On the design of widening operators *"If you widen without principles you may converge without precision"*

Nicolas Halbwachs

Verimag/CNRS Grenoble Automatic verification mainly consists in computing fixpoints of monotone functions on lattices.

Example: computation of reachable states

Reach = Init \cup post(Reach) $L = 2^{S}$

Model-checking = exact fixpoint computation, generally in finite or finite-depth lattices.

(Finite) abstraction [Cousot-Cousot, POPL'77]

$$L_c \stackrel{\alpha}{\stackrel{\sim}{\stackrel{\rightarrow}{\rightarrow}}} L_a$$
 (finite)

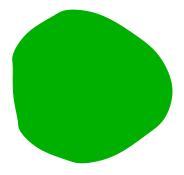
Let $F_a = \alpha \circ F_c \circ \gamma$, then $Ifp(F_c) \sqsubseteq \gamma(Ifp(F_a))$

 \longrightarrow fixpoint approximation, conservative verification

now routinely used in model-checking [Clarke-Grumberg-Long, TOPLAS'94] [Graf-Loiseaux, CAV'93]

Example: Predicate abstraction

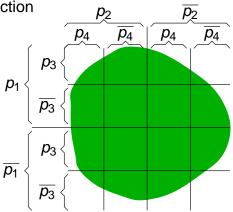
S, a set of states



Example: Predicate abstraction

S, a set of states

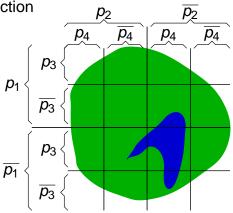
P a finite set of predicates $p_i : S \mapsto \{0, 1\}$

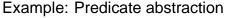


Example: Predicate abstraction

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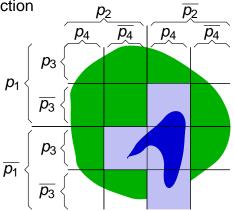
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S, a set of states

P a finite set of predicates $p_i: S \mapsto \{0, 1\}$



$$L_{c} = 2^{S} , \quad L_{a} = 2^{Mon(P)}$$
$$\alpha(X) = \{ \phi \in Mon(P) \mid \exists x \in X, x \models \phi \}$$
$$\gamma(Y) = \{ x \mid \exists \phi \in Y, x \models \phi \}$$

So the main remaining difference between Model-Checking and Abstract Interpretation is the use of widening.

Widening often considered as a dirty heuristic!

So the main remaining difference between Model-Checking and Abstract Interpretation is the use of widening.

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Outline of the talk

- Introduction
- Reminders about Widening
- Widening on convex polyhedra
 - Reminders about linear relation analysis
 - Classical widenings
 - Correct and incorrect attempts for improvement
 - Taking the program into account
- Avoiding widening
 - Acceleration
 - Exact abstract solution
 - Can we combine the two?

Widening: basic idea

[Cousot-Cousot, POPL'77]:

- stay in an infinite lattice
- iterative computations of $lfp(F) = \bigsqcup F^n(\bot)$ may be infinite

n∈ℕ

- try to guess the limit from its first terms $(X_0 = \bot, X_1 = F(X_0), X_2 = F(X_1)...)$
- this guess is made through the computation of

$$Y_0 = X_0, \quad Y_{n+1} = Y_n \nabla F(Y_n)$$

where ∇ is a widening operator.

Widening: definition

- $(L, \sqsubseteq, \sqcap, \sqcup, \bot, \top)$ a complete lattice.
- $\nabla: L \times L \mapsto L$ is a widening iff
 - $\forall x, y \in L, x \sqcup y \sqsubseteq x \nabla y$
 - [chain condition]

for all increasing chain $x_0 \sqsubseteq x_1 \sqsubseteq \ldots \sqsubseteq x_n \ldots$ in *L*, the increasing chain $y_0 = x_0, \ldots y_{n+1} = y_n \nabla x_{n+1}, \ldots$ is not strictly increasing (i.e., stabilizes after a finite number of terms)

Widening: use

Instead of computing the (infinite) sequence

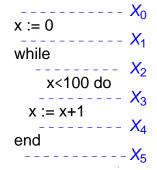
$$X_0 = \bot, X_{n+1} = F(X_n)$$

compute the finite sequence

$$Y_0 = X_0, Y_{n+1} = Y_n \nabla F(Y_n)$$

which limit is greater than lfp(F)

