

A Crash Course on Data Compression

1. Introduction

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Overview

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

What is Data Compression?

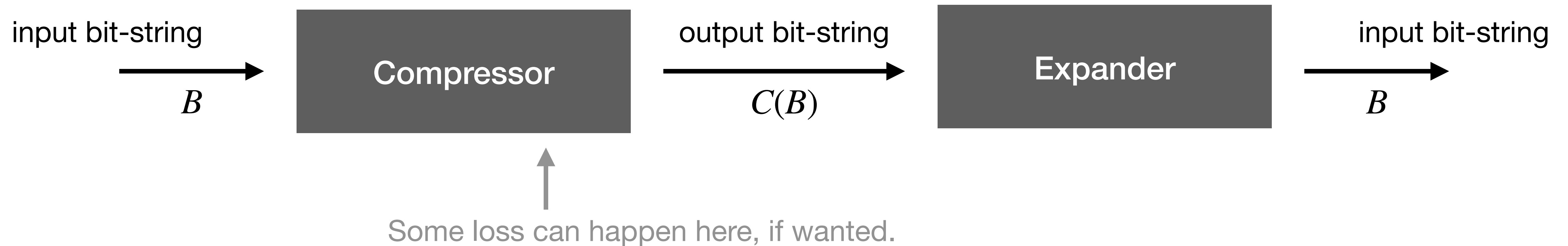
- The process for which data is transformed into another representation that takes *less storage space*:
 - *save space* when storing data,
 - *save time* when transmitting data.
- The process must be *reversible* (exactly or admitting some loss) to be useful.

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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*.

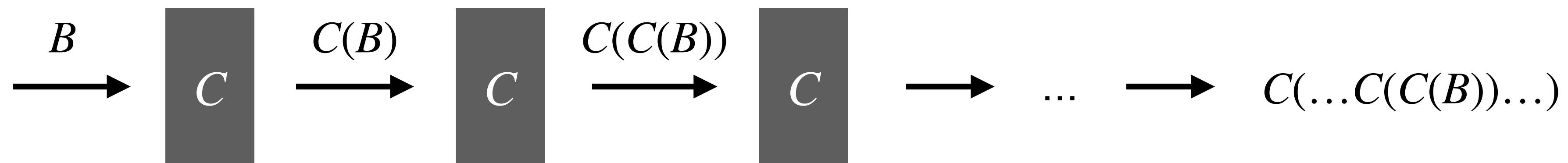
Limit

- **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.



