

# A Crash Course in Compilers for Parallel Computing

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# Overview of "Crash Course"

- L1: Data Dependence Analysis and Parallelization (Oct. 30)
- L2 & L3: Loop Reordering Transformations, Reuse Analysis and Locality Optimization (Nov. 6)
- L4: Autotuning Compiler Technology (Nov. 13)

# Outline of Lecture

- I. Summary of Last Week
- II. Reuse Analysis
- III. Two Loop Reordering Transformations
  - Permutation
  - Tiling (aka blocking)
- IV. Locality Optimization

# I. Summary: Data Dependence

True (flow) dependence

a =  
= a

Anti-dependence

a = a  
=

Output dependence

a =  
a =

*Input dependence (for locality)*

= a  
= a

**Definition:** Data dependence exists from a reference instance  $i$  to  $i'$  iff

either  $i$  or  $i'$  is a write operation  
 $i$  and  $i'$  refer to the same variable  
 $i$  executes before  $i'$

# Restrict to an Affine Domain

```
for (i=1; i<N; i++)
  for (j=1; j<N j++) {
    A[i+2*j+3, 4*i+2*j, 3*i] = ...;
    ... = A[1, 2*i+1, j];
  }
```

- Only use loop bounds and array indices which are integer linear functions of loop variables.
- **Non-affine example:**

```
for (i=1; i<N; i++)
  for (j=1; j<N j++) {
    A[i*j] = A[i*(j-1)];
    A[B[i]] = A[B[j]];
  }
```

# Distance Vectors

```
N = 6;  
for (i=1; i<N; i++)  
    for (j=1; j<N; j++)  
        A[i+1,j+1] = A[i,j] * 2.0;
```

- Distance vector =  $[1, 1]$
- A loop has a distance vector  $D$  if there exists data dependence from iteration vector  $I$  to a later vector  $I'$ , and  $D = I' - I$ .
- Since  $I' > I$ ,  $D \geq 0$ .  
( $D$  is lexicographically greater than or equal to 0).

# Equivalence to Integer Programming

- Need to determine if  $F(i) = G(i')$ , where  $i$  and  $i'$  are iteration vectors, with constraints  $i, i' \geq L, U \geq i, i'$

- **Example:**

```
for (i=2; i<=100; i++)  
  A[i] = A[i-1];
```

- **Inequalities:**

$0 \leq i_1 \leq 100,$        $i_2 = i_1 - 1,$        $i_2 \leq 100$   
*integer vector I,*       $AI \leq b$

$$\begin{bmatrix} -1 & 0 \\ 1 & 0 \\ -1 & 1 \\ 1 & -1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} i_1 \\ i_2 \end{bmatrix} \leq \begin{bmatrix} 0 \\ 100 \\ -1 \\ 1 \\ 100 \end{bmatrix}$$

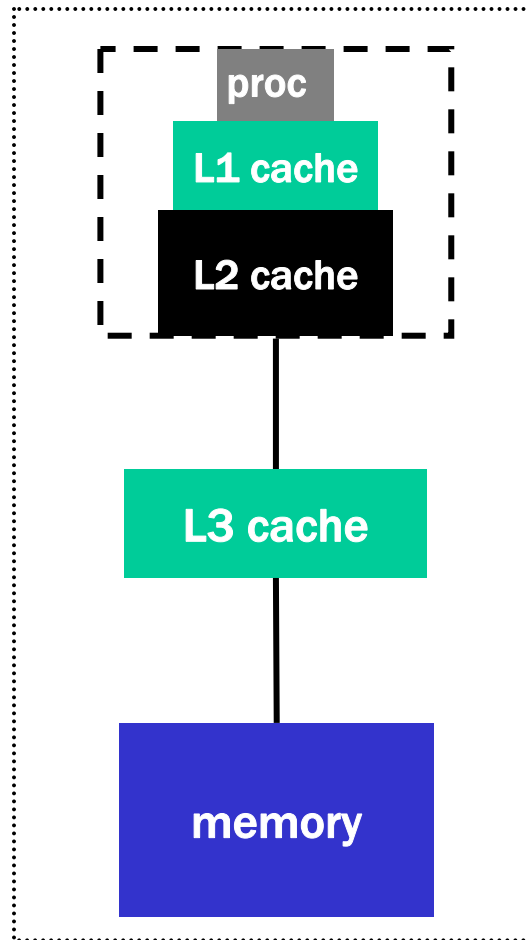
Solution exist?  
Yes  $\rightarrow$  dependence

# Fundamental Theorem of Dependence

- **Theorem 2.2:**
  - Any reordering transformation that preserves every dependence in a program preserves the meaning of that program.



## II. Introduction to Locality Optimization and Reuse Analysis



- Large memories are slow, fast memories are small
- Hierarchy allows fast and large memory on average
- Managing *locality* crucial for achieving high performance

# Cache basics: a quiz

- **Cache hit:**
    - in-cache memory access—cheap
  - **Cache miss:**
    - non-cached memory access—expensive
    - need to access next, slower level of hierarchy
- 
- **Cache line size:**
    - # of bytes loaded together in one entry
    - typically a few machine words per entry
  - **Capacity:**
    - amount of data that can be simultaneously in cache
  - **Associativity**
    - direct-mapped: only 1 address (line) in a given range in cache
    - $n$ -way:  $n \geq 2$  lines w/ different addresses can be stored

Parameters to optimization

# How do we get locality (in caches)?

- Data locality:
  - data is reused and is present in cache
  - same data or same cache line
- Data *reuse*:
  - data used multiple times
  - intrinsic in computation
- If a computation has reuse, what can we do to get locality?
  - code reordering transformations (today)
  - data layout

# Temporal Reuse

- Same data used in distinct iterations  $I$  and  $I'$

```
for (i=1; i<N; i++)  
  for (j=1; j<N; j++)  
    A[j]= A[j]+A[j+1]+A[j-1]
```

- $A[j]$  has self-temporal reuse in loop  $i$

# Spatial Reuse

- Same cache line used in distinct iterations  $I$  and  $I'$

```
for (i=1; i<N; i++)  
  for (j=1; j<N; j++)  
    A[j]= A[j]+A[j+1]+A[j-1]
```

- $A[j]$  has self-spatial reuse in loop  $j$
- **Multi-dimensional array note:** C arrays are stored in row-major order, while FORTRAN arrays are stored in column-major order)

# Group Reuse

- Same data used by distinct references

```
for (i=1; i<N; i++)  
  for (j=1; j<N; j++)  
    A[j] = A[j] + A[j+1] + A[j-1]
```

- $A[j]$ ,  $A[j+1]$  and  $A[j-1]$  have group reuse (spatial and temporal) in loop  $j$

## III. Reordering Transformations

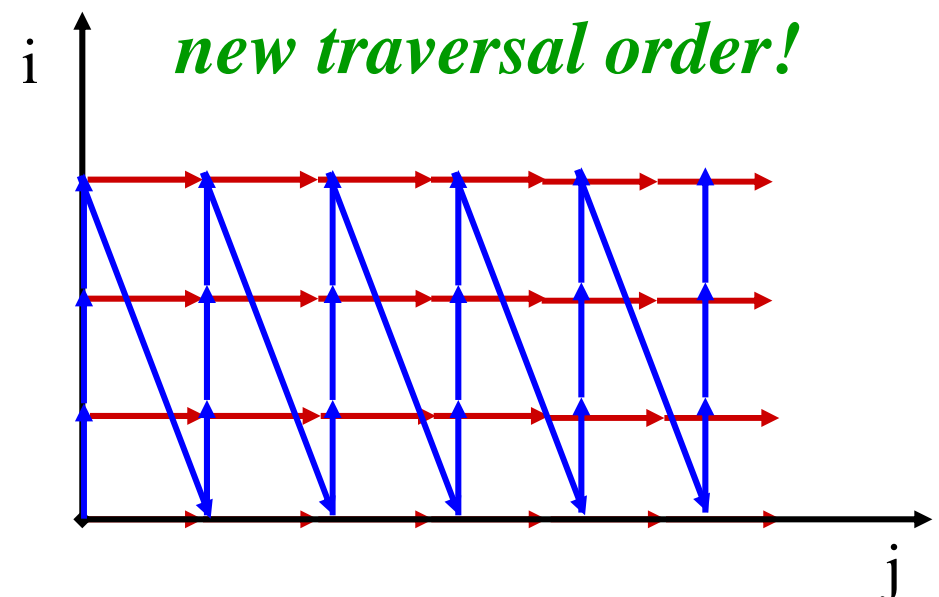
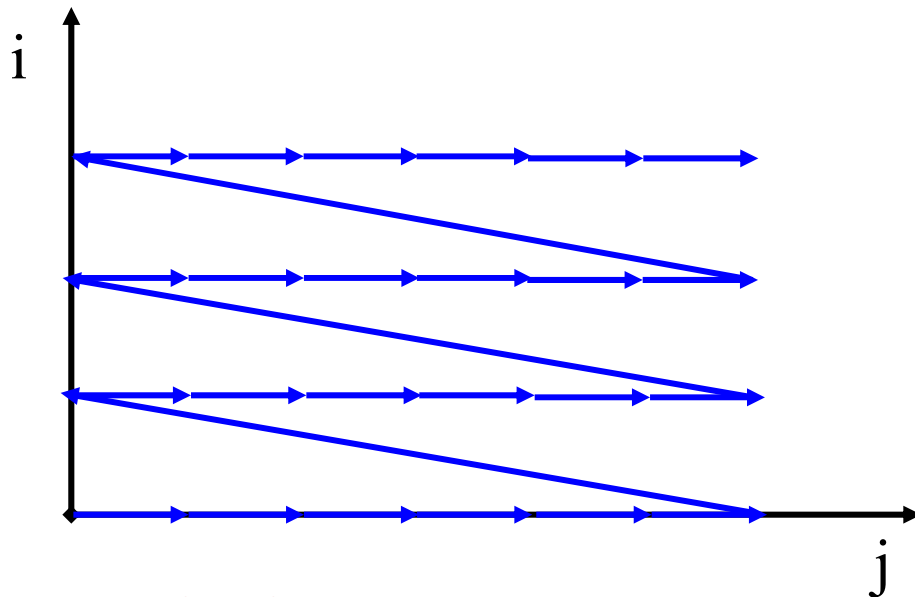
- Analyze reuse in computation
- Apply loop reordering transformations to improve locality based on reuse
- With any loop reordering transformation, always ask
  - **Safety?** (doesn't reverse dependences)
  - **Profitability?** (improves locality)

# Loop Permutation: A Reordering Transformation

Permute the order of the loops to modify the traversal order

```
for (i= 0; i<3; i++)  
  for (j=0; j<6; j++)  
    A[i,j+1]=A[i,j]+B[j]
```

```
for (j=0; j<6; j++)  
  for (i= 0; i<3; i++)  
    A[i,j+1]=A[i,j]+B[j]
```



**Which one is better for row-major storage?**



# Safety of Permutation

- **Intuition:** Cannot permute two loops  $i$  and  $j$  in a loop nest if doing so reverses the direction of any dependence.
- Loops  $i$  through  $j$  of an  $n$ -deep loop nest are *fully permutable* if for all dependences  $D$ , either

$$(d_1, \dots, d_{i-1}) > 0$$

or

$$\text{for all } k, i \leq k \leq j, d_k \geq 0$$

- **Stated without proof:** Within the affine domain,  $n-1$  inner loops of  $n$ -deep loop nest can be transformed to be fully permutable.

# Simple Examples: 2-d Loop Nests

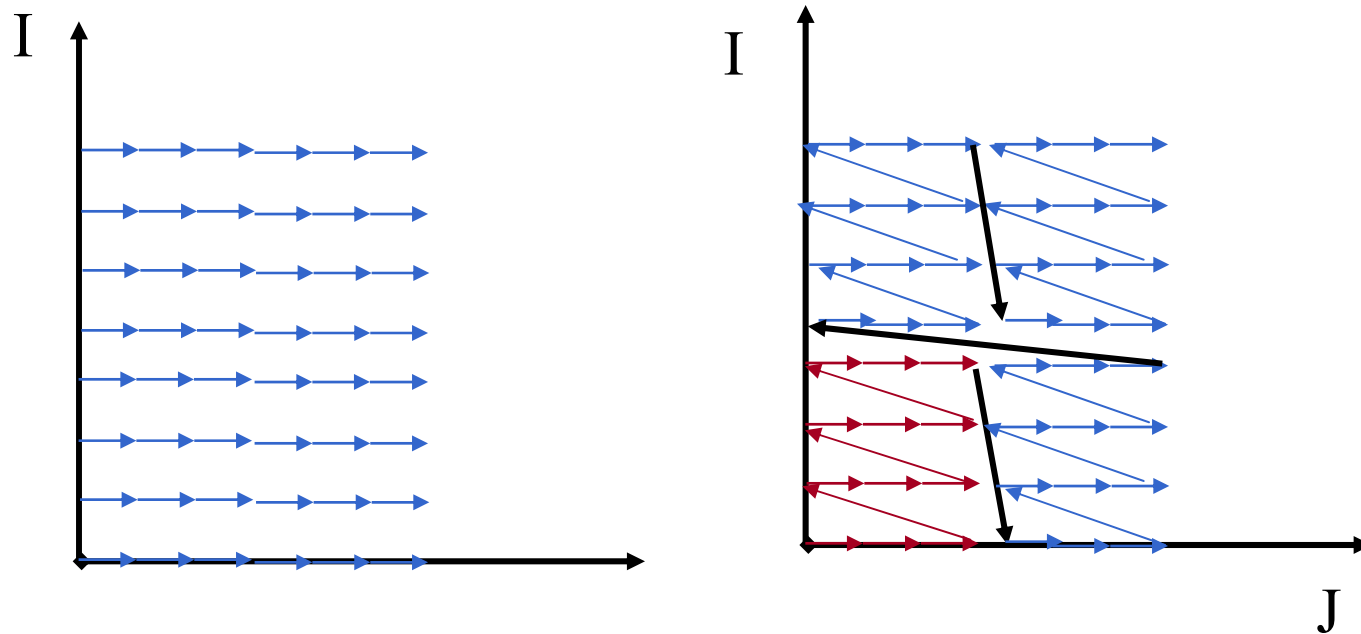
```
for (i= 0; i<3; i++)  
  for (j=0; j<6; j++)  
    A[i,j+1]=A[i,j]+B[j]
```

```
for (i= 0; i<3; i++)  
  for (j=0; j<6; j++)  
    A[i+1,j-1]=A[i,j]  
                +B[j]
```

- Distance vectors
- Ok to permute?

# Tiling (Blocking): Another Loop Reordering Transformation

- Blocking reorders loop iterations to bring iterations that reuse data closer in time



# Tiling Example

```
for (j=1; j<M; j++)  
  for (i=1; i<N; i++)  
    D[i] = D[i] +B[j,i]
```

Strip  
mine

```
for (j=1; j<M; j++)  
  for (i=1; i<N; i+=s)  
    for (ii=i, min(i+s-1,N)  
        D[ii] = D[ii] +B[j,ii]
```

Permute

```
for (i=1; i<N; i++)  
  for (j=1; j<M; j++)  
    for (ii=i, min(i+s-1,N)  
        D[ii] = D[ii] +B[j,ii]
```

## Legality of Tiling

- Tiling = strip-mine and permutation
  - Strip-mine does not reorder iterations
  - Permutation must be legal
- OR
  - strip size less than dependence distance

## IV. Locality Optimization

- Reuse analysis can be formulated in a manner similar to dependence analysis
  - Particularly true for temporal reuse
  - Spatial reuse requires special handling of most quickly varying dimension
- Simplification for today's lecture
  - Estimate cache misses for different scenarios
  - Select scenario that minimizes misses

# Reuse Analysis: Use to Estimate Cache Misses

```
for (i=0; i<N; i++)
  for (j=0; j<M; j++)
    A[i]=A[i]+B[j,i]
```

```
for (j=0; j<M; j++)
  for (i=0; i<N; i++)
    A[i]=A[i]+B[j,i]
```

<i>reference</i>	<i>loop J</i>	<i>loop I</i>
A[i]	1	N
B[j,i]	M	N*M

<i>reference</i>	<i>loop I</i>	<i>loop J</i>
A[i]	$N/\text{cls}^{(*)}$	$M*N/\text{cls}$
B[j,i]	$N/\text{cls}$	$M*N/\text{cls}$

