

Developments in Economics Education Conference 2023



Language models and AI in economic education: Unpacking the risks and opportunities

Tomasz Kopczewski
Ewa Weychert

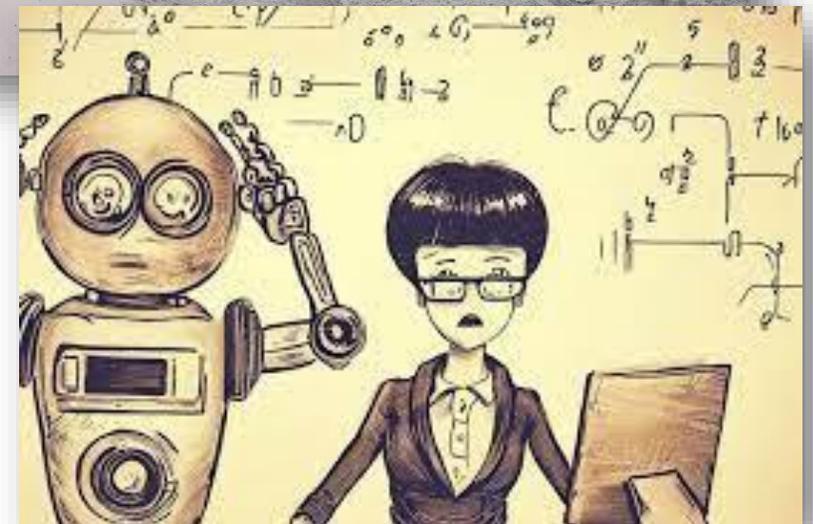


Surprise, surprise,

There is a Polish saying: The road services were surprised by winter again. There is also a version: Students were surprised by the session.

[It is a situation when we all know something is going to happen, we should be prepared, but we are not]

This paper aims to show how to avoid surprises using AI in economics and its teaching (in the long run). We should learn how AI will affect our perception/narration of economic reality.



We were surprised by AI too

The recent final exam sessions forced us to quickly adapt the grading rules to the unexpected emergence of ChatGPT as a student's toolkit.

We created the experimental final assignment consisting of three parts:

- i) students writing a short essay demonstrating their understanding of the given economic problem;**
- ii) providing a transcript of the chat conversation on the topic;**
- iii) giving a short critical analysis of chat use;**

We were surprised by AI too

This approach to changing the student assessment setup seems worth considering because it allows quantitative and qualitative analysis of students' work with the chat.

It provides three data samples of texts. With these samples, it's possible to examine the extent of similarity between student texts and the chat's text and investigate how students construct their narratives and how closely they align with the narrative presented by the chat.

Observation

We obtained a sample of text data from the final assignments in which students were asked to describe the fundamental problems: the market mechanism and equilibrium. The student's narrative in terms of language was much worse than ChatGPT, but in terms of comprehension, they (ChatGPT and students) presented variations on the content of primary microeconomics textbooks.

Quantitative research requires substantial data sets, and the experiment will be extended. Still, the student final assignments inspired the qualitative reflections presented in this article.

Observation

While the unauthorized use of these tools and cheating remains a didactic concern, creating a narrative where students assume a passive role towards AI as uncritical consumers of information generated by these tools can have even worse consequences on creating collective knowledge.

We will lose our knowledge completely in information (data) as predicted by Mirowski and Nik-Khah (2017)

Observation

The emergence of ChatGPT has shown us how much the two aspects of human interaction are neglected by economics.

- The creation of collective knowledge and the social learning process. [I will focus on the topic now]**
- The narrative as a carrier of collective knowledge.**

A simple idea -> lots of sophisticated tools

ChatGPT is 'just' generative AI. It is not a singularity, which is expected to happen according to futurists around 2035 [Kurzweil 2006, Modis, 2006]. Chat GPT is the result of evolution, not revolution !!!

The Postmodernism Generator - the IT extension of Alan Sokal's provocation is one of the first known examples of the use of algorithms that, improved and reinforced constitute algorithmic/generative AI today.

Try the generator. Isn't this type of text familiar? The artificiality of this text hurts - the ChatGPT texts hurt less.



<https://www.elsewhere.org/pomo/>

A simple idea -> lots of sophisticated tools

Generative artificial intelligence create content based on existing data.

Despite the use of very sophisticated tools, the idea behind generative intelligence (ChatGPT) is simple: based on the language corpus, determine the most likely word that follows the sentence of other words ->

The last word in the sentence "The mouse ate the _____"

-> and using one parameter (temperature) we can determine whether the text should be very conservative (cheese) or more creative (pizza) or nonsensical (cat).



