

Algorithms find solutions, so what's the problem?

Mathijs de Weerd, June 7, 2023





But we aren't using algorithms for finding the solution to some very relevant problems.



Outline

1. What is an algorithm?
2. What is the problem?
3. How can we address this?



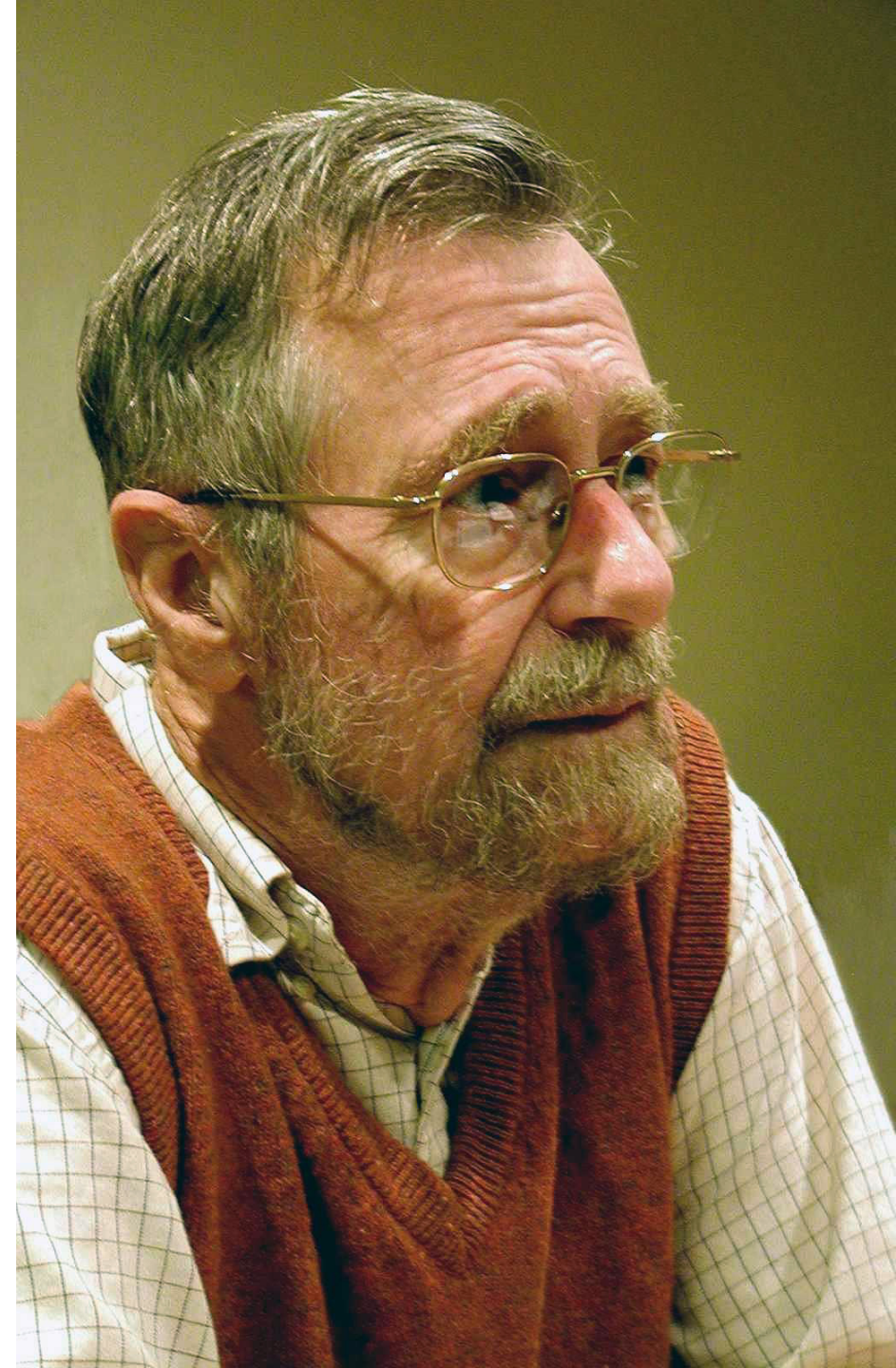
efficient

**algorithm = recipe to be executed by a
computer**

Dijkstra's algorithm (1959)

““The question of whether a computer can think is no more interesting than the question of whether a submarine can swim.”

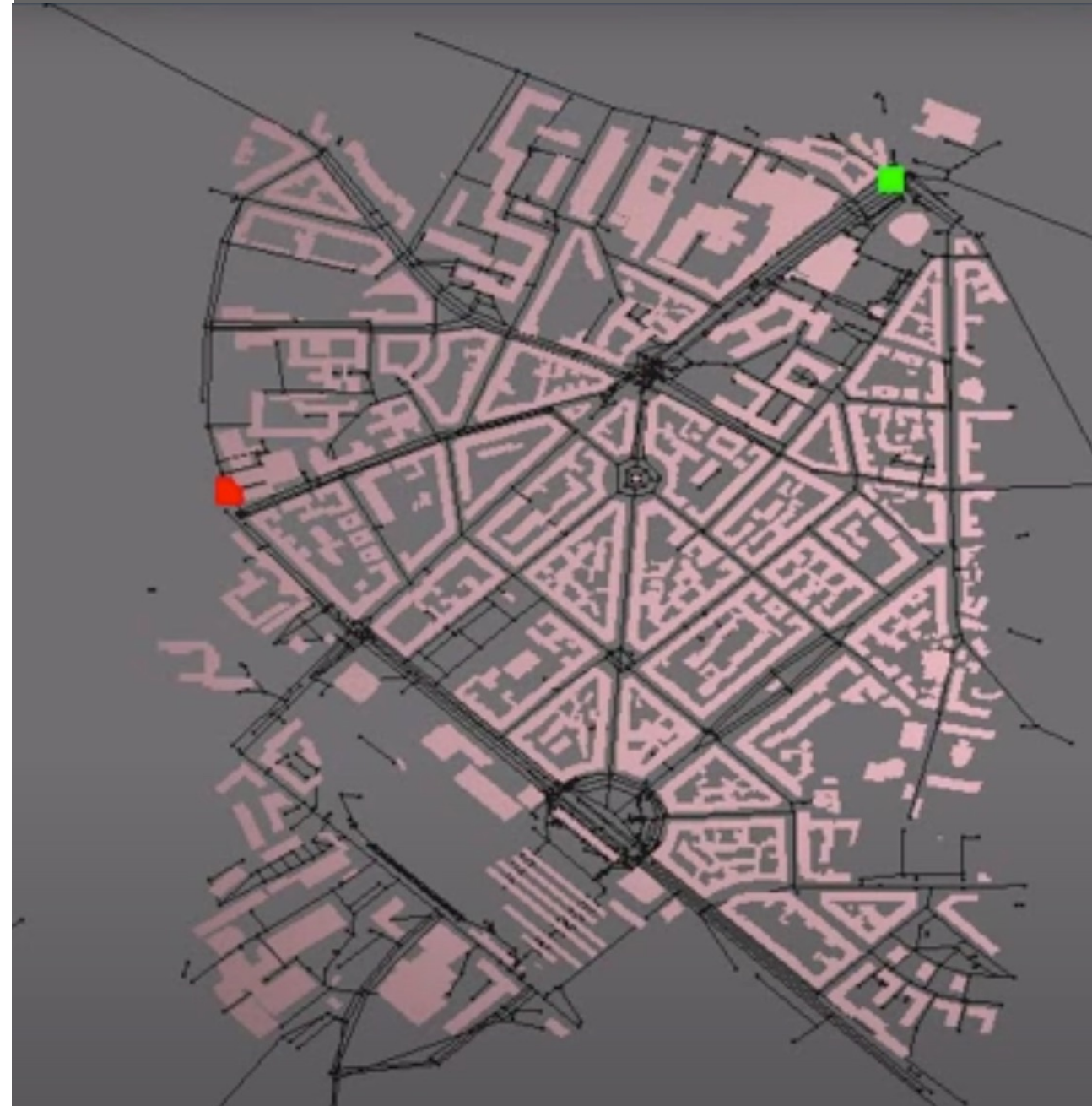
— Edsger W. Dijkstra




```

function DIJKSTRA( $V, E$ )
   $d[v] \leftarrow \infty$  for all  $v \in V$ 
   $d[s] \leftarrow 0$  for start vertex
  while  $V$  is not empty do
     $m \leftarrow \arg \min_{v \in V} d[v]$ 
    Remove  $m$  from  $V$ 
