Association for Computational Linguistics

The Bright Future of ACL/NLP

Dr. Ming Zhou, ACL president
Microsoft Research Asia

ACL 2019

Florence, Italy, July 29, 2019

A big thanks to

- General Chair Lluís Màrquez, program chairs Anna Korhonen and David Traum, local organization chairs *Alessandro Lenci, Bernardo Magnini, Simonetta Montemagni, other chairs, and everyone of their teams*

- ACL 2019 Coordinating Committee(Marti Hearst, David Yarowsky, Priscilla Rasmussen and all others)

Galileo showed the Doge of Venice how to use the telescope

Outline »

ACL business update

NLP technical development

Galileo showed the Doge of Venice how to use the telescope

Outline »

ACL business update

NLP technical development

ACL: the premier scientific and professional society for CL/NLP

Annual conference
ACL

Conferences by
regional chapters

SIGs

21 SIGs
including SIGDAT,
SIGHAN, SIGMT,
SIGDIAL...

ACL Anthology
https://www.aclweb.org/anthology/

1962
ACL

1982
EACL

2000
NAACL

2018
AACL

The discipline of CL/NLP has been formed with solid theoretical framework, systematic technologies and important applications. Now CL/NLP has been widely viewed as the holy grail of AI.

ACL executive board's duties

Handle finance and membership

Select and negotiate venues for the main conferences

Help organize the various components of conferences

Coordinate 5 main conferences (ACL, EMNLP, EACL, NAACL, AACL)

Coordinate 21 SIGs and 50 workshops

Sponsor 2 journals (CL and TACL) and ACL anthology

Manage centralized IT

Make and execute policy that matters

Handle rising problems

Make strategic planning

Current ACL executive board members

Ming Zhou
President

Shiqi Zhao
Secretary (2016 - 2020)

Sharon Goldwater
EACL chair

Marti Hearst
Past President

Nitin Madnani
At-large (2019 - 2021)

Julia Hockenmaier
NAACL chair (2018 - 2019)

Hinrich Schütze
Vice-President

Barbara Di Eugenio
At-large (2018 - 2020)

Haifeng Wang
AACL Chair (2018 - 2020)

Rada Mihalcea
Vice-President-Elect (2019)

Jennifer Foster
At-large (2017 - 2019)

Supported by business manager:

David Yarowsky
Treasurer (2018 - 2022)

Hwee Tou Ng
CL Journal editor (2018-)

Priscilla Rasmussen
Business Manager

https://www.aclweb.org/adminwiki/index.php?title=ACL_Officers

Thanks to outgoing ACL execs who finished their terms

Joakim Nivre
Past President

Graeme Hirst
Past Treasurer

Jing-Shin Chang
At-large

Paola Merlo
CL Editor

Walter Daelemans
Chair of the EACL

ACL fellows in 2018

Robert Dale

For significant contributions to research in the generation of referring expressions and in natural language generation more broadly.

Jason Eisner

For significant contributions to probabilistic models and algorithms for finding linguistic structure, especially lexicalized syntax and morphology.

Mari Ostendorf

For significant contributions to prosody, pronunciation, acoustic, language modeling, and developments in using out-of-domain data and discourse structure.

Dragomir Radev

For significant contributions to text summarization and question answering, as well as large scale efforts to expand and diversify the computational linguistics pipeline.
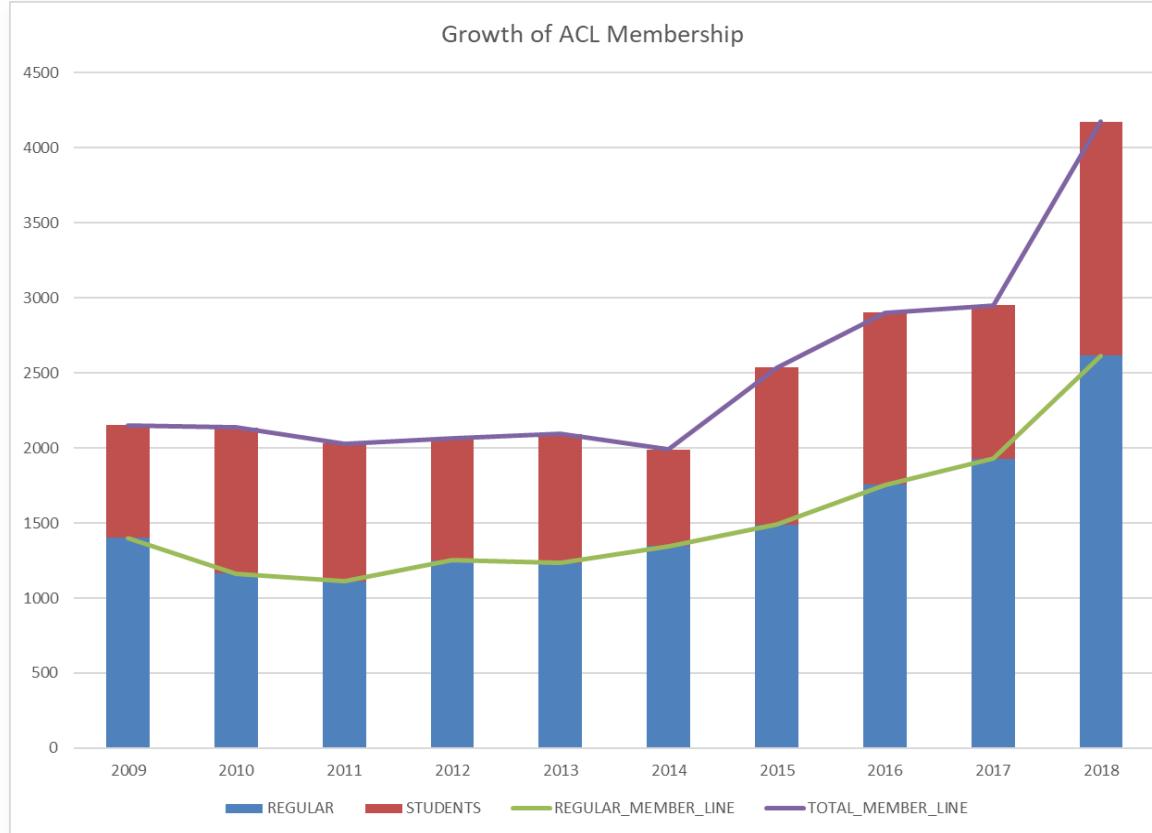
Ellen Riloff

For significant contributions to information extraction, and the analysis of sentiment, subjectivity and affect.
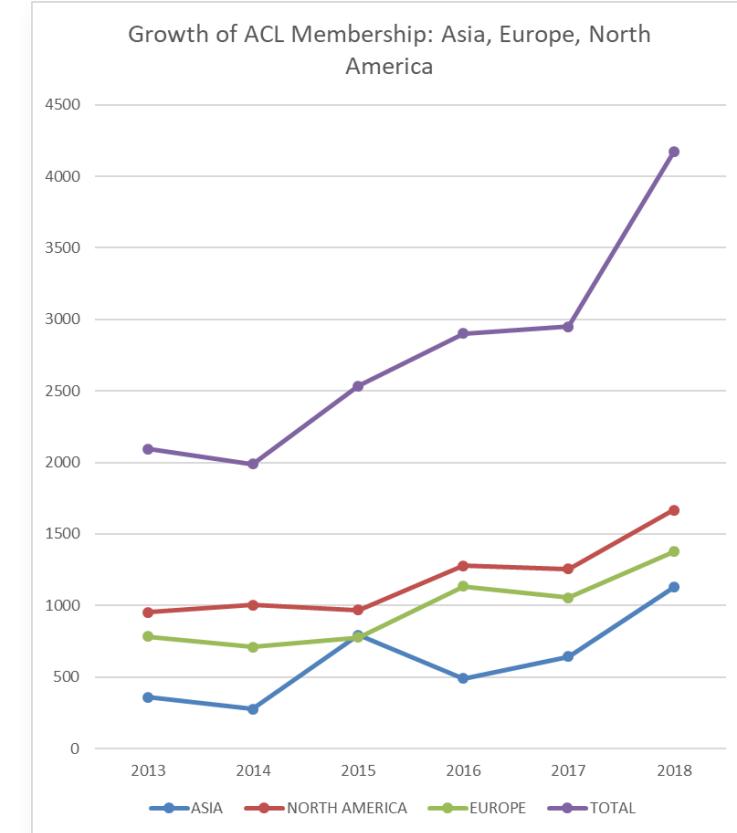
The ACL Fellows program has been established in 2011 by the ACL. The Fellows program recognizes ACL members whose contributions to the field have been most extraordinary in terms of scientific and technical excellence, service to the association and the community and/or educational or outreach activities with broader impact.

