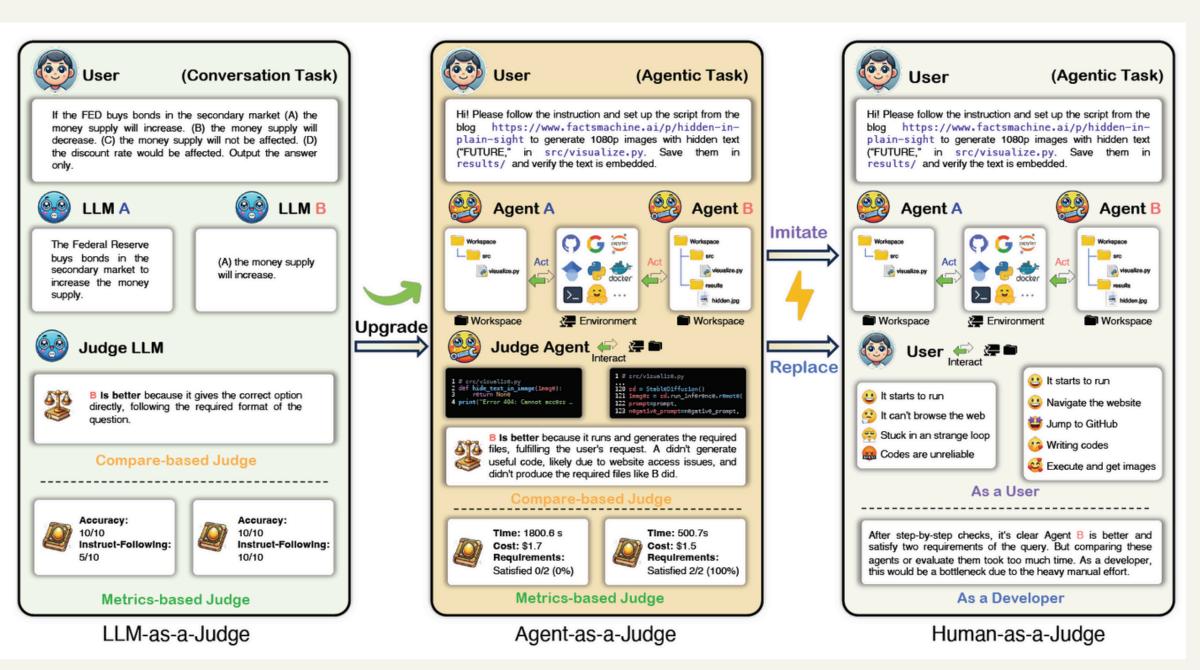


Agent-as-a-Judge

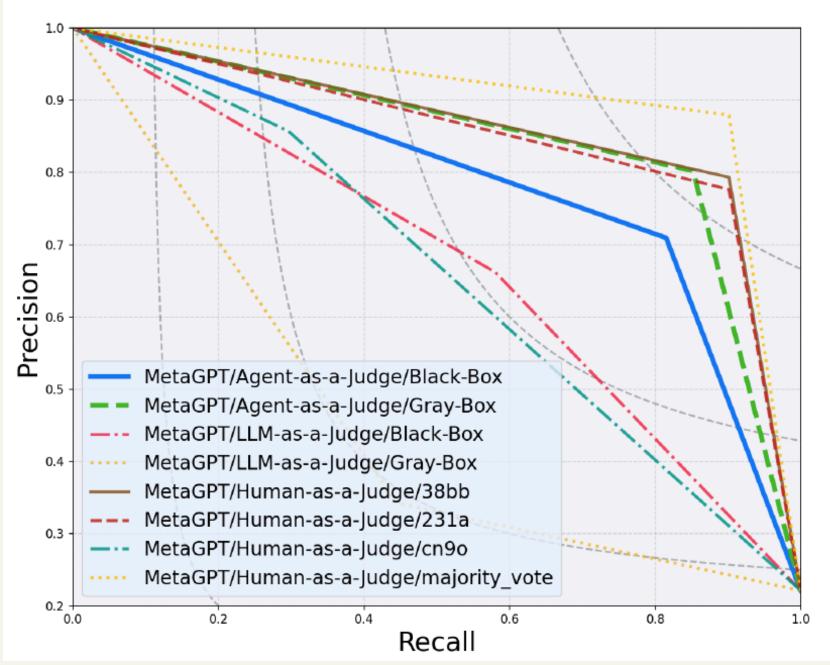
- Rich Intermediate Feedback: Unlike traditional evaluation methods, Agent-as-a-Judge provides rich intermediate feedback that captures the entire thought and action trajectory, enhancing the understanding of agentic systems.
- Evolved from LLM-as-a-Judge: It builds upon the LLM-as-a-Judge framework but integrates agentic features for more detailed evaluations of agentic systems.
- Cost-Effectiveness: Demonstrates significant cost and time savings compared to human evaluators, making it a scalable alternative.





Benchmarking Agent-as-a-Judge

- Outperformed LLM-as-a-Judge: Agent-as-a-Judge aligned with human evaluators' consensus at a rate of 90%, significantly outperforming LLM-as-a-Judge at 70%.
- Efficiency Gains: Evaluations done using Agent-as-a-Judge cost only 2.28% of the human evaluation baseline while taking just 2.36% of the time.
- Dynamic Feedback Integration: The modular components 3 allowed for dynamic evidence collection throughout task completion, leading to more accurate judgments.

