

INTRODUCTION

- Contextual Commonsense Inference (CCI): inferring causal relations between events in text
- GLUCOSE [1] dataset is valuable, but GLUCOSE task conflates CCI and generation (NLG)
- Finetuning T5 [2] LM \rightarrow model relies on NLG
- Evaluating with BLEU \rightarrow partial matches count

Parameter	Text Input
Story	Fred woke up late. He just missed his bus. He then went to his mom's room. His mom then drives him to school. He makes it to first class on time.
Selected Sentence (X)	Fred woke up late.
Dimension	6
Specific Rule	Fred wakes up late >Causes/Enables> Fred misses his bus
General Rule	Someone _A wakes up late >Causes/Enables> Someone _A misses Something _A

Output

METHODS

1. Diagnose extent to which CCI and NLG are conflated in the GLUCOSE task

Task	Input
ORIGINAL	1: My mother told me to fix the car. I was
	unable to do this right away. * I could not find
	my tools. * I looked everywhere for them. It
	turns out they were stolen the night before.
HISTORY	1: My mother told me to fix the car. I was
	unable to do this right away.
Mask X	My mother told me to fix the car. I was unable
	to do this right away. <masked> I looked</masked>
	everywhere for them. It turns out they were
	stolen the night before.
HISTORY+X	1: My mother told me to fix the car. I was
	unable to do this right away. * I could not find
	my tools. *
Output:	

They were stolen the night before >Causes/Enables> I **could not find my tools** ** Something_A is stolen >Causes/Enables> Someone_A cannot find Something_A

CIS²: A Simplified Commonsense Inference Evaluation for Story Prose Bryan Li, Lara J. Martin, and Chris Callison-Burch RESULTS Transformers trained on with generation commonsense inference Tas ORI tasks will rely on their HIS⁻ MA HIS⁻ generation ability, if given the chance, which 70% 61.9% 60% 50% 40% inflates performance. 30% 20% 12.9%

2. Suggest an alternative evaluation: CIS² (Contextual Commonsense Inference in Sentence Selection)

Abstract away from NLG, only consider causal links between sentences

- *Convert* GLUCOSE to CIS² labels using SBERT[3] similarity metric
- 2. Train T5 on CIS² converted data to compare to GLUCOSE models
- *Evaluate* using **exact match** CIS² accuracy

CONVERSION

Corresponding	GLUCOSE Input	G
Tags	6:	
<s0></s0>	* Fred woke up late. * story	>
<s1></s1>	He just missed his bus.	٠.
<s2></s2>	He then went to his mom's room.	
<\$3>	His mom then drives him to school.	
<\$4>	He makes it to the first class on time.	and the second se

Output: <s₄> >Causes/Enables> <s₂>

GLUCOSE Output (Specific Rule)

Fred wakes up late

>Causes/Enables>

Fred misses his bus

Resulting CIS² Tags <s₀>

>causes/enables>

<s1>

10% 0%

Fred wakes up late >Causes/Enables> Fred misses his bus ** Someone_△ wakes up late >Causes/Enables> Someone_△ misses Something

• T5 will rely on generation if it is an option. • CCI tasks need to put less emphasis on generation in order to evoke the "reasoning" capabilities of Transformers.

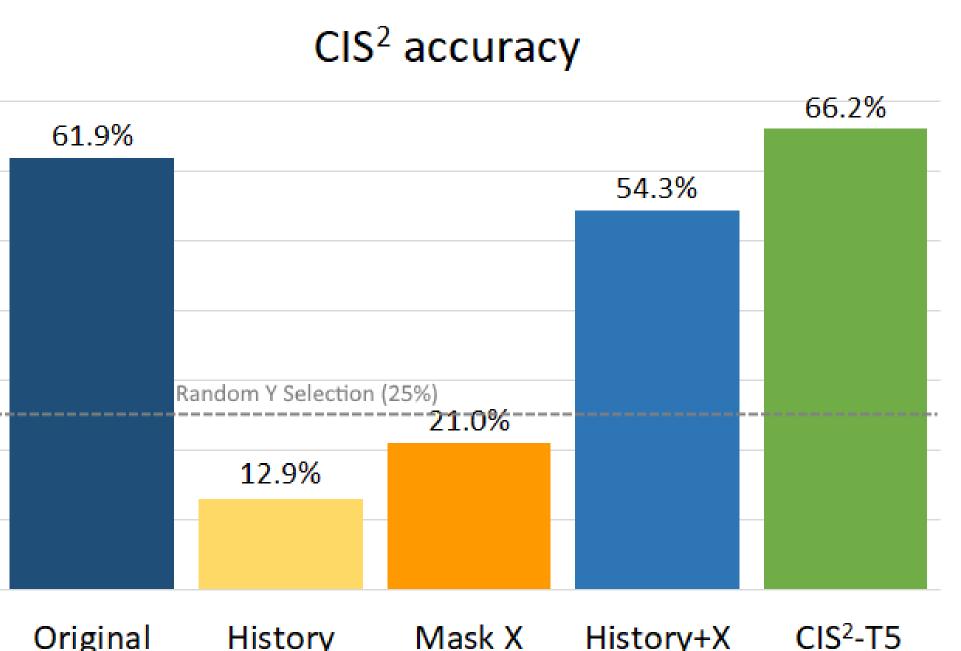


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1. Diagnose where CCI is being conflated

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s k	Specific Rule (SacreBLEU)	General Rule (SacreBLEU)
IGINAL	70.7	66.2
TORY	35.9	50.4
SK X	41.6	49.6
TORY+X	68.3	65.5

2. Suggest alternative evaluation: CIS²



DISCUSSION

Input #6: *Fred woke up late. * He just missed his bus. He then went to his mom's room. His mom then drives him to school. He makes it to first class on time. Output

[1] Nasrin Mostafazadeh et al. <u>GLUCOSE: GeneraLized and COntextualized</u> story explanations. In EMNLP, 2020.

[2] Colin Raffel, et al. Exploring the limits of transfer learning with a unified text-to-text transformer. Journal of Machine Learning Research, 2020. [3] Nils Reimers and Iryna Gurevych. <u>Sentence-BERT: Sentence Embeddings</u> using Siamese BERT-Networks. EMNLP-IJCNLP, 2019.