A Crash Course on Data Compression

1. Introduction

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@jermp

Overview

- What is Data Compression and why do we need it?
- Fundamental questions and undecidability
- Some applications
- Technological limitations
- Warmup

What is Data Compression?

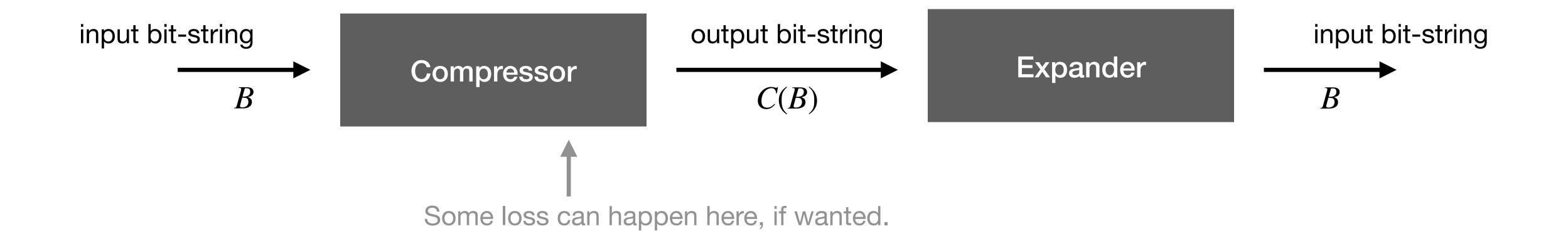
- The process for which data is transformed into another representation that takes *less storage space*:
 - save space when storing data,
 - save time when transmitting data.
- The process must be reversible (exactly or admitting some loss) to be useful.

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- Q. What is this "process"?
 - A. A computer program that takes data as input and produces a data structure that takes less space than the input.
- We seek/need efficient programs that build such data structures.

Example: command line utility gzip.

Basic Model



- Compression ratio. The compression ratio is defined as CR = |B|/|C(B)|.
- If CR = r, then the size of the compressed output |C(B)| is r times smaller than the input size |B|.

Space vs. Time Trade-Off

- The compression ratio depends on many factors: wanted compression/decompression speed, related to the amount of energy spent (CPU power); loss of precision; (...)
- The most common one: the trade-off between the space of the compressed data structure and the efficiency of the operations that we want to support on the data.

Example: gzip has 9 compression "levels" (1 is fastest but "worst" compression; 9 is slower but "best").

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Example: gzip has 9 compression "levels" (1 is fastest but "worst" compression; 9 is slower but "best").

- This trade-off is becoming more and more important: nowadays, we cannot afford the naive approach "decompress and compute".
- Ultimate goal: allow direct computation over compressed data.

Limit

• Proposition. No algorithm can compress every bit-string.

Proof.

