

## Reductions and completeness

- $\#3SAT$  is  $\#P$ -complete under **parsimonious** reductions (Exercise).

# Reductions and completeness

- **#3SAT** is #P-complete under **parsimonious** reductions (Exercise).
- **PERMANENT** is #P-complete under **Turing** reductions (Not an exercise).

# Reductions and completeness

- $\#3SAT$  is  $\#P$ -complete under **parsimonious** reductions (Exercise).
- **PERMANENT** is  $\#P$ -complete under **Turing** reductions (Not an exercise).
  - ▶ If **PERMANENT** is  $\#P$ -complete under parsimonious reductions, then  $P = NP$  (Exercise).

# Reductions and completeness

- $\#3SAT$  is  $\#P$ -complete under **parsimonious** reductions (Exercise).
- **PERMANENT** is  $\#P$ -complete under **Turing** reductions (Not an exercise).
  - ▶ If **PERMANENT** is  $\#P$ -complete under parsimonious reductions, then  $P = NP$  (Exercise).
- Every  $\#P$ -complete problem under parsimonious reductions has an NP-complete decision version (Exercise).

# Reductions and completeness

- $\#3SAT$  is  $\#P$ -complete under **parsimonious** reductions (Exercise).
- **PERMANENT** is  $\#P$ -complete under **Turing** reductions (Not an exercise).
  - ▶ If **PERMANENT** is  $\#P$ -complete under parsimonious reductions, then  $P = NP$  (Exercise).
- Every  $\#P$ -complete problem under parsimonious reductions has an NP-complete decision version (Exercise).
- **Conjecture:** Every NP-complete problem has a  $\#P$ -complete counting version.

## Some basic inclusions

- $FP \subseteq \#P \subseteq FPSPACE$ .
- $NP \subseteq P^{\#P[1]}$ .
- If  $FP = \#P$ , then  $P = NP$ .
- **Toda's Theorem:**  $PH \subseteq P^{\#P[1]}$ .

# Overview

## 1 Introduction to Counting Complexity

- The class  $\#P$
- **Three classes of counting problems**
- Holographic transformations

## 2 Matchgates and Holographic Algorithms

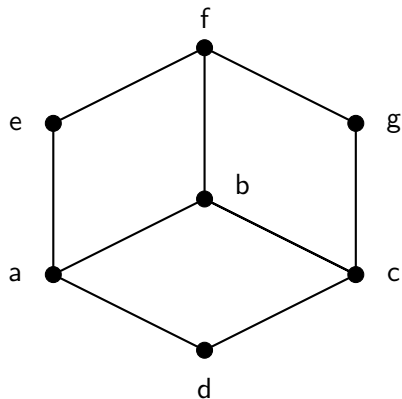
- Kasteleyn's algorithm
- Matchgates
- Holographic algorithms

## 3 Polynomial Interpolation

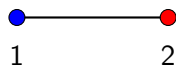
## 4 Dichotomy Theorems for counting problems

# Graph Homomorphism Problems

Input graph **G**



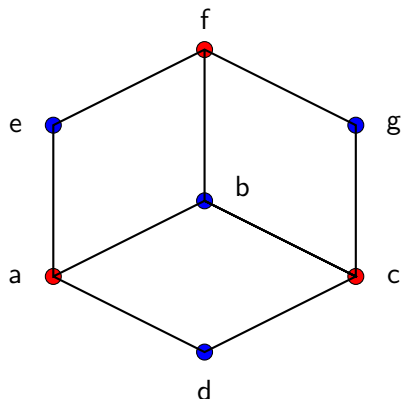
Target graph **H**



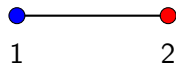


# Graph Homomorphism Problems

Input graph  $G$



Target graph  $H$

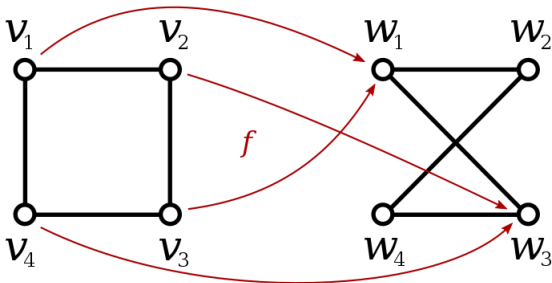


- Every homomorphism from  $G$  to  $H$  is a 2-coloring of  $G$ .
- The number of homomorphisms from  $G$  to  $H$  is equal to  $\#2\text{-COLORINGS}(G)$ .

# Graph homomorphisms preserve structure

## Definition

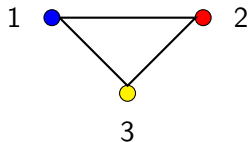
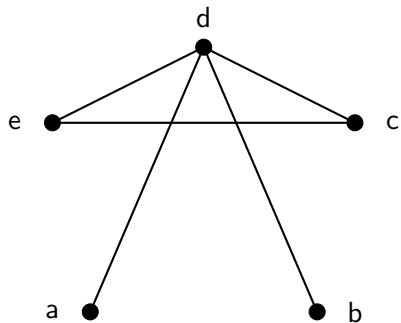
Given graphs  $G$  and  $H$ , a homomorphism from  $G$  to  $H$  is a function  $f : V(G) \rightarrow V(H)$  such that every edge  $(u, v) \in E(G)$  is mapped to an edge  $(f(u), f(v)) \in E(H)$ .



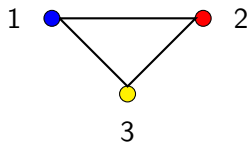
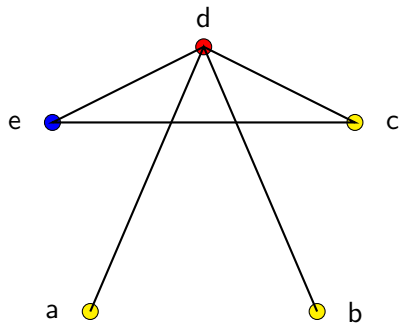
# Counting graph homomorphisms or $H$ -colorings of $G$

- We denote by  $\text{Hom}(G, H)$  the number of homomorphisms from  $G$  to  $H$ .
- Graph homomorphisms from  $G$  to  $H$  are also called  $H$ -colorings of  $G$ .
- We denote by  $\#\text{HOMSTO}H$  (or  $\#\text{H-COLORINGS}$ ) the problem of counting the number of homomorphisms from an input graph to a fixed graph  $H$ .

## Other examples – #3-COLORINGS

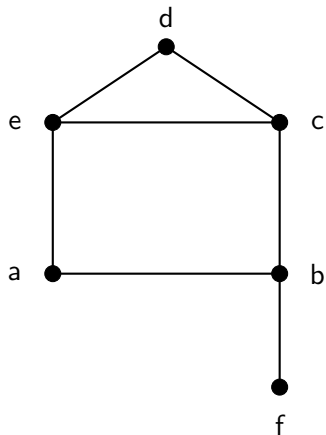


## Other examples – #3-COLORINGS

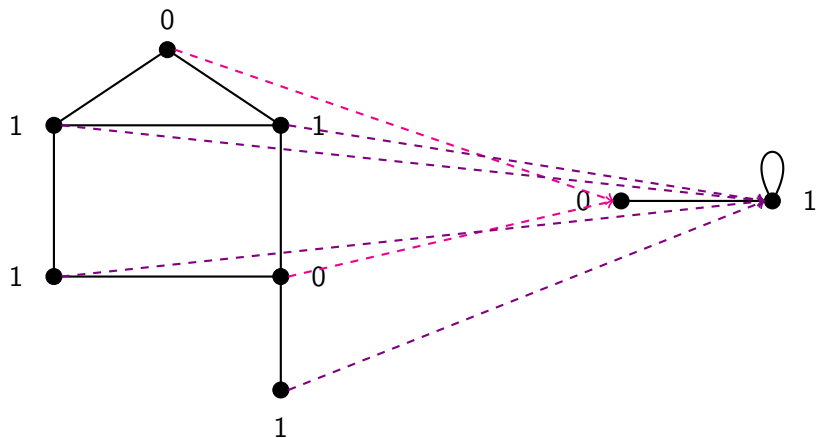


- $\text{Hom}(G, K_3) = \#3\text{-COLORINGS}(G)$ .
- $\text{Hom}(G, K_q) = \#q\text{-COLORINGS}(G)$  for any  $q \geq 2$ .

