

BlendX : Complex Multi-intent Detection with Blended Patterns

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Yejin Yoon

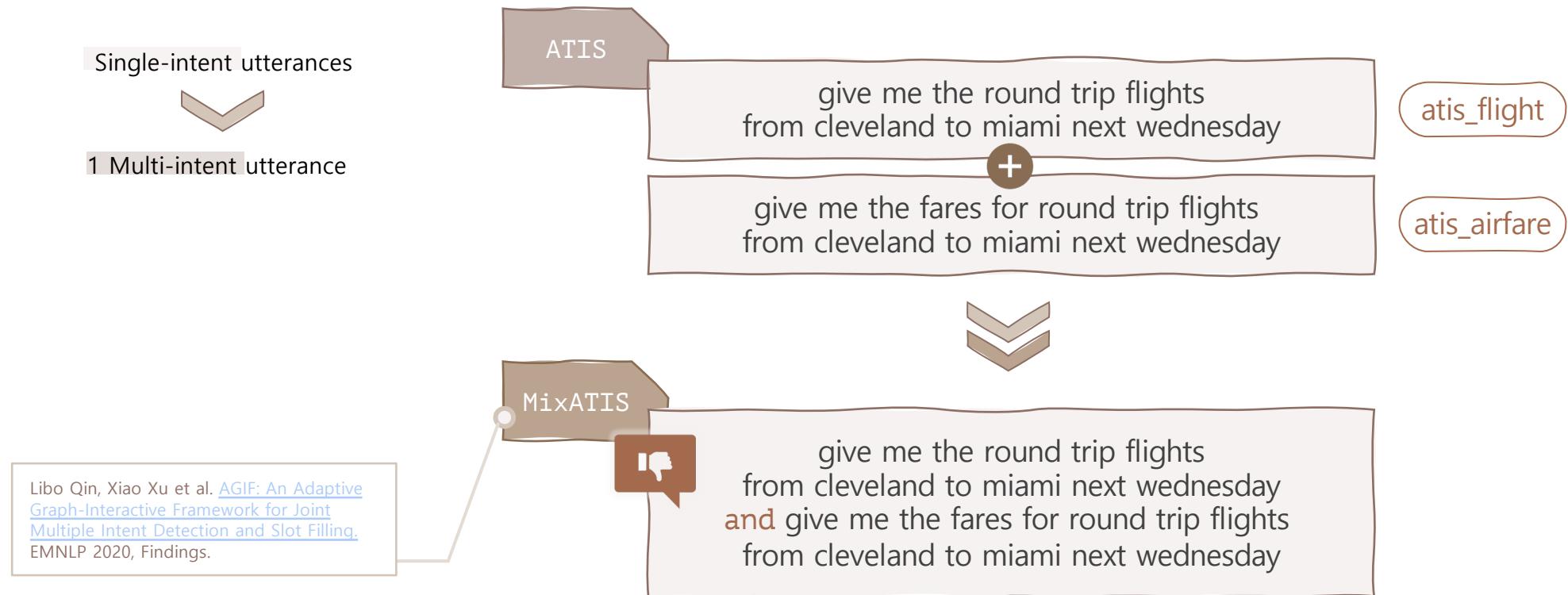
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Current Trends in MID

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.

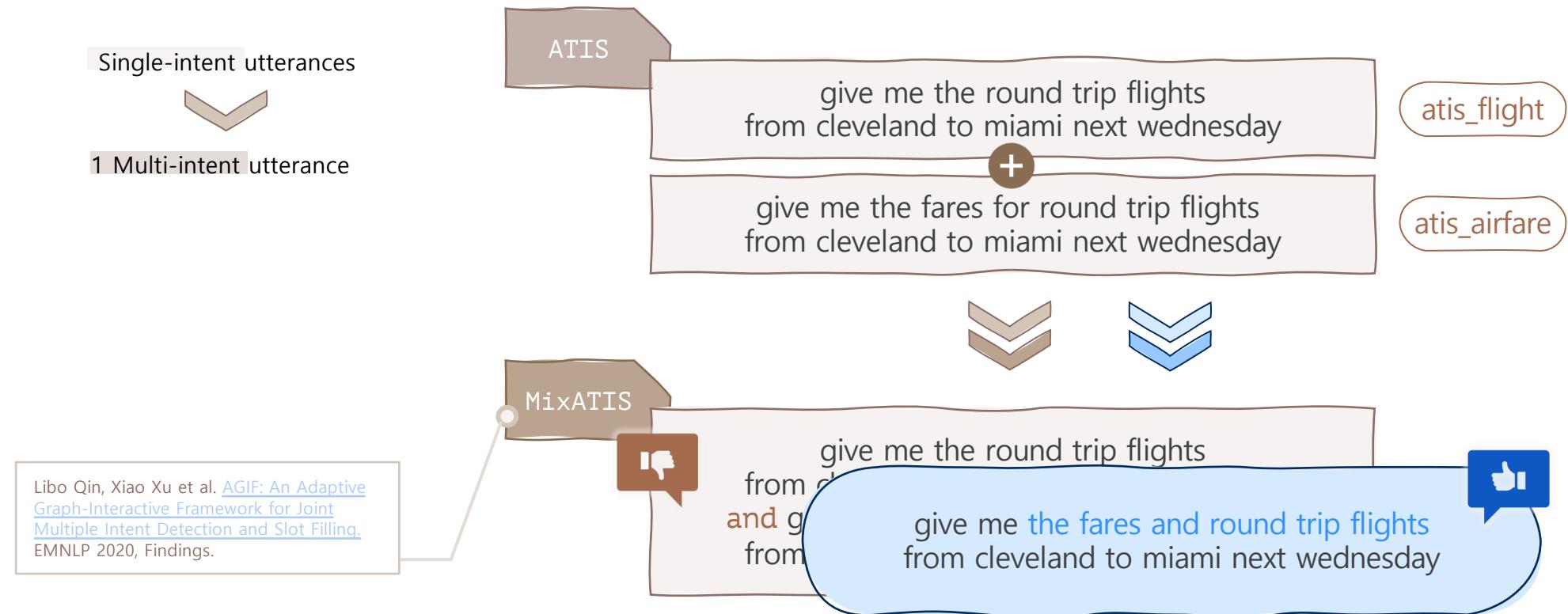
Benchmark Datasets Analysis (1/2)



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)