[Cousot-Cousot, ISP'76]



[Cousot-Cousot, ISP'76] $\cdots X_0 = [-\infty, +\infty]$ x := 0----- $X_1 = X_0[x := 0]$ while $---- X_2 = X_1 \sqcup X_4$ x<100 do $X_3 = X_2 \sqcap [-\infty, 99]$ x := x+1 $----X_4 = X_3[x := x+1]$ end $----X_5 = X_2 \sqcap [100, +\infty]$

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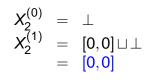
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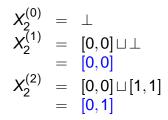
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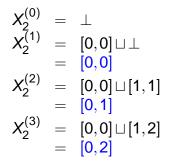
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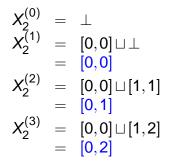


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Exact computation



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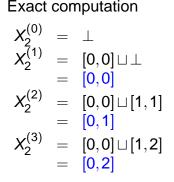
With widening

$$\begin{array}{rcl} X_2^{(0)} &=& \bot \\ X_2^{(1)} &=& [0,0] \sqcup \bot \\ &=& [0,0] \\ X_2^{(2)} &=& [0,0] \sqcup [1,1] \\ &=& [0,1] \\ X_2^{(3)} &=& [0,0] \sqcup [1,2] \\ &=& [0,2] \end{array}$$

. . .

$$X_2^{(0)} = \bot$$

$$X_2 = [0,0] \sqcup ((X_2 \cap [-\infty,99])[x := x+1))$$

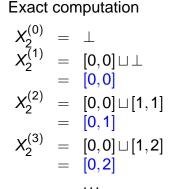


With widening

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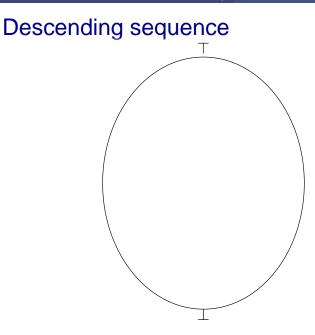


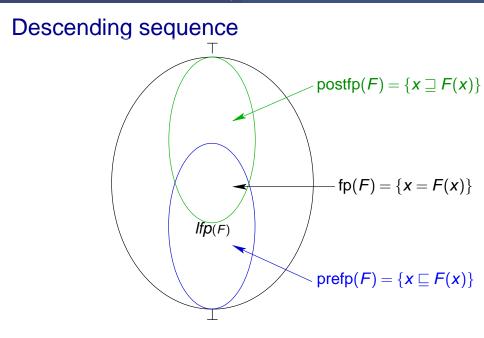
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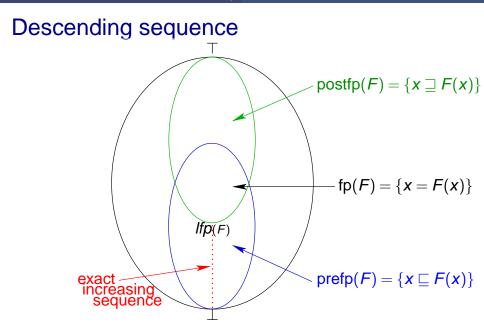
$$\begin{array}{rcl} X_2^{(0)} &=& \bot \\ X_2^{(1)} &=& \bot \nabla ([0,0] \sqcup \bot) \\ &=& \bot \nabla [0,0] \\ &=& [0,0] \\ X_2^{(2)} &=& [0,0] \nabla ([0,0] \sqcup [1,1]) \\ &=& [0,0] \nabla [0,1] \\ &=& [0,\infty] \\ \text{convergence!} \end{array}$$

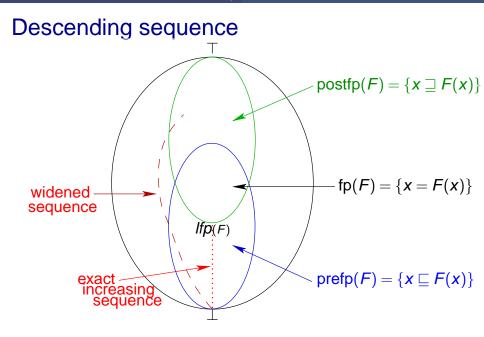
Widening on intervals:

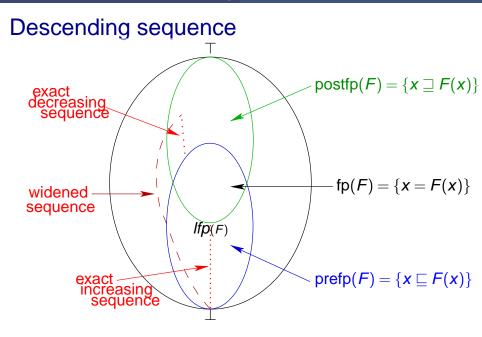
$$\begin{split} & \perp \nabla I = I \\ [a,b] \, \nabla \, [c,d] = \left[\text{if } c < a \, \text{then } -\infty \, \text{else } a \, , \, \text{if } d > b \, \text{then } \infty \, \text{else } b \right] \\ & \text{So,} \\ & \quad \perp \nabla [0,0] = [0,0] \\ & \text{and} \\ & \quad [0,0] \nabla [0,1] = [0,\infty] \end{split}$$











Descending sequence: intervals

The widened sequence converged at $X_2^{(2)} = [0, \infty]$ Descending sequence:

$$\begin{array}{rcl} X_2^{(3)} &=& [0,0] \sqcup \left(\left(X_2^{(2)} \cap [-\infty,99] \right) [x := x+1] \right) \\ &=& [0,0] \sqcup ([0,99][x := x+1]) \\ &=& [0,0] \sqcup [1,100] \\ &=& [0,100] \\ X_2^{(4)} &=& [0,0] \sqcup \left(\left(X_2^{(3)} \cap [-\infty,99] \right) [x := x+1] \right) \\ &=& X_2^{(3)} & \text{Fixpoint!} \end{array}$$

Another old example: Karp & Miller

[Karp-Miller, J. Comput. Syst. Sci. 69] Boundedness of Petri nets

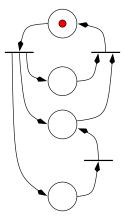
A Petri net with *p* places. Markings = \mathbb{N}^p Order on markings: $M \leq M' \Leftrightarrow \forall i = 1..p, M_i \leq M'_i$ Obvious property: a set of mutually incomparable mark

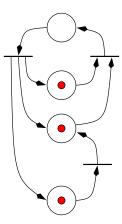
Obvious property: a set of mutually incomparable markings cannot be infinite

Enumerate the reachable markings, and whenever some marking *M* leads to a strictly greater marking *M'*, replace *M'* by $M\nabla M'$:

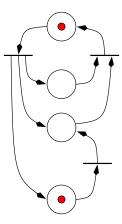
$$(M\nabla M')_{i} = \begin{cases} M_{i} & \text{if } M_{i} = M'_{i} \\ \infty & \text{if } M_{i} < M'_{i} \end{cases}$$



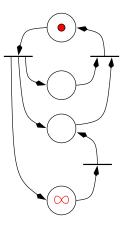


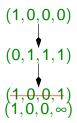




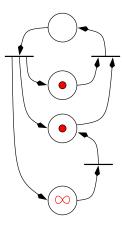


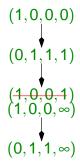




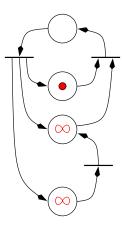


Karp & Miller: example



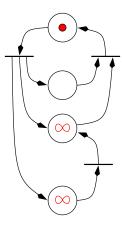


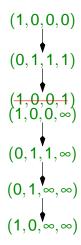
Karp & Miller: example





Karp & Miller: example





A more general definition

[Ball-Podelski-Rajamani, TACAS'02]

- Infinite states, infinite set of atomic predicates φ
- Abstract values = predicates in DNF: $\bigvee_{i \in I} \bigwedge_{j \in J_i} \varphi_{ij}$
- Widening (hint): keep in X∇Y only the conjuncts of X which are still in Y:

$$\left(\bigvee_{i\in I}\bigwedge_{j\in J_i}\varphi_{ij}\right)\nabla\left(\bigvee_{i\in I}\bigwedge_{j\in J'_i}\varphi_{ij}\right) = \left(\bigvee_{i\in I}\bigwedge_{j\in J_i\cap J'_i}\varphi_{ij}\right)$$

Obvious need of canonical form.

