

Inferring Sufficient Conditions with Backward Polyhedral Under-Approximations

Antoine Miné

CNRS & École normale supérieure
Paris, France

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Introduction

Inferring **invariants** is a **well-studied** problem,
with many applications

(proving correctness, optimising, ...)

Favorite method: **abstract interpretation** using abstract domains
(scalable, terminates thanks to ∇ , modular, flexible)

We want to infer **sufficient conditions**
for a given property to hold on a given program...
...and still use **abstractions**

Focus on **numeric** properties

⇒ numeric abstract domains (**polyhedra**)

- sufficient conditions
 - transition systems
 - programs
 - applications
- design effective abstract under-approximations
 - general algebraic properties
 - **polyhedral operators**
(tests, assignments, loops, non-linear expressions)

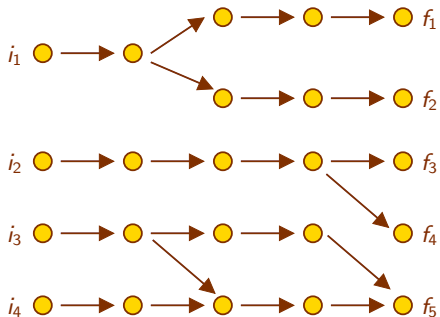
Disclaimer

Work in [progress](#), much research to be done

My goal: convince you that it **might** be **interesting**

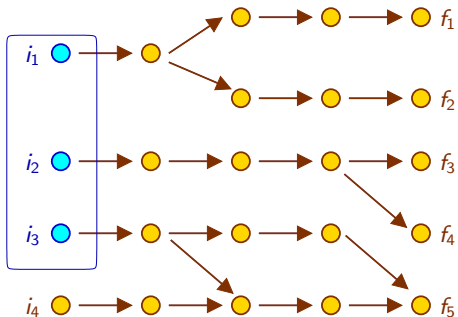
Sufficient Conditions

Transition Systems



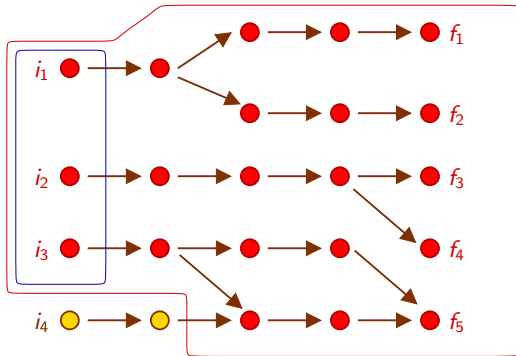
- Σ : set of **states**
- $\tau \subseteq \Sigma \times \Sigma$: **transition relation**, possibly non-deterministic
- initial states i_1, i_2, \dots and final states f_1, f_2, \dots

Transition Systems: Invariants



Given a set I of **initial** states,
 compute the set $\text{inv}(I)$ of **reachable states**

Transition Systems: Invariants

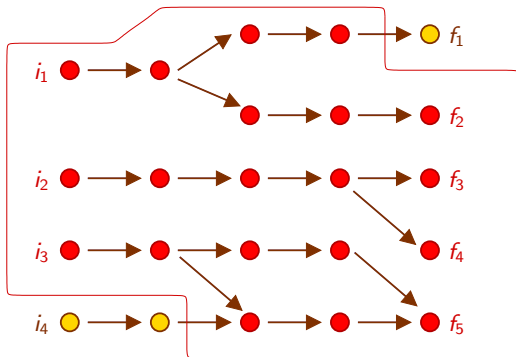


$$\text{inv}(I) = \text{lfp}_I \lambda X. X \cup \text{post}(X)$$

$$\text{where } \text{post}(X) \stackrel{\text{def}}{=} \{ \sigma \in \Sigma \mid \exists \sigma' \in X : (\sigma', \sigma) \in \tau \}$$

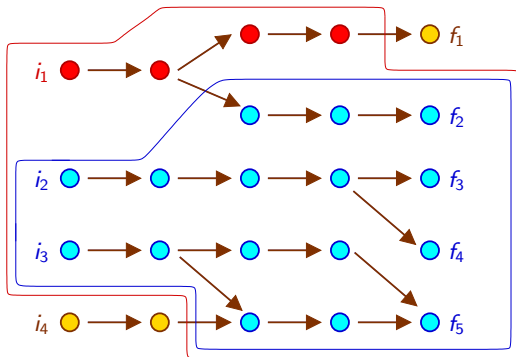
Smallest set containing I and invariant by τ

