

Relating Complexity and Precision in Control Flow Analysis

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Introduction

We investigate the *precision* of static, compile-time analysis, and the necessary analytic *tradeoff* with the computational *resources* that go into the analysis.

Outline

- What is Control Flow Analysis?
- What is k CFA?
- What is an exact analysis?
- Symmetric boolean logic gates
- **P**TIME-completeness of 0CFA
- **NP**-hardness of k CFA
- **EXPTIME**-completeness of n CFA
- A sketch of what's left in the paper

CFA Primer

1. For every application, which abstractions can be applied?
2. For every abstraction, to which arguments can it be applied?

Preliminaries: the Language

The language:

$e ::= t^l$ expressions (labeled terms)

$t ::= x \mid (e e) \mid (\lambda x.e)$ terms (unlabeled expressions)

For example:

$((\lambda f.((f^1 f^2)^3 (\lambda y.y^4)^5)^6)^7 (\lambda x.x^8)^9)^{10}$

Preliminaries: Contours

Contours are strings of @-labels of length $\leq k$.

Contour environments map variable names to contours.

$$\delta \in \Delta = \mathbf{Lab}^{\leq k}$$

$$ce \in \mathbf{CEnv} = \mathbf{Var} \rightarrow \Delta$$

Contours describe the context in which a term evaluates.

$(e_0([\lambda x.e_1])e_2)^{\ell_1}{}^{\ell_2} \Rightarrow$ “ e_1 evaluates in contour $\ell_2\ell_1$.”

Contour environments describe the context in which a variable was bound.

$(e_0([\lambda x.e_1])e_2)^{\ell_1}{}^{\ell_2} \Rightarrow x \mapsto \ell_2\ell_1 \Rightarrow$ “ x bound in contour $\ell_2\ell_1$.”

Preliminaries: the Analysis

An analysis is a table \hat{C} that maps a label and contour to a set of abstract closures.

$$\hat{C} : \mathbf{Lab} \times \Delta \rightarrow \mathcal{P}(\mathbf{Term} \times \mathbf{CEnv})$$

$$\hat{C}(\ell, \delta) = \{ \langle (\lambda y. \dots), ce \rangle, \langle (\lambda z. \dots), ce' \rangle \}$$

In contour δ , the term labeled ℓ evaluates to either the closure $\langle (\lambda y. \dots), ce \rangle$, or $\langle (\lambda z. \dots), ce' \rangle$.

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$$\hat{C}(\ell, \delta) = \{ \langle \lambda y, ce \rangle, \langle \lambda z, ce' \rangle \} \quad \text{shorthand}$$

In contour δ , the term labeled ℓ evaluates to either the closure $\langle \lambda y, ce \rangle$, or $\langle \lambda z, ce' \rangle$.

Preliminaries: the Analysis

An analysis is a table \hat{C} that maps a label and contour to a set of abstract closures.

$$\hat{C} : \mathbf{Lab} \times \Delta \rightarrow \mathcal{P}(\mathbf{Term} \times \mathbf{CEnv})$$

$$\hat{C}(x, \delta) = \{ \langle \lambda y, ce \rangle, \langle \lambda z, ce' \rangle \} \quad \text{overloading}$$

In contour δ , x is bound to $\langle \lambda y, ce \rangle$, or $\langle \lambda z, ce' \rangle$...

... λx is applied to either $\langle \lambda y, ce \rangle$, or $\langle \lambda z, ce' \rangle$.

Decision problem

Control Flow Problem (*k*CFA): Given a closure and a label ℓ and contour δ , does that closure flow into the program point labeled ℓ under δ ?

$$\langle \lambda x, ce \rangle \in \widehat{C}(\ell, \delta)?$$

Acceptability

The analysis is acceptable for e , closed by ce , in contour δ :

$$\hat{C} \models_{\delta}^{ce} e$$

At the top-level (for a closed program), ce and δ are empty:

$$\hat{C} \models_{\epsilon}^{[]} e$$

What do δ and ce mean when non-empty?

