



# Transfer Learning for Biomedical Relation Extraction

Block seminar

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# Outline

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- Introduction
  - Evaluation setting
  - Results (Preview)
- Solution presentations
  - Group 3: CNNs with pre-trained biomedical word embeddings (Hanjo, Danielle, Dennis)
  - Group 1: (Bio-) BERT for Relation Extraction (Phuc, Duy)
- Feedback

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# Introduction

# Training data

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- You have **two distinct** training data sets for the competition
  - Both corpora contain **protein-protein interactions**
  - Usage of a uniform, consistent XML-format

	<b>AIMed</b>	<b>BioInfer</b>
#Documents	180	669
#Sentences	1554	894
#Entities	3407	3611
Distinct	933	952
#Pairs	4680	8043
Positive	800	2109
Negative	3880	5934

# Evaluation setting

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- Your approaches will be evaluated on **hold-out sets** (~ test data) of the two corpora
  - The (XML-) format is **equivalent** to the training data set
  - Of course, the test sets **do not contain** the gold standard labels
    - => No *interaction* attribute!

```
<pair e1="AIMed.d0.s7.e1" e2="AIMed.d0.s7.e3" id="AIMed.d0.s7.p4" />
```

```
<pair e1="AIMed.d0.s7.e2" e2="AIMed.d0.s7.e3" id="AIMed.d0.s7.p5" />
```

# Evaluation setting

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- Your approaches will be evaluated on **hold-out sets** (~ test data) of the two corpora

	<b>AIMed</b>	<b>BioInfer</b>
#Documents	44	167
#Sentences	389	286
#Entities	795	810
Distinct	288	298
#Pairs	1095	1623
Positive	191	425
Negative	904	1198
	17% Positive	26% Positive

# Results (preview)

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- AiMed

Team	Model	P	R	F1
Group 3 (CNN)	Without-Embs	0.250	0.330	0.284
Group 1 (BERT)	Alibaba-SciBert	<b>0.711</b>	0.759	0.734
	Lee-BioBert	0.672	<b>0.859</b>	<b>0.754</b>
	Lin-Bert	0.674	0.780	0.723

- BioInfer

Team	Model	P	R	F1
Group 3 (CNN)	Without-Embs	0.365	0.191	0.250
Group 1 (BERT)	Alibaba-SciBert	<b>0.831</b>	0.671	<b>0.742</b>
	Lee-SciBert	<b>0.831</b>	0.649	0.729
	Lin-BioBert	0.815	<b>0.675</b>	0.739

# Comparison with competitors

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- AiMed

Author	Model	P	R	F1
	Lee-BioBert	0.672	0.859	0.754
Quan et al. [1]	Multi-Channel CNN	0.764	0.690	0.725
Hsieh et al. [2]	Bi-LSTM	0.788	0.752	0.769
Yadav et al. [3]	Bi-LSTM on DT + Words	0.911	0.822	<b>0.865</b>

- BioInfer

Author	Model	P	R	F1
	Alibaba-SciBert	0.831	0.671	0.742
Quan et al. [1]	Multi-Channel CNN	0.813	0.781	0.796
Hsieh et al. [2]	Bi-LSTM	0.870	0.874	<b>0.872</b>
Yadav et al. [3]	Bi-LSTM on DT + Words	0.724	0.831	0.774

\* Figures taken from original published results



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# Feedback

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**Thank you for your attention!**

# References

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- [1] Quan, Chanqin, et al. "*Multichannel convolutional neural network for biological relation extraction.*" *BioMed research international* 2016 (2016).
- [2] Hsieh, Yu-Lun, et al. "Identifying protein-protein interactions in biomedical literature using recurrent neural networks with long short-term memory." *Proceedings of the eighth international joint conference on natural language processing (volume 2: short papers)*. 2017.
- [3] Yadav, Shweta, et al. "Feature Assisted bi-directional LSTM Model for Protein-Protein Interaction Identification from Biomedical Texts." *arXiv preprint arXiv:1807.02162* (2018).