

Representations and their Matching

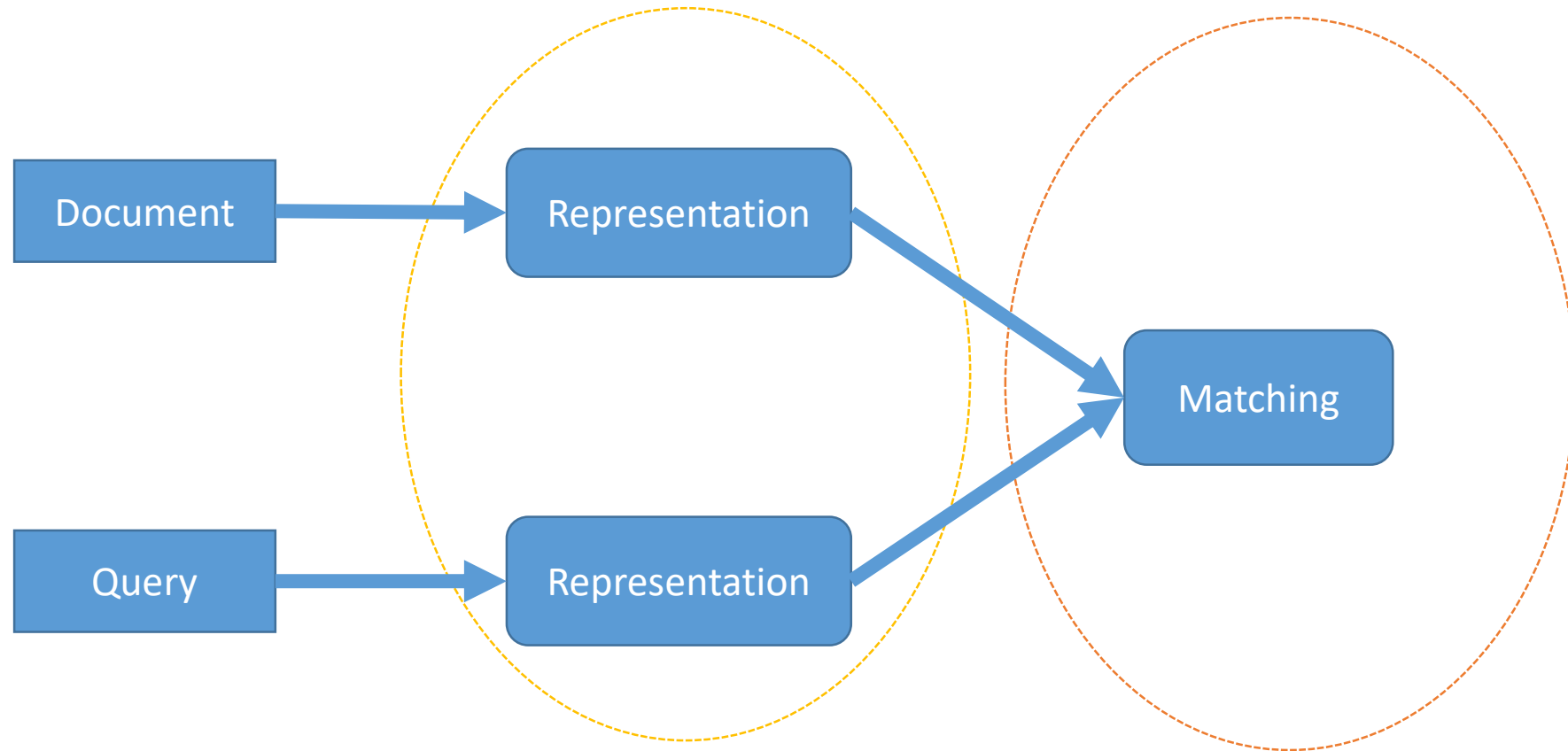
Overview of my research



Benyou Wang

Supervised by Prof. Massimo Melucci

Examples for Representation & Matching



Key concerns

- How to build a good **representation**

- Language model
- Neural representation
- Multi-modality

[End-2-End QLM AAAI 2018] [QMWF-LM AAAI 2018]

