

« An Informal Overview of Abstract Interpretation »

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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  ← j < n0 so no upper overflow
      {n0=n,j>=1,i>=j,n0>=i}
    od;
    {n0=n,i=j,i>=1,n0>=i}
    i := i - 1  ← i > 0 so no lower overflow
    {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)

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.



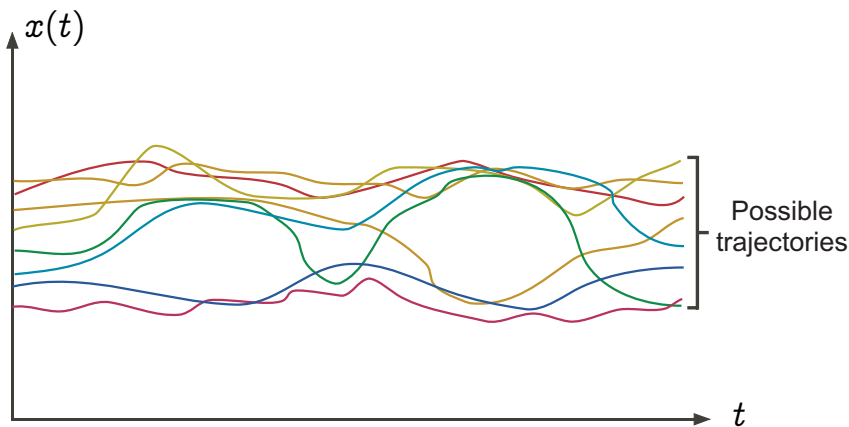
¹ See for example the Microsoft Security Bulletins MS02-065, MS04-011, etc.

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.

A very informal introduction
to the principles of
abstract interpretation

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

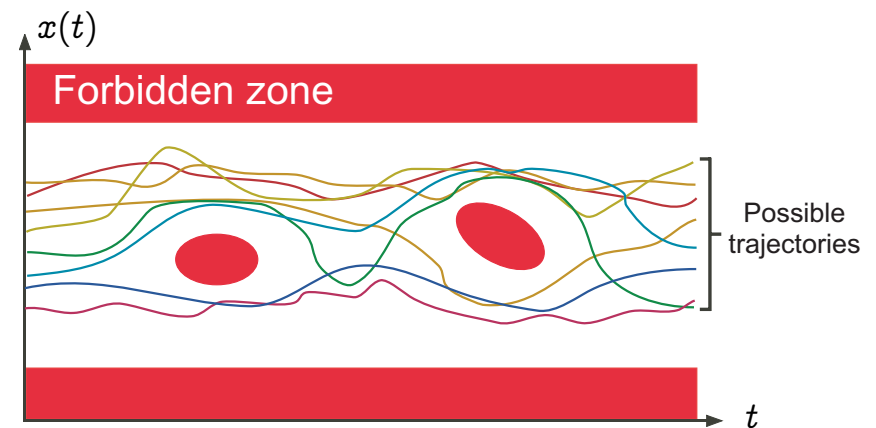
- Assume $\text{termination}(P)$ would always terminate and returns true iff P always terminates on all input data;
- The following program yields a contradiction

```
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

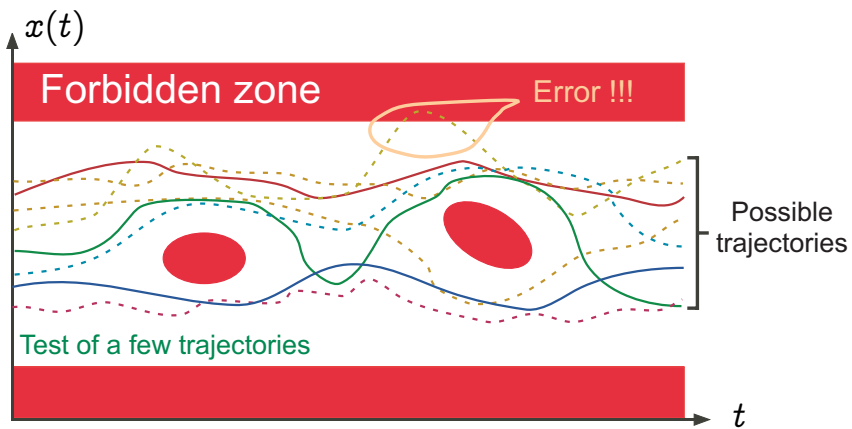
- A *safety proof* consists in proving that the intersection of the program concrete semantics and the forbidden zone is empty;
- *Undecidable* problem (the concrete semantics is not computable);
- Impossible to provide completely automatic answers with finite computer resources and neither human interaction nor uncertainty on the answer².

² e.g. probabilistic answer.

Test/debugging

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

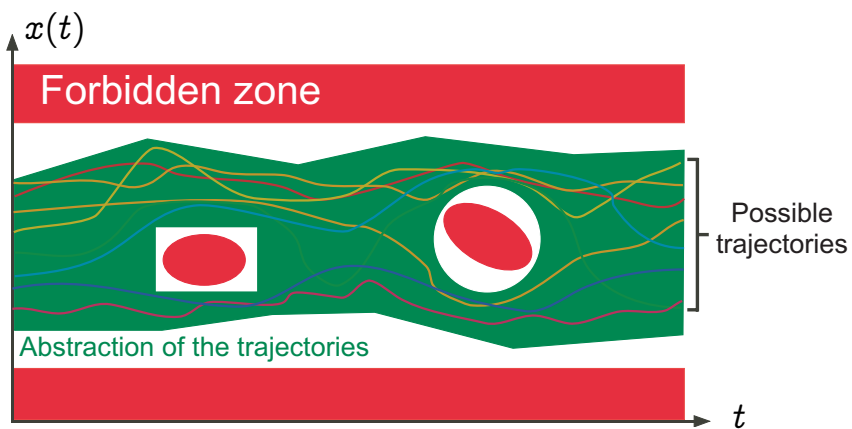
Graphic example: Property test/simulation



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:

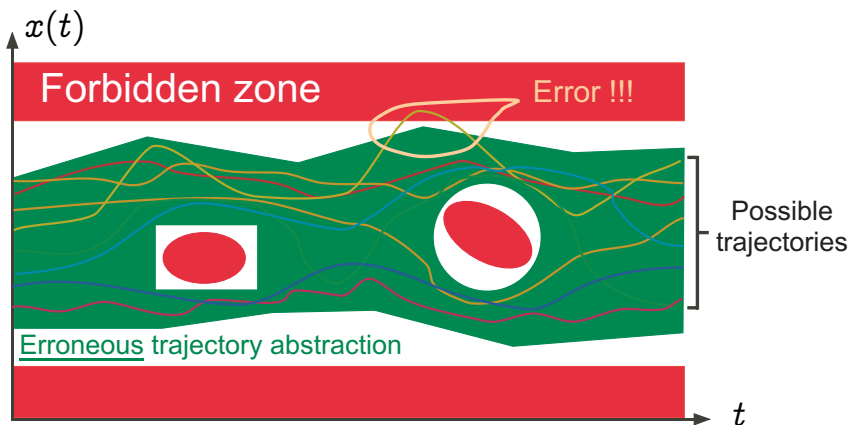
- “*model checking*”:
 - the abstract semantics is given manually by the user;
 - in the form of a finitary model of the program execution;
 - can be computed automatically, by techniques relevant to static analysis.

- “*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).

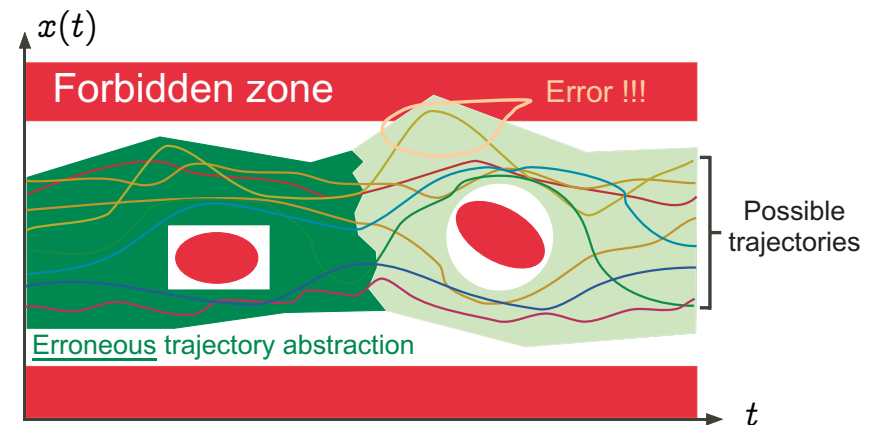
Required properties of the abstract semantics

- **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).

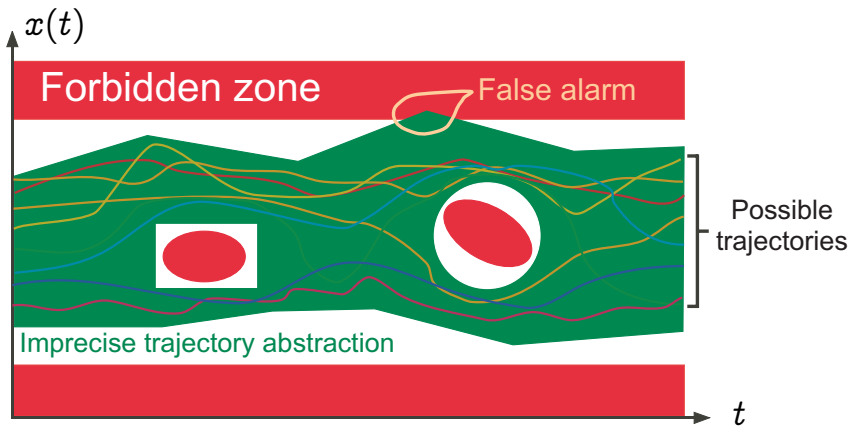
Graphic example: Erroneous abstraction — I



Graphic example: Erroneous abstraction — II



Graphic example: Imprecision \Rightarrow false alarms

