

« A Tutorial on Abstract Interpretation »

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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 (example: program semantics);

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.



A very informal introduction
to the principles of
abstract interpretation

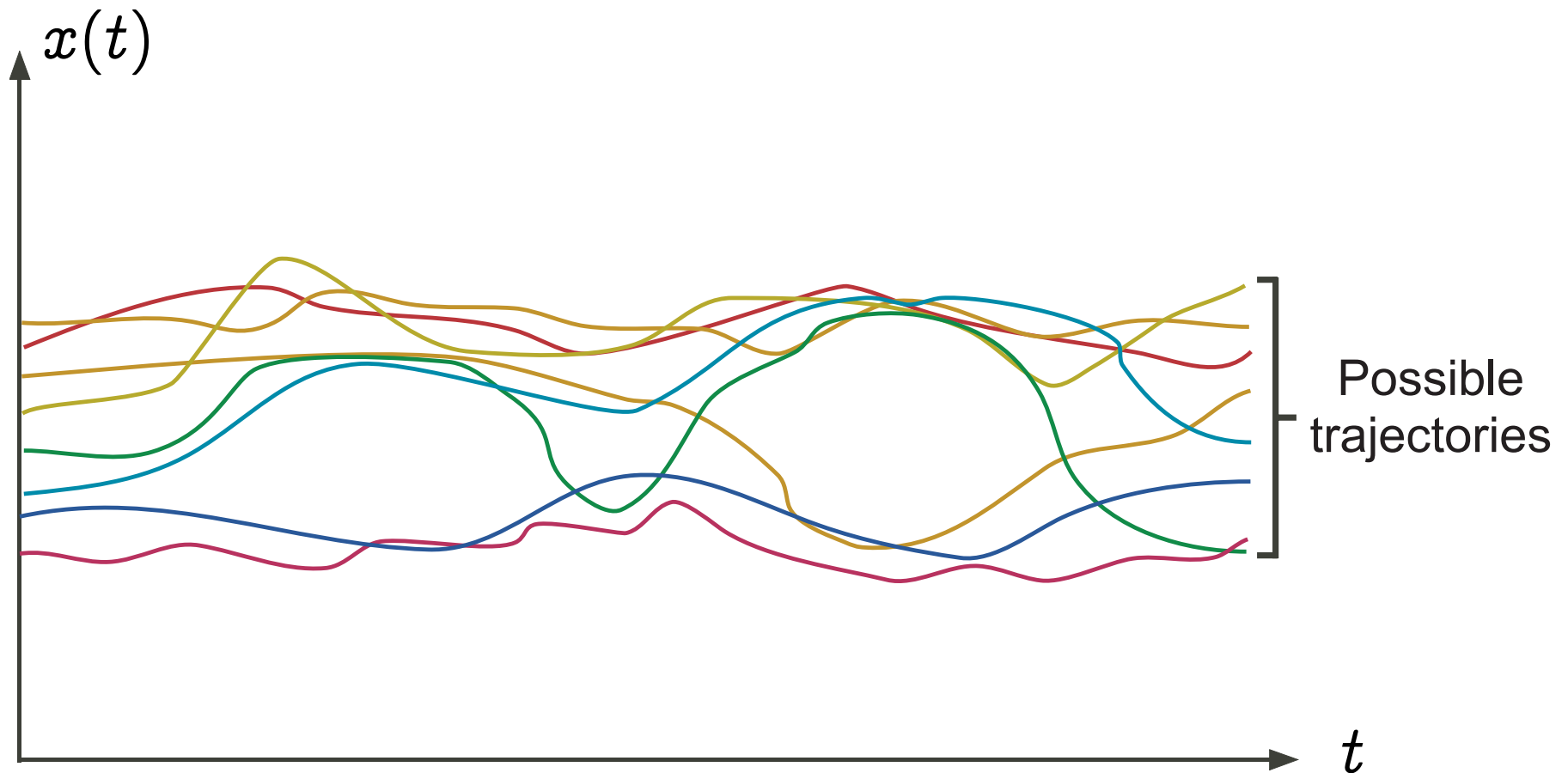


Semantics

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: termination

- Assume `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.
```

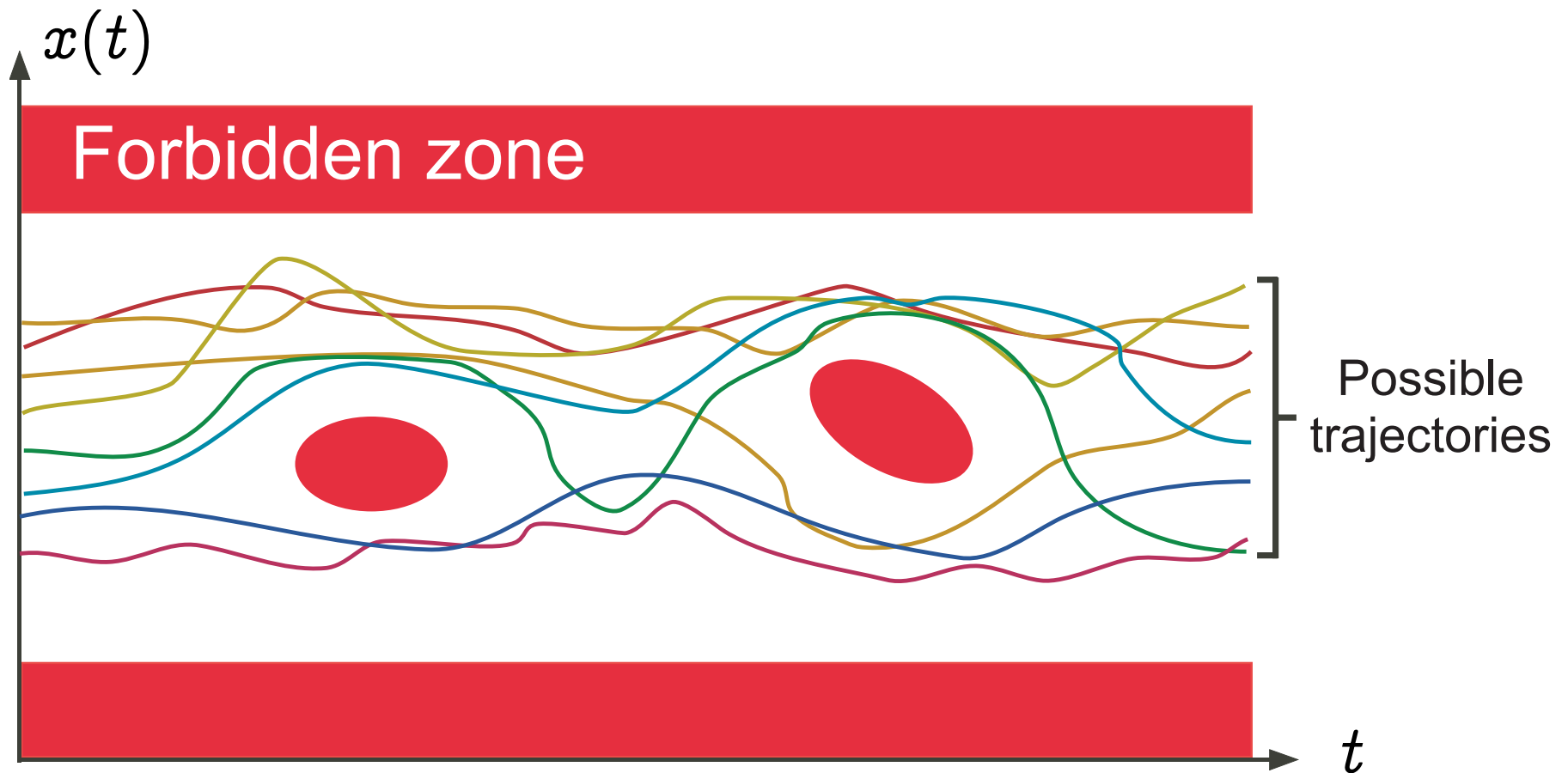


Graphic example: Safety properties

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 answer¹.

¹ e.g. probabilistic answer.

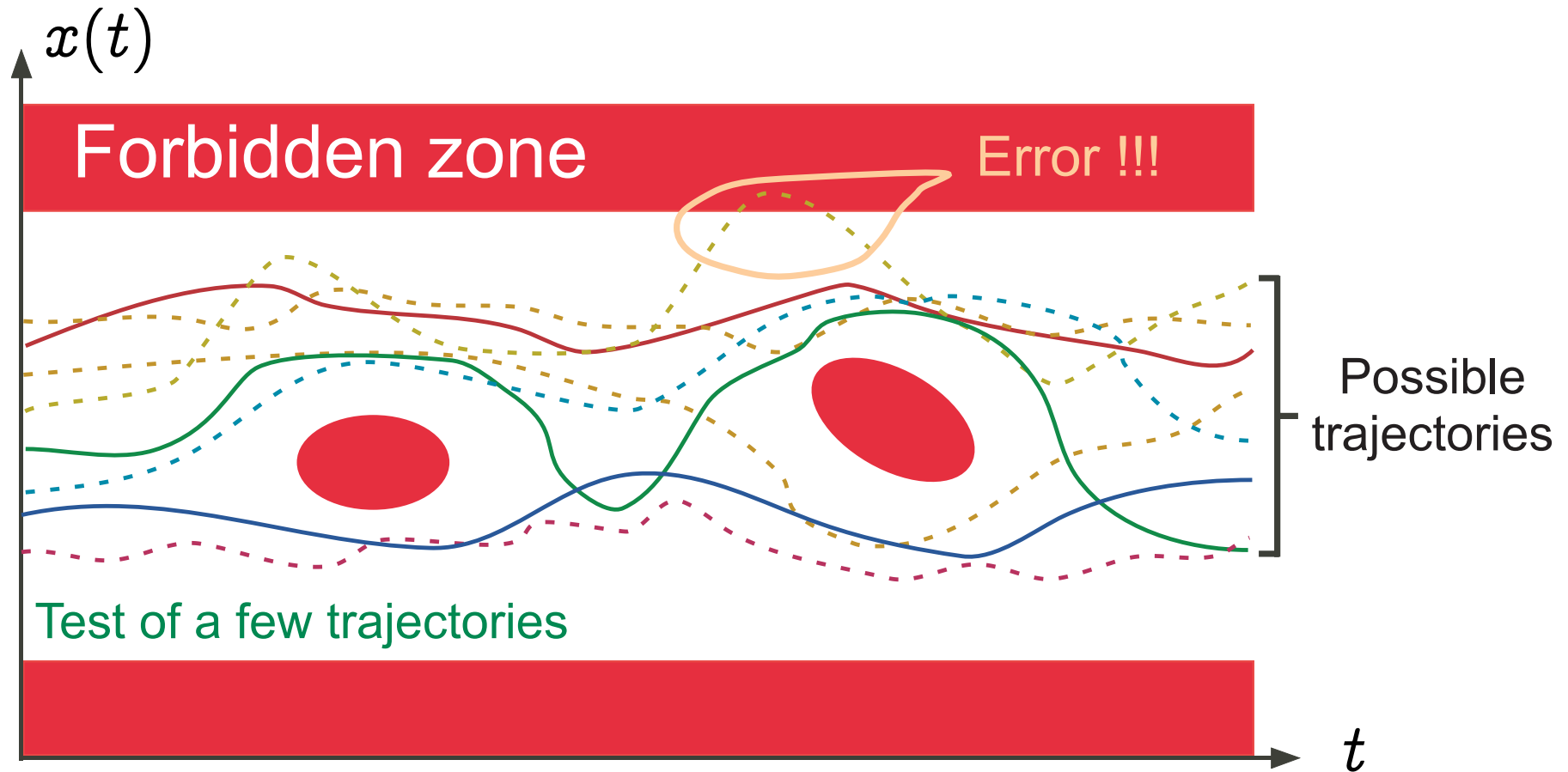


Test/debugging

- 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

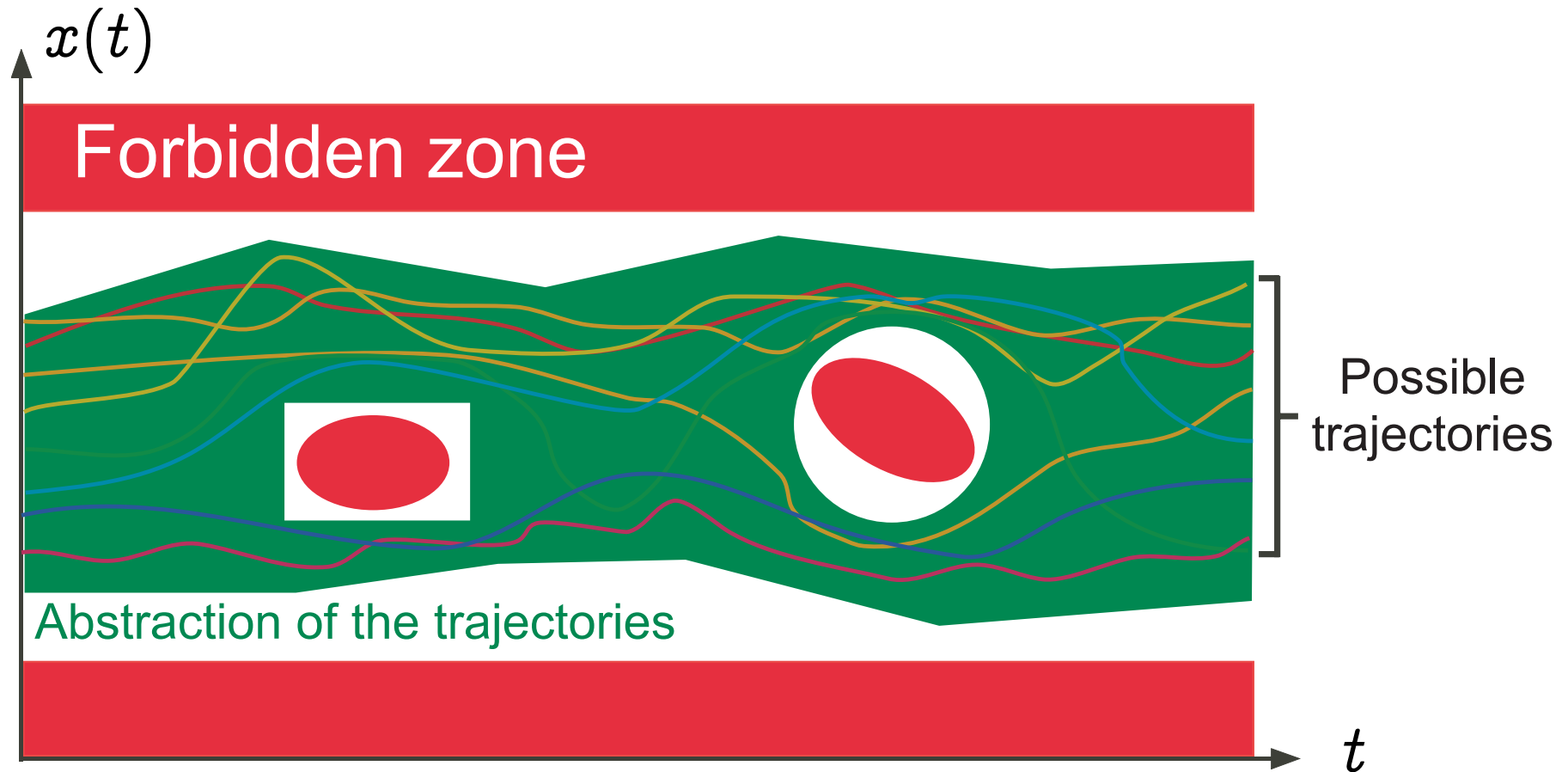


Abstract interpretation

- 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).

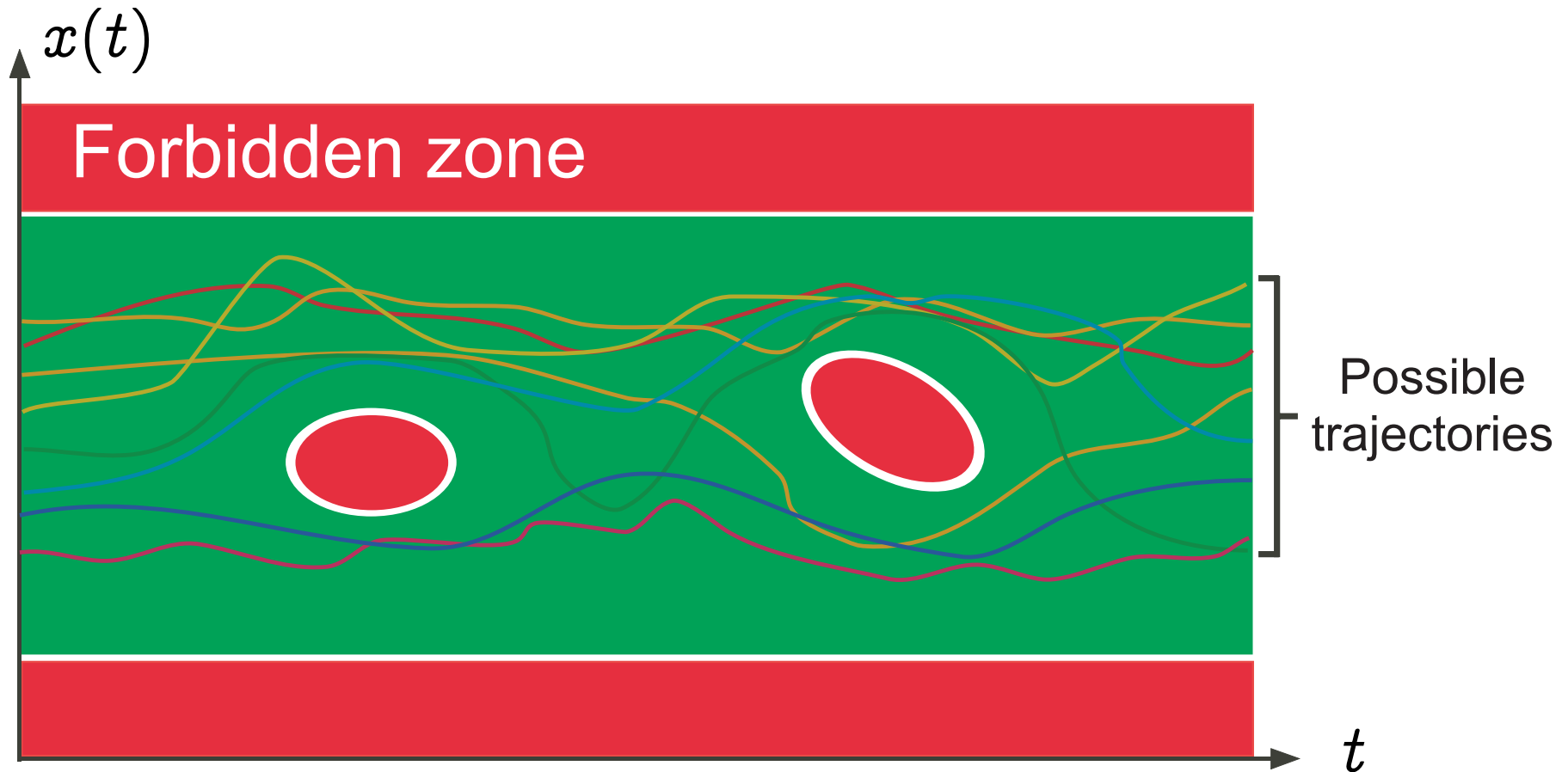


Required properties of the abstract semantics

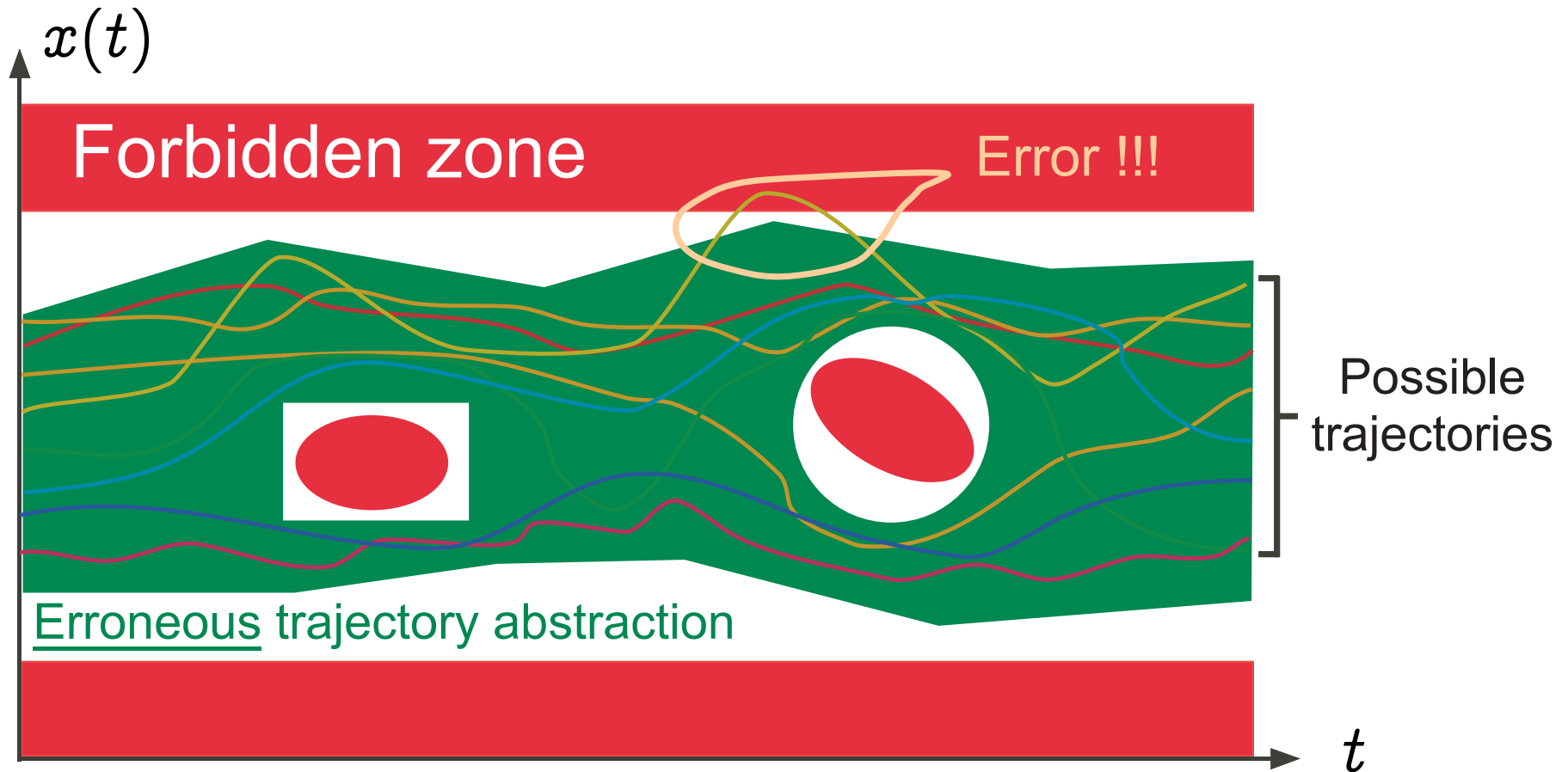
- **sound** so that no possible error can be forgotten;
- **precise** enough (to avoid false alarms);
- as **simple/abstract** as possible (to avoid combinatorial explosion phenomena).



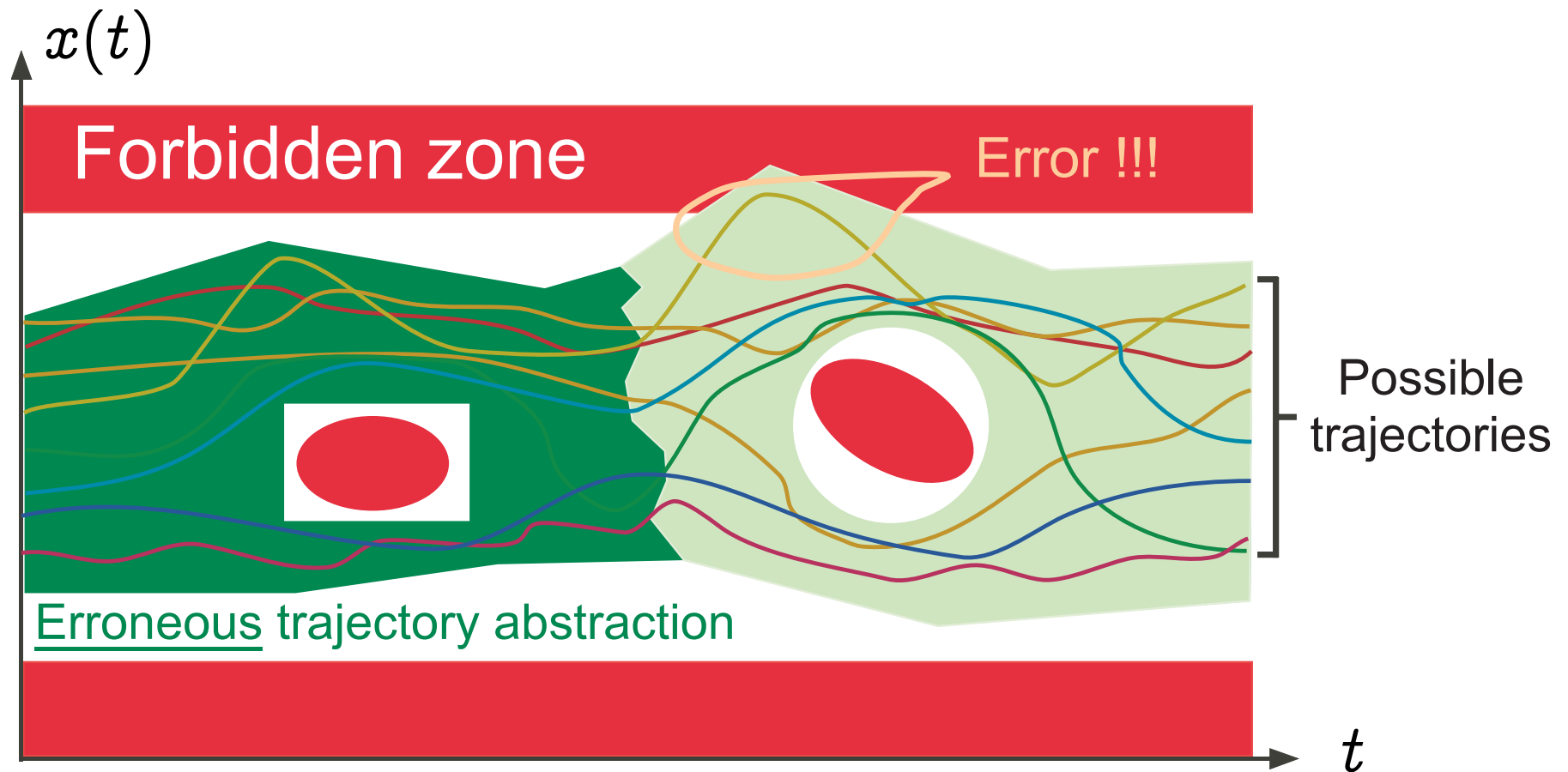
Graphic example: The most abstract correct and precise semantics



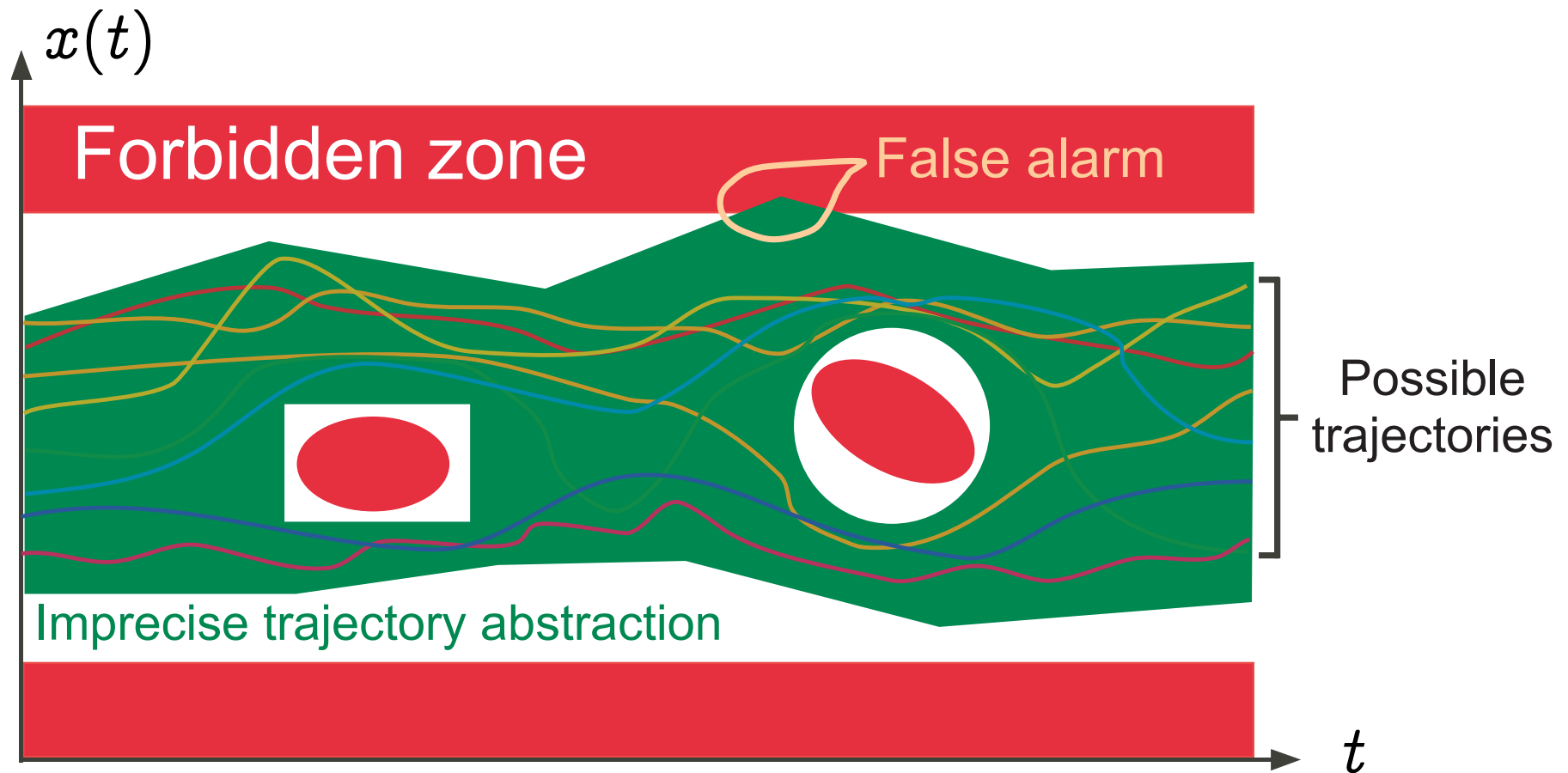
Graphic example: Erroneous abstraction — I



Graphic example: Erroneous abstraction — II



Graphic example: Imprecision \Rightarrow false alarms



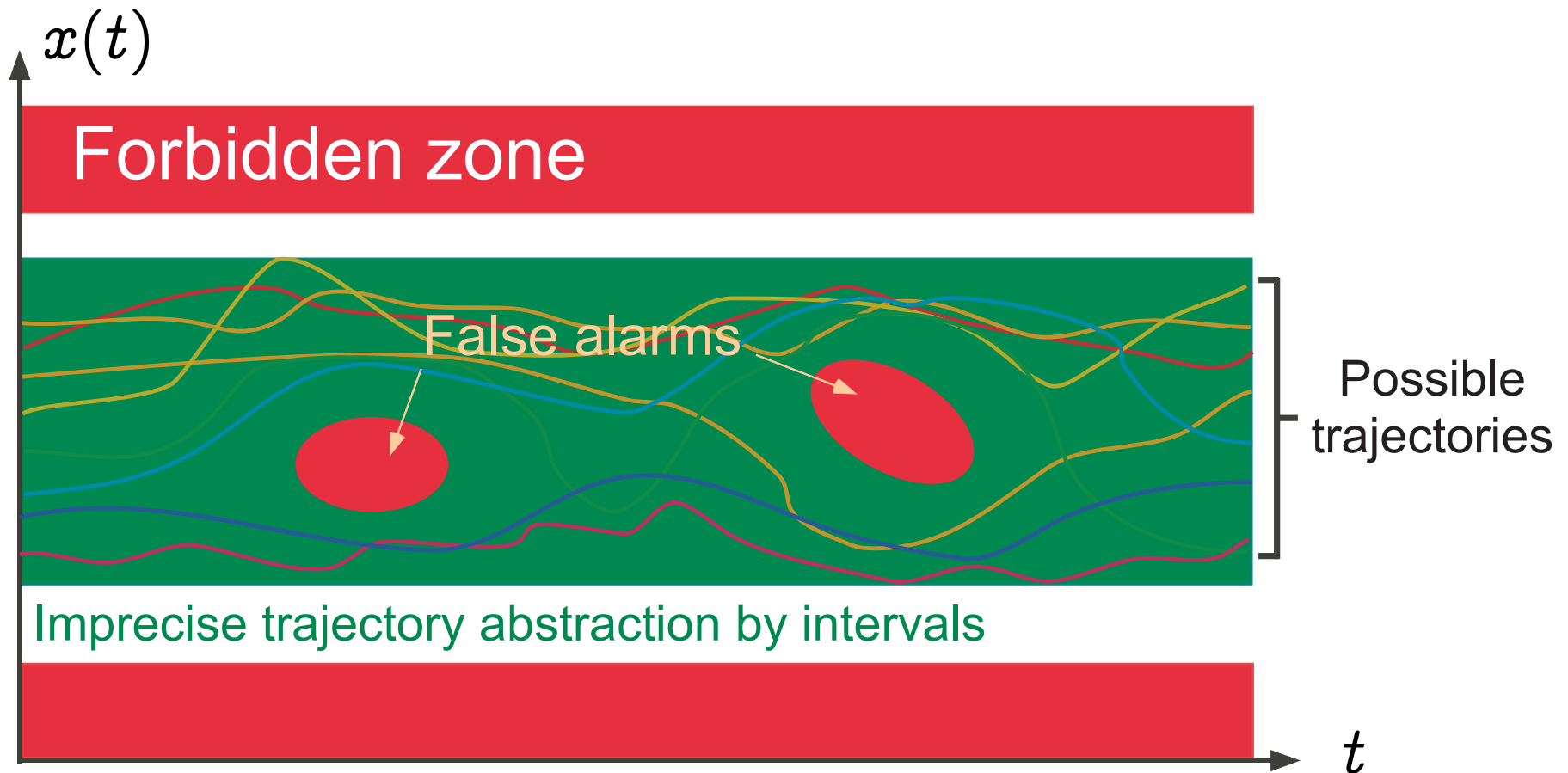
Abstract domains

Standard abstractions

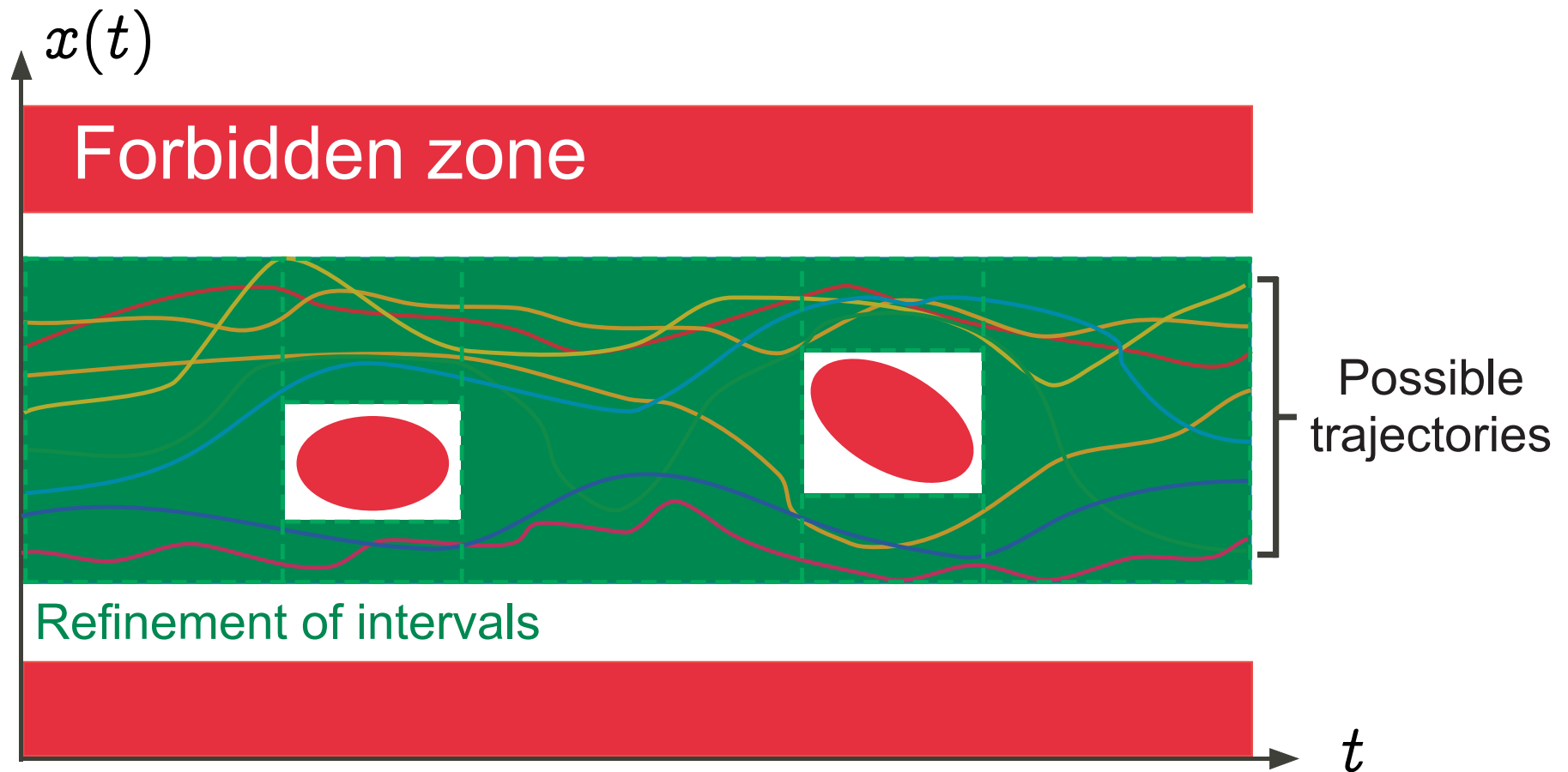
- that serve as a **basis** for the design of static analyzers:
 - abstract program data,
 - abstract program basic operations;
 - abstract program control (iteration, procedure, concurrency, ...);
- can be **parametrized** to allow for manual adaptation to the application domains.



Graphic example: Standard abstraction by intervals



Graphic example: A more refined abstraction



A very informal introduction to static analysis algorithms



Standard operational semantics

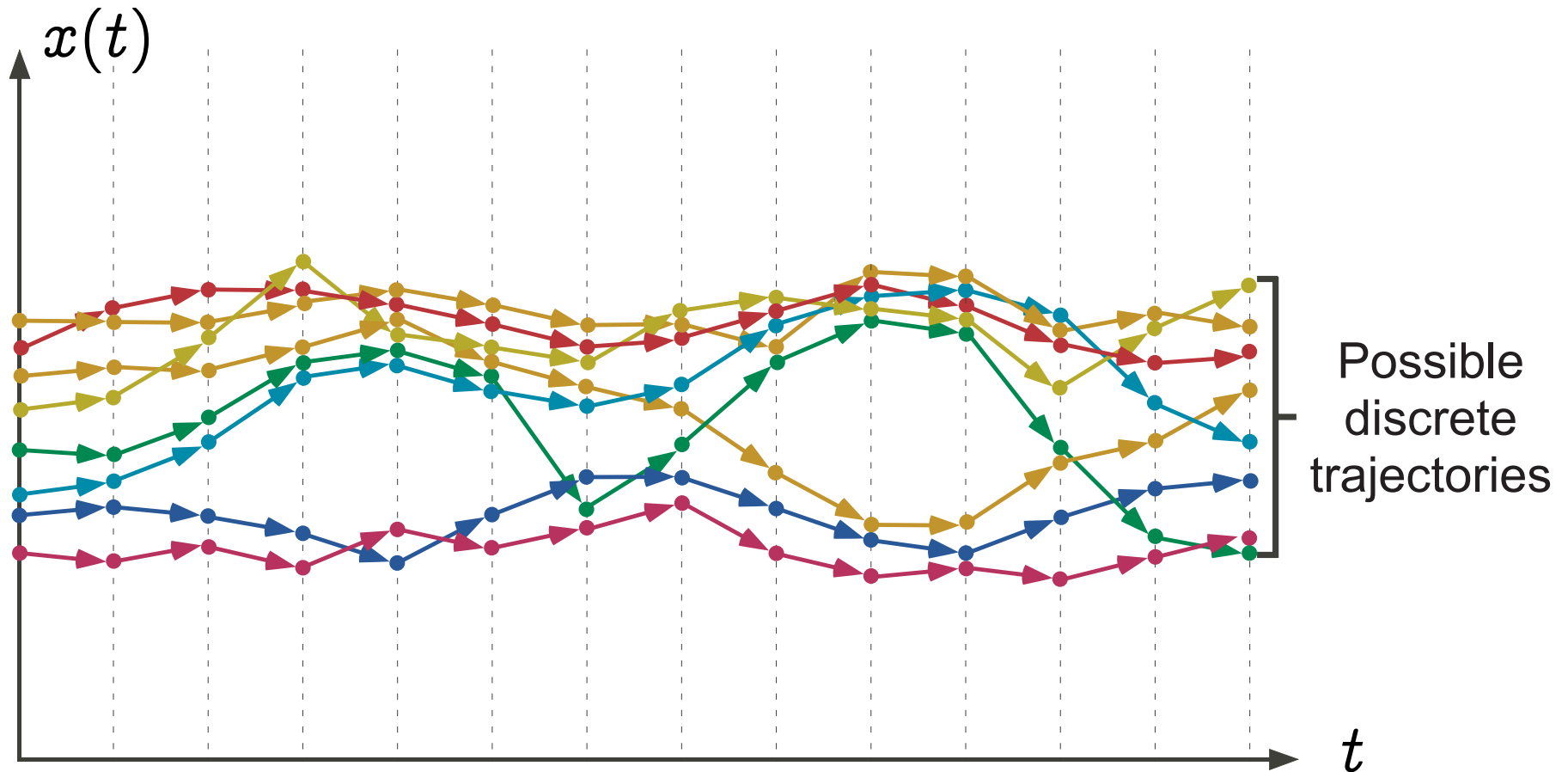


Standard semantics

- Start from a **standard operational semantics** that describes formally:
 - **states** that is data values of program variables,
 - **transitions** that is elementary computation steps;
- Consider **traces** that is successions of states corresponding to executions described by transitions (possibly infinite).



Graphic example: Small-steps transition semantics



Example: Small-steps transition semantics of an assignment

