

Hpi Workshop Helsingor, 2024

# AI in Science: SWOT analysis

Strength, Weaknesses, Opportunities, Threats

# Important Definitions:

- “Generative” AI: Large language models (LLM), natural language processing (NLP) such as ChatGPT,

Translations, music/voice generators, image/movie generators

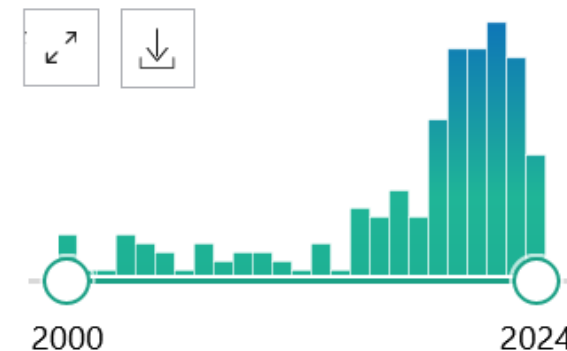
(statistics) Closed/Blackbox (original data not known, usually based on large training datasets; strong “admixture” = mosaics texts)

- Logical and logical data-based AI: alignments, -Omics data analysis, evolutionary distance analysis, structural models (alphafold)

More “open box” (original data known and accessible), includes machine-learning algorithms

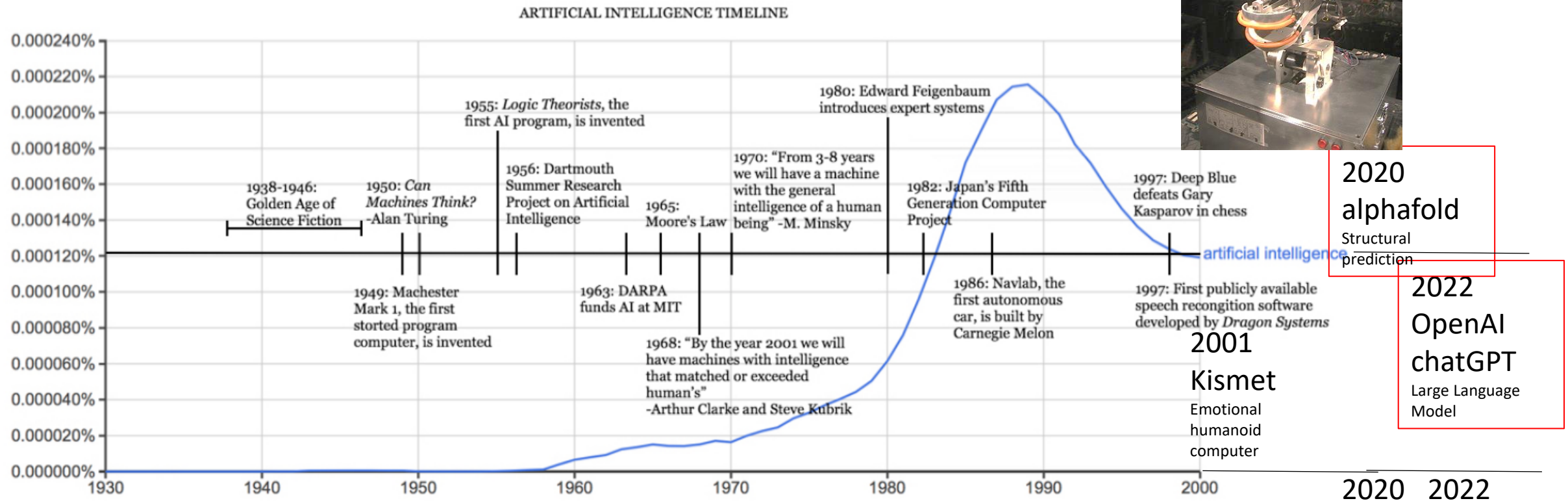
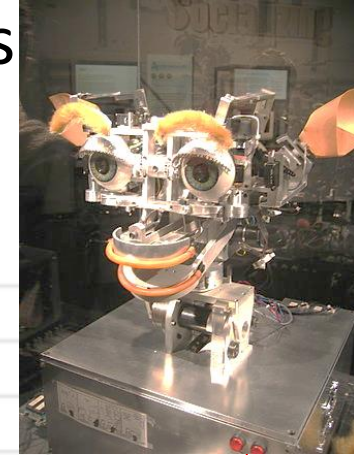
*H. pylori*: 164 papers, mostly logical AI use  
(clinical, meta reviews)

RESULTS BY YEAR



# History:

- Old models and algorithms (<30 years); current major difference is the amount of training data going in and the increased calculation speed and storage capacity of large supercomputers and computing networks



