

# Compiler Design

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1402-1403

# Local Optimization

- **The Use of Algebraic Identities**

- Arithmetic identities can be applied to **eliminate computations** from a basic block

$x + 0 = 0 + x = x$	$x - 0 = x$
$x \times 1 = 1 \times x = x$	$x / 1 = x$

- **Local reduction in strength**, that is, replacing a more expensive operator by a cheaper one

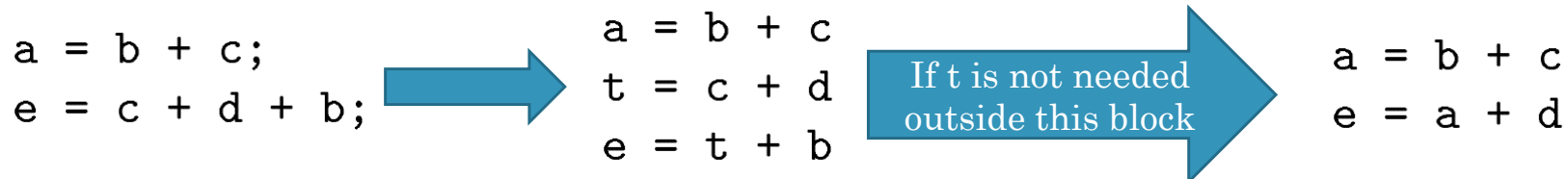
EXPENSIVE		CHEAPER
$x^2$	=	$x \times x$
$2 \times x$	=	$x + x$
$x / 2$	=	$x \times 0.5$

- **Constant folding**, that is, evaluating constant expressions at compile time and replace the constant expressions by their values
  - The expression **2 \* 3.14** would be replaced by **6.28**

# Local Optimization

- **The Use of Algebraic Identities**

- We can apply algebraic transformations such as **commutativity** and **associativity**
  - For example,  $*$  is commutative; that is,  $x * y = y * x$ 
    - Before we create a new node labeled  $*$  with left child  $M$  and right child  $N$ , we always check whether such a node already exists
    - However, because  $*$  is commutative, we should then check for a node having operator  $*$ , left child  $N$ , and right child  $M$
- The **relational operators** sometimes generate unexpected common subexpressions
  - For example, the condition  $x > y$  can also be tested by subtracting
- **Associative laws** might also be applicable to expose common subexpressions



# Local Optimization

- **Representation of Array References**

- We might be tempted to optimize by replacing the third instruction  $z = a[i]$  by the simpler  $z = x$
- However, since  $j$  could equal  $i$ , the middle statement may in fact change the value of  $a[i]$ ; thus, it is not legal to make this change

```
x = a[i]
a[j] = y
z = a[i]
```

