

BlendX : Complex Multi-intent Detection with Blended Patterns

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Yejin Yoon

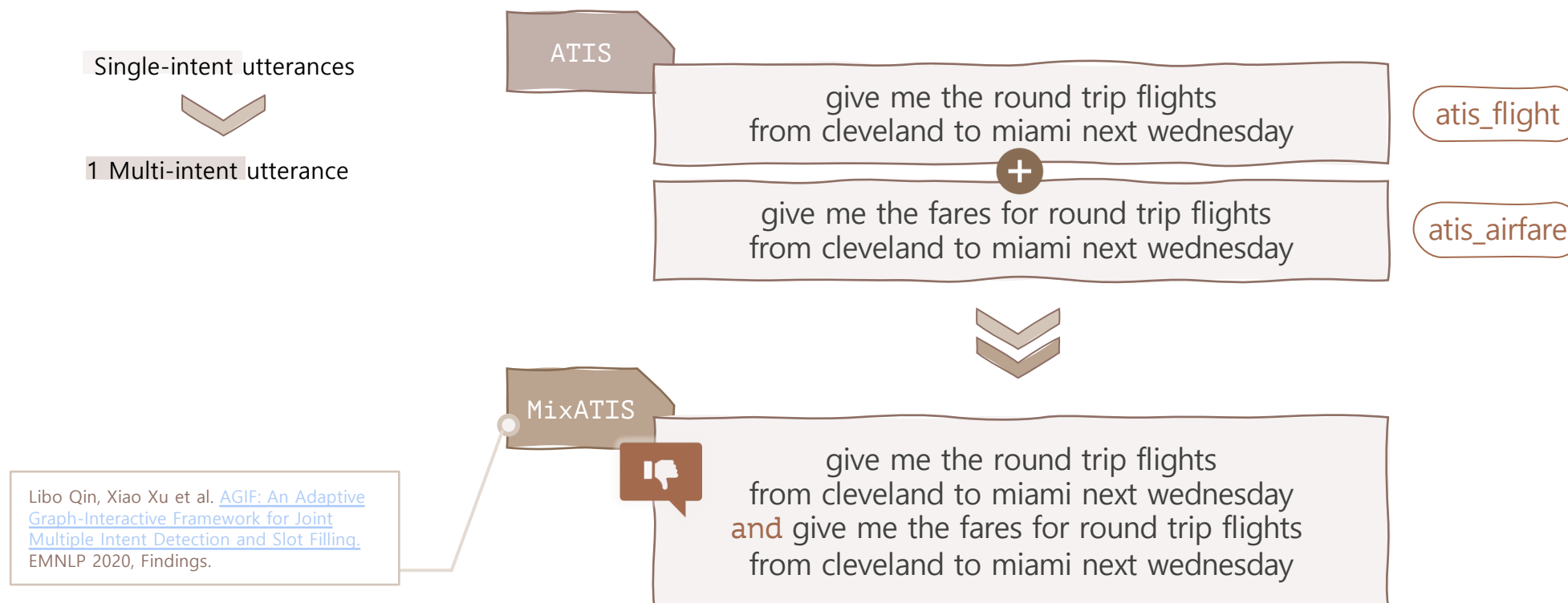
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Current Trends in MID

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)	88.3	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
TFMN (Bert-base)	88.0	79.8	50.2	96.4	97.7	84.7

- **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.

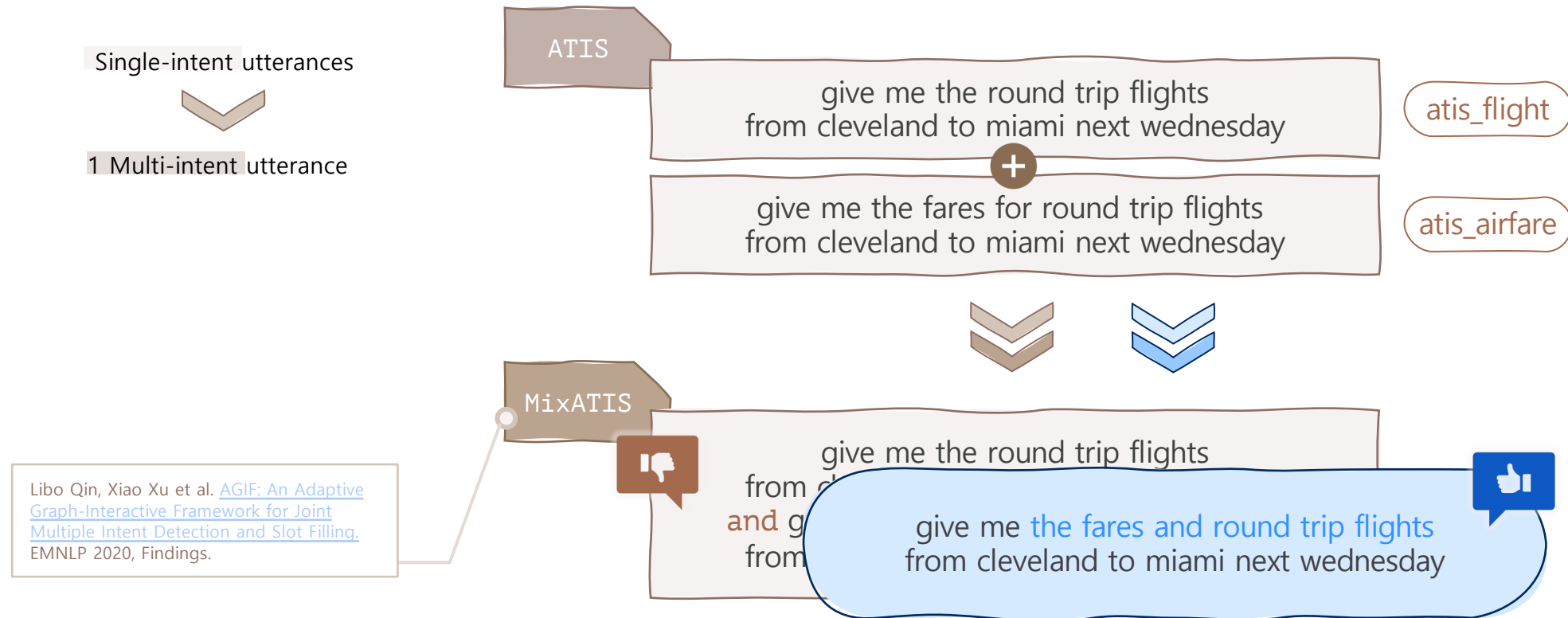
Benchmark Datasets Analysis (1/2)



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)



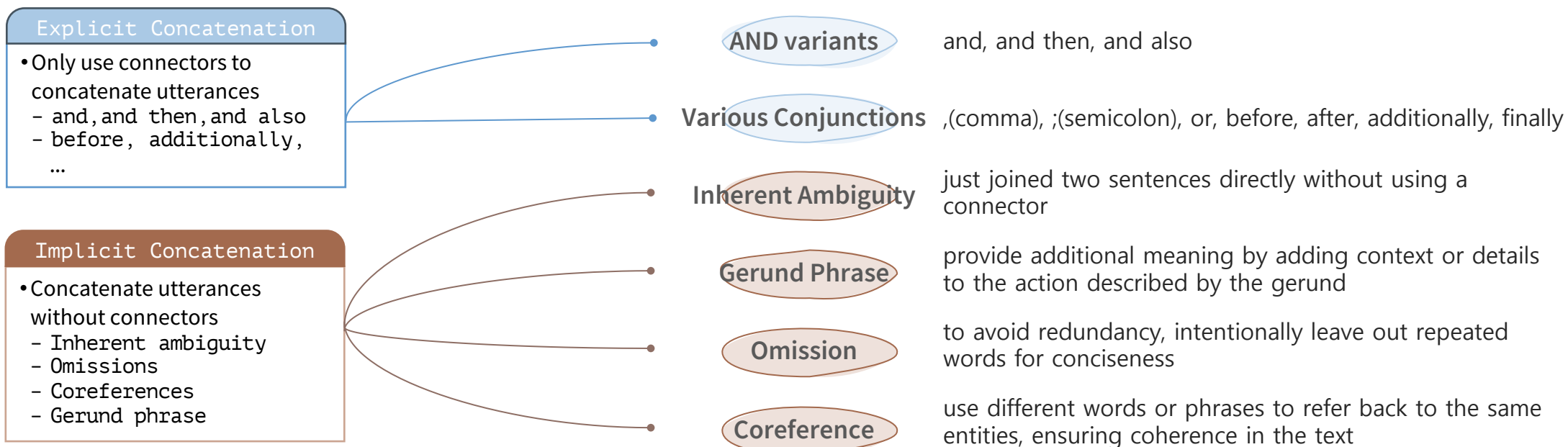
We are focused on constructing our own set that better mirrors natural language usage to provide more **challenging** and **realistic** resources for training and testing multi-intent detection models.

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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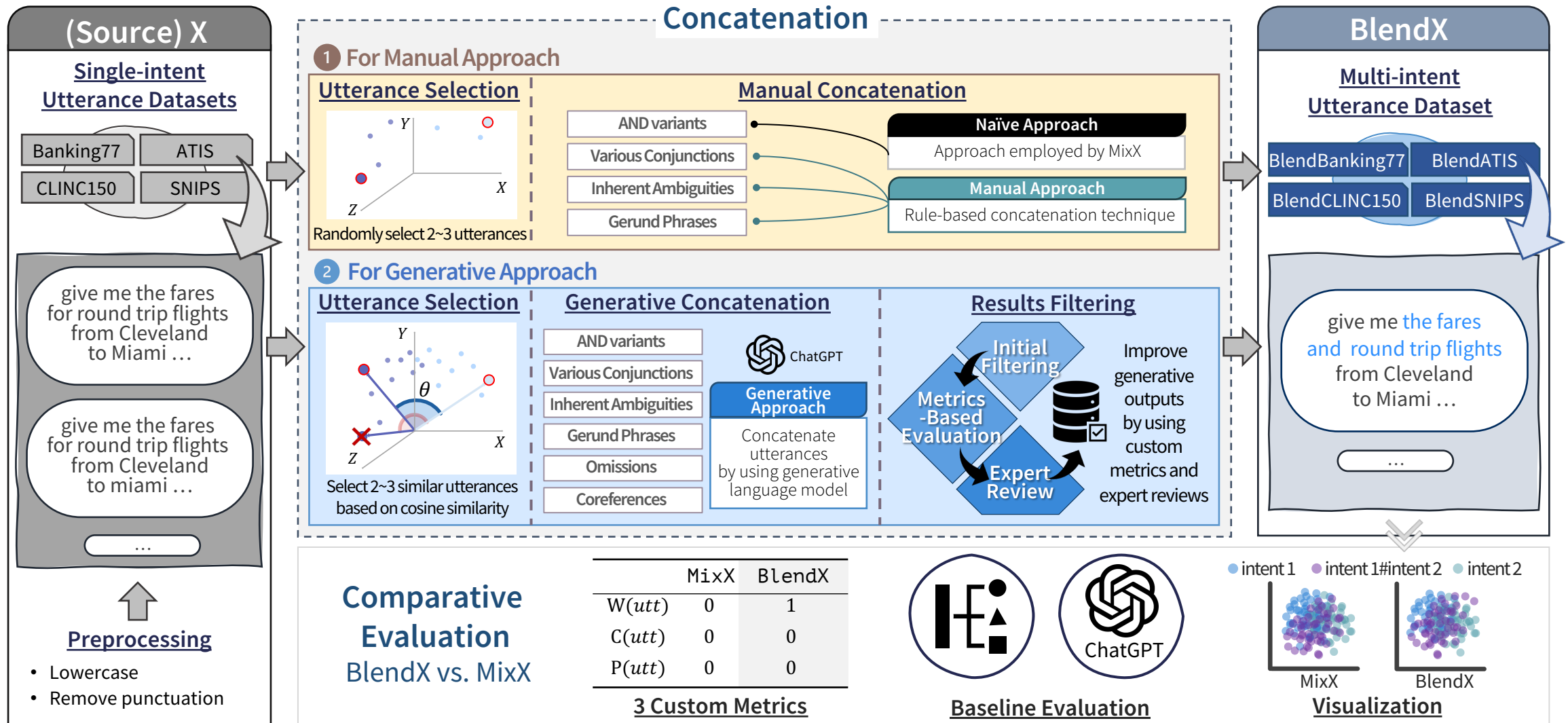
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Overview of 2 Methods to Utterance Concatenation