Polyvariance

During reduction, a function may copy its argument:

$$((\lambda f. \dots (f e_1)^{\ell_1} \dots (f e_2)^{\ell_2} \dots)) (\lambda x. e)$$

Contours and environments let us talk about each copy of e :

$$\widehat{C} \Vdash_{\ell_1}^{x \mapsto \ell_1} e \quad \widehat{C} \Vdash_{\ell_2}^{x \mapsto \ell_2} e$$

The analysis is *polyvariant*. Contours and environments describe which *instance* (copy) of a term we are talking about.

Acceptability

The analysis is acceptable for e , closed by ce , in contour δ :

$$\hat{C} \Vdash_{\delta}^{ce} e$$

The analysis is acceptable for the copy of e that occurs in context described by δ , closed by the environment ce which says what copy of a term each variable is bound to.

Finally, let's look at what is acceptable...

Acceptability

$$\widehat{C} \models_{\delta}^{ce} x^{\ell} \quad \text{iff} \quad \widehat{C}(x, ce(x)) \subseteq \widehat{C}(\ell, \delta)$$

$$\widehat{C} \models_{\delta}^{ce} (\lambda x.e)^{\ell} \quad \text{iff} \quad \langle (\lambda x.e), ce_0 \rangle \in \widehat{C}(\ell, \delta)$$

where $ce_0 = ce|_{\mathbf{fv}(\lambda x.e)}$

$$\widehat{C} \models_{\delta}^{ce} (t_1^{\ell_1} t_2^{\ell_2})^{\ell} \quad \text{iff} \quad \widehat{C} \models_{\delta}^{ce} t_1^{\ell_1} \wedge \widehat{C} \models_{\delta}^{ce} t_2^{\ell_2} \wedge$$
$$\forall \langle (\lambda x.t_0^{\ell_0}), ce_0 \rangle \in \widehat{C}(\ell_1, \delta) :$$
$$\widehat{C} \models_{\delta_0}^{ce'_0} t_0^{\ell_0} \wedge$$
$$\widehat{C}(\ell_2, \delta) \subseteq \widehat{C}(x, \delta_0) \wedge$$
$$\widehat{C}(\ell_0, \delta_0) \subseteq \widehat{C}(\ell, \delta)$$

where $\delta_0 = \lceil \delta, \ell \rceil_k$ and $ce'_0 = ce_0[x \mapsto \delta_0]$

Acceptability

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where $\delta_0 = \lceil \delta, \ell \rceil_k$ and $ce'_0 = ce_0[x \mapsto \delta_0]$



Mr. Yuck: Ingesting formalisms

may cause *rigor mortis*

Exact analysis

An analysis is a table \hat{C} that maps label-contour pairs to sets of abstract closures.

$$\hat{C}(\ell, \delta) = \{ \langle \lambda y, ce \rangle, \langle \lambda z, ce' \rangle \}$$

In contour δ , the term labeled ℓ evaluates to either the closure $\langle \lambda y, ce \rangle$, or $\langle \lambda z, ce' \rangle$.

Exact analysis

An *exact* analysis is a table \hat{C} that maps label-contour pairs to *an* abstract closure.

$$\hat{C}(\ell, \delta) = \{ \langle \lambda y, ce \rangle \}$$

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An *exact* analysis is a table \hat{C} that maps label-contour pairs to *an* abstract closure.

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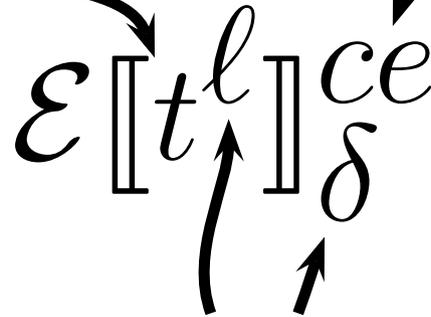
In contour δ , the term labeled ℓ evaluates to either the closure $\langle \lambda y, ce \rangle$.

Pick any nose you want! (You only have one)



Evaluator

Evaluate the term t , which is closed under environment ce .



Write the result into location (l, δ) of the table.

