



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

CCD Computational Cognition & Development Lab

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. ...

Write here

Add answer Next

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

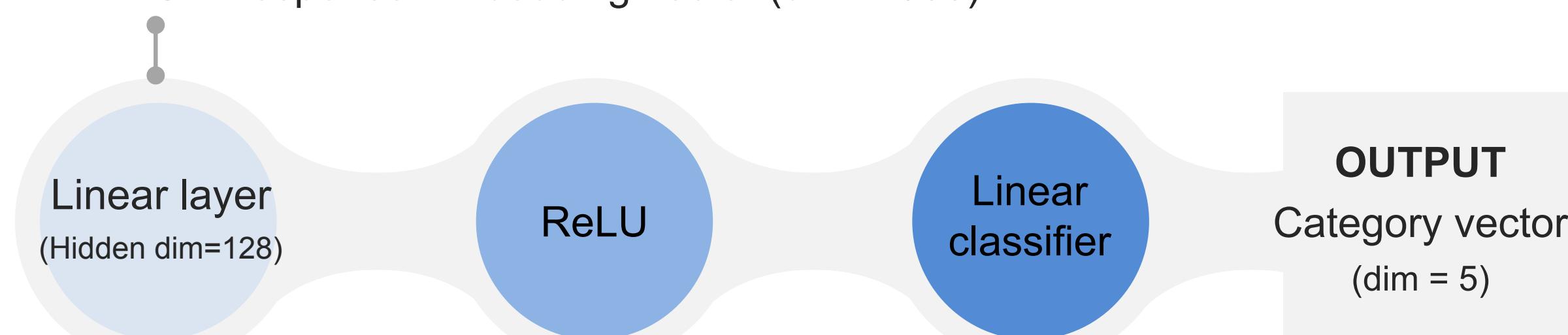
Hierarchical Clustering

MDS Visualization

Similar Words

Category Predict Model

INPUT Response Embedding vector (dim = 300)



Normal/Obese Group Predict Model

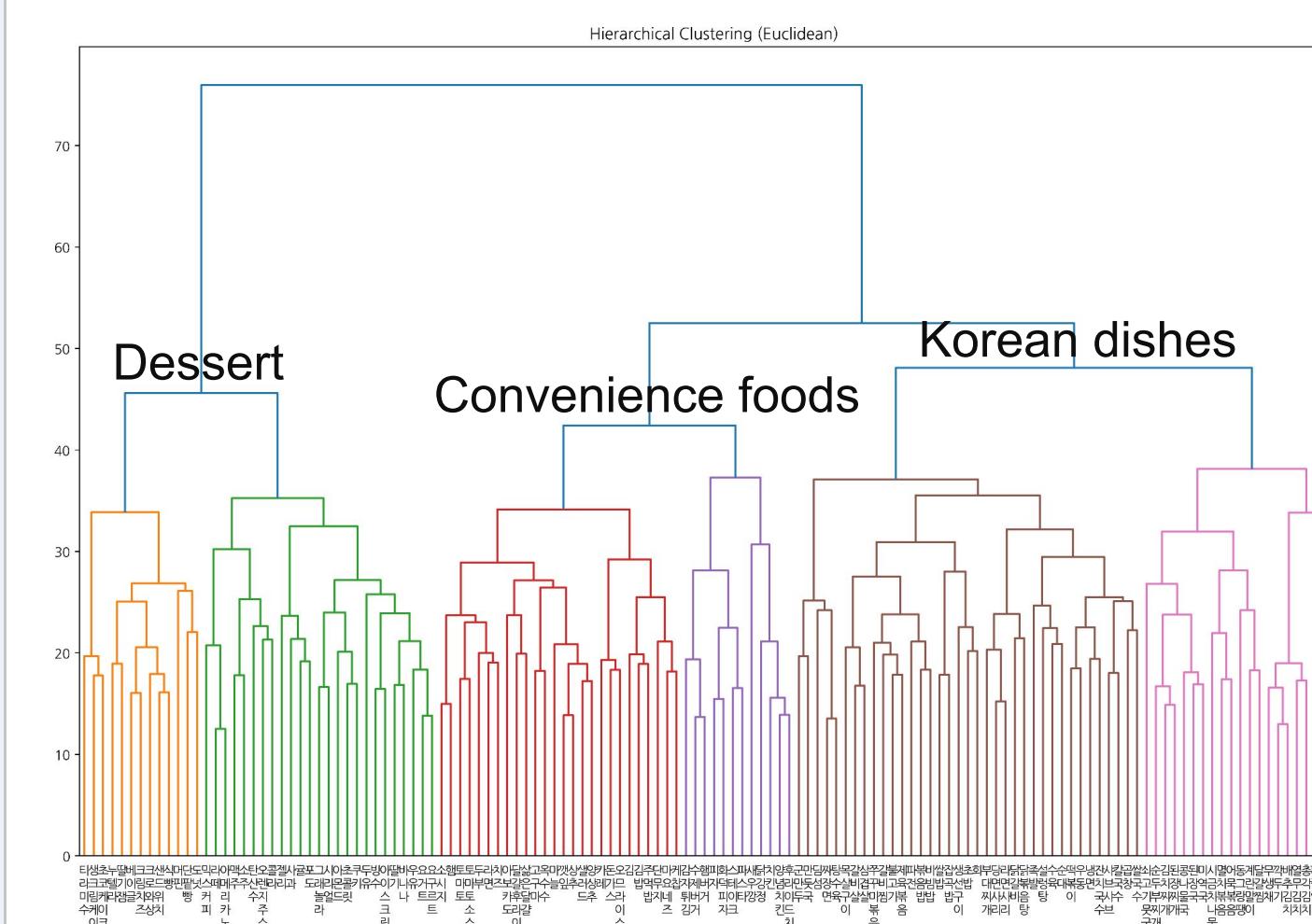
INPUT : Cue+Response Embedding vector (sequence length = 2)

