

Loop Transformations: Convexity, Pruning and Optimization

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Compiler Optimizations for Performance

- ▶ **High-level loop transformations are critical for performance...**
 - ▶ Coarse-grain parallelism (OpenMP)
 - ▶ Fine-grain parallelism (SIMD)
 - ▶ Data locality (reduce cache misses)

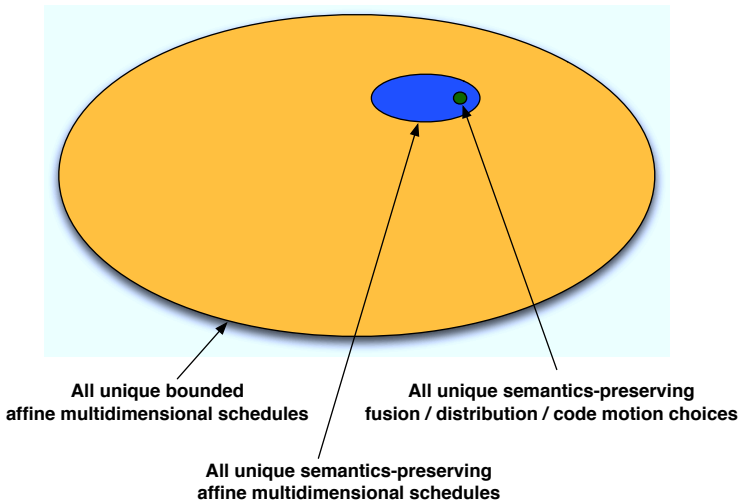
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- ▶ **... But deciding the best sequence of transformations is hard!**
 - ▶ Conflicting objectives: more SIMD implies less locality, etc.
 - ▶ It is machine-dependent and of course program-dependent
 - ▶ Expressive search spaces are required, but challenge the search!

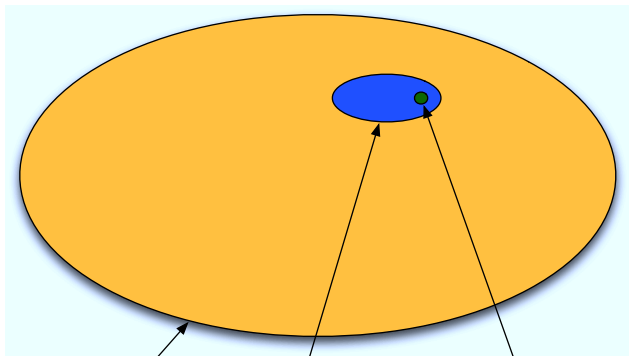
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 - ▶ Expressive search spaces are required, but challenge the search!
- ▶ **Our approach:**
 - ▶ **Convexity:** model optimization spaces as convex set (ILP, scan, project, etc.)
 - ▶ **Pruning:** make our spaces contain all and only semantically equivalent programs in our framework
 - ▶ **Optimization:** decompose in two more tractable sub-problems without any loss of expressiveness, empirical search + ILP models

Spaces of Affine Loop transformations



Spaces of Affine Loop transformations



All unique bounded
affine multidimensional schedules

All unique semantics-preserving
fusion / distribution / code motion choices

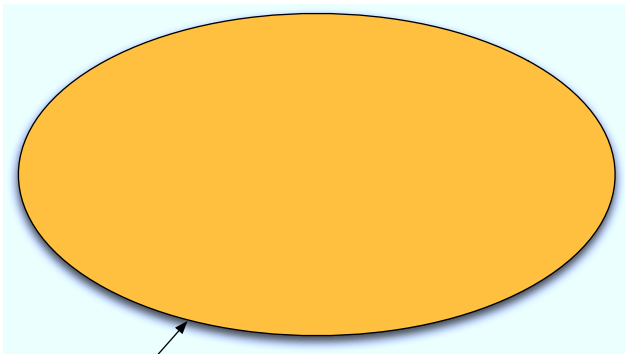
All unique semantics-preserving
affine multidimensional schedules

