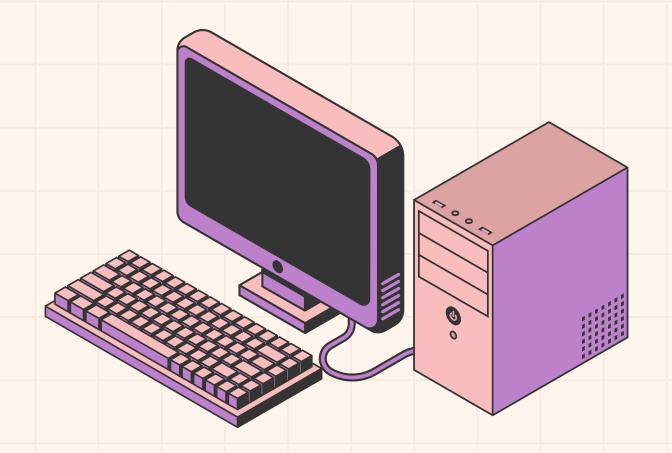
UNPACKING LEGAL REASONING IN LLMS

Chain-of-Thought as a Key to Human-Machine Alignment in Legal Essay Tasks

ESSLI WORKSHOP - NALOMA 2025

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MOTIVATION

Why Compare Human and LLM Legal Reasoning?

- Legal reasoning requires **multi-step logic**, statutory interpretation, and precise subsumption.
- Benchmarks like LexGLUE focus on MCQs, not complex reasoning chains
- LLMs often "guess correctly" without following legal logic.
- Goal: Examine and align LLM reasoning flow with human legal thought.

RESEARCH GOALS

How Can Chain-of-Thought Prompting Help?

- Make LLMs' reasoning traceable and interpretable.
- Improve alignment with human reasoning patterns.
- Diagnose reasoning failures, not just accuracy.
- Contribute a structured evaluation framework for legal NLU tasks.

EVALUATION DIMENSIONS

Three Axes of Human-LLM Alignment

- Issue Coverage Did the model identify the core disputes?
- Statutory Application Were the correct laws cited and applied?
- Reasoning Clarity Was the logic coherent and structured?

METHOD OVERVIEW

Two-Stage Evaluation Framework









- Stage 1: Decomposed reasoning on 68 sub-questions.
- Stage 2: Full-length answers scored by legal experts.
- Models tested: ChatGPT-4o, Gemini, Grok3, Copilot.
- Chain-of-Thought (CoT) prompting vs. direct answering.

CHAT-BASED EXAMPLE

What CoT Prompting Looks Like in Action

- <u>u</u> LLM: The divorce was not valid because the witness signatures were forged...
- ☑
 ♣ User: If the divorce was invalid, was the marriage to E lawful?

- <u>math</u> LLM: E does not, since the marriage was void and E was not in good faith...

STAGE 1 - SUB-QUESTION DESIGN

Fine-Grained Probing of Legal Reasoning

- 6 essay questions from Taiwanese status law exams.
- Decomposed into yes/no, numeric, and short-answer questions.
- Evaluates factual classification, statutory mapping, inheritance logic.

STAGE 1 RESULTS

Accuracy Gains from CoT Prompting

Model	Raw Accuracy	CoT Accuracy	t-value	p-value
ChatGPT	0.842	0.866	-0.92	0.398
Gemini	0.833	0.9445	-3.71	0.013
Copilot	0.822	0.864	-2.14	0.089
Grok3	0.843	0.895	-2.98	0.031

CoT prompting led to statistically significant gains in Gemini and Grok3.

STAGE 2 - FULL ANSWER EVALUATION

Can CoT Prompting Enhance Holistic Legal Writing?

- 6 full-length answers scored on 0–10 scale.
- Evaluated blindly by professor and law student.
- Scoring criteria: issue coverage, law accuracy, clarity.

STAGE 2 RESULTS

Human Ratings: CoT vs. Baseline

Model	Raw Average Score	CoT Average Score	Average Improveme nt
ChatGPT	6.5	9.17	2.67
Gemini	6.12	8.04	1.92
Copilot	5.83	7.42	1.58
Grok3	6.25	8.08	1.83

ChatGPT had the most consistent performance boost.

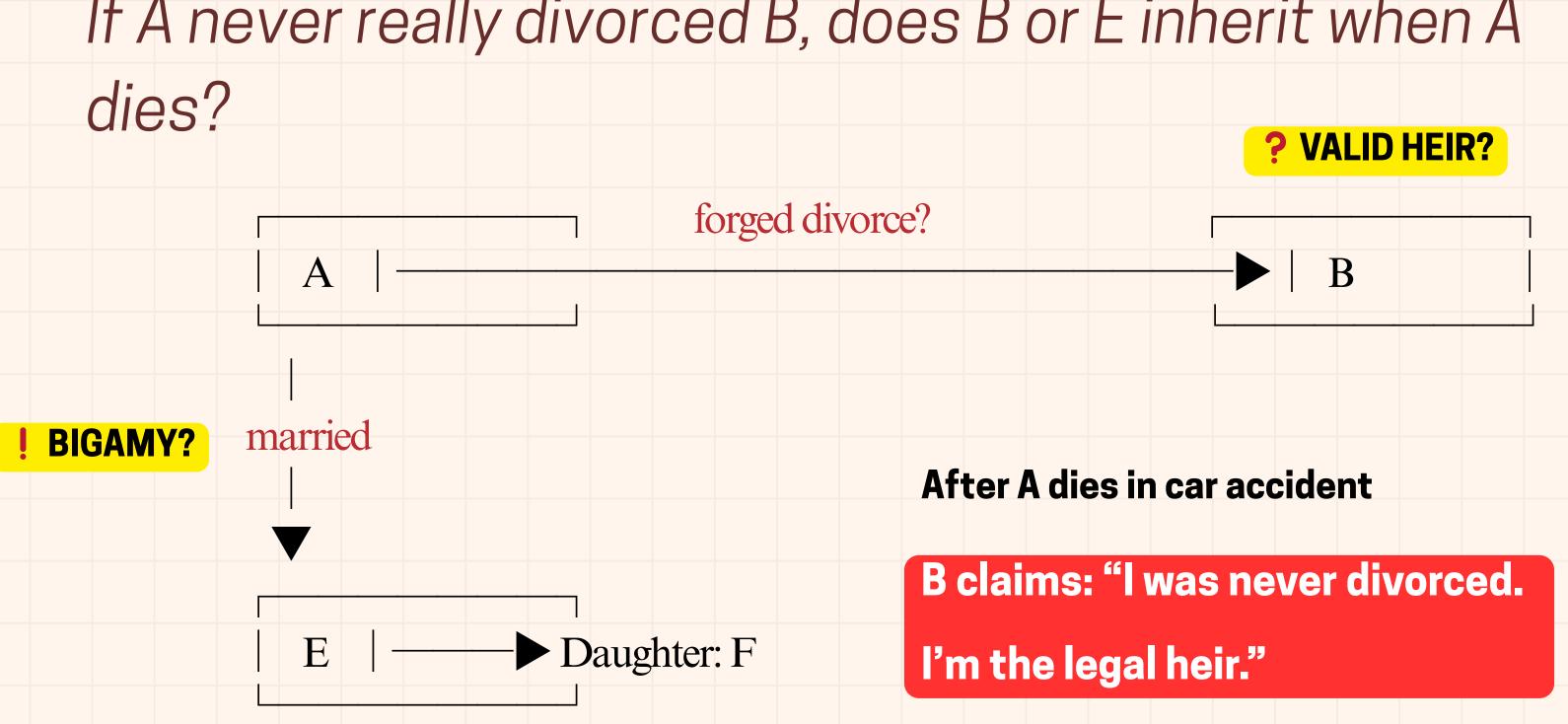
WHICH QUESTIONS IMPROVED MOST?

Task-Level Gains from CoT Prompting

Question	Raw Avg. Score	CoT Avg. Score	Score Gain	p-value
1	4.5	7.75	3.25	0.068
2	6.12	7.12	1	0.43
3	6.25	6.38	0.12	0.919
4	3.88	5.5	1.62	0.08
5	3.88	7	3.12	0.002
6	4.63	7.5	2.88	0.011

CASE STUDY - GEMINI ON Q5

If A never really divorced B, does B or E inherit when A



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CASE STUDY - GEMINI ON Q5

Key Questions: Divorce Validity Marriage Law Inheritance & Registration

- 1. Was the divorce between A and B legally valid, or was it fake?
- 2. If it was fake, is A's second marriage to E still valid?
- 3. Can E legally inherit as a spouse?
- 4. Does F, their daughter, still have inheritance rights?
- 5. Can B (the first wife) cancel the registration done by E and F?

How Human Lawyers Would Reason through This Case

Was the divorce between A and B valid?

X No (forged witness)

Is A's second marriage to E valid?

X No (bigamy, no good faith)

Does E inherit?

X No (marriage void)

Does Finherit?

V Yes (non-marital child, legally recognized)



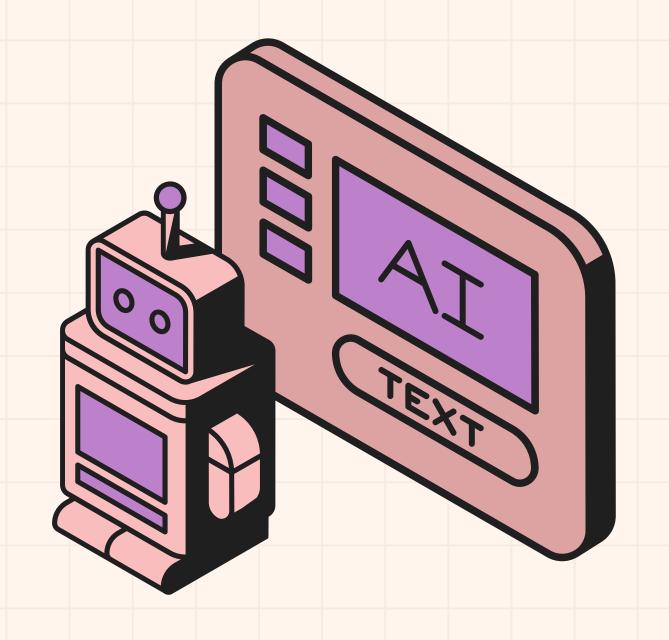
- Registration Cancellation:
- 1
- **☑** E's registration canceled
- **V** F's registration retained

A REASONING COMPARISON

"A human lawyer would never cite §92 here— it applies to intent defects, not procedural flaws in divorce registration."

CONCLUSION WHAT DID WE LEARN?

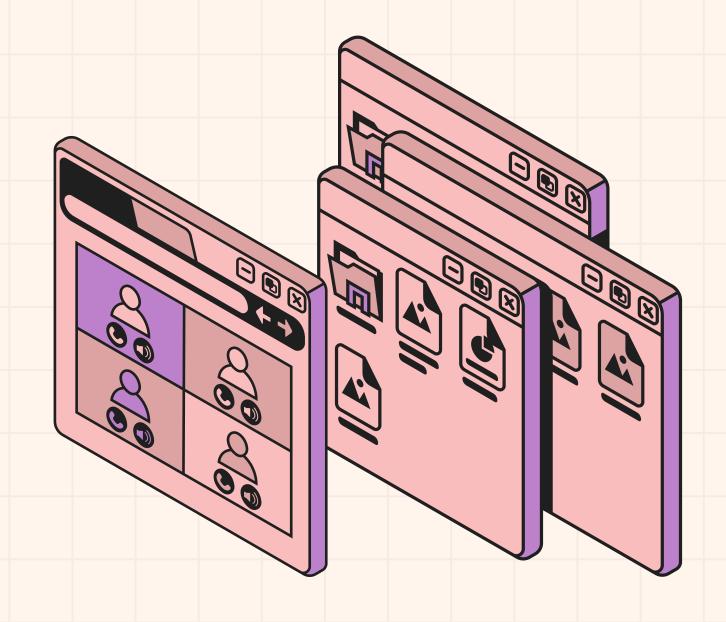
- LLMs can generate fluent but misleading answers.
- CoT prompting improves alignment with legal logic.
- Still gaps in statutory precision and edgecase reasoning.
- Our framework opens the black box for legal NLU.

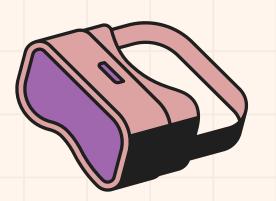


FUTURE WORK

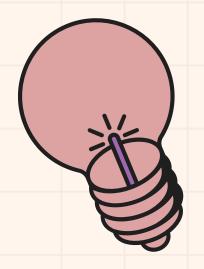
TOWARD MORE RELIABLE LEGAL AI

- Build typology of reasoning failures.
- Formalize sub-question annotation and flowchart templates.
- Scale to multilingual, cross-jurisdictional settings.
- Integrate CoT into fine-tuning pipelines.





THANK YOU!



Questions and feedback welcome.

Contact Us:

