

# Dichotomy Theorems for Counting Constraint Satisfaction Problems

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# Constraint Satisfaction Problem

- Let  $D = \{1, \dots, d\}$  be a domain.
- A **language** is a finite set of relations  $\Gamma = \{\Theta_1, \dots, \Theta_h\}$ .  
An instance of  $\text{CSP}(\Gamma)$  consists of a set of variables  $x_1, \dots, x_n$  and a set of constraints from  $\Gamma$ . It defines an  $n$ -ary relation  $R \subseteq D^n$ , where  $(x_1, \dots, x_n) \in R$  if all constraints are satisfied.

$$R \subseteq D^4 : \Theta_1(x_1, x_3, x_2) \wedge \Theta_2(x_4, x_3) \wedge \Theta_2(x_2, x_3)$$

- **Decide** if  $R$  is **empty** or not.

# Examples

- $d$ -COLORING:  $D = \{1, \dots, d\}$  and  $\Gamma = \{\Theta\}$ , where

$$\Theta = \{(i, j) : i, j \in D \text{ and } i \neq j\}$$

- INDEPENDENT SET:  $D = \{1, 2\}$  and  $\Gamma = \{\Theta\}$ , where

$$\Theta = \{(1, 1), (1, 2), (2, 1)\}$$

- 2-SAT:  $D = \{0, 1\}$  and

$$\Gamma = \{x_1 \vee x_2, \bar{x}_1 \vee x_2, x_1 \vee \bar{x}_2, \bar{x}_1 \vee \bar{x}_2\}$$

- 3-SAT ...

One of the most important classes of problems in TCS:

- **Decision**: whether a solution exists?  
The CSP dichotomy conjecture of Feder and Vardi is open.
- **Optimization**: satisfy as many constraints as possible, and more generally, the valued constraint satisfaction problem to find an assignment to maximize the total weight.
- **Counting**: This talk.

# Counting Constraint Satisfaction Problem

- Let  $D = \{1, \dots, d\}$  be a domain.
- A **language** is a finite set of relations  $\Gamma = \{\Phi_1, \dots, \Phi_h\}$ .  
An instance of  $\#\text{CSP}(\Gamma)$  consists of variables  $x_1, \dots, x_n$  and a set of constraints from  $\Gamma$ . It defines an  $n$ -ary relation  $R \subseteq D^n$ , where  $(x_1, \dots, x_n) \in R$  if all constraints are satisfied.
- Compute  $|R|$ .

# Examples

- #  $d$ -COLORING:  $D = \{1, \dots, d\}$  and  $\Gamma = \{\Theta\}$ , where

$$\Theta = \{(i, j) : i, j \in D \text{ and } i \neq j\}.$$

- # INDEPENDENT SET:  $D = \{1, 2\}$  and  $\Gamma = \{\Theta\}$ , where

$$\Theta = \{(1, 1), (1, 2), (2, 1)\}.$$

- # 2-SAT:  $D = \{0, 1\}$  and

$$\Gamma = \{x_1 \vee x_2, \bar{x}_1 \vee x_2, x_1 \vee \bar{x}_2, \bar{x}_1 \vee \bar{x}_2\}$$

- # 3-SAT ...

- A **weighted** language  $\mathcal{L} = \{g_1, \dots, g_h\}$  with  $g_i : D^{r_i} \rightarrow \mathbb{C}$ .
- An instance of  $\#CSP(\mathcal{L})$  consists of variables  $x_1, \dots, x_n$  over  $D$  and a set of functions from  $\mathcal{L}$ . It defines an  $n$ -ary function  $F$ : for any assignment  $\mathbf{x} = (x_1, \dots, x_n) \in D^n$ ,  $F(\mathbf{x})$  is the product of the constraint function evaluations. E.g.,

$$F(x_1, x_2, x_3, x_4) = g_1(x_1, x_3, x_2) \cdot g_2(x_2, x_4) \cdot g_2(x_3, x_2)$$

- Compute  $\sum_{\mathbf{x} \in D^n} F(\mathbf{x})$ .

