

Navigating Inclusive Tourism: An AI-Powered Guide for Non-Japanese Tourists



BACKGROUND

In 2023, the number of inbound visitors traveling to Japan amounted to approximately 25.07 million, recovering considerably compared to the previous year. However, there are several challenges for foreign tourists when they travel in Japan

Majority of tourists don't understand Japanese

Food ingredients might be different from tourists' country

➤ **Ingredient Confusion**
Unfamiliar ingredients make it hard to identify dietary restrictions

➤ **Allergen Alerts**
Identifying food allergies and intolerances is challenging

➤ **Hard to comply diet restriction based on religions**

Halal for Muslims	Kosher for Jews	Vegetarian for Hindus
No pork on non halal meat products, alcohol	Avoid pork	No meat, dairy, or animal products

AI Utilization | Food Classifier

1. User profiling

Users input their diet restriction and preferences

2. Capture Image

Snap a photo of the ingredient's label

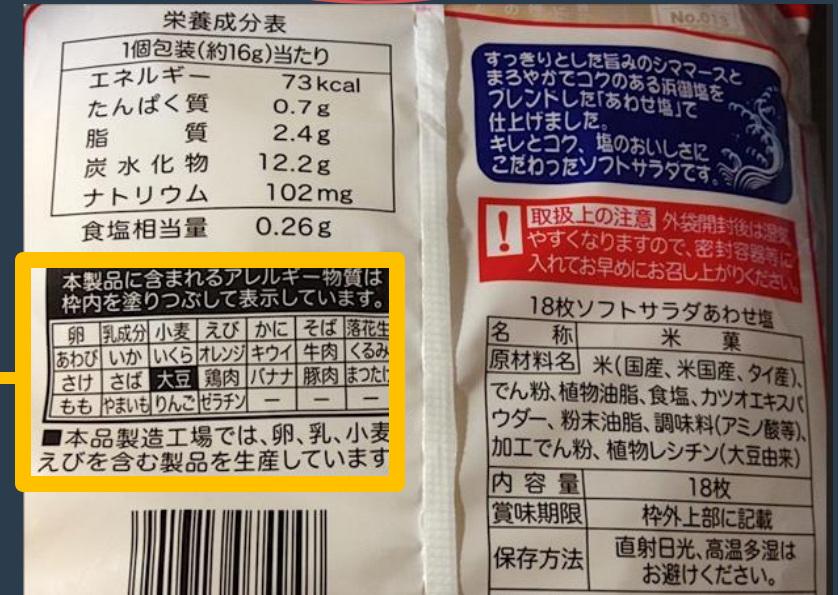
3. Image inference

Using multimodal-multilanguage LLM, we will analyze the captured image and the user profile information

4. Result

LLM will inform the users in English or their preferred language if they can eat the food or not

Example:



Ingredients

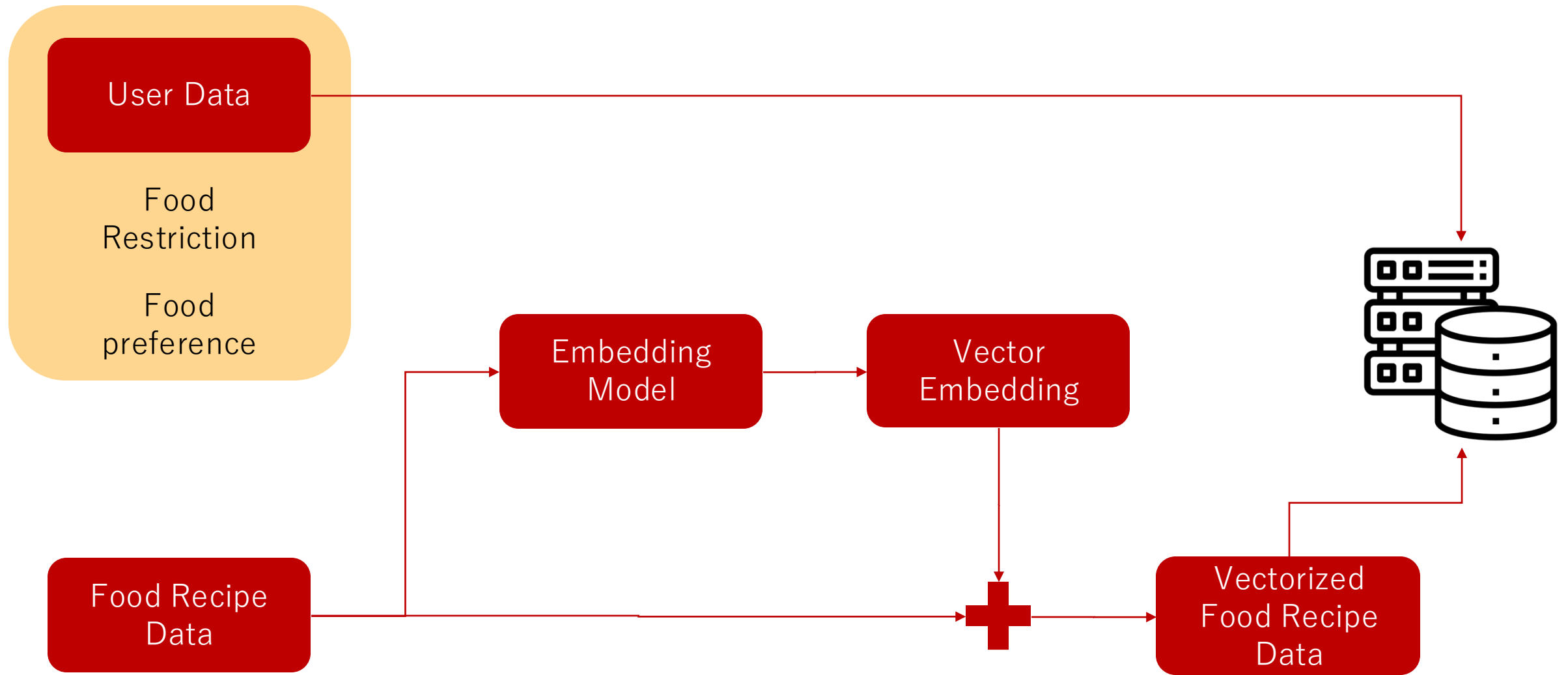
egg milk ingredients
wheat
Shrimp
Crab
Soba
peanut
Abalone Salmon
squid
orange
kiwi
Beef
Walnut
Mackerel

soybeans
chicken meat
