Performance: More than Asymptotic Complexity

**Constant factors matter too!**
- You can easily see 10:1 performance range depending on how your code is written.
- Must optimize at multiple levels: algorithm, data representations, procedures, loops.

**Must understand the system to optimize performance.**
- How programs are compiled and executed.
- How to measure program performance and identify bottlenecks.
- How to improve performance without destroying code modularity and generality.

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**Optimizing Compilers**

Provide efficient mapping of program to machine:
- register allocation
- code selection and ordering
- eliminating minor inefficiencies

Don't (usually) improve asymptotic efficiency.
- It's up the programmer to select best overall algorithm.
- Big-O savings are often more important than constant factors.
- But constant factors also matter.

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**Limitations of Optimizing Compilers**

Optimizing compilers have difficulty overcoming "optimization blockers":
- potential memory aliasing
- potential procedure side-effects.

Compilers operate under a fundamental constraint:
- They must not cause any change in program behavior under any possible condition.
- This often prevents making optimizations when they would only affect behavior under pathological conditions.
Limitations of Optimizing Compilers

- Behavior obvious to the programmer may be obfuscated by languages and coding styles.
  - e.g., data ranges may be more limited than the variable type suggests.
- Most analysis is performed only within procedures; whole-program analysis is too expensive in most cases.
- Most analysis is based only on static information.
- When in doubt, the compiler must be conservative.

Machine-Independent Optimizations

Some optimizations you should do regardless of the processor / compiler.

**Code Motion:**
- Reduce frequency with which computation is performed, if it will always produce the same result.
- Move code out of loops if possible.

**The unoptimized version:**
```c
for (i=0; i<n; i++)
  for (j=0; j<n; j++)
    a[n*i + j] = b[j];
```

**The optimized version:**
```c
int ni = n;  
for (i=0; i<n; i++)
  {  
    int *p = a+ni;  
    for (j=0; j<n; j++)
      *p++ = b[j];  
  }
```

Compiler-Generated Code Motion

Most compilers do a good job with array code and simple loop structures.

Code generated by gcc:
```c
testl %edx, %edx
jle .L1
movslq %edx, %r9
xorl %r8d, %r8d
salq $2, %r9
.L3: xorl %eax, %eax
.L5: movl (%rsi,%rax,4), %ecx
movl %ecx, (%rdi,%rax,4)
addq $1, %rax
cmpl %eax, %edx
jg .L5
addl $1, %r8d
addq %r9, %rdi
cmpl %edx, %r8d
jne .L3
.L1: ret
```

Compiler generates the equivalent of:
```c
for (i=0; i<n; i++) {  
  int ni = n*i;  
  for (j=0; j<n; j++)  
    *p++ = b[j];  
}
```

**Reduction in Strength**

Replace costly operations with simpler ones.
- Shift, add instead of multiply or divide: $16 \times x$ becomes $x \ll 4$.
- The utility of this is machine dependent; depends on the cost of multiply and divide instructions.
- On Pentium II or III, integer multiply only requires 4 CPU cycles.

Recognize a sequence of products:
```c
for (i=0; i<n; i++)
  {  
    int ni = 0;  
    for (j=0; j<n; j++)  
      a[ni + j] = b[j];  
    ni += n;  
  }
```
Make Use of Registers

Reading and writing registers is much faster than reading / writing memory.

Limitations:
- Compiler is not always able to determine whether a variable can be held in a register.
- There’s the possibility of aliasing.

Simple Program

```c
int adder( int *p, int *q ) {
    *p = 2;
    *q = 3;
    return (*p + *q);
}
```

What value is returned? Couldn’t we just return 5 and save two memory references?

Not so fast! What if p and q point to the same location (i.e., contain the same address)?

Aliasing means that a location may have multiple names. Often, the compiler must assume that aliasing is possible.

