

Lecture 25

Memory Hierarchy Optimizations & Locality Analysis

Caches: A Quick Review

- How do they work?
- Why do we care about them?
- What are typical configurations today?
- What are some important cache parameters that will affect performance?

Optimizing Cache Performance

- Things to enhance:
 - temporal locality
 - spatial locality
- Things to minimize:
 - conflicts (i.e. bad replacement decisions)

What can the *compiler* do to help?

Two Things We Can Manipulate

- Time:
 - When is an object accessed?
- Space:
 - Where does an object exist in the address space?

How do we exploit these two levers?

Time: Reordering Computation

- What makes it difficult to know *when* an object is accessed?
- How can we predict a *better time* to access it?
 - What information is needed?
- How do we know that this would be *safe*?

Space: Changing Data Layout

- What do we know about an object's *location*?
 - scalars, structures, pointer-based data structures, arrays, code, etc.
- How can we tell what a *better layout* would be?
 - how many can we create?
- To what extent can we *safely* alter the layout?

Types of Objects to Consider

- Scalars
- Structures & Pointers
- Arrays

Scalars

- Locals
- Globals
- Procedure arguments

- Is cache performance a concern here?
- If so, what can be done?

```
int x;  
double y;  
foo(int a) {  
    int i;  
    ...  
    x = a*i;  
    ...  
}
```

Structures and Pointers

- What can we do here?
 - within a node
 - across nodes

```
struct {  
    int count;  
    double velocity;  
    double inertia;  
    struct node *neighbors[N];  
} node;
```

- What limits the compiler's ability to optimize here?

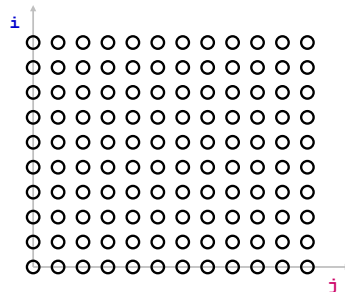
Arrays

```
double A[N][N], B[N][N];  
...  
for i = 0 to N-1  
    for j = 0 to N-1  
        A[i][j] = B[j][i];
```

- usually accessed within **loops nests**
 - makes it easy to understand "time"
- what we know about **array element addresses**:
 - start of array?
 - relative position within array

Handy Representation: "Iteration Space"

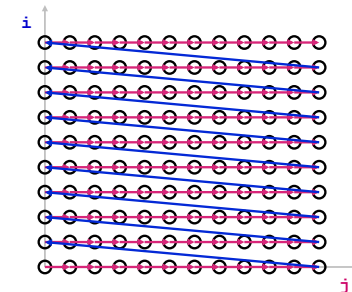
```
for i = 0 to N-1  
    for j = 0 to N-1  
        A[i][j] = B[j][i];
```



- each position represents an iteration

Visitation Order in Iteration Space

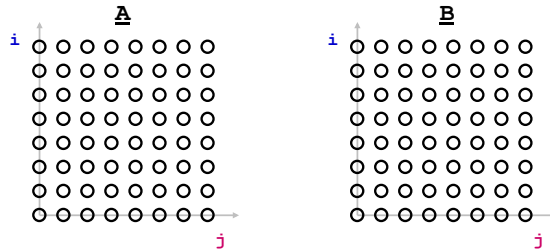
```
for i = 0 to N-1  
    for j = 0 to N-1  
        A[i][j] = B[j][i];
```



- Note: **iteration space** \neq **data space**

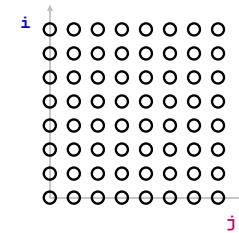
When Do Cache Misses Occur?

```
for i = 0 to N-1
  for j = 0 to N-1
    A[i][j] = B[j][i];
```



When Do Cache Misses Occur?

```
for i = 0 to N-1
  for j = 0 to N-1
    A[i+j][0] = i*j;
```



Optimizing the Cache Behavior of Array Accesses

- We need to answer the following questions:
 - when do cache misses occur?
 - use "locality analysis"
 - can we change the order of the iterations (or possibly data layout) to produce better behavior?
 - evaluate the cost of various alternatives
 - does the new ordering/layout still produce correct results?
 - use "dependence analysis"

Examples of Loop Transformations

- Loop Interchange
- Cache Blocking
- Skewing
- Loop Reversal
- ...