Bounded: 10^{200}

Legal: 10^{50}

Empirical search: 10

Spaces of Affine Loop transformations



All unique bounded
affine multidimensional schedules

1 point \leftrightarrow 1 unique transformed program

Polyhedral Representation of Programs

Static Control Parts

- ▶ Loops have affine control only (over-approximation otherwise)

Polyhedral Representation of Programs

Static Control Parts

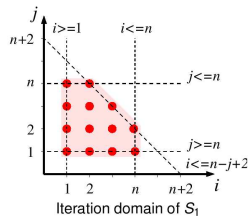
- ▶ Loops have affine control only (over-approximation otherwise)
- ▶ Iteration domain: represented as integer polyhedra

```

for (i=1; i<=n; ++i)
. for (j=1; j<=n; ++j)
. . if (i<=n-j+2)
. . . s[i] = ...

```

$$\mathcal{D}_{S_1} = \begin{bmatrix} 1 & 0 & 0 & -1 \\ -1 & 0 & 1 & 0 \\ 0 & 1 & 0 & -1 \\ -1 & 0 & 1 & 0 \\ -1 & -1 & 1 & 2 \end{bmatrix} \cdot \begin{pmatrix} i \\ j \\ n \\ 1 \end{pmatrix} \geq \vec{0}$$



Polyhedral Representation of Programs

Static Control Parts

- ▶ Loops have affine control only (over-approximation otherwise)
- ▶ Iteration domain: represented as integer polyhedra
- ▶ Memory accesses: static references, represented as affine functions of \vec{x}_S and \vec{p}

```

for (i=0; i<n; ++i) {
. s[i] = 0;
. for (j=0; j<n; ++j)
. . s[i] = s[i]+a[i][j]*x[j];
}

```

$$f_s(\vec{x}_{S2}) = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \cdot \begin{pmatrix} \vec{x}_{S2} \\ n \\ 1 \end{pmatrix}$$

$$f_a(\vec{x}_{S2}) = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \cdot \begin{pmatrix} \vec{x}_{S2} \\ n \\ 1 \end{pmatrix}$$

$$f_x(\vec{x}_{S2}) = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \cdot \begin{pmatrix} \vec{x}_{S2} \\ n \\ 1 \end{pmatrix}$$

Polyhedral Representation of Programs

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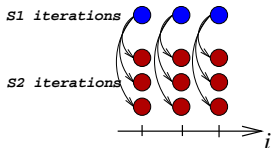
- ▶ Loops have affine control only (over-approximation otherwise)
- ▶ Iteration domain: represented as integer polyhedra
- ▶ Memory accesses: static references, represented as affine functions of \vec{x}_S and \vec{p}
- ▶ Data dependence between S1 and S2: a subset of the Cartesian product of \mathcal{D}_{S1} and \mathcal{D}_{S2} (**exact analysis**)

```

for (i=1; i<=3; ++i) {
. s[i] = 0;
. for (j=1; j<=3; ++j)
. . s[i] = s[i] + 1;
}

```

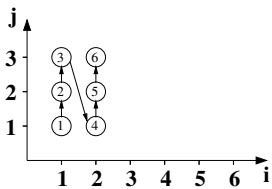
$$\mathcal{D}_{S1 \& S2} : \begin{bmatrix} 1 & -1 & 0 & 0 \\ 1 & 0 & 0 & -1 \\ -1 & 0 & 0 & 3 \\ 0 & 1 & 0 & -1 \\ 0 & -1 & 0 & 3 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & -1 & 3 \end{bmatrix} \cdot \begin{pmatrix} i_{S1} \\ i_{S2} \\ j_{S2} \\ 1 \end{pmatrix} \begin{matrix} \equiv 0 \\ \geq 0 \end{matrix}$$



Affine Transformations for Iteration Reordering

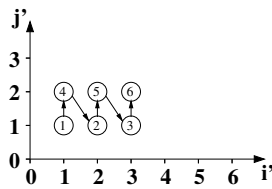
Interchange Transformation

The transformation matrix is the identity with a permutation of two rows.



