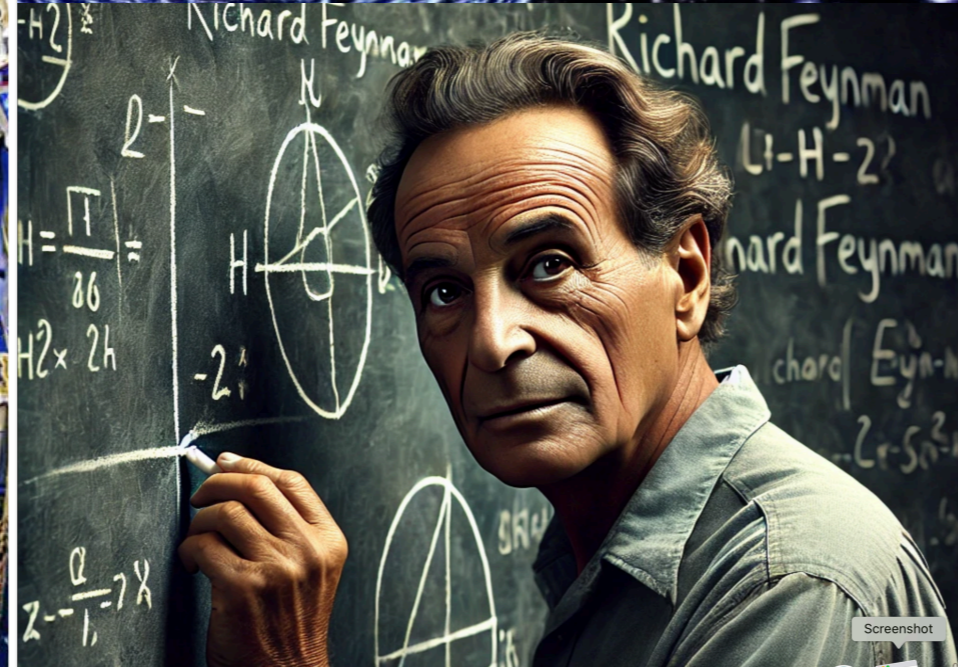
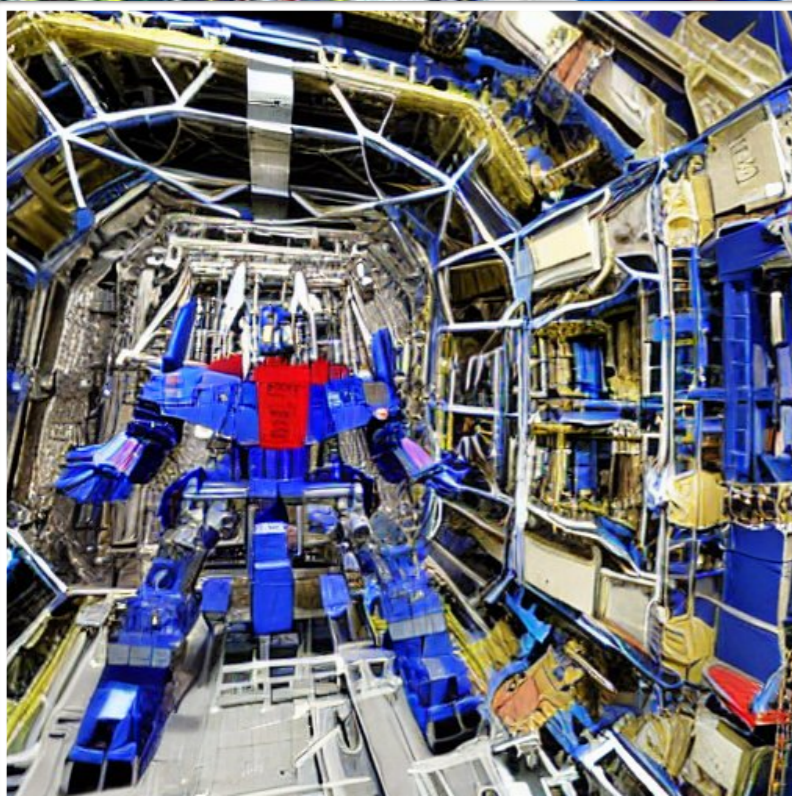
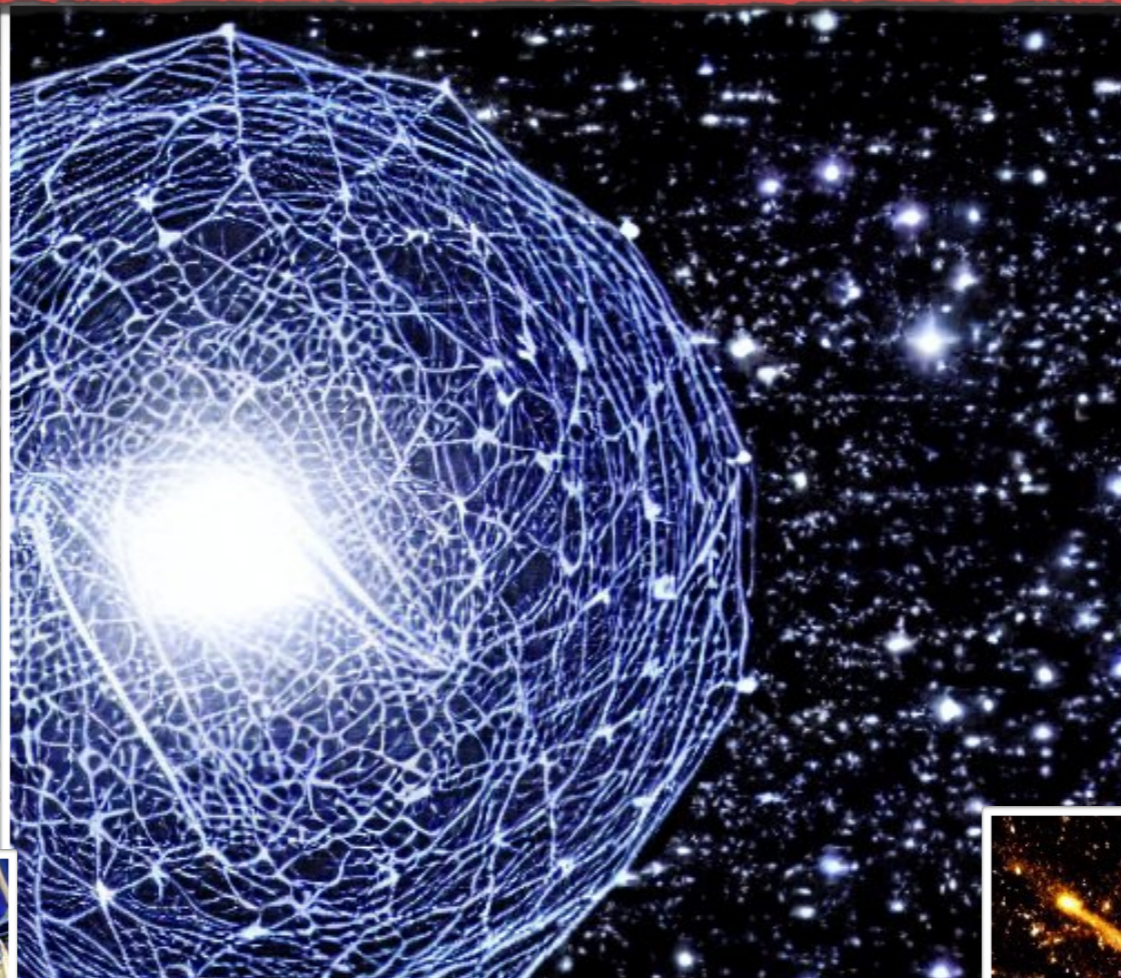


