^σ satisfying (4). This distribution is indeed the weak limit of $(\mathbf{P}_{G_n}^\sigma)_{n \in \mathbb{N}}$ obtained from any sequence $(G_n)_{n \in \mathbb{N}}$ that converges to ϕ . Equation (4) is especially

useful when $\phi^\sigma(a) \geq 0$ with probability one. In this case, (4) yields $\phi(\llbracket a \rrbracket_\sigma) \geq 0$. In particular, we will use that for any $\phi \in \text{Hom}^+(\mathcal{A}, \mathbb{R})$ and any $a \in \mathcal{A}^\sigma$

$$\phi(\llbracket a \cdot a \rrbracket_\sigma) \geq 0.$$

In the simplified notation, we could just write $\llbracket a \cdot a \rrbracket_\sigma \geq 0$. Note, the inequality $\llbracket a^2 \rrbracket_\sigma \geq 0$ can be also obtained from the following flag version of Cauchy-Schwarz inequality

$$\llbracket a^2 \rrbracket_\sigma \cdot \llbracket b^2 \rrbracket_\sigma \geq \llbracket ab \rrbracket_\sigma^2. \quad (5)$$

We try another proof of Mantel's theorem. Recall, we consider only triangle-free graphs. We denote the type of one labeled vertex simply by 1.

11: Complete the following inequality into an inequality between unlabeled triangle-free graphs (do not multiply unlabeled graphs)

$$\begin{array}{c} \bullet \\ | \\ \bullet \end{array}^2 = \left[\begin{array}{c} \bullet \\ | \\ \bullet \\ \text{1} \end{array} \right]_1^2 \leq \left[\begin{array}{c} \bullet \\ | \\ \bullet \\ \text{1} \end{array} \right]_1^2 = \left[\begin{array}{c} \bullet \\ / \quad \backslash \\ \bullet \quad \bullet \\ \text{1} \end{array} \right]_1 = \frac{1}{3} \begin{array}{c} \bullet \\ | \\ \bullet \end{array}$$

12: Use (1) to show $\frac{1}{3} \begin{array}{c} \bullet \\ / \quad \backslash \\ \bullet \quad \bullet \end{array} \leq \frac{1}{2} \begin{array}{c} \bullet \\ | \\ \bullet \end{array}$.

Solution:

$$\begin{array}{c} \bullet \\ | \\ \bullet \end{array} = \frac{1}{3} \begin{array}{c} \bullet \\ / \quad \backslash \\ \bullet \quad \bullet \end{array} + \frac{2}{3} \begin{array}{c} \bullet \\ / \quad \backslash \\ \bullet \quad \bullet \end{array} \geq \frac{2}{3} \begin{array}{c} \bullet \\ / \quad \backslash \\ \bullet \quad \bullet \end{array}$$

13: Combine the previous two questions to show Mantel's theorem.

Solution:

$$\begin{array}{c} \bullet \\ | \\ \bullet \end{array}^2 \leq \frac{1}{3} \begin{array}{c} \bullet \\ / \quad \backslash \\ \bullet \quad \bullet \end{array} \leq \frac{1}{2} \begin{array}{c} \bullet \\ | \\ \bullet \end{array} \quad \text{hence} \quad \begin{array}{c} \bullet \\ | \\ \bullet \end{array} \leq \frac{1}{2}$$

Generating $a \geq 0$ using \mathcal{A}^σ and $\llbracket \cdot \rrbracket_\sigma$

Recall that a symmetric real matrix $M \in \mathbb{R}^{n \times n}$ is *positive semidefinite* if $x^T M x \geq 0$ for all $x \in \mathbb{R}^n$. Denoted by $M \succcurlyeq 0$.

14: How to check if $M \in \mathbb{R}^{n \times n}$ is positive semidefinite?

Solution: All eigenvalues of M are non-negative. $M = U^T U$ for some matrix U . All principal minors are non-negative.

Main observation Let σ be fixed. If $M \succcurlyeq 0$ and X is a vector $(\mathcal{F}_\ell^\sigma)^n$, then for any ϕ^σ holds

$$0 \leq \phi^\sigma(X^T M X) \quad \text{hence} \quad \llbracket X^T M X \rrbracket_\sigma \geq 0.$$

Next we show, that using a semidefinite matrices is analogous to using sum of squares.

Let $a \in \mathcal{A}^\sigma$ be a linear combination of σ -flags. Let X be vector of flags in a and let v be a vector of coefficients.

15: Show that $a^2 = X^T M X$ for some matrix $M \succcurlyeq 0$. Hint: $M = v v^T$.

Solution: Just expand it: $X^T M X = X^T v v^T X = X^T v \cdot (X^T v) = a \cdot a = a^2$.

16: In light of the previous exercise, what is $X^T M X$ for general $M \succeq 0$.

Solution: Since we can decompose every $M \in \mathbb{R}^{m \times m}$ into $U^T U$, we can decompose $X^T M X$ into sum of (at most) m squares.

17: Let $M \in \mathbb{R}^{2 \times 2}$, σ be one labeled vertex and X be a vector containing all labeled graphs in \mathcal{F}_2^σ . Evaluate $X^T M X$, which is can be also written as: (now we are NOT triangle-free)

$$\begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \begin{pmatrix} a & c \\ c & b \end{pmatrix} \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix}^T$$

Solution:

$$\begin{aligned} & \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \begin{pmatrix} a & c \\ c & b \end{pmatrix} \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix}^T = a \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix}^2 + 2c \begin{pmatrix} \bullet & \bullet & \bullet \\ | & | & | \\ 1 \blacksquare & 1 \blacksquare & 1 \blacksquare \end{pmatrix} + b \begin{pmatrix} \bullet & \bullet & \bullet \\ | & | & | \\ 1 \blacksquare & 1 \blacksquare & 1 \blacksquare \end{pmatrix} \\ & = a \left(\begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} + \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right) + 2c \cdot \frac{1}{2} \left(\begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} + \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right) + b \left(\begin{pmatrix} \bullet & \bullet & \bullet \\ | & | & | \\ 1 \blacksquare & 1 \blacksquare & 1 \blacksquare \end{pmatrix} + \begin{pmatrix} \bullet & \bullet & \bullet \\ | & | & | \\ 1 \blacksquare & 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right) \\ & = a \left(\begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} + \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right) + c \left(\begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} + \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right) + b \left(\begin{pmatrix} \bullet & \bullet & \bullet \\ | & | & | \\ 1 \blacksquare & 1 \blacksquare & 1 \blacksquare \end{pmatrix} + \begin{pmatrix} \bullet & \bullet & \bullet \\ | & | & | \\ 1 \blacksquare & 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right) \end{aligned}$$

18: Use the solution of the previous exercise to evaluate

$$\left[\left[\begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \begin{pmatrix} a & c \\ c & b \end{pmatrix} \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix}^T \right] \right]_\sigma$$

Solution:

$$\begin{aligned} & \left[\left[\begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \begin{pmatrix} a & c \\ c & b \end{pmatrix} \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix}^T \right] \right]_\sigma \\ & = \left[a \left(\begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} + \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right) + c \left(\begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} + \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right) + b \left(\begin{pmatrix} \bullet & \bullet & \bullet \\ | & | & | \\ 1 \blacksquare & 1 \blacksquare & 1 \blacksquare \end{pmatrix} + \begin{pmatrix} \bullet & \bullet & \bullet \\ | & | & | \\ 1 \blacksquare & 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right) \right]_\sigma \\ & = a \left(\left[\begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right]_\sigma + \left[\begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right]_\sigma \right) + c \left(\left[\begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right]_\sigma + \left[\begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right]_\sigma \right) + b \left(\left[\begin{pmatrix} \bullet & \bullet & \bullet \\ | & | & | \\ 1 \blacksquare & 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right]_\sigma + \left[\begin{pmatrix} \bullet & \bullet & \bullet \\ | & | & | \\ 1 \blacksquare & 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right]_\sigma \right) \\ & = a \left(\frac{1}{3} \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} + \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right) + c \left(\frac{2}{3} \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} + \frac{2}{3} \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right) + b \left(\begin{pmatrix} \bullet & \bullet & \bullet \\ | & | & | \\ 1 \blacksquare & 1 \blacksquare & 1 \blacksquare \end{pmatrix} + \frac{1}{3} \begin{pmatrix} \bullet & \bullet & \bullet \\ | & | & | \\ 1 \blacksquare & 1 \blacksquare & 1 \blacksquare \end{pmatrix} \right) \\ & = b \cdot \begin{pmatrix} \bullet & \bullet & \bullet \\ | & | & | \\ 1 \blacksquare & 1 \blacksquare & 1 \blacksquare \end{pmatrix} + \frac{b+2c}{3} \cdot \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} + \frac{a+2c}{3} \cdot \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} + a \cdot \begin{pmatrix} \bullet & \bullet \\ | & | \\ 1 \blacksquare & 1 \blacksquare \end{pmatrix} \end{aligned}$$

19: Recall that $\llbracket X^T M X \rrbracket_\sigma \geq 0$ is true for ANY $M \succeq 0$. Prove that inequality (3) is valid by finding a suitable M . Hint: Use the last solution.

Solution: Take

$$M = \begin{pmatrix} a & c \\ c & b \end{pmatrix} = \begin{pmatrix} 3 & -3 \\ -3 & 3 \end{pmatrix}$$

This matrix has eigenvalues 0 and 6 with eigenvectors $(1, 1)$ and $(1, -1)$. Plugging it into the last line of previous question gives exactly (3).

It is also possible to see that M is positive semidefinite since all principal minors are non-negative.

Bonus: A sufficient condition for M being positive semidefinite is that leading principal minors are positive except the last one, which is non-negative. Our matrix satisfies this too. It is easier to verify.

20: We try to prove Mantel's theorem again. Start by summing

$$0 \leq b \cdot \begin{array}{c} \bullet \\ | \\ \bullet \end{array} + \frac{b+2c}{3} \cdot \begin{array}{c} \bullet \text{---} \bullet \\ | \\ \bullet \end{array} + \frac{a+2c}{3} \cdot \begin{array}{c} \bullet \text{---} \bullet \\ \diagdown \\ \bullet \end{array} + a \cdot \begin{array}{c} \bullet \\ \diagup \quad \diagdown \\ \bullet \end{array}$$

and

$$\begin{array}{c} \bullet \\ | \\ \bullet \end{array} = \frac{1}{3} \cdot \begin{array}{c} \bullet \text{---} \bullet \\ | \\ \bullet \end{array} + \frac{2}{3} \cdot \begin{array}{c} \bullet \text{---} \bullet \\ \diagdown \\ \bullet \end{array} + \begin{array}{c} \bullet \\ \diagup \quad \diagdown \\ \bullet \end{array}.$$

Use $\begin{array}{c} \bullet \\ \diagup \quad \diagdown \\ \bullet \end{array} = 0$ and try to get $\begin{array}{c} \bullet \\ | \\ \bullet \end{array} \leq \text{function}(a, b, c)$. *Hint: function containing max.* Can you guess a, b, c ?