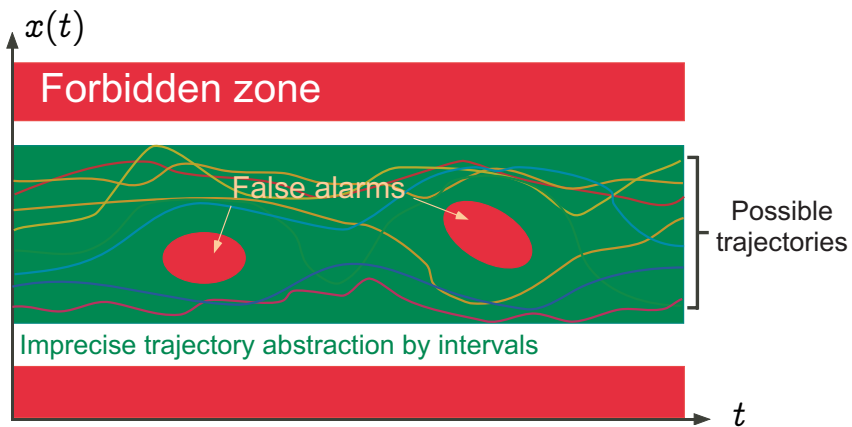


Abstract domains

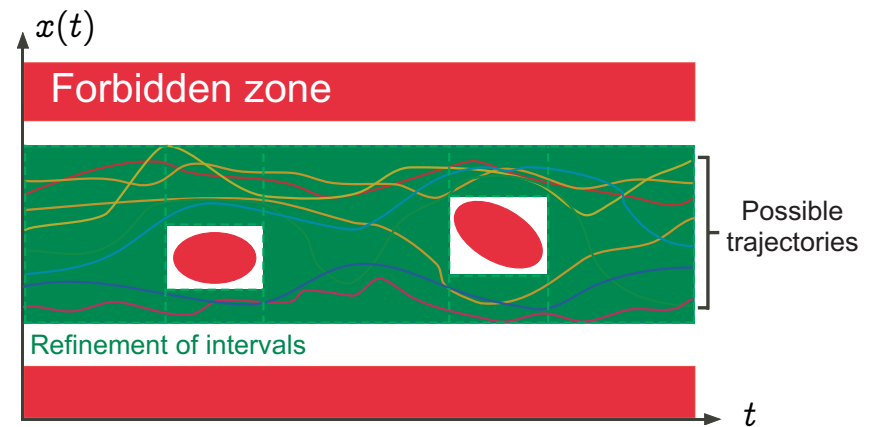
Standard abstractions

- 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



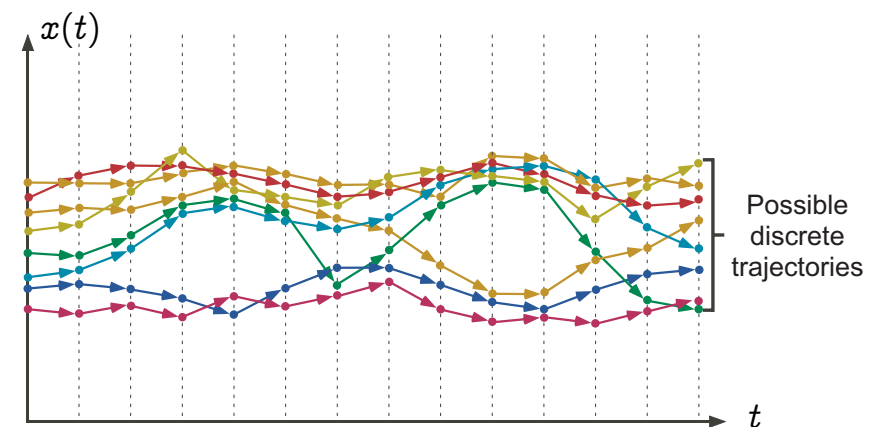
A very informal introduction to static analysis algorithms

Trace semantics

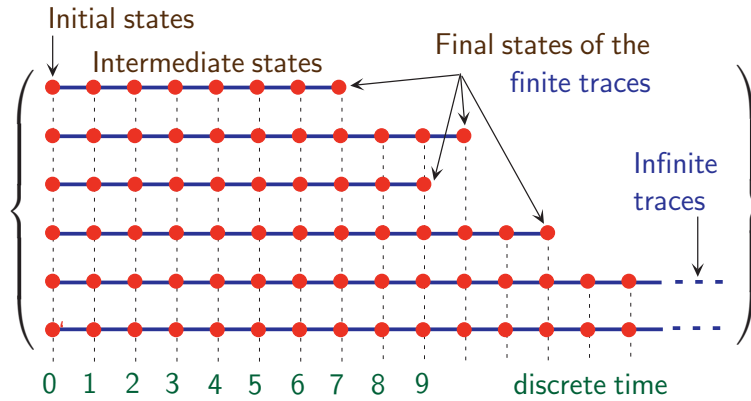
Trace semantics

- 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

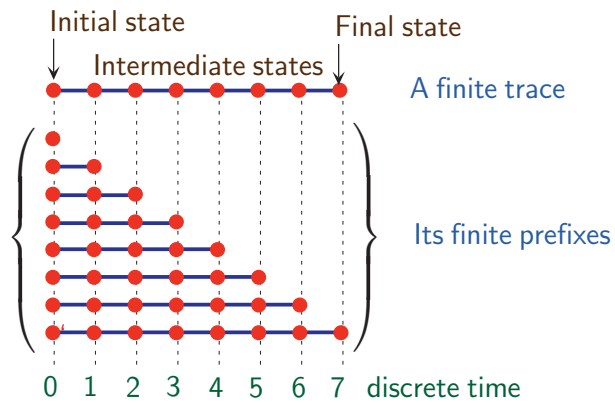


Trace semantics, intuition

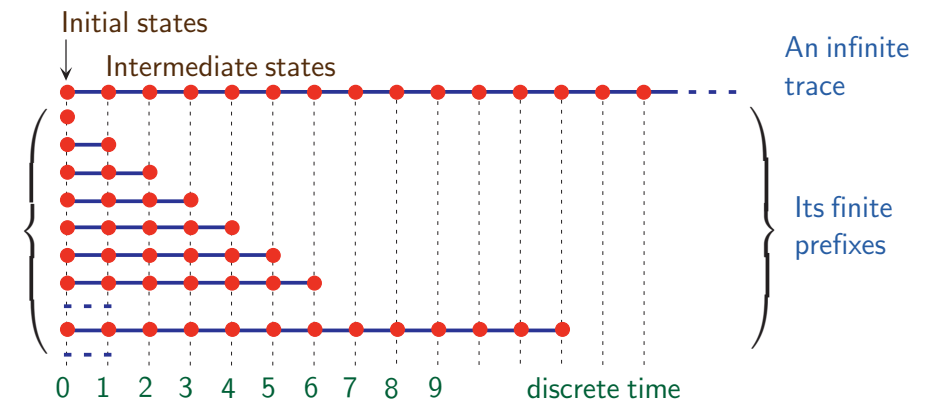


Prefix trace semantics

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

Prefixes = $\{\bullet \mid \bullet \text{ is an initial state}\}$

$\cup \{\bullet \text{---}\dots\text{---}\bullet \mid \bullet \text{---}\dots\text{---}\bullet \in \text{Prefixes}\}$

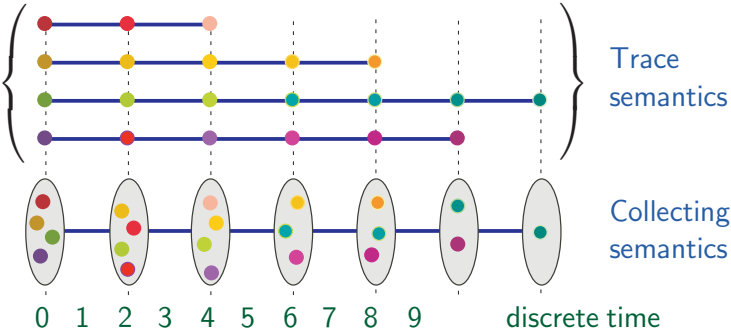
& $\bullet \text{---}\bullet \text{ is a transition step}\}$

- 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

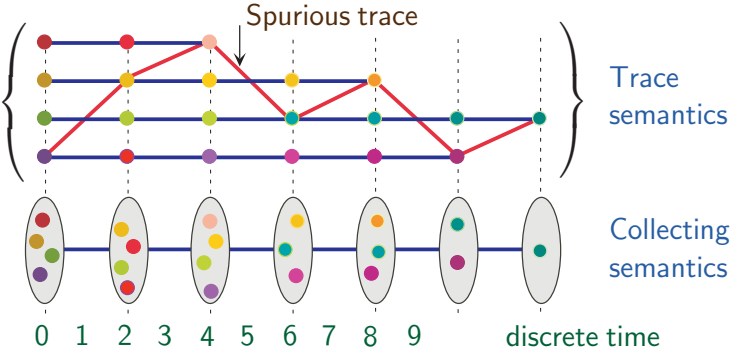
Collecting semantics

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

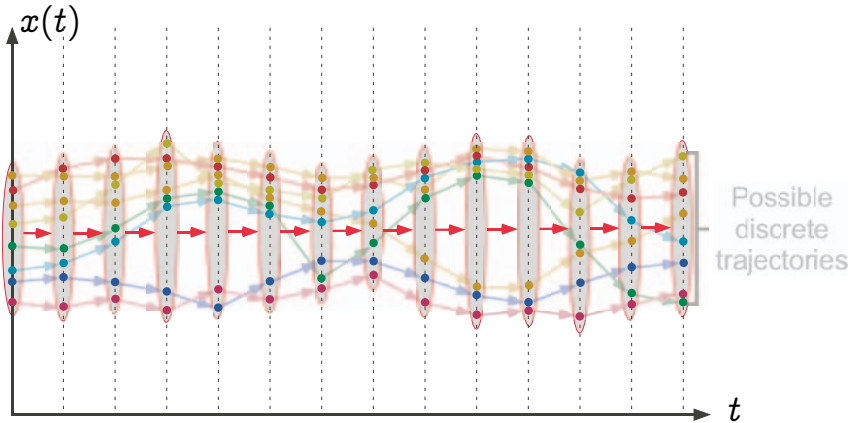


Collecting abstraction

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

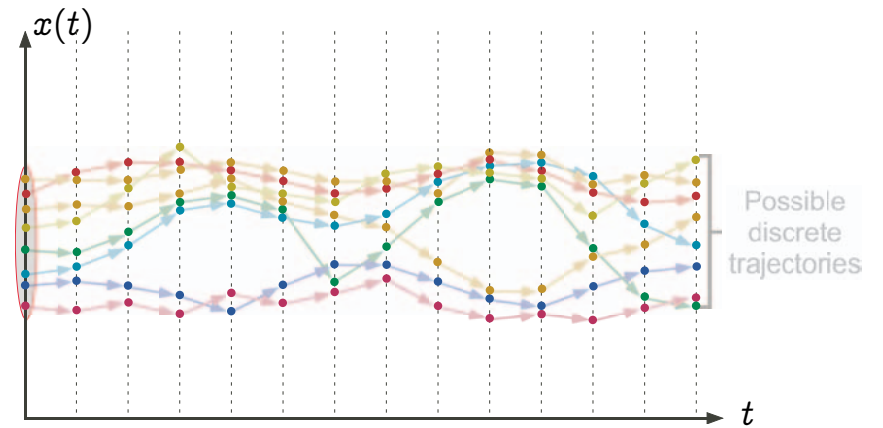


Graphic example: collecting semantics

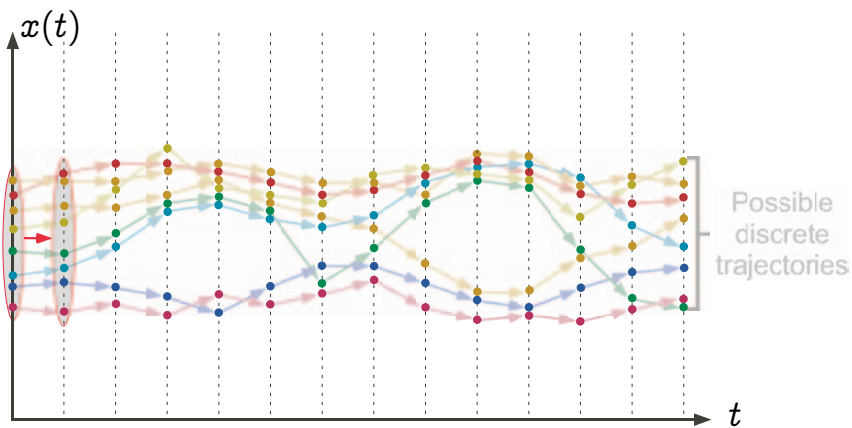


Collecting semantics
in fixpoint form

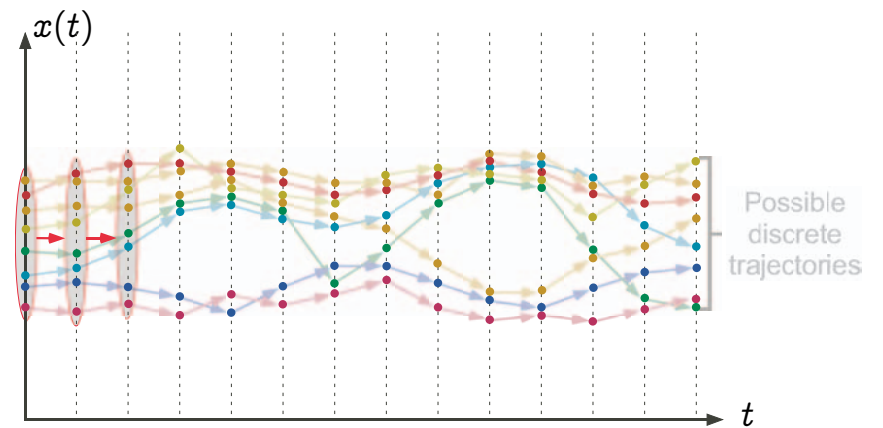
Graphic example: collecting semantics
in fixpoint form



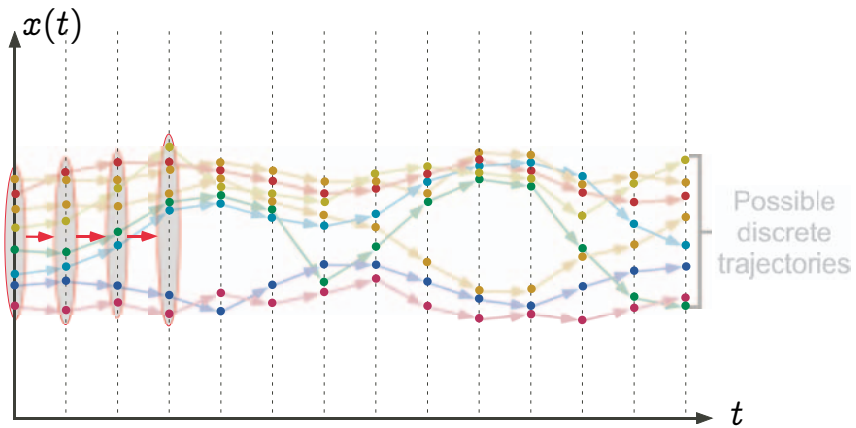
Graphic example: collecting semantics
in fixpoint form



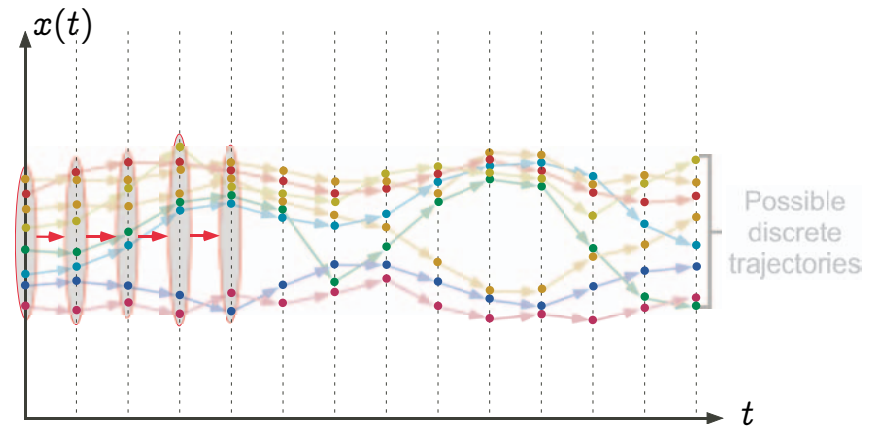
Graphic example: collecting semantics
in fixpoint form



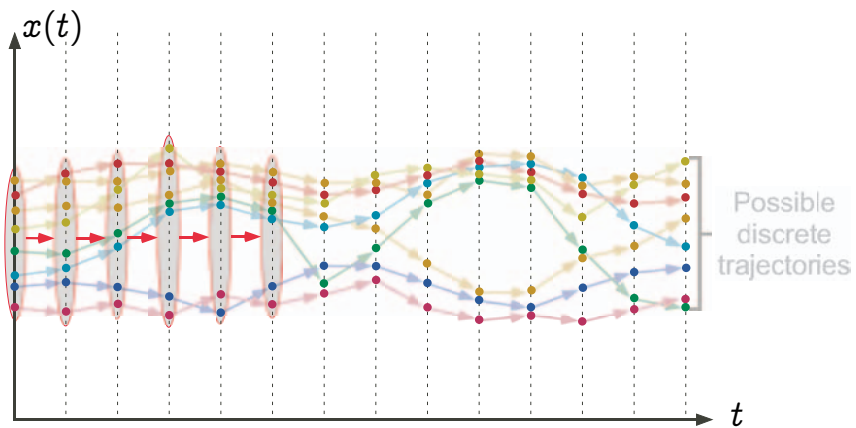
Graphic example: collecting semantics in fixpoint form



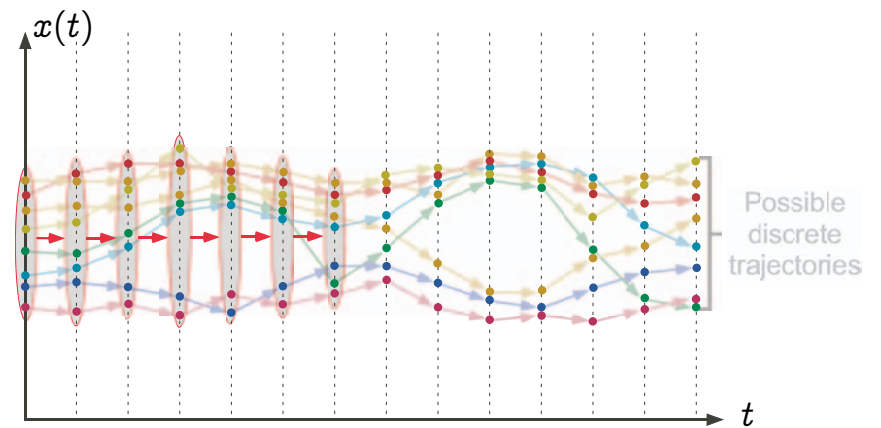
Graphic example: collecting semantics in fixpoint form



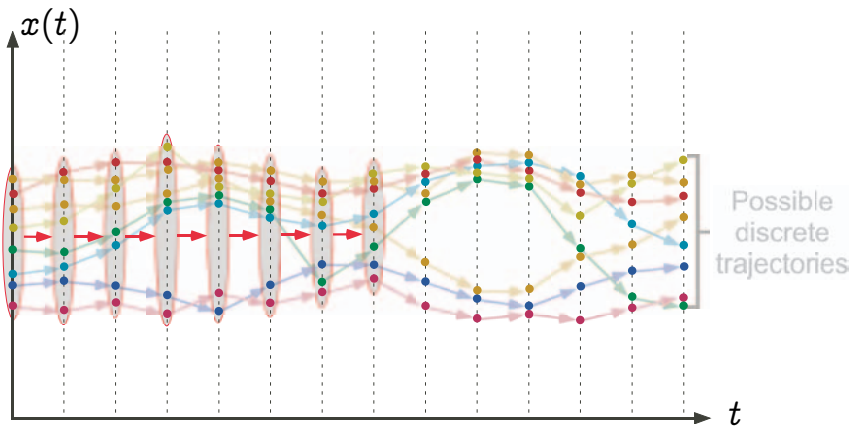
Graphic example: collecting semantics in fixpoint form



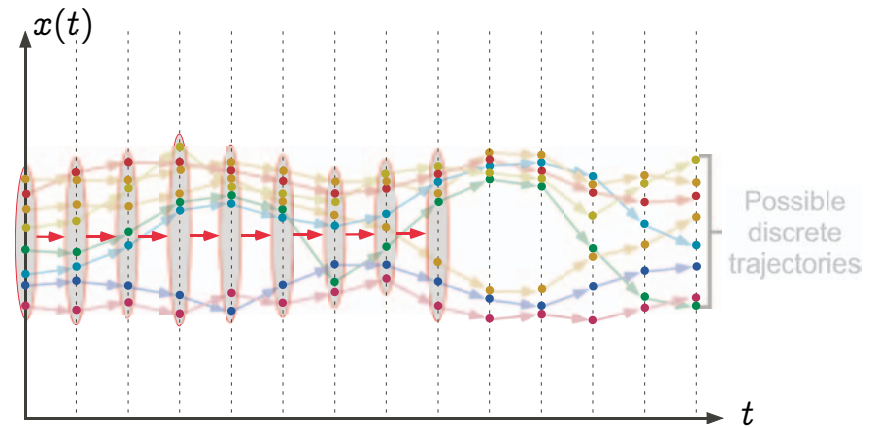
Graphic example: collecting semantics in fixpoint form



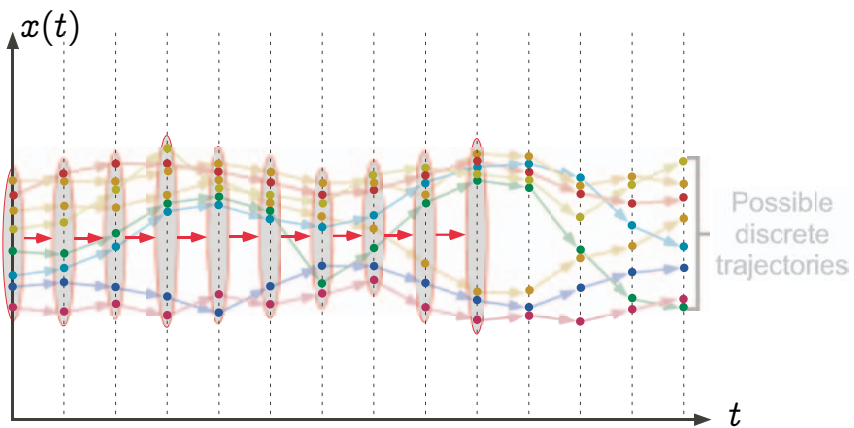
Graphic example: collecting semantics in fixpoint form



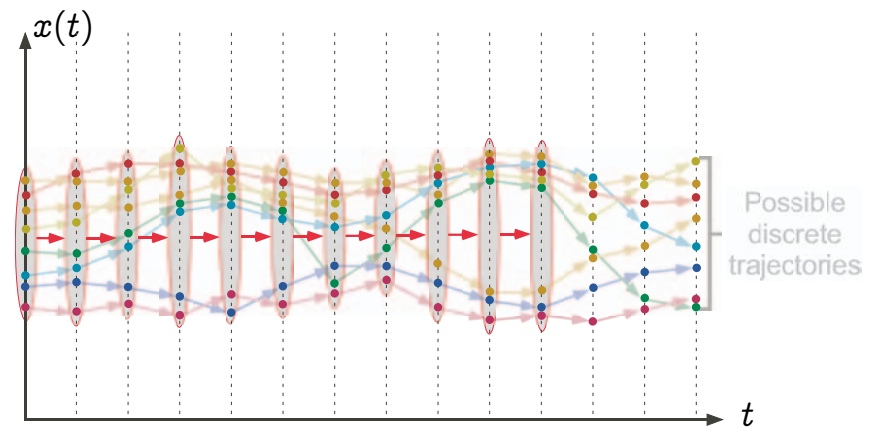
Graphic example: collecting semantics in fixpoint form