(a) original polyhedron

\Rightarrow



(c) target polyhedron

$$\begin{bmatrix} 1 & 0 \\ -1 & 0 \\ 0 & 1 \\ 0 & -1 \end{bmatrix} \begin{pmatrix} i \\ j \end{pmatrix} + \begin{pmatrix} -1 \\ 2 \\ -1 \\ 3 \end{pmatrix} \geq \vec{0}$$

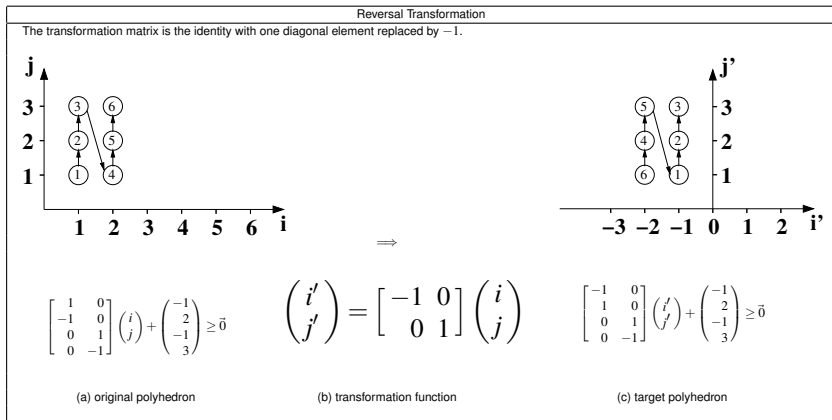
$$\begin{pmatrix} i' \\ j' \end{pmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{pmatrix} i \\ j \end{pmatrix}$$

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```
do i = 1, 2
  do j = 1, 3
    S(i, j)
```

```
do i' = 1, 3
  do j' = 1, 2
    S(i=j', j=i')
```

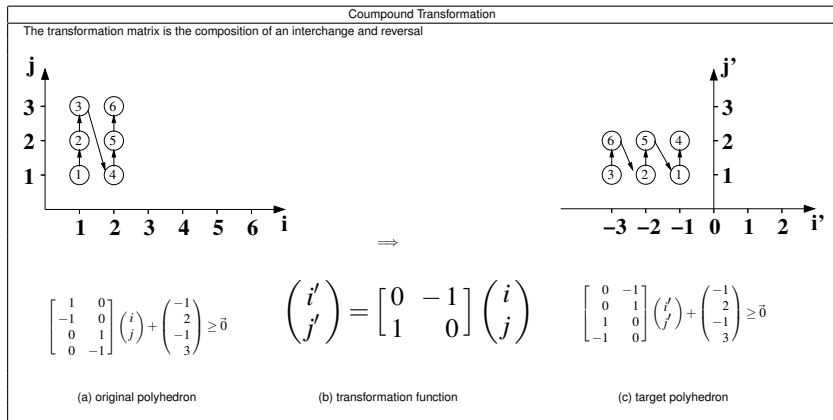
Affine Transformations for Iteration Reordering



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do i = 1, 2
  do j = 1, 3
    S(i, j)
```

```
do i' = -1, -2, -1
  do j' = 1, 3
    S(i=3-i', j=j')
```

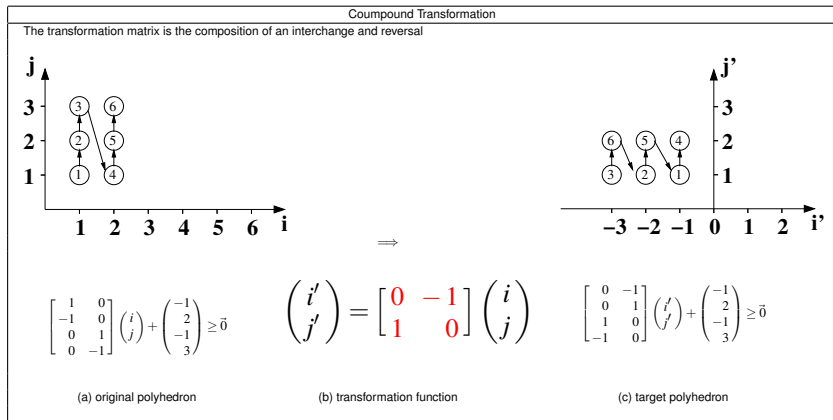
Affine Transformations for Iteration Reordering



