# Modern C++ Programming

# 16. CODE OPTIMIZATION I

Federico Busato

University of Verona, Dept. of Computer Science 2020, v3.0



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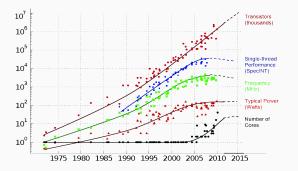
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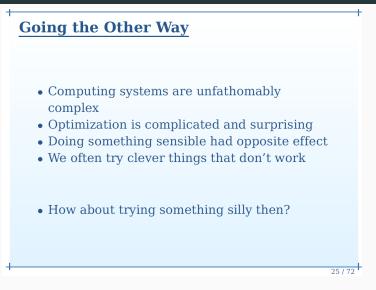
C++ Objects Optimizations

#### Preface

#### Reasons for optimizing:

- In the first decades, the *computer performance was extremely limited*. Low-level optimizations were required to fully exploit the hardware
- Modern systems provide much higher performance, but we cannot more rely on hardware improvement on short-period





from "Speed is Found in the Minds of People", Andrei Alexandrescu, CppCon 2019

- Optimized C++, Kurt Guntheroth
- Awesome C/C++ performance optimization resources, Bartlomiej Filipek
- Optimizing C++, wikibook
- Optimizing software in C++, Agner Fog
- Hacker Delight (2nd), Henry S. Warren

# **General Concepts**

The **asymptotic analysis** refers to estimate the execution time or memory usage as function of the input size (the *order of growing*)

The *asymptotic behavior* is opposed to a *low-level analysis* of the code (instruction/loop counting/weighting, cache accesses, etc.)

#### Drawbacks:

- The worst-case is not the average-case
- Asymptotic complexity does not consider small inputs (think to *insertion sort*)
- The hidden constant can be relevant in practice
- Asymptotic complexity does not consider instructions cost and hardware details

One example out of them all is the *Strassen*'s matrix multiplication algorithm... but arXiv:1808.07984: *Implementing Strassen's Algorithm with CUTLASS on NVIDIA Volta GPUs, J. Huang et. al* 8/88 Be aware that only **real-world problems** with a small asymptotic complexity or small size can be solved in a *"user" acceptable time* 

Two examples:

- Sorting: O(n log n), try to sort an array of one billion elements (4GB)
- Diameter of a (sparse) graph: O (V<sup>2</sup>), just for graphs with a few hundred thousand vertices it becomes impractical without advanced techniques

*"If you're not writing a program, don't use a programming language"* 

Leslie Lamport, Turing Award

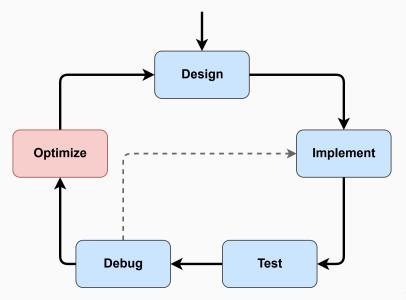
"Inside every large program is an algorithm trying to get out" **Tony Hoare**, Turing Award

"Premature optimization is the root of all evil" **Donald Knuth**, Turing Award

"Code for correctness first, then optimize!"

"First solve the problem, then write the code"

#### **Optimization Cycle**



- One of the most important phase of the optimization cycle is the application profiling for finding regions of code that are *critical for performance* (hotspot)
  - Expensive code region (absolute)
  - Code regions executed many times (cumulative)
- Most of the times, there is no the perfect algorithm for all cases (e.g. insertion, merge, radix sort). Optimizing refers also in finding the correct heuristics for different program inputs instead of modifying the existing code

#### Ahmdal Law

The **Ahmdal law** expresses the maximum improvement possible by improving a particular part of a system

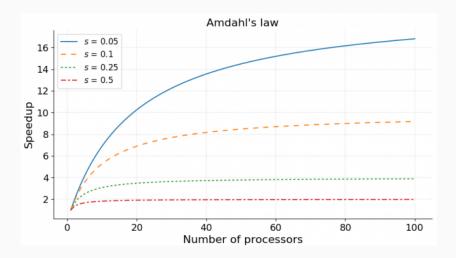
*Observation:* The performance of any system is constrained by the speed or capacity of the slowest point

$$Improvement(S) = \frac{1}{(1-P) + \frac{P}{S}}$$

- ${\it P}$  : portion of the system that can be improved
- S : improvement factor

$$1 - P \quad \boxed{\frac{P}{s}} \quad P$$

Ahmdal Law



note:  $\boldsymbol{s}$  is the portion of the system that cannot be improved

The performance of a program is *bounded* by one or more aspects of its computation. This is also strictly related to the underlying hardware

- Memory-bound. The program spends its time primarily in performing *memory accesses*. The progress is limited by the *memory bandwidth* (sometime memory-bound also refers to the amount of memory available)
- **Compute-bound**. The program spends its time primarily in computing *arithmetic instructions*. The progress is limited by the *speed of the CPU*

- Latency-bound. The program spends its time primarily in waiting the data are ready (instruction/memory dependencies). The progress is limited by the latency of the CPU/memory
- I/O Bound. The program spends its time primarily in performing I/O operations (network, user input, storage, etc.). The progress is limited by the speed of the I/O subsystem

#### **Arithmetic Intensity**

#### **Arithmetic Intensity**

Arithmetic/Operational intensity is the ratio of total operations to total data movement (bytes)

The naive matrix multiplication algorithm requires  $n^3 \cdot 2$ floating-point operations (multiplication + addition), while it involves  $(n^2 \cdot 4B) \cdot 3$  data movement

$$R = \frac{ops}{bytes} = \frac{2n^3}{12n^2} = \frac{n}{6}$$

which means that for every byte accessed, the algorithm performs  $\frac{n}{6}$  operations  $\rightarrow$  compute-bound

• *Example:* 
$$N = 10240, R = \frac{210 GFlops}{1.2GB} \approx 1706$$

A modern CPU performs 100 GFlops, and has about 50 GB/s memory bandwidth

Modern processor architectures are deeply pipelined Instruction-level parallelism (ILP) is a measure of how many instructions in a computer program can be executed simultaneously by issuing independent instructions in sequence (*out-of-order*)

**Instruction pipelining** is a technique for implementing ILP within a single processor

for (int i = 0; i < N; i++) // with no optimizations the loop
 sum += A[i] \* B[i]; // is exectued in sequence</pre>

can be rewritten as:

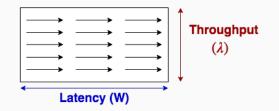
for (int i = 0; i < N; i += 4) { // here, there are
 sum += A[i] \* B[i]; // four independent
 sum += A[i + 1] \* B[i + 1]; // multiplications
 sum += A[i + 2] \* B[i + 2]; // per iteration
 sum += A[i + 3] \* B[i + 3];
}</pre>

#### ILP and Little's Law

The **Little's Law** expresses the relation between *latency* and *throughput*. The throughput of a system  $\lambda$  is equal to the number of elements in the system divided by the average time spent W for each elements in the system:

$$L = \lambda W \quad \rightarrow \quad \lambda = \frac{L}{W}$$

- L: average number of customers in a store
- λ: arrival rate (throughput)
- W: average time spent (*latency*)



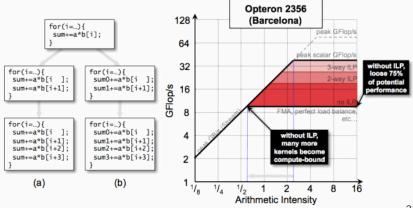
The **time-memory trade-off** is a way of solving a problem or calculation in less time by using more storage space (less often the opposite direction)

Examples:

- *Memoization* (e.g. used in dynamic programming): returning the cached result when the same inputs occur again
- Hash table: number of entries vs. efficiency
- Lookup tables: precomputed data instead branches
- Uncompressed data: bitmap image vs. jpeg

#### **Roofline Model**

The **Roofline model** is a visual performance model used to provide performance estimates of a given application by showing hardware limitations, and potential benefit and priority of optimizations



# I/O Operations

# I/O Operations are orders of magnitude slower than memory accesses

In general, Input/Output are one of the most expensive operations

- Use endl for ostream only when it is strictly necessary (prefer \n)
- Disable synchronization with printf/scanf: std::ios\_base::sync\_with\_stdio(false)
- Disable IO *flushing* when mixing istream/ostream calls:
   <istream\_obj>.tie(nullptr);
- Increase IO *buffer size*:

file.rdbuf()->pubsetbuf(buffer\_var, buffer\_size);

```
#include <iostream>
int main() {
   std::ifstream fin;
   // -----
   std::ios_base::sync_with_stdio(false); // sync disable
   fin.tie(nullptr);
                                       // flush disable
                                       // buffer increase
   const int BUFFER_SIZE = 1024 * 1024; // 1 MB
   char buffer[BUFFER SIZE];
   fin.rdbuf()->pubsetbuf(buffer, BUFFER_SIZE);
   // ____
   fin.open(filename); // Note: open() after optimizations
```

#### // IO operations

```
fin.close();
```

#### printf

- printf is faster than ostream (see speed test link)
- A printf call with the format string %s\n is converted to a puts() call printf("%s\n", string);
- A printf call with a simple format string ending with \n is converted to a puts() call printf("Hello World\n");
- No optimization if the string is not ending with  $\n$
- No optimization if one or more % are detected in the format string

Reference: www.ciselant.de/projects/gcc\_printf/gcc\_printf.html 25/88

A **memory-mapped file** is a segment of virtual memory that has been assigned a direct byte-for-byte correlation with some portion of a file

#### **Benefits:**

- Orders of magnitude faster than system calls
- Input can be "cached" in RAM memory (page/file cache)
- A file requires disk access only when a new page boundary is crossed
- Memory-mapping may bypass the page file completely
- Load and store raw data (no parsing/conversion)