# Counting Graph Homomorphisms

The special case when  $\mathcal{L}$  consists of a single symmetric binary function [Dyer and Greenhill 00], [Bulatov and Grohe 05], [Goldberg, Grohe, Jerrum and Thurley 09], [Cai, C and Lu 11].



# Complexity Dichotomies Arise for $\#CSP$

## Theorem (Bulatov 08)

*$\#CSP(\Gamma)$  either can be solved in P-time or is  $\#P$ -complete.*

## Theorem (Dyer and Richerby 10)

*An alternative proof; the tractability criterion is decidable in NP.*

Further extended to nonnegative rational languages [Bulatov, Dyer, Goldberg, Jalsenius, Jerrum and Richerby 10], and nonnegative algebraic languages [Cai, C and Lu 11].

## Theorem (Cai and C 12)

*A dichotomy for  $\#CSP(\mathcal{L})$  with complex weights.*

- ① Dichotomy for Unweighted #CSP:
  - Tractability criterion: Strong balance
  - Mal'tsev polymorphisms and Witness functions
  - The main counting algorithm.
- ② Dichotomy for Nonnegative and Complex #CSP

## Definition

An  $n \times m$  nonnegative matrix is *rectangular* if  $A_{i,k}, A_{i,\ell}, A_{j,k} > 0$  imply  $A_{j,\ell} > 0$  (or block-diagonal where every block is all positive).

$$\begin{pmatrix} 1 & 2 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 2 & 2 & 0 \\ 0 & 0 & 3 & 3 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

## Definition

A nonnegative matrix is *block-rank-1* if one can permute its rows and columns to make it block-diagonal and every block is rank 1.

$$\begin{pmatrix} 1 & 2 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 2 & 2 & 0 \\ 0 & 0 & 3 & 3 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$



# Strong Rectangularity and Strong Balance

- Let  $\Gamma = \{\Theta_1, \dots, \Theta_h\}$  be a language over  $D$ .
- Given an  $n$ -ary relation  $R \subseteq D^n$  derived by an instance of  $\#\text{CSP}(\Gamma)$  and integers  $k, \ell, r$  such that  $k + \ell + r = n$ , we are interested in the following  $|D|^k \times |D|^\ell$  matrix  $\mathbf{M}$ :

$$M(\mathbf{u}, \mathbf{v}) = \left| \{ \mathbf{w} \in D^r : (\mathbf{u}, \mathbf{v}, \mathbf{w}) \in R \} \right|,$$

with rows indexed by  $\mathbf{u} \in D^k$ , columns indexed by  $\mathbf{v} \in D^\ell$ .

# Strong Rectangularity and Strong Balance

Definition (Dyer and Richerby 10)

$\Gamma$  is *strongly rectangular* if every such matrix  $\mathbf{M}$  is rectangular;  
 $\Gamma$  is *strongly balanced* if every such matrix  $\mathbf{M}$  is block-rank-1.

Strong balance implies strong rectangularity.

equivalent to [congruence singularity](#) [Bulatov 08].

### Theorem (Bulatov 08)

If  $\Gamma$  is *congruence singular*, then  $\#CSP(\Gamma)$  is solvable in  $P$ -time; otherwise  $\#CSP(\Gamma)$  is  $\#P$ -hard.

### Theorem (Dyer and Richerby 10)

If  $\Gamma$  is *strongly balanced*, then  $\#CSP(\Gamma)$  is solvable in  $P$ -time; otherwise  $\#CSP(\Gamma)$  is  $\#P$ -hard.

### Proof of the Hardness Part:

**Gadget construction:** A reduction from  $\text{EVAL}(\mathbf{A})$  to  $\#CSP(\Gamma)$  for a nonnegative  $\mathbf{A}$  that violates the condition of [Bulatov-Grohe 05].

