

PLUTO+: NEAR-COMPLETE MODELING OF AFFINE TRANSFORMATIONS FOR PARALLELISM AND LOCALITY

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- 2 PLUTO
- 3 MOTIVATION FOR NEGATIVE COEFFICIENTS
- 4 PLUTO+
- 5 EXPERIMENTAL RESULTS
- 6 RELATED WORK

AFFINE TRANSFORMATIONS

- Examples of affine functions of i, j : $i + j, i - j, i + 1, 2i + 5$
- Not affine: $ij, i^2, i^2 + j^2, a[j]$

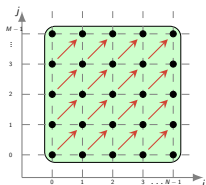


FIGURE: Iteration space

```
for (i = 0; i < N; i++){
  for (j = 0; j < M; j++){
    A[i+1][j+1] = f(A[i][j]);
  }
}
```

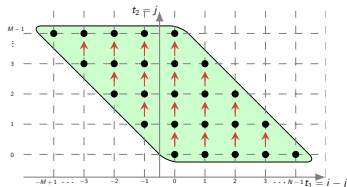


FIGURE: Transformed space

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  for (t2=max(0,-t1); t2<=min(M-1,N-1-t1); t2++){
    A[t1+t2+1][t2+1]=f(A[t1+t2][t2]);
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- Transformation: $(i, j) \rightarrow (i - j, j)$

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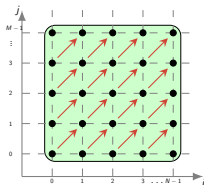


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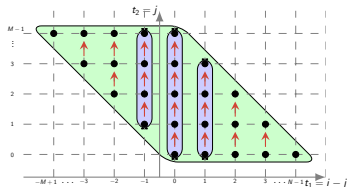


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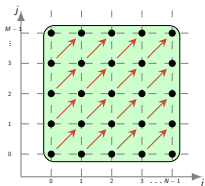


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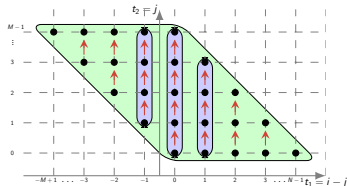


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- Affine transformations are attractive because:
 - Preserve **collinearity** of points and **ratio of distances** between points
 - Code generation with affine transformations has thus been studied well (CLoop, ISL, OMEGA+)

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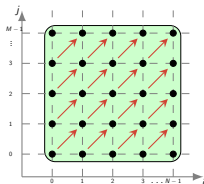


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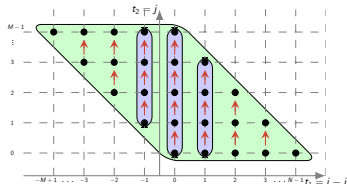


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- Affine transformations are attractive because:
 - Preserve **collinearity** of points and **ratio of distances** between points
 - Code generation with affine transformations has thus been studied well (CLoG, ISL, OMEGA+)
 - Model a very rich class of loop re-orderings
 - Useful for several domains like dense linear algebra, stencils, image processing pipelines, Lattice Boltzmann Method

AFFINE TRANSFORMATIONS

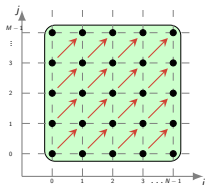


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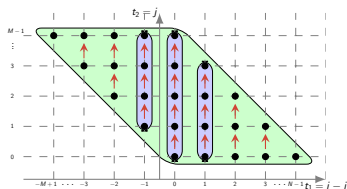


FIGURE: Transformed space

- Can express complex compositions of simpler transformations like permutation, skewing, reversal, scaling, shifting, tiling, fusion, distribution
- Affine transformations can improve **parallelism** and **locality** (Feautrier 1992, Lim and Lam 1997, Pluto 2008)