(\*) cls = Cache Line Size (in elements)

# Allen & Kennedy: Innermost memory cost

- Innermost memory cost:  $C_M(L_i)$ 
  - assume  $L_i$  is innermost loop
    - $l_i$  = loop variable,  $N$  = number of iterations of  $L_i$
  - for each array reference  $r$  in loop nest:
    - $r$  does not depend on  $l_i$ :  $\text{cost}(r) = 1$
    - $r$  such that  $l_i$  strides over a non-contiguous dimension:  
 $\text{cost}(r) = N$
    - $r$  such that  $l_i$  strides over a contiguous dimension:  
 $\text{cost}(r) = N/\text{cls}$
  - $C_M(L_i) = \text{sum of cost}(r)$

Implicit in this cost function is that  $N$  is sufficiently large that cache capacity is exceeded by data footprint in innermost loop



# Canonical Example: matrix multiply

## Selecting Loop Order

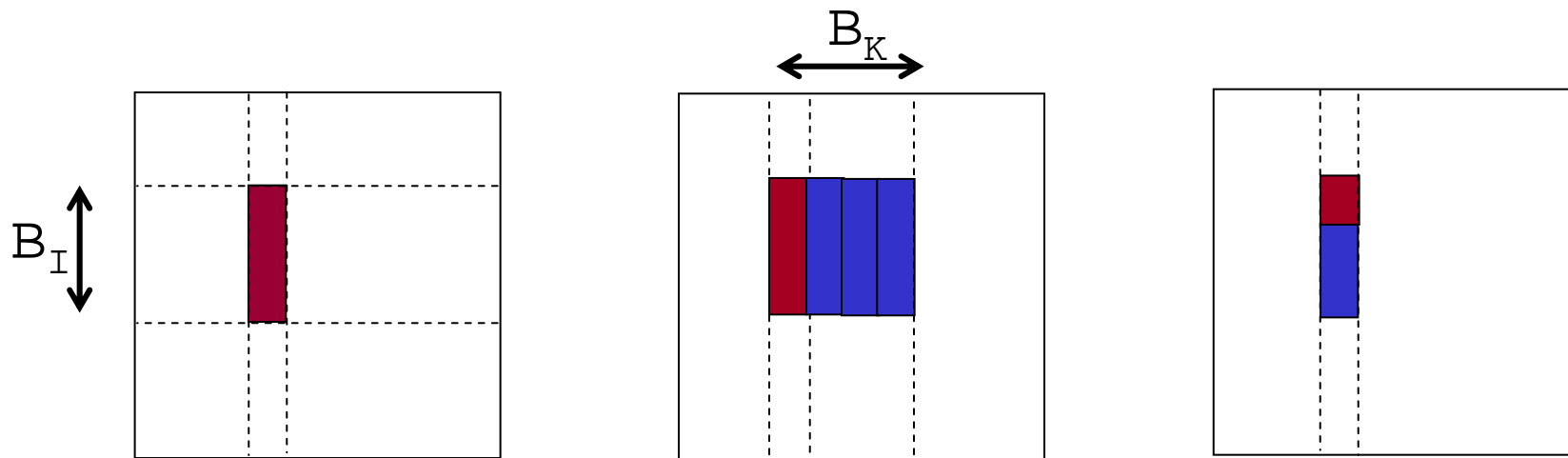
```
DO I = 1, N
  DO J = 1, N
    DO K = 1, N
      C(I,J) = C(I,J) + A(I,K) * B(K,J)
```

- $C_M(I) = 2N^3/\text{cls} + N^2$
- $C_M(J) = 2N^3 + N^2$
- $C_M(K) = N^3 + N^3/\text{cls} + N^2$
- Ordering by innermost loop cost: (J, K, I)