<https://lukesalamone.github.io/posts/what-is-temperature/>

A simple idea -> lots of sophisticated tools

Generative models average the knowledge that has been generated by humans and is found in language corpora, graphic and film collections.

What is the most likely word in a typical text about a mouse?

What does the face of a surprised person look like? Average all images containing the faces of surprised people. These images must have been described by people.

.....

Averaging is a mechanism of collective knowledge (Surowiecki 2005). Generative AI creates something new based on provided databases. It is a representation (averaging) of the collective knowledge of humanity.

A simple idea -> lots of sophisticated tools

If we have little data or it is difficult to find links between terms in a searched statement it is easy to show that our AI is mindless. We have a "garbage in, garbage out" situation.

Example of four galleries:

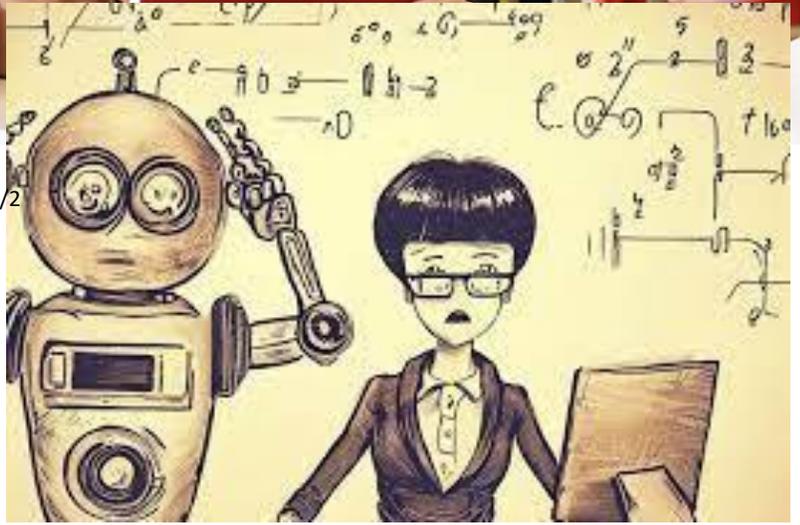
The first two galleries were created by searching graphics for two statements on Google:

- *The road services were surprised by the winter again*
- *Teachers were surprised by AI*

The next two galleries were created to reproduce similar images by the most advanced graphical AI [DALL-E].



<https://www.forbes.com/sites/rashishrivastava/2022/12/12/teachers-fear-chatgpt-will-make-cheating-easier-than-ever/?sh=100351861eef>

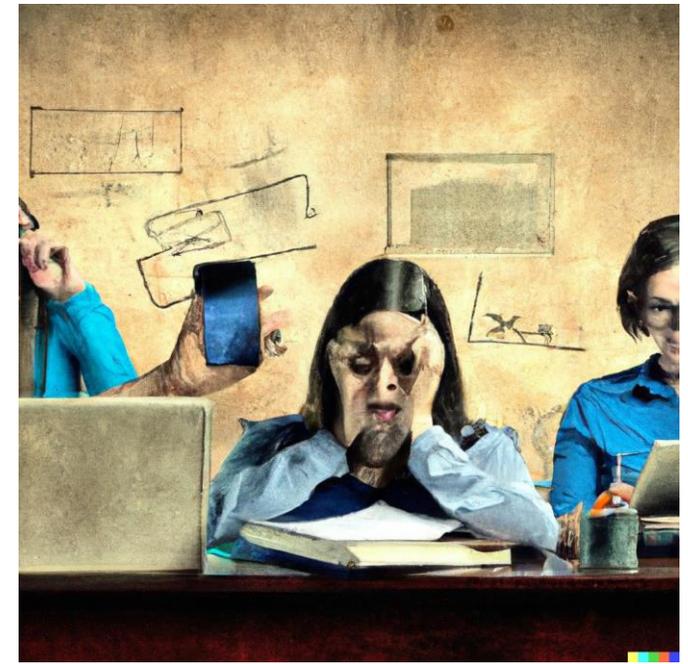


<https://learnlearn.uk/ai/2023/05/10/how-to-assess-student-work-in-the-age-of-chatgpt/>



The road services were surprised by the winter again





Teachers were surprised by AI



AI errors and the economics

Hallucinations of AI are not dangerous when they are so obvious. The problem arises when there is a lot of data and artificial intelligence produces an epistemologically coherent message, but at the same time simplified and skewed.