Future prospective of involving AI in research

Hrayr Matevosyan



AI APPLICATIONS IN PHYSICS

Applications of LLMs and optimisation:

▶ *Language Models:*

◆ *Literature search in scientific articles:*

Questions Answering / Summarisation.

◆ *Analysis code: Code generation / documentation.*

▶ *Optimisation:*

◆ *Experiment Shift Personnel Scheduling.*

◆ *Detector Design Optimisation.*

NATURAL LANGUAGE PROCESSING (NLP)

Creating programs that “understand” and process natural language such as text and speech.

- *Tasks:*

- Named Entity Recognition
- Question Answering
- Machine Translation
- Text Generation
- Sentiment Analysis
- Summarisation

- *Extremely Challenging:*

Often need additional context: “The Silver Arrows were trampled by the Prancing Horse this season”.

CONTEXT MATTERS

- *Extremely Challenging: often need “world knowledge”*

“The Prancing Horse trampled the Silver Arrows on the track”.



LARGE LANGUAGE MODELS (LLM)

Breakthroughs: Transformers (2017) and Transfer Learning (2018)
Generative AI / “Reasoning Models” - Inference time compute

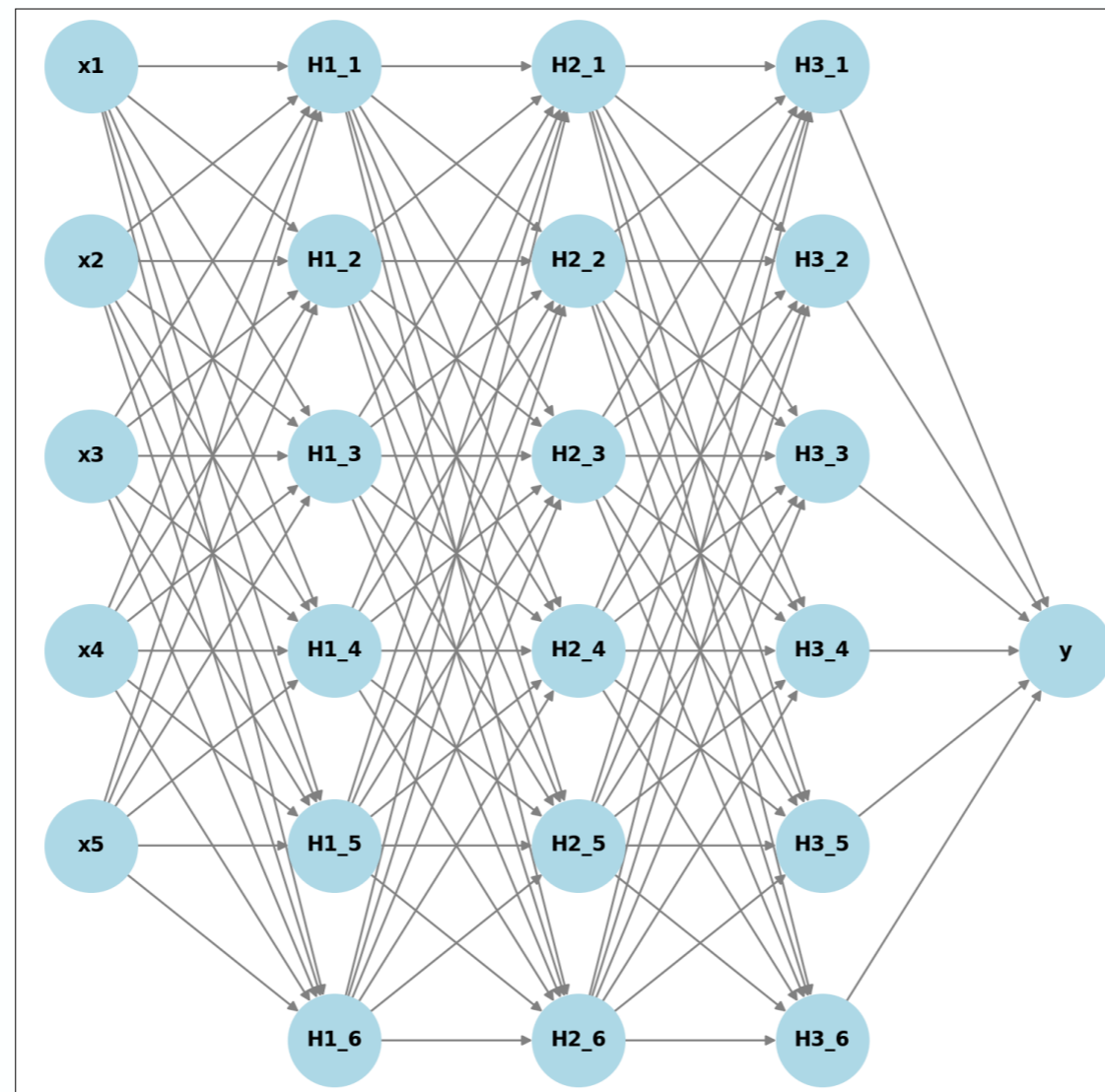
- ▶ **Fundamentally: “next token prediction” models**
 - Pre-training (Unsupervised Learning) - next token prediction
 - Fine-Tuning (Supervised/Task-Specific Training) - follow instructions
 - Alignment and Reinforcement Learning with Human Feedback (RLHF)
- ▶ **Very large up to $\sim 10^{12}$ parameters !**
- ▶ **Encode “world knowledge”** - knows what “Prancing Horse” refers to.
- ▶ **Hallucinations** - will always produce an answer, can give an answer that “mimics” the correct one.

