

# Data science – an occupation should become a profession?


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# What to expect in the next minutes

- Some definitions, but no mathematics!
- Original artwork copyright Ursula Garczarek!
- Some provocative statements.
- Interruptions (a real debate would be perfect).
- A line of reasoning, why it is essential that data science develops professional ethics and becomes a profession.

# Definition of data science (Donoho 2017)

*Data science is the science of learning from data; it studies the methods involved in the analysis and processing of data and proposes technology to improve methods in an evidence-based manner. The scope and impact of this science will expand enormously in coming decades as scientific data and data about science itself become ubiquitously available.*

1. Data gathering, preparation, and exploration,
2. data representation and transformation,
3. computing with data,
4. data modelling,
5. data visualization and presentation,
6. science about data science.

# Occupation vs Profession (Airaksinen, 2009)

A profession is different from an occupation through

- Scientific training  
*Knowing what is to be done by understanding the rational, epistemological foundations of professional action*
- Autonomy  
*A profession can influence the social decisions that regulate its members' work and their related rights and obligations*
- Professional ethics  
*If the public needs this expertise and therefore cannot unproblematically reject, challenge, or ignore the professional advice and the influence of their work, professional ethics becomes a key issue when the public evaluates the potential bias of professional work in relation to the quality of their life. (Trust!)*

# Does data science form a profession?

- The Code of Conduct of the RSS defines statistics as a profession.
- Data science is only in part statistics. It draws a lot from i.e. computer science.
- Data science was born as a tool in a commercial surrounding, where data was abundant, cheap to store, cheap to process.
- No common rational, epistemological foundation of data science education. Some start from the inferential framework of statistics, some start from computational learning theory in machine learning, or even economics!
- A very weak education in ethical aspects of data science, but improving. Most new study programs include some ethics courses.

**No! But why not and why would it help?**

# Our statement for debate

*Data science is in the focal point of current societal development.*

*Without becoming a profession with professional ethics,*

*data science will fail in building trust*

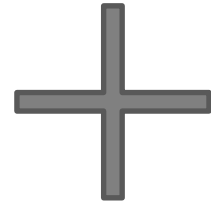
*in its interaction with*

*and it's much needed contributions to*

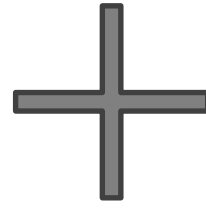
*society!*

**Line of reasoning**

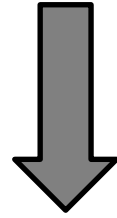
Societal  
impact



Ethical  
issues



Influence



Responsibility

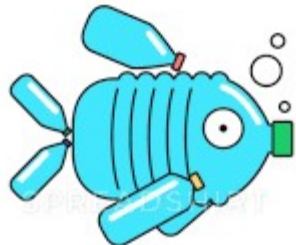
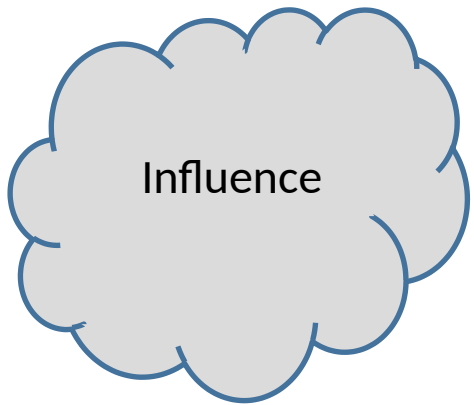
Lack of  
knowledge

