



University of
Central Lancashire
UCLan

CO3519

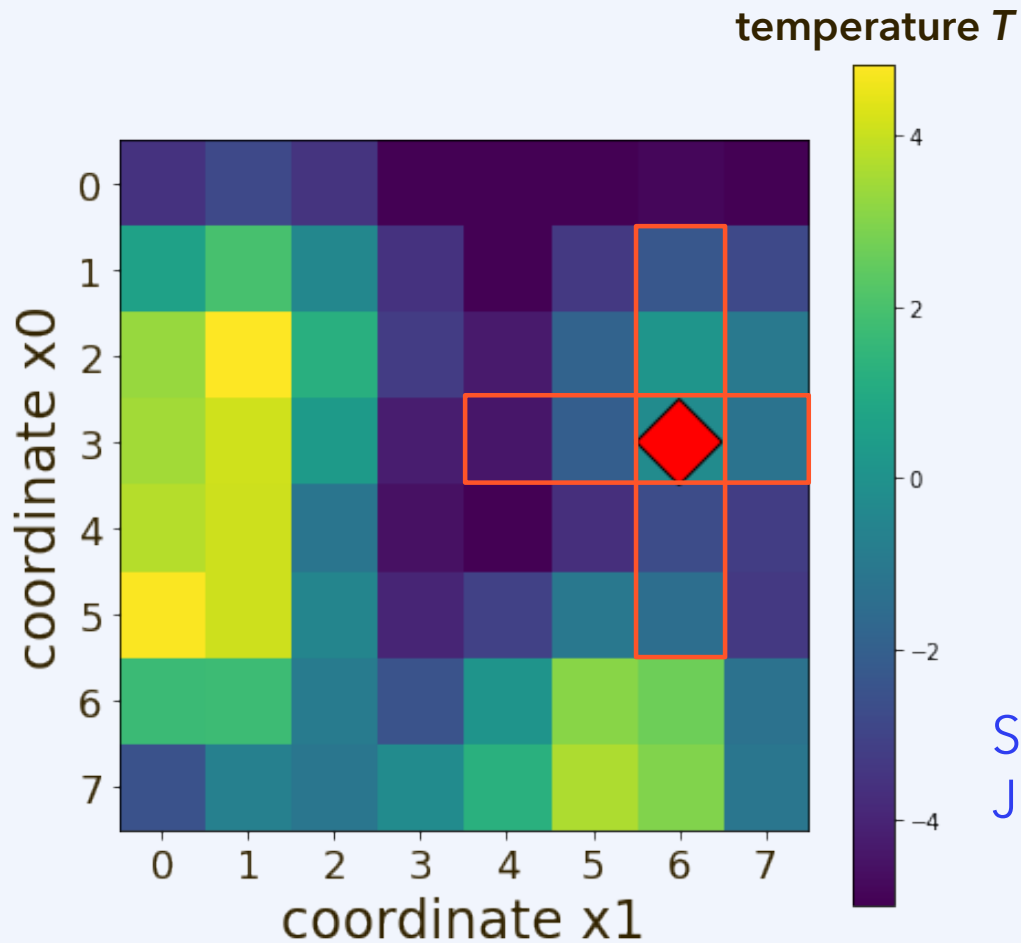
Artificial Intelligence

Tutorial 2.3 discussion
Review of parts 1 to 3
End-of-year reflection

