# Locality, Matrix Multiplication, and Affine Transformations

Advanced Compiler Techniques

Source:

https://www.youtube.com/playlist?list=PLf3ZkSCyj1tf3rPAkOKY5hUzDrDoekAc7

Videos: 125 - 130

# Part 1: Locality

#### Locality: What is it? Why do we want it?

- Data locality refers to data accessed being near to each other, either in the spatial dimension, or the temporal dimension.
- Spatial locality: Addresses which are spatially near each other get accessed.
   Example: A-1, A, A+1, A+2, etc.
- Temporal locality: Same address is accessed again and again. Example: A, A, B, A, C, A, etc.
- Caches exploit locality to reduce the (average) memory latency observed by the execution pipeline.
- Typically more locality = More hits in caches = Faster Execution!

#### Interchanging Loops can affect locality

```
for(i = 0; i < 5; i++)
  for(j = 0; j < 10; j++)
      x = a[j]</pre>
```

#### Reordering Loops can affect locality

```
for(i = 0; i < 5; i++)
  for(j = 0; j < 10; j++)
     x = a[j]</pre>
```

High Spatial locality, low Temporal locality

a[6] a[7] a[8]	a[9] a	a[0] a[1]
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**Miss** 

#### Reordering Loops can affect locality

```
for(j = 0; j < 10; j++)

for(i = 0; i < 5; i++)

x = a[j]
```

High Spatial locality, high Temporal locality

a[0]
------

Miss

#### What parameters affect locality?

- Spatial: Prefetchers + CL size
- Temporal: Replacement Policy

# Part 2: Matrix Multiplication

#### Matrix Matrix Multiplication

```
C[row, col]: Can be reg allocated. One time cost.

A[row, idx]: No point in reg allocating. Might need to pay for every access.

B[row, idx]: No point in reg allocating. Might need to pay for every access.
```

#### Matrix Matrix Multiplication

```
for(row = 0; row < A_ROW_MAX; row++)
    for(col = 0; col < B_COL_MAX; col++)
        for(idx = 0; idx < A_COL_MAX; idx++)
        C[row, col] += A[row, idx] + B[idx, col]</pre>
```

Worst case: A -> Column major, B -> Row major Best case: A -> Row major, B -> Column major

Realistically, if both A and B are row major, and B\_COL\_MAX is sufficiently large, every access to B[idx, col] misses in cache. O(n³) misses.

#### Matrix Matrix Multiplication

```
for(row = 0; row < A_ROW_MAX; row++)
    for(col = 0; col < B_COL_MAX ; col++)
        for(idx = 0; idx < A_COL_MAX; idx++)
        C[row, col] += A[row, idx] + B[idx, col]</pre>
```

#### However, if

- There are more cache lines than there are rows in B
- 2. More than 1 element of B can completely in a cache line

The inner two loops will only have to face  $O(n^2/C)$  misses when the first column would be accessed where C = Cache Line Size / Size of one element of B = number of columns of B that can fit in the cache.

In best case when we have sufficiently high number of cache lines, all columns of B will fit, reducing the total penalty to  $O(n^2/C)$  for all three loops

#### What about A and C?

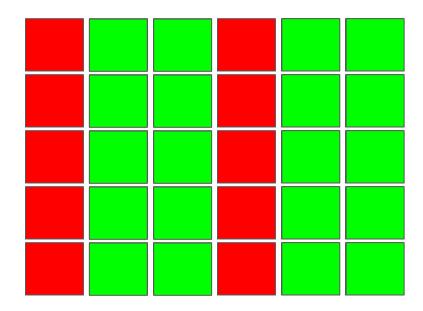
```
for(row = 0; row < A_ROW_MAX; row++)
    for(col = 0; col < B_COL_MAX ; col++)
        for(idx = 0; idx < A_COL_MAX; idx++)
        C[row, col] += A[row, idx] + B[idx, col]</pre>
```

Fastest moving index for both A and C is the columns. Therefore, if A and C are row-major, then we only need to pay  $O(n^2/C)$  for each of them.

