

## Data Handling: Import, Cleaning and Visualisation

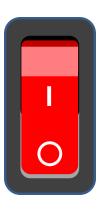
Lecture 4:

Data Storage and Data Structures

Prof. Dr. Ulrich Matter

### The binary system

Microprocessors can only represent two signs (states):



# The binary counting frame

Only two signs: 0, 1.

Base 2.

Columns:  $2^0 = 1$ ,  $2^1 = 2$ ,  $2^2 = 4$ , and so forth.

# Decimal numbers in a computer

Number 128 64 32 16 8 4 2 1

# Decimal numbers in a computer

Number	128	64	32	16	8	4	2	1
0 =	0	0	0	0	0	0	0	0
1 =	0	0	0	0	0	0	0	1
2 =	0	0	0	0	0	0	1	0
3 =	0	0	0	0	0	0	1	1
139 =	1	0	0	0	1	0	1	1

### Computers and text

How can a computer understand text if it only understands 0s and 1s?

**Standards** define how 0s and 1s correspond to specific letters/characters of different human languages.

These standards are usually called **character encodings**.

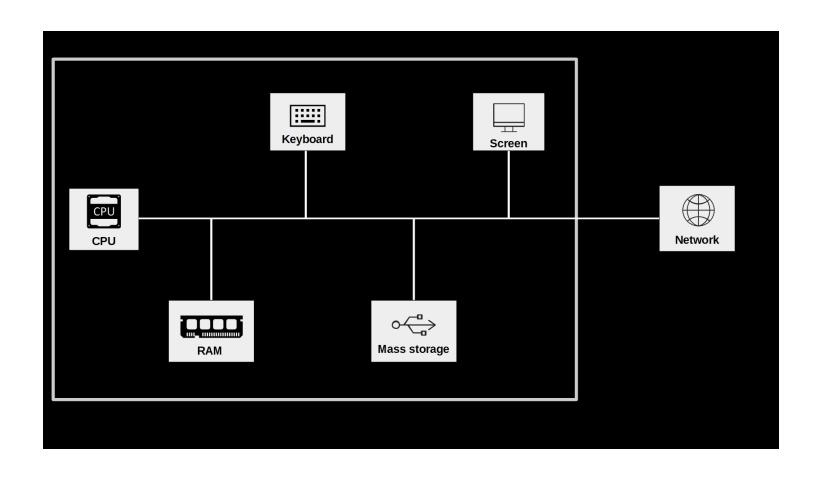
Coded character sets that map unique numbers (in the end in binary coded values) to each character in the set.

For example, ASCII (American Standard Code for Information Interchange).



ASCII logo. (public domain).

# Digital data processing



## Computer code

Instructions to a computer, in a language it understands... (R)

Code is written to **text files** 

Text is 'translated' into 0s and 1s which the CPU can process.

### Data storage

Data usually stored in **text files** 

Read data from text files: data import.

Write data to text files: data export.

#### Unstructured data in text files

Store Hello World! in helloworld.txt.

Allocation of a block of computer memory containing Hello World!. Simply a sequence of 0s and 1s...

.txt indicates to the operating system which program to use when opening this file.

Encoding and format tell the computer how to interpret the 0s and 1s.

### Inspect a text file

Interpreting 0s and 1s as text...

```
cat helloworld.txt; echo
## Hello World!
```

Or, from the R-console:

```
system("cat helloworld.txt")
```

### Inspect a text file

#### Directly looking at the 0s and 1s...

```
xxd -b helloworld.txt
```

### Inspect a text file

#### Similarly we can display the content in hexadecimal values:

xxd data/helloworld.txt

## 00000000: 4865 6c6c 6f20 576f 726c 6421 Hello World!

# **Encoding issues**

cat hastamanana.txt; echo

## Hasta Ma?ana!

What is the problem?

# **Encoding issues**

#### Inspect the encoding

file -b hastamanana.txt

## ISO-8859 text

### Use the correct encoding

Read the file again, this time with the correct encoding

```
iconv -f iso-8859-1 -t utf-8 hastamanana.txt | cat ## Hasta Mañana!
```

### **UTF** encodings

'Universal' standards.

Contain broad variety of symbols (various languages).

Less problems with newer data sources...

