

Maintaining parallelism in reverse-mode automatic differentiation on functional parallel array languages

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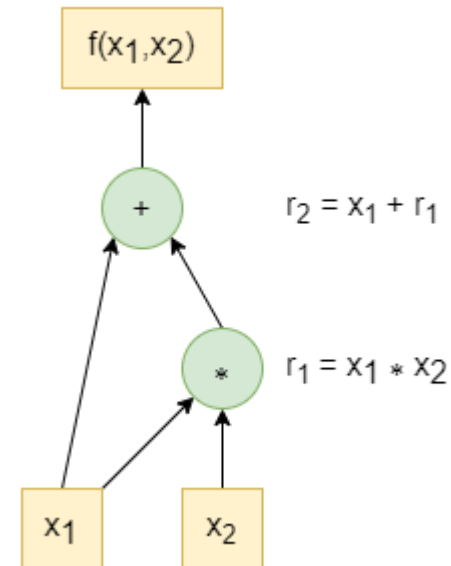
The Main Idea

- Tracing
 - How does the program work?
- Automatic Differentiation
 - Finding the derivative of the program
- Parallelism
 - Find the derivative *quickly*

Tracing

Tracing

- Single-Line program
- Function Elimination
- Generalized Type Elimination



Tracing

- Generalized Type Elimination
- Trace Consistency
- Trace Minimalism

$$\begin{aligned} &\forall s \in S \\ &\forall i \in I \\ &\text{trace}(s, i) = (t \in T, o \in O) \\ &(t \in S \cap T) \rightarrow (\text{eval}(s, i) = \text{eval}(t, i) = o) \end{aligned}$$

$$\begin{aligned} &\forall s \in S \\ &\forall i \in I \\ &\text{trace}(s, i) = (t \in T, o \in O) \\ &\text{trace}(t, i) = (t, o) \end{aligned}$$

Tracing

Function Tracing example

- Insert tracing into lambda expression
- Rediscover on application

Tracing Arrays

- Instantiation
- Map
- Fold

```
1 trace :: TEnvironment -> Expression -> (Value, Trace)
2 trace n (EApply e1 e2) =
3   -- First trace e1 and e2
4   let (v1, t1) = trace n e1
5       (v2, t2) = trace n e2
6   -- Check if v1 actually returns a function
7   in case v1 of
8     -- Do the application, return the result and the combined trace
9     TFunc f -> let (vf, tf) = f v2
10                in (vf, tf ++ t2 ++ t1)
11    _         -> error "Type mismatch in trace/EApply"
12
13 trace n (ELambda s e1) =
14   -- Define the function, insert value x as variable s into the environment that is currently
15   -- present, and trace the body
16   let f = TFunc (\x -> trace (insert s x) e1)
17   -- Return the function as abstracted function as a value, and no trace
18   in (f, [])
```

Automatic Differentiation

Automatic Differentiation

By hand

Finite Differencing

Automatic Differentiation

- Forward Mode
- Backwards/Reverse Mode

```
x1 = 15
dx1 = 1
x2 = 7
dx2 = 0
r1 = x1 + x2
dr1 = dx1 + dx2
y = r1 × x2
dy = r1 × dx2 + dr1 × x2
```

Listing 1: An example of forward mode AD by source transformation, with the AD statements in red

```
x1 = 15
x2 = 7
r1 = x1 + x2
y = r1 × x2

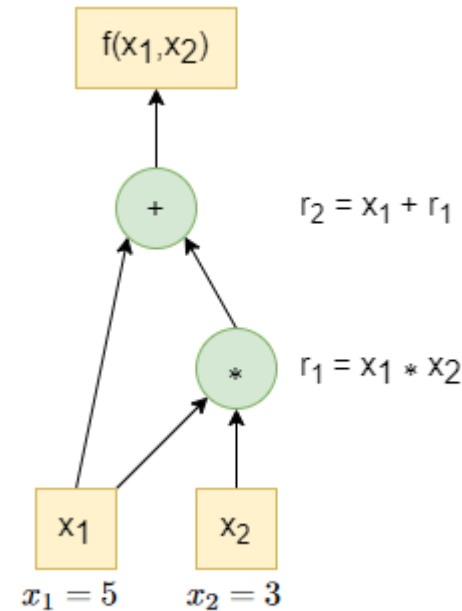
dy = 1
dr1 = dy × x2
dx2 = dy × r1 + dr1 × 1
dx1 = dr1 × 1
```

Listing 2: An example of reverse mode AD by source transformation, with the AD statements in red