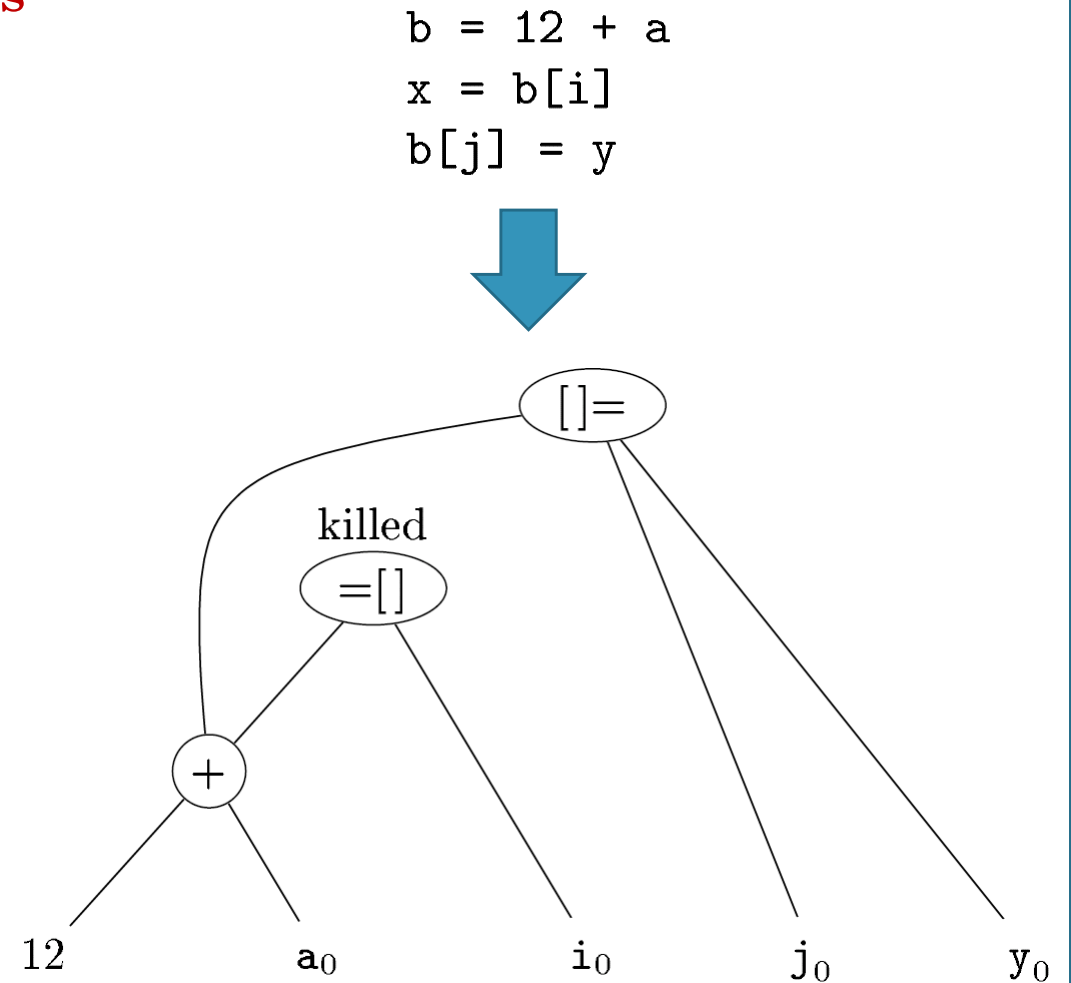
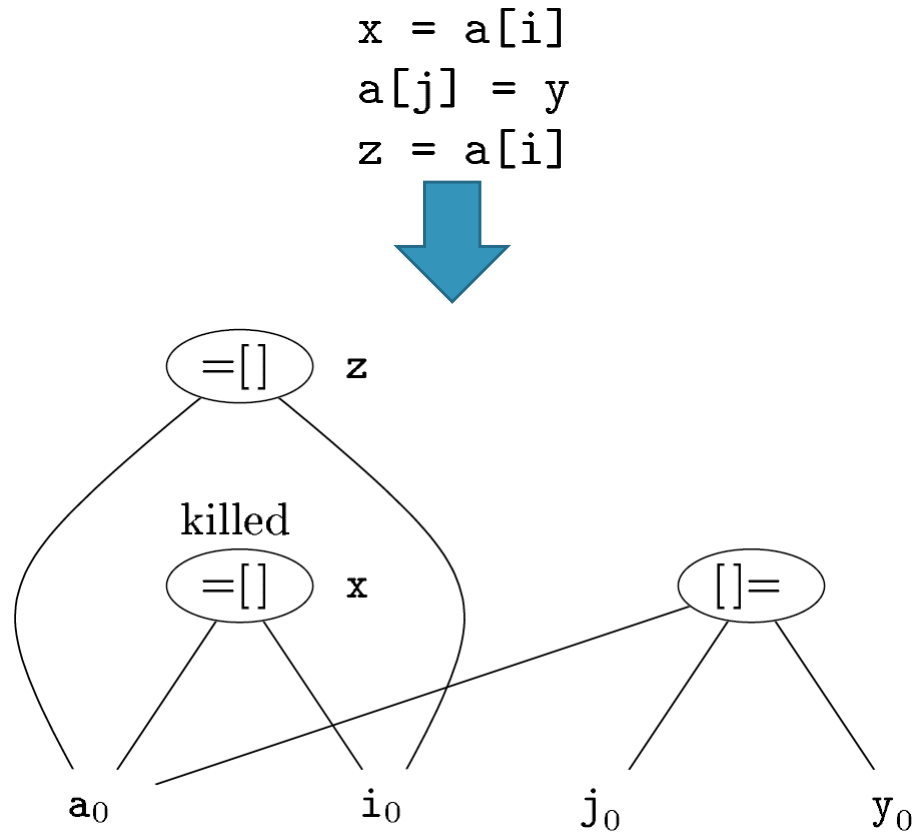
# Local Optimization

