
New Widening Operators for Convex Polyhedra

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<http://www.cs.unipr.it/pp1/>

MOTIVATIONS

- **Linear Relation Analysis** is a key component of many static analysis and (semi-) automatic verification tools.
- Since it has infinite chains, the domain of convex polyhedra has to be provided with **widening operators**.
- The **standard widening** (**Cousot and Halbwachs, POPL'78**) is the one and only champion: since then, no challenger has been proposed.
- But some applications need **more precision**. Solutions include:
 - ① the **widening delay** technique (**Cousot, '81**);
 - ② the **widening 'up to'** technique (**Halbwachs, CAV'93**);
 - ③ various **extrapolation operators** (no **convergence guarantee**).
- **Our goal**: *provide a **framework** for the definition of **new widening operators** on convex polyhedra improving upon the **precision** of the **standard widening**.*

PLAN OF THE TALK

- ① Problems in the Approximated Computation of Semantics
- ② Widening Operators Are the Solution
- ③ The Standard Widening on Convex Polyhedra
- ④ Some Techniques to Obtain Better Approximations
- ⑤ A New Framework for Improving Upon a Fixed Widening
- ⑥ Heuristic Techniques Improving the Standard Widening
- ⑦ Experimental Results
- ⑧ Conclusion

COMPUTING THE CONCRETE SEMANTICS

```
x := 0; b := true;
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while (b) do
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    x := x+2;
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    read(b);
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`while (b) do`

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`x := x+2;`

`read(b);`

`endwhile`

Let $\mathcal{F}: \wp(\mathbb{R}) \rightarrow \wp(\mathbb{R})$ be such that

$$\mathcal{F}(X) \stackrel{\text{def}}{=} \{0\} \cup \{n+2 \mid n \in X\}$$

The concrete semantics S is computed as the least fixpoint of \mathcal{F} on the complete lattice $\langle \wp(\mathbb{R}), \subseteq, \emptyset, \mathbb{R}, \cup, \cap \rangle$.

COMPUTING THE CONCRETE SEMANTICS

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$$\mathcal{F}(X) \stackrel{\text{def}}{=} \{0\} \cup \{n + 2 \mid n \in X\}$$

`while (b) do`

`$x \in S = 2\mathbb{N}$`

`x := x+2;`

$$X_0 = \emptyset;$$

$$X_1 = \mathcal{F}(\emptyset) = \{0\};$$

$$X_2 = \mathcal{F}(\mathcal{F}(\emptyset)) = \{0, 2\};$$

...

`read(b);`

`endwhile`

$$S = X_\omega = \text{lfp}(\mathcal{F}) = 2\mathbb{N}.$$

THE DOMAIN \mathbb{CP}_n OF CLOSED CONVEX POLYHEDRA

A lattice $\langle \mathbb{CP}_n, \subseteq, \emptyset, \mathbb{R}^n, \uplus, \cap \rangle$, with **infinite chains**.

Constraint Representation: $\mathcal{P} = \text{con}(\mathcal{C})$

- \mathcal{C} is a finite set of **linear non-strict inequality** (resp., **equality**) constraints.
- No redundant constraint + max number of equalities \implies **minimal form**.
- Inequalities orthogonal wrt equalities \implies **orthogonal form**.

Generator Representation: $\mathcal{P} = \text{gen}(\mathcal{G})$

- $\mathcal{G} = (L, R, P)$, where
 - P is a finite set of **points** of \mathcal{P} ;
 - R is a finite set of **rays** (directions of infinity) of \mathcal{P} ;
 - L is a finite set of **lines** (bidirectional rays) of \mathcal{P} .
- No redundant generator + max number of lines \implies **minimal form**.
- Points and rays orthogonal wrt lines \implies **orthogonal form**.

APPROXIMATING THE SEMANTICS ON \mathbb{CP}_1

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x := 0; b := true;
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Correctness of $\mathcal{F}^\#$ wrt \mathcal{F} :

$$X \subseteq \mathcal{P} \implies \mathcal{F}(X) \subseteq \mathcal{F}^\#(\mathcal{P}).$$

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The concrete semantics $S \in \mathbb{R}$ is approximated by computing a post-fixpoint $Q \in \mathbb{CP}_1$ of the abstract semantic function $\mathcal{F}^\#$.

APPROXIMATING THE SEMANTICS ON \mathbb{CP}_1

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$$\mathcal{Q} = [0, +\infty).$$

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- ① The “limit” of the approximated computation may not be representable in the abstract domain (e.g., a circle is not a polyhedron);

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- ③ Even when the computation is intrinsically finite, it may be **practically unfeasible** if it requires too many approximated iterations; for instance,

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Widening operators try to solve all of these problems at once.

DEFINITION OF WIDENING OPERATOR

A variant of the classical one (see [Cousot and Cousot, PLILP'92](#)):

→ Let $\langle L, \sqsubseteq, \perp, \sqcup \rangle$ be a join-semi-lattice. Then, the operator

$\nabla: L \times L \rightarrow L$ is a **widening** on L if

① $\forall x, y \in L : x \sqsubseteq y \implies y \sqsubseteq x \nabla y$;

② for all increasing chains $y_0 \sqsubseteq y_1 \sqsubseteq \dots$, the chain defined by

$x_0 \stackrel{\text{def}}{=} y_0, \dots, x_{i+1} \stackrel{\text{def}}{=} x_i \nabla (x_i \sqcup y_{i+1}), \dots$ is not strictly increasing.

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→ The upward iteration sequence with widenings (starting from $x_0 = \perp$)

$$x_{i+1} = \begin{cases} x_i, & \text{if } \mathcal{F}^\#(x_i) \sqsubseteq x_i; \\ x_i \nabla (x_i \sqcup \mathcal{F}^\#(x_i)), & \text{otherwise;} \end{cases}$$

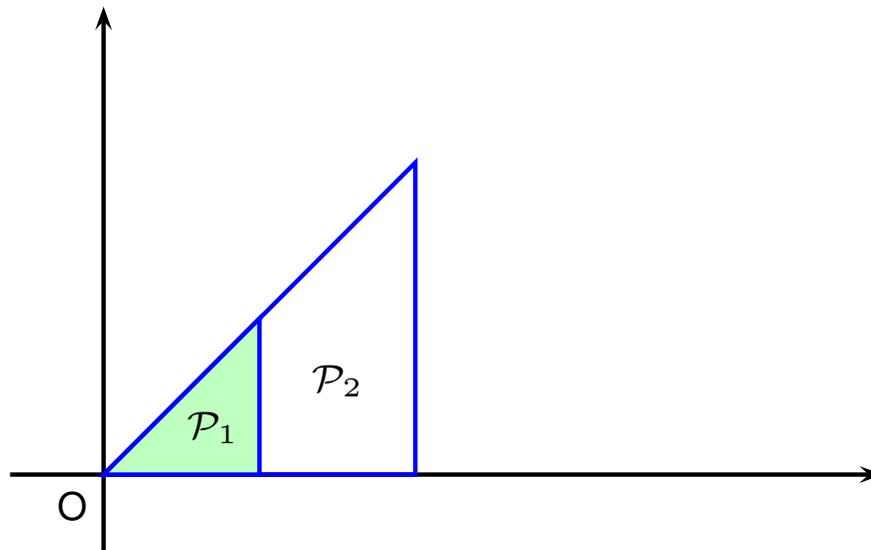
converges (to a post-fixpoint of $\mathcal{F}^\#$) after a **finite number of iterations**.

THE STANDARD WIDENING ∇_s ON \mathbb{CP}_n

- Initially proposed in **Cousot and Halbwachs, POPL'78**.