Transition Systems: Sufficient Conditions



Given a **target invariant** set T ,
 compute the **largest** set I such that $\text{inv}(I) \subseteq T$

Transition Systems: Sufficient Conditions



$$\text{cond}(T) = \text{gfp}_T \lambda X. X \cap \widetilde{\text{pre}}(X)$$

where $\widetilde{\text{pre}}(X) \stackrel{\text{def}}{=} \{ \sigma \in \Sigma \mid \forall \sigma' \in \Sigma: (\sigma, \sigma') \in T \implies \sigma' \in X \}$

$\implies \text{cond}(T)$ is the largest set I such that $\text{inv}(I) \subseteq T$

Example Program Analysis

Simple (non-deterministic) loop example

```
j := [0;10]
for (i = 0; i < 100; i++) {
  j = j + [0;1];
}
assert (j <= 105);
```

Example Program Analysis

Simple (non-deterministic) loop example

```
j := [0;10]
for (i = 0; i < 100; i++) {
  j = j + [0;1];
}
assert (j <= 105);
```

Forward invariant inference:

final values of j

$\implies j \in [0; 110]$, the assertion can be violated

Example Program Analysis

Simple (non-deterministic) loop example

```
j := [0;10]
for (i = 0; i < 100; i++) {
  j = j + [0;1];
}
assert (j <= 105);
```

Classic backward analysis:

initial values of j that **may** lead to correct program termination

$\implies j \in [0; 10]$ (we can always choose $0 \in [0; 1]$)

Example Program Analysis

Simple (non-deterministic) loop example

```
j := [0;10]
for (i = 0; i < 100; i++) {
  j = j + [0;1];
}
assert (j <= 105);
```

Backward sufficient condition analysis:

initial values of j that **always** lead to correct program termination

$\implies j \in [0; 5]$ (works even if we choose always $1 \in [0; 1]$)

Analysis Specificities

- backwards
- handling of non-determinism
- under-approximation

Analysis Specificities

- backwards
- handling of non-determinism

$$\widetilde{\text{pre}}(X) \neq \text{pre}(X) \stackrel{\text{def}}{=} \{ \sigma \in \Sigma \mid \exists \sigma' \in X : (\sigma, \sigma') \in \tau \}$$

we could use $\widetilde{\text{pre}}(X) = \neg(\text{pre}(\neg X))$

but abstract domains are seldom closed under \neg

\implies classic backward operators cannot be recycled easily

- under-approximation

Analysis Specificities

- backwards
- handling of non-determinism
- **under-approximation**

$$I' \subseteq \text{cond}(T) \implies \text{inv}(I') \subseteq T$$

→ soundness requires **under-approximations**

abstract domains seldom have best under-approximations
(vs. best over-approximations via Galois connection)

we **could** use an abstract X to represent $\neg X$

→ turns over-approximations into under-approximations

but $\neg X$ seldom represents an interesting property

⇒ classic backward operators **cannot be recycled easily**

Analysis Specificities

- backwards
- handling of non-determinism
- under-approximation

⇒ design new, non-optimal operators on existing domains

Applications

- sufficient condition for correctness
application to contract inference
- run-time check hoisting
- counter-example inference

Applications

- sufficient condition for correctness
- run-time check hoisting

original

```
for (i=0; i<n; i++)  
    assert (i≥0 && i<N);  
    t[i] = 1;
```

→

optimized

```
if (n≤N)  
    for (i=0; i<n; i++)  
        t[i] = 1;  
else  
    original program
```

- counter-example inference

Applications

- sufficient condition for correctness
- run-time check hoisting
- counter-example inference
 - sufficient initial conditions for the assertion to fail
→ under-approximate $\text{cond}(T)$ as before
 - proof that the assertion is **eventually** reached
liveness property!
→ instrument with a decreasing counter
(idea from Cousot Cousot POPL'12)

Backward Polyhedral Under-Approximations

Backward Versions

Programs are decomposed into **atomic** instructions i
with **well-known** concrete **transfer functions** $\tau\{i\}$

\implies derive concrete **backwards** transfer functions $\overleftarrow{\tau}\{i\}$
from concrete **forward** ones $\tau\{i\}$

Backward version of $f : \mathcal{P}(X) \rightarrow \mathcal{P}(Y)$

$$\overleftarrow{f} : \mathcal{P}(Y) \rightarrow \mathcal{P}(X)$$

$$\overleftarrow{f}(B) \stackrel{\text{def}}{=} \{a \in X \mid f(\{a\}) \subseteq B\}$$