Where opportunity creates success

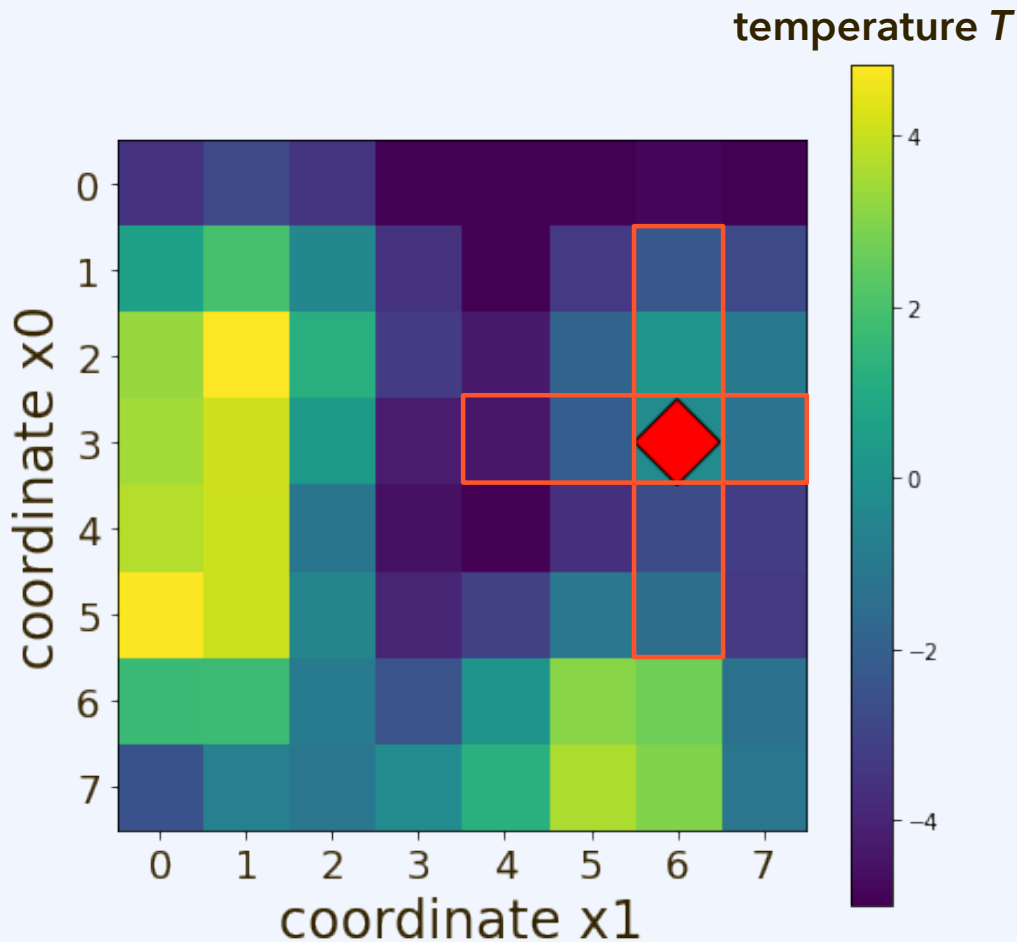
Tutorial 2.3 discussion

Agent escaping extreme temperatures



See [agent-function-evaluation](#)
Jupyter Notebook.