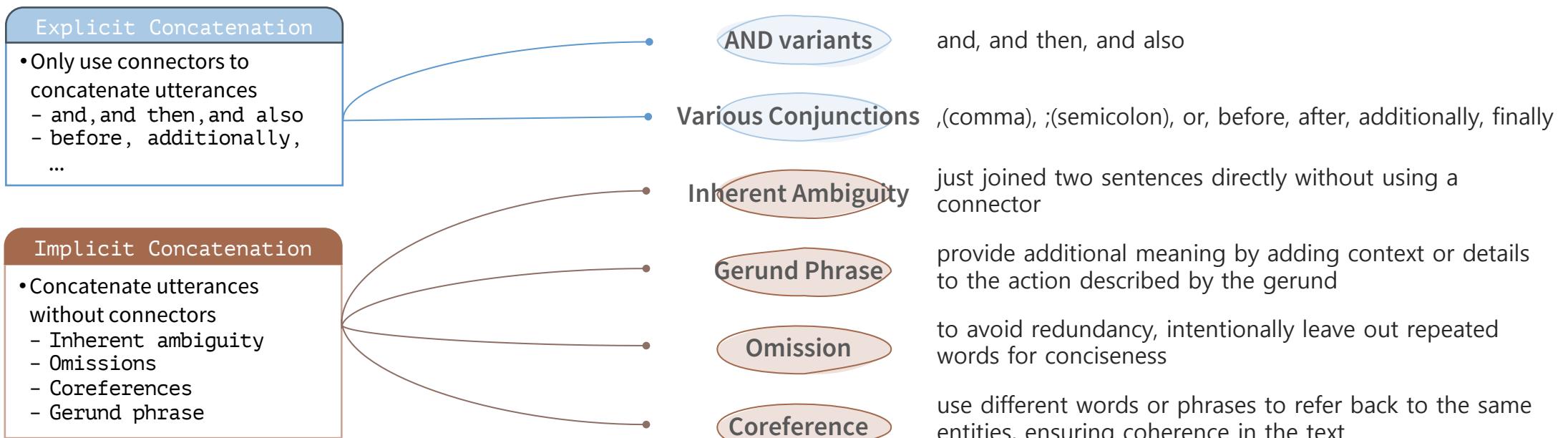
We are focused on constructing our own set that better mirrors natural language usage to provide more **challenging** and **realistic** resources for training and testing multi-intent detection models.