Overview of 2 Methods to Utterance Concatenation

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

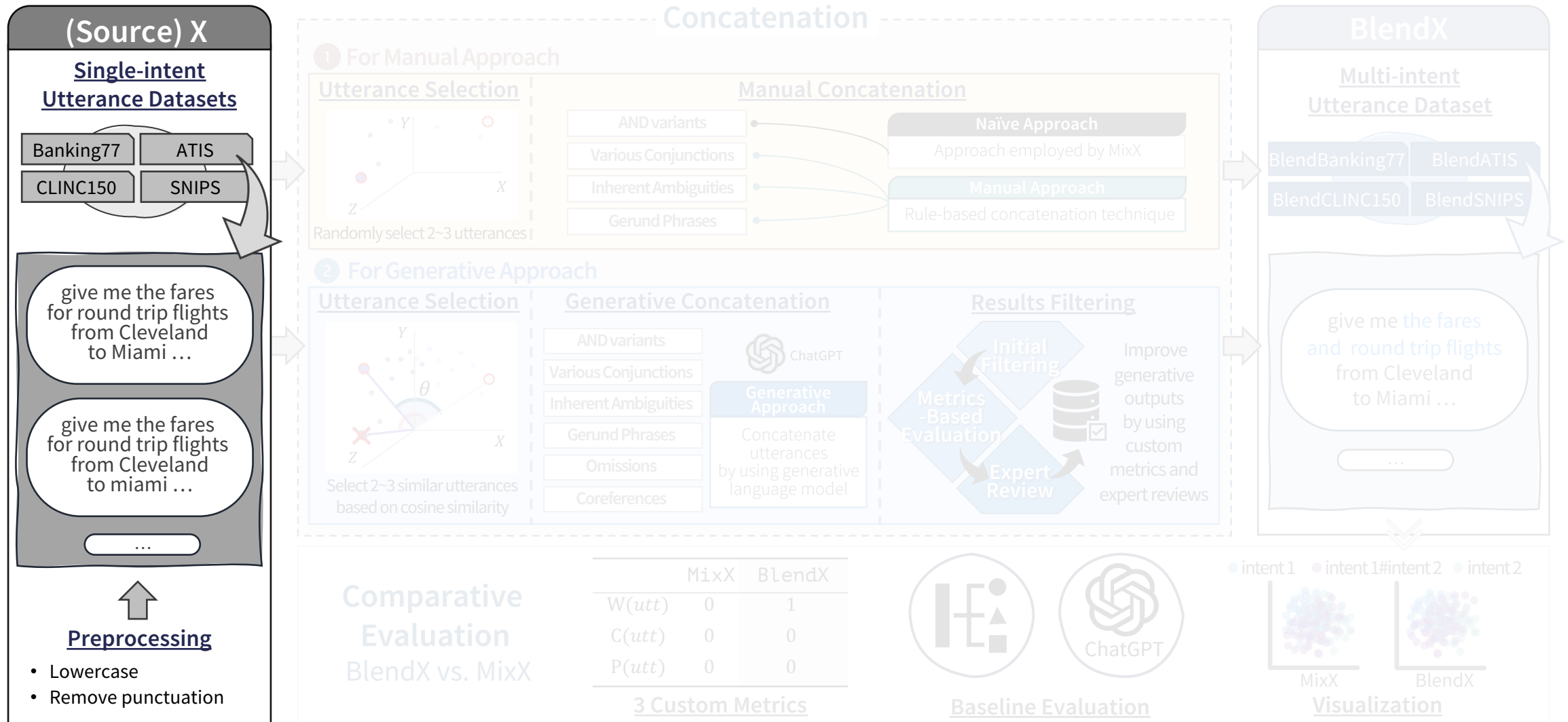
1 Manual Approach

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

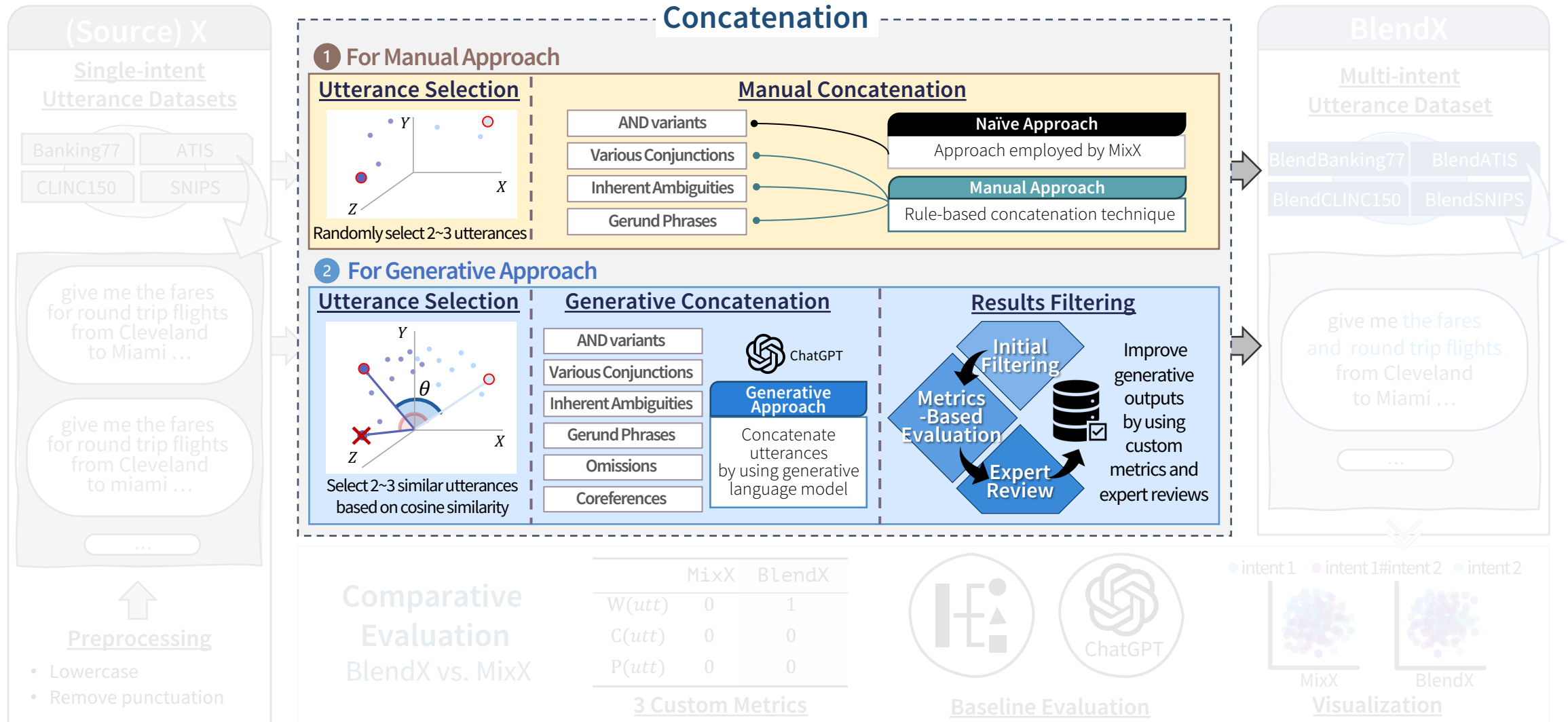
