# Divide-and-Conquer Checkpointing for Arbitrary Programs with No User Annotation

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Joint work with Barak Avrum Pearlmutter

## Barak and my Work

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AD in functional programs.

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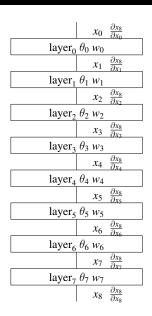
AD in functional programs.

AD is easier in functional programs.

#### A Neural Network

	$x_0$	$\frac{\partial x_8}{\partial x_0}$
layer <sub>0</sub>	$\theta_0 w_0$	,
	$x_1$	$\frac{\partial x_8}{\partial x_1}$
layer <sub>1</sub>	$\theta_1 w_1$	
	$x_2$	$\frac{\partial x_8}{\partial x_2}$
layer <sub>2</sub>	$\theta_2 w_2$	_
	$x_3$	$\frac{\partial x_8}{\partial x_3}$
layer <sub>3</sub>	$\theta_3 w_3$	
	<i>x</i> <sub>4</sub>	$\frac{\partial x_8}{\partial x_4}$
layer <sub>4</sub>	$\theta_4 w_4$	
	<i>x</i> <sub>5</sub>	$\frac{\partial x_8}{\partial x_5}$
layer <sub>5</sub>	$\theta_5 w_5$	Ť
	<i>x</i> <sub>6</sub>	$\frac{\partial x_8}{\partial x_6}$
layer <sub>6</sub>	$\theta_6 w_6$	
	<i>x</i> <sub>7</sub>	$\frac{\partial x_8}{\partial x_7}$
layer <sub>7</sub>	$\theta_7 w_7$	
	<i>x</i> <sub>8</sub>	$\frac{\partial x_8}{\partial x_8}$

## A Neural Network is a (Functional) Program



```
net [\theta_0,\ldots,\theta_7] [w_0,\ldots,w_7] x_0 \stackrel{\triangle}{=}
     let x_1 = \operatorname{layer}_0 \theta_0 w_0 x_0
             x_2 = \operatorname{layer}_1 \theta_1 w_1 x_1
             x_3 = \text{layer}_2 \theta_2 w_2 x_2
             x_4 = \text{layer}_3 \theta_3 w_3 x_3
             x_5 = \text{layer}_4 \theta_4 w_4 x_4
             x_6 = \text{layer}_5 \theta_5 w_5 x_5
             x_7 = \text{layer}_6 \theta_6 w_6 x_6
             x_8 = \text{layer}_7 \theta_7 w_7 x_7
     in x_8
```

## A (Functional) Program

$$f \begin{bmatrix} w_0, w_1 \end{bmatrix} \begin{bmatrix} x_0, x_1 \end{bmatrix} \stackrel{\triangle}{=}$$

$$\mathbf{let} \quad t_0 = w_0 \times x_0$$

$$t_1 = w_1 \times x_1$$

$$y = t_0 + t_1$$

$$\mathbf{in} \quad y$$

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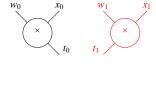
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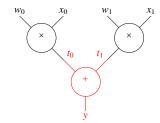
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$$in y$$



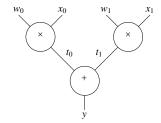
$$f \begin{bmatrix} w_0, w_1 \end{bmatrix} \begin{bmatrix} x_0, x_1 \end{bmatrix} \stackrel{\triangle}{=}$$

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- ► Can perform backpropagation on (functional) programs by having an execution of the program generate a network. This is called reverse-mode automatic differentiation (AD).

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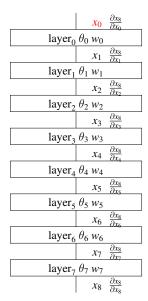
B. Speelpenning, *Compiling Fast Partial Derivatives of Functions Given by Algorithms*, Department of Computer Science, University of Illinois at Urbana-Champaign, 1980.

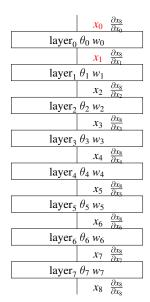
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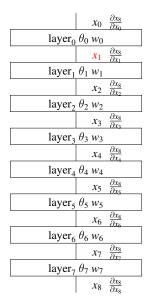
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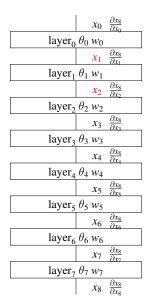
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- A.E. Bryson, Jr. and Y.-C. Ho, Applied optimal control, Blaisdell, 1969.

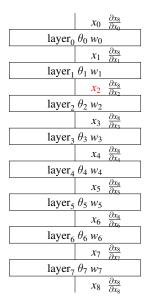


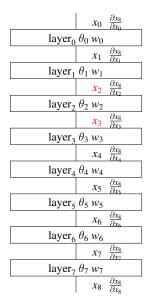


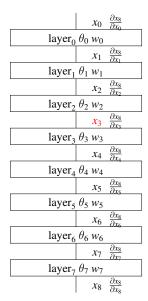


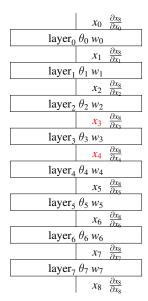


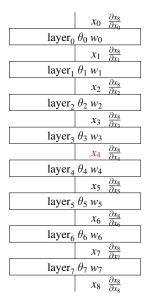
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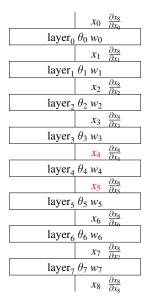


