

Scaling up in static verification by abstract interpretation

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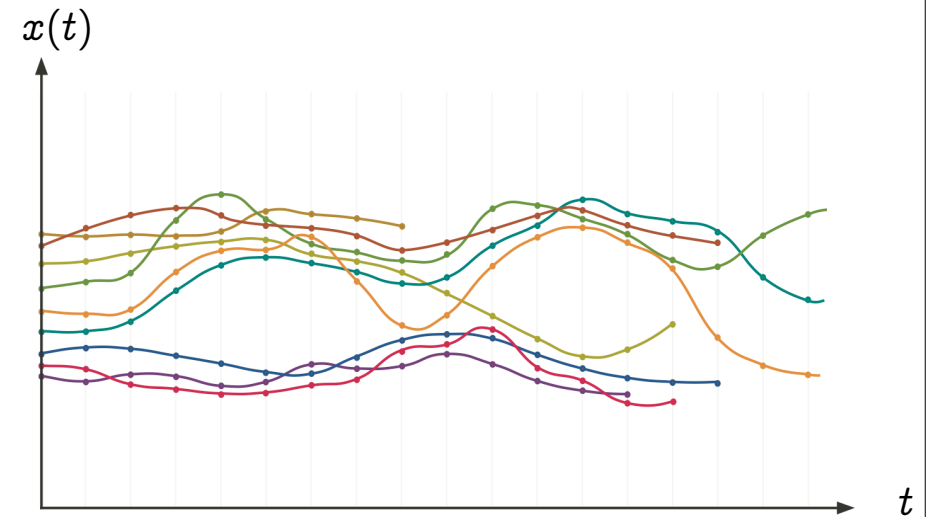
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Semantics

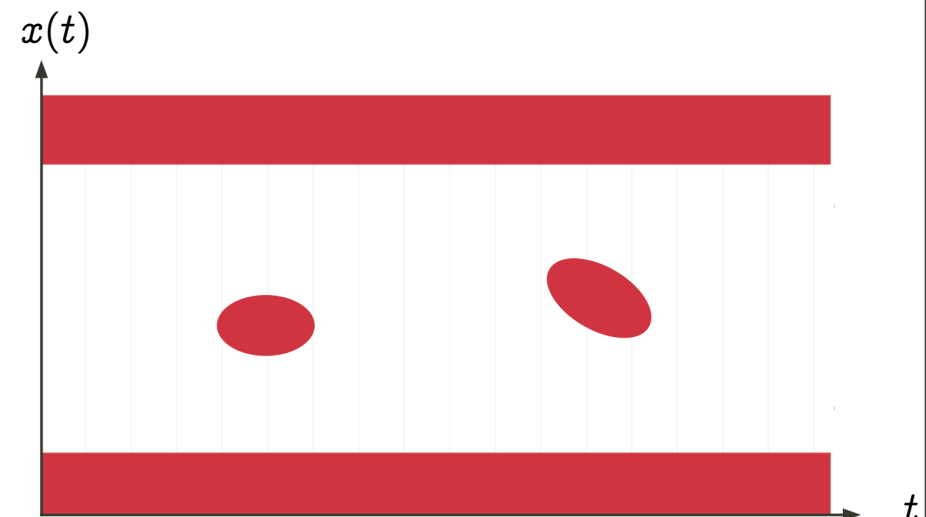


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A very short intuitive and
informal introduction to
abstract interpretation

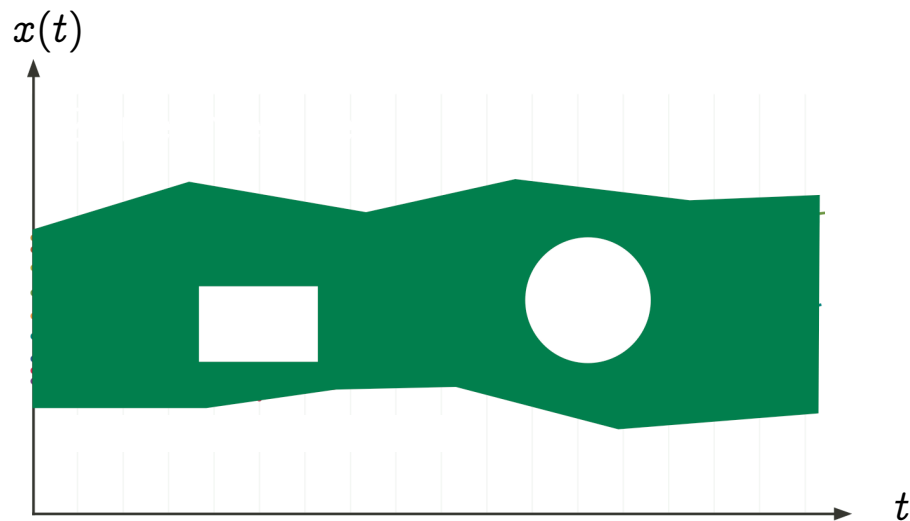
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Safety specification