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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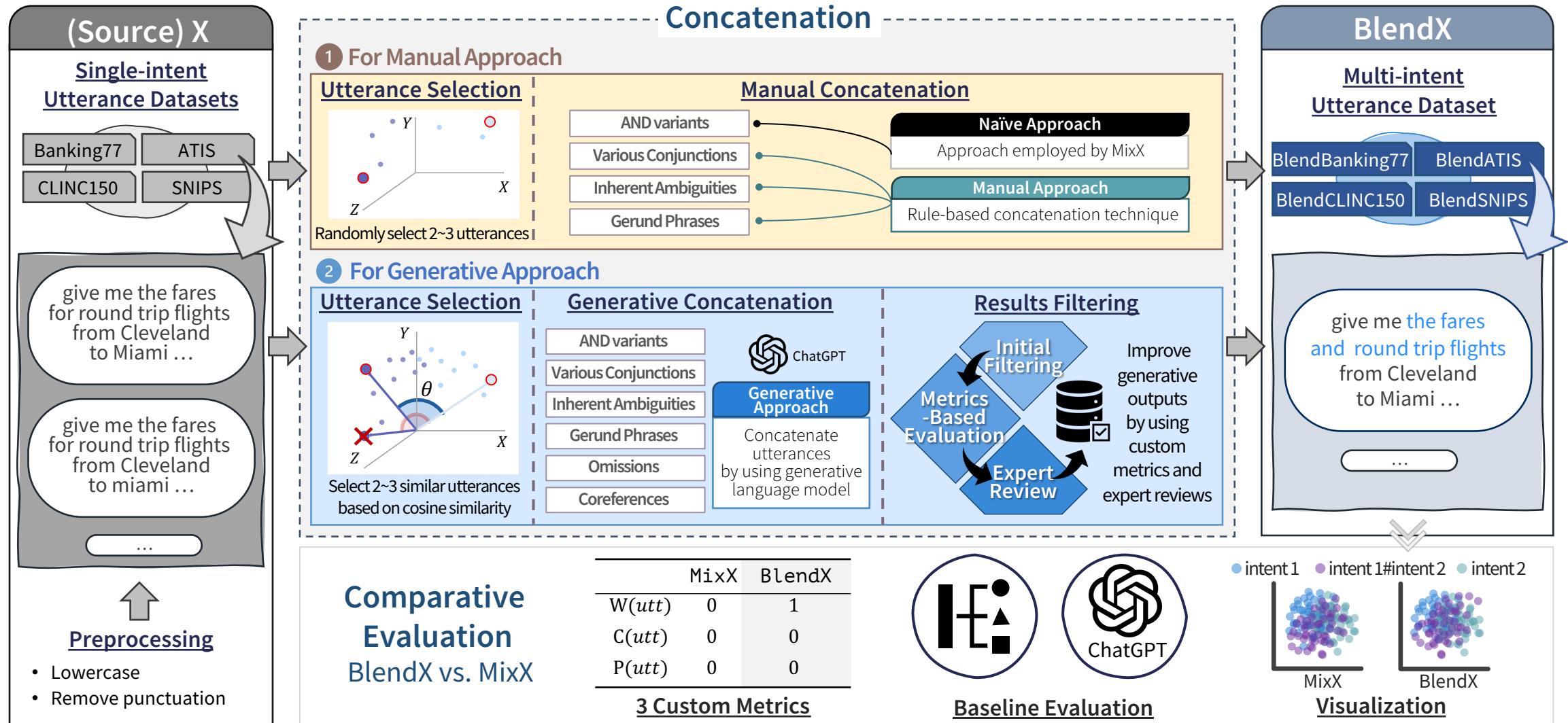
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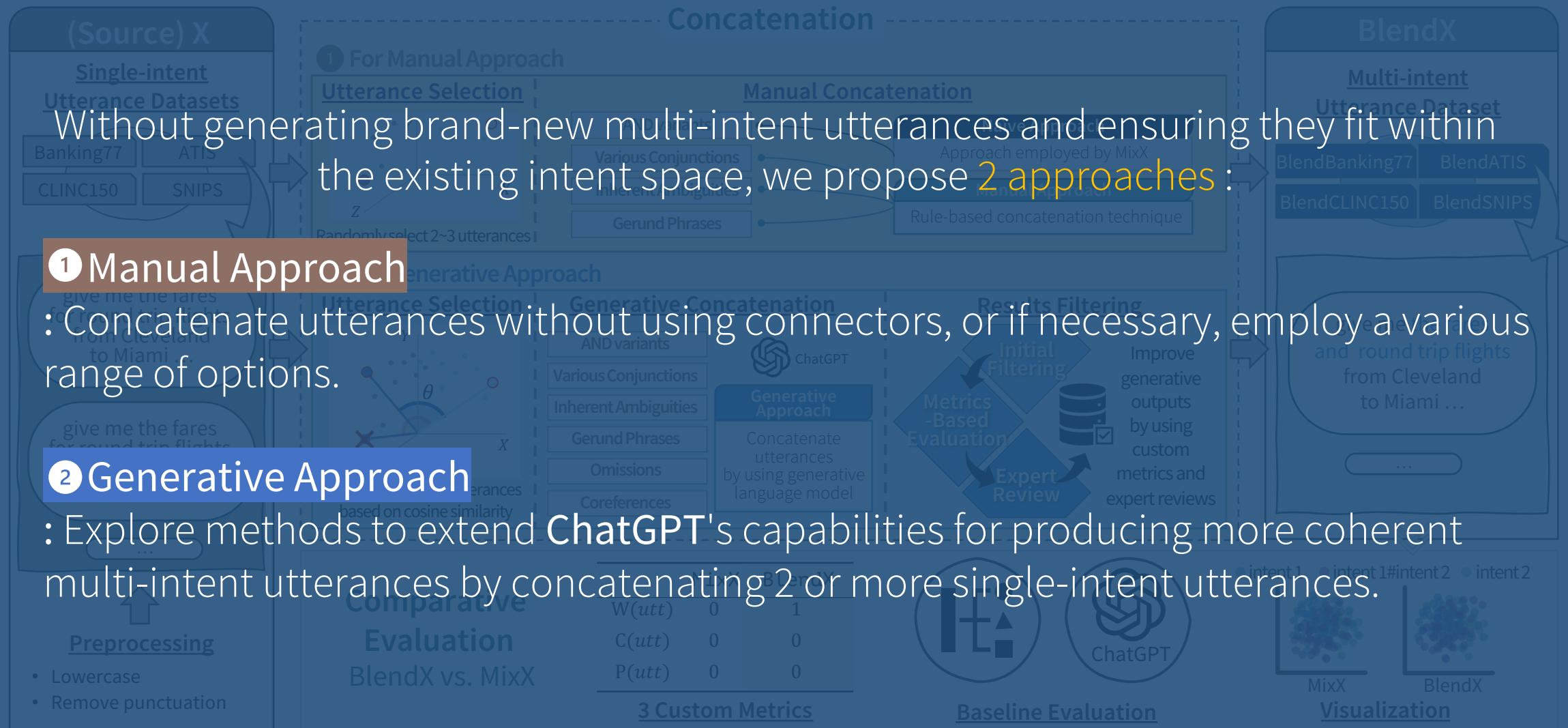
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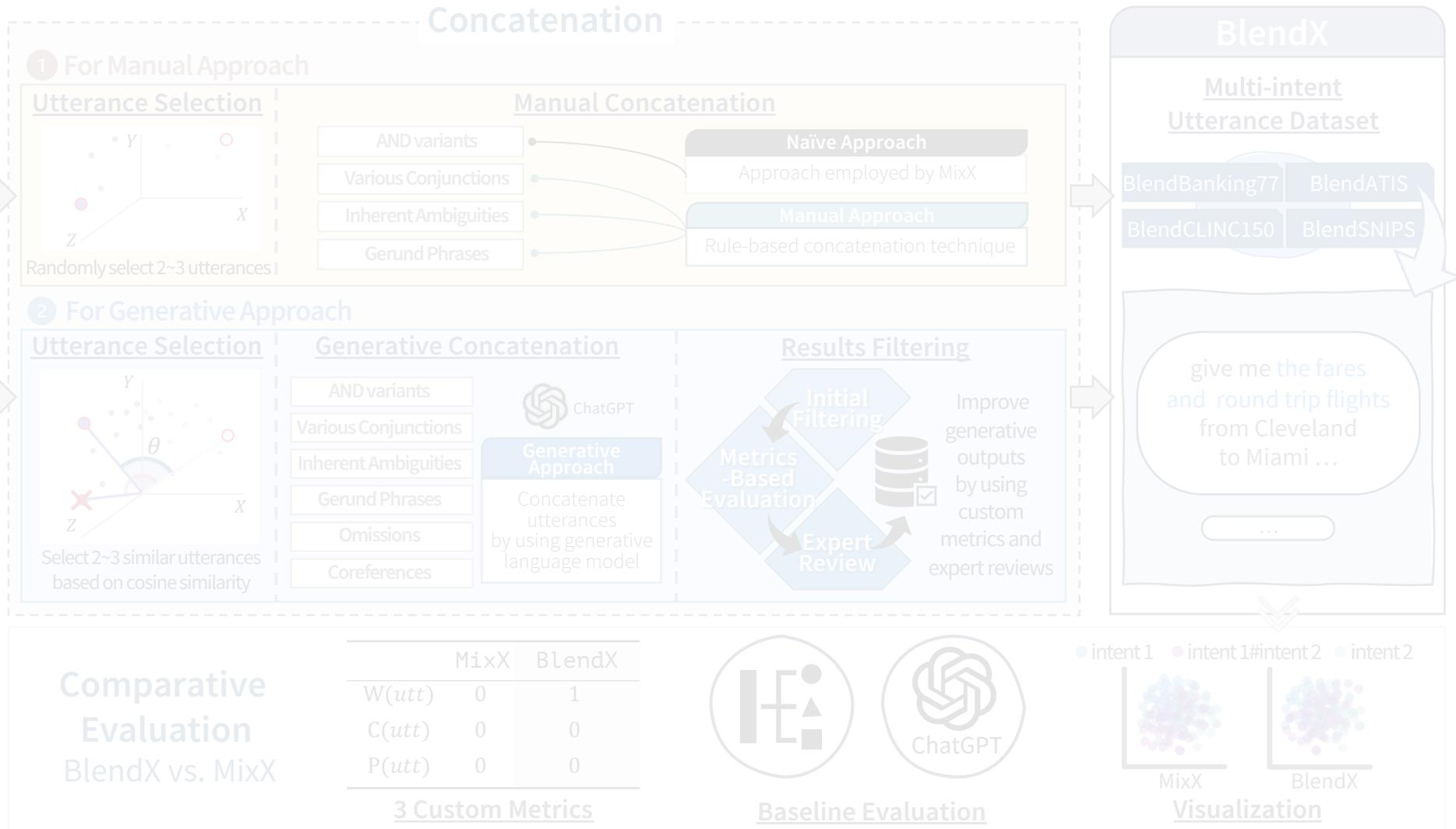
Overview of 2 Methods to Utterance Concatenation



