

# Flag Algebras

An informal introduction

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Today, our goal is to be able to understand the following:

$$\begin{aligned}
 0 &\leq \left[ \left[ \left( \begin{array}{c} \bullet \\ \circ \end{array} - \begin{array}{c} \bullet \\ \circ \end{array} \right)^2 \right] \right] \\
 &= \left[ \left[ \begin{array}{c} \bullet^2 \\ \circ \end{array} - 2 \begin{array}{c} \bullet \\ \circ \times \bullet \\ \circ \end{array} + \begin{array}{c} \bullet^2 \\ \circ \end{array} \right] \right] \\
 &= \left[ \left[ \left( \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \begin{array}{c} \bullet \bullet \\ \circ \end{array} \right) - 2 \left( \frac{1}{2} \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \frac{1}{2} \begin{array}{c} \bullet \bullet \\ \circ \end{array} \right) + \left( \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \begin{array}{c} \bullet \bullet \\ \circ \end{array} \right) \right] \right] \\
 &= \left[ \left[ \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \begin{array}{c} \bullet \bullet \\ \circ \end{array} - \begin{array}{c} \bullet \bullet \\ \circ \end{array} - \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \begin{array}{c} \bullet \bullet \\ \circ \end{array} \right] \right] \\
 &= \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \frac{1}{3} \begin{array}{c} \bullet \bullet \\ \circ \end{array} - \frac{2}{3} \begin{array}{c} \bullet \bullet \\ \circ \end{array} - \frac{2}{3} \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \frac{1}{3} \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \begin{array}{c} \bullet \bullet \\ \circ \end{array} \\
 &= \begin{array}{c} \bullet \bullet \\ \circ \end{array} - \frac{1}{3} \begin{array}{c} \bullet \bullet \\ \circ \end{array} - \frac{1}{3} \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \begin{array}{c} \bullet \bullet \\ \circ \end{array} \\
 \frac{1}{3} \left( \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \begin{array}{c} \bullet \bullet \\ \circ \end{array} \right) &\leq \frac{4}{3} \left( \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \begin{array}{c} \bullet \bullet \\ \circ \end{array} \right) \\
 \frac{1}{4} &\leq \begin{array}{c} \bullet \bullet \\ \circ \end{array} + \begin{array}{c} \bullet \bullet \\ \circ \end{array}
 \end{aligned}$$

Let's get started!

## Unlabeled Flags

**Definition 1.** Given a subset of vertices  $S \subseteq V(G)$  in graph  $G$ , the subgraph *induced* in  $G$  by  $S$  is the graph with vertex set  $S$  and edge set  $\{\{u, v\} \mid u, v \in S, \{u, v\} \in G\}$ .

**Definition 2.** The *density* of graph  $F$  in graph  $G$ , denoted  $p(F; G)$ , is the probability that the subgraph induced in  $G$  by  $|V(F)|$  random vertices is  $F$ .

We also denote by  $p(F_1, F_2; G)$  the probability that  $|V(F_1)|$  random vertices and  $|V(F_2)|$  random remaining vertices in  $G$  induce  $F_1$  and  $F_2$ , respectively.

Unlabeled flags are graphs equipped with the following operations:

–  $\bullet = 1$ .

– Vector space operations and axioms, e.g., we can write  $\frac{1}{3} \begin{matrix} \bullet & \bullet \\ & \diagdown \diagup \\ & \bullet \end{matrix} - \frac{1}{2} \begin{matrix} \bullet \\ | \\ \bullet \end{matrix}$ .

– We can multiply flags, as follows:

$$F_1 \times F_2 = \sum_{|V(F)|=|V(F_1)|+|V(F_2)|} p(F_1, F_2; F) F$$

For example,

$$\begin{matrix} \bullet & \bullet \\ | & | \\ \bullet & \bullet \end{matrix} \times \begin{matrix} \bullet & \bullet \\ | & | \\ \bullet & \bullet \end{matrix} = \frac{1}{2} \begin{matrix} \bullet & \bullet & \bullet \\ | & | & | \\ \bullet & \bullet & \bullet \end{matrix} + \frac{1}{2} \begin{matrix} \bullet & \bullet & \bullet \\ | & | & | \\ \bullet & \bullet & \bullet \end{matrix} + \frac{1}{3} \begin{matrix} \bullet & \bullet & \bullet \\ | & | & | \\ \bullet & \bullet & \bullet \end{matrix} + \frac{1}{3} \begin{matrix} \bullet & \bullet & \bullet \\ | & | & | \\ \bullet & \bullet & \bullet \end{matrix} + \frac{1}{6} \begin{matrix} \bullet & \bullet & \bullet & \bullet \\ | & | & | & | \\ \bullet & \bullet & \bullet & \bullet \end{matrix} + \frac{1}{6} \begin{matrix} \bullet & \bullet & \bullet & \bullet \\ | & | & | & | \\ \bullet & \bullet & \bullet & \bullet \end{matrix} + \frac{1}{6} \begin{matrix} \bullet & \bullet & \bullet & \bullet \\ | & | & | & | \\ \bullet & \bullet & \bullet & \bullet \end{matrix}$$

A flag  $F$  when used in equations represents  $p(F; G)$  for a very large graph  $G$  (more formally, a sequence of graphs of increasing size, where  $p(F; G)$  converges for all  $F$ ).

