CONTEXTUAL COINDUCTION AND CLASSICAL LOGIC

OR: FEARLESSLY OBSERVING THE INFINITE

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COINDUCTION, EH?

So when do we start talking about categories?

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COINDUCTION, EH?

SO WHEN DO WE START TALKING ABOUT CATEGORIES?





(STRUCTURAL) INDUCTION FOR CLASSIC ALGORITHMS

The Workhorse of Computer Science Theories A & B

Many classic (theory A) algorithms follow inductive structure:

Sorting Searching Tree-based data structures

Language foundations (theory B): induction all the way down

Syntax Semantics Compiliers

Type systems

Proof assistants (the magnum opus of theory B) are effectively big induction engines.

INTERACTING WITH OTHERS

Programs that need to interact with the outside world while they run are inherently coinductive:

Operating systems & User Interfaces Web servers & Networks Distributed systems & IoT Real time & Embedded systems Control software & robotics

Coinduction is essiential to understand modern systems

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....
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How To Coinduction?

COINDUCTION WITH CONFIDENCE

WHAT IS THE COINDUCTIVE HYPOTHESIS?

AND WHEN CAN I USE IT?

Coq's coinduction founded on <u>begging the question</u> $cofix : (\forall x.P(x) \rightarrow P(x)) \rightarrow \forall x.P(x) ???$

<u>Viciously</u> circular logic

Assume exactly what you are trying to provebut only use it sometimes?

CIH should be guarded/inside a constructor

unless you eliminate it too soon ...

CIH should be guarded/behind a projection

unless you use another projection in the "wrong spot" ...

... or pass the result to another function (???)

<u>"Guardedness"</u> is trying to say something about the context's structure and how it shrinks

STRUCTURAL (Co)INDUCTION

COINDUCTION = INDUCTION ON THE CONTEXT

Principle (Induction on Natural Number Values)

Given property P: Nat \rightarrow Prop on natural numbers, P(n) holds for all numbers n: Nat if and only if

P(0) holds, and

for all n: Nat, P(n) implies P(n + 1).

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Principle (Coinduction on Stream Observations*)

Given property P: (Stream $A \rightarrow B$) \rightarrow Prop on stream observations, P(f) holds for all observations f: Stream $A \rightarrow B$ if and only if

for all $g : A \to B$, $P(g \circ \text{Head})$ holds, and for all $h : \text{Stream } A \to B$, P(h) implies $P(h \circ \text{Tail})$.

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COINDUCTIVE HYPOTHESIS = LABELING THE OBSERVER

Principle (Contextual Equivalence) Given x : A and y : A, x = yIF AND ONLY IF for all f, f(x) = f(y).

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Corollary (Contextual Stream Equality) *Given xs* : Stream *A and ys* : Stream *A*,

xs = ys IF AND ONLY IF for all g, g(Head(xs)) = g(Head(ys)) AND for all h, h(xs) = h(ys) IMPLIES h(Tail(xs)) = h(Tail(ys))

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By Begging the Question (Boo...)

always
$$x = \text{Cons } x \text{ (always } x)$$

repeat $f x = \text{Cons } x \text{ (repeat } f \text{ (} f x\text{))}$

Theorem *repeat* $(\lambda y.y) x = always x$

Proof.

Assume coinductive hypothesis (CoIH): repeat $(\lambda y.y) x = always x$. Now prove repeat $(\lambda y.y) x = always x$:

repeat
$$(\lambda y.y) x = always x$$
 (CoIH?)

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BY CONTEXTUAL STREAM EQUALITY (NO GUESSWORK!)

