

The Geometry of Categorical and Hierarchical Concepts in Large Language Models

ICLR 2025
April 25, Singapore



Kiho Park
UChicago

Graduating next year!



Yo Joong "YJ" Choe
UChicago / INSEAD



Yibo Jiang
UChicago

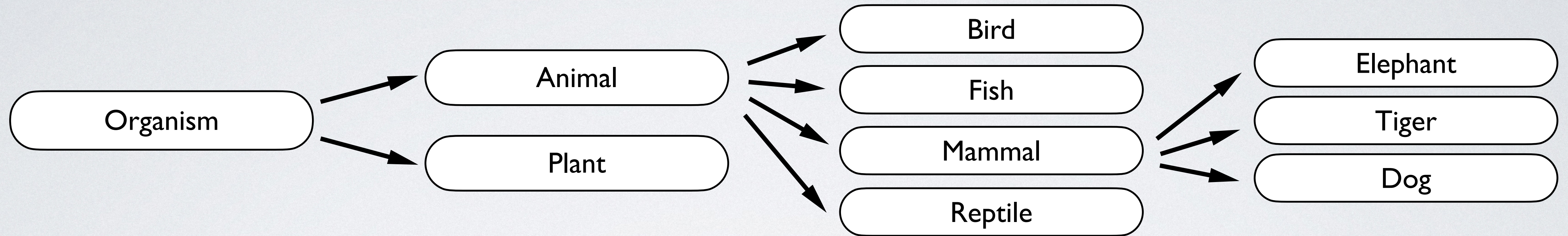


Victor Veitch
UChicago / Google

The “Big Picture” Question

*How is semantic meaning encoded in
the representation spaces of LLMs?*

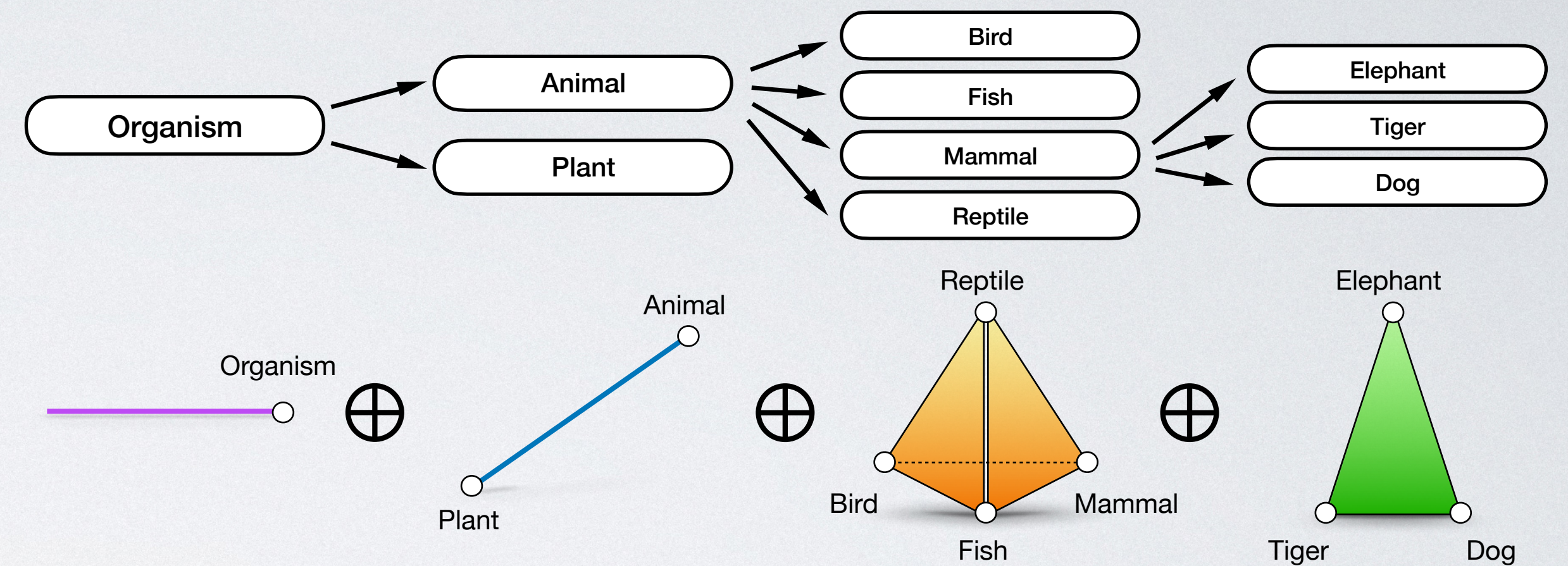
Key Questions in This Work



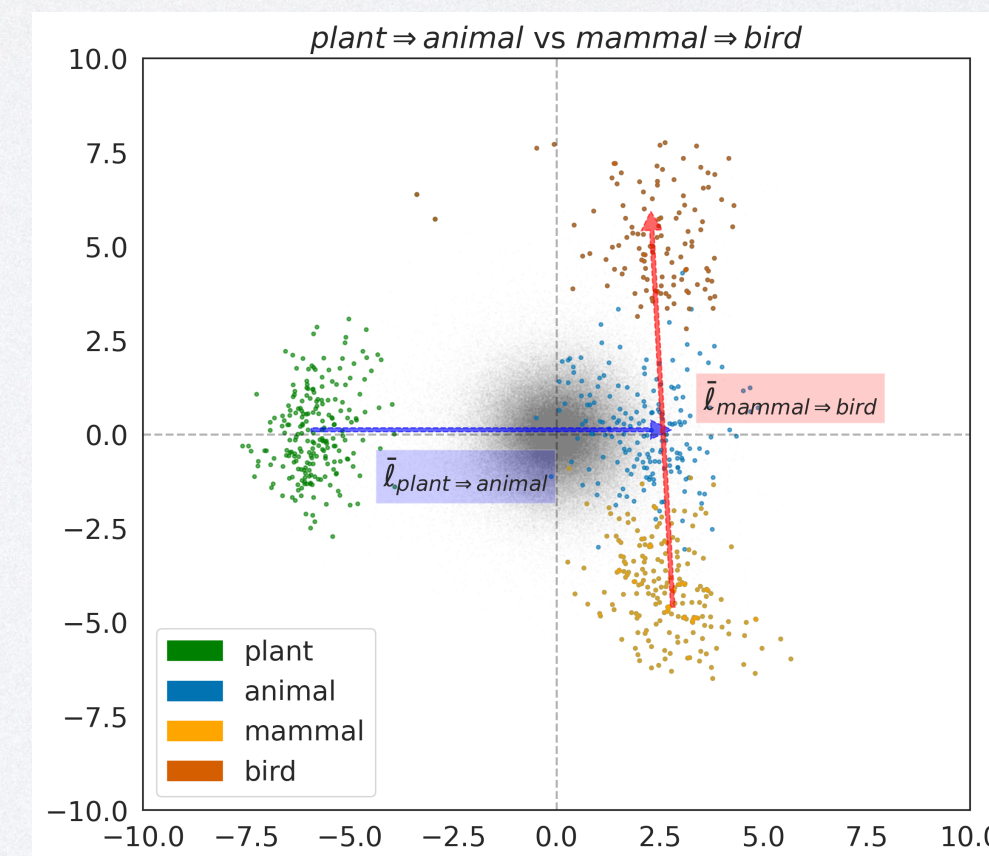
1. What is a representation of a single feature (e.g., *is_animal*)?
2. How are categorical concepts represented?
3. How are hierarchical relations between concepts represented?