Proceed by contradiction. Assume that

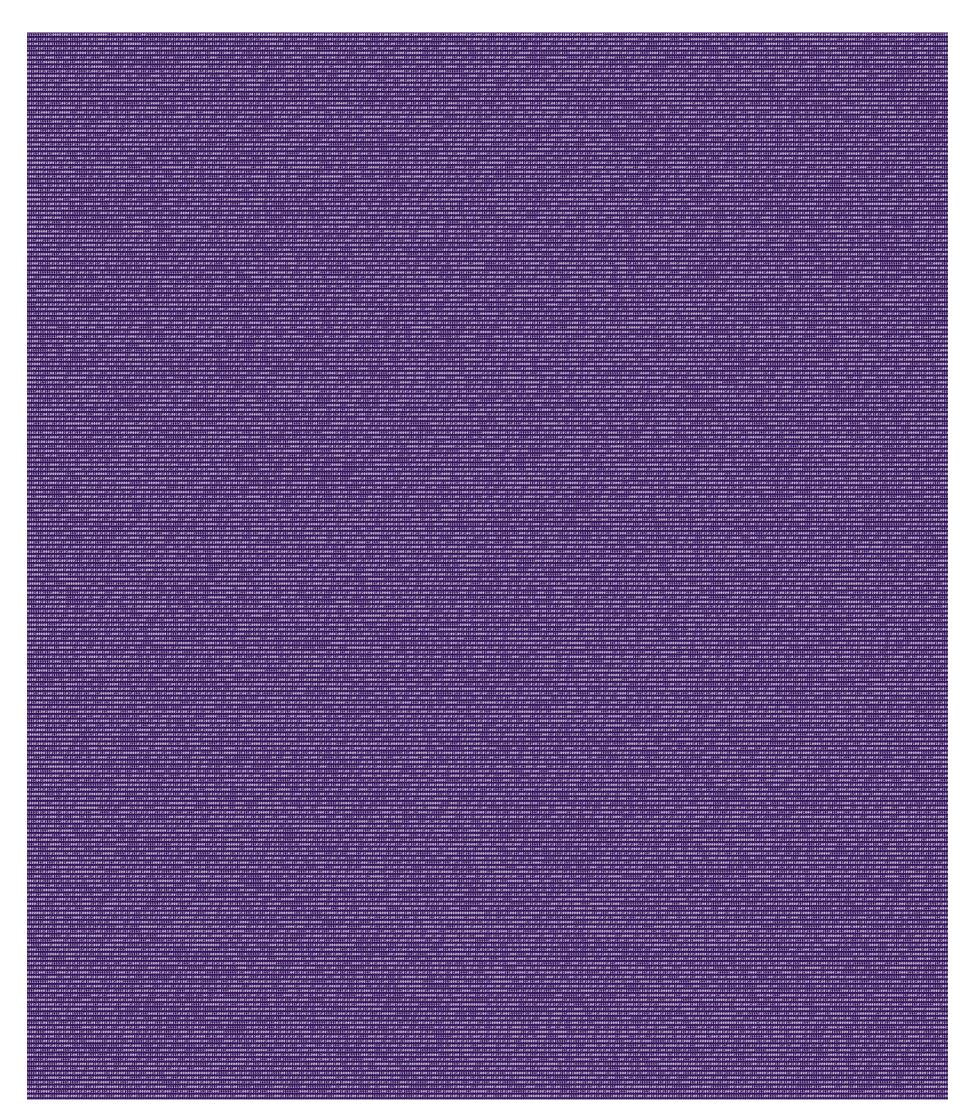
$$|B| > |C(B)| > |C(C(B))| > |C(C(C(B)))| > \dots$$
, like in the picture below.

Then all possible bit-strings could be compressed to 0 bits — absurd. ■

Fundamental Question(s)

- Q. What is the best way of compressing a file for my application?
 A. This is an undecidable problem.
- An extreme, but very common, example: to compress a file, you may replace it by the program that originated the file (related to the so-called Kolmogorov complexity).
 But how would you "find" (i.e., write) such a program?
- If you think: most of the data we deal with (Web pages, log files, sequencing data, ecc.) is created by programs, *not* by humans.

