

CS1-LLM: Integrating LLMs into CS1 Instruction

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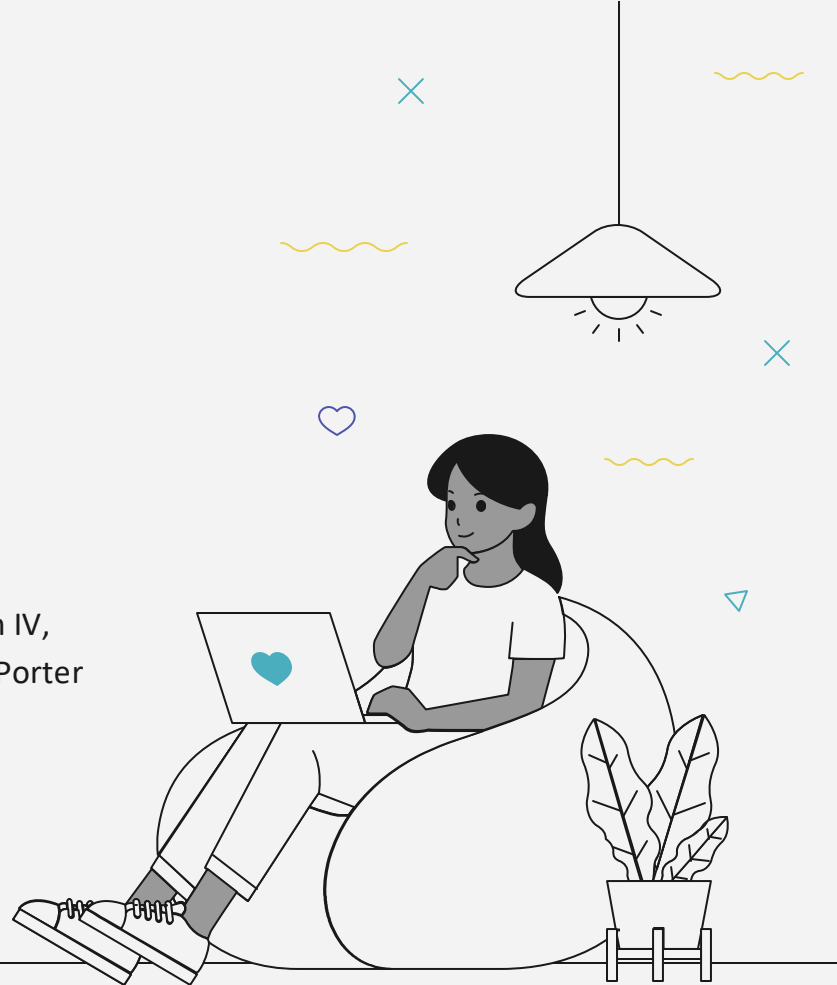




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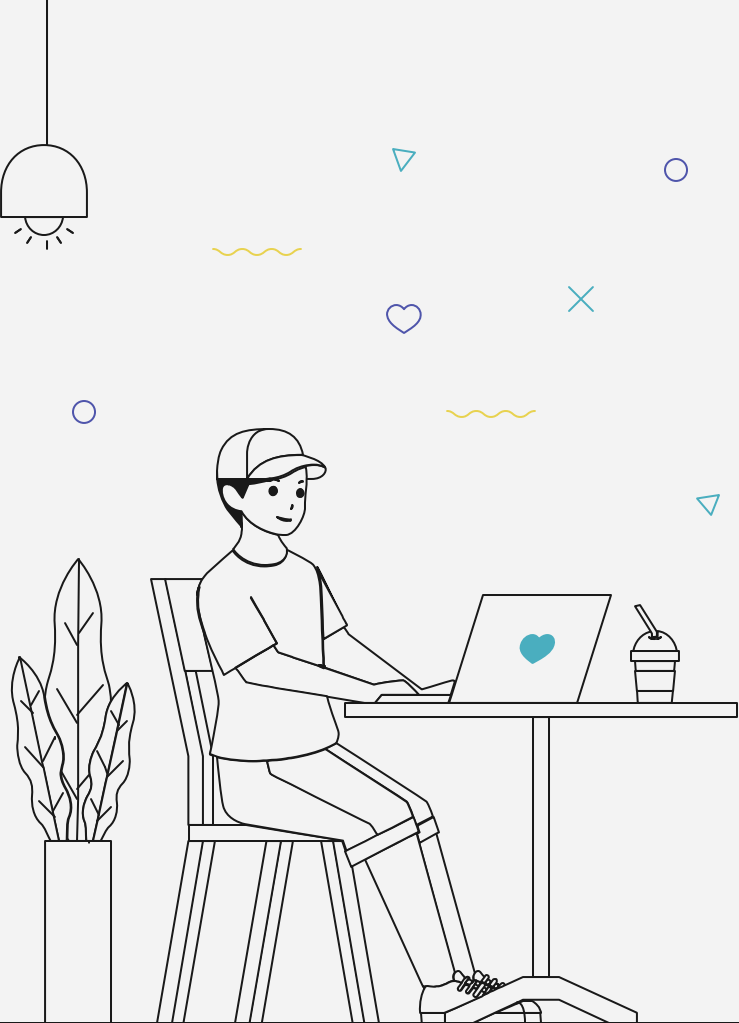
Discussing key lessons learned from the course and provide recommendations for future iterations