Lack of  
skills

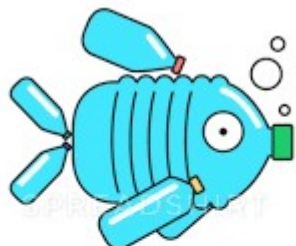
Lack of  
power

Lack of  
interest

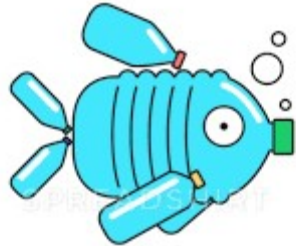




Lack of knowledge



Lack of Skills



Lack of power



**Reasoning step by step**



## Agree or Disagree?

**“Societal impact of data science is large!”**

- Vast amount of data (on individuals and the general public)
- Increased computational power
- Fast, low cost processing of large databases
- Used for decision making on all aspects of (human) life  
(Weapons of Math Destruction, Cathy O’Neill)
- (In general the progress of computational power stresses human imagination and human scales of time. In my opinion this changes the **quality** of applying an algorithm, **not only** the **quantity**.)



Ethical  
issues

## Agree or disagree?

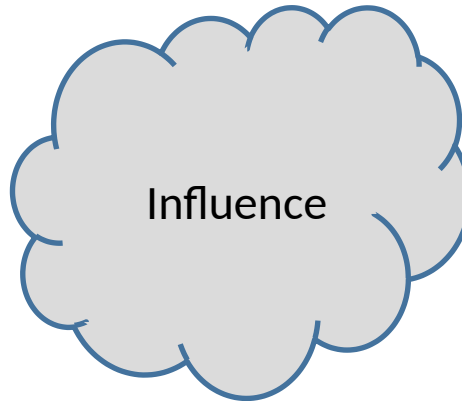
**“Using algorithms can raise ethical issues!”**

- Autonomous machines: a threat to free will and responsibility
- Bias, discrimination and exclusion (not only black box algorithms)
- Algorithmic profiling: personalisation versus collective benefits
- Preventing massive files while enhancing AI: seeking a new balance
- Quality, quantity, relevance: the challenges of data curated for AI
- Human identity before the challenge of artificial intelligence

## Agree or disagree?

**“Data scientists have influence on design and application of algorithms”**

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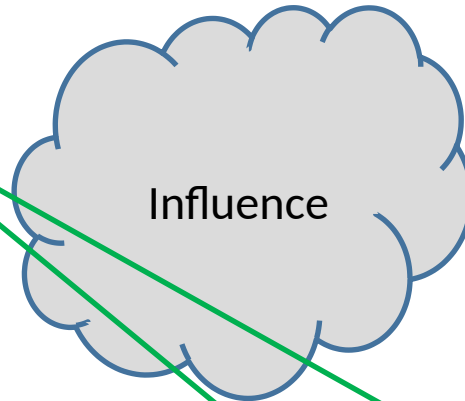


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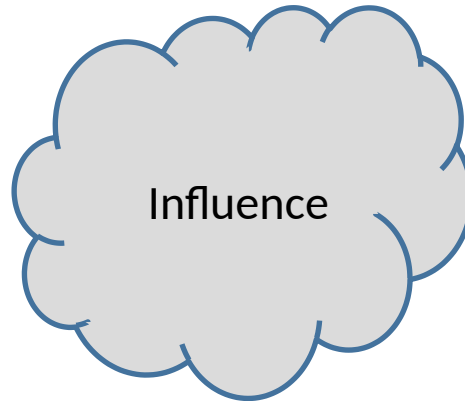