2 Generative Approach

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

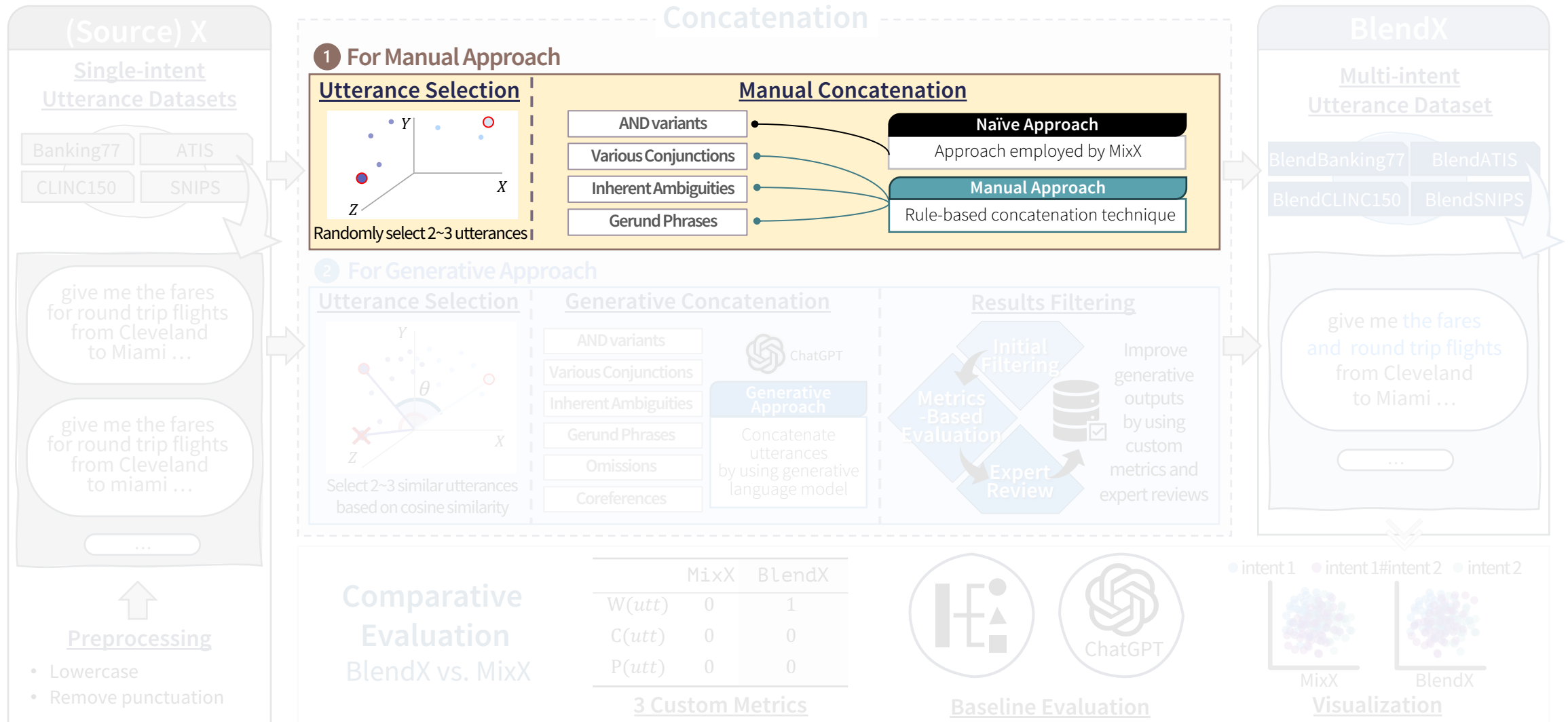
1) Preprocessing



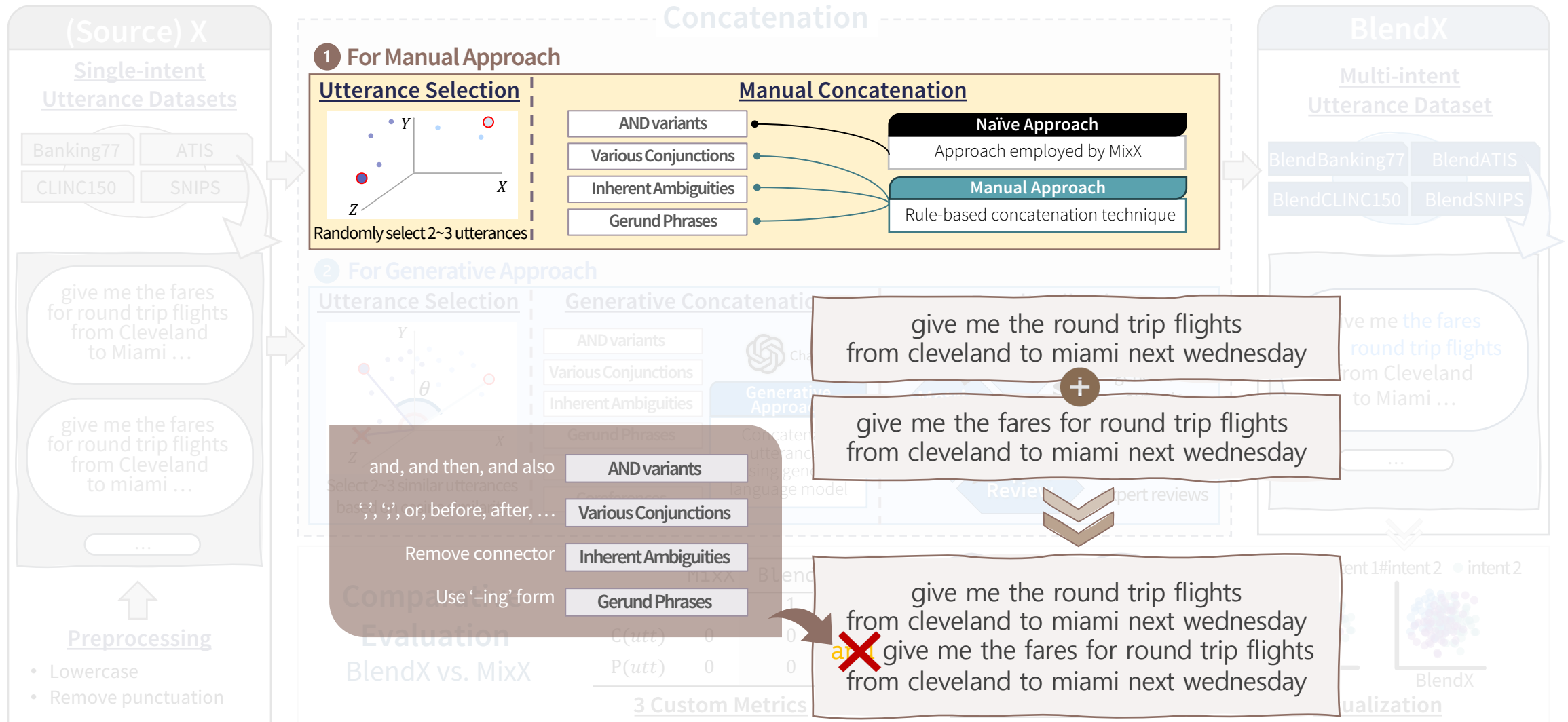
2) Concatenation



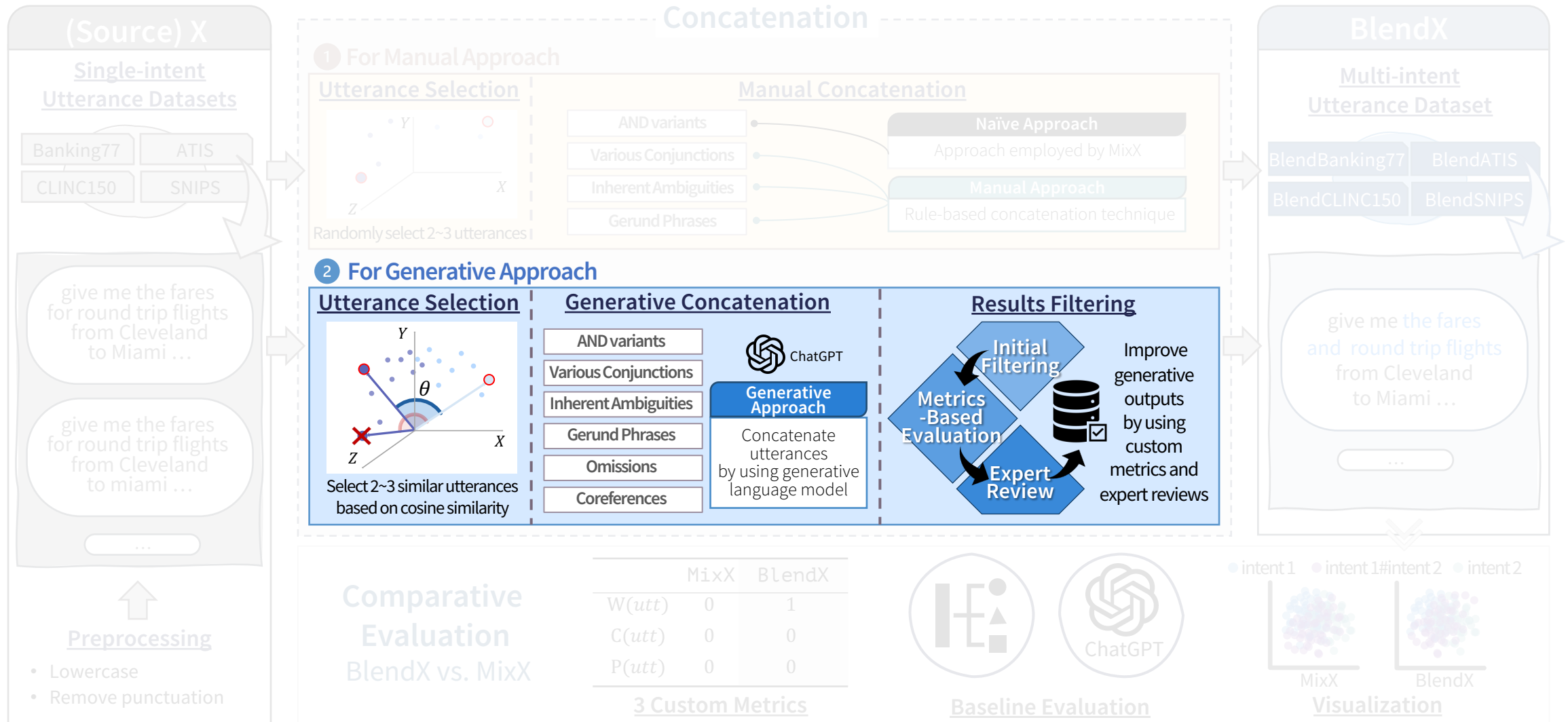
