

The background of the slide is a photograph of a modern office or cafe. It features a brick wall, wooden desks, and a coffee machine. A large window is visible on the left side. The text is overlaid on this background in two speech bubble shapes.

# Language Models That Teach Themselves

## Augmenting Training Data for Topic Classification Using GPT-3

Salvador Balkus  
Program in Data Science,  
UMass Dartmouth

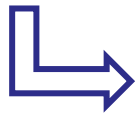


# Short-Text Topic Classification



**Sal Balkus** Today at 11:10 PM

Hey everyone, which language do you prefer for machine learning - python or R?

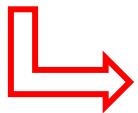


**Topic: Data Science**



**Sal Balkus** Today at 11:14 PM

Hey everyone, what's your favorite animal? I like pythons!

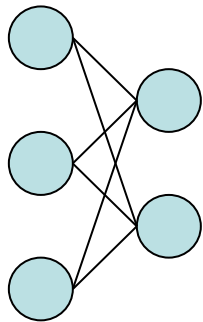


**Topic: Other**

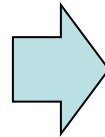
- Gathered 72 questions from UMass Dartmouth Big Data Club Discord Server



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



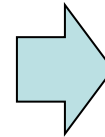
Large Transformer Model (~175 billion parameters)



WIKIPEDIA  
The Free Encyclopedia

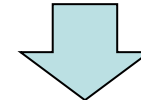


Train on Common Crawl Dataset (internet text)



It's a bird!  
No, it's a  
plane! No,  
it's \_\_\_\_

Train to predict next token in text string



Topic: Data Science

Transfer to other tasks (classification, etc.)

# Completion Endpoint

Decide whether the topic of the question is "Data" or "Other".

What is the best way to learn Tableau and PowerBI?  
Topic: Data

Does anyone know if non-library buildings are open on campus?  
Topic: Other

Is it possible to set up an API in AWS?  
Topic: Data

What are some libraries for data visualization in python?  
Topic: Data

What are some people's favorite movies?  
Topic: Other

Neural networks can be programmed in both Tensorflow and PyTorch, true or false?  
Topic: Data

Submit ↻ ↺ ↻ ↻ ↻

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Mode  
⋮ ⬇ ⋮

Engine  
text-ada-001

Temperature 0.7  
⎓

Maximum length 256  
⎓

Stop sequences  
Enter sequence and press Tab

Top P 1  
⎓

Frequency penalty 0  
⎓

Presence penalty 0  
⎓

- + Only requires minimal training examples
- Sensitive to example selection

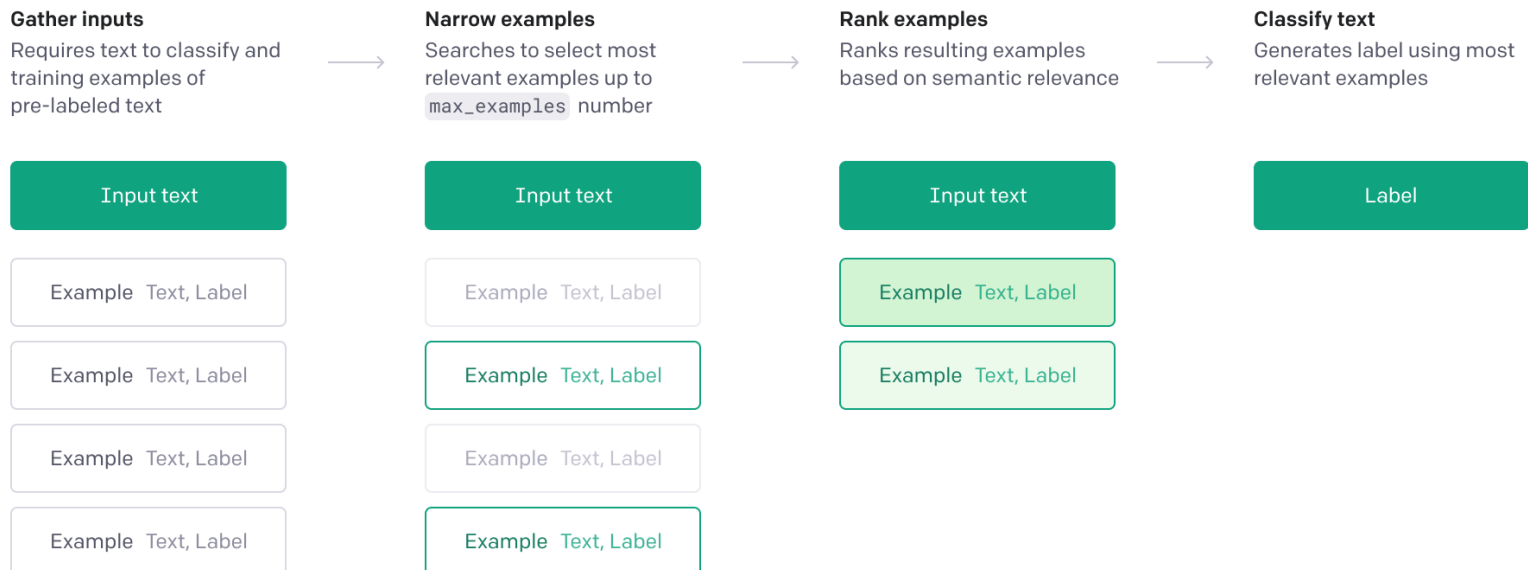
How to choose correct examples?



# Classification Endpoint

- + Specifically designed for task
- Good performance requires more examples

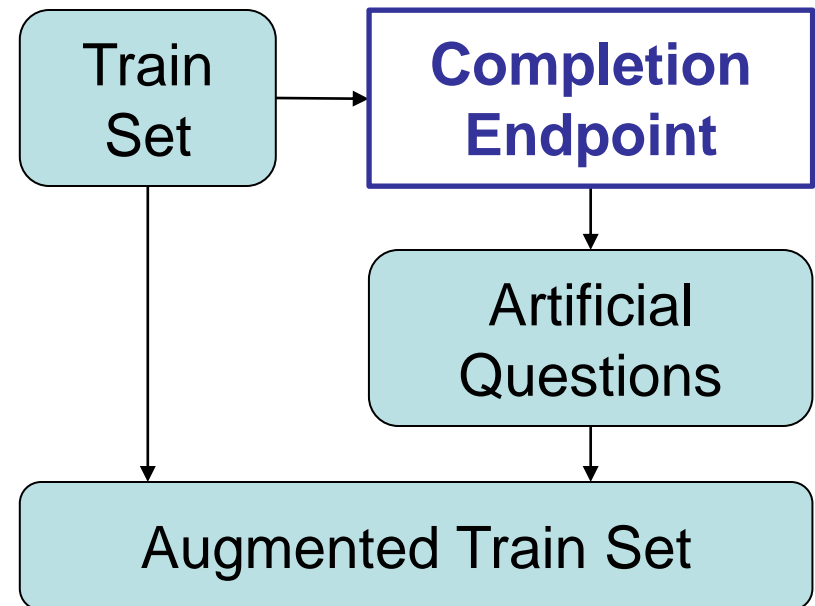
How to use with limited data?





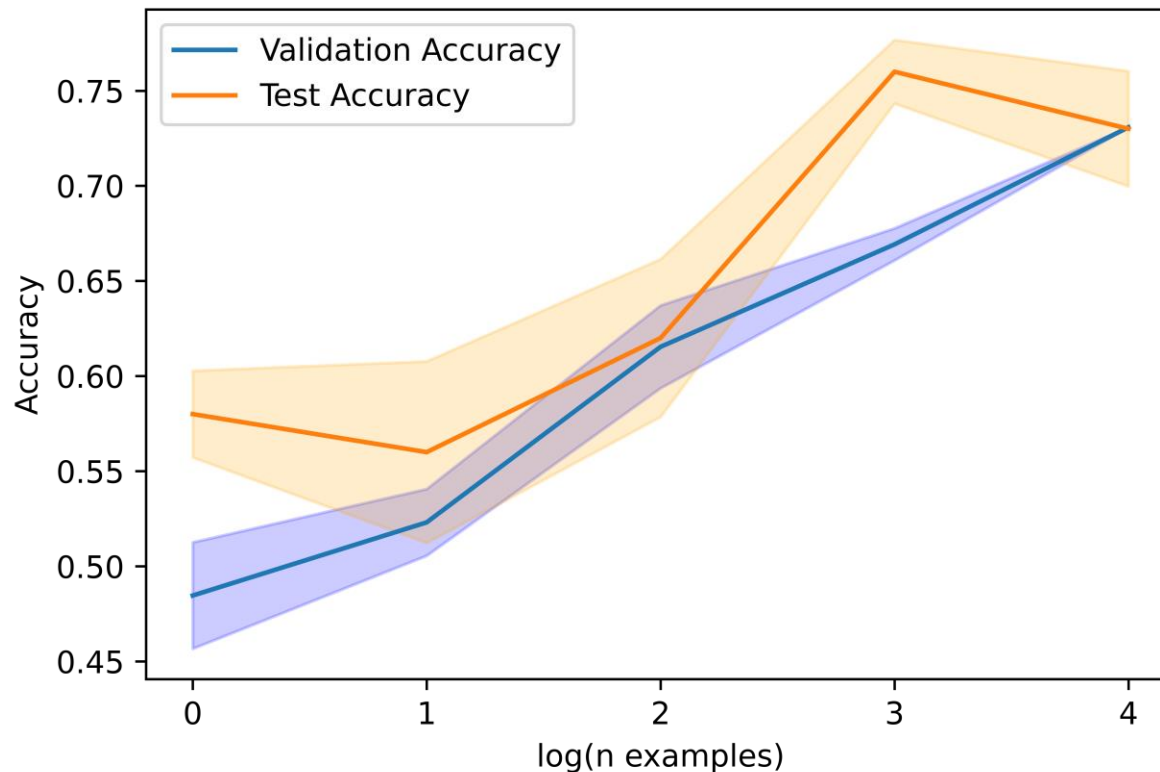
# Solution: Augment the training data

- 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		
	$\mu$	SE	$p$	$\mu$	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.



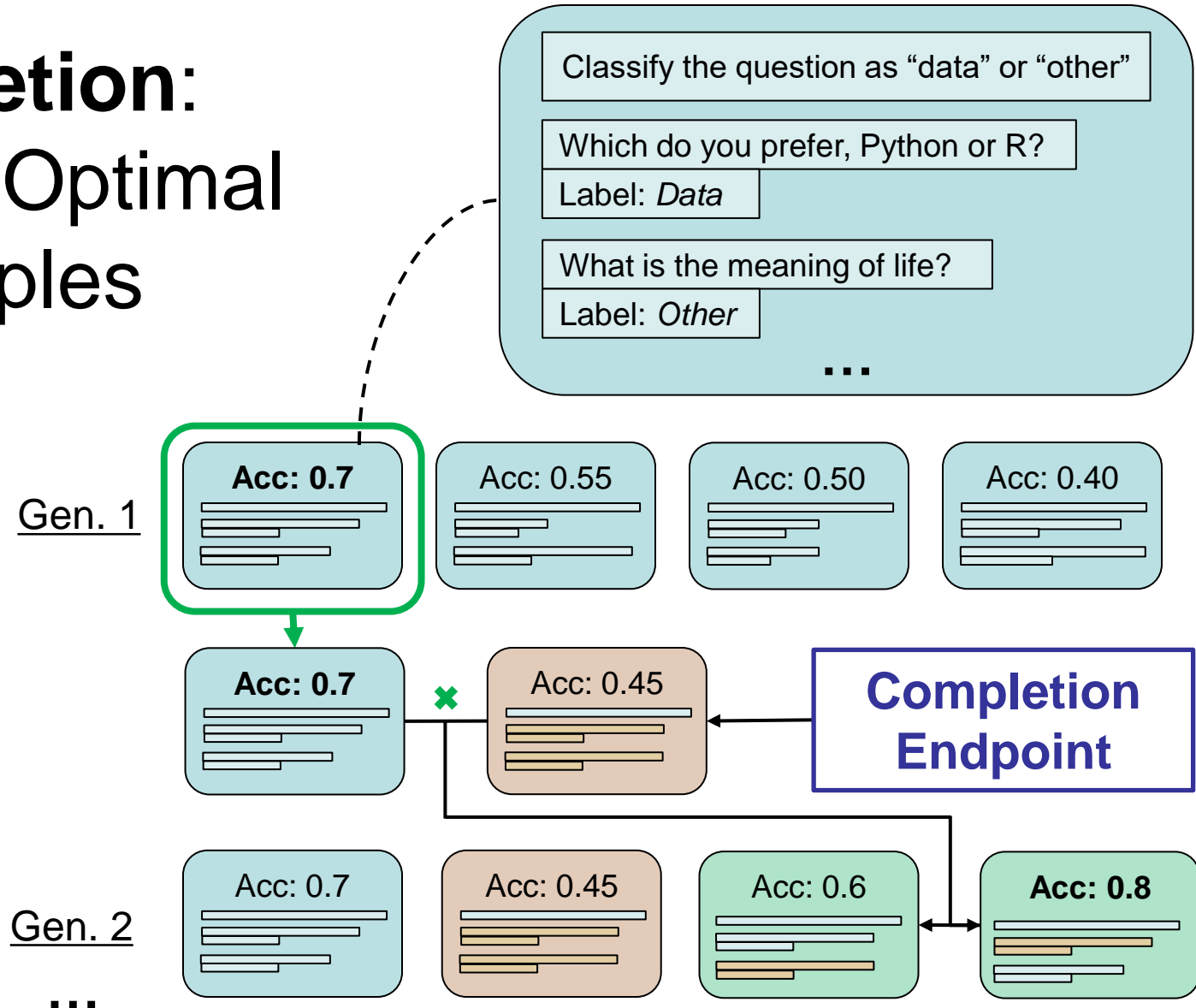


# What about the Completion Endpoint?



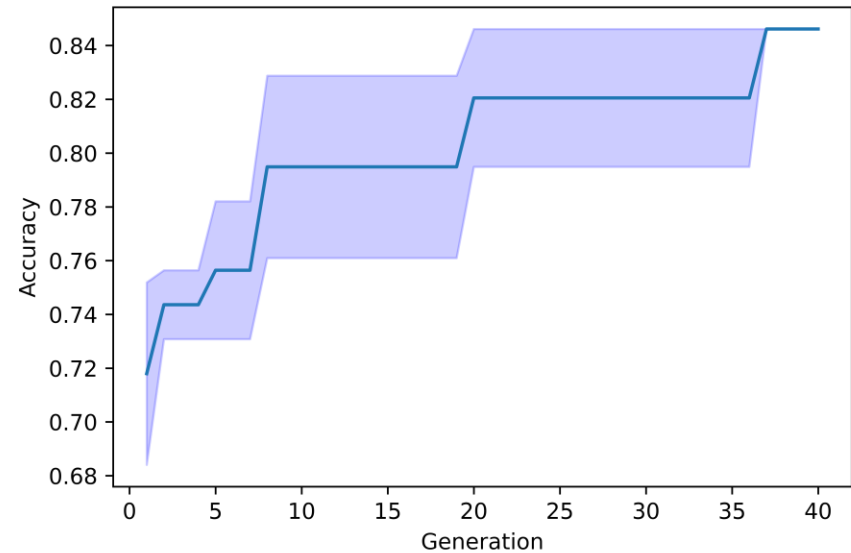
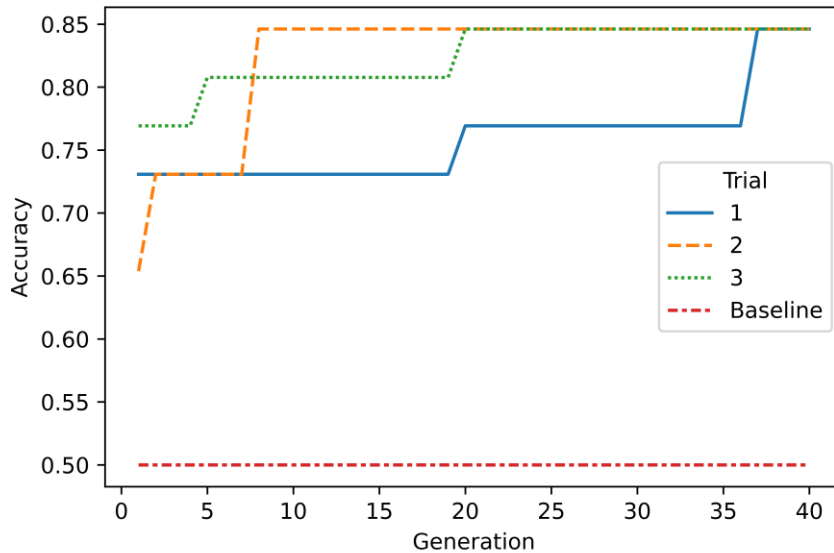
# Completion: Selecting Optimal Examples

- **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)

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

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Drastic underperformance  
compared to validation set!



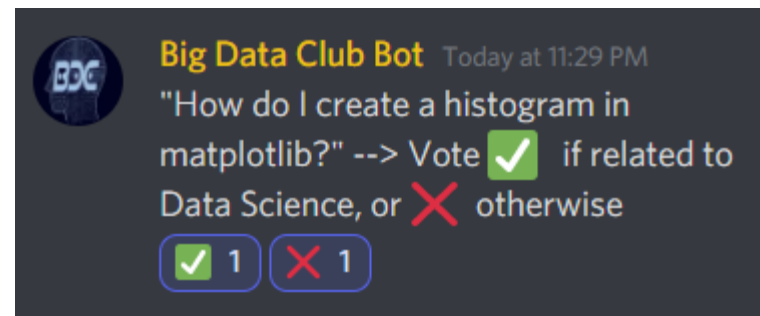
# Discussion & Future Work

Subjective human language → imperfections

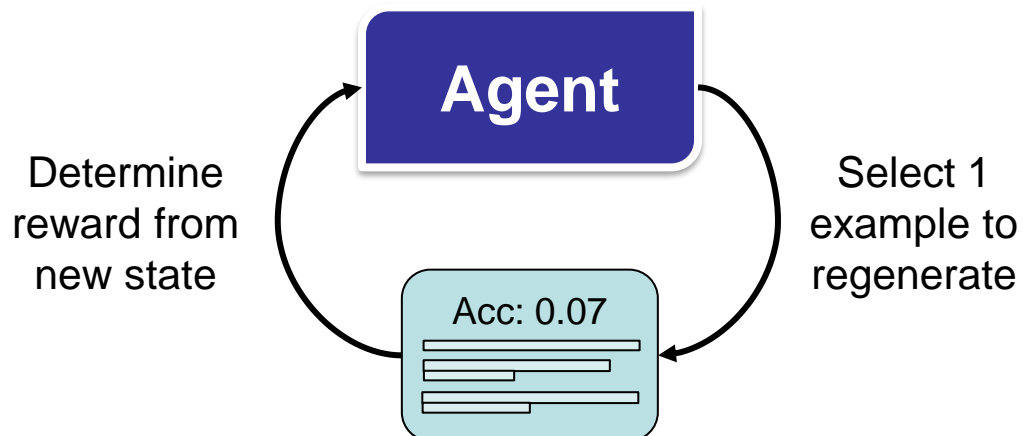
## *Common Mistakes:*

- “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