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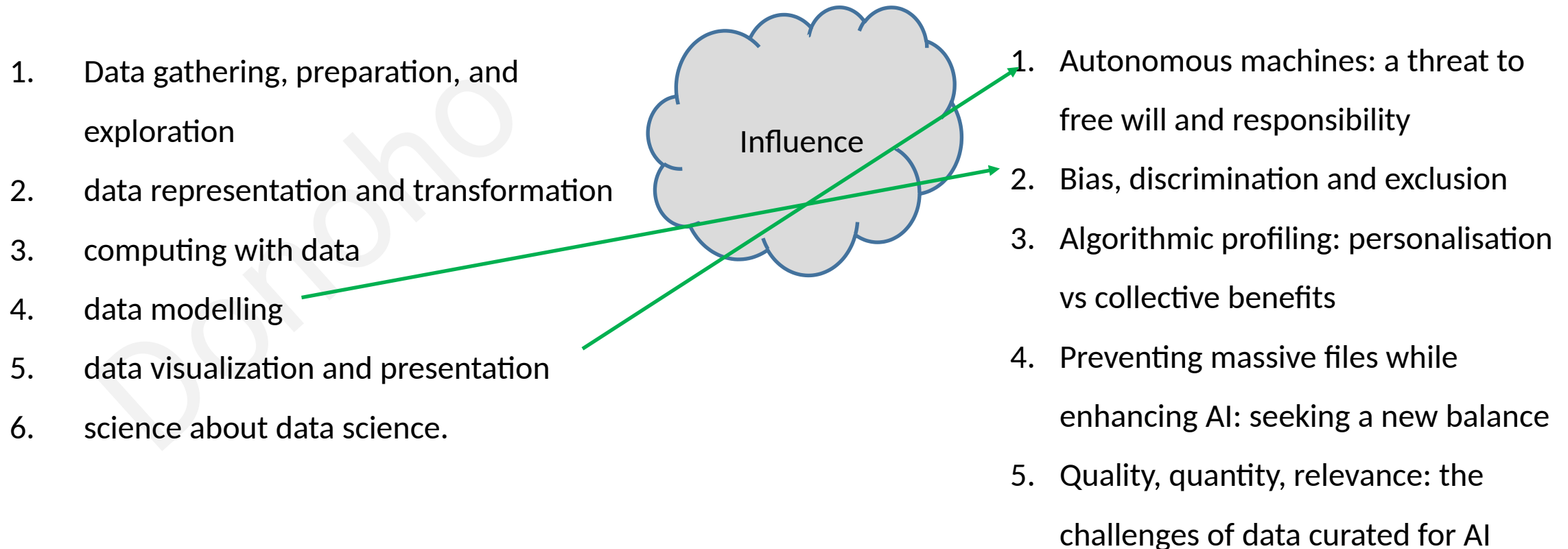
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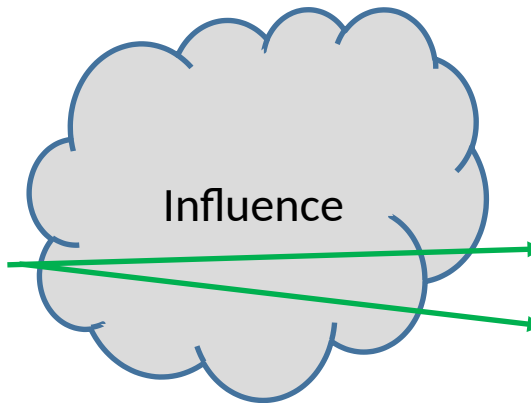




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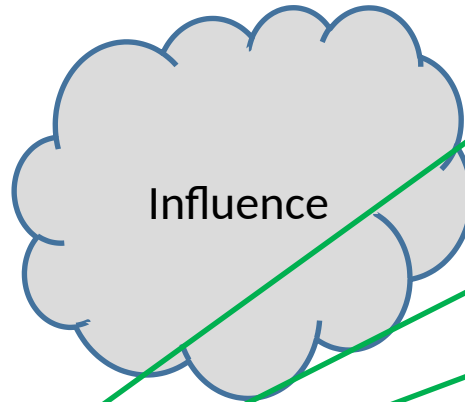


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## Agree or disagree?

