

Resolving and Exploiting the k -CFA Paradox

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Plan

- ★ What is program analysis (aka abstract interpretation)?
- ★ What is k -CFA?
- ★ The paradox of k -CFA
- ★ A resolution: OO vs. functional k -CFA
- ★ An exploitation: m -CFA
- ★ Evaluation and conclusion

What is abstract interpretation?

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AI is the sound computable approximation of evaluation.

Useful for:

- ★ Safety
- ★ Error detection
- ★ Optimization
- ★ Transformation
- ★ Termination
- ★ ...

Kinds:

- ★ Numerical analysis
- ★ Polyhedral analysis
- ★ Congruence analysis
- ★ Strictness analysis
- ★ Control-flow analysis
- ★ ...

What is control flow analysis?

Control flow analysis answers the question: where does control transfer to at a given procedure call?

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$$(\text{if } P (\lambda (y) \dots) (\lambda (z) \dots))$$

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○ . m (x) ;

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n(o) { return o.m(x); }
```

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```
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```

```
...
```

```
n(P ? q : r);
```

What is control flow analysis?

Control flow analysis answers the question: where does control transfer to at a given procedure call?

But this is not easy to decide in a *higher-order* language.

Which languages are higher-order? Almost all of them: Scheme, ML, Haskell, JavaScript, Java, Ruby, Python, etc.

Fun. programming in Java

```
class Calculus {

    interface Fun<X,Y> { Y apply(X x); }

    Fun<Fun<Double,Double>,Fun<Double,Double>> deriv =
        new Fun<Fun<Double,Double>,Fun<Double,Double>>() {
            double d = 0.001;
            public Fun<Double,Double> apply(final Fun<Double,Double> f) {
                return new Fun<Double,Double>() {
                    public Double apply(Double x) {
                        return (f.apply(x + d) - f.apply(x - d)) / (d * 2);}}};}};

    Fun<Double,Double> sin = new Fun<Double,Double>() {
        public Double apply(Double x) { return Math.sin(x); }
    };

    Fun<Double,Double> cos = deriv.apply(sin);
}
```

Fun. programming in Java

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class Calculus {

    interface Fun<X,Y> { Y apply(X x); }

    deriv =
        new Fun<Fun<Double,Double>,Fun<Double,Double>>() {
            double d = 0.001;
            apply(f) {
                return new Fun<Double,Double>() {
                    apply(x) {
                        return (f.apply(x + d) - f.apply(x - d)) / (d * 2);}}}};

    sin = new Fun<Double,Double>() {
        apply(x) { return Math.sin(x); }
    };

    cos = deriv.apply(sin);
}
```

Fun. programming in Scheme

```
(define (deriv f) ; (Number → Number) → (Number → Number)
  (let ((d 0.001))
    (λ (x) ; Number → Number
      (/ (- (f (+ x d))
            (f (- x d)))
         (* d 2)))))
```

```
(define cos (deriv sin))
(define dbl (deriv sqr))
```

What is k -CFA?

Approximation in OCFA

OCFA approximates closures by their code component.

$$\begin{aligned}(\text{deriv } \text{sin}) &= \langle (\lambda (x) \dots f..f\dots), [f \mapsto \text{sin}] \rangle \\ &\approx (\lambda (x) \dots f..f\dots) \\ f &\approx \{\text{sin}, \text{sqr}\}\end{aligned}$$

$$(\text{cos } \text{pi}) \approx \{-0.999, 6.2831, \dots\}$$

What is k -CFA?

k -CFA uses *calling contexts* to increase analysis precision:

`(deriv sin)1 ... (deriv sqr)2`

1-CFA:

★ $\langle (\lambda (x) \dots f..f\dots), [f \mapsto 1] \rangle$ and $[1 \mapsto \text{sin}]$.

★ $\langle (\lambda (x) \dots f..f\dots), [f \mapsto 2] \rangle$ and $[2 \mapsto \text{sqr}]$.

