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Semantic Change in the Lexicon:
Models of Inference
Based on Patterns of Coexpression

**ALT-2026: 16th International Conference of
the Association for Linguistic Typology**

WS6 - Semantic shifts and the dynamics of lexification patterns

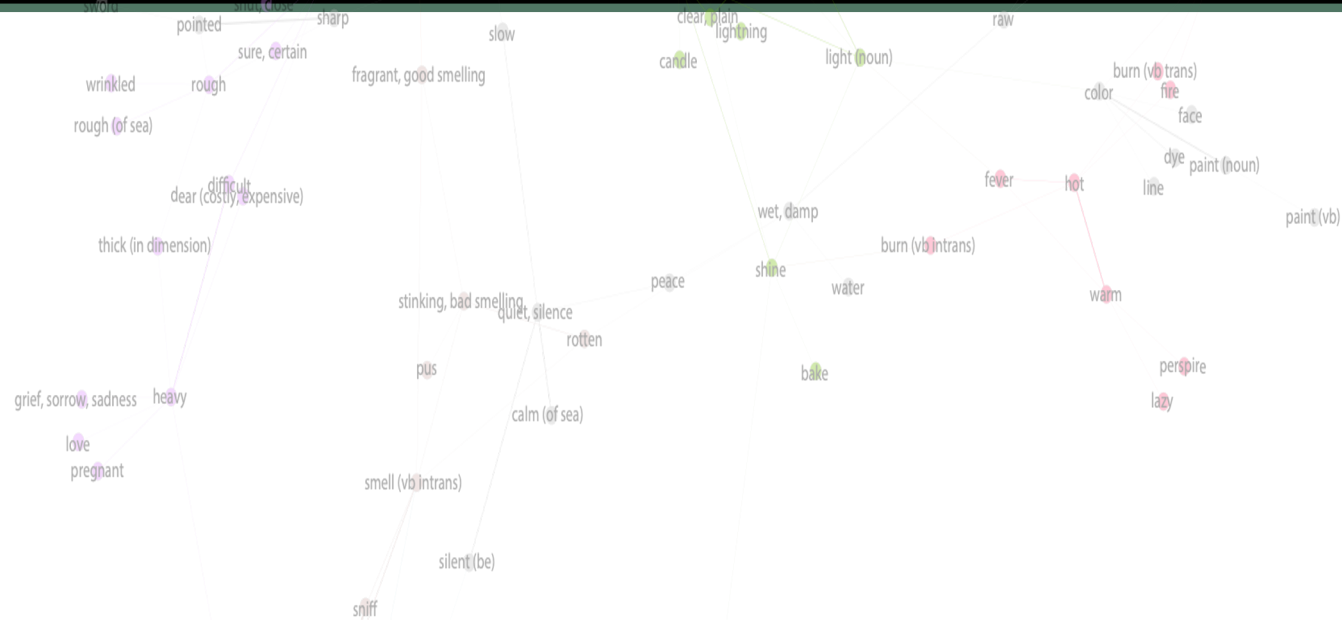
1-3 July 2026 Lyon (France)



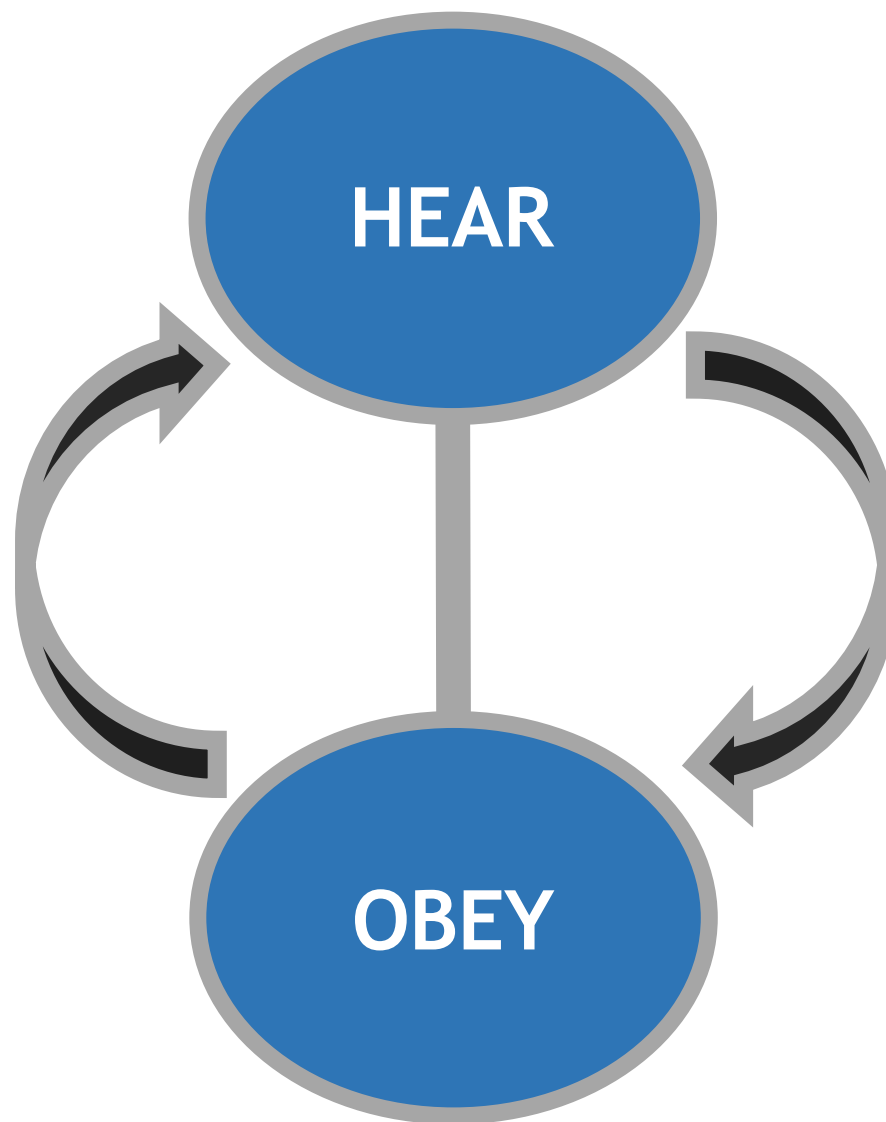
Goal of the talk

**Can semantic change be
predicted from synchronic data?**

Introductory remarks



Introductory remarks



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Introductory remarks

Why is semantic change difficult to model?

- Meanings change in many possible directions:
 - specific → general
 - general → specific
 - positive → negative
 - negative → positive
 - concrete → abstract
 - concrete → concrete
- Semantic change has often appeared:
 - Irregular, unpredictable, less systematic than phonological or morphological change

“by and large, semantic change operates in a rather random fashion, affecting one word here (in one way) and another form there (in another way)” (Hock, 1991)

Introductory remarks

From unpredictability to regularity

- **Research in the last 30 years**
 - Studies have identified regular patterns in semantic evolution (e.g., Sweetser, 1990; Traugott & Dasher, 2002)
- **Semantic change is frequently unidirectional**
 - **Examples:**
 - objective → subjective
 - concrete → abstract
 - lexical → grammatical
- **Mechanisms are increasingly well understood**
 - subjectification
 - metaphor
 - asymmetric priming (Jäger & Rosenbach, 2008)

Bottom line: Semantic change is not entirely random

(see Baumann & Hartmann, 2026; Li & Siew, 2022; List, 2023; Urban, 2011; Winter & Srinivasan, 2022; Xu et al., 2017)

Introductory remarks

- **What we still cannot do well**

We still lack robust predictive models of semantic change

- **What do we need**

1. Reliable predictors
2. Access to large-scale synchronic resources with broad lexical and conceptual coverage for statistical modelling
3. Extensive diachronic datasets documenting historical semantic shifts across languages

Today's presentation

1. Comparing two predictive approaches

a) Approach 1 - Causal Inferences

Relies directly on coexpression patterns, applying causal inference algorithms to derive diachronic trajectories from synchronic distributions (Dellert, 2016; 2019; 2024)

b) Approach 2 - PredictMe

Directionality is predicted from psycholinguistic properties attached to the meanings themselves

(cf. Winter & Srinivasan, 2022; Carling et al., 2023; Baumann & Hartmann, 2026)

2. How well do the predictions agree with independent diachronic evidence?

- Greek, Ancient Egyptian, DatSemShift

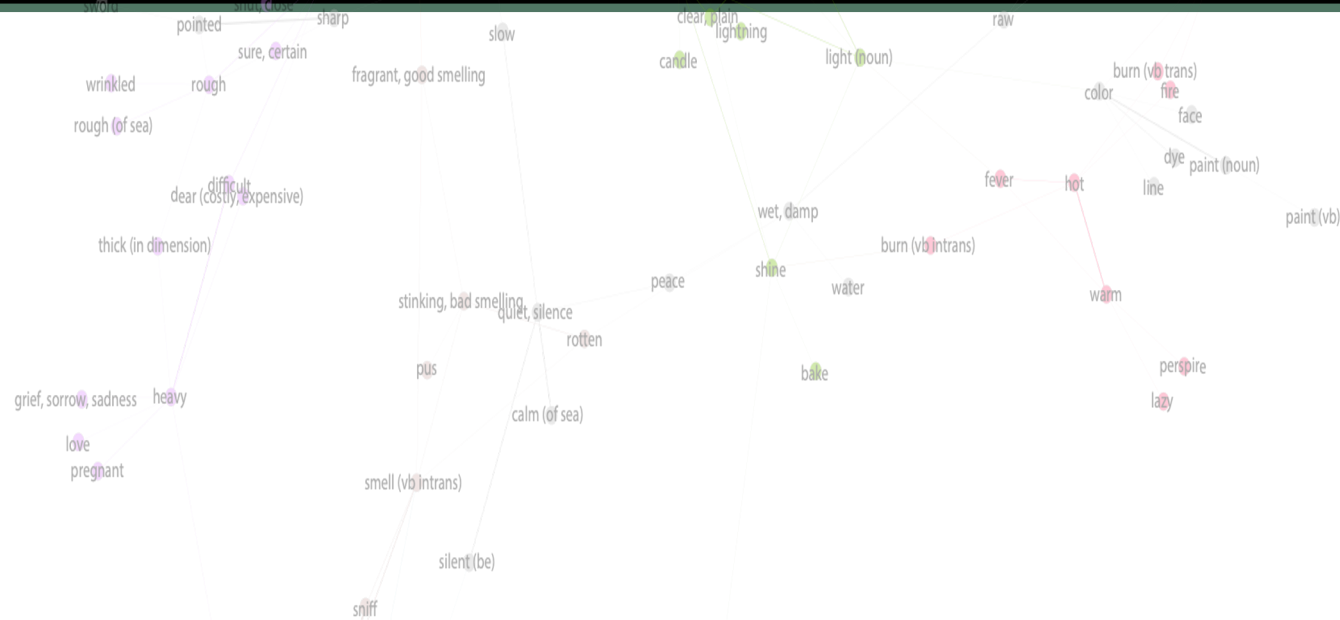
Case study

PERCEPTION-COGNITION

(cf. Georgakopoulos et al., 2022; Norcliffe & Majid, 2024)

