

Abstract Induction

ETH Zürich
October 2–3, 2015

Patrick Cousot

pcousot@cs.nyu.edu cs.nyu.edu/~pcousot

Concrete Induction

Software correctness proofs

- Any formal proof of a non-trivial program requires a reasoning by **mathematical induction** (e.g., following Turing, on the number of program execution steps):
- Invent an **inductive argument** (e.g. invariant, variant function), **the hardest part**
- Prove the **base case** and **inductive case** (e.g. true on loop entry and preserved by one more loop iteration)
- Prove that the inductive argument is **strong-enough**, that is, it implies the program property to be verified

Avoiding the difficulties:

(I) finitary methods

Avoiding the difficulty

- **Unsoundness**: not for scientists
- **Model-checking**: finite enumeration, no induction needed
- **Deductive methods** (theorem provers, proof verifiers, SMT solvers): avoid (part of) the difficulty since the inductive argument must be provided by the end-user (\Rightarrow still difficult, shame is on the prover)
- **Finitary abstractions** (predicate abstraction \equiv any finite abstract domain): only finitely many possible statements to be checked to be inductive

Limitations of finite abstractions

- A sound and complete finite abstraction exists to prove any property of any **program**:

$x=0; \text{ while } x<1 \text{ do } x++ \longrightarrow \{\perp, [0,0], [0,1], [-\infty, \infty]\}$

$x=0; \text{ while } x<2 \text{ do } x++ \longrightarrow \{\perp, [0,0], [0,1], [0,2], [-\infty, \infty]\}$

...

$x=0; \text{ while } x<n \text{ do } x++ \longrightarrow \{\perp, [0,0], [0,1], [0,2], [0,3], \dots, [0,n], [-\infty, \infty]\}$

...

- Not true for a **programming language** !
- Finite abstractions fail on infinitely many programs on which infinitary abstractions do succeed

Avoiding the difficulty

(II) Refinement in finite domains

Verification/static analysis by abstract interpretation

- Define the **abstraction**:

$$\langle \wp(\mathcal{D} \llbracket P \rrbracket), \subseteq \rangle \xrightleftharpoons[\alpha \llbracket P \rrbracket]{\gamma \llbracket P \rrbracket} \langle \mathcal{A} \llbracket P \rrbracket, \sqsubseteq \rangle$$

- Calculate the **abstract semantics**:

$$S^\# \llbracket P \rrbracket = \alpha \llbracket P \rrbracket (\{ S \llbracket P \rrbracket \}) \quad \text{exact abstraction}$$

$$S^\# \llbracket P \rrbracket \sqsupseteq \alpha \llbracket P \rrbracket (\{ S \llbracket P \rrbracket \}) \quad \text{approximate abstraction}$$

- Soundness** (by construction):

$$\forall P \in \mathbb{L}: \forall Q \in \mathcal{A}: S^\# \llbracket P \rrbracket \sqsubseteq Q \implies S \llbracket P \rrbracket \in \gamma \llbracket P \rrbracket (Q)$$

Refinement: good news

- **Problem**: how to prove a valid abstract property $\alpha(\{\text{lfp } F \llbracket P \rrbracket\}) \sqsubseteq Q$ when $\alpha \circ F \sqsubseteq F^\# \circ \alpha$ but $\text{lfp } F^\# \llbracket P \rrbracket \not\sqsubseteq Q$? (i.e. strongest inductive argument too weak)
- It is **always** possible to refine $\langle \mathcal{A}, \sqsubseteq \rangle$ into a most abstract more precise abstraction $\langle \mathcal{A}', \sqsubseteq' \rangle$ such that

$$\langle \wp(\mathcal{D}), \sqsubseteq \rangle \xrightleftharpoons[\alpha']{\gamma'} \langle \mathcal{A}', \sqsubseteq' \rangle$$

and $\alpha' \circ F = F' \circ \alpha$ with $\text{lfp } F' \llbracket P \rrbracket \sqsubseteq' \alpha' \circ \gamma(Q)$

(thus proving $\text{lfp } F \llbracket P \rrbracket \in \gamma'(Q)$ which implies $\text{lfp } F \llbracket P \rrbracket \in \gamma(Q)$)

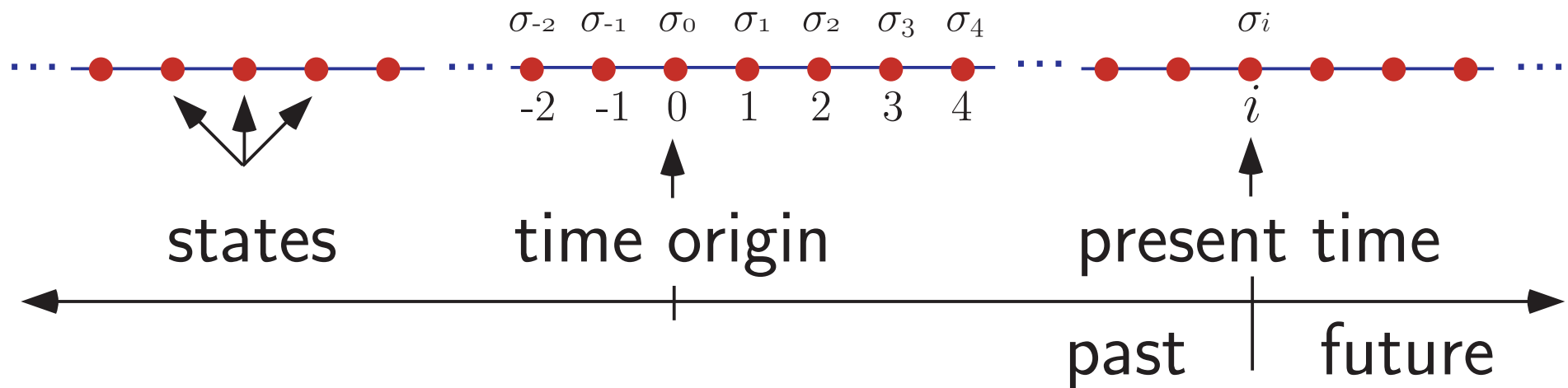
Roberto Giacobazzi, Francesco Ranzato, Francesca Scozzari: Making abstract interpretations complete. J. ACM 47(2): 361-416 (2000)

Refinement: bad news

- But, refinements of an abstraction can be **intrinsically incomplete**
- The only complete refinement of that abstraction for the collecting semantics is :
the identity (i.e. no abstraction at all)
- In that case, the only complete refinement of the abstraction is to the collecting semantics and any other refinement is always imprecise

Example of intrinsic approximate refinement

- Consider executions **traces** $\langle i, \sigma \rangle$ with infinite past and future:



Example of intrinsic approximate refinement

- Consider the temporal specification language μ^* (containing LTL, CTL, CTL*, and Kozen's μ -calculus as fragments):

$\varphi ::=$	σ_S	$S \in \wp(\mathbb{S})$	state predicate
	π_t	$t \in \wp(\mathbb{S} \times \mathbb{S})$	transition predicate
	$\oplus \varphi_1$		next
	φ_1^\sim		reversal
	$\varphi_1 \vee \varphi_2$		disjunction
	$\neg \varphi_1$		negation
	X	$X \in \mathbb{X}$	variable
	$\mu X \cdot \varphi_1$		least fixpoint
	$\nu X \cdot \varphi_1$		greatest fixpoint
	$\forall \varphi_1 : \varphi_2$		universal state closure