#### Memory Mapped I/O (Example 1/2)

```
#if !defined( linux )
   #error It works only on linux
#endif
#include <fcntl.h> //::open
#include <sys/mman.h> //::mmap
#include <sys/stat.h> //::open
#include <sys/types.h> //::open
#include <unistd.h> //::lseek
// usage: ./exec <file> <byte_size> <mode>
int main(int argc, char* argv[]) {
  size_t file_size = std::stoll(argv[2]);
  auto is_read = std::string(argv[3]) == "READ";
  int fd = is_read ? ::open(argv[1], O_RDONLY) :
                     ::open(argv[1], O_RDWR | O_CREAT | O_TRUNC,
                            S IRUSR | S IWUSR);
  if (fd == -1)
      ERROR("::open")
                                      // try to get the last byte
  if (::lseek(fd, static_cast<off_t>(file_size - 1), SEEK_SET) == -1)
      ERROR("::lseek")
  if (!is_read && ::write(fd, "", 1) != 1) // try to write
                                                                    27/88
      ERROR("::write")
```

#### Memory Mapped I/O (Example 2/2)

```
auto mm_mode = (is_read) ? PROT_READ : PROT_WRITE;
```

```
// Open Memory Mapped file
auto mmap_ptr = static_cast<char*>(
                                ::mmap(nullptr, file_size, mm_mode, MAP_SHARED, fd, 0) );
```

```
if (mmap_ptr == MAP_FAILED)
    ERROR("::mmap");
// Advise sequential access
```

```
if (::madvise(mmap_ptr, file_size, MADV_SEQUENTIAL) == -1)
    ERROR("::madvise");
```

// MemoryMapped Operations
// read from/write to "mmap\_ptr" as a normal array: mmap\_ptr[i]

```
// Close Memory Mapped file
if (::munmap(mmap_ptr, file_size) == -1)
    ERROR("::munmap");
if (::close(fd) == -1)
    The file of the set of the set
```

```
ERROR("::close");
```

}

#### Consider using optimized (low-level) numeric conversion routines:

```
template<int N, unsigned MUL, int INDEX = 0>
struct fastStringToIntStr;
inline unsigned fastStringToUnsigned(const char* str, int length) {
    switch(length) {
        case 10: return fastStringToIntStr<10, 100000000>::aux(str);
             9: return fastStringToIntStr< 9, 100000000>::aux(str);
        case
        case 8: return fastStringToIntStr< 8, 10000000>::aux(str);
        case 7: return fastStringToIntStr< 7, 1000000>::aux(str);
        case 6: return fastStringToIntStr< 6, 100000>::aux(str);
        case 5: return fastStringToIntStr< 5, 10000>::aux(str);
        case 4: return fastStringToIntStr< 4, 1000>::aux(str);
        case 3: return fastStringToIntStr< 3, 100>::aux(str);
        case 2: return fastStringToIntStr< 2, 10>::aux(str);
        case 1: return fastStringToIntStr< 1, 1>::aux(str);
        default: return 0:
    }
```

```
template<int N, unsigned MUL, int INDEX>
struct fastStringToIntStr {
    static inline unsigned aux(const char* str) {
        return static_cast<unsigned>(str[INDEX] - '0') * MUL +
               fastStringToIntStr<N - 1, MUL / 10, INDEX + 1>::aux(str);
    }
};
template<unsigned MUL, int INDEX>
struct fastStringToIntStr<1, MUL, INDEX> {
    static inline unsigned aux(const char* str) {
        return static_cast<unsigned>(str[INDEX] - '0');
    }
};
```

Faster parsing: lemire.me/blog/tag/simd-swar-parsing

#### Speed Up Raw Data Loading

- Hard disk is orders of magnitude slower than RAM
- Parsing is faster than hard disk data reading
- Parsing can be avoided by using *binary* storage and mmap
- Decreasing the number of hard disk accesses improves the performance  $\rightarrow$  compression

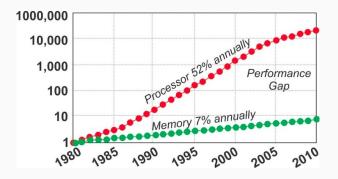
**LZ4** is lossless compression algorithm providing *extremely fast decompression* up to 35% of memcpy and good compression ratio

github.com/lz4/lz4

# Locality and Memory Access Patterns

# Access to memory dominates other costs in a processor

#### The Memory Wall:



# Memory Hierarchy

Modern architectures rely on complex memory hierarchy (primary memory, caches, registers, scratchpad memory, etc.). Each level has different characteristics and constrains (size, latency, bandwidth, concurrent accesses, etc.)



1 byte of RAM (1946)



IBM 5MB hard drive (1956)

Memory hierarchies		Intel Haswell E5-2650 v3	Intel KNL 7250 DDR5   MCDRAM	ARM Cortex A57
		10 cores 368 Gflop/s 105 Watts	68 cores 2662 Gflop/s 215 Watts	4 cores 32 Gflop/s 7 Watts
	REGISTERS		32/core AVX-512	32/core
	L1 CACHE & GPU SHARED MEMORY	32 KB/core	32 KB/core	32 KB/core
	L2 CACHE L3 CACHE		1024 KB/2cores	2 MB
			016 GB	N/A
	MAIN MEMORY	64 GB	384  16 GB	4 GB
		68 GB/s	115   421 GB/s	26 GB/s
	MAIN MEMORY BW	5.4 flops/byte	23   6 Flops/byte	1.2 flops/byte
	MAIN MEMORY BW PCI EXPRESS GEN3x16 NVLINK			
	PCI EXPRESS GEN3x16	5.4 flops/byte 16 GB/s	23   6 Flops/byte 16 GB/s	1.2 flops/byte 16 GB/s

