# Introdution to medical informatics. Data and information.

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#### Outline

- Course overview
- Medical informatics. Why? What for?
- Fundamental role of informatics in modern science.
- Information. Data. Knowledge.

#### Course overview

- Information. Data. Knowledge
- Biostatistics
- Formal Logic. Expert and decision support systems in medicine
- Evidence-based medicine

Medical informatics – the application of informatics to the field of biology and medicine/healthcare.

# Informatics, Computers science

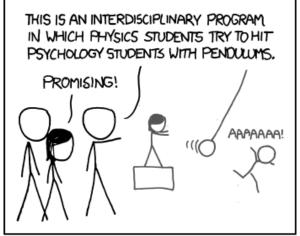
- Informatics the study of the processing, management, and retrieval of information
- Computers science is the study of the theoretical foundations of information and computation and of practical techniques for their implementation and application in computer systems
- **Cybernetics** is the interdisciplinary study of the structure of regulatory systems.

# Medical informatics. Why? What for?

Medical informatics borrows concepts/ideas from:

- Medicine
- Informatics
- Logic
- Statistics

# Medical informatics – interdisciplinary study



MY PROFESSORS HAD AN ONGOING COMPETITION TO GET THE WEIRDEST THING TAKEN SERIOUSLY UNDER THE LABEL "INTERDISCIPLINARY PROGRAM,"

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- Any object in physical reality can be described with some information
- Any physical process can be described as informational process
- ullet Physical laws  $\sim$  program/algorithm
- Universe is a gigantic computer!
  - Seth Lloyd, Programming the Universe
  - Roger Penrose, The Emperor's New Mind

# Fundamental role of informatics in modern society

That's why informatics is an essential part of science disciplines in any University.

You will need basic informatics skills in modern society:

- We live in so called information society. (or at least tend to develop it)
- You may need to analyze systems/products
- You may need apply informatics skills in your research projects
- You may need to implement prototypes
- You may need to oversee programmers in your future job

# Fundamental concepts of informatics

- Information
- Computation
- Data
- Knowledge

# **Information** – primitive notion<sup>1</sup> of modern science.

So it is not defined formally in terms of previously defined concepts, but is only motivated informally, usually by an appeal to intuition and everyday experience

<sup>&</sup>lt;sup>1</sup>fundamental undefined concept; primitive, undefined, term  $\bullet \bigcirc \bullet \bullet \supseteq \bullet \bullet \supseteq \bullet \circ \bigcirc \bigcirc$ 

# **Properties**

# Properties of information:

- Objectivity information doesn't depend upon storage method or your opinion
- Authenticity how trustworsy it is
- Accuracy
- Actual

#### Transmission of Information



All communicative situations include the presence of the sender information, its recipient and channel. Between any sender and a receiver there is always a transmission channel through which the message must pass.

#### Measure

Function  $\mu: \Omega \to \mathbb{R}$  is called **measure**, if:

- $\forall A \in \Omega : \mu(A) \geq 0$
- $\forall A, B \in \Omega, A \cap B = \emptyset : \mu(A \cup B) = \mu(A) + \mu(B)$
- $A_n \searrow \varnothing : \mu(A_n) \longrightarrow 0$

Measure is mathematics's formalization/generalisation of such trivial common notion as **volume**.

#### Measures of Information. Shannon's measure

Well, we don't *exactly* know what the information is. But for our practical goals we can define *measures of information*, depending on our goals.

Shannon's measure<sup>2</sup>:  $I(x) = \log(\frac{1}{P(x)}) = -\log(P(x))$  Here  $x \in X$  – some message from the source X, P(x) – probability of receiving message x.

#### Example:

On tossing a coin, the chance of 'tail' is 0.5. When it is proclaimed that indeed 'tail' occurred, this amounts to  $I('tail') = \log_2(\frac{1}{0.5}) = \log_2 2 = 1$  bits of information.



<sup>&</sup>lt;sup>2</sup>also called self-information

#### Shannon's measure. WTF?

Meaning behind the stages of all this math. Why Shannon has designed this measure to look so? His motives were simple:

- **9**  $P(x) \searrow \Rightarrow I(x) \nearrow$  more unexpected message contains more information.
- **②** For independent messages we can add their informational measures:  $\forall x, y \in X; x \cap y = \emptyset : I(x \cup y) = I(x) + I(y)$
- $I(x) = 0 \Rightarrow P(x) = 1$  If we already know smt, then we don't receive any information at all.