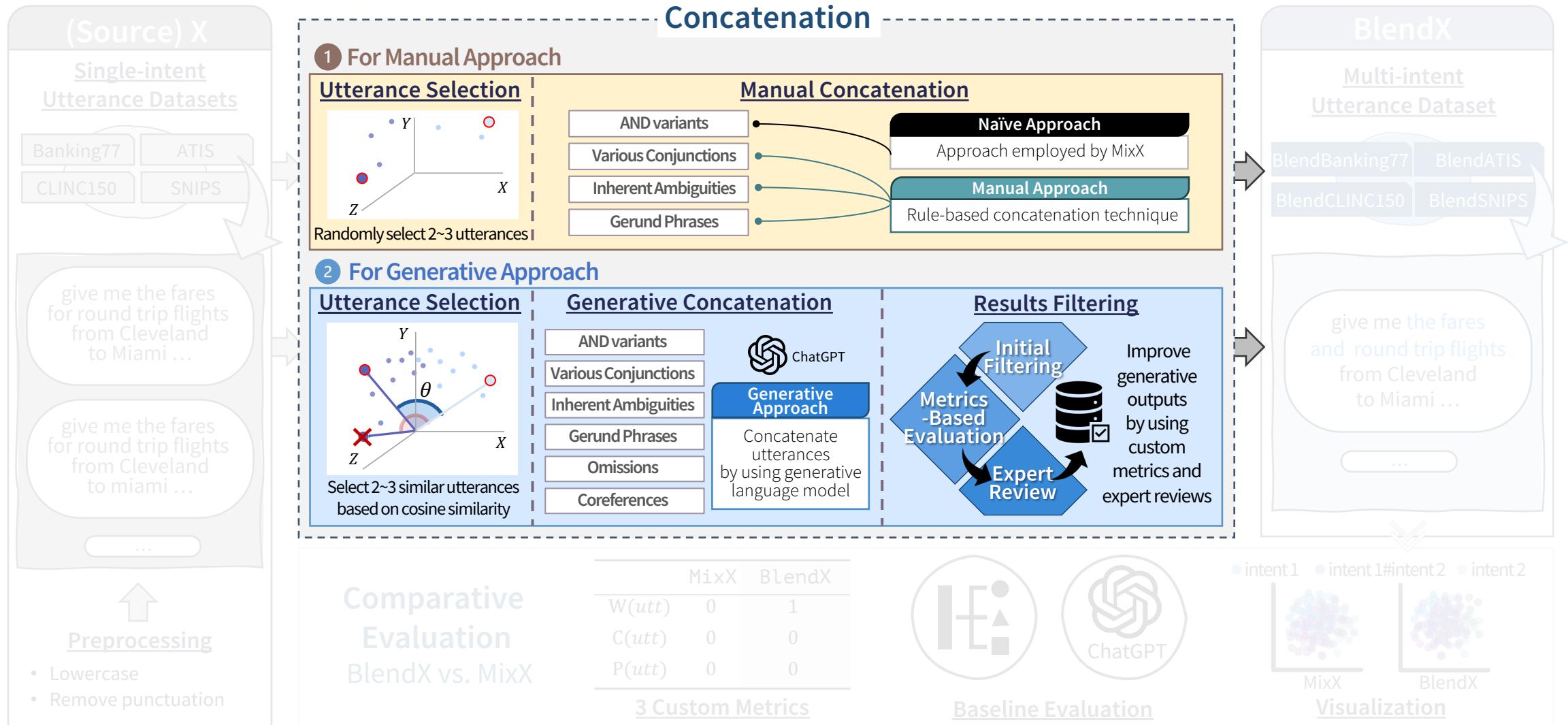
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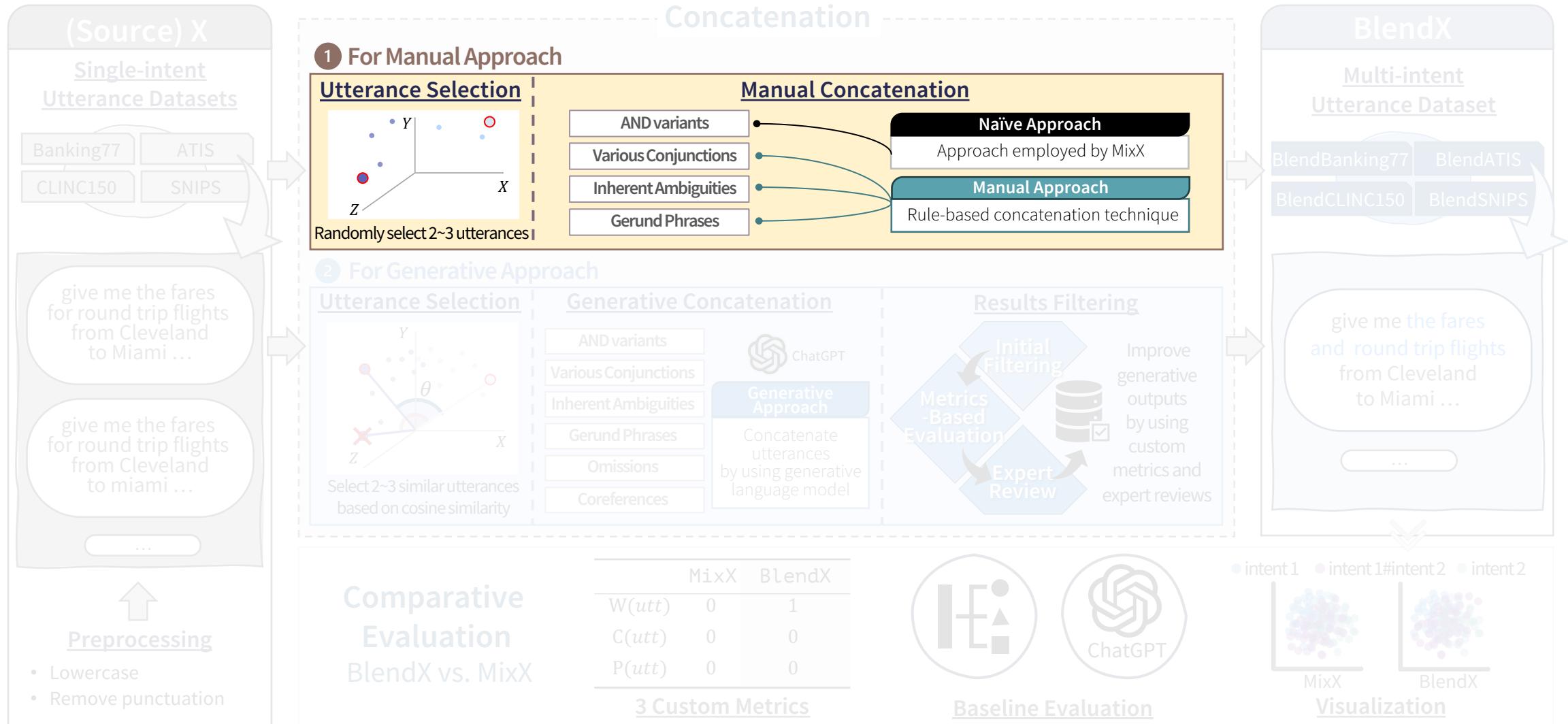
1) Preprocessing