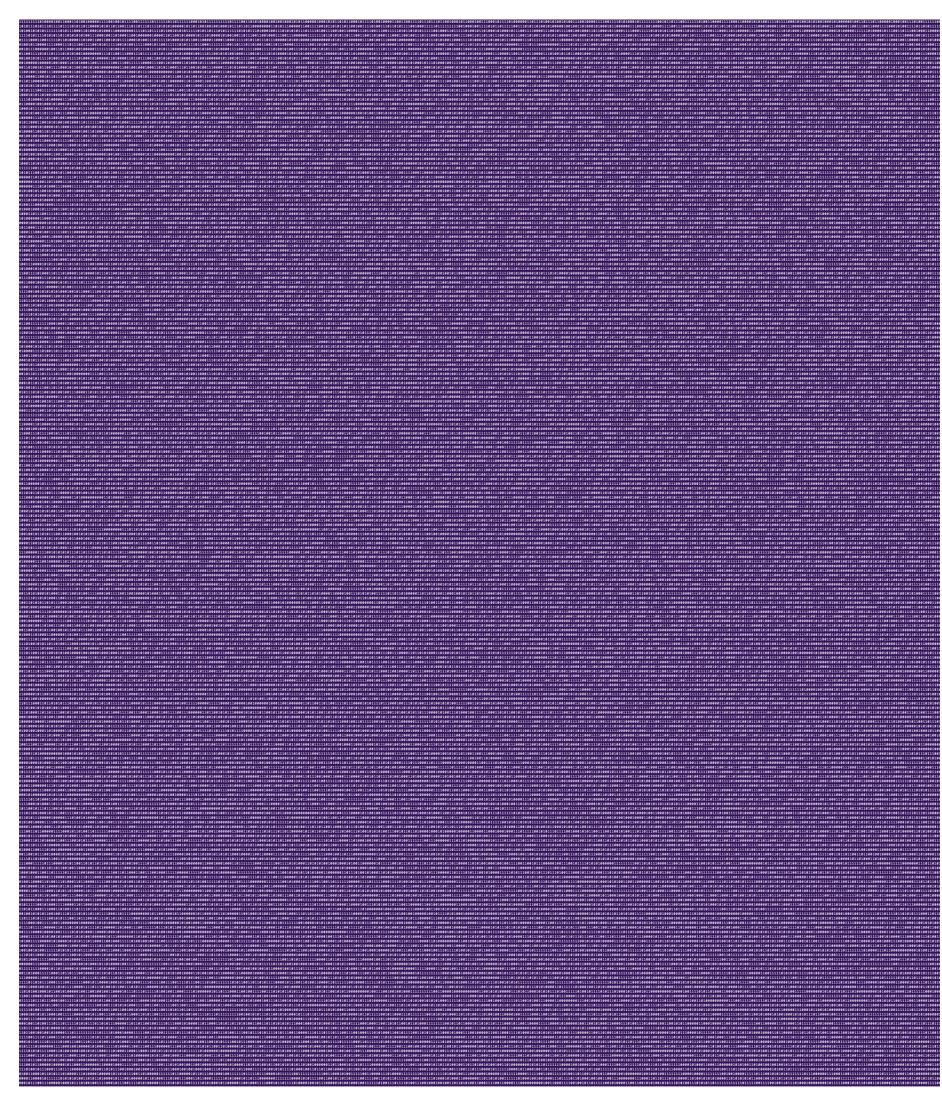
Undecidability



100,000 pseudo-random bits

Q. How would you compress these bits?

Undecidability



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Q. How would you compress these bits?A. With the following piece of code.

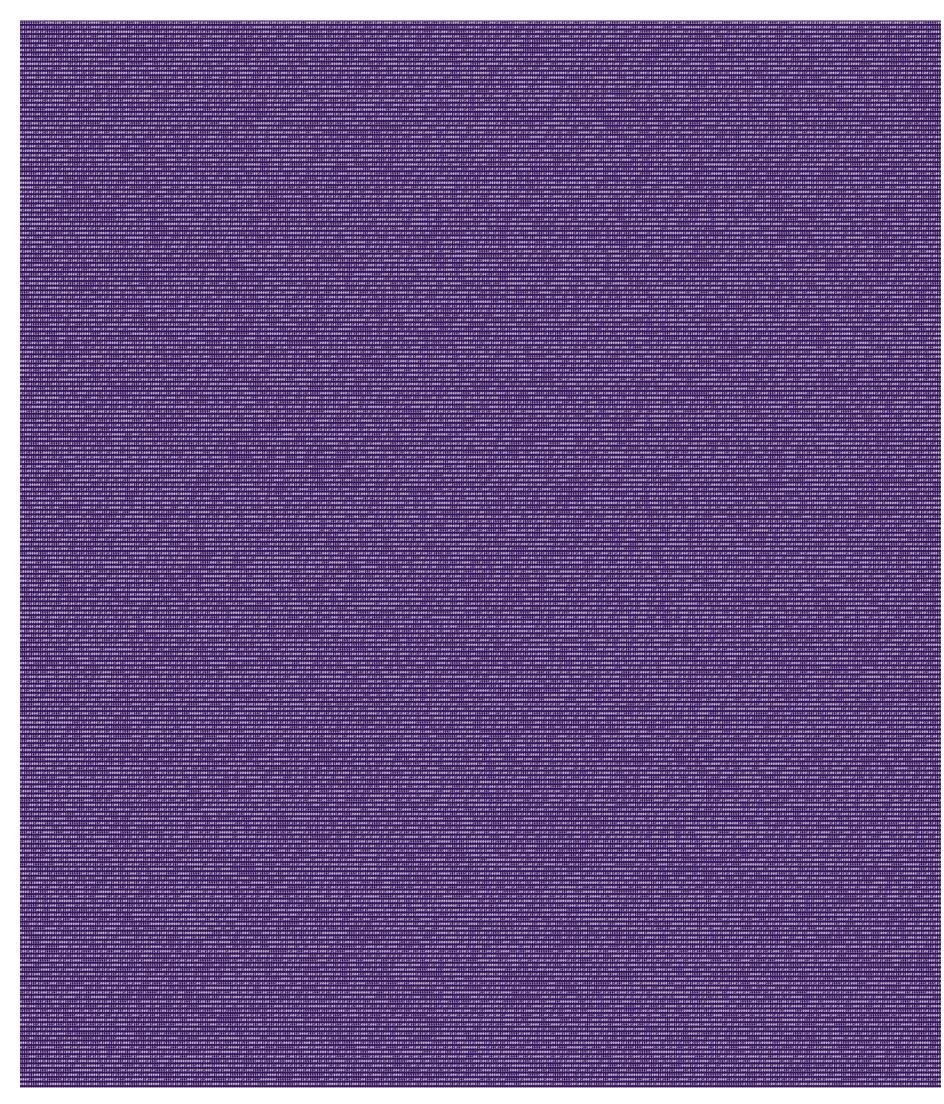
```
#include <iostream>
int main() {
    int n = 100000;
    int x = 989511;
    for (int i = 0; i != n; ++i) {
        x = x * 312523 + 852596;
        std::cout << int(x > 0);
    }
    std::cout << '\n';
    return 0;
}</pre>
```