Agent escaping extreme temperatures

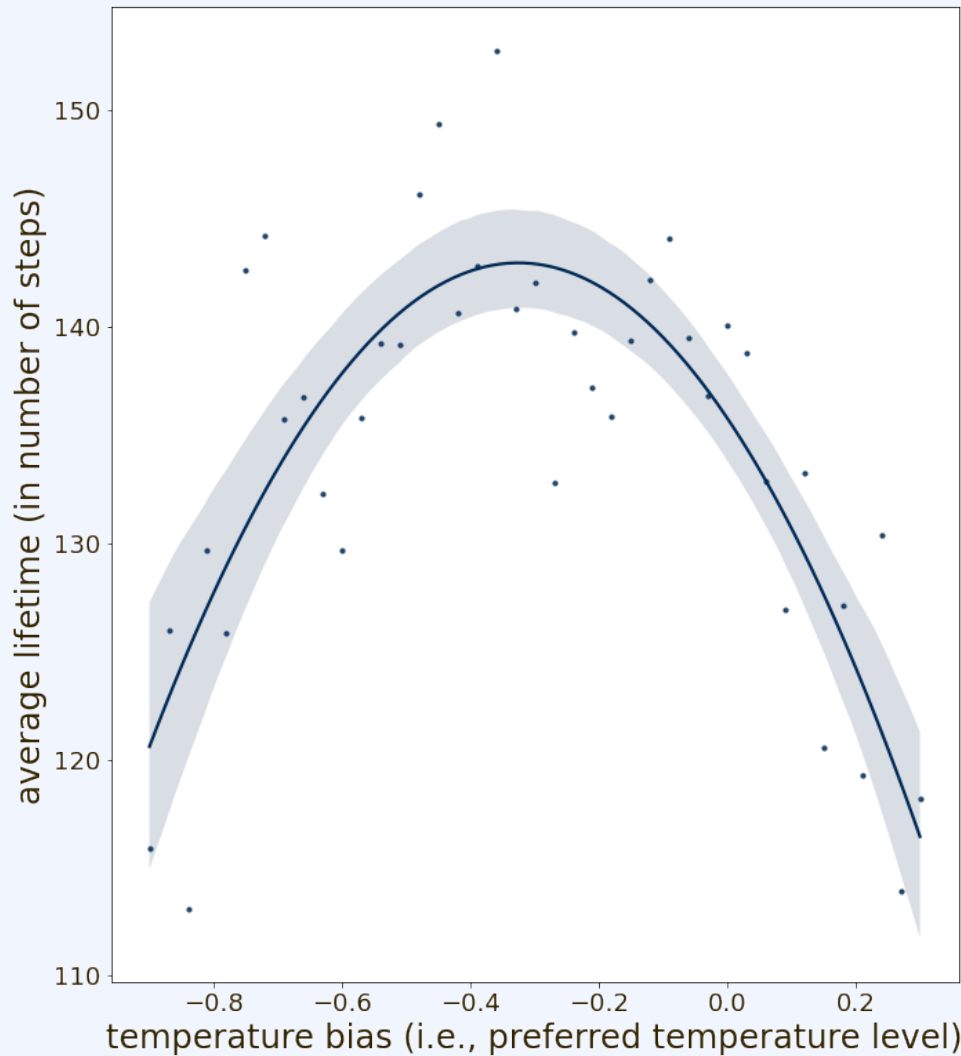


Good and comparably simple agent function:

- Check which of the immediately adjacent fields have $|T|$ closest to zero.
- Go in that direction, or stay put if they all have a greater $|T|$ than the present location.

Average life time: 134 steps.

Agent escaping extreme temperatures



Agent function optimization,
with one free parameter, yields:

- Check which of the immediately adjacent fields have $|T + \frac{1}{3}|$ closest to zero.
- Go in that direction, or stay put if they all have a greater $|T + \frac{1}{3}|$ than the present location.

Average life time: 146 steps.

Review of parts 1 to 3

Module structure

Upon successful completion of this module, a student will be able to:

- 1) Explain the theoretical underpinnings of algorithms and techniques specific to artificial intelligence;
- 2) Critically evaluate the principles and algorithms of artificial intelligence;
- 3) Analyse and evaluate the theoretical foundations of artificial intelligence and computing;
- 4) Implement artificial intelligence algorithms.

optimization

**agents and
decisions**

modelling

**game
theory**

**knowledge
representation**

**reasoning
and learning**

**uncertainty
quantification**

Part 1: Optimization

On the field of **optimization**, we have:

- Discussed single-parameter, single-objective problems as the base case in optimization; there, $y = f(x)$ is minimized or maximized;
- Looked at local and global optimization algorithms and the way in which local optimization is based on the functions' derivatives;
- Introduced multivariate and multicriteria optimization.

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hill climbing (local)

multivariate optimization

Newton's method (local)