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do i = 1, 2
  do j = 1, 3
    S(i, j)
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```
do j' = -1, -3, -1
  do i' = 1, 2
    S(i=4-j', j=j')
```

Affine Transformations for Iteration Reordering



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```

Affine Schedule

Definition (Affine multidimensional schedule)

Given a statement S , an affine schedule Θ^S of dimension m is an affine form on the d outer loop iterators \vec{x}_S and the p global parameters \vec{n} .

$\Theta^S \in \mathbb{Z}^{m \times (d+p+1)}$ can be written as:

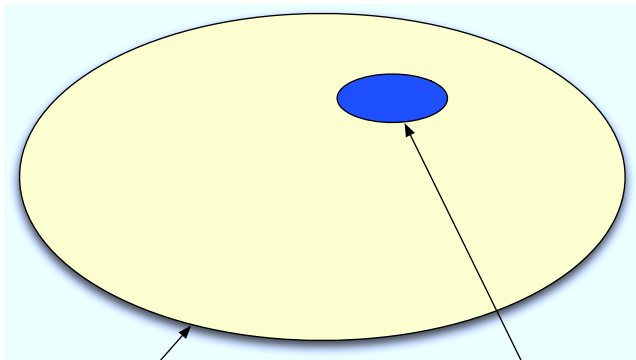
$$\Theta^S(\vec{x}_S) = \begin{pmatrix} \theta_{1,1} & \cdots & \theta_{1,d+p+1} \\ \vdots & & \vdots \\ \theta_{m,1} & \cdots & \theta_{m,d+p+1} \end{pmatrix} \cdot \begin{pmatrix} \vec{x}_S \\ \vec{n} \\ 1 \end{pmatrix}$$

Θ_k^S denotes the k^{th} row of Θ^S .

Definition (Bounded affine multidimensional schedule)

Θ^S is a bounded schedule if $\theta_{i,j}^S \in [x, y]$ with $x, y \in \mathbb{Z}$

Space of Semantics-Preserving Affine Schedules



All unique bounded
affine multidimensional schedules

All unique semantics-preserving
affine multidimensional schedules

1 point \leftrightarrow 1 unique semantically equivalent program
(up to affine iteration reordering)

Semantics Preservation

Definition (Causality condition)

Given Θ^R a schedule for the instances of R , Θ^S a schedule for the instances of S . Θ^R and Θ^S preserve the dependence $\mathcal{D}_{R,S}$ if $\forall \langle \vec{x}_R, \vec{x}_S \rangle \in \mathcal{D}_{R,S}$:

$$\Theta^R(\vec{x}_R) \prec \Theta^S(\vec{x}_S)$$

\prec denotes the *lexicographic ordering*.

$(a_1, \dots, a_n) \prec (b_1, \dots, b_m)$ iff $\exists i, 1 \leq i \leq \min(n, m)$ s.t. $(a_1, \dots, a_{i-1}) = (b_1, \dots, b_{i-1})$
and $a_i < b_i$

Lexico-positivity of Dependence Satisfaction

- ▶ $\Theta^R(\vec{x}_R) \prec \Theta^S(\vec{x}_S)$ is equivalently written $\Theta^S(\vec{x}_S) - \Theta^R(\vec{x}_R) \succ \vec{0}$

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- ▶ Considering the row p of the scheduling matrices:

$$\Theta_p^S(\vec{x}_S) - \Theta_p^R(\vec{x}_R) \geq \delta_p$$

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- ▶ $\delta_p \geq 1$ implies no constraints on $\delta_k, k > p$
- ▶ $\delta_p \geq 0$ is required if $\nexists k < p, \delta_k \geq 1$

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- ▶ Schedule lower bound:

Lemma (Schedule lower bound)

Given Θ_k^R, Θ_k^S such that each coefficient value is bounded in $[x, y]$. Then there exists $K \in \mathbb{Z}$ such that:

$$\Theta_k^S(\vec{x}_S) - \Theta_k^R(\vec{x}_R) > -K \cdot \vec{n} - K$$

Convex Form of All Bounded Affine Schedules

Lemma (Convex form of semantics-preserving affine schedules)

Given a set of affine schedules $\Theta^R, \Theta^S \dots$ of dimension m , the program semantics is preserved if the three following conditions hold:

$$(i) \quad \forall \mathcal{D}_{R,S}, \delta_p^{\mathcal{D}_{R,S}} \in \{0, 1\}$$

$$(ii) \quad \forall \mathcal{D}_{R,S}, \sum_{p=1}^m \delta_p^{\mathcal{D}_{R,S}} = 1$$

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$$\Theta_p^S(\vec{x}_S) - \Theta_p^R(\vec{x}_R) \geq \delta_p^{\mathcal{D}_{R,S}} - \sum_{k=1}^{p-1} \delta_k^{\mathcal{D}_{R,S}} \cdot (K \cdot \vec{n} + K)$$

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→ Use **Farkas lemma** to build all non-negative functions over a **polyhedron** (here, the dependence polyhedra) [Feautrier,92]

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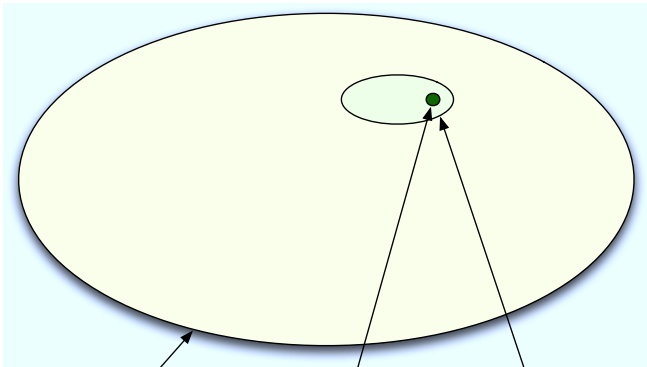
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- Use **Farkas lemma** to build all non-negative functions over a **polyhedron** (here, the dependence polyhedra) [Feautrier,92]
- Bounded coefficients required [Vasilache,07]

Space of Semantics-Preserving Fusion Choices



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**All unique semantics-preserving
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1 point \leftrightarrow 1 unique semantically equivalent program
(up to "partial" statement reordering)

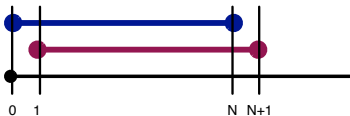
Fusion in the Polyhedral Model