Is widening useful?

• [Hankin-Hunt, ESOP'92]: All results that you get with widening can also be obtained by computing in a suitable finite lattice.

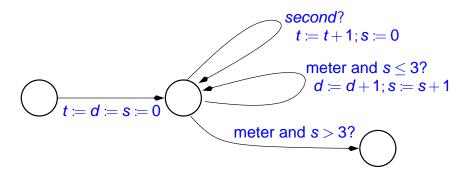
Is widening useful?

- [Hankin-Hunt, ESOP'92]: All results that you get with widening can also be obtained by computing in a suitable finite lattice.
- [Cousot-Cousot, PLILP'92]:
 - For each program, there exists a finite lattice which can be used for this program to obtain results equivalent to those obtained using widening;
 - No such finite lattice will do for all programs;
 - For a particular program, it is not possible to infer the set of needed abstract values by a simple inspection of the text of the program.

The case of convex polyhedra

Linear relation analysis: compute, in each point of a program, a set of linear inequalities invariantly satisfied by the numerical variables.

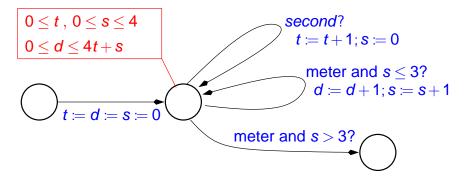
Example: a speedometer (speed limit: 4m/s)



The case of convex polyhedra

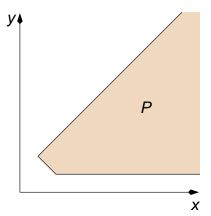
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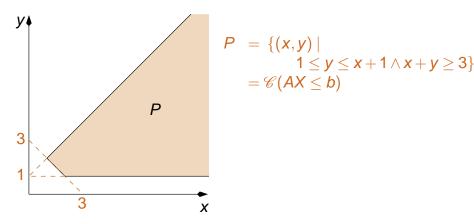
Computing over convex polyhedra (1/2)

The double representation



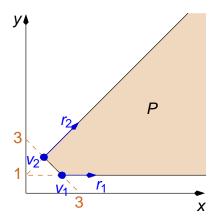
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Computing over convex polyhedra (1/2)

The double representation



$$P = \{(x, y) \mid \\ 1 \le y \le x + 1 \land x + y \ge 3\} \\ = \mathscr{C}(AX \le b)$$

$$P = \{ \lambda v_1 + (1 - \lambda) v_2 + \mu_1 r_1 + \mu_2 r_2 \mid \\ \lambda \in [0, 1], \ \mu_1, \mu_2 \ge 0 \} \\ = \mathscr{S}(V, R) \}$$

F

Computing over convex polyhedra (2/2)

Common operations:

• intersection:

 $\mathscr{C}(AX \leq B) \cap \mathscr{C}(A'X \leq B') = \mathscr{C}(AX \leq B \land A'X \leq B')$

- convex hull (approximation of union): $\mathscr{S}(V,R) \sqcup \mathscr{S}(V',R') = \mathscr{S}(V \cup V',R \cup R')$
- affine transformation: $CP + D = \{CX + D \mid X \in P\}$ $\mathscr{S}(C\mathscr{S}(V, R) + D) = \{Cv + D \mid v \in V\}, \{Cr \mid r \in R\}$
- test for inclusion:

 $\mathscr{S}(V, R) \subseteq \mathscr{C}(AX \leq B)$ iff $Av \leq B, \forall v \in V$ and $Ar \leq 0, \forall r \in R$

• test for emptyness: $\mathscr{S}(V, R) = \emptyset$ iff $V = \emptyset$

[Cousot-Halbwachs, POPL'78]

Basic idea: keep for $P\nabla Q$ the constraints of P which are still satisfied by Q

$\boldsymbol{P} = \mathscr{C}(0 \leq \boldsymbol{y} \leq \boldsymbol{x} \leq 1)$



[Cousot-Halbwachs, POPL'78]