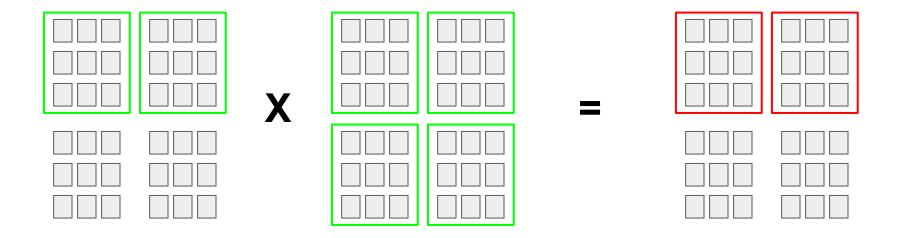
Therefore in total we pay around  $3n^2/C$ .

If C > n, we don't get any benefit as then useless data gets fetched into cache. If C = n, we pay around 3n penalty. This makes sense as we need n access for each matrix to bring the matrix into the cache. n accesses because each access brings a complete row, and there are n rows.

## Matrix Multiplication



## **Blocking Matrix Multiplication**



#### **Blocking Matrix Multiplication**

Best case, each block faces 3B<sup>2</sup>/C misses (B<sup>2</sup>/C for each sub-matrix). If each element is one byte long:

C = cache line size = l

Total (n/B)<sup>3</sup> block operations.

Therefore total cost =  $3n^3/BC$ 

B <= n and C <= n. Therefore, best case we get 3n penalty. This matches the best case estimate from earlier.

## Part 3: Affine Loop Transformations

#### What does affine mean?

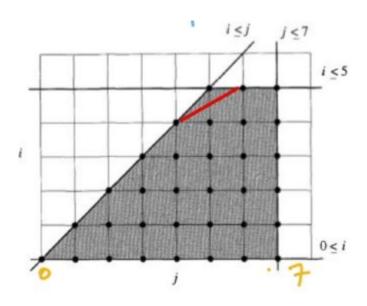
- An affine expression is a linear expression of the inputs.
- $f(x_1, x_2, x_3, ..., x_n) = c_0 + c_1x_1 + c_2x_2 + .... c_nx_n$

#### When can we use polyhedral optimizations?

- Even if accesses are affine, there might be dependencies.
- Upper and lower bounds of loops are affine functions of outer loop variables.
- Increments are by 1. This can be achieved by using a placeholder variable in loop and multiplying it by a constant before use.
- Under this assumption, the iteration space will always be convex.

#### Example

```
for (i = 0; i <= 5; i++)
for (j = i; j <= 7; j++)
Z[j,i] = 0;
```



#### Categories of Affine Transformations

- Splitting iteration space into independent slices which can be executed parallely.
- Blocking to create a hierarchy of iterations to improve locality.

### Example: Splitting Into Parallel Slices

```
for(idx = 0; idx < N; idx ++)
a[i] = b[i]</pre>
```

#### Example: Splitting Into Parallel Slices

```
block_size = m
p = ceil(n/m)
for(local_idx = m*p; local_idx < min(m*(p+1), n); local_idx++)
    a[local_idx] = b[local_idx]</pre>
```

#### Affine Transform Theory: Three Spaces

- 1. Iteration space: Set of all dynamic execution instances, i.e. all possible combinations of iterators. May/May not be rectangular.
- Data space: Set of array elements accessed. Typically defined as an affine functions of the iteration space.
- 3. Processor space: Set of all processors in the system. We create an affine function map from iteration space to processor space.

```
for (i = 0; i <= 5; i++)
for (j = i; j <= 7; j++)
Z[j,i] = 0;
```

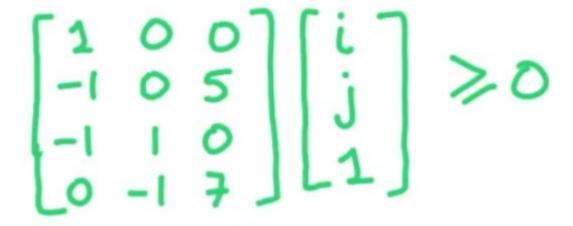
- i >= 0
- i <= 5
- j >= i
- j <= 7

```
for (i = 0; i <= 5; i++)
for (j = i; j <= 7; j++)
Z[j,i] = 0;
```

• 
$$-i + 5 >= 0$$

$$\bullet$$
 -i + j >= 0

• 
$$-j + 7 >= 0$$



$$\{ \underline{i} \in \mathbb{Z}^d \mid B\underline{i} + \underline{b} \geqslant 0 \}$$
 $\underline{i} = (\underline{i}) B = \begin{pmatrix} 1 & 0 \\ -1 & 0 \\ 0 & -1 \end{pmatrix} \underline{b} = \begin{pmatrix} 0 \\ 5 \\ 9 \end{pmatrix}$ 