Perception-Cognition

Concepts and synchronic data



Concepts and synchronic data

Classification of concepts (semantic field) Concepticon (see List et al., 2016)

Concept sets

Showing 1 to 21 of 21 entries (filtered from 4,034 total entries)



ID	Name	Definition	Semantic field	Ontological category
Search	Search	Search	Sense perception	Action/Process
21	TASTE (SOMETHING)	To try the flavor of something.	--any-- Agriculture and vegetation Animals Basic actions and technology Clothing and grooming Cognition Emotions and values Food and drink Kinship Law Miscellaneous function words Modern world Motion Possession Quantity Religion and belief Sense perception	
252	FRIGHTEN	To make someone afraid or anxious.		
375	SMELL	To make a short, audible inhalation, through the nose, as if to smell something.		
378	PINCH	To squeeze a small amount of a person's skin and flesh, making it hurt.		
708	SHINE	To emit light.		
1039	SHOW	To have somebody see something.		
1079	FEEL	To be in some emotional state.		
1408	HEAR	To perceive sound with the ear, without necessarily paying attention to it.		

Concepts and synchronic data

	A	B	C
1	concepticon_id	concept_label	Concepticon_sf_label
2	1890	BELIEVE	Cognition
3	1902	BETRAY	Cognition
4	1975	CHECK	Cognition
5	3230	DECEIVE	Cognition
6	1626	EXPLAIN	Cognition
7	1079	FEEL	Sense perception
8	1523	FORGET	Cognition
9	1932	GUESS	Cognition
10	1408	HEAR	Sense perception
11	2411	HURT (SOMEBODY)	Sense perception
12	1410	KNOW (SOMETHING)	Cognition
13	504	LEARN	Cognition
14	1608	LISTEN	Sense perception
15	1819	LOOK	Sense perception
16	378	PINCH	Sense perception
17	1409	SEE	Sense perception
18	2124	SMELL	Sense perception
19	1586	SMELL (PERCEIVE)	Sense perception
20	1587	SMELL (STINK)	Sense perception
21	1929	STUDY	Cognition
22	1841	SUSPECT	Cognition
23	21	TASTE (SOMETHING)	Sense perception
24	709	TEACH	Cognition
25	2271	THINK	Cognition
26	1892	TOUCH	Sense perception
27	1812	TRY	Cognition
28	1536	UNDERSTAND	Cognition
29	1635	OBEY	Social political relations
30	2559	FEEL (TACTUALLY)	Basic actions technology
31	2893	NOTICE (SOMETHING)	Basic actions technology

✓ Semantic field

- At least one concept of the pair classified as COGNITION or SENSE PERCEPTION (three exceptions)

✓ Ontology

- Both concepts classified as ACTION/PROCESS

→ 30 concepts, the shared basis for both methods

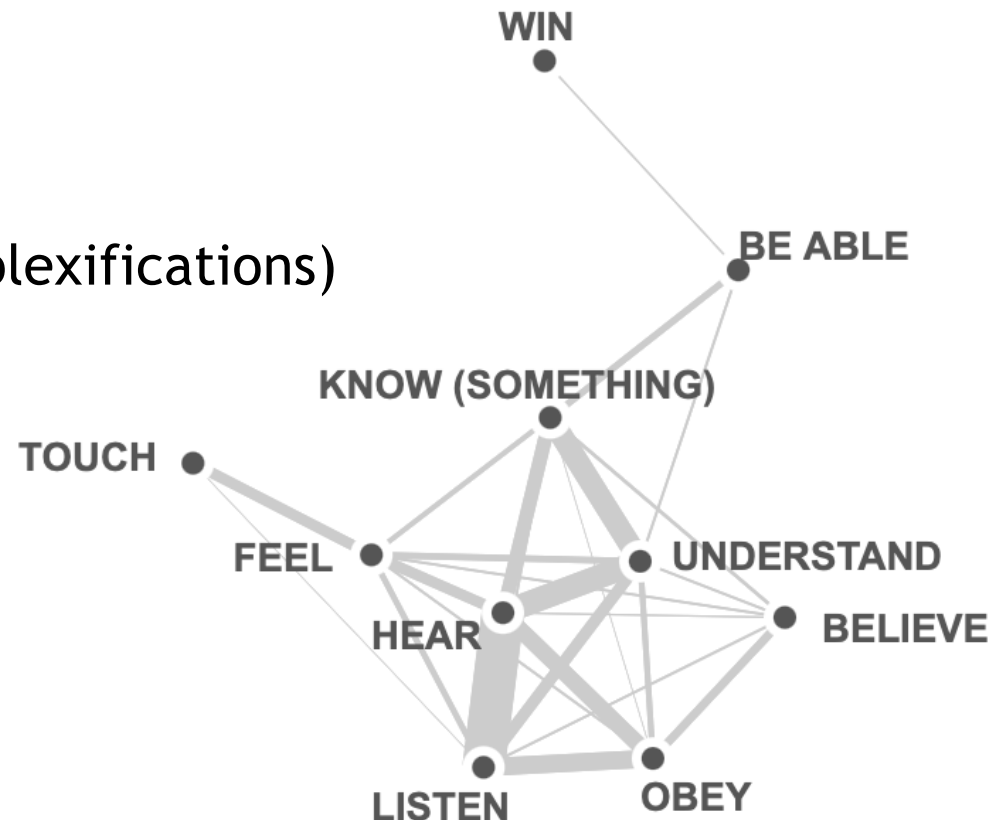
Table 1. The 30 concepts from the domains of COGNITION and SENSE PERCEPTION included in the analysis

Concepts and synchronic data

Synchronic data sourced from



- CLICS³ database ([see Rzymiski et al., 2020](#))
 - 3156 language varieties
 - Network:
 - 2,919 nodes
 - 4,228 edges (colexifications)



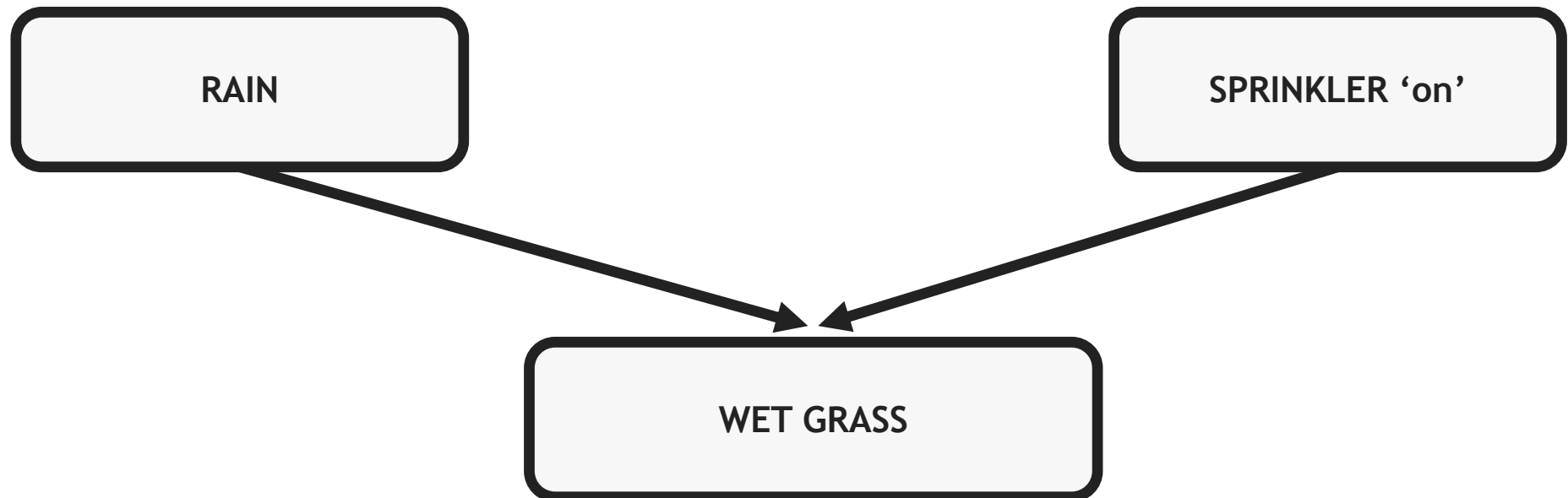
Approach 1:
*Causal inference on
coexpression patterns*

What is Causal Inference?

Causal inference: recover causal structure from observation alone, using patterns of (conditional) independence.

The collider signature (“explaining away”): two independent causes become dependent once their common effect is observed.

Rain and Sprinkler are independent – until we observe that the grass is wet.



No collider → different structures can be statistically indistinguishable: direction is not always recoverable.

From the principle to language

- Meanings = variables; colexification across languages = co-occurrence.
- The PC* algorithm (Spirtes, Glymour & Scheines 2000), applied to colexification by Dellert (2016, 2024).
- It orients an edge only where a collider licenses it; otherwise the edge is left bidirectional.
- Stability via bootstrap resampling (N = 1,000).