(we will briefly discuss the first two)

Loop Interchange

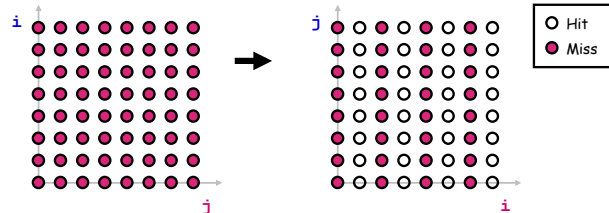
```

for i = 0 to N-1
  for j = 0 to N-1
    A[j][i] = i*j;
  
```

→

```

for j = 0 to N-1
  for i = 0 to N-1
    A[j][i] = i*j;
  
```



- (assuming N is large relative to cache size)

Cache Blocking (aka "Tiling")

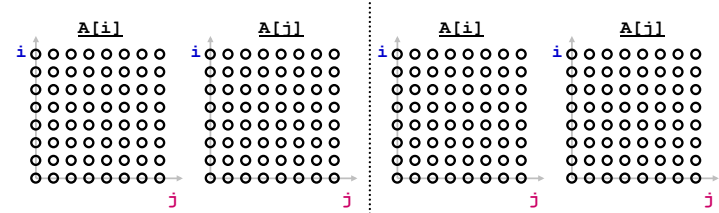
```

for i = 0 to N-1
  for j = 0 to N-1
    f(A[i], A[j]);
  
```

→

```

for JJ = 0 to N-1 by B
  for i = 0 to N-1
    for j = JJ to max(N-1, JJ+B-1)
      f(A[i], A[j]);
    
```



now we can exploit temporal locality

Impact on Visitation Order in Iteration Space

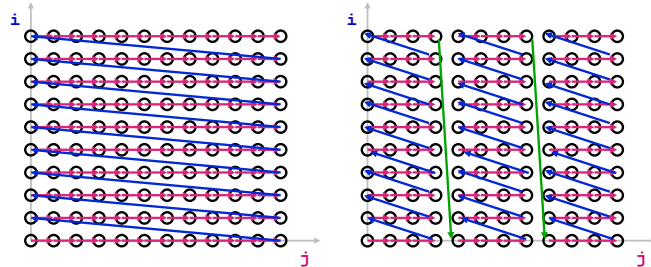
```

for i = 0 to N-1
  for j = 0 to N-1
    f(A[i], A[j]);
  
```

→

```

for JJ = 0 to N-1 by B
  for i = 0 to N-1
    for j = JJ to max(N-1, JJ+B-1)
      f(A[i], A[j]);
    
```



Cache Blocking in Two Dimensions

```

for i = 0 to N-1
  for j = 0 to N-1
    for k = 0 to N-1
      c[i,k] += a[i,j]*b[j,k];
    
```

→

```

for JJ = 0 to N-1 by B
  for KK = 0 to N-1 by B
    for i = 0 to N-1
      for j = JJ to max(N-1, JJ+B-1)
        for k = KK to max(N-1, KK+B-1)
          c[i,k] += a[i,j]*b[j,k];
        
```

- brings square sub-blocks of matrix "b" into the cache
- completely uses them up before moving on

Predicting Cache Behavior through "Locality Analysis"

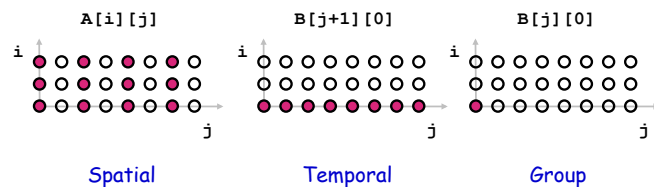
- Definitions:
 - Reuse:**
 - accessing a location that **has been accessed in the past**
 - Locality:**
 - accessing a location that is **now found in the cache**
- Key Insights
 - Locality only occurs when there is reuse!**
 - BUT, reuse does not necessarily result in locality.
 - why not?

Steps in Locality Analysis

- Find data reuse
 - if caches were infinitely large, we would be finished
- Determine "localized iteration space"
 - set of inner loops where the data accessed by an iteration is expected to fit within the cache
- Find data locality:
 - $\text{reuse} \cap \text{localized iteration space} \Rightarrow \text{locality}$

Types of Data Reuse/Locality

```
for i = 0 to 2
  for j = 0 to 100
    A[i][j] = B[j][0] + B[j+1][0];
```



Reuse Analysis: Representation

```
for i = 0 to 2
  for j = 0 to 100
    A[i][j] = B[j][0] + B[j+1][0];
```

- Map n loop indices into d array indices via array indexing function:

$$\vec{f}(\vec{i}) = H\vec{i} + \vec{c}$$

$$A[i][j] = A\left(\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} i \\ j \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \end{bmatrix}\right)$$

$$B[j][0] = B\left(\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} i \\ j \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \end{bmatrix}\right)$$

$$B[j+1][0] = B\left(\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} i \\ j \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix}\right)$$