- 220 bytes (1 char = 1 byte)
- $CR = 100000/(8 \cdot 220) = 56.8$

```
Compile with:
g++ random_bits.cpp -o random_bits

Run with:
./random_bits
```

Undecidability



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Data and Information

- Data and information are not the same.
- Information is the knowledge coming from interpreting data according to a specific semantic scheme.
- In the future, it is foreseen that data will grow much faster than information: data will become more and more redundant.
- So there are great possibilities for compression!
 Huge amount of research being actively carried out.

Why Data Compression?

- We use compression everywhere/anytime; even without being aware of it.
- Ever-increasing demand for storage and large-scale computing.
 - Generic file compression: gzip, bzip, LZ4, Zstd, ecc.
 - Multimedia: images (jpeg, png, gif); sound (MP3); video (MPEG, DVD); Spotify, Netflix, ecc.
 - Search engines (Google, Yandex, Bing,...);
 - Distributed storage (Dropbox, Google Drive,...);
- Communication cost.
 - Skype, Zoom, FaceTime, WhatsApp, ecc.
 - Social networks (Facebook, Instagram, Twitter,...);

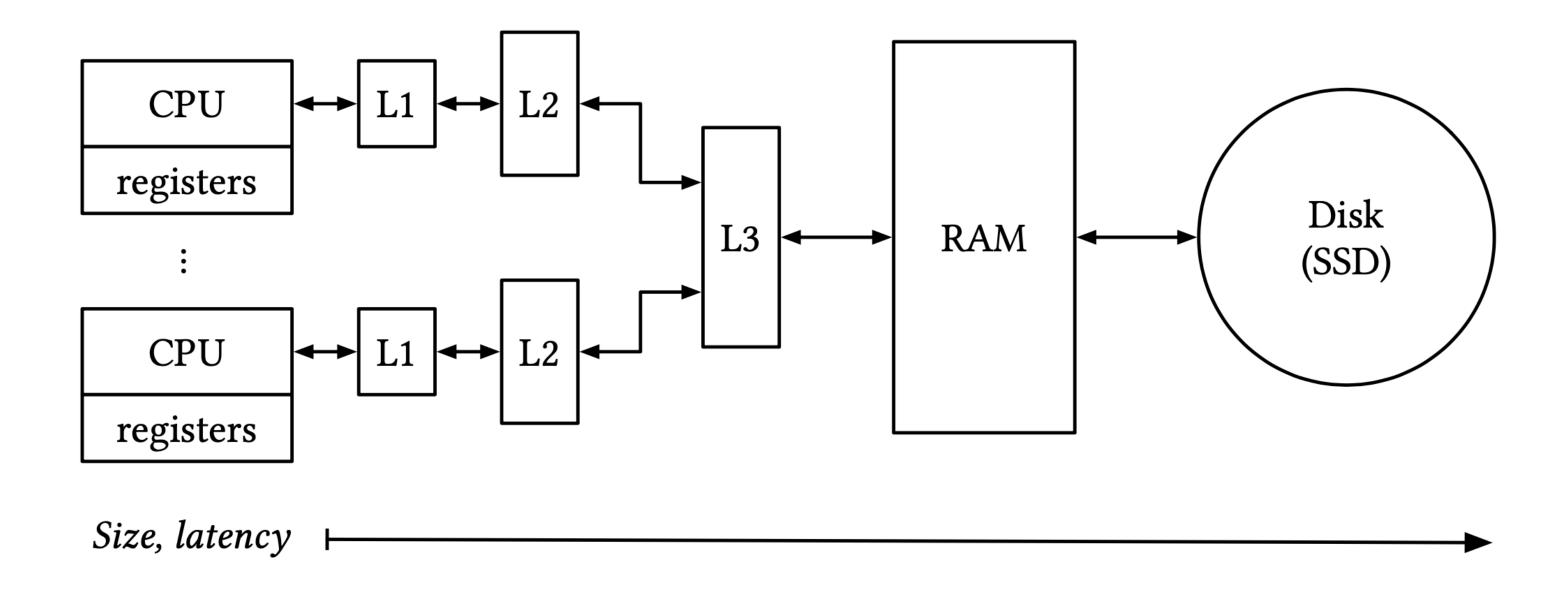
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 - Search engines (Google, Yandex, Bing,...);
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- Communication cost.
 - Skype, Zoom, FaceTime, WhatsApp, ecc.
 - Social networks (Facebook, Instagram, Twitter,...);
- Increased software performance.