"Accelerating Linear Algebra on Small Matrices from Batched BLAS to Large Scale Solvers", ICL, University of Tennessee

#### Intel Coffee Lake Core-i7-8700 example:

Cache level	Size	Latency	Bandwidth
L1 cache	192 KB	$\sim 1.5~{ m ns}$	$\sim$ 1,600 GB/s
L2 cache	1.5 MB	$\sim$ 4 ns	$\sim$ 570 GB/s
L3 cache	12 MB	$\sim$ 12 - 40 ns	$\sim$ 320 GB/s
DRAM	/	$\sim$ 60 ns	$\sim$ 40 GB/s

en.wikichip.org/wiki/WikiChip

www.forrestthewoods.com/blog/memory-bandwidth-napkin-math/ 35/88

- Spatial Locality refers to the use of data elements within <u>relatively close storage locations</u> e.g. scan arrays in increasing order, matrices by row. It involves mechanisms such as <u>memory prefetching</u> and <u>access granularity</u>
- Temporal Locality refers to the reuse of specific data within a relatively <u>small time duration</u>, and, as consequence, exploit lower levels of the memory hierarchy (caches)
   Heavily used memory locations can be accessed more quickly.

Heavily used memory locations can be accessed more quickly than less heavily used locations

# Spatial Locality Example

```
A, B, C matrices of size N × N
C = A * B
for (int i = 0; i < N; i++) {
   for (int j = 0; j < N; j++) {
      int sum = 0;
      for (int k = 0; k < N; k++)
           sum += A[i][k] * B[k][j]; // row × column
      C[i][j] = sum;
   }
}</pre>
```

```
C = A * B<sup>T</sup>
for (int i = 0; i < N; i++) {
   for (int j = 0; j < N; j++) {
      int sum = 0;
      for (int k = 0; k < N; k++)
           sum += A[i][k] * B[j][k]; // row × row
      C[i][j] = sum;
   }
}</pre>
```

#### Benchmark:

N	64	128	256	512	1024
A * B	$< 1 \ { m ms}$	5 ms	29 ms	141 ms	1,030 ms
$A * B^T$	$< 1 \ { m ms}$	2 ms	6 ms	48 ms	385 ms
Speedup	/	2.5x	4.8x	2.9×	2.7×

# Cache Optimization Example

#### Speeding up a random-access function

```
for (int i = 0; i < N; i++) { // V2
for (int K = 0; K < N; K += CACHE) {
    auto x = hash(i);
    if (x >= K && x < K + CACHE)
        out_array[i] = in_array[x];
    }
}</pre>
```

V1 : 436 ms, V2 : 336 ms  $\rightarrow$  1.3x speedup .. but it needs a careful evaluation of CACHE and it can even decrease the performance for other sizes

```
pre-sorted hash(i) : 135 ms \rightarrow 3.2x speedup
```

- Dynamic heap allocation is expensive: implementation dependent and interaction with the operating system
- Many small heap allocations are more expensive than one large memory allocation
   The default page size on Linux is 4KB. For smaller/multiple sizes, C++ uses a suballocator
- Allocations within the page size is faster than larger allocations (suballocator)

- *Stack memory is faster than heap memory*. The stack memory provides high locality
- **static** stack allocations produces better code. It avoids filling the stack each time the function is reached

```
// constexpr int array[] = {1,2,3,4,5,6,7,8,9};
// faster in general
static constexpr int array[] = {1,2,3,4,5,6,7,8,9};
for (const auto& x : array)
    printf("%d", x);
```

#### Maximize cache utilization:

- Prefer small data types
- Prefer std::vector<bool> over array of bool
- Prefer std::bitset<N> over std::vector<bool> if the data size is known in advance or bounded

note: modern processors have several MBs of (L1) cache

# **Internal Structure Alignment**

struct A1 {	<pre>struct A2 { // internal alignment</pre>
char x1; // offset 0	char x1; // offset 0
<pre>double y1; // offset 8!! (not 1)</pre>	char x2; // offset 1
char x2; // offset 16	char x3; // offset 2
double y2; // offset 24	char x4; // offset 3
char x3; // offset 32	char x5; // offset 4
double y3; // offset 40	<pre>double y1; // offset 8</pre>
char x4; // offset 48	<pre>double y2; // offset 16</pre>
double y4; // offset 56	double y3; // offset 24
<b>char</b> x5; // offset 64 (byte 65)	<pre>double y4; // offset 32 (byte 40)</pre>
}	}

Considering an array of structures, there are two problems:

- We are wasting 40% of memory in the first case (A1)
- In common x64 processors the cache line is 64 bytes. For the first structure A1, every access involves two cache line operations (2x slower)

see also #pragma pack(1)

Considering the previous example for the structure A2, random loads from an array of structure A2 leads to one or two cache line operations depending on the alignment at a specific index, e.g.

