

# Automatic Optimization Of Arrays In Affine Loop-Nests

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## Outline

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- 1 Introduction
- 2 Intra-Array Storage Optimization Problem
- 3 Conflicts, Conflict Satisfaction
- 4 Experimental Evaluation
- 5 Summary

## Storage Optimization

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**Basic Goal** Reuse memory locations for values without overlapping lifetimes

- Reuse within a given array or across different arrays
- Crucial for data-intensive programs
  - run larger problem size with a fixed amount of main memory
  - stencils, image processing applications, DSL compilers
    - affine loop-nests

## Contracting A Particular Array

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```

for(t=1;t<=N;i++)
for(i=1;i<=N;i++)
/*S*/ A[t,i]=f(A[t-1,i-1]
               +A[t-1,i]
               +A[t-1,i+1]);

```

(a) 1-d stencil using  $N^2$  storage

**Dependences**  $(1, -1)$ ,  $(1, 0)$  and  $(1, 1)$

**Live-out**  $A[T, *]$

```

for(t=1;t<=N;i++)
for(i=1;i<=N;i++)
/*S*/ A[t%2,i]=f(A[(t-1)%2,i-1]
                 +A[(t-1)%2,i]
                 +A[(t-1)%2,i+1]);

```

(b) Array contracted to size  $2 \times N$

```

for(t=1;t<=N;i++)
for(i=1;i<=N;i++)
/*S*/ A[(i-t+N)%(N+1)]=f(A[(i-t+N)%(N+1)]
                         +A[(i-t+1+N)%(N+1)]
                         +A[(i-t+2+N)%(N+1)]);

```

(c) Array contracted to  $N+1$  cells.

Storage optimal!

## Reuse Across Arrays - Image Processing Applications

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```

#define isbound(i,j) (i==0)||(i==(N-1))
                    ||(j==0)||(j==(N-1))
for(int i=0; i<N; ++i)
    for(int j=0; j<N; ++j)
/*S0*/ A0[i,j]= isbound(i,j) ? a[i,j]
        :a[i,j]+(a[i-1,j]+a[i+1,j]
        +a[i,j-1]+a[i][j+1]);

for(int i=0; i<N; ++i)
    for(int j=0; j<N; ++j)
/*S1*/ A1[i,j]=isbound(i,j) ? A0[i,j]
        :A0[i,j]+(A0[i-1,j]+A0[i+1,j]
        +A0[i,j-1]+A0[i,j+1]);

for(int i=0; i<N; ++i)
    for(int j=0; j<N; ++j)
/*S2*/ A2[i,j]=!isbound(i,j) ? A1[i][j]
        :A1[i,j]+(A1[i-1,j]+A1[i+1,j]
        +A1[i,j-1]+A1[i,j+1]);

```

$$S_0 : A_0[i, j] \rightarrow A[(i + 3) \bmod (N + 2), j \bmod N]$$

$$S_1 : A_1[i, j] \rightarrow A[(i + 1) \bmod (N + 2), j \bmod N]$$

$$S_2 : A_2[i, j] \rightarrow A[(i - 1) \bmod (N + 2), j \bmod N]$$

(b) The storage mapping enabling inter-array reuse. Overall storage requirement is reduced from  $3N^2$  to  $N^2 + N$ .

(a)  $A_0, A_1$  are just intermediate arrays which are not live-out.

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## General Approach To The Problem

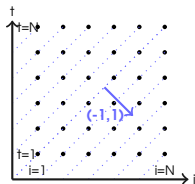
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- Contract array along one or more directions to fixed sizes
  - Step 1: Determine good directions**
    - Canonical directions need not be good ones
    - Affects dimensionality and storage size
    - Can be the difference between  $N^2$ ,  $2N$ ,  $N + 1$  storage for a given  $N \times N$  array
  - Step 2: Minimize the array size along these directions**
    - Thoroughly studied by Lefebvre and Feautrier(1998)
- No good heuristics for **Step 1**
- Darte et al(2005), Lefebvre and Feautrier(1998)
  - work with canonical basis or assume that directions are given.