    if  $d[m] = \infty$  then
      return  $\infty$ 
    for every  $e \leftarrow (m, u) \in E$  do
      if  $d[m] + w(e) < d[u]$  then
         $d[u] \leftarrow d[m] + w(e)$ 
  return  $d$ 

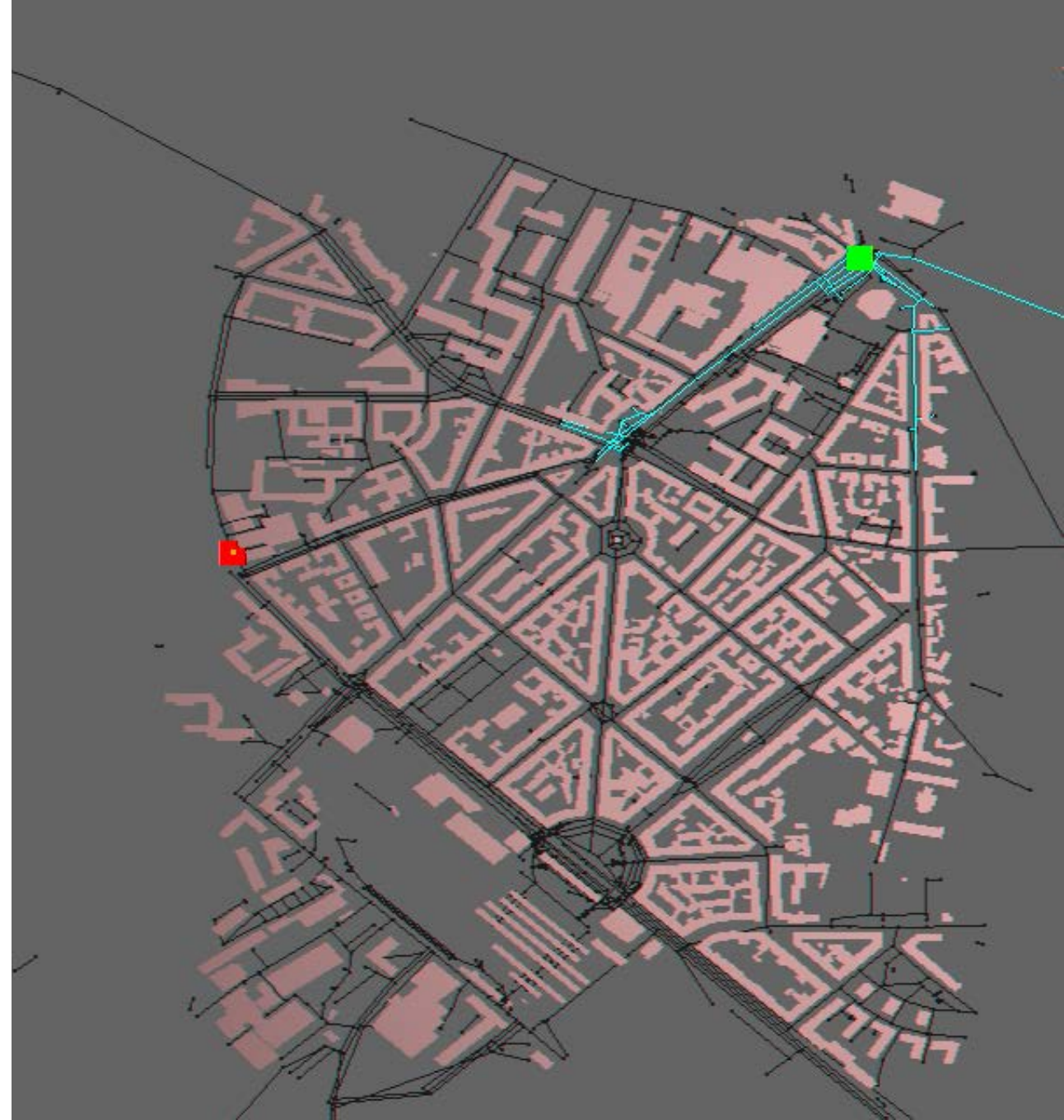
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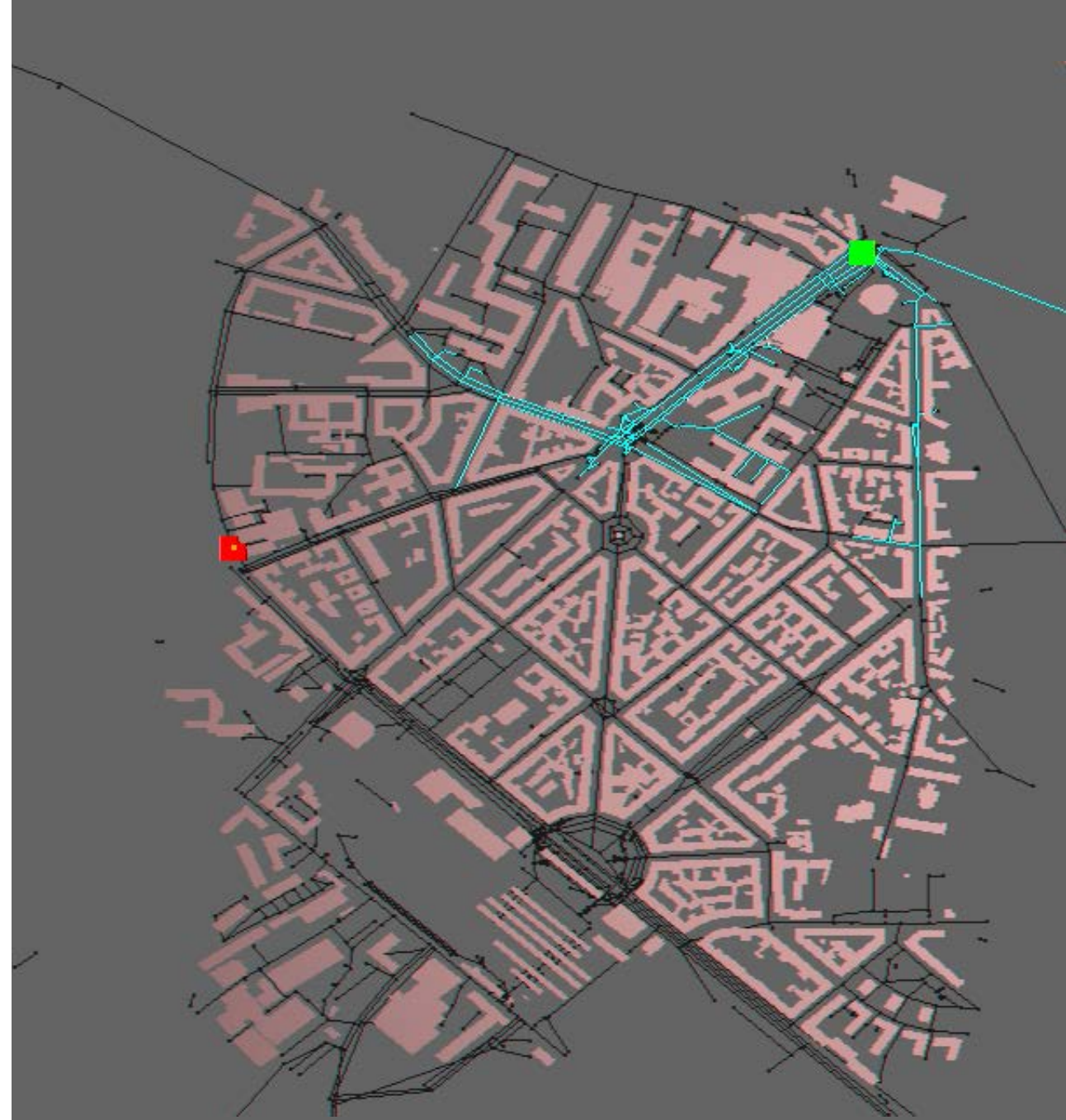


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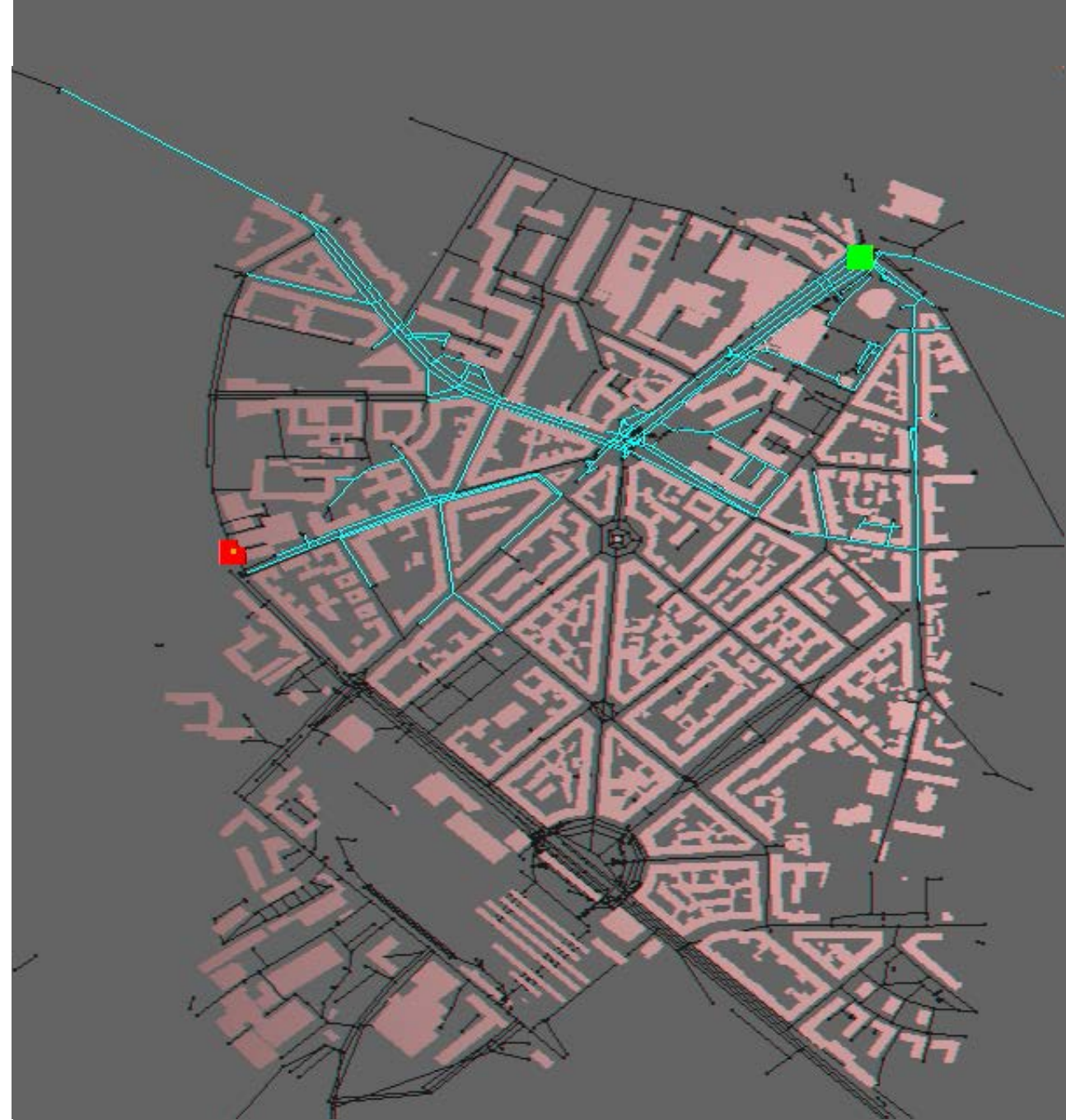
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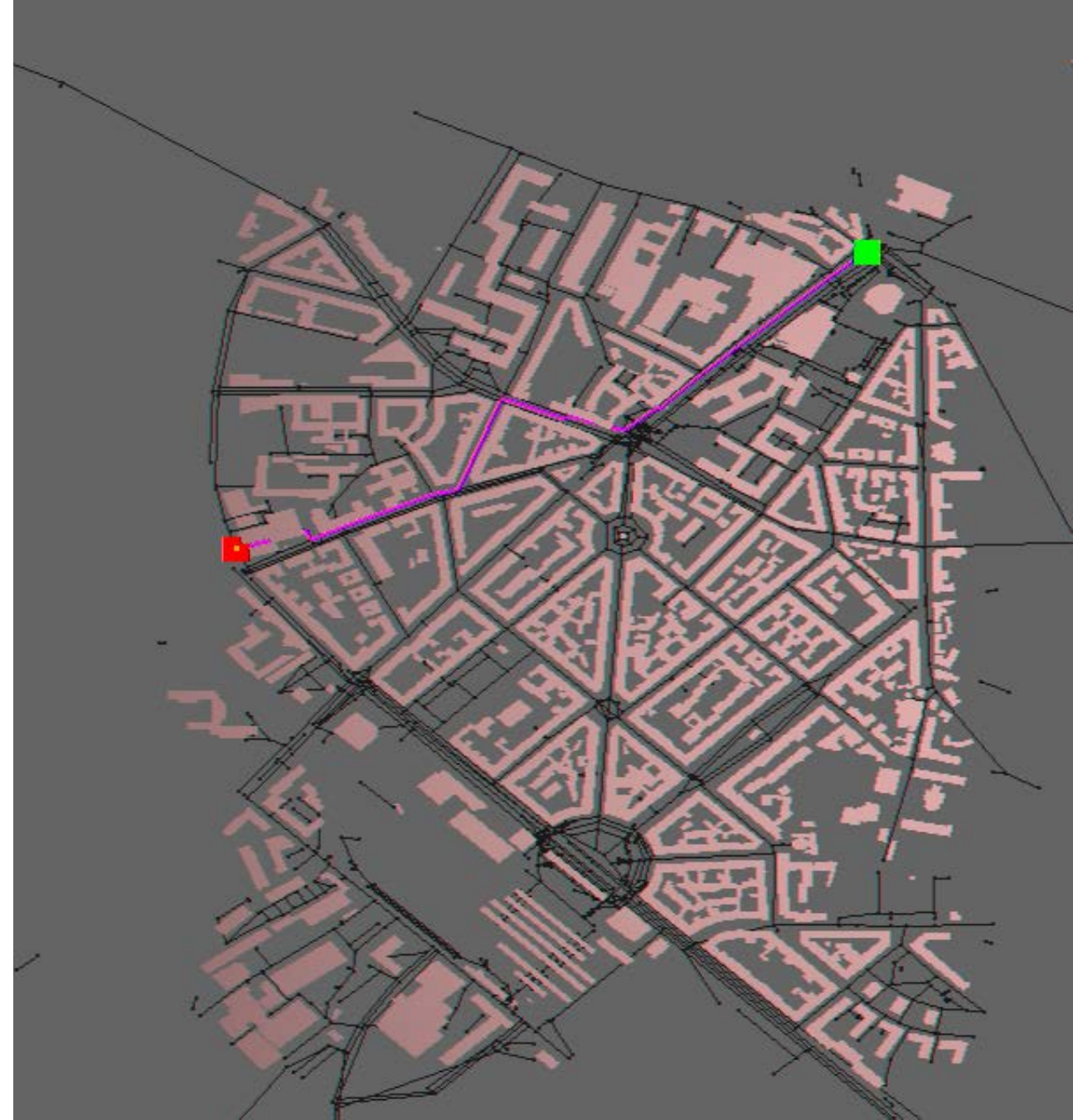
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```



Data = ingredients


- structured (e.g. a map)
- unstructured:
 - patterns trained and used (machine learning)
 - collected by the algorithm itself (reinforcement learning)



An example with a *lot of* data: large language models