```
for (i = 0; i <= N; ++i) {  
    Blue(i);  
    Red(i);  
}
```

Perfectly aligned fusion

Fusion in the Polyhedral Model



```

Blue(0);
for (i = 1; i <= N; ++i) {
    Blue(i);
    Red(i-1);
}
Red(N);

```

Fusion with shift of 1
Not all instances are fused

Fusion in the Polyhedral Model



```

for (i = 0; i < P; ++i)
  Blue(i);
for (i = P; i <= N; ++i) {
  Blue(i);
  Red(i-P);
}
for (i = N+1; i <= N+P; ++i)
  Red(i-P);

```

Fusion with parametric shift of P
Automatic generation of prolog/epilog code

Fusion in the Polyhedral Model



```

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  Blue(i);
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}
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  Red(i-P);

```

Many other transformations may be required to enable fusion: interchange, skewing, etc.

Affine Constraints for Fusibility

- ▶ **Two statements can be fused if their timestamp can overlap**

Definition (Generalized fusibility check)

Given v_R (resp. v_S) the set of vertices of \mathcal{D}_R (resp. \mathcal{D}_S). R and S are fusible at level p if, $\forall k \in \{1 \dots p\}$, there exist two semantics-preserving schedules Θ_k^R and Θ_k^S such that

$$\exists(\vec{x}_1, \vec{x}_2, \vec{x}_3) \in v_R \times v_S \times v_R, \quad \Theta_k^R(\vec{x}_1) \leq \Theta_k^S(\vec{x}_2) \leq \Theta_k^R(\vec{x}_3)$$

- ▶ Intersect \mathcal{L} with fusibility and distribution constraints
- ▶ **Completeness:** if the test fails, then there is no sequence of affine transformations that can implement this fusion structure

Fusion / Distribution / Code Motion

Our strategy:

- 1 Build a set containing all unique fusion / distribution / code motion combinations
- 2 Prune all combinations that do not preserve the semantics

Given two statements R and S, three choices:

- 1 R is *fully before* S \rightarrow distribution + code motion
 - 2 R is *fully after* S \rightarrow distribution + code motion
 - 3 otherwise \rightarrow fusion
- \Rightarrow It corresponds to all total preorders of R and S

Affine Encoding of Total Preorders

Principle:

- ▶ Model a total preorder with 3 binary variables

$$p_{i,j} : i < j \quad s_{i,j} : i > j \quad e_{i,j} : i = j$$

- ▶ Enforce totality and mutual exclusion
- ▶ Enforce all cases of transitivity through affine inequalities connecting some variables. Ex: $e_{i,j} = 1 \wedge e_{j,k} = 1 \Rightarrow e_{i,k} = 1$

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- ▶ **This set contains one and only one point per distinct total preorder of n elements**

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- ▶ Easy pruning: just bound the sum of some variables

$$\text{e.g., } e_{1,2} + e_{4,5} + e_{8,12} < 3$$

- ▶ Automatic removal of supersets of unfusable sets

Convex set of All Unique Total Preorders

$$O = \left\{ \begin{array}{l} 0 \leq p_{i,j} \leq 1 \\ 0 \leq e_{i,j} \leq 1 \\ 0 \leq s_{i,j} \leq 1 \end{array} \right\} \quad \text{constrained to:} \quad O = \left\{ \begin{array}{l} \forall k \in]j, n] \\ \forall k \in]i, j[\\ \forall k \in]j, n] \\ \forall k \in]i, j[\\ \forall k \in]j, n] \end{array} \right. \left. \begin{array}{l} \left. \begin{array}{l} 0 \leq p_{i,j} \leq 1 \\ 0 \leq e_{i,j} \leq 1 \end{array} \right\} \text{Variables are binary} \\ \left. \begin{array}{l} p_{i,j} + e_{i,j} \leq 1 \\ e_{i,j} + e_{i,k} \leq 1 + e_{j,k} \\ e_{i,j} + e_{j,k} \leq 1 + e_{i,k} \end{array} \right\} \text{Relaxed mutual exclusion} \\ \left. \begin{array}{l} p_{i,k} + p_{k,j} \leq 1 + p_{i,j} \\ e_{i,j} + p_{i,k} \leq 1 + p_{j,k} \\ e_{i,j} + p_{j,k} \leq 1 + p_{i,k} \end{array} \right\} \text{Basic transitivity on } p \\ \left. \begin{array}{l} e_{k,j} + p_{i,k} \leq 1 + p_{i,j} \\ e_{i,j} + p_{i,j} + p_{j,k} \leq 1 + p_{i,k} + e_{i,k} \end{array} \right\} \text{Complex transitivity on } p \text{ and } e \\ \left. \begin{array}{l} e_{i,j} + p_{i,j} + p_{j,k} \leq 1 + p_{i,k} + e_{i,k} \end{array} \right\} \text{Complex transitivity on } s \text{ and } p \end{array} \right.$$

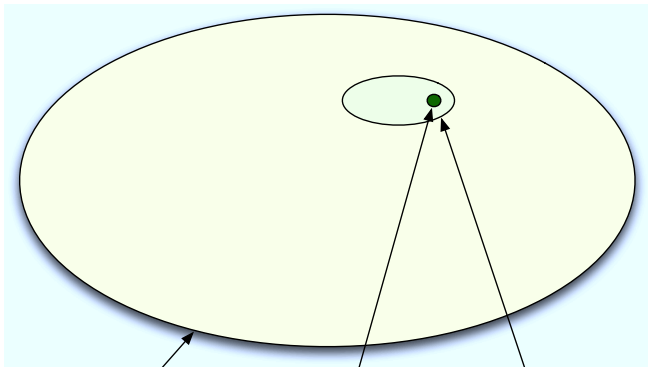
- ▶ Systematic construction for a given n , needs n^2 Boolean variables
- ▶ **Enable ILP modeling, enumeration, etc.**
- ▶ Extension to multidimensional total preorders (i.e., multi-level fusion)

Pruning for Semantics Preservation

Intuition: enumerate the smallest sets of unfusible statements

- ▶ Use an intermediate structure to represent sets of statements
 - ▶ Graph representation of maybe-unfusible sets (1 node per statement)
 - ▶ Enumerate sets from the smallest to the largest
- ▶ Leverage dependence graph + properties of fusion / distribution
- ▶ Compute properties by intersecting \mathcal{L} with additional fusion / distribution / code motion affine constraints
- ▶ Any individual point can be removed from O

Space of Semantics-Preserving Fusion Choices



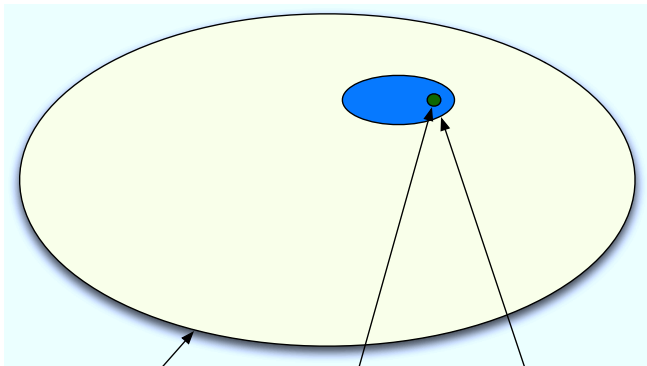
All unique bounded
affine multidimensional schedules

All unique semantics-preserving
affine multidimensional schedules

All unique semantics-preserving
fusion / distribution / code motion choices

1 point \leftrightarrow 1 unique semantically equivalent program
(up to statement reordering)

Space of Semantics-Preserving Fusion Choices



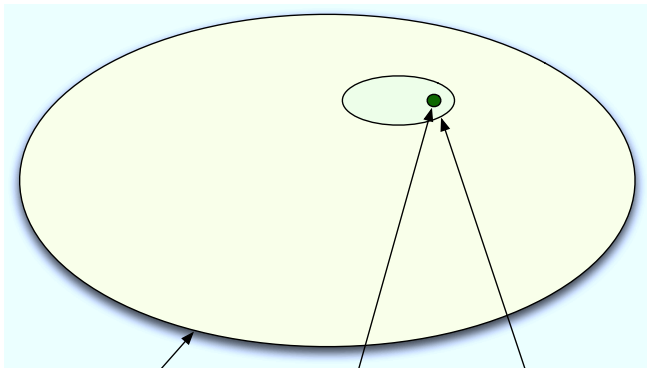
All unique bounded
affine multidimensional schedules

All unique semantics-preserving
affine multidimensional schedules

All unique semantics-preserving
fusion / distribution / code motion choices

1 point \leftrightarrow **many** unique semantically equivalent programs
(up to iteration reordering)

Space of Semantics-Preserving Fusion Choices



All unique bounded
affine multidimensional schedules

All unique semantics-preserving
affine multidimensional schedules

All unique semantics-preserving
fusion / distribution / code motion choices

1 point \leftrightarrow 1 unique semantically equivalent program
(up to limited iteration reordering)

Objectives for Effective Optimization

Objectives:

- ▶ Achieve efficient coarse-grain parallelization
- ▶ Combine iterative search of profitable transformations for tiling
 - loop fusion and loop distribution

Tiling Hyperplane method [Bondhugula,08]

- ▶ Model-driven approach for automatic parallelization + locality improvement
- ▶ Tiling-oriented
- ▶ Poor model-driven heuristic for the selection of loop fusion (not portable)
- ▶ Overly relaxed definition of fused statements

Fusibility Restricted to Non-negative Schedules

- ▶ Fusibility is not a transitive relation!
 - ▶ Example: sequence of matrix-by-vector products $x = Ab$, $y = Bx$, $z = Cy$
 - ▶ $x = Ab$, $y = Bx$ can be fused, also $y = Bx$, $z = Cy$
 - ▶ They cannot be fused all together
- ▶ **Determining the Fusibility of a group of statements is reducible to exhibiting compatible pairwise loop permutations**
 - ▶ Extremely easy to compute all possible loop permutations that lead to fuse a pair of statements
 - ▶ Never check \mathcal{L} on more than two statements!
- ▶ **Stronger definition of fusion**
 - ▶ Guarantee **at most c instances are not fused**

$$-c < \Theta_k^R(\vec{0}) - \Theta_k^S(\vec{0}) < c$$

- ▶ No combinatorial choice

The Optimization Algorithm in a Nutshell

Proceeds from the outer-most loop level to the inner-most:

- 1 Compute the space of valid fusion/distribution/code motion choices
- 2 **Select a fusion/distribution/code motion scheme** in this space
- 3 Compute an affine schedule **that implements this scheme**
 - ▶ Static cost model to select the schedule
 - ▶ Compound of skewing, shifting, fusion, distribution, interchange, tiling and parallelization (OpenMP)
 - ▶ **Maximize locality** for each set of statements to be fused

Experimental Results

Benchmark	#loops	#stmts	#refs	O			\mathcal{F}^1			Time	perf-Intel	perf-AMD
				#dim	#cst	#points	#dim	#cst	#points			
advect3d	12	4	32	12	58	75	9	43	26	0.82s	1.47×	5.19×
atax	4	4	10	12	58	75	6	25	16	0.06s	3.66×	1.88×
bicg	3	4	10	12	58	75	10	52	26	0.05s	1.75×	1.40×
gemver	7	4	19	12	58	75	6	28	8	0.06s	1.34×	1.33×
ludcmp	9	14	35	182	3003	$\approx 10^{12}$	40	443	8	0.54s	1.98×	1.45×
doitgen	5	3	7	6	22	13	3	10	4	0.08s	15.35×	14.27×
varcovar	7	7	26	42	350	47293	22	193	96	0.09s	7.24×	14.83×
correl	5	6	12	30	215	4683	21	162	176	0.09s	3.00×	3.44×

Table: Search space statistics and performance improvement

- ▶ **Performance portability:** empirical search on the target machine of the optimal fusion structure
- ▶ Outperforms state-of-the-art cost models
- ▶ Full implementation in the source-to-source polyhedral compiler PoCC

Conclusion

Take-home message:

- ⇒ **Clear formalization of loop fusion** in the polyhedral model
- ⇒ **Formal definition of all semantically equivalent programs** up to:
 - ▶ statement reordering
 - ▶ limited affine iteration reordering
 - ▶ arbitrary affine iteration reordering

- ⇒ **Effective and portable hybrid empirical optimization algorithm**
(parallelization + data locality)

Future work:

- ▶ Develop static cost models for fusion / distribution / code motion
- ▶ Use statistical techniques to learn optimization algorithms