



University
of Glasgow

Lift: Code Generation by Rewriting Algorithmic Skeletons

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and the LIFT team

**INSPIRING
PEOPLE**



1968 - Go To Statement Considered Harmful

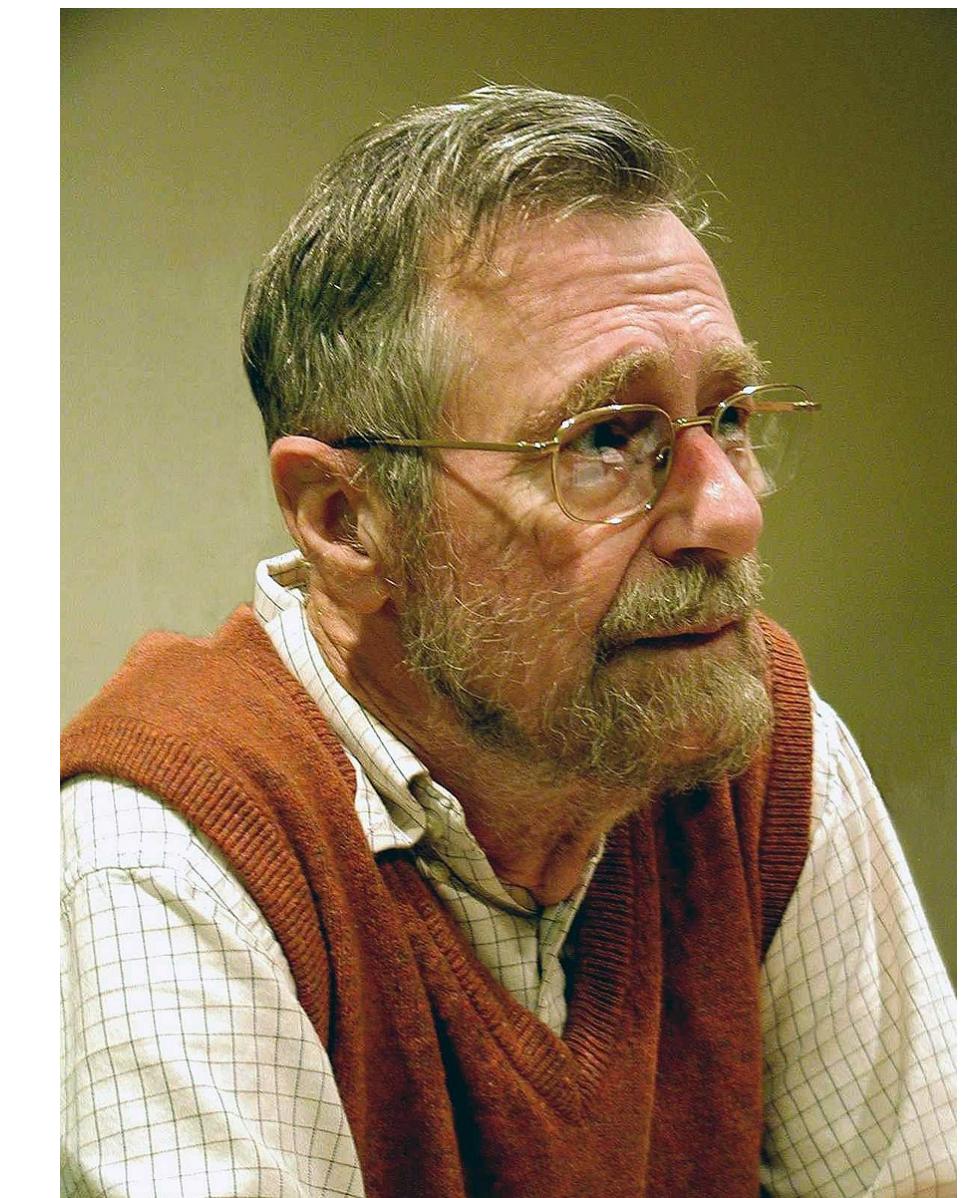
EWD215 - 0

A Case against the GO TO Statement.

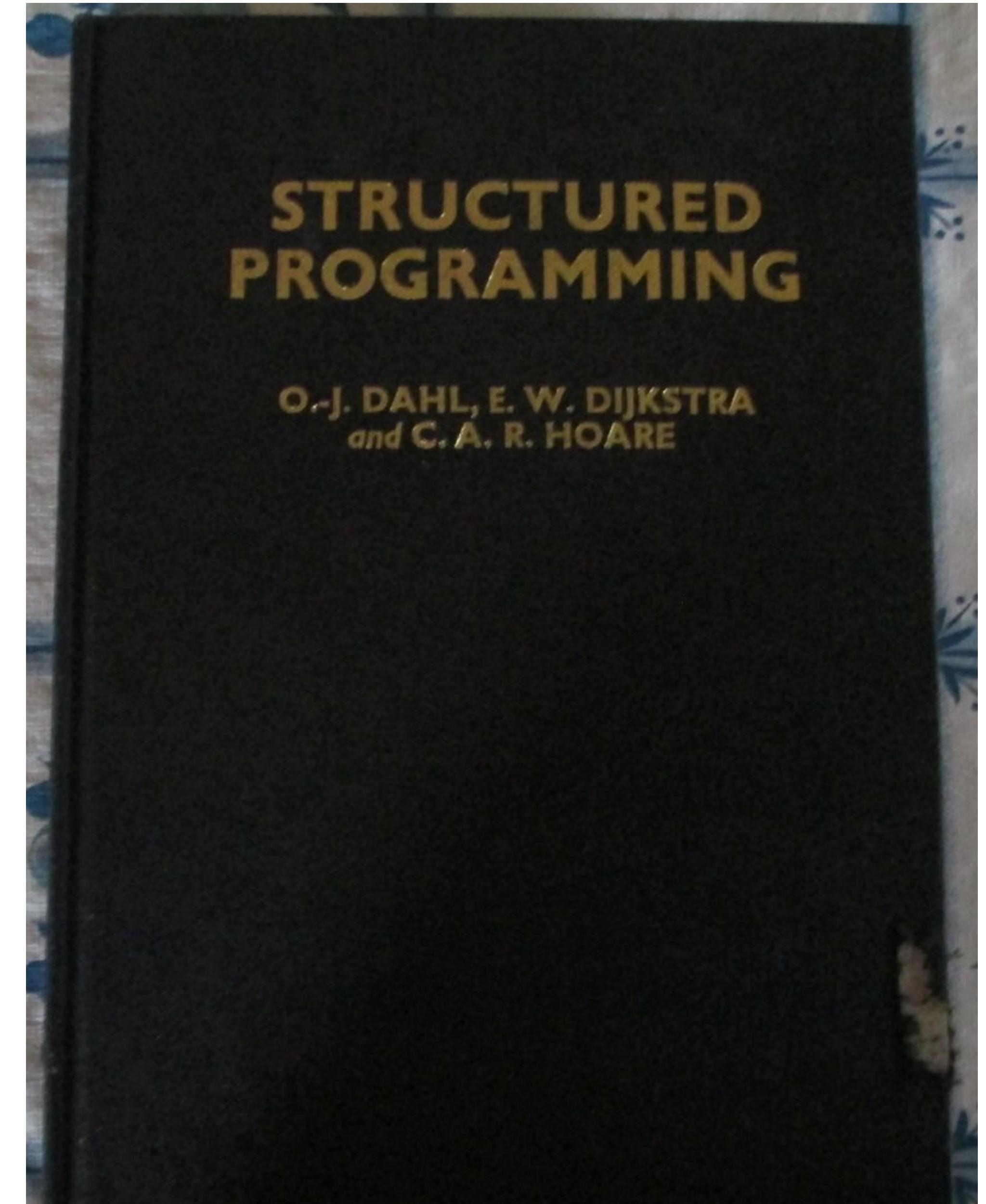
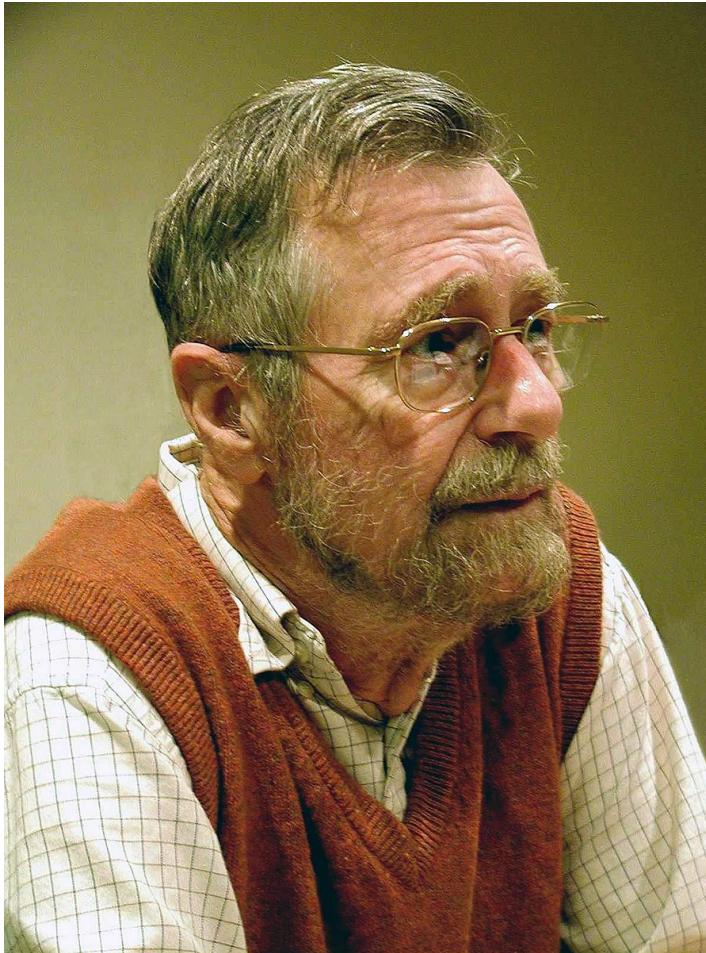
by Edsger W. Dijkstra
Technological University
Eindhoven, The Netherlands

Since a number of years I am familiar with the observation that the quality of programmers is a decreasing function of the density of go to statements in the programs they produce. Later I discovered why the use of the go to statement has such disastrous effects and did I become convinced that the go to statement should be abolished from all "higher level" programming languages (i.e. everything except -perhaps- plain machine code). At that time I did not attach too much importance to this discovery; I now submit my considerations for publication because in very recent discussions in which the subject turned up, I have been urged to do so.

My first remark is that, although the programmer's activity ends when he has constructed a correct program, the process taking place under control



1972 - Structured Programming



```
procedure matmult (A, B, C, m, n, p);
    array A, B, C; integer m, n, p;
begin integer i, j, k;
    for i := 1 step 1 until m do
    for j := 1 step 1 until n do
    begin C[i, j] := 0;
        for k := 1 step 1 until p do
            C[i, j] := C[i, j] + A[i, k] × B[k, j]
        end
    end;
```

1989 - Structured *Parallel* Programming



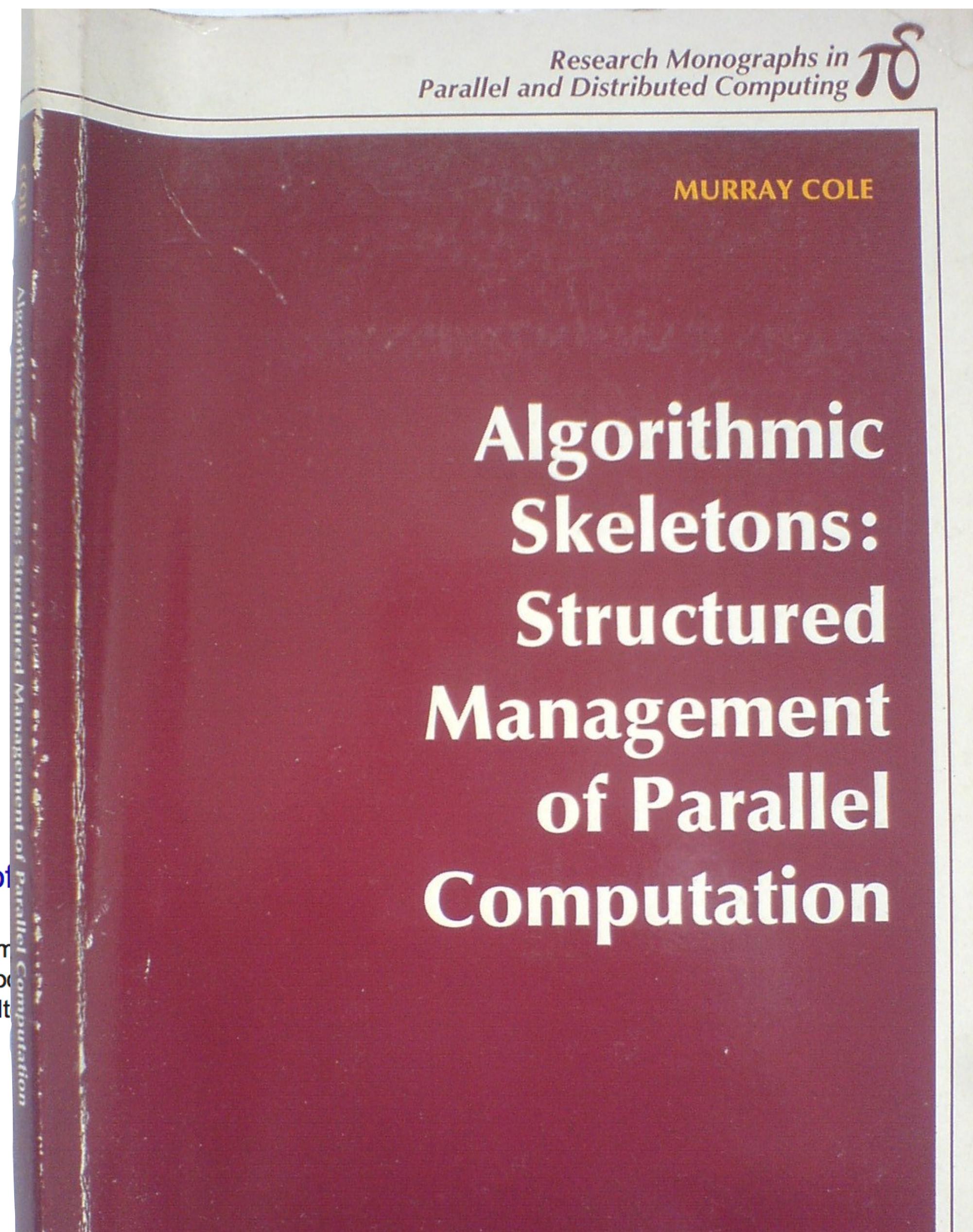
D_C indivisible split join $f = F$

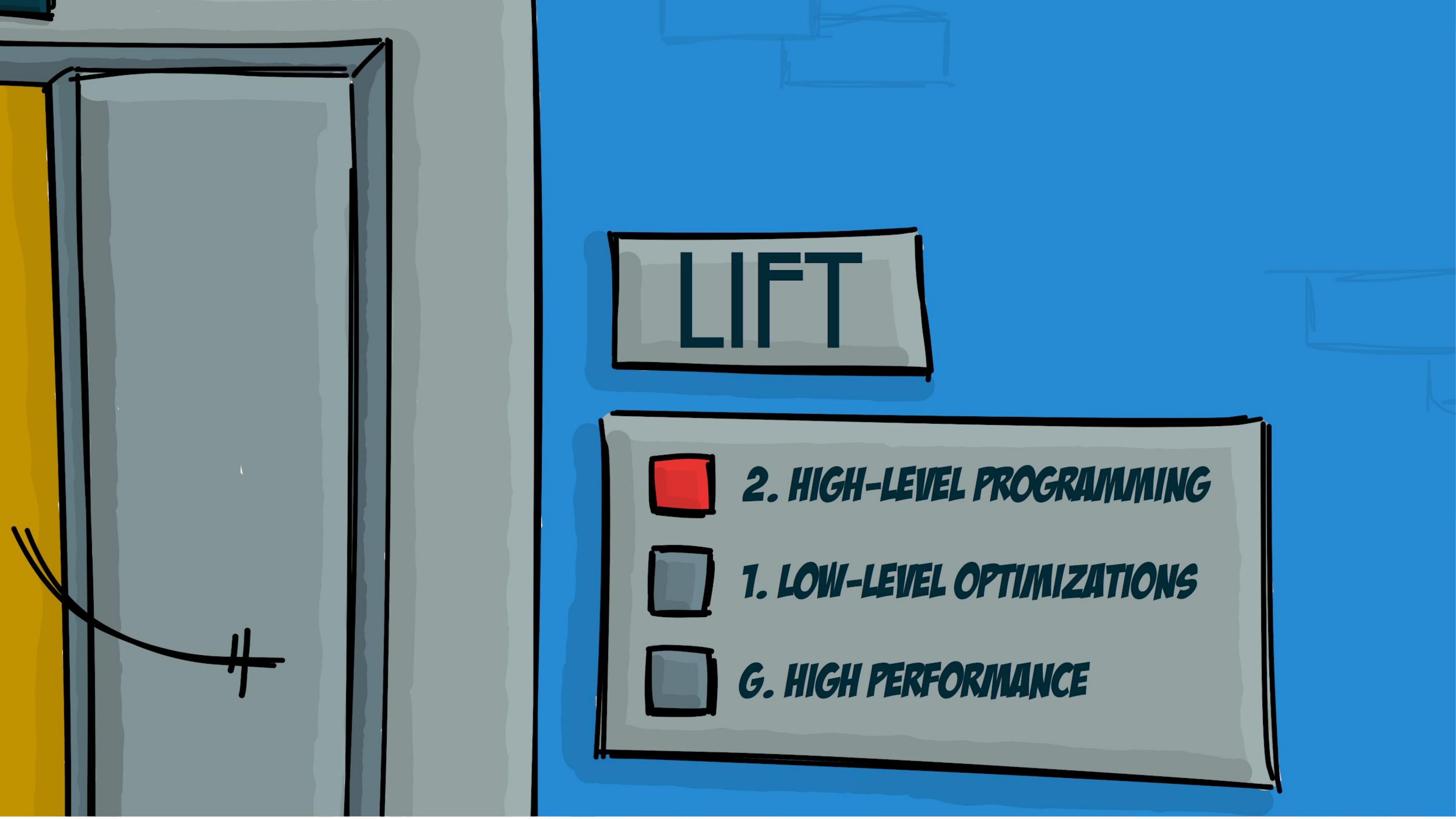
where $F P = \begin{cases} f P, & \text{if } \text{indivisible } P \\ \text{join}(\text{map } F (\text{split } P)), & \text{otherwise} \end{cases}$

[BOOK] Algorithmic skeletons: structured management of parallel computation
MI Cole - 1989 - homepages.inf.ed.ac.uk

Abstract In the past, most significant improvements in computer performance have been achieved as a result of advances in simple device technology. The introduction of large-scale parallelism at the inter-processor level now represents a viable alternative to sequential processing.

☆ 77 Cited by 1304 Related articles All 6 versions ☰





LIFT

2. HIGH-LEVEL PROGRAMMING

1. LOW-LEVEL OPTIMIZATIONS

G. HIGH PERFORMANCE



[ICFP'15]

DSL

DSL

DSL

High-Level IR

Explore Optimizations
by rewriting

[GPGPU'16]
[CASES'16]

Low-Level Program

Code Generation
[CGO'17]

Multicore
CPU

GPU

HPC
Mobile

Xeon
Phi

KNC
KNL

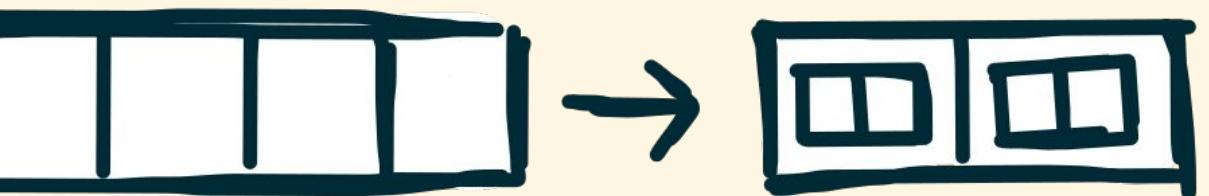
...

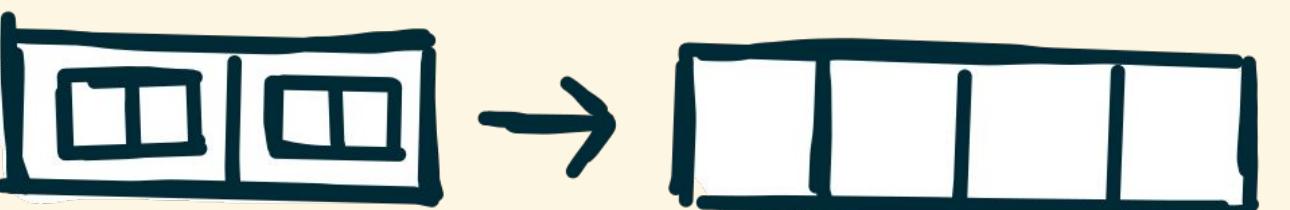
Hardware

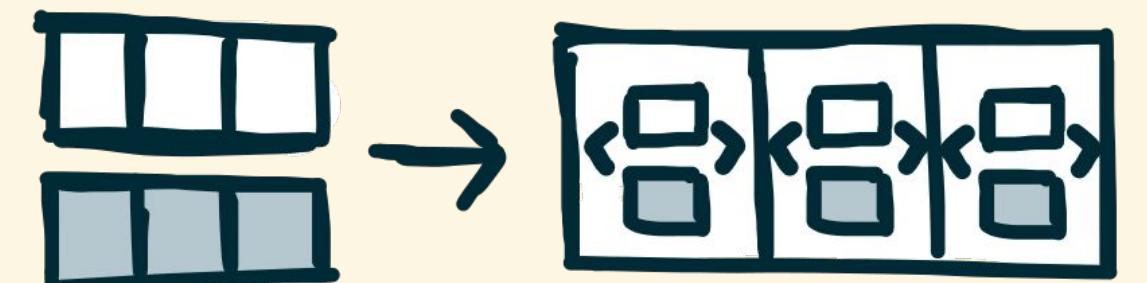
LIFT'S HIGH-LEVEL PRIMITIVES

map($\square \rightarrow \square$) 

reduce(\oplus) 

split(n) 

join 

zip 

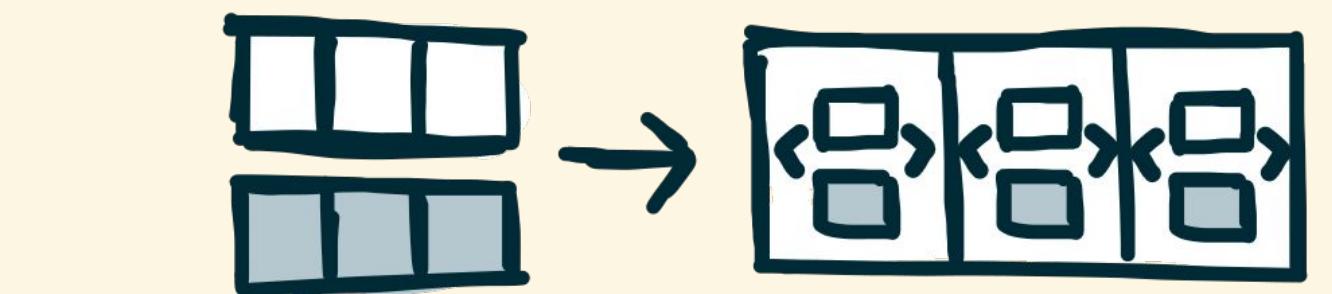
LIFT'S HIGH-LEVEL PRIMITIVES

map($\square \rightarrow \square$) 