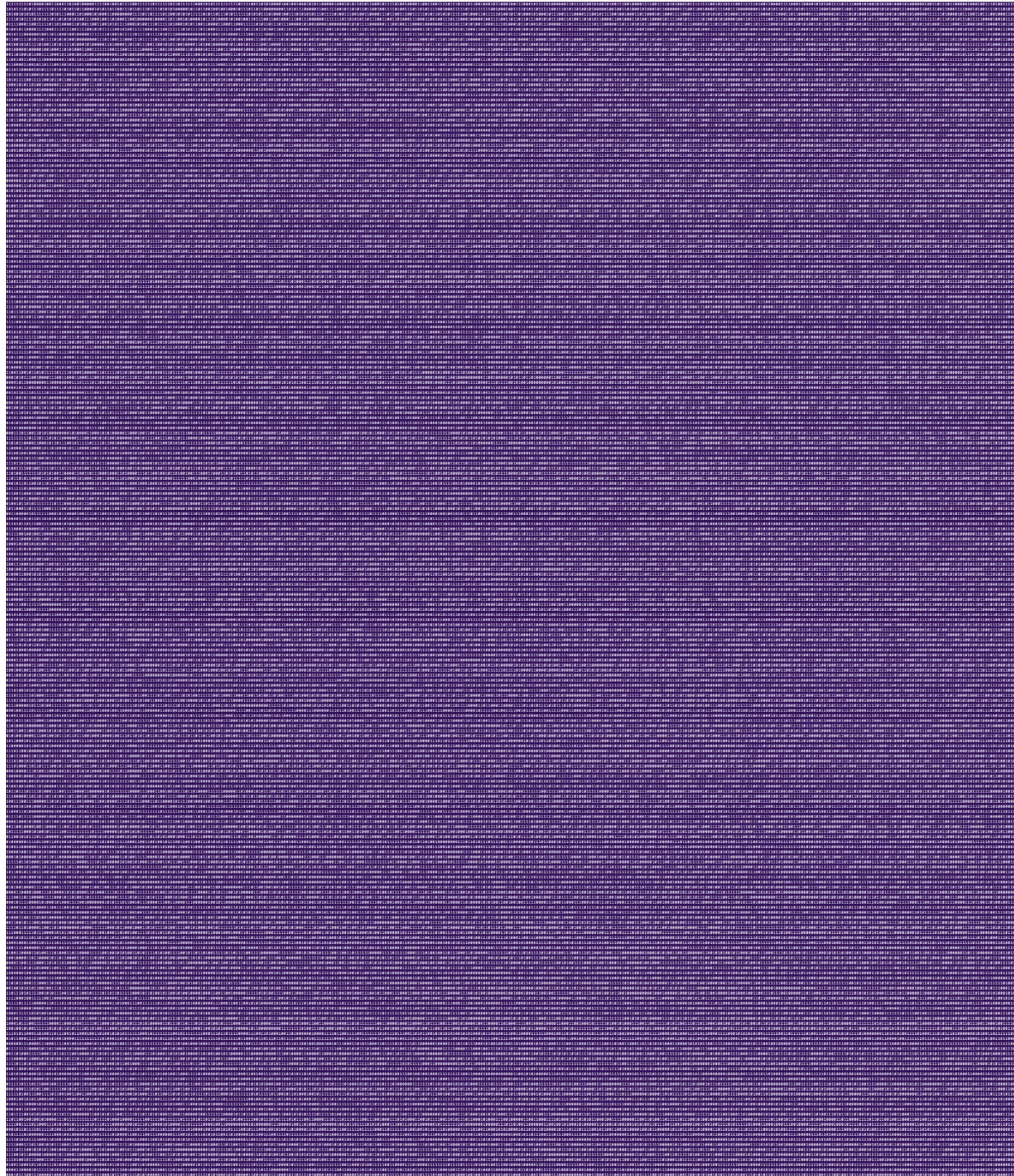
Then all possible bit-strings could be compressed to **0 bits** — absurd. ■

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

Q. How would you compress these bits?



100,000 pseudo-random bits

Undecidability

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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;
}
```

- 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

100,000 pseudo-random bits

Undecidability

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```
        x = x * 312523 + 852596;
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```
        std::cout << int(x > 0);
```

```
    }
```

```
    std::cout << '\n';
```

```
    return 0;
```

```
}
```

What if $n = 1000000$?

What happens now to the *CR*?

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

Compile with:

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g++ random_bits.cpp -o random_bits
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Run with:

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```

100,000 pseudo-random bits

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,...);