2-1) Manual Approach



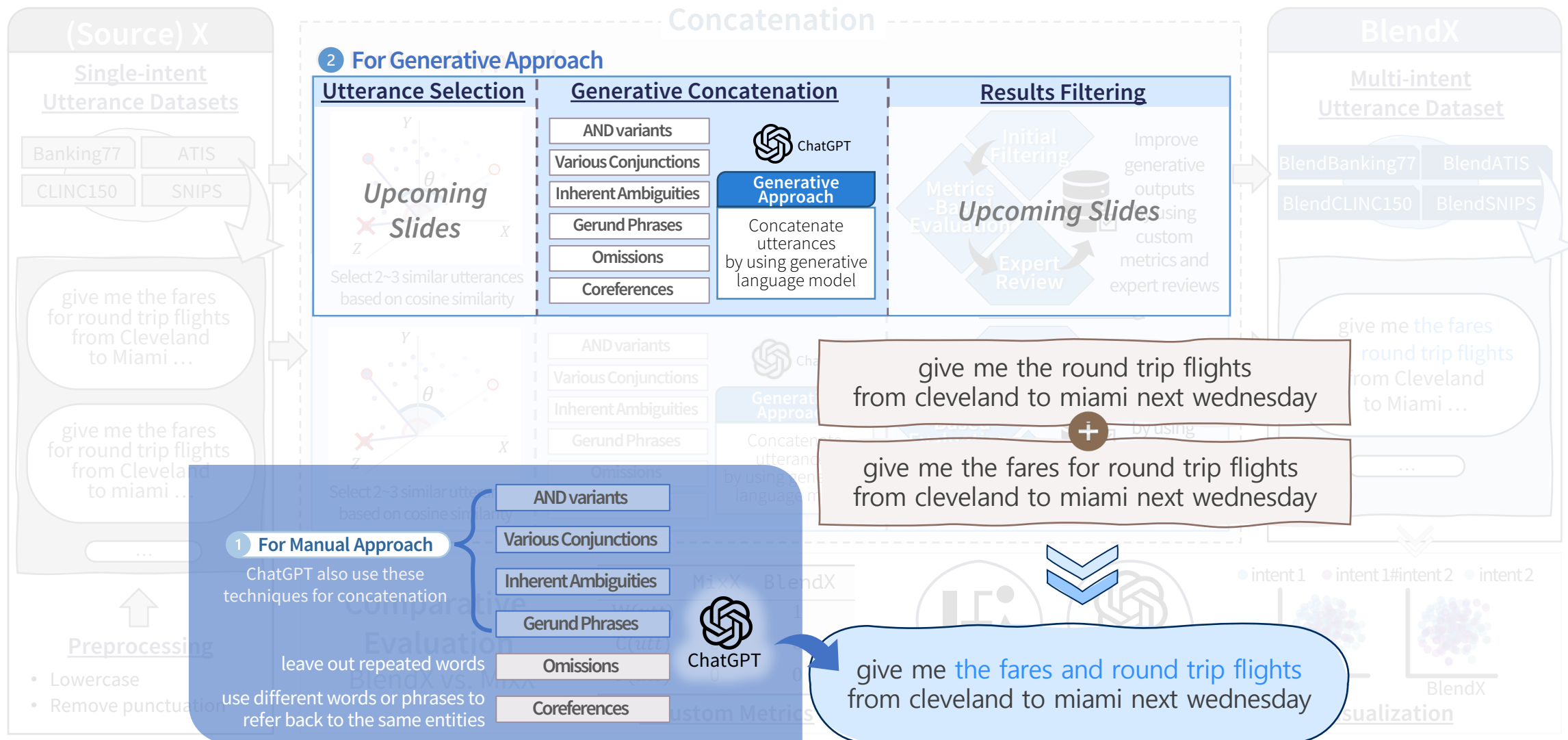
2-1) Manual Approach



2-2) Generative Approach



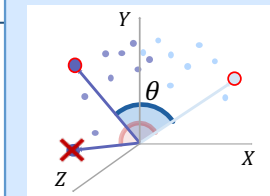
2-2) Generative Approach



Utterance Selection for 2-2) Generative Approach

2 For Generative Approach