Machine-Independent Optimizations (Continued)

Share Common Subexpressions:
- Reuse portions of expressions.
- Compilers often are not very sophisticated in exploiting arithmetic properties.

```c
/* Sum neighbors of i, j */
up = val[(i-1)*n + j];
down = val[(i+1)*n + j];
left = val[i*n + j-1];
right = val[i*n + j+1];
sum = up + down + left + right;
```

Uses 1 multiplication:

```c
int inj = i*n + j;
up = val[inj - n];
down = val[inj + n];
left = val[inj - 1];
right = val[inj + 1];
sum = up + down + left + right;
```

Uses 3 multiplications:

```c
leal -1(%edx),%ecx
imull %ebx,%ecx
leal 1(%edx),%eax
imull %ebx,%eax
imull %ebx,%edx
```
**Time Scales**

**Absolute time:** Typically uses nanoseconds ($10^{-9}$ seconds).

**Clock cycles:**
- Most computers are controlled by a high frequency clock signal.
- Typical range:
  - Low end: 100 MHz: $10^8$ cycles per second; clock period = 10 ns.
  - High end: 2 GHz: $2 \times 10^9$ cycles per second; clock period = 0.5 ns.

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**Example of Performance Measurement**

**Loop unrolling:** Perform more in each iteration of the loop. (Assume even number of elements.)

Original loop:

```c
void vsum1( int n ) {
    int i;
    for ( i = 0; i < n; i++)
        c[i] = a[i] + b[i];
}
```

Loop unrolled:

```c
void vsum2( int n ) {
    int i;
    for ( i = 0; i < n; i+=2 ) {
        c[i] = a[i] + b[i];
        c[i+1] = a[i+1] + b[i+1];
    }
}
```

Why would this make any difference in performance?

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**Cycles Per Element**

CPE is a convenient way to express performance of a program that operates on vectors or lists.

If the vector length = $n$, then

$$T = \text{CPE} \times n + \text{Overhead}$$

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**Code Motion Example**

**Procedure to convert a string to lower case:**

```c
void lower( char *s )
{
    int i;
    for ( i = 0; i < strlen(s); i++)
        if ( s[i] >= 'A' && s[i] <= 'Z' )
            s[i] -= ('A' - 'a');
}
```

Time quadruples when string length doubles (quadratic performance: $O(n^2)$). Why would that be?
void lower(char *s) {
    int i = 0;
    if (i >= strlen(s))
        goto done;
loop:
    if (s[i] >= 'A' && s[i] <= 'Z')
        s[i] = ('A' - 'a');
    i++;
    if (i < strlen(s))
        goto loop;
done:
}

So what is the issue?

- strlen is executed every iteration.
- strlen is linear in length of the string; must scan string until it finds \'\0\'. Why is that?
- Overall performance is quadratic. What do you do?

Improving Performance

Can move the call to strlen outside of loop, since the result does not change from one iteration to another. This is a form of code motion.

void lower(char *s) {
    int i;
    int len = strlen(s);
    for (i = 0; i < len; i++)
        if (s[i] >= 'A' && s[i] <= 'Z')
            s[i] = ('A' - 'a');
}

Now, the run time doubles when the string length doubles (linear performance: \(O(n)\)).

Can you see other obvious optimizations in this code?

Optimization Blocker: Procedure Calls

Why couldn't the compiler move strlen out of the inner loop?
- Procedures may have side effects. E.g., might alter global state each time called.
- Function may not return the same value for given arguments; might depend on other parts of the global state.
- Procedure lower could interact with strlen.

Why doesn't the compiler just look at the code for strlen?
- The linker might overload with a different version (unless it's declared static.
- Inter-procedural optimization is rare because of the cost.

Warning:
- The compiler treats a procedure call as a black box.
- It applies weak optimizations in and around procedures.
Create a vector abstract data type similar to array implementations in Pascal, ML, Java. E.g., always do bounds checking.

**Procedures:**

- `vec_ptr new_vec( int len )`
  Create vector of specified length

- `int get_vec_element( vec_ptr v, int index, int *dest )`
  Retrieve vector element, store at *dest
  Return 0 if out of bounds, 1 if successful

- `int *get_vec_start( vec_ptr v )`
  Return pointer to start of vector data

**Reduction in Strength**

- `void combine2( vec_ptr v, int *dest )`

  ```c
  { 
  int i;
  int length = vec_length(v);
  int *data = get_vec_start(v);
  *dest = 0;
  for( i = 0; i < length; i++ )
    *dest += data[i];
  }
  ```

**Optimization**

- Avoid procedure call to retrieve each vector element.
- Get pointer to start of array before loop.
- Within the loop just do pointer reference.
- Not as clean in terms of data abstraction.
- CPE: 6.00 (compiled -O2)
- Procedure calls are expensive!
- Bounds checking is expensive!