Example of intrinsic approximate refinement

- Consider **universal model-checking abstraction**:

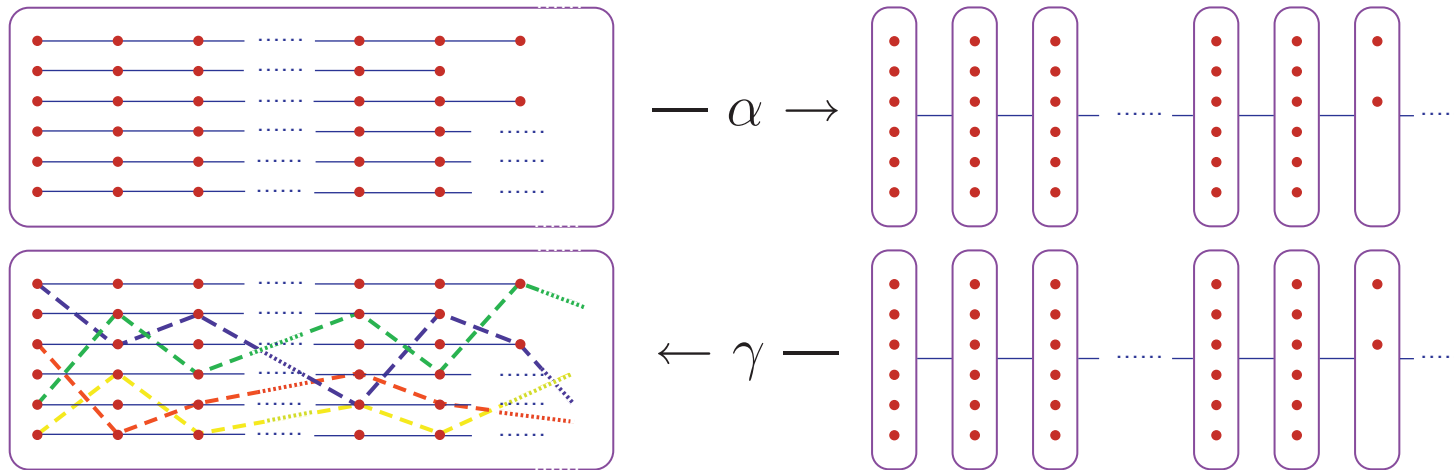
$$\begin{aligned}\text{MC}_M^\forall(\phi) &= \alpha_M^\forall(\llbracket \phi \rrbracket) \in \wp(\text{Traces}) \rightarrow \wp(\text{States}) \\ &= \{s \in \text{States} \mid \forall \langle i, \sigma \rangle \in \text{Traces}_M. (\sigma_i = s) \Rightarrow \\ &\quad \langle i, \sigma \rangle \in \llbracket \phi \rrbracket\}\end{aligned}$$

where M is defined by a transition system

(and dually the existential model-checking abstraction)

Example of intrinsic approximate refinement

- The abstraction from a set of traces to a trace of sets is sound but *incomplete*, even for finite systems (*)

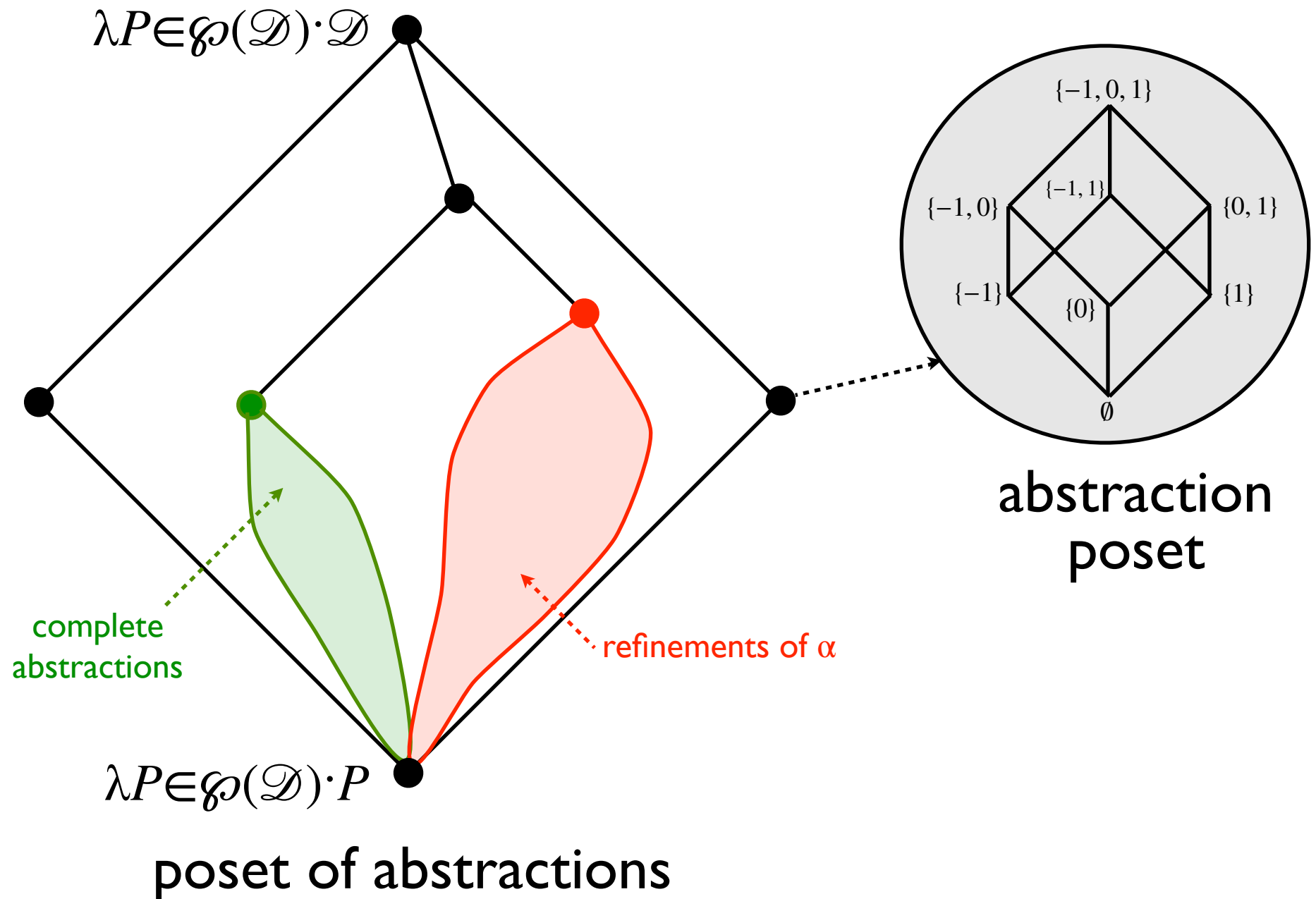


- Any refinement of this abstraction is *incomplete* (but to the infinite past/future trace semantics itself) (**)

(*) Patrick Cousot, Radhia Cousot: Temporal Abstract Interpretation. POPL 2000: 12-25

(**) Roberto Giacobazzi, Francesco Ranzato: Incompleteness of states w.r.t. traces in model checking. Inf. Comput. 204(3): 376-407 (2006)

Intrinsic approximate refinement



In general refinement does not terminate

- Narrowing is needed to stop **infinite iterated automatic refinements**:

e.g. SLAM stops refinement after 20mn, now abandoned (despite complete success claimed in 98% of studied cases ^(*))

- **Intelligence is needed** for refinement:

e.g. human-driven refinement of Astrée ^(**)

(*) Thomas Ball, Vladimir Levin, Sriram K. Rajamani: A decade of software model checking with SLAM. *Commun. ACM* 54(7): 68-76 (2011)

(**) Julien Bertrane, Patrick Cousot, Radhia Cousot, Jérôme Feret, Laurent Mauborgne, Antoine Miné, & Xavier Rival. Static Analysis and Verification of Aerospace Software by Abstract Interpretation. In *AIAA Infotech@Aerospace 2010*, Atlanta, Georgia. American Institute of Aeronautics and Astronautics, 20—22 April 2010. © AIAA.

