

**Comparing the Galois Connection
and Widening/Narrowing Approaches
to Abstract Interpretation**

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ABSTRACT INTERPRETATION (IN THEORY)

ABSTRACT INTERPRETATION is method for deriving conservative approximations of the semantics of programming languages.

ABSTRACT INTERPRETATION is used to:

- Specify hierarchies of semantics of programming languages at different levels of abstraction.
- Design program proof methods.
- Specify automatic program analyzers (by interpretation of programs in abstract domains).
- Etc.

ABSTRACT INTERPRETATION (IN PRACTICE)

ABSTRACT INTERPRETATION is a method for the automatic, static and conservative determination of dynamic properties of programs:

- Automatic: no human intervention during the analysis (as opposed to proof methods).
- Static: without considering all possible runs (as opposed to model-checking).
- Conservative/sound: without omitting some runs (as opposed to debugging).
- Dynamic properties: semantic properties of the runtime behaviors (as opposed to program metrology).

PART 1

**The Galois Connection Approach
to Abstract Interpretation**

COLLECTING SEMANTICS

- For a given program, the problem is the effective computation of a sound approximation A of the collecting semantics, specifying the exact properties of concern. For simplicity:
 - The collecting semantics is $\text{lfp}_{\perp} F$ where $\perp \in L$, $F \in L \xrightarrow{\text{con}} L$ and $L(\sqsubseteq, \perp, \sqcup)$ is a cpo.
 - Soundness of the approximation A is defined by: $\text{lfp}_{\perp} F \sqsubseteq A$.

EXAMPLE: DECLARATIVE SEMANTICS OF A LOGIC PROGRAM

- B_P : Herbrand universe for program P .
- $\text{ground}(P)$: set of all ground instances of clauses in P .
- The immediate consequence operator $T_P \in \wp(B_P) \xrightarrow{\text{con}} \wp(B_P)$:

$$T_P(X) = \left\{ A \mid A \leftarrow B_1, \dots, B_n \in \text{ground}(P) \right. \\ \left. \wedge \forall i = 1, \dots, n : B_i \in X \right\}$$
- A model of P is $I \subseteq B_P$, such that $T_P(I) \subseteq I$.
- Characterization theorem of the least model M_P (van Emden and Kowalski):

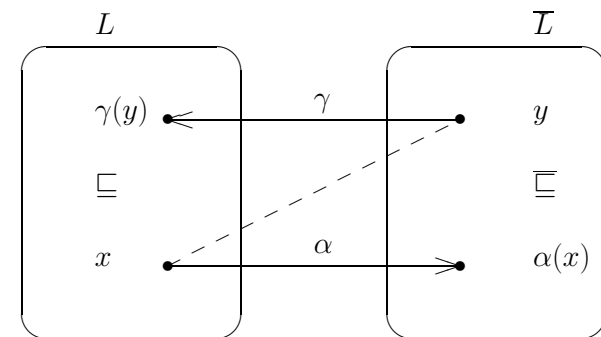
$\wp(B_P)(\subseteq, \emptyset, \cup)$ is a complete lattice.

$$M_P = \text{lfp}_{\emptyset} T_P = \cup_{n \in \mathbb{N}} T_P^n(\emptyset).$$

PROPERTY APPROXIMATION USING GALOIS CONNECTIONS

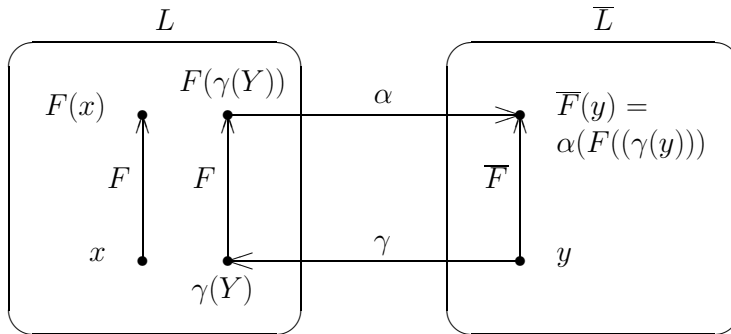
- Chose an abstract version \bar{L} of the concrete properties L .
- Chose an abstract version $\bar{\sqsubseteq}$ of the concrete approximation relation \sqsubseteq .
- For each abstract property $y \in \bar{L}$ chose its concrete meaning $\gamma(y) \in L$.
- Decide once for all of the abstract approximation $\alpha(x) \in \bar{L}$ of any concrete property $x \in L$.