Model Architecture : RNN (300-128-1)

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

Results

Corpus Construction



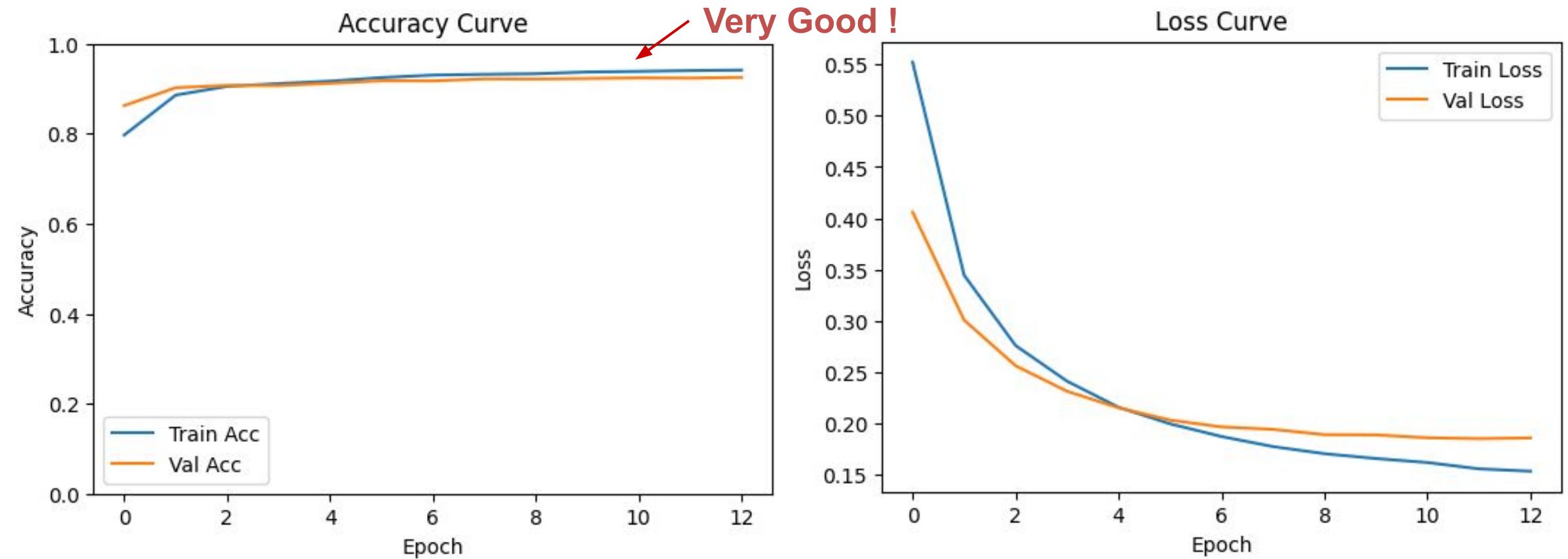
Hierarchical clustering food cue embeddings

