

Modifying the Writing Style in Goal-Oriented Dialog Generation

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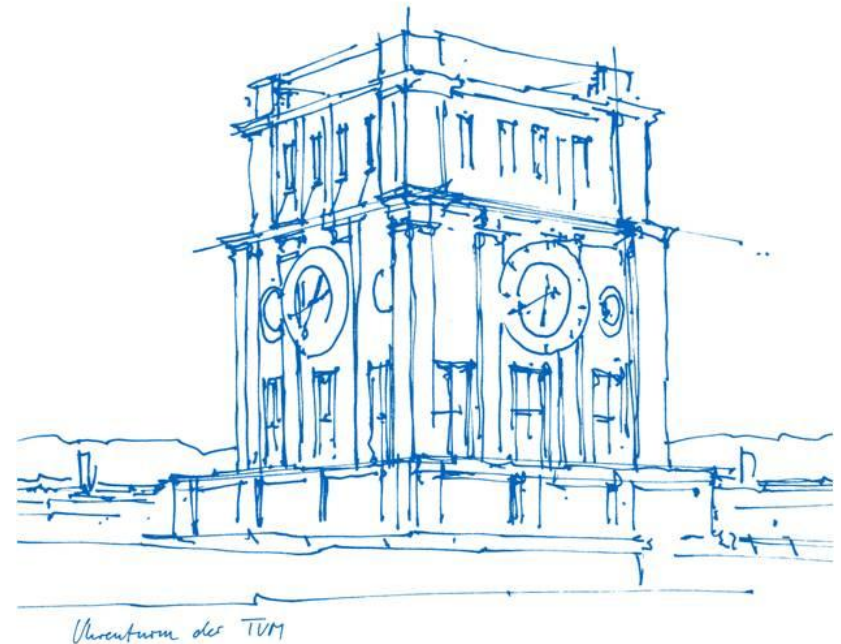
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- Motivations and Goals
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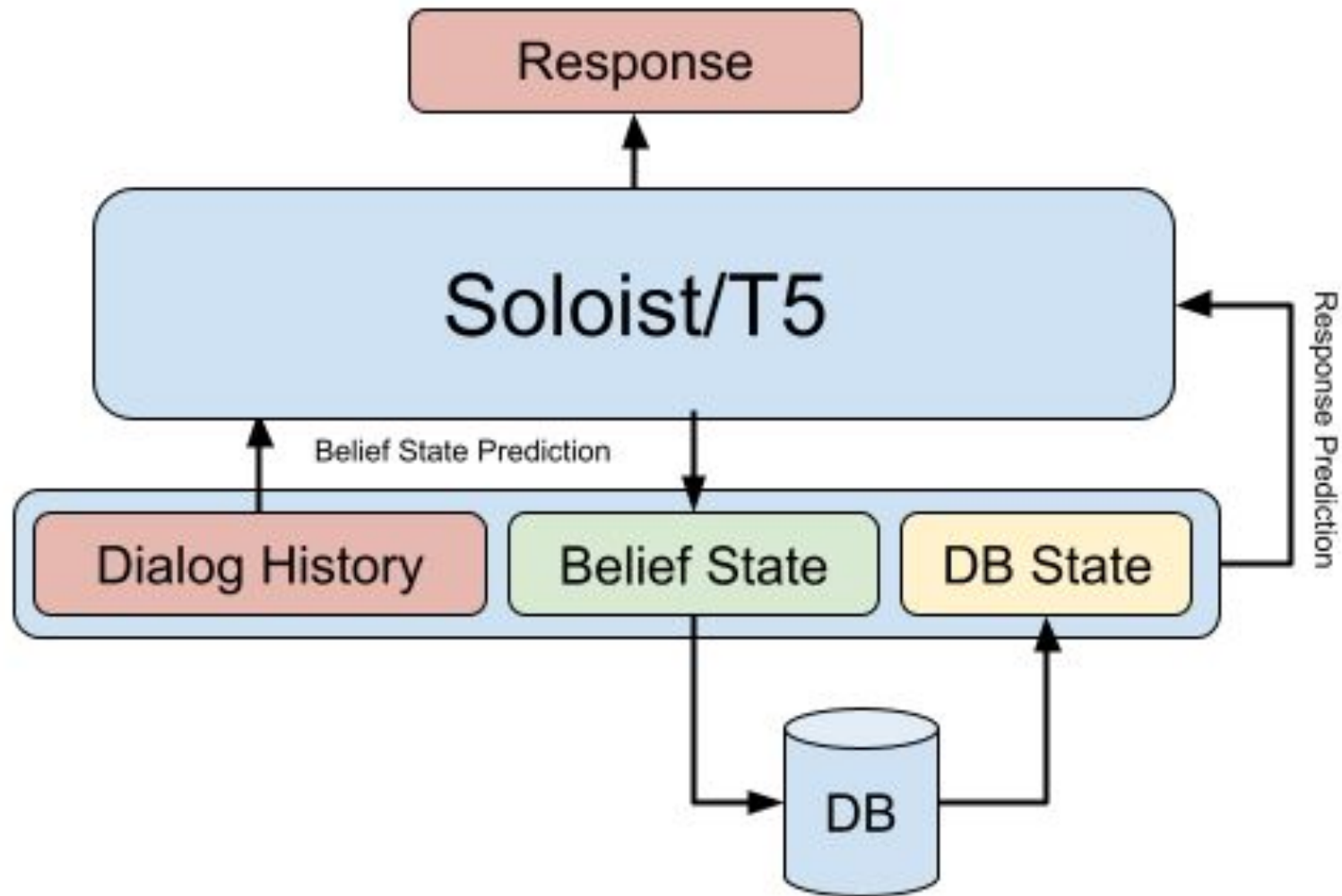
Motivation