Utterance Selection



Select 2~3 similar utterances based on cosine similarity

Generative Concatenation

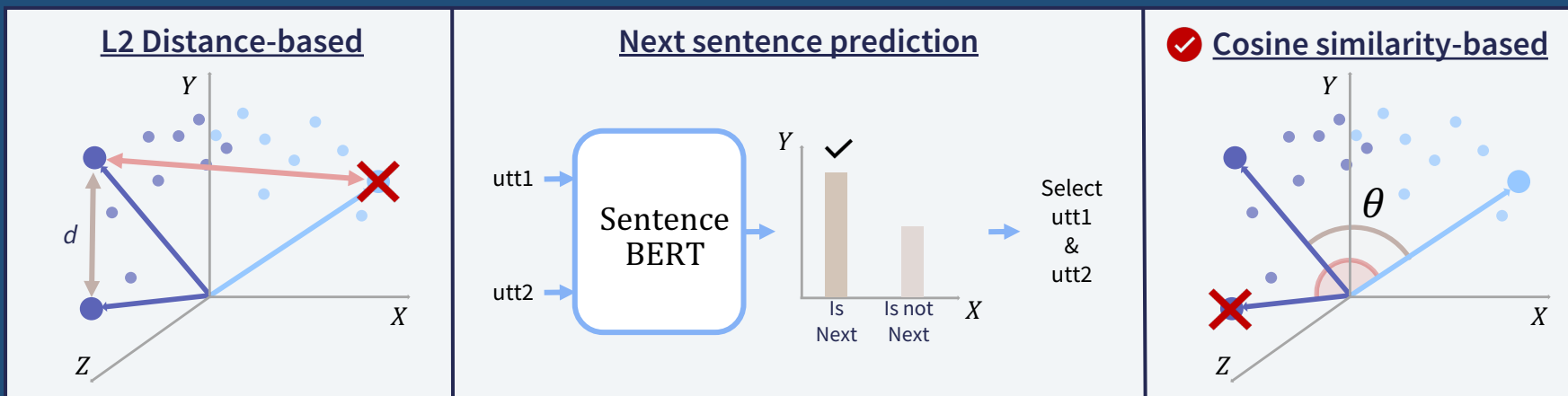
- AND variants
- Various Conjunctions
- Inherent Ambiguities
- Gerund Phrases
- Omissions
- Coreferences

