

Generating Stylistically Consistent Dialog Responses with Transfer Learning

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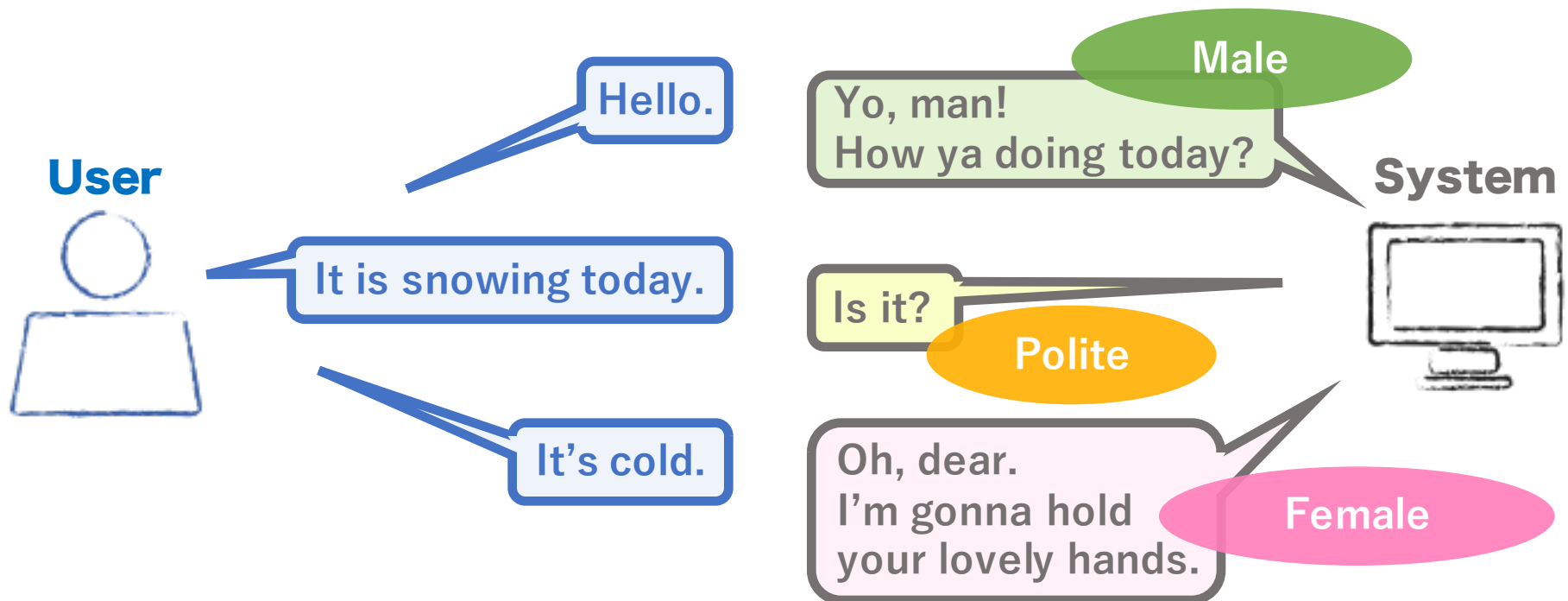
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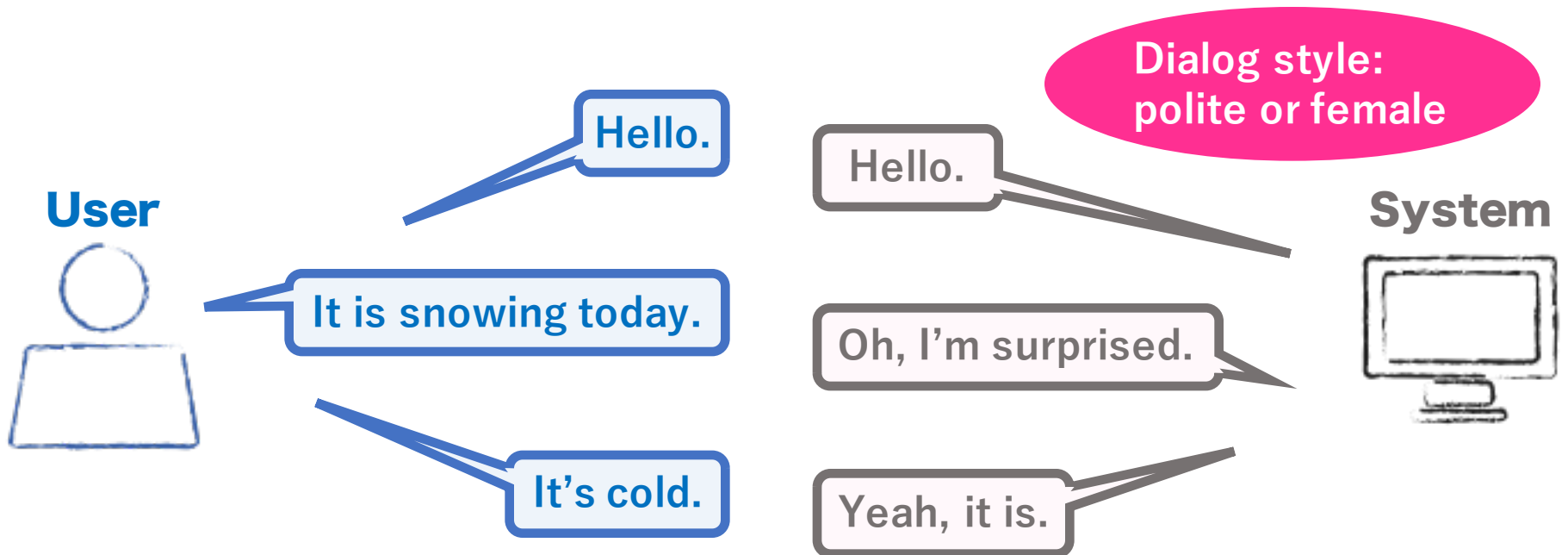
Prior approaches

