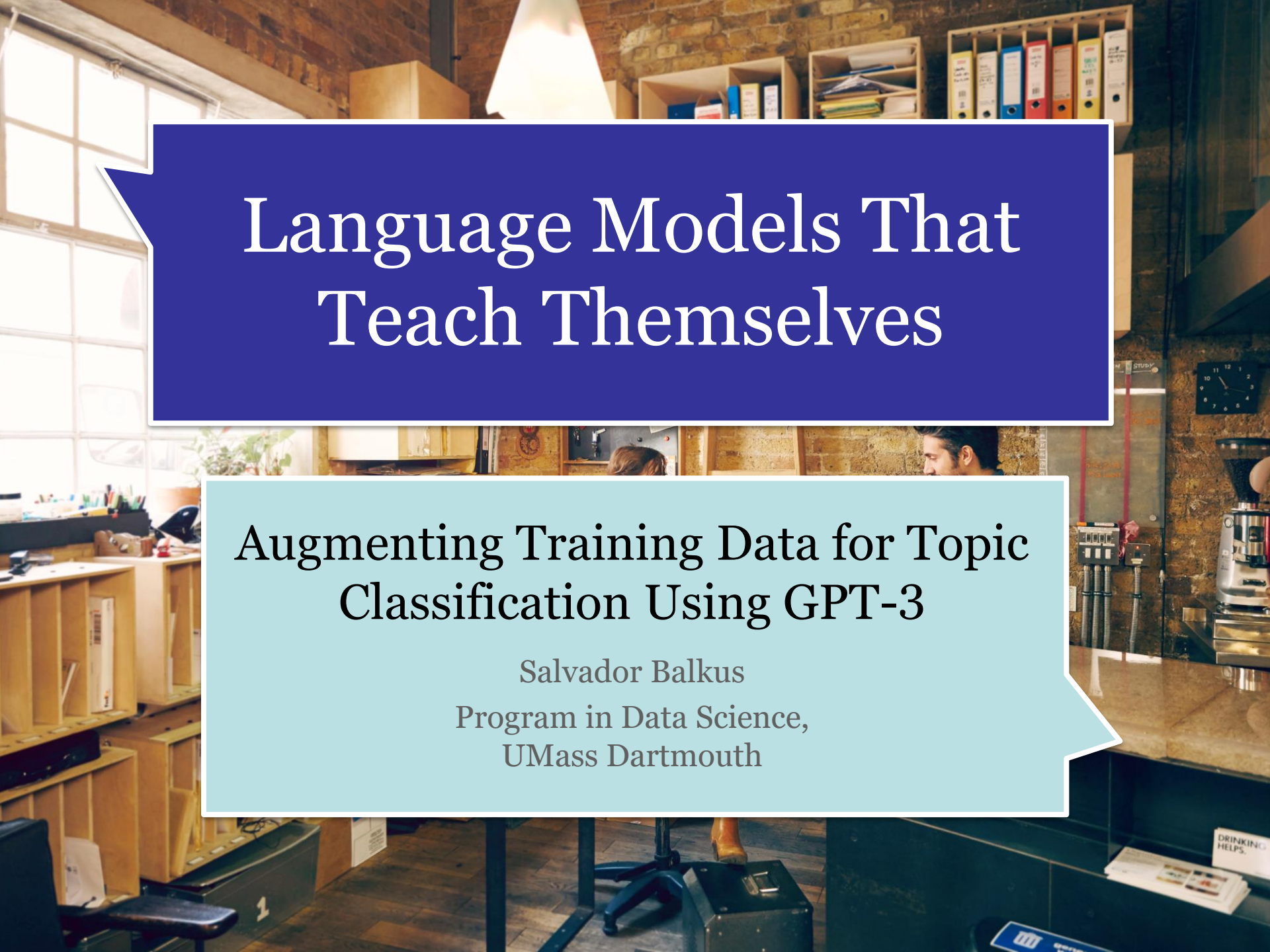


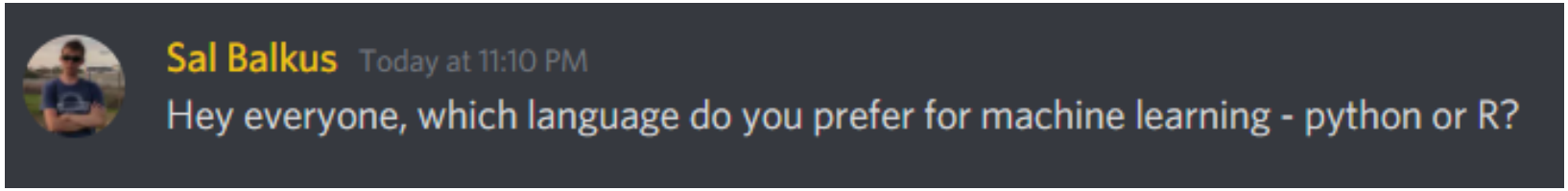
Language Models That Teach Themselves

Augmenting Training Data for Topic Classification Using GPT-3

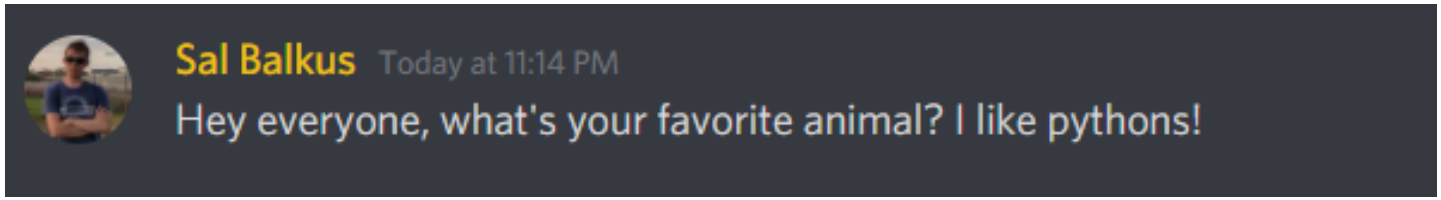
Salvador Balkus

Program in Data Science,
UMass Dartmouth

Short-Text Topic Classification



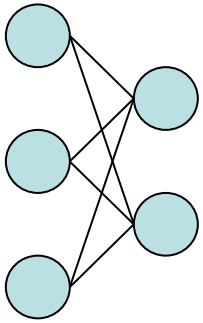
→ `Topic: Data Science`



