



Embedding the World of Food in Our Minds : Constructing a Food Schema



Hyunsu Jung, and Inhwa Oh, School of Data Science, Hanyang University
Hyungwook Yim



Research Background

Why Do we Need to study Food Schema?

- People hold different food schema for the same food, and studies suggest that these differences may relate to eating habits and weight management.

Free-association Task

Write down **all** the words that come into your mind when reading the word below.

cue : Kimchi

response :

1. Rice
2. Egg
3. ...

Prior Work

- Responses were manually labeled into five semantic categories.
 - Cue one-hot vectors + category proportions were used to train a normal vs. obese classifier.
- Indicates group-level differences in food schema.

Limitation of Prior Work

- Lack of Domain-Specific Embeddings**
Korean food-specific terms (e.g., *pajeorim*, *chonggak-kimchi*) were not covered by existing embeddings.
- Manual Category Tagging**
Labor-intensive and not scalable.

Our Approach & Why It Matters

Food Domain corpus construction

- Food-related blog texts : contain rich relationships between foods & daily contexts (places, sensations, situations).

Word Embedding

- Train embeddings for Korean food/brand/place that were not covered in general models.

Automatic Category Prediction

- Propose an automated tagging pipeline suitable for large datasets.

Embedding based Normal/Obese Group Prediction

- Pilot test of whether embedding-based representations yield improved predictive performance.

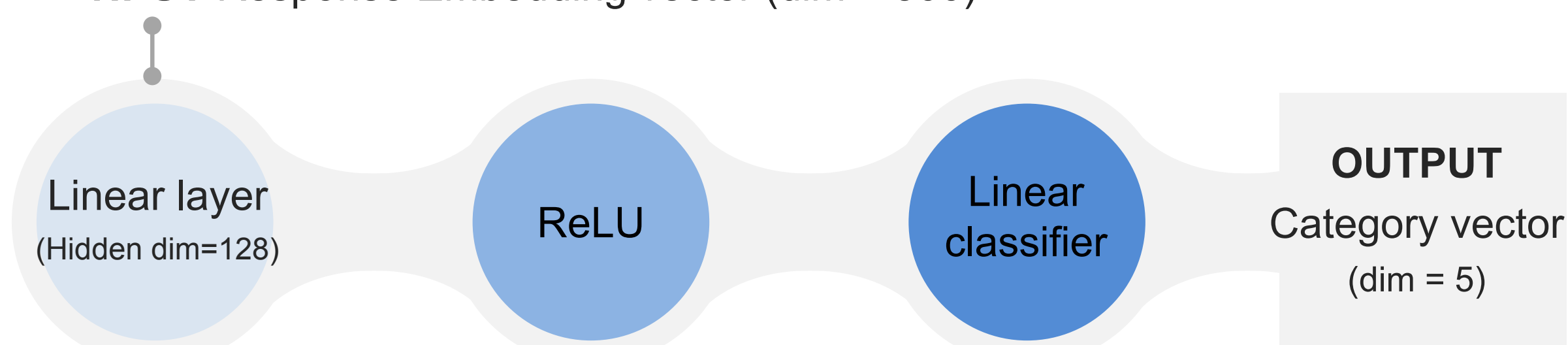
Methods

Corpus & Word Embedding

- Corpus Construction**
Crawling Food-Related Blog Texts
Search keyword : 135 cue, 1,632 response / Total documents : 424,080
- Embedding Training**
Word2Vec
Captures semantic relations in food-related contexts
Output: 300-dimensional vectors
- Embedding Quality Check**
Hierarchical Clustering
MDS Visualization
Similar Words

Category Predict Model

INPUT Response Embedding vector (dim = 300)