multicriteria optimization

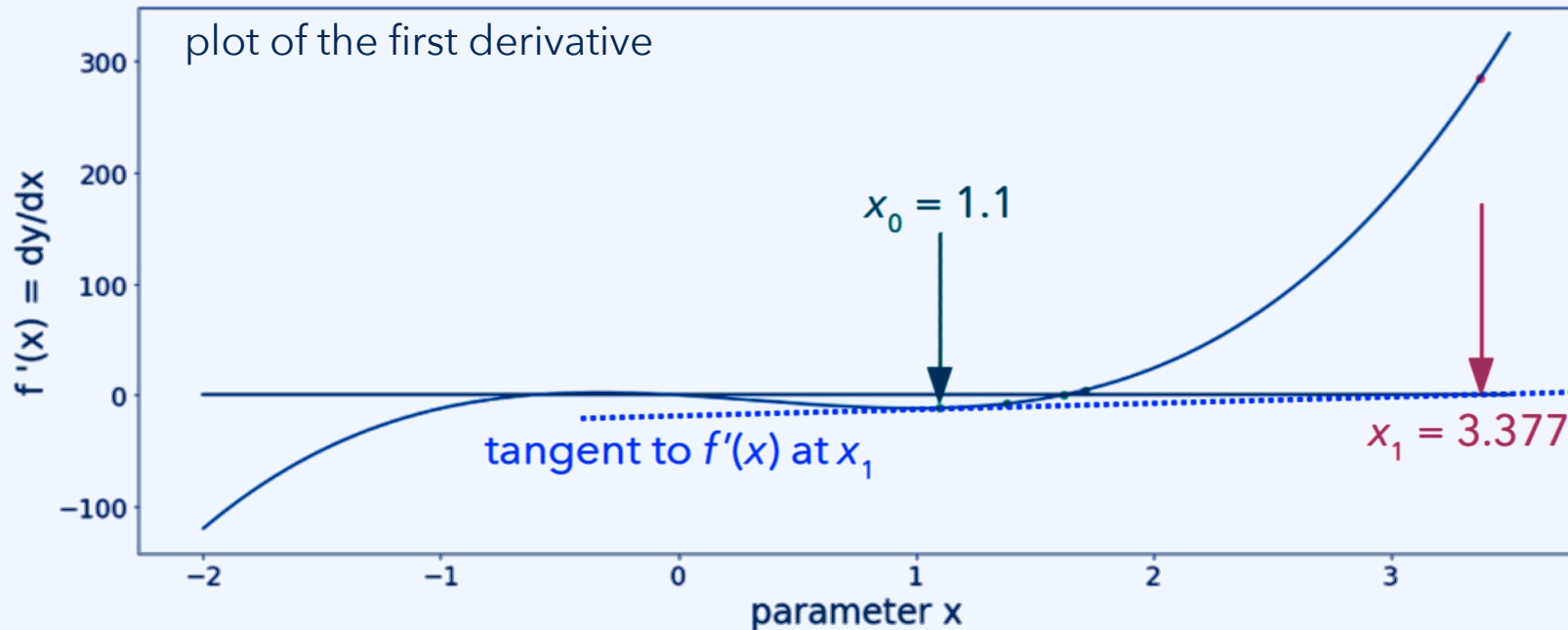
global optimization

Pareto front

Part 1: Optimization

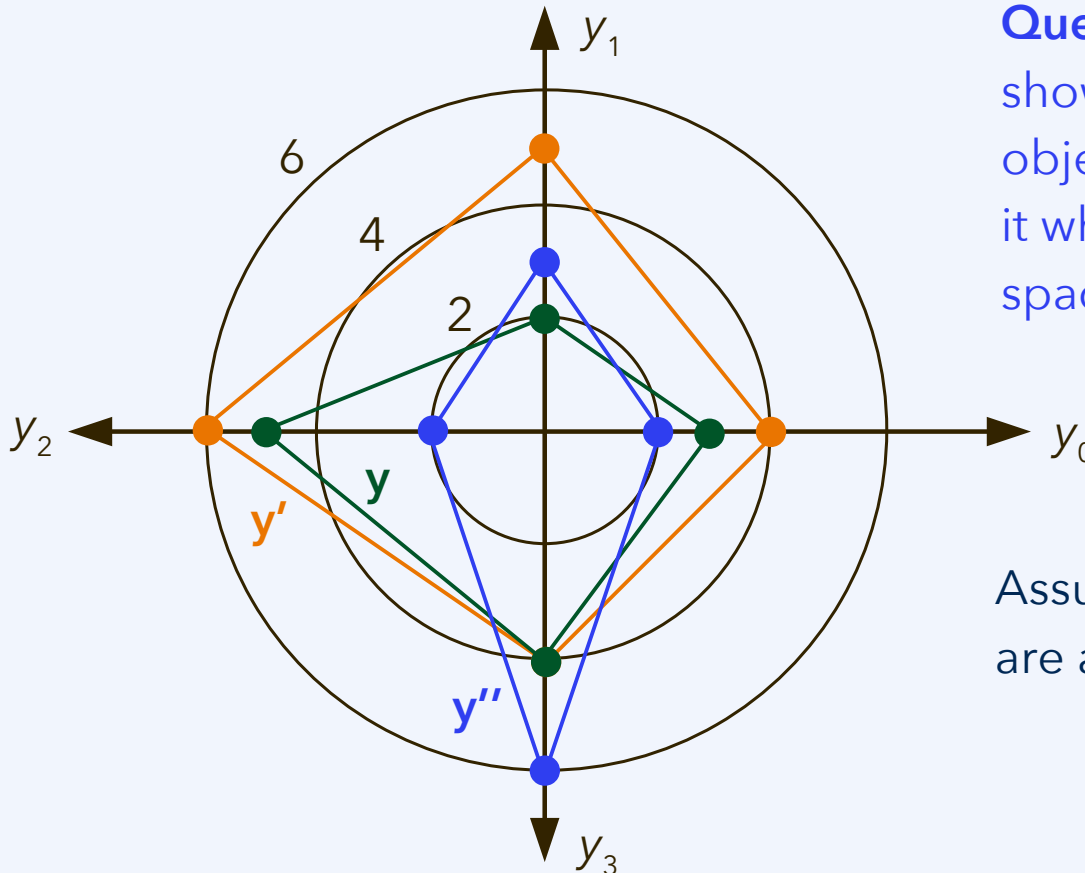
Revising the concepts

Question: What property does a function $y = f(x) = \dots$ need to have so that, using **Newton's method**, we find the minimum **exactly** within a **single step**?



Part 1: Optimization

Revising the concepts



Question: In a spider diagram showing the values for multiple objectives, how can we recognize it when one point in objective space **dominates** another?

Assume that y_0 , y_1 , y_2 , and y_3 are all minimization objectives.