- **Representation of Array References**

- The proper way to represent array accesses in a DAG is as follows
  - An assignment from an array, like  $x = a[i]$ , is represented by creating a node with operator  $= []$  and two children representing the initial value of the array,  $a0$  in this case, and the index  $i$ 
    - Variable  $x$  becomes a label of this new node
  - An assignment to an array, like  $a[j] = y$ , is represented by a new node with operator  $[] =$  and three children representing  $a0$ ,  $j$  and  $y$ 
    - There is no variable labeling this node
    - The creation of this node kills all currently constructed nodes whose value depends on  $a0$

# Local Optimization

- **Representation of Array References**
  - **Example**



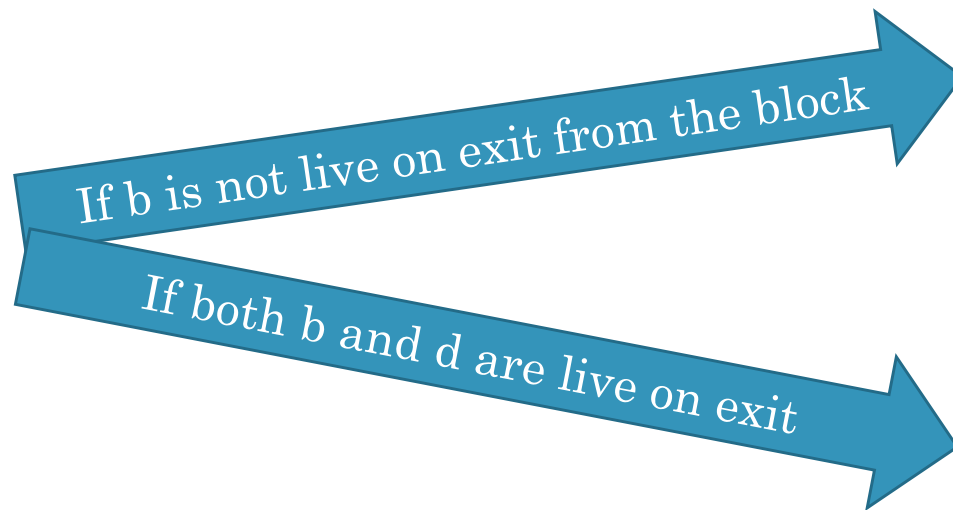
# Local Optimization

- **Reassembling Basic Blocks From DAGs**

- For each node that has one or more attached variables, we construct a three-address statement that computes the value of one of those variables
- If we do not have global live-variable information to work from, we need to assume that every variable of the program (*but not temporaries that are generated by the compiler to process expressions*) is live on exit from the block

- **Example**

a = b + c  
b = a - d  
c = b + c  
d = a - d



a = b + c  
d = a - d  
c = d + c

a = b + c  
d = a - d  
b = d  
c = d + c

# Global Optimization

- Most global optimizations are based on data-flow analyses
- A compiler optimization must preserve the semantics of the original program
- Except in very special circumstances, the compiler cannot understand enough about the program to replace it with a substantially different and more efficient algorithm
- *A compiler knows only how to apply relatively low-level semantic transformations*