Summary of Contributions

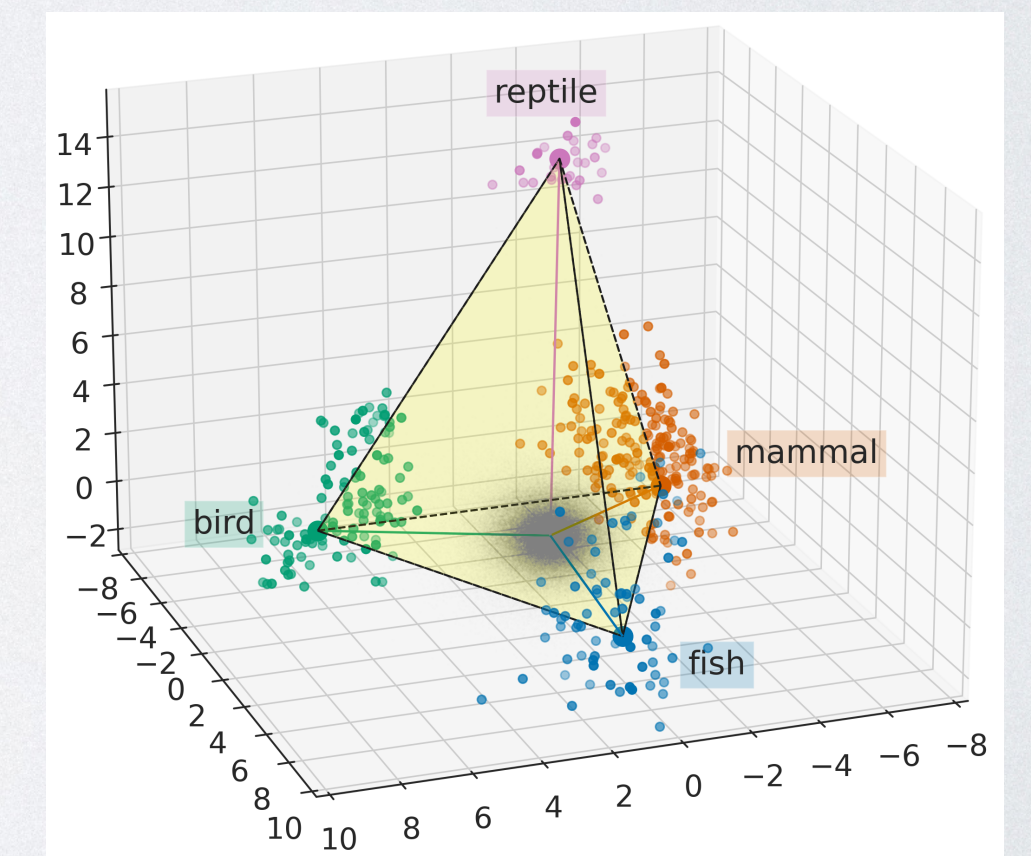
1. Single features are represented as *vectors*
2. Categorical concepts are represented as *polytopes*
3. Hierarchical relations are represented as *orthogonality*



(a) Pictorial depiction of the representation of hierarchically related concepts.



(b) Hierarchy is encoded as orthogonality in Gemma.