This is what happens in economics, a new type of epistemological risk (uncertainty) emerges that economics cannot manage.

For decades economics has neglected the problems of narrative, social learning, and the creation of collective knowledge. Understanding the essence of this risk is a condition for avoiding its consequences, but we must do so using economic concepts.

Positive effects

Generative AI aggregates information and thus reduces information overload. Information overload (Simon) means that too much information (different sources) reduces the ability to understand the message and make rational decisions.

But the solution to the problem creates another huge problem. Generative AI becomes an algorithmic troll inhibiting the emergence of collective knowledge.

Collective wisdom and AI

Three of the five conditions for the emergence of collective knowledge (Surowiecki 2005) will be significantly weakened if generative AI becomes the dominant source of information:

- **Diversity of opinion:** Everyone should have private information (**everyone will have the same information generated by AI and, in the future, AI models will feed on themselves, which will make it mad (Alemohammad et al. 2023)**)
- **Independence:** People's opinions are not determined by the views of those around them. (**AI will influence all**)
- **Decentralization:** People can specialize and draw on local knowledge. (**There will be an oligopoly of AI firms, and all models will average accessible information**)

AI as an algorithmic troll

An algorithmic troll is an agent in a network that at least influences one other agent in the network but is not itself influenced by other agents.

Does generative AI meet this assumption? Yes, it does. If the information in the network is well established, then the influence of an individual on the network's data resources will tend to be zero. And the AI itself influences a significant number of agents.

The 'algorithmic troll' model was used as a computational and didactic experiment in which we show how generative AI can reduce plurality and diversity in the network and weak collective knowledge in the long term. We uses a modified DeGrott social learning model.

DeGroot model

- It is a simple social learning model that provides a basis for creating complex models of social interaction and finding consensus or social polarization.
- It is an interesting example of the application of Markov chains to analyse social behaviour.

$$\mathbf{x} = (x_1, x_2, \dots, x_n) \in [0,1]^n$$

Agents' opinion/belief vector

0 - do not believe in

1 - I believe in

$$W = \begin{bmatrix} w_{1,1} & \cdots & w_{1,n} \\ \vdots & \ddots & \vdots \\ w_{n,1} & \cdots & w_{n,n} \end{bmatrix}$$

Social interaction matrix;

Who and how influences to whom?

Model DeGroota

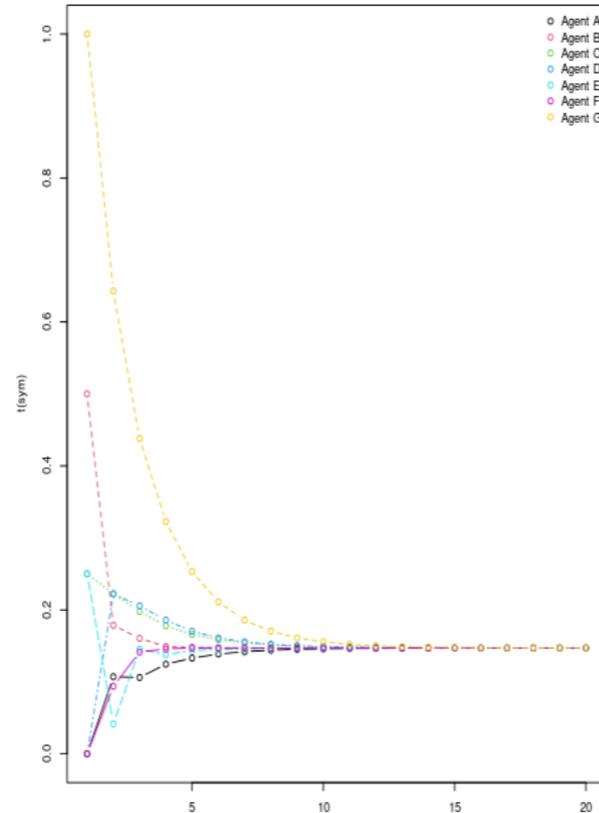
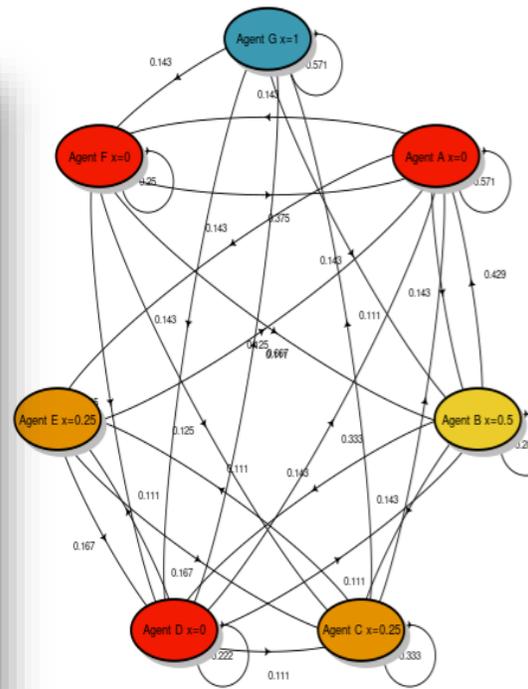
The model tends towards a stationary state. By averaging, we come to a consensus - we learn from others.