Core properties:

- $\widetilde{\text{pre}} = \overleftarrow{\text{post}}$
- \overleftarrow{f} is monotonic and a complete \cap -morphism
- $\mathcal{P}(X) \xrightleftharpoons[f]{\overleftarrow{f}} \mathcal{P}(Y)$ (if f is a complete \cup -morphism)

Algebraic Properties

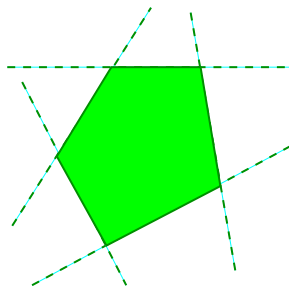
- $\overleftarrow{f \cup g} = \overleftarrow{f} \cap \overleftarrow{g}$ (but $\overleftarrow{f \cap g} \supset \overleftarrow{f} \cup \overleftarrow{g}$)
(control-flow join)
- $\overleftarrow{f \circ g} = \overleftarrow{g} \circ \overleftarrow{f}$
(instruction sequence)
- $f \subseteq g \iff \overleftarrow{g} \subseteq \overleftarrow{f}$
(approximation)
- $\overleftarrow{\lambda x. \text{lfp}_x(\lambda z. z \cup f(z))} = \lambda y. \text{gfp}_y(\lambda z. z \cap \overleftarrow{f}(z))$
(loops)

\implies **break-down**, **abstract** and **combine** backward functions

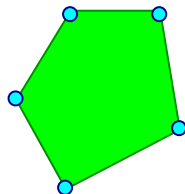
Polyhedra

domain of convex closed polyhedra

Dual representations:



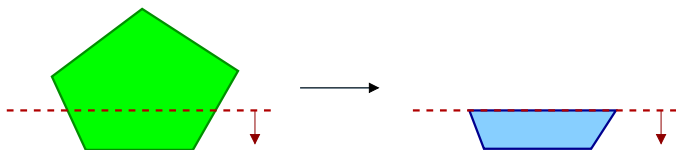
Constraints



Generators

(Cousot & Halbwachs, POPL 1978)

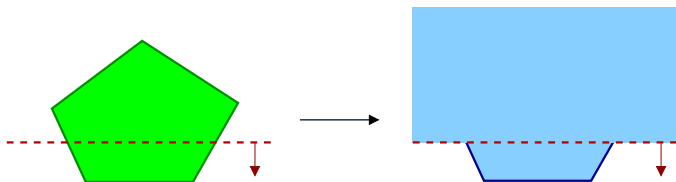
Modeling Affine Tests



Concrete Semantics:

we have $\tau \{ \mathbf{a} \cdot \mathbf{x} \leq b \} R = \{ \rho \in R \mid \mathbf{a} \cdot \rho(\mathbf{x}) \leq b \}$

Modeling Affine Tests

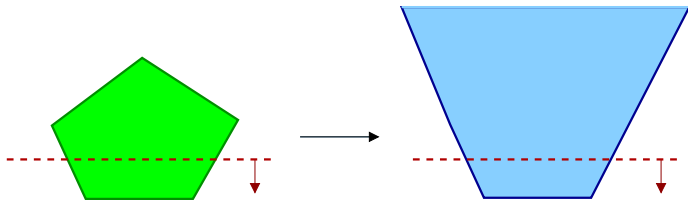


Concrete Semantics:

we have $\tau \{ \mathbf{a} \cdot \mathbf{x} \leq b \} R = \{ \rho \in R \mid \mathbf{a} \cdot \rho(\mathbf{x}) \leq b \}$

and so $\overleftarrow{\tau} \{ \mathbf{a} \cdot \mathbf{x} \leq b \} R = R \cup \{ \rho \mid \mathbf{a} \cdot \rho(\mathbf{x}) > b \}$

Modeling Affine Tests



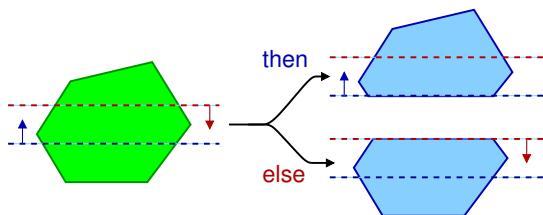
Abstract Polyhedral Semantics: $\overleftarrow{\tau} \{ \mathbf{a} \cdot \mathbf{x} \leq b? \}^\# P$

