## Petalisp

A Common Lisp Library for Data Parallel Programming

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The library **Petalisp**<sup>1</sup> is a new approach to data parallel computing.

The Goal: Elegant High Performance Computing

- Programs that are beautiful and fast
- A programming model that is safe and productive

#### Drawbacks:

- Limited to operations on structured data
- Significant run-time overhead

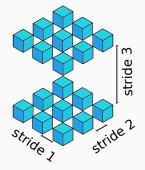
<sup>&</sup>lt;sup>1</sup>https://github.com/marcoheisig/Petalisp

- 1. Using Petalisp
- 2. Implementation
- 3. Performance
- 4. Conclusions

## **Using Petalisp**

A **strided array** in n dimensions is a function from elements of the cartesian product of nranges to a set of Common Lisp objects.

A range with the lower bound  $x_L$ , the step size sand the upper bound  $x_U$ , with  $x_L$ , s,  $x_U \in \mathbb{Z}$ , is the set of integers  $\{x \in \mathbb{Z} \mid x_L \le x \le x_U \land (\exists k \in \mathbb{Z}) [x = x_L + ks] \}.$ 

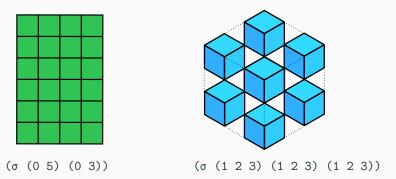


Objects of type cl:simple-array are a special case of strided arrays.

## API 1/5 — First Class Index Spaces

To introduce parallelism, Petalisp always operates on index spaces, not on individual array elements.

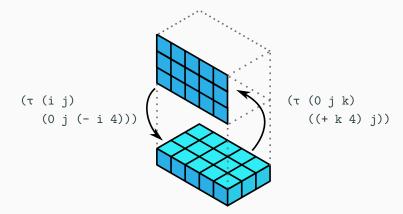
Notation: ( $\sigma$  (START [STEP] END) ...)



**Implementation detail:** Petalisp can compute the union, difference and intersection of arbitrary index spaces.

## API 2/5 — Transformations

A transformation is an affine-linear mapping from indices to indices. Notation:  $(\tau (INDEX ...) (EXPRESSION ...))$ 



**Implementation detail:** Petalisp can compute the inverse and composition of arbitrary transformations.

The -> operator allows to select, transform or broadcast data.

(-> 0 ( $\sigma$  (0 9) (0 9))) ; a 10  $\times$  10 array of zeros

 $(-> #(2 3) (\sigma (0 0)))$ ; the first element only

(-> A ( $\tau$  (i j) (j i))) ; transposing A

Admittedly, -> is a mediocre function name. Better suggestions are most welcome!

The fuse and fuse\* operator combine multiple arrays into one.

For fuse, the arguments must be non-overlapping. For fuse\*, the value of the rightmost array takes precedence on overlap.

(defvar B (-> #(2) ( $\tau$  (i) ((1+ i)))))

(fuse #(1) B) ; equivalent to (-> #(1 2))

(fuse #(1 3) B) ; an error!

(fuse\* #(1 3) B) ; equivalent to (-> #(1 2))

The final piece: Application of Common Lisp functions to strided arrays.

- The function  $\alpha$  is basically cl:map for strided arrays.
- The function β is basically cl:reduce applied to the *last* dimension of a strided array.

(α #'+	23)	; adding two numbers
(α <b>#</b> '+	A B C)	; adding three arrays element-wise
(β <b>#</b> '+	#(2 3)	) ; adding two numbers
(defvar	B #2/	((1 2 3) (4 5 6)))
(β <b>#</b> '+	B)	; summing the rows of B

**Remark:** No guarantees are made about when and how often the functions passed to  $\alpha$  and  $\beta$  are invoked.

All core functions at a glance:

- Index spaces, e.g. ( $\sigma$  (a b))
- Transformations, e.g. ( $\tau$  (x y) (y (- x)))
- Data motion, e.g. (-> A ( $\sigma$  (2 5)))
- Data combination, e.g. (fuse\* A B C)
- Parallel map, e.g. (α #'\* A B)
- Parallel reduce, e.g. ( $\beta$  #'+ A)

This API is purely functional and declarative.

But how do we obtain values?

Petalisp provides two functions to trigger evaluation.

The compute function converts strided arrays into regular Common Lisp arrays.

 $(compute (-> 0.0 (\sigma (0 1)))) => #(0.0 0.0)$ 

(defvar A #(1 2 3))

 $(compute (\beta \#' + A)) \implies 6$ 

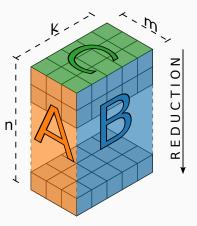
(compute (-> A ( $\tau$  (i) ((- i))))) => #(3 2 1)

Remark: There is also a schedule function for asynchronous evaluation.