[Complex Word Embedding ACL 2018][TextZoo Arxiv 2018][Quantum Attention]

[Image caption IJCAI2018][Sentimental analysis TCS]



- How to make a good matching

- Ad hoc retrieval
- Question Answering
- GAN for Matching
- Recommendation
- Customer Service

[Quantum Query Expansion Entropy 2018]

[End-2-End QLM AAAI 2018]

[IRGAN SIGIR 2017]

[Long + short-term Profile IJCAI 2018]

[QA NLPCC 2016&NLPCC 2018]



- Tools

- Quantum Concepts or Quantum-inspired method
- GAN/Multitask/DNN/Eye tracking





Faster language model with CNN



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Model	Test PPL	Hardware
Sigmoid-RNN-2048 (Ji et al., 2015)	68.3	1 CPU
Interpolated KN 5-Gram (Chelba et al., 2013)	67.6	100 CPUs
Sparse Non-Negative Matrix LM (Shazeer et al., 2014)	52.9	-
RNN-1024 + MaxEnt 9 Gram Features (Chelba et al., 2013)	51.3	24 GPUs
LSTM-2048-512 (Jozefowicz et al., 2016)	43.7	32 GPUs
2-layer LSTM-8192-1024 (Jozefowicz et al., 2016)	30.6	32 GPUs
BIG GLSTM-G4 (Kuchaiev & Ginsburg, 2017)	23.3*	8 GPUs
LSTM-2048 (Grave et al., 2016a)	43.9	1 GPU
2-layer LSTM-2048 (Grave et al., 2016a)	39.8	1 GPU
GCNN-13	38.1	1 GPU
GCNN-14 Bottleneck	31.9	8 GPUs

Yin W, Kann K, Yu M, et al. Comparative study of cnn and rnn for natural language processing[J]. arXiv preprint arXiv:1702.01923, 2017.

https://github.com/wabyking/Gated_CNN_for_language_Modeling,

Dauphin Y N, Fan A, Auli M, et al. Language Modeling with Gated Convolutional Networks. **ICML 2017**: 933-941.

TextZOO [Benyou et.al Arkiv 2018]

Models

- ✓ FasText
- ✓ CNN (Kim CNN, Multi-Layer CNN, Multi-perspective CNN, Inception CNN)
- ✓ LSTM (BILSTM, StackLSTM, LSTM with Attention)
- ✓ Hybrids between CNN and RNN (RCNN, C-LSTM)
- ✓ Attention (Self Attention / Quantum Attention)
- ✓ Transformer - Attention is all you need
- ✓ Capsule
- ✓ Quantum-inspired NN
- ConS2S
- Memory Network

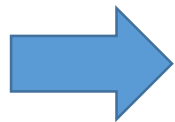
Datasets

- **IMDB**
- **MR**
- CR
- MPQA
- SST1
- SST2
- Subj
- TREC

Understand the content of a movie [Wei et.al IJCAI 2018]