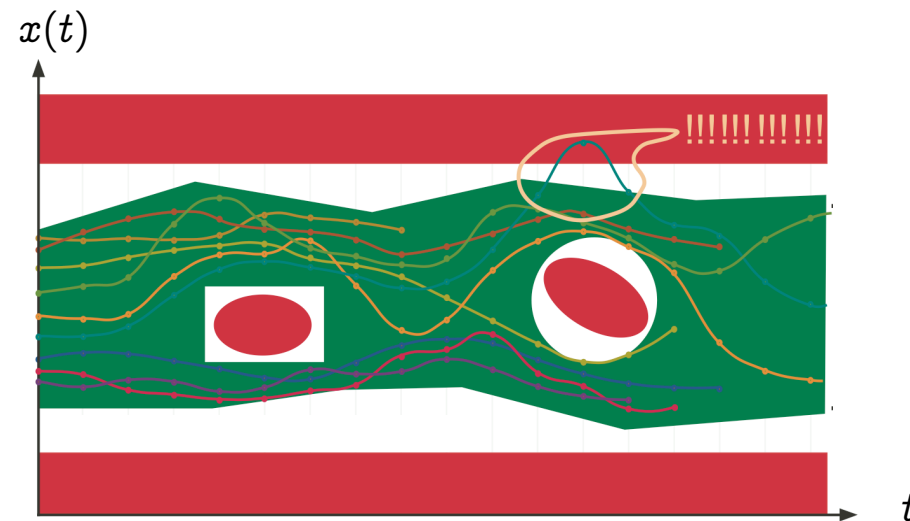
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Abstraction



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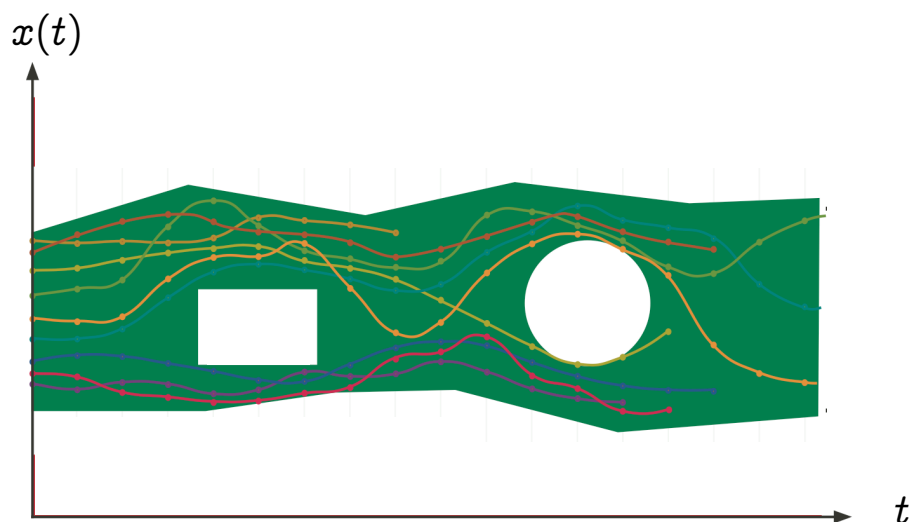
Example of unsound abstraction⁽¹⁾