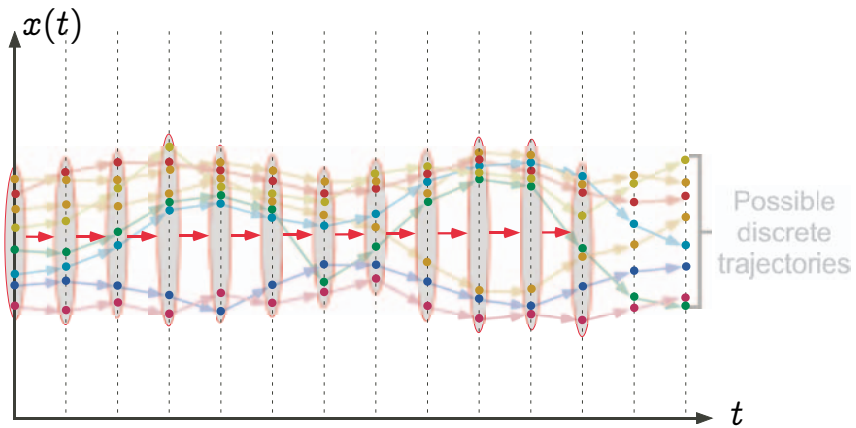
Graphic example: collecting semantics in fixpoint form



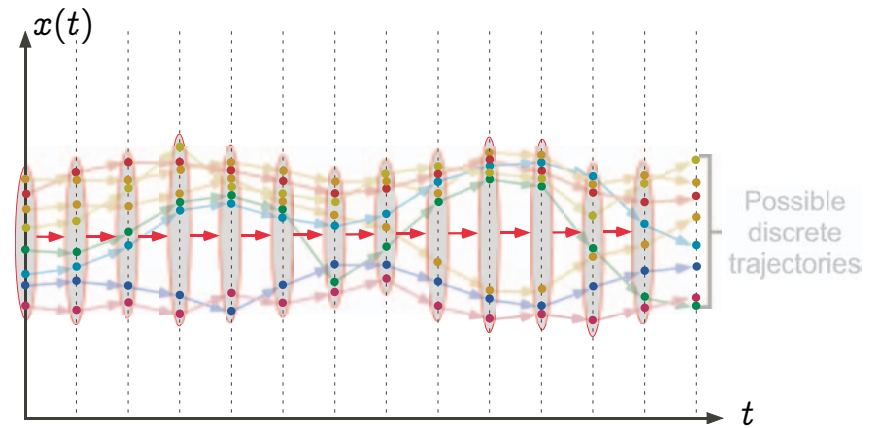
Graphic example: collecting semantics in fixpoint form



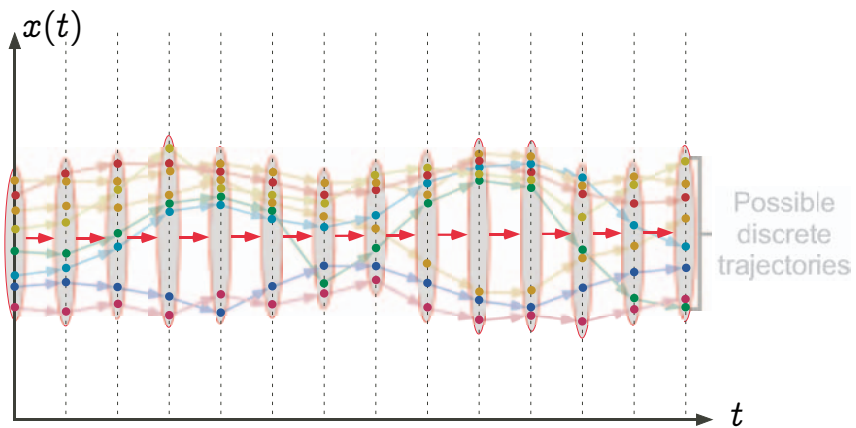
Graphic example: collecting semantics in fixpoint form



Graphic example: collecting semantics in fixpoint form

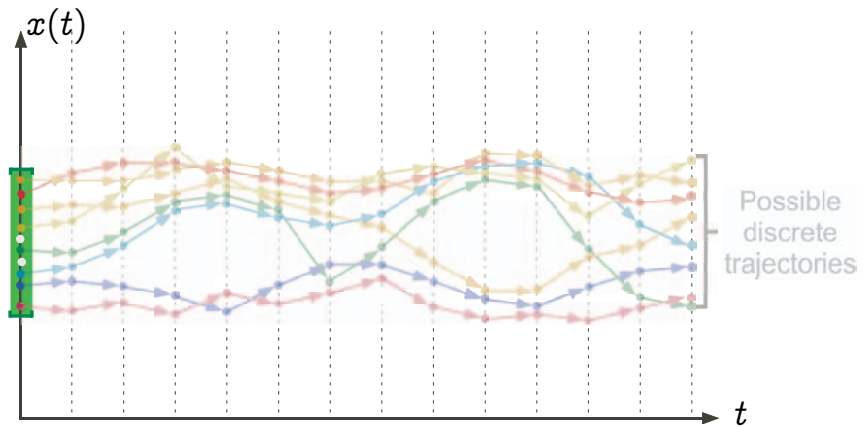


Graphic example: collecting semantics in fixpoint form

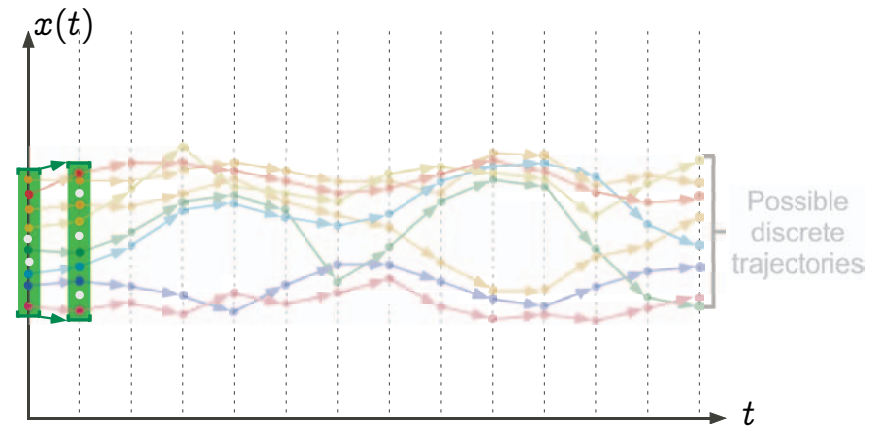


Interval Abstraction
(in iterative fixpoint form)

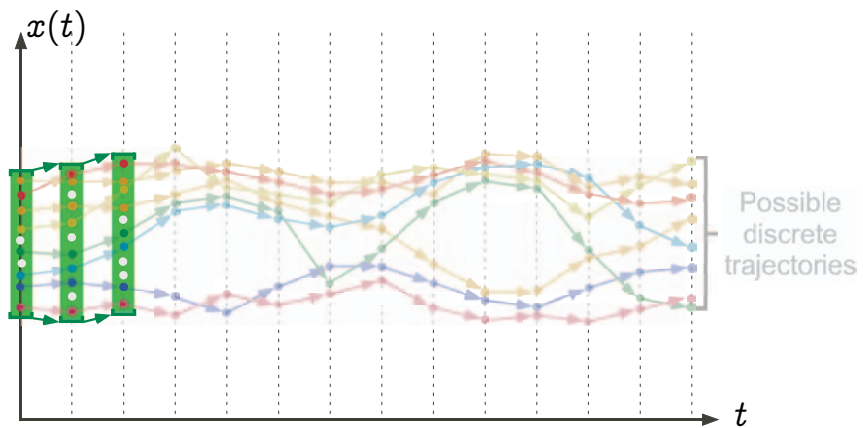
Graphic example: traces of intervals in fixpoint form



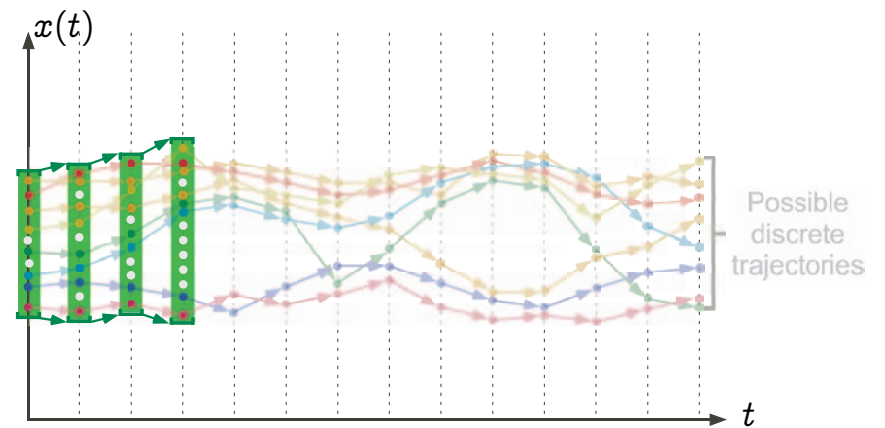
Graphic example: traces of intervals in fixpoint form



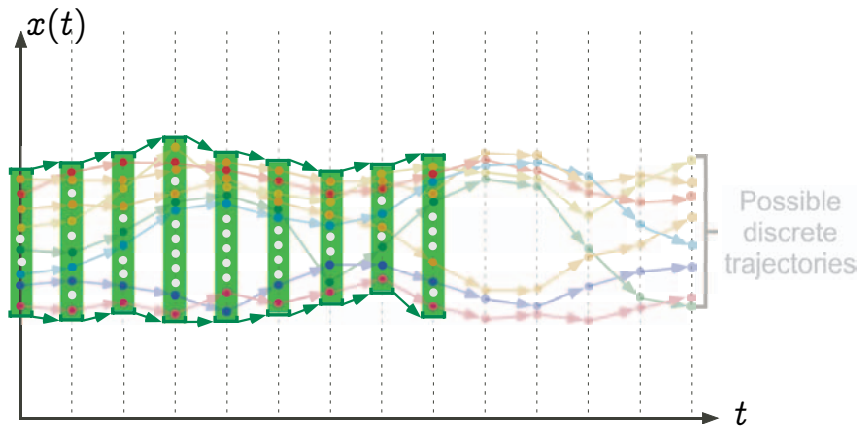
Graphic example: traces of intervals in fixpoint form



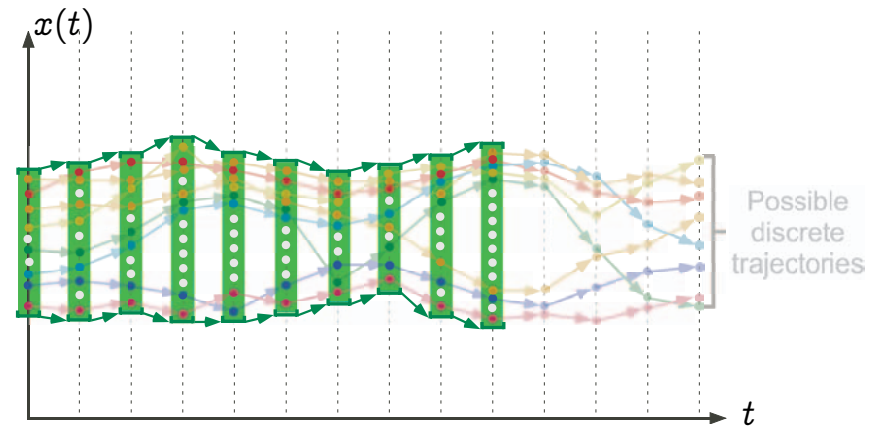
Graphic example: traces of intervals in fixpoint form



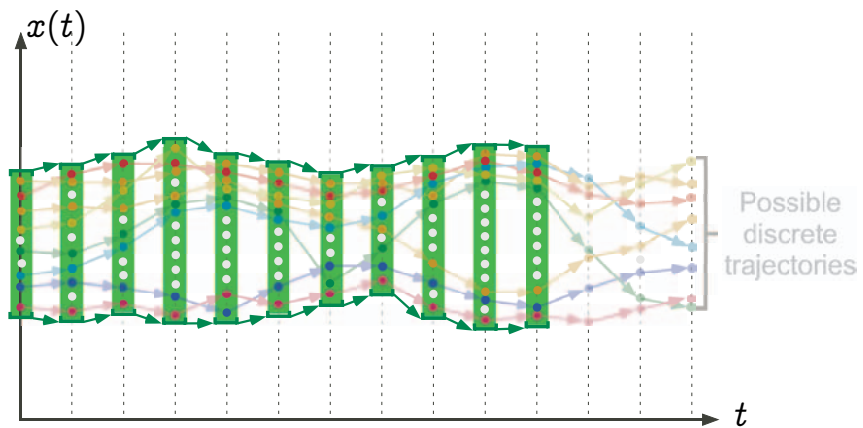
Graphic example: traces of intervals in fixpoint form



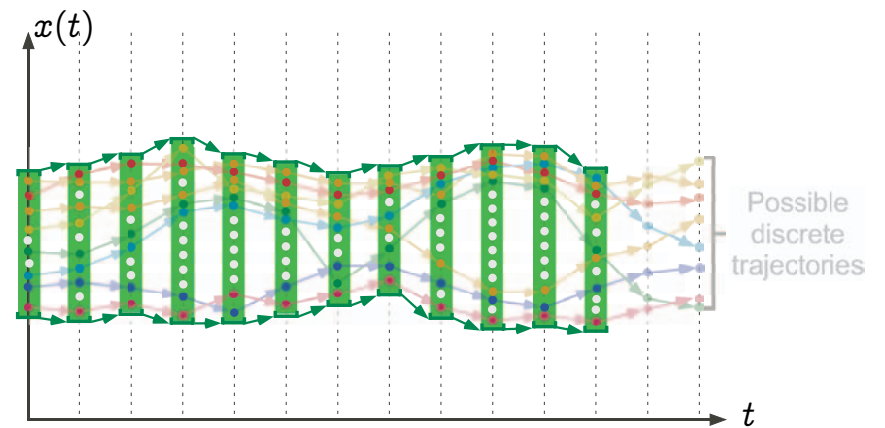
Graphic example: traces of intervals in fixpoint form



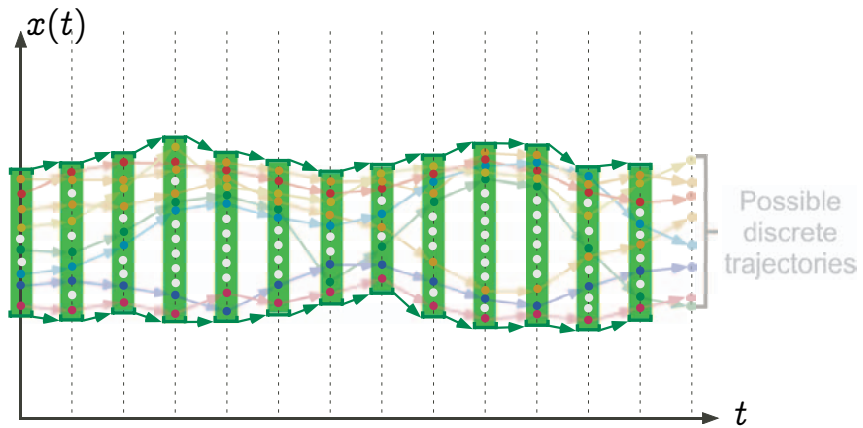
Graphic example: traces of intervals in fixpoint form



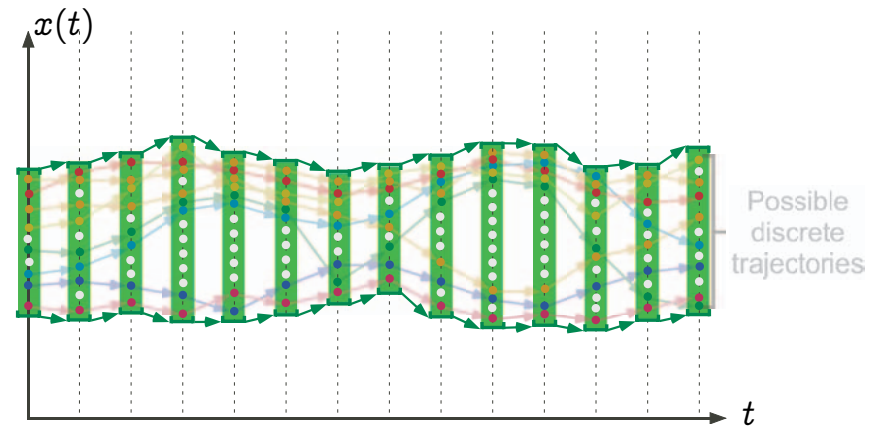
Graphic example: traces of intervals in fixpoint form



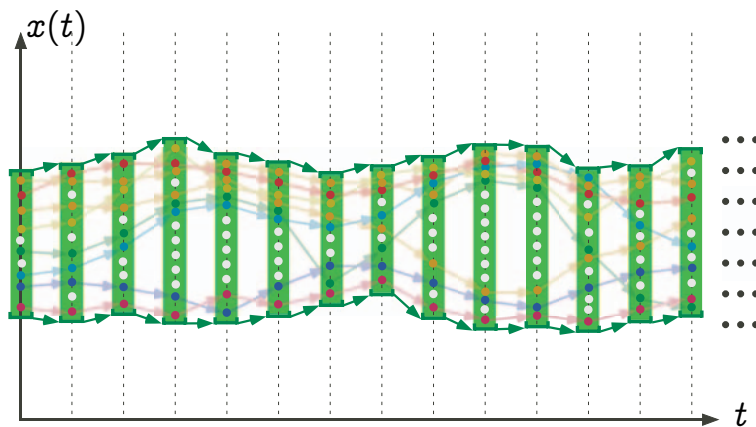
Graphic example: traces of intervals in fixpoint form



Graphic example: traces of intervals in fixpoint form



Graphic example: traces of intervals in fixpoint form



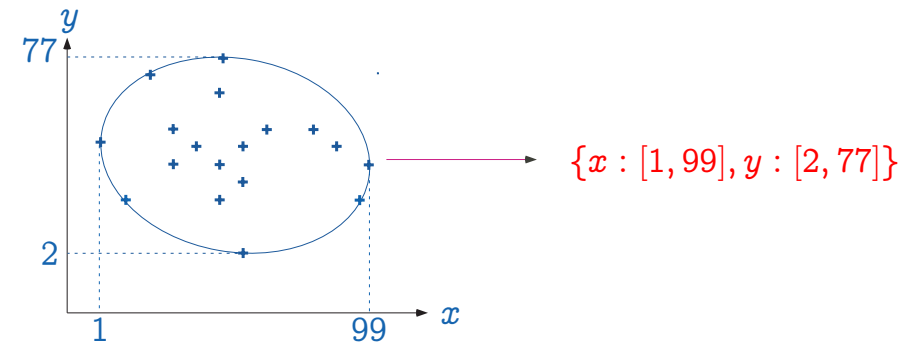
Abstraction by Galois connections



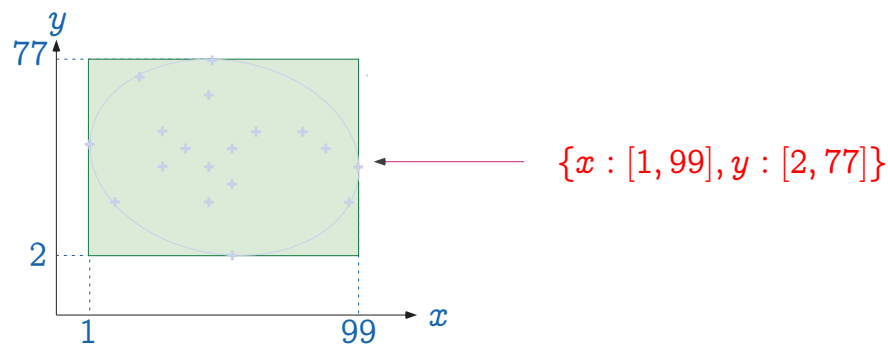
Abstracting sets (i.e. properties)

- 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) $S \subseteq \gamma(\alpha(S))$.

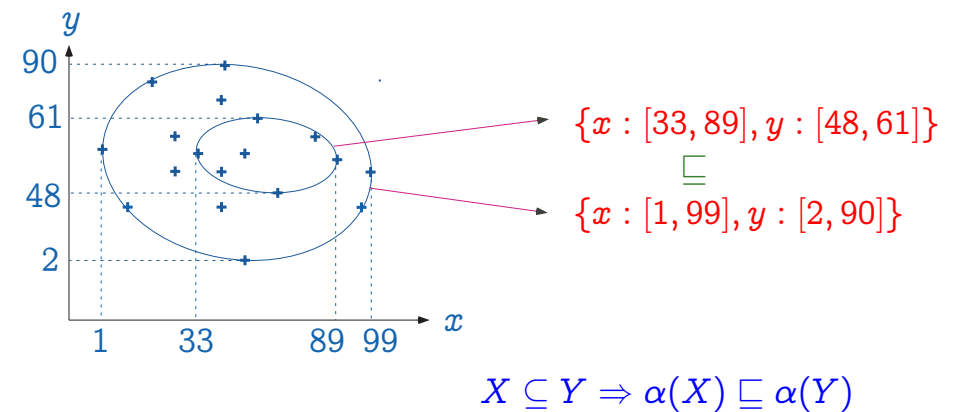
Interval abstraction α



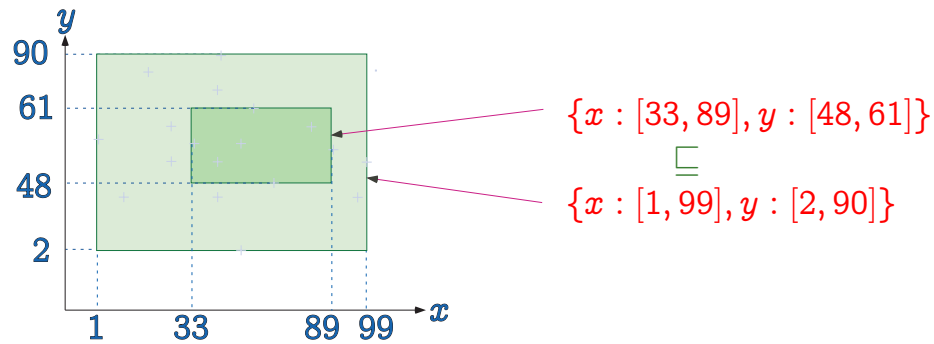
Interval concretization γ



The abstraction α is monotone



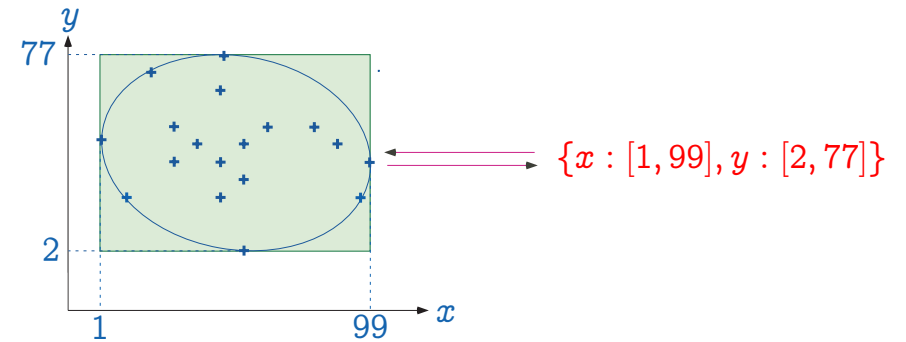
The concretization γ is monotone



$$\{x : [33, 89], y : [48, 61]\} \sqsubseteq \{x : [1, 99], y : [2, 90]\}$$

$$X \sqsubseteq Y \Rightarrow \gamma(X) \subseteq \gamma(Y)$$

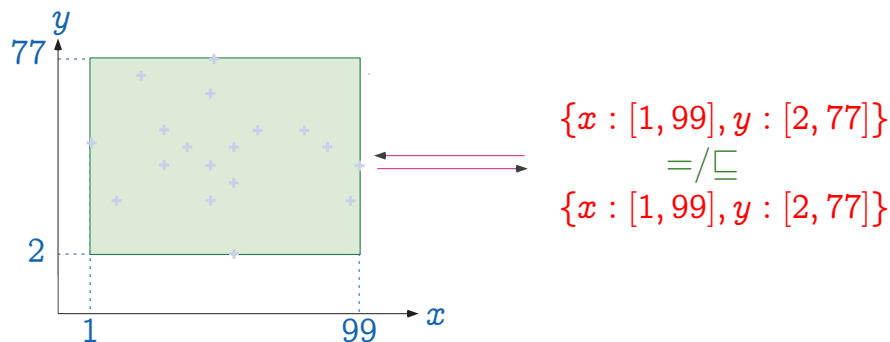
The $\gamma \circ \alpha$ composition is extensive



$$\{x : [1, 99], y : [2, 77]\}$$

