Data-Flow Analysis for Hot-Spot Program Optimization

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x := a+b

n := c+b ○

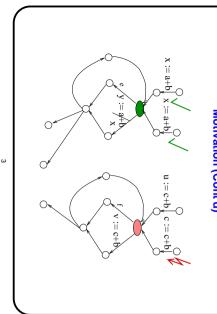
c :: c+b

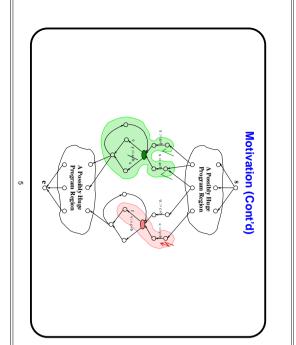
Motivation

Technische Universität Wien



Motivation (Cont'd)



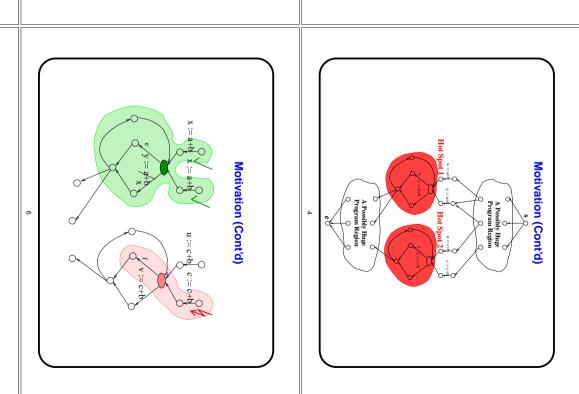


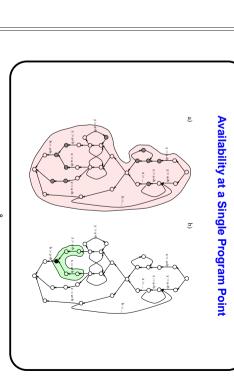
Why Standard Data-Flow Analysis Fails

Availability of terms..

$$\mathsf{AVAIL}(n) \quad = \quad \left\{ \begin{array}{ll} false & \text{if } n = \mathbf{s} \\ \prod\limits_{m \, \in \, pred(n)} \mathbb{I}\left(m,n\right) \mathbb{I}\left(\mathsf{AVAIL}(n)\right) & \text{otherwise} \end{array} \right.$$

$$[\![\ (m,n)\]\!](b) = (\mathsf{COMP}_{(m,n)} + b) * \mathsf{TRANSP}_{(m,n)}$$





Outline of the Talk

Standard vs. Reverse Data-Flow Analysis..

- Background
- Essentials
- The Connecting Link
- The Clou: Why does it work?
- Applications
- Conclusions

9

Agrawal (2000) Horwitz, Reps, Sagiv (1994+) Duesterwald, Gupta, Soffa (1995+)

Demand-Driven Data-Flow Analysis..

Background

Knoop (Euro-Par 1999, KPS 2007)

10

Reverse Data-Flow Analysis: The Basics

(Standard) Data-Flow Analysis...

- Data-Flow Lattice $\hat{\mathcal{C}} = (\mathcal{C},\sqcap,\sqcup,\sqsubseteq,\perp,\top)$
- ullet Data-Flow Functional $[\![\]\!]:E
 ightarrow (\mathcal{C}
 ightarrow \mathcal{C})$

Reverse Data-Flow Analysis...

- Reverse Data-Flow Functional (Hughes, Launchbury 1992+)
- $[\![\]\!]_R:E \mathop{\rightarrow} (\mathcal{C} \mathop{\rightarrow} \mathcal{C})$ defined by

$$\forall e \in E \ \forall \ c \in \mathcal{C}. \ \llbracket \ e \ \rrbracket_R(c) =_{d\!f} \bigcap \{ \ c' \mid \llbracket \ e \ \rrbracket(c') \ \sqsupseteq \ c \}$$

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semantics for availability o

Availability of Terms

- Abstract semantics for availability of terms
- 1. Data-Flow Lattice:

 $(\mathcal{C},\sqcap,\sqcup,\sqsubseteq,\perp,\top){=}_{\mathit{df}}\left(\mathcal{B},\,\wedge\,,\,\vee\,,\leq,\mathit{false},\mathit{true}\right)$

2. Data-Flow Functional: $[\![\]\!]_{av}: E
ightarrow (\mathcal{B}
ightarrow \mathcal{B})$ defined by

$$\forall\,e\in E.\, [\![\,e\,]\!]_{av} =_{df} \left\{ \begin{array}{ll} Cst_{true} & \text{if } \mathsf{Comp}_{\,e} \wedge \mathsf{Transp}_{\,e} \\\\ \mathit{Id}_{B} & \text{if } \neg\mathsf{Comp}_{\,e} \wedge \mathsf{Transp}_{\,e} \\\\ Cst_{false} & \text{otherwise} \end{array} \right.$$

12

On the Relationship of $[\![\]\!]$ and $[\![\]\!]_R$

Lemma

- 1. $[\![e]\!]_R$ is well-defined and monotonic.
- 2. $[\![e\,]\!]_R$ is additive, if $[\![e\,]\!]$ is distributive

13

Monotonicity, Distributivity, and Additivity

...of data-flow functions.