(1) excluded by abstract interpretation theory

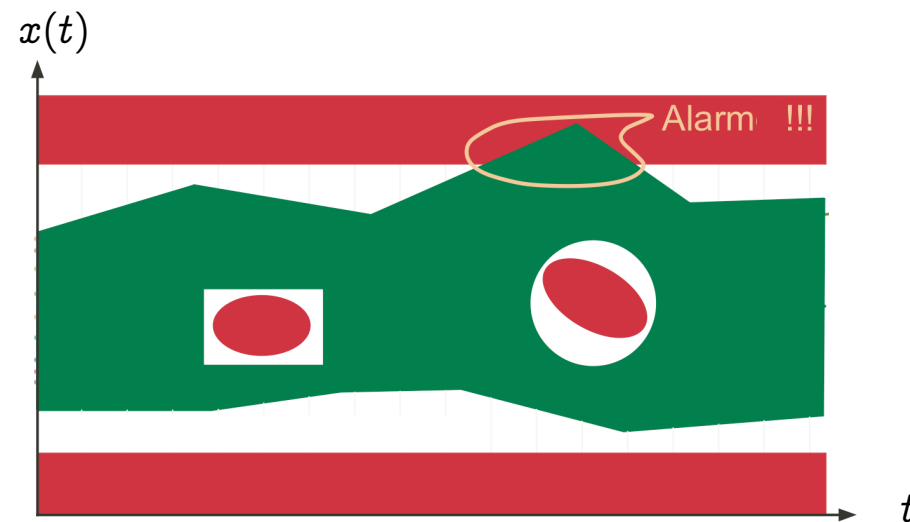
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Soundness of the abstraction



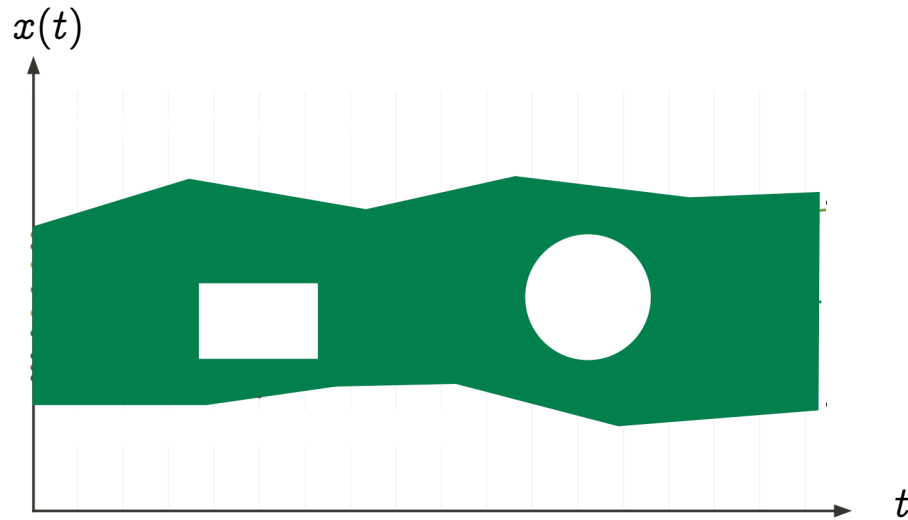
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True or false alarm?



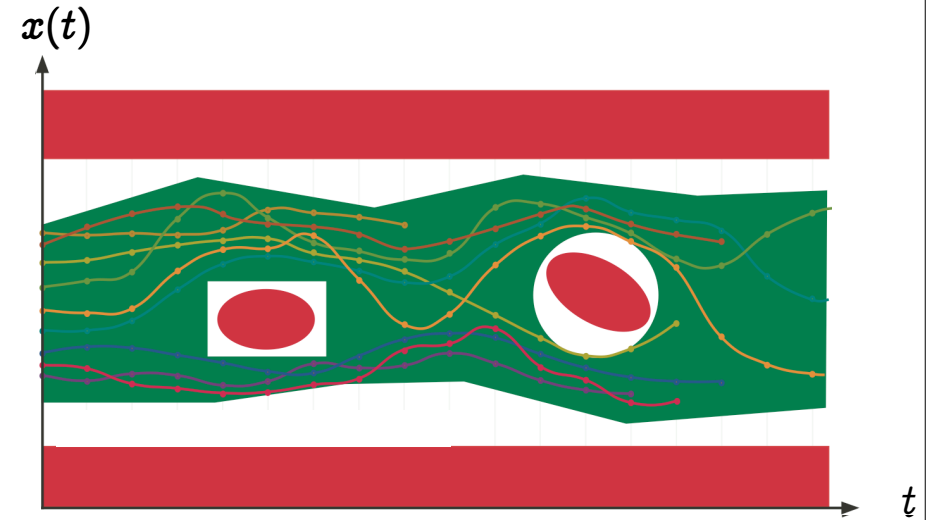
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Refinement



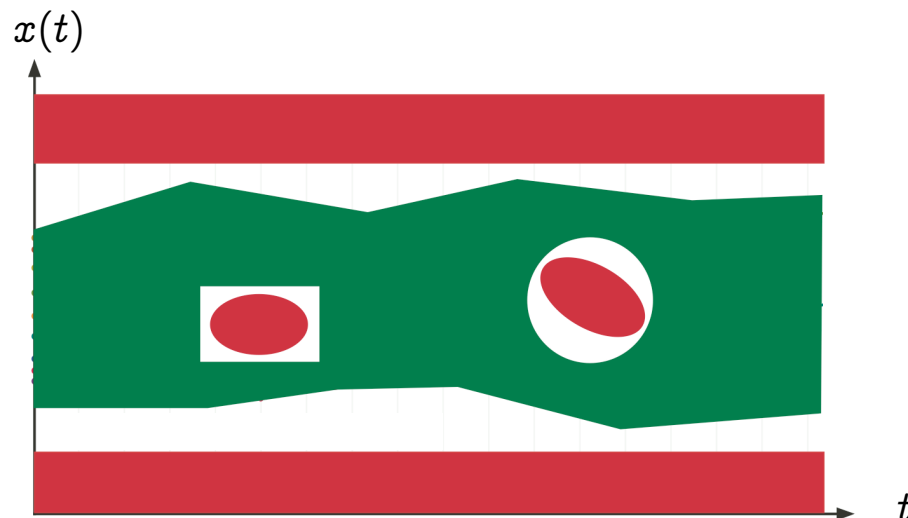
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Formal proof in the abstract is sound



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Formal proof in the abstract



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Example: inference of an invariant

```

{y ≥ 0} ← hypothesis
x := y
{I(x, y)} ← loop invariant
while (x > 0) do
  x := x - 1;
od
    
```

Abstract fixpoint equation:

$$I(x, y) = x \geq 0 \wedge (x = y \vee I(x + 1, y)) \quad (\text{i.e. } I = \mathbf{F}^\sharp(I)^{(1)})$$

Equivalent Floyd-Naur-Hoare verification conditions:

$$\begin{aligned}
 (y \geq 0 \wedge x = y) &\implies I(x, y) && \text{initialisation} \\
 (I(x, y) \wedge x > 0 \wedge x' = x - 1) &\implies I(x', y) && \text{iteration}
 \end{aligned}$$

⁽¹⁾ We look for the most precise invariant I , implying all others, that is $\text{fp} \equiv \mathbf{F}^\sharp$.

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Iterates with widening

$$I = \bigsqcup_{n \rightarrow \infty} F^{\sharp n}(\text{false})$$

$$I^0(x, y) = \text{false}$$

$$I^1(x, y) = x \geq 0 \wedge (x = y \vee I^0(x+1, y)) \\ = 0 \leq x = y$$

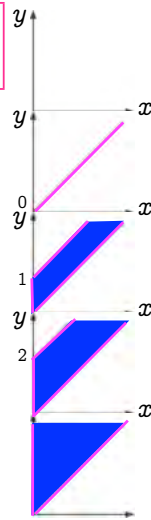
$$I^2(x, y) = x \geq 0 \wedge (x = y \vee I^1(x+1, y)) \\ = 0 \leq x \leq y \leq x+1$$

$$I^3(x, y) = x \geq 0 \wedge (x = y \vee I^2(x+1, y)) \\ = 0 \leq x \leq y \leq x+2$$

$$I^4(x, y) = I^2(x, y) \nabla I^3(x, y) \leftarrow \text{widening} \\ = 0 \leq x \leq y$$

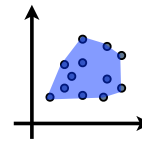
$$I^5(x, y) = x \geq 0 \wedge (x = y \vee I^4(x+1, y)) \\ = I^4(x, y) \text{ fixed point!}$$

The invariants are computer representable with octagons!

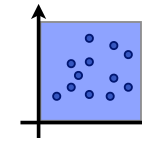


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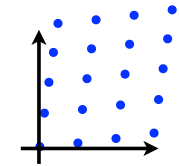
Examples of abstractions not used by Astrée



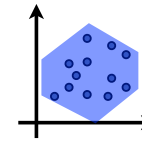
Polyhedra (too expensive)



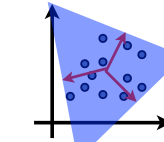
Signs (too imprecise)



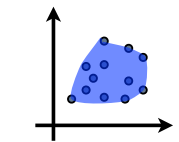
Linear congruences (too expensive)



Zonotopes (inclusion?)