GALOIS CONNECTIONS



- y is an approximation of x
- $\Leftrightarrow x \sqsubseteq \gamma(y)$
- $\Leftrightarrow \alpha(x) \bar{\sqsubseteq} y$

EXTENSION OF GALOIS CONNECTIONS TO FUNCTIONS



- \bar{F} is an approximation of F
- $\Leftrightarrow \alpha \circ F \circ \gamma \sqsubseteq \bar{F}$
- $\Leftrightarrow F \sqsubseteq \gamma \circ \bar{F} \circ \alpha$

EXTENSION OF GALOIS CONNECTIONS FROM PROPERTIES TO HIGHER-ORDER PROPERTY TRANSFORMERS

- if $L \xrightarrow[\alpha]{\gamma} \bar{L}$ is a Galois connection, then:

$$\vec{\alpha} \in (L \mapsto L) \mapsto (\bar{L} \mapsto \bar{L})$$

$$\vec{\alpha}(\varphi) \stackrel{\text{def}}{=} \alpha \circ \varphi \circ \gamma$$

$$\vec{\gamma} \in (\bar{L} \mapsto \bar{L}) \mapsto (L \mapsto L)$$

$$\vec{\gamma}(\bar{\varphi}) \stackrel{\text{def}}{=} \gamma \circ \bar{\varphi} \circ \alpha$$

is a Galois connection:

$$(L \xrightarrow{\text{mon}} L) \xrightarrow[\vec{\alpha}]{\vec{\gamma}} (\bar{L} \xrightarrow{\text{mon}} \bar{L})$$

FIXPOINT APPROXIMATION USING GALOIS CONNECTIONS

- $L(\sqsubseteq, \perp, \sqcup)$ is a cpo of concrete properties,
 $F \in (L \xrightarrow{\text{con}} L)$ is continuous,
 $\text{lfp}_{\perp} F = \sqcup_{n \geq 0} F^n(\perp)$ is not computable.
- Chose a cpo $\bar{L}(\sqsubseteq, \bar{\perp}, \bar{\sqcup})$ of abstract properties such that $L \xrightarrow[\alpha]{\gamma} \bar{L}$.
- Define $\bar{F} = \alpha \circ F \circ \gamma$.
 and $\bar{\perp} = \alpha(\perp)$.
- then $\text{lfp}_{\perp} F \sqsubseteq \gamma(\text{lfp}_{\bar{\perp}} \bar{F})$.

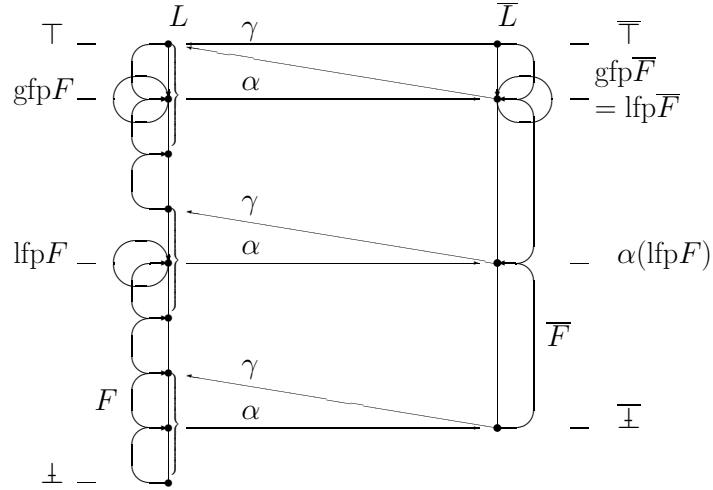
FIXPOINT APPROXIMATION ALGORITHM USING GALOIS CONNECTIONS

- If \bar{L} is finite (or satisfies the ascending chain condition), you have got an effective program analysis algorithm:

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⟨ $\bar{\perp}$ ,  $\bar{F}$ ⟩ := analysis(Program);
%%  $\alpha(\perp) \sqsubseteq \bar{\perp} \wedge \alpha \circ F \circ \gamma \sqsubseteq \bar{F}$ 
X :=  $\bar{\perp}$ ;
repeat
    Y := X;
    X :=  $\bar{F}(X)$ 
until Y = X;
%%  $\text{lfp}_{\perp} F \sqsubseteq \gamma(X) \wedge \text{lfp}_{\bar{\perp}} \bar{F} \sqsubseteq X$ 
    
```

FIXPOINT APPROXIMATION USING GALOIS CONNECTIONS



A FEW CLASSICAL EXAMPLES: EXAMPLE 1: RULE OF SIGNS

- $L = \wp(\mathbb{Z})$ set of possible values of an integer variable.
- $\bar{L} =$
- $\alpha(X) = \sqcup \{\text{sign}(x) \mid x \in X\}$

EXAMPLE 2: MYCROFT'S STRICTNESS ANALYSIS IN FUNCTIONAL PROGRAMMING

- $\mathbb{Z}_\perp = \mathbb{Z} \cup \perp$ \perp represent non-termination
- f is strict $\Leftrightarrow f(\perp) = \perp$
- $\Leftrightarrow f^*(\{\perp\}) \subseteq \{\perp\}$ where $f^*(X) = \{f(x) \mid x \in X\}$
- $L = \wp(\mathbb{Z}_\perp) \mapsto \wp(\mathbb{Z}_\perp)$
- $\bar{L} = \mathbb{B} \mapsto \mathbb{B}$ where $\mathbb{B} = \{0, 1\}$

SOUNDNESS OF STRICTNESS ANALYSIS

- $\alpha(X) = 0$ if $X \subseteq \{\perp\}$
 $\quad = 1$ if $X \not\subseteq \{\perp\}$
- $\gamma(0) = \{\perp\}$
 $\quad \gamma(1) = \mathbb{Z}_\perp$
- $\bar{\alpha}(f^*) = \alpha \circ f^* \circ \gamma$
 $\quad \bar{\gamma}(\bar{f}) = \gamma \circ \bar{f} \circ \alpha$
- $\bar{f}(0) = 0 \Rightarrow \alpha \circ f^* \circ \gamma(0) = 0 \Rightarrow \alpha \circ f^*(\{\perp\}) = 0 \Rightarrow f^*(\{\perp\}) \subseteq \{\perp\} \Leftrightarrow f$ is strict.

EXAMPLE 3 : MANNILA AND UKKONEN GROUNDNESS ANALYSIS IN LOGIC PROGRAMMING

- $\alpha(S) = \{\alpha_S(s) \mid s \in S\}$ set of states
- $\alpha_S(\langle g, \theta \rangle) = \alpha_g(g)$ state
- $\alpha_g(\square) = \emptyset$ goal
- $\alpha_g(a_1 \dots a_n \square) = \{\alpha_a(a_i) \mid i = 1, \dots, n\}$
- $\alpha_a(p(t_1, \dots, t_n)) = p(\alpha_t(t_1), \dots, \alpha_t(t_n))$ predicate
- $\alpha_t(X) = \text{NG}$ variable
- $\alpha_t(c) = G$ constant
- $\alpha_t(f(t_1, \dots, t_n)) = G$ if $\forall i = 1, \dots, n : \alpha_t(t_i) = G$ term
 $= \text{NG}$ if $\exists i = 1, \dots, n : \alpha_t(t_i) = \text{NG}$

ON THE GALOIS CONNECTION APPROACH TO ABSTRACT INTERPRETATION

- The approximation is done a priori, once for all $(L \xrightarrow{\gamma} \overline{L})$.
- The approximation α may be very rough.
- Usefulness of the approximation is shown by experience.
- The approximation is applied at each iteration step for $\overline{F} = \alpha \circ F \circ \gamma$.
- The approximation is independent of the iterates.
- \overline{L} must satisfy the ascending chain condition.

PART 2

The Widening/Narrowing Approach to Abstract Interpretation

WIDENING OPERATOR

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

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

- $\hat{X}^0 = \perp$
- $\hat{X}^{i+1} = \hat{X}^i$ 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 $\text{lfp}_{\perp} F$.

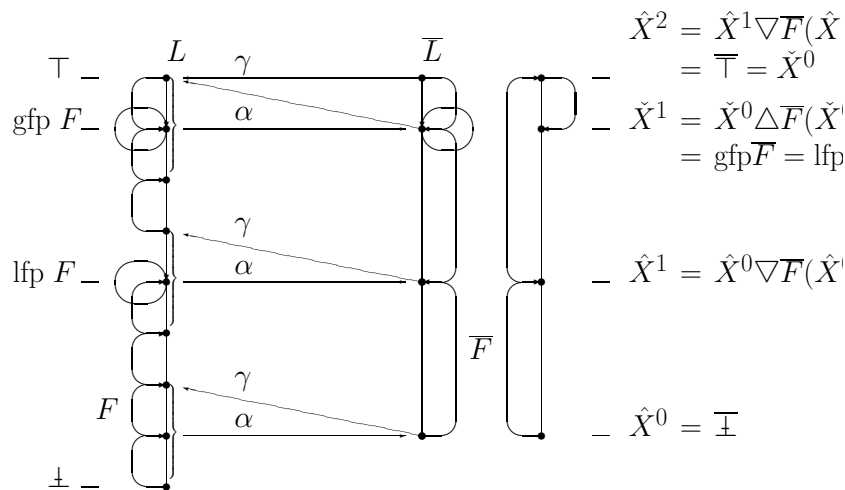
$$\text{lfp}_{\perp} F \sqsubseteq \hat{A}$$

PROGRAM ANALYSIS ALGORITHM WITH WIDENING