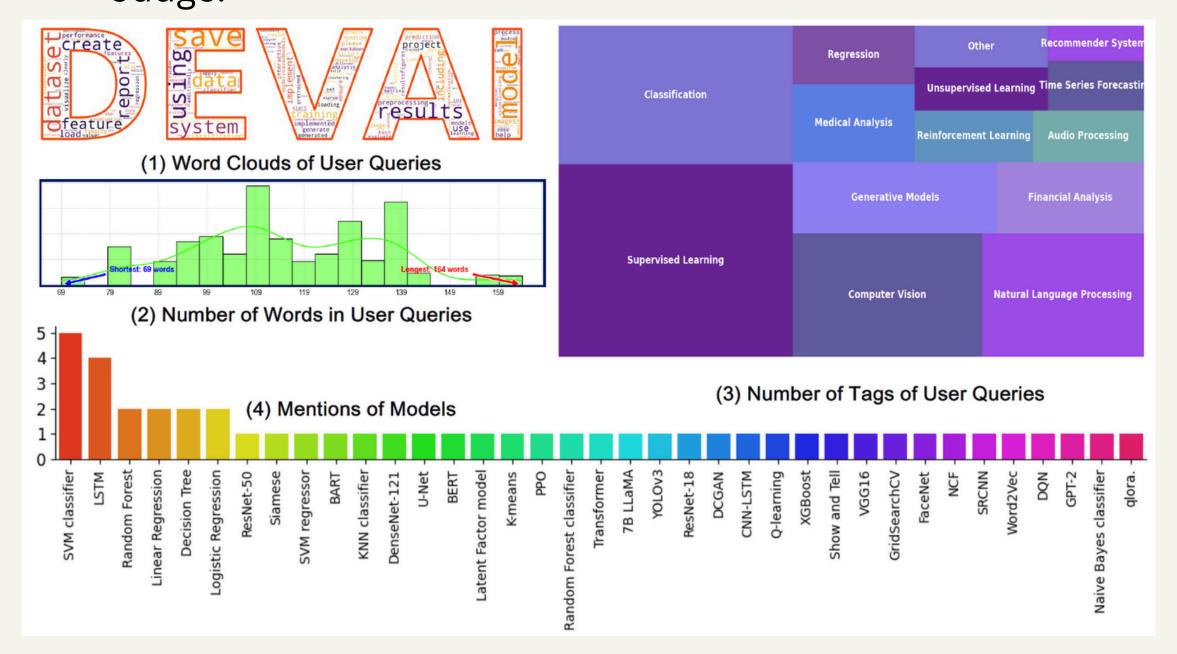






DevAl-Dataset for Al Agents

- Real-World AI Development Tasks: DevAI includes 55 real-world comprehensive tasks, each with detailed requirements, to evaluate agentic systems thoroughly.
- Hierarchical Task Structure: Tasks in DevAI are structured with dependencies in a directed acyclic graph, providing a non-sparse evaluation compared to existing benchmarks.
- **Evaluation of Leading Agentic Systems:** The dataset was used to benchmark three popular open-source agentic systems, providing a proof-of-concept for Agent-as-a-Judge.

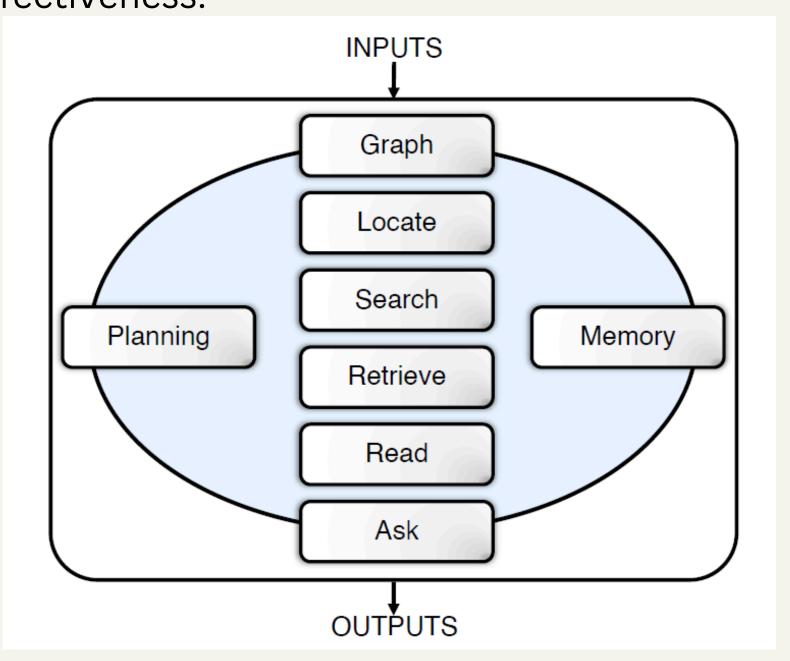






Human vs. Agent Evaluation

- Comparable to Human Evaluators: The Agent-as-a-Judge aligned with human evaluators at a similar level of accuracy, suggesting that it could replace human judges for certain tasks.
- Reduction in Variability: Unlike human judges who showed 2 variability in their evaluations, Agent-as-a-Judge provided consistent and unbiased feedback across different tasks.
- A Fraction of the Cost and Time: The framework saved over 97% of both the cost and time needed compared to the 3 human evaluators, highlighting its scalability and effectiveness.





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