- Fixed writing style
 - cannot react on changes in writing style of the used
 - writing style cannot be changed after training
- when aiming at a certain style → find task data with task data

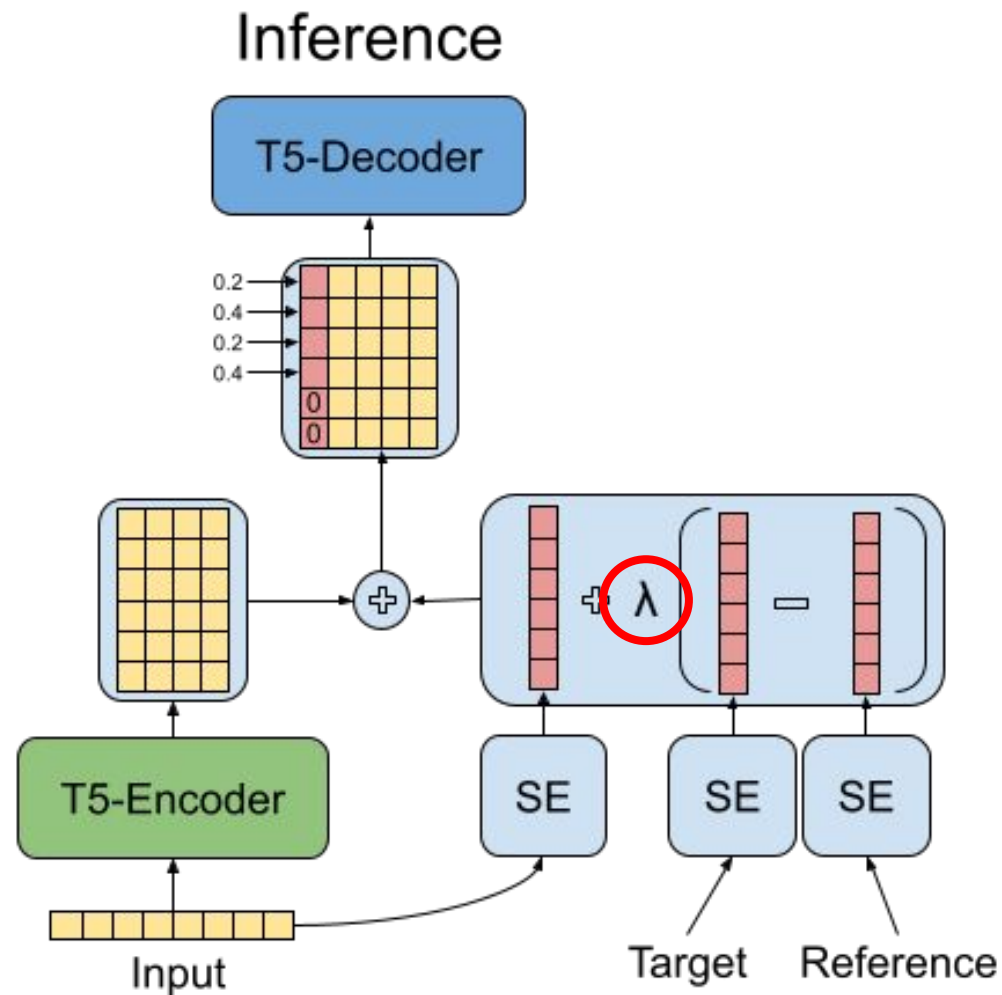
Goals

1. develop a dialogue agent
2. develop a model that modifies the writing style with a given style
3. or combine the agent and modification model

