

# Data Dependences

- Essential constraints:

**S1:  $a = b + c$**   
**S2:  $d = a * 2$**   
**S3:  $a = c + 2$**   
**S4:  $e = d + c + 2$**

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# Data Dependences

- Essential constraints:

<b>S1:</b>	<b><math>a = b + c</math></b>
<b>S2:</b>	<b><math>d = a * 2</math></b>
<b>S3:</b>	<b><math>a = c + 2</math></b>
<b>S4:</b>	<b><math>e = d + c + 2</math></b>

- S1 and S2 cannot execute concurrently
- S2 and S3 cannot execute concurrently
- S1 and S3 cannot execute concurrently
- But S3 and S4 can execute concurrently
  
- Execution conditions due to Bernstein (1966)

# Types of Dependences

- Flow-dependence occurs when a variable which is assigned a value in one statement say S1 is read in another statement, say S2 later.

<b>S1:</b>	<b>a = b + c</b>
<b>S2:</b>	<b>d = a * 3</b>

# Types of Dependences

- Anti-dependence occurs when a variable which is read in one statement say S1 is assigned a value in another statement, say S2, later.

<b>S1:</b>	<b>d = a * 3</b>
<b>S2:</b>	<b>a = b + c</b>



# Types of Dependences

- Output-dependence occurs when a variable which is assigned a value in one statement say S1 is later reassigned in another statement, say S2.

<b>S1:</b>	<b>a = b + c</b>
<b>S2:</b>	<b>a = d * 3</b>

# Types of Dependences

- Input-dependence occurs when a variable is read in two different statements say S1 and S2. Relative ordering of S1 and S2 is not important for input dependence.

<b>S1:</b>	<b>a = b + c</b>
<b>S2:</b>	<b>d = b * 3</b>

# Data Dependences in Loops

- Associate a dynamic instance to each statement. For example

```
For i = 1 to 50
S1:   A(i) = B(i-1) + C(i)
S2:   B(i) = A(i+2) + C(i)
EndFor
```

- Statements S1 and S2 are executed 50 times. We say S2(10) to mean the execution of S2 when  $i = 10$ .
- Dependences are based on dynamic instances of statements.

# Data Dependences in Loops

- Unrolling loops can help one figure out dependences:

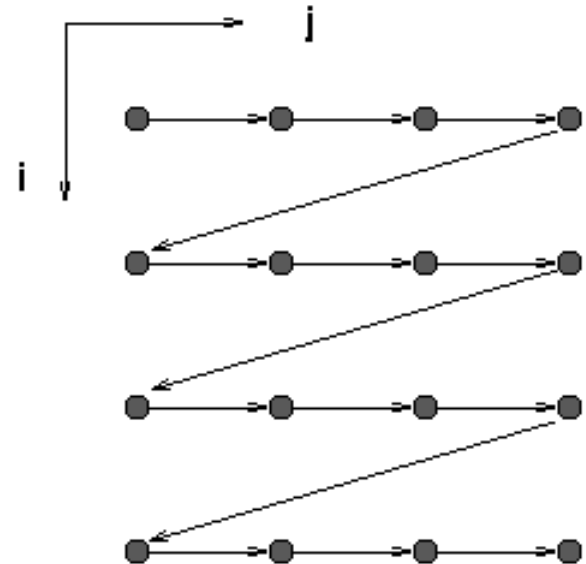
```
S1 (1) :      A (1) = B (0) + C (1)
S2 (1) :      B (1) = A (3) + C (1)
S1 (2) :      A (2) = B (1) + C (2)
S2 (2) :      B (2) = A (4) + C (2)
S1 (3) :      A (3) = B (2) + C (3)
S2 (3) :      B (3) = A (5) + C (3)
.....
S1 (50) :     A (50) = B (49) + C (50)
S2 (50) :     B (50) = A (52) + C (50)
```

# Iteration Spaces

- Nested loops define an iteration space:

```
For i = 1 to 4
  for j = 1 to 4
    A(i,j) = A(i,j) + C(j)
  Endfor
Endfor
```

- Sequential execution (traversal order):
- Dimensionality of iteration space = loop nest level; arbitrary convex shapes are allowed
- Change in order of execution is valid if no dependences are violated



