



CS398: Computational Education

Insight Tracing

Story of Riley



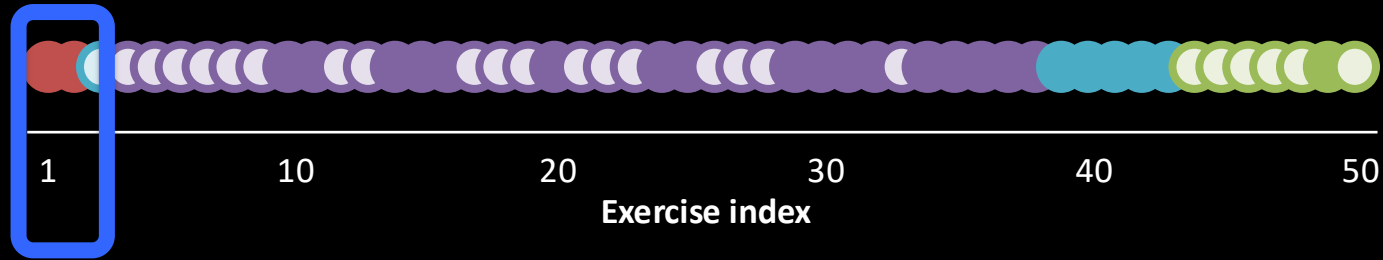
Exercise Type:

- Solving for x-intercept
- Solving for y-intercept
- Graphing linear equations
- Square roots
- Slope of a line

Answer:

- Correct
- Incorrect

Story of Riley



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Story of Riley



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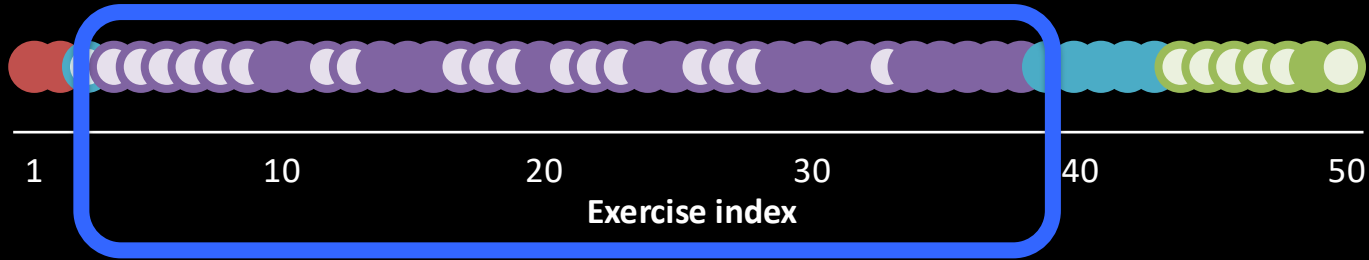
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Story of Riley



What does Riley know?

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Story of Riley



What should Riley do next?

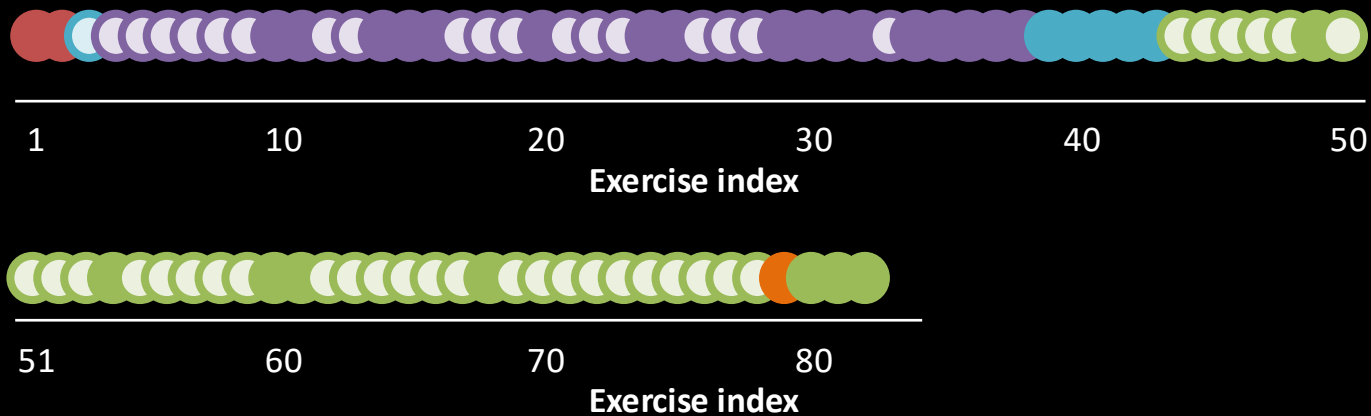
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Story of Riley