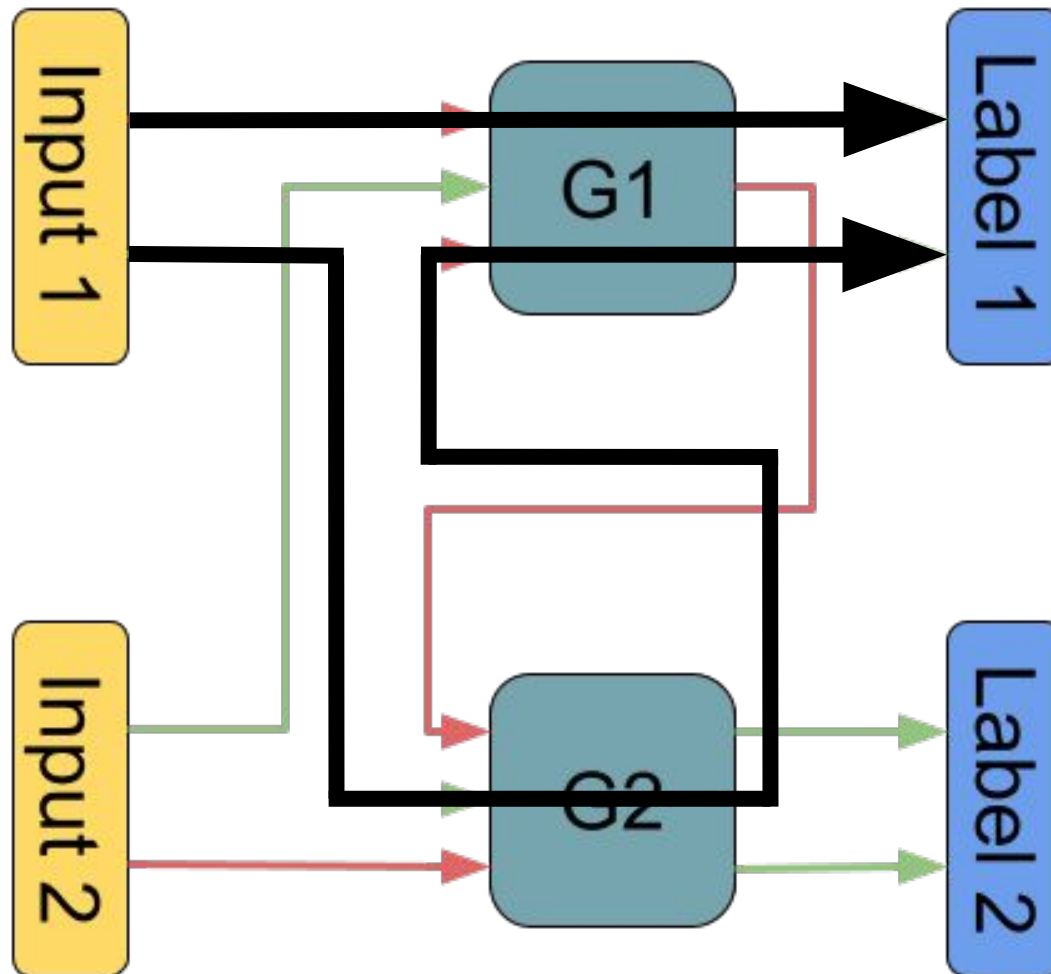
Approaches - Soloist



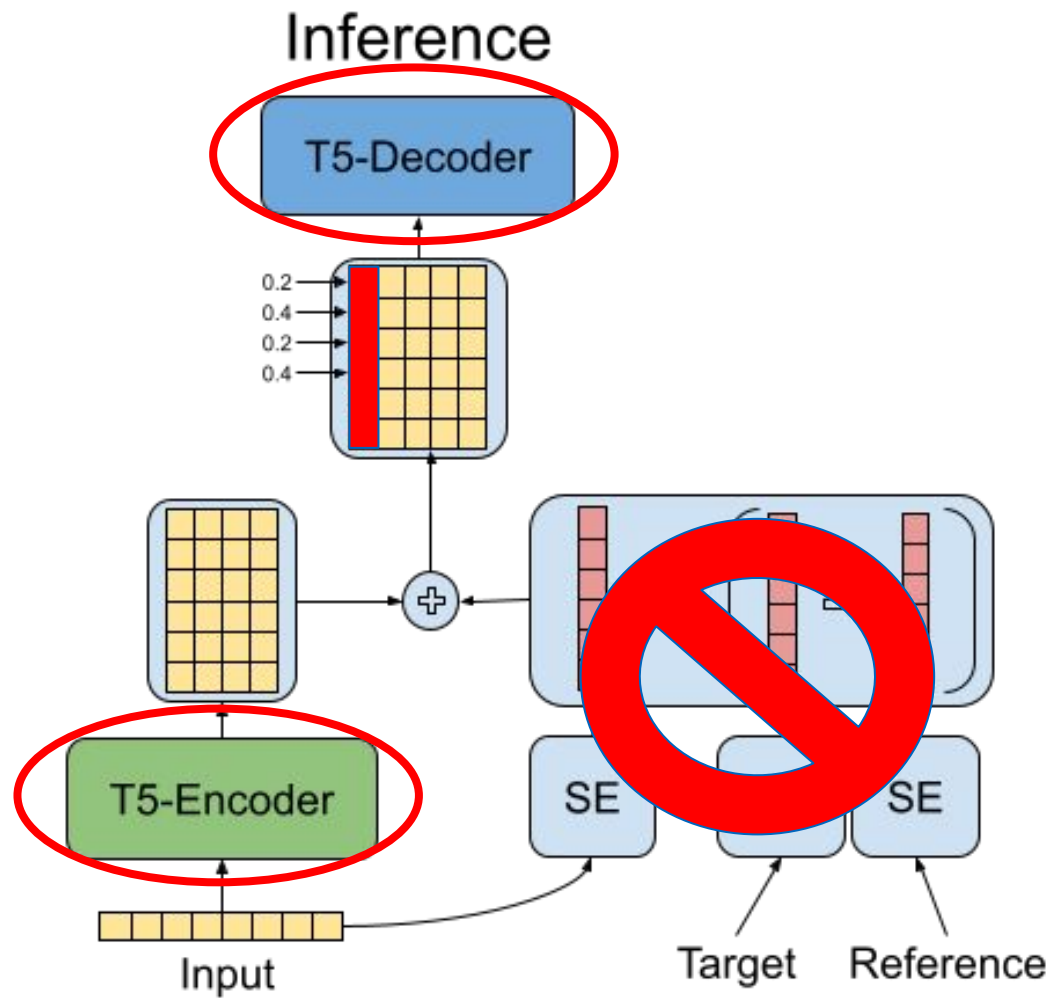
Approaches - TextSETTR



Approaches - DGST



Merged TextSETTR



Parameters

| Original Model | Parameters | Modified Model | Parameters |
|---------------------|------------|------------------|------------|
| Soloist (GPT-2) | 117M | Small Soloist | 60M |
| TextSETTR | 1B | Small TextSETTR | 96M |
| TextSETTR + Soloist | 1.1B | Merged TExtSETTR | 96M |

Datasets

Taskmaster Dataset (~46k entries):

user: Hello, I am looking for movie theaters near Hudson, Wyoming.

=> Belief State : Movie Tickets
{ location = near Hudson, Wyoming } query { find_theaters } <EOB>

DB: find_theaters {
name.theater_1 = Gem Theater; ... } <EOKB>

Hello, I have two nearby examples. One is Gem Theater, which is 15.2 miles away ...