`(cos pi) ≈ {-0.999}`

Concrete semantics: k -CFA

$((f\ e)^\ell, \beta, \sigma, a) \Rightarrow (e', \beta', \sigma', a')$, where

$$\mathcal{E}(f, \beta, \sigma) = \langle (\lambda\ (x)\ e'), \beta'' \rangle \quad a' = \ell \cdot a$$

$$\mathcal{E}(e, \beta, \sigma) = d \quad \beta' = \beta''[x \mapsto a']$$

$$\sigma' = \sigma[a' \mapsto d]$$

And:

$$\mathcal{E}(x, \beta, \sigma) = \sigma(\beta(x))$$

$$\mathcal{E}((\lambda\ (x)\ e), \beta, \sigma) = \langle (\lambda\ (x)\ e), \beta \rangle$$

Abstract semantics: k -CFA

$((f\ e)^\ell, \hat{\beta}, \hat{\sigma}, \hat{a}) \rightsquigarrow (e', \hat{\beta}', \hat{\sigma}', \hat{a}')$, where

$$\hat{\mathcal{E}}(f, \hat{\beta}, \hat{\sigma}) \ni \langle (\lambda\ (x)\ e'), \hat{\beta}'' \rangle \qquad \hat{a}' = \lfloor \ell \cdot \hat{a} \rfloor_k$$

$$\hat{\mathcal{E}}(e, \hat{\beta}, \hat{\sigma}) = \hat{d} \qquad \hat{\beta}' = \hat{\beta}''[x \mapsto \hat{a}']$$

$$\hat{\sigma}' = \hat{\sigma} \sqcup [\hat{a}' \mapsto \hat{d}]$$

And:

$$\hat{\mathcal{E}}(x, \hat{\beta}, \hat{\sigma}) = \hat{\sigma}(\hat{\beta}(x))$$

$$\hat{\mathcal{E}}((\lambda\ (x)\ e), \hat{\beta}, \hat{\sigma}) = \{ \langle (\lambda\ (x)\ e), \hat{\beta} \rangle \}$$

The Paradox of k -CFA

k -CFA is hard

It did not take long to discover that the basic analysis, for any $k > 0$, was intractably slow for large programs.

Shivers, Higher-order control-flow analysis in retrospect: Lessons learned, lessons abandoned (2004)

What makes k -CFA hard?

Closures.

$((\lambda (f) (f u)^1 (f v)^2) (\lambda (x) (\lambda (y) x)))$

1-CFA:

- ★ $\langle (\lambda (y) x), [x \mapsto 1] \rangle$ and $[1 \mapsto u]$.
- ★ $\langle (\lambda (y) x), [x \mapsto 2] \rangle$ and $[2 \mapsto v]$.

What makes k -CFA hard?

$((\lambda (f_1) (f_1 0)^1 (f_1 1)^2)$
 $(\lambda (x_1)$
 \dots
 $((\lambda (f_n) (f_n 0)^{2n-1} (f_n 1)^{2n})$
 $(\lambda (x_n)$
 $(\lambda (z) (z x_1 \dots x_n))) \dots))$

1-CFA:

- ★ $[x_1 \mapsto 0, \dots, x_n \mapsto 0]$
- ★ $[x_1 \mapsto 1, \dots, x_n \mapsto 0]$
- ★ \dots
- ★ $[x_1 \mapsto 0, \dots, x_n \mapsto 1]$
- ★ $[x_1 \mapsto 1, \dots, x_n \mapsto 1]$

k -CFA is complete for EXPTIME (Van Horn and Mairson, '08).

k-CFA is easy

k-CFA of object-oriented programs is in PTIME.

Implemented in Datalog (Bravenboer and Smaragdakis, '09).

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PTIME \subsetneq EXPTIME

A Resolution: OO vs. functional k -CFA

FP in OO and back again

$A \rightarrow B$

\equiv

```
interface Fun<A,B> { B apply(A a); }
```

FP in OO and back again

$((\lambda (f) (f u) (f v)) (\lambda (x) (\lambda (y) x)))$

\equiv

```
new Fun() { apply(f) { f.apply(u); f.apply(v); }}  
.apply(new Fun() { apply(x) { new Fun() { apply(y) { x; }}}});
```

\equiv

```
class Lam1 { apply(f) { f.apply(u); f.apply(v); }}  
class Lam2 { apply(x) { new Lam3(x); }}  
class Lam3 { Lam3(x){this.x = x}; apply(y) { x; }}  
new Lam1().apply(new Lam2());
```

\equiv

$((\lambda (f) (f u) (f v)) (\lambda (x) (\text{let } x' = x \text{ in } (\lambda (y) x'))))$

\equiv

$((\lambda (f) (f u) (f v)) (\lambda (x) ((\lambda (x') (\lambda (y) x')) x)))$

An Exploitation: *m*-CFA

The idea: Flat closures

Change the representation of closures:

$$Clo = Lam \times (Var \rightarrow Addr)$$

$$Clo' = Lam \times Addr$$

At call sites, copy values of closure into the store.

