

# Optimal Inapproximability of Promise Equations over Finite Groups

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# Systems of equations

3LIN(G)

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$$x_{m1} \cdot x_{m2}^{-1} \cdot x_{m3} = g_m$$

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## Theorem (Goldman, Russell 2002)

*Finding a solution to a satisfiable system of equations over a finite group  $G$  is*

- *solvable in PTIME if  $G$  is Abelian;*
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An algorithm gives an  $\alpha$ -approximation for a maximization problem  $P$  if it outputs a solution of value at least  $\alpha \cdot \text{OPT}(\mathcal{I})$  on every instance  $\mathcal{I}$  of  $P$ .

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For 3-LIN( $G$ ), the *random assignment* over  $G$  gives a  $1/|G|$ -approximation.  
Can we do any better?

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## Approximation

### Theorem (Håstad 2001)

Let  $G$  be a finite **Abelian** group. It is NP-hard to approximate **almost**-satisfiable instances of  $3\text{LIN}(G)$  better than the random assignment threshold over  $G$ .

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### Theorem (Bhargale, Khot 2021)

Let  $G$  be a finite  $\text{SL}(2, \mathbb{F}_q)$  group. It is NP-hard to approximate  $\epsilon$ -satisfiable instances of  $3\text{LIN}(G)$  better than the random assignment threshold over  $[G, G]$  (and this is optimal).

# Constraint Satisfaction Problems

Equations in context

Let  $\mathbb{A}, \mathbb{B}$  be fixed finite relational structures.

## CSP( $\mathbb{A}$ )

Given an input structure  $\mathbb{X}$ , decide if  $\mathbb{X} \rightarrow \mathbb{A}$ .

→ e.g. CSP( $K_3$ ), CSP(3SAT), CSP(3LIN) ...

## PCSP( $\mathbb{A}, \mathbb{B}$ )

Given an input structure  $\mathbb{X}$  with a promise that  $\mathbb{X} \rightarrow \mathbb{A}$ , find a homomorphism  $\mathbb{X} \rightarrow \mathbb{B}$ .

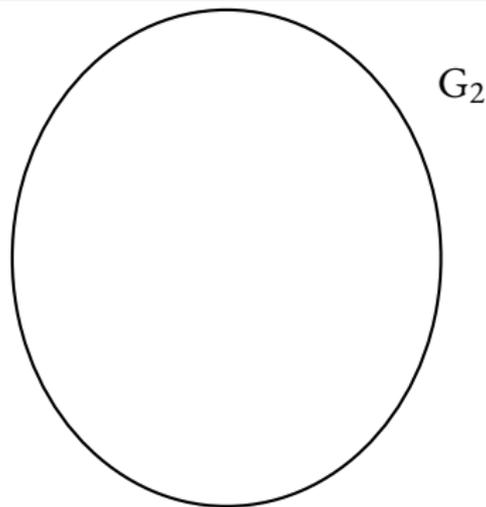
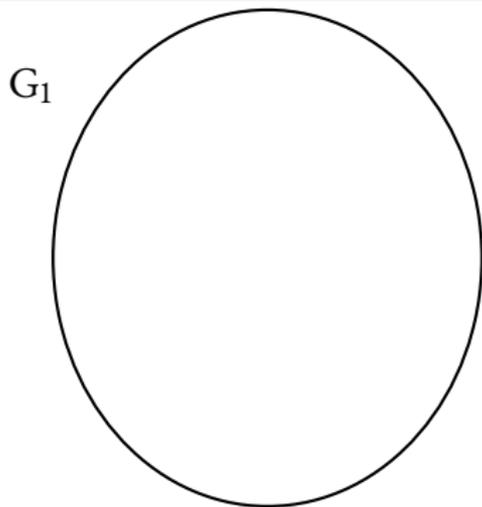
→ e.g. PCSP( $K_3, K_{100}$ ), PCSP(1in3, NAE), ...

# Systems of promise equations

Decision

3-LIN( $G_1, G_2, \varphi$ )

Let  $G_1 \rightarrow G_2$ . Given a system of linear equations that is promised to be satisfiable in  $G_1$ , find a satisfying assignment in  $G_2$ .