Trump Twitter Dataset (~34k entries):

To all of those who have asked, I will not be going to the Inauguration on January 20th.

I WON THIS ELECTION, BY A LOT!

Fine Food Dataset (~500k entries):

I mix this with nonfat plain yogurt and it's the best dip around -- I can eat veggies all day with this stuff. Hard to find in grocery stores so I'm THRILLED it's on Amazon!

Training

- DGST, Soloist, Small TextSETTR and Merged TextSETTR:
 - Modified seq2seq task
 - on one Tesla T4 GPU
- Full sized TextSETTR:
 - Modified seq2seq task
 - model parallelism on 4 Tesla K80

Evaluation - Soloist

- BLEU Score
- Inform \rightarrow {location : Munich, ... }
- Success \rightarrow The theater is located in *Munich*
- Combined \rightarrow (Inform + Success) * 0.5 + BLEU Score

Results - Soloist

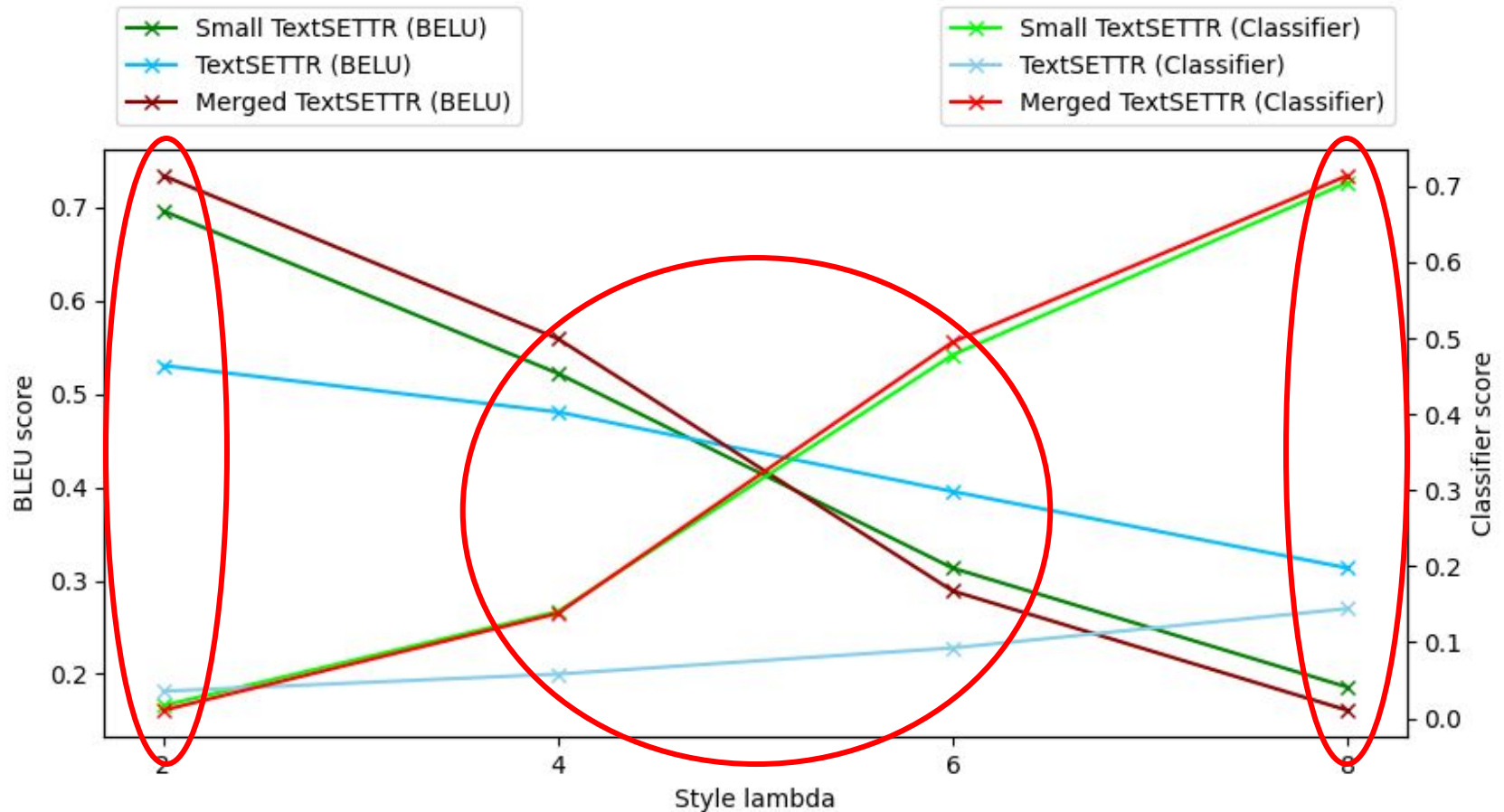
| Model | Dataset | Inform | Success | BLEU | Combined |
|-----------------|----------------------------|--------|---------|-------|----------|
| Soloist (large) | MultiWOZ (6 Domains) | 85.50 | 72.90 | 16.54 | 102.49 |
| Soloist (small) | Taskmaster (Movie Tickets) | 84.24 | 67.30 | 60.79 | 136.56 |

