







## AGI-Elo: How far are we from mastering a task?

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Shuo SUN 21 Oct 2025

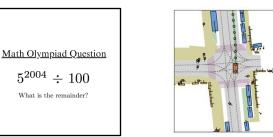
## **Motivation**

Understanding Task Difficulty

How difficult is a task (to AI or Human)?









Tasks

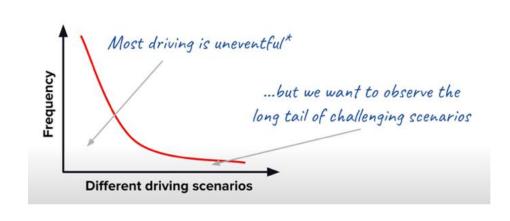
Playing Chess

Detecting Objects

Answering Questions

Driving Vehicles

- Have current SOTA models fully mastered a task?
  - Aggregate metrics ≠ Expected performance
  - Easy and difficult cases should not be treated equally
  - Data distribution is important
  - Quantifying the notion of difficulty is the key!

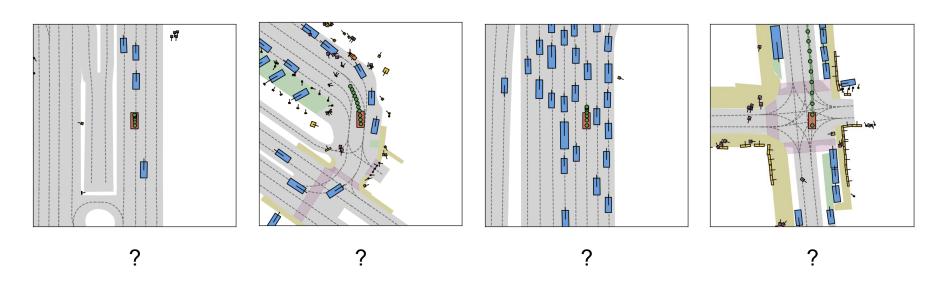


# Research Gap

Difficulty

## Understanding Test Case Difficulty

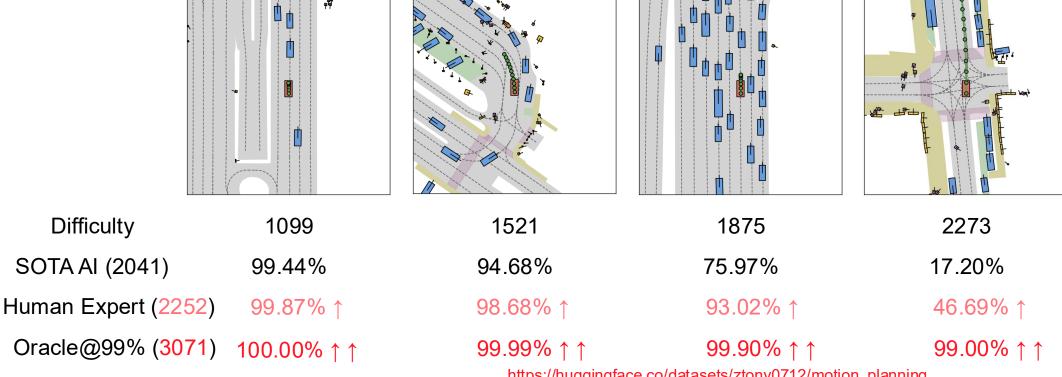
- What is the difficulty of a test case within a task (or dataset)?
- What is the competency of an Al model (or a human) on a given task?
- How far are the current SOTA models from fully mastering a task?



# Research Gap

### Understanding Test Case Difficulty

- What is the difficulty of a test case within a task (or dataset)?
- What is the competency of an Al model (or a human) on a given task?
- How far are the current SOTA models from fully mastering a task?

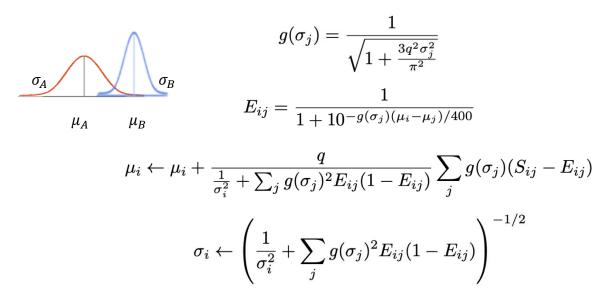


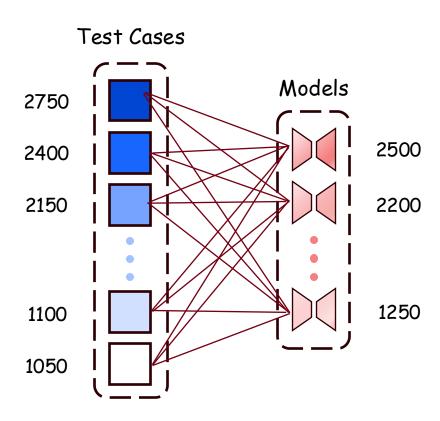
https://huggingface.co/datasets/ztony0712/motion\_planning