Definition [Monotonicity, Distributivity, Additivity] Let $\hat{\mathcal{C}} = (\mathcal{C}, \sqcap, \sqcup, \sqsubseteq, \bot, \top)$ be a complete lattice and $f: \mathcal{C} \to \mathcal{C}$ a

function on ${\mathcal C}.$ Then: f is

1. monotonic iff $\forall c, c' \in \mathcal{C}. \ c \sqsubseteq c' \Rightarrow f(c) \sqsubseteq f(c')$

(Preserving the order of elements)

- 2. distributive iff \forall $C' \subseteq \mathcal{C}$. $f(\Box C') = \Box \{f(c) \mid c \in C'\}$
- (Preserving greatest lower bounds) 3. additive iff $\forall C' \subseteq \mathcal{C}.\ f(\Box C') = \Box \ \{f(c) \ | \ c \in C'\}$

(Preserving least upper bounds)

14

Often useful

...the following equivalent characterization of monotonicity:

Lemma

Let $\hat{\mathcal{C}}=(\mathcal{C},\sqcap,\sqcup,\sqsubseteq,\perp,\top)$ be a complete lattice and $f:\mathcal{C}\to\mathcal{C}$ a function on \mathcal{C} . Then:

 $f \text{ is monotonic} \Longleftrightarrow \forall \, C' \subseteq \mathcal{C}. \, \, f(\square C') \sqsubseteq \, \square \, \{ f(c) \, | \, c \in C' \}$ $(\Longleftrightarrow \forall \, C' \subseteq \mathcal{C}. \, \, f(\square C') \supseteq \, \square \, \{ f(c) \, | \, c \in C' \})$

On the Relationship of $[\![\]\!]$ and $[\![\]\!]_R$ (Cont'd)

Lemma

- 1. $[\![e]\!]_R \circ [\![e]\!] \sqsubseteq \mathit{Id}_{\mathcal{C}}$, if $[\![e]\!]$ is monotonic.
- 2. $\llbracket \, e \, \rrbracket \circ \llbracket \, e \, \rrbracket_R \sqsupseteq \mathit{Id}_{\mathcal{C}},$ if $\llbracket \, e \, \rrbracket$ is distributive.

In terms of the theory of "abstract interpretation":

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16

Reverse DFA: The $R ext{-}MinFP ext{-}$ Approach

The R-MinFP-Equation System:

$$\textit{regInf}\left(n\right) = \left\{ \begin{array}{l} c_q & \text{if } n = \mathbf{q} \\ \bigsqcup \big\{ \left[\!\left[\left(n,m\right) \right]\!\right]_R \big(\textit{regInf}\left(m\right) \big) \, \big| \, m \in succ(n) \, \big\} \\ & \text{otherwise} \end{array} \right.$$

The R-MinFP-Solution:

$$orall \, c_q \in \mathcal{C} \, orall \, n \in N. \, R\text{-}MinFP_{c_q}(n) =_{df} ext{regInf} \, ^*_{c_q}(n)$$

where $\textit{regInf}_{c_q}^r$ denotes the least solution of the R-MinFP-equation system wrt $c_q \in \mathcal{C}$.

17

Standard DFA: The MaxFP -Approach

The ${\it MaxFP} ext{-}{\it Equation System:}$

$$\mathit{inf}(n) = \left\{ egin{array}{ll} c_{\mathbf{s}} & \mathit{if} \ n = \mathbf{s} \\ & \prod \left\{ \llbracket \ (m,n) \ \rrbracket (\mathit{inf}(m)) \ | \ m \in \mathit{pred}(n) \right\} \\ & & \mathit{otherwise} \end{array} \right.$$

The $\mathit{MaxFP} ext{-Solution}$:

$$orall \, c_{\mathbf{s}} \in \mathcal{C} \, orall \, n \in N. \, \mathit{MaxFP}_{(\llbracket \ \rrbracket, c_{\mathbf{s}})}(n) =_{df} \mathit{inf}^{\, *}_{c_{\mathbf{s}}}(n)$$

where $\inf_{c_s}^*$ denotes the greatest solution of the $Max\!FP$ -equation system wrt $c_s\in\mathcal{C}.$