Part 2: Agents and decisions

On the field of **agents and decisions**, we have:

- Reviewed common definitions of agency and intelligent agents;
- Discussed the use of AI in assisting human decision making;
- Explored philosophical issues related to the field.

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important types of agents:

decision support

agent

agent function

Turing test

goal-oriented agent

intelligent agent

rational agent

knowledge-based agent

Part 2: Agents and decisions

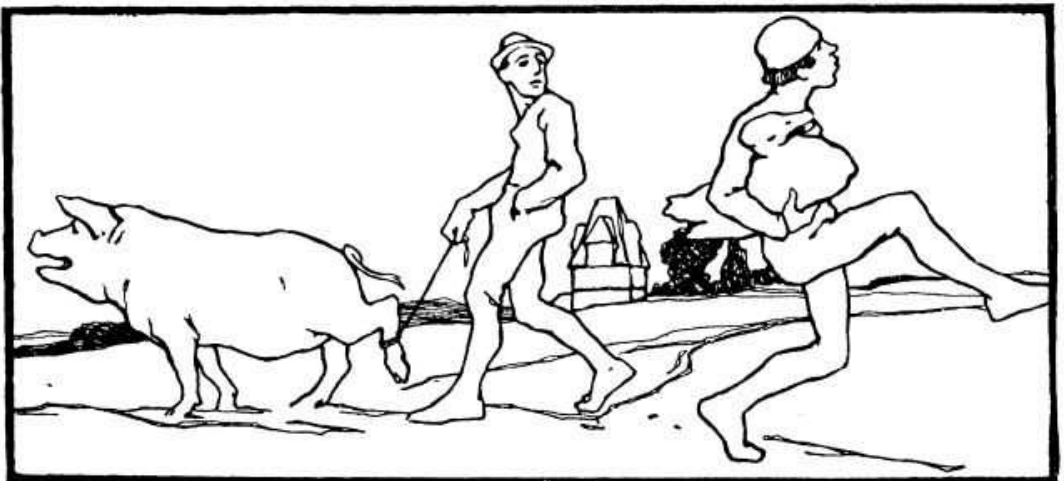
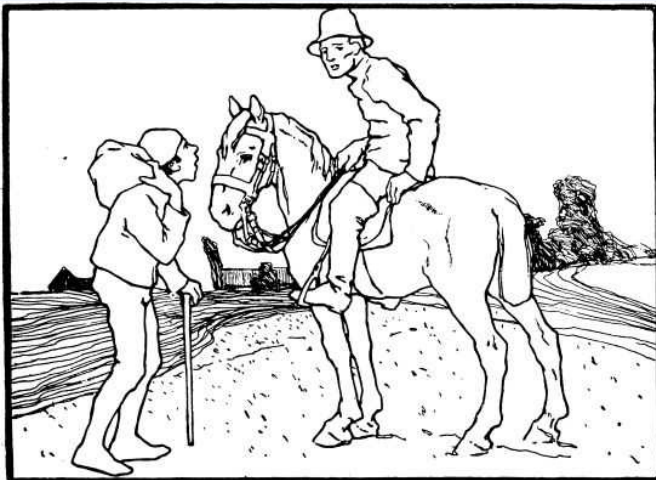
Hans moves to the city to work, earn money, and make savings.