index 0  $\rightarrow$  one cache line load

index 1 ightarrow two cache line loads

It is possible to fix the structure alignment in two ways:

- The **memory padding** refers to introduce extra bytes at the end of the data structure to enforce the memory alignment e.g. add a char array of size 24 to the structure A2. It can be also extended to 2D (or *N*-D) data structures such as dense matrices
- Align keyword or attribute allows specifying the alignment requirement of a type or an object (next slide)

 $C{++}$  allows specifying the alignment requirement in three ways:

- C++11 alignas(N) only for variable / struct declaration
- C++17 aligned new (e.g. new int[2, N])
- Compiler Intrinsic only for variables / struct declaration
  - GCC/Clang: \_\_attribute\_\_((aligned(N)))
  - MSVC: \_\_declspec(align(N))
- Compiler Intrinsic for dynamic pointer
  - GCC/Clang: \_\_builtin\_assume\_aligned(x)
  - Intel: \_\_assume\_aligned(x)

Data alignment is essential for exploiting hardware vector instructions (SIMD) like SSE, AVX, etc.

```
struct alignas(16) A1 { // C++11
    int x, y;
};
struct __attribute__((aligned(16))) A2 { // require compiler
    int x, y;
                                         // support
};
auto ptr1 = new int[100, 16]; // 16B alignment
auto ptr2 = new int[100]; // 4B alignment guarantee
auto ptr3 = __builtin_assume_aligned(ptr2, 16);
// require compiler support
```

# Arithmetic

### Hardware Notes

- Instruction throughput greatly depends on processor model and characteristics
- Addition, subtraction, and bitwise operations are computed by the ALU and they have very similar throughput
- In modern processors, multiplication and addition are computed by the same hardware component for decreasing circuit area → multiplication and addition can be fused in a single operation fma (floating-point) and mad (integer)
- Modern processors provide separated units for floating-point computation (FPU)

# Data Types

- 32-bit integral vs. floating-point: in general, integral types are faster, but it depends on the processor characteristics
- 32-bit types are faster than 64-bit types
  - 64-bit integral types are slightly slower than 32-bit integral types. Modern processors widely support native 64-bit instructions for most operations, otherwise they require multiple operations
  - Single precision floating-points are up to three times faster than double precision floating-points
- Small integral types are slower than 32-bit integer, but they require less memory → cache/memory efficiency

- In modern architectures, arithmetic increment/decrement
   ++ / -- has the same performance of add / sub
- Prefer prefix operator (++var) instead of the postfix operator (var++) \*
- Use the assignment composite operators (a += b) instead of operators combined with assignment (a = a + b) \*
- Keep near constant values/variables → the compiler can merge their values

<sup>\*</sup> the compiler automatically applies such optimization whenever possible (this is not ensured for object types)

# Power-of-Two Multiplication/Division/Modulo

- Prefer shift for power-of-two multiplications ( a << b ) and divisions ( a >> b ) only for run-time values \*
- Some unsigned operations are faster than signed operations (deal with negative number), e.g. x / 2
- Prefer bitwise AND a % b → a & (b 1) for power-of-two modulo operations only for run-time values \*
- Constant multiplication and division can be heavily optimized by the compiler, even for non-trivial values

<sup>\*</sup> the compiler automatically applies such optimizations if **b** is known at compile-time. Bitwise operations make the code harder to read

From	То	Cost
Signed	Unsigned	no cost, bit representation is the same
Unsigned	Larger Unsigned	no cost, register extended
Signed	Larger Signed	1 clock-cycle, register $+$ sign extended
Integer	Floating-point	4-16 clock-cycles Signed $\rightarrow$ Floating-point is faster than Unsigned $\rightarrow$ Floating-point (except AVX512 instruction set is enabled)
Floating-point	Integer	fast if SSE2, slow otherwise (50-100 clock-cycles)

Reference: Optimizing software in C++, Agner Fog

#### Multiplication is much faster than division\*

```
not optimized:
// "value" is floating-point (dynamic)
for (int i = 0; i < N; i++)
        A[i] = B[i] / value;</pre>
```

\* Multiplying by the inverse is not the same as the division see lemire.me/blog/2019/03/12 CPU processors perform computations with a larger register size than the original data type (e.g. 48-bit for 32-bit floating-point) to achieve this result

Modern processors allow performing a \* b + c in a single operation, called **fused multiply-add** (std::fma in C++11). This implies better performance and accuracy

Compiler behavior:

- GCC 9 and ICC 19 produce a single instruction for std::fma and for a \* b + c with -O3 -march=native
- Clang 9 and MSVC 19.\* produce a single instruction for std::fma but not for a \* b + c

FMA: solve quadratic equation

FMA: extended precision addition and multiplication by  $\mathrm{constant}\,^{53/88}$ 

**Compiler intrinsics** are highly optimized functions directly provided by the compiler instead of external libraries

## Advantages:

- Directly mapped to hardware functionalities if available
- Inline expansion
- Do not inhibit high-level optimizations and they are portable contrary to asm code