18

The Connecting Link

Link Theorem

For distributive data-flow functionals $[\![\]\!], q\in N,$ and $c_{\mathbf{s}}, c_q\in \mathcal{C},$ we have:

$$R\text{-}MinFP_{c_q}(\mathbf{s}) \sqsubseteq c_\mathbf{s} \iff MaxFP_{c_\mathbf{s}}(q) \sqsupseteq c_q$$

19

Continuing the Analogy

...of Standard and Reverse Data-Flow Analysis regarding

Soundness & Completeness (in terms of program verification) /
 Safety & Coincidence (Precision) (in terms of data-flow analysis)

20

Essential

...the extensibility of data-flow functionals to paths

$$\llbracket \, p \, \rrbracket \! = _{df} \left\{ \begin{array}{ll} \mathit{Id}_{\mathcal{C}} & \text{if } q \! < \! 1 \\ \\ \llbracket \, \left\langle e_2, \ldots, e_q \right\rangle \, \rrbracket \circ \llbracket \, e_1 \, \rrbracket & \text{otherwise} \end{array} \right.$$

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Standard DFA: Main Results

Theorem [Soundness / Safety]

$$\forall c_{\mathbf{s}} \in \mathcal{C} \ \forall n \in N. \ MaxFP_{c_{\mathbf{s}}}(n) \sqsubseteq MOP_{c_{\mathbf{s}}}(n)$$

if the data-flow functional $[\hspace{.1em}]$ is monotonic.

Theorem [Completeness / Coincidence (Precision)]

$$\forall \, c_{\mathbf{s}} \in \mathcal{C} \, \forall \, n \in N. \, \mathit{MaxFP}_{c_{\mathbf{s}}}(n) = \mathit{MOP}_{c_{\mathbf{s}}}(n)$$

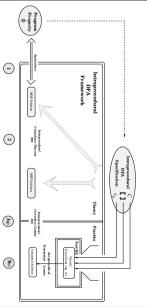
if the data-flow functional $[\![\]\!]$ is distributive

The MOP-Approach

$$\forall \, c_{\mathbf{s}} \in \mathcal{C} \, \forall \, n \in N. \, MOP_{c_{\mathbf{s}}}(n) = \bigcap \, \big\{ \left[\! \left[p \, \right] \! \right] \! \left(c_{\mathbf{s}} \right) \, \middle| \, p \in \mathbf{P}[\mathbf{s}, n] \, \big\}$$

22





24

Of course...

Reverse data-flow functionals can be extended to paths, too:

$$\llbracket p \rrbracket_R =_{df} \left\{ \begin{array}{l} \operatorname{id}_C & \text{if } q < 1 \\ & \llbracket \left\langle e_1, \dots, e_{q-1} \right\rangle \rrbracket_R \circ \llbracket \left. e_q \right. \rrbracket_R & \text{otherwise} \end{array} \right.$$

 $\forall c_q \in \mathcal{C} \,\forall \, n \in N. \, R\text{-}JOP_{c_q}(n) = df \, \sqcup \, \{ \, \llbracket \, p \, \rrbracket_R(c_q) \, | \, p \in \mathbf{P}[n, \mathbf{q}] \}$

The $R\text{-}J\!O\!P\text{-}$ Solution:

The $R ext{-}J\!O\!P ext{-}$ Approach

25

26

Reverse DFA: Main Results

Theorem [Soundness / Reverse Safety]

$$\forall c_q \in \mathcal{C} \ \forall n \in N. \ R-MinFP_{c_q}(n) \supseteq R-JOP_{c_q}(n)$$

Theorem [Completeness / Reverse Coincidence (Precision)]

$$\forall c_q \in \mathcal{C} \ \forall n \in N. \ R-MinFP_{c_q}(n) = R-JOP_{c_q}(n)$$

if $[\hspace{-1.5pt}[\hspace{-1.5pt}]$ is distributive.

27

Putting it together...