banana
pork
Matsutake
Peaches
Yam
apple
gelatin

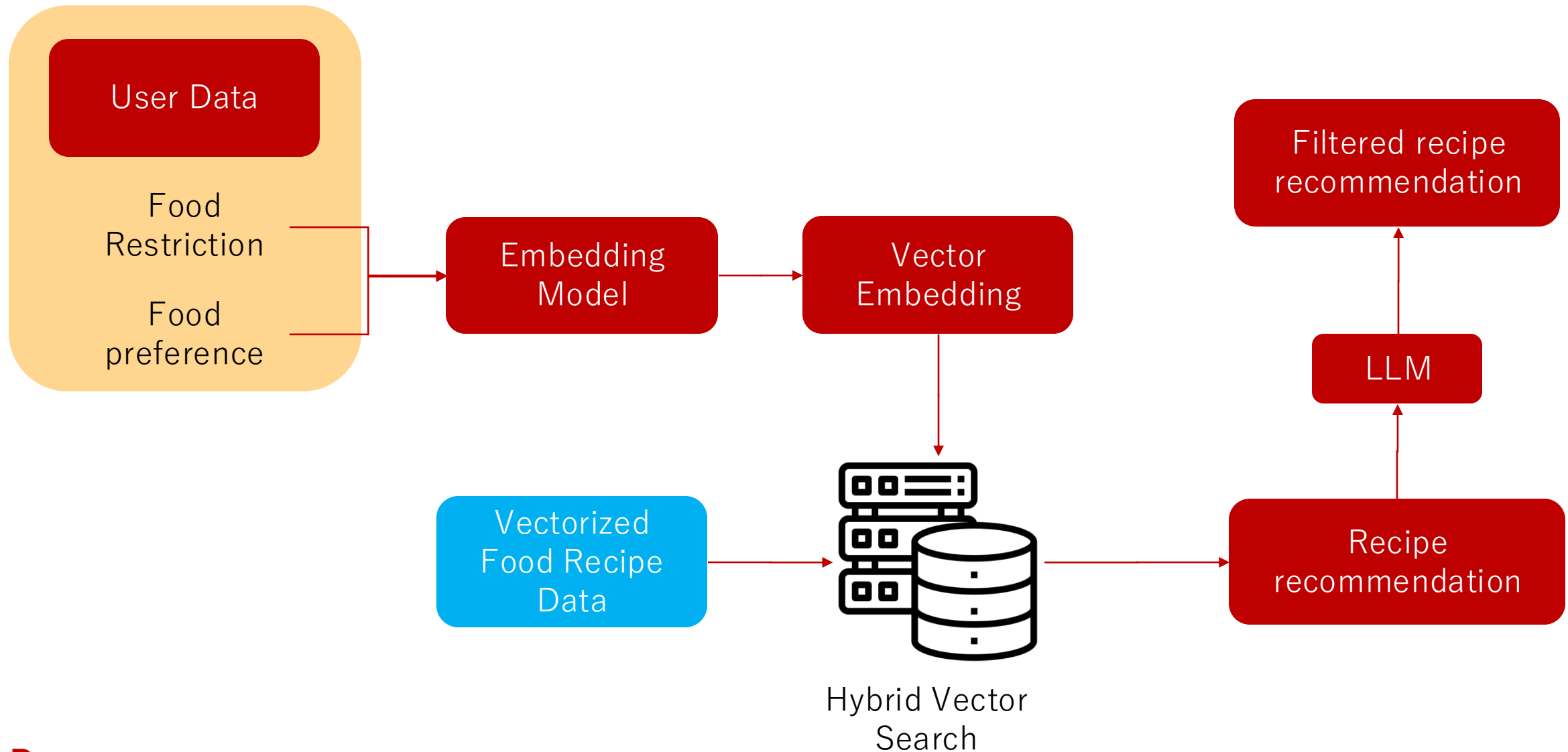
■ This product manufacturing factory produces products containing eggs, milk, wheat, and shrimp.