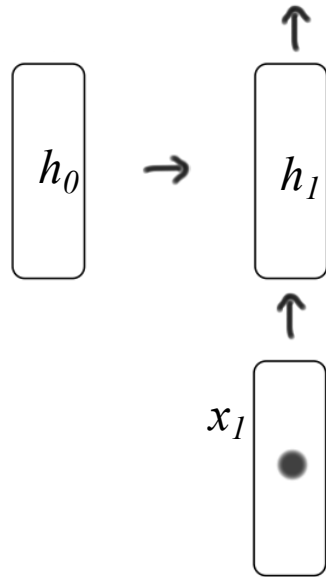
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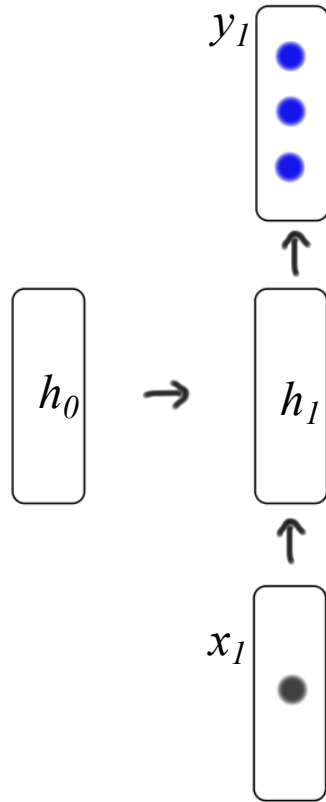
Answer:

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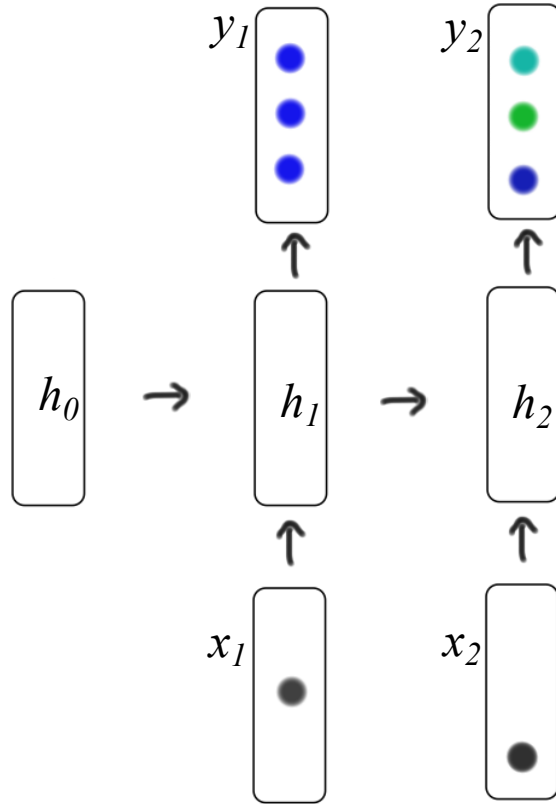
Deep Knowledge Tracing