- Feed large-scale conversational data into sequence-to-sequence (seq2seq) model [Vinyals+'15, Li+'16, etc.]
- **Problem:** generates **NOT stylistically consistent responses**



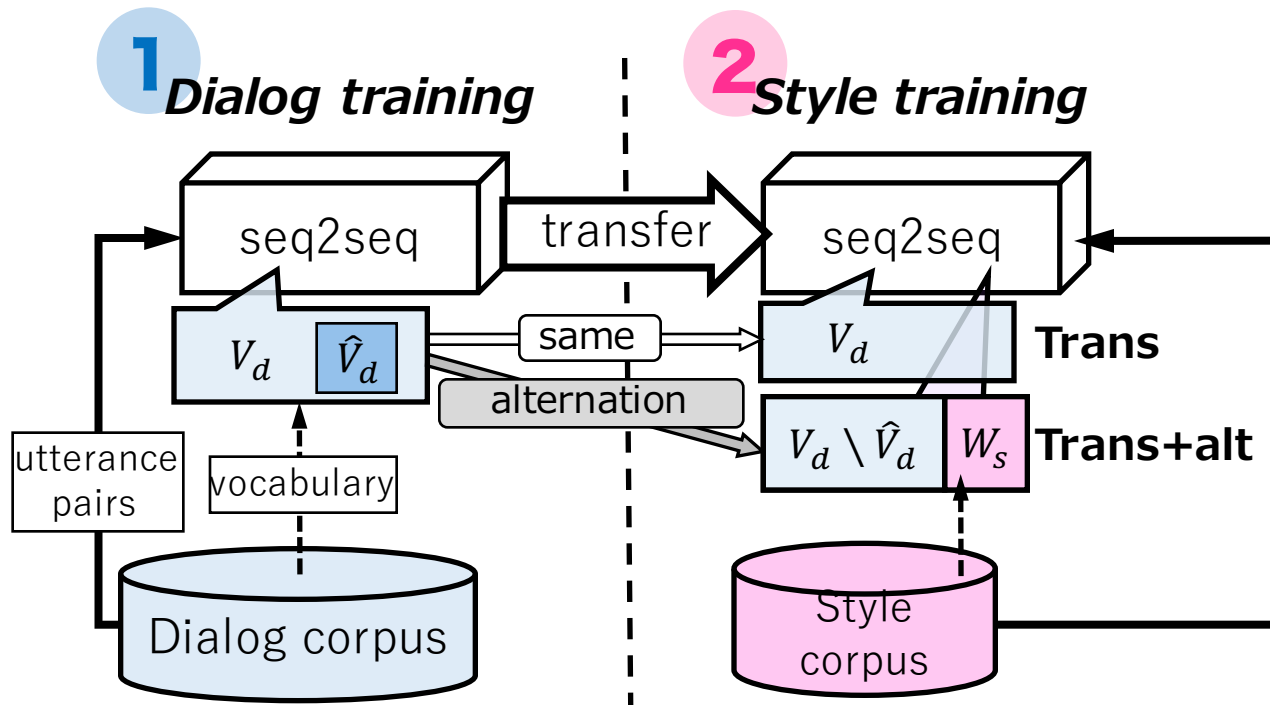
Our goal