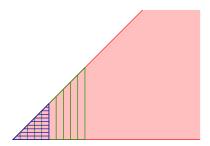
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$$P = \mathscr{C}(0 \le y \le x \le 1)$$
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Basic idea: keep for $P\nabla Q$ the constraints of P which are still satisfied by Q



$$P = \mathscr{C}(0 \le y \le x \le 1)$$
$$Q = \mathscr{C}(0 \le y \le x \le 2)$$
$$P \nabla Q = \mathscr{C}(0 \le y \le x)$$

Problem: None of the representations is canonical

 $\boldsymbol{P} = \mathscr{C}(\boldsymbol{y} = \boldsymbol{0} \; , \; \boldsymbol{0} \leq \boldsymbol{x} \leq \boldsymbol{1})$

Problem: None of the representations is canonical

$$P = \mathscr{C}(y = 0, 0 \leq x \leq 1)$$

 $Q = \mathscr{C}(0 \leq y \leq x \leq 2)$



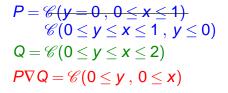
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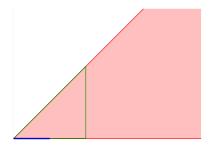
$$P = \mathscr{C}(y = 0, 0 \le x \le 1)$$

 $\begin{aligned} & \mathsf{Q} = \mathscr{C}(0 \leq y \leq x \leq 2) \\ & \mathsf{P} \nabla \mathsf{Q} = \mathscr{C}(0 \leq y \;,\; 0 \leq x) \end{aligned}$

Problem: None of the representations is canonical



Problem: None of the representations is canonical



$$P = \mathscr{C}(\underline{y=0, 0 \le x \le 1})$$

$$\mathscr{C}(0 \le y \le x \le 1, y \le 0)$$

$$Q = \mathscr{C}(0 \le y \le x \le 2)$$

$$P\nabla Q = \mathscr{C}(0 \le y, 0 \le x)$$

$$\mathscr{C}(0 \le y \le x)$$

Solution [Halbwachs, Thesis 1979]: keep for $P\nabla Q$ the constraints of Q which are mutually redundant with constraints of P

(i.e., can replace some constraints of P without changing it) 2 constraints are mutually redundant for P if they are saturated by the same generators of P.

 $\boldsymbol{P} = \mathscr{C}(\boldsymbol{y} = \boldsymbol{0} \;, \; \boldsymbol{0} \leq \boldsymbol{x} \leq \boldsymbol{1})$

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 $P = \mathscr{C}(y = 0, 0 \le x \le 1)$ $Q = \mathscr{C}(0 \le y \le x \le 2)$ $0 \le y \text{ already in } P$ $y \le x \text{ mut. red. with } 0 \le x \text{ in } P$ $P \nabla Q = \mathscr{C}(0 \le y \le x)$

Chain condition: either the number of constraints decreases, or the dimension increases

Improving the precision

- delaying the widening
- improve the operator
- take the program into account

[Folk!], [Halbwachs, CAV'93], [Goubault, SAS'01], [Blanchet et al., PLDI'03] Instead of computing

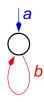
 $X_0 = \bot, X_1 = X_0 \nabla F(X_0), X_2 = X_1 \nabla F(X_1) \dots X_{n+1} = X_n \nabla F(X_n)$

fix k > 0 and compute

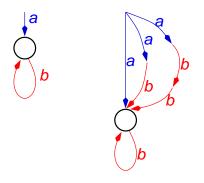
$$X_n = \begin{cases} \perp & \text{if } n = 0\\ F(X_{n-1}) & \text{if } n \le k\\ X_{n-1} \nabla F(X_{n-1}) & \text{if } n > k \end{cases}$$

or, more generally, apply the widening sporadically but infinitely often.

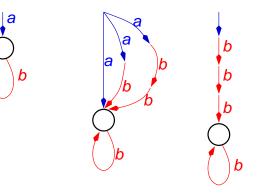
Delay, loop unrolling



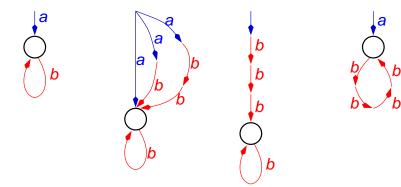
Delay, loop unrolling



Delay, loop unrolling



Delay, loop unrolling



Easy, but often expensive.