# An Array Space Partitioning Approach

## Storage Partitioning Hyperplane

Partitions the iteration space such that each partition uses a single memory location.



Hyperplane  $(-1, 1)$  creating  $(2N - 1)$  partitions.

**Good Directions?** Storage hyperplanes with good orientations

**Contraction?** Minimize the number of partitions created

- Affects the resulting storage size

**Dimensionality?** Number of storage hyperplanes found

- Iteratively found until some criterion is met



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## Conflicts Within An Array Space

### Conflicting indices $\vec{i} \bowtie \vec{j}$

Two array indices  $\vec{i}, \vec{j}$ , ( $\vec{i} \neq \vec{j}$ ), conflict with each other and the conflict relation  $\vec{i} \bowtie \vec{j}$  holds if the corresponding array elements are simultaneously live under the given schedule  $\theta$ .

```
for(i=2;i<=n;i++)
  fib[i]=fib[i-1]+fib[i-2];
  result=fib[n];
```

(a) Before contraction

```
for(i=2;i<=n;i++)
  fib[i%2]=fib[(i-1)%2]+fib[(i-2)%2];
  result=fib[n%2];
```

(a) After contraction

**Dependences?**  $(i-2) \rightarrow_{RAW} i, (i-1) \rightarrow_{RAW} i$

**Live Out?**  $fib(n)$

**Conflicts?** Each array index conflicts with its adjacent index:  $i \bowtie (i-1)$

$\Rightarrow$  fib can be contracted to a 2-element array  
 Modulo storage mapping:  $fib[i] \rightarrow fib[i \bmod 2]$

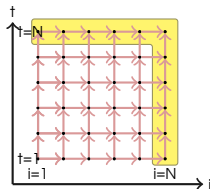
## Array Space Partitioning Through Conflict Satisfaction

```

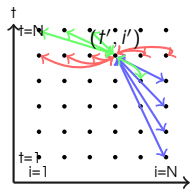
for (t=1; t<=N; i++)
  for (i=1; i<=N; i++)
    /*S*/ A[t,i]=A[t,i-1]+A[t-1,i];
for (i=1; i<=N; i++)
  result=result+A[i,N]+A[N,i];

```

(a) A producer-consumer loop-nest



The flow dependences.  
Live-out portion in yellow.



Conflicts in different  
conflict polyhedra.

### Conflict Satisfaction

A conflict  $\vec{i} \bowtie \vec{j}$  is said to be satisfied by a hyperplane  $\vec{\Gamma}$  if  $\vec{\Gamma} \cdot \vec{i} - \vec{\Gamma} \cdot \vec{j} \neq 0$ .

- Conflicting indices must be mapped to different partitions

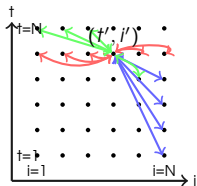
## Example Revisited

```

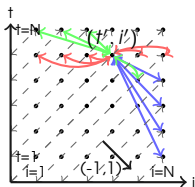
for (t=1; t<=N; i++)
  for (i=1; i<=N; i++)
    /*S*/ A[t,i]=A[t,i-1]+A[t-1,i];
for (i=1; i<=N; i++)
  result=result+A[i,N]+A[N,i];

```

(a) A producer-consumer loop-nest



Conflicts in different  
conflict polyhedra



$(-1, 1)$  satisfies all conflicts

Candidate hyperplanes ...

$(1, 0)$  Satisfies only blue, green conflicts

$(0, 1)$  Satisfies only red, green conflicts

$(-1, 1)$  Satisfies all conflicts creating  $2N - 1$  partitions

$(-2, 1)$  Satisfies all conflicts creating  $3N - 2$  partitions

$(-3, 1)$  Satisfies all conflicts creating  $4N - 2$  partitions

**Modulo Storage Mapping**  $A[t, i] \rightarrow A[(i - t) \bmod (2N - 1)]$

Storage as well as dimension optimal!

