

The Interprocedural Coincidence Theorem*

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Abstract

We present an interprocedural generalization of the well-known (intraprocedural) Coincidence Theorem of Kam and Ullman, which provides a sufficient condition for the equivalence of the *meet over all paths* (*MOP*) solution and the *maximal fixed point* (*MFP*) solution to a data flow analysis problem. This generalization covers arbitrary imperative programs with recursive procedures, global and local variables, and formal value parameters. In the absence of procedures, it reduces to the classical intraprocedural version. In particular, our stack-based approach generalizes the coincidence theorems of Barth and Sharir/Pnueli for the same setup, which do not properly deal with local variables of recursive procedures.

1 Motivation

Data flow analysis is a classical method for the static analysis of programs that supports the generation of efficient object code by “optimizing” compilers (cf. [He, MJ]). For imperative languages, it provides information about the program states that may occur at some given program points during execution.

Theoretically well-founded are data flow analyses that are based on *abstract interpretation* (cf. [CC1]). The point of this approach is to replace the “full” semantics by a simpler more abstract version, which is tailored to deal with a specific problem. Usually, the abstract semantics is specified by a local semantic functional, which gives abstract meaning to every program statement in terms of a transformation function from a lattice \mathcal{C} into itself. The elements of \mathcal{C} express the data flow information of interest. The (global) abstract semantics then results from one of the following two globalization strategies; the “operational” *meet over all paths* (*MOP*) strategy, and the “denotational” *maximal fixed point* (*MFP*) strategy¹ in the sense of Kam and Ullman [KU].²

The *MOP*-strategy directly mimics possible program executions: it “meets” (intersects) all information corresponding to program paths reaching the program point under consideration. This specifies the optimal result of a globalization but is in general not effective.

The *MFP*-strategy iteratively approximates the greatest solution of a system of equations that express consistency between pre-conditions and post-conditions that are given in terms of data flow information: the pre-condition of a statement must be implied by each of the post-conditions of the predecessors, and the post-condition must be implied by the result of transforming the pre-condition according to the (abstract) meaning of the statement. In general, this leads to a suboptimal but algorithmic description.

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¹These are the strategies that lead to the *MOP*-solution and *MFP*-solution, respectively.

²The operational and denotational flavour of these two strategies becomes particularly apparent in the interprocedural setting.

The (intraprocedural) Coincidence Theorem of [KU] states the coincidence of the *MOP*-solution and the *MFP*-solution in the case of *distributive* (local) semantic functionals (see Section 2). In this paper, we present an *interprocedural* generalization of this theorem (cf. Interprocedural Coincidence Theorem 5.3), which covers arbitrary imperative programs with recursive procedures, global and local variables, and formal value parameters. In the absence of procedures, it reduces to the classical intraprocedural version of Kam and Ullman. The point of our generalization is the introduction of stacks of lattice elements as data flow information, which is necessary to properly deal with local variables of recursive procedures. These stacks directly mimic the run-time stacks used by run-time systems for maintaining the activation records of different procedure incarnations. Whereas the operational (*MOP*) strategy exhaustively makes use of the stack structure, it turns out that the denotational (*MFP*) strategy only needs stacks of length at most two (Remark 4.7).

Related Work

Semantically based reasoning about interprocedural data flow analysis was first considered by Cousot/ Cousot [CC2], Rosen [Ro], Jones and Muchnick [JM], and more recently by Bourdoncle [Bo].

Rosen [Ro] proved the correctness of an interprocedural data flow analysis algorithm computing a maximal fixed point solution that provides information whether a variable is used, modified or preserved. Thus his algorithm is tailored to specific problems, and it is not clear how to generalize his approach to arbitrary abstract interpretations. Such a general situation was investigated in [CC2] and [JM], leading to highly technical definitions that are difficult to apply. Moreover, as in [Ro], these two papers only address the correctness aspect but no optimality. This holds also for the approach of [Bo], which considers correctness of data flow analyses with respect to the collecting semantics of a program.

Both correctness and optimality were considered by Barth [Ba1, Ba2] and Sharir/Pnueli [SP], who independently proposed an interprocedural version of the Coincidence Theorem of Kam and Ullman. However, their approaches do not properly deal with local variables of recursive procedures, which would require to store information about the local variables when treating a recursive procedure call.

Structure of the Paper

After sketching the intraprocedural setting in Section 2, we recall the formal framework for interprocedural data flow analysis in Section 3, and define two versions of abstract semantics for programs with procedures in Section 4: an “operational” one caused by the *interprocedural meet over all paths* strategy and a “denotational” one caused by the *interprocedural maximal fixed point* strategy. Subsequently, Section 5 presents the main results, Section 6 sketches some applications, and Section 7 contains our conclusions. Finally, the Appendix provides the detailed data flow analysis algorithms.

2 The Intraprocedural Setting

In this section we summarize the intraprocedural setting for data flow analysis, which is characterized by a separate and independent investigation of the procedures of a program. Here it is common to represent procedures as *directed flow graphs* $G = (N, E, \mathbf{s}, \mathbf{e})$ with node set N and edge set E .³ Nodes $n \in N$ represent the statements and edges $(n, m) \in E$ the nondeterministic branching structure of the corresponding procedure. $pred_G(n) =_{df} \{ m \mid (m, n) \in E \}$ and $succ_G(n) =_{df} \{ m \mid (n, m) \in E \}$ denote the set of all immediate predecessors and successors of

³The construction of flow graphs is described in [All].

a node n , respectively. \mathbf{s} and \mathbf{e} denote the unique *start node* and *end node* of G , which are assumed to possess no predecessors and successors, respectively. A *finite path* in G is a sequence (n_1, \dots, n_q) of nodes such that $(n_j, n_{j+1}) \in E$ for $j \in \{1, \dots, q-1\}$. $\mathbf{P}[m, n]$ denotes the set of all finite paths from m to n , and $\mathbf{P}[m, n)$ the set of all finite paths from m to a predecessor of n . Moreover, $lgt h(p)$ denotes the number of node occurrences in p , and ε the unique path of length 0. Finally, we assume that every node $n \in N$ lies on a path from \mathbf{s} to \mathbf{e} .

Given a complete semi-lattice $(\mathcal{C}, \sqcap, \sqsupseteq, \perp, \top)$, whose elements are intended to express the relevant data flow information, the *local* abstract semantics of a flow graph G is given by a semantic functional

$$\llbracket \cdot \rrbracket : N \rightarrow (\mathcal{C} \rightarrow \mathcal{C})$$

which gives meaning to every node $n \in N$ in terms of a transformation on \mathcal{C} . For simplicity, it is assumed that \mathbf{s} and \mathbf{e} are associated with the identity on \mathcal{C} .