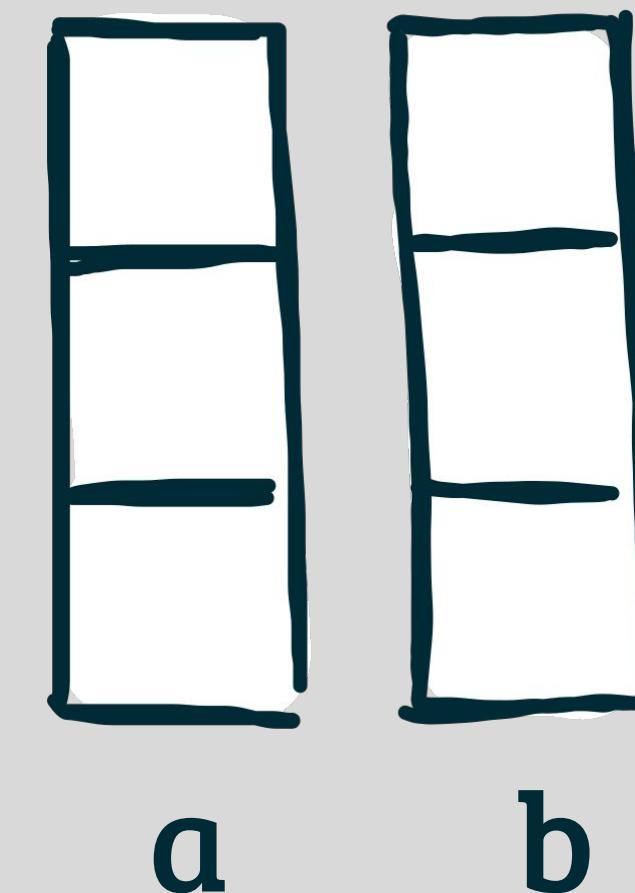
reduce(\oplus) 

split(n) 

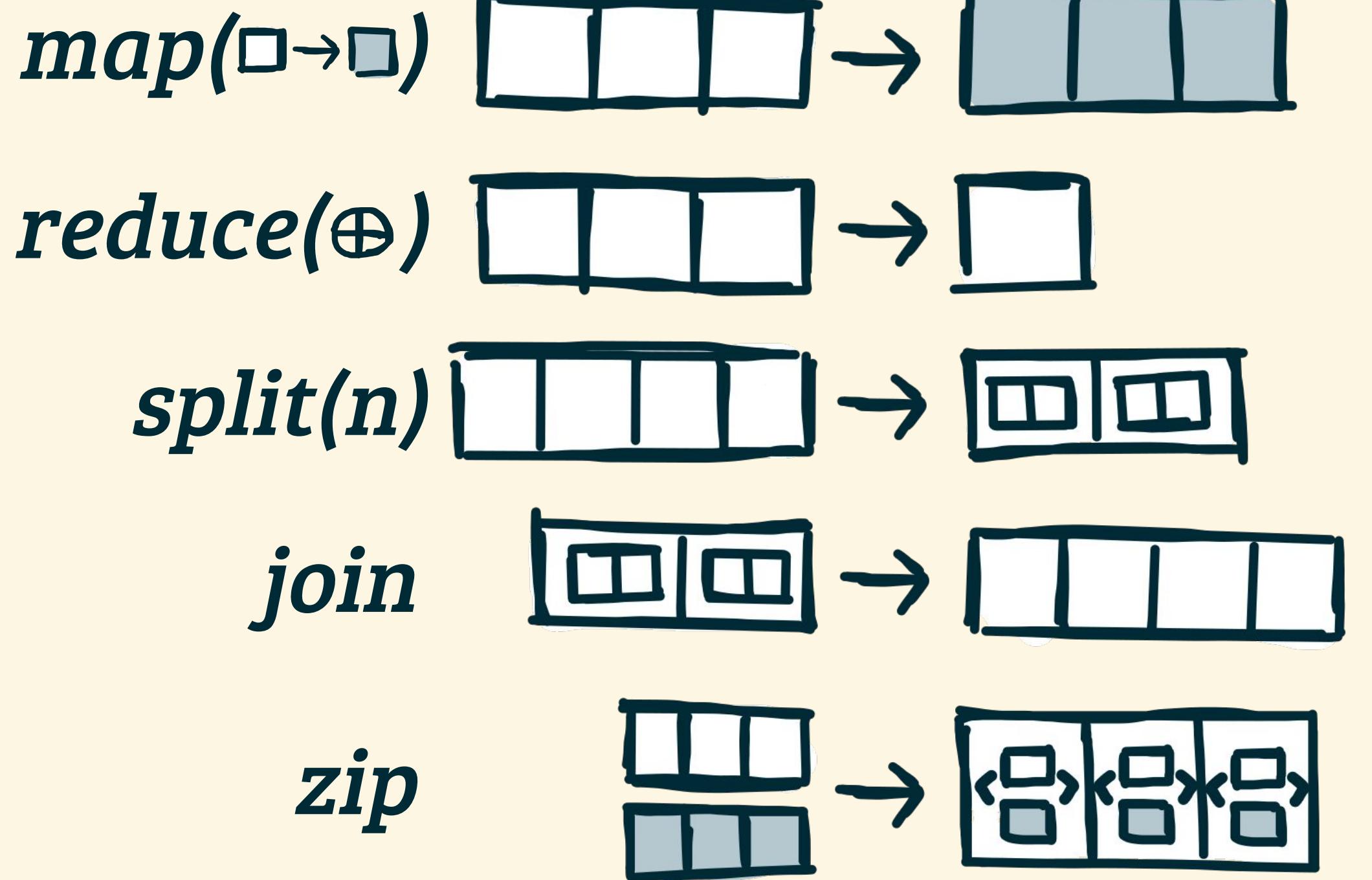
join 

zip 

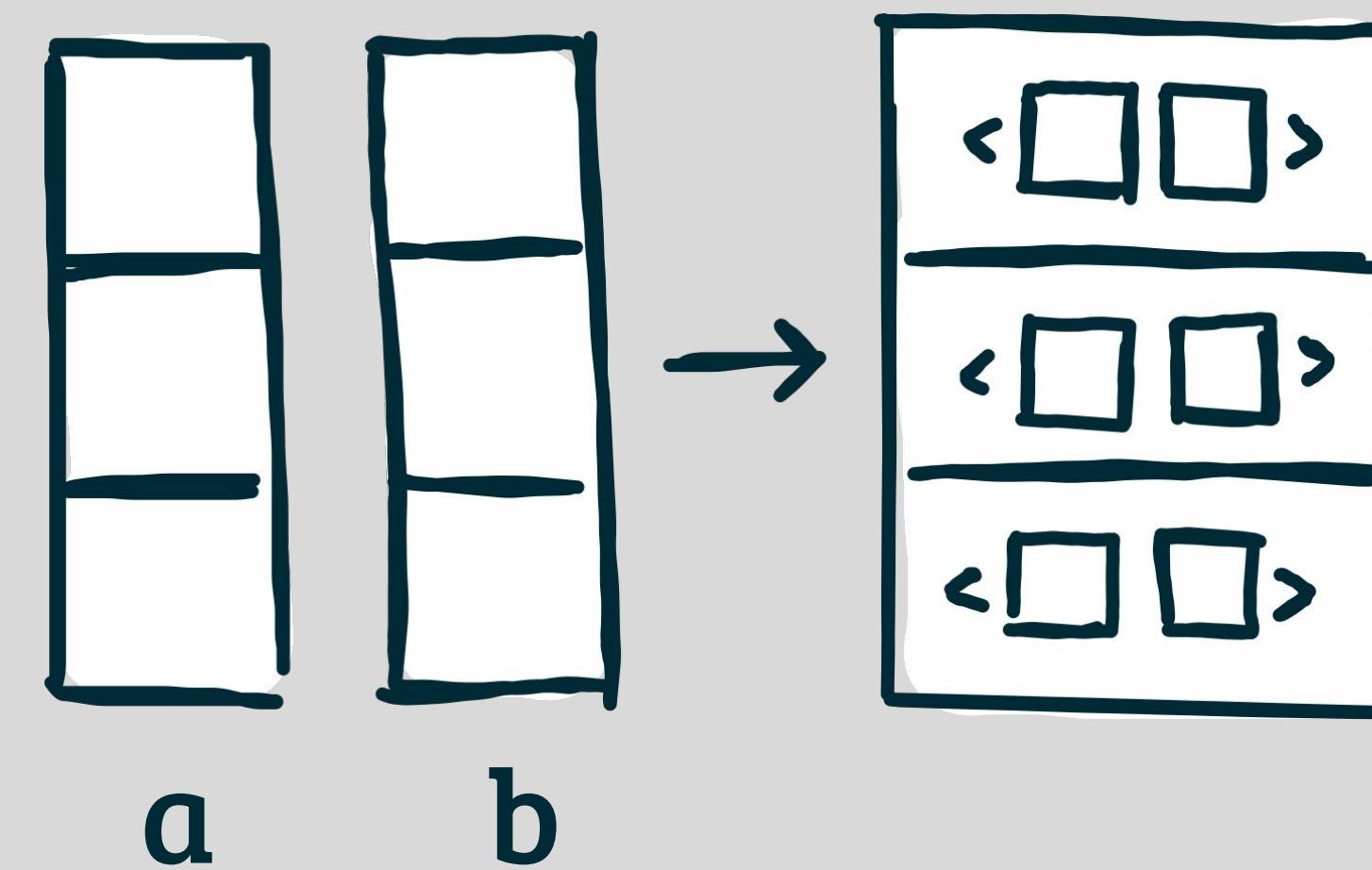
dotproduct.lift



LIFT'S HIGH-LEVEL PRIMITIVES

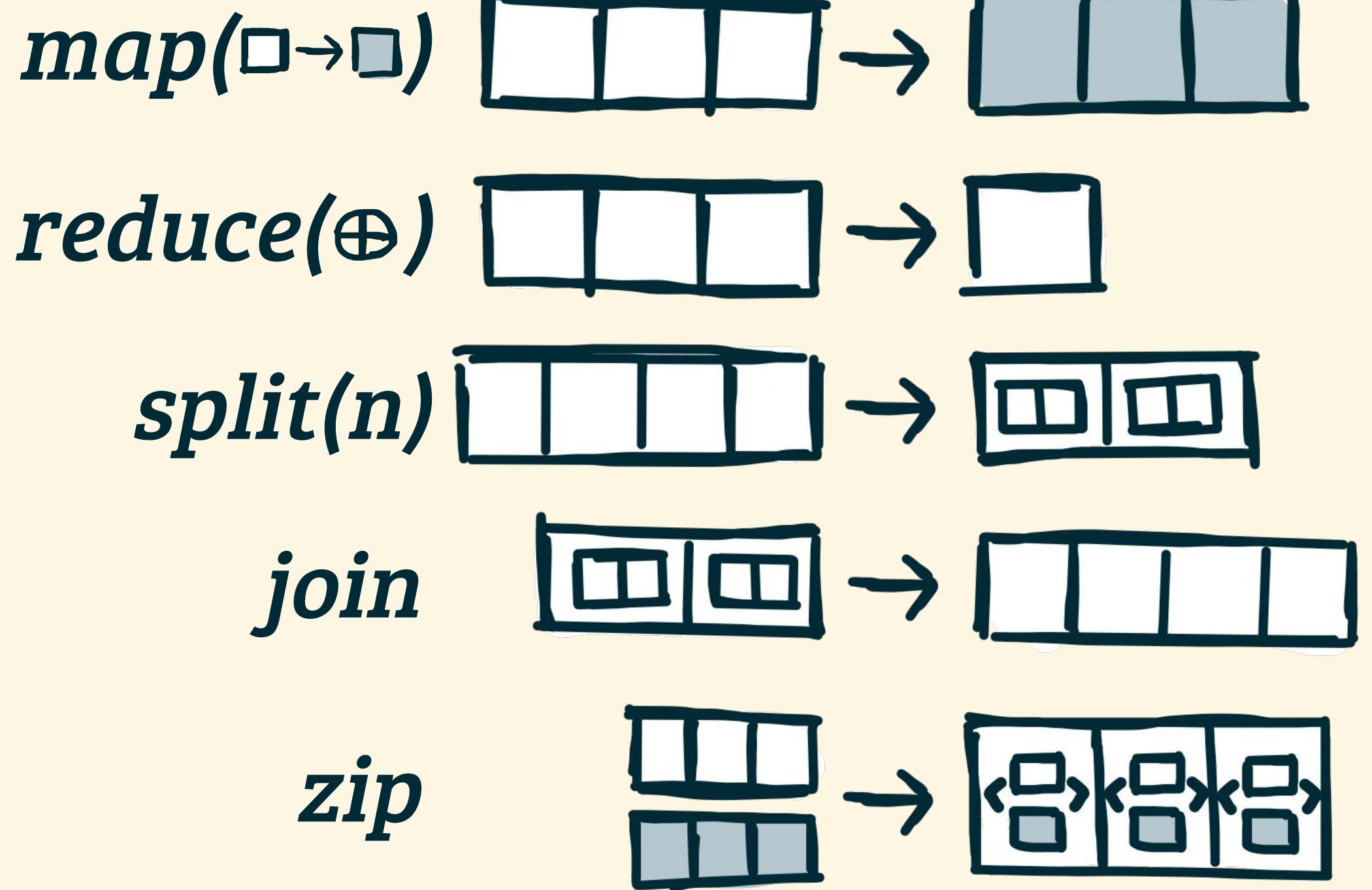


dotproduct.lift

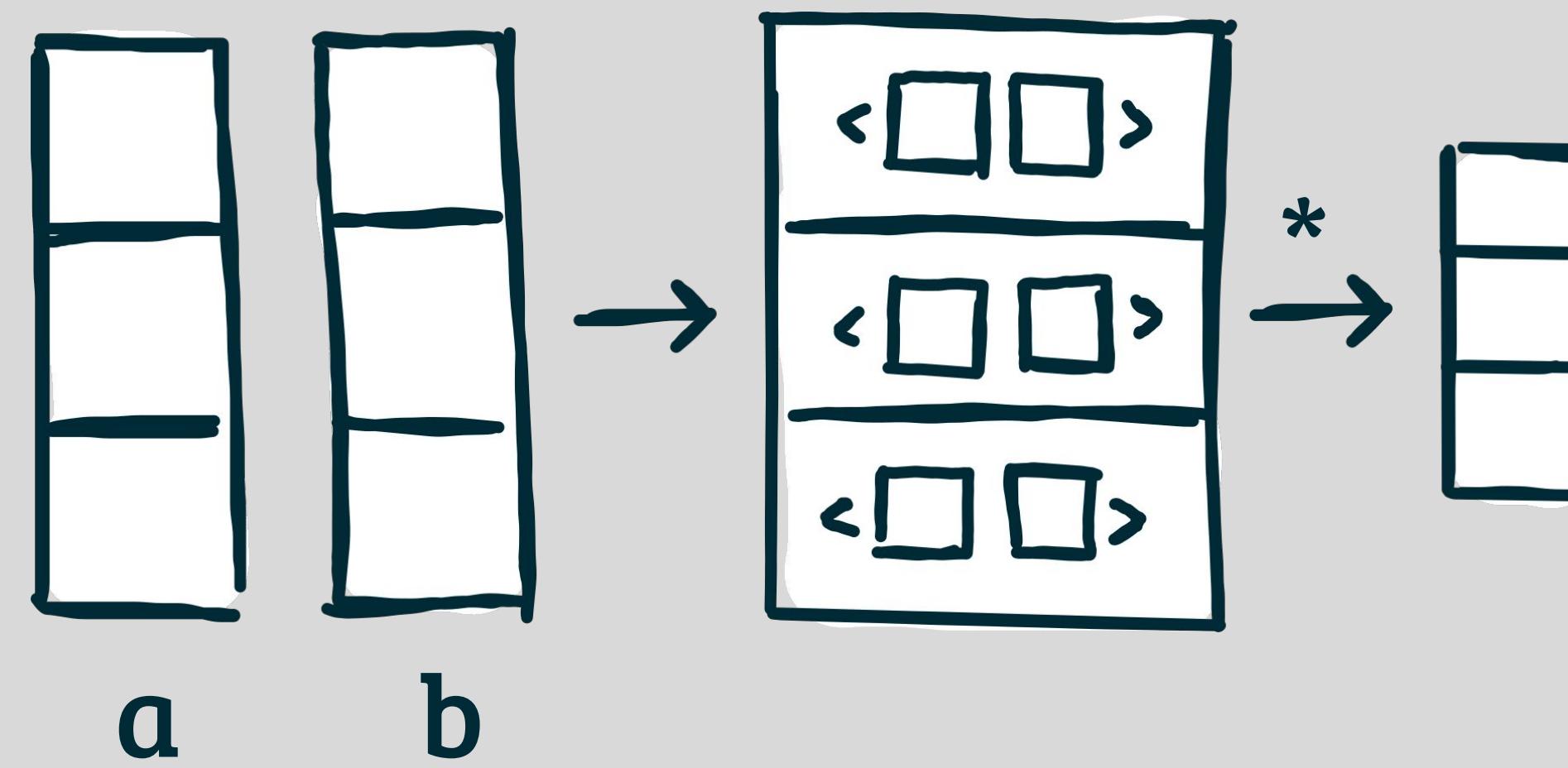


zip(a, b)

LIFT'S HIGH-LEVEL PRIMITIVES

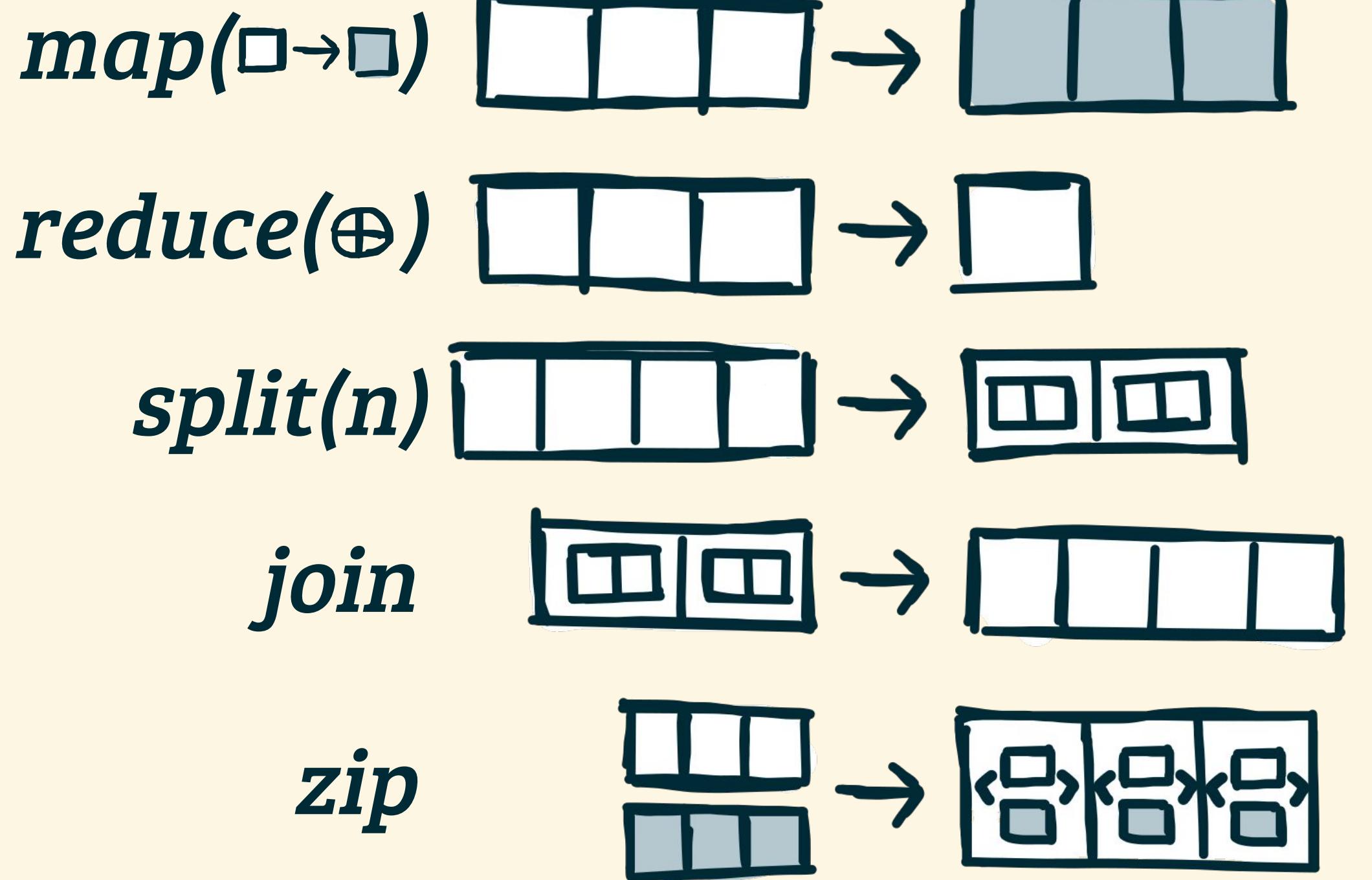


dotproduct.lift

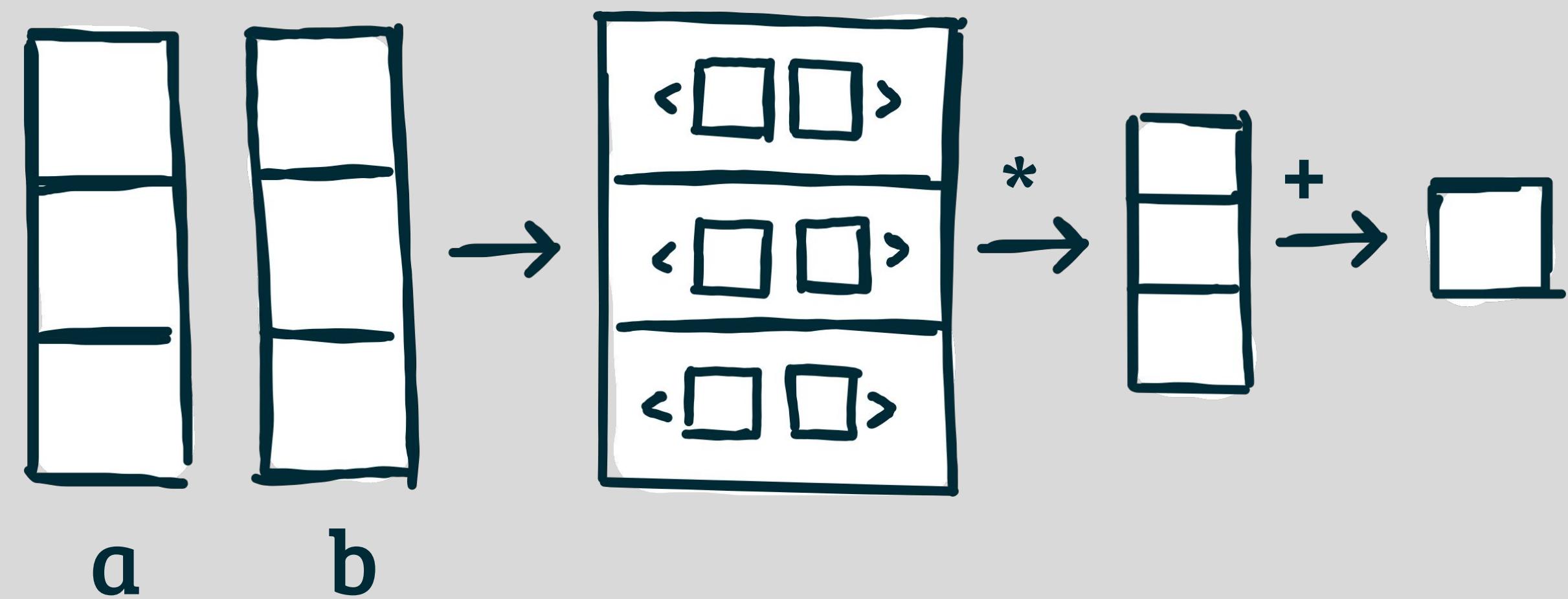


map(* , *zip*(a, b))