Number of iteration

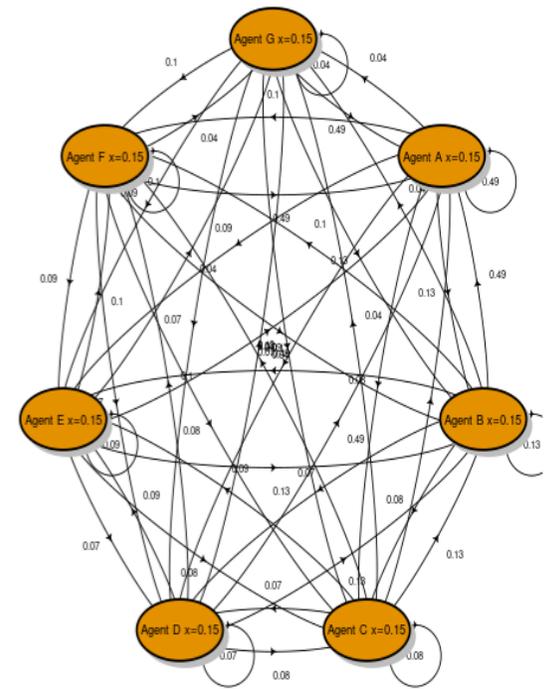


DeGroot model of social learning

Social learning/influence network t = 0



Social learning/influence network t = 20



$$x(t + 1) = W^t x(1) = W^{t+1} x(0)$$



DeGroot model with troll

We assumed that one agent believes **only in itself**, **no one influences the agent**, and **the agent influences at least one agent**. We called this agent an algorithmic troll.

$$W = \begin{bmatrix} 1/3 & 1/6 & 1/3 & 1/6 \\ 1/3 & 2/3 & 0 & 0 \\ 0 & 1/6 & 5/6 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

DeGroot model with troll

The introduction of the troll revealed a fundamental assumption of the DeGroot model: the matrix W does not create a strongly connected graph. We analysed the impact of abolishing this assumption on the model's behaviour.

$$\lim_{t \rightarrow \infty} x(t) = \lim_{t \rightarrow \infty} W^t x(0) = \begin{bmatrix} 0 & 0 & \dots & 1 \\ 0 & \ddots & 0 & \vdots \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} x(0)$$