Evaluation - DGST and TextSETTR models

- BLEU Score
- Classifier Score:
 - Distilbert + classifier head
 - Taskmaster response vs. Donald Trump response
 - 99.6% accuracy after one epoch

Only heuristics !!!

Results - TextSETTR



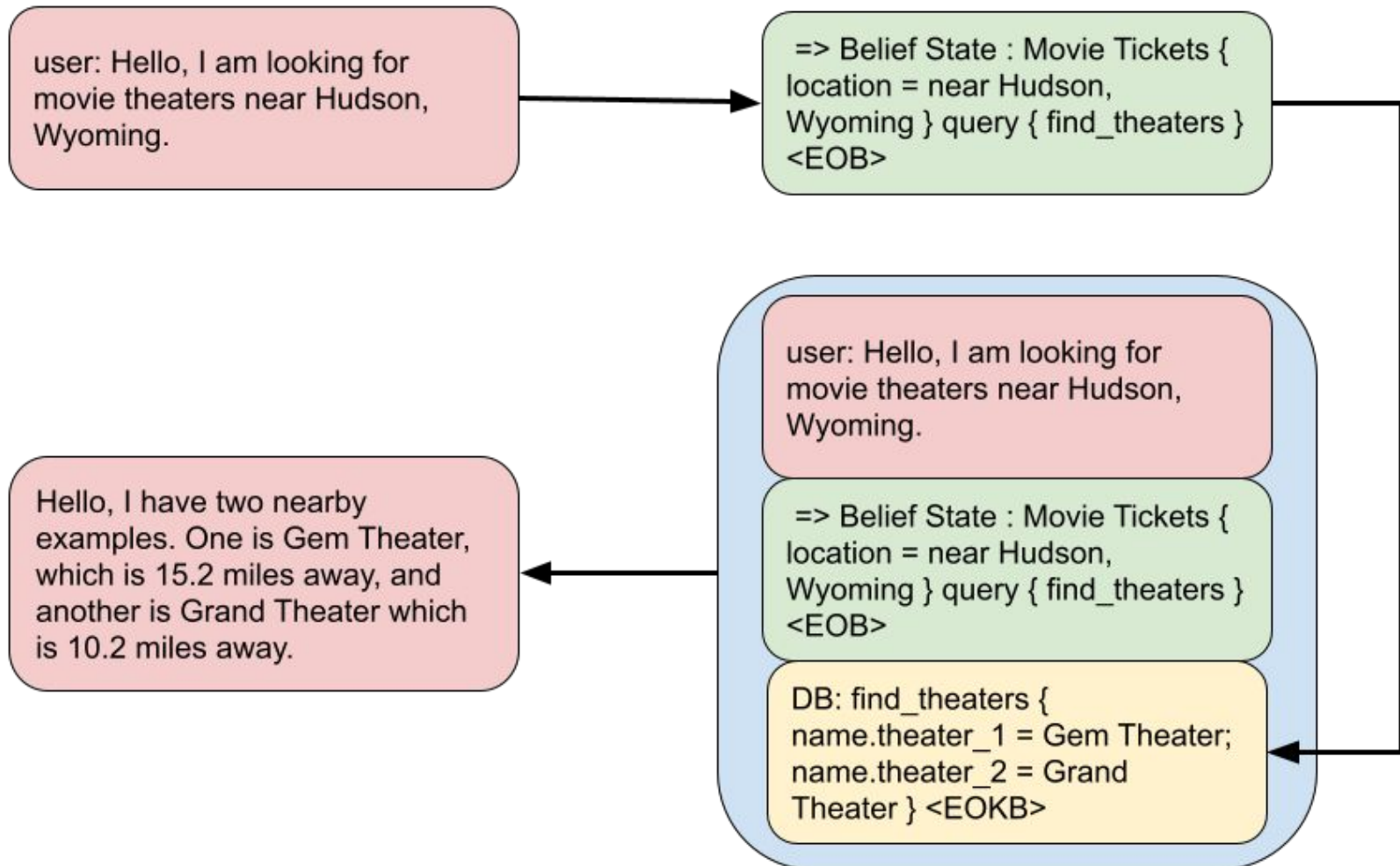
Results - DGST

- BLEU score: 82.3
- Classifier score: 0.037
- non modified dataset score: 0.033

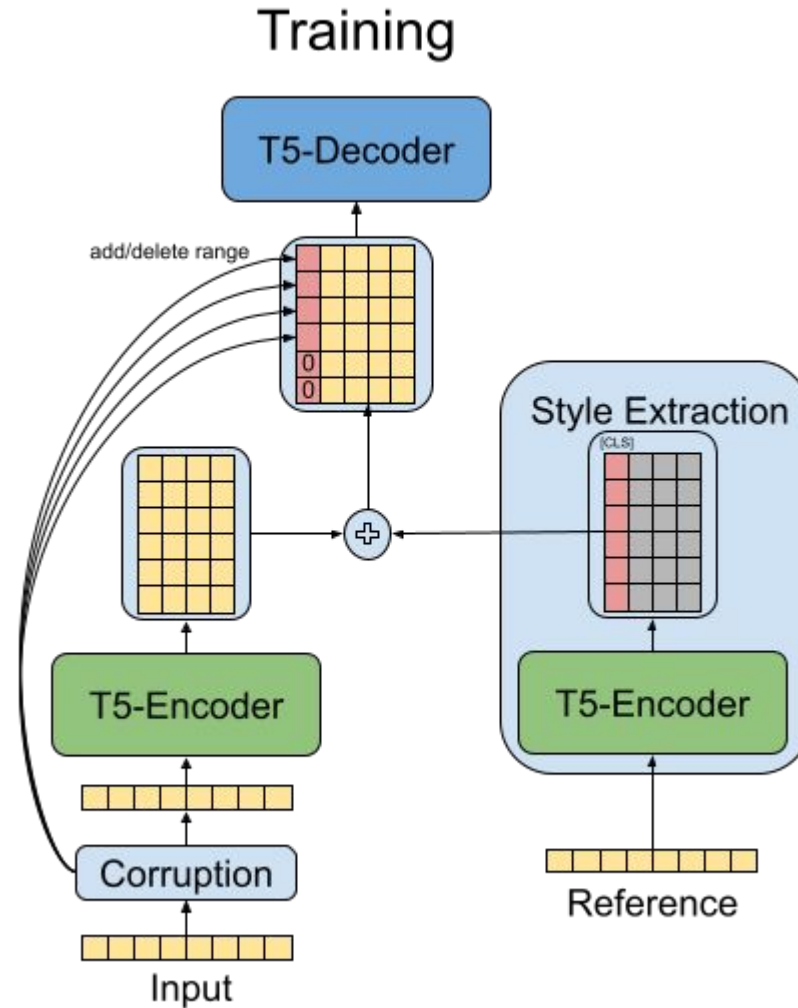
Conclusion

- Good performance of the small Soloist on a reduced domain
- TextSETTR and DGST does not perform sufficiently
- TextSETTR is style adaptive

Approaches - Soloist Data



Approaches - TextSETTR Training



Prediction Examples

TextSETTR lambda 4:

- "Family Jewels" is currently available
- i am a "Family Jewels" currently

Large TextSETTR lambda 4:

- Okay let me check. Yes there is one available in the Filmore district. Is that okay with you?
- We have a dog named Mr. Weimaraner.. Yes there is one available in the Filmore district.. Is that okay with you?