LIFT'S HIGH-LEVEL PRIMITIVES



dotproduct.lift



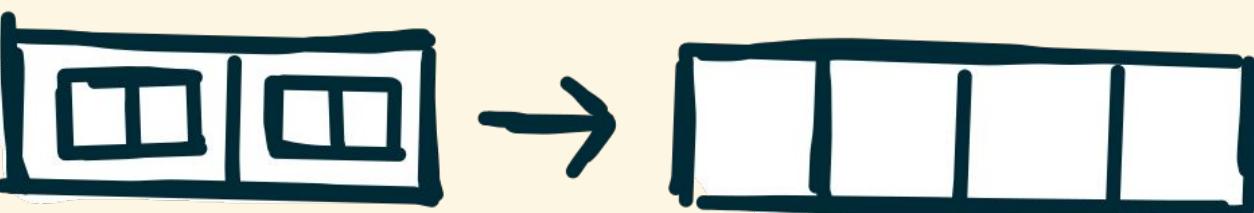
reduce(+, 0, map(, zip(a,b)))*

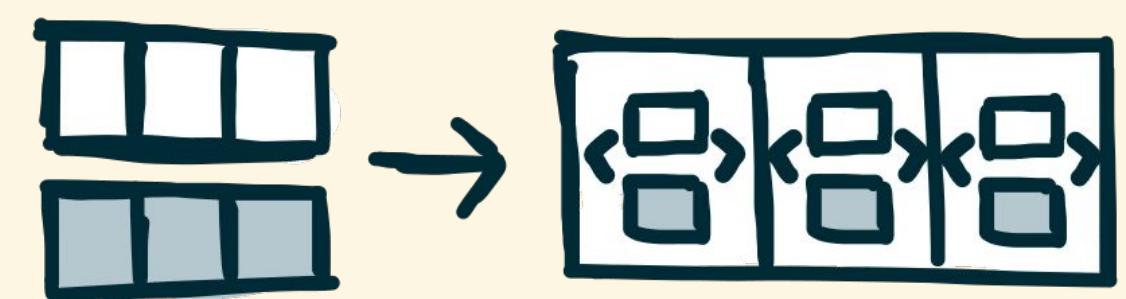
LIFT'S HIGH-LEVEL PRIMITIVES

map($\square \rightarrow \square$) 

reduce(\oplus) 

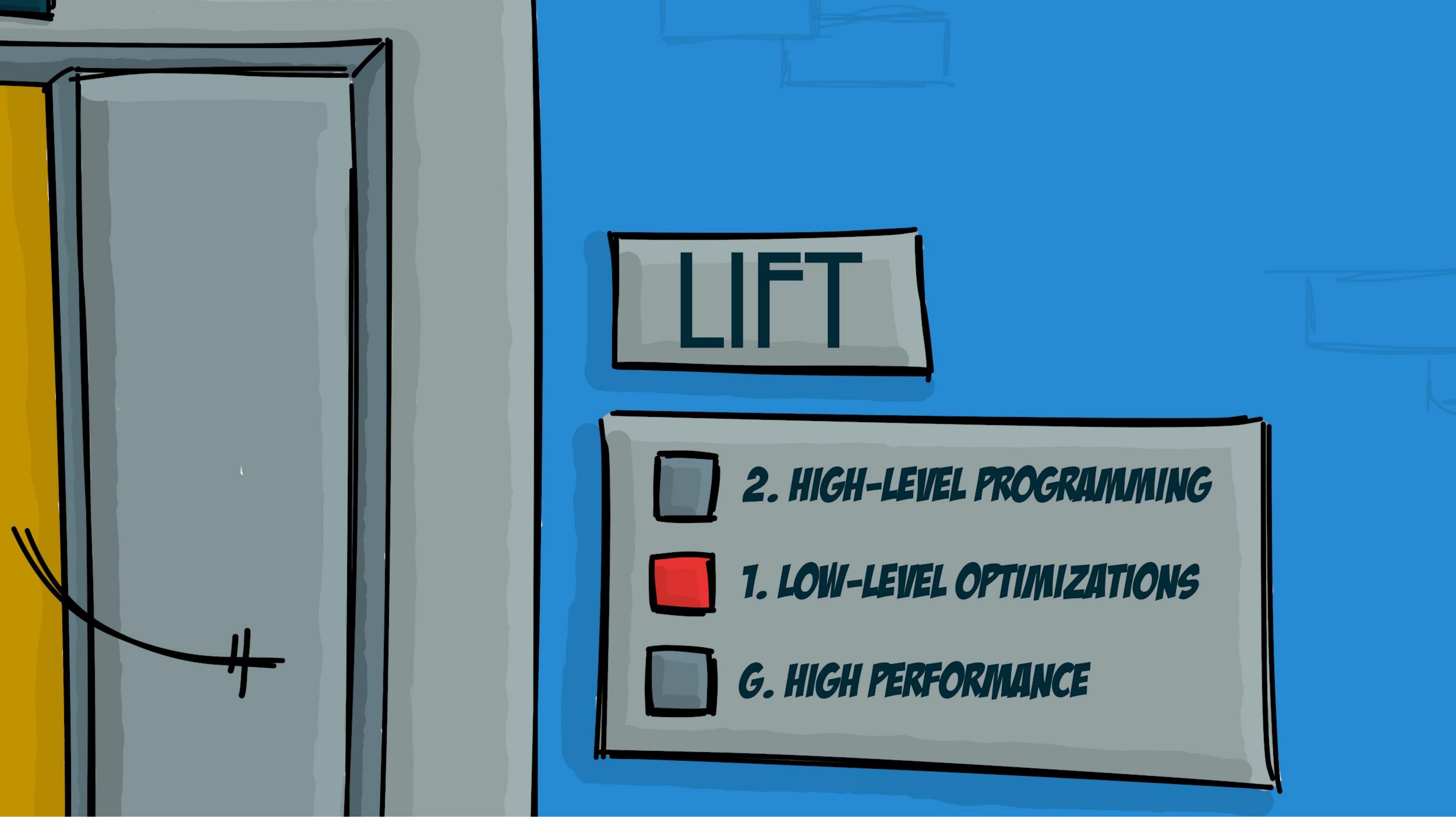
split(n) 

join 

zip 

matrixMult.lift

```
map( $\lambda$  rowA  $\mapsto$ 
    map( $\lambda$  colB  $\mapsto$ 
        dotProduct(rowA, colB)
        , transpose(B))
        , A)
```



LIFT

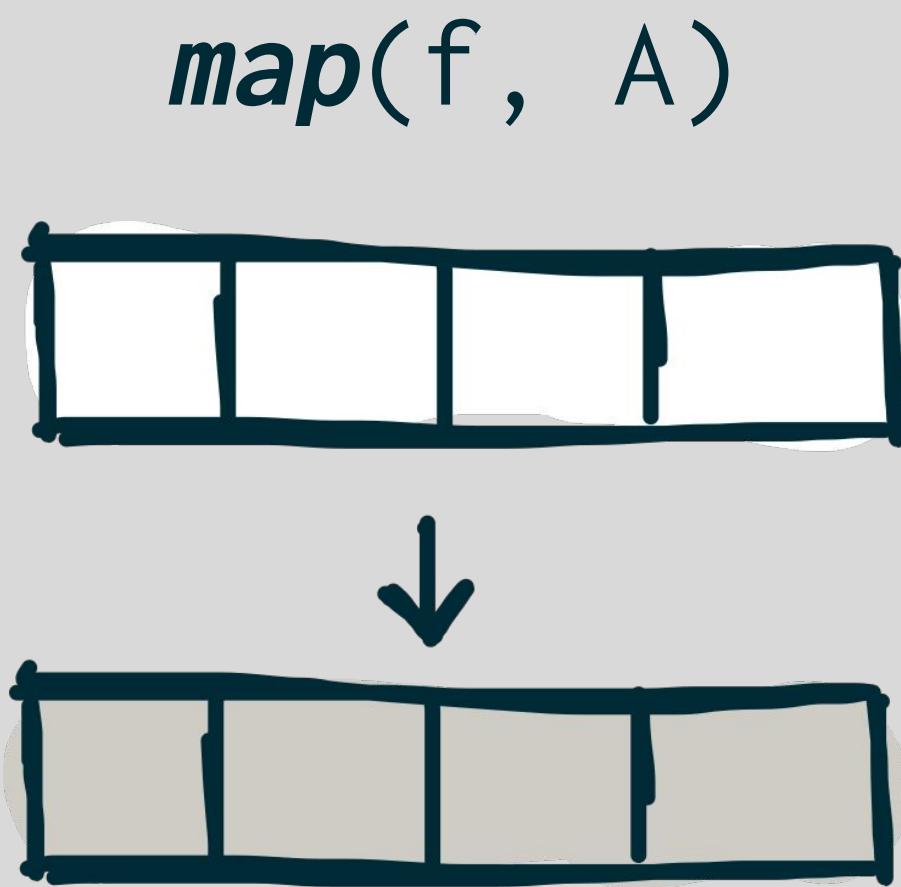
2. HIGH-LEVEL PROGRAMMING

1. LOW-LEVEL OPTIMIZATIONS

G. HIGH PERFORMANCE

IMPLEMENTATION CHOICES AS REWRITE RULES

Divide & Conquer



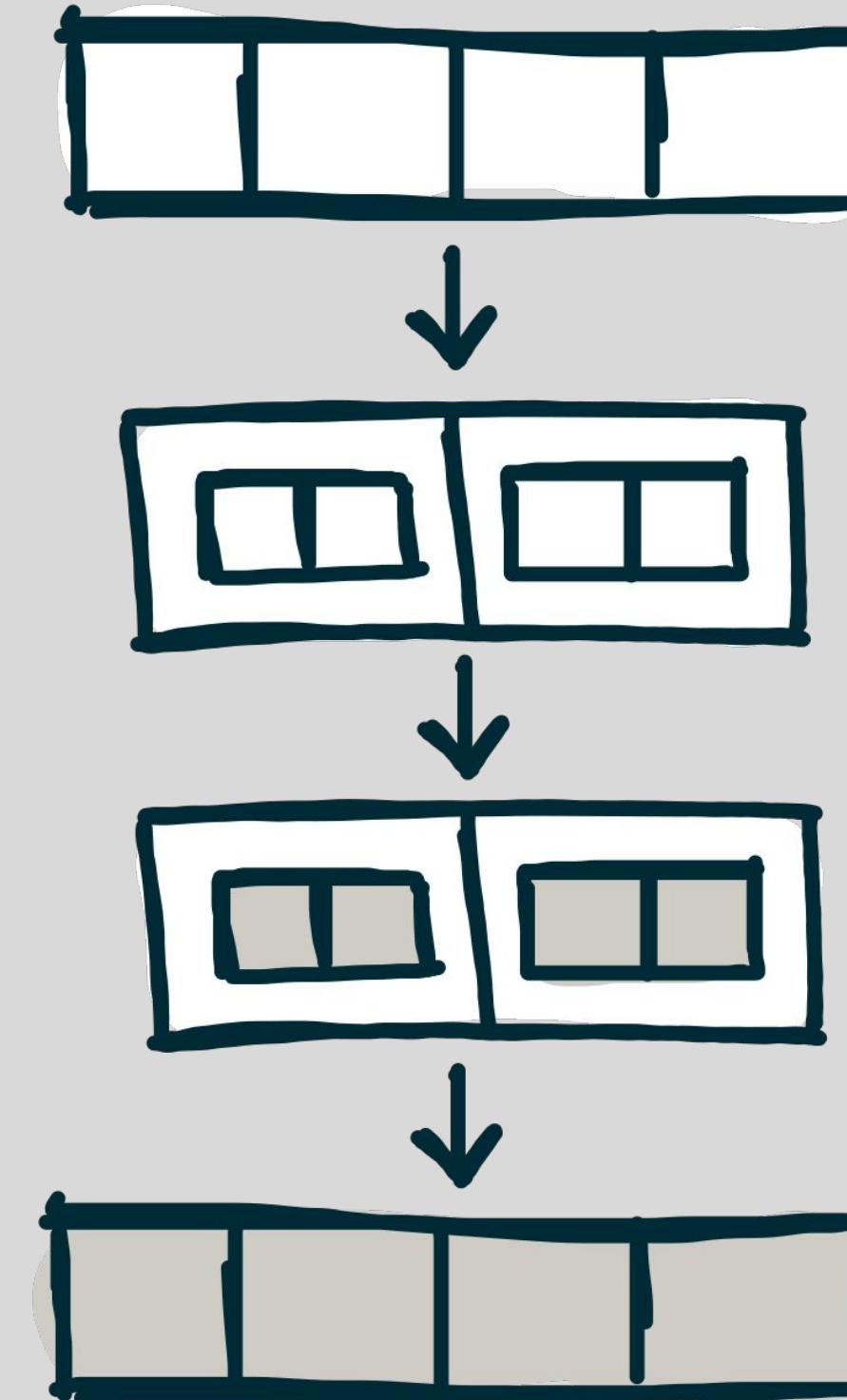
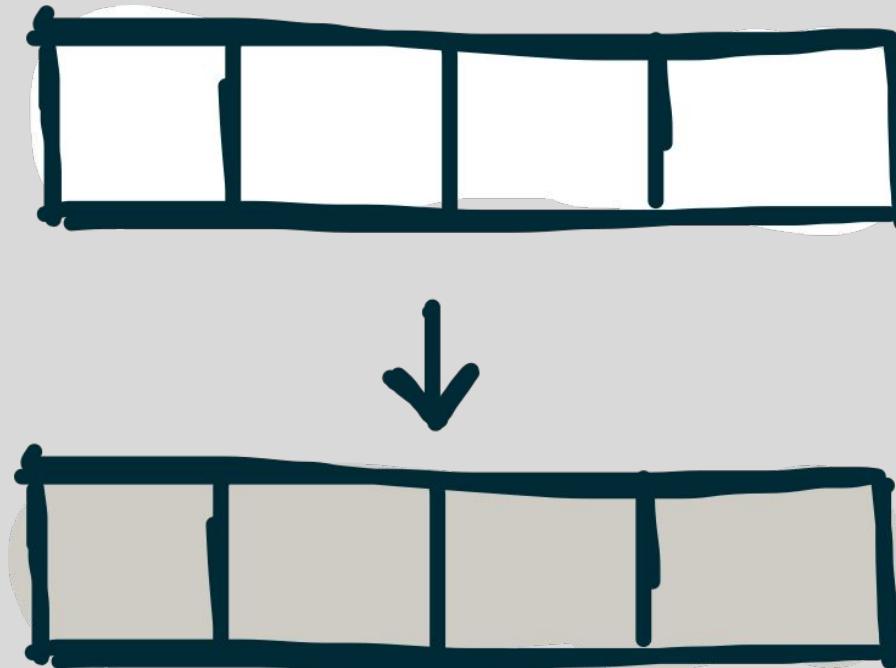
IMPLEMENTATION CHOICES AS REWRITE RULES

Divide & Conquer

map(f, A)



*join(map(map(f),
split(n, A)))*



LIFT'S LOW LEVEL (OPENCL) PRIMITIVES

Lift primitive

mapGlobal

mapWorkgroup

mapLocal

mapSeq

reduceSeq

toLocal, toGlobal

mapVec, splitVec, joinVec

OpenCL concept

Work-items

Work-groups

Sequential implementations

Memory areas

Vectorisation

REWRITING INTO OPENCL

Map rules:

$\text{map } f \mapsto \text{mapGlobal } f \mid \text{mapWorkgroup } f \mid \text{mapLocal } f \mid \text{mapSeq } f$

Local / global memory:

$\text{mapLocal } f \mapsto \text{toLocal } (\text{mapLocal } f)$

$\text{mapLocal } f \mapsto \text{toGlobal } (\text{mapLocal } f)$

Vectorization:

$\text{map } f \mapsto \text{joinVec } \circ \text{map } (\text{mapVec } f) \circ \text{splitVec } n$

OPTIMIZATIONS AS MACRO RULES

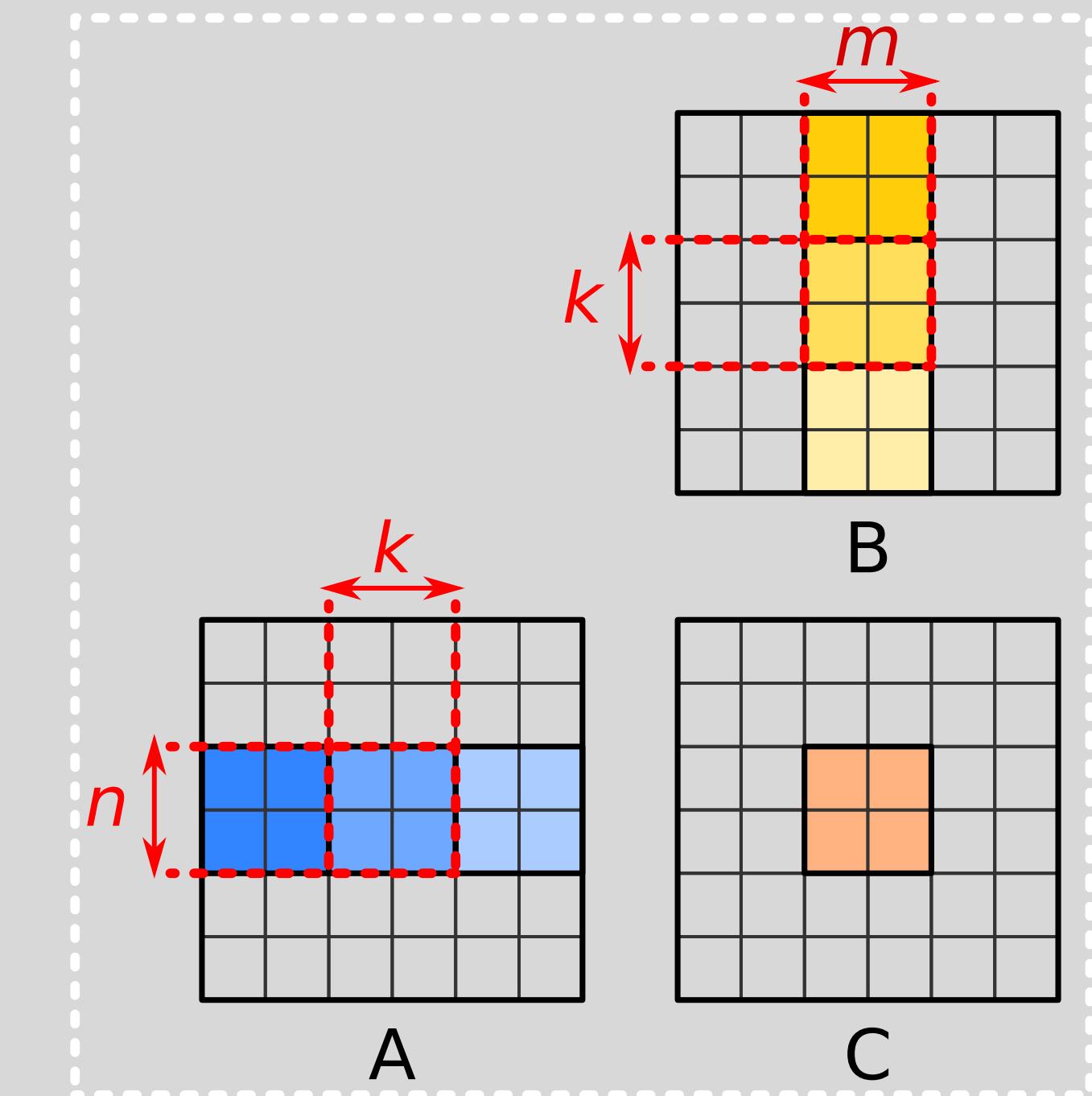
2D Tiling

Naïve matrix multiplication

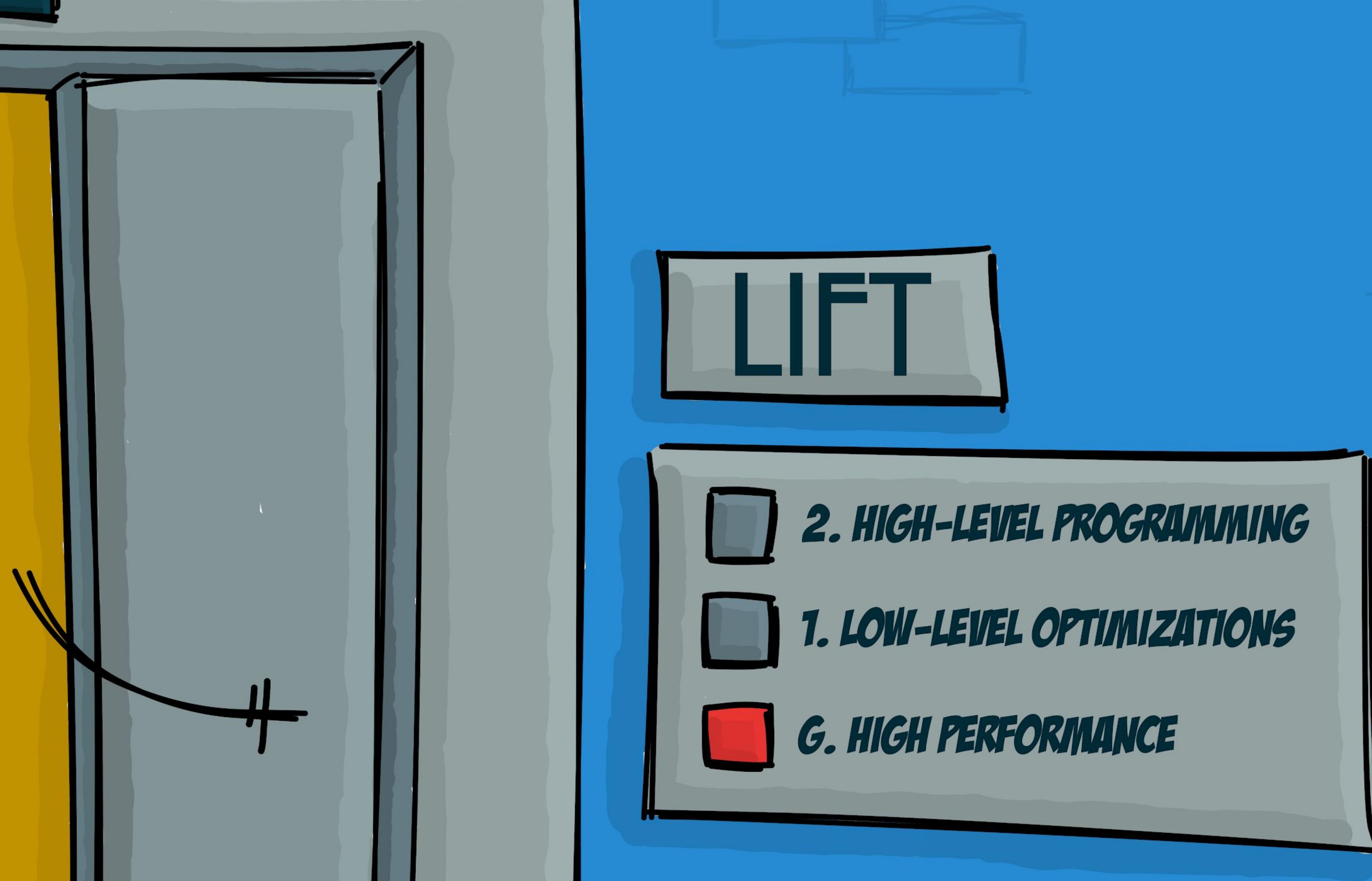
```
1 map(λ arow .  
2   map(λ bcol .  
3     reduce(+, 0) ○ map(×) ○ zip(arow, bcol)  
4     , transpose(B))  
5   , A)
```