✓ This product contains pork, gelatin and dairy product, so muslim cannot eat it

AI-Utilization | Recommender System | Data Collection



AI-Utilization | Recommender System | Inference



Data Collection Methodology

Open source data:

```
@misc{shuyang_li_2019, title={Food.com Recipes and Interactions},
url={https://www.kaggle.com/dsv/783630},
DOI={10.34740/KAGGLE/DSV/783630}, publisher={Kaggle},
author={Shuyang Li}, year={2019} }
```

```
recipe_data.head()
```

✓ 0.0s

Python

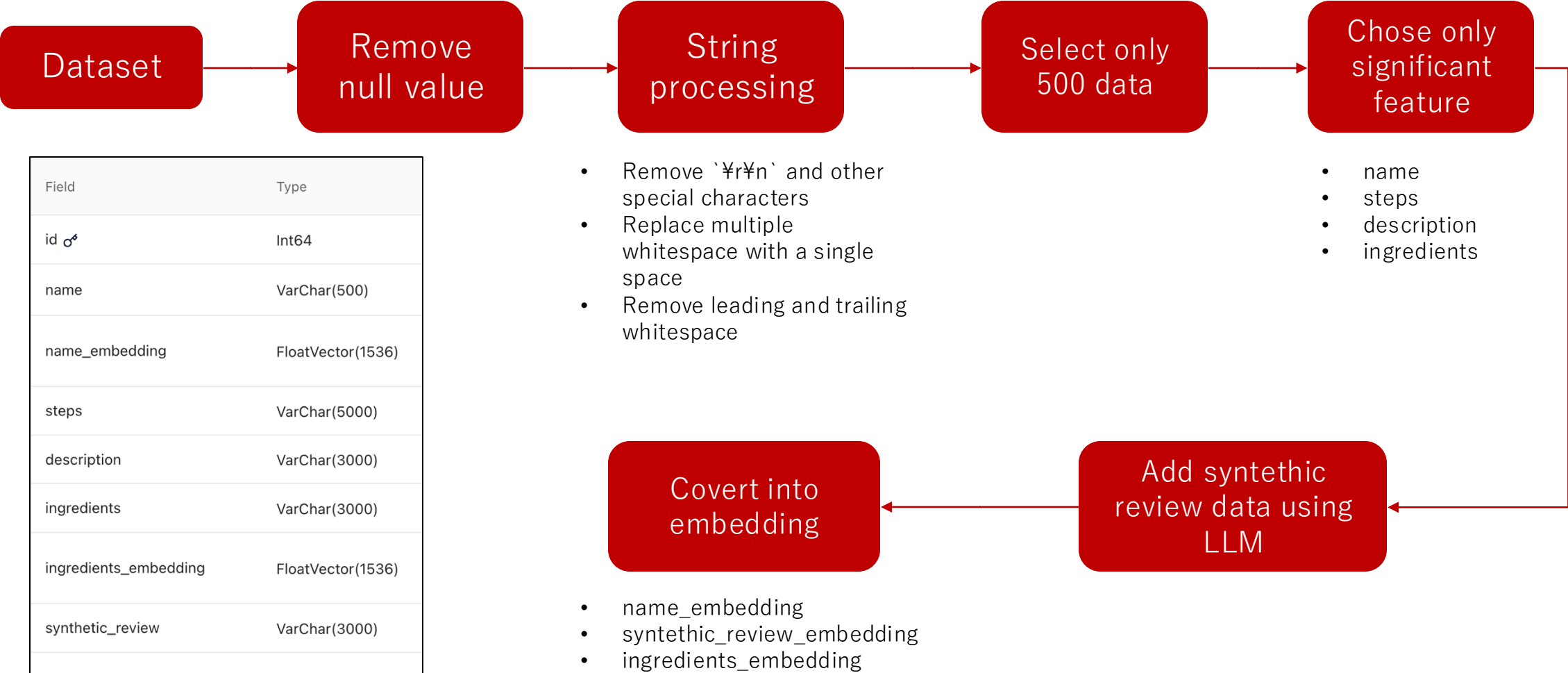
	name	id	minutes	contributor_id	submitted	tags	nutrition	n_steps	steps	description	ingredients	n_ingredients
0	arriba baked winter squash mexican style	137739	55	47892	2005-09-16	['60-minutes-or-less', 'time-to-make', 'course...']	[51.5, 0.0, 13.0, 0.0, 2.0, 0.0, 4.0]	11	['make a choice and proceed with recipe', 'dep...']	autumn is my favorite time of year to cook! th...	['winter squash', 'mexican seasoning', 'mixed ...']	7
1	a bit different breakfast pizza	31490	30	26278	2002-06-17	['30-minutes-or-less', 'time-to-make', 'course...']	[173.4, 18.0, 0.0, 17.0, 22.0, 35.0, 1.0]	9	['preheat oven to 425 degrees f', 'press dough...']	this recipe calls for the crust to be prebaked...	['prepared pizza crust', 'sausage patty', 'egg...']	6
2	all in the kitchen chili	112140	130	196586	2005-02-25	['time-to-make', 'course', 'preparation', 'mai...']	[269.8, 22.0, 32.0, 48.0, 39.0, 27.0, 5.0]	6	['brown ground beef in large pot', 'add choppe...']	this modified version of 'mom's' chili was a h...	['ground beef', 'yellow onions', 'diced tomato...']	13

```
recipe_data.info()
```

✓ 0.0s

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 231637 entries, 0 to 231636
Data columns (total 12 columns):
Column Non-Null Count Dtype
--- -
0 name 231636 non-null object
1 id 231637 non-null int64
2 minutes 231637 non-null int64
3 contributor_id 231637 non-null int64
4 submitted 231637 non-null object
5 tags 231637 non-null object
6 nutrition 231637 non-null object
7 n_steps 231637 non-null int64
8 steps 231637 non-null object
9 description 226658 non-null object
10 ingredients 231637 non-null object
11 n_ingredients 231637 non-null int64
dtypes: int64(5), object(7)

