

Opinion Summarization from Customer Reviews

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Megagon Labs

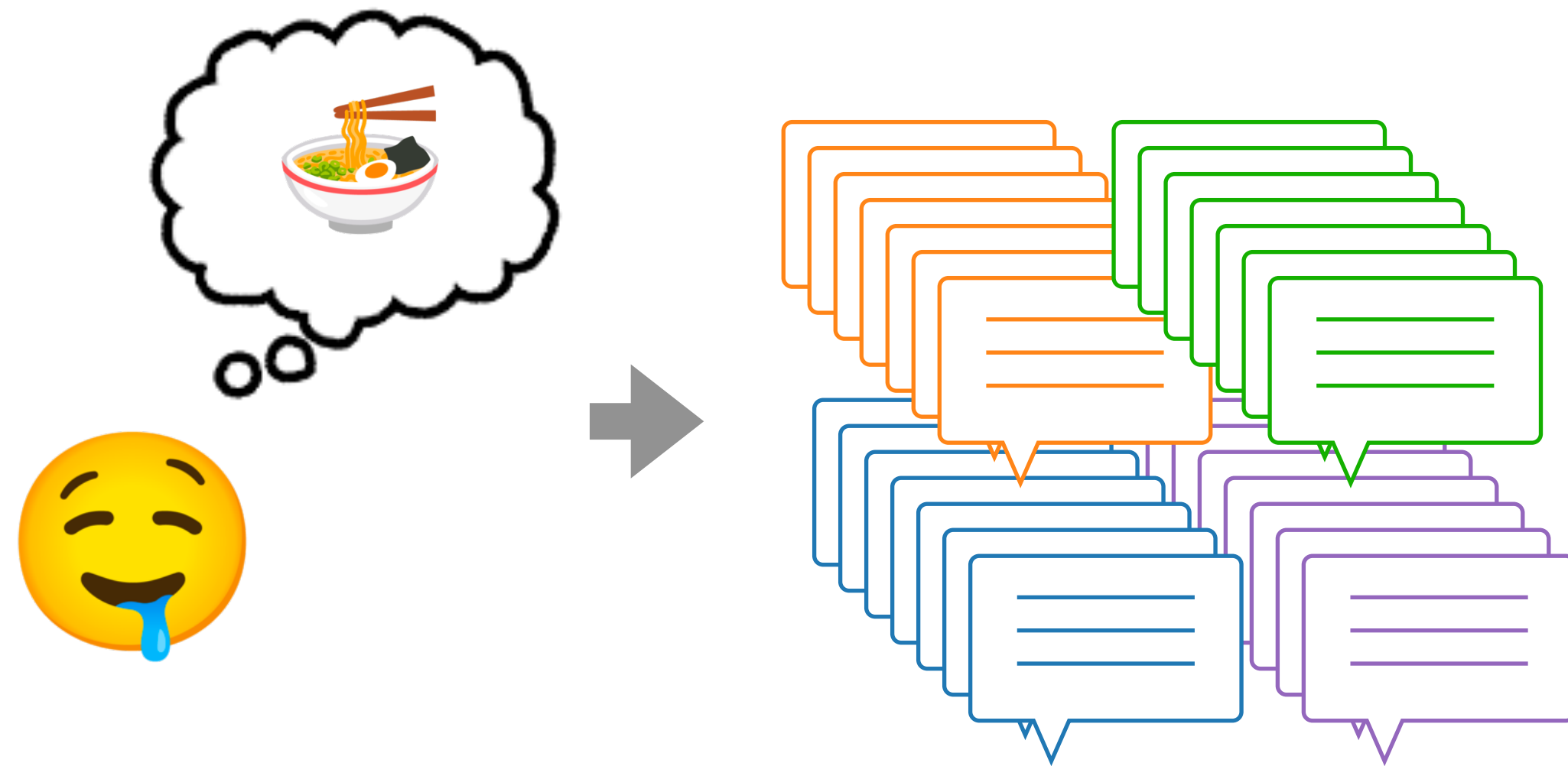
Opinion Summarization



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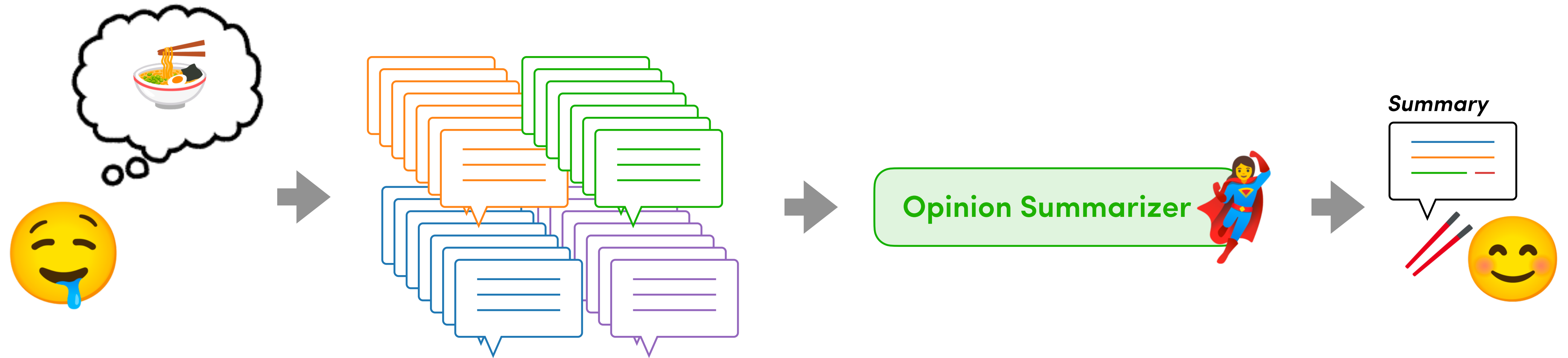
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Outline

1. How to build opinion summarization system without training pairs
2. How to make it more comparable between entities

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2. How to make it more comparable between entities

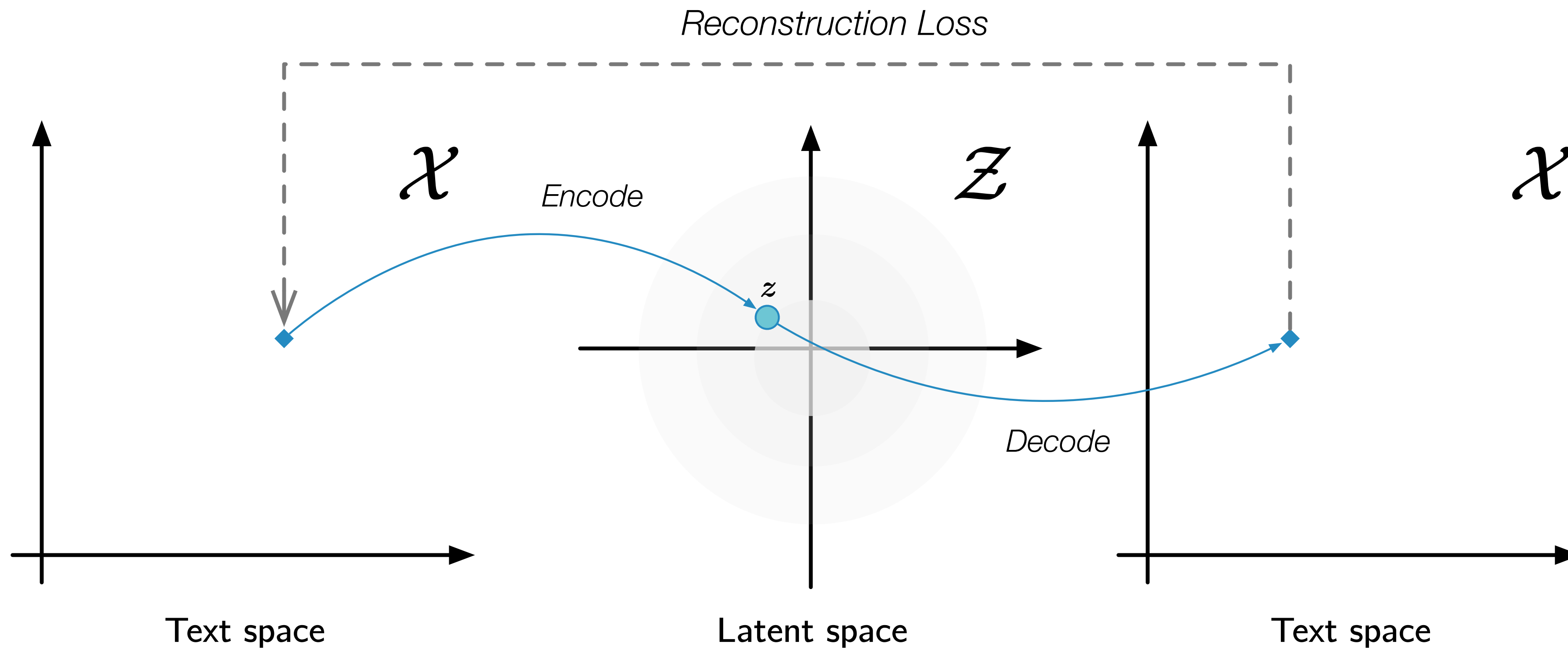
Not Enough Data for Supervised Learning

- Human written reference summaries are extremely limited
- Large amounts of reviews are available