### Drawbacks:

- Portability is limited to a specific compiler
- Some intrinsics do not work on all platforms
- The same instricics can be mapped to a non-optimal instruction sequence depending on the compiler

Most compilers provide intrinsics **bit-manipulation functions** for SSE4.2 or ABM (Advanced Bit Manipulation) instruction sets for Intel and AMD processors

GCC examples:

\_\_builtin\_popcount(x) count the number of one bits

\_\_builtin\_clz(x) (count leading zeros) counts the number of zero bits following the most significant one bit

\_\_builtin\_ctz(x) (count trailing zeros) counts the number of zero bits preceding the least significant one bit

Reference: gcc.gnu.org/onlinedocs/gcc/Other-Builtins.html

```
• Compute integer log2
inline unsigned log2(unsigned x) {
    return 31 - __builtin_clz(x);
}
```

• Check if a number is a power of 2

```
inline bool is_power2(unsigned x) {
    return __builtin_popcount(x) == 1;
}
```

Bit search and clear

# 4/5

#### Example of intrinsic portability issue:

\_\_builtin\_popcount() GCC produces \_\_popcountdi2
instruction while Intel Compiler (ICC) produces 13 instructions

mm\_popcnt\_u32 GCC and ICC produce popcnt instruction, but it is available only for processor with support for SSE4.2 instruction set

#### More advanced usage

- Compute CRC: \_mm\_crc32\_u32
- AES cryptography: \_mm256\_aesenclast\_epi128
- Hash function: \_mm\_sha256msg1\_epu32

Reference: software.intel.com/sites/landingpage/IntrinsicsGuide/ 57/88

Using intrinsic instructions are <u>extremely dangerous</u> if the target processor does not natively support such instructions

Example:

"If you run code that uses the intrinsic on hardware that doesn't support the *lzcnt* instruction, the results are unpredictable" - MSVC

on the contrary, GNU and clang \_\_builtin\_\* instructions are always well-defined. The instruction is translated to a non-optimal operation sequence in the worst case

The instruction set support should be checked at *run-time* (e.g. with \_\_cpuid function on MSVC), or, when available, by using compiler-time macro (e.g. \_\_AVX\_\_ )

#### Value in a Range

Checking if a non-negative value x is within a range [A, B] can be optimized if B > A (useful when the condition is repeated multiple times)

if  $(x \ge A \&\& x \le B)$ // STEP 1: subtract A if  $(x - A) = A - A \&\& x - A \leq B - A$ // --> if  $(x - A) = 0 \& x - A \le B - A$  // B - A is precomputed // STEP 2 // - convert " $x - A \ge 0$ " --> (unsigned) (x - A) // - ensure that "B - A" is not less than zero if  $((unsigned) (x - A) \le (unsigned) (B - A))$ // works even if A, B are negative  $(B \ge A)$ 59/88 Check if a value is an uppercase letter:

 $\begin{array}{cccc} \texttt{uint8_t } \texttt{x} = & \dots & \texttt{uint8_t } \texttt{x} = & \dots & \\ \texttt{if } (\texttt{x} \geq \texttt{'A'} \&\& \texttt{x} \leq \texttt{'Z'}) & \longrightarrow & \texttt{if } (\texttt{x} - \texttt{'A'} \leq \texttt{'Z'}) & \\ & \dots & & & \dots & \end{array}$ 

A more general case:

 $\begin{array}{ccc} \text{int } x = \dots & \\ \text{if } (x \geq -10 \&\& x \leq 30) & \rightarrow & \\ & &$ 

The compiler applies this optimization only in some cases (tested with GCC/Clang 9-03)

# Lookup Table

**Lookup table** is a *memoization* technique which allows replacing *runtime* computation with precomputed values

Example: a function that computes the logarithm base 10 of a number in the range [1-100]

```
#include <array> // the code requires C++17
#include <cmath>
template<int SIZE, typename Lambda>
constexpr std::array<float, SIZE> build(Lambda lambda) {
    std::array<float, SIZE> array{};
    for (int i = 0; i < SIZE; i++)</pre>
        array[i] = lambda(i);
    return array;
}
float log10(int value) {
    constexpr auto lamba = [](int i) { return std::log10f((float) i); };
    static constexpr auto table = build<100>(lambda);
    return table[value];
                                                                          61/88
}
```

Collection of low-level implementations/optimization of common operations:

Bit Twiddling Hacks

 $graphics.stanford.edu/\sim$ seander/bithacks.html

- The Aggregate Magic Algorithms aggregate.org/MAGIC
- Hackers Delight Book
   www.hackersdelight.org

The same instruction/operation may take different clock-cycles on different architectures/CPU type

- Agner Fog Instruction tables (latencies, throughputs)
   www.agner.org/optimize/instruction\_tables.pdf
- Latency, Throughput, and Port Usage Information uops.info/table.html

# **Control Flow**

# Computation is faster than decision

# **Control Flow**

- Avoid run-time recursion (very expensive). Prefer *iterative* algorithms instead (see next slides)
- Prefer switch statements instead of multiple if . If the compiler does not use a jump-table, the cases are evaluated in order of appearance → the most frequent cases should be placed before
- Prefer square brackets syntax [] over pointer arithmetic operations for array access to facilitate compiler loop optimizations (polyhedral loop transformations)
- Prefer signed integer for loop indexing. The compiler optimizes more aggressively such loops since integer overflow is not defined

