

Conversational Korean-Vietnam translator for interpreters

(Neural machine translation)

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Introduction

Conversational

- **Improvement of translation quality**

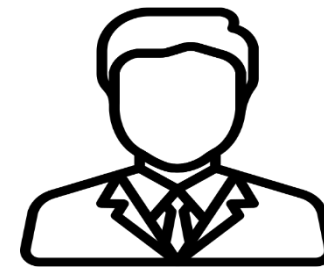
- ✓ Translating natural expression of foreigners into appropriate Korean for situation
- ✓ Improving smooth communication in real-life interpreting situations



It's not rocket science



그건 쉬운 일이야





Related works

02. Related works

WMT22



Seventh Conference on Machine Translation

Confirmed language pairs:

- English – Chinese
- English – Czech
- English – German
- English – Japanese
- English – Russian
- French – German
- Croatian – English
- Livonian – English
- Yakut – Russian

Korean-Vietnamese Neural Machine Translation System With Korean Morphological Analysis and Word Sense Disambiguation

TABLE 11. Statistics of the Korean-Vietnamese parallel corpus.

		#Sent.	#Avg. Len.	#Tokens	#Vocabulary
Vietnamese	Original	454,751	19.3	8,790,197	40,090
	Segmented		16.3	7,409,163	49,208
Korean	Original		12.0	5,435,686	397,130
	Morph. Ana. WSD		21.4	9,728,801	63,735 68,856

02. Related works

Conversational Data

Parallel Corpus

< 영어 >

Because you're **all plans and clockworks**

I just got outsmarted by Mr.**Potato Head**.

Even though it's **once in a blue moon**, there are these moments... Moments?

So we're just gonna **bite the bullet** and just do it.

의역

< 한국어 >

당신은 모든게 **계획대로 순조롭게** 진행
되야하잖아

나 방금 저 **꼴통**한테 당한 건가.

아주 가끔이지만 순간 순간

그래서 **울며 겨자 먹기**로 당장 해치우
려고

English (*source text*)

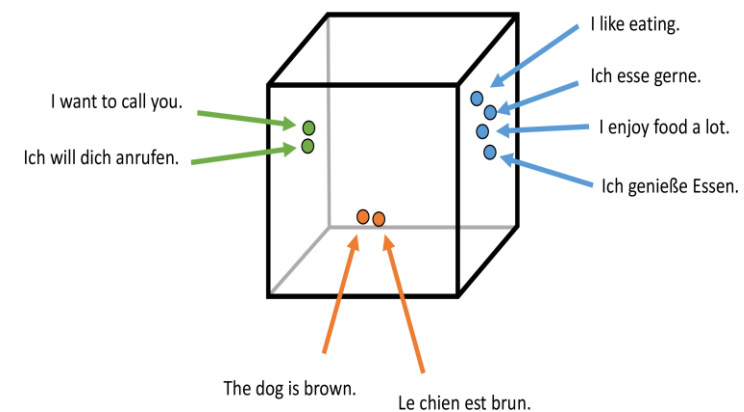
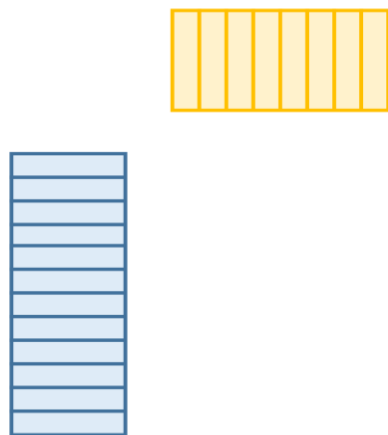
Transformer (*E2E*)
Machine Translation
model

Korean (*target text*)

03. Experience

Data Collection

Use Vecalign to make parallel corpus



03. Experience

Data Collection

- Conversational parallel corpus configuration

Language	Number of Sentences	Number of words
Korean	11,878,865	49,532,476
Vietnam	11,878,865	96,670,925

- Conversational parallel corpus example

Korean	Vietnam
그런데 그걸 집에 와서 봤더니 갑자기 어떤 여자가 나와서.	Vậy mà khi anh ta bật lên để xem lại, thì nhìn thấy một người phụ nữ trong cuốn băng.
그런데 지방이라 동경하고 채널이 다르잖아.	Nhưng không hề có kênh đó phát từ Tokyo.
부인이 병원에 계속 있었어요.	Phải. Vợ ông đã ở trong bệnh viện.
사과하는 걸 수도 있잖아요	Đó có thể là một lời xin lỗi theo như chúng ta biết.

03. Experience

Experience configuration

: Configure hyper parameters for each model size

Base model

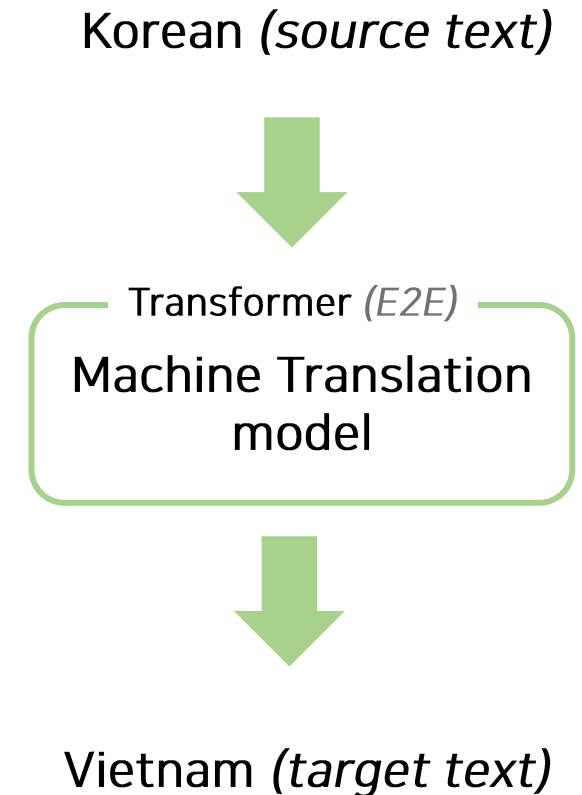
- Text Encoder/Decoder(MT):
Transformer (6-block, 256-hidden, 4-head, noam-learning rate)

Large model

- Text Encoder/Decoder(MT):
Transformer (6-block, 1,024-hidden, 16-head, noam-learning rate)

Evaluation matrix

- BLEU score



03. Experience

Experience configuration

: Configuration for each experiment

Experience	Model size	Data	Training data	Validation data
Exp1	Base	Written	2 million	3,000
Exp2	Base	Conversational	2 million	
Exp3	Base	Conversational	11 million	
Exp4	Large	Conversational	11 million	

: Evaluation data configuration

Language	Number of Sentence	Number of words
Korean	3,000	12,005
Vietnam	3,000	27,002



Result

04. Results

Results 1

: Pre-experimentation

- Quantitative analysis

Model	Written eval set	Conversational eval set	[BLEU Score]
Written model (2M)	26.2	8.8	
Conversational model (2M)	23.4	13.1	

More difficult



: Performance comparison of conversational and written models

- Quantitative analysis

Model	Conversational eval set	[BLEU Score]
Exp1: Written data	7.13	
Exp2: Conversational data	9.35	

04. Results

Results 1

: Performance comparison of conversational and written models

- Quantitative analysis

[BLEU Score]

Model	Conversational eval set
Exp1: Written data	7.13
Exp2: Conversational data	9.35

- Static analysis

Reference	Hypothesis of Exp1	Hypothesis of Exp2
Tay nắm cửa. (The handle is ...)	Bạn có tay cầm không?	Tay cầm...
tìm bà cô. (I'm helping to find aunt)	Tôi đang giúp bạn tìm dì của tôi.	Tôi đang giúp cô tìm dì.

04. Results

Results 2

: Pre-experimentation

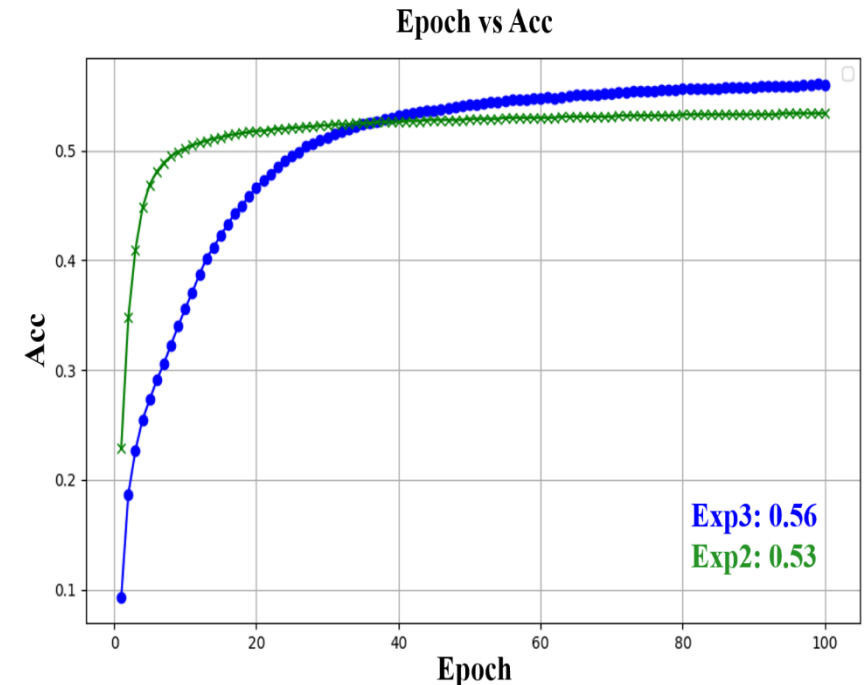
- Quantitative analysis – 2 million VS 11 million

Just increasing the data did not make a difference about quality of translation.

: Performance comparison of increase data

- Quantitative analysis

Model	Conversational eval set	[BLEU Score]
Exp2: 2 million		9.35
Exp3: 12 million		11.83



04. Results

Results 3

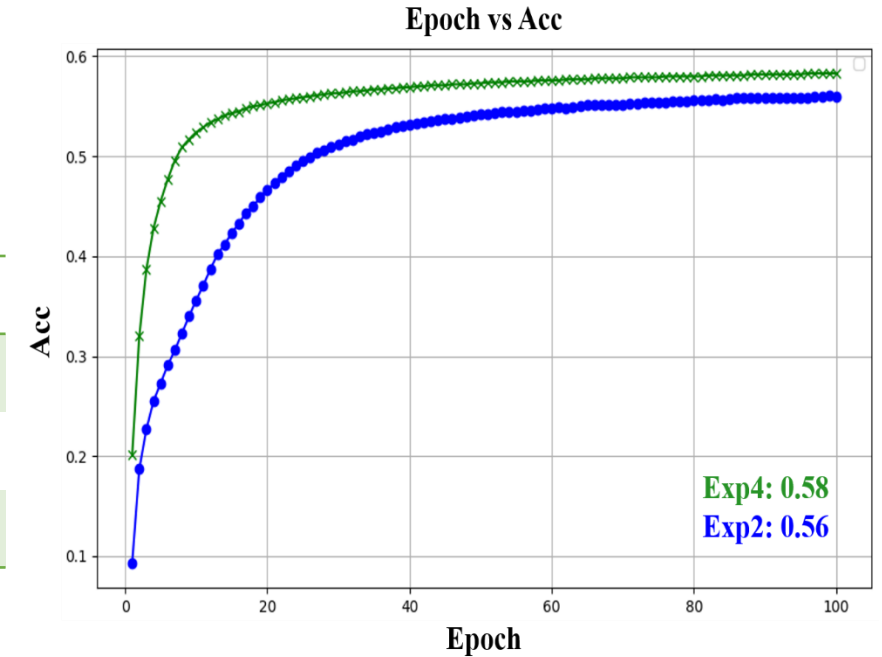
: Performance comparison of increase data & model size

- Quantitative analysis

[BLEU Score]

Model	Conversational eval set
Exp2: 2 million, Base	9.35
Exp3: 12 million, Base	11.83
Exp4: 12 million, Large	13.02

- Static analysis



Reference	Hypothesis of Exp2	Hypothesis of Exp4
chính là nó.	Đây rồi.	Chính là nó.
Tôi đang ở đâu?	Chúng ta đang ở đâu?	Tôi đang ở đâu?
Đây không hẳn là kiếp sau.	Đây không phải là thế giới sau cái chết.	Đây không hẳn là thế giới sau khi chết.

04. Results

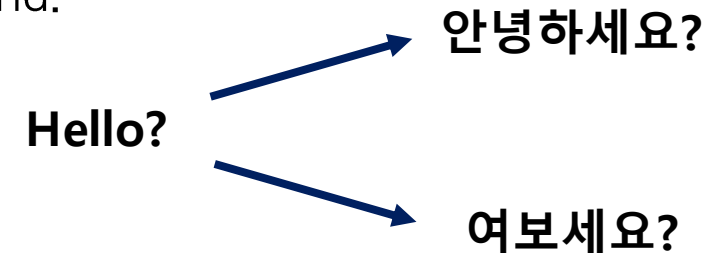
Analysis

: Feature of conversational data

- Conversational data is difficult to translate, because it contains paraphrases.
- Conversational data increases, the ambiguity of the translation increases.
- Compared to other translators, the result is a bit easier to understand.

: Improving massively parallel corpus about MT

- Modified hyperparameters to be suitable for huge data.
- Reduce the likelihood of output
 - Learning contextual information from translation models.





Side Experience



Data configuration

: Korean – English machine translation

- Source language: English
- Target language: Korean

: Parallel corpus

- Training data: Conversational Data, 40 million
- Evaluating data: 4,000

Model	Data info	Domain
Korean -> English	38 million sentence	Conversational (Drama, movie scripts)
	10 million sentence	Written (Trip, daily)
	6.9 million sentence	Written (Trip, daily)

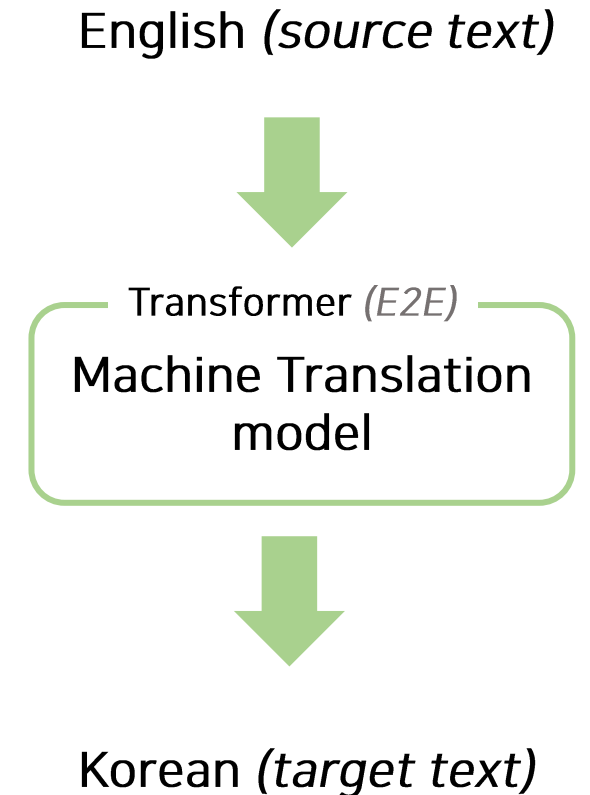
Experience configuration

: Korean – English machine translation

- Text Encoder (MT):
Transformer (6-block, 512-hidden, 8-head, noam-optimizer)
- Text Decoder (MT):
Transformer (6-block, 512-hidden, 8-head, noam-optimizer)

: Eval matrix

- DeepL, Papago
- BLEU score



Result

: My model vs DeepL vs Papago

- Quantitative analysis

[BLEU Score]		
Model	Conversational eval set	Written eval set
My model	34.0	32.3
DeepL	36.6	28.6
Papago	32.6	33.9

Conversational eval set: Papago < My model < DeepL

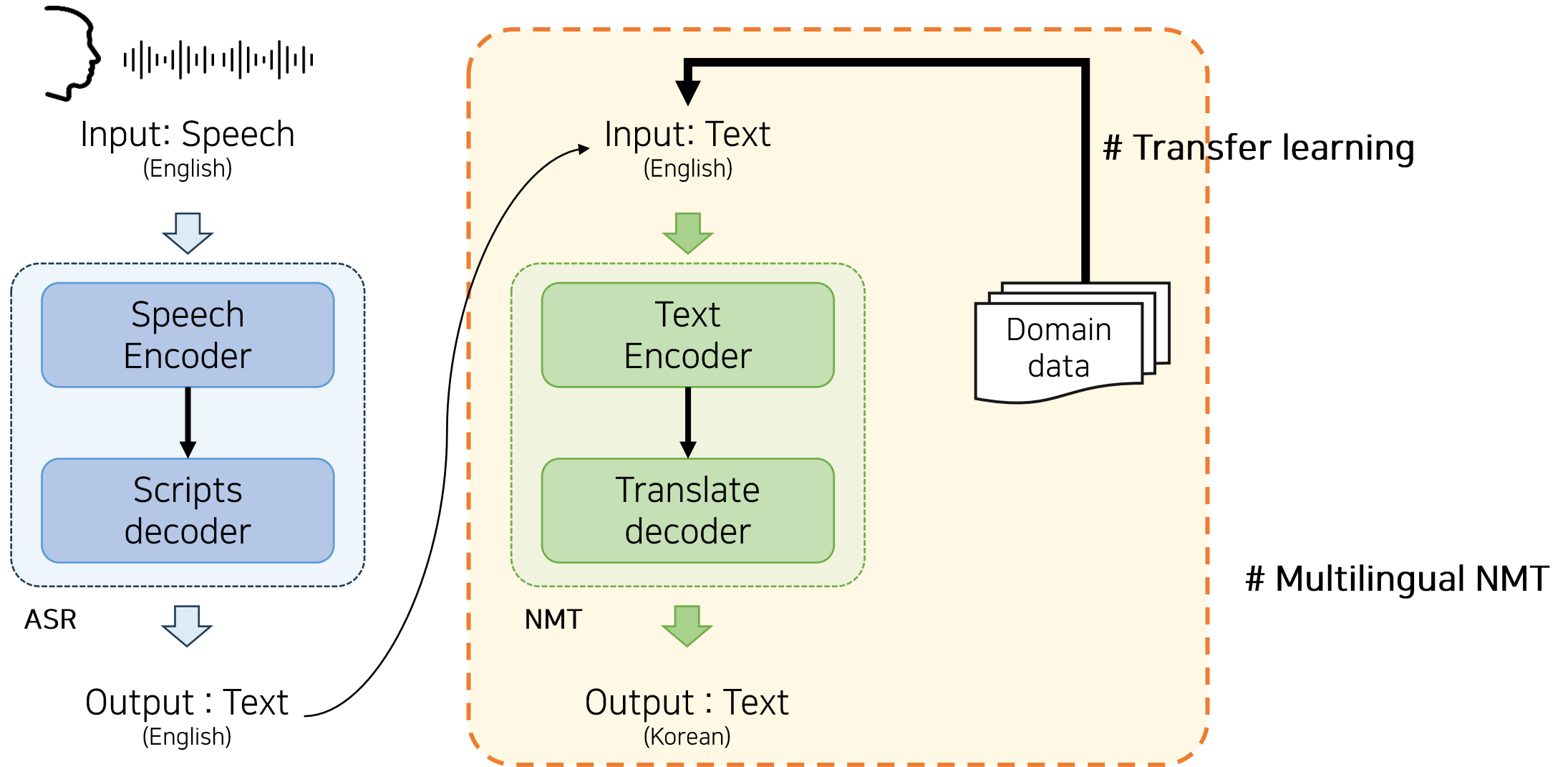
Written eval set: DeepL < My model < Papago

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Future work

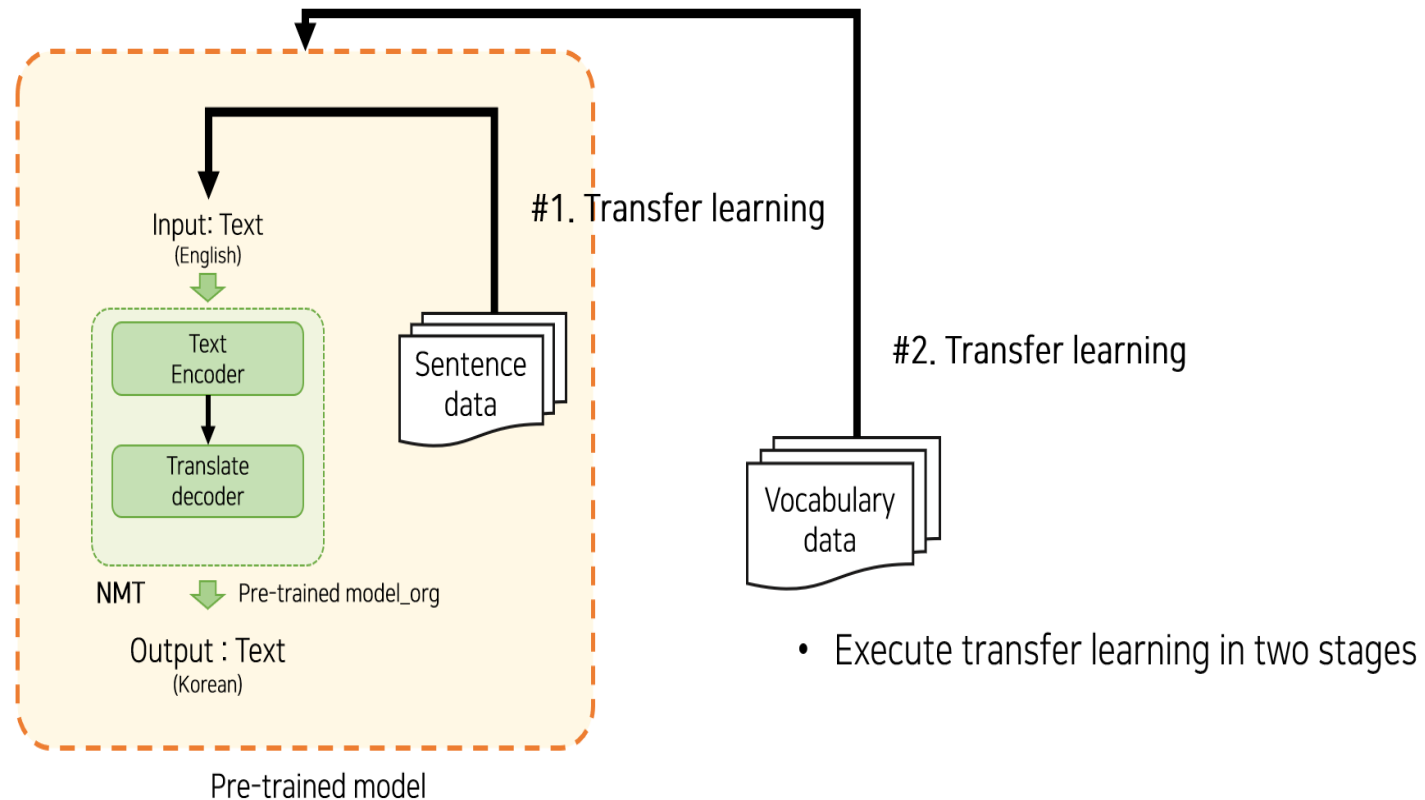
03. Motivation

Conversational NMT model applicable in real-life scenarios- structure



03. Motivation

Domain adaptation



Transfer learning

- Domain: conference (CV)
- Consider: copy-mechanism
- Use data: sentences & words

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Future work

Q & A

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