- Generate dialog responses with **particular style consistently** while maintaining goodness of contents and fluency



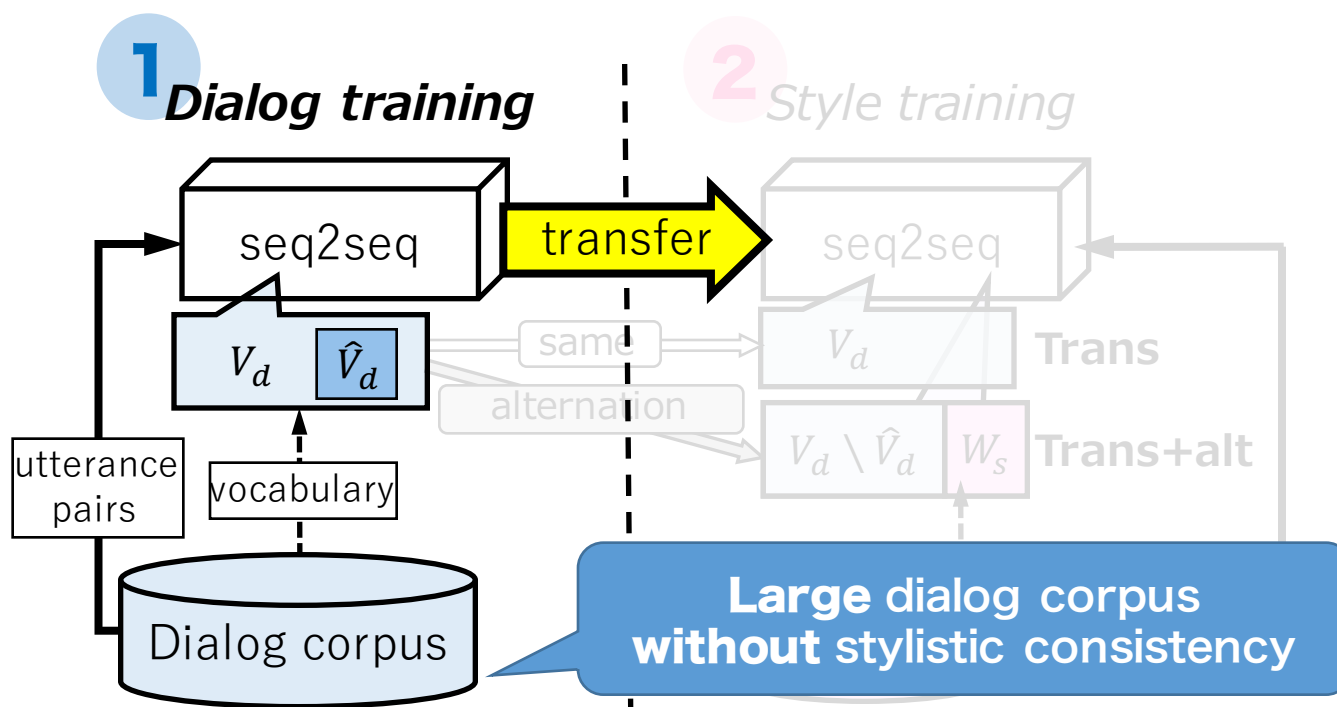
Proposed method

- Apply **transfer learning** to seq2seq response generation model
- Propose **two-staged** training framework