**Pipelines** are an essential element in modern processors. Some processors have up to 20 pipeline stages (14/16 typically)

The downside to long pipelines includes the danger of **pipeline stalls** that waste CPU time, and the time it takes to reload the pipeline on **conditional branch** operations (if, while, for)

## Minimize Branch Overhead

- Branch prediction: technique to guess which way a branch takes. It requires hardware support and it is generically based on dynamic history of code executing
- Branch predication: a conditional branch is substituted by a sequence of instructions from both paths of the branch. Only the instructions associated to a *predicate* (boolean value), that represents the direction of the branch, are actually executed

```
int x = (condition) ? A[i] : B[i];
P = (condition) // P: predicate

P x = A[i];

P x = B[i];
```

• **Speculative execution**: execute both sides of the conditional branch to better utilize the computer resources and commit the results associated to the branch taken

# Loop Hoisting

#### Loop hoisting optimization

Wrong:	Correct:
<pre>for (int i = 0; i &lt; 100; i++)     a[i] = x + y;</pre>	<pre>v = x + y for (int i = 0; i &lt; 100; i++) a[i] = v;</pre>

Loop hoisting is also important in the evaluation of loop conditions Wrong: Correct:

// "x" never changes	<pre>int limit = f(x)</pre>		
<pre>for (int i = 0; i &lt; f(x); i++)</pre>	<pre>for (int i = 0; i &lt; limit; i++)</pre>		
a[i] = y;	a[i] = y;		

In the worst case, f(x) is evaluated every iteration (especially when it belongs to another translation unit)

the compiler already applies such optimization  $\underline{when \ it \ is \ safe}$  (it does not change the program semantic)

# Loop Unrolling

**Loop unrolling** (or **unwinding**) is a loop transformation technique which optimizes the code by removing (or reducing) loop iterations

The optimization produces better code at the expense of binary size

Example:

```
for (int i = 0; i < N; i++)
    sum += A[i];</pre>
```

can be rewritten as:

```
for (int i = 0; i < N; i += 8) {
    sum += A[i];
    sum += A[i + 1];
    sum += A[i + 2];
    sum += A[i + 3];
    ...
} // we suppose N is a multiple of 8</pre>
```

### Loop unrolling notes:

- + Improve instruction-level parallelism (ILP)
- + Allow vector (SIMD) instructions
- + Reduce control instructions and branches
  - Increase compile-time/binary size
  - Require more instruction decoding
  - Use more memory and instruction cache

**Unroll directive** The Intel, IBM, and clang compilers (but not GCC) provide the preprocessing directive **#pragma unroll** (to insert above the loop) to force loop unrolling. The compiler already applies the optimization in most cases

# Loop Unswitching and Fusion

### Loop Unswitching

```
for (i = 0; i < N; i++) {
    if (x)
        a[i] = 0;
    else
        b[i] = 0;
}</pre>
```

```
if (x) {
   for (i = 0; i < N; i++)
        a[i] = 0;
}
else {
   for (i = 0; i < N; i++)
        b[i] = 0;
}</pre>
```

#### Loop Fusion (Jamming)

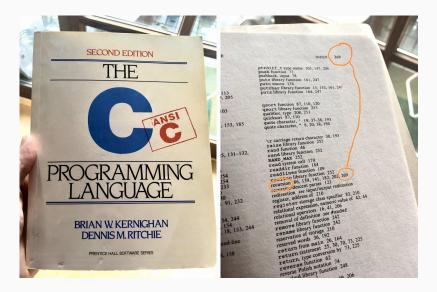
for (i = 0; i < 300; i++)
 a[i] = a[i] + 3;
for (i = 0; i < 300; i++)
 b[i] = b[i] + 4;
}
for (i = 0; i < 300; i++)
 b[i] = b[i] + 4;
}</pre>

Loop unswitching and loop fusion do not produce better code, but loop merging/splitting has implications on cache usage

In many cases, the compiler already applies these optimizations

**Recursion cost:** The program must store all variables (snapshot) at each recursion iteration on the stack, and remove them when the control return to the caller instance

The **tail recursion** optimization avoids to maintain caller stack and pass the control to the next iteration. The optimization is possible only if all computation can be executed before the recursive call



# **Functions**

#### Function call methods:

Direct: Function address is known at compile-timeIndirect: Function address is known only at run-timeInline: The function code is fused in the caller code

#### Function call cost:

- The caller pushes the arguments on the stack in reverse order
- Jump to function address
- The caller clears (pop) the stack

**pass by-value** small data types ( $\leq$  8/16 bytes).

The data are copied into registers, instead of stack

pass by-pointer introduces one level of indirection.
They should be used only for raw pointers
(potentially NULL)

pass by-reference may introduce one level of indirection. pass-by-reference is more efficient than pass-by-pointer as it facilitates variable elimination by the compiler, and the function code does not require checking for NULL pointer

Most compilers optimize **pass by-value** with **pass by-reference** for passive data structures