## Definition

Let  $\Theta \subseteq D^r$  be an  $r$ -ary relation, and  $\psi : D^3 \rightarrow D$  be a map. Then we say  $\psi$  is a **polymorphism** of  $\Theta$  if  $\mathbf{u}, \mathbf{v}, \mathbf{w} \in \Theta$  implies

$$\left( \psi(u_1, v_1, w_1), \psi(u_2, v_2, w_2), \dots, \psi(u_r, v_r, w_r) \right) \in \Theta.$$

$u_1$	$u_2$	$\dots$	$u_r$
$v_1$	$v_2$	$\dots$	$v_r$
$w_1$	$w_2$	$\dots$	$w_r$

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$$\psi(u_1, v_1, w_1) \quad \psi(u_2, v_2, w_2) \quad \dots \quad \psi(u_r, v_r, w_r)$$



## Definition

Map  $\psi$  is a *Mal'tsev polymorphism* of  $\Theta$  if it also satisfies

$$\psi(a, b, b) = \psi(b, b, a) = a, \quad \text{for all } a, b \in D.$$

We say  $\psi$  is a *Mal'tsev polymorphism* of  $\Gamma = \{\Theta_1, \dots, \Theta_h\}$  if  $\psi$  is a *Mal'tsev polymorphism* of every relation  $\Theta_i$ .

## Observation

If  $\psi$  is a Mal'tsev polymorphism of  $\Gamma$ , then it is also a Mal'tsev polymorphism of every relation  $R$  derived by a  $\#CSP(\Gamma)$  instance.

$$R \subseteq D^4 : \Theta_1(x_1, x_2, x_3) \wedge \Theta_2(x_3, x_4) \wedge \Theta_2(x_4, x_2)$$

$u_1$	$u_2$	$u_3$	$u_4$
$v_1$	$v_2$	$v_3$	$v_4$
$w_1$	$w_2$	$w_3$	$w_4$

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$$\psi(u_1, v_1, w_1) \quad \psi(u_2, v_2, w_2) \quad \psi(u_3, v_3, w_3) \quad \psi(u_4, v_4, w_4)$$

## Theorem

$\Gamma$  is strongly rectangular iff it has a Mal'tsev polymorphism  $\psi$ .

Proof of the easier direction  $\Leftarrow$ .

Let  $R$  be a relation derived by a  $\#CSP(\Gamma)$  instance. Let  $\mathbf{u}, \mathbf{u}' \in D^k$   
 $\mathbf{v}, \mathbf{v}' \in D^\ell$ . If the  $(\mathbf{u}', \mathbf{v}), (\mathbf{u}, \mathbf{v}), (\mathbf{u}, \mathbf{v}')$  entries of  $\mathbf{M}$  are positive:

$$\begin{array}{cccccccccc} u'_1 & \dots & u'_k & v_1 & \dots & v_\ell & w_1 & \dots & w_r & \in R \\ u_1 & \dots & u_k & v_1 & \dots & v_\ell & w'_1 & \dots & w'_r & \in R \\ u_1 & \dots & u_k & v'_1 & \dots & v'_\ell & w''_1 & \dots & w''_r & \in R \\ \hline u'_1 & \dots & u'_k & v'_1 & \dots & v'_\ell & w_1^* & \dots & w_r^* & \in R \end{array}$$

This implies that the  $(\mathbf{u}', \mathbf{v}')$  entry of  $\mathbf{M}$  is also positive. □

Assume that  $\psi$  is given.

Let  $\psi$  be a Mal'tsev polymorphism of  $R \subseteq D^n$ . For  $i \in [n]$ :

- 1  $\text{Pr}_i R$ : the projection of  $R$  on the  $i$ th coordinate, i.e.,  
 $a \in \text{Pr}_i R$  if there exists  $\mathbf{u} \in R$  with  $u_i = a$  (called a **witness**).
- 2  $\sim_i$  over  $\text{Pr}_i R$ :  $a \sim_i b$  if there exist  $\mathbf{u} \in D^{i-1}$ ,  $\mathbf{w}, \mathbf{w}' \in D^{n-i}$ :

$$(\mathbf{u}, a, \mathbf{w}) \in R \quad \text{and} \quad (\mathbf{u}, b, \mathbf{w}') \in R.$$

### Lemma

If  $R$  has a Mal'tsev polymorphism,  $\sim_i$  is an **equivalence** relation.