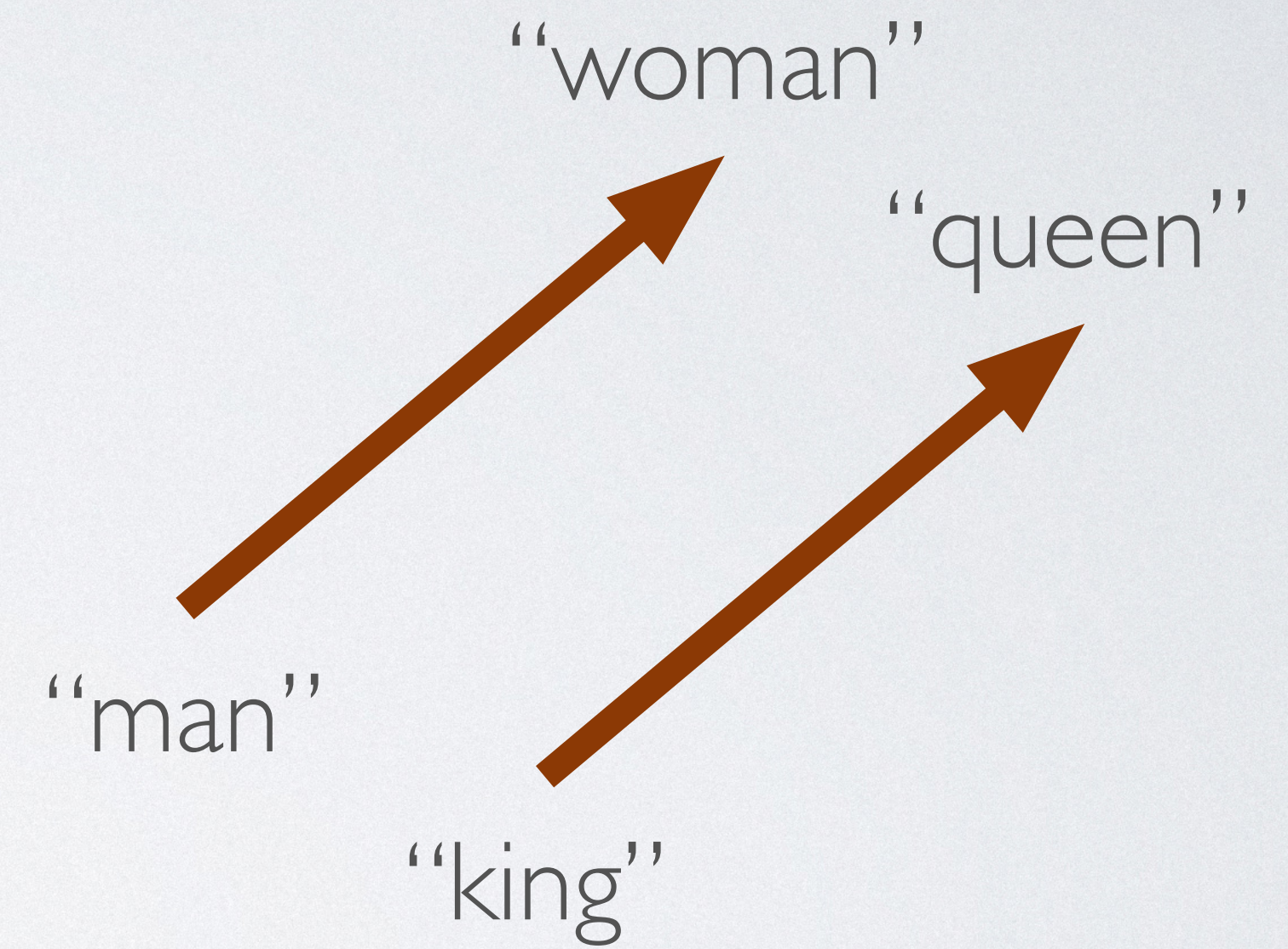


(c) Categorical concepts are represented as polytopes in Gemma.

Background

Background I. Linear Representation Hypothesis