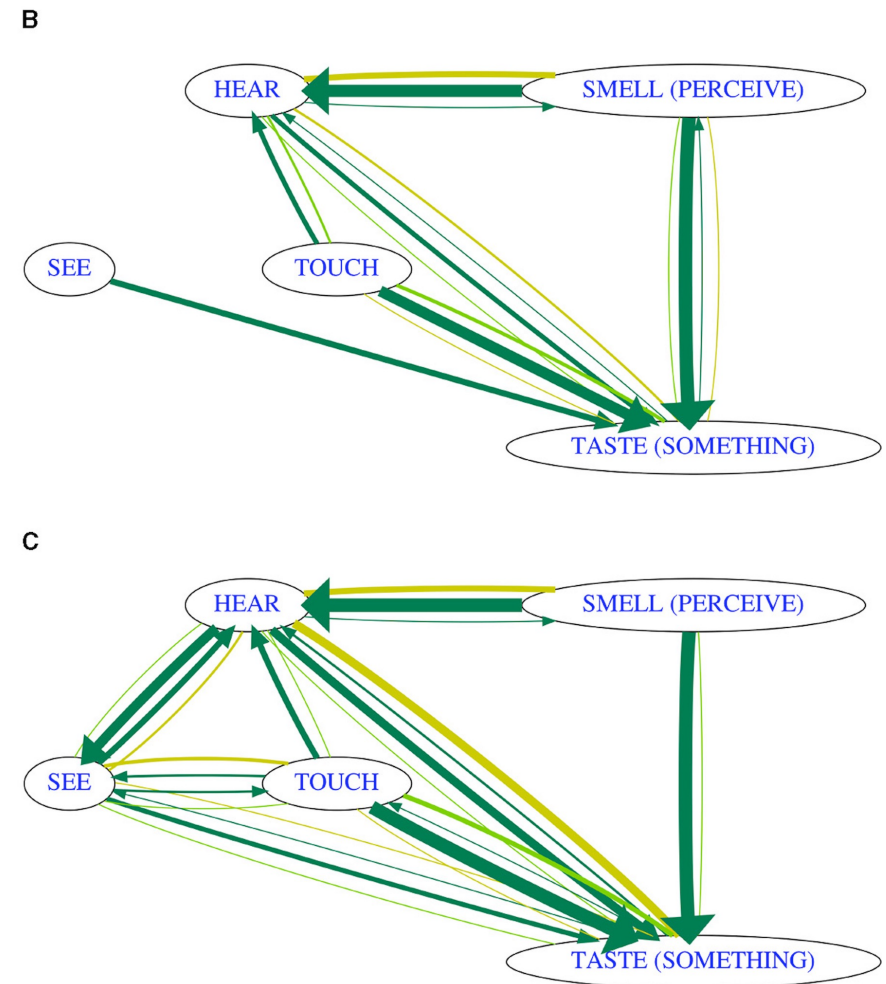


Figure 1. Dellert (2024: Fig. 4). (b) 293 languages, all meanings attested; (c) 1,071 languages, at most two meanings missing.

Causal Inference - mincolex & open network

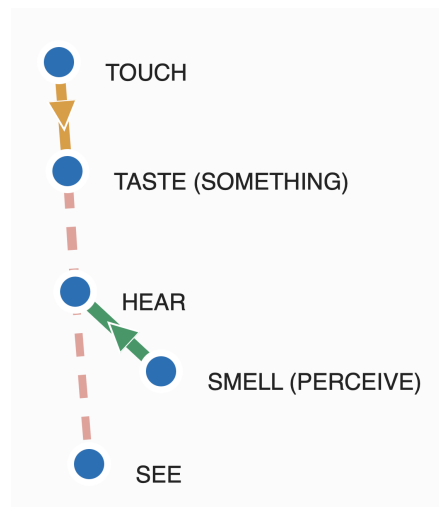


Figure 2a. mincolex 10

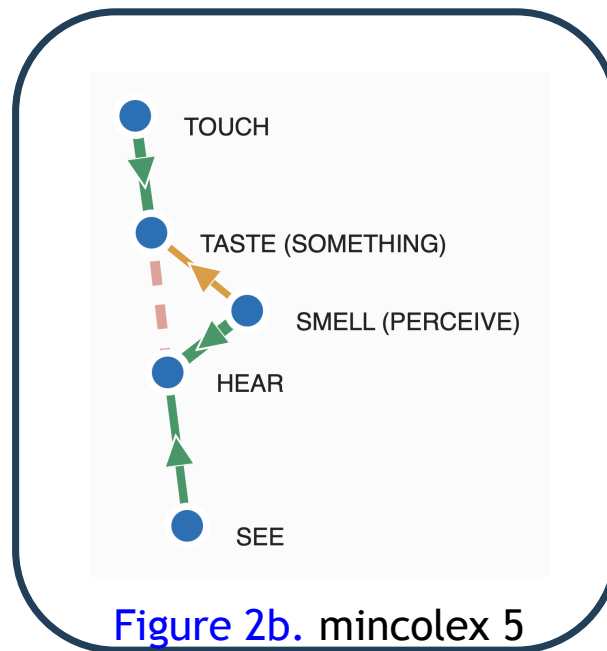


Figure 2b. mincolex 5

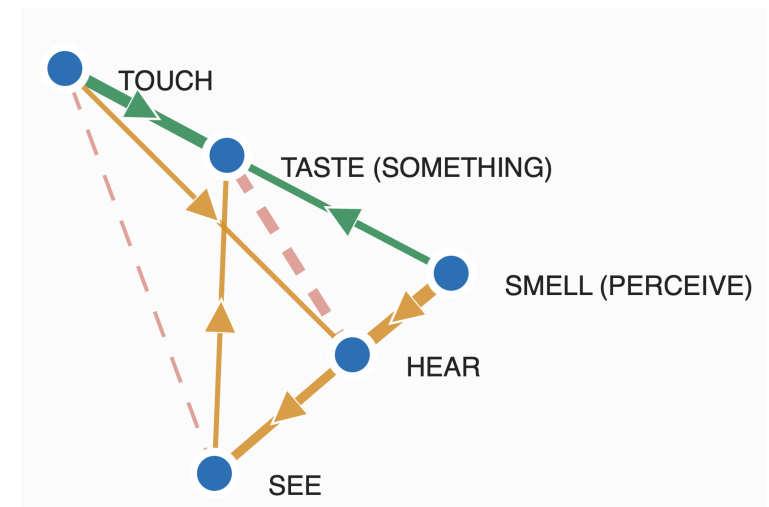


Figure 2c. mincolex 1

Causal Inference - mincolex & open network

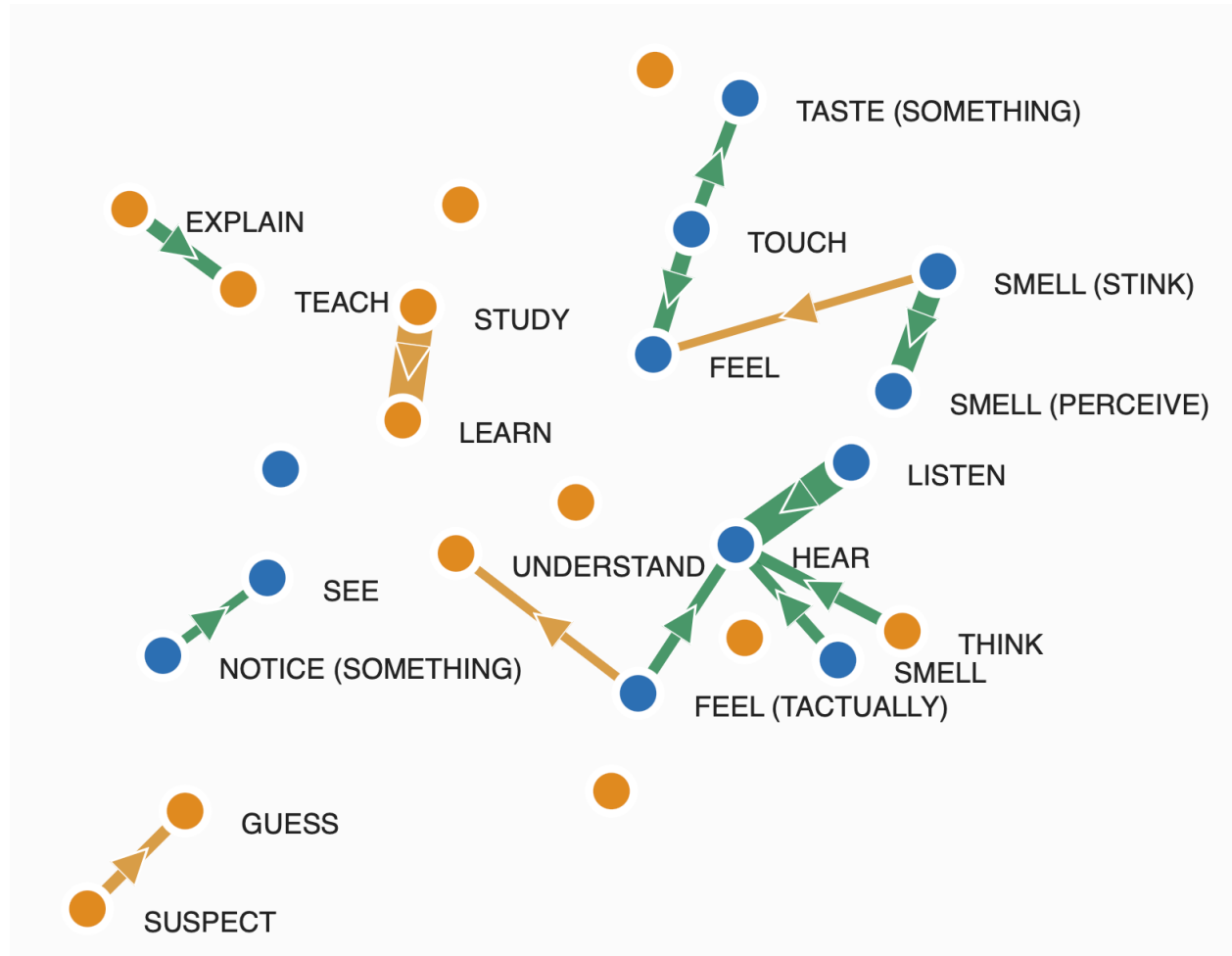


Figure 3. Closed network – directed edges only (mincolex 5)

