## Data-Flow Analysis

Data-flow analysis (DFA) provides information about how the execution of a program may manipulate its data.

Many different problems can be formulated as data-flow problems, for example:

- Which assignments to variable $\mathbf{v}$ may influence a use of $\mathbf{v}$ at a certain program position?
- Is a variable $v$ used on any path from a program position $p$ to the exit node?
- The values of which expressions are available at program position $\mathbf{p}$ ?

Data-flow problems are stated in terms of

- paths through the control-flow graph and
- properties of basic blocks.

Data-flow analysis provides information for global optimization.

## Data-flow analysis does not know

- which input values are provided at run-time,
- which branches are taken at run-time.

Its results are to be interpreted pessimistic

## Data-Flow Equations

A data-flow problem is stated as a system of equations for a control-flow graph.
System of Equations for forward problems (propagate information along control-flow edges):

Example Reaching definitions:
A definiton $d$ of a variable $v$ reaches the begin of a block B if
there is a path from d to B on which $\mathbf{v}$ is not assigned again

## In, Out, Gen, Kill represen

analysis information:
sets of statements
sets of variables,
sets of expressions
depending on the analysis problem
2 equations for each basic block:

$$
\text { Out } \begin{aligned}
(B) & =f_{B}(\ln (B)) \\
& =\operatorname{Gen}(B) \cup(\ln (B)-\text { Kill }(B))
\end{aligned}
$$

$$
\text { In }(B)=\underset{h \in \underset{\operatorname{pred}(B)}{\Theta}}{ } \text { Out (h) }
$$



In, Out variables of the system of equations for each block
Gen, Kill a pair of constant sets that characterize a block w.r.t. the DFA problem $\Theta$ meet operator; e. g. $\Theta=\cup$ for „reaching definitions", $\Theta=\cap$ for „available expressions"

## Variants of DFA Problems

- forward problem

DFA information flows along the control flow
$\ln (\mathrm{B})$ is determined by Out $(\mathrm{h})$ of the predecessor blocks
backward problem (see C-2.23):
DFA information flows against the control flow
Out( B ) is determined by $\ln (\mathrm{h})$ of the successor blocks

- union problem:
problem description: „there is a path";
meet operator is $\Theta=\cup$
solution: minimal sets that solve the equations
intersect problem:
problem description: „for all paths"
meet operator is $\Theta=?$
solution: maximal sets that solve the equations
- optimization information: sets of certain statements, of variables, of expressions

Further classes of DFA problems over general lattices instead of sets are not considered here.

Example Reaching Definitions

Description of DFA-Problem
Gen

| $B_{1}$ | $d_{1}, d_{2}, d_{3}$ | $d_{4}, d_{5}, d_{6}, d_{7}, d_{8}$ |
| :--- | :--- | :--- |
| $B_{2}$ | $d_{4}$ | $d_{2}, d_{6}$ |
| $B_{3}$ | $d_{5}$ | $d_{3}, d_{7}$ |
| $B_{4}$ | $d_{6}, d_{7}$ | $d_{2}, d_{3}, d_{4}, d_{5}$ |
| $B_{5}$ | $d_{8}$ | $d_{1}$ |

## Iterative Solution of Data-Flow Equations

Input: the CFG; the sets $\operatorname{Gen}(\mathrm{B})$ and $\operatorname{Kill}(\mathrm{B})$ for each basic block B Output: the sets $\operatorname{In}(B)$ and $\operatorname{Out}(B)$

> Initialization Union: empty sets for all B do begin In (B) $:=\varnothing$; Out (B) $:=$ Gen (B) end; Intersect: full sets for all B do begin In (B) $:=\mathrm{U} ;$ Out (B) $:=$ Gen (B) $\cup$ end; $\quad \begin{aligned} & \text { (U Kill (B) ) }\end{aligned}$

Complexity: $\mathrm{O}\left(\mathrm{n}^{3}\right)$ with n number of basic blocks
$O\left(n^{2}\right)$ if $|\operatorname{pred}(B)| \leq k \ll n$ for all $B$

## Backward Problems

System of Equations for backward problems
propagate information against control-flow edges:

2 equations for each basic block:

## Example Live variables:

1. Description: Is variable $\mathbf{v}$ alive at a given point $p$ in the program, i. e. is there a path from p to the exit where v is used but not defined before the use?
2. backward problem
3. optimization information: sets of variables
4. meet operator: $\Theta=\cup$ union

5. Gen $(B)$ : variables that are used in $B$, but not defined before they are used there.
6. Kill (B): variables that are defined in $B$, but not used before they are defined there.