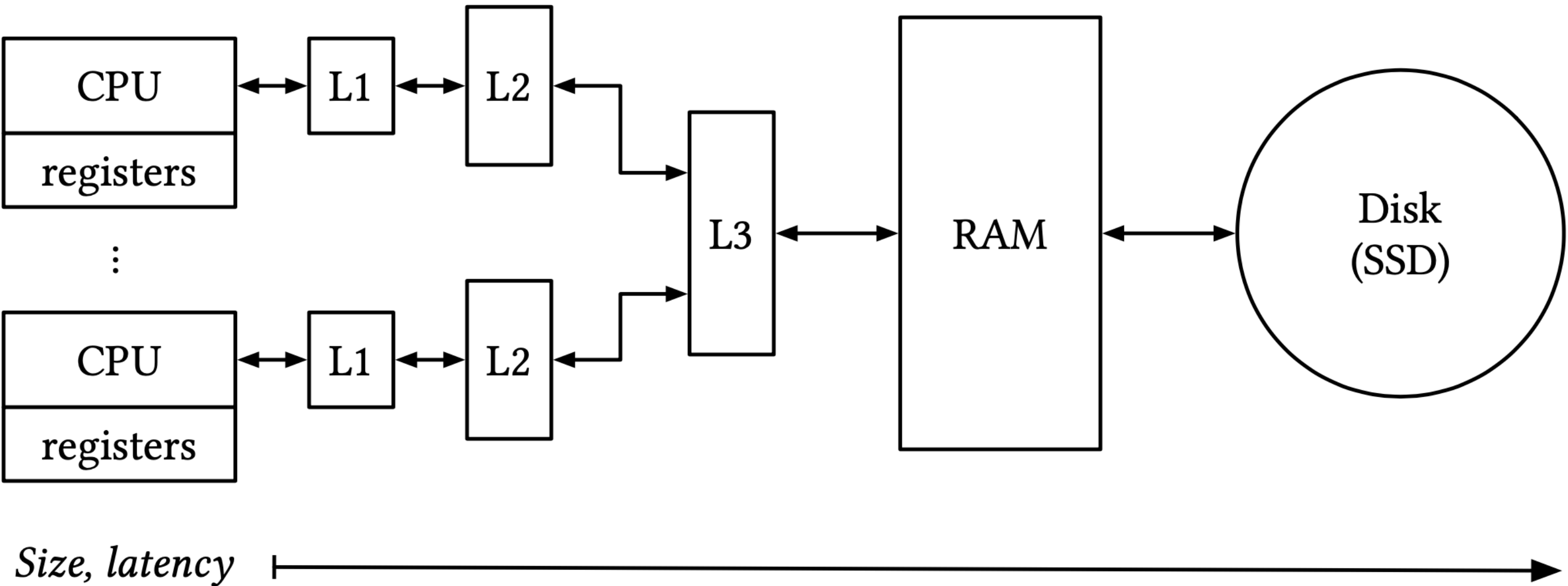
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 - Search engines (Google, Yandex, Bing,...);
 - Distributed storage (Dropbox, Google Drive,...);
- **Communication cost.**
 - Skype, Zoom, FaceTime, WhatsApp, ecc.
 - Social networks (Facebook, Instagram, Twitter,...);
- **Increased software performance.**