MaxFP Coincidence || Theorem Data-flow Analysis

 $\llbracket \mathbf{e} \rrbracket \in \llbracket \, C \xrightarrow{\text{distributive}} C \rrbracket$

Program Verification

Strongest Postcondition View

 $\{p\} \mathrel{\pi} \{?\}$

 $\llbracket e \rrbracket_R(c) =_{\mathrm{df}} \, \bigcap \, \{c^i | \, \llbracket \, e \, \rrbracket(c^i) \, \trianglerighteq \, c\}$ $\{?\}\pi\{q\}$

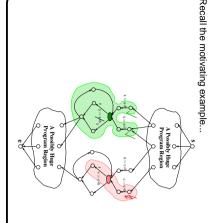
Weakest Precondition View

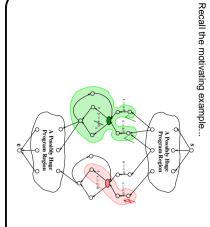
 $\overrightarrow{\text{R-JOP}}_{c_{\mathbf{q}}}(s) \sqsubseteq c_{\mathbf{s}}$

Link Theorem

R-MinFP

Are We Done?





29

Changing the Perspective

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Mastering the Road to Success

pre-conditions" on "strongest post-conditions". ...requires more. It requires us to conclude from "weakest

verification problem. ...essentially, this means to replace the analysis problem by a

30

Changing the Perspective: The Standard Taxonomy

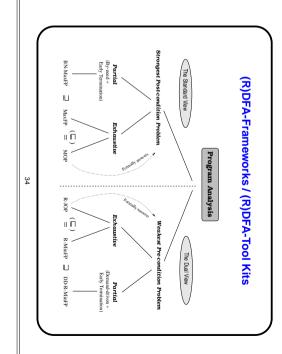
Conventional Classification of DFA Techniques

Demand-Driven \mathbf{DFA}

Exhaustive DFA

32

Changing the Perspective: Conclusions Derived The specification problem: The implementation problem: {**p**} π {?} {p} π {q} ? ... the domain of exhaustive DFA ... the domain of den 33 '-driven DFA



Gen/Kill-Problems

problem boils down to a WPC-verification problem. ...allow us to master the road to success: The SPC-analysis

This is important because...

Redundant Expression/Assignment Elimination

2. Data-flow functional: $[\![]\!]_{av}: E \to (\mathcal{B}_X \to \mathcal{B}_X)$ defined by

 Cst_{true}^{X} if $\mathit{Comp}_{e} \wedge \mathit{Transp}_{e}$

with $\bot = false \ \Box \ true \ \Box \ failure = \top$

 $(C, \sqcap, \sqcup, \sqsubseteq, \bot, \top) =_{df} (\mathcal{B}_X, \wedge, \vee, \leq, false, failure)$

 $\forall e \in E. \llbracket e \rrbracket_{av} =_{df}$

 $egin{pmatrix} \mathsf{Id}_{\mathcal{B}_X} \ Cst_{false}^X \end{pmatrix}$

otherwise

if $\neg \textit{Comp}_e \wedge \textit{Transp}_e$

36

Data-flow lattice:

Abstract semantics for availability

Concluding the Example: Availability

- Dead-Code Elimination
- Strength Reduction

are based on Gen-Kill-problems.

Reverse Availability

Reverse abstract semantics for availability

- Data-flow lattice:
- $(C, \sqcap, \sqcup, \sqsubseteq, \bot, \top) =_{df} (\mathcal{B}_X, \land, \lor, \leq, false, failure)$
- 2. Reverse data-flow functional: $[\![\]\!]_{av_R}: E \to (\mathcal{B}_X \to \mathcal{B}_X)$ defined by

$$\forall\,e\in E.\,\llbracket\,e\,\rrbracket_{av_R} =_{df}\,\left\{\begin{array}{ll} R\text{-}Cst^X_{true} & \text{if } \llbracket\,e\,\rrbracket_{av} = Cst^X_{true} \\\\ R\text{-}log_X & \text{if } \llbracket\,e\,\rrbracket_{av} = Idg_X \\\\ R\text{-}Cst^X_{false} & \text{if } \llbracket\,e\,\rrbracket_{av} = Cst^X_{false} \end{array}\right.$$

37

Summing Up / Extensions

In this talk...

• The intraprocedural basic setting of (R)DFA (Knoop, KPS 2007)

Extensions are possible...

- Interprocedural setting (Knoop, CC 1992, LNCS 1428 (1998))
- Parallel setting (Knoop, Euro-Par 1999)

Supporting Functions

$$\forall \, b \in \mathcal{B}_X. \; R\text{-} \, Cst^X_{false}(b) =_{d\!f} \left\{ \begin{array}{ll} false & \text{if } b = false \\ failure & \text{otherwise} \end{array} \right.$$

R-Id
$$_{\mathcal{B}_X}$$
 $=_{df}$ Id $_{\mathcal{B}_X}$

38

(R)DFA-Frameworks / (R)DFA-Tool Kits (Cont'd)

...the general pattern:

