

# Multi-Intent Utterance Dataset Construction for Spoken Language Understanding

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# Contents

1

## PRE-REQUISITE

Key Terms

2

## INTRODUCTION

Problem States / Background

3

## DISCUSSION 1 : Dataset Construction

Utterance Concatenation / Utterance Selection

4

## DISCUSSION 2 : Experiments and Analysis

3 Custom Metrics / SOTA Models Evaluation / Visualization

5

## CONCLUSION

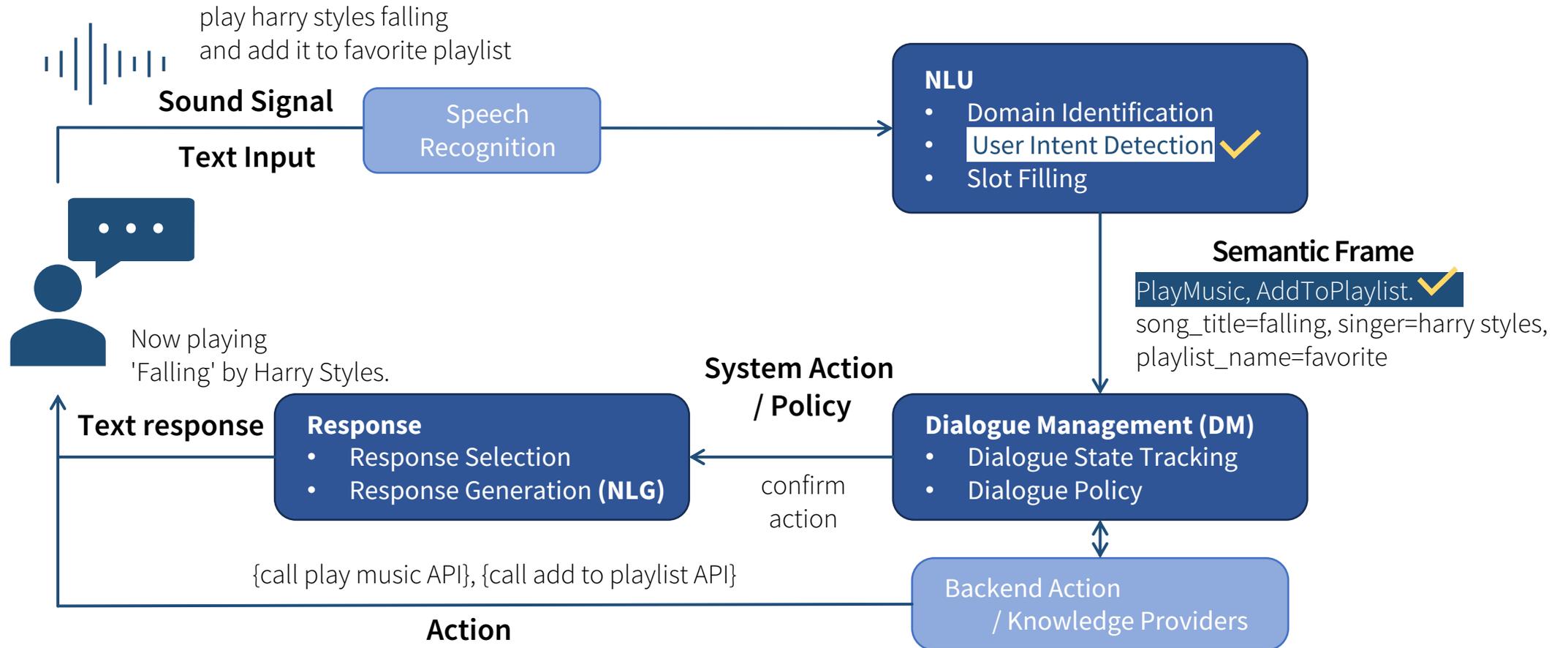
Contribution

# PRE-REQUISITE

# Task-oriented Dialogue System

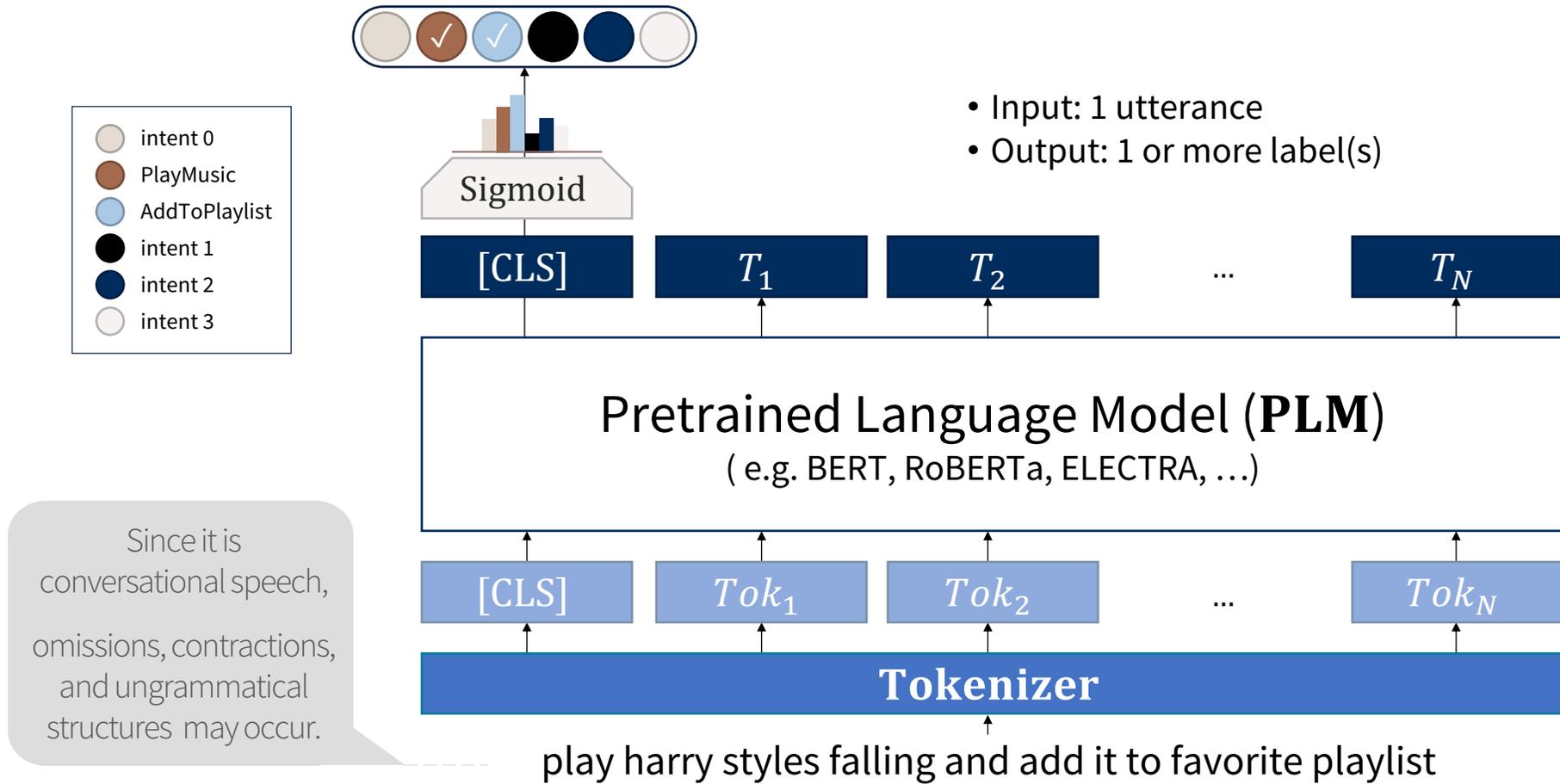
# Multi-intent Detection

# Task-oriented Dialogue System



- Help users achieve their specific goals
- Focus on **understanding users**, tracking states, and generating next actions

# Multi-intent Detection



Identify and respond to multiple intentions or requests within a single user utterance