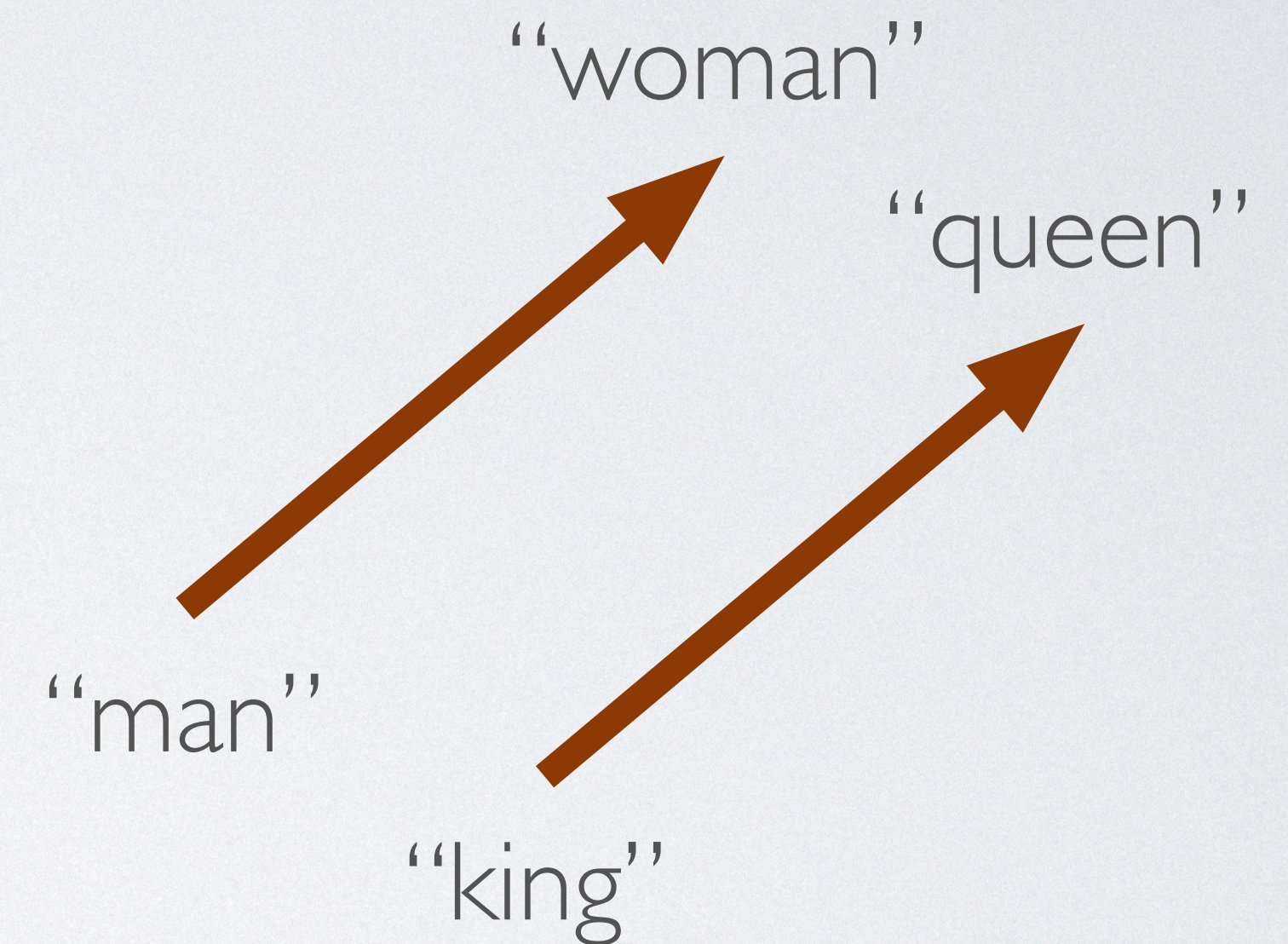
“High-level concepts are represented linearly as *directions* in the representation space”