# Global Optimization

- **Causes of Redundancy**

- Sometimes the redundancy is available at the **source level**
  - For instance, a programmer may find it more direct and convenient to recalculate some result
- But more often, the redundancy is a side effect of **having written the program in a high-level language**

# Global Optimization

- **Example: Quicksort**

```
void quicksort(int m, int n)
    /* recursively sorts a[m] through a[n] */
{
    int i, j;
    int v, x;
    if (n <= m) return;
    /* fragment begins here */
    i = m-1; j = n; v = a[n];
    while (1) {
        do i = i+1; while (a[i] < v);
        do j = j-1; while (a[j] > v);
        if (i >= j) break;
        x = a[i]; a[i] = a[j]; a[j] = x; /* swap a[i], a[j] */
    }
    x = a[i]; a[i] = a[n]; a[n] = x; /* swap a[i], a[n] */
    /* fragment ends here */
    quicksort(m,j); quicksort(i+1,n);
}
```

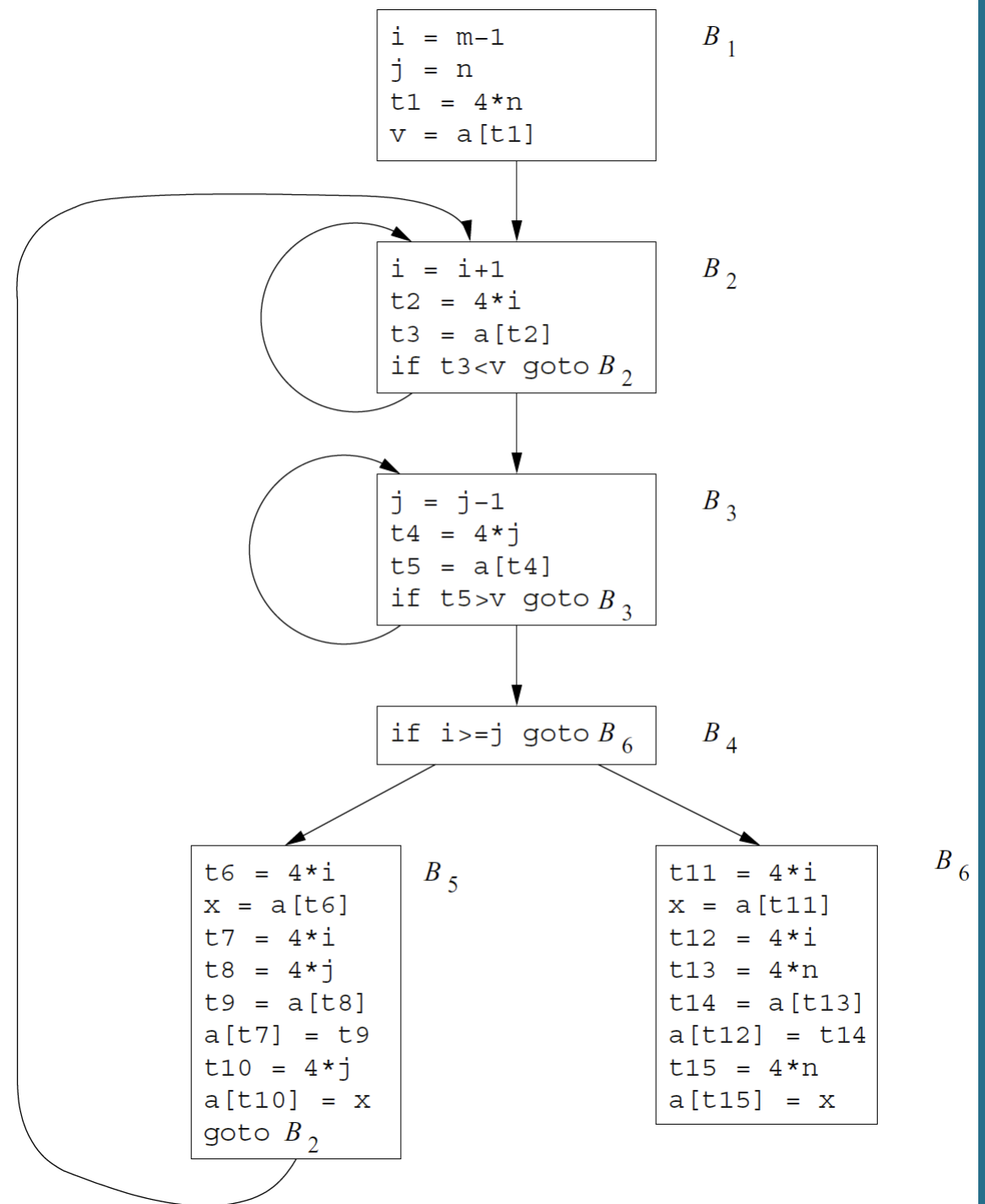
# Global Optimization

- **Three-address code for the quicksort fragment**