### “Influence leads to responsibility”

- GI (Gesellschaft für Informatik): No data science specific aspects, „Do no harm!“
- GfKI (Data Science Society): Trying to establish a working group on ethical guidelines
- ASA: Because society depends on informed judgments supported by statistical methods, all practitioners of statistics—regardless of training and occupation or job title—have an obligation to work in a professional, competent, respectful, and ethical manner.
- ACM: Computing professionals' actions change the world. To act responsibly, they should reflect upon the wider impacts of their work, consistently supporting the public good.
- RSS/IfoA: Data science can be both beneficial and detrimental to individuals and/or society ...members may seek to understand the impact of their work. (Nov 19)

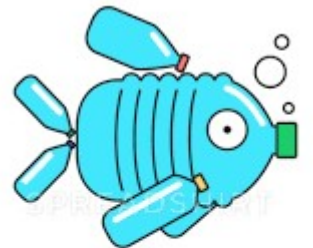
## Agree or disagree?

**“Currently among those that apply data science there are too many that lack knowledge on data science and on ethics to act responsibly”**

- Deficits in data science education on data science (esp. statistics)

*“Knowing what is to be done by understanding the rational, epistemological foundations of professional action”*

- Data scientists trained in 3 month courses
- No „Science about data science“ at all for many
- Statisticians SaDS on World → Data → Knowledge
- Machine Learners SaDS covers Data → Representation → Knowledge → Representation
- Engineers SaDS Algorithm? → World?
- Deficits in data science education on ethics
  - No training in ethics or ethical reasoning, (moral) debate, politics,...

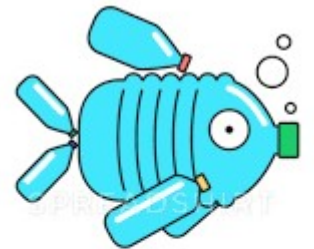


Lack of  
knowledge

## Agree or disagree?

**“Data scientists need to become better in communicating with lay people on data science and related ethical issues to act responsibly”**

- Those that have all the knowledge are still limited in their effectiveness to take responsibility if they can not talk with lay people about it.
  - Talking about methods and methodology with lay people
    - Education in general consulting skills is still neglected in academic training
  - Discuss ethics and own moral views related to data science
    - Data scientists have to overcome shyness and ignorance

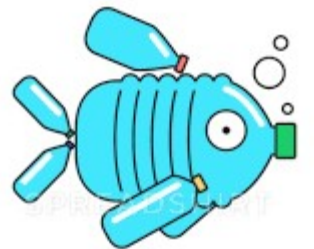


Lack of Skills

## Agree or disagree?

**“Currently, the community of those that want to practice responsible data science lack power to fight irresponsible data science”**

- There are very many companies that sell expensive and less than useful data science services.
- Such poor data science **ruins** the reputation of data science and **ruins** companies and institutions.
- Data science sharlatanes also do harm to society! (No only outside academia!)
- In companies, departments for data science, analytics, statistics, etc. fight against each other for funding.
- In the academic world ML and statisticians often fight fruitless fights about who is fundamentally wrong or right.
- **Together we are strong!**