## Algebraic Foundation of DFA

DFA performs computations on a lattice (dt. Verband) of values,
e. g. bit-vectors representing finite sets. It guarantees termination of computation and well-defined solutions. see [Muchnick, pp 223-228]

A lattice $L$ is a set of values with two operations: $\cap$ meet and $\cup$ join
Required properties:

1. closure: $\quad x, y \in L$ implies $x \cap y \in L, x \cup y \in L$
2. commutativity: $x \cap y=y \cap x$ and $x \cup y=y \cup x$
3. associativity: $(x \cap y) \cap z=x \cap(y \cap z)$ and $(x \cup y) \cup z=x \cup(y \cup z)$
4. absorption: $\quad x \cap(x \cup y)=x=x \cup(x \cap y)$
5. unique elements bottom $\perp$, top $T$ :

$$
\mathrm{x} \cap \perp=\perp \text { and } \mathrm{x} \cup \mathrm{~T}=\mathrm{T}
$$

In most DFA problems only a semilattice is used with $L, \cap, \perp$ or $L, \cup, T$

$$
\begin{array}{lll}
\text { Partial order } & \text { defined by meet, } & \text { defined by join: } \\
& x \subseteq y: x \cap y=x & x \supseteq y: x \cup y=x \\
& \text { (transitive, antisymmetric, reflexive) }
\end{array}
$$

## Monotone Functions Over Lattices

The effects of program constructs on DFA information are described by functions over a suitable lattice,
e. g. the function for basic block $B_{3}$ on C-2.22:

$$
f_{3}\left(<x_{1} x_{2} x_{3} x_{4} x_{5} x_{6} x_{7} x_{8}>\right)=\left\langle x_{1} x_{2} 0 x_{4} 1 x_{6} 0 x_{8}>\in B V^{8}\right.
$$

## Gen-Kill pair encoded as function

$f: L \rightarrow L$ is a monotone function over the lattice $L$ if

$$
\forall x, y \in L: x \subseteq y \Rightarrow f(x) \subseteq f(y)
$$

Finite height of the lattice and monotonicity of the functions guarantee termination of the algorithms.

Fixed points $z$ of the function $f$, with $f(z)=z$, is a solution of the set of DFA equations
MOP: Meet over all paths solution is desired, i. e. the „best" with respect to L
MFP: Maximum fixed point is computed by algorithms, if functions are monotone
If the functions $f$ are additionally distributive, then MFP = MOP.
$f: L \rightarrow L$ is a distributive function over the lattice $L$ if
$\forall x, y \in L: f(x \cap y)=f(x) \cap f(y)$

## Some DFA Lattices

| Bool | T = true |
| :---: | :---: |
| $n=$ and |  |
| $\cup=$ or | $\perp$ = false |

Variable usage
\{defined, used\}
\{defined $\}$

5
Range Lattice: $[\mathrm{lo}, \mathrm{hi}] \in(\mathrm{Z} \cup\{-\infty, \infty\})^{2}$
$\perp=$ [ ] empty range, $T=[-\infty, \infty]$,
$x \subseteq y: x$ is contained in $y$
$\cap:[11, \mathrm{~h} 1] \cap[12, \mathrm{~h} 2]=\mathrm{x}$ let $\mathrm{I}=\max (11,12)$,
$\mathrm{h}=\min (\mathrm{h} 1, \mathrm{~h} 2)$,
$x=$ if $h<1$ then $\perp$ else $[l, h]$
$\cup:[11, \mathrm{~h} 1] \cup[12, \mathrm{~h} 2]=$ $[\min (11, \mid 2), \max (h 1, h 2)]$

4


4 ICP Integer Constant Propagation Lattice
false $\cdots \cdots$
$\mathrm{n} \cap \perp=\perp \quad \mathrm{n} \cap \mathrm{n}=\mathrm{n} \quad \mathrm{n} \cap \mathrm{m}=\perp$ if $\mathrm{n} \neq \mathrm{m}$ $n \cup T=T \quad n \cup n=n \quad n \cup m=T \quad$ if $n \neq m$


## Variants of DFA Algorithms

## Heuristic improvement:

Goal: propagate changes in the In and Out sets as fast as possible.
Technique: visit CFG nodes in topological order in the outer for-loop \{*\}.
Then the number of iterations of the outer repeat-loop is only determined. by back edges in the CFG

## Algorithm for backward problems:

Exchange In and Out sets symmetrically in the algorithm of C-2.22b
The nodes should be visited in topological order as if the directions of edges were flipped.