## Iteration Space Example: Execution order

Fastest moving variable is lexicographically smaller.

$$\{ \underline{i} \in \mathbb{Z}^d \mid B\underline{i} + \underline{b} \geqslant \underline{0} \}^{\frac{1}{2}}$$
  
 $\underline{i} = (\underline{i}) B = (\underline{1}, 0) \underline{b} = (\underline{0}, 0)$   
 $\underline{i} = (\underline{0}) B = (\underline{1}, 0) \underline{b} = (\underline{0}, 0)$ 

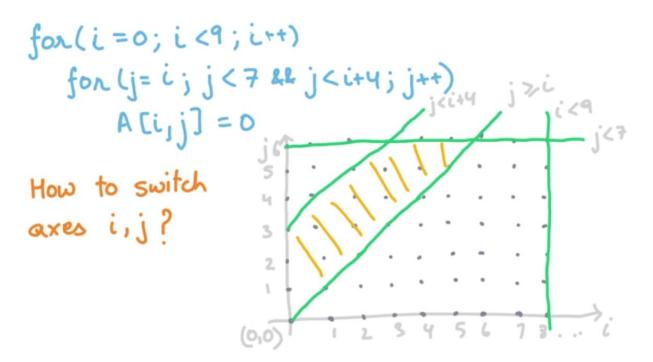
#### Iteration Space: Loop Invariant

for (i=0; i
\underbrace{i \in \mathbb{Z} \mid Bi+b \geqslant 0} ?

$$\underline{i} = (i) \quad B = \begin{pmatrix} 1 \\ -1 \end{pmatrix} \quad \underline{b} = \begin{pmatrix} 0 \\ n-1 \end{pmatrix}$$

#### **Exchanging Variables**

Suppose we want to exchange i and j



#### **Exchanging Variables**

- 1. Project the iteration space on j to find the range of j.
- 2. For each j, find i in terms of j.

We are guaranteed to have another convex polyhedron after projection.

#### **Exchanging Variables: Projection**

- The set of points  $(x_1, x_2, ..., x_m)$  will be in the projection of S on m dimensions if for some  $(x_{m+1}, x_{m+2}, ..., x_n)$ ,  $(x_1, x_2, ..., x_n)$  lies in S.
- Effective for every point in the projection, you should be able to find some set of values for the rest of n-m dimensions such that the n dimensional point lies in S.

#### Exchanging variables: Idea

- Project the iteration space onto all the dimensions except the desired innermost variable's dimension.
- Then project the rest onto all the dimensions except the desired second-innermost variable's dimension.
- Then project the rest onto all the dimensions except the desired third-innermost variable's dimension.

... and so on until all variables are exhausted.

#### Fourier Motzkin Method

#### Input:

- 1. A convex polyhedron S in m dimensions.
- 2. A variable  $x_m$  to eliminate

#### Output:

S', a projection of S on all dimensions except the m<sup>th</sup> dimension.

## Fourier Motzkin Method: Terminology

### Fourier Motzkin Method: Algorithm

Algorithm: 
$$S = \{ \underline{x} \mid \underline{B} \underline{x} + \underline{b} \geq 0 \}$$
 $C = \text{constraints in } S \text{ involving } x_m$ 

For every pair of lower bound and upper bound on  $x_m$  in  $C$  such that

 $L \leq C_1 * x_m$ 
 $C_1, C_2 \geq 0$ 
 $C_2 x_m \leq U$ 

add  $C_2 L \leq C_1 U$  to  $S^1$ 

Also add  $S-C$  to  $S^1$ 

#### Fourier Motzkin Method: Example Setup

for (i=0; i<9; i++)

for (j=i; j< min(7, i+4); j++)

A[i,j]=0

$$S = \{ \underline{i} \mid B \underline{i} + \underline{b} \ge 0 \} \qquad \underline{i} = (\underline{i})$$

$$B = \begin{pmatrix} 1 & 0 \\ -1 & 0 \\ 0 & -1 \\ 1 & -1 \end{pmatrix} \qquad \underline{b} = \begin{pmatrix} 0 \\ 8 \\ 0 \\ 6 \\ 3 \end{pmatrix} \qquad \text{Eliminate } i \text{ (project on } j)$$

#### Fourier Motzkin Method: Example Solution

#### **Example Solution**

# Thanks!

-Setu Gupta