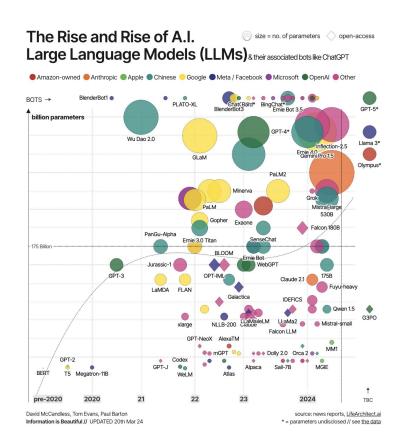
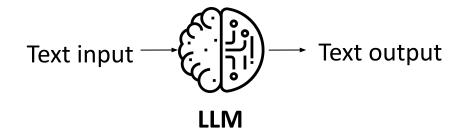
CS 294/194-196: Large Language Model Agents

Teaching Staff

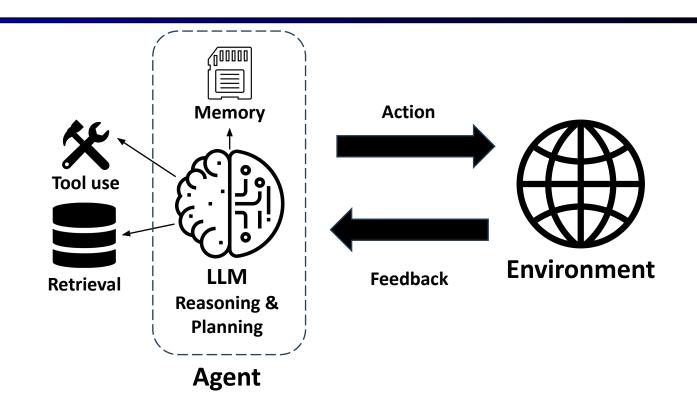
- Instructor: Prof. Dawn Song
- (guest) Co-instructor: Dr. Xinyun Chen
- GSIs: Alex Pan & Sehoon Kim
- Readers: Tara Pande & Ashwin Dara

Accelerated development of large language models (LLMs)

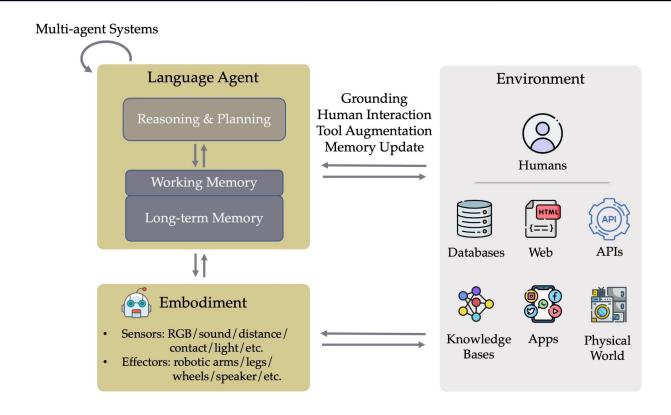




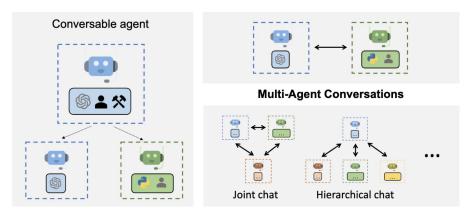
LLM agents: enabling LLMs to interact with the environment



LLM Agents in Diverse Environments



Multi-agent collaboration: division of labor for complex tasks



Agent Customization

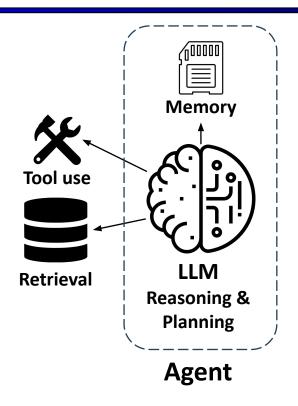
Flexible Conversation Patterns

Specialized agents for different subtasks Autogen, CrewAl, CAMEL, Mixture-of-Agents,...



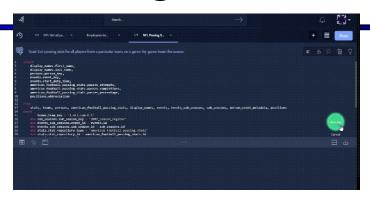
Emergence of social behaviors with role-play LLMs
Generative agents, Project Sid,...

Why empowering LLMs with the agent framework



- Solving real-world tasks typically involves a trial-and-error process
- Leveraging external tools and retrieving from external knowledge expand LLM's capabilities
- Agent workflow facilitates complex tasks
 - Task decomposition
 - Allocation of subtasks to specialized modules
 - Division of labor for project collaboration
 - Multi-agent generation inspires better responses

LLM agents transformed various applications



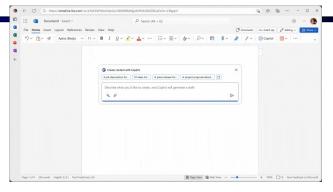
Code generation

Cursor, GitHub Copilot, Devin, Replit,...



Personal assistant

Google Astra, OpenAI GPT-40,...