### Take-away message

#### Recognize an encoding issue when it occurs!

Problem occurs right at the beginning of the data pipeline!

Rest of pipeline affected...

... cleaning of data fails ...

... analysis suffers.

#### **Structured Data Formats**

Still text files, but with standardized **structure**.

**Special characters** define the structure.

More complex syntax, more complex structures can be represented...

#### Table-like formats

#### Example ch\_gdp.csv.

```
year,gdp_chfb
1980,184
1985,244
1990,331
1995,374
2000,422
2005,464
```

What is the structure?

#### Table-like formats

What is the reocurring pattern?

Special character,

New lines

Table is visible from structure in raw text file...

How can we instruct a computer to read this text as a table?

A simple parser algorithm

\_

#### CSVs and fixed-width format

```
'Comma-Separated Values' (therefore .csv)
```

commas separate values
other delimiters (;, tabs, etc.) possible
new lines separate rows/observations
(many related formats with other separators)
Instructions of how to read a .csv-file: CSV parser.

#### CSVs and fixed-width format

Common format to store and transfer data.

Very common in a data analysis context.

Natural format/structure when the dataset can be thought of as a table.

How does the computer know that the end of a line is reached?

### More complex formats

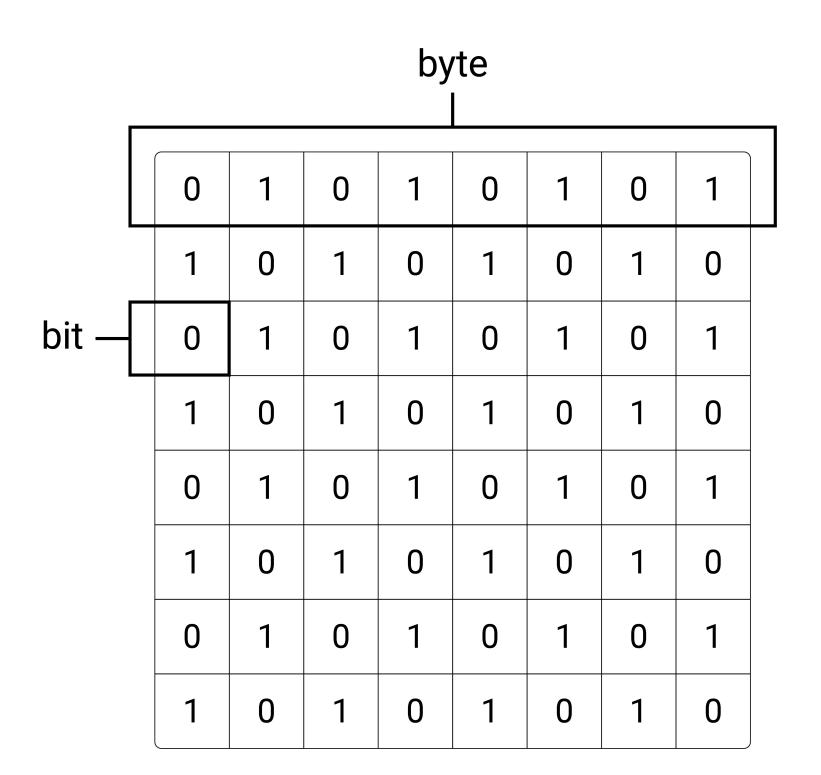
N-dimensional data

Nested data

XML, JSON, YAML, etc.

Often encountered online!

(Next lecture!)



## Bigger units for storage capacity

 $1 \text{ kilobyte (KB)} = 1000^1 \text{ bytes}$ 

 $1 \text{ megabyte (MB)} = 1000^2 \text{ bytes}$ 

1 gigabyte (GB) =  $1000^3$  bytes

#### Structures to work with...

Data structures for storage on hard drive (e.g., csv).

Representation of data in RAM (e.g. as an R-object)?

What is the representation of the 'structure' once the data is parsed (read into RAM)?