→ `Topic: Other`

- Gathered 72 questions from UMass Dartmouth Big Data Club Discord Server

GPT-3: A Transfer Learning NLP Model [1]



Large Transformer Model (~175 billion parameters)

Train on Common Crawl Dataset (internet text)

```
It's a bird!
No, it's a
plane! No,
it's ____
```

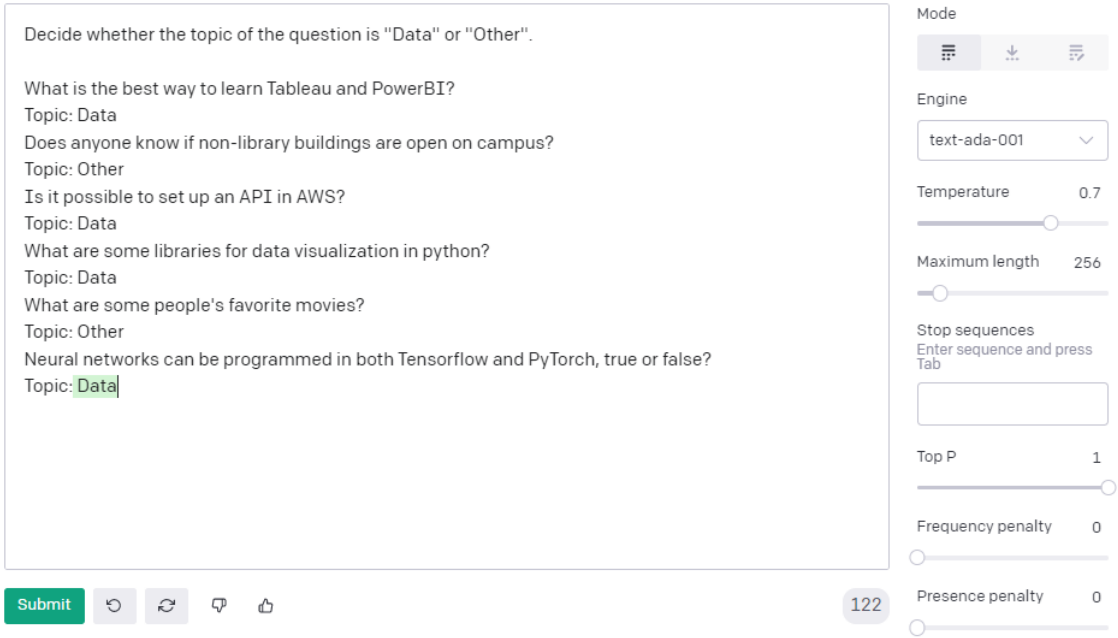
Train to predict next token in text string