# Introduction

# Problem States

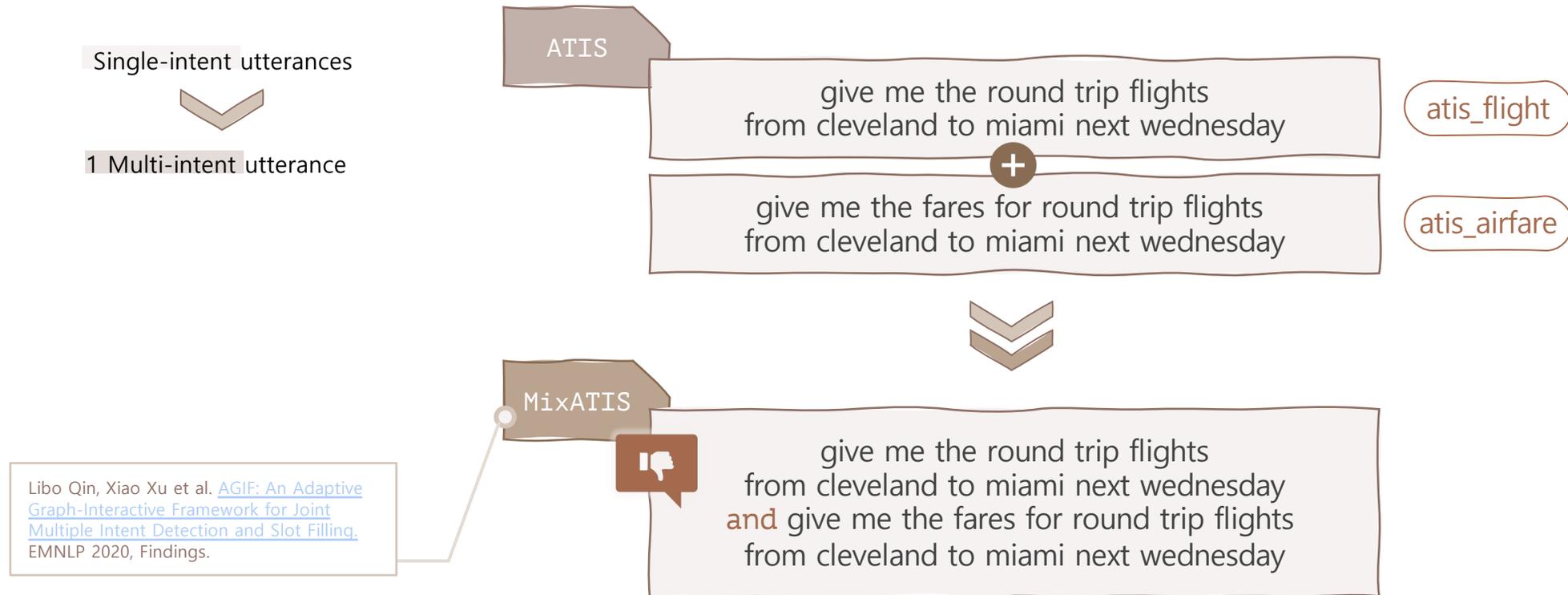
# Background

# Current Trends in MID

Model	MixATIS			MixSNIPS		
	Slot (F1)	Intent (Acc)	Overall (Acc)	Slot (F1)	Intent (Acc)	Overall (Acc)
SF-ID ( <i>concat</i> ) (2019)	87.4	66.2	34.9	90.6	95.0	59.9
Stack-Propagation ( <i>thresh</i> = 0.5) (2019)	87.8	72.1	40.1	94.2	96.0	72.9
Joint Multiple ID-SF ( <i>thresh</i> = 0.5) (2019)	84.6	73.4	36.1	90.6	95.1	62.9
AGIF ( <i>thresh</i> = 0.5)(2020)	86.7	74.4	40.8	94.2	95.1	74.2
GL-GIN ( <i>thresh</i> = 0.5)(2021)	<b>88.3</b>	76.3	43.5	94.9	95.6	75.4
SDJN ( <i>thresh</i> = 0.5)(2022a)	88.2	77.1	44.6	94.4	96.5	75.7
SDJN+BERT ( <i>thresh</i> = 0.5)(2022a)	87.5	78.0	46.3	95.4	96.7	79.3
Bert-baseline ( <i>thresh</i> = 0.3)	83.1	74.8	42.6	95.5	95.7	80.2
Bert-baseline ( <i>thresh</i> = 0.5)	86.3	74.5	44.8	95.5	95.6	80.1
Bert-baseline ( <i>thresh</i> = 0.8)	85.6	75.8	43.5	95.2	96.7	80.6
<b>TFMN (Bert-base)</b>	88.0	<b>79.8</b>	<b>50.2</b>	<b>96.4</b>	<b>97.7</b>	<b>84.7</b>

- **Transformer-Based** Models in MID: Achieving Unprecedented High Performance.
- Competitive Edge: Latest Models Contending in Subtle Decimal Point Differences.
- MID and Slot Filling: **Jointly-learning** (multi-task learning) in Advanced Research Fields.

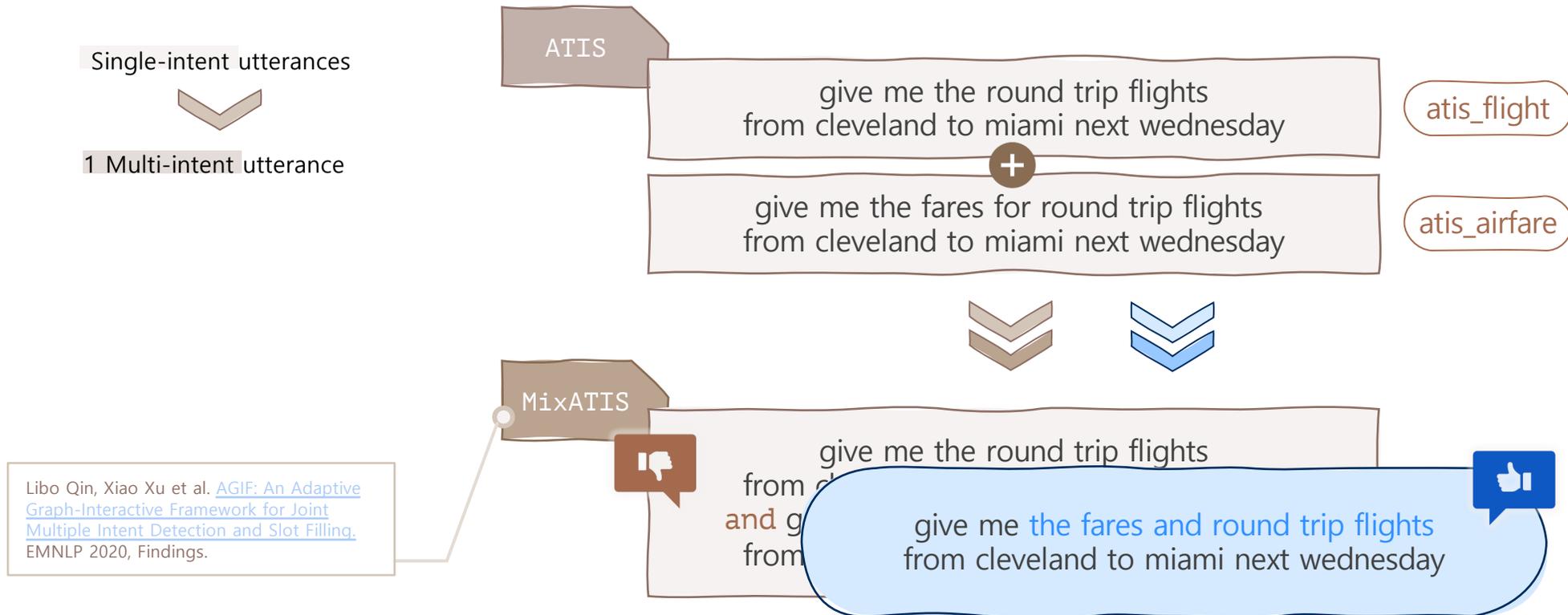
# Benchmark Datasets Analysis (1/2)



The dataset relies on **only** a few specific **connectors** ("and", "and then", "and also") when concatenating 2 or more single-intent utterances.

→ Real-world conversations often involve **more varied and complex ways of combining intents**

# Benchmark Datasets Analysis (2/2)



We are focused on constructing our own set that better mirrors natural language usage to provide more **challenging** and **realistic** resources for training and testing multi-intent detection models.

# Discussion 1 : Dataset Construction

# Concatenation: Single- to Multi- intent utterance

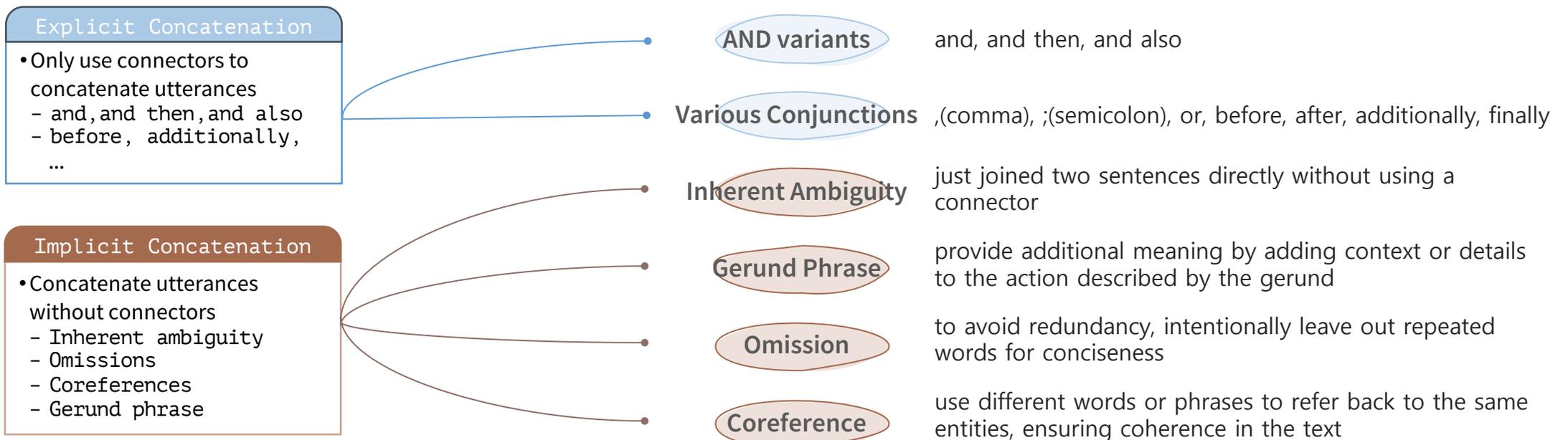
# Utterance Selection

# Categories of Concatenation Complexity-side

## • Complexity side

- Explicit Concatenation: use connectors during concatenation
  - AND variants / Various Conjunctions
- Implicit Concatenation: do NOT use connectors during concatenation
  - Inherent Ambiguity / Gerund Phrase / Omission / Coreference