Causal Inference - mincolex & open network

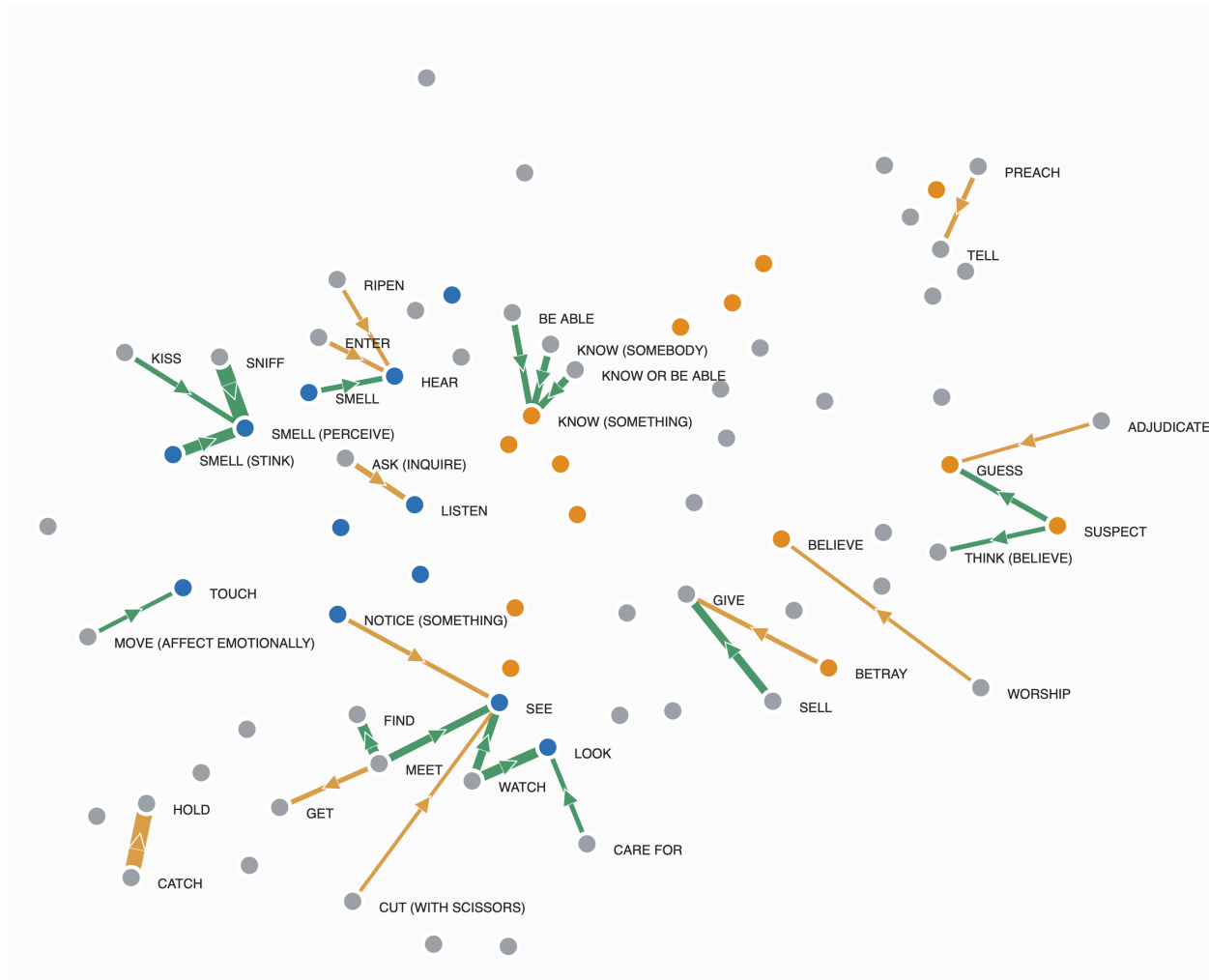


Figure 4. Open network — directed edges only
(mincolex 5)

Admitting the periphery is productive.

At Action-Process / mincolex 5 (227 edges): 18 directed, 194 bidirectional. The robust orientations are coherent *specific* → *general* extensions:

- **Perception** (cline parallels a drop in agentivity): WATCH → SEE 0.99, SNIFF → SMELL 1.00, WATCH → LOOK 0.96, KISS → SMELL 0.98, CARE FOR → LOOK 0.89.
- **Cognition**: SUSPECT → GUESS 0.94, SUSPECT → THINK 0.91.
- Signal converges by $N \approx 250$ resamples.

Causal Inference - take away

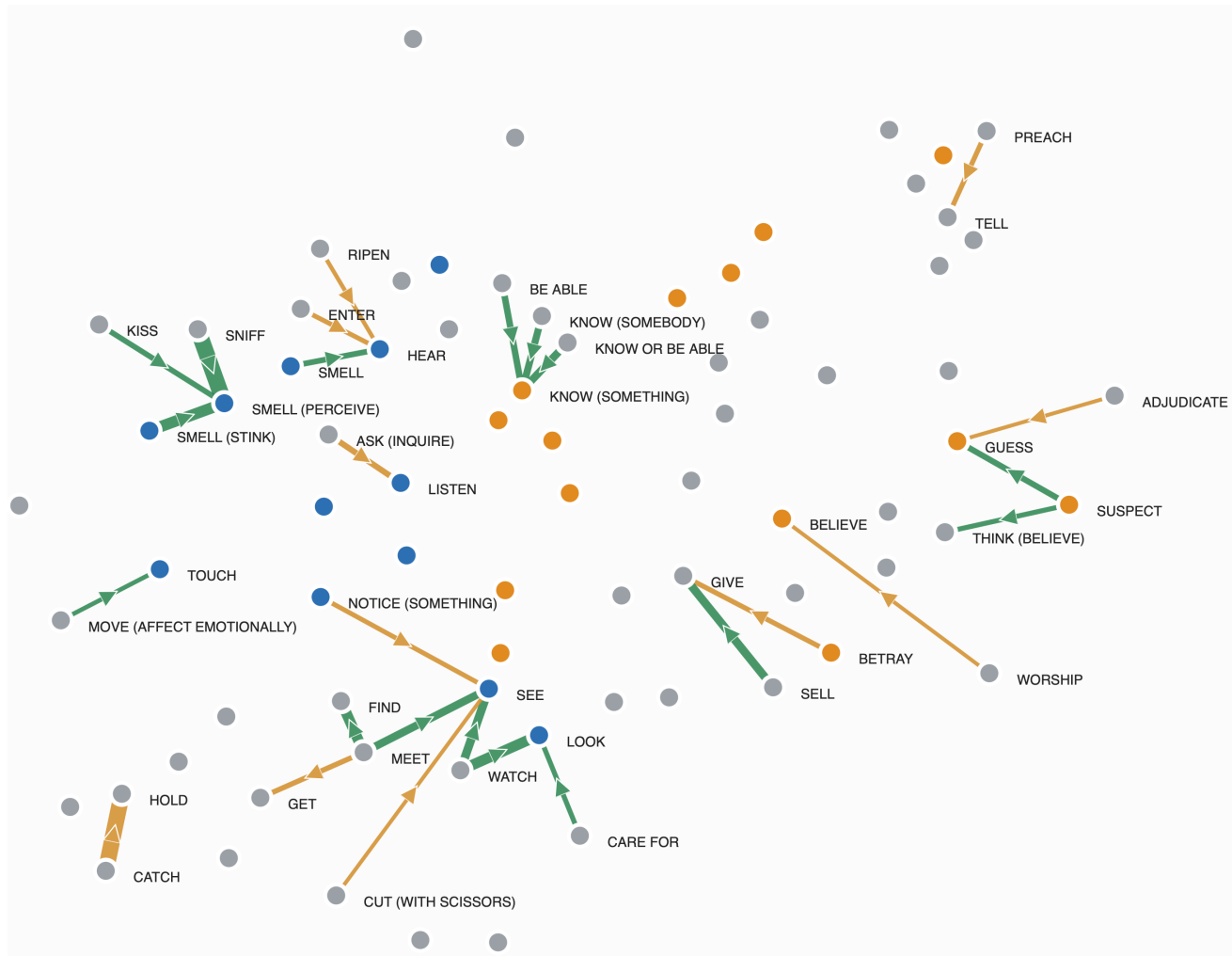


Figure 5. Open network — directed edges only (mincolex 5)

- Recovers a semantic skeleton and *credible source* → *domain* extensions that needs to be evaluated against diachronic data
- But abstains on every perception ↔ cognition bridge (0 oriented)

PredictMe addresses these questions

PredictME

From Networks to Semantic Change:
A Multifactorial Approach to the
Predictability of Meaning Evolution



Approach 2: *Concept properties*

**Predict directionality from
psycholinguistic properties**

Predictive model construction: The synchronic data

NoRaRe (see Tjuka et al., 2022)

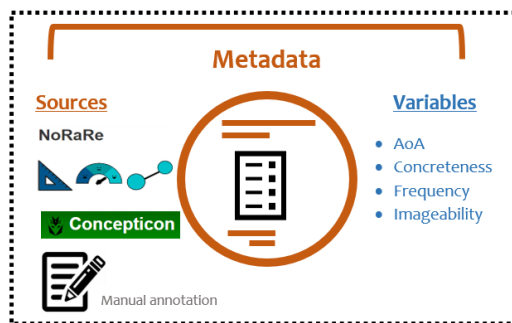


Focus on four psycholinguistic variables

- a) Concreteness
- b) Imageability
- c) Frequency
- d) Age of Acquisition



- These factors serve as predictors of semantic change direction
- Each concept must have at least one value for each predictor.