```

 $\langle \underline{\perp}, \overline{F} \rangle := \text{analysis}(\text{Program});$ 
%%  $\alpha(\perp) \sqsubseteq \underline{\perp} \wedge \alpha \circ F \circ \gamma \sqsubseteq \overline{F}$ 
 $X := \underline{\perp};$ 
repeat
   $Y := X;$ 
   $X := \overline{F}(X)$ 
if  $X \sqsubseteq Y$  then  $C := \text{true}$ 
else  $C := \text{false}; X := Y \nabla X$ 
until  $C;$ 
%%  $\text{lfp}_{\perp} F \sqsubseteq \gamma(Y) \wedge \text{lfp}_{\perp} \overline{F} \sqsubseteq Y$ 
    
```

FIXPOINT APPROXIMATION WITH WIDENING/NARROWING



A FEW CLASSICAL EXAMPLES: EXAMPLE 1: INTERVAL ANALYSIS

INTERVAL ANALYSIS (CONTINUED)

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

$$\begin{aligned} \perp \nabla X &= X \\ X \nabla \perp &= X \\ [\ell_0, u_0] \nabla [\ell_1, u_1] &= [\text{if } \ell_1 < \ell_0 \text{ then } -\infty \text{ else } \ell_0, \\ &\quad \text{if } u_1 > u_0 \text{ then } +\infty \text{ else } u_0] \end{aligned}$$

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

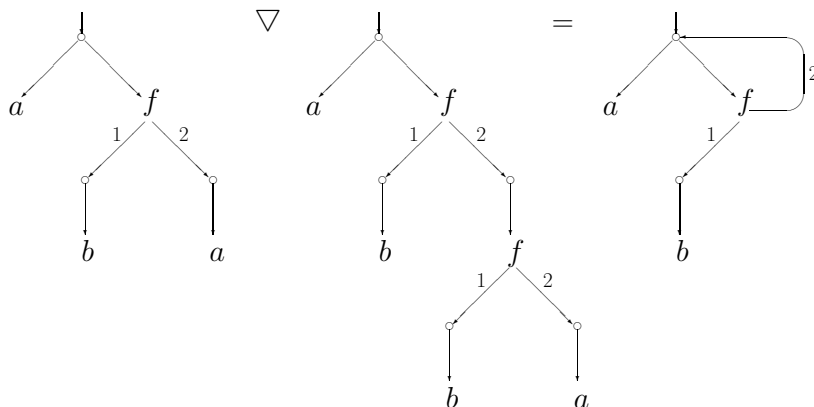
IMPROVED WIDENING FOR INTERVAL ANALYSIS

- Extrapolate to zero, one or infinity:

$$\begin{aligned} \perp \nabla X &= X \\ X \nabla \perp &= X \\ [\ell_0, u_0] \nabla [\ell_1, u_1] &= [\text{if } \ell \leq \ell_1 < \ell_0 \wedge \ell \in \{1, 0, -1\} \text{ then } 1 \\ &\quad \text{elseif } \ell_1 < \ell_0 \text{ then } -\infty \\ &\quad \text{else } \ell_0, \\ &\quad \text{if } u_0 < u_1 \leq u \wedge u \in \{-1, 0, 1\} \text{ then } u \\ &\quad \text{elseif } u_0 < u_1 \text{ then } +\infty \\ &\quad \text{else } u_0] \end{aligned}$$

- So the analysis is always as good as the sign analysis.

EXAMPLE 2: BRUYNNOOGHE'S TYPE GRAPH WIDENING



EXAMPLE 3: LINEAR INEQUALITIES & APPLICATION TO ARGUMENT SIZE ANALYSIS IN LOGIC PROGRAMMING

- Approximation of a term by its size:

$$\begin{aligned} \sigma(c) &= \sigma(X) = 1 \\ \sigma(f(t_1, \dots, t_n)) &= 1 + \sum_{i=1}^n \sigma(t_i) \end{aligned}$$

- Approximation a set of points in \mathbb{Z}^n by its convex hull:

$$\alpha_A(X) = \lambda p. \text{ConvexHull}(\{\langle \sigma(t_1), \dots, \sigma(t_n) \rangle \mid p(t_1, \dots, t_n) \in X\})$$

- Approximation of a set of states by upper bounds of the argument sizes of the atoms occurring in these states:

$$\begin{aligned} \alpha_g(a_1 \dots a_n \square) &= \{a_i \mid i = 1, \dots, n\} \quad (\emptyset \text{ if } n = 0) \\ \alpha_S(\langle g, \theta \rangle) &= \alpha_g(g) \\ \alpha(S) &= \alpha_A(\cup \{\alpha_S(s) \mid s \in S\}) \end{aligned}$$

EXAMPLE OF ARGUMENT SIZE ANALYSIS

- Program testing for inequality of natural numbers $n \geq 0$ represented as successors $s^n(0)$ of zero:

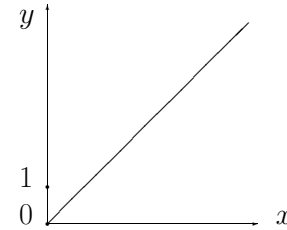
$$\begin{aligned} p(X, X) & \rightarrow ; \\ p(X, s(Y)) & \rightarrow p(X, Y) ; \end{aligned}$$

- Set of atoms: $\{p(X, s^n(X)) \mid n \geq 0\}$
- Approximation: $\{p(x, y) \mid x \geq 0 \wedge y \geq 0 \wedge x \leq y\}$