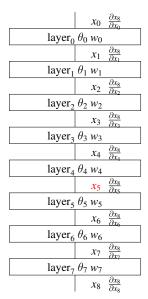


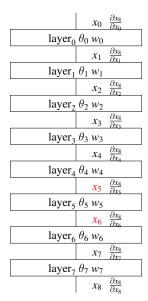


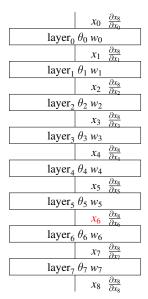


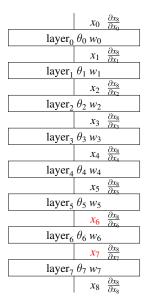


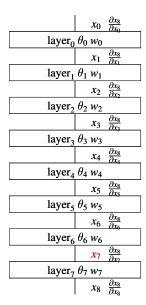


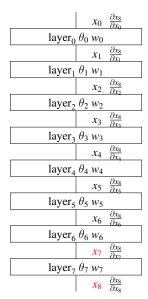


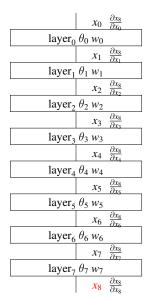


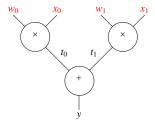


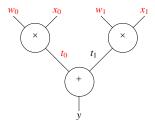


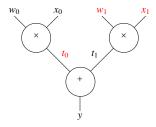


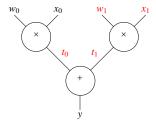


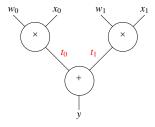


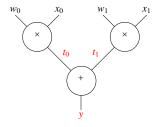


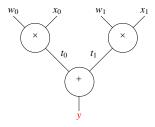












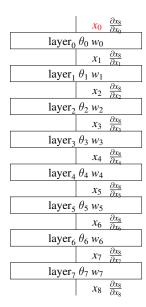
### Some Observations

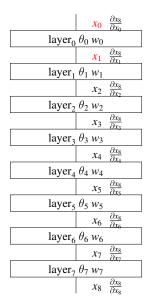
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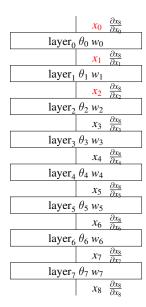
Only need to store live variables.

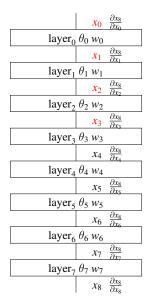
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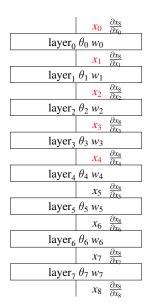
- Only need to store live variables.
- Most deep learning frameworks store all intermediate variables to allow subsequent backpropagation.

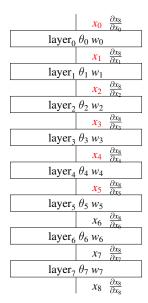


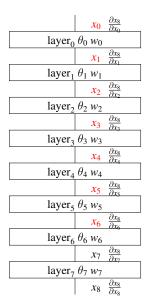


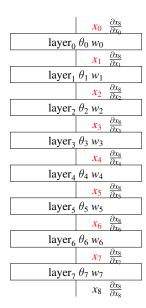


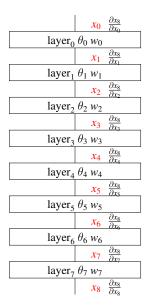


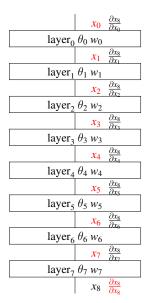


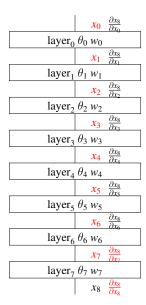


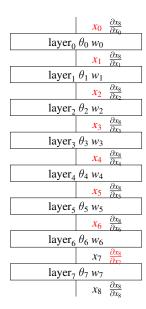


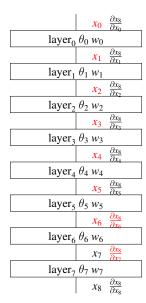


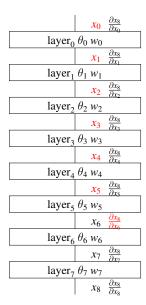


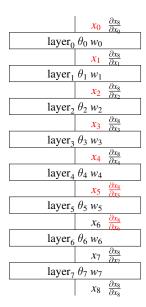


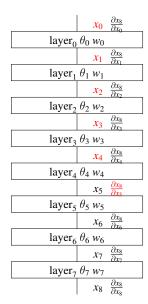




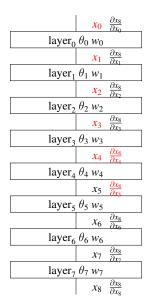


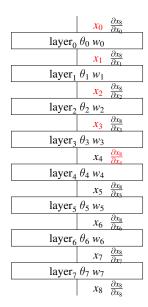


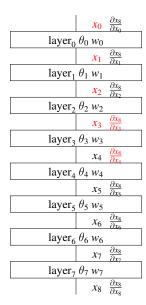




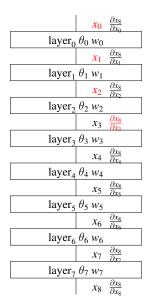
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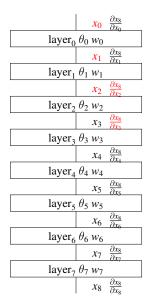


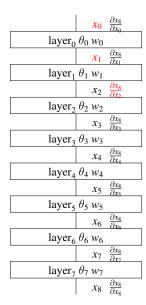


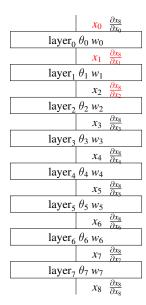


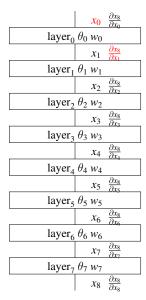
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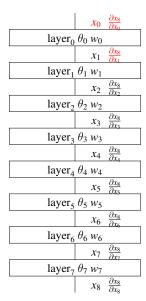


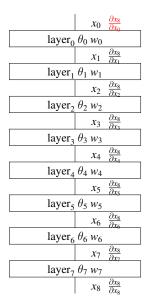












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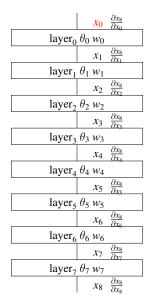
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- ▶ It doesn't matter because storage use is dominated by maximal use.
- ▶ Maximal use is proportional to the depth of the network *i.e.*, the running time of the program.