• 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

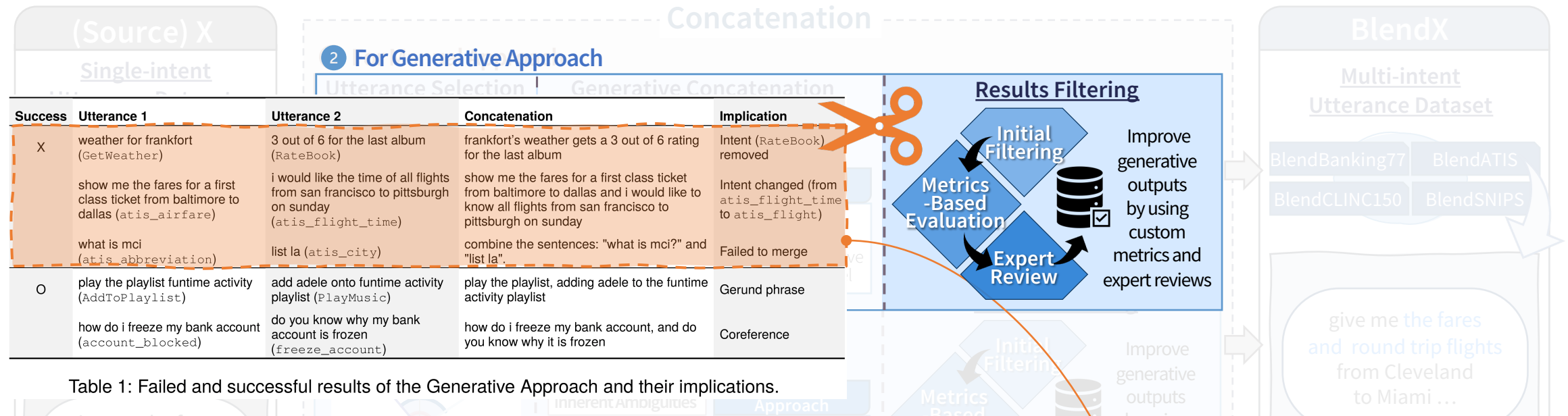
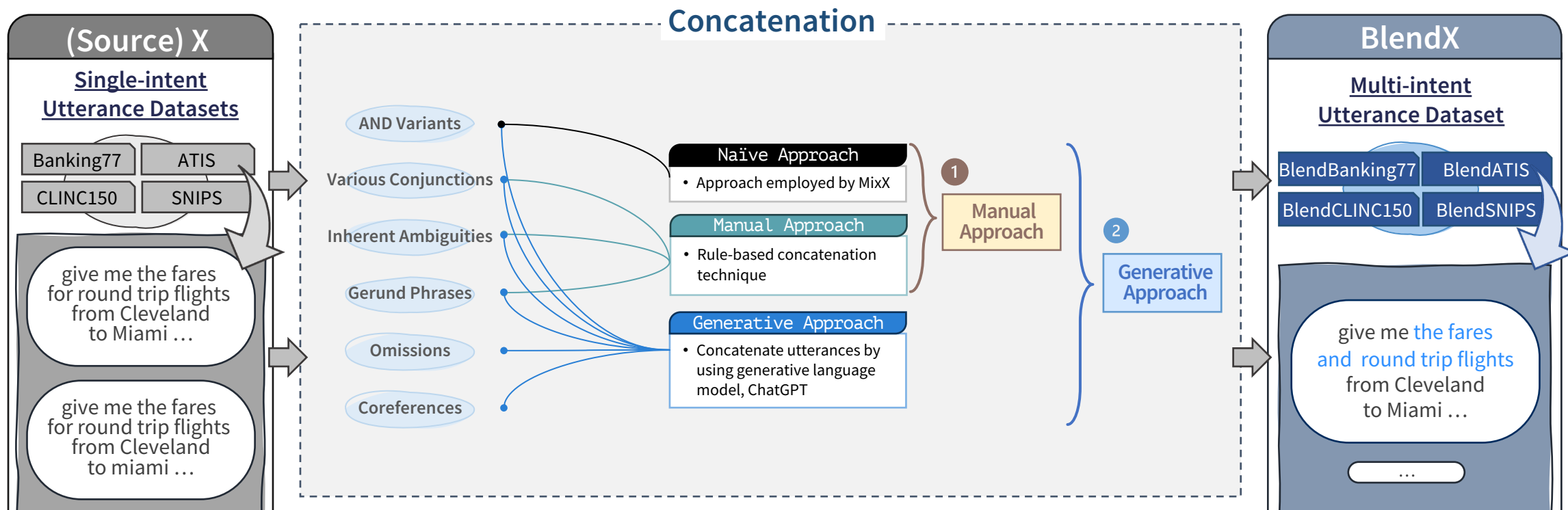


Table 1: Failed and successful results of the Generative Approach and their implications.