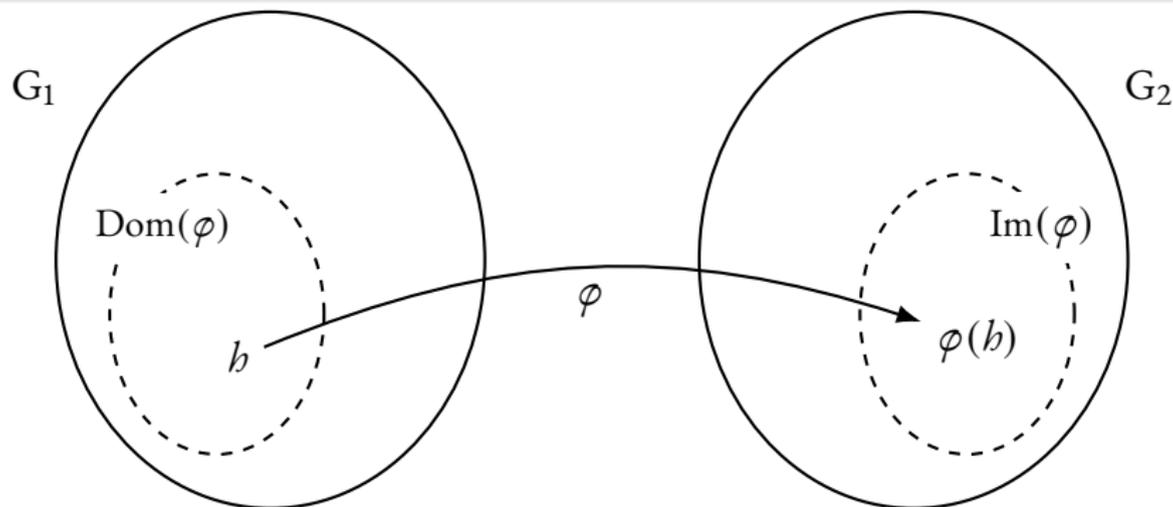


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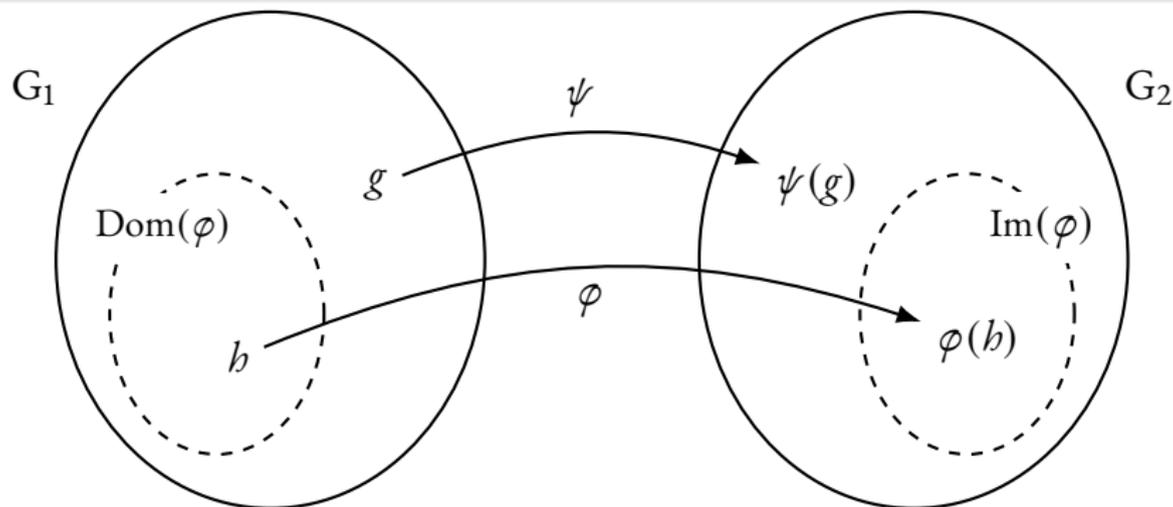