e.g. ChatGPT: GPT-4 was trained on 570 gigabytes of data

570 000 000 000 bytes
(characters)



thick book

An example with a *lot of* data: large language models

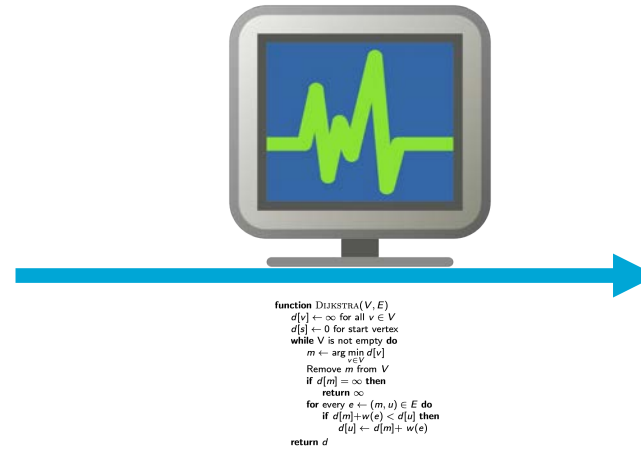
“all” data on the internet estimated to be 5 billion gigabytes (by former CEO of Google, Eric Schmidt, interview in 2023)

5 000 000 000 000 000 000 000 bytes
(characters)

thick book

movie

The quality of the ingredients counts!



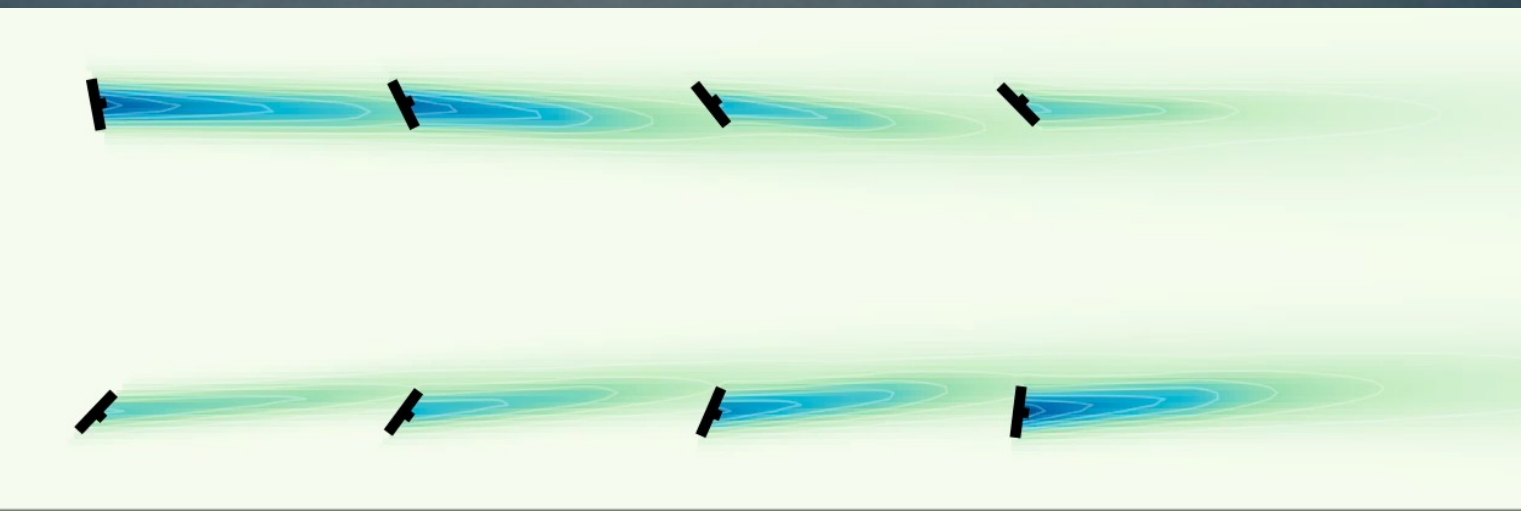
Wind farm wake control



Greg Neustroev

<https://github.com/AlgTUDelft/wind-farm-env>

Greg Neustroev, S. Andringa, R. Verzijlbergh, MdW (2022), **Deep Reinforcement Learning for Active Wake Control**, In *AAMAS 2022* p.944-953



Algorithms should be used for problems that matter!

Many of these are **planning or scheduling** problems!

