

Induced Operations

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- Motivation
- Inducing along the Abstraction function
 - Correctness
 - Fixed Points
 - Application to Data Flow Analysis
 - Example
- Inducing along the Concretisation function
 - Introduction
 - Concretisation process
 - Widening Operator
 - Correctness

Key Idea

Use Galois connections to transform computations into more approximate computations with better time-, space-, or termination behavior.

Two possible ways:

Inducing along the Abstraction function

- Replace a computation using L by a computation using M :
Analysis using M is an upper approximation to the analysis induced by L (loss of precision).

Inducing along the Concretisation function

- Use M for approximating the fixed point computations in L :
Ensure convergence of fixed points by using the more approximate complete lattice M while maintaining precision of the analysis.

Assumptions

- Galois connections $(L_1, \alpha_1, \gamma_1, M_1)$ and $(L_2, \alpha_2, \gamma_2, M_2)$
- Analysis $f_p : L_1 \rightarrow L_2$

Goal

Replace f_p by new more approximate analysis $g_p : M_1 \rightarrow M_2$.

- Candidate for g_p : $\alpha_2 \circ f_p \circ \gamma_1$

Inducing along the Abstraction Function

Example

Consider analysis $f_{plus}(\mathbb{ZZ}) = \{z_1 + z_2 \mid (z_1, z_2) \in \mathbb{ZZ}\}$ using complete lattices $(\mathcal{P}(\mathbb{Z}), \subseteq)$ and $(\mathcal{P}(\mathbb{Z} \times \mathbb{Z}), \subseteq)$.

Galois Connections

- $(\mathcal{P}(\mathbb{Z}), \alpha_{sign}, \gamma_{sign}, \mathcal{P}(\mathbf{Sign}))$
 - $\alpha_{sign}(Z) = \{sign(z) \mid z \in Z\}$
 - $\gamma_{sign}(S) = \{z \in \mathbb{Z} \mid sign(z) \in S\}$
- $(\mathcal{P}(\mathbb{Z} \times \mathbb{Z}), \alpha_{SS'}, \gamma_{SS'}, \mathcal{P}(\mathbf{Sign} \times \mathbf{Sign}))$
 - $\alpha_{SS'}(ZZ) = \{(sign(z_1), sign(z_2)) \mid (z_1, z_2) \in ZZ\}$
 - $\gamma_{SS'}(SS) = \{(z_1, z_2) \mid (sign(z_1), sign(z_2)) \in SS\}$

Inducing along the Abstraction Function

Example (cont'd)

Construct analysis $g_{plus} : \mathcal{P}(\mathbf{Sign} \times \mathbf{Sign}) \rightarrow \mathcal{P}(\mathbf{Sign})$ from f_{plus} , using candidate $g_{plus} = \alpha_{sign} \circ f_{plus} \circ \gamma_{SS'}$.

$$\begin{aligned}g_{plus}(SS) &= \alpha_{sign}(f_{plus}(\gamma_{SS'}(SS))) \\ &= \alpha_{sign}(f_{plus}(\{(z_1, z_2) \mid (sign(z_1), sign(z_2)) \in SS\})) \\ &= \alpha_{sign}(\{z_1 + z_2 \mid (sign(z_1), sign(z_2)) \in SS\}) \\ &= \{sign(z_1 + z_2) \mid (sign(z_1), sign(z_2)) \in SS\} \\ &= \bigcup \{s_1 \oplus s_2 \mid (s_1, s_2) \in SS\}\end{aligned}$$

where \oplus is the “addition” on signs.

Inducing along the Abstraction Function

Correctness

Recap

- Galois connections $(L_i, \alpha_i, \gamma_i, M_i)$, $i \in \{1, 2\}$
- Analysis $f_p : L_1 \rightarrow L_2$
- Analysis $g_p : M_1 \rightarrow M_2$

Correctness relations

- Representation functions $\beta_i : V_i \rightarrow L_i$
- Correctness relations $R_i : V_i \times L_i \rightarrow \{true, false\}$ generated by $\beta_i : V_i \rightarrow L_i$
- Correctness relations $S_i : V_i \times M_i \rightarrow \{true, false\}$ generated by $\alpha_i \circ \beta_i : V_i \rightarrow M_i$

Inducing along the Abstraction Function

Correctness cont'd

Lemma 4.41

If $(L_i, \alpha_i, \gamma_i, M_i)$ are Galois connections and $\beta_i : V_i \rightarrow L_i$ are representation functions, then

$$((\alpha_1 \circ \beta_1) \rightarrow (\alpha_2 \circ \beta_2))(\rightsquigarrow) = \alpha_2 \circ ((\beta_1 \rightarrow \beta_2)(\rightsquigarrow)) \circ \gamma_1$$

holds for all \rightsquigarrow .