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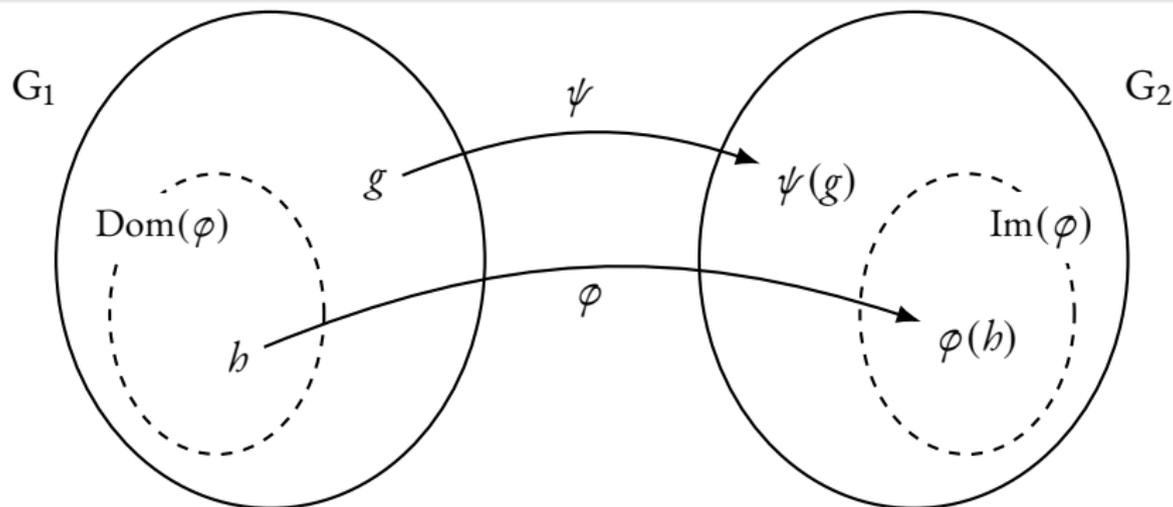


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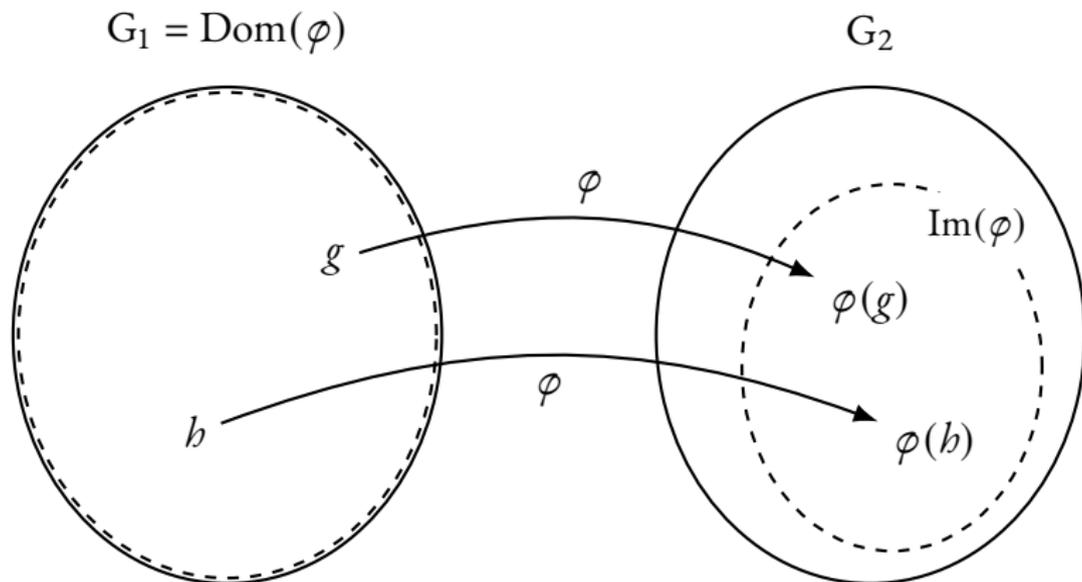
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Equations:  $x y z = b$  for  $b \in \text{Dom}(\varphi)$  interpreted via  $\varphi$  in  $G_2$ .