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

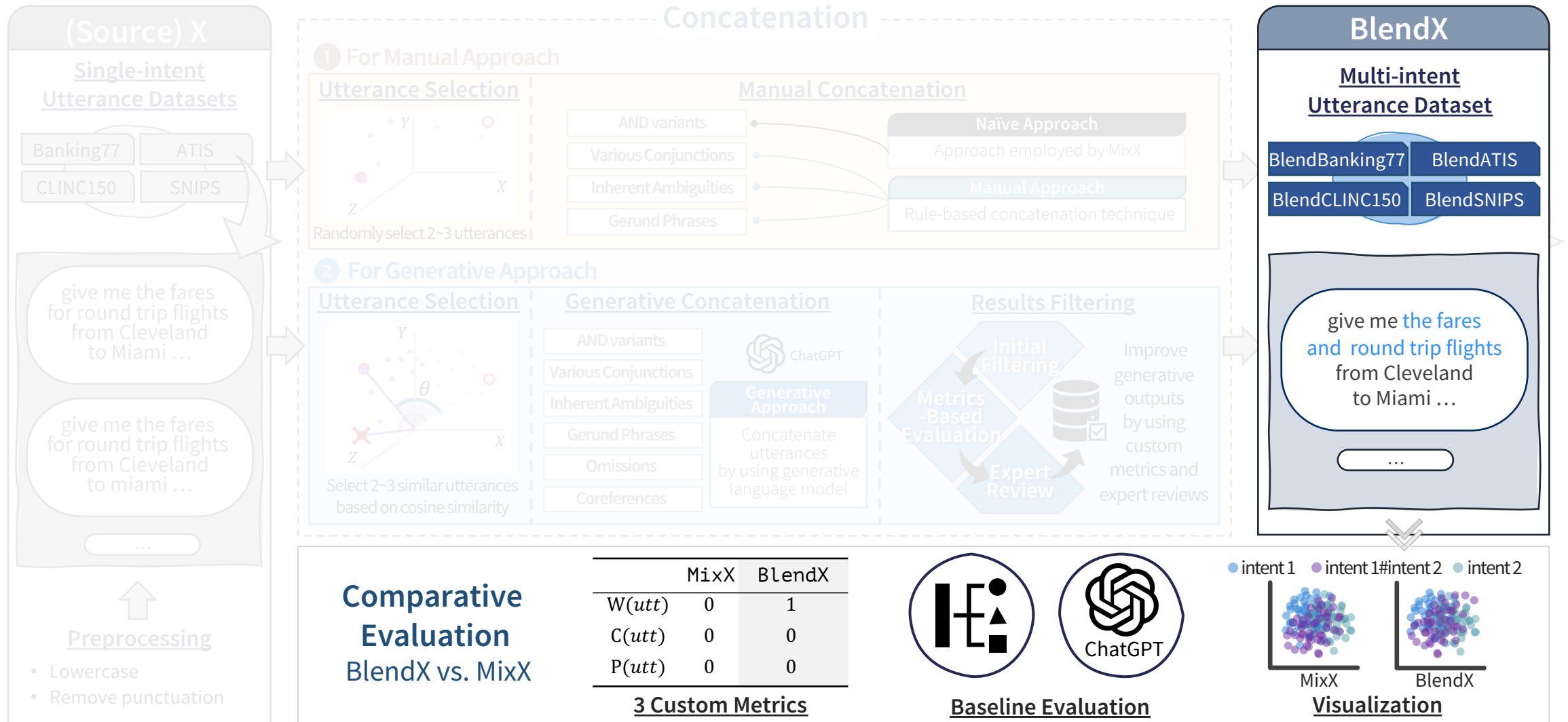


BlendX: Complex multi-intent detection with blended patterns

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

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	$W(utt, 2)$	$C(utt, 2)$	$P(utt, 2)$
Explicit Concatenation	play my 88 keys playlist and also add another song to my 88 keys playlist	0	0	0
Implicit Concatenation				
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
example #1	add another song to my 88 keys playlist playing it	play my 88 keys playlist	add another song to my 88 keys playlist		
Words	10	5	8	$10 - (5 + 8) = -3$	$W(\cdot, 2) = 1$
Conjunctions	0	0	0	$0 - (0 + 0) = 0$	$C(\cdot, 2) = 1$
Pronouns	1	0	0	$1 - (0 + 0) = 1$	$P(\cdot, 2) = 1$
example #2	i need to clear my to-do list and then repeat it	clear my to do list	repeat my to do list		
Words	11	5	5	$11 - (5 + 5) = 1$	$W(\cdot, 2) = 0$
Conjunctions	1	0	0	$1 - (0 + 0) = 1$	$C(\cdot, 2) = 0$
Pronouns	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.

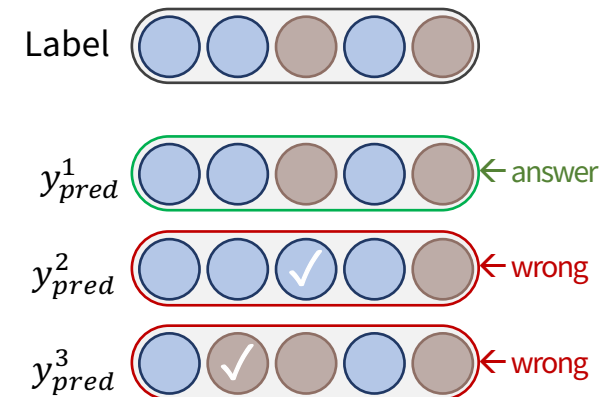
- Notably, **MixX** did not involve **implicit** concatenation. □
(‘Naïve’ refers to the original construction method of **MixX**, meaning concatenation using only **and**, **and then**, and **and also**.)
- Particularly, **manual concatenation** often resulted in shorter utterance lengths.
- Conversely, **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* ± 0.57	77.98* ± 0.57	76.61 ± 1.17	85.88 ± 1.03
	MixX	BlendX	52.51 ± 1.86	42.51 ± 1.48	37.31 ± 0.81	42.45 ± 2.40
	BlendX	BlendX	94.93 ± 0.85	76.50 ± 0.83	63.99 ± 0.81	77.96 ± 0.82
SLIM	MixX	MixX	95.97* ± 0.23	77.10* ± 0.28	83.71 ± 0.88	88.67 ± 0.56
	MixX	BlendX	93.51 ± 0.18	72.80 ± 1.48	69.89 ± 0.46	73.39 ± 2.46
	BlendX	BlendX	95.73 ± 0.86	76.92 ± 0.84	75.30 ± 0.71	85.62 ± 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**)

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
CLINC150

ATIS ✓
SNIPS

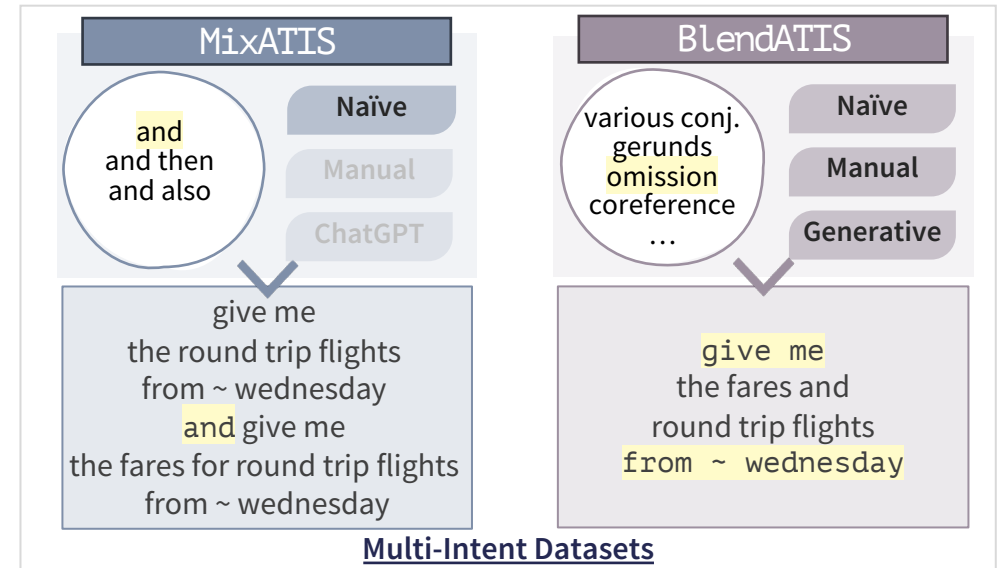
give me the round trip flights
from ~ wednesday

atis_flight

give me the fares for round trip flights
from ~ wednesday

atis_airfare

#2 Concatenation



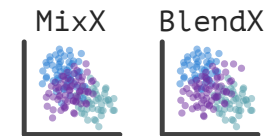
#3 Evaluation

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

custom metric



baseline evaluation



visualization

Thank You

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arXiv
2403.18277



GitHub
HYU-NLP/BlendX