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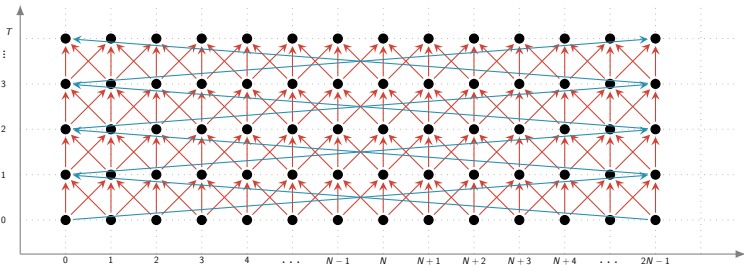
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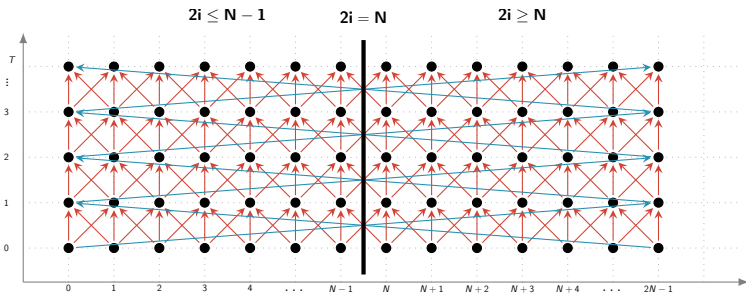
When transformation coefficients are negative, the above constraints miss useful solutions

- Near-neighbor dependences and some long wraparound dependences
- Applications in fluid simulations of infinite domains
- Periodic Lattice Boltzmann Methods (LBM) used in fluid dynamics, Swim (shallow water equations) fall into this category

TILING PERIODIC DOMAINS

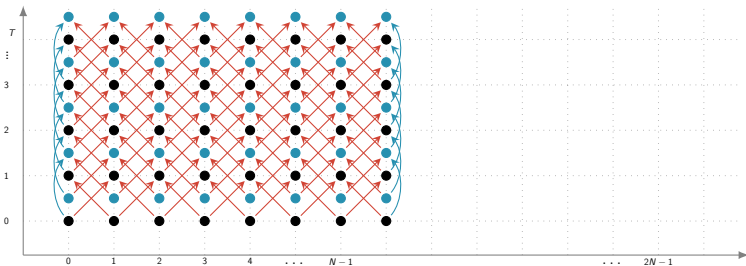


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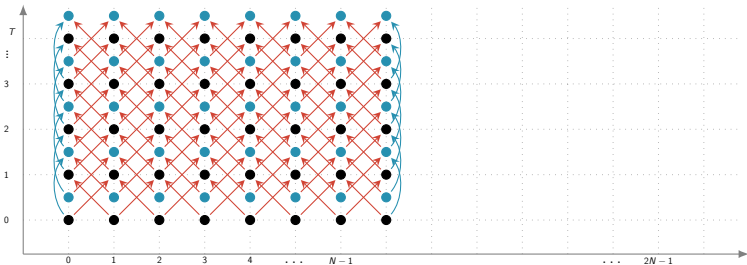
- Find a cut close to the mid point: ($2i = N$)

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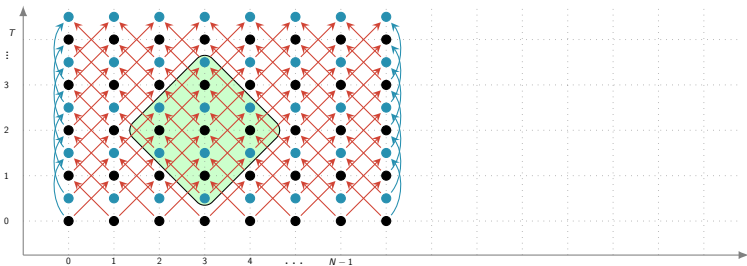
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 $(t, i) \rightarrow (t, N - i)$
- Now all dependences are **short**
- Tile the time dimension (parallelogram, diamond)

CHALLENGE 1: AVOIDING THE TRIVIAL SOLUTION

Transformation coefficients are **non-negative integers**.

