

« An Informal Overview of Abstract Interpretation »

Patrick Cousot

Jerome C. Hunsaker Visiting Professor
Massachusetts Institute of Technology
Department of Aeronautics and Astronautics

`cousot@mit.edu`
`www.mit.edu/~cousot`

Course 16.399: “Abstract interpretation”

<http://web.mit.edu/afs/athena.mit.edu/course/16/16.399/www/>



What is static analysis
by abstract interpretation?



Example of static analysis (input)

```
n := n0;
i := n;
while (i <> 0 ) do
  j := 0;
  while (j <> i) do
    j := j + 1
  od;
  i := i - 1
od
```



Example of static analysis (output)

```
{n0>=0}
  n := n0;
{n0=n,n0>=0}
  i := n;
{n0=i,n0=n,n0>=0}
  while (i <> 0 ) do
    {n0=n,i>=1,n0>=i}
    j := 0;
    {n0=n,j=0,i>=1,n0>=i}
    while (j <> i) do
      {n0=n,j>=0,i>=j+1,n0>=i}
      j := j + 1
      {n0=n,j>=1,i>=j,n0>=i}
    od;
    {n0=n,i=j,i>=1,n0>=i}
    i := i - 1
    {i+1=j,n0=n,i>=0,n0>=i+1}
  od
{n0=n,i=0,n0>=0}
```



Example of static analysis (safety)

```

{n0>=0}
  n := n0;
{n0=n,n0>=0}
  i := n;
{n0=i,n0=n,n0>=0}
  while (i < 0) do
    {n0=n,i>=1,n0>=i}
    j := 0;
    {n0=n,j=0,i>=1,n0>=i}
    while (j < i) do
      {n0=n,j>=0,i>=j+1,n0>=i}
      j := j + 1
      {n0=n,j>=1,i>=j,n0>=i}
    od;
    {n0=n,i=j,i>=1,n0>=i}
    i := i - 1
    {i+1=j,n0=n,i>=0,n0>=i+1}
  od
{n0=n,i=0,n0>=0}

```

n0 must be initially nonnegative
(otherwise the program does not
terminate properly)

← j < n0 so no upper overflow

← i > 0 so no lower overflow



Static analysis by abstract interpretation

Verification: define and prove automatically a **property** of the **possible behaviors** of a complex computer program;

Abstraction: the reasoning/calculus can be done on an **abstraction** of these behaviors dealing only with those elements of the behaviors related to the considered property;

Theory: abstract interpretation.



Example of static analysis

Verification: absence of runtime errors;

Abstraction: polyhedral abstraction (affine inequalities);

Theory: abstract interpretation.



Potential impact of runtime errors

- 50% of the **security attacks** on computer systems are through **buffer overruns**¹!
- Embedded computer system **crashes** easily result from **overflows** of various kinds.



¹ See for example the Microsoft Security Bulletin MS02-065, MS04-011, etc.



A very informal introduction to the principles of abstract interpretation

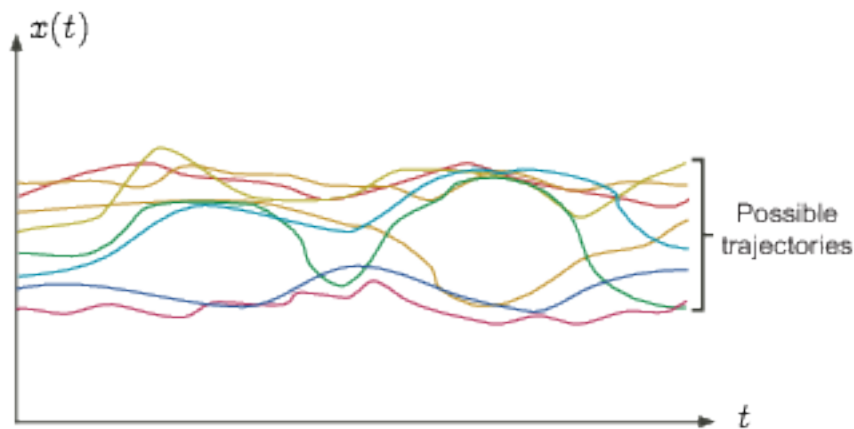


Semantics

The *concrete semantics* of a program formalizes (is a mathematical model of) the set of all its possible executions in all possible execution environments.



Graphic example: Possible behaviors



Undecidability

- The concrete mathematical semantics of a program is an “infinite” mathematical object, *not computable*;
- All non trivial questions on the concrete program semantics are *undecidable*.

Example: Kurt Gödel argument on termination

- Assume `termination(P)` would always terminates and returns true iff P always terminates on all input data;
- The following program yields a contradiction

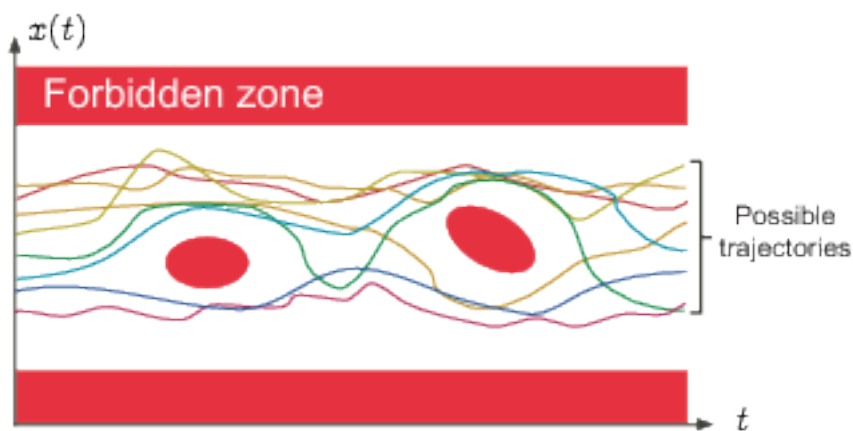
```
P ≡ while termination(P) do skip od.
```

Graphic example: Safety properties

The *safety properties* of a program express that no possible execution in any possible execution environment can reach an *erroneous state*.



Graphic example: Safety property



Safety proofs

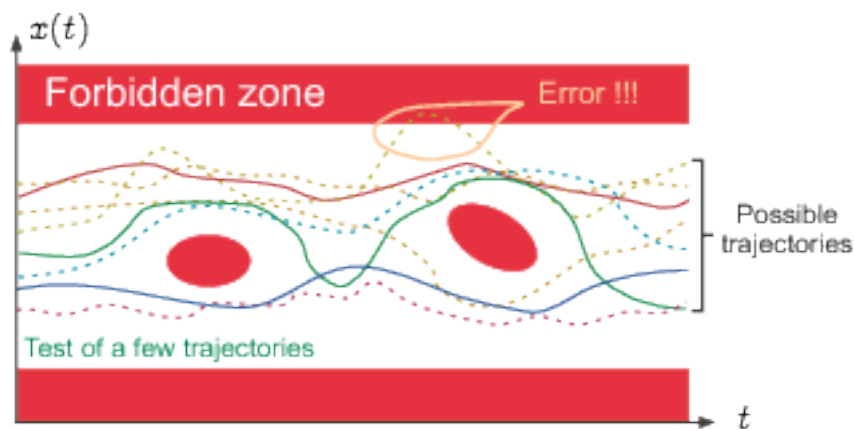
- A **safety proof** consists in proving that the intersection of the program concrete semantics and the forbidden zone is empty;
- **Undecidable** problem (the concrete semantics is not computable);
- Impossible to provide completely automatic answers with finite computer resources and neither human interaction nor uncertainty on the answer ².

² e.g. probabilistic semantics.

Test/debugging

- consists in considering a subset of the possible executions;
- not a correctness proof;
- **absence of coverage** is the main problem.

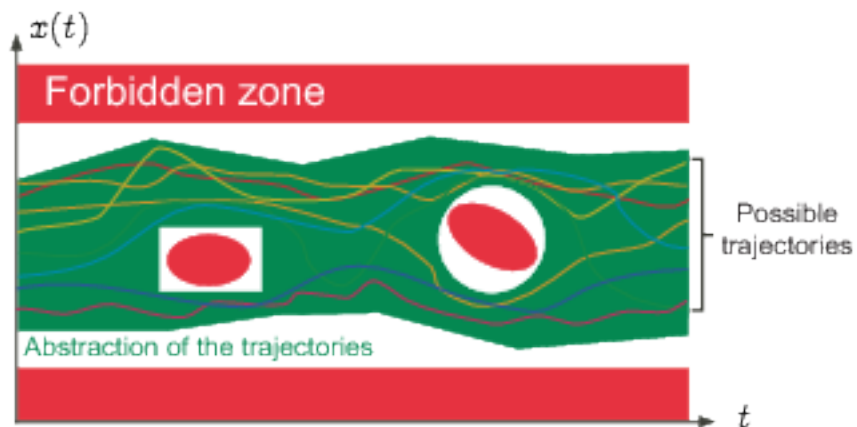
Graphic example: Property test/simulation



Abstract interpretation

- consists in considering an *abstract semantics*, that is to say a superset of the concrete semantics of the program;
- hence the abstract semantics *covers all possible concrete cases*;
- *correct*: if the abstract semantics is safe (does not intersect the forbidden zone) then so is the concrete semantics.

Graphic example: Abstract interpretation



Formal methods

Formal methods are abstract interpretations, which differ in the way to obtain the abstract semantics:

- “*model checking*”:
 - the abstract semantics is given manually by the user;
 - in the form of a finitary model of the program execution;
 - can be computed automatically, by techniques relevant to static analysis.

- “*deductive methods*”:
 - the abstract semantics is specified by verification conditions;
 - the user must provide the abstract semantics in the form of inductive arguments (e.g. invariants);
 - can be computed automatically by methods relevant to static analysis.
- “*static analysis*”: the abstract semantics is computed automatically from the program text according to pre-defined abstractions (that can sometimes be tailored automatically/manually by the user).

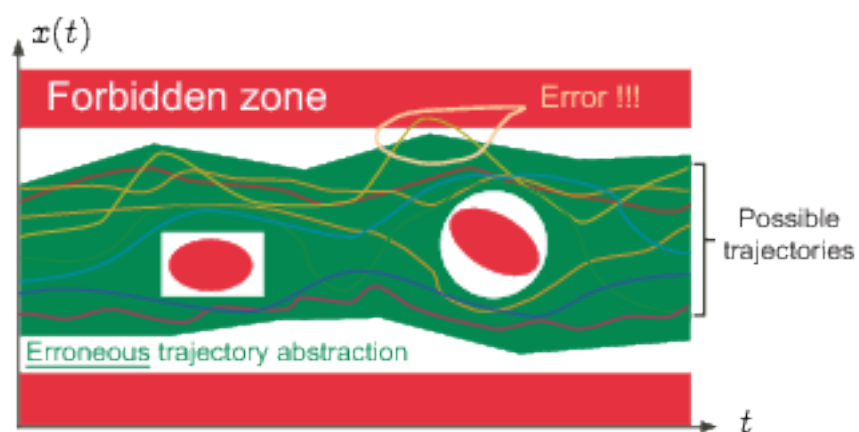


Required properties of the abstract semantics

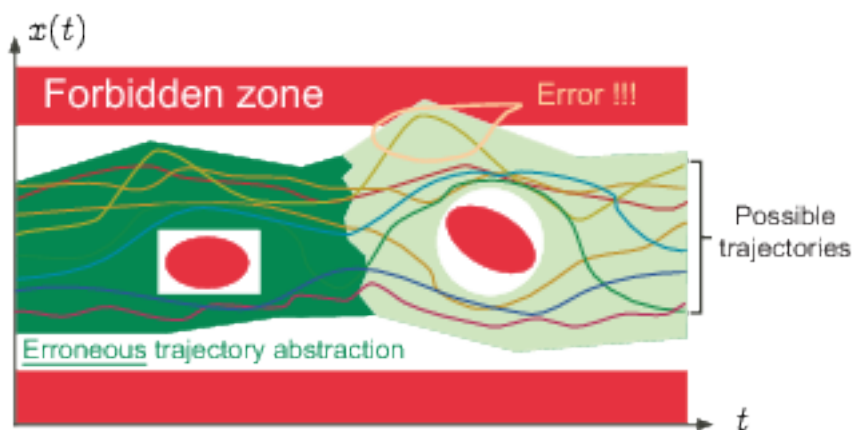
- *sound* so that no possible error can be forgotten;
- *precise* enough (to avoid false alarms);
- as *simple/abstract* as possible (to avoid combinatorial explosion phenomena).



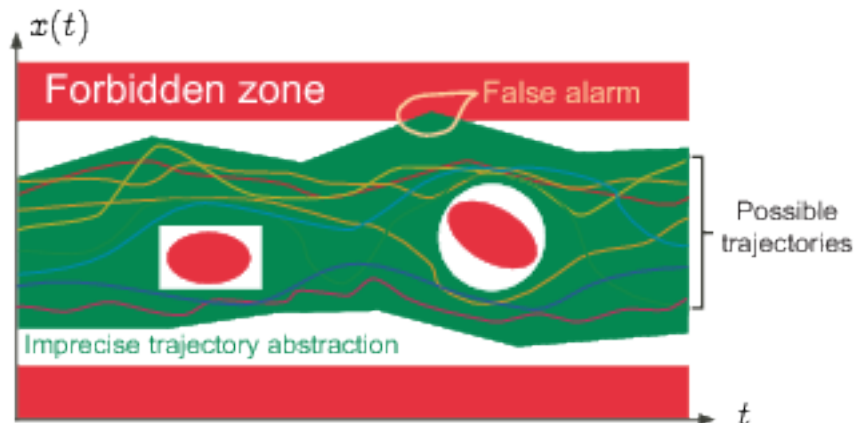
Graphic example: Erroneous abstraction — I



Graphic example: Erroneous abstraction — II



Graphic example: Imprecision \Rightarrow false alarms

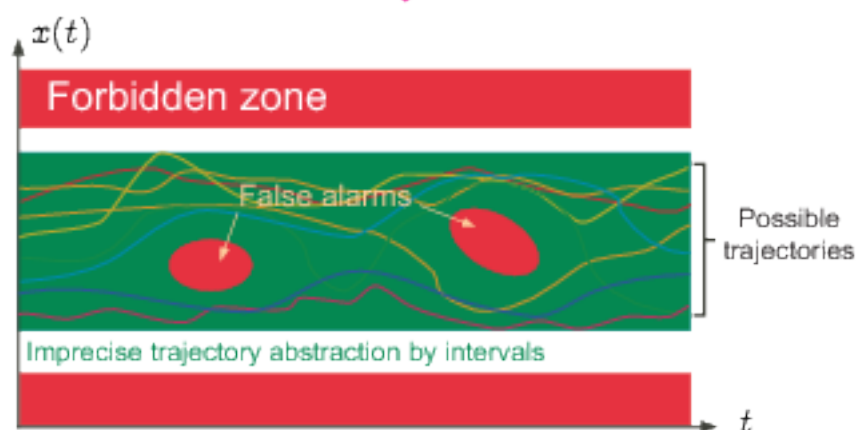


Abstract domains

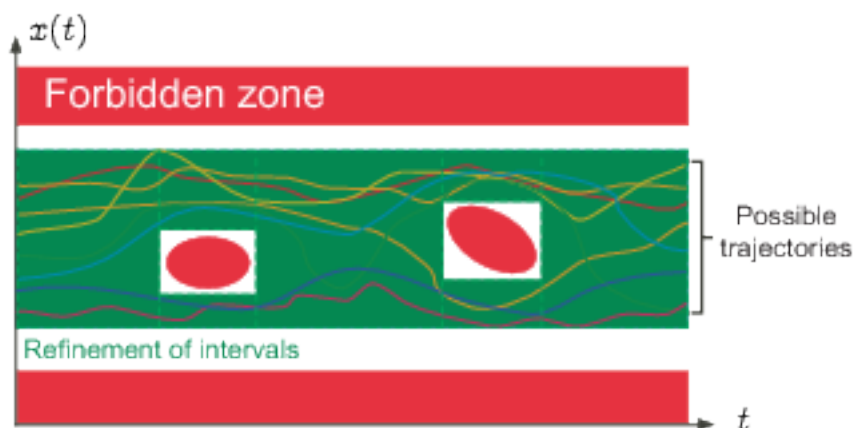
Standard abstractions

- that serve as a **basis** for the design of static analyzers:
 - abstract program data,
 - abstract program basic operations;
 - abstract program control (iteration, procedure, concurrency, ...);
- can be **parametrized** to allow for manual adaptation to the application domains.