From his savings, he buys a horse.

He exchanges the horse against a cow, then a pig, a goose, and a grindstone.

But the grindstone is too heavy, so he throws it into a well.

Question: Why do we say that Hans is not a rational agent?



"Hans im Glück" (Hans in luck); artist: O. Ubbelohde (1909).

Part 3: Modelling

On the field of **modelling**, we have:

- Discussed the use of models in optimization and decision support;
- Used optimization to parameterize models;
- Assessed model quality by validation and testing.

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supervised learning

training

hypotheses

linear regression

validation

overfitting

***p* value**

testing

margin of error

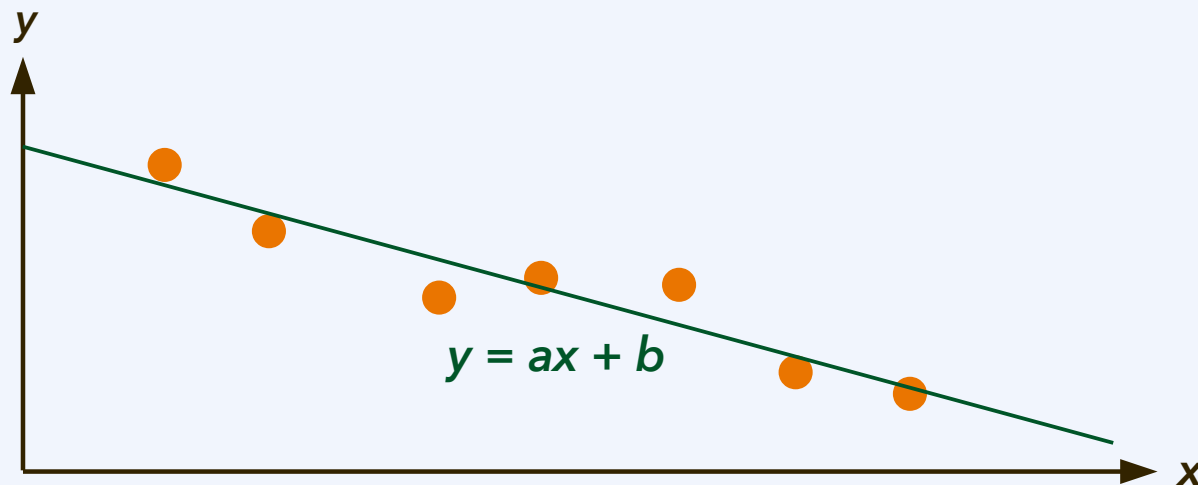
Part 3: Modelling

Revising the concepts

Model parameterization: Determine coefficients such as a , b , c in $ax^2 + bx + c$.

- Used optimization to parameterize models;

Question: How and where did we use **optimization** to parameterize models?



