



AI Adoption Training

Workshop#5: Advanced AI Topics – Coding,
Models and Agents

December 12th, 2025



Today's Agenda

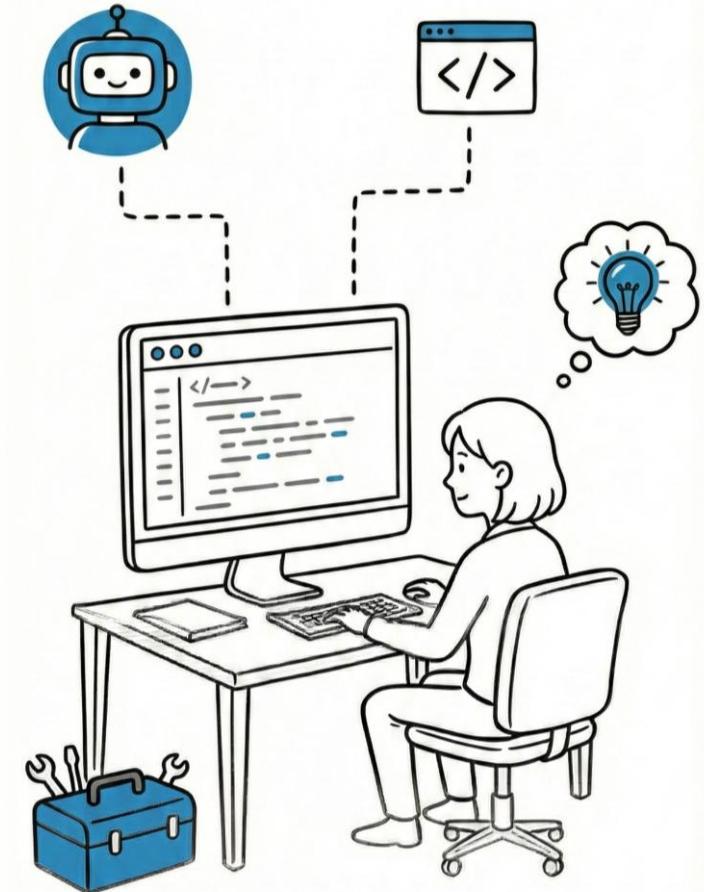
1. **Intro & Context** (5 mins)
2. **Understanding LLM Coding & Agents** (20 mins)
3. **LLM Coding Techniques** (30 mins)
4. **Agents Beyond Coding** (10 minutes)
5. **Learnings and Additional Resources** (5 mins)
6. **Wrap Up** (5 mins)

Materials are available at: https://github.com/mullinsean/ai_training

Goals for Today's Session

Exposure to AI coding tools, not mastery

- Goal: Understand what is possible with advanced applications of LLM tools
 - Admittedly, more complex than previous four sessions
 - For those that have some coding knowledge: provide a roadmap for how to use LLM tools to dramatically enhance your productivity
 - For those who do not have prior coding knowledge: provide an overview of what is possible and how to edge into this space
 - Provide resources for further learning



The LLM coding revolution

No industry is changing faster due to LLMs than software engineering

- In 2024, 25% of all code at Google was generated by LLM
- Cursor (AI coding tool) has grown from \$0 to >\$1B in annual revenue since 2023
- Anthropic has disclosed that more than half of their revenue (estimated at \$10B for 2025) is coming from coding and software engineering tasks
 - Claude Code (launched in Feb 2025) is already generating more than \$1B in annual revenue
- Early evidence that demand for *junior* software engineers is declining
- Performance on software benchmarks are becoming key factors in evaluating new frontier models (eg: Claude Sonnet 4.5, Gemini 3.0, ChatGPT 5.1)

"The future is already here – it's just not evenly distributed."