- Intuitively, $\mathcal{P}_1 \nabla_s \mathcal{P}_2$ *is defined by all the non-redundant constraints of \mathcal{P}_1 that are also satisfied by \mathcal{P}_2 .*

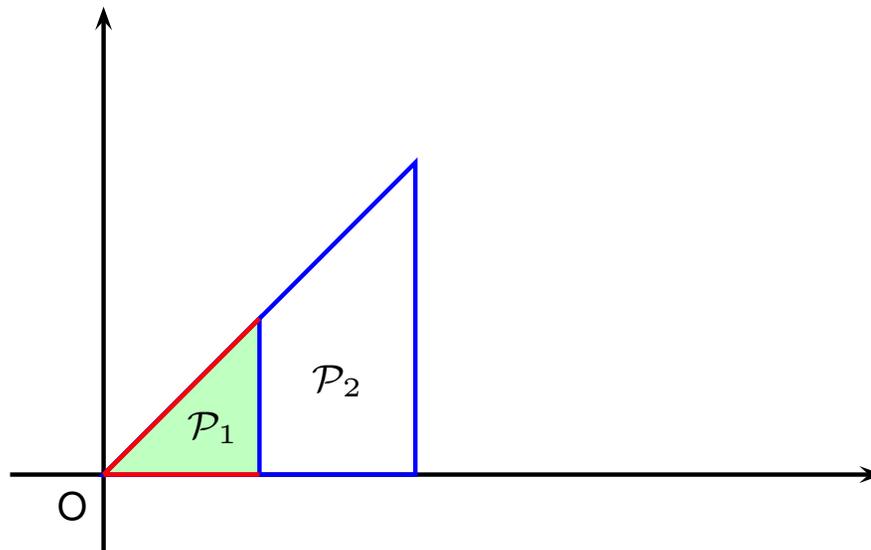
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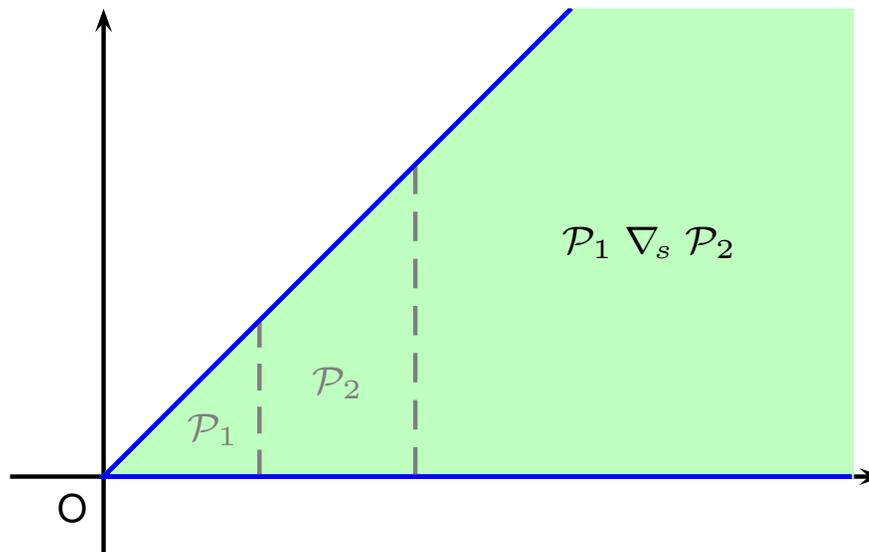
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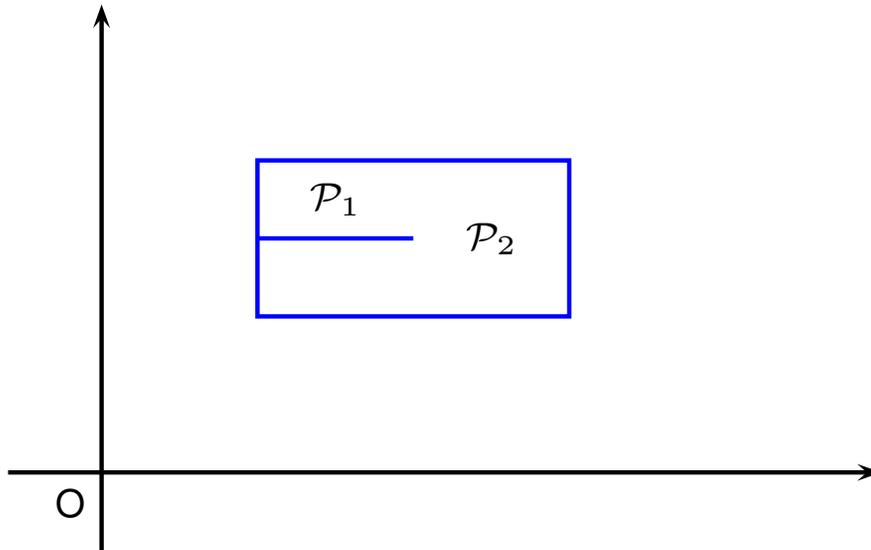


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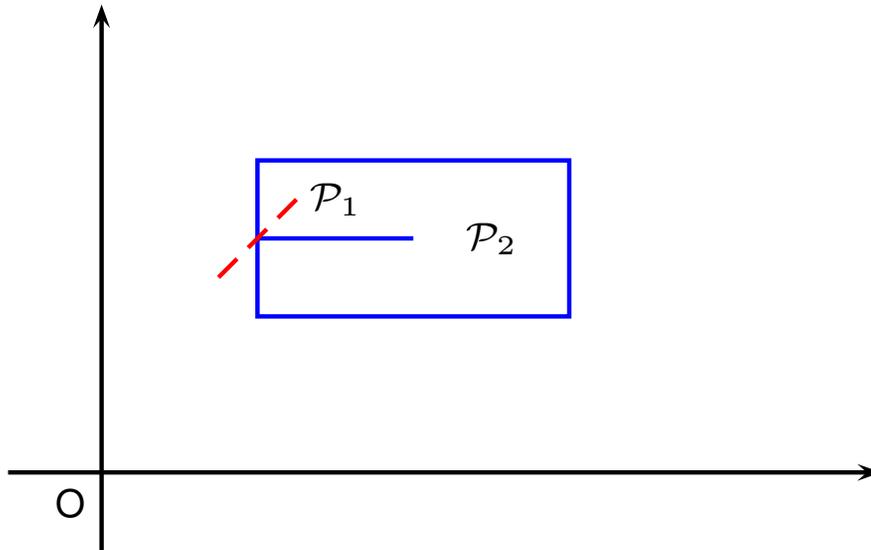
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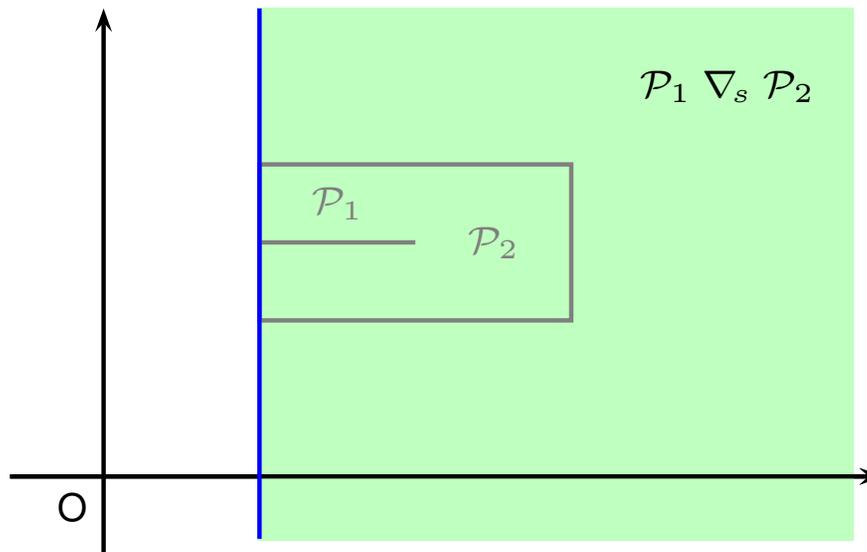
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- For an increasing number of applications, this precision level is not sufficient. **Can we further improve upon the precision of the standard widening?** (Perhaps, trading some efficiency.)

∇ -COMPATIBLE LIMITED GROWTH ORDERING

- Let $\langle L, \sqsubseteq, \perp, \sqcup \rangle$ be a join-semi-lattice.
- A **limited growth ordering** (lgo) is the strict version of a finitely computable preorder relation that satisfies the ascending chain condition on L .
 - ① preorder: reflexive and transitive;
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- A **∇ -compatible lgo** formalizes the notion of **computable convergence guarantee** for the widening ∇ .

A FRAMEWORK FOR IMPROVING UPON A FIXED WIDENING

Suppose that

- $\nabla: L \times L \rightarrow L$ is a widening on the join-semi-lattice $\langle L, \sqsubseteq, \perp, \sqcup \rangle$;
- $\curvearrowright \subseteq L \times L$ is a ∇ -compatible lgo;
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For all $x, y \in L$ such that $x \sqsubseteq y$, define

$$x \tilde{\nabla} y \stackrel{\text{def}}{=} \begin{cases} h(x, y), & \text{if } x \curvearrowright h(x, y) \sqsubseteq x \nabla y; \\ x \nabla y, & \text{otherwise.} \end{cases}$$

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- Then $\tilde{\nabla}$ is a **widening operator at least as precise as ∇** .

A FINE-GRAINED LGO ON $\mathbb{C}\mathbb{P}_n$

- Variant of a well-founded preorder defined in [Besson *et al.*, SAS'99](#). It is obtained as the **lexicographic product** of five lgo's.

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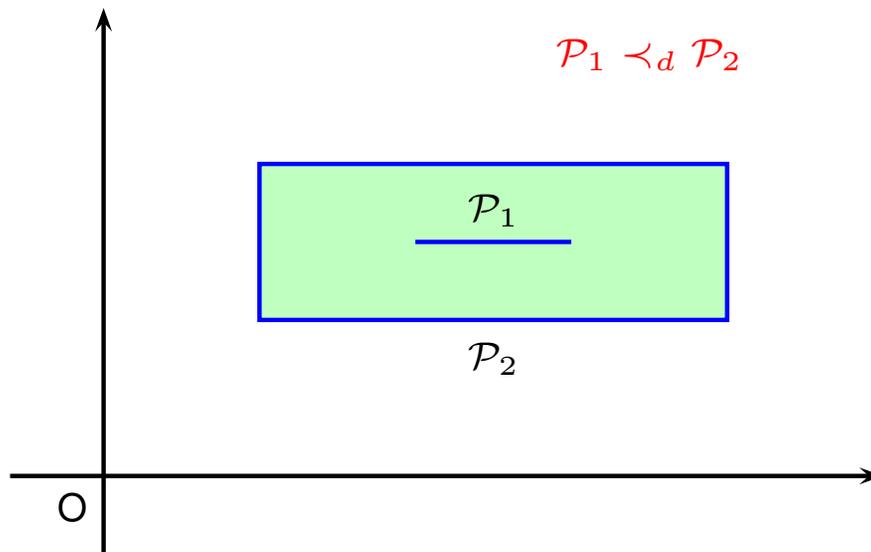
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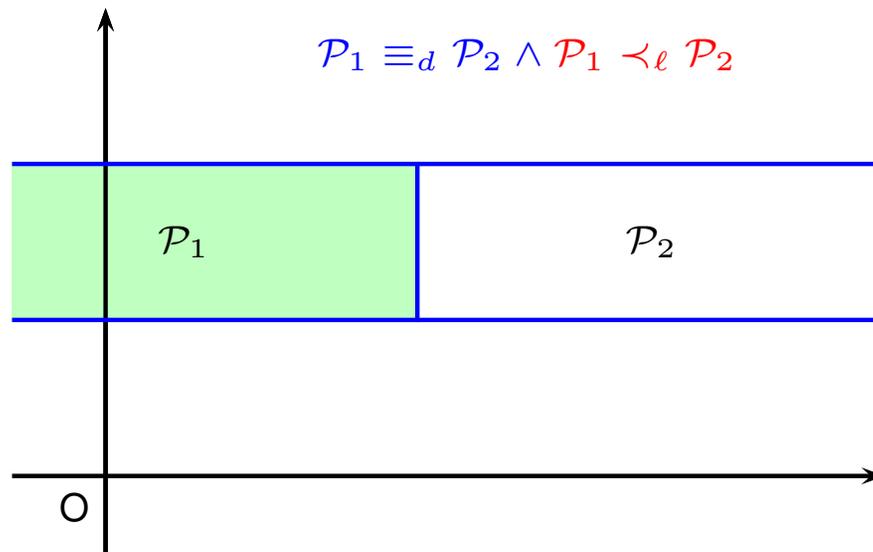
→ We denote by \curvearrowright_n the strict version of the lexicographic product

$$\mathcal{P}_1 \preceq_n \mathcal{P}_2 \quad \stackrel{\text{def}}{\iff} \quad \mathcal{P}_1 \preceq_{dlcpr} \mathcal{P}_2.$$

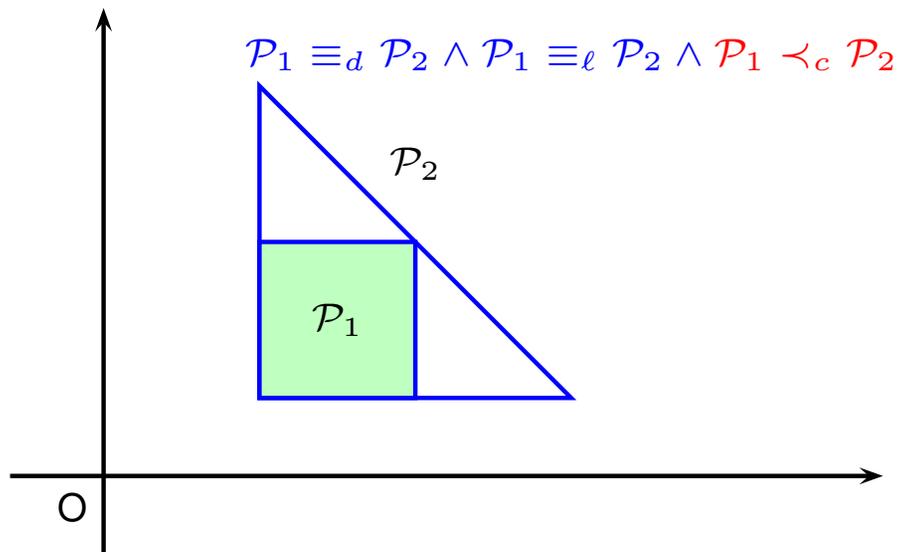
EXAMPLES FOR $\mathcal{P}_1 \curvearrowright_n \mathcal{P}_2$: CASE 1



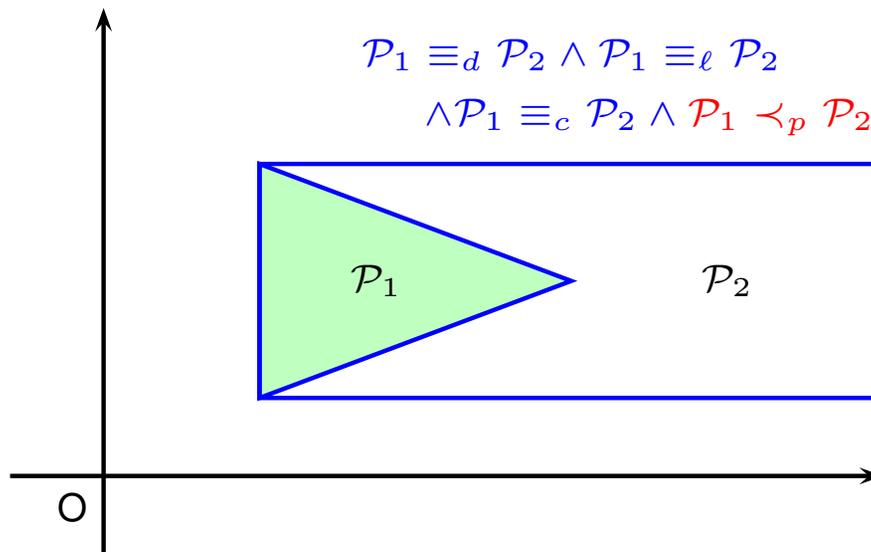
EXAMPLES FOR $\mathcal{P}_1 \approx_n \mathcal{P}_2$: CASE 2



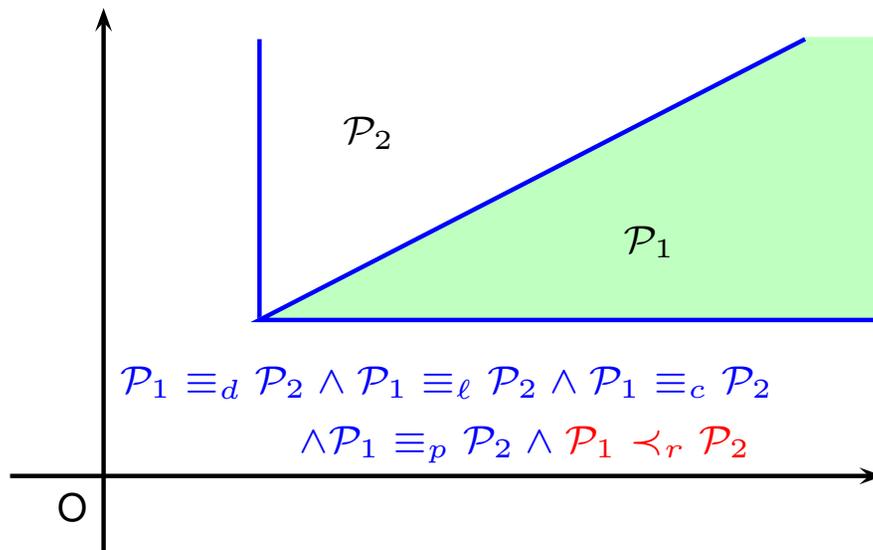
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EXAMPLES FOR $\mathcal{P}_1 \curvearrowright_n \mathcal{P}_2$: CASE 4



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The key result.

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- For any **upper bound operator** $h: \mathbb{CP}_n \times \mathbb{CP}_n \rightarrow \mathbb{CP}_n$, the framework will return a proper widening operator on \mathbb{CP}_n improving on the standard widening.
- In our attempt to improve precision, we can consider any finite set of such heuristic techniques: our new widening will use **four** upper bounds.

1ST HEURISTICS: DO NOT WIDEN

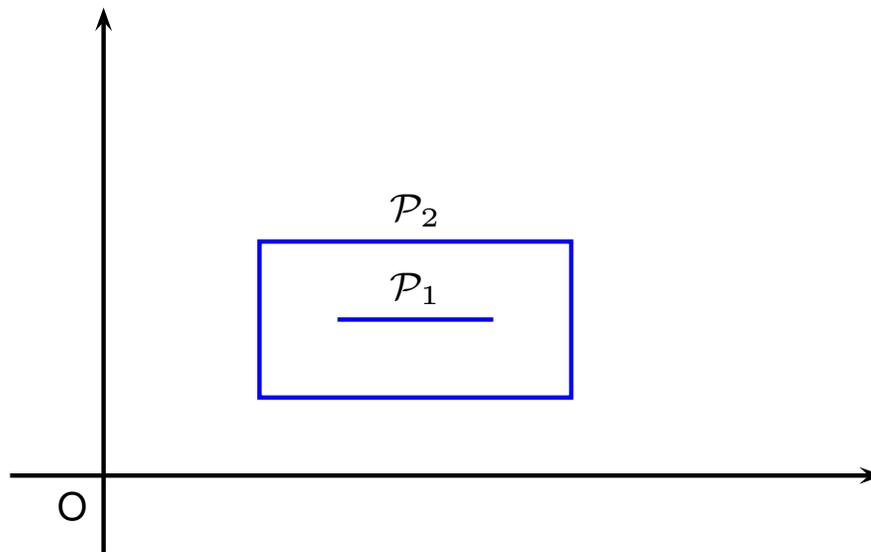
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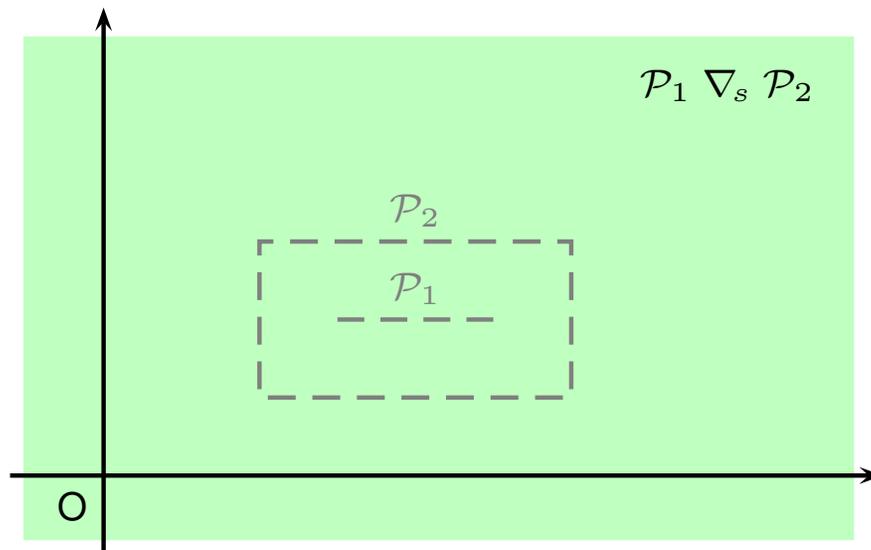
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- Applicable whenever $\mathcal{P}_1 \sqsupseteq \mathcal{P}_2$.
- **No precision loss**: to be tried before all other techniques.
- Already suggested by **Cousot and Cousot, PLILP'92**.

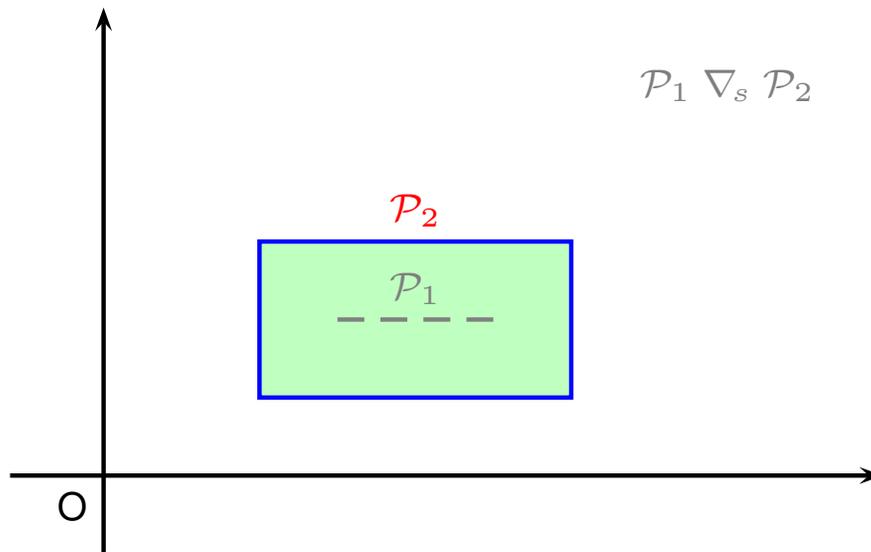
STANDARD WIDENING VS. DO NOT WIDEN (I)



STANDARD WIDENING VS. DO NOT WIDEN (II)



STANDARD WIDENING VS. DO NOT WIDEN (III)



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Let h be the least upper bound, so that $h(\mathcal{P}_1, \mathcal{P}_2) = \mathcal{P}_2$.

- Applicable whenever $\mathcal{P}_1 \curvearrowright \mathcal{P}_2$.
- No precision loss: to be tried before all other techniques.
- Already suggested by Cousot and Cousot, PLILP'92.
- **All the other techniques** may safely assume $\mathcal{P}_1 \not\curvearrowright \mathcal{P}_2$.
- Since by hypothesis $\mathcal{P}_1 \subseteq \mathcal{P}_2$, we can also assume

$$\text{aff.hull}(\mathcal{P}_1) = \text{aff.hull}(\mathcal{P}_2),$$

$$\text{lin.space}(\mathcal{P}_1) = \text{lin.space}(\mathcal{P}_2).$$

2ND HEURISTICS: COMBINING CONSTRAINTS

Let $h_c(\mathcal{P}_1, \mathcal{P}_2) \stackrel{\text{def}}{=} \text{con}(\mathcal{C}_\oplus) \cap (\mathcal{P}_1 \nabla_s \mathcal{P}_2)$, where

→ \mathcal{C}_∇ are the constraints of the standard widening;

→ $\mathcal{C}_\oplus \stackrel{\text{def}}{=} \left\{ \oplus(\mathcal{C}_p) \left| \begin{array}{l} p \in P_1, \text{sat_con}(p, \text{ineq}(\mathcal{C}_\nabla)) = \emptyset, \\ \mathcal{C}_p = \text{sat_con}(p, \text{ineq}(\mathcal{C}_2)) \neq \emptyset \end{array} \right. \right\}$.

→ \oplus is a (deliberately left unspecified) **convex combination**.