- train shunting/path finding
- energy system investments
- heat network control
- efficient (bio)manufacturing

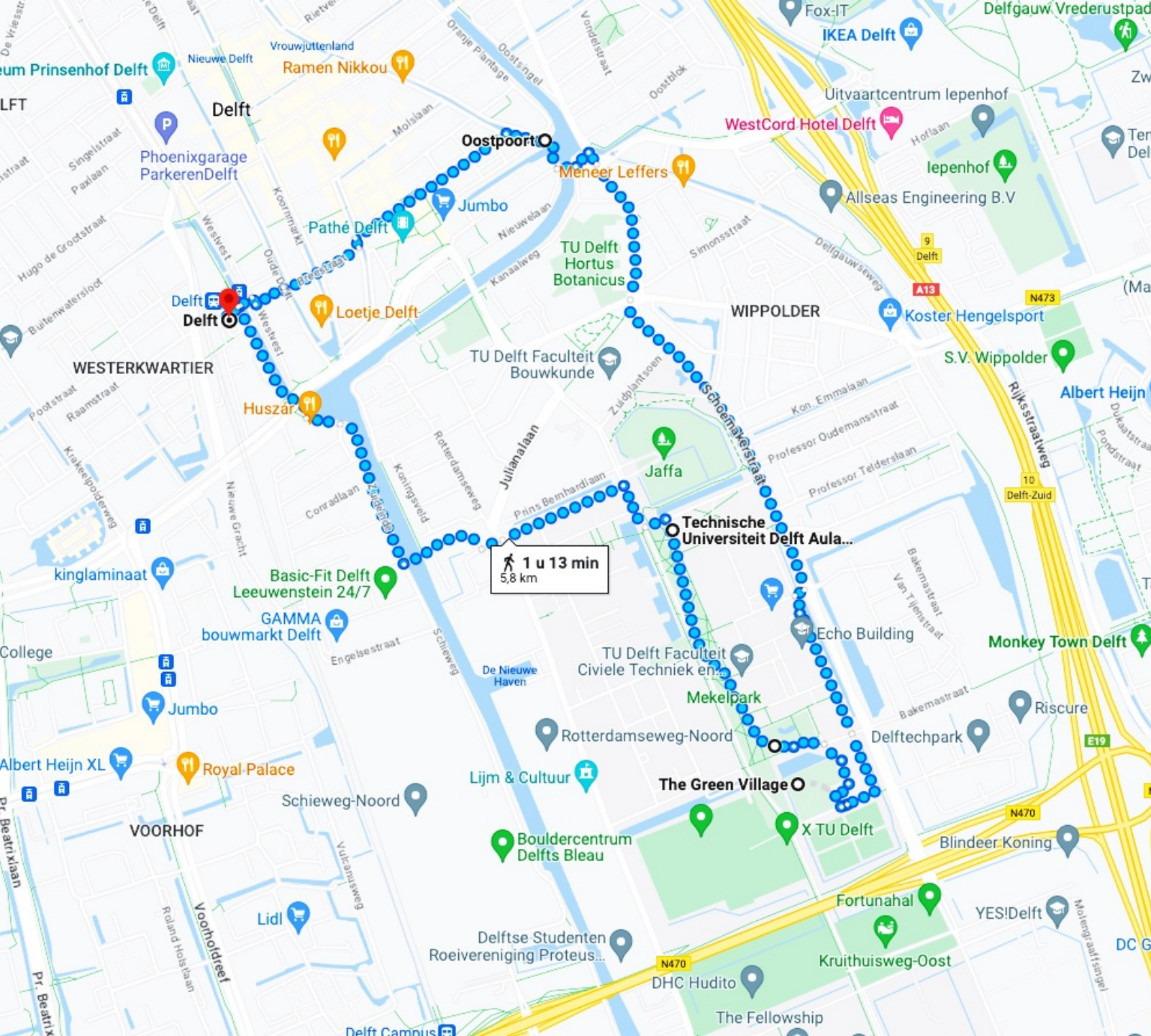


Planning & Scheduling Methods in Practice

Surprisingly little use of such algorithmic insights in practice!

- “the majority of derived mathematical solutions have not been used in practice by manufacturing businesses” (Jackson et al., 2004)
- “**Spreadsheet application** still dominates enterprise resource planning and advanced planning systems” (De Man, Strandhagen, 2018)
- “most schedulers prefer the use of a simulation-based software or **manual decision**, which result to suboptimal solutions” (Georgiadis et al., 2019)

So what's the problem?



Traveling Salesman

find the shortest route via
a number of locations



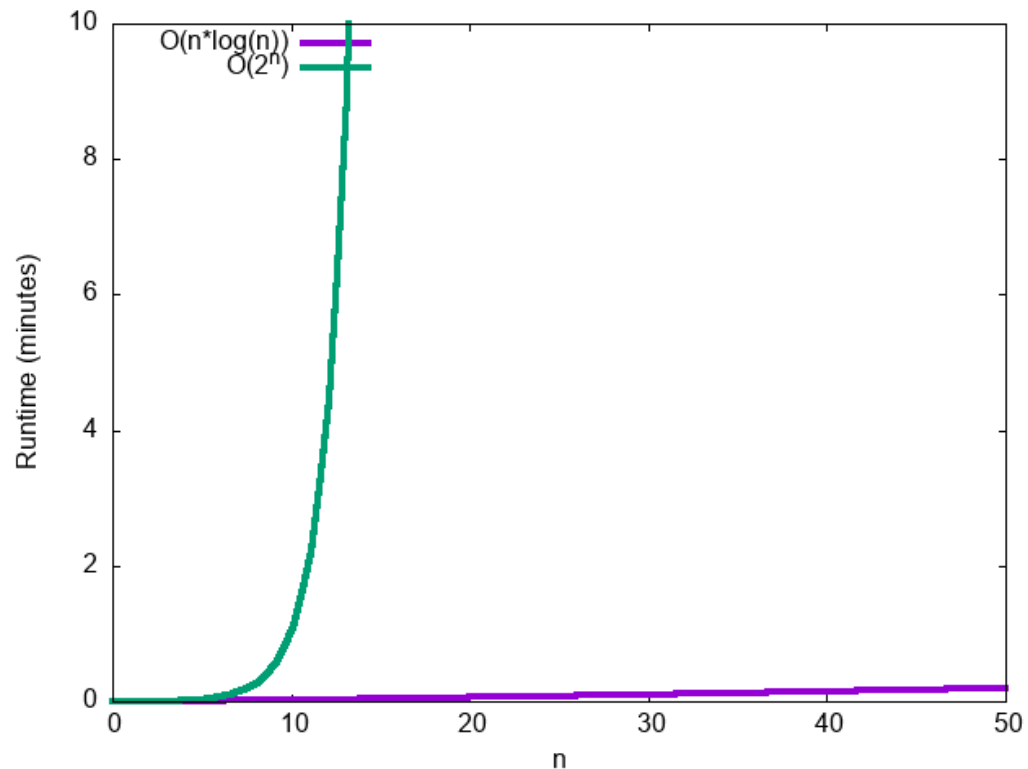
Traveling Salesman Package delivery

find the shortest route via
a number of locations


Problem complexity

Time needed for the computation
(runtime)

Input size of the problem: n



$$2^n$$



18 446 744 073 709 552 000
5 000 000 000 000 000 000 000

P (easy)

P is the class of problems that can be solved in polynomial time.

Examples

- Shortest path
- Scheduling with unit processing times

NP (-hard)

NP-hard are those problems for which we have no polynomial-time algorithms.

Examples

- Traveling salesman problem
- Scheduling with arbitrary processing times, release times and deadlines



So what now?

- algorithmic innovations (possibly not optimal), or
- don't solve the whole problem...

Choosing the right model:



Choosing the right model requires...

algorithmic understanding

1. what is computationally challenging?
2. which algorithm to use?

and expertise on the domain

1. what's important for (in)feasibility?
2. what can we fix later?

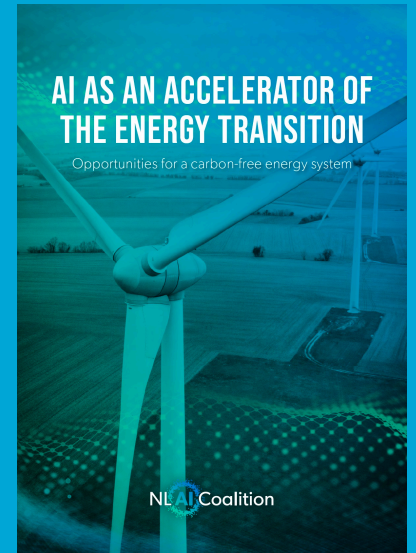
How to address this?

