

Polysemy interpretation by using similarity based estimation

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The Linguistic Society of Korea
한국언어학회

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Personal Profile



Seongmin Mun

Experience & Education

PostDoc – Chosun University

- NLP
- Web / server development
- Deep learning
- Image processing

Ph.D. – Université Paris Nanterre

- NLP
- Data visualization
- Neural network
- Linguistics
- Statistics
- Language models
- Machine learning
- Web-based system

M.S. – Ajou University

- Data visualization
- Machine learning
- Web-based system
- Statistics

<https://seongminmun.com/>



Seongmin Mun

Skills & Endorsements

Research Knowledge

- NLP
- Linguistics
- Data Visualization
- Data analysis
- Machine Learning
- Web Development

Computer Language

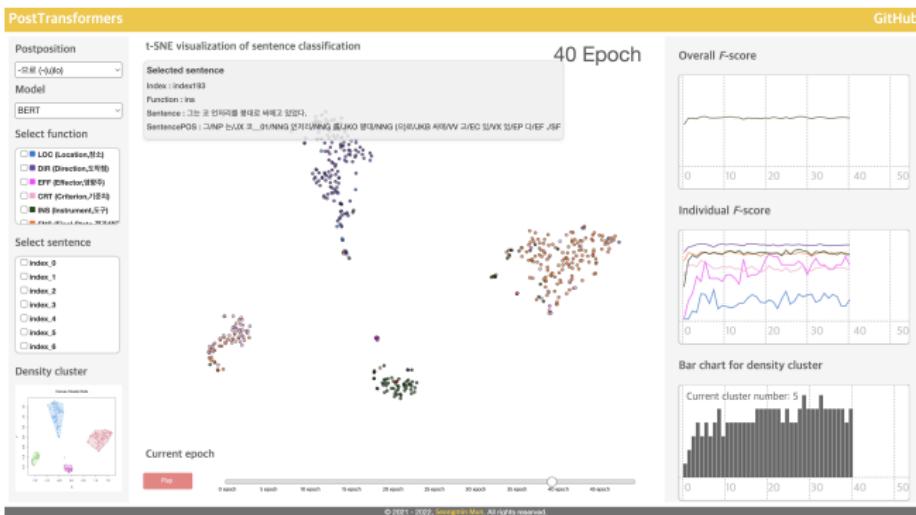
- Java
- JavaScript
- HTML/CSS
- Python
- SQL
- PHP
- R

Statistics Software

- R
- MATLAB
- SAS
- SPSS

<https://seongminmun.com/>

Mun, 2021

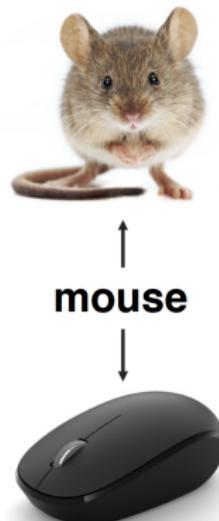


Mun, S. (2021). Polysemy resolution with word embedding models and data visualization: the case of adverbial postpositions -ey, -eyse, and -(u)lo in Korean. presented at IMPRS2020 (MaxPlanck), ICCG11, and ACL 2022

Introduction

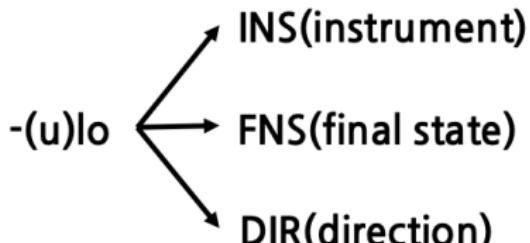
Polysemy

Polysemy, one type of ambiguity, occurs when one form delivers multiple meanings/functions (Glynn and Robinson, 2014).



Korean language

Korean is a Subject-Object-Verb language, which marks grammatical information with dedicated postpositions (Sohn, 1999).



Polysemy in Korean

Polysemy in Korean adverbial postposition

-(u)lo as INS (instrument)

na-nun kamca-lul khal-lo ssel-ess-ta.

I-TOP potato-ACC knife-INS cut-PST-DECL

‘I cut a potato with a knife.’

Figure: An example sentence involving the postposition -(u)lo as a function of INS (instrument)

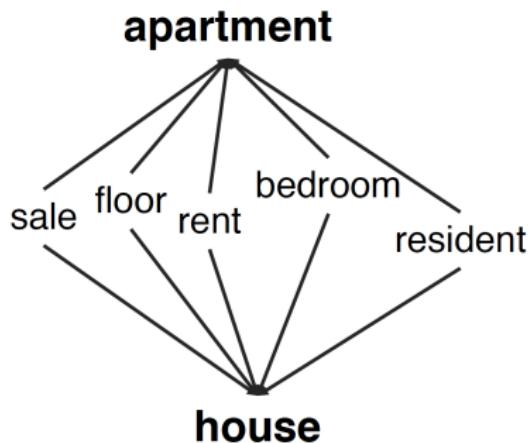
Polysemy in Korean

Question: How a speaker can understand the function of postposition?