Technological Limitations

- Whatever space we have available: we are going to fill it up, by virtue of our human, eager, nature.
- Moore's Law. Number of transistors on a chip doubles every 1.5 2 years.
- So we get faster processors...
- But not faster memories!

Memory Hierarchies

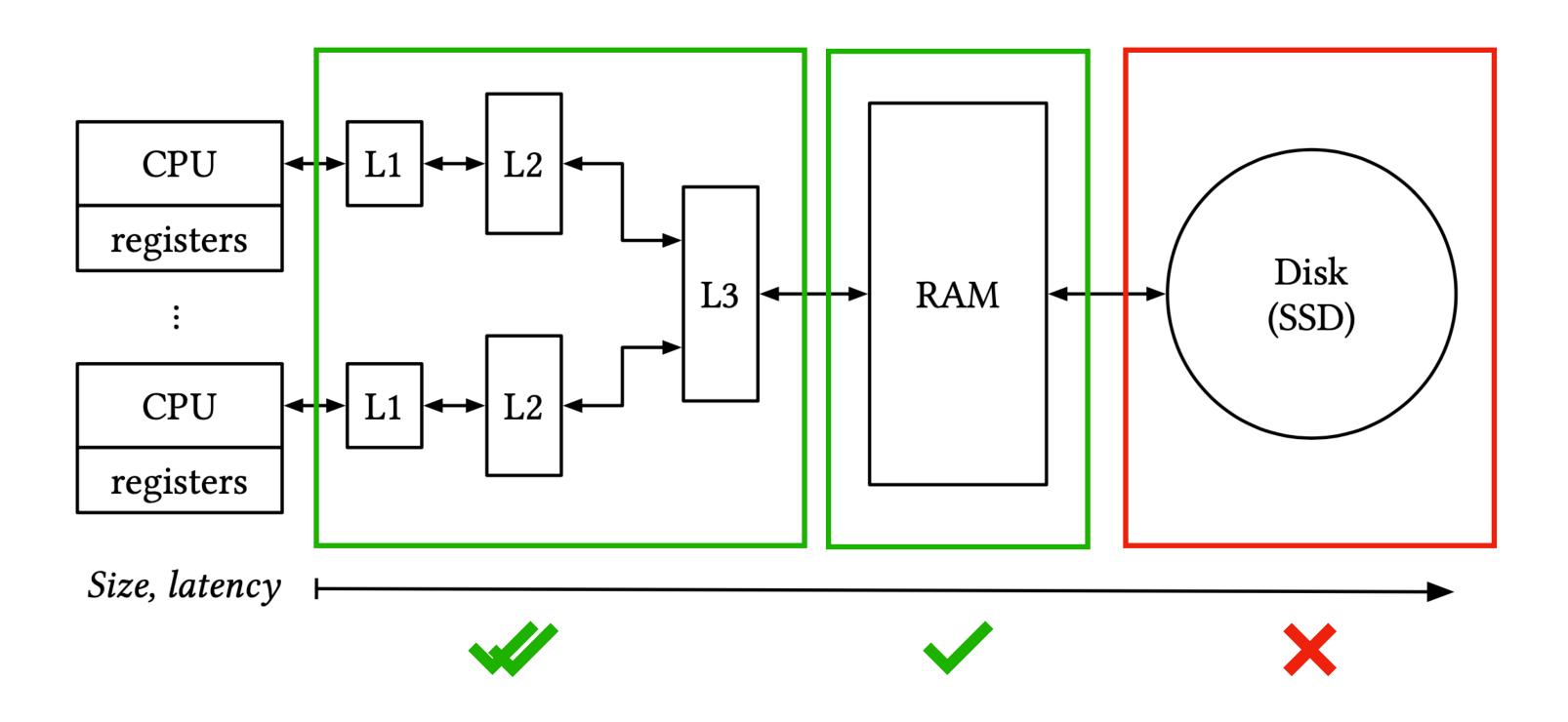


- If a program stalls...it is likely that it is waiting for memory.
- Thus, it is more important than ever to trade-off processor time for RAM/disk access time.
- Action of compression: transfer more data to the processor.