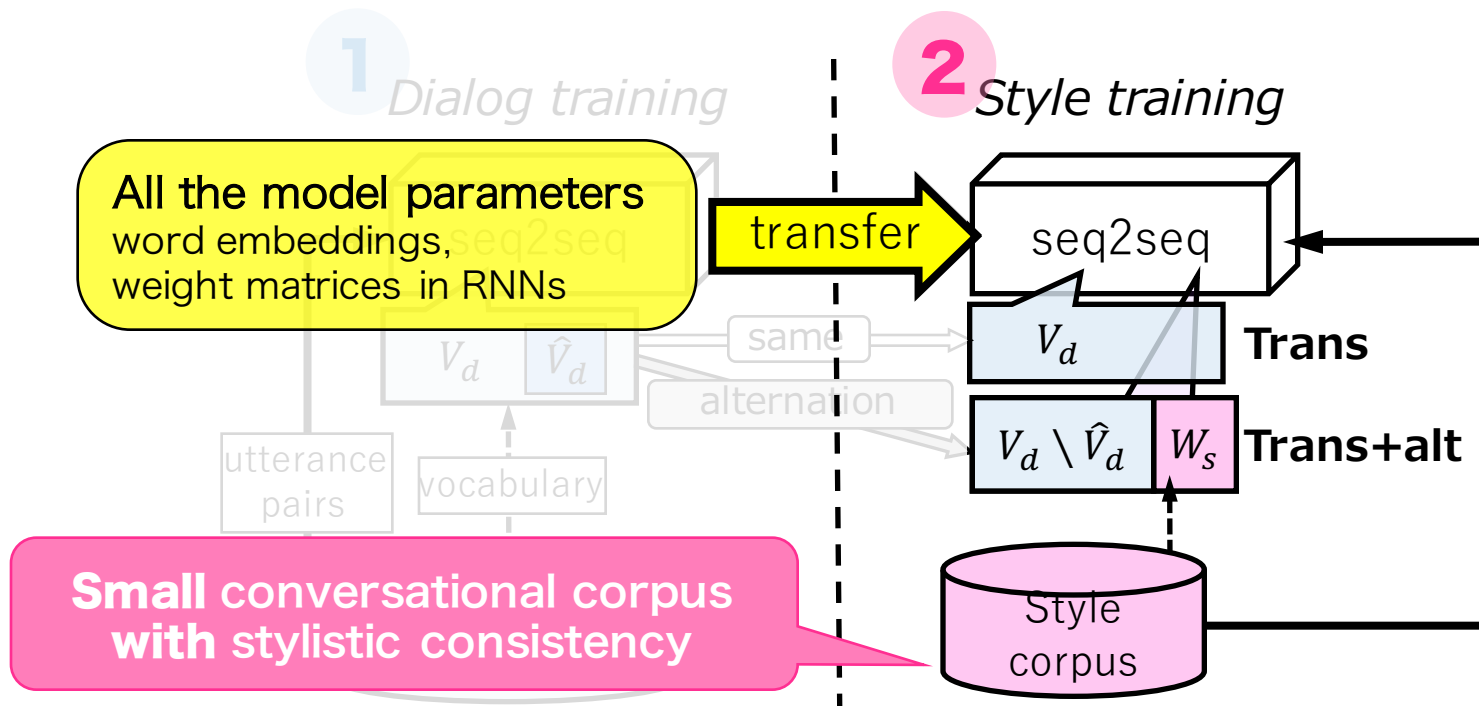
1st stage: Dialog training

- Focus on learning **what-to-say** and **fluent** responses
- Exploit cheap, large-scale but *stylistically inconsistent* dialog corpus



2nd stage: Style training

- Focus on learning **stylistically consistent** responses
- Fine-tune seq2seq model trained in dialog training
- Exploit expensive, small-scale but *stylistically consistent* corpus

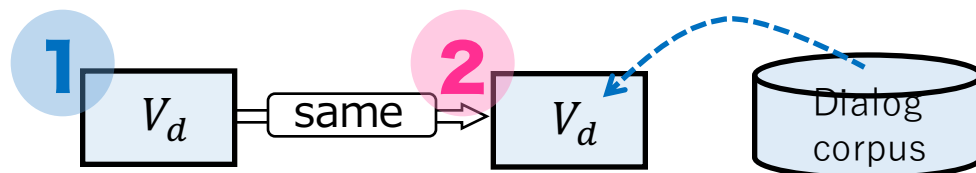


Seq2seq vocabulary in style training

Issue: how to create the seq2seq vocab. for style training?