# Single Processor Performance Enhancement

- Two fundamental issues:
  - Adequate fine-grained parallelism
    - Exploit vector instructions sets (SSE, AVX, AVX-512, ...)
    - Multiple pipelined functional units in each core
  - Minimize memory-access costs (about an order of magnitude higher than clock cycle)
- Useful loop transformations:
  - Loop Permutation
  - Loop Unrolling
  - Loop Blocking (tiling)
  - Loop Fusion/Distribution

# Access Stride and Spatial Locality

- Access stride: Separation between successively accessed memory locations
- Unit access stride maximizes spatial locality (only one miss per cache line)
- 2-D arrays have different linearized representations in Fortran and C

a	b	c	d
e	f	g	h
i	j	k	l
m	n	o	p

a	b	c	d	e	f		o	p
---	---	---	---	---	---	--	---	---

Row Major Order (C)

a	e	i	m	b	f		l	p
---	---	---	---	---	---	--	---	---

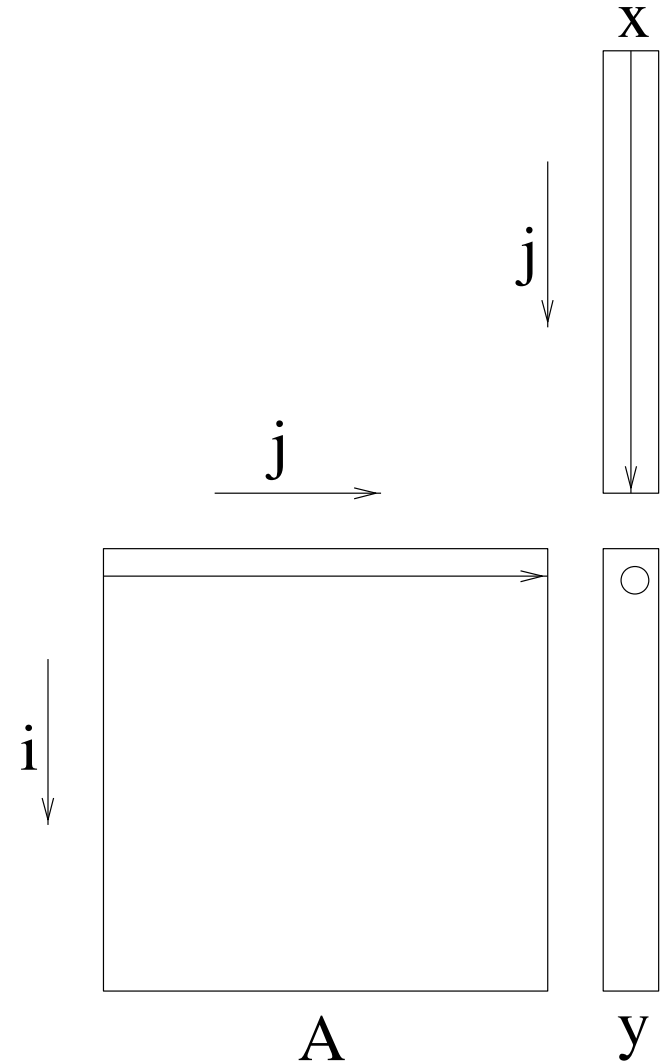
Column-major Order (Fortran)

# Matrix-Vector Multiplication: Dot-Product

```
For I = 1, N
  For J = 1, N
    y(I) = y(I) + A(I, J) * x(J)
  EndFor
EndFor
```