- remove $\mathbf{a} \cdot \mathbf{x} \leq b$ from the constraint set
- remove all constraints redundant with $\mathbf{a} \cdot \mathbf{x} \leq b$

\implies under-approximation, not optimal

Note: $\lambda P.P$ always under-approximates $\overleftarrow{\tau} \{ e? \}$

Modeling If-Then-Else



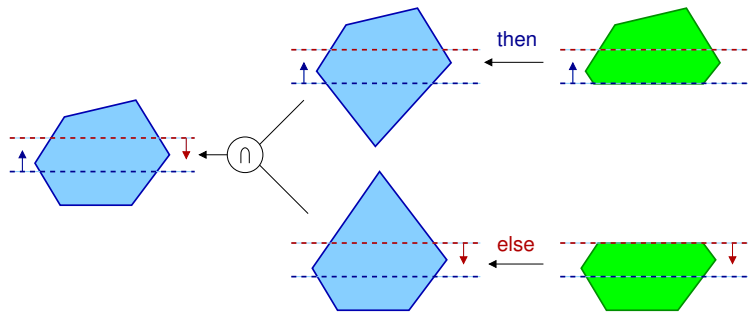
Forward semantics: $y + [0; 1] \geq 0$

$$\tau\{\text{if } (y + [0; 1] \geq 0) \{t\} \text{ else } \{e\}\} = \\ (\tau\{t\} \circ \tau\{y + [0; 1] \geq 0\}) \cup (\tau\{e\} \circ \tau\{y + [0; 1] < 0\})$$

where $\tau\{y + [0; 1] \geq 0\} R = \{(x, y) \in R \mid y \geq -1\}$

$\tau\{y + [0; 1] < 0\} R = \{(x, y) \in R \mid y < 0\}$

Modeling If-Then-Else

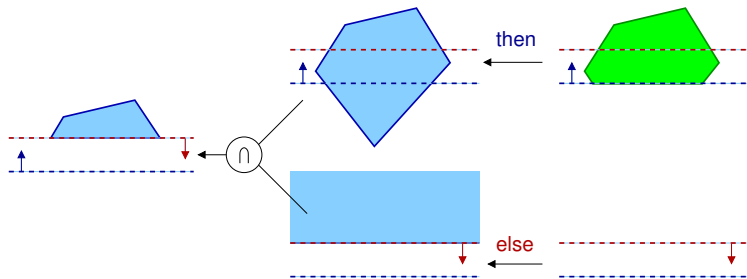
**Backward semantics:**

$$\begin{aligned} \overleftarrow{\mathcal{T}} \{ \text{if } (y + [0; 1] \geq 0) \{ t \} \text{ else } \{ e \} \} = \\ (\overleftarrow{\mathcal{T}} \{ y + [0; 1] \geq 0 \} \circ \overleftarrow{\mathcal{T}} \{ t \}) \cap \overleftarrow{\mathcal{T}} \{ y + [0; 1] < 0 \} \circ \overleftarrow{\mathcal{T}} \{ e \} \end{aligned}$$

where $\overleftarrow{\mathcal{T}} \{ y + [0; 1] \geq 0 \} R = R \cup \{ (x, y) \mid y < -1 \}$

$\overleftarrow{\mathcal{T}} \{ y + [0; 1] < 0 \} R = R \cup \{ (x, y) \mid y \geq 0 \}$

Modeling If-Then-Else

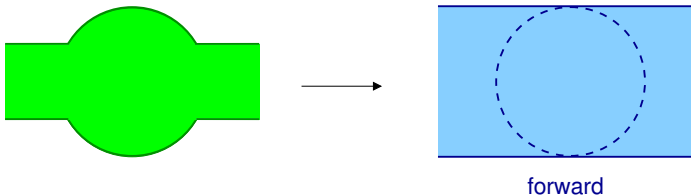


$\leftarrow \{ \cdot \}$ is not strict

\implies we can recover from a coarse under-approximation, even \emptyset

Example: the analysis finds the else branch dead
but continues nevertheless

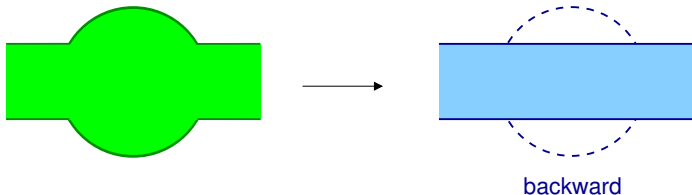
Modeling Projections



Concrete Semantics:

we have $\tau \{x := ?\} R = \{(x, y) \mid \exists v: (v, y) \in R\}$

Modeling Projections

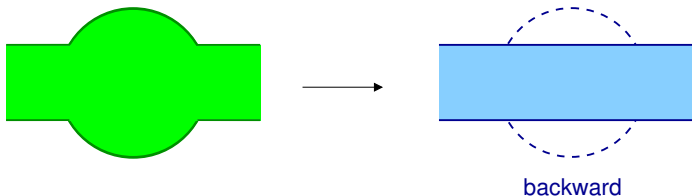


Concrete Semantics:

we have $\tau \{x := ?\} R = \{(x, y) \mid \exists v: (v, y) \in R\}$

and so $\overleftarrow{\tau} \{x := ?\} R = \{(x, y) \mid \forall v: (v, y) \in R\}$

Modeling Projections



Theorem

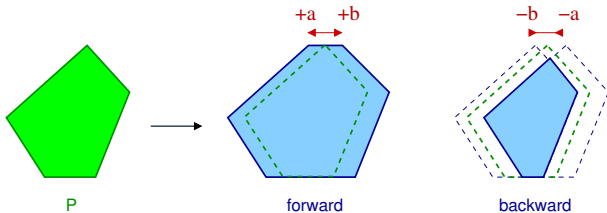
If R is convex closed, then $\overleftarrow{\tau} \{x := ?\} R$ is either R or \emptyset

Abstract Polyhedral Semantics:

$$\overleftarrow{\tau} \{x := ?\}^\# P = \begin{cases} P & \text{if } \tau \{x := ?\}^\# P = P \\ \emptyset & \text{otherwise} \end{cases} \quad (\text{exact})$$

Modeling Assignments

Example: $x := x + [a; b]$ (exact)



$$\tau \{ x := x + [a; b] \} R = \{ (x + v, y) \mid (x, y) \in R, v \in [a; b] \}$$

$$\overleftarrow{\tau} \{ x := x + [a; b] \} R = \{ (x, y) \mid \forall v \in [a; b]: (x + v, y) \in R \}$$

$$\overleftarrow{\tau} \{ x := x + [a; b] \}^\# P = \tau \{ x := x - a \}^\# P \cap^\# \tau \{ x := x - b \}^\# P$$

General case:

- $\tau \{ x := e \}$ can be synthesized using tests and projections
 \implies so can $\overleftarrow{\tau} \{ x := e \}$ and $\overleftarrow{\tau} \{ x := e \}^\#$
- $\tau \{ x := ? \}$ over-approximates $\tau \{ x := e \}$, so $\overleftarrow{\tau} \{ x := ? \}$
 under-approximates $\overleftarrow{\tau} \{ x := e \} \implies$ use $\overleftarrow{\tau} \{ x := ? \}^\#$

Loops

Concrete Semantics:

$$\tau \{ \text{while } (e) \{ b \} \} X \stackrel{\text{def}}{=} \tau \{ \neg e? \} (\text{lfp}_X \lambda Y. Y \cup (\tau \{ b \} \circ \tau \{ e? \})) Y$$

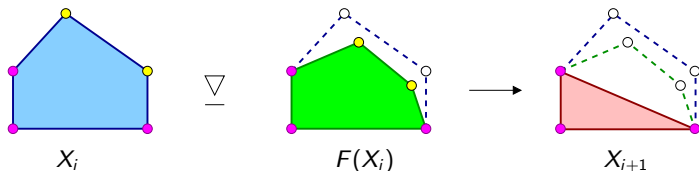
$$\overleftarrow{\tau} \{ \text{while } (e) \{ b \} \} X = \text{gfp}_{(\overleftarrow{\tau} \{ \neg e? \} X)} \lambda Y. Y \cap (\overleftarrow{\tau} \{ e? \} \circ \overleftarrow{\tau} \{ b \}) Y$$

Forward Abstract Semantics:

abstract $\text{lfp}_X F$ as the limit of $\left\{ \begin{array}{l} X_0^\# \stackrel{\text{def}}{=} X^\# \\ X_{i+1}^\# \stackrel{\text{def}}{=} X_i^\# \nabla F^\#(X_i^\#) \end{array} \right.$ where

- $X^\#$ and $F^\#$ over-approximate X and F
- the widening ∇ ensures convergence
(e.g., remove unstable constraints)