### Structures to work with (in R)

We distinguish two basic characteristics:

1. Data types: integers; real numbers ('numeric values', floating point numbers); text ('string', 'character values').

### Structures to work with (in R)

We distinguish two basic characteristics:

- 1. Data types: integers; real numbers ('numeric values', floating point numbers); text ('string', 'character values').
- 2. Basic data structures in RAM:

**Vectors** 

**Factors** 

**Arrays/Matrices** 

Lists

**Data frames** (very R-specific)

# Data types: numeric

```
a <- 1.5
b <- 3
```

R interprets this data as type double (class 'numeric'):

```
typeof(a)

## [1] "double"

class(a)

## [1] "numeric"
```

## Data types: numeric

Given that these bytes of data are interpreted as numeric, we can use operators (here: math operators) that can work with such functions:

```
a + b
## [1] 4.5
```

# Data types: character

```
a <- "1.5"
b <- "3"

typeof(a)

## [1] "character"

class(a)

## [1] "character"</pre>
```

# Data types: character

Now the same line of code as above will result in an error:

```
a + b
## Error in a + b: non-numeric argument to binary operator
```

### Data structures: vectors

#### Data structures: vectors

```
persons <- c("Andy", "Brian", "Claire")
persons

## [1] "Andy" "Brian" "Claire"

ages <- c(24, 50, 30)
ages

## [1] 24 50 30</pre>
```

## Data structures: factors

Male

Male

Female

#### Data structures: factors

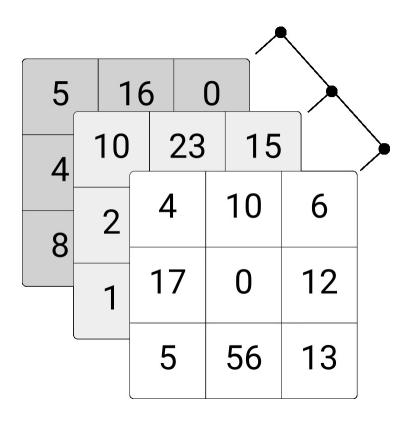
```
gender <- factor(c("Male", "Male", "Female"))
gender

## [1] Male Male Female
## Levels: Female Male</pre>
```

### Data structures: matrices

1	4	7	
2	5	8	
3	6	9	

## Data structures: arrays



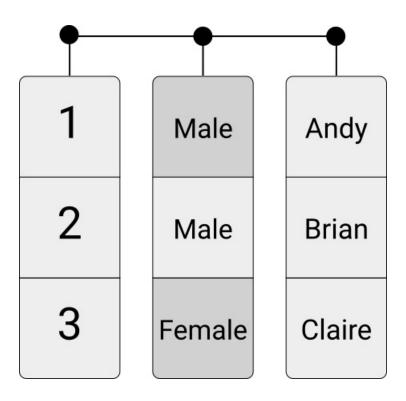
## Data structures: matrices/arrays

```
my_matrix <- matrix(c(1,2,3,4,5,6), nrow = 3)
my_matrix

## [,1] [,2]
## [1,] 1 4
## [2,] 2 5
## [3,] 3 6</pre>
```

## Data structures: matrices/arrays

## Data frames, tibbles, and data tables

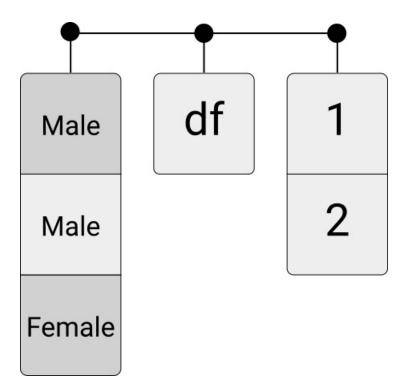


### Data frames, tibbles, and data tables

```
df <- data.frame(person = persons, age = ages, gender = gender)
df

## person age gender
## 1 Andy 24 Male
## 2 Brian 50 Male
## 3 Claire 30 Female</pre>
```

### Data structures: lists



#### Data structures: lists

```
my list <- list(my array, my matrix, df)</pre>
my list
## [[1]]
## , , 1
##
## [,1] [,2]
## [1,] 1 3
## [2,] 2 4
## , , 2
## [,1] [,2]
## [1,] 5 7
## [2,] 6 8
##
## [[2]]
  [,1] [,2]
## [1,] 1 4
## [2,] 2 5
## [3,] 3 6
##
## [[3]]
```

### References