Inducing along the Abstraction Function

Correctness cont'd

Proof (Lemma 4.41)

Simply calculate:

$$\begin{aligned} & ((\alpha_1 \circ \beta_1) \twoheadrightarrow (\alpha_2 \circ \beta_2))(\sim)(m_1) \\ &= \bigsqcup \{ \alpha_2(\beta_2(v_2)) \mid \alpha_1(\beta_1(v_1)) \sqsubseteq m_1 \wedge v_1 \sim v_2 \} \\ &= \alpha_2 \left(\bigsqcup \{ \beta_2(v_2) \mid \beta_1(v_1) \sqsubseteq \gamma_1(m_1) \wedge v_1 \sim v_2 \} \right) \\ &= \alpha_2((\beta_1 \twoheadrightarrow \beta_2)(\sim)(\gamma_1(m_1))) \\ &= (\alpha_2 \circ ((\beta_1 \twoheadrightarrow \beta_2)(\sim))) \circ \gamma_1(m_1) \end{aligned}$$



Inducing along the Abstraction Function

Correctness cont'd

Lemma 4.41 yields:

$$\begin{aligned} (p \vdash \cdot \rightsquigarrow \cdot)(R_1 \twoheadrightarrow R_2)f_p \wedge \alpha_2 \circ f_p \circ \gamma_1 \sqsubseteq g_p \\ \Rightarrow (p \vdash \cdot \rightsquigarrow \cdot)(S_1 \twoheadrightarrow S_2)g_p \end{aligned}$$

In words: if f_p is correct and g_p is an upper approximation to the induced analysis $\alpha_2 \circ f_p \circ \gamma_1$ then also g_p is correct.

Inducing along the Abstraction Function

Correctness cont'd

Proof

- 1 Suppose $(p \vdash \cdot \rightsquigarrow \cdot)(R_1 \rightarrow R_2)f_p$ and $\alpha_2 \circ f_p \circ \gamma_1 \sqsubseteq g_p$.
- 2 Since $(L_i, \alpha_i, \gamma_i, M_i)$ are Galois connections and f_p and g_p are monotone we get $f_p \sqsubseteq \gamma_2 \circ g_p \circ \alpha_1$.
- 3 Using the first assumption and Lemma 4.8:

$$\begin{aligned} & (p \vdash \cdot \rightsquigarrow \cdot)(R_1 \rightarrow R_2)f_p \wedge f_p \sqsubseteq \gamma_2 \circ g_p \circ \alpha_1 \\ & \Rightarrow (\beta_1 \rightarrow \beta_2)(p \vdash \cdot \rightsquigarrow \cdot) \sqsubseteq f_p \wedge f_p \sqsubseteq \gamma_2 \circ g_p \circ \alpha_1 \\ & \Rightarrow (\beta_1 \rightarrow \beta_2)(p \vdash \cdot \rightsquigarrow \cdot) \sqsubseteq \gamma_2 \circ g_p \circ \alpha_1 \\ & \Rightarrow \alpha_2 \circ (\beta_1 \rightarrow \beta_2)(p \vdash \cdot \rightsquigarrow \cdot) \circ \gamma_1 \sqsubseteq g_p \\ & \Rightarrow (\alpha_1 \circ \beta_1 \rightarrow \alpha_2 \circ \beta_2)(p \vdash \cdot \rightsquigarrow \cdot) \sqsubseteq g_p \\ & \Rightarrow (p \vdash \cdot \rightsquigarrow \cdot)(S_1 \rightarrow S_2)g_p \end{aligned}$$



Inducing along the Abstraction Function

Optimality

Definition

A function $f_p : L_1 \rightarrow L_2$ is *optimal* for the program p if and only if correctness of a function $f' : L_1 \rightarrow L_2$ amounts to $f_p \sqsubseteq f'$

Equivalently, f_p is *optimal* if and only if $(\beta_1 \rightarrow \beta_2)(p \vdash \cdot \rightsquigarrow \cdot) = f_p$

Lemma 4.41 may then be read as saying that if $f_p : L_1 \rightarrow L_2$ is optimal then so is $\alpha_2 \circ f_p \circ \gamma_1 : M_1 \rightarrow M_2$.