$$X \subseteq \gamma \circ \alpha(X)$$

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



$$\{x : [1, 99], y : [2, 77]\} \neq \sqsubseteq \{x : [1, 99], y : [2, 77]\}$$

$$\alpha \circ \gamma(Y) \neq \sqsubseteq Y$$

Correspondance between concrete and abstract properties

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

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

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

Galois connection

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

iff

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

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

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

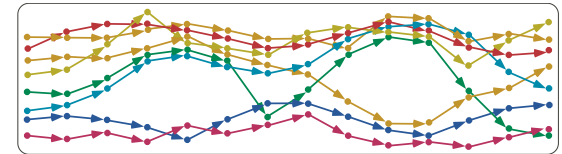
$$\wedge \forall \bar{y} \in \overline{\mathcal{D}} : \alpha(\gamma(\bar{y})) \sqsubseteq \bar{y}$$

iff

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

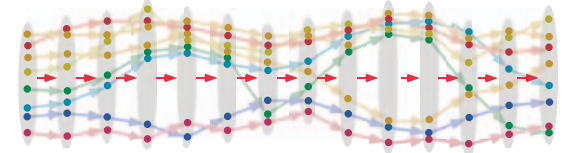
Example: Set of traces to trace of intervals abstraction

Set of traces:



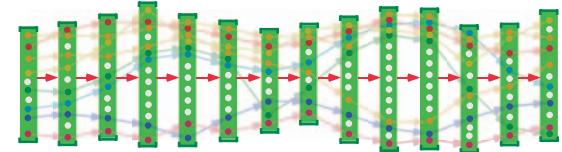
$\alpha_1 \downarrow$

Trace of sets:



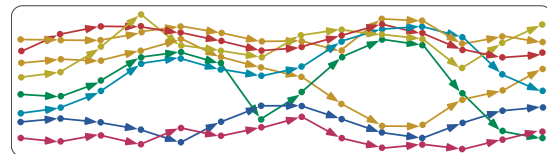
$\alpha_2 \downarrow$

Trace of intervals



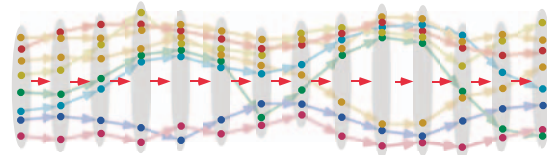
Example: Set of traces to reachable states abstraction

Set of traces:



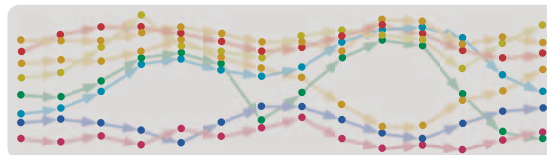
$\alpha_1 \downarrow$

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:

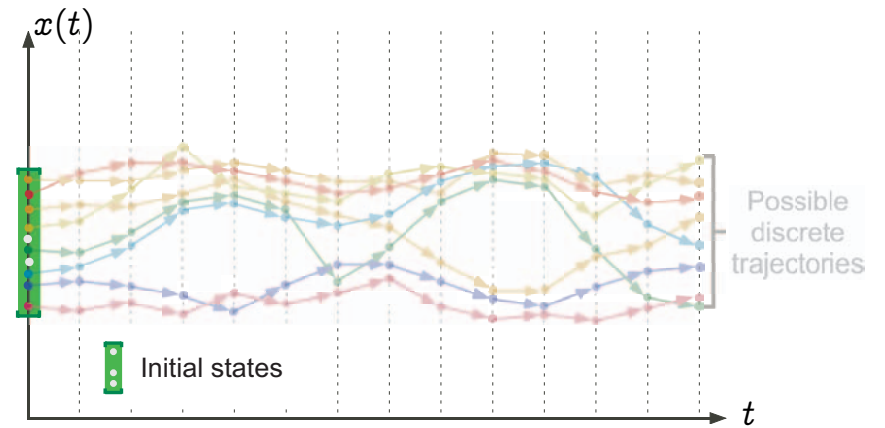
$$\langle M, \sqsubseteq \rangle \xrightarrow[\alpha_2]{\gamma_2} \langle N, \preceq \rangle$$

is a Galois connection:

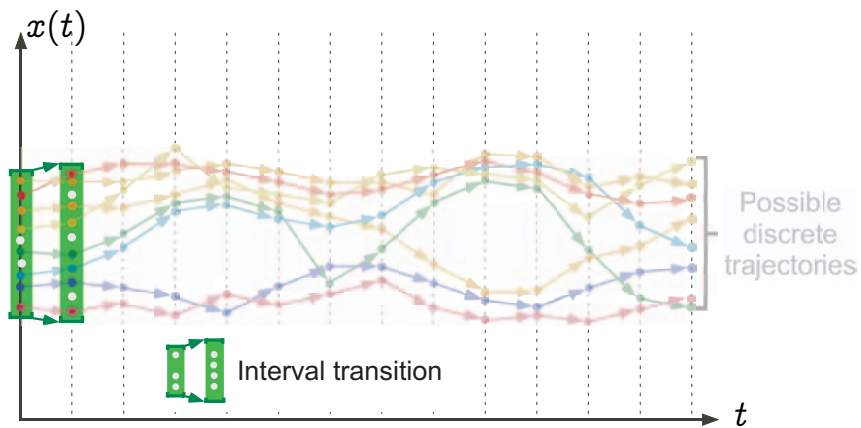
$$\langle L, \leq \rangle \xrightarrow[\alpha_2 \circ \alpha_1]{\gamma_1 \circ \gamma_2} \langle N, \preceq \rangle$$

Convergence acceleration
by widening/narrowing

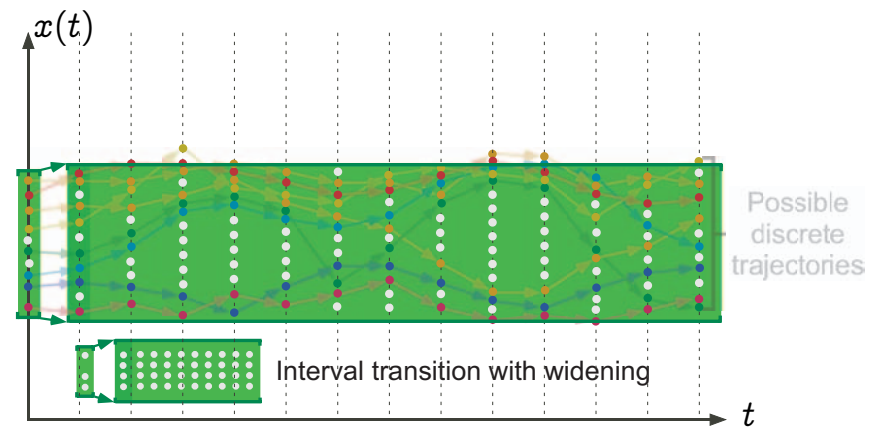
Graphic example: upward iteration
with widening



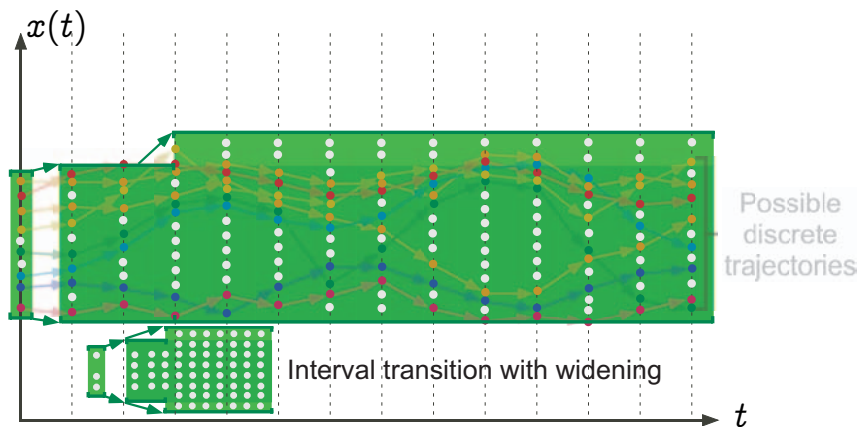
Graphic example: upward iteration
with widening



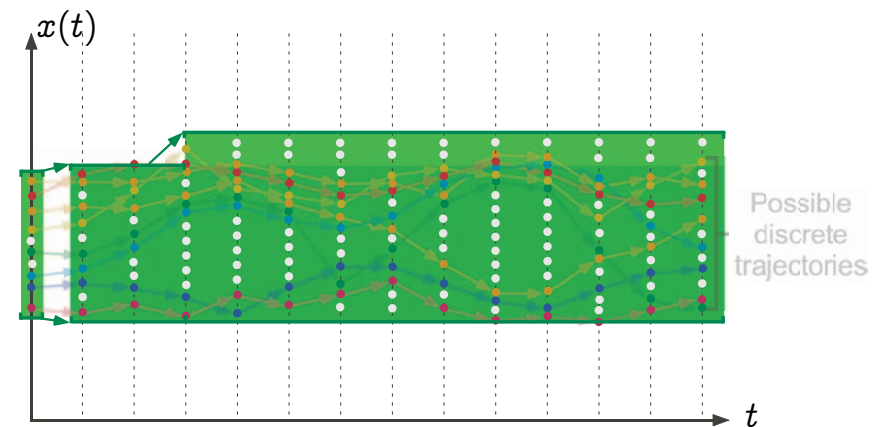
Graphic example: upward iteration
with widening



Graphic example: upward iteration with widening



Graphic example: stability of the upward iteration



Interval widening

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

$$\perp \nabla X = X$$