**Eliminate Unneeded Memory Refs**

- `void combine3( vec_ptr v, int *dest )`

  ```c
  { 
  int i;
  int length = vec_length(v);
  int *data = get_vec_start(v);
  int sum = 0;
  for( i = 0; i < length; i++ )
    sum += data[i];
  *dest = sum;
  }
  ```

**Optimization**

- Don’t need to store result in destination until the end.
- Local variable sum will be held in a register.
- Avoids 1 memory read and 1 memory write per cycle.
- CPE: 2.00 (compiled -O2)
- Memory references are expensive!
Detecting Unneeded Memory Refs

Combine2

.L18:
    movl (%ecx,%edx,4),%eax
    addl %eax,(%edi)
    incl %edx
    cmpl %esi,%edx
    jle .L18

Combine3

.L24:
    addl (%eax,%edx,4),%ecx
    incl %edx
    cmpl %esi,%edx
    jle .L24

Performance:
- Combine2: 5 instructions in 6 clock cycles; addl must read and write memory.
- Combine3: 4 instructions in 2 clock cycles.

Aliasing: two different memory references specify a single location.

Example:
- let v: [3, 2, 17]
  - combine2( v, get_vec_start(v)+2 ) → ?
  - combine3( v, get_vec_start(v)+2 ) → ?

Observations:
- This can easily occur in C, since you’re allowed to do address arithmetic.
- You have direct access to storage structures.
- Get into the habit of introducing local variables and accumulating within loops.
- This is your way of telling the compiler not to check for potential aliasing.

Previous Best Combining Code

```c
void combine3( vec_ptr v, int *dest )
{
    int i;
    int length = vec_length(v);
    int *data = get_vec_start(v);
    int sum = 0;
    for( i = 0; i < length; i++ )
        sum += data[i];
    *dest = sum;
}
```

Task:
- Compute sum of all elements in vector.
- Vector is represented by C-style abstract data type.
- Achieved cycles per element (CPE) of 2.00.

Previous Best Combining Code

```c
void abstract_combine3( vec_ptr v, data_t *dest )
{
    int i;
    int length = vec_length(v);
    data_t *data = get_vec_start(v);
    data_t t = IDENT;
    for( i = 0; i < length; i++ )
        t = t OP data[i];
    *dest = t;
}
```

Data Types: Use different declarations for data_t (int, float, double, etc.)

Operations: Use different definitions of OP and IDENT (+/0, */1, etc.)
### Machine Independent Optimization Results

<table>
<thead>
<tr>
<th>Method</th>
<th>Integer</th>
<th>Floating Point</th>
</tr>
</thead>
<tbody>
<tr>
<td>abstract -g</td>
<td>42.06</td>
<td>41.86</td>
</tr>
<tr>
<td>abstract -O2</td>
<td>31.25</td>
<td>33.25</td>
</tr>
<tr>
<td>move vec_length</td>
<td>20.66</td>
<td>21.25</td>
</tr>
<tr>
<td>data access</td>
<td>6.00</td>
<td>9.00</td>
</tr>
<tr>
<td>accum in temp</td>
<td>2.00</td>
<td>4.00</td>
</tr>
</tbody>
</table>

Optimizations: reduce function calls and memory references within loop.

Performance anomaly:
- Computing FP product of all elements exceptionally slow.
- Very large speedup when accumulate in temporary.
  - Caused by quirk in IA32 floating point.
  - Memory uses 64-bit format; register uses 80-bit format.
  - Benchmark data caused overflow in 64 bits, but not in 80 bits.

### Pointer Code

```c
void combine3p( vec_ptr v, int *dest )
{
    int length = vec_length(v);
    int *data = get_vec_start(v);
    int *dend = data + length;
    int sum = 0;
    while (data < dend) {
        sum += *data;
        data++;
    }
    *dest = sum;
}
```

### Optimization I

**Optimization:**
- Use pointers rather than array references.
- CPE: 3.00 (compiled -O2) – Oops! We’re making reverse progress.

Warning: Some compilers do a better job of optimizing array code.

### Pointer vs. Array Code Inner Loops

**Array Code:**
```
.L24:
    addl (%eax,%edx,4), %ecx # Loop
    incl %edx # sum += data[i]
    cmpl %esi,%edx # i++
    jl .L24 # i:length
```

**Pointer Code:**
```
.L30:
    addl (%eax), %ecx # Loop
    addl $4,%eax # sum += *data[i]
    addl %edx,%eax # data++
    cmpl %edx,%eax # data:dend
    jl .L30 # if < goto Loop
```

Performance:
- Array code: 4 instructions in 2 clock cycles
- Pointer code: almost same 4 instructions in 3 clock cycles

### Machine-Independent Optimization Summary

**Code Motion**
- Compilers are good at this for simple loop/array structures
- They don’t do well in the presence of procedure calls and potential memory aliasing.