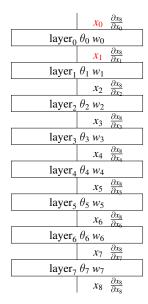
▶ If running time of primal is O(t)

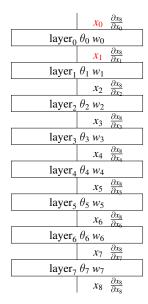
If running time of primal is O(t) and primal has maximal live storage O(w)

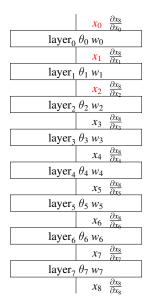
- If running time of primal is O(t) and primal has maximal live storage O(w)
- then reverse mode takes O(wt) space

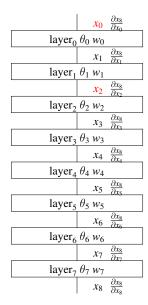
- If running time of primal is O(t) and primal has maximal live storage O(w)
- then reverse mode takes O(wt) space and O(t) time.

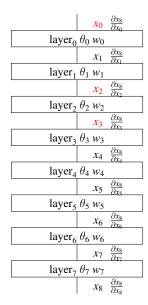


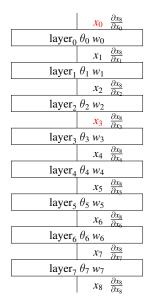


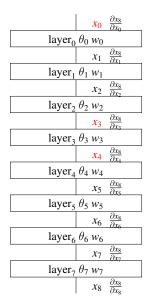


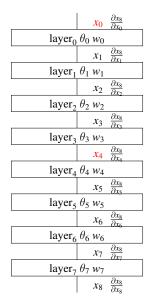


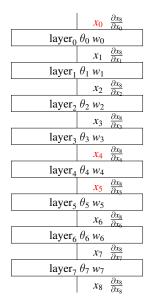


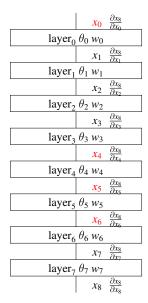


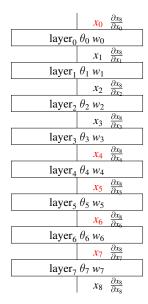


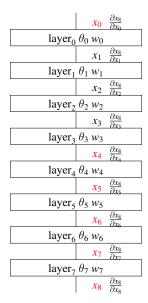


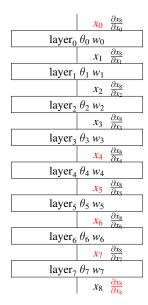


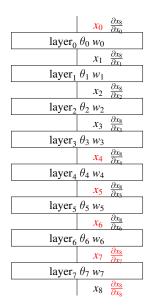


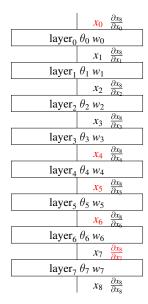


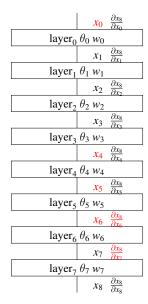


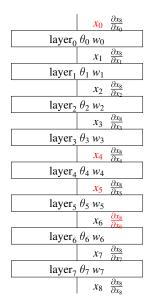


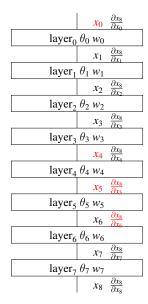


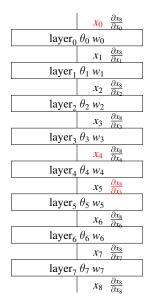


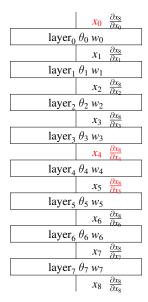


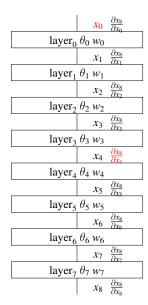


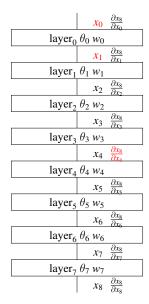


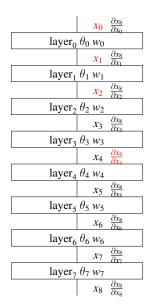


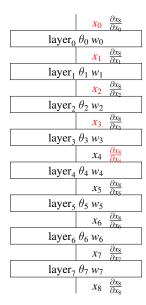


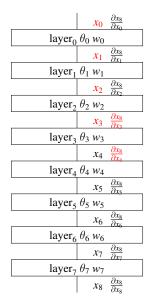


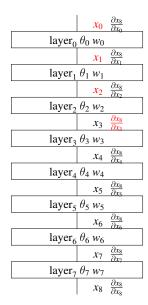


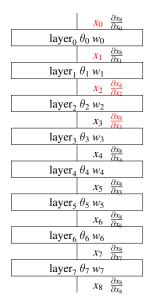


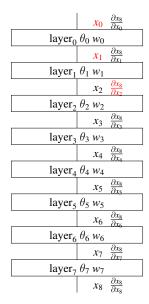


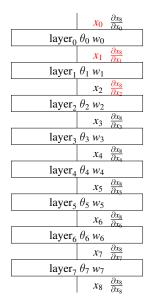


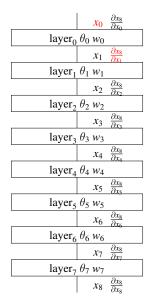


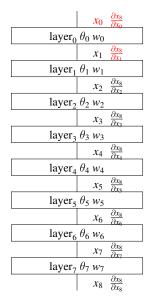


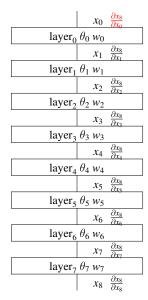












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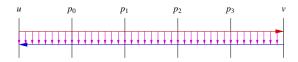
- ► Trades off extra running time for reduction in space.
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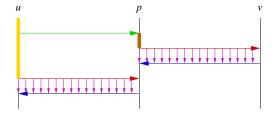
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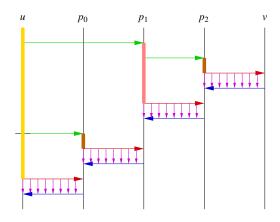