Evaluator (exact)

$$\begin{aligned}\mathcal{E}[[x^\ell]_\delta^{ce}] &= \widehat{C}(\ell, \delta) \leftarrow \widehat{C}(x, ce(x)) \\ \mathcal{E}[(\lambda x.e_0)^\ell]_\delta^{ce} &= \widehat{C}(\ell, \delta) \leftarrow \langle \lambda x.e_0, ce_0 \rangle \\ &\quad \text{where } ce_0 = ce|_{\mathbf{fv}(\lambda x.e_0)} \\ \mathcal{E}[(t_1^{\ell_1} t_2^{\ell_2})^\ell]_\delta^{ce} &= \mathcal{E}[t_1^{\ell_1}]_\delta^{ce}; \mathcal{E}[t_2^{\ell_2}]_\delta^{ce}; \\ &\quad \text{let } \langle \lambda x.t_0^{\ell_0}, ce_0 \rangle = \widehat{C}(\ell_1, \delta) \text{ in} \\ &\quad \widehat{C}(x, \delta, \ell) \leftarrow \widehat{C}(\ell_2, \delta); \\ &\quad \mathcal{E}[t_0^{\ell_0}]_{\delta, \ell}^{ce_0[x \mapsto \delta, \ell]}; \\ &\quad \widehat{C}(\ell, \delta) \leftarrow \widehat{C}(\ell_0, \delta, \ell)\end{aligned}$$

If e has an exact k CFA analysis, then $\mathcal{E}[[e]_\epsilon^{\llbracket \cdot \rrbracket}]$ constructs it.

Evaluator (exact)

$$\mathcal{E}[[x^\ell]_\delta^{ce}] = \widehat{C}(\ell, \delta) \leftarrow \widehat{C}(x, ce(x))$$

$$\mathcal{E}[(\lambda x.e_0)^\ell]_\delta^{ce} = \widehat{C}(\ell, \delta) \leftarrow \langle \lambda x.e_0, ce_0 \rangle$$

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$$\mathcal{E}[(t_1^{\ell_1} t_2^{\ell_2})^\ell]_\delta^{ce} = \mathcal{E}[[t_1^{\ell_1}]_\delta^{ce}; \mathcal{E}[[t_2^{\ell_2}]_\delta^{ce};$$

let $\langle \lambda x.t_0^{\ell_0}, ce_0 \rangle = \widehat{C}(\ell_1, \delta)$ in

$$\widehat{C}(x, \delta, \ell) \leftarrow \widehat{C}(\ell_2, \delta);$$

$$\mathcal{E}[[t_0^{\ell_0}]_{\delta, \ell}^{ce_0[x \mapsto \delta, \ell]}];$$

$$\widehat{C}(\ell, \delta) \leftarrow \widehat{C}(\ell_0, \delta, \ell)$$



Mr. Natural: Exact analysis is normalization.

If e has an exact k CFA analysis, then $\mathcal{E}[[e]_\epsilon^[]]$ constructs it.

Evaluator (inexact)

$$\begin{aligned}
 \mathcal{E}[[x^\ell]_\delta^{ce}] &= \widehat{\mathcal{C}}(\ell, \delta) \leftarrow \widehat{\mathcal{C}}(x, ce(x)) \\
 \mathcal{E}[(\lambda x.e_0)^\ell]_\delta^{ce} &= \widehat{\mathcal{C}}(\ell, \delta) \leftarrow \langle \lambda x.e_0, ce_0 \rangle \\
 &\quad \text{where } ce_0 = ce|_{\mathbf{fv}(\lambda x.e_0)} \\
 \mathcal{E}[(t_1^{\ell_1} t_2^{\ell_2})^\ell]_\delta^{ce} &= \mathcal{E}[[t_1^{\ell_1}]_\delta^{ce}; \mathcal{E}[[t_2^{\ell_2}]_\delta^{ce}; \\
 &\quad \forall \langle \lambda x.t_0^{\ell_0}, ce_0 \rangle \in \widehat{\mathcal{C}}(\ell_1, \delta) : \\
 &\quad \widehat{\mathcal{C}}(x, [\delta, \ell]_k) \leftarrow \widehat{\mathcal{C}}(\ell_2, \delta); \\
 &\quad \mathcal{E}[[t_0^{\ell_0}]_{[\delta, \ell]_k}^{ce_0[x \mapsto [\delta, \ell]_k]}]; \\
 &\quad \widehat{\mathcal{C}}(\ell, \delta) \leftarrow \widehat{\mathcal{C}}(\ell_0, [\delta, \ell]_k)
 \end{aligned}$$