- Fixpoint equation:

$$F(X) = \{\langle x, y \rangle \mid x \geq 0 \wedge y \geq 0 \wedge ((x = y) \vee (\langle x, y - 1 \rangle \in X))\}$$
- The iterative computation of the least fixpoint does not converge in finitely many steps.

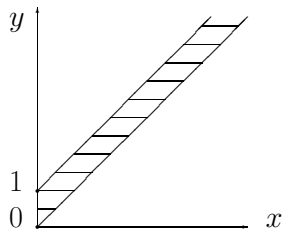
ITERATION WITH WIDENING (1)

- $\hat{X}^0 = \emptyset$
- $\hat{X}^1 = F(\hat{X}^0)$
 $= \{\langle x, y \rangle \mid x \geq 0 \wedge x = y\}$



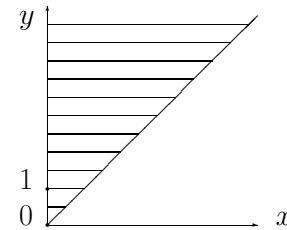
ITERATION WITH WIDENING (2)

- $F(\hat{X}^1) = \{\langle x, y \rangle \mid 0 \leq x \leq y \leq x + 1\}$



ITERATION WITH WIDENING (3)

- $\hat{X}^2 = \hat{X}^1 \nabla F(\hat{X}^1)$
 $= \{\langle x, y \rangle \mid 0 \leq x \leq y\}$



- $\hat{X}^3 = F(\hat{X}^2) = \hat{X}^2$

WIDENING OF POLYHEDRA

- Polyhedron P_1 is given by inequalities $S_1 = \{\beta_1, \dots, \beta_n\}$
- P_2 is represented by $S_2 = \{\gamma_1, \dots, \gamma_m\}$
- $P_1 \nabla P_2$ is $S'_1 \cup S'_2$ where:
 - S'_1 is the set of inequalities $\beta_i \in S_1$ satisfied by all points of P_2
 - S'_2 is the set of linear inequalities $\gamma_i \in S_2$ which can replace some $\beta_j \in S_1$ without changing polyhedron P_1

Example:

$$P_1 = \{\langle x, y \rangle \mid x \geq 0 \wedge x \leq y \wedge y \leq x\}$$

$$P_2 = \{\langle x, y \rangle \mid 0 \leq x \leq y \leq x + 1\}$$

$$P_1 \nabla P_2 = \{\langle x, y \rangle \mid 0 \leq x \leq y\}$$

ON THE FIXPOINT APPROXIMATION USING WIDENING OPERATORS

- The approximation is done a priori, once for all ($L \xrightarrow{\alpha} \bar{L}$ and ∇).
- The approximation α may be precise while ∇ may be very rough.
- Usefulness of the approximation is shown by experience (precision/cost can be tuned with ∇).
- The approximation is applied at each iteration step for \bar{F} .
- The approximation is dependent of the iterates.
- \bar{L} need not satisfy the ascending chain condition (since ∇ will be used to enforce convergence).

PART 3

**Comparing
the Galois Connection
and
The Widening/Narrowing
Approaches to Abstract Interpretation**

A COMMON BELIEVE ABOUT WIDENINGS

- Given an infinite abstract domain together with specific widening (and narrowing) operators, it is possible to find a finite lattice and a Galois connection which will give the same results.
- Hence the widening/narrowing approach to abstract interpretation is a useless trick.

WHAT IS PROVED IN THE PAPER ?

1. For each program there exists a finite lattice which can be used for this program to obtain results equivalent to those obtained using widening/narrowing operators;
2. No such a finite lattice will do for all programs;
3. For all programs, infinitely many abstract values are necessary;
4. For a particular program it is not possible to infer the set of needed abstract values by a simple inspection of the text of the program.

EXAMPLE 1 : LINEAR INEQUALITY ANALYSIS