Distributional semantic models (DSMs)

Concept of DSMs

The concept of distributional semantic models (DSMs) is that **a word meaning is closely tied to a context** that is created by a group of neighborhood words, dubbed the distributional hypothesis (Firth, 1957; Harris, 1954).



Distributional semantic models (DSMs)

Context window

A range of words surrounding a target word, affecting the determination of its characteristics (Mun, 2021)

Window1

밭/NNG 에서/JKB 채소/NNG (으)로_INS/JKB 가꾸/VV 다/EF
슬로/NNG 모션/NNG (으)로_FNS/JKB 보이_01/VV 다/EF

Window2

밭/NNG 에서/JKB 채소/NNG (으)로_INS/JKB 가꾸/VV 다/EF
슬로/NNG 모션/NNG (으)로_FNS/JKB 보이_01/VV 다/EF

Word-level embedding model

- ▶ Model training: Positive Pointwise Mutual Information (PPMI; Church and Hanks, 1989) and Singular Value Decomposition (SVD; Eckart and Young, 1936).
- ▶ Classification model: similarity-based estimate (Dagan et al., 1993) by calculating cosine similarity scores between -(u) lo and its co-occurring content words.

Methods

Corpus: Adverbial Postpositions In Korean (APIK)

- ▶ Sejong corpus, with semantic annotations of three adverbial postpositions -ey, -eyse, and -(u)lo cross-verified by three native speakers of Korean (Mun & Desagulier, 2022)
- ▶ Available at:
<https://github.com/seongmin-mun/Corpora/tree/main/APIK>

Corpus: Adverbial Postpositions In Korean (APIK)

Index ### Label ### Function ### Sentence_POS ### Sentence

- 1 ### 0 ### FNS ### 이_05/MM 넥타이/NNG 는/JX 수제품/NNG (으)로/JKB 우리나라/NNG 에서/JKB 는/J;
- 2 ### 2 ### DIR ### 나/NP 의/JKG 마음_01/NNG 의/JKG 움직임/NNG 이/JKS 위_01/NNG 에서부터/JKE
- 3 ### 1 ### INS ### 굿/NNG 무당_01/NNG 이/JKS 노래/NNG 나/JC 춤_01/NNG (으)로/JKB 귀신_01,
- 4 ### 0 ### FNS ### 모든/MM 주장_03/NNG 이/JKS 나름/NNB 대로/JKB 의/JKG 근거/NNG 를/JKO 갖추/
- 5 ### 3 ### EFF ### 기억/NNG 이/JKS 스스로/NNG 의/JKG 부력_01/NNG (으)로/JKB 떠오르/VV 았/EP ·
- 6 ### 2 ### DIR ### 신축_03/NNG 전원주택/NNG 위쪽/NNG (으)로/JKB 는/JX 집_01/NNG 이/JKS 없/·
- 7 ### 0 ### FNS ### 명명/XR 하/XSA ㄴ-/ETM 채_09/NNB (으)로/JKB 시간_04/NNG 이/JKS 흘러가/VV
- 8 ### 1 ### INS ### 수한/NNP 이/JKS 저/NP 의/JKG 손_01/NNG (으)로/JKB 저/NP 의/JKG 가슴_01,
- 9 ### 2 ### DIR ### 쇠전_01/NNG 꾼/XSN 들/XSN 이/JKS 술청/NNG (으)로/JKB 돌아오/VV 았/EP 다/E
- 10 ### 3 ### EFF ### 그리고/MAJ 그/MM 결과_02/NNG (으)로/JKB 오줌/NNG 이/JKS 나오/VV ㄴ다/EF
- 11 ### 5 ### LOC ### "/SS 집_01/NNG 들/XSN 이/JKS 다/MAG 어디/NP (으)로/JKB 가/VV 았/EP 나도
- 12 ### 5 ### LOC ### 바로/MAG 앞/NNG (으)로/JKB 소달구지/NNG 바퀴_01/NNG 자국_01/NNG 이/JKS

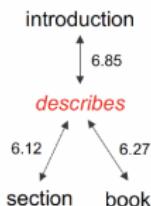
Corpus: Adverbial Postpositions In Korean (APIK)

밭/NNG 에서/JKB 채소/NNG (으)로_{_}INS/JKB 가꾸/MV 다/EF
슬로/NNG 모션/NNG (으)로_{_}FNS/JKB 보이_01/VV 다/EF
우리/NP 그만/MAG 포항/NNP (으)로_{_}DIR/JKB 가/VV 자/EF