**Reduction in Strength**
- Shift, add instead of multiply, divide
  - Compilers (generally) good at this.
- The exact trade off is machine-dependent.

**Keep data in registers rather than memory.**
- Compilers are not good at this, since they are concerned with potential aliasing.

**Share Common Subexpressions**
- Compilers have limited algebraic reasoning capabilities.
Measurement

- Accurately compute time taken by code.
  - Most modern machines have built-in cycle counters.
  - Using them to get reliable measurements is tricky.
- Profile procedure calling frequencies (Unix tool *gprof*).

Observation: Generating assembly code:

- lets you see what optimizations the compiler can make;
- allows you to understand the capabilities / limitations of a particular compiler.

Code Profiling Example

**Task**

- Count word frequencies in a text document.
- Produce sorted list of words from most frequent to least.

**Steps**

- Convert strings to lowercase.
- Apply hash function.
- Read words and insert into hash table:
  - Mostly list operations.
  - Maintain counter for each unique word
- Sort the results.

**Data Set**

- Collected works of Shakespeare.
- 946,596 total words; 26,596 unique words.
- Initial implementation: 9.2 seconds.

**Profiling Results**

<table>
<thead>
<tr>
<th>% time</th>
<th>cumulative</th>
<th>self</th>
<th>calls</th>
<th>self</th>
<th>total</th>
<th>name</th>
</tr>
</thead>
<tbody>
<tr>
<td>86.60</td>
<td>8.21</td>
<td>8.21</td>
<td>1</td>
<td>8210.00</td>
<td>8210.00</td>
<td>sort_words</td>
</tr>
<tr>
<td>5.80</td>
<td>8.76</td>
<td>0.55</td>
<td>946596</td>
<td>0.00</td>
<td>0.00</td>
<td>lower1</td>
</tr>
<tr>
<td>4.75</td>
<td>9.21</td>
<td>0.45</td>
<td>946596</td>
<td>0.00</td>
<td>0.00</td>
<td>fine_ele_rec</td>
</tr>
<tr>
<td>1.27</td>
<td>9.33</td>
<td>0.12</td>
<td>946596</td>
<td>0.00</td>
<td>0.00</td>
<td>h_add</td>
</tr>
</tbody>
</table>

**Call Statistics:** Number of calls and cumulative time for each function.

**Performance Limiter:**

- Using inefficient sorting algorithm.
- Single call uses 87% of CPU time.

The first obvious step in optimization is to use a more efficient sorting algorithm. Replacing the initial slow sort with the library function `qsort` (QuickSort), brought the time down from 9 seconds to around 1 second!
Further Optimizations

- Iter first: use iterative function to insert elements into the linked list; actually causes code to slow down.
- Iter last: iterative function that places new entries at end of the list rather than front; tends to place common words near the front of the list.
- Big table: increase the number of hash functions.
- Better hash: use a more sophisticated hash function.
- Linear lower: move strlen out of the loop.

By applying these optimizations successively and profiling the result, the overall runtime was reduced to around 0.5 seconds.

Profiling Observations

**Benefits**
- Helps identify performance bottlenecks.
- Especially useful for complex systems with many components.

**Limitations**
- Only shows performance for the data tested.
- E.g., linear lower did not show a big gain, since words are short.
  - Quadratic inefficiency could remain lurking in the code.
- The timing mechanism is fairly crude; it only works for programs that run for > 3 seconds.

Role of the Programmer

*How should I write my programs, given that I have a good optimizing compiler?*

- **Don’t**: Smash code into oblivion.
  - Becomes hard to read, maintain, and assure correctness.
- **Do**:
  - Select the best algorithm.
  - Write code that’s readable and maintainable.
    - Use procedures and recursion and eliminate built-in limits.
    - Even though these factors can slow down code.
  - Eliminate optimization blockers to allow the compiler to do its job.
  - Focus on inner loops.
    - Do detailed optimizations where code will be executed repeatedly.
    - You’ll get the most performance gain here.

Summary

- Optimization blocker: procedure calls
- Optimization blocker: memory aliasing
- Tools (profiling) for understanding performance