(1)	i = m-1	(16)	t7 = 4*i
(2)	j = n	(17)	t8 = 4*j
(3)	t1 = 4*n	(18)	t9 = a[t8]
(4)	v = a[t1]	(19)	a[t7] = t9
(5)	i = i+1	(20)	t10 = 4*j
(6)	t2 = 4*i	(21)	a[t10] = x
(7)	t3 = a[t2]	(22)	goto (5)
(8)	if t3<v goto (5)	(23)	t11 = 4*i
(9)	j = j-1	(24)	x = a[t11]
(10)	t4 = 4*j	(25)	t12 = 4*i
(11)	t5 = a[t4]	(26)	t13 = 4*n
(12)	if t5>v goto (9)	(27)	t14 = a[t13]
(13)	if i>=j goto (23)	(28)	a[t12] = t14
(14)	t6 = 4*i	(29)	t15 = 4*n
(15)	x = a[t6]	(30)	a[t15] = x

# Global Optimization

- Flow graph for the quicksort fragment



# Global Optimization

- **Semantics-Preserving Transformations**
  - A program will include several calculations of the same value
  - **Example:** Local common-subexpression elimination

```
t6 = 4*i  
x = a[t6]  
t7 = 4*i  
t8 = 4*j  
t9 = a[t8]  
a[t7] = t9  
t10 = 4*j  
a[t10] = x  
goto B2
```

$B_5$

(a) Before.

```
t6 = 4*i  
x = a[t6]  
t8 = 4*j  
t9 = a[t8]  
a[t6] = t9  
a[t8] = x  
goto B2
```