↓ Apply tiling rules

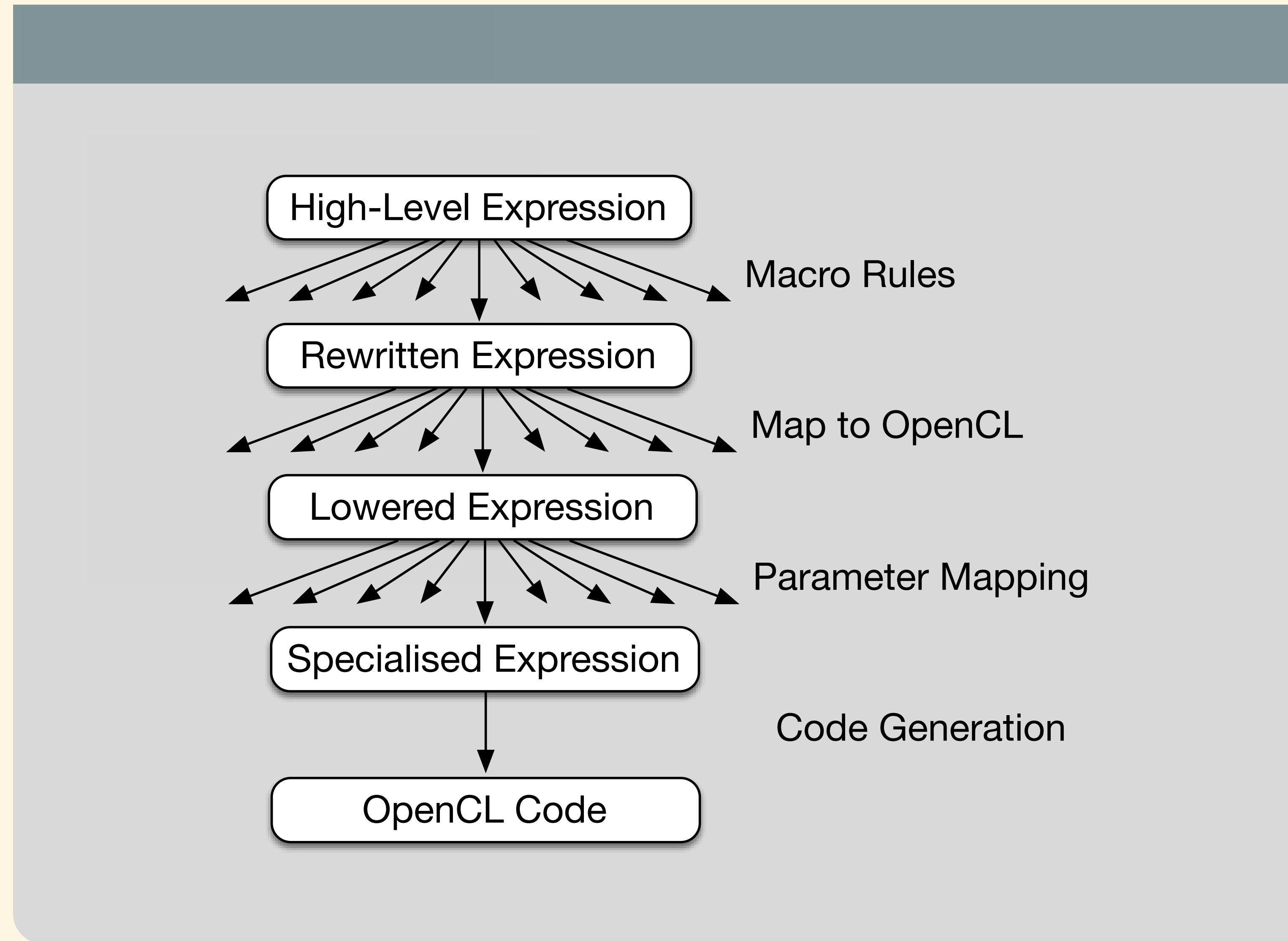
```
1 untile ○ map(λ rowOfTilesA .  
2   map(λ colOfTilesB .  
3     toGlobal(copy2D) ○  
4     reduce(λ (tileAcc, (tileA, tileB)) .  
5       map(map(+)) ○ zip(tileAcc) ○  
6       map(λ as .  
7         map(λ bs .  
8           reduce(+, 0) ○ map(×) ○ zip(as, bs)  
9           , toLocal(copy2D(tileB)))  
10          , toLocal(copy2D(tileA)))  
11          , 0, zip(rowOfTilesA, colOfTilesB))  
12        ) ○ tile(m, k, transpose(B))  
13      ) ○ tile(n, k, A)
```



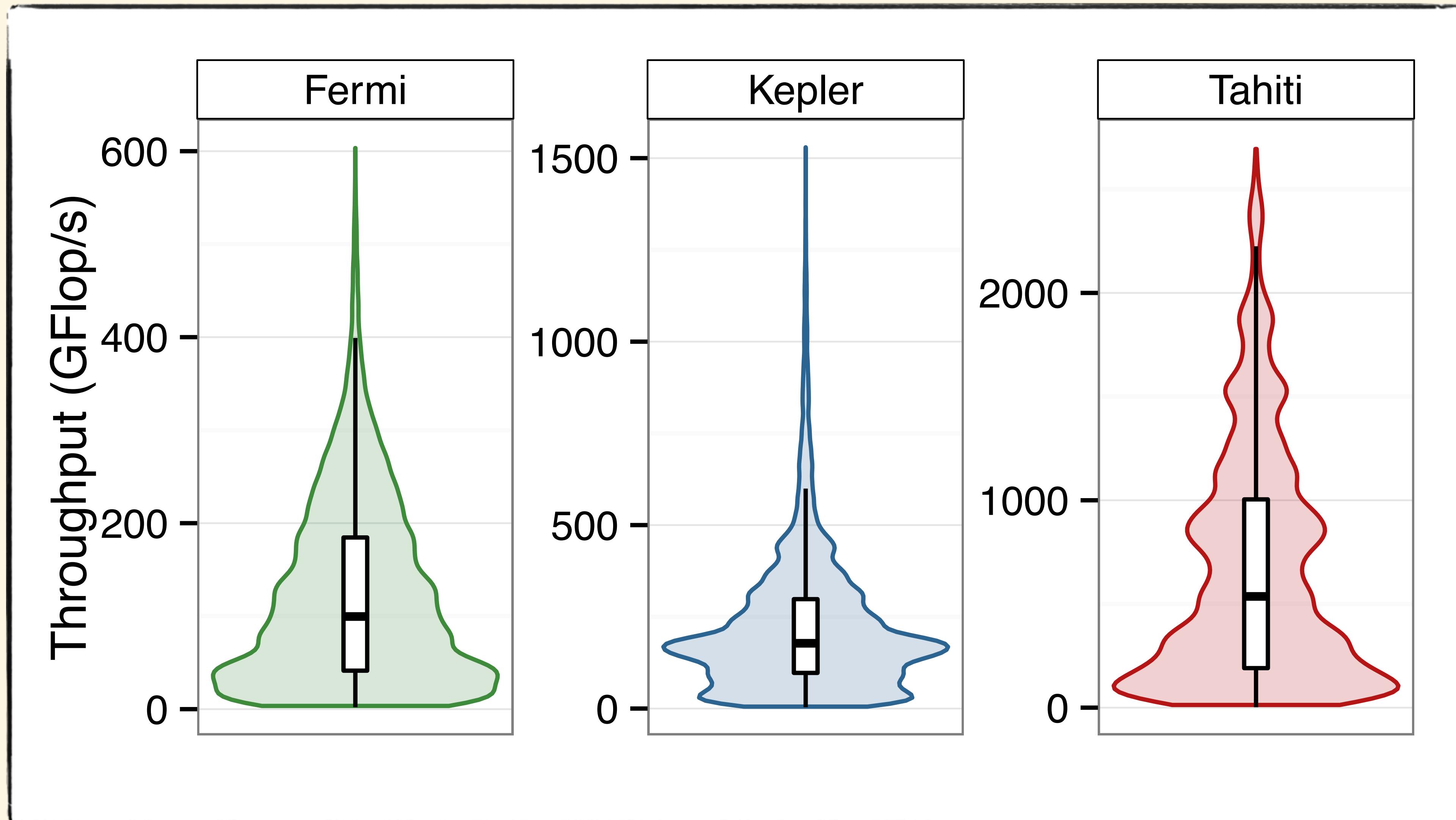
[GPGPU'16]



EXPLORATION BY REWRITING



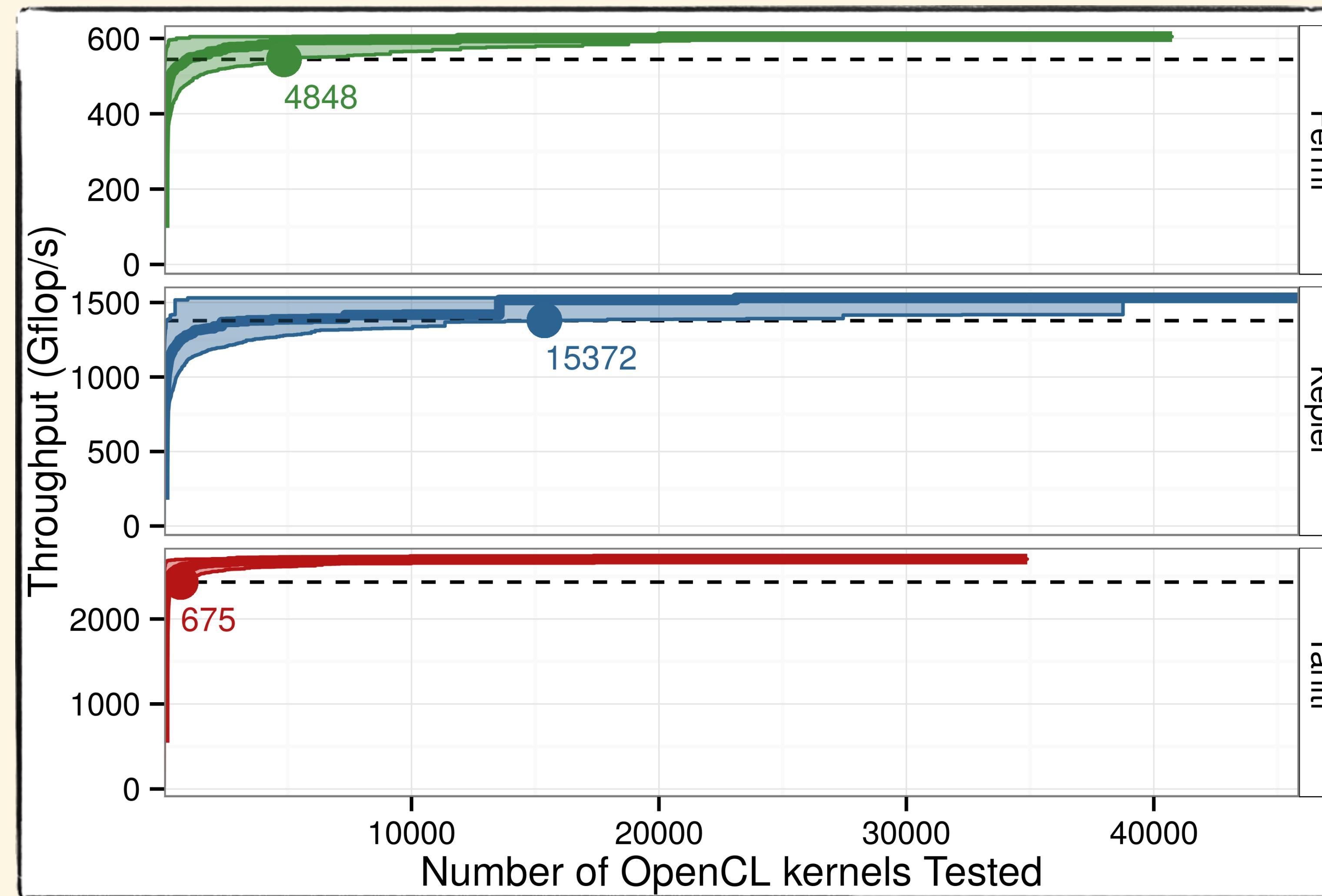
EXPLORATION SPACE MATRIX MULTIPLICATION



Only few generated code with very good performance

[GPGPU'16]

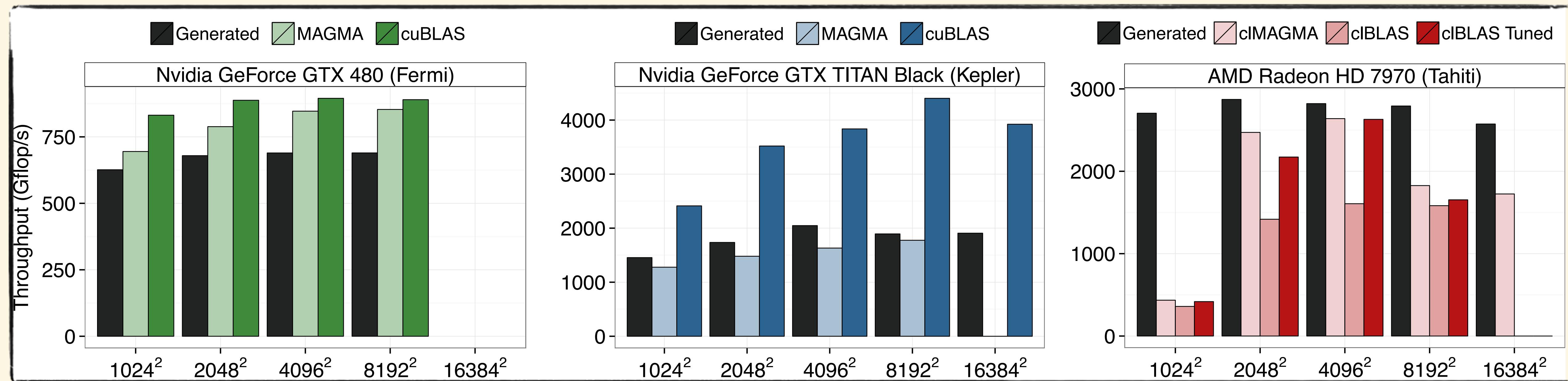
EVEN RANDOMISED SEARCH WORKS WELL!



Still: One can expect to find a good performing kernel quickly!

[GPGPU'16]

PERFORMANCE RESULTS MATRIX MULTIPLICATION

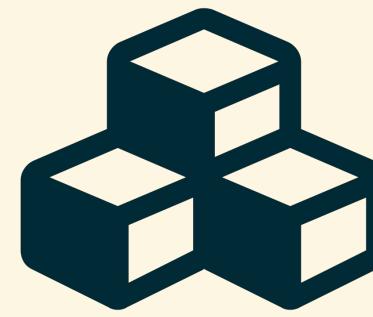


Performance close or better than hand-tuned MAGMA library

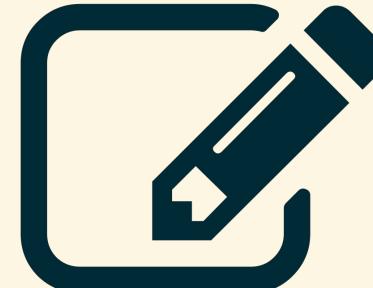
STENCIL COMPUTATIONS IN LIFT

[CGO'18] Best Paper Award

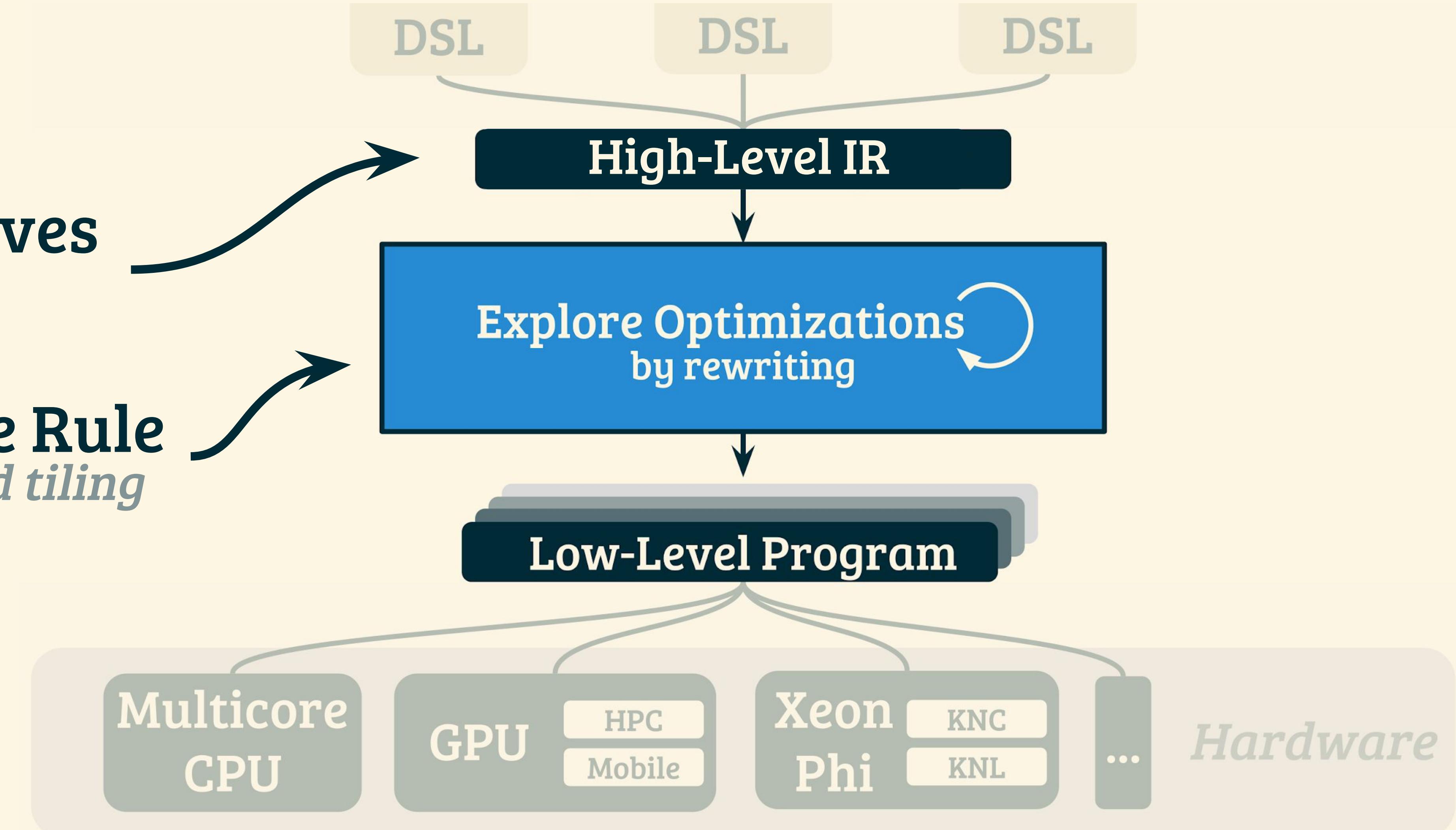
We added:



2 Primitives
pad, slide



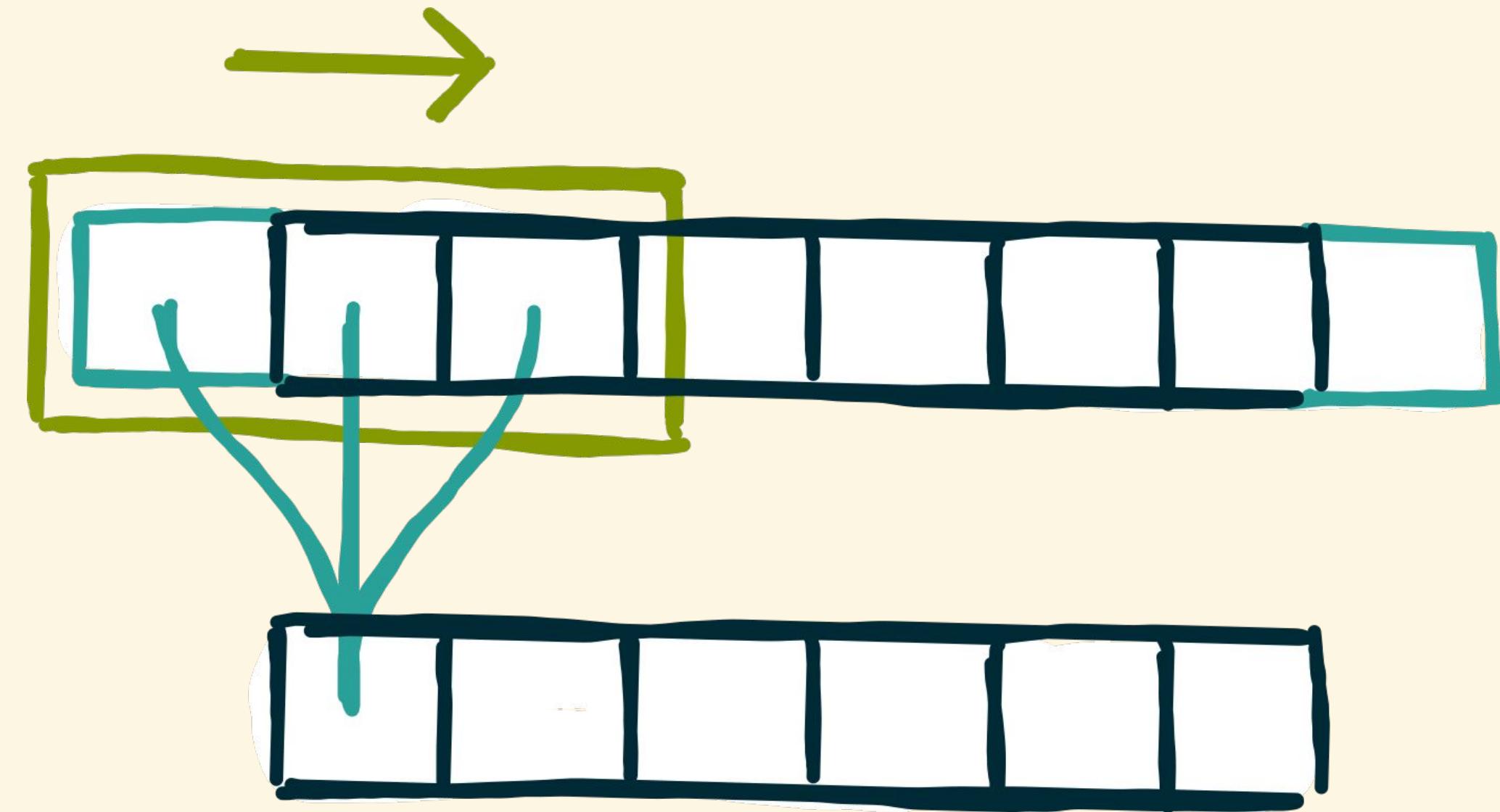
1 Rewrite Rule
overlapped tiling



DECOMPOSING STENCIL COMPUTATIONS

3-point-stencil.c

```
for (int i = 0; i < N ; i++) {  
    int sum = 0;  
    for ( int j = -1; j <= 1; j ++ ) { // ( a )  
        int pos = i + j;  
        pos = pos < 0 ? 0 : pos;  
        pos = pos > N - 1 ? N - 1 : pos;  
        sum += A[ pos ]; }  
    B[ i ] = sum ; }
```