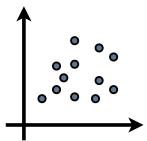
Support functions (widening?)



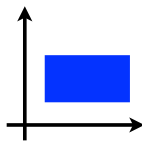
Convex sets (algorithmics?)

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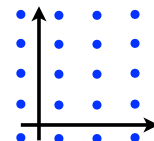
Examples of abstractions used by Astrée



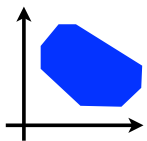
semantics
set of points



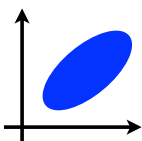
intervals
 $x \in [a, b]$



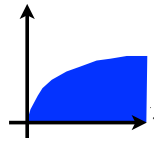
simple congruences
 $x \equiv a[b]$



octagons
 $\pm x \pm y \leq a$



ellipsoids
 $ax^2 + by^2 + cxy \leq d$



exponentials
 $x(t) \leq a^{bt}$

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Example of analysis by Astrée

```
% cat retro.c
typedef enum {FALSE=0, TRUE=1} BOOL;
BOOL FIRST;
volatile BOOL SWITCH;
volatile float E;
float P, X, A, B;

void dev( )
{ X=E;
  if (FIRST) { P = X; }
  else
    { P = (P - (((2.0 * P) - A) - B)
            * 5.0e-03)); };
  B = A;
  if (SWITCH) {A = P;}
  else {A = X;}
}
```

```
void main()
{ FIRST = TRUE;
  while (TRUE) {
    dev( );
    FIRST = FALSE;
    __ASTREE_wait_for_clock();
  }
}

% cat retro.config
__ASTREE_volatile_input((E [-15.0, 15.0]));
__ASTREE_volatile_input((SWITCH [0,1]));
__ASTREE_max_clock((3600000));

astree -exec-fn main -config-sem retro.config
retro.c |& grep "|P|" | tail -n 1
|P| <=1.0000002*((15. +
5.8774718e-39/(1.0000002-1))*(1.0000002)clock -
5.8774718e-39/(1.0000002-1) + 5.8774718e-39 <=
23.039353
```

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Scaling up in static verification by abstract interpretation

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Implicit versus explicit specifications

Implicit specifications:

- e.g. absence of runtime errors
- no extra burden on the end-user, cost effective

Explicit specifications:

- e.g. temporal specification
- available specification may not be exploitable (Simulink model)
- additional cost to maintain both the system, its specification and a formal specification

Size scalability through specification restriction:

- e.g. verification → bug-finding

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Scalability in verification

Many different **measures of scalability** of an automatic verifier:

- **size scalability** : size of the systems to be verified
- **design scalability** : required efforts from the verifier designer
- **use scalability** : required efforts from the verifier end-user
- **re-use scalability** : reproducibility under small changes (maintenance)

Scalability → can only be measured with respect to well-specified and often incomparable objectives

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Universal versus domain specific systems

Universal systems:

- e.g. programs in a universal programming language

Domain specific systems:

- e.g. synchronous real-time control command programs
- **excludes**
 - recursive function calls,
 - dynamic memory allocations,
 - backward gotos,
 - long jumps,
 - concurrency,
 - very complicated data structures,
 - or highly nested loops.

Size scalability through application domain restriction

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Precision versus incompleteness

Precise abstractions:

- always exist for a given system (but not an infinite family of systems)
- very hard to find

General abstractions:

- universal use (e.g. intervals)
- hard to avoid **false alarms**

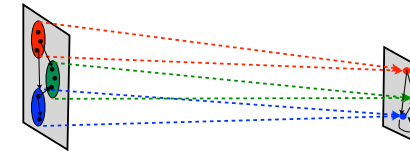
Size-scalability through abstraction → how to get rid of false alarms?

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Models versus properties (cont'd)

Model abstraction:

- **Concrete model** → **Abstract model**
e.g. transition system abstract transition system



The same tools can be used in the concrete and the abstract

.../...

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System modelization

Abstract model generation:

- **System** → **Model**
e.g. C program transition system

Abstract model generator:

- **System generator** → **Model generator**
e.g. C language C program → abstract model

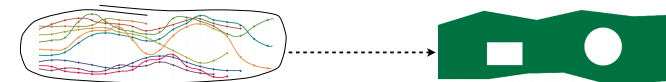
Use-scalability through automatic model generation → more burden is put on the static analyzer designer

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Models versus properties (cont'd)

Property abstraction:

- **Concrete properties generator** **Abstract properties generator**
e.g. trace collecting semantics polyhedral analysis



Use-scalability through more general abstractions → more burden is put on the static analyzer designer

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Universal versus domain specific abstractions

Universal abstractions:

- e.g. invariance
- hard to automatize

Domain-specific abstractions:

- e.g. filters, integrators in system control
- domain-specific knowledge can be incorporated in the verification process

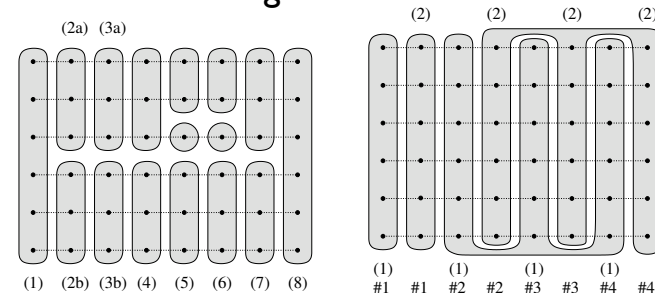
Scalability by domain-specific abstractions → helpful (essential?), more burden on the static analyzer designer

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Global versus local abstractions (cont'd)

Local abstractions:

- e.g. invariance through local trace abstraction



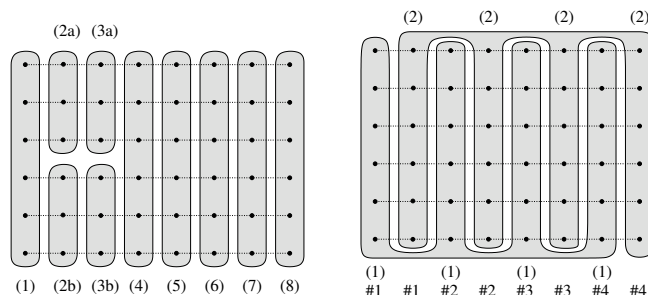
Scalability through local abstraction → much more powerful in the abstract, more burden of the static analyzer designer

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Global versus local abstractions

Global abstractions:

- e.g. invariance through reachability
- the same abstraction everywhere in time and space
- provably sound and complete (Cook proved relative completeness of Hoare logic)



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Finite versus infinite abstractions

Finite abstractions:

- Universal representations/algorithms
- No approximation in the abstract

Infinite abstractions:

- No universal representations/algorithms
- Require further approximation in the abstract to ensure termination (widening/narrowing)

Equivalent for the static analysis of a given system, provably more powerful for an infinite family of systems

Scalability through infinite abstractions → more burden of the static analyzer designer

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Unique versus multiple abstraction encoding

Universal representation of abstractions:

- e.g. terms for provers, BDDs, etc

Abstraction-specific representations:

- e.g. no efficient universal representation of geometric objects

Scalability through multiple abstraction representations

→ more burden on the analyzer designer but very efficient algorithms require specific data structures

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Refinement

- **Refinement** is necessary (very hard to find good abstractions)
- **Automatic abstraction-refinement** is costly and may not terminate
- **Manual refinement** by end-user & static analyzer designer

e.g. in Astrée:

- Parameterized abstract domains
- Parameterized widenings
- (Automated) analysis directives
- Local improvements of the abstract predicate transformers
- Inclusion of new abstract domains

Scalability through manual refinement → more burden of the analyzer designer (& user)

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Handling disjunction with multiple abstractions

Unique abstraction representation:

- embeds conjunction

Multiple abstraction-specific representations:

- conjunction of multiple abstractions with different abstractions is not immediate
- reduced product (optimal in precision)
- partial reduction is cheaper (although sub-optimal)

e.g. Astrée:

- fixed reduction order among abstract domains
- some domains broadcast information to all domains
- some domains ask questions to others, in turn

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Conclusion

- Finding the **appropriate abstraction** is undecidable and extremely difficult
- Universal/all purpose tools with automatic refinement **only have the program as guidelines for automatic refinement, this is hardly sufficient** (e.g. linear program → ellipsoidal abstraction for programs)
- Human guided parametrization/directives/choice of **combinations of domain specific abstractions** is an alternative which proved successful but ...
- Ultimately **new abstractions** (hence computer property representations & manipulation algorithms) **must be designed manually but are reusable**

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