### Complexity side



# (Intuitive) ChatGPT in Concatenation (1/2)

## • Prompt Engineering for ChatGPT Concatenation

```
You are a native English speaker.
[Task Definition] Combine 2 or 3 sentences as one single sentence.
[Goal] The focus is on creating a single sentence that captures the essence of both ideas without unnecessary redundancy.
[Instructions] - Avoid adding just punctuation.
               - Don't paraphrase.
               - Don't compromise the meaning of each sentence.
               - Don't capitalize all characters.
               - Don't replace numbers with radix.
               - Maintain the intent of each sentence.
               - Don't forget that if a sentence starts with a verb, it's a statement.
               - Do NOT use conjunctions like 'and'
               - Don't print '[Good Answer]'
               - Don't print intent directly.

[Example 1]
My dog is playful (dog's feature) + My dog loves chasing balls (dog's feature)
[Good Answer] My playful dog loves chasing balls
[Bad Answer] My dog is playful, and my dog loves chasing balls
[Bad Answer] My dog is playful, and also loves chasing balls.

[Example 2]
They finished the project(project done) + They had time(taking time)
[Good Answer] With time on their hands, they finished the project
[Bad Answer] They finished the project, and they had time
[Bad Answer] They finished the project and had time

[Example 3]
She answered the phone (answering)) + She was making dinner (cooking)
[Good Answer] While answering the phone, she was making dinner
[Bad Answer] She answered the phone, and She was making dinner

Combine the following sentences naturally. Inside the parentheses is the intent of each sentence. :
{utt1} (intent: {intents[0]}) + {utt2} (intent: {intents[1]})
```

Returning results that don't follow the explicit constraints we gave ChatGPT

Few-shot setting

# (Intuitive) ChatGPT in Concatenation (2/2)

## • Failure of Using ChatGPT

- Over 90% of the 100 randomly sampled instances were connected by 'and'
- Failed to maintain intent in at least 15 instances
  - Intent removed / changed
  - Fail to merge the utterances

Success	Utterance 1	Utterance 2	Concatenation	Implication
X	weather for frankfort (GetWeather)	3 out of 6 for the last album (RateBook)	frankfort's weather gets a 3 out of 6 rating for the last album	Intent (RateBook) removed
	show me the fares for a first class ticket from baltimore to dallas (atis_airfare)	i would like the time of all flights from san francisco to pittsburgh on sunday (atis_flight_time)	show me the fares for a first class ticket from baltimore to dallas and i would like to know all flights from san francisco to pittsburgh on sunday	Intent changed (from atis_flight_time to atis_flight)
	what is mci (atis_abbreviation)	list la (atis_city)	combine the sentences: "what is mci?" and "list la".	Failed to merge
O	play the playlist funtime activity (AddToPlaylist)	add adele onto funtime activity playlist (PlayMusic)	play the playlist, adding adele to the funtime activity playlist	Gerund phrase
	how do i freeze my bank account (account_blocked)	do you know why my bank account is frozen (freeze_account)	how do i freeze my bank account, and do you know why it is frozen	Coreference

Table 1: Failed and successful results of the Generative Approach and their implications.

Only few cherry-picked examples were, brilliantly, implicitly concatenated, which is what we intended.

# (Intuitive) ChatGPT in Concatenation (2/2)

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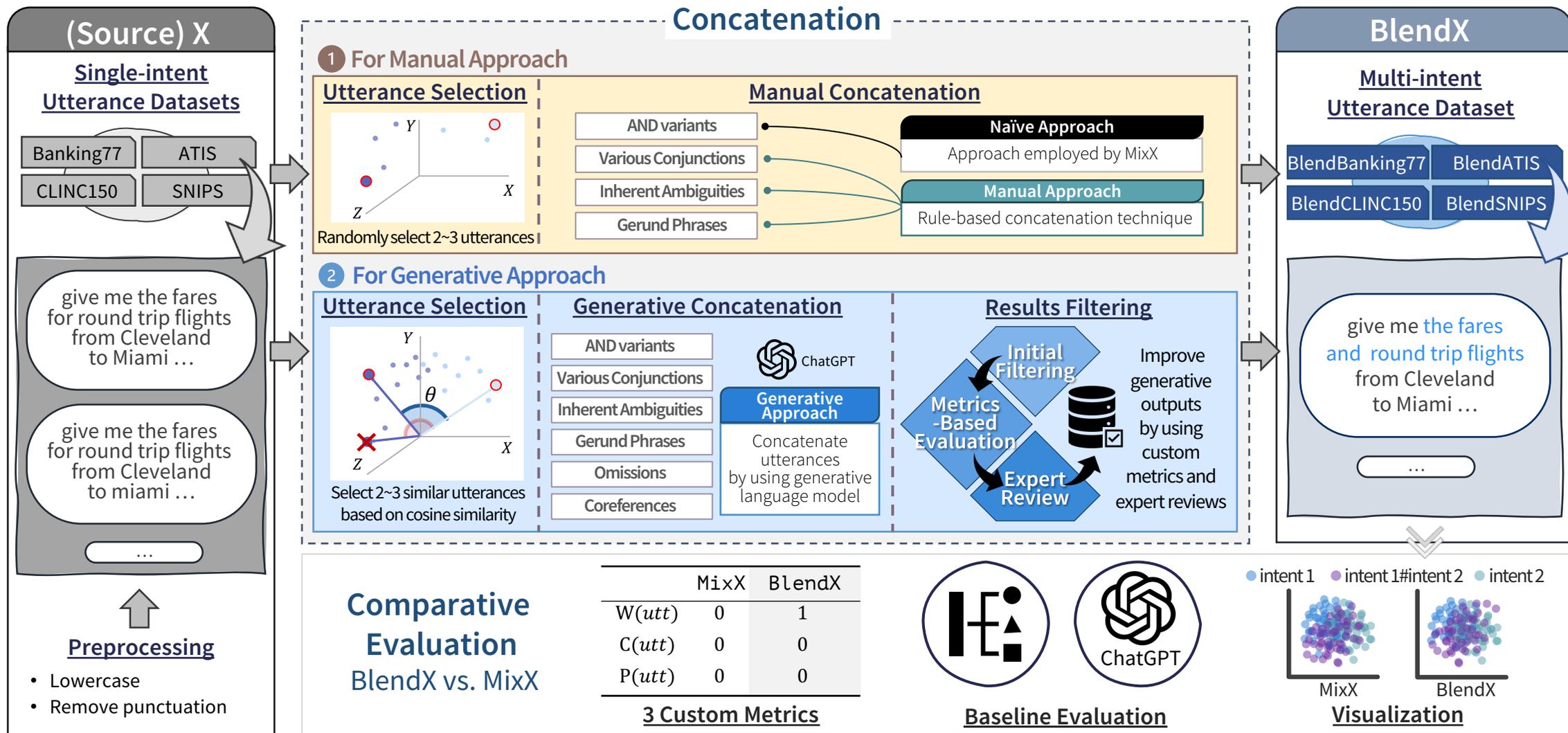
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# Overview of 2 Methods to Utterance Concatenation



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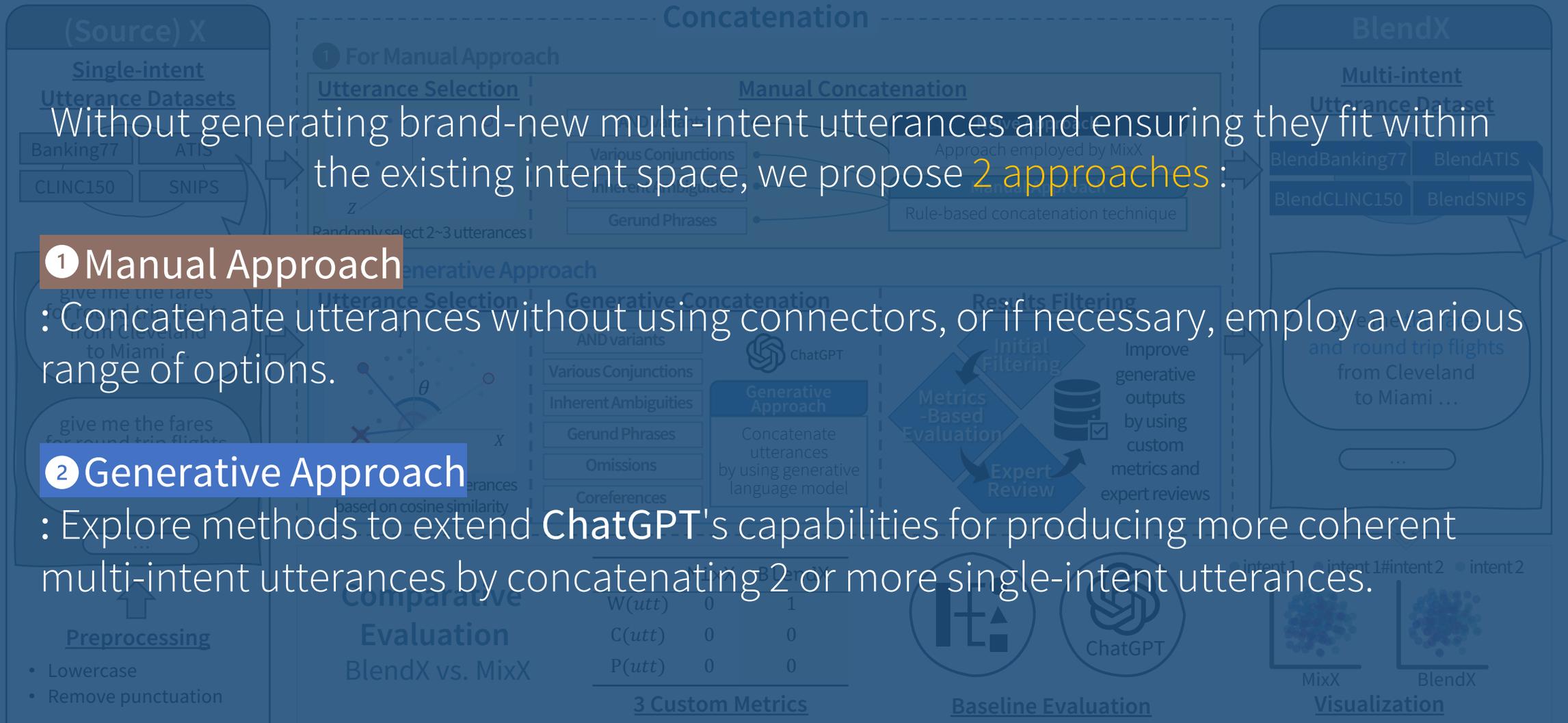
Without generating brand-new multi-intent utterances and ensuring they fit within the existing intent space, we propose **2 approaches**:

## 1 Manual Approach

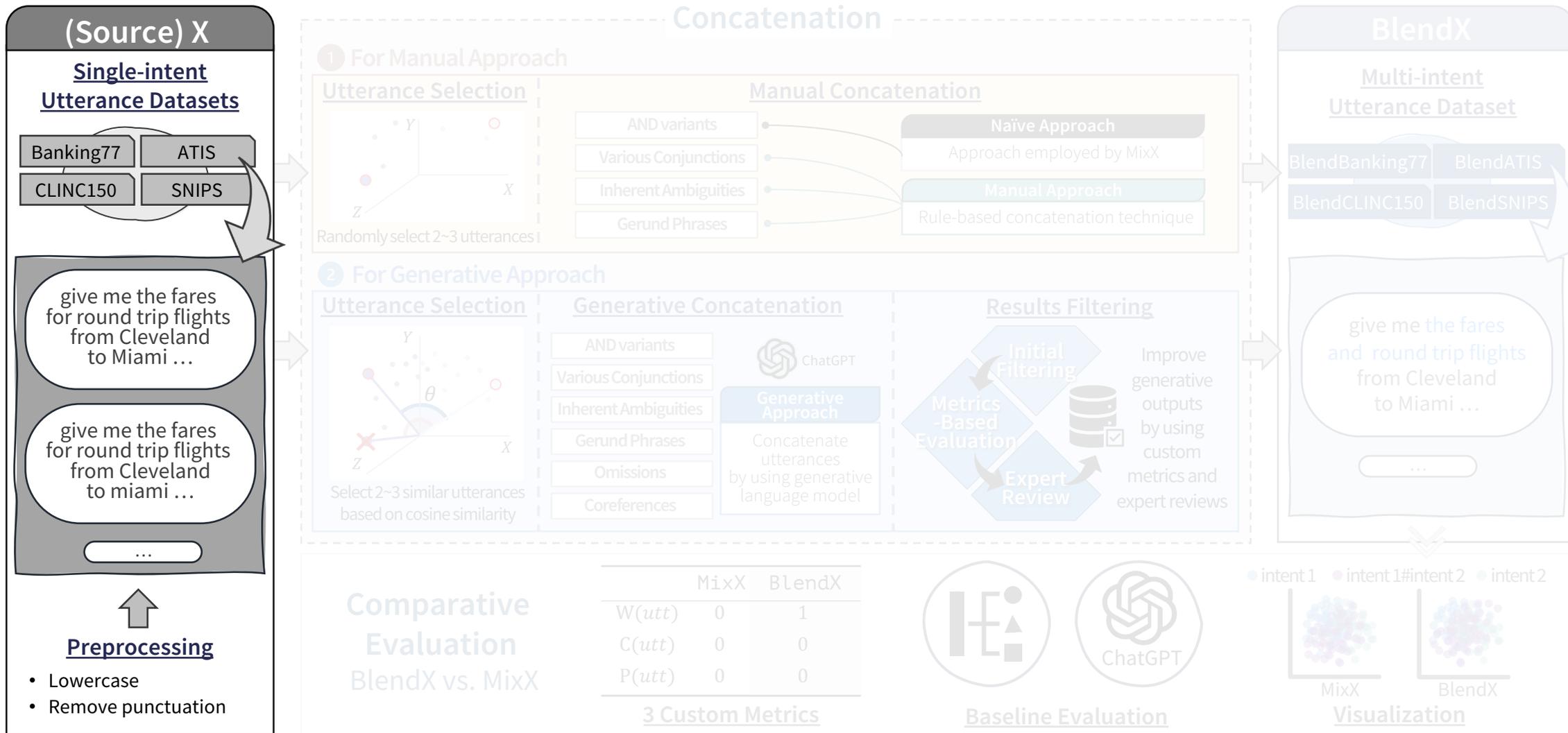
: Concatenate utterances without using connectors, or if necessary, employ a various range of options.

## 2 Generative Approach

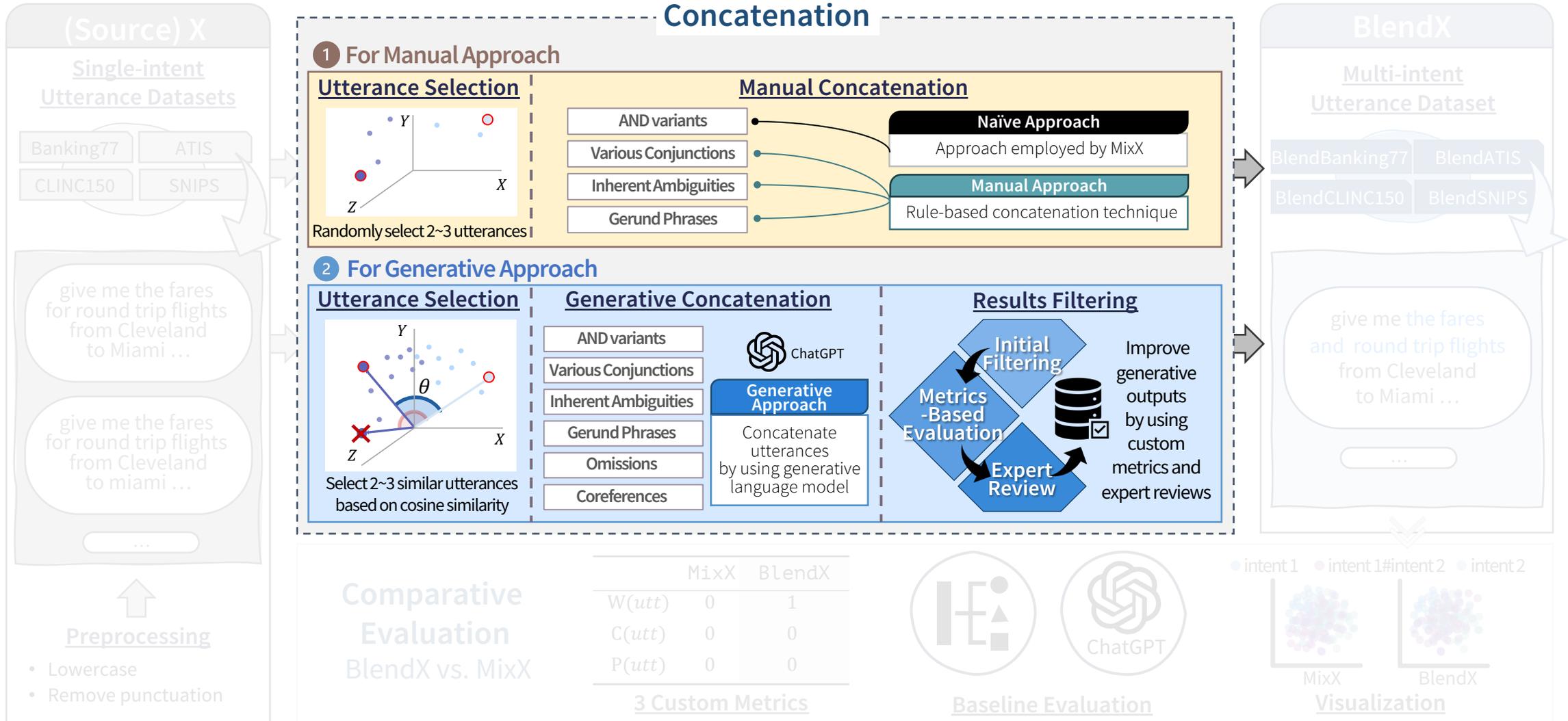
: Explore methods to extend **ChatGPT's** capabilities for producing more coherent multi-intent utterances by concatenating 2 or more single-intent utterances.