$B_5$

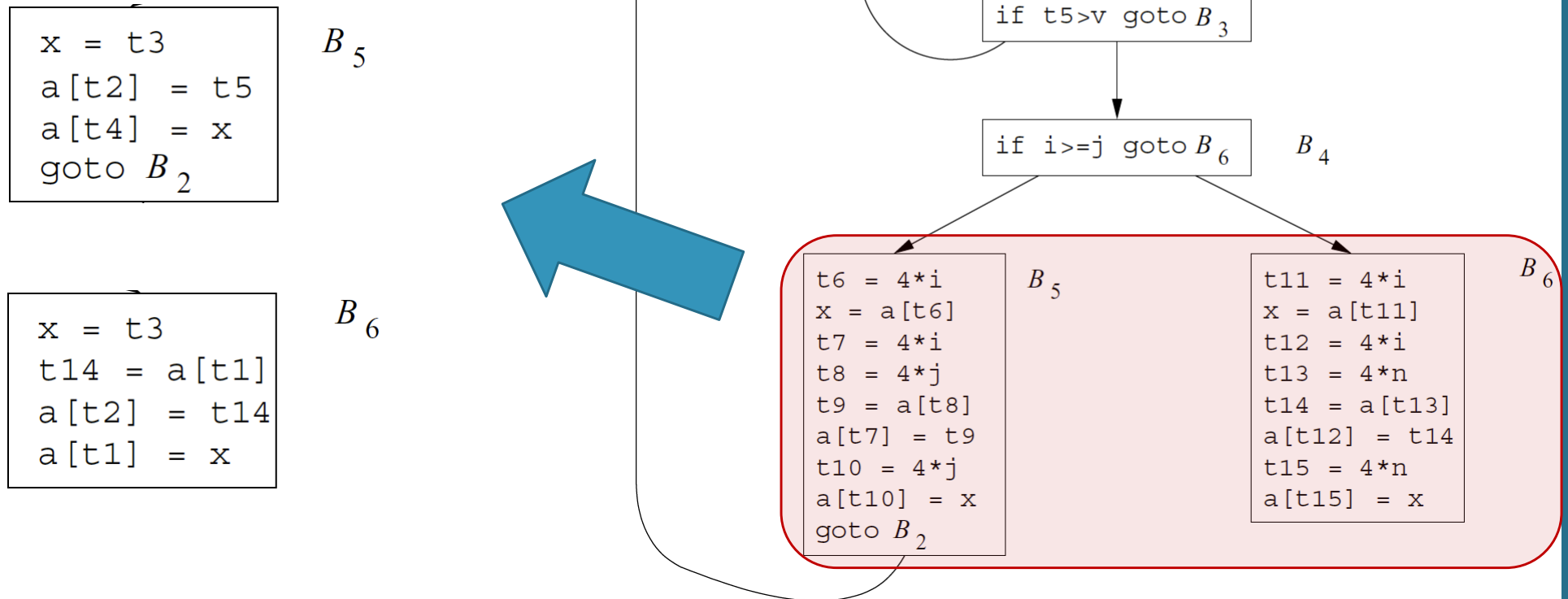
(b) After.

# Global Optimization

- **Global Common Subexpressions**
  - An occurrence of an expression  $E$  is called a common subexpression if  $E$  was previously computed and the values of the variables in  $E$  have not changed since the previous computation

# Global Optimization

- Example:** Eliminating both global and local common subexpressions from blocks B5 and B6



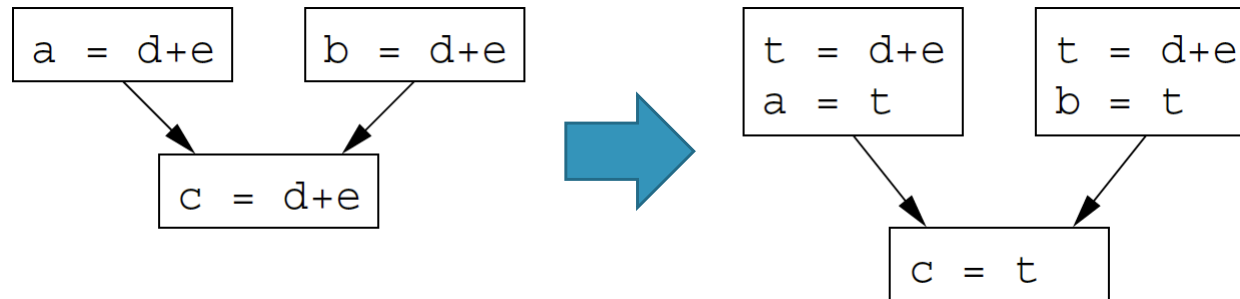
# Global Optimization

- **Copy Propagation**

- The idea behind the copy-propagation transformation is to use  $v$  for  $u$ , wherever possible after the copy statement  $u = v$

- **Example**

- Since control may reach  $c = d+e$  either after the assignment to  $a$  or after the assignment to  $b$ , it would be incorrect to replace  $c = d+e$  by either  $c = a$  or by  $c = b$





# Global Optimization

- **Copy Propagation**

```
x = t3  
a[t2] = t5  
a[t4] = x  
goto B2
```



Basic block B5 after copy  
propagation

```
x = t3  
a[t2] = t5  
a[t4] = t3  
goto B2
```

# Global Optimization

- **Dead-Code Elimination**


- A variable is live at a point in a program if its value can be used subsequently; otherwise, it is dead at that point