The local abstract semantics $\llbracket \cdot \rrbracket$ can easily be extended to cover finite paths as well. For every path $p = (n_1, \dots, n_q) \in \mathbf{P}[m, n]$, we define:

$$\llbracket p \rrbracket =_{df} \begin{cases} id_{\mathcal{C}} & \text{if } p \equiv \varepsilon \\ \llbracket (n_2, \dots, n_q) \rrbracket \circ \llbracket n_1 \rrbracket & \text{otherwise} \end{cases}$$

The (global) abstract semantics then results from one of the following two globalization strategies: the “operational” *meet over all paths* (*MOP*) strategy, and the “denotational” *maximal fixed point* (*MFP*) strategy in the sense of Kam and Ullman [KU].

The *MOP*-strategy directly mimics possible program executions: it “meets” (intersects) all informations, which belong to a program path reaching the program point under consideration:

$$\mathbf{The\ MOP-Solution:} \quad \forall n \in N \quad \forall c_0 \in \mathcal{C}. \quad MOP_{c_0}(n) = \sqcap \{ \llbracket p \rrbracket(c_0) \mid p \in \mathbf{P}[\mathbf{s}, n] \}$$

This directly reflects our desires but is in general not effective.

The *MFP*-strategy iteratively approximates the greatest solution of a system of equations which specifies the consistency between pre-conditions and post-conditions that are expressed in terms of \mathcal{C} :

Equation System 2.1

$$\begin{aligned} \mathbf{pre}(n) &= \begin{cases} c_0 & \text{if } n = \mathbf{s} \\ \sqcap \{ \mathbf{post}(m) \mid m \in \mathit{pred}_G(n) \} & \text{otherwise} \end{cases} \\ \mathbf{post}(n) &= \llbracket n \rrbracket(\mathbf{pre}(n)) \end{aligned}$$

Denoting the greatest solution of Equation System 2.1 with respect to the start information $c_0 \in \mathcal{C}$ by \mathbf{pre}_{c_0} and \mathbf{post}_{c_0} , the solution of the *MFP*-strategy is defined by:

$$\mathbf{The\ MFP-Solution:} \quad \forall n \in N \quad \forall c_0 \in \mathcal{C}. \quad MFP_{c_0}(n) = \mathbf{pre}_{c_0}(n)$$

In general, this leads to a suboptimal but algorithmic description.

Thus we have two global notions of semantics here, an operational one, which precisely mimics our intention, and a denotational one, which has an algorithmic character. In fact, we consider the *MOP*-strategy as a mean for the direct specification of data flow analysis problems, and the

MFP-strategy as an algorithmic realization of such problems.⁴ This view rises the question of *correctness* (safety) or even *completeness* (optimality) of such algorithms. For the elegant answer to these questions we need two further notions: given a complete semi-lattice $(\mathcal{C}, \sqcap, \sqsupseteq, \perp, \top)$, a function $f : \mathcal{C} \rightarrow \mathcal{C}$ is called

- *monotonic* iff $\forall c, c' \in \mathcal{C}. c \sqsubseteq c' \text{ implies } f(c) \sqsubseteq f(c')$
- *distributive* iff $\forall C' \subseteq \mathcal{C}. f(\sqcap C') = \sqcap \{f(c) \mid c \in C'\}$

It is well-known that distributivity is a stronger requirement than monotonicity in the following sense:

Lemma 2.2

A function $f : \mathcal{C} \rightarrow \mathcal{C}$ is monotonic iff $\forall C' \subseteq \mathcal{C}. f(\sqcap C') \sqsubseteq \sqcap \{f(c) \mid c \in C'\}$

As in this lemma, \mathcal{C} will always denote a complete semi-lattice. We have (cf. [KU]):

Theorem 2.3 (Safety Theorem)

Given a flow graph $G = (N, E, \mathbf{s}, \mathbf{e})$, the *MFP-solution* is a correct (or safe) approximation of the *MOP-solution*, i.e. $\forall n \in N \forall c_0 \in \mathcal{C}. MFP_{c_0}(n) \sqsubseteq MOP_{c_0}(n)$, if all the semantic functions $\llbracket n \rrbracket$ are monotonic.

Distributivity of the semantic functions yields completeness (optimality). This follows from the well-known intraprocedural Coincidence Theorem 2.4 of [KU]:

Theorem 2.4 (Coincidence Theorem)

Given a flow graph $G = (N, E, \mathbf{s}, \mathbf{e})$, the *MFP-solution* is complete (or optimal) for the *MOP-solution*, i.e. $\forall n \in N \forall c_0 \in \mathcal{C}. MOP_{c_0}(n) = MFP_{c_0}(n)$, if all the semantic functions $\llbracket n \rrbracket$ are distributive.

3 Interprocedural Notions

In the interprocedural setting we represent programs Π as systems $(\pi_0, \pi_1, \dots, \pi_k)$ of (recursive) procedure definitions, where every $\pi \in \Pi$ has a list of formal value parameters and a list of local variables. π_0 is assumed to denote the *main program* and therefore cannot be called. π_1 up to π_k are the *procedure declarations* of Π . For simplicity we assume that there is no (static) procedure nesting except that π_0 encloses π_1 up to π_k .⁵ Thus, the variables of the main program are global variables of the procedures, and can be accessed by them.

The denotational (*IMFP*) approach and the operational (*IMOP*) approach require different representations of programs Π : *flow graph systems* and *interprocedural flow graphs*.

Flow Graph Systems

The denotational approach works on systems $S = (G_0, G_1, \dots, G_k)$ of *flow graphs* with disjoint sets of nodes N_i and edges E_i , in which every procedure π of Π (including the main program π_0) is represented as a *directed flow graph* $G = (N, E, \mathbf{s}, \mathbf{e})$ in the sense of Section 2. $\mathbf{N}^S =_{df} \bigcup \{N_i \mid i \in \{0, \dots, k\}\}$ denotes the set of all nodes of S , $\mathbf{E}^S =_{df} \bigcup \{E_i \mid i \in \{0, \dots, k\}\}$ the set of all edges of S , and $\mathbf{N}_C^S \subseteq \mathbf{N}^S$ the set of all nodes representing procedure calls. Finally, we need the following functions, where \mathcal{P} denotes the power set operator:

- $fg : \mathbf{N}^S \rightarrow S$ with $fg(n) =_{df} G_i$ iff $n \in N_i$,

⁴Explicit algorithms are presented in Appendix A.

⁵Integrating static procedure nesting is straightforward.

- $callee : \mathbf{N}_C^S \rightarrow S$ with $callee(n) =_{df} G_i$ iff n represents a procedure call of G_i ,
- $caller : S \rightarrow \mathcal{P}(\mathbf{N}_C^S)$ with $caller(G_i) =_{df} \{n \mid callee(n) = G_i\}$,
- $start : S \rightarrow \{\mathbf{s}_0, \dots, \mathbf{s}_k\}$ with $start(G_i) =_{df} \mathbf{s}_i$ for all $i \in \{0, \dots, k\}$ and
- $end : S \rightarrow \{\mathbf{e}_0, \dots, \mathbf{e}_k\}$ with $end(G_i) =_{df} \mathbf{e}_i$ for all $i \in \{0, \dots, k\}$.