– William Gibson

Code + Tools: The Language of Computers

LLMs as the bridge between intention and execution

- Problem: Often, prompts written in English are not specific enough for LLMs to implement sophisticated, detailed tasks (eg: build an economic model of X)
 - How do you know what the LLM is doing “under the hood” of a task?
Important for verification and reproducibility
 - For repeated problems, we may not want to reinvent the solution each time
- LLMs writing code can solve for these problems:
 - Code is the “language” of how we specify to computers how to execute tasks
 - LLMs are language experts → this especially includes software code!
 - Code be reviewed, edited, shared and reused; it also allows you to work on problems step-by-step
- LLMs also write code for themselves to solve problems and enhance tool use
(eg: how Claude creates and manipulates [Excel files](#))



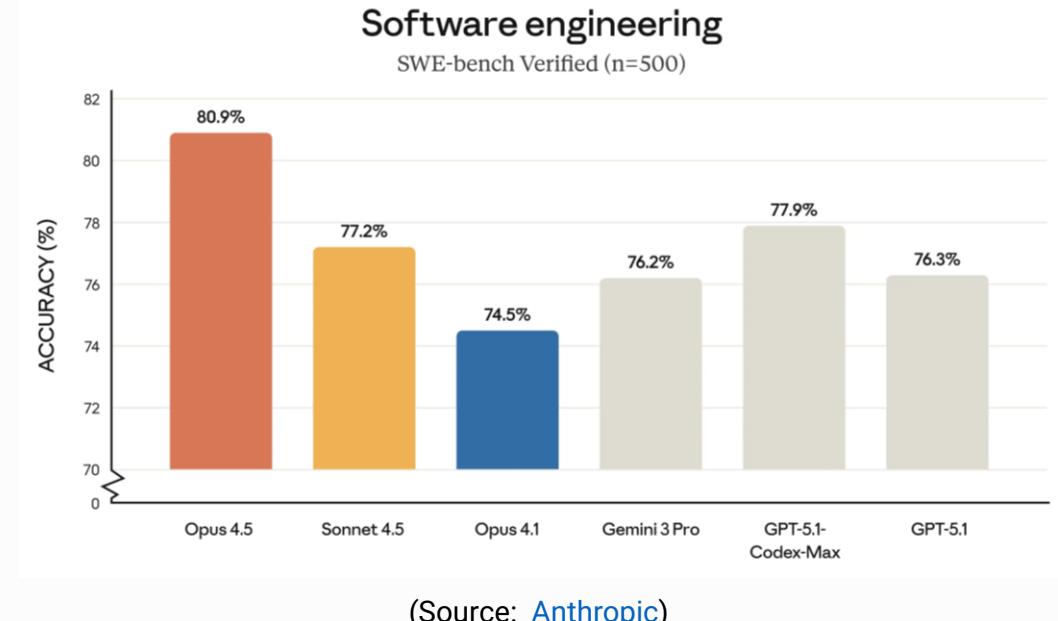
< ⌂ ⌂ G7 gdp chart .JSX

```
1 import React, { useState, useEffect } from 'react';
2 import { LineChart, Line, XAxis, YAxis, CartesianGrid, Tooltip, Legend
3
4 const G7GDPChart = () => {
5   const [chartData, setChartData] = useState([]);
6   const [loading, setLoading] = useState(true);
7   const [error, setError] = useState(null);
8
9   // 67 country codes, names, and colors
10  const countries = [
11    'USA': { name: 'United States', color: '#1f77b4' },
12    'CAN': { name: 'Canada', color: '#ff7f0e' },
13    'GBR': { name: 'United Kingdom', color: '#2ca02c' },
14    'DEU': { name: 'Germany', color: '#d62728' },
15    'FRA': { name: 'France', color: '#9467bd' },
16    'ITA': { name: 'Italy', color: '#8c564b' },
17    'JPN': { name: 'Japan', color: '#e377c2' }
18  ];
19
20  useEffect(() => {
21    const fetchData = async () => {
22      try {
23        setLoading(true);
```

The Race to Build the Best Coders...

Frontier models are being increasingly optimized for coding

- All major AI companies are now training and optimizing frontier models for coding:
 - OpenAI: GPT5.2 and GPT5.2-Codex (agentic coding)
 - Anthropic: Claude Sonnet 4.5 and Claude Opus 4.5 (complex problems)
 - Google: Gemini Pro 3.0
- Why it works so well:
 - Vast amount of publicly available code for training data; verifiable benchmarks for performance
 - Code is more structured, less ambiguous than regular language (by design)
 - AI labs use these models themselves → huge incentive to improve performance
- Emergence of specialized tools: Claude Code, Codex, Github Copilot, Cursor, Windsurf