Workflow automation

Microsoft Copilot, Multi-On,...



Robotics

Figure AI, Tesla Optimus,...

Education

- Law
- Finance
- Healthcare
- Cybersecurity

. . .

LLM agents are improving

04/2024

05/2024

05/2024

09/2023

04/2024

03/2024

10/2023

01/2024

ont-40-2024-05-13

Patel et al + 2024

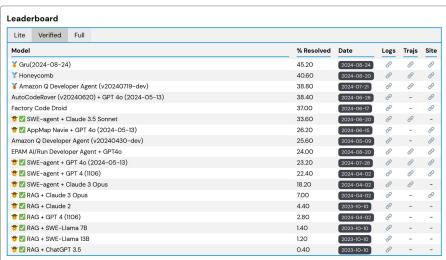
Llama3-chat-708

Agent Flan

Agenti M-70b

AgentLM-7b

Codel Jama-instruct-7b



SWE-bench Lite is a subset of SWE-bench that's been curated to make evaluation less costly and more accessible [Post].

SWE-bench Verified is a human annotator filtered subset that has been deemed to have a ceiling of 100% resolution rate [Post].

- The % Resolved metric refers to the percentage of SWE-bench instances (2294 for test, 500 for verified, 300 for lite) that were resolved by the model.
- Checked indicates that we, the SWE-bench team, received access to the system and were able to reproduce the patch generations.
- 😻 Open refers to submissions that have open-source code. This does not necessarily mean the underlying model is open-source.
- The leaderboard is updated once a week on Monday.
- If you would like to submit your model to the leaderboard, please check the submission page.
- All submissions are Pass@1, do not use hints text, and are in the unassisted setting.

SWE-Bench (Jimenez*, Yang*, et al.) swebench.com



Patel et al + 2024

Llama3

Agent Flan

Agent Tuning

Agent Tuning

WebArena

WebArena Team

WebArena

Patel et al + 2024

Patel et al + 2024

WebArena Team

Agent Flan

WebArena Team WebArena Team

Agent Tuning

Agent Flan

WebArena Tear

7.02

WebArena (Zhou et al.) webarena.dev

when "not achievable" hint is provided

Selected tasks by template

Challenges for LLM agent deployment in the wild

- Reasoning and planning
 - LLM agents tend to make mistakes when performing complex tasks end-to-end
- Embodiment and learning from environment feedback
 - LLM agents are not yet efficient at recovering from mistakes for long-horizon tasks
 - Continuous learning, self-improvement
 - Multimodal understanding, grounding and world models
- Multi-agent learning, theory of mind
- Safety and privacy
 - LLMs are susceptible to adversarial attacks, can emit harmful messages and leak private data
- Human-agent interaction, ethics
 - How to effectively control the LLM agent behavior, and design the interaction mode between humans and LLM agents

Topics covered in this course

- Model core capabilities
 - Reasoning
 - Planning
 - Multimodal understanding
- LLM agent frameworks
 - Workflow design
 - Tool use
 - Retrieval-augmented generation
 - Multi-agent systems
- Applications
 - Software development
 - Workflow automation
 - Multimodal applications
 - Enterprise applications
- Safety and ethics

Large Language Model Agents MOOC





Dawn Song

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UNIVERSITY OF CALIFORNIA



Xinyun Chen

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Denny Zhou

DeepMind



Shunyu Yao Shunyu Yao



Chi Wang

DeepMind



Jerry Liu
LlamaIndex



Burak Gokturk Google





Graham Neubig Carnegie Mellon University



Nicolas Chapados Servicenow



Meta Al



Jim Fan

NUDIA



Percy Liang

Stanford
University



Ben Mann
ANTHROP\C



Course Work

- Weekly Reading Assignment
 - Due midnight PT Sunday before the next Monday's lecture
- 1 hands-on Lab
- Semester-long course project

Grading

lecture attendance & weekly reading assignment

+

- 1 unit: article about the topic of a lecture (at least 2 pages)
- 2 units: lab + project (implementation not required)
- 3 units: lab + project with implementation
- 4 units: lab + project with significant implementation and end-to-end demo

Grading

	1 unit	2 units	3/4 units
Participation	45%	20%	10%
Reading Summaries & Q/A	10%	4%	2%
Article	45%		
Lab		16%	8%
Project			
Proposal		10%	10%
Milestone 1		10%	10%
Milestone 2		10%	10%
Presentation		15%	15%
Report		15%	15%
Implementation			20%

Class Project

5 students per group; can be part of a hackathon (more details later)

Applications Track

Build LLM agent applications in novel domains

Benchmarks Track

Create and improve benchmarks for LLM agents

Fundamentals Track

Enhance core agent capabilities (memory, planning, tool use)

Safety Track

Address safety concerns in deployment (misuse, privacy, etc.)

Decentralized and Multi-agent Track

Enhance decentralized multi-agent systems

Timeline

	Released	Due
Project group formation	9/9	9/16
Project proposal	9/16	9/30
Lab	9/23	10/7
Project milestone #1	10/8	10/21
Project milestone #2	10/29	11/18
Project final presentation	11/19	12/12
Project final report	11/19	12/12