Intuitively, fg maps every node of a flow graph system to its corresponding flow graph, $callee$ every call node to the called procedure, $caller$ every procedure to its set of call nodes, and $start$ and end every procedure to its start node and its end node, respectively.

An illustrative flow graph system is given in Figure 1.

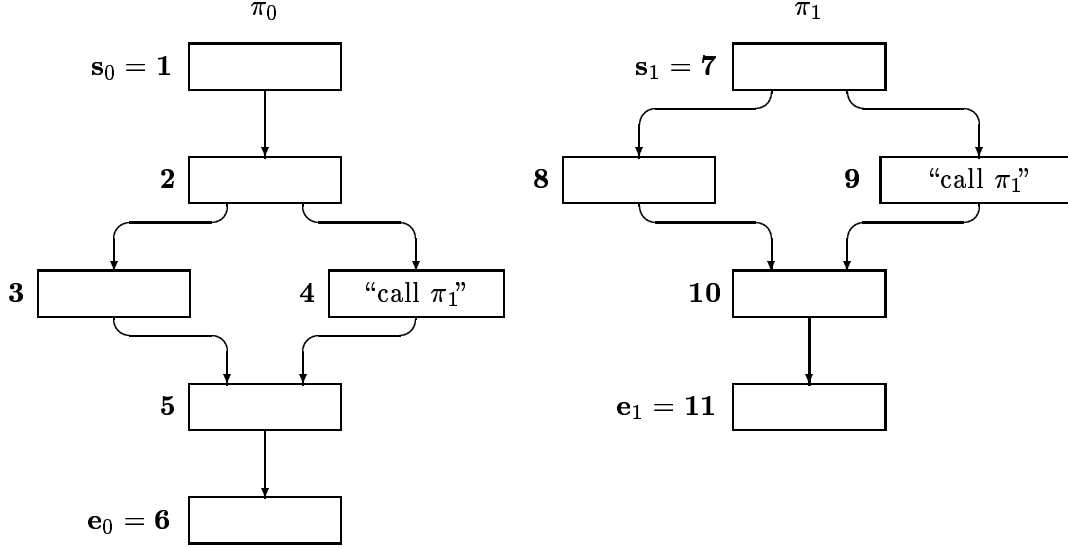


Figure 1: The Flow Graph System

Interprocedural Flow Graphs

The operational approach requires an explicit representation of the interprocedural control flow caused by procedure calls. This is achieved by combining the flow graphs of S to an *interprocedural flow graph* $G^* = (N^*, E^*, \mathbf{s}^*, \mathbf{e}^*)$, where \mathbf{s}^* is given by \mathbf{s}_0 and \mathbf{e}^* by \mathbf{e}_0 (cf. [My, SP]). In detail, G^* results from S by applying the following two step procedure to every node n of \mathbf{N}_C^S :

Algorithm 3.1 Let S be a flow graph system, and $n \in \mathbf{N}_C^S$. Then

1. Replace n by two new nodes, the call node n_C and the return node n_R such that n_C has the same set of predecessors as n but no successors and n_R has the same set of successors as n but no predecessors.⁶
2. Draw an edge from n_C to $start(callee(n))$ and from $end(callee(n))$ to n_R .

N_C^* and N_R^* denote the set of all call nodes and return nodes in N^* , respectively, and $pred^*(n) =_{df} \{m \mid (m, n) \in E^*\}$ the set of all immediate interprocedural predecessors of n . In the following, we will identify the set \mathbf{N}^S of nodes of S with the set $N^* \setminus N_R^*$ of nodes of G^* to get an interpretation independent notion of program point.

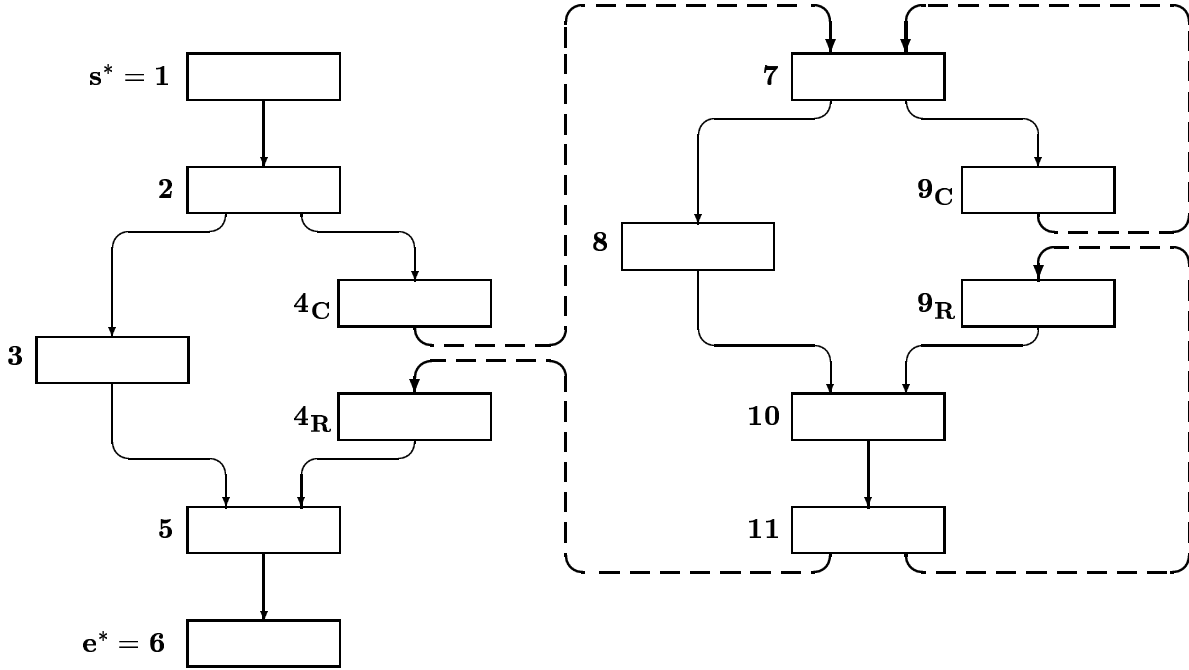


Figure 2: The Interprocedural Flow Graph

Figure 2 shows the interprocedural flow graph that corresponds to the flow graph system of Figure 1.