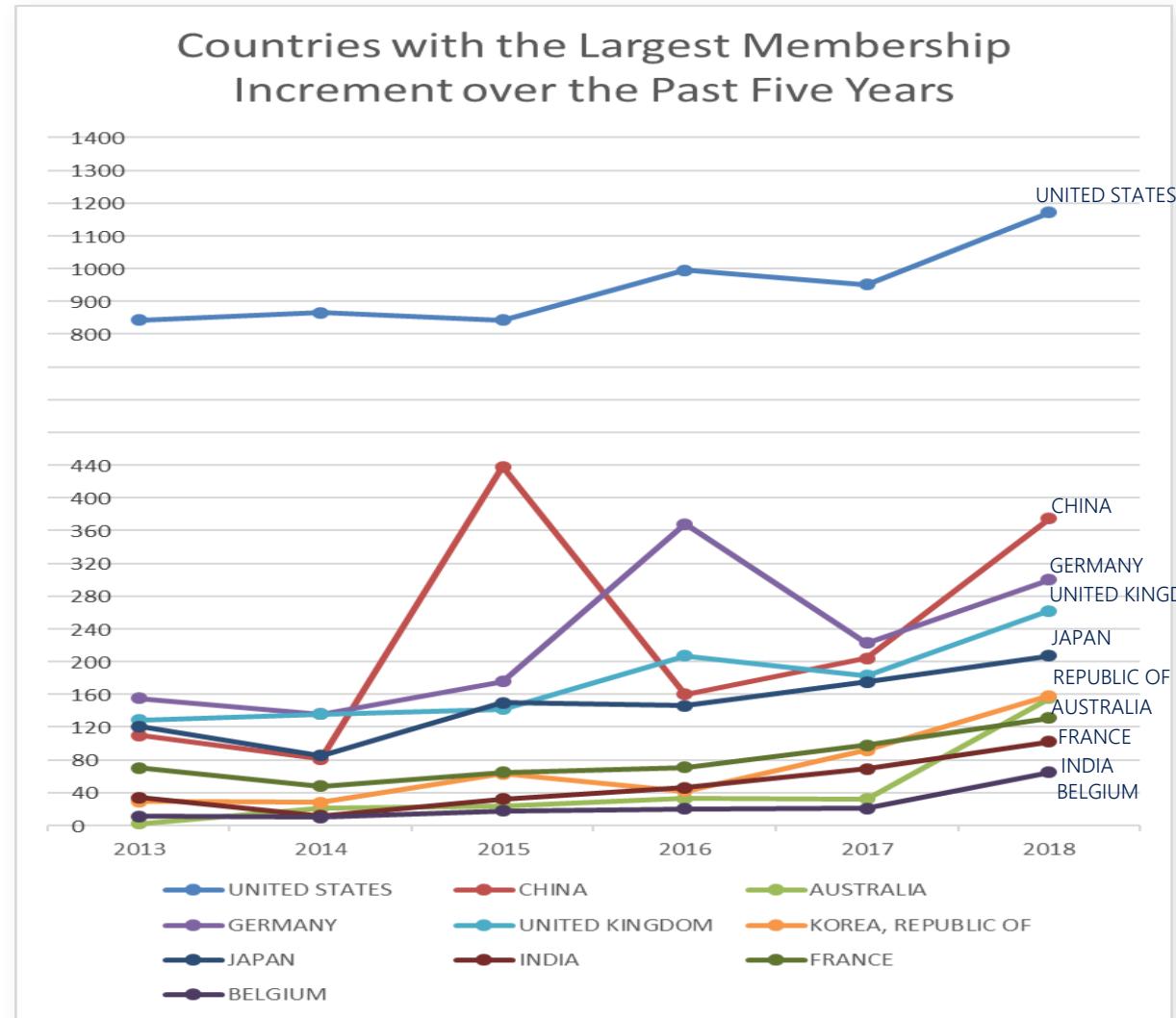https://aclweb.org/aclwiki/ACL_Fellows

Smooth growth of ACL membership

Growth of ACL Membership

Growth of ACL Membership: Asia, Europe, North America

REGULAR   REGULAR_MEMBER_LINE

STUDENTS   TOTAL_MEMBER_LINE
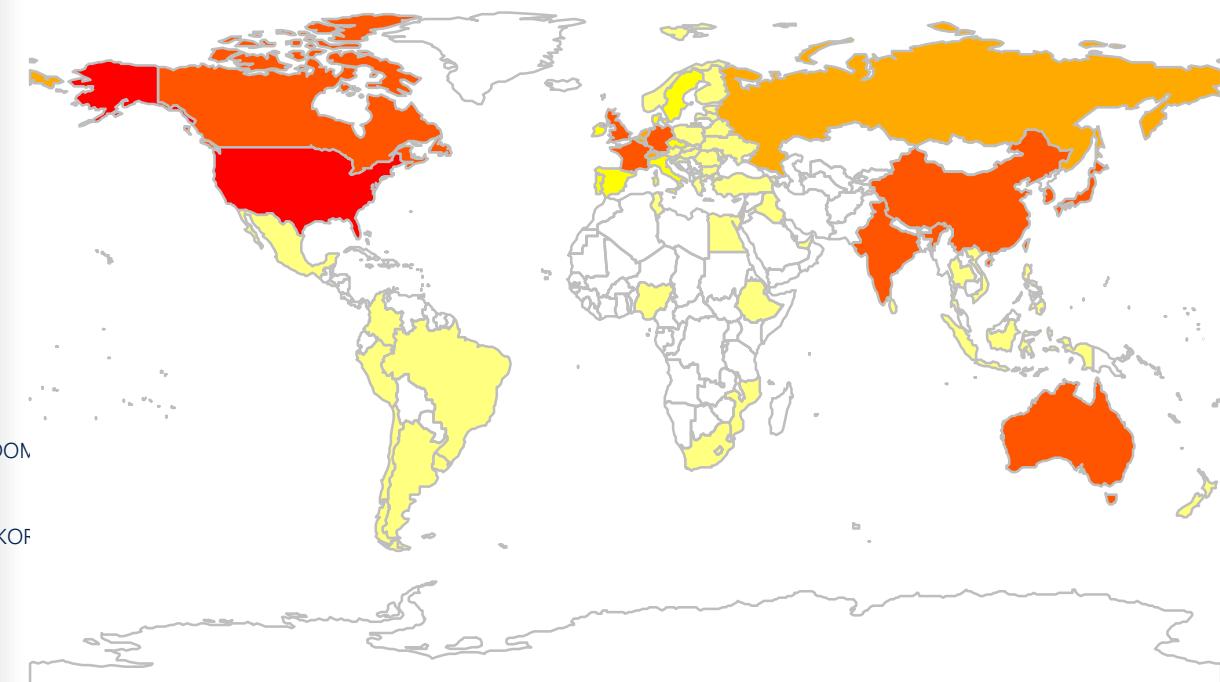
ASIA   EUROPE

NORTH AMERICA   TOTAL
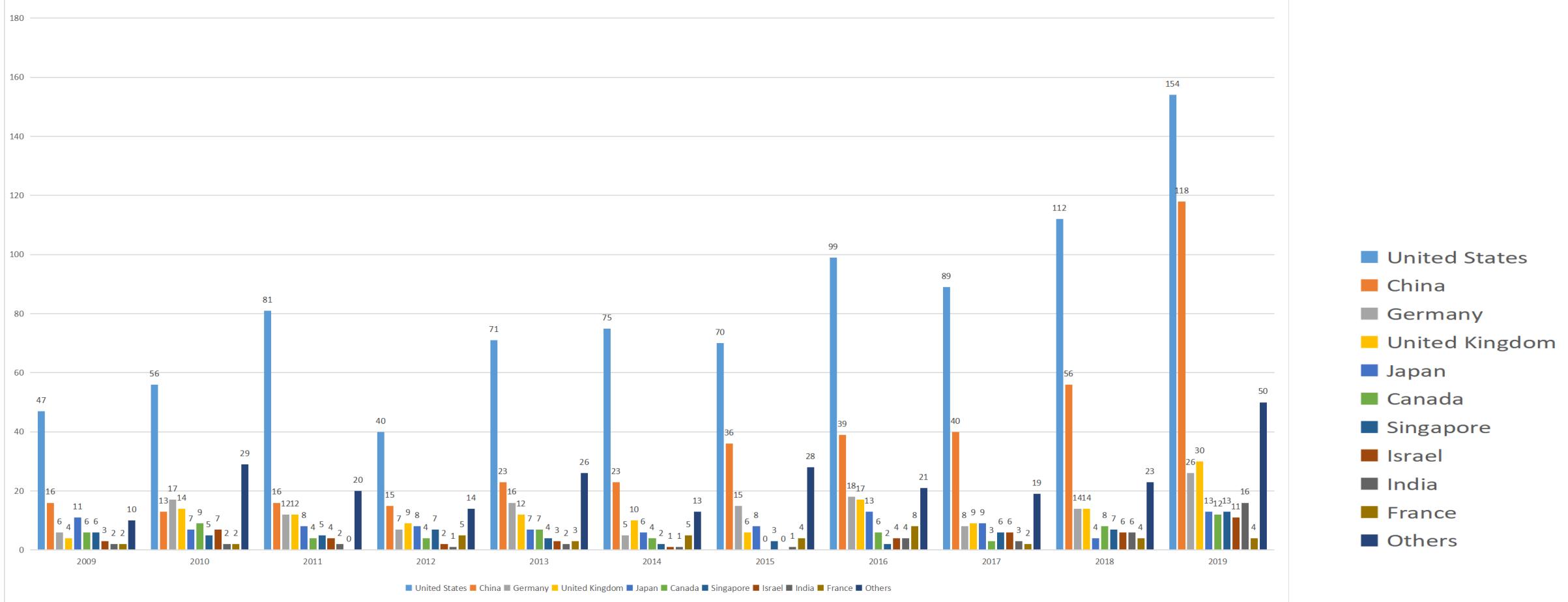
Distribution of membership (statistics in 2013-2018)

Countries with the Largest Membership Increment over the Past Five Years

ACL membership density

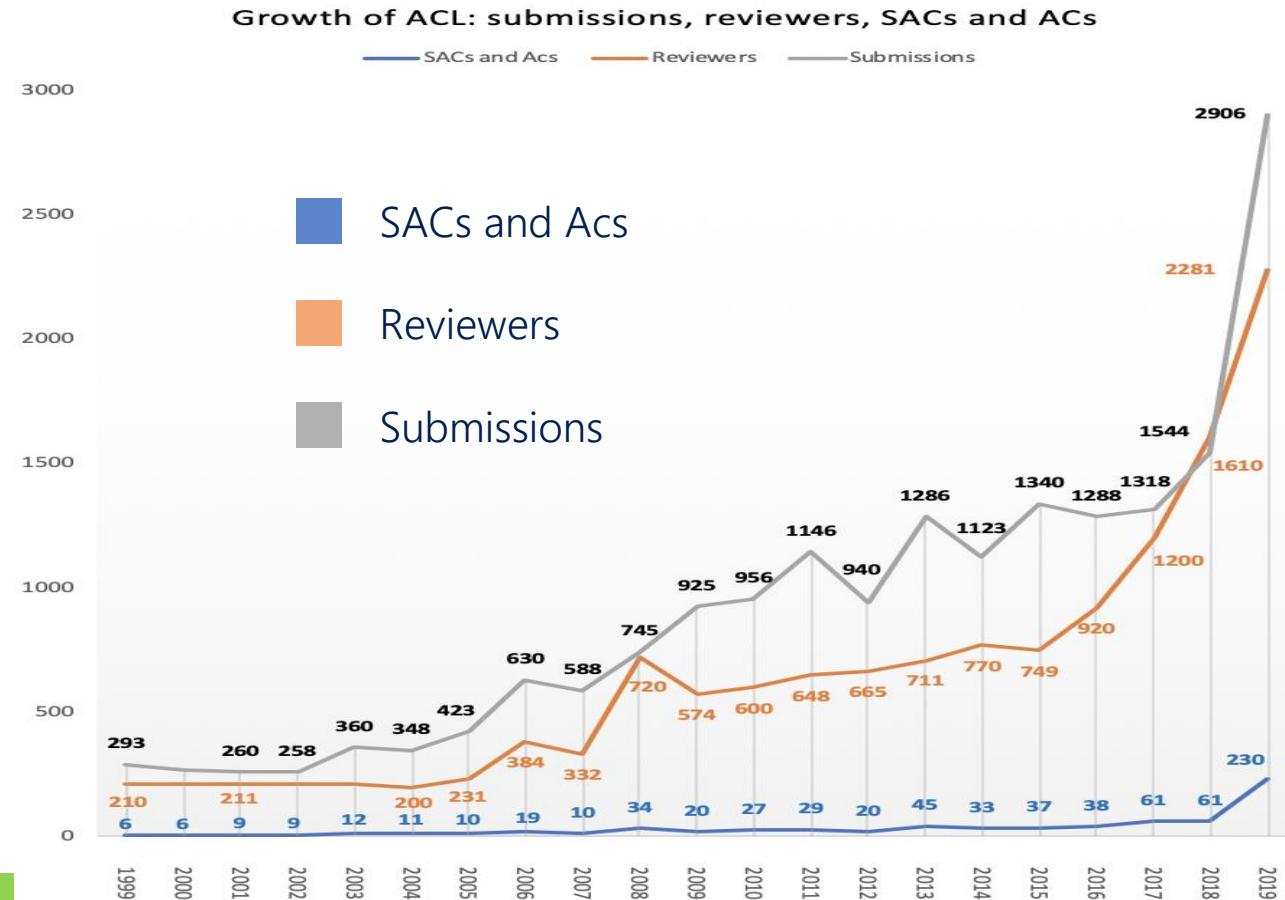# Imbalance of accepted papers over countries/regions

Based on ACL full papers (2009-2019)

Soaring growth of submission causes huge challenges to paper review

$|papers|$

$\propto |conferences| \times |topics| \times |datasets|$

$\times |network\ structures| \times |learning\ algorithms|$

$\times |languages|$

$\times |researchers + professors + students|$

$\times |rejection\ times|$

$\times |high\ salary\ of\ AI\ jobs| \times |\dots|$

There will be a special discussion on paper reviewing at business meeting on July 30

Growth of ACL: submissions, reviewers, SACs and ACs

SACs and Acs — Reviewers — Submissions

SACs and Acs
Reviewers
Submissions

http://acl2019pcblog.fileli.unipi.it/?p=156

ACL/NLP growth in Asia-Pacific
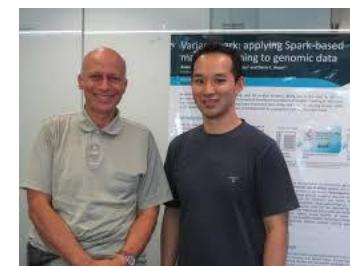
Diverse societies in Asia-Pacific

- Australasian Language Technology Association (ALTA), Sydney, Australia
- Natural Language Processing Association India (NLPAI), Hyderabad, India
- Indonesian Association of Computational Linguistics (INACL), Jakarta, Indonesia
- The Association for Natural Language Processing (ANLP), Tokyo, Japan
- SIG-HLT (Special Interest Group of Human Language Technology) of KIISE (Korea Institute of Information Scientists and Engineers), Pohang, Korea
- Chinese and Oriental Languages Information Processing Society (COLIPS), Singapore
- The Association for Computational Linguistics and Chinese Language Processing (ACLCLP), Chinese Taipei
- Society for Natural Language Processing (SNLP), Lahore, Pakistan
- Chinese Information Processing Society
- China Computer Federation
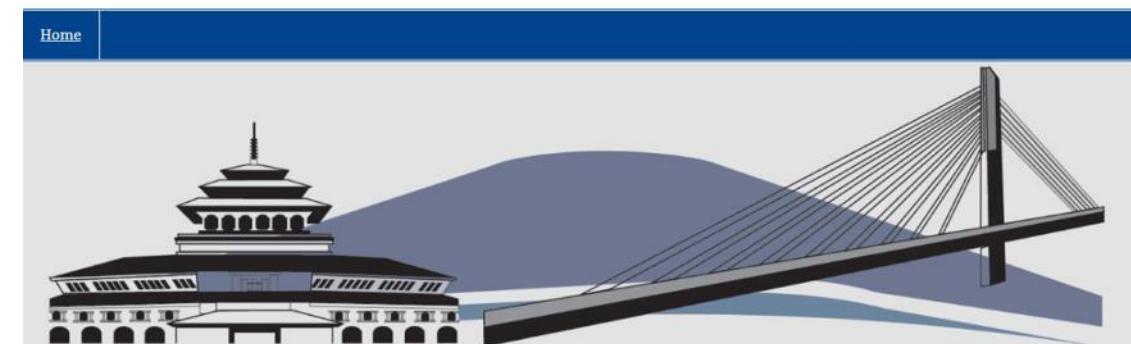- China Association of Artificial Intelligence