- **Example:**

- It may be possible for the compiler to deduce that each time the program reaches this statement, the value of `debug` is `FALSE`
- We can eliminate both the test and the print operation from the object code

`if (debug) print ...`

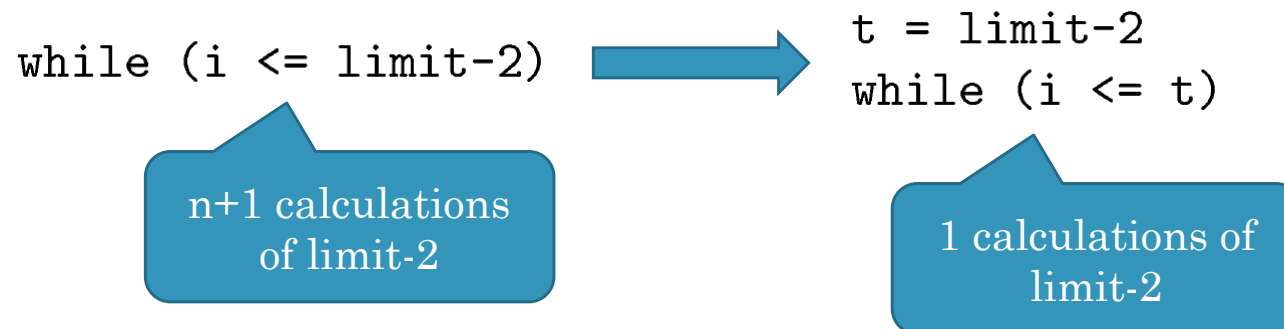
- Deducing at compile time that the value of an expression is a constant and using the constant instead is known as *constant folding*
- One advantage of copy propagation is that it often turns the copy statement into dead code

<code>x = t3</code>		
<code>a[t2] = t5</code>		
<code>a[t4] = t3</code>		<code>a[t2] = t5</code>
<code>goto B<sub>2</sub></code>		<code>a[t4] = t3</code>
		<code>goto B<sub>2</sub></code>

# Global Optimization

- **Code Motion**

- Loops are a very important place for optimizations
  - Especially the inner loops where programs tend to spend the bulk of their time
  - The running time of a program may be improved if we decrease the number of instructions in an inner loop, even if we increase the amount of code outside that loop
- An important modification that decreases the amount of code in a loop is code motion
  - This transformation takes an expression that yields the same result independent of the number of times a loop is executed (a loop-invariant computation) and evaluates the expression before the loop
- **Example:** Evaluation of `limit-2` is a loop-invariant computation



# Global Optimization

- **Induction Variables and Reduction in Strength**

- Another important optimization is to find induction variables in loops and optimize their computation
- A variable  $x$  is said to be an "**induction variable**" if there is a positive or negative constant  $c$  such that each time  $x$  is assigned, its value increases by  $c$
- **Example:**  $i$  and  $t2$  are induction variables in the loop containing  $B_2$

<pre>i = i+1 t2 = 4*i t3 = a[t2] if t3 &lt; v goto B<sub>2</sub></pre>	$B_2$
--	-------