```
int x;
```

```
...
```

```
l:
```

```
    x := x + 1;
```

```
l':
```

$$\{l : x = v \rightarrow l' : x = v + 1 \mid v \in [\text{min_int}, \text{max_int} - 1]\} \\ \cup \{l : x = \text{max_int} \rightarrow l' : x = \Omega\} \quad (\text{runtime error})$$



Example: Small-steps transition semantics of a loop

```

11: x := 1;
12: while x < 10 do
13:   x := x + 1
14: od
15:

```

```

11 : ...
11 : x = -1
11 : x = 0
11 : x = 1
11 : ...
12 : x = 1 → 13 : x = 1
13 : x = 1 → 14 : x = 2
14 : x = 2 → 13 : x = 2
13 : x = 2 → 14 : x = 3
...
14 : x = 10 → 15 : x = 10

```

a loop



Example: Trace semantics of loop

```
11: x := 1;  
12: while x < 10 do  
13:   x := x + 1  
14: od  
15:
```

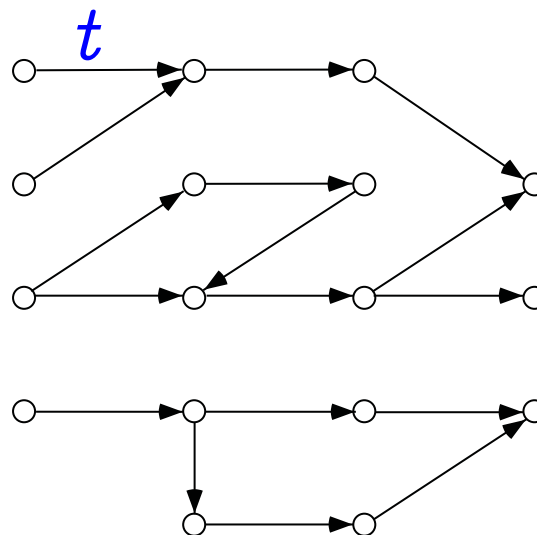
```
11 : ...  
11 : x = -1  
11 : x = 0  
11 : x = 1  
11 : ...  
13 : x = 2 → 14 : x = 3 ... → 14 : x = 10 → 15 : x = 10
```

Diagram illustrating the trace semantics of a loop. A vertical list of states (labeled 11) is shown on the left, with a large blue bracket on its right side. Two arrows originate from the bracket: one points down to the state `12 : x = 1` and another points up to the state `14 : x = 2`. A horizontal sequence of states follows: `12 : x = 1 → 13 : x = 1 → 14 : x = 2 →`



Transition systems

- $\langle S, \xrightarrow{t} \rangle$ where:
 - S is a set of states/vertices/...
 - $\xrightarrow{t} \in \wp(S \times S)$ is a transition relation/set of arcs/...



Collecting semantics in fixpoint form

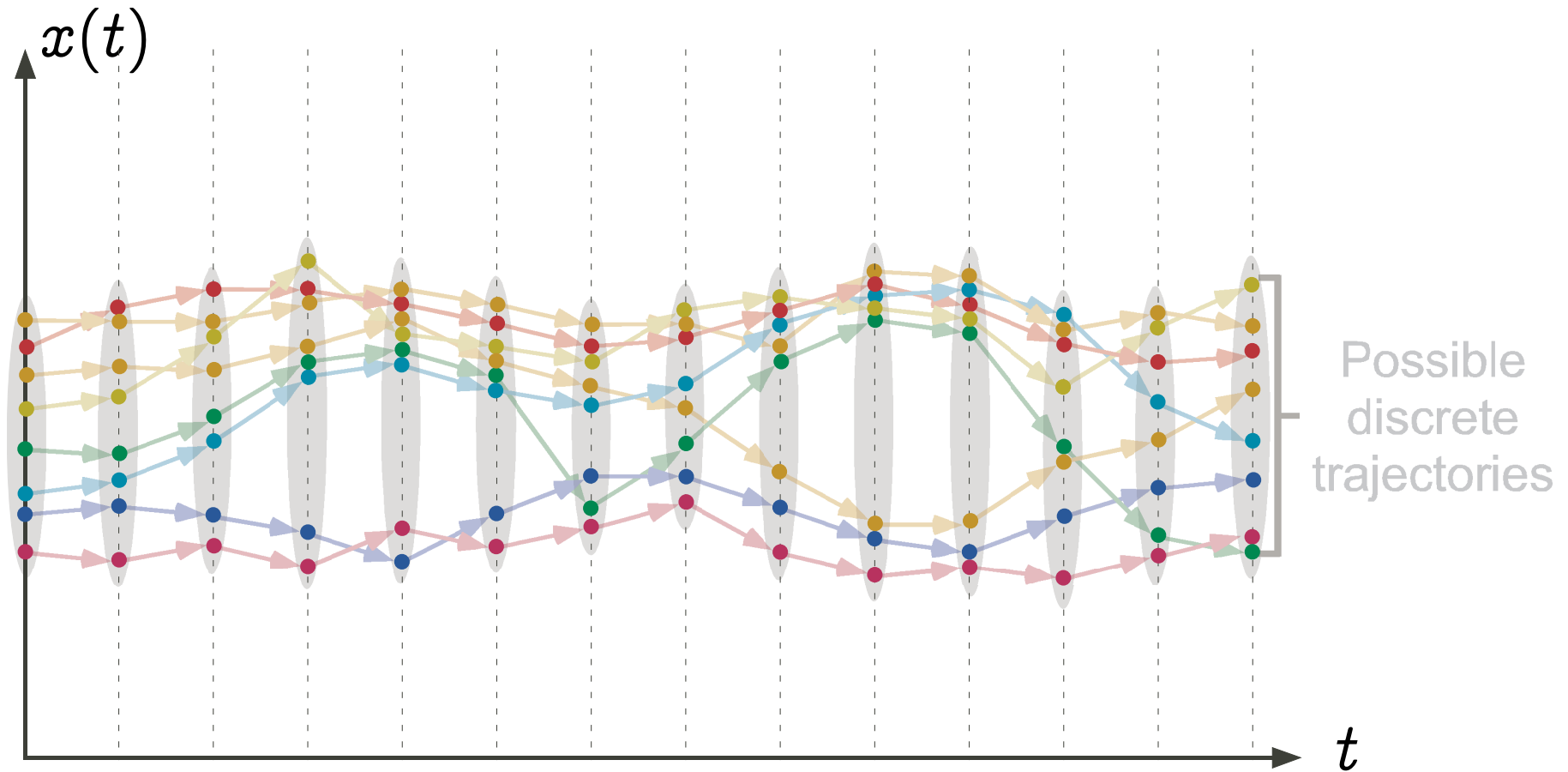


Collecting semantics

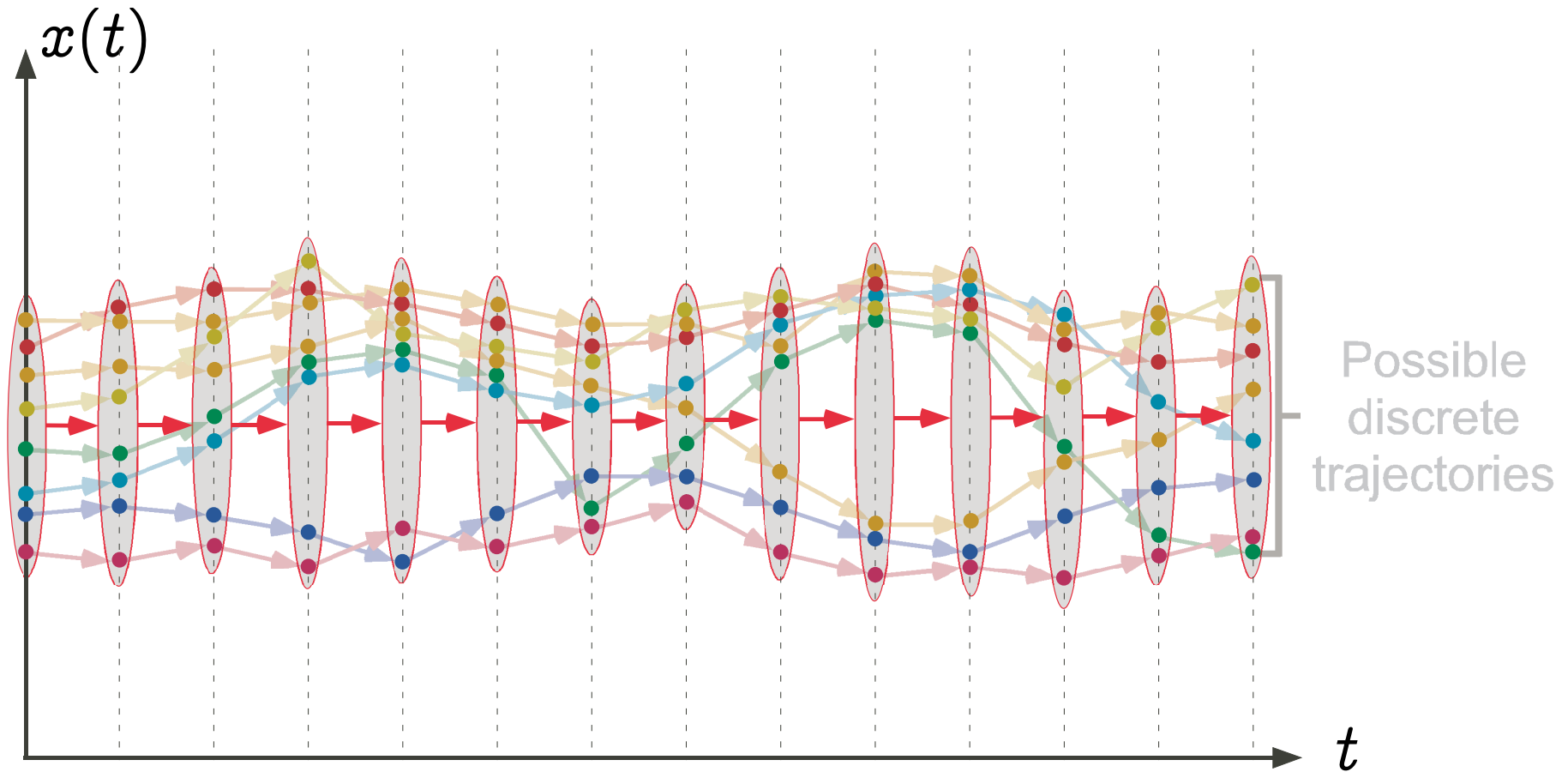
- consider all traces simultaneously;
- collecting semantics:
 - sets of states that describe data values of program variables on all possible trajectories;
 - set of states transitions that is simultaneous elementary computation steps on all possible trajectories;



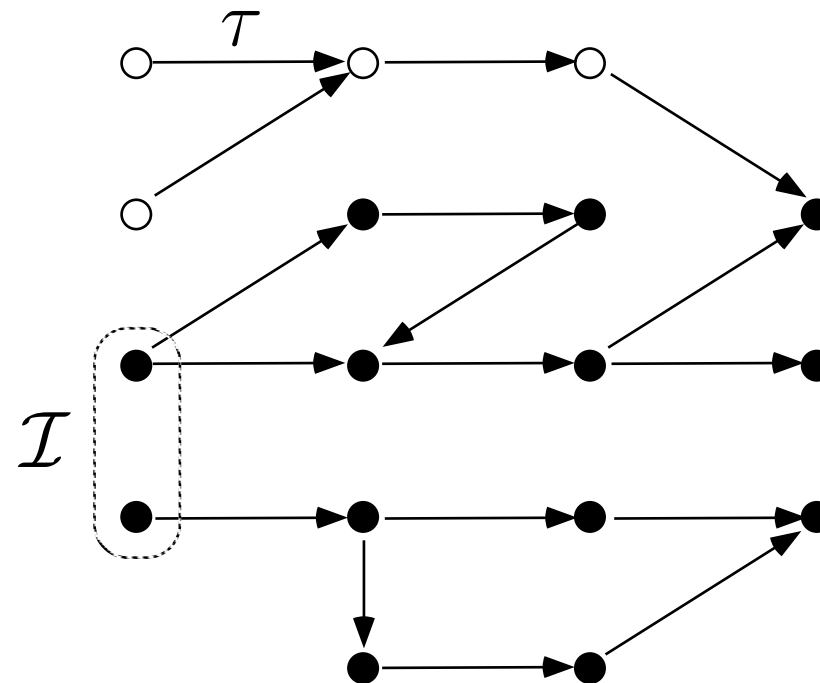
Graphic example: sets of states



Graphic example: set of states transitions



Example: Reachable states of a transition system



Reachable states in fixpoint form

$$F(X) = \mathcal{I} \cup \{s' \mid \exists s \in X : s \xrightarrow{t} s'\}$$

$$\mathcal{R} = \text{lfp}_{\emptyset}^{\subseteq} F$$

$$= \bigcup_{n=0}^{+\infty} F^n(\emptyset)$$

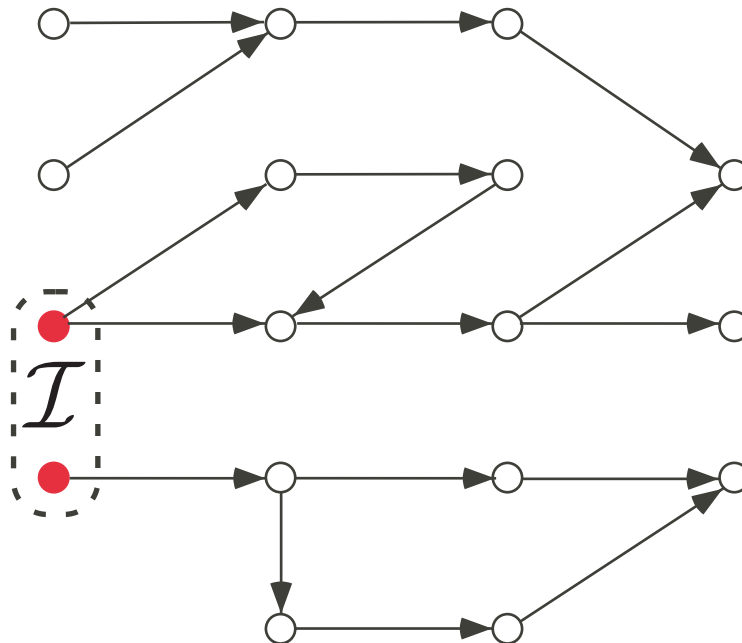
where

$$\begin{aligned} f^0(x) &= x \\ f^{n+1}(x) &= f(f^n(x)) \end{aligned}$$



Example of fixpoint iteration

for reachable states $\text{lfp}_{\emptyset}^{\subseteq} \lambda X. \mathcal{I} \cup \{s' \mid \exists s \in X : s \xrightarrow{t} s'\}$

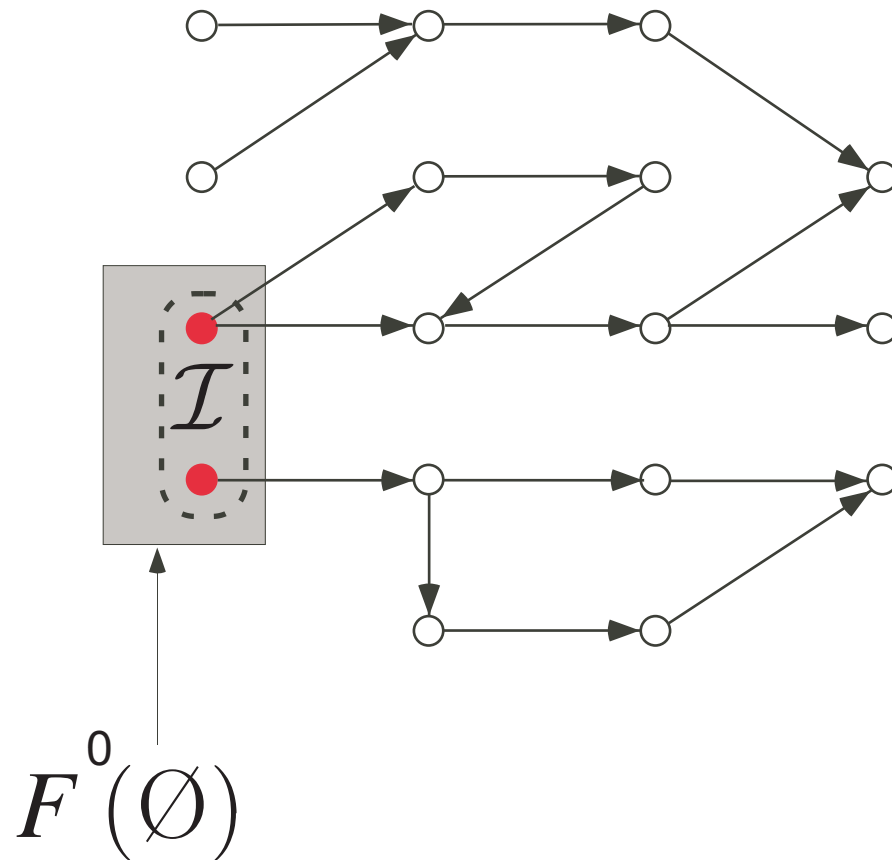


\emptyset



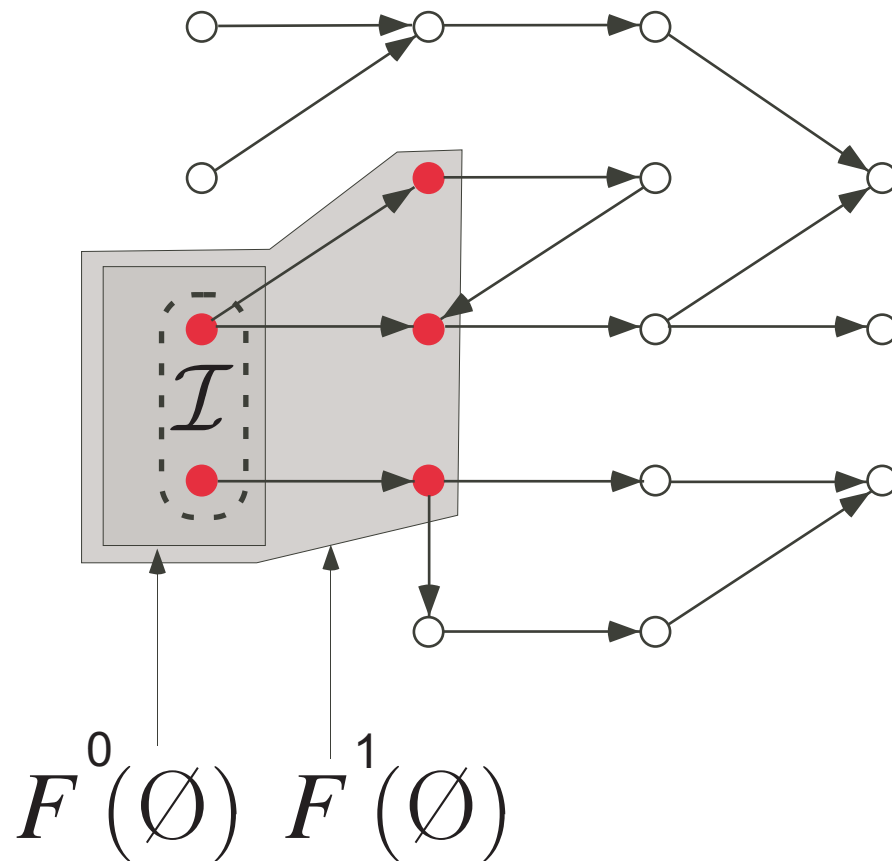
Example of fixpoint iteration

for reachable states $\text{lfp}_{\emptyset}^{\subseteq} \lambda X. \mathcal{I} \cup \{s' \mid \exists s \in X : s \xrightarrow{t} s'\}$



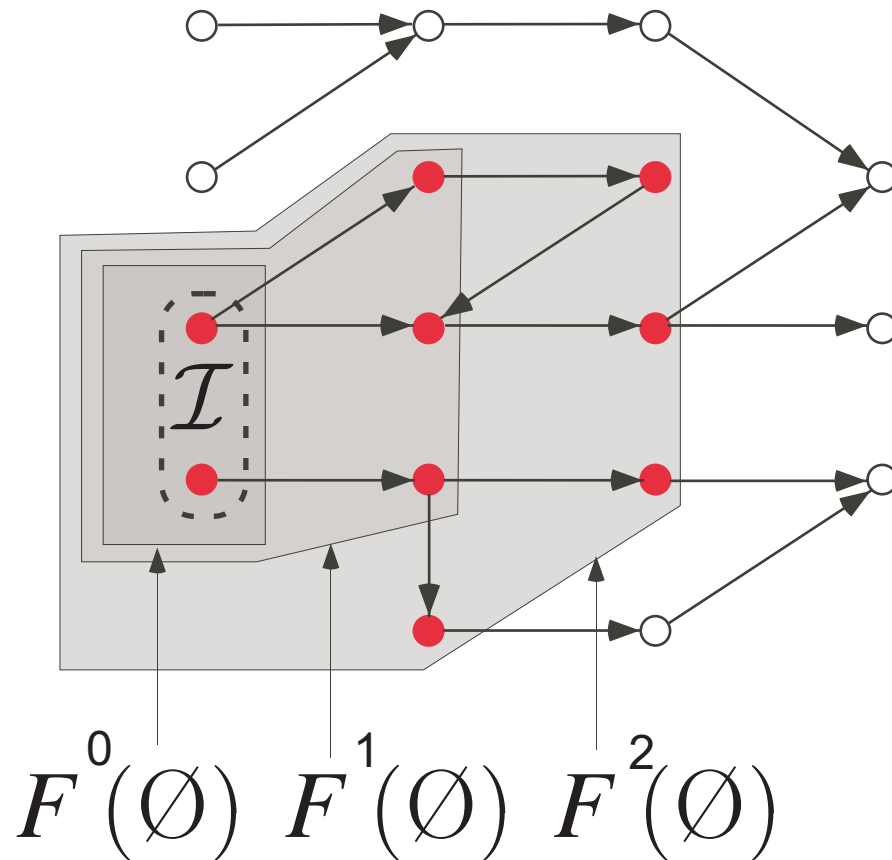
Example of fixpoint iteration

for reachable states $\text{lfp}_{\emptyset}^{\subseteq} \lambda X. \mathcal{I} \cup \{s' \mid \exists s \in X : s \xrightarrow{t} s'\}$



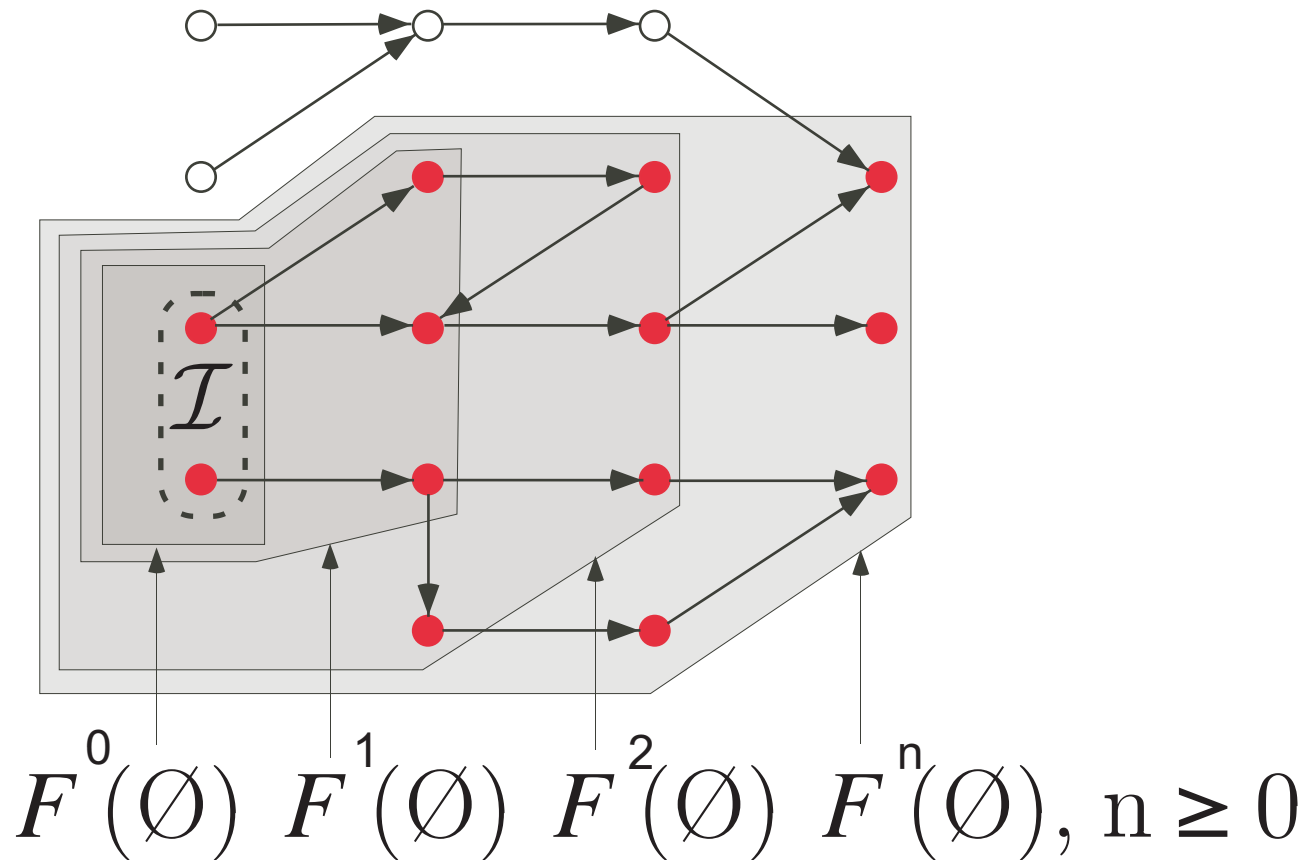
Example of fixpoint iteration

for reachable states $\text{lfp}_{\emptyset}^{\subseteq} \lambda X. \mathcal{I} \cup \{s' \mid \exists s \in X : s \xrightarrow{t} s'\}$



Example of fixpoint iteration

for reachable states $\text{lfp}_{\emptyset}^{\subseteq} \lambda X. \mathcal{I} \cup \{s' \mid \exists s \in X : s \xrightarrow{t} s'\}$



Abstraction by Galois connections

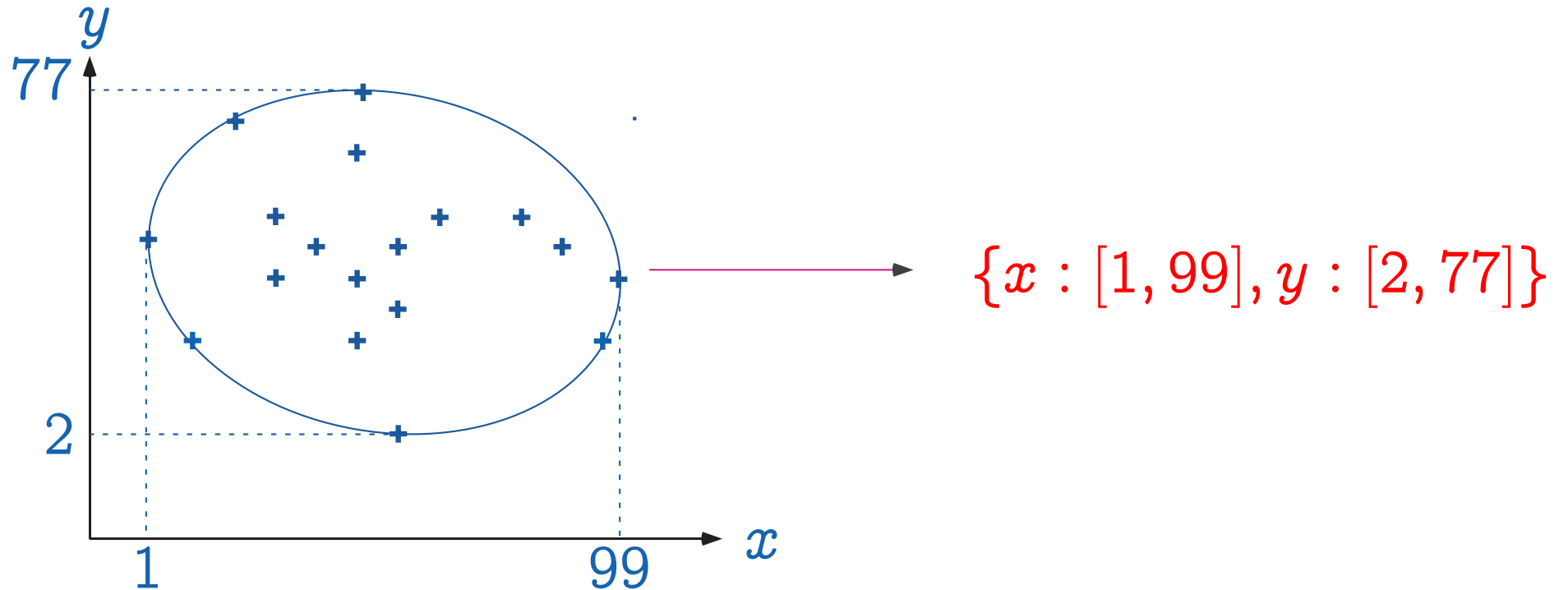


Abstracting sets (i.e. properties)

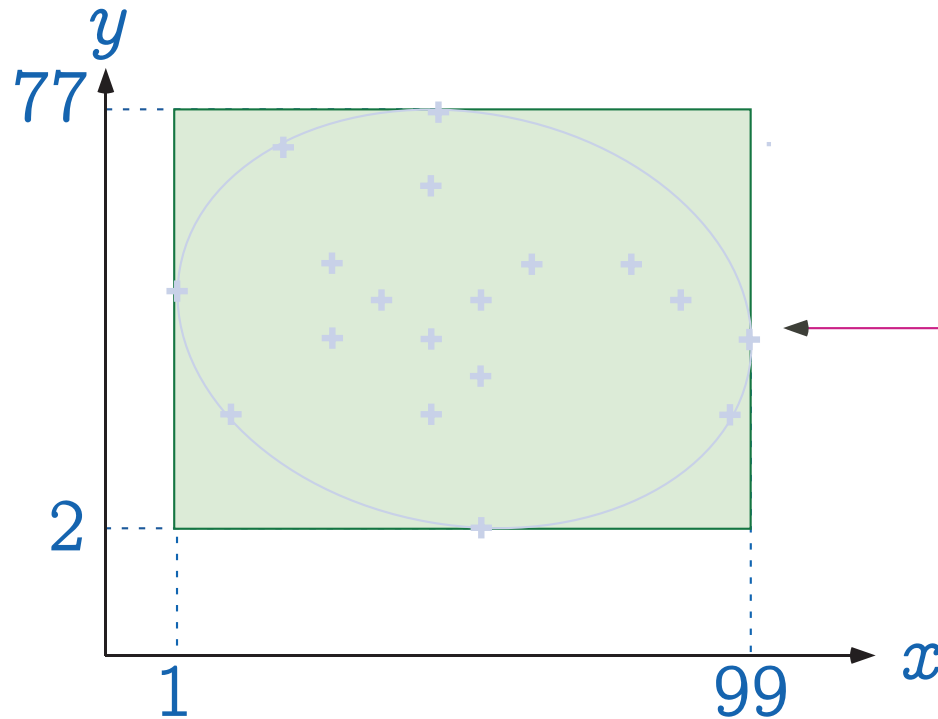
- Choose an **abstract domain**, replacing sets of objects (states, traces, ...) S by their abstraction $\alpha(S)$
- The **abstraction function** α maps a set of concrete objects to its abstract interpretation;
- The inverse **concretization function** γ maps an abstract set of objects to concrete ones;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



Interval abstraction α



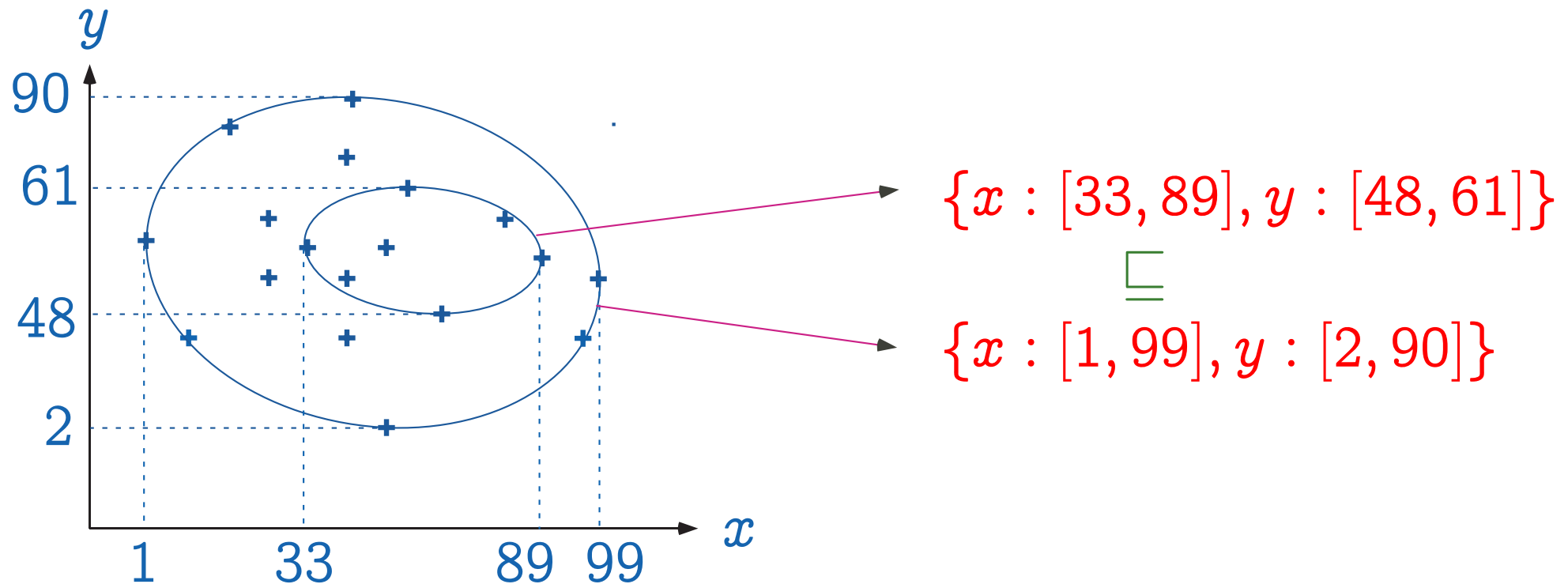
Interval concretization γ



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



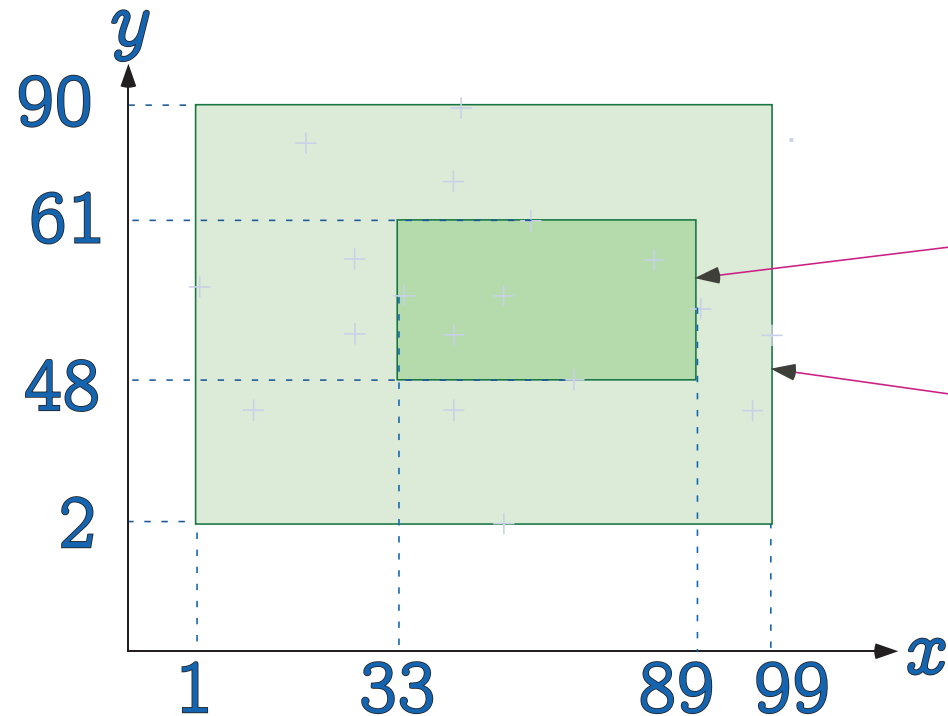
The abstraction α is monotone



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



The concretization γ is monotone



$$\{x : [33, 89], y : [48, 61]\}$$

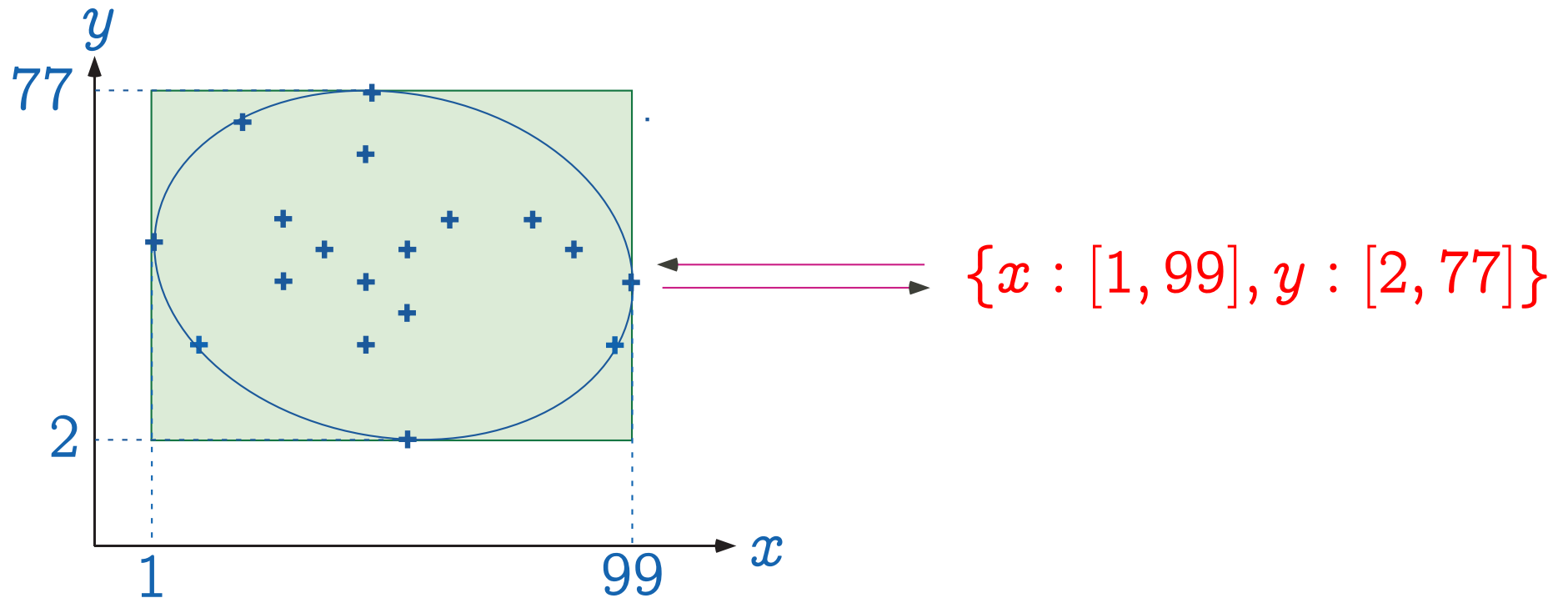
$$\subseteq$$

$$\{x : [1, 99], y : [2, 90]\}$$

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



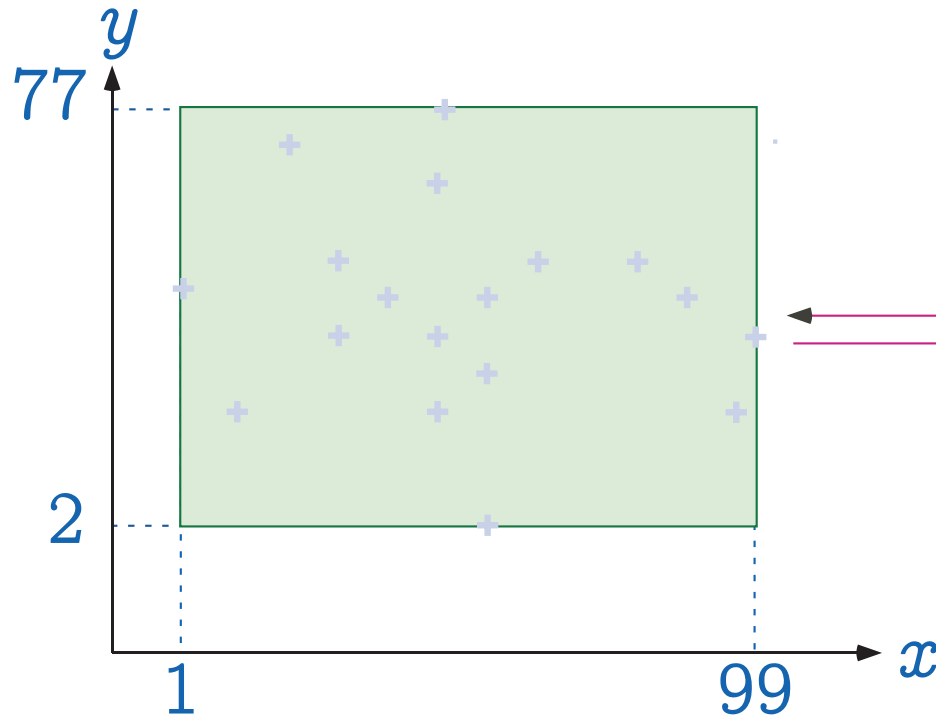
The $\gamma \circ \alpha$ composition is extensive



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



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



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

$$\alpha \circ \gamma(Y) =/\sqsubseteq Y$$



Correspondance between concrete and abstract properties

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

$$\langle \wp(S), \subseteq \rangle \begin{matrix} \xleftarrow{\gamma} \\ \xrightarrow{\alpha} \end{matrix} \langle \mathcal{D}, \sqsubseteq \rangle$$

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



Galois connection

$$\langle \mathcal{D}, \subseteq \rangle \begin{array}{c} \xleftarrow{\gamma} \\ \xrightarrow{\alpha} \end{array} \langle \overline{\mathcal{D}}, \sqsubseteq \rangle$$

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

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

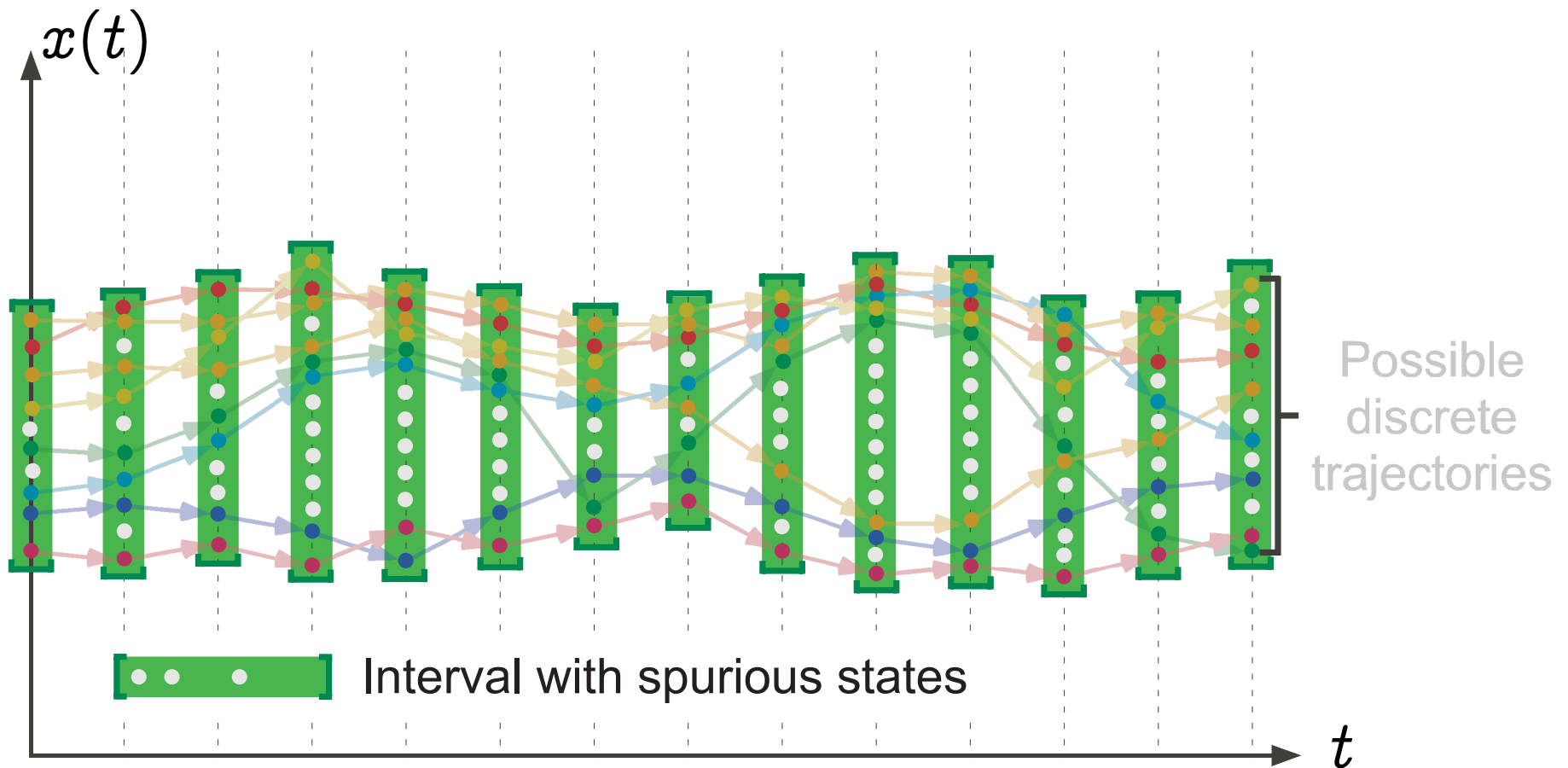
$\wedge \forall x \in \mathcal{D} : x \subseteq \gamma(\alpha(x))$

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

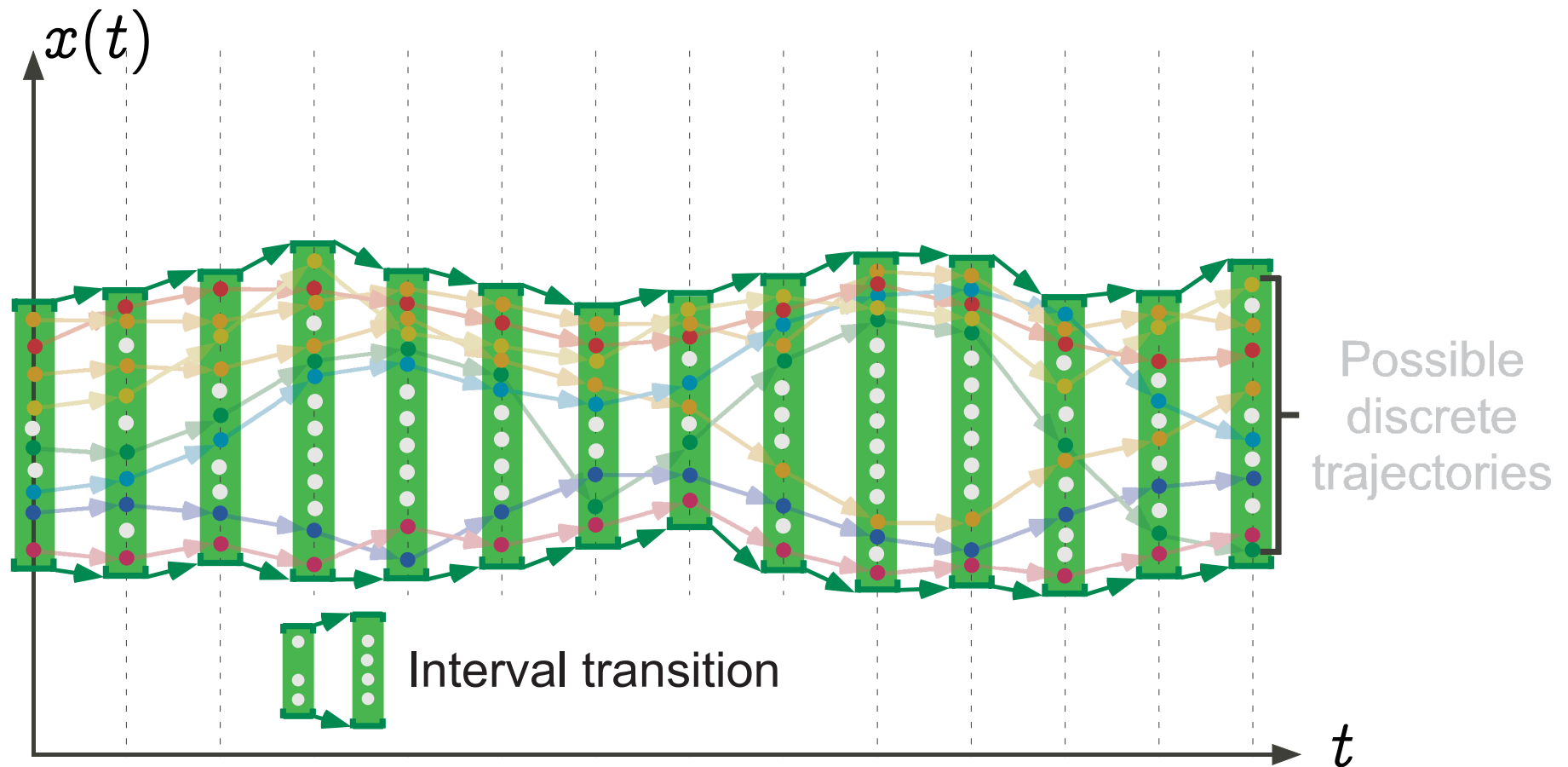
iff $\forall x \in \mathcal{D}, \bar{y} \in \overline{\mathcal{D}} : \alpha(x) \sqsubseteq \bar{y} \iff x \subseteq \gamma(\bar{y})$



Graphic example: Interval abstraction



Graphic example: Abstract transitions



Example: Interval transition semantics of assignments

```
int x;
```

```
...
```

```
l:
```

```
    x := x + 1;
```

```
l':
```

$$\{l : x \in [\ell, h] \rightarrow l' : x \in [l + 1, \min(h + 1, \text{max_int})] \cup \{\Omega \mid h = \text{max_int}\} \mid \ell \leq h\}$$

where $[\ell, h] = \emptyset$ when $h < \ell$.



Abstract domain



Concrete domain



Function abstraction

$$F^\# = \alpha \circ F \circ \gamma$$

$$\text{i.e. } F^\# = \rho \circ F$$

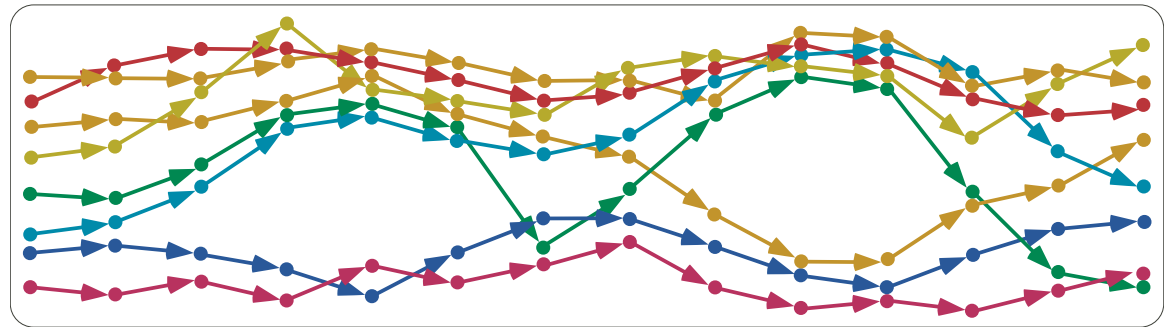
$$\langle P, \subseteq \rangle \begin{matrix} \xleftarrow{\gamma} \\ \xrightarrow{\alpha} \end{matrix} \langle Q, \sqsubseteq \rangle \Rightarrow$$

$$\langle P \xrightarrow{\text{mon}} P, \dot{\subseteq} \rangle \begin{matrix} \xleftarrow{\lambda F^\# \cdot \gamma \circ F^\# \circ \alpha} \\ \xrightarrow{\lambda F \cdot \alpha \circ F \circ \gamma} \end{matrix} \langle Q \xrightarrow{\text{mon}} Q, \dot{\sqsubseteq} \rangle$$



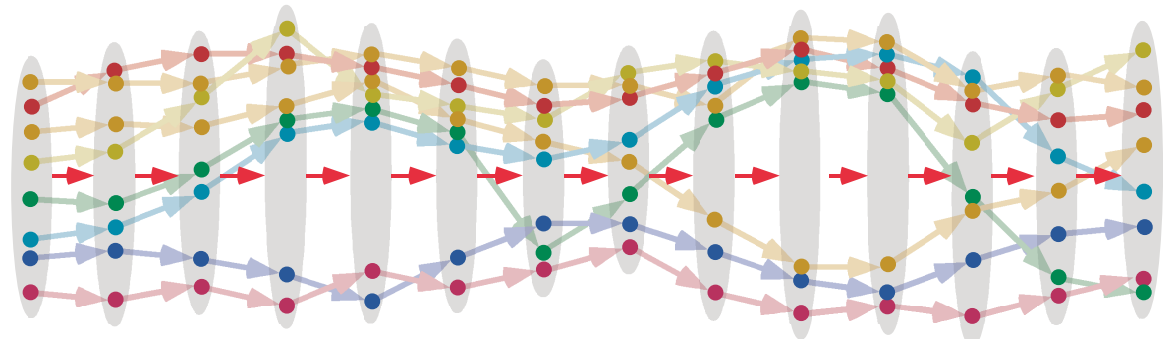
Example: Set of traces to trace of intervals abstraction

Set of traces:



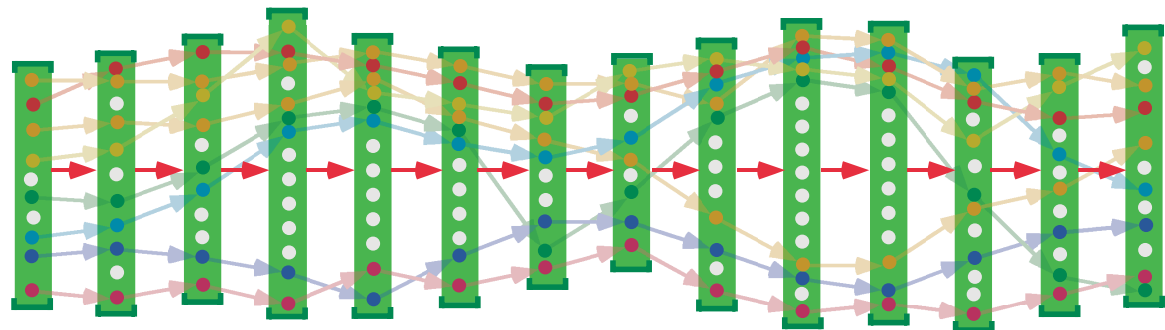
$\alpha_1 \downarrow$

Trace of sets:



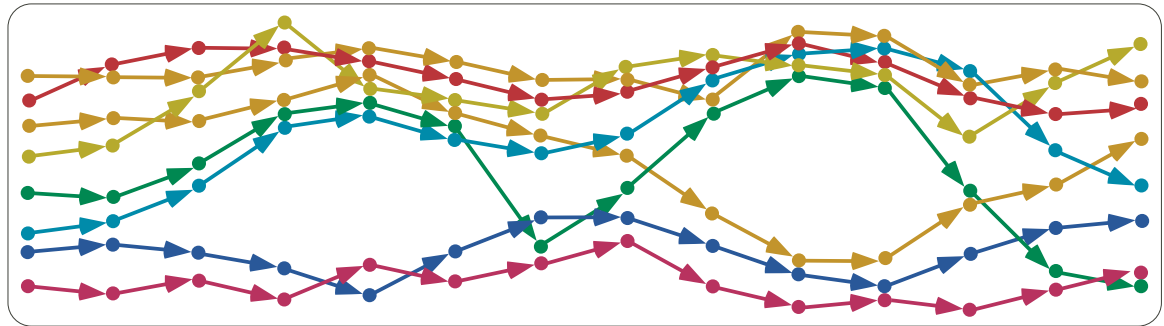
$\alpha_2 \downarrow$

Trace of intervals



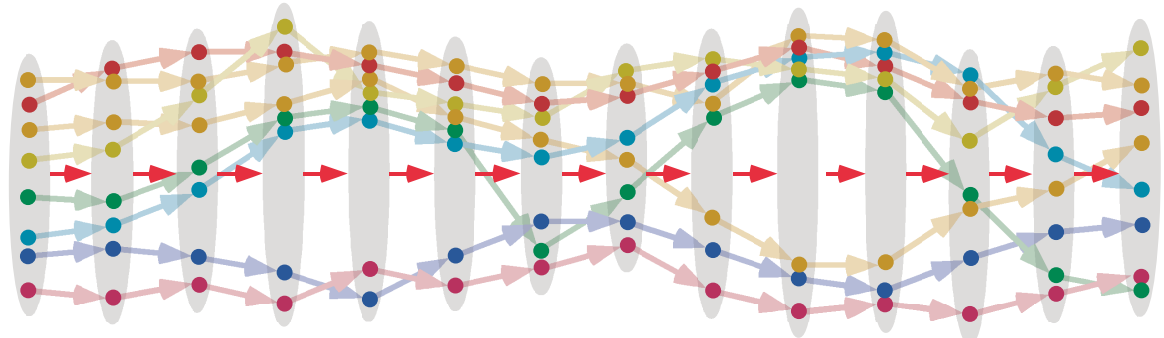
Example: Set of traces to reachable states abstraction

Set of traces:



$\alpha_1 \downarrow$

Trace of sets:



$\alpha_3 \downarrow$

Reachable states



Composition of Galois Connections

The composition of Galois connections:

$$\langle L, \leq \rangle \begin{array}{c} \xleftarrow{\gamma_1} \\ \xrightarrow{\alpha_1} \end{array} \langle M, \sqsubseteq \rangle$$

and:

$$\langle M, \sqsubseteq \rangle \begin{array}{c} \xleftarrow{\gamma_2} \\ \xrightarrow{\alpha_2} \end{array} \langle N, \preceq \rangle$$

is a Galois connection:

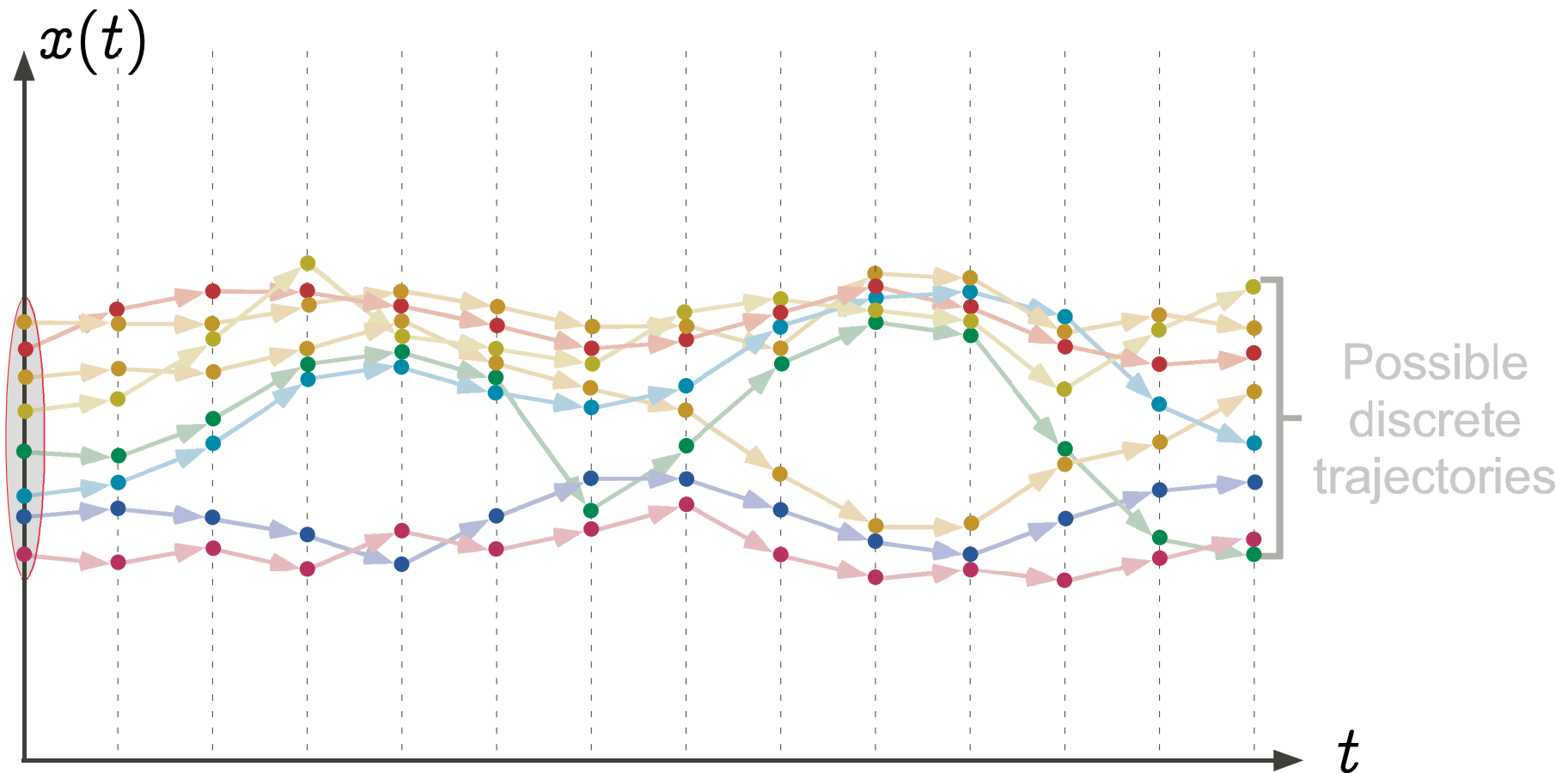
$$\langle L, \leq \rangle \begin{array}{c} \xleftarrow{\gamma_1 \circ \gamma_2} \\ \xrightarrow{\alpha_2 \circ \alpha_1} \end{array} \langle N, \preceq \rangle$$



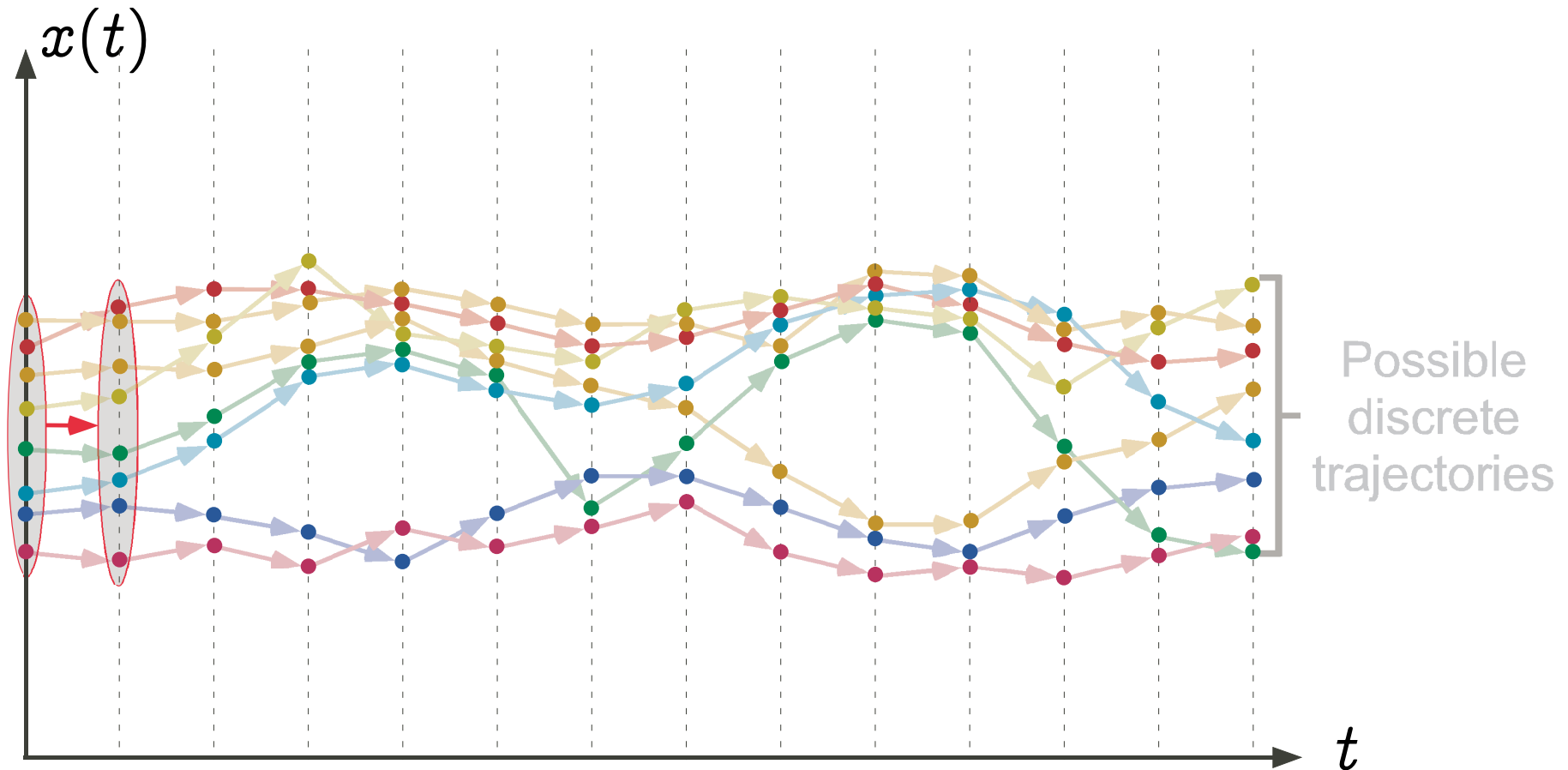
Abstract semantics in fixpoint form



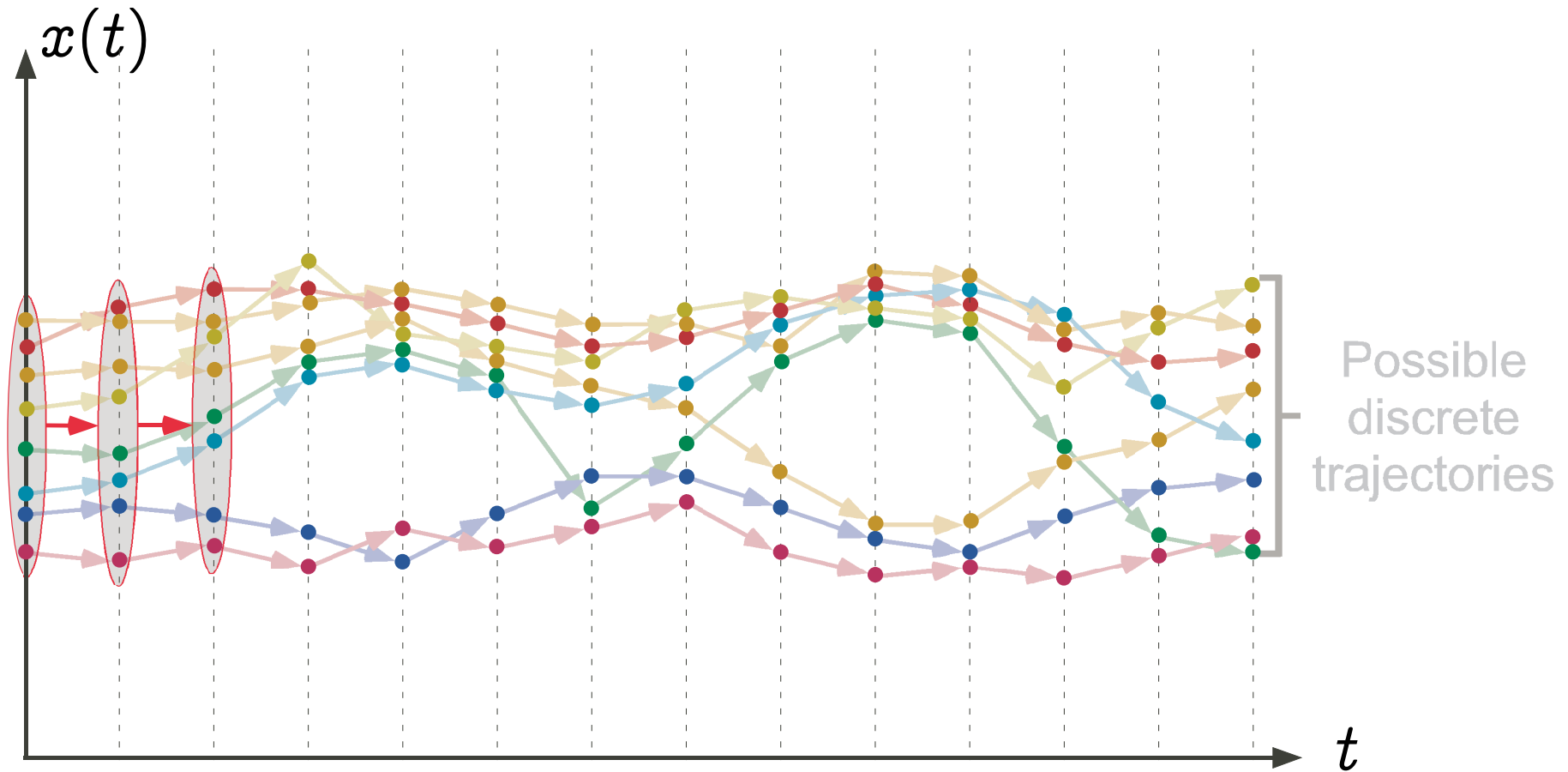
Graphic example: traces of sets of states in fixpoint form



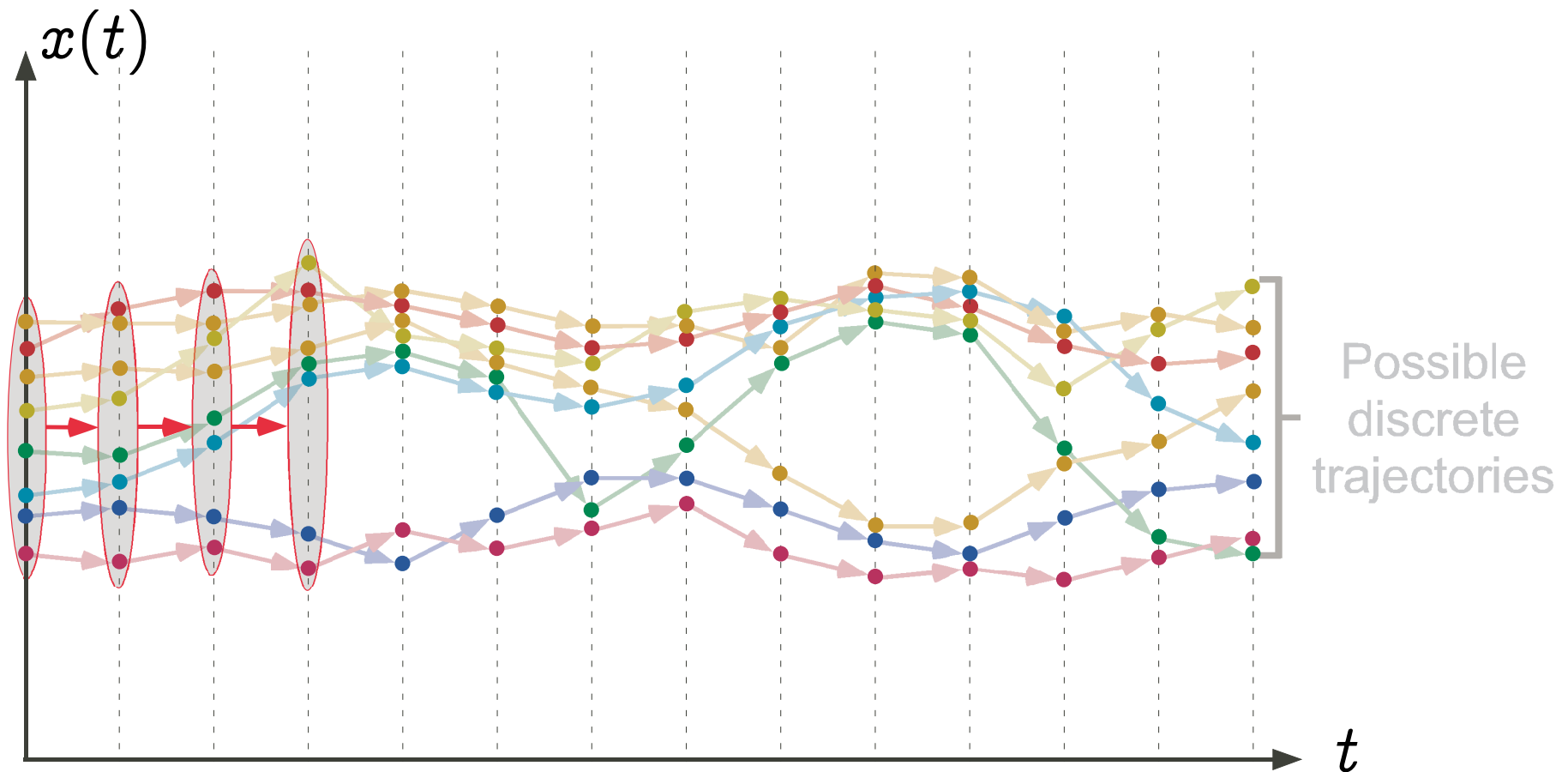
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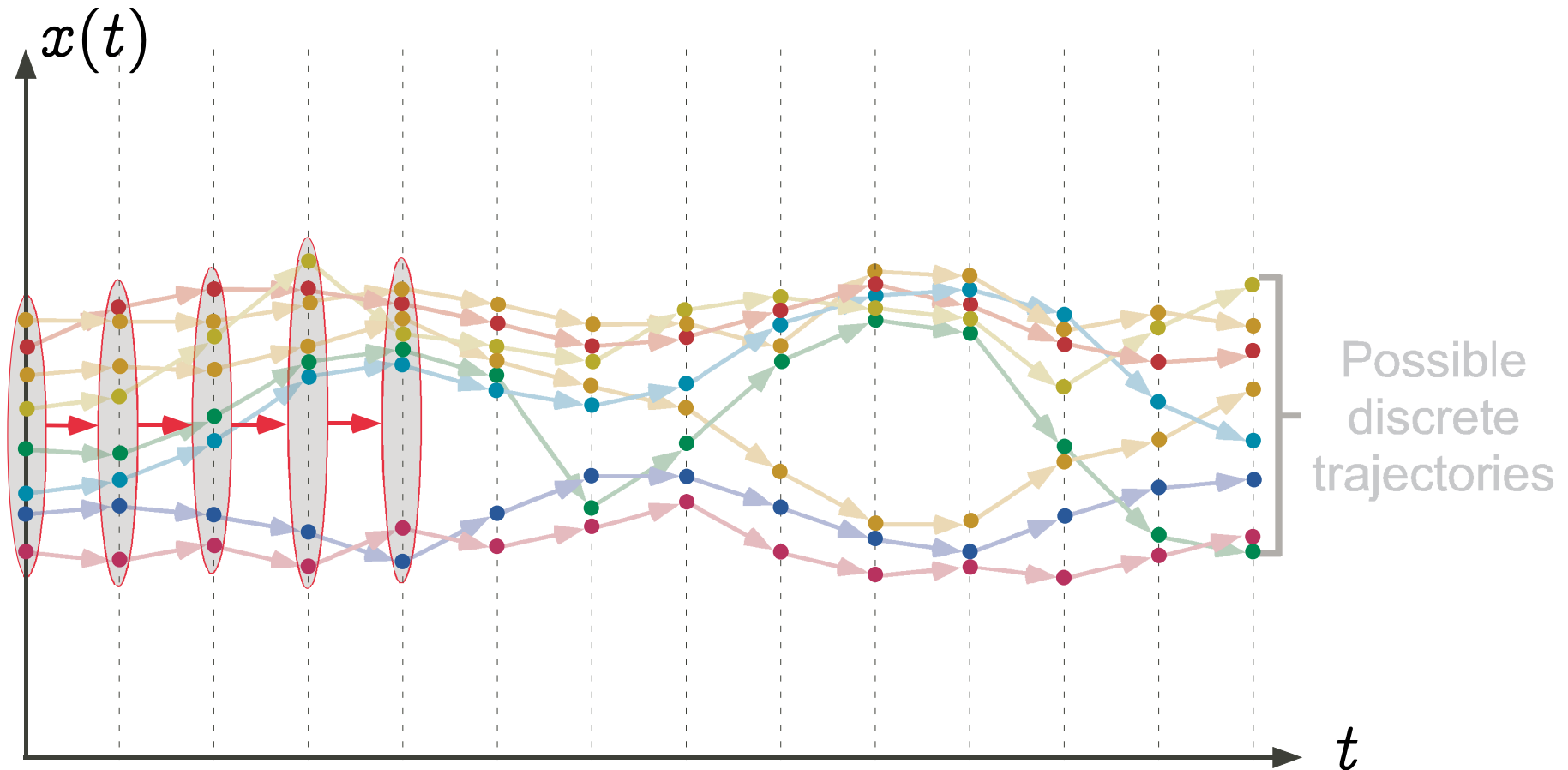
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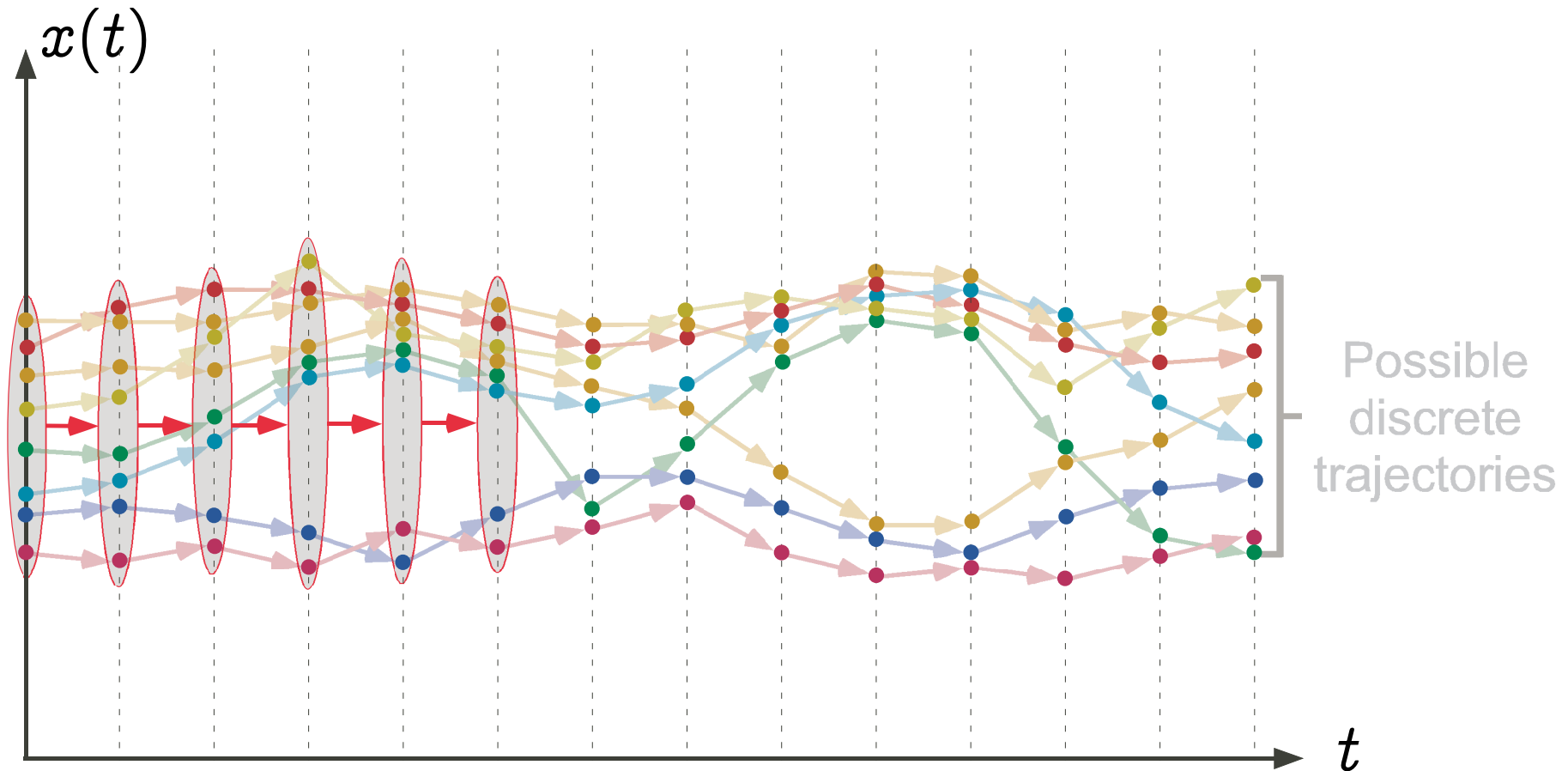
Graphic example: traces of sets of states in fixpoint form



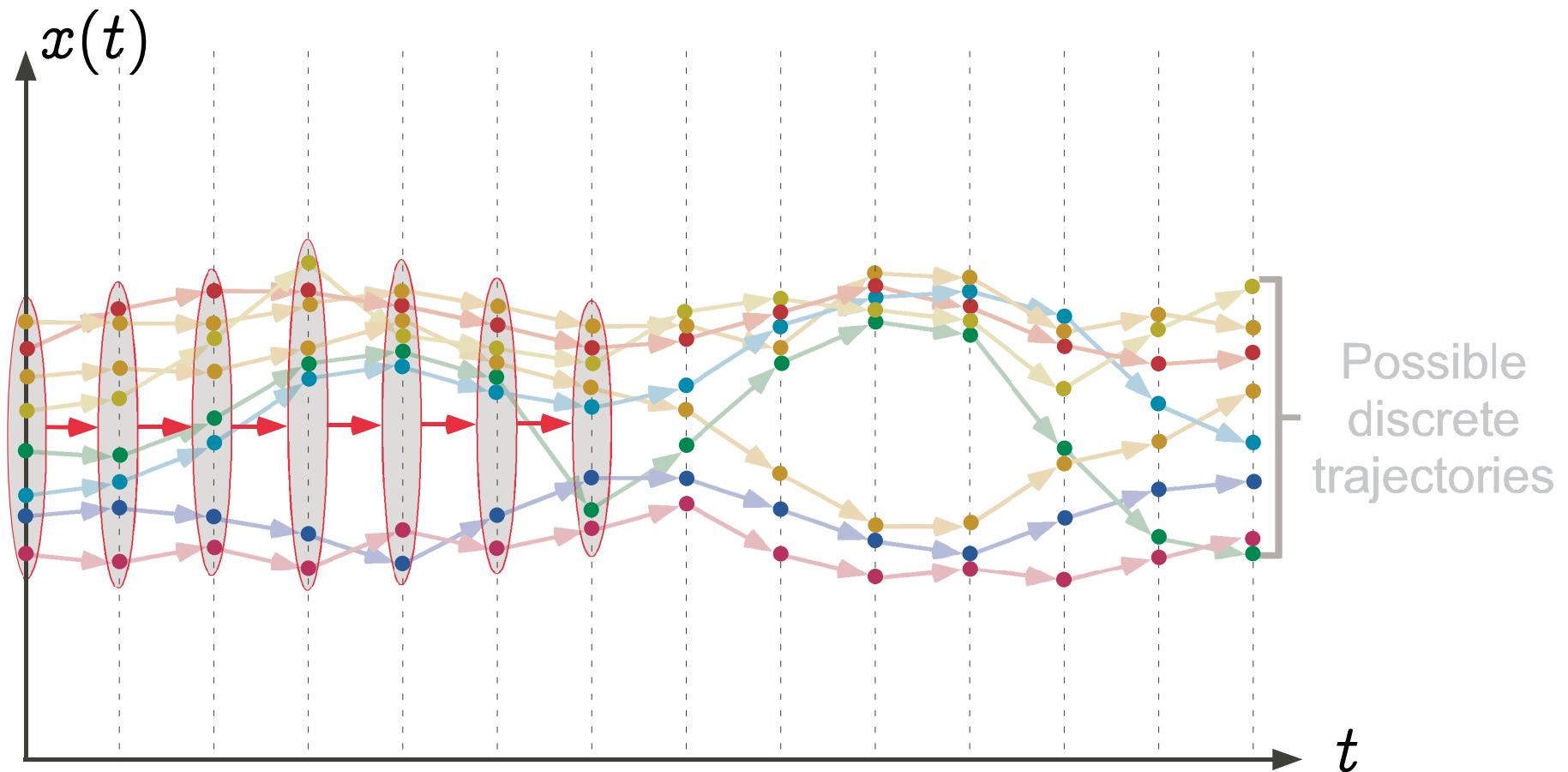
Graphic example: traces of sets of states in fixpoint form



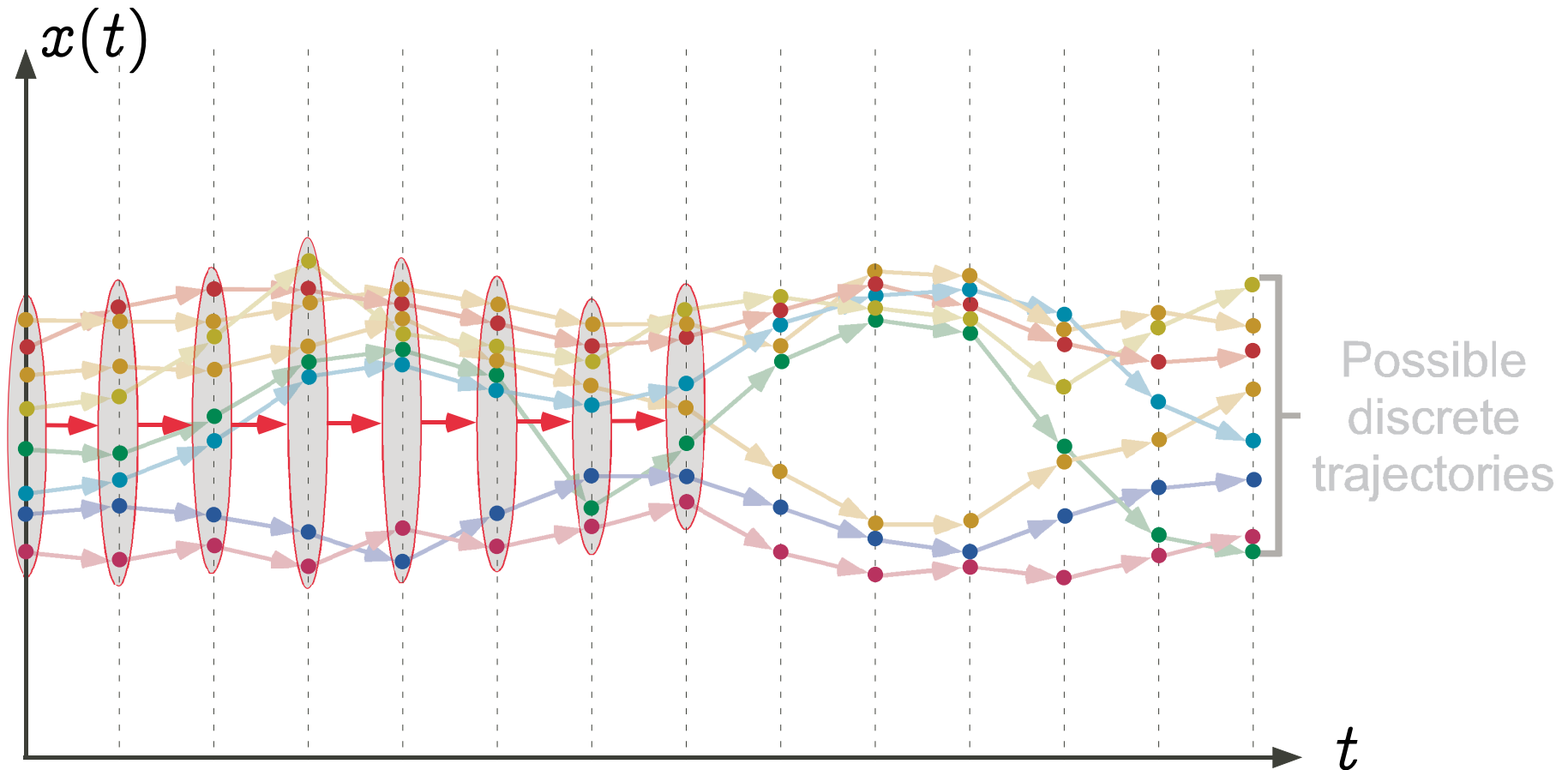
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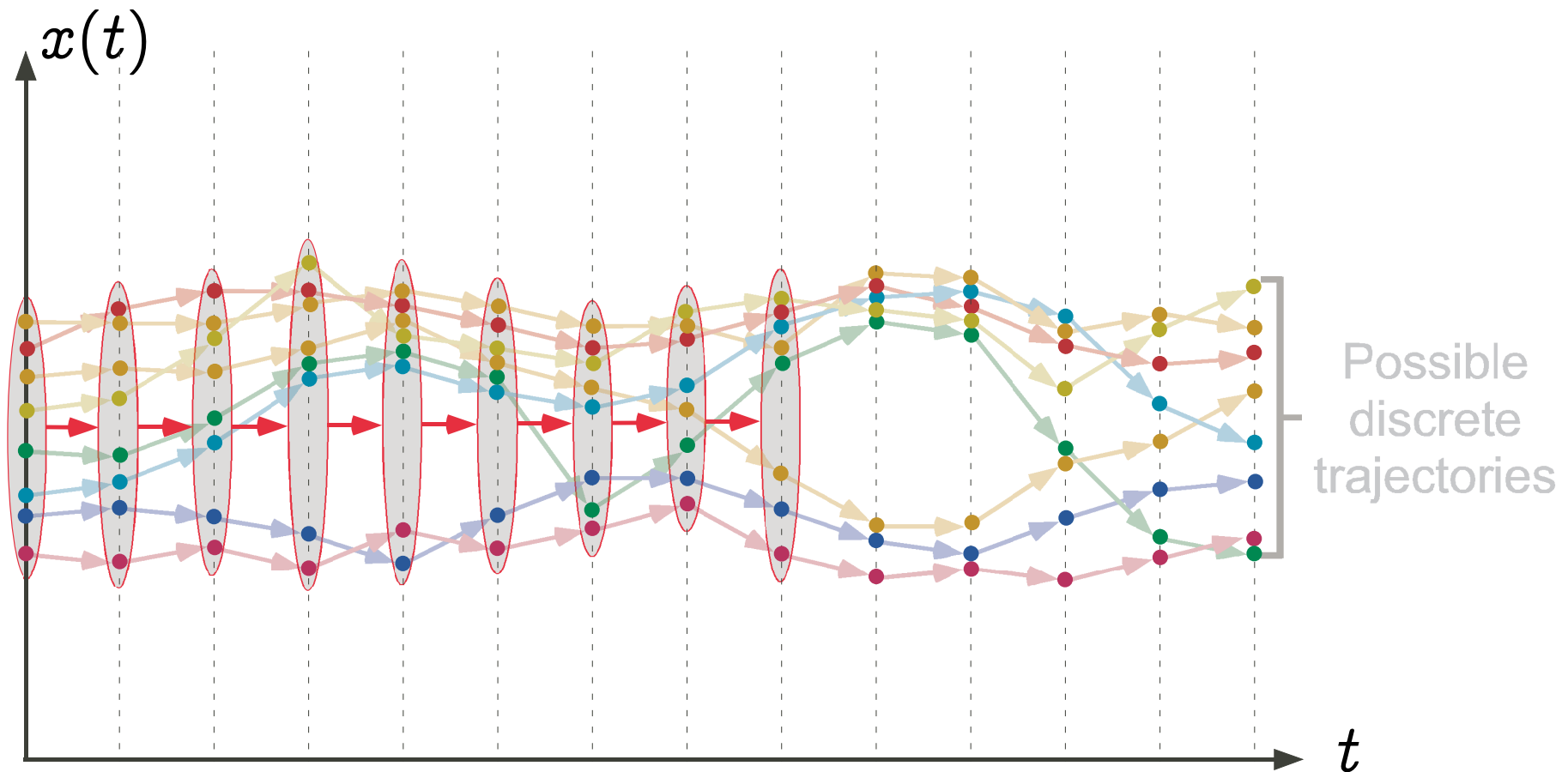
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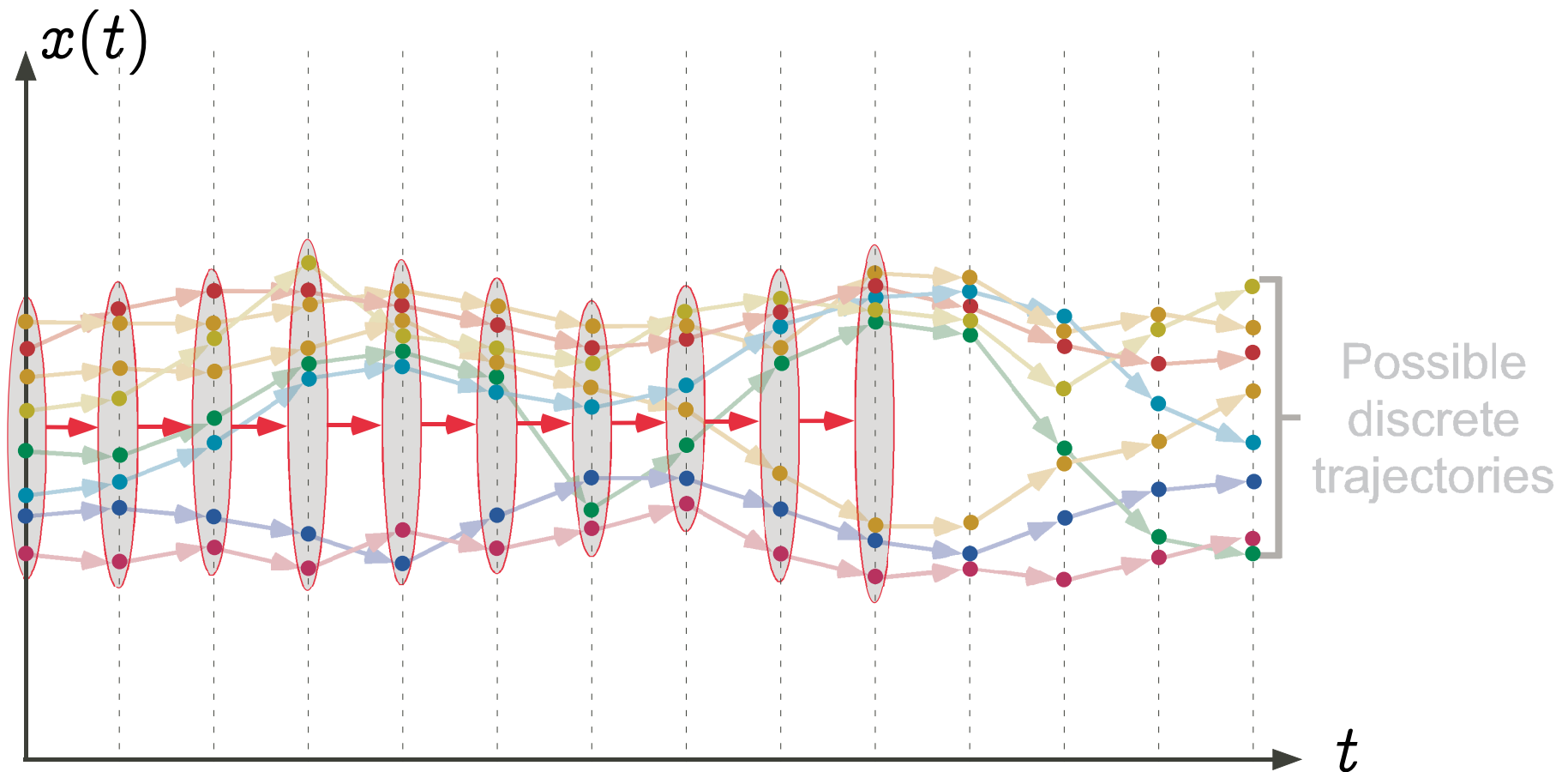
Graphic example: traces of sets of states in fixpoint form



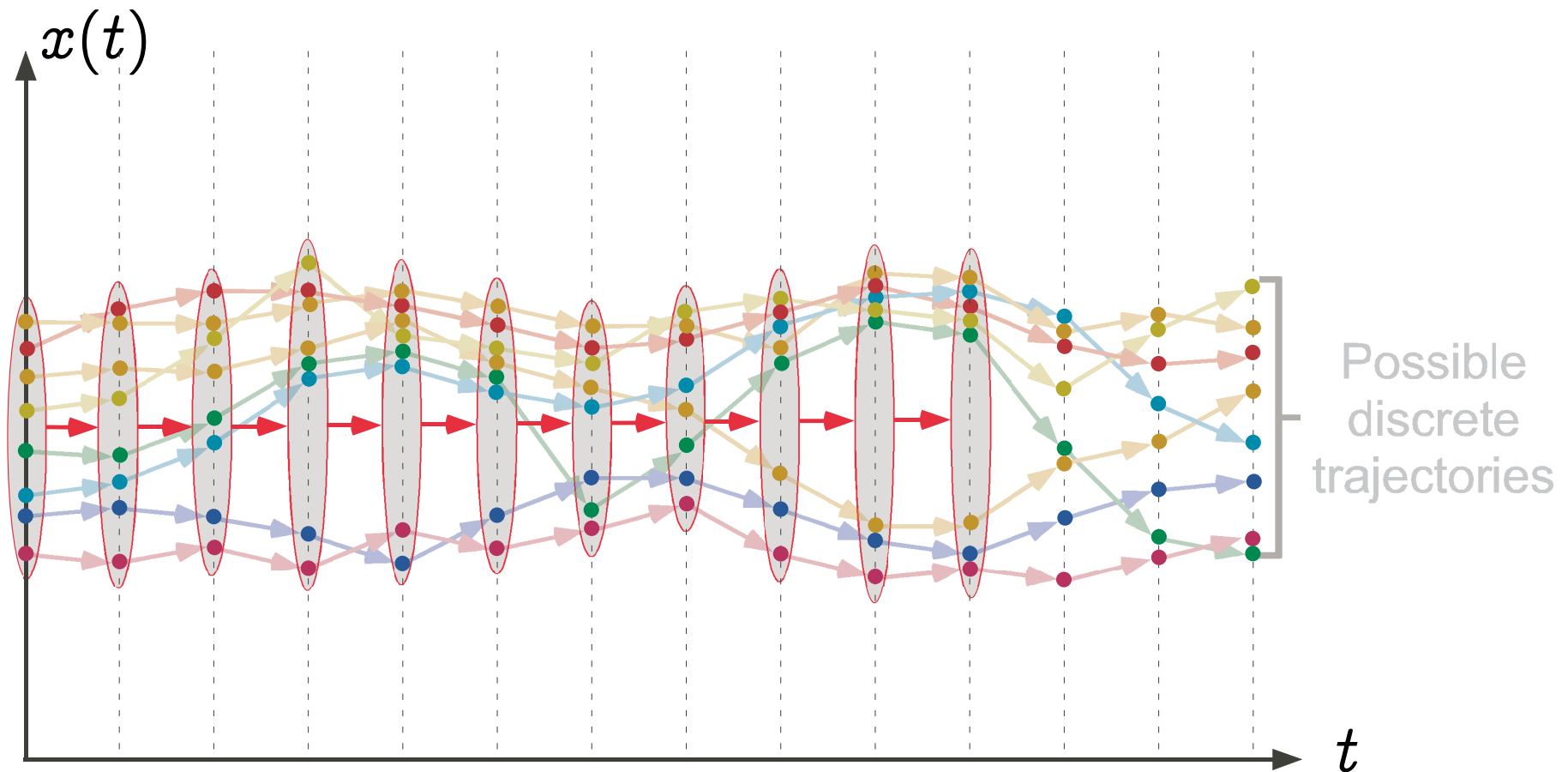
Graphic example: traces of sets of states in fixpoint form



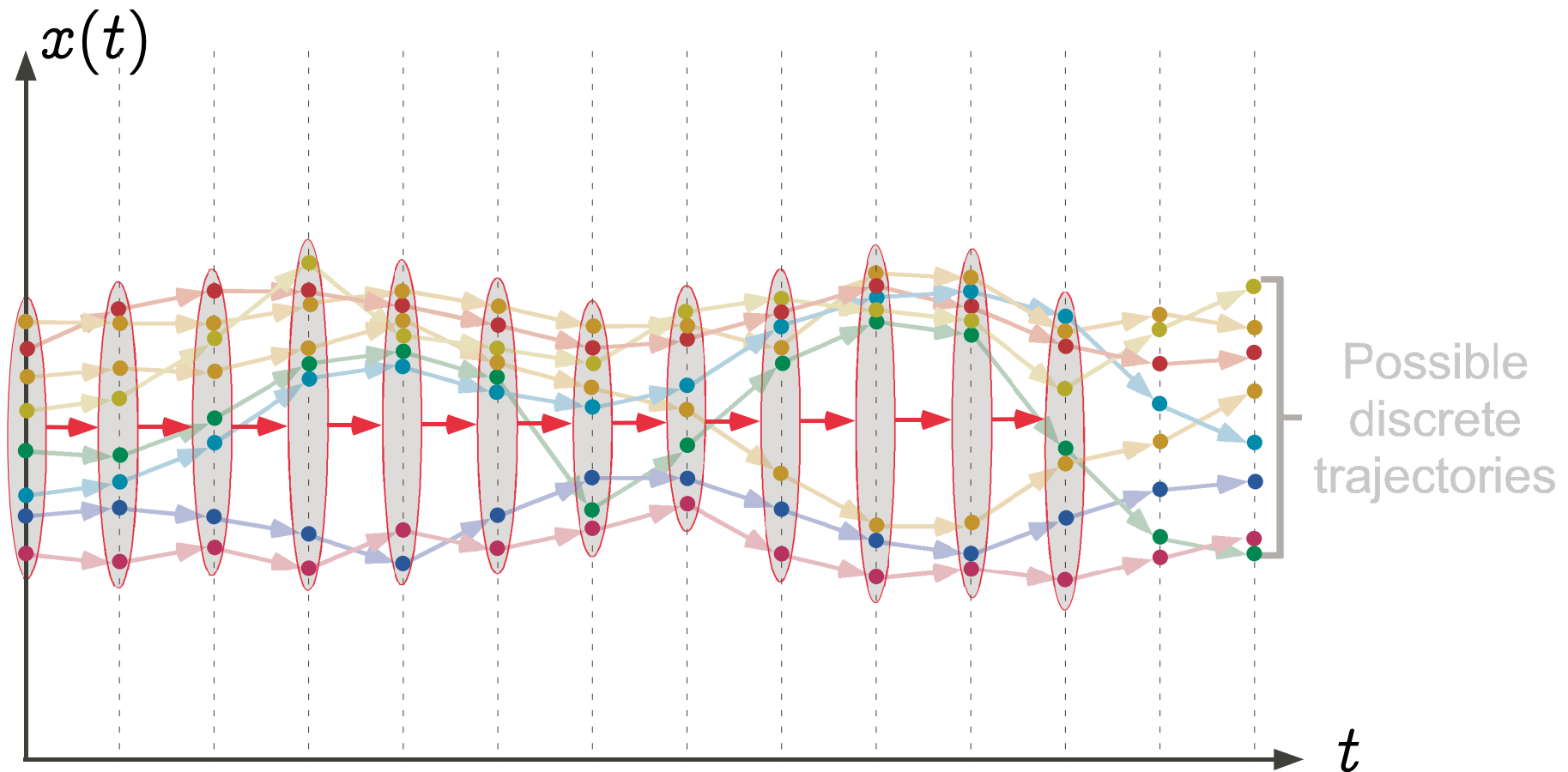
Graphic example: traces of sets of states in fixpoint form



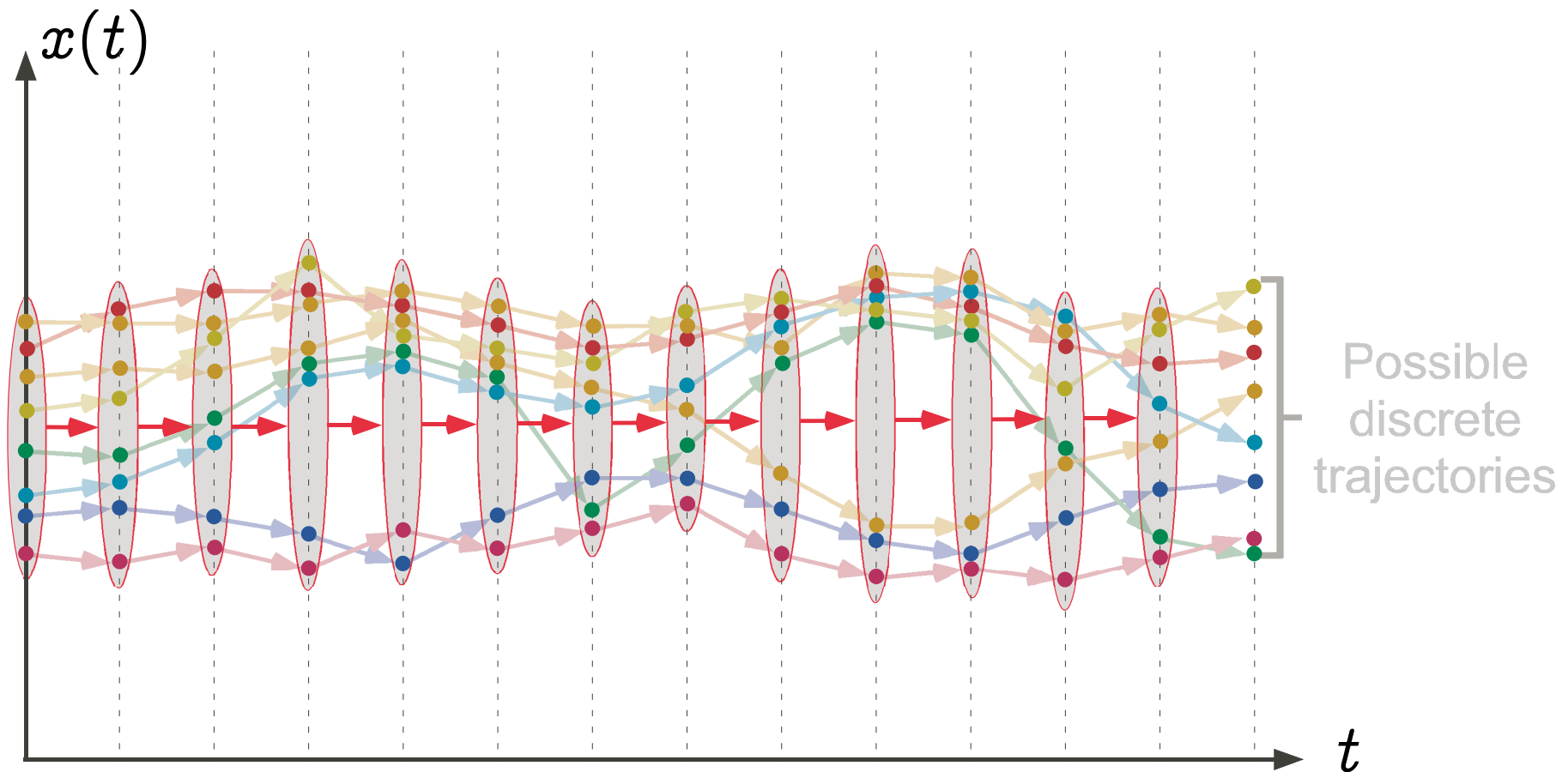
Graphic example: traces of sets of states in fixpoint form



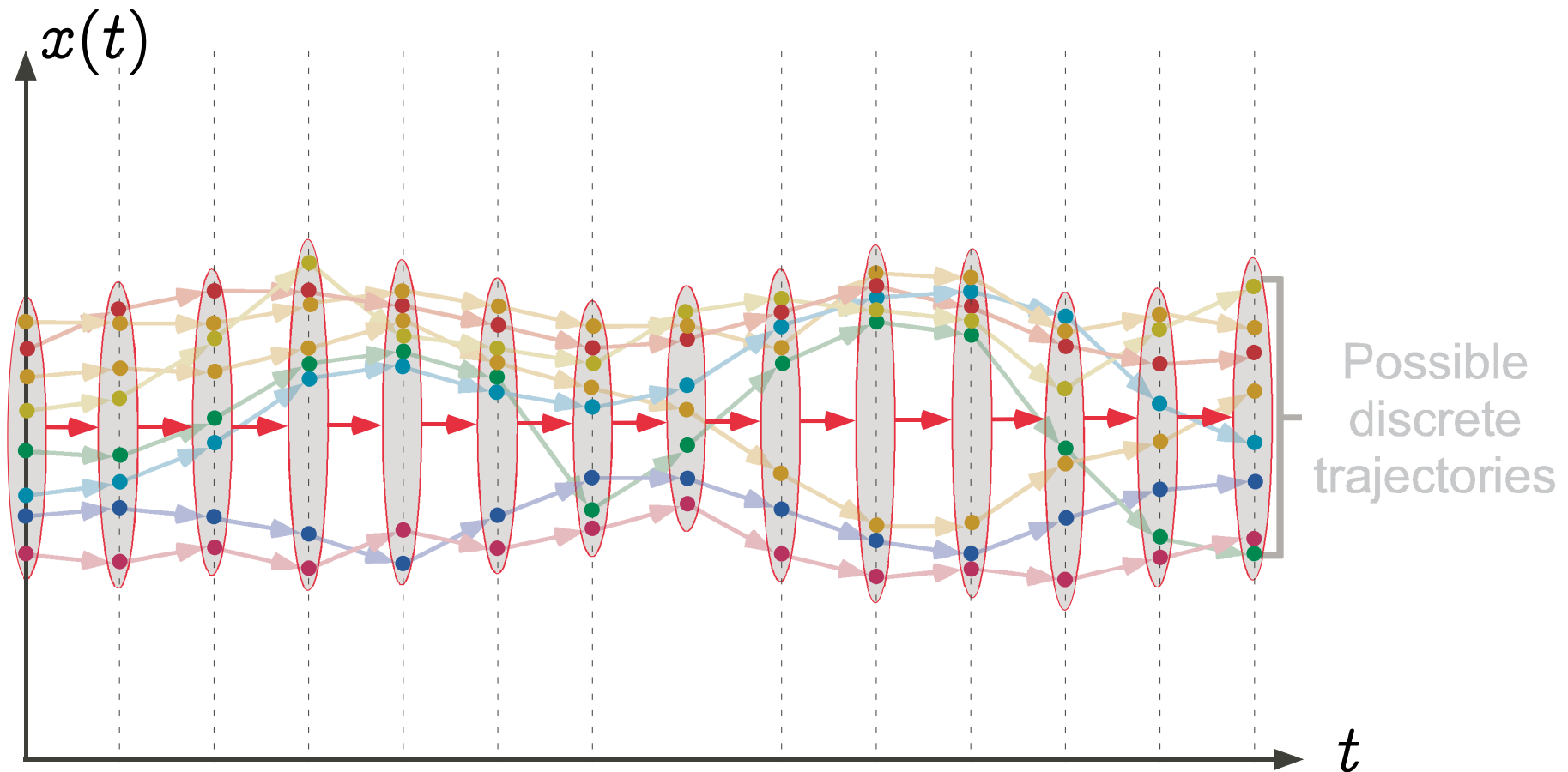
Graphic example: traces of sets of states in fixpoint form



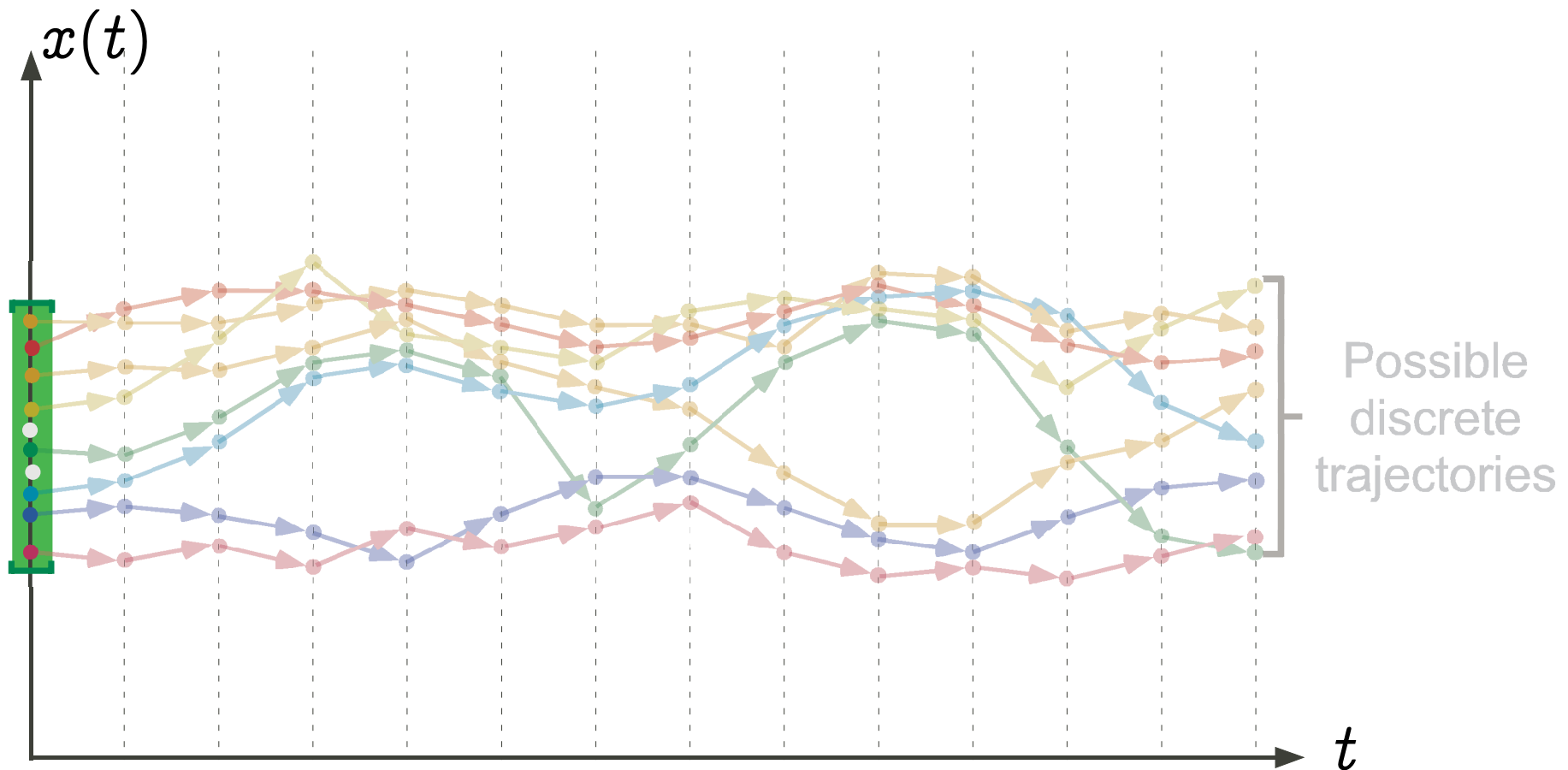
Graphic example: traces of sets of states in fixpoint form



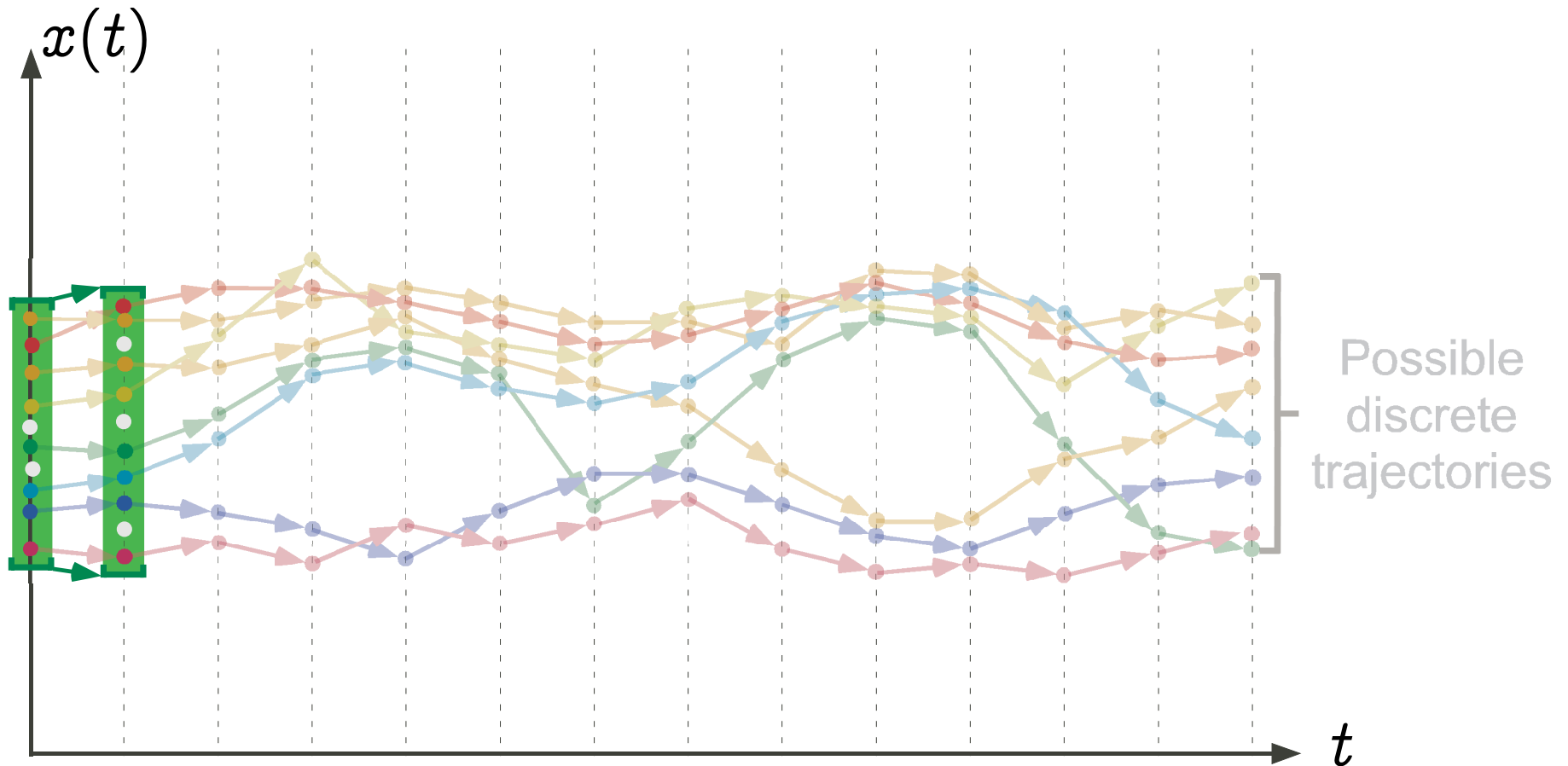
Graphic example: traces of sets of states in fixpoint form



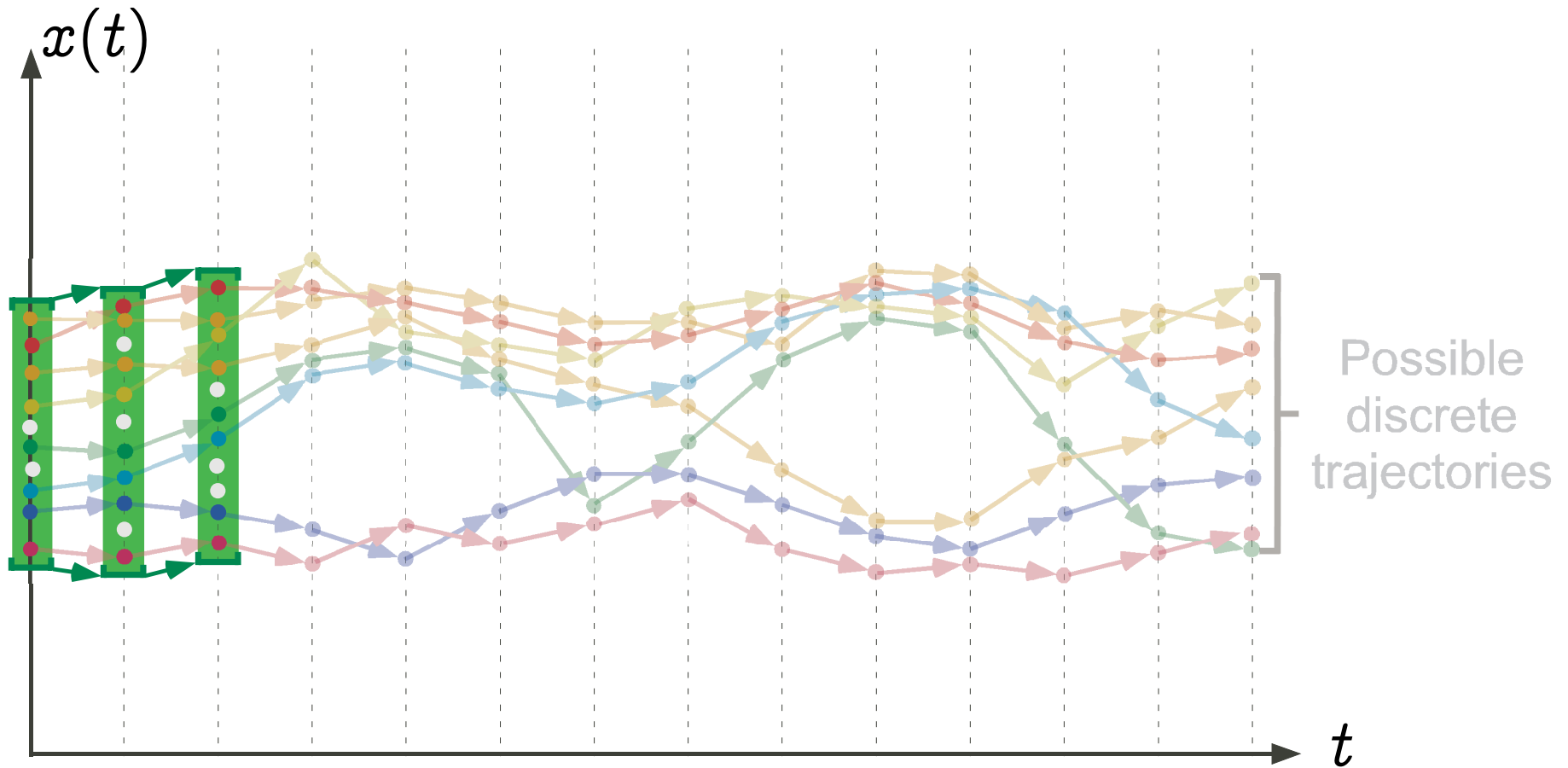
Graphic example: traces of intervals in fixpoint form



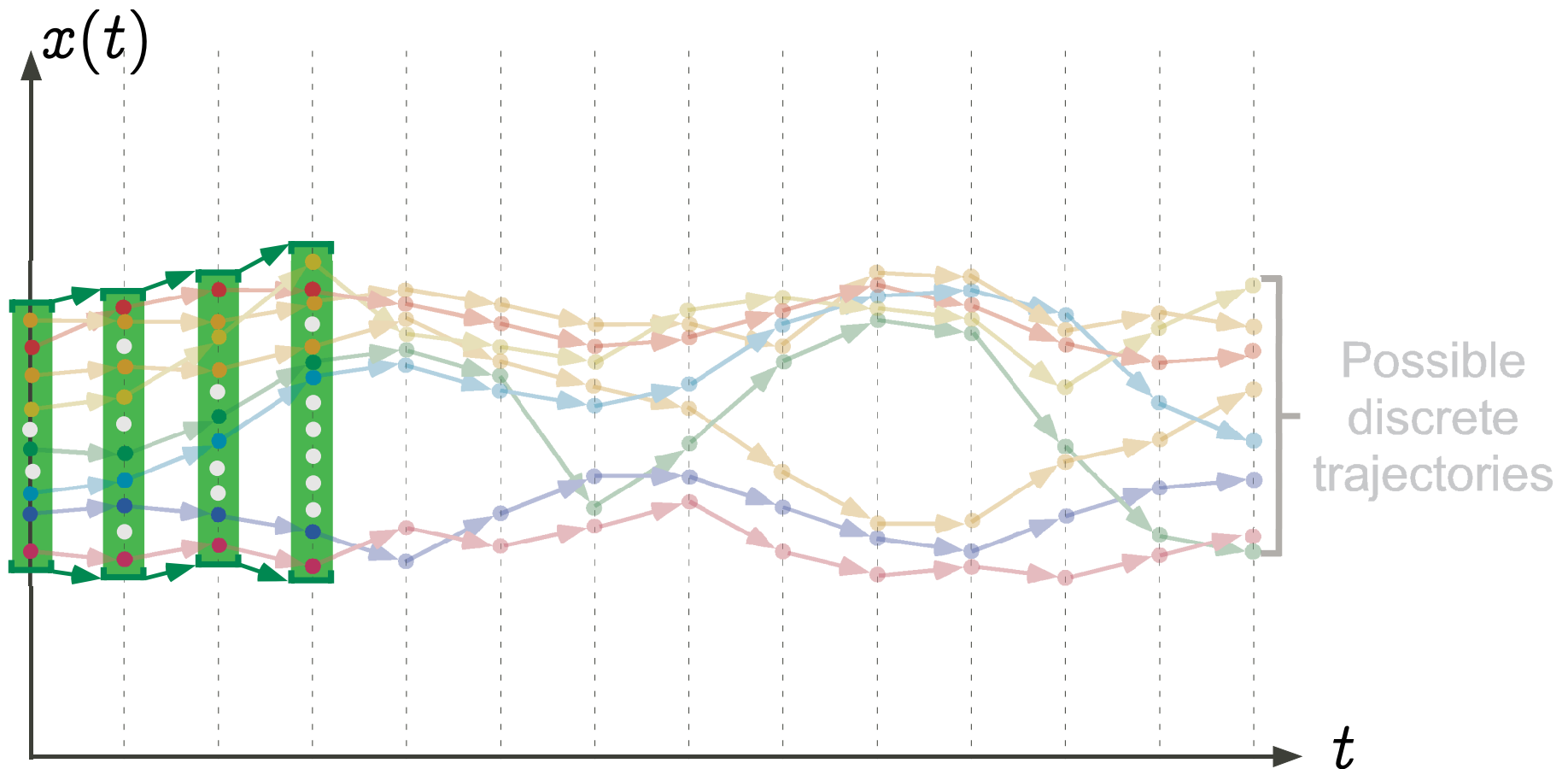
Graphic example: traces of intervals in fixpoint form



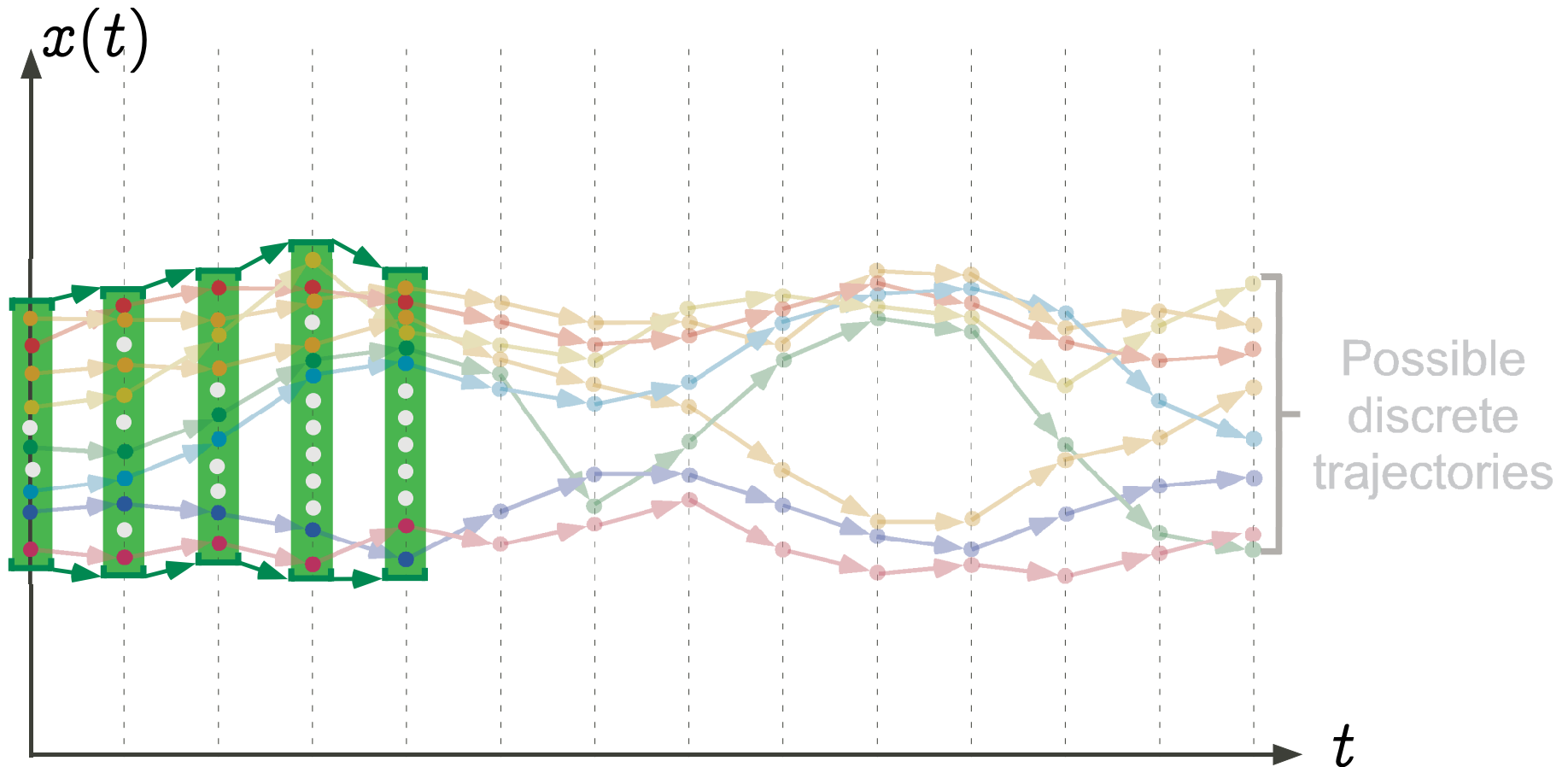
Graphic example: traces of intervals in fixpoint form



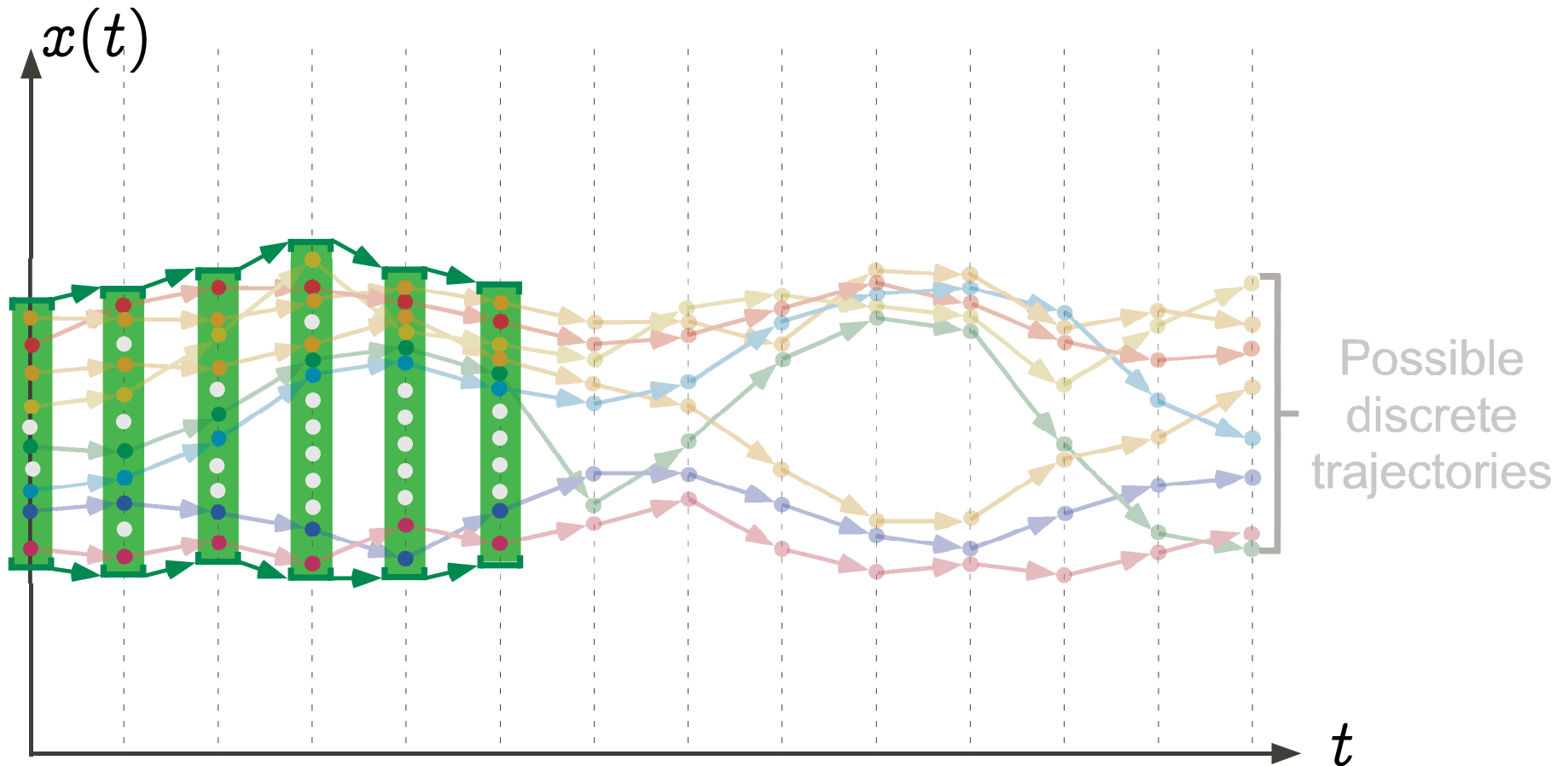
Graphic example: traces of intervals in fixpoint form



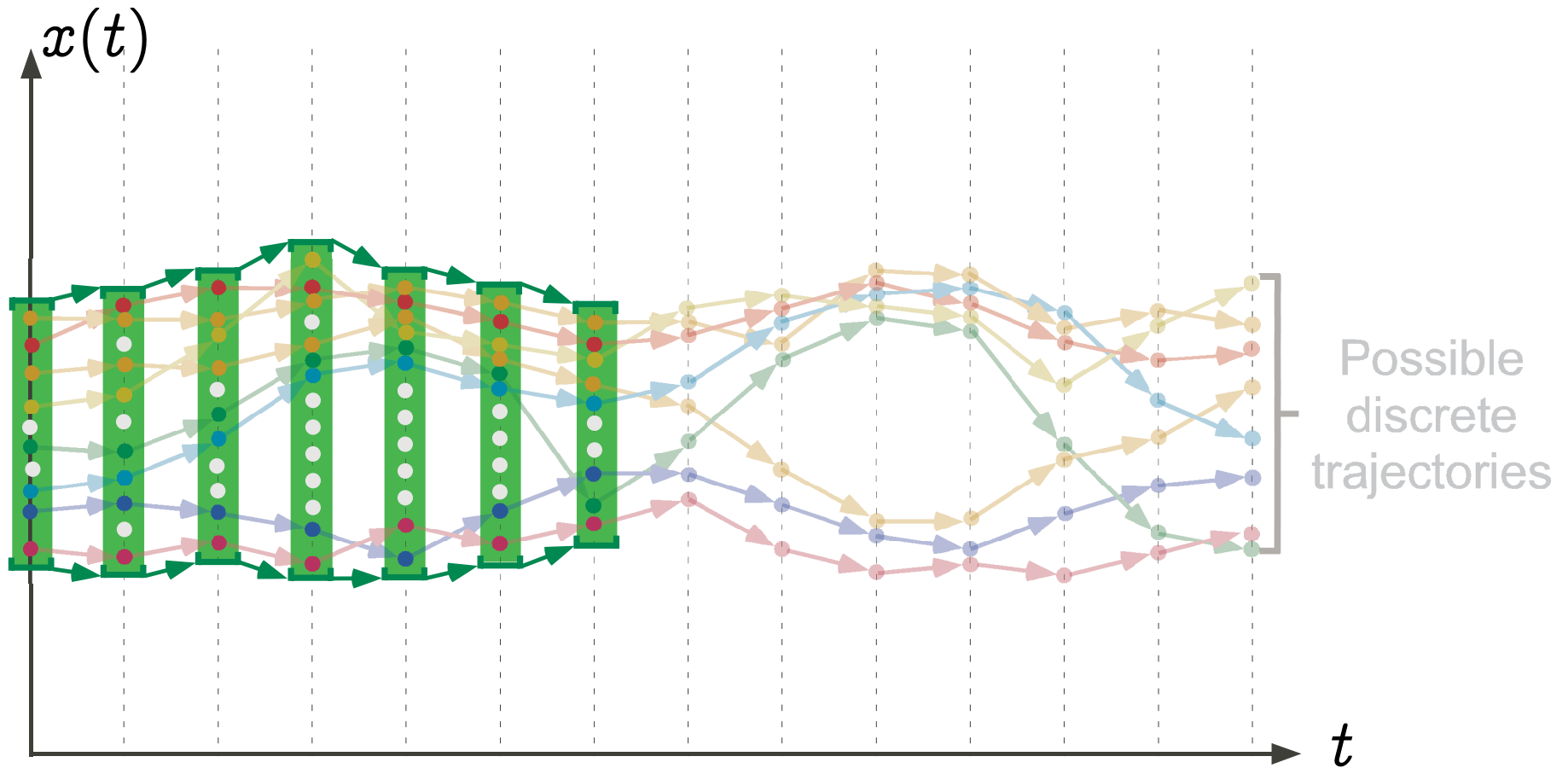
Graphic example: traces of intervals in fixpoint form



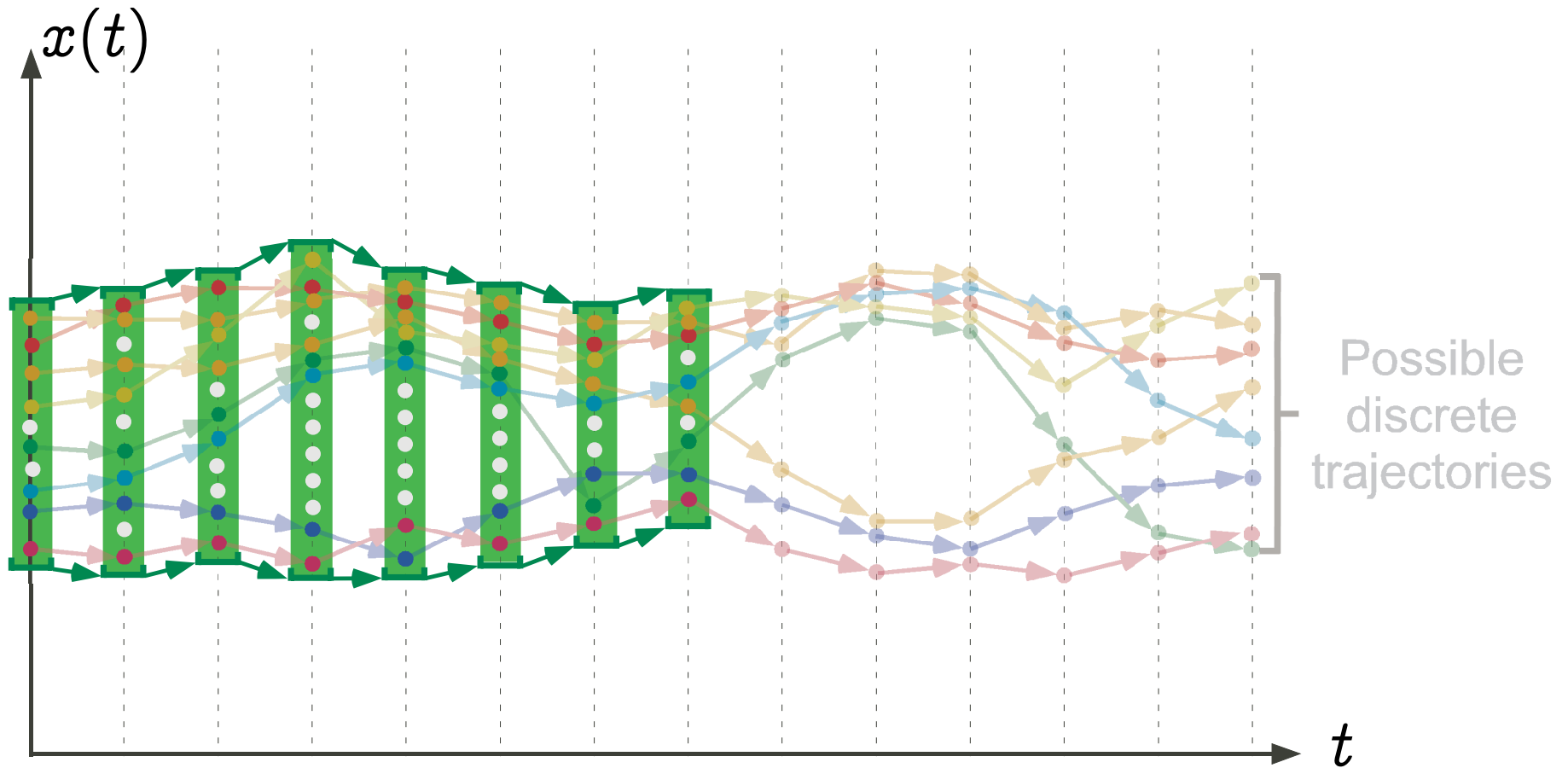
Graphic example: traces of intervals in fixpoint form



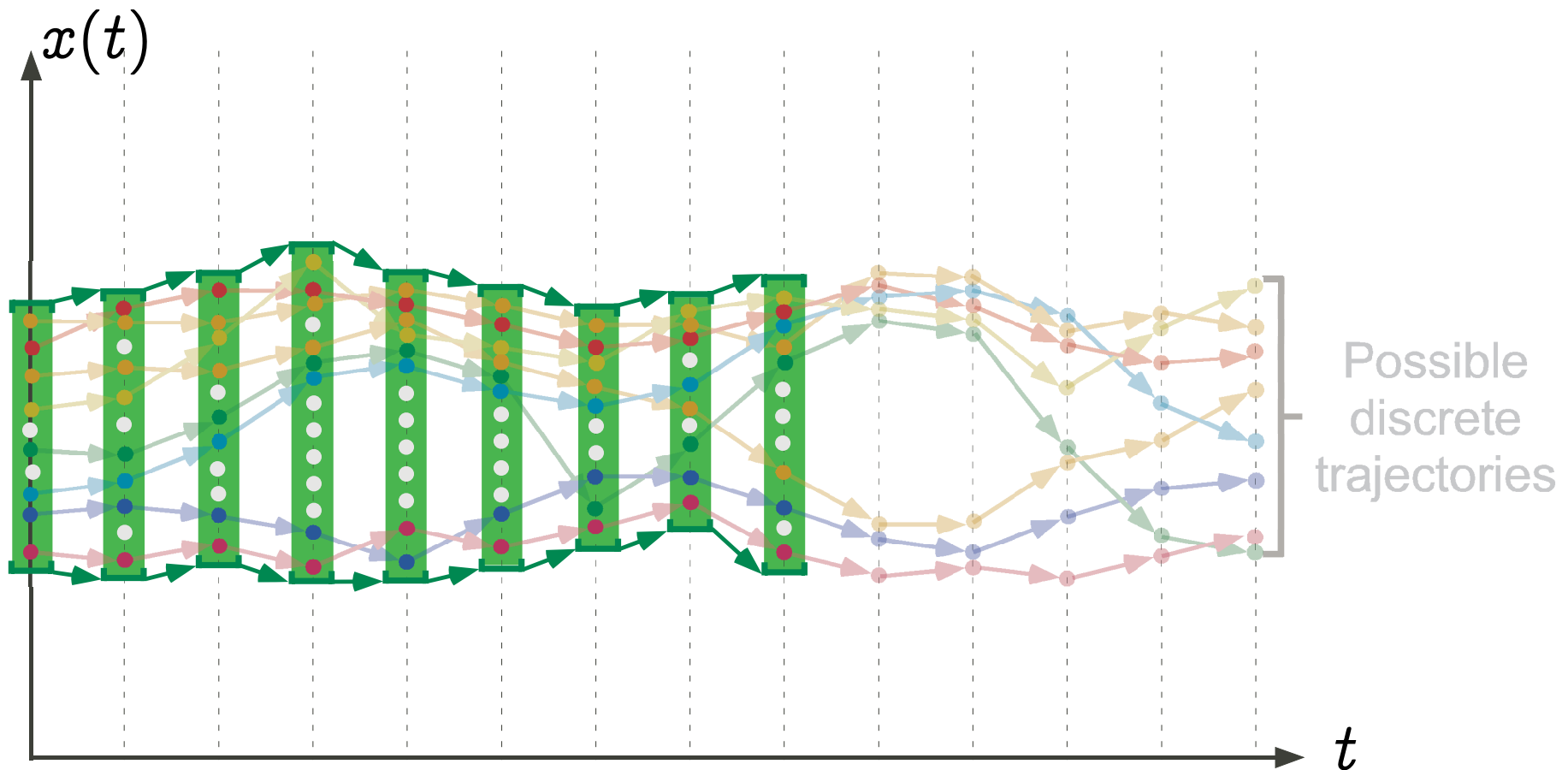
Graphic example: traces of intervals in fixpoint form



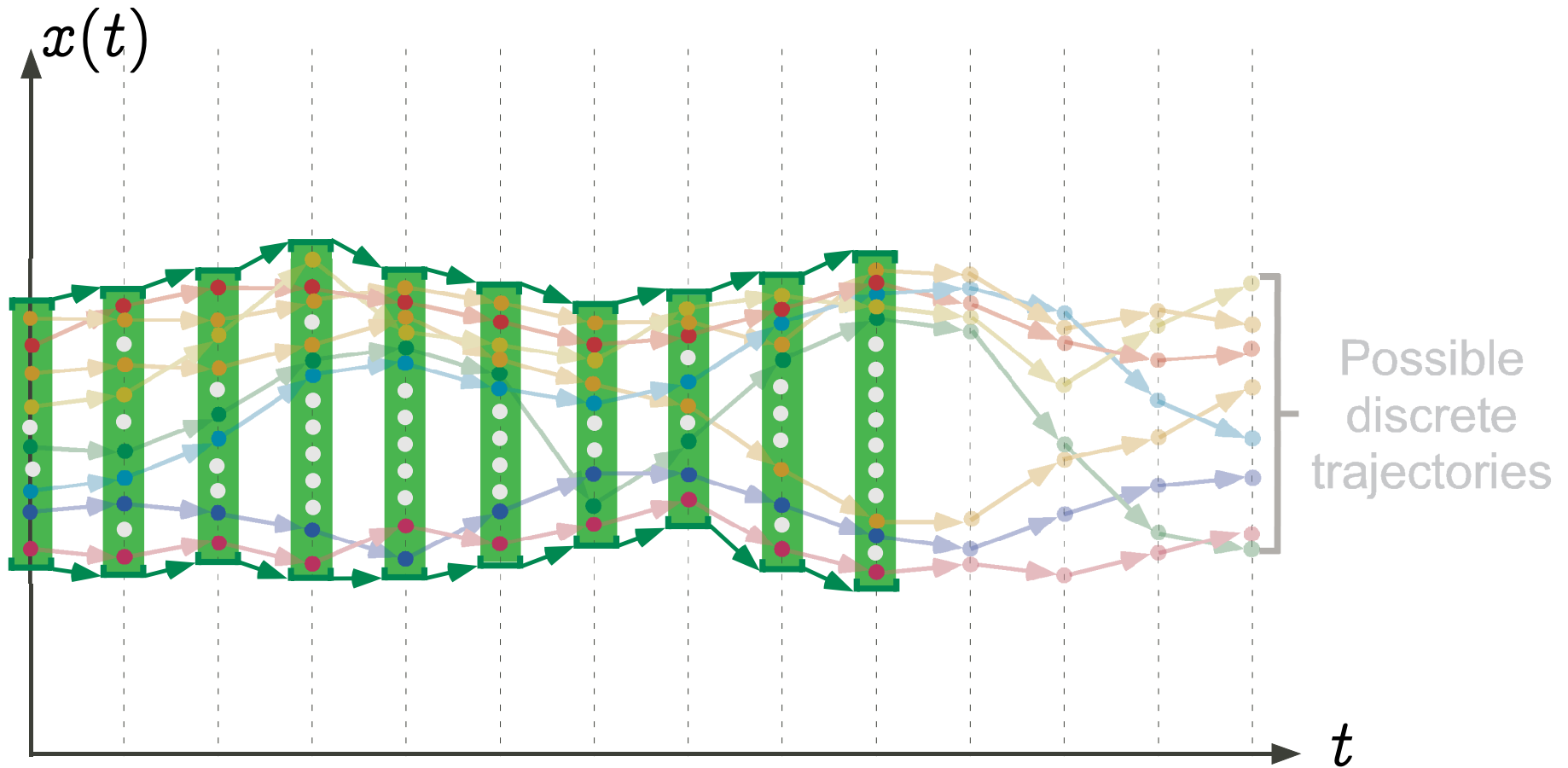
Graphic example: traces of intervals in fixpoint form



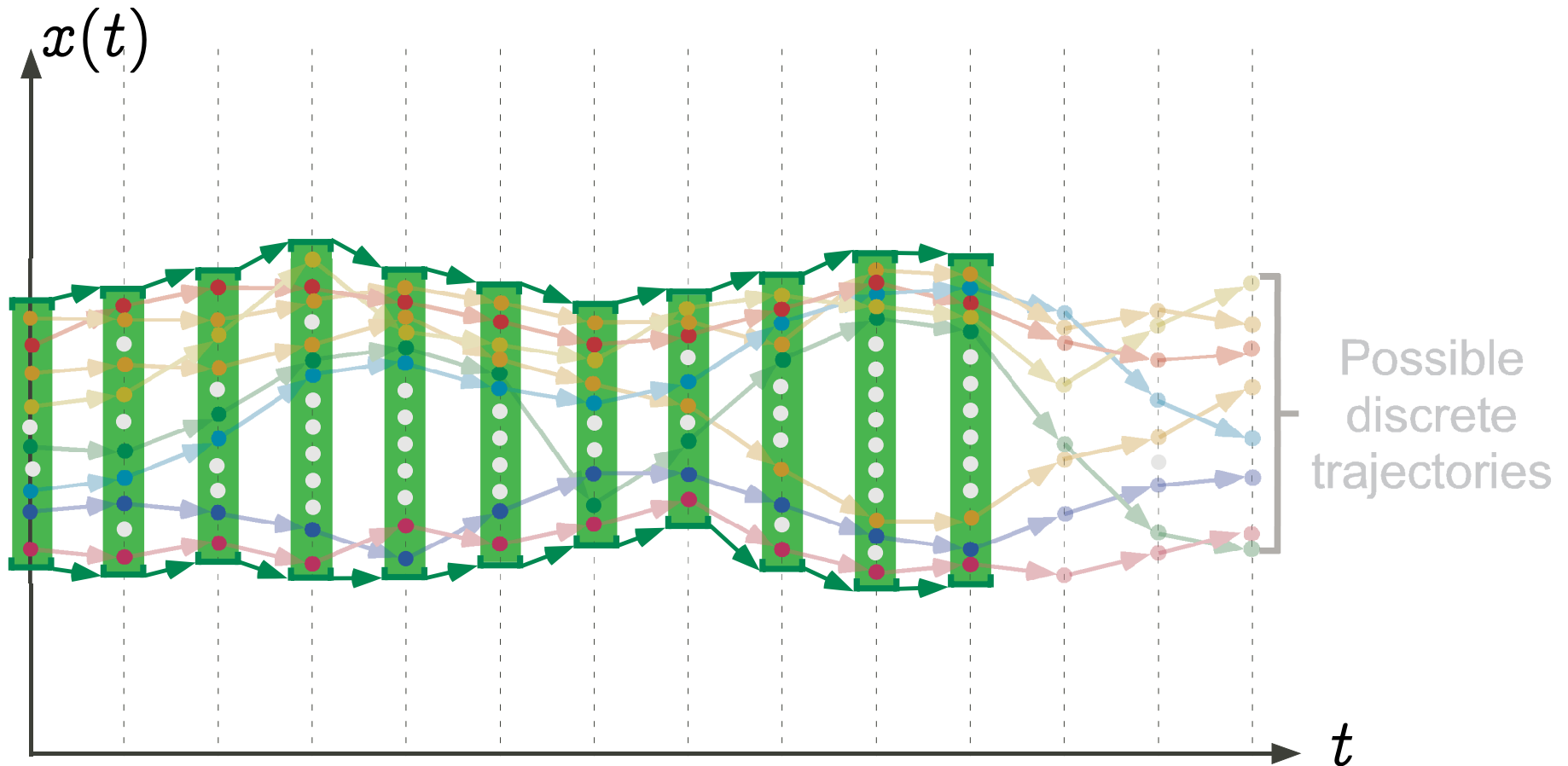
Graphic example: traces of intervals in fixpoint form



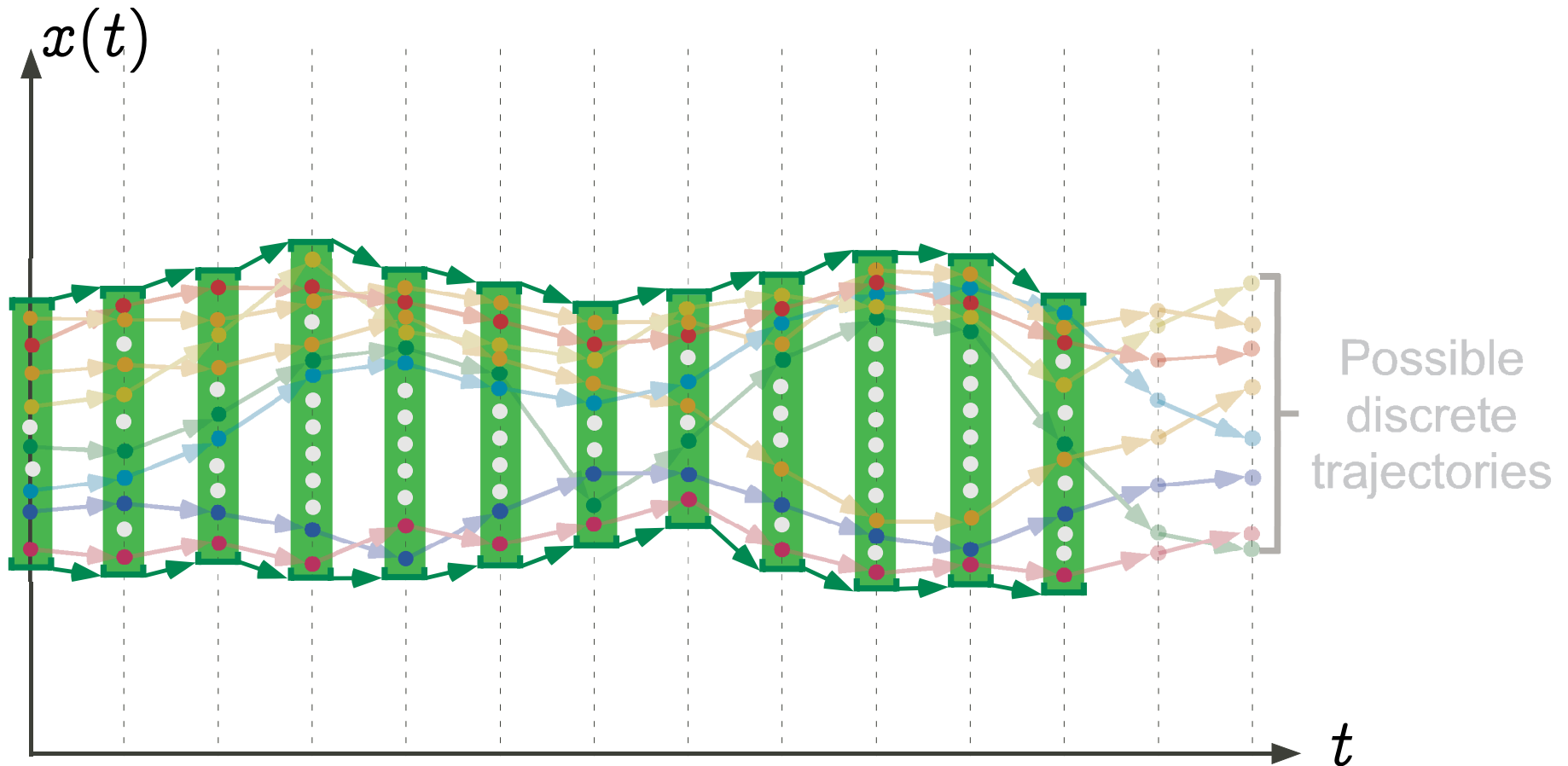
Graphic example: traces of intervals in fixpoint form



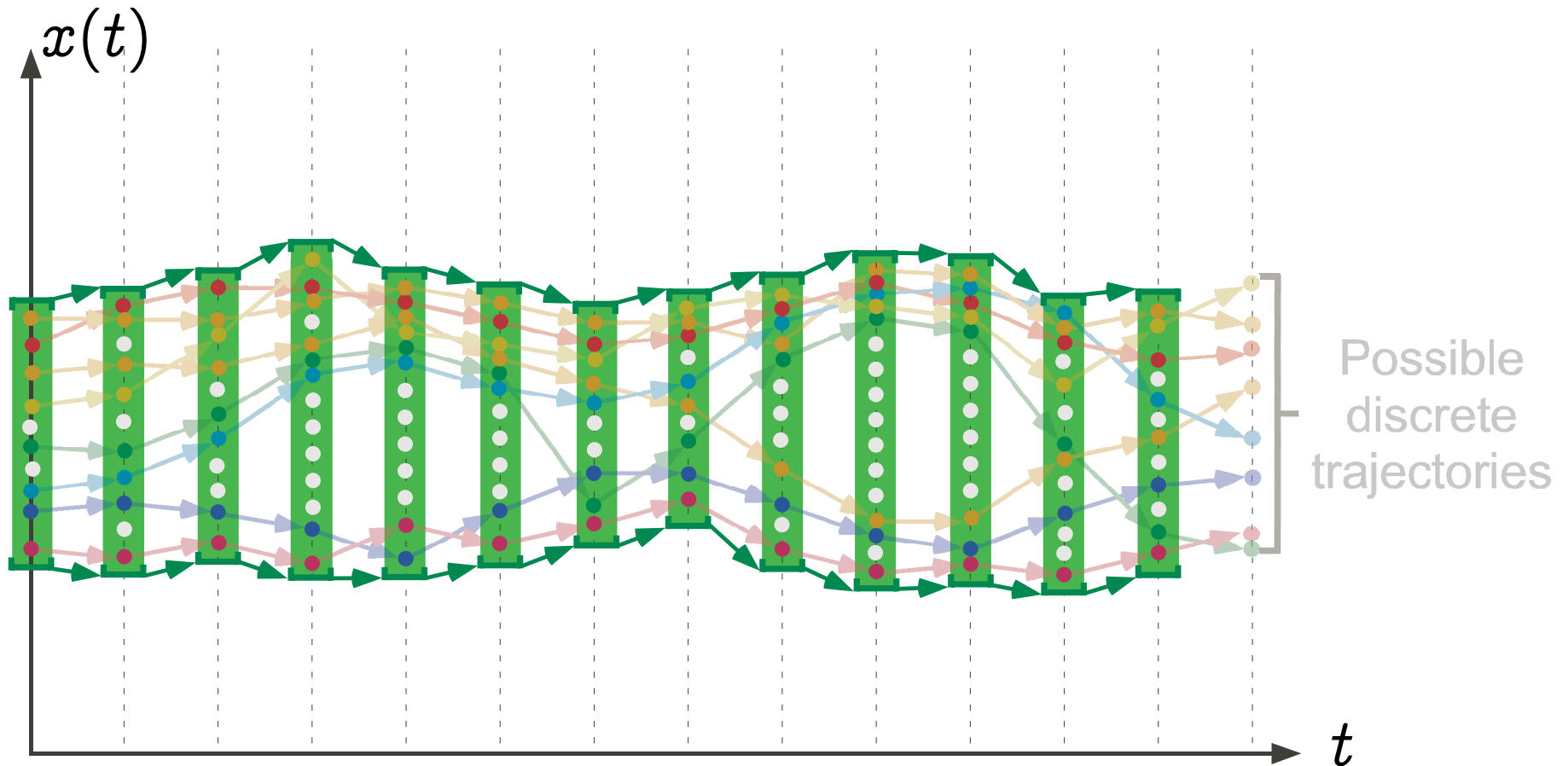
Graphic example: traces of intervals in fixpoint form



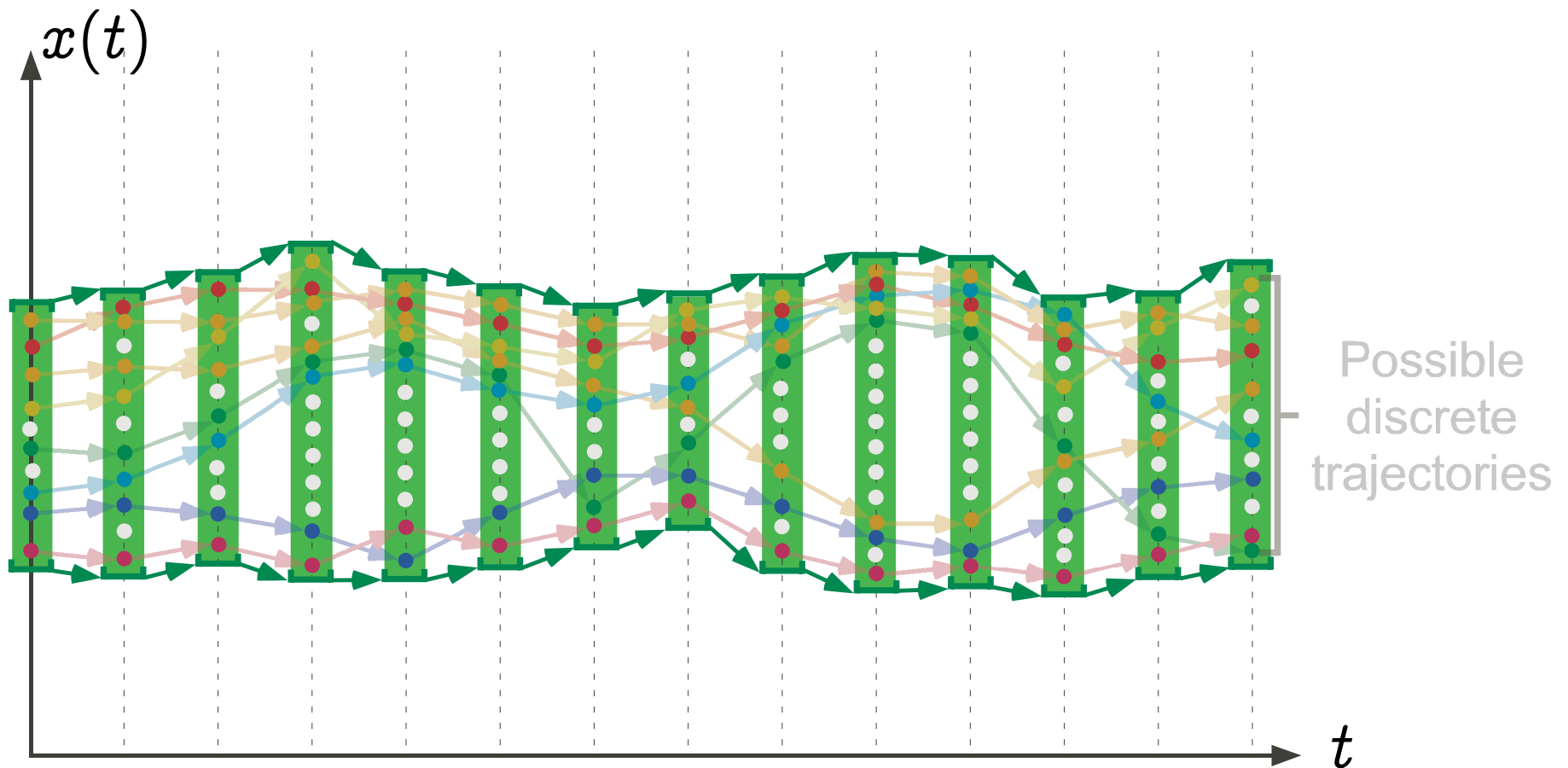
Graphic example: traces of intervals in fixpoint form



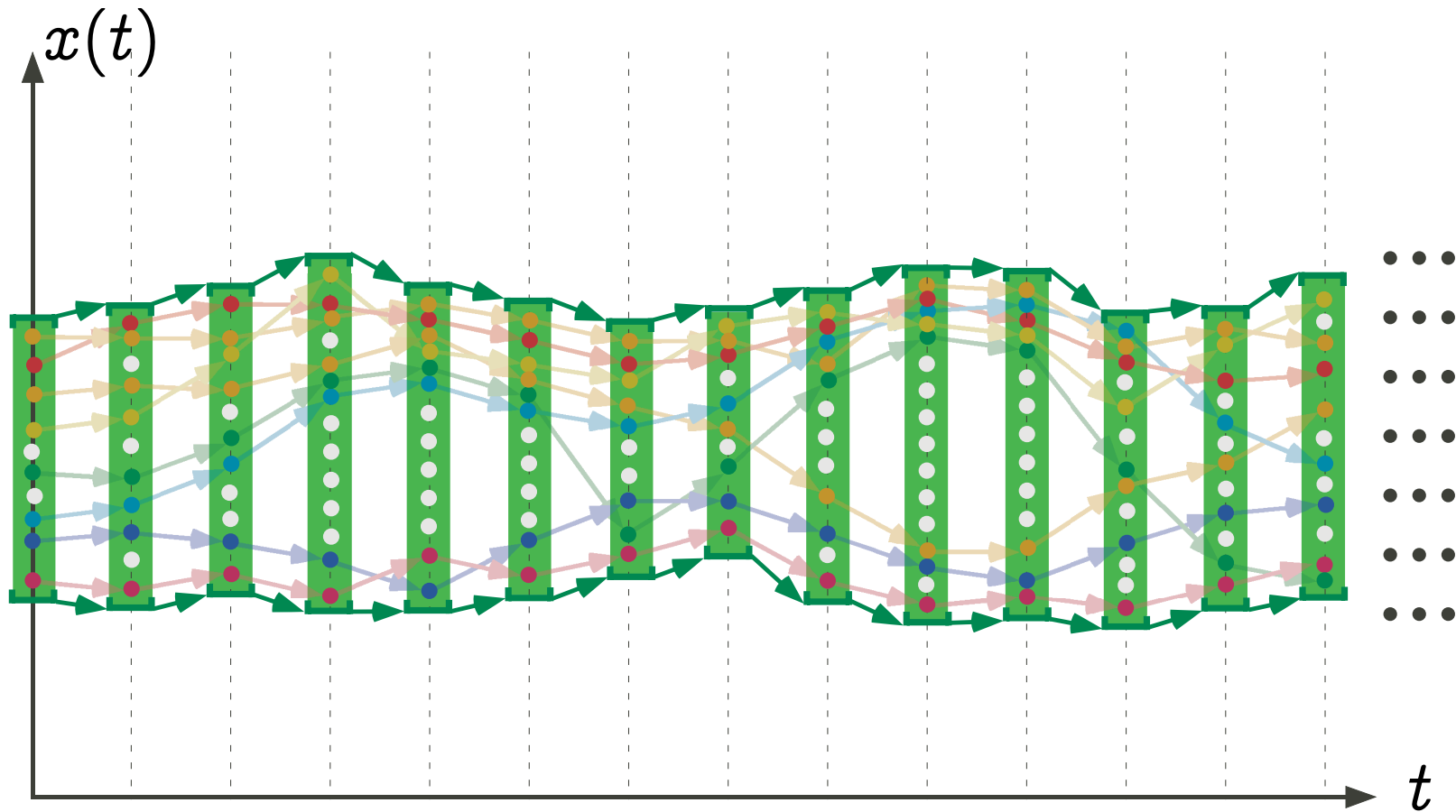
Graphic example: traces of intervals in fixpoint form



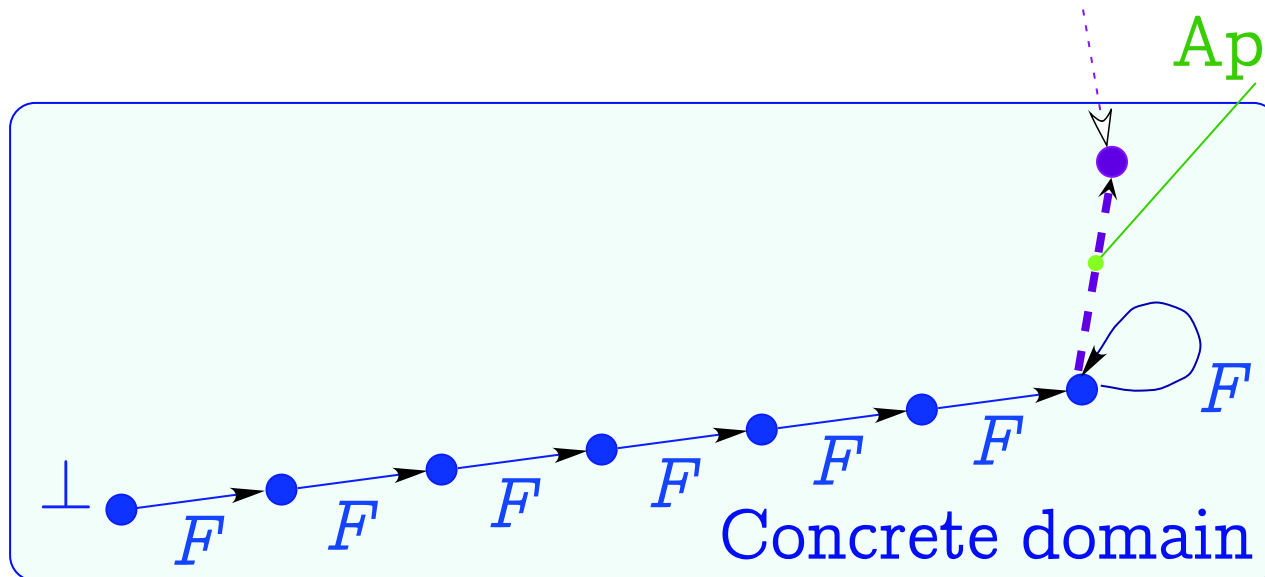
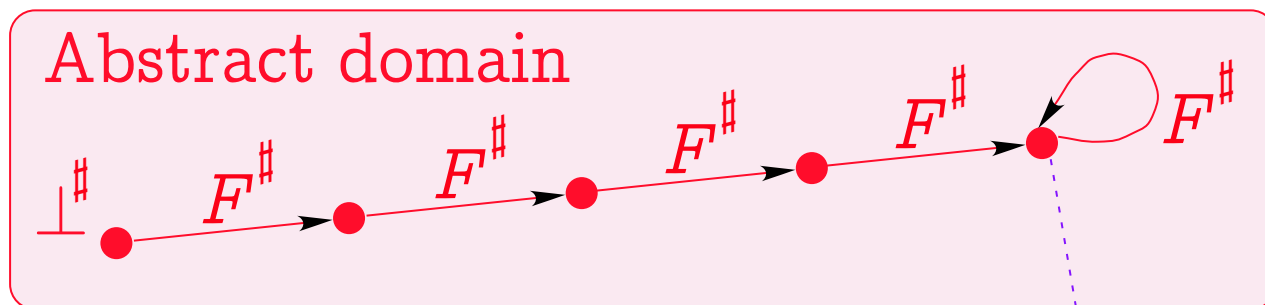
Graphic example: traces of intervals in fixpoint form



Graphic example: traces of intervals in fixpoint form



Approximate fixpoint abstraction



Approximation relation \sqsubseteq

$$\alpha(\text{lfp } F) \sqsubseteq \text{lfp } F^\#$$



approximate/exact fixpoint abstraction

Exact Abstraction:

$$\alpha(\text{lfp } F) = \text{lfp } F^\#$$

Approximate Abstraction:

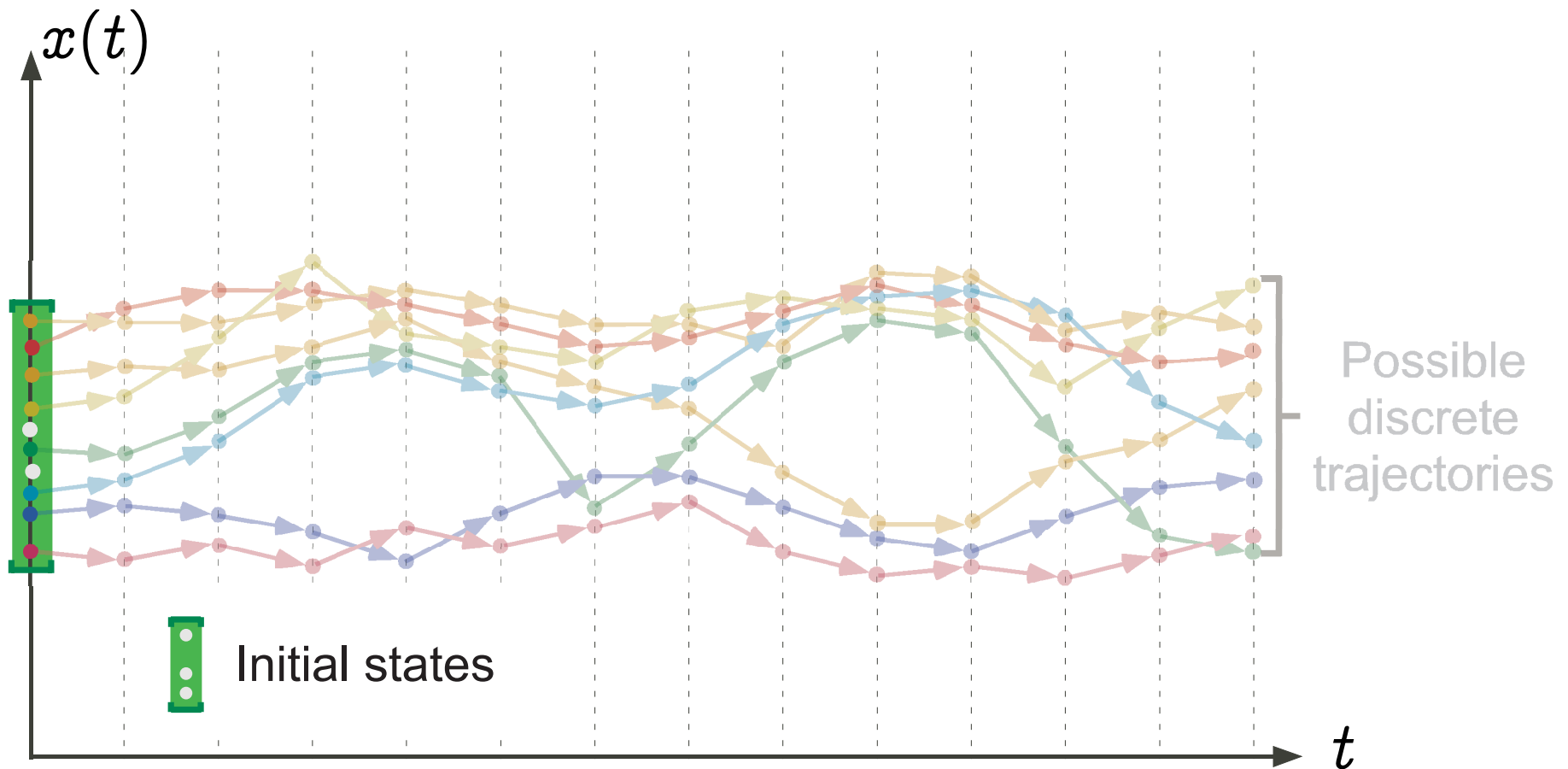
$$\alpha(\text{lfp } F) \sqsubseteq^\# \text{lfp } F^\#$$



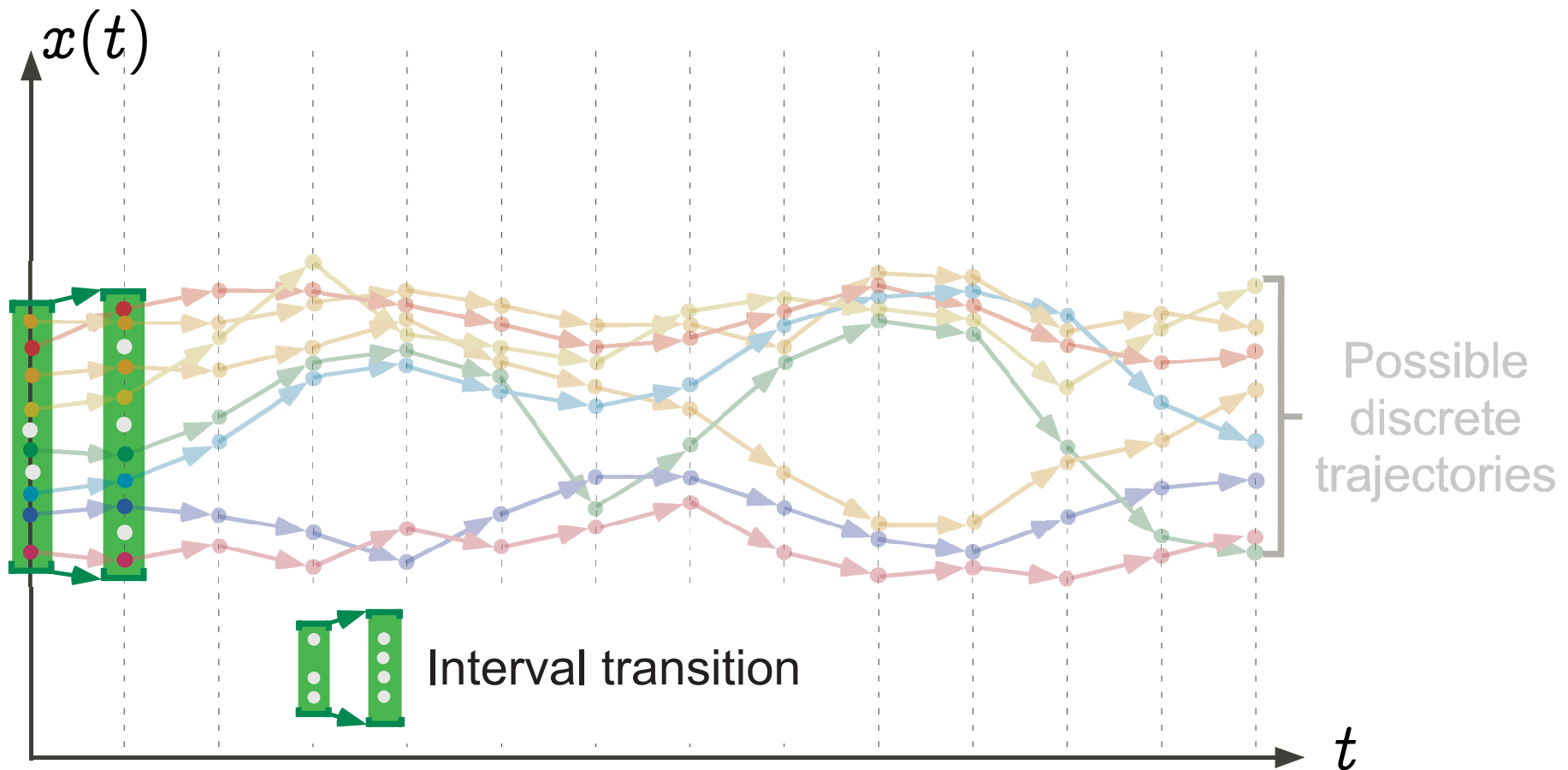
Convergence acceleration by widening/narrowing



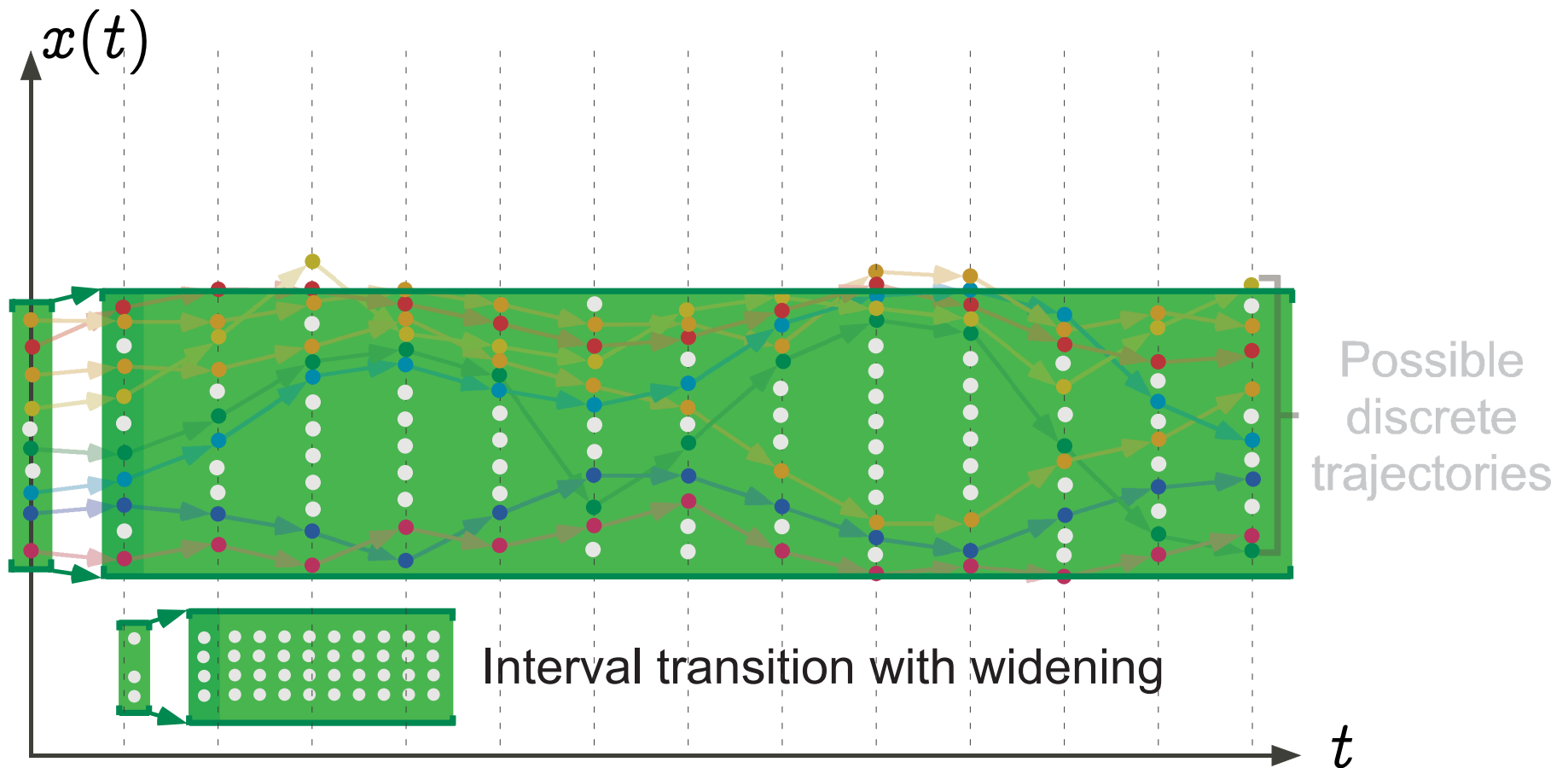
Graphic example: upward iteration with widening



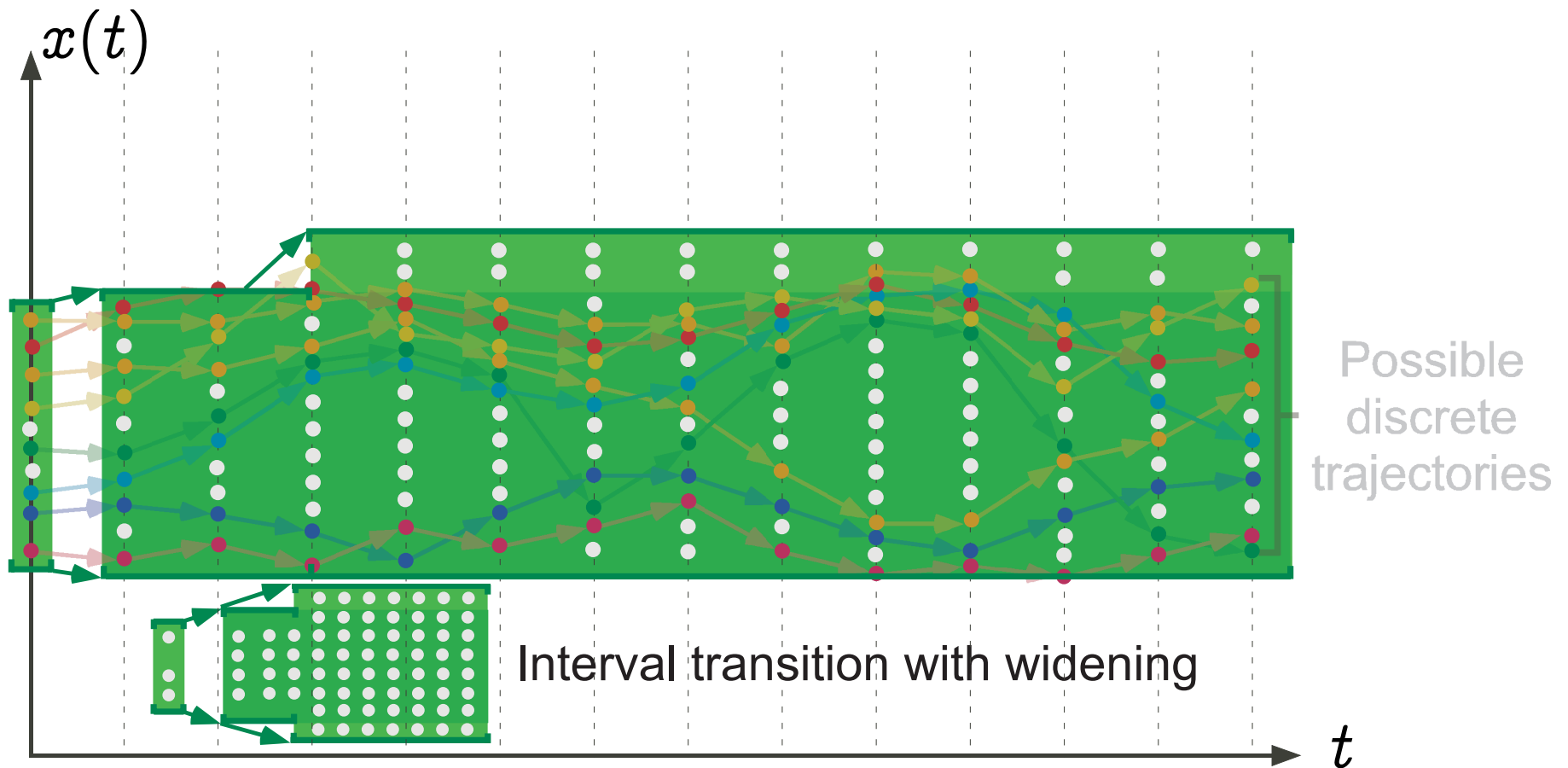
Graphic example: upward iteration with widening



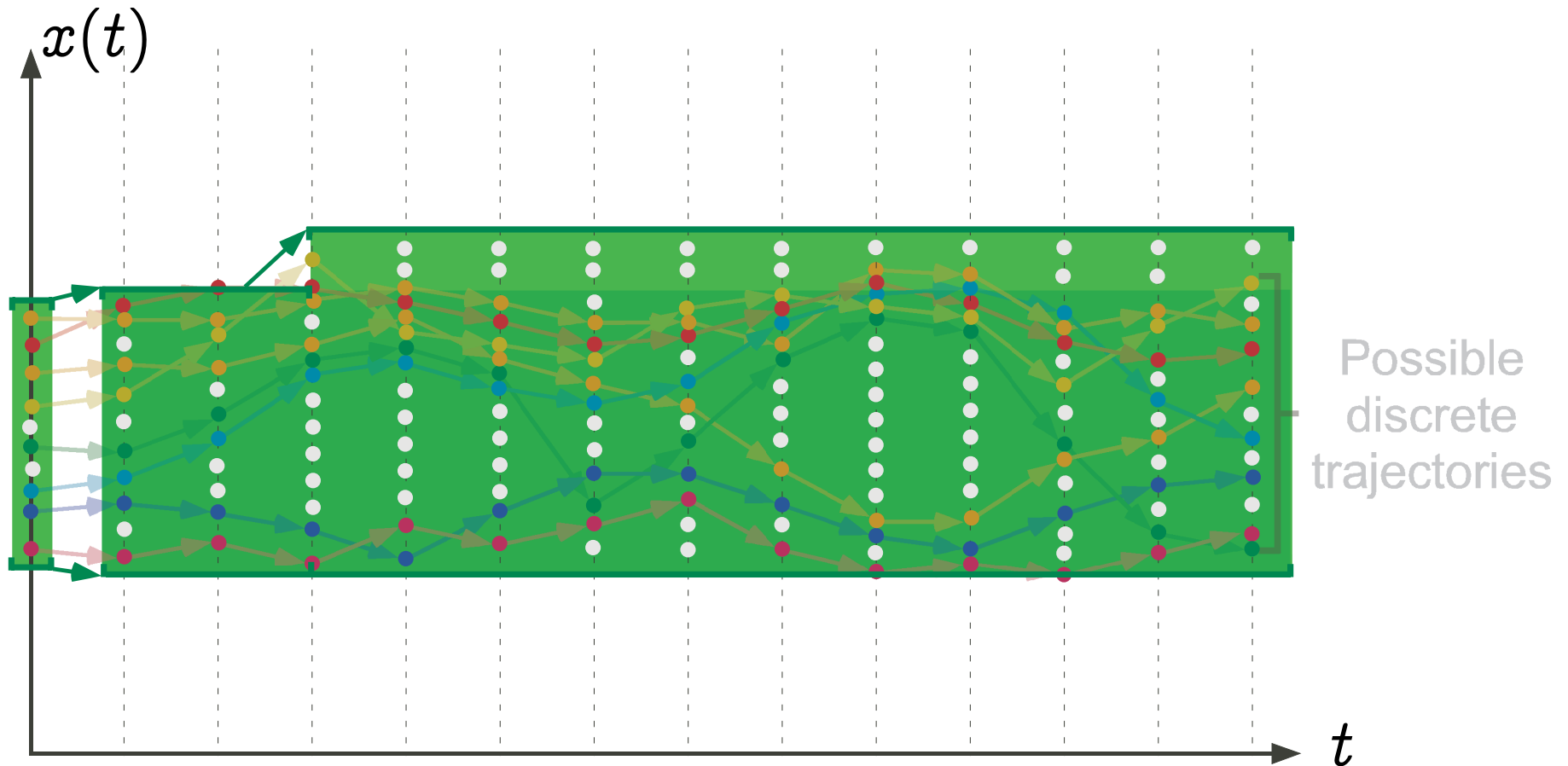
Graphic example: upward iteration with widening