	A	x	y
C	1	1	0
Fortran	n	1	0

Access Stride for Arrays



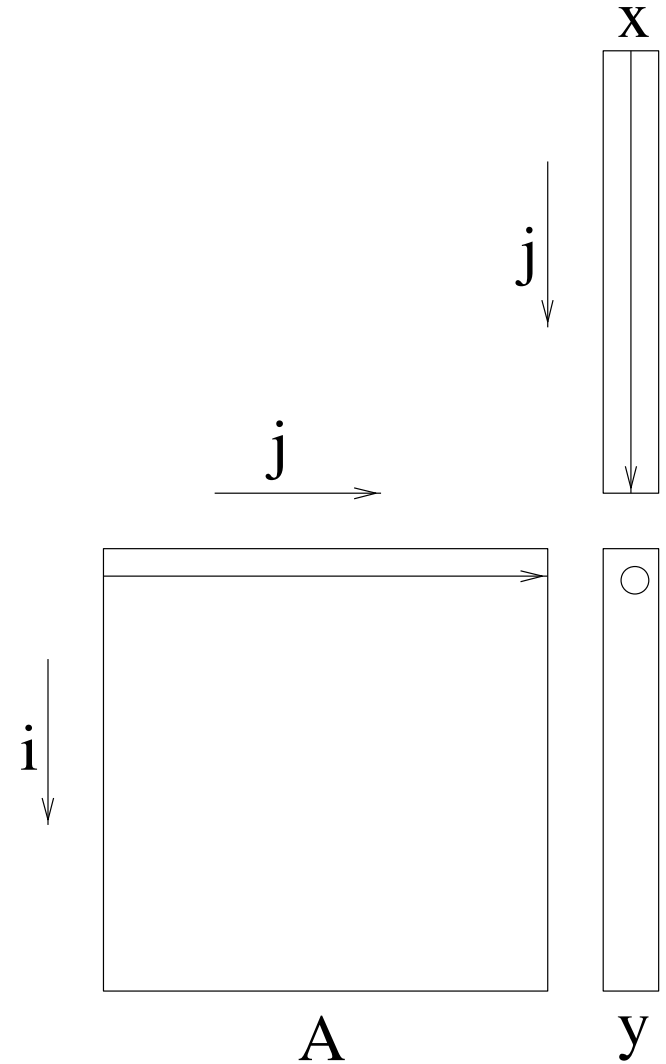


# Matrix-Vector Multiplication: SAXPY

```
For J = 1, N
  For I = 1, N
    y(I) = y(I) + A(I, J) * x(J)
  EndFor
EndFor
```

	A	x	y
C	n	0	1
Fortran	1	0	1

Access Stride for Arrays



# Loop Permutation: Matrix Multiplication

```

for(i=0;i<n;i++)
  for(j=0;j<n;j++)
    for(k=0;k<n;k++)
      c[i][j] = c[i][j] + a[i][k]*b[k][j];
  
```

Reference	ikj	kij	jik	ijk	jki	kji
C(i,j)	1	1	0	0	n	n
A(i,k)	0	0	1	1	n	n
B(k,j)	1	1	n	n	0	0
	Best	Best			Worst	Worst

Access Stride for Arrays (C: Row-Major)

# Loop Permutation: Matrix Multiplication

Reference	ikj	kij	jik	ijk	jki	kji
C(i,j)	1	1	0	0	n	n
A(i,k)	0	0	1	1	n	n
B(k,j)	1	1	n	n	0	0
	Best	Best			Worst	Worst

Access Stride for Arrays (C: Row-Major)

Compiler/Opt	ikj	kij	jik	ijk	jki	kji
icc -fast	17.0	17.0	17.0	17.0	17.0	17.0
icc -O3	5.0	5.0	5.0	5.0	5.0	5.0
icc -O2	7.8	7.8	7.8	7.8	7.8	7.8
icc -O1	2.0	2.0	.95	1.0	.29	.29
gcc -O3	6.1	7.6	.94	1.0	.29	.29
gcc -O2	2.0	2.0	.94	1.0	.29	.29
gcc -O1	1.9	1.9	.94	1.0	.29	.29

Performance on one core of Intel Xeon x5650 (GFLOPS)

# Permutation: Non-Rectangular Loops

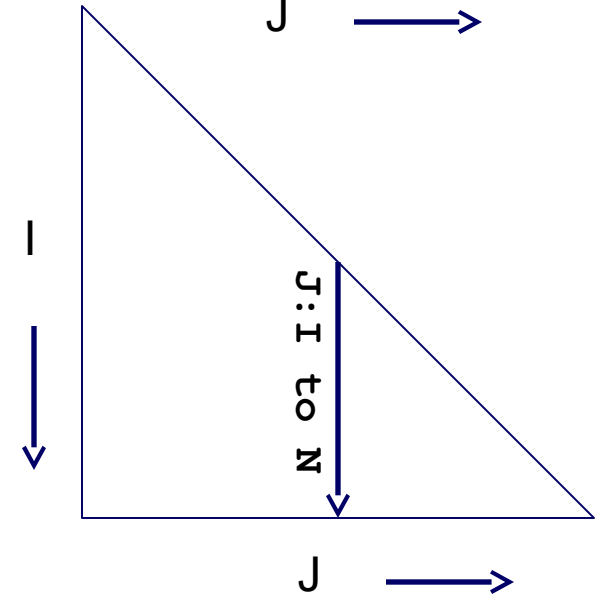
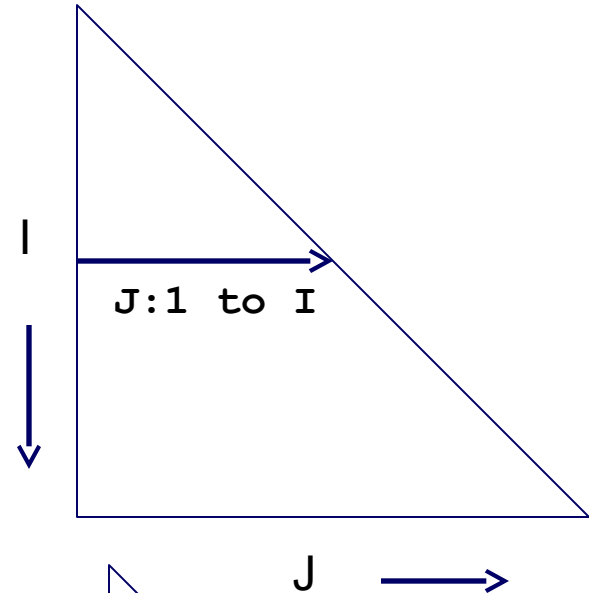
```
For I = 1, N
  For J = 1, I
    y(I) = y(I) + A(I, J) * x(J)
  EndFor
EndFor
```