Annual China-Japan joint NLP workshop (since 2001)

The 18th China Japan Natural Language Processing Joint Research Promotion Conference (CJNLP2018)

第六届中日自然语言处理专家研讨会
The 6th China-Japan Natural Language processing Joint Research Promotion Conference
中国 上海交通大学 2006.11.13-15   Shanghai Jiao Tong University,China Nov 13-15,2006

第3回 日中自然言語処理共同研究促進会議
主催 独立行政法人 通信総合研究所   会期 平成15年11月11日火・12日水

Conferences organized by NLP societies in Asia-Pacific
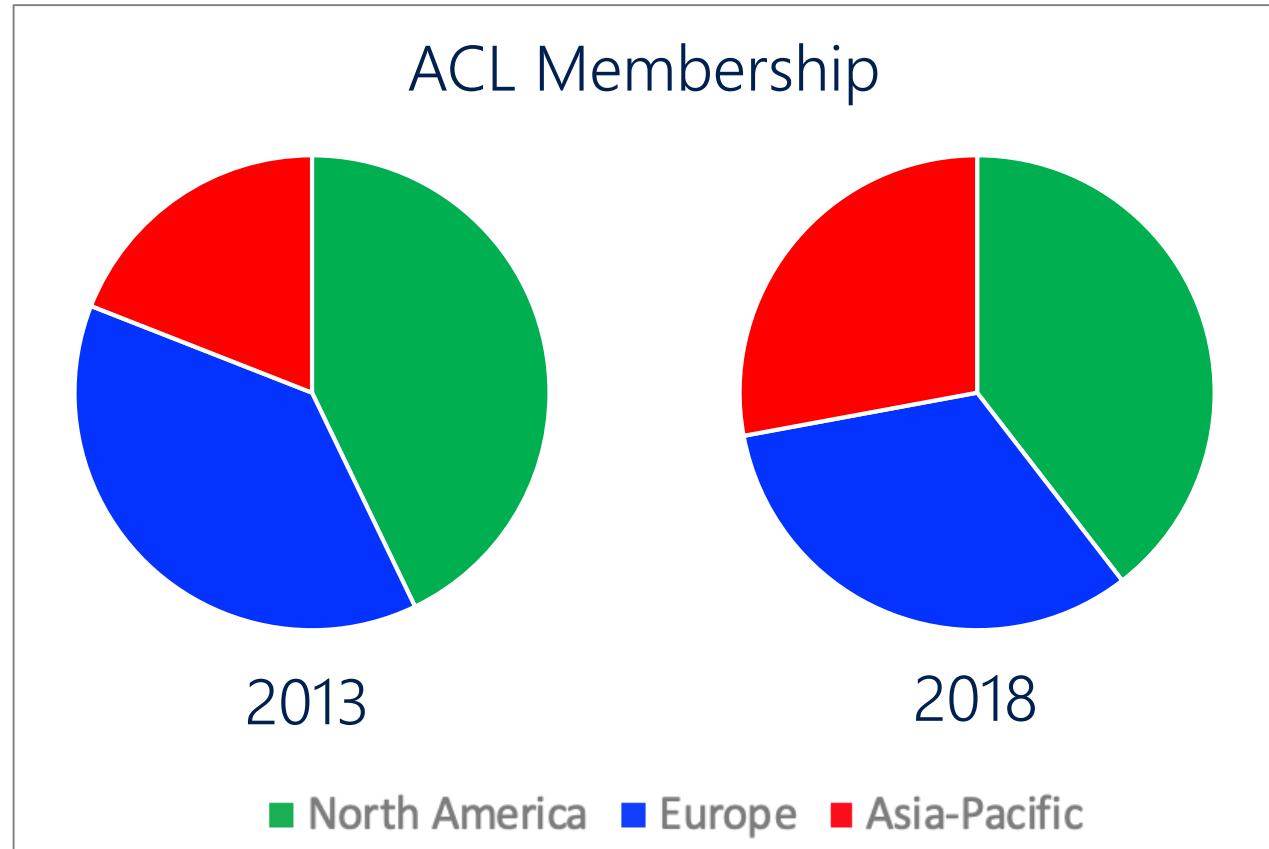
Conferences, summer schools, tutorials in China

Fast growth of ACL membership in Asia-Pacific

- The impact of ACL has been dramatically increasing in Asia-Pacific in recent years

ACL Membership
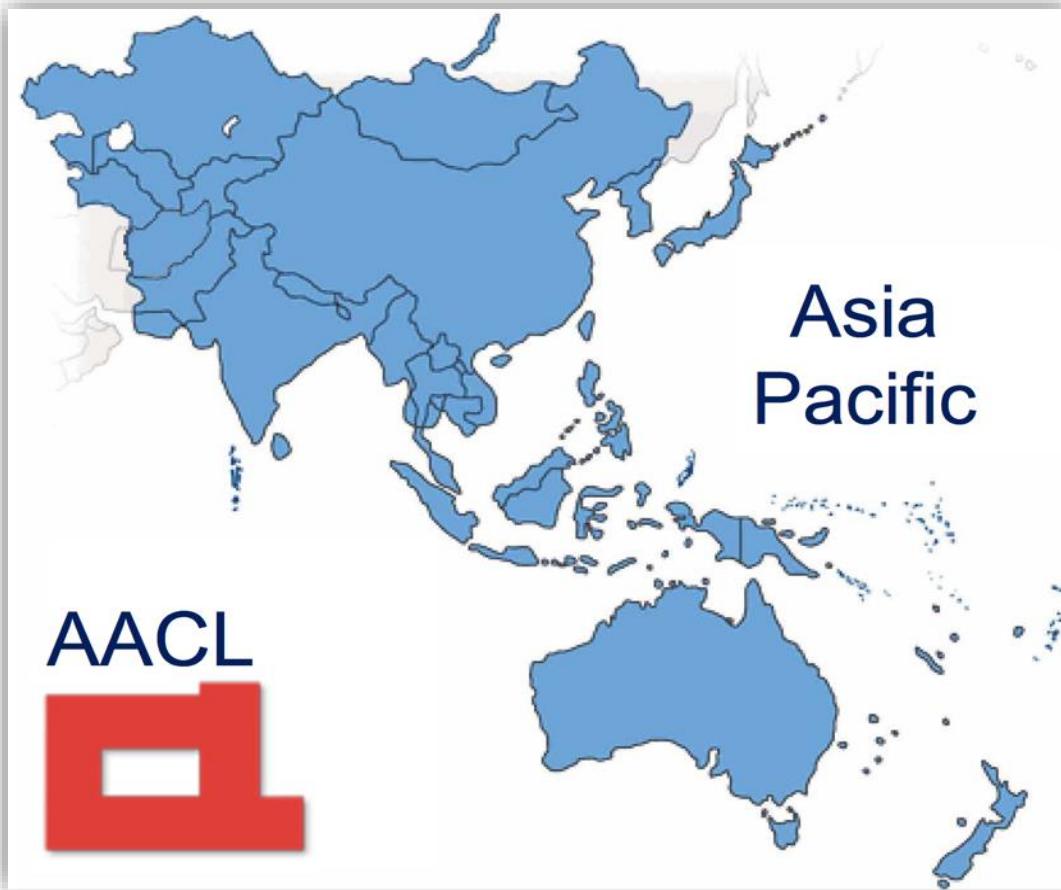
2013   2018

■ North America   ■ Europe   ■ Asia-Pacific

Year	Host
2018	Melbourne
2015	Beijing
2012	Jeju Island
2009	Singapore
2006	Sydney
2003	Sapporo
2000	Hong Kong

ACLs held in Asia-Pacific

AACL: the Asia-Pacific Chapter of ACL (launched in 2018)

- Serves ACL members from 57 countries/regions in Asia-Pacific
- Builds a new bridge with AFNLP and all NLP societies in Asia-Pacific

AACL Executive Board

Asia Pacific

AACL

Haifeng Wang
Chair
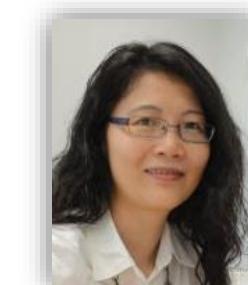
Keh-Yih Su
Chair-elect

Yang Liu
Secretary

Seung-won Hwang
Treasurer

Yusuke Miyao
At-large

Jian Su
At-large

Mark Dras
At-large

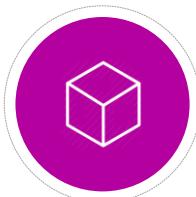# Towards balanced, inclusive and diverse development of ACL/NLP
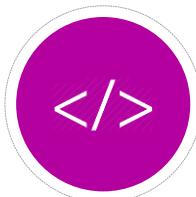
Better membership service by ACL and its chapters

Talent fostering by summer schools, mentoring programs, internship programs, language training

Conferences and activities in diverse venues

Strong support to low-resource languages

WiNLP/EquiCL/BIG to encourage diversity and inclusion

IT system, review system, coordination across chapters, SIGs and conferences

New committees

Information Committee Director

Nitin Madnani

Equity Committee Director

Under search led by
Rada Mihalcea

Publicity Committee Director

Barbara Plank

Professional Conduct Committee Directors

Emily M Bender

Graeme Hirst

Anthology Committee Director

Matt Post

Business Meeting
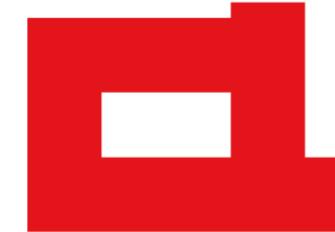## Plenary Hall
July 30, 2019

- Reports from ACL functional units (secretary, treasurer, office, IT, CL, TACL)
- Updates on EACL, NAACL and AACL
- Progress on setting up ACL2020, ACL2021
- Special panel on paper reviewing