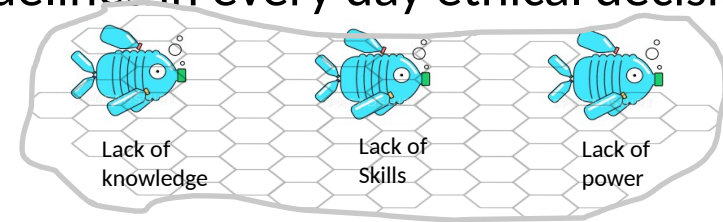


Lack of power

## Agree or disagree?

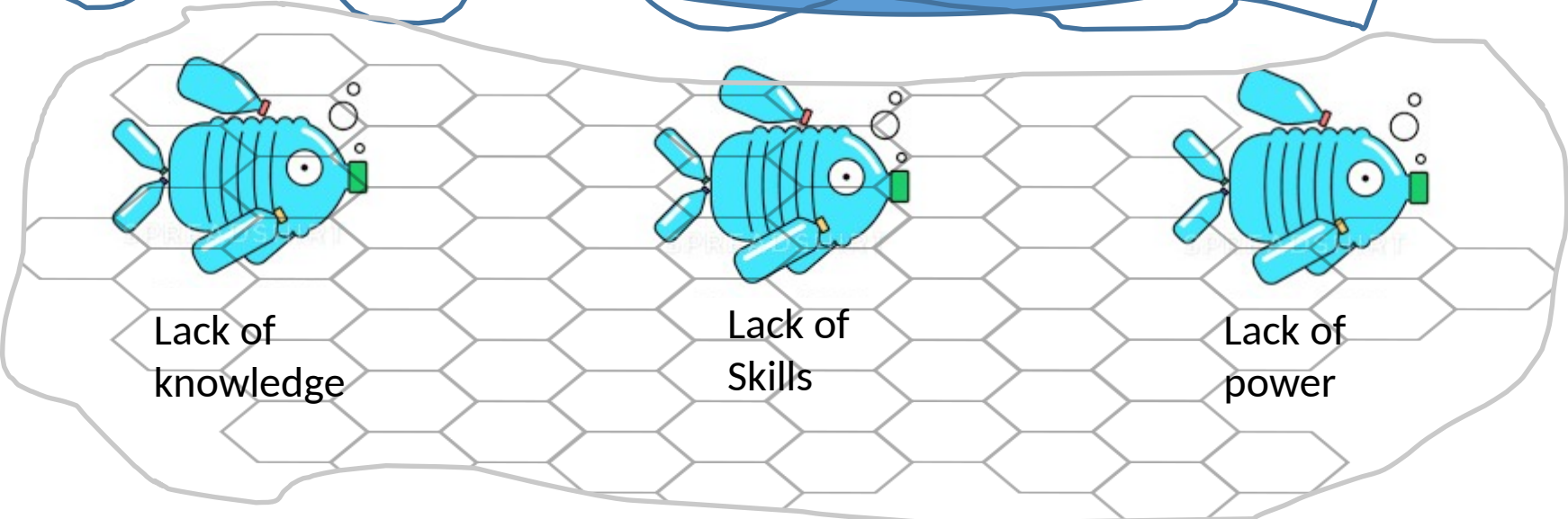
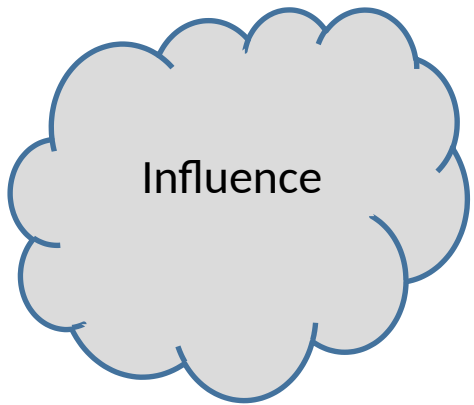
**“Lack of knowledge, skills, and power can be addressed by data science as a profession with professional ethics”**

- Support the individual data scientist by education and guidelines in every day ethical decision making .



- No need to re-invent the wheel for the individual.

- Written rules of conduct for data science services help to establish a relationship of trust between data scientists, their clients, their employers, and society.
- Being trusted as a profession increases social status, reputation and power for a professional.
- In case of conflicts of interests an ethical guideline under the maintainership of some professional society may offer an arbitration process.







## **Agree with any of those reasons for lack of interest in professional ethics?**

- Professional societies with memberships, codes of conducts asf are elitarian and limit scientific freedom and hinder innovation.
- What is right and wrong is defined by law.
- What is right and wrong is my personal belief.
- Corporate responsibility is more important than professional ethics, as those who implement algorithms are not responsible for their societal impact.
- Paper does not blush, and most code of conducts are hypocritical.
- My opinion: All those are cheap invalid excuses!

# Help establishing the debate!

*Data science is in the focal point of current societal development.*

*Without becoming a profession with professional ethics,*

*data science will fail in building trust*

*in its interaction with*

*and its much needed contributions to*

*society!*

Thank you!