Graphic example: upward iteration with widening



Graphic example: stability of the upward iteration



Convergence acceleration with widening



Widening operator

A widening operator $\nabla \in \overline{L} \times \overline{L} \mapsto \overline{L}$ is such that:

– Correctness:

$$- \forall x, y \in \overline{L} : \gamma(x) \sqsubseteq \gamma(x \nabla y)$$

$$- \forall x, y \in \overline{L} : \gamma(y) \sqsubseteq \gamma(x \nabla y)$$

– Convergence:

- for all increasing chains $x^0 \sqsubseteq x^1 \sqsubseteq \dots$, the increasing chain defined by $y^0 = x^0, \dots, y^{i+1} = y^i \nabla x^{i+1}, \dots$ is not strictly increasing.



Fixpoint approximation with widening

The upward iteration sequence with widening:

- $\hat{X}^0 = \perp$ (infimum)
- $\hat{X}^{i+1} = \hat{X}^i$ if $\overline{F}(\hat{X}^i) \sqsubseteq \hat{X}^i$
 $= \hat{X}^i \nabla F(\hat{X}^i)$ otherwise

is ultimately stationary and its limit \hat{A} is a sound upper approximation of $\text{lfp}_{\perp} \overline{F}$:

$$\text{lfp}_{\perp} \overline{F} \sqsubseteq \hat{A}$$



Interval widening

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

$$\perp \nabla X = X$$

$$X \nabla \perp = X$$

$$[l_0, u_0] \nabla [l_1, u_1] = [\text{if } l_1 < l_0 \text{ then } -\infty \text{ else } l_0, \\ \text{if } u_1 > u_0 \text{ then } +\infty \text{ else } u_0]$$

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):

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



Example: Interval analysis (1975)

Resolution by chaotic increasing iteration:

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

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



Example: Interval analysis (1975)

Increasing chaotic iteration:

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

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



Example: Interval analysis (1975)

Increasing chaotic iteration:

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

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



Example: Interval analysis (1975)

Increasing chaotic iteration:

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

$$\left\{ \begin{array}{l} X_1 = [1, 1] \\ X_2 = [1, 1] \\ X_3 = [2, 2] \\ X_4 = \emptyset \end{array} \right.$$



Example: Interval analysis (1975)

Increasing chaotic iteration:

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

$$\left\{ \begin{array}{l} X_1 = [1, 1] \\ X_2 = [1, 2] \\ X_3 = [2, 2] \\ X_4 = \emptyset \end{array} \right.$$



Example: Interval analysis (1975)

Increasing chaotic iteration: **convergence !**

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

$$\left\{ \begin{array}{l} X_1 = [1, 1] \\ X_2 = [1, 2] \\ X_3 = [2, 3] \\ X_4 = \emptyset \end{array} \right.$$



Example: Interval analysis (1975)