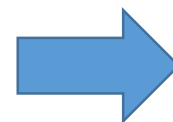
Keyframe
extraction



Key frames



Image Caption



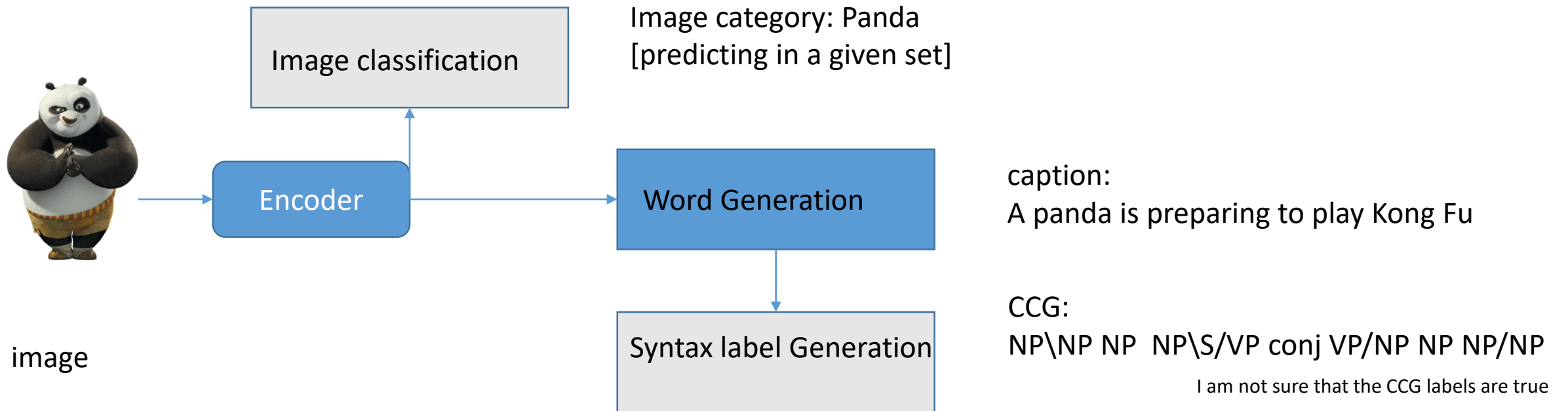
Textual features

A panda is preparing to play Kong Fu

A panda is eating noodles on the table

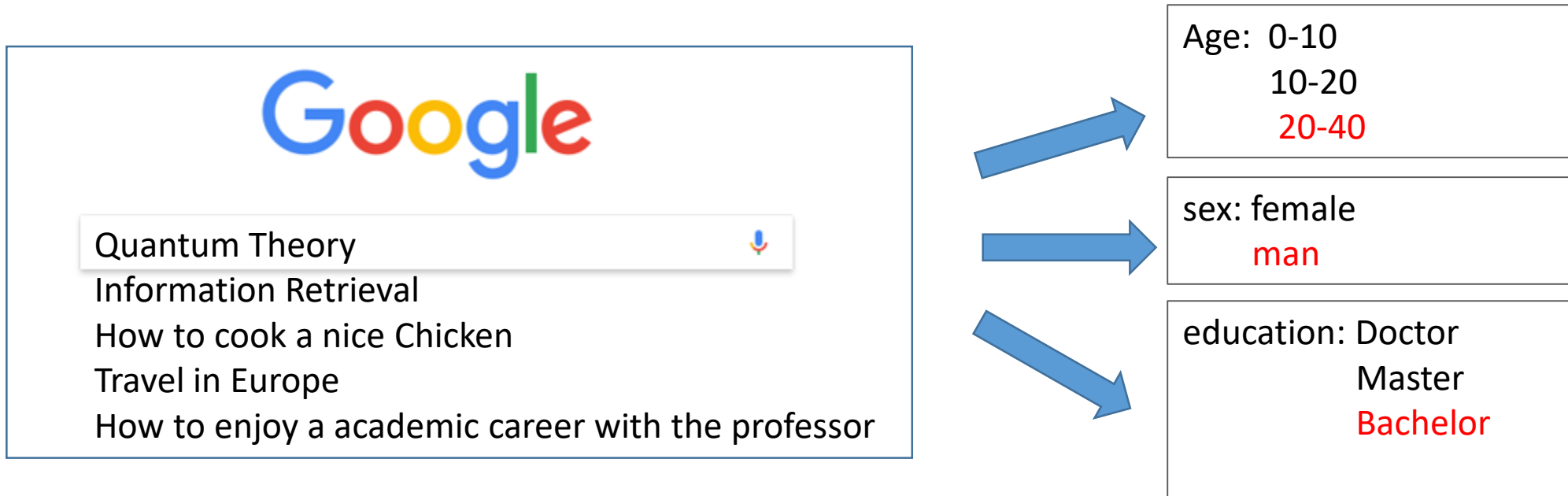
Two pandas are laughing

Multi-task for encoder and decoder

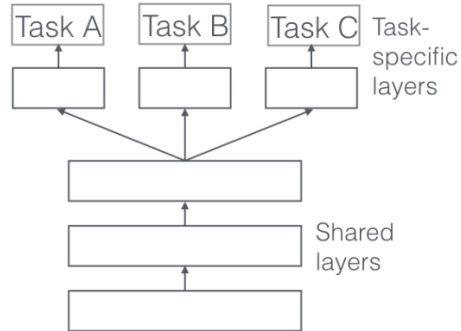


Multi-task approach

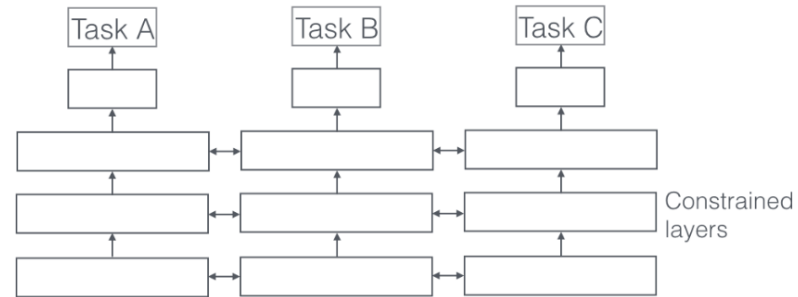
Example: Using **query logs** to infer the user profile like age, sex and education background



Two Typical Multi-task Paradigms



Hard parameter sharing



Soft parameter sharing

The distance between the parameters of the model is then regularized in order to encourage the parameters to be similar

Representation inspired by Quantum

- Quantum Probability Space
- Quantum Many-body wave function and Tensor language model
- Quantum Capsule Models (Using direction, instead of numerical number)
- Quantum two-state Formalism

Melucci M. An Algorithm to Calculate a Quantum Probability Space[J]. arXiv preprint arXiv:1710.10158, 2017.

Zhan Su, Peng Zhang, Lipeng Zhang, Benyou Wang, et.al. A Quantum Many-body Wave Function Inspired Language Modeling Approach, submitted to CIKM 2018.

Pestun V, Vlassopoulos Y. Tensor network language model[J]. arXiv preprint arXiv:1710.10248, 2017.

Sabour S, Frosst N, Hinton G E. Dynamic routing between capsules[C]//Advances in Neural Information Processing Systems. 2017: 3856-3866.