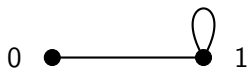
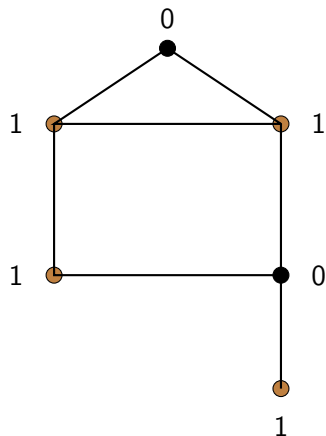
## Other examples – #VERTEXCOVERS



## Other examples – #VERTEXCOVERS



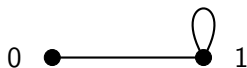
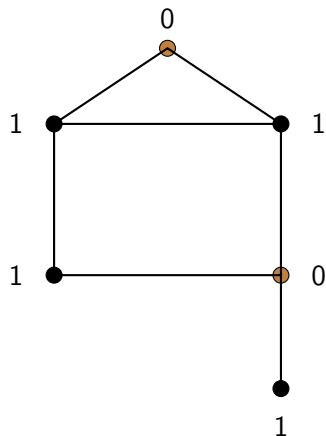
## Other examples – $\# \text{VERTEXCOVERS}$



- The subset of vertices mapped to 1 form a vertex cover.
- $\# \text{HOMSTO}H(G) = \# \text{VERTEXCOVERS}(G)$ .



## Other examples – #IS



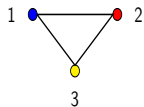
- The subset of vertices mapped to 0 form an independent set.
- $\# \text{HOMsToH}(G) = \# \text{IS}(G)$ .

- Let  $A$  be the (0-1) adjacency matrix of  $H$ .
- Given a graph  $G$ ,  $\#\text{HOMSTO}H$  is the problem of computing the following sum

$$Z_A(G) = \sum_{\sigma: V(G) \rightarrow V(H)} \prod_{(u,v) \in E(G)} A(\sigma(u), \sigma(v)).$$

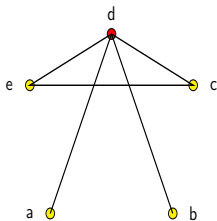
- Sometimes we write  $Z_H(G)$  instead of  $Z_A(G)$ .

For example,



has  $A = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}$ .

So, the following  $\sigma : V \rightarrow [3]$



contributes 0, since

$$A(\sigma(e), \sigma(c)) = A(3, 3) = 0.$$

## Weighted graph homomorphisms

- In general,  $H$  can be a weighted graph, and so  $A$  can be a matrix over the real or complex numbers.

## Weighted graph homomorphisms

- In general,  $H$  can be a weighted graph, and so  $A$  can be a matrix over the real or complex numbers.
- The product  $\prod_{(u,v) \in E(G)} A(\sigma(u), \sigma(v))$  is the weight of assignment  $\sigma$ .

## Weighted graph homomorphisms

- In general,  $H$  can be a weighted graph, and so  $A$  can be a matrix over the real or complex numbers.
- The product  $\prod_{(u,v) \in E(G)} A(\sigma(u), \sigma(v))$  is the weight of assignment  $\sigma$ .
- The sum over all assignments

$$Z_A(G) = \sum_{\sigma: V(G) \rightarrow V(H)} \prod_{(u,v) \in E(G)} A(\sigma(u), \sigma(v))$$

is called the **partition function**.

# Weighted graph homomorphisms

- Motivation and applications come from statistical physics.
- To capture the computational complexity of such functions, we consider the closure of  $\#P$  under Turing reductions.

That is the class  $FP^{\#P}$ .

