# **Deep Learning for NLP 2020**



# **Shared Task Group 30, Yi Cui, Zixuan Chen**

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WBS	Task description	Start date	Finish date	<b>Progress</b>	22	23	24 2	25	26	27 2	8 2	29 3	0 1	2	3	4	5	6	7	8	9	10	11	12	13	3 14
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1	Task understanding	2020.6.27	2020.6.28	100%						= :	=															
1.1	research literature	2020.6.27	2020.6.27	100%						=																
1.2	Template implementation	2020.6.27	2020.6.28	100%						= :	=															
_																										
2	Modeling and Development	2020.6.28	2020.7.3	81%						:	=	= =	=	=												
2.1	expand model	2020.6.28	2020.7.2	75%							=	= =	=													
2.2	Hyperparameter optimization	2020.6.28	2020.7.2	100%						-	=	= =	=	<b>=</b> =												
2.3	Result evaluation and optimization	2020.6.28	2020.7.3	70%							=	= =	=											_		
_																					Ш					
3	Report and Presentation	2020.7.9	2020.7.9	100%																						
3.1	write report	2020.7.9	2020.7.9	100%																	1					
3.2	prepare presentation	2020.7.9	2020.7.9	100%																	1					

43	qwerqwer	5	07/03/20	0.2960 (41)
44	m-bodensohn	1	07/01/20	0.2925 (42)



# **Agenda**



- Task
- Model
- Result

#### **Task**



- Semantic similarity:
  - a metric defined over a set of documents or terms, where the idea of distance between items is based on the likeness of their meaning or semantic content as opposed to lexicographical similarity [1]
- Adversarial attacks:
  - disemvoweling: Some of the vowels have been removed. [2]
  - visual attacks: Similar-looking characters have been replaced by each other. [2]

Three dgs playing in th snow.



Three sogs run in the sno $\mathbb{C}$ .



#### **Task**

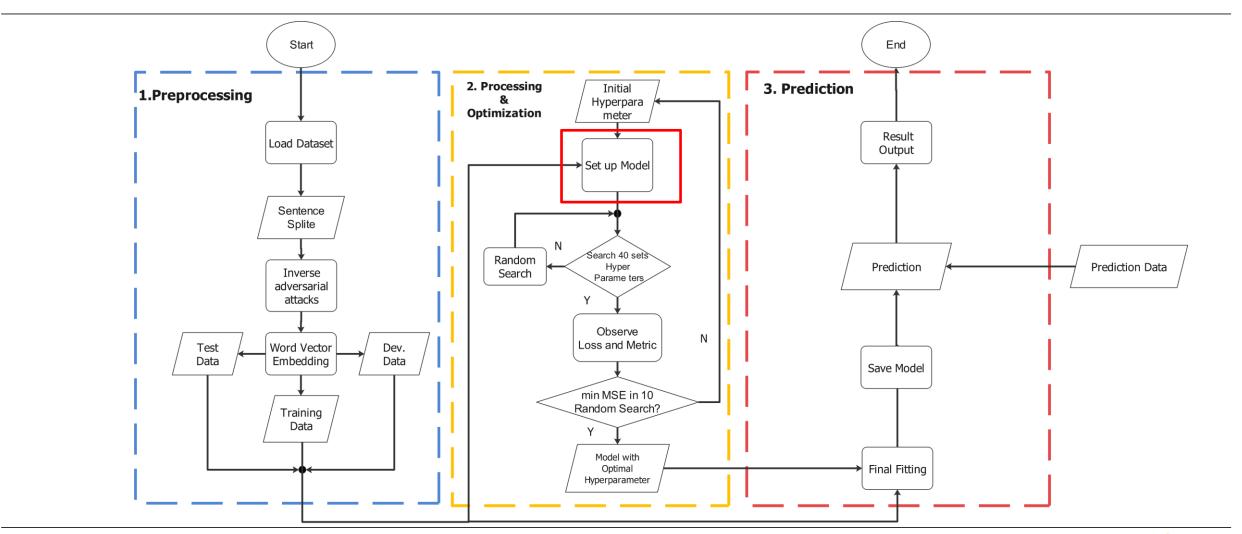


- Approach
  - Preprocessing
    - inverse adversarial attacks → reduce OOV
    - **Word vector** → Embedding words with wiki-news-300d-1M.vec
  - Processing
    - apply MLP with Random Search → find appropriate mapping
    - extend MLP with LSTM → prognosticate semantic similarity with context



