

# A Simple Method for Commonsense Reasoning

Trieu and Quoc



Editorial Introduction to the Special Articles in the Spring Issue

## **Beyond the Turing Test**

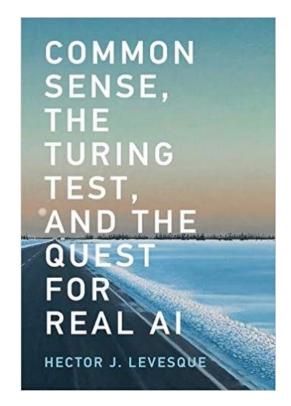
Gary Marcus, Francesca Rossi, Manuela Veloso

## On our best behaviour

Hector J. Levesque

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#### ??

??



The **racecar** zoomed by the **school bus** because **[it]** was going so fast.

The racecar zoomed by the school bus because [it] was going so fast.

**Comment:** From (Levesque 2009); deliberately created as a non-example. "Fast" is associated with racecars.

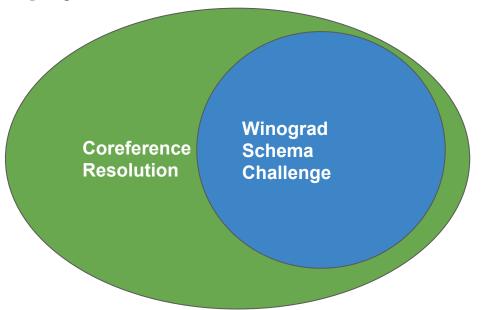
The racecar zoomed by the school bus because [it] was going so fast.

**Comment:** From (Levesque 2009); deliberately created as a non-example. "Fast" is associated with racecars.

Frank was pleased when Bill said [he] was the winner of the competition.

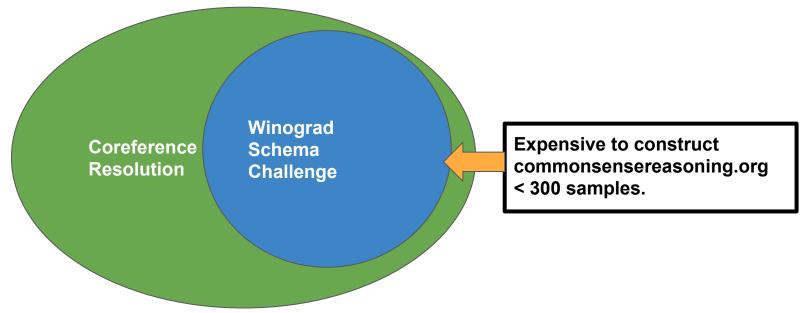
**Comment:** From (Levesque 2009); deliberately created as a non-example. The version with "pleased" is genuinely ambiguous (i.e. to the human reader). Frank might well be pleased on learning that Bill was the winner.

• The trophy cannot fit in the suitcase because it is too big.



#### On our best behaviour

Hector J. Levesque
Dept. of Computer Science
University of Toronto



• The **trophy** cannot fit in the **suitcase** because *it* is too big.

Random Guess: ~50%

Human: ~90%

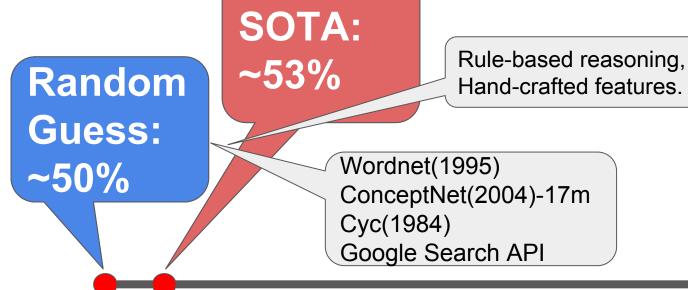
• The **trophy** cannot fit in the **suitcase** because *it* is too big.