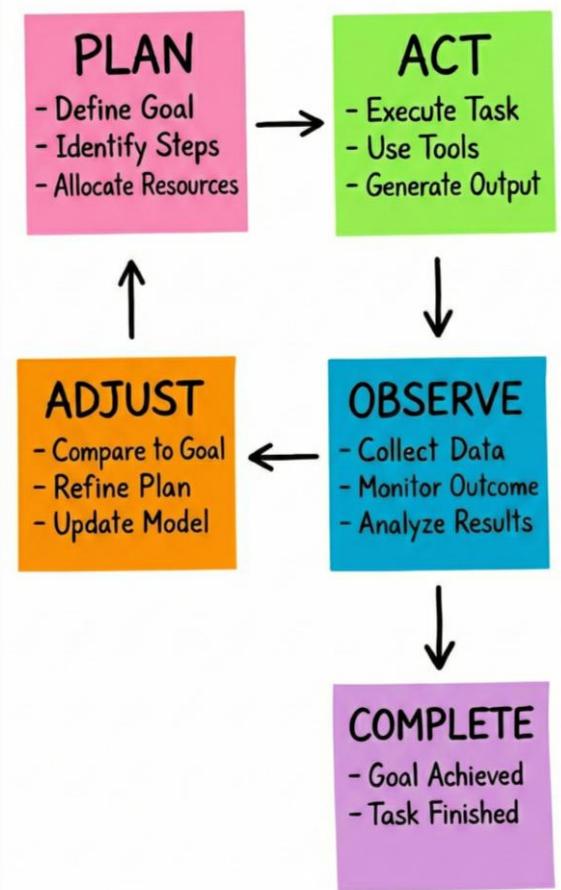


What Are AI Agents?

“LLM + tools in a loop”

- Tremendous source of hype right now: “2025 is the year of AI agents”
 - More realistically, we are looking at the “decade of agents” (Karpathy)
- AI Agent = **LLM + tools + ability to act autonomously to achieve a goal**
 - Agents can plan multi-step tasks, use tools, delegate to “sub-agents”, decide when task is finished
 - Deep Research feature is an example of an agent: submit a research request, tool goes off and conducts research, notifies you when complete
- Conceptually:
 - LLMs are *stateless*; the appearance of an interactive chat session is an illusion; in practice, they are series of API calls
 - The output of one LLM call can generate input for the next LLM call; allows for branching, monitoring outcomes and deciding when goal has been achieved.

Agentic AI Loop

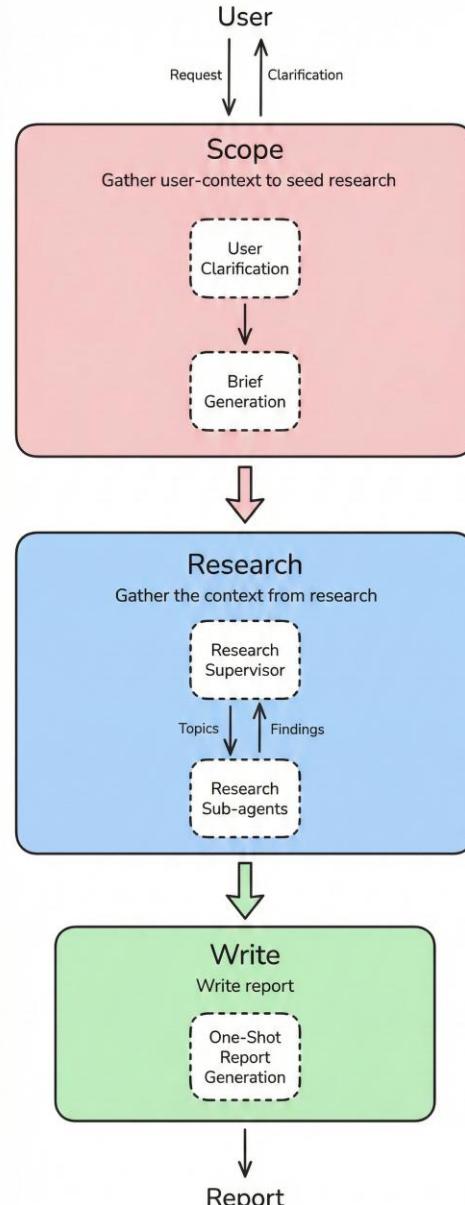


AI Agents in Action

Parallel processing and problem decomposition

- Under the hood, AI agents are just sequences of calls to LLMs:
 - They can break problems into subproblems and call multiple “subagents” in parallel
 - Each LLM call can have its own instructions and context, isolated from the rest of the process
 - Generally, one “supervisor” LLM maintains general context across the task
- Example: [Open Deep Research](#) tool by LangChain:
 1. Refine and clarify research brief with user; generate research brief
 2. Use research brief to break down task into multiple, smaller research queries; stop research when sufficient amount of information has been collected
 3. Combine findings from each subagent and write final research report