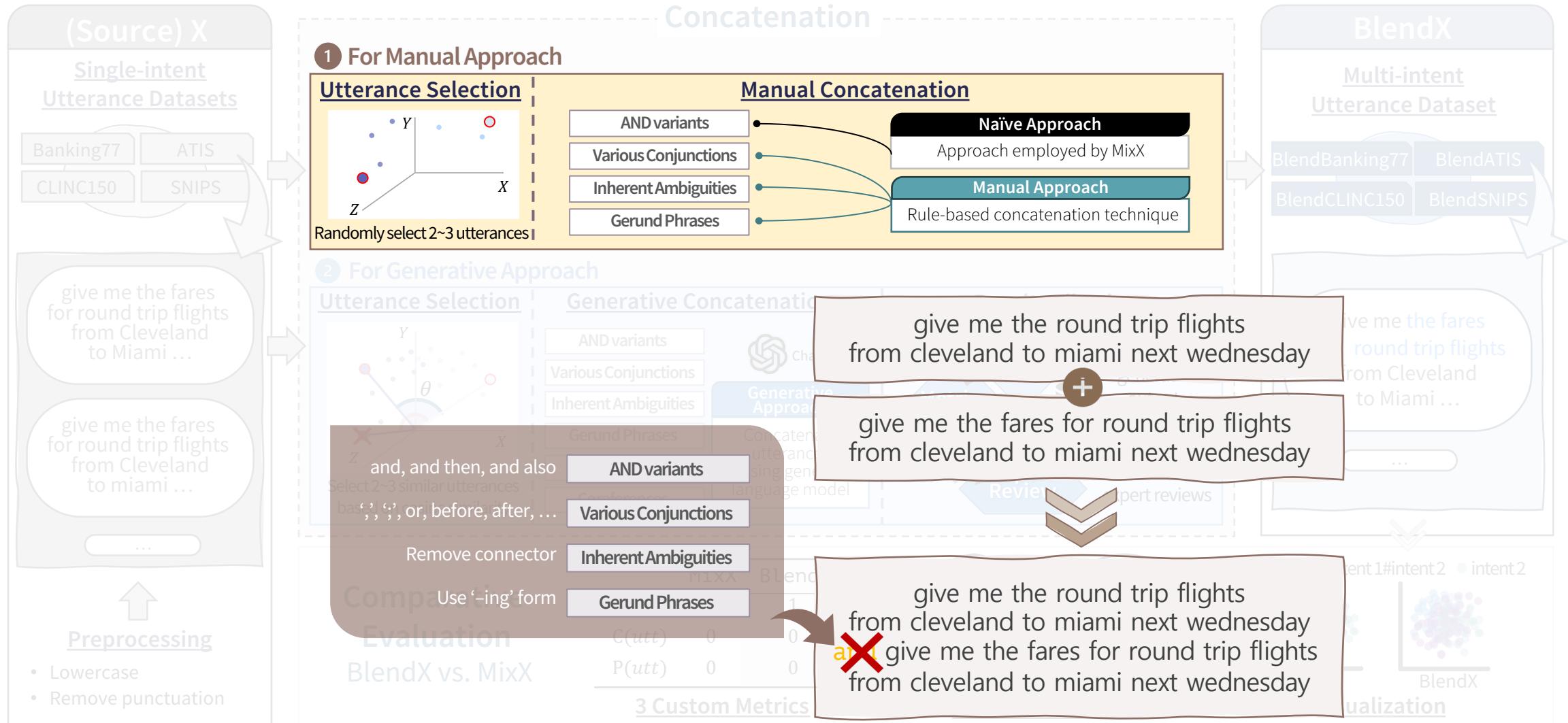
2) Concatenation



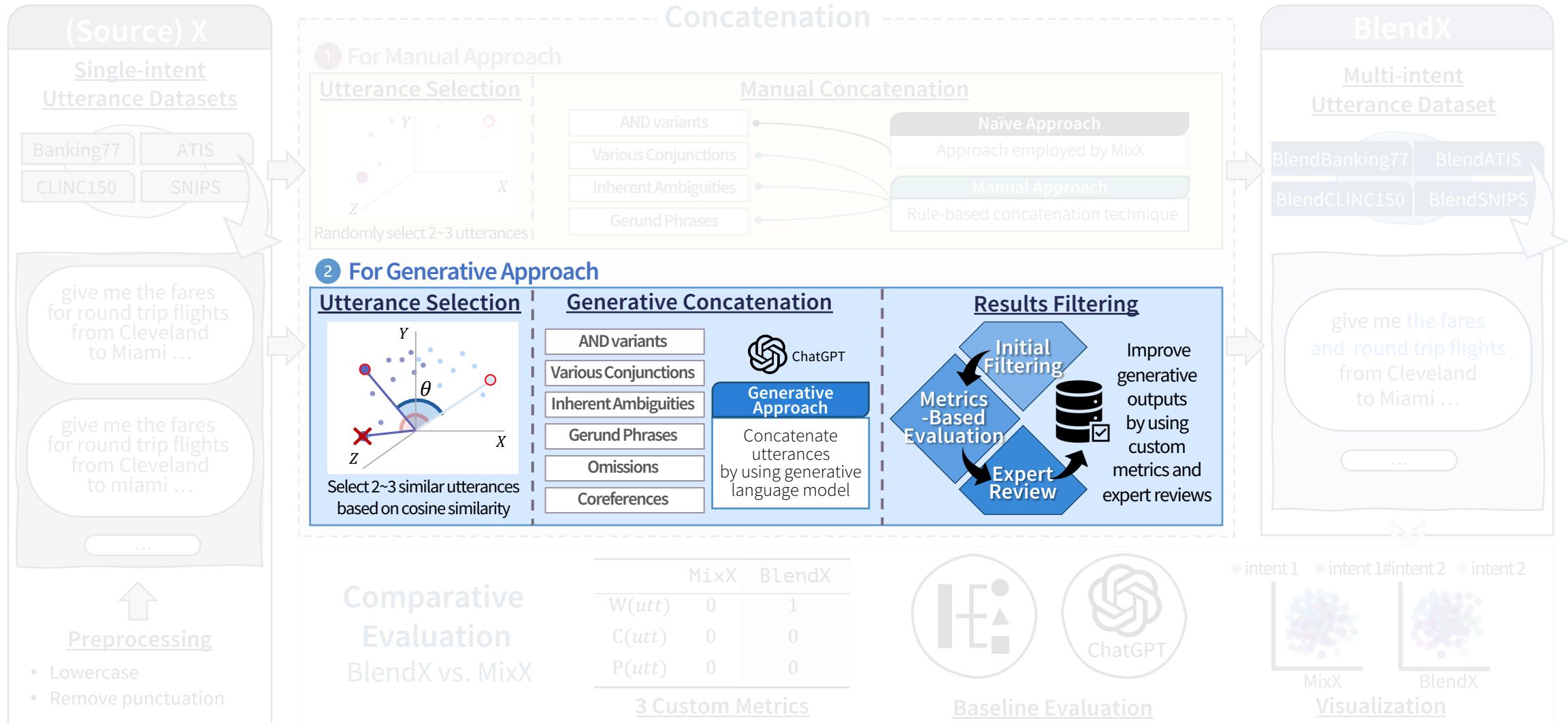
2-1) Manual Approach



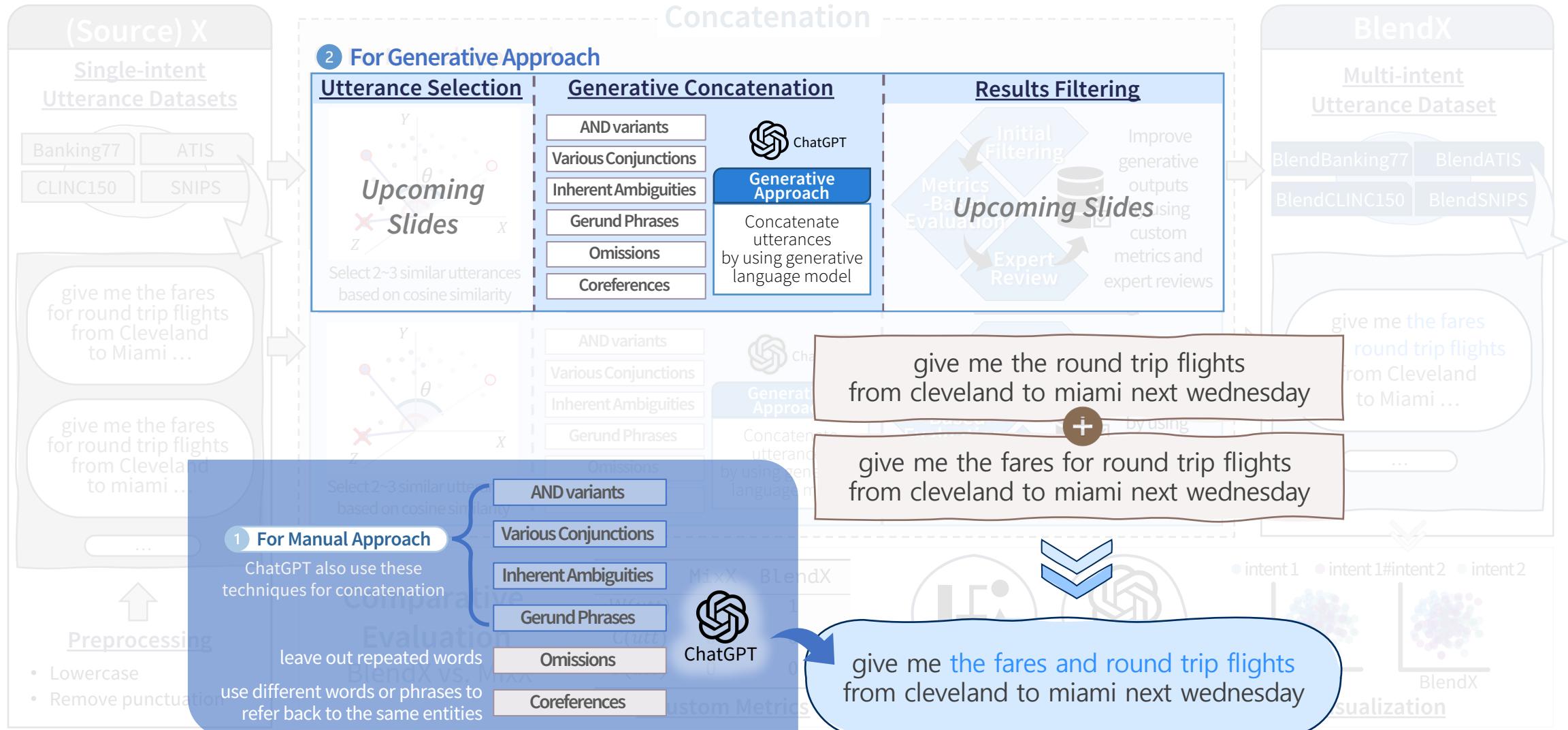
2-1) Manual Approach



2-2) Generative Approach



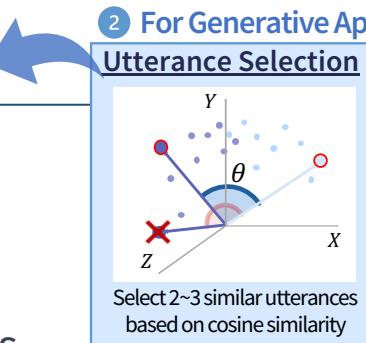
2-2) Generative Approach