 $\begin{aligned} \text{Head}(always x) &= x \\ \text{Head}(repeat f x) &= x \end{aligned} \qquad \begin{aligned} \text{Tail}(always x) &= always x \\ \text{Head}(repeat f x) &= x \end{aligned} \qquad \begin{aligned} \text{Tail}(repeat f x) &= repeat f (f x) \end{aligned}$

Theorem

repeat $(\lambda y.y) x = always x$

Proof. By contextual stream equality:

(Head) Show Head(*repeat* $(\lambda y.y) x$) = Head(*always x*)

 $Head(repeat (\lambda y.y) x) = x = Head(always x)$

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$$map f (repeat f x) = repeat f (f x)$$

 $\begin{aligned} & \text{Head}(map \ f \ xs) = f \ (\text{Head} \ xs) \\ & \text{Head}(repeat \ f \ x) = x \end{aligned} \qquad \begin{aligned} & \text{Tail}(map \ f \ xs) = map \ f \ (\text{Tail} \ xs) \\ & \text{Tail}(repeat \ f \ x) = repeat \ f \ (f \ x) \end{aligned}$

Theorem

for all x, map f (repeat f x) = repeat f (f x)

Proof. By contextual stream equality:

(Head) Show $\forall x$, Head(map f (repeat f x)) = Head(repeat f (f x)). Head(map f (repeat f x)) = f(Head(repeat f x)) = f(x) = Head(repeat f (f x))

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MUTUAL COINDUCTION

EVENS AND ODDS

 $evens (x_0, x_1, x_2, x_3, x_4, x_5, \dots) = x_0, x_2, x_4, \dots$ $odds (x_0, x_1, x_2, x_3, x_4, x_5, \dots) = x_1, x_3, x_5, \dots$ $merge (x_0, x_1, x_2, \dots) (y_0, y_1, y_2, \dots) = x_0, y_0, x_1, y_1, x_2, y_2, \dots$

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$$odds (x_0, x_1, x_2, x_3, x_4, x_5, \dots) = x_1, x_3, x_5, \dots$$

$$merge (x_0, x_1, x_2, \dots) (y_0, y_1, y_2, \dots) = x_0, y_0, x_1, y_1, x_2, y_2, \dots$$

Head(merge xs ys) = Head xsHead(Tail(merge xs ys)) = Head ysTail(Tail(merge xs ys)) = merge (Tail xs) (Tail ys)

PROOF BY MUTUAL COINDUCTION

Theorem

for all xs and ys, evens (merge xs ys) = xs AND odds (merge xs ys) = ys

Proof. By mutual contextual stream equality:
(Head) Head(evens (merge xs ys)) = Head(merge xs ys) = Head xs Head(odds (merge xs ys)) = Head(evens (Tail(merge xs ys))) = Head(Tail(merge xs ys)) = Head ys

PROOF BY MUTUAL COINDUCTION

Theorem

for all xs and ys, evens (merge xs ys) = xs AND odds (merge xs ys) = ys

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(Head) Head(evens (merge xs ys)) = Head(merge xs ys) = Head xs Head(odds (merge xs ys)) = Head(evens (Tail(merge xs ys))) = Head(Tail(merge xs ys)) = Head ys

(Tail) $\forall xs, ys, ColH_1 : h(evens(merge xs ys)) = h(xs), AND$ $ColH_2 : h(odds(merge xs ys)) = h(ys).$

h(Tail(evens (merge xs ys))) = h(evens (Tail(Tail(merge xs ys))))= h(evens (merge (Tail xs) (Tail ys))) = h(Tail xs) (ColH_1[(Tail xs)/xs, (Tail ys)/ys])

h(Tail(odds (merge xs ys))) = h(odds (Tail(merge xs ys))))= h(odds (merge (Tail xs) (Tail ys)))= $h(\text{Tail } ys) \qquad (ColH_2[(\text{Tail } xs)/xs, (\text{Tail } ys)/ys])$

BASE CASES

Theorem

for all xs, merge (evens xs) (odds xs) = xs.

Proof. By strong contextual stream equality:

(Head) Head(merge (evens xs) (odds xs)) = Head xs

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BASE CASES

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 $(\text{Head} \circ \text{Tail}) \quad \text{Head}(\text{Tail}(merge (evens xs) (odds xs))) = \text{Head}(\text{Tail } xs)$

... CONTINUED

Theorem

for all xs, merge (evens xs) (odds xs) = xs.

Proof. By strong contextual stream equality:

(Tail \circ Tail) Assume *ColH* : $\forall xs$, h(merge (evens xs) (odds xs)) = h(xs). Show $\forall xs$, h(Tail(Tail(merge (evens xs) (odds xs)))) = h(Tail(Tail(xs))).