**Exercise 3.** What is  $\begin{matrix} \bullet & \bullet \\ | & | \\ \bullet & \bullet \end{matrix}^2$ ?

**Exercise 4.** Explain why

$$p(F_1; G) \times p(F_2; G) \approx \sum_{|V(F)|=|V(F_1)|+|V(F_2)|} p(F_1, F_2; F) p(F; G)$$

in a very large graph  $G$ .

**Exercise 5.** Using flag operations, show that

$$\sum_{|V(F)|=n} F = 1$$

and explain why this makes sense given the interpretation of flags above. *Hint: Look at  $\bullet^n$ .*

**Exercise 6.** Using flag operations, show that

$$F = \sum_{|V(F')|=n} p(F; F') F'$$

and explain why this makes sense given the interpretation of flags above. *Hint: Look at  $\bullet^m \times F$ .*

**Exercise 7.** Extend  $p$  to  $p(F_1, \dots, F_m; G)$  so that

$$\prod_{i=1}^m F_i = \sum_{|V(F)|=\sum_{i=1}^m |V(F_i)|} p(F_1, \dots, F_m; F) F.$$

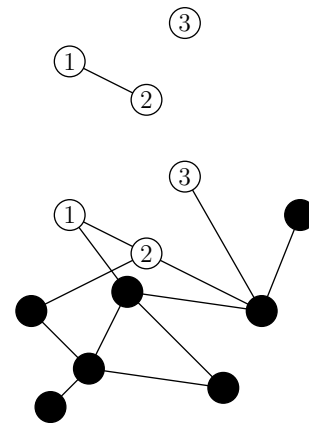
Also conclude that flag multiplication is associative.

### $\sigma$ -Flags

For a labeled graph  $\sigma$ , a  $\sigma$ -graph is a graph with some labeled vertices inducing  $\sigma$  and otherwise unlabeled vertices. We do not include the labeled vertices in the vertex set, nor the edges between them (edges from labeled to unlabeled vertices are included, however).

**Definition 8.** The *density* of a  $\sigma$ -graph  $F_\sigma$  in a  $\sigma$ -graph  $G_\sigma$ , denoted  $p_\sigma(F_\sigma; G_\sigma)$ , is the probability that the  $\sigma$ -subgraph induced in  $G_\sigma$  by the labeled vertices and  $|V(F_\sigma)|$  random vertices is  $F_\sigma$ .

We also denote by  $p_\sigma(F_{1\sigma}, F_{2\sigma}; G_\sigma)$  the probability that  $\sigma$  along with both  $|V(F_{1\sigma})|$  random vertices and  $|V(F_{2\sigma})|$  random remaining vertices in  $G_\sigma$  induce  $F_{1\sigma}$  and  $F_{2\sigma}$ , respectively.



A labeled graph  $\sigma$  (top) and a  $\sigma$ -graph (bottom).

$\sigma$ -flags are  $\sigma$ -graphs equipped with the following operations:

- $\sum_{|V(F_\sigma)|=1} F_\sigma = 1$ .
- Vector space operations and axioms, e.g., we can write  $\frac{1}{3} \begin{matrix} \bullet & \bullet \\ | & | \\ \circ & \circ \end{matrix} - \frac{1}{2} \begin{matrix} \bullet \\ | \\ \circ \end{matrix}$ .
- We can multiply  $\sigma$ -flags, as follows:

$$F_{1\sigma} \times F_{2\sigma} = \sum_{|V(F_\sigma)|=|V(F_{1\sigma})|+|V(F_{2\sigma})|} p_\sigma(F_{1\sigma}, F_{2\sigma}; F_\sigma) F_\sigma$$

For example,

$$\begin{matrix} \circ & \circ \\ | & | \\ \bullet & \bullet \end{matrix} \times \begin{matrix} \circ & \circ \\ | & | \\ \bullet & \bullet \end{matrix} = \frac{1}{2} \begin{matrix} \circ & \circ & \circ \\ | & | & | \\ \bullet & \bullet & \bullet \end{matrix} + \frac{1}{2} \begin{matrix} \circ & \circ & \circ \\ | & | & | \\ \bullet & \bullet & \bullet \end{matrix}$$

$$\begin{matrix} \bullet \\ | \\ \circ \end{matrix} \times \begin{matrix} \bullet & \bullet \\ | & | \\ \circ & \circ \end{matrix} = \frac{2}{3} \begin{matrix} \bullet & \bullet & \bullet \\ | & | & | \\ \circ & \bullet & \bullet \end{matrix} + \frac{2}{3} \begin{matrix} \bullet & \bullet & \bullet \\ | & | & | \\ \circ & \bullet & \bullet \end{matrix} + \frac{1}{3} \begin{matrix} \bullet & \bullet & \bullet \\ | & | & | \\ \circ & \bullet & \bullet \end{matrix} + \frac{1}{3} \begin{matrix} \bullet & \bullet & \bullet \\ | & | & | \\ \circ & \bullet & \bullet \end{matrix}$$

A  $\sigma$ -flag  $F_\sigma$  when used in equations represents  $p_\sigma(F_\sigma; G_\sigma)$  for a very large  $\sigma$ -graph  $G_\sigma$  (more formally, a sequence of  $\sigma$ -graphs of increasing size, where  $p_\sigma(F_\sigma; G_\sigma)$  converges for all  $F_\sigma$ ).