# Permutation: Non-Rectangular Loops

```
For I = 1, N
  For J = 1, I
    y(I) = y(I) + A(I, J) * x(J)
  EndFor
EndFor
```

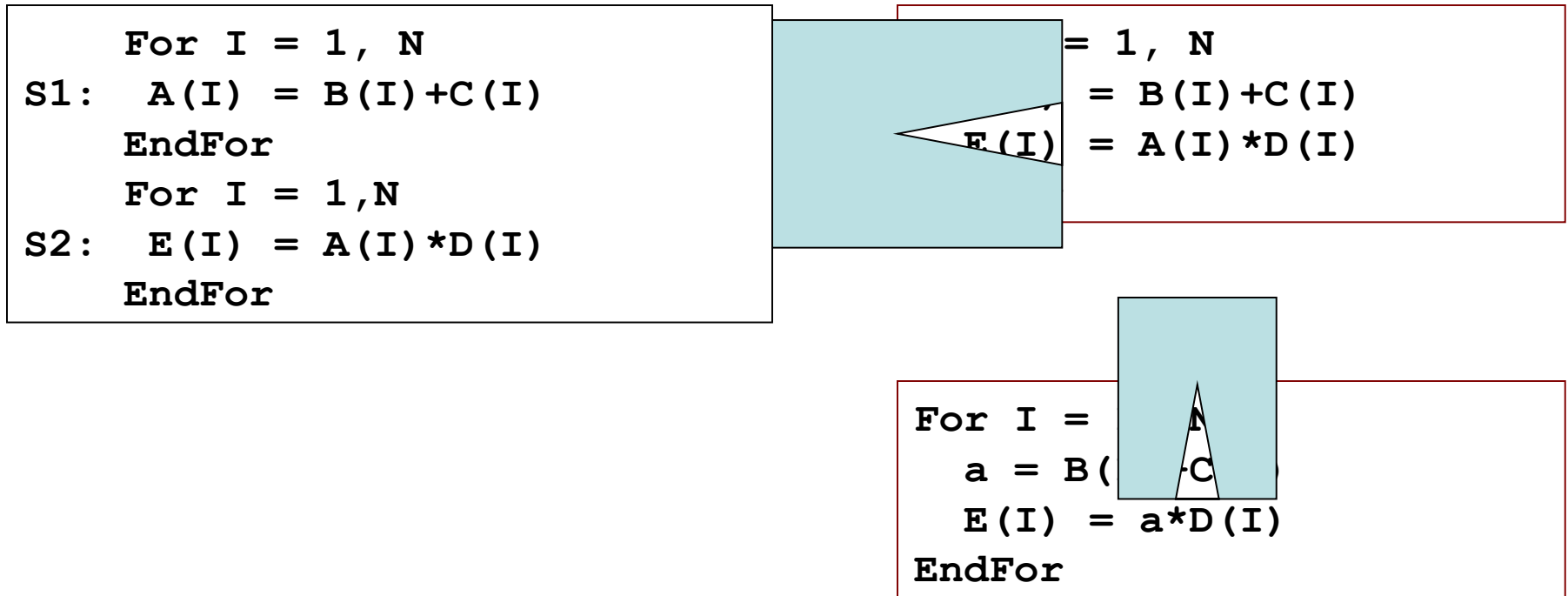


```
For J = 1, N
  For I = J, N
    y(I) = y(I) + A(I, J) * x(J)
  EndFor
EndFor
```