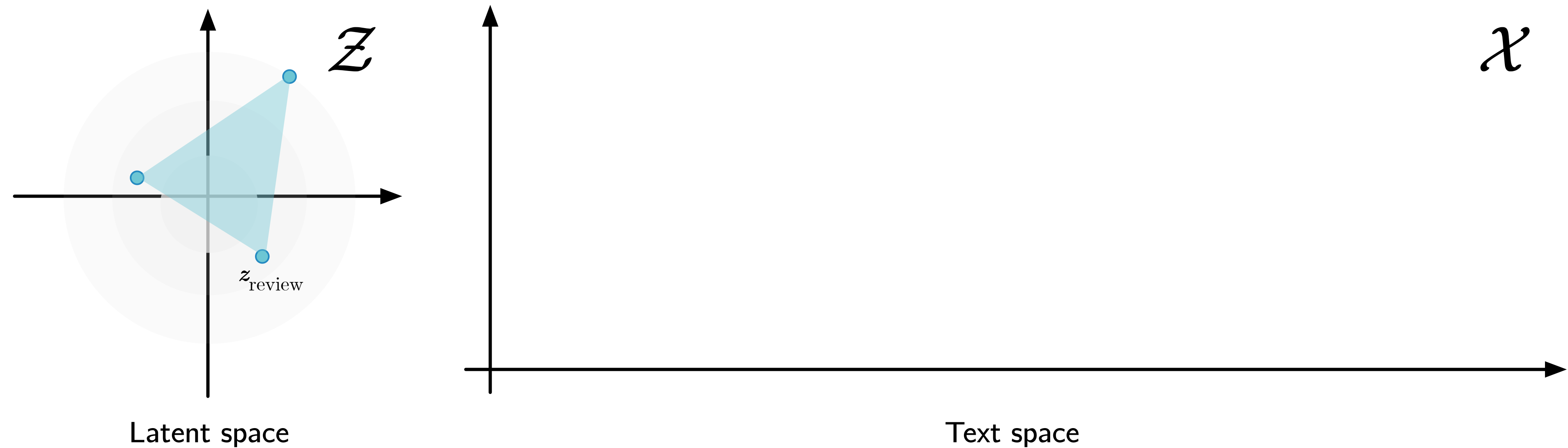


Unsupervised Learning for Opinion Summarization

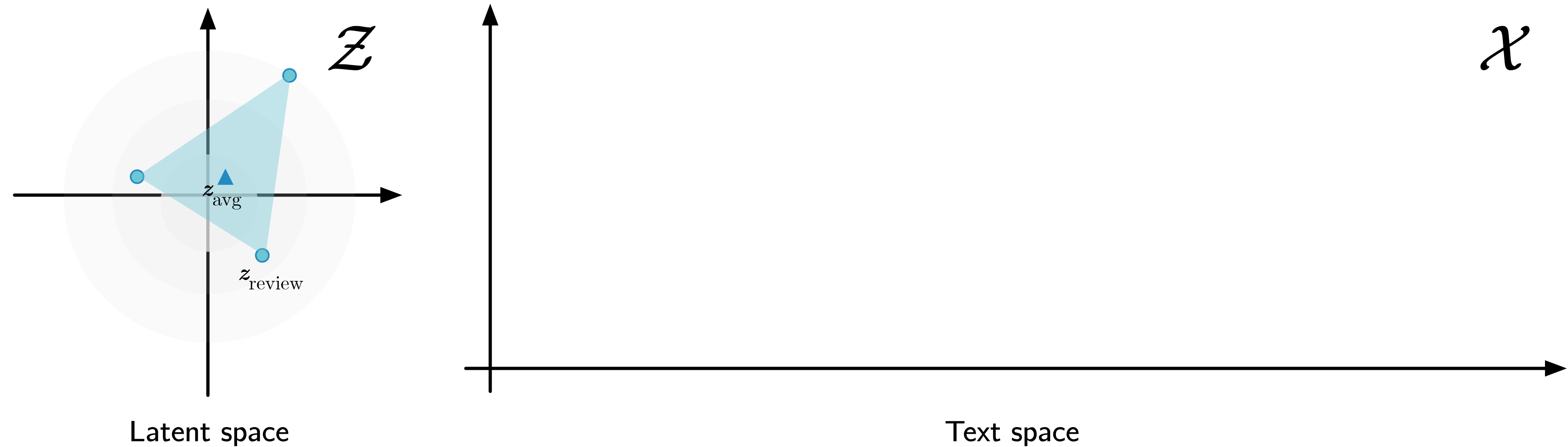
Standard Approach: Auto Encoder



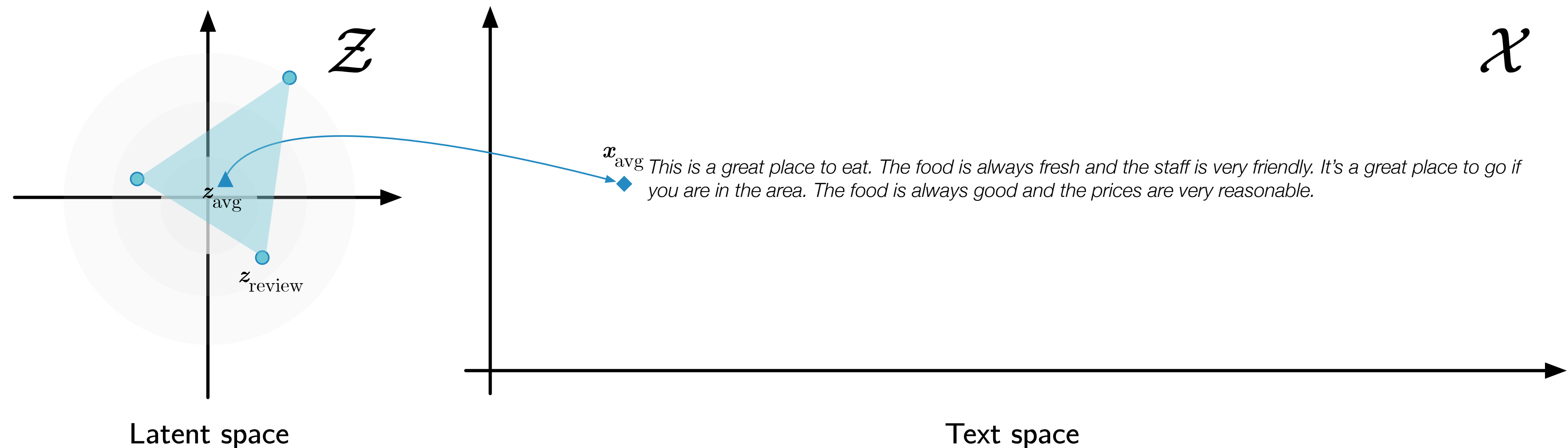
Standard Approach: Summarizing in Latent Space



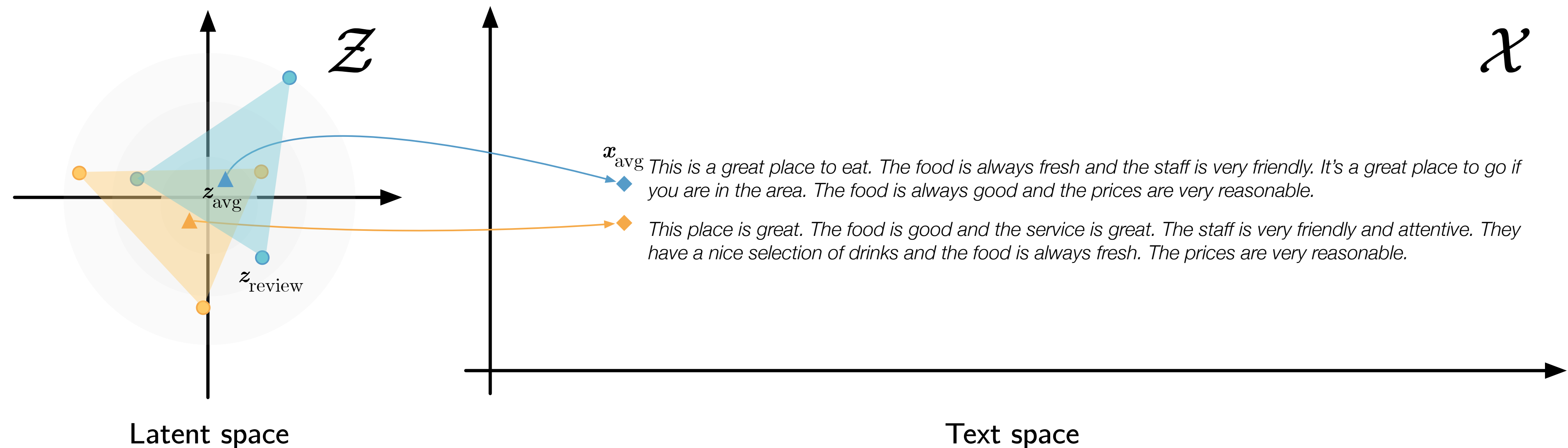
Standard Approach: Summarizing in Latent Space