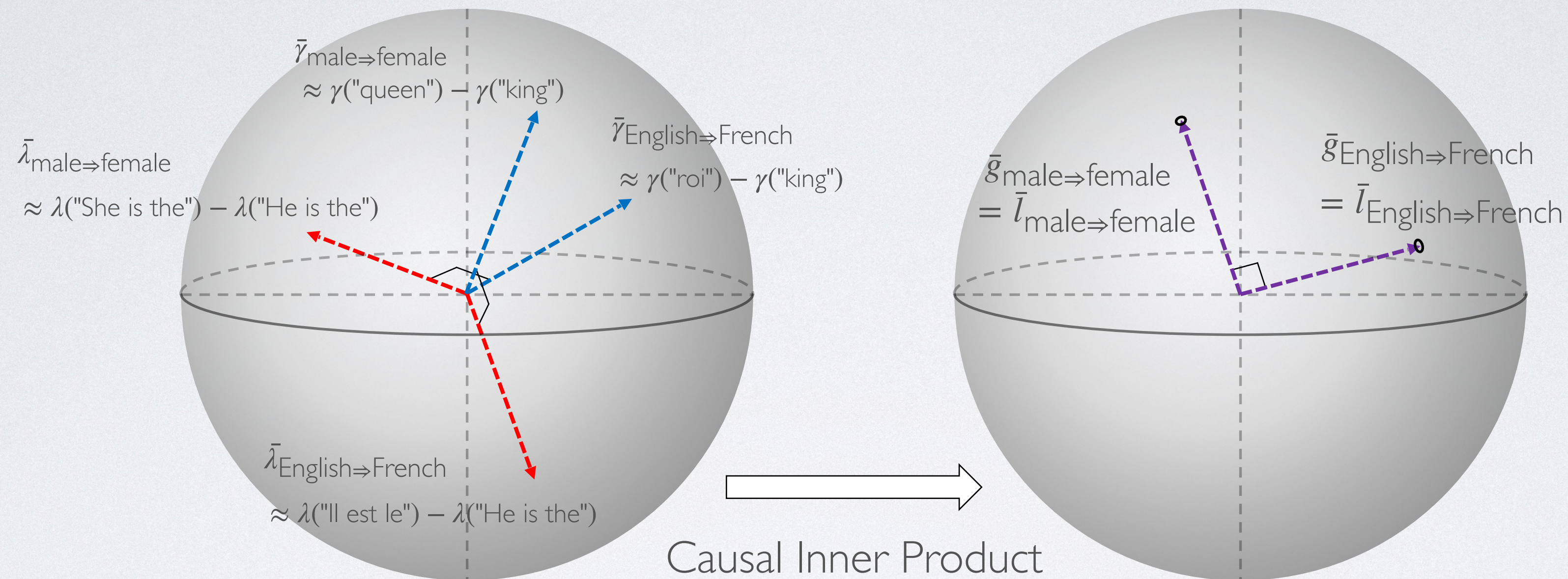
Background I. Linear Representation Hypothesis

“High-level concepts are represented linearly as *directions* in the representation space”

*But... a vector = direction + **magnitude***



Background 2. Causal Inner Product



Embedding

$$l(x) \in \mathbb{R}^d$$

Softmax

$$\mathbb{P}(y \mid x) \propto \exp(l(x)^\top g(y))$$

Unembedding

$$g(y) \in \mathbb{R}^d$$

Result 1. Vector Representations of Binary Features

Linear Representations of Binary Concepts

Desideratum: *If a linear representation exists as a direction, moving an embedding vector in this direction should modify the probability of the target concept **in isolation***

Result 1. Vector Representations of Binary Features



Logits: $l("I \text{ have } a")^T g(y) \xrightarrow{?} (l("I \text{ have } a") + \alpha \bar{l}_{animal})^T g(y)$

Result I. Vector Representations of Binary Features



Result I. Vector Representations of Binary Features



Result 1. Vector Representations of Binary Features