Facing the difficulties: Abstract induction

Sound software static analysis

- The **mathematical induction** must be performed in the **abstract** (e.g. the inductive argument must belong to an abstract domain with a finite computer representation)
- (and imply the mathematical induction in the **concrete**)

Abstract induction

- The **inductive argument** must be expressible in the abstract domain (complex abstract domains favored)
- It must be **strong enough** to imply the program property (complex abstract domains favored)
- It must be **inferable in the abstract** (simple abstract domains favored)

Abstract induction in infinite domains

Abstract Interpreters

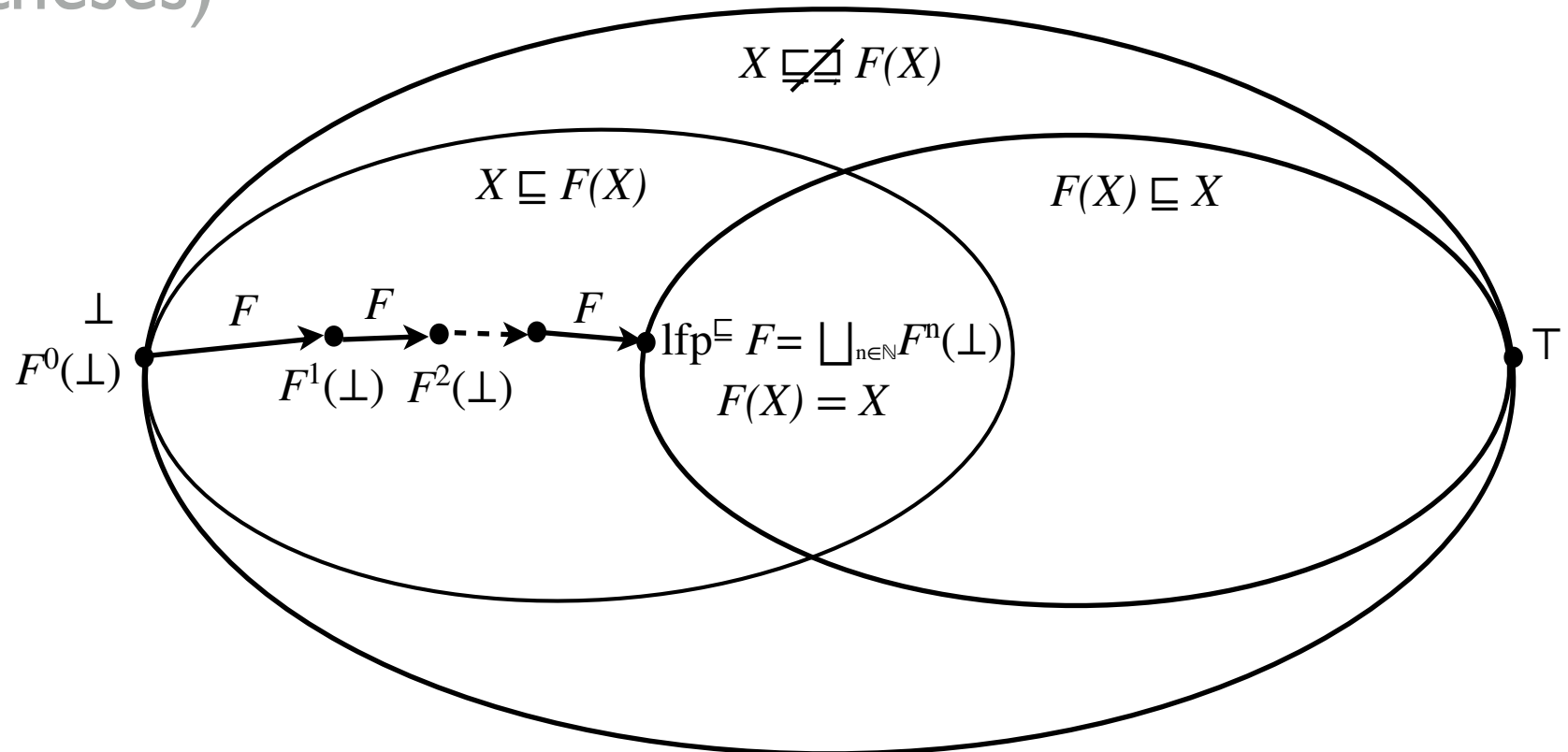
- **Transitional abstract interpreters**: proceed by induction on program steps
- **Structural abstract interpreters**: proceed by induction on the program syntax
- **Common main problem**: over/under-approximate fixpoints in non-Noetherian^(*) abstract domains ^(**)

(*) Iterative fixpoint computations may not converge in finitely many steps

(**) Or convergence may be guaranteed but too slow.

Fixpoints

- Poset (or pre-order) $\langle D, \sqsubseteq, \perp, \top \rangle$
- Transformer (increasing in the concrete) $F \in D \mapsto D$
- Least fixpoint: $\text{lfp}^{\sqsubseteq} F = \bigsqcup_{n \in \mathbb{N}} F^n(\perp)$ (under appropriate hypotheses)



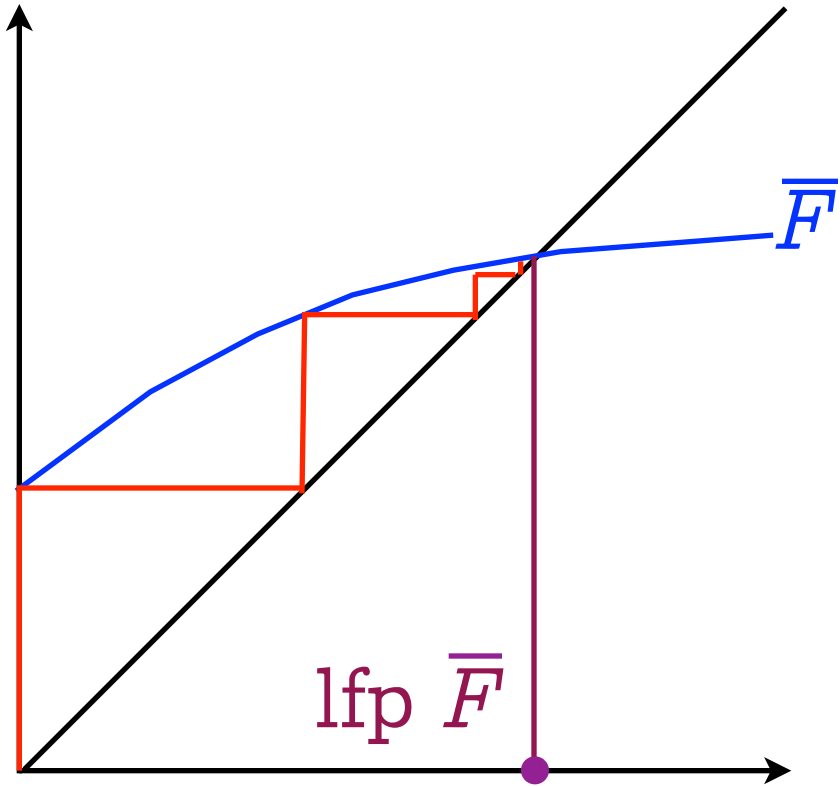
Convergence criterion

- By Tarski (or variants)

$$F(X) \sqsubseteq X \quad \Rightarrow \quad \text{lfp}^\sqsubseteq F \sqsubseteq X$$