Meaningful groupings emerged

MDS projection of Cue word

Cue words formed clusters with semantically related terms (foods, places, store names, etc.).

Category Predict Model



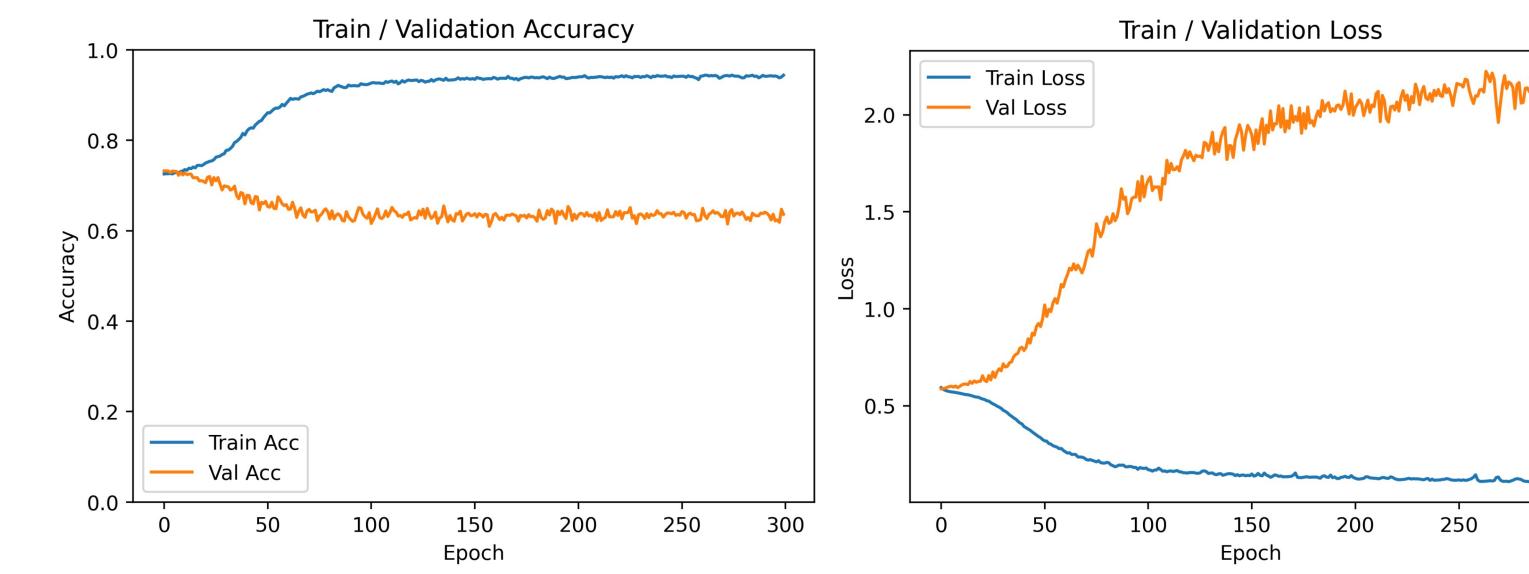
MLP Category Classifier

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.