# Transformations: Loop Fusion

- Fusion: Fuses two loops, also known as jamming (useful for locality enhancement). In example below, after fusion, you cannot have dependencies from S2 to S1



# Illegal Loop Fusion Example

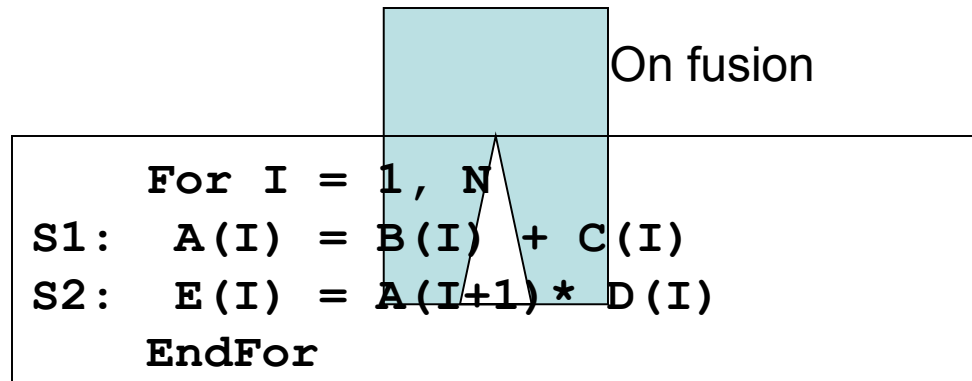
```
    For I = 1, N
S1:  A(I) = B(I) + C(I)
    EndFor
    For I = 1,N
S2:  E(I) = A(I+1)* D(I)
    EndFor
```

We have flow dependences from S1 to S2

# Illegal Loop Fusion Example

```
For I = 1, N
S1:  A(I) = B(I) + C(I)
EndFor
For I = 1, N
S2:  E(I) = A(I+1) * D(I)
EndFor
```

We have flow dependences from S1 to S2



**Illegal fusion:** On fusing the two loops, we have a violation of original data dependence



# Transformations: Loop Distribution

- ***Loop Distribution***: Splits a single loop nest into many, also known as ***loop fission***.

```
      For I = 1, N
S1:   A(I) = B(I)+C(I)
S2:   E(I) = A(I)*D(I)
      EndFor
```

```
For I = 1, N
    A(I) = B(I)+C(I)
EndFor
For I = 1,N
    E(I) = A(I)*D(I)
EndFor
```

- Like loop fusion, distribution is not always legal – must ensure that no data dependences are violated.
- Needed for vectorization

# Loop Unrolling

- Reduce number of iterations of loop but add statement(s) to loop body to do work of missing iterations
- Increases amount of instruction-level parallelism in loop body

```
for(j=0; j< 2*m; j++)  
{  
    Loop-Body(j)  
}
```



```
for(j=0; j< 2*m; j+=2)  
{  
    Loop-Body(j)  
    Loop-Body(j+1)  
}
```

```
for(i=0; i< n; i++)  
    for(j=0; j< 2*m; j++)  
    {  
        Loop-Body(i,j)  
    }
```

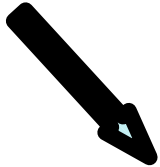


```
for(i=0; i< n; i++)  
    for(j=0; j< 2*m; j+=2)  
    {  
        Loop-Body(i,j)  
        Loop-Body(i,j+1)  
    }
```

# Example: Inner Loop Unrolling

```
// Assumes n is a multiple of 4
for(i=0;i<n;i++)
  for(j=0;j<n;j+=4) {
    y[i]=y[i]+a[i][j]*x[j];
    y[i]=y[i]+a[i][j+1]*x[j+1];
    y[i]=y[i]+a[i][j+2]*x[j+2];
    y[i]=y[i]+a[i][j+3]*x[j+3]; }

```




```
for(i=0;i<n;i++)
  for(j=0;j<n;j++)
    y[i]=y[i]+a[i][j]*x[j];

```

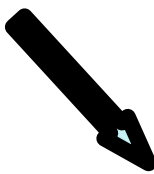
```
for(i=0;i<n;i++)
  for(j=0;j<n;j+=4) {
    y[i]=y[i]+a[i][j]*x[j];
    +a[i][j+1]*x[j+1];
    +a[i][j+2]*x[j+2];
    +a[i][j+3]*x[j+3]; }