Solution: Since $\begin{array}{c} \bullet \\ \diagup \quad \diagdown \\ \bullet \end{array} = 0$, we just ignore it in the calculation.

$$\begin{array}{c} \bullet \\ | \\ \bullet \end{array} = 0 \cdot \begin{array}{c} \bullet \\ | \\ \bullet \end{array} + \frac{1}{3} \cdot \begin{array}{c} \bullet \text{---} \bullet \\ | \\ \bullet \end{array} + \frac{2}{3} \cdot \begin{array}{c} \bullet \text{---} \bullet \\ \diagdown \\ \bullet \end{array}$$

$$0 \leq b \cdot \begin{array}{c} \bullet \\ | \\ \bullet \end{array} + \frac{b+2c}{3} \cdot \begin{array}{c} \bullet \text{---} \bullet \\ | \\ \bullet \end{array} + \frac{a+2c}{3} \cdot \begin{array}{c} \bullet \text{---} \bullet \\ \diagdown \\ \bullet \end{array}$$

Summing together we get

$$\begin{array}{c} \bullet \\ | \\ \bullet \end{array} = b \cdot \begin{array}{c} \bullet \\ | \\ \bullet \end{array} + \frac{1+b+2c}{3} \cdot \begin{array}{c} \bullet \text{---} \bullet \\ | \\ \bullet \end{array} + \frac{2+a+2c}{3} \cdot \begin{array}{c} \bullet \text{---} \bullet \\ \diagdown \\ \bullet \end{array}$$

$$\leq \max \left\{ b, \frac{1+b+2c}{3}, \frac{2+a+2c}{3} \right\} \cdot \left(\begin{array}{c} \bullet \\ | \\ \bullet \end{array} + \begin{array}{c} \bullet \text{---} \bullet \\ | \\ \bullet \end{array} + \begin{array}{c} \bullet \text{---} \bullet \\ \diagdown \\ \bullet \end{array} \right)$$

$$= \max \left\{ b, \frac{1+b+2c}{3}, \frac{2+a+2c}{3} \right\}$$

Since we want the upper bound to be $\leq 1/2$, we can go with $b = 1/2$. To make the whole thing fit, we put also $a = 1/2$ and $c = -1/2$. Then we get

$$\begin{array}{c} \bullet \\ | \\ \bullet \end{array} \leq \max \left\{ \frac{1}{2}, \frac{1}{6}, \frac{1}{2} \right\} = \frac{1}{2}.$$

Notice that a part of the last solution for Mantel's theorem can be stated as

$$t \leq \max \left\{ b, \frac{1+b+2c}{3}, \frac{2+a+2c}{3} \right\}$$

while a, b, c had to form a positive semidefinite matrix. It is not completely obvious how to guess a, b, c . This can be written as a semidefinite program. One could think of it as a linear program with a bonus constraint that the variables together form a positive semidefinite matrix. A writeup of the program follows.

$$(SDP) \begin{cases} \text{Minimize} & t \\ \text{subject to} & a \leq t \\ & \frac{1+a+2c}{3} \leq t \\ & \frac{2+b+2c}{3} \leq t \\ & \begin{pmatrix} a & c \\ c & b \end{pmatrix} \succeq 0 \\ & t \geq 0 \end{cases}$$

(*SDP*) can be solved on computers using freely available software CSDP or SDPA. What many flag algebra applications do is set up a semidefinite program and try to solve it. We will explore semidefinite programming next time.

21: What constraints must be satisfied by a, b, c to guarantee that

$$\begin{pmatrix} a & c \\ c & b \end{pmatrix} \succeq 0$$

Solution: Positive semidefinite matrix must have entries in the diagonal ≥ 0 . Hence $a, b \geq 0$. Then also the determinant must be non-negative. So $ab - c^2 \geq 0$. Notice that this is giving some kind of a quadratic constraint on c .

22: Calculate the following product as a linear combination of graphs on 3 vertices.

$$K_1 \cdot \begin{array}{c} \bullet \\ | \\ \bullet \\ | \\ \bullet \end{array} = 0 \cdot \begin{array}{c} \bullet \\ \bullet \\ \bullet \end{array} + \frac{1}{3} \cdot \begin{array}{c} \bullet \text{---} \bullet \\ | \\ \bullet \end{array} + \frac{2}{3} \cdot \begin{array}{c} \bullet \text{---} \bullet \\ \diagdown \quad \diagup \\ \bullet \end{array} + 1 \cdot \begin{array}{c} \bullet \text{---} \bullet \\ \diagdown \quad \diagup \\ \bullet \end{array}$$

Solution: Notice that that the right-hand side is *identical* to (1).

