

# A Unifying View On Task-oriented Dialogue Annotation

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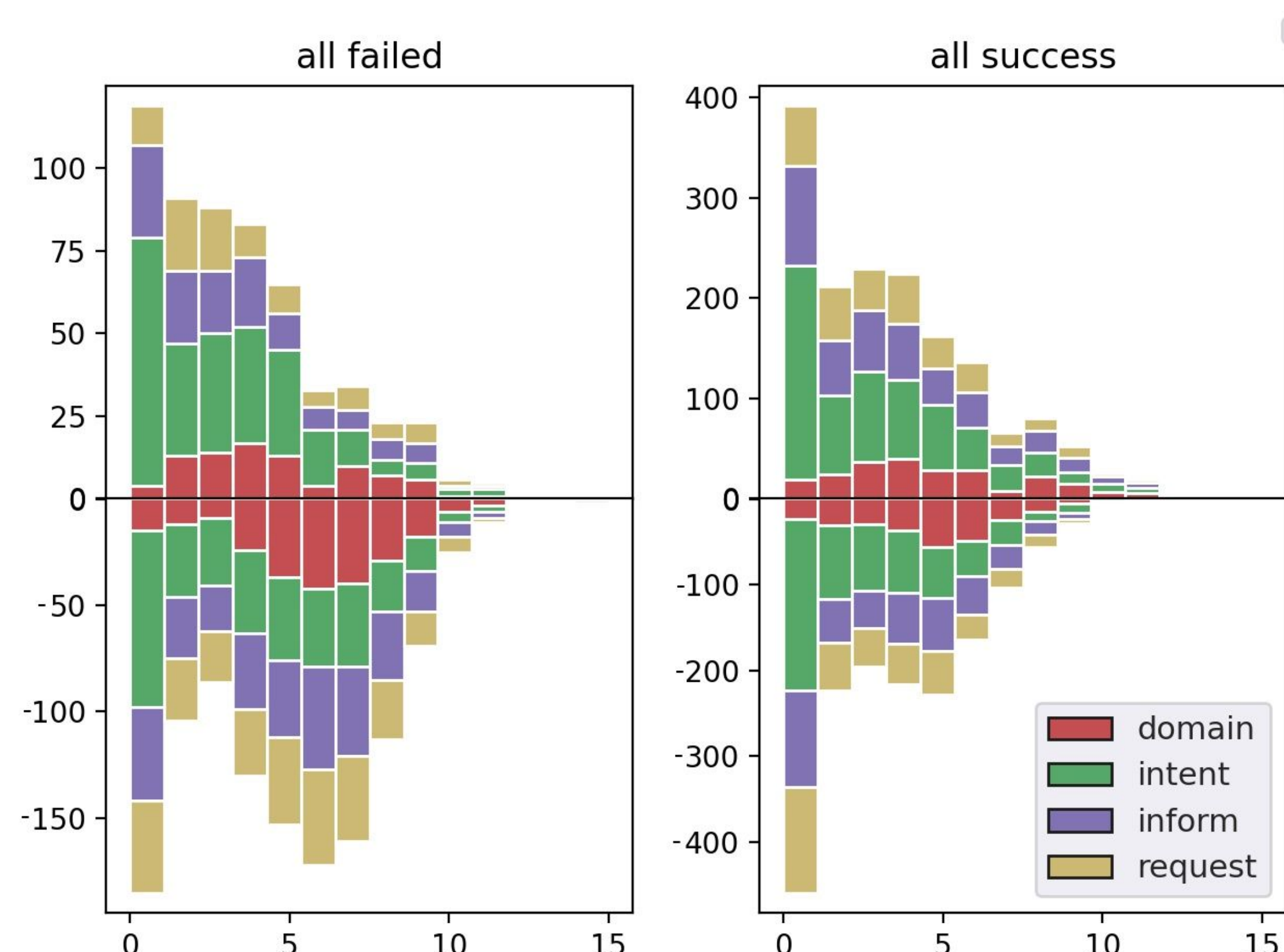
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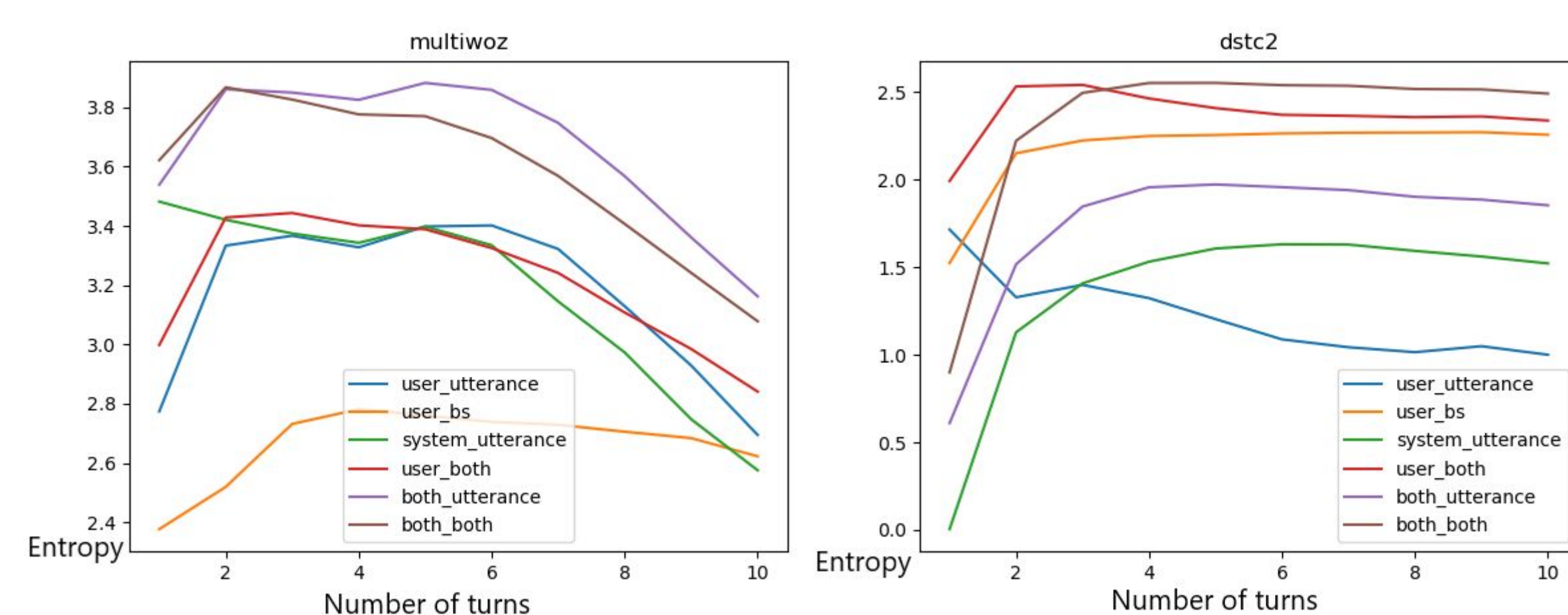
## Overview