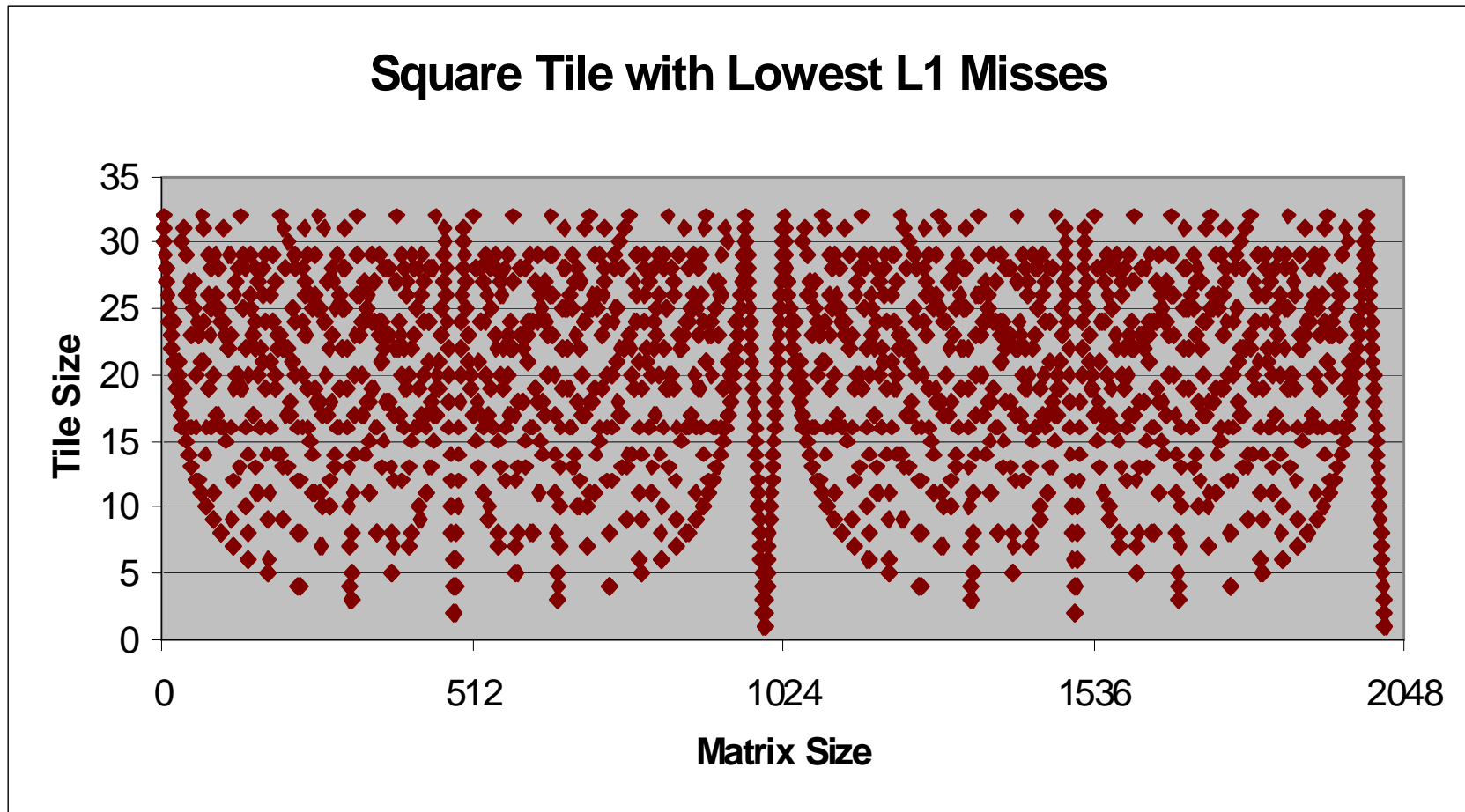
# Canonical Example: Matrix Multiply Selecting Tile Size

Choose  $T_i$  and  $T_k$  such that data footprint does not exceed cache capacity

```
DO K = 1, N by  $T_k$ 
  DO I = 1, N by  $T_i$ 
    DO J = 1, N
      DO KK = K, min(KK+  $T_k$ , N)
        DO II = I, min(II+  $T_i$ , N)
          C(II, J) = C(II, J) + A(II, KK) * B(KK, J)
```



# How to select optimal tile size ? (topic for next week)



Slide source: Jacqueline Chame

## Next Week

- How to use loop reordering transformations in an auto-tuning optimization system?