Random Guess: ~50%

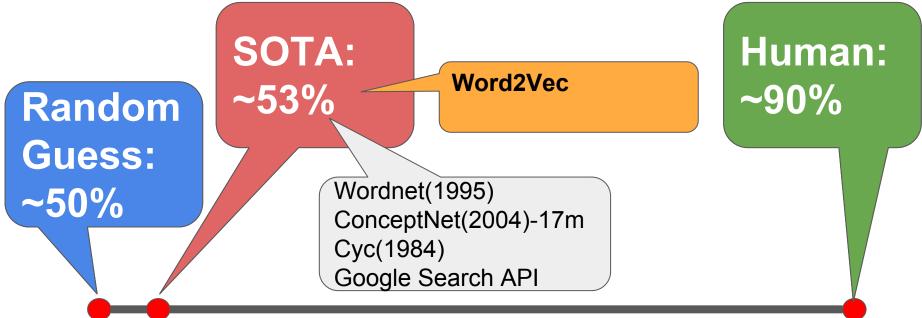
SOTA: ~53%

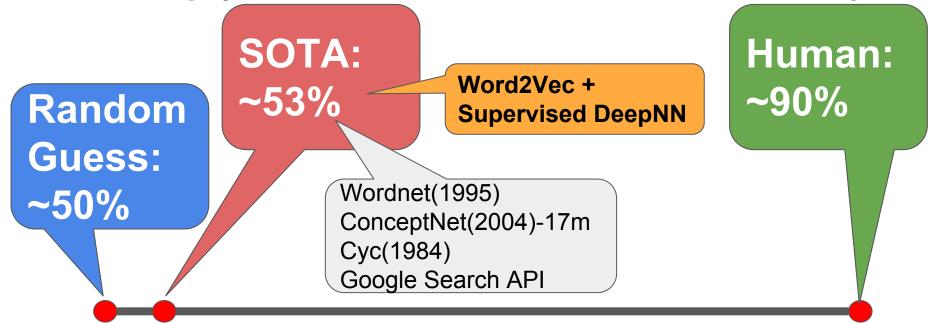
Human: ~90%

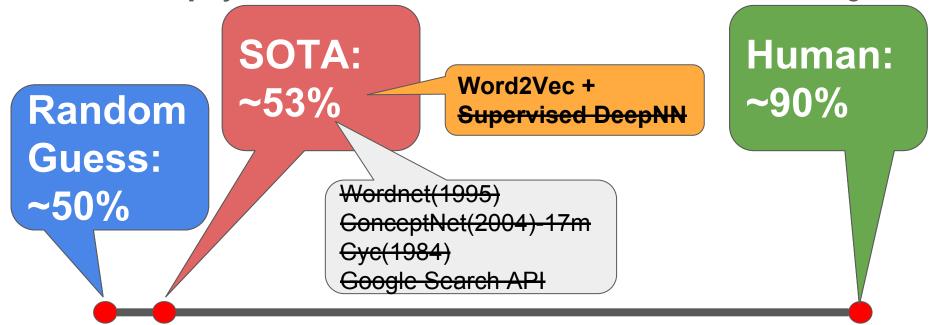
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Human: ~90%







## Winograd Schema Challenge • The trophy cannot fit in the suitcase because it is too big. SOTA: **Human:** Ours LM: ~53% ~90% Random ~64% **Guess:** ~50%

• The **trophy** cannot fit in the **suitcase** because *it* is too big.

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The **trophy** cannot fit in the **suitcase** because **the trophy** is too big .

• The **trophy** cannot fit in the **suitcase** because *it* is too big.

#### More **probable**?

The **trophy** cannot fit in the **suitcase** because **the trophy** is too big .

• The **trophy** cannot fit in the **suitcase** because *it* is too big.

2. Then use *big data*: search all the English text on the web to determine which is the more common pattern:

The trop

- -x does not fit in y + x is so small vs.
- -x does not fit in y + y is so small

The trophy cannot fit in the suitcase because the suitcase is too big .

big .

The treatise the only man alive who still remembers my father as an infant. When Fred first saw my father, he was twelve years old. Who was twelve years old?
 Fred
 my father (Special=years; other=months) is too big.

• The **trophy** cannot fit in the **suitcase** because *it* is too big.

#### More **probable**?

The **trophy** cannot fit in the **suitcase** because **the trophy** is too big .

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P(substitution|Human knowledge)

The **trophy** cannot fit in the **suitcase** because **the trophy** is too big .

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**Language Model** 

 $P(substitution|Human\ knowledge)$ 

The **trophy** cannot fit in the **suitcase** because **the trophy** is too big .