Informally, we ensure that each **non-redundant point** $p \in \mathcal{P}_1$ that was lying on a **facet of** \mathcal{P}_2 will still lie on a **facet of** $h_c(\mathcal{P}_1, \mathcal{P}_2)$.

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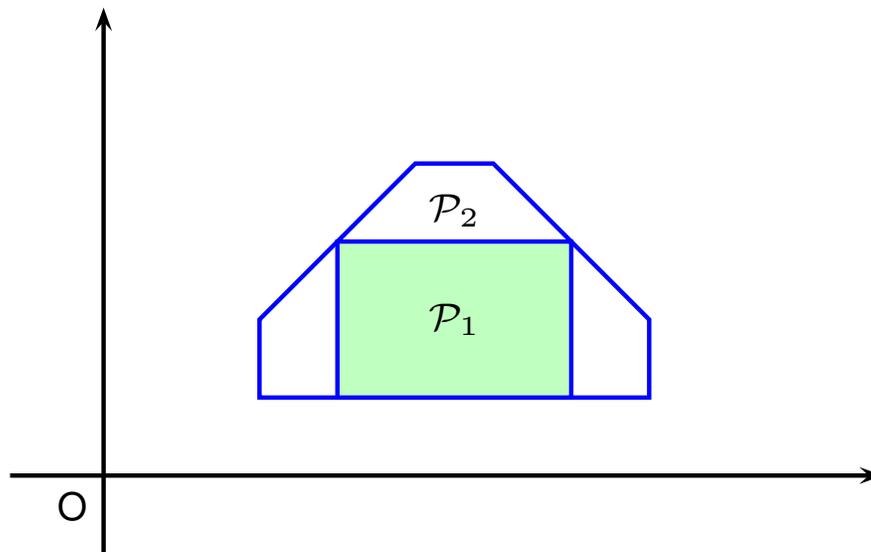
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→ Besson et al., SAS'99 suggest to **average** the constraints in \mathcal{C}_p .

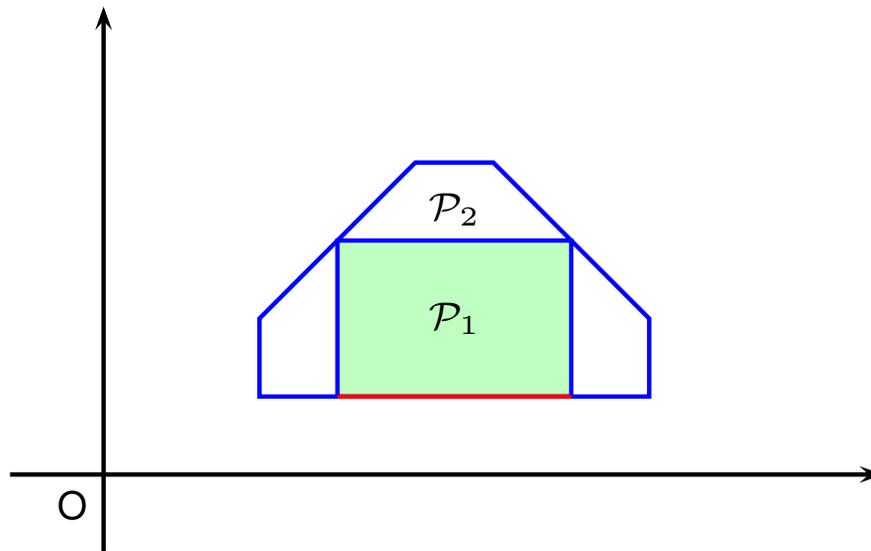
→ Afterall, the choice of \oplus is arbitrary: we opted for a simpler combination.

→ A similar heuristics, with **no convergence guarantee**, was proposed by Henzinger et al., CDC'01.

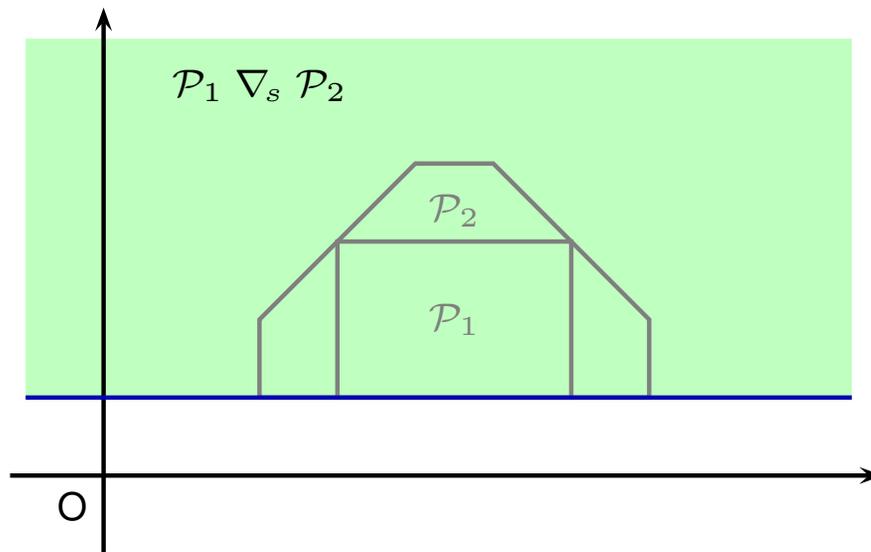
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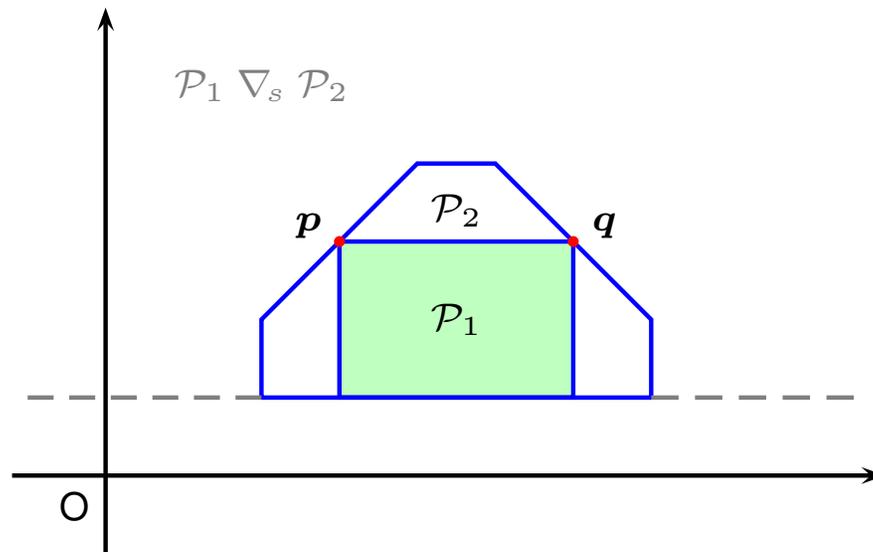
STANDARD WIDENING VS. COMBINING CONSTRAINTS (II)



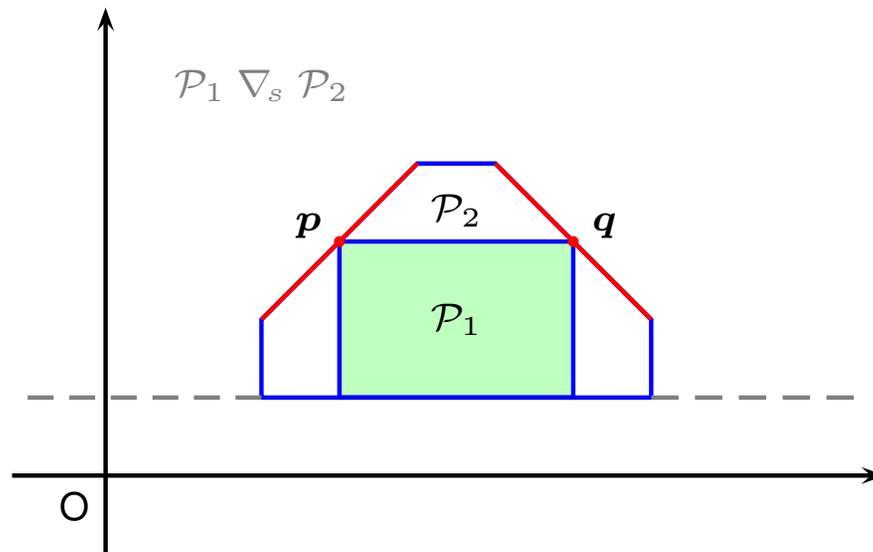
STANDARD WIDENING VS. COMBINING CONSTRAINTS (III)



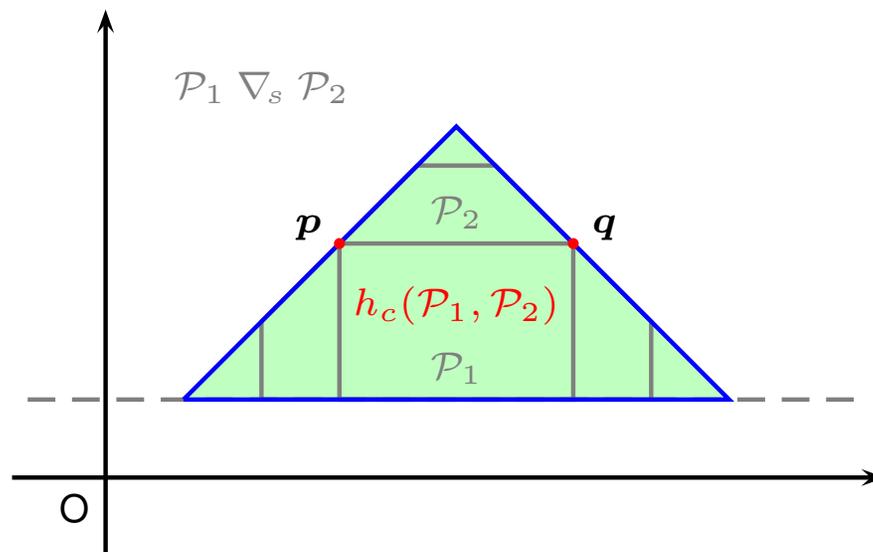
STANDARD WIDENING VS. COMBINING CONSTRAINTS (IV)



STANDARD WIDENING VS. COMBINING CONSTRAINTS (V)



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3RD HEURISTICS: EVOLVING POINTS

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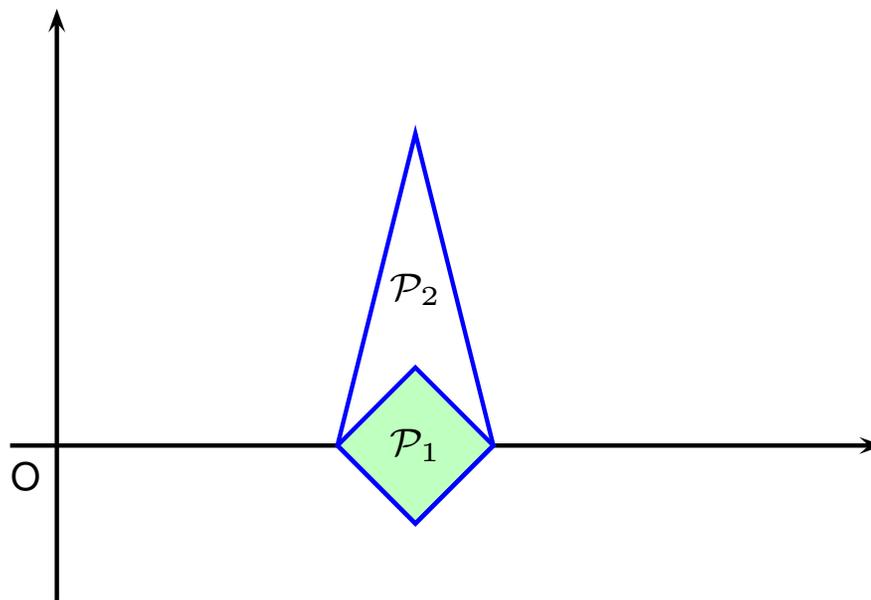
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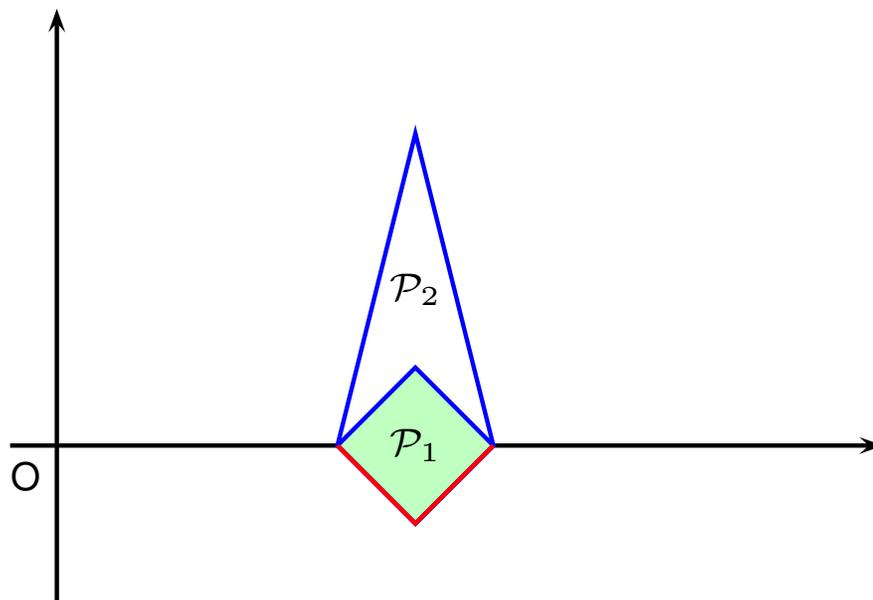