Graphic example: Standard abstraction by intervals



Graphic example: A more refined abstraction



A very informal introduction to static analysis algorithms



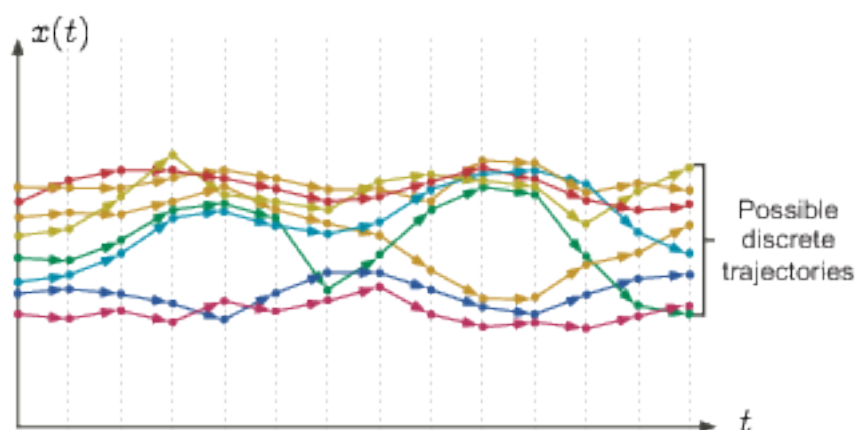
Trace semantics



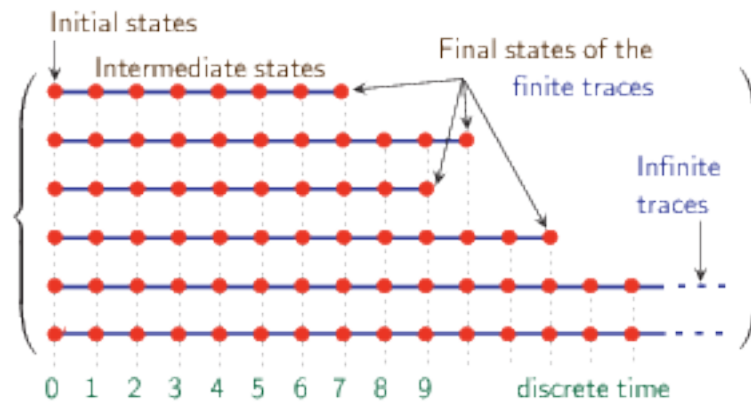
Trace semantics

- Consider (possibly infinite) **traces** that is series of states corresponding to executions described by discrete transitions;
- The collection of all such traces, starting from the initial states, is the **trace semantics**.

Graphic example: Small-steps transition semantics

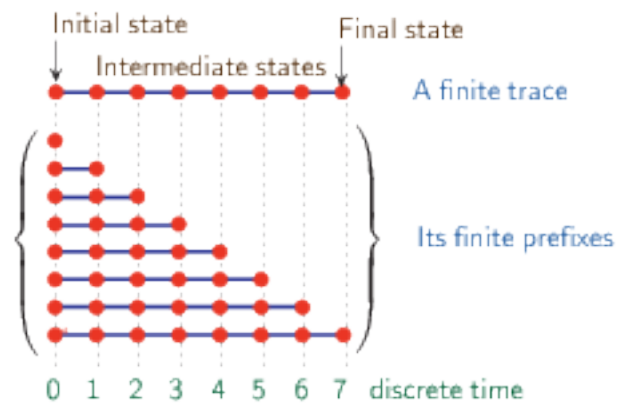


Trace semantics, intuition

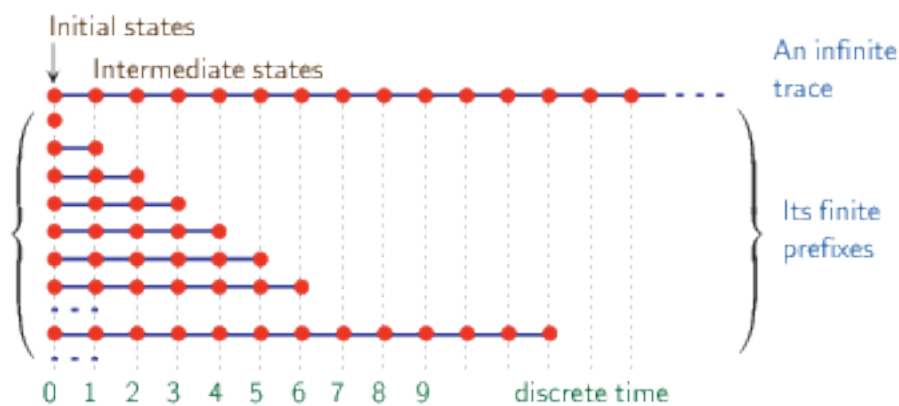


Prefix trace semantics

Prefixes of a finite trace



Prefixes of an infinite trace



Prefix trace semantics

Trace semantics: maximal finite and infinite behaviors

Prefix trace semantics: finite prefixes of the maximal behaviors



Abstraction

This is an **abstraction**. For example:

Trace semantics: $\{a^n b \mid n \geq 0\}$

Prefix trace semantics: $\{a^n \mid n \geq 0\} \cup \{a^n b \mid n \geq 0\}$

Is there of possible behavior with infinitely many successive **a**?

- Trace semantics: **no**
- Prefix trace semantics: **I don't know**



Prefix trace semantics in fixpoint form



Least Fixpoint Prefix Trace Semantics

$\text{Prefixes} = \{ \bullet \mid \bullet \text{ is an initial state} \}$
 $\cup \{ \bullet \xrightarrow{\quad} \dots \xrightarrow{\quad} \bullet \mid \bullet \xrightarrow{\quad} \dots \xrightarrow{\quad} \bullet \in \text{Prefixes} \}$
 $\&\& \bullet \xrightarrow{\quad} \bullet \text{ is a transition step} \}$

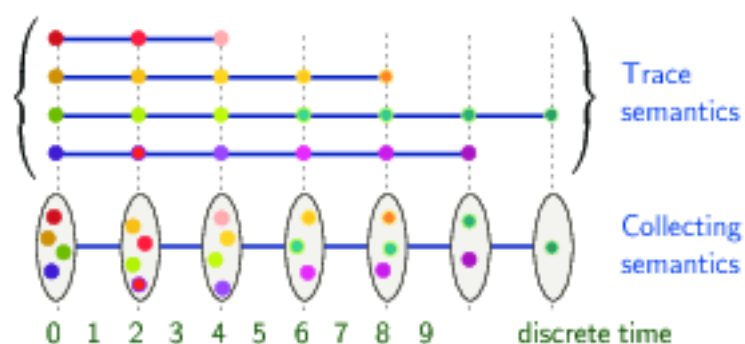
- In general, the equation $\text{Prefixes} = F(\text{Prefixes})$ may have multiple solutions;
- Choose the least one for **subset inclusion** \subseteq .
- **Abstractions** of this equation lead to **effective iterative analysis algorithms**.



Collecting semantics

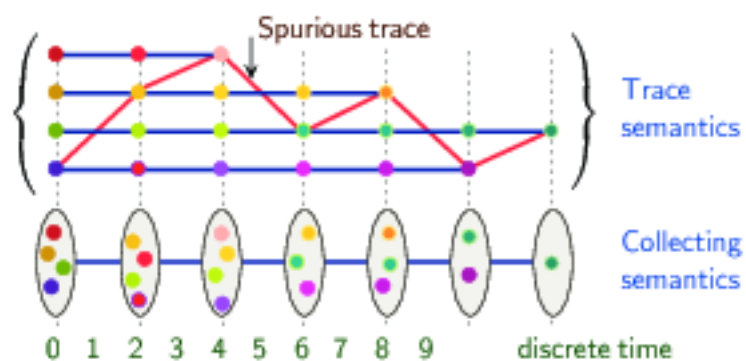
Collecting semantics

- Collect all states that can appear on some trace at any given discrete time:

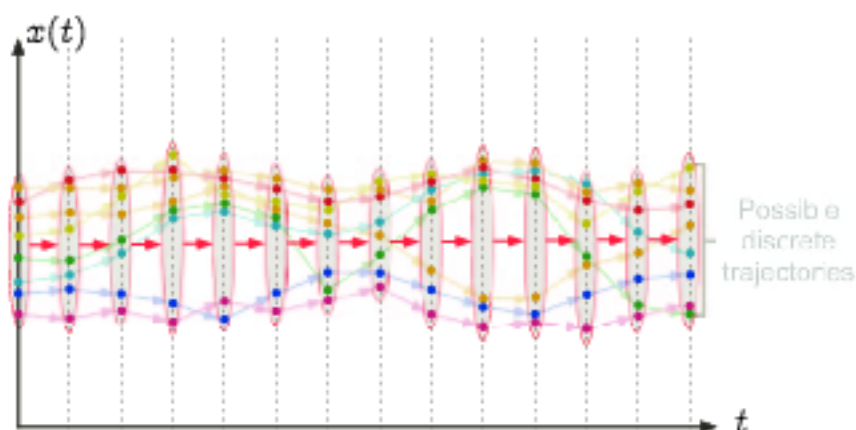


Collecting abstraction

- This is an **abstraction**. Does the red trace exists?
Trace semantics: **no**, collecting semantics: **I don't know**.



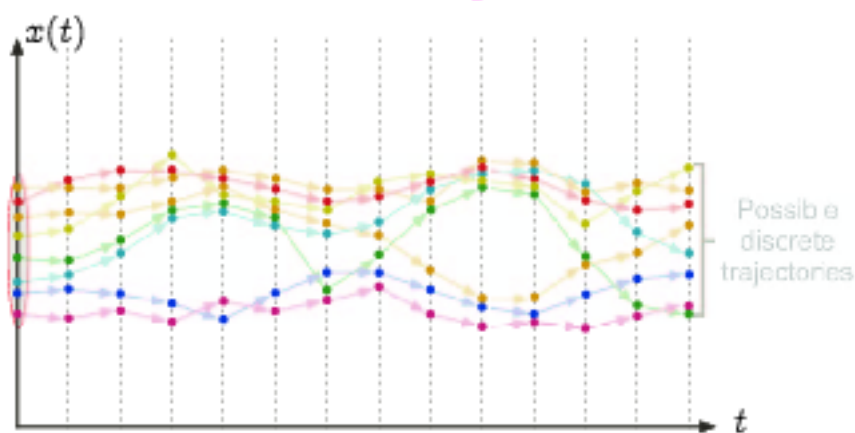
Graphic example: collecting semantics



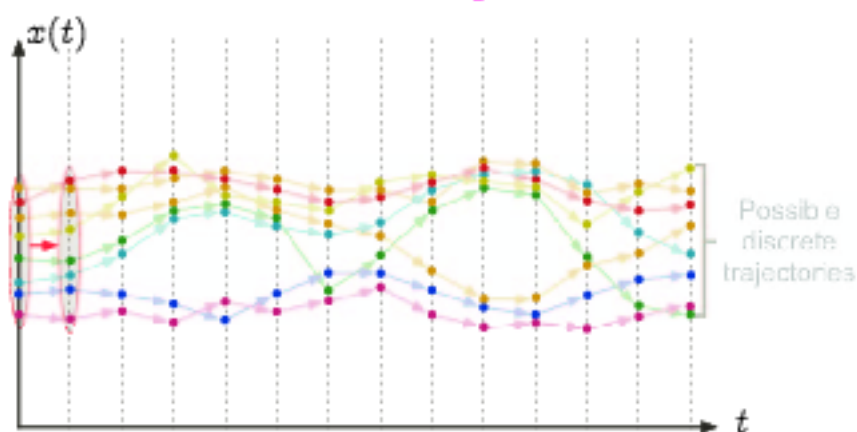
Collecting semantics in fixpoint form



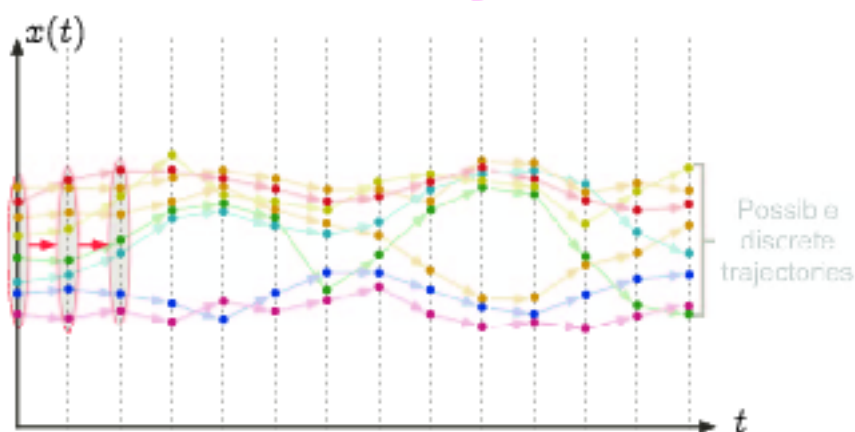
Graphic example: collecting semantics in fixpoint form



