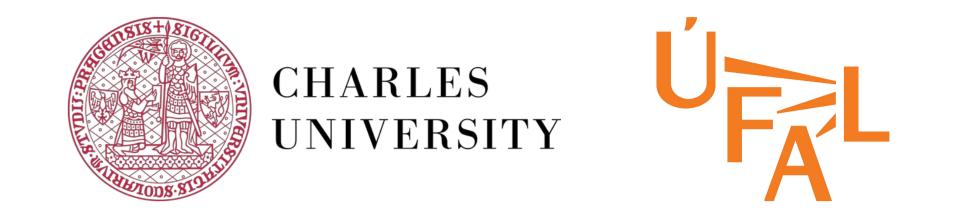
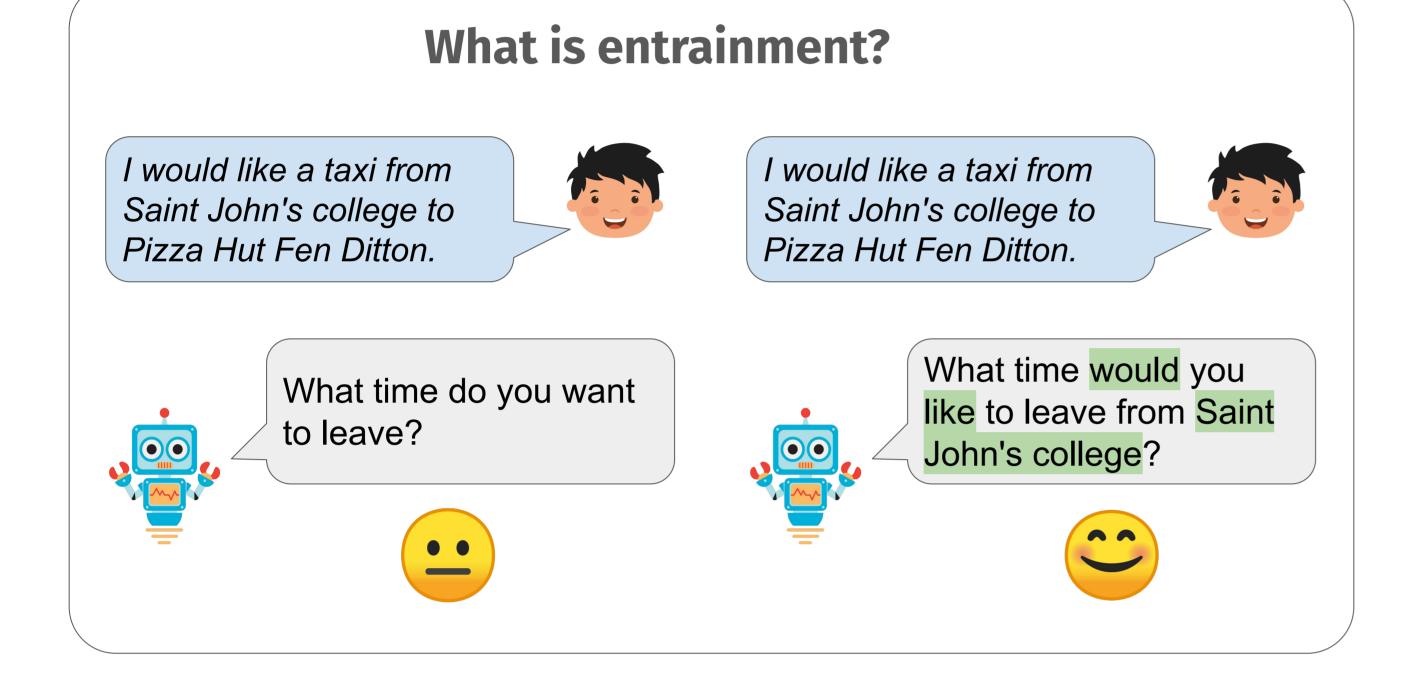
# Entrainment increases naturalness in dialogue systems, without hurting task success

# LEEETs-Dial: Linguistic Entrainment in End-to-End Task-oriented Dialogue systems

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# **Objective & Approaches**

- Lexical entrainment for E2E Task-oriented DS
- Entrainment = alignment/adaptation of conversational partners
   occurs at various linguistic levels
- Entrainment → More natural responses → Better user experience