ACC = accusative case marker; DAT = dative marker; DECL = declarative; EF = final ending; JKB = adverbial case marker; MAG = general adverb; NNG = common noun; NNP = proper noun; NOM = nominative case marker; NP = pronoun; PST = past tense marker; TOP = topic; VV = verb

Similarity-based estimate (Dagan et al., 1993)

Network from the training set



Q: How to calculate the similarity score between '*describes*' and '*chapter* (unknown)?

describes ← → ? *chapter* (unknown)



(w_1, w_2)	$\bar{I}(w_1, w_2)$	$f(w_1, w_2)$	$f(w_1)$	$f(w_2)$
(introduction, describes)	6.85	5	464	277
(book, describes)	6.27	13	1800	277
(section, describes)	6.12	6	923	277
Average:	6.41			

Table 1: The similarity based estimate as an average on similar pairs: $\bar{I}(\text{chapter}, \text{describes}) = 6.41$

Input as a test item

[introduction,
chapter (unknown),
book,
section]

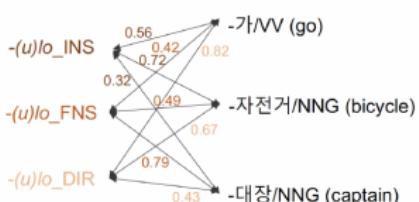


describes ← → **6.41** *chapter* (unknown)

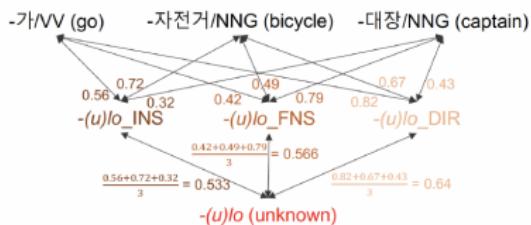
Approach (adapted from Dagan et al., 1993)

Three functions of -(u)lo : INS (instrumental), FNS (final state), DIR (directional)

Network from the training set
(window size: 1; normalized cosine)



Q: Which function is the intended function of -(u)lo?



Input as a test item

[-가/VV (go), -(u)lo (unknown),
-자전거/NNG (bicycle), -대장/NNG (captain)]

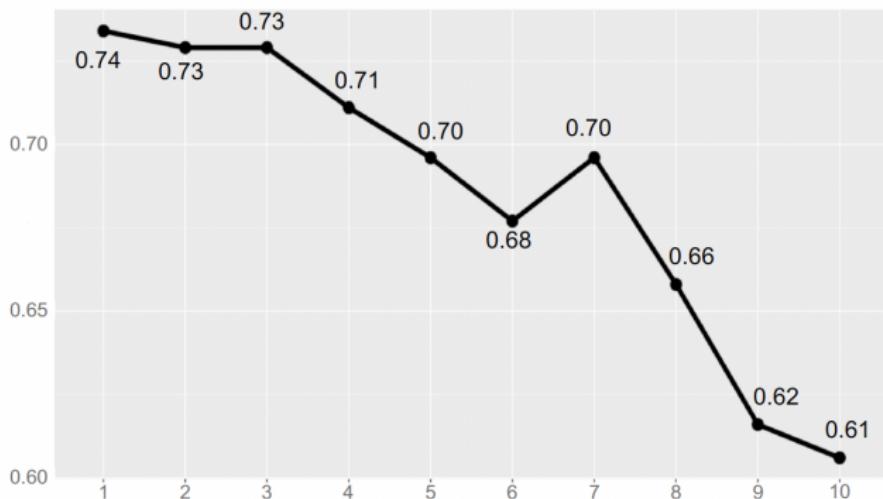
-(u)lo_INS: 0.533

-(u)lo_FNS: 0.566

-(u)lo_DIR: 0.64

Result & Discussion

Result



X-axis: window sizes
Y-axis: accuracy (%)

Our model achieved the highest classification accuracy rate in the window size of one, and the accuracy rates decreased as the window size increased.

Interpretation

- ▶ This trend aligns with advantages of small window sizes (Bullinaria Levy, 2007).
- ▶ Considering that a narrower range of context window relates more to syntactic than to semantic information (Patel et al., 1997), our model may have employed structural, more than semantic, characteristics of tri-grams (word-target-word) for the best classification performance.

Personal Profile



Introduction



Methods



Result & Discussion



Appendix



Appendix

Data processing by using Python

- ▶ Colab: Python code

Web-based System

The screenshot shows a web-based application interface. At the top left is a dark header bar with a three-line menu icon. Below it, the main title "Similarity Based Estimation: -(u)lo" is displayed in blue text. Underneath the title, there is a "Context window size" label followed by a dropdown menu set to "window 1". A "Input Sentence" label is positioned above a large input text area containing the placeholder "Input your sentence ...". At the bottom right of the input area is a blue "Analyze" button.

Thank you for listening.