Tjuka, Annika, Robert Forkel, and Johann-Mattis List. 2022. Linking Norms, Ratings, and Relations of Words and Concepts Across Multiple Language Varieties. *Behavior Research Methods* 54. 864–884. DOI: [10.3758/s13428-021-01650-1](https://doi.org/10.3758/s13428-021-01650-1)

Predictive model construction: The synchronic data

NoRaRe



Variable	Number of datasets	Number of languages	Languages
Frequency	11	20	Basque, Chilean Spanish, Chinese, Czech, Dutch, English, Estonian, Finnish, French, German, Greek, Italian, Māori, Polish, Portuguese, Russian, Spanish, Swahili, Tok Pisin, Turkish
Concreteness	9	6	Chinese, English, French, Italian, Polish, Spanish
Imageability	11	6	Chinese, English, German, Italian, Polish, Spanish
Age of Acquisition (AoA)	7	26	Arabic, Catalan, Croatian, Danish, Dutch, English, French, German, Greek, Hebrew, Hungarian, Italian, Kannada, Malay, Mandarin Chinese, Norwegian, Persian, Polish, Portuguese, Serbian, Slovak, Spanish, Swedish, Turkish, Ukrainian, Vietnamese

Table 2. Datasets and language coverage of the psycholinguistic predictors

Predictive model construction: The synchronic data

- **Summary: final dataset for analysis**

Inclusion criteria

✓ Semantic field

- At least one concept classified as **COGNITION** or **SENSE PERCEPTION**

✓ Ontology

- Both concepts classified as **ACTION/PROCESS**

✓ Psycholinguistic data

- Each concept must have at least one value for each predictor

→ 1919 colexification pairs

Predictors of semantic change

Predictors of semantic change

Concreteness

H1: In a colexification pair, the concept with the highest concreteness score will most likely be the source meaning in a process of change (see Xu et al., 2017; cf. Winter and Srinivasan, 2022)



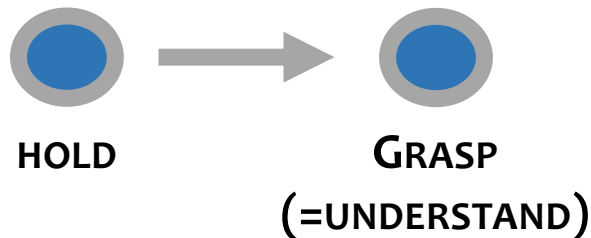
See the asymmetrical relationship between concrete and abstract entities, as theorized in Conceptual Metaphor Theory (e.g., Lakoff & Johnson, 1980; cf. Dunn 2015; on diachrony, see Heine et al. 1991)

Predictors of semantic change



Concreteness

H1: In a colexification pair, the concept with the highest concreteness score will most likely be the source meaning in a process of change (see Xu et al., 2017; cf. Winter and Srinivasan, 2022)



Dataset in NoRaRe	Property	HOLD	GRASP (=UNDERSTAND)	RATING SCALE
Scott-2019- Concreteness-ENGLISH	Concreteness	4.303	3.853	7-point

Variable ENGLISH_CONCRETENESS_MEAN

Number of Scale Points: 7 Concreteness is a measure of how concrete or abstract something is..

Predictors of semantic change



Imageability

H2: In a colexification pair, the concept with the highest imageability score will most likely be the source meaning in a process of change



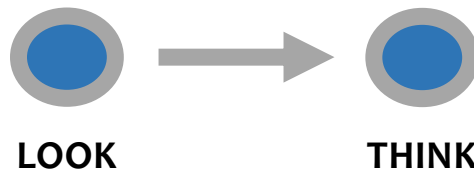
Domains that are less imageable (e.g., TIME) generally depend on more imageable domains (e.g., SPACE). In line with many Conceptual Metaphor Theorists (e.g., Lakoff & Johnson, 1980)

Predictors of semantic change



Imageability

H2: In a colexification pair, the concept with the highest imageability score will most likely be the source meaning in a process of change



Dataset in NoRaRe	Property	LOOK	THINK	RATING SCALE
Clark-2004-ImageryFamiliarity-ENGLISH_IMAGEABILITY_MEAN	Imageability	3	2.59	7-point

Variable ENGLISH_IMAGEABILITY_MEAN

Mean value of imageability ratings based on a 7-point scale.

Predictors of semantic change



Frequency

H3: In a colexification pair, the concept with the highest frequency score* is more likely to be the source meaning

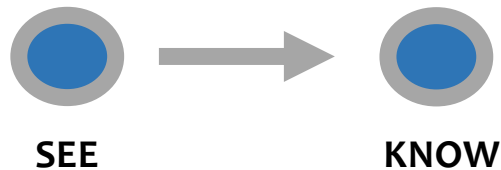
* As indexed by word frequency metrics from various languages

- Important role in the studies of the lexicon ([Zipf, 1949](#))
 - Significant factor in understanding various aspects of language behavior (see [Bybee, 2007](#))
 - May function as an important driver of language change ([Pagel et al. 2007](#); cf. [Mooers & Pappas, 2011](#) for a critique).
 - Higher-frequency words tend to be more polysemous ([Zipf, 1945](#); [Casas et al., 2019](#))
- Frequent meanings are more likely to serve as source meanings for semantic extensions
(see [Winter & Srinivasan, 2022](#))

Predictors of semantic change

💡 Frequency

H3: In a colexification pair, the concept with the highest frequency score is more likely to be the source meaning
(see [Winter & Srinivasan, 2022](#))



Dataset in NoRaRe	Property	SEE [SP._VER]	KNOW [SP._SABER]
Cuetos-2011-Frequency-SPANISH	frequency	4.659535907154217	4.319189369204891

Variable SPANISH_FREQUENCY_LOG

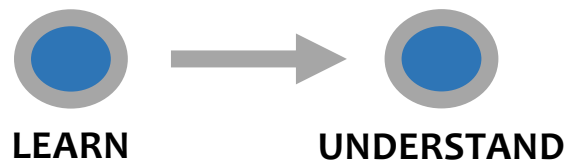
Log10 of subtitle word frequency.

Predictors of semantic change

💡 Age of Acquisition

H4: In a colexification pair, the concept that is acquired earlier will most likely be the source meaning.

(see also Baumann & Hartmann, 2026)



In a process of change that involves the pair <LEARN, UNDERSTAND>, we assume that LEARN will most likely be the source meaning, since this is acquired earlier than UNDERSTAND

Dataset in NoRaRe	Property	LEARN [SP._APRENDER]	UNDERSTAND [SP._ENTENDER]
Alonso-2015-AoA-SPANISH	AoA	4.54	5.44
Variable SPANISH_AOA_MEAN			

Average score of AoA ratings on a scale from 1 to 11.

Predictor preprocessing and source score computation

Predictor preprocessing & source score computation

Data preprocessing

1. Data harmonization

a. Scale alignment

- Reverse-code variables whose direction differed
 - Imbir (2021) concreteness:
 - original: higher = **less** concrete

b. Standardization

- z-transformation
- performed **within each dataset**

Predictor preprocessing & source score computation

Data preprocessing

2. Aggregation

Computing concept-level predictor scores

For each predictor

- retain all available standardized scores
- average across datasets

Result

- Frequency_mean
- Concreteness_mean
- Imageability_mean
- AoA_mean

Finally

- reverse-code AoA_mean
- higher values = earlier acquisition

Predictor preprocessing & source score computation

Source score

- Predicting the originating sense (each predictor contributes equally)

$$S_i = \frac{C_i + F_i + A_i + I_i}{4}$$

where

- C = concreteness
- F = frequency
- A = reverse-coded AoA
- I = imageability

For each colexification pair

- compare the two source scores
- **higher score** → **predicted source**

Predictor preprocessing & source score computation

Confidence threshold

- Confidence-based prediction

For each pair,

$$\Delta = |S_1 - S_2|$$

Large Δ

→ clear prediction

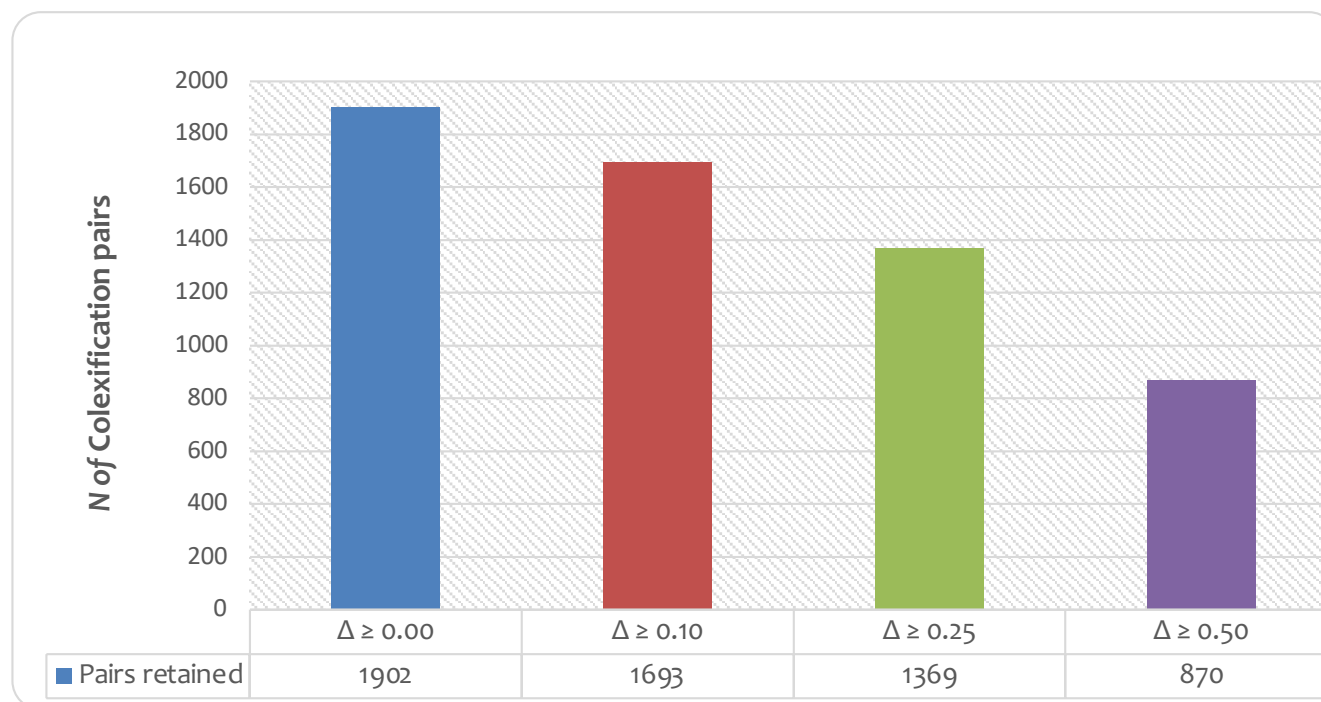
Small Δ

→ concepts receive very similar scores
(excluded from directional interpretation)

Predictor preprocessing & source score computation

Confidence threshold

- Confidence-based prediction



- Final choice:

$$\Delta \geq 0.25$$

- Removes weak predictions
- Preserves most of the dataset

Table 3. Number and proportion of colexification pairs retained under alternative confidence thresholds (Δ)

Predictor preprocessing & source score computation

How the predictive model works: two examples

→ Correct prediction | Arrow = predicted direction (from left to right concept) | ● = Concept

$$\text{Source score (S)} = (C + F + A + I) / 4$$

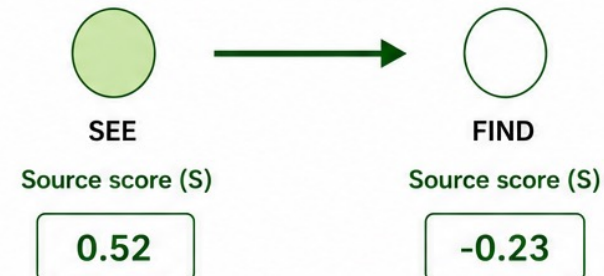
C = Concreteness, F = Frequency, A = reverse-coded AoA, I = Imageability

1 KNOW (SOMETHING) → ADMIT



✓ -0.50 > -1.02 → Predict KNOW (SOMETHING) as source (matches attested direction)

2 SEE → FIND



✓ 0.52 > -0.23 → Predict SEE as source (matches attested direction)

i How the source score is computed

- For each concept, obtain standardized (z-scored) values for: Concreteness (C), Frequency (F), Reverse-coded AoA (A), Imageability (I)
- Average the four aggregated z-scores:
$$S = (C + F + A + I) / 4$$
- For each pair, compare the two source scores:
Higher score = predicted source of semantic change.