Transfer to other tasks (classification, etc.)

OpenAI

```
Topic: Data Science
```

Completion Endpoint



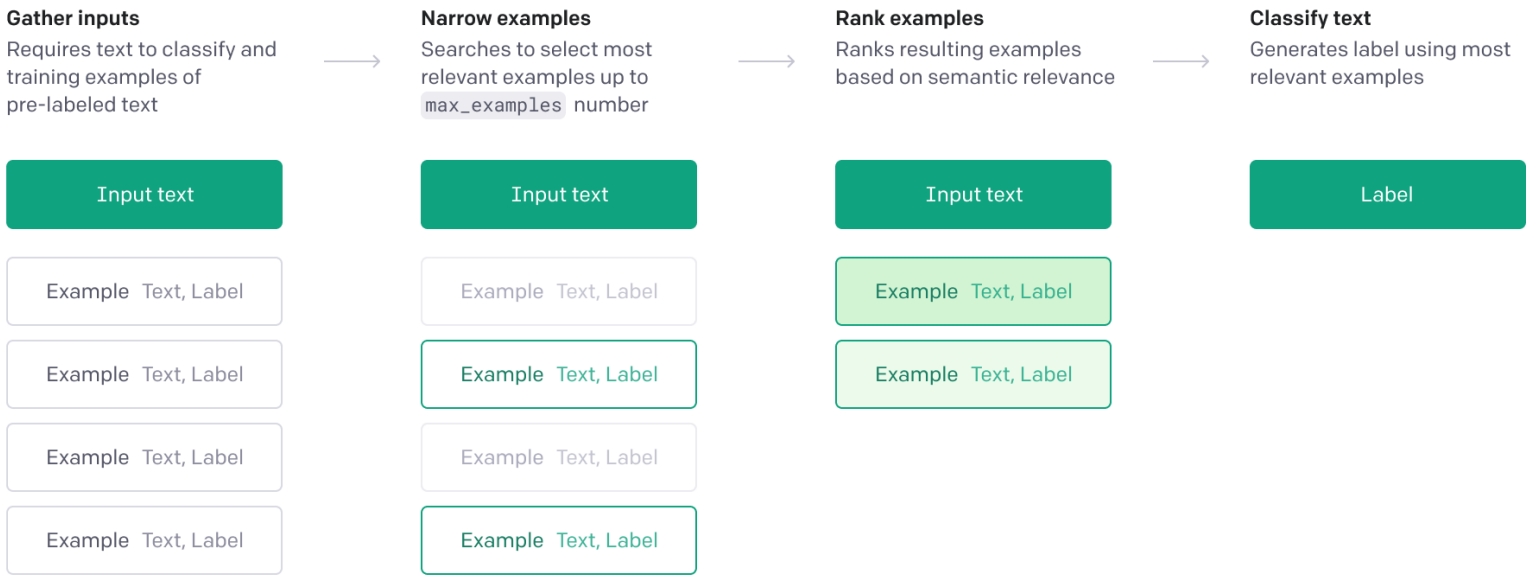
+ Only requires minimal training examples
- Sensitive to example selection

How to choose correct examples?