The suitable number of delays or unrolling may depend on the program.

Improving the widening operator (1/4)

Correct and incorrect improvements of the standard widening on polyhedra

[Bagnara-Hill-Ricci-Zaffanella, SAS'03 and SCP'05] Hint:

$$P\nabla Q = \begin{cases} P \sqcup Q \\ P \sqcup Q \\ P \bigcup Q \\ P \cup Q \\ P \cup$$

- Correct widening
- $\forall P, Q, P \nabla Q \subseteq P \nabla_S Q$
- Generally better than the standard widening.

Improving the widening operator (2/4)

- The widening is generally not monotonic.
- So, the fact that PV₁Q ⊆ PV₂Q does not imply that the limit computed with V₁ will be better that the one provided by V₂.
- May slow down the convergence.
- Ideal goal: Converge fast towards a precise limit.

Improving the widening operator (3/4)

An attractive idea:

We look for the limit of the sequence $(X_n)_{n>0}$.

• Add a new variable k (loop counter)

• Set
$$X'_n = X_n \wedge (k = n)$$

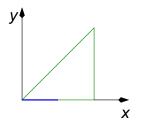
• express X'_n as a function of k, and let k tend to ∞ .

More precisely,

- Compute $X'_1 = (X_0 \land k = 0) \sqcup (X_1 \land k = 1)$.
- Forget the constraint $k \le 1$ in X'_1
- Eliminate k by existential quantification.

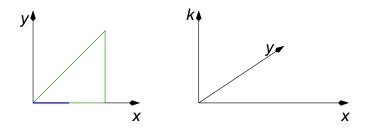
Improving the widening operator (4/4)

Example:



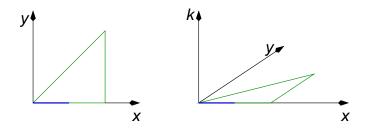
Improving the widening operator (4/4)

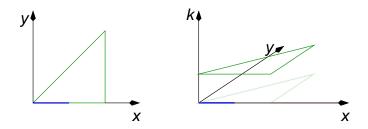
Example:

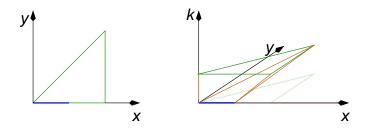


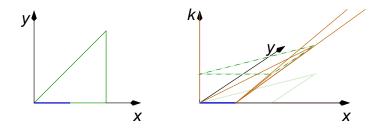
Improving the widening operator (4/4)

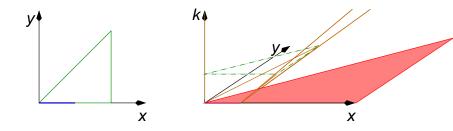
Example:

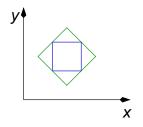


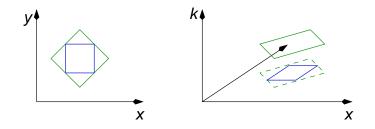


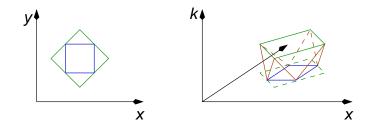


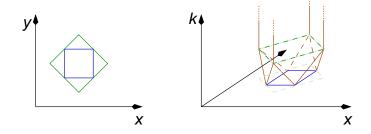


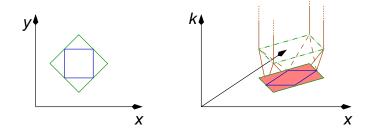




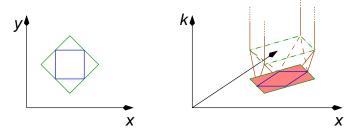








Unfortunately incorrect (does not fulfill the chain condition)



But adding loop counters often improves the precision of normal widening.

Taking the program into account

The widening $X\nabla F(X)$ uses the result F(X) of F, but doesn't look into F.