# Model (Chart Flow)



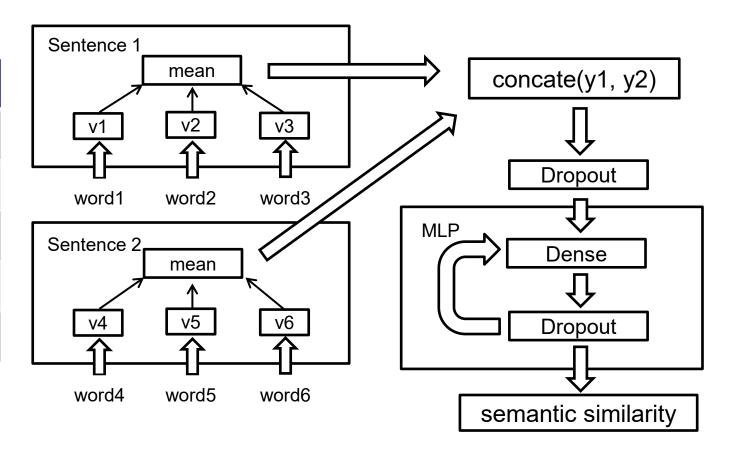


# Model (MLP)



#### Hyper Parameter

Туре	Value						
Dense size	337, 171						
Dropout rate	0.5, 0.1, 0.2						
Activation	'relu', 'selu', 'selu'						
Loss	MSE						
Metric	MSE						



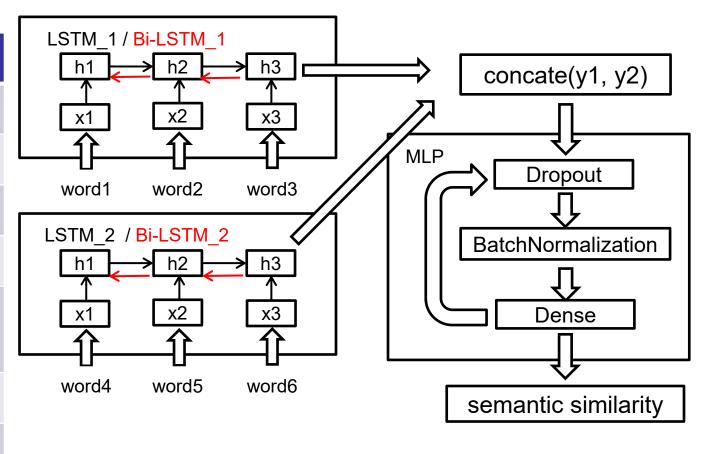


# Model (LSTM and Bi-LSTM)



Hyper Parameter

Туре	Value
Emb_size	56
LSTM Units	300
Dense size	440, 201, 107, 58,
Dropout rate	0.2, 0.2, 0.1, 0.2, 0.2
Activation	'selu', 'hard sigmoid', 'relu', 'hard sigmoid',
Loss	MSLE
Metric	MSE





#### **Overview of result**



Score from CodaLab:

MLP > LSTM > Bi-LSTM

	MLP	LSTM + MLP	Bi-LSTM + MLP					
Development	0.3982	0.3092	0.2844					
Test	0.2960	None	None					

Hidden Layer (Dense) less than 3 → Otherwise Overfitting

- Model is sensitive to Data (OOV)
  - > inverse attacks are more important than modeling
- Time management are essential



#### Reference



- [1]: <a href="https://en.wikipedia.org/wiki/Semantic similarity">https://en.wikipedia.org/wiki/Semantic similarity</a>
- [2]: Deep Learning for NLP 2020 Shared Task Organization



# Thanks for your attention



# Back up