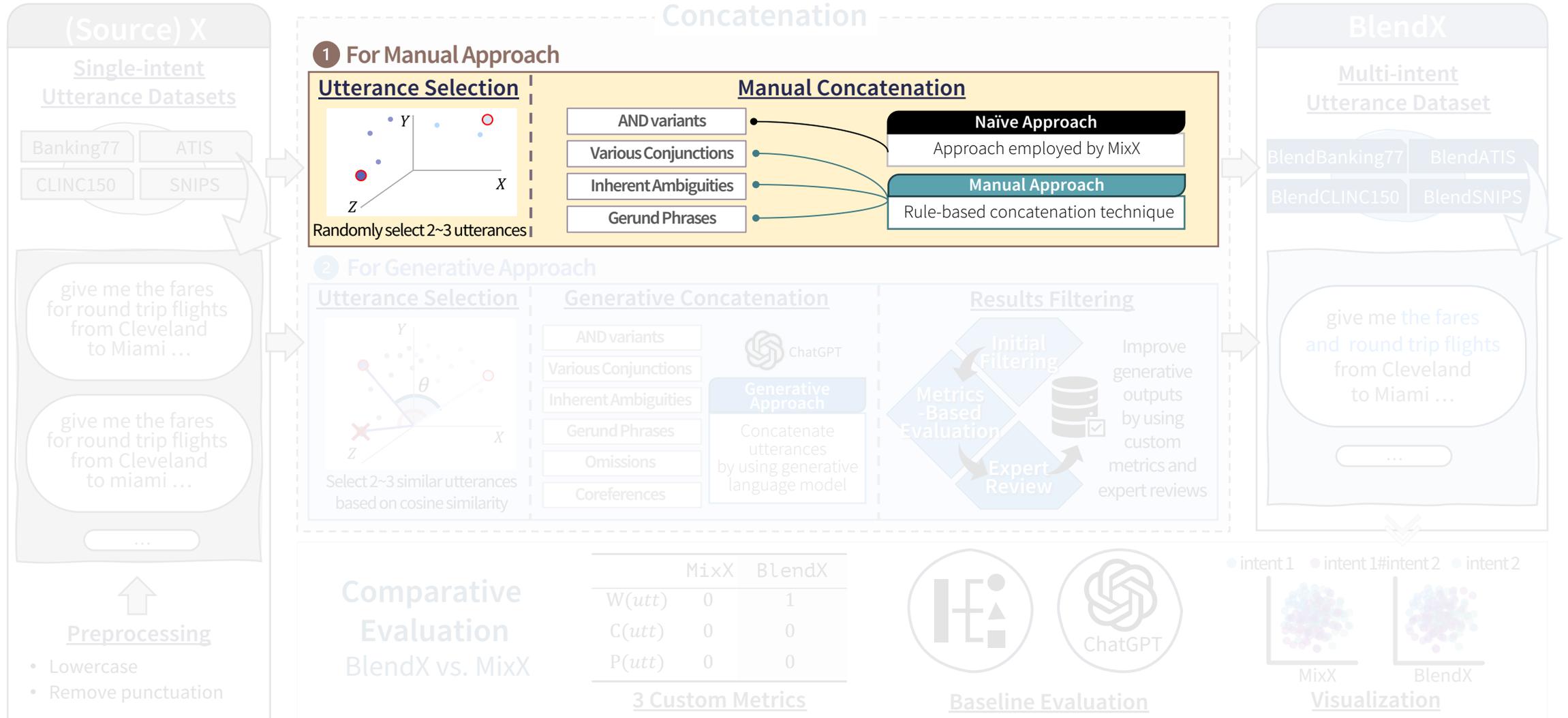
# 1) Preprocessing



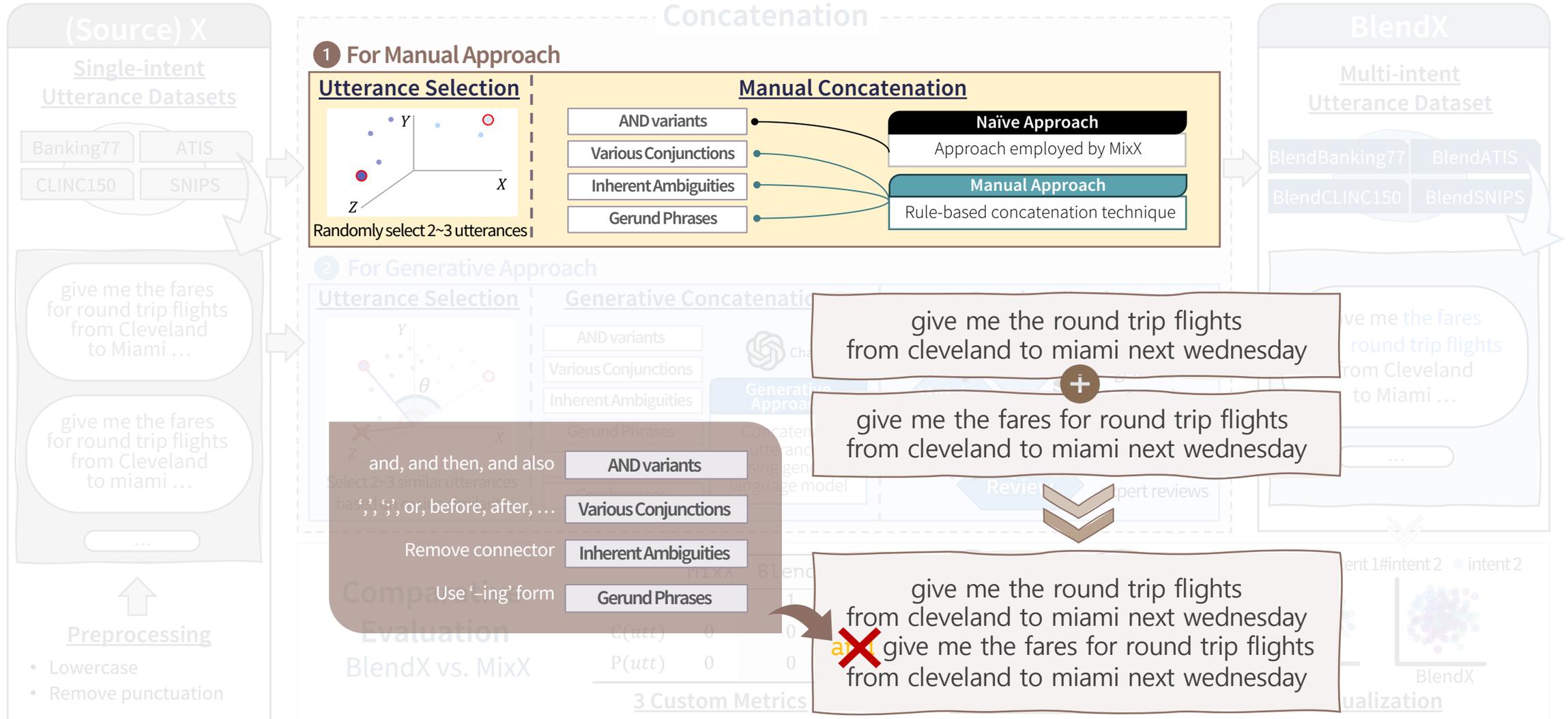
# 2) Concatenation



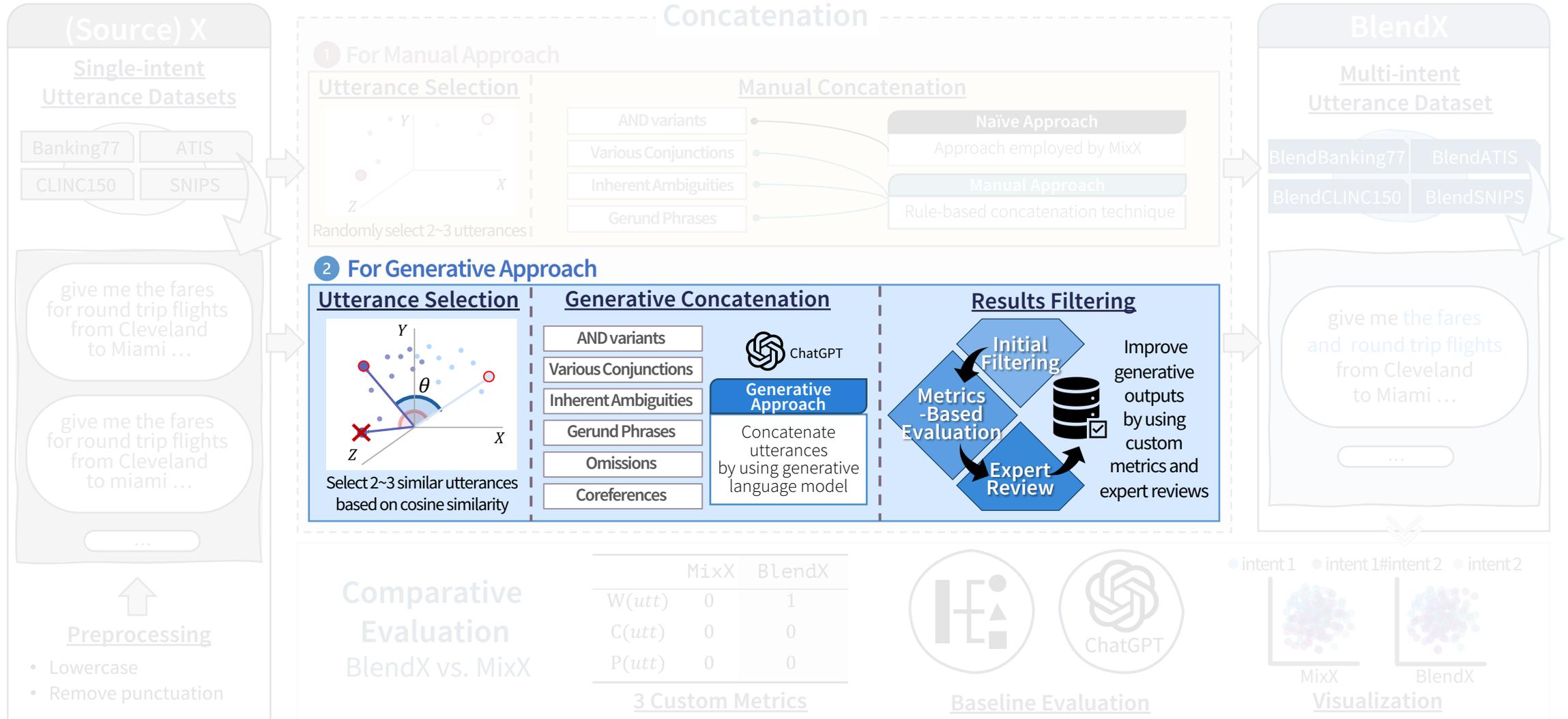
# 2-1) Manual Approach



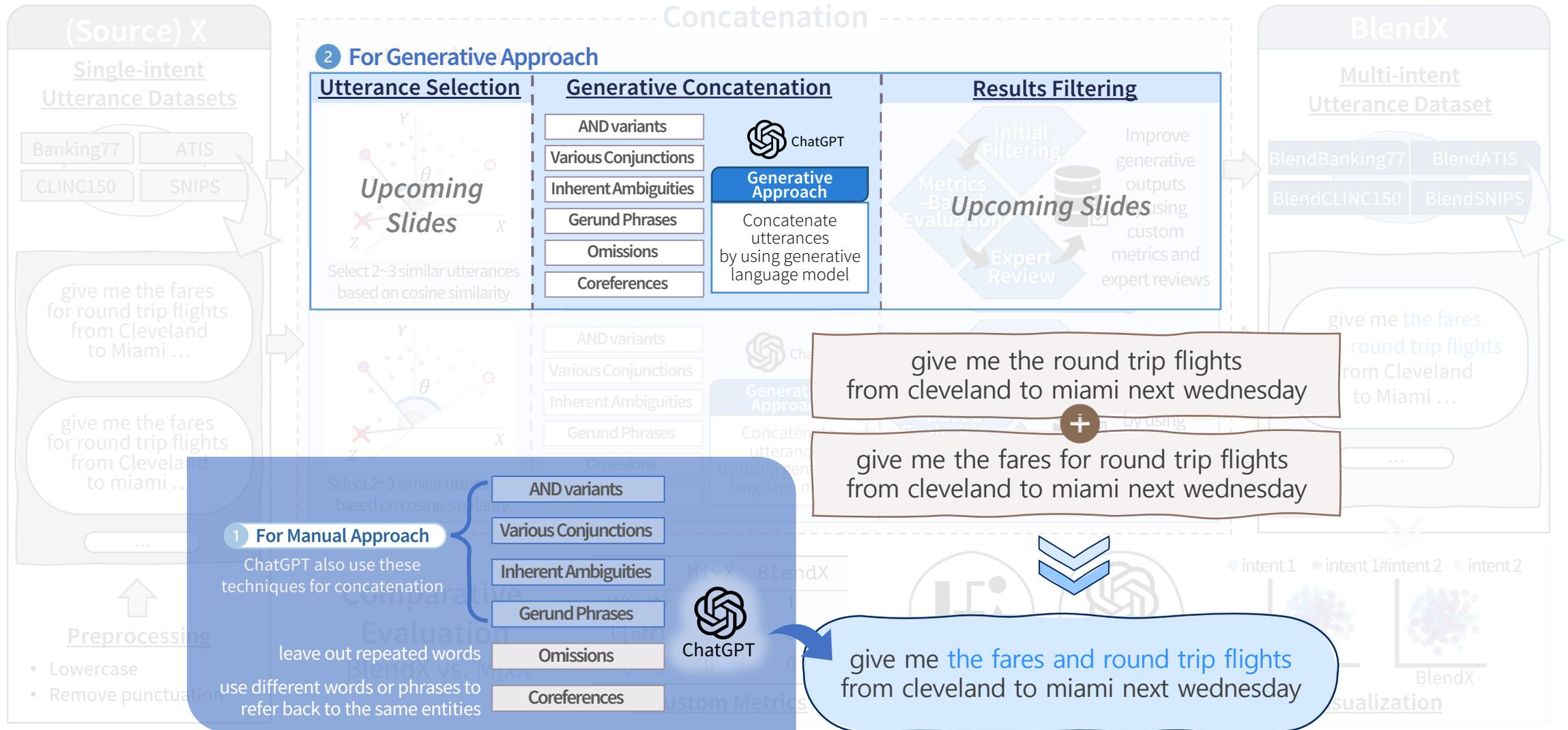
# 2-1) Manual Approach



# 2-2) Generative Approach



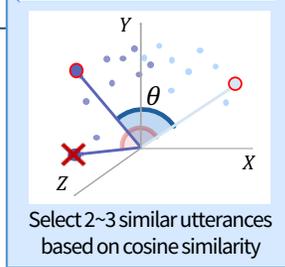
# 2-2) Generative Approach



# Utterance Selection for 2-2) Generative Approach

2 For Generative Approach

Utterance Selection



Generative Concatenation

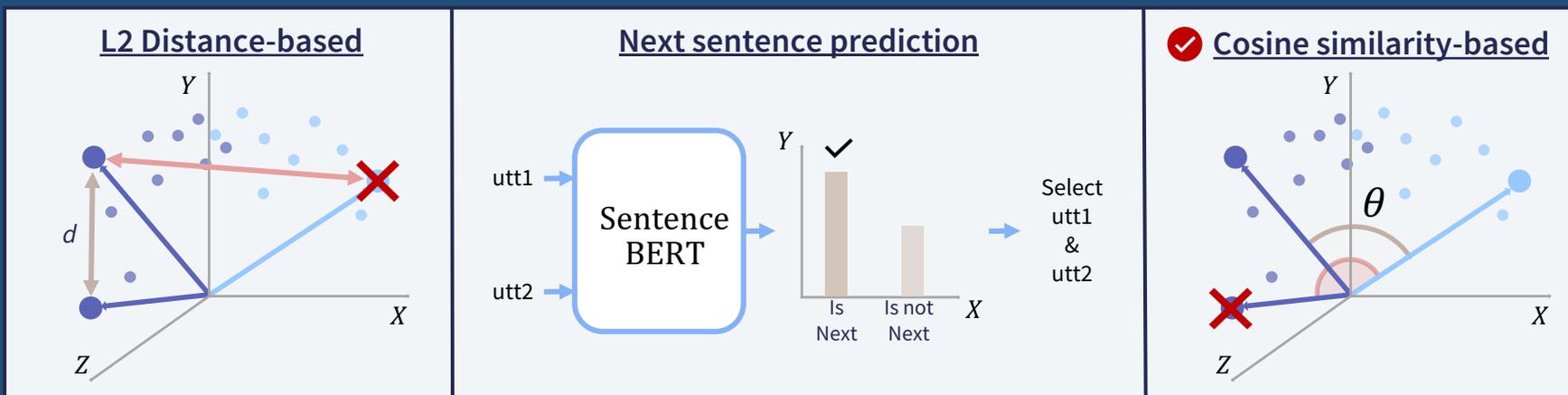
AND variants	Concatenation by using language
Various Conjunctions	
Inherent Ambiguities	
Gerund Phrases	
Omissions	
Coreferences	

## • Process

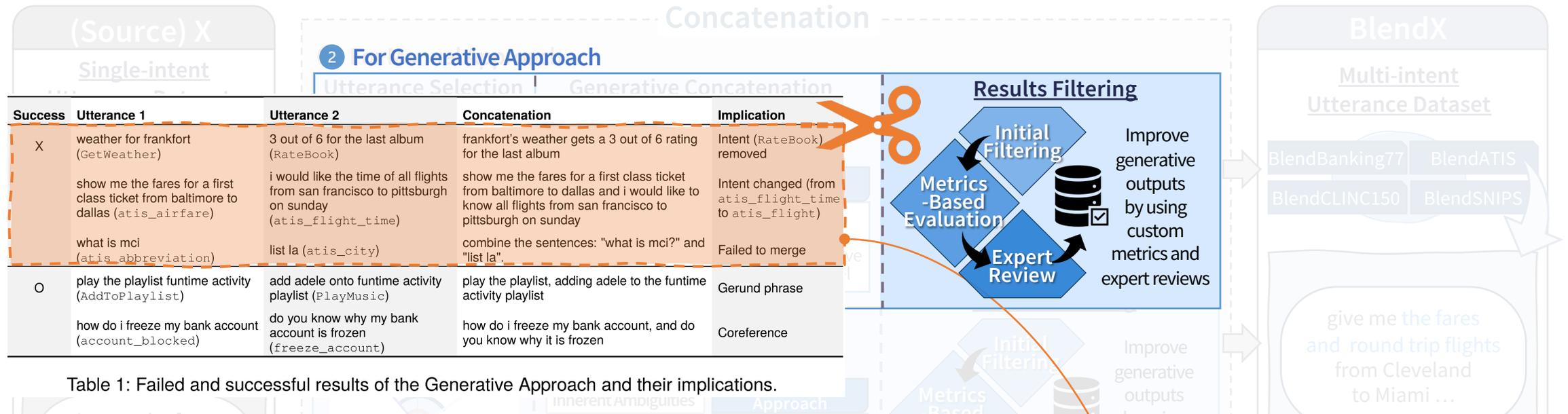
