

# Agents, Inference and Hardware

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#### **Technical Trends In Al**

- On-device Al. Instead of running an LLM in the cloud, run it on your own laptop, phone or industrial PC.
- Image/Video analysis. LLMs brought us the text processing revolution. The visual processing revolution is coming – not just generation, but analysis. This will affect, manufacturing, life sciences, self-driving, retail, etc.
- Al Agentic Workflows. Given an instruction ("research topic X for me")
  software that can carry out a sequence of steps to generate a result.

# **LLM-Based Agents**

#### Non-agentic workflow (zero-shot)

Please type out an essay on topic X from start to finish in one go, without using backspace.

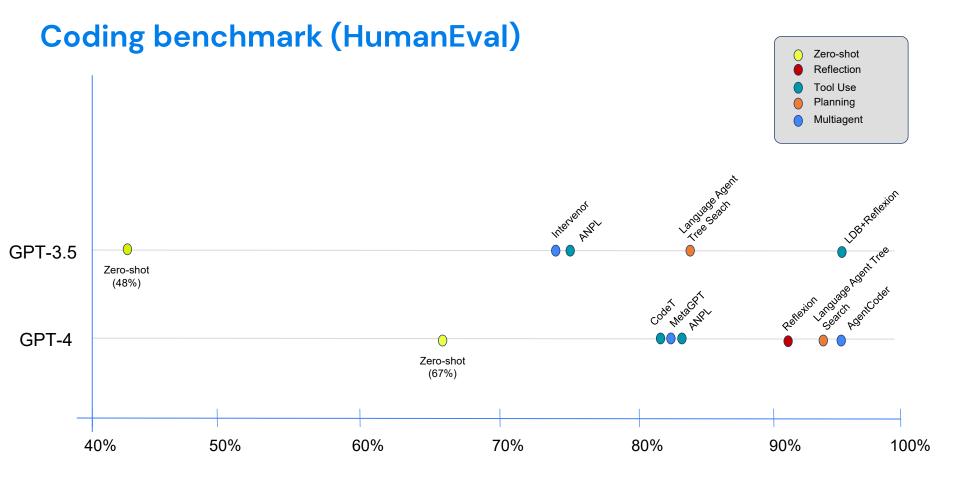


#### Agentic workflow

- Write an essay outline on topic X
- Do you need any web research?
- Write a first draft.
- Consider what parts need revision or more research.
- Revise your draft

. . . .

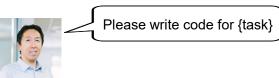


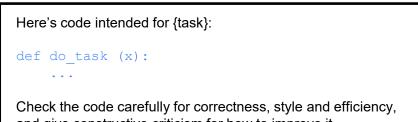


[Thanks to Joaquin Dominguez and John Santerre (DeepLearning.AI) for help with analysis.]

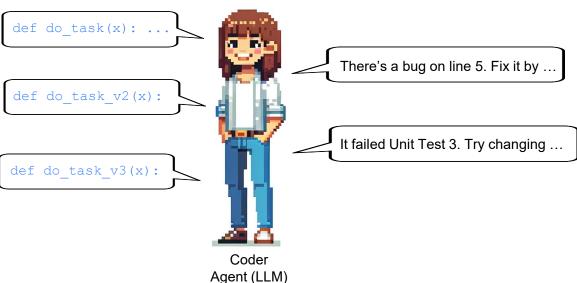
**&** Landing Al

#### **Reflection with LLMs**





and give constructive criticism for how to improve it.

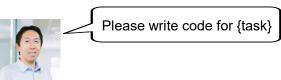


#### Related work:

- Self-Refine: Iterative Refinement with Self-Feedback, Madaan et al. (2023)
- Reflexion: Language Agents with Verbal Reinforcement Learning, Shinn et al., (2023)



#### **Reflection with LLMs**



Here's code intended for {task}:

```
def do_task (x):
```

Check the code carefully for correctness, style and efficiency, and give constructive criticism for how to improve it.



There's a bug on line 5. Fix it by ...

It failed Unit Test 3. Try changing ..



#### Recommended reading:

- Self-Refine: Iterative Refinement with Self-Feedback, Madaan et al. (2023)
- Reflexion: Language Agents with Verbal Reinforcement Learning, Shinn et al., (2023)

# **Agentic Reasoning Design Patterns**

1. Reflection

2. Multi-agent collaboration

3. Tool use (API calls)

4. Planning (decide on steps for task)

# The importance of inference

- 1. Agentic workloads use many more tokens than zero-shot prompting.
- 2. Fast, low cost token generation will be a huge performance driver. Open weight models (like Llama 3.1) also make it easier for providers to compete directly on inference price and speed.
- 3. Training remains important, but I hope our community will also invest significantly in inference. (E.g., SambaNova, Cerebras, Groq). Also see benchmarks at artificialanalysis.ai

## **LMM-Based Agents**

#### Non-agentic workflow (zero-shot)

Watch this video and tell me if any surfer is within 10m of a shark



#### Agentic workflow

- Detect the sharks (bounding boxes)
- Detect the surfers (bounding boxes)
- 3. Compute distances between sharks & surfers
- 4. Determine if any are <10m
- 5. Iterate through Steps 1-4 for all frames in video



Video credit: Crazy Nature [youtu.be/2\_HfdtljBKk?si=3rKt7StDATeuFYjS]





### Demo:

# Mercedes Logo



# **Tool use: Computer Vision Models**

+222.9% Y/Y

 $144 \rightarrow 465$ 

+186.4% Y/Y

<200 → 842

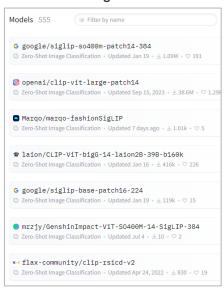
+384.2% Y/Y

 $514 \rightarrow 2.5k$ 

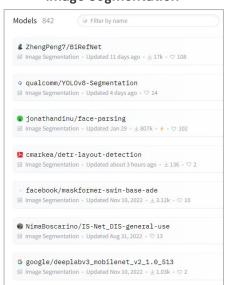
+214.3% Y/Y

 $4.3k \rightarrow 13.6k$ 

#### **Zero-Shot Image Classification**



#### **Image Segmentation**



#### **Object Detection**



#### **Image Classification**



Source: Huggingface.co





#### Resources

- Try VisionAgent for free at va.landing.ai
- Open source code on github

LandingAl team here today



**Steve Ackley** 



**Ian Rysdale** 





# Fireside Chat

Andrew Ng
Executive Chairman, LandingAl

Baskar Sridharan
VP of AI/ML services & infrastructure at AWS at