PANEL DISCUSSION

Galileo showed the Doge of Venice how to use the telescope

Outline »
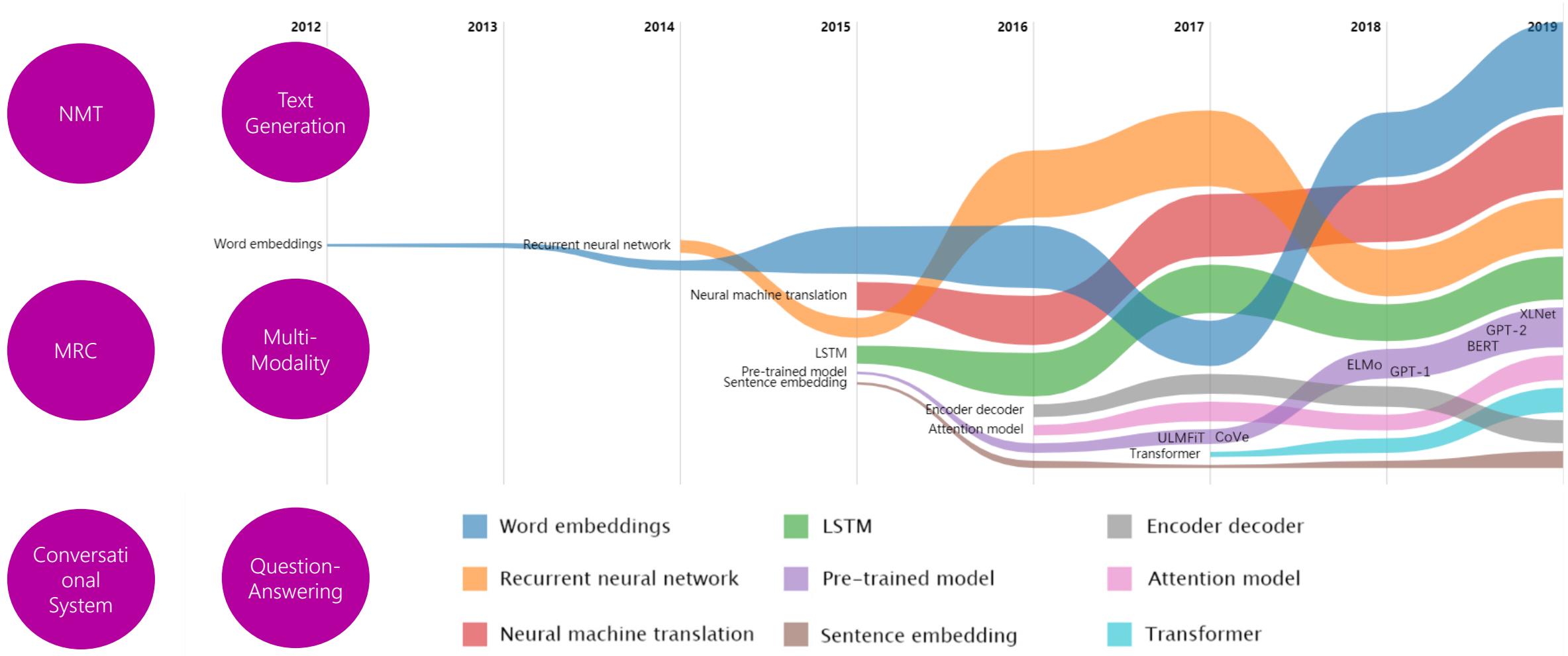
ACL business update
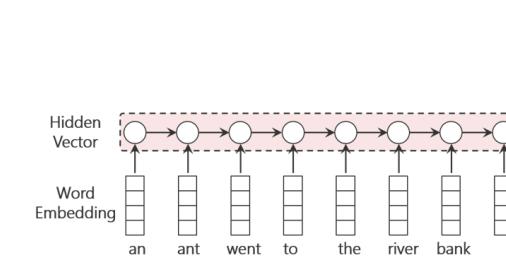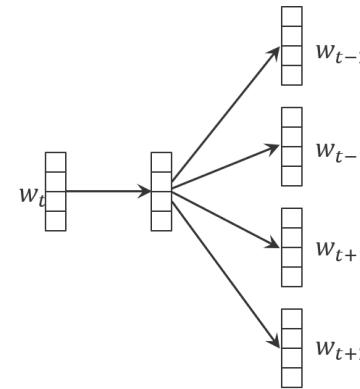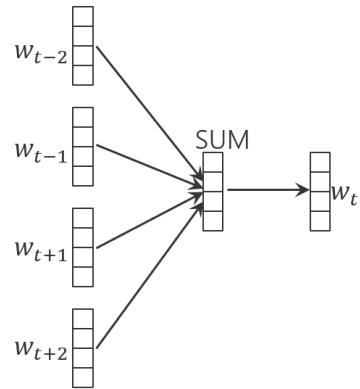
NLP technical development
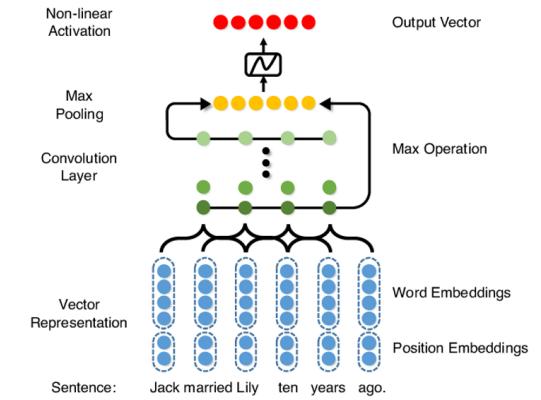
DNN-NLP progress

Key techs of DNN-NLP

CBOW (Continuous Bag-of-Words): using the context words in a window to predict the central word.

Skip-gram (Continuous Skip-gram): using the central word to predict the context words in a window.

Word embedding (Mikolov et al., 2013)

Recurrent Neural Networks (RNN)

Convolutional Neural Networks (CNN)

Sentence Embedding

Encoder-Decoder with attention (Bahdanau et al., 2014)

Transformer (Vaswani et al., 2016)

Pre-training + finetune, a new paradigm of NLP

Where is the future direction of NLP?

- Are we satisfied with current DNN-NLP?

- DNN-NLP deeply relies on huge cost of computer power and annotated data and suffers from big challenges in modelling, reasoning and interpretability.

- Linguistics, knowledge, common sense and symbolic reasoning should still play important roles to solve these challenges.

- I would like to analyze challenges in typical tasks and share my views on the technical developments.

Dataset: high cost, bias, noises, privacy and discrepancy from real scenarios

Racial Analogies

black → criminal	caucasian → lawful
caucasian → hillbilly	asian → yuppie
asian → engineer	black → killer

Religious Analogies

christian → conservative	jew → liberal
muslim → terrorist	jew → journalist
christian → conservative	muslim → regressive
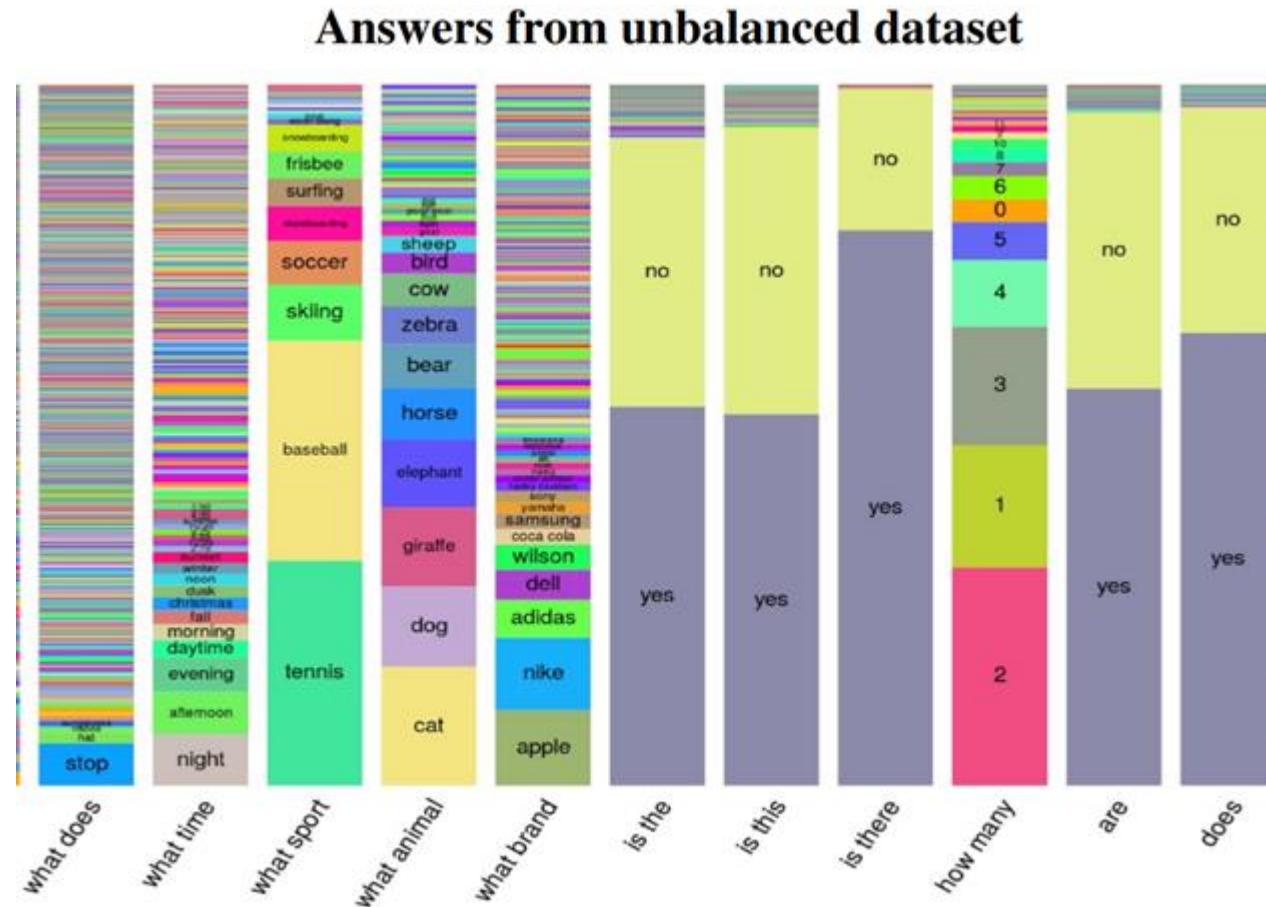
Table 1: Examples of racial and religious biases in analogies generated from word embeddings trained on the Reddit data from users from the USA.

Answers from unbalanced dataset

Thomas Manzini, Yao Chong Lim, Yulia Tsvetkov, Alan W Black. Black is to Criminal as Caucasian is to Police: Detecting and Removing Multiclass Bias in Word Embeddings. NAACL, 2019.

Yash Goyal, Tejas Khot, Douglas Summers-Stay, Dhruv Batra, Devi Parikh. Making the V in VQA Matter: Elevating the Role of Image Understanding in Visual Question Answering. CVPR, 2017.

Fierce computing power arm races

- New arms race over AI chips and computing power

- High computation cost for big models

- Low return on investment(sometimes)

- New type of environmental pollution

Model	Hardware	Power (W)	Hours	kWh·PUE	CO$_2$e	Cloud compute cost
Transformer$_{base}$	P100x8	1415.78	12	27	26	$41–$140
Transformer$_{big}$	P100x8	1515.43	84	201	192	$289–$981
ELMo	P100x3	517.66	336	275	262	$433–$1472
BERT$_{base}$	V100x64	12,041.51	79	1507	1438	$3751–$12,571
BERT$_{base}$	TPUv2x16	—	96	—	—	$2074–$6912
NAS	P100x8	1515.43	274,120	656,347	626,155	$942,973–$3,201,722
NAS	TPUv2x1	—	32,623	—	—	$44,055–$146,848
GPT-2	TPUv3x32	—	168	—	—	$12,902–$43,008

Emma Strubell, Ananya Ganesh and Andrew McCallum. Energy and Policy Considerations for Deep Learning in NLP. ACL, 2019.
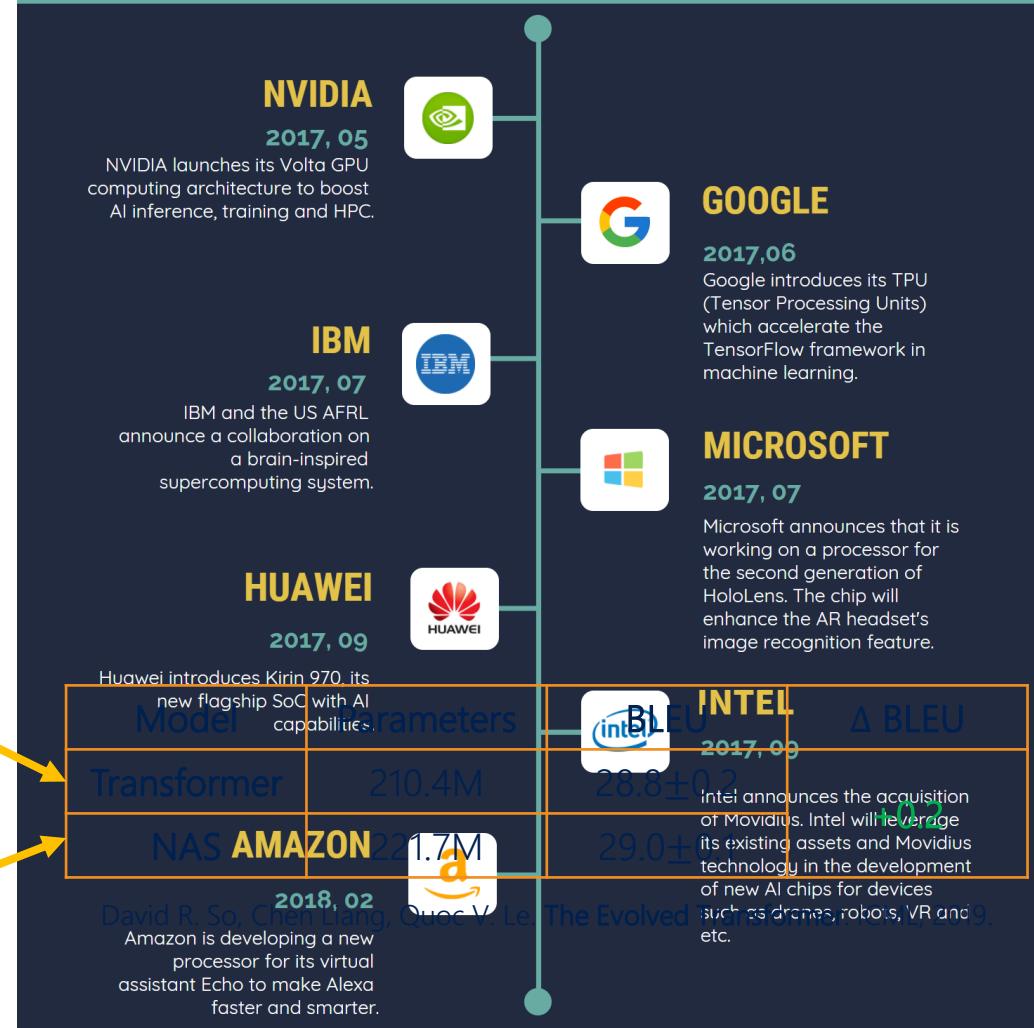
Big Corps AI Chips Strategy

NVIDIA
2017, 05
NVIDIA launches its Volta GPU computing architecture to boost AI inference, training and HPC.