1. Generate embeddings for each single-intent utterance using SentenceBERT.
2. Select utterances for concatenation based on high similarity between embeddings.
  - \* Chosen utterances will have different intents.

## • Selection approach

- L2 Distance-based: Select utterances with close proximity in embedding space.
- Next sentence prediction: Binary classification of whether a given pair of utterances are sequential.
- ✓ - **Cosine similarity-based**: Choose utterances with high cosine similarity between embeddings.



# Results Filtering for 2-2) Generative Approach



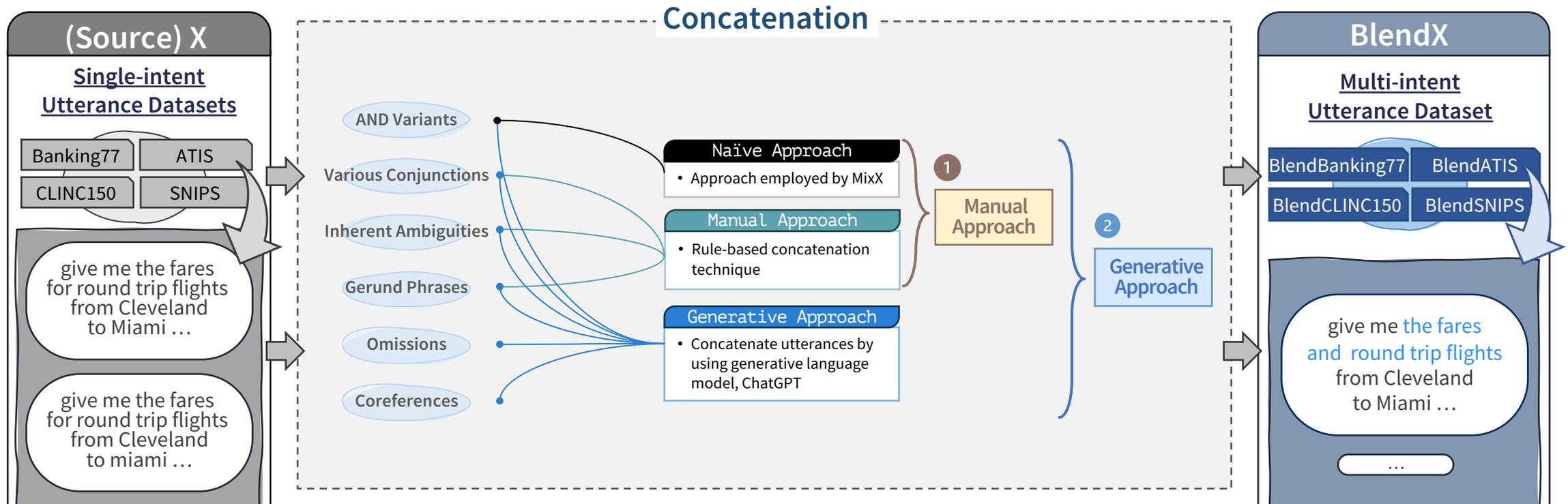
## 1. Remove clear failures generated by ChatGPT

: explicit mentions of an intent label, unnecessary punctuation, ...

