

# Optimization 2

**15-411/15-611 Compiler Design**

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# Common loop optimizations

- |  |   |
|--|---|
| <ul style="list-style-type: none"><li>• Hoisting of loop-invariant computations<ul style="list-style-type: none"><li>– pre-compute before entering the loop</li></ul></li><li>• Elimination of induction variables<ul style="list-style-type: none"><li>– change <math>p=i*w+b</math> to <math>p=b, p+=w</math>, when <math>w, b</math> invariant</li></ul></li><li>• Loop unrolling<ul style="list-style-type: none"><li>– to improve scheduling of the loop body</li></ul></li></ul> | Scalar opts,<br>DF analysis,<br>Control flow analysis |
| <ul style="list-style-type: none"><li>• Software pipelining<ul style="list-style-type: none"><li>– To improve scheduling of the loop body</li></ul></li><li>• Loop permutation<ul style="list-style-type: none"><li>– to improve cache memory performance</li></ul></li></ul>  | Requires<br>understanding<br>data<br>dependencies     |

# Loop Terminology

Loop: Strongly Connected Component of CFG

Entry Edge: tail not in loop, head in loop.

Exit edge: tail in loop, head not in loop

Loop Header: target of entry edge

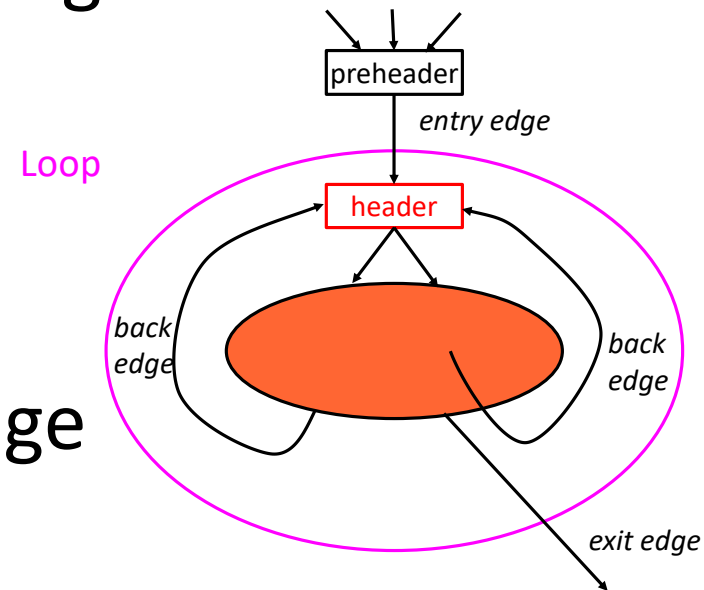
Back Edge: target is header,  
source is in loop

Preheader:

Source of the only entry edge

**Natural Loop:**

A Loop with only a single loop header



# **Loop optimizations: Hoisting of loop-invariant computations**

# Loop-invariant computations

- A definition

$$t = x \text{ op } y$$

in a loop is (conservatively) loop-invariant if

- x and y are constants, or
- all reaching definitions of x and y are outside the loop, or
- only one definition reaches x (and y), and that definition is loop-invariant
  - so keep marking iteratively

# Loop-invariant computations

- If not in SSA Be careful

```
t = expr;  
for () {  
    s = t * 2;  
    t = loop_invariant;  
    x = t + 2;  
    ...  
}
```

Of course, not an issue in SSA

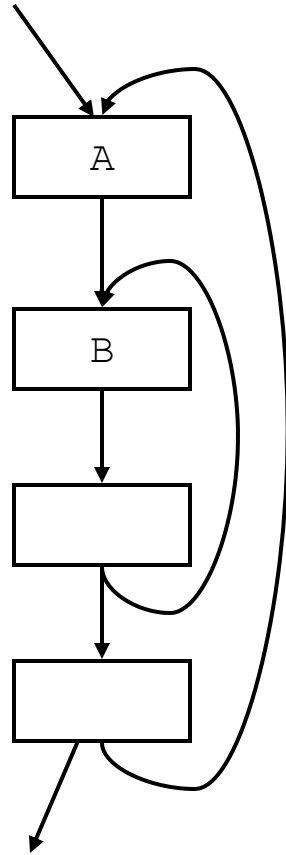
```
t1 = expr;  
L1:  
    brc L2;  
    t2 = phi(t1, t3);  
    s = t2 * 2;  
    t3 = loop_invariant_expr;  
    x1 = t3 * 2;  
    ...  
    jmp L1;  
L2:
```

- Even though t's two reaching expressions are each invariant, s is not invariant...