```




# Outer Loop Unrolling (Unroll/Jam)

- Reduce number of iterations of an outer loop
- Simply replicating inner-loop structures will not increase op-level parallelism; need to fuse ("jam") replicated inner-loops
- Changes memory access order
  - Could reduce cache misses
  - Hence must verify validity of transformation



```
for(i=0;i<2*n;i++)  
  for(j=0;j< m;j++)  
    Loop-Body(i,j)
```

```
for(i=0;i<2*n;i+=2)  
{for(j=0;j<m;j++)  
  Loop-Body(i,j)    // 2-way outer-unroll  
  for(j=0;j<m;j++)  // does not increase  
    Loop-Body(i+1,j) // op-lvl parallelism  
}
```



```
for(i=0;i<2*n;i+=2)  
{for(j=0;j<m;j++)  
  {Loop-Body(i,j)    // unroll-jam increases  
  Loop-Body(i+1,j)} // op-lvl parallelism  
}
```

# Example: Outer Loop Unrolling

```
for(i=0;i<n;i++)  
  for(j=0;j<n;j++)  
    y[i]=y[i]+a[i][j]*x[j];
```



```
// Assumes n is a multiple of 4  
for(i=0;i<n;i+=4)  
  for(j=0;j<n;j++) {  
    y[i]=y[i]+a[i][j]*x[j];  
    y[i+1]=y[i+1]+a[i+1][j]*x[j];  
    y[i+2]=y[i+2]+a[i+2][j]*x[j];  
    y[i+3]=y[i+3]+a[i+3][j]*x[j];  
  }
```

# Improving Temporal Locality by Blocking

## Example: Blocked matrix multiplication

- “block” (in this context) does not mean “cache block”.
- Instead, it means a sub-block within the matrix.
- Example:  $N = 8$ ; sub-block size = 4

$$\begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} = \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix}$$

Key idea: Sub-blocks (i.e.,  $A_{xy}$ ) can be treated just like scalars.

$$C_{11} = A_{11}B_{11} + A_{12}B_{21}$$

$$C_{12} = A_{11}B_{12} + A_{12}B_{22}$$

$$C_{21} = A_{21}B_{11} + A_{22}B_{21}$$

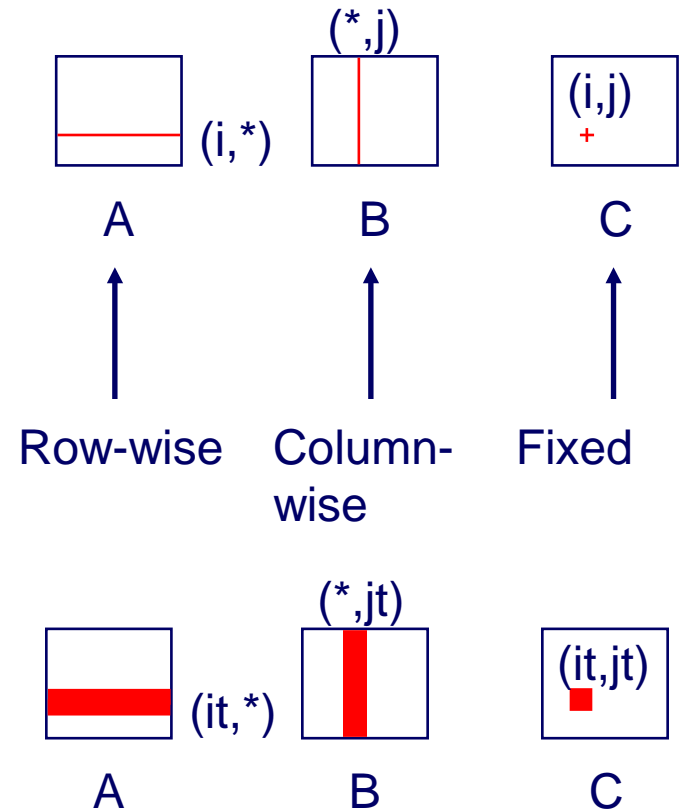
$$C_{22} = A_{21}B_{12} + A_{22}B_{22}$$