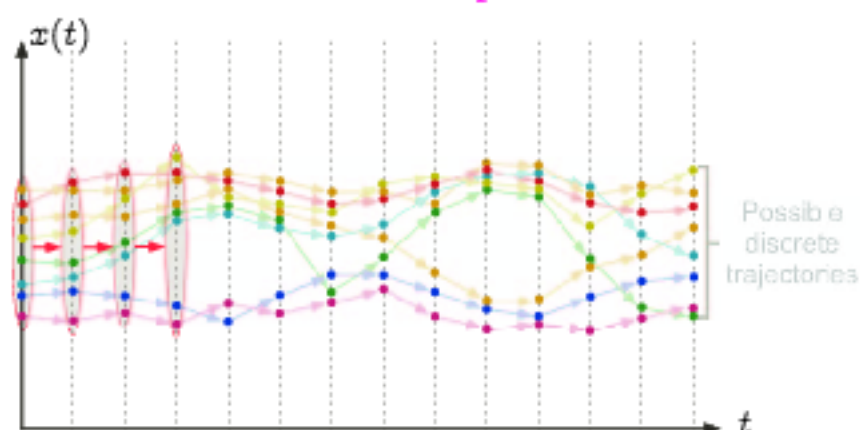
Graphic example: collecting semantics
in fixpoint form



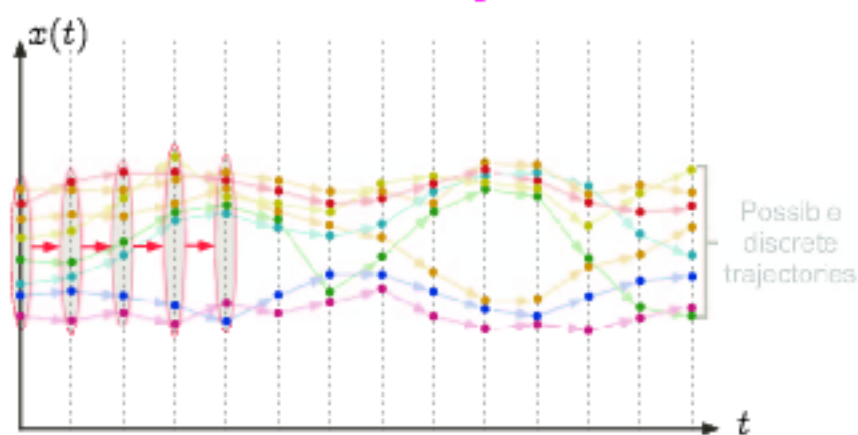
Graphic example: collecting semantics
in fixpoint form



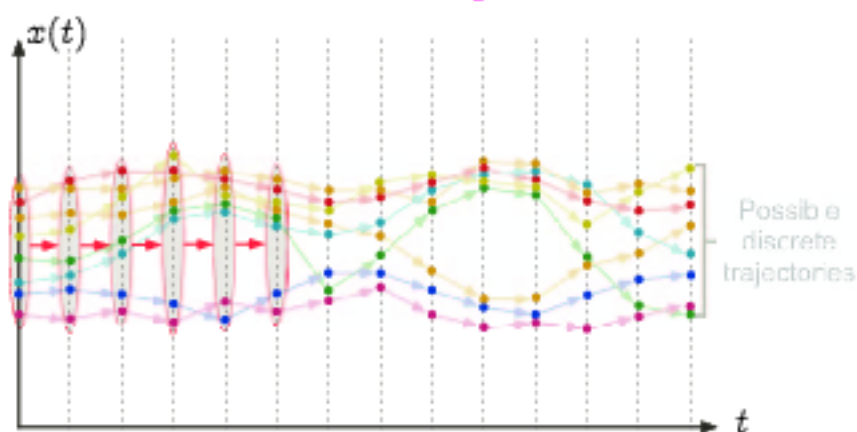
Graphic example: collecting semantics
in fixpoint form



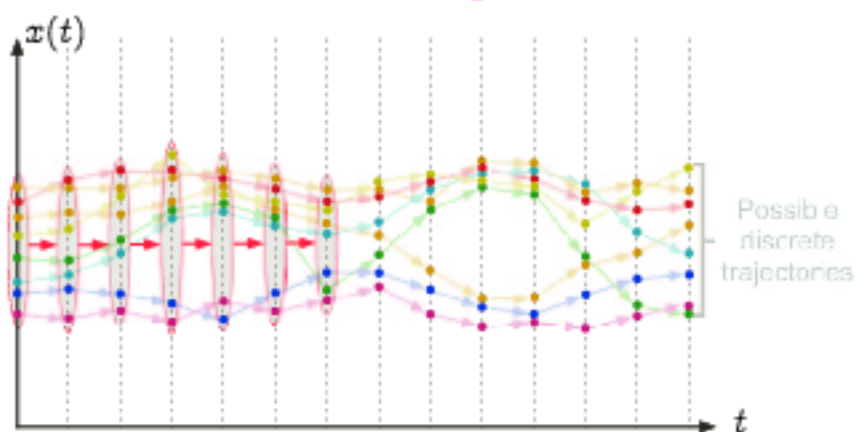
Graphic example: collecting semantics
in fixpoint form



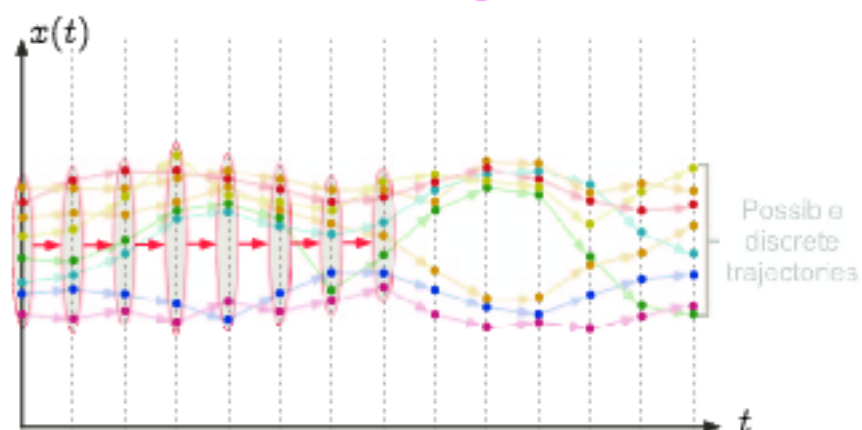
Graphic example: collecting semantics
in fixpoint form



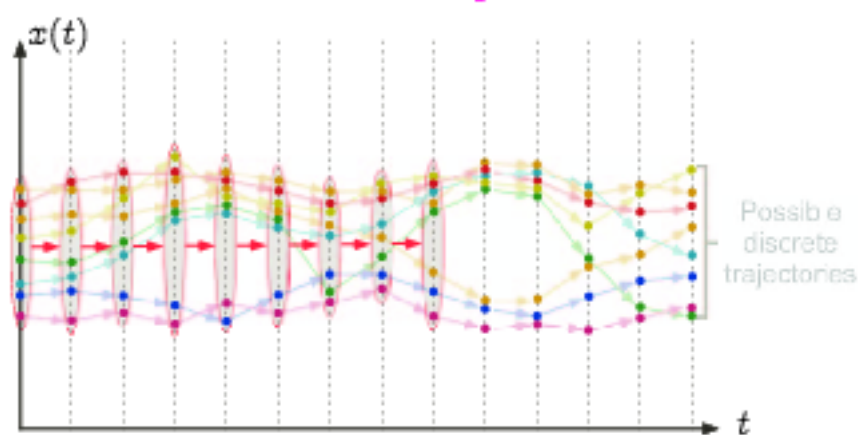
Graphic example: collecting semantics
in fixpoint form



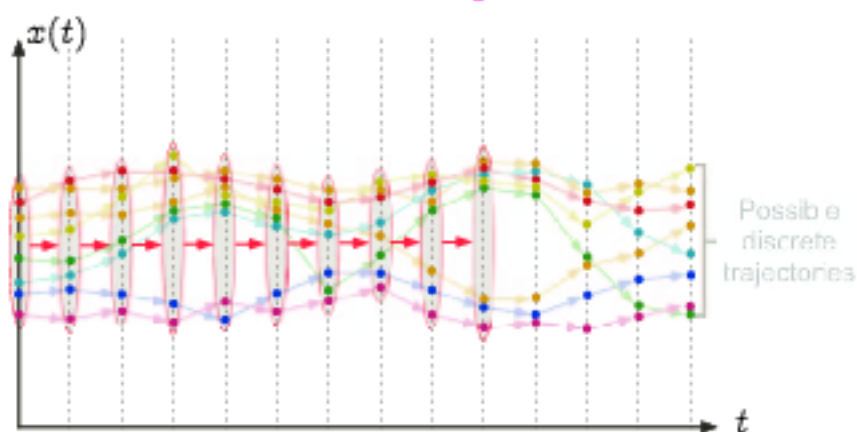
Graphic example: collecting semantics
in fixpoint form



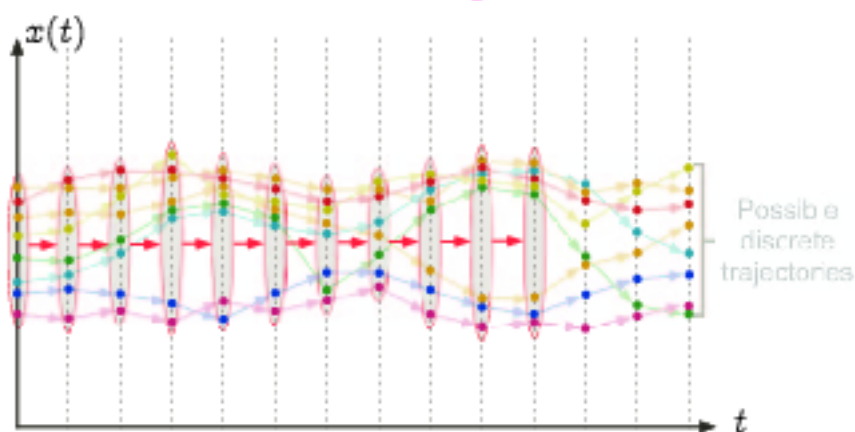
Graphic example: collecting semantics
in fixpoint form



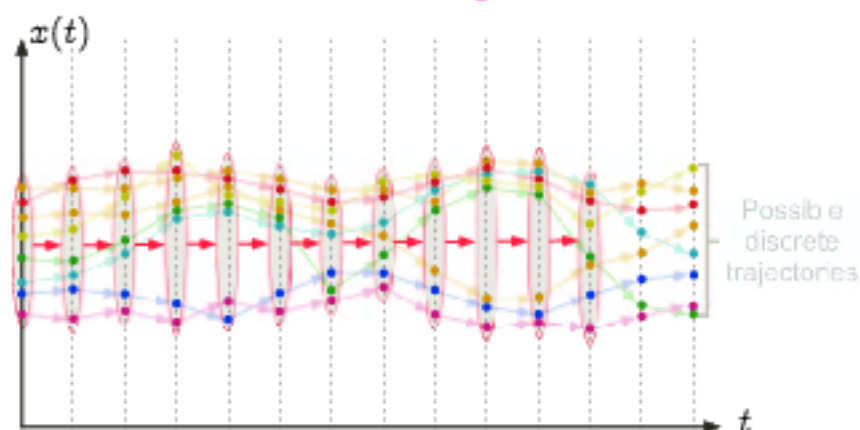
Graphic example: collecting semantics in fixpoint form



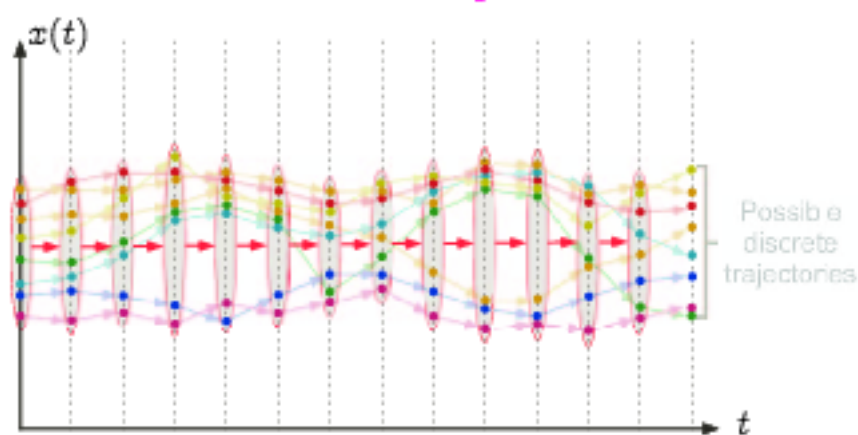
Graphic example: collecting semantics in fixpoint form



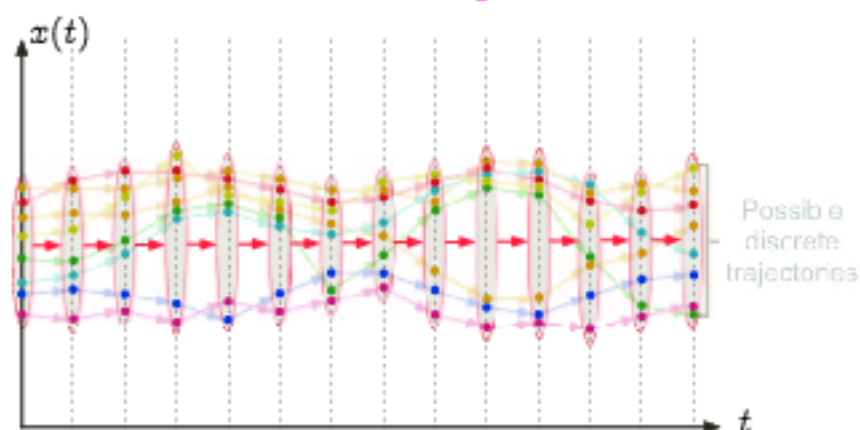
Graphic example: collecting semantics
in fixpoint form



Graphic example: collecting semantics
in fixpoint form

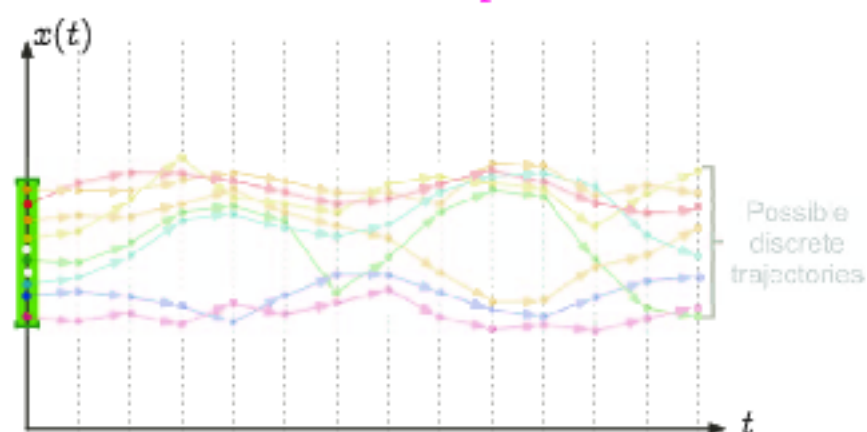


Graphic example: collecting semantics
in fixpoint form

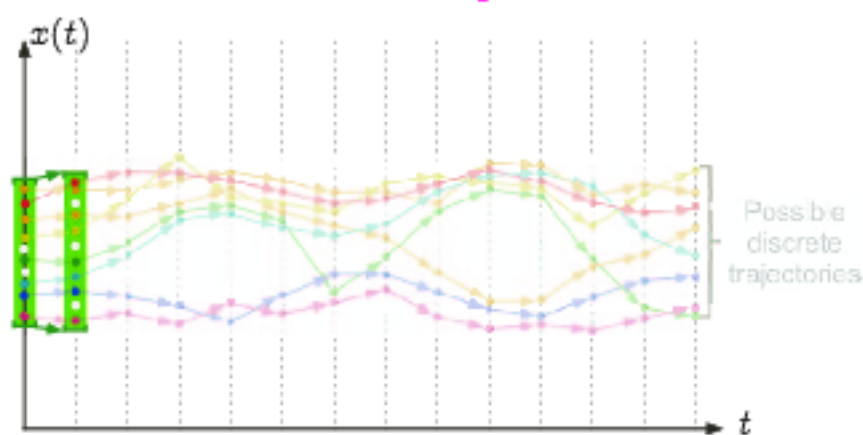


Interval Abstraction
(in iterative fixpoint form)