The k CFA analysis of e is constructed by iterating $\mathcal{E}[[e]_\epsilon^{\lceil \cdot \rceil}]$ until $\widehat{\mathcal{C}}$ reaches a fixed point.

Evaluator (inexact, $k = 0$)

$$\begin{aligned}\mathcal{E}[[x^\ell]_\delta^{ce} &= \widehat{\mathcal{C}}(\ell, \delta) \leftarrow \widehat{\mathcal{C}}(x, ce(x)) \\ \mathcal{E}[(\lambda x.e_0)^\ell]_\delta^{ce} &= \widehat{\mathcal{C}}(\ell, \delta) \leftarrow \langle \lambda x.e_0, ce_0 \rangle \\ &\quad \text{where } ce_0 = ce|_{\mathbf{fv}(\lambda x.e_0)} \\ \mathcal{E}[(t_1^{\ell_1} t_2^{\ell_2})^\ell]_\delta^{ce} &= \mathcal{E}[t_1^{\ell_1}]_\delta^{ce}; \mathcal{E}[t_2^{\ell_2}]_\delta^{ce}; \\ &\quad \forall \langle \lambda x.t_0^{\ell_0}, ce_0 \rangle \in \widehat{\mathcal{C}}(\ell_1, \delta) : \\ &\quad \quad \widehat{\mathcal{C}}(x, [\delta, \ell]_0) \leftarrow \widehat{\mathcal{C}}(\ell_2, \delta); \\ &\quad \quad \mathcal{E}[t_0^{\ell_0}]_{[\delta, \ell]_0}^{ce_0[x \mapsto [\delta, \ell]_0]}; \\ &\quad \quad \widehat{\mathcal{C}}(\ell, \delta) \leftarrow \widehat{\mathcal{C}}(\ell_0, [\delta, \ell]_0)\end{aligned}$$

But $[\delta, \ell]_0 = \epsilon$, so all contours are empty and all environments map all variables to ϵ . We can safely discard both.

Evaluator (inexact, $k = 0$)

$$\begin{aligned}\mathcal{E}[[x^\ell]] &= \widehat{C}(\ell) \leftarrow \widehat{C}(x) \\ \mathcal{E}[(\lambda x.e_0)^\ell] &= \widehat{C}(\ell) \leftarrow (\lambda x.e_0) \\ \mathcal{E}[(t_1^{\ell_1} t_2^{\ell_2})^\ell] &= \mathcal{E}[[t_1^{\ell_1}]]; \mathcal{E}[[t_2^{\ell_2}]]; \\ &\quad \forall (\lambda x.t_0^{\ell_0}) \in \widehat{C}(\ell_1) : \\ &\quad \quad \widehat{C}(x) \leftarrow \widehat{C}(\ell_2, \delta); \\ &\quad \quad \mathcal{E}[[t_0^{\ell_0}]]; \\ &\quad \quad \widehat{C}(\ell) \leftarrow \widehat{C}(\ell_0)\end{aligned}$$

But $[\delta, \ell]_0 = \epsilon$, so all contours are empty and all environments map all variables to ϵ . We can safely discard both.