- Unifying annotation & ontologies in 4 task-oriented dialogue datasets:
  - MultiWOZ, SGD, DSTC2, CamRest**
  - one of the largest annotated sets to date
- Data analysis and visualization
- Baseline model training and comparison

## Success/failure analysis



## Conditional entropy



## Analysis

- 3 dialogue phases** (based on conditional entropy)
  - information growth, stagnation, information deprecation
- Human-human vs. human-machine** entropy evolution differs
- Most **dialogue failures** due to missing information
  - Recoverable
  - It is correlated with entropy evolution.
- Best model** – trained on full data
  - SGD → better BLEU
  - MultiWOZ → better state tracking

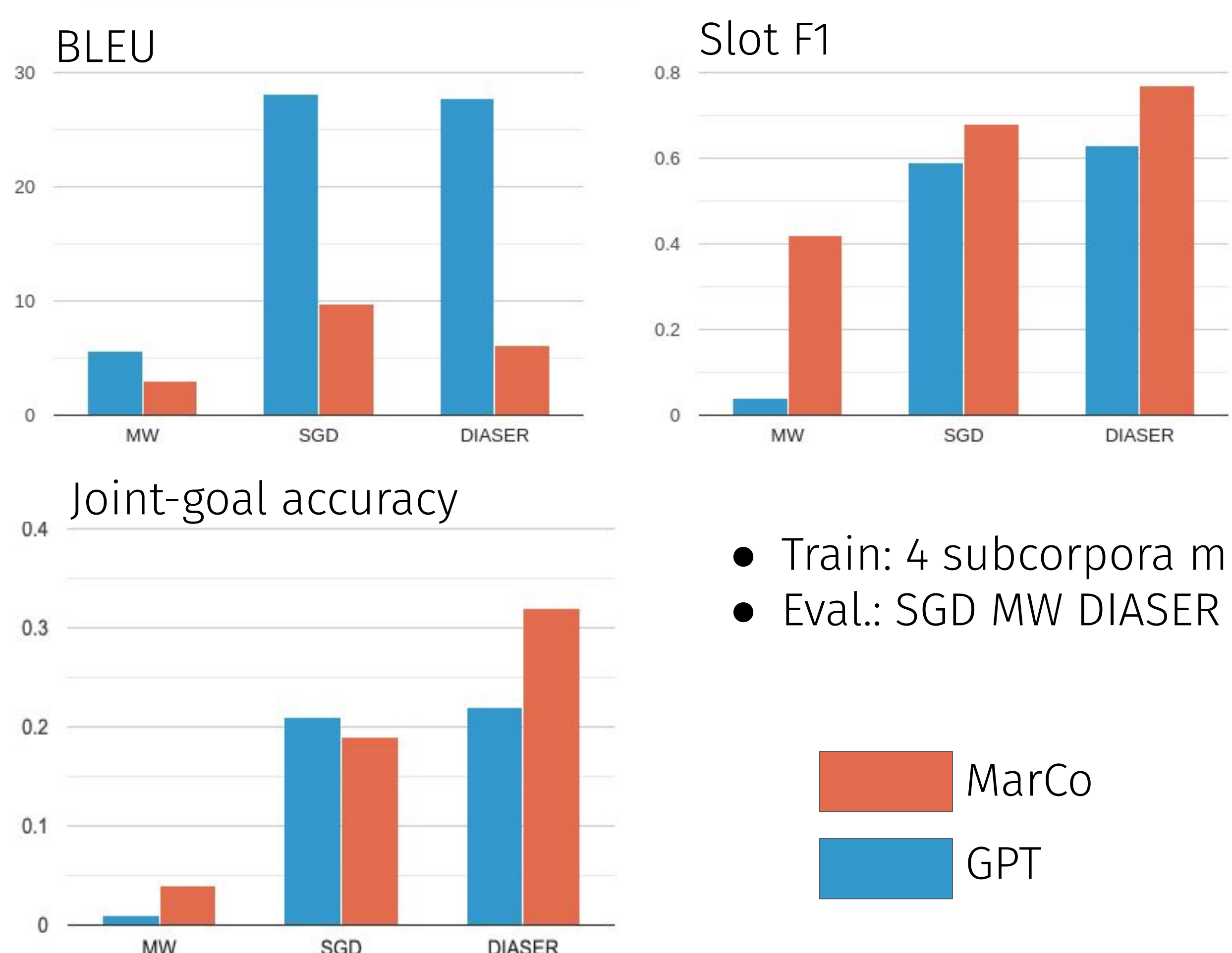
## Data Statistics

	SGD	MultiWOZ	DSTC2	Camrest	Total
Domains	18	7	1	1	19
Slots	145	29	10	7	166
Dialogues	22.8k	10.4k	3.2k	700	37.1k
Avg. utt. length	9.9	13.2	8.5	10.7	10.5
Entropy	4.8	4.4	2.1	3.0	4.8

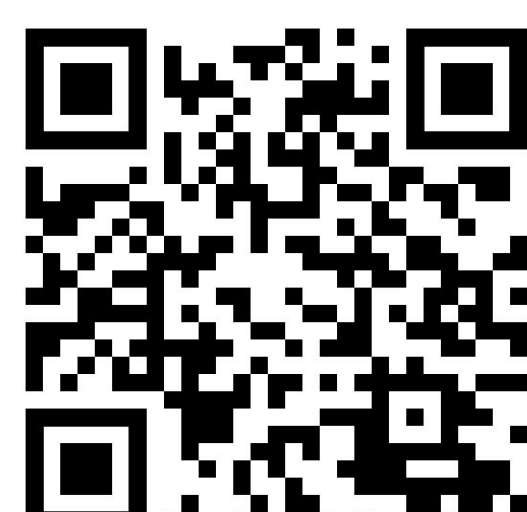
## Example

dialogue\_id: MUL0674.json, original\_dataset: multiwoz, ← origin IDs  
domains: train, hotel  
goal: hotel: { book: { day: thursday, people: 7, stay: 3 }  
info: { name: aylesbray lodge guest house } ... ← dialogue goal descriptions  
train: { book: { people: 7 }, info: { arriveBy ... }  
message: You are looking for information in Cambridge. You are looking for a train. The train should arrive by ...  