Interprocedural Paths

The notion of finite path as introduced in Section 2 naturally applies to interprocedural flow graphs as well. However, due to the special nature of procedure calls not every finite path $p \in \mathbf{P}[m, n]$ represents a valid execution. For example in Figure 2 the path $(2, 4_C, 7, 8, 10, 11, 4_R)$ is possible, while the path $(2, 4_C, 7, 8, 10, 11, 9_R)$ is not. This led to the following definition of interprocedural (valid) paths [SP]:

Definition 3.2 (Interprocedural Path)

1. Let $p \in \mathbf{P}[m, n]$. Then p is an interprocedural path iff the tuple (m_1, \dots, m_r) , which results from p by deleting all nodes in $N^* \setminus (N_C^* \cup N_R^*)$, is well-formed in the following sense:
 - if there is no return node in (m_1, \dots, m_r) , then (m_1, \dots, m_r) is well-formed,
 - otherwise, let j be the smallest index in $\{1, \dots, r\}$ such that m_j is a return node. Then (m_1, \dots, m_r) is well-formed iff $j > 1$ and m_{j-1} is a call node matching m_j , and the remaining sequence after deleting m_{j-1} and m_j is well-formed too.
2. A call node and a return node of p are said to correspond to each other, if they are eliminated simultaneously in the procedure above.
3. $\mathbf{IP}[m, n]$ denotes the set of all interprocedural paths from m to n , and $\mathbf{IP}[m, n)$ the set of all interprocedural paths from m to a predecessor of n .

⁶ n_C is called to match n_R and vice versa.

Complete Interprocedural Paths

In order to determine the semantics of procedure calls, we need to deal with complete interprocedural paths p from $start(fg(n))$ to n , which are characterized by the fact that all procedure calls in p have been completed by a subsequent return. This guarantees that the occurrences of $start(fg(n))$ and n belong to the same procedure incarnation.

Definition 3.3 (Complete Interprocedural Path)

1. An interprocedural path $p = (n_1, \dots, n_k) \in \mathbf{IP}[start(fg(n)), n]$ is called complete if it possesses equally many occurrences of procedure call and return nodes:

$$|\{i \mid n_i \in N_C^*\}| = |\{i \mid n_i \in N_R^*\}|$$

2. $\mathbf{CIP}[start(fg(n)), n]$ and $\mathbf{CIP}[start(fg(n)), n]$ denote the set of all complete interprocedural paths from $start(fg(n))$ to n , and from $start(fg(n))$ to a predecessor of n , respectively.

That this actually realizes our intention is a consequence of the following property of interprocedural paths:

Lemma 3.4 *Let $p = (n_1, \dots, n_k) \in \mathbf{IP}[m, n]$ be an interprocedural path and (n_i, n_j) and $(n_{i'}, n_{j'})$ two of its pairs of corresponding call and return nodes. Then the integer intervals $[i : j]$ and $[i' : j']$ are either disjoint or one is included in the other.*

This pattern is illustrated in Figure 3, where $\{(n_{C_i}, n_{R_i}) \mid i \in \{1, \dots, 4\}\}$ are assumed to be pairs of corresponding call and return nodes of p .

$$p = (m, \dots, n_{C_1}, \dots, n_{C_2}, \dots, n_{R_2}, \dots, n_{C_3}, \dots, n_{C_4}, \dots, n_{R_4}, \dots, n_{R_3}, \dots, n_{R_1}, \dots, n) \in \mathbf{IP}[m, n]$$

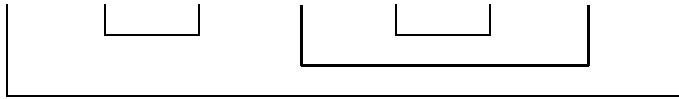


Figure 3: Complete Interprocedural Paths

The following lemma, which can easily be proved, will be important:

Lemma 3.5 *Let $s \in \{s_0, \dots, s_k\}$, $p = (n_1, \dots, n_k) \in \mathbf{IP}[s, n]$ and (n_i, n_j) a pair of corresponding call and return nodes. Then we have:*

$$(n_{i+1}, \dots, n_{j-1}) \in \mathbf{CIP}[start(callee(n_i)), end(callee(n_i))]$$

Remark 3.6 If the underlying program Π has no procedures, the flow graph system S and the interprocedural flow graph G^* collapse to the flow graph G_0 of π_0 . In this special case our framework coincides with the standard intraprocedural framework.

Conventions: Throughout the rest of this paper we assume an arbitrary but fixed program $\Pi = (\pi_0, \pi_1, \dots, \pi_k)$ with flow graph system $S = (G_0, G_1, \dots, G_k)$ and interprocedural flow graph $G^* = (N^*, E^*, s^*, e^*)$, where every node n of G^* is assumed to lie on an interprocedural path from s^* to e^* . Moreover, m and n , possibly indexed, are nodes of S or G^* , and for every node $n \in N_C^S$, n_C and n_R denote its corresponding call node and return node in N^* , respectively.

4 Abstract Semantics

In this section we present new *interprocedural* versions of the meet over all paths strategy and the maximal fixed point strategy. They define the (global) semantics for interprocedural flow graphs and flow graph systems, respectively. The point of this presentation is the extension of the data flow information in a way that mimics run-time stacks as used in run-time systems.

4.1 The Local Semantic Functional

As its intraprocedural counterpart, the interprocedural meet over all paths (*IMOP*) solution directly records all possible program executions that lead to a particular program point. However, in the presence of recursive procedures it is necessary to work on stacks of lattice elements instead of the lattice itself, in order to record the part of the history which will become relevant after returning from (nested) procedure calls. Thus the local semantic functional has type

$$\llbracket \cdot \rrbracket : N^* \rightarrow (STACK \rightarrow STACK)$$

where *STACK* denotes the set of all non empty stacks with components of \mathcal{C} , which can be manipulated by means of the following operations:

- *newstack* : $\mathcal{C} \rightarrow STACK$
- *push* : $STACK \times \mathcal{C} \rightarrow STACK$
- *pop* : $STACK \rightarrow STACK$
- *top* : $STACK \rightarrow \mathcal{C}$

Intuitively, *newstack*(*c*) creates a new stack with single component *c*, *push* puts a new component on the top of the argument stack, *pop* removes the top component, and *top* delivers the content of the top component, while not affecting the argument stack.⁷ Thus only the top components of the stacks can be affected by these operations.

STACK is an abstract version of the run-time stacks used by run-time systems for maintaining the activation records of different procedure incarnations. Intuitively, the top component of a stack contains the data flow information corresponding to the currently valid activation record,⁸ while the data flow informations of the remaining stack components correspond to activation records of preceding but not yet finished procedure calls. However, in contrast to a concrete run-time stack, where variables that are global for the currently activated procedure are accessed by means of static and dynamic link chains, the components of a data flow analysis stack are assumed to contain all information related with the current procedure incarnation, i.e. also the information related to global variables.⁹ Thus the data flow analysis stacks directly reflect the nesting of procedure incarnations according to the current call sequence.