Utterance Selection for 2-2) Generative Approach

• Process

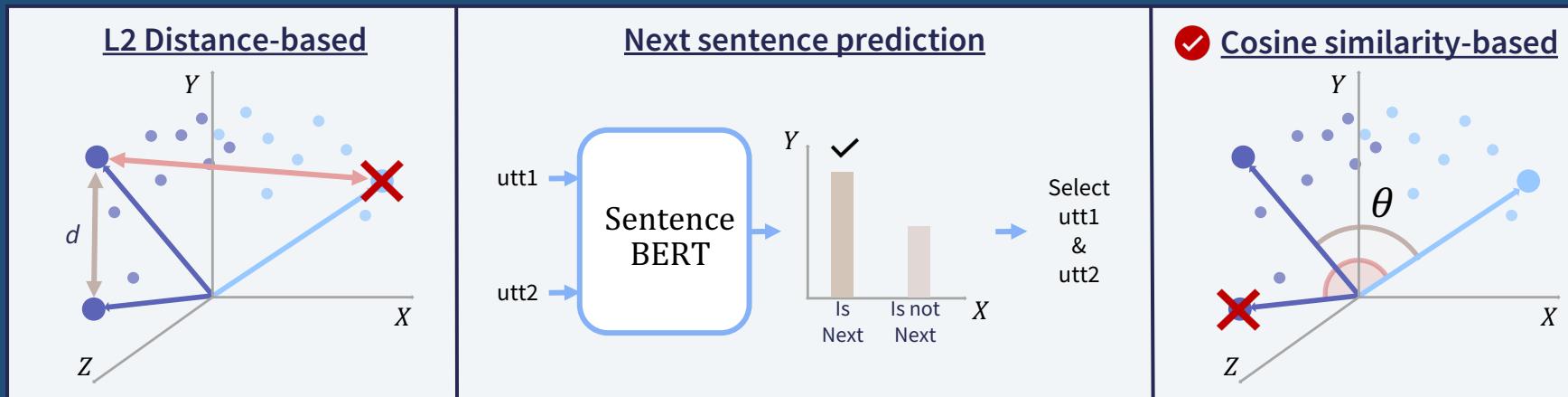
1. Generate embeddings for each single-intent utterance using SentenceBERT.
2. Select utterances for concatenation based on high similarity between embeddings.
 - * Chosen utterances will have different intents.



