GenAl HW7 Understanding what Al is thinking

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Deadline: 2024/5/16 23:59:59 (UTC+8)

Outline

- Introduction
- Task 1: Token Importance Analysis
- Task 2: LLM Explanation
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Link

Colab

COOL Quiz

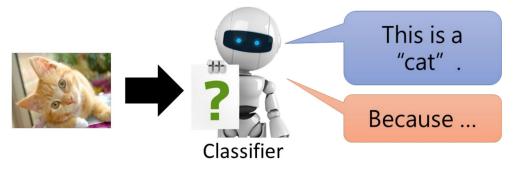
Questions

Introduction

Why should we know what generative AI is thinking?

- Correct answer ≠ Intelligent
- Explanation is essential in high-stakes applications, e.g., medicine and law.
- We can improve our model based on our explanation.

Model Explanation





Is this review positive? Why or why not? "A fantastic movie directed by the famous Lucas Johnson, who has a track-record of producing terrific novel adaptions."

This review is very positive, as judged by the positive words "fantastic" and "terrific".



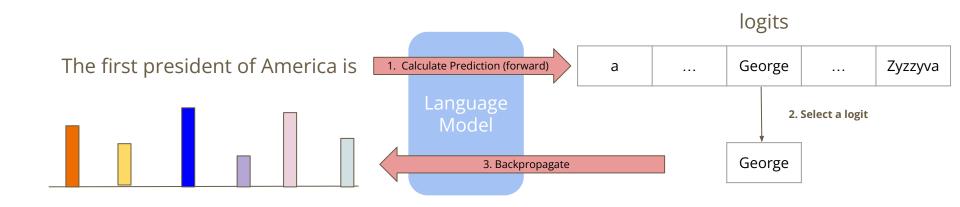
Task 1: Token Importance Analysis

Task Description

- In this task, we aim to understand what tokens play important roles in generating the response.
- We utilize feature attribution methods to analyze the importance.
 - Gradient-based approach
 - Attention-mechanism
- Run the sample code and finish question 1 to 7.

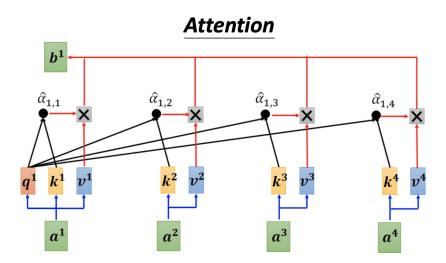
Gradient-based Approach (saliency)

Compute the gradient of the target logit with respect to the input tokens.



Attention-mechanism

- Commonly used in transformer-based models.
- Shows which tokens the model attends to when generating the output.



Token Visualization

- In this task, we use https://github.com/inseq-team/inseq/ to visualize the importance of token when generating the response.
- It supports many feature attribution methods, including gradient and attention, which we will use in this homework.

Inseq

```
method
                                    model
    model = inseq.load_model("gpt2", "saliency")
    out = model.attribute(
         "Hello ladies and"
Input
         generation_args={"max_new_tokens": 9},
         n steps=500,
         internal batch size=50
    out.show()
The attention mask and the pad token id were not set. As a consequence, you may observe unex
    Setting 'pad_token_id' to 'eos_token_id':50256 for open-end generation.
    WARNING:inseq.attr.feat.attribution_utils:Unused arguments during attribution: {'n_steps': !
    Attributing with saliency...: 100% 2000 12/12 [00:00<00:00, 36.16it/s]
    0th instance:
        Target Saliency Heatmap
        x: Generated tokens, y: Attributed tokens
                   gentlemen
                                                      member
                                                               of
                                                                     the
                                                                          Board
          Hello
                     0.302
                             0.234 0.171
                                         0.249 0.106
                                                       0.199
                                                              0.132 0.092
                                                                          0.156
          ladies
                     0.509
                             0.283 0.227
                                         0.292
                                               0.119
                                                       0.243
                                                              0.138
                                                                    0.093
                                                                           0.151
           and
                     0.189
                              0.21
                                  0.213 0.085
                                               0.255
                                                       0.082
                                                              0.09
                                                                    0.185
                                                                           0.075
         gentlemen
                             0.274 | 0.148 | 0.189 | 0.099
                                                      0.195
                                                              0.108 0.094
                                                                           0.156
                                                                           0.054
                                    0.24
                                         0.073
                                               0.223
                                                       0.057
                                                              0.071
                                                                    0.154
                                         0.112
                                               0.083
                                                       0.061
                                                              0.063
                                                                    0.046
                                                                           0.054
           am
                                               0.115
                                                       0.08
                                                              0.054
                                                                    0.034
                                                                           0.05
                                                              0.086
                                                                    0.131
                                                                           0.052
                                                       0.084
                                                                     0.09
                                                                           0.172
         member
                                                              0.258
            of
                                                                    0.081
                                                                           0.029
           the
                                                                           0.049
          Board
```

Saliency map of machine translation task

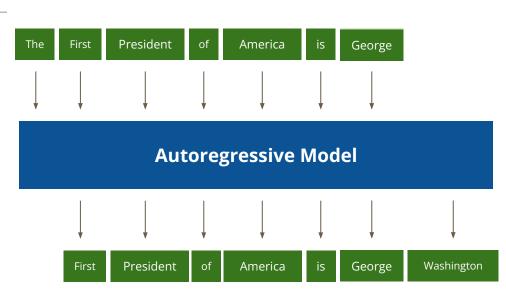
Source Saliency Heatmap Output x: Generated tokens, y: Attributed tokens 美国 第一 总统 乔治 </s> 0.052 0.089 0.204 0.193 0.056 0.035 0.113 0.02 0.066 0.109 _The 0.01 __first 0.108 0.274 0.178 0.103 0.064 0.033 0.064 0.008 0.016 0.032 0.093 president 0.191 0.202 0.285 0.093 0.057 0.072 0.014 0.023 0.036 0.129 0.036 0.105 0.082 0.103 0.089 0.057 0.026 0.02 0.019 0.023 0.057 _of 0.073 0.061 0.089 0.09 0.066 0.047 0.022 0.027 0.031 0.107 America 0.308 _is 0.079 0.057 0.04 0.059 0.071 0.051 0.023 0.028 0.039 0.084 0.341 0.052 0.035 0.044 0.079 0.114 George 0.048 0.142 0.364 0.139 0.079 0.061 Wash 0.046 0.03 0.041 0.035 0.08 0.206 0.282 0.473 0.135 0.102 in 0.017 0.012 0.033 0.02 0.025 0.065 0.08 0.189 0.139 0.1 0.045 0.015 0.017 0.044 0.032 0.02 0.049 0.066 0.072 0.142 0.401 0.075 ton 0.028 0.049 0.076 0.026 0.018 0.049 0.046 0.031 0.018 0.035 0.084 </s> 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 probability 0.696 0.495 0.925 0.889 0.683 0.802 0.603 0.485 0.397 0.897 0.583

Input

The first president ... **Translation** Model 美國第一任總統...

Saliency map of sentence completion task

Target Saliency Heatmap Generated tokens x: Generated tokens, y: Attributed tokens _George _Washington _Unterscheidung between __The 0.229 0.176 0.255 0.194 0.06 0.079 __first 0.107 0.093 0.059 0.031 _president 0.174 0.15 0.114 0.09 0.026 Input 0.142 0.116 _of 0.141 0.085 0.083 __America 0.219 0.181 0.104 0.072 0.024 0.129 0.12 0.145 0.1 0.069 _is 0.103 __George 0.139 0.097 0.034 0.084 __Washington 0.103 0.031 0.2 0.353 _Unterscheidung 0.29 __between probability 0.66 0.908 0.992 0.023 0.535



Importance score

Example of saliency map

Q: When generating the word "Washington", what's the importance score of "America"?

A: 0.181

Target Saliency Heatmap

x: Generated tokens, y: Attributed tokens

	George	Washington		Unterscheidung	_between
_The	0.229	0.176	0.255	0.194	0.06
first	0.107	0.093	0.079	0.059	0.031
president	0.174	0.15	0.114	0.09	0.026
_of	0.141	0.142	0.116	0.085	0.083
America	0.219	0.181	0.104	0.072	0.024
_is	0.129	0.12	0.145	0.1	0.069
George		0.139	0.103	0.097	0.034
Washington			0.084	0.103	0.031
				0.2	0.353
Unterscheidung					0.29
_between					
probability	0.908	0.992	0.66	0.023	0.535

Task 2: LLM Explanation

LLM Explanation

- LLMs have the ability to explain in natural language.
- It is much more straightforward to understand than prior methods.



Is this review positive? Why or why not? "A fantastic movie directed by the famous Lucas Johnson, who has a track-record of producing terrific novel adaptions."

This review is very positive, as judged by the positive words "fantastic" and "terrific".



Task Description

- In this task, we aim to assess the effectiveness of LLM explanation.
- We will explore two LLM explanation approaches.
 - Providing the explanation for the model's answer.
 - Simulating the feature attribution method in task 1 to see the importance of tokens.
- Run the given prompts on ChatGPT and finish Question 8 to 10.
 (No need of Colab)

Explain the model's answer

Directly ask the LLM to explain its answer.

Prompt:

You are a creative and intelligent movie review analyst, whose purpose is to aid in sentiment analysis of movie reviews. Determine whether the review below is positive or negative, and explain your answers.

Review: This film is a compelling drama that captivates audiences with its intricate storytelling and powerful performances.

Simulate feature attribution methods with LLM explanation

 Ask the LLM to explain the importance of the input tokens in contributing to the answer, similar to what we do in task 1.

2310.11207.pdf (arxiv.org)

Simulate feature attribution methods with LLM explanation

Prompt:

You are a movie review analyst tasked with sentiment analysis. For each review, provide a list of tuples representing the importance of each word and punctuation, with values ranging from -1 (negative) to 1 (positive). Then, classify the review as positive (1) or negative (-1). The review is within <review> tags.

Example output:

[(<word or punctuation>, <float importance>), ...]

<int classification>

<review> This film is a compelling drama that captivates audiences with its intricate storytelling and powerful performances. <review>

Note: ChatGPT's responses may vary due to randomness. If the format isn't as desired, please try again.

2310.11207.pdf (arxiv.org)

Submission & Deadline

Submission

- Finish questions on NTU COOL Quiz
- Unlimited times of submissions for the quiz, but only the latest submission will be considered when grading
- No late submission is allowed

Important dates

Deadline for Submission (NTU Cool)

2024/05/16 23:59:59 (UTC+8)

Grading Release Date

2024/05/31 23:59:59 (UTC+8)

Contact

If You Have Any Questions

- NTU Cool HW7 作業討論區
 - 如果同學的問題不涉及作業答案或隱私,請**一律使用**NTU Cool 討論區
 - 助教們會優先回答NTU Cool討論區上的問題
- Email: ntu-gen-ai-2024-spring-ta@googlegroups.com
 - Title should start with [GenAl 2024 Spring HW7]
 - Email with the wrong title will be moved to trash automatically
- TA Hours
 - o Time:
 - **5/3, 16:30 ~ 17:20**
 - **5/10 13:20~14:10, 16:30 ~ 17:20**
 - Location: 綜合大講堂