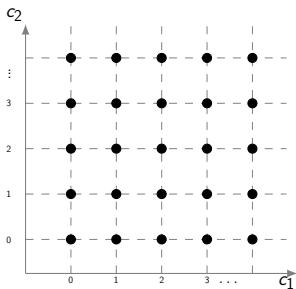
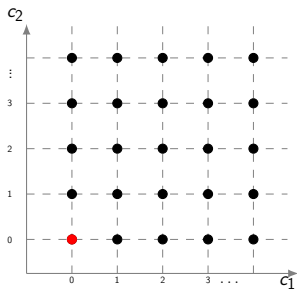


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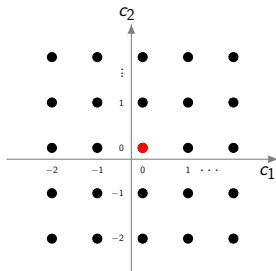


$$\begin{aligned}c_1, c_2 &\geq 0, \\c_1 + c_2 &\geq 1.\end{aligned}$$

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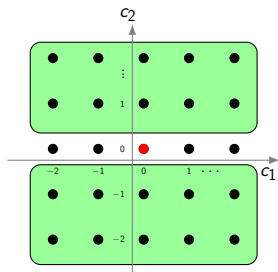
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With coefficients being **negative**, we may miss valid solutions. Eg:
 $c_1 = 1, c_2 = -1$

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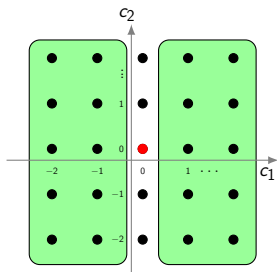


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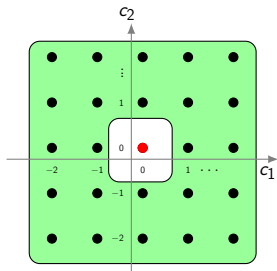


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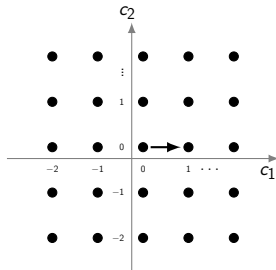
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This constraint results in a **non convex space**. Approach **does not scale**.

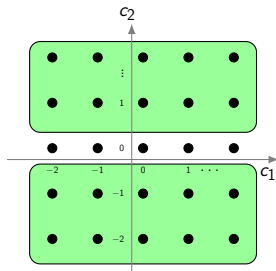
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Assume that c_0, c_1, c_2 are bounded by -4 and $+4$.

- c_0, c_1, c_2 can be considered to be in base 5
- If $(c_0, c_1, c_2) = \vec{0}$, then $5^2c_2 + 5c_1 + c_0 = 0$ and vice versa

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- Since the coefficients are in base 5,
the maximum value of $5^2c_2 + 5c_1 + c_0$ is $5^3 - 1 = 124$.
the minimum value of $5^2c_2 + 5c_1 + c_0$ is $1 - 5^3 = -124$.
- Hence, upper and lower bounds for $5^2c_2 + 5c_1 + c_0$ are known

- Introduce a decision variable to obtain a convex space representing the constraint on the absolute value
- We then have:

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Consider the following expressions where $\delta \in \{\mathbf{0}, \mathbf{1}\}$ is a decision variable:

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Negative half-space

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- Get a convex space corresponding to the absolute value using a decision variable.

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- Eg: Transformation obtained for LBM D2Q9 (periodic)

Original schedule	Transformed schedule
S1: (t, i, j)	S1: $(t-i, t+i, t+j)$
	S2: $(t+i-N, t-i+N, t+j)$
	S3: $(t-i, t+i, t-j+N)$
	S4: $(t+i-N, t-i+N, t-j+N)$

- Performance evaluation: Heat equation benchmarks with periodic conditions from Pochoir, Swim from SPEC 2000fp, several Lattice Boltzmann Method (LBM) simulations
- Comparison with Intel C compiler, Palabos in addition for LBM