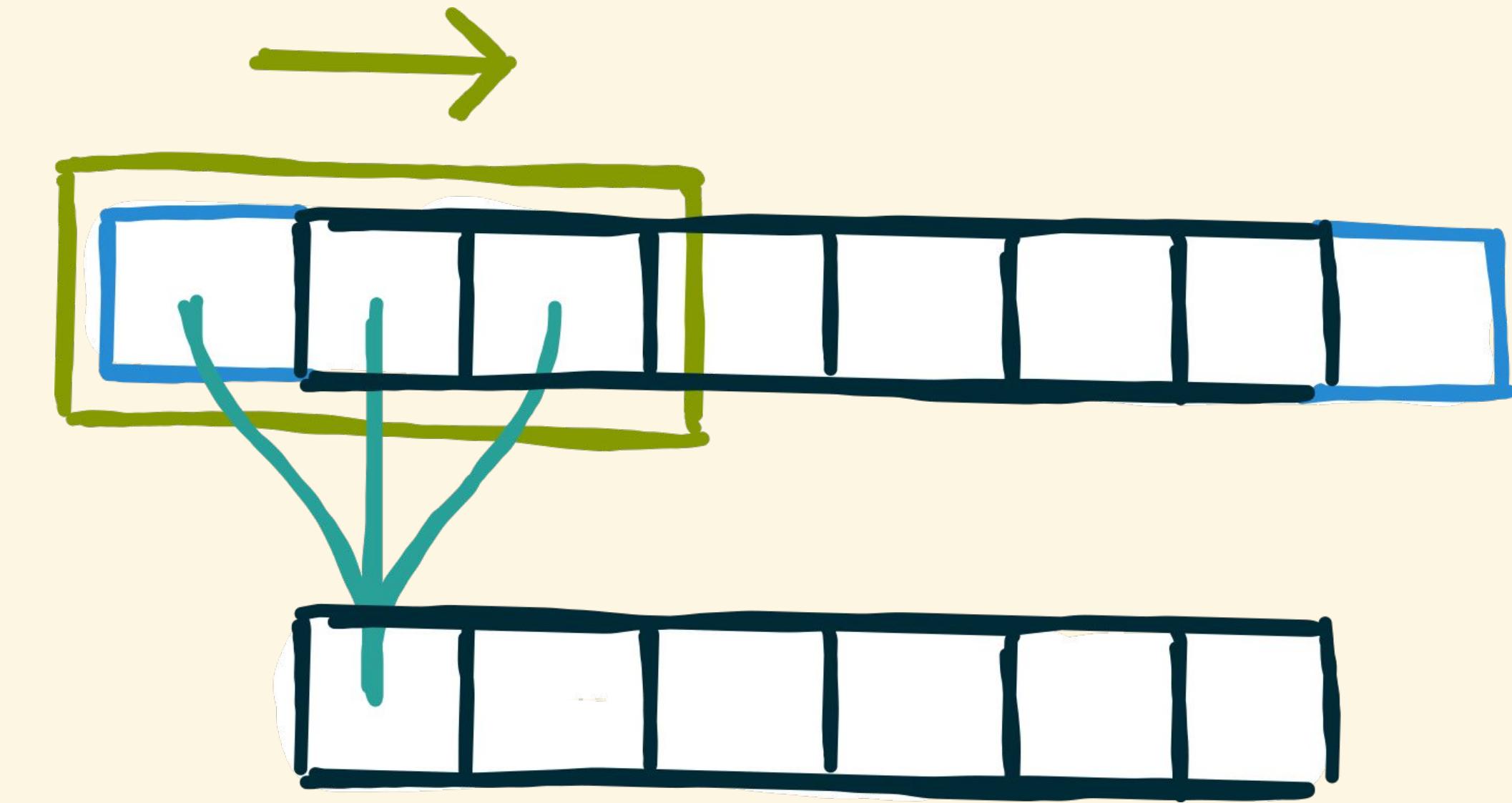


(a) access neighborhoods for every element

DECOMPOSING STENCIL COMPUTATIONS

3-point-stencil.c

```
for (int i = 0; i < N ; i++) {  
    int sum = 0;  
    for ( int j = -1; j <= 1; j ++ ) { // ( a )  
        int pos = i + j;  
        pos = pos < 0 ? 0 : pos;           // ( b )  
        pos = pos > N - 1 ? N - 1 : pos;  
        sum += A[ pos ]; }  
    B[ i ] = sum ; }
```

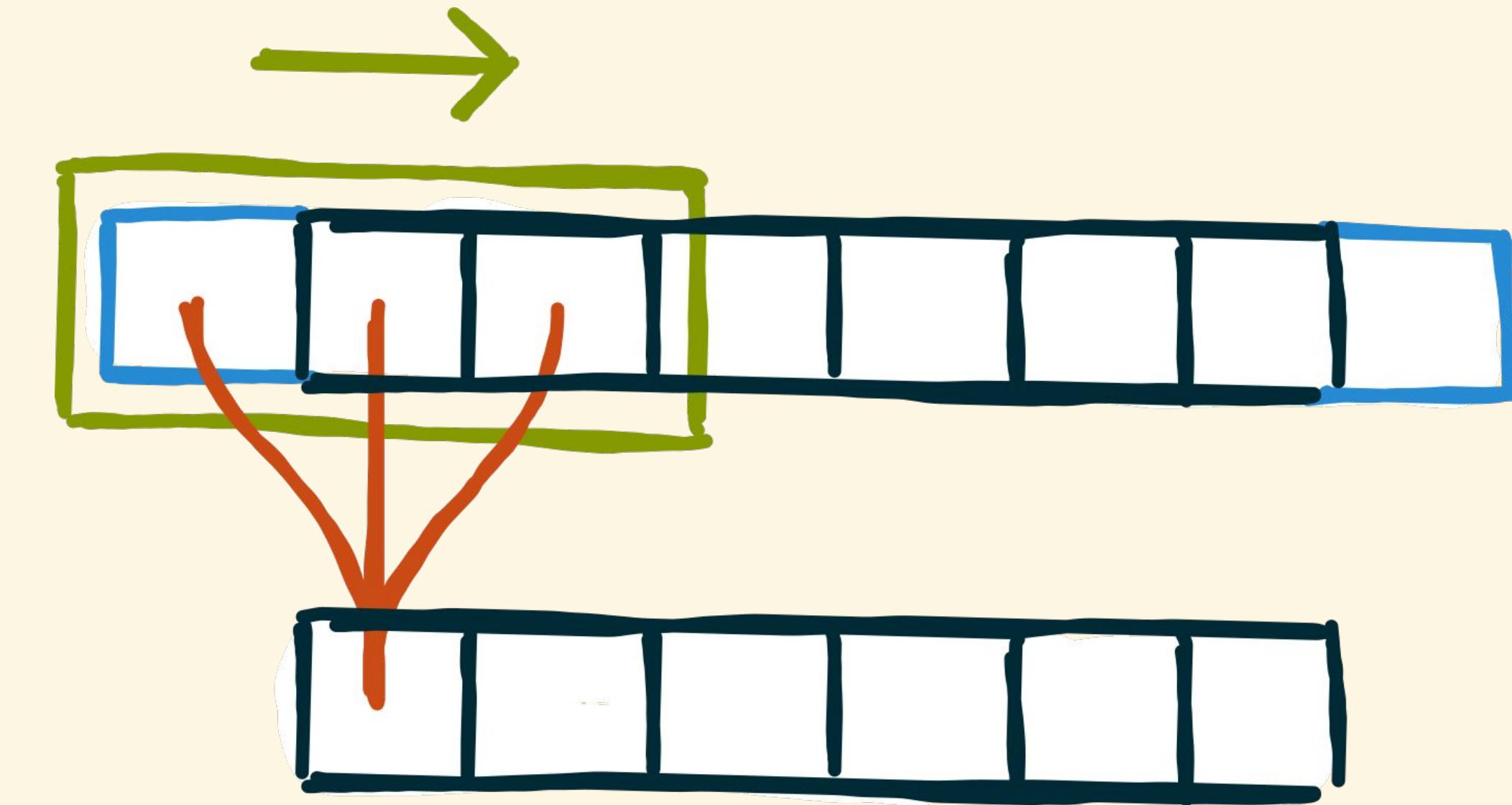


- (a) access neighborhoods for every element
- (b) specify boundary handling

DECOMPOSING STENCIL COMPUTATIONS

3-point-stencil.c

```
for (int i = 0; i < N ; i++) {  
    int sum = 0;  
    for ( int j = -1; j <= 1; j ++ ) { // ( a )  
        int pos = i + j;  
        pos = pos < 0 ? 0 : pos;           // ( b )  
        pos = pos > N - 1 ? N - 1 : pos;  
        sum += A[ pos ]; }                // ( c )  
    B[ i ] = sum ; }
```

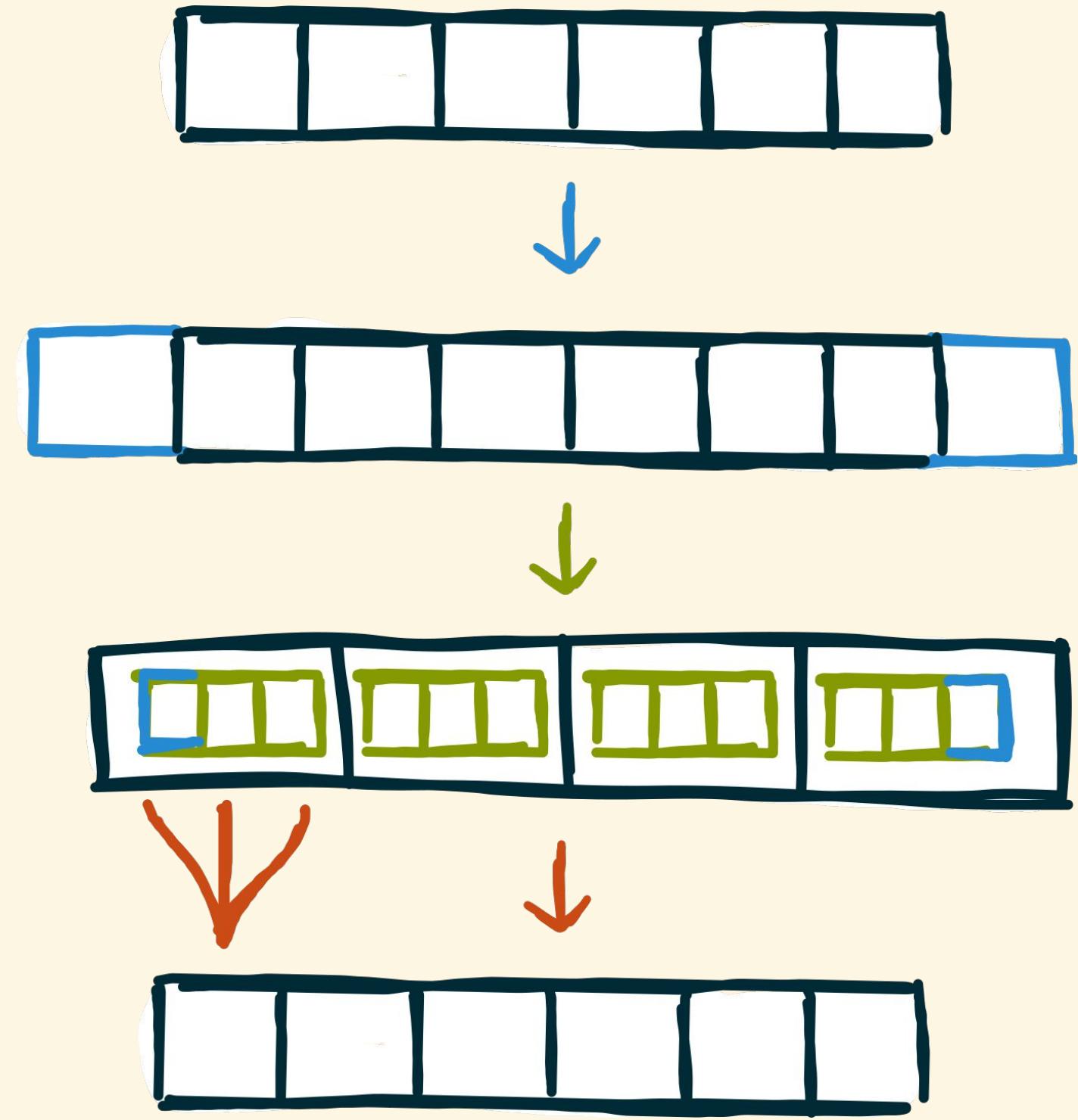


- (a) access neighborhoods for every element
- (b) specify boundary handling
- (c) apply stencil function to neighborhoods

DECOMPOSING STENCIL COMPUTATIONS

3-point-stencil.c

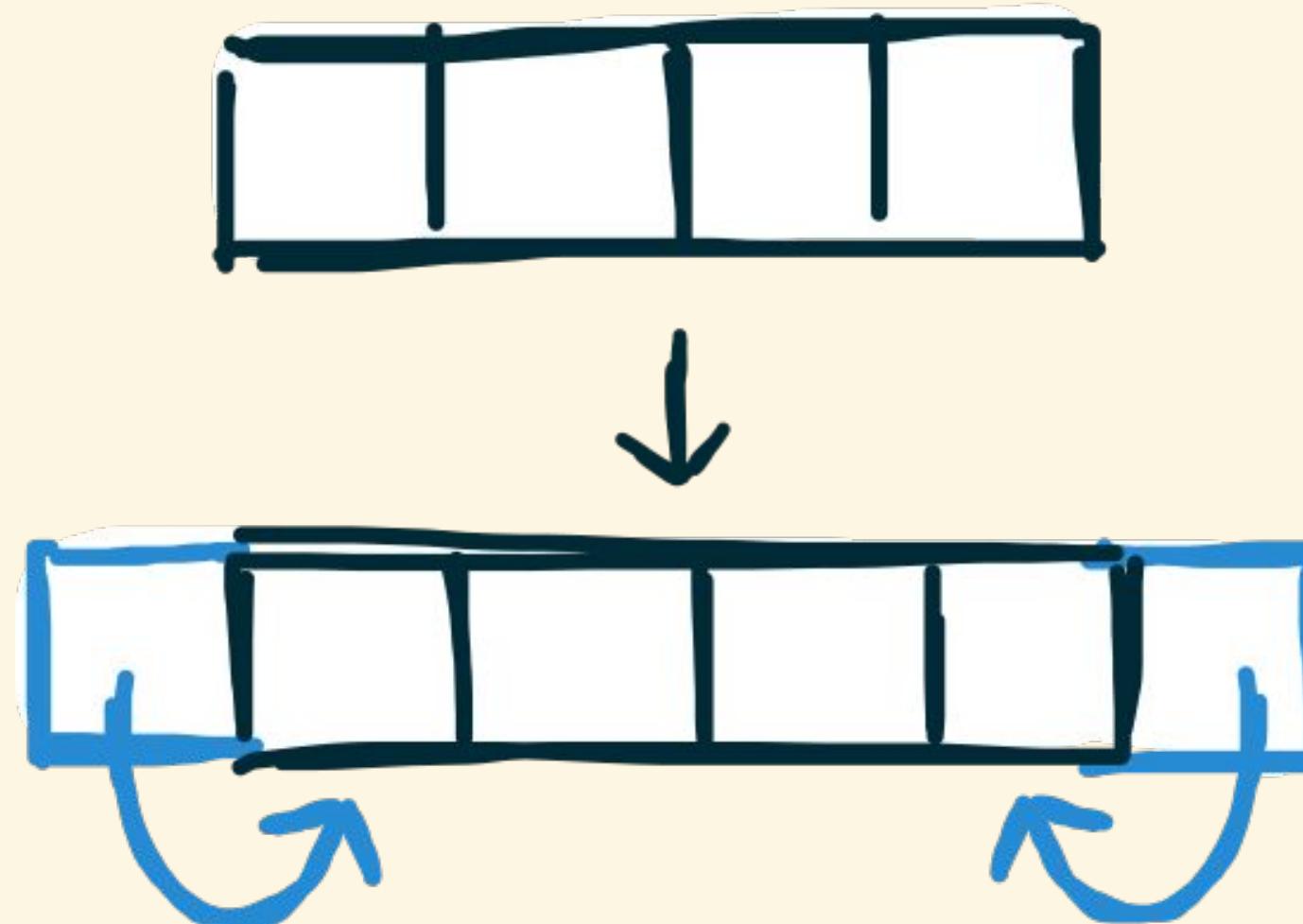
```
for (int i = 0; i < N ; i++) {  
    int sum = 0;  
    for ( int j = -1; j <= 1; j ++ ) { // ( a )  
        int pos = i + j;  
        pos = pos < 0 ? 0 : pos;           // ( b )  
        pos = pos > N - 1 ? N - 1 : pos;  
        sum += A[ pos ]; }               // ( c )  
    B[ i ] = sum ; }
```



- (a) access neighborhoods for every element
- (b) specify boundary handling
- (c) apply stencil function to neighborhoods

BOUNDARY HANDLING USING PAD

pad (reindexing)

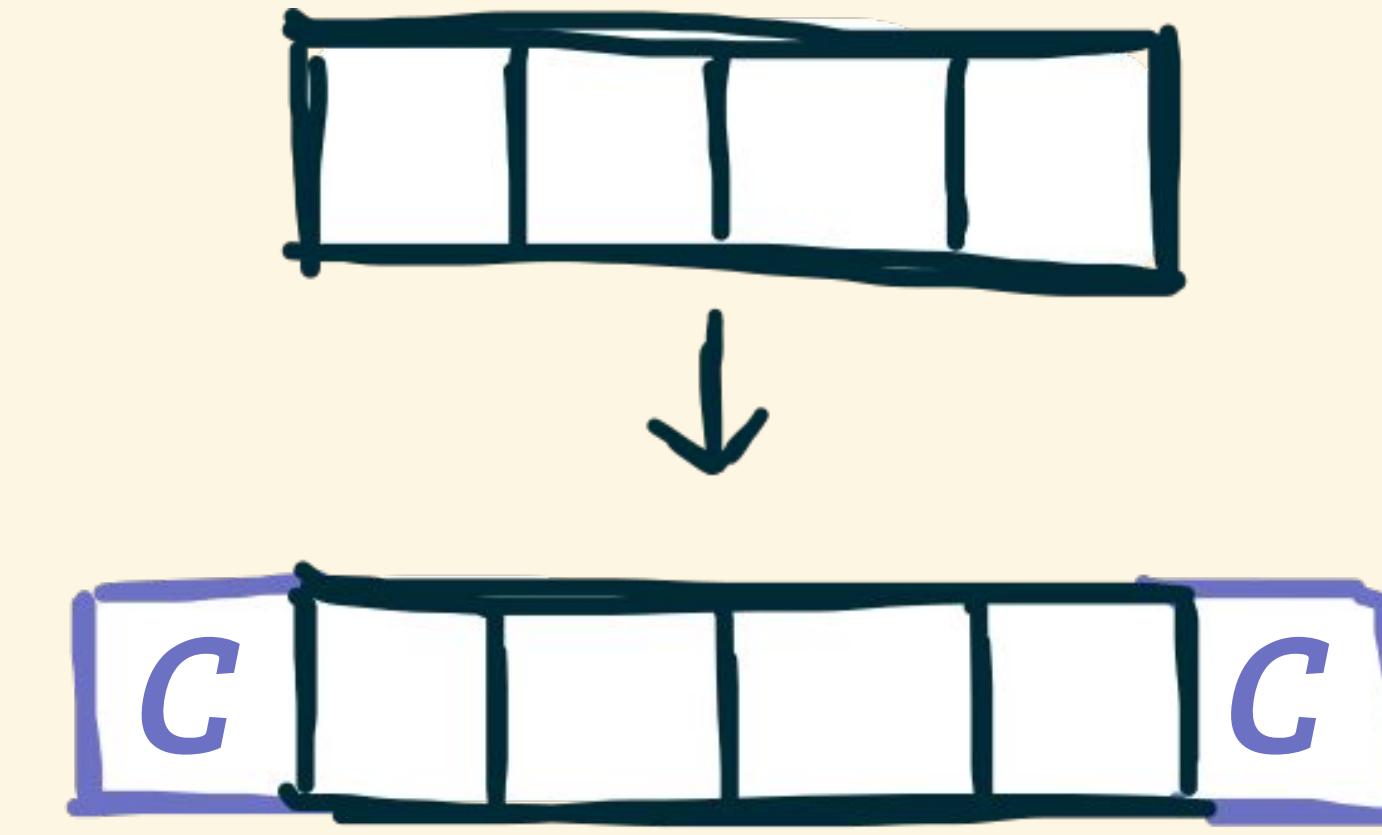


pad-reindexing.lift

```
clamp(i, n) = (i < 0) ? 0 :  
               ((i >= n) ? n-1:i)
```

```
pad(1,1,clamp, [a,b,c,d]) =  
[a,a,b,c,d,d]
```

pad (constant)

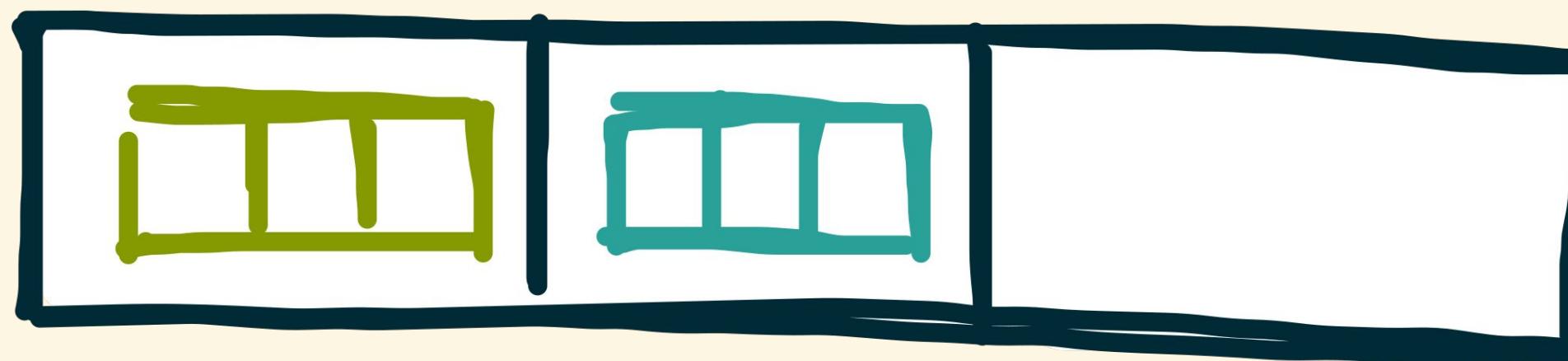
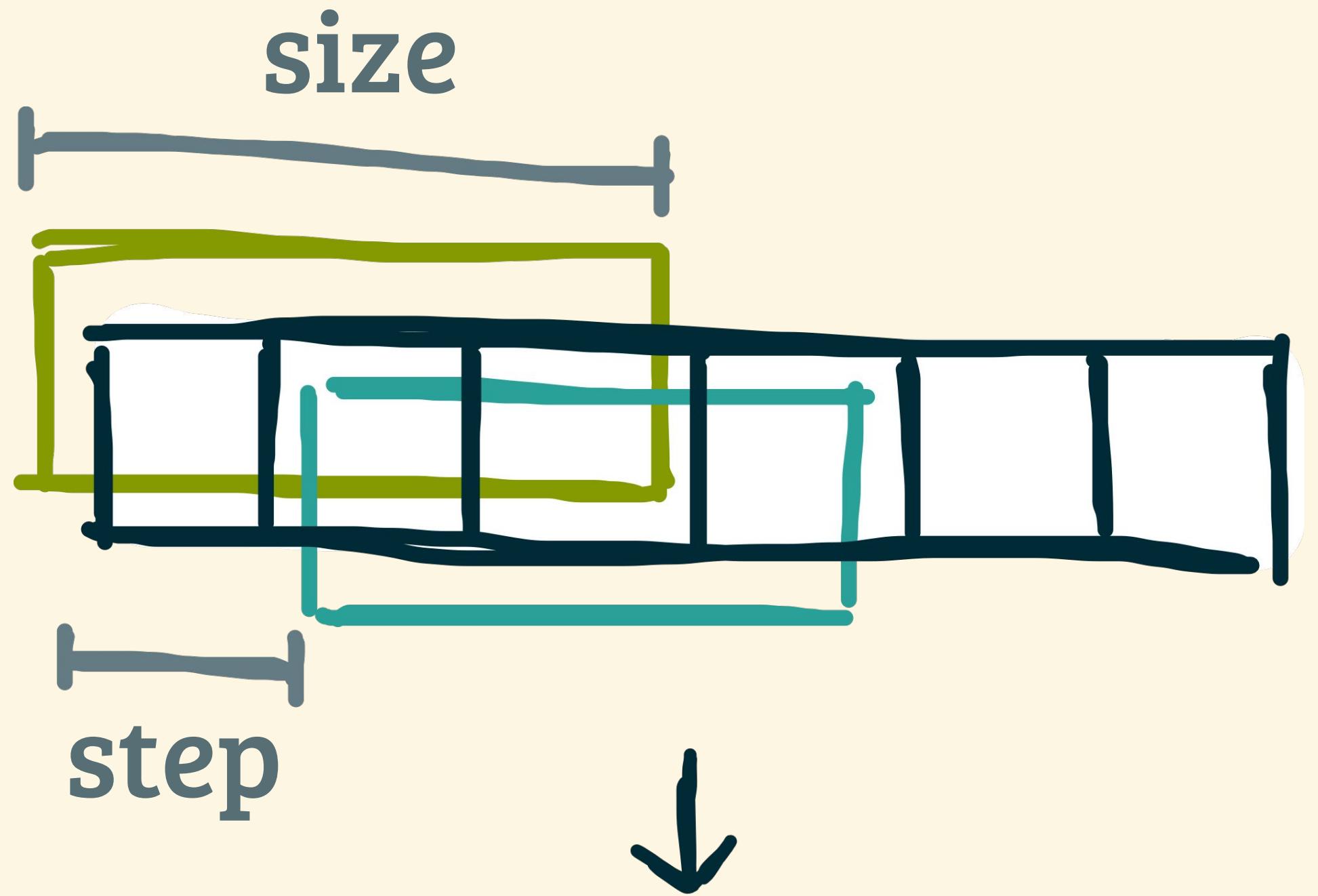


pad-constant.lift

```
constant(i, n) = C
```

```
pad(1,1,constant, [a,b,c,d]) =  
[C,a,b,c,d,C]
```