Data Collection Methodology: Data Prerocessing

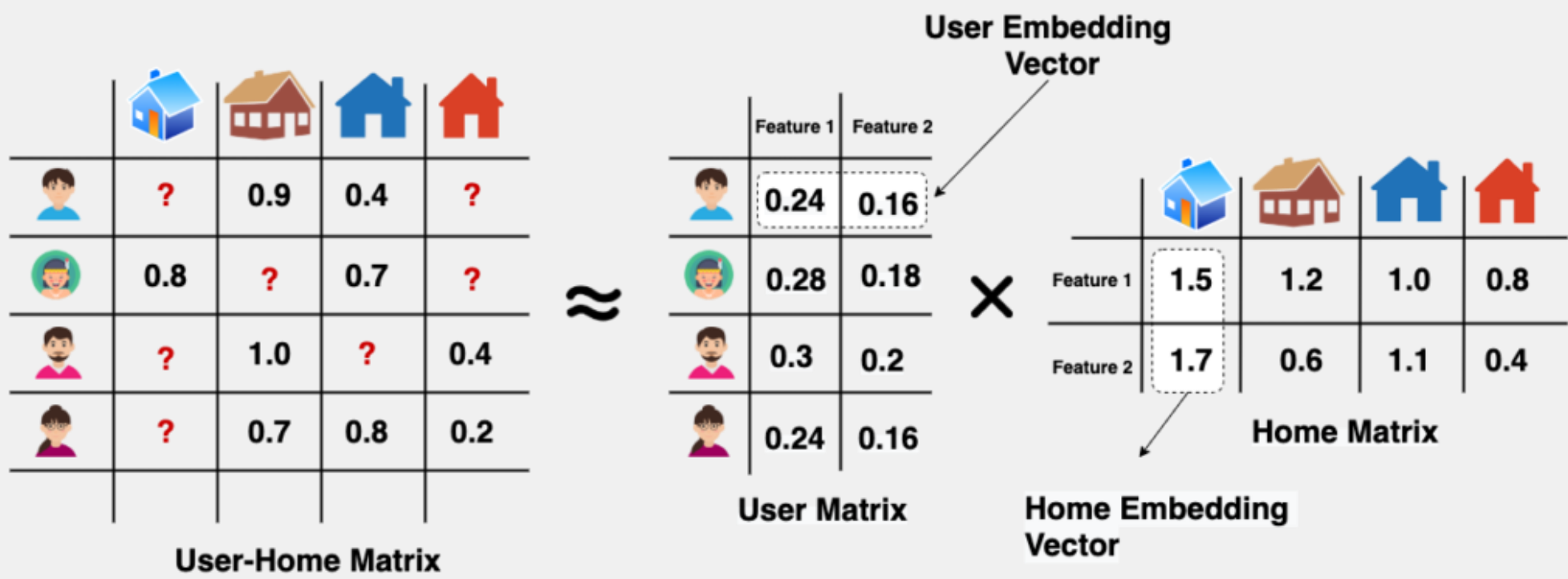


Further improvement point: Combine Neural Collaborative Filtering with Current System

Our approach: content based filtering -> uses item features to recommend other items similar to what the user likes

CF: In a broad sense, it is the process of filtering for information or patterns using techniques involving collaboration among multiple users, agents, and data sources.

Matrix Factorization

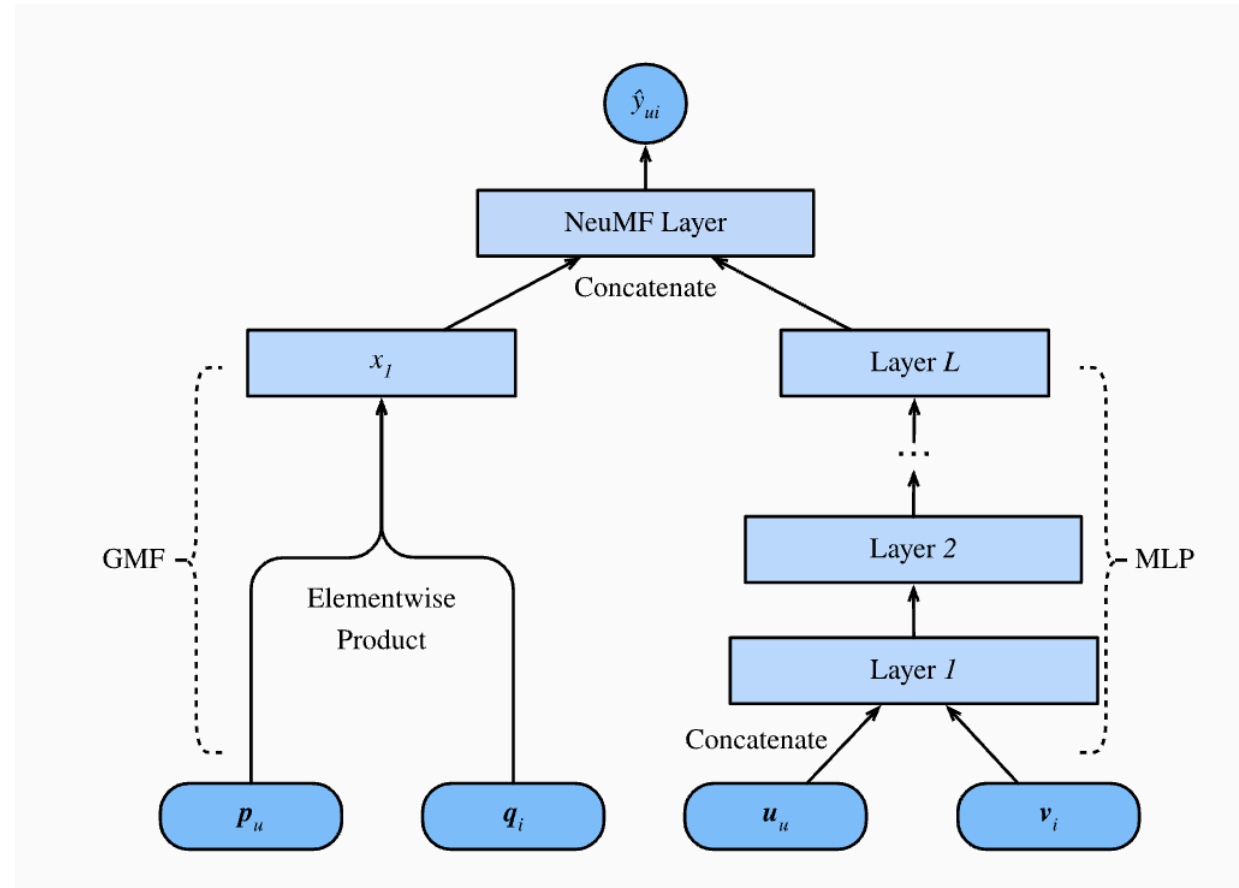


Further improvement point: Combine Neural Collaborative Filtering with Current System

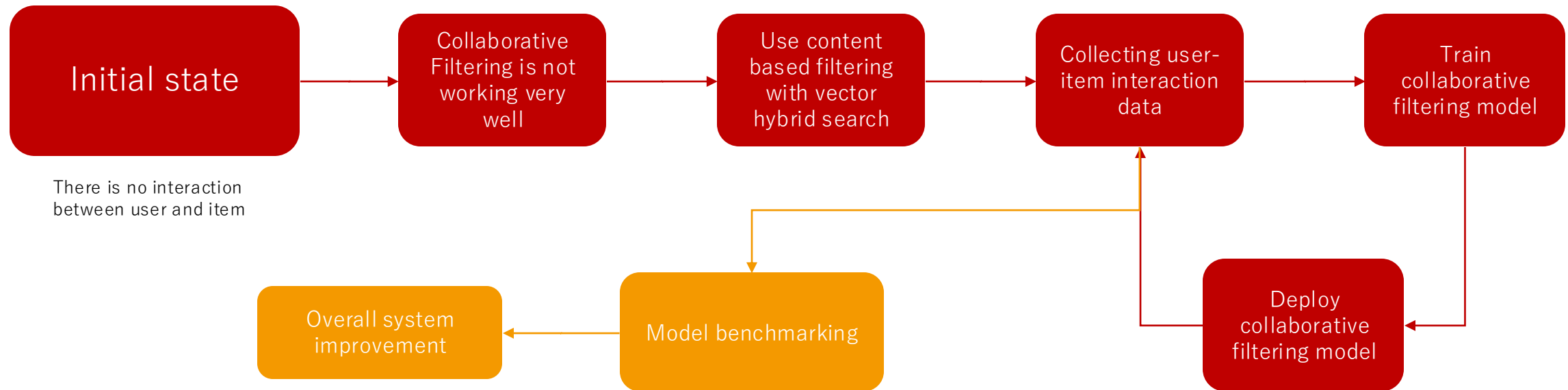
Neural Collaborative Filtering: Combination of matrix factorization and neural network

Matrix factorization
deficiency:

- Lack of expressiveness:
The inner product, which simply combines the multiplication of latent features linearly, may not be sufficient to capture the complex structure of user interaction data.
- Only use explicit feedback
->cannot incorporate implicit feedback



Further improvement point: Combine Neural Collaborative Filtering with Current System



References

1. <https://medium.com/@rebirth4vali/implementing-matrix-factorization-technique-for-recommender-systems-from-scratch-7828c9166d3c>
2. Dive deep into deep learning books
3. Icon: flaticon.com