$$X \nabla \perp = X$$

$$[l_0, u_0] \nabla [l_1, u_1] = \begin{cases} -\infty & \text{if } l_1 < l_0 \\ l_0 & \text{if } l_1 \geq l_0 \end{cases} \quad \begin{cases} +\infty & \text{if } u_1 > u_0 \\ u_0 & \text{if } u_1 \leq u_0 \end{cases}$$

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):

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

Example: Interval analysis (1975)

Resolution by chaotic increasing iteration:

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

$$\left\{ \begin{array}{l}
 X_1 = \emptyset \\
 X_2 = \emptyset \\
 X_3 = \emptyset \\
 X_4 = \emptyset
 \end{array} \right.$$

Example: Interval analysis (1975)

Increasing chaotic iteration:

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

$$\left\{ \begin{array}{l}
 X_1 = [1, 1] \\
 X_2 = \emptyset \\
 X_3 = \emptyset \\
 X_4 = \emptyset
 \end{array} \right.$$

Example: Interval analysis (1975)

Increasing chaotic iteration:

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

$$\left\{ \begin{array}{l}
 X_1 = [1, 1] \\
 X_2 = [1, 1] \\
 X_3 = \emptyset \\
 X_4 = \emptyset
 \end{array} \right.$$

Example: Interval analysis (1975)

Increasing chaotic iteration:

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

Example: Interval analysis (1975)

Increasing chaotic iteration:

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

Example: Interval analysis (1975)

Increasing chaotic iteration: **convergence !**

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

Example: Interval analysis (1975)

Increasing chaotic iteration: **convergence !!**

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

Example: Interval analysis (1975)

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

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

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

Example: Interval analysis (1975)

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

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

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

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 !!!!!!!**

$$\begin{array}{l}
 x := 1; \\
 1: \\
 \text{while } x < 10000 \text{ do} \\
 2: \\
 \quad x := x + 1 \\
 3: \\
 \text{od;} \\
 4:
 \end{array}
 \left\{ \begin{array}{l}