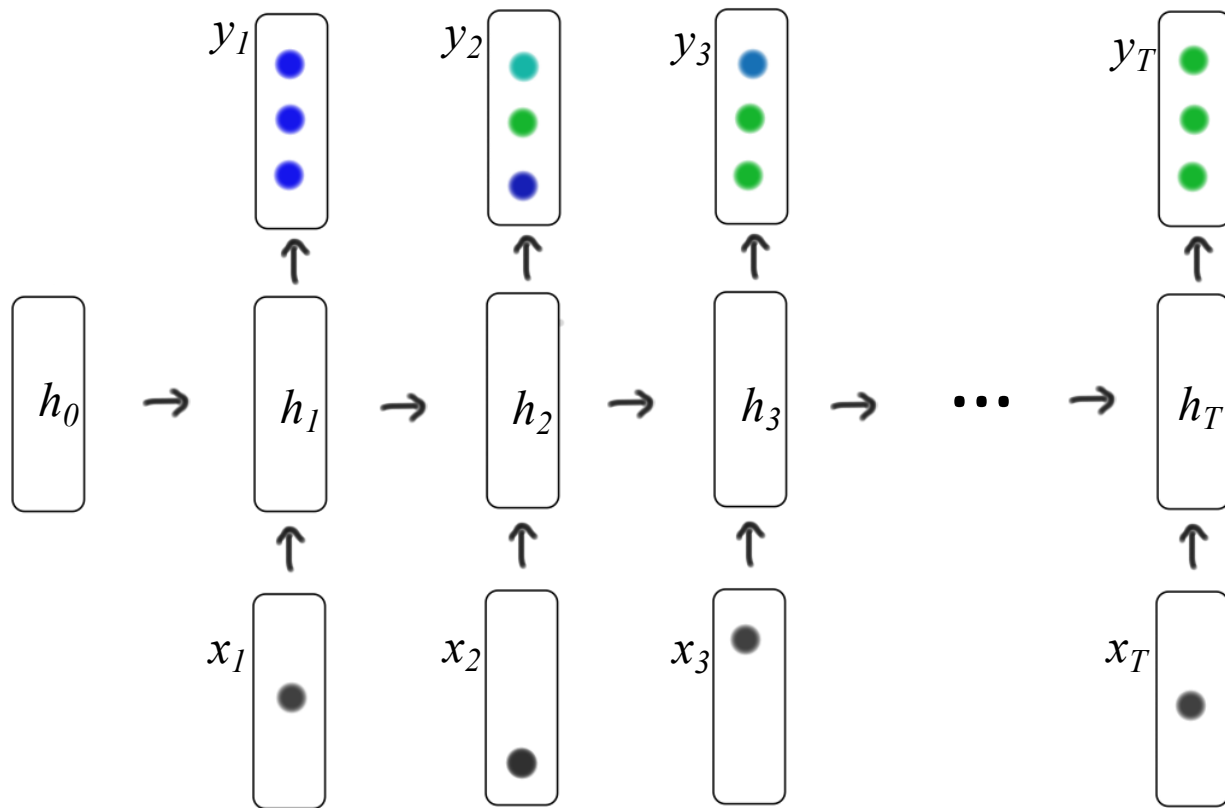
Deep Knowledge Tracing



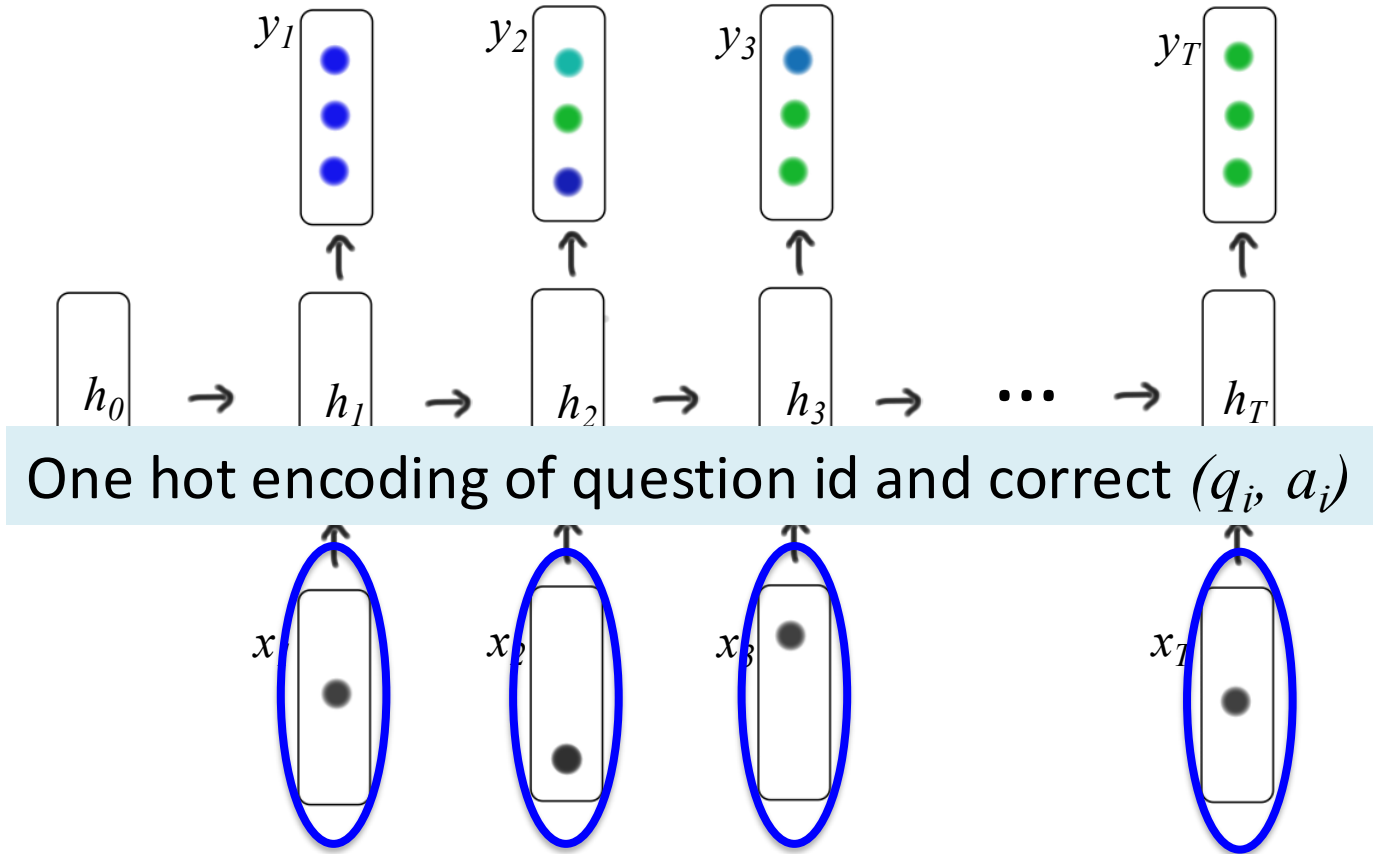
Deep Knowledge Tracing



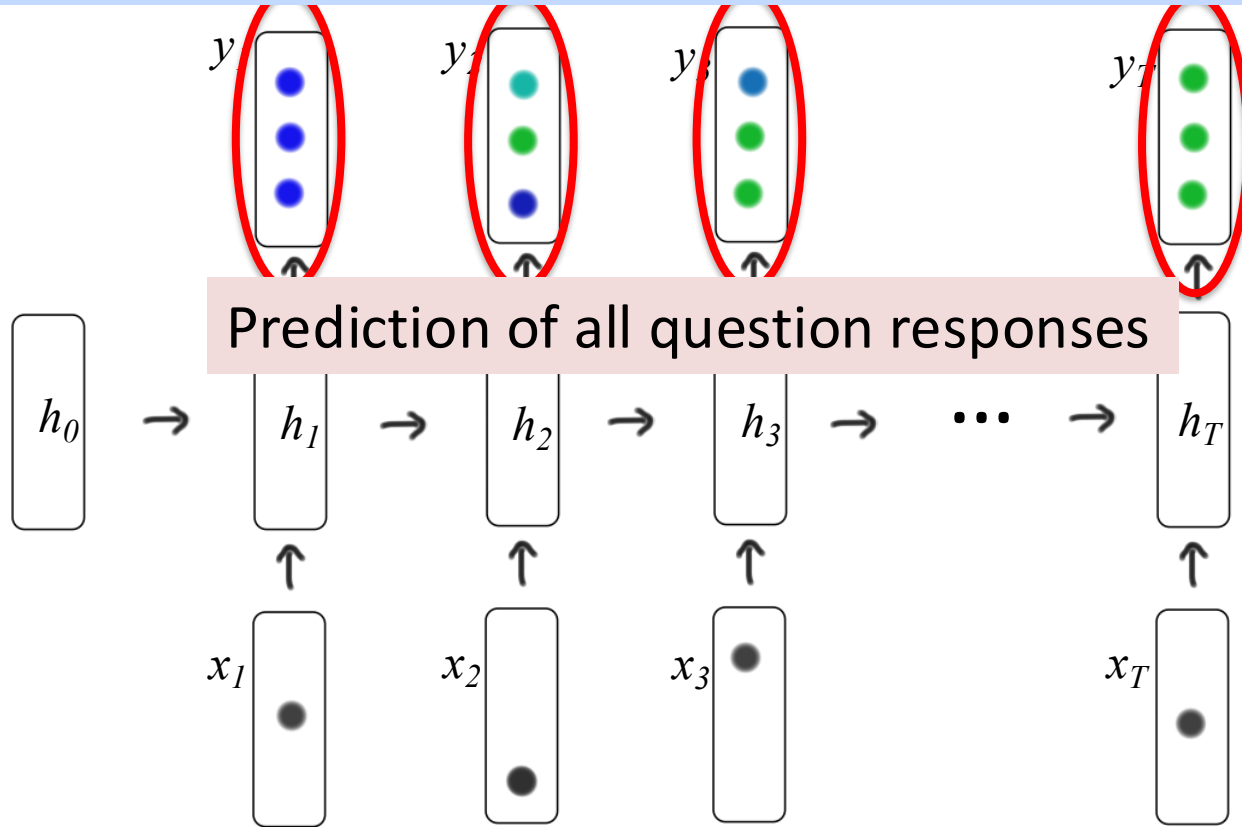