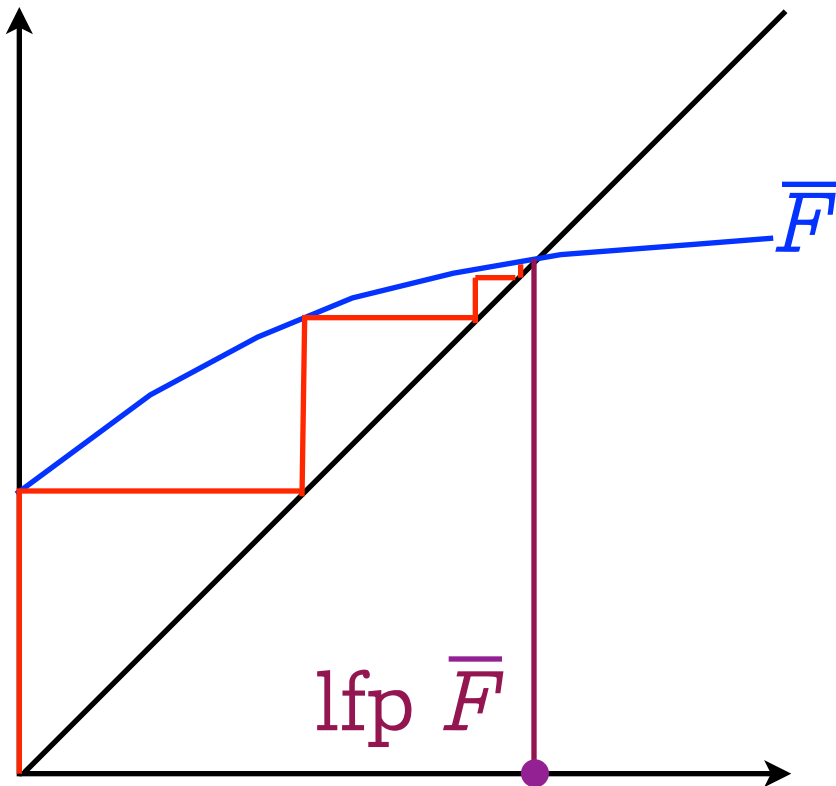
Widening

Convergence acceleration with widening

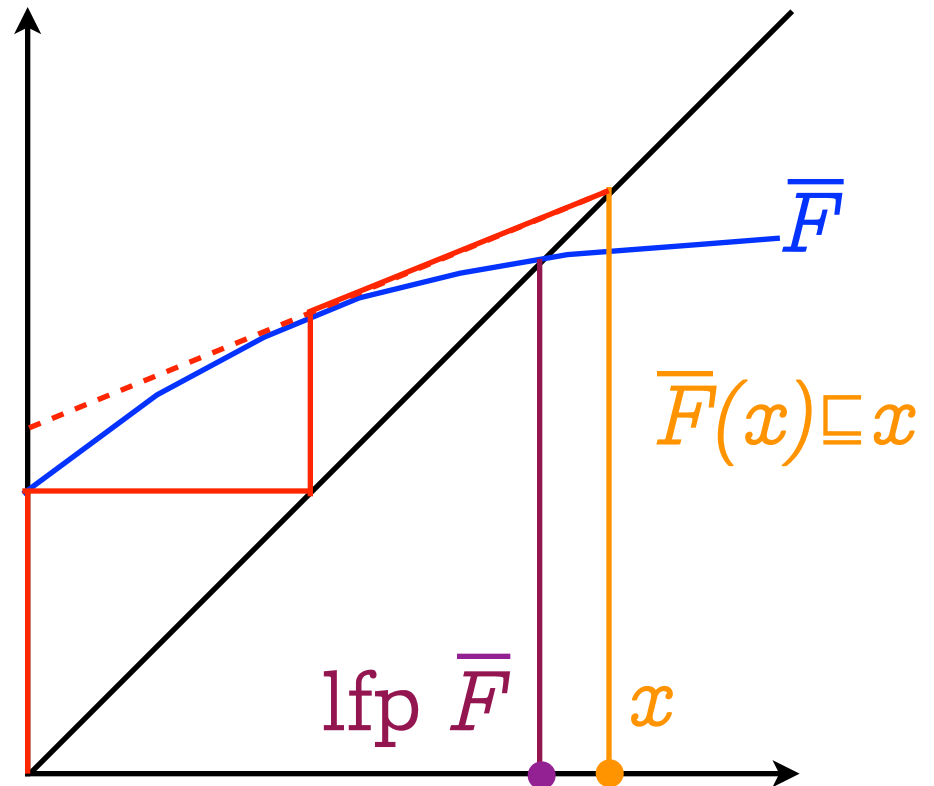


Infinite iteration

Convergence acceleration with widening



Infinite iteration



Accelerated iteration with widening
(e.g. with a widening based on the derivative
as in Newton-Raphson method^(*))

^(*) Javier Esparza, Stefan Kiefer, Michael Luttenberger: Newtonian program analysis. J. ACM 57(6): 33 (2010)

Extrapolation by Widening

- $X^0 = \perp$ (increasing iterates with widening)
- $X^{n+1} = X^n \nabla F(X^n)$ when $F(F(X^n)) \not\subseteq F(X^n)$
- $X^{n+1} = F(X^n)$ when $F(F(X^n)) \subseteq F(X^n)$
- Widening ∇ , two independent *hypotheses*:
- $Y \subseteq X \nabla Y$ (*extrapolation*)
- Enforces *convergence* of increasing iterates with widening (to a limit X^ℓ)

The oldest widenings

- Primitive widening [1,2]

```

(x  $\bar{\nabla}$  y) = cas x  $\in$  Va, y  $\in$  Va dans
|
|  $\square$ , ?  $\Rightarrow$  y ;
| ? ,  $\square$   $\Rightarrow$  x ;
| [n1, m1], [n2, m2]  $\Rightarrow$ 
|   [si n2 < n1 alors - $\infty$  sinon n1 fsi ;
|     si m2 > m1 alors + $\infty$  sinon m1 fsi] ;
| fincas ;
  
```

```

[a1, b1]  $\bar{\nabla}$  [a2, b2] =
  [if a2 < a1 then - $\infty$  else a1 fi,
   if b2 > b1 then + $\infty$  else b1 fi]
  
```

- Widening with thresholds [3]