GOOGLE
2017,06
Google introduces its TPU (Tensor Processing Units) which accelerate the TensorFlow framework in machine learning.

IBM
2017, 07
IBM and the US AFRL announce a collaboration on a brain-inspired supercomputing system.

MICROSOFT
2017, 07
Microsoft announces that it is working on a processor for the second generation of HoloLens. The chip will enhance the AR headset's image recognition feature.

HUAWEI
2017, 09
Huawei introduces Kirin 970, its new flagship SoC with AI capabilities.

INTEL
2017, 09
Intel announces the acquisition of Movidius. Intel will leverage its existing assets and Movidius technology in the development of new AI chips for devices such as drones, robots, VR and etc.

AMAZON
2018, 02
Amazon is developing a new processor for its virtual assistant Echo to make Alexa faster and smarter.

Model	Parameters	BLEU	Δ BLEU
Transformer	210.4M	28.8±0.3	
NAS	221.7M	29.0±0.2	+0.2

David R. So, Chen Liang, Quoc V. Le. The Evolved ...

Analysis on typical tasks
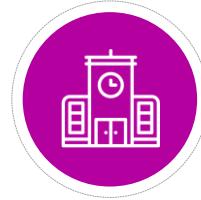
Rich-Resource

Low-Resource

Multi-Turn

Analysis on typical tasks

Rich-Resource

Low-Resource

Multi-Turn

Error analysis of NMT results (Ch-En)

Error Category	Fraction [%]	
Incorrect Words	7.64	
Ungrammatical	6.33	
Missing Words	5.46	
Named Entity	4.38	
Person		1.53
Location		1.53
Organization		0.66
Event		0.22
Other		0.44
Word Order	0.87	
Factoid	0.66	
Word Repetition	0.22	
Collocation	0.22	
Unknown Words	0	

Error distribution, as fraction of sentences that contain specific error categories.

The translation of 500 sentence were manually checked and error types were labeled

This analysis indicate that there is still big room to improve the translation quality

Achieving Human Parity on Automatic Chinese to English News Translation
https://arxiv.org/abs/1803.05567

Missing words

英国最高法院随后裁定引渡他到瑞典。但阿桑奇于2012年6月在保释期间进入位于伦敦的厄瓜多尔驻英国使馆寻求庇护至今。阿桑奇创建的"维基揭秘"网站因公布大量美国有关阿富汗和伊拉克战争的秘密文件，引起轰动和争议。

The British Supreme Court subsequently decided to extradite him to Sweden. But Assange entered the British embassy in London in June 2012 on bail to seek refuge so far. The WikiLeaks website, created by Mr. Assange, has caused a stir and controversy by publishing a large number of U.S. secret documents about the war in Afghanistan and Iraq.

"厄瓜多尔" (Ecuador) is missed

Acronym understanding

德国在参与打击极端组织的多国联合行动时，向土耳其空军基地派驻约250名军人。土耳其政府此前指责德国为参与去年7月土耳其未遂政变的人员提供政治避难。作为报复，**土方**禁止德国议员探视德国驻军。

Germany has deployed about 250 troops to the Turkish Air Force base in its multinational operations against extremist groups. The Turkish government has previously accused Germany of providing political asylum for those who participated in last July's attempted coup in Turkey. In retaliation, the Earth forbids German MPs to visit the German garrison.

"土方" (contextualized acronym of 土耳其 /Turkey) is wrongly translated

Unknown named entities

日前闭幕的"一带一路"国际合作高峰论坛收获了丰硕成果，达成多个合作项目，提出了一系列合作举措，赢得广泛赞誉。

The recent closing of the "area all the way" International Cooperation Summit Forum Harvest fruitful results, reached a number of cooperation projects, put forward a series of cooperation initiatives, won wide acclaim.

国务院总理李克强19日在中南海紫光阁会见菲律宾众议长阿尔瓦雷兹。

Chinese Premier Li Keqiang met with Philippine Chancellor Alvarez in Zhongnanhai 19th.
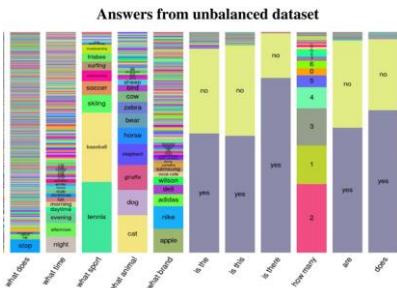
"一带一路" (one belt one road) is wrongly translated

"众议长" (house speaker) is wrongly translated perhaps due to noises of training data
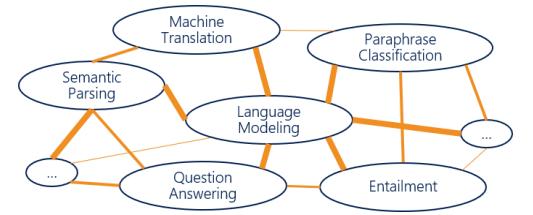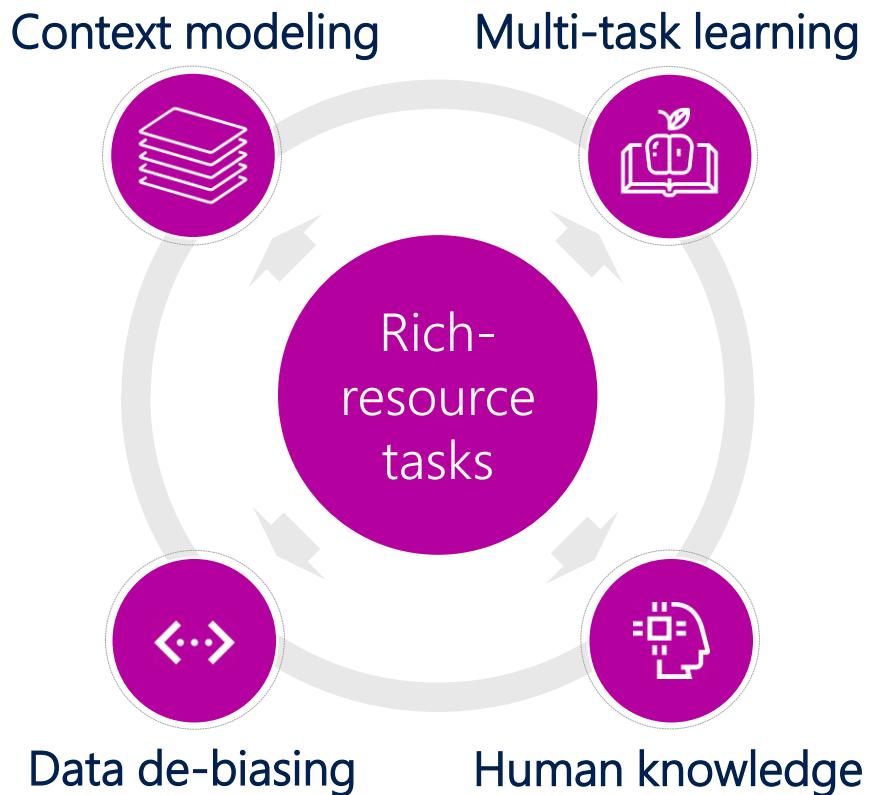
"紫光阁" (Ziguangge) is wrongly missed
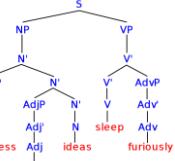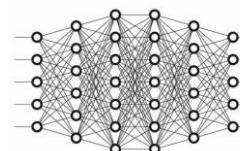
Important topics for rich-resource tasks

Model longer context for document MT, cross-document summarization dialogue system and chatbot.

Context modeling

Multi-task learning

Rich-resource tasks

Data de-biasing

Human knowledge

Further strengthen models with multi-task learning

Alleviate bias issues of training and evaluation datasets for robust models.

Leverage linguistic knowledge and domain knowledge in modelling

Analysis on typical tasks

Rich-Resource

Low-Resource

Multi-Turn

Low-resource scenarios

A task with little training data but highly related to other rich resource tasks

Transfer Learning: learn from other tasks

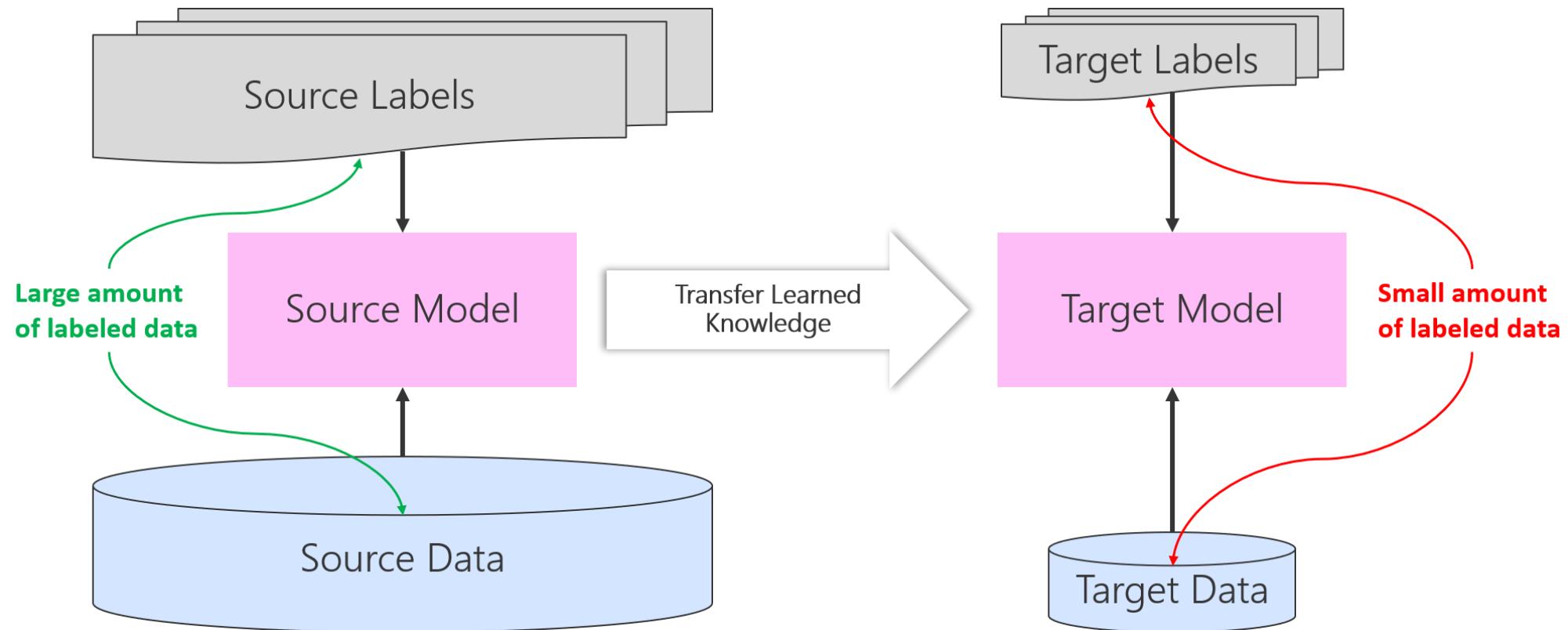A task with little training data in one language but with rich training data in other languages