Inducing along the Abstraction Function

Fixed Points

Consider analysis $f_p : L_1 \rightarrow L_2$ requires computation of the least fixed point of a monotone function $F : (L_1 \rightarrow L_2) \rightarrow (L_1 \rightarrow L_2)$ so that $f_p = \text{lfp}(F)$.

- $(L_i, \alpha_i, \gamma_i, M_i)$ give rise to $(L_1 \rightarrow L_2, \alpha, \gamma, M_1 \rightarrow M_2)$
- Let $G : (M_1 \rightarrow M_2) \rightarrow (M_1 \rightarrow M_2)$ be an upper approximation to $\alpha \circ F \circ \gamma$
- Take $g_p : M_1 \rightarrow M_2$ to be $g_p = \text{lfp}(G)$

Fact

Correctness of f_p carries over to g_p .

Inducing along the Abstraction Function

Fixed Points - Correctness

Lemma 4.42

Assume

- (L, α, γ, M) is a Galois connection
- $f : L \rightarrow L$ and $g : M \rightarrow M$ are monotone functions
- g is an upper approximation to f (i.e. $\alpha \circ f \circ \gamma \sqsubseteq g$)

Then follows

- $\forall m \in M : g(m) \sqsubseteq m \Rightarrow f(\gamma(m)) \sqsubseteq \gamma(m)$
- and furthermore $lfp(f) \sqsubseteq \gamma(lfp(g))$ and $\alpha(lfp(f)) \sqsubseteq lfp(g)$

Inducing along the Abstraction Function

Fixed Points - Correctness

Proof

Show $\forall m \in M : g(m) \sqsubseteq m \Rightarrow f(\gamma(m)) \sqsubseteq \gamma(m)$

$$\begin{aligned}g(m) \sqsubseteq m \wedge \alpha(f(\gamma(m))) \sqsubseteq g(m) \\ \Rightarrow \alpha(f(\gamma(m))) \sqsubseteq m \\ \Rightarrow f(\gamma(m)) \sqsubseteq \gamma(m)\end{aligned}$$



Inducing along the Abstraction Function

Fixed Points - Correctness

Proof cont'd

From the previous result follows $\{\gamma(m) \mid g(m) \sqsubseteq m\} \subseteq \{l \mid f(l) \sqsubseteq l\}$ and hence (using Lemma 4.22)

$$\gamma\left(\bigsqcap\{m \mid g(m) \sqsubseteq m\}\right) = \bigsqcap\{\gamma(m) \mid g(m) \sqsubseteq m\} \sqsupseteq \bigsqcap\{l \mid f(l) \sqsubseteq l\}$$

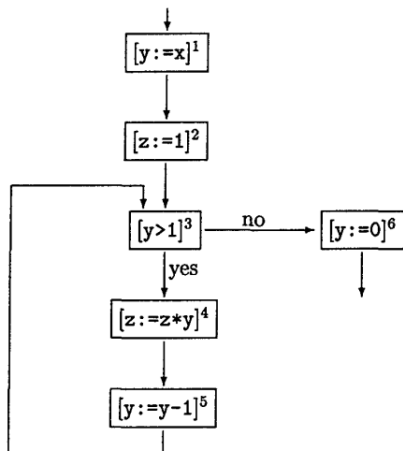
Using Tarski's theorem and that a Galois connection is an adjunction:

$$\begin{aligned} \gamma\left(\bigsqcap\{m \mid g(m) \sqsubseteq m\}\right) &\sqsupseteq \bigsqcap\{l \mid f(l) \sqsubseteq l\} \\ &\Rightarrow \gamma(\text{Red}(g)) \sqsupseteq \text{Red}(f) \\ &\Rightarrow \text{lfp}(f) \sqsubseteq \gamma(\text{lfp}(g)) \\ &\Rightarrow \alpha(\text{lfp}(f)) \sqsubseteq \text{lfp}(g) \end{aligned}$$

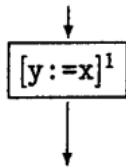


Excursion - Data Flow Analysis

- Model how the data flows through a program
- Construct Constraints describing the program
- Analysis describes state-changes with transfer functions