**Exercise 9.** What is  $\begin{matrix} \bullet^2 \\ | \\ \circ \end{matrix}$ ?

**Exercise 10.** Explain why

$$p_\sigma(F_{1\sigma}; G_\sigma) \times p_\sigma(F_{2\sigma}; G_\sigma) \approx \sum_{|V(F_\sigma)|=|V(F_{1\sigma})|+|V(F_{2\sigma})|} p_\sigma(F_{1\sigma}, F_{2\sigma}; F_\sigma) p_\sigma(F_\sigma; G_\sigma)$$

in a very large graph  $G_\sigma$ .

**Exercise 11.** Using  $\sigma$ -flag operations, show that

$$\sum_{|V(F_\sigma)|=n} F_\sigma = 1$$

and explain why this makes sense given the interpretation of  $\sigma$ -flags above. *Hint: Look at  $\left(\sum_{|V(F_\sigma)|=1} F_\sigma\right)^n$ .*

**Exercise 12.** Using  $\sigma$ -flag operations, show that

$$F_\sigma = \sum_{|V(F'_\sigma)|=n} p_\sigma(F_\sigma; F'_\sigma) F_\sigma$$

and explain why this makes sense given the interpretation of  $\sigma$ -flags above. *Hint: Look at  $\left(\sum_{|V(F_\sigma)|=1} F_\sigma\right)^m \times F_\sigma$ .*

**Exercise 13.** Extend  $p_\sigma$  to  $p_\sigma(F_{1\sigma}, \dots, F_{m\sigma}; G_\sigma)$  so that

$$\prod_{i=1}^m F_{i\sigma} = \sum_{|V(F_\sigma)|=\sum_{i=1}^m |V(F_{i\sigma})|} p_\sigma(F_{1\sigma}, \dots, F_{m\sigma}; F_\sigma) F_\sigma.$$

Also conclude that  $\sigma$ -flag multiplication is associative.

### The Averaging Operator

**Definition 14.** The *averaging operator* (a.k.a. *downward operator*) on a  $\sigma$ -flag  $F_\sigma$ , denoted  $\llbracket F_\sigma \rrbracket$ , is  $PF$ , where  $F$  is the unlabeled flag formed by removing the labels from  $F_\sigma$  and  $P$  and is the probability that putting  $|V(\sigma)|$  labels randomly on  $F$  gives  $F_\sigma$ .

Additionally, we make the averaging operator a linear operator.

For example,  $\llbracket \begin{array}{c} \circ \\ \diagdown \quad \diagup \\ \bullet \quad \bullet \end{array} \rrbracket = \frac{2}{3} \begin{array}{c} \bullet \quad \bullet \\ \diagdown \quad \diagup \\ \bullet \quad \bullet \end{array}.$

An interpretation of the averaging operator is that  $\llbracket F_\sigma \rrbracket$  is the probability that labeling  $|V(\sigma)|$  vertices of a very large graph  $G_\sigma$  then choosing  $|F_\sigma|$  vertices induces  $F_\sigma$ .

**Theorem 15.** (Razborov) For any two linear combinations of flags  $\mathcal{F}_1$  and  $\mathcal{F}_2$ ,

$$\llbracket \mathcal{F}_1^2_\sigma \rrbracket \times \llbracket \mathcal{F}_2^2_\sigma \rrbracket \geq \llbracket \mathcal{F}_1 \mathcal{F}_2 \rrbracket^2$$

**Exercise 16.** Show that  $\llbracket \mathcal{F}_\sigma^2 \rrbracket \geq 0$  for all linear combinations of flags.

Now let's look at the equation on page 1.

**Exercise 17.** What have we proved? Make sure you are comfortable with this proof.

**Exercise 18.** Show that the number of triangles in a graph  $G$  is at least  $(1 - o(1)) \cdot \left(\frac{4|E(G)|^2}{3|V(G)|} - \frac{|E(G)||V(G)|}{3}\right)$ .

Here,  $1 - o(1)$  denotes a value that gets arbitrarily close to 1 as  $G$  grows large.

*Hint: Expand:  $2 \llbracket \begin{array}{c} \bullet \\ \diagdown \quad \diagup \\ \bullet \quad \bullet \end{array} \rrbracket^2 - \begin{array}{c} \bullet \\ \bullet \end{array}$*