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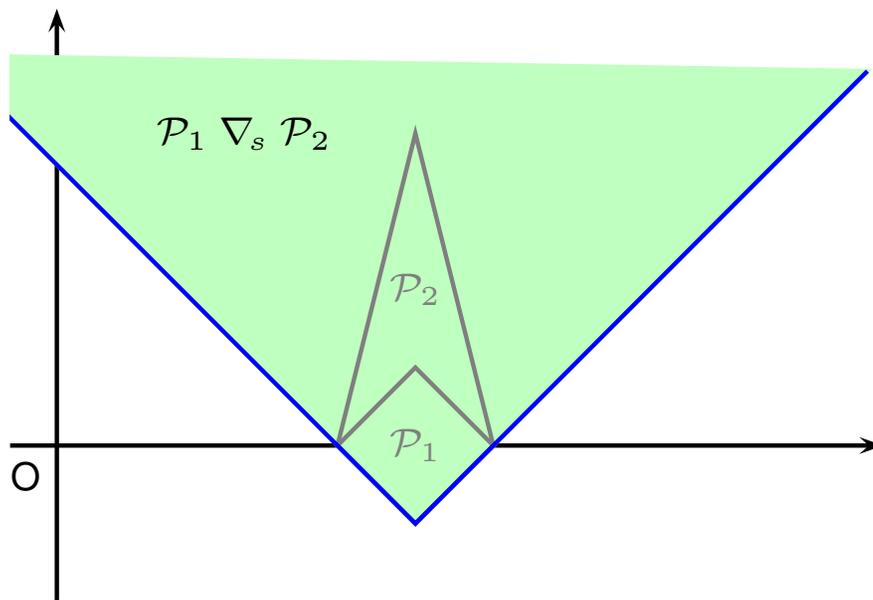
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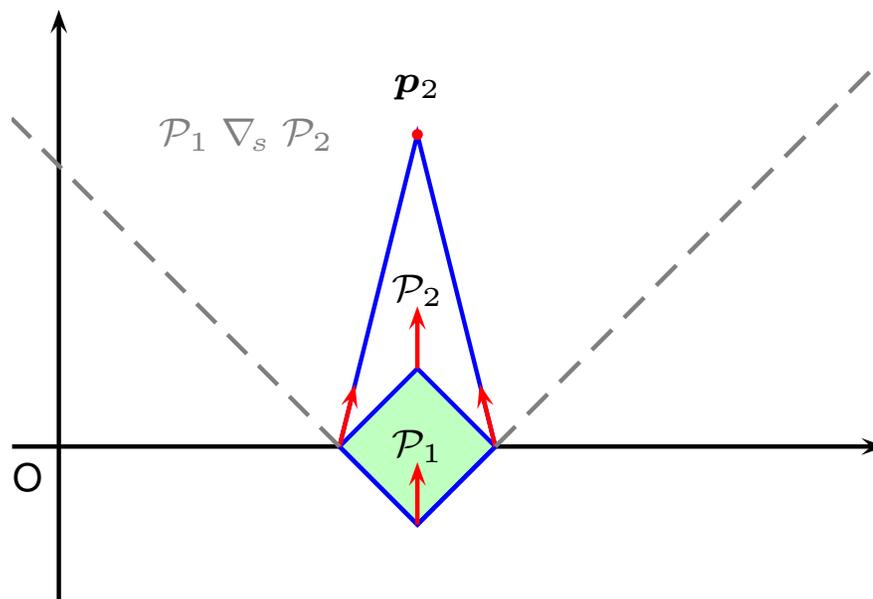
STANDARD WIDENING VS. EVOLVING POINTS (II)



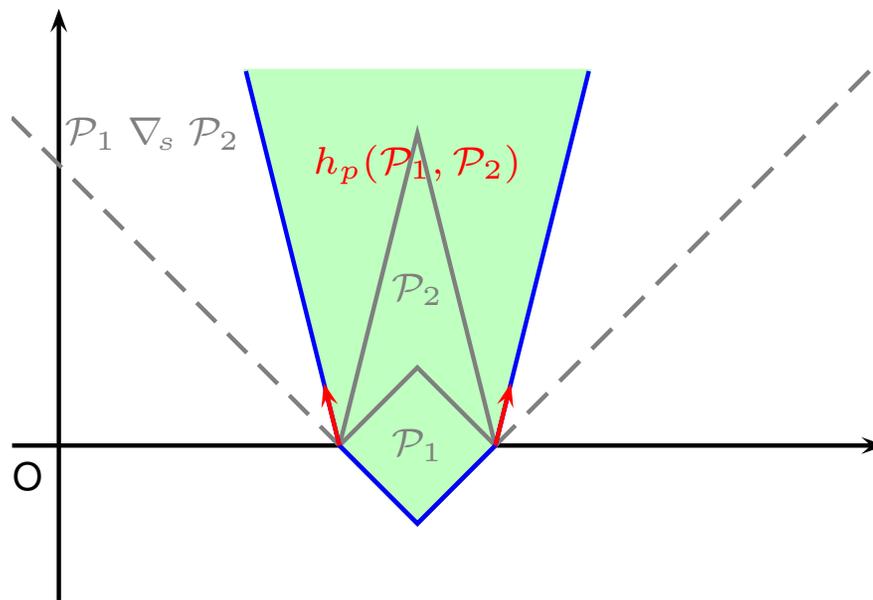
STANDARD WIDENING VS. EVOLVING POINTS (III)



STANDARD WIDENING VS. EVOLVING POINTS (V)



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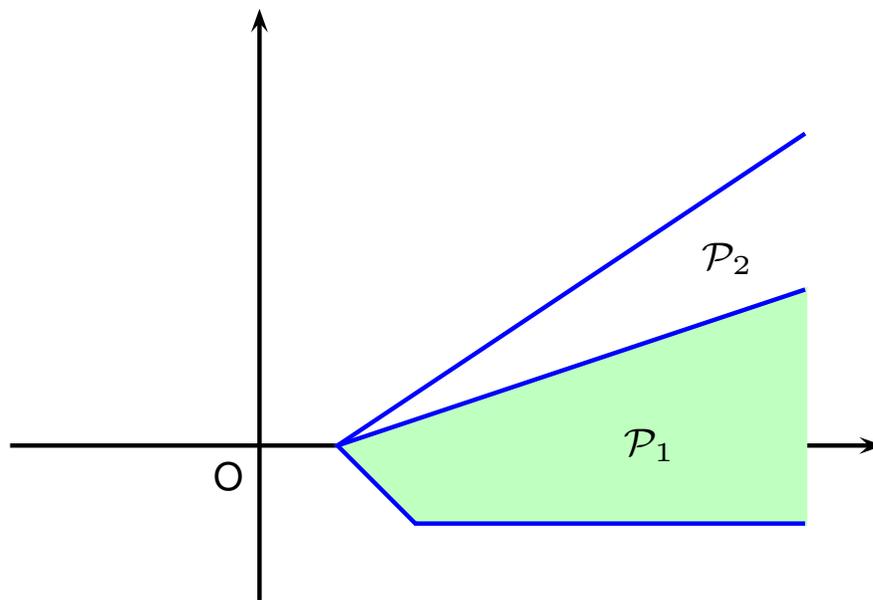
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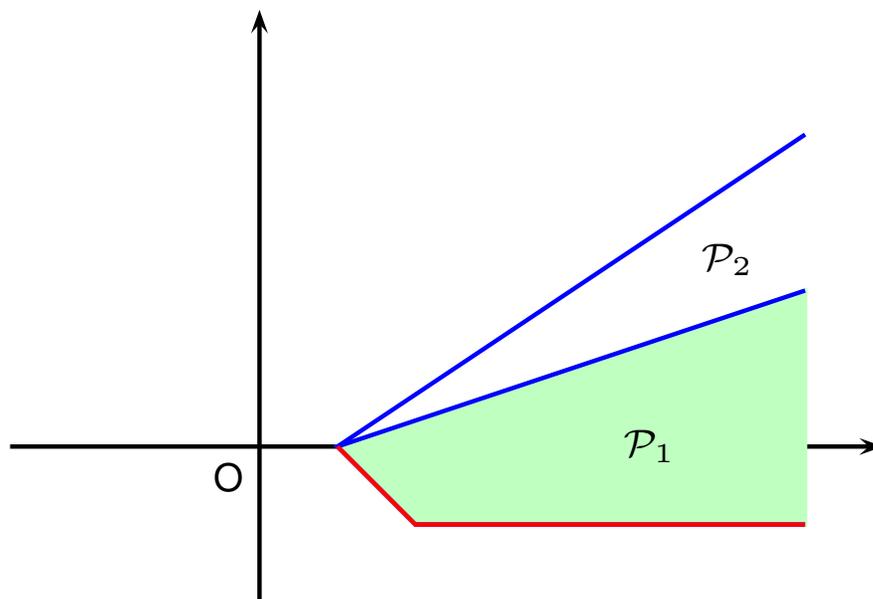
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- The extrapolation will decrease the total number of non-zero coordinates of the ray \implies hopefully satisfying the last case in the definition of the lgo \curvearrowright_n :

$$\mathcal{P}_1 \prec_r h_r(\mathcal{P}_1, \mathcal{P}_2).$$

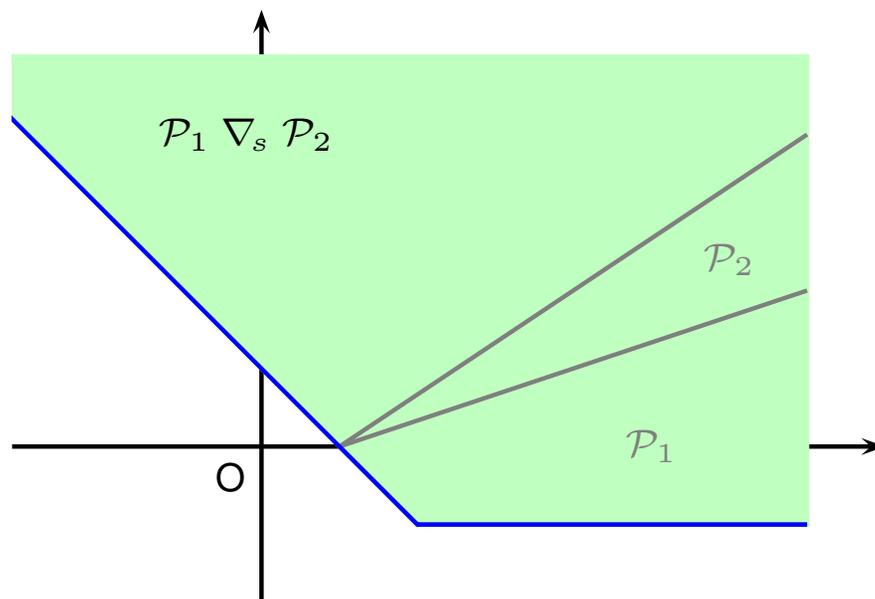
STANDARD WIDENING VS. EVOLVING RAYS (I)



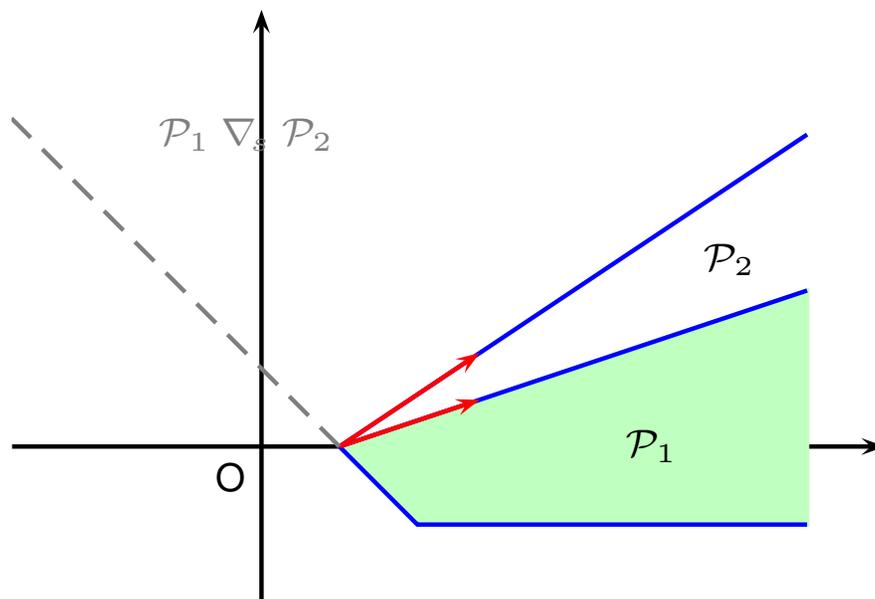
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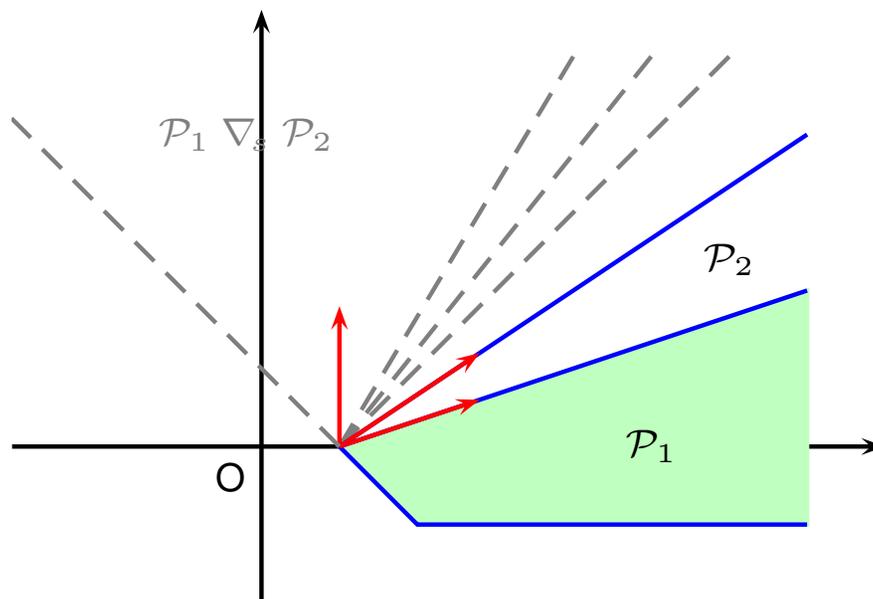
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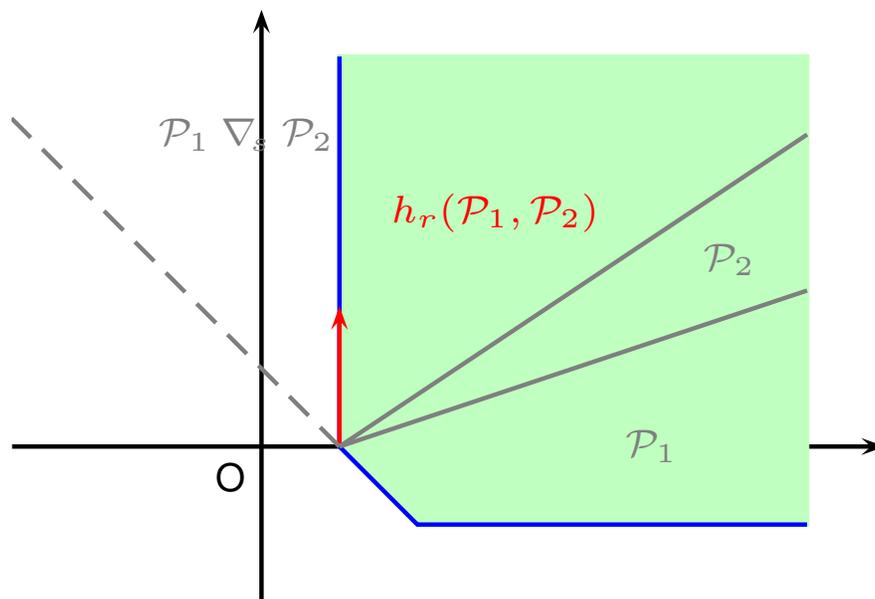
STANDARD WIDENING VS. EVOLVING RAYS (IV)



STANDARD WIDENING VS. EVOLVING RAYS (v)