Cross-lingual Learning: learn from other languages

A task with little training data, without related tasks, without rich training data in other languages

Less or unsupervised Learning: learn from seeds/dictionaries/rules/…
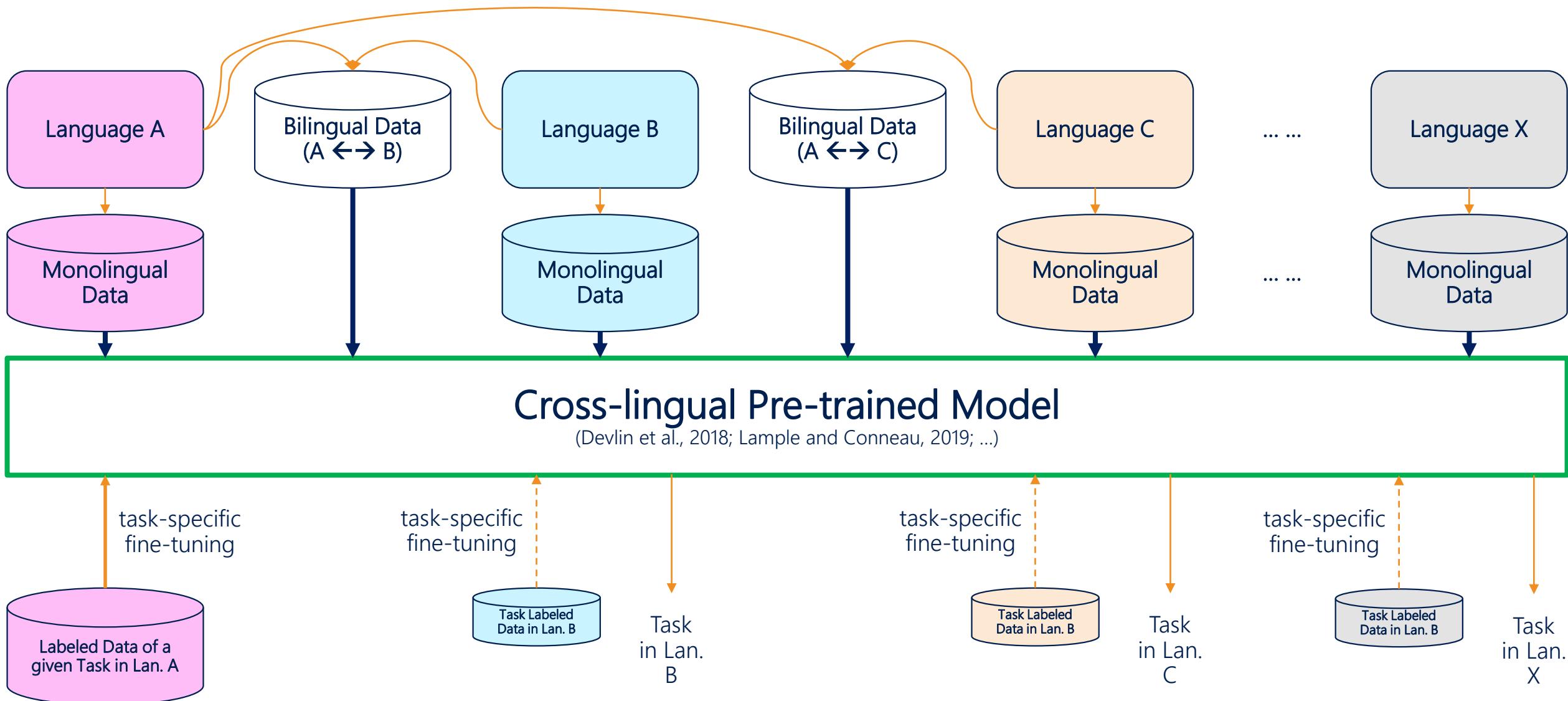
Transfer learning: learn from other tasks
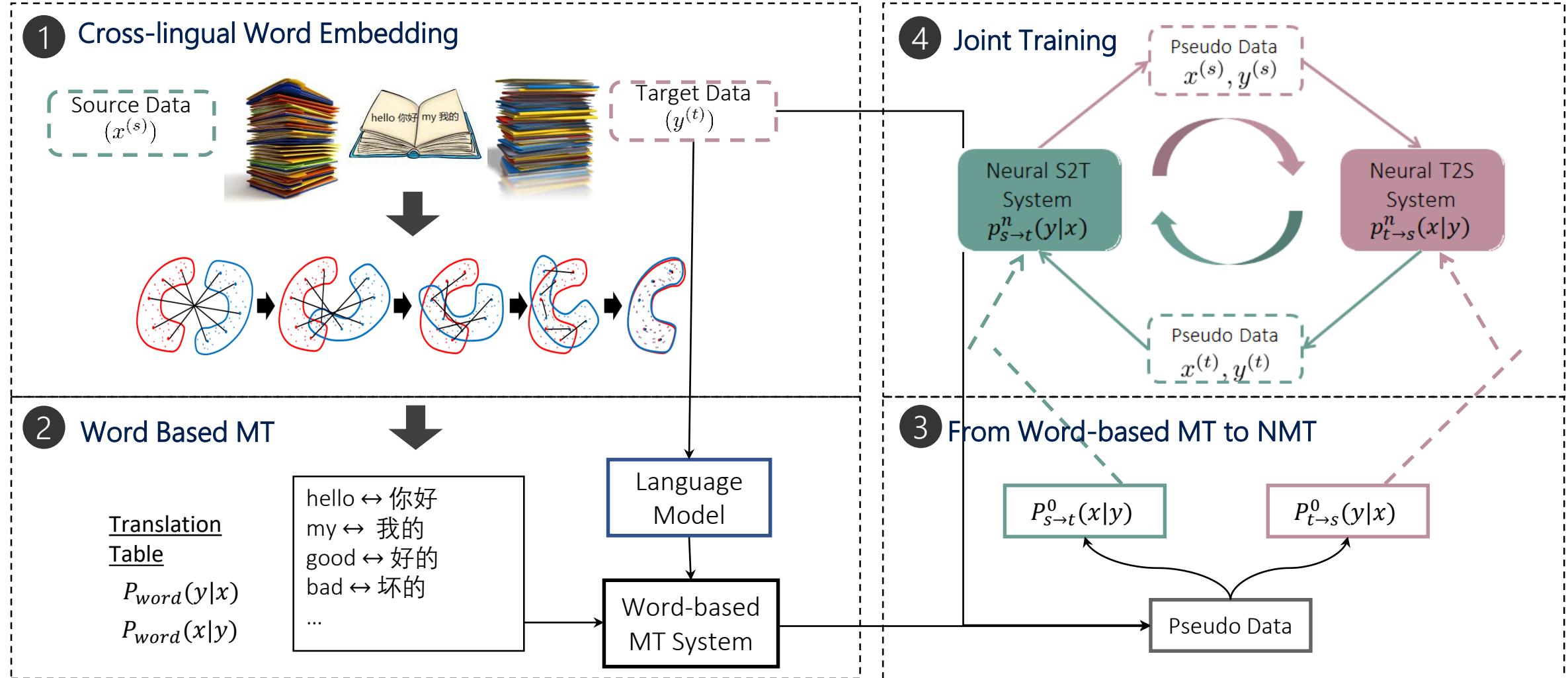

Successful Cases

- LM pre-training (BERT, GPT, XLNet) → various NLP tasks such as QA/MRC, summarization, paraphrase classification, etc.

- ImageNet pre-training (VGGNet, ResNet) → various CV tasks such as visual QA, object detection, scene graph generation, etc.

- …
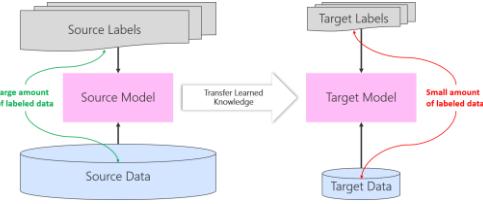
Cross-lingual learning: learn from other languages

Learning with seeds (lexicon, rules, small annotated data)
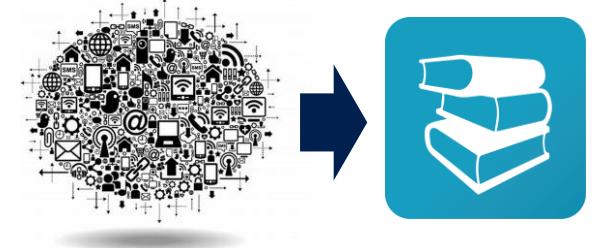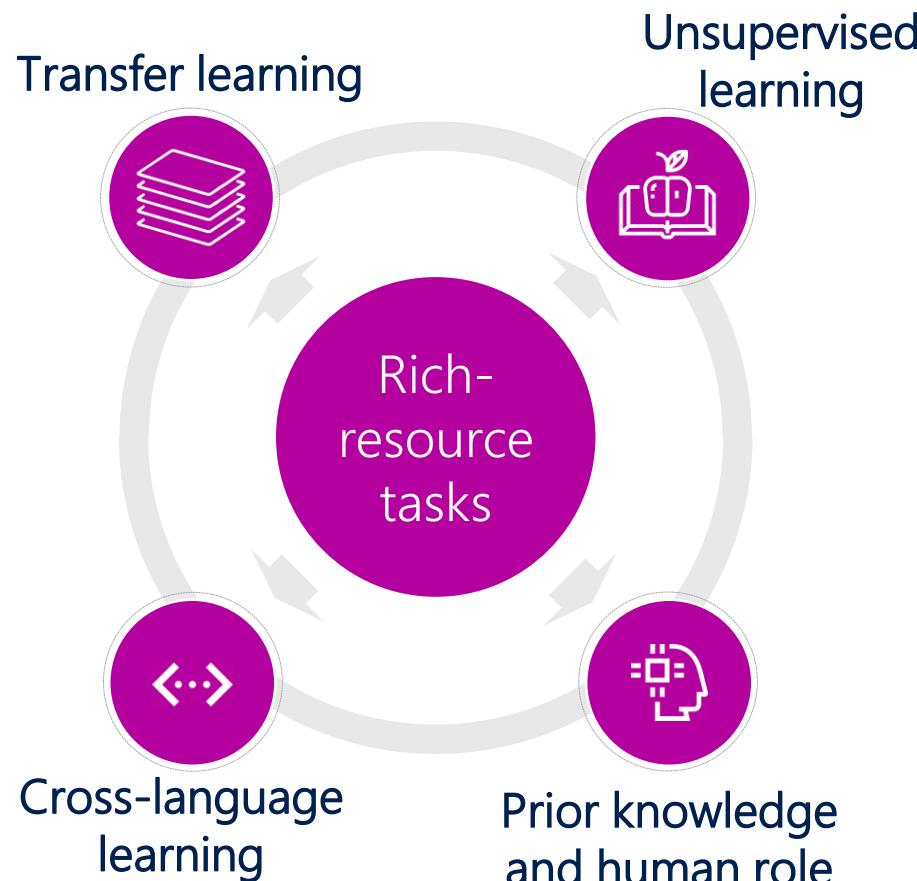
Important topics for low-resource tasks

Transfer knowledge learnt from rich-resource tasks to low-resource tasks, such as BERT and ResNet.

Discover knowledge from unannotated data based on distribution and patterns.

Learn mappings and relationships among langauges for cross-lingual NLP tasks.

Cold-start with seeds such as rules and dictionary, active learning, reinforcement learning

Transfer learning

Unsupervised learning

Rich-resource tasks

Cross-language learning

Prior knowledge and human role

Analysis on typical tasks

Rich-Resource

Low-Resource

Multi-Turn

Weak in modelling common sense and conducting reasoning

Common sense and *reasoning* are required.

Fact: ACL 2019 is held in Florence

Florence	**Q-1**: Where is ACL 2019 held?	Florence
No, because...	**Q-2**: Is ACL 2019 held in France?	
Yes, if...	**Q-3**: Can I attend this conference without an accepted paper?	
Because...	**Q-4**: Why ACL 2019 is held in Florence?	

What kind reasoning is needed?