Standard Approach: Summarizing in Latent Space



Summary Vector Degeneration

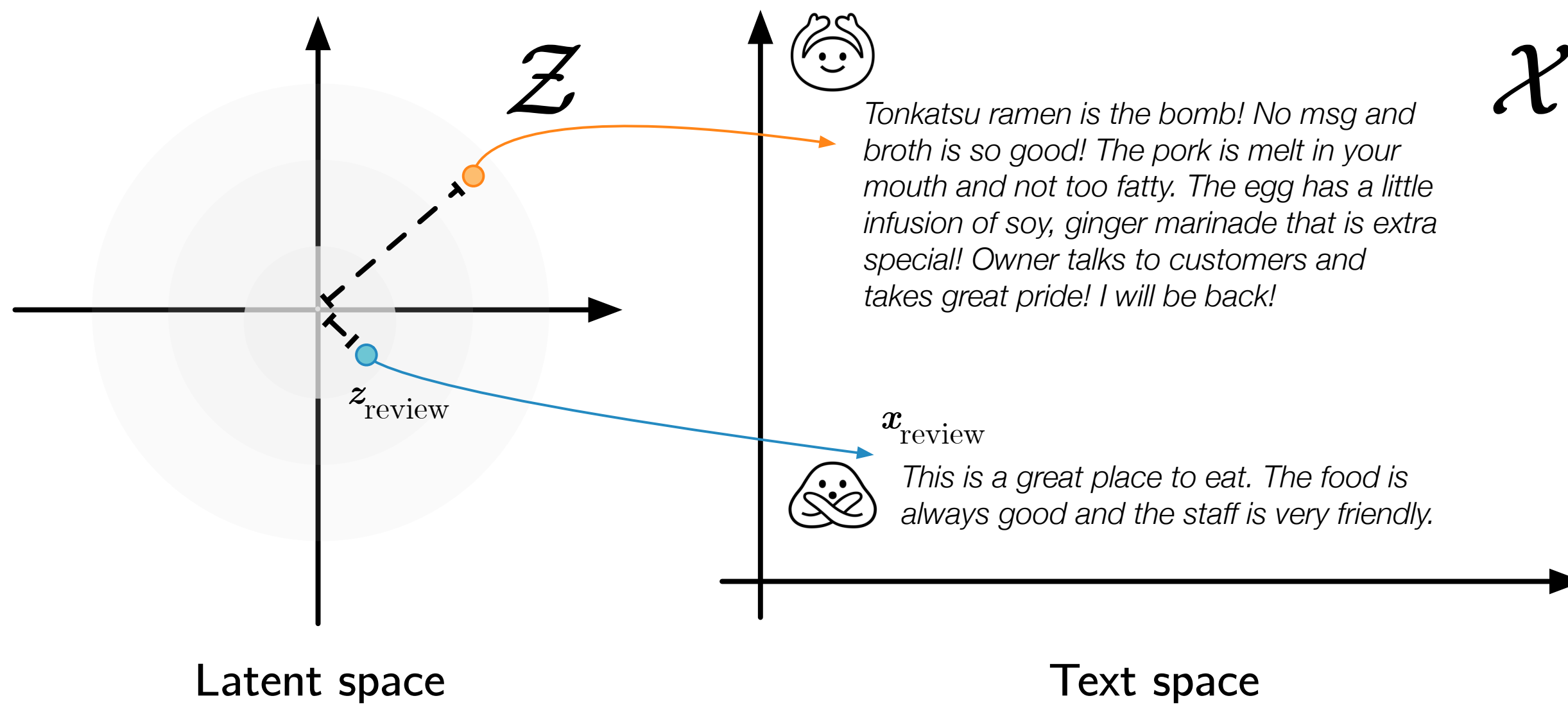


😞 Models tend to generate generic summaries from averaged vector

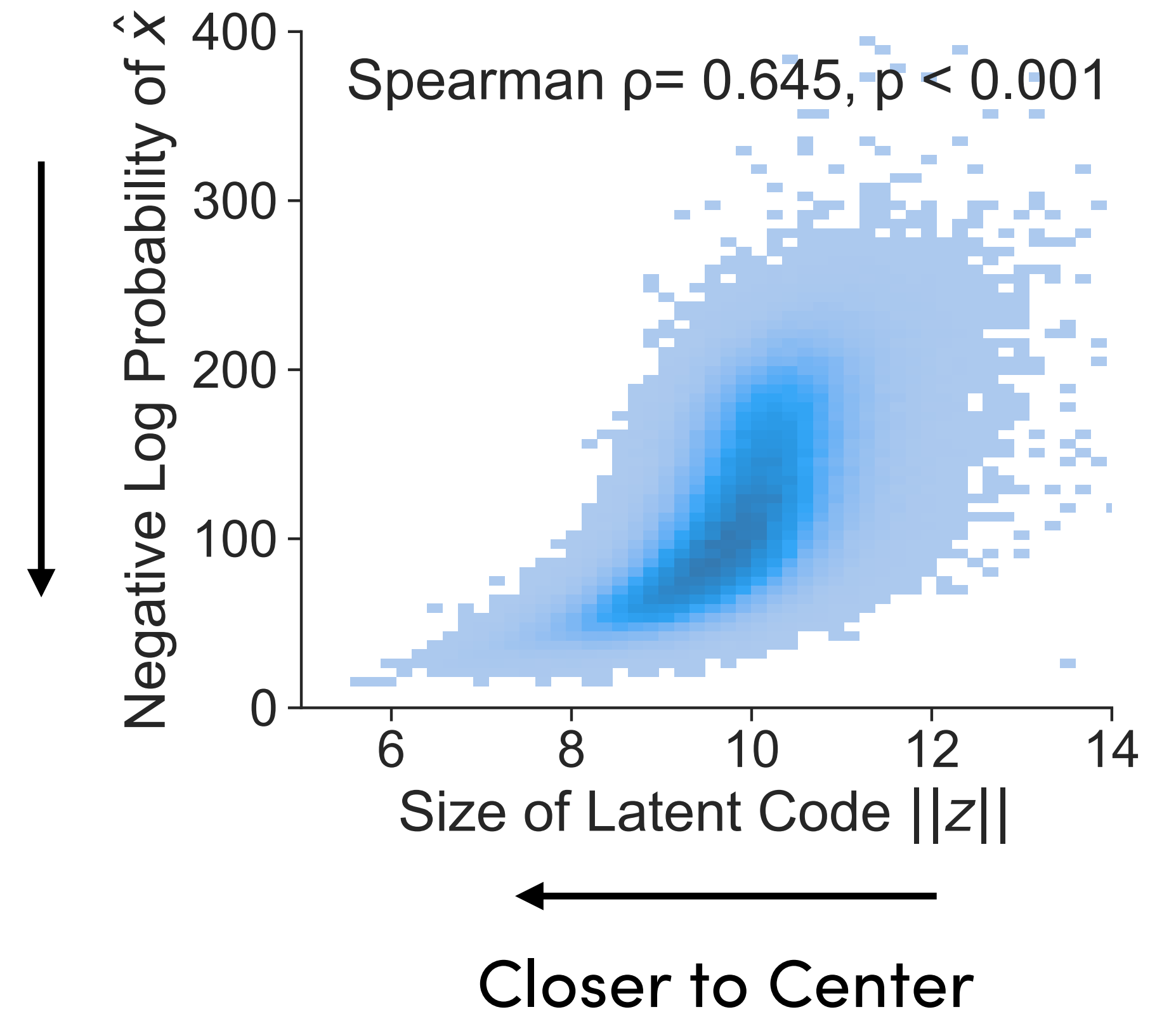
Objective:

- Why *summary vector degeneration* happen?
- How to generate more *informative* opinion summaries?