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THE NEW WIDENING ∇_n

→ An **instance of the framework**: try the four heuristics in the given order, eventually falling back to the standard widening.

$$\mathcal{P}_1 \nabla_n \mathcal{P}_2 \stackrel{\text{def}}{=} \begin{cases} \mathcal{P}_2, & \text{if } \mathcal{P}_1 \curvearrowright \mathcal{P}_2; \\ h_c(\mathcal{P}_1, \mathcal{P}_2), & \text{if } \mathcal{P}_1 \curvearrowright h_c(\mathcal{P}_1, \mathcal{P}_2) \subset \mathcal{P}_1 \nabla_s \mathcal{P}_2; \\ h_p(\mathcal{P}_1, \mathcal{P}_2), & \text{if } \mathcal{P}_1 \curvearrowright h_p(\mathcal{P}_1, \mathcal{P}_2) \subset \mathcal{P}_1 \nabla_s \mathcal{P}_2; \\ h_r(\mathcal{P}_1, \mathcal{P}_2), & \text{if } \mathcal{P}_1 \curvearrowright h_r(\mathcal{P}_1, \mathcal{P}_2) \subset \mathcal{P}_1 \nabla_s \mathcal{P}_2; \\ \mathcal{P}_1 \nabla_s \mathcal{P}_2, & \text{otherwise.} \end{cases}$$

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- **Uniformly more precise** than the standard widening.
- In general, this does **not** hold for the **final result of upward iteration sequences**, because neither the standard widening nor the new one are **monotonic operators**.

PRECISION COMPARISON

Argument size relations for Prolog programs using [China + PPL](#).

Note: carefully chosen widening strategy ([Bourdoncle, FMPTA'93](#))