$\forall x \in \bar{L}_2, \perp \nabla_2(j) x = x \nabla_2(j) \perp = x$

$[l_1, u_1] \nabla_2(j) [l_2, u_2]$

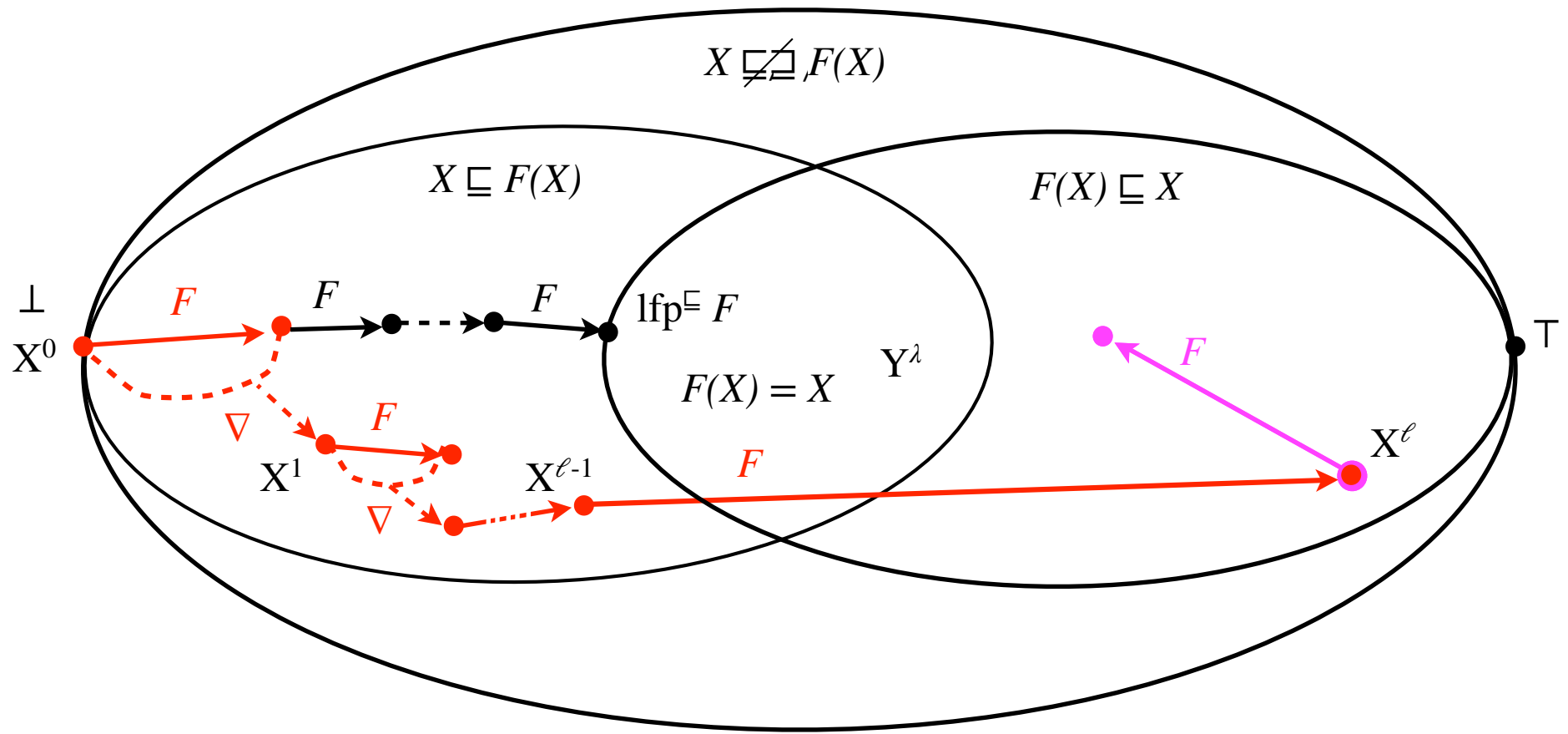
$=$ [if $0 \leq l_2 < l_1$ then 0 elsif $l_2 < l_1$ then $-b - 1$ else l_1 fi,
 if $u_1 < u_2 \leq 0$ then 0 elsif $u_1 < u_2$ then b else u_1 fi]

[1] Patrick Cousot, Radhia Cousot: Vérification statique de la cohérence dynamique des programmes, Rapport du contrat IRIA-SESORI No 75-032, 23 septembre 1975.

[2] Patrick Cousot, Radhia Cousot: Abstract Interpretation: A Unified Lattice Model for Static Analysis of Programs by Construction or Approximation of Fixpoints. POPL 1977: 238-252

[3] Patrick Cousot, Semantic foundations of program analysis, Ch. 10 of Program flow analysis: theory and practice, N. Jones & S. Muchnick (eds), Prentice Hall, 1981.

Extrapolation with widening



Widenings are not increasing

- A **well-known** fact

$$[1,1] \subseteq [1,2] \text{ but } [1,1] \nabla [1,2] = [1, \infty] \not\subseteq [1,2] \nabla [1,2] = [1,2]$$

- A widening cannot both:
- Be **increasing** in its first parameter
- Enforce **termination** of the iterates
- Avoid **useless over-approximations** as soon as a solution is found^(*)

^(*) A counter-example is $x \nabla y = \top$

Narrowing

Interpolation with narrowing

- $Y^0 = X^\ell$ (decreasing iterates with narrowing)

$$Y^{n+1} = Y^n \Delta F(Y^n) \quad \text{when } F(F(Y^n)) \sqsubset F(Y^n)$$

$$Y^{n+1} = F(Y^n) \quad \text{when } F(F(Y^n)) = F(Y^n)$$

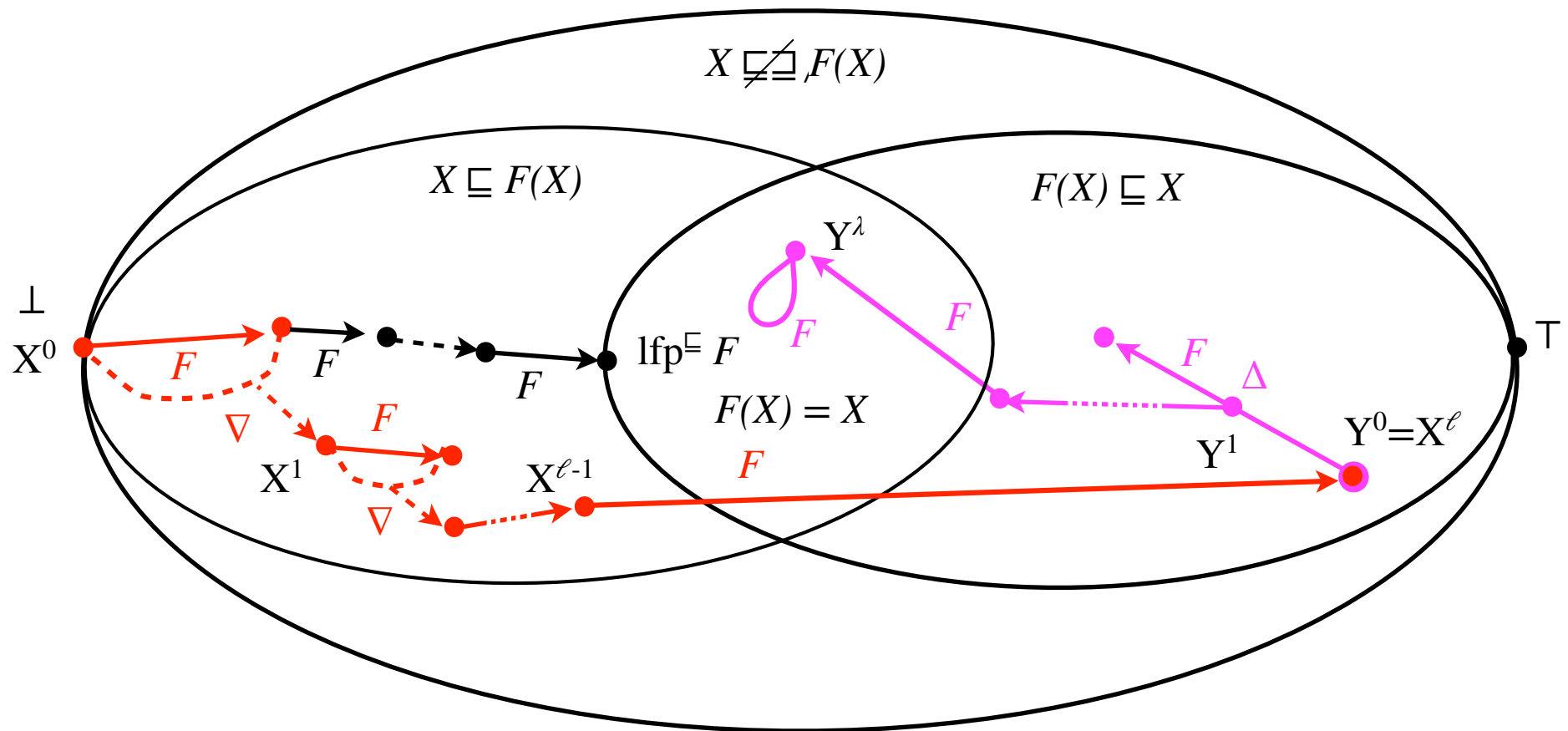
- **Narrowing** Δ , two independent *hypotheses*:
- $Y \sqsubseteq X \implies Y \sqsubseteq X \Delta Y \sqsubseteq X$ (*interpolation*)
- Enforces *convergence* of decreasing iterates with narrowing (to a limit Y^λ)

The oldest narrowing

- [2]

$$\begin{aligned} [a_1, b_1] \bar{\Delta} [a_2, b_2] = \\ [\text{if } a_1 = -\infty \text{ then } a_2 \text{ else } \text{MIN } (a_1, a_2), \\ \text{if } b_1 = +\infty \text{ then } b_2 \text{ else } \text{MAX } (b_1, b_2)] \end{aligned}$$

Interpolation with narrowing



Could stop when $F(X) \not\subseteq X \wedge F(F(X)) \subseteq F(X)$ but not the current practice.

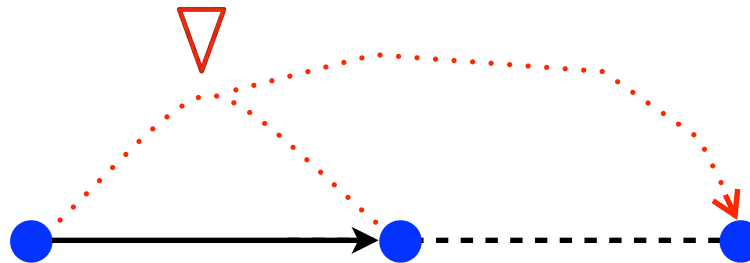
Duality

Duality

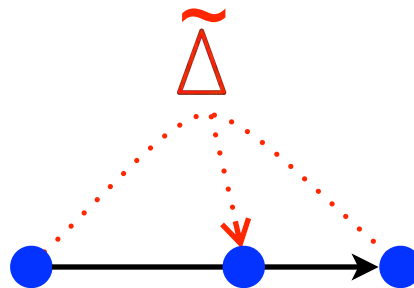
●	Convergence above the limit	Convergence below the limit
Increasing iteration	Widening ∇	Dual-narrowing $\tilde{\Delta}$
Decreasing iteration	Narrowing Δ	Dual widening $\tilde{\nabla}$

Extrapolators ($\nabla, \tilde{\nabla}$) and interpolators ($\Delta, \tilde{\Delta}$)

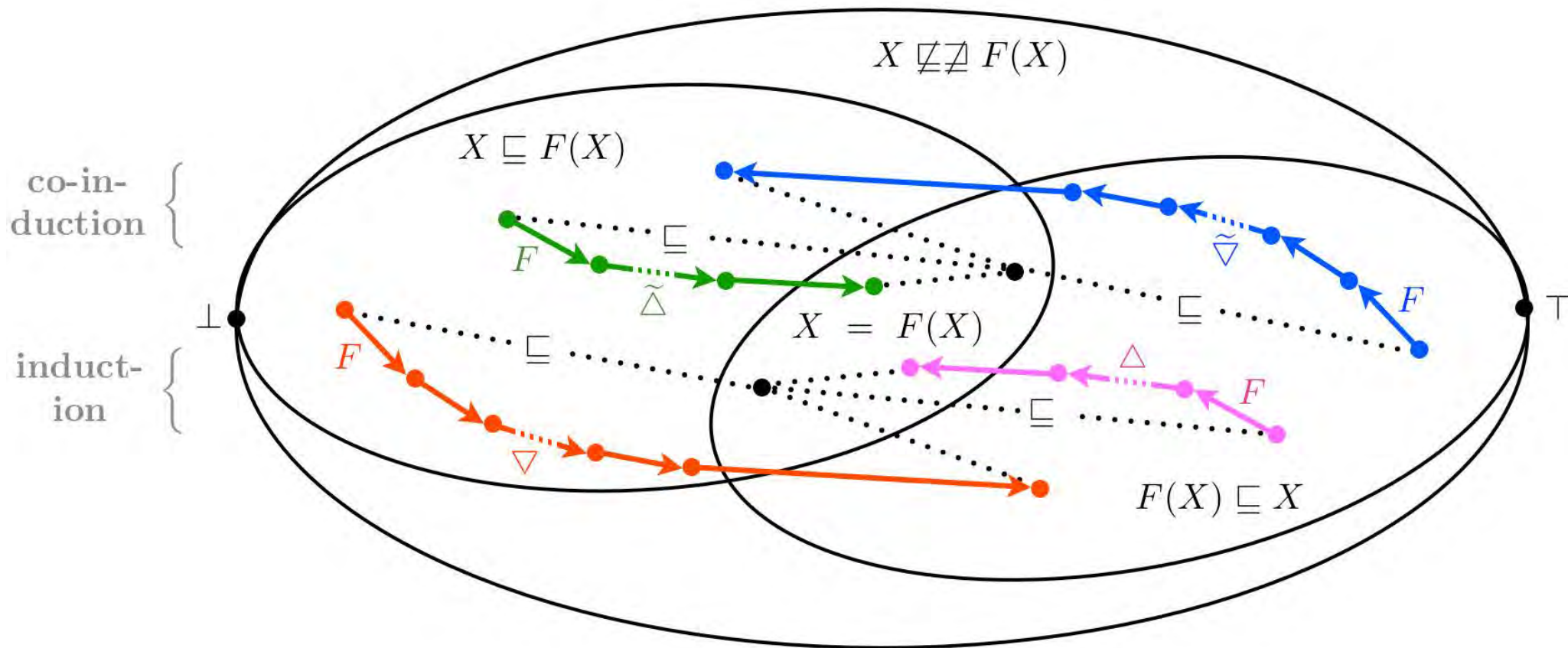
● Extrapolators:



● Interpolators:



Extrapolators, Interpolators, and Duals



Multi-step extrapolators/interpolators

- The extrapolators/interpolators can be on
 - the last two iterates
 - a bounded number of previous iterates
 - all **previous iterates**
- Examples:
 - loop unrolling
 - delayed widening
 - etc

Dual narrowing

Interpolation with dual narrowing

- $Z^0 = \perp$ (increasing iterates with dual-narrowing)

$$Z^{n+1} = F(Z^n) \tilde{\Delta} Y^\lambda \quad \text{when } F(F(Z^n)) \not\sqsubseteq F(Z^n)$$

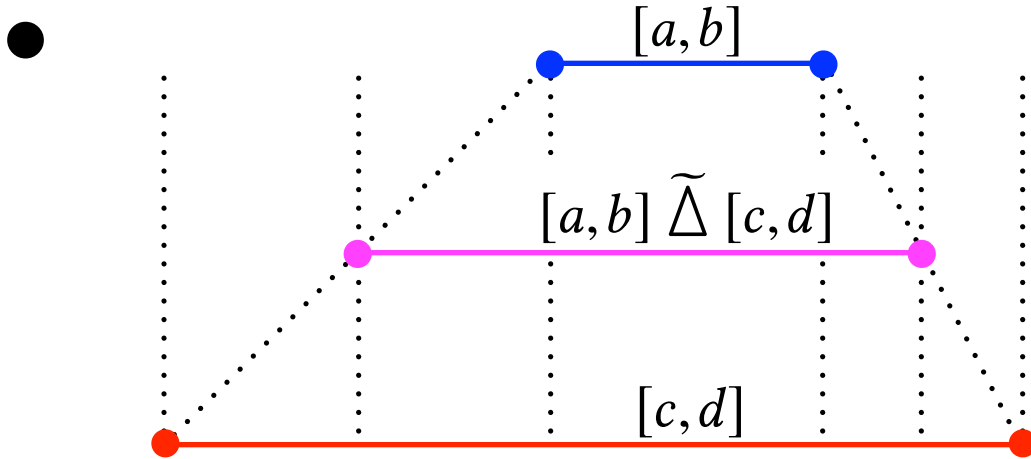
$$Z^{n+1} = F(Z^n) \quad \text{when } F(F(Z^n)) \sqsubseteq F(Z^n)$$

- Dual-narrowing $\tilde{\Delta}$, two independent *hypotheses*:

- $X \sqsubseteq Y \implies X \sqsubseteq Y \tilde{\Delta} X \sqsubseteq Y$ (*interpolation*)

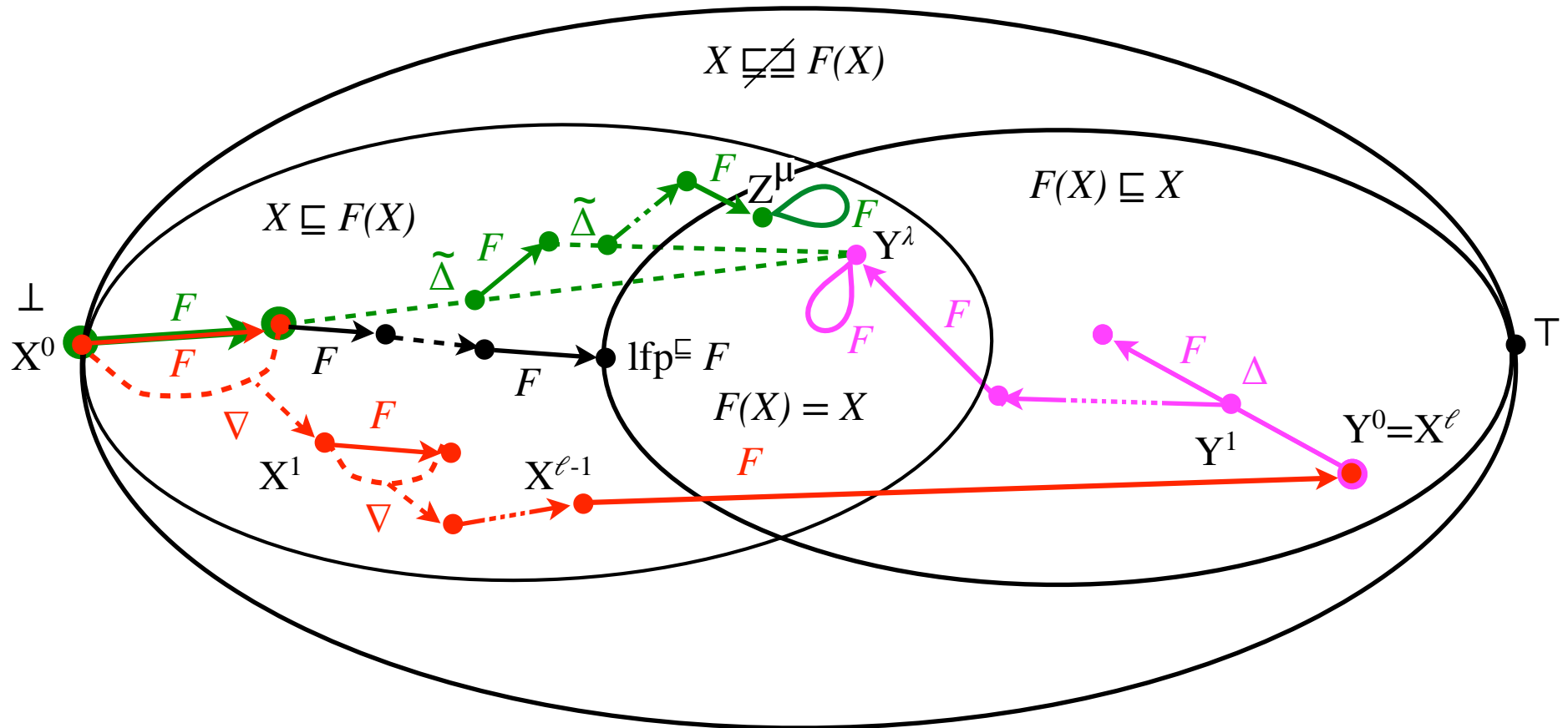
- Enforces *convergence* of increasing iterates with dual-narrowing

Example of dual-narrowing



- $[a, b] \tilde{\Delta} [c, d] \triangleq [(\lfloor c = -\infty \text{ ? } a : \lfloor (a + c)/2 \rfloor), (\lceil d = \infty \text{ ? } b : \lceil (b + d)/2 \rceil)]$
- The first method we tried in the late 70's with Radhia
 - Slow
 - Does not easily generalize (e.g. to pointer analysis)

Interpolation with dual-narrowing



- Refine widening/narrowing iterations Y^λ
- Refine a user-defined specification (Craig interpolation)

Craig interpolation

- Craig interpolation:

Given $P \implies Q$ find I such that $P \implies I \implies Q$ with
 $\text{var}(I) \subseteq \text{var}(P) \cap \text{var}(Q)$

is a **dual narrowing** (already observed by Vijay D'Silva and Leopold Haller as a narrowing [indeed inversed narrowing!])