Evaluator (exact, $k = 0$)

$$\begin{aligned}\mathcal{E}[[x^\ell]] &= \widehat{\mathbf{C}}(\ell) \leftarrow \widehat{\mathbf{C}}(x) \\ \mathcal{E}[[\lambda x.e_0]^\ell] &= \widehat{\mathbf{C}}(\ell) \leftarrow (\lambda x.e_0) \\ \mathcal{E}[[t_1^{\ell_1} t_2^{\ell_2}]^\ell] &= \mathcal{E}[[t_1^{\ell_1}]]; \mathcal{E}[[t_2^{\ell_2}]]; \\ &\text{let } (\lambda x.t_0^{\ell_0}) = \widehat{\mathbf{C}}(\ell_1) \text{ in} \\ &\quad \widehat{\mathbf{C}}(x) \leftarrow \widehat{\mathbf{C}}(\ell_2, \delta); \\ &\quad \mathcal{E}[[t_0^{\ell_0}]]; \\ &\quad \widehat{\mathbf{C}}(\ell) \leftarrow \widehat{\mathbf{C}}(\ell_0)\end{aligned}$$

*This is an evaluator, **but for what language?***

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*This is an evaluator, **but for what language?***

The linear λ -calculus.

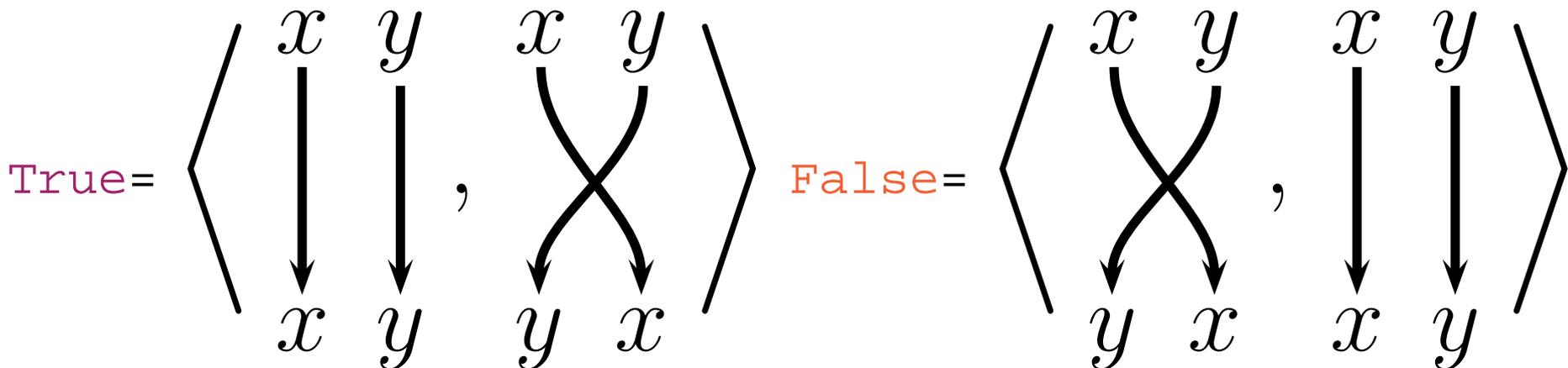
Symmetric logic gates

```
- fun TT(x:'a,y:'a)= (x,y);  
val TT= fn : 'a * 'a -> 'a * 'a  
- fun FF(x:'a,y:'a)= (y,x);  
val FF= fn : 'a * 'a -> 'a * 'a
```

Booleans built out of
constants **TT**, **FF**

```
- val True= (TT: ('a * 'a -> 'a * 'a), FF: ('a * 'a -> 'a * 'a));  
val True = (fn,fn) : ('a * 'a -> 'a * 'a) * ('a * 'a -> 'a * 'a)
```

```
- val False= (FF: ('a * 'a -> 'a * 'a), TT: ('a * 'a -> 'a * 'a));  
val False = (fn,fn) : ('a * 'a -> 'a * 'a) * ('a * 'a -> 'a * 'a)
```



Symmetric garbage is self-annihilating

And $(p, p') (q, q') \equiv (p \wedge q, p' \vee q') \equiv (p \wedge q, \neg(p' \wedge q'))$

- fun And (p,p') (q,q')=

let val ((u,v),(u',v')) = (p (q,FF), p' (TT,q'))

in (u,Compose (Compose (u',v),Compose (v',FF))) end;

val And = fn

: ('a * ('b * 'b -> 'b * 'b) -> 'c * ('d -> 'e))