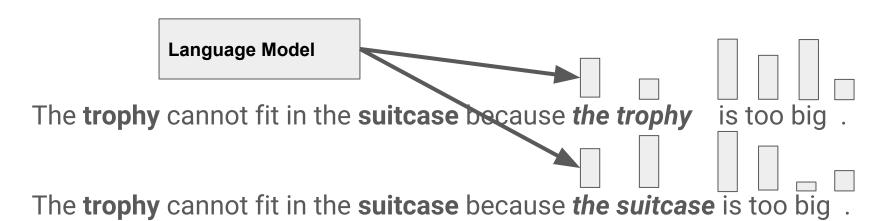
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Language Model 
$$P(substitution | \hat{ heta})$$

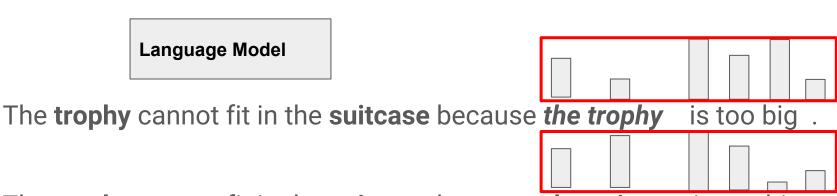
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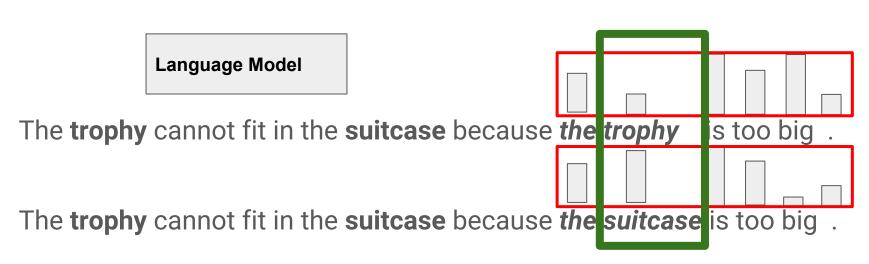
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The trophy cannot fit in the suitcase because it is too big.





• The **trophy** cannot fit in the **suitcase** because *it* is too big.

**Language Model** 

The **trophy** cannot fit in the **suitcase** because **the trophy** 

is too big .

• The trophy cannot fit in the suitcase because it is too big.

**Language Model** 

The **trophy** cannot fit in the **suitcase** because **the trophy** is too big

is too big.

• The **trophy** cannot fit in the **suitcase** because *it* is too big.

1		Wrong prediction	Corrected	Correction percentage	
The	PDP-60	30 55 102	10 33 64	33.3% 60.0% 62.7%	too big .

• The trophy cannot fit in the suitcase because it is too big.

Language Model

The **trophy** cannot fit in the **suitcase** because **the trophy** 

is too big .

The **trophy** cannot fit in the **suitcase** because **the suitcase** is too big .

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**Language Model** 

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The **trophy** cannot fit in the **suitcase** because **the suitcase** is too big .



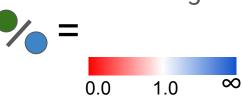
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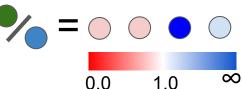
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**Language Model** 

