

AI Mentors for Student Projects: Spotting Early Issues in Computer Science Proposals



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Summary

Problem: Evaluating project proposals and student readiness for project-based learning (PBL) is time-consuming and difficult to scale. Quick and early evaluation could help educators determine which students need support to thrive in PBL.

Findings from user study (n=36):

- GPT-4o's ratings of project proposals show high agreement with educator ratings
- Novices struggle with writing high-quality proposals and learning objectives
- Users perceived the system as helpful for writing project proposals and identifying tools and technologies to learn more about

Task & Evaluation Criteria

Project Proposal Task: adapted from existing high school CS Career Pathways PBL program

- **Topic** (Web Dev, Video Game Dev, Useful Scripts)
- **Background / Problem, Objectives**
- **Find + Analyze Inspiring Examples**
- **Evaluation Plan**
- Name 3 skills to develop by working on project
- Pair each skill with mentor (I would like advice from...), job tasks, and technologies (source: O*Net Online)

Evaluation: Each project proposal was independently evaluated by two human experts (college CS teaching assistants) and GPT-4o according to a **29-item rubric**

- **1-item x 3 Skill Quality Classification:** Is skill good? (NOT irrelevant to project, non-technical, or vague)
- **3-item x 3 Skill Pairing Classifications:** Is pairing (mentor-skill, job task-skill, technologies-skill) a *good fit*?
- **10-item Quality Checklist:** adapted from college CS mobile dev. project proposal rubric

Item 1. This proposal describes a specific focus and motivation (beyond describing the project topic)	Item 6. The design hypothesis describes a specific feature that will be built in the project
Item 2. This proposal describes a good use of computer science skills	Item 7. The predicted effects of the design hypothesis can be tested quickly
Item 3. This proposal describes specific tangible features that will be built in the project	Item 8. The project has an objective measure of success or learning
Item 4. Working on the project is relevant to learning about project topic	Item 9. The design hypothesis has an objective measure of success or learning
Item 5. The proposal analyzes similar products, papers, or applications	Item 10. The evaluation plan can be carried out within a 4 week sprint

- **1-item Recommend for Resume:** General measure for quality of project (goes beyond tutorial, understandable)

Educators vs. GPT-4o Evaluate Project Proposals

- 😊 **Skill Quality:** High Agreement
- 😢 **Skill Pairings:** Low Agreement (hard to grade vague skills)
- 😐 **Quality Checklist:** Slightly Low Agreement (GPT-4o is stricter)

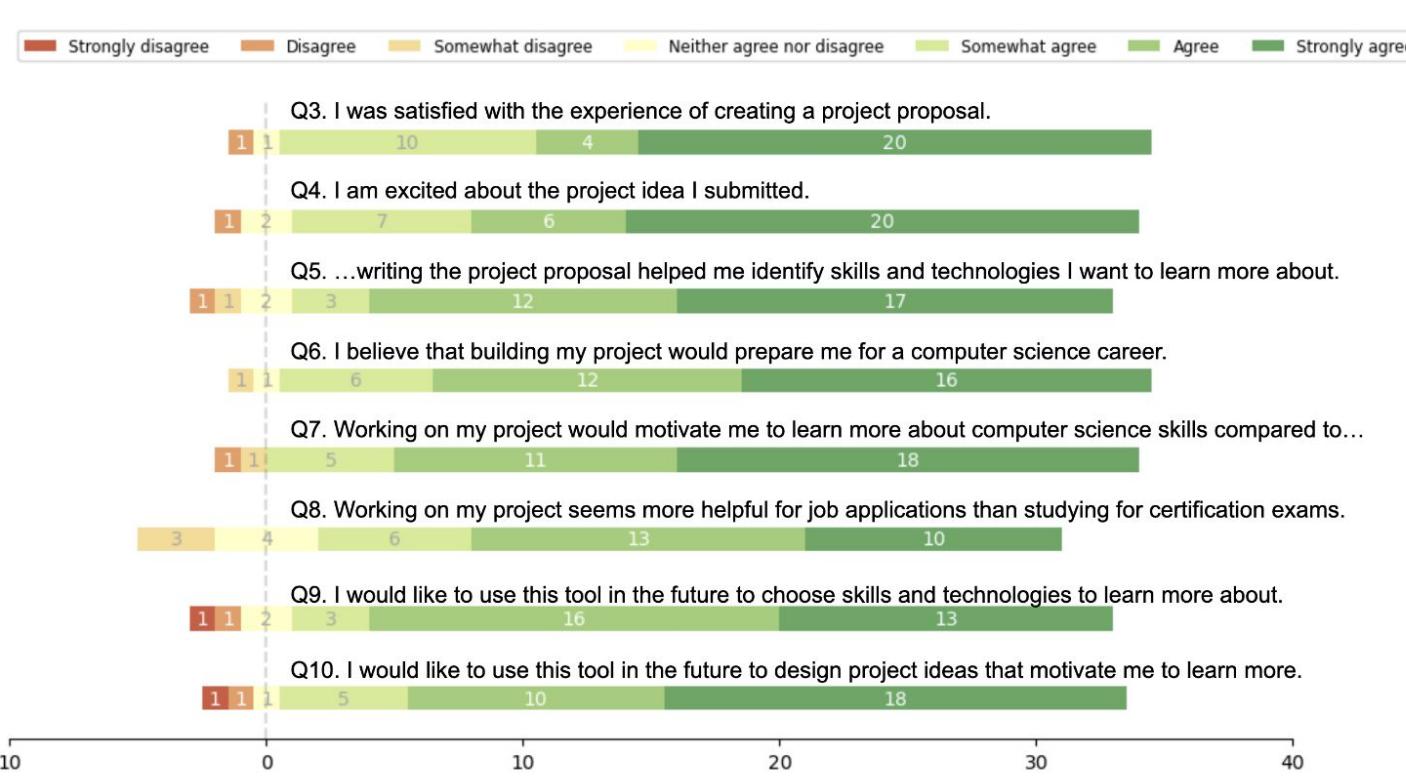
	TA1 / TA2	TA1 / GPT-4o	TA2 / GPT-4o	Avg. κ
Skill Quality Classification	86.1%, $\kappa = 0.72$	84.3%, $\kappa = 0.68$	81.5%, $\kappa = 0.63$	0.68
Skill Pairing Classification	68.6%, $\kappa = 0.26$	74.2%, $\kappa = 0.46$	67.2%, $\kappa = 0.20$	0.29
Quality Checklist	87.2%, $\kappa = 0.49$	71.4%, $\kappa = 0.28$	74.7%, $\kappa = 0.28$	0.38
Recommend for Resume	80.6%, $\kappa = 0.50$	75.0%, $\kappa = 0.43$	77.8%, $\kappa = 0.45$	0.46

	TA 1	TA 2	GPT-4o	Self
TA 1	1.00	0.74	0.70	0.16
TA 2	0.74	1.00	0.53	0.38
GPT-4o	0.70	0.53	1.00	0.23
Self	0.16	0.38	0.23	1.00

- 😊 Spearman correlations show that though GPT-4o is a *stricter* grader than the TAs, **it's grades preserves the rank order of the Human Experts' Quality Checklist scores** (better than students' self-evaluations)

Task	Experience	TA1	TA2	GPT-4o	Self-Rating
Skill Classification	Novice	35.3%	39.2%	31.4%	-
	Experienced	70.2%	68.4%	68.4%	-
Skill Pairing Classification	Novice	60.8%	86.9%	57.5%	-
	Experienced	53.7%	92.4%	64.9%	-
Quality Checklist	Novice	83.5%	82.4%	58.2%	85.9%
	Experienced	84.7%	90.0%	71.1%	95.8%
Recommend for Resume	Novice	58.8%	70.6%	58.8%	-
	Experienced	78.9%	84.2%	73.7%	-

- 😊 **Sanity Check** — Students with prior CS knowledge tend to have higher quality submissions, e.g., on average, novices can propose 1 good skill, experienced students can propose 2 good skills.



- 😊 Majority of students enjoyed completing the project proposal activity and believed it benefited their learning.

Future Directions

GPT-4o achieves high agreement with human expert evaluations and hence enables scalable support for PBL

- **Classroom studies:** Does LLM agreement transfer to teachers' rubrics? Student-led rubrics?
- **Long-term studies:** Does early detection of issues spotted by GPT-4o lead to a smoother, more educational PBL experience?