LARGE LANGUAGE MODELS (LLM)

x

y

The measurements of the azimuthal asymmetries in two-hadron production in semi-inclusive deep inelastic scattering (SIDIS) process have been recently analyzed to extract the nucleon transversity



LARGE LANGUAGE MODELS (LLM)

Breakthroughs: Transformers (2017) and Transfer Learning (2018)
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ADVANCED SEARCH QUESTION ANSWERING (QA)

Answers to scientific question:

<u>LLMs (GPT, Claude)</u>	<u>+Search (Perplexity)</u>	<u>Generative QA (RAG)</u>	<u>Extractive QA</u>
<i>Snapshot of internet.</i>	<i>Search “fresh” data.</i>	<i>Use relevant texts.</i>	<i>Use relevant texts.</i>
<i>Generate answers.</i>	<i>Generate answers.</i>	<i>Generate answers.</i>	<i>Highlight the answers.</i>
<i>Hallucinations</i>	<i>Hallucinations</i>	<i>Hallucinations</i>	<i>Cannot combine parts</i>
<i>Slow</i>	<i>Slow</i>	<i>Slow</i>	<i>Fast</i>

*Generative Models can represent the “average opinion”
NO critical reasoning!*

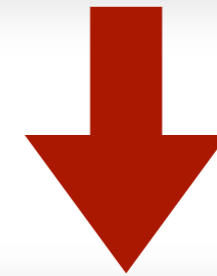
Extractive + *Generative QA*

CODE GENERATION

- *The latest generation of “reasoning” models can generate high quality code in Python / Java / C++, etc.*
- *Large “context windows” $\sim 10^5$: include the relevant parts of existing code / documentation.*
- *The generated code can be auto “tested” using “agents”:
Check for code execution / results consistent with the physics.*
- *Document existing convoluted code.*

OPTIMISING PERSONNEL SCHEDULING

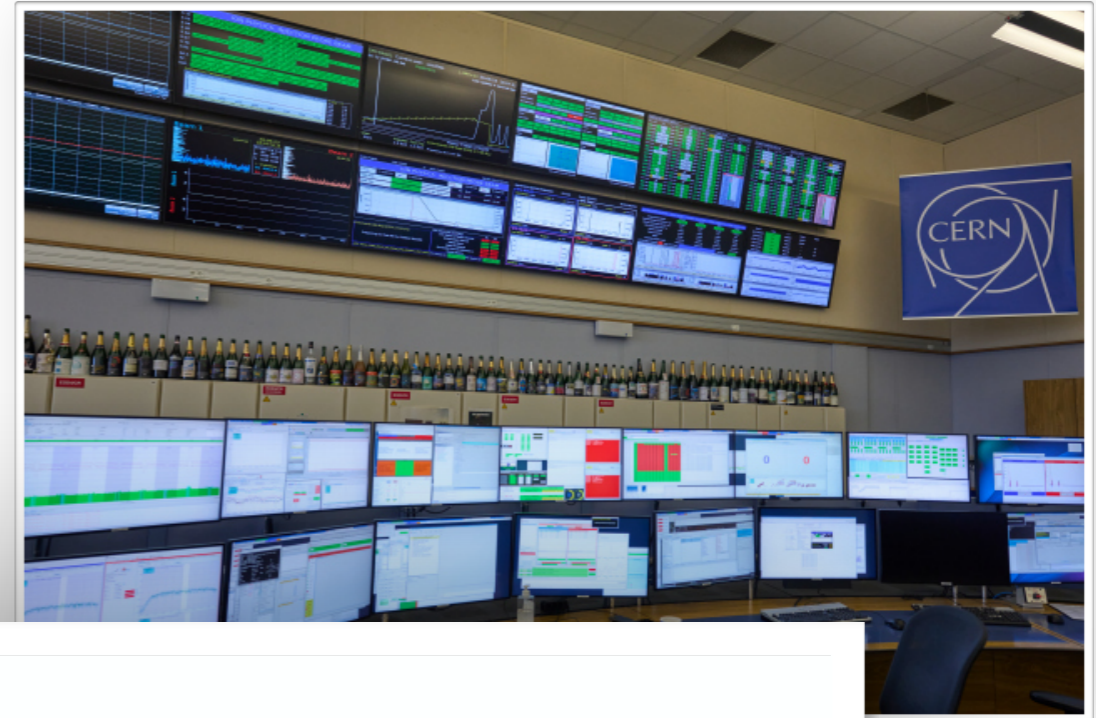
- **The Need:**
 - *Cover all the stations w/ specialists*
 - *Peoples availability / expertise*
 - *Cove the breaks (meals/rest)*
 - *Preference / Fairness*
- **Re-Scheduling:**
 - *Disruptions in experiments*
 - *Absences / change of availability*
 - *People leaving / joining the experiment*
- **Outcomes:**
 - *Reduced Planning time / effort*
 - *Fairness*
 - *Happy(er) scientists*