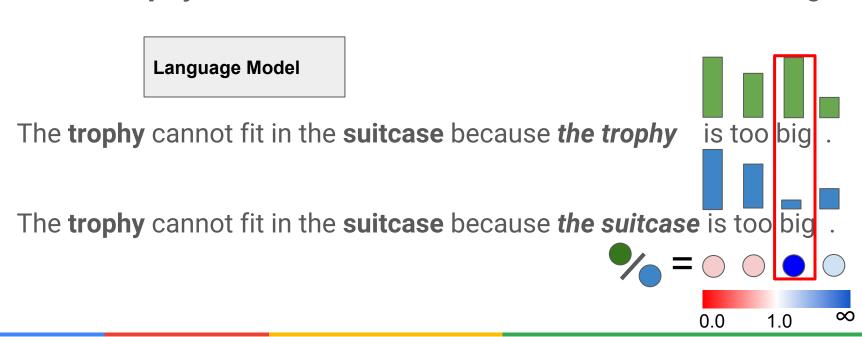
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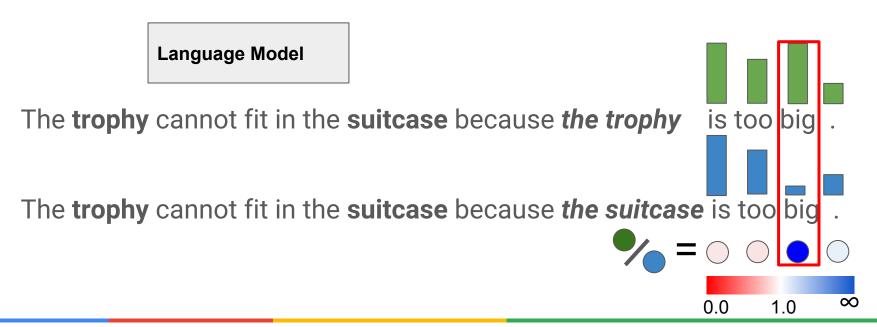
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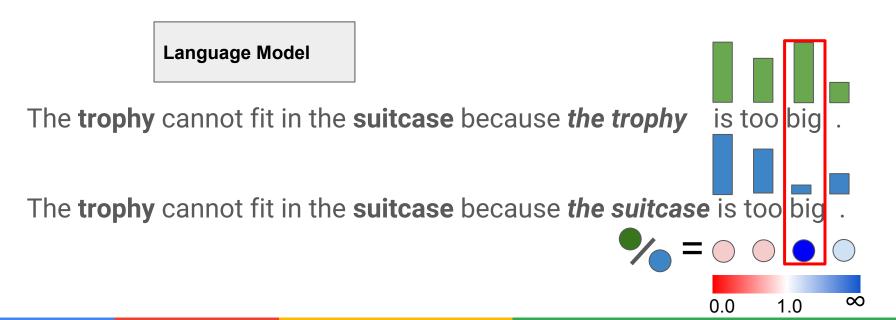
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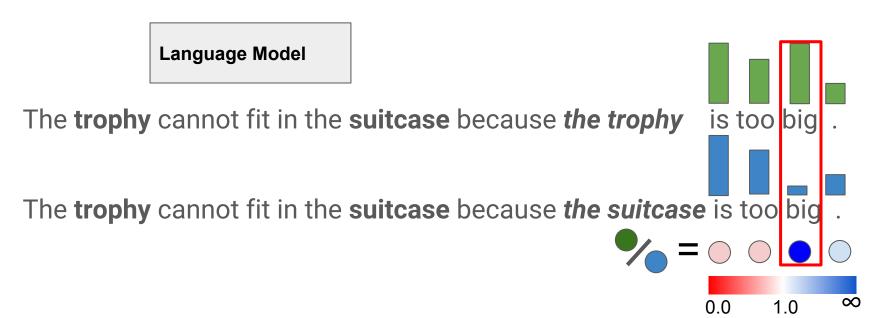
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- The **trophy** cannot fit in the **suitcase** because *it* is too small.
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# Capturing special words

```
Full: Paul tried to call George on the phone , but (Paul/ George*) wasn 't [available] .

Partial: Paul tried to call George on the phone , but (Paul/ George*) wasn 't [available] .
```

# Capturing special words

```
Full: Paul tried to call George on the phone , but (Paul/ George*) wasn 't [available] .

Partial: Paul tried to call George on the phone , but (Paul/ George*) wasn 't [available] .

Full: The drain is clogged with hair . The (drain*/ hair) has to be [cleaned] .

Partial: The drain is clogged with hair . The (drain*/ hair) has to be [cleaned] .

Full: Sam took French classes from Adam , because (Sam*/ Adam) was [eager] to speak it fluently .

Partial: Sam took French classes from Adam , because (Sam*/ Adam) was [eager] to speak it fluently .

Full: Jim [yelled at] Kevin because (Jim*/ Kevin) was so upset .

Partial: Jim [yelled at] Kevin because (Jim*/ Kevin) was so upset .
```