01

Introduction

The motivation for integrating LLMs into the CS1 course, the changing landscape of AI, and how tools like GitHub Copilot are transforming programming education



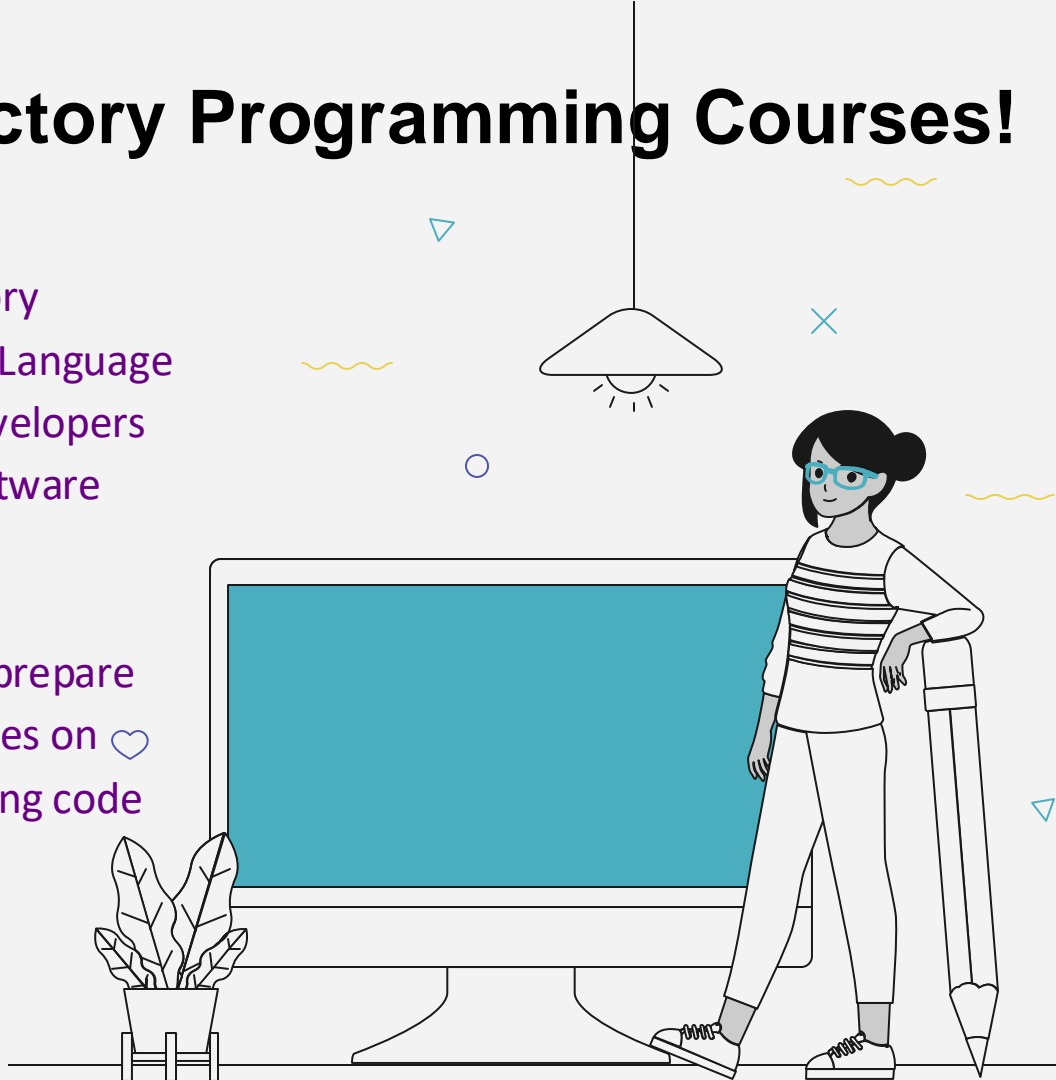
A stylized line drawing of a study desk. On the left, a laptop is open, displaying a blue screen with a white icon of an open book. To the right of the laptop is a desk lamp with a circular base and an adjustable arm. Further right is a simple mug. In the foreground, there are two spiral-bound notebooks with lined pages. The background is decorated with various geometric shapes: a blue triangle at the top, a blue 'X' and a blue circle on the left, and a blue 'X' on the right. Yellow wavy lines are scattered across the scene.


In 2023, research highlighted the growing impact of Large Language Models (LLMs) and Generative AI (GenAI) on computing education. Tools like GPT-4 can now solve programming problems at the **level of top students**, prompting educators to rethink course designs. Some are restricting AI use, while others are embracing the changes.

Redesigning Introductory Programming Courses!