- # Generative AI (ChatGPT), example
- PROMPT: The study of **Helicobacter pylori** (H. pylori) has advanced significantly, but several important questions remain open:
- Mechanisms of Pathogenesis**:
    - How exactly does H. pylori cause gastric cancer and other diseases? Knowing detailed molecular pathways and interactions between the bacteria and host cells is crucial.
  - Antibiotic Resistance**:
    - How can we effectively manage and overcome the increasing AMR in H. pylori strains? Alternative therapies?
  - Immune Response and Host Interaction**:
    - How does H. pylori evade the host immune system? Role of host genetic factors?
  - Transmission and Epidemiology**:
    - What are the precise routes and mechanisms of H. pylori transmission? Public health interventions?
  - Microbiome Interactions**:
    - How does H. pylori interact with the rest of the gastric and intestinal microbiota?
  - Vaccine Development**:
    - What are the challenges for effective vaccine against H. pylori and how can this be solved?
  - Strain Variability and Virulence Factors**:
    - How do different H. pylori strains vary in terms of virulence and disease-causing potential?
  - Environmental and Lifestyle Factors**:
    - What environmental and lifestyle factors influence the risk of H. pylori infection and the severity of associated diseases?
  - Long-term Consequences of Infection**:
    - What are the long-term health consequences of H. pylori infection beyond gastric diseases?
  - Diagnosis and Screening**:
    - How can diagnostic methods and screening be improved to detect H. pylori (non-invasively)?
- Addressing these questions requires interdisciplinary research combining microbiology, immunology, genetics, epidemiology, and

# SWOT – generative AI (**G**), logical AI (if x- then y; (**L**))

## Strengths (Science):

- Can possibly be time-saving (**G**, **L**)
- Analysis of large datasets, publication collections (**G**)
- Facilitate all data analysis (e.g. Omics) (**L**)
- Facilitate structural discovery (proteins, others molecules) (**L**)
- Facilitate drug discovery (**L**)
- Facilitate image analysis and medical image analysis (e.g. pathology)

## Opportunities (Science):

- Improve output
- Improve detectability of mistakes/plagiarism
- Provide open code

## Weaknesses (Science, General):

- (growing) Business models behind it (**G**, **L**)
- Often no sources given (**G**) “It is a mix”
- Sources not transparent (**G**)
- No open code available (**G**, **L**)
- May support or induce plagiarism (**G**)
- May infringe copyrights/confidentiality (**G**)
- May contain filters/biases (**G**)
- Degeneration and mistakes of output and code (**G**)
- Appropriation of content/data/art created by others (**G**, **L**)
- [May inhibit intellectual training]

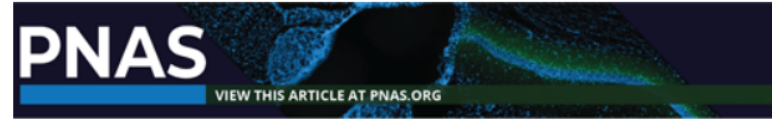
## Threats (to Science, but also to the models themselves):

- Degeneration of training data (**G**) (vicious cycle)
- Degeneration and mistakes of output and code (**G**)
- Faulty data, faulty interpretation (garbage in, garbage out)
- Large energy and resource expenditure (**G**, **L**) 100 households-1 d, 4 Mio liters water for pre-training ChatGPT

# Current Policies/Measures

- Recent papers for use guidelines of generative AI in science/medicine (NIH authors)

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[Proc Natl Acad Sci U S A.](#) 2024 May 28; 121(22): e2407886121.

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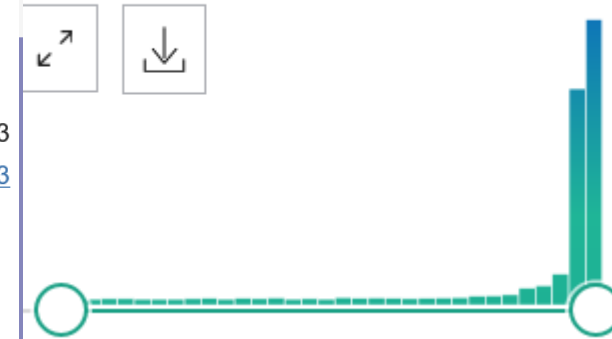
Published online 2024 May 21. doi: [10.1073/pnas.2407886121](https://doi.org/10.1073/pnas.2407886121)

PMID: [38771193](https://pubmed.ncbi.nlm.nih.gov/38771193/)

## Protecting scientific integrity in an age of generative AI

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RESULTS BY YEAR



- Journal policies: allow no use of AI (LL) for text generation or without precisely giving sources; only for text polishing; humans remain responsible for content; no use of AI (LL) for generating manuscript reviews; give AI use for selected purposes. No use for image generation
- Funding agency policies (similar to journals); no use of AI (LL) for making reviews

- More SWOT
- More comments
- Legitimate/ GSP Uses