Vocabulary of 800K words, including common names

## Capturing special words

```
Full: Paul tried to call George on the phone , but (Paul/ George*) wasn 't [available] .

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Full: Jim [yelled at] Kevin because (Jim*/ Kevin) was so upset .

Partial: Jim [yelled at] Kevin because (Jim*/ Kevin) was so upset .
```

	Special word retrieved
Forward scoring	97 / 133
Backward scoring	18 / 45

Table 4: Accuracy on Winograd Schema Challenge

Method	Accuracy
Random guess USSM + Knowledge Base USSM + Supervised DeepNet + Knowledge Base	50.0% 52.0 % 52.8 %
Char-LM	51.3%
Word-LM	56.4%
Ensemble of 10 Unsupervised LMs	61.5 %

Table 4: Accuracy on Winograd Schema Challenge

	Method	Accuracy
Gutenberg Books LM-1-Billion CommonCrawl	Random guess USSM + Knowledge Base USSM + Supervised DeepNet + Knowledge Base	50.0% 52.0 % 52.8 %
SQuAD	Char-LM	51.3%
	Word-LM	56.4%
	Ensemble of 10 Unsupervised LMs	61.5 %

#### STORIES = CommonCrawl documents w/ largest overlapping n-grams

Gutenberg Books
LM-1-Billion
CommonCrawl
SQuAD
+ STORIES

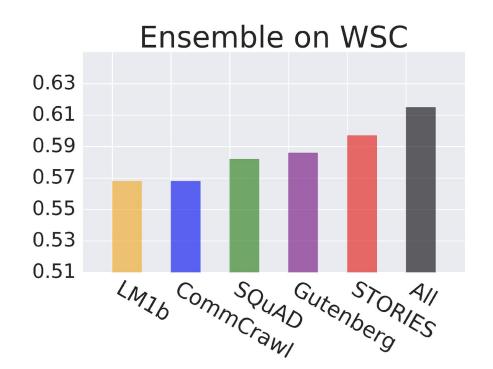
One day when John and I had been out on some - business of our master 's, and were returning gently on a long, straight road, at some distance we saw a boy trying to leap a pony over a gate; the pony would not take the leap, and the boy cut him with the whip, but the only turned off on one side. He whipped him again, but the pony turned off on the other side. Then the boy got off and gave him a hard thrashing, and knocked him about the head ...

#### STORIES = CommonCrawl documents w/ largest overlapping n-grams

Gutenberg Books
LM-1-Billion
CommonCrawl
SQuAD
+ STORIES

Method	Accuracy
USSM + Supervised DeepNet + Knowledge Base	52.8 %
Char-LM-partial	57.9%
Word-LM-partial	62.6%
Ensemble of 14 LMs	63.7 %

# **Diversity of Training Data**



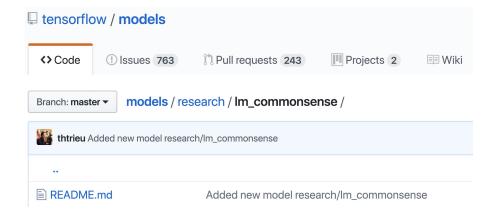
# Preprint and Code for Reproducing

#### A Simple Method for Commonsense Reasoning

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#### Abstract

Commonsense reasoning is a long-standing challenge for deep learning. For example, it is difficult to use neural networks to tackle the Winograd Schema dataset [1]. In this paper, we present a simple method for commonsense reasoning with neural networks, using unsupervised learning. Key to our method is the use of language models, trained on a massive amount of unlabled data, to score multiple choice questions posed by commonsense reasoning tests. On both Pronoun Disambiguation and Winograd Schema challenges, our models outperform previous state-of-the-art methods by a large margin, without using expensive annotated knowledge bases or hand-engineered features. We train an array of large RNN language models that operate at word or character level on LM-1-Billion, CommonCrawl, SQuAD, Gutenberg Books, and a customized corpus for this task and show that diversity of training data plays an important role in test performance. Further analysis also shows that our system successfully discovers important features of the context that decide the correct answer, indicating a good grasp of commonsense knowledge.



# Takeaway

Deep NN can capture Commonsense.

 Commonsense representation does not have to be Graphs or Tuples, but might as well be *vectors*.

# Thank you!

Q&A