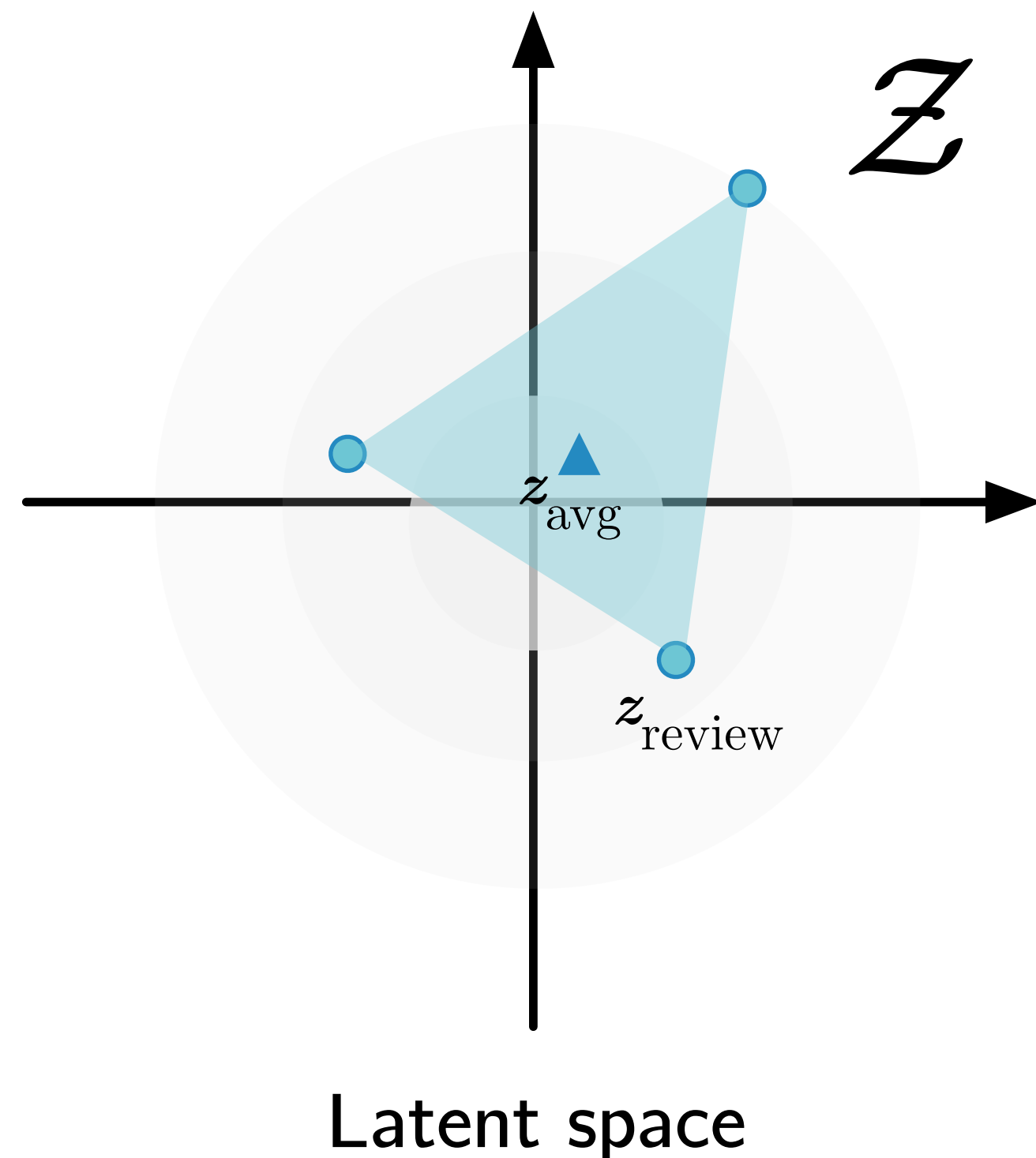
Smaller vectors are less informative



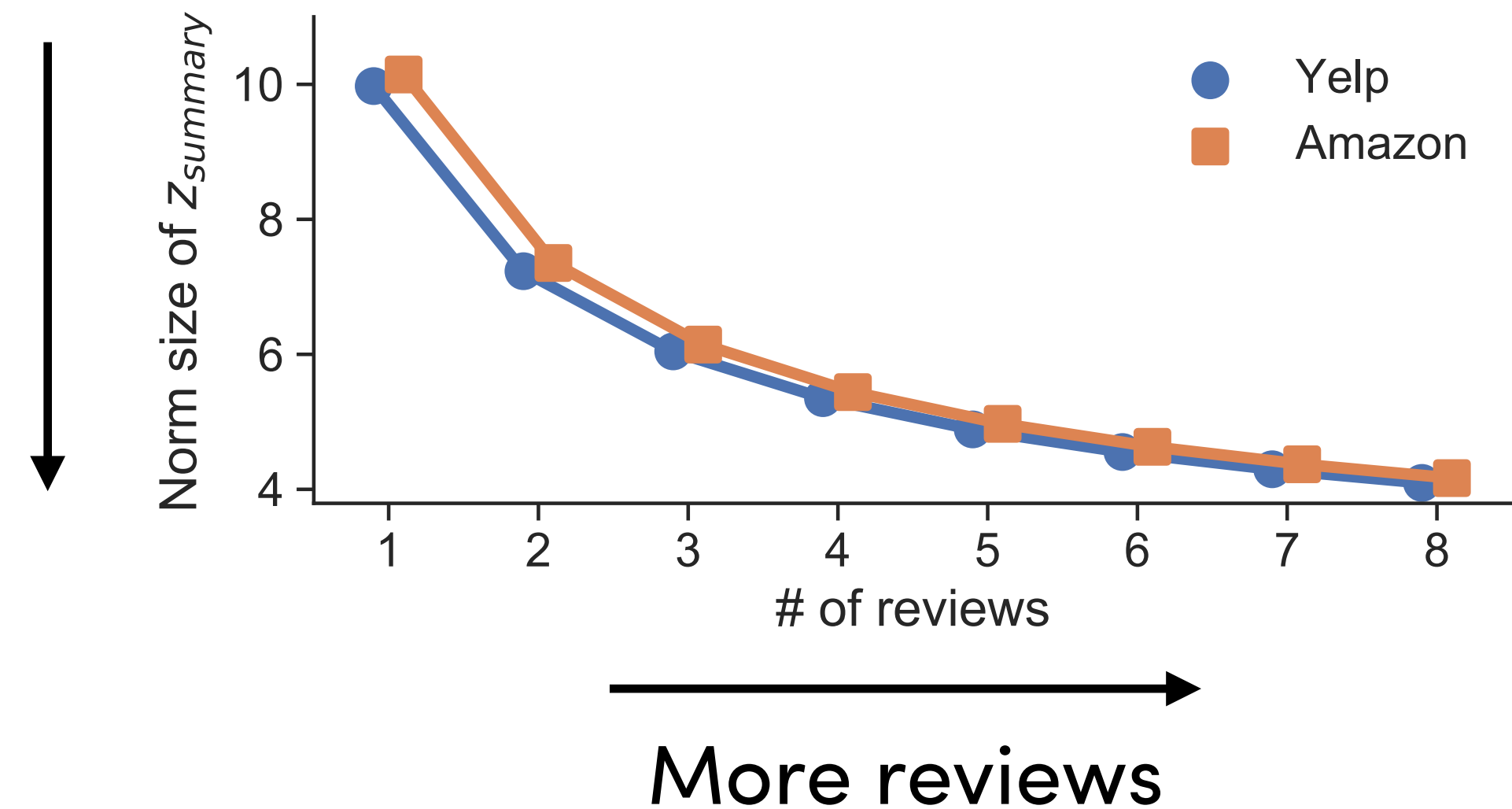
Less Informative



Averaged vectors are small

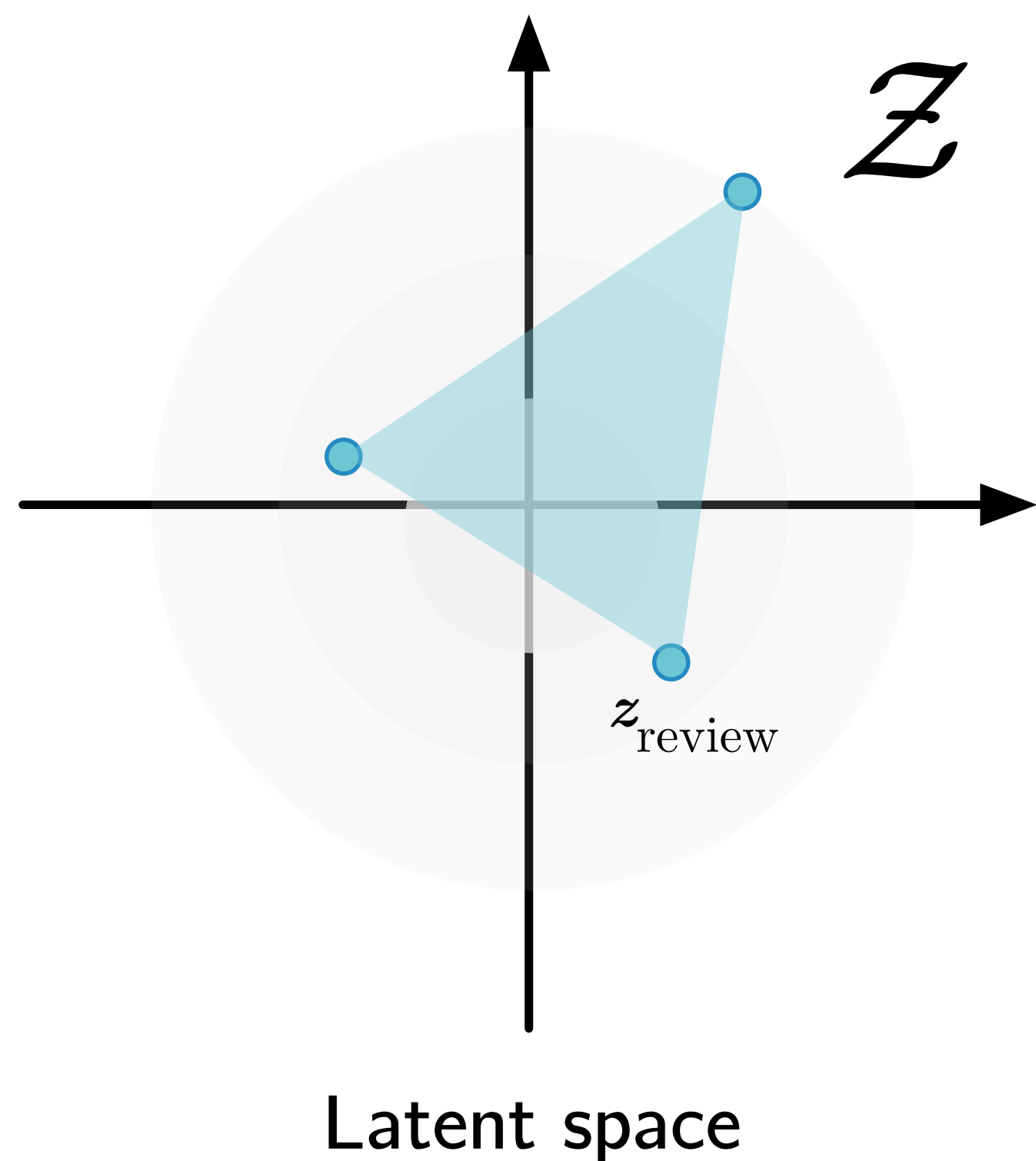


Closer to Center

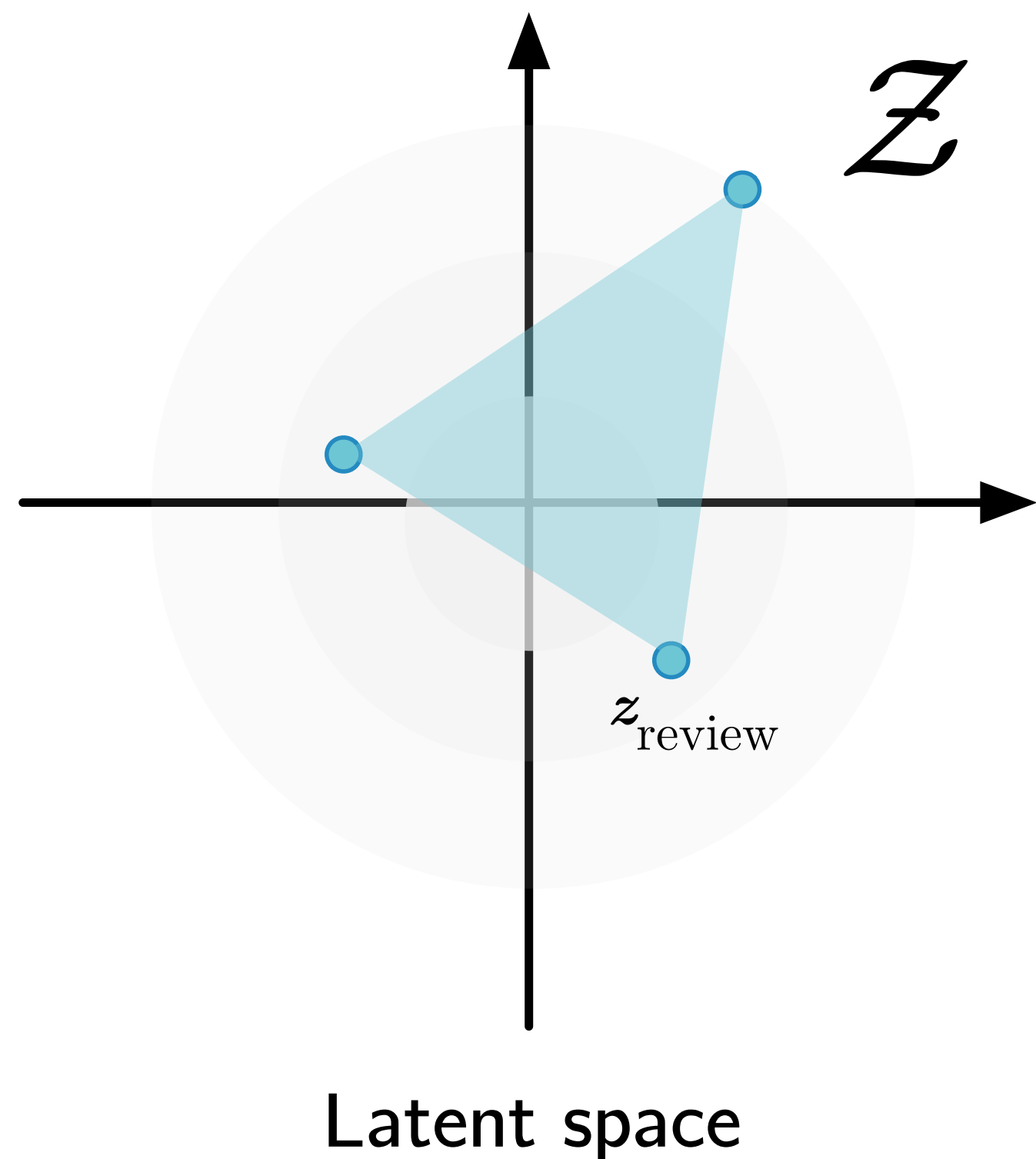


😞 Averaged vector tends to generate less informative summary

Our Approach: *Finding Better Summary Vector*

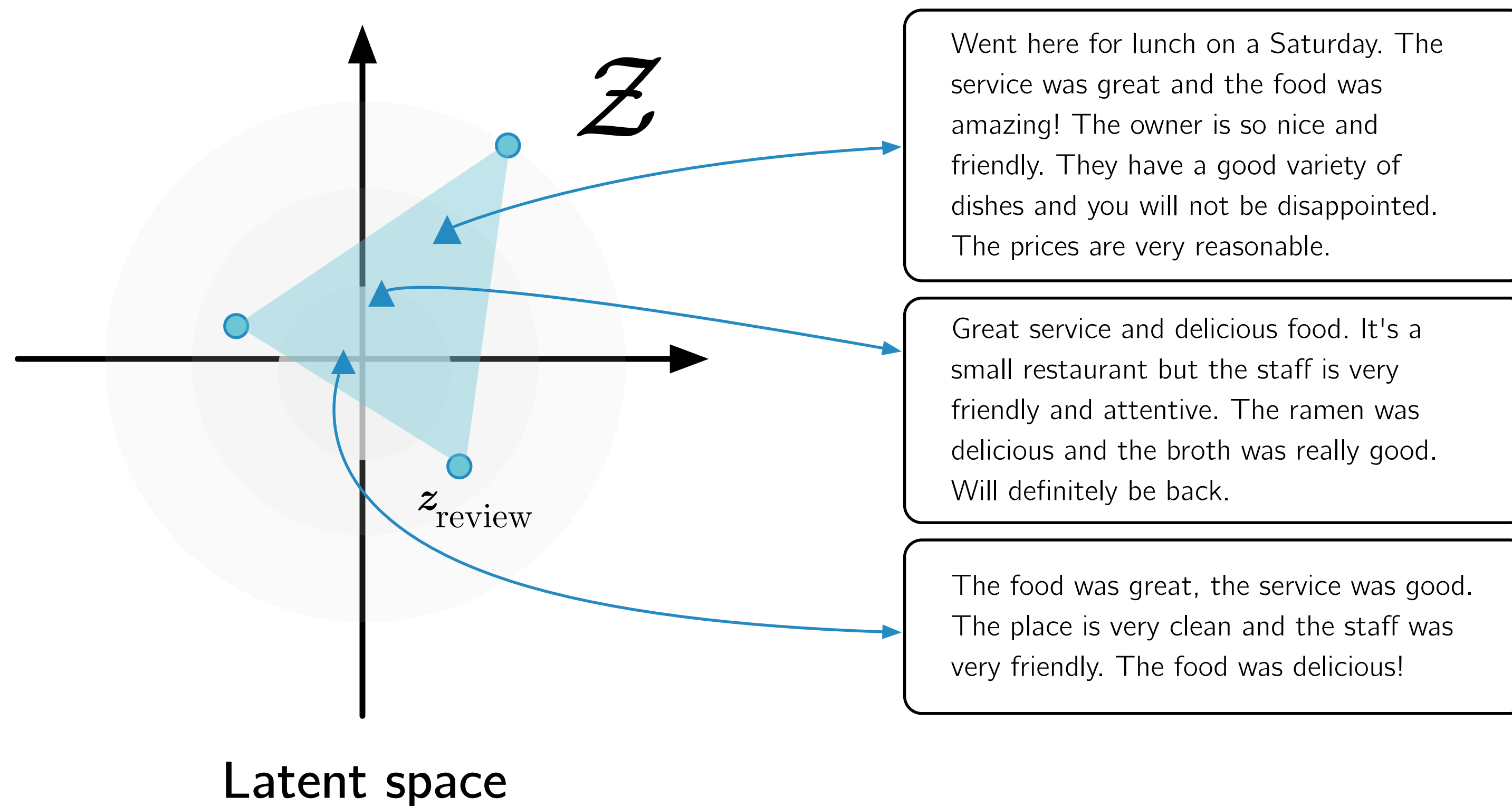


Our Approach: *Finding Better Summary Vector*

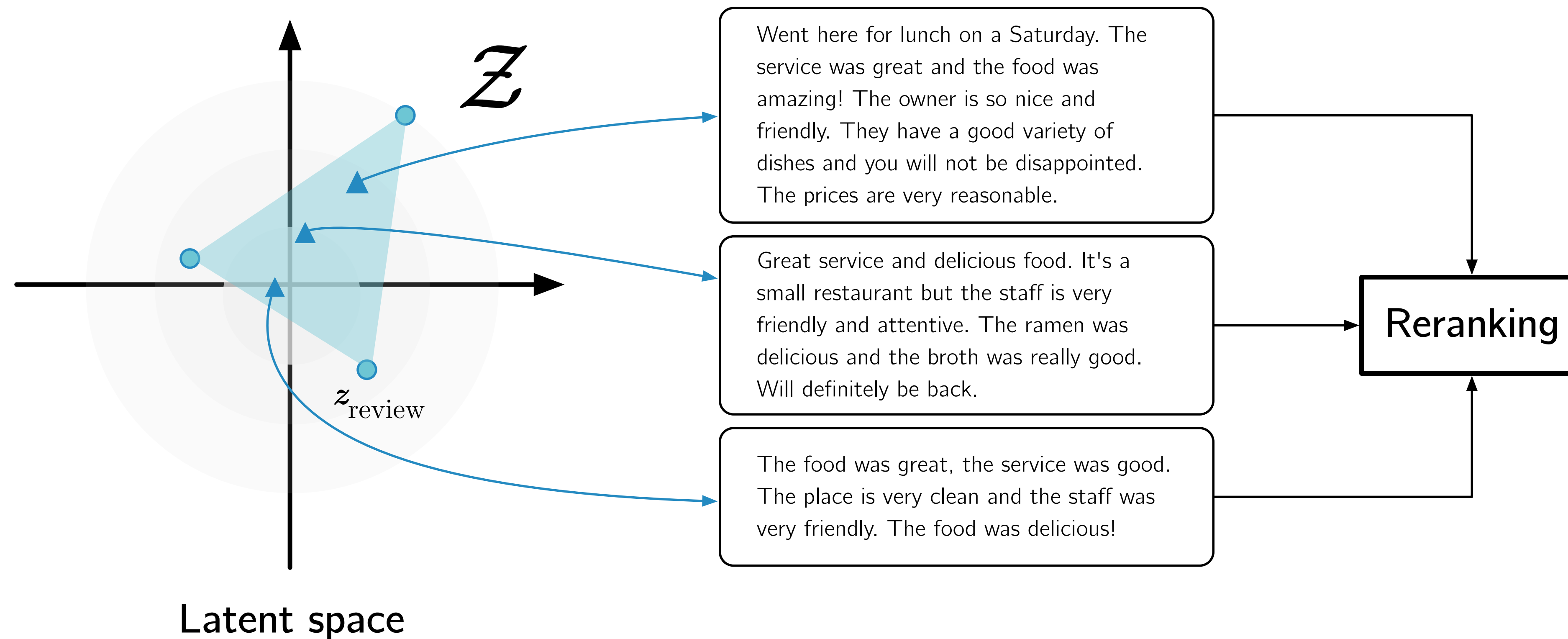


Assumption:
Good summary vector would be
in the convex hull of reviews.

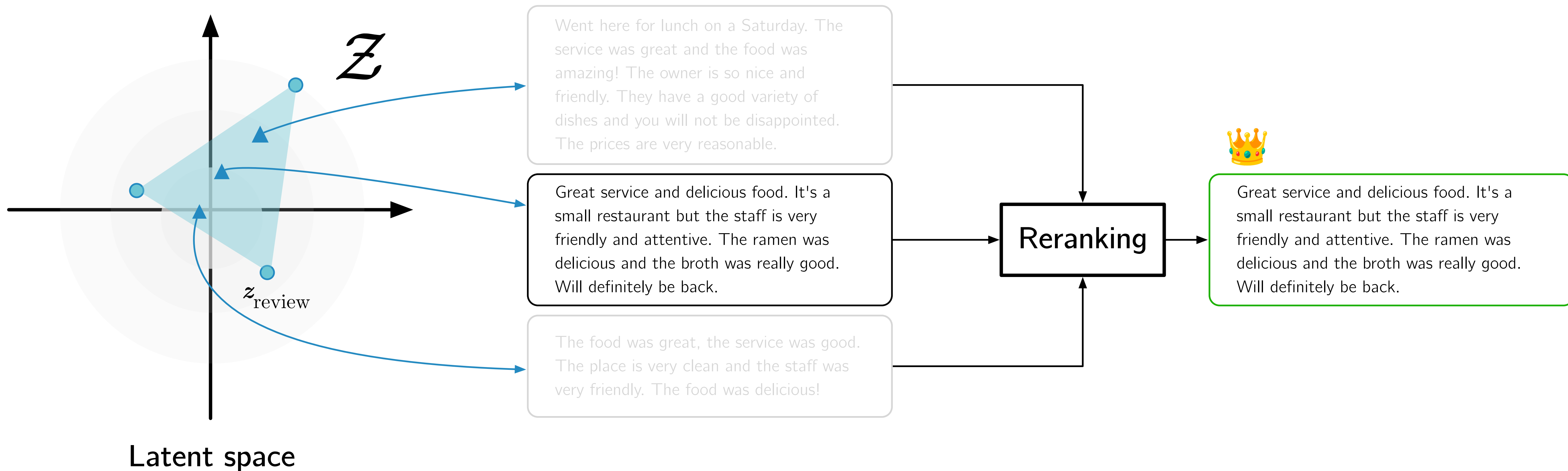
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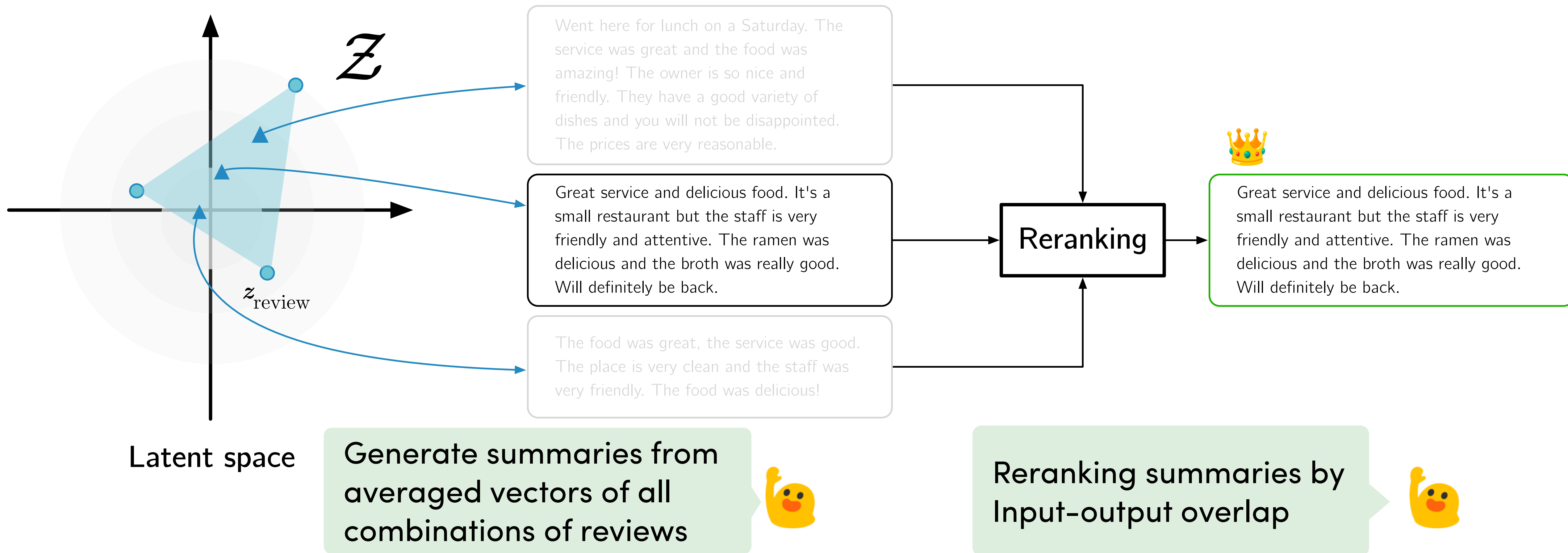