Formally, the local semantic functional $\llbracket \cdot \rrbracket^*$ for this setting is defined by

$$\forall n \in N^* \forall stk \in STACK.$$

$$\llbracket n \rrbracket^*(stk) =_{df} \begin{cases} push(pop(stk), \llbracket n \rrbracket'(top(stk))) & \text{if } n \in N^* \setminus (N_C^* \cup N_R^*) \\ push(stk, \llbracket n \rrbracket'(top(stk))) & \text{if } n \in N_C^* \\ push(pop(pop(stk)), \mathcal{R}_n(top(pop(stk))), \llbracket n \rrbracket'(top(stk))) & \text{if } n \in N_R^* \end{cases}$$

⁷We consider the operation *newstack* instead of the usual *emptystack* : $\rightarrow STACK$ here, in order to exclude empty stacks, which are irrelevant in our framework.

⁸Therefore, we are never dealing with empty stacks.

⁹Static and dynamic link chains are just a technical mean for getting efficient implementations of run-time systems. In our abstract framework, however, this aspect can be neglected without any harm (cf. Remark 4.7). Moreover, it allows us to work with local semantic functionals that affect only the top components of data flow analysis stacks.

where $\llbracket \cdot \rrbracket' : N^* \rightarrow (\mathcal{C} \rightarrow \mathcal{C})$ denotes the straightforward extension of the semantic functional of Section 2 to interprocedural flow graphs,¹⁰ and $\mathcal{R}_n : \mathcal{C} \times \mathcal{C} \rightarrow \mathcal{C}$ is a function as described below.

The intuition behind this definition is as follows:

The execution of an ordinary statement (i.e. $n \in N^* \setminus (N_C^* \cup N_R^*)$) only affects the currently valid activation record. Thus it can be modelled by simply modifying the top component of the stack representing the current data flow information.

A procedure call (i.e. $n \in N_C^*$) requires the generation of a new activation record. This is reflected by pushing a new element on the top of the stack, which results from modifying the top component of the stack according to the parameter transfer.

The treatment of return statements (i.e. $n \in N_R^*$) demonstrates the necessity of introducing stacks into the framework. Returning from a procedure call (i.e. $n \in N_R^*$) essentially requires removal of the activation record belonging to the called procedure and reactivation of its predecessor. However, one observation is important here. The effect of a (directly) recursive procedure to a global variable needs to be maintained, whereas the local variables must be reset to their values at call time. Thus we need to consider the data flow information valid immediately before entering the procedure (available in $top(pop(stk))$), as well as the information valid after executing its body (available in $\llbracket n \rrbracket'(top(stk))$), in order to compute the data flow information being valid after returning from the called procedure. The function $\mathcal{R}_n : \mathcal{C} \times \mathcal{C} \rightarrow \mathcal{C}$ models this computation. Thus popping the top component of the stack and replacing the subsequent component by

$$\mathcal{R}_n(top(pop(stk)), \llbracket n \rrbracket'(top(stk)))$$

reflects the whole process of completing a procedure call.

4.2 The Structure of the Semantic Functions

Let $\mathcal{F} =_{df} [STACK \rightarrow STACK]$ denote the set of all functions from $STACK$ to $STACK$ and

$$\mathcal{F}_O =_{df} \{ f \in \mathcal{F} \mid \forall stk \in STACK. pop(f(stk)) = pop(stk) \}$$

$$\mathcal{F}_C =_{df} \{ f \in \mathcal{F} \mid \forall stk \in STACK. pop(f(stk)) = stk \}$$

$$\mathcal{F}_R =_{df} \{ f \in \mathcal{F} \mid \forall stk \in STACK. pop(f(stk)) = pop(pop(stk)) \}$$

Then we have:

Lemma 4.1

1. $\forall n \in N^* \setminus (N_C^* \cup N_R^*). \llbracket n \rrbracket^* \in \mathcal{F}_O$
2. $\forall n \in N_C^*. \llbracket n \rrbracket^* \in \mathcal{F}_C$
3. $\forall n \in N_R^*. \llbracket n \rrbracket^* \in \mathcal{F}_R$

Intuitively this means that the semantic function of an ordinary statement only affects the top component of the argument stack, that the semantic function of a call statement simply adds a new top component to the argument stack, and that a return statement replaces the upper two components of the argument stack by a new component. The following lemma is an easy consequence of these properties of \mathcal{F}_O , \mathcal{F}_C and \mathcal{F}_R .

¹⁰ $\llbracket n \rrbracket^*(stk)$, $n \in N_R^*$, is only defined for stacks with at least two components, a fact, which is automatically taken care of in any reasonable analysis context.

Lemma 4.2 $\forall f_r \in \mathcal{F}_R \quad \forall f_o, f'_o \in \mathcal{F}_O \quad \forall f_c \in \mathcal{F}_C. \quad f_o \circ f'_o, \quad f_r \circ f_o \circ f_c \in \mathcal{F}_O$

The formal development of the paper requires the following derived notions of monotonicity and distributivity:

Definition 4.3 (S-Monotonicity, S-Distributivity)

A function $f \in \mathcal{F}_O \cup \mathcal{F}_C \cup \mathcal{F}_R$ is called

- s-monotonic iff f_s is monotonic
- s-distributive iff f_s is distributive

where f_s , the significant part of f , is defined according to the following two cases:

- $f \in \mathcal{F}_O \cup \mathcal{F}_C$: here $f_s : \mathcal{C} \rightarrow \mathcal{C}$ is defined by: $f_s(c) =_{df} \text{top}(f(\text{newstack}(c)))$
- $f \in \mathcal{F}_R$: here $f_s : \mathcal{C} \times \mathcal{C} \rightarrow \mathcal{C}$ is defined by:¹¹ $f_s(c_1, c_2) =_{df} \text{top}(f(\text{push}(\text{newstack}(c_1), c_2)))$

The following lemma shows that the effort for checking the preconditions of the Interprocedural Safety Theorem 5.2 and the Interprocedural Coincidence Theorem 5.3 is comparable to the effort necessary for their intraprocedural counterparts (cf. Section 2 and 5).

Lemma 4.4 For all $n \in N^*$ we have that $\llbracket n \rrbracket^*$ is s-monotonic (s-distributive) if

- $n \in N_R^* : \llbracket n \rrbracket'$ and \mathcal{R}_n are monotonic (distributive)
- $n \notin N_R^* : \llbracket n \rrbracket'$ is monotonic (distributive)

Conventions: In the following we consider s-monotonicity (s-distributivity) as a generalization of the usual monotonicity (distributivity) by identifying lattice elements with their unique representations as one-component stacks. Moreover, we extend the meet operation \sqcap to work on stacks in the following way:

$$\forall STK \subseteq STACK. \quad \sqcap STK =_{df} \text{newstack}(\sqcap \{ \text{top}(stk) \mid stk \in STK \})$$

Thus, the meet over a set of stacks is just the one-component stack containing the meet of all the top components in its single component.

4.3 The Interprocedural Meet Over all Paths Solution

Analogously to Section 2 the local abstract semantics $\llbracket \cdot \rrbracket^*$ can be extended to cover finite interprocedural paths. For every path $p = (n_1, \dots, n_q) \in \mathbf{IP}[m, n]$, we define $\llbracket p \rrbracket^* : STACK \rightarrow STACK$ by

$$\llbracket p \rrbracket^* =_{df} \begin{cases} id_{STACK} & \text{if } p \equiv \varepsilon \\ \llbracket (n_2, \dots, n_q) \rrbracket^* \circ \llbracket n_1 \rrbracket^* & \text{otherwise} \end{cases}$$

Now, as its intraprocedural counterpart (cf. [KU]), the interprocedural meet over all paths (*IMOP*) solution directly records all possible program executions leading to a particular program point. Here it is important that for any interprocedural path $p \in \mathbf{IP}[s^*, n]$ and any stack $stk \in STACK$, $\text{top}(\llbracket p \rrbracket^*(stk))$ is the only data flow information relevant for node n after executing p , since all other components of $\llbracket p \rrbracket^*(stk)$ correspond to activation records that are not valid after p . Identifying one-component stacks with the content of their unique component, the formal definition of the interprocedural meet over all paths solution is given by:

The *IMOP*-Solution:

$$\forall n \in N^* \quad \forall c_0 \in \mathcal{C}. \quad IMOP_{c_0}(n) = \sqcap \{ \llbracket p \rrbracket^*(\text{newstack}(c_0)) \mid p \in \mathbf{IP}[s^*, n] \}$$

¹¹Note that $\mathcal{C} \times \mathcal{C}$ is a lattice, whenever \mathcal{C} is.

4.4 The Interprocedural Maximal Fixed Point Solution

In addition to the equational characterization of the intraprocedural case (Equation System 2.1), flow graph systems need a preprocess, which determines the meaning of call nodes in terms of the meaning of the called procedures. This requires the introduction of an auxiliary semantic functional $\llbracket \cdot \rrbracket$, which gives meaning to whole flow graphs. Essentially, $\llbracket n \rrbracket$ transforms data flow information that is assumed to be valid at the entry of the procedure that contains n into the corresponding data flow information being valid before an execution of n . In particular, $\llbracket e_i \rrbracket$ is the meaning function of the i -th procedure.¹² Formally, the full preprocess for determining the meaning $\llbracket n \rrbracket$ of call nodes $n \in \mathbf{N}_C^S$ is characterized by:

Definition 4.5 $\llbracket \cdot \rrbracket : \mathbf{N}^S \rightarrow (STACK \rightarrow STACK)$ and $\llbracket \cdot \rrbracket : \mathbf{N}^S \rightarrow (STACK \rightarrow STACK)$ are defined as the greatest solution of the equation system given by:

$$\llbracket n \rrbracket =_{df} \begin{cases} id_{STACK} & \text{if } n \in \{s_0, \dots, s_k\} \\ \sqcap \{ \llbracket m \rrbracket \circ \llbracket m \rrbracket \mid m \in pred_{fg(n)}(n) \} & \text{otherwise} \end{cases}$$

and

$$\llbracket n \rrbracket =_{df} \begin{cases} \llbracket n \rrbracket^* & \text{if } n \in \mathbf{N}^S \setminus \mathbf{N}_C^S \\ \llbracket n_R \rrbracket^* \circ \llbracket end(callee(n)) \rrbracket \circ \llbracket n_C \rrbracket^* & \text{otherwise} \end{cases}$$

where id_{STACK} denotes the identity on $STACK$, and \sqcap the “componentwise” meet operation on \mathcal{F}_O .¹³

The effect of a procedure call $n \in \mathbf{N}_C^S$ is determined in three steps reflecting the three phases of its execution:

- *Entering* the called procedure: $\llbracket n_C \rrbracket^*$ creates a new activation record by transforming the content of the top component of the stack according to the semantics of the call node and pushing it onto the stack. – Usually, the semantics of call nodes will reflect the parameter transfer.
- *Evaluating* the call: $\llbracket end(callee(n)) \rrbracket$ computes the effect of the procedure body. Note that this affects the top component of the argument stack only.
- *Leaving* the called procedure: $\llbracket n_R \rrbracket^*$ removes the activation record related with the current procedure call by popping the top component from the stack, and replacing its subsequent component by the data flow information representing the effect of the procedure call relative to its call site.

Applying Lemma 4.2, we obtain

Lemma 4.6 $\forall n \in \mathbf{N}^S. \llbracket n \rrbracket, \llbracket n \rrbracket \in \mathcal{F}_O$

Remark 4.7 Lemma 4.6 is important, since it shows that all the stacks occurring during the iterative computation of the *IMFP*-solution will have at most two components.¹⁴ This is in contrast to the *IMOP*-strategy, where the size of stacks contributing to the *IMOP*-solution is in general unbounded. Moreover, it allows us to prove termination in the usual way.

¹²Remember, $\llbracket e_i \rrbracket' =_{df} id_C$. Thus, e_i is related to the identity on $STACK$.

¹³ $\forall f, f' \in \mathcal{F}_O. f \sqcap f' =_{df} f'' \in \mathcal{F}_O$ with $\forall stk \in STACK. top(f''(stk)) = top(f(stk)) \sqcap top(f'(stk))$. As usual, “ \sqcap ” induces an inclusion relation “ \sqsubseteq ” on \mathcal{F}_O by: $f \sqsubseteq f'$ iff $f \sqcap f' = f$.

¹⁴This is because the computation starts from a one-component stack $newstack(c_0)$ (cf. Equation System 4.8 and Algorithm A.3).

After fixing the meaning of call nodes, $\llbracket \cdot \rrbracket$ plays essentially the same role as the local (abstract) semantic functional of Section 2. Formally, the interprocedural maximal fixed point strategy is characterized by Equation System 4.8. As its intraprocedural counterpart, this strategy labels every node n of \mathbf{N}^S with a pre-information $\mathbf{pre}_{c_0}(n)$ and a post-information $\mathbf{post}_{c_0}(n)$, whose top components are the greatest solution of this equation system with respect to $c_0 \in \mathcal{C}$.