Unfortunately Shannon's measure is not suitable for complex<sup>3</sup> messages.

### Entropy.

**Entropy** – amount of disorder in a system In *information theory*, **entropy** is a measure of the uncertainty associated with a random variable(in our case received message); Entropy is an average amount of information in one message.

$$H(X) = M(I(x)) = \sum_{x \in X} p(x) \log p(x)$$

# Entropy. Example.

| Message               | Mark   | Number of students   |
|-----------------------|--|--|
| $x_1$                 | 5  | 125  |
| <i>x</i> <sub>2</sub> | 4  | 250  |
| <i>X</i> <sub>3</sub> | 3  | 500  |
| <i>x</i> <sub>4</sub> | 2  | 125  |
|                       | $ \begin{array}{ccc} x_i & P(\\ x_1 & \frac{1}{8} \\ x_2 & \frac{1}{4} \\ x_3 & \frac{1}{2} \\ x_4 & \frac{1}{8} \end{array} $ | $     \begin{cases}       x_i \\       x_i     \end{cases}     $ $     \begin{cases}       I(x_i) \\       2 \\       1 \\       3     \end{cases}   $ |
| H(X) =                | $\frac{1}{8} \cdot 3 +$  | $\frac{1}{4} \cdot 2 + \frac{1}{2} \cdot 1 + \frac{1}{8} \cdot 3$  |

Let's have a break!

#### Data

**Data** refers to qualitative or quantitative attributes of a variable or set of variables.

- For example results of measurements: (35.7, 36.7, 42, 38, 39.4, 37.5, 40.1)
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- Raw data must be interpreted and take on a meaning
- Example: (35.7, 36.7, 42, 38, 39.4, 37.5, 40.1) results of patients thermometrics

# Knowledge

**Knowledge** is a familiarity with someone or something, that can include facts/data, descriptions, information, and/or skills acquired through experience or education.

Main difference between knowledge and data: knowledge has dynamic nature by itself, it can change via learning. This the highest level of information abstraction.

# Computation

• **Computation** – any type of information processing.

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- To describe a computation we need some formal way to state rules for information processing algorithm.

# Algorithm

**Algorithm** is an effective method expressed as a finite list of *well-defined*<sup>4</sup> instructions for computation.

Starting from an initial state and initial input (perhaps empty), the instructions describe a computation that, when executed, will proceed through a finite number of well-defined successive states, using finite amount of resources, eventually producing "output"

<sup>&</sup>lt;sup>4</sup>Because "well-defined" is not strict, formal term – the whole definition, presented here, is **informal**. Though it is sufficient for basic intuitive understanding of the topic

# Properties of Algorithms

- Finiteness An algorithm must always terminate after a finite number of steps
- Definiteness Each step of an algorithm must be precisely defined; the actions to be carried out must be rigorously and unambiguously specified for each case
- Discrete moving in separate(discrete) steps: 1, 2, 3, ...
- Input
- Output

# Computability and computational complexity

Some task is called **computable** if there exists an algorithm for it.<sup>5</sup> Not all tasks are computable. **Computational complexity** – informally,

amount of resources we'll need to complete the task using selected algorithm. This characteristics helps us do determinate the most optimal algorithm to solve our problems.



<sup>&</sup>lt;sup>5</sup>Literally: it can be solved in general case.

# Example – herbal drug foalfoot

- 10 g of herbal drug (2 spoons) place to enamel ware,
- 4 Add 200 ml (1 cup) of boiled room-temperature water
- close cover and infuse on the water bath for 15 min
- Cool down for 45 min at the room temperature
- strain
- press out the rest to the strained infusion
- The infusion adds up with boiled water to 200 ml
- Take 0, 5 cup of warm infusion two times a time 15 min before meal

# Flowcharts

| Name                     | Drawing    |
|--------------------------|------------|
| Start and end symbols    |            |
| Input/Output             |            |
| Generic processing steps |            |
| Conditional or decision  | $\Diamond$ |

# Types of Algorithms and Their Flowcharts

- Linear algorithm
- Branch algorithm
- Cycle algorithm

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