For *active* objects with <u>non-trivial</u> (and expensive) copy constructor or destructor:

by-value Expensive, hard to optimize by-pointer/reference ok. Prefer pass-by- const -X ( const overloading can also be cheaper)

For *passive* objects with trivial copy constructor *and* destructor:

**by-const-value** Always produce the optimal code (converted in pass-by-ref if needed)

**by-value** Produce optimal code except for GCC (tested with GCC 9.2)

by-reference Could introduce a level of indirection

- Pass by-value built-in types and passive data structured (no side-effect. The compiler already applies heuristics to determine the most efficient way to pass the parameter (by-value or by-reference). Pass by-reference does not allow the compiler to optimize in pass by-value (if not inline)
- Keep small the number of function parameters. The parameters can be passed by using the registers instead filling and emptying the stack
- Consider *combining several function parameters* in a structure

- const modifier applied to pointers and references *does not* produce better code in most cases, but it is useful for ensuring read-only accesses
- const applied to pass by-value does not change the function signature and, for this reason, should be avoided in function declaration

# **Inlining Constrains**

Compilers have different heuristics for function inlining:

- Number of lines (even comments: How new-lines affect the Linux kernel performance)
- Number of assembly instructions
- Inlining depth (recursive)

#### A function is integrated into the caller if:

- It satisfies the compiler heuristics
- The function must have *internal linkage* (static or anonymous namespace)
- The inline keyword helps to decrease the heuristic threshold
- Compiler-specific decorators

( \_\_attribute\_\_((always\_inlinee)) , \_\_forceinline ) allow skipping compiler heuristics

- An Inline Function is As Fast As a Macro
- Inlining Decisions in Visual Studio

All compilers, except MSVC, export all function symbols  $\rightarrow$  slow, the symbols can be used in other translation units

Alternatives:

- Use static functions
- Use anonymous namespace (functions and classes)
- Use GNU extension (also clang)
  \_\_attribute\_\_((visibility("hidden")))

# **Pointers Aliasing**

Consider the following example:

```
// suppose f() is not inline
void f(int* input, int size, int* output) {
   for (int i = 0; i < size; i++)
        output[i] = input[i];
}</pre>
```

- The compiler <u>cannot</u> unroll the loop (sequential execution, no ILP) because output and input pointers can be aliased,
   e.g. output = input + 1
- The aliasing problem is even worse for more complex code and inhibits all kind of optimization from code re-ordering to common sub-expression elimination

Most compilers (included GCC/Clang/MSVC) provide **restricted** pointers (<u>\_\_restrict</u>) so that the programmer asserts that the pointers are not aliased

Potential benefits:

- Instruction-level parallelism
- Less instructions executed
- Merge common sub-expressions

# Benchmarking matrix multiplication

<pre>void matrix_mul_v1(const</pre>	<pre>int*</pre>	A,			
const	<pre>int*</pre>	Β,			
int		N,			
int*		C)	{		

<pre>void matrix_mul_v2</pre>	const	int*	restrict	Α,	
	const	<pre>int*</pre>	restrict	Β,	
	int			N,	
	int*		restrict	C)	{

Optimization	-01	-02	-03
v1	1,030 ms	777 ms	777 ms
v2	513 ms	510 ms	761 ms
Speedup	2.0x	1.5×	1.02×

# C++ Objects

# Variable/Object Scope

#### Declare local variable in the inner most scope

- the compiler will be able to fit them into registers instead stack
- it improves readability

#### Wrong:

#### Correct:

```
int i, x;
for (i = 0; i < N; i++) {
    x = value * 5;
    sum += x;
}

for (int i = 0; i < N; i++) {
    int x = value * 5;
    sum += x;
}
```

**Exception!** Built-in type variables and passive structures should be placed in the inner most loop, while objects with constructors should be placed outside loops

```
for (int i = 0; i < N; i++) {
    std::string str("prefix_");
    std::cout << str + value[i];
    // str call CTOR/DTOR N times
}
std::cout << str + value[i];
}</pre>
```

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- Prefer direct initialization and *full object constructor* instead of two-step initialization (also for variables)
- Prefer move semantic instead of copy constructor. Mark copy constructor as =delete (sometimes it is hard to see, e.g. implicit)
- Mark final all *virtual* functions that are not overridden
- Avoid dynamic operations: exceptions\* (and use noexcept), dynamic\_cast, smart pointer
- Virtual calls are slower than standard functions

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- Use static for all members that do not use instance member (avoid passing this pointer)
- Avoid multiple + operations between objects to avoid temporary storage (need example)
- Prefer ++obj / --obj (return &obj ), instead of obj++ , obj-- (return old obj )
- Prefer x += obj , instead of x = x + obj  $\rightarrow$  avoid the object copy

## **Object Implicit Conversion**

```
#include <algorithm> // std::copy
struct A { // big object
    int array[10000];
};
struct B {
  int array[10000];
  B(const A& a) {
      std::copy(a.array, a.array + 10000, array);
  }
};
//---
void f(const B& b) {}
int main() {
  B b;
  f(b); // no cost
  A a;
  f(a); // very costly
```