## Hierarchical algorithms, interval analysis

Regions of the CFG are considered nodes of a CFG on a higher level. That abstraction is recursively applied until a single root node is reached The Gen, Kill sets are combined in upward direction
the In , Out sets are refined downward

## Program Analysis: Call Graph (context-insensitive)

## Nodes: defined functions

Arc $g->h$ : function $g$ contains a call $h()$,
i. e. a call $g()$ may cause the execution of a call $h()$
void a () \{...b()...c()...f()...\}
void b () \{...d()...c()...\}
void c() \{...e()...\}
void $d() \quad\{\ldots\}$
void e() \{...v++; ...b()...\}

void $f() \quad\{. . . d() . .$.
Analysis of structure
b, c, e are recursive,
a, d, f are non-recursive

Propagation of properties:
assume a call e() may modify a global variable v
then calls $a(), b(), c()$ may indirectly cause modification of $v$
v = f(); cnt = 0; while(...)\{...b(); cnt += v; \}

## Program Analysis: Call Graph (context-sensitive)

## Nodes: defined functions and calls (bipartite)

Arc $\mathrm{g}->\mathrm{h}$ : function g contains a call h() , i.e a call g() may cause the execution of a call h() or call g() leads to function g
void a () \{...b()...c()...f()...\}
void b () \{...d()...c()...\}
void $c() \quad\{. . . e() . .$.
void $d() \quad\{. .$.
void e() \{...v++; ...b()...\}
void f() \{...d()...\}


Calls of the same function in different contexts are distinguished by different nodes, e.g. the call of $c$ in $a$ and in $b$

Analysis can be more precise in that aspect.

## Calls Using Function Variables

## Contents of function variables is assigned at run-time.

Static analysis does not know (precisely) which function is called.
Call graph has to assume that any function may be called

$$
\begin{aligned}
& \text { void a() } \\
& \qquad\{\ldots(* h)(0.3,27) \ldots\}
\end{aligned}
$$

Analysis for a better approximation of potential callees:
only those functions which

1. fit to the type of $h$
2. are assigned somewhere in the program
3. can be derived from the reaching definitions at the call


$$
\begin{aligned}
& \text { void } m \text { (int } j)\{\ldots\} \\
& \text { void } g(f l o a t ~ \\
& x, \text { int i) }\{\ldots\} \\
& \ldots k=m ; \ldots f(g) ; \ldots \\
& \text { void a() } \\
& \quad\{\text { void (*h) (float,int) }=g ; \\
& \quad \ldots \\
& \quad \text { if(...) } h=s ; \\
& \quad \ldots(* h)(0.3,27) \ldots
\end{aligned}
$$

## Analysis of Object-Oriented Programs

Aspects specific for object-oriented analysis:

1. hierarchy of classes and interfaces specifies a complex system of subtypes
2. hierarchy of classes and interfaces specifies inheritance and overriding relation for methods
3. dynamic method binding
for method calls v.m ( . . ) the callee is determined at run-time good object-oriented style relies on that feature
4. many small methods are typical object-oriented style
5. library use and reuse of modules
complete program contains many unused classes and methods

Static predictions for dynamically bound method calls are essential for most analyses


## Call Graphs Constructed by Class Hierarchy Analysis (CHA)

The call graph is reduced to a set of reachable methods using the
class hierarchy and the static type of the receiver expression in the call:
If a method $F-p$ is reachable and
if it contains a dynamically bound call v.m(...) and
$T$ is the static type of $v$,
then every method $m$ that is inherited by $\mathbf{T}$ or by a subtype of $\mathbf{T}$
is also reachable, and arcs go from F-p to them.


Call graph for F-p containing v.m(...) static type: F v;
A-m

## Results of Analysis of Dynamically Bound Calls



| analysis module | purpose | category |
| :---: | :---: | :---: |
| ClassMemberVisibility | examines visibility levels of declarations | visualization |
| MethodSizeStatistics | examines length of method implementations in bytecode operations and frequency of different bytecode operations |  |
| ExternalEntities | histogram of references to program entities that reside outside a group of classes |  |
| InheritanceBoundary | histogram of lowest superclass outside a group of classes |  |
| SimplesetterGetter | recognizes simple access methods with bytecode patterns |  |
| Methodinspector | decomposes the raw bytecode array of a method implementation into a list of instruction objects | auxiliary analysis |
| ControlFlow | builds a control flow graph for method implementations | fundamental analyses |
| Dominator | constructs the dominator tree for a control flow graph |  |
| Loop | uses the dominator tree to augment the control flow graph with loop and loop nesting information |  |
| InstrDefuse | models operand accesses for each bytecode instruction |  |
| LocalDefuse | builds intraprocedural def/ use chains |  |
| LifeSpan | analyzes lifeness of local variables and stack locations |  |
| DefuseTypeInfo | infers type information for operand accesses | analysis of incomplete programs |
| Hierarchy | class hierarchy analysis based on a horizontal slice of the hierarchy |  |
| PreciseCallGraph | builds call graph based on inferred type information, copes with incomplete class hierarchy |  |
| ParamEscape | transitively traces propagation of actual parameters in a method call (escape $=$ leaves analyzed library) |  |
| ReadWriteFields | transitive liveness and access analysis for instance fields accessed by a method call |  |

Table 0-1. Analysis plug-ins in our framework
[ Michael Thies: Combining Static Analysis of Java Libraries with Dynamic Optimization, Dissertation, Shaker Verlag, April 2001]