- Statement with label $l = 1$
- Assigns x to y
- For some analysis A we write
 - $A_o(l)$ for the **entry**-state
 - $A_\bullet(l)$ for the **exit**-state
 - $f_l^A(\mathbf{State})$ for the transfer function
- constraints are modeled for example with
 - $A_\bullet(1) \sqsupseteq f_1^A(A_o(1))$
 - $A_o(2) \sqsupseteq A_\bullet(1)$



Inducing along the Abstraction Function

Application to Data Flow Analysis

Generalized Monotone Framework A

- complete lattice L
- finite flow $F \subseteq \mathbf{Lab} \times \mathbf{Lab}$
- finite set of extremal labels $E \subseteq \mathbf{Lab}$
- extremal value $i \in L$
- a mapping f from the labels of F and E to monotone transfer functions $L \rightarrow L$
- Constraints $A \sqsupseteq$

$$A_o(l) \sqsupseteq \bigsqcup \{A_\bullet(l') \mid (l', l) \in F\} \sqcup i'_E \quad \text{where } i'_E = \begin{cases} i & l \in E \\ \perp & l \notin E \end{cases}$$

$$A_\bullet(l) \sqsupseteq f_l(A_o(l))$$

Inducing along the Abstraction Function

Application to Data Flow Analysis

Generalized Monotone Framework A

- $(A_o, A_\bullet) \models A^\sqsupseteq$ whenever A_o, A_\bullet is a solution to the constraints A^\sqsupseteq
- consider the associated monotone function
 $\vec{f}(A_o, A_\bullet) = (\lambda l. A_o(l), \lambda l. A_\bullet(l))$
- $(A_o, A_\bullet) \sqsupseteq \vec{f}(A_o, A_\bullet)$ is equivalent to $(A_o, A_\bullet) \models A^\sqsupseteq$

Inducing along the Abstraction Function

Application to Data Flow Analysis

Generalized Monotone Framework B

- let (L, α, γ, M) be a Galois Connection
- B is as A , but has
 - the mapping g from labels of F and E to monotone transfer functions $M \rightarrow M$, that satisfies $g_l \sqsupseteq \alpha \circ f_l \circ \gamma$
 - the extremal value $j \sqsupseteq \alpha(i)$
- As in A we get the constraints B^{\sqsupseteq} for B and the associated monotone function \vec{g}

Fact

$$(B_{\circ}, B_{\bullet}) \models B^{\sqsupseteq} \implies (\gamma \circ B_{\circ}, \gamma \circ B_{\bullet}) \models A^{\sqsupseteq}$$

Inducing along the Abstraction Function

A Worked Example

Sets of States Analysis SS

- complete lattice $(\mathcal{P}(\mathbf{State}), \subseteq)$
- flow $F = flow(S_*)$
- set $E = \{init(S_*)\}$ of extremal labels
- extremal value $i = \mathbf{State}$
- transfer functions given by f_i^{SS} :

$$f_i^{SS}(\Sigma) = \begin{cases} \{\sigma[x \mapsto \mathcal{A}[[a]]\sigma] \mid \sigma \in \Sigma\} & \text{if } [x := a]' \text{ is in } S_* \\ \Sigma & \text{if } [skip]' \text{ is in } S_* \\ \Sigma & \text{if } [b]' \text{ is in } S_* \end{cases}$$

where $\Sigma \subseteq \mathbf{State}$

Inducing along the Abstraction Function

A Worked Example

Fact

The *SS* analysis is correct

Inducing along the Abstraction Function

A Worked Example

Constant Propagation Analysis

- complete lattice $\widehat{\mathbf{State}}_{\text{CP}} = ((\mathbf{Var} \rightarrow \mathbf{Z}^{\top})_{\perp}, \sqsubseteq)$
- flow $F = \text{flow}(S_*)$
- extremal labels $E = \{\text{init}(S_*)\}$
- extremal value $i = \lambda x. \top$
- transfer functions of the constant propagation analysis ^a given by f_i^{CP}

^aPrinciples of Program Analysis, page 71, Table 2.7

Inducing along the Abstraction Function

A Worked Example

The relationship between the two analyses is established by the representation function

$$\beta_{CP} : \mathbf{State} \rightarrow \widehat{\mathbf{State}}_{CP}$$
$$\beta_{CP}(\sigma) = \sigma$$