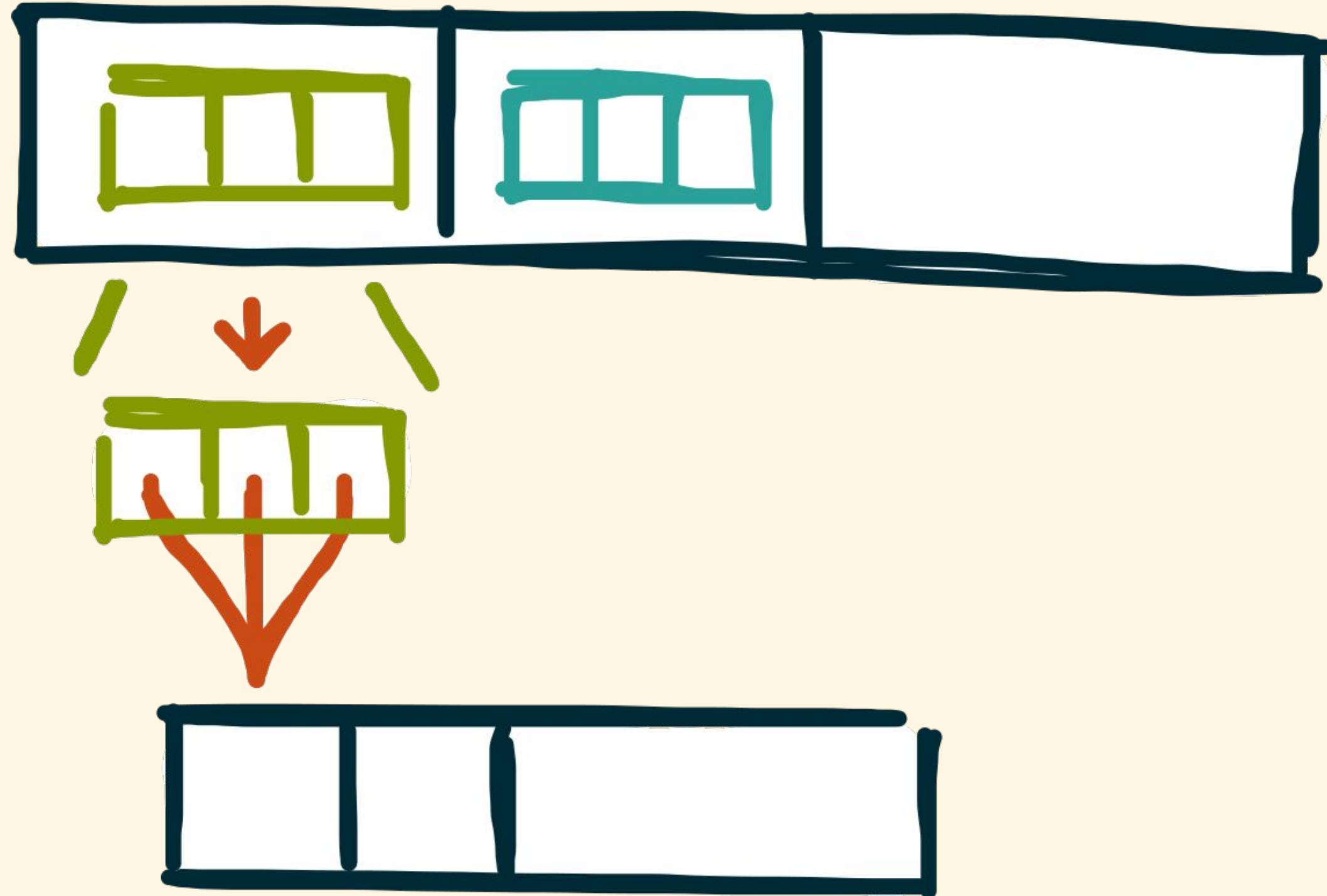
NEIGHBORHOOD CREATION USING SLIDE



slide-example.lift

```
slide(3,1,[a,b,c,d,e]) =  
[[a,b,c],[b,c,d],[c,d,e]]
```

APPLYING STENCIL FUNCTION USING MAP



sum-neighborhoods.lift

```
map(nbh =>  
    reduce(add, 0.0f, nbh))
```

PUTTING IT TOGETHER

map($\square \rightarrow \square$)



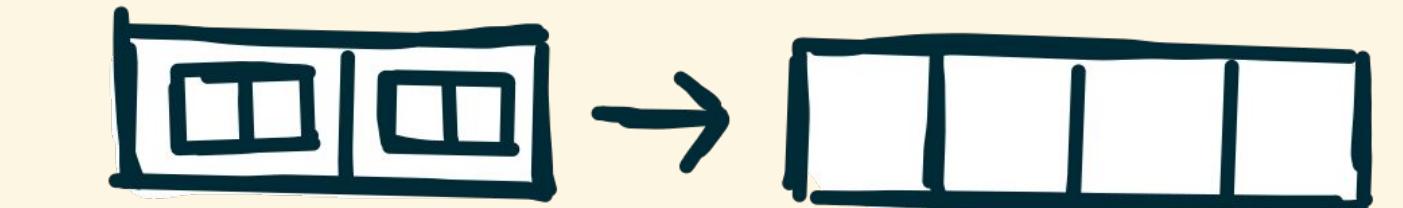
reduce(\oplus)



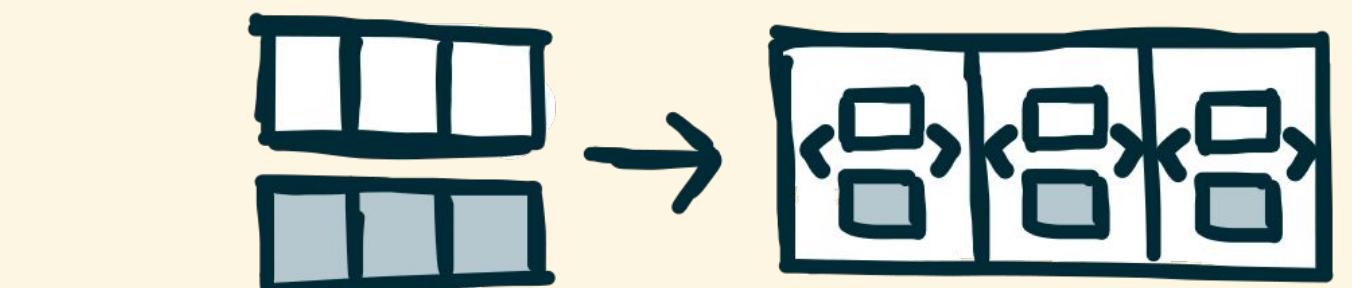
split(n)



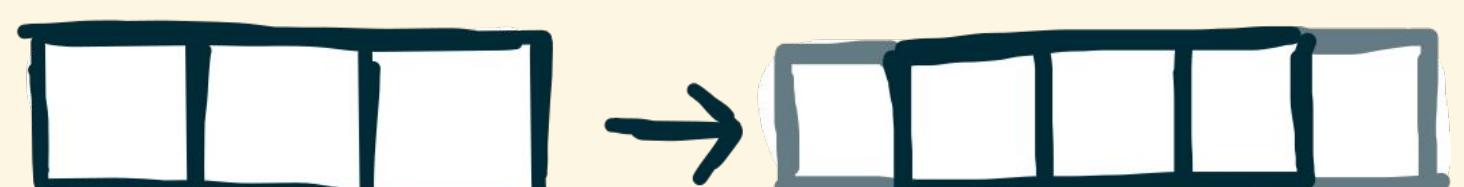
join



zip



pad(l,r,b)

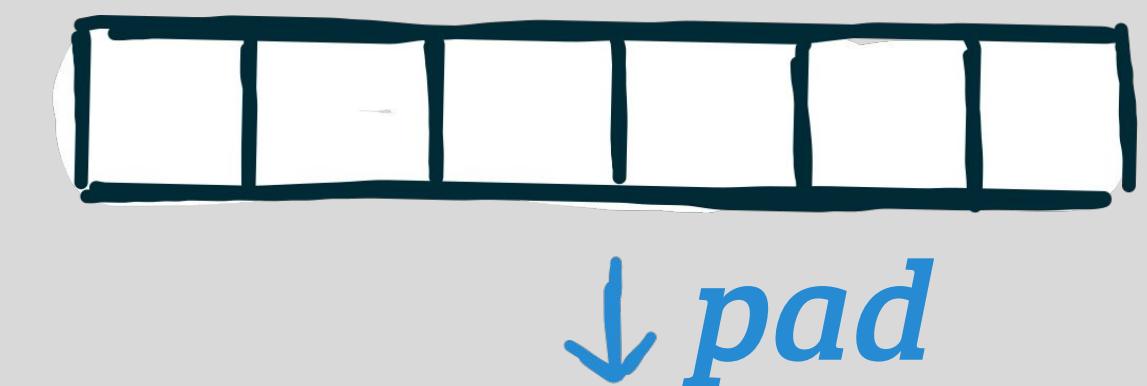


slide(n,s)

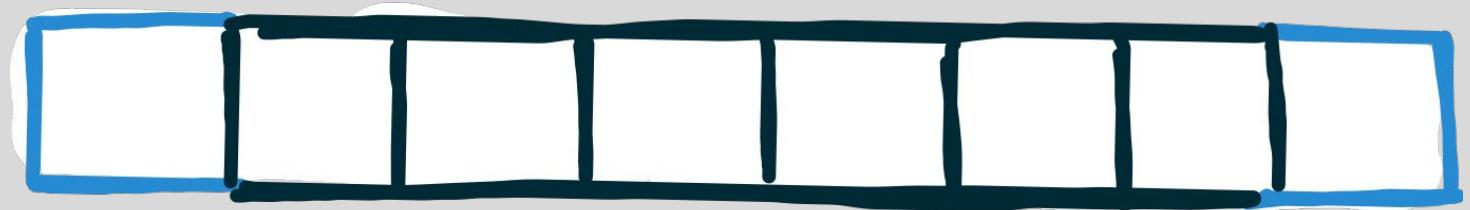


stencil1D.lift

```
def stencil1D =
  fun(A =>
    map(reduce(add, 0.0f),
        slide(3,1,
              pad(1,1,clamp,A))))
```



$\downarrow \text{pad}$



$\downarrow \text{slide}$

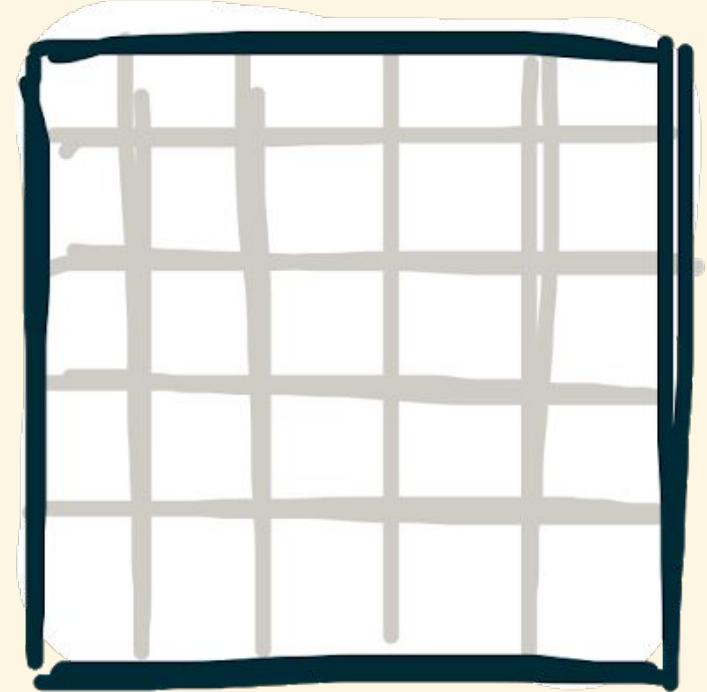


$\downarrow \text{map}$



MULTIDIMENSIONAL STENCIL COMPUTATIONS

are expressed as compositions of intuitive, generic 1D primitives

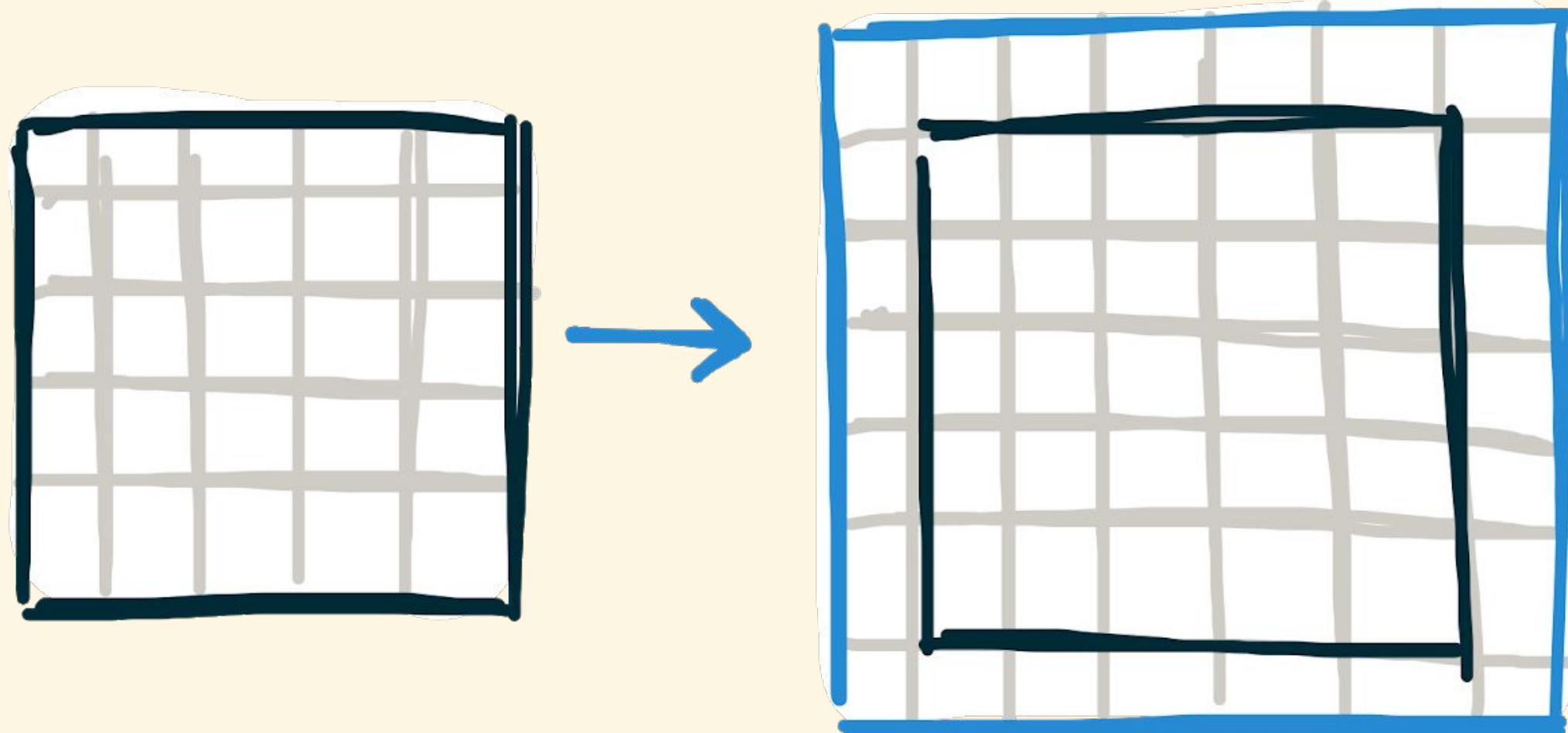


Decompose to Re-Compose

MULTIDIMENSIONAL STENCIL COMPUTATIONS

are expressed as compositions of intuitive, generic 1D primitives

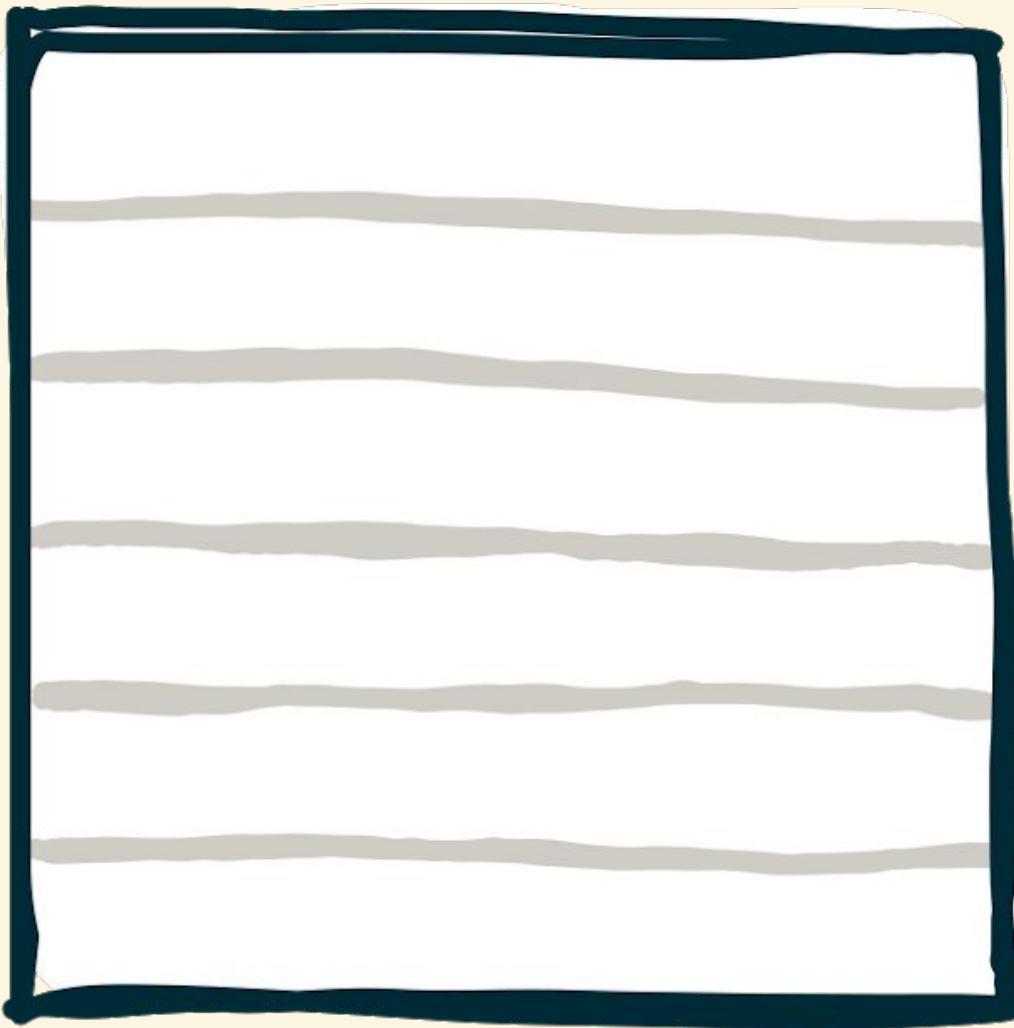
Decompose to Re-Compose



$\text{pad}_2(1, 1, \text{clamp}, \text{input})$

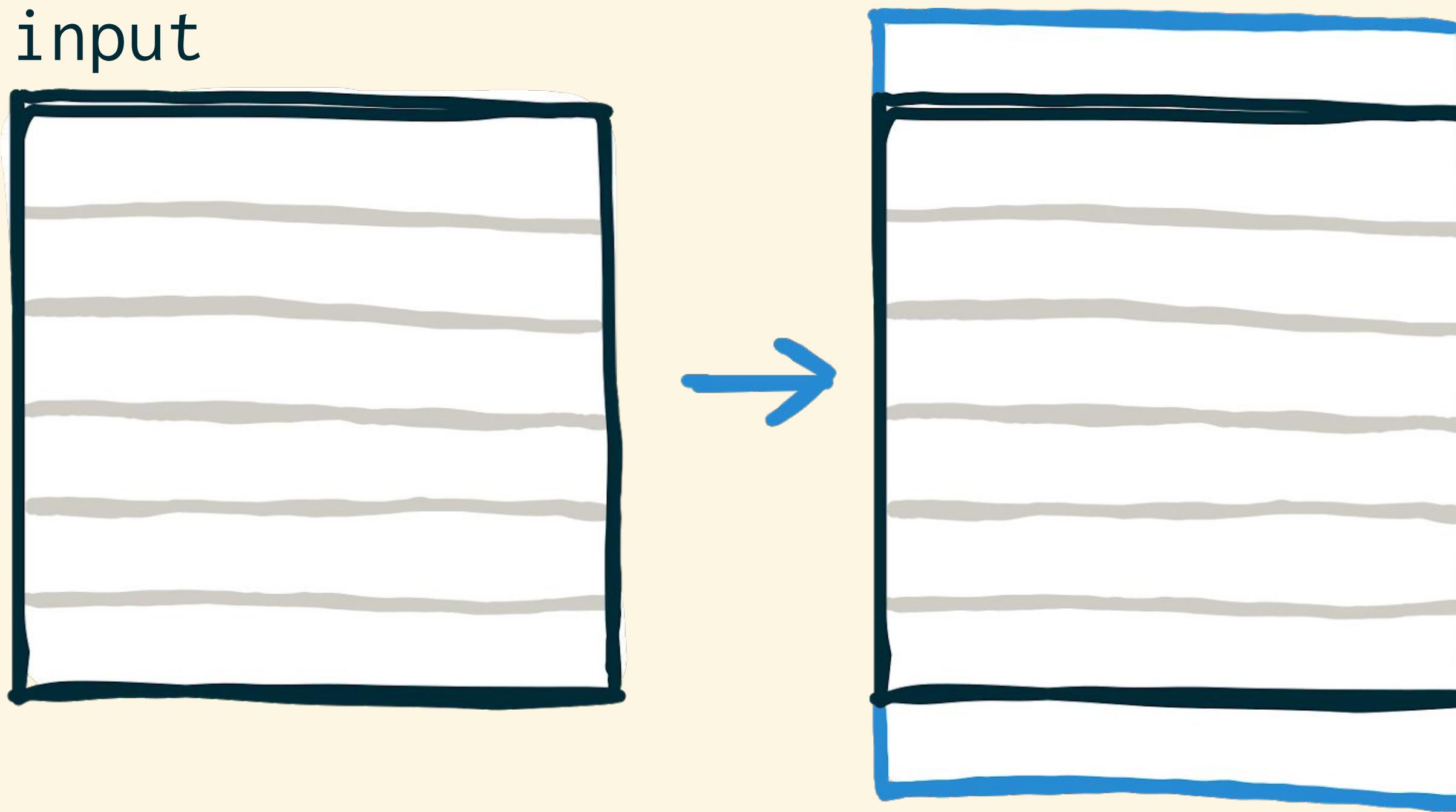
MULTIDIMENSIONAL BOUNDARY HANDLING USING pad_2

input



$pad_2 =$

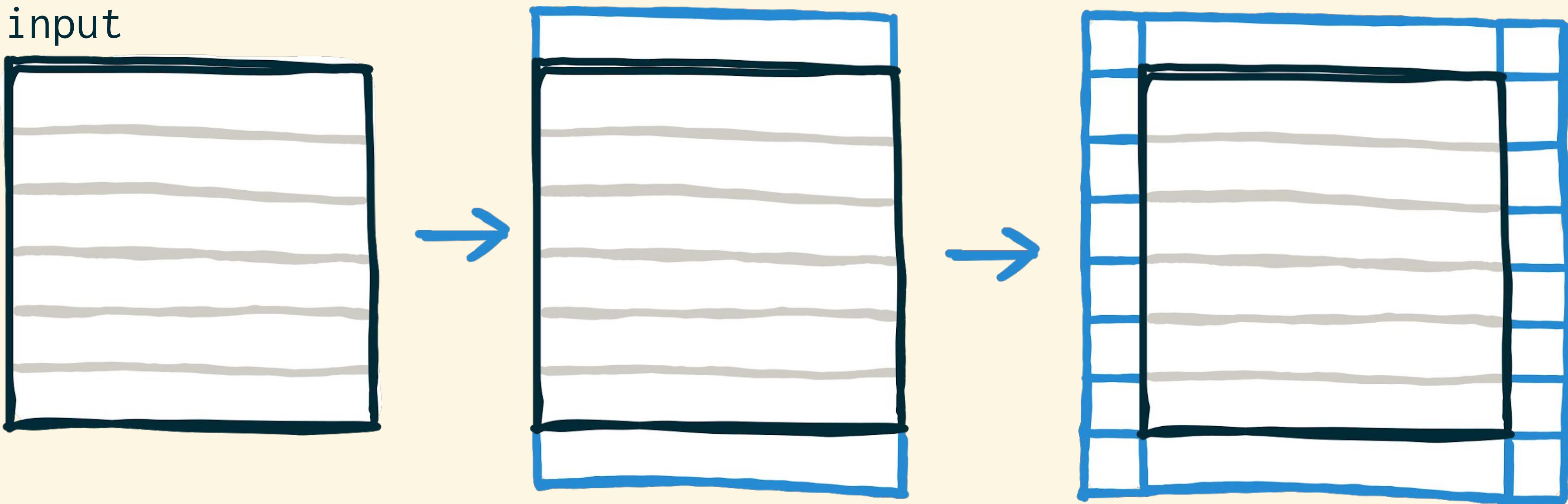
MULTIDIMENSIONAL BOUNDARY HANDLING USING pad_2