Proof.

Goal:  $a \sim_i b$  and  $b \sim_i c$  imply  $a \sim_i c$ .

$a \sim_i b \Rightarrow$  there exist  $\mathbf{u}, \mathbf{v}, \mathbf{v}'$  such that  $(\mathbf{u}, a, \mathbf{v}), (\mathbf{u}, b, \mathbf{v}') \in R$ .

$b \sim_i c \Rightarrow$  there exist  $\mathbf{u}', \mathbf{w}, \mathbf{w}'$  such that  $(\mathbf{u}', b, \mathbf{w}), (\mathbf{u}', c, \mathbf{w}') \in R$ .

$$\mathbf{u} \quad a \quad \mathbf{v} \quad \in R$$

$$\mathbf{u} \quad b \quad \mathbf{v}' \quad \in R$$

$$\mathbf{u}' \quad b \quad \mathbf{w} \quad \in R$$

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$$\mathbf{u}' \quad a \quad \mathbf{w}^* \quad \in R$$

Since  $(\mathbf{u}', c, \mathbf{w}') \in R$ , we have  $a \sim_i c$ . □

# Witness Function: A Succinct Representation

## Definition (Dyer and Richerby 10)

Suppose  $R \subseteq D^n$  has a Mal'tsev polymorphism  $\psi$ . We say

$$\omega : [n] \times D \rightarrow D^n \cup \{\text{NIL}\}$$

is a *witness function* of  $R$  if for every  $i \in [n]$ :

- 1  $a \notin \text{Pr}_i R \Rightarrow \omega(i, a) = \text{NIL}$ ;
- 2  $a \in \text{Pr}_i R, \omega(i, a) \in R$  and its  $i$ th entry is  $a$ ;
- 3 If  $a \sim_i b$ ,  $\omega(i, a)$  and  $\omega(i, b)$  share the same  $(i - 1)$ -prefix.

Similar to the [compact representation](#) of [Bulatov and Dalmau 06].  
A witness function  $\omega$  of  $R \subseteq D^n$  is of polynomial length.

# Witness Function: A Succinct Representation

## Lemma

*Suppose  $R \subseteq D^n$  has a Mal'tsev polymorphism  $\psi$ . Given  $\omega$  and a tuple  $\mathbf{u} \in D^n$ , one can decide if  $\mathbf{u} \in R$  or not in P-time.*

## Lemma

*Suppose  $R \subseteq D^n$  has a Mal'tsev polymorphism  $\psi$ . Given  $\omega$  and  $\mathbf{u} \in D^t$  for some  $t \leq n$ , one can decide if  $\mathbf{u} \in \text{Pr}_{[t]} R$  in P-time.*

Round 1: Check if  $u_1 \in \text{Pr}_1 R$ ; if so find a witness.

- If  $\omega(1, u_1) = \text{NIL}$ , reject.

Otherwise, let  $\omega(1, u_1) = (u_1, v_2, \mathbf{w}) \in R$  (a witness).



Let  $\omega(1, u_1) = (u_1, v_2, \mathbf{w}) \in R$ .

**Round 2:** Check if  $(u_1, u_2) \in \text{Pr}_{[2]} R$ ; if so find a witness.

- 1 If  $\omega(2, u_2) = \text{NIL}$ , reject.
- 2 If  $\omega(2, u_2)$  and  $\omega(2, v_2)$  have different first entries, reject.  
As  $(u_1, u_2) \in \text{Pr}_{[2]} R$  would imply that  $u_2 \sim_2 v_2$ .
- 3 Otherwise, let  $\omega(2, u_2) = (w_1, u_2, \mathbf{w}')$ ,  $\omega(2, v_2) = (w_1, v_2, \mathbf{w}^*)$ .

$$\begin{array}{rcll} u_1 & v_2 & \mathbf{w} & \in R \\ w_1 & v_2 & \mathbf{w}^* & \in R \\ w_1 & u_2 & \mathbf{w}' & \in R \\ \hline u_1 & u_2 & \mathbf{w}'' & \in R \end{array}$$

The result  $(u_1, u_2, \mathbf{w}'')$  is a witness for  $(u_1, u_2) \in \text{Pr}_{[2]} R$ .