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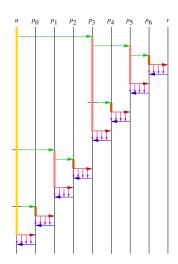
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   Only need saved intermediate variables from forward pass for current stage.
- Can perform divide-and-conquer.











▶ If running time of primal is O(t)

If running time of primal is O(t) and primal has maximal live storage O(w)

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- then reverse mode takes  $O(w \log t)$  space and  $O(t \log t)$  time.

#### A (Brief) History of Divide-and-Conquer Checkpointing

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A. Griewank, *Achieving Logarithmic Growth of Temporal and Spatial Complexity in Reverse Automatic Differentiation*, Optimization Methods and Software, 1:35-54, 1992.

L. Hascoët and V. Pascual, *TAPENADE 2.1 User's Guide*, Rapport technique 300, INRIA, 2004.

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10 continue

L. Hascoët and V. Pascual, *TAPENADE 2.1 User's Guide*, Rapport technique 300, INRIA, 2004.

$$\begin{array}{c} \text{do 10 i=1, n} \\ \text{...} \\ 10 \quad \text{continue} \end{array} \right\} \rightsquigarrow \begin{cases} \text{c$ad binomial-ckp n+1 30 1}} \\ \begin{array}{c} \text{do 10 i=1, n} \\ \\ \text{...} \\ 10 \quad \text{continue} \end{array}$$

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https://www-sop.inria.fr/tropics/tapenade/faq.html
Assuming that the final number of iterations N is known, and
assuming that each iteration has the same runtime cost,

#### Desiderata

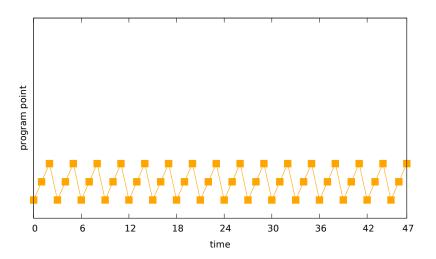
#### Desiderata

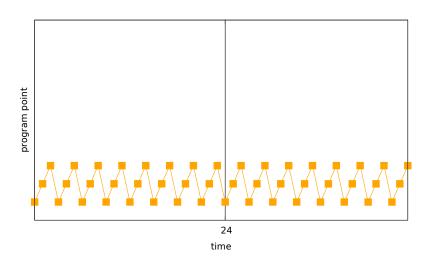
A (deep) neural network has no loops (except inside primitives).

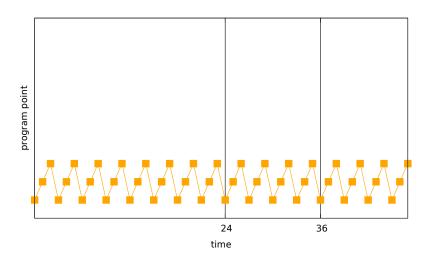
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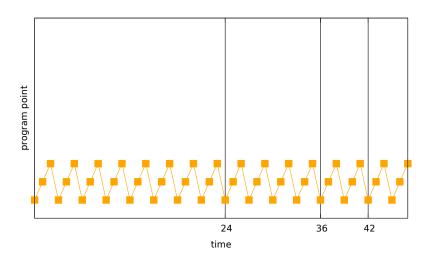
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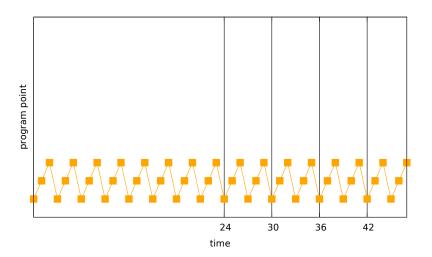
Want to implement for arbitrary code (not just a single DO loop).

