# Neural Adversarial Training for Semi-supervised Japanese **Predicate-argument Structure Analysis**



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### Introduction & Task

We propose a novel GAN-like model: generator and validator networks that enable the model to learn from an unlabeled corpus.

- Japanese predicate-argument structure (PAS) analysis is a task to find an argument for each case of a predicate.
- PAS analysis relies on the numerous pairs of predicates and their arguments depending on their contexts. However, annotated corpora for PAS analysis are very limited.
- Our validator scores the generator outputs and enables the generator to learn predicate and argument pairs from unlabeled corpora.
- タクシー<u>がNOM</u> 客<u>をACC</u> 駅にDAT 送った takushi-ga kyaku-wo eki-ni okutta. A taxi carried passengers to the station. (2) その 列車は 荷物をACC 運んだ。 sono ressha-wa nimotsu-wo hakonda. The train also carried baggages (3) タクシー<u>がNOM</u> 客<u>をACC</u> 乗せた とき 事故<u>にDAT</u> 巻き込まれた。 takushi-ga kyaku-wo noseta toki jiko-ni makikomareta When the taxi picked up passengers, it was involved in the accident. (4) この列車には乗れません。 kono ressha-ni-wa noremasen.

	Predicate	NOM	ACC	DAT	
	送った	タクシー	客	駅	
	okutta	takushi	kyaku	eki	
	sent/carried	taxi	passenger	station	
	運んだ	「 列車	荷物		
	hakonda	ressha	nimotsu	NULL	
	carried	_ train _	baggage		
	乗せた	タクシー	客		
a.	noseta	takushi	kyaku	NULL	
	picked up	taxi	passenger		
	巻き込まれた	「タクシー】		事故	
	makikomareta	takushi	NULL	jiko	
	was involved	_ taxi _		$\operatorname{accident}$	
	乗れません	「あなた ]		列車	
	noremasen	anata	NULL	ressha	
	can not take	you		$\operatorname{train}$	

 $p_{\mathrm{arg}_3,\mathrm{pred}}^{\mathrm{case}_k}$ 

(1, 1, 1)

You can not take this train Examples of Japanese PAS analysis. For predicates in a sentence, the model predicts NOM, ACC and DAT case roles of arguments.

### Model

Our entire model consists of the generator network that predicts the arguments for each predicate and the validator network that scores the outputs of the generator network.

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train

Bi-LSTM

Т

**Bi-LSTM** 

т

この

this

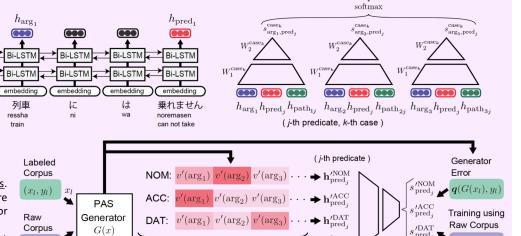
embedding

#### Generator

- · The generator consists of two neural networks: the bi-LSTM based sentence encoder and the FNNs for argument selection.
- There are three ENNs for NOM ACC and DAT cases.

#### Validator & Overall Model

- We propose the validator network that has inputs from the generator network and outputs the validity of the generator outputs.
- The generator and the validator networks are coupled by the weighted sum of the validator embeddings (attention mechanism).
- We firstly train the generator by a supervised method. Then we train the validator and the generator using this validator



Attention mechanism to

validator embeddings v'(\*)

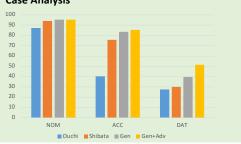
## Experiments & Results

- We use the KWDLC dataset [Hangyo+ 12] for evaluations, following [Hangyo+ 13, Shibata+ 16].
- We evaluate our model in two tasks: case analysis and zero anaphora resolution.
- Gen is the generator network trained with the supervised learning method, while Gen+Adv is the proposal model trained with the validator, compared with [Ouchi+15, Shibata+ 16].



#### **Case Analysis**

 $x_{ul}$ 



· We obverse large increases of scores in ACC and DAT cases. They have fewer training instances and relies on external knowledge resources of predicates and arguments.

FNN of

Validator V(x)

 $\mathbf{h}'_{\mathrm{pred}_i}$ 

 $p_{\arg_2, \operatorname{pred}_j}^{\operatorname{case}_k}$ 

 $p_{\arg_1, \operatorname{pred}_j}^{\operatorname{cacc}_k}$ 

#### **Zero Anaphora Resolution**



### Conclusion

We propose novel adversarial training model for PAS analysis. Our validator enables the generator to learn from an

[1] I. J. Goodfellow, J. Pouget-Abadie, M. Mirza, B. Xu, D. Warde-Farley, S. Ozair, A. C. Courville and Y. Bengio. Generative adversarial nets (2014). [2] T. Salimans, I. Goodfellow, W. Zaremba, V. Cheung, A. Radford, X. Chen and X. Chen. Improved techniques for training gans (2016).
[3] H. Ouchi, H. Shindo, K. Duh and Y. Matsumoto. Joint case argument identification for japanese predicate argument structure analysis (2015). [4] T. Shibata, D. Kawahara, and S. Kurohashi. Neural network-based model for japanese predicate argument structure analysis (2016). [5] M. Hangyo, D. Kawahara and S. Kurohashi. Building a diverse document leads corpus annotated with semantic relations (2012)

unlabeled corpus as an external knowledge resource. We achieve SOTA scores in all cases of KWDLC.

References