$pad_2 =$

$pad(1, r, b, \text{input})$

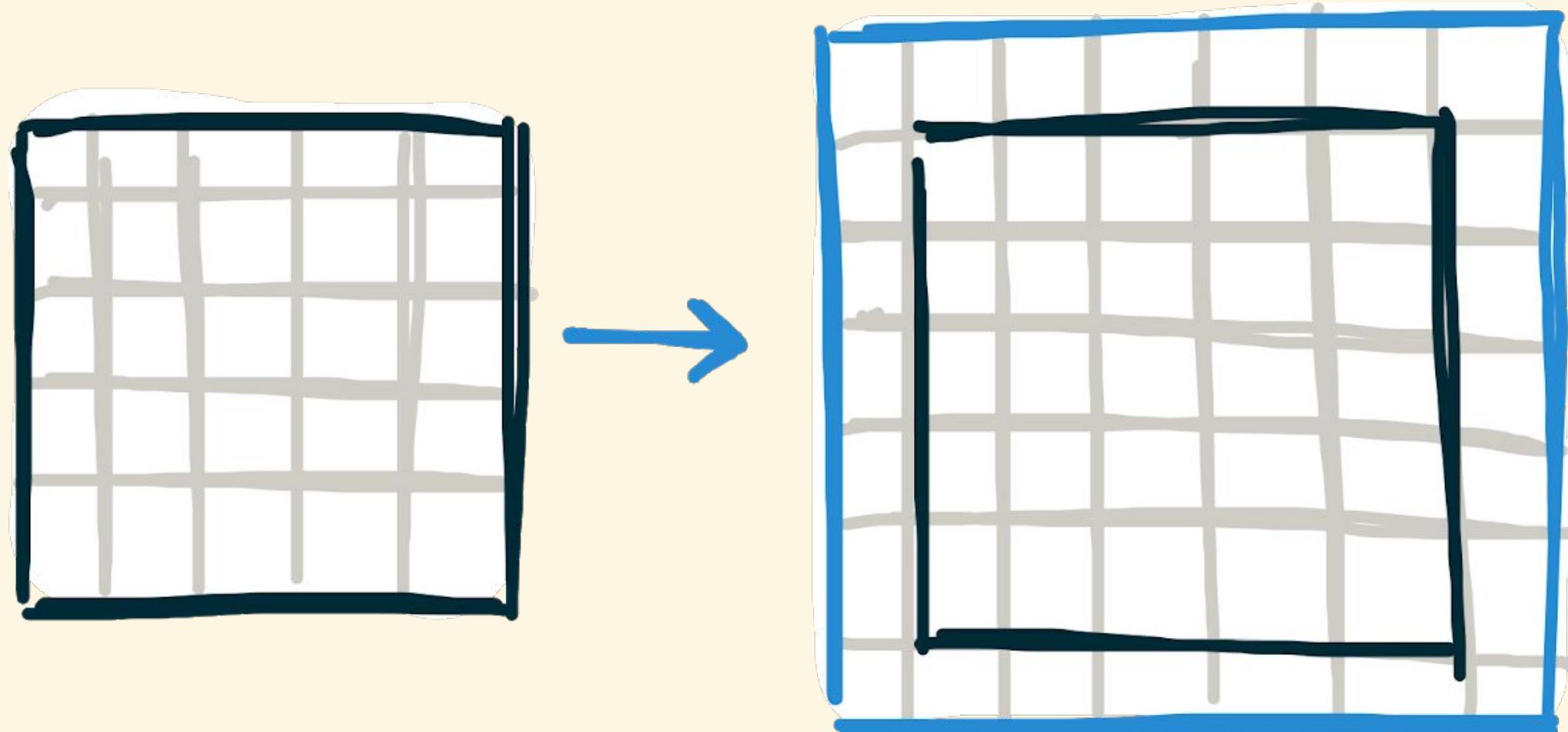
MULTIDIMENSIONAL BOUNDARY HANDLING USING pad_2


$$pad_2 = map(pad(1, r, b, pad(1, r, b, input)))$$

MULTIDIMENSIONAL STENCIL COMPUTATIONS

are expressed as compositions of intuitive, generic 1D primitives

Decompose to Re-Compose

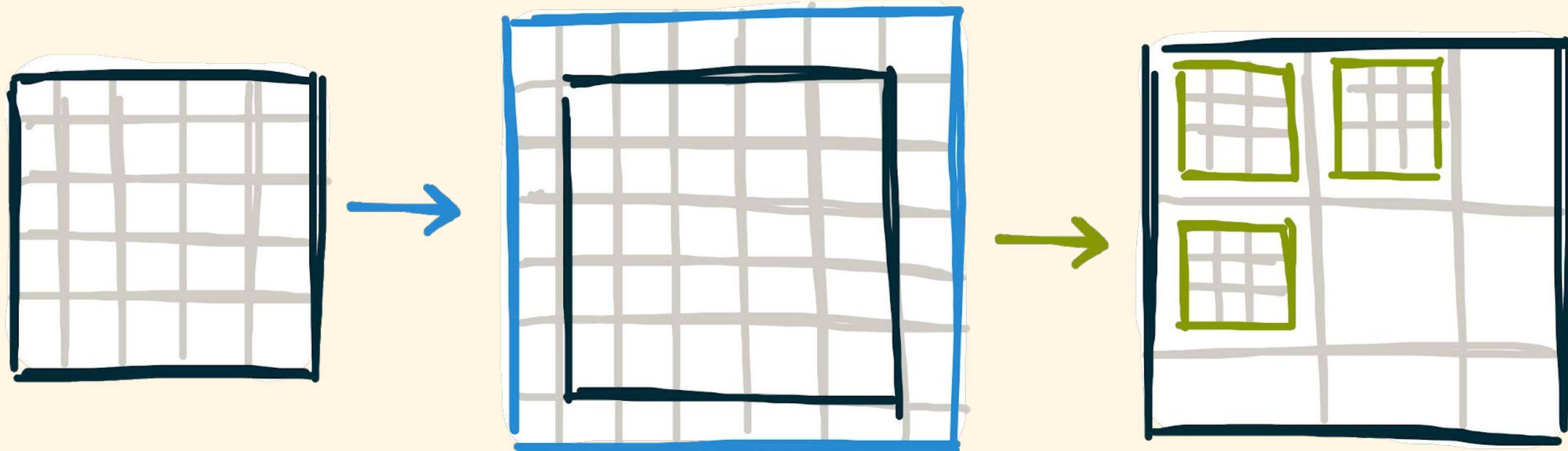


$\text{pad}_2(1, 1, \text{clamp}, \text{input})$

MULTIDIMENSIONAL STENCIL COMPUTATIONS

are expressed as compositions of intuitive, generic 1D primitives

Decompose to Re-Compose

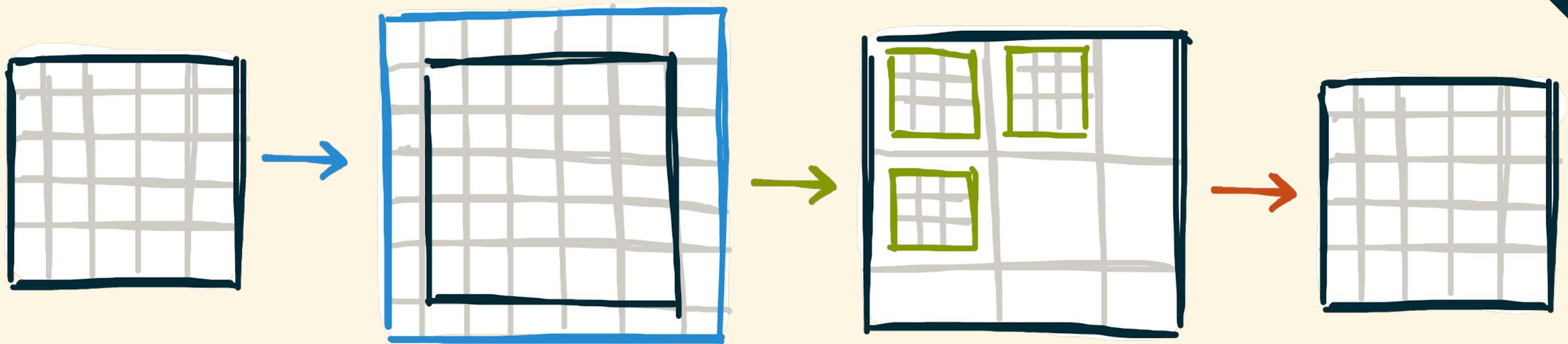


slide₂(3, 1, pad₂(1, 1, clamp, input))

MULTIDIMENSIONAL STENCIL COMPUTATIONS

are expressed as compositions of intuitive, generic 1D primitives

Decompose to Re-Compose

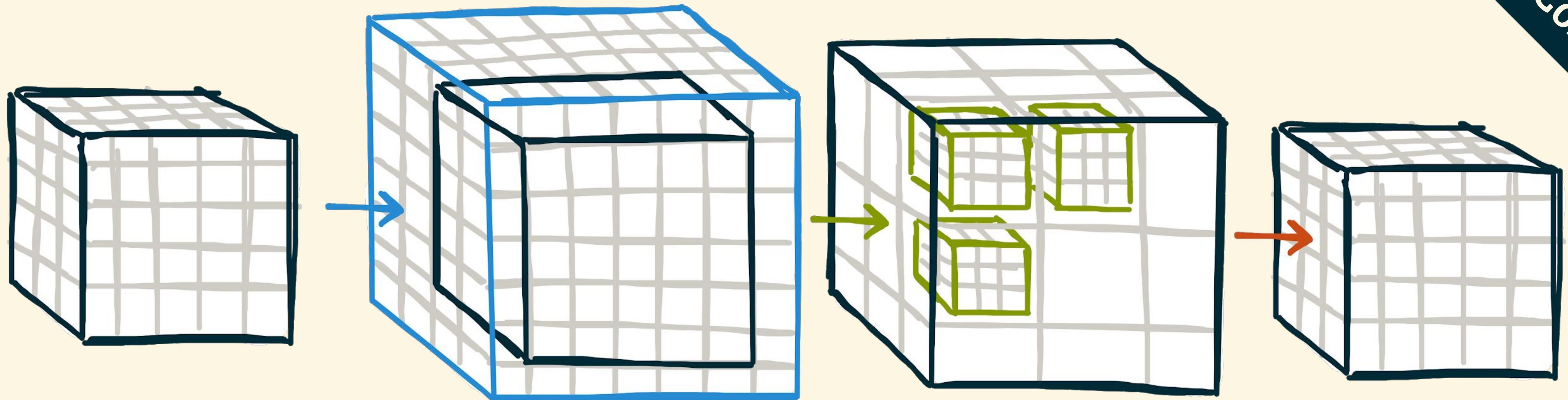


$\text{map}_2(\text{sum}, \text{slide}_2(3, 1, \text{pad}_2(1, 1, \text{clamp}, \text{input})))$

MULTIDIMENSIONAL STENCIL COMPUTATIONS

are expressed as compositions of intuitive, generic 1D primitives

Decompose to Re-Compose

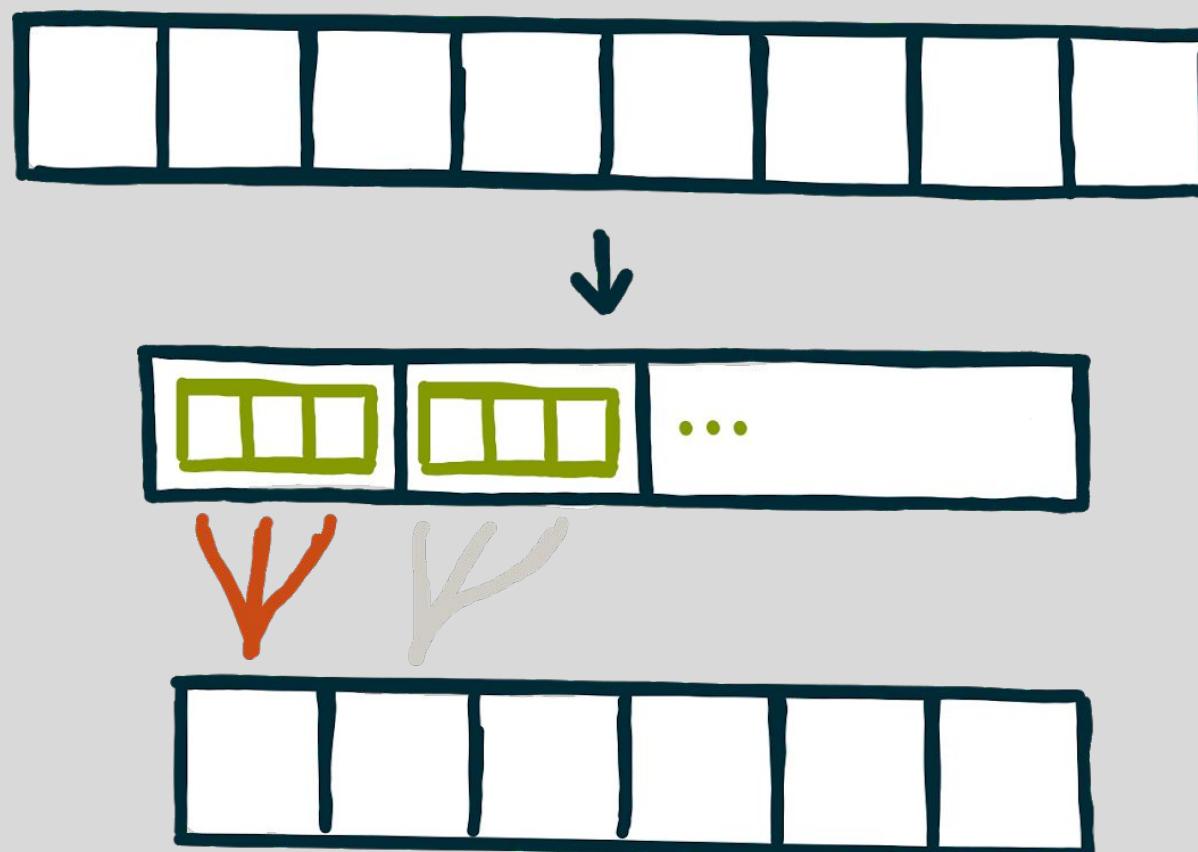


$\text{map}_3(\text{sum}, \text{slide}_3(3, 1, \text{pad}_3(1, 1, \text{clamp}, \text{input})))$

OVERLAPPED TILING AS A REWRITE RULE

overlapped tiling rule

map(f, slide(3,1,input))



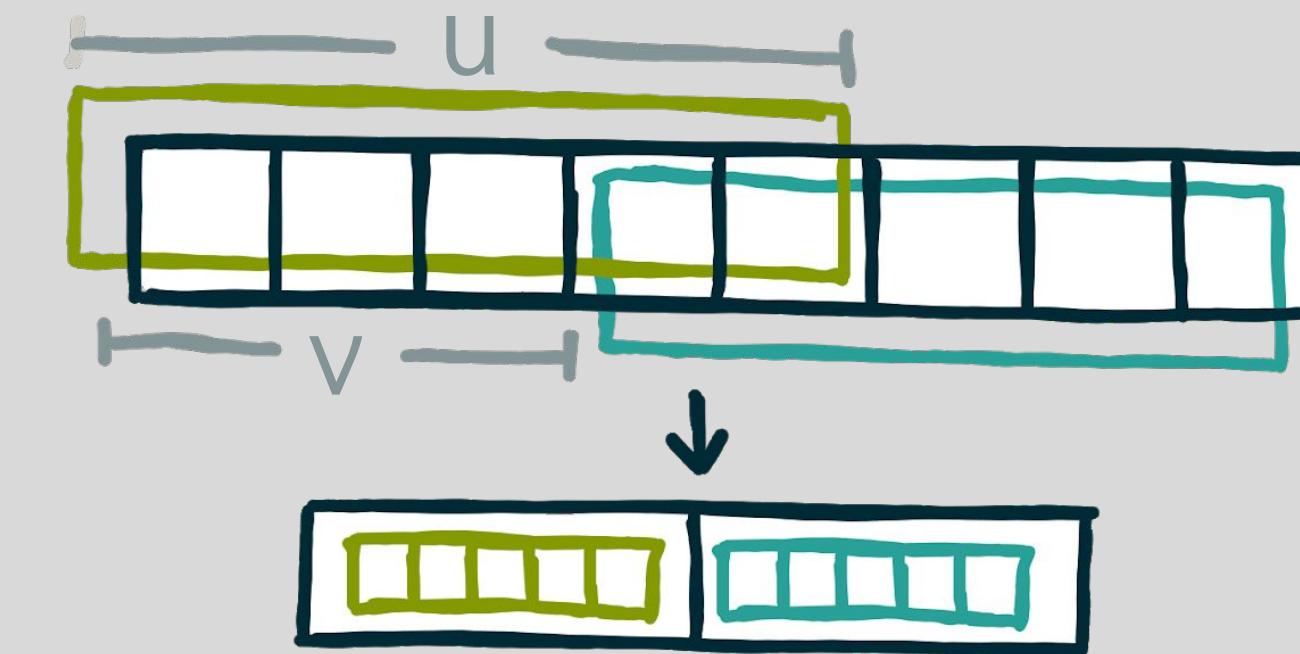
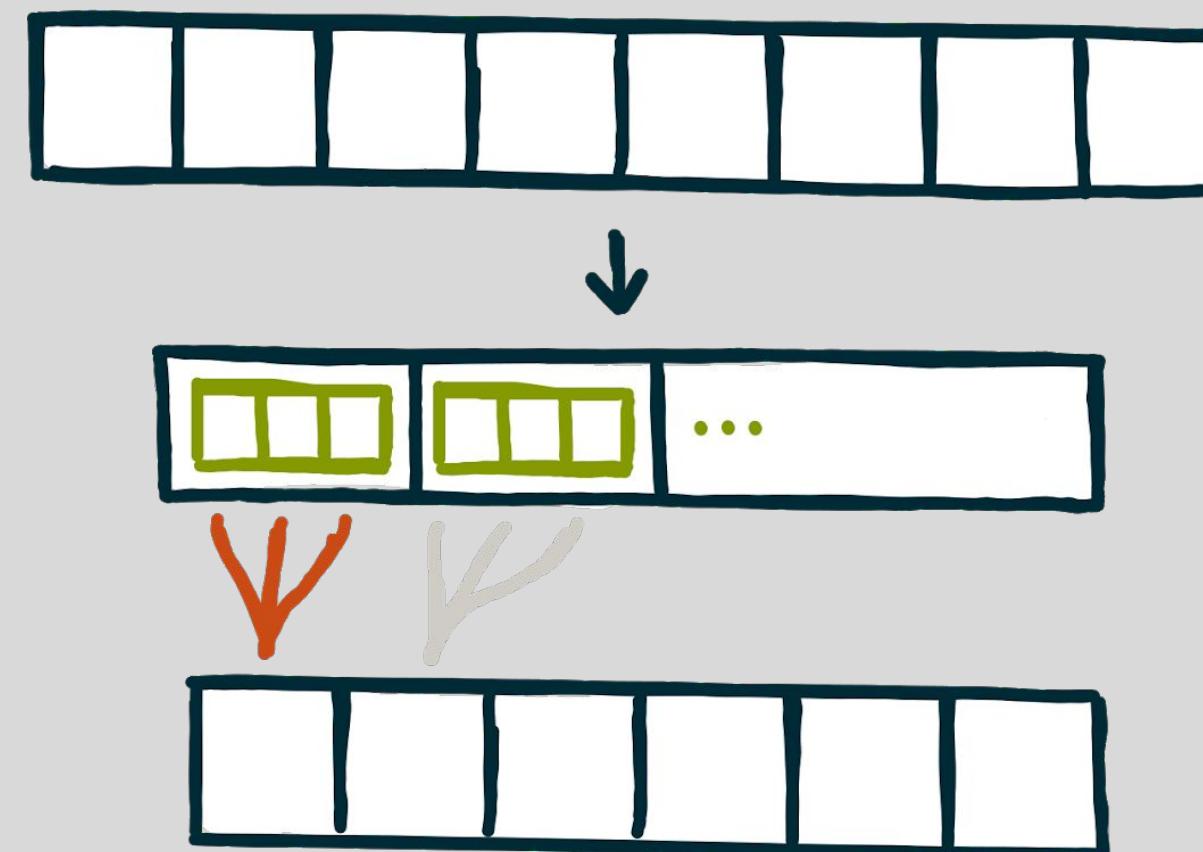
OVERLAPPED TILING AS A REWRITE RULE

overlapped tiling rule

map(f, slide(3,1,input))

⇒

slide(u,v,input)