In my own research



Supporting others

EnergySHR



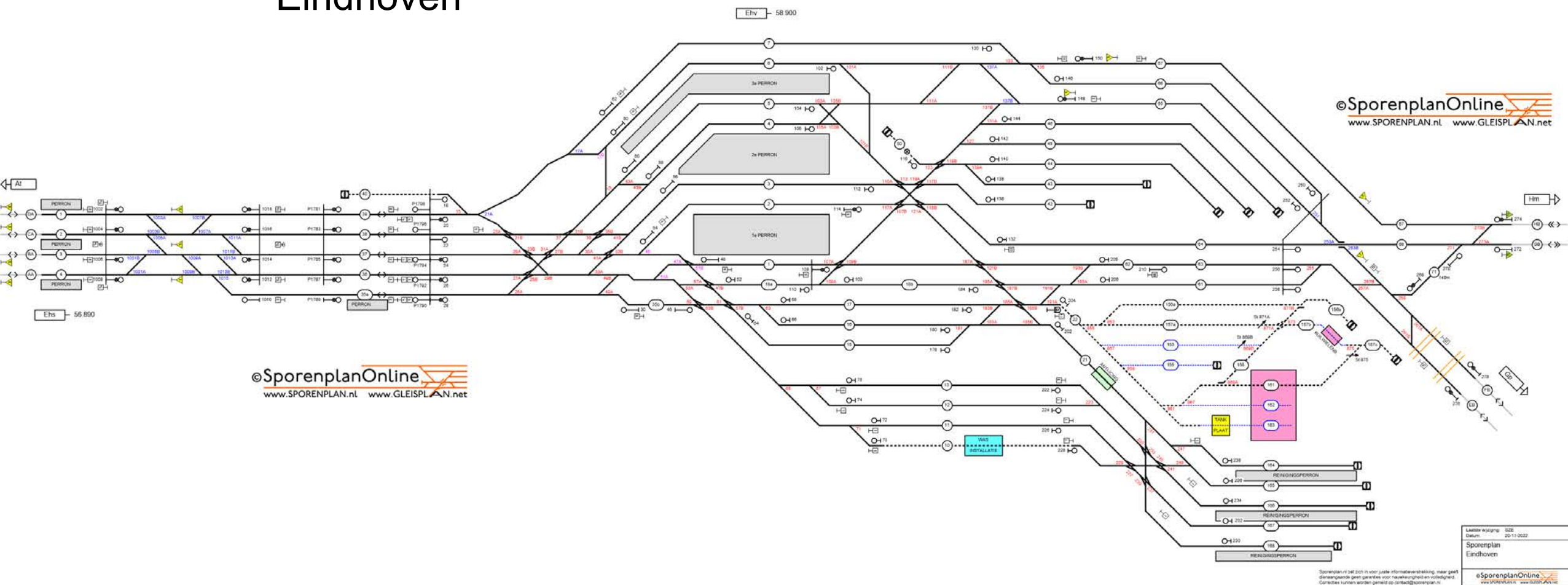
Shunting at railway hubs



<https://nieuws.ns.nl/media-archief/>



Eindhoven





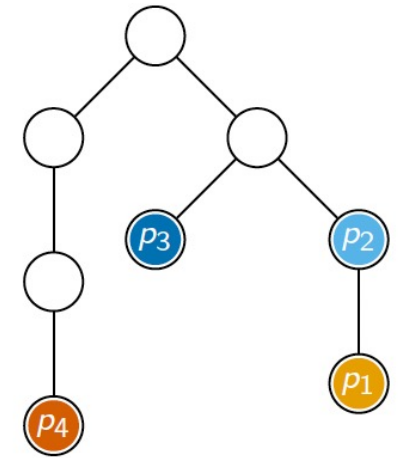
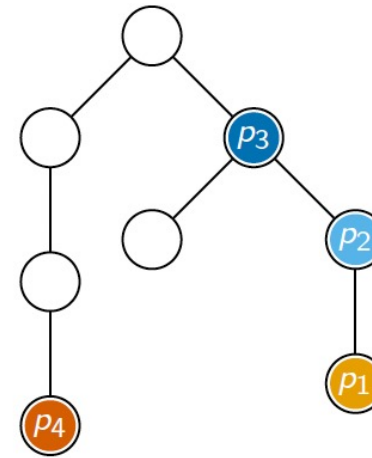
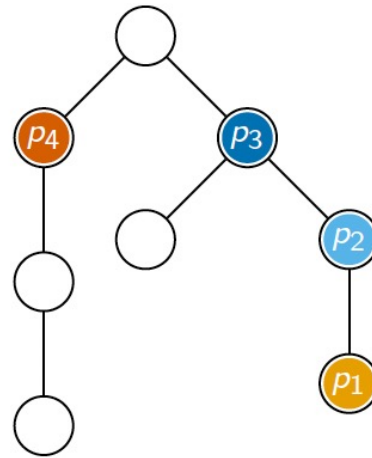
<https://commons.wikimedia.org/wiki/File:15-Puzzle.jpg>



Using a more simple model can help!



Issa Hanou

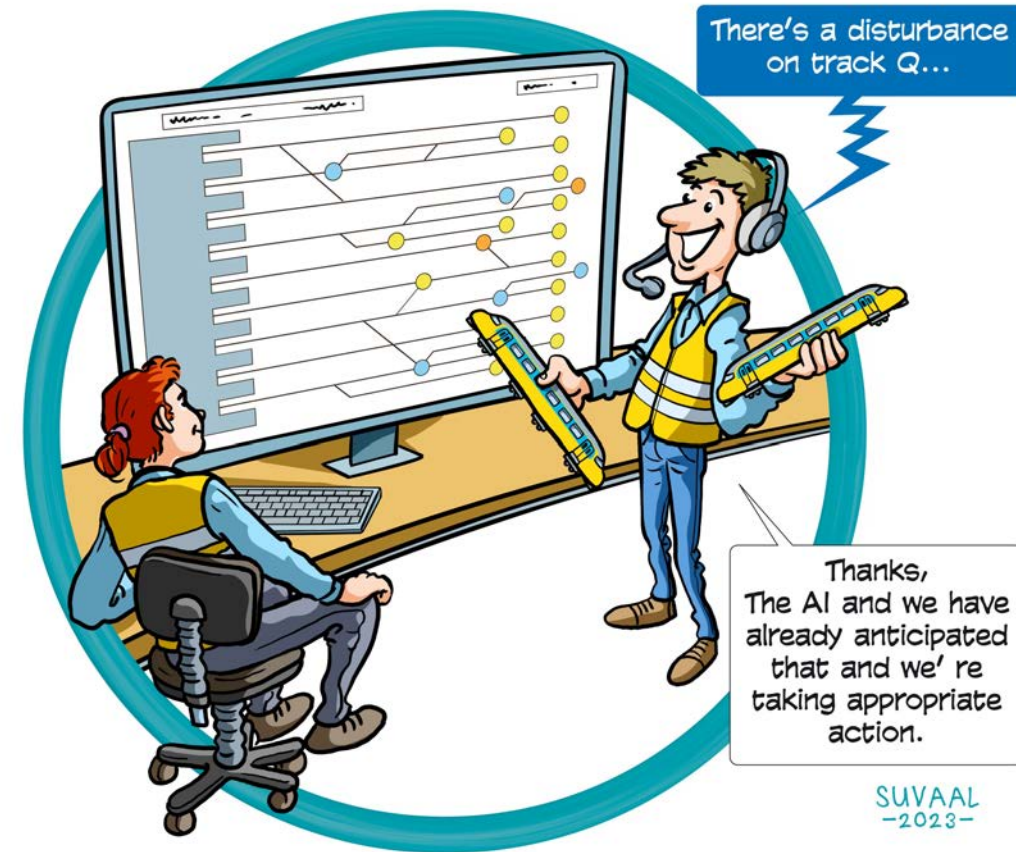


Example: RAIL lab – LPT Robust (2023-2027)

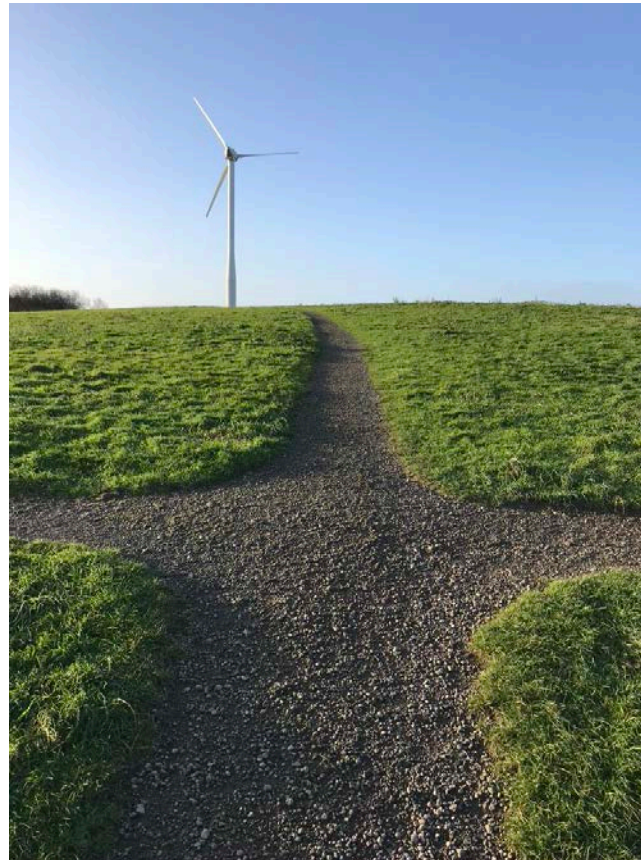
ICAI Lab, Scientific directors: Marjan van den Akker (UU), Mathijs de Weerd (TU Delft)

Five PhD projects:

1. Cooperation between human and AI planners
2. Robust planning
3. Quickly reacting to changes and disruptions
