

# Learning General Language Processing Agents

Dani Yogatama

# Language and Intelligence

A uniquely human ability that is a **core component** of our intelligence, independent of the surface forms it manifests in (Hockett, 1960).

ହାଲ୍ ପେର୍ଶେନ୍ଦେତ୍ଜେ **Halo**

Aloha こんにちは Sveiki ଶ୍ଲୋ

Ciao Ahoj **Hello** Сайн уу  
ନମସ୍କାର

KAMUSTA Γειά σου 여보세요 Salve

Здравствуйте مرحبا Merhaba

Hej 你好 Hola xin chào

# Language and Intelligence

A primary medium through which we **acquire** new skills and knowledge (+visual perception).



# Language and Intelligence

The **most effective** form of communication to **transmit** information and knowledge to others.

(Language for communication; Wittgenstein, 1953; Austin, 1975)



# Language and Intelligence

A mechanism with which we **formulate our thought process**. (Language for thinking; Spelke, 2003)



# Language and Intelligence

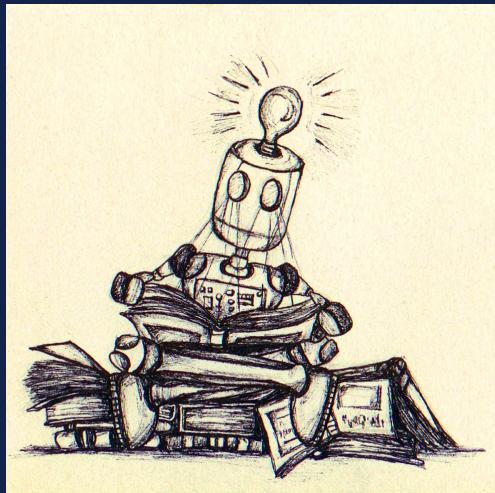
**Language** is key to **human intelligence** and is important for  
**artificial intelligence.**

# General Linguistic Intelligence

The ability to **acquire, store, and reuse** knowledge (about a language's lexicon, syntax, semantics, and pragmatic conventions) to **adapt** to new tasks **quickly without forgetting** old ones.

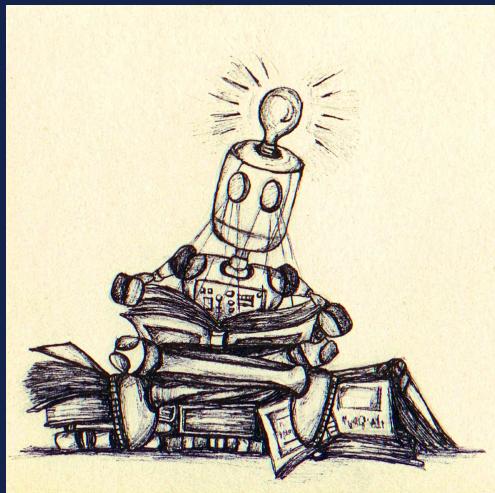
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Здравствуйте اب حرم Merhaba  
Hej 你好 Hola xin chào



# The State of Natural Language Processing

State-of-the-art models are based on increasingly larger transformers.

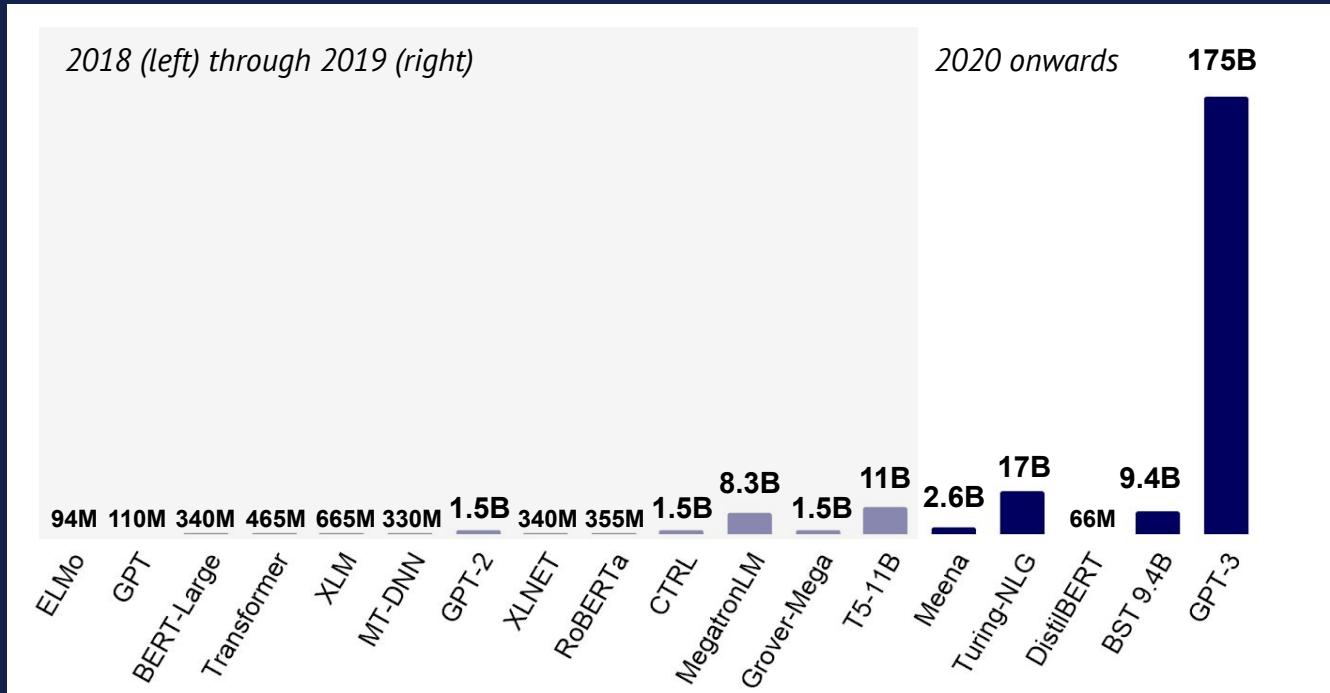


Figure taken from [State of AI Report 2020](#).

# Challenges: Human Learning vs. Machine Learning



Human

“Large” datasets

Acquisition

# Challenges: Human Learning vs. Machine Learning



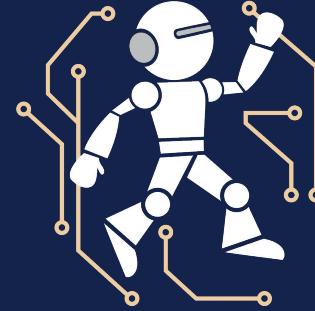
Human	
``Large'' datasets	<b>Acquisition</b>
Few examples	<b>Task Training</b>

# Challenges: Human Learning vs. Machine Learning



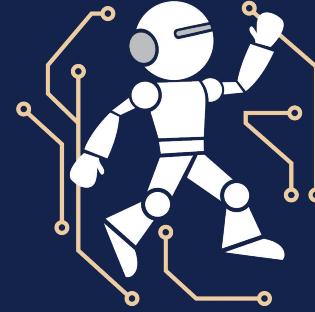
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Generalizable to new tasks	<b>Generalization</b>

# Challenges: Human Learning vs. Machine Learning



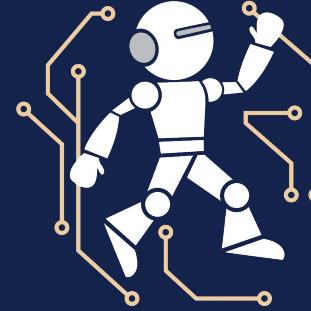
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# Challenges: Human Learning vs. Machine Learning



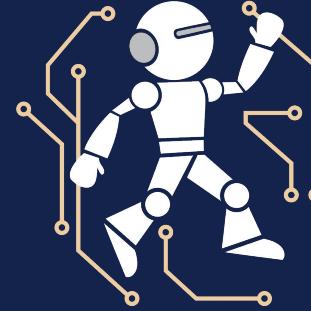
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# Challenges: Human Learning vs. Machine Learning



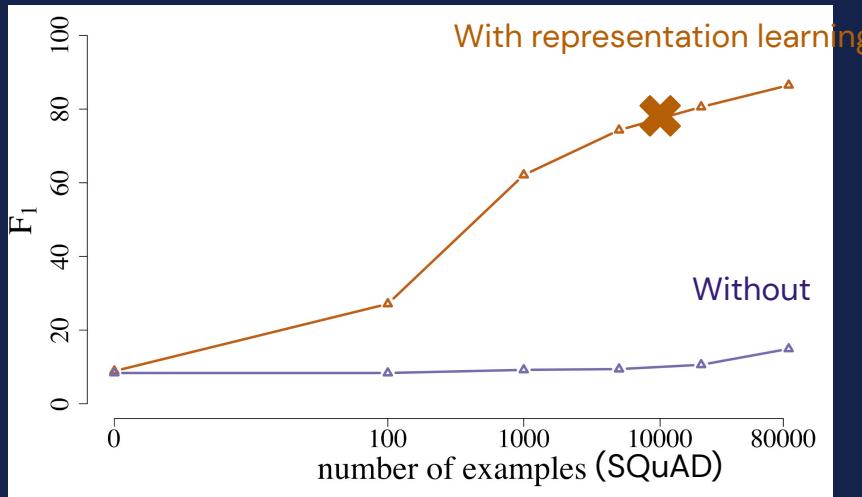
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# Challenges: Human Learning vs. Machine Learning



Human		Machine
“Large” datasets	<b>Acquisition</b>	Large datasets (representation learning)
Few examples	<b>Task Training</b>	Large datasets (supervised fine tuning)
Dataset agnostic	<b>Linguistic knowledge</b>	Dataset specific
Generalizable to new tasks	<b>Generalization</b>	Forget previous tasks given a new task

# The State of Natural Language Processing

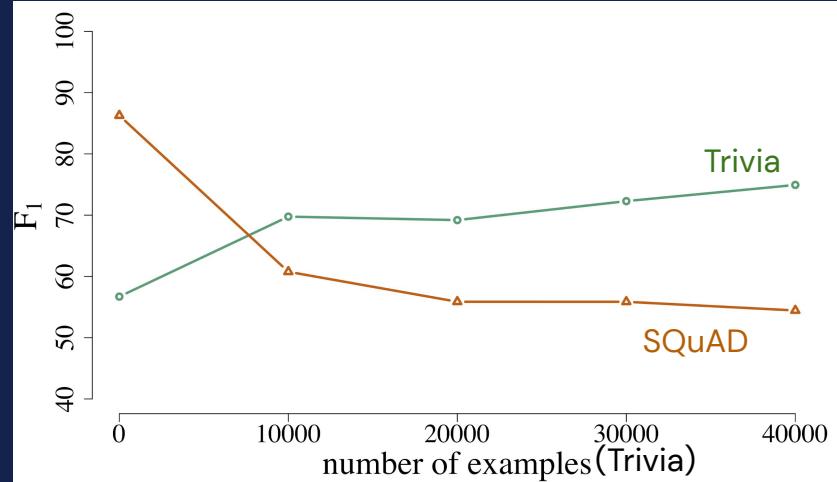
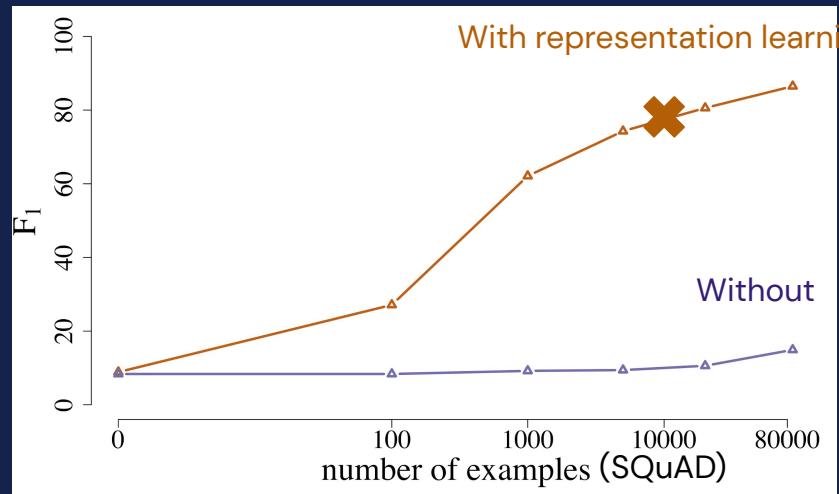


**Yogatama et al., arXiv 2019**

Model: BERT, [Devlin et al. 2019](#)

QA dataset: SQuAD, [Rajpurkar et al., 2016](#)

# The State of Natural Language Processing



Model: BERT, [Devlin et al. 2019](#)

QA dataset: SQuAD, [Rajpurkar et al., 2016](#)

QA dataset 2: Trivia, [Joshi et al., 2017](#)

**Yogatama et al., arXiv 2019**

# Research Areas



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.

# Research Areas



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.

**Training Paradigms**

**Model Architectures**

# Research Areas



## Training Paradigms

Representation Learning

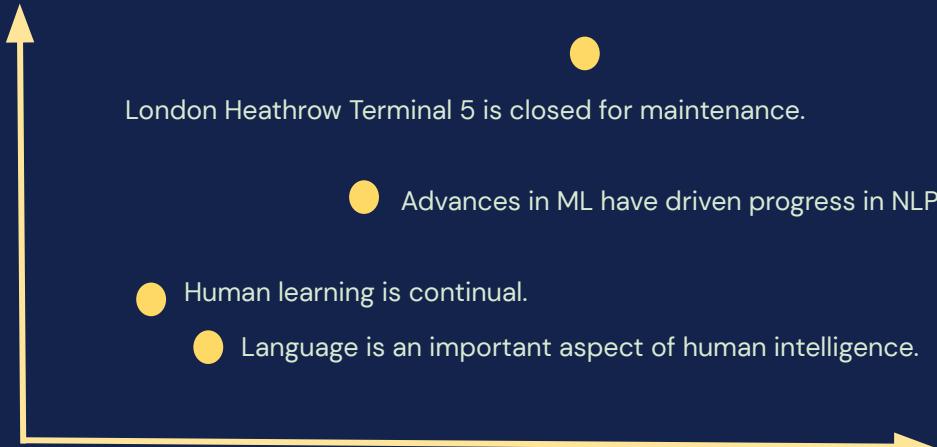
**Yogatama and Smith, ACL 2014**

**Yogatama et al., ACL 2015**

**Yogatama and Smith; ICML 2015**

**Kong, de Masson d'Autume, Ling, Yu, Dai, Yogatama; ICLR 2020**

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# Research Areas



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.

## Training Paradigms

Generative Training

**Yogatama et al., TACL 2014**

**Yogatama et al., arXiv 2017**

**Kong, Melis, Ling, Yu, and Yogatama, ICLR 2018**

**Cao and Yogatama, arXiv 2020**

$$\mathcal{L} = \log p(\mathbf{x}, y)$$

# Research Areas



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.

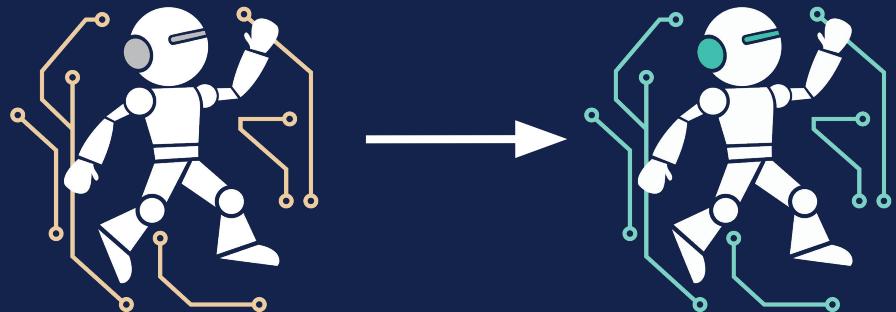
## Training Paradigms

Few-shot and Transfer Learning

**Yogatama and Mann, AISTATS 2014**

**Yogatama et al., EMNLP 2015**

**Artetxe, Ruder, Yogatama, ACL 2020**



# Research Areas



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.

## Model Architectures

Memory Networks

**Yogatama et al., ICLR 2017**

**Yogatama et al., ICLR 2018**

**de Masson d'Autume, Ruder, Kong, Yogatama, NeurIPS 2019**

**Yogatama et al., TACL 2021**

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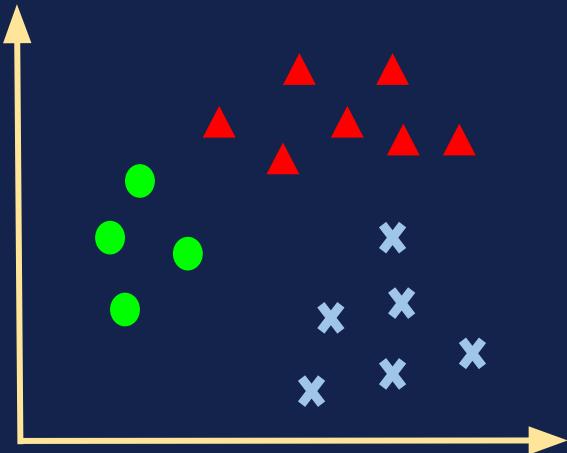
**de Masson d'Autume, Ruder, Kong, Yogatama, NeurIPS 2019**

**Yogatama et al., TACL 2021**

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## Model Architectures

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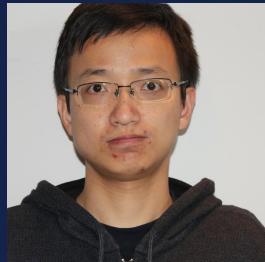
Yogatama et al., TACL 2021

# This Talk

- A framework for self-supervised language representation learning methods.  
**Kong et al., ICLR 2020**
- Semiparametric (memory-augmented) language models.  
**Yogatama et al., TACL 2021**

# A Mutual Information Maximization Perspective of Language Representation Learning

Kong et al., ICLR 2020



Lingpeng



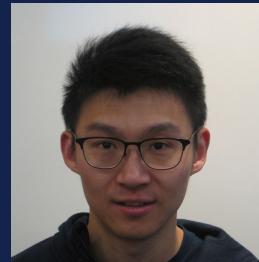
Cyprien



Wang



Lei



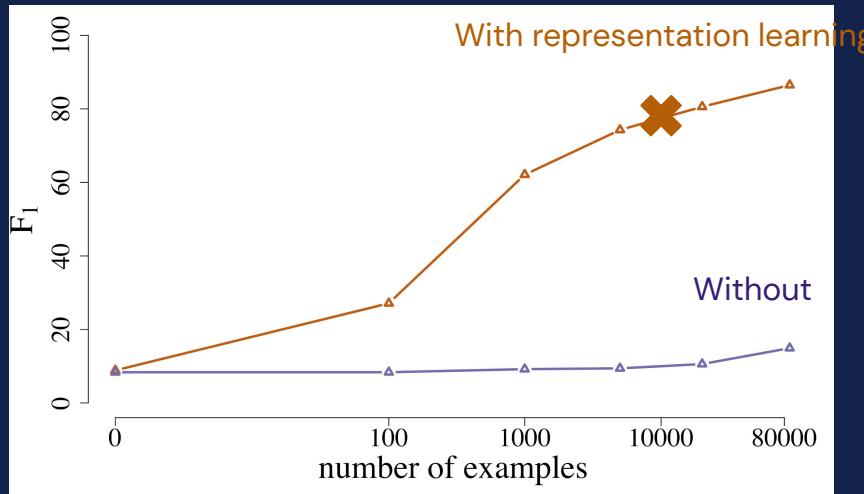
Zihang



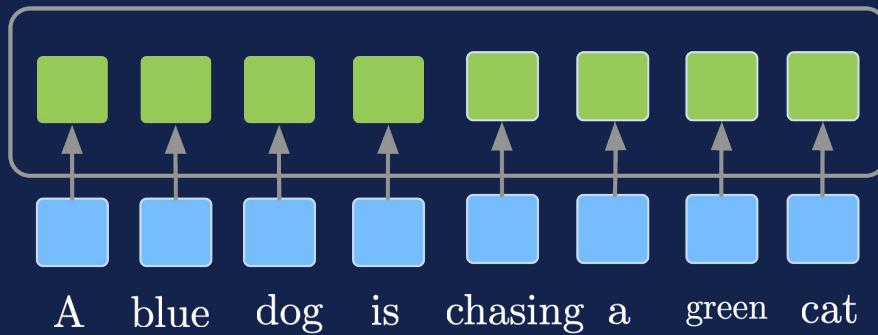
Dani

# Text Representations

Good representations facilitate more efficient transfer.