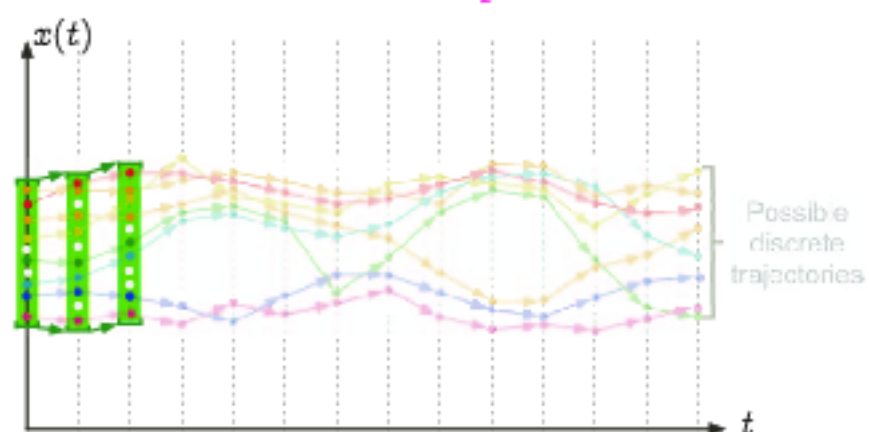
Graphic example: traces of intervals
in fixpoint form



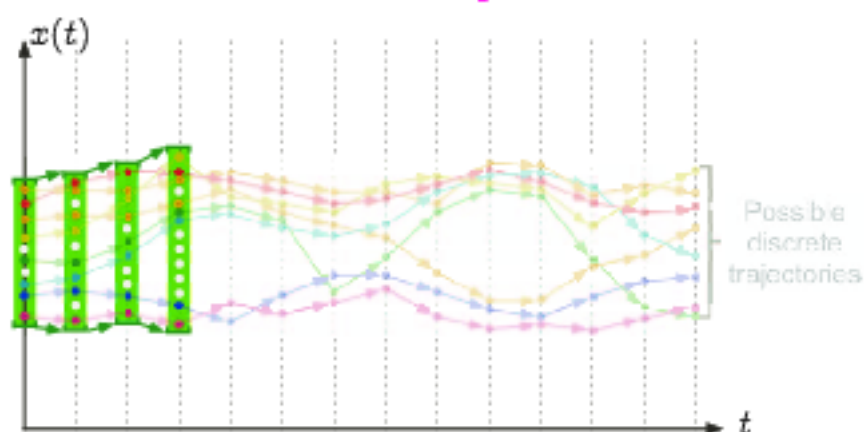
Graphic example: traces of intervals
in fixpoint form



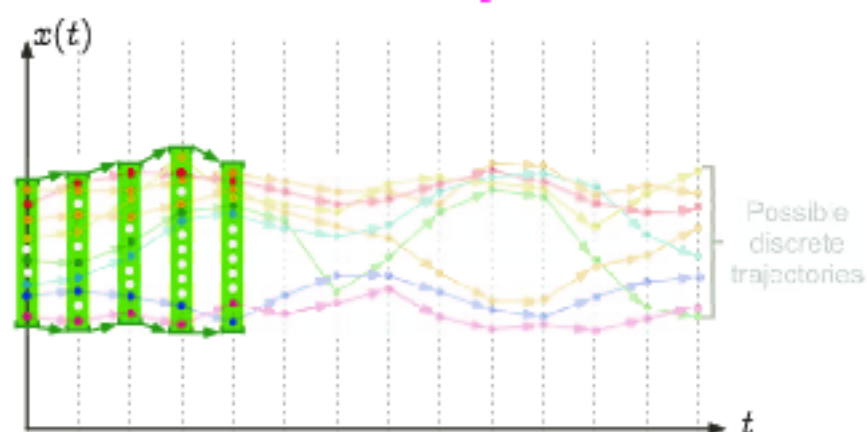
Graphic example: traces of intervals
in fixpoint form



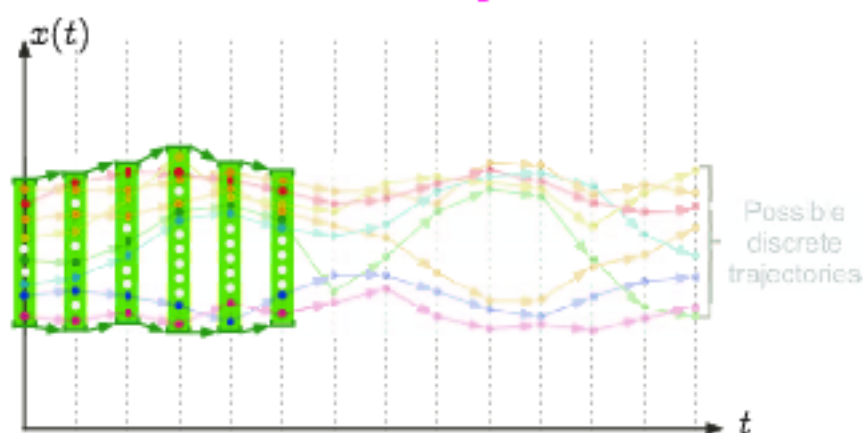
Graphic example: traces of intervals
in fixpoint form



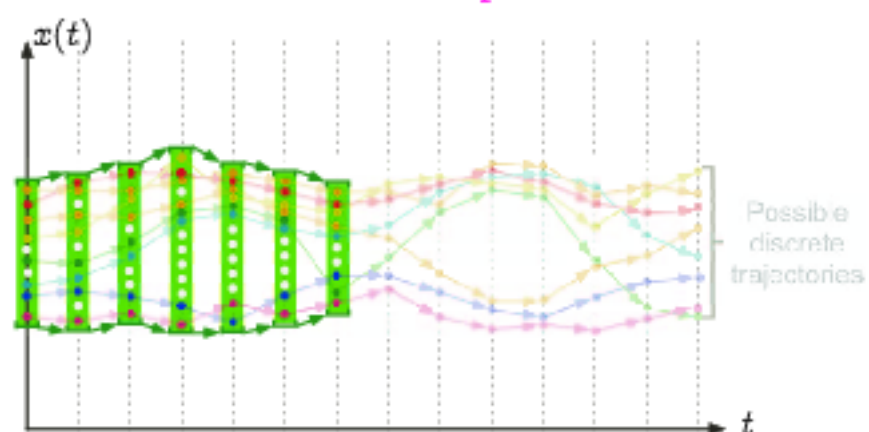
Graphic example: traces of intervals
in fixpoint form



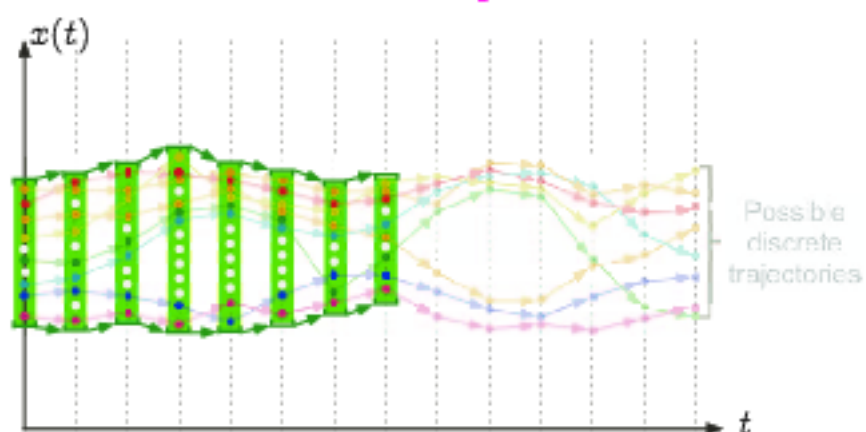
Graphic example: traces of intervals
in fixpoint form



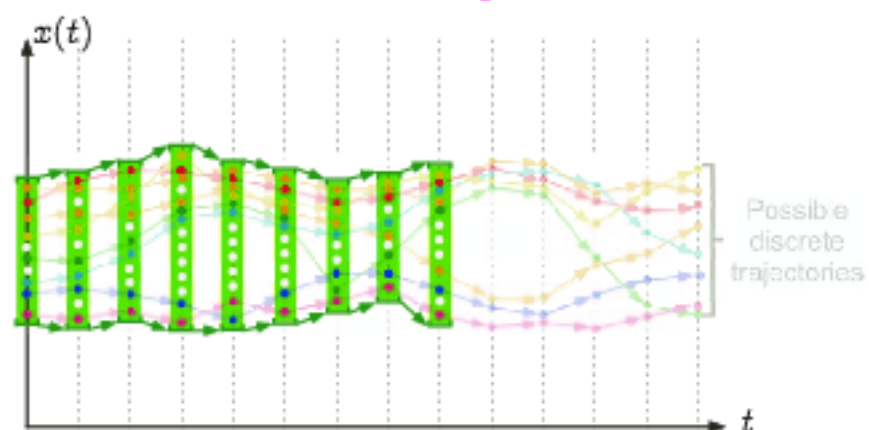
Graphic example: traces of intervals
in fixpoint form



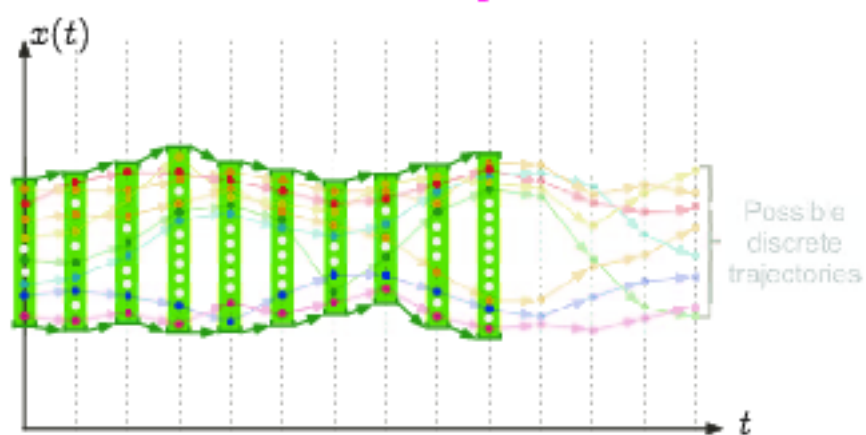
Graphic example: traces of intervals
in fixpoint form



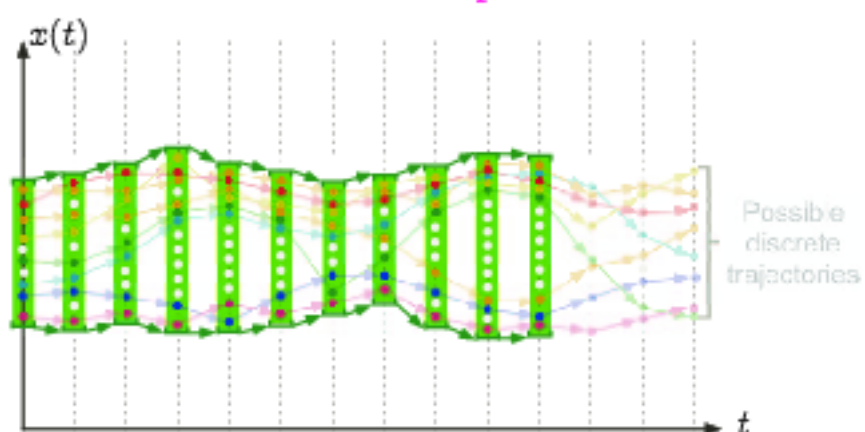
Graphic example: traces of intervals
in fixpoint form



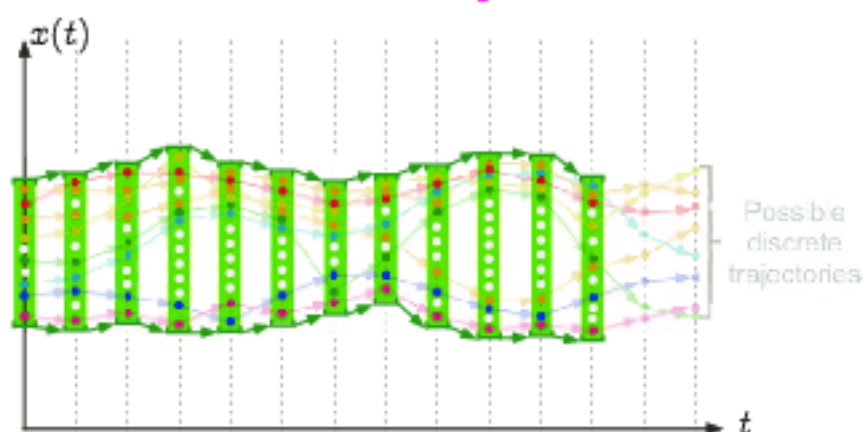
Graphic example: traces of intervals
in fixpoint form



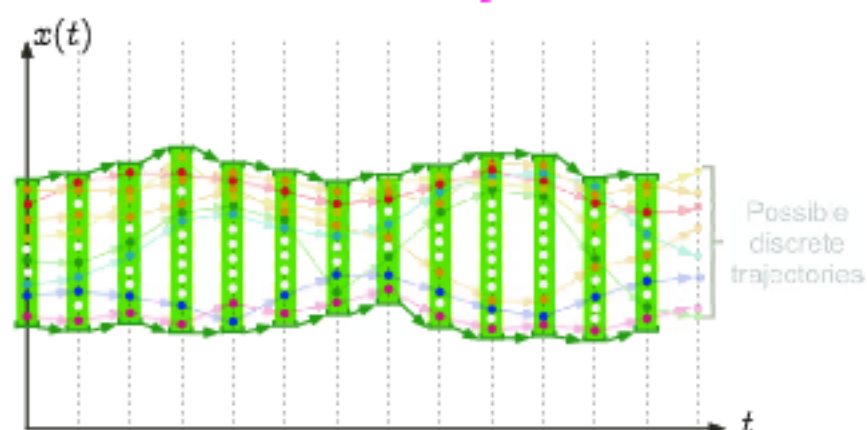
Graphic example: traces of intervals
in fixpoint form