Galois Connection

β_{CP} gives rise to a Galois connection $(\mathcal{P}(\mathbf{State}), \alpha_{CP}, \gamma_{CP}, \widehat{\mathbf{State}}_{CP})$

$$\alpha_{CP}(\Sigma) = \bigsqcup \{\beta_{CP}(\sigma) \mid \sigma \in \Sigma\}$$

$$\gamma_{CP}(\hat{\sigma}) = \{\sigma \mid \beta_{CP}(\sigma) \sqsubseteq \hat{\sigma}\}$$

Inducing along the Abstraction Function

A Worked Example

Conclusion

one can now show

$$\forall l \in \mathbf{Lab} : f_l^{CP} \sqsupseteq \alpha_{CP} \circ f_l^{SS} \circ \gamma_{CP}$$
$$\gamma_{CP}(\lambda x. \top) = \mathbf{State}$$

and hence CP is an upper approximation to the analysis induced from SS by the Galois connection and therefore correct.

Why?

Inducing by abstraction function has a critical disadvantage. It loses precision along the analysis.

Inducing by Concretisation Function

instead of replacing the analysis using L with analysis using M ;

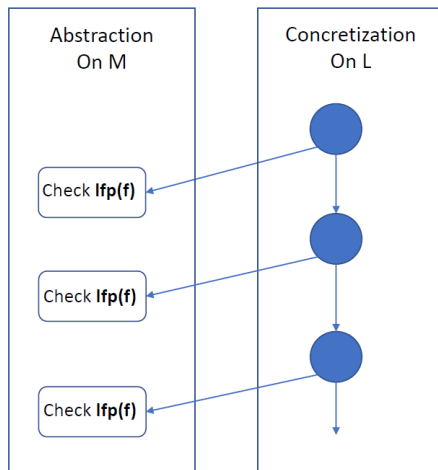
- We perform normally on L (to not lose precision).
- but we only use M to approximate the fixed point computations done in L (to ensure convergence of the fixed points).

Inducing by Concretisation Function

Using widening operator $\nabla_M : M \times M \rightarrow M$

- to define $\nabla_L : L \times L \rightarrow L$
by using the formula $l_1 \nabla_L l_2 = \gamma(\alpha(l_1) \nabla_M \alpha(l_2))$
- we can approximate **lfp**(f) over L .

Concretisation process



Inducing along the concretization function

Why Widening Operator?

- We can't guarantee reaching stability eventually.
- or reaching least upper bound that equals **lfp(f)**.

Widening Operator

- used to obtain approximations of the least fixed points.
- used to limit the number of computation steps needed.

lemma 4.45

If $(\mathbf{L}, \alpha, \gamma, \mathbf{M})$ is a Galois insertion such that

$\gamma(\perp_{\mathbf{M}}) = \perp_{\mathbf{L}}$ and if $\nabla_{\mathbf{M}} : \mathbf{M} \times \mathbf{M} \rightarrow \mathbf{M}$ is a widening operator.

Then $\nabla_{\mathbf{L}} : \mathbf{L} \times \mathbf{L} \rightarrow \mathbf{L}$ is a widening operator defined by the formula

$$\mathbf{l}_1 \nabla_{\mathbf{L}} \mathbf{l}_2 = \gamma(\alpha(\mathbf{l}_1 \nabla_{\mathbf{M}} \alpha(\mathbf{l}_2))).$$

this satisfies $\mathbf{lfp}_{\nabla_{\mathbf{L}}}(\mathbf{f}) = \gamma(\mathbf{lfp}_{\nabla_{\mathbf{M}}}(\alpha \circ \mathbf{f} \circ \gamma))$ for all monotone functions $\mathbf{f} : \mathbf{L} \rightarrow \mathbf{L}$.

Proof

- given $\nabla_{\mathbf{L}}$ is a widening operator, $\exists n_f \geq 0, \mathbf{lfp}_{\nabla_{\mathbf{L}}}(\mathbf{f}) = \mathbf{f}_{\nabla_{\mathbf{L}}}^{n_f} = \mathbf{f}_{\nabla_{\mathbf{L}}}^n$
- given $\nabla_{\mathbf{M}}$ is a widening operator, $\exists n_g \geq 0, \mathbf{lfp}_{\nabla_{\mathbf{M}}}(\mathbf{g}) = \mathbf{g}_{\nabla_{\mathbf{M}}}^{n_g} = \mathbf{g}_{\nabla_{\mathbf{M}}}^n$
- if we can prove that: $\mathbf{f}_{\nabla_{\mathbf{L}}}^n = \gamma(\mathbf{g}_{\nabla_{\mathbf{M}}}^n)$
- we can obtain that: $\mathbf{lfp}_{\nabla_{\mathbf{L}}}(\mathbf{f}) = \gamma(\mathbf{lfp}_{\nabla_{\mathbf{M}}}(\mathbf{g}))$

The Proof

by induction on n :

- base case: $n = 0$.