Looking at F, i.e., at the program, can help in improving the precision:

- Iimited widening
- new control path
- (abstract) acceleration

[Halbwachs, CAV'93], [Blanchet et al., PLDI'03]

Let *C* be a fixed finite set of linear constraints, then ∇_C^L defined by:

$$P\nabla^L_{\mathbf{C}}\mathsf{Q} = (P\nabla\mathsf{Q}) \cap \mathscr{C}(\{c \in \mathbf{C} \mid P \models c \land \mathsf{Q} \models c\})$$

is a widening

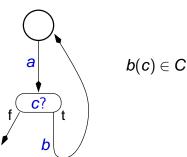
[Halbwachs, CAV'93], [Blanchet et al., PLDI'03]

Let *C* be a fixed finite set of linear constraints, then ∇_C^L defined by:

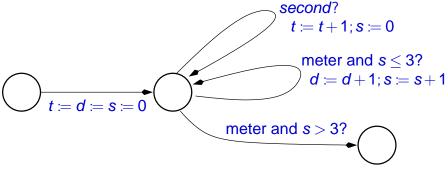
$$P\nabla^L_{\boldsymbol{C}}\mathsf{Q} = (P\nabla \mathsf{Q}) \cap \mathscr{C}(\{\boldsymbol{c} \in \boldsymbol{C} \mid \boldsymbol{P} \models \boldsymbol{c} \land \boldsymbol{Q} \models \boldsymbol{c}\})$$

is a widening

Choice of limiting constraints:

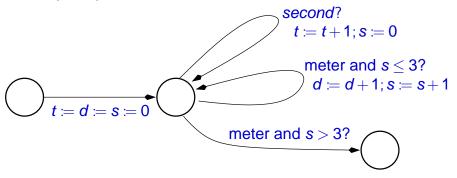


Example: speedometer



Limit the widening with $[s := s+1](s \le 3) = (s \le 4)$.

Example: speedometer



Limit the widening with $[s = s+1](s \le 3) = (s \le 4)$.

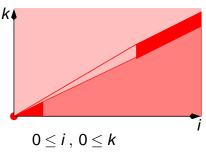
Very efficient, often makes the descending sequence useless, sometimes gives better results than the descending sequence.

$$i := j := 0 := k := 0;$$

while $i' <= 100$ do
 $i := i+2; j := j+1;$
if j then $k := k+1;$
if $j = 10$ then
 $j := 0; k := k+1;$
end

$$i := j := 0 := k := 0;$$

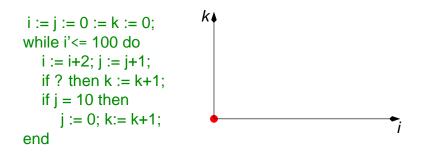
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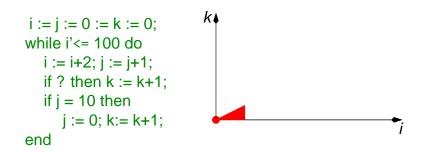


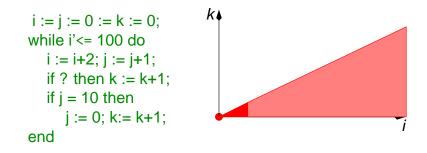
Solutions: If, at step *n*, a previously unfeasible path in the loop becomes feasible,

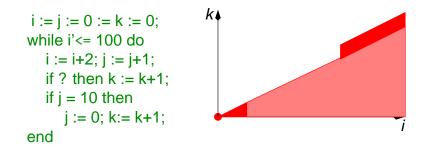
- either don't widen at this step [Astrée]
- or take $X_0 \nabla X_n$, instead of $X_{n-1} \nabla X_n$ [Halbwachs, CAV'93]

Can occur only finitely many times!









i := j := 0 := k := 0;
while i'<= 100 do
i := i+2; j := j+1;
if ? then k := k+1;
if j = 10 then
j := 0; k:= k+1;
end

$$0 \le k \le \frac{11}{20}i$$

Exact computations

Are there cases where the effect of a loop can be computed exactly?