utterances: [  
{ ... actor: user ... I am looking for a train arriving by 21:45 and departing from cambridge. ... turn: 1, intent: train },  
{ ... actor: system ... What day are you making this trip, and where would you like to travel to? ... turn: 1, intent: train },  
{ actor: user,  
utterance: Hi I would like to go to kings lynn on sunday please.  
delex\_utterance: Hi I would like to go to [destination] on [day] please. ← delexicalized versions  
nlu: Train-Inform(day=Sunday, destination=kings lynn) ← consistent NLU & state format  
state: train-day: Sunday, train-departure: cambridge,  
train-destination: kings\_lynn, train-end\_time: 21:45 ← unified domain, intent, slot names  
turn: 2, intent: train },  
{ actor: system,  
utterance: Train TR1600 leaves Cambridge at 20:11 and arrives in Kings Lynn at 20:58. Would that work?  
delex\_utterance: Train [train id] leaves [departure] at [leave at] and arrives in [destination] at [arrive by]. Would that work?  
nlu: Train-Inform(arriveby=20:58, departure: Cambridge ...) &general-reqmore(),  
turn: 2, ... } ]

## Baseline results



- Train: 4 subcorpora mixes
- Eval.: SGD MW DIASER



Presented at LREC 2022, Marseille.

<https://github.com/ufal/diaser>

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