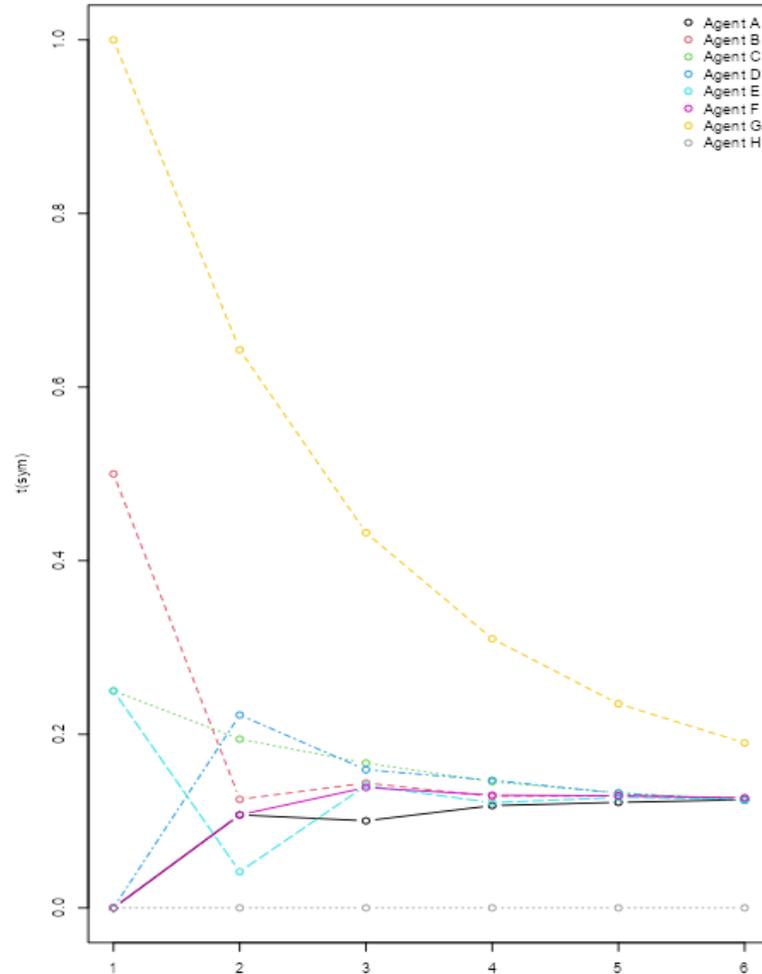
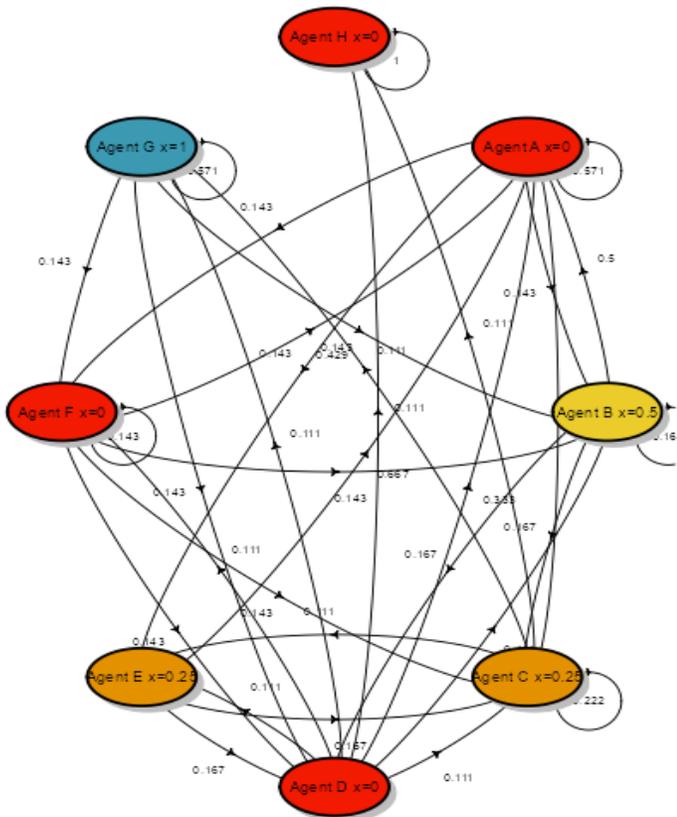
It is not an interesting algebraic solution, but it is interesting how the system behaves. We can distinguish two states of the system:

- The first phase of merging agents
- The second phase of graph degeneracy

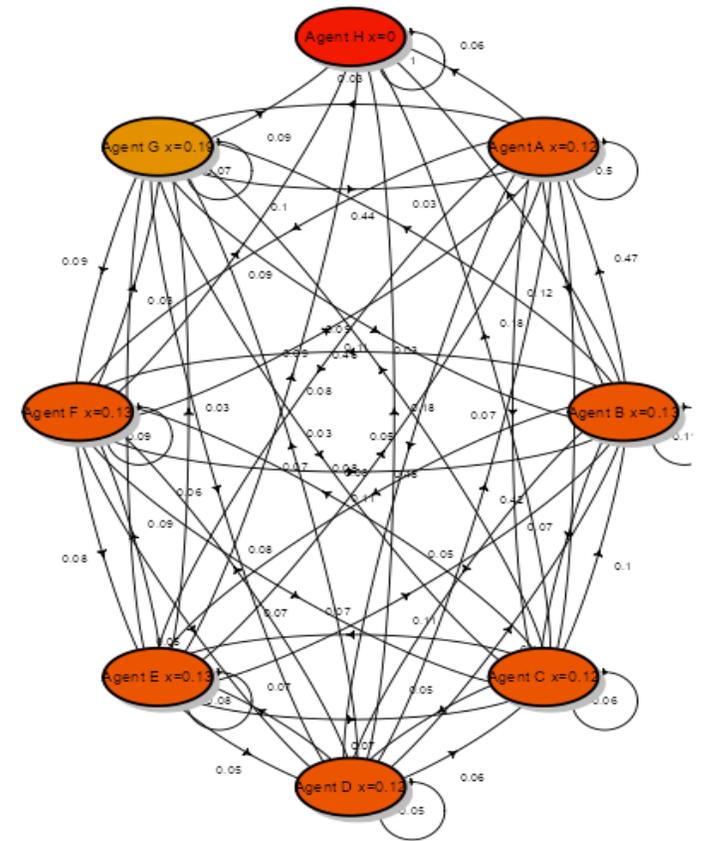
DeGroot model with troll

- Merging phase - the graph becomes complete

Social learning/influence network $t = 0$



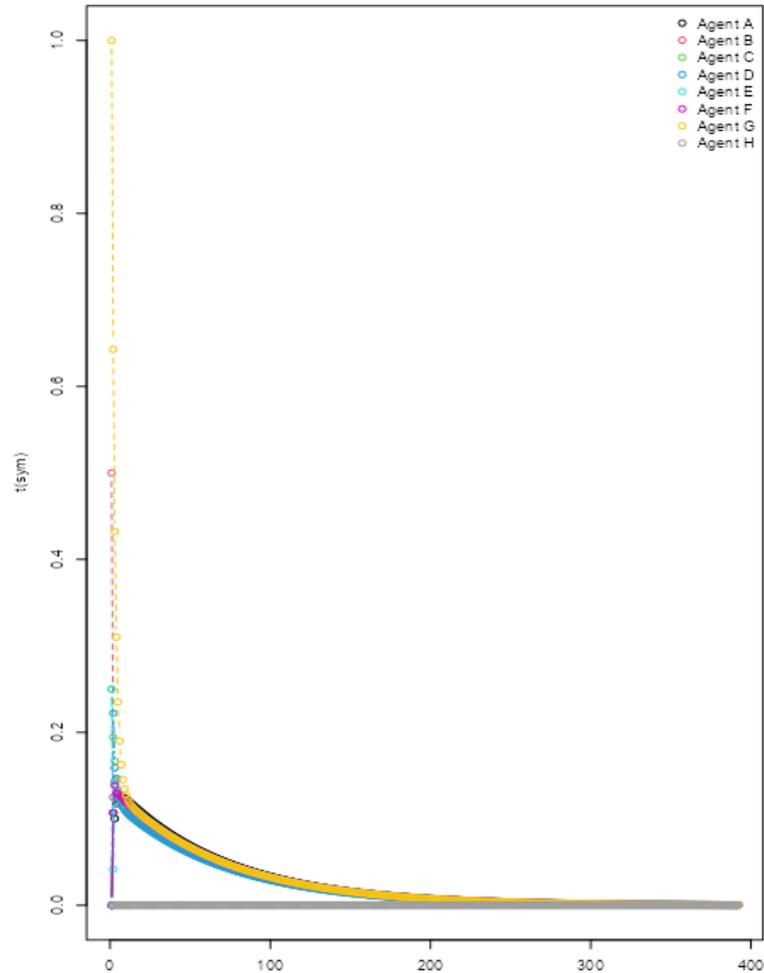
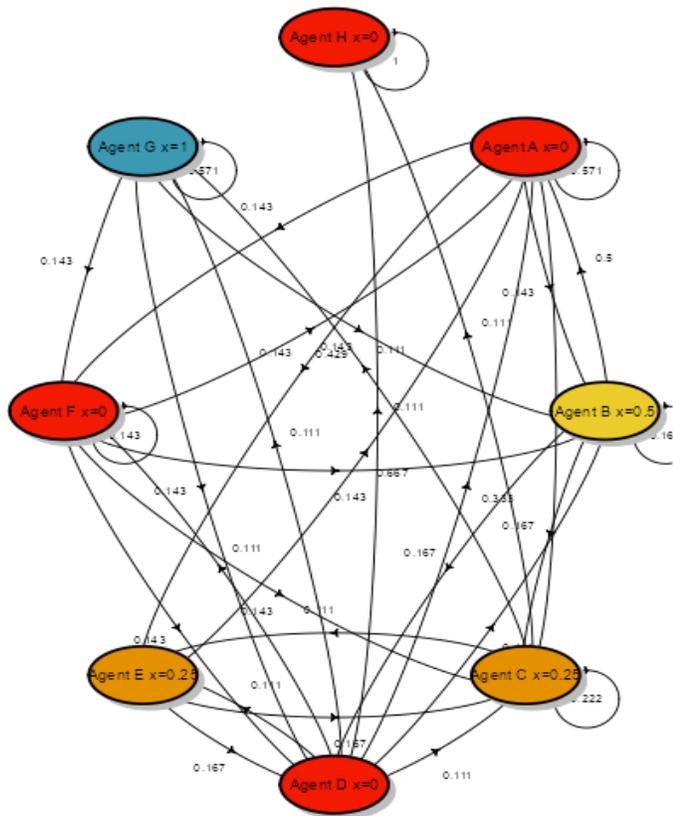
Social learning/influence network $t = 6$