- Ioop acceleration
- exact abstract computation

Loop acceleration (1/3)

Exact computations in Presburger arithmetic (for a widening in Presburger arithmetic, see [Bultan-Gerber-Pugh, CAV'97])

Convergence using loop acceleration [Boigelot-Wolper, CAV'94 and CAV'98], [Common-Jurski, CAV'98], [Finkel-Sutre, MFCS'00], [Bardin et al., CAV'03]

 τ being a relation in $\mathbb{N}^n \times \mathbb{N}^n$, in which cases can we compute exactly $\tau^* = \bigcup_{k \in \mathbb{N}} \tau^k$ or $\tau^*(X_0)$?



Loop acceleration (2/3)

- Restrictions on τ (e.g., ℓ(X) ∧ X' = X + B, or ℓ(X) ∧ X' ♯X + B, ♯ ∈ {≤,=≥}), flat automata (without nested loops), ...
- Semi-algorithms for more general cases [FAST]

But,

- Exact computation does not scale up
- What to do with programs which don't meet the restrictions?

Loop acceleration (3/3)



Loop acceleration (3/3)

$$\bigcirc \ell?X := X + B$$

$au^*(X_0) = \ell(X_0) \land \exists i \ge 0, X = X_0 + iB \land \ell(X_0 + (i-1)B)$ is a Presburger formula

Loop acceleration (3/3)



 $\tau^*(X_0) = \ell(X_0) \land \exists i \geq 0, \ X = X_0 + iB \land \ell(X_0 + (i-1)B)$

is a Presburger formula

Not true for interleaving of transitions

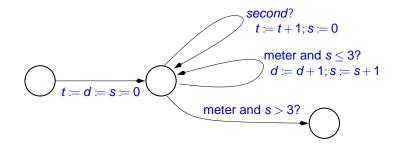
In which cases can we solve exactly the abstract fixpoint equation?

[Su-Wagner, TACAS'04]: Exact computation of the least fixpoint in interval analysis (in polynomial time)

using, basically, a suitable limited widening

Abstract acceleration on polyhedra [Gonnord-Halbwachs06]

Example: the speedometer in one abstract acceleration:

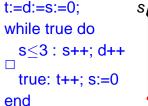


The speedometer in one abstract acceleration:

```
t:=d:=s:=0; s
while true do
s≤3 : s++; d++
true: t++; s:=0
end
```

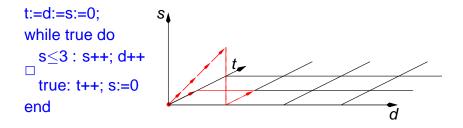
C

The speedometer in one abstract acceleration:

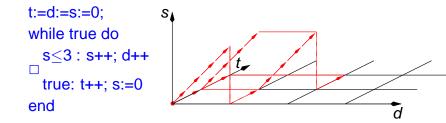




The speedometer in one abstract acceleration:

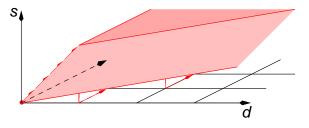


The speedometer in one abstract acceleration:



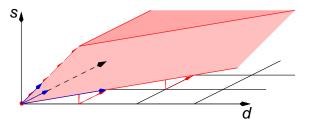
The speedometer in one abstract acceleration:

t:=d:=s:=0; while true do s≤3 : s++; d++ true: t++; s:=0 end



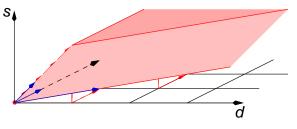
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The speedometer in one abstract acceleration:

t:=d:=s:=0; while true do s≤3 : s++; d++ true: t++; s:=0 end



$$\begin{array}{ll} {\sf P} &= (\{(0,0,0)\} \nearrow \{(1,0,0),(0,1,1),(1,4,0)\}) \cap (s \le 4) \\ &= (t \ge 0, 0 \le s \le 4, 0 \le d \le 4t + s) \end{array}$$

Obvious advantages:

- when it applies, it is both more precise and more efficient
- easy to combine with widening

Conclusions

- Widening is more than a dirty heuristic
 - it is based on a reasonable hypothesis:
 - Regular, foreseeable programs should be easy to analyse
 - It allows to work in quite expressive lattices.
 - It allows general programs to be analysed.
- It may be more effective to use well-chosen application strategies, (e.g., limited widening, specific cases when new pathes are discovered), than to endlessly look for "better" widening.
- Looking at the program can lead to significant improvements.