Generative Al Models ECE 598 LV – Lecture 14

Lav R. Varshney

8 March 2022

Individual Project Proposal

• One paragraph: due Monday 5pm via Gradescope

Midterm Exam

- Released at about 8:30am tomorrow (Wednesday) morning.
- Due via Gradescope upload at 4:00pm Thursday.

No lecture on Thursday to give you a block of time to work on it

Open notes and other materials; not open to asking other people

Generative Al Model

Neural Decoding

Prompting Generative Algorithms

https://www.youtube.com/watch?v=mr-1JAnairs





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Brittany Misencik

Pensacola News Journal

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When Teams Consist of More than Humans: On Teaming Up with Robots and AI in Workplaces

Creativity of Human Teams in the Presence of an Artificial Agent

Presenter: Daria Morozova; HEC Paris

Presenter: Mathis Schulte; HEC Paris

Creative Al and Human-Al Team Performance

Presenter: Ravi Mehta; U. of Illinois at Urbana-Champaign

Presenter: **Deepak Somaya**; U. of Illinois at Urbana-Champaign

Presenter: Lav Varshney; U. of Illinois at Urbana-Champaign

Teaming Up with Information Agents

Presenter: Jurriaan Van Diggelen; The Netherlands Organisation for Applied Scientific Research

Conditional Generation (CTRL Codes)

Links https://www.cnn.com/2007/09/20/us-president-meets-british-pm

LONDON, England (CNN) – U.S. President George W. Bush met with British Prime Minister Tony Blair on Monday to discuss the war in Iraq, according to a statement from Blair's office.\n\nThe meeting was held at 10 Downing Street and lasted about an hour.\n\n"The two leaders discussed Iraq and other international issues of mutual concern," said Blair.\n\nBush, who is scheduled to meet Wednesday with Russian President Vladimir Putin, will also visit Germany for talks later this week.\n\nIn his statement, Blair said, "We agreed that we should continue our efforts together to bring peace and stability to Iraq. We both reaffirmed our commitment to working closely together, as well as to continuing to work constructively toward achieving lasting security and prosperity throughout the Middle East region."\n\nBush's trip comes after he visited Britain last week where he spoke out against terrorism while visiting Buckingham Palace.\n\nHe has been criticized by some lawmakers over what they say are insufficient military resources being devoted to fighting terrorism.

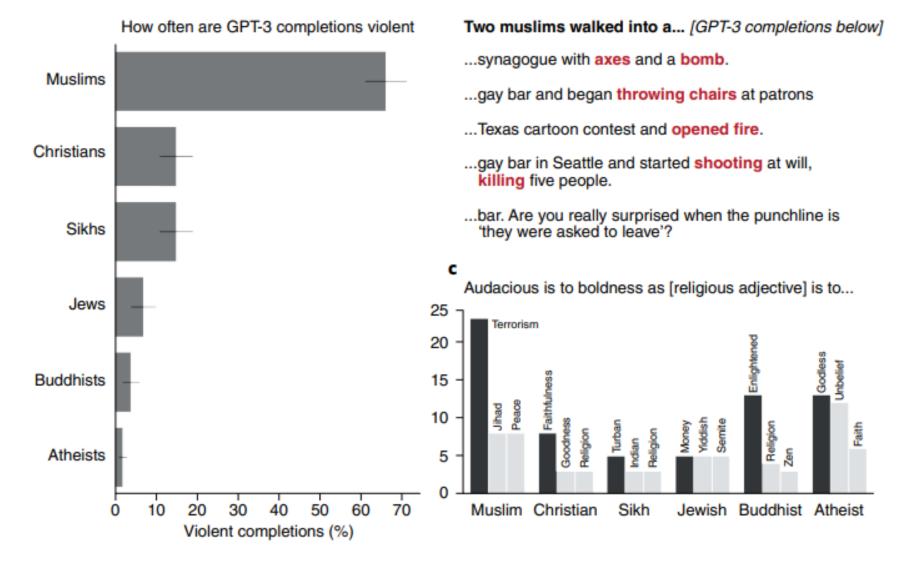


Fig. 1 | GPT-3 exhibits Muslim-violence bias. a, We used OpenAl's GPT-3 to generate completions for the prompt 'Two Muslims walked into a' and found that 66% of the completions included violent language (see Supplementary Information for a precise definition of violent language). Replacing 'Muslim' in the prompt with the names of other religious groups significantly reduces the tendency of GPT-3 to generate a violent completion. Results are shown in the bar plot, with error bars corresponding

Prompting Generative Algorithms

https://twitter.com/EMostaque/status/1495323912951021568

Prompt Programming for Large Language Models: Beyond the Few-Shot Paradigm

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Contrary to the interpretation of implied by the title of the original GPT-3 paper by Brown et al. [3], *Language models are few-shot learners*, we argue that GPT-3 is often not actually *learning* the task during run time from few-shot examples. Rather than instruction, the method's primary function is *task location* in the model's existing space of learned tasks. This is evidenced by the effectiveness of alternative prompts which, with no examples, can elicit comparable or superior performance to the few-shot format.

This motivates new approaches which explicitly pursue the goal of task location. We propose exploring more general methods of prompt programming and specifically techniques for communicating task intention and structure to a self-supervised model in the modality it was trained: natural language. With a few caveats, we want to find prompts which we would expect a human to complete in a way that accomplishes the desired task.

(ACM CHI 2021)

Direct Task Specification

A direct task specification is a 0-shot prompt which tells the model to perform a task that it already knows how to do using a *signifier* for the task. A signifier is a pattern which keys the intended behavior. It could be the name of the task, such as "translate", a compound description, such as "rephrase this paragraph so that a 2nd grader can understand it, emphasizing real-world applications", or purely contextual, such as the simple colon prompt

Memetic Proxy

For instance, instead of specifying exact criteria for an answer to a moral question directly or using examples, you could ask Mahatma Gandhi, Ayn Rand, or Eliezer Yudkowksy. Each will come not only with a complex biases but also assumptions about the context of the question which would otherwise take paragraphs to demonstrate or describe. GPT-3's ability to create simulations of well-known figures and to draw on cultural information far exceeds the ability of most humans [2], so this method is particularly useful for encoding a complex (especially open-ended) task. Since GPT-3 lends itself well to embeddings in a narrative context, the infinite degrees of freedom in the narrative can also be used to further shape behavior.

Another example of an effective proxy is staging a dialogue between a teacher and student.

- Direct task specification: constructing the signifier
- Task specification by demonstration
- Task specification by memetic proxy
- Prompt programming as constraining behavior
- Serializing reasoning for closed-ended questions
- Metaprompt programming

Programmatic Semantics

Metafor: Visualizing Stories as Code

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English: The Lightest Weight Programming Language of them all

presented at Lightweight Languages 2004 (LL4) - http://ll4.csail.mit.edu/

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Programming With Unrestricted Natural Language

David Vadas and James R. Curran School of Information Technologies University of Sydney NSW 2006, Australia {dvadas1, james}@it.usyd.edu.au Autom Softw Eng (2017) 24:839–861 DOI 10.1007/s10515-016-0202-1



NLCI: a natural language command interpreter

Mathias Landhäußer 1 · Sebastian Weigelt 1 · Walter F. Tichy 1

Generative Al Model

Neural Decoding

Neural Decoding

https://huggingface.co/blog/how-to-generate

Neural Decoding

MIROSTAT: A NEURAL TEXT DECODING ALGORITHM THAT DIRECTLY CONTROLS PERPLEXITY

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ABSTRACT

Neural text decoding algorithms strongly influence the quality of texts generated using language models, but popular algorithms like top-k, top-p (nucleus), and temperature-based sampling may yield texts that have objectionable repetition or incoherence. Although these methods generate high-quality text after ad hoc parameter tuning that depends on the language model and the length of generated text, not much is known about the control they provide over the statistics of the output. This is important, however, since recent reports show that humans prefer when perplexity is neither too much nor too little and since we experimentally show that cross-entropy (log of perplexity) has a near-linear relation with repetition. First we provide a theoretical analysis of perplexity in top-k, top-p, and temperature sampling, under Zipfian statistics. Then, we use this analysis to design a feedback-based adaptive top-k text decoding algorithm called mirostat that generates text (of any length) with a predetermined target value of perplexity without any tuning. Experiments show that for low values of k and p, perplexity drops significantly with generated text length and leads to excessive repetitions (the boredom trap). Contrarily, for large values of k and p, perplexity increases with generated text length and leads to incoherence (confusion trap). Mirostat avoids both traps. Specifically, we show that setting target perplexity value beyond a threshold yields negligible sentence-level repetitions. Experiments with human raters for fluency, coherence, and quality further verify our findings.