 X_1 = [1, 1] \\
 X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\
 X_3 = X_2 \oplus [1, 1] \\
 X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \\
 \\
 X_1 = [1, 1] \\
 X_2 = [1, 5] \\
 X_3 = [2, 6] \\
 X_4 = \emptyset
 \end{array} \right.$$

Example: Interval analysis (1975)

Convergence speed-up by widening:

$$\begin{array}{l}
 x := 1; \\
 1: \\
 \text{while } x < 10000 \text{ do} \\
 2: \\
 \quad x := x + 1 \\
 3: \\
 \text{od;} \\
 4:
 \end{array}
 \left\{ \begin{array}{l}
 X_1 = [1, 1] \\
 X_2 = (X_1 \cup X_3) \cap [-\infty, 9999] \\
 X_3 = X_2 \oplus [1, 1] \\
 X_4 = (X_1 \cup X_3) \cap [10000, +\infty] \\
 \\
 X_1 = [1, 1] \\
 X_2 = [1, +\infty] \leftarrow \text{widening} \\
 X_3 = [2, 6] \\
 X_4 = \emptyset
 \end{array} \right.$$

Example: Interval analysis (1975)

Decreasing chaotic iteration:

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

Example: Interval analysis (1975)

Decreasing chaotic iteration:

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

Example: Interval analysis (1975)

Decreasing chaotic iteration:

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

Example: Interval analysis (1975)

Final solution:

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

Example: Interval analysis (1975)

Result of the interval analysis:

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

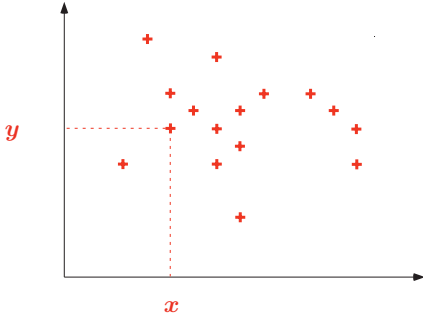
Example: Interval analysis (1975)

Checking absence of runtime errors with interval analysis:

$$\begin{array}{l}
 x := 1; \\
 1: \{x = 1\} \\
 \text{while } x < 10000 \text{ do} \\
 2: \{x \in [1, 9999]\} \\
 \quad x := x + 1 \quad \leftarrow \text{no overflow} \\
 3: \{x \in [2, +10000]\} \\
 \text{od;} \\
 4: \{x = 10000\}
 \end{array}$$

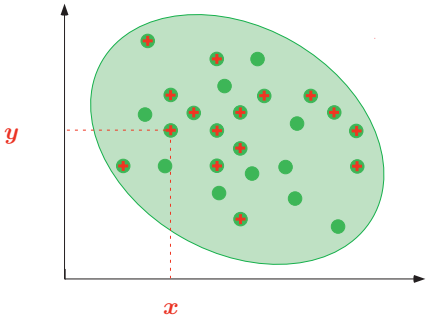
Refinement of abstractions

Approximations of an [in]finite set of points:



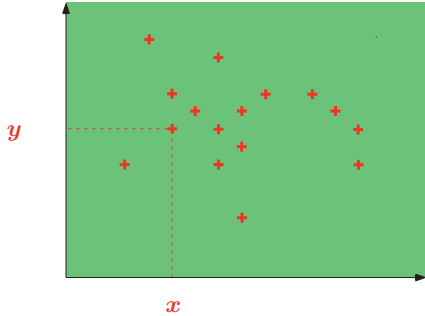
$$\{\dots, \langle 19, 77 \rangle, \dots, \langle 20, 03 \rangle, \dots\}$$

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



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

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



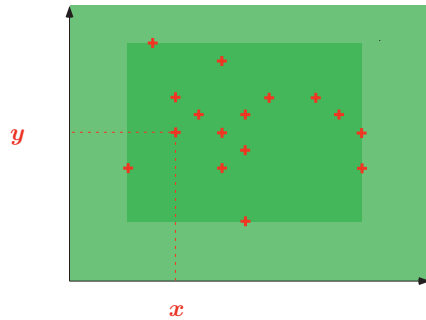
$$\begin{cases} x \geq 0 \\ y \geq 0 \end{cases}$$

From Below: dual³ + combinations.