The mathematical definition

$$C_{ij} = \sum_{p=1}^{n} A_{ip} B_{pj}$$

The corresponding Petalisp code

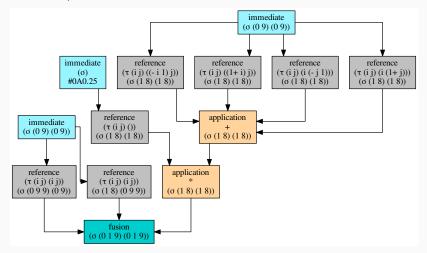


# Implementation

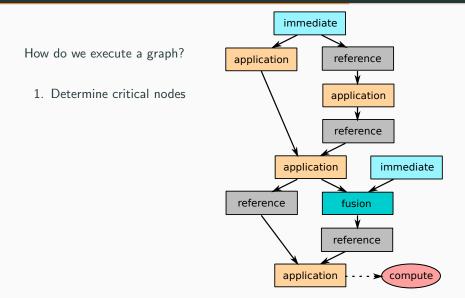
#### Lazy Arrays are Data Flow Graphs

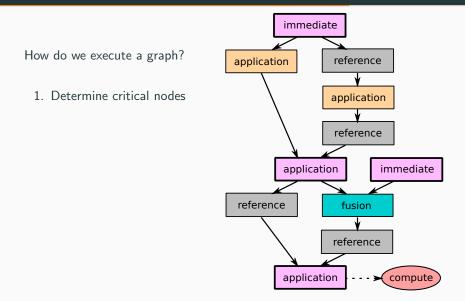
(jacobi u 1) => #<strided-array-fusion t ( $\sigma$  (0 9) (0 9))>

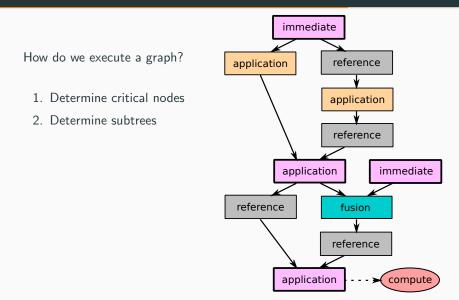
Internal representation:

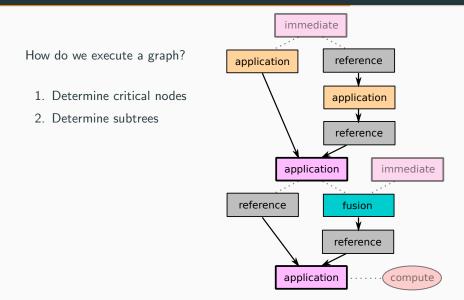


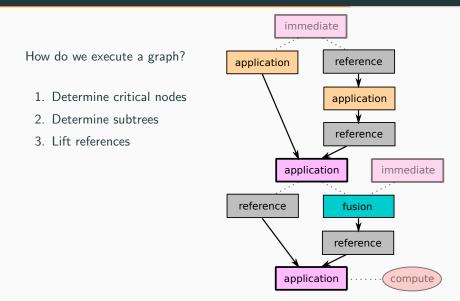
immediate How do we execute a graph? reference application application reference immediate application reference fusion reference application ---> compute