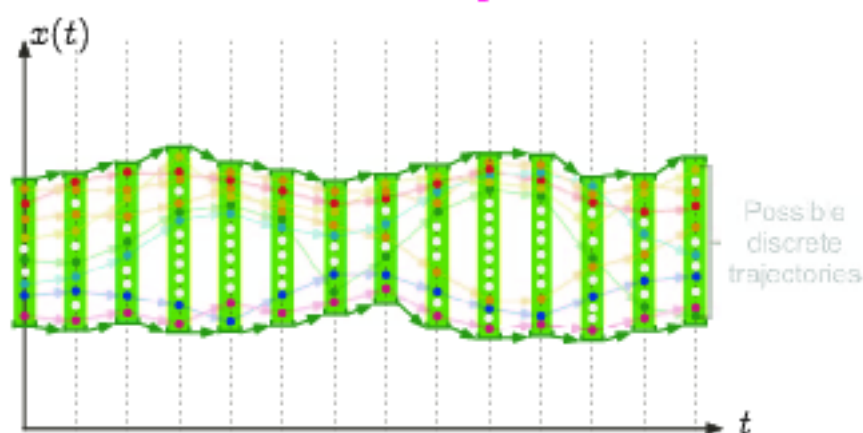
Graphic example: traces of intervals
in fixpoint form



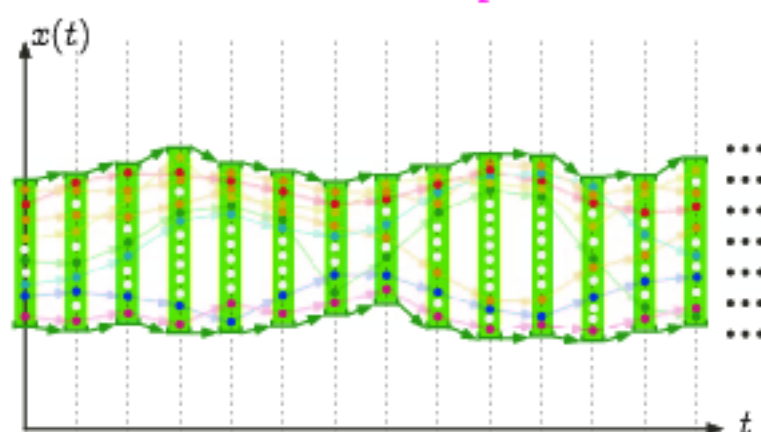
Graphic example: traces of intervals
in fixpoint form



Graphic example: traces of intervals
in fixpoint form



Graphic example: traces of intervals in fixpoint form

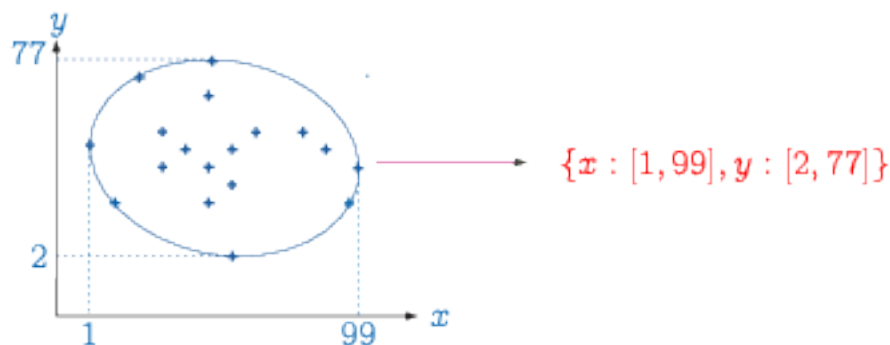


Abstraction by Galois connections

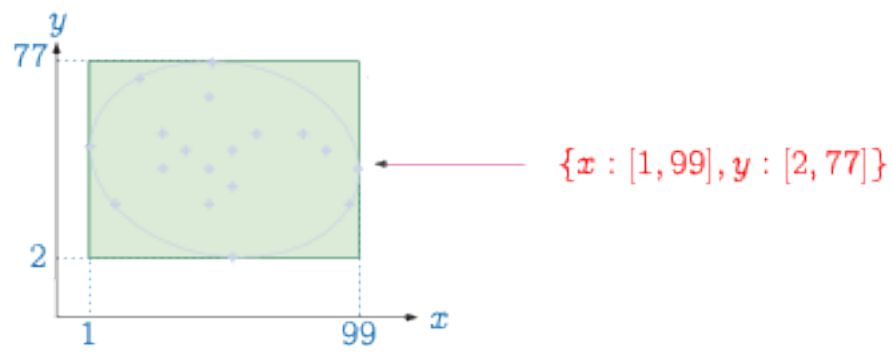
Abstracting sets (i.e. properties)

- Choose an **abstract domain**, replacing sets of objects (states, traces, ...) S by their abstraction $\alpha(S)$
- The **abstraction function** α maps a set of concrete objects to its abstract interpretation;
- The inverse **concretization function** γ maps an abstract set of objects to concrete ones;
- **Forget no concrete objects**: (abstraction from above) $S \subseteq \gamma(\alpha(S))$.

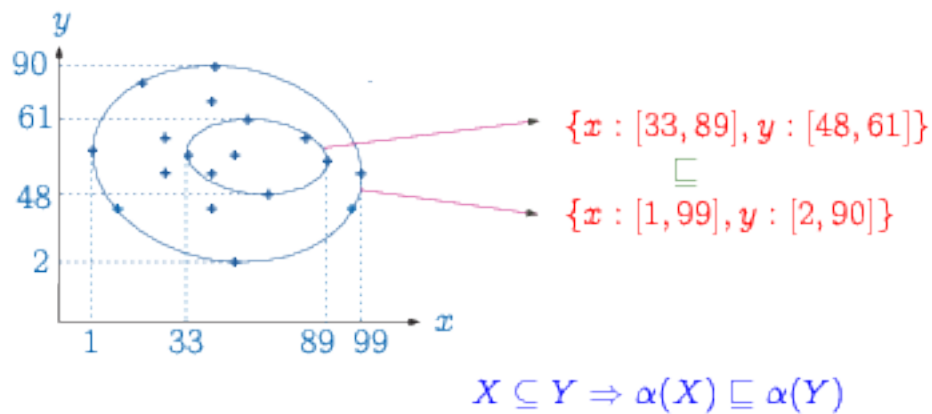
Interval abstraction α



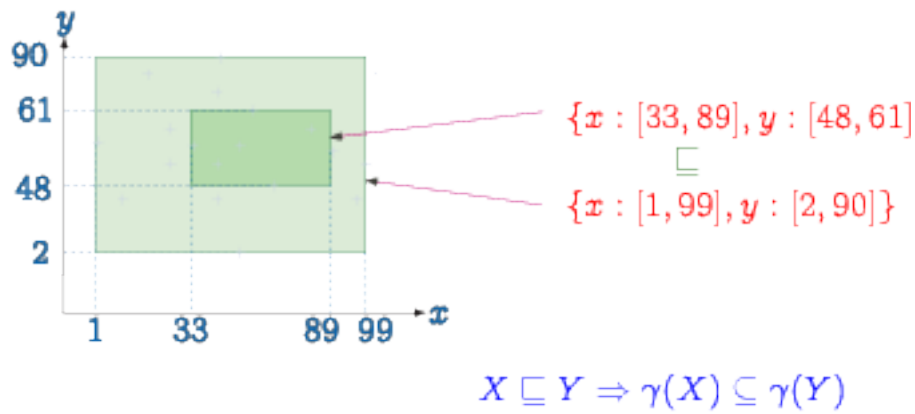
Interval concretization γ



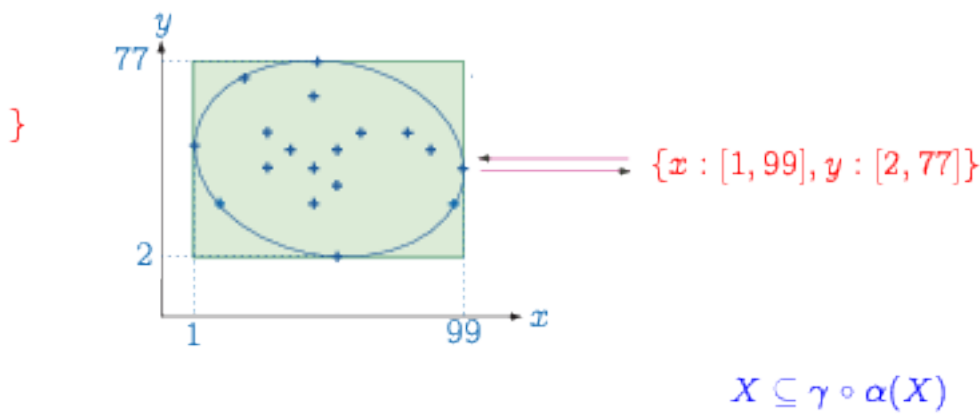
The abstraction α is monotone



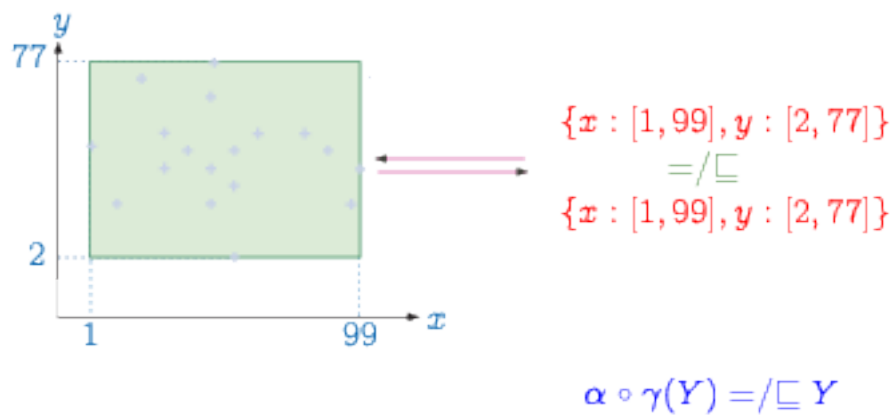
The concretization γ is monotone



The $\gamma \circ \alpha$ composition is extensive



The $\alpha \circ \gamma$ composition is reductive



Correspondance between concrete and abstract properties

- The pair $\langle \alpha, \gamma \rangle$ is a **Galois connection**:

$$\langle \wp(S), \subseteq \rangle \xrightarrow[\alpha]{\gamma} \langle \mathcal{D}, \sqsubseteq \rangle$$

- $\langle \wp(S), \subseteq \rangle \xrightarrow[\alpha]{\gamma} \langle \mathcal{D}, \sqsubseteq \rangle$ when α is onto (equivalently $\alpha \circ \gamma = 1$ or γ is one-to-one).

Galois connection

$$\langle \mathcal{D}, \subseteq \rangle \xrightarrow[\alpha]{\gamma} \langle \overline{\mathcal{D}}, \sqsubseteq \rangle$$

$$\text{iff} \quad \forall x, y \in \mathcal{D} : x \subseteq y \implies \alpha(x) \sqsubseteq \alpha(y)$$

$$\wedge \forall \overline{x}, \overline{y} \in \overline{\mathcal{D}} : \overline{x} \sqsubseteq \overline{y} \implies \gamma(\overline{x}) \subseteq \gamma(\overline{y})$$

$$\wedge \forall x \in \mathcal{D} : x \subseteq \gamma(\alpha(x))$$

$$\wedge \forall \overline{y} \in \overline{\mathcal{D}} : \alpha(\gamma(\overline{y})) \sqsubseteq \overline{y}$$

$$\text{iff} \quad \forall x \in \mathcal{D}, \overline{y} \in \overline{\mathcal{D}} : \alpha(x) \sqsubseteq \overline{y} \iff x \subseteq \gamma(\overline{y})$$



Example: Set of traces to trace of intervals abstraction

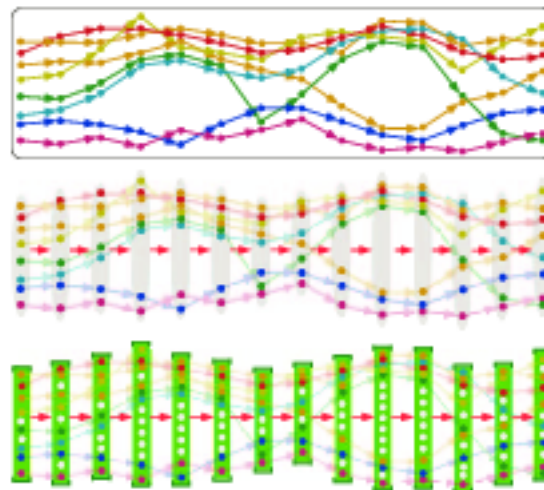
Set of traces:

$\alpha_1 \downarrow$

Trace of sets:

$\alpha_2 \downarrow$

Trace of intervals



Example: Set of traces to reachable states abstraction

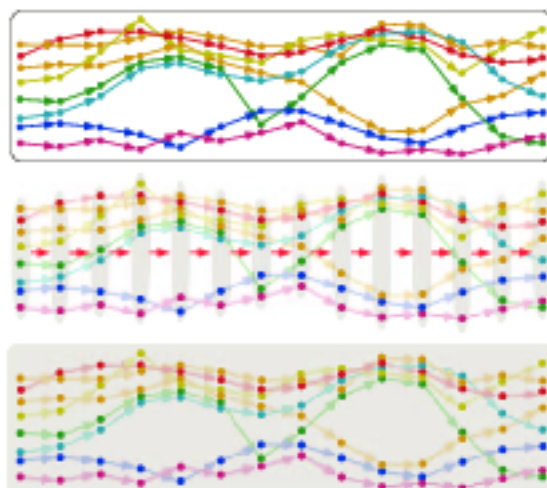
Set of traces:

$\alpha_1 \downarrow$

Trace of sets:

$\alpha_3 \downarrow$

Reachable states



Composition of Galois Connections

The composition of Galois connections:

$$\langle L, \leq \rangle \xrightarrow[\alpha_1]{\gamma_1} \langle M, \sqsubseteq \rangle$$

and:

$$\langle M, \sqsubseteq \rangle \xrightarrow[\alpha_2]{\gamma_2} \langle N, \preceq \rangle$$

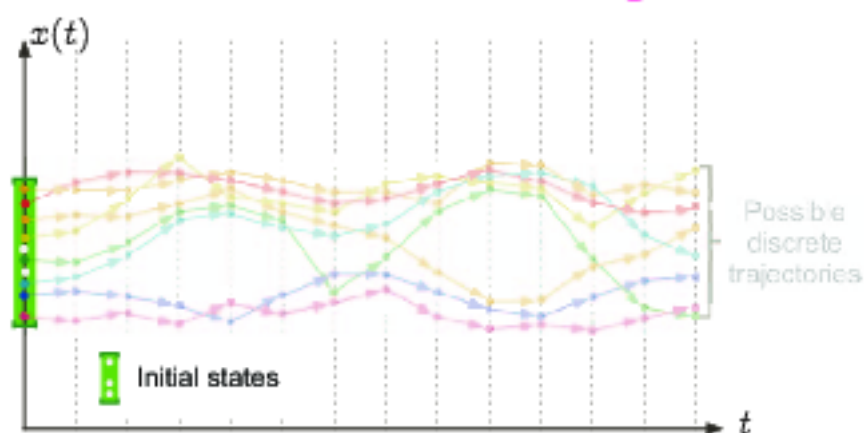
is a Galois connection:

$$\langle L, \leq \rangle \xrightarrow[\alpha_2 \circ \alpha_1]{\gamma_1 \circ \gamma_2} \langle N, \preceq \rangle$$

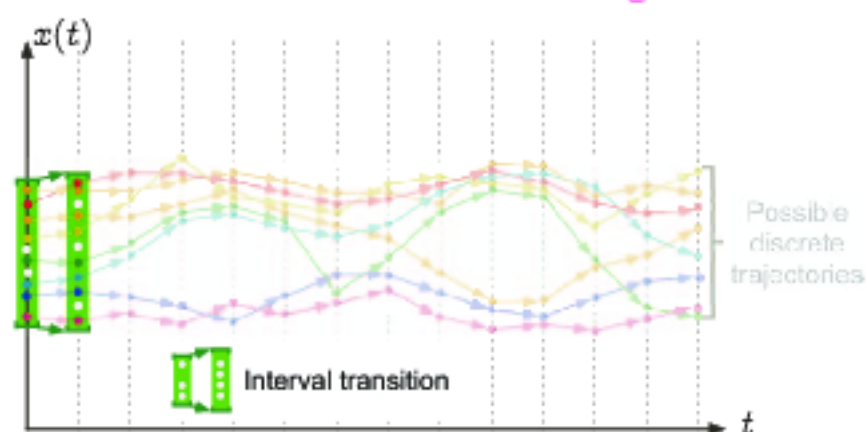


Convergence acceleration by widening/narrowing

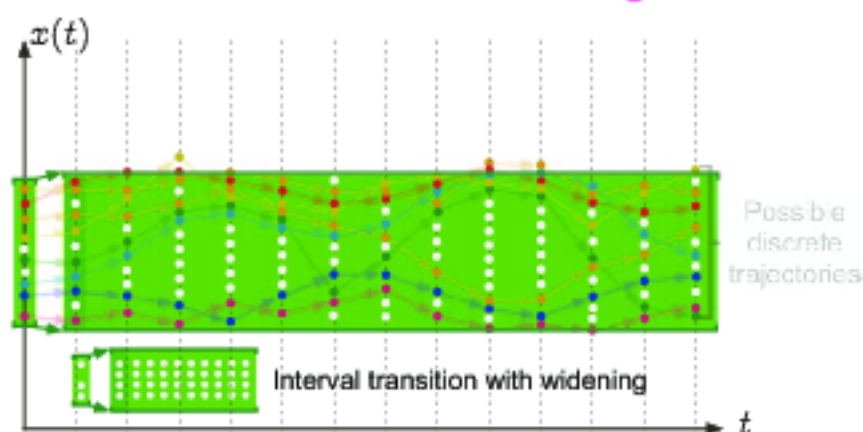
Graphic example: upward iteration
with widening



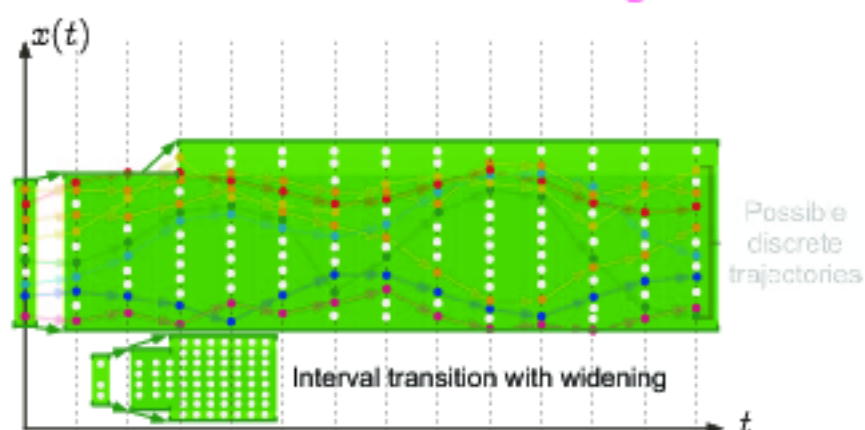
Graphic example: upward iteration with widening



Graphic example: upward iteration with widening



Graphic example: upward iteration
with widening



Graphic example: stability of the
upward iteration



Interval widening

- $\overline{\mathbb{L}} = \{\perp\} \cup \{[\ell, u] \mid \ell, u \in \mathbb{Z} \cup \{-\infty\} \wedge u \in \mathbb{Z} \cup \{\infty\} \wedge \ell \leq u\}$
- The **widening** extrapolates unstable bounds to infinity:

$$\perp \nabla X = X$$

$$X \nabla \perp = X$$

$$[\ell_0, u_0] \nabla [\ell_1, u_1] = \begin{cases} -\infty & \text{if } \ell_1 < \ell_0 \\ \ell_0 & \text{if } \ell_1 \geq \ell_0 \end{cases} \quad \begin{cases} +\infty & \text{if } u_1 < u_0 \\ u_0 & \text{if } u_1 \geq u_0 \end{cases}$$

Not monotone. For example $[0, 1] \sqsubseteq [0, 2]$ but $[0, 1] \nabla [0, 2] = [0, +\infty] \not\sqsubseteq [0, 2] = [0, 2] \nabla [0, 2]$

Example: Interval analysis (1975)

Program to be analyzed:

```
x := 1;
1:
  while x < 10000 do
2:
    x := x + 1
3:
  od;
4:
```


Example: Interval analysis (1975)

Equations (abstract interpretation of the semantics):

```
x := 1;
1:   while x < 10000 do
2:       x := x + 1
3:   od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$


Example: Interval analysis (1975)

Resolution by chaotic increasing iteration:

```
x := 1;
1:   while x < 10000 do
2:       x := x + 1
3:   od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = 0 \\ X_2 = 0 \\ X_3 = 0 \\ X_4 = 0 \end{cases}$$


Example: Interval analysis (1975)

Increasing chaotic iteration:

```
x := 1;
1: while x < 10000 do
2:   x := x + 1
3:   od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = \emptyset \\ X_3 = \emptyset \\ X_4 = \emptyset \end{cases}$$


Example: Interval analysis (1975)

Increasing chaotic iteration:

```
x := 1;
1: while x < 10000 do
2:   x := x + 1
3:   od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, 1] \\ X_3 = \emptyset \\ X_4 = \emptyset \end{cases}$$


Example: Interval analysis (1975)

Increasing chaotic iteration:

```
x := 1;
1: while x < 10000 do
2:   x := x + 1
3: od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, 1] \\ X_3 = [2, 2] \\ X_4 = \emptyset \end{cases}$$

Example: Interval analysis (1975)

Increasing chaotic iteration:

```
x := 1;
1: while x < 10000 do
2:   x := x + 1
3: od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, 2] \\ X_3 = [2, 2] \\ X_4 = \emptyset \end{cases}$$

Example: Interval analysis (1975)

Increasing chaotic iteration: **convergence I**

```

x := 1;
1: while x < 10000 do
2:   x := x + 1
3: od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, 2] \\ X_3 = [2, 3] \\ X_4 = \emptyset \end{cases}$$

Example: Interval analysis (1975)

Increasing chaotic iteration: **convergence II**

```

x := 1;
1: while x < 10000 do
2:   x := x + 1
3: od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, 3] \\ X_3 = [2, 3] \\ X_4 = \emptyset \end{cases}$$

Example: Interval analysis (1975)

Increasing chaotic iteration: **convergence III**

```
x := 1;
1: while x < 10000 do
2:   x := x + 1
3: od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, 3] \\ X_3 = [2, 4] \\ X_4 = \emptyset \end{cases}$$


Example: Interval analysis (1975)

Increasing chaotic iteration: **convergence IIII**

```
x := 1;
1: while x < 10000 do
2:   x := x + 1
3: od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, 4] \\ X_3 = [2, 4] \\ X_4 = \emptyset \end{cases}$$


Example: Interval analysis (1975)

Increasing chaotic iteration: **convergence !!!!!**

```

x := 1;
1: while x < 10000 do
2:   x := x + 1
3: od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, 4] \\ X_3 = [2, 5] \\ X_4 = \emptyset \end{cases}$$

Example: Interval analysis (1975)

Increasing chaotic iteration: **convergence !!!!!**

```

x := 1;
1: while x < 10000 do
2:   x := x + 1
3: od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, 5] \\ X_3 = [2, 5] \\ X_4 = \emptyset \end{cases}$$

Example: Interval analysis (1975)

Increasing chaotic iteration: **convergence** |||||

```
x := 1;
1: while x < 10000 do
2:   x := x + 1
3: od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, 5] \\ X_3 = [2, 6] \\ X_4 = \emptyset \end{cases}$$

Example: Interval analysis (1975)

Convergence speed-up by widening:

```
x := 1;
1: while x < 10000 do
2:   x := x + 1
3: od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, +\infty] \leftarrow \text{widening} \\ X_3 = [2, 6] \\ X_4 = \emptyset \end{cases}$$

Example: Interval analysis (1975)

Decreasing chaotic iteration:

```

x := 1;
1: while x < 10000 do
2:   x := x + 1
3: od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, +\infty] \\ X_3 = [2, +\infty] \\ X_4 = \emptyset \end{cases}$$

Example: Interval analysis (1975)

Decreasing chaotic iteration:

```

x := 1;
1: while x < 10000 do
2:   x := x + 1
3: od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, 9999] \\ X_3 = [2, +\infty] \\ X_4 = \emptyset \end{cases}$$

Example: Interval analysis (1975)

Decreasing chaotic iteration:

```
x := 1;
1: while x < 10000 do
2:   x := x + 1
3: od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, 9999] \\ X_3 = [2, +10000] \\ X_4 = \emptyset \end{cases}$$


Example: Interval analysis (1975)

Final solution:

```
x := 1;
1: while x < 10000 do
2:   x := x + 1
3: od;
4:
```

$$\begin{cases} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{cases}$$
$$\begin{cases} X_1 = [1, 1] \\ X_2 = [1, 9999] \\ X_3 = [2, +10000] \\ X_4 = [+10000, +10000] \end{cases}$$


Example: Interval analysis (1975)

Result of the interval analysis:

$$\begin{array}{l} x := 1; \\ 1: \{x = 1\} \\ \quad \text{while } x < 10000 \text{ do} \\ 2: \{x \in [1, 9999]\} \\ \quad \quad x := x + 1 \\ 3: \{x \in [2, +10000]\} \\ \quad \quad \text{od;} \\ 4: \{x = 10000\} \end{array} \quad \left\{ \begin{array}{l} X_1 = [1, 1] \\ X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\ X_3 = X_2 \oplus [1, 1] \\ X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \end{array} \right.$$



Example: Interval analysis (1975)

Checking absence of runtime errors with interval analysis:

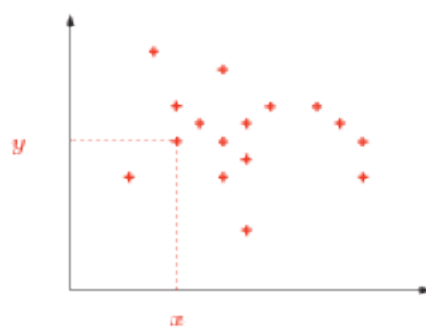
```
x := 1;
1: {x = 1}
   while x < 10000 do
2: {x ∈ [1, 9999]}
   x := x + 1      ← no overflow
3: {x ∈ [2, +10000]}
   od;
4: {x = 10000}
```



Refinement of abstractions



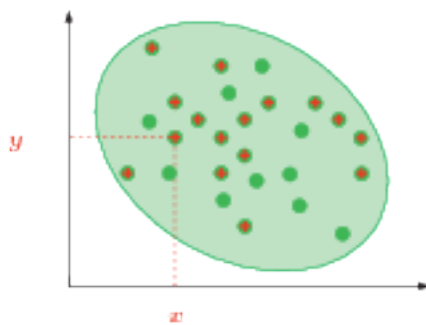
Approximations of an [in]finite set of points:



$\{\dots, \langle 19, 77 \rangle, \dots,$
 $\langle 20, 03 \rangle, \dots\}$



Approximations of an [in]finite set of points:
from above



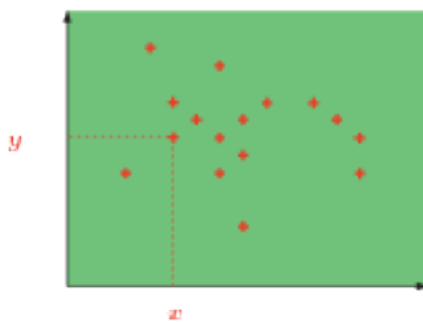
$\{\dots, \langle 19, 77 \rangle, \dots,$
 $\langle 20, 03 \rangle, \langle ?, ? \rangle, \dots\}$

From Below: dual³ + combinations.

³ Trivial for finite states (liveness model-checking), more difficult for infinite states (variant functions).



Effective computable approximations of an
[in]finite set of points; Signs⁴

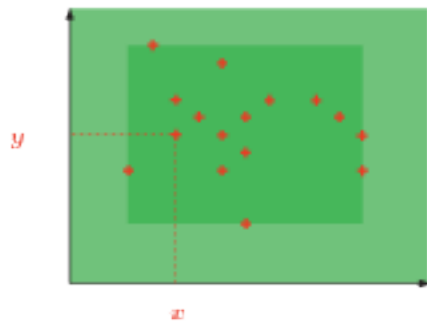


$\begin{cases} x \geq 0 \\ y \geq 0 \end{cases}$

⁴ P. Cousot & R. Cousot. Systematic design of program analysis frameworks. ACM POPL'79, pp. 269-282, 1979.



Effective computable approximations of an [in]finite set of points; Intervals⁵

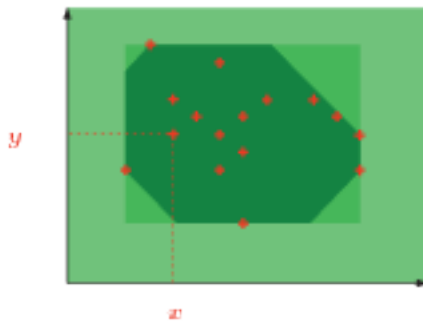


$$\begin{cases} x \in [19, 77] \\ y \in [20, 03] \end{cases}$$

⁵ P. Cousot & R. Cousot. *Static determination of dynamic properties of programs*. Proc. 2nd Int. Symp. on Programming, Dunod, 1976.



Effective computable approximations of an [in]finite set of points; Octagons⁶

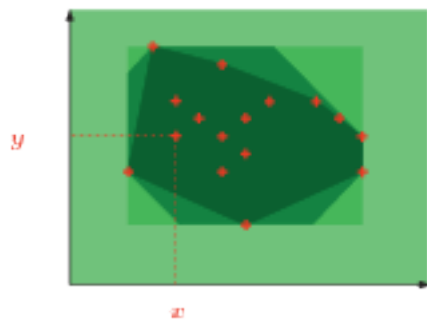


$$\begin{cases} 1 \leq x \leq 9 \\ x + y \leq 77 \\ 1 \leq y \leq 9 \\ x - y \leq 99 \end{cases}$$

⁶ A. Miné. *A New Numerical Abstract Domain Based on Difference-Bound Matrices*. PADO'2001. LNCS 2083, pp. 158–172. Springer 2001. See the *The Octagon Abstract Domain Library* on <http://www.di.sns.fr/~mine/oct/>



Effective computable approximations of an [in]finite set of points; Polyhedra⁷

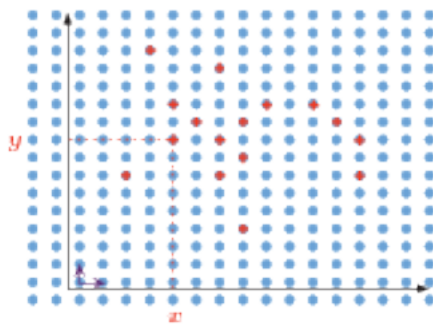


$$\begin{cases} 19x + 77y \leq 2004 \\ 20x + 03y \geq 0 \end{cases}$$

⁷ P. Cousot & H. Halbwachs. *Automatic discovery of linear restraints among variables of a program*. ACM POPL-4, 1978, pp. 84–97.



Effective computable approximations of an [in]finite set of points; Simple congruences⁸

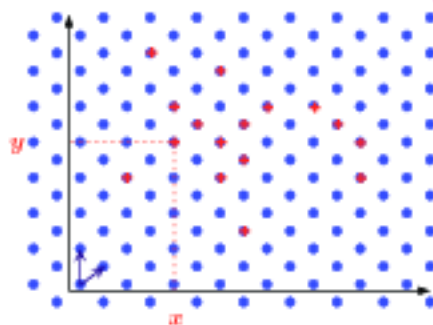


$$\begin{cases} x = 19 \bmod 77 \\ y = 20 \bmod 99 \end{cases}$$

⁸ Ph. Gauger. *Static Analysis of Arithmetical Congruences*. Int. J. Comput. Math. 30, 1989, pp. 165–190.



Effective computable approximations of an
[in]finite set of points; Linear
congruences⁹

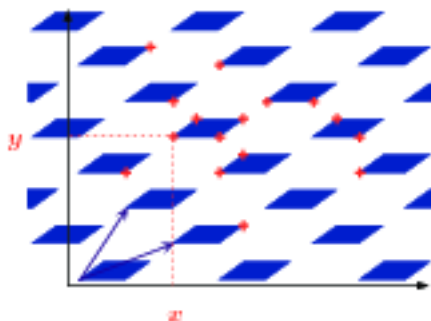


$$\begin{cases} 1x + 9y = 7 \bmod 8 \\ 2x - 1y = 9 \bmod 9 \end{cases}$$

⁹ Ph. Changer. *Static Analysis of Linear Congruence Equations among Variables of a Program*. TAPSOFT'91, pp. 184-192. LNCS 493, Springer, 1991.



Effective computable approximations of an
[in]finite set of points; Trapezoidal lin-
ear congruences¹⁰



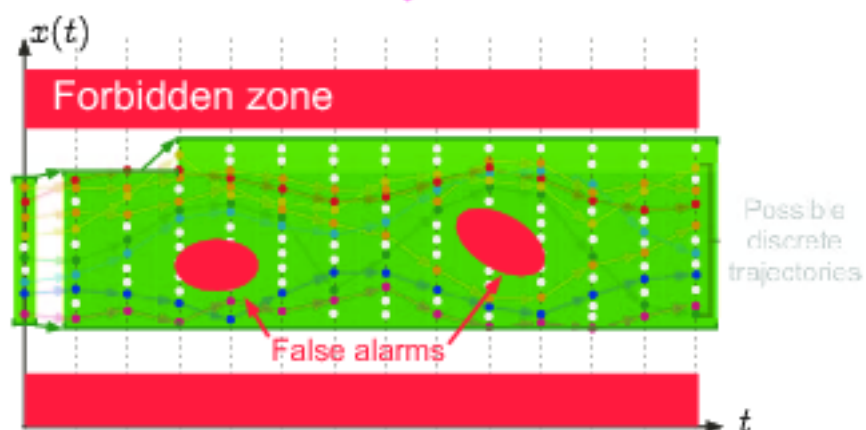
$$\begin{cases} 1x + 9y \in [0, 77] \bmod 10 \\ 2x - 1y \in [0, 99] \bmod 11 \end{cases}$$

¹⁰ F. Masmoudy. *Array Operations Abstraction Using Semantic Analysis of Trapezoid Congruences*. ACM ICS'02.

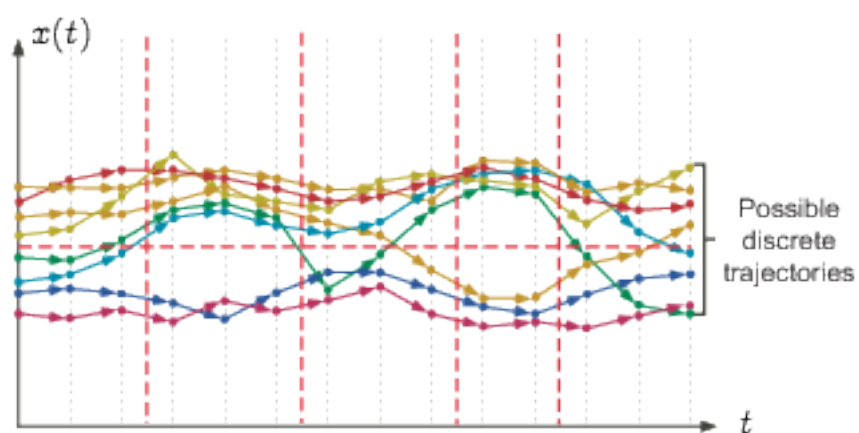


Refinement of iterates

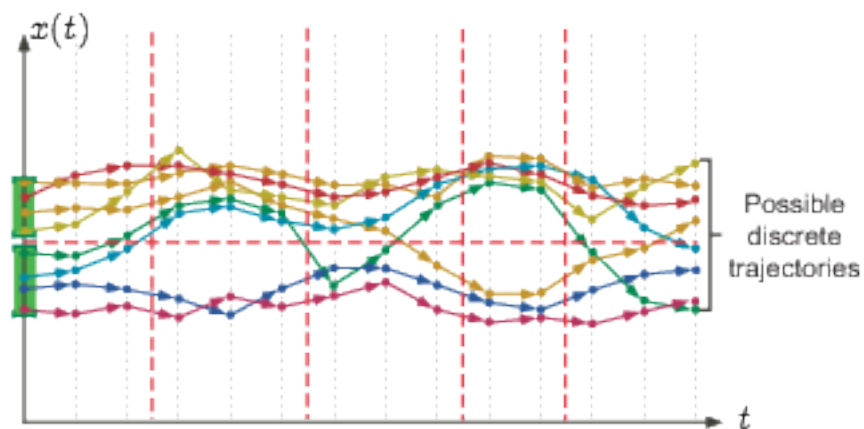
Graphic example: Refinement required
by false alarms



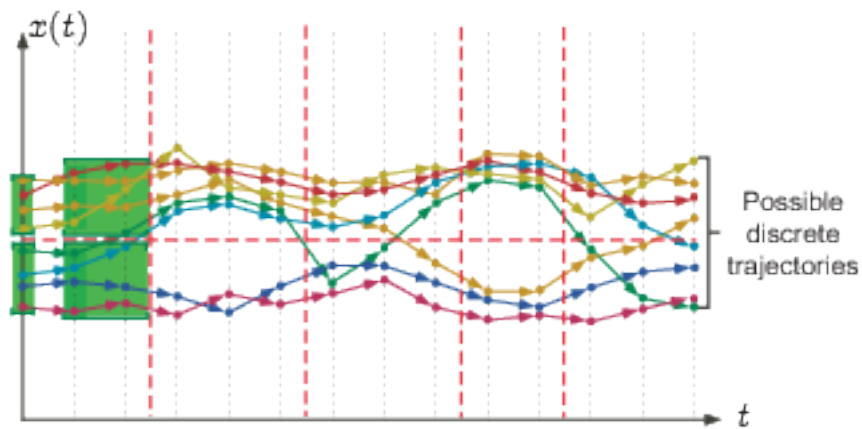
Graphic example: Partitionning



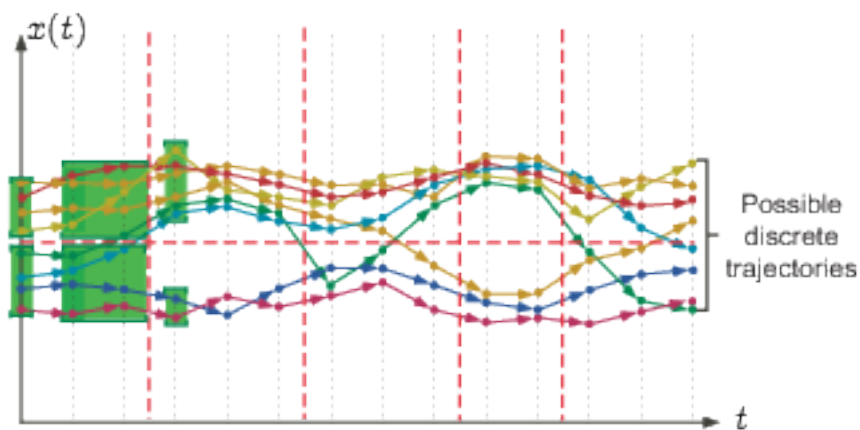
Graphic example: partitionned upward iteration with widening



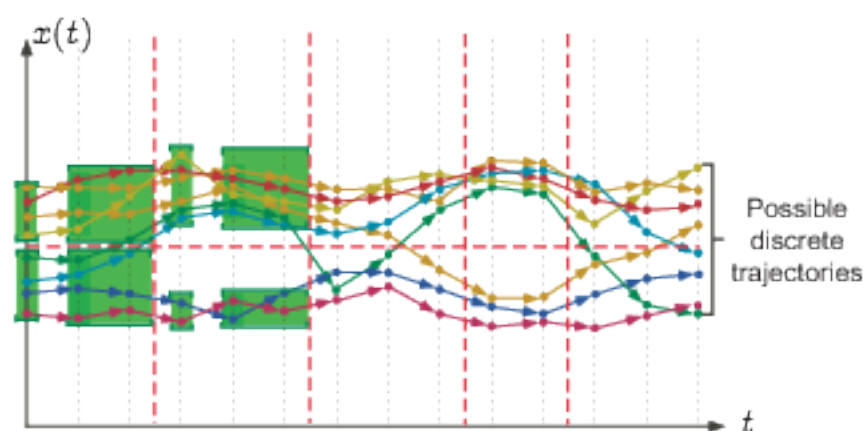
Graphic example: partitionned upward iteration with widening



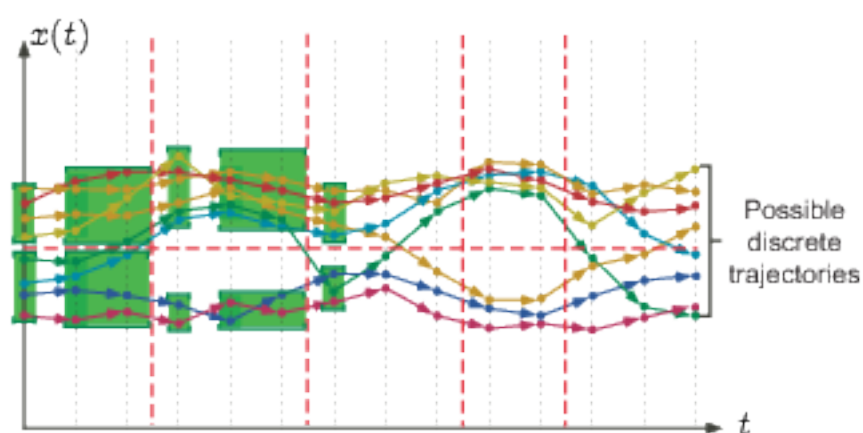
Graphic example: partitionned upward iteration with widening



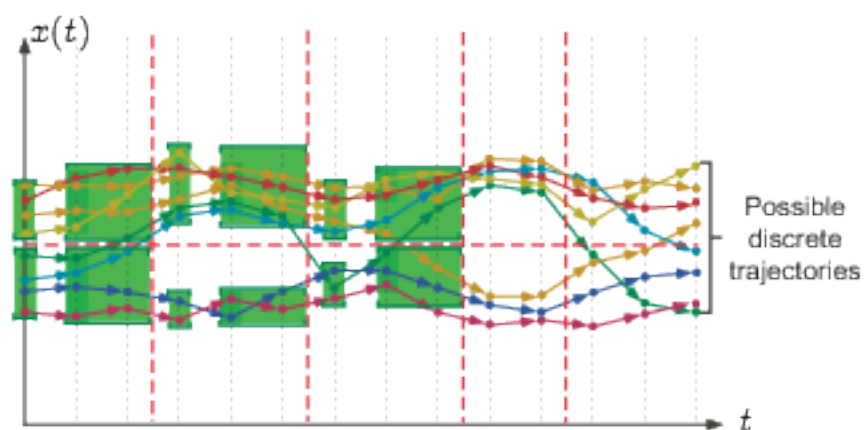
Graphic example: partitionned upward iteration with widening



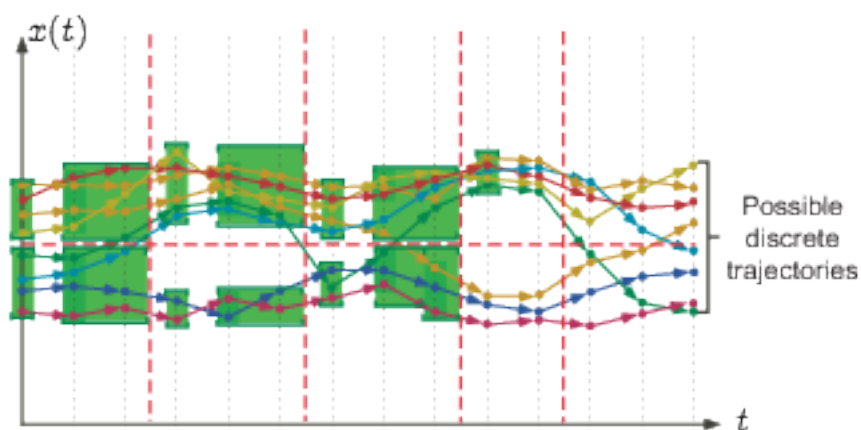
Graphic example: partitionned upward iteration with widening



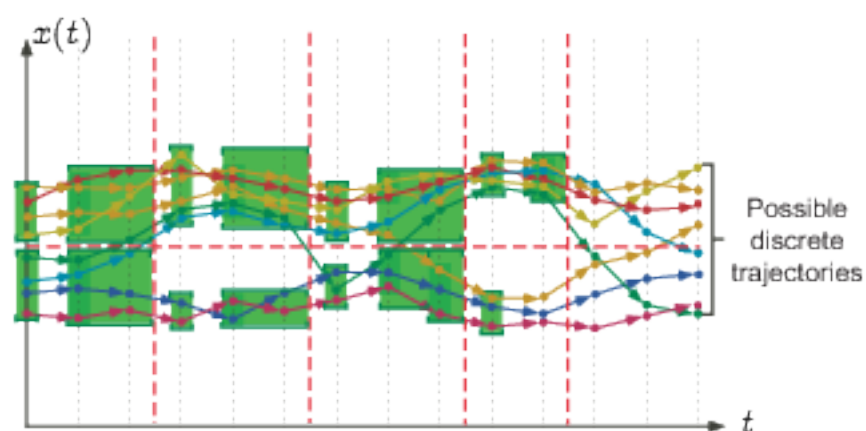
Graphic example: partitionned upward iteration with widening



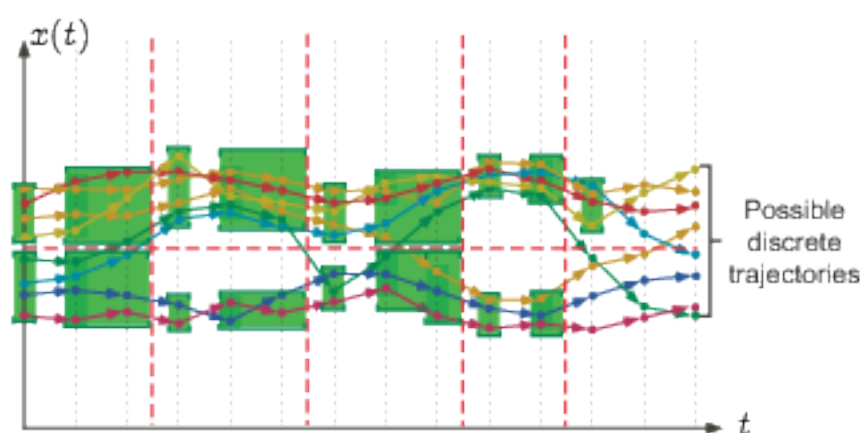
Graphic example: partitionned upward iteration with widening



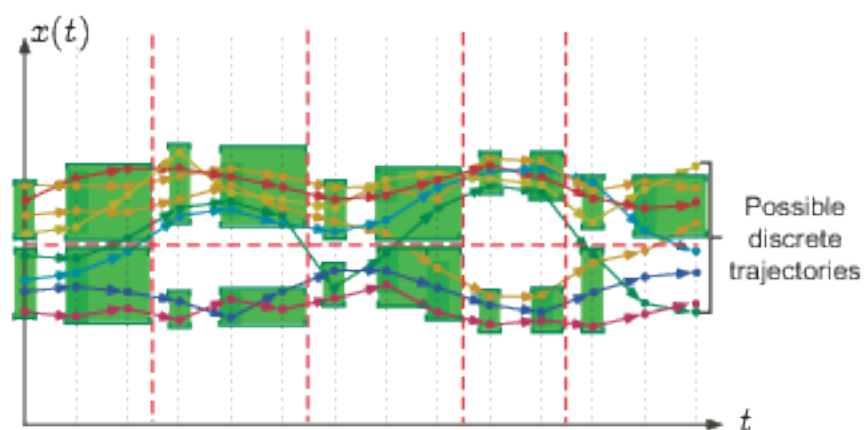
Graphic example: partitionned upward iteration with widening



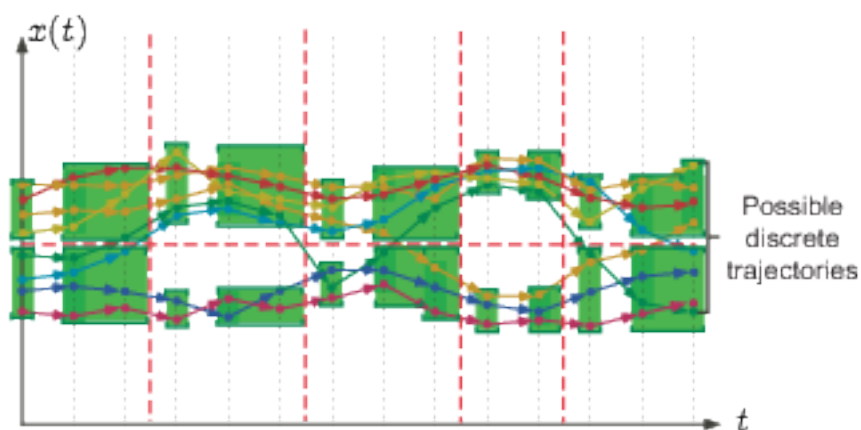
Graphic example: partitionned upward iteration with widening



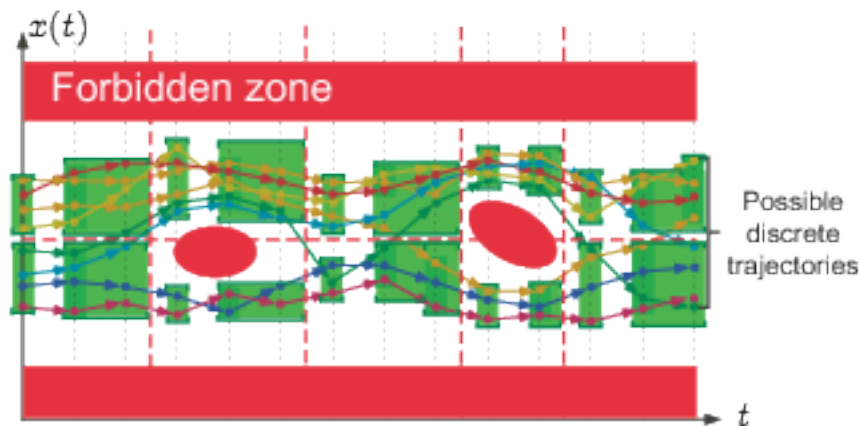
Graphic example: partitionned upward iteration with widening



Graphic example: partitionned upward iteration with widening



Graphic example: safety verification



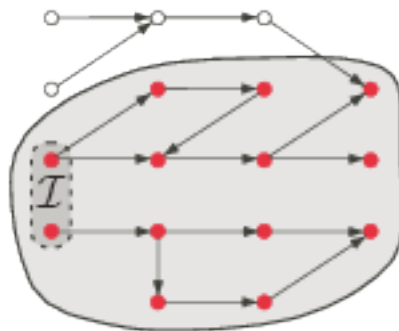
Interval widening with threshold set

- The **threshold set** T is a finite set of numbers (plus $+\infty$ and $-\infty$),
- $[a, b] \nabla_T [a', b'] = [\text{if } a' < a \text{ then } \max\{\ell \in T \mid \ell \leq a'\} \text{ else } a, \\ \text{if } b' > b \text{ then } \min\{h \in T \mid h \geq b'\} \text{ else } b]$.
- Examples (intervals):
 - sign analysis: $T = \{-\infty, 0, +\infty\}$;
 - strict sign analysis: $T = \{-\infty, -1, 0, +1, +\infty\}$;
- T is a **parameter** of the analysis.

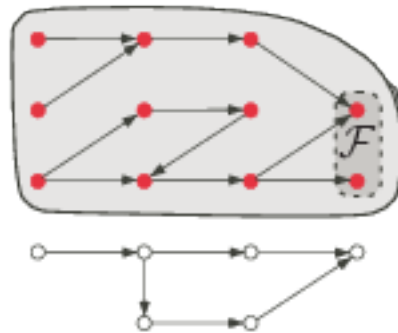
Combinations of abstractions



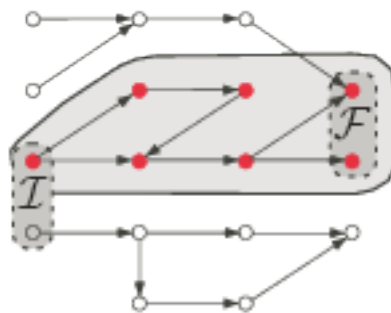
Forward/reachability analysis



Backward/ancestry analysis



Iterated forward/backward analysis



Example of iterated forward/backward analysis

Arithmetical mean of two integers x and y :