```
program PL;
  var I, J : integer;
begin
  I := 2; J := 0;
  while ... do begin
    {  $2J+2 \leq I \wedge 0 \leq J$  }
    if ... then begin
      I := I + 4;
      {  $2J+6 \leq I \wedge 0 \leq J$  }
    end else begin
      I := I + 2; J := J + 1;
      {  $2J+2 \leq I \wedge 1 \leq J$  }
    end;
    {  $2J+2 \leq I \wedge 6 \leq I+2J \wedge 0 \leq J$  }
  end;
end.
```

EXAMPLE 2 : RATIONAL CONGRUENCE ANALYSIS (GRANGER)

```
program PC;
  var X : real;
begin
  X := 2.8542;
  while ... do begin
    {  $X \equiv 1/5000 \pmod{1/500}$  }
    X := X + 1/500;
  end;
end.
```

EXAMPLE 3 : INTERVAL ANALYSIS

```
program Function9ofMcCarthy;
  var X, Y : integer;
  function F(X : integer) : integer;
  begin
    if X > n then
      F := X - 10;
    else
      F := F(F(X + 11));
    end;
  end;
begin
  readln(X);
  Y := F(X);
  {  $Y \in [n-9, \text{maxint}-10]$  }
end.
```

```

program Function91ofMcCarthy;
  var X, Y : integer;
  function F(X : integer) : integer;
  begin
    if X > 100 then
      F := X - 10
      { F ∈ [91, maxint - 10] }
    else
      F := F(F(F(X + 33)));
      { F ∈ [91, 93] }
    { F ∈ [91, maxint - 10] }
  end;
begin
  readln(X);
  Y := F(X);
  { Y ∈ [91, maxint - 10] }
end.

```

CONCLUSION

- The Galois connection approach is the basic method of abstract interpretation.
- Combination with the widening/narrowing is the key to practical success:
 - Rich domain of information,
 - Convergence acceleration.
- Ideas for designing widenings/narrowings are given in the paper together with examples.