# Week 2: Data Management



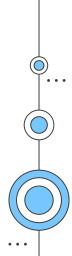


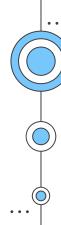
## Dialogue Sum Train Set

#### 4 Columns: ID, Dialogue, Summary, Topic:

- test\_id corresponds to the unique test of each row Will change this to a single integer ID
- **Dialogue** column contains the interaction between entities
- Summary column contains multiple summaries for a given dialogue
- **Topic** is a categorical variable that identifies the type of dialogue it corresponds to.

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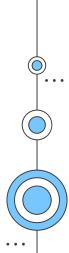


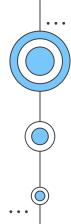
## **Dialogue Sum Insights**

#### **Topics column and duplicate dialogues/summaries:**

- Row 1626 and 9118 have the same dialogue but...they have different summaries
  of the dialogue. Only two cases of duplicate dialogues have been found in
  training dataset. Test set contains more.
- Likewise, we have found a total of 24 summary duplicates
- There are 7,434 different topics not sure if this will be a useful attribute to include
- Top 5 most popular are: shopping: 174, job interview: 161, daily casual talk: 125, phone call: 89, order food: 79

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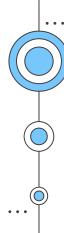




## **Example: Dialogue Duplicates**

 We can see that despite the duplicate dialogue, there contains a different topic categorization and summary.

	id	dialogue	summary	topic
1626	train_1626	#Person1#: Any plans tonight? \n#Person2#: Not	#Person1# invites #Person2# to have a drink be social	al communication
9118	train_9118	#Person1#: Any plans tonight? \n#Person2#: Not	#Person1# invites #Person2# to take a hang-out	after inteview (
	i	d dialog	gue summary	topic
7618	train_761	8 #Person1#: I'm searching for an old music box	#Person1# is searching for an old music box ma	old music box
11492	train_1149	2 #Person1#: I'm searching for an old music box	#Person1# wants to buy an old music box with d	music box



#### **Example: Duplicate Summary**

- Same topic, written differently.
- Dialogue for each summary are identical BUT were not detected due to small character differences (i.e. spaces)

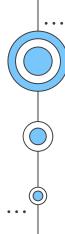
	id	dialogue	summary	topic	dialogue_tokens	summary_tokens	sentiment
108	train_1085	#Person1#: I'd like to take this opportunity $t$	#Person1# thanks #Person2# for #Person2#'s help.	gratitude	61.0	12.0	1
960	train 9605	#Person1#: I'd like to take this opportunity t	#Person1# thanks #Person2# for #Person2#'s help.	thanks	62.0	12.0	0

#### train\_df.iloc[1085, 1]

'#Person1#: I'd like to take this opportunity to thank you for everything you did for me.\n#Person2#: It's my pleasure. I enjoyed working with you.\n#Person1#: I wouldn't be able to make it without your help.\n#Person2#: Then keep up the good work.'

#### train\_df.iloc[9605, 1]

'#Person1#: I'd like to take this opportunity to thank you for everything you did for me. \n#Person2#: It's my pleasure. I enjoyed working with you. \n#Person1#: I wouldn't be able to make it without your help. \n#Person2#: Then keep up the good work.



#### **Data Management**

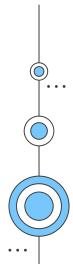
- Group data based on Dialogue.
- May need to use ML techniques to process dialogue for similarity - make uniqueness reliable and reduce redundancy
- Currently, thoughts are on creating 4 separate entity tables:

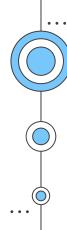
LLM's { Ilmid; Type; } Dialogue {
 Did;
 Dialogue;
 dtokens;
 }

Summary {
 Sid;
did (references dialogue);
 stokens;
}

Scores {
LImid (references LLMs);
Rogue;
Bleu;
Meteor;
}

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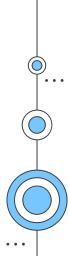


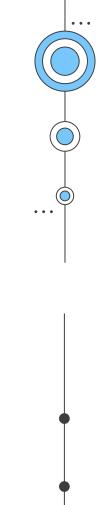


## **Final Thoughts**

- Dialogues and summaries seem to have strong similarities that are sometimes the exact same minus a character or so
- Inclusion of sentiment column with the appropriate labels
- Topics columns seem too varied
- Inclusion of token counts might be useful for dialogue and summary tables

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# Thank you