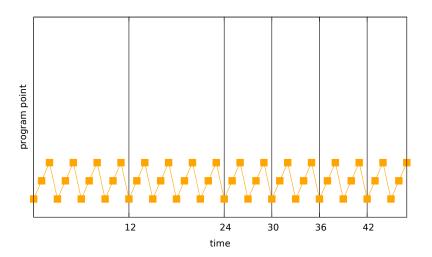


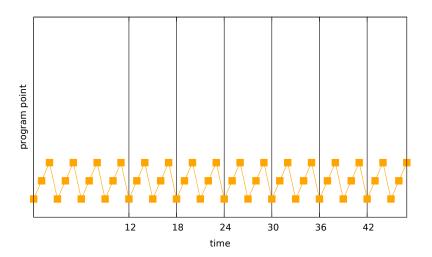


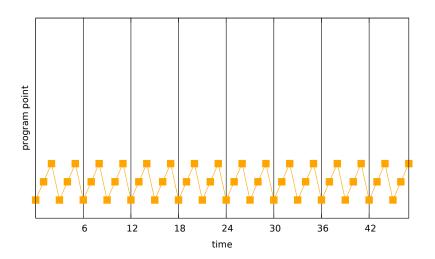


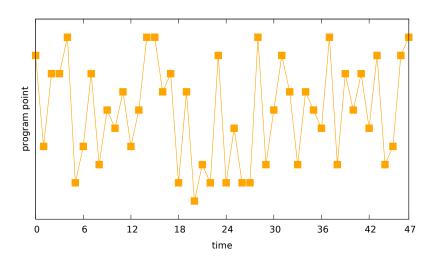


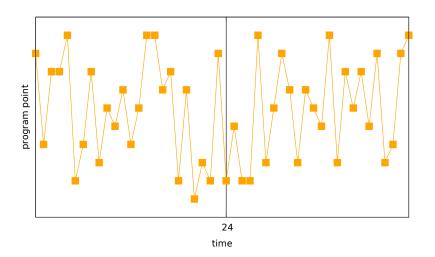


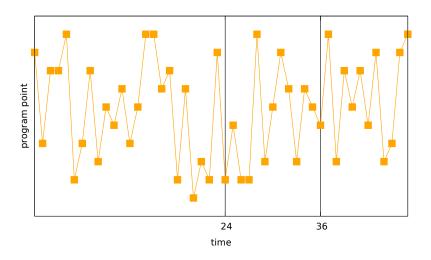


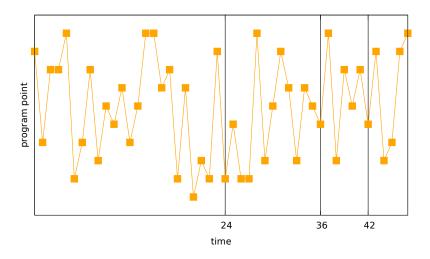


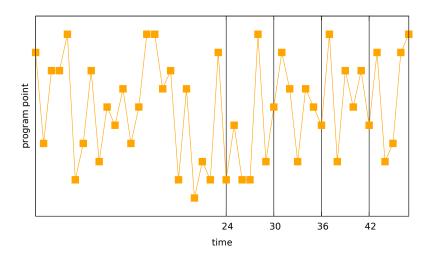


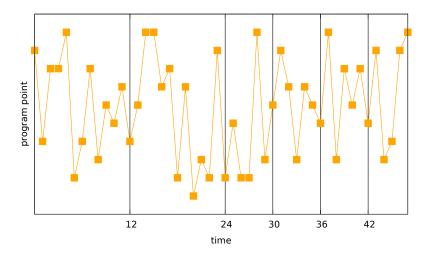


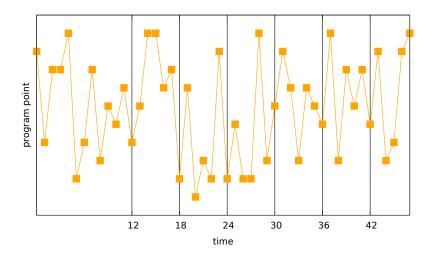


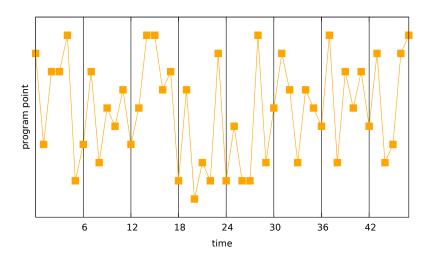












### **Key Challenges**

Need to interleave generation of the network with forward and backward passes through the network.

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Need to interleave generation of the network with forward and backward passes through the network.

Portions of the network need to be (re)generated, and (re)evaluated with forward and backward passes, multiple times and out of order.