OPTIMISING PERSONNEL SCHEDULING

- **The Need:**

- *Cover all the stations w/ specialists*
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- **Re-So**

- *Disrup*
- *Absenc*
- *People*

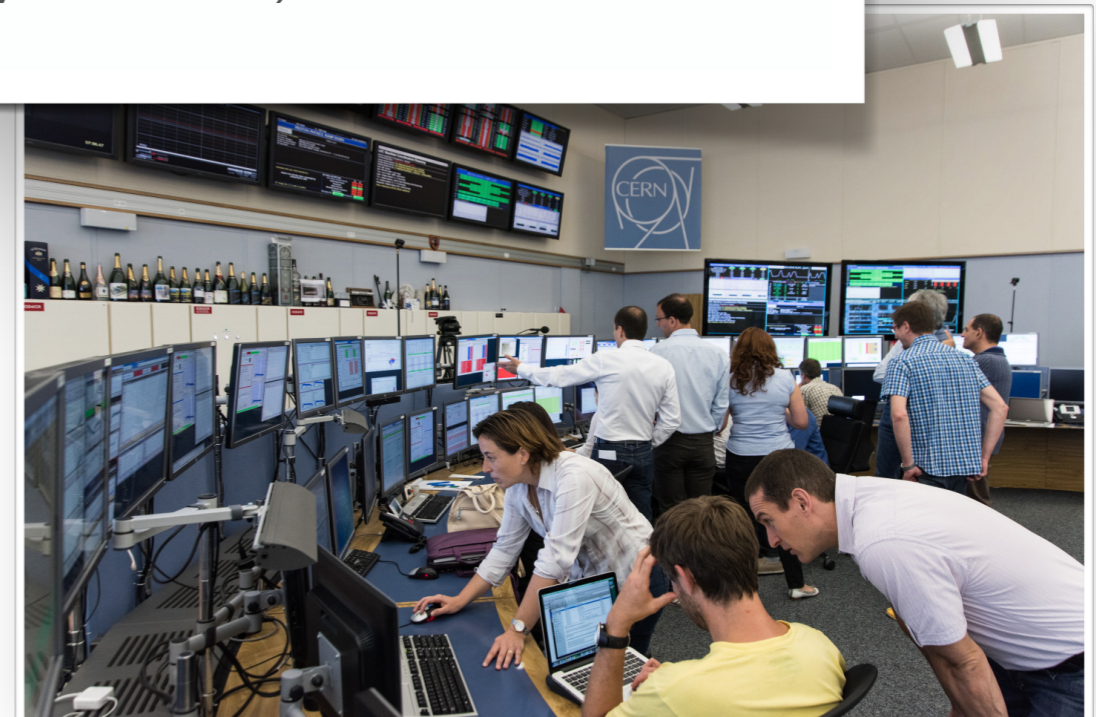
NP*-HARD Problem:

Number of shift combinations grows exponentially !

**NP (nondeterministic polynomial time)*

- **Outcomes:**

- *Reduced Planning time / effort*
- *Fairness*
- *Happy(er) scientists*



“NURSE SCHEDULING PROBLEM”

- Well Researched in Operations Research

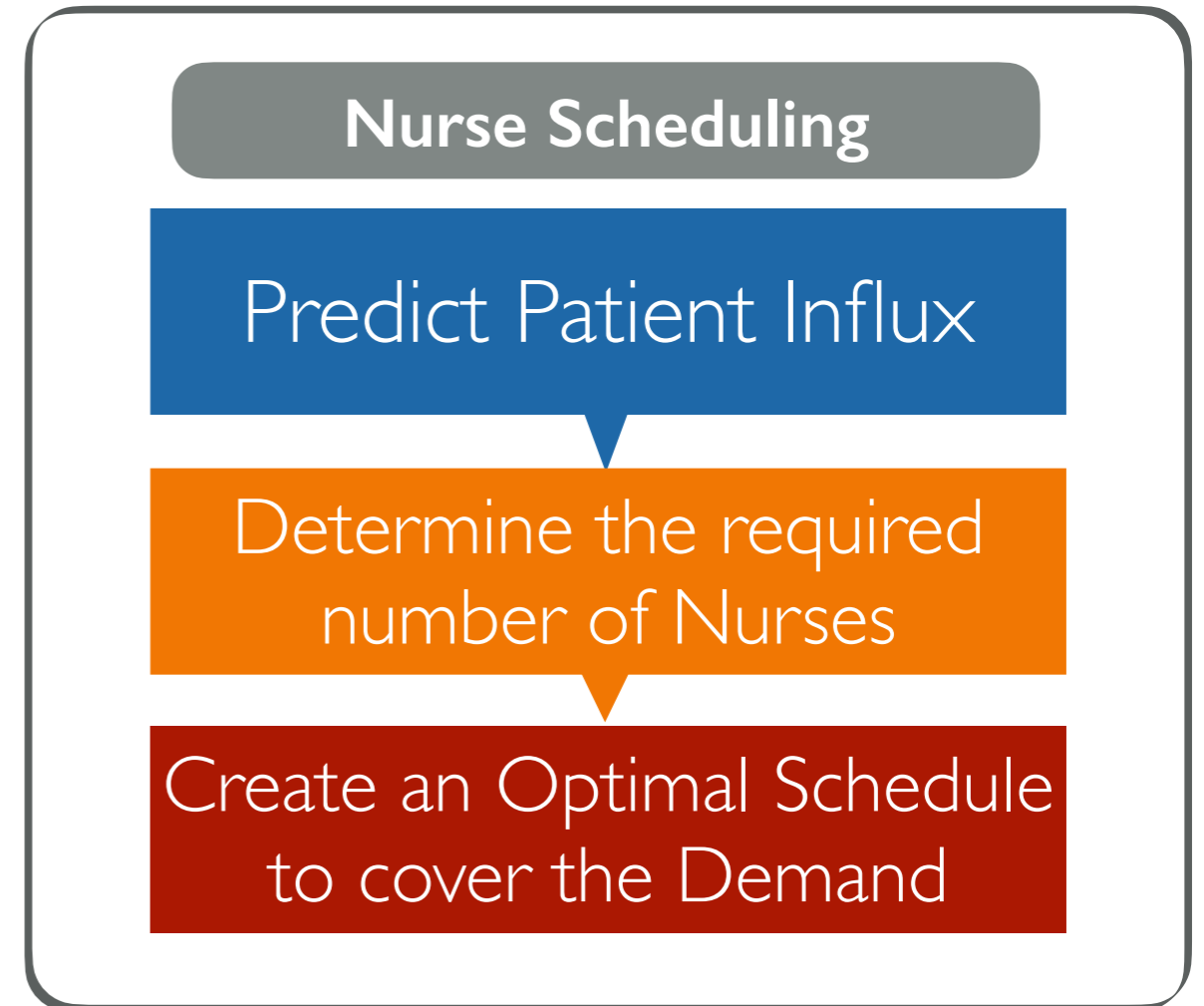
- *Nurse Scheduling*
- *Pilots / drivers*
- *Restaurant staff*
- *Emergency services workers*

- Re-Scheduling:

- *Dynamic Optimisation*
- *Continuous Planning*

- Approaches:

- *Integer Programming*
- *Constraint Programming*
- *Heuristics*



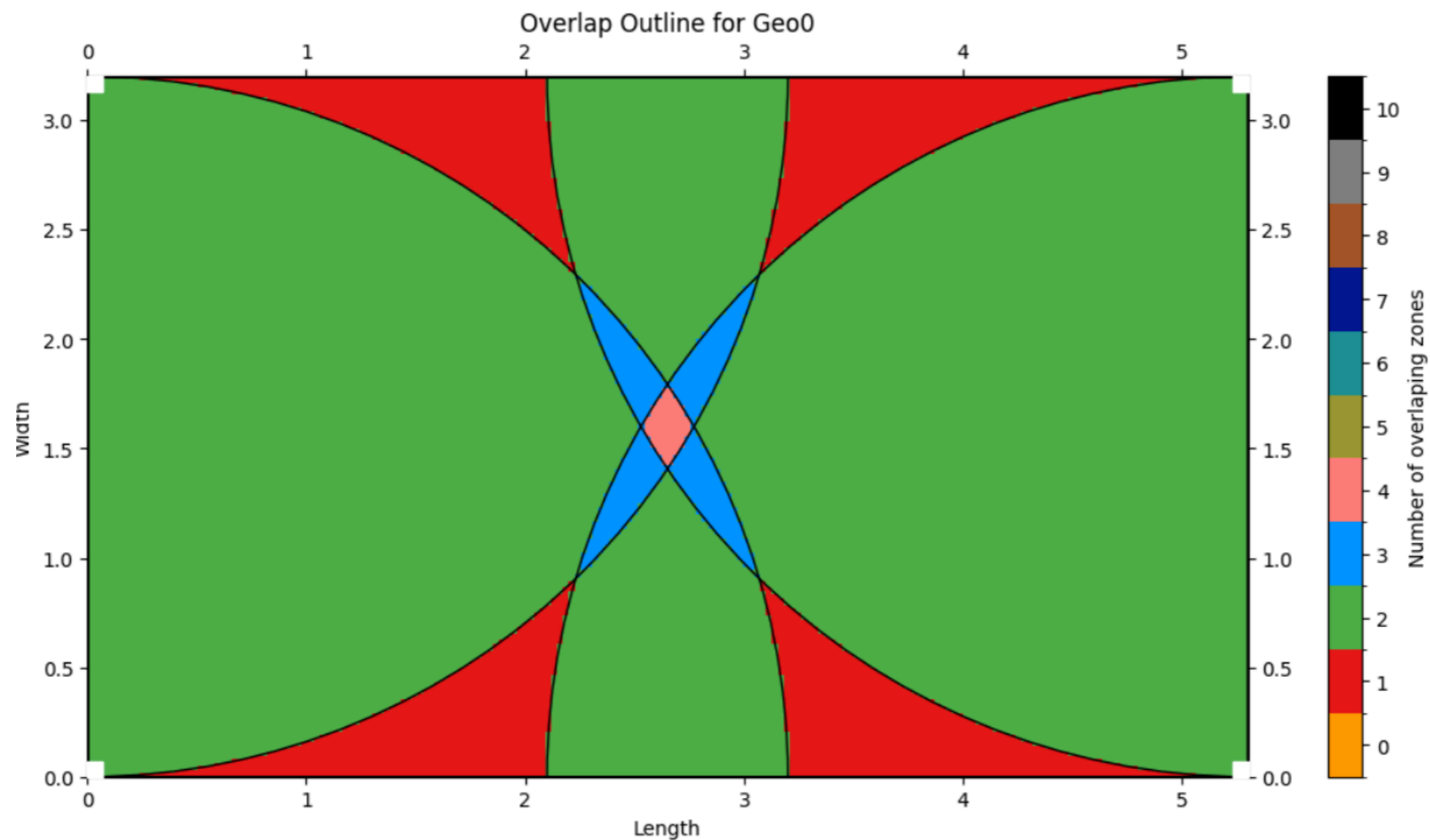
Nurse (Specialty)	Mon	Tue	Wed	Thu	Fri	Sat	Sun
Alice 🧑‍⚕️	🟢🟡	✖	🟢🟡	✖	🟢🟡	✖	✖
Mark 🧑‍⚕️	🟢🟡	✖	✖	🟢🟡	✖	🟢🟡	✖
Emily 🧑‍⚕️	✖	🟡🟤	✖	🟡🟤	✖	🟡🟤	✖
Robert ❤️	✖	🟡	✖	🟡	✖	🟡	✖
Sophia 🦷	🟤	✖	🟤	✖	🟤	✖	🟤