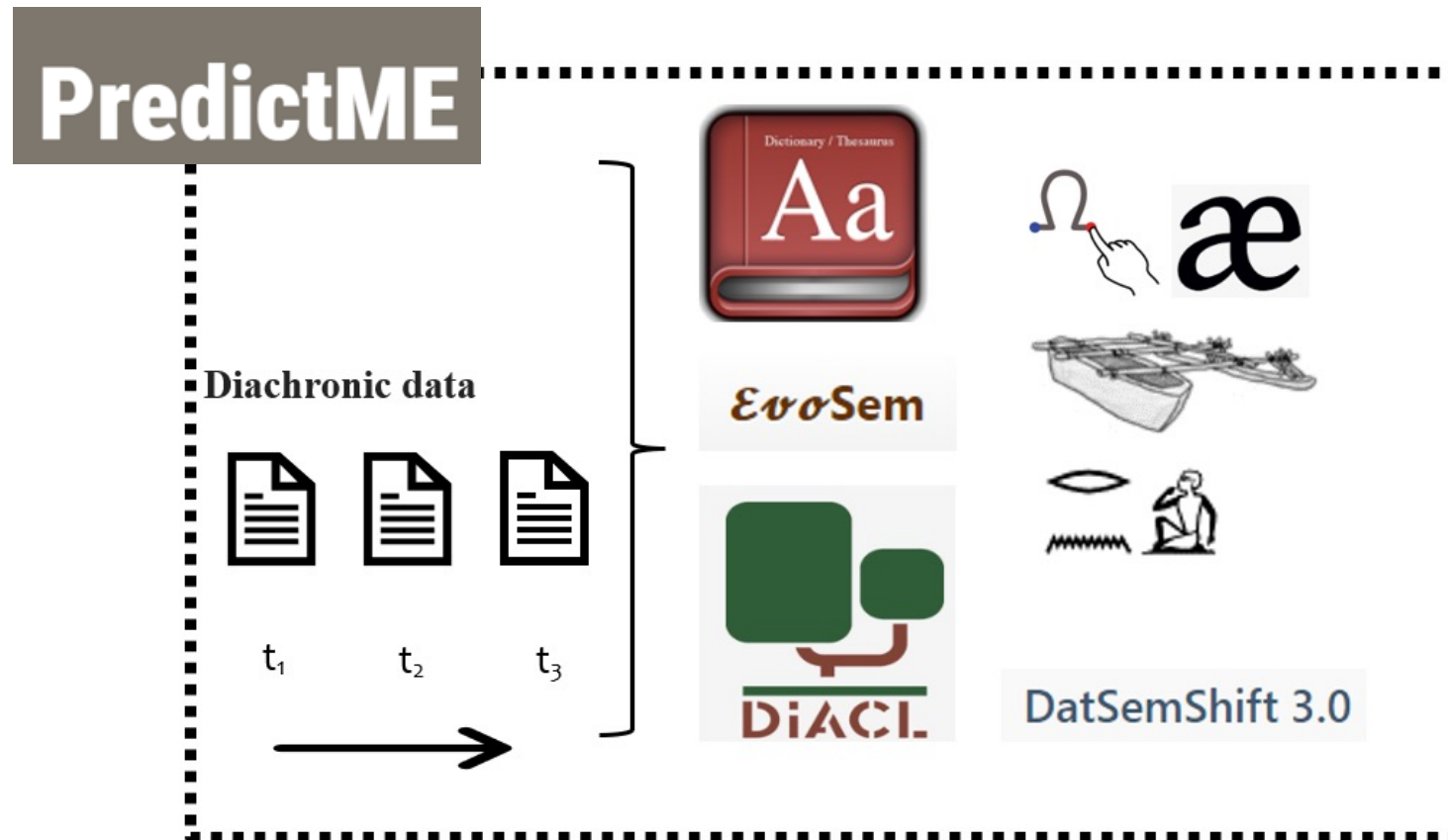
Example: KNOW (SOMETHING)

Predictor	Aggregated z-score
Frequency (F)	1.08
Concreteness (C)	-1.81
Imageability (I)	-1.19
Reverse-coded AoA (A)	-0.08

$$\begin{aligned}\text{Source score (S)} &= (C + F + A + I) / 4 \\ &= (-1.81 + 1.08 - 0.08 - 1.19) / 4 \\ &= (-2.00) / 4 \\ &= -0.50\end{aligned}$$

Evaluation of the Models

Model evaluation: Diachronic validation data



Model evaluation: Diachronic validation data

TODAY: data from DatSemShift

- Database of Semantic Shifts in languages of the world
(Zalizniak et al., 2012; Zalizniak, 2018; DatSemShift 3.0,
<https://datsemshift.ru/>)

Meaning pair	Form(s)	Language(s)	Realization type
hear → obey	šēmû	Akkadian	Synchronic polysemy
hear → understand	entendre → entendre	French → Spanish	Cognate
arc → rainbow	Bogen → Regenbogen	German	Morphological derivation
moon → Monday	mānadagher → maanantai	Old Swedish → Finnish	Borrowing
perceive smell → know	sapio → savoir	Latin → French	Diachronic sem. evolution

Table 4. The different realization types of semantic shifts in the DatSemShift

Database of Semantic Shifts
in languages of the world

Model evaluation: Diachronic validation data

TODAY: data from Greek

- Proceed **onomasiologically**: identify the lexemes in Ancient Greek that express meanings in SENSORY PERCEPTION and COGNITION
- Time span: (8th c. BC - 17th c. AD)

Lexical Resources

- Dictionaries

GE

The Brill Dictionary
of Ancient Greek

- Brill Dictionary ([Montanari, 2015](#))

H. G. Liddell and R. Scott

Greek-English
Lexicon

With a Revised Supplement

- LSJ ([Liddell & Scott, 1996](#))



- Cambridge Greek Lexicon ([Faculty of Classics, University of Cambridge & Diggle, 2021](#))

dictionnaire
étymologique
de la langue grecque

- [Chantraine's \(1968\)](#) Etymological dictionary

ΛΕΞΙΚΟ
ΤΗΣ ΜΕΣΑΙΩΝΙΚΗΣ ΕΛΛΗΝΙΚΗΣ
ΔΗΜΩΔΟΥΣ ΓΡΑΜΜΑΤΕΙΑΣ
—οργάνο—
1100–1669

- [Kriaras \(1968-\)](#) Dictionary of Medieval Greek Vernacular Literature (1100–1669)

- Relevant studies on (ancient) Greek PERCEPTION and COGNITION verbs ([e.g. Luraghi, 2020; Luraghi & Sausa, 2019](#))

Model evaluation: Diachronic validation data

Domain	Concepts	Ancient Greek Words
PERCEPTION	SEE	<i>horáō</i>
PERCEPTION	SEE	<i>dérkomai</i>
PERCEPTION	SEE	<i>leússō</i>
PERCEPTION	SEE	<i>sképtomai</i>
PERCEPTION	HEAR	<i>klúō</i>
PERCEPTION	HEAR	<i>aíō</i>
PERCEPTION	HEAR	<i>akoúō</i>
COGNITION	THINK (REFLECT)	<i>phronéō</i>
COGNITION	UNDERSTAND	<i>ennoéō</i>
COGNITION	KNOW	<i>oîda, gignóskō</i>
COGNITION	LEARN	<i>manthánō</i>

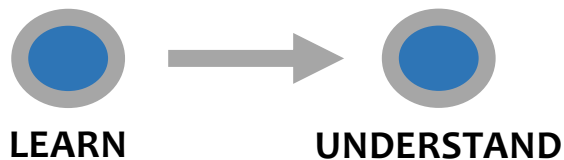
Table 5. A subset of Ancient Greek lexemes expressing the concepts in the domains of SENSE PERCEPTION and COGNITION

Model evaluation: Diachronic validation data

- **Proceed semasiologically**
 - List the different meanings of the lexemes identified (in different stages)

Diachronic data: Greek

Lexeme	Diachronic stage	Meaning 1 'understand'	Meaning 2 'learn'
<i>manthánō</i>	1_Homeric Greek	0	1
	2_Classical Greek	1	1



Model evaluation: Diachronic validation data

- **Proceed semasiologically**

- List the different meanings of the lexemes identified (in different stages)

(1) ἄλλ' ἐπεὶ οὖν δὴ ἔργα κάκ' ἔμαθεν
all' epei oûn dè érga kák' émathen
but when PTCL PTCL deed(N):ACC.PL evil:ACC.PL.N learn:AOR.3SG
'But since he has learned only deeds of evil' (Hom., Od. 22.226)

(2)
Dionysus μανθάνεις;
mantháneis
learn:PRS.2SG
Xanthias πάνυ μανθάνω
pánu manthánō
ADV learn:PRS.1SG
'**Dionysus:** do you understand?