Normal/Obese Group Predict Model

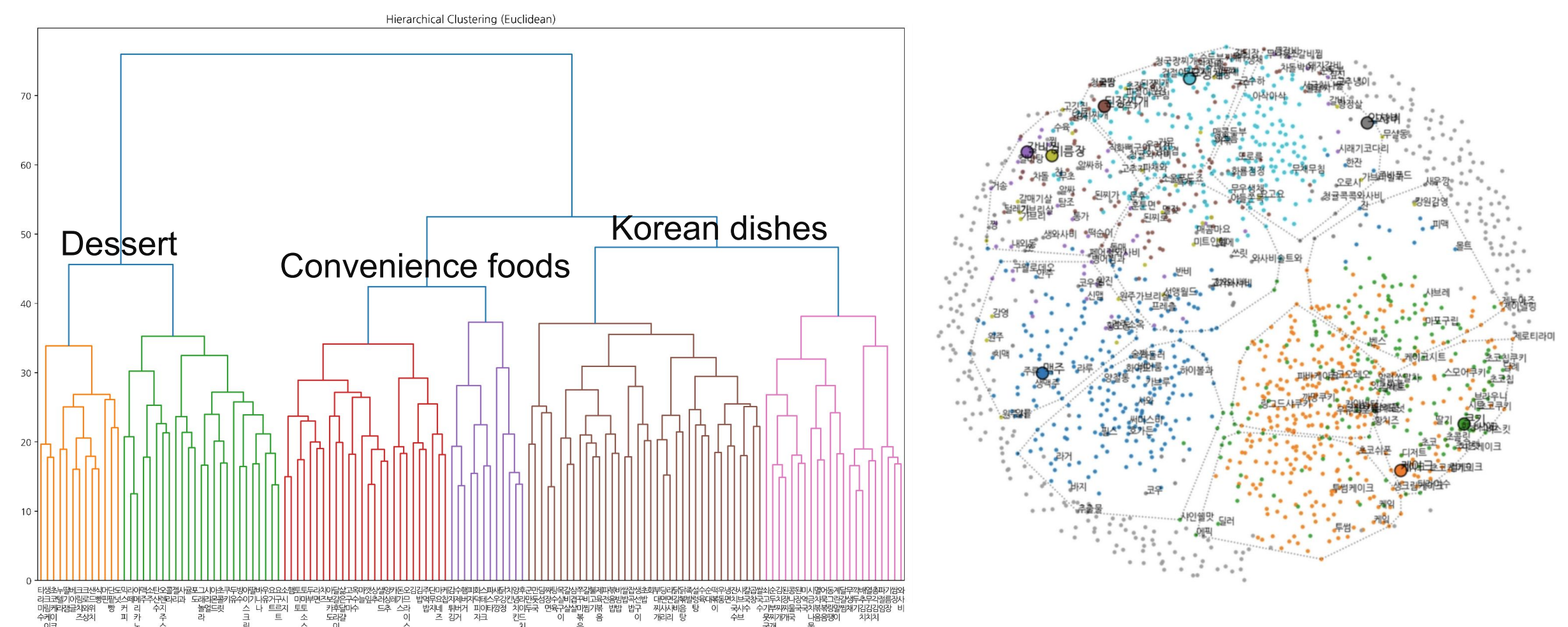
INPUT : Cue+Response Embedding vector (sequence length = 2)

Model Architecture : RNN (300-128-1)