Lower Widening

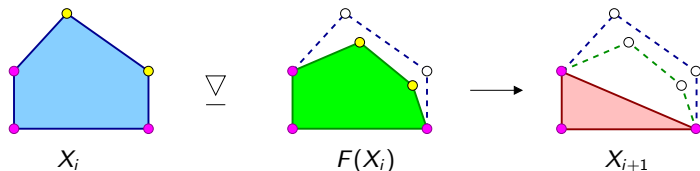


Backward Abstract Semantics:

abstract $\text{gfp}_X F$ as the limit of $\left\{ \begin{array}{l} X_0^\# \stackrel{\text{def}}{=} X^\# \\ X_{i+1}^\# \stackrel{\text{def}}{=} X_i^\# \underline{\nabla} F^\#(X_i^\#) \end{array} \right.$ where

- $X^\#$ and $F^\#$ under-approximate X and F
- $\underline{\nabla}$ is a **lower widening**
 - $A^\# \underline{\nabla} B^\#$ under-approximates $A^\# \cap B^\#$
 - $\forall (X_n^\#)_{n \in \mathbb{N}}$: the sequence $Y_0^\# \stackrel{\text{def}}{=} X_0^\#, Y_{i+1}^\# \stackrel{\text{def}}{=} Y_i^\# \underline{\nabla} X_{i+1}^\#$ stabilizes in finite time: $\exists i: Y_{i+1}^\# = Y_i^\#$

Lower Widening



Example Polyhedral Lower Widening:

- remove unstable generators
- optionally use thresholds

∇ introduced formally by Cousot, along ∇
but no instance until now

∇ *is not a narrowing!*

Expression Abstraction

Forward Expression Abstraction:

Idea: replace $\tau \{ x := e \} R$ with $\tau \{ x := f \} R$ when

- $x := f$ is simpler to abstract than $x := e$
- $\forall \rho \in R: \llbracket e \rrbracket \rho \subseteq \llbracket f \rrbracket \rho$ (soundness)

Backward Expression Abstraction:

Theorem

$$\forall X: \overleftarrow{\tau} \{ x := e \} X \supseteq (\overleftarrow{\tau} \{ v := f \} X) \cap R$$

\implies replace $\overleftarrow{\tau} \{ x := e \}^\# X^\#$ with $(\overleftarrow{\tau} \{ x := f \}^\# X^\#) \cap^\# R^\#$

e.g.: $\overleftarrow{\tau} \{ x := y * z \}^\# P \longrightarrow (\overleftarrow{\tau} \{ x := [0; 1] * z \}^\# P) \cap Q$
if $Q \Rightarrow y \in [0; 1]$

over-approximating e under-approximates $\overleftarrow{\tau} \{ x := e \}!$

(also works on tests)

Prototype

extremely simple proof-of-concept

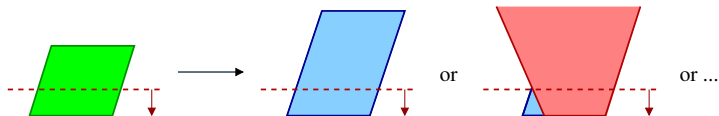
- analyzes a **toy**-language
- forward **invariant** + backward **sufficient condition** analysis
- abstract interpretation by induction on the syntax
- polyhedral abstractions based on **Apron**
- **on-line analysis**: <http://www.di.ens.fr/~mine/banal>
- **sources freely available** (in Ocaml)

Example analyses:

- simple loops (`for (...) j = j + [0;1]`)
- bubble sort (Cousot Halbwachs 78)

Short-Comings

- Abstraction of **tests**



⇒ how to choose **now** to **maximize** the **end result**?

- **Lower widening**

⇒ widening is the usual suspect

- Experiments

Related Work

- $w(l)p$ calculus (Dijkstra 75, Morris 97)
- (finite) model-checking (Dams 96)
- over-approximating backward abstract interpretation (Bourdoncle 93, Rival 05)
- under-approximations
 - exact disjunctive completions
 - path enumeration (Moy 08)
 - domains closed by complement (Lev-Ami et al. 07)
- higher-order abstract interpretation
 - abstract lower closure operators (Massé 02)
 - under-approximation on powersets (Schmidt 04)

Conclusion

It **seems possible** to:

- infer **sound sufficient conditions**
- for **non-deterministic, infinite-state** programs
- using non-trivial **under-approximations**
- on **infinite-state** abstract domains

Our contribution:

- 1 **general properties** for compositional analysis design
- 2 example abstract transfer functions on **polyhedra**

Much more work to do to make it practical!