4. Supporting strategic decisions regarding the infrastructure capacity
5. Learning from previous situations and produce recognizable plans



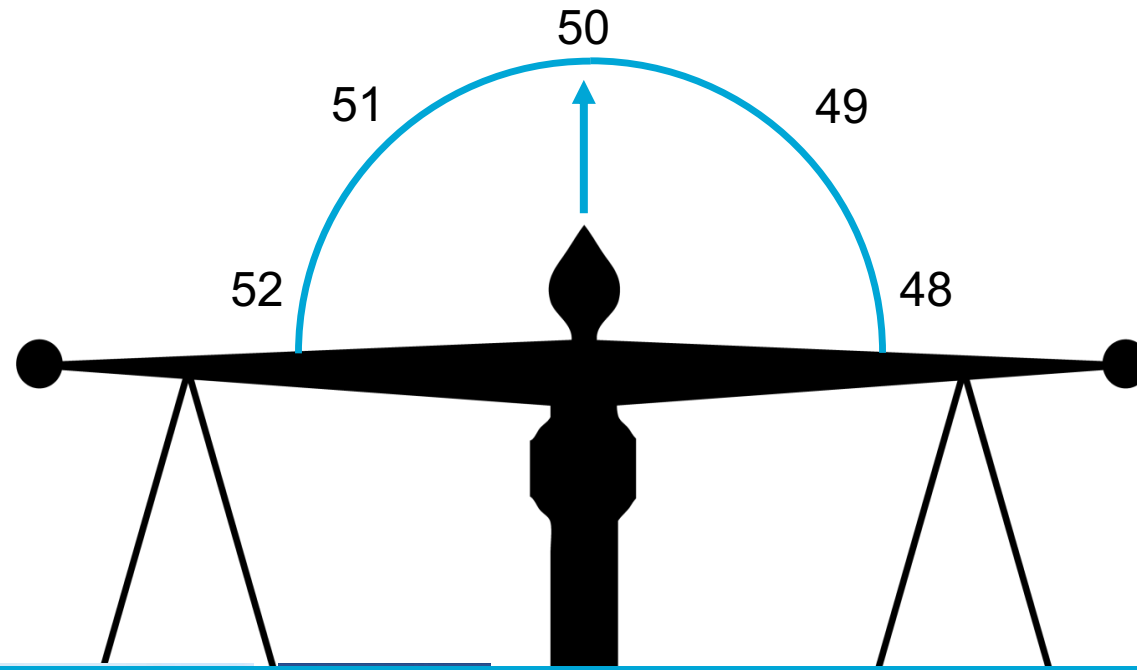
CVIII Ontwerpers



- not completely controllable,
- variable,
- not perfectly predictable

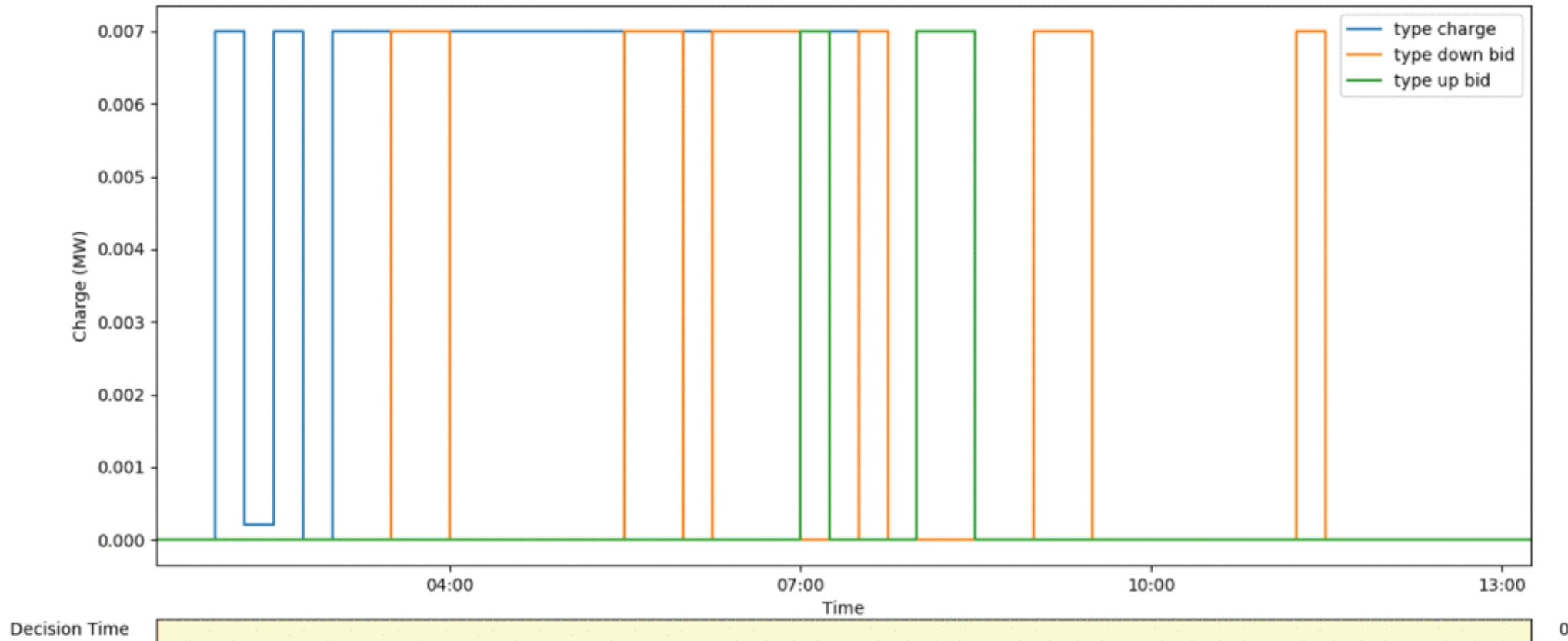
supply

demand



- need for energy storage
- demand management
- more active role for end users
- new business models and markets

Scheduling what to bid & when to charge

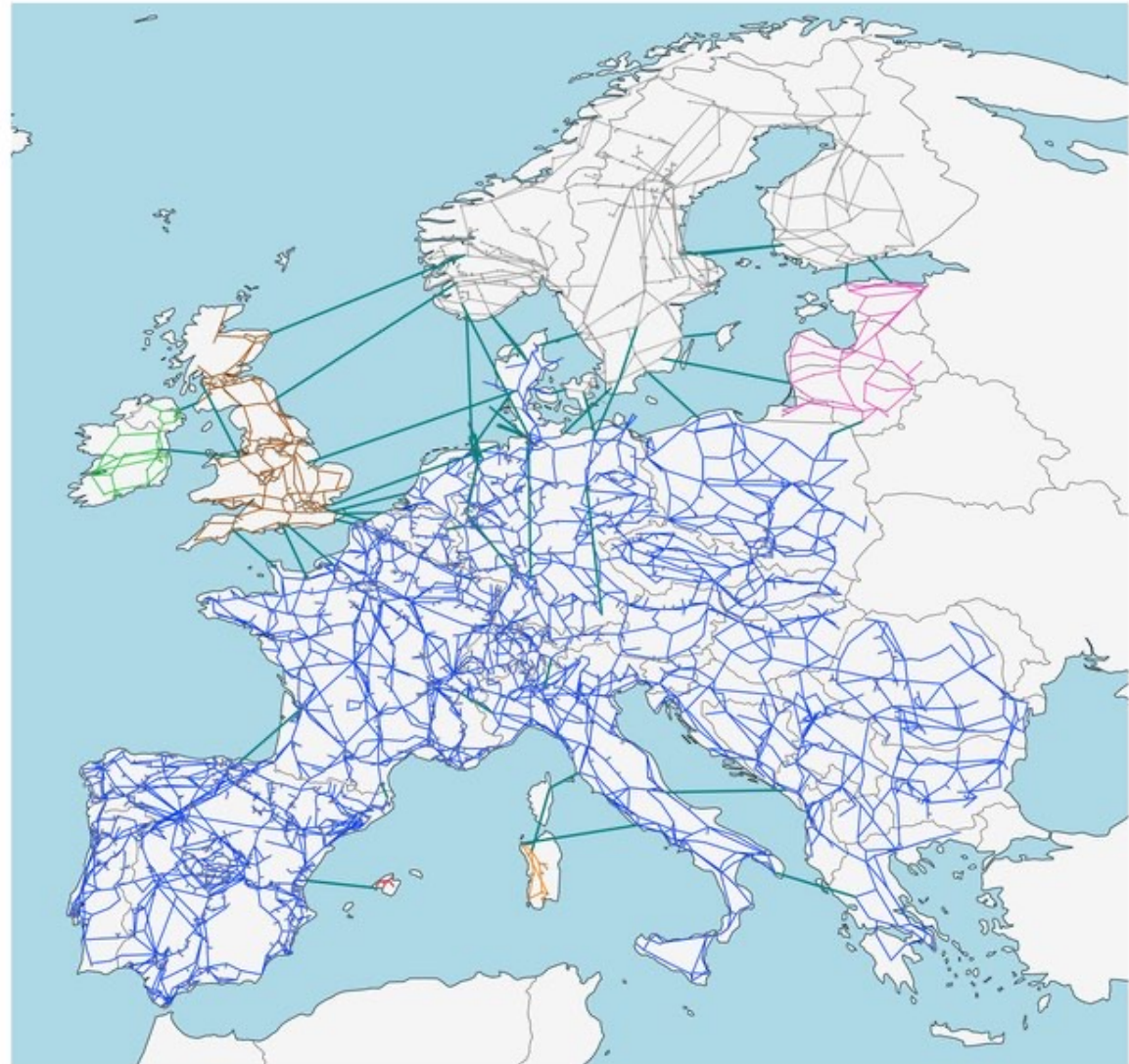


Koos
van der Linden

<https://github.com/AlgTUDelft/B-FELSA>

Optimal Investments in and Operation of the Energy System

- whole of Europe
- 20+ million constraints
- no optimal solution after a week
- just for the power system...