## 2. Use metrics to evaluate integration diversity and complexity and collect results above a set threshold.

## 3. Human-expert review of the generated output filtered by the above two steps and remove if a generated utterance deviated significantly from its original intents.

### 3) BlendX – Multi-intent utterance datasets



#### BlendX : Complex Multi-intent Detection with Blended Patterns

Dataset	Intents #	Training	Dev	Test	Total
SNIPS	7	50,625	2,613	2,615	55,853
ATIS	18	20,250	1,125	1,125	22,500
Banking77	77	36,390	2,009	2,021	40,420
CLINC150	147	54,896	2,889	2,977	60,762

- Source Dataset : SNIPS, ATIS, Banking77, CLINC150
- Random selection for **Manual** Concatenation Approach
- Cosine Similarity-based selection for **Generative** Concatenation Approach

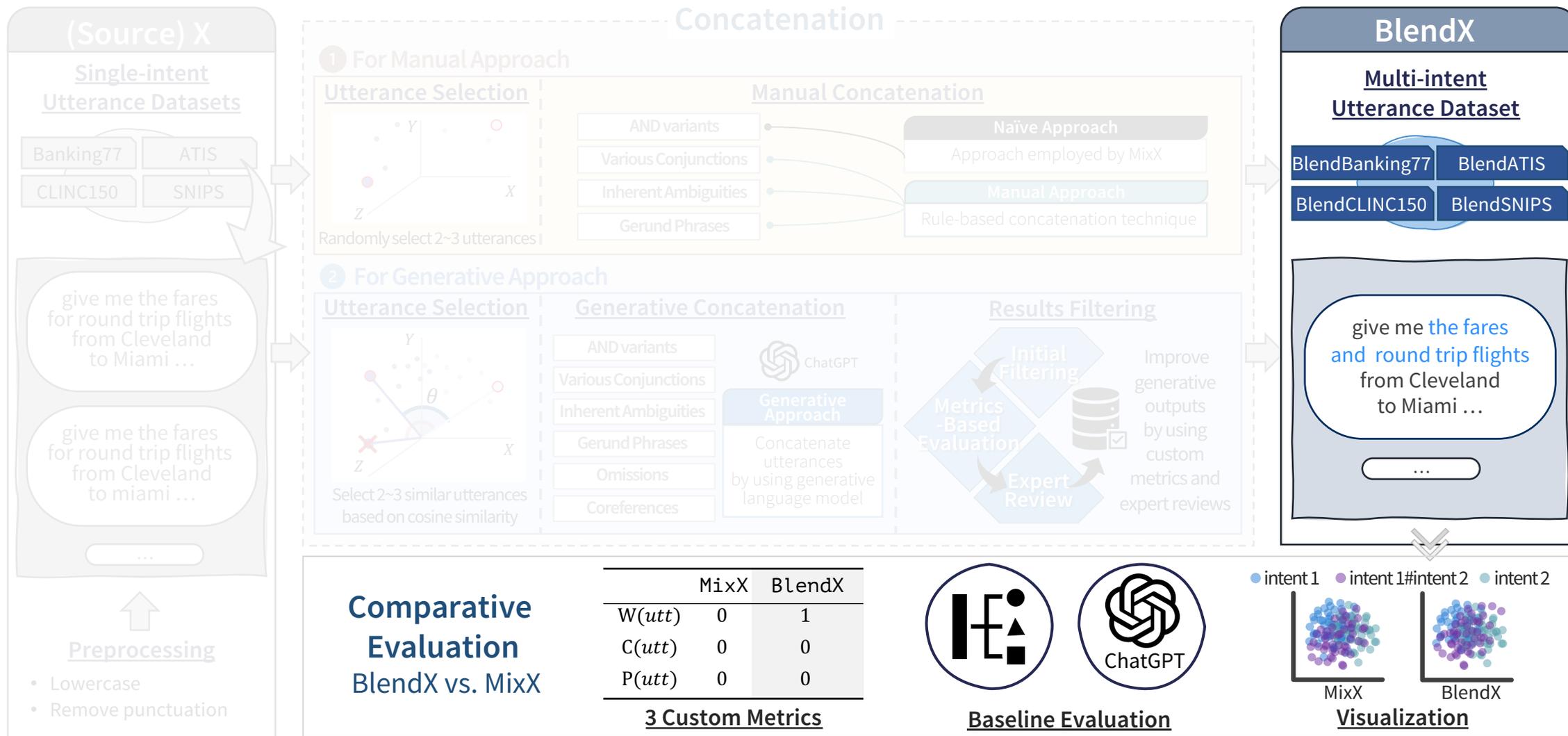
# Discussion 2: Experiment and Analysis

# 3 Custom Metrics Evaluation

# SOTA Models Evaluation

# Visualization of BlendX

# 4) Comparative Evaluation – BlendX vs. MixX



## 4-1) 3 Custom Metrics (1/3)

### • 3 Custom Metrics

- *utt*: concatenated utterance with 2 or more intents
- *n*: Number of single-intent utterances used for concatenation

#### $W(utt, n)$

Word count

$$W(utt, n) \stackrel{\text{def}}{=} \mathbf{1}_{\mathbb{Z}-\mathbb{N}} \left( |utt|_{word} - \sum_{i=1}^n |utt_i|_{word} \right).$$

Check if the **word count** difference before and after an utterance concatenation is zero or negative

( to ascertain a decrease in word count )

#### $C(utt, n)$

Conjunction

$$C(utt, n) \stackrel{\text{def}}{=} \mathbf{1}_{\mathbb{Z}-\mathbb{N}} \left( |utt|_{conj} - \sum_{i=1}^n |utt_i|_{conj} \right).$$

Verify if the number of **conjunctions** before and after an utterance changes to zero or less

( to determine the elimination or reduction of conjunctions )

\* **conjunctions** such as:  
and, or, before, after, additionally,  
finally, ‘,’, ‘;’

#### $P(utt, n)$

Pronoun

$$P(utt, n) \stackrel{\text{def}}{=} \mathbf{1}_{\mathbb{N}} \left( |utt|_{pron} - \sum_{i=1}^n |utt_i|_{pron} \right).$$

Assess if the difference in **pronoun count** before and after an utterance is one or more

( to identify the usage of pronouns )

\* **pronoun** such as :  
it, them, their, theirs, this, that,  
those, these

An **implicitly** concatenated utterance is likely to receive **1** in the metrics evaluation.

## 4-1) 3 Custom Metrics (2/3)

### • Example of applying 3 metrics

Utterance 1	play my 88 keys playlist (PlayMusic)			
Utterance 2	add another song to my 88 keys playlist (AddToPlaylist)			
Strategies	Concatenation Results	$W(utt, 2)$	$C(utt, 2)$	$P(utt, 2)$
<b>Explicit Concatenation</b>	play my 88 keys playlist <b>and</b> also add another song to my 88 keys playlist	0	0	0
<b>Implicit Concatenation</b>				
Inherent Ambiguity	play my 88 keys playlist add another song to my 88 keys playlist	1	1	0
Omissions	play my 88 keys playlist and add another song	1	0	0
Coreferences	play my 88 keys playlist and add another song to it	1	0	1
Gerund Phrase	add another song to my 88 keys playlist playing it	1	1	1

Table 3: Various concatenation classes, accompanied by their examples and respective metric values.

	Concatenated Utterance	Utterance 1	Utterance 2	Difference	Metric
<b>example #1</b>	add another song to my 88 keys playlist playing it	play my 88 keys playlist	add another song to my 88 keys playlist		
<b>Words</b>	10	5	8	$10 - (5 + 8) = -3$	$W(\cdot, 2) = 1$
<b>Conjunctions</b>	0	0	0	$0 - (0 + 0) = 0$	$C(\cdot, 2) = 1$
<b>Pronouns</b>	1	0	0	$1 - (0 + 0) = 1$	$P(\cdot, 2) = 1$
<b>example #2</b>	i need to clear my to-do list and then repeat it	clear my to do list	repeat my to do list		
<b>Words</b>	11	5	5	$11 - (5 + 5) = 1$	$W(\cdot, 2) = 0$
<b>Conjunctions</b>	1	0	0	$1 - (0 + 0) = 1$	$C(\cdot, 2) = 0$
<b>Pronouns</b>	1	0	0	$1 - (0 + 0) = 1$	$P(\cdot, 2) = 1$

## 4-1) 3 Custom Metrics (3/3)

### • Results using 3 metrics for each approach

Metric	SNIPS			ATIS			Banking77			CLINC150		
	Naïve	Manual	Generative	Naïve	Manual	Generative	Naïve	Manual	Generative	Naïve	Manual	Generative
$W(utt, 2)(\uparrow)$	0%	37%	29%	0%	36%	18%	0%	46%	37%	0%	48%	28%
$C(utt, 2)(\uparrow)$	0%	56%	10%	0%	52%	15%	0%	50%	27%	0%	56%	32%
$P(utt, 2)(\uparrow)$	0%	0%	7%	0%	0%	8%	0%	0%	13%	0%	0%	6%

Table 4: Comparative analysis of the three concatenation approaches: Naïve, Manual, and Generative. Notably, the Manual method demonstrates pronounced efficiency in reducing utterance length.

Our approach, incorporating both **manual** and **generative** methods, achieves a more **diverse range** of explicit and **implicit** concatenation compared to existing techniques.