OUTPUT : normal(0) / obese(1) group

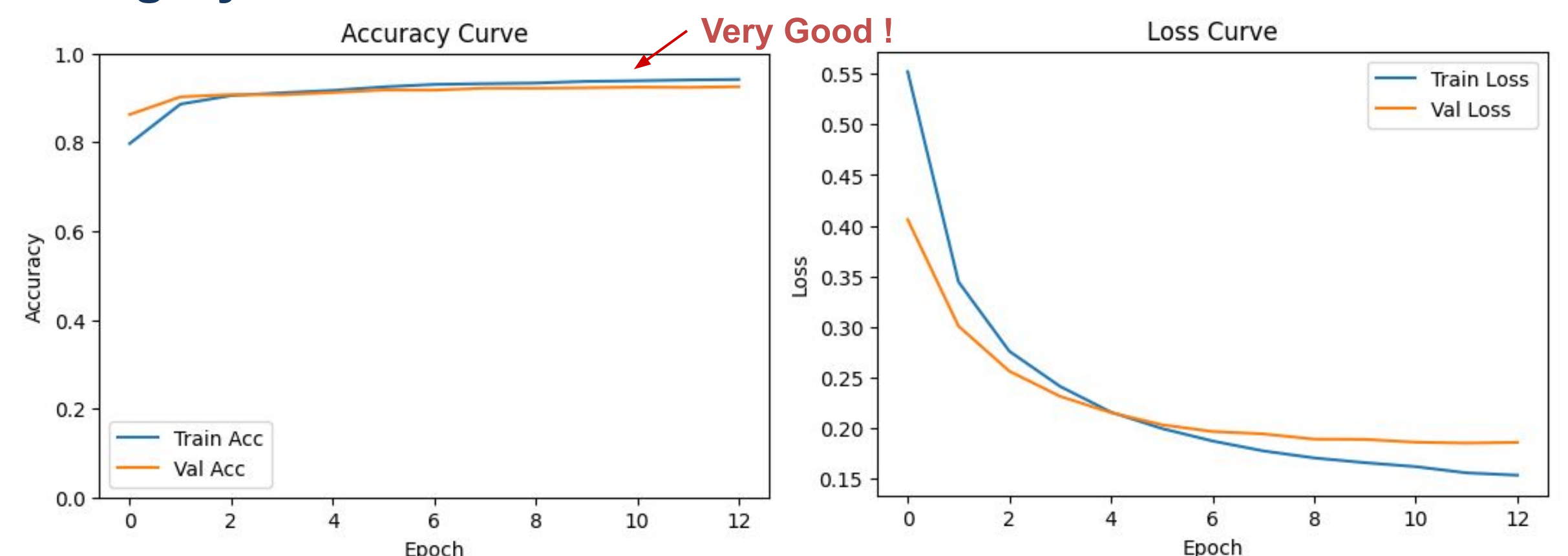
Results

Corpus Construction



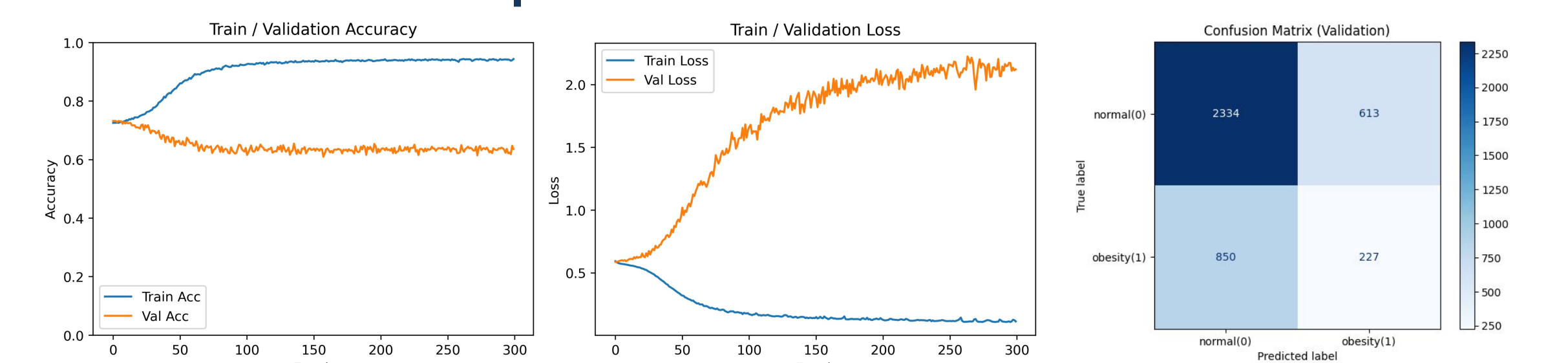
- Hierarchical clustering food cue embeddings**
Meaningful groupings emerged
- MDS projection of Cue word**
Cue words formed clusters with semantically related terms (foods, places, store names, etc.).

Category Predict Model



- MLP Category Classifier**
e.g
Input: embedding vector of "rice" (300-dim)
Output: 10000 (5dim) = food word category

Normal/Obese Group Predict Model



- RNN normal/obese group classifier result

DISCUSSION

Interpretation & Contribution

- Establishing a Basis for Representing Food Schema in Vector Space**
→ Provides a reusable infrastructure for future food-schema research.
- Automated Category-Tagging Pipeline**
High prediction accuracy demonstrates that the embedding effectively captures association-based meaning structure.
→ Suggests the potential to replace manual tagging in large-scale free-association datasets.

Future Directions

- Potential for Normal vs. Obese Group Classification**
Validation performance fluctuated due to individual differences.
→ More participants would allow clearer group patterns and better classification performance.
- Corpus Expansion**
Expand the corpus with SNS posts and reviews
→ Enables more refined food-schema representations.