Memory type	Size	
registers	64	bits
L1	32	KB
L2	1	MB
L3	30	MB
RAM	64	GB
Disk	1	ТВ

Event	10 ⁻⁹ secs
L1 cache reference	1
L2 cache reference	4
RAM reference	100
SSD random read	16,000
Disk random read	3,000,000

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uint64_t is a primitive data type
for unsigned 64-bit ints;
uint8_t for unsigned 8-bit ints

```
struct large_record {
    uint64_t weight;
    uint64_t height;
    uint64_t day;
    uint64_t month;
    uint64_t year;
};
struct small_record {
    uint8_t weight;
    uint8_t height;
    uint8_t day;
    uint8_t month;
    uint16_t year;
```

With a b-bit unsigned integer, we can represent all values in $[0,2^b)$.

large_record consumes 40 bytes overall;
small_record consumes 6 bytes overall (slight lie)

Experiment methodology.

- 1. Allocate two vectors of the same size, one holding large_record objects and the other holding small_record objects.
- 2. Fill the two vectors with the same data.
- 3. Sort the two vectors (say, on the day attribute).

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```
initialise the pseudo-random generator
constexpr unsigned seed = 13; ←
                                     with a fixed seed to reproduce the results
std::srand(seed);
std::vector<large_record> large_records;
                                                 create the vectors
std::vector<small_record> small_records;
                                                 and reserve space
large_records.reserve(vector_size);
small_records.reserve(vector_size);
for (uint64_t i = 0; i != vector_size; ++i) {
   uint64_t weight = std::rand() % 256;
   uint64_t height = std::rand() % 256;
                                                  fill the vectors
   uint64_t day = std::rand() % 32;
    uint64_t month = std::rand() % 16;
    uint64_t year = std::rand() % 4096;
    large_records.emplace_back(weight, height, day, month, year);
    small_records.emplace_back(weight, height, day, month, year);
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```
typedef std::chrono::high_resolution_clock clock_t;
typedef std::chrono::milliseconds duration_t;
    auto start = clock_t::now();
    std::sort(large_records.begin(), large_records.end(),
               [](large_record const& x, large_record const& y) {
                   return x.day < y.day;</pre>
              });
    auto stop = clock_t::now();
    auto elapsed = std::chrono::duration_cast<duration_t>(stop - start);
    std::cout << "sorting vec took: " << elapsed.count() << " millisecs"</pre>
               << std::endl;
                                               use the std::sort algorithm to sort
                                               the vectors, using a lambda function
                                               to implement the comparison
    auto start = clock_t::now();
    std::sort(small_records.begin(), small_records.end(),
               [](small_record const& x, small_record const& y) {
                   return x.day < y.day;</pre>
              });
    auto stop = clock_t::now();
    auto elapsed = std::chrono::duration_cast<duration_t>(stop - start);
    std::cout << "sorting vec took: " << elapsed.count() << " millisecs"</pre>
               << std::endl;
```

std::chrono to measure time

- Q. Which sort will take less time?
- Hint. Remember! The smaller the data, the more data can be transferred to the processor.

```
Compile with:
g++ -std=c++11 -03 sort_bench.cpp -o sort_bench

Run with:
./sort_bench 10000000
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```
→ data_compression_course git:(master) x cd 1_introduction/code
→ code git:(master) x g++ -std=c++11 -03 sort_bench.cpp -o sort_bench
→ code git:(master) x ./sort_bench 100000000
sorting vec took: 348 millisecs
sorting vec took: 190 millisecs
→ code git:(master) x ./sort_bench 1000000000
sorting vec took: 3375 millisecs
sorting vec took: 1784 millisecs
→ code git:(master) x
```

The size of the data matters!

Further Readings

- Preface and Chapter 1 of:
 Alistair Moffat and Andrew Turpin. 2002. Compression and coding algorithms.

 Springer Science & Business Media, ISBN 978-1-4615-0935-6.
- Chapter 5.5 (pages 810-825) of: Robert Sedgewick and Kevin Wayne. 2011. Algorithms. 4-th Edition. Addison-Wesley Professional, ISBN 0-321-57351-X.