# Hoisting

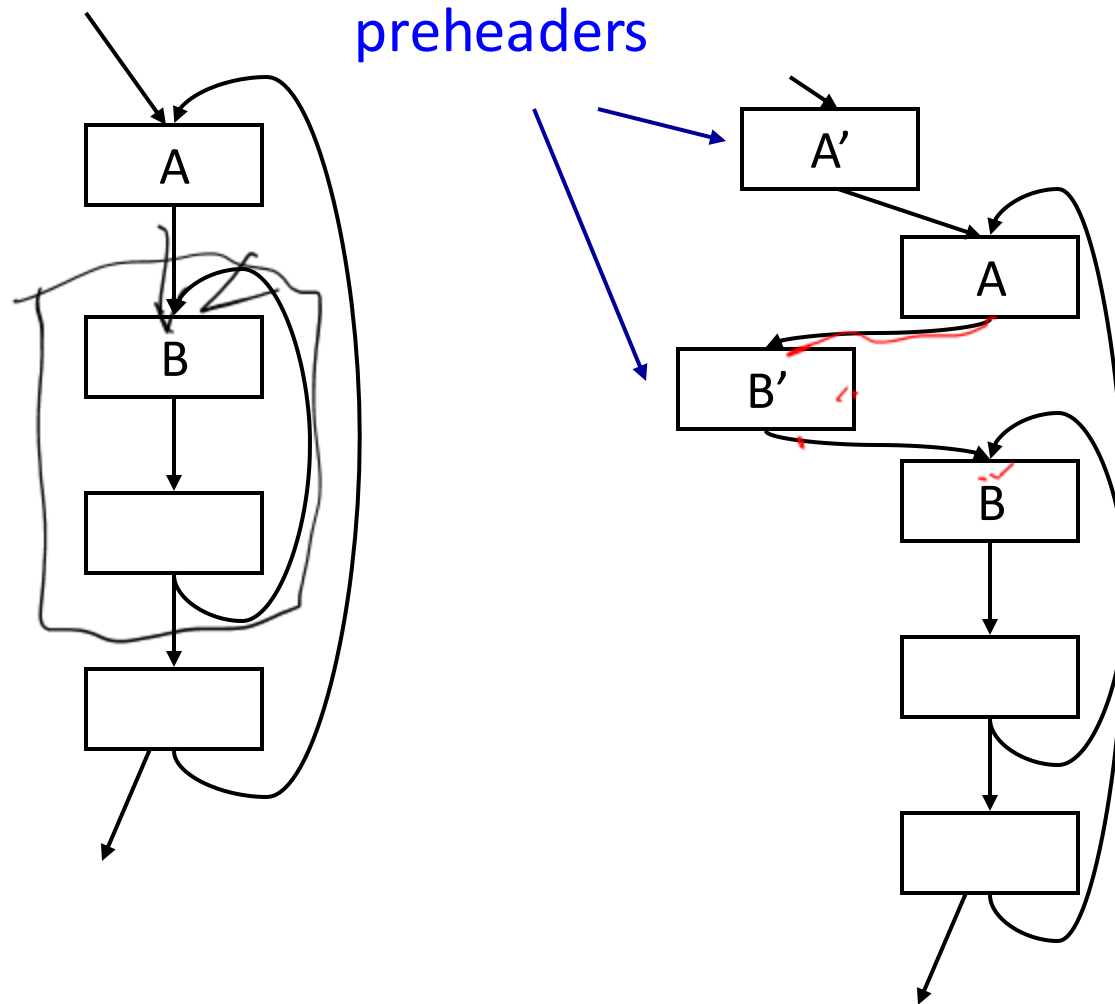
- In order to “hoist” a loop-invariant computation out of a loop, we need a place to put it
- We could copy it to all immediate predecessors (except along the back-edge) of the loop header...
- ...But we can avoid code duplication by ensuring there is a **pre-header**

# Hoisting Uses Pre-Headers





# Hoisting Uses Pre-Headers



# General Hoisting conditions

- For a loop-invariant definition

$d: t = x \text{ op } y$

- we can hoist  $d$  into the loop's pre-header only if

1.  $d$ 's block dominates all loop exits at which  $t$  is live-out,  
and

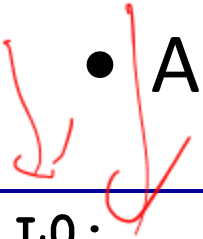
2.  $d$  is the only definition of  $t$  in the loop, and

3.  $t$  is not live-out of the pre-header



# We need to be careful...

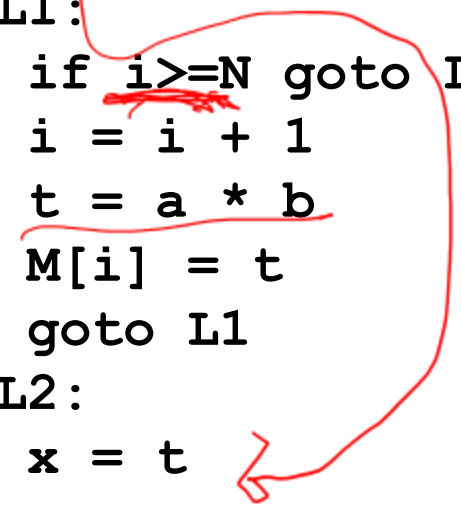
- All hoisting conditions must be satisfied!



```
L0:  
  t = 0  
L1:  
  i = i + 1  
  t = a * b  
  M[i] = t  
  if i < N goto L1  
L2:  
  x = t
```


OK

```
L0:  
  t = 0  
L1:  
  if i >= N goto L2  
  i = i + 1  
  t = a * b  
  M[i] = t  
  goto L1  
L2:  
  x = t
```



violates 1,3

```
L0:  
  t = 0  
L1:  
  i = i + 1  
  t = a * b  
  M[i] = t  
  t = 0  
  M[j] = t  
  if i < N goto L1  
L2:
```



violates 2

# We need to be careful...

- All hoisting conditions must be satisfied!

```
L0:  
  t = 0  
L1:  
  i = i + 1  
  t = a * b  
  M[i] = t  
  if i < N goto L1  
L2:  
  x = t
```

OK

```
L0:  
  t = 0  
L1:  
  if i >= N goto L2  
  i = i + 1  
  t = a * b  
  M[i] = t  
  goto L1  
L2:  
  x = t
```

violates 1,3

```
L0:  
  t = 0  
L1:  
  i = i + 1  
  t = a * b  
  M[i] = t  
  t = 0  
  M[j] = t  
  if i < N goto L1  
L2:
```

violates 2

# SSA Hoisting conditions

- For a loop-invariant definition

$d: t = x \text{ op } y$

- we can hoist  $d$  into the loop's pre-header only if

1.  $d$ 's block dominates all loop exits at which  $t$  is live-out, and

easy

2.  $d$  is the only definition of  $t$  in the loop, and

trivial

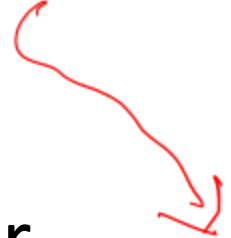
3.  $t$  is not live-out of the pre-header

easy

Condition 1:

- Can be violated if?
- Why would you?