- May not be unique
- May not terminate

Relationship between narrowing and dual-narrowing

- $\tilde{\Delta} = \Delta^{-1}$
- $Y \sqsubseteq X \implies Y \sqsubseteq X \Delta Y \sqsubseteq X$ (narrowing)
- $Y \sqsubseteq X \implies Y \sqsubseteq Y \tilde{\Delta} X \sqsubseteq X$ (dual-narrowing)

Note: effectiveness and termination conditions may be different

Bounded widening

Dual-narrowing versus bounded widening

- Dual-narrowing $\tilde{\Delta}$:

$$F(X) \sqsubseteq B \implies F(X) \sqsubseteq F(X) \tilde{\Delta} B \sqsubseteq B$$

Induction on $F(X)$ and B

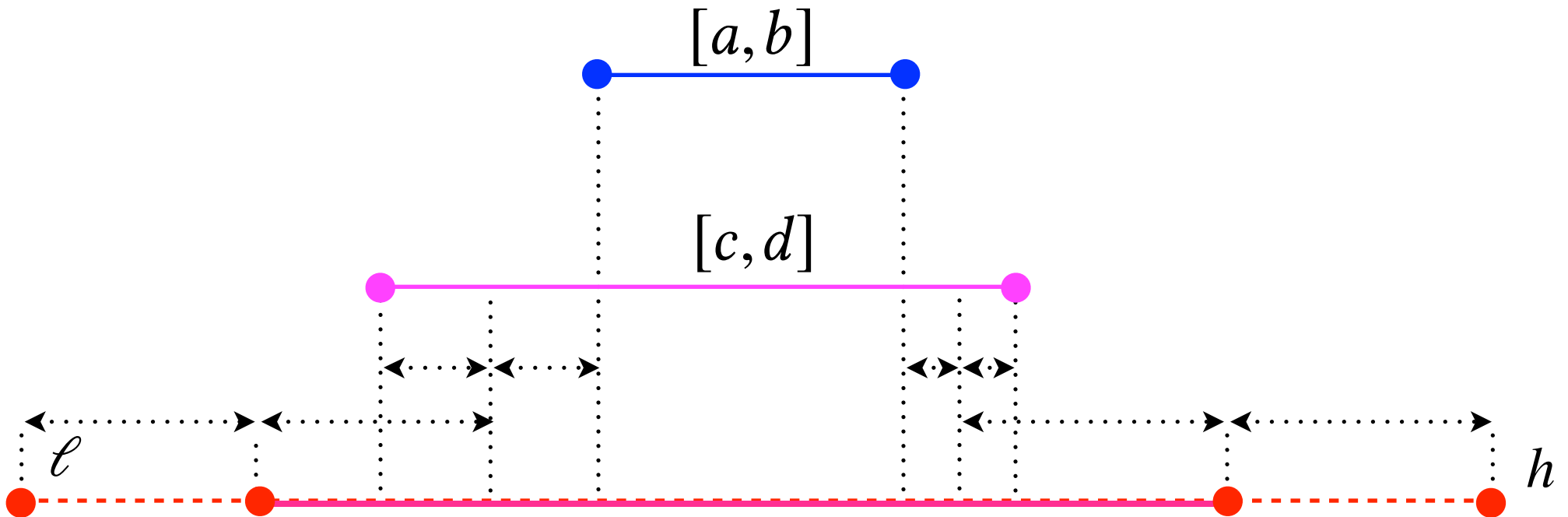
- Bounded widening ∇_B :

$$X \sqsubseteq F(X) \sqsubseteq B \implies F(X) \sqsubseteq X \nabla_B F(X) \sqsubseteq B$$

Induction on X , $F(X)$, and B

Example of widenings (cont'd)

- Bounded widening (in $[\ell, h]$):



$$[a, b] \nabla_{[\ell, h]} [c, d] \triangleq \left[\frac{c+a-2\ell}{2}, \frac{b+d+2h}{2} \right]$$

Soundness

Soundness

- Fixpoint approximation soundness theorems can be expressed with **minimalist hypotheses** (*):
- No need for complete lattices, complete partial orders (CPO's):
 - The concrete domain is a poset
 - The abstract domain is a pre-order
 - The concretization is defined for the abstract iterates only.

(*) Patrick Cousot. Abstracting Induction by Extrapolation and Interpolation In Deepak D'Souza, Akash Lal, and Kim Guldstrand Larsen (Eds), *16th International Conference on Verification, Model Checking, and Abstract Interpretation*, Mumbai, India, January 12—14, 2015. Lecture Notes in Computer Science, vol. 8931, pp. 19—42, © Springer 2015.

Soundness (cont'd)

- No need for increasingness/monotony hypotheses for fixpoint theorems (Tarski, Kleene, etc)
 - The concrete transformer is increasing and the limit of the iterations does exist in the concrete domain
 - No monotonicity hypotheses on the abstract transformer (no need for fixpoints in the abstract)
 - Soundness hypotheses on the extrapolators/interpolators with respect to the concrete
- In addition, the independent termination hypotheses on the extrapolators/interpolators ensure convergence in finitely many steps

Conclusion

The challenge of verification

- Infer the inductive argument
- Without deep knowledge about the program (e.g. very precise, quasi-inductive, quasi-strong enough specification)
- Scale

Infer the abstract inductive argument

```
typedef enum {FALSE = 0, TRUE = 1} BOOLEAN;
BOOLEAN INIT; float P, X;
void filter () {
    static float E[2], S[2];
    if (INIT) { S[0] = X; P = X; E[0] = X; }
    else { P = (((((0.5 * X) - (E[0] * 0.7)) + (E[1] * 0.4))
                + (S[0] * 1.5)) - (S[1] * 0.7)); }
    E[1] = E[0]; E[0] = X; S[1] = S[0]; S[0] = P;
    /* S[0], S[1] in ?????????????????????????????? */
}

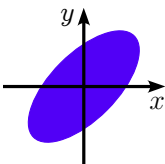
void main () { X = 0.2 * X + 5; INIT = TRUE;
    while (1) {
        X = 0.9 * X + 35; /* simulated filter input */
        filter (); INIT = FALSE; }
}
```

Infer the abstract inductive argument

```
typedef enum {FALSE = 0, TRUE = 1} BOOLEAN;
BOOLEAN INIT; float P, X;

void filter () {
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    else { P = (((((0.5 * X) - (E[0] * 0.7)) + (E[1] * 0.4))
                + (S[0] * 1.5)) - (S[1] * 0.7)); }
    E[1] = E[0]; E[0] = X; S[1] = S[0]; S[0] = P;
    /* S[0], S[1] in [-1327.02698354, 1327.02698354] */
}

void main () { X = 0.2 * X + 5; INIT = TRUE;
    while (1) {
        X = 0.9 * X + 35; /* simulated filter input */
        filter (); INIT = FALSE; }
}
```



Extrapolation/Interpolation

- Abstract interpretation in infinite domains is traditionally by iteration with widening/narrowing.
- We have shown how to use iteration with dual-narrowing.
- These ideas of the 70's generalize Craig interpolation from logic to arbitrary abstract domains.
- Can be used to improve precision when a fixpoint is reached after the widening/narrowing iterations

The End, Thank You