Repeat for  $t$  rounds ...

## Lemma (Dyer and Richerby 10)

*Suppose  $\Gamma$  has a Mal'tsev polymorphism  $\psi$ . Given an #CSP instance, a witness function for its relation can be built in P-time.*

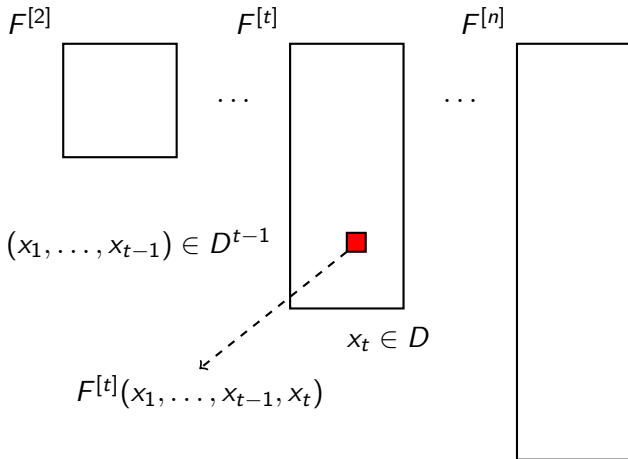
Assume that  $\Gamma = \{\Theta_1, \dots, \Theta_h\}$  is strongly balanced and thus, has a Mal'tsev polymorphism  $\psi$ .

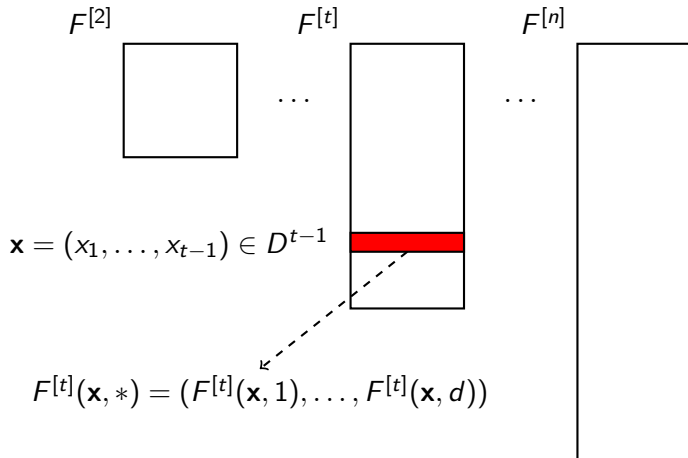
Given a  $\#\text{CSP}(\Gamma)$  instance with  $n$  variables that defines  $R \subseteq D^n$ :

- For each  $t = 1, \dots, n$ , let

$$F^{[t]}(x_1, \dots, x_t) = \left| \{ \mathbf{w} \in D^{n-t} : (x_1, \dots, x_t, \mathbf{w}) \in R \} \right|.$$

View each  $F^{[t]}$ ,  $t \geq 2$ , as a  $d^{t-1} \times d$  matrix.





For each  $t \geq 2$ , access to a **data structure** for  $F^{[t]}$ :

- One can send a  $(t - 1)$ -tuple  $\mathbf{x} \in D^{t-1}$  to the data structure. If  $F^{[t]}(\mathbf{x}, *) = \mathbf{0}$ , return  $\mathbf{v} = \mathbf{0}$ ; otherwise, return a nonzero vector  $\mathbf{v}$  **linearly dependent** with  $F^{[t]}(\mathbf{x}, *)$ , in P-time.