### Key Idea

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function main(w)
  local x = f(w)
  local y = h(g(x))
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```
function main(w)
   for i = 1, 5
   if i==1 then
      local x = f(w)
   elseif i==2 then
      local t = q(x)
   elseif i==3 then
      local v = h(t)
   elseif i==4 then
      local z = p(y)
   elseif i==5 then
     return z
   end
end
```

#### Core Language

$$e := c \mid x \mid \lambda x.e \mid e_1 \mid e_2 \mid \mathbf{if} \mid e_1 \mathbf{then} \mid e_2 \mathbf{else} \mid e_3 \mid \diamond e \mid e_1 \bullet e_2$$

### Adding AD Operators to the Core Language

$$\overleftarrow{\mathcal{J}}:fx\grave{y}\mapsto (y,\grave{x})$$

$$\mathcal{J}: f \ x \ \dot{y} \mapsto (y, \dot{x})$$

To compute 
$$(y, \dot{x}) = \mathcal{J} f x \dot{y}$$
:

**base case** (
$$f x \text{ fast}$$
):  $(y, \dot{x}) = \overleftarrow{\mathcal{J}} f x \dot{y}$  (step 0)

**inductive case**: 
$$h \circ g = f$$
 (step 1)

$$z = g x (step 2)$$

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measure the length of the primal computation

- measure the length of the primal computation
- interrupt the primal computation at a portion of the measured length

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- save the state of the interrupted computation as a capsule

- measure the length of the primal computation
- interrupt the primal computation at a portion of the measured length
- save the state of the interrupted computation as a capsule
- resume an interrupted computation from a capsule

PRIMOPS  $f x \mapsto l$ 

Return the number l of evaluation steps needed to compute y = f(x).

INTERRUPT  $f x l \mapsto z$ 

Run the first l steps of the computation of f(x) and return a capsule z.

RESUME  $z \mapsto y$ 

PRIMOPS  $f x \mapsto l$ 

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PRIMOPS  $f x \mapsto l$  Return the number l of evaluation steps needed to compute y = f(x).

INTERRUPT  $f \ x \ l \mapsto z$  Run the first l steps of the computation of f(x) and return a capsule z.

RESUME  $z \mapsto y$  If z = (INTERRUPT f x l), return y = f(x).

PRIMOPS  $f x \mapsto l$ 

Return the number l of evaluation steps needed to compute y = f(x).

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To compute 
$$(y, \dot{x}) = \mathcal{J} f x \dot{y}$$
:

**base case** (
$$f x \text{ fast}$$
):  $(y, \dot{x}) = \overleftarrow{\mathcal{J}} f x \dot{y}$  (step 0)

**inductive case**: 
$$h \circ g = f$$
 (step 1)

$$z = g x (step 2)$$

$$(y, \dot{z}) = \int h z \dot{y}$$
 (step 3)

$$(z, \dot{x}) = \int g \, x \, \dot{z}$$
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$$l = PRIMOPS f x$$
 (step 1)

$$z = \text{INTERRUPT } f x \left\lfloor \frac{l}{2} \right\rfloor$$
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$$(y, \dot{z}) = \mathcal{J}(\lambda z. \text{RESUME } z) z \dot{y}$$
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 (step 3)

$$(z, \dot{x}) = \mathcal{J}(\lambda x.INTERRUPT f x \lfloor \frac{1}{2} \rfloor) x \dot{z}$$
 (step 4)

#### **Example of CPS Conversion**

```
\left.\begin{array}{c} \text{function } f(c, \ x) \\ \text{return } g(\text{function}(t1) \\ \text{return } h(\text{function}(t2) \\ \text{return } p(\text{function}(t3) \\ \text{return } q(c, \ t3) \\ \text{end}, \ t1, \ t2) \\ \\ \text{end}, \ x) \\ \end{array}\right\} \sim \left\{\begin{array}{c} \text{function } f(c, \ x) \\ \text{return } g(\text{function}(t1) \\ \text{return } p(\text{function}(t3) \\ \text{return } q(c, \ t3) \\ \text{end}, \ x) \\ \\ \text{end}, \ x) \\ \\ \text{end} \end{array}\right.
```

• Convert source program to CPS.

- Convert source program to CPS.
- Thread step count and limit.

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$$[x|k] \rightsquigarrow k x$$

$$[(\lambda x.e)|k] \rightsquigarrow k (\lambda k' x.[e|k'])$$

$$[(e_1 e_2)|k] \rightsquigarrow [e_1|(\lambda x_1.[e_2|(\lambda x_2.(x_1 k x_2))])]$$

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#### Implementation

- Onvert source program to CPS.
- Thread step count and limit.
- Translate CPS to C.
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$$[x|k,n] \Rightarrow k(n+1) x$$

$$[(\lambda x.e)|k,n] \Rightarrow k(n+1) (\lambda k n x.[e|k,n])$$

$$[(e_1 e_2)|k,n] \Rightarrow [e_1|(\lambda n x_1.$$

$$[e_2|(\lambda n x_2.$$

$$(x_1 k n x_2)),$$

$$n ]),$$

$$(n+1) ]$$

$$[x|k, n, l] \rightarrow k (n+1) l x$$

$$[(\lambda x.e)|k, n, l] \rightarrow k (n+1) l (\lambda k n l x.[e|k, n, l])$$

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$$n, l],$$

$$(n+1), l]$$

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 $\langle\!\langle e \rangle\!\rangle_{k,n,l} \leadsto \mathbf{if} \ n = l \ \mathbf{then} \ [\![k, \lambda k \ n \ l \ \_.e]\!] \ \mathbf{else} \ e$ 

$$\lceil x|k, n, l \rceil \rightsquigarrow \langle \langle k (n+1) l x \rangle \rangle_{k,n,l}$$

$$\lceil (\lambda x.e)|k, n, l \rceil \rightsquigarrow \langle \langle k (n+1) l (\lambda k n l x. \lceil e | k, n, l \rceil) \rangle \rangle_{k,n,l}$$

$$\lceil (e_1 e_2)|k, n, l \rceil \rightsquigarrow \langle \langle \lceil e_1 | (\lambda n l x_1. \\ \lceil e_2 | (\lambda n l x_2. \\ (x_1 k n l x_2)), \\ n, l \rceil \rangle,$$

$$(n+1), l \rceil \rangle_{k,n,l}$$

 $\langle \langle e \rangle \rangle_{k,n,l} \rightsquigarrow \text{if } n = l \text{ then } [\![k, \lambda k \ n \ l \ \_.e]\!] \text{ else } e$ 

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$$[(e_1 e_2)|k, n, l] \rightsquigarrow \langle \langle [e_1|(\lambda n l x_1. \\ [e_2|(\lambda n l x_2. \\ (x_1 k n l x_2)), \\ n, l]),$$

$$(n+1), l] \rangle_{k,n,l}$$

$$\vdots$$

$$\langle \langle e \rangle_{k,n,l} \rightsquigarrow \text{if } n = l \text{ then } [k, \lambda k n l ...e] \text{ else } e$$

PRIMOPS 
$$f x = \mathcal{A} (\lambda n \ l \ v.n) \ 0 \infty f \ x$$
  
INTERRUPT  $f x \ l = \mathcal{A} (\lambda n \ l \ v.v) \ 0 \ l f \ x$   
RESUME  $[\![k,f]\!] = \mathcal{A} \ k \ 0 \infty f \ \bot$ 

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$$f$$
  $x = \mathcal{A}$  ( $\lambda n \ l \ v.n$ )  $0 \infty f \ x$   
INTERRUPT  $f$   $x \ l = \mathcal{A}$  ( $\lambda n \ l \ v.v$ )  $0 \ l f \ x$   
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  $x$  =  $\mathcal{A}$  ( $\lambda n \ l \ v.n$ )  $0 \infty f$   $x$ 

INTERRUPT  $f$   $x \ l$  =  $\mathcal{A}$  ( $\lambda n \ l \ v.v$ )  $0 \ lf$   $x$ 

RESUME  $[\![k,f]\!]$  =  $\mathcal{A}$   $k$   $0 \infty f$   $\bot$ 

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#### Implementation

- Oconvert source program to CPS.
- Thread step count and limit.
- Translate CPS to C.
- ① Combine with general-purpose interruption and resumption interface and  $\mathcal{J}$  written in C.
- Compile to machine code.

#### **Code Generation**