Deep Knowledge Tracing



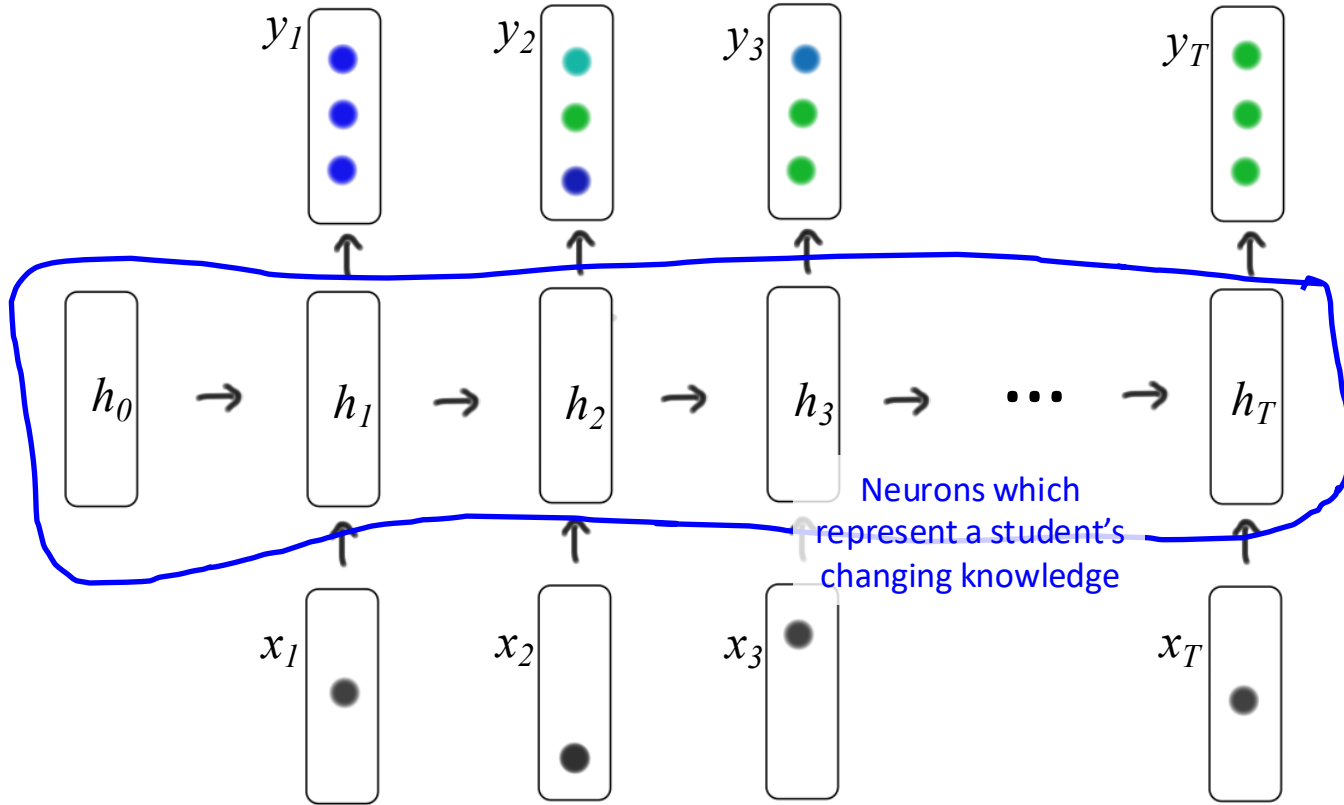
Deep Knowledge Tracing



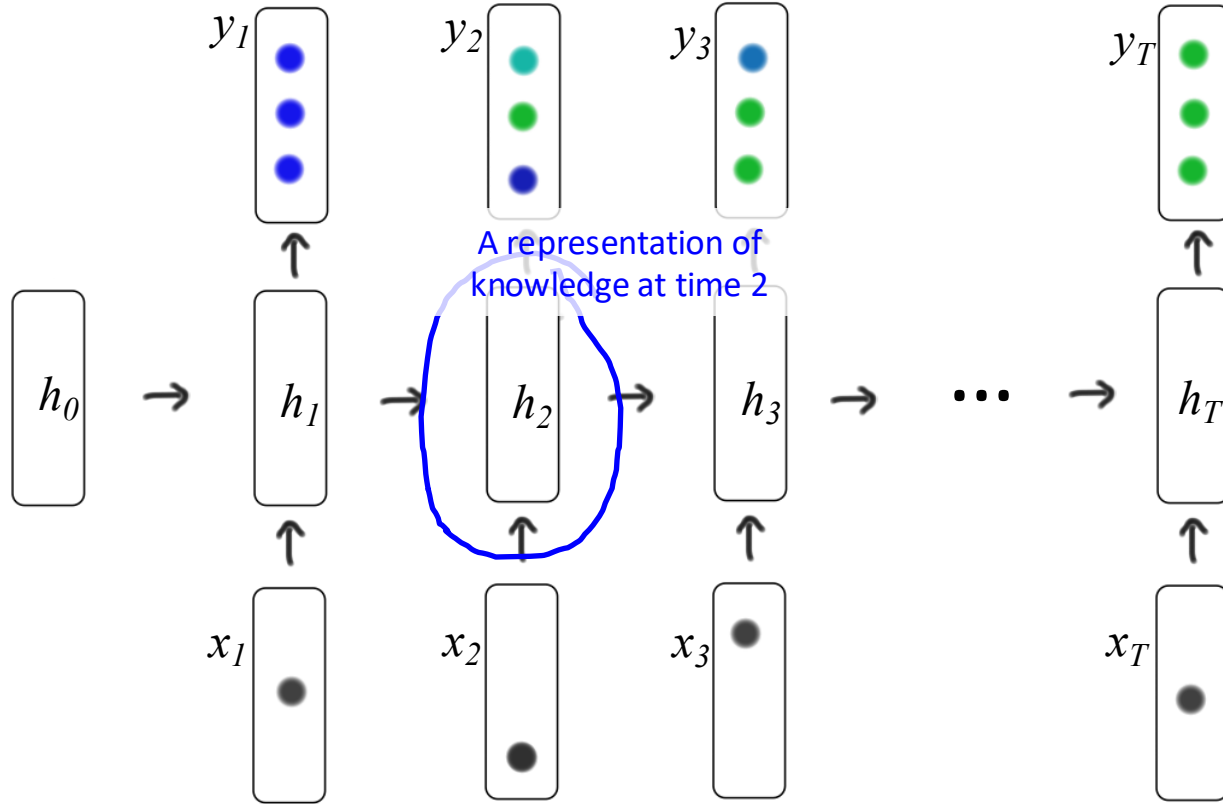
Deep Knowledge Tracing



Deep Knowledge Tracing



Deep Knowledge Tracing



Deep Knowledge Tracing

KHAN
Student



1

10

Exercise index

Exercise Type:

Answer:



Solving for x-intercept



Correct



Solving for y-intercept



Incorrect



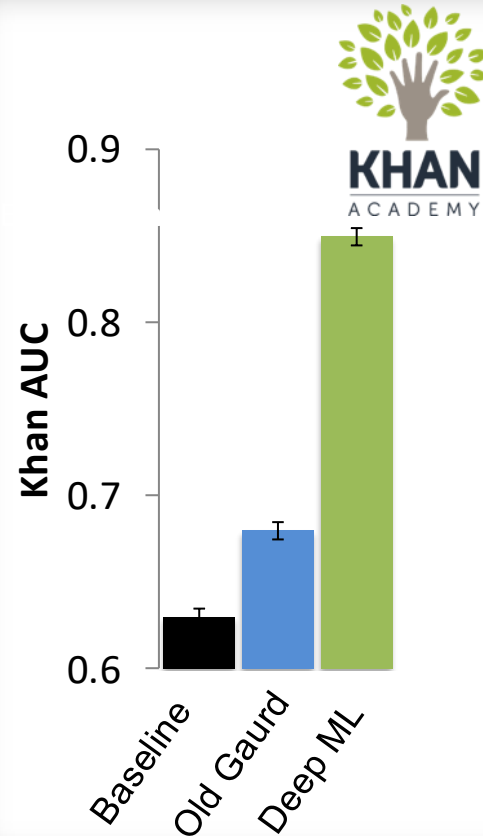
Graphing linear equations



Square roots

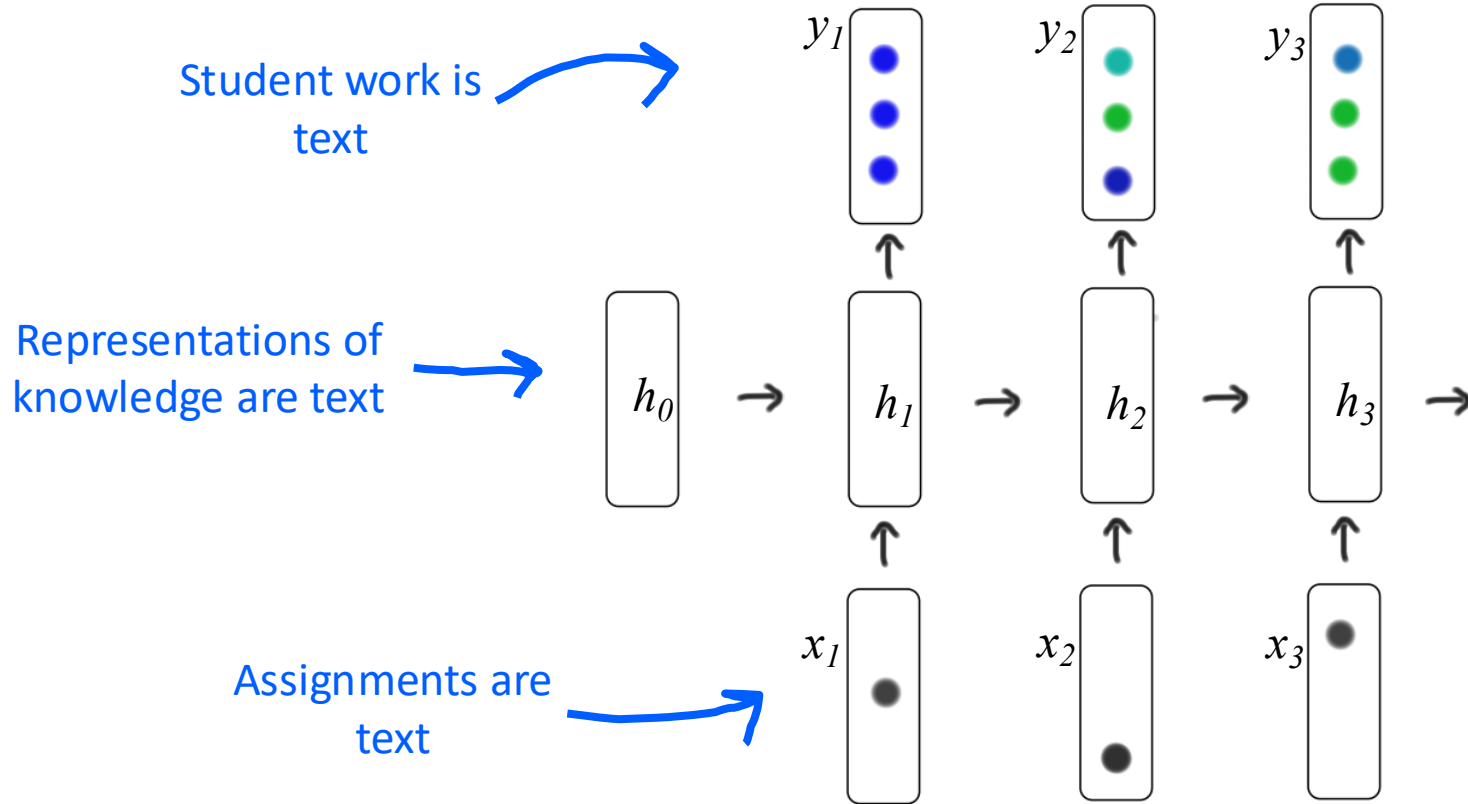


Slope of a line



Powers DuoLingo Bird Brain (as of 2023)

Insight Tracing



Insight Tracing

Text based
representation of
what the student
misconceptions
are



Predict what the
student will write
next



Code 11



Code 1



Code 2

...