```
{x>=y}
while (x <> y) do
  {x>=y+2}
  x := x - 1;
  {x>=y+1}
  y := y + 1
  {x>=y}
od
{x=y}
```

Necessarily $x \geq y$ for proper termination



Example of iterated forward/backward analysis

Adding an auxiliary counter k decremented in the loop body and asserted to be null on loop exit:

```
{x=y+2k, x>=y}
while (x <> y) do
  {x=y+2k, x>=y+2}
  k := k - 1;
  {x=y+2k+2, x>=y+2}
  x := x - 1;
  {x=y+2k+1, x>=y+1}
  y := y + 1
  {x=y+2k, x>=y}
od
{x=y, k=0}
assume {k = 0}
{x=y, k=0}
```

Moreover the difference of x and y must be even for proper termination



Applications of abstract interpretation



Theoretical applications of abstract interpretation

- **Static Program Analysis** [POPL '77,78,79] including **Data-flow Analysis** [POPL '79,00], **Set-based Analysis** [FPCA '95], etc
- **Syntax Analysis** [TCS 290(1) 2002]
- **Hierarchies of Semantics (Including Proofs)** [POPL '92, TCS 277(1–2) 2002]
- **Typing** [POPL '97]
- **Model Checking** [POPL '00]
- **Program Transformation** [POPL '02]
- **Software watermarking** [POPL '04]



Industrial applications of abstract interpretation

- **Program analysis and manipulation:** a small rate of false alarms is acceptable
 - **AiT:** worst case execution time¹¹
 - **StackAnalyser:** stack usage analysis¹¹
- **Program verification:** no false alarms is acceptable
 - **TVLA:** A system for generating abstract interpreters
 - **Astrée:** verification of absence of run-time errors¹¹

¹¹ applied to the primary flight control software of the Airbus A340/350 and A380 fly-by-wire systems

Bibliography

Seminal papers

- Patrick Cousot & Radhia Cousot. [Abstract interpretation: a unified lattice model for static analysis of programs by construction or approximation of fixpoints](#). In 4th Symp. on Principles of Programming Languages, pages 238—252. ACM Press, 1977.
- Patrick Cousot & Nicolas Halbwachs. [Automatic discovery of linear restraints among variables of a program](#). In 5th Symp. on Principles of Programming Languages, pages 84—97. ACM Press, 1978.
- Patrick Cousot & Radhia Cousot. [Systematic design of program analysis frameworks](#). In 6th Symp. on Principles of Programming Languages pages 269—282. ACM Press, 1979.



Recent surveys

- Patrick Cousot. [Interprétation abstraite](#). Technique et Science Informatique, Vol. 19, Nb 1-2-3. Janvier 2000, Hermès, Paris, France. pp. 155-164. ■
- Patrick Cousot. [Abstract Interpretation Based Formal Methods and Future Challenges](#). In Informatics, 10 Years Back — 10 Years Ahead, R. Wilhelm (Ed.), LNCS 2000, pp. 138-156, 2001.
- Patrick Cousot & Radhia Cousot. [Abstract Interpretation Based Verification of Embedded Software: Problems and Perspectives](#). In Proc. 1st Int. Workshop on Embedded Software, EMSOFT 2001, T.A. Henzinger & C.M. Kirsch (Eds.), LNCS 2211, pp. 97–113. Springer, 2001.



Course Content



Anticipated Content of Course 16.399:

Abstract Interpretation

- Today : an informal [overview of abstract interpretation](#);
- The [software verification problem](#) (undecidability, complexity, test, simulation, specification, formal methods (deductive methods, model-checking, static analysis) and their limitations, intuitive notion of approximation, false alarms);
- [Mathematical foundations](#) (naive set theory, first order classical logic, lattice theory, fixpoints);



- **Semantics of programming languages** (abstract syntax, operational semantics, inductive definitions, example of a simple imperative language, grammar and interpreter of the language, trace semantics);
- **Program specification and manual proofs** (safety properties, Hoare logic, predicate transformers, liveness properties, linear-time temporal logic (LTL));
- **Order-theoretic approximation** (abstraction, closures, Galois connections, fixpoint abstraction, widening, narrowing, reduced product, absence of best approximation, refinement);

- **Principle of static analysis by abstract interpretation** (reachability analysis of a transition system, finite approximation, model-checking, infinite approximation, static analysis, program-based versus language-based analysis, limitations of finite approximations);
- **Design of a generic structural abstract interpreter** (collecting semantics, non-relational and relational analysis, convergence acceleration by widening/narrowing);
- **Static analysis** (forward reachability analysis, backward analysis, iterated forward/backward analysis, inevitability analysis, termination)

- **Numerical abstract domains** (intervals, affine equalities, congruences, octagons, polyhedra);
- **Symbolic abstract domains** (abstraction of sequences, trees and graphs, BDDs, word and tree automata, pointer analysis);
- **Case studies** (abstractions used in ASTREE and TVLA);

Anticipated Home Work of Course 6.862: Abstract Interpretation

- A **reading assignment** of the slides for each course and of a recommended recently published research article related to that course;
- A **personal project** on the design and implementation of a **static analyser** of numerical programs (which frontend will be provided)

Assigned reading for course 1

Patrick Cousot.

**Abstract Interpretation Based Formal Methods and
Future Challenges.**

In *Informatics, 10 Years Back — 10 Years Ahead*,
R. Wilhelm (Ed.), Lecture Notes in Computer Science
2000, pp. 138–156, 2001.

Anticipated Grading of Course 16.399:

Abstract Interpretation

The course is letter graded.

- 10% Class participation
- 15% Presentation 1
- 15% Presentation 2
- 40% Personal project
- 20% Final written exam

The two **presentations of research papers** are in CS conference format (25mn of talk and 5mn of questions) to be selected by the students in the list of assigned readings; to be held outside of lecture hours — times TBA.

THE END

My MIT web site is <http://www.mit.edu/~csaver/>

The course web site is <http://web.mit.edu/6.034/www-mit.edu/course/18/18.300/www/>.



Course 6.0342 "Abstract Interpretation", Thursday, February 24, 2016

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