#### How to facilitate entrainment? Ideas:

+ IW: Promote "nice" data instances during training
+ ULL: Maximize the probability of "desired" tokens
+ LK: Assist the model by adding hints to the input

## **Experiments & Results**

#### Model: AuGPT (Kulhánek et al. 2021)

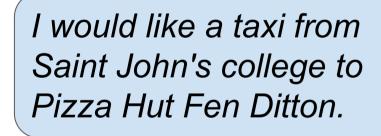
- end-to-end, GPT-2 based
- context  $\rightarrow$  belief state  $\rightarrow$  DB  $\rightarrow$  response

#### Baselines

- **Base-CE**: AuGPT finetuned on MultiWOZ 2.1
- **D&J16**: Base-CE + n-gram reranking while decoding

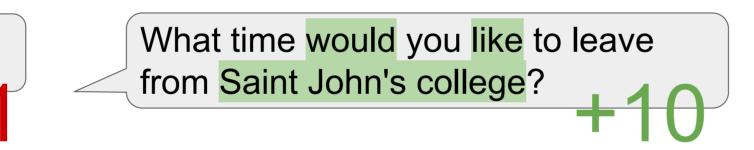
### Evaluation on MultiWOZ 2.1

- MultiWOZ metrics: inform/succes rate & BLEU
- Entrainment metrics:
  - 1-gram precision & recall (lexical)
- 2/3-gram precision on POS tags (syntactic)
  Human: Relative ranking of naturalness



## 1. Instance Weighting (IW)

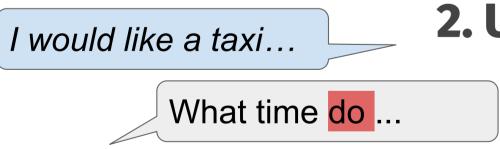
What time do you want to leave?



- Higher weights for better user-system overlap (1-gram precision)
  - Discrete (*IW1*): **1** if 1-gram precision < 0.25, else **10**

+'

• Continuous (*IW2*): modified sigmoid function with values 1-10



#### 2. User Likelihood Loss (ULL)

Reuse {I, would, like, a, taxi, from, Saint, John's, college, to, Pizza, Hut, Fen, Ditton.}

- Maximize probability of user tokens  $u \in U$
- Additional loss with weight  $\alpha: -\alpha \cdot \log(\sum_{u \in U} p(u|x_t))$

## 3. Conditioning on Lexical Keywords (LK)

I would like a taxi...

Use these keywords: {would, like,

What time would you like the taxi to leave from Saint John's college?

#### Results

system	inform	success	bleu	lex-p1	lex-r1	syn-p2	syn-p3	Ørank
Base-CE	83.5	63.8	15.7	20.7	24.5	14.8	5.0	4.18
D&J16	85.7	63.6	10.6	31.9	26.1	23.1	10.4	5.35
IW1	84.5	68.6	14.9	22.9	30.9	16.4	5.9	3.16
IW2	82.6	67.7	15.3	22.9	29.8	16.4	5.8	3.77
ULL (0.20)	81.6	65.3	15.3	23.7	29.4	16.2	5.7	4.17
ULL (0.25)	81.6	63.6	14.6	24.7	31.6	16.9	6.1	4.33
LK (0.05)	83.3	66.3	12.8	25.8	33.6	17.0	6.5	3.25

- **D&J16** shorter, less polite, less interactive outputs
- IW1 consistently natural-looking outputs
- **LK** better entrained outputs
- **ULL** fragile, hallucinates with higher weight
- Context: [...]

taxi, from, Saint, John's, college}

- Assisting model with "important" user tokens
- Training keywords = overlapping 1-grams in user-system
   Inference keywords = user tokens based on self-attn scores
   Blending: expose model to inference keywords during training with probability σ = {0, 0.05, 0.5}

**User**: I would like one in the centre please. **Ground Truth**: Christ's College, Corpus Christi, King's College, and Pembroke College are some of the colleges in that area

Base-CE: Christ's College is located in the centre and has free admission
D&J16: Christ's College is in the centre of town.
IW1: There are 13 colleges in the centre. I recommend Christ's College. Would you like more information on it?
ULL (0.25): I would recommend Christ's College. Would you like the address?
LK-CE (0.05): Christ's College is one of the best college in the centre. Would you like more information?



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https://github.com/knalin55/LEEETs-Dial

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