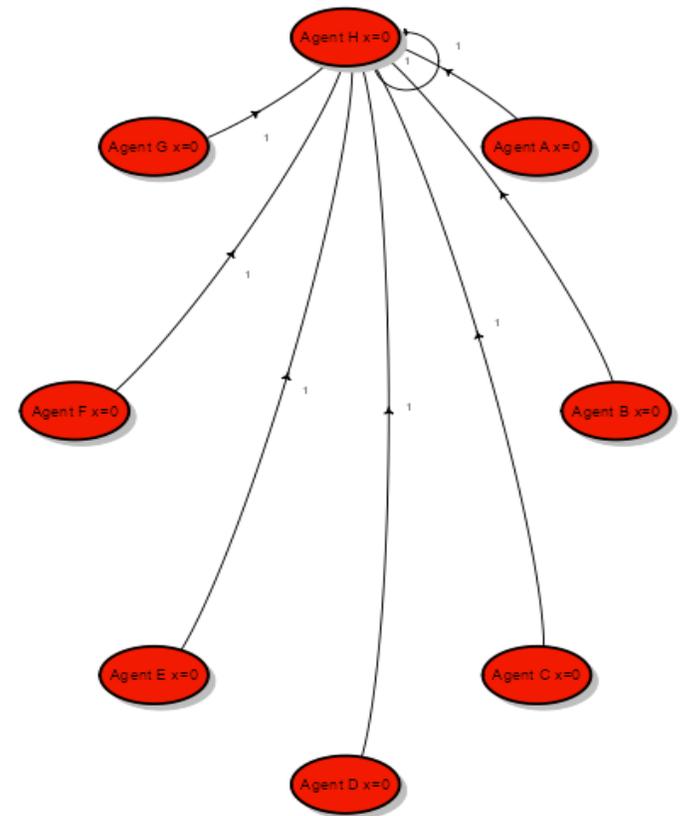
DeGroot model with troll

The second phase: graph degenerating

Social learning/influence network $t = 0$



Social learning/influence network $t = 392$

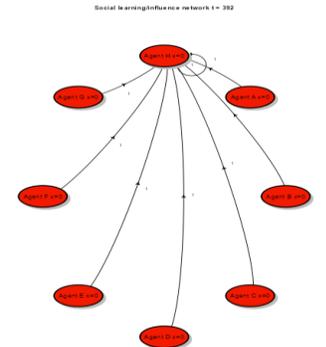
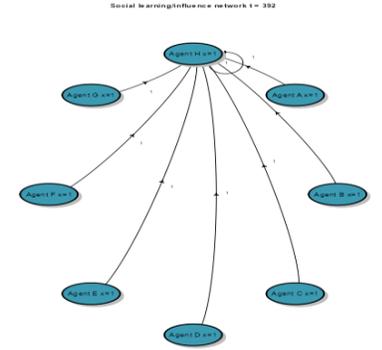


DeGroot model with troll

It does not matter whether the troll believes (1) or not (0) - the system is heading towards the troll's opinion anyway.

The model shows two phenomena:

- The illusory truth effect (repetition effect) - This is the tendency to believe that false information is correct when repeatedly repeated (Hasher et. al, 1977)
- The salami tactic - slow change that goes unnoticed - is socially acceptable (Enh, 2010).



Epistemic Uncertainty

Epistemic uncertainty - mainly refers to uncertainty arising from a lack of knowledge or statistical evidence. But in the case of an algorithmic troll, it will be the propagation of the 'statistically' strongest narrative too.

AI can propagate i) misconceptions about economics (dominance of orthodox economics) ii) discrimination against social groups that are misrepresented in available databases (AI will omit information about these groups as 'statistically insignificant), iii) misinterpretation of facts

Epistemic Uncertainty

Misinterpretation of facts may be caused by even minimal distortions.

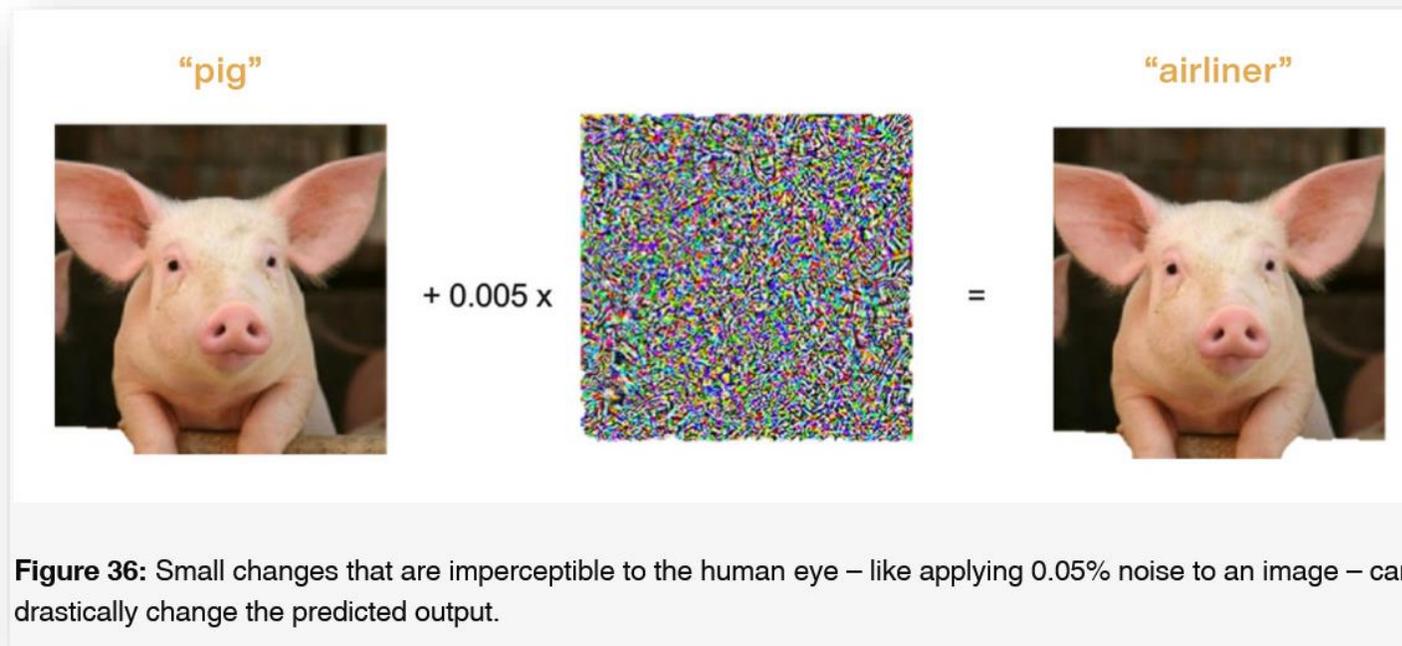


Figure 36: Small changes that are imperceptible to the human eye – like applying 0.05% noise to an image – can drastically change the predicted output.

<https://c3.ai/introduction-what-is-machine-learning/ensuring-algorithm-robustness/>

AI in the classroom_(non-technical approach)

How to deal with AI in the classroom

- Use economic tools to analyze AI (attention economics, social learning models, collective knowledge) - understanding is key.
- Change narratives about AI:
 - Passive use of generative AI is the first step to unemployment.
 - You must be a "critical miner" of generative AI. Don't stop at acquiring the ore (information) - turn it into knowledge and share it with others. Your even imperfect interpretation of information is needed to enhance diversity and, thus, collective knowledge.

AI in the classroom

The experimental final assignment is standard now. It is teaching students critical „mining of information” from AI. It consisting of three parts:

- students writing a short essay demonstrating their understanding of the given economic problem
- providing a transcript of the chat conversation on the topic
- giving a short critical analysis of chat use

The evaluation covered student statements and the manner and quality of questions in the chat conversation.

Our future work (we are going to share)

Paper about philosophical and methodological issues about AI in the class (attentional economics, social learning models, collective knowledge, narrative economics)

**Empirical paper based on text samples on how the distances between
i) ChatGPT ii) students iii) economics textbooks narration change over time.**

Course: Use ChatGTP to learn econometrics and R-CRAN (materials)

**Thank you and be tuned
(tkopczewski@wne.uw.edu.pl)**