End-of-year reflection

Help improve the module for the coming year

Go to www.menti.com and use the code 8857 6306

Rate these hypotheses about CO3519:

Mentimeter

Strongly disagree

The lecture slides are too packed or too many

I prefer more practical rather than theoretical content

I prefer a focus on concepts, not on implementation

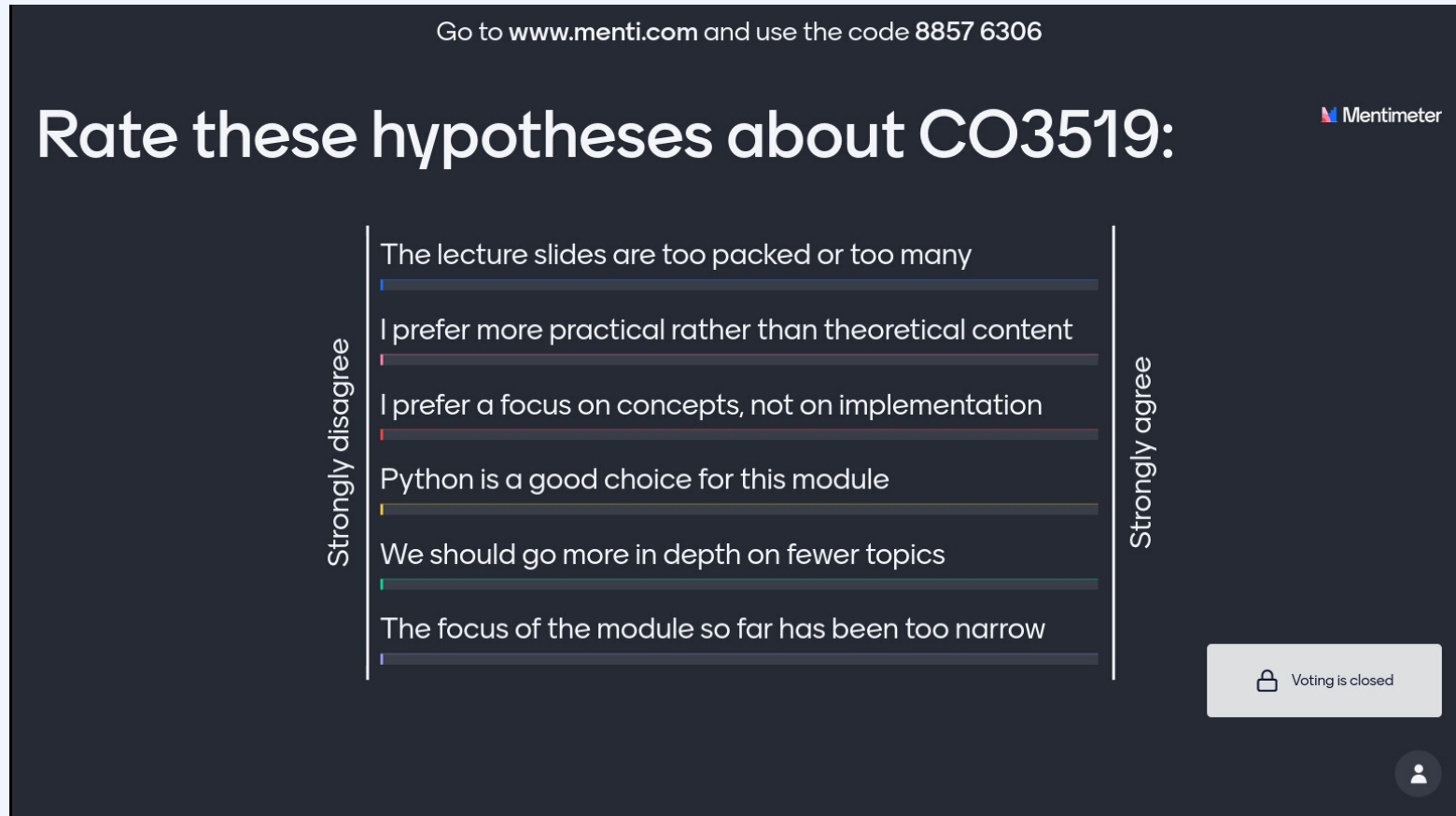
Python is a good choice for this module

We should go more in depth on fewer topics

The focus of the module so far has been too narrow

Strongly agree

Voting is closed



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