NextGenOpt (2023-2027)

Next Generation Sector-Coupling Models for Optimal Investments and Operation

Work in a team with algorithmic & domain expertise

Core aims:

- also **use algorithms to find models** with a good trade-off between accuracy and computation
- and to allow **making trade-offs interactively with the user**



Maaïke Elgersma



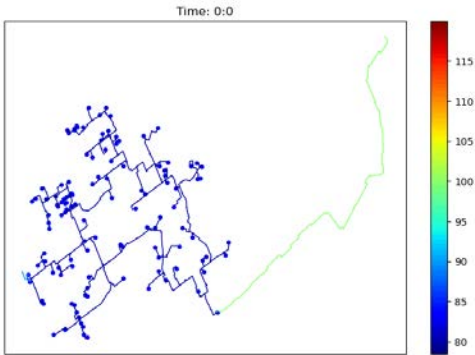
Greg Neustroev



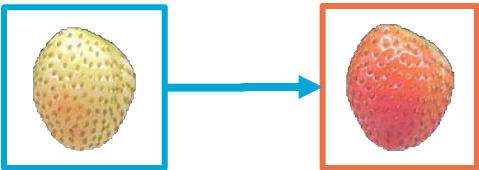
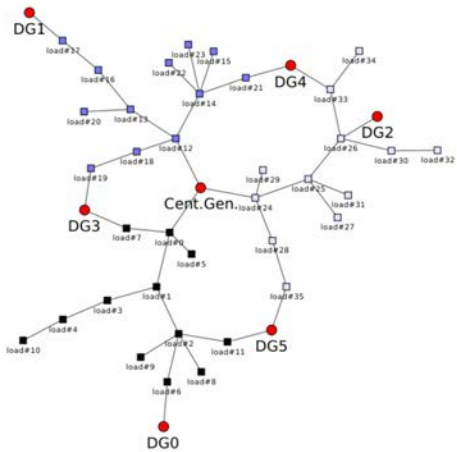
PBL Netherlands Environmental
Assessment Agency

Other projects

<https://github.com/ftbv/grid-penguin>



by Jichen Wu



Flex
Technologies



Kim van der Houten



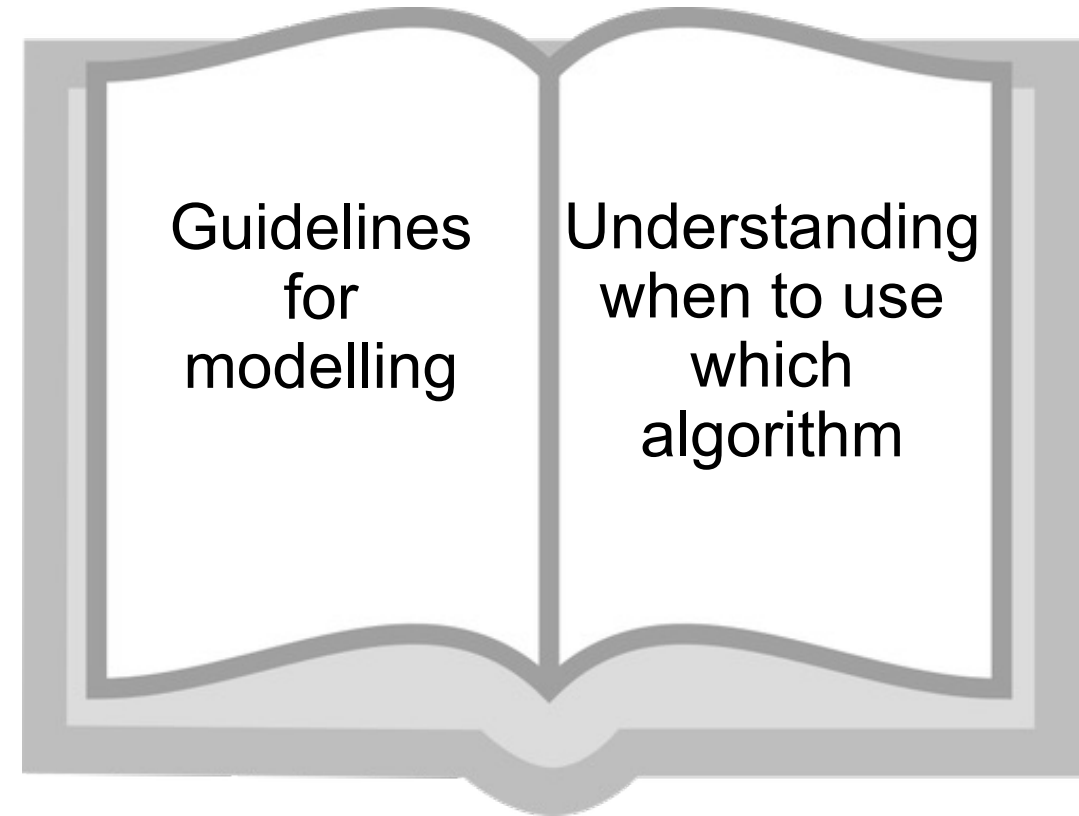
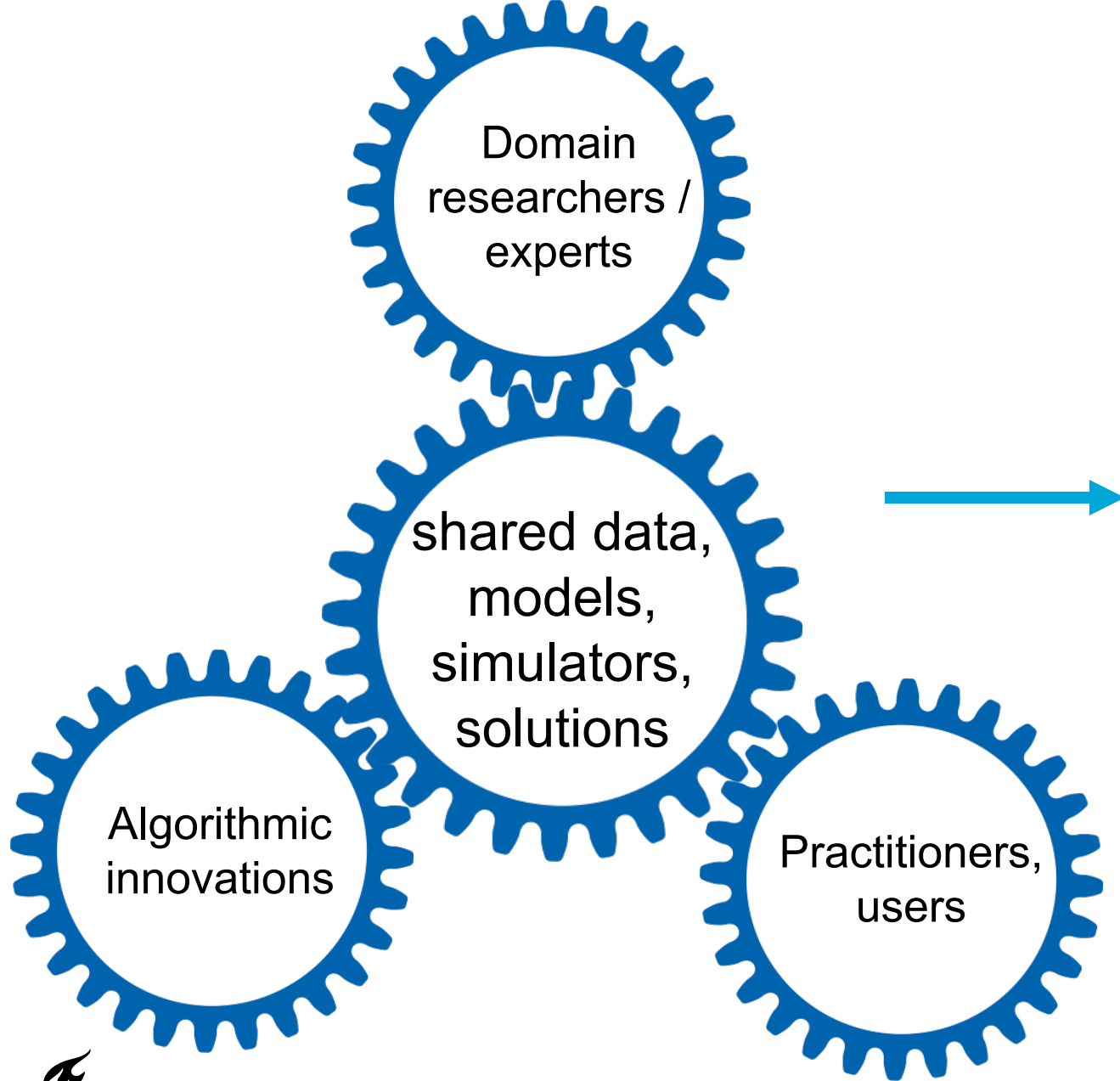
Ksenija Stepanovic



Grigorii Vevyurko



Junhan Wen



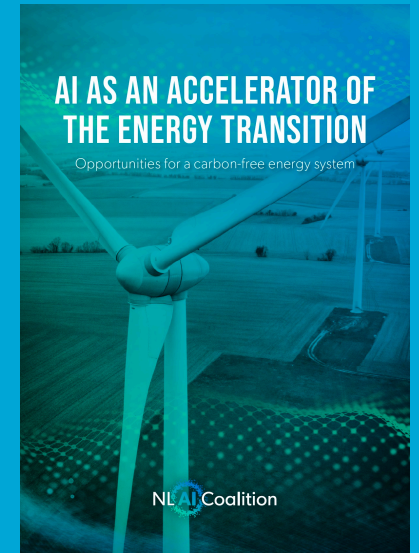
How to address this?

In my own research



Supporting others

EnergySHR



co-chaired with Yashar Ghiassi-Farrokhfal (RSM)

EnergySHR

Sharing data and code by and for
researchers

- For practitioners: to benefit more directly from new algorithms
- For researchers: working with real data; solving the right problems

Contact me (or Caroline Duterloo)!