Equation System 4.8

$$\begin{aligned} \mathbf{pre}(n) &= \begin{cases} \mathit{newstack}(c_0) & \text{if } n = \mathbf{s}_0 \\ \sqcap \{ \llbracket m_C \rrbracket^*(\mathbf{pre}(m)) \mid m \in \mathit{caller}(fg(n)) \} & \text{if } n \in \{\mathbf{s}_1, \dots, \mathbf{s}_k\} \\ \sqcap \{ \mathbf{post}(m) \mid m \in \mathit{pred}_{fg(n)}(n) \} & \text{otherwise} \end{cases} \\ \mathbf{post}(n) &= \llbracket n \rrbracket(\mathbf{pre}(n)) \end{aligned}$$

As before, identifying a stack having a single component only with the content of this component, we obtain as in the intraprocedural case:

The *IMFP*-Solution: $\forall n \in \mathbf{N}^S \forall c_0 \in \mathcal{C}. \mathit{IMFP}_{c_0}(n) = \mathbf{pre}_{c_0}(n)$

5 Main Results

The main step in the proof of our main results is taken by proving the following Main Lemma 5.1, whose proof is given in full detail in [KS1].

Lemma 5.1 (The Main Lemma)

For all $n \in \mathbf{N}_{\mathcal{C}}^S$, we have, if the semantic functions $\llbracket m \rrbracket^*$, $m \in N^*$, are

1. *s-monotonic*: $\llbracket n \rrbracket \sqsubseteq \sqcap \{ \llbracket p \rrbracket^* \mid p \in \mathbf{CIP}[n_C, n_R] \}$
2. *s-distributive*: $\llbracket n \rrbracket = \sqcap \{ \llbracket p \rrbracket^* \mid p \in \mathbf{CIP}[n_C, n_R] \}$

After having established this result, the Interprocedural Safety Theorem 5.2 and the Interprocedural Coincidence Theorem 5.3 can be proved almost as in the intraprocedural case. Thus we omit these proofs here.¹⁵

As in the intraprocedural case, the first theorem states that the *IMFP*-solution is a correct approximation of the *IMOP*-solution, whenever all the local abstract semantic functions are *s-monotonic*:

Theorem 5.2 (Interprocedural Safety Theorem)

Given a flow graph system $S = (G_0, G_1, \dots, G_k)$ and its corresponding interprocedural flow graph $G^* = (N^*, E^*, \mathbf{s}^*, \mathbf{e}^*)$, the *IMFP*-solution is a correct approximation of the *IMOP*-solution, i.e. $\forall n \in \mathbf{N}^S \forall c_0 \in \mathcal{C}. \mathit{IMFP}_{c_0}(n) \sqsubseteq \mathit{IMOP}_{c_0}(n)$, if the abstract semantics $\llbracket n \rrbracket^*$ of all nodes $n \in N^*$ is given by an *s-monotonic* function.

Again, as in the intraprocedural case, *s-distributivity* of the semantic functions yields optimality (or completeness):

Theorem 5.3 (Interprocedural Coincidence Theorem)

Given a flow graph system $S = (G_0, G_1, \dots, G_k)$ and its corresponding interprocedural flow graph $G^* = (N^*, E^*, \mathbf{s}^*, \mathbf{e}^*)$, the *IMFP*-solution and the *IMOP*-solution coincide, i.e. $\forall n \in \mathbf{N}^S \forall c_0 \in \mathcal{C}. \mathit{IMOP}_{c_0}(n) = \mathit{IMFP}_{c_0}(n)$, if the abstract semantics $\llbracket n \rrbracket^*$ of all nodes $n \in N^*$ is given by an *s-distributive* function.

¹⁵Both proofs are given in [KS1].

Note that Lemma 4.4 allows to check the s-monotonicity or s-distributivity of the semantic functions $\llbracket n \rrbracket^*$ simply by checking these properties for the semantic functions $\llbracket n \rrbracket'$ and the reduction functions \mathcal{R}_n . Thus the only additional effort in comparison to the intraprocedural case arises from checking the reduction functions.

6 Applications

In this section we sketch two applications of the Interprocedural Coincidence Theorem 5.3. We omit details here, since both examples require their own setup.

In [SK2] we propose an algorithm for *interprocedural constant propagation* and *constant folding*, which generalizes and improves all previous techniques for interprocedural constant propagation (cf. [CC2, CCKT, JM]). This algorithm determines all *finite interprocedural constants*, which are the interprocedural analogue to the set of finite constants introduced in [SK1]. As in the intraprocedural case, finite interprocedural constants have a purely *operational* characterization in the sense of the *IMOP*-strategy, and a purely *denotational* characterization in the sense of the *IMFP*-strategy. The Interprocedural Coincidence Theorem 5.3 yields the equivalence of these characterizations.

The second example concerns the interprocedural versions of the classical bit-vector data flow analyses, e.g. determining available expressions, reaching definitions, live variables, very busy (anticipatable) expressions (cf. [He]),¹⁶ and, more sophisticatedly, the optimal elimination of interprocedural partial redundancies.¹⁷ In all these cases, the Interprocedural Coincidence Theorem 5.3 allows us to prove the optimality of our algorithms for programs with recursive procedures, global and local variables, and formal value parameters [KS2].¹⁸

7 Conclusions

We have presented an interprocedural generalization of the well-known intraprocedural Coincidence Theorem of Kam and Ullman [KU], which covers arbitrary programs with recursive procedures, global and local variables, and formal value parameters. Our theorem, which reduces to the classical intraprocedural version in the absence of procedures, delivers a sufficient condition for the coincidence of the *interprocedural meet over all paths* strategy and the *interprocedural maximal fixed point* strategy, and it generalizes previous results (cf. [Ba1, Ba2, SP]), which do not deal properly with local variables of recursive procedures. Our results are formulated within the framework of abstract interpretation, thus covering a wide range of data flow analyses.

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¹⁶This application has also been suggested (without giving any details) by Sharir and Pnueli [SP].

¹⁷This problem is heuristically dealt with in [Mo, MR].

¹⁸A detailed presentation of interprocedural bit-vector data flow analyses is given in [KS3].

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A Algorithms

This section provides algorithms that compute the *IMFP*-solution. First, we have an algorithm for the preprocess that determines the semantics of call nodes:

Algorithm A.1 (Computation of the Semantic Functionals $\llbracket \cdot \rrbracket$ and $\llbracket \cdot \rrbracket$)

Input: A flow graph system $S = (G_0, G_1, \dots, G_k)$, a complete semi-lattice \mathcal{C} , for every node $n \in \mathbf{N}^S \setminus \mathbf{N}_C^S$ an s -monotonic function $\llbracket n \rrbracket^* : STACK \rightarrow STACK \in \mathcal{F}_O$, which is the identity for all nodes in $\{\mathbf{s}_0, \dots, \mathbf{s}_k, \mathbf{e}_0, \dots, \mathbf{e}_k\}$. Moreover, for every node $n \in \mathbf{N}_C^S$ two s -monotonic functions $\llbracket n_C \rrbracket^* : STACK \rightarrow STACK \in \mathcal{F}_C$ and $\llbracket n_R \rrbracket^* : STACK \rightarrow STACK \in \mathcal{F}_R$.