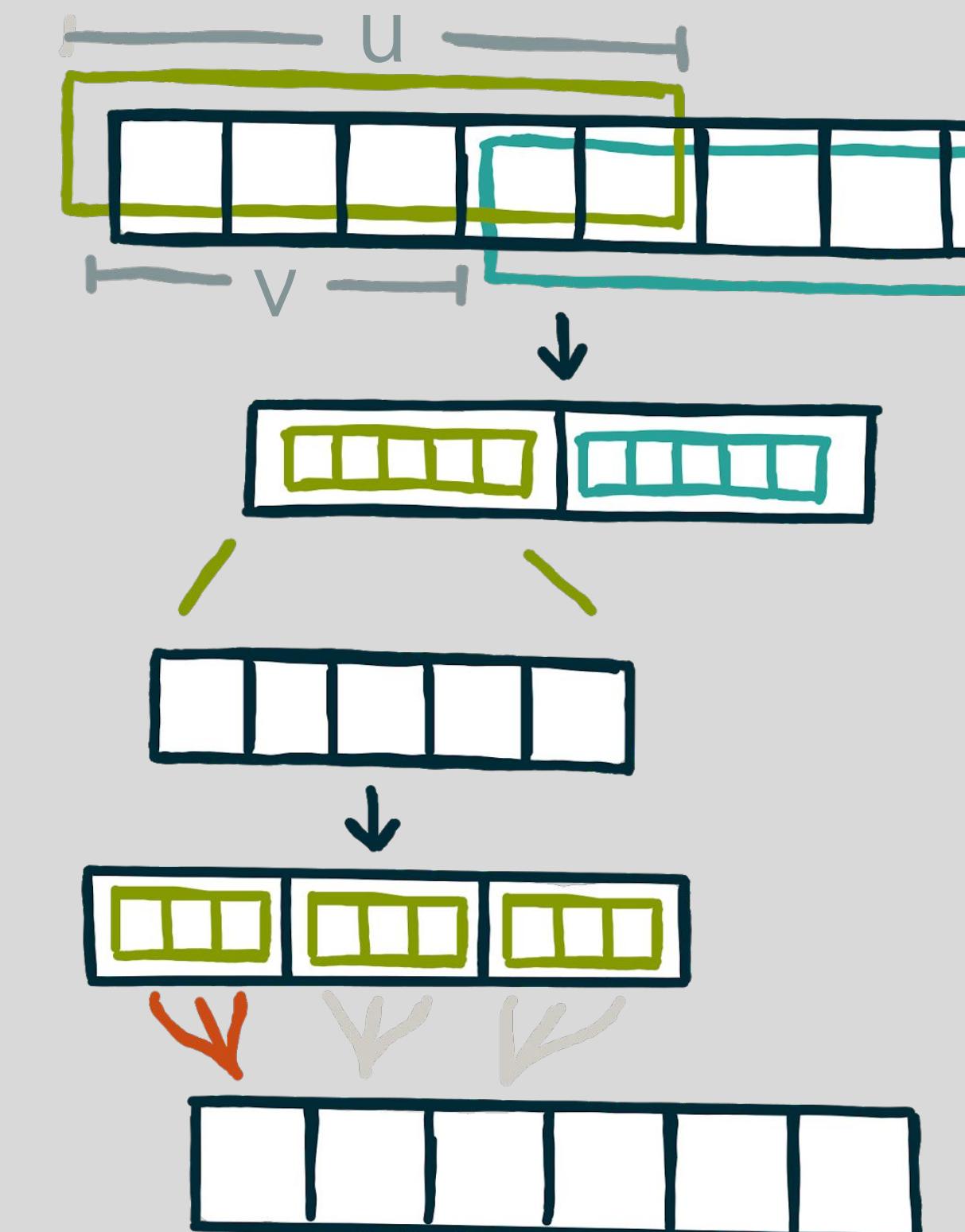
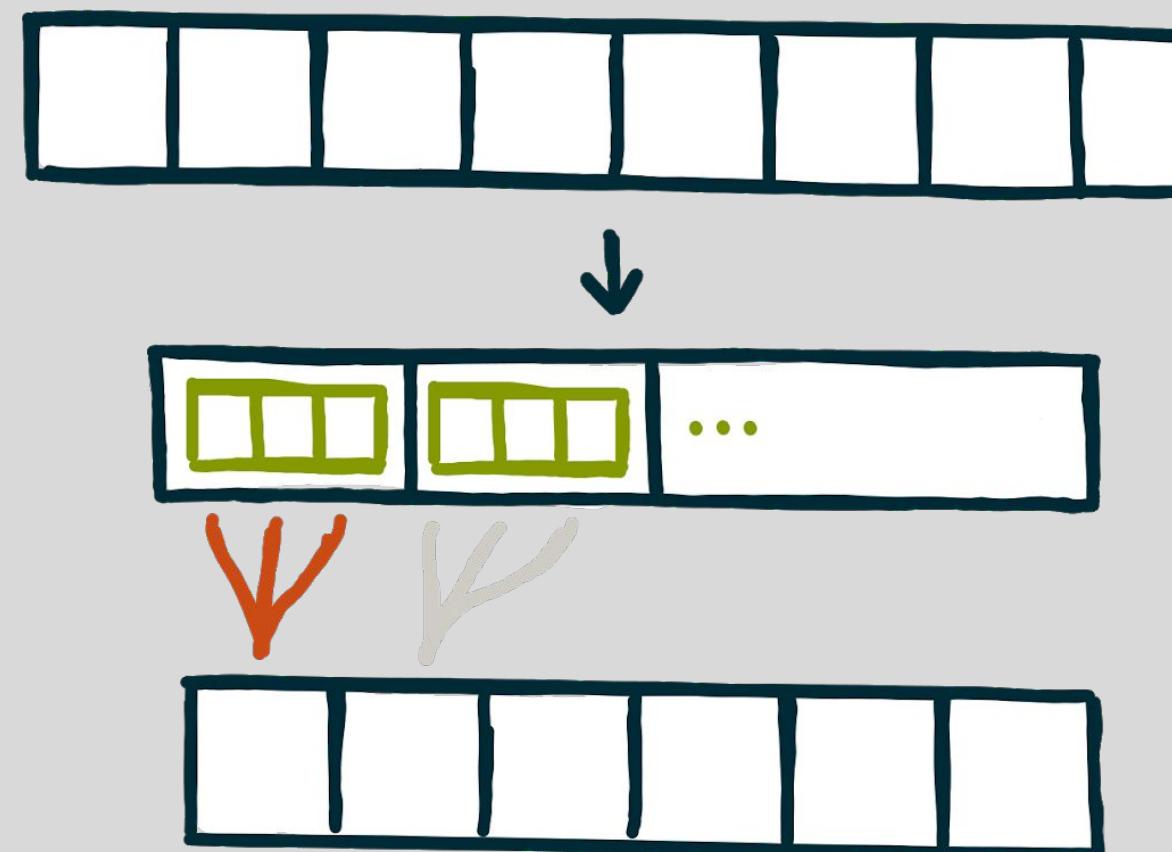
OVERLAPPED TILING AS A REWRITE RULE

overlapped tiling rule

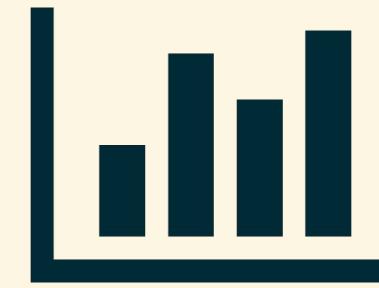
$\text{map}(f, \text{slide}(3,1,\text{input}))$

⇒

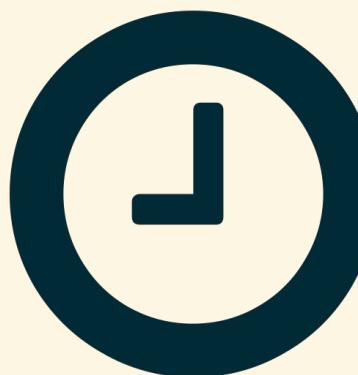
$\text{join}(\text{map}(\text{tile} \Rightarrow$
 $\text{map}(f, \text{slide}(3,1,\text{tile})),$
 $\text{slide}(u,v,\text{input})))$



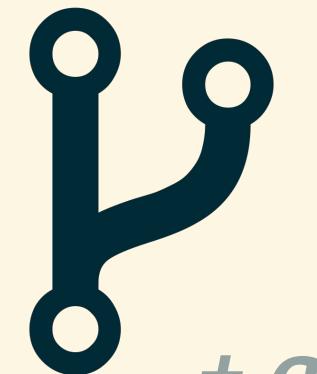
EXPERIMENTAL EVALUATION



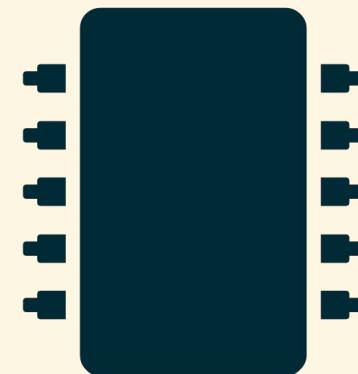
14 Benchmarks
*6 hand-optimized
8 polyhedral compilation*



< 3h Exploration
per benchmark



**up to 20 algorithmically
different variants**
+ auto-tuning of numerical parameters



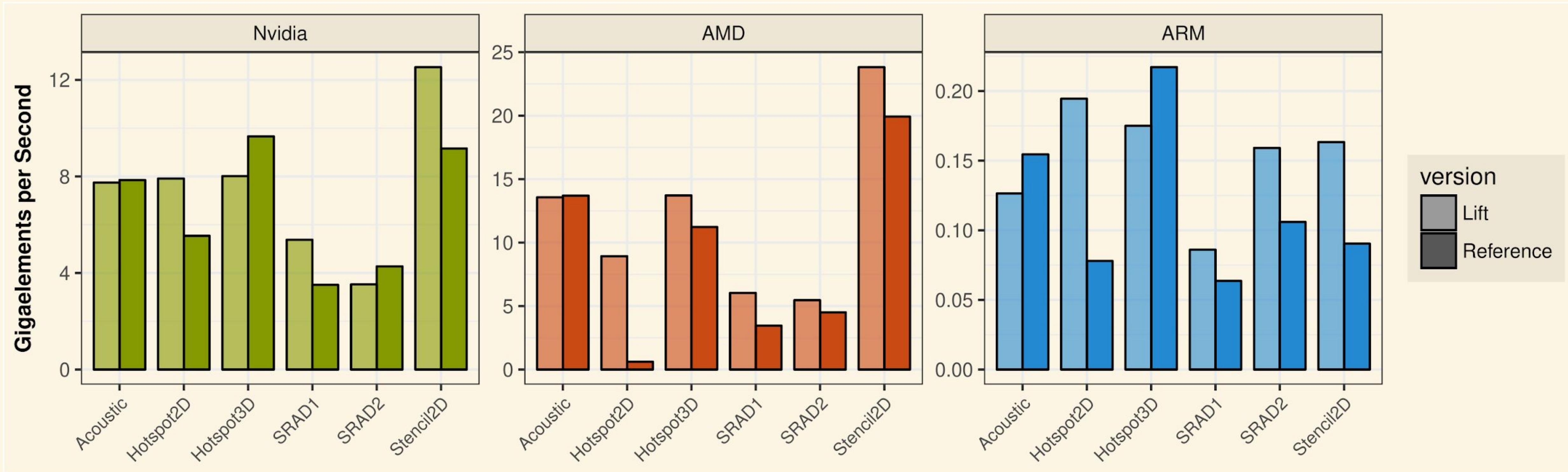
3 GPU Architectures
*2 Desktop GPUs
1 Mobile GPU*

Multicore
CPU →



COMPARISON WITH HAND-OPTIMIZED CODES

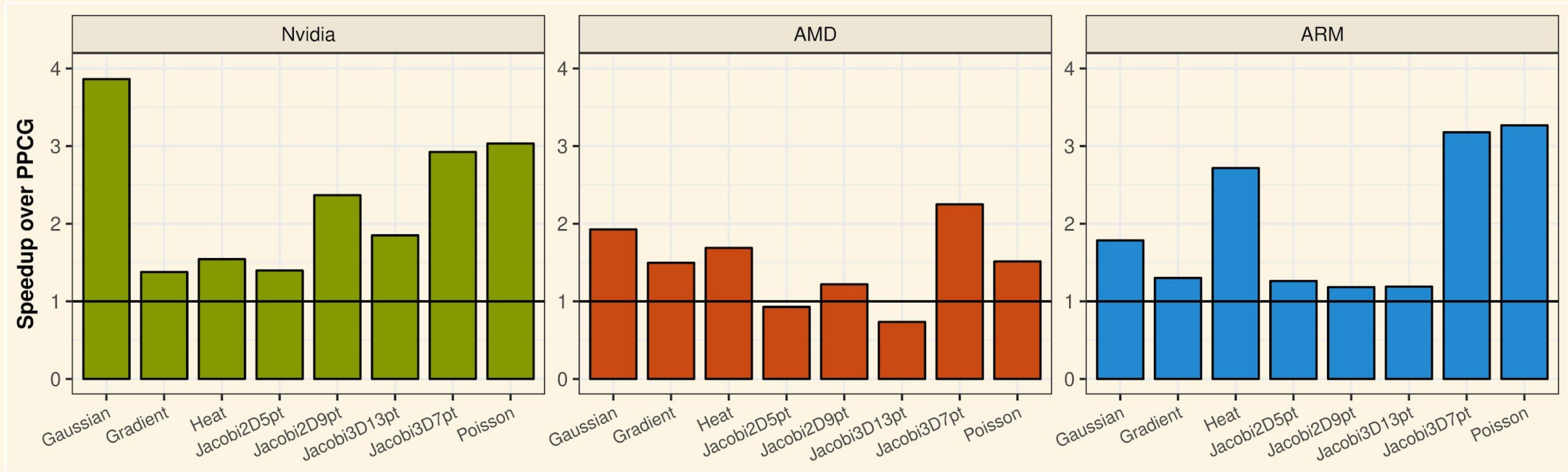
higher is better



Lift achieves the same performance
as hand optimized code

COMPARISON WITH POLYHEDRAL COMPILATION

higher is better



Lift outperforms state-of-the-art
optimizing compilers

LIFT IS OPEN SOURCE!



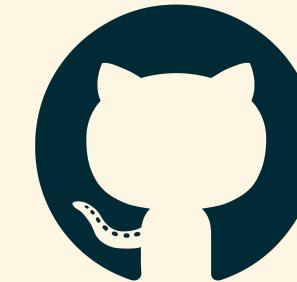
more info at:

lift-project.org

„ Paper



Artifacts



Source Code



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Larisa Stoltzfus

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Automatic Matching of Legacy Code to Heterogeneous APIs: An Idiomatic Approach

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Abstract

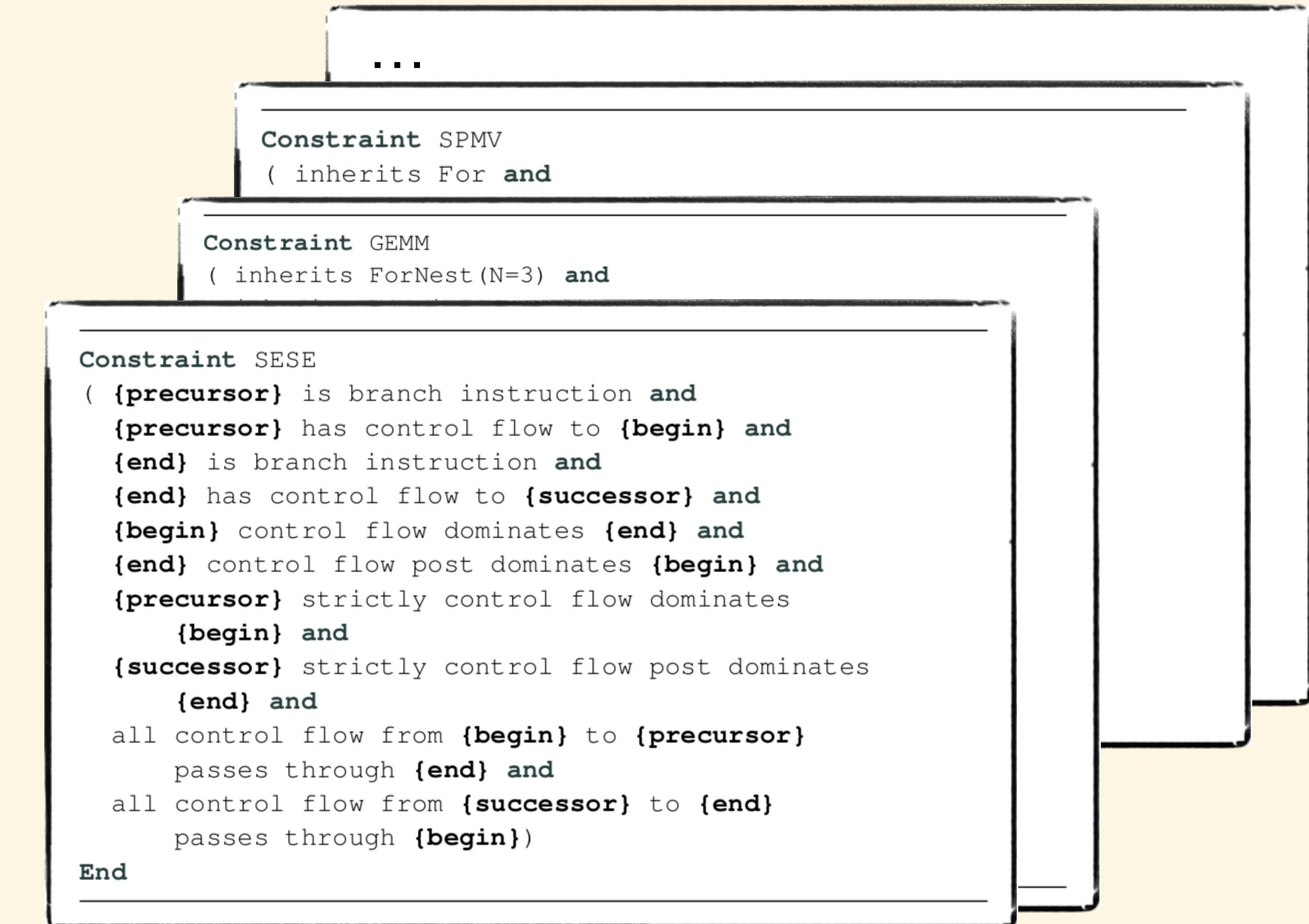
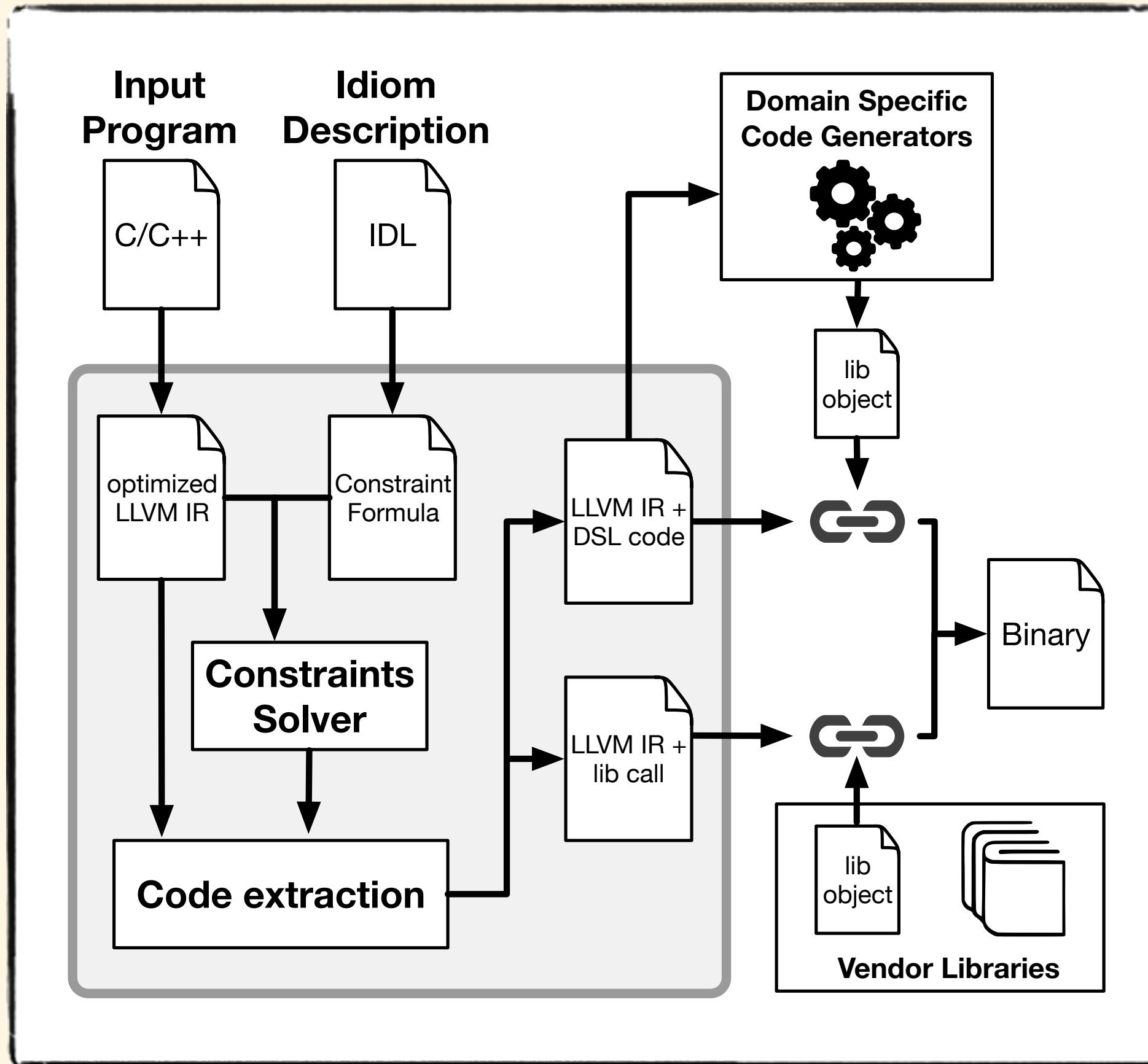
Heterogeneous accelerators often disappoint. They provide the prospect of great performance, but only deliver it when using vendor specific optimized libraries or domain specific languages. This requires considerable legacy code modifications, hindering the adoption of heterogeneous computing.

This paper develops a novel approach to automatically

1 Introduction

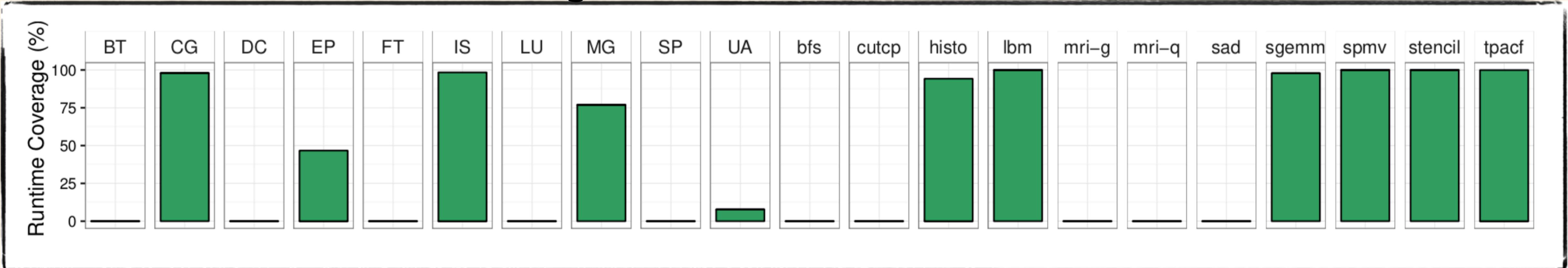
Heterogeneous accelerators provide the potential for great performance. However, achieving that potential is difficult. General purpose languages such as OpenCL [36] provide portability, but the achieved performance often disappoints [29]. This shortfall has led vendors to deliver specialized libraries to bridge the gap [2]. Alternatively, domain specific

IDIOM DETECTION VIA CONSTRAINT LANGUAGE

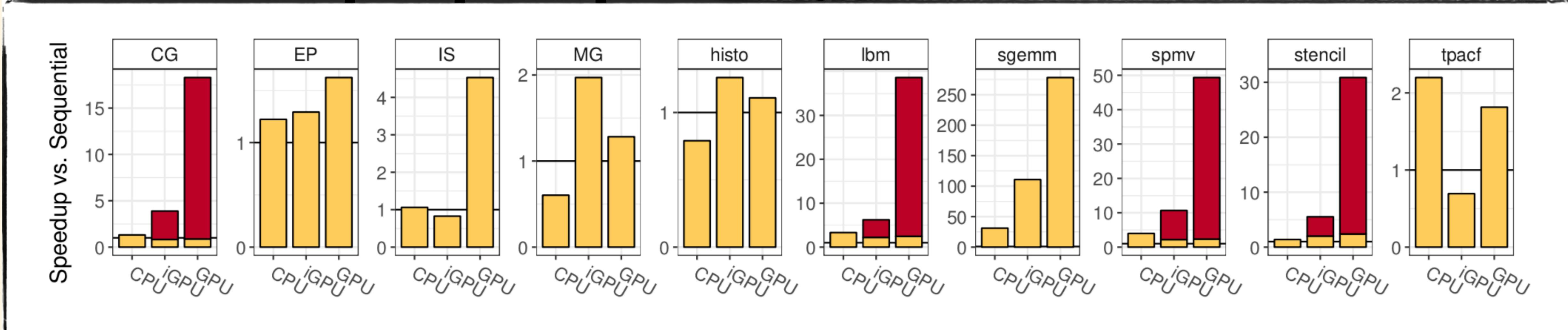


PERFORMANCE RESULTS

Runtime Coverage of detected Idioms (NAS PB + Parboil)



Speedup vs. Sequential (using BLAS, Halide, Lift as backends)



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 @LIFTlang

**INSPIRING
PEOPLE**

#UofGWorldChangers



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