


DeepMind

Language Team Overview

ACL 2020

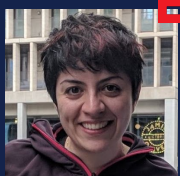


DeepMind Language Team

 is at ACL 2020



Adhi
Kuncoro



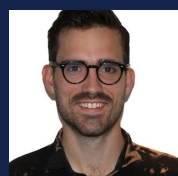
Aida
Nematzadeh



Aishwarya
Agrawal



Amy
Wu



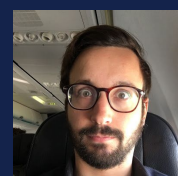
Andrew
Trask



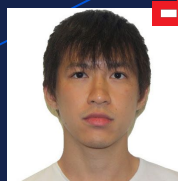
Angeliki
Lazaridou



Chris
Dyer



Cyprien
de Masson
d'Autume



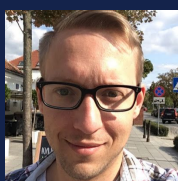
Dani
Yogatama



Devang
Agrawal



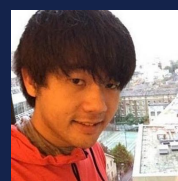
Gábor
Melis



Wojciech
Stokowiec



John
Hale



Kazuya
Kawakami



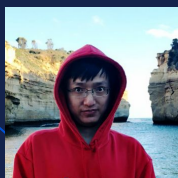
Kris
Cao



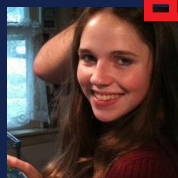
Laura
Rimell



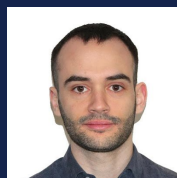
Lei
Yu



Lingpeng
Kong



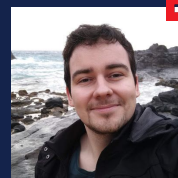
Lisa Anne
Hendricks



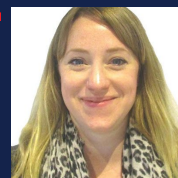
Laurent
Sartran



Phil
Blunsom



Sebastian
Ruder



Susannah
Young



Tomas
Kocisky



Wang
Ling

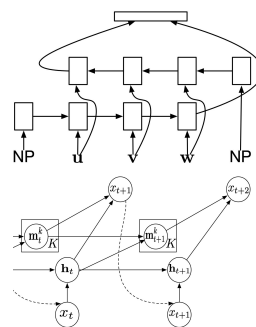
Language Modelling

Language modelling—assigning probabilities to sequences of tokens—is a fundamental NLP task, and better language models have many useful applications. We are interested in the following questions:

- What mechanisms allow language models to generalize better?
- How can we imbue language models with prior knowledge?
- How do we best incorporate longer-range context in our models?

Recent selected publications

1. [On the State of the Art of Evaluation in Neural Language Models](#) (ICLR 2018)
2. [Unsupervised Recurrent Neural Network Grammars](#) (NAACL 2019)
3. [Variational Smoothing in RNN LMs](#) (ICLR 2019)
4. [Mogrifier LSTM](#) (ICLR 2020)



[Recurrent Neural Network Grammars](#)
(NAACL 2016)

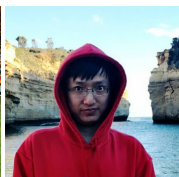
[Memory Architectures in RNN LMs](#)
(ICLR 2018)



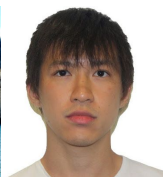
Adhi
Kuncoro



Gábor
Melis



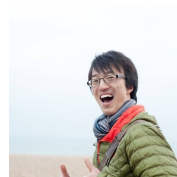
Lingpeng
Kong



Dani
Yogatama



Phil
Blunsom



Kris
Cao



Chris
Dyer



Kazuya
Kawakami

+ previous intern:
Yoon Kim



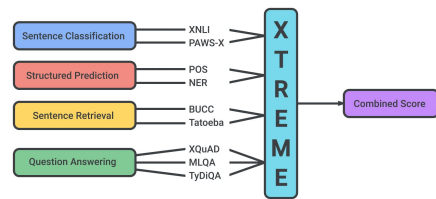
Machine Translation and Multilingual NLP