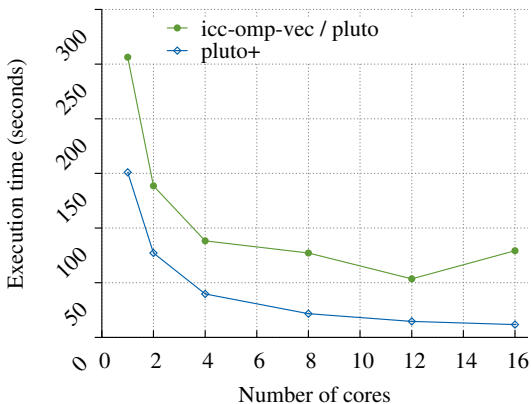
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- Analysis of impact on polyhedral automatic transformation time and overall compilation time

- Codes were run on Intel SandyBridge Machine with the following configuration.

	Intel Xeon E5-2680
Clock	2.7 GHz
Cores / socket	8
Total cores	16
L1 cache / core	32 KB
L2 cache / core	512 KB
L3 cache / socket	20 MB
Peak GFLOPs	172.8

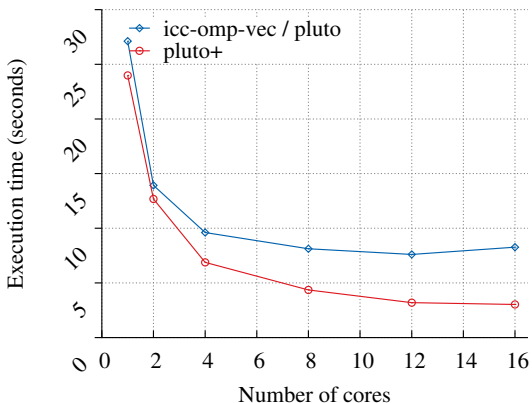
Compiler	Intel C compiler (icc) 14.0.1
Compiler flags	-O3 -xHost -ipo -restrict -fno-alias -ansi-alias -fp-model precise -fast-transcendentals
Linux kernel	3.8.0-44

PERFORMANCE: HEAT-2D



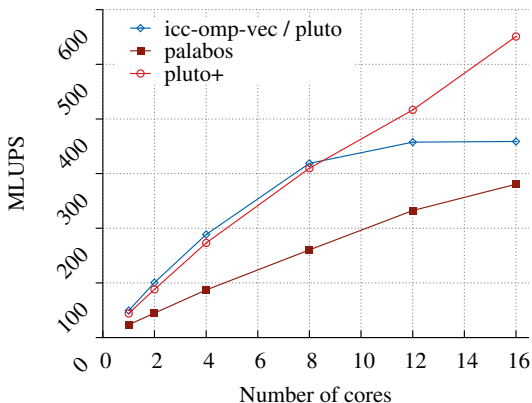
- Speedup of $1.7\times$ on single core and $6.7\times$ on 16 cores against icc

SWIM BENCHMARK (SPEC 2000FP)



- Speedup of $2.73\times$ over on 16 cores icc

PERFORMANCE: LBM D2Q9



- Speedup of $1.5\times$ over icc and $1.9\times$ over Palabos

- Comparison of Pluto+ with Pluto ($\frac{\text{Pluto+ time}}{\text{Pluto time}}$)

Benchmark	Auto-transformation	Total Time
Polybench	0.89	1.15
Heat equation	0.39	2.25
Swim (SPEC2000fp)	9.71	2.83
LBM benchmarks	0.49	1.80

- Pluto+ scales very well
- Improvement in auto-transformation time in several cases due to bounds on transformation coefficients
- In most cases, the increase in compile time was due to an increase in code generation time
- Total compilation time varied from 0.013s (jacobi-1d-imper) to 56.36s (LBM D3Q27)

PERFORMANCE SUMMARY

Benchmarks	Increase in compilation time	Speedup in running time
Heat equation	2.25×	2.91×
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- **Availability:** Code and benchmarks available at <http://mcl.csa.iisc.ernet.in/>

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- [R-STREAM compiler \[Encl. of Par. Computing 2011\]](#): larger number of decision variables per statement.

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Acknowledgments

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- Albert Cohen, INRIA, for discussions
- Intel Labs, India, for equipment and software used for experimentation
- Microsoft Research, India and ACM SIGPLAN for travel grants