* (('f * 'f -> 'f * 'f) * 'g -> ('e -> 'h) * ('i * 'i -> 'd))

-> 'a * 'g -> 'c * ('i * 'i -> 'h)

When $p=TT$ (identity),

$(u, v) = (q, FF)$

$(u', v') = (q', TT)$

When $p=FF$ (twist),

$(u, v) = (FF, q)$

$(u', v') = (TT, q')$

So, $\{v, v'\} = \{TT, FF\}$, and

$\text{Compose } (v, \text{Compose } (v', FF)) = TT$

$\text{Compose } (\text{Compose } (u', v), \text{Compose } (v', FF)) = u'$

Symmetric logic gates

```
- fun Copy (p,p')= (p (TT,FF), p' (FF,TT));
```

```
val Copy = fn
```

```
  : (('a * 'a -> 'a * 'a) * ('b * 'b -> 'b * 'b) -> 'c)  
    * (('d * 'd -> 'd * 'd) * ('e * 'e -> 'e * 'e) -> 'f)  
    -> 'c * 'f
```

```
[p= TT]: Copy (p,p') = ((TT,FF), (TT,FF)) [second component reversed]
```

```
[p= FF]: Copy (p,p') = ((FF,TT), (FF,TT)) [first component reversed]
```

```
- fun Not (x,y)= (y,x);
```

```
val Not = fn : 'a * 'b -> 'b * 'a
```

Or is symmetric to And

PTIME Hardness of OCFA

Code the boolean circuit as a linear term. Either (u, u') is bound to (TT, FF) or (FF, TT) (but not both, by linearity).

```
let val (u, u') =  $\phi(\vec{v})$  in
let val ((x, y), (x', y')) = (u (f, g), u' (f', g'))
    ((x a, y b), (x' a', y' b')) end end;
```

We know u is either TT (identity) or FF (twist)... so $(u (f, g))$ is either (f, g) or (g, f) ... therefore (x, y) is bound to either (f, g) or (g, f) ...

so $f = x$ iff $\phi(\vec{v}) = \text{True}$

$(f = y$ iff $\phi(\vec{v}) = \text{False}$)

so f is applied to a iff $\phi(\vec{v}) = \text{True}$.

OCFA is PTIME-hard

PTIME Inclusion of 0CFA

Well known, eg. PPA Nielson et al. (1999):

- 0CFA computes a binary relation over a *fixed structure* (the graph description of a program).
- The computation of the relation is *monotone*: begins empty and is added to incrementally.
- A *fixed point* must be reached by this incremental computation (structure is finite).
- The binary relation can be at most *polynomial in size*, and each increment is *computed in polynomial time*.

0CFA is PTIME-complete

Closures

Because CFA makes approximations, many closures can flow to a single program point and contour. In 1CFA, for example,

$$(\lambda w. w x_1 x_2 \dots x_n)$$

Has n free variables, with 2^n possible associated environments mapping these variables to program points (contours of length 1).

1CFA as SAT solver

$(\lambda f_1.(f_1 \text{ True})(f_1 \text{ False}))$

$(\lambda x_1.$

$(\lambda f_2.(f_2 \text{ True})(f_2 \text{ False}))$

$(\lambda x_2.$

$(\lambda f_3.(f_3 \text{ True})(f_3 \text{ False}))$

$(\lambda x_3.$

\dots

$(\lambda f_n.(f_n \text{ True})(f_n \text{ False}))$

$(\lambda x_n.$

$C[(\lambda v.\phi v)(\lambda w.wx_1x_2 \dots x_n)] \dots])$

1CFA as SAT solver

$$\begin{aligned} &(\lambda f_1.(f_1 \text{ True})(f_1 \text{ False})) \\ &(\lambda x_1. \\ &\quad (\lambda f_2.(f_2 \text{ True})(f_2 \text{ False})) \\ &\quad (\lambda x_2. \\ &\quad\quad (\lambda f_3.(f_3 \text{ True})(f_3 \text{ False})) \\ &\quad\quad (\lambda x_3. \\ &\quad\quad\quad \dots \\ &\quad\quad\quad (\lambda f_n.(f_n \text{ True})(f_n \text{ False})) \\ &\quad\quad\quad (\lambda x_n. \\ &\quad\quad\quad\quad C[(\lambda v.\phi v)(\lambda w.wx_1x_2 \cdots x_n)])) \cdots)) \end{aligned}$$

Approximation allows us to bind each x_i to either of the closed λ -terms for **True** and **False**.