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

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

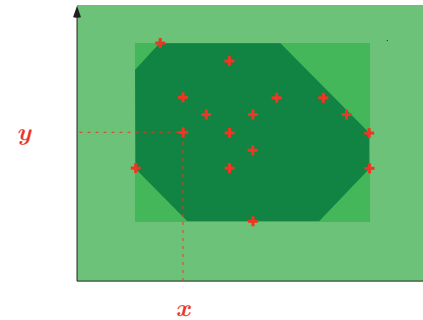
Effective computable approximations of an [in]finite set of points; Intervals⁵



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

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

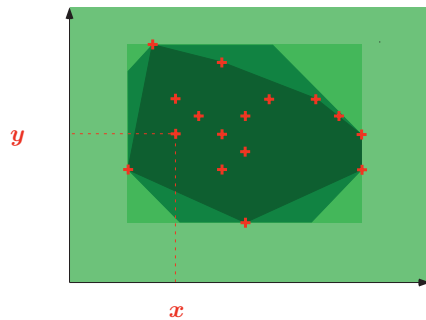
Effective computable approximations of an [in]finite set of points; Octagons⁶



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

⁶ A. Miné. *A New Numerical Abstract Domain Based on Difference-Bound Matrices*. PADO'2001. LNCS 2053, pp. 155–172. Springer 2001. See the *The Octagon Abstract Domain Library* on <http://www.di.ens.fr/~mine/oct/>

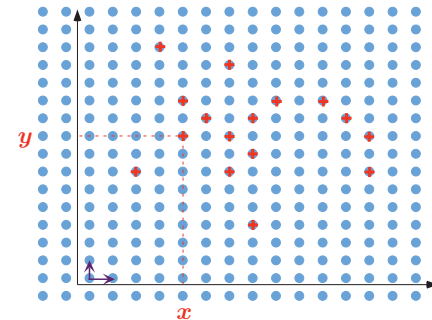
Effective computable approximations of an [in]finite set of points; Polyhedra⁷



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

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

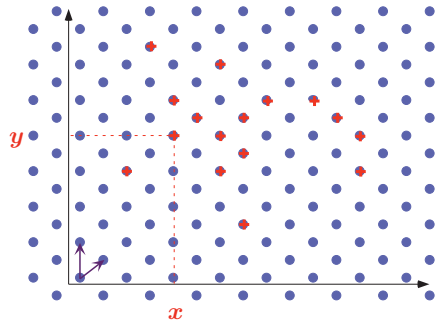
Effective computable approximations of an [in]finite set of points; Simple congruences⁸



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

⁸ Ph. Granger. *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

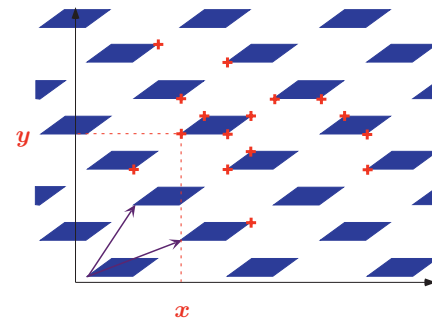


$$\begin{cases} 1x + 9y = 7 \pmod{8} \\ 2x - 1y = 9 \pmod{9} \end{cases}$$

⁹ Ph. Granger. *Static Analysis of Linear Congruence Equalities among Variables of a Program*. TAPSOFT'91, pp. 169–192. LNCS 493, Springer, 1991.



Effective computable approximations of an [in]finite set of points; Trapezoidal linear congruences¹⁰



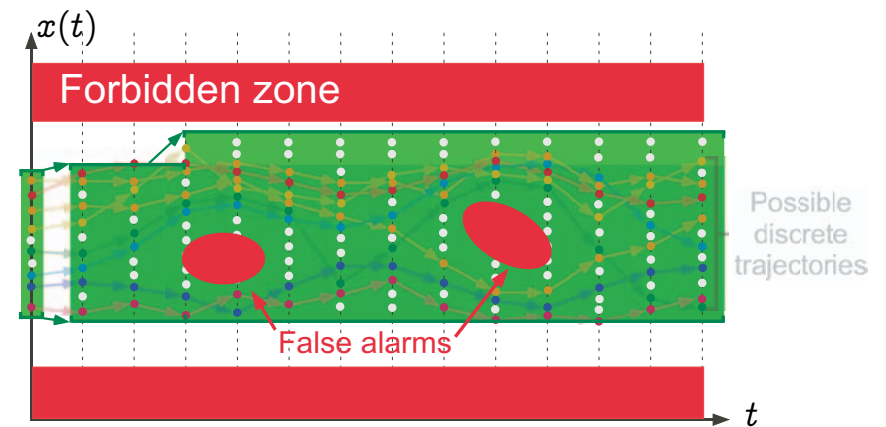
$$\begin{cases} 1x + 9y \in [0, 77] \pmod{10} \\ 2x - 1y \in [0, 99] \pmod{11} \end{cases}$$

¹⁰ F. Masdupuy. *Array Operations Abstraction Using Semantic Analysis of Trapezoid Congruences*. ACM ICS'92.

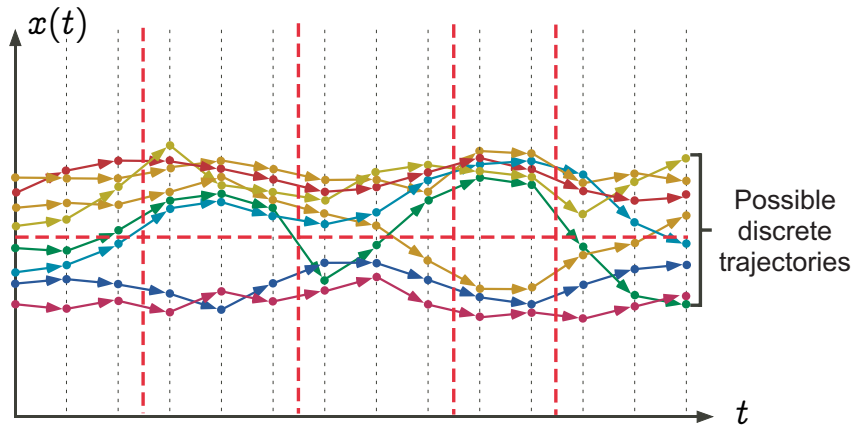


Refinement of iterates

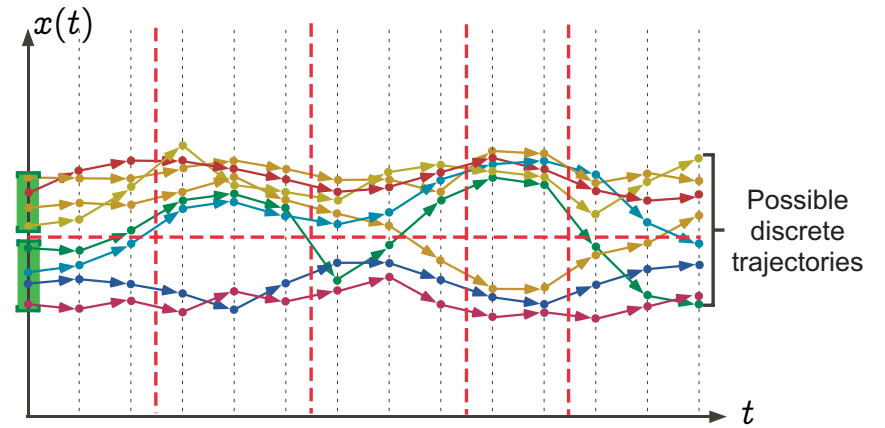
Graphic example: Refinement required by false alarms



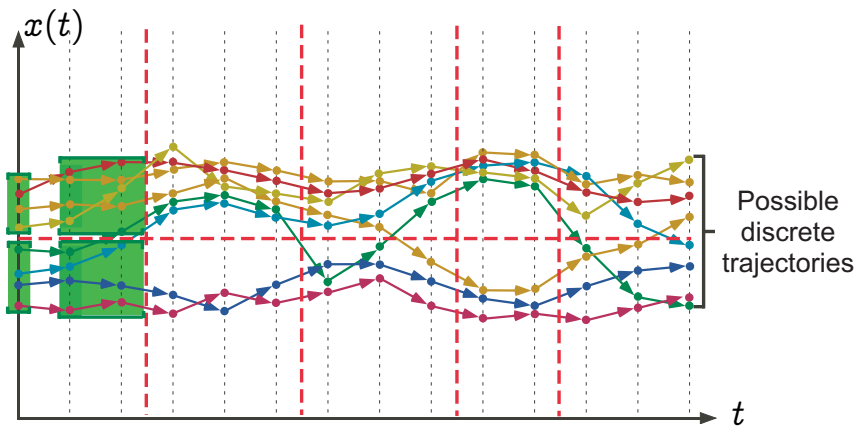
Graphic example: Partitioning



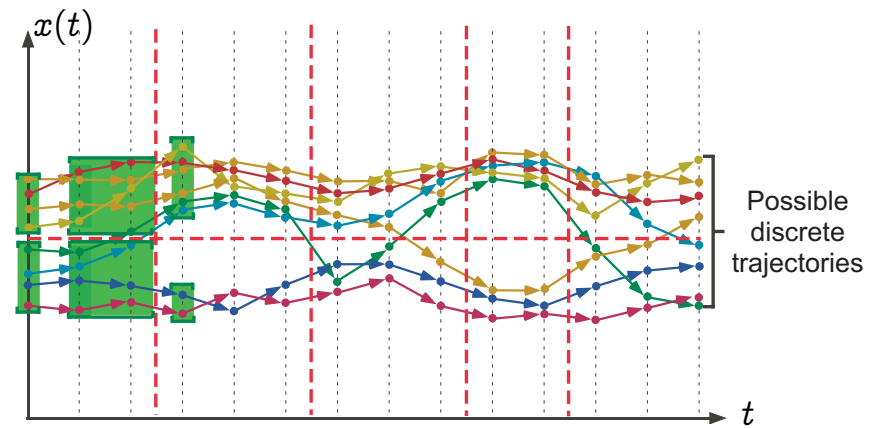
Graphic example: partitionned upward iteration with widening



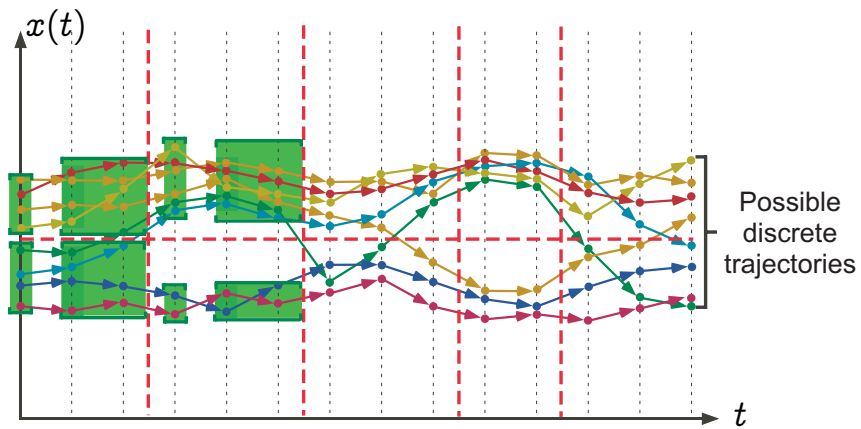