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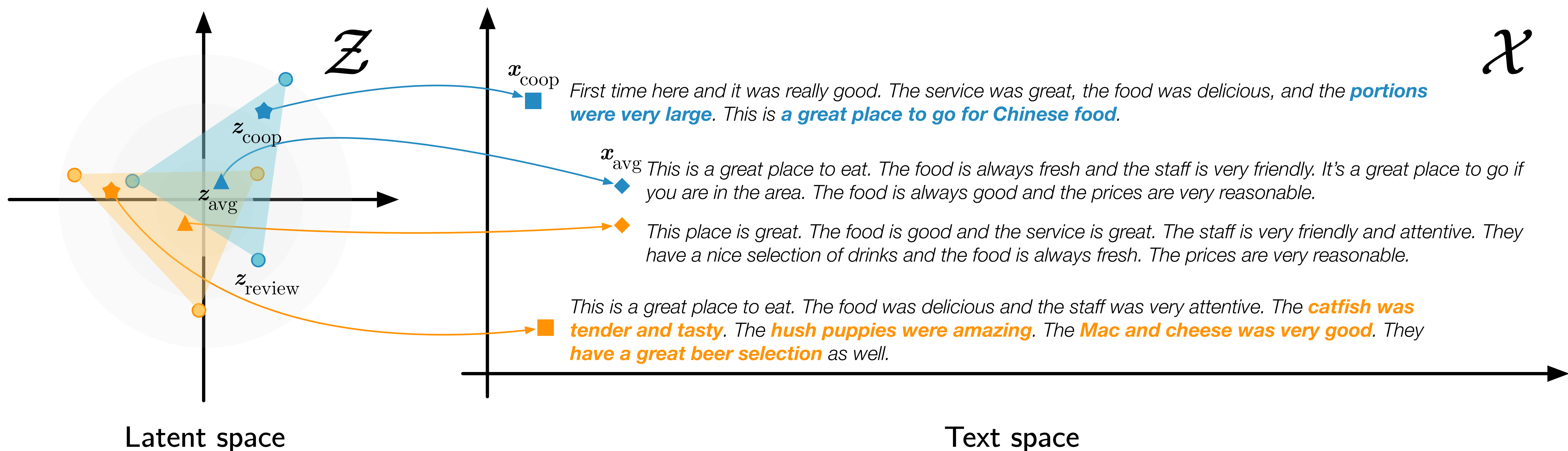
Our Approach: *Finding Better Summary Vector*




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Our Approach: *Finding Better Summary Vector*



Results

		Yelp			Amazon			
Method		R1	R2	RL	R1	R2	RL	#Param
	COOP							
	BIMEANVAE	35.37	7.35	19.94	36.57	7.23	21.24	13M
	SimpleAvg							
	BIMEANVAE	32.87	6.93	<u>19.89</u>	33.60	6.64	<u>20.87</u>	13M
	MeanSum [†]	28.46	3.66	15.57	29.20	4.70	18.15	28M
	CopyCat [†]	29.47	5.26	18.09	31.97	5.81	20.16	21M
	<i>Weakly Supervised</i>							
	PlanSum [‡]	<u>34.79</u>	<u>7.01</u>	19.74	32.87	6.12	19.05	