Increasing chaotic iteration: **convergence !!**

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

$$\left\{ \begin{array}{l} X_1 = [1, 1] \\ X_2 = [1, 3] \\ X_3 = [2, 3] \\ X_4 = \emptyset \end{array} \right.$$



Example: Interval analysis (1975)

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

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

$$\left\{ \begin{array}{l} X_1 = [1, 1] \\ X_2 = [1, 3] \\ X_3 = [2, 4] \\ X_4 = \emptyset \end{array} \right.$$



Example: Interval analysis (1975)

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

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



Example: Interval analysis (1975)

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

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



Example: Interval analysis (1975)

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

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

$$\left\{ \begin{array}{l} X_1 = [1, 1] \\ X_2 = [1, 5] \\ X_3 = [2, 5] \\ X_4 = \emptyset \end{array} \right.$$



Example: Interval analysis (1975)

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

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



Example: Interval analysis (1975)

Convergence speed-up by widening:

1:	<code>x := 1;</code>	{	$X_1 = [1, 1]$	
	<code>while x < 10000 do</code>		$X_2 = (X_1 \cup X_3) \cap [-\infty, 9999]$	
	<code> x := x + 1</code>	$X_3 = X_2 \oplus [1, 1]$		
	<code>od;</code>	$X_4 = (X_1 \cup X_3) \cap [10000, +\infty]$		
2:				
3:		{	$X_1 = [1, 1]$	
			$X_2 = [1, +\infty] \quad \Leftarrow \text{widening}$	
		$X_3 = [2, 6]$		
4:		$X_4 = \emptyset$		



Example: Interval analysis (1975)

Decreasing chaotic iteration:

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

$$\left\{ \begin{array}{l} X_1 = [1, 1] \\ X_2 = [1, +\infty] \\ X_3 = [2, +\infty] \\ X_4 = \emptyset \end{array} \right.$$



Example: Interval analysis (1975)

Decreasing chaotic iteration:

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

$$\left\{ \begin{array}{l} X_1 = [1, 1] \\ X_2 = [1, 9999] \\ X_3 = [2, +\infty] \\ X_4 = \emptyset \end{array} \right.$$



Example: Interval analysis (1975)

Decreasing chaotic iteration:

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

$$\left\{ \begin{array}{l} X_1 = [1, 1] \\ X_2 = [1, 9999] \\ X_3 = [2, +10000] \\ X_4 = \emptyset \end{array} \right.$$



Example: Interval analysis (1975)

Final solution:

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



Example: Interval analysis (1975)

Result of the interval analysis:

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

$$\left\{ \begin{array}{l} X_1 = [1, 1] \\ X_2 = [1, 9999] \\ X_3 = [2, +10000] \\ X_4 = [+10000, +10000] \end{array} \right.$$



Example: Interval analysis (1975)

Checking absence of runtime errors with interval analysis:

```
x := 1;
```

1: {x = 1}

```
while x < 10000 do
```

2: {x ∈ [1, 9999]}

```
    x := x + 1
```

← no overflow

3: {x ∈ [2, +10000]}

```
od;
```

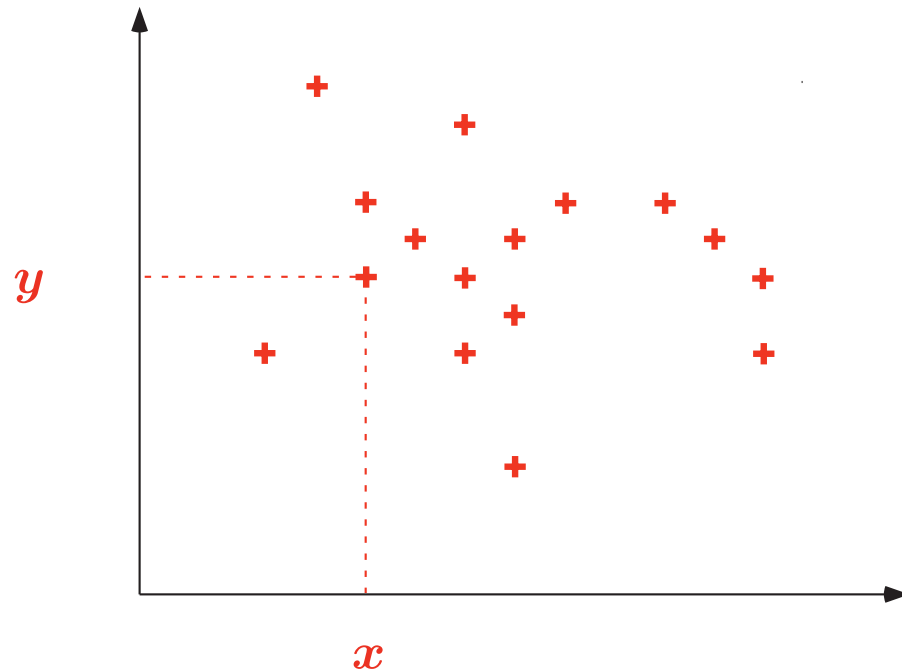
4: {x = 10000}



Refinement of abstractions



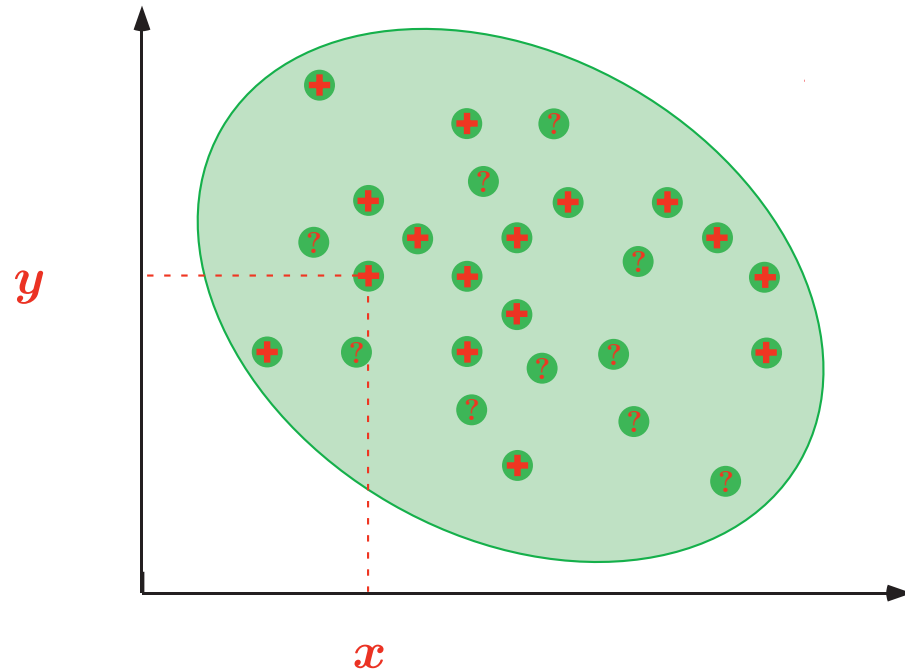
Approximations of an [in]finite set of points:



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



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



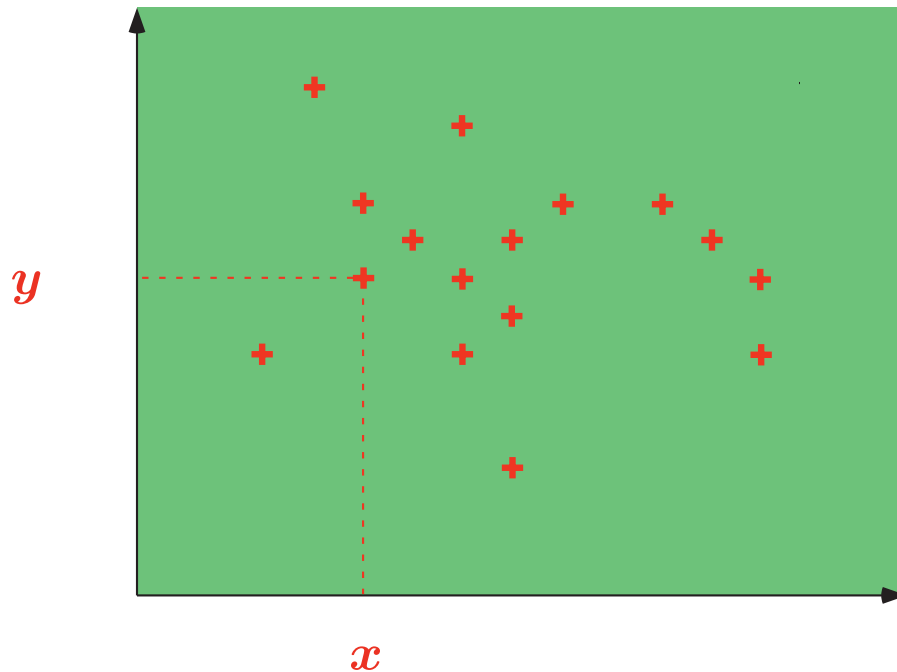
$\{\dots, \langle 19, 77 \rangle, \dots,$
 $\langle 20, 03 \rangle, \langle ?, ? \rangle, \dots\}$

From Below: dual² + combinations.

² Trivial for finite states (liveness model-checking), more difficult for infinite states (variant functions).



Effective computable approximations of an [in]finite set of points; Signs³

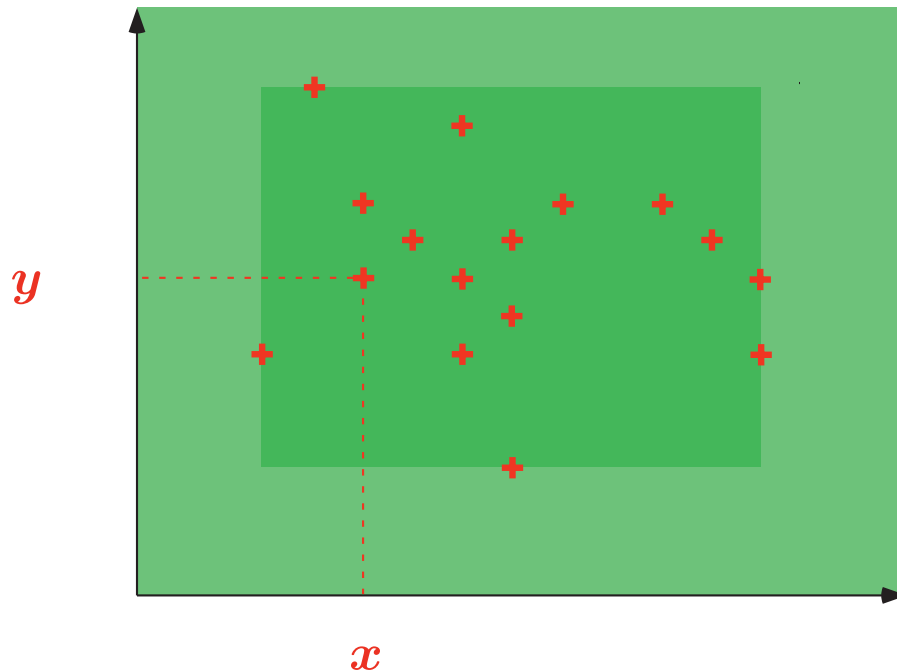


$$\begin{cases} x \geq 0 \\ y \geq 0 \end{cases}$$

³ 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; Intervals⁴

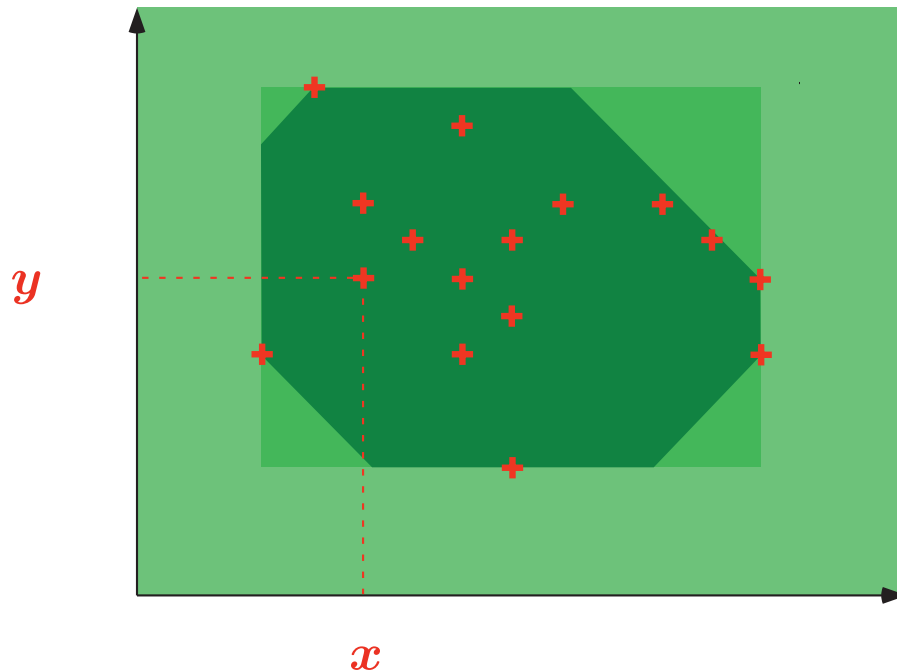


$$\begin{cases} x \in [19, 77] \\ y \in [20, 03] \end{cases}$$

⁴ 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; Octagons⁵

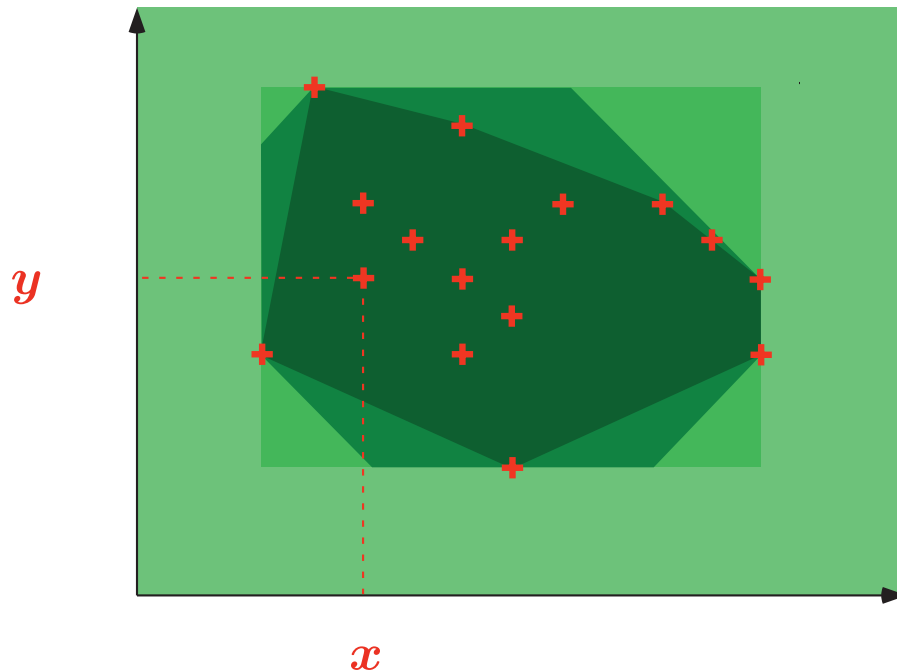


$$\begin{cases} 1 \leq x \leq 9 \\ x + y \leq 77 \\ 1 \leq y \leq 9 \\ x - y \leq 99 \end{cases}$$

⁵ 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; Polyhedra⁶

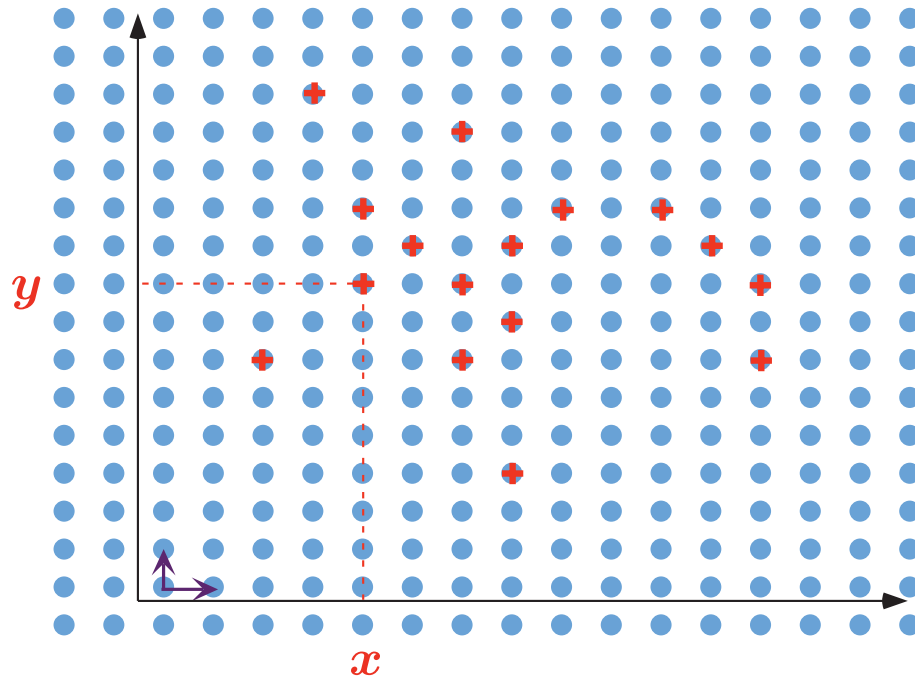


$$\begin{cases} 19x + 77y \leq 2004 \\ 20x + 03y \geq 0 \end{cases}$$

⁶ 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 congruences⁷

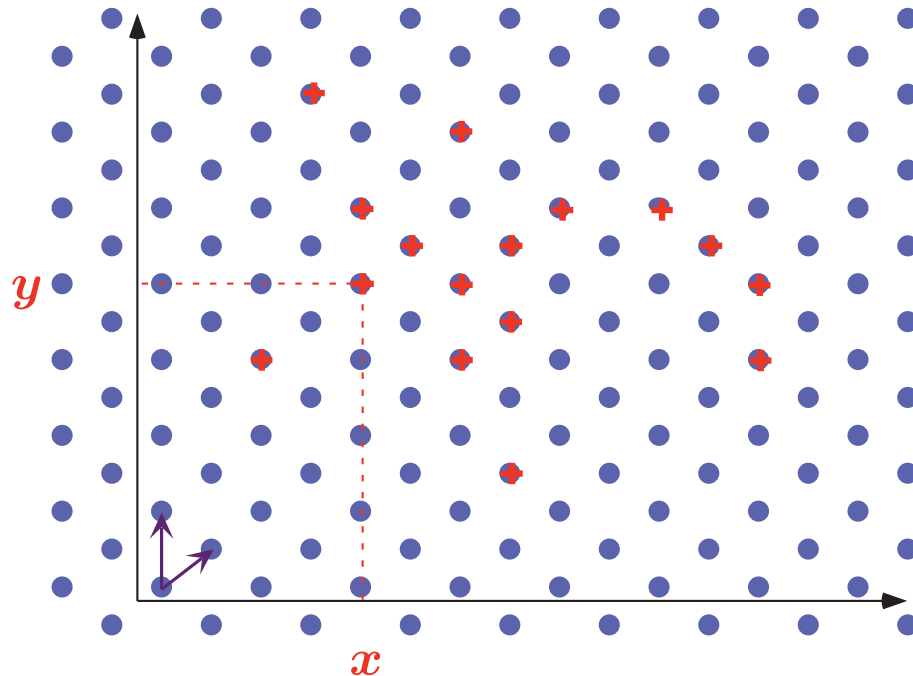


$$\begin{cases} x = 19 \pmod{77} \\ y = 20 \pmod{99} \end{cases}$$

⁷ 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 congruences⁸

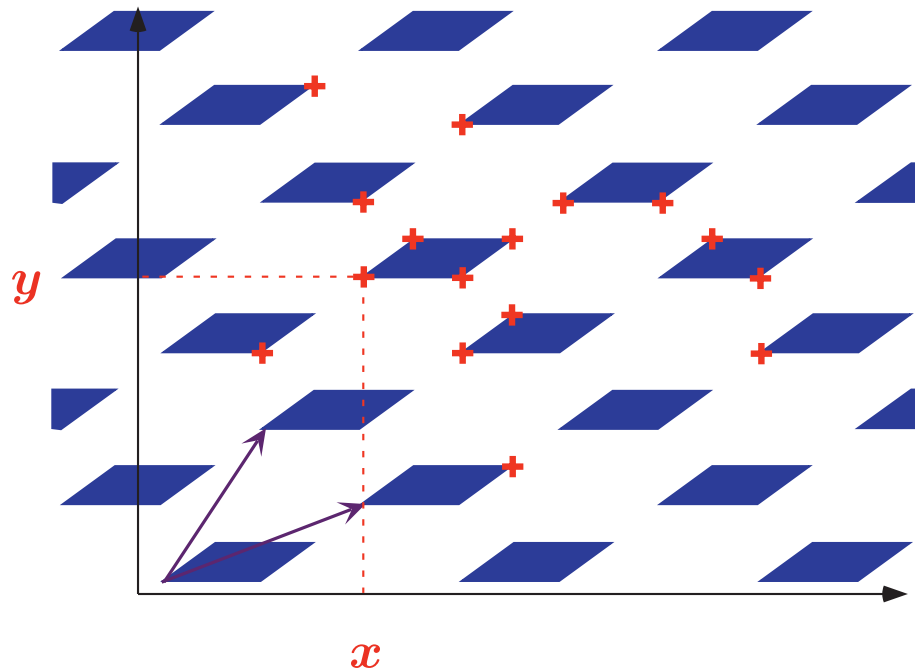


$$\begin{cases} 1x + 9y = 7 \pmod{8} \\ 2x - 1y = 9 \pmod{9} \end{cases}$$

⁸ 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 congruences⁹



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

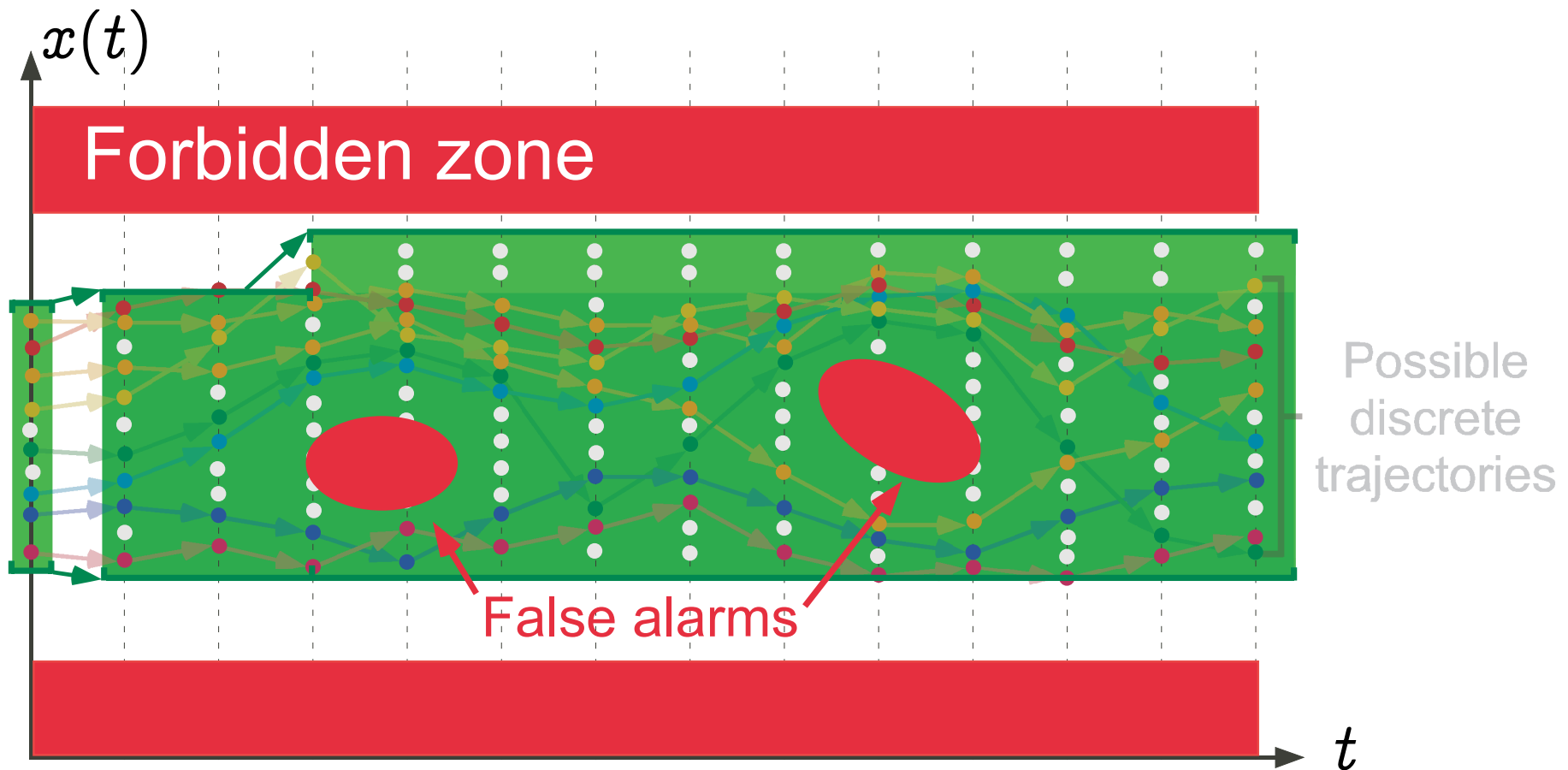
⁹ F. Masdupuy. *Array Operations Abstraction Using Semantic Analysis of Trapezoid Congruences*. ACM ICS '92.



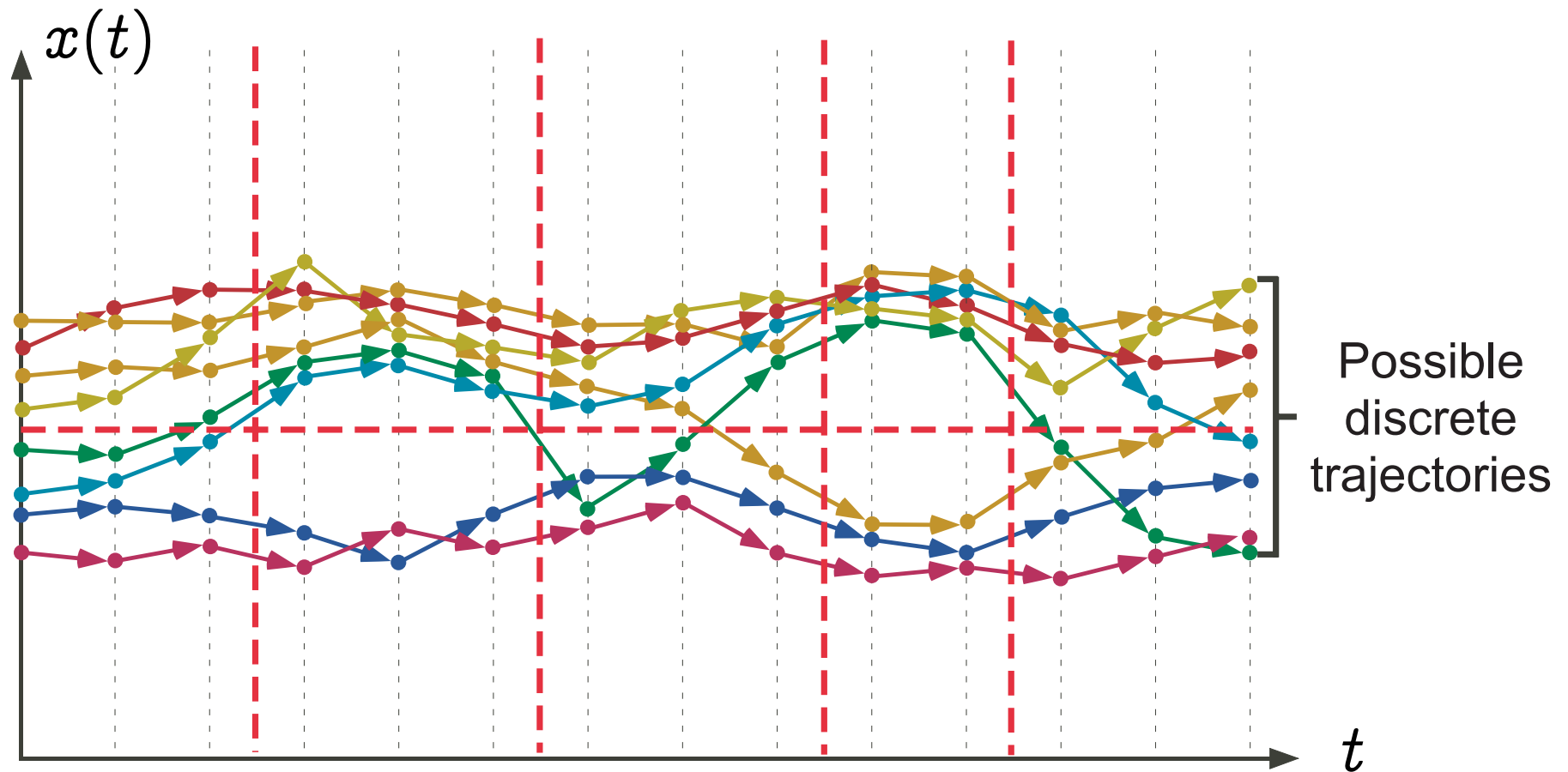
Refinement of iterates



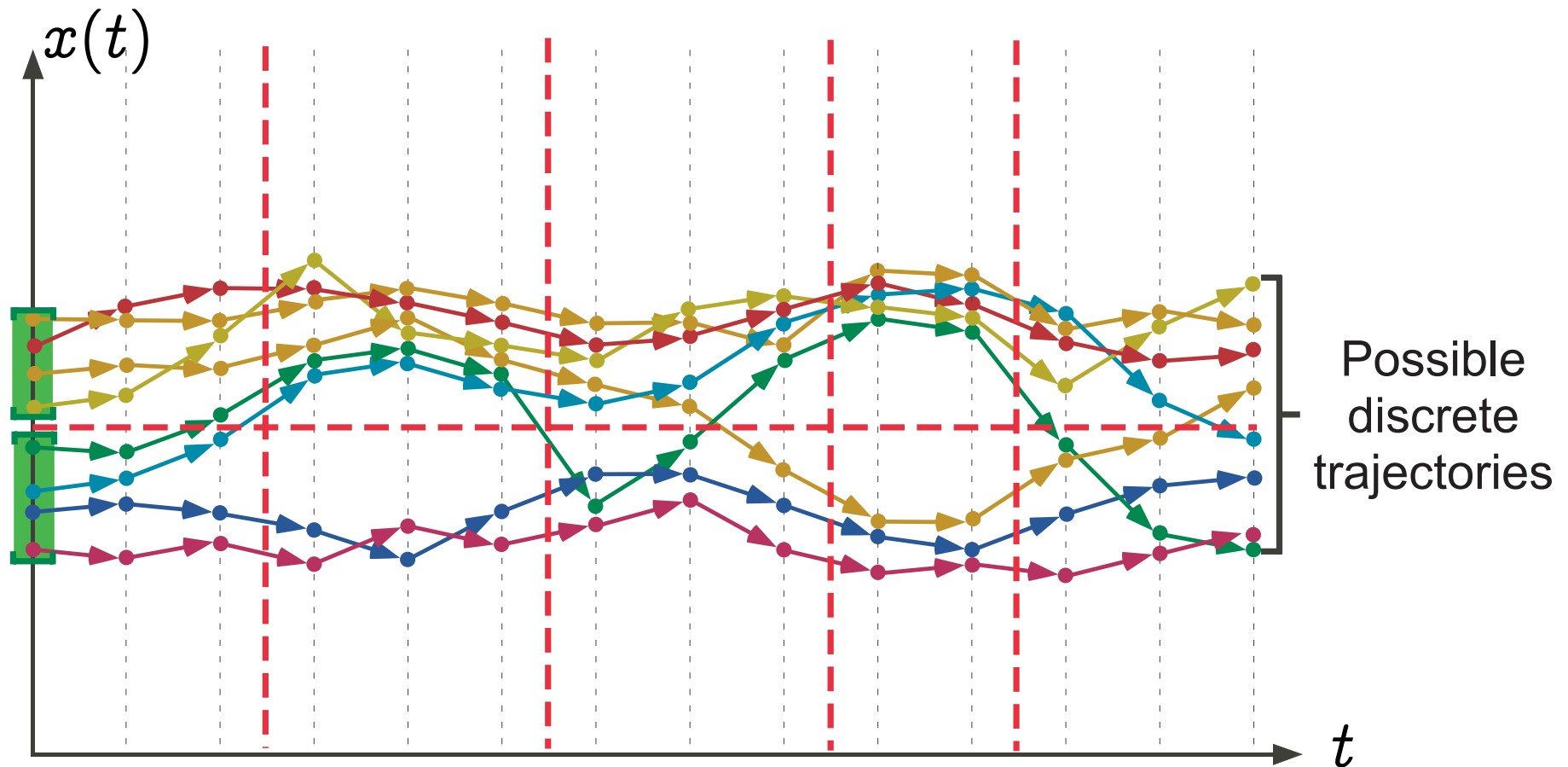
Graphic example: Refinement required by false alarms



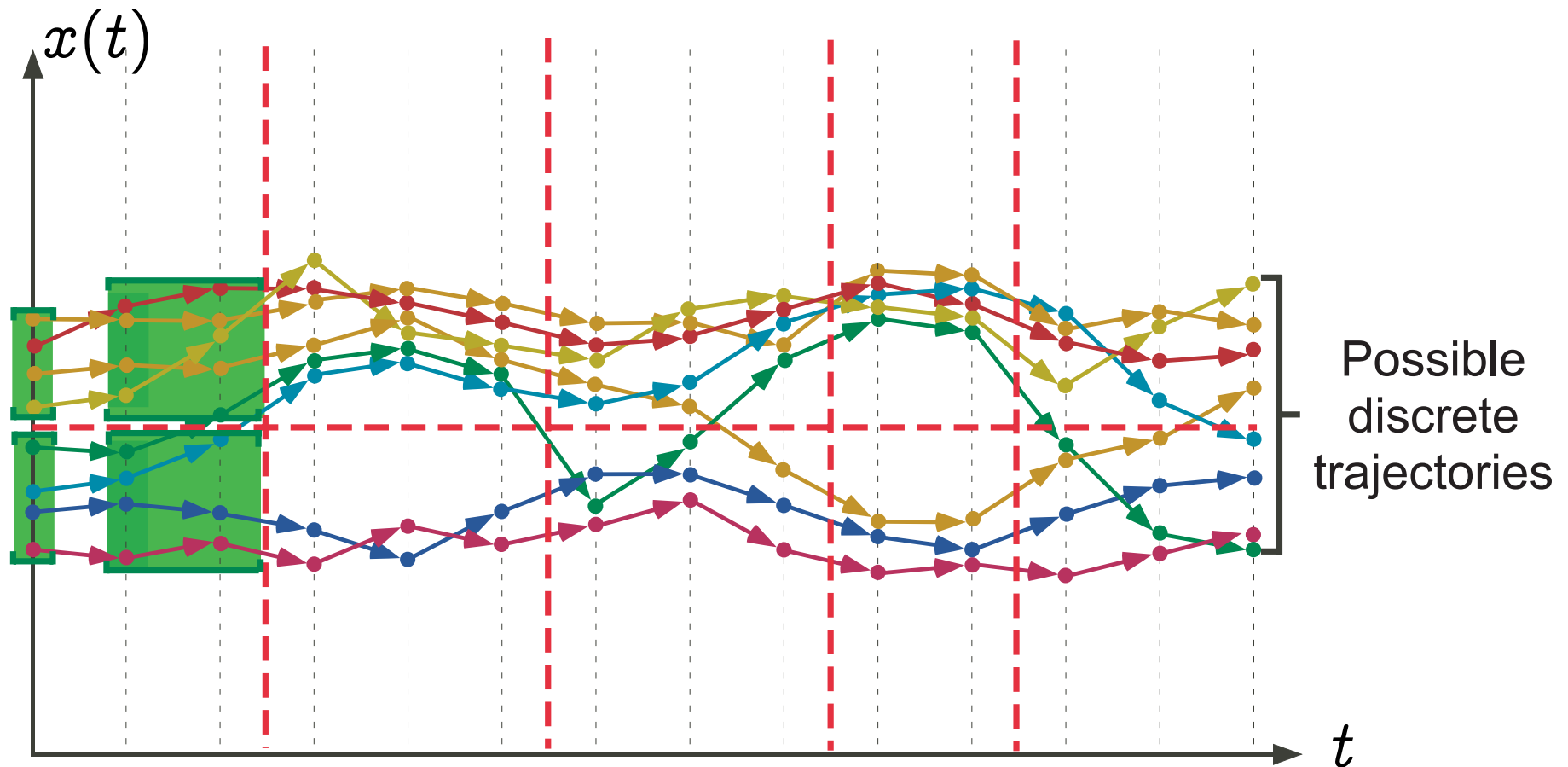
Graphic example: Partitionning



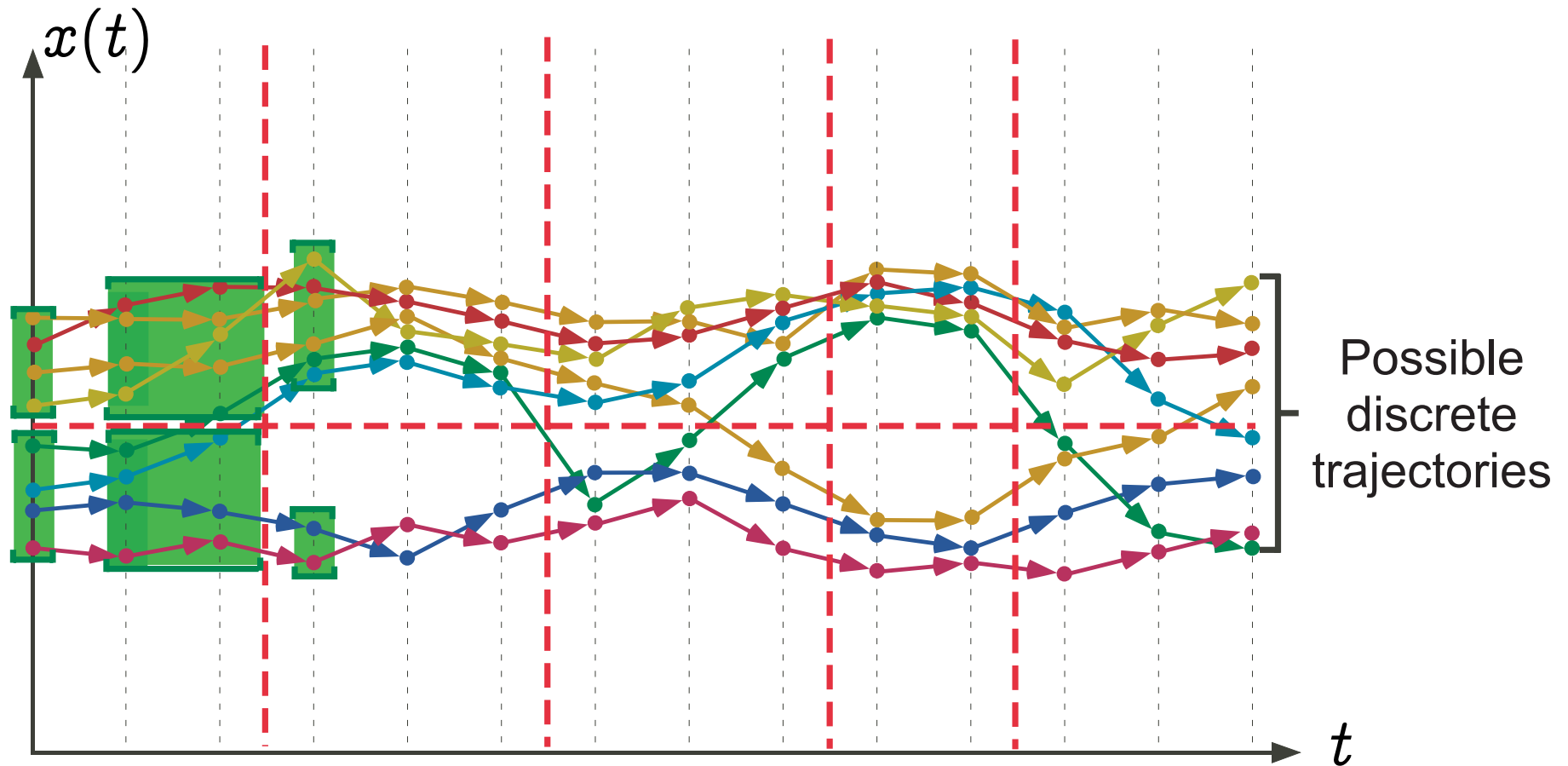
Graphic example: partitionned upward iteration with widening



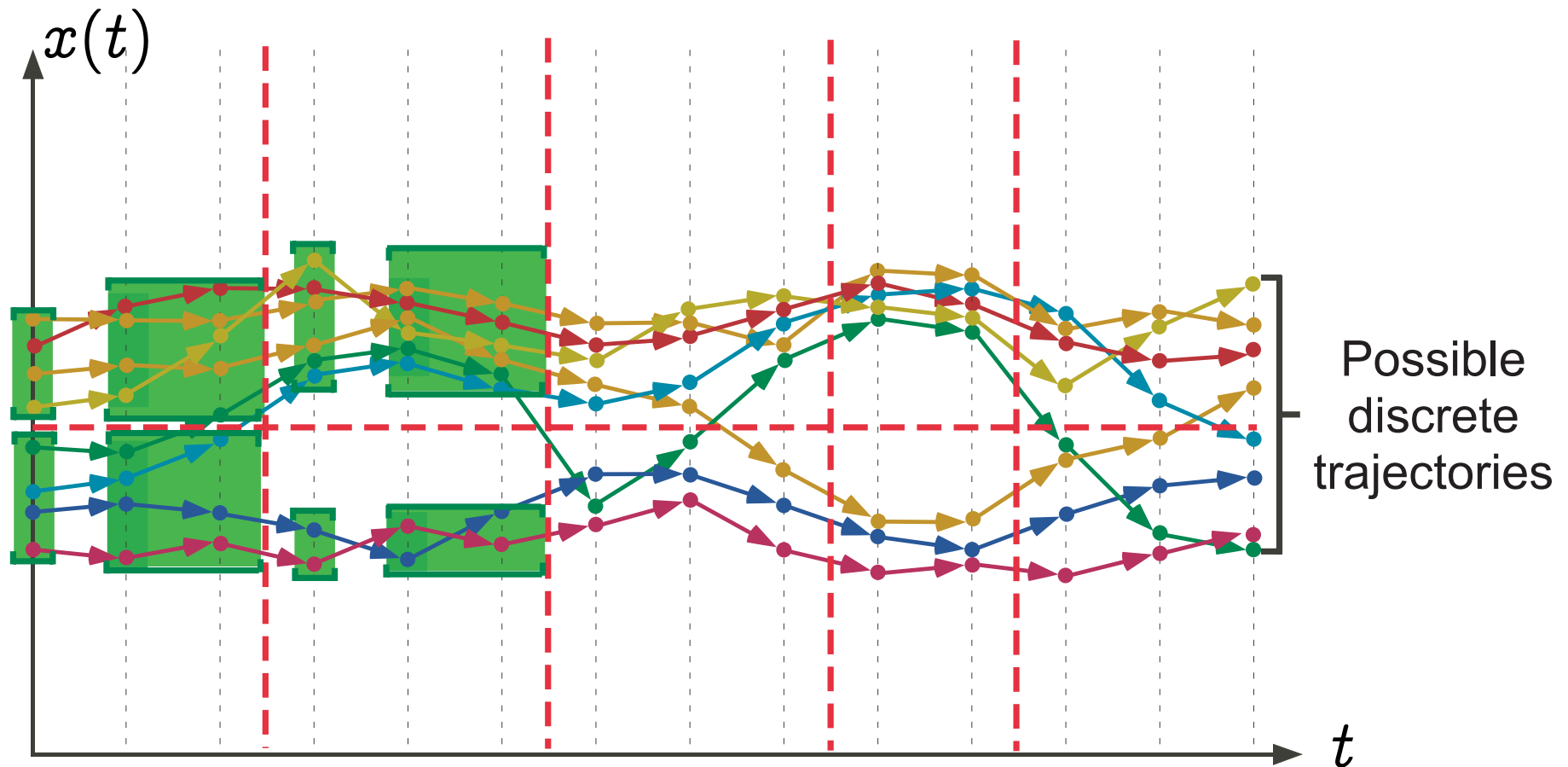
Graphic example: partitionned upward iteration with widening



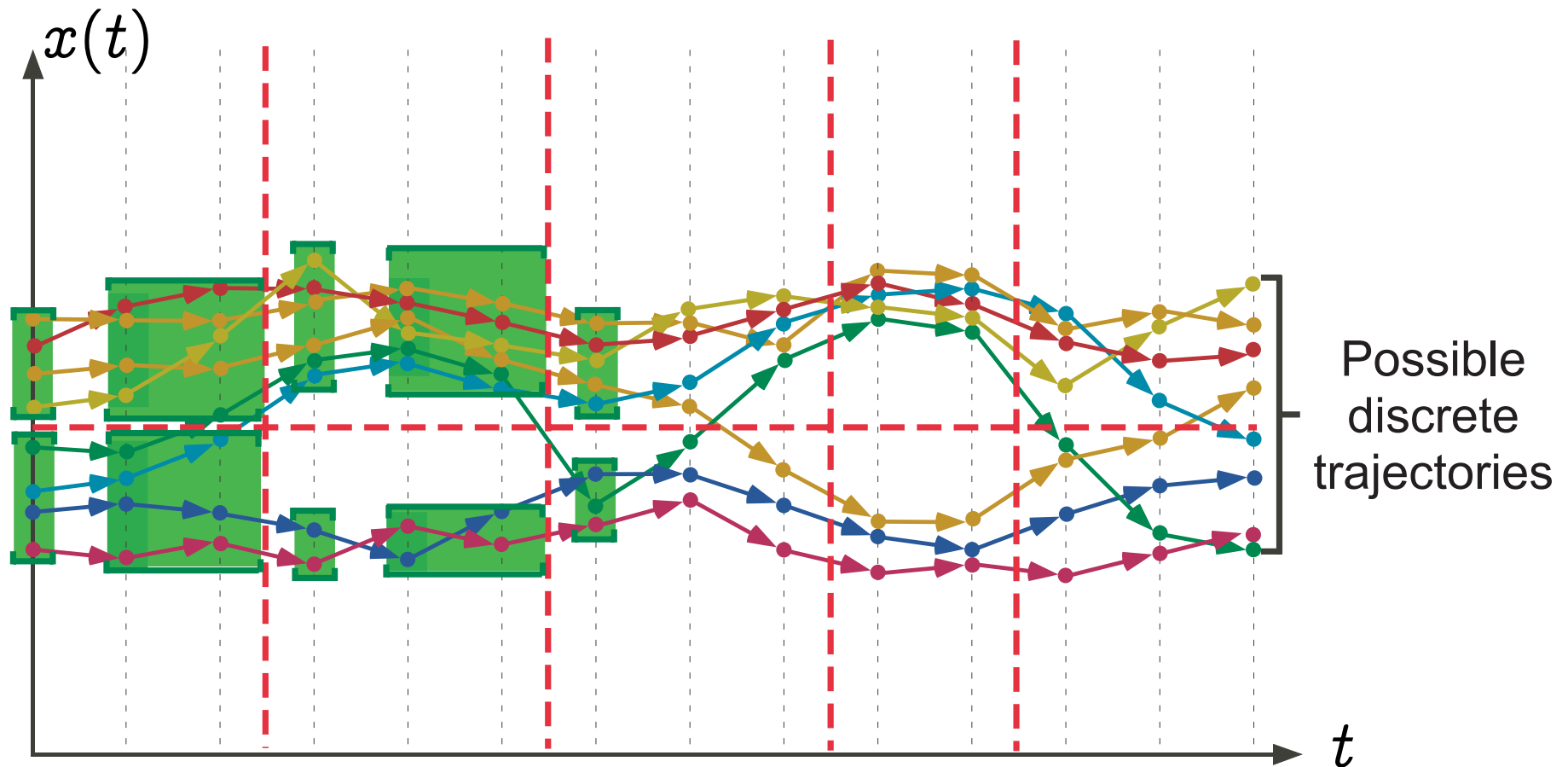
Graphic example: partitionned upward iteration with widening



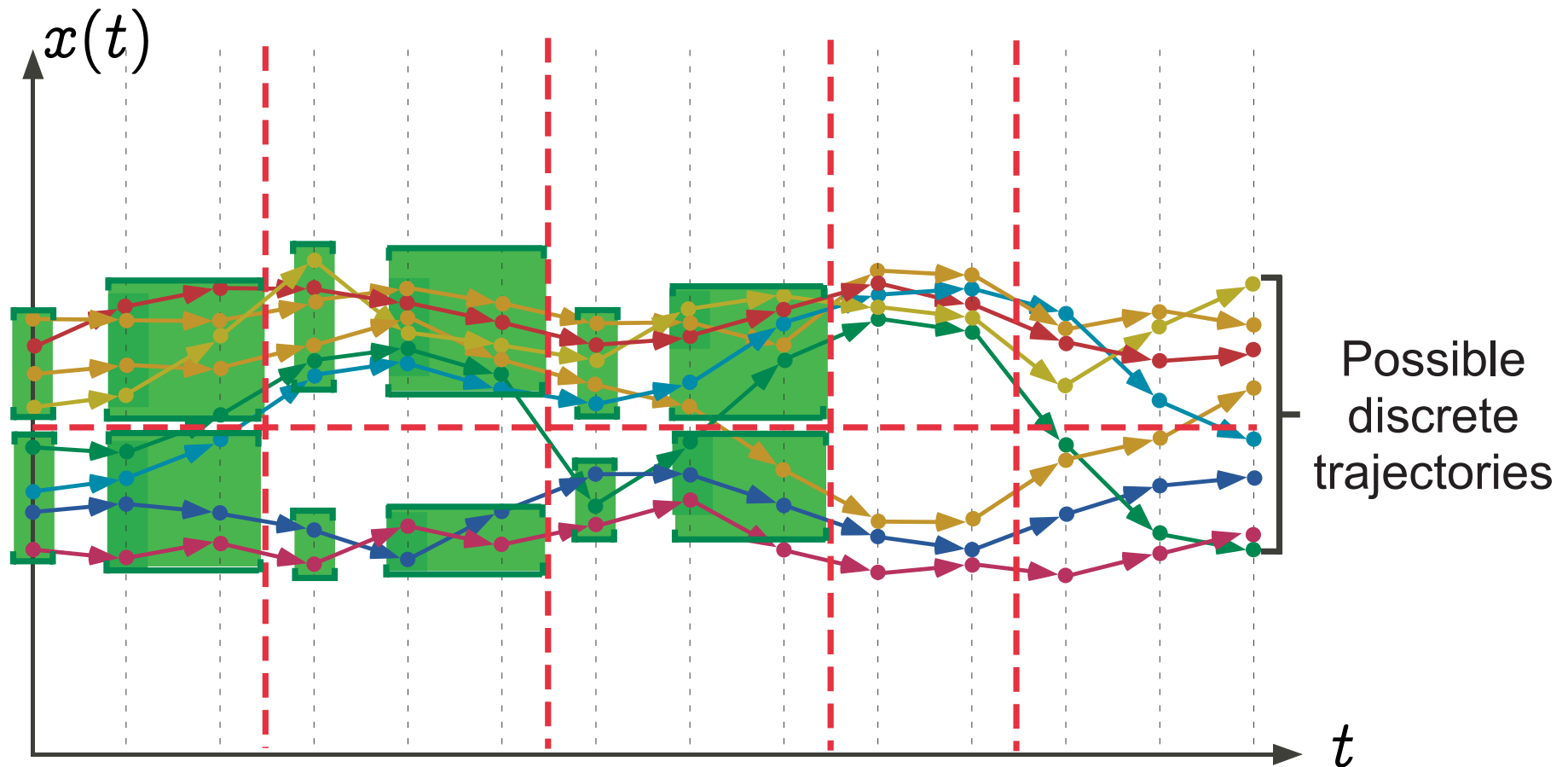
Graphic example: partitionned upward iteration with widening



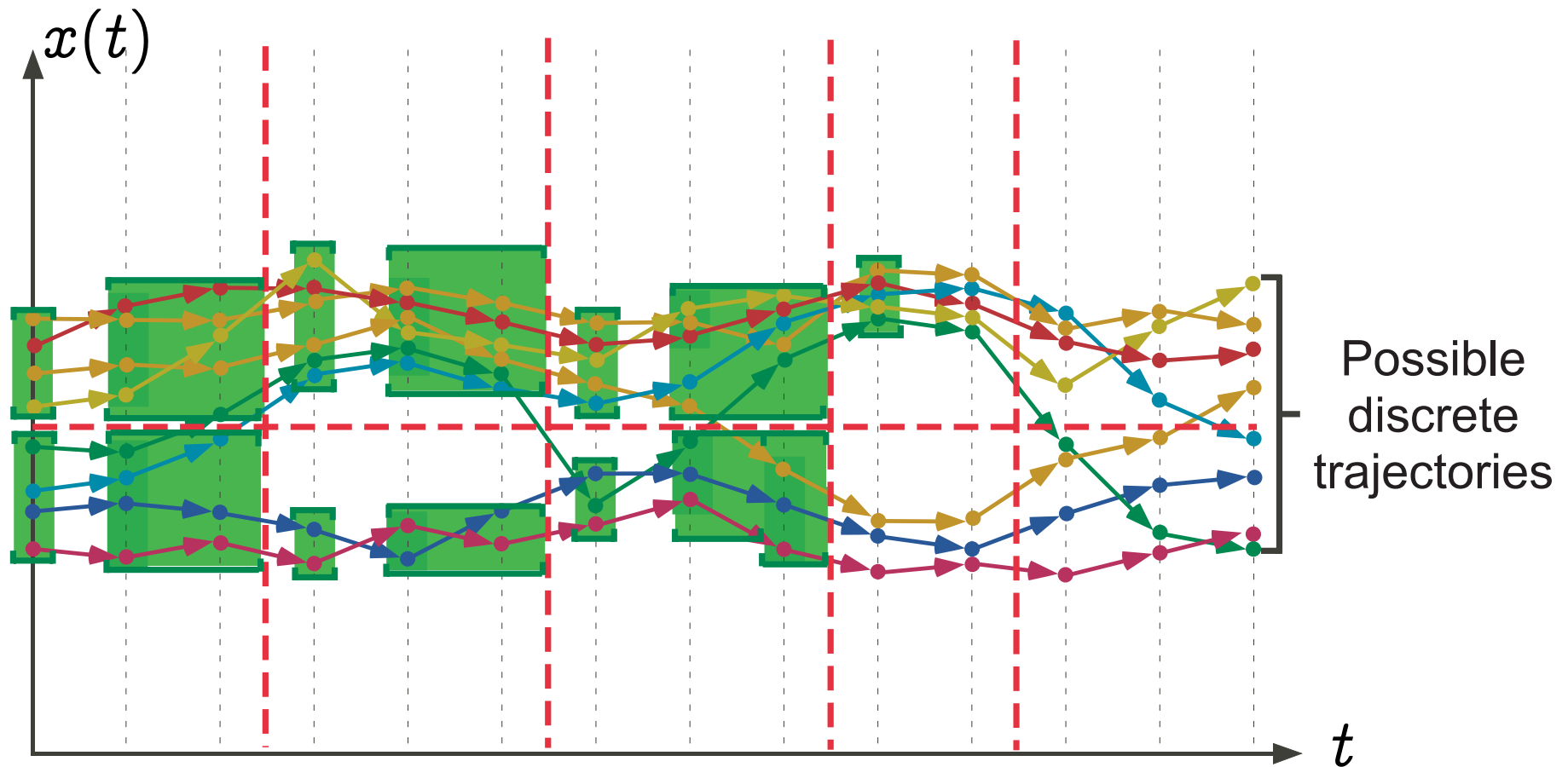
Graphic example: partitionned upward iteration with widening



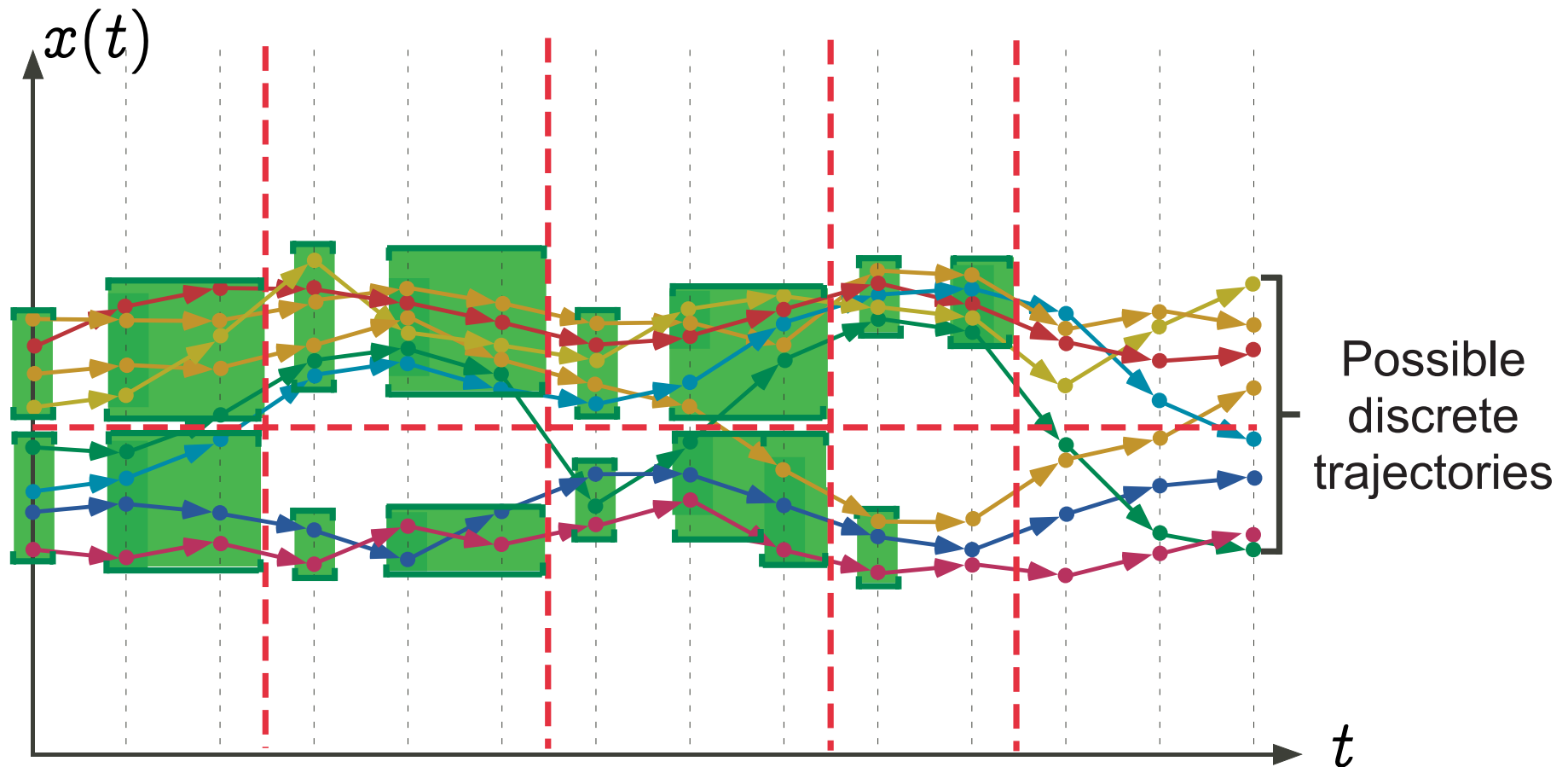
Graphic example: partitionned upward iteration with widening



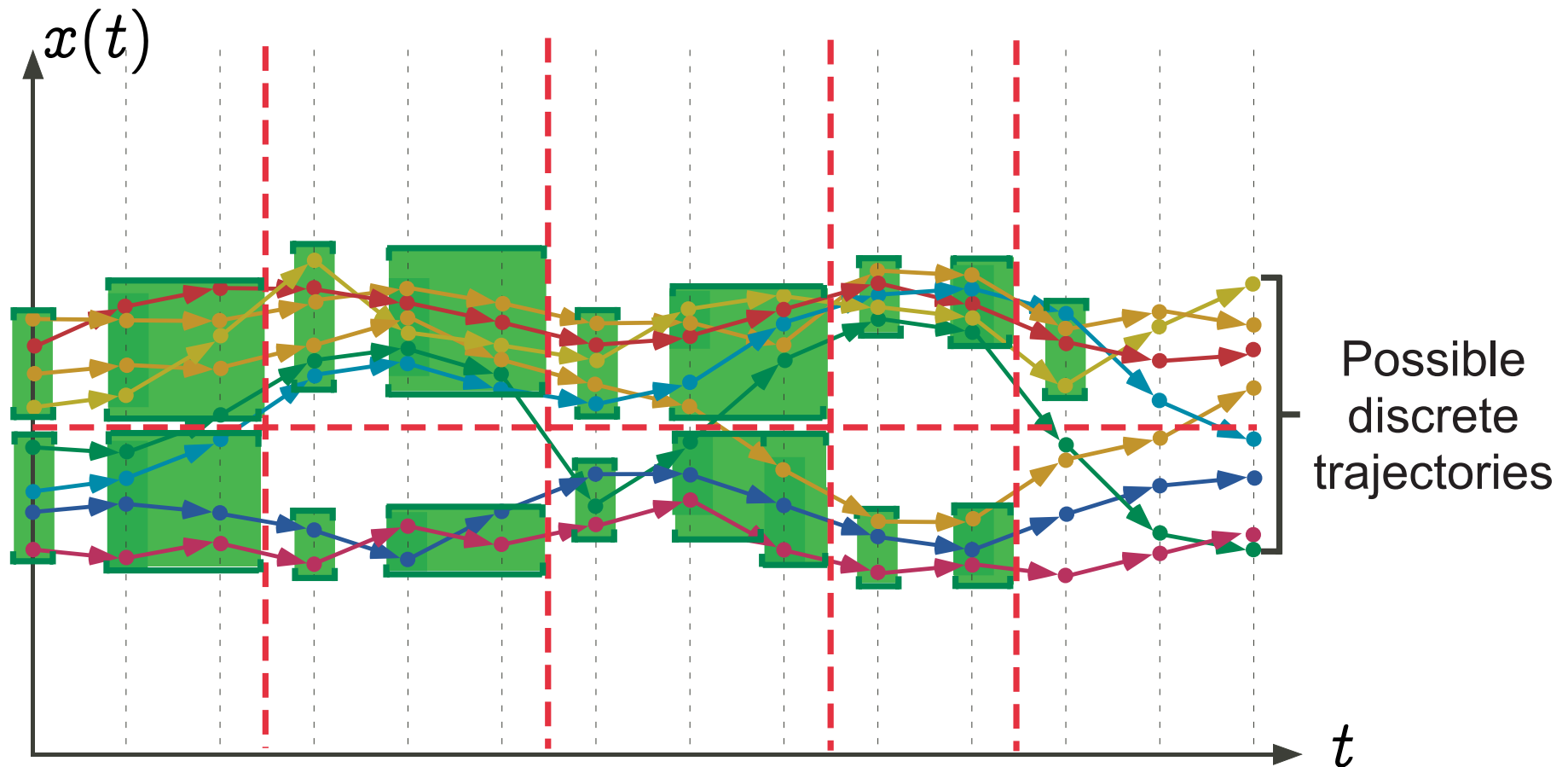
Graphic example: partitionned upward iteration with widening



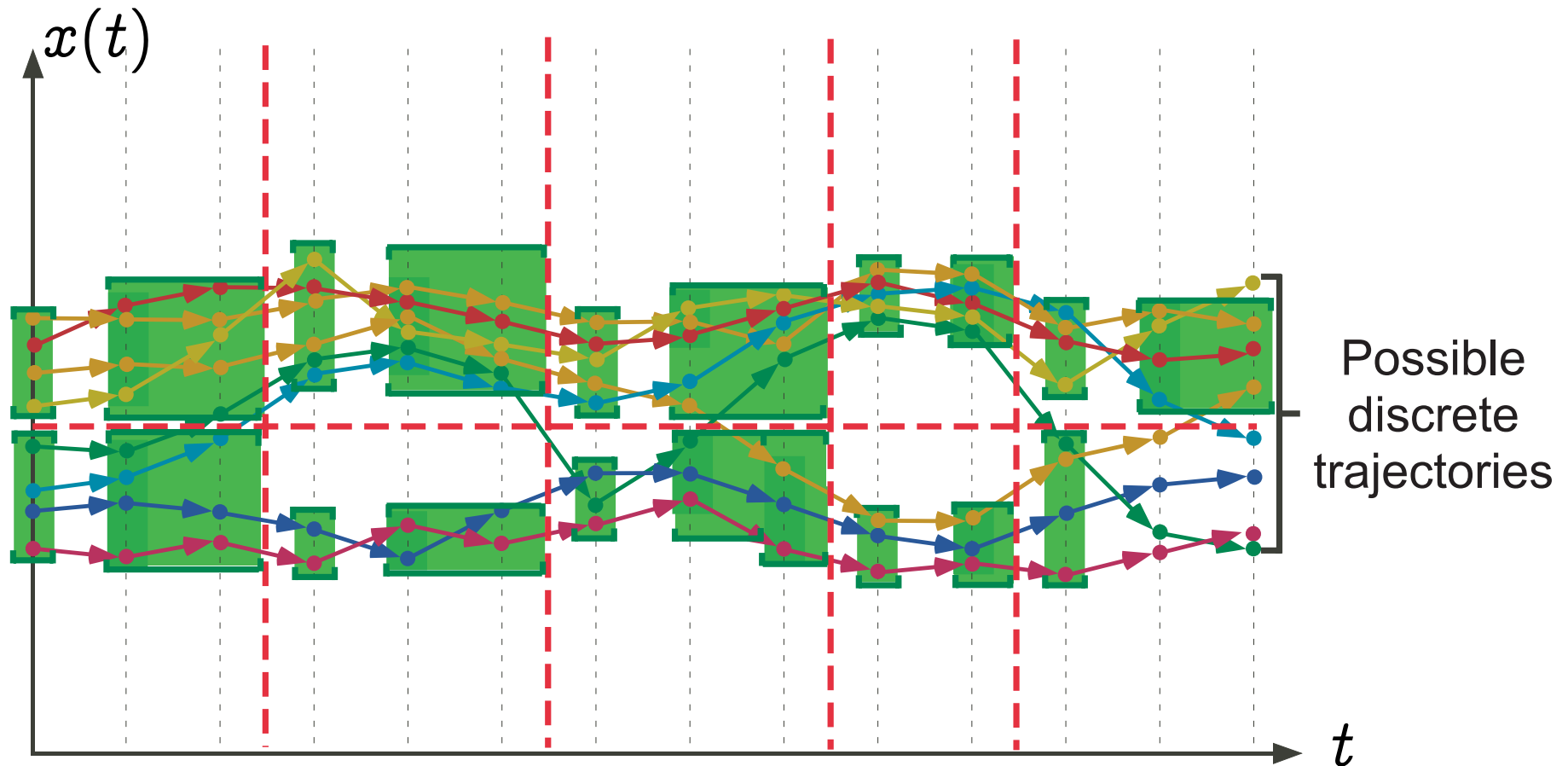
Graphic example: partitionned upward iteration with widening



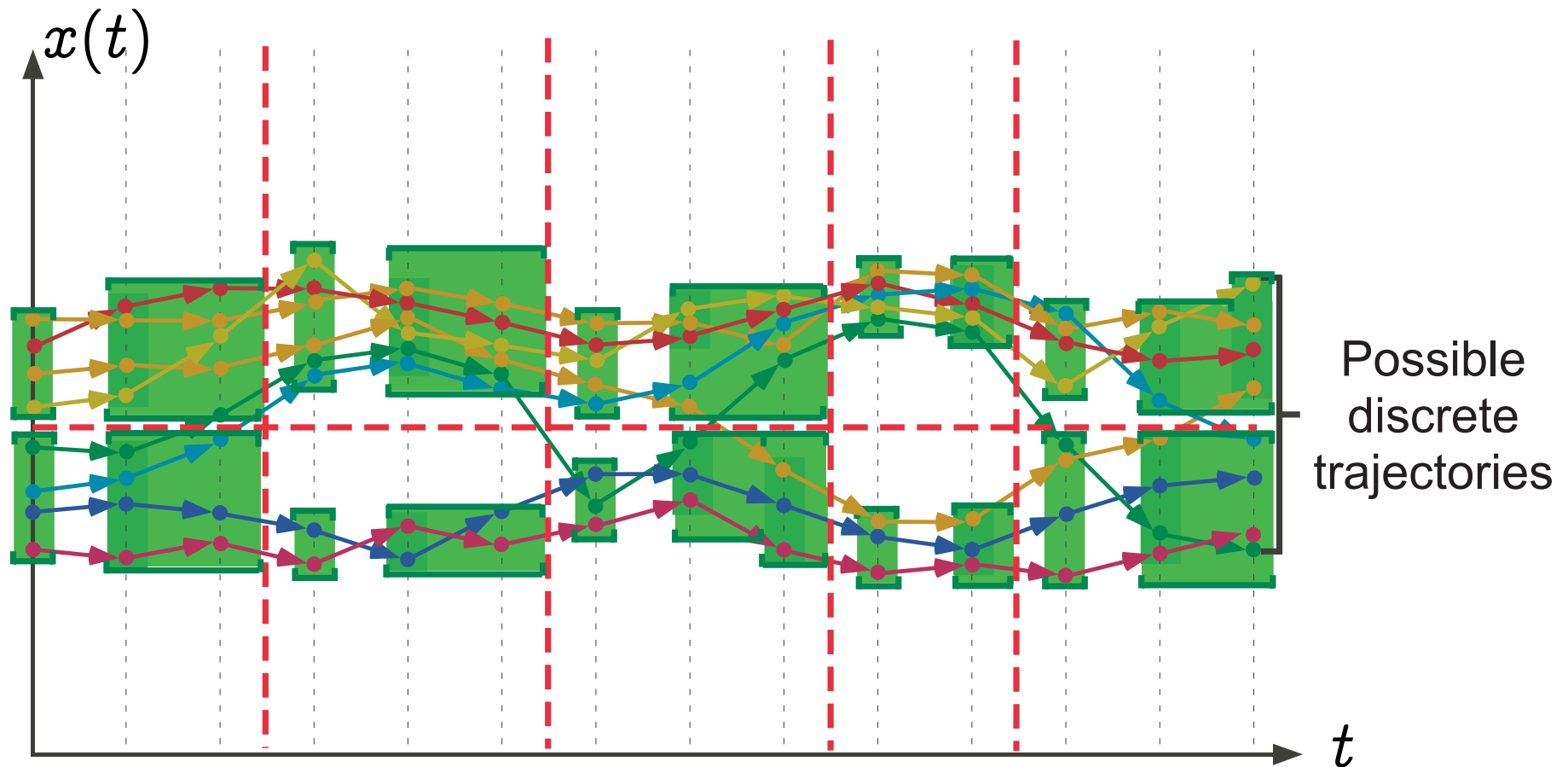
Graphic example: partitionned upward iteration with widening



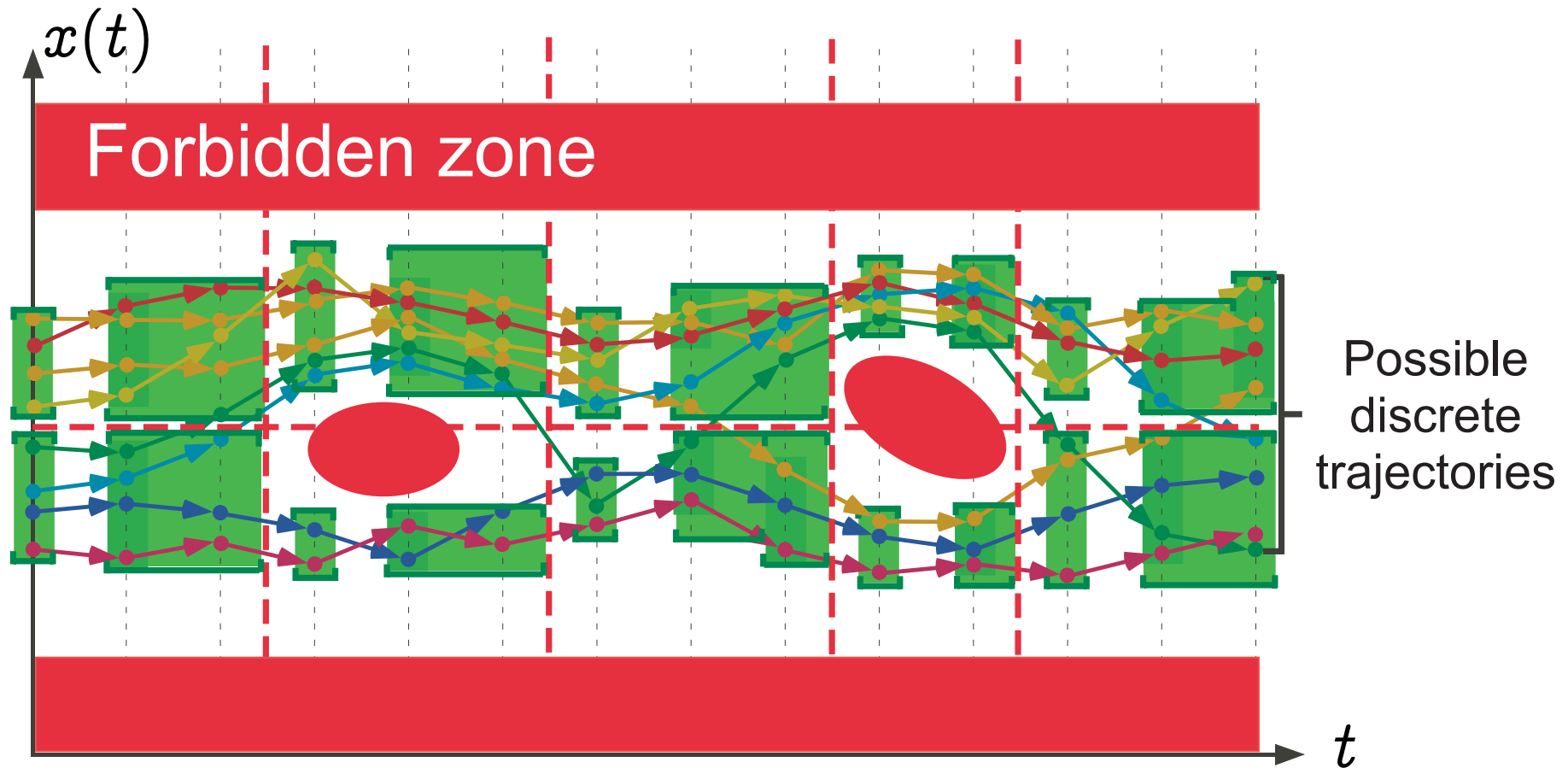
Graphic example: partitionned upward iteration with widening



Graphic example: partitionned upward iteration with widening



Graphic example: safety verification



Examples of partitionnings

- **sets of control states**: attach local information to program points instead of global information for the whole program/procedure/loop
- **sets of data states**:
 - case analysis (test, switches)
- **fixpoint iterates**:
 - widening with threshold set



Interval widening with threshold set

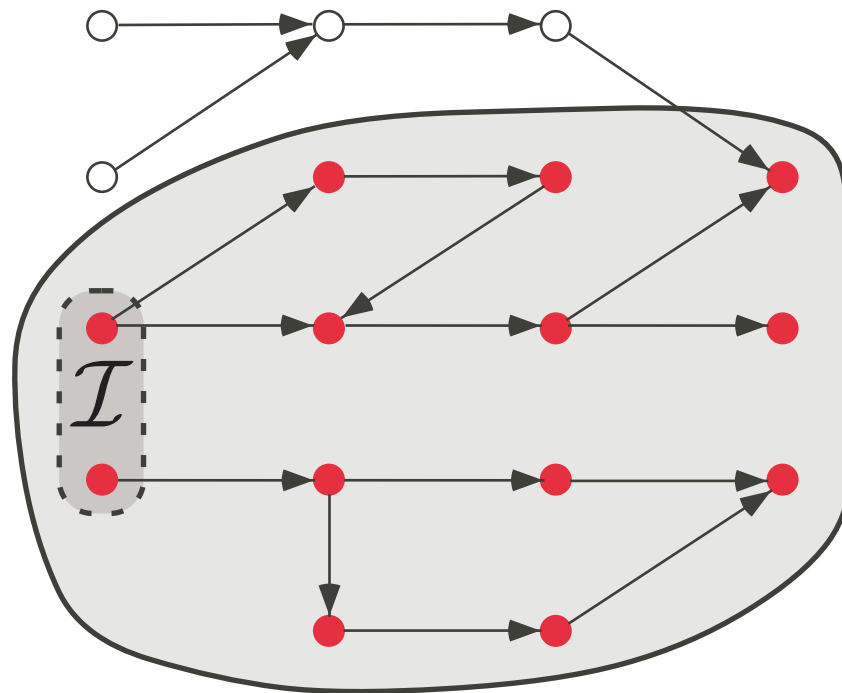
- The **threshold set** T is a finite set of numbers (plus $+\infty$ and $-\infty$),
- $[a, b] \nabla_T [a', b'] = [if\ a' < a\ then\ \max\{l \in T \mid l \leq a'\}$
else $a,$
if $b' > b\ then\ \min\{h \in T \mid h \geq b'\}$
else $b]$.
- Examples (intervals):
 - sign analysis: $T = \{-\infty, 0, +\infty\}$;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



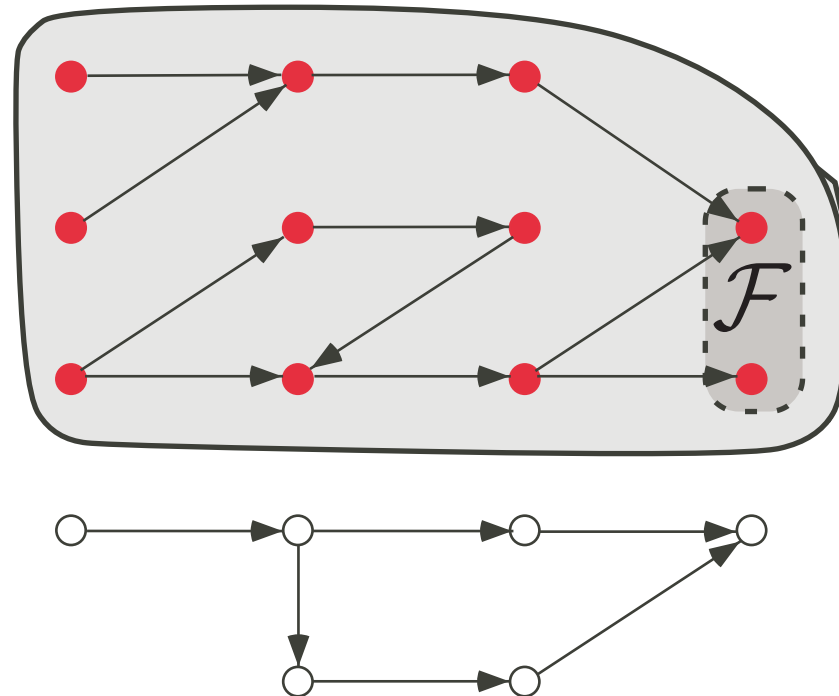
Combinations of abstractions



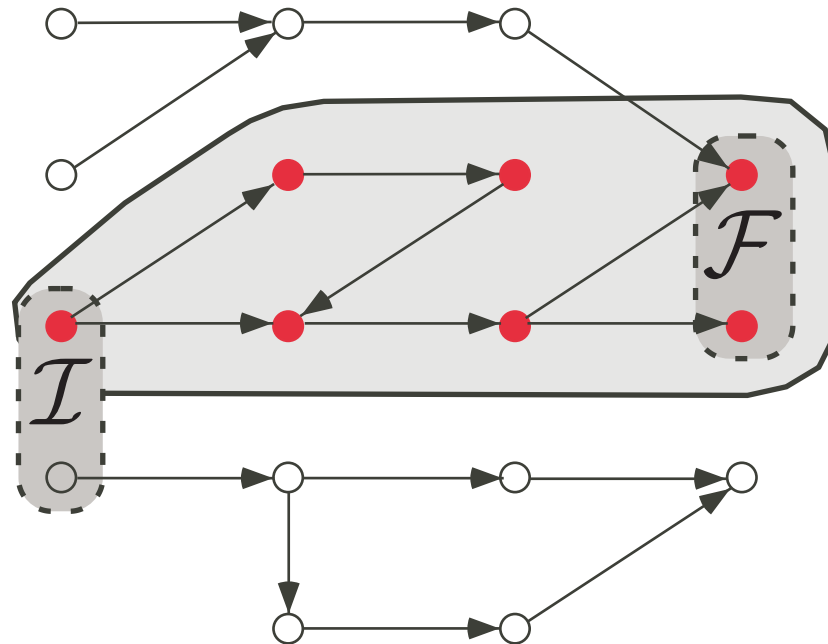
Forward/reachability analysis



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



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.



Conclusion



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]



Practical applications of abstract interpretation

- **Program analysis and manipulation**: a small rate of false alarms is acceptable
 - **AiT: worst case execution time** – Christian Ferdinand
- **Program verification**: no false alarms is acceptable
 - **TVLA: A system for generating abstract interpreters**
 - Mooly Sagiv
 - **Astrée: verification of absence of run-time errors** – Laurent Mauborgne



Industrial applications of abstract interpretation

- Both to **Program analysis and verification**
- Experience with the industrial use of abstract interpretation-based static analysis tools – Jean Souyris (Airbus France)



THE END

More references at URL www.di.ens.fr/~cousot.