23: Let H be a fixed graph on ℓ vertices. Calculate the following product as a linear combination of graphs on $\ell + 1$ vertices.

$$K_1 \cdot H = \sum_{F \in \mathcal{F}_{\ell+1}} P(H, F) \cdot F$$

Solution: Recall that

$$H = \sum_{F \in \mathcal{F}_{\ell+1}} P(H, F) \cdot F$$

24: Let H be a fixed graph and \mathcal{F} be a class of graphs closed under taking subgraphs. For example, triangle-free graphs. Show that

$$\lim_{n \rightarrow \infty} \max\{P(H, G) : G \in \mathcal{F}_\ell\}$$

exists.

Solution: We show that $\max\{P(H, G) : G \in \mathcal{F}_\ell\}$ is monotone non-increasing. Suppose that $\max\{P(H, G) : G \in \mathcal{F}_\ell\} = a \in [0, 1]$. Suppose for contradiction that there exists $G' \in \mathcal{F}_{\ell'}$, where $\ell' > \ell$ and $P(H, G') > a$. Then

$$a < P(H, G') = \sum_{F \in \mathcal{F}_\ell} P(H, F) \cdot P(F, G') \leq \sum_{F \in \mathcal{F}_\ell} a \cdot P(F, G') = a,$$

which is a contradiction.

25: Use (5), i.e., $\llbracket a^2 \rrbracket_\sigma \cdot \llbracket b^2 \rrbracket_\sigma \geq \llbracket ab \rrbracket_\sigma^2$, to show

$$\llbracket a^2 \rrbracket_\sigma \geq \frac{\llbracket a \rrbracket_\sigma^2}{\llbracket \sigma \rrbracket_\sigma}.$$

Solution: We use $b = \sigma$ and get

$$\llbracket a^2 \rrbracket_\sigma \cdot \llbracket \sigma \rrbracket_\sigma = \llbracket a^2 \rrbracket_\sigma \cdot \llbracket \sigma^2 \rrbracket_\sigma \geq \llbracket a\sigma \rrbracket_\sigma^2 = \llbracket a \rrbracket_\sigma^2.$$

26: Let $(G_n)_{n \in \mathbb{N}}$ be a convergent sequence, where G_n is disjoint union of a clique on n vertices and n isolated vertices ($v(G_n) = 2n$). Denote the corresponding homomorphism by ϕ .

- Show that $(G_n)_{n \in \mathbb{N}}$ is indeed convergent.

- Let σ be $\begin{array}{c} 2 \\ | \\ 1 \end{array}$. Determine \mathbf{P}_ϕ^σ .

- Compute $\mathbb{E}_{\mathbf{P}_\phi^\sigma} \left[\phi^\sigma \left(\begin{array}{c} 2 \text{ } \blacksquare \text{ } \bullet \\ \diagdown \quad \diagup \\ 1 \text{ } \blacksquare \end{array} \right) \right]$, $\phi \left(\left[\begin{array}{c} 2 \text{ } \blacksquare \text{ } \bullet \\ \diagdown \quad \diagup \\ 1 \text{ } \blacksquare \end{array} \right]_\sigma \right)$, and $\phi(\llbracket \sigma \rrbracket_\sigma)$. Compare with (4), which states

$$\phi(\llbracket \sigma \rrbracket_\sigma) \cdot \mathbb{E}_{\mathbf{P}_\phi^\sigma} [\phi^\sigma(a)] = \phi(\llbracket a \rrbracket_\sigma).$$

27: The proof of Mantel's theorem can be written as just one square instead of the whole positive semidefinite matrix. Can you find how?