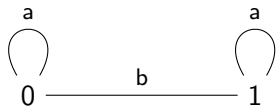
# Partition functions in statistical physics

- In statistical physics, partition functions can be represented as weighted graph homomorphisms.
- In this case,  $H$  is a weighted graph with both edge and vertex weights.
- Let  $A$  be the adjacency matrix of  $H$ , where  $A(u, v)$  is the weight of the edge  $(u, v) \in E(H)$  and let  $\{\lambda_v\}_{v \in V(H)}$  be the vertex weights.
- Weighted  $\#HOMSTO H$  is the problem of computing the following sum

$$\sum_{\sigma: V(G) \rightarrow V(H)} \left( \prod_{(u,v) \in E(G)} A(\sigma(u), \sigma(v)) \prod_{v \in V(G)} \lambda_{\sigma(v)} \right).$$



## 2-spin systems – Ising model



**Adjacency matrix**

$$A = \begin{bmatrix} a & b \\ b & a \end{bmatrix}$$

**Vertex weights**

$$\lambda_0 = 1 \text{ and } \lambda_1 = \lambda$$

$$Z_{a,b,\lambda}(G) =$$

$$\sum_{\sigma: V \rightarrow \{0,1\}} a^{|\{(u,v) \in E: \sigma(u) = \sigma(v)\}|} \cdot b^{|\{(u,v) \in E: \sigma(u) \neq \sigma(v)\}|} \cdot \lambda^{|\{u \in V: \sigma(u) = 1\}|}$$

## 2-spin systems – Hardcore model



**Adjacency matrix**

$$A = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$$

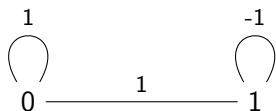
**Vertex weights**

$$\lambda_0 = 1 \text{ and } \lambda_1 = \lambda$$

$$Z_\lambda(G) = \sum_{I \in \mathcal{I}} \lambda^{|I|}$$

where  $\mathcal{I}$  is the set of all independent sets in  $G$

# Counting induced subgraphs with an even number of edges



**Adjacency matrix**

$$A = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

- The term  $\prod_{(u,v) \in E} A(\sigma(u), \sigma(v))$  is 1 if the subgraph of  $G$  induced by vertices mapped to 1 has an even number of edges and it is -1, otherwise.
- $Z_A(G) = X - Y$ , where  $X$  (resp.  $Y$ ) is the number of induced subgraphs of  $G$  with an even (resp. odd) number of edges.
- $X + Y = 2^n$
- So,

$$X = \frac{2^n + Z_A(G)}{2}.$$

# Constraint Satisfaction Problems

3SAT as a CSP:

- Variables  $x_1, \dots, x_n$
- Domain  $\{0, 1\}$
- Constraints  $(C_i, x_{i_1}, x_{i_2}, x_{i_3})$

Clause	Relation
$x_1 \vee x_2 \vee x_3$	$C_0 = \{0, 1\}^3 \setminus (0, 0, 0)$
$\neg x_1 \vee x_2 \vee x_3$	$C_1 = \{0, 1\}^3 \setminus (1, 0, 0)$
$\neg x_1 \vee \neg x_2 \vee x_3$	$C_2 = \{0, 1\}^3 \setminus (1, 1, 0)$
$\neg x_1 \vee \neg x_2 \vee \neg x_3$	$C_3 = \{0, 1\}^3 \setminus (1, 1, 1)$

# Constraint Satisfaction Problems

## Decision version

*Input:*

- A set of variables  $x_1, \dots, x_n$ .
- A domain  $[q]$  of size  $q$ .
- A set  $C$  of constraints  $(C_i, x_{i_1}, \dots, x_{i_k})$ , where  $C_i$  are relations on the domain of arity  $k$ .

*Output:* Is there an assignment of values to the variables such that all constraints are satisfied?

# Counting Constraint Satisfaction Problems

- A **counting** constraint satisfaction problem is parameterized by a set of local constraint **functions**  $\mathcal{F}$ .

# Counting Constraint Satisfaction Problems

- A **counting** constraint satisfaction problem is parameterized by a set of local constraint **functions**  $\mathcal{F}$ .
- It is denoted by  $\#\text{CSP}_q(\mathcal{F})$  when the constraint functions in  $\mathcal{F}$  are defined over a domain  $[q]$  of size  $q$ .