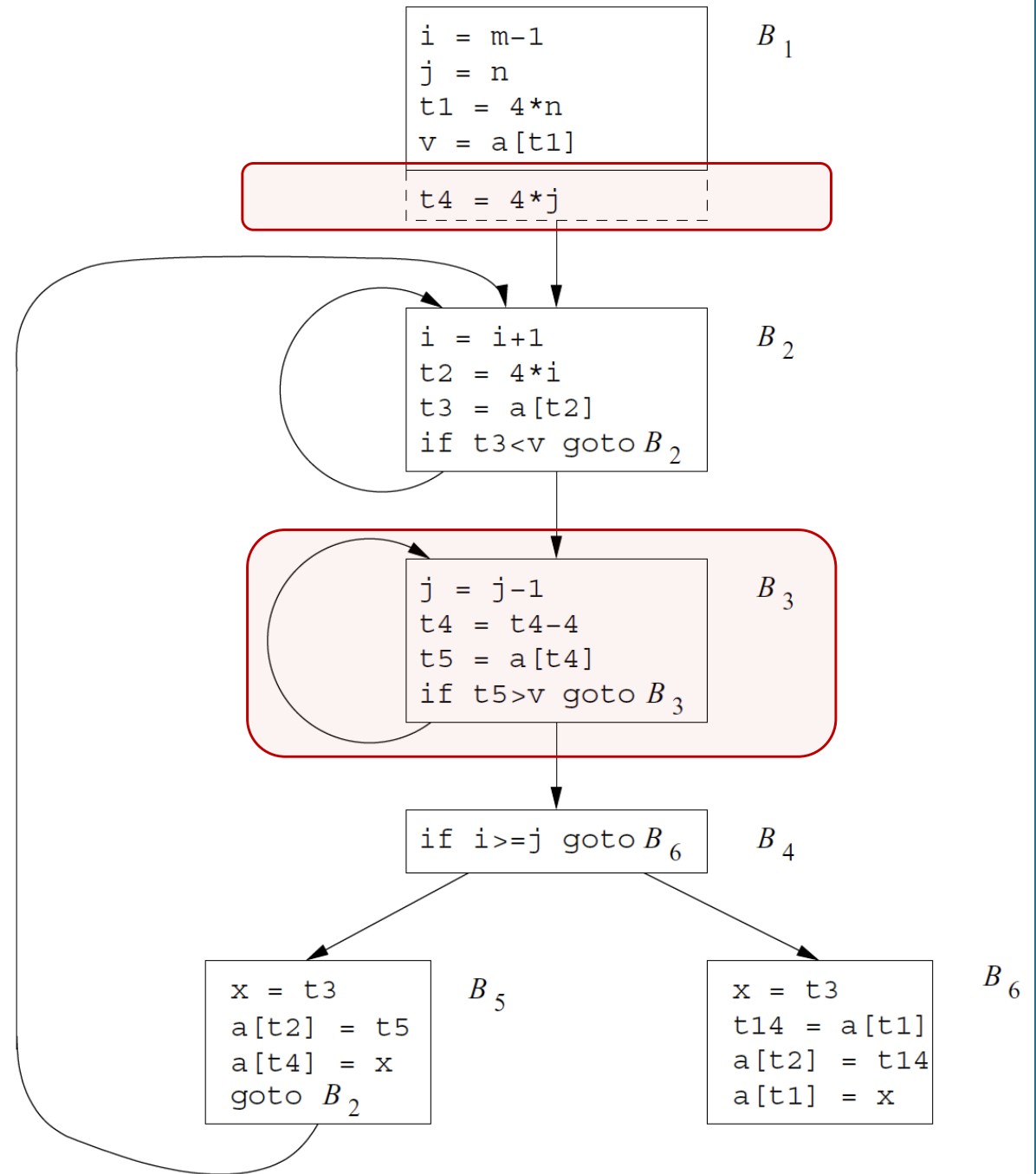
- The transformation of replacing an expensive operation, such as multiplication, by a cheaper one, such as addition, is known as "**strength reduction**"

# Global Optimization

- **Induction Variables and Reduction in Strength**
  - When processing loops, it is useful to work "**inside-out**"; that is, we shall start with the inner loops and proceed to progressively larger, surrounding loops
  - When there are two or more induction variables in a loop, it may be possible to get rid of all but one
  - However, we can illustrate reduction in strength

# Global Optimization

- **Induction Variables and Reduction in Strength**
  - **Example:** Strength reduction applied to  $4*j$  in block  $B_3$



# Global Optimization

- **Induction Variables and Reduction in Strength**

- **Example:** Flow graph after induction-variable elimination

- The test  $t2 \geq t4$  can substitute for  $i \geq j$ .
- Once this replacement is made,  $i$  in block  $B_2$  and  $j$  in block  $B_3$  become dead variables, and the assignments to them in these blocks become dead code that can be eliminated