Context: Dionysus is checking whether Xanthias has caught his point/joke

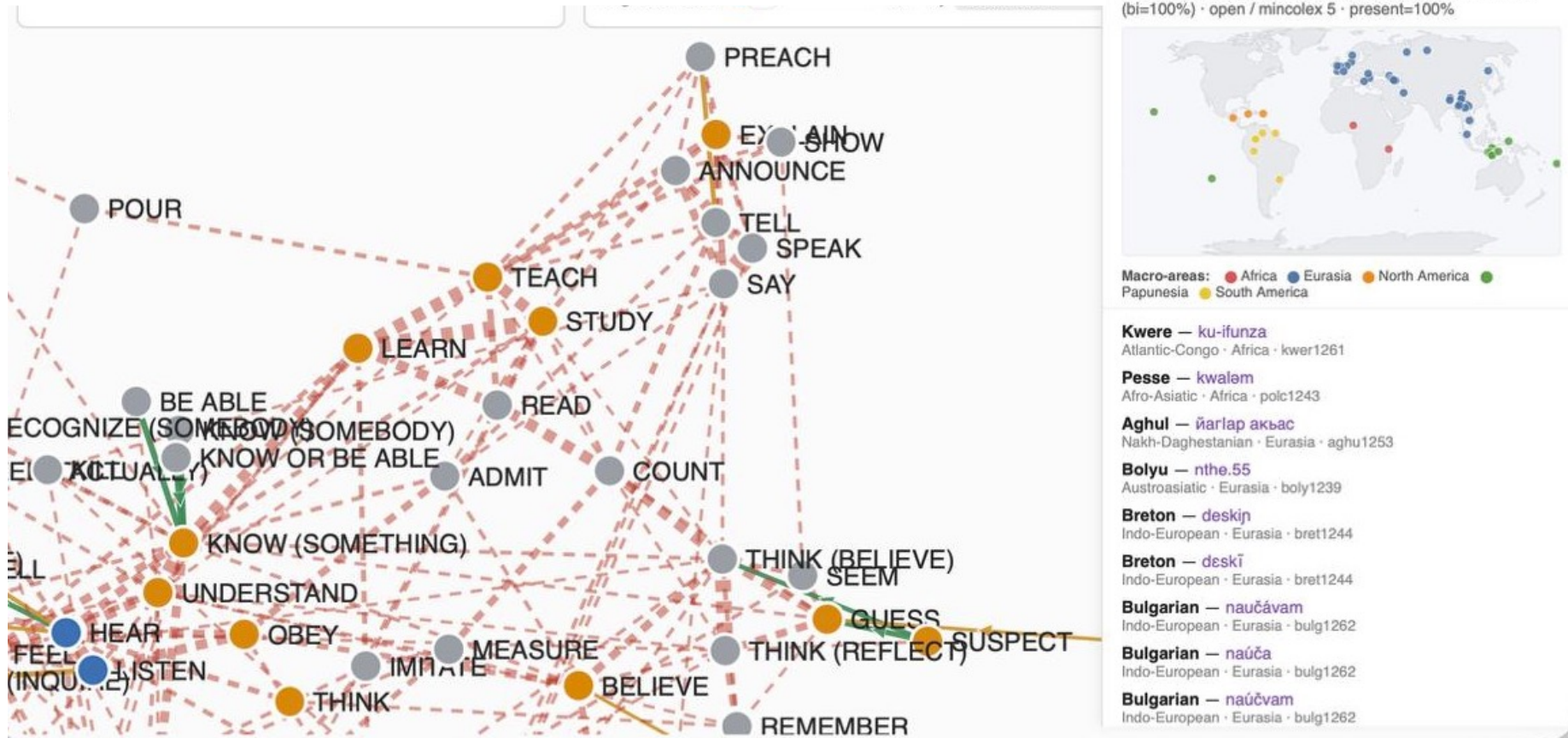
Xanthias: I understand perfectly' (Aristoph. *Frogs* 195)

Approx.
8th c. BC

5th c. BC

Model evaluation: Greek & DatSemShift record vs. Causal inference

Figure 5. Open network - All edges, with selection of the bidirectional link between LEARN and TEACH.

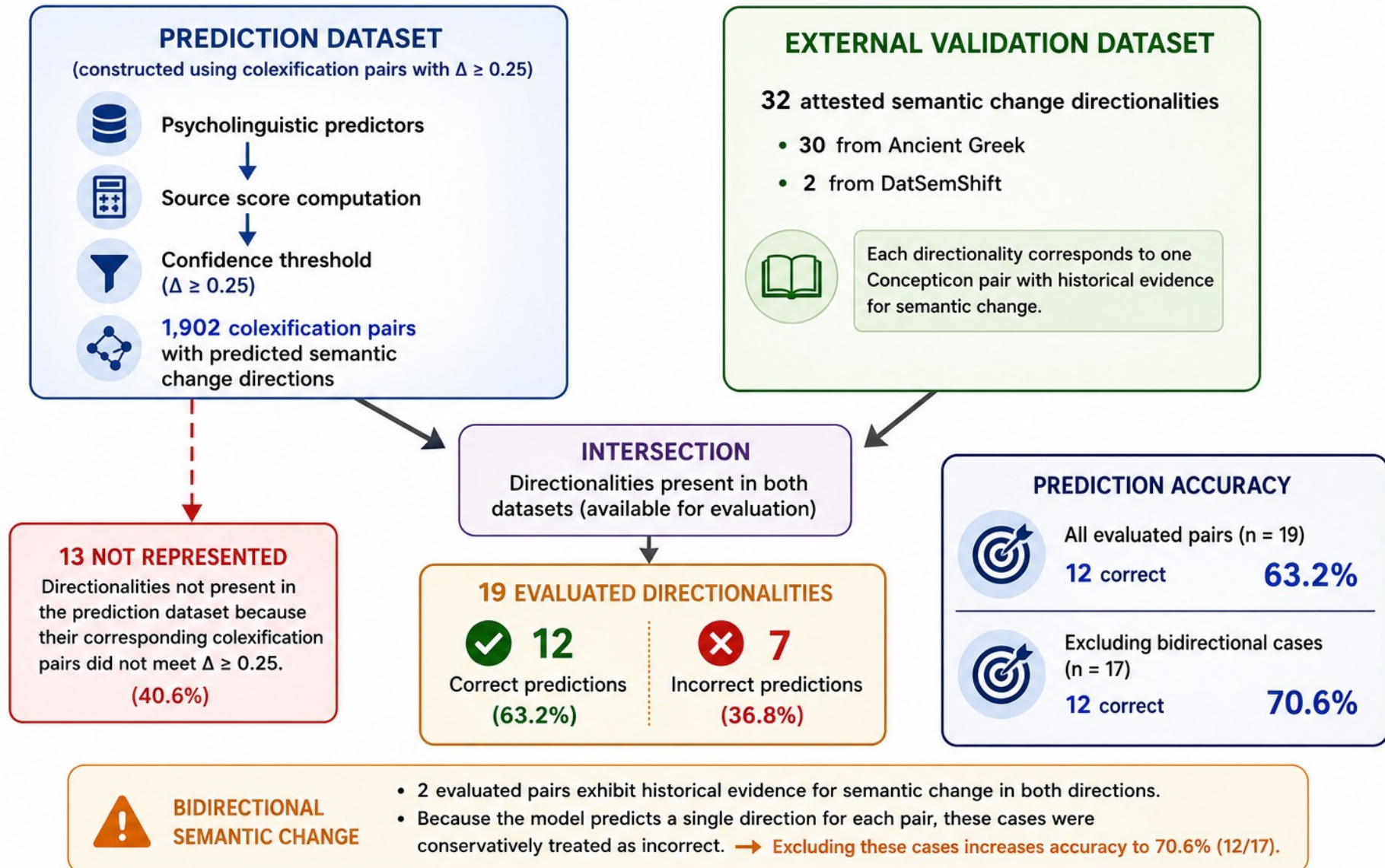


- Of the 32 attested directionalities in Greek & DatSemShift: 0 correspond to oriented edges.

- Where Greek & DatSemShift attest both directions (KNOW ↔ LEARN, LEARN ↔ TEACH), the bidirectional verdict is faithful to the record.

Model evaluation: Greek & DatSemShift record vs. PredictMe

External Validation of PredictME

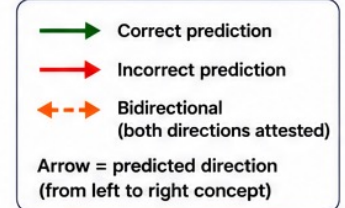


Model evaluation: Greek & DatSemShift record vs. PredictMe

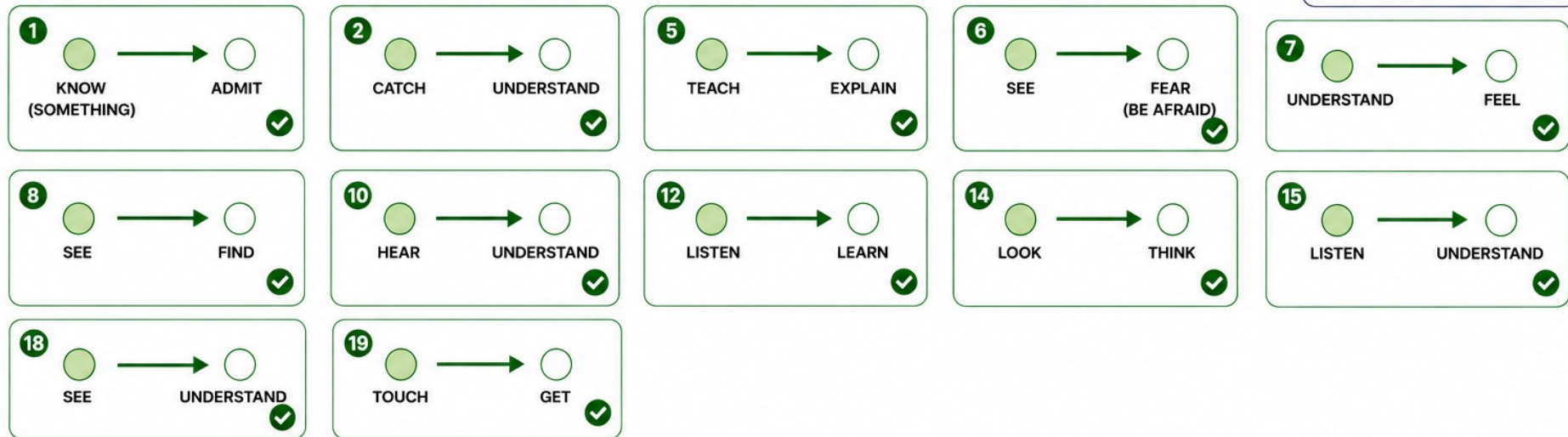
External validation using semantic change directionalities

(matches Ancient Greek and DatSemShift evidence)