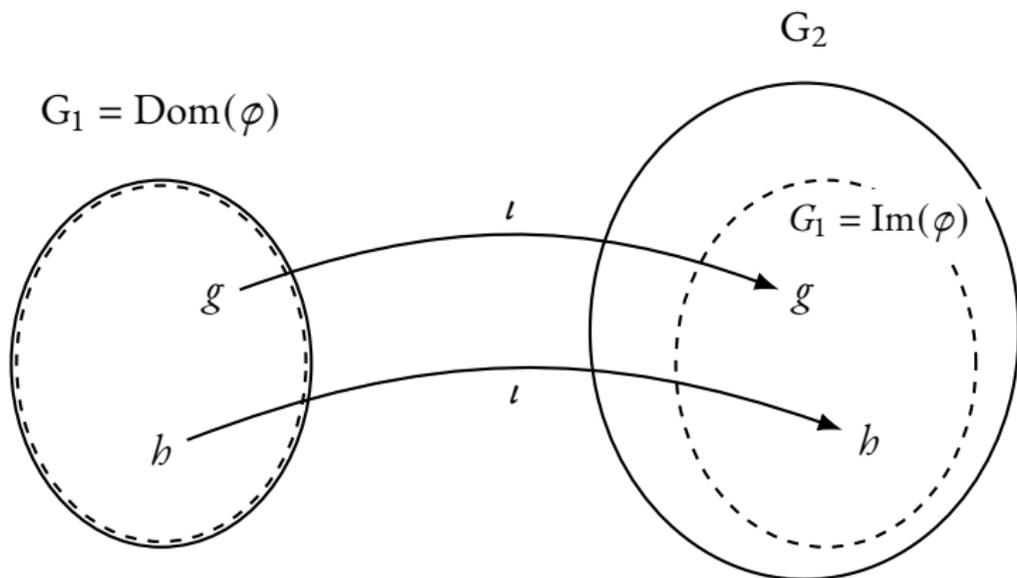
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- *solvable in PTIME if  $G$  is Abelian;*
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## Theorem (Larrauri, Živný 2024)

*3-LIN( $G_1, G_2, \varphi$ ) is*

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# Approximation of promise equations

3-LIN( $G_1, G_2, \varphi, c, s$ )

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Let  $G_1, G_2$  be finite groups,  $\varphi$  a homomorphism between subgroups that extends to a full homomorphism, and  $s \leq c$ . Given a system of equations with constants in  $\text{Dom}(\varphi)$  that is  $c$ -satisfiable in  $G_1$ , find an  $s$ -satisfying assignment in  $G_2$ .

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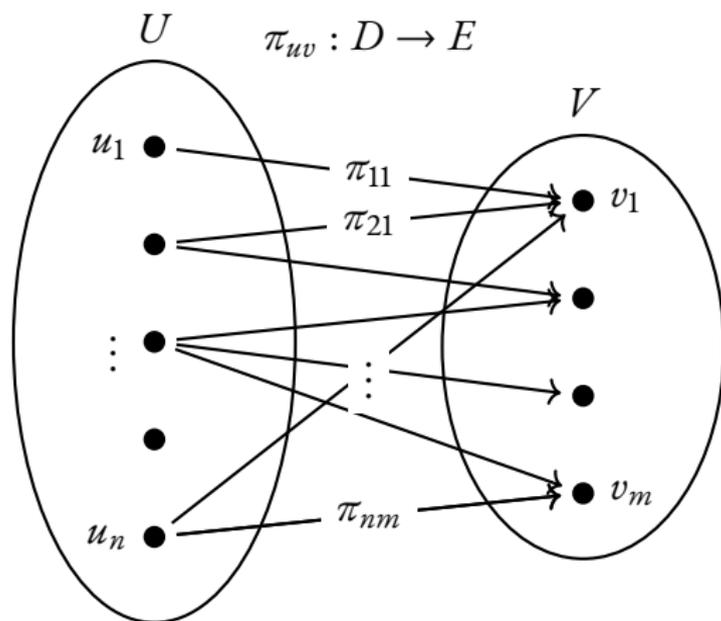
Can we do any better than this?

### Theorem (B., Larrauri, Živný 2025)

*Let  $G_1, G_2$  be finite groups and  $\varphi$  a homomorphism between subgroups that extends to a full homomorphism. Given a system of equations with constants in  $\text{Dom}(\varphi)$  that is almost-satisfiable in  $G_1$ , it is NP-hard to find an assignment in  $G_2$  that satisfies a  $1/|\text{Im}(\varphi)| + \delta$  fraction of the equations.*

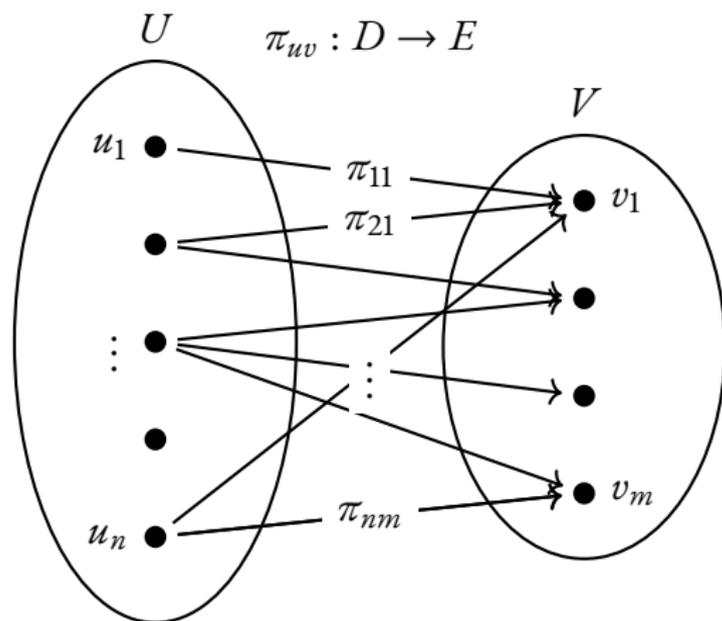
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## Hardness of Gap Label Cover