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

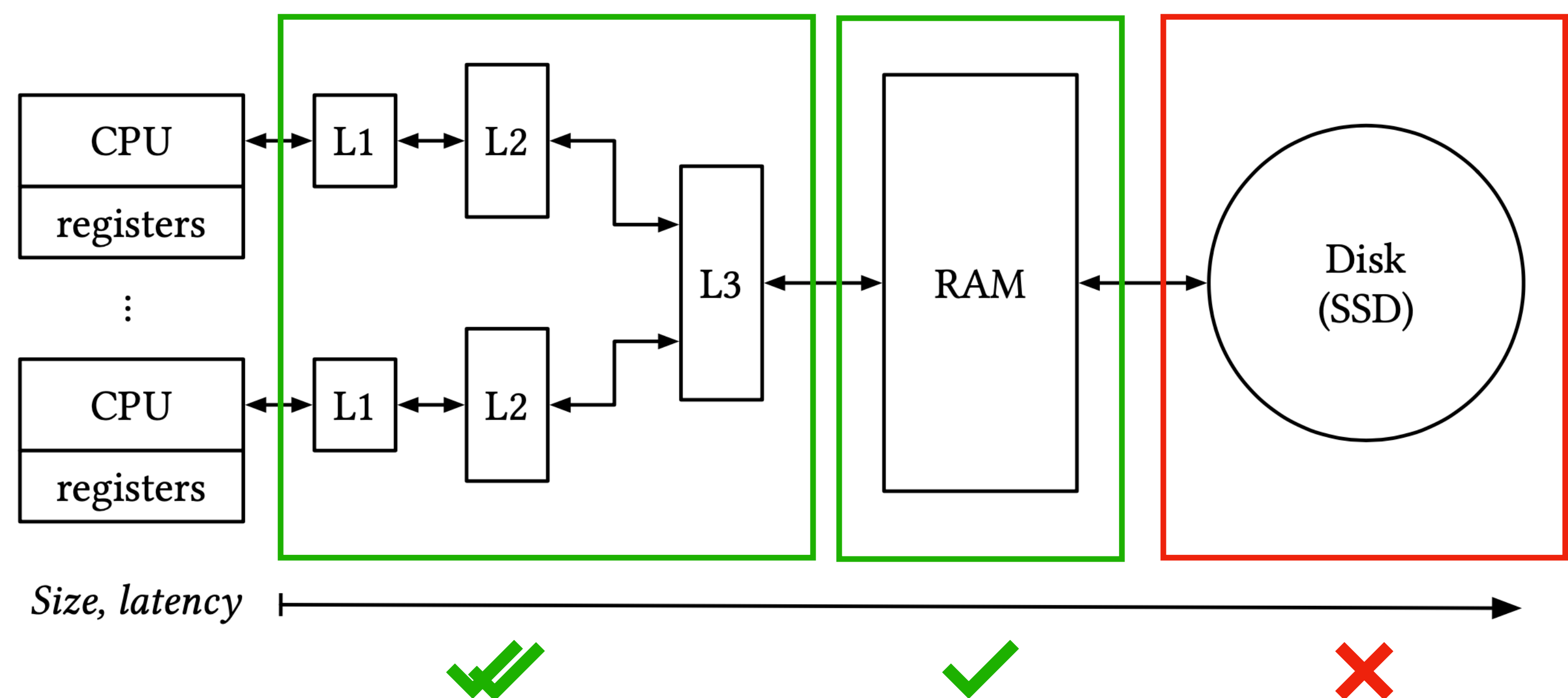


Memory type	Size	
registers	64	bits
L1	32	KB
L2	1	MB
L3	30	MB
RAM	64	GB
Disk	1	TB

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

- 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.*

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A Simple Experiment

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;  
};
```

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).

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

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```
constexpr unsigned seed = 13;
std::srand(seed);

std::vector<large_record> large_records;
std::vector<small_record> small_records;
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;
    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);
}
```

← initialise the pseudo-random generator
with a fixed seed to reproduce the results

← create the vectors
and reserve space

← fill the vectors

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}
```

← fill the vectors

1+2

3

```
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;
        });
    auto stop = clock_t::now();
    auto elapsed = std::chrono::duration_cast<duration_t>(stop - start);
    std::cout << "sorting vec took: " << elapsed.count() << " millisecs"
        << std::endl;
}

{
    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;
        });
    auto stop = clock_t::now();
    auto elapsed = std::chrono::duration_cast<duration_t>(stop - start);
    std::cout << "sorting vec took: " << elapsed.count() << " millisecs"
        << std::endl;
}
```

std::chrono to measure time

use the `std::sort` algorithm to sort the vectors, using a lambda function to implement the comparison

A Simple Experiment

- **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 -O3 sort_bench.cpp -o sort_bench
```

Run with:

```
./sort_bench 10000000
```

A Simple Experiment

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Run with:

```
./sort_bench 10000000
```

```
→ data_compression_course git:(master) x cd 1_introduction/code
→ code git:(master) x g++ -std=c++11 -O3 sort_bench.cpp -o sort_bench
→ code git:(master) x ./sort_bench 10000000
sorting vec took: 348 millisecs
sorting vec took: 190 millisecs
→ code git:(master) x ./sort_bench 100000000
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.