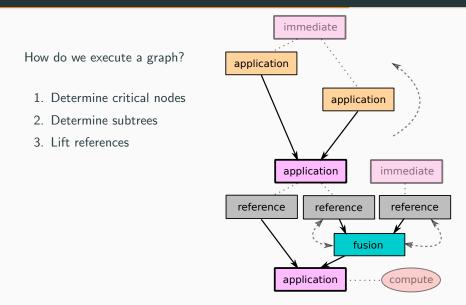


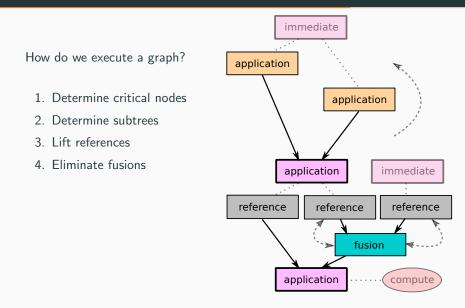


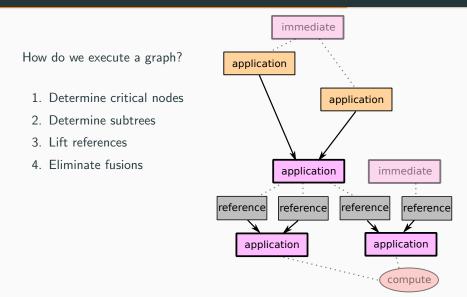


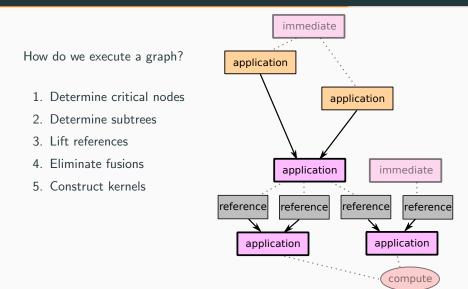


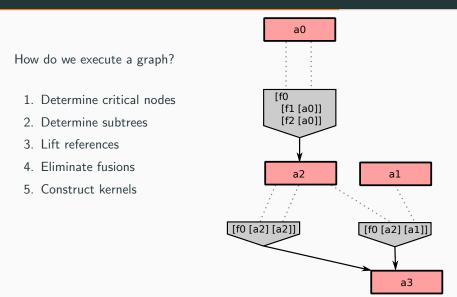


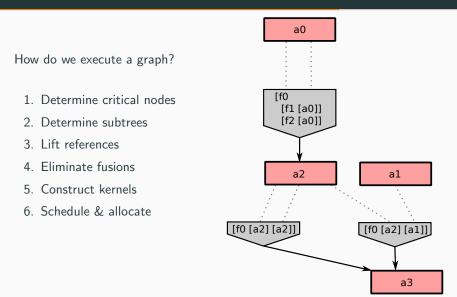






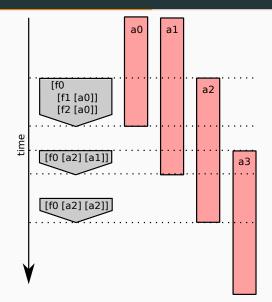






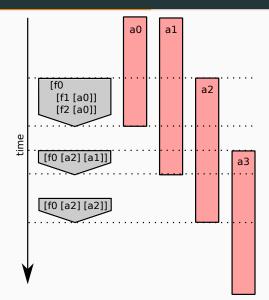
How do we execute a graph?

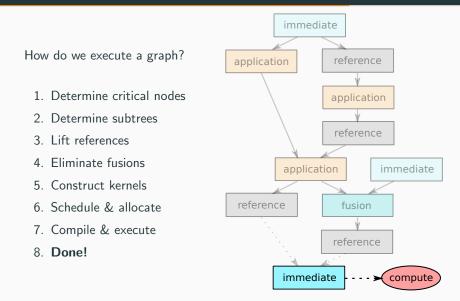
- 1. Determine critical nodes
- 2. Determine subtrees
- 3. Lift references
- 4. Eliminate fusions
- 5. Construct kernels
- 6. Schedule & allocate



How do we execute a graph?

- 1. Determine critical nodes
- 2. Determine subtrees
- 3. Lift references
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- 5. Construct kernels
- 6. Schedule & allocate
- 7. Compile & execute





Performance

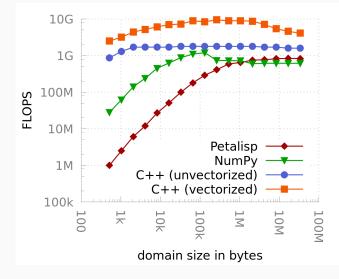
The long-term goal of Petalisp is to provide a programming model for Petascale ( $10^{15}$  operations per second) systems.

The constant, high overhead of analysis and JIT-compilation seems to be at odds with this goal.

#### However:

- Do not underestimate the power of memoization, hash-consing and CLOS wizardry.
- Scheduling can often be done asynchronously.
- Petalisp's analysis is independent of the problem size.

#### Jacobi's method: Python vs. C++ vs. Petalisp



Hardware: Intel Xeon E3-1275 CPU 3.6GHz

## Conclusions

**Main Result:** Our compilation strategy is feasible, with just about 10 – 500 microseconds overhead when calling compute.

#### **Benefits:**

- Clean separation between notation and execution.
- Unprecedented potential for optimization.
- Already faster than NumPy.

... all in just about 5000 lines of maintainable code.

My preliminary roadmap for the next years:

- More applications (simulations, image processing, machine learning)
- API finalization
- Sophisticated Scheduling
- Better Shared-Memory Parallelization
- Auto-Vectorization
- Distributed Parallelization
- ...
- Make this a PhD thesis

# Thank you!



# Questions or remarks?