> h(Tail(Tail(merge (evens xs) (odds xs)))) = h(merge (Tail(evens xs)) (Tail(odds xs))) = h(merge (evens (Tail(Tail xs))) (odds (Tail(Tail xs))))

... CONTINUED

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COINDUCTIVE RULES IN CLASSICAL LOGIC

FINITE INDUCTION

Consider property P : Bool \rightarrow Prop Is P(x) **true** for any value x : Bool? All the cases of x:

$$x = tt$$

 $x = ff$

FINITE INDUCTION

Consider property P : Bool \rightarrow Prop Is P(x) **true** for any value x : Bool? All the cases of x:

$$x = \text{tt}$$

$$x = \text{ff}$$

$$\frac{\Gamma \vdash P(\text{tt}) \bullet \vdash P(\text{ff})}{\Gamma, x : \text{Bool} \vdash P(x)} \text{Bool Ind}$$

$$\approx$$

$$P(\text{tt}) \implies P(\text{ff}) \implies \forall x : \text{Bool} . P(x)$$

INFINITE INDUCTION?

TOO MANY CASES

Consider property P : Nat \rightarrow Prop Is P(x) **true** for any value x : Nat? All the cases of x:

$$x = 0$$
$$x = 1$$
$$x = 2$$

. . .

INFINITE INDUCTION?

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TOO MANY CASES
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Consider property P: Nat \rightarrow Prop Is P(x) true for any value x : Nat? All the cases of x: $\mathbf{x} = \mathbf{0}$ x = 1x = 2. . . $\Gamma \vdash P(0) \quad \Gamma \vdash P(1) \quad \Gamma \vdash P(2) \quad \dots$ $\Gamma, x : \operatorname{Nat} \vdash P(x)$

AN INDUCTION PRINCIPLE

Consider property P : Nat \rightarrow Prop Is P(x) true for any value x : Nat? All the cases of x: $\mathbf{x} = \mathbf{0}$ x = y + 1 for some other y : Nat $\frac{\Gamma \vdash P(0) \quad \Gamma, \mathbf{y} : \operatorname{Nat}, P(\mathbf{y}) \vdash P(\mathbf{y}+1)}{\Gamma, \mathbf{x} : \operatorname{Nat} \vdash P(\mathbf{x})} \operatorname{Nat} Ind$

$$P(0) \implies (\forall y : \text{Nat} . P(y) \implies P(y+1)) \implies \forall x : \text{Nat} . P(x)$$

MAKING COINDUCTION LESS VICIOUS

$$\frac{\Gamma, xs : \text{Stream } A, P(xs) \vdash P(xs)}{\Gamma, xs : \text{Stream } A \vdash P(xs)} \text{ vicious!}$$

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$$\frac{\Gamma, xs : \text{Stream } A, P(xs) \vdash P(xs)}{\Gamma, xs : \text{Stream } A \vdash P(xs)} \text{ vicious!}$$

$$\frac{xs: \operatorname{Stream} A \vdash P\left(\operatorname{Head} xs, \operatorname{Head}(\operatorname{Tail} xs), \operatorname{Head}(\operatorname{Tail}(\operatorname{Tail} xs)), \ldots\right)}{xs: \operatorname{Stream} A \vdash P(xs)}$$

MAKING COINDUCTION LESS VICIOUS

$$\frac{\Gamma, xs : \text{Stream } A, P(xs) \vdash P(xs)}{\Gamma, xs : \text{Stream } A \vdash P(xs)} \text{ vicious!}$$

$$\frac{xs: \text{Stream } A \vdash P\left(\begin{array}{c} \text{Head } xs, \text{Head}(\text{Tail } xs), \\ \text{Head}(\text{Tail}(\text{Tail } xs)), \ldots\right)}{xs: \text{Stream } A \vdash P(xs)}$$

Key Idea: Move focus away from stream values xs: Stream A, and consider the cases for any observation α ÷ Stream A that might look at xs

THE STRUCTURE OF OBSERVATIONS

And observation $\alpha \div \textit{StreamA}$ of streams might be:

 $\alpha =$ Head(observe element 0) $\alpha =$ Head \circ Tail(observe element 1) $\alpha =$ Head \circ Tail \circ Tail \circ Tail \circ Tail \circ (observe element 2) $\alpha =$ Head \circ Tail \circ Tail \circ Tail \circ Tail \circ Observe element 3)

(observe element i)

 $\alpha = \mathsf{Head} \circ \mathsf{Tail}^i$

. . .