Finding Temporal Reuse

- Temporal reuse occurs between iterations \vec{i}_1 and \vec{i}_2 whenever:

$$H\vec{i}_1 + \vec{c} = H\vec{i}_2 + \vec{c}$$

$$H(\vec{i}_1 - \vec{i}_2) = \vec{0}$$

- Rather than worrying about individual values of \vec{i}_1 and \vec{i}_2 we say that reuse occurs along **direction vector** \vec{r} when:

$$H(\vec{r}) = \vec{0}$$

- Solution:** compute the **nullspace** of H

Temporal Reuse Example

```
for i = 0 to 2
  for j = 0 to 100
    A[i][j] = B[j][0] + B[j+1][0];
```

- Reuse between iterations (i_1, j_1) and (i_2, j_2) whenever:

$$\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} i_1 \\ j_1 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} i_2 \\ j_2 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

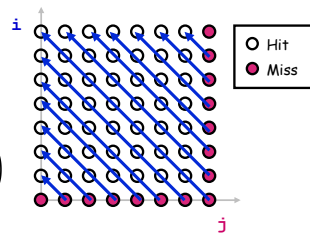
$$\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} i_1 - i_2 \\ j_1 - j_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

- True whenever $j_1 = j_2$, and regardless of the difference between i_1 and i_2 .
 - i.e. whenever the difference lies along the nullspace of $\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$, which is $\text{span}\{(1,0)\}$ (i.e. the outer loop).

More Complicated Example

```
for i = 0 to N-1
  for j = 0 to N-1
    A[i+j][0] = i*j;
```

$$A[i+j][0] = A\left(\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} i \\ j \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \end{bmatrix}\right)$$



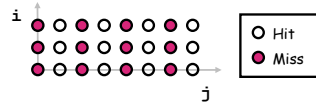
- Nullspace of $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} = \text{span}\{(1,-1)\}$.

Computing Spatial Reuse

- Replace last row of H with zeros, creating H_s
- Find the **nullspace** of H_s
- Result:** vector along which we access the same row

Computing Spatial Reuse: Example

```
for i = 0 to 2
  for j = 0 to 100
    A[i][j] = B[j][0] + B[j+1][0];
```

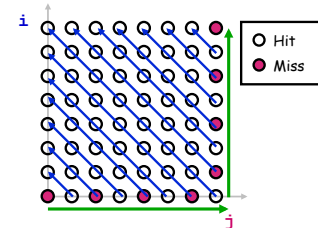


$$A[i][j] = A \left(\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} i \\ j \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \end{bmatrix} \right)$$

- $H_s = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$
- Nullspace of $H_s = \text{span}\{(0,1)\}$
 - i.e. access same row of $A[i][j]$ along inner loop

Computing Spatial Reuse: More Complicated Example

```
for i = 0 to N-1
  for j = 0 to N-1
    A[i+j] = i*j;
```



$$A[i+j] = A \left(\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} i \\ j \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \end{bmatrix} \right)$$

- $H_s = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$
- Nullspace of $H = \text{span}\{(1,-1)\}$
- Nullspace of $H_s = \text{span}\{(1,0), (0,1)\}$

Group Reuse

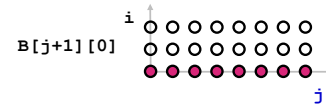
```
for i = 0 to 2
  for j = 0 to 100
    A[i][j] = B[j][0] + B[j+1][0];
```

- Only consider "uniformly generated sets"
 - index expressions differ only by constant terms
- Check whether they actually do access the same cache line
- Only the "leading reference" suffers the bulk of the cache misses

Localized Iteration Space

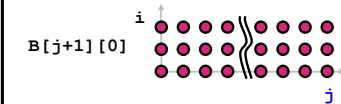
- Given finite cache, when does reuse result in locality?

```
for i = 0 to 2
  for j = 0 to 8
    A[i][j] = B[j][0] + B[j+1][0];
```



Localized: both i and j loops
(i.e. $\text{span}\{(1,0), (0,1)\}$)

```
for i = 0 to 2
  for j = 0 to 1000000
    A[i][j] = B[j][0] + B[j+1][0];
```




Localized: j loop only
(i.e. $\text{span}\{(0,1)\}$)

- Localized if accesses less data than effective cache size

Computing Locality

- Reuse Vector Space \cap Localized Vector Space \Rightarrow Locality Vector Space

• Example: `for i = 0 to 2`
 `for j = 0 to 100`
 `A[i][j] = B[j][0] + B[j+1][0];`



- If both loops are localized:
 - $\text{span}\{(1,0)\} \cap \text{span}\{(1,0),(0,1)\} \Rightarrow \text{span}\{(1,0)\}$
 - i.e. temporal reuse *does* result in temporal locality
- If only the innermost loop is localized:
 - $\text{span}\{(1,0)\} \cap \text{span}\{(0,1)\} \Rightarrow \text{span}\{\}$
 - i.e. *no* temporal locality