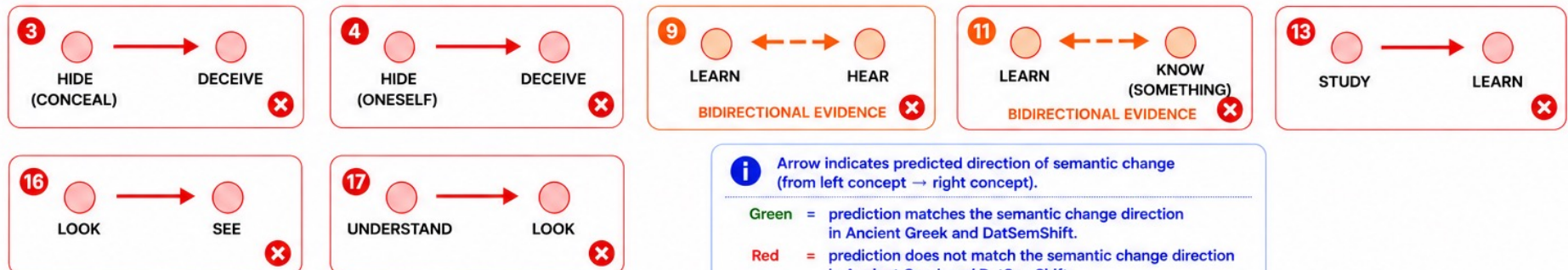
✓ 19 pairs found | 12 correct | 7 incorrect | Accuracy: 63.2%



✓ CORRECT PREDICTIONS (12 pairs)



✗ INCORRECT PREDICTIONS (7 pairs)



i Arrow indicates predicted direction of semantic change (from left concept → right concept).
 Green = prediction matches the semantic change direction in Ancient Greek and DatSemShift.
 Red = prediction does not match the semantic change direction in Ancient Greek and DatSemShift.
 Orange = semantic change attested in both directions in Ancient Greek and DatSemShift.

Model evaluation: Greek & DatSemShift record vs. PredictMe

External validation using semantic change directionalities

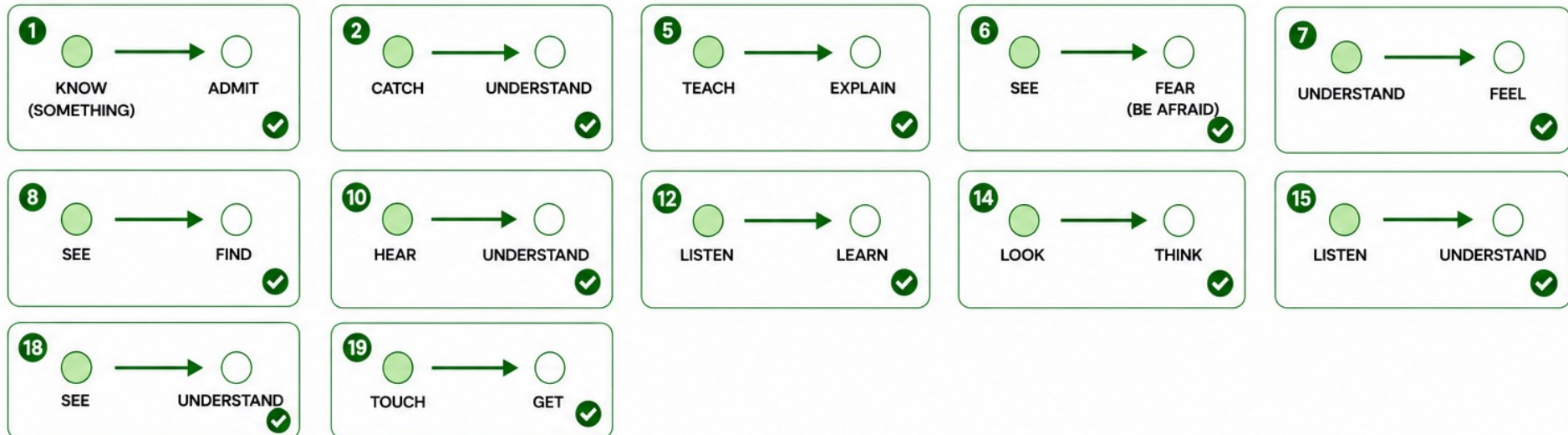
(matches Ancient Greek and DatSemShift evidence)

— excludes bidirectionalities —

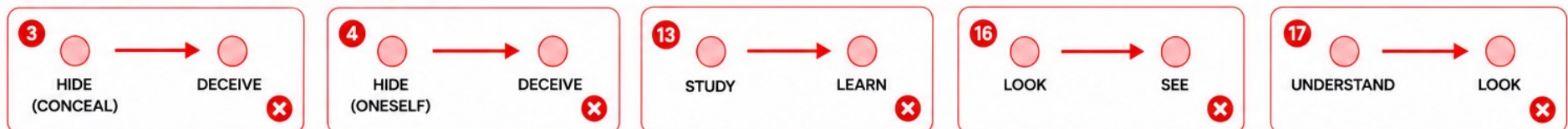
✓ 17 pairs found | 12 correct | 5 incorrect | Accuracy: 70.6%

→ Correct prediction
→ Incorrect prediction
Arrow = predicted direction
(from left to right concept)

✓ CORRECT PREDICTIONS (12 pairs)



✗ INCORRECT PREDICTIONS (5 pairs)



i Arrow indicates predicted direction of semantic change
(from left concept → right concept).

Green = prediction matches the semantic change direction
in Ancient Greek and DatSemShift.

Red = prediction does not match the semantic change direction
in Ancient Greek and DatSemShift.

Model evaluation - Take away (Greek & DatSemShift)

0 % vs ~70 %

Causal inference: 0 / 32 attested directions are oriented

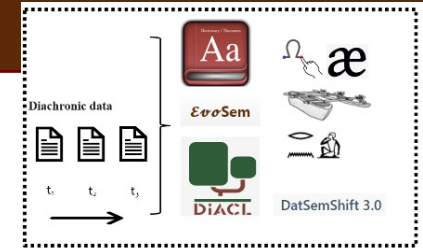
PredictME: ~70% correct

- Because PredictME ignores colexification frequency, it can be scored on far more Greek pairs than causal inference

mincolex	unidirectional correct	accuracy
≥ 1	12 / 17 (+2 bi-attested)	70.6 %
≥ 3	7 / 10 (+1)	70.0 %
≥ 5	7 / 9 (+1)	77.8 %
≥ 10	4 / 6 (+1)	66.7 %

Limitations and future directions

Limitations and future directions



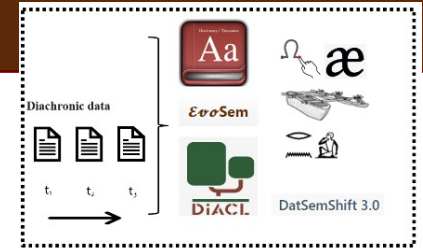
1. Optimizing predictor contributions

- Estimate the relative contribution of each predictor included in the model
- Determine which psycholinguistic properties have the greatest predictive validity for the directionality of semantic change
- Replace equal weighting with empirically derived weights
- Examine issues of multicollinearity

2. Refining source score estimation

- Examine cases with small score differences
- Assess whether predictions are driven by a single predictor
- Introduce robustness and sensitivity analyses

Limitations and future directions



3. Incorporating network structure

- Integrate network-based predictors (e.g. degree, betweenness, closeness, eigenvector centrality)
- Exploit the topology of colexification networks

4. Broadening semantic coverage

- Expand beyond the domains of COGNITION and SENSE PERCEPTION
- Extend the model to all semantic domains represented in CLICS and Concepticon

5. Expanding diachronic validation

- Extend diachronic validation with additional languages (including data from Ancient Egyptian, English, Greek)
- Incorporate evidence from additional resources
(e.g. [DatSemShift](#), [Zalizniak, 2018](#); [DiACL](#), [Carling et al., 2018](#) [EvoSem](#), [Dehouck et al., 2023](#))

Limitations and future directions

- PredictME orients HEAR → OBEY and LISTEN → OBEY

→ Egyptian corroborates where Greek is silent with
sdm “to hear” > “to obey”



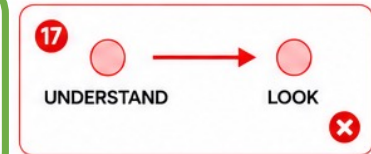
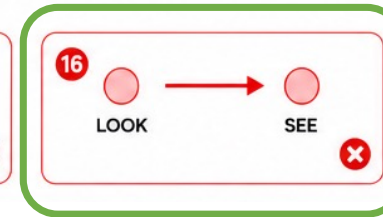
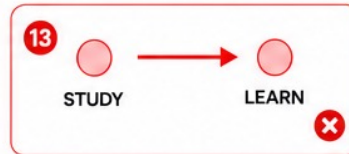
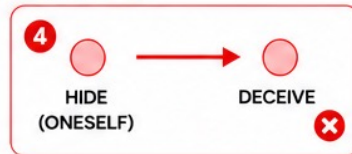
Limitations and future directions

- PredictME orients HEAR → OBEY and LISTEN → OBEY

→ Egyptian corroborates where Greek is silent with
sdm “to hear” > “to obey”



✗ INCORRECT PREDICTIONS (5 pairs)



- Ancient Egyptian also validate predictions that are currently invalidated by the Greek data



ptr “to look” → “to see”



nw “to watch, to look” → “to see”

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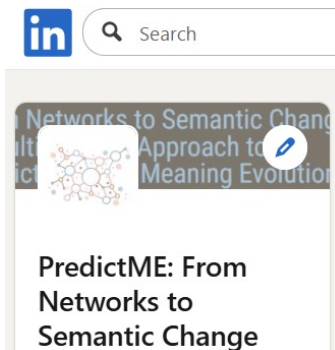
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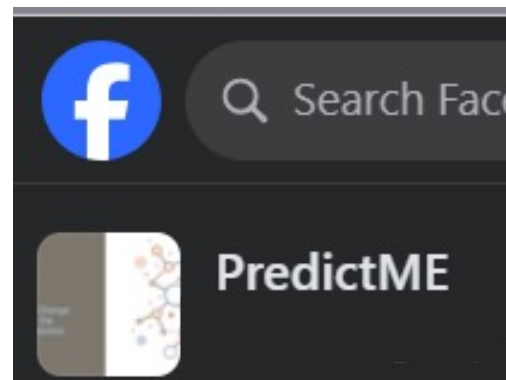


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Thank you!



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