# **Rating System**

### AGI-Elo: Test Cases vs. Agents

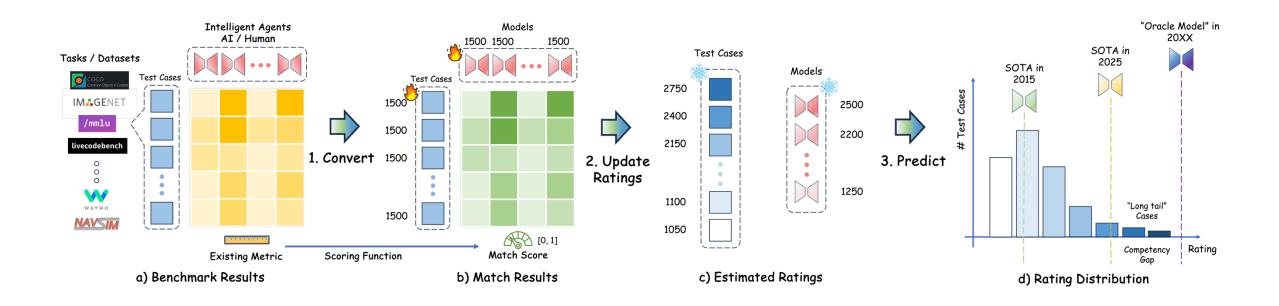
- Jointly estimate the relative difficulty and competency:
  - Record model performance on each test case
  - Convert performance metric to match scores
  - Update ratings based on pairwise match outcomes
- Rating update (Glicko System):





## Method

## Test Case vs. Agents



## **Findings**

## Rating Distribution

#### Vision Task:

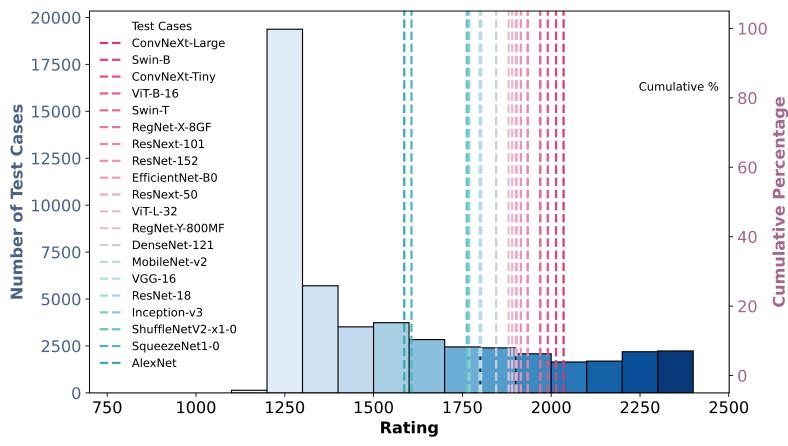
Image Classification

#### Dataset:

ImageNet val set

### Experiment:

- 5,000 test cases
- 20 models
- 1,000,000 matches







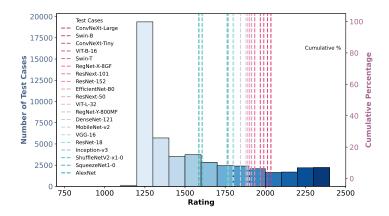
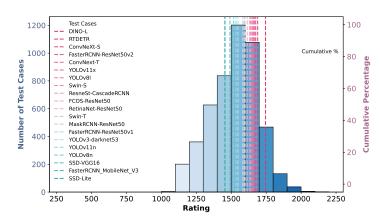
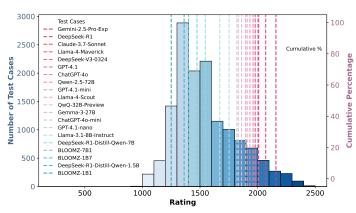


Image Classification: ImageNet

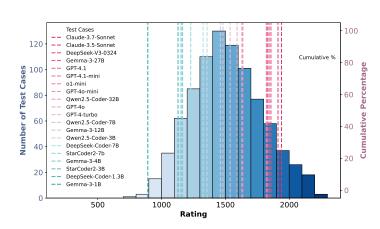


Object Detection: COCO



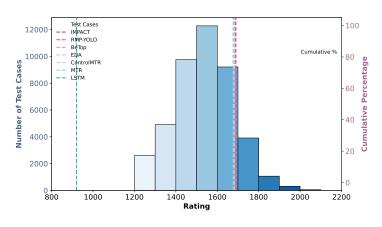


**Question Answering: MMLU** 

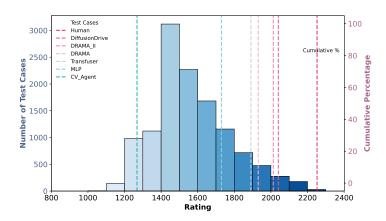


Code Generation: LiveCodeBench





Motion Prediction: Waymo



Motion Planning: NAVSIM

## Conclusion

AGI-Elo provides a robust, predictive, and comprehensive framework for evaluating AGI capabilities and limitations. By jointly modeling test case difficulty and model competency, we enable a deeper understanding of AI performance and trustworthiness across diverse domains.

### **Difficutly-aware Evaluation**

Move beyond aggregate metrics to gain fine-grained insights into task difficulty distributions and model progression across vision, language, and action domains.

### **Task-agnostic Benchmarking**

Provide researchers and practitioners with a principled, task-agnostic framework for evaluating and comparing AI systems across diverse benchmarks.

### **Predictive Insights**

Quantitatively predict agent performance on individual test cases and identify long-tail challenges that require future progress for full task mastery.

### **Gap to Full Task Mastery**

Identify outstanding challenges and competency gaps that remain on the path to achieving full task mastery at defined confidence levels.

#### **Open Source & Community**

Code and evaluated test case ratings on all six datasets are publicly available

https://ss47816.github.io/AGI-Elo/