THE STRUCTURE OF OBSERVATIONS

. . .

And observation $\alpha \div \textit{StreamA}$ of streams might be:

$$\begin{array}{ll} \alpha = {\rm Head} & ({\rm observe\ element\ 0}) \\ \alpha = {\rm Head} \circ {\rm Tail} & ({\rm observe\ element\ 1}) \\ \alpha = {\rm Head} \circ {\rm Tail} \circ {\rm Tail} & ({\rm observe\ element\ 2}) \\ \alpha = {\rm Head} \circ {\rm Tail} \circ {\rm Tail} \circ {\rm Tail} & ({\rm observe\ element\ 3}) \\ \dots \\ \alpha = {\rm Head} \circ {\rm Tail}^i & ({\rm observe\ element\ i}) \end{array}$$

ALL stream observations $\alpha \div StreamA$ are one of: $\alpha = \beta \circ \text{Head}$ for some observation $\beta \div A$, or

 $\alpha = \delta \circ \mathsf{Tail}$ for some other $\delta \div \mathsf{Stream} A$

A COINDUCTION PRINCIPLE

Consider property P : - Stream $A \rightarrow$ Prop Is $P(\alpha)$ **true** for any observation $\alpha \div$ Stream AAll the cases of α :

 $\alpha = \beta \circ \text{Head for some observation } \beta \div A$ $\alpha = \delta \circ \text{Tail for some other } \delta \div \text{Stream } A$

$$\frac{\Gamma, \beta \div A \vdash P(\beta \circ \text{Head}) \quad \Gamma, \delta \div \text{Stream } A, P(\delta) \vdash P(\delta \circ \text{Tail})}{\Gamma, \alpha \div \text{Stream } A \vdash P(\alpha)} \text{ Stream } Colnd$$

BUT WHAT ABOUT BISIMULATION?

Given any binary relation R: Stream $A \times$ Stream $A \rightarrow$ Prop,

 $(\forall xs, ys : \text{Stream } A. R(xs, ys) \implies \text{Head } xs = \text{Head } ys) \implies$ $(\forall xs, ys : \text{Stream } A. R(xs, ys) \implies R(\text{Tail } xs, \text{Tail } ys)) \implies$ $\forall xs, ys. R(xs, ys) \implies xs = ys$

BUT WHAT ABOUT BISIMULATION?

Given any binary relation R: Stream $A \times$ Stream $A \rightarrow$ Prop,

$$(\forall xs, ys : \text{Stream } A. \ R(xs, ys) \implies \text{Head } xs = \text{Head } ys) \implies$$

 $(\forall xs, ys : \text{Stream } A. \ R(xs, ys) \implies R(\text{Tail } xs, \text{Tail } ys)) \implies$
 $\forall xs, ys. \ R(xs, ys) \implies xs = ys$

Bisimulation is derivable from Stream *Colnd* with help from observational equivalence:

$$\frac{\Gamma, \alpha \div A \vdash \langle v \| \alpha \rangle = \langle w \| \alpha \rangle}{\Gamma \vdash v = w : A} \text{ Obs.Equiv.}$$

 $\langle \mathbf{v} \| \alpha \rangle$ is a <u>computation</u> where α observes \mathbf{v} :

$$\frac{\Gamma \vdash v : A \quad \Gamma \vdash \alpha \div A}{\Gamma \vdash \langle v \| \alpha \rangle} Cut$$

Computing With Contextual Coinduction

DUALITIES OF COMPUTATION

Embodying the Context



- A producer *v* gives an answer
- A consumer *e* asks a question
- A command $\langle v \| e \rangle$ is an interaction