Classification Endpoint

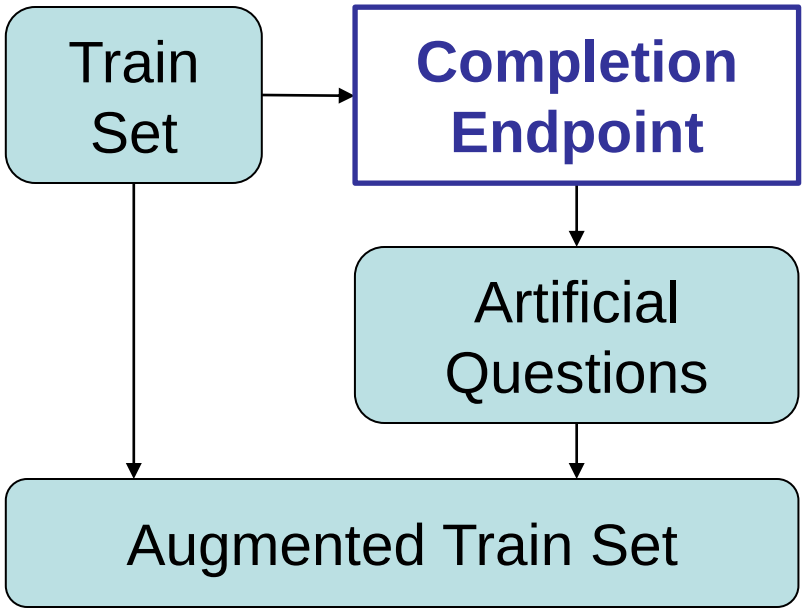
+ Specifically designed for task
- Good performance requires more examples

How to use with limited data?

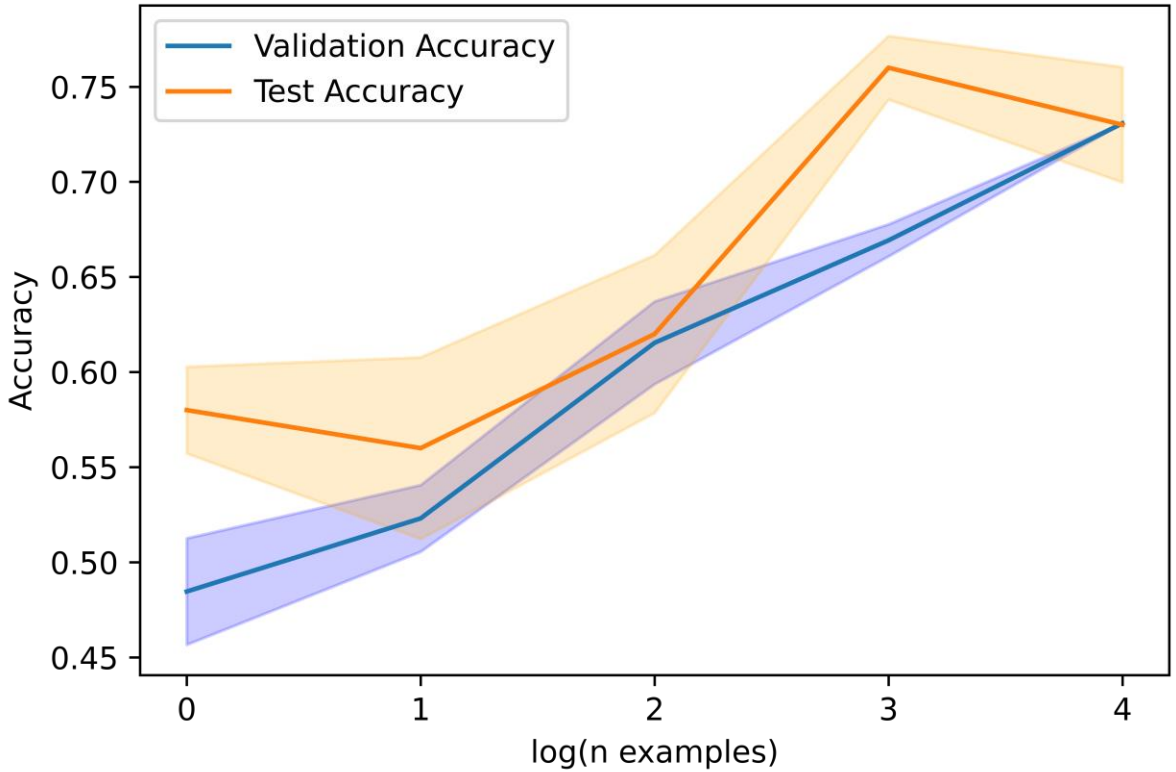


Solution: **Augment** the training data

- GPT-3 exceptional at *generating* text
- Idea: create new training examples using Completion endpoint



Results – Classification Endpoint



Classification Endpoint accuracy on Validation (n = 26) and Test (n = 20) question sets given different numbers of additional examples added to Train (n = 26) set. All questions posed by Big Data Club Discord Server.