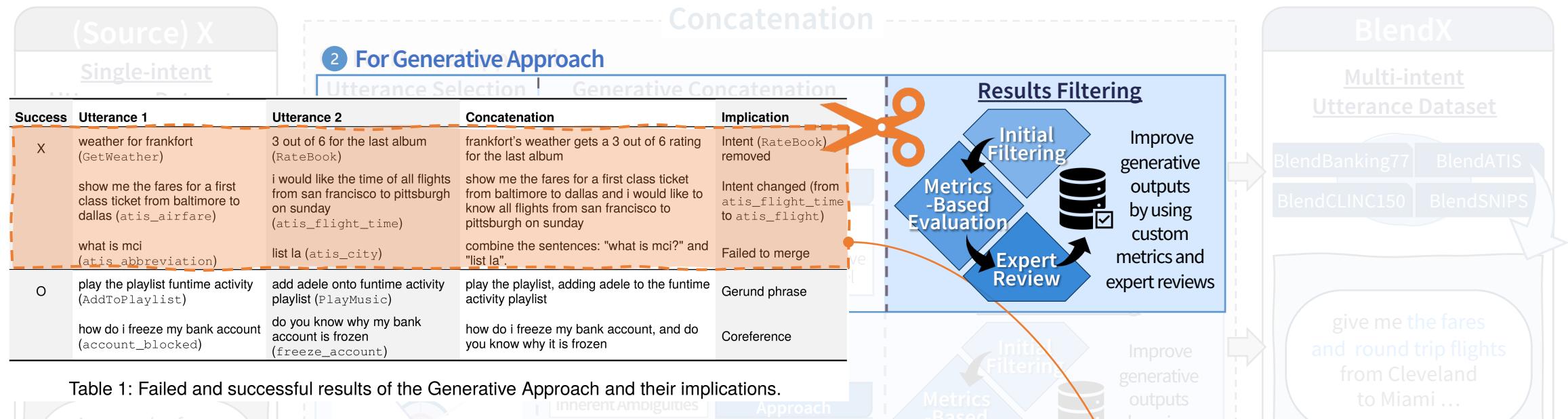
Generative Concatenation
AND variants
Various Conjunctions
Inherent Ambiguities
Gerund Phrases
Omissions
Coreferences

• 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.

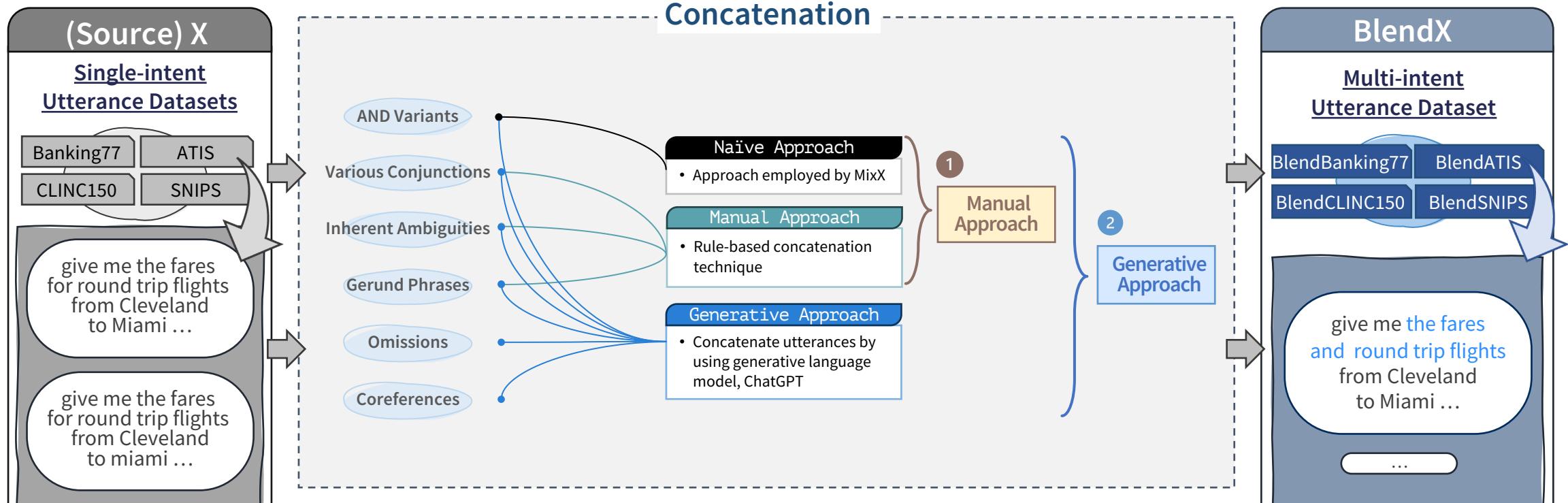


Results Filtering for 2-2) Generative Approach



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

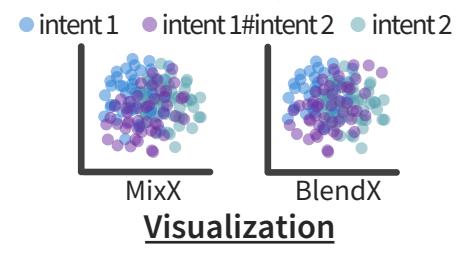
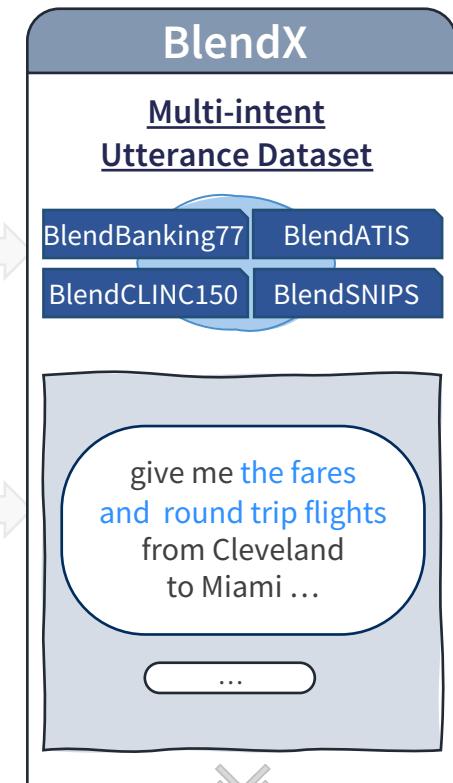
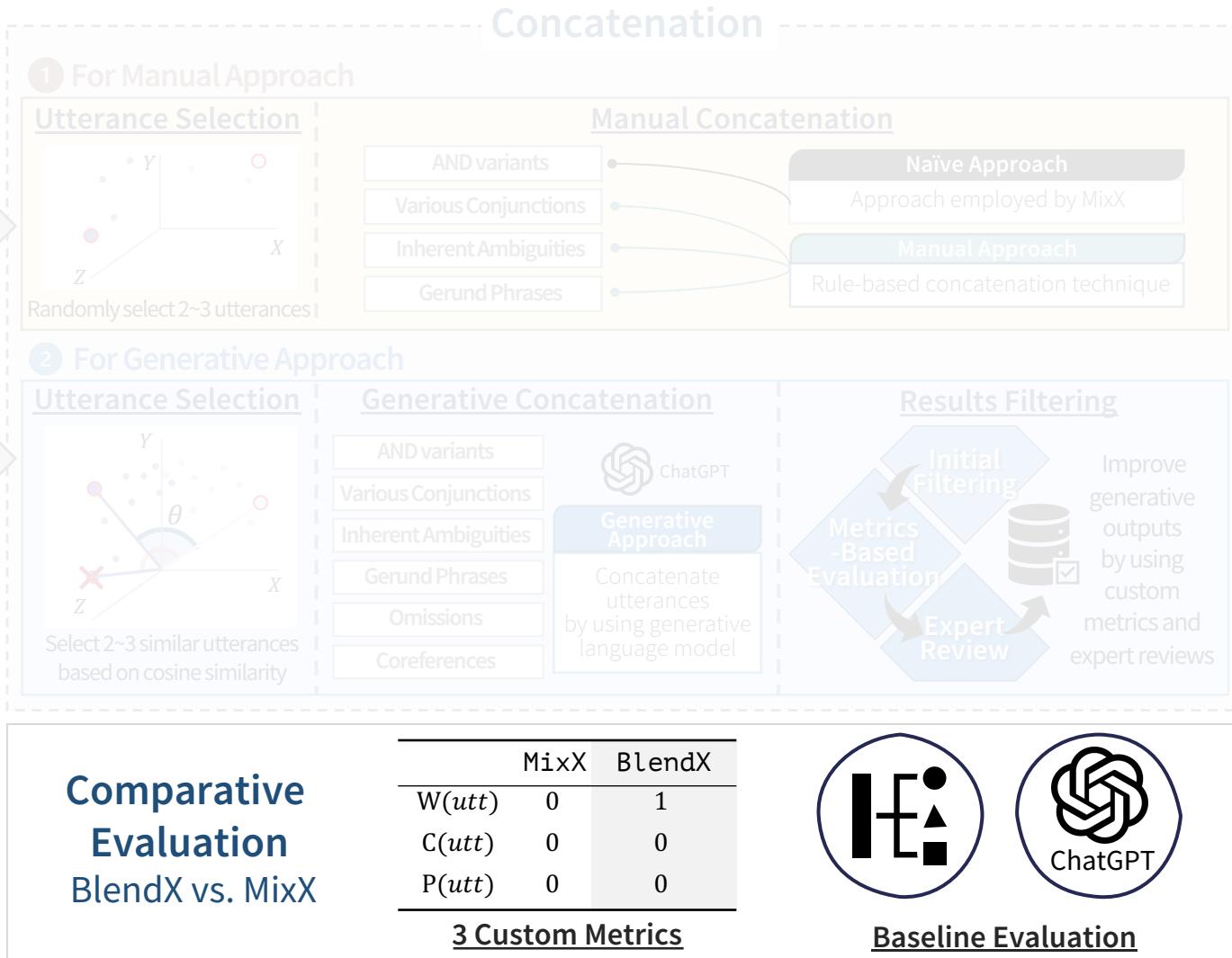
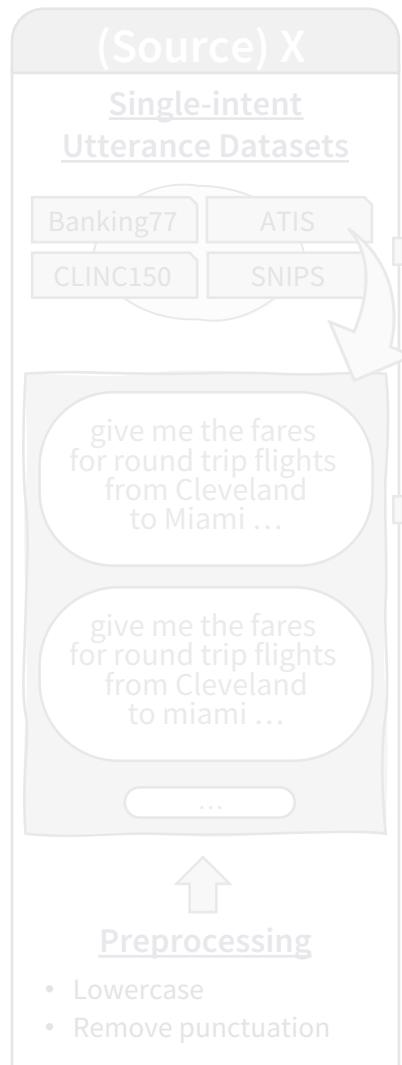


BlendX : Complex multi-intent detection with blended patterns

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

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		$W(utt, 2)$	$C(utt, 2)$
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	$10 - (5 + 8) = -3$	$W(\bullet, 2) = 1$
Conjunctions	0	0	0	$0 - (0 + 0) = 0$	$C(\bullet, 2) = 1$
Pronouns	1	0	0	$1 - (0 + 0) = 1$	$P(\bullet, 2) = 1$
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	$11 - (5 + 5) = 1$	$W(\bullet, 2) = 0$
Conjunctions	1	0	0	$1 - (0 + 0) = 1$	$C(\bullet, 2) = 0$
Pronouns	1	0	0	$1 - (0 + 0) = 1$	$P(\bullet, 2) = 1$

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(<i>utt</i> , 2)(\uparrow)	0%	37%	29%	0%	36%	18%	0%	46%	37%	0%	48%	28%
C(<i>utt</i> , 2)(\uparrow)	0%	56%	10%	0%	52%	15%	0%	50%	27%	0%	56%	32%
P(<i>utt</i> , 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.

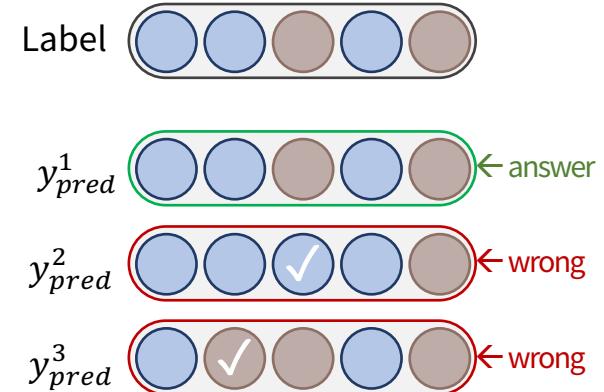
- Notably, **MixX** did not involve **implicit** concatenation.
(‘Naïve’ refers to the original construction method of **MixX**, meaning concatenation using only **and**, **and then**, **and also**.)
- Particularly, **manual concatenation** often resulted in shorter utterance lengths.
- Conversely, **generative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

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

MixATIS

and
and then
and also

Naïve
Manual
ChatGPT

give me
the round trip flights
from ~ wednesday
and give me
the fares for round trip flights
from ~ wednesday

BlendATIS

various conj.
gerunds
omission
coreference
...

Naïve
Manual
Generative

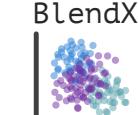
give me
the fares and
round trip flights
from ~ wednesday

Multi-Intent Datasets

#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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