$t = a * b$



# Enabling Transformations

- Convert while into repeat-until

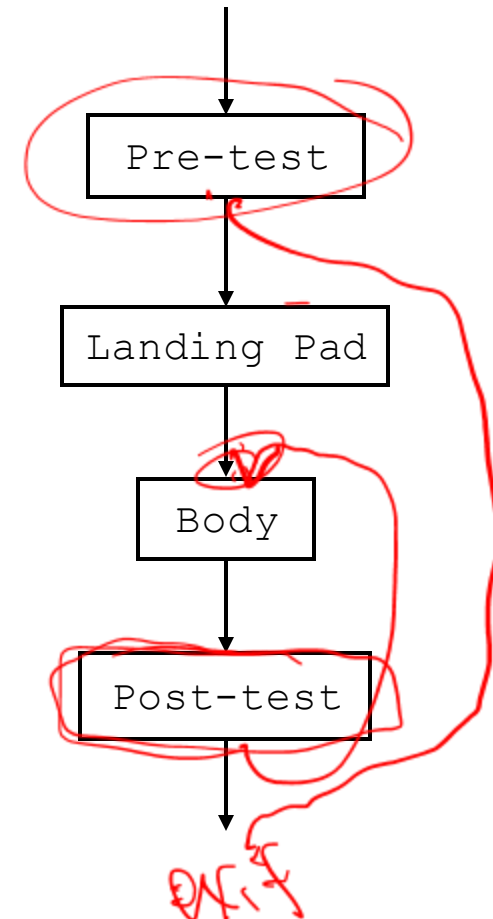
```
while (e) {  
  T  
  j = loopinv // does not dominate all loop exits  
  S  
}
```




```
If (e) {  
  repeat {  
    T  
    j = loopinv  
    S  
  } until (!e)  
}
```

# Enabling Transformations

- More Generally, add landing pad
  - For any speculative code:  
add test before pre-header



# Should You?

- Does Loop Body always execute?
- Do we speculate?
  - Use profiling information?
- Register Pressure? 



# LICM subsumed by PRE

- Don't have to implement Loop invariant code motion if you implement PRE, since PRE subsumes it anyway!
- (But, PRE is difficult)

**Loop optimizations:**  
**Induction-variable Elimination**  
**Strength reduction**

# The basic idea of IVE

- Suppose we have a loop variable
  - $i$  initially 0; each iteration  $i = i + 1$
- and a variable that linearly depends on it:  
 $x = i * c1 + c2$
- In such cases, we can try to
  - initialize  $x = i_0 * c1 + c2$  (execute once)
  - increment  $x$  by  $c1$  each iteration

# Induction Variable

- Basic Induction Variable has the form:

$$X = X \pm C$$

where  $C$  is constant or loop-invariant

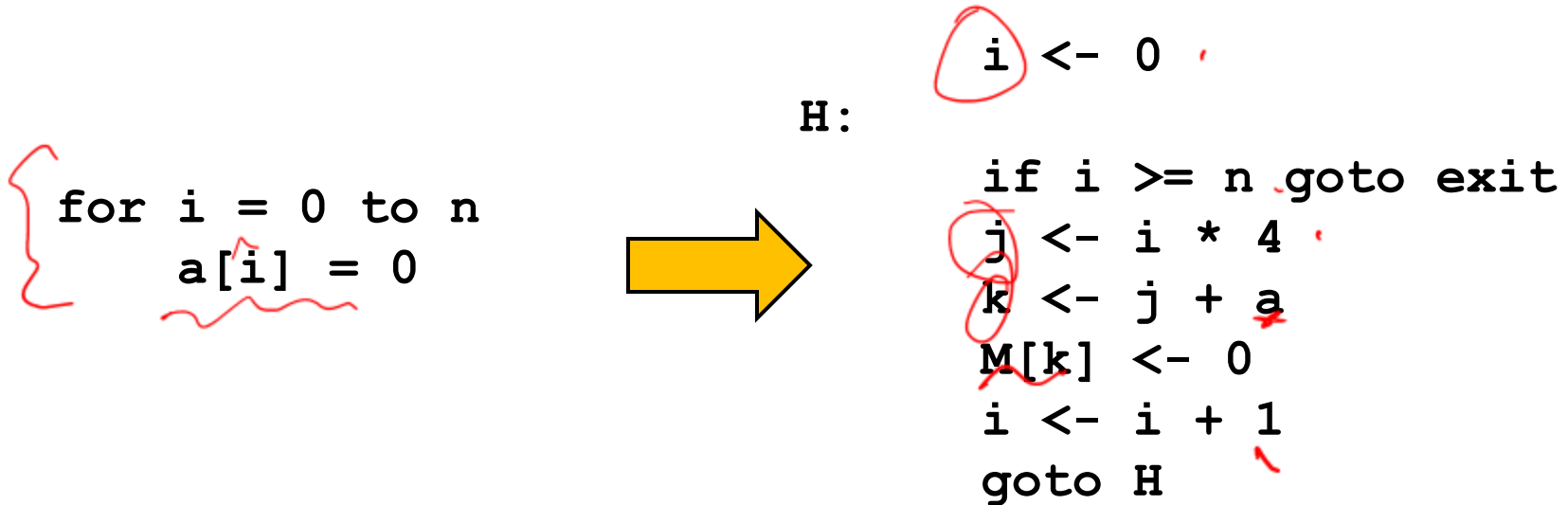
- Derived Induction Variable has form:

$$X = C_1 * Y \pm C_2$$

where

- $Y$  is a Basic induction variable
- $C_1$  and  $C_2$  are constants

# Simple Example of IVE



Clearly, `j` & `k` do not need to be computed anew each time since they are related to `i` and `i` changes linearly.