They proposed reshaping introductory programming courses around Large Language Models (LLMs). With 92% of U.S. developers using LLMs, the skills needed for software development are evolving.

Students should learn with LLMs to prepare for a workforce that increasingly relies on these tools, shifting focus from writing code from scratch.

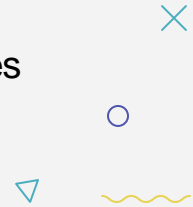











LLMs enable students to tackle larger, open-ended projects, improving engagement and real-world programming skills. Our new CS1 course uses LLMs to help students overcome syntax issues and focus on creative programming

Design Goals:

The CS1-LLM course was designed with five key principles to help students benefit from Large Language Models (LLMs) like GitHub Copilot.

1. **LLM Integration:** Students are encouraged to use LLMs throughout their coursework, including some supervised quizzes and tests. The course balances teaching coding fluency both with and without LLM assistance.
 1. **Industry Preparation:** The course bridges the gap between academic coding practices and real-world industry needs, focusing on code reading, modification, and feature development rather than just writing standalone programs.
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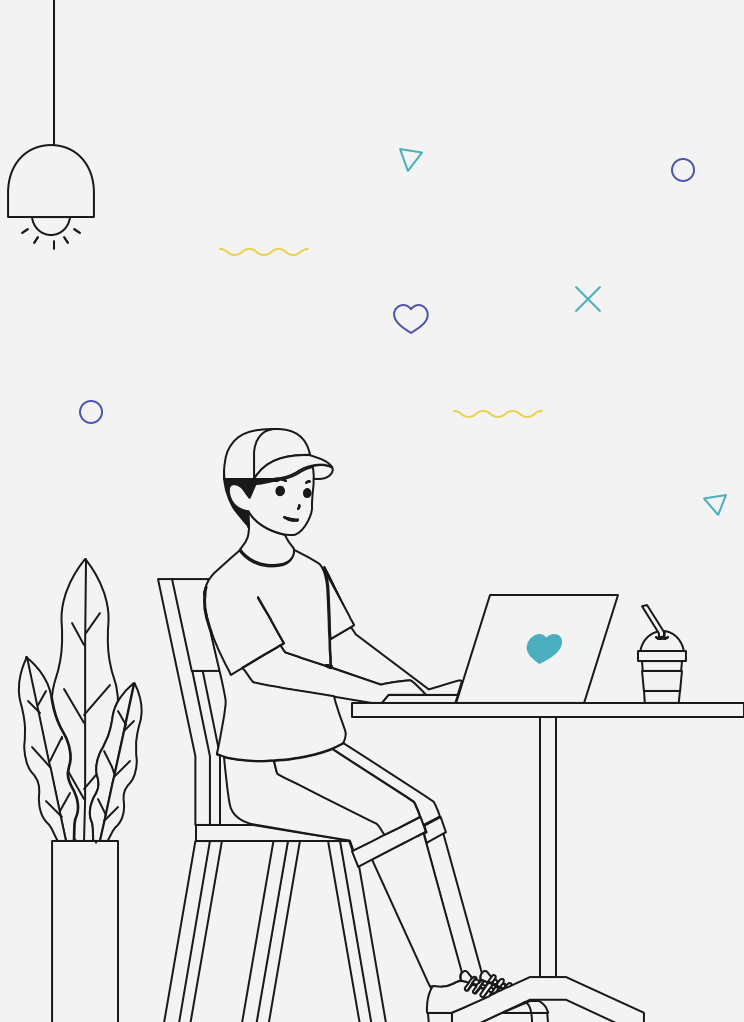
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3. **Supporting Underrepresented Groups:** Best practices, like Peer Instruction and pair programming, are employed to improve outcomes for underrepresented students.
4. **Creativity in Assignments:** The course promotes open-ended projects to foster creativity, moving away from constrained problems typically used in auto-tested assignments.
5. **Serving All Students:** The course serves both future CS professionals and those who won't pursue more CS courses, emphasizing high-level skills and practical applications over low-level syntax
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02

Course CS1-LLM

Course context, learning goals, structure,
scheduling and how LLMs were
integrated to achieve set goals

[Course Materials Link](#)



Course Context

Table 1: Course Demographics

Group	Yes	No	Decline
Computing Major	33.7%	66.3%	–
Gender			
Male	44.3%	52.2%	3.5%
Female	50%	46.5%	3.5%
Nonbinary	2.2%	94.3%	3.5%
Race/Ethnicity			
Hispanic or Latine	27.0%	70.2%	2.9%
Native American	2.5%	77.8%	19.7%
Black or African American	3.2%	77.1%	19.7%
East or Southeast Asian	43.8%	36.5%	19.7%
Indian or other South Asian	9.5%	70.8%	19.7%
Pacific Islander	0.6%	79.7%	19.7%
North African/Middle-Eastern	3.2%	77.1%	19.7%
White or Caucasian	22.5%	57.8%	19.7%
Student Status			
Transfer Student	7.3%	91.1%	1.6%
First-Gen. College Student	43.2%	50.8%	6.0%
Pell Grant Eligible	47.0%	30.8%	22.2%

Classes include

- 3 hours of lecture
- 1 hour of in-person discussions
- 1 hour of closed labs

Diverse student population: including Computer Science, Computer Engineering, Bioinformatics, and Data Science. The demographics also reveal that 47% of students were Pell Grant eligible, indicating many from low-income backgrounds.

This course covers fundamental programming concepts like variables, functions, conditionals, loops, strings, lists and dictionaries in Python.





Course Learning Goals



01

Knowledge

Define nondeterminism, LLM, prompt, and related terms.

02

Comprehension

Illustrate AI workflow
Describe key Python features

03

Application

Use prompt engineering to guide AI code output.

04

Analysis

Analyze, divide, and debug Python programs effectively.

05

Synthesis

Design, test, fix, modify, and write Python programs.

06

Evaluation

Use testing evidence to verify program correctness.



Course Structure

Week 1-4:

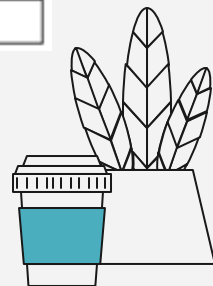
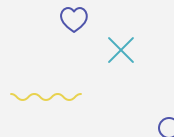
- How to read, trace, and explain code
- How to use GitHub Copilot
- Basics (variables, conditionals, loops, functions, strings, and lists)

Week 5-10:

Transitioned to a more software engineering-focused approach, teaching skills like problem decomposition, testing, and debugging across three different domains: data science, image manipulation, and game development.

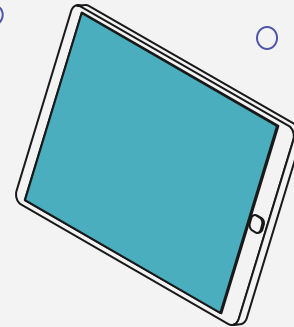
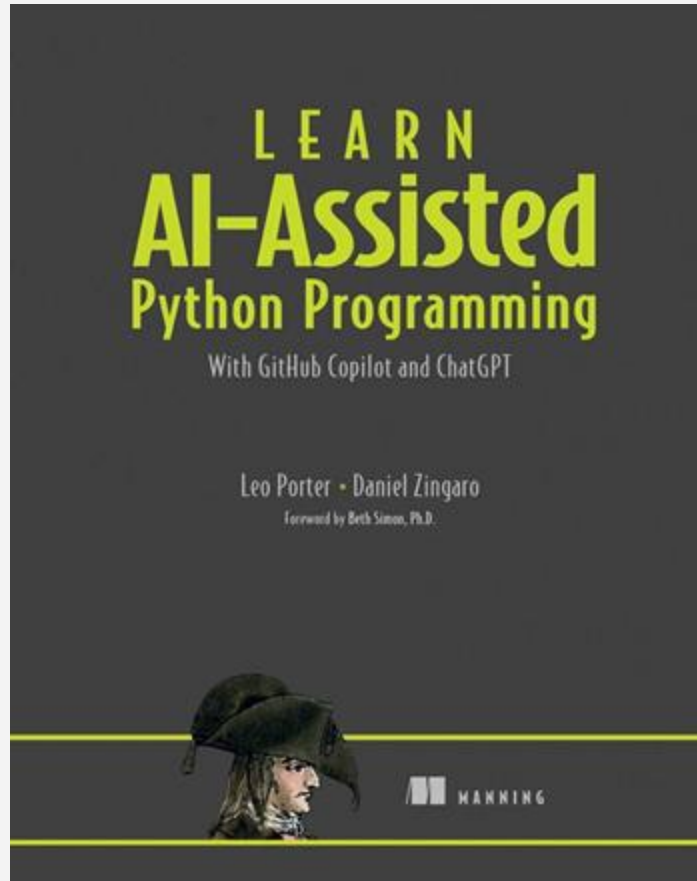
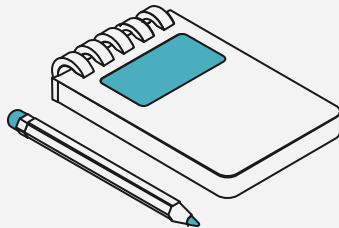
Table 2: Course Schedule

Week	Topic(s)
1	Functions and Working with Copilot
2	Variables, Conditionals, Memory Models
3	Loops, Strings, Testing, VSCode Debugger
4	Loops, Lists, Files, Problem Decomposition
5	Intro to Data Science, Dictionaries
6	Revisit Problem Decomposition and Testing
7	Intro to Images, PIL, Image Filters
8	Copying Images, Intro to Games and Randomness
9	Large Game Example
10	Python Modules and Automating Tedious Tasks



Textbook

Learn AI-Assisted Python
Programming with
GitHub Copilot and
ChatGPT



Nov, 2023

By Leo Porter,
Daniel Zingaro



Lectures



Combination

- ☐ Mini-lectures
- ☐ Live Coding
- ☐ Peer Instruction



Changes

- ☐ Discussions on Copilot interactions
- ☐ Examples of incorrect Copilot responses
- ☐ Conversations with Copilot Chat



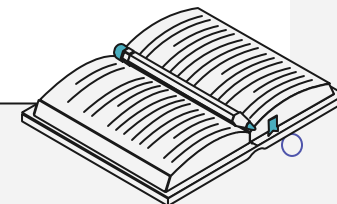


Assessments

Formative assessments accounted for 35% of the grade;
Summative assessments comprised the remaining 65%

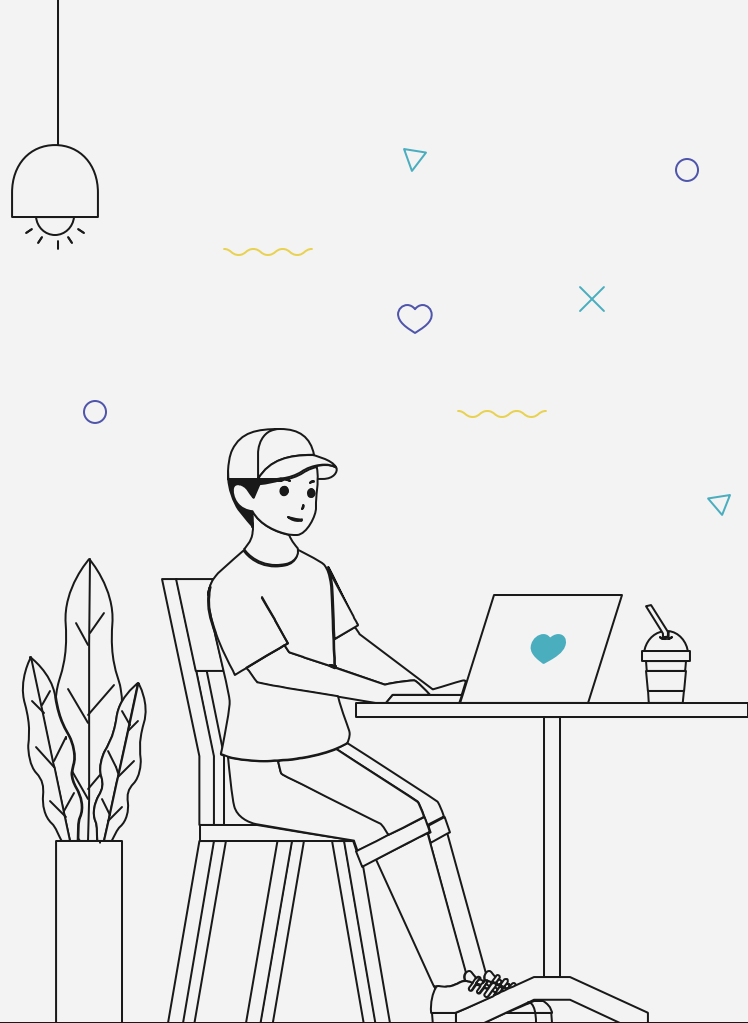
Homework	15% Weekly, Multiple question types, No restrictions on using LLMs
Quizzes	30% Total four, 50-minute, Mostly without access to Copilot
Projects	10% One project per domain of data science, image manipulation, and games
Labs	10% Weekly, Synchronous/at home, some work with Copilot and some not
Final Exam	25-30% 90mins, 70%, Multi-choice component 45mins, 15%, Code writing tasks 45mins, 15%, One large new problem to be completed with Copilot

Readings and other online quizzes/surveys: 5% and Participation: 0-5%



03 Student Perceptions

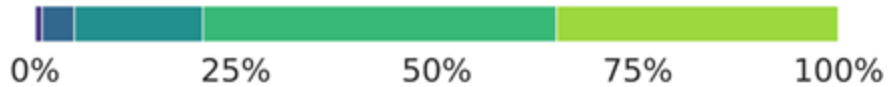
To evaluate the impact of the redesigned CS1 course, instructors surveyed students, focusing on their experiences with LLMs





Student Comfort with Copilot

How comfortable or uncomfortable are you in using GenAI tools to program?



- Not at all comfortable
- Slightly comfortable
- Neutral
- Comfortable
- Strongly comfortable

Figure 2: Student Comfort Programming with GenAI

79%

Comfortable

The majority of students felt comfortable using GenAI tools like Copilot.



Impact of Copilot on Student Learning

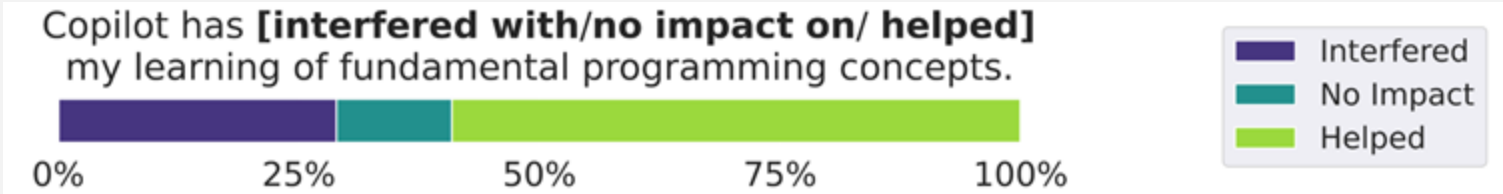


Figure 3: Student Perceptions of how Copilot impacted their learning.



59%

Helped

A slight majority of students reported that it had helped them understand programming concepts.

"It was really nice having an assistant that could always help me the moment I needed it and made programming a lot less daunting."

"If I were asked to code without Copilot, I wouldn't feel very confident in myself despite doing well in the course."

Student Confidence

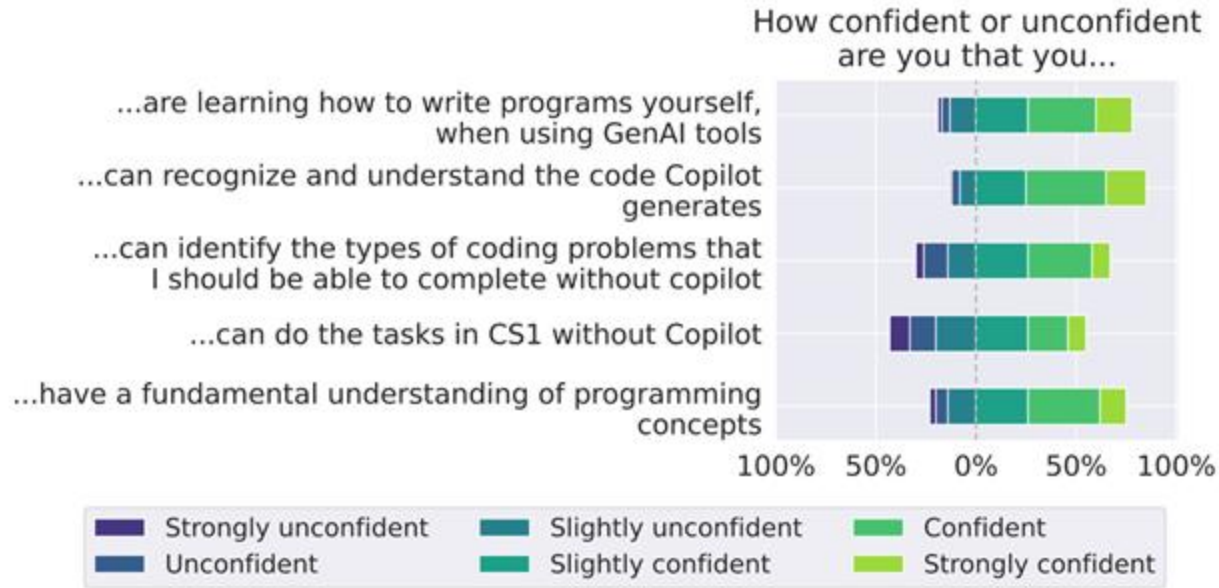


Figure 4: Student confidence in their ability and understanding at the end of the course.



31.1%

Low
Confidence



Some students expressed low confidence in identifying coding problems they should be able to solve independently.

Some students found the inconsistency in using Copilot frustrating, as it was allowed in certain parts of the course but not in others, such as quizzes

How Students Interacted with Copilot



Please select the approximate percentage of time that you do the following. When I get a response from Copilot, how often do you:

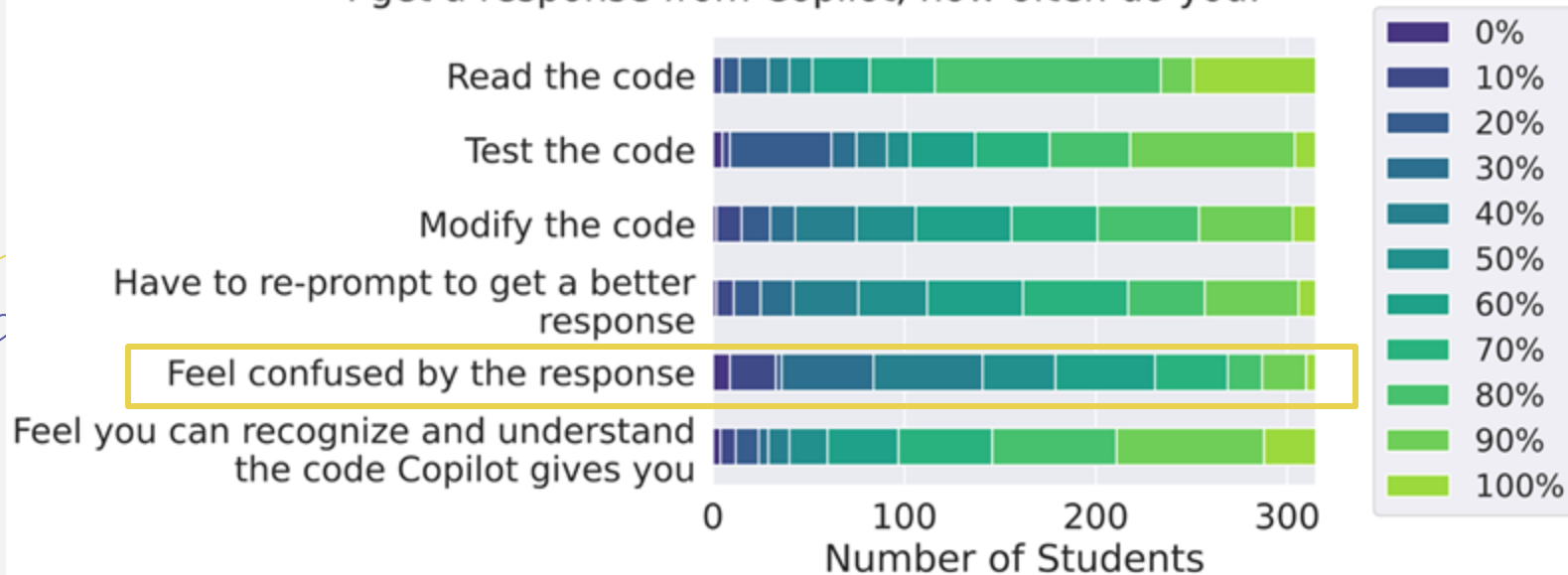


Figure 5: Student perceptions on how they interact with Copilot.



Student Comfort with Copilot

How helpful were the programming projects for your learning?

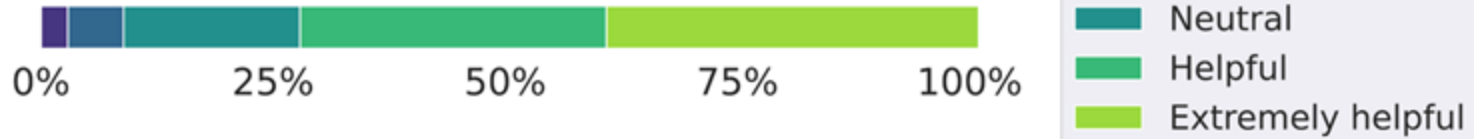


Figure 6: Students' View on Learning with Projects



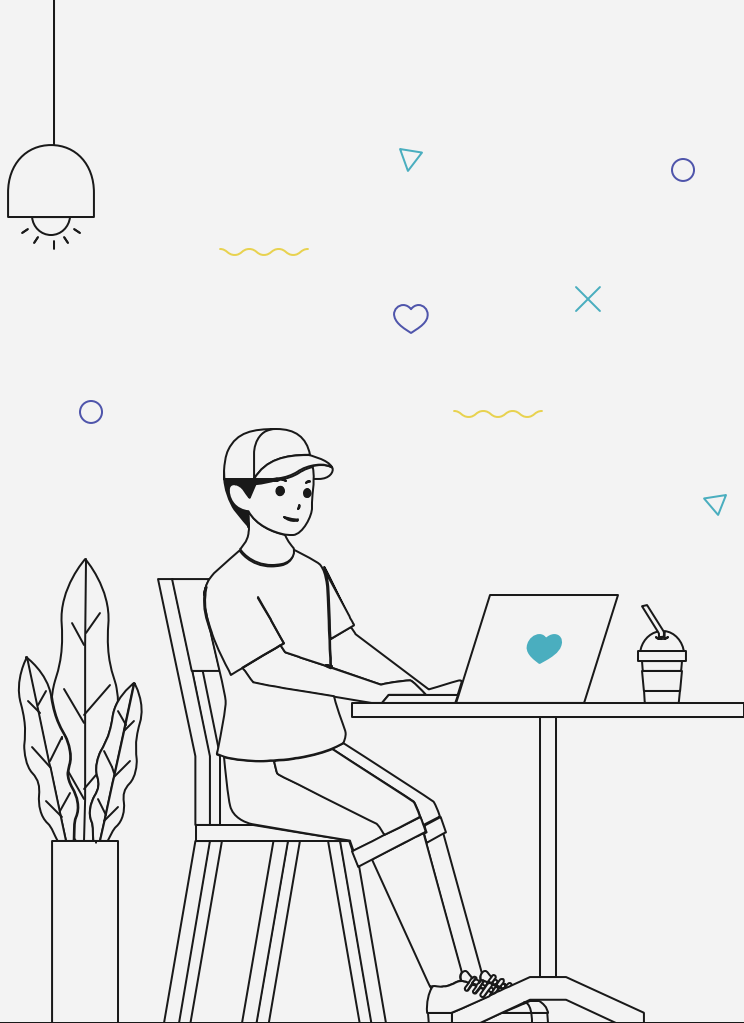
70%

Helpful

Students generally found these projects valuable for their learning

"Though they were a little frustrating at times, I was really impressed by what I was able to do for each project... I definitely felt more comfortable with coding in general after each project."

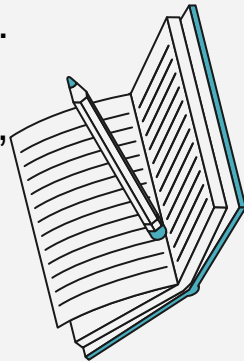
04 Discussion

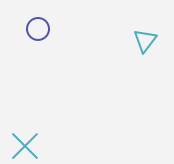




Discussion

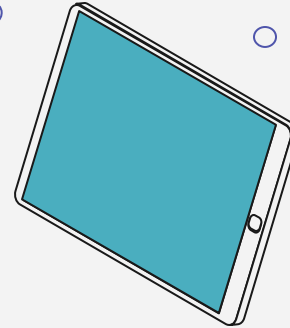
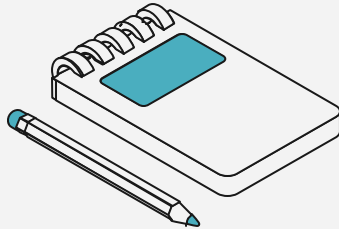
The CS1-LLM course saw most students (79%) feeling capable of programming with LLMs, while a slight majority (59%) found LLMs helpful for learning. However, some students expressed concerns about over-reliance on the tool and gaps in fundamental skills. Despite these concerns, students created more advanced projects than typically expected in a CS1 course.

- **Student Performance:** Students performed similarly to past CS1 courses on code tracing and reading, but slightly lower on code writing tasks. However, their project scope and ability to solve large problems were impressive.
- **Essential Components:** Key elements include projects utilizing LLMs, teaching problem decomposition, testing, and incorporating LLM responses into class examples.

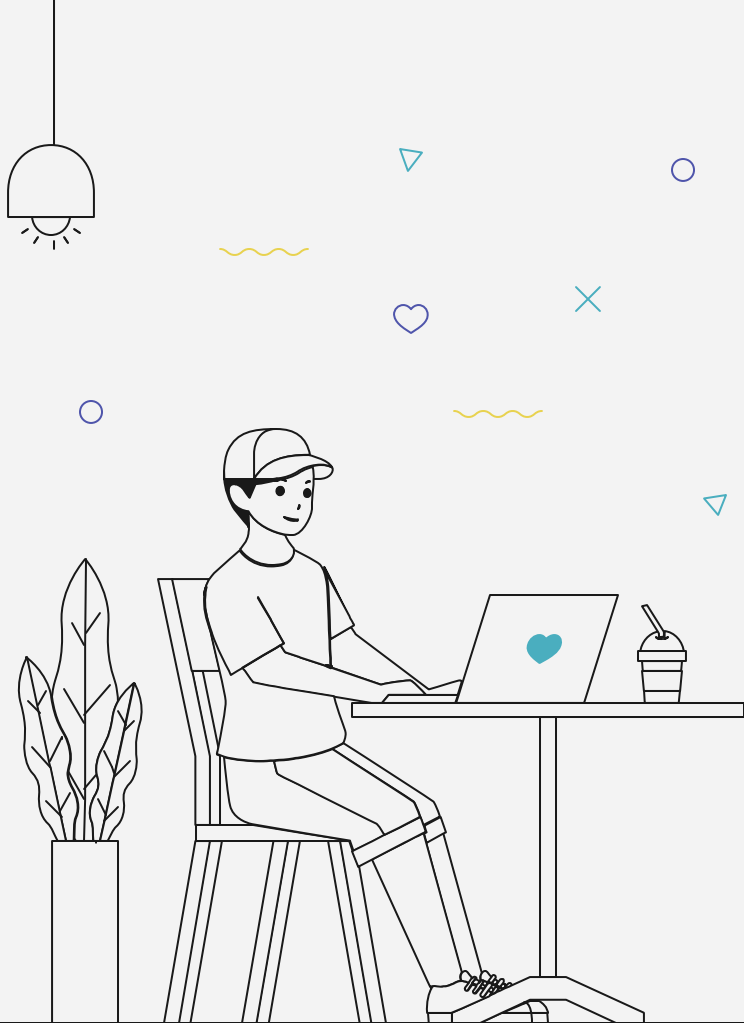


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- **Course Changes:** Most changes involved LLM interaction, problem decomposition, and larger projects, though basic coding skills were still covered.
 - **When to Introduce LLMs:** Introducing LLMs in the first week was challenging, so delaying their introduction slightly in future offerings is recommended.
 - **Clear Expectations:** Students were unclear about what they should accomplish with and without LLMs. Future courses will provide clearer guidance.
- 

- **Learning Goals Alignment:** Assessments must be aligned with updated course goals, avoiding past questions that don't fit the new objectives.
- **Non-Determinism in Live Coding:** LLM responses varied across class sections, prompting the team to pre-paste Copilot responses into slides.
- **Login Issues in Exams:** Signing into GitHub for exams was time-consuming due to credential and verification issues. Future exams should allow extra time or personal device use for easier access.



05 Conclusion



Thanks!