Inspired by The Green Village: a living lab
for sustainability and energy research

NL AIC

- transfer of knowledge
<https://energie.ai-cursus.nl/home>
- bringing together problem owners and AI expertise (researchers & developers)

Contact me if you're looking for AI expertise!

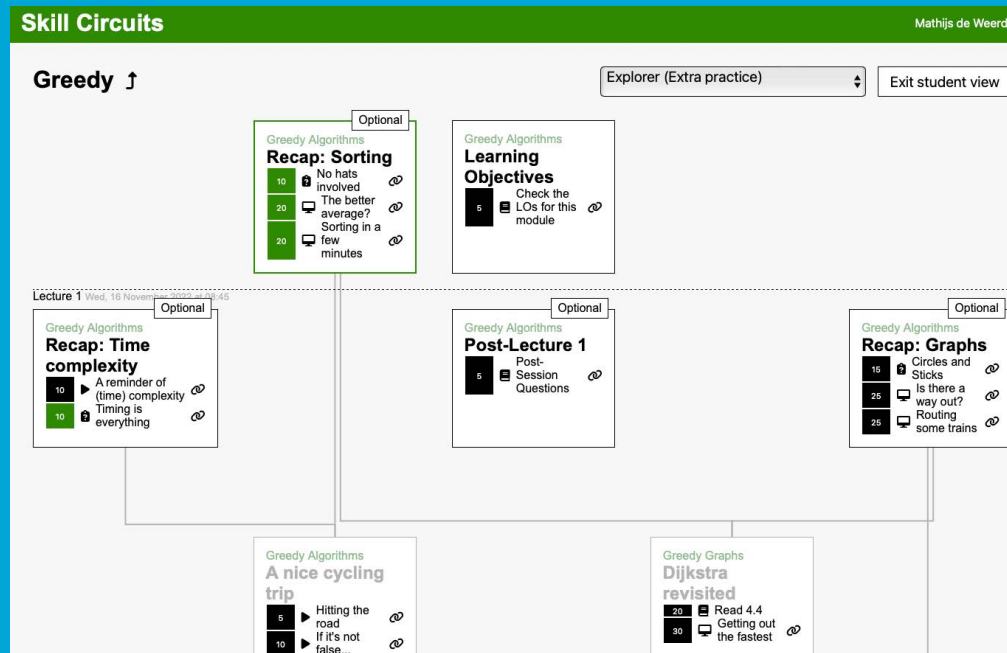
AI AS AN ACCELERATOR OF THE ENERGY TRANSITION

Opportunities for a carbon-free energy system

Education

Students are all different

Teach bachelor students about research



More than 250 papers in less than 2 months

NEWS - 23 JULY 2021

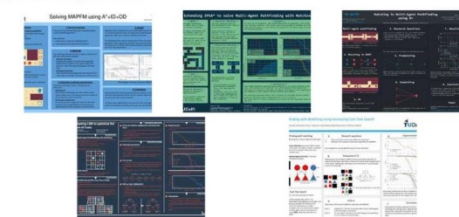
Vibrant groups full of creativity and interesting new research, but above all: a new generation of bachelor students ready for all the possible academic hurdles they will face during their master's. The new "Research Project" of the renewed Computer Science and Engineering curriculum is a great success! But what made that possible?

A big change

You don't change a successful curriculum just like that. 'We thought about it very carefully,' says Mathijs de Weerd, 'because a lot of what we were doing before also worked well'. Until a year ago, the bachelor students of Computer Science Engineering did their graduation research in groups of four at a company, to gain practical experience. 'That was an incredibly valuable experience and we didn't really want to give it up. But ultimately we also want to train scientists.' The group assignment at a company was therefore retained, but moved to the second year. And indeed, that went very well. In their third and final year of the bachelor's, the students now do a graduation assignment, in which they set up and complete a high-quality research project in two months, including a paper. 'That is quite a task, and it was a little exciting whether everyone would succeed.'

But they did: the 250 studies are full of interesting results, from programmes that enable a computer to defend itself better in a game of AI football to algorithms that automatically analyse web comics. 'Supervisors, including professors, were more than positively surprised by the quality', says Mathijs enthusiastically, 'some are even working on getting their research published.'

Matching in Multi-Agent Path Finding
Supervisors: J. Mulderij, M.M. de Weerd



Are you curious about all the results? Then click [here](#).

An unexpected puzzle

For Mathijs and his team there was also a major task: how do you ensure that all those students set up a good research project

Greedy ↗

Explorer (Extra practice)

Exit student view

Optional

Greedy Algorithms

Recap: Sorting

- 10 No hats involved
- 20 The better average?
- 20 Sorting in a few minutes

Greedy Algorithms

Learning Objectives

- 5 Check the LOs for this module

Lecture 1 Wed, 16 November 2022 at 08:45

Optional

Greedy Algorithms

Recap: Time complexity

- 10 A reminder of (time) complexity
- 10 Timing is everything

Optional

Greedy Algorithms

Post-Lecture 1

- 5 Post-Session Questions

Optional

Greedy Algorithms

Recap: Graphs

- 15 Circles and Sticks
- 25 Is there a way out?
- 25 Routing some trains

Greedy Algorithms

A nice cycling trip

- 5 Hitting the road
- 10 If it's not false...

Greedy Graphs

Dijkstra revisited

- 20 Read 4.4
- 30 Getting out the fastest

with
Stefan Hugtenburg
Ivo van Kreveld

<https://skills.ewi.tudelft.nl/>

Education: the bachelor research project

Challenge:

- experience research
- write a paper individually
- over 300 students per year, 60-90 projects

Next:

- connect CS students (more) to other disciplines
- allow other students to learn (more) about algorithms

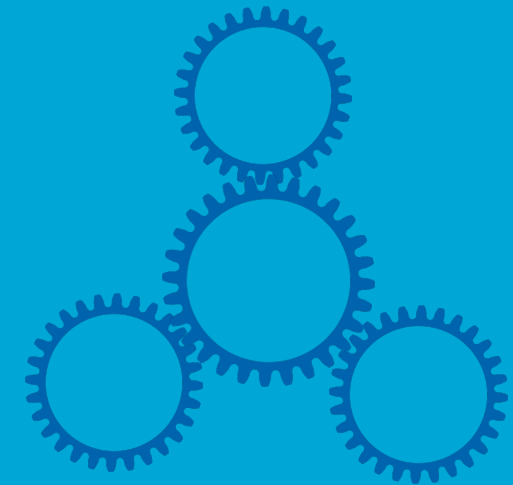
| Project Popularity | | | | |
|---|--------------|----------|----------|-----------|
| Project | Popularity ? | In top 3 | In top 5 | In top 10 |
| 2582 - Cooperative AI for Overcooked! | 10.65 | 51 | 72 | 124 |
| 2440 - A Virtual Reality Game to Explore Hyperbolic Geometry | 11.98 | 38 | 56 | 84 |
| 2467 - Find biological markers for human disease and allergies | 11.77 | 37 | 63 | 87 |
| 2545 - Advances and Challenges in Traffic Forecasting with Deep Learning... | 12.62 | 31 | 41 | 67 |
| 2471 - Person identification using heart rate and activity from consumer... | 12.0 | 30 | 50 | 88 |
| 2439 - Comics Illustration Synthesizer using Deep Generative Models | 12.98 | 28 | 35 | 58 |
| 2466 - "We need to learn how to teach Machine Learning" | 12.67 | 28 | 45 | 65 |
| 2485 - Shining a light on Material Appearance | 12.78 | 26 | 39 | 63 |
| 2518 - Mining Software Testing Knowledge | 12.74 | 23 | 38 | 67 |
| 2524 - Combinatorial Optimisation for Scheduling (NP-Hard Algorithms) | 12.8 | 23 | 39 | 62 |
| 2539 - Revealing the Secret to Successful Virtual Meetings: How Personal... | 12.99 | 21 | 32 | 61 |
| 2520 - Github Mining: Discover the Descriptive Metrics of the Context in... | 13.06 | 20 | 33 | 59 |
| 2433 - Creation of Gang Territories and other Patterns | 13.23 | 19 | 31 | 52 |

Conclusion

Algorithmic insights *can* help solve society's problems.

But it is important that an algorithm

1. uses the right model
2. uses the right data
3. works in interaction with users



Not alone

- My promotor: great to have you here!
- PhD students: current and former
- Post-docs, research software engineers
- Business developers, project support, etc.



Konstantin
Sidorov



Ksenija
Stepanovic



Issa Hanou



Maaïke
Elgersma



Koos
van der
Linden



Kim van der
Houten



Grigorii
Vevurko



Junhan Wen

Not alone

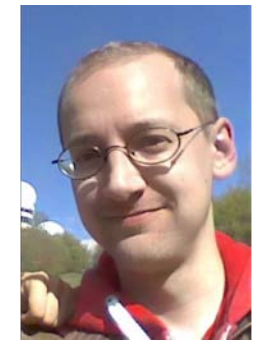
- Other staff of the section Algorithmics
- All colleagues of the two departments
- My collaborators over the years
- Industry partners
- Could not mention all in this talk
- Fantastic to work with you!



Anna Lukina
safety and
explainability of AI



Matthijs Spaan
planning under uncertainty



Wendelin Böhmer
deep RL



Peter Bosman
evolutionary intelligence



Sicco Verwer
learning interpretable models



Sebastijan Dumančić
probabilistic program synthesis



Emir Demirović
optimisation & ML



Neil Yorke-Smith
data-driven optimisation