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	# programs (361)			# predicates (23279)		
k (delay)	improve	degr	incomp	improve	degr	incomp
0	121	-	2	1340	3	2
1	34	-	-	273	-	-
2	29	-	-	222	-	-
3	28	-	-	160	-	-
4	25	-	2	126	2	-
10	25	-	-	124	-	-

EFFICIENCY COMPARISON

Argument size relations for Prolog programs using **China + PPL**.

Total analysis time

k (delay)	$k\nabla_s$		$k\nabla_n$	
	all	top 20	all	top 20
0	1.00	0.72	1.05	0.77
1	1.09	0.79	1.11	0.80
2	1.16	0.83	1.18	0.84
3	1.23	0.88	1.25	0.89
4	1.32	0.95	1.34	0.95
10	1.82	1.23	1.85	1.24

CONCLUSION

- We have defined a **domain independent framework** for improving upon the precision of a fixed widening operator;
- We have **instantiated** the framework on the domain of convex polyhedra improving on the precision of the **standard widening**;
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CURRENT AND FUTURE WORK

- Widening operators are the corner stone for both the **feasibility** and **precision** of static analyses adopting **accurate abstract domains**:
 - ① We have defined (generic) widenings for disjunctive domains, such as **finite sets of polyhedra** (see the last planned seminar);
 - ② Many interesting domains are still missing (non-trivial) widening operators (e.g., \mathbb{Z} -polyhedra).