Example: Open Deep Research Agent



In Practice: Coding with LLMs

From “vibe coding” to professional software development

- Coding with LLMs is rapidly changing as the technology is continuously evolving
- Right now, three “levels” of how to code with LLMs:
 - Level 1: “Vide Coding” – coding in a chat window → copy/pasting code
 - Level 2: IDE integration – VS Code, Cursor, etc. → coding with autocomplete and light LLM tool use
 - Level 3: Advanced Agentic Coding – Claude Code/Codex, IDE for review, Git/Github for version control



Level 1: “Vibe Coding”

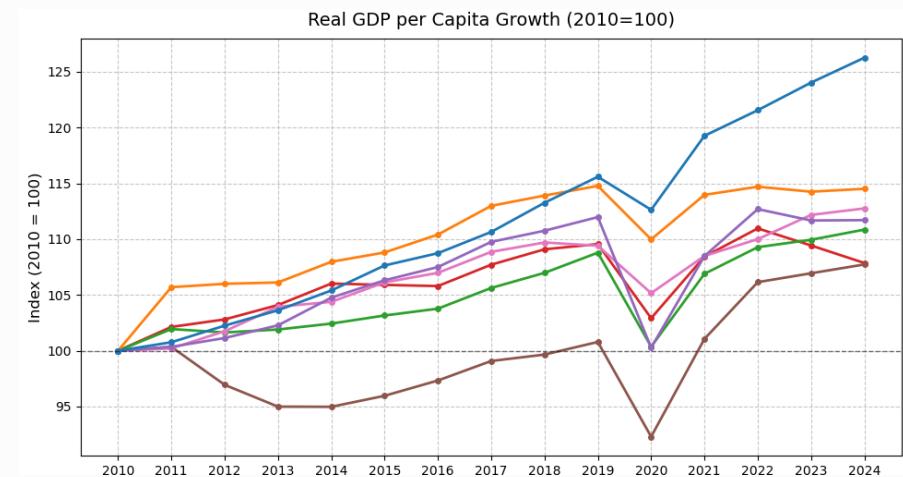
Quick scripts in the chat window

- Simplest entry point for coding: prompting the LLM to generate code directly in the chat window
 - Ask LLM to write code (in language X) to solve a particular problem
 - Inspect code
 - Copy/paste code into your environment and run
 - Interpret any error messages or bugs
 - Iterate via chat conversation
- **Good for:** small, discrete tasks; one-off scripts; testing ideas or simple prototypes
- **Limitations:** no version control; clunky to iterate and fix bugs; rapidly runs into context window limits; can only work on one file at a time;
- While useful in certain circumstances, pure “vibe coding” is generally not feasible for larger, more complex projects.

Example:

Write code in Python to plot real GDP per capita for all G7 countries from 2010 to 2025. Normalize data so 2010=100.

([chat](#), [code](#), [chart](#))



Level 2: IDE Integration

LLM-enhanced development environments

- Code in a professional IDE, but with LLM-enhanced tools: [VS Code](#), [Cursor](#), [Github Copilot](#), [Windsurf](#), Google's [Antigravity](#)
- Key features:
 - Autocomplete: simplest use; start typing code and use tab key to accept autocomplete suggestions
 - Generate code via chat window
 - Use “Plan Mode” for complex tasks
 - Accept/reject edits via “diffs”
 - Can help setup environments and other “coding overhead” requirements
- **Good for:** Medium-sized projects, learning to code, augmenting productivity of experienced coders

