

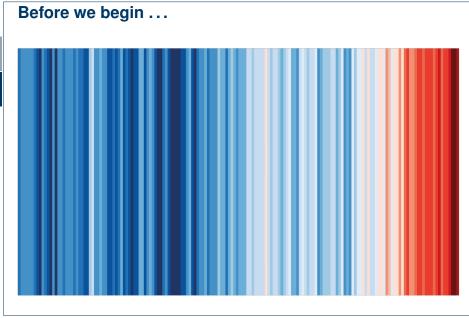
FACULTY OF ENGINEERING

# Lazy, Parallel Multiple Value Reductions in Common Lisp

Marco Heisig Chair for System Simulation, FAU Erlangen-Nürnberg 01.04.2019









### **Table of contents**

### 1. Motivation

### 2. The Function $\boldsymbol{\beta}$

### 3. Implementation



FACULTY OF ENGINEERING

# Motivation





### **Reductions are Awesome!**

```
(defun fold (f z l)
  (if (null l)
        z
        (fold f (funcall f (first l) z) (rest l))))
```

```
• sum
(fold #'+ 0 numbers)
```

maximum

(fold #'max 0 non-negative-numbers)

reversal

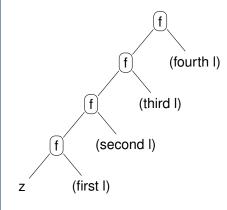
(fold #'cons '() list)

filtering

(fold (lambda (i j) (if (oddp i) (cons i j) j)) '() list)



### Problem #1 - Parallelism



- Long serial chain of dependencies.
- Execution time will always be time(f) · length(l).
- Exascale computers expected in 2021.

"foldl and foldr Considered Slightly Harmful" – Guy Steele



### Problem #2 - Multiple values

• Life, as it should be:

(reduce #'fn values :initial-value iv)

Life, as it is:

```
(loop for value in values
  for elt across aux
  for idx from 0
  for acc-1 = (fn-1 value elt idx)
  for acc-2 = (fn-2 acc-1 elt idx)
  finally (return (values acc-1 acc-2)))
```

Goal: Reductions on multiple streams of data at once.



# The Goals

### Parallelism

 $O(\log(N))$  runtime on a sufficiently parallel machine.

#### Multiple values

Gather multiple quantities from multiple sources.

### Laziness

Programmers should not have to cripple their source code to avoid allocation of intermediate data.

### Array Programming

Support for multi-dimensional arrays.

### Performance

Competitive to a good cl:reduce.



### **Context: The Petalisp Project**

- A Common Lisp library for elegant parallel programming.
- The core data structures are lazy, strided arrays.
- All operations are deterministic and purely functional.
- Petalisp has only four core operators. Parallel reduction is one of them.
- Arrays are evaluated by calling compute.

### Interested?

(ql:quickload :petalisp)

https://github.com/marcoheisig/Petalisp

/join #petalisp



FACULTY OF ENGINEERING

# The Function $\beta$





# Definition

(defun  $\beta$  (f array &rest more-arrays) ...)

- *f* must accept 2*k* arguments and return *k* values, where *k* is the number of supplied arrays.
- All supplied arrays must have the same shape  $S = r_1 \times \ldots \times r_n$ , where each range  $r_k$  is a set of integers,  $\{0, 1, \ldots, m\}$ .
- Returns *k* arrays of shape  $s = r_2 \times ... \times r_n$ , whose elements are a combination of the elements along the first axis of each array.

It remains to clarify how we combine elements of the first axis.



### **The Reduction Rules**

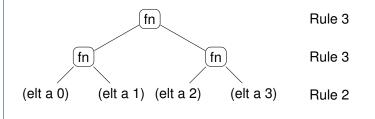
- k arrays of dimension n and shape S
- f is a function from 2k arguments to k values
- *n* 1 dimensional output shape *s*
- 1. If the given arrays are empty, signal an error.\*
- 2. If the first axis of each given array contains exactly one element, drop that axis and return the resulting *k* arrays of shape *s*.
- 3. Otherwise
  - Split each array into a lower and an upper half.
  - Recurse into each of the two halves.
  - Combine the 2k resulting arrays of shape s element-wise with f.
  - Return the resulting *k* arrays of shape *s*.



### A Simple Example

#### **Example:** ( $\beta$ #'fn (vector a b c d))

- The number of arrays k is 1.
- The input shape *S* is ({0, 1, 2, 3}).
- The output shape s is (), i.e. the result is a scalar.





# Parallelism

	(reduce #'+ array)	(β <b>#'</b> + array)
Number of Additions	<i>N</i> – 1	<i>N</i> – 1
Dependency Tree Depth	<i>N</i> – 1	$\lceil \log_2(N) \rceil$

 $\Rightarrow$  The function  $\beta$  is well suited for parallel computing.



### **Multiple Values**

Computing both the maximum element and its index:

Look Ma, no loops!



# **Multiple Values and Multiple Dimensions**

Computing both the maximum element and its index:

... works for multi-dimensional arrays, too!



FACULTY OF ENGINEERING

# Implementation





# Implementing $\beta$ is Hard

The function  $\beta$  has many degrees of freedom:

- The number of supplied arrays k.
- The rank of the supplied arrays d.
- The element type of each supplied array.

And this is without taking lazy evaluation into account!

Our reference implementation is terribly slow, with gems like

```
(values-list
  (subseq
    (multiple-value-list
        (multiple-value-call f
        (divide-and-conquer ls le)
        (divide-and-conquer us ue)))
    0 k))
```



# Making $\beta$ Fast

- For classical sequence functions, it is common to define multiple specialized versions.
- We cannot use this trick, because we'd require *DE<sup>k</sup>* versions, where *D* is the supported number of dimensions and *E* is the number of specialized array element types.

What we do instead:

- Compute a normalized problem description.
- Turn this problem description into efficient Lisp code.
- Use cl:compile to generate a fast evaluator.
- Invoke the compiled function on the supplied arrays.
- Cache the compiled function, using the normalized problem description as key.

Result: ( $\beta$  #'+ v) can actually inline #'+!



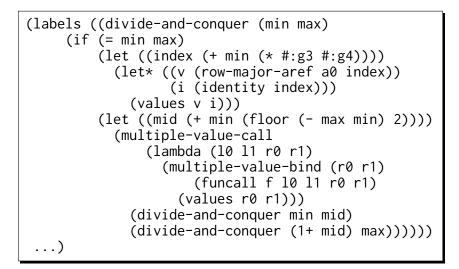
# The Petalisp JIT-Compiler

Each Petalisp evaluation consists of the following steps:

- 1. Broadcasting, Type Inference, Shape Checking
- 2. Data-flow Optimization
- 3. IR-Conversion
- 4. Normalization
- 5. Scheduling
- 6. Code Generation
- 7. Compilation
- 8. Execution
  - Thanks to memoization, the steps 6. and 7. can usually be skipped.
  - The steps 5. and 8. can usually overlap.
  - The challenge is getting the steps 1. to 4. fast enough.
  - We are down to a few microseconds, but need to get better.



### **Optimization Showcase - a call to max\***





### **Future Challenges**

Challenges for the next months:

- Reduce the latency of compute.
- Add proper multi-threading.
- Further tweak the generated code.

Challenges for the next few years:

- Distributed Computing
- GPU offloading