One side moves first in a predictable pattern

The other side responds to first move

Data = patterns of answers

Codata = patterns of questions

(Co)INDUCTION AS STRUCTURAL (CO)RECURSION

A call stack $x \cdot \alpha$ contains an:

argument x

return pointer α

map is well-founded because its argument shrinks:

repeat is well-founded because its return pointer shrinks:

$$\langle repeat || f \cdot x \cdot \alpha \circ Head \rangle = \langle x || \alpha \rangle$$
$$\langle repeat || f \cdot x \cdot \alpha \circ Tail \rangle = \langle repeat || f \cdot f x \cdot \alpha \rangle$$

More Adventurous Structural (Co)Recursion

evens and *odds* are <u>mutually</u> well-founded because *even*'s return pointer always shrinks and *odds* return pointer stays the same:

$$\langle evens \| xs \cdot \alpha \circ \mathsf{Head} \rangle = \langle xs \| \alpha \circ \mathsf{Head} \rangle$$
$$\langle evens \| xs \cdot \alpha \circ \mathsf{Tail} \rangle = \langle odds \| \mathsf{Tail} xs \cdot \alpha \rangle$$
$$\langle odds \| xs \cdot \alpha \rangle = \langle evens \| \mathsf{Tail} xs \cdot \alpha \rangle$$

merge is well-founded by <u>strong</u> corecursion because its return pointer shrinks by 2, and the first 2 base cases are covered:

CONSISTENCY OF EQUALITY

DO THE SYNTACTIC RULES MEAN ANYTHING?

Theorem *If* $\Gamma \vdash \langle v_1 || e_1 \rangle = \langle v_2 || e_2 \rangle$, *then* $\langle v_1 || e_1 \rangle$ *and* $\langle v_2 || e_2 \rangle$ *are contextually equivalent (as usual, per the operational semantics).*

Proof. By a logical relation based on orthogonal fixed points in a subtyping lattice. Key: Knaster-Tarski and Kleene fixed points coincide.

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Proof. By a logical relation based on orthogonal fixed points in a subtyping lattice. Key: Knaster-Tarski and Kleene fixed points coincide.

Corollary
If
$$\alpha \div \text{Bool} \vdash \langle v_1 || e_2 \rangle = \langle v_2 || e_2 \rangle$$
, then either
 $\langle v_1 || e_2 \rangle \mapsto \langle \text{tt} || \alpha \rangle \nleftrightarrow \langle v_2 || e_2 \rangle$ or
 $\langle v_1 || e_2 \rangle \mapsto \langle \text{ff} || \alpha \rangle \twoheadleftarrow \langle v_2 || e_2 \rangle.$

Corollary • \vdash tt = ff : Bool *is not derivable.*

WHAT ABOUT EFFECTS?

Programs can do some funny things

Conventional side effects Mutable state / references Input / Output Exceptions and Jumps Infinite loops

Suprising wringle: Information effects Dual to control effects (manipulating control flow) Erasing answers Duplicating answers

Both can cause (co)inductive reasoning principles to go awry For example, they can cause inconsistency

(Co)INDUCTION AND EVALUATION STRATEGY

ADJUSTING STRENGTH TO SAVE CONSISTENCY

Induction principles (like Nat *Ind*) + Effects are Fully consistent under call-by-value evaluation Safe for strict properties in call-by-name evaluation

Strict on $x \ni \Psi(x) ::= \langle x || E \rangle = \langle x || E' \rangle$ $(E, E' \in Eval.Cxt.)$ $| \forall y : A. \Psi(x)$ $(x \neq y)$ $| P \Longrightarrow \Psi(x)$ $(x \notin FV(P))$ $| \cdots$

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Coinduction principles (like Stream *Colnd*) + Effects are Fully consistent under call-by-name evaluation Safe for productive properties in call-by-value evaluation

Productive on $\alpha \ni$

Ψ

$$(\alpha) ::= \langle V \| \alpha \rangle = \langle V' \| \alpha \rangle \qquad (V, V' \in Value) \\ | \forall \beta \div A. \Psi(\alpha) \qquad (\alpha \neq \beta) \\ | P \Longrightarrow \Psi(\alpha) \qquad (\alpha \notin FV(P)) \\ | \dots$$

Answers $\langle Me || You \rangle$ Questions

Downen & Ariola, Structures for Structural Recursion, ICFP '16.

Downen & Ariola, <u>A Computational Understanding of Classical (Co)Recursion</u>, PPDP '20.

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