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$$\text{assume } \perp_L = \gamma(\perp_M)$$

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$$f_{\nabla_L}^0 = \perp_L \text{ and } g_{\nabla_M}^0 = \perp_M$$

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$$\Rightarrow f_{\nabla_L}^0 = \gamma(g_{\nabla_M}^0)$$

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by induction on n :

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$$\text{assume } \perp_L = \gamma(\perp_M)$$

$$\Rightarrow f_{\nabla_L}^0 = \gamma(g_{\nabla_M}^0)$$

- for induction step over n .

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$$\Rightarrow f_{\nabla_L}^0 = \gamma(g_{\nabla_M}^0)$$

- for induction step over n .

$$f(f_{\nabla_L}^n) \sqsubseteq f_{\nabla_L}^n \Leftrightarrow g(g_{\nabla_M}^n) \sqsubseteq g_{\nabla_M}^n$$

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$$f(f_{\nabla_L}^n) \sqsubseteq f_{\nabla_L}^n \Leftrightarrow g(g_{\nabla_M}^n) \sqsubseteq g_{\nabla_M}^n$$

$$f(f_{\nabla_L}^n) \sqsubseteq f_{\nabla_L}^n \Rightarrow \alpha(f(f_{\nabla_L}^n)) \sqsubseteq \alpha(f_{\nabla_L}^n)$$

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- base case: $n = 0$.

$$f_{\nabla_L}^0 = \perp_L \text{ and } g_{\nabla_M}^0 = \perp_M$$

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$$\Rightarrow f_{\nabla_L}^0 = \gamma(g_{\nabla_M}^0)$$

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$$\Rightarrow \alpha(f(\gamma(g_{\nabla_M}^n))) \sqsubseteq \alpha(\gamma(g_{\nabla_M}^n))$$

The Proof

by induction on n :

- base case: $n = 0$.

$$f_{\nabla_L}^0 = \perp_L \text{ and } g_{\nabla_M}^0 = \perp_M$$

$$\text{assume } \perp_L = \gamma(\perp_M)$$

$$\Rightarrow f_{\nabla_L}^0 = \gamma(g_{\nabla_M}^0)$$

- for induction step over n .

$$f(f_{\nabla_L}^n) \sqsubseteq f_{\nabla_L}^n \Leftrightarrow g(g_{\nabla_M}^n) \sqsubseteq g_{\nabla_M}^n$$

$$f(f_{\nabla_L}^n) \sqsubseteq f_{\nabla_L}^n \Rightarrow \alpha(f(f_{\nabla_L}^n)) \sqsubseteq \alpha(f_{\nabla_L}^n)$$

$$\Rightarrow \alpha(f(\gamma(g_{\nabla_M}^n))) \sqsubseteq \alpha(\gamma(g_{\nabla_M}^n))$$

$$\Rightarrow g(g_{\nabla_M}^n) \sqsubseteq \alpha(\gamma(g_{\nabla_M}^n))$$

The Proof

by induction on n :

- base case: $n = 0$.

$$f_{\nabla_L}^0 = \perp_L \text{ and } g_{\nabla_M}^0 = \perp_M$$

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$$f(f_{\nabla_L}^n) \sqsubseteq f_{\nabla_L}^n \Rightarrow \alpha(f(f_{\nabla_L}^n)) \sqsubseteq \alpha(f_{\nabla_L}^n)$$

$$\Rightarrow \alpha(f(\gamma(g_{\nabla_M}^n))) \sqsubseteq \alpha(\gamma(g_{\nabla_M}^n))$$

$$\Rightarrow g(g_{\nabla_M}^n) \sqsubseteq \alpha(\gamma(g_{\nabla_M}^n))$$

$$\Rightarrow g(g_{\nabla_M}^n) \sqsubseteq g_{\nabla_M}^n$$

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$$f(f_{\nabla_L}^n) \sqsubseteq f_{\nabla_L}^n \Leftrightarrow g(g_{\nabla_M}^n) \sqsubseteq g_{\nabla_M}^n$$

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- for induction step over n .