# Blocked Matrix Multiplication

```
/* ijk */  
for (i=0; i<n; i++)  
  for (j=0; j<n; j++)  
    for (k=0; k<n; k++)  
      c[i][j] += a[i][k]*b[k][j];
```

```
/* Tiled; assume n multiple of T */  
for (it=0; it<n; it+=T)  
  for (jt=0; jt<n; jt+=T)  
    for (kt=0; kt<n; kt+=T)  
      for (i=it; i<it+T; i++)  
        for (j=jt; j<jt+T; j++)  
          for (k=kt; k<kt+T; k++)  
            c[i][j] += a[i][k]*b[k][j];
```

Inner loop:



# Cache Misses: Blocked Mat-Mult

```
/* Tiled; assume n multiple of T */  
for (it=0; it<n; it+=T)  
  for (jt=0; jt<n; jt+=T)  
    for (kt=0; kt<n; kt+=T)  
      for (i=it; i<it+T; i++)  
        for (j=jt; j<jt+T; j++)  
          for (k=kt; k<kt+T; k++)  
            c[i][j] += a[i][k]*b[k][j];
```

Assume fully associative  
Cache of size  $> 3 \cdot T \cdot T$

sub-mat-mult

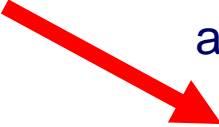
- Each sub-mat-mult involves product of two  $T \times T$  sub-matrices of A, B to contribute to a  $T \times T$  sub-matrix of C
- Each sub-mat-mult has at most  $3 \cdot (T^2/B)$  cache misses (no evictions during computation;  $T^2$  elements for each array)
- Number of result blocks of C:  $(N/T) \cdot (N/T) = N^2/T^2$
- Each C-block requires  $(N/T)$  sub-mat-mults
- Total cache misses  $\leq 3 \cdot (T^2/B) \cdot (N/T) \cdot N^2/T^2 = 3N^3/(B \cdot T)$
- T can be as large as  $\text{sqrt}(\text{CacheSize}/3)$



# Tiling = Loop-Split+Permutation

```
/* ijk */  
for (i=0; i<n; i++)  
  for (j=0; j<n; j++)  
    for (k=0; k<n; k++)  
      c[i][j] += a[i][k]*b[k][j];
```

Strip-mine each loop into  
a pair of equivalent loops



```
for (it=0; it<n; it+=T)  
  for (i=it; i<it+T; i++)  
    for (jt=0; jt<n; jt+=T)  
      for (j=jt; j<jt+T; j++)  
        for (kt=0; kt<n; kt+=T)  
          for (k=kt; k<kt+T; k++)  
            c[i][j] += a[i][k]*b[k][j];
```

```
for (it=0; it<n; it+=T)  
  for (jt=0; jt<n; jt+=T)  
    for (kt=0; kt<n; kt+=T)  
      for (i=it; i<it+T; i++)  
        for (j=jt; j<jt+T; j++)  
          for (k=kt; k<kt+T; k++)  
            c[i][j] += a[i][k]*b[k][j];
```

Loop Permutation



# Total Cache Miss Analysis: IJK

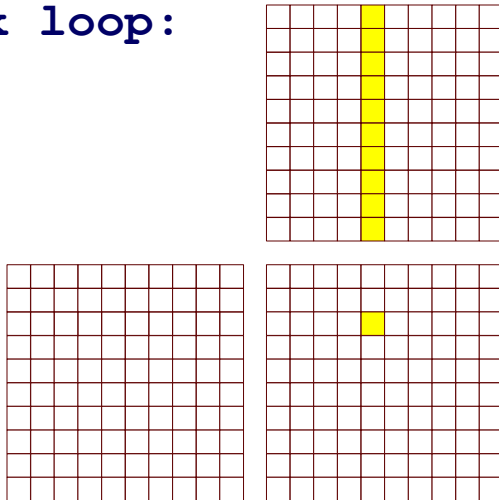
I  
J  
K

```

for ( i = 0; j < N; i++ )
  for ( j = 0; j < N; j++ )
    for ( k = 0; k < N; k++ )
      C[ i ][ j ] += A[ i ][ k ] x B[ k ][ j ]
    
```

let:  $C < B \cdot N$   
fully associative cache

k loop:



	<u>A</u>	<u>B</u>	<u>C</u>
I	N	N	N
J	1	N	$\frac{N}{B}$
K	$\frac{N}{B}$	N	1
	<hr/>		<hr/>
	$\frac{N^2}{B}$	N3	$\frac{N^2}{B}$