# Simple Example of IVE

H:

```
i <- 0
```

```
if i >= n goto exit
```

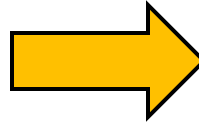
```
j <- i * 4
```

```
k <- j + a
```

```
M[k] <- 0
```

```
i <- i + 1
```

```
goto H
```



H:

```
i <- 0
```

```
j' <- 0
```

```
k' <- a
```

```
if i >= n goto exit
```

```
j <- j'
```

```
k <- k'
```

```
M[k] <- 0
```

```
i <- i + 1
```

```
j' <- j' + 4
```

```
k' <- k' + 4
```

```
goto H
```

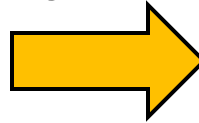
But, then we don't even need j (or j')

# Simple Example of IVE

```
i <- 0  
j' <- 0  
k' <- a
```

H:

```
if i >= n goto exit  
j <- j'  
k <- k'  
M[k] <- 0  
i <- i + 1  
j' <- j' + 4  
k' <- k' + 4  
goto H
```



```
i <- 0  
k' <- a
```

H:

```
if i >= n goto exit  
k <- k'  
M[k] <- 0  
i <- i + 1  
k' <- k' + 4  
goto H
```

Do we need i?

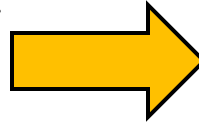
# Simple Example of IVE

Rewrite comparison

```
i <- 0  
k' <- a
```

H:

```
if i >= n goto exit  
k <- k'  
M[k] <- 0  
i <- i + 1  
k' <- k' + 4  
goto H
```



H:

```
i <- 0  
k' <- a
```

```
if k' >= a + (n*4) goto exit  
k <- k'  
M[k] <- 0  
k' <- k' + 4  
goto H
```

Red arrows point to the expression  $a + (n*4)$  in the rewritten code, indicating it is the loop invariant.

But,  $a + (n*4)$  is loop invariant



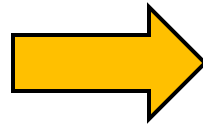
# Simple Example of IVE

Invariant code motion on  $a+(n*4)$

```
i <- 0  
k' <- a
```

H:

```
if k' >= a+(n*4) goto exit  
k <- k'  
M[k] <- 0  
k' <- k' + 4  
goto H
```



H:

```
k' <- a  
n' <- a + (n * 4)  
  
if k' >= n' goto exit  
k <- k'  
M[k] <- 0  
k' <- k' + 4  
goto H
```

now, we do copy propagation and eliminate k

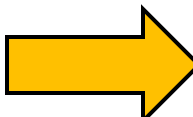
# Simple Example of IVE

## Copy propagation

H:

```
k' <- a
n' <- a + (n * 4)

if k' >= n' goto exit
k <- k'
M[k] <- 0
k' <- k' + 4
goto H
```



H:

```
k' <- a
n' <- a + (n * 4)

if k' >= n' goto exit
M[k'] <- 0
k' <- k' + 4
goto H
```

Voila!

# Simple Example of IVE

Compare original and result of IVE

```
    i <- 0
H:   if i >= n goto exit
      j <- i * 4
      k <- j + a
      M[k] <- 0
      i <- i + 1
      goto H
```

```
    k' <- a
    n' <- a + (n * 4)
H:   if k' >= n' goto exit
      M[k'] <- 0
      k' <- k' + 4
      goto H
```

Voila!

# What we did

- identified induction variables  $(i, j, k)$
- strength reduction (changed  $*$  into  $+$ )
- dead-code elimination ( $j \leftarrow j'$ )
- useless-variable elimination ( $j' \leftarrow j' + 4$ )  
(This can also be done with ADCE)
- loop invariant identification & code-motion
- almost useless-variable elimination (i)
- copy propagation

# Is it faster?

- On some hardware, adds are much faster than multiplies
- Fewer instructions (better \$ behavior)
- Furthermore, one fewer value is computed,
  - thus potentially saving a register
  - and decreasing the possibility of spilling
- Can be used to eliminate bounds checking in loop

# Loop preparation

- Before attempting IVE, it is best to first perform :
  - constant propagation & constant folding
  - copy propagation
  - loop-invariant hoisting

# How to do it, step 1

- First, find the **basic IVs**
  - scan loop body for defs of the form
$$x = x + c \text{ or } x = x - c$$
where  $c$  is loop-invariant
  - record these basic IVs as

$$x = (x, 0, c)$$

- this represents the IV:  $x = x * c$

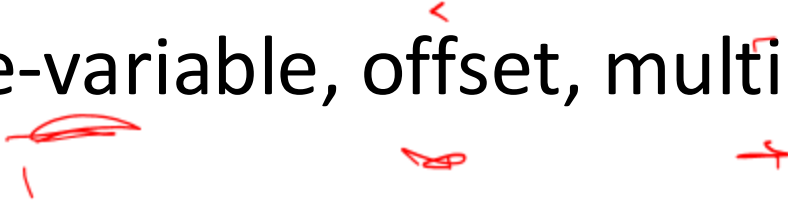
$$x = \phi(x_j, x^r)$$

$$x' = x + c$$

# Representing IVs

- Characterize all induction variables by:

(base-variable, offset, multiple)



– where the offset and multiple are loop-invariant

- IOW, after an induction variable is defined it equals:

offset + multiple \* base-variable





## How to do it, step 2

- Scan for **derived IVs** of the form

$$k = i * c1 + c2$$

- where  $i$  is a basic IV,

this is the only def of k in the loop, and  
c1 and c2 are loop invariant

- We say  $k$  is in the family of  $i$

- Record as  $k = (i, c_2, c_1)$

# How to do it, step 3

- Iterate, looking for derived IVs of the form

$$k = j * c1 + c2$$

- where IV  $j = (i, a, b)$ , and
  - this is the only def of  $k$  in the loop, and
  - there is no def of  $i$  between the def of  $j$  and the def of  $k$
  - $c1$  and  $c2$  are loop invariant
- Record as  $k = (i, a*c1, b*c1+c2)$

# Simple Example of IVE

**i** <- 0

H:

```
if i >= n goto exit
j <- i * 4
k <- j + a
M[k] <- 0
i <- i + 1
goto H
```

**i: (i, 0, 1)**      i.e.,  $i = 0 + 1 * i$

**j: (i, 0, 4)**      i.e.,  $j = 0 + 4 * i$

**k: (i, a, 4)**      i.e.,  $k = a + 4 * i$

So, j & k are in family of i

# Identifying Induction Variables

- Two steps:
  - Find Basic IVs of form  $i \leftarrow i \pm c$
  - Find Derived IVs of form  $k \leftarrow j * c$  or  $k \leftarrow j \pm c$

# Finding Basic IVs

- Maintain two tables:
  - basic: Holds all vars that can be basic IV
  - other: Holds all vars that cannot be basic IV
- Scan stmts in loop:
  - if  $i \leftarrow i \pm c$  and  $i \notin \text{other}$ , then put in basic
  - if  $i \leftarrow$  anything else, then remove from basic and put in other

# Finding Derived IVs

- Scan statements to create worklist W
  - if var defined more than once and var  $\notin$  basic, then, put into other
  - if stmt uses any var  $\in$  basic, insert into W
- Repeat until W is empty:
  - if s has form “ $k \leftarrow j * x$ ” or “ $k \leftarrow j \pm x$ ” AND  $k \notin \text{other}$  AND  $x$  is loop invariant, then
    - if  $j \in \text{basic}$ , then  $k$  is derived IV  
enter  $k$  into derivedTable  
put all stmts using  $k$  into W

# Finding Derived IVs

- Repeat until W is empty:
  - if s has form “ $k \leftarrow j * x$ ” or “ $k \leftarrow j \pm x$ ” AND  
k  $\notin$  other AND  
x is loop invariant, then
    - if  $j \in \text{basic}$ , then k is derived IV  
enter k into derivedTable  
put all stmts using k into W
    - else if  $j \in \text{derivedTable}$ , then
      - if only def of j reaching k is in loop AND  
only 1 def reaches k AND  
no assignment to i between j & k, then  
put k in derivedTable  
put all stmts using k into W



# Tracking tuples

- As we gather IVs we record:  
(base, offset, multiple) for each one
- For IV  $k$ :
  - if it is basic, the record:  $(k, 0, c)$
  - else if defined as “ $k \leftarrow j * x$ ” AND  $j$  has  $(i, a, b)$   
record:  $(i, a*x, b*x)$
  - else if defined as “ $k \leftarrow j \pm x$ ” AND  $j$  has  $(i, a, b)$   
record:  $(i, a \pm x, b)$



# IV Optimizations

- Once we have identified all IVs and recorded their tuples, we perform 3 optimizations:
  - strength reduction
  - useless-variable elimination
  - Comparison rewriting

# How to do it, step 4

- This is the strength reduction step
- For an induction variable  $k = (i, c1, c2)$ 
  - initialize  $k = i * c2 + c1$  in the preheader
  - replace  $k$ 's def in the loop by
$$k = k + c2$$
  - make sure to do this after  $i$ 's def

# How to do it, step 5

- This is the comparison rewriting step
- For an induction variable  $k = (i, a_k, b_k)$ 
  - If  $k$  used only in definition and comparison
  - There exists another variable,  $j$ , in the same class and is not “useless” and  $j = (i, a_j, b_j)$
- Rewrite  $k < n$  as
$$j < (b_j/b_k)(n - a_k) + a_j$$
- Note: since they are in same class:
$$(j - a_j)/b_j = (k - a_k)/b_k$$

# Notes

- Are the  $c1$ ,  $c2$  constant, or just invariant?
  - if constant, then you can keep folding them: they're always a constant even for derived IVs
  - otherwise, they can be expressions of loop-invariant variables
- But if constant, can find IVs of the type
$$x = i/b$$
and know that it's legal, if  $b$  evenly divides the stride...

# Is it faster? (2)

- On some hardware, adds are much faster than multiplies
- But...not always a win!
  - Constant multiplies might otherwise be reduced to shifts/adds that result in even better code than IVE
  - Scaling of addresses ( $i*4$ ) might come for free on your processor's address modes
- So maybe: only convert  $i*c1+c2$  when  $c1$  is loop invariant but not a constant
- Or, can be used to eliminate bound check!