$$f(f_{\nabla_L}^n) \sqsubseteq f_{\nabla_L}^n \Leftrightarrow g(g_{\nabla_M}^n) \sqsubseteq g_{\nabla_M}^n$$

$$g(g_{\nabla_M}^n) \sqsubseteq g_{\nabla_M}^n \Rightarrow \gamma(g(g_{\nabla_M}^n)) \sqsubseteq \gamma(g_{\nabla_M}^n)$$

The Proof

by induction on n :

- base case: $n = 0$.

$$f_{\nabla_L}^0 = \perp_L \text{ and } g_{\nabla_M}^0 = \perp_M$$

$$\text{assume } \perp_L = \gamma(\perp_M)$$

$$\Rightarrow f_{\nabla_L}^0 = \gamma(g_{\nabla_M}^0)$$

- for induction step over n .

$$f(f_{\nabla_L}^n) \sqsubseteq f_{\nabla_L}^n \Leftrightarrow g(g_{\nabla_M}^n) \sqsubseteq g_{\nabla_M}^n$$

$$g(g_{\nabla_M}^n) \sqsubseteq g_{\nabla_M}^n \Rightarrow \gamma(g(g_{\nabla_M}^n)) \sqsubseteq \gamma(g_{\nabla_M}^n)$$

$$\Rightarrow \gamma(\alpha(f(\gamma(g_{\nabla_M}^n)))) \sqsubseteq \gamma(g_{\nabla_M}^n)$$

The Proof

by induction on n :

- base case: $n = 0$.

$$f_{\nabla_L}^0 = \perp_L \text{ and } g_{\nabla_M}^0 = \perp_M$$

$$\text{assume } \perp_L = \gamma(\perp_M)$$

$$\Rightarrow f_{\nabla_L}^0 = \gamma(g_{\nabla_M}^0)$$

- for induction step over n .

$$f(f_{\nabla_L}^n) \sqsubseteq f_{\nabla_L}^n \Leftrightarrow g(g_{\nabla_M}^n) \sqsubseteq g_{\nabla_M}^n$$

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$$\Rightarrow \gamma(\alpha(f(\gamma(g_{\nabla_M}^n)))) \sqsubseteq \gamma(g_{\nabla_M}^n)$$

$$\Rightarrow \gamma(\alpha(f(f_{\nabla_L}^n))) \sqsubseteq f_{\nabla_L}^n$$

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$$f(f_{\nabla_L}^n) \sqsubseteq f_{\nabla_L}^n \Leftrightarrow g(g_{\nabla_M}^n) \sqsubseteq g_{\nabla_M}^n$$

$$g(g_{\nabla_M}^n) \sqsubseteq g_{\nabla_M}^n \Rightarrow \gamma(g(g_{\nabla_M}^n)) \sqsubseteq \gamma(g_{\nabla_M}^n)$$

$$\Rightarrow \gamma(\alpha(f(\gamma(g_{\nabla_M}^n)))) \sqsubseteq \gamma(g_{\nabla_M}^n)$$

$$\Rightarrow \gamma(\alpha(f(f_{\nabla_L}^n))) \sqsubseteq f_{\nabla_L}^n$$

$$\Rightarrow f(f_{\nabla_L}^n) \sqsubseteq f_{\nabla_L}^n$$

The Proof

- induction step: $n > 0$.

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$$f_{\nabla_L}^n = \begin{cases} f_{\nabla_L}^{n-1} & \text{if } f(f_{\nabla_L}^{n-1}) \sqsubseteq f_{\nabla_L}^{n-1} \\ f_{\nabla_L}^{n-1} \nabla_L f(f_{\nabla_L}^{n-1}) & \text{otherwise} \end{cases}$$

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- induction step: $n > 0$.

$$\begin{aligned} f_{\nabla_L}^n &= \begin{cases} f_{\nabla_L}^{n-1} & \text{if } f(f_{\nabla_L}^{n-1}) \sqsubseteq f_{\nabla_L}^{n-1} \\ f_{\nabla_L}^{n-1} \nabla_L f(f_{\nabla_L}^{n-1}) & \text{otherwise} \end{cases} \\ &= \begin{cases} f_{\nabla_L}^{n-1} & \text{if } g(g_{\nabla_M}^{n-1}) \sqsubseteq g_{\nabla_M}^{n-1} \\ f_{\nabla_L}^{n-1} \nabla_L f(f_{\nabla_L}^{n-1}) & \text{otherwise} \end{cases} \end{aligned}$$

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The Proof

- induction step: $n > 0$.