Example: Cursor IDE

The screenshot shows the Cursor IDE interface. On the left, there is a code editor with a Python file named `g7_gdp_chart.py`. The code imports pandas, matplotlib.pyplot, datetime, and warnings, and defines a function to fetch G7 GDP data from wbdata and pandas_datareader. It also lists the G7 countries and their ISO codes. On the right, a sidebar titled "Resolve unexpected keyword argue" provides context about the task: "write code in python to create a chart of real GDP per capita for all G7 countries from 2010 to present. use annual data. normalize data so 2015=100". Below this, a "G7 GDP Chart Implementation" section describes the goal: "Create a Python script that fetches real GDP per capita data for all G7 countries from 2010 to present, normalizes the data so 2015=100, and generates a visualization chart". A "To-dos" list includes items like "Verify and improve data fetching logic for both wbdata and pandas_datareader, ensuring annual data extraction works correctly". The bottom status bar shows the environment is "WSL: Ubuntu" and the file is "coding_example".

```
#!/usr/bin/env python3
"""
Create a chart of real GDP per capita for all G7 countries from 2010 to present.
Data is normalized so 2015 = 100.
"""

import pandas as pd
import matplotlib.pyplot as plt
from datetime import datetime
import warnings
warnings.filterwarnings('ignore')

# Try to import data readers
try:
    import pandas_datareader.data as web
    HAS_PANDAS_DATAREADER = True
except ImportError:
    HAS_PANDAS_DATAREADER = False

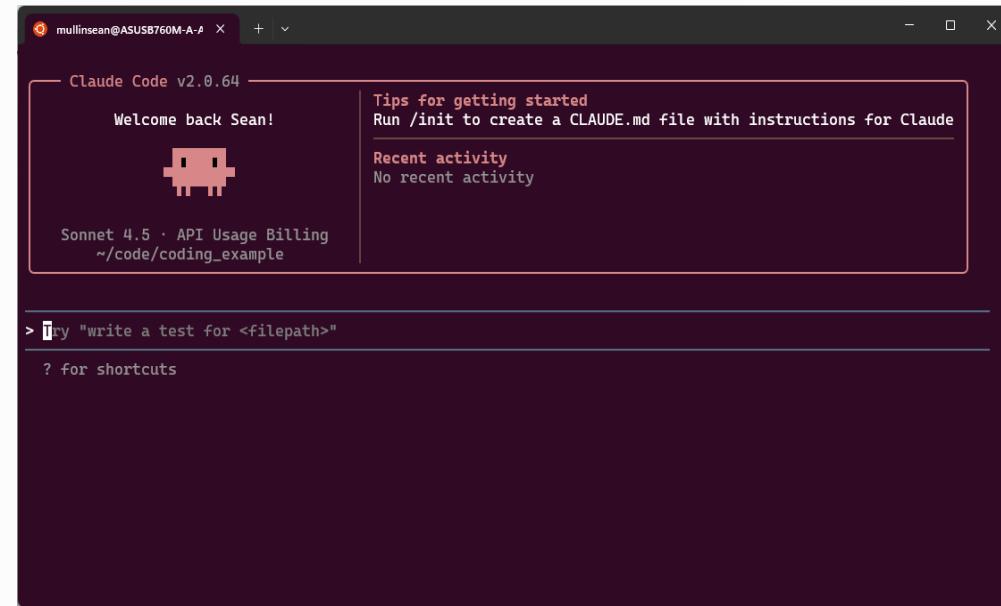
try:
    import wbdata
    HAS_WBDATA = True
except ImportError:
    HAS_WBDATA = False

# G7 countries and their ISO country codes
G7_COUNTRIES = {
    'Canada': 'CAN',
    'France': 'FRA',
    'Germany': 'DEU',
    'Italy': 'ITA',
    'Japan': 'JPN',
    'United Kingdom': 'GBR',
}
```

Level 3: Advanced Agentic Coding

Command-line agents with version control

- Command-line interface (CLI) coding agents have emerged as most powerful tools for complex, production-level codebases: [Claude Code](#) and [Codex](#) (OpenAI) are the two leading models
 - Can work over multiple files, making edits across entire code bases, run tests and debug errors
- Use Git/Github for version control – mistakes can be easily rolled back
- Features:
 - Planning mode maps out workplan for your approval before coding
 - CLAUDE.MD (or AGENTS.MD) maintains state of project and any coding style patterns or other customizations
 - Explicit management of context window
 - Choose which model to use (eg: Sonnet, Haiku, Opus)
 - Directly edits files in your filesystem; can backup to Github; run command line tools



Workflow:

Generate feature → Inspect → Test → Commit

Deep Dive into Claude Code

Professional agentic coding in action

- Example 1: Setting up a new project in Claude Code
 - Initializing environment
 - Navigating Claude Code
 - Generating code to solve a problem
 - Committing to Git
- Example 2: Making a simple change to a file
 - Fetching repo from Github
 - Requesting edits
 - Reviewing and pushing to Github
 - Merging changes
- Example 3: Adding a new sophisticated feature to existing code base
 - Using [Claude Code on the web](#) to plan and implement code change



Documentation: [Claude Code Overview](#)

Techniques for Effective LLM Coding

Best practices across all three levels

- Create “bite-sized” features:** Build incrementally, test as you go
- Frequent commits:** Version control lets you easily roll back LLM errors
- Use the planning feature with the most advanced models:** Let the LLM first write out a plan for approval before starting coding
- Parse error messages:** Use LLM to troubleshoot and explain what went wrong; use this to build your knowledge as you go
- Stay within your understanding:** Don’t try to build things beyond your knowledge level (but experimentation is okay!)
- Focus inspection on most critical parts of the code:** eg: review the core economic model code, not the chart generation or file saving boilerplate
- Stick to popular languages:** Python, Javascript/Typescript, Java, C/C++ all have vast amounts of training data on the web
- Build your knowledge gradually:** LLMs are great learning tools; expand your knowledge as you test more complex use cases for coding

Cost and Token Usage:

Using LLM-based tools to generate code can rapidly hit usage limits for most plans.

Many professional coders create API keys and get billed for coding based on [per token usage](#).

This allows costs to be allocated *per project*, if desired.

LLM Coding: Shift in Mentality

From writing code to software architect

- Coding in the AI era is now much more about **co-developing the architecture for software** with an LLM and delegating the actual coding to the model.
 - Old mentality: Write code line by line
 - New mentality: Architect solutions, guide AI implementation
- Spend more time thinking through the problem you are trying to solve and what are the potential steps to solve it, rather than coding.
- Because writing code is cheap, experiment more: try new features, build multiple prototypes, create something just to learn
 - You can always throw away code or roll back mistakes that don't work.
- Final, "mission critical" code should be carefully tested

LLM Coding Mental Process:

- Define the problem clearly
- Break it into logical pieces
- Build features one at a time
- Test and evaluate
- Integrate and refine
- Move to the next step of the problem

Using Agents Beyond Coding

Practical applications for LLM-based agents

- Agents are not just for coding tasks and have a wider range of general-purpose uses.
- Claude Code is actually a general-purpose command-line tool. For example, it can:
 - Rename all the files in a folder or find all the PDF files in range of folders and move them
 - Summarize a series of PDFs in a folder and write the output to a text file
- Platforms like [Zapier](#) or [Gumloop](#) allow you to connect “AI agents” with existing tools to automate workflows
- MS Office Co-Pilot has advanced agent features for data analysis, task automation and research within the Office suite.
- Can be built into messaging/work communication tools like [Slack](#) or MS Teams to automate notifications, communication and provide an easy way to interact with agents.

Limits of Agents

- Errors can compound over a long horizon task; reducing likelihood of a successful outcome
- Context can degrade over long horizon tasks, limiting complexity of an agentic task
- Retrieving relevant data for task is still a work in progress
- Can be expensive (ie: use a lot of tokens)

Additional Resources

Resources for further exploring LLM-assisted coding

Tools:

- [Claude Code](#), [Codex](#)
- [Cursor](#), [Windsurf](#)
- [Google AI Studio](#), [Vercel](#), [Replit](#)
- [VS Code](#), [Git](#), [GitHub](#)

Guides:

- [Claude Code](#)
- [Codex](#)

Articles and People:

- Simon Williamson ([link](#))
- Andrej Karpathy ([YouTube](#) and [X](#))
- Claude Code in Action ([course](#))
- Writing a good CLAUDE.md file ([link](#))
- How to Use Cursor - for non-technical people ([link](#))
- Social Science Tech Stack (Kevin Bryan) ([link](#))

Final Thoughts

Using LLMs for success and happiness... :)

- LLMs are powerful tools that, if used properly, can significantly **augment** your productivity
 - They are especially helpful for knowledge work.
- LLMs are not a replacement for thinking:
 - Used properly, they can allow you to spend **more time thinking critically** about the most important aspects of your work;
 - But careful review of their output is frequently necessary
- They are incredible **teaching machines**:
 - Experiment often, continuously test their capabilities and use LLMs to learn
- The technology is **changing very fast**:
 - What is not possible now, may be a simple feature in 3-6 months
- Everyone is learning what they are capable of; no one is really an expert



Thank You!



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