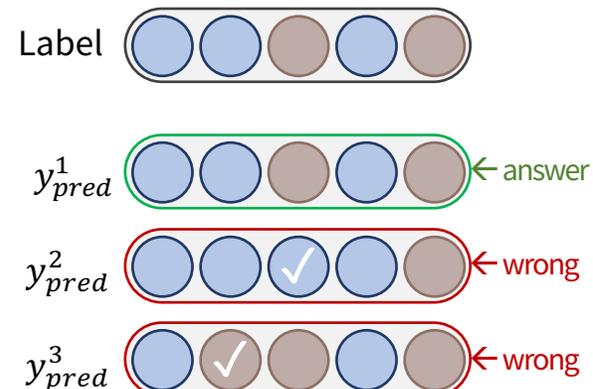
- Notably, **MixX** did not involve **implicit** concatenation.   
(‘Naïve’ refers to the original construction method of **MixX**, meaning concatenation using only **and**, **and then**, and **and also**.)
- Particularly, **manual concatenation** often resulted in shorter utterance lengths.
- Conversely, **generative concatenation** uniquely led to the use of pronouns.

## 4-2) SOTA Models with BlendX

### • Evaluate Baseline

Model	Option		Dataset (Metric: accuracy)			
	Training	Test	SNIPS	ATIS	Banking77	CLINC150
TFMN	MixX	MixX	95.68* $\pm$ 0.57	77.98* $\pm$ 0.57	76.61 $\pm$ 1.17	85.88 $\pm$ 1.03
	MixX	BlendX	52.51 $\pm$ 1.86	42.51 $\pm$ 1.48	37.31 $\pm$ 0.81	42.45 $\pm$ 2.40
	BlendX	BlendX	94.93 $\pm$ 0.85	76.50 $\pm$ 0.83	63.99 $\pm$ 0.81	77.96 $\pm$ 0.82
SLIM	MixX	MixX	95.97* $\pm$ 0.23	77.10* $\pm$ 0.28	83.71 $\pm$ 0.88	88.67 $\pm$ 0.56
	MixX	BlendX	93.51 $\pm$ 0.18	72.80 $\pm$ 1.48	69.89 $\pm$ 0.46	73.39 $\pm$ 2.46
	BlendX	BlendX	95.73 $\pm$ 0.86	76.92 $\pm$ 0.84	75.30 $\pm$ 0.71	85.62 $\pm$ 0.51
gpt-3.5-turbo	-	MixX	81.68	40.30	30.90	49.22
	-	BlendX	76.18	38.84	22.67	37.55

### Accuracy in MID = exact match



For (un)supervised SOTA models, we consistently observe a **performance drop** on our **BlendX** datasets with explicit as well as implicit concatenations.

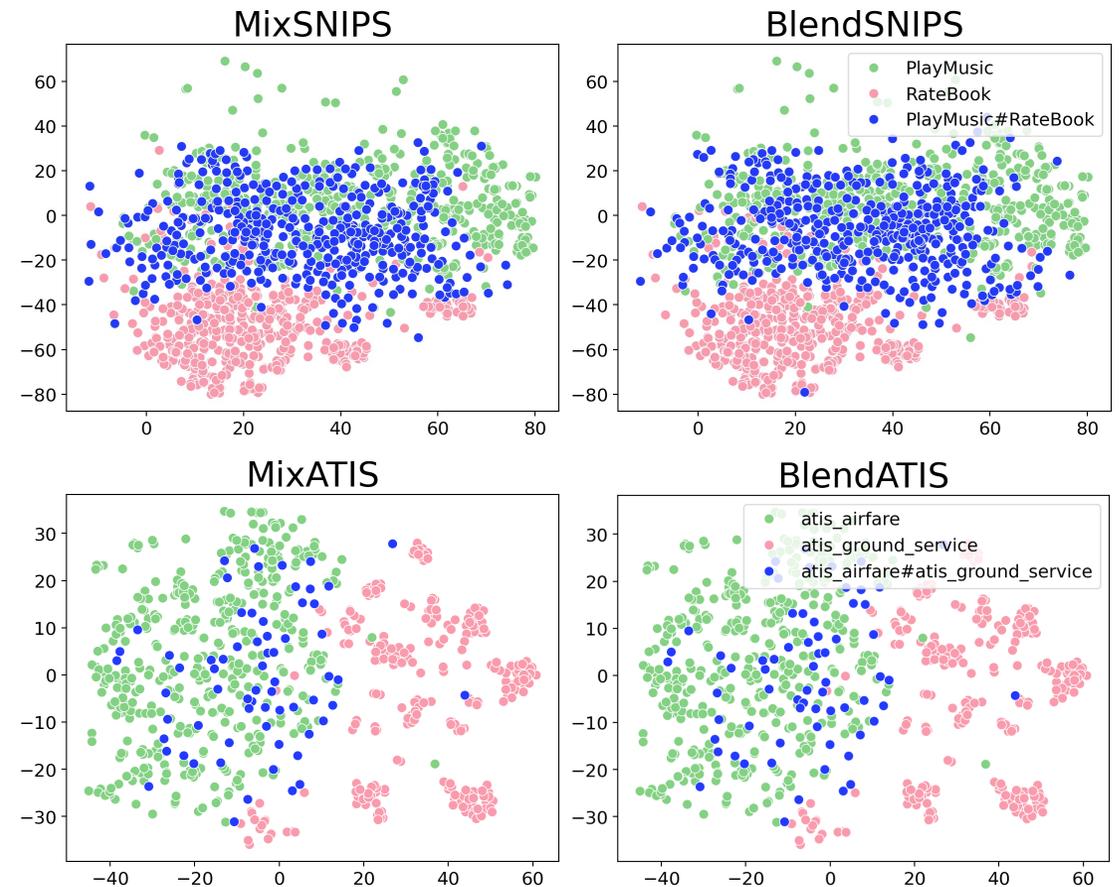
- 3-Baseline: implemented without slot-filling part
  - ✓ **TFMN** : predict # of intents  $k$ , and then top- $k$  intents over the probability distribution
  - ✓ **SLIM** : threshold-based classification model using sigmoid function
  - ✓ **ChatGPT** : OpenAI's generative model (**gpt-3.5-turbo-0613**)

## 4-3) Visualization of MixX and BlendX

### • Visualization of MixX and BlendX in the $\mathbb{R}^2$ space

- datasets
  - MixSNIPS & BlendSNIPS (Top)
  - MixATIS & BlendATIS (Bottom)
- Techniques
  - Sentence embeddings via SBERT
  - 2-dim. Projection with t-SNE

- Both MixX and BlendX visualizations show **similar patterns**.
- Despite the additional complexity in BlendX, the embeddings map the semantic space without causing significant distortions, **capturing all intents effectively**.



# Conclusion

# Contribution

# Conclusion - Main Findings

## • Identified limitations in existing multi-intent datasets

- **MixX**: Reliance on explicit concatenation through the 'and' connector.
- **BlendX**: Constructing a more complex and realistic multi-intent dataset
  - Proposed 3 novel concatenation approaches : Naïve, Manual, Generative
  - Beyond random sentence selection, applied a similarity-based strategy in the **generative** concatenation approach.
  - Designed 3 statistical metrics for comparing and validating **BlendX** against the existing **MixX**: W, C, P
  - Upcoming dataset release : Extensions of **MixX** (CLINC150/Banking77) and new publication of the **BlendX** dataset.

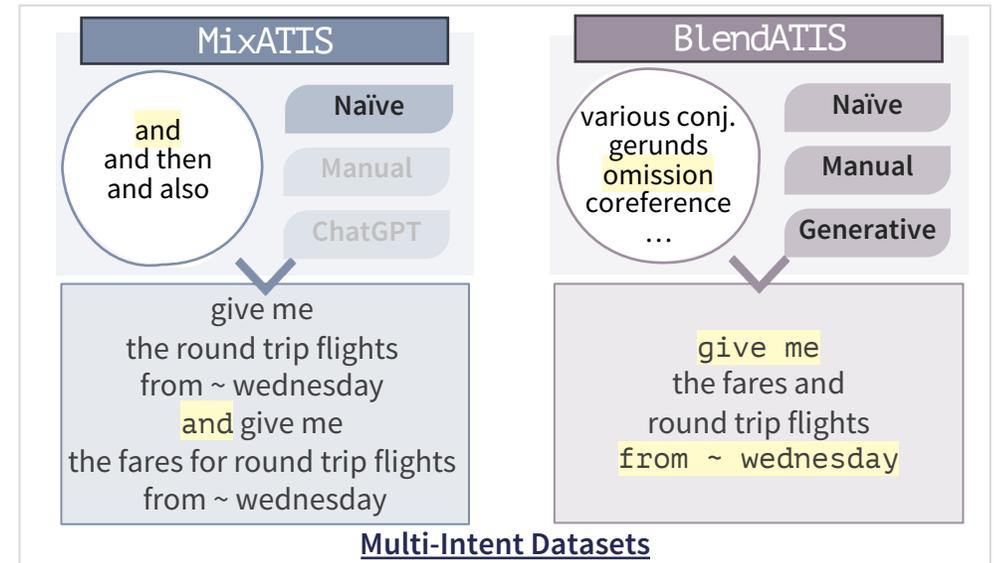
### #1 Selection

#### Single-Intent Datasets

Banking77 ATIS ✓  
CLINC150 SNIPS

give me the round trip flights  
from ~ wednesday **atis\_flight**  
give me the fares for round trip flights  
from ~ wednesday **atis\_airfare**

### #2 Concatenation



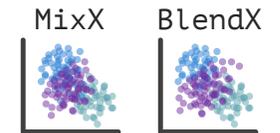
### #3 Evaluation

	Mix	Blend
$W(utt)$	0	1
$C(utt)$	0	0
$P(utt)$	0	0

**custom metric**



**baseline evaluation**



**visualization**

# Thank You

**Yejin Yoon**

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