Code 10



Example of Insight Tracing

Student **before** TeachNow session

```
3 def main():
7     num2 = random.randint(10, 99)
8     print(f"What is {num1} + {num2}?")
9     answer = int(input("Your answer: "))
10    if answer = num1 + num2:
11        print("Correct!")
12    else:
13        print("Incorrect.")
14    if __name__ == '__main__':
15        main()
```

Auto Generated Insight:

"The learner misunderstands equality vs assignment with variables"

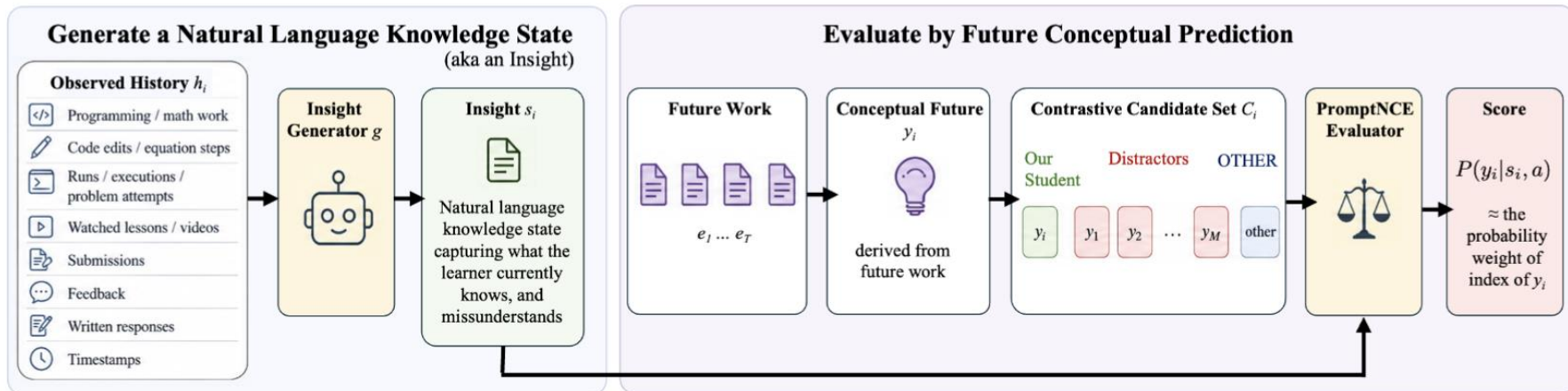
Student **after** TeachNow session

```
3 def main():
    {num2}?")
9     answer = int(input("Your
    answer: "))
10    if answer == num1 + num2:
11        print("Correct!")
12    else:
13        print("Incorrect.")
14    if __name__ == '__main__':
15        main()
```

Auto Generated Insight:

"The learner correctly fixed an = to be an == but may have a lingering misconception with assignment"

Overview



PromptNCE: Conditional Probabilities and PMI Using Only LLMs and Contrastive Estimation Prompts

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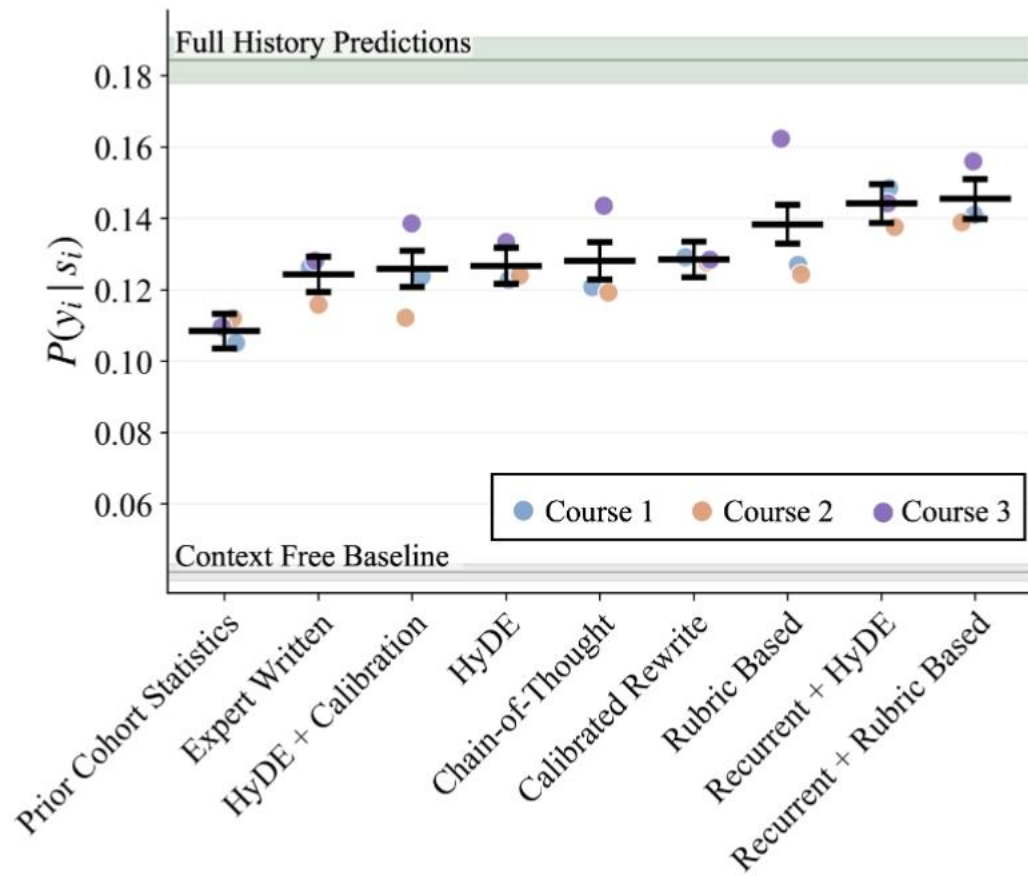
Abstract