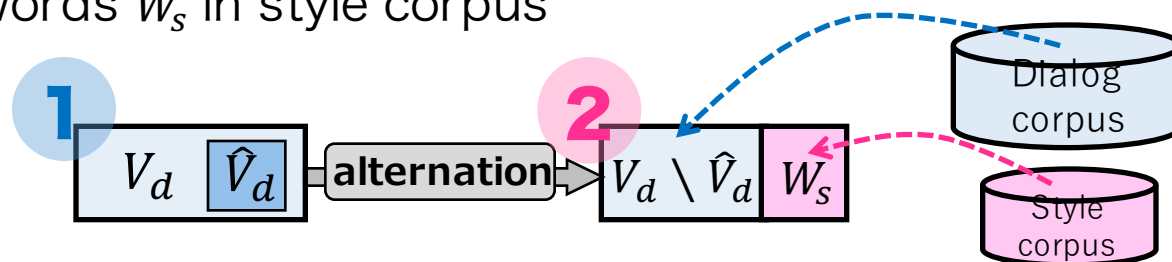
1. Trans

- Simply use the same vocabulary V_d as Dialog training
- V_d : the top N_d most frequent words in dialog corpus



2. Trans+alt


- Alter the vocabulary before style training
- Replace infrequent words \hat{V}_d from dialog corpus with frequent words W_s in style corpus



Dataset

Dataset

- Japanese single-turn dialog (i.e., utterance pairs)
- **Dialog corpus**
 - **3.7 M** utterance pairs extracted from tweet-reply chains
- **Style corpus**
 - **Feminine and polite style**
 - **12 K** utterance pairs extracted from TV subtitles
 - Only 0.3% of dialog corpus
 - TV program where *Tetsuko Kuroyanagi* chats with some guests

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- * Japanese TV personality
 - * Elderly woman
 - * Speaks **feminine and polite words**

Experiments setup

Models

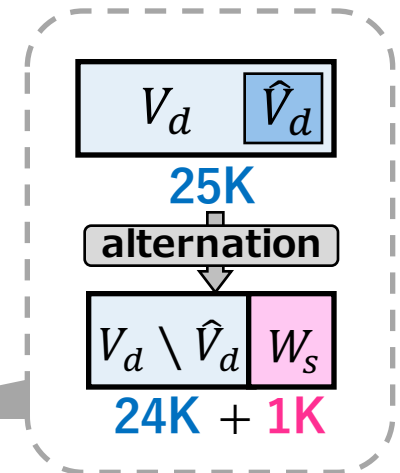
		Baselines		Proposals	
		Base	Mix	Trans	Trans+alt
features	models				
Dialog corpus		✓	✓	✓	✓
Style corpus			✓	✓	✓
Vocabulary alternation			✓		✓
Transfer learning				✓	✓

Settings

- Seq2seq: 2 layer LSTMs with 2048 units
- Optimization: Adam (mini-batch size 64)
- Word embeddings: 1024 dim
- Vocab. size: 25K

Vocabulary alternation (Trans+alt)

- Vocab. size: 1K words from style corpus (W_s)