1CFA as SAT solver

$$\begin{aligned} &(\lambda f_1.(f_1 \text{ True})(f_1 \text{ False})) \\ &(\lambda x_1. \\ &\quad (\lambda f_2.(f_2 \text{ True})(f_2 \text{ False})) \\ &\quad (\lambda x_2. \\ &\quad\quad (\lambda f_3.(f_3 \text{ True})(f_3 \text{ False})) \\ &\quad\quad (\lambda x_3. \\ &\quad\quad\quad \dots \\ &\quad\quad\quad (\lambda f_n.(f_n \text{ True})(f_n \text{ False})) \\ &\quad\quad\quad (\lambda x_n. \\ &\quad\quad\quad\quad C[(\lambda v.\phi v)(\lambda w.wx_1x_2 \cdots x_n)] \cdots))))) \end{aligned}$$

Applying a Boolean function necessitates computation of all 2^n bindings to compute the flow out of the application.

1CFA as SAT solver

$$\begin{aligned} &(\lambda f_1.(f_1 \text{ True})(f_1 \text{ False})) \\ &(\lambda x_1. \\ &\quad (\lambda f_2.(f_2 \text{ True})(f_2 \text{ False})) \\ &\quad (\lambda x_2. \\ &\quad\quad (\lambda f_3.(f_3 \text{ True})(f_3 \text{ False})) \\ &\quad\quad (\lambda x_3. \\ &\quad\quad\quad \dots \\ &\quad\quad\quad (\lambda f_n.(f_n \text{ True})(f_n \text{ False})) \\ &\quad\quad\quad (\lambda x_n. \\ &\quad\quad\quad\quad C[(\lambda v.\phi v)(\lambda w.wx_1x_2 \cdots x_n)])) \cdots)) \end{aligned}$$

True flows out of the apply iff the Boolean function is satisfied by some truth valuation.

1CFA as SAT solver

$(\lambda f_1.(f_1 \text{ True})(f_1 \text{ False}))$

$(\lambda x_1.$

$(\lambda f_2.(f_2 \text{ True})(f_2 \text{ False}))$

$(\lambda x_2.$

$(\lambda f_3.(f_3 \text{ True})(f_3 \text{ False}))$

$(\lambda x_3.$

...

$(\lambda f_n.(f_n \text{ True})(f_n \text{ False}))$

$(\lambda x_n.$

$C[(\lambda v.\phi v)(\lambda w.wx_1x_2 \cdots x_n)] \cdots))$



Approximation of closures as non-deterministic computation!

The Widget

$C =$

```
let val (u,u') = [ ] in
```

```
let val ((x,y),(x',y')) = (u (f,g), u' (f',g')) in
```

```
((x a, y b), (x' a', y' b')) end end;
```

In $C[(\lambda v.\phi v)(\lambda w.wx_1x_2 \cdots x_n)]$:

f is applied to a iff ϕ is satisfiable.

1CFA is NP-hard

The Widget

$C =$

```
let val (u,u') = [ ] in
```

```
let val ((x,y),(x',y')) = (u (f,g), u' (f',g')) in  
  ((x a, y b), (x' a', y' b')) end end;
```

In $C[(\lambda v.\phi v)(\lambda w.wx_1x_2 \cdots x_n)]$:

f is applied to a iff ϕ is satisfiable.