※BiMeanVAE: Text VAE (Bowman+, 2016) with BiLSTM and MeanPooling

Few lines to run VAE-based Summarizer!

```
# pip install git+https://github.com/megagonlabs/coop.git
from coop import VAE

# Automatically download from Huggingface 🤗 model hub
vae = VAE("megagonlabs/bimeanvae-yelp")

# Encode List[str] into Tensor
reviews = [
    "I love this ramen shop!! Highly recommended!!",
    "Here is one of my favorite ramen places! You must try!"
]
zs = vae.encode(reviews) # Tensor: [num_reviews * latent_size]

# Averaging to summarize!
z = zs.mean(dim=0, keepdim=True)
vae.generate(zs)

# Output
['I love this place! Food is amazing!!']
```

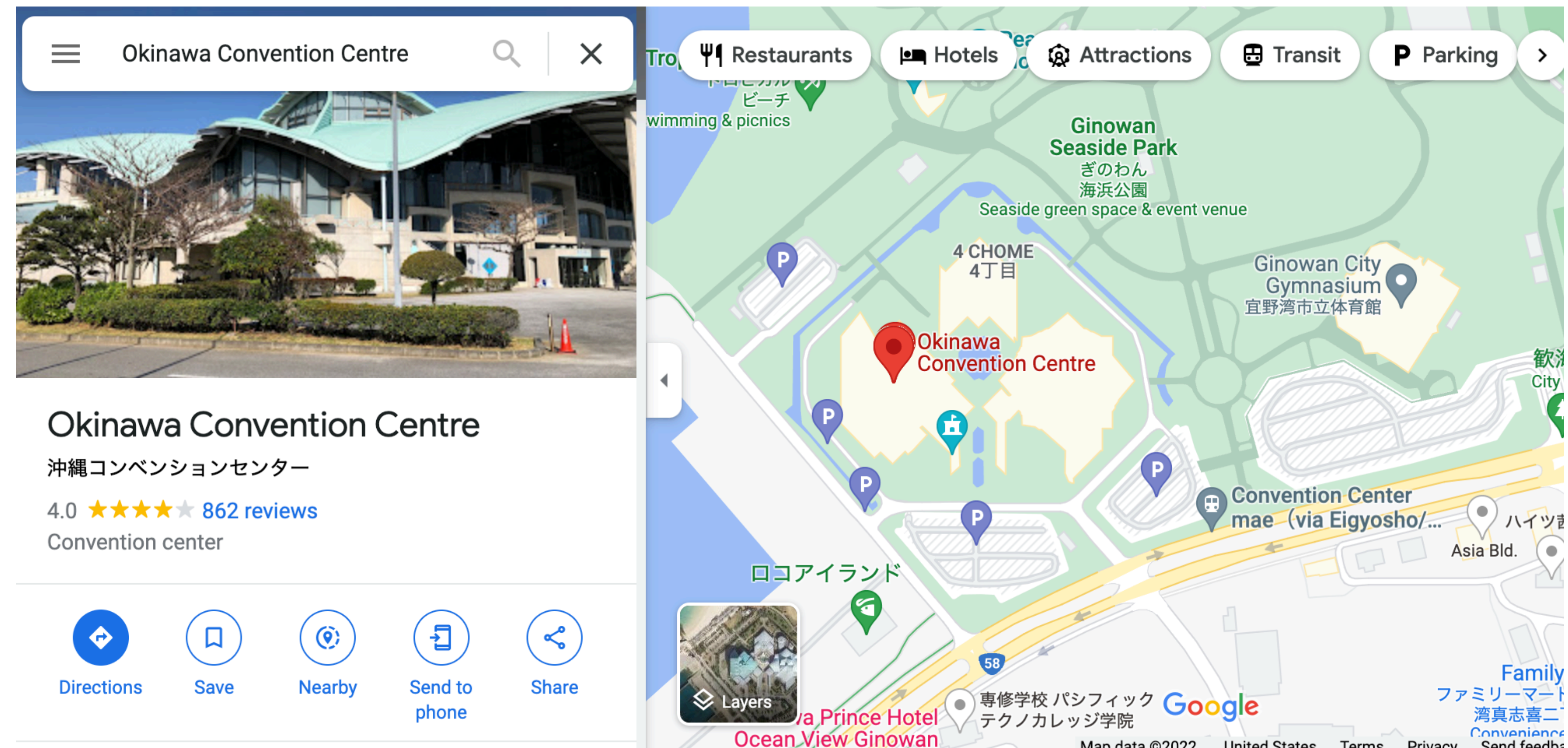
<https://github.com/megagonlabs/coop>

Outline

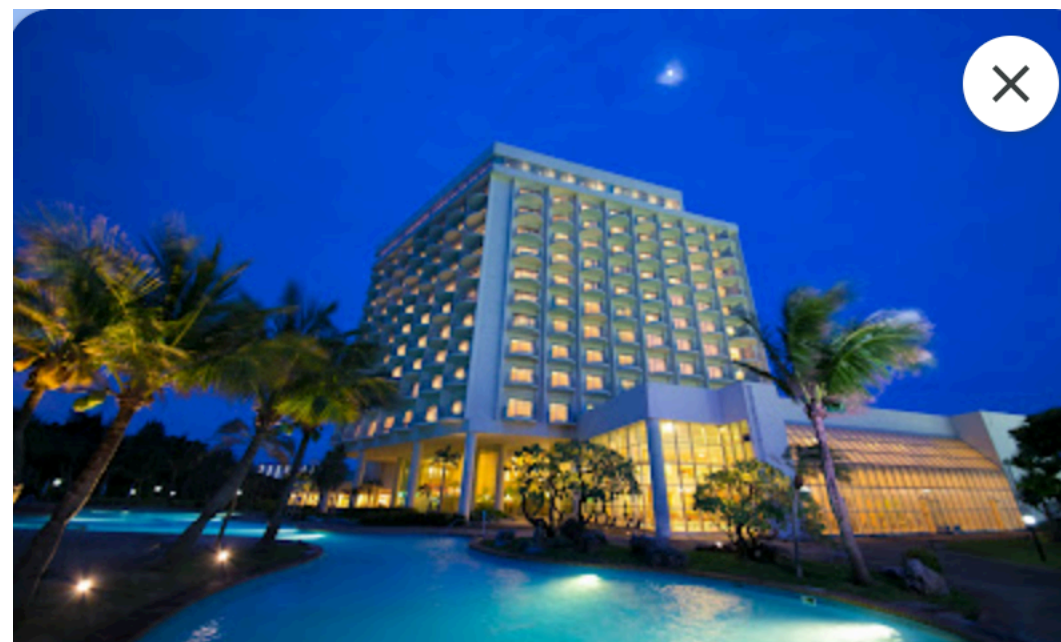
1. How to build a good opinion summarization system without training pairs
- 2. How to make it more comparable between entities**

Comparison is the key in decision making process

Where should I stay
for NLP2023??



Comparison is the key in decision making process



Laguna Garden Hotel Okinawa

ラグナガーデンホテル

4.2 ★★★★★ 1,636 reviews · 4-star hotel



Directions



Save



Nearby



Send to
phone



Share

CHECK AVAILABILITY

These two sound reasonable, but
which one should I pick?



Moon Ocean Ginowan Hotel & Residence

ムーンオーシャン宜野湾 ホテル&レジデンス

4.4 ★★★★★ 1,137 reviews · 4-star hotel



Directions



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CHECK AVAILABILITY

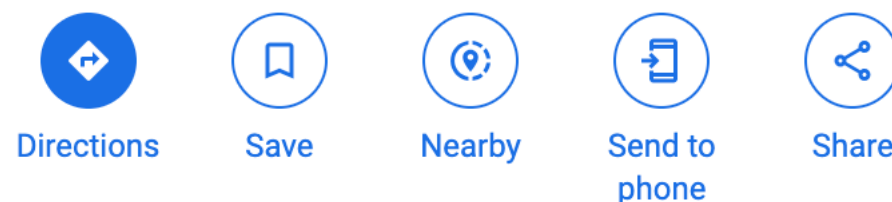
Single entity summarization models are not sufficient!



Laguna Garden Hotel Okinawa

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4.2 ★★★★★ 1,636 reviews · 4-star hotel



CHECK AVAILABILITY

“Large and clean rooms with great views. Convenient location near city center.”

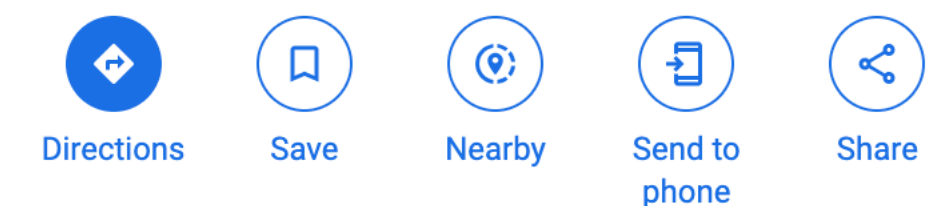
“Good location near city center and shopping. Rooms are large and clean. Rooms have views.”



Moon Ocean Ginowan Hotel & Residence

ムーンオーシャン宜野湾 ホテル&レジデンス

4.4 ★★★★★ 1,137 reviews · 4-star hotel



CHECK AVAILABILITY

Single entity summarization models often generates generic opinions

Summaries FAIL to include common/contrastive opinions

Contrastive summary contain many common opinions and vise versa.

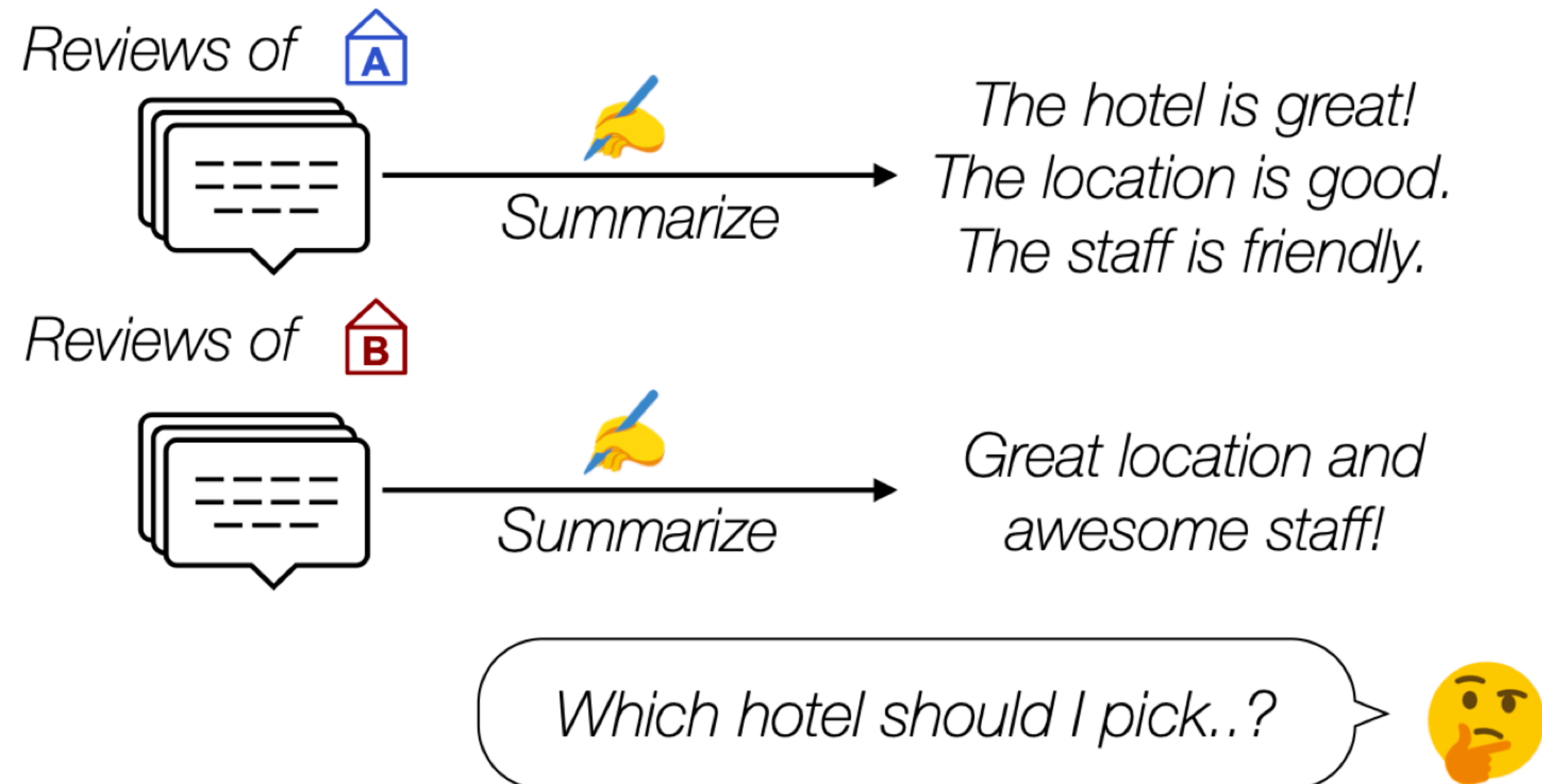
Text highlighted refers to common opinions (Bad content)

This is a perfect hotel for any type of stay and you will want to keep coming back for the tranquillity, *unbeatable price* and the great service. This hotel is in a really bustling area of Rome and close to the main sights of the city. **The rooms in the hotel are a good size, with spacious bathrooms and even some really great chocolates on the bed.** The hotel staff are very helpful and always willing to help out with their polite manners. The breakfast provided by the hotel was really good, *although a little bit basic.* **The elevator in this hotel is a little bit old but it's in good condition.**

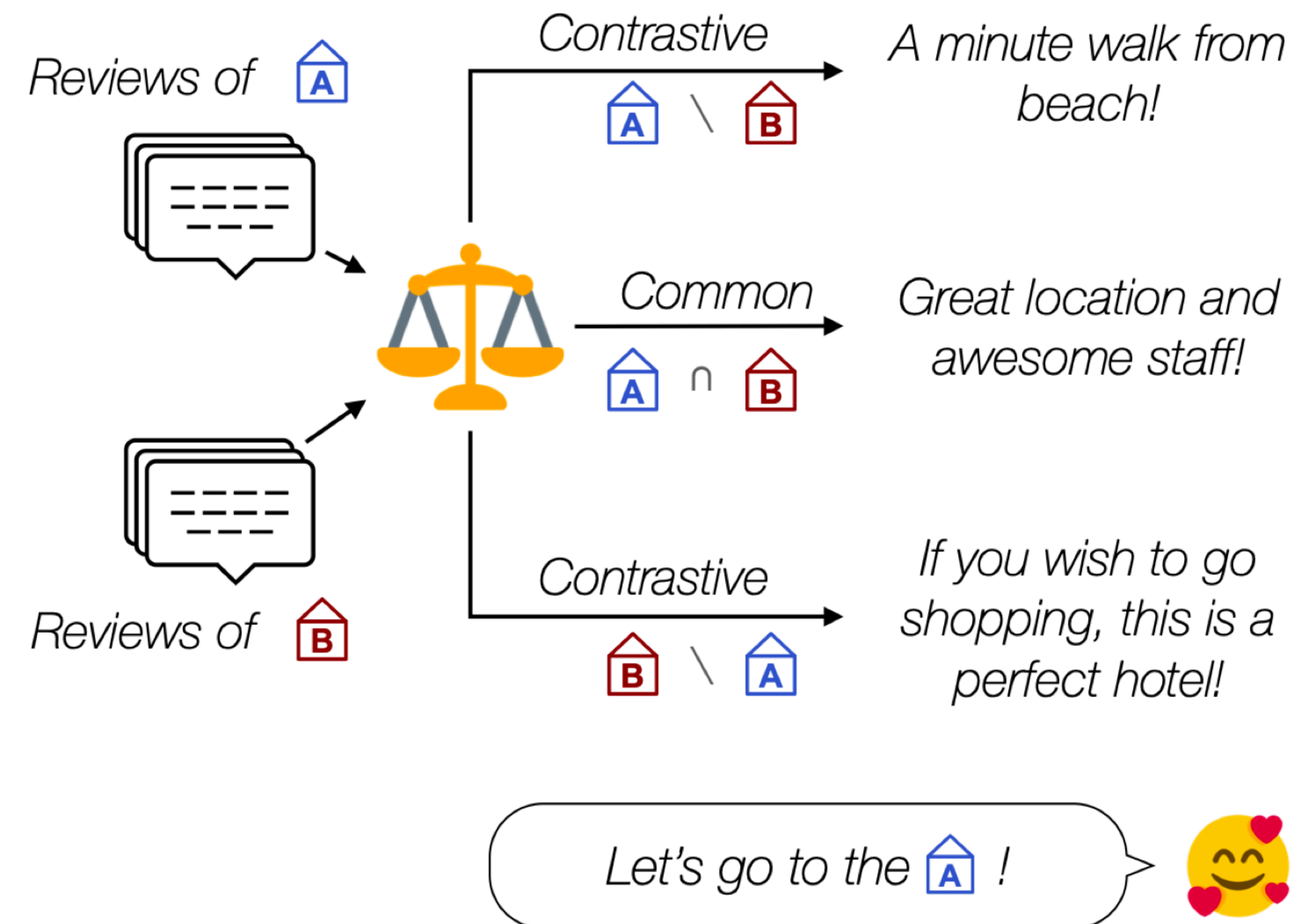
Text in blue refers to contrastive opinions (Good content)

Introducing *Comparative Opinion Summarization*

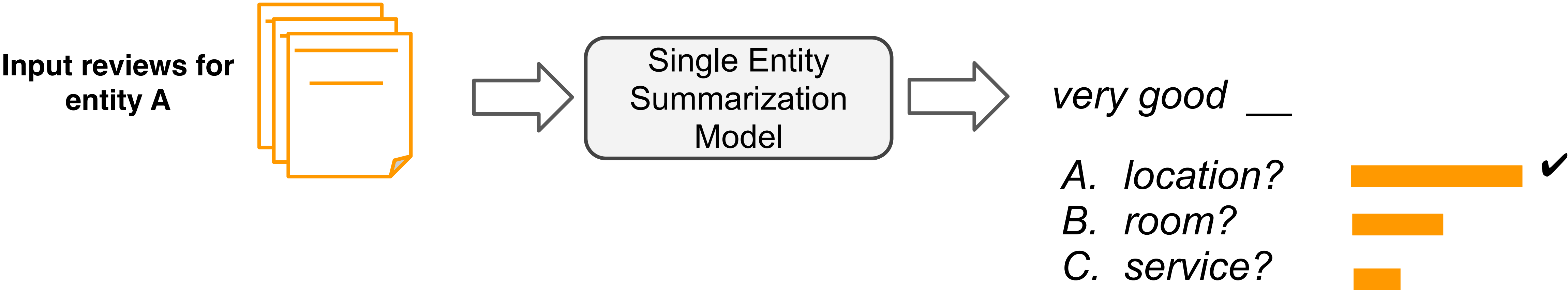
Single Entity Opinion Summarization



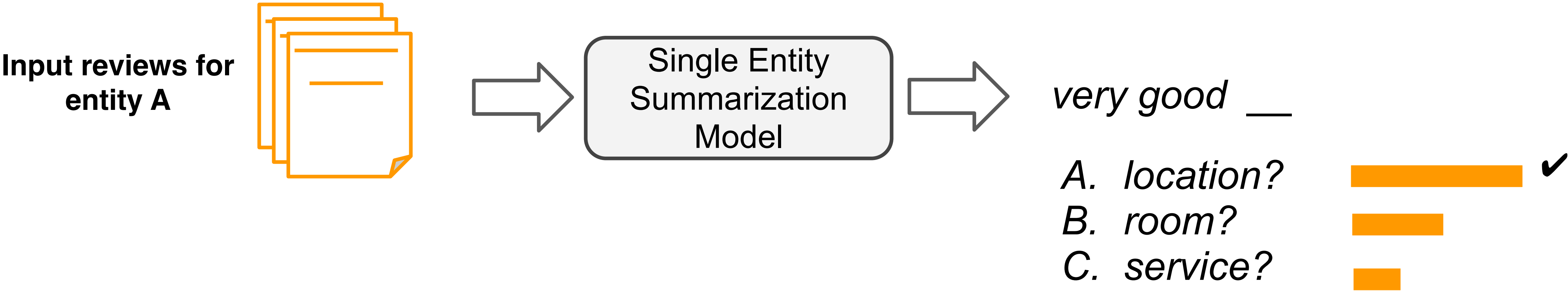
Comparative Opinion Summarization



Use Collaborative-Decoding for summary generation

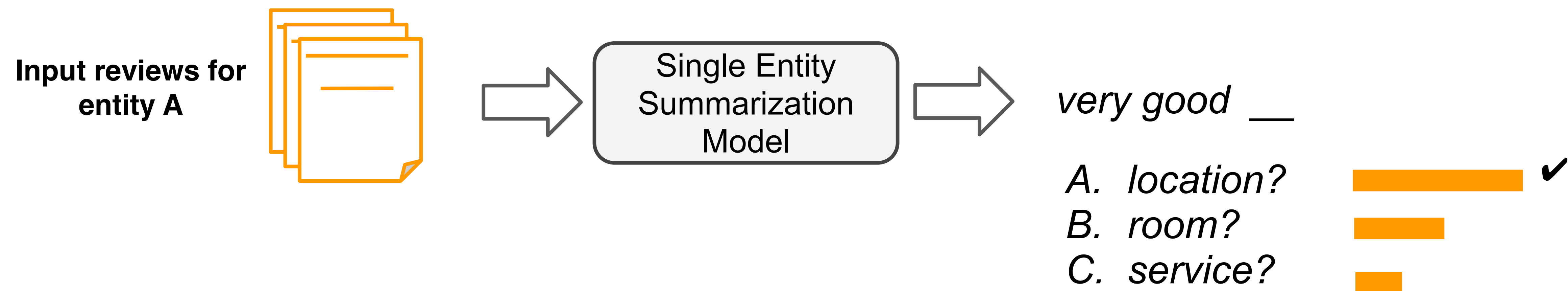


Use Collaborative-Decoding for summary generation



Problem: “very good location” may also appear in entity B’s reviews!

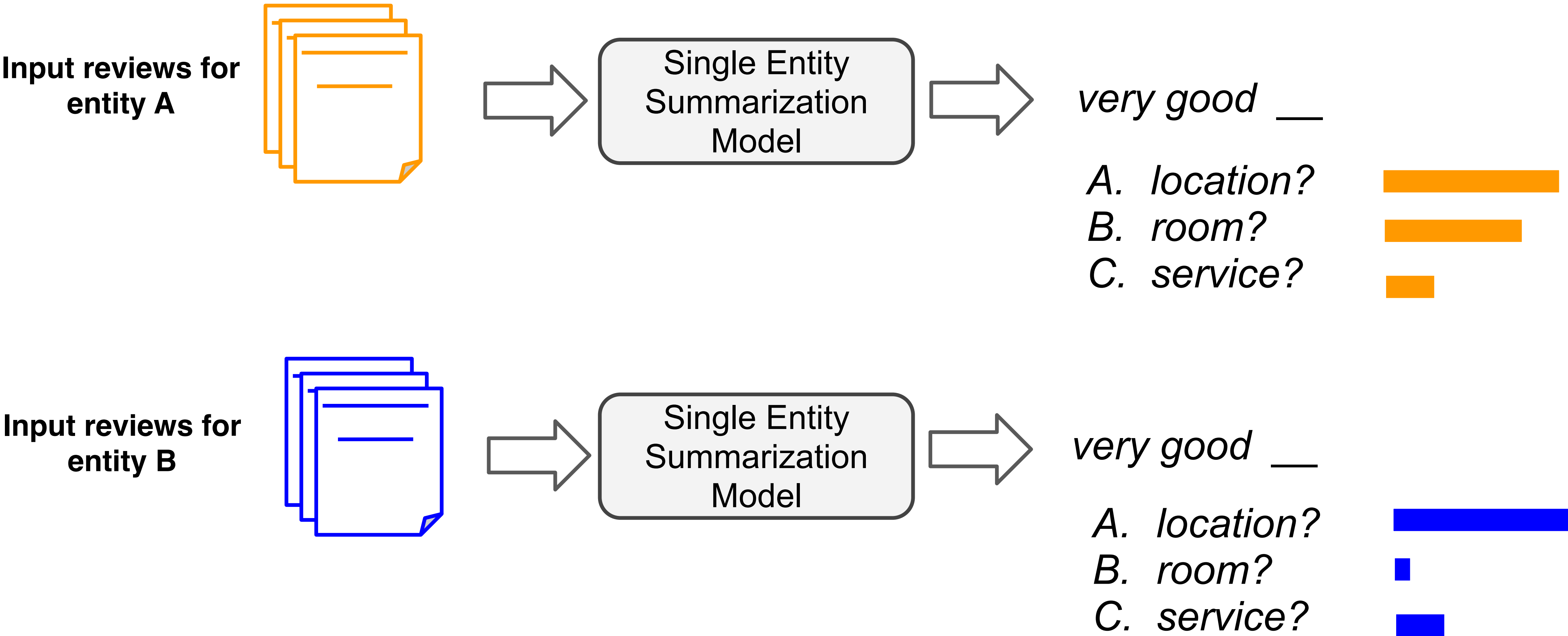
Use Collaborative-Decoding for summary generation



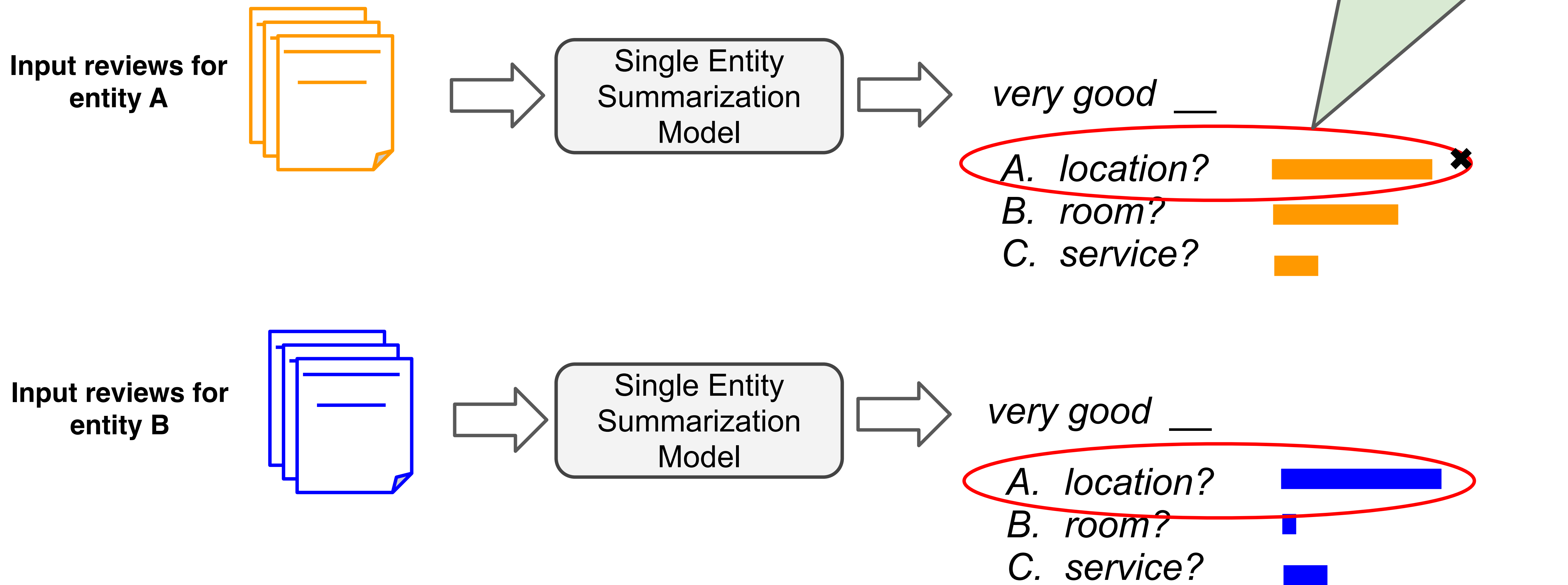
How do we know if “very good location” also appear in entity B or not?

Problem: “very good location” may also appear in entity B’s reviews!

Use Collaborative-Decoding for summary generation



Use Collaborative-Decoding for sum



Use Collaborative-Decoding for summary generation

very good __

A. location?



B. room?



C. service?



very good __

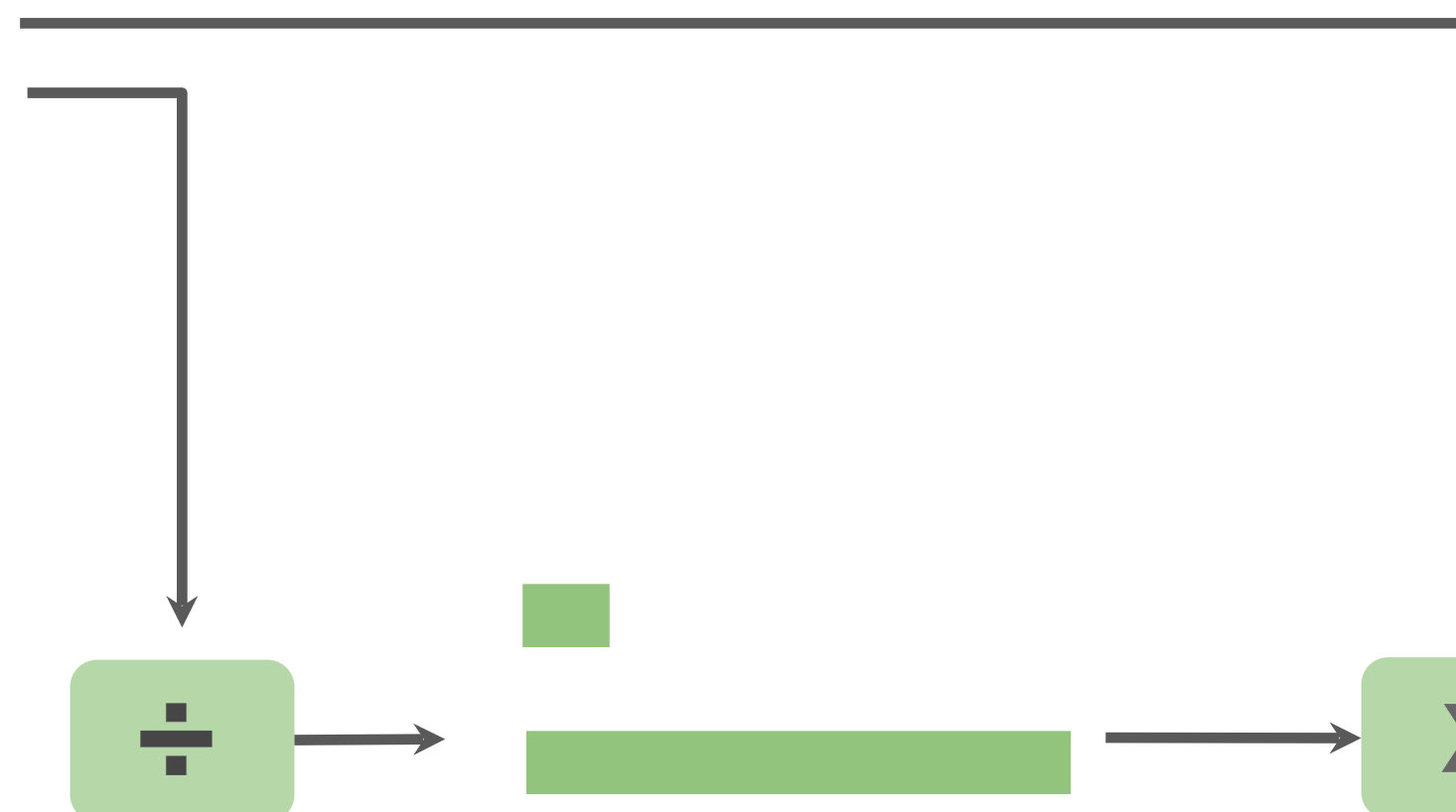
A. location?



B. room?



C. service?



A. location?



B. room?



C. service?



Use Collaborative-Decoding for summary generation


- Encourage models to produce contrastive summaries via the decoding step.

Less bad content

The hotel was available with a deal via the hotel , but there were some issues with the elevator and lines were a bit plain. Overall this is a perfect hotel for solo stays in Rome and not far from Campiano Airport. The rooms in the hotel are not huge but comfortable and clean. The bathrooms are gorgeous and the rooms make the day extra special. The hotel upgraded rooms to have Boschari toiletries on the bed each day. The elevator was a bit plain and the lines were too lines. The hotel staff are always courteous and helpful. Every member of staff have loads of great advice and recommendations for local attractions and sight-seeing. The hotel provides a good size buffet and on roof top garden you can enjoy a nice shower.

More good content

Automatic Evaluation on CoCoTrip

	BERTScore		Distinct. Score
	Contrastive	Common	
LexRank _{BERT}	20.51	24.83	65.56
OpinionDigest [2]	21.77	17.06	64.87
BiMeanVAE [1]	24.33	50.98	42.55
 CoCoSum	35.63	61.52	74.02
w/o Co-decoding	35.42	59.32	71.69

Dataset & Code are available!

<https://github.com/megagonlabs/cocosum>

Takeaways

- Autoencoder is a great model to build an opinion summarizer but keep in mind to choose a “good” summary vector
- Collaborative-Decoding is an inference time decoding algorithm that encourages generating more distinctive summaries