Tell me the movies with Tom Hanks and Meg Ryan

$\lambda x. \text{film\_film\_actor}(x, \text{Tom Hanks}) \wedge \text{film\_film\_actor}(x, \text{Meg Ryan})$

*Reasoning by **semantic parsing** using **open domain knowledge***

	film_film_actor	
Sleepless in Seattle	film_film_actor	Tom Hanks
Sleepless in Seattle	film_film_actor	Meg Ryan
You've Got Mail	film_film_actor	Tom Hanks
You've Got Mail	film_film_actor	Meg Ryan

Sleepless in Seattle, You've Got Mail,...

When was he born ?

$\lambda x. \text{people\_person\_dateofbirth}(\text{Tom Hanks}, x)$
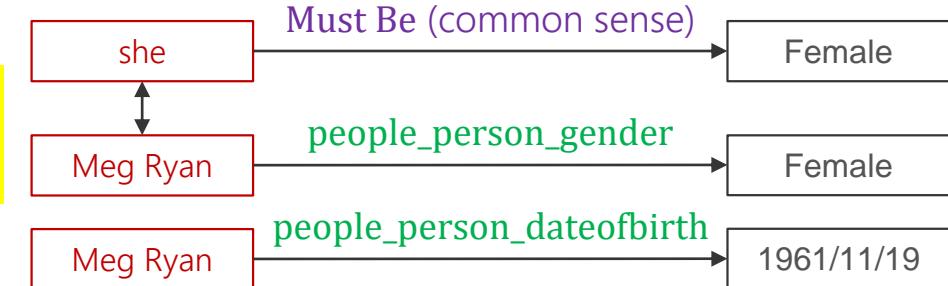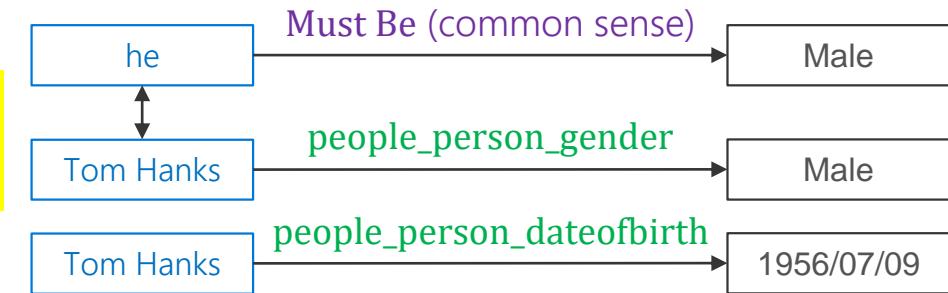
*Reasoning by **semantic parsing** and **coreference resolution** using **common sense**, **open domain knowledge** and **context***

1956/07/09

he	Must Be (common sense)	Male
Tom Hanks	people_person_gender	Male
Tom Hanks	people_person_dateofbirth	1956/07/09

How about her ?

$\lambda x. \text{people\_person\_dateofbirth}(\text{Meg Ryan}, x)$

*Reasoning by **semantic parsing**, **coreference resolution** and **ellipsis resolution** using **common sense**, **open domain knowledge** and **context***

1961/11/19

she	Must Be (common sense)	Female
Meg Ryan	people_person_gender	Female
Meg Ryan	people_person_dateofbirth	1961/11/19

Key components in reasoning

Context

Bought Read Listened Went Clicked Watched Said ...
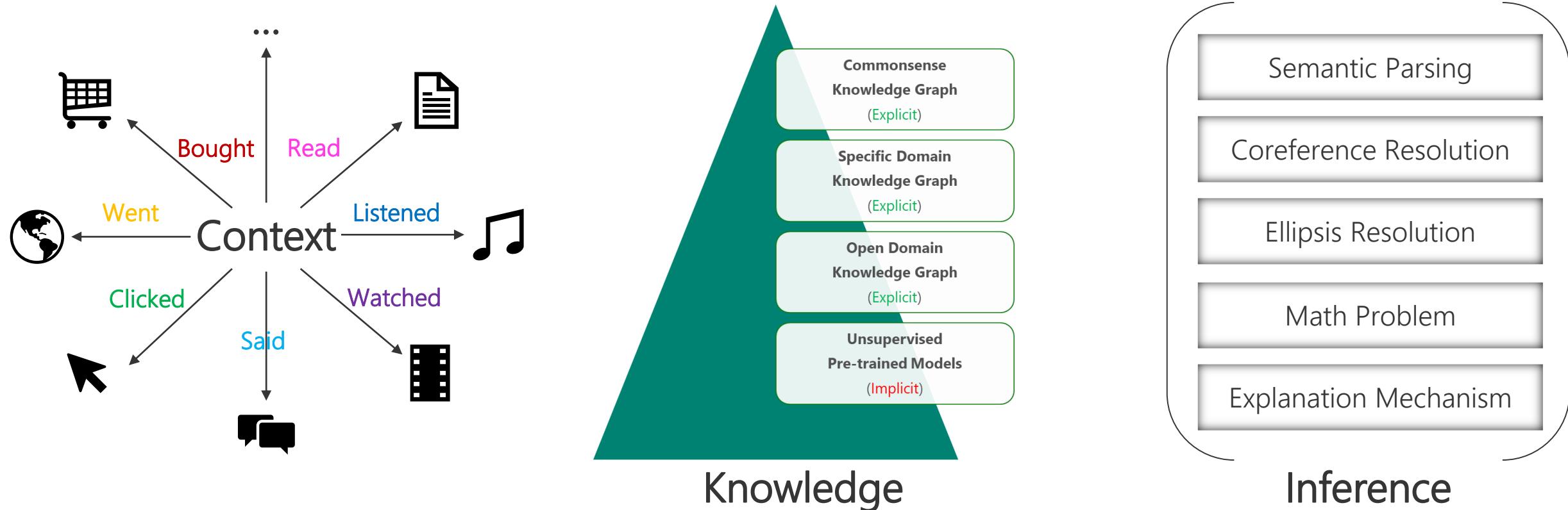
Knowledge

- Commonsense Knowledge Graph (Explicit)
- Specific Domain Knowledge Graph (Explicit)
- Open Domain Knowledge Graph (Explicit)
- Unsupervised Pre-trained Models (Implicit)

Inference

- Semantic Parsing
- Coreference Resolution
- Ellipsis Resolution
- Math Problem
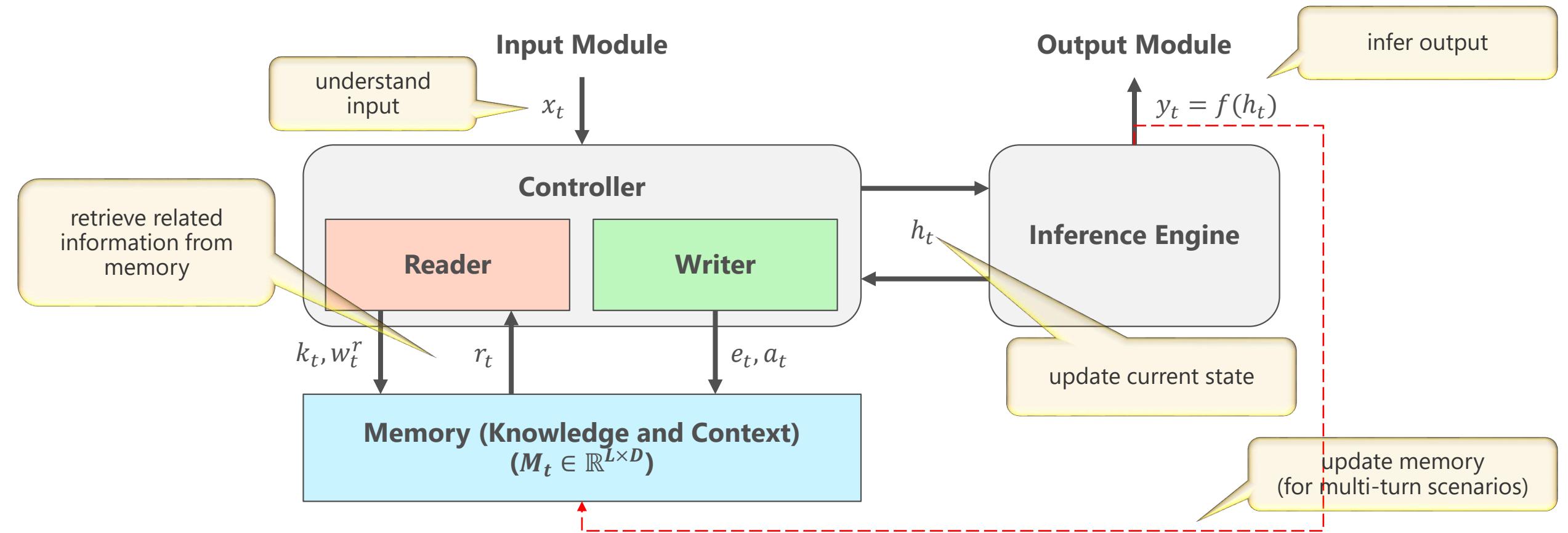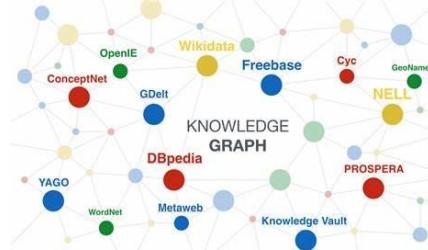- Explanation Mechanism

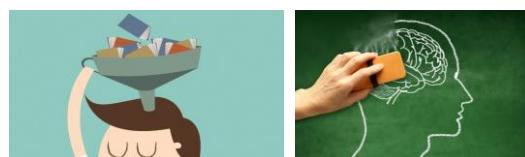# Concept model of reasoning with memory-augmented network

- Alex Graves, Greg Wayne, Ivo Danihelka. **Neural Turing Machines**. arXiv, 2014.
- Sainbayar Sukhbaatar, Arthur Szlam, Jason Weston, Rob Fergus. **End-To-End Memory Networks**. NeurIPS, 2015.
- Alexander Miller, Adam Fisch, Jesse Dodge, Amir-Hossein Karimi, Antoine Bordes, Jason Weston. **Key-Value Memory Networks for Directly Reading Documents**. EMNLP, 2016.
- Adam Santoro, Sergey Bartunov, Matthew Botvinick, Daan Wierstra, Timothy Lillicrap. **Meta-Learning with Memory-Augmented Neural Networks**. ICML, 2016.
- Drew A. Hudson and Christopher D. Manning. **Compositional Attention Networks for Machine Reasoning**. ICLR, 2018.
- ...

Important topics for multi-turn tasks

Extract, represent, conflate and use different types of knowledge and common sense..