$$\begin{aligned} f_{\nabla_L}^n &= \begin{cases} f_{\nabla_L}^{n-1} & \text{if } f(f_{\nabla_L}^{n-1}) \sqsubseteq f_{\nabla_L}^{n-1} \\ f_{\nabla_L}^{n-1} \nabla_L f(f_{\nabla_L}^{n-1}) & \text{otherwise} \end{cases} \\ &= \begin{cases} f_{\nabla_L}^{n-1} & \text{if } g(g_{\nabla_M}^{n-1}) \sqsubseteq g_{\nabla_M}^{n-1} \\ f_{\nabla_L}^{n-1} \nabla_L f(f_{\nabla_L}^{n-1}) & \text{otherwise} \end{cases} \\ &= \begin{cases} \gamma(g_{\nabla_M}^{n-1}) & \text{if } g(g_{\nabla_M}^{n-1}) \sqsubseteq g_{\nabla_M}^{n-1} \\ \gamma(\alpha(\gamma(g_{\nabla_M}^{n-1}) \nabla_M f(\gamma(g_{\nabla_M}^{n-1})))) & \text{otherwise} \end{cases} \\ &= \gamma \left(\begin{cases} (g_{\nabla_M}^{n-1}) & \text{if } g(g_{\nabla_M}^{n-1}) \sqsubseteq g_{\nabla_M}^{n-1} \\ g_{\nabla_M}^{n-1} \nabla_M g(g_{\nabla_M}^{n-1}) & \text{otherwise} \end{cases} \right) \end{aligned}$$

The Proof

- induction step: $n > 0$.

$$\begin{aligned} f_{\nabla_L}^n &= \begin{cases} f_{\nabla_L}^{n-1} & \text{if } f(f_{\nabla_L}^{n-1}) \sqsubseteq f_{\nabla_L}^{n-1} \\ f_{\nabla_L}^{n-1} \nabla_L f(f_{\nabla_L}^{n-1}) & \text{otherwise} \end{cases} \\ &= \begin{cases} f_{\nabla_L}^{n-1} & \text{if } g(g_{\nabla_M}^{n-1}) \sqsubseteq g_{\nabla_M}^{n-1} \\ f_{\nabla_L}^{n-1} \nabla_L f(f_{\nabla_L}^{n-1}) & \text{otherwise} \end{cases} \\ &= \begin{cases} \gamma(g_{\nabla_M}^{n-1}) & \text{if } g(g_{\nabla_M}^{n-1}) \sqsubseteq g_{\nabla_M}^{n-1} \\ \gamma(\alpha(\gamma(g_{\nabla_M}^{n-1}) \nabla_M f(\gamma(g_{\nabla_M}^{n-1})))) & \text{otherwise} \end{cases} \\ &= \gamma \left(\begin{cases} (g_{\nabla_M}^{n-1}) & \text{if } g(g_{\nabla_M}^{n-1}) \sqsubseteq g_{\nabla_M}^{n-1} \\ g_{\nabla_M}^{n-1} \nabla_M g(g_{\nabla_M}^{n-1}) & \text{otherwise} \end{cases} \right) \\ &= \gamma(g_{\nabla_M}^n) \end{aligned}$$

Proof

- given $\nabla_{\mathbf{L}}$ is a widening operator, $\exists n_f \geq 0, \mathbf{lfp}_{\nabla_{\mathbf{L}}}(\mathbf{f}) = \mathbf{f}_{\nabla_{\mathbf{L}}}^{n_f} = \mathbf{f}_{\nabla_{\mathbf{L}}}^n$
- given $\nabla_{\mathbf{M}}$ is a widening operator, $\exists n_g \geq 0, \mathbf{lfp}_{\nabla_{\mathbf{M}}}(\mathbf{g}) = \mathbf{g}_{\nabla_{\mathbf{M}}}^{n_g} = \mathbf{g}_{\nabla_{\mathbf{M}}}^n$
- We have proven that: $\mathbf{f}_{\nabla_{\mathbf{L}}}^n = \gamma(\mathbf{g}_{\nabla_{\mathbf{M}}}^n)$
- which prove that: $\mathbf{lfp}_{\nabla_{\mathbf{L}}}(\mathbf{f}) = \gamma(\mathbf{lfp}_{\nabla_{\mathbf{M}}}(\mathbf{g}))$

So, we can perform our analysis over \mathbf{L} without losing precision.

Thank you for your attention,
Questions?