Given access to such data structures, compute

$$|R| = \sum_{a_1 \in D} F^{[1]}(a_1).$$

# The Main Counting Algorithm

To compute  $F^{[1]}(a_1)$  for some  $a_1 \in D$ :

- 1 send  $(a_1)$  to the **data structure** for  $F^{[2]}$
- 2 receive  $\mathbf{v}$  that is linearly dependent with  $F^{[2]}(a_1, *)$
- 3 if  $\mathbf{v} = \mathbf{0}$ ,  $F^{[2]}(a_1, *) = \mathbf{0} \Rightarrow F^{[1]}(a_1) = 0$
- 4 otherwise, let  $v_{a_2}$  be a nonzero entry of  $\mathbf{v}$ ,  $a_2 \in D$ :

$$F^{[1]}(a_1) = \sum_{b \in D} F^{[2]}(a_1, b) = F^{[2]}(a_1, a_2) \left( \frac{1}{v_{a_2}} \sum_{b \in D} v_b \right)$$

# The Main Counting Algorithm

To compute  $F^{[2]}(a_1, a_2)$ :

- 1 send  $(a_1, a_2)$  to the **data structure** for  $F^{[3]}$
- 2 receive  $\mathbf{w}$  that is linearly dependent with  $F^{[3]}((a_1, a_2), *)$
- 3 if  $\mathbf{w} = \mathbf{0}$ , then  $F^{[2]}(a_1, a_2) = 0$
- 4 so  $\mathbf{w} \neq \mathbf{0}$ ; let  $w_{a_3}$  be a nonzero entry of  $\mathbf{w}$ ,  $a_3 \in D$ :

$$F^{[2]}(a_1, a_2) = F^{[3]}(a_1, a_2, a_3) \left( \frac{1}{w_{a_3}} \sum_{b \in D} w_b \right)$$



# The Main Counting Algorithm

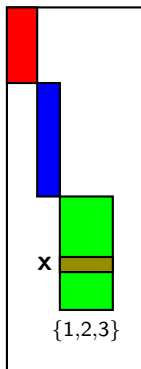
After  $n - 1$  steps, the algorithm reduces  $F^{[1]}(a_1)$  to

$$F^{[n]}(a_1, a_2, \dots, a_n)$$

for some appropriate  $a_2, \dots, a_n \in D$ .  $F = F^{[n]}$  is easy to evaluate.

Rest of the tractability proof: *How to build the data structures?*

Strong balance  $\Rightarrow F^{[t]}$  is a block-rank-1 matrix.



Each equivalence class  $\mathcal{E}_i$  of  $\sim_t$  corresponds to a block:

$$a \sim_t b \Rightarrow F^{[t]}(\mathbf{x}, a) > 0 \text{ and } F^{[t]}(\mathbf{x}, b) > 0 \text{ for some } \mathbf{x}$$

To build the data structure for  $F^{[t]}$ , it suffices to compute a **representative vector**  $\mathbf{v}_j$  for each block (equivalently, each equivalence class  $\mathcal{E}_j$  of  $\sim_t$ ).

- For a query  $\mathbf{x} \in D^{t-1}$ ,  $\mathbf{x}$  is in the block of  $\mathcal{E}_j$

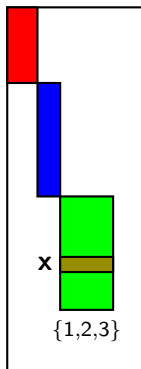


$$F^{[t]}(\mathbf{x}, a) > 0 \text{ for some } a \in \mathcal{E}_j$$



$$(\mathbf{x}, a) \in \Pr_{[t]} R \text{ for some } a \in \mathcal{E}_j$$

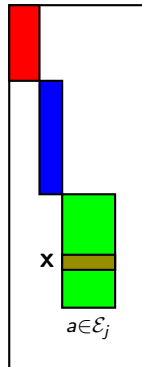
- Return  $\mathbf{v}_j$  if  $\mathbf{x}$  is in the block of  $\mathcal{E}_j$ ; return  $\mathbf{0}$  if it does not belong to any block.



# Building the Data Structures Backwards

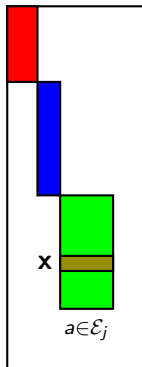
Initialization: The data structure for  $F^{[n]}$ .

- For each equivalence class  $\mathcal{E}_j$  of  $\sim_n$ , pick an  $a \in \mathcal{E}_j$ . Let  $\omega(n, a) = (\mathbf{x}, a) \in R$ . Set the representative vector  $\mathbf{v}_j$  of  $\mathcal{E}_j$  to be  $F^{[n]}(\mathbf{x}, *)$ .



Induction: The data structure for  $F^{[t]}$ .

- For each equivalence class  $\mathcal{E}_j$  of  $\sim_j$ , pick an  $a \in \mathcal{E}_j$ . Let  $\omega(t, a) = (\mathbf{x}, a, \mathbf{v}) \in R$ . Set the representative vector  $\mathbf{v}_j$  to be  $F^{[t]}(\mathbf{x}, *)$ .
- Use **data structures** for  $F^{[t+1]}, \dots, F^{[n]}$  and the **main counting algorithm** to evaluate  $F^{[t]}$ .



# Strong Balance for Nonnegative Languages

- Let  $\mathcal{L} = \{g_1, \dots, g_h\}$  be a nonnegative language.
- Given an  $n$ -ary function  $F$  derived by an instance of  $\#\text{CSP}(\mathcal{L})$  and integers  $k, \ell, r$  such that  $k + \ell + r = n$ , we are interested in the following  $|D|^k \times |D|^\ell$  matrix  $\mathbf{M}$ :

$$M(\mathbf{u}, \mathbf{v}) = \sum_{\mathbf{w} \in D^r} F(\mathbf{u}, \mathbf{v}, \mathbf{w}).$$

## Definition

$\mathcal{L}$  is *strongly balanced* if every such matrix  $\mathbf{M}$  is block-rank-1.

## Theorem (Cai, C and Lu)

If a nonnegative language  $\mathcal{L}$  is strongly balanced, then  $\#\text{CSP}(\mathcal{L})$  is solvable in  $P$ -time; otherwise, it is  $\#\text{P}$ -hard.

- 1 Dichotomy for Unweighted #CSP:
  - Tractability criterion: Strong balance
  - Mal'tsev polymorphisms and Witness functions
  - The counting algorithm
  - Generalization to Nonnegative #CSP
- 2 Dichotomy for #CSP with Complex Values



Cancellations ( $\{\pm 1\}$  or even roots of unity) may sometimes lead to efficient algorithms and more tractable cases (e.g., Permanent vs Determinant and Holographic algorithms [Valiant 04]).

Let  $\mathcal{L}$  be a complex-valued language, and  $F : D^n \rightarrow D$  be an  $n$ -ary function derived by a  $\#CSP(\mathcal{L})$  instance. Let  $R \subseteq D^n$ :

$$\mathbf{x} \in R \iff F(\mathbf{x}) \neq 0.$$

Even with a **witness function**  $\omega$  of  $R$ , not clear how to use  $\omega$  to decide efficiently if  $F^{[t]}(x_1, \dots, x_t) = 0$  or not, where

$$F^{[t]}(x_1, \dots, x_t) = \sum_{\mathbf{w} \in D^{n-t}} F(x_1, \dots, x_t, \mathbf{w}).$$

# The Same Data Structures

For each  $t \geq 2$ , build a **data structure** for  $F^{[t]}$ :

- One can send a  $(t - 1)$ -tuple  $\mathbf{x} \in D^{t-1}$  to the data structure. If  $F^{[t]}(\mathbf{x}, *) = \mathbf{0}$ , return  $\mathbf{v} = \mathbf{0}$ ; otherwise, return a nonzero vector  $\mathbf{v}$  **linearly dependent** with  $F^{[t]}(\mathbf{x}, *)$ , in P-time.

Then a similar **main counting algorithm** can compute efficiently

$$\sum_{\mathbf{x} \in D^n} F(\mathbf{x}).$$

# The First Difficulty

- 1 An  $d^{t-1} \times d$  matrix may have  $d^{t-1}$  pairwise linearly independent rows. Cannot even afford to store this many **representative vectors**.