```
\mathcal{S}\pi () = null_constant \mathcal{S}\pi true = true_constant \mathcal{S}\pi false = false_constant \mathcal{S}\pi (c_1, c_2) = cons ((\mathcal{S}\pi c_1), (\mathcal{S}\pi c_1)) \mathcal{S}\pi n \mathcal{S}\pi 'k' = continuation \mathcal{S}\pi 'n' = count \mathcal{S}\pi '1' = limit \mathcal{S}\pi 'x' = argument \mathcal{S}\pi x = as_closure (target) ->environment [\pix]
```

#### **Code Generation**

```
S \pi (\lambda_3 n \, l \, x.e) = (\{
                 thing function(thing target,
                                   thing count,
                                   thing limit,
                                   thing argument) {
                 return (S(\phi e)e);
                 thing lambda = (thing)GC_malloc(sizeof(struct {
                   enum tag tag;
                   struct {
                      thing (*function)();
                      unsigned n;
                      thing environment [|\phi e|];
                   }))
                 set_closure(lambda);
                 as_closure(lambda) ->function = &function;
                 as_closure(lambda)->n = |\phi e|:
                 as_closure(lambda) ->environment[0] = S\pi(\phi e)_0
                 as_closure(lambda)->environment[|\phi e|-1] = \mathcal{S}\pi(\phi e)_{|\phi e|-1}
                 lambda:
              })
```

#### **Code Generation**

```
S\pi (\lambda_4 k \, n \, l \, x.e) = (\{
                   thing function(thing target,
                                     thing continuation,
                                     thing count,
                                     thing limit,
                                     thing argument) {
                   return (S(\phi e)e);
                   thing lambda = (thing)GC_malloc(sizeof(struct {
                     enum tag tag;
                     struct {
                        thing (*function)();
                        unsigned n;
                        thing environment [|\phi e|];
                     }))
                   set_closure(lambda);
                   as_closure(lambda) ->function = &function;
                   as_closure(lambda) ->n = |\phi e|;
                  as_closure(lambda) ->environment[0] = S\pi(\phi e)_0
                   as_closure(lambda) ->environment[|\phi e| - 1] = \mathcal{S} \pi (\phi e)_{|\phi e| - 1}
                   lambda;
                })
```

#### Code Generation

```
S\pi(e_1 e_2 e_3 e_4) = \text{continuation\_apply}((S\pi e_1),
                                                                                 (S \pi e_2),
                                                                                 (S \pi e_3),
                                                                                 (S \pi e_1)
         S\pi(e_1e_2e_3e_4e_5) = \text{converted\_apply}((S\pi e_1),
                                                                           (S \pi e_2),
                                                                           (S \pi e_3),
                                                                           (S \pi e_A).
                                                                           (S \pi e_5)
S \pi (if e_1 then e_2 else e_3) = (!is_false((S \pi e_1))?(S \pi e_2):(S \pi e_3))
                        S \pi (\diamond e) = (\mathcal{N} \diamond) ((S \pi e))
                  S \pi (e_1 \bullet e_2) = (\mathcal{N} \bullet) ((S \pi e_1), (S \pi e_2))
            S\pi(\overrightarrow{J}e_1e_2e_3) = (\mathcal{N}\overrightarrow{J})((S\pi e_1), (S\pi e_2), (S\pi e_3))
            S\pi(\overleftarrow{\mathcal{J}}e_1e_2e_3) = (\mathcal{N}\overleftarrow{\mathcal{J}})((S\pi e_1), (S\pi e_2), (S\pi e_3))
            S \pi (\mathcal{T} e_1 e_2 e_3) = (\mathcal{N} \mathcal{T}) ((S \pi e_1), (S \pi e_2), (S \pi e_3))
```

#### Implementation

- Convert source program to CPS.
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# Implementation of the General-Purpose Interruption and Resumption Interface with Threaded CPS

PRIMOPS 
$$f x = \mathcal{A} (\lambda n \ l \ v.n) \ 0 \infty f \ x$$
  
INTERRUPT  $f x \ l = \mathcal{A} (\lambda n \ l \ v.v) \ 0 \ l f \ x$   
RESUME  $[\![k,f]\!] = \mathcal{A} \ k \ 0 \infty f \ \bot$ 

## Implementation of the General-Purpose Interruption and Resumption Interface with Threaded CPS

written in C

```
static thing lambda_expression_that_returns_x
(thing f, thing n, thing l, thing x) {
  return x;
static thing lambda_expression_that_returns_n
(thing f, thing n, thing l, thing x) {
  return n;
static thing lambda_expression_that_resumes
(thing f, thing continuation, thing n, thing l, thing x) {
  if (!is_interrupt(x)) internal_error();
  return converted_apply(as_interrupt(x)->closure,
                         as_interrupt(x)->continuation,
                         make_real(0.0),
                         1.
                         null constant);
```

# Implementation of the General-Purpose Interruption and Resumption Interface with Threaded CPS

```
static unsigned long primops(thing f, thing x) {
  thing result = converted apply(f,
                                  continuation that returns n.
                                  make real (0.0).
                                  make real (HUGE VAL).
                                  x);
 else if (is real(result)) return (unsigned long) as real(result);
static thing interrupt(thing f, thing x, thing l) {
 thing result = converted_apply(f,
                                  continuation_that_returns_x,
                                  make real(0.0).
                                  1,
                                  x);
  if (!is interrupt(result)) internal error();
  return result:
```

### Algorithm for Divide-and-Conquer Checkpointing

To compute 
$$(y, \dot{x}) = \mathcal{J} f x \dot{y}$$
:

**base case** (
$$f x \text{ fast}$$
):  $(y, \dot{x}) = \overleftarrow{\mathcal{J}} f x \dot{y}$  (step 0)

**inductive case**: 
$$h \circ g = f$$
 (step 1)

$$z = g x (step 2)$$

$$(y, \dot{z}) = \int h z \dot{y}$$
 (step 3)

$$(z, \dot{x}) = \int g \, x \, \dot{z}$$
 (step 4)

### Algorithm for Divide-and-Conquer Checkpointing

written in C

```
static thing checkpoint starj(thing f, thing x, thing y cotangent)
 thing loop(thing f, thing x, thing y cotangent, unsigned long 1) {
    if (1<=base case duration) return ternary star; (f, x, y cotangent);
   else {
      thing u = interrupt(f, x, make real(1/2));
      thing v u cotangent = loop(closure that resumes, u, v cotangent, 1-1/2);
      if (!is_pair(y_u_cotangent)) internal_error();
     thing u x cotangent =
        loop (make closure for interrupt (f, 1/2),
             х,
             as_pair(y_u_cotangent)->cdr,
             1/2);
      if (!is_pair(u_x_cotangent)) internal_error();
      return cons(as pair(v u cotangent)->car,
                  as pair(u x cotangent) -> cdr);
 return loop(f, x, y cotangent, primops(f, x));
```

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#### **Determinant Example**

```
(define (car (cons car cdr)) car)
(define (cdr (cons car cdr)) cdr)
(define (matrix-rows a)
(if (null? a) 0 (+ (matrix-rows (cdr a)) 1)))
(define (list-ref l i)
(if (zero? i) (car l) (list-ref (cdr l) (- i 1))))
(define (matrix-ref a i i) (list-ref (list-ref a i) i))
(define (list-set l i x)
(if (zero? i)
    (cons x (cdr 1))
    (cons (car 1) (list-set (cdr 1) (- i 1) x))))
(define (matrix-set a i i x)
(list-set a i (list-set (list-ref a i) j x)))
(define (map-n f n)
(if (zero? n) '() (cons (f (- n 1)) (map-n f (- n 1)))))
(define (identity-matrix n)
(map-n (lambda (i) (map-n (lambda (j) (if (= i j) 1 0)) n)) n))
```

#### Determinant Example

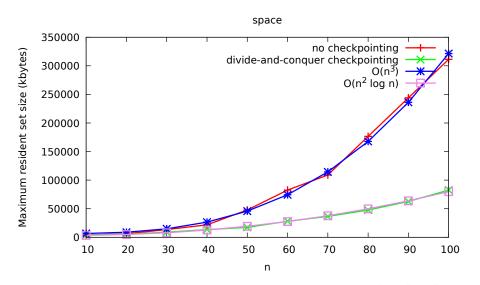
```
(define (determinant a)
(let ((n (matrix-rows a)))
  (let loop ((i 0) (b a) (d 1))
  (if (= i n)
       d
       (let* ((c (matrix-ref b i i))
              (b (let loop ((j i) (b b))
                  (if (= j n)
                      b
                      (loop (+ j 1) (matrix-set b i j (/ (matrix-ref b i j) c))))))
        (loop (+ i 1)
              (let loop ((j (+ i 1)) (b b))
               (if (= j n)
                   h
                   (loop (+ j 1)
                          (let ((e (matrix-ref b j i)))
                           (let loop ((k (+ i 1)) (b b))
                           (if (= k n)
                               h
                                (loop (+ k 1)
                                      (matrix-set b j k
                                                  (- (matrix-ref b i k)
                                                      (* e (matrix-ref b i k))))))))))
              (* d c)))))))
```

(write-real (determinant (cdr (checkpoint-\*j determinant (identity-matrix (read-real)) 1))))

#### Complexity Analysis

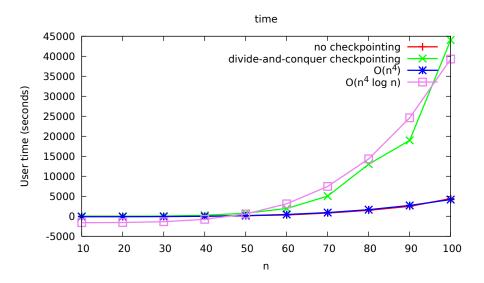
	space	time
primal	$O(n^2)$	$O(n^4)$
no checkpointing	$O(n^3)$	$O(n^4)$
divide-and-conquer checkpointing	$O(n^2 \log n)$	$O(n^4 \log n)$

#### Space Usage of the Determinant Example



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#### Time Usage of the Determinant Example



Interpreter using CPS evaluator

- Interpreter using CPS evaluator
- Hybrid compiler/interpreter using CPS conversion followed by direct-style evaluator

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All three support exact same source language. No exceptions. Same space and time complexity. Differ only in constant factors.

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Will release when manuscript is accepted.



#### Divide-and-Conquer checkpointing

• is traditionally formulated around loop iterations

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metaphor: a CPU is an instruction-execution loop

### Thank You