# Loop Unrolling

- For loops with a small body:
  - significant portion of time spent incrementing and testing induction variables
  - May be stalled due to dependencies (more on this later)
- Loop unrolling reduces overhead (and increases opportunity for superscalar to tolerate latencies) by copying body of loop

# Unroll Mechanism

- A loop L with header h and backedges  $s_i \rightarrow h$ 
  - copy L to a new loop L' with header h' and backedges  $s'_i \rightarrow h'$
  - changes edges  $s_i \rightarrow h$  in L to  $s_i \rightarrow h'$
  - change backedges in L' from  $s'_i \rightarrow h$
- Change IVs
- Must deal with potential left over iterations in an epilogue



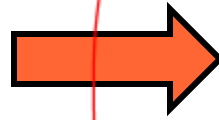
# IV changes for unrolling

- Eliminate IV in L
- create new IV,  $i' \leftarrow i+c$  that dominates all back edges of new loop
- Change uses of IV,  $i$ , to be proper offset
- change final test of IV to account for  $\Delta$  unrolls.
- Finally, insert epilogue to deal with left overs.



# Simple Example

```
i <- 0  
H:  
  cmp i, n  
  jg exit  
  sum <- sum + a[i]  
  i <- i + 1  
  jmp H  
exit:
```



```
i <- 0  
H:  
  cmp i, n  
  jg exit  
  sum <- sum + a[i]  
  i <- i + 1  
  jmp H1:  
H1:  
  cmp i, n  
  jg exit  
  sum <- sum + a[i]  
  i <- i + 1  
  jmp H:
```

# Simple Example

```
i <- 0
```

```
H:
```

```
  cmp i, n
```

```
  jg exit
```

```
  sum <- sum + a[i]
```

```
  i <- i + 1
```

```
  jmp H1:
```

```
H1:
```

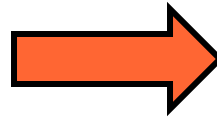
```
  cmp i, n
```

```
  jg exit
```

```
  sum <- sum + a[i]
```

```
  i <- i + 1
```

```
  jmp H:
```



```
i <- 0
```

```
H:
```

```
  cmp i, n
```

```
  jg exit
```

```
  sum <- sum + a[i]
```

```
  cmp i, n
```

```
  jg exit
```

```
  sum <- sum + a[i+1]
```

```
  i <- i + 2
```

```
  jmp H:
```

# Simple Example

```
i <- 0
```

```
H:
```

```
cmp i, n
```

```
jg exit
```

```
sum <- sum + a[i]
```

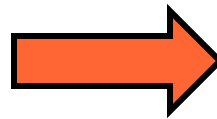
```
{ cmp i, n
```

```
jg exit
```

```
sum <- sum + a[i+1]
```

```
i <- i + 2
```

```
jmp H:
```



```
i <- 0
```

```
H:
```

```
cmp i, n-1
```

```
jg exit
```

```
sum <- sum + a[i]
```

```
sum <- sum + a[i+1]
```

```
i <- i + 2
```

```
jmp H:
```

```
exit:
```

```
H1:
```

```
cmp i, n
```

```
jg exit1
```

```
sum <- sum + a[i]
```

```
i <- i + 1
```

```
jmp H1:
```

```
exit1:
```

# Common loop optimizations

- Hoisting of loop-invariant computations
    - pre-compute before entering the loop
  - Elimination of induction variables
    - change  $p=i*w+b$  to  $p=b, p+=w$ , when  $w, b$  invariant
  - Loop unrolling
    - to improve scheduling of the loop body
  - Software pipelining
    - To improve scheduling of the loop body
  - Loop permutation
    - to improve cache memory performance
- Requires understanding data dependencies

# Dependencies in Loops

- Loop independent data dependence occurs between accesses in the **same** loop iteration.
- Loop-carried data dependence occurs between accesses across **different** loop iterations.
- There is data dependence between access **a** at iteration **i-k** and access **b** at iteration **i** when:
  - a and b access the same memory location
  - There is a path from **a** to **b**
  - Either **a** or **b** is a write

*dependence distance*

$$\text{for } (i=1; i < n; i++)$$

$$A[i] = \dots$$

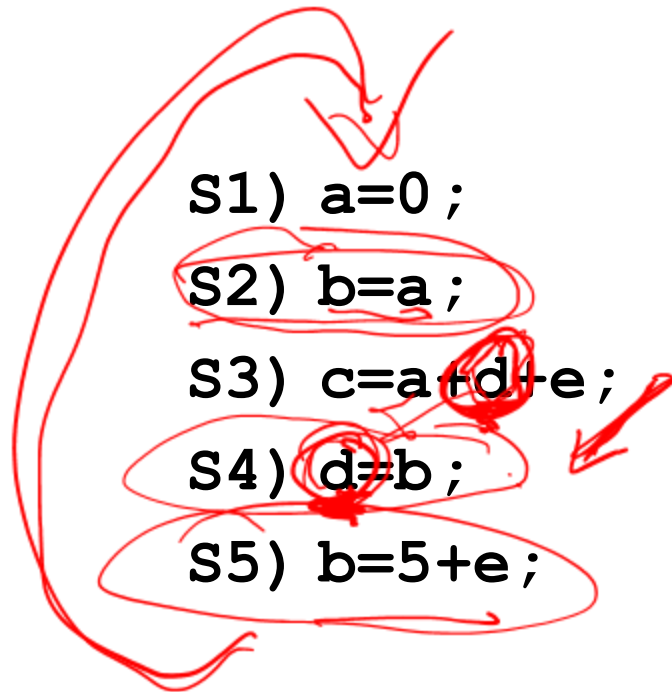

---


$$A[i] = A[i-2] + 2$$

# Defining Dependencies

- Flow Dependence
- Anti-Dependence
- Output Dependence

$W \rightarrow R$	$\delta^f$	} true
$R \rightarrow W$	$\delta^a$	
$W \rightarrow W$	$\delta^o$	} false

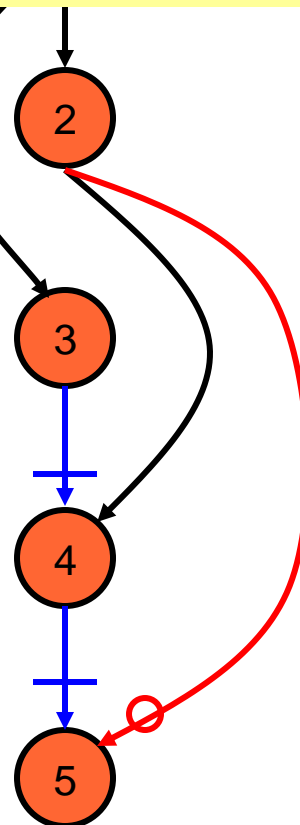


# Example Dependencies

S1)  $a=0$  ;  
 S2)  $b=a$  ;  
 S3)  $c=a+d+e$  ;  
 S4)  $d=b$  ;  
 S5)  $b=5+e$  ;

These are scalar dependencies. The same idea holds for memory accesses.

<u>source</u>	<u>type</u>	<u>target</u>	<u>due to</u>
S1	$\delta^f$	S2	a
S1	$\delta^f$	S3	a
S2	$\delta^f$	S4	b
S3	$\delta^a$	S4	d
S4	$\delta^a$	S5	b
S2	$\delta^o$	S5	b



- What can we do with this information?
- What are anti- and flow- called “false” dependences?

# Data Dependence in Loops

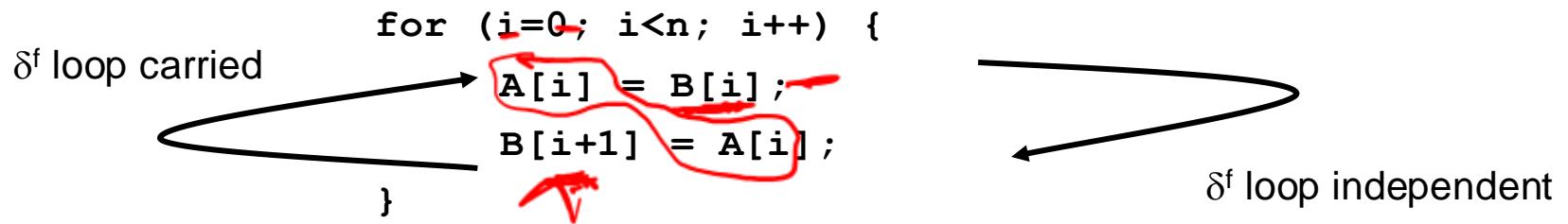
- Dependence can flow across iterations of the loop.
- Dependence information is annotated with iteration information.
- If dependence is across iterations it is **loop carried** otherwise **loop independent**.

```
for (i=0; i<n; i++) {  
    A[i] = B[i];  
    B[i+1] = A[i];  
}
```



# Data Dependence in Loops

- Dependence can flow across iterations of the loop.
- Dependence information is annotated with iteration information.
- If dependence is across iterations it is **loop carried** otherwise **loop independent**.



# Unroll Loop to Find Dependencies

$\delta^f$  loop carried

```
for (i=0; i<n; i++) {  
    A[i] = B[i];  
    B[i+1] = A[i];  
}
```

$\delta^f$  loop independent

$A[0] = B[0];$   
 $B[1] = A[0];$   
 $A[1] = B[1];$   
 $B[2] = A[1];$   
 $A[2] = B[2];$   
 $B[3] = A[2];$   
...

$i=0$   
 $i=1$   
 $i=2$

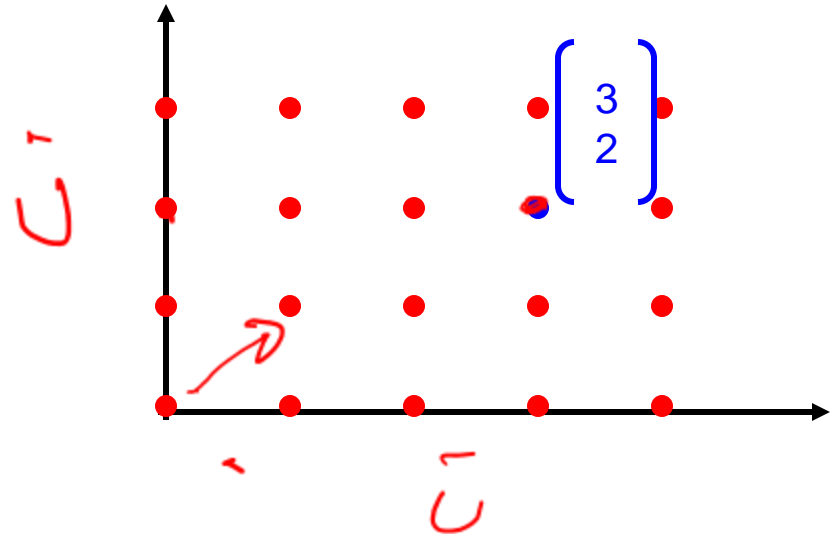
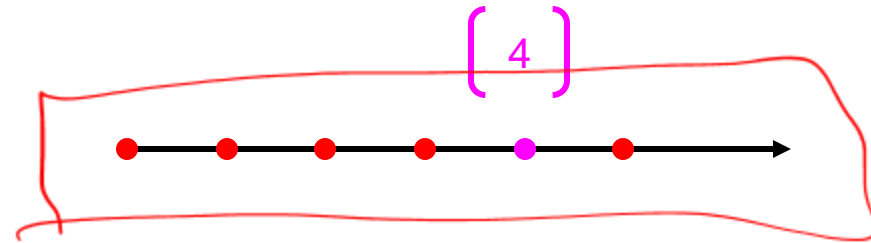
Distance/Direction of the dependence is also important.