Key concerns

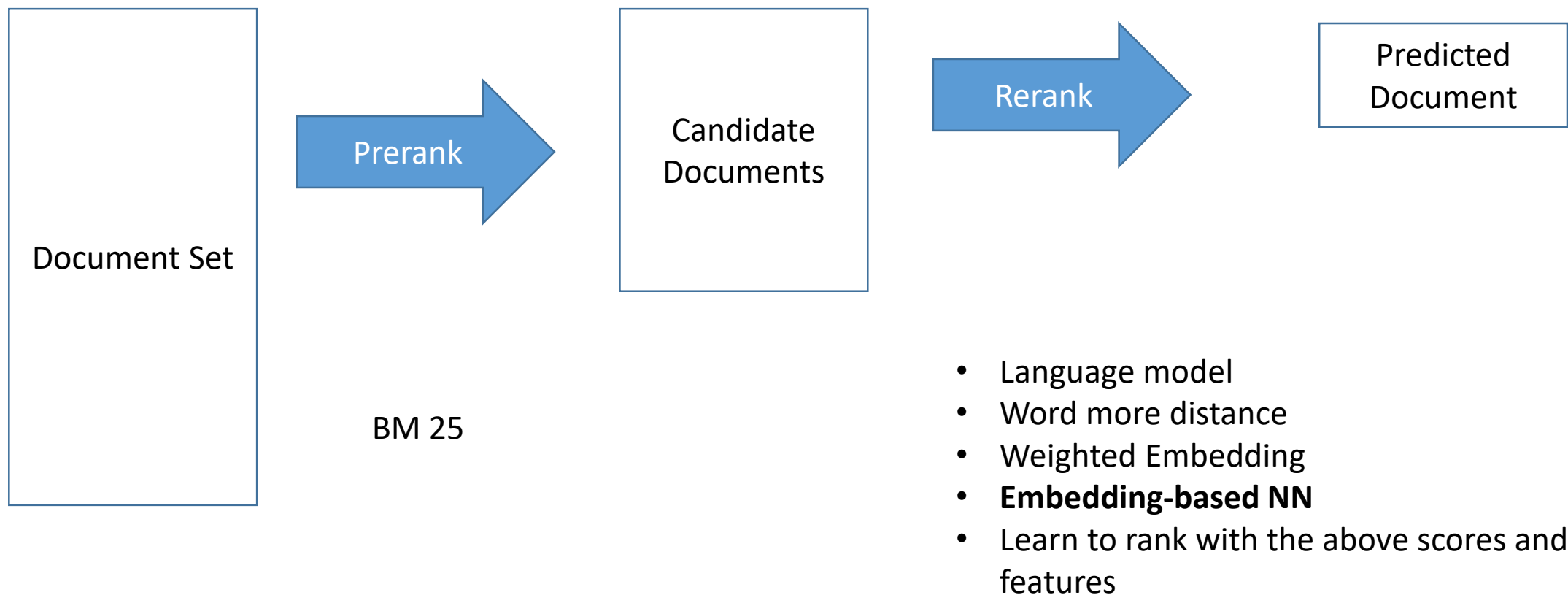
- How to make a good matching

- Ad hoc retrieval [Quantum Query Expansion Entropy 2018]
- Question Answering [End-2-End QLM AAAI 2018]
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Customer Service in Tencent



Given a set of Frequent Question Answer Pairs, and answer a new question from the give QA collection.

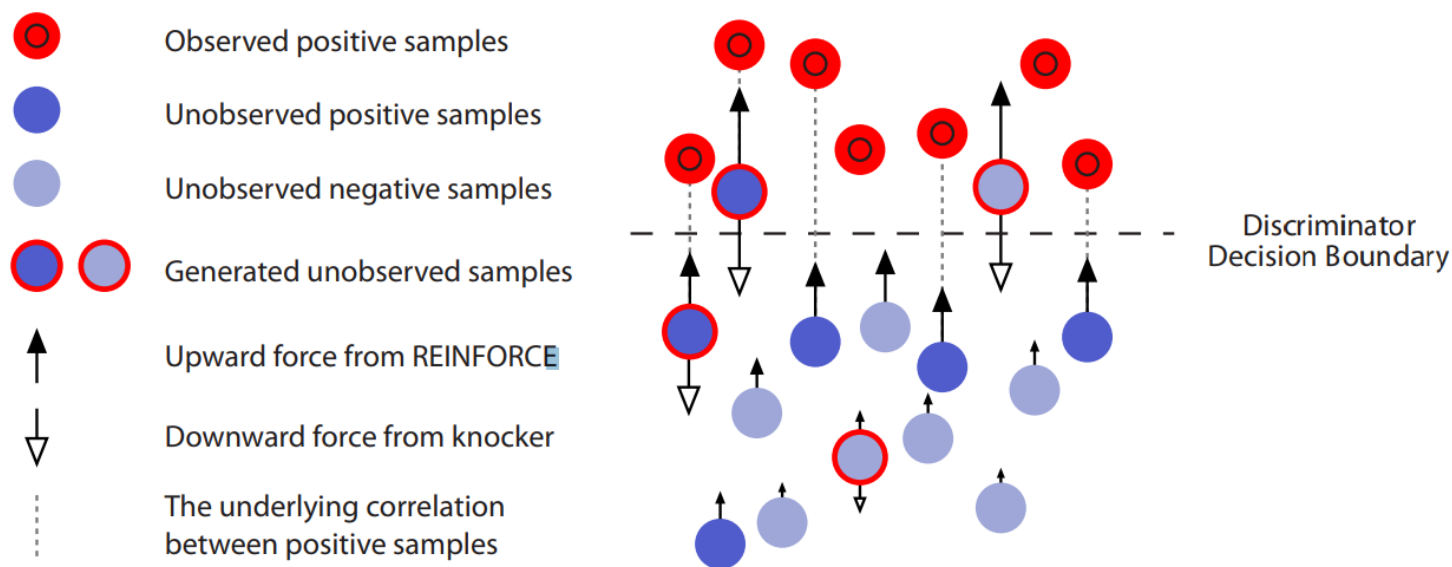


QA matching tasks from a industrial view

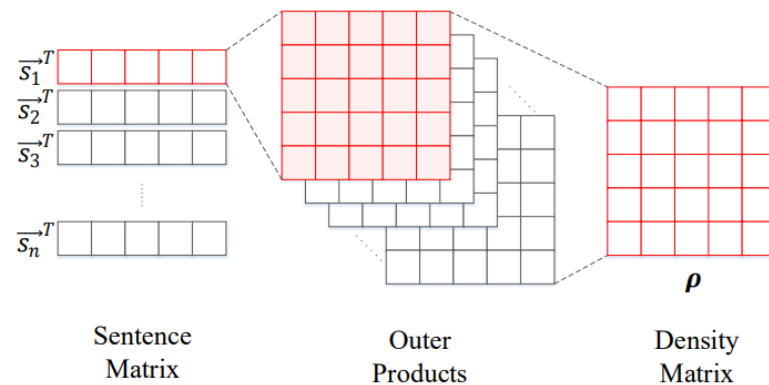
- Count-based VS Embedding-based
 - Count-based bag-of-word models are more robust
 - Embedding-based models needs supervised corpus.
- If you have **enough more high-quality supervised matching pair**. It should achieve much better performance

IRGAN [Jun et.al. SIGIR 2017]

Adjust the original unsupervised models via the feedback from the supervised ones



End2end Language model [Peng et.al AAI 2018]

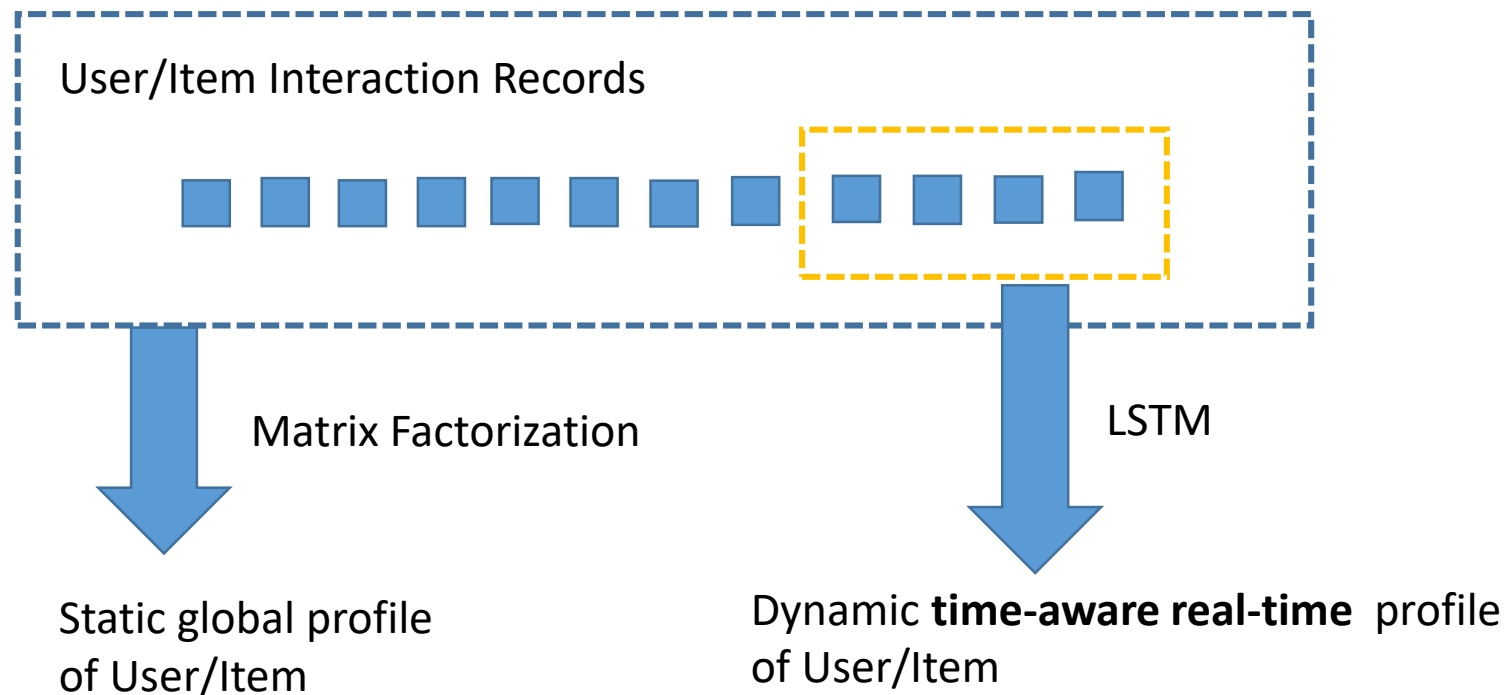


Matching with two matrices

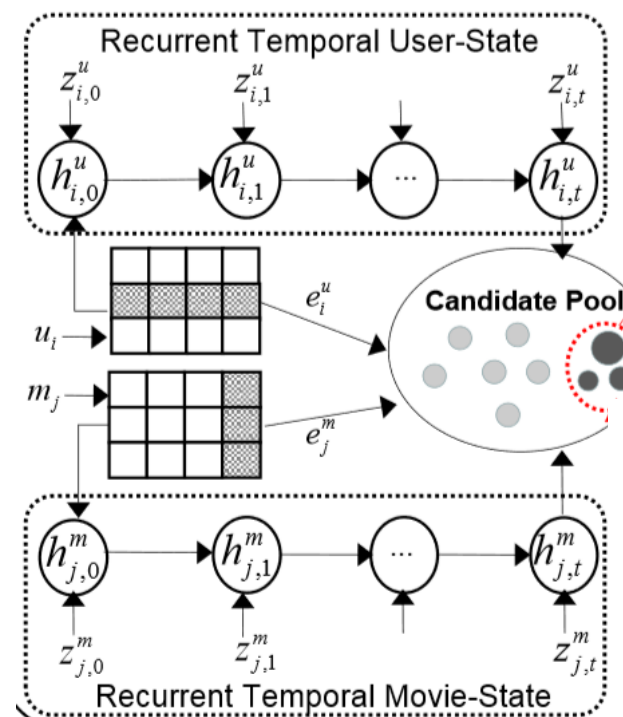
- $tr(\rho_1\rho_2)$
- CNN over $\rho_1\rho_2$

*More **non-quantumnic**, the better in performance*

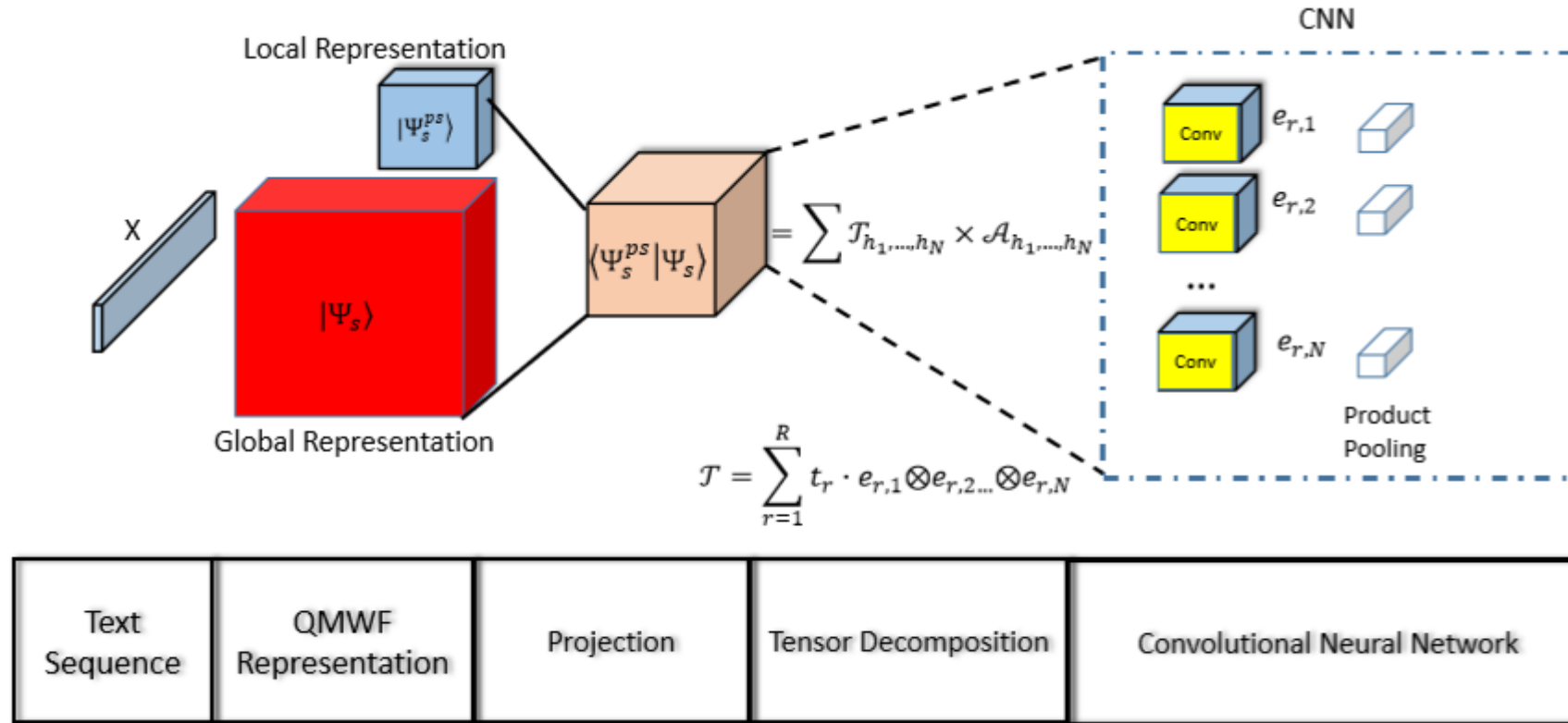
Long/short profile for Item/User [Wei et.al IJCAI 2018]



Long/short profile for Item/User [Wei et.al IJCAI 2018]



Quantum Many body language for NN



Use CNN to **approximate** Tensor Decomposition in the projection of Quantum Many-Body Language Function

Future

I am Open with the research topics

- Quantum Probability Space
- Contextual Quantum language model in Dynamics
- Capsule Network with Quantum mechanism
- Develop unsupervised IR models with adversarial method
-

Thanks