# Advice to young Al scientists

Current Trends in Al

13 lessons learnt







LAB

Ceci n'est pas d'intelligence

• Add MATHS!





### Exercise #1

- Is ChatGPT3 intelligent?
- How does ChatGPT3 deal with deciding... relevance?
- Can ChatGPT3 perform **common sense** reasoning?
- How does it model domain knowledge? E.g. physical knowledge?





### **#1 – Be critical**

# We are all part of a culture that boosts learning but also often inhibits fundamental progress.





#### Exercise

• Everyone tell one thing about/related to chairs

• How would you programme a machine to answer such questions?





### **#2 - Be humble.**

#### We are nowhere yet, there is still a LOT to investigate.





#### Exercises

- How does Chat GPT3 represent a chair?
- Design a neural network to divide two numbers.
- Design a minimal neural network by hand that models XOR
  - Hard limiter activation function
  - Two teams: 1) bits encoded as (-1, +1), 2) bits encoded as (0,1)





### **#3 - Representations are key**





### The Iris dataset...





### Exercise

- What do('n t) you know about the Iris dataset?
  - How were the measurements performed?
  - How do you define "length" and "width" of a leaf
  - Where were the samples collected? In one place?
  - By one person? In one year?
  - What is the meaning of the labels?





### #4 - Try to bring real impact

Reality is complex, problems are ill-defined, everyone talks a different language.

And intriguing!!!





### But start with the lowest possible goal.

You can always increase complexity if you move faster than expected\*.

(\*this never happens)





#### Exercise leads to higher cholesterol levels







### Causality

- Age
  - Has a causal link to exercise
  - And potentially to cholesterol
- Age is thus a confounder of exercise and cholesterol
- So we should control for it!





#### Exercise leads to lower cholesterol levels







### **#5 – Remain curious**

Education starts when you graduate.

JUDEA PEARL WINNER OF THE TURING AWARD AND DANA MACKENZIE THE BOOK OF WHY WHY





### Quiz

- When was the term "Machine Learning" designed?
  - 1959 by Arthur Samuel
- When was Object Oriented Programming (+/-) invented?
  - Late 50ies in the context of LISP programming
- When were CNNs invented?
  - 50ies, 80, 89, ...
- Who can be considered the pioneer of NNs?
  - Frank Rosenblatt, a psychologist. Perceptron: 1957





### New is really not always better

- A lot of garbage
- Big ideas come slowly
- Publication pressure
- Roots are great sources of insipiration
  - How did the original researchers come to this idea?







### #6 -Go back to the roots

- Read the classics!
- Find the original ideas.
- Research = a very asymmetric tree
- Unexplored ideas are often valuable
- Follow people first, then papers.









### **#7 - Break the cat**

#### Romanian saying. Start with what you fear the most

You think it will be too slow, you don't know how to implement this piece, ...?





#### Exercise

- The stable matching problem, in its most basic form, takes
  - as input equal numbers of two types of participants (n medical students and n internships, for example),
  - an ordering for each participant giving their preference for whom to be matched to among the participants of the other type.
- A matching is *not* stable if:
  - there is an element A of the first matched set which prefers some given element B of the second matched set over the element to which A is already matched, and
  - B also prefers A over the element to which B is already matched.





### Gale-Shapley algorithm

```
Initialize all men and women to free
while there exist a free man m who still has a woman w to propose to
{
    w = m's highest ranked such woman to whom he has not yet proposed
    if w is free
        (m, w) become engaged
    else some pair (m', w) already exists
    if w prefers m to m'
        (m, w) become engaged
        m' becomes free
    else
        (m', w) remain engaged
}
```





#### **#8 - Problem complexity ≠ Solution complexity**

Problems that can be formulated in simple terms, can have complex solutions. And vice versa.

#### DO NOT confuse the solution with the problem!





### **#9 - Move slowly. Slower...**

Take more time to understand the problem than the solution.

"Once you *really* understand the problem, you're probably smart enough to find a solution" (Michiel Steyaert)





# **#10 - On methodology**

How to cope with too complex problems?





### The scientific method

- 1. Define questions
- 2. Think how to answer them
- 3. Code / perform experiment
- 4. Learn & update





### Coming up with questions

- **Don't assume** you know: turn every certainty into a (hierarchy) of research question(s).
- Simplify questions & scope till it seems **trivial** to implement (which it will not be)
- What is the **simplest next** experiment you can think of from which we can learn something?
- To finally arrive to an algorithm/implementation that can be evaluated with a ground truth





### Coming up with questions

- An experiment can be any activity that provides you with ANSWERS: building a prototype and measuring outcomes, literature review, strength & weaknesses, brainstorm for new scientific ideas, having a conversation with the customer
- Some thoughts how to generate **hypotheses**: (a) state the obvious (better too many than too few), (b) verbalize you intuition: describe what you think and rephrase in questions, (c) operationalise definitions, verbs





### Organising results

- For each question, create a notebook / document with a fixed structured and a section for:
  - (a) the question + more specific description,
  - (b) the experiment you construct to answer it (and support it),
  - (c) hypotheses: what do you expect to see in the answers,
  - (d) the implementation itself,
  - (e) observations from the results,
  - (f) link back the answer, and anything else you have learnt





### **#10 - Build or think?**

# Decide carefully when you need to think things through, and when you need to build!





### Build or think?

- Build
  - When you do not know all factors influencing the problem/solution
  - When there are too many assumptions
  - When moving to the solution phase
  - Because it forces you to go all the way
  - Because we always underestimate the complexity of things

#### • Think

- When you need to get things tight.
- When you start to get a **full** picture
- To keep things **clean**
- After building
- To get a **fresh** mathematical view





### Grasping a tomato

• Would you consider that an AI task?







### **Social robotics**

- Playing cards
- Looking into the eyes





# **#11 - Learn from building**

- Traverse bottom-up and top-down
- End-to-end
- Measure bottlenecks, do not assume them





### #12 - Go for elegance





### The problem with extrinsic evaluation

- If you evaluate the correctness of an answer only
  - System A learns everything by heart
  - System B learns the underlying structure
- Which system is the "smartest"?
- Al is heavily influenced by behaviourism (Turing!). Not necessarily a good thing...





### Is a traffic sign Al?

- It's technology...
- And it makes us smarter!







### **#13 - Don't forget embodiment**





## **Questions?**