1CFA is NP-hard

$(k > 1)$ CFA is just as hard. The construction just needs to be “padded” to undo the added precision of longer contours.

k CFA is NP-hard

Naïve exponential algorithm for k CFA

- The \hat{C} table is finite and has n^{k+1} entries.
- Each entry contains a set of closures.
- The environment maps p free variables to any one of n^k contours.
- There are n possible λx terms and n^{kp} environments, so each entry contains at most n^{1+kp} closures.
- Approximate evaluation is monotonic, and there are at most $n^{1+(k+1)p}$ updates to \hat{C}
- $p \leq n$ so k CFA in **EXPTIME**

k CFA in NP?

$$\mathcal{E}[[x^\ell]_\delta^{ce}] = \widehat{C}(\ell, \delta) \leftarrow \widehat{C}(x, ce(x))$$

$$\mathcal{E}[(\lambda x.e_0)^\ell]_\delta^{ce} = \widehat{C}(\ell, \delta) \leftarrow \langle \lambda x.e_0, ce_0 \rangle$$

where $ce_0 = ce|_{\mathbf{fv}(\lambda x.e_0)}$

$$\mathcal{E}[(t_1^{\ell_1} t_2^{\ell_2})^\ell]_\delta^{ce} = \mathcal{E}[t_1^{\ell_1}]_\delta^{ce}; \mathcal{E}[t_2^{\ell_2}]_\delta^{ce};$$

$$\forall \langle \lambda x.t_0^{\ell_0}, ce_0 \rangle \in \widehat{C}(\ell_1, \delta) :$$

$$\widehat{C}(x, [\delta, \ell]_k) \leftarrow \widehat{C}(\ell_2, \delta);$$

$$\mathcal{E}[t_0^{\ell_0}]_{[\delta, \ell]_k}^{ce_0[x \mapsto [\delta, \ell]_k]};$$

$$\widehat{C}(\ell, \delta) \leftarrow \widehat{C}(\ell_0, [\delta, \ell]_k)$$

Can we guess our way through the computation to answer the k CFA decision problem?

Exact analysis for non-linear terms

0CFA is exact for linear terms...

... When is k CFA exact for non-linear terms?

Suppose ϕ is a linear term coding the transition function of a Turing machine and I is the (linear) initial machine configuration.

$$((\bar{2}\phi)I) \equiv (((\lambda s.(\lambda z.(s^1(s^2 z))))))\phi)I)$$

1CFA analyzes each application of ϕ distinctly in contour 1 and 2, and therefore is *exact*...

... So analysis simulates 2 steps of the TM.

Exact analysis for non-linear terms

0CFA is exact for linear terms...

... When is k CFA exact for non-linear terms?

Scaling up, consider:

$$\begin{aligned} ((\bar{2}(\bar{2}\phi))I) &\equiv (((\lambda s.(\lambda z.(s^3(s^4 z)))))) \\ &\quad (((\lambda s.(\lambda z.(s^1(s^2 z))))\phi))I) \end{aligned}$$

2CFA analyzes each application of ϕ distinctly in contour 31, 32, 41 and 42, and therefore is *exact*...

... So analysis simulates 4 steps of the TM.

Exact analysis for non-linear terms

0CFA is exact for linear terms...

... When is k CFA exact for non-linear terms?

In general, n CFA is *exact* for:

$$((\bar{2}^n \phi) I)$$

... So analysis simulates 2^n steps of the TM.

Exact analysis for non-linear terms

0CFA is exact for linear terms...

... When is k CFA exact for non-linear terms?

In general, n CFA is *exact* for:

$$((\bar{2}^n \phi) I)$$

... So analysis simulates 2^n steps of the TM.

n CFA is EXPTIME-complete

The Doggie Bag

What I want you take home:



- Linearity subverts approximation in static analysis.
- It doesn't matter how big your k is (as long as it's constant), but how you use your closures.
- Either you run the program or you do something stupid.

What else?

Preprint available:

Relating Complexity and Precision in Control Flow Analysis

<http://www.cs.brandeis.edu/~dvanhorn/pubs/vanhorn-mairson-07.pdf>

- A graph formulation of k CFA
- Allows for (direct-style) k CFA of $\lambda + \text{call/cc}$, λ_μ , symmetric λ , (any calculus w/ a CPS semantics).
- Draws connection between k CFA and linear logic.
- 0CFA of simply-typed, η -expanded programs is LOGSPACE-complete.

The End



Citations

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