# Counting Constraint Satisfaction Problems

- A **counting** constraint satisfaction problem is parameterized by a set of local constraint **functions**  $\mathcal{F}$ .
- It is denoted by  $\#\text{CSP}_q(\mathcal{F})$  when the constraint functions in  $\mathcal{F}$  are defined over a domain  $[q]$  of size  $q$ .
- Every constraint is a function  $f \in \mathcal{F}$  of some arity  $k$  together with a sequence of  $k$  variables  $x_{i_1}, \dots, x_{i_k} \in \{x_1, \dots, x_n\}$ .



# Counting Constraint Satisfaction Problems

- A **counting** constraint satisfaction problem is parameterized by a set of local constraint **functions**  $\mathcal{F}$ .
- It is denoted by  $\#\text{CSP}_q(\mathcal{F})$  when the constraint functions in  $\mathcal{F}$  are defined over a domain  $[q]$  of size  $q$ .
- Every constraint is a function  $f \in \mathcal{F}$  of some arity  $k$  together with a sequence of  $k$  variables  $x_{i_1}, \dots, x_{i_k} \in \{x_1, \dots, x_n\}$ .
- *Output:* How many assignments of values to the variables are there such that all constraints are satisfied?  
Equivalently, compute the following sum

$$\sum_{x_1, \dots, x_n \in [q]} \prod_{(f, x_{i_1}, \dots, x_{i_k}) \in \mathcal{C}} f(x_{i_1}, \dots, x_{i_k})$$

## Examples of $\#CSP_q(\mathcal{F}) - \#SAT$

- Variables  $x_1, \dots, x_n$
- Domain  $\{0, 1\}$
- $\mathcal{F} = \{OR_k \mid k \geq 1\} \cup \{\neq_2\}$ , where

$$OR_k(x_1, \dots, x_k) = \begin{cases} 0, & \text{if } x_1 = \dots = x_k = 0 \\ 1, & \text{otherwise} \end{cases} \quad \text{and}$$

$$\neq_2(x_1, x_2) = \begin{cases} 0, & \text{if } x_1 = x_2 \\ 1, & \text{otherwise} \end{cases} .$$

## Examples of $\#CSP_2(\mathcal{F}) - \#SAT$

For example,  $(x_1 \vee x_2 \vee x_3) \wedge (\neg x_3 \vee x_4)$  corresponds to the following instance of  $\#CSP_2(\mathcal{F})$ :

- 1 Variables  $x_1, x_2, x_3, x_4, x_5$ .
- 2 Constraints  $(OR_3, x_1, x_2, x_3), (OR_2, x_5, x_4), (\neq_2, x_3, x_5)$ .

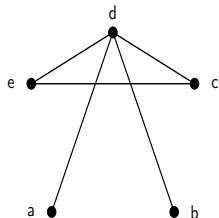
## Examples of $\#CSP_2(\mathcal{F})$ – Counting satisfying assignments

Problem	Constraint functions
$\#3SAT$	$\mathcal{F} = \{OR_3, \neq_2\}$
$\#NAE-3SAT$	$\mathcal{F} = \{Not-All-Equal_3, \neq_2\}$
$\#MONSAT$	$\mathcal{F} = \{OR_k \mid k \geq 1\}$
$\#MON3SAT$	$\mathcal{F} = \{OR_3\}$
$\#MON1-IN-3SAT$	$\mathcal{F} = \{Exact-One_3\}$

# An Example of $\#CSP_3(\mathcal{F})$ – $\#3$ -COLORINGS

- Variables  $x_1, \dots, x_n$
- Domain  $\{1, 2, 3\}$
- $\mathcal{F} = \{\neq_2\}$

For example,



corresponds to the following instance of  $\#CSP_3(\mathcal{F})$ :

- 1 Variables  $x_a, x_b, x_c, x_d, x_e$ .
- 2 Constraints  $(\neq_2, x_a, x_d), (\neq_2, x_b, x_d), (\neq_2, x_d, x_e), (\neq_2, x_c, x_d), (\neq_2, x_c, x_e)$ .

## Computing $\#\text{HOMSToH}$ is a $\#\text{CSP}_q(\mathcal{F})$

For graphs  $G$  and  $H$ , we construct a CSP instance with only one kind of constraint, as follows.

- The variables are the vertices of  $G$ .
- The domain is the vertex set of  $H$ .
- The constraints are  $((u, v), E(H))$  for every edge  $(u, v) \in E(G)$ .

An assignment  $\sigma$  of values to the variables is a function from  $V(G)$  to  $V(H)$  such that  $(\sigma(u), \sigma(v)) \in E(H)$  for every  $(u, v) \in E(G)$ .

## Computing $\#HOMSToH$ is a $\#CSP_q(\mathcal{F})$

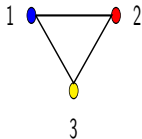
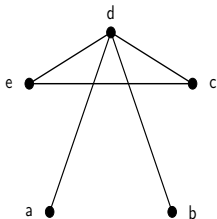
For graphs  $G$  and  $H$ , we construct a CSP instance with only one kind of constraint, as follows.

- The variables are the vertices of  $G$ .
- The domain is the vertex set of  $H$ .
- The constraints are  $((u, v), E(H))$  for every edge  $(u, v) \in E(G)$ .

An assignment  $\sigma$  of values to the variables is a function from  $V(G)$  to  $V(H)$  such that  $(\sigma(u), \sigma(v)) \in E(H)$  for every  $(u, v) \in E(G)$ .

For the counting version, we consider the corresponding constraint function  $E_H : V(H) \times V(H) \rightarrow \{0, 1\}$ .

## #3-COLORINGS revisited



- 1 Variables  $x_a, x_b, x_c, x_d, x_e$ .
- 2 Domain  $\{1, 2, 3\}$
- 3 Constraints  $(E_H, x_a, x_d), (E_H, x_b, x_d), (E_H, x_d, x_e), (E_H, x_c, x_d), (E_H, x_c, x_e),$

$$\text{where } E_H(x, y) = \begin{cases} 1, & \text{if } (x, y) \in E(H) \\ 0, & \text{otherwise} \end{cases}.$$



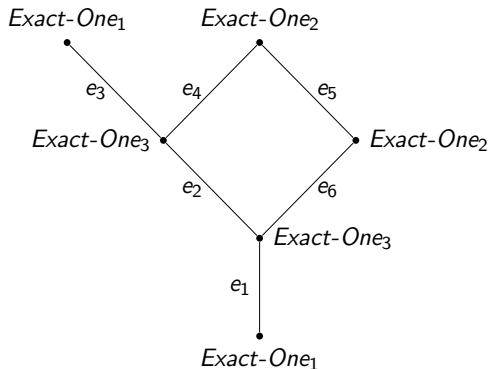
- $\#\text{CSP}_q(\mathcal{F})$  is a generalization of counting graph homomorphisms.

- $\#\text{CSP}_q(\mathcal{F})$  is a generalization of counting graph homomorphisms.
- Every  $\#\text{CSP}_q(\mathcal{F})$  is a problem of counting homomorphisms between two relational structures  $\mathcal{G}$  and  $\mathcal{H}$ .

- $\#\text{CSP}_q(\mathcal{F})$  is a generalization of counting graph homomorphisms.
- Every  $\#\text{CSP}_q(\mathcal{F})$  is a problem of counting homomorphisms between two relational structures  $\mathcal{G}$  and  $\mathcal{H}$ .
- Weighted counting CSP are defined by considering constraint functions over the rational, real or complex numbers.

- $\#\text{CSP}_q(\mathcal{F})$  is a generalization of counting graph homomorphisms.
- Every  $\#\text{CSP}_q(\mathcal{F})$  is a problem of counting homomorphisms between two relational structures  $\mathcal{G}$  and  $\mathcal{H}$ .
- Weighted counting CSP are defined by considering constraint functions over the rational, real or complex numbers.
- Boolean CSP are CSP with Boolean domain  $\{0, 1\}$ .

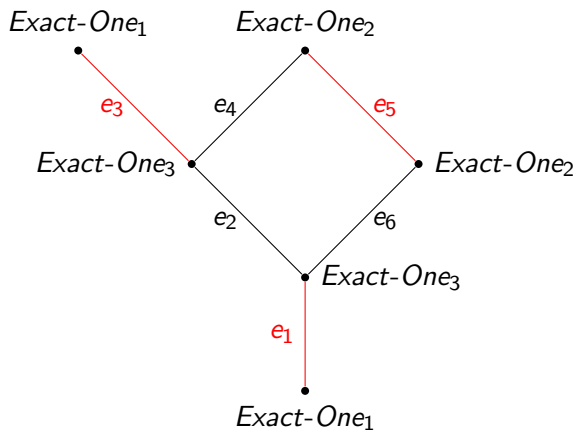
# Holant problems



$$Exact-One_2(e_i, e_j) = \begin{cases} 1, & \text{if } (e_i, e_j) = (0, 1) \text{ or } (e_i, e_j) = (1, 0) \\ 0, & \text{otherwise} \end{cases}$$

The subscript  $k$  in  $Exact-One_k$  denotes the arity of the function.

# Holant problems



# Holant problems

- A Holant problem is parameterized by a set  $\mathcal{F}$  of local constraint functions, also called **signatures**.

## Holant problems

- A Holant problem is parameterized by a set  $\mathcal{F}$  of local constraint functions, also called **signatures**.
- A **signature grid**  $\Omega = (G, \pi)$  over  $\mathcal{F}$  consists of a graph  $G = (V, E)$  and a mapping  $\pi$  that assigns to each vertex  $u \in V$  an  $f_u \in \mathcal{F}$  and a linear order on the incident edges at  $u$ .



## Holant problems

- A Holant problem is parameterized by a set  $\mathcal{F}$  of local constraint functions, also called **signatures**.
- A **signature grid**  $\Omega = (G, \pi)$  over  $\mathcal{F}$  consists of a graph  $G = (V, E)$  and a mapping  $\pi$  that assigns to each vertex  $u \in V$  an  $f_u \in \mathcal{F}$  and a linear order on the incident edges at  $u$ .
- The arity of  $f_u$  is equal to the degree of  $u$ , and the incident edges at  $u$  are associated with the input variables of  $f_u$ .

# Holant problems

- A Holant problem is parameterized by a set  $\mathcal{F}$  of local constraint functions, also called **signatures**.
- A **signature grid**  $\Omega = (G, \pi)$  over  $\mathcal{F}$  consists of a graph  $G = (V, E)$  and a mapping  $\pi$  that assigns to each vertex  $u \in V$  an  $f_u \in \mathcal{F}$  and a linear order on the incident edges at  $u$ .
- The arity of  $f_u$  is equal to the degree of  $u$ , and the incident edges at  $u$  are associated with the input variables of  $f_u$ .
- A signature is **symmetric** if its value is invariant under permutation of its variables.

## Holant problems

- A Holant problem is parameterized by a set  $\mathcal{F}$  of local constraint functions, also called **signatures**.
- A **signature grid**  $\Omega = (G, \pi)$  over  $\mathcal{F}$  consists of a graph  $G = (V, E)$  and a mapping  $\pi$  that assigns to each vertex  $u \in V$  an  $f_u \in \mathcal{F}$  and a linear order on the incident edges at  $u$ .
- The arity of  $f_u$  is equal to the degree of  $u$ , and the incident edges at  $u$  are associated with the input variables of  $f_u$ .
- A signature is **symmetric** if its value is invariant under permutation of its variables.
- If all signatures in  $\mathcal{F}$  are symmetric, then there is no need to assign an order for incident edges at any  $u$ .

# Holant problems

## Definition

For a set  $\mathcal{F}$  of signatures over a domain  $[q]$ , we define  $\text{Holant}_q(\mathcal{F})$  as the problem with

*Input:* A signature grid  $\Omega = (G, \pi)$  over  $\mathcal{F}$

*Output:*

$$\text{Holant}_q(\Omega, \mathcal{F}) = \sum_{\sigma: E \rightarrow [q]} \prod_{u \in V} f_u(\sigma \upharpoonright_{E(u)})$$

- $G = (V, E)$  and  $E(u)$  denotes the incident edges of  $u$ .
- $\sigma \upharpoonright_{E(u)}$  denotes the restriction of  $\sigma$  to  $E(u)$ .
- $f_u(\sigma \upharpoonright_{E(u)})$  is the evaluation of  $f_u$  on the ordered input tuple  $\sigma \upharpoonright_{E(u)}$ .
- We use  $\text{Holant}(\mathcal{F})$  to denote  $\text{Holant}_2(\mathcal{F})$ , the Holant problems over the Boolean domain.

## Some notation

- A **signature**  $f$  of arity  $k$  over the Boolean domain can be denoted by  $(f_0, f_1, \dots, f_{2^k-1})$ , where  $f_x$  is the output of  $f$  on  $x \in [2]^k$ , ordered lexicographically.

## Some notation

- A **signature**  $f$  of arity  $k$  over the Boolean domain can be denoted by  $(f_0, f_1, \dots, f_{2^k-1})$ , where  $f_x$  is the output of  $f$  on  $x \in [2]^k$ , ordered lexicographically.
- A signature  $f$  of arity  $k$  over the Boolean domain is **symmetric** if for every  $x, y \in \{0, 1\}^k$  of equal Hamming weight,  $f(x) = f(y)$ .
- In this case,  $f$  can also be expressed as  $[f_0, f_1, \dots, f_k]$ , where  $f_w$  is the value of  $f$  on inputs of Hamming weight  $w$ .

## Some notation

- A **signature**  $f$  of arity  $k$  over the Boolean domain can be denoted by  $(f_0, f_1, \dots, f_{2^k-1})$ , where  $f_x$  is the output of  $f$  on  $x \in [2]^k$ , ordered lexicographically.
- A signature  $f$  of arity  $k$  over the Boolean domain is **symmetric** if for every  $x, y \in \{0, 1\}^k$  of equal Hamming weight,  $f(x) = f(y)$ .
- In this case,  $f$  can also be expressed as  $[f_0, f_1, \dots, f_k]$ , where  $f_w$  is the value of  $f$  on inputs of Hamming weight  $w$ .
- For example, *Equality* signature of arity  $k$ ,

$$(\text{=}_k) = [1, 0, \dots, 0, 1]$$

and *Disequality* signature of arity 2,

$$(\text{\neq}_2) = [0, 1, 0].$$

# Holant problems – Examples

Examples of counting problems in  $k$ -regular graphs.

- Over the Boolean domain:

$$\text{Holant}(G; f) \text{ counts } \begin{cases} \text{matchings} & \text{in } G \text{ when } f = \text{AT-MOST-ONE}_k; \\ \text{perfect matchings} & \text{in } G \text{ when } f = \text{EXACT-ONE}_k; \\ \text{cycle covers} & \text{in } G \text{ when } f = \text{EXACT-TWO}_k; \\ \text{edge covers} & \text{in } G \text{ when } f = \text{OR}_k. \end{cases}$$

- Over domain of size  $q$ ,  $\text{Holant}_q(G, \text{All-Distinct}_k)$  is the problem  $\#q\text{-EDGECOLORINGS}$ .



## Planar / bipartite signature grids

- A **planar signature grid** is a signature grid s.t. its underlying graph is planar. We use  $\text{PI-Holant}_q(\mathcal{F})$  to denote the restriction of  $\text{Holant}_q(\mathcal{F})$  to planar signature grids.

## Planar / bipartite signature grids

- A **planar signature grid** is a signature grid s.t. its underlying graph is planar. We use  $\text{Pl-Holant}_q(\mathcal{F})$  to denote the restriction of  $\text{Holant}_q(\mathcal{F})$  to planar signature grids.
- A **bipartite signature grid** over  $(\mathcal{F} \mid \mathcal{G})$  is a signature grid  $\Omega = (H, \pi)$  over  $\mathcal{F} \cup \mathcal{G}$ , where  $H = (V_1 \cup V_2, E)$  is a bipartite graph, s.t.  $\pi(V_1) \subseteq \mathcal{F}$  and  $\pi(V_2) \subseteq \mathcal{G}$ . We use  $\text{Holant}_q(\mathcal{F} \mid \mathcal{G})$  to denote the restriction of  $\text{Holant}_q(\mathcal{F} \cup \mathcal{G})$  to bipartite signatures over  $(\mathcal{F} \mid \mathcal{G})$ .