DETECTOR DESIGN

- **The Need:**

- *Find the best geometry / material combinations*
- *Maximise the detector sensitivity / coverage.*
- *Constrained by physical properties / geometry / cost.*

- **A toy example: find optimal coverage by circle segments:**

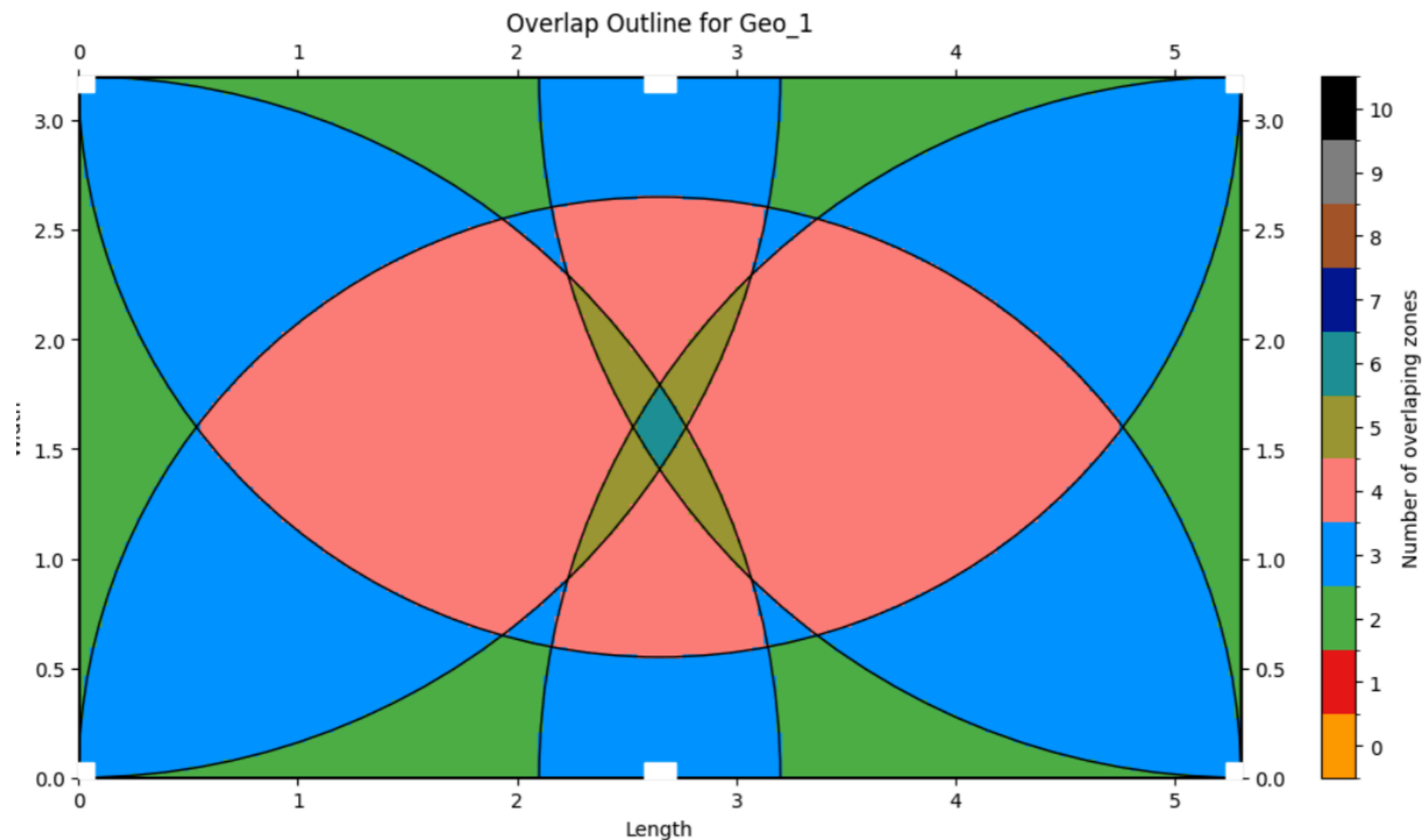


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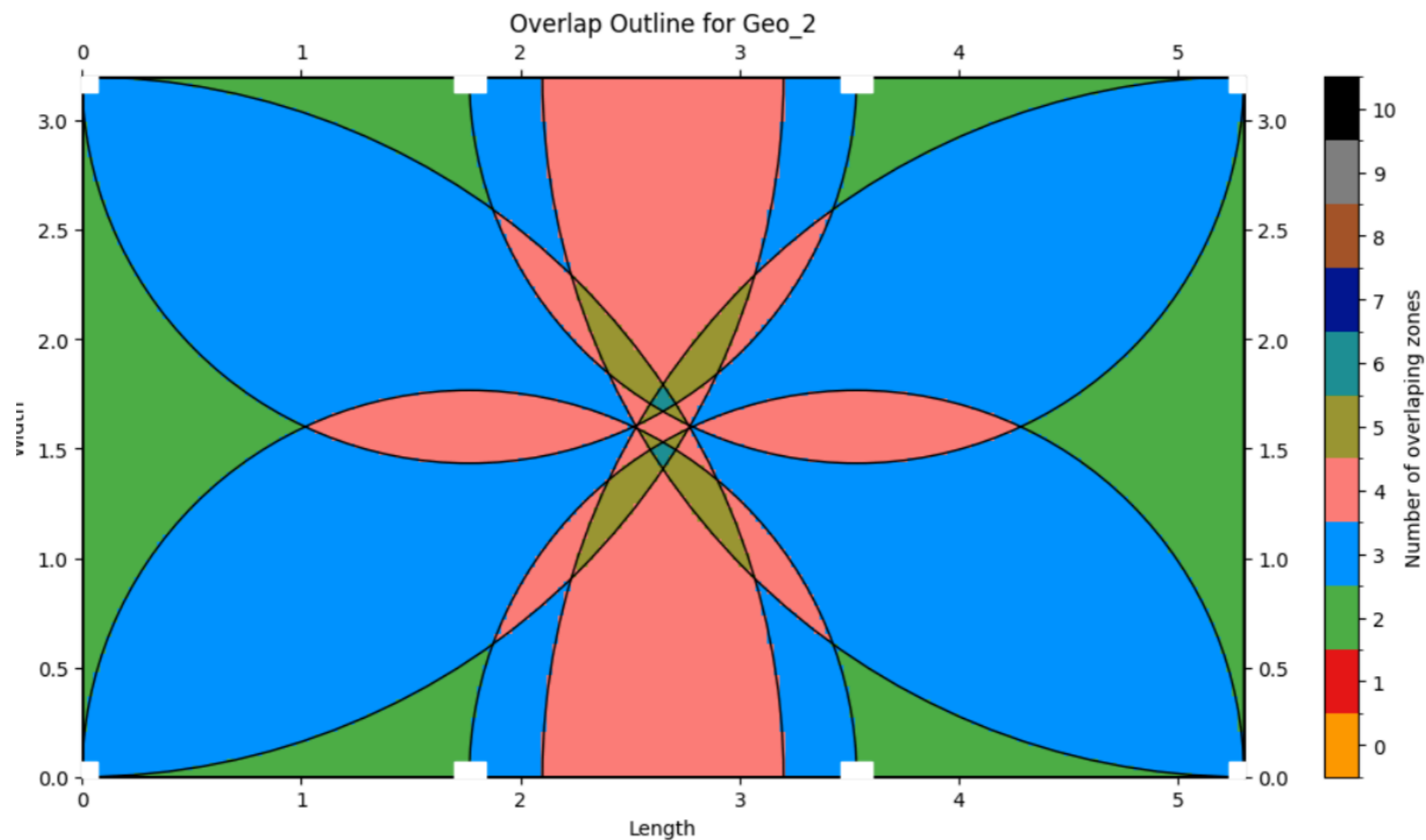


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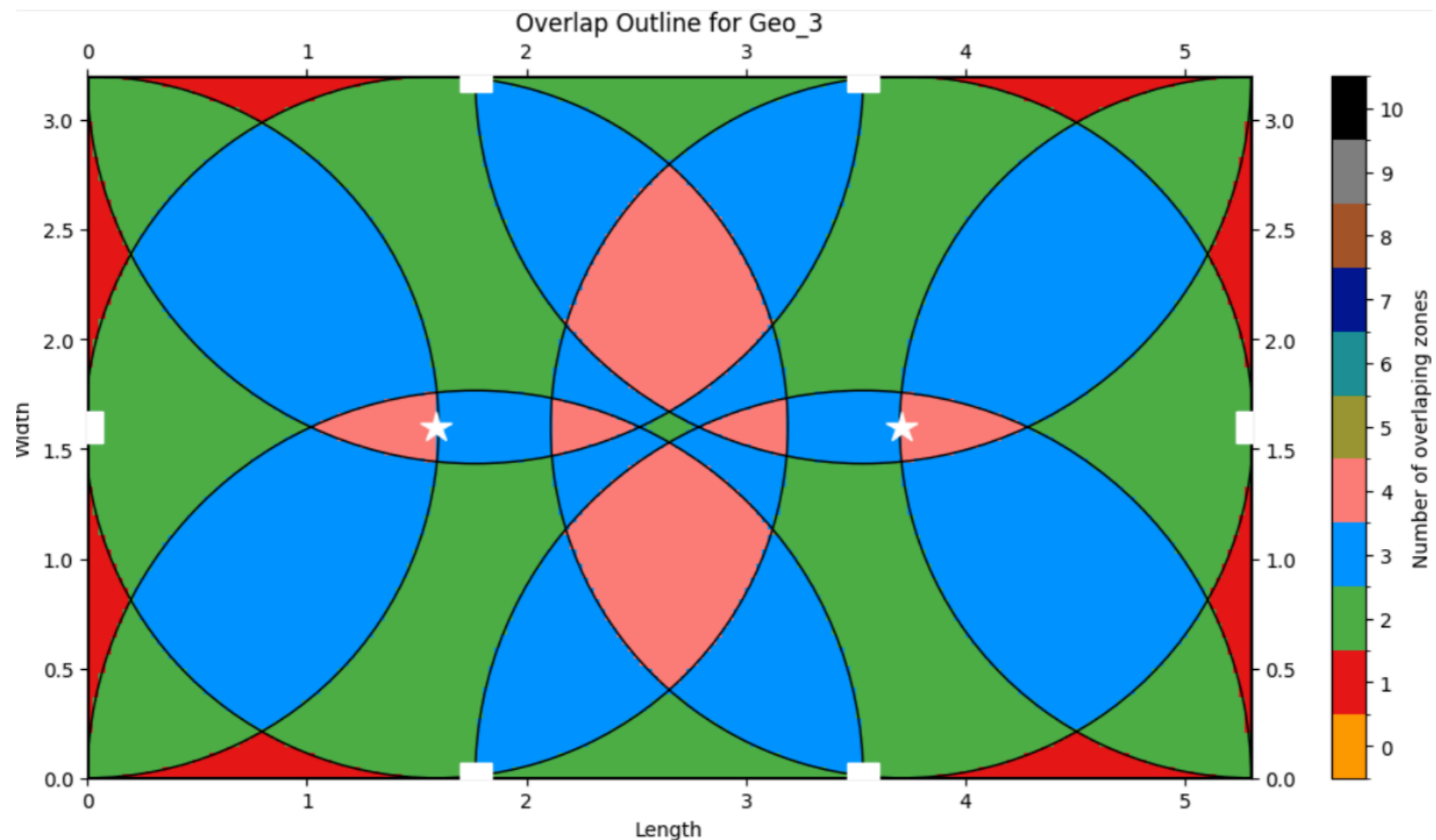