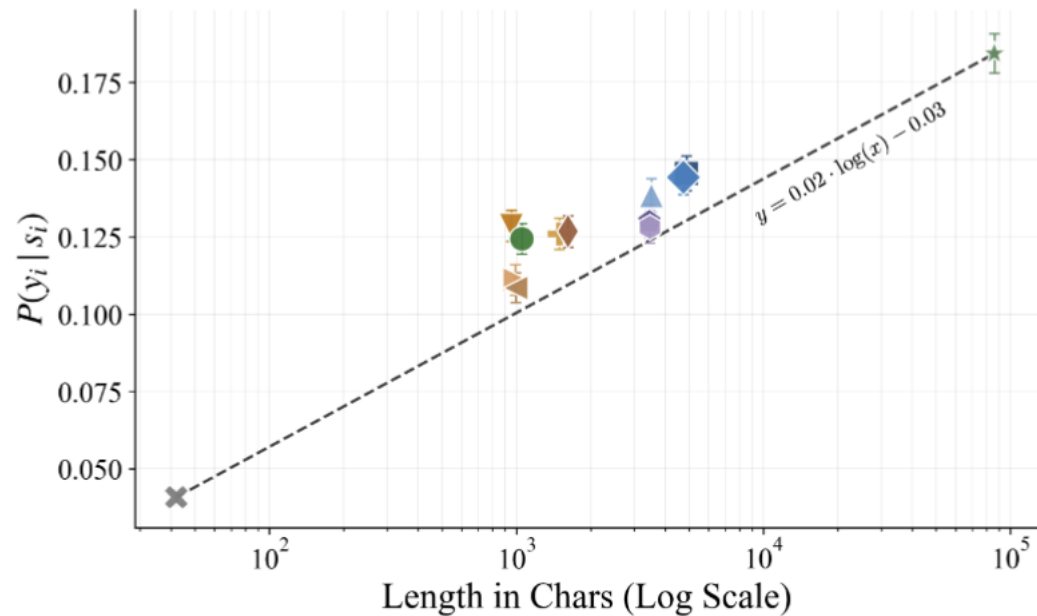
Estimating mutual information from text usually requires training a task-specific critic, which limits its use in low-data settings. We ask whether large language models can instead estimate pointwise mutual information zero-shot, using only prompts and elicited probabilities. We construct a benchmark from three publicly available human-annotated datasets with ground-truth PMI, and evaluate five information-theoretic prompting-based estimators. Our main method, PROMPTNCE, frames conditional probability estimation as a contrastive task and augments the candidate set with an explicit OTHER category. The OTHER category allows the model to assign probability mass outside the candidate set, avoiding the closed-set normalization of standard contrastive prompts. PROMPTNCE gives the best conditional probability estimates on all three datasets. For full PMI, we find that estimating label base rates is the primary bottleneck on two of the three datasets, with the best methods reaching Spearman correlation up to 0.78. We also present a case study in computer science education showing how these estimators can be used to score student knowledge summaries in a low-data setting. We release our code and prompts.



Insight Tracing: Now We Can...

- ✓ Trace when **misconceptions** started and stopped for students
- ✓ Give **summaries to teachers** before they interact with students.
- ✓ Ask **arbitrary questions** about the student state
- ✓ Measure **impact of teaching** experiences (TeachNow, Section, Lectures)
- ✓ Novel **teacher puzzles** (with principled scoring)
- ✓ **Transfer insights** from one platform to another (including offline)





Leaderboard

Rank	Team	Best model	Best accuracy	Submissions
1	MKN	gradient-boosted tree model	66.7%	3
2	AMF	DKT + IRT head (GRU-64)	65.6%	3
3	Big E	DKT	65.2%	1
4	rube_goldberg	irt	64.5%	1
5	peach	Kalman filter	63.5%	3
6	Ehm	IRT	62.6%	1