Automatic Differentiation

Forward Pass

- Operations performed
- Reference counters
- Intermediate values



```

1 f :: Value -> Value -> Expression
2 f x1 x2 = ELet "x1" (ELift x1) (
3   ELet "x2" (ELift x2) (
4     EOp2 Add (ERef "x1") (
5       EOp2 Mul (ERef "x1") (ERef "x2")
6     )
7   ))
8
9 trace_result :: (TValue, Trace)
10 trace_result = (TReal "r2" 20.0, [
11   ("x1", TLift (TReal "x1" 5.0)),
12   ("x2", TLift (TReal "x2" 3.0)),
13   ("r1", TOp2 Mul "x1" "x2"),
14   ("r2", TOp2 Add "x1" "r1")
15 ])

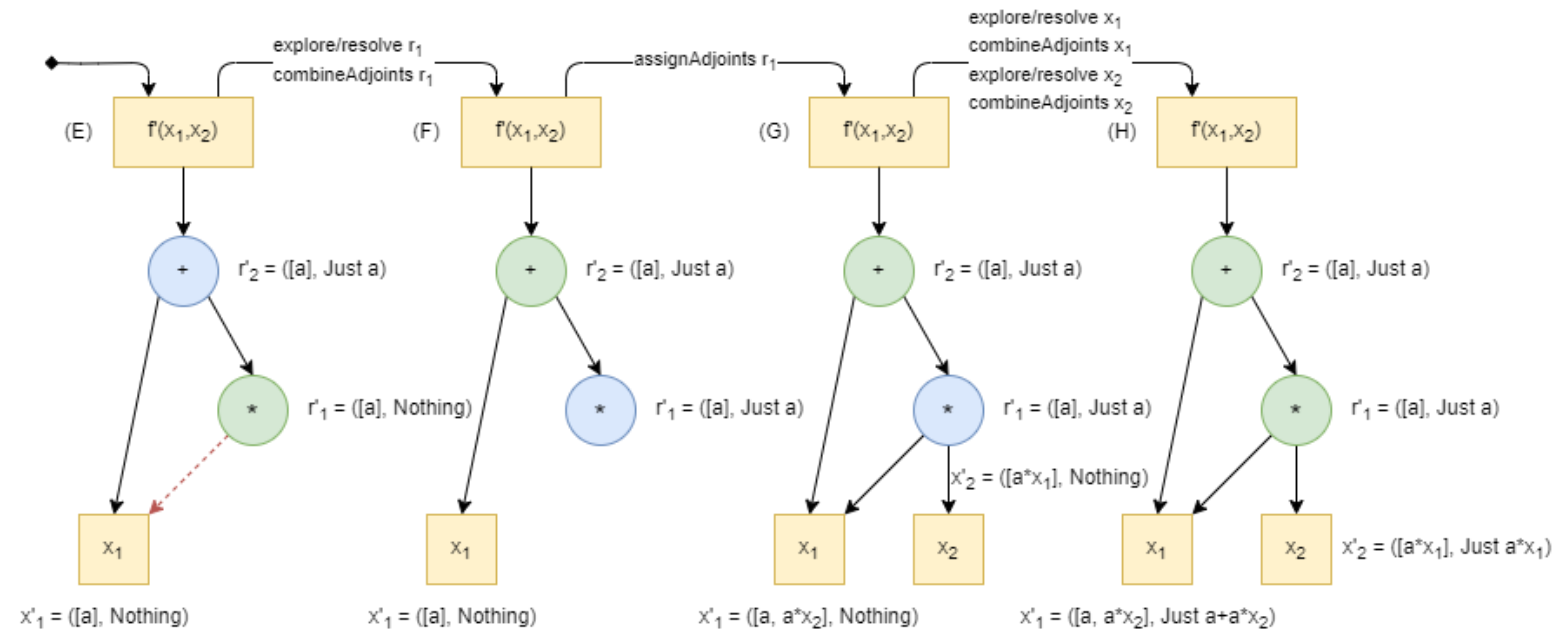
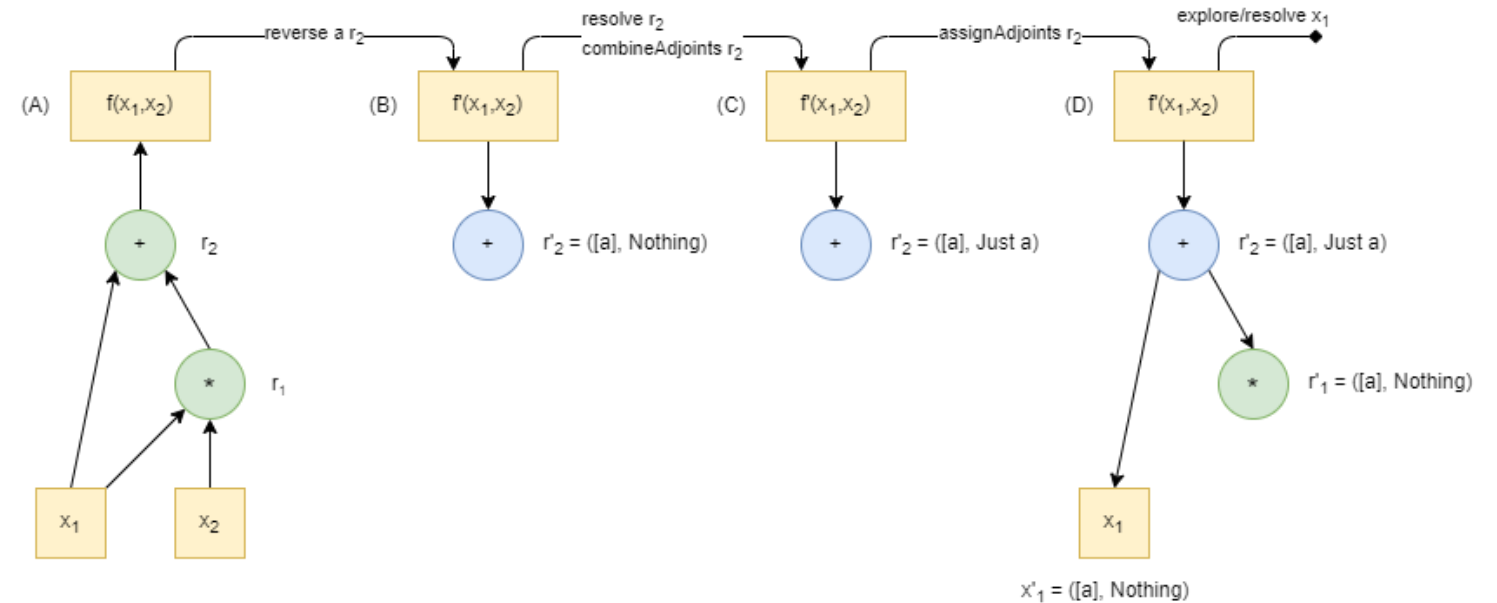
```

Listing 21: DSL definition of f and its trace

Automatic Differentiation

Reverse Pass

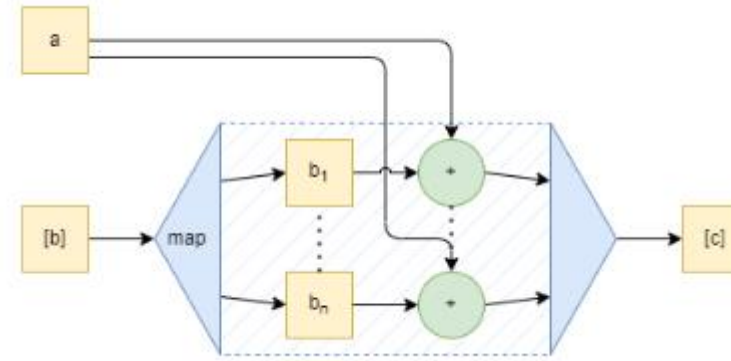
- Resolve node
- Combine adjoints
- Assign adjoints to ancestors
- Explore and resolve ancestors



Automatic Differentiation

Reverse Pass on arrays

- Map
- Fold
- Closures



Parallelism

Parallelism

Data Parallelism

Task Parallelism

Distributed/Machine Parallelism

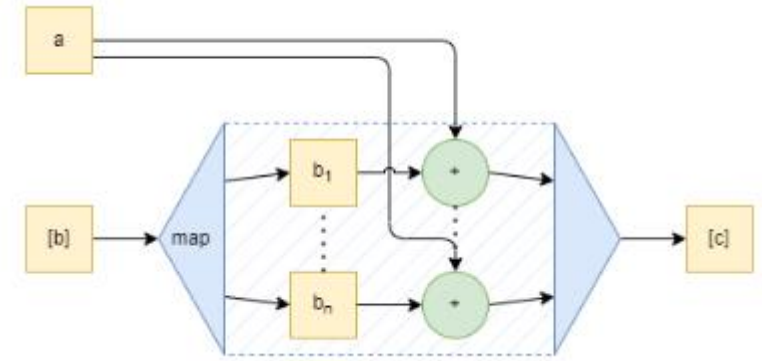
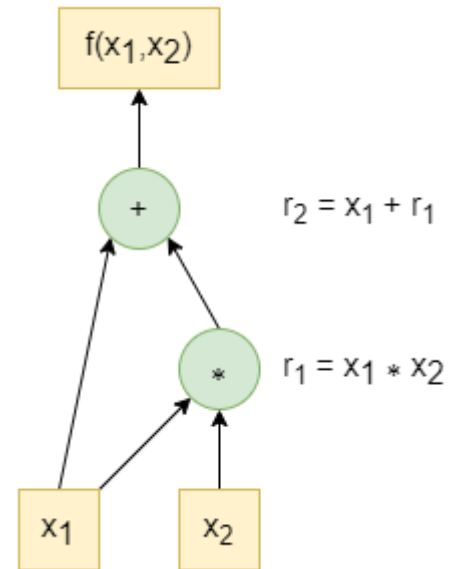
Parallelism

Task parallelism

- Concurrency
- Naming

Data parallelism

- Vectorization
- Segmentation



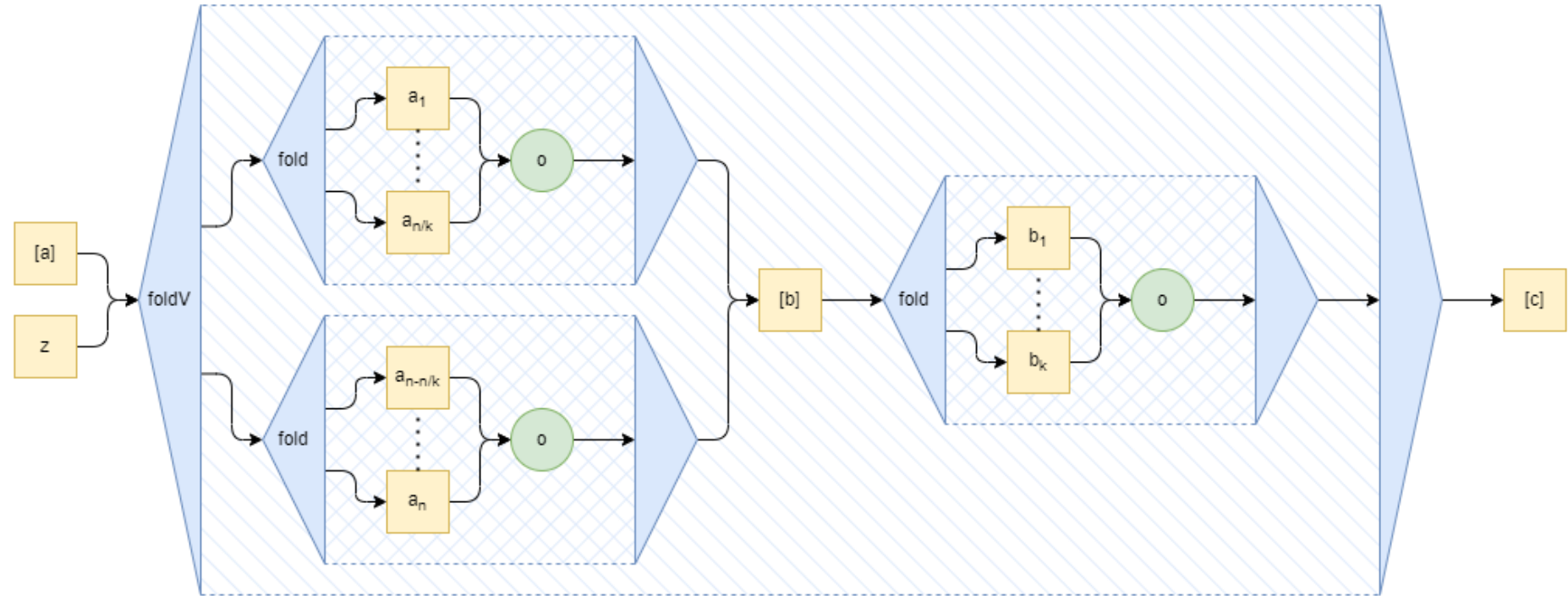
Parallelism

Task parallelism

- Concurrency
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Data parallelism

- Vectorization
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Conclusion

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Tracing

- Formal Definition, Consistency, Minimalism

Automatic Differentiation

- Reference counting, intermediate values

Parallelism

- Task and data parallelism

That's it!

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**Utrecht
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shaping tomorrow