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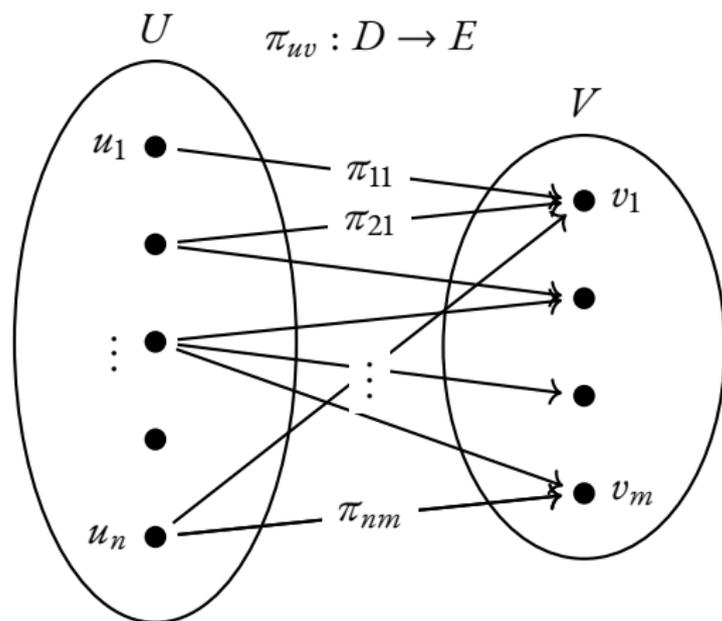


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Distinguish  $c$ -satisfiable  
from not  $s$ -satisfiable  
label cover instances.

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label cover instances.

Theorem (PCP + Parallel repetition)

For every  $\alpha > 0$  there exists finite sets  $D, E$  such that  $GLC_{D,E}(1, \alpha)$  is NP-hard.

# The reduction

## Dictatorship test

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- (2) Sample elements  $\mathbf{a} \in G_1^E$ ,  $\mathbf{b} \in G_1^D$  uniformly at random.
- (3) Sample  $\eta \in G_1^D$  so that for each  $d \in D$ , independently,  
 $\eta(d) = 1_{G_1}$  with probability  $1 - \epsilon$ , and  
 $\eta(d)$  selected uniformly at random from  $G_1$  with probability  $\epsilon$ .  
Set  $\mathbf{c} = \mathbf{b}^{-1}(\mathbf{a} \circ \pi_{uv})^{-1}\eta \in G_1^D$ .

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Set  $\mathbf{c} = \mathbf{b}^{-1}(\mathbf{a} \circ \pi_{uv})^{-1} \eta \in G_1^D$ .
- (4) Test  $A_\varphi(\mathbf{a})B(\mathbf{b})B(\mathbf{c}) = 1_{G_2}$ .

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Folding over  $\varphi$  changes equations

$$x \cdot y \cdot z = 1 \mapsto x' \cdot y \cdot z = h$$

such that

$$A(x')B(y)B(z) = h \iff A_\varphi(x)B(y)B(z) = 1_{G_2}$$

and  $h \in \text{Dom}(\varphi)$ !

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$$\begin{aligned} A_v(\mathbf{a})B_u(\mathbf{b})B_u(\mathbf{c}) &= \mathbf{a}(f(v))\mathbf{b}(f(u))\mathbf{b}^{-1}(f(u))(\mathbf{a} \circ \pi_{uv})^{-1}(f(u))\eta(f(u)) \\ &= \mathbf{a}(\pi_{uv}(f(u)))(\mathbf{a}(\pi_{uv}(f(u))))^{-1}\eta(f(u)) \\ &= \eta(f(u)) = 1_{G_1} \text{ with probability } \geq 1 - \epsilon. \end{aligned}$$

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- We use **Fourier analysis** over non-Abelian groups + **representation theory** for the soundness analysis.
  - Given better-than-random assignments  $A : G_1^E \rightarrow G_2$ ,  $B : G_1^D \rightarrow G_2 \dots$
  - We analyse complex-valued functions

$$\mathcal{A} := \omega \circ A : G_1^E \rightarrow \mathbb{C}^{d_\omega \times d_\omega}, \quad \mathcal{B} := \omega \circ B : G_1^D \rightarrow \mathbb{C}^{d_\omega \times d_\omega}$$

where  $\omega$  is a ‘nice’ representation of  $G_2$ .

$$\text{tr } \mathbb{E} \left[ \mathcal{A}(\mathbf{a})(\mathcal{B} * \mathcal{B})((\mathbf{a}\pi)^{-1}\mu) \right]$$

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 = & \operatorname{tr} \mathbb{E} \left[ \left( \sum_{\tau \in \widehat{G}_1^E} \dim_{\tau} \widehat{\mathcal{A}}_{\tau} \tau(\mathbf{a}) \right) \left( \sum_{\rho \in \widehat{G}_1^D} \dim_{\rho} \overline{(\mathcal{B} * \mathcal{B})}_{\rho} \rho((\mathbf{a}\pi)^{-1}\mu) \right) \right] \\
 = & \operatorname{tr} \mathbb{E} \left[ \widehat{\mathcal{A}}_1 \left( \sum_{\rho \in \widehat{G}_1^D} \dim_{\rho} \overline{(\mathcal{B} * \mathcal{B})}_{\rho} \rho((\mathbf{a}\pi)^{-1}\mu) \right) \right] \leq \langle 1, \chi_{\omega} | \operatorname{Im}(\varphi) \rangle \\
 + & \operatorname{tr} \mathbb{E} \left[ \left( \sum_{\tau \neq 1} \dim_{\tau} \widehat{\mathcal{A}}_{\tau} \tau(\mathbf{a}) \right) \left( \sum_{|\rho| \text{ large}} \dim_{\rho} \overline{(\mathcal{B} * \mathcal{B})}_{\rho} \rho((\mathbf{a}\pi)^{-1}\mu) \right) \right] \leq \dim_{\omega} \delta / 2 \\
 + & \operatorname{tr} \mathbb{E} \left[ \left( \sum_{\tau \neq 1} \dim_{\tau} \widehat{\mathcal{A}}_{\tau} \tau(\mathbf{a}) \right) \left( \sum_{|\rho| \text{ small}} \dim_{\rho} \overline{(\mathcal{B} * \mathcal{B})}_{\rho} \rho((\mathbf{a}\pi)^{-1}\mu) \right) \right] \text{ 'large' }
 \end{aligned}$$

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 \end{aligned}$$

→ Allows us to devise a good sampling strategy for GLC:

- Sample reps of small degree according to Fourier coefficients;
- Sample u.a.r. from non-trivial subrepresentations.

# The big picture

	3-LIN(G)	3-LIN( $G_1, G_2, \varphi$ )
Abelian, almost-sat	$1/ G $ [Håstad'01]	$1/ \text{Im}(\varphi) $ [this work]
non Abelian, almost-sat	$1/ G $ [EHR'04]	$1/ \text{Im}(\varphi) $ [this work]
non Abelian, satisfiable	$1/ [G, G] $ [BK'21]	

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3-LIN( $G_1, G_2, \varphi, c, s$ ) is a special case of a **Valued Promise CSP**.

# Promise Valued CSPs

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**PCSP( $\mathbb{A}, \mathbb{B}, c, s$ )**

Given instance  $\mathbb{X}$  that is  $c$ -satisfiable in  $\mathbb{A}$ , find an  $s$ -satisfying assignment in  $\mathbb{B}$ .

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## **Algebraic Approach to Approximation**

*Valued minion homomorphisms give reductions between Valued Promise CSPs.*

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## Algebraic Approach to Approximation

*Valued minion homomorphisms give reductions between Valued Promise CSPs.*

Theorem (B., Larrauri, Živný 2025)

*There is a valued minion homomorphism*

$$\text{Plu}(3\text{-LIN}(G_1, G_2, \varphi, 1 - \varepsilon, 1/|\text{Im}(\varphi)| + \delta)) \rightsquigarrow \text{Plu}(\text{GLC}(1, \alpha)).$$

Thank You!