Knowledge and Common sense

Inference mechanism
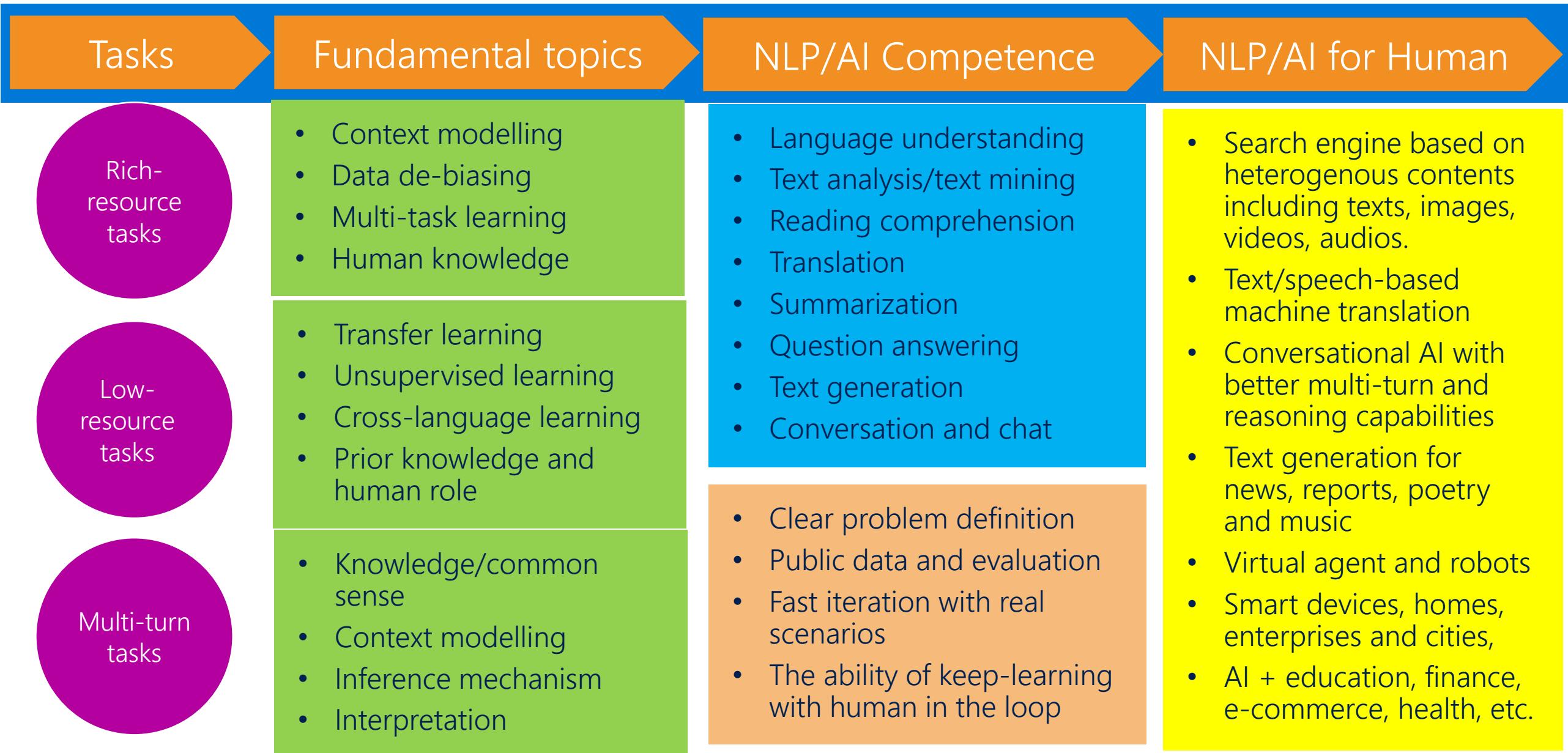
Annotate, model and evaluate the inference procedure.

Rich-resource tasks

Represent, memorize and forget context information in reasoning.
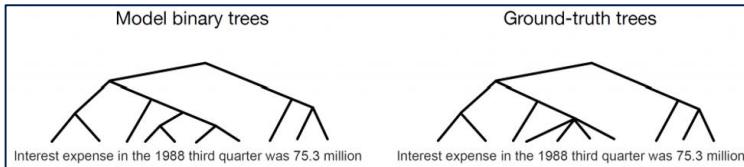
Context modeling

Explainability

Mechanism, debugging, evaluation, visualization

Towards interpretable, knowledgeable, ethical, economical and non-stop-learnable NLP

Tasks	Fundamental topics	NLP/AI Competence	NLP/AI for Human

Rich-resource tasks

- Context modelling
- Data de-biasing
- Multi-task learning
- Human knowledge

Low-resource tasks

- Transfer learning
- Unsupervised learning
- Cross-language learning
- Prior knowledge and human role

Multi-turn tasks

- Knowledge/common sense
- Context modelling
- Inference mechanism
- Interpretation

NLP/AI Competence (blue)

- Language understanding
- Text analysis/text mining
- Reading comprehension
- Translation
- Summarization
- Question answering
- Text generation
- Conversation and chat

(orange)

- Clear problem definition
- Public data and evaluation
- Fast iteration with real scenarios
- The ability of keep-learning with human in the loop

NLP/AI for Human (yellow)

- Search engine based on heterogenous contents including texts, images, videos, audios.
- Text/speech-based machine translation
- Conversational AI with better multi-turn and reasoning capabilities
- Text generation for news, reports, poetry and music
- Virtual agent and robots
- Smart devices, homes, enterprises and cities,
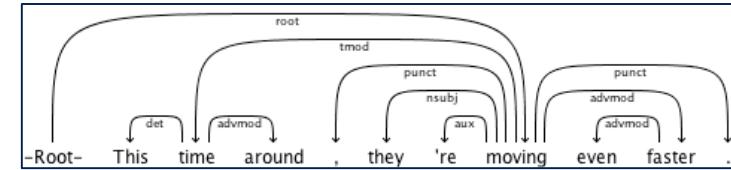- AI + education, finance, e-commerce, health, etc.

Deep learning and linguistics boost each other

Deep learning models can find hidden syntactic tree structures of natural language sentences in an unsupervised way.
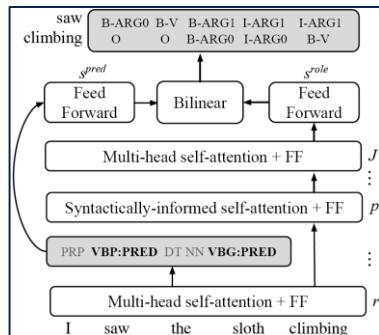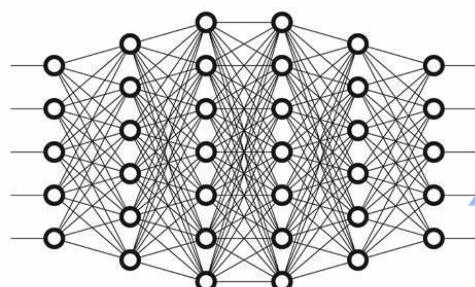
Emma Strubell, Patrick Verga, Daniel Andor, David Weiss, Andrew McCallum. *Linguistically-Informed Self-Attention for Semantic Role Labeling*. EMNLP, 2018.

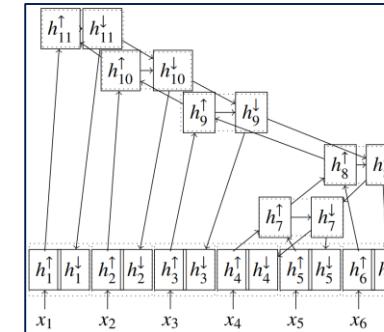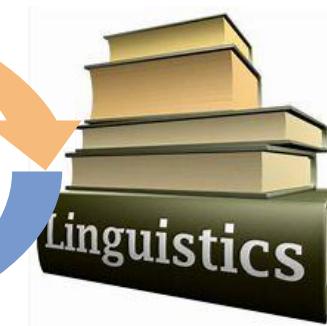Deep learning models can predict better syntactic tree structures of natural language sentences in a supervised way.

Danqi Chen and Christopher Manning. *A Fast and Accurate Dependency Parser Using Neural Networks*. EMNLP, 2014.

NN helps Linguistics

Linguistics helps NN

Linguistic information can improve NLP tasks as input signals.

Yikang Shen, Shawn Tan, Alessandro Sordoni, Aaron Courville. *Ordered Neurons: Integrating Tree Structures into Recurrent Neural Networks*. ICLR, 2019.

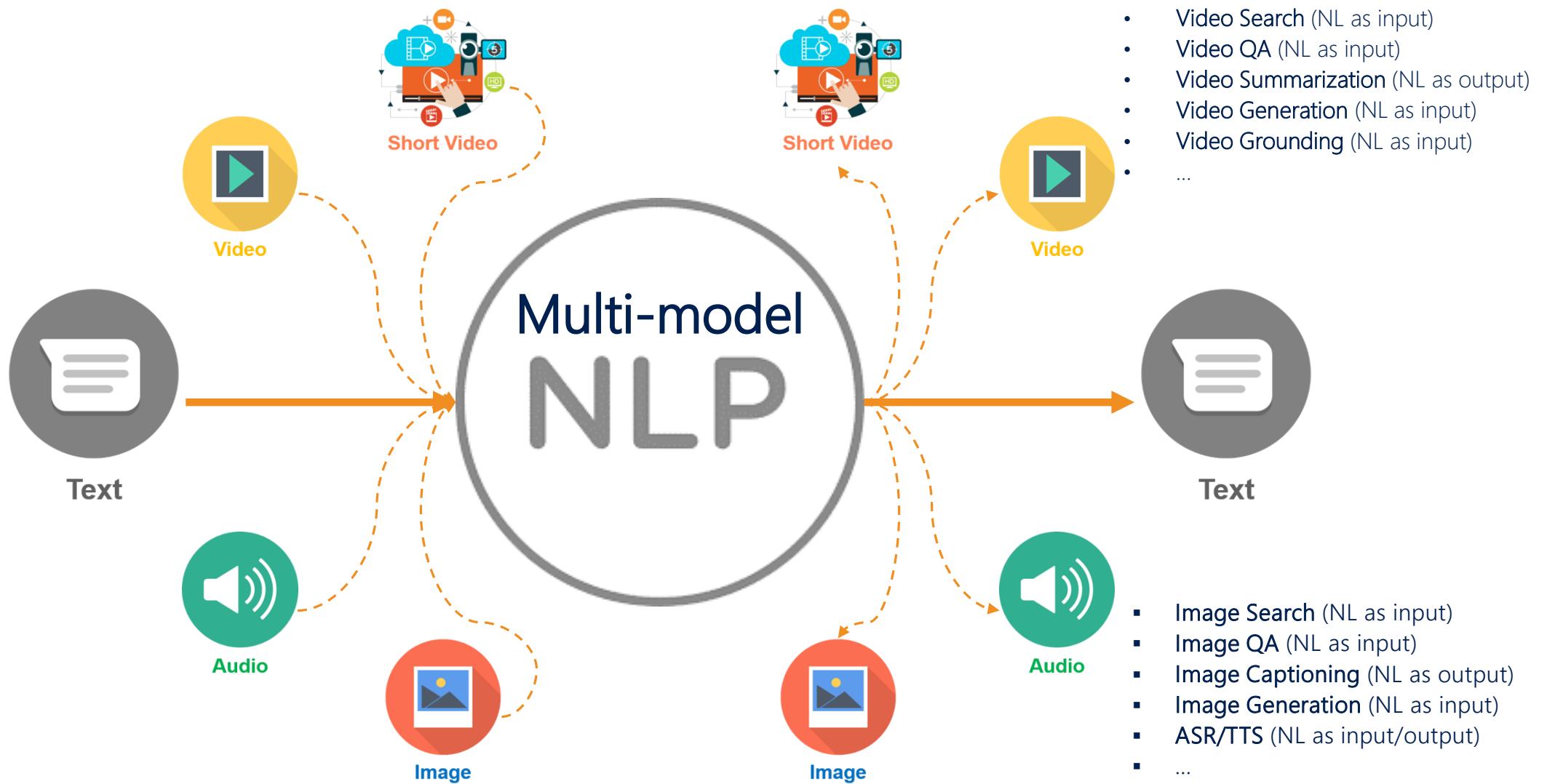Linguistic information can improve NLP tasks by designing syntactic-aware neural network structures..

Huadong Chen, Shujian Huang, David Chiang, Jiajun Chen. *Improved Neural Machine Translation with a Syntax-Aware Encoder and Decoder*. ACL, 2017.

Multi-modality processing to enrich input and output

Short Video

Video

Text

Audio

Image

Multi-model NLP

- Video Search (NL as input)
- Video QA (NL as input)
- Video Summarization (NL as output)
- Video Generation (NL as input)
- Video Grounding (NL as input)
- …

- Image Search (NL as input)
- Image QA (NL as input)
- Image Captioning (NL as output)
- Image Generation (NL as input)
- ASR/TTS (NL as input/output)
- …

Embrace the bright future with efforts from the whole society

Computing power
- Advanced chip and machine
- Powerful architecture and cloud computing
- Efficient resources management
- Model compression and acceleration

Talent
- Reform the curriculum
- Emphasize the system building capability
- Balance on following the trends and challenging the trends
- International view

Data
- Open-source data and shared tasks
- Efficient collection and annotation
- Data de-biasing and de-noising
- Privacy preserved learning

Collaboration
- University-enterprise
- Multi-domain and disciplinary(multi-modal processing, linguistics, brain science, ethics, big data,...)
- International partnership
- Eco-system with technical provider and users

Models
- New methods of supervised, less supervised and unsupervised
- Further development of pre-trained models
- Incorporating NN + Knowledge
- Reasoning and interpretability

Application
- Understand the needs of real scenarios of various verticals
- Result-oriented problem solving
- Human in the loop
- Market analysis and business model

> "We climbed together until I saw some beautiful things appearing in the vault through a circular hole. So we went out to see the stars.
>
> -- Dante Alighieri "

Thank you ~