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## Storage Mappings Obtained Using SMO tool

**Table:** Our approach (SMO) compared to the baseline (Lefebvre and Feautrier(1998)) with  $B$  being the loop blocking factor

Benchmark		Modulo storage mapping	Reduction (approx.)	SMO time
produce-consume	baseline	$A[t \bmod N, i \bmod N]$	$\frac{N}{2}$	0.17s
	SMO	$A[(i - t) \bmod (2N - 1)]$		
blur-interleaved	baseline	$blurx[y \bmod 3, x \bmod N]$	1.5	0.14s
	SMO	$blurx[(2x - y) \bmod (2N + 1)]$		
blur-tiled	baseline	$A[tx, ty, x \bmod B, y \bmod B]$	$\frac{B}{3}$	0.11s
	SMO	$A[tx, ty, (y - 2x) \bmod (3B - 2)]$		
harris-corner-tiled	baseline	$sobel[tx, ty, x \bmod B, y \bmod B]$	$\frac{B}{3}$	0.12s
	SMO	$sobel[tx, ty, (y - 2x) \bmod (3B - 2)]$		
unsharp-mask-tiled	baseline	$A[z, tx, ty, x \bmod B, y \bmod B]$	$\frac{B}{5}$	0.82s
	SMO	$A[z, tx, ty, (y - 4x) \bmod (5B - 4)]$		
LBM-D2Q9	baseline	$A[t \bmod 2, i \bmod N, j \bmod N]$	2	0.61s
	SMO	$A[(i - 2t) \bmod (N + 2), j \bmod N]$		
LBM-D3Q19	baseline	$A[t \bmod 2, i \bmod N, j \bmod N, k \bmod N]$	2	3.32s
	SMO	$A[(i - 2t) \bmod (N + 2), j \bmod N, k \bmod N]$		
LBM-D3Q27	baseline	$A[t \bmod 2, i \bmod N, j \bmod N, k \bmod N]$	2	3.33s
	SMO	$A[(i - 2t) \bmod (N + 2), j \bmod N, k \bmod N]$		
diamond-tile	baseline	$A_B[tt \bmod B, ii \bmod (2B - 1)]$	$\frac{B}{3}$	0.44s
	SMO	$A_B[(tt - 3ii) \bmod (6B - 5)]$		
stencil-1d-llelogram-tile	baseline	$A_B[tt \bmod B, ii \bmod B]$	$\frac{B}{3}$	0.29s
	SMO	$A_B[(tt - ii) \bmod (3B - 2)]$		
stencil-1d-hex-tile	baseline	$A_B[tt \bmod B, ii \bmod (3B - 2)]$	$\frac{B}{3}$	1.15s
	SMO	$A_B[(-tt + 3ii) \bmod (9B - 7)]$		

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## Summary

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- Intra-array and inter-array storage reuse
  - Array space partitioning by finding good storage hyperplanes
- Heuristic driven by a fourfold objective function.
  - greedy conflict satisfaction (impacts dimensionality)
  - minimizes the partitions (minimizes dimension sizes)
  - factors in inter-statement conflicts (exploits inter-statement reuse)
- Developed SMO tool—a polyhedral storage optimizer.
  - Effective on several real-world examples.
  - Storage mappings which are asymptotically better than those by existing techniques.



## Acknowledgements And Publications

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- INRIA (France) for an associate team award POLYFLOW
  - National Instruments
- 
- 1 Somashekaracharya G. Bhaskaracharya, Uday Bondhugula, Albert Cohen *Automatic Storage Optimization for Arrays*, ACM Transactions on Programming Languages and Systems (TOPLAS), accepted in 2015.
  - 2 Somashekaracharya G. Bhaskaracharya, Uday Bondhugula, Albert Cohen *SMO: An Integrated Approach to Intra-Array and Inter-Array Storage Optimization*, ACM SIGPLAN-SIGACT Symposium on Principles of Programming Languages (POPL), St.Petersberg, USA, pages 526 - 538, Jan 2016.