# Iteration Space

Every iteration generates a point in an n-dimensional space, where n is the depth of the loop nest.

```
for (i=0; i<n; i++) {  
    ...  
}
```

```
for (i=0; i<n; i++)  
    for (j=0; j<4; j++) {  
        ...  
    }
```



(1)

# Distance Vector



```
for (i=0; i<n; i++) {  
    A[i] = B[i];  
    B[i+1] = A[i];  
}
```

Distance vector is the difference between the target and source iterations.

$$\mathbf{d} = \mathbf{l}_t - \mathbf{l}_s$$

Exactly the distance of the dependence, i.e.,

$$\mathbf{l}_s + \mathbf{d} = \mathbf{l}_t$$

**A**[0] = B[0];

**B**[1] = **A**[0];

**A**[1] = **B**[1];

**B**[2] = **A**[1];

**A**[2] = **B**[2];

**B**[3] = **A**[2];

•  
•  
•

i=0

i=1

i=2



# Example of Distance Vectors

$\begin{pmatrix} 3 \\ 1 \end{pmatrix} \begin{pmatrix} 2 \\ 2 \end{pmatrix}$

```
for (i=0; i<n; i++)
  for (j=0; j<m; j++){
    A[i,j] = ;
    = A[i,j];
    B[i,j+1] = ;
    = B[i,j];
    C[i+1,j] = ;
    = C[i,j+1] ;
  }
```

j

A <sub>0,2</sub> = =A <sub>0,2</sub> B <sub>0,3</sub> = =B <sub>0,2</sub> C <sub>1,2</sub> = =C <sub>0,3</sub>	A <sub>1,2</sub> = =A <sub>1,2</sub> B <sub>1,3</sub> = =B <sub>1,2</sub> C <sub>2,2</sub> = =C <sub>1,3</sub>	A <sub>2,2</sub> = =A <sub>2,2</sub> B <sub>2,3</sub> = =B <sub>2,2</sub> C <sub>3,2</sub> = =C <sub>2,3</sub>
A <sub>0,1</sub> = =A <sub>0,1</sub> B <sub>0,2</sub> = =B <sub>0,1</sub> C <sub>1,1</sub> = =C <sub>0,2</sub>	A <sub>1,1</sub> = =A <sub>1,1</sub> B <sub>1,2</sub> = =B <sub>1,1</sub> C <sub>2,1</sub> = =C <sub>1,2</sub>	A <sub>2,1</sub> = =A <sub>2,1</sub> B <sub>2,2</sub> = =B <sub>2,1</sub> C <sub>3,1</sub> = =C <sub>2,2</sub>
A <sub>0,0</sub> = =A <sub>0,0</sub> B <sub>0,1</sub> = =B <sub>0,0</sub> C <sub>1,0</sub> = =C <sub>0,1</sub>	A <sub>1,0</sub> = =A <sub>1,0</sub> B <sub>1,1</sub> = =B <sub>1,0</sub> C <sub>2,0</sub> = =C <sub>1,1</sub>	A <sub>2,0</sub> = =A <sub>2,0</sub> B <sub>2,1</sub> = =B <sub>2,0</sub> C <sub>3,0</sub> = =C <sub>2,1</sub>

$\begin{pmatrix} 0 \\ 0 \end{pmatrix}$

$\begin{pmatrix} 0 \\ 1 \end{pmatrix}$

i

$\begin{pmatrix} 1 \\ -1 \end{pmatrix}$

# Example of Distance Vectors

```

for (i=0; i<n; i++)
  for (j=0; j<m; j++) {
    A[i,j] =      ;
    = A[i,j];
    B[i,j+1] =    ;
    = B[i,j];
    C[i+1,j] =    ;
    = C[i,j+1] ;
  }
    
```

$A_{0,2} = A_{0,2}$ $B_{0,3} = B_{0,2}$ $C_{1,2} = C_{0,3}$	$A_{1,2} = A_{1,2}$ $B_{1,3} = B_{1,2}$ $C_{2,2} = C_{1,3}$	$A_{2,2} = A_{2,2}$ $B_{2,3} = B_{2,2}$ $C_{3,2} = C_{2,3}$
$A_{0,1} = A_{0,1}$ $B_{0,2} = B_{0,1}$ $C_{1,1} = C_{0,2}$	$A_{1,1} = A_{1,1}$ $B_{1,2} = B_{1,1}$ $C_{2,1} = C_{1,2}$	$A_{2,1} = A_{2,1}$ $B_{2,2} = B_{2,1}$ $C_{3,1} = C_{2,2}$
$A_{0,0} = A_{0,0}$ $B_{0,1} = B_{0,0}$ $C_{1,0} = C_{0,1}$	$A_{1,0} = A_{1,0}$ $B_{1,1} = B_{1,0}$ $C_{2,0} = C_{1,1}$	$A_{2,0} = A_{2,0}$ $B_{2,1} = B_{2,0}$ $C_{3,0} = C_{2,1}$

A yields:  $\begin{pmatrix} 0 \\ 0 \end{pmatrix}$

B yields:  $\begin{pmatrix} 0 \\ 1 \end{pmatrix}$

C yields:  $\begin{pmatrix} 1 \\ -1 \end{pmatrix}$