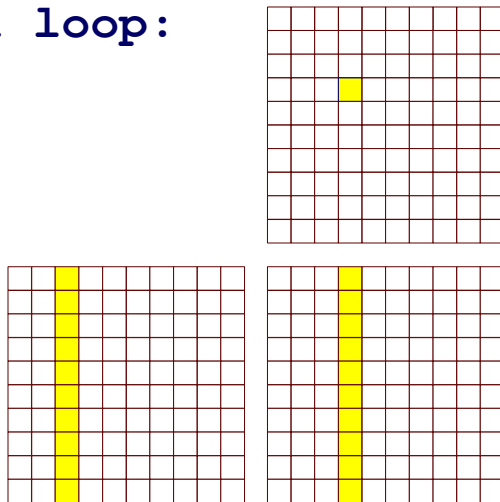
# Total Cache Miss Analysis: JKI

```

J for (j = 0; j < N; j++)
K   for (k = 0; k < N; k++)
I     for (i = 0; i < N; i++)
      C[i][j] += A[i][k] x B[k][j]
    
```

let:  $C < B \cdot N$   
 fully associative cache

i loop:



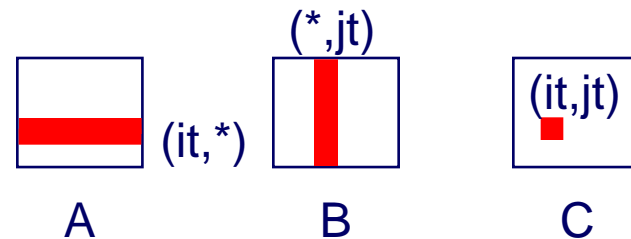
	<u>A</u>	<u>B</u>	<u>C</u>
J	N	N	N
K	N	N	N
I	N	1	N
	N3	N2	N3

# Blocked Matrix Multiply: Cache Misses

```

/* Tiled; assume n multiple of T */
for (it=0; it<n; it+=T)
  for (jt=0; jt<n; jt+=T)
    for (kt=0; kt<n; kt+=T)
      for (i=it; i<it+T; i++)
        for (j=jt; j<jt+T; j++)
          for (k=kt; k<kt+T; k++)
            c[i][j] += a[i][k]*b[k][j];

```



Assume fully associative  
Cache: size  $> 3 \cdot T \cdot T$   
But size  $< T \cdot N$

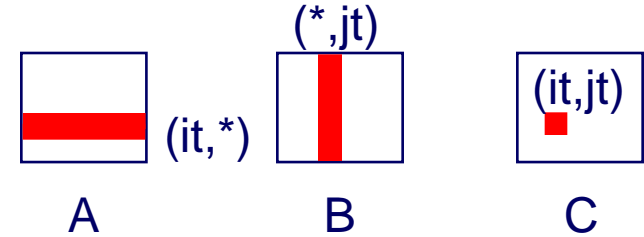
Loop	A	B	C
it			
jt			
kt			
i			
j			
k			

# Blocked Matrix Multiply: Cache Misses

```

/* Tiled; assume n multiple of T */
for (it=0; it<n; it+=T)
  for (jt=0; jt<n; jt+=T)
    for (kt=0; kt<n; kt+=T)
      for (i=it; i<it+T; i++)
        for (j=jt; j<jt+T; j++)
          for (k=kt; k<kt+T; k++)
            c[i][j] += a[i][k]*b[k][j];

```



Assume fully associative  
Cache: size  $> 3 \cdot T \cdot T$   
But size  $< T \cdot N$

Loop	A	B	C
it	N/T	N/T	N/T
jt	N/T	N/T	N/T
kt	N/T	N/T	1
i	T	1	T
j	1	T/B	T/B
k	T/B	T	1
Total	$N^2/(T^2)$	$N^2/(T^2)$	$N^2/(T^2)$

# Tiling: Arbitrary Bounds and Tilesize

```
/* ijk */  
for (i=0; i<m; i++)  
  for (j=0; j<n; j++)  
    for (k=0; k<p; k++)  
      c[i][j] += a[i][k]*b[k][j];
```

```
for (it=0; it<n; it+=Ti)  
  for (jt=0; jt<m; jt+=Tj)  
    for (kt=0; kt<p; kt+=Tk)  
      for (i=it; i< min(it+Ti,n); i++)  
        for (j=jt; j< min(jt+Tj,m); j++)  
          for (k=kt; k< min(kt+Tk,p); k++)  
            c[i][j] += a[i][k]*b[k][j];
```