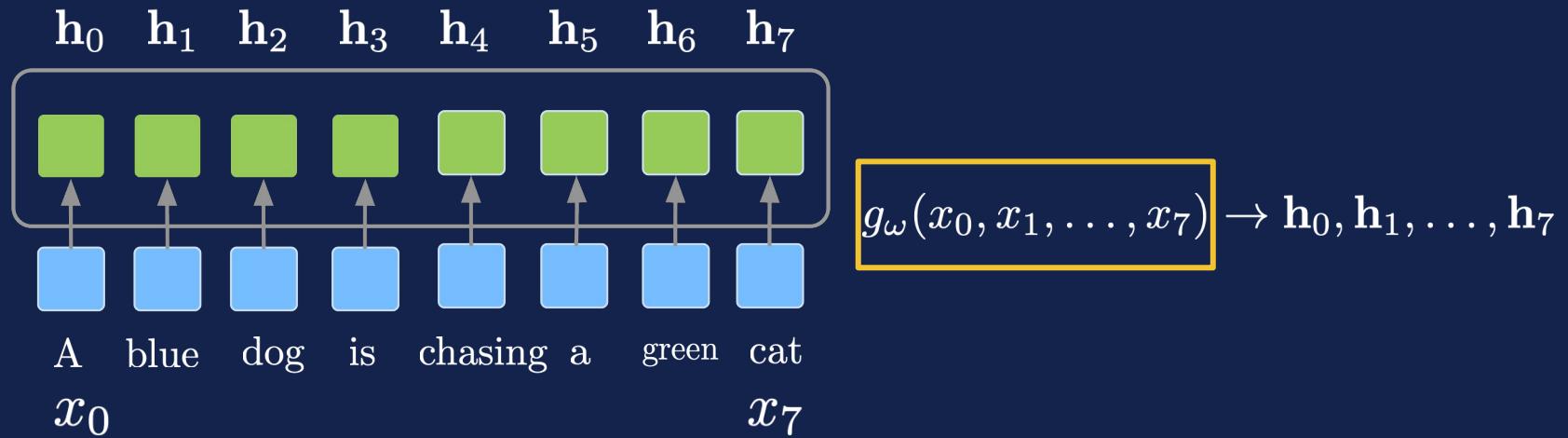
# Text Representations



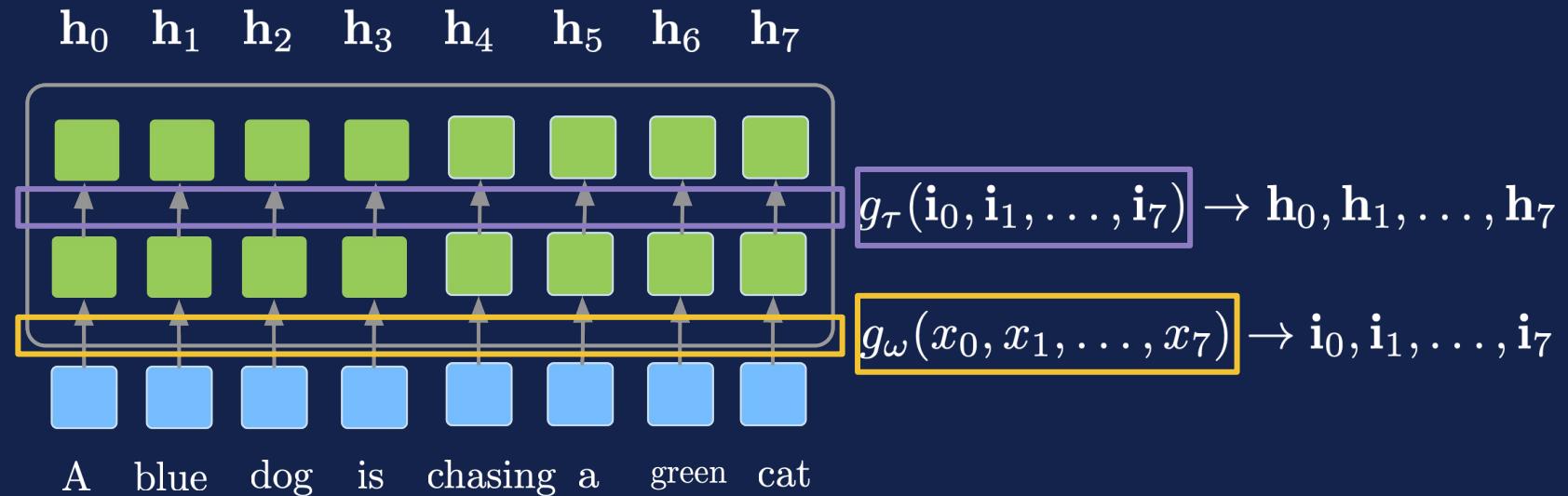
**a sequence of vectors  
(word representations)**

**a sequence of words**

# Text Representations



# Text Representations



# Text Representations



<https://twitter.com/SmithaMilli/status/837153616116985856/>

**Bag of words**

**Word embeddings**

**Contextual word embeddings**

Skip gram, Mikolov et al., 2013.  
GloVe, Pennington et al., 2014.

ELMo, Peters et al., 2018.  
BERT, Devlin et al., 2019.

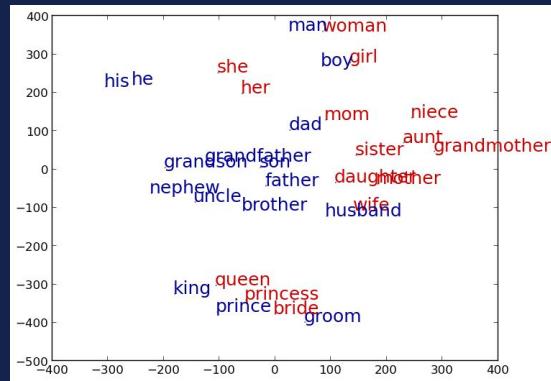


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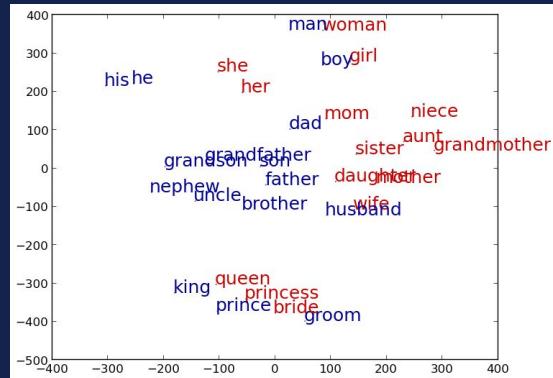


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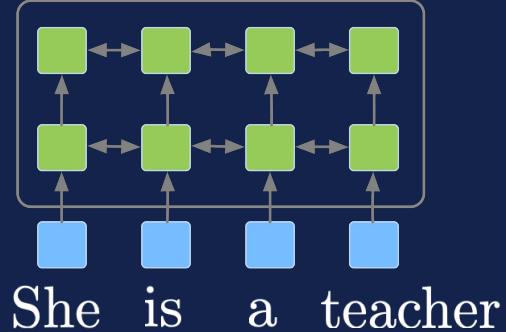
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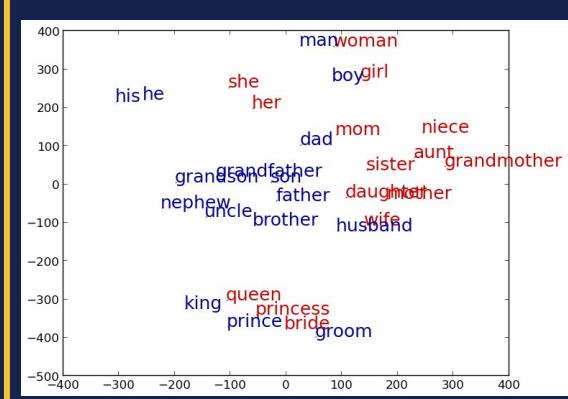
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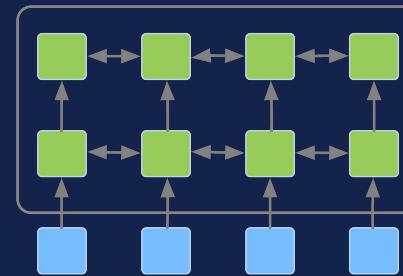
What has been the main driver of progress so far?



**Word embeddings**

Skip gram, Mikolov et al., 2013.

GloVe, Pennington et al., 2014.



**Contextual word embeddings**

ELMo, Peters et al., 2018.

BERT, Devlin et al., 2019.



# Contrastive Learning

**Main assumption:** representations should capture similarity (Arora et al., 2019).

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Human learning is continual.

Advances in ML have driven progress in NLP.  
Logistic regression can be used for classification.  
Transformer uses self attention.

There are many direct flights between London and Tokyo.  
London Heathrow Terminal 5 is closed for maintenance.

# Contrastive Learning with InfoNCE

**Main assumption:** representations should capture similarity (Arora et al., 2019).

$$I(A, B) \geq \mathbb{E}_{p(A, B)} \left[ \mathbb{E}_{p(C)} \left[ \log \frac{\exp f_{\theta}(a, b)}{\exp f_{\theta}(a, b) + \sum_{c \neq b} \exp f_{\theta}(a, c)} \right] \right]$$

InfoNCE objective  
Logeswaran and Lee, 2018  
van den Oord, et al., 2019

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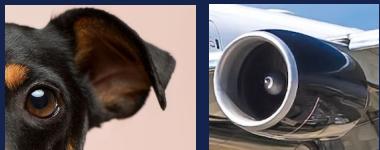
High when **a** and **b** go together

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InfoNCE objective  
Logeswaran and Lee, 2018  
van den Oord, et al., 2019



Low when **a** and **c** do not go together



# Contrastive Learning with InfoNCE

$$\mathbb{E}_{p(A,B)} \left[ \mathbb{E}_{p(C)} \left[ \log \frac{\exp f_{\theta}(a,b)}{\exp f_{\theta}(a,b) + \sum_{c \neq b} \exp f_{\theta}(a,c)} \right] \right]$$

The University of Waterloo is located in Canada

# Contrastive Learning with InfoNCE

$$\mathbb{E}_{p(A,B)} \left[ \mathbb{E}_{p(C)} \left[ \log \frac{\exp[f_{\theta}(a,b)]}{\exp f_{\theta}(a,b) + \sum_{c \neq b} \exp f_{\theta}(a,c)} \right] \right]$$

*a*      *b*

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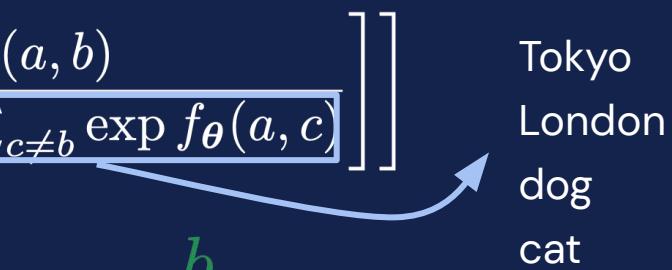
$$f_{\theta}(a,b) = g_{\psi}(b)^{\top} g_{\omega}(a)$$

# Contrastive Learning with InfoNCE

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The University of Waterloo is located in Canada

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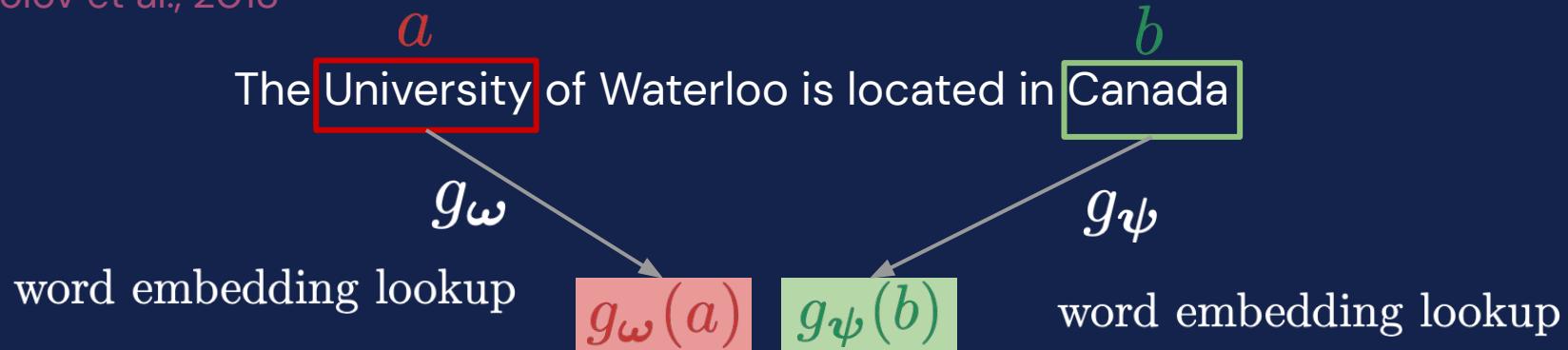
# Skip-gram

Mikolov et al., 2013

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# Skip-gram

Mikolov et al., 2013



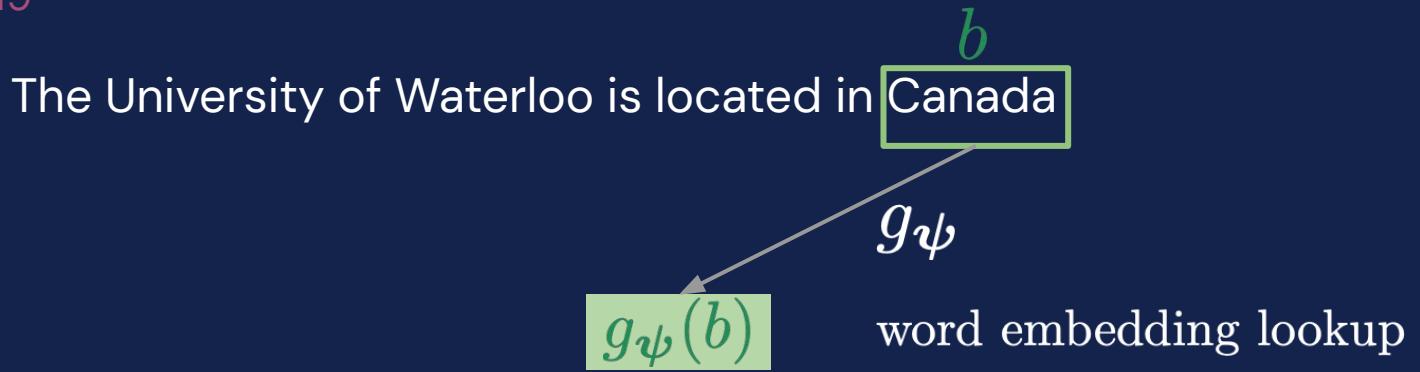
# BERT

Devlin et al., 2019

The University of Waterloo is located in Canada

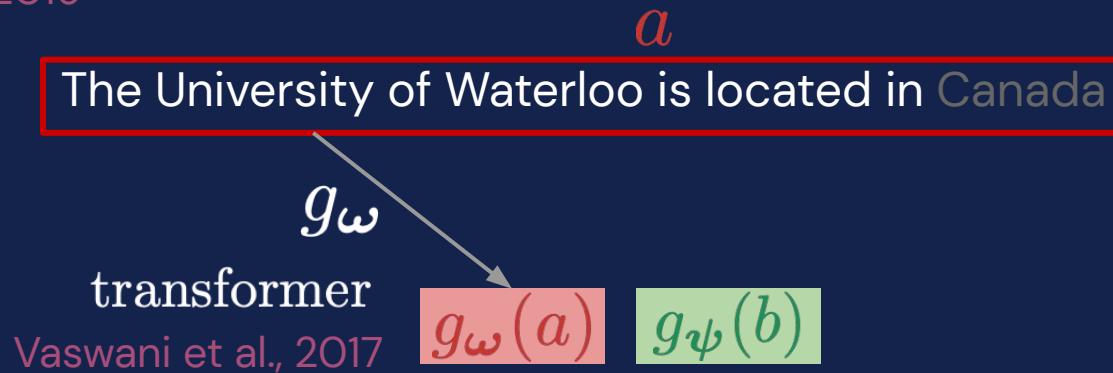
# BERT

Devlin et al., 2019



# BERT

Devlin et al., 2019



# Why is this interesting?

- A framework that unifies classical and modern word embedding methods.

		$a$	$b$	$g_{\omega}$	$g_{\psi}$
Mikolov et al., 2013	<b>Skip-gram</b>	word	word	lookup	lookup
Devlin et al., 2019	<b>BERT</b>	context	word	transformer	lookup
Yang et al., 2019	<b>XLNet</b>	context	word	TXL++	lookup

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- Provides connections to methods used in other domains (vision, speech).
- Facilitates exchanges of ideas on how to improve representation learning models.

# Model

Deep InfoMax (DIM; [Hjelm et al., 2019](#))



# Model

Deep InfoMax (DIM; Hjelm et al., 2019)



$$g_{\omega} \rightarrow g_{\omega}(a)$$

# Model

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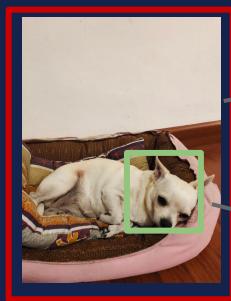


$$g_{\omega} \rightarrow g_{\omega}(a)$$

$$g_{\psi} \rightarrow g_{\psi}(b)$$

# Model

Deep InfoMax (DIM; Hjelm et al., 2019)

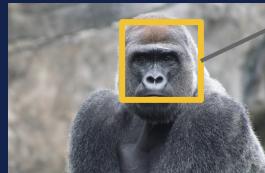


$$g_{\omega} \rightarrow g_{\omega}(a)$$

$$g_{\psi} \rightarrow g_{\psi}(b)$$



$$g_{\psi} \rightarrow g_{\psi}(c_1)$$



$$g_{\psi} \rightarrow g_{\psi}(c_2)$$

# Model

Deep InfoMax (DIM; Hjelm et al., 2019)



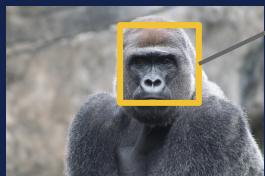
$$g_{\omega} \rightarrow g_{\omega}(a)$$

$$g_{\psi} \rightarrow g_{\psi}(b)$$



$$g_{\psi} \rightarrow g_{\psi}(c_1)$$

$$g_{\psi} \rightarrow g_{\psi}(c_2)$$



$$\mathcal{I}_{\text{DIM}} = \mathbb{E}_{p(A,B)} \left[ \mathbb{E}_{p(C)} \left[ \log \frac{\exp[g_{\omega}(a)^\top g_{\psi}(b)]}{\exp[g_{\omega}(a)^\top g_{\psi}(b)] + \sum_{c \neq b} \exp[g_{\omega}(a)^\top g_{\psi}(c)]} \right] \right]$$

# Model

Deep InfoMax (DIM; Hjelm et al., 2019)

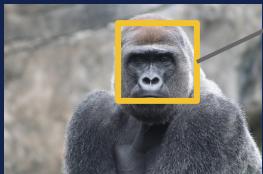


$$g_{\omega} \rightarrow g_{\omega}(a)$$

$$g_{\psi} \rightarrow g_{\psi}(b)$$



$$g_{\psi} \rightarrow g_{\psi}(c_1)$$

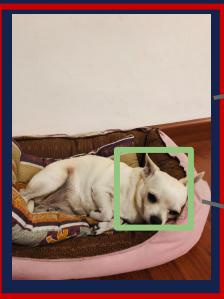


$$g_{\psi} \rightarrow g_{\psi}(c_2)$$

UWaterloo is located in Canada

# Model

Deep InfoMax (DIM; Hjelm et al., 2019)



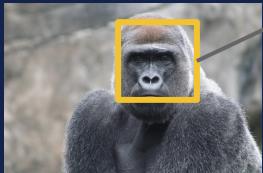
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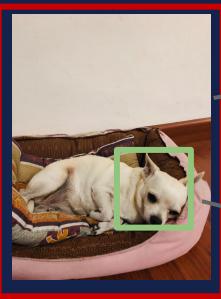


UWaterloo is located in Canada

$$\begin{array}{l} g_{\psi} \\ \text{transformer} \\ \downarrow \\ g_{\psi}(b) \end{array}$$

# Model

Deep InfoMax (DIM; Hjelm et al., 2019)



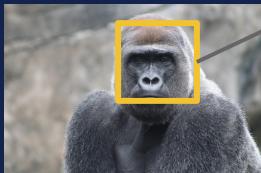
$$g_{\omega} \rightarrow g_{\omega}(a)$$

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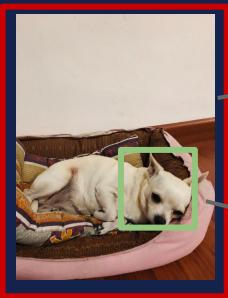
UWaterloo is located in Canada

$g_{\omega}$   
transformer

$$g_{\omega}(a) \quad g_{\psi}(b)$$

# Model

Deep InfoMax (DIM; Hjelm et al., 2019)



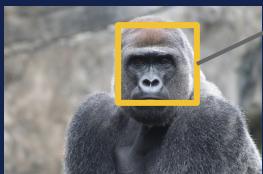
$$g_{\omega} \rightarrow g_{\omega}(a)$$

$$g_{\psi} \rightarrow g_{\psi}(b)$$



$$g_{\psi} \rightarrow g_{\psi}(c_1)$$

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UWaterloo is located in Canada

$$g_{\omega}(a) \quad g_{\psi}(b)$$

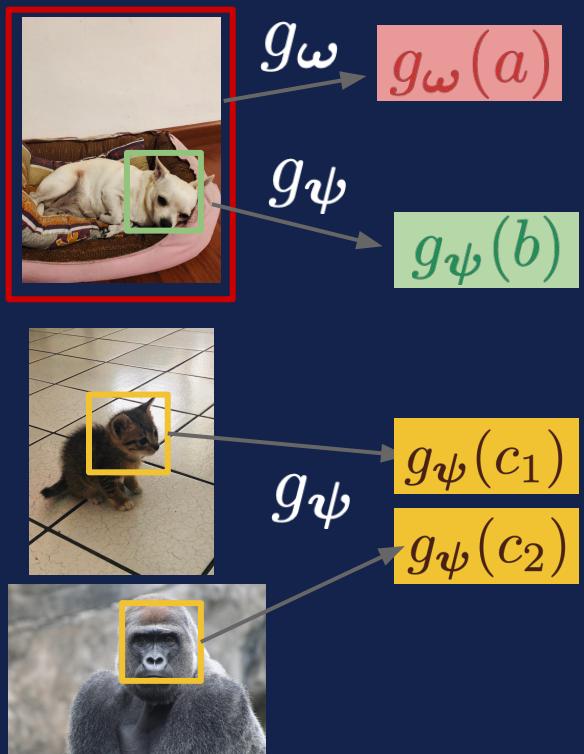
Starcraft II is a fun game

Cristiano Ronaldo scores an own goal

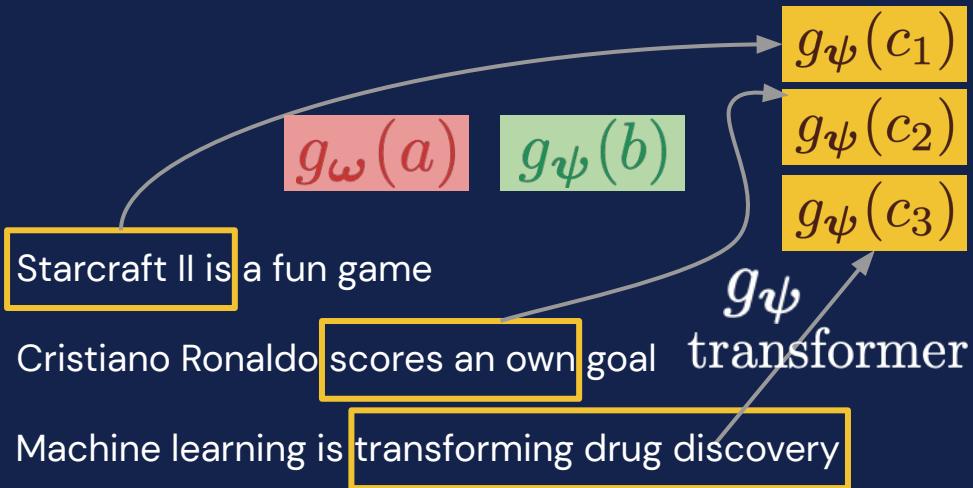
Machine learning is transforming drug discovery

# Model

Deep InfoMax (DIM; Hjelm et al., 2019)

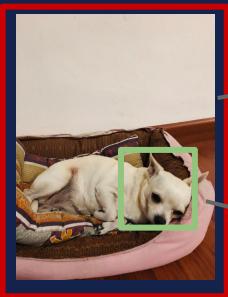


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# Model

Deep InfoMax (DIM; Hjelm et al., 2019)



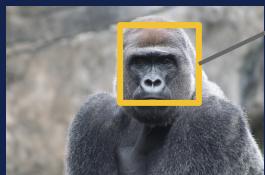
$$g_{\omega} \rightarrow g_{\omega}(a)$$

$$g_{\psi} \rightarrow g_{\psi}(b)$$



$$g_{\psi} \rightarrow g_{\psi}(c_1)$$

$$g_{\psi} \rightarrow g_{\psi}(c_2)$$



$$\mathcal{I}_{\text{DIM}} = \mathbb{E}_{p(A,B)} \left[ \mathbb{E}_{p(C)} \left[ \log \frac{\exp[g_{\omega}(a)^\top g_{\psi}(b)]}{\exp[g_{\omega}(a)^\top g_{\psi}(b)] + \sum_{c \neq b} \exp[g_{\omega}(a)^\top g_{\psi}(c)]} \right] \right]$$

UWaterloo is located in Canada

$$g_{\psi}(c_1)$$

$$g_{\psi}(c_2)$$

$$g_{\psi}(c_3)$$

$$g_{\omega}(a) \quad g_{\psi}(b)$$

Starcraft II is a fun game

Cristiano Ronaldo scores an own goal

Machine learning is transforming drug discovery

# Experiments

Question answering on SQuAD ([Rajpurkar et al., 2016](#)).

		F1
Small Model	BERT	90.9
	Ours	<b>91.4</b>
Large Model	BERT	92.7
	Ours	<b>93.1</b>

F1 scores (0-100), higher is better.

BERT: [Devlin et al., 2019](#).

# Takeaways

- It is possible to transfer ideas across domains when designing self-supervised tasks.
- Progress in language representation learning has largely been driven by advances in model architectures (and training objectives).

# Adaptive Semiparametric Language Models

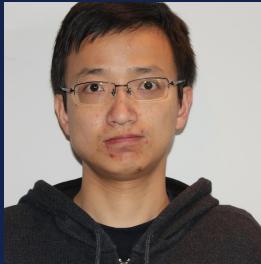
Yogatama et al., TACL 2021



Dani



Cyprien



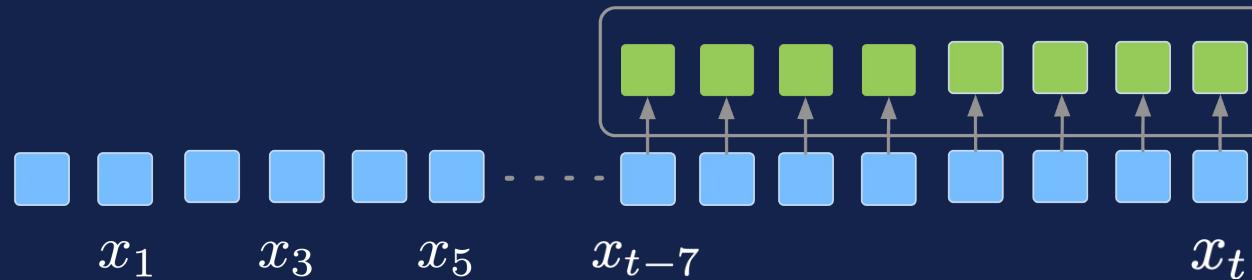
Lingpeng

# Background

What are core limitations of existing architectures?

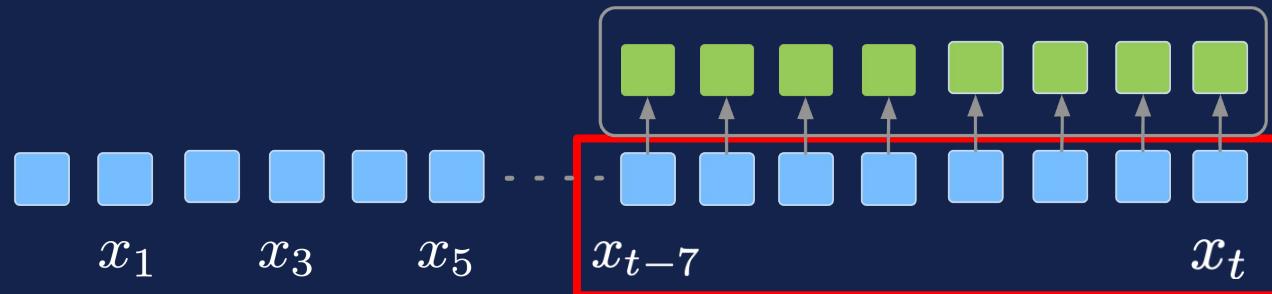
# Background

State of the art architectures (transformers) are limited by the input sequence length.



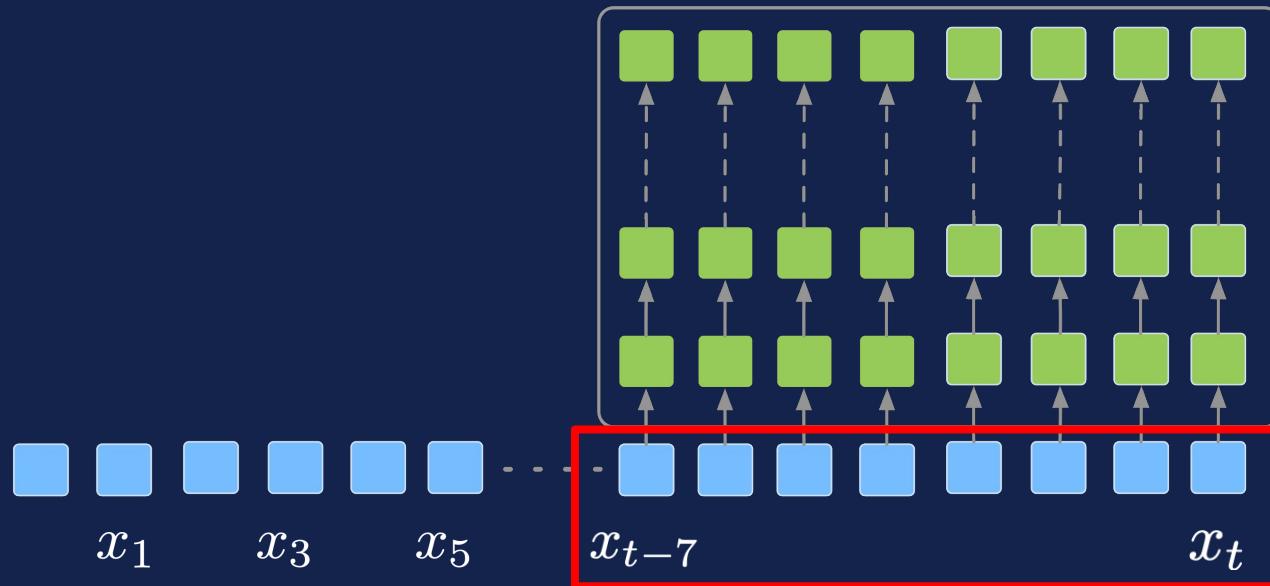
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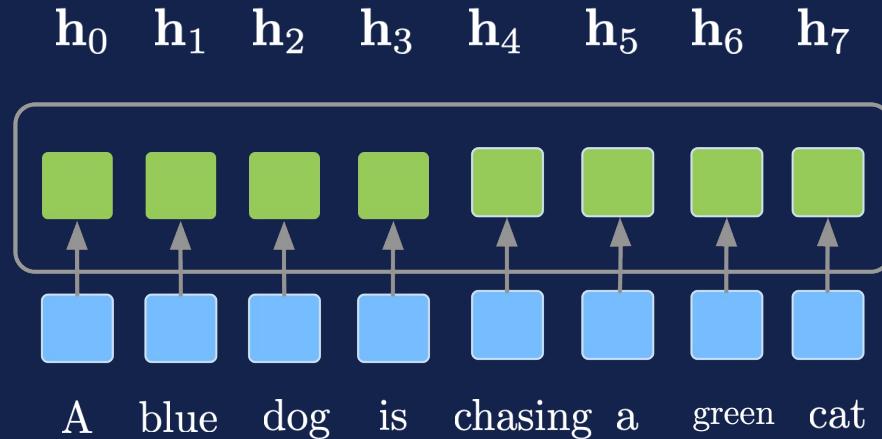


# Background

Knowledge is encoded in the weights of a parametric neural network.

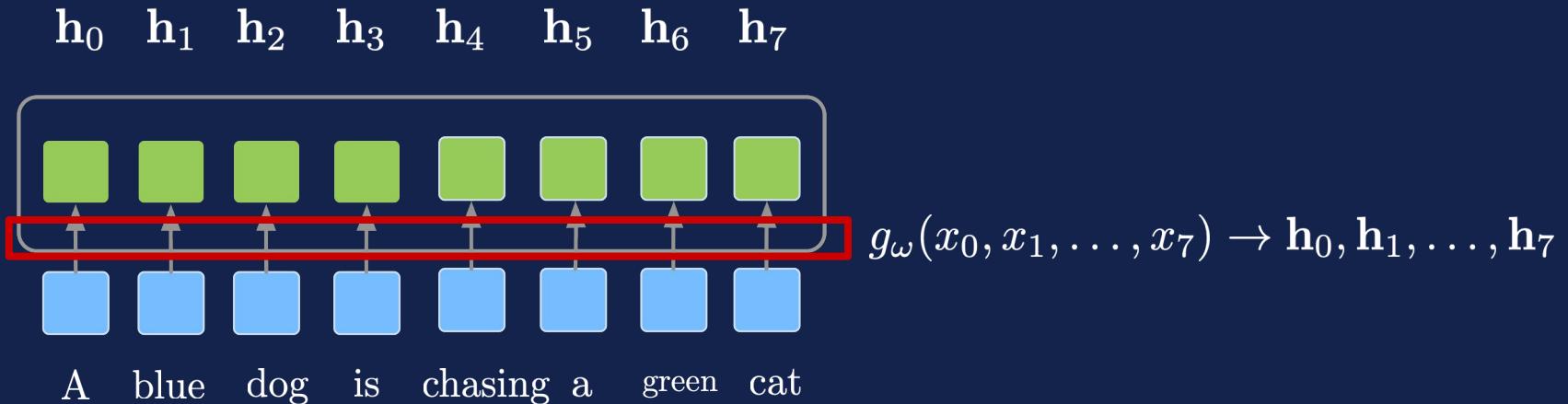
# Background

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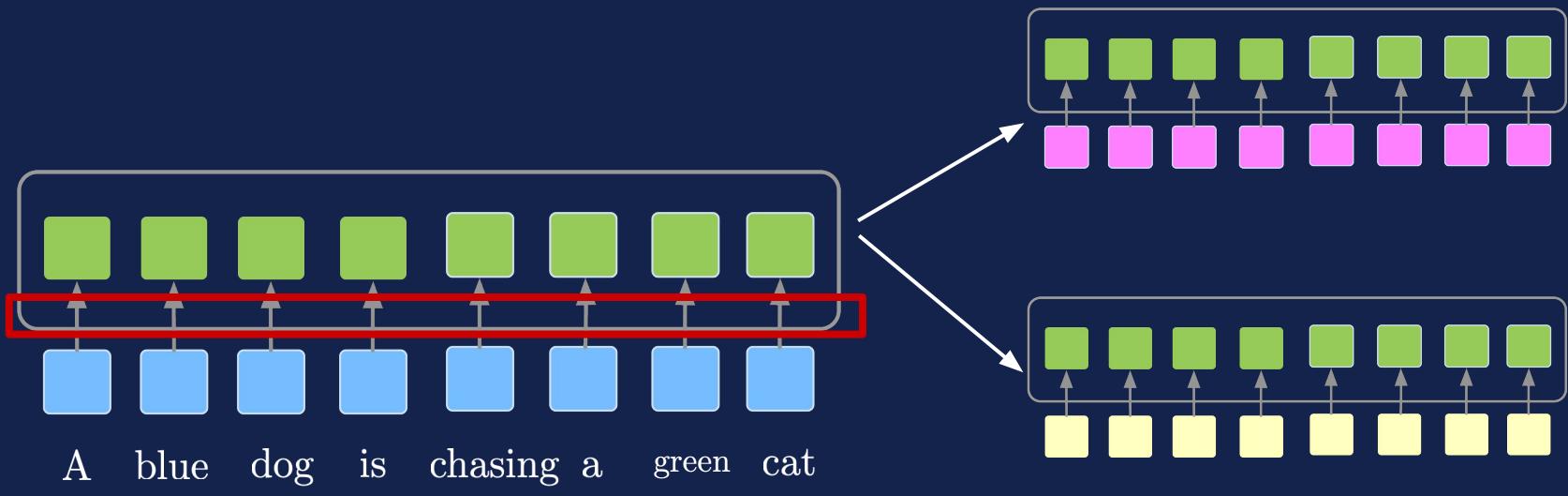
# Background

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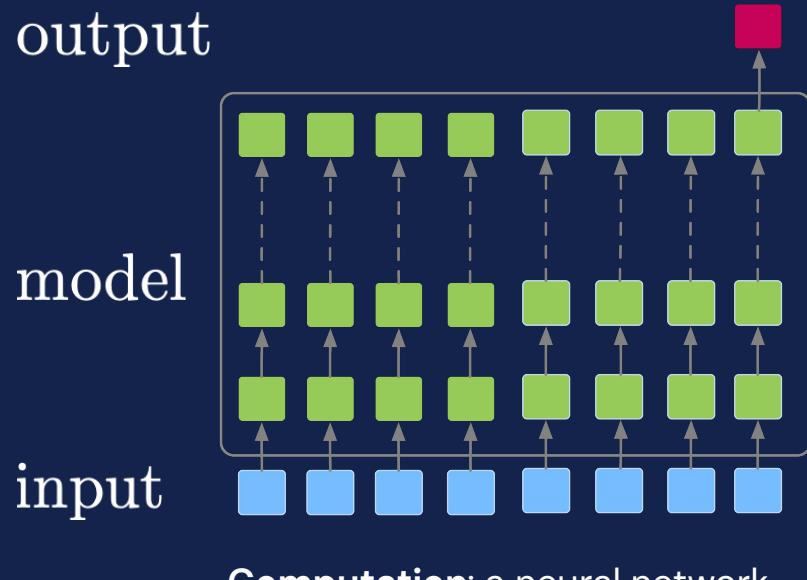
Update weights with new knowledge → changes affect all examples (sequences).

# Semiparametric Models

Separation of computation and storage as an architectural bias.

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Separation of computation and storage as an architectural bias.



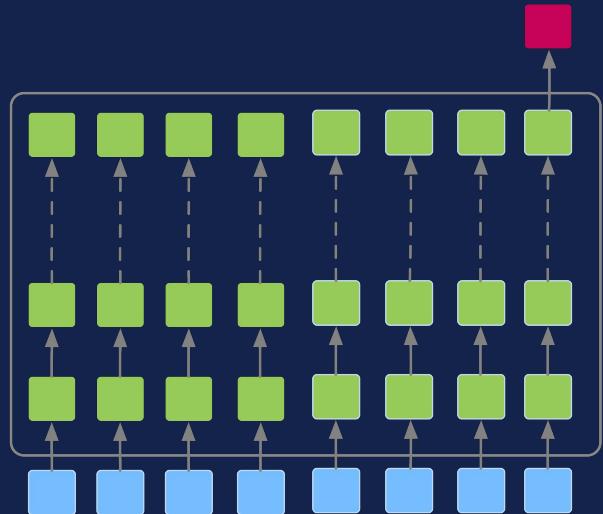
# Semiparametric Models

Separation of computation and storage as an architectural bias.

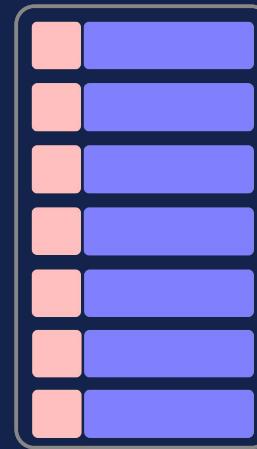
output

model

input

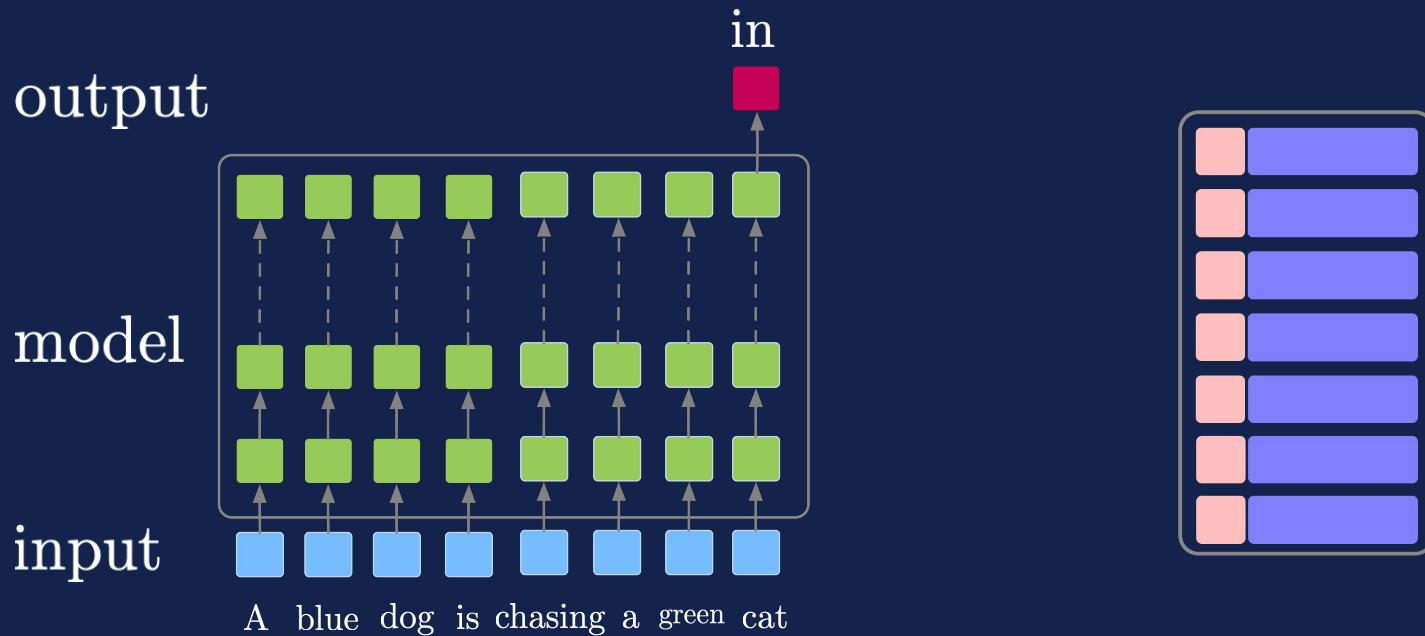


**Computation:** a neural network



**Memory (storage):** a key-value database

# Semiparametric Language Models



# Problem Setup

## University of Waterloo Wikipedia

The University of Waterloo (commonly referred to as Waterloo, UW, or UWaterloo) is a public research university with a main campus in Waterloo, Ontario, Canada. The main campus is on 404 hectares of land adjacent to **Uptown**

# Problem Setup

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# Language Model

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Current context

(computation)

# Language Model

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Current context  
(computation)

Extended context  
(short-term memory)

# Language Model

University of Waterloo Wikipedia

The University of Waterloo (commonly referred to as Waterloo, UW, or UWaterloo) is a public research university with a main campus in Waterloo, Ontario, Canada. The main campus is on 404 hectares of land adjacent to Uptown Waterloo and Waterloo ???

Current context  
(computation)

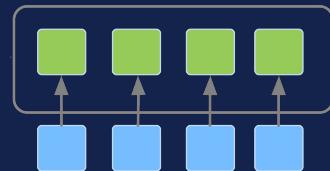
Extended context  
(short-term memory)

Long-term memory

Waterloo Park Wikipedia

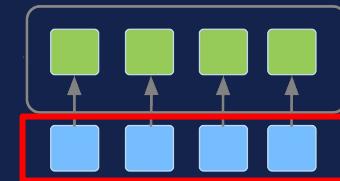
Waterloo Park is an urban park situated in Waterloo, Ontario, Canada.

# Language Model



UWaterloo is a public

# Language Model



UWaterloo is a public

**Input:** a sequence of tokens.

# Language Model

**Encoder:** transformer  
(Vaswani et al., 2017)



UWaterloo is a public

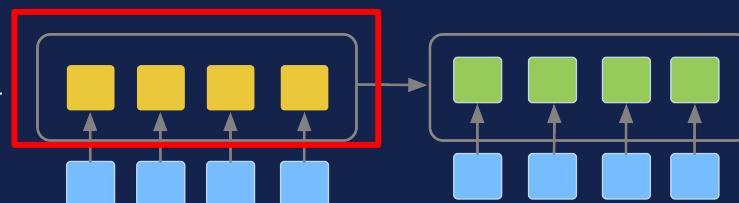
encoder  
(computation)

# Language Model

**Short-term memory:**

transformer-XL (Dai et al., 2019)

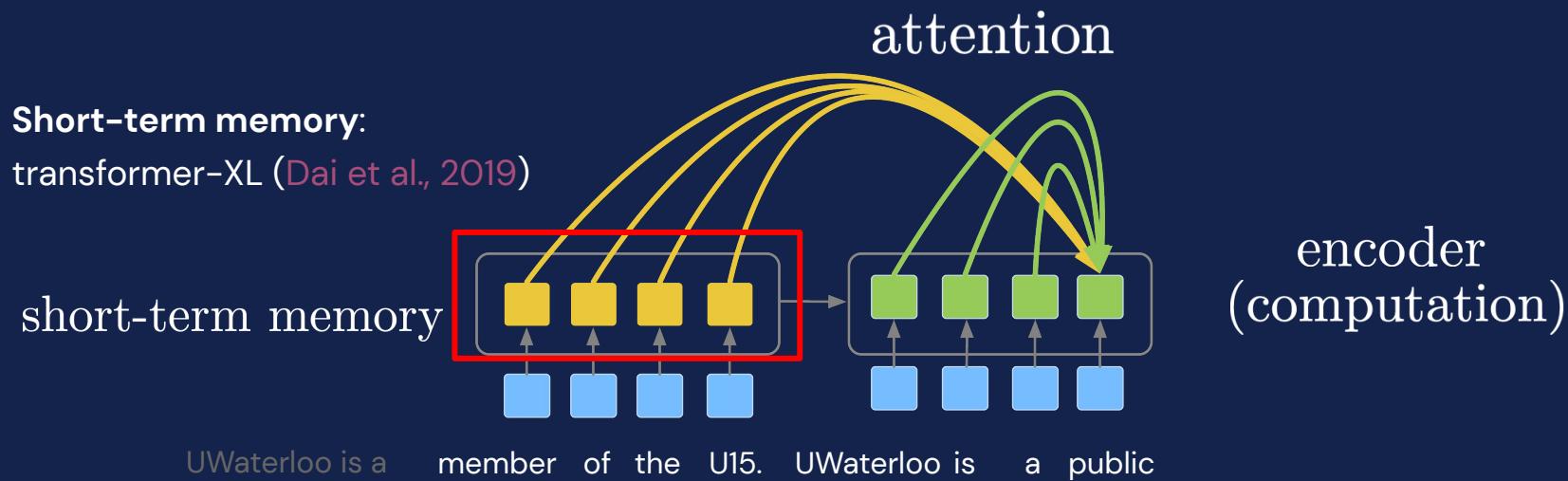
short-term memory



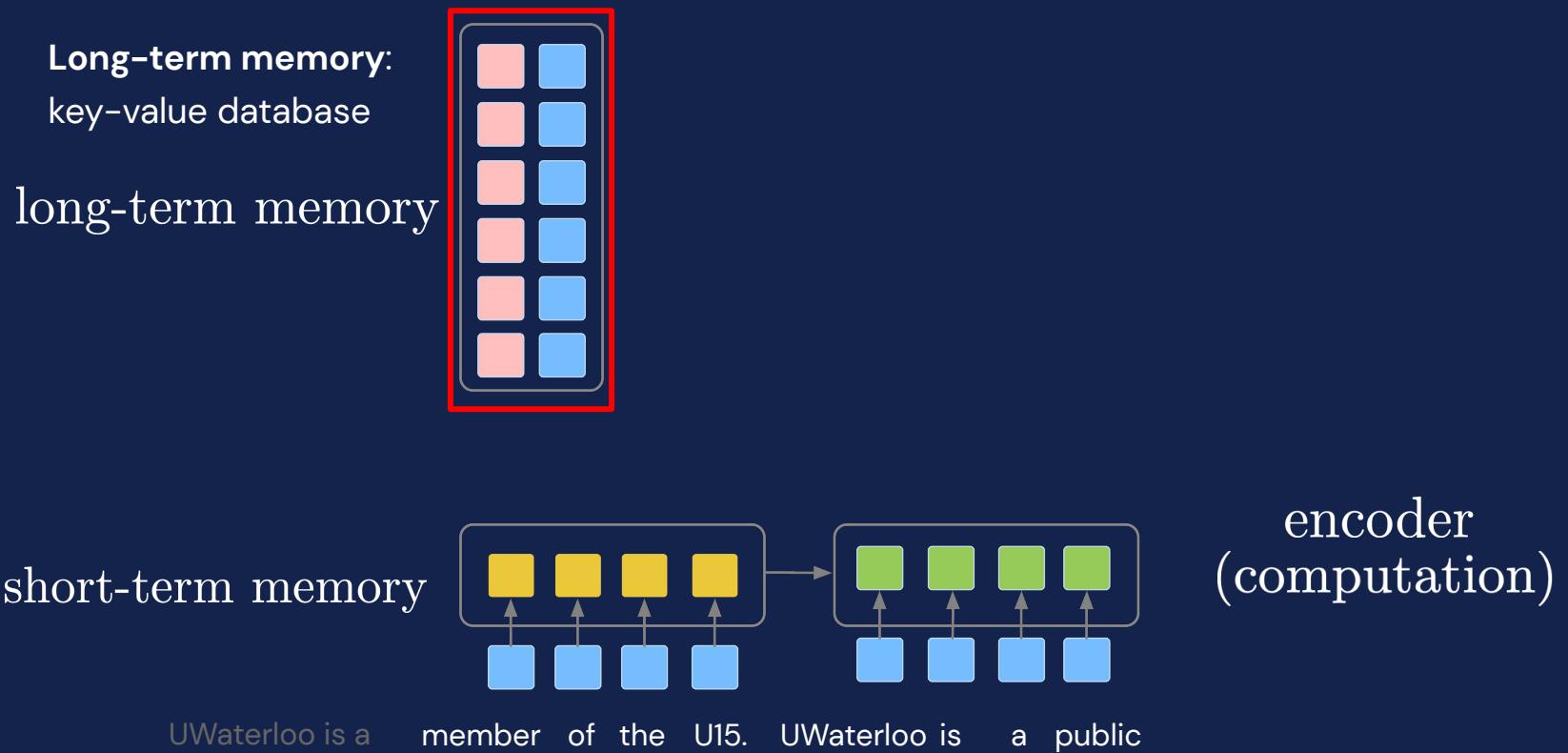
encoder  
(computation)

UWaterloo is a member of the U15. UWaterloo is a public

# Language Model

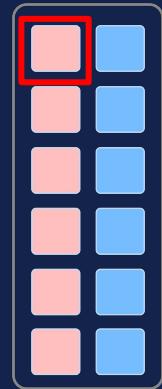


# Language Model



# Language Model

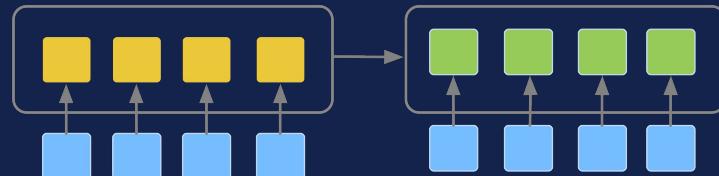
long-term memory



**Key:** compressed long-term context

Canada is a beautiful

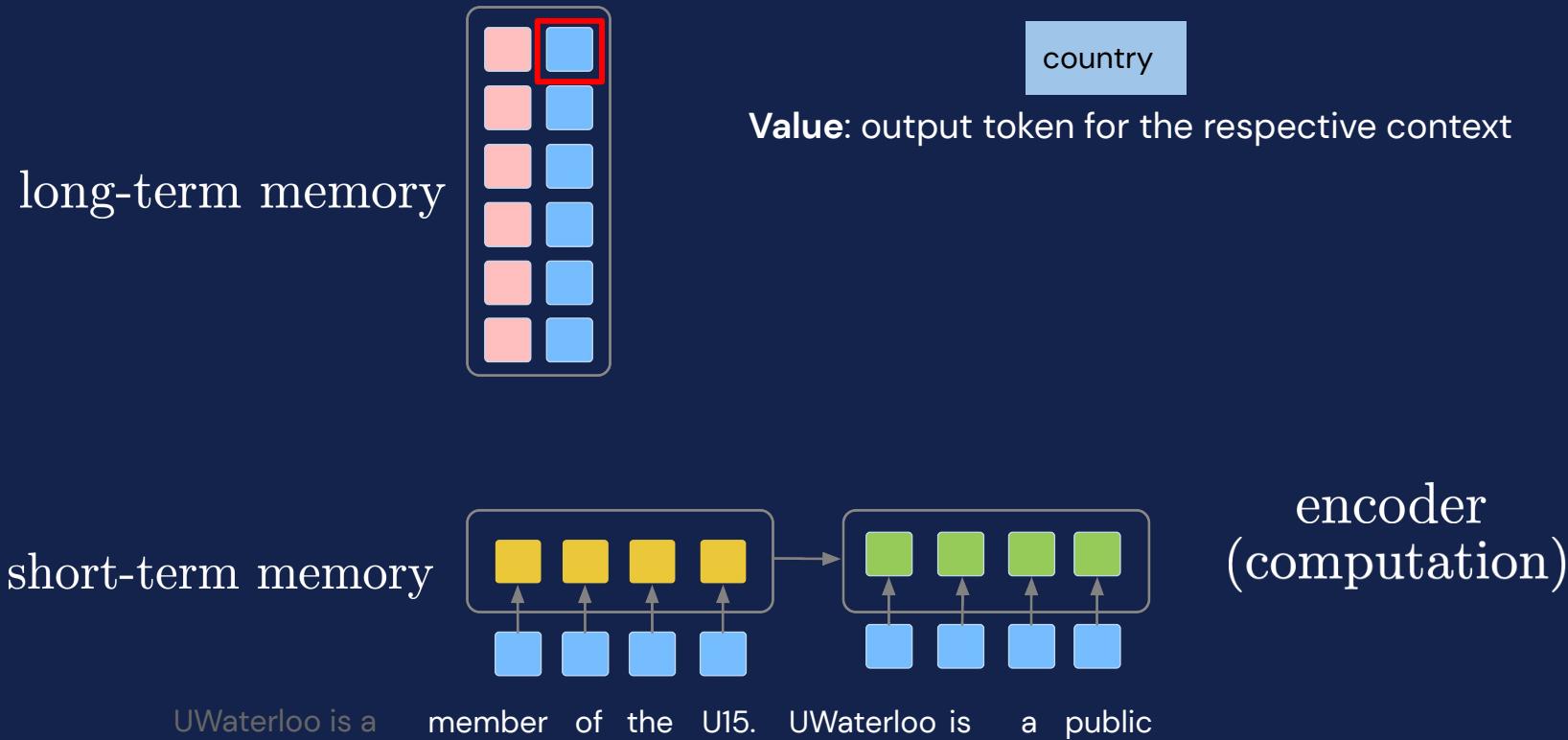
short-term memory



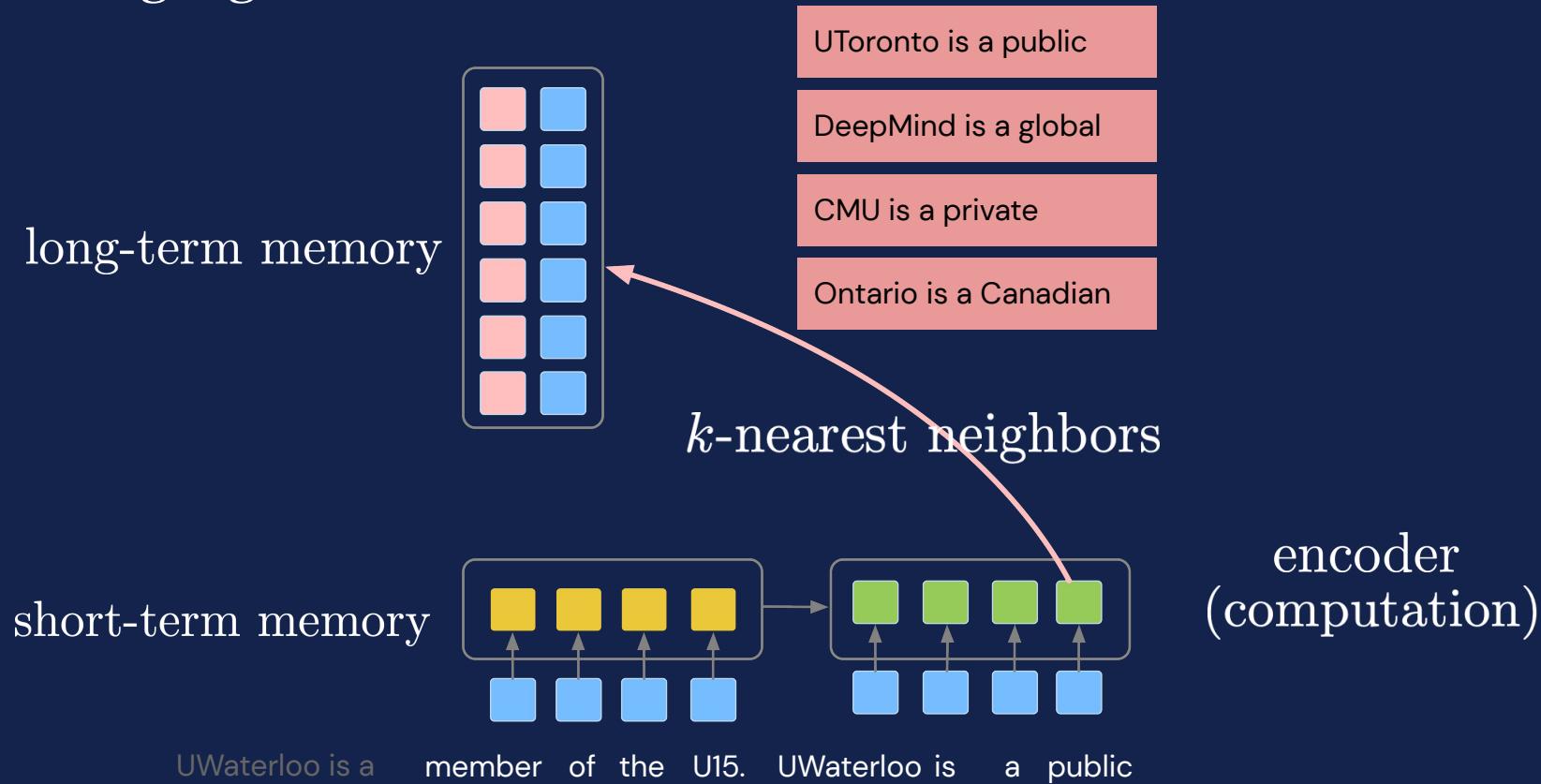
UWaterloo is a member of the U15. UWaterloo is a public

encoder  
(computation)

# Language Model



# Language Model



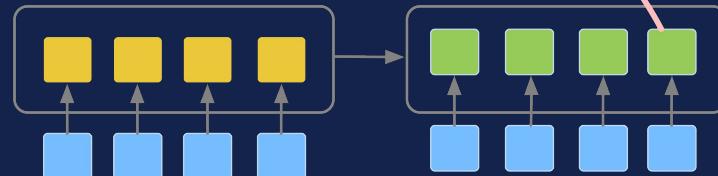
# Language Model

long-term memory

UToronto is a public	research
DeepMind is a global	research
CMU is a private	university
Ontario is a Canadian	province

short-term memory

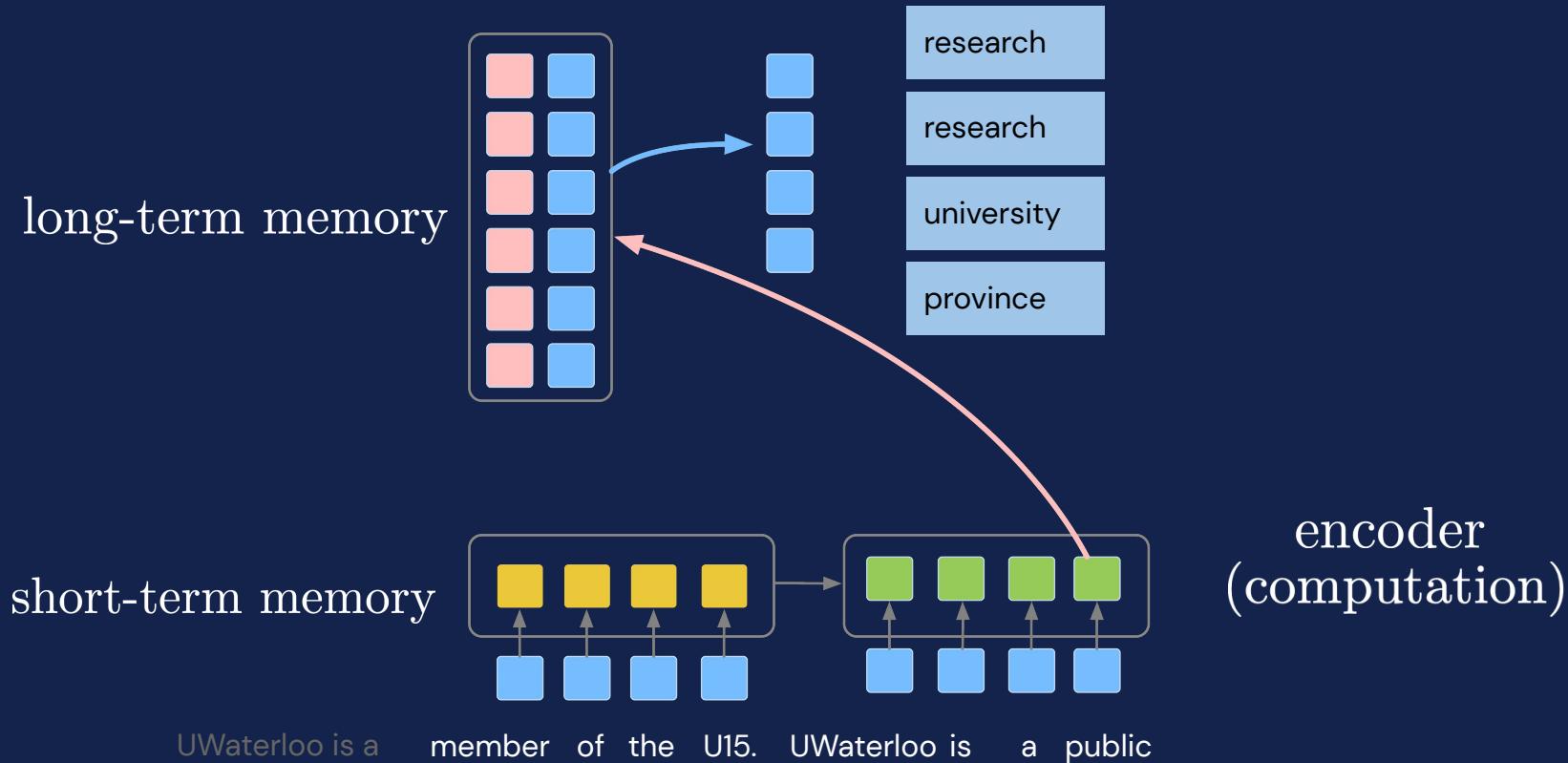
UWaterloo is a member of the U15. UWaterloo is a public



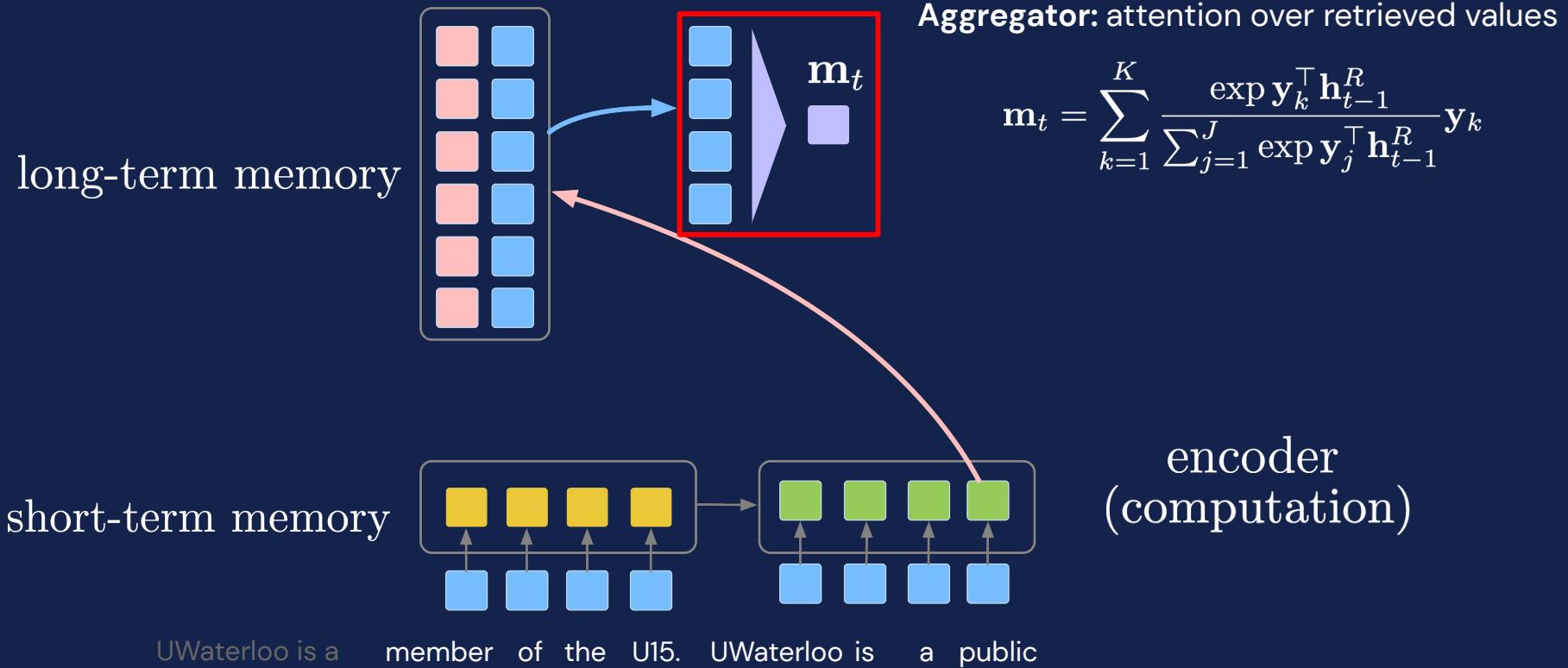
encoder  
(computation)

$k$ -nearest neighbors

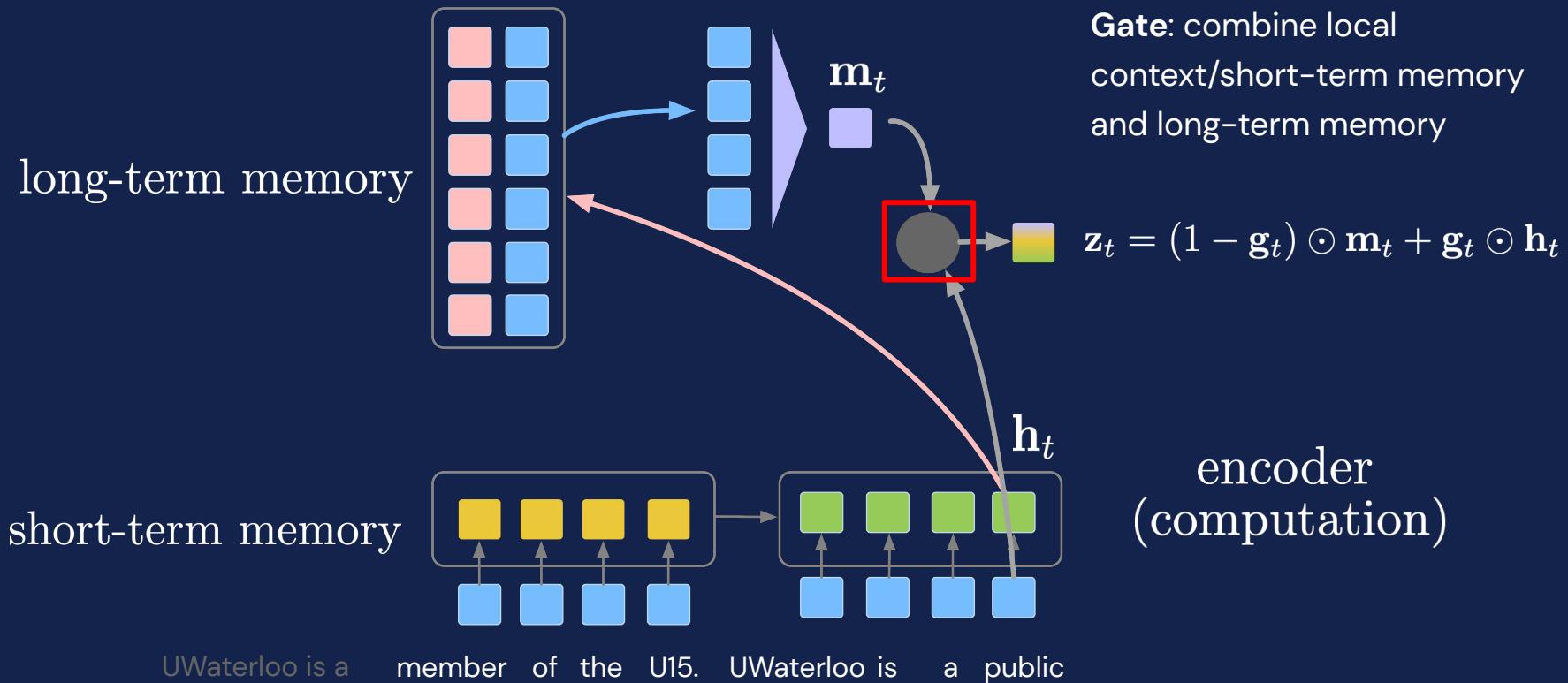
# Language Model



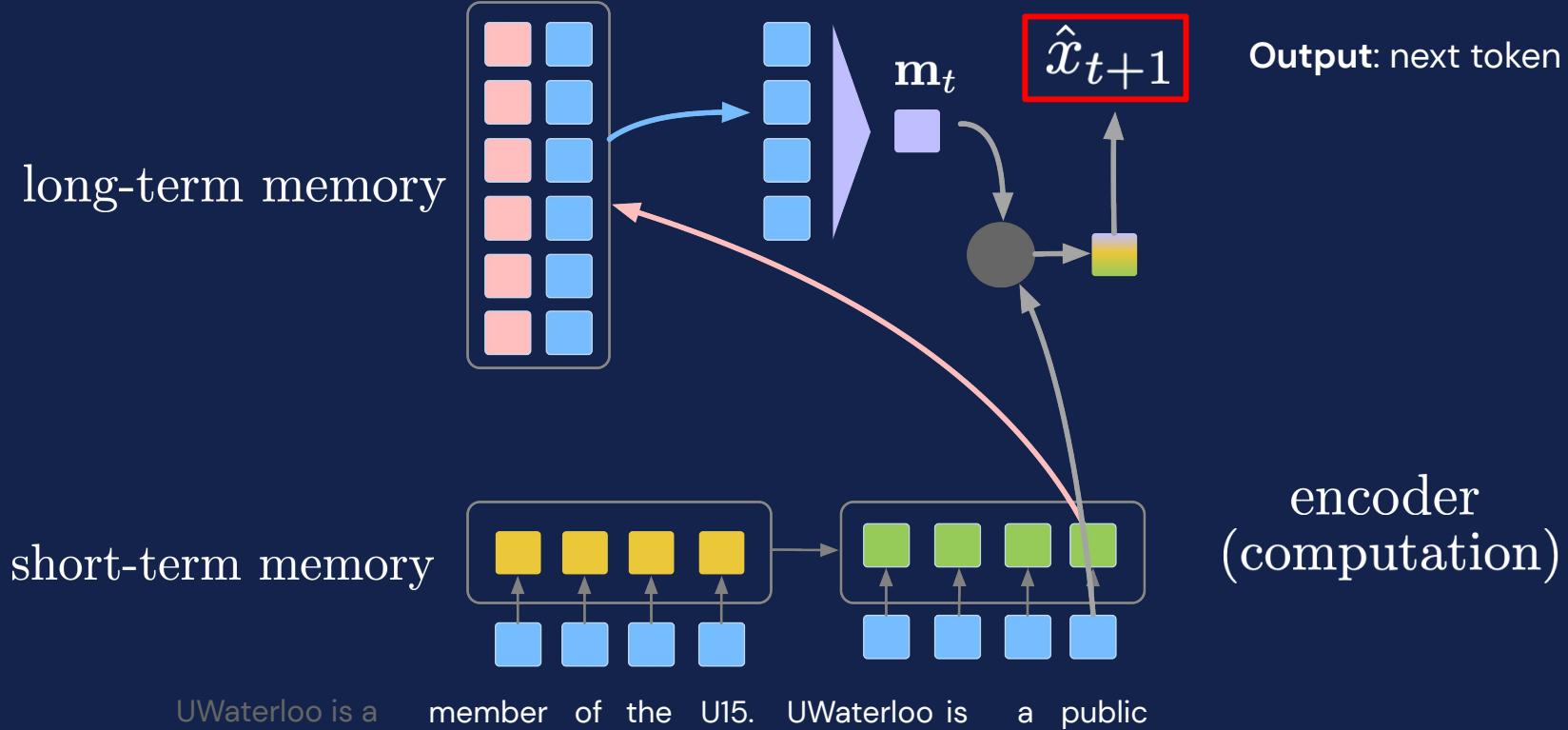
# Language Model



# Language Model



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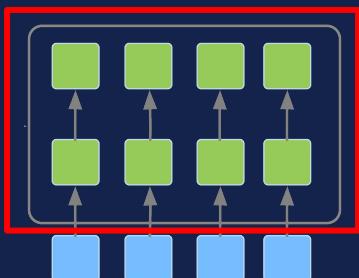


# Language Model



**Input:** a sequence of tokens.

# Language Model



**Encoder:** transformer  
(Vaswani et al., 2017)  
encoder  
(computation)

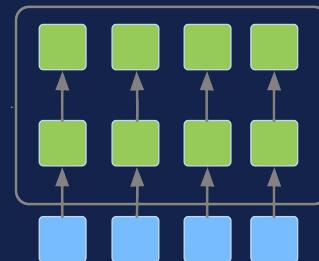
# Language Model

**Short-term memory:**

transformer-XL (Dai et al., 2019)

**Encoder:** transformer  
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encoder  
(computation)



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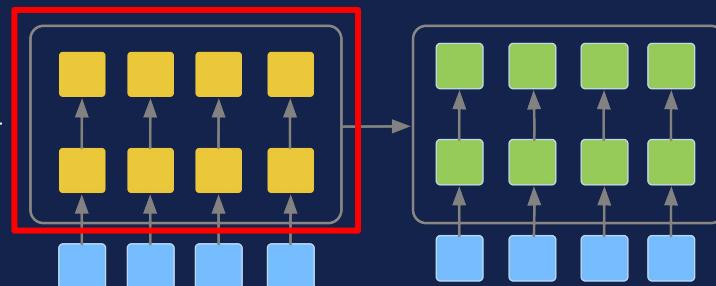
UWaterloo is a public

# Language Model

**Short-term memory:**

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short-term memory

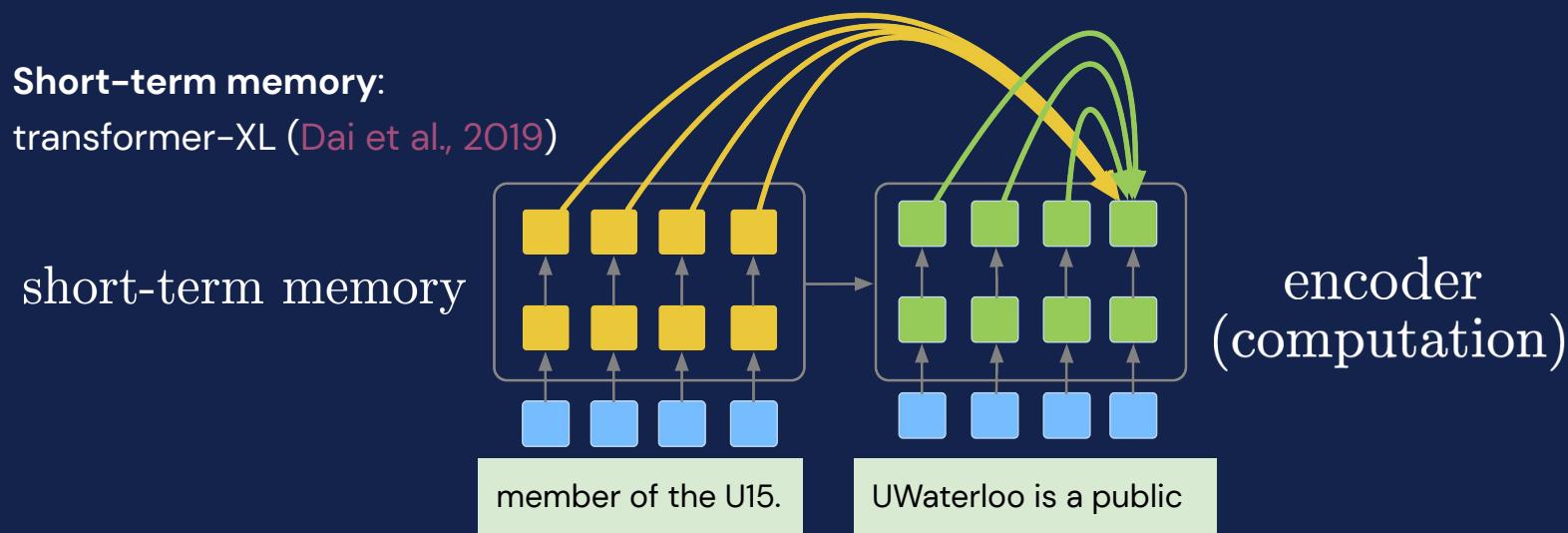


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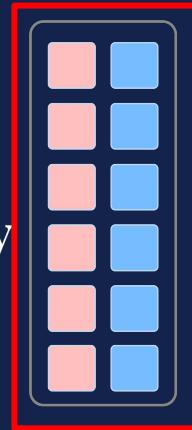
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# Language Model

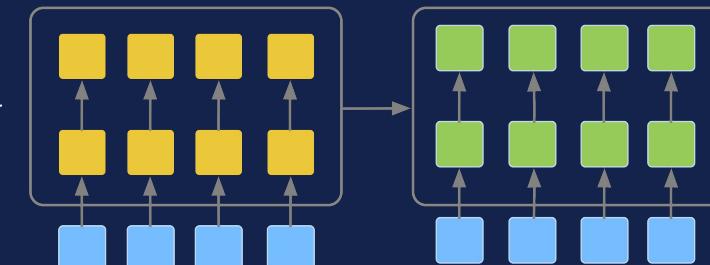


# Language Model

**Long-term memory:**  
key-value database  
  
long-term memory



short-term memory



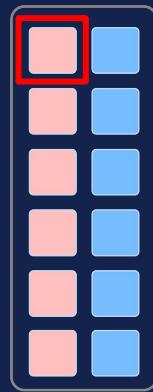
member of the U15.

UWaterloo is a public

encoder  
(computation)

# Language Model

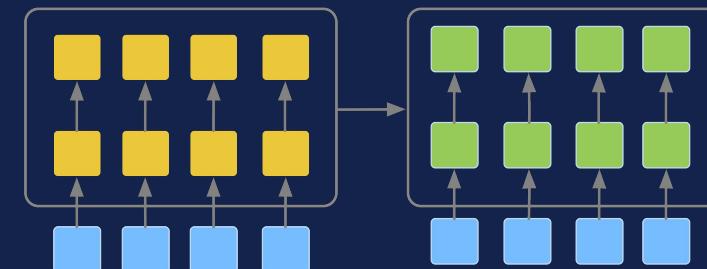
long-term memory



**Key:** compressed long-term context

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short-term memory



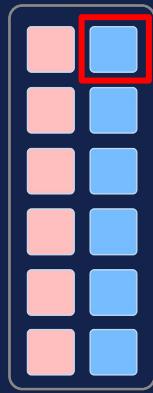
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# Language Model

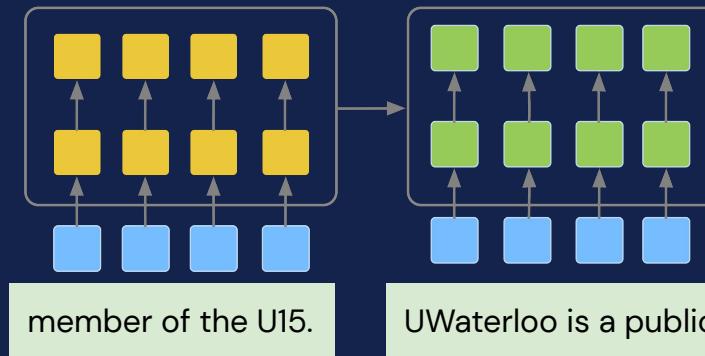
long-term memory



country

**Value:** output token for the respective context

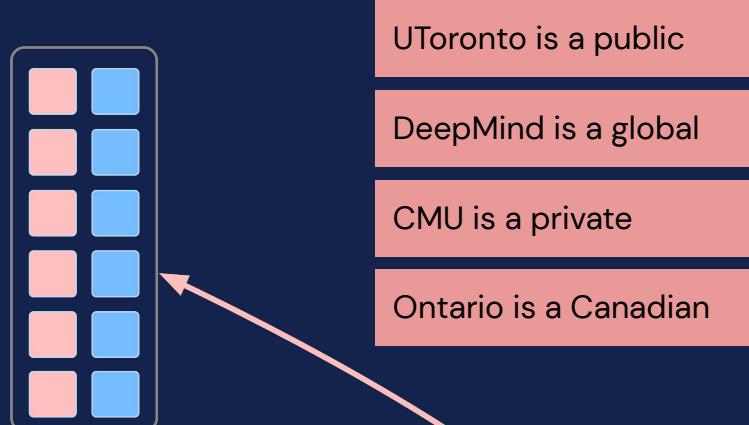
short-term memory



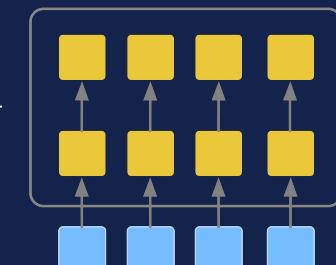
encoder  
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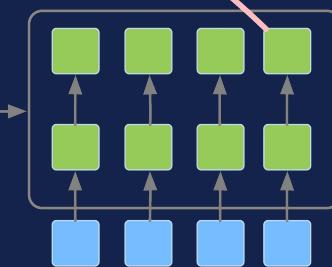
long-term memory



short-term memory



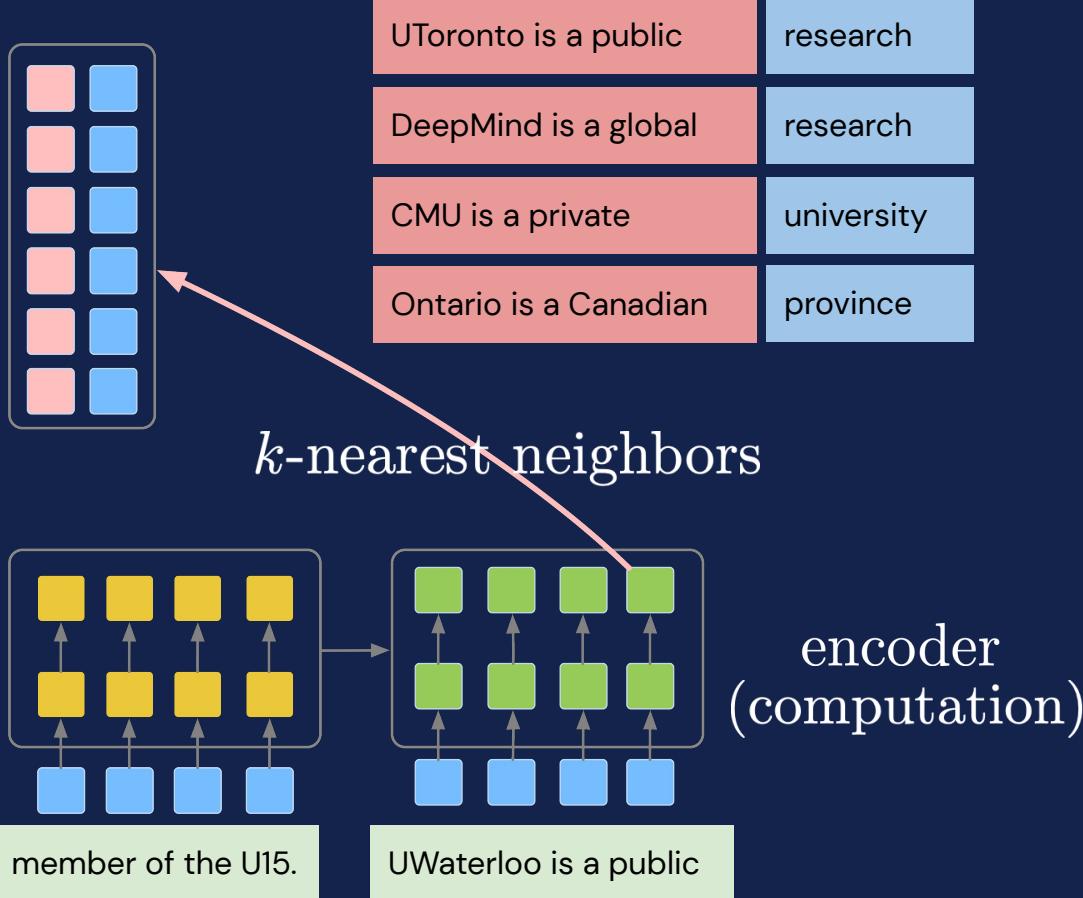
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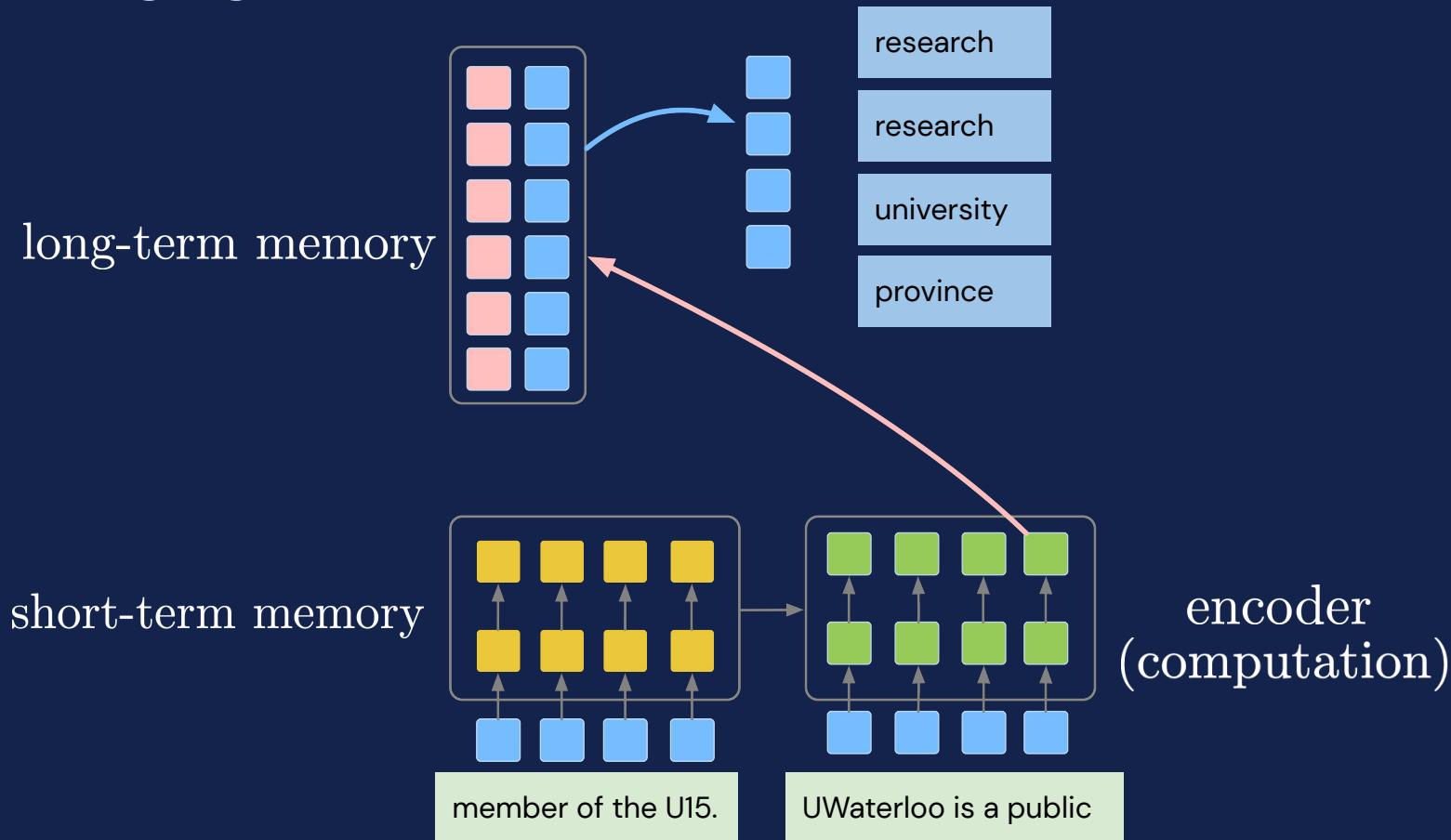
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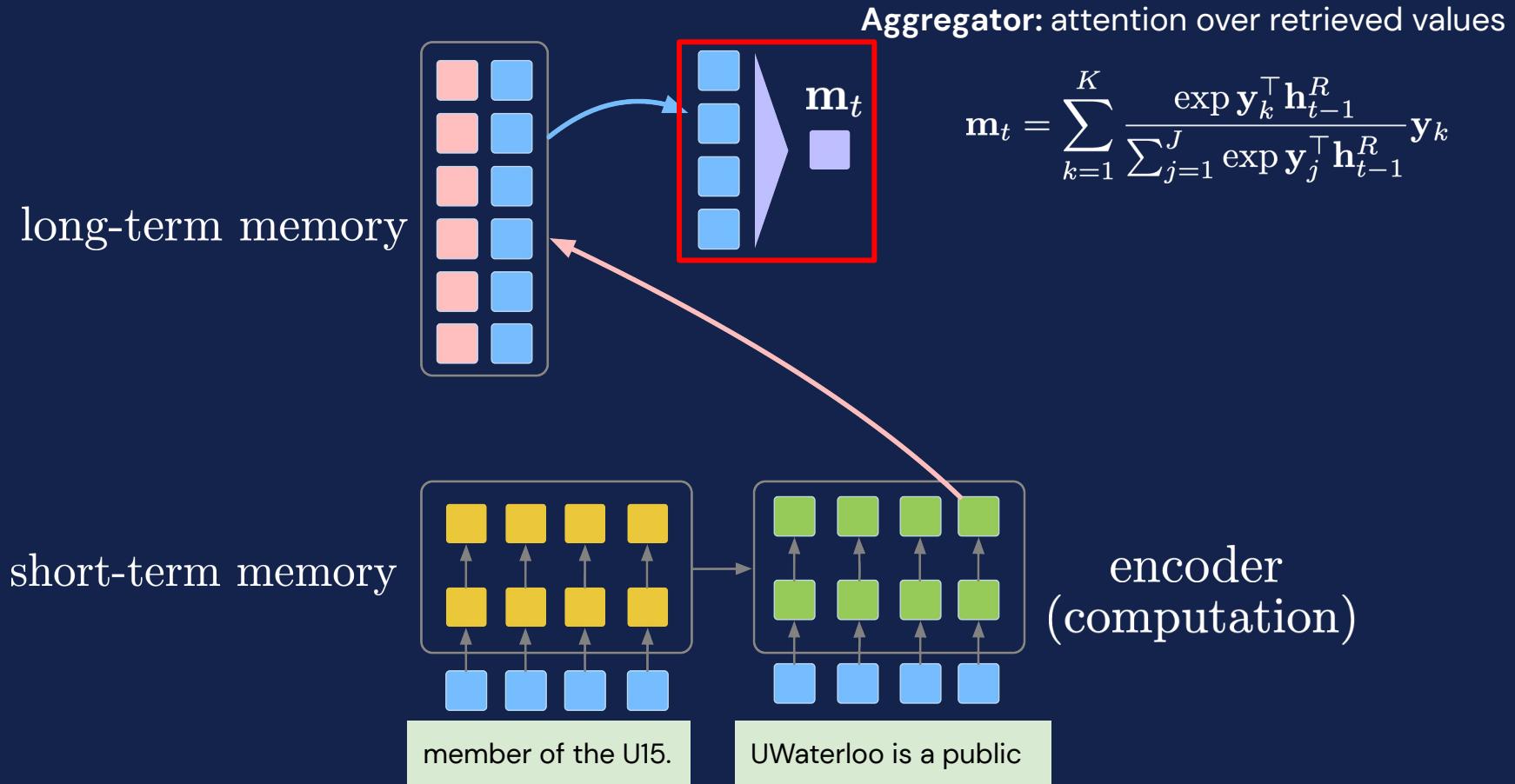
long-term memory



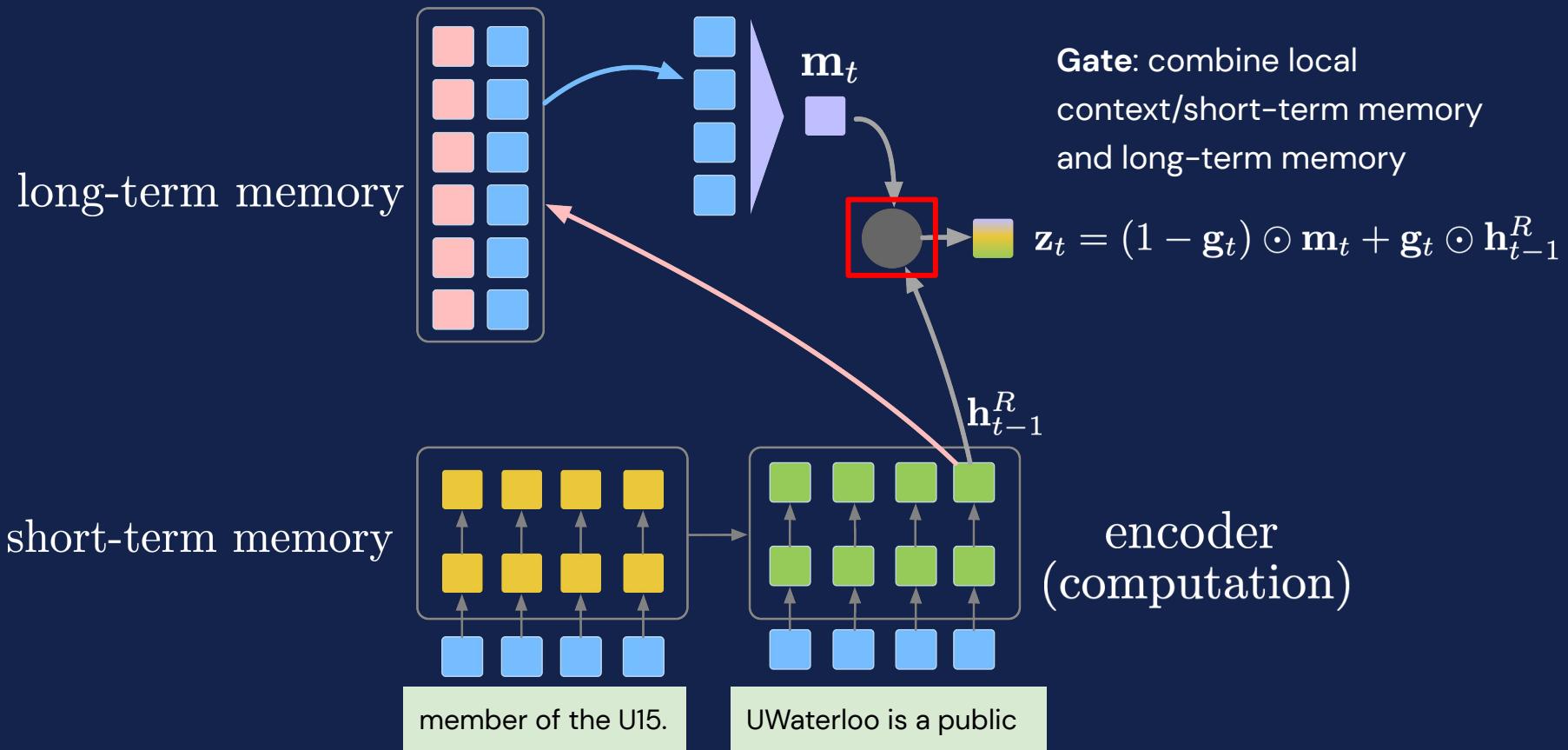
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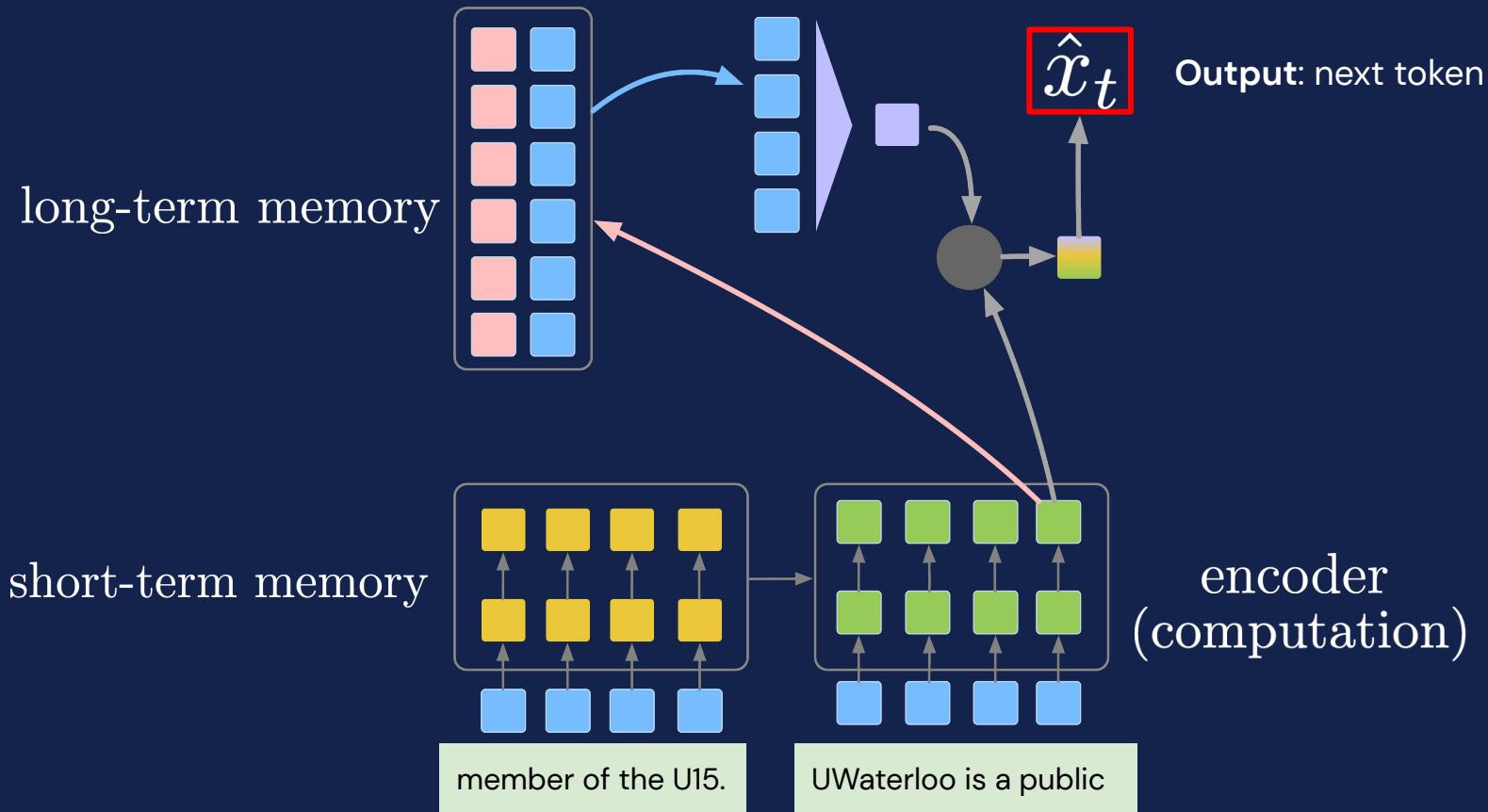
# Language Model



# Language Model



# Language Model



# Experiments

- Word-level language modeling.
  - WikiText-103 English (Merity et al., 2016).
  - WMT 2019 English: <http://www.statmt.org/wmt19/>.
- Character-level language modeling.
  - enwik8: <http://prize.hutter1.net>.

# Experiments

Perplexity (1-inf), lower is better

	Base	TXL	kNN-LM	Ours
WikiText-103	21.8	19.1	18.0	<b>17.6*</b>
WMT	16.5	15.5	15.2	<b>14.1</b>

Transformer: Vaswani et al., 2017

Transformer-XL: Dai et al., 2019

kNN-LM: Khandelwal et al., 2020

# Experiments

BPC (0-inf), lower is better

	Base	TXL	kNN-LM	Ours
enwik8	1.05	1.01	1.02	<b>1.00</b>

Transformer: Vaswani et al., 2017

Transformer-XL: Dai et al., 2019

kNN-LM: Khandelwal et al., 2020

# Analysis

Liberal Democrat leader Jo Swinson has said she would work with Donald Trump in government as

# Analysis

What's in the long-term memory?

Elizabeth Warren on Friday proposed \$20 trillion in spending over the next decade to provide health care for every American without raising taxes on the middle class.

# Analysis

What's in the long-term memory?

For

Perhaps  
Like  
Elizabeth Warren

on Friday proposed \$ 20 trillion in

spending over the next decade to provide health care

every American without raising taxes on the middle class

# Analysis

What's in the long-term memory?

For Warren  
Warren  
Perhaps Warren  
Like Warren  
Elizabeth Warren on Friday proposed \$20 trillion in

spending over the next decade to provide health care

every American without raising taxes on the middle class

# Analysis

What's in the long-term memory?

For Warren &  
Perhaps Warren may  
Like Warren has  
Elizabeth Warren ,  
spending over the next decade to provide health care  
every American without raising taxes on the middle class

# Analysis

What's in the long-term memory?

For Warren & Wednesday  
Warren may Tuesday  
Perhaps Warren has Sunday  
Like Warren , Monday  
Elizabeth Warren on Friday proposed \$ 20 trillion in

spending over the next decade to provide health care

every American without raising taxes on the middle class

# Analysis

What's in the long-term memory?

For Warren & Wednesday briefly a 5 billion to  
Warren may Tuesday praised wiping 16 trillion in  
Perhaps Warren has Sunday stood breaking 10 billion for  
Like Warren , Monday defended using 166 trillion in  
Elizabeth Warren on Friday proposed \$ 20 trillion in

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What's in the long-term memory?

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grants in 10 course eight . fight even care  
funding over the next three . upgrade them cover  
funds over 10 next five in improve American -  
, over a next 10 , invest a insur.  
spending over the next decade to provide health care

every American without raising taxes on the middle class

# Analysis

What's in the long-term memory?

For	Warren	&	Wednesday	briefly	a	5	billion	to
	Warren	may	Tuesday	praised	wiping	16	trillion	in
Perhaps	Warren	has	Sunday	stood	breaking	10	billion	for
Like	Warren	,	Monday	defended	using	166	trillion	in
Elizabeth	Warren	on	Friday	proposed	\$	20	trillion	in
grants	in	10	course	eight	.	fight	even	care
funding	over	the	next	three	.	upgrade	them	cover
funds	over	10	next	five	in	improve	American	-
,	over	a	next	10	,	invest	a	insur.
spending	over	the	next	decade	to	provide	health	care
more	community	as	the	rates	.	the	middle	class
everyone	child	,	a	taxes	on	the	wealthy	class
some	baby	,	co	taxes	.	the	middle	class
every	American	by	triggering	taxes	on	all	middle	class
every	American	without	raising	taxes	on	the	middle	class

# Takeaways

- A language model that adaptively combines local context, short-term memory, and long-term memory.

# Takeaways

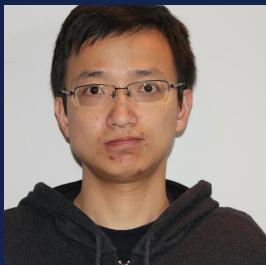
- A language model that adaptively combines local context, short-term memory, and long-term memory.
- A variant of the models for question answering (**de Masson d'Autume et al., NeurIPS 2019**)



Cyprien



Sebastian



Lingpeng



Dani

# Future Directions



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.

# Future Directions



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.

**Training Paradigms**

**Model Architectures**

# Future Directions



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.

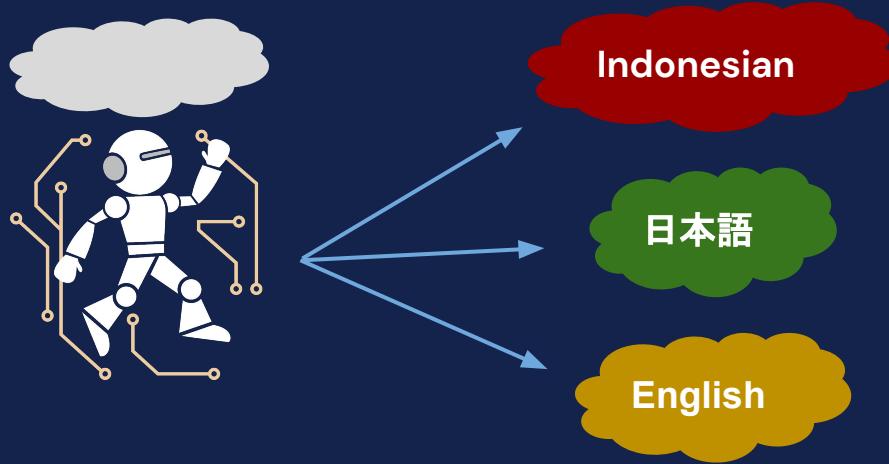
**Training Paradigms**

**Model Architectures**

# Future Directions



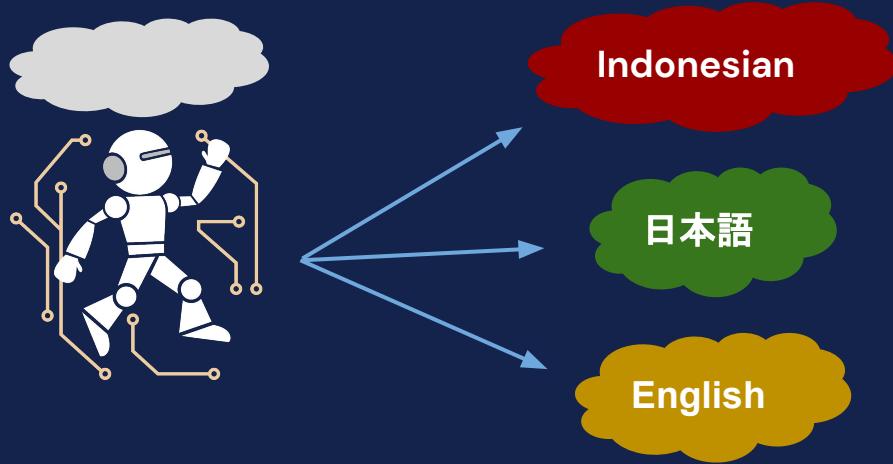
A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.



# Future Directions



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.



Learning cross-lingual  
transferable representations

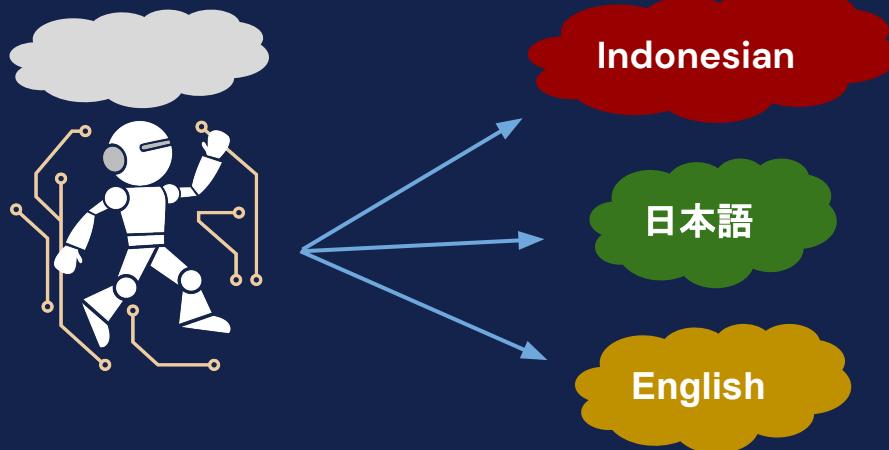
Artetxe et al., ACL 2020



# Future Directions



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.



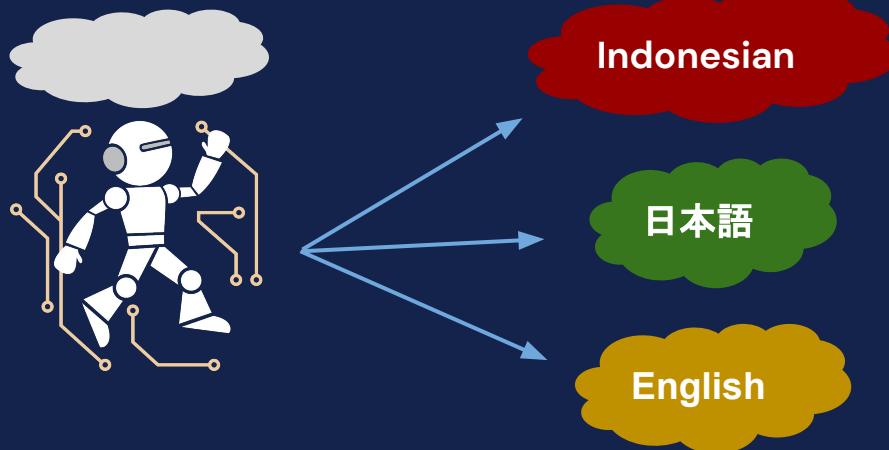
Distributionally Robust Optimization

$$\min_{\theta} \sup_q \mathbb{E}_{(x,y) \sim q} \mathcal{L}_{\theta}(x, y)$$

# Future Directions



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.



Distributionally Robust Optimization

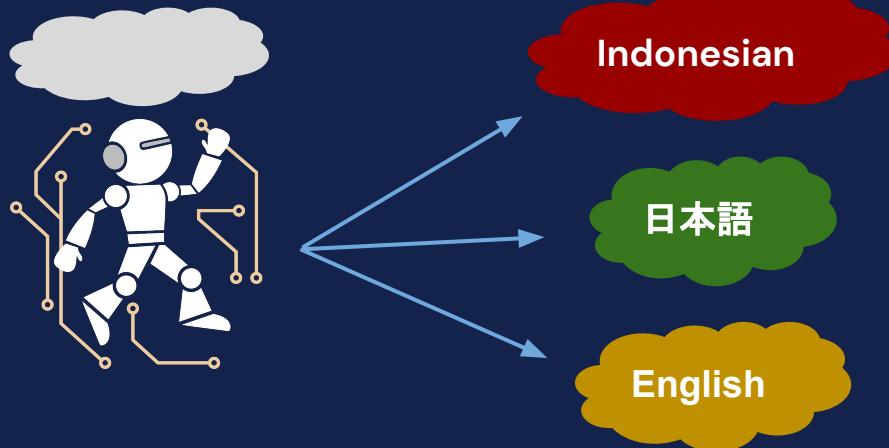
$$\min_{\theta} \sup_q \mathbb{E}_{(x,y) \sim q} \mathcal{L}_{\theta}(x, y)$$

Language

# Future Directions



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.



Distributionally Robust Optimization

$$\min_{\theta} \sup_q \mathbb{E}_{(x,y) \sim q} \mathcal{L}_{\theta}(x, y)$$

Ensuring that a language model works equally well across languages (important for fairness)

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**Training Paradigms**

**Model Architectures**

# Future Directions



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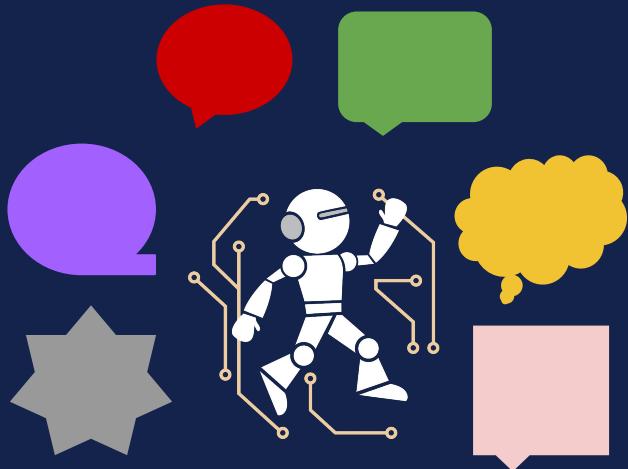
Training Paradigms

Model Architectures

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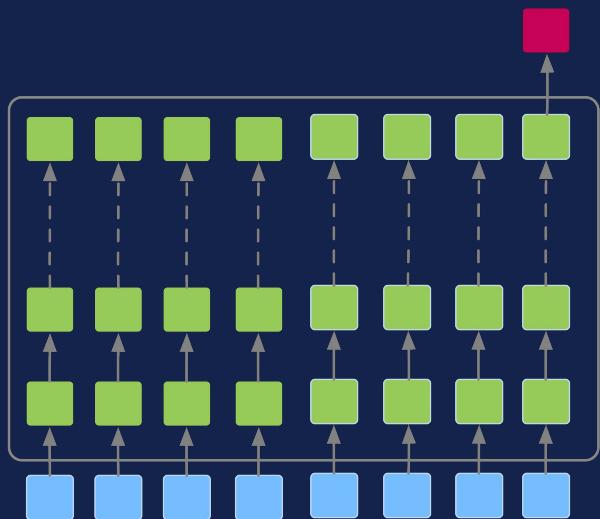


Integration of data from various sources and modalities.

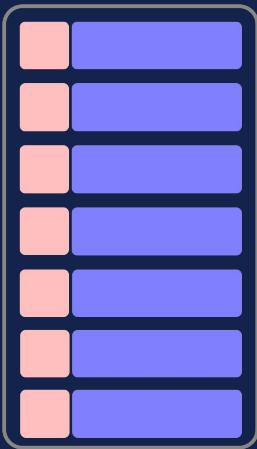
# Future Directions



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.



Computation

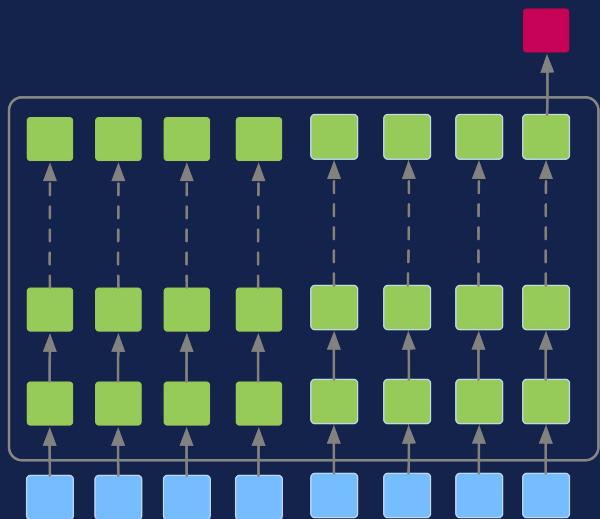


Storage

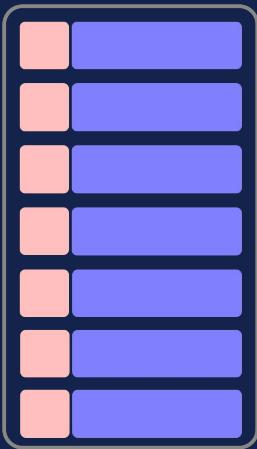
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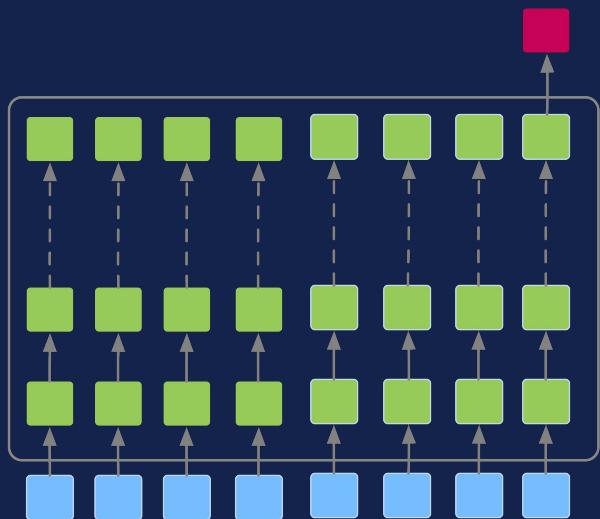
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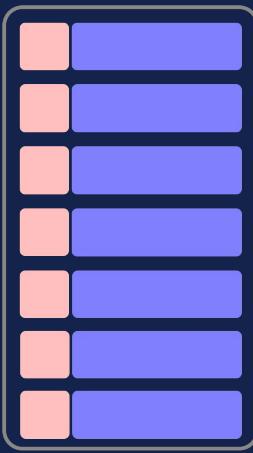
# Future Directions



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.



Computation



Storage

A screenshot of a profile page for Cristiano Ronaldo. At the top, there is a grid of five small images showing him in various poses. Below the images, his name "Cristiano Ronaldo" is displayed in a large, bold, black font, followed by the subtitle "Portuguese footballer". A link "cristianoronaldo.com" is provided. In the bottom right corner of the profile area, there is a "More images" button with a camera icon. The main content area contains the following biographical information:

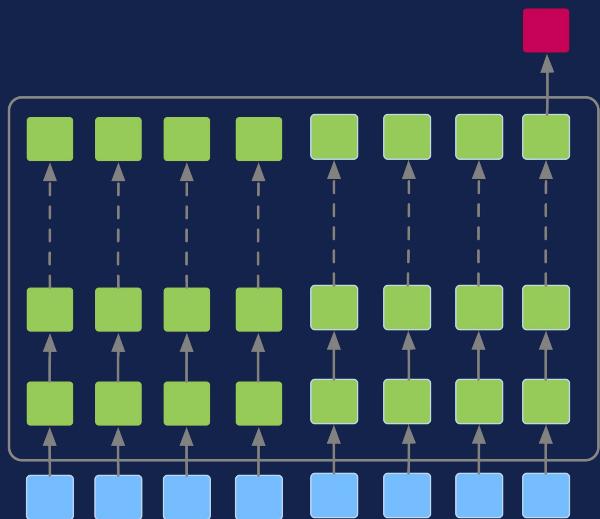
- Born: 5 February 1985 (age 36 years), Hospital Dr. Nélio Mendonça, Funchal, Portugal
- Height: 1.87 m
- Partner: Georgina Rodríguez (2017–)
- Salary: 31 million EUR (2019)
- Children: Cristiano Ronaldo Jr., Alana Martina dos Santos Aveiro, Eva Maria Dos Santos, Mateo Ronaldo
- Current teams: Juventus F.C. (#7 / Forward), Portugal national football team (#7 / Forward)

At the bottom right of the slide, there is a small "Share" icon.

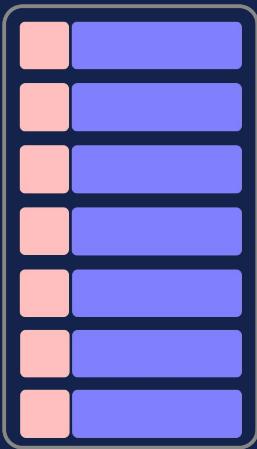
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Computation



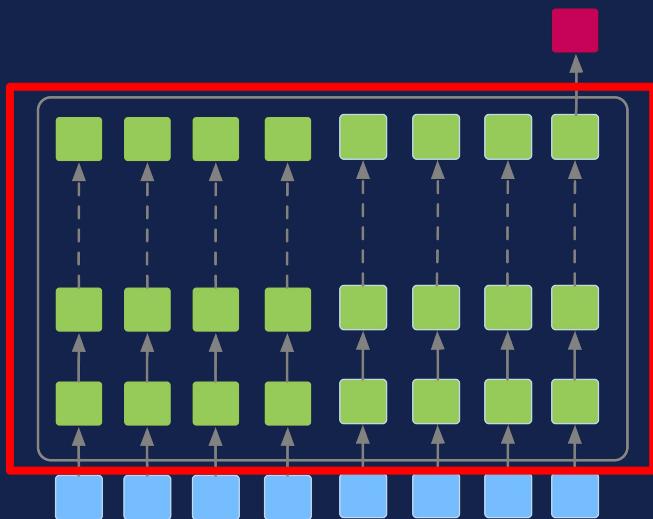
Storage

↑ computational efficiency

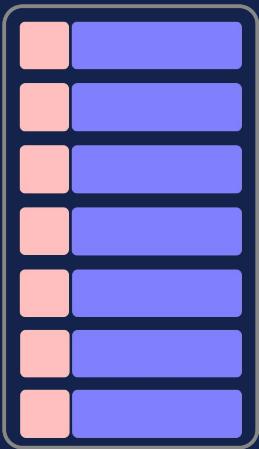
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Computation



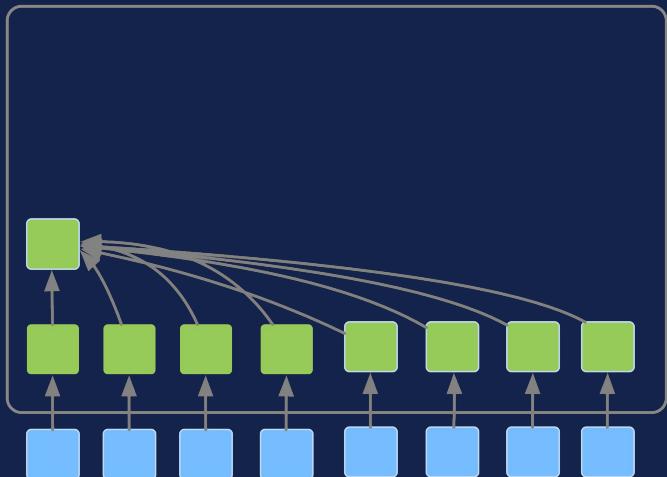
Storage

↑ computational efficiency

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Random Feature Attention

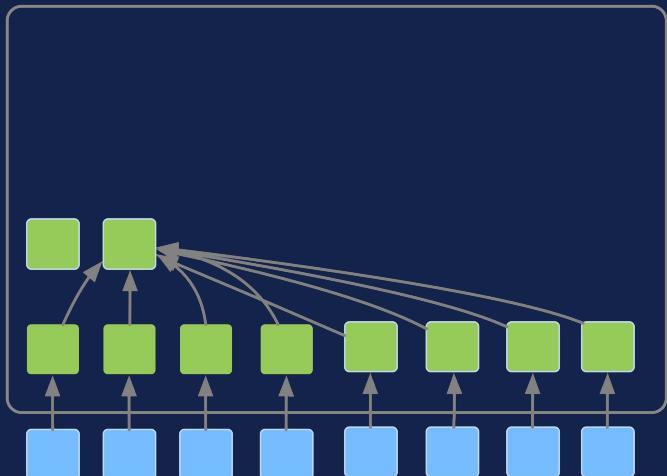
**Peng et al., ICLR 2021**



# Future Directions



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Random Feature Attention

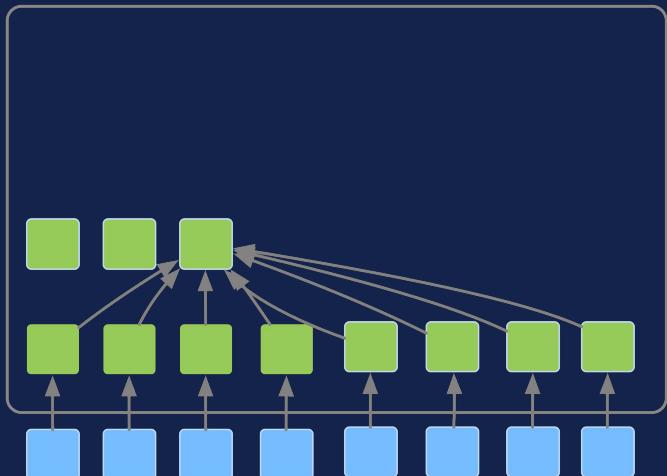
Peng et al., ICLR 2021



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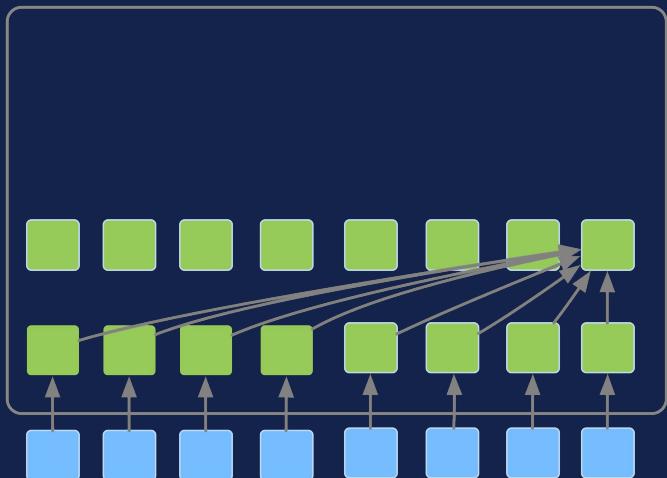
Peng et al., ICLR 2021



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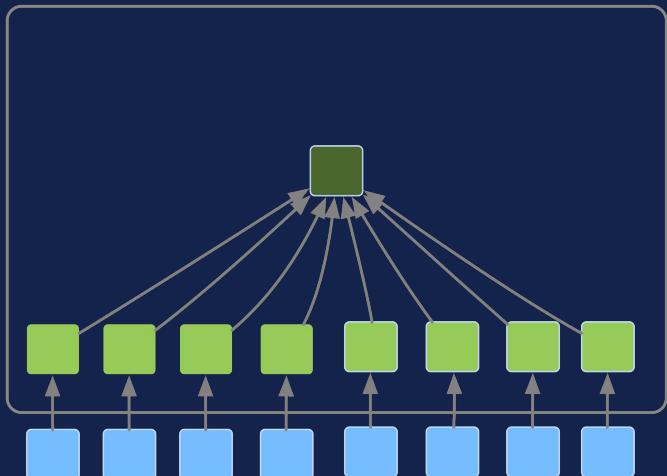
Peng et al., ICLR 2021



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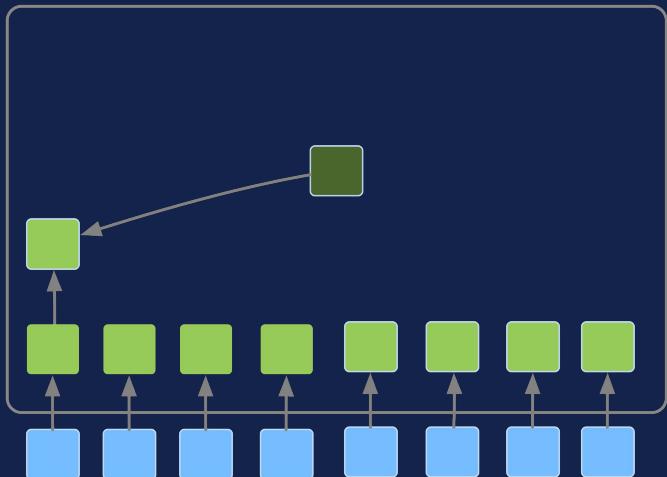
Peng et al., ICLR 2021



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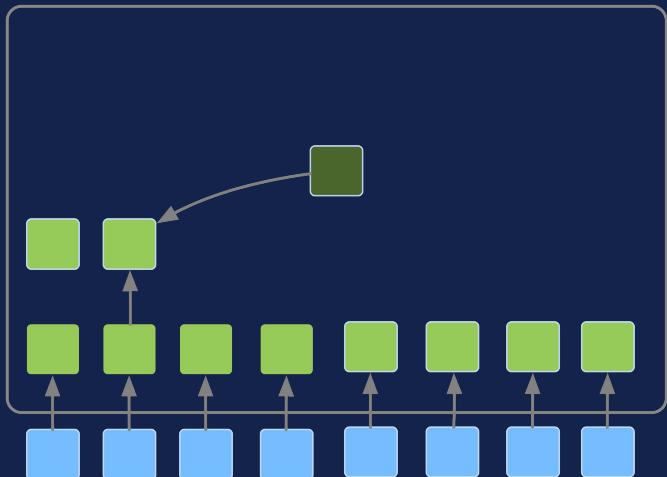
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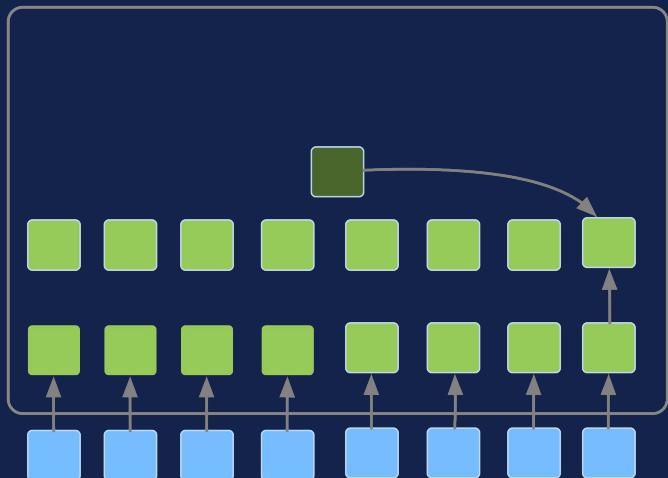
Peng et al., ICLR 2021



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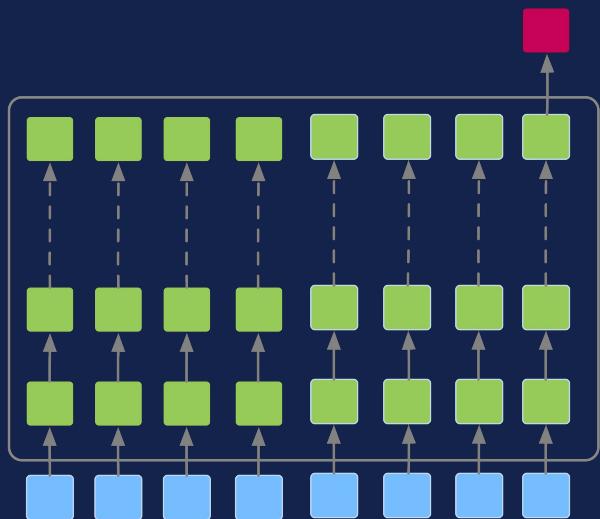
Peng et al., ICLR 2021



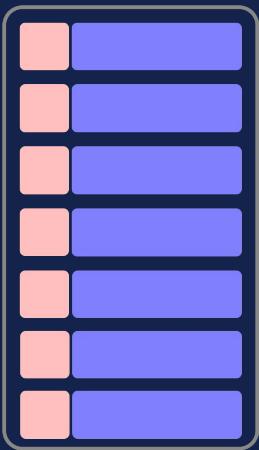
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Computation



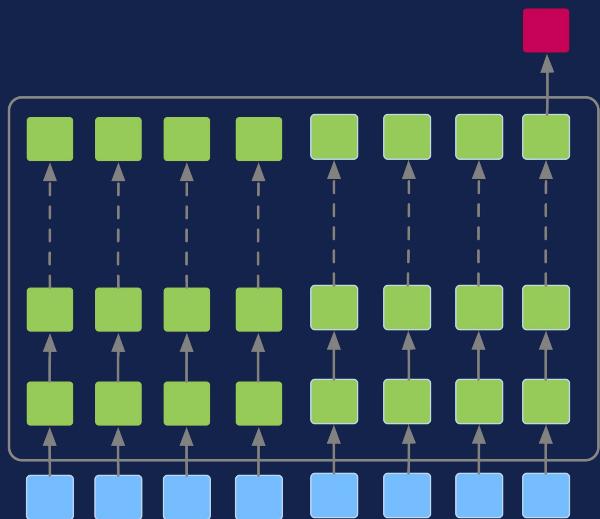
Storage

↑ computational efficiency

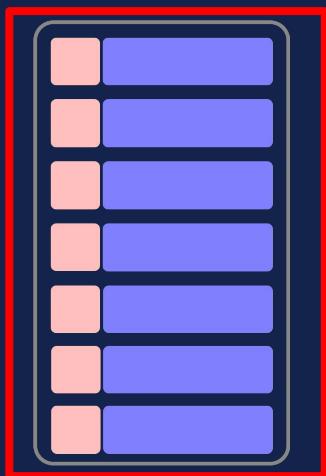
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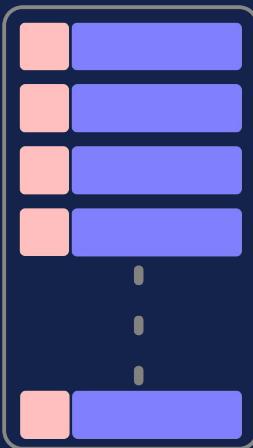
Storage

↑ computational efficiency

# Future Directions



A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.



Storage

Learning what to remember and forget

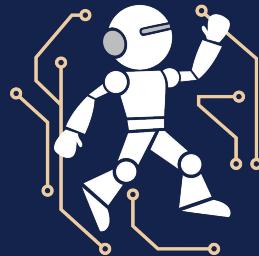


Constant-size memory

# Future Directions



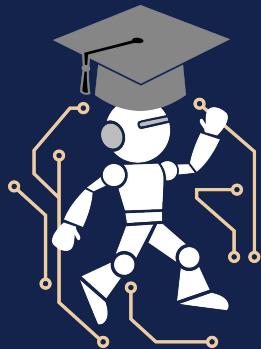
A language model that continually learns in an efficient way to perform multiple complex tasks in many languages.



# Future Directions



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tack გნორჩაჲალითჟიოს Danke  
ありがとうございました Salamat  
**grazie** **Thank you** multumesc நன்றி  
ধন্যবাদ Terima kasih Dankie 감사합니다 Merci  
Спасибо شکرا جزیلا σας ευχαριστώ  
teşekkür ederim 谢謝 cảm ơn bạn

<https://dyogatama.github.io>  
dyogatama@google.com