

Prague Strategies

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The Prague Theorem

Theorem (Barto, Kozik 2009)

Let \mathbf{A} be an idempotent algebra. TFAE

- ▶ *\mathbf{A} is an $\text{SD}(\wedge)$ algebra
(= lies in a variety omitting **1** and **2**)*
- ▶ *Every Prague strategy over \mathbf{A} has a solution*

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Plan:

- ▶ k -intersection property
- ▶ $SD(\wedge)$
- ▶ $CSP(\mathbf{A})$
- ▶ (k, l) -minimal instance
- ▶ Prague strategy

All algebras are finite and idempotent

Definition (k -equal relations)

$R_1, R_2 \subseteq A^n$ are *k -equal*, if $\forall J \subseteq [n], |J| \leq k$, the projections of R_1 and R_2 to J are equal.

k-intersection property

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Definition (*k*-intersection property, Valeriote)

A finite algebra \mathbf{A} satisfies the *k*-intersection property, if $\forall n$ every collection of pairwise *k*-equal non-empty subuniverses $R_1, \dots, R_m \leq \mathbf{A}^n$ has nonempty intersection.

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Observation

$\mathbf{B} \in \text{HSP}(\mathbf{A})$. Then

\mathbf{A} has the k -intersection property $\Rightarrow \mathbf{B}$ has the k -intersection prop.

k -IP, modules are bad

Observation

If \mathbf{A} is a reduct of a module and $|A| > 1$, then \mathbf{A} fails the k -intersection property for every k .

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Proof.

For $a \in A$ let

$$R_a = \{(a_1, \dots, a_{k+1}) : a_1 + a_2 + \dots + a_{k+1} = a\}$$

Clearly

- ▶ R_a is a subuniverse of \mathbf{A}^{k+1}
- ▶ any projection to less than $k + 1$ coordinates is full
- ▶ if $a \neq b$ then $R_a \cap R_b = \emptyset$



k -IP, a necessary condition and a conjecture

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Conjecture (Valeriote)

The other implication is also true.

Theorem (Hobby, Maróti, McKenzie, Valeriote, Willard)

Let \mathbf{A} be an algebra. TFAE

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- ▶ $\text{HSP}(\mathbf{A})$ is congruence meet semi-distributive, i.e.

if $\mathbf{B} \in \text{HSP}(\mathbf{A})$, $\alpha, \beta_1, \beta_2 \in \text{Con}(\mathbf{B})$

then $\alpha \wedge \beta_1 = \alpha \wedge \beta_2 \Rightarrow \alpha \wedge (\beta_1 \vee \beta_2) = \alpha \wedge \beta_1$

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Definition

\mathbf{A} is SD(\wedge), if it satisfies the equivalent conditions above

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- ▶ If \mathbf{A} has a short (3-terms) chain of Jónsson terms, then \mathbf{A} has the 2-intersection property [Kiss, Valeriote](#) and 2 is the optimal number

Definition (CSP(\mathbf{A}))

Let \mathbf{A} be an algebra. An *instance of CSP(\mathbf{A})* is a pair (V, \mathcal{C}) , where

- ▶ V is a finite set (elements are called *variables*)
- ▶ \mathcal{C} is a finite set of *constraints*

Constraint is a subuniverse C of \mathbf{A}^D , where

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The aim is to find a solution fast (in poly-time).

(k, l) -minimal instance

Definition (\pm Bulatov, Jeavons)

Let $k \leq l$ be natural numbers.

An instance (V, \mathcal{C}) of $\text{CSP}(\mathbf{A})$ is called (k, l) -minimal if

- ▶ Every l -element subset of V is a subset of the scope of some constraint in \mathcal{C}
- ▶ For every $J \subseteq V$, $|J| \leq k$ and every pair $C_1, C_2 \in \mathcal{C}$ whose scopes contain J , the projections of C_1 and C_2 onto J are the same.

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An instance (V, \mathcal{C}) is called k -minimal, if it is (k, k) -minimal.

Observation

If $k' \leq k$ and $l' \leq l$ then (k, l) -minimal instance is (k', l') -minimal.

Observation

Every instance of CSP(**A**) can be converted into an equivalent (k, l) -minimal instance in poly-time.

(Two instances are equivalent if they have the same set of solutions.)

Bounded relational width

Observation

Every instance of $\text{CSP}(\mathbf{A})$ can be converted into an equivalent (k, l) -minimal instance in poly-time.

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Definition

\mathbf{A} has *relational width* (k, l) if every (k, l) -minimal instance, whose constraints are non-empty, has a solution.

\mathbf{A} has *bounded relational width* if it has relational width (k, l) for some k, l .

The bounded relational width conjecture

Theorem (Larose, Zádori, Bulatov)

If \mathbf{A} has bounded relational width, then \mathbf{A} is an $\text{SD}(\wedge)$ algebra.

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- ▶ If \mathbf{A} has a short chain of Jónsson terms (4 terms), then \mathbf{A} has “bounded width” [Carvalho, Dalmau, Marković, Maróti](#)

Corollaries of the Prague Theorem

Corollary

If \mathbf{A} is an $SD(\wedge)$ algebra, then \mathbf{A} has relational width $(2, 3)$.
(The parameters $(2, 3)$ are optimal.)

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If \mathbf{A} is an $\text{SD}(\wedge)$ algebra, then \mathbf{A} satisfies the 2-intersection property. (Recall that 2 is optimal.)

Proof.

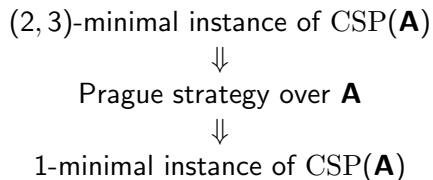
- ▶ Let $R_1, \dots, R_m \leq \mathbf{A}^n$ be nonempty and 2-equal
- ▶ Let $V = [n]$, $\mathcal{C} = \{R_1, \dots, R_m\}$
- ▶ (V, \mathcal{C}) is a $(2, n)$ -minimal instance of $\text{CSP}(\mathbf{A})$



No idea about the title

I am finally going to introduce Prague strategies.

Comparison with known notions:



Patterns

Let (V, \mathcal{C}) be an instance of $\text{CSP}(\mathbf{A})$

Patterns

For $x, y \in V$ and $C \in \mathcal{C}$ and $a, b \in A$ we write $a \xrightarrow{x, y, C} b$, if

- ▶ x, y are in the scope of C
- ▶ The mapping $x \rightarrow a, y \rightarrow b$ is in the projection of C to $\{x, y\}$

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Definition

A *pattern* w is a tuple (x_1, C_1, \dots) :

$$x_1 \xrightarrow{C_1} x_2 \xrightarrow{C_2} \dots \xrightarrow{C_i} x_{i+1},$$

where $x_j \in V$ and $C_j \in \mathcal{C}$.

We write $a \xrightarrow{w} b$, if there exist $a = a_1, a_2, \dots, a_{i+1} = b$ such that

$$a = a_1 \xrightarrow{x_1, x_2, C_1} a_2 \xrightarrow{x_2, x_3, C_2} a_3 \rightarrow \dots \rightarrow a_i \xrightarrow{x_i, x_{i+1}, C_i} a_{i+1} = b$$

The *scope* of w is $[[w]] = \{x_1, \dots, x_{i+1}\}$

Prague strategy

If patterns w_1, w_2 start and end with the same variable x , we can form their concatenation $w_1 \circ w_2$.

$$w^K = w \circ w \circ \dots \circ w \text{ (} K\text{-times)}$$

Definition (!!!!!!!)

A Prague strategy over \mathbf{A} is an instance (V, \mathcal{C}) of $\text{CSP}(\mathbf{A})$ such that

- ▶ (V, \mathcal{C}) is 1-minimal
- ▶ For every $x \in V$,
 - every pattern v starting and ending with x ,
 - every $a, b \in A$ such that $a \xrightarrow{v} b$ and
 - every pattern w starting and ending with x s.t. $[[v]] \subseteq [[w]]$,there exists a natural number K such that $a \xrightarrow{w^K} b$

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Observation

Every (2, 3)-min. instance of $\text{CSP}(\mathbf{A})$ is a Prague strategy over \mathbf{A} .

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Implication \uparrow follows from Larose, Zádori, Bulatov



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Two cases

- ▶ When we have a proper absorbing set of the projection to some singleton



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Two cases

- ▶ When we have a proper absorbing set of the projection to some singleton
- ▶ When we don't have ...



thAnk yoU FOr youR ATtentiON!

ThANK you fOR your atTENTion?

thank you foR yoU AtteNTion?