Graphic example: partitionned upward iteration with widening



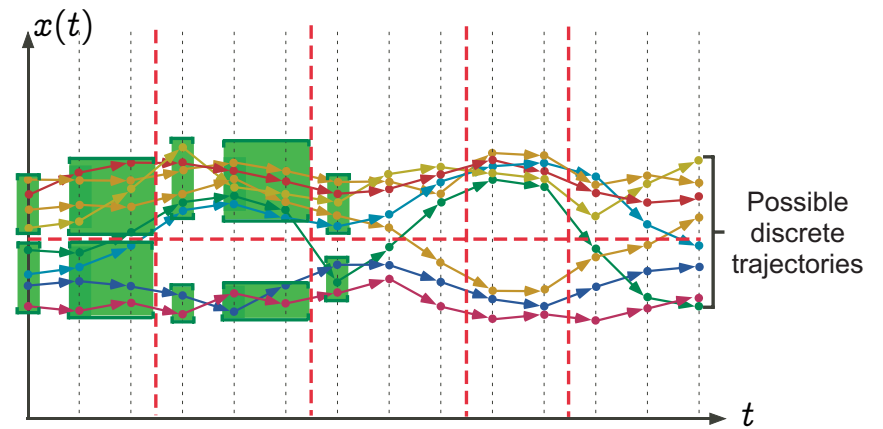
Graphic example: partitionned upward iteration with widening



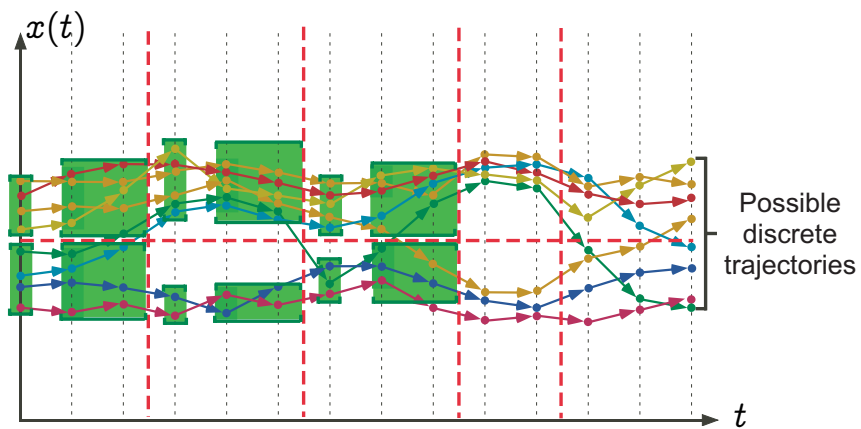
Graphic example: partitionned upward iteration with widening



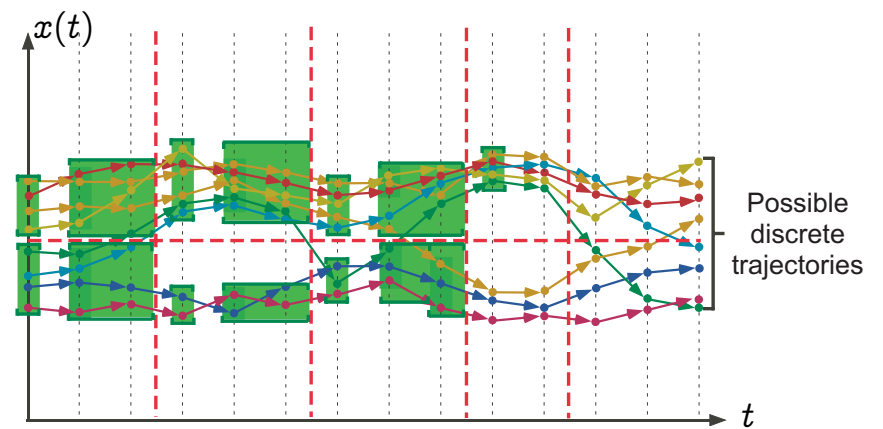
Graphic example: partitionned upward iteration with widening



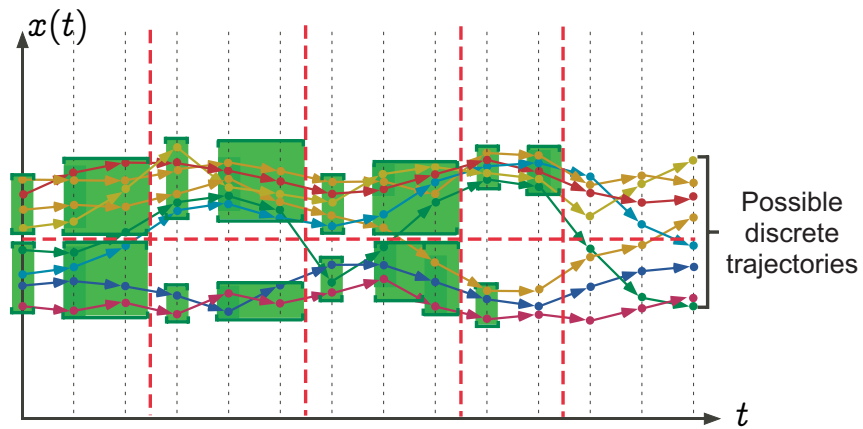
Graphic example: partitionned upward iteration with widening



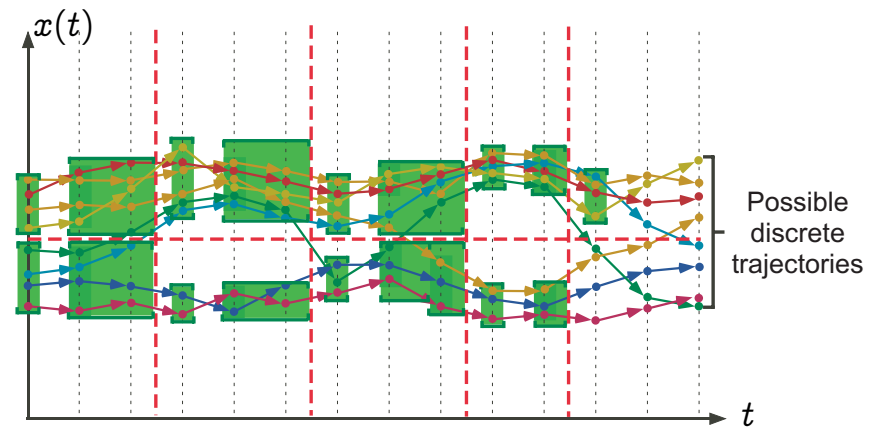
Graphic example: partitionned upward iteration with widening



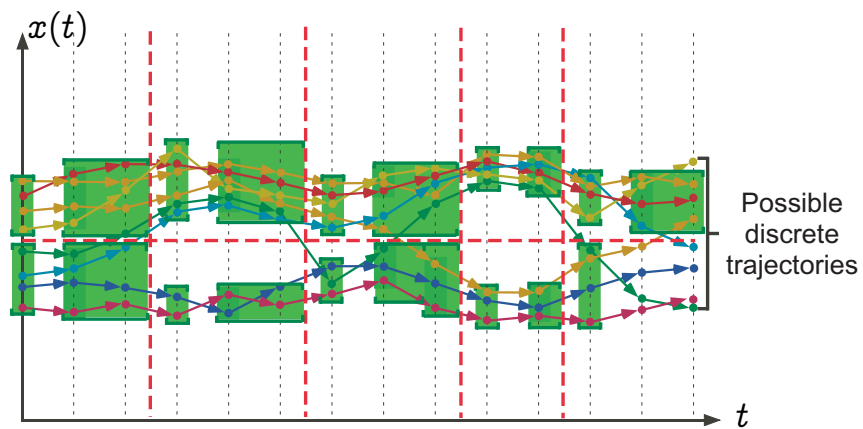
Graphic example: partitionned upward iteration with widening



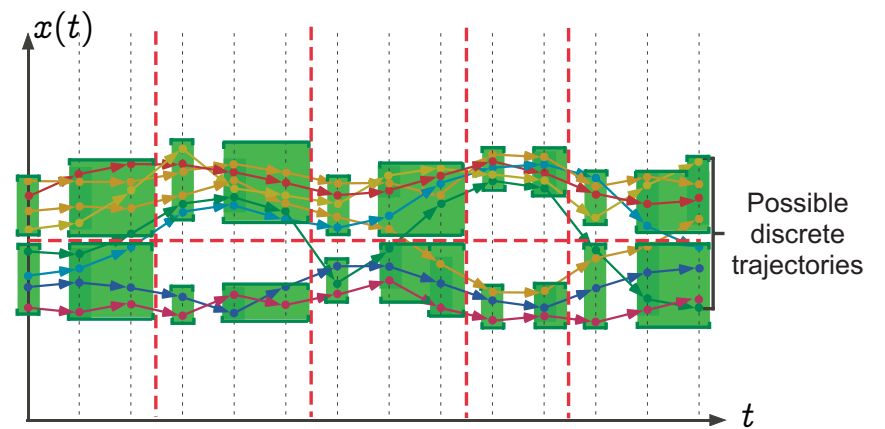
Graphic example: partitionned upward iteration with widening



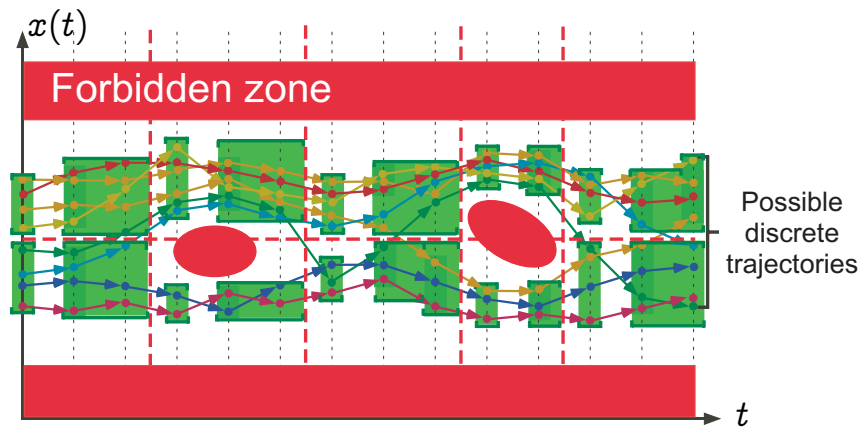
Graphic example: partitionned upward iteration with widening



Graphic example: partitionned upward iteration with widening



Graphic example: safety verification

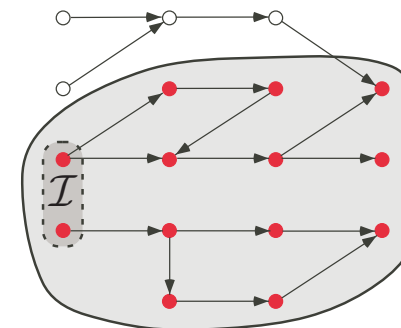


Interval widening with threshold set

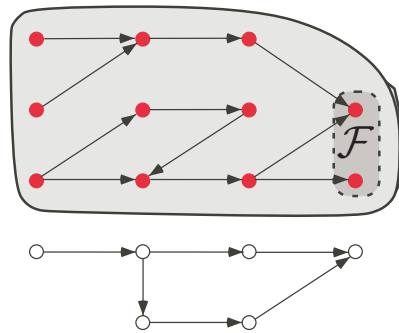
- The **threshold set** T is a finite set of numbers (plus $+\infty$ and $-\infty$),
- $[a, b] \nabla_T [a', b'] = [\text{if } a' < a \text{ then } \max\{l \in T \mid l \leq a'\} \text{ else } a, \text{ if } b' > b \text{ then } \min\{h \in T \mid h \geq b'\} \text{ else } b]$.
- Examples (intervals):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

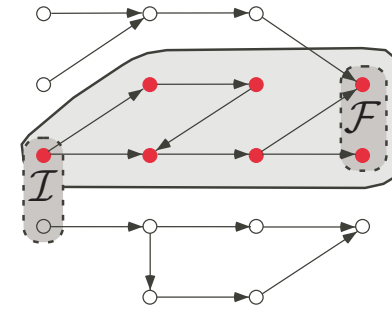
Forward/reachability analysis



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
 - **StackAnalyzer**: 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

Bibliography

¹¹ applied to the primary flight control software of the Airbus A340/600 and A380 fly-by-wire systems

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);



- [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);



- [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)



Anticipated Home Work of Course 16.399: 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 analyzer](#) 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/~cousot/>

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