



# Commonsense AI: *Myth* and *Truth*

— 🍔 *cheeseburger stabbing redux, 2021 edition* —

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# Commonsense AI: Myth and Truth

— 🍔 *cheeseburger stabbing redux, 2021 edition* —



# Year 2020

## ACL Commonsense Tutorial

— 2nd most popular (among 8 tutorials) —

<https://homes.cs.washington.edu/~msap/acl2020-commonsense/>

5k = # of the main conference registration

1.3k = # of our tutorial registration

3.4k = # of view counts on our recorded lectures



Vered  
Shwartz



Maarten  
Sap



Antoine  
Bosselut



Dan Roth



Yejin  
Choi

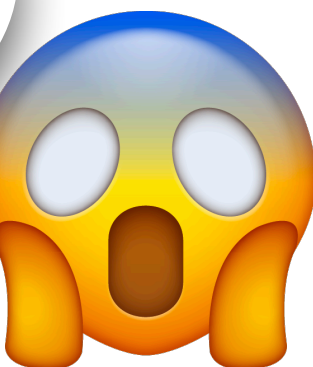


# Circa 2017...

That's a research topic of 70s & 80s



Shouldn't work on it  
since it won't work



Commonsense ?



Don't even say that word



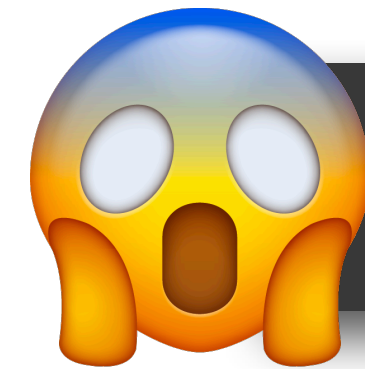
# 2017 — 2020

Truth or Myth ?



## What is commonsense?

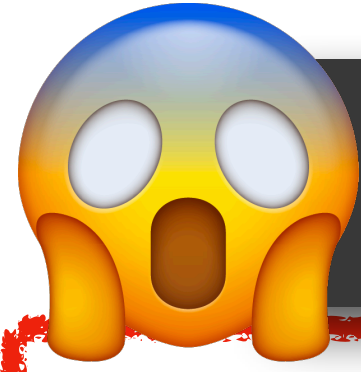
- It's what everyone knows and agrees on?
- language is irrelevant to commonsense?
- Too hard to define precisely, thus shouldn't work on it



## Should we or should we not?

- Commonsense AI is an impossible goal (ever)
- That's a research topic of 70s and 80s
- Maybe only possible in the faraway future

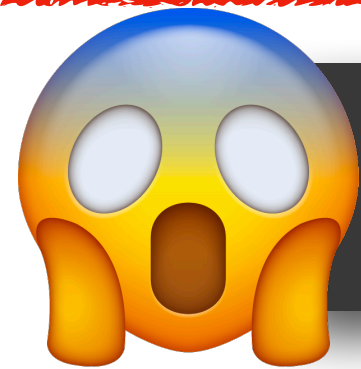




## Commonly held beliefs

- Knowledge and reasoning are distinct and exclusive
- Language is in the way of reasoning; let's do formal logics
- Language is not symbols. Words and numbers are, but not language at large
- Humans acquire commonsense completely un-/self-supervised, thus so should machines

Truth or Myth ?



## What is commonsense?

- It's what everyone knows and agrees on?
- language is irrelevant to commonsense?
- Too hard to define precisely, thus shouldn't work on it



## Should we or should we not?

- Commonsense AI is an impossible goal (ever)
- That's a research topic of 70s and 80s
- Maybe only possible in the faraway future



# The Curious Case of Cheeseburger Stabbing

— *An example repeatedly appeared my talks between Mar 2017 and May 2018* —



# The Curious Case of 🍔 “Cheeseburger Stabbing”

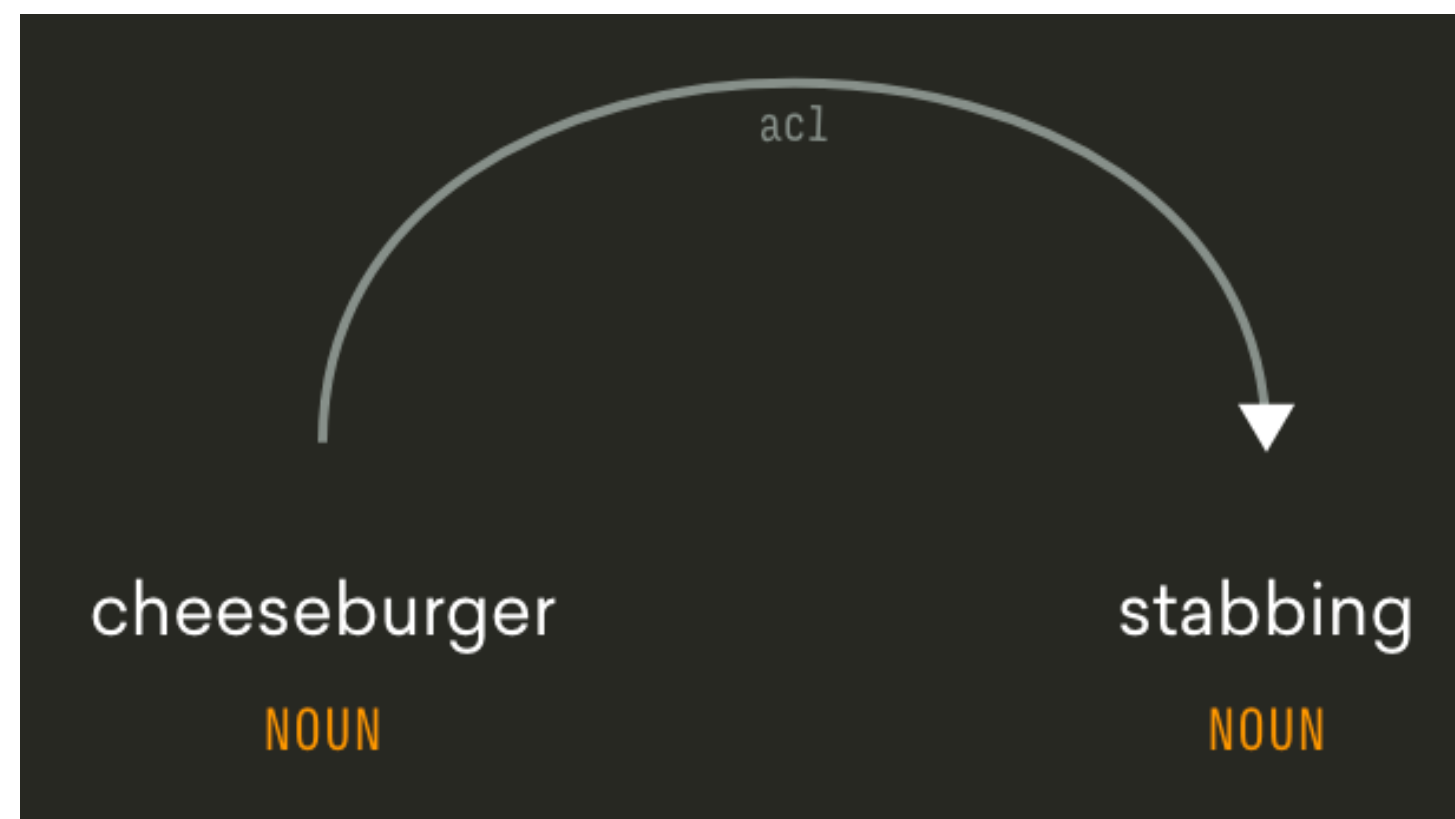


Someone stabbed a cheeseburger?

A cheeseburger stabbed someone?

A cheeseburger stabbed another cheeseburger?

Someone stabbed someone else over a cheeseburger?



# The Curious Case of 🍔 “Cheeseburger Stabbing”

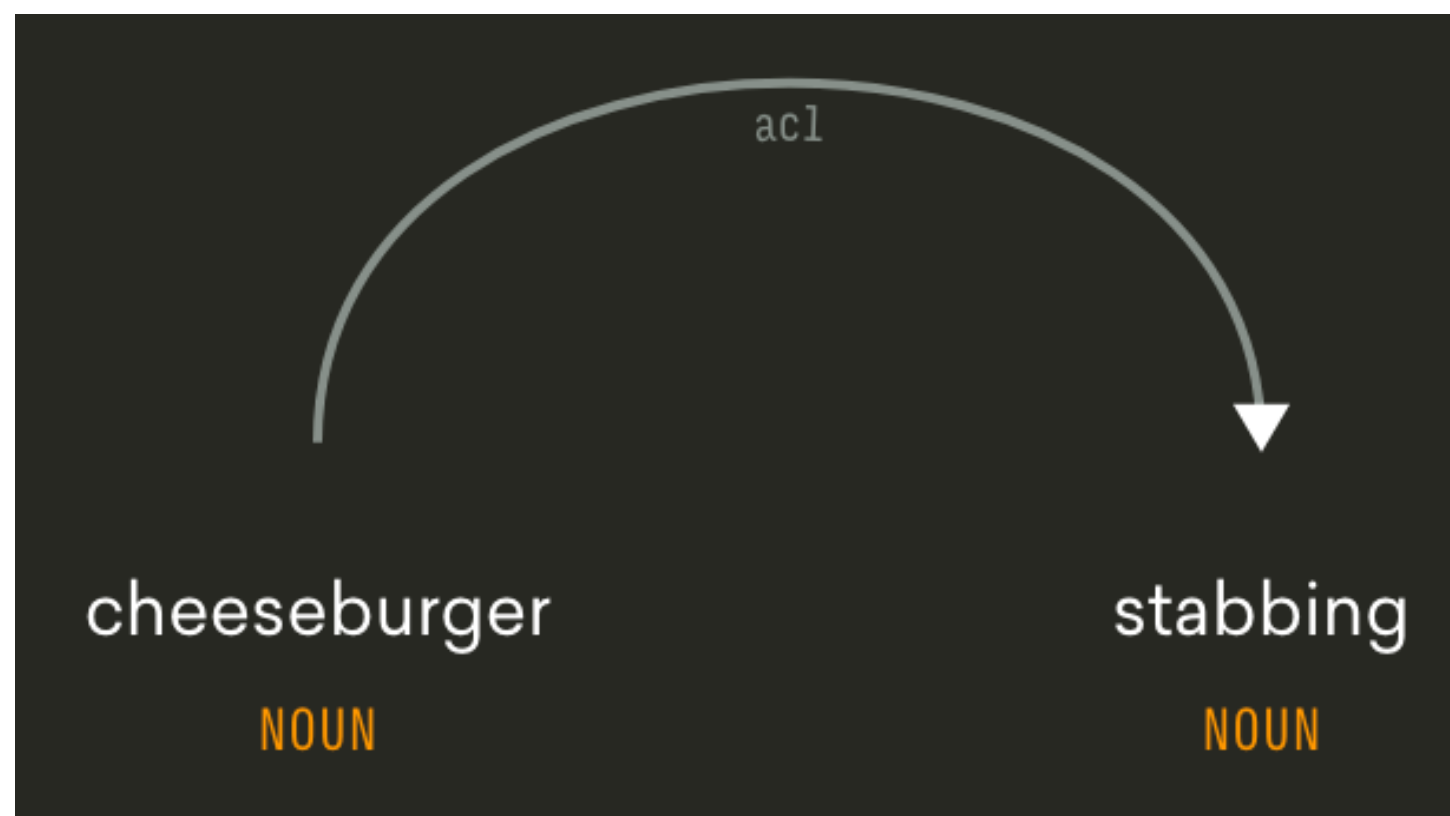


**Physical Commonsense:** not possible to stab using a burger

**Social Commonsense:** stabbing someone is bad

what is said +

what is not said



Someone stabbed a cheeseburger?

A cheeseburger stabbed someone?

A cheeseburger stabbed another cheeseburger?

Someone stabbed someone else over a cheeseburger?



# The Curious Case of 🍔 "Cheeseburger Stabbing"



**Vered Schwartz**  
@VeredShwartz

Replying to @shacharmirkin and @AllThingsLing

Thanks! I first heard it last year in @YejinChoinka 's talk, and I used it as a difficult example to test my noun

generators. I was surprised by the results. As a result, I decided to

cheeseburger" 😂

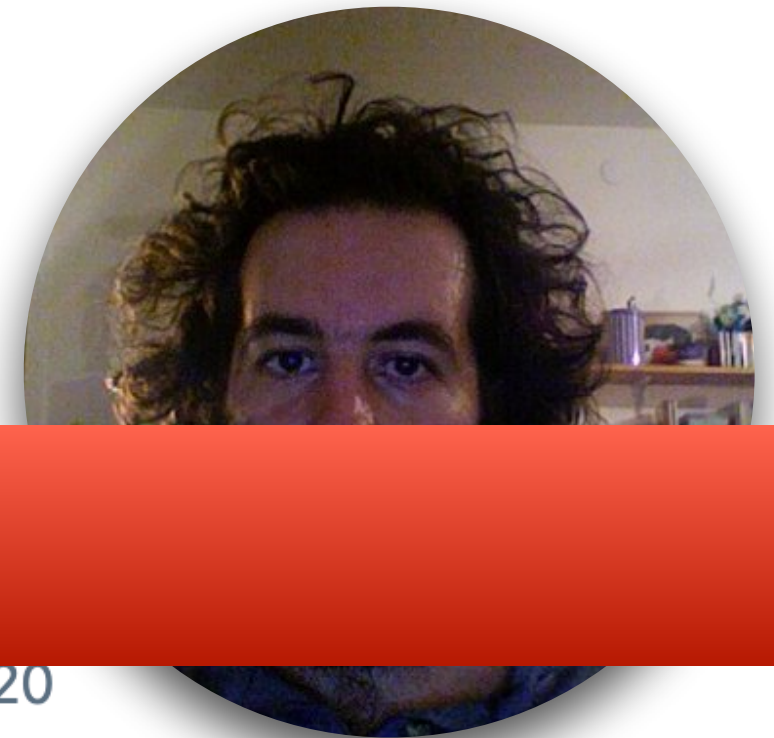
12:10 AM · Aug 31, 2018 · Twitter for Android

**2018:**

**"Stabbing is a crime punishable by  
cheeseburger" 😂**

**2020:**

**"stabbing of a cheeseburger"  
per GPT-3**



(((yoav' ())) @yoavgo · Jul 18, 2020

Q: parsley cake

A: cake made of parsley

Q: cheeseburger stabbing

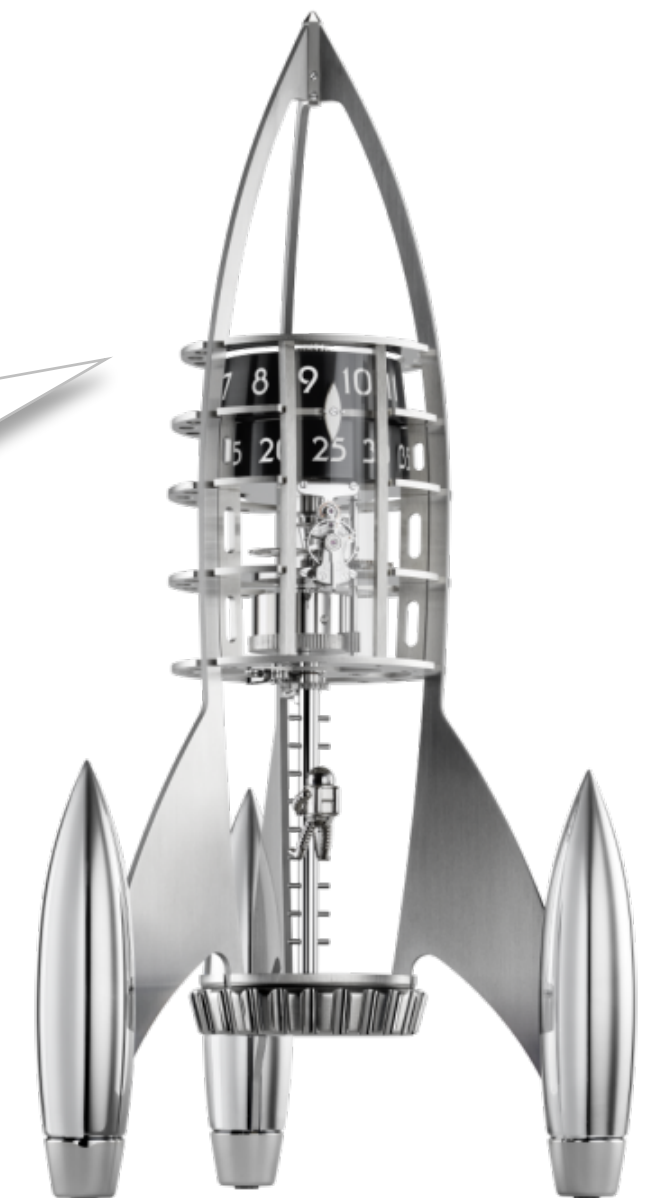
A: stabbing of a cheeseburger

# Path to commonsense?

Brute force larger networks with deeper layers?



You don't reach to the moon  
by making the tallest building in the world  
taller





# Path to commonsense?



Obligatory



Controversial Remarks of the Day



of the Day

Neural

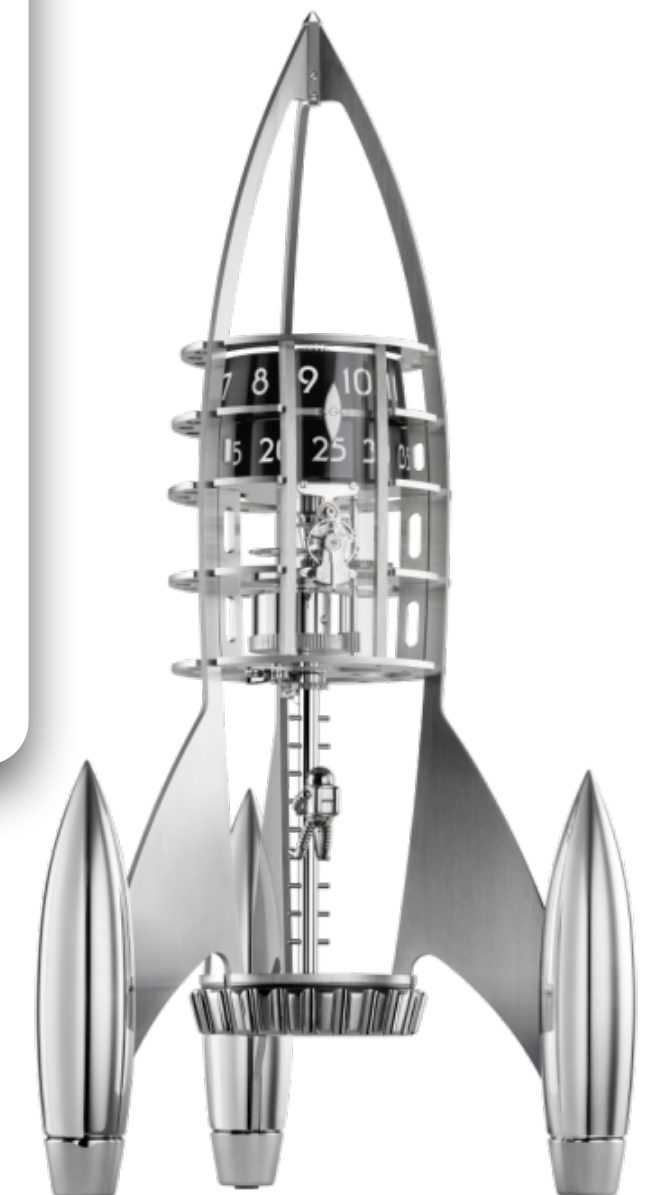
Symbolic

Language

Knowledge

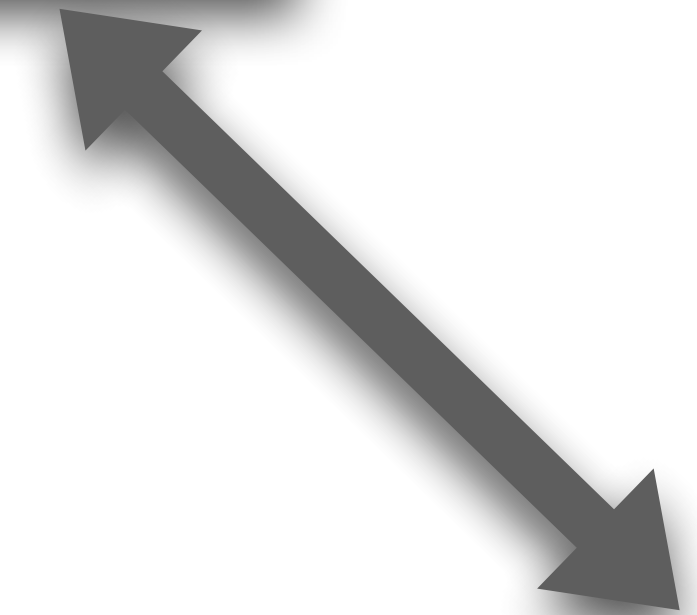
Reasoning

1. the continuum between knowledge and reasoning
2. the interplay between reasoning and language generation
3. the blend between neural vs symbolic representation





Language



Reasoning

Induction — from specific to general

Deduction — from general to specific

**Abduction** — “*why* something happened”  
(or reasoning about the the probable explanation)



Peirce 1965

**Counterfactual** — “*what if* something else happened”





Back to the Future:

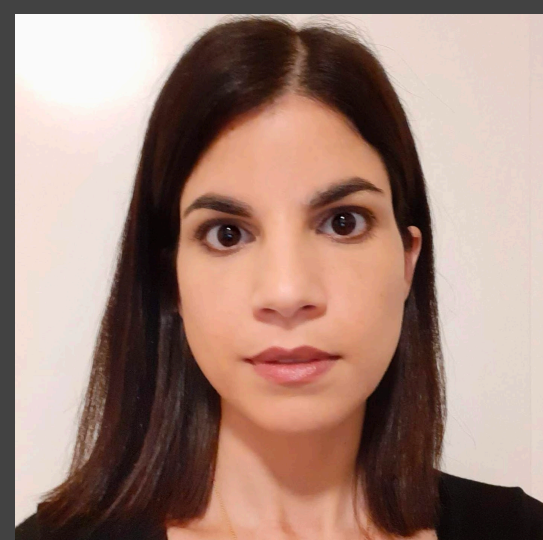
# Unsupervised Backprop-based Decoding for Counterfactual and Abductive Commonsense Reasoning

EMNLP 2020

Lianhui Qin



Vered  
Shwartz



Peter  
West



Chandra  
Bhagavatula



Jena  
Hwang



Ronan  
LeBras



Antoine  
Bosselut



Me



# Abductive Reasoning

(Bhagavatula et al., 2019)



## *Past Observation*

Ray hung a tire on a rope to make his daughter a swing.

# What happened in between?



## *Future Observation*

Ray ran to his daughter to make sure she was okay.



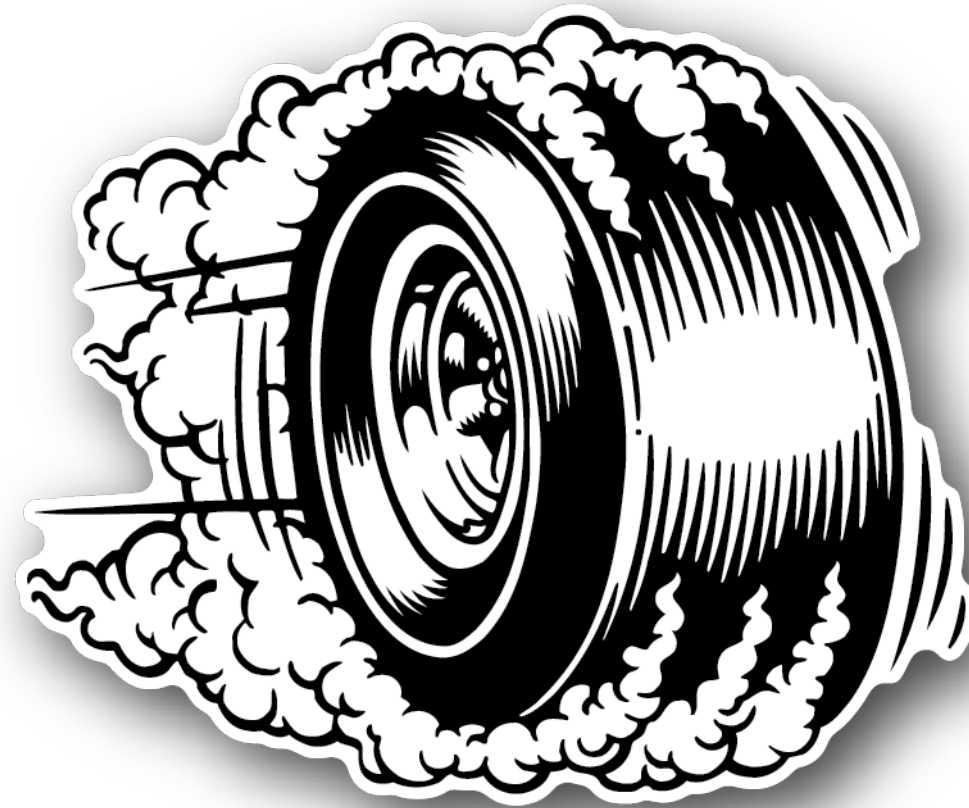
# Abductive Reasoning

(Bhagavatula et al., 2019)



*Past Observation*

Ray hung a tire on a rope to make his daughter a swing.



*Hypothesis*

*She hit the rope and the tire fell on top of her.*



*Future Observation*

Ray ran to his daughter to make sure she was okay.

**Abduction** — “*why* something happened”

(or reasoning about the the probable explanation)



Peirce 1965

# Abductive Reasoning

(Bhagavatula et al., 2019)

## *Past Observation*

Ray hung a tire on a rope to make his daughter a swing.

## *Hypothesis*

*She hit the rope and the tire fell on top of her.*

## *Future Observation*

Ray ran to his daughter to make sure she was okay.

# Counterfactual Reasoning

(Qin et al., 2019)





An example from the  
“TimeTravel” dataset  
(Qin et al., EMNLP 2019)

Zeke was throwing a party.

All his friends were dressing up for this Halloween party. Story context changes...

Zeke thought about being a vampire or a wizard.

Then he decided on a scarier costume.

Zeke dressed up like a skeleton.

What if this is a **Game of Thrones themed party** instead of a **Halloween party**? 🤔





An example from the  
“TimeTravel” dataset  
(Qin et al., EMNLP 2019)

Story ending doesn't  
make sense now...

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What if this is a **Game of Thrones themed party** instead of a **Halloween party**?



## Counterfactual Reasoning

(Qin et al., 2019)

Reasoning about the alternative **future**  
based on counterfactual **past**.



An example from the  
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(Qin et al., EMNLP 2019)

Story ending doesn't  
make sense now...

Zeke was throwing a party.

All his friends were dressing up for this Halloween party.

Story context changes...

Zeke thought about ~~being a vampire or a wizard.~~

~~Then he decided on a scarier costume.~~

Zeke dressed up like a ~~skeleton.~~

What if this is a **Game of Thrones themed party** instead of a **Halloween party**?



Zeke was throwing a party.

All his friends were dressing up for this Halloween party.

Only do minimal edit!

Zeke thought about *Lannister, but he didn't want to look like a Lannister.*

Consistent  
now!!

*He wanted to look like a Stark.*

Zeke dressed up like a *Stark.*

# Abductive Reasoning

(Bhagavatula et al., 2019)

## Past Observation

Ray hung a tire on a rope to make his daughter a swing.

## Hypothesis

*She hit the rope and the tire fell on top of her.*

## Future Observation

Ray ran to his daughter to make sure she was okay.

## Story Context

Zeke was throwing a party.

All his friends were dressing up for this Halloween party.

All his friends were dressing up for this Game of Thrones themed party.

## Rewritten Ending

*Zeke thought about Lannister, but he didn't want to look like a Lannister.*

*He wanted to look like a Stark.*

*Zeke dressed up like a Stark.*

## Original Ending

Zeke thought about being a vampire or a wizard.

Then he decided on a scarier costume.

Zeke dressed up like a skeleton.

# Counterfactual Reasoning

(Qin et al., 2019)



# Abductive Reasoning

(Bhagavatula et al., 2019)

Both involve ***nonmonotonic reasoning*** with  
past context  $X$  and future constraint  $Z$



# Counterfactual Reasoning

(Qin et al., 2019)

Pretrained Language Models are successful on many tasks...

How are Pretrained LMs on  
the Nonmonotonic Reasoning?

Let's first see the abductive case...

*Y*

The little girl liked it and was thrilled at it.



Pre-trained GPT2

*X*

Ray hung a tire on a rope to make his daughter a swing.

Not able to do right to left!

*Z*

Ray ran to his daughter to make sure she was okay.



Why not just concatenate both direction?

$Y$

But the swing didn't go off, so they moved down the slope towards the



Doesn't make sense!

Pre-trained GPT2

$Z[s]X$

Ray ran to his daughter to make sure she was okay.

Ray hung a tire on a rope to make his daughter a swing.

Try again?

$Y$

As the swing moved, the girl's cries sounded in his ears.



Doesn't make sense!

Pre-trained GPT2

$Z[s]X$

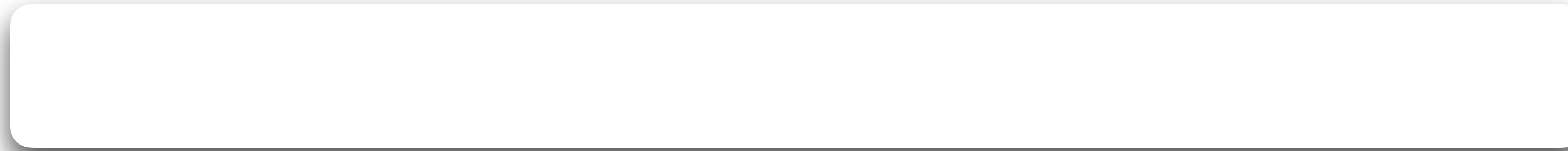
Ray ran to his daughter to make sure she was okay.

Ray hung a tire on a rope to make his daughter a swing.

Something might have been missing here...



$Y$



Backpropagation!!

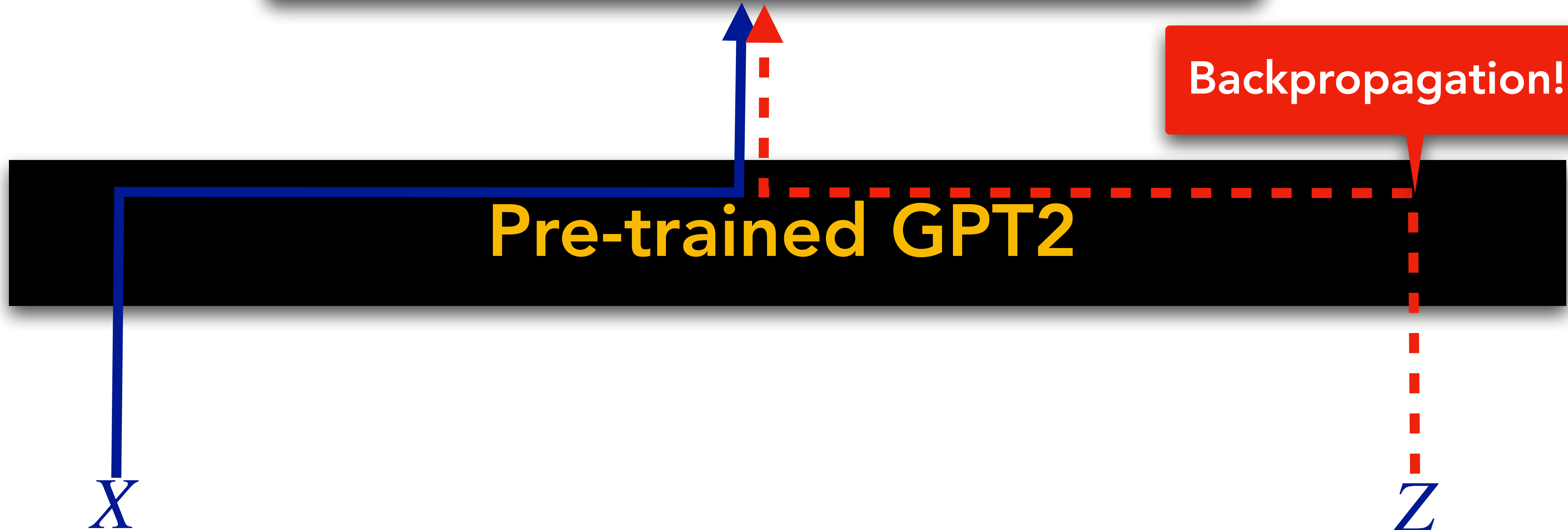
Pre-trained GPT2

$X$

Ray hung a tire on a rope to make his daughter a swing.

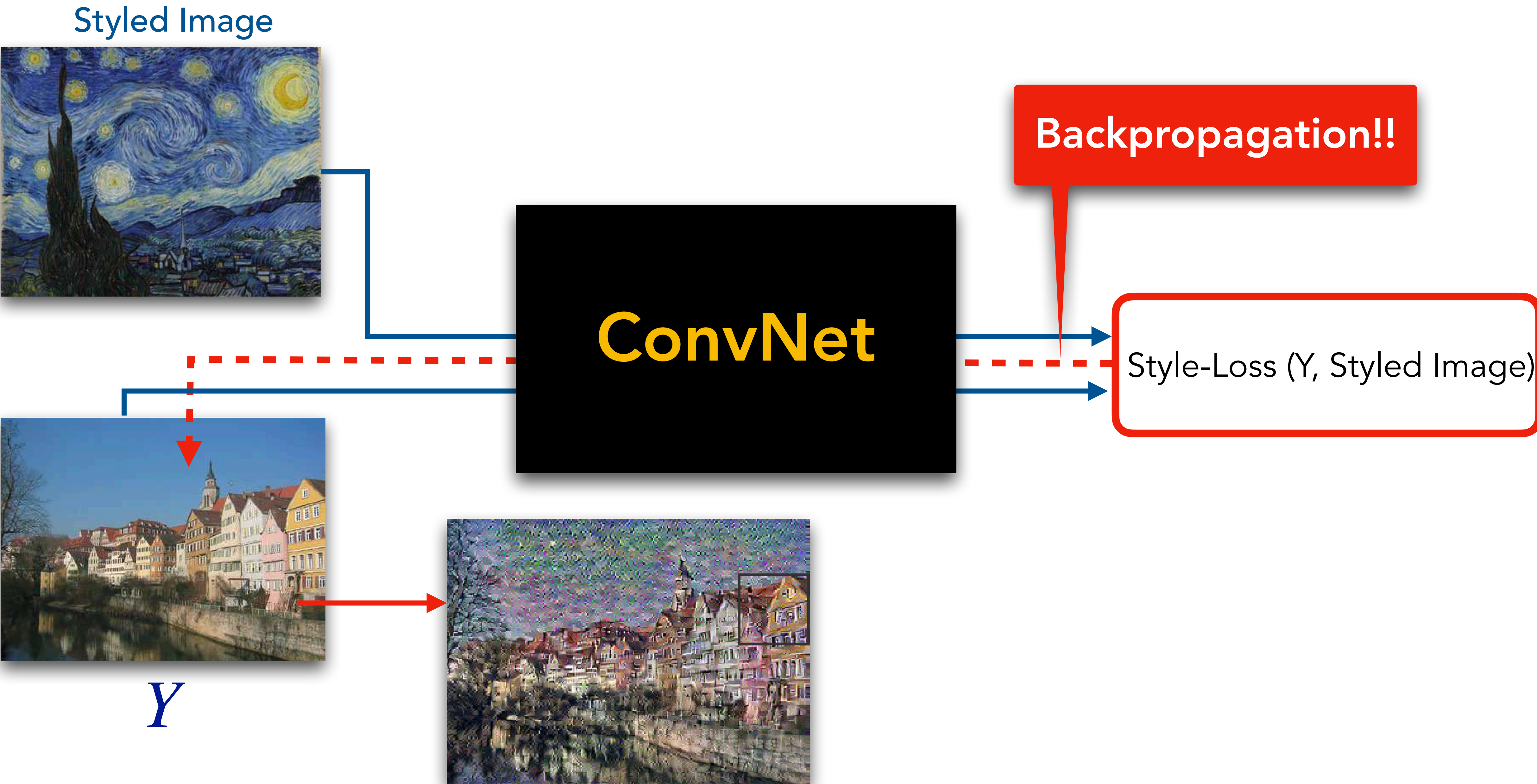
$Z$

Ray ran to his daughter to make sure she was okay.

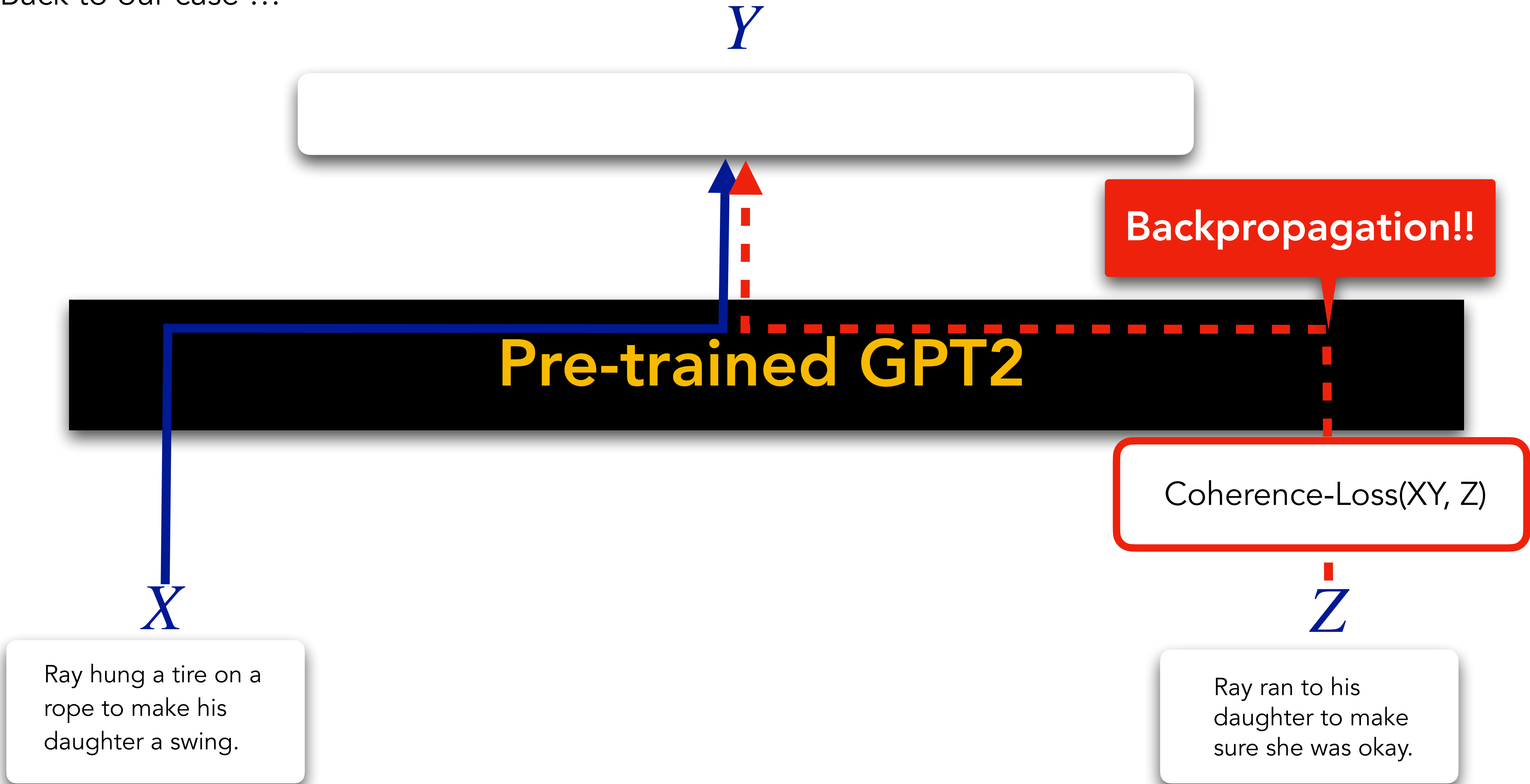




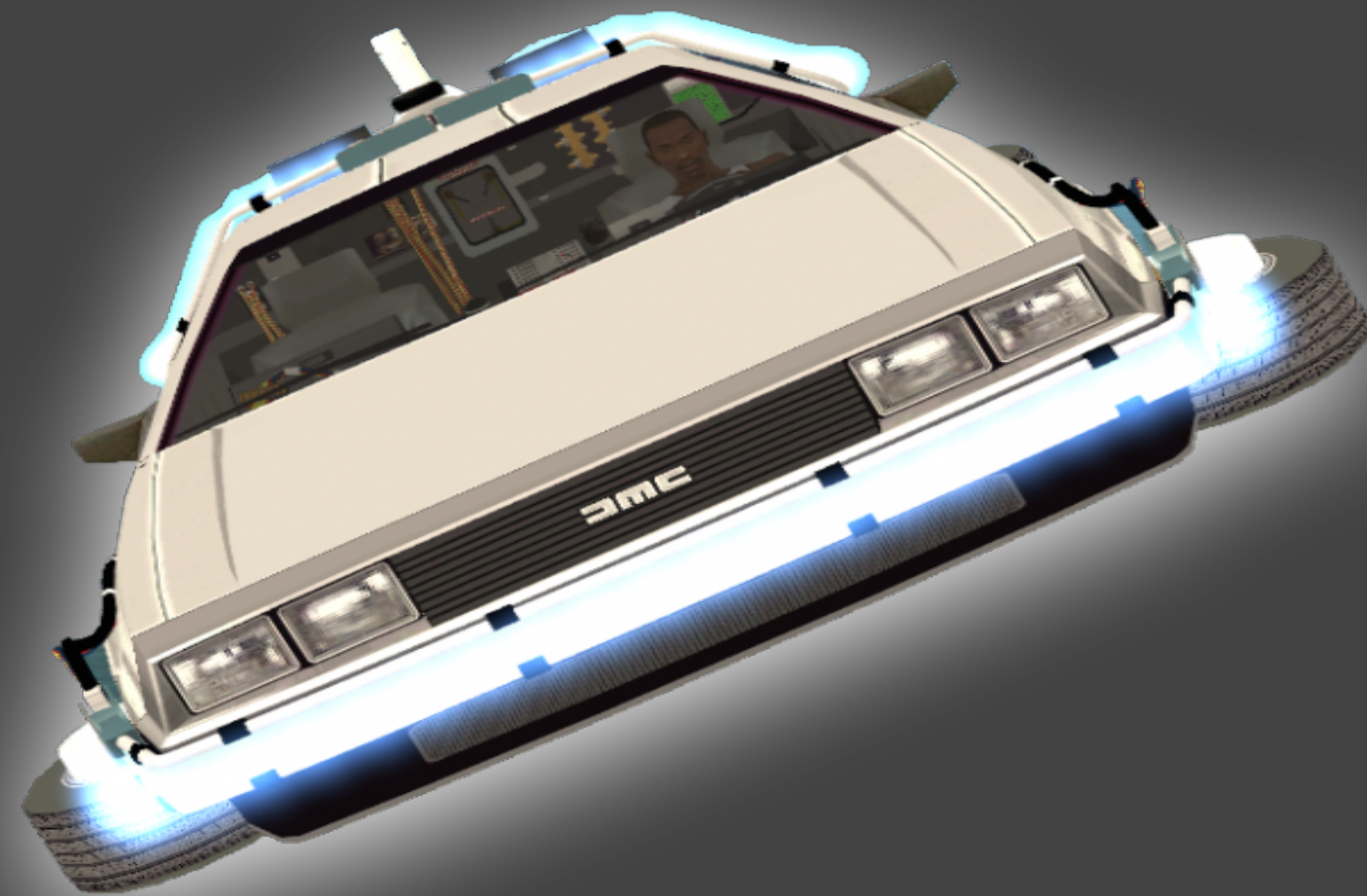
Inspired by “Image Style Transfer” (Gatys et al, 2016)...



Back to our case ...







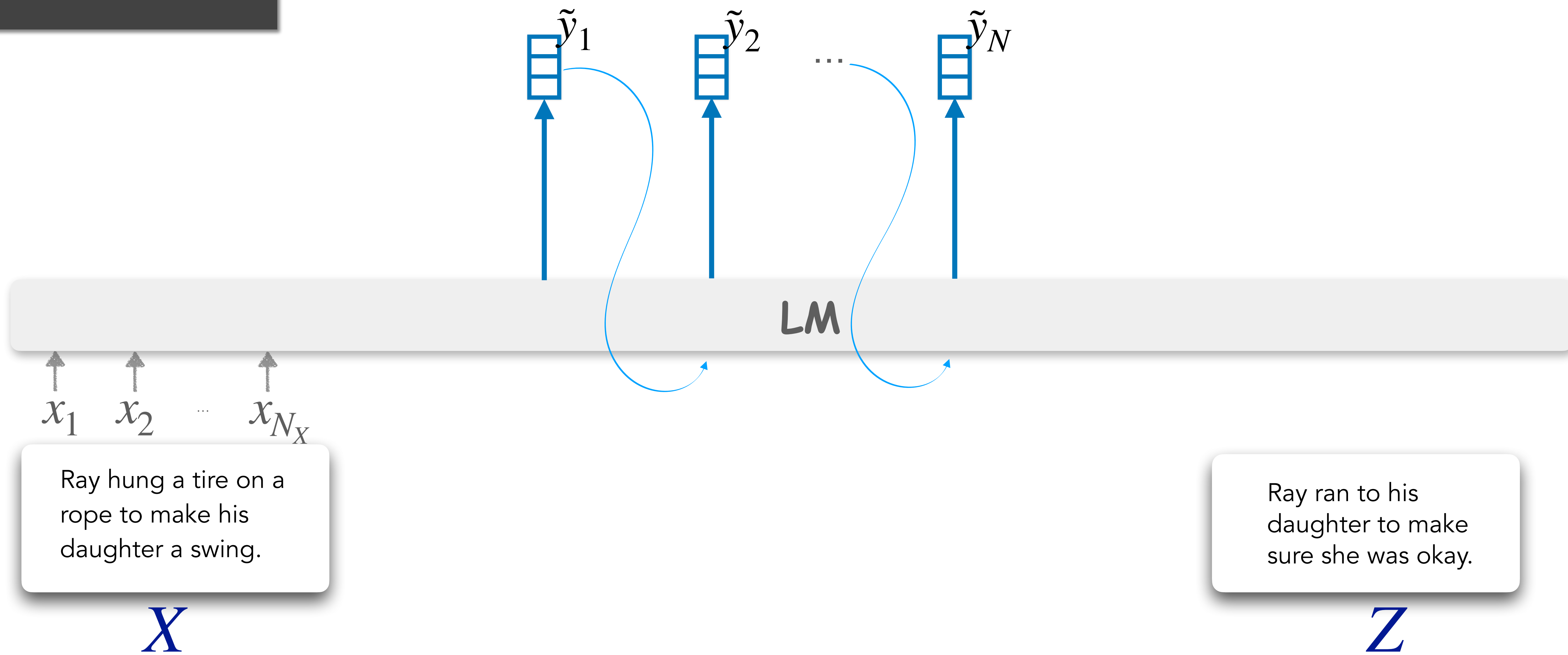
# DELOREAN

(DEcoding for nonmonotonic LOgical REAsoNing)



## Initialization

Just as how you do  
regular decoding



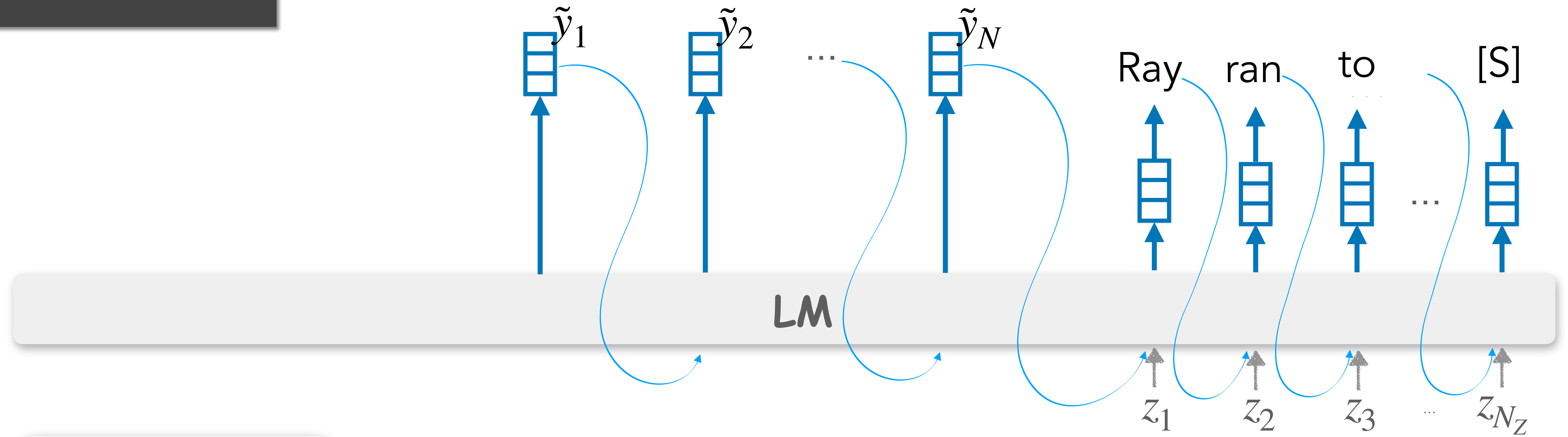




# Backward Pass

Backpropagate  
future information

$Y$



Ray hung a tire on a rope to make his daughter a swing.

$X$

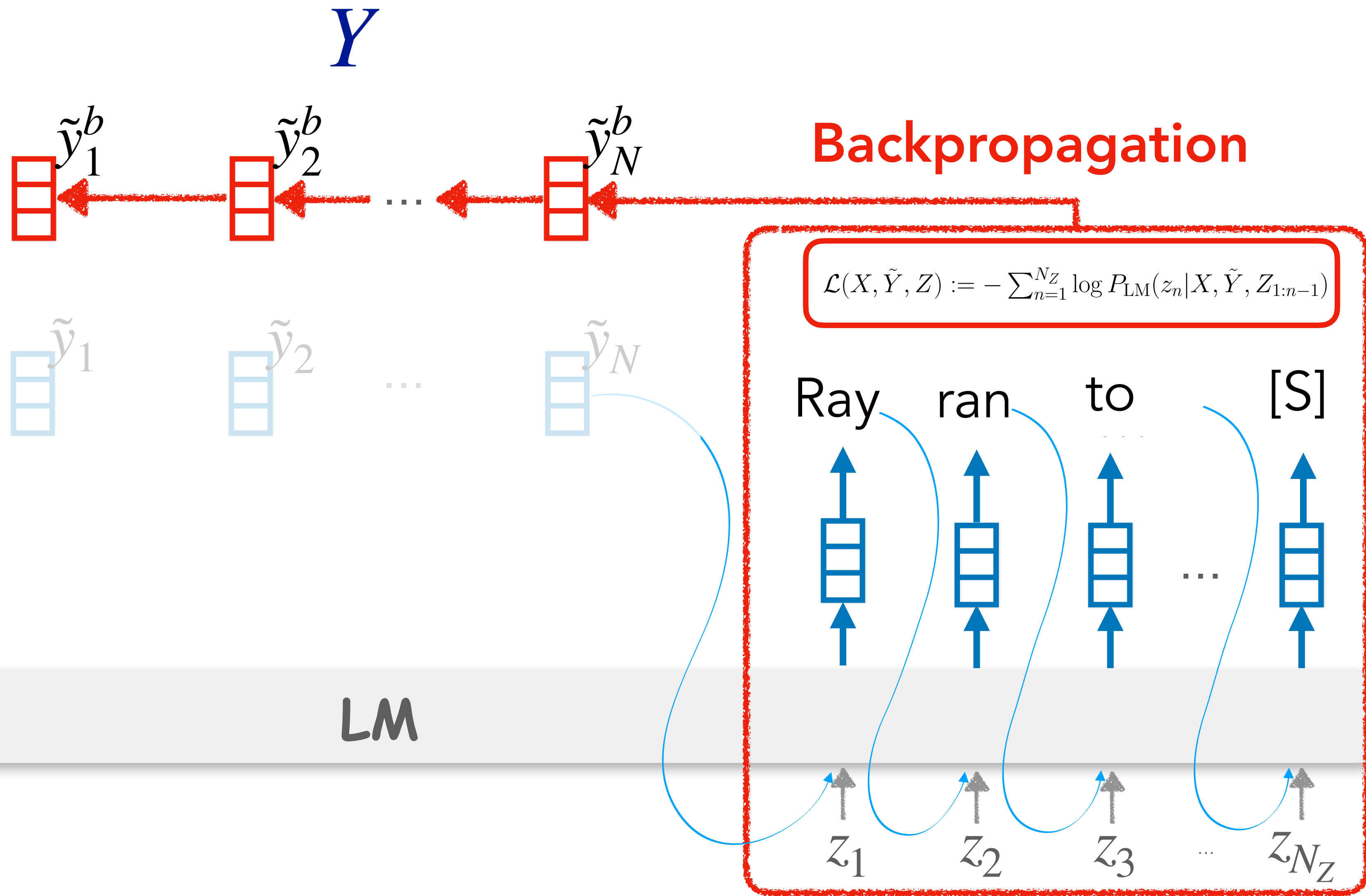
Ray ran to his daughter to make sure she was okay.

$Z$



# Backward Pass

Backpropagate  
future information  
 $\text{Loss}(Z | X, Y)$



Ray hung a tire on a  
rope to make his  
daughter a swing.

$X$

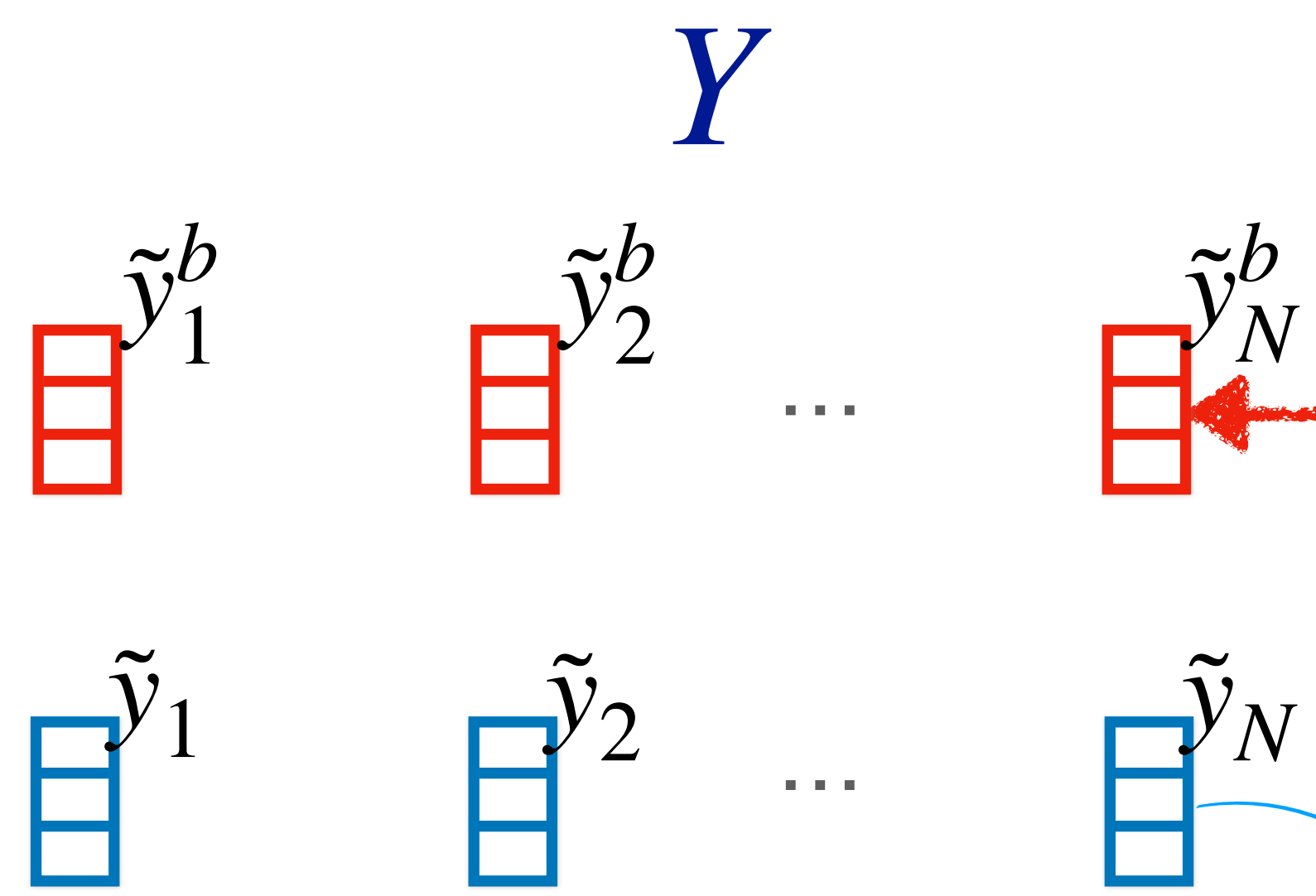
Ray ran to his  
daughter to make  
sure she was okay.

$Z$



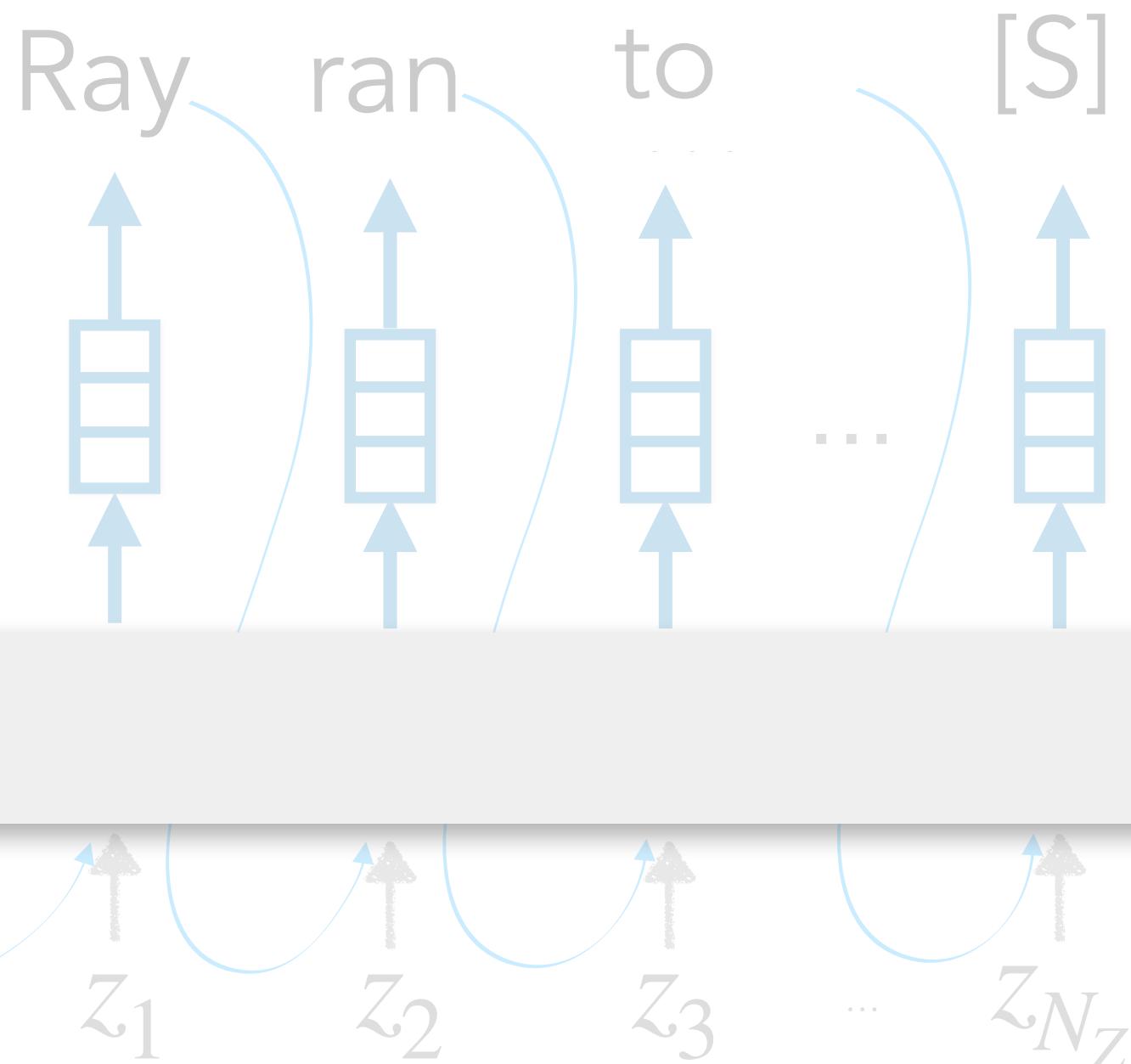
# Forward Pass

Mix both past and future information



# Backpropagation

$$\mathcal{L}(X, \tilde{Y}, Z) := - \sum_{n=1}^{N_Z} \log P_{\text{LM}}(z_n | X, \tilde{Y}, Z_{1:n-1})$$



Ray hung a tire on a rope to make his daughter a swing.

$X$

Ray ran to his daughter to make sure she was okay.

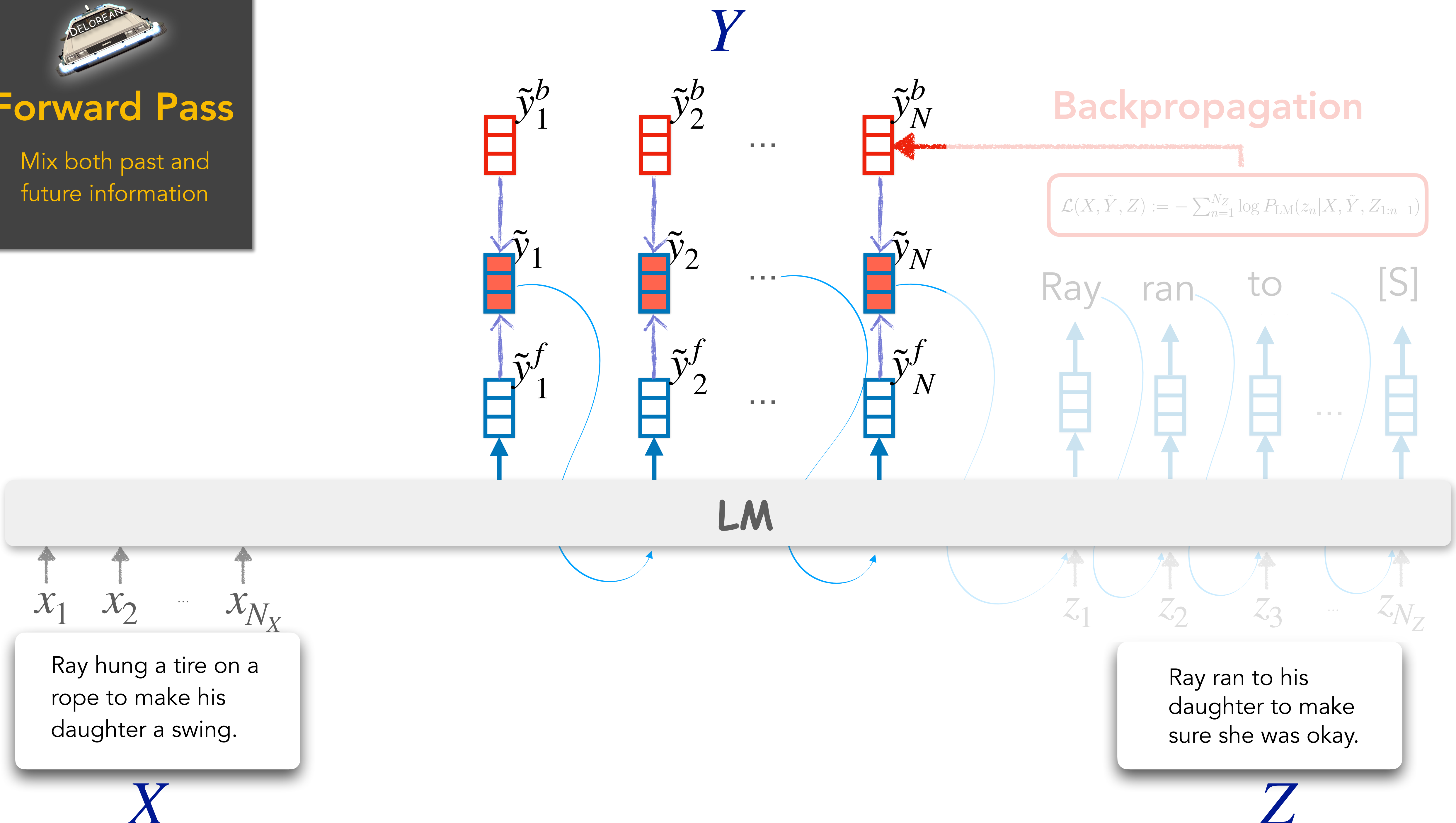
$Z$





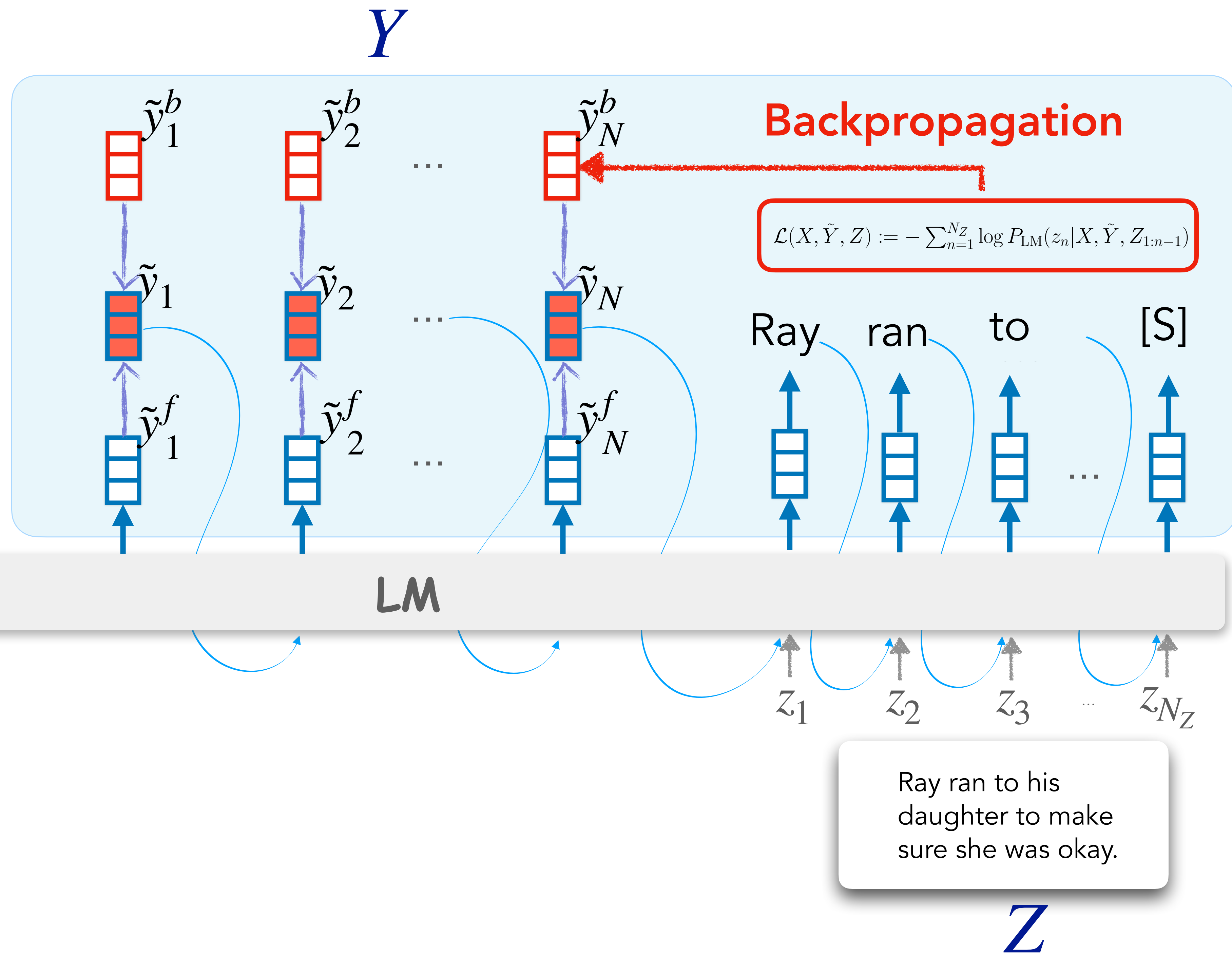
# Forward Pass

Mix both past and future information





Repeat  
 $T$  times





Sampling

$Y$

Output: She hit the rope and the tire fell on top of her.

Backpropagation

$$\mathcal{L}(X, \tilde{Y}, Z) := - \sum_{n=1}^{N_Z} \log P_{\text{LM}}(z_n | X, \tilde{Y}, Z_{1:n-1})$$

$\tilde{y}_1^b$

$\tilde{y}_2^b$

...

$\tilde{y}_N^b$

$\tilde{y}_1$

$\tilde{y}_2$

...

$\tilde{y}_N$

$\tilde{y}_1^f$

$\tilde{y}_2^f$

...

$\tilde{y}_N^f$

Ray

ran

to

[S]

LM

$x_1$   $x_2$  ...  $x_{N_X}$

Ray hung a tire on a rope to make his daughter a swing.

$X$

$z_1$   $z_2$   $z_3$  ...  $z_{N_Z}$

Ray ran to his daughter to make sure she was okay.

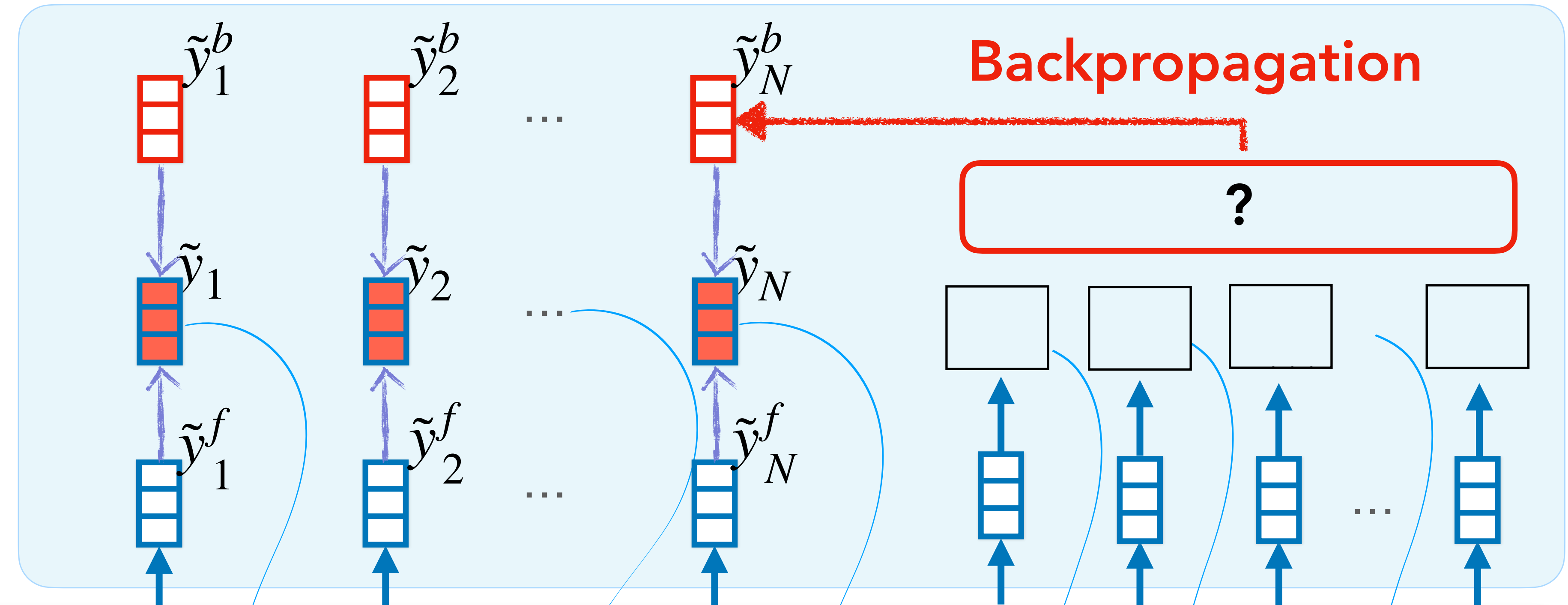
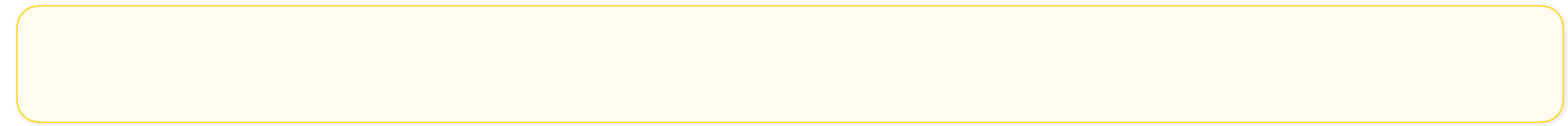
$Z$





# Counterfactual Reasoning?

$Y$



$x_1 \quad x_2 \quad \dots \quad x_{N_X}$

$z_1 \quad z_2 \quad z_3 \quad \dots \quad z_{N_Z}$

$X$

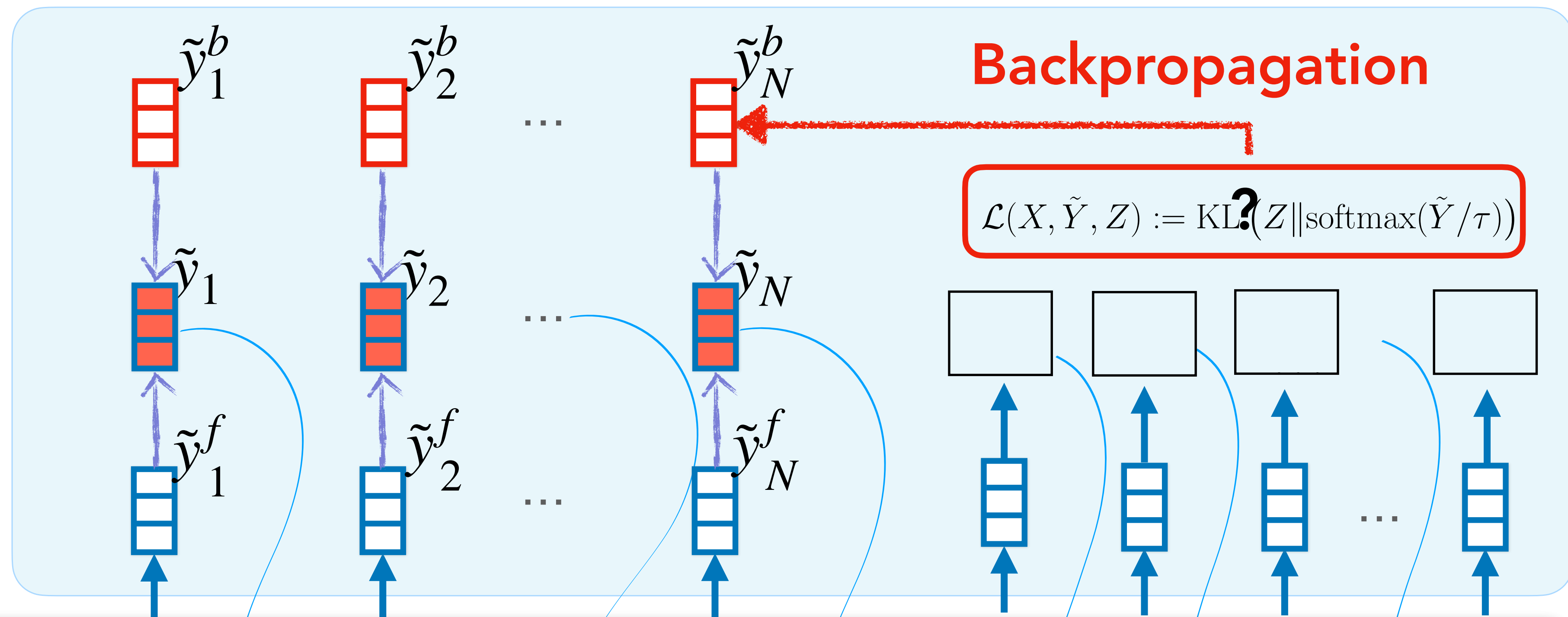
$Z$



# Counterfactual Reasoning?

Distance-Loss:  $(Y, Z)$

$Y$



LM

$x_1$   $x_2$  ...  $x_{N_X}$

$z_1$   $z_2$   $z_3$  ...  $z_{N_Z}$

$X$

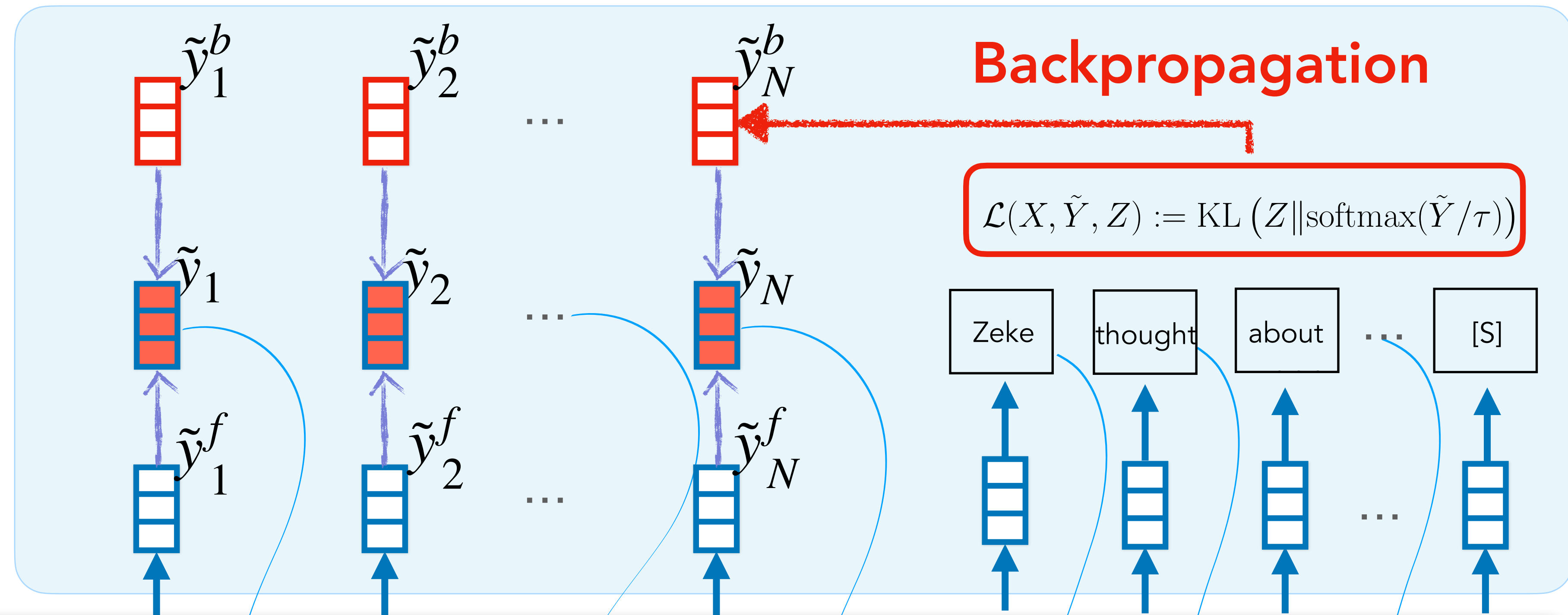
$Z$



# Counterfactual Reasoning?

Distance-Loss:  $(Y, Z)$

$Y$



$x_1$   $x_2$  ...  $x_{N_X}$

Zeke was throwing a party.  
[Counterfactual] All his friends  
were dressing up for this Game  
of Thrones themed party.

$X$

$z_1$   $z_2$   $z_3$  ...  $z_{N_Z}$

Zeke thought about being  
a vampire or a wizard. Then  
he decided on a scarier  
costume. Zeke dressed up  
like a skeleton.

$Z$





# Counterfactual Reasoning?

Distance-Loss:  $(Y, Z)$

$Y$

Zeke thought about **Lannister**, but he didn't want to look like a Lannister. He wanted to look like a Stark. Zeke dressed up like a **Stark**.

Backpropagation

$$\mathcal{L}(X, \tilde{Y}, Z) := \text{KL} (Z \| \text{softmax}(\tilde{Y} / \tau))$$

LM

$x_1$   $x_2$  ...  $x_{N_X}$

Zeke was throwing a party.  
[Counterfactual] All his friends were dressing up for this Game of Thrones themed party.

$X$

$z_1$   $z_2$   $z_3$  ...  $z_{N_Z}$

Zeke thought about being a vampire or a wizard. Then he decided on a scarier costume. Zeke dressed up like a skeleton.

$Z$

$\tilde{y}_1^b$

$\tilde{y}_2^b$

$\tilde{y}_N^b$

$\tilde{y}_1$

$\tilde{y}_2$

$\tilde{y}_N$

$\tilde{y}_1^f$

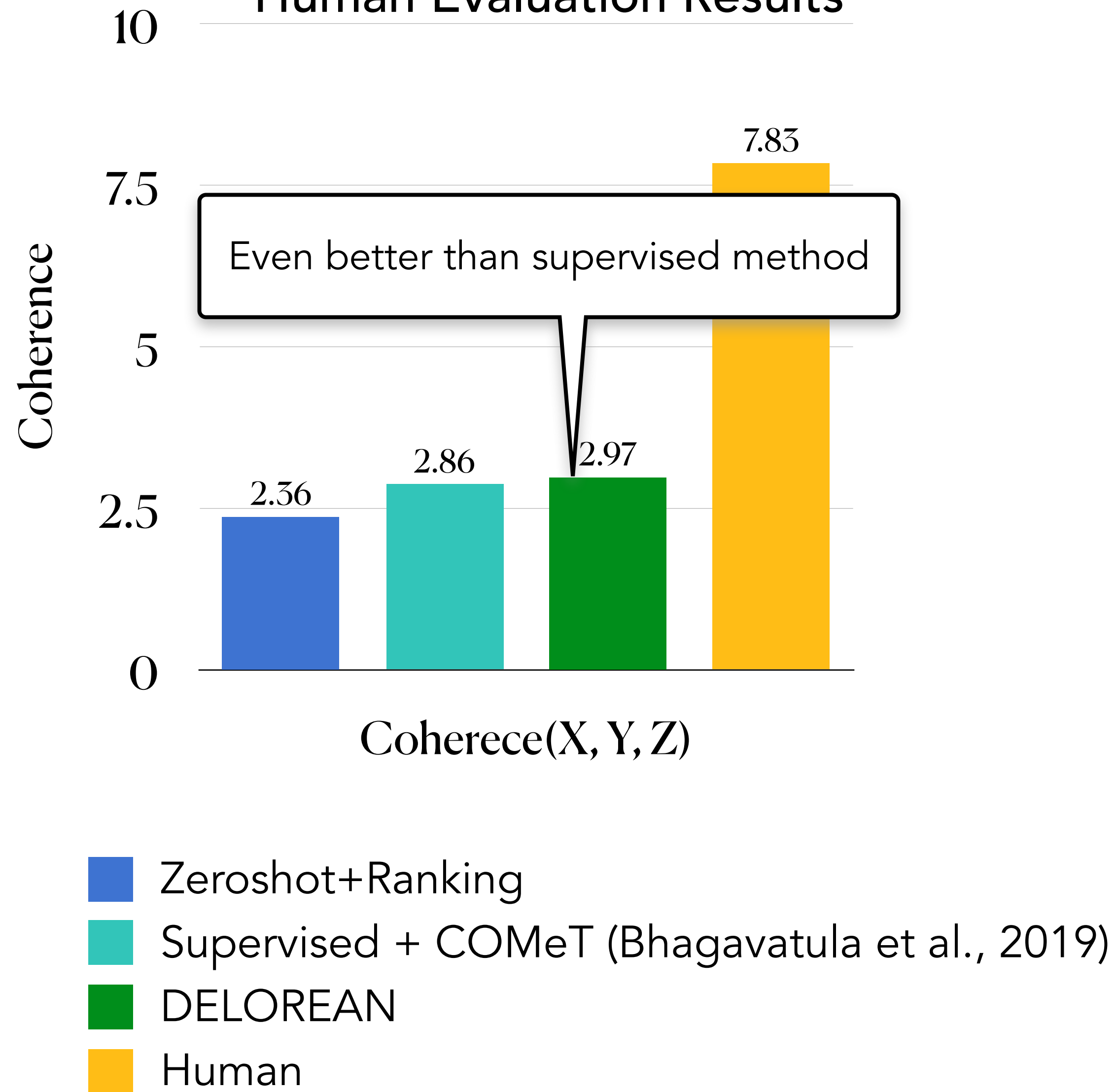
$\tilde{y}_2^f$

$\tilde{y}_N^f$

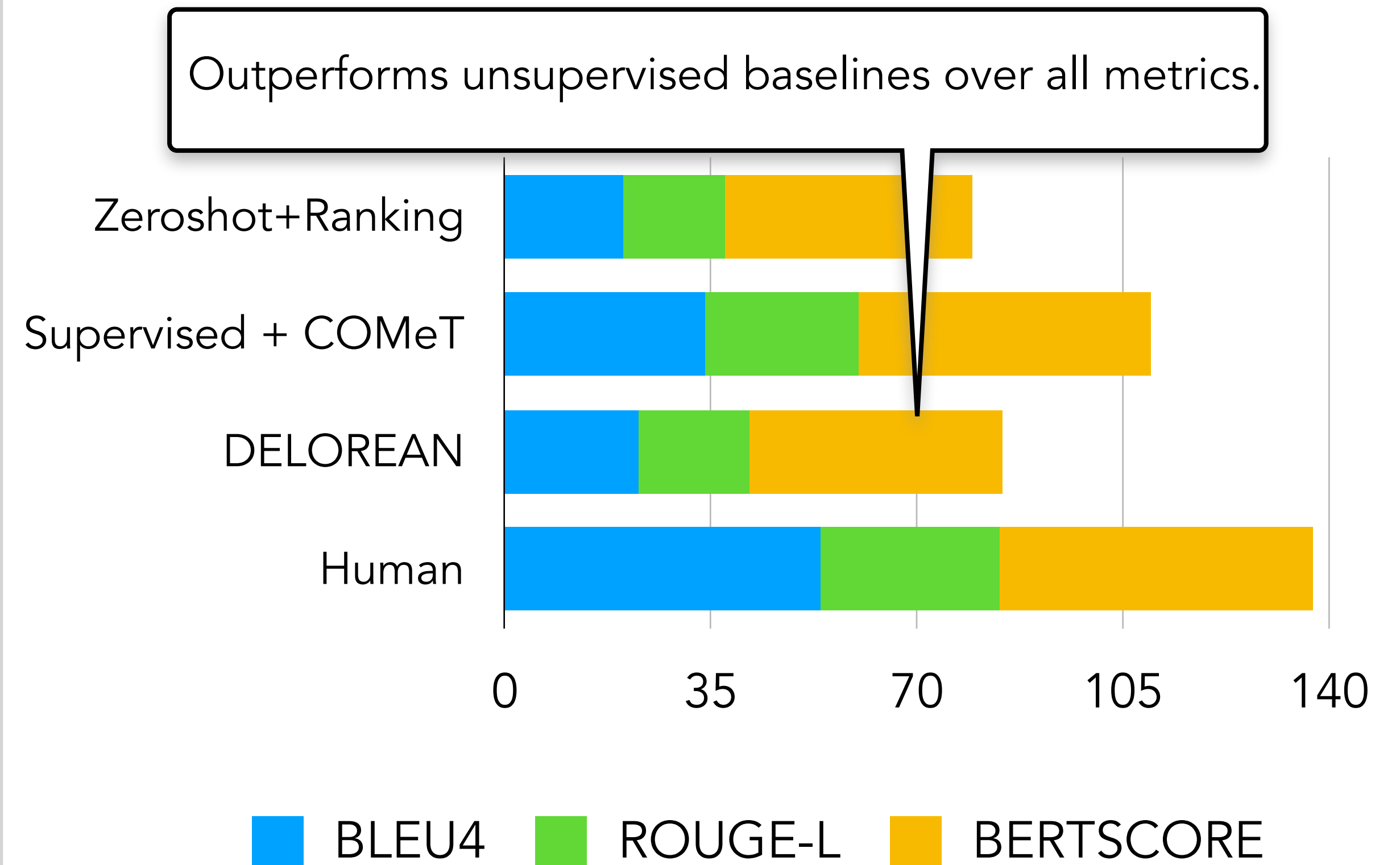
Zeke thought about ... [S]

# Abductive Reasoning

## Human Evaluation Results



## Automatic Evaluation Results



Please check the paper for more baselines ...

Language

Induction

Deduction

Abduction

Counterfactual

Reasoning

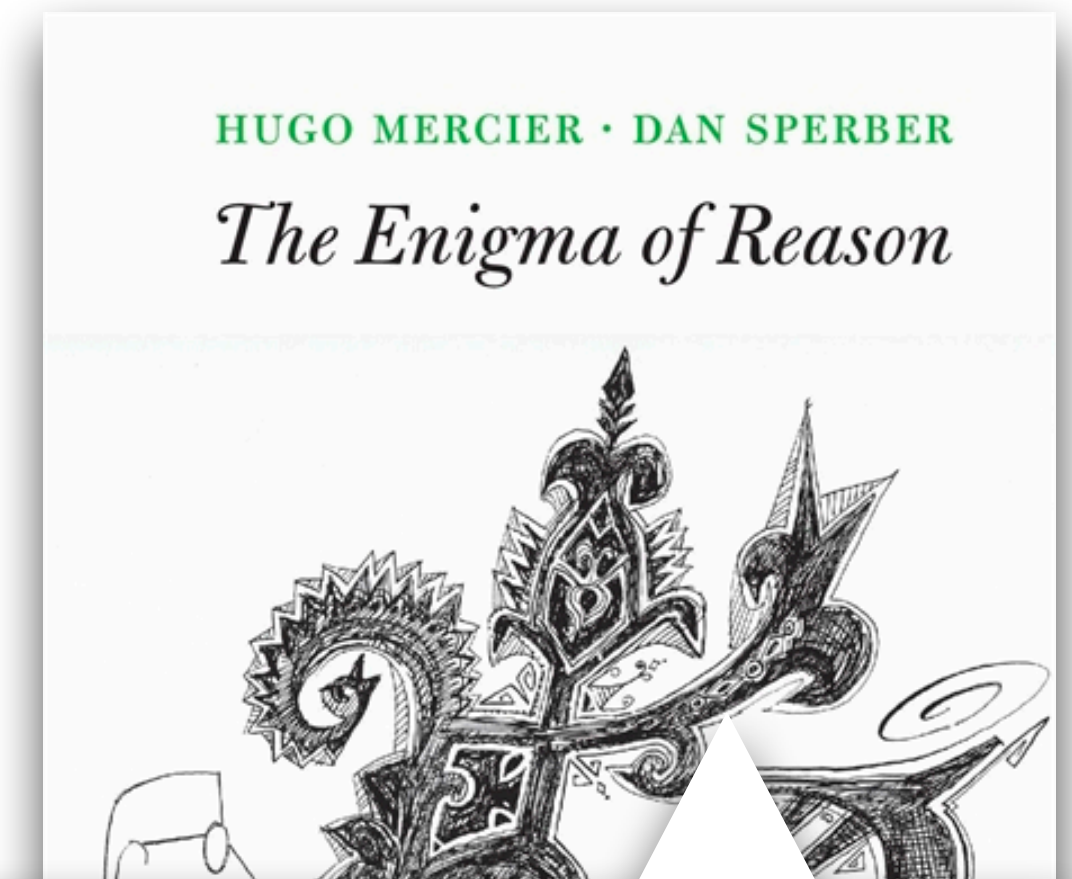
Reasoning as **generative tasks**

- As opposed to **discriminative tasks** (i.e., categorization)
- Because the space of reasoning in language is **infinite**

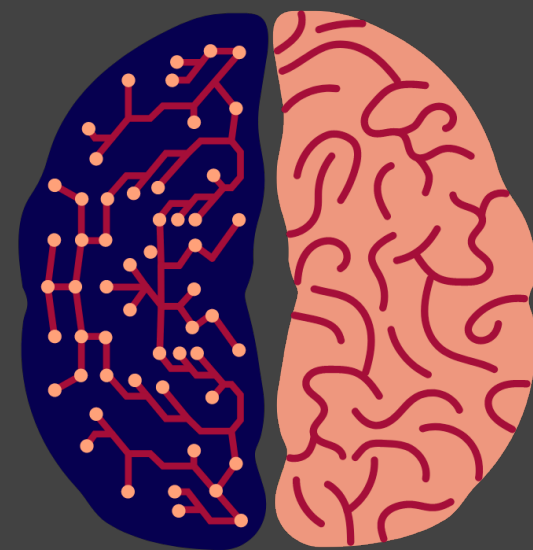
“thinking out loud”

We often think as we speak, on the fly, word-by-word without enumerating all possible alternative sentences

**Reasoning** serves the purpose of communication







# NEUROLOGIC DECODING

(Un)supervised Neural Text Generation with Predicate Logic Constraints

NAACL 2021

Ximing Lu



Peter  
West



Rowan  
Zellers



Ronan  
LeBras



Chandra  
Bhagavatula



Yejin  
Choi



# Seq2Seq

## Machine Translation

*X*

The physician told the baker that she had cancer.

*Y*

Der Arzt sagte dem Bäckerin, dass er Krebs habe.

## Dialogue Response

*X*

type	hotel
count	182
dogs allowed	don't care

*Y*

There are 182 hotels if you do not care whether dogs are allowed .

*Y*



## Language Model

## COMMONGEN

*X*

{ food, table, sit, front }

*Y*

The man sat with his food at the front of the table.

## Image Captioning

*X*



*Y*

Man in blue wetsuit is surfing on wave.

*X*



# COMMONGEN (Lin et al. EMNLP 2020)

missing keyword { **lose**, **ride** }



A man is trying to keep his **balance** as he **falls** off a **board** .

should use **all** given keywords ....

**Fine-tuned Language Model**

$X$

{ **board**, **lose**, **ride**, **fall**, **balance** }





# NEUROLOGIC DECODING

$C$

Decoding

Logical Constraint in CNF form

$$\underbrace{(\mathcal{D}_1 \vee \mathcal{D}_2 \cdots \vee \mathcal{D}_i)}_{\mathcal{C}_1} \wedge \cdots \wedge \underbrace{(\mathcal{D}_k \vee \mathcal{D}_{k+1} \cdots \vee \mathcal{D}_l)}_{\mathcal{C}_m}$$

$Y$

Advanced beam search with diverse partial solutions of CNF in consideration of four dynamic states of clauses:

- reversible unsatisfaction
- irreversible unsatisfaction
- reversible satisfaction
- irreversible satisfaction

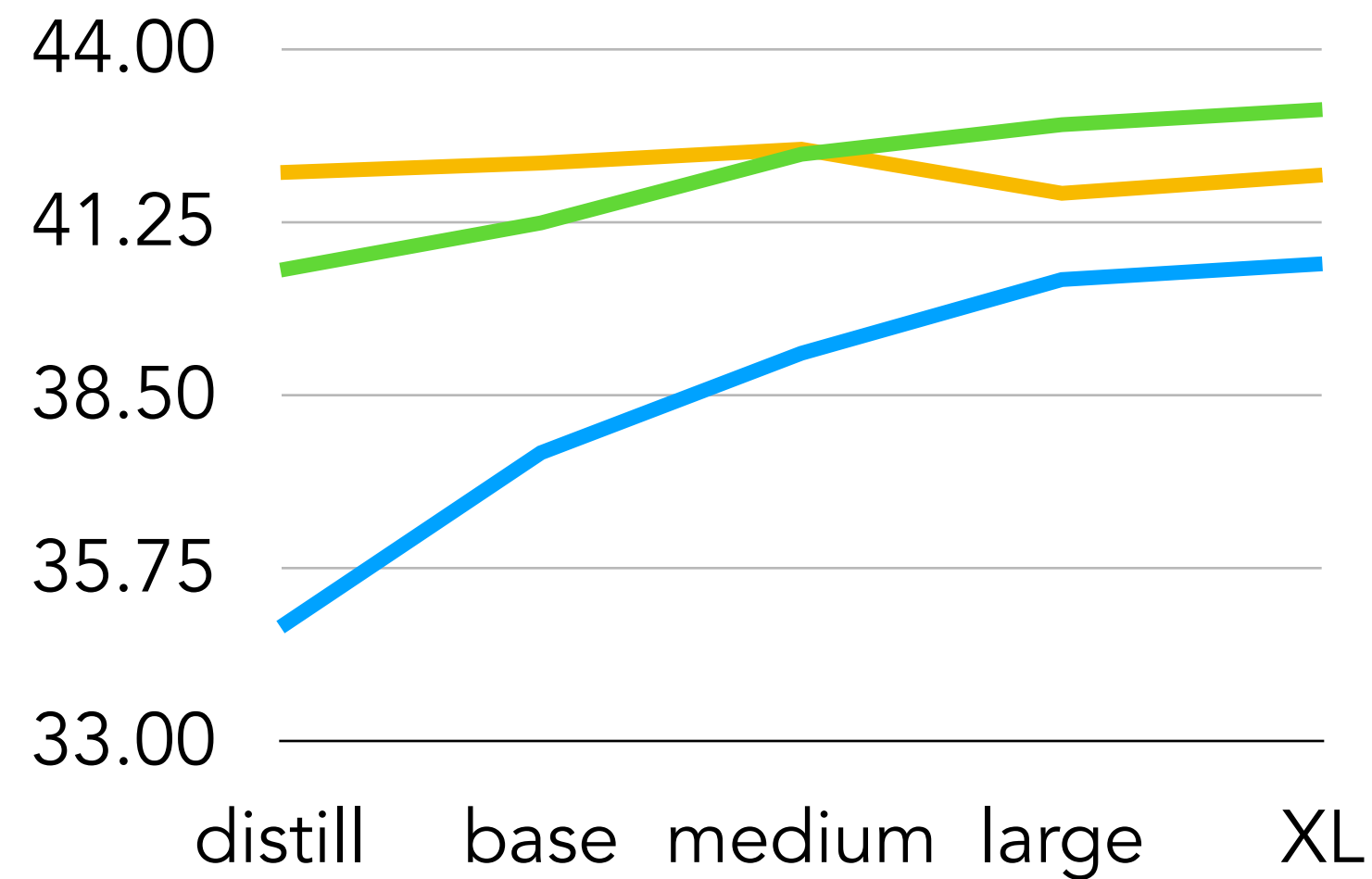
Language Model

$X$

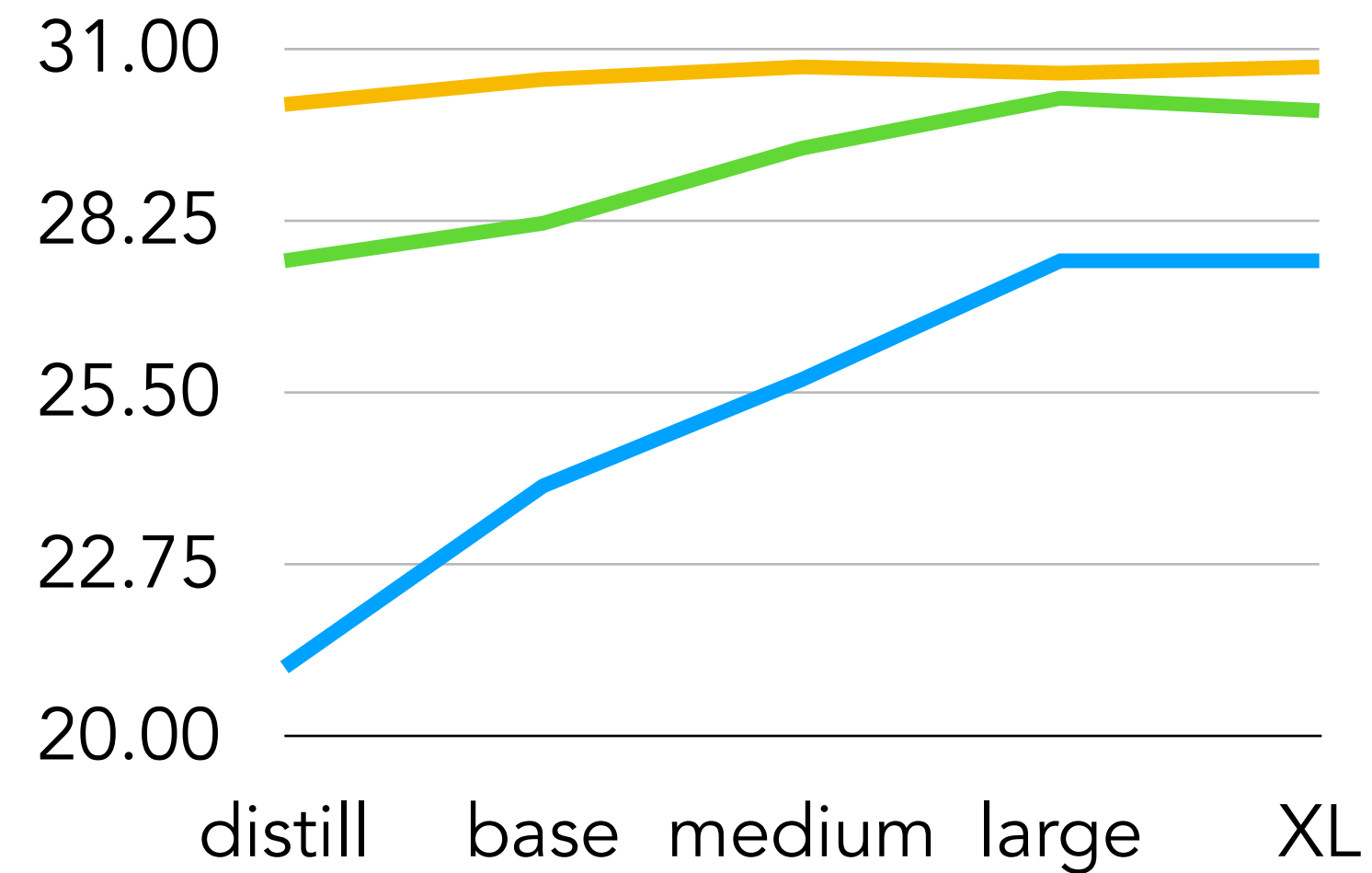
# COMMONGEN (Zero-shot)

- beam search (supervised)
- NeuroLogic (supervised)
- NeuroLogic (zero-shot)

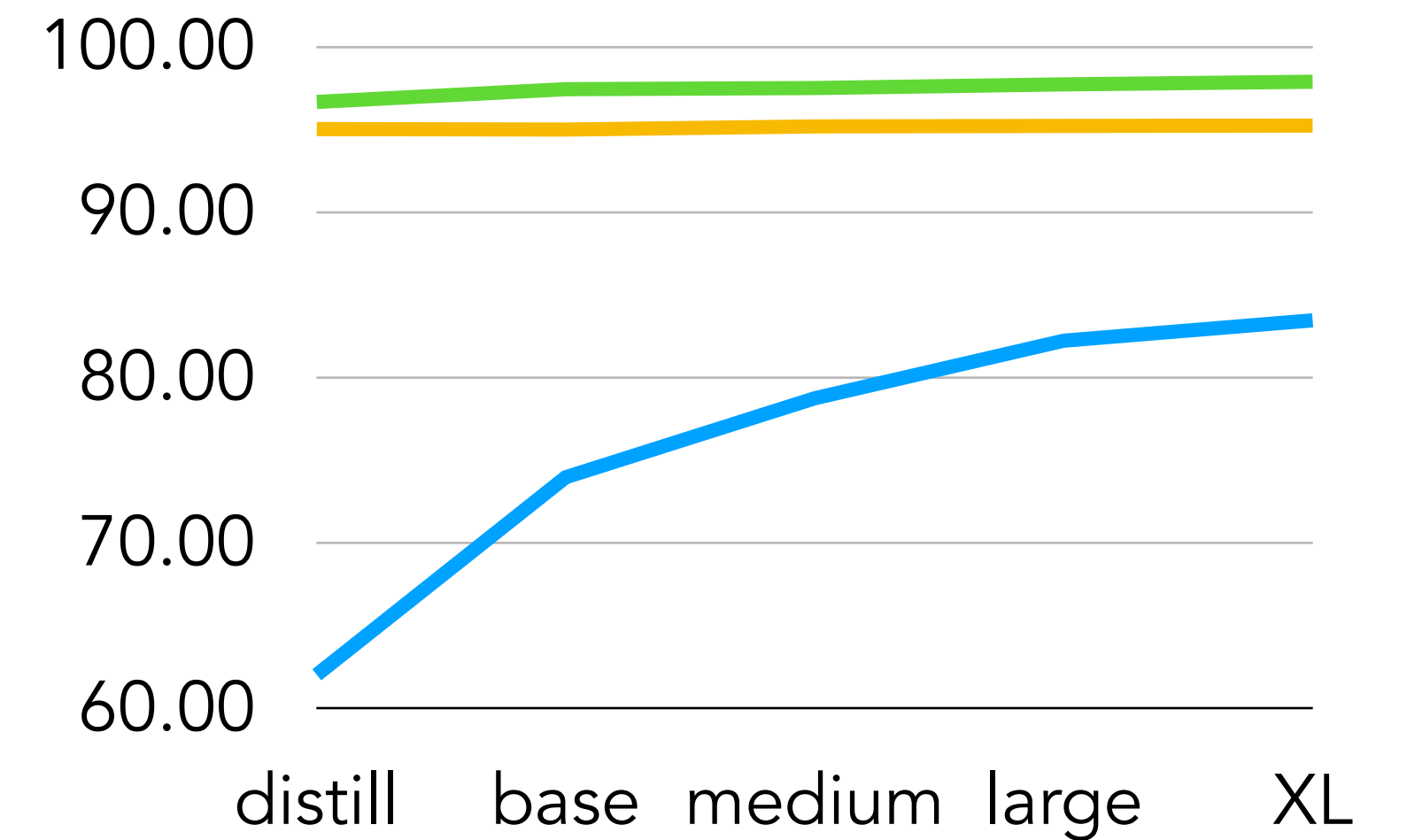
## ROUGE-L



## METEOR



## Coverage



Unsupervised NeuroLogic  
outperforms  
supervised approaches

Unsupervised NeuroLogic on **smaller**  
**networks** outperforms  
supervised approaches on **larger** networks!

# Path to commonsense?



Obligatory



Controversial Remarks of the Day



of the Day

Neural

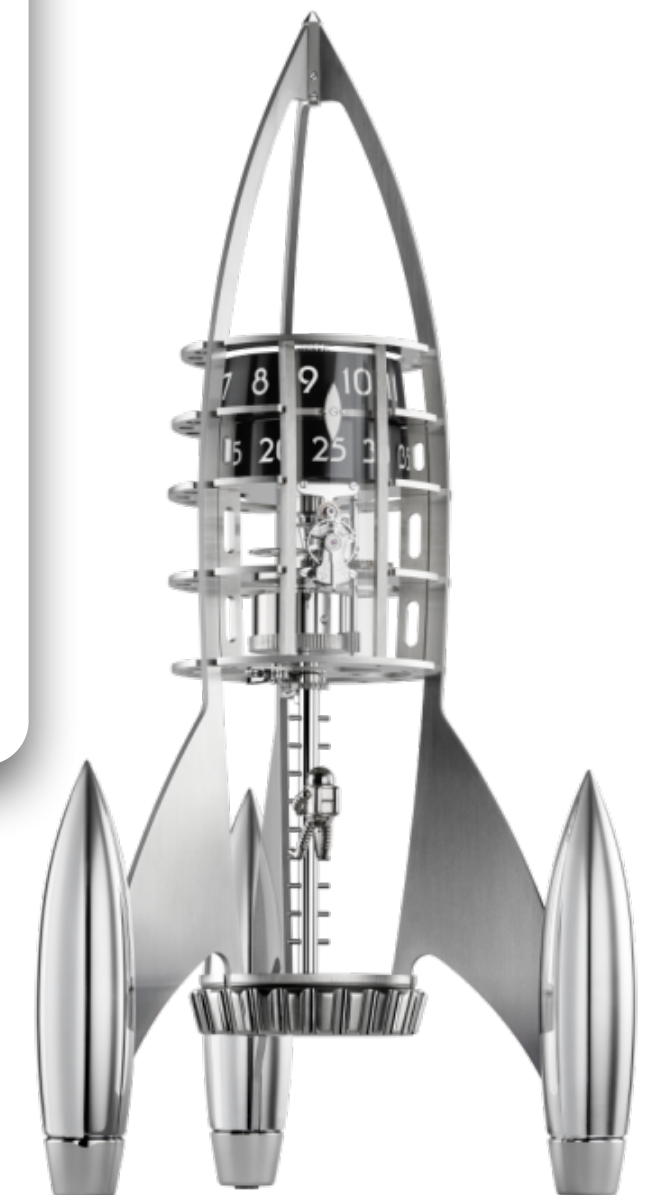
Symbolic

Language

Knowledge

Reasoning

1. the continuum between knowledge and reasoning
- the interplay between reasoning and language generation
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# Path to commonsense?



Obligatory



Controversial Remarks of the Day



of the Day

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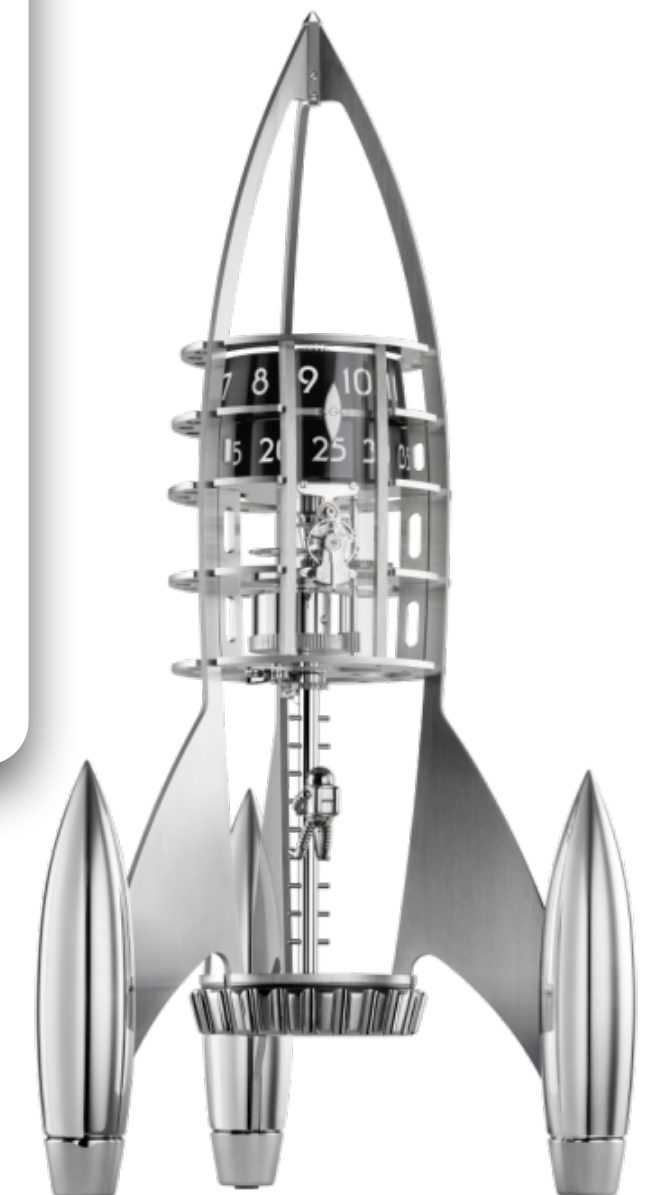
Symbolic

Language

Knowledge

Reasoning

1. the continuum between knowledge and reasoning
  2. the interplay between reasoning and language generation
- the blend between neural vs symbolic representation





# (COMET-) ATOMIC<sub>20</sub><sup>20</sup>:

## On Symbolic and Neural Commonsense Knowledge Graphs

— wait, doesn't GPT-3 know everything? —

AAAI 2021

Jena  
Hwang



Chandra  
Bhagavatula



Ronan  
Le Bras



Jeff  
Da



Keisuke  
Sakaguchi



Antoine  
Bosseult

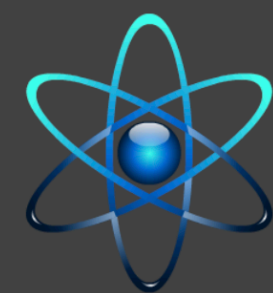


Me





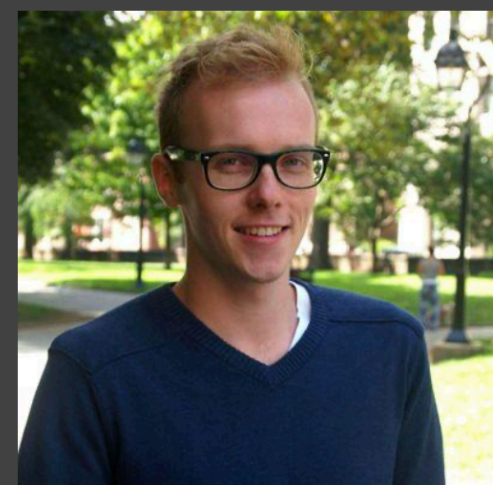
# Language models != knowledge models



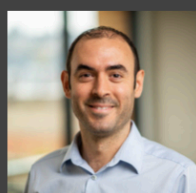
## ATOMIC: An Atlas of Machine Commonsense for If-Then Reasoning

AAAI 2019

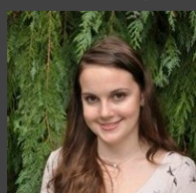
Maarten Sap



Ronan LeBras



Emily Allaway



Chandra Bhagavatula



Nicholas Lourie



Hannah Rashkin



Brendan Roof



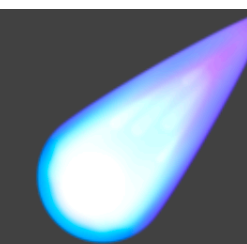
Noah Smith



Me



**Symbolic** commonsense knowledge graph



## COMET: Commonsense Transformers for Automatic Knowledge Graph Construction

ACL 2019

Antoine Bosselut



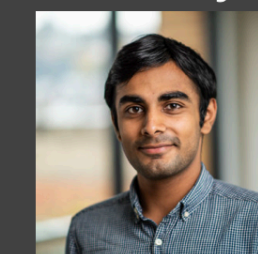
Hannah Rashkin



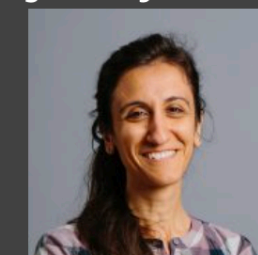
Maarten Sap



Chaitanya Malaviya



Asli Çelikyilmaz

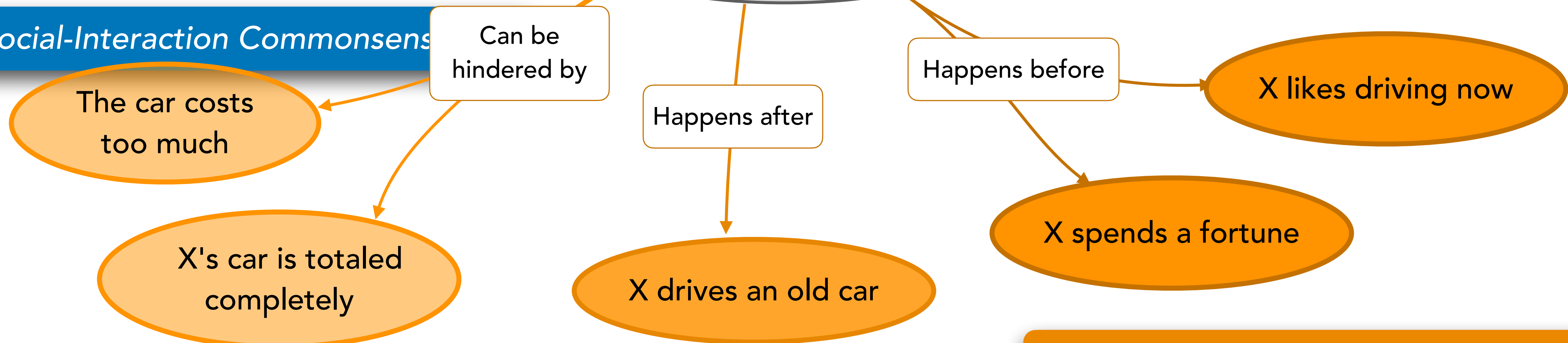
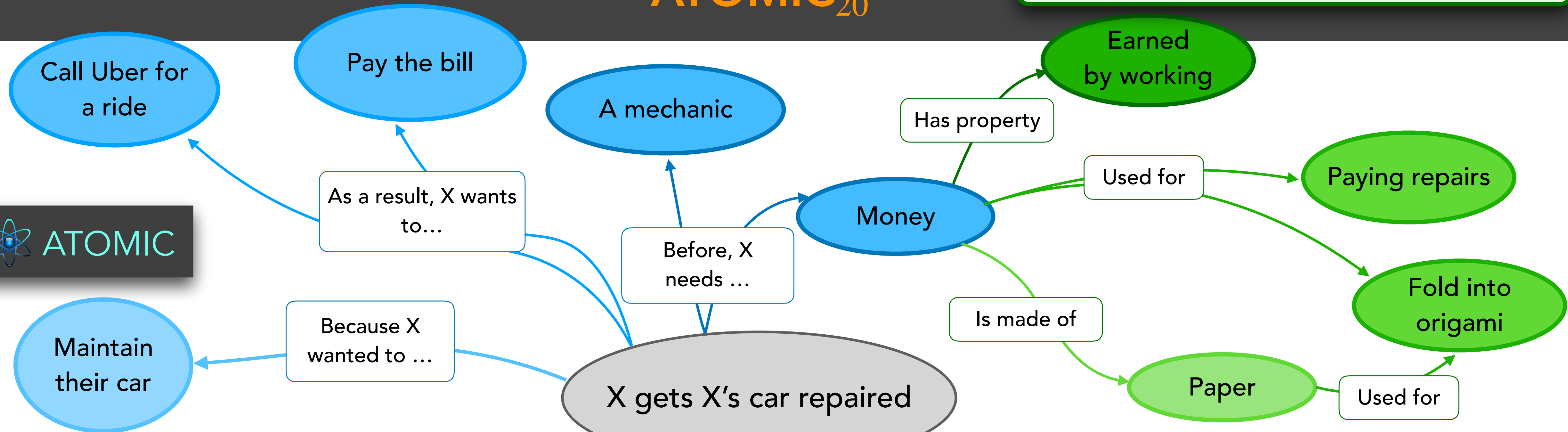


Me

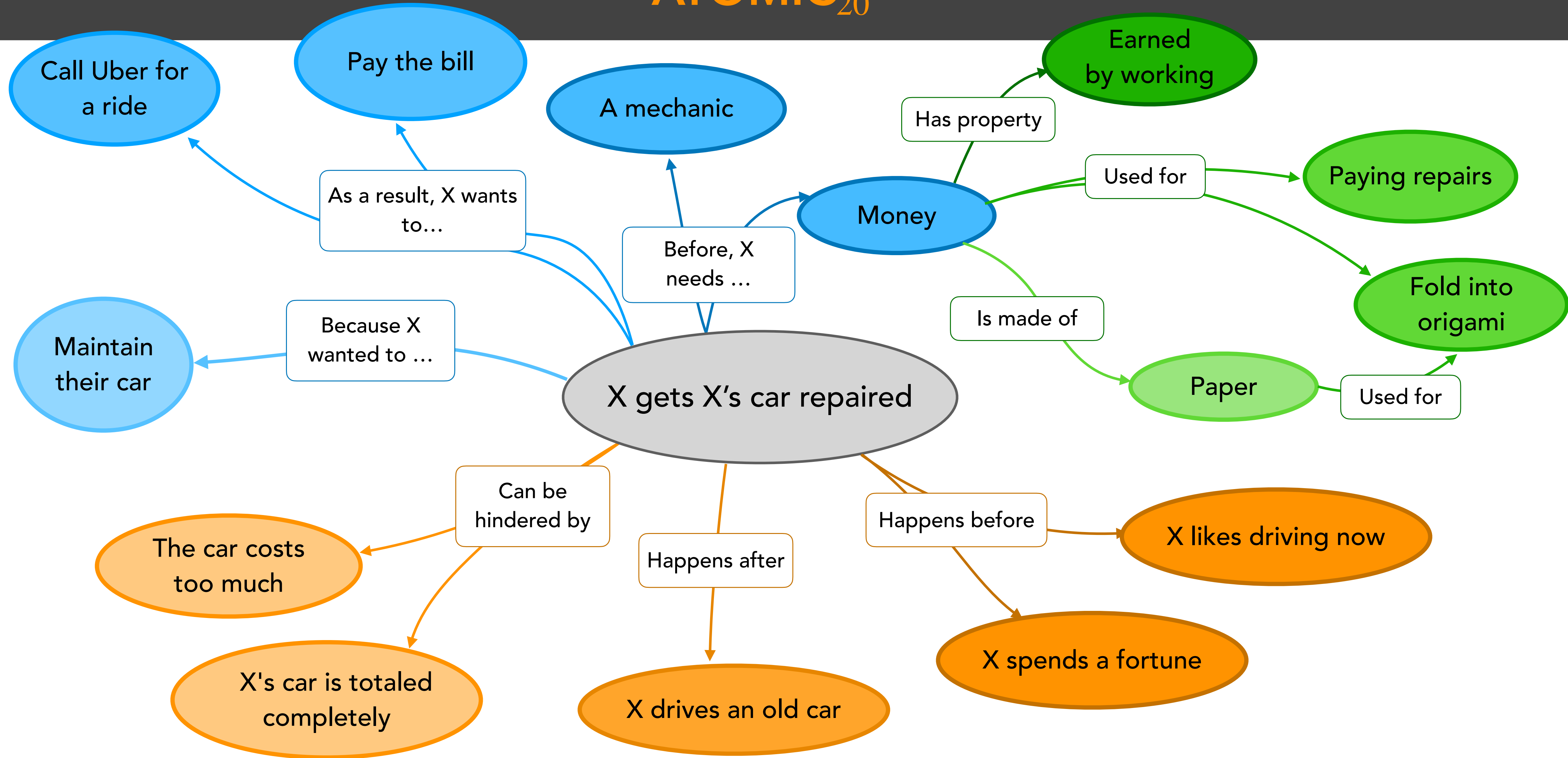


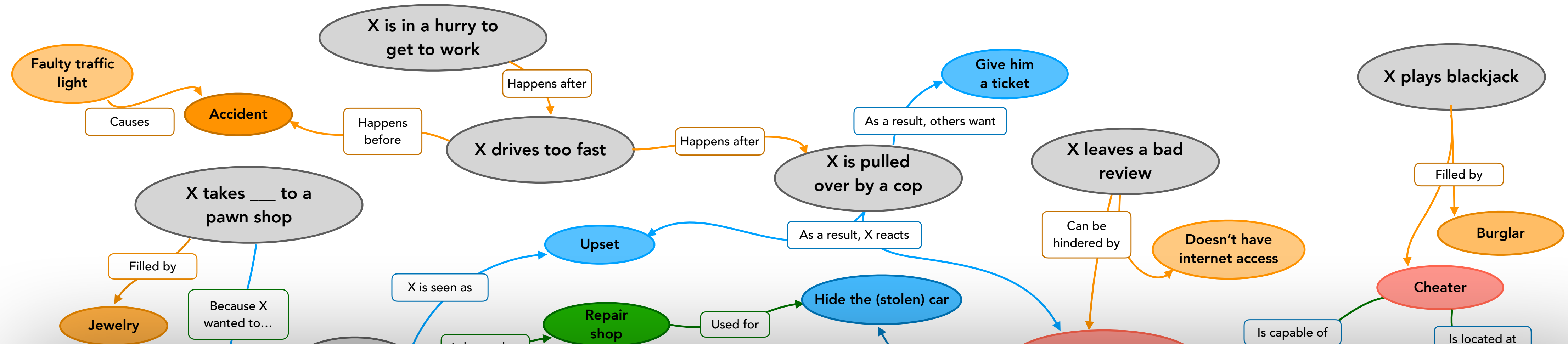
**Neural** commonsense model



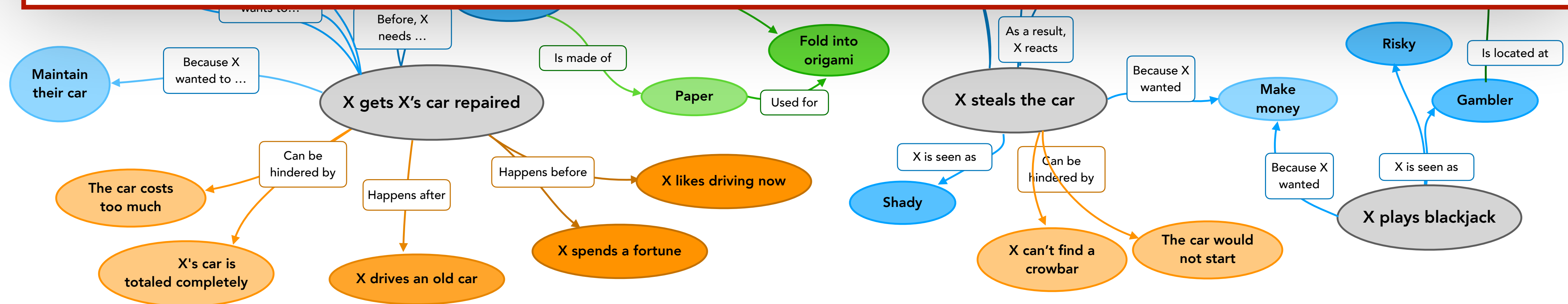


# ATOMIC<sup>20</sup><sub>20</sub>





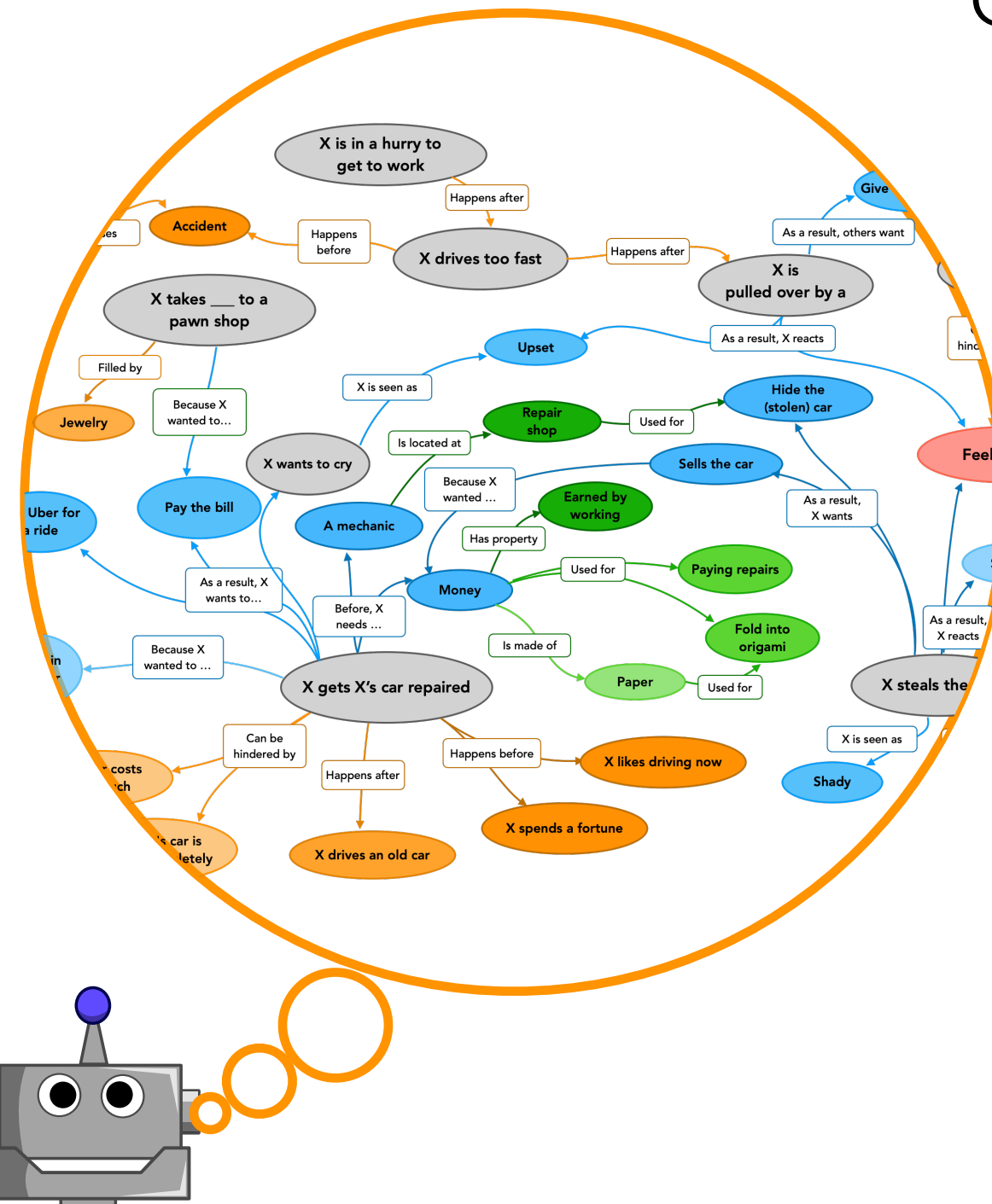
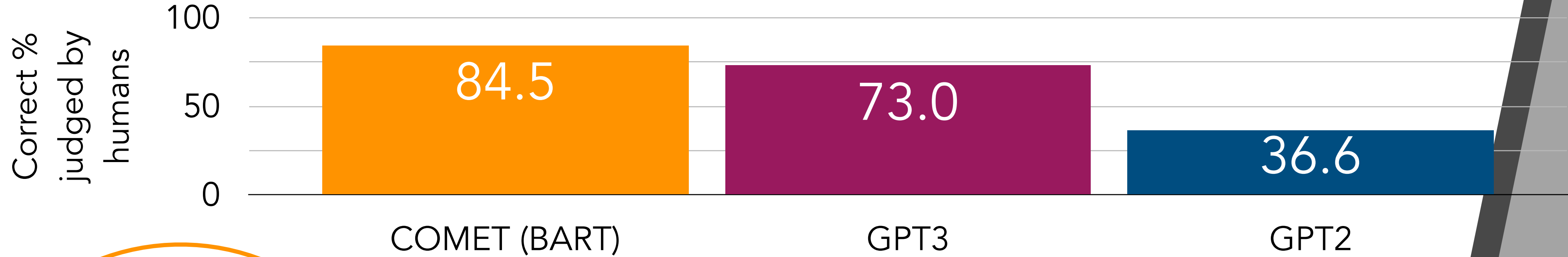
**1.33M commonsense if-then inferences**  
**23 relations (or inference types)**





## Knowledge Models

## Off-the-shelf Language Models



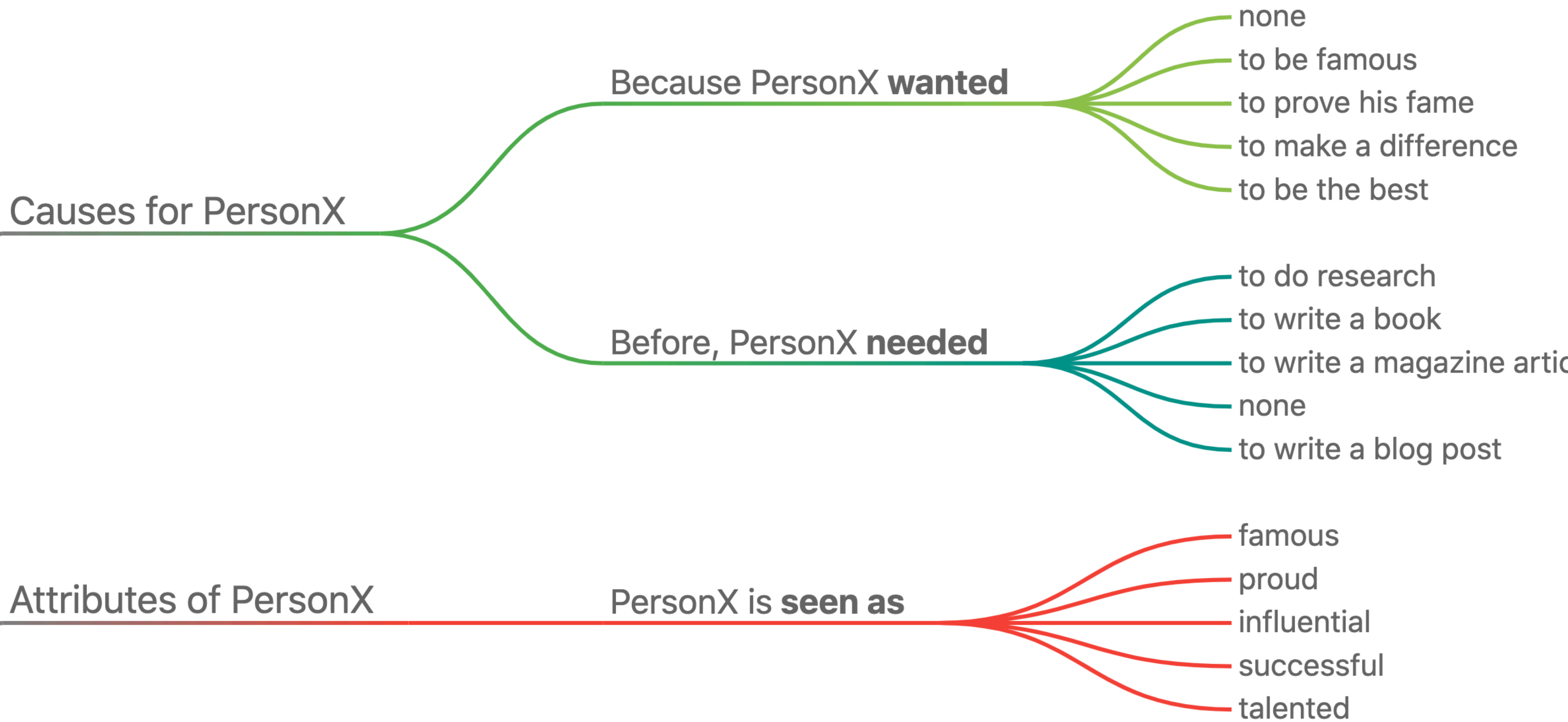
COMeT (BART): x435 smaller model (~400M parameters),  
informed by **ATOMIC**<sub>20</sub><sup>20</sup>

**GPT-3 (Few Shot): 175B parameters!!**  
pre-trained with a ton of web text (~500B tokens)

Turns out, COMET generalizes well  
on out-of-domain examples

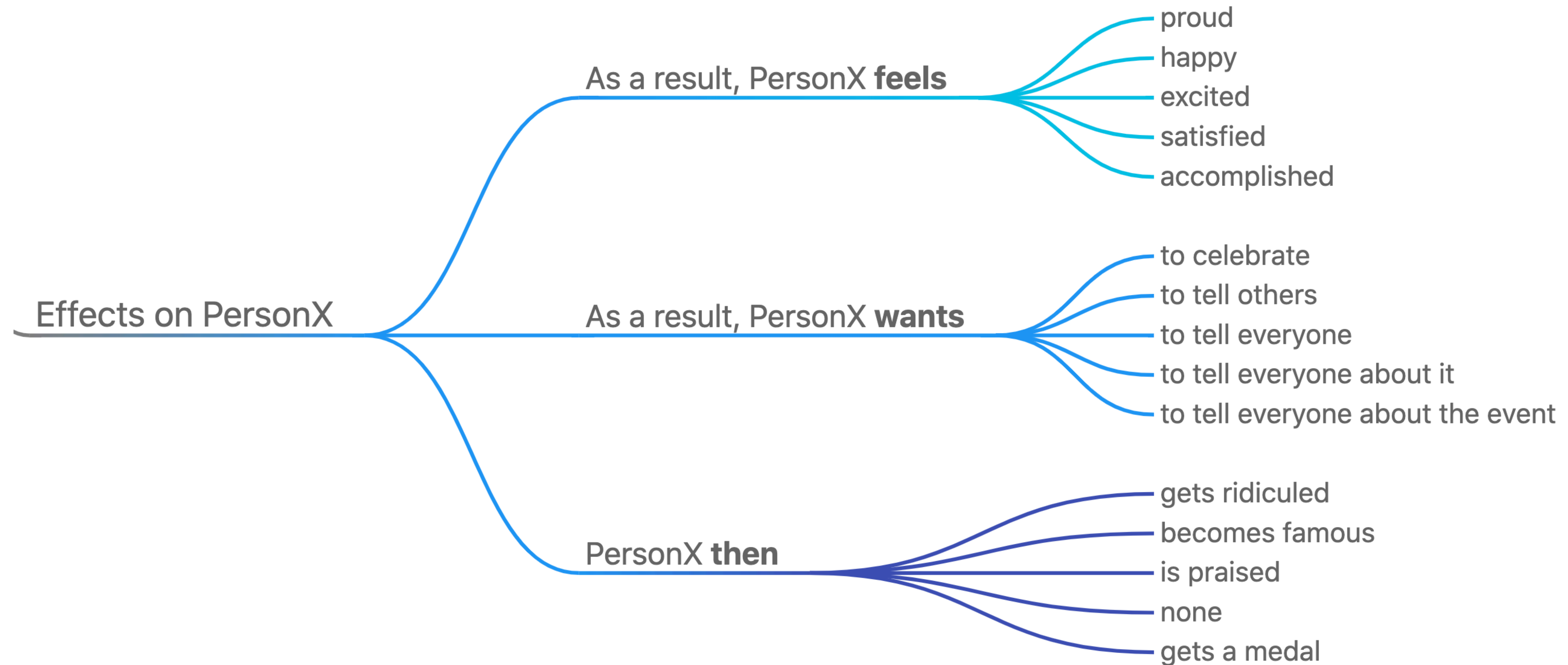
(which we realized only after publishing our ACL 2019 paper... )

# Gary breaks the world record for most controversial tweet.





# Gary breaks the world record for most controversial tweet.





**Gary Marcus** @GaryMarcus · 1h



ps i realize there is a typo in my query. fixing the typo doesn't much help though:

## Completion

**what happens when you stack kindling and logs in a fireplace and then drop some matches is that you typically start a** ick. So, it's kind of ironic that the second day after my son was born, the fire in the living room had melted through the kindling. It's pretty neat."





**Yejin Choi**  
@YejinChoinka

Replying to @GaryMarcus

Gary, try [mosaickg.apps.allenai.org](https://mosaickg.apps.allenai.org) by typing "Gary stacks kindling and logs and drops some matches". Sorry I used deep learning... :)





focus on  
"causes and effects"  
(causal knowledge)

COMeT



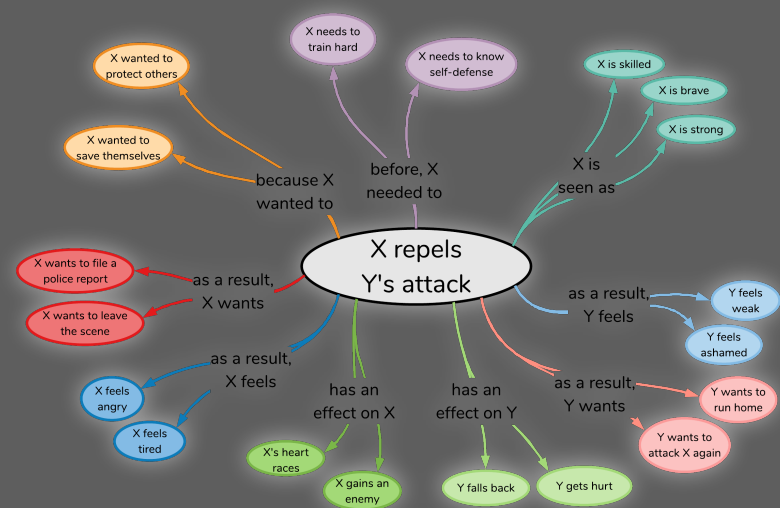
NEURAL  
(generalizes well to  
compositional &  
unseen events)



(semi-) supervised learning of  
declarative knowledge

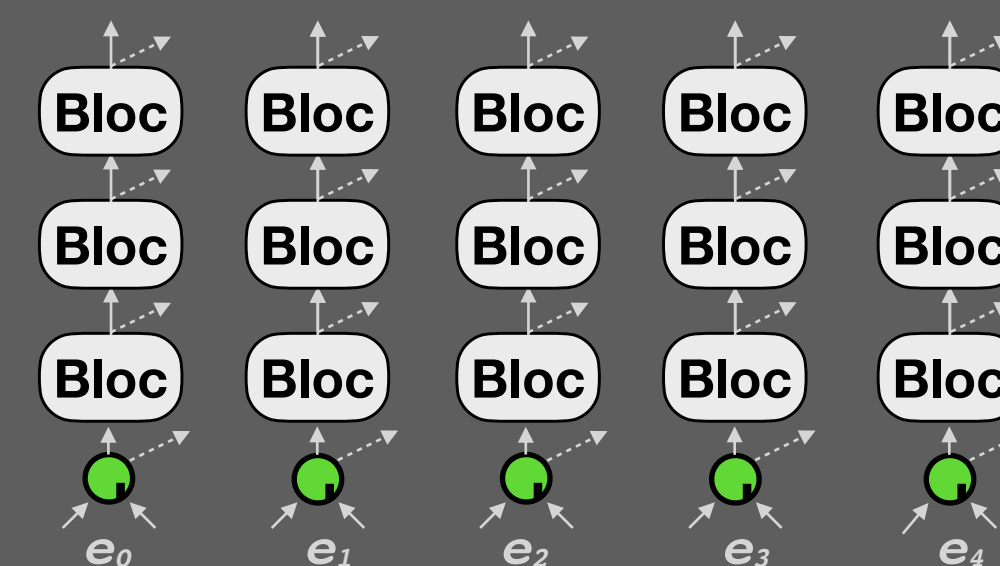
self supervised learning of  
observed knowledge

ATOMIC  
[tinyurl.com/atomic-](https://tinyurl.com/atomic-)



SYMBOLIC  
but in LANGUAGE  
(instead of LOGIC)

Language Models



# Scruples:

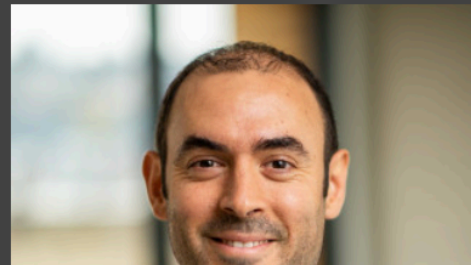
A Corpus of Community Ethical Judgments  
on 32,000 Real-Life Anecdotes

Nicholas  
Lourie

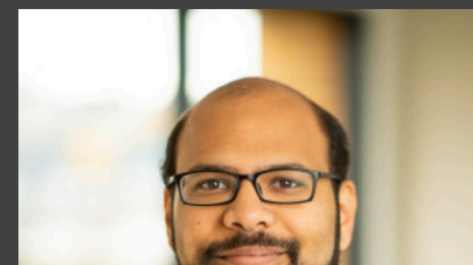


AAAI 2021

Ronan  
Le Bras



Chandra  
Bhagavatula

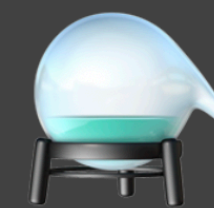


Yejin  
Choi



32,000 complex real-life  
anecdotes with community's  
ethical judgments

Extension of Social Commonsense to Social and Moral Norms



## Social Chemistry 101

Learning to Reason about Social and Moral Norms

EMNLP 2020

Maxwell Forbes



Jena Hwang



Vered Shwartz



Maarten Sap



Me



300,000 Rules of Thumb  
for social/moral norms  
grounded on 104k real life  
situations





# The Curious Case of Cheeseburger Stabbing

— *An example repeatedly appeared my talks between Mar 2017 and May 2018* —



# The Curious Case of 🍔 "Cheeseburger Stabbing"

*(GPT-3 Davinchi with sampling temp=0.4)*

## Breaking News: Cheeseburger Stabbing.

A man has been arrested after stabbing a cheeseburger. (...) He stabbed the cheeseburger and then stabbed himself in the stomach. The man was taken to the hospital and is expected to live. The cheeseburger was taken to the morgue.

## Breaking News: Cheeseburger Stabbing.

A man has been arrested after allegedly stabbing another man with a cheeseburger in a fight over ketchup.

2020:

**"stabbing of a cheeseburger"  
per GPT-3**



((yoav' ())) @yoavgo · Jul 18, 2020

Q: parsley cake

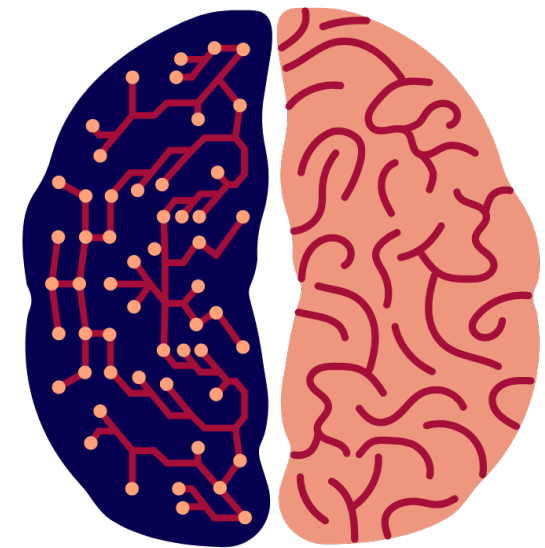
A: cake made of parsley

Q: cheeseburger stabbing

A: stabbing of a cheeseburger

# The Curious Case of 🍔 "Cheeseburger Stabbing"

**NeuroLogic Decoding** from **GPT-2** with logic constraints to include words derived from {stab, cheeseburger}



He was ***stabbed*** in the neck with a ***cheeseburger*** fork.

A man has been arrested after police say he ***stabbed*** a ***cheeseburger*** delivery man in the face.



# The Curious Case of 🍔 “Cheeseburger Stabbing”

**(moral) social Commonsense:**

which is *morally* worse?  
(so that it's news worthy)



Someone stabbed a cheeseburger?  
A cheeseburger stabbed someone?

A cheeseburger stabbed another cheeseburger?



Someone stabbed someone else over a cheeseburger?



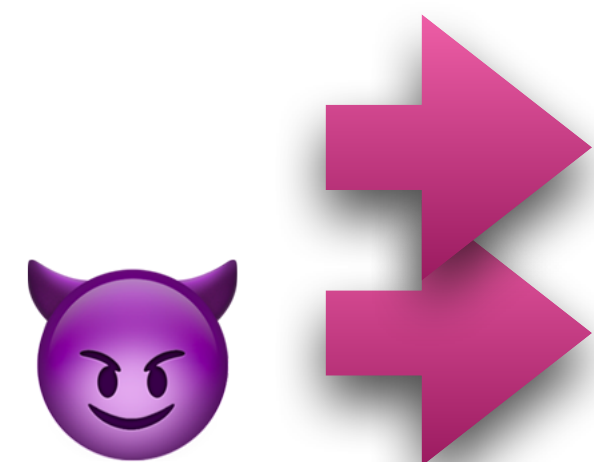
# The Curious Case of 🍔 “Cheeseburger Stabbing”

Sruples Demo:

<https://norms.apps.allenai.org/>



Someone stabbed a cheeseburger?  
A cheeseburger stabbed someone?



A cheeseburger stabbed another cheeseburger?  
Someone stabbed someone else over a cheeseburger?



# The Curious Case of 🍔 “Cheeseburger Stabbing”

Sruples Demo:

<https://norms.apps.allenai.org/>



Someone stabbed a cheeseburger?  
A cheeseburger stabbed someone?

A cheeseburger stabbed another cheeseburger?

👹 ➡ Someone stabbed someone else **over** a cheeseburger?  
➡ Someone stabbed someone else **with** a cheeseburger?



# Commonsense AI: Closing Remarks & Open Research Questions

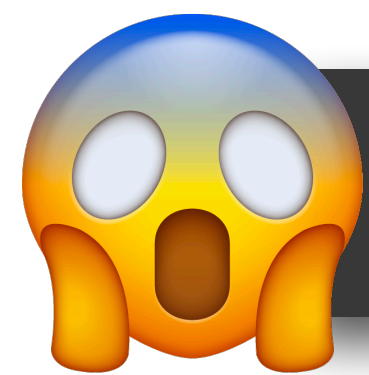




## Commonly held beliefs

- Knowledge and reasoning are distinct and exclusive
- Language is in the way of reasoning; let's do formal logics
- Language is not symbols. Words and numbers are, but not language at large
- Humans acquire commonsense completely un-/self-supervised, thus so should machines

Truth or Myth ?



## What is commonsense?

- It's what everyone knows and agrees on
- My dog (cat, baby, ...) has commonsense without language, thus commonsense AI doesn't need language



## Should we or should we not?

- Commonsense AI is an impossible goal (ever)
- That's a research topic of 70s and 80s
- Maybe only possible in the faraway future
- Can't define commonsense, thus don't work on it

I was told not to speak the word commonsense...

Past failures (in 70s – 80s) are inconclusive

- weak computing power
- not much data
- no crowdsourcing
- not as strong models
- not ideal conceptualization

We can't define the "standard" English either, and yet, we have no problem studying language models

Truth or Myth ?



Should we or should we not?

- Commonsense AI is an impossible goal (ever)
- That's a research topic of 70s and 80s
- Maybe only possible in the faraway future
- Can't define commonsense, thus don't work on it



# Scruples:

A Corpus of Community Ethical Judgments  
on 32,000 Real-Life Anecdotes

Nicholas  
Lourie

AAAI 2021



Ronan  
Le Bras

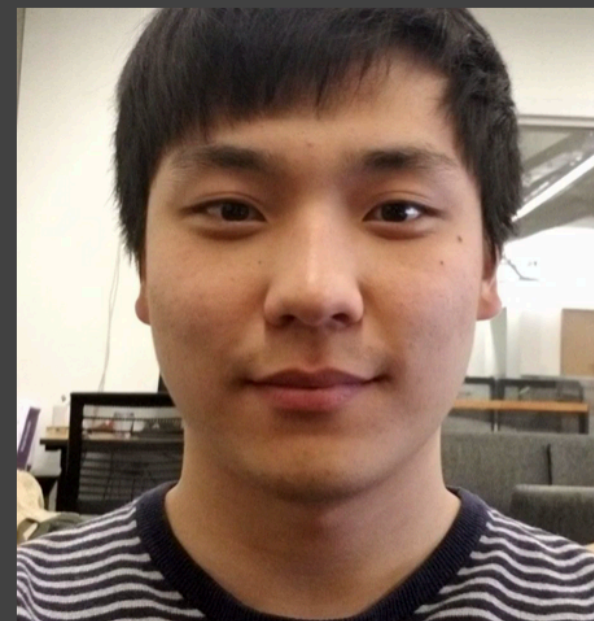


## Visual COMET:

Reasoning about the *Dynamic* Context of a *Still* Image

ECCV 2020

Jae Sung (James) Park



Chandra  
Bhagavatula



Roosbeh  
Mottaghi



Ali  
Farhadi



Yejin  
Choi



EMNLP 2020

Maxwell Forbes



Jena Hwang



Vered Shwartz



Maarten Sap



Me

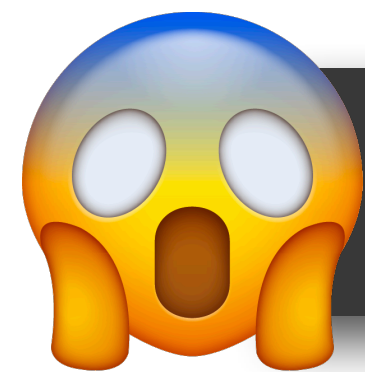


ministry 101  
Social and Moral Norms



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Truth or Myth ?



Language

Knowledge

Reasoning

- the continuum between knowledge and reasoning
- the interplay between reasoning and language generation
- the blend between neural vs symbolic knowledge





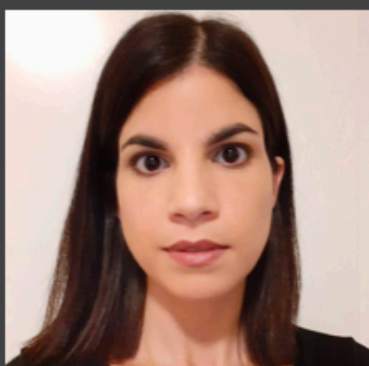
# Back to the Future:

## Unsupervised Backprop-based Decoding for Cou Cor

Lianhui Qin



Vered  
Shwartz



## (COMET-) ATOMIC<sub>20</sub><sub>20</sub> : On Symbolic and Neural Commonsense Knowledge Graphs — wait, doesn't GPT-3 know everything? —

AAAI 2021

Jena  
Hwang



## NEUROLOGIC DECODING

(Un)supervised Neural Text Generation with Predicate Logic Constraints

NAACL 2021

Ximing Lu



Peter  
West



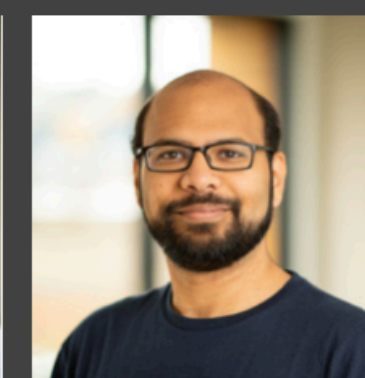
Rowan  
Zellers



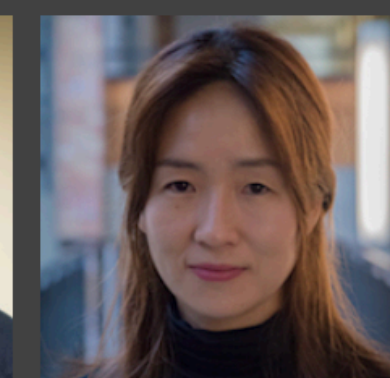
Ronan  
LeBras



Chandra  
Bhagavatula



Yejin  
Choi





Thanks! Questions?