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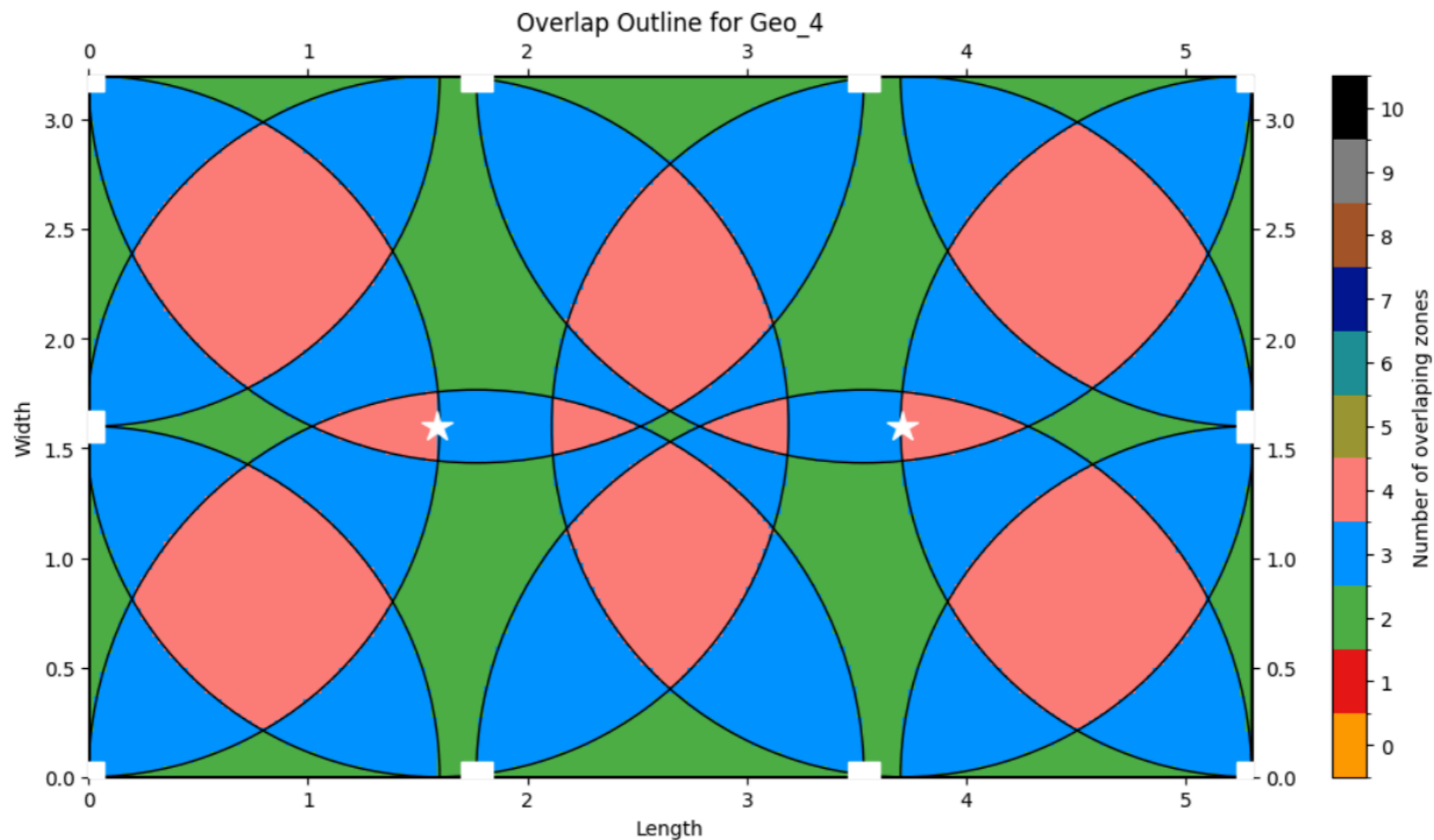


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Thank you!