Output: An annotation of S with functions $\llbracket n \rrbracket : \mathcal{C} \rightarrow \mathcal{C}$ (stored in *gtr*, which stands for global transformation), and $\llbracket n \rrbracket : \mathcal{C} \rightarrow \mathcal{C}$ (stored in *ltr*, which stands for local transformation) representing the greatest solution of Equation System 4.5.

Remark: $\top_{\mathcal{F}_O} : STACK \rightarrow STACK \in \mathcal{F}_O$ denotes the “universal” function which is assumed to “contain” every function $f \in \mathcal{F}_O$, and id_{STACK} is the identity on $STACK$. The variable *workset* controls the iterative process. Its elements are tuples, whose first components are nodes $m \in \mathbf{N}^S$ of the flow graph system S , and whose second components are functions $f : STACK \rightarrow STACK \in \mathcal{F}_O$ that specify a new approximation for the function $\llbracket m \rrbracket$ of the node of the first component. Note that due to the mutual interdependence of the definitions of $\llbracket \cdot \rrbracket$ and $\llbracket \cdot \rrbracket$ the iterative approximation of $\llbracket \cdot \rrbracket$ is superposed by an interprocedural iteration step which updates the semantics $\llbracket \cdot \rrbracket$ of call nodes.

(Initialization of the annotation arrays *gtr* and *ltr* and the variable *workset*)

FORALL $m \in \mathbf{N}^S$ **DO**

$gtr[m] := \top_{\mathcal{F}_O}$;

IF $m \in \mathbf{N}_C^S$

THEN $ltr[m] := \llbracket m_R \rrbracket^* \circ \top_{\mathcal{F}_O} \circ \llbracket m_C \rrbracket^*$

ELSE $ltr[m] := \llbracket m \rrbracket^*$

FI

OD;

$workset := \{(\mathbf{s}, id_{STACK}) \mid \mathbf{s} \in \{\mathbf{s}_0, \dots, \mathbf{s}_k\}\} \cup$
 $\{(n, f) \mid f = \sqcap \{ltr[m] \circ gtr[m] \mid m \in pred_{fg}(n)(n)\} \sqsubset \top_{\mathcal{F}_O}\}$

(Iterative fixed point computation)

WHILE $workset \neq \emptyset$ **DO**

LET $(m, f) \in workset$

```

BEGIN
   $workset := workset \setminus \{(m, f)\};$ 
  IF  $gtr[m] \sqsupseteq gtr[m] \sqcap f$ 
    THEN
       $gtr[m] := gtr[m] \sqcap f;$ 
      IF  $m \in \{e_i \mid i \in \{0, \dots, k\}\}$ 
        THEN
          FORALL  $l \in caller(fg(m))$  DO
             $ltr[l] := ltr[l] \sqcap \llbracket l_R \rrbracket^* \circ gtr[m] \circ \llbracket l_C \rrbracket^*;$ 
             $workset := workset \cup \{(n, ltr[l] \circ gtr[l]) \mid n \in succ_{fg}(l)\}$ 
          OD
        ELSE
           $workset := workset \cup \{(n, ltr[m] \circ gtr[m]) \mid n \in succ_{fg}(m)\}$ 
        FI
      FI
    END
  OD.

```

In order to simplify the formulation of the central property of this algorithm, we abbreviate the values of $ltr[n]$ and $gtr[n]$ after the k -th execution of the while-loop by $ltr^k[n]$ and $gtr^k[n]$, respectively. The following theorem can now be proved in a straightforward fashion (cf. [Ki]):

Theorem A.2 $\forall n \in \mathbf{N}^S. \llbracket n \rrbracket = \sqcap \{ltr^k[n] \mid k \geq 0\} \wedge \llbracket n \rrbracket = \sqcap \{gtr^k[n] \mid k \geq 0\}$
In particular, we have $\forall n \in \mathbf{N}^S. \llbracket n \rrbracket = gtr[n] \wedge \llbracket n \rrbracket = ltr[n]$ after termination of Algorithm A.1.

The second algorithm computes the *IMFP*-Solution:

Algorithm A.3 (Computation of the *IMFP*-Solution)

Input: A flow graph system $S = (G_0, G_1, \dots, G_k)$, the semantic functional $\llbracket \cdot \rrbracket$, for every node $n \in \mathbf{N}_C^S$ the function $\llbracket n_C \rrbracket^*$, and a start information $c_0 \in \mathcal{C}$.

Output: An annotation of S with data flow informations, i.e. an annotation with pre-informations (stored in *pre*) and post-informations (stored in *post*) of one-component stacks that characterize valid data flow information at the entry and at the exit of every node.

Remark: $newstack(\top)$ denotes the “universal” data flow information, which is assumed to “contain” every data flow information. The variable *workset* controls the iterative process. Its elements are tuples whose first components are nodes $m \in \mathbf{N}^S$ of the flow graph system S and whose second components are elements of *STACK* specifying a new approximation for the pre-information of the node of the first component.

(Initialization of the annotation arrays *pre* and *post* and the variable *workset*)

```

FORALL  $m \in \mathbf{N}^S$  DO ( $pre[m], post[m]$ ) := ( $newstack(\top), \llbracket m \rrbracket(newstack(\top))$ ) OD;
 $workset := \{(s_0, newstack(c_0))\} \cup$ 
   $\{(n, stk) \mid stk = \sqcap \{post[m] \mid m \in pred_{fg}(n)\} \sqcap newstack(\top)\} \cup$ 
   $\{(n, stk) \mid n \in \{s_1, s_2, \dots, s_k\} \wedge$ 
     $stk = \sqcap \{\llbracket m_C \rrbracket^*(pre[m]) \mid m \in caller(fg(n))\} \sqcap newstack(\top)\};$ 

```


(Iterative fixed point computation)

WHILE $workset \neq \emptyset$ **DO**

LET $(m, stk) \in workset$

BEGIN

$workset := workset \setminus \{(m, stk)\};$

IF $pre[m] \sqsupseteq pre[m] \sqcap stk$

THEN

$pre[m] := pre[m] \sqcap stk;$

$post[m] := \llbracket m \rrbracket(pre[m]);$

$workset := workset \cup \{(n, post[m]) \mid n \in succ_{fg(m)}(m)\};$

IF $m \in \mathbf{N}_C^S$

THEN $workset := workset \cup \{(start(callee(m)), \llbracket m_C \rrbracket^*(pre[m]))\}$

FI

FI

END

OD.

As before, given a start information c_0 , we abbreviate the values of $pre[n]$, and $post[n]$ after the k -th execution of the while-loop by $pre^k[n]$, and $post^k[n]$. In analogy to Theorem A.2 we have:

Theorem A.4 $\forall n \in \mathbf{N}^S. IMFP_{c_0}(n) = \sqcap \{pre^k[n] \mid k \geq 0\}$

In particular, we have $\forall n \in \mathbf{N}^S. IMFP_{c_0}(n) = pre[n]$ after termination of Algorithm A.3.