# A Hint from Counting Graph Homomorphisms

Real matrices [Goldberg, Grohe, Jerrum and Thurley 09]  
and complex matrices [Cai, C and Lu 11]

$$\begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & 1 & -1 \\ 1 & -1 & -1 & 1 \end{pmatrix} \quad \text{and} \quad \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & \zeta & \zeta^2 & \zeta^3 & \zeta^4 \\ 1 & \zeta^2 & \zeta^4 & \zeta & \zeta^3 \\ 1 & \zeta^3 & \zeta & \zeta^4 & \zeta^2 \\ 1 & \zeta^4 & \zeta^3 & \zeta^2 & \zeta^1 \end{pmatrix}$$

Wishful thinking: What if any two rows of  $F^{[t]}$  are **either linearly dependent or orthogonal**  $\Rightarrow$  **At most  $d$**  representative vectors.

## The Block Orthogonality condition

Let  $F$  be a function defined by a  $\#CSP(\mathcal{L})$  instance. Then every two rows of  $F^{[t]}$  are **either linearly dependent or orthogonal**.

## Lemma

*If  $\mathcal{L}$  violates this condition, then  $\#CSP(\mathcal{L})$  is  $\#P$ -hard.*

# The Second Difficulty

- ② Let  $\{\mathbf{v}_i\}$  be the representative vectors of  $F^{[t]}$ . Let  $S_j$  denote the set of rows  $\mathbf{u} \in D^{t-1}$  that are linearly dependent with  $\mathbf{v}_i$ . Given a query  $\mathbf{u} \in D^{t-1}$ , how to decide if  $\mathbf{u} \in S_j$  or not?

*A witness function for  $R$  no longer helps!*



Wishful thinking: Every  $S_j \subseteq D^{t-1}$  has a **Mal'tsev polymorphism**.  
If so, one can hope to build a **witness function** for each  $S_j$ .

### The Mal'tsev condition

Let  $F$  be a function defined by a  $\#CSP(\mathcal{L})$  instance. Then all such sets  $S_j \subseteq D^{t-1}$  share a Mal'tsev polymorphism  $\psi$ .

### Lemma

*If  $\mathcal{L}$  violates this condition, then  $\#CSP(\mathcal{L})$  is  $\#P$ -hard.*



# The Last Difficulty

- ③ How to build the data structures for  $F^{[t]}$  inductively? Need to compute the representative vectors (at most  $d$  many) and to compute a witness function  $\omega_j$  for each  $S_j \subseteq D^{t-1}$ .

## Type Partition condition

Manipulate relations that share a Mal'tsev polymorphism.

## Lemma

*If  $\mathcal{L}$  violates the Type Partition condition,  $\#CSP(\mathcal{L})$  is  $\#P$ -hard.*

# Putting the Pieces Together ...

For  $t = n, \dots, 2$ , build inductively a data structure for  $F^{[t]}$ :

- 1 Compute the representative vectors. Number of representative vectors is at most  $d$ : **the Block Orthogonality condition**.
- 2 Compute a witness function  $\omega_j$  of  $S_j \subseteq D^{t-1}$  for each representative vector  $\mathbf{v}_j$ . Existence: **the Mal'tsev condition**.

Computation of these objects uses **the Type Partition condition**.

## Theorem

*If  $\mathcal{L}$  satisfies all these three conditions,  $\#\text{CSP}(\mathcal{L})$  can be solved in polynomial time; otherwise,  $\#\text{CSP}(\mathcal{L})$  is  $\#\text{P}$ -hard.*

- 1 Determine the decidability of the tractability criterion:
  - Given a finite set of complex-valued functions  $\mathcal{L}$ , decide whether  $\mathcal{L}$  satisfies the tractability criterion.
  - Counting graph homomorphisms: in P.
  - Dichotomy for nonnegative #CSP: in NP.
  - Dichotomy for complex #CSP: decidable?

Jin-Yi's take: The land is logically conquered, but one does not really know what treasures lie within.

- 2 Possibility to apply the ideas elsewhere?

Thanks!