Concrete semantics: m -CFA

$((f\ e)^\ell, a, \sigma) \Rightarrow (e', a', \sigma')$, where

$$\mathcal{E}(f, a, \sigma) = \langle (\lambda\ (x)\ e'), a'' \rangle \quad a' = \ell \cdot a$$

$$\mathcal{E}(e, a, \sigma) = d \quad d'_j = \sigma(a'', y_j)$$

$$\{y_1, \dots, y_n\} = fv(e') \setminus \{x\} \quad \sigma' = \sigma[a' \mapsto (d, d'_1, \dots, d'_n)]$$

And:

$$\mathcal{E}(x, a, \sigma) = \sigma(a, x)$$

$$\mathcal{E}((\lambda\ (x)\ e), a, \sigma) = \langle (\lambda\ (x)\ e), a \rangle$$

Abstract semantics: m -CFA

$((f\ e)^\ell, \hat{a}, \hat{\sigma}) \rightsquigarrow (e', \hat{a}', \hat{\sigma}')$, where

$$\hat{\mathcal{E}}(f, \hat{a}, \hat{\sigma}) \ni \langle (\lambda\ (x)\ e'), \hat{a}'' \rangle \quad \hat{a}' = \lfloor \ell \cdot \hat{a} \rfloor_m$$

$$\hat{\mathcal{E}}(e, \hat{a}, \hat{\sigma}) = \hat{d} \quad \hat{d}'_j = \hat{\sigma}(\hat{a}'', y_j)$$

$$\{y_1, \dots, y_n\} = fv(e') \setminus \{x\} \quad \hat{\sigma}' = \hat{\sigma} \sqcup [\hat{a}' \mapsto (\hat{d}, \hat{d}'_1, \dots, \hat{d}'_n)]$$

And:

$$\hat{\mathcal{E}}(x, \hat{a}, \hat{\sigma}) = \hat{\sigma}(\hat{a}, x)$$

$$\hat{\mathcal{E}}((\lambda\ (x)\ e), \hat{a}, \hat{\sigma}) = \{ \langle (\lambda\ (x)\ e), \hat{a} \rangle \}$$

Evaluation and conclusion

Speed: Worst-case benchmark

Terms	$k = 1$	$m = 1$	poly., $k=1$	$k=0$
69	ϵ	ϵ	ϵ	ϵ
123	ϵ	ϵ	ϵ	ϵ
231	46 s	ϵ	2 s	ϵ
447	∞	3 s	5 s	2 s
879	∞	48 s	1 m 8 s	15 s
1743	∞	51 m	∞	3 m 48 s

Speed and precision

Prog/Terms	$k = 1$	$m = 1$	poly., $k=1$	$k=0$
eta / 49	€ 7	€ 7	€ 3	€ 3
map / 157	€ 8	€ 8	€ 8	€ 6
sat / 223	∞ -	€ 12	1s 12	€ 12
regex / 1015	4s 25	3s 25	14s 25	2s 25
scm2java / 2318	5s 86	3s 86	3s 79	4s 79
interp / 4289	5s 123	4s 123	9s 123	5s 123
scm2c / 6219	179s 136	143s 136	157s 131	55s 131

Column 1: analysis time, Column 2: inlinings enabled

Doggie bag

- ★ k -CFA of λ -programs is hard because of closures.
- ★ k -CFA of OO-programs is easy because of flat closures.
- ★ m -CFA: similar precision, less cost.
- ★ m -CFA is always in PTIME.



The End

Thank you.

Preprint, implementation, benchmarks:

<http://www.ucombinator.org/projects/mcfa/>