In order to develop general NLU methods, we need to learn abstractions that can generalise to any language. This will also help overcome language barriers and enable wider access to information. We are interested in the following questions:

- How can we incorporate additional context and leverage monolingual data in neural machine translation?
- How can we learn representations that generalise to extremely low-resource scenarios?
- What structural biases do we need for cross-lingual generalisation?

$$p(\mathbf{y}, \mathbf{z} | \mathbf{x}) \approx \prod_{j=1}^{|\mathbf{y}|} \underbrace{p(z_j | z_{j-1}, \mathbf{x}_1^{z_j}, \mathbf{y}_1^{j-1})}_{\text{alignment probability}} \underbrace{p(y_j | \mathbf{x}_1^{z_j}, \mathbf{y}_1^{j-1})}_{\text{word probability}}$$

The Neural Noisy Channel (ICLR 2017)



XTREME (ICML 2020)

Recent selected publications

1. [Better Document-level Machine Translation with Bayes' Rule](#) (TACL 2020)
2. [On the Cross-lingual Transferability of Monolingual Representations](#) (ACL 2020)
3. [Capturing document context inside sentence-level NMT models with self-training](#) (arXiv 2020)
4. [A Call for More Rigor in Unsupervised Cross-lingual Learning](#) (ACL 2020)



Lei Yu



Chris Dyer



Sebastian Ruder



Dani Yogatama



Phil Blunsom



Tomas Kocisky



Laurent Sartran



Wojciech Stokowiec

+ previous interns:
Mikel Artetxe &
Elman Mansimov



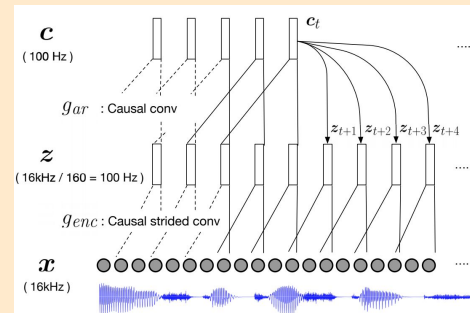
Representation and Transfer Learning

We are interested in learning robust and transferable representations. Our research directions in this area include:

- What biases and architectures allow us to deal with the full complexity of natural language across a variety of tasks?
- How can we effectively store and reuse representations?
- How can we adapt to new tasks in environments with limited amounts of data?

Recent selected publications

1. [Sentence Encoding with Tree-constrained Relation Networks](#) (arXiv 2018)
2. [Learning and Evaluating General Linguistic Intelligence](#) (arXiv 2019)
3. [Episodic Memory in Lifelong Language Learning](#) (NeurIPS 2019)
4. [On Mutual Information Maximization for Representation Learning](#) (ICLR 2020)
5. [Modelling Latent Skills for Multitask Language Generation](#) (arXiv 2020)
6. [Syntactic Structure Distillation Pretraining For Bidirectional Encoders](#) (arXiv 2020)



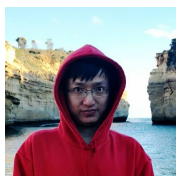
[Learning Robust and Multilingual Speech Representations](#)
(arXiv 2020)



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Kuncoro



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Yogatama



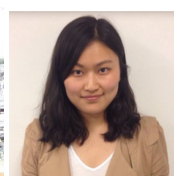
Lingpeng
Kong



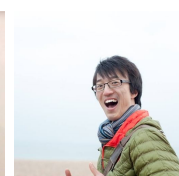
Sebastian
Ruder



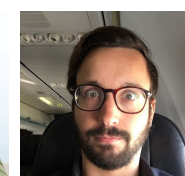
Kazuya
Kawakami



Lei
Yu



Kris
Cao



Cyprien de
Masson d'Autume



Document Modelling and Question Answering

We want to go from understanding sentences to documents and beyond. We are interested in the following questions:

- What is the best way to test document understanding and how to construct such test sets?
- How can we understand and integrate information across a long text or multiple documents?
- How can we reason about events, entities, and their relationships over time?

Selected publications

1. [Teaching Machines to Read and Comprehend](#) (NeurIPS 2015)
2. [Program Induction by Rationale Generation: Learning to Solve and Explain Algebraic Word Problems](#) (ACL 2017)
3. [Reference-Aware Language Models](#) (EMNLP 2017)
4. [The NarrativeQA Reading Comprehension Challenge](#) (TACL 2018)



Tomas
Kocisky



Wang
Ling



Dani
Yogatama



Phil
Blunsom



Chris
Dyer



Gábor
Melis

+ previous intern:
Zichao Yang



Multi-agent Communication and Dialogue

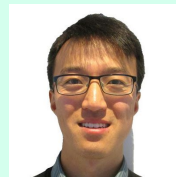
One of the fundamental functions of natural language is that of communication. Some of the questions that we are currently tackling are:

- How can we leverage multi-agent and reinforcement learning framework to tackle problems related to communication and dialogue?
- What are the properties of communication protocols emerging within multi-agent interactions grounded to a particular task?
- Can we nudge these properties to resemble those found in human communication?
- Can we use pre-trained and large scale language models as a key component of dialogue systems? What's the most appropriate way to do so?

Angeliki
Lazaridou




Kris
Cao



Laura
Rimell



Recent selected publications

1. [Emergent communication through negotiation](#) (ICLR 2018)
2. [Social Influence as Intrinsic Motivation for Multi-Agent Deep Reinforcement Learning](#) (ICML 2019)
3. [Biases for Emergent Communication in Multi-agent Reinforcement Learning](#) (NeurIPS 2019)
4. [Multi-agent Communication meets Natural Language: Synergies between Functional and Structural Language Learning](#) (ACL 2020) 

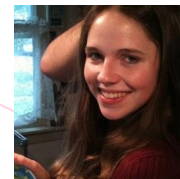


Language + Vision

In order to communicate about the physical world, AI agents must learn across multiple modalities. We are interested in the following questions:

- How can we learn general purpose multimodal features for downstream tasks such as image captioning or visual question answering?
- How robust are multimodal features to domain transfer and how can we ensure models do not amplify dataset bias during training?
- How can we effectively leverage data from single modality sources (e.g., unlabeled images or uncurated text) for multimodal tasks?

Lisa Anne Hendricks



Angeliki Lazaridou



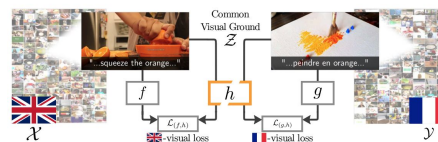
Aishwarya Agrawal



Aida Nematzadeh




Learning to Segment Actions



Visual Grounding in Video

Recent selected publications

1. [Learning to Segment Actions from Observation and Narration](#) (ACL 2020) 
2. [Visual Grounding in Video for Unsupervised Word Translation](#) (CVPR 2020)



Language + Cognitive Science

We take inspiration from cognitive science and linguistics to both better understand child language acquisition and also improve existing NLP systems. Broadly, we aim to:

- Shed light on mechanisms involved in human language acquisition by developing computational models and evaluating them against behavioral data.
- Identify the shortcoming of existing models by designing evaluation tasks inspired by experiments from developmental psychology.
- Improve performance/sample-efficiency of models by integrating them with inductive biases inspired by human cognition.

Recent selected publications

1. [Evaluating Theory of Mind in Question Answering](#) (EMNLP 2018)
2. [Finding syntax in Human Encephalography with Beam Search](#) (ACL 2018 - **Best Paper Award**)
3. [Text Genre and Training Data Size in Human-like Parsing](#) (EMNLP 2019)
4. [On Memory in Human and Artificial Language Processing Systems](#) (The BAICS workshop at ICLR 2020)



Adhi Kuncoro



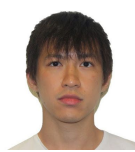
Aida Nematzadeh



Angeliki Lazaridou



Chris Dyer



Dani Yogatama



John Hale



Laura Rimell



Sebastian Ruder