Results – Classification Endpoint

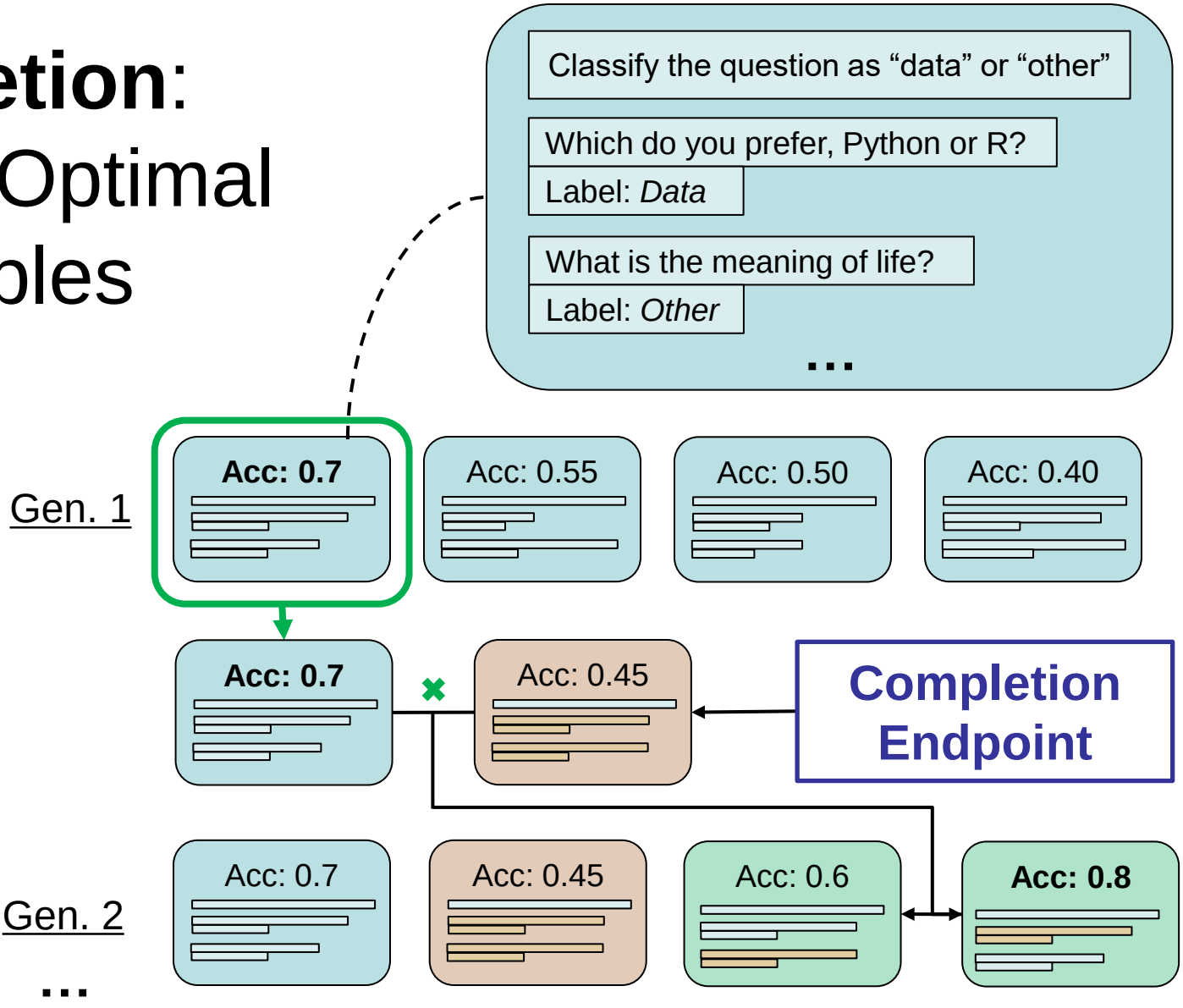
n Additional Examples	Validation Accuracy			Test Accuracy		
	μ	SE	p	μ	SE	p
0	0.485	($\pm$0.028)	-	0.58	($\pm$0.023)	-
10	0.523	($\pm$0.018)	0.328	0.56	($\pm$0.048)	0.744
100	0.615	($\pm$0.022)	0.012	0.62	($\pm$0.041)	0.471
1000	0.669	($\pm$0.008)	0.001	0.76	($\pm$0.017)	0.001
10000	0.731	($\pm$0.000)	0.001	0.73	($\pm$0.030)	0.008

GPT-3 Classification Endpoint performance on data science question topic classification, additional examples generated using GPT-3 Davinci Completion. p-values test for significance from results with no additional examples.

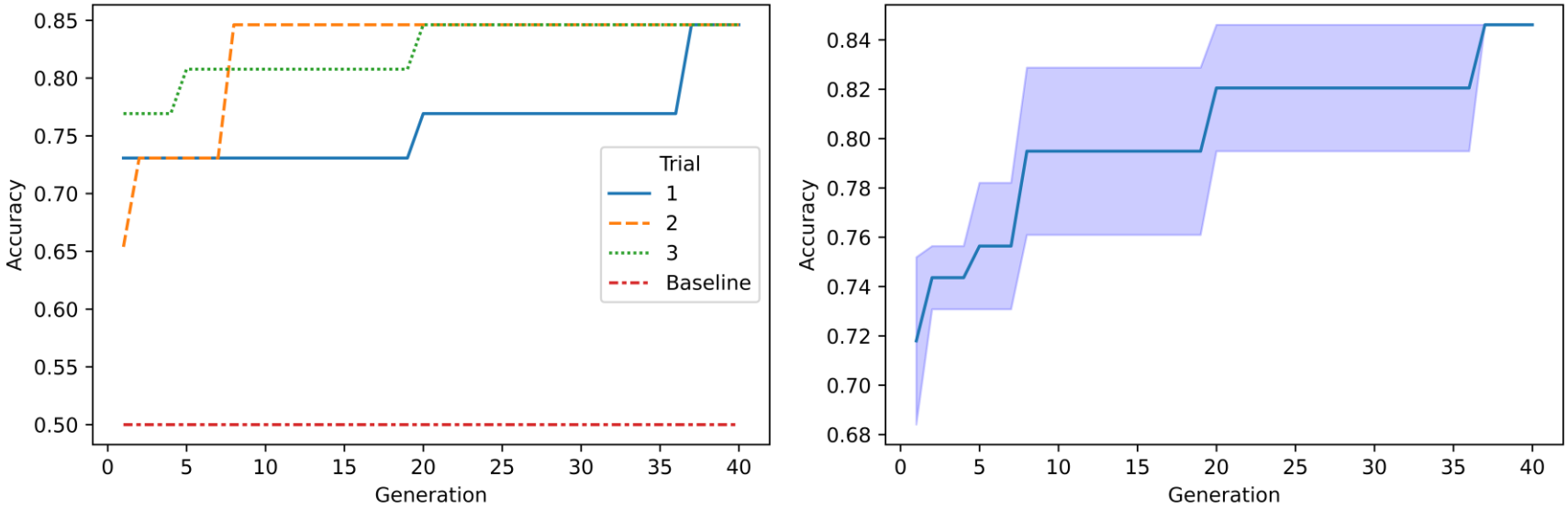
What about the Completion Endpoint?

Completion: Selecting Optimal Examples

- **Genetic algorithm** selects best examples
- GPT-3 creates *new* candidates at each generation



Validation Results – Completion Endpoint (Optimizing via Genetic Algorithm)



Completion Endpoint accuracy on Validation (n = 26) question sets over 40 generations using random subsets of 8 examples from Train (n = 26) set as initial population. All questions posed by Big Data Club Discord Server.

Test Results – Completion Endpoint (Optimizing via Genetic Algorithm)

Best Val. Accuracy	0.85 (+/- 0.00)
Test Accuracy	0.67 (+/- 0.02)

Drastic underperformance compared to validation set!

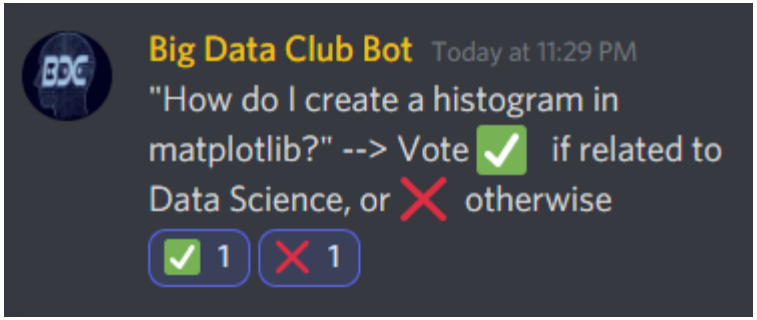
Discussion & Future Work

Subjective human language → imperfections

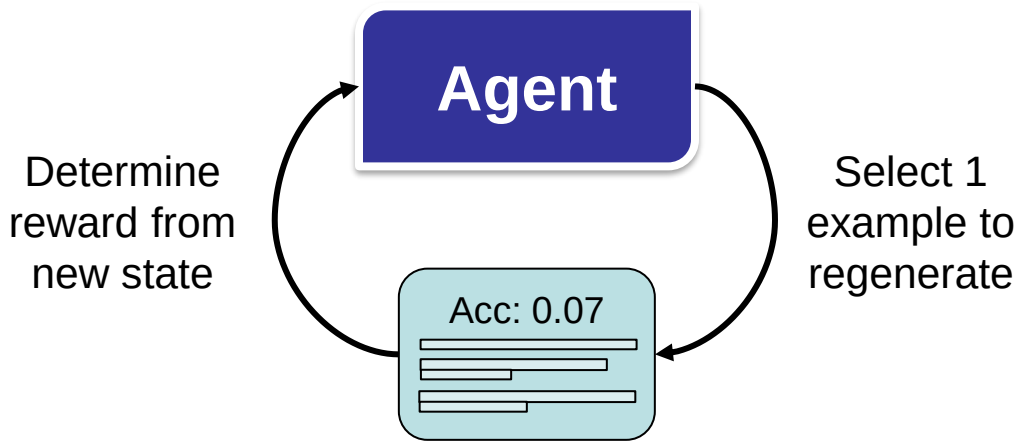
Common Mistakes:

- `“Can someone help me understand how to SSH into the computing cluster on Friday?” [Data]`
- `“Here's a link on how to make custom Jupyter notebook themes. Big Data Club-themed notebooks, anyone?” [Data]`
- `“Are you coming to Big Data Club tomorrow?” [Other]`

Want data quality control



Discussion & Future Work



Completion accuracy loss *validation* → *test* – why?

- Overfitting
- Does not "understand" prompt [2]
- GPT-3 weakest in classification [1, 3]

➢ *Potential Improvement:* **Reinforcement learning**

➢ Test other applications (sentiment, summarization, etc.)

Acknowledgements

Thank you to…

- The UMass Dartmouth **Program in Data Science**, for financial support.
- The UMass Dartmouth **Big Data Club** members for submitting questions used in this study

Thank you!

Questions?

References

[1] T Brown et al. “Language Models are Few-Shot Learners,” 2020. Available: https://arxiv.org/pdf/2005.14165.pdf.

[2] A Webson and E Pavlick. “Do Prompt-Based Models Really Understand the Meaning of their Prompts?” 2021. Available: https://arxiv.org/pdf/2109.01247.pdf

[3] R Habib “How good is GPT-3 in practice?” 2021. Available: https://humanloop.com/blog/how-good-is-gpt-3-in-practice

Appendix: GA Parameters

Parameter	Value	Parameter	Value
Encoding	String	Population size	32
Selection method	Tournament	Tournament size	4
Crossover method	Partially-matched	Crossover probability	1.0
Mutation method	Uniform	Mutation rate	0.1
Fitness function	Accuracy	Test set size	26
GPT-3 Engine	Ada	GPT-3 Temperature	0
Header	Yes	Number of alleles	8

Table 1: GA parameters for Completion Endpoint context optimization