Evaluation setup

Human evaluation

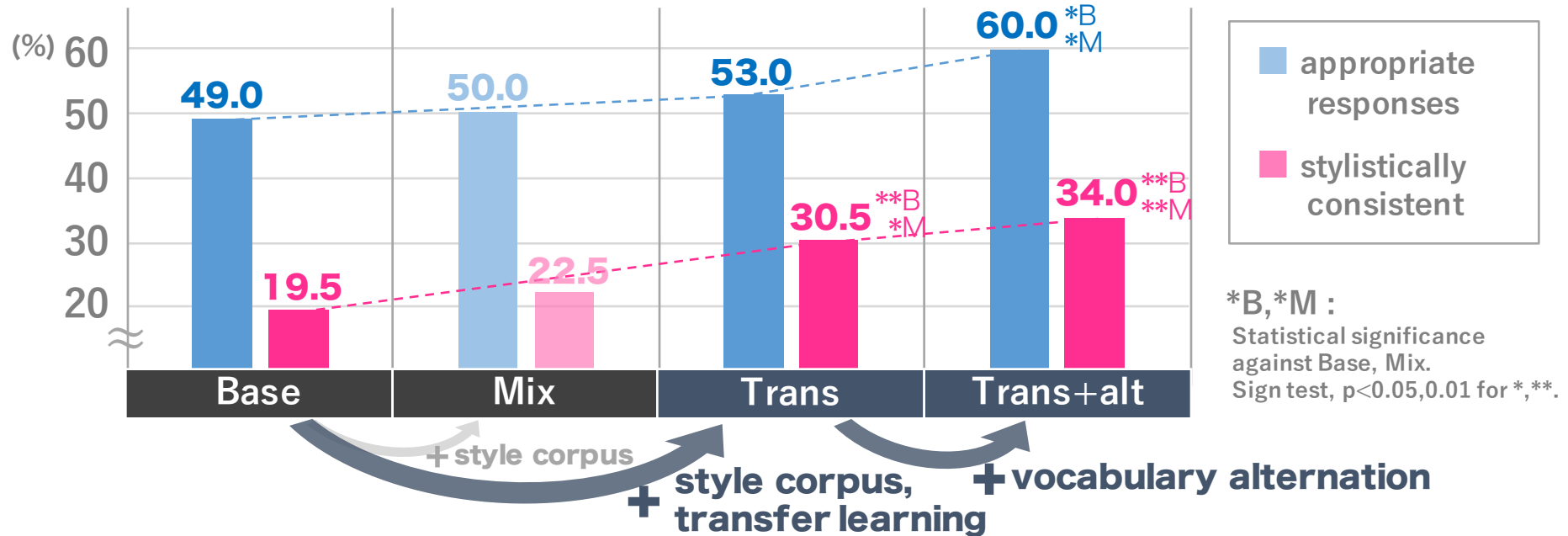
- 5 crowd workers judged each responses generated by each model
- Given input, generated response, style description, and **2 questions**:
 - Q1.** Whether the response is grammatically and semantically **appropriate**
 - Q2.** Whether **the style** of response matches the given description



Evaluation metrics

- The percentage of responses judged as
 - 1. appropriate responses**
 - 2. stylistically consistent**

Results



- **Transfer learning** frameworks are successful
 - Increase stylistic consistency of generated responses
 - Maintain the appropriateness of responses
- **Vocabulary alternation** helps to make the more stylistically consistent responses

Conclusion

Summary

- Presented novel end-to-end framework to build **stylistically consistent** dialog response generation system
- Apply transfer learning to seq2seq response generation model
- Human evaluation demonstrated that proposed method produces stylistically consistent responses while maintaining appropriateness of responses

Future work

- Improve style training so that it can learn only the style of responses
- Explore effective way of creating style corpus
 - e.g., automatically collecting polite utterance from a large Twitter corpus

Appendixes

References

- Diederik Kingma and Jimmy Ba. 2015. “Adam: A method for stochastic optimization.” In *The International Conference on Learning Representations (ICLR)*.
- Ilya Sutskever, Oriol Vinyals, and Quoc V. Le. 2014. “Sequence to sequence learning with neural networks.” In *Advances in neural information processing systems*, pages 3104–3112.
- Oriol Vinyals and Quoc Le. 2015. “A neural conversational model.” In *International Conference on Machine Learning (ICML) Deep Learning Workshop 2015*.

Generated responses

- Generated responses by our best model Trans+alt
- Feminine and polite style

input	generated responses
お住まいはどちらでしょうか？ Where do you live?	京都です I live in Kyoto.
ドル安ですね The dollar is weak, isn't it?	そうですね Yes, it is.
月が綺麗 The moon is beautiful.	すごい Marvelous!

Human evaluation

- Each model generated 200 responses for evaluation
 - input: 50 utterances
 - generated 4 responses for each utterance
- Each response is judged by 5 crowd workers
 - Via Yahoo! Crowd Sourcing (<https://crowdsourcing.yahoo.co.jp/>)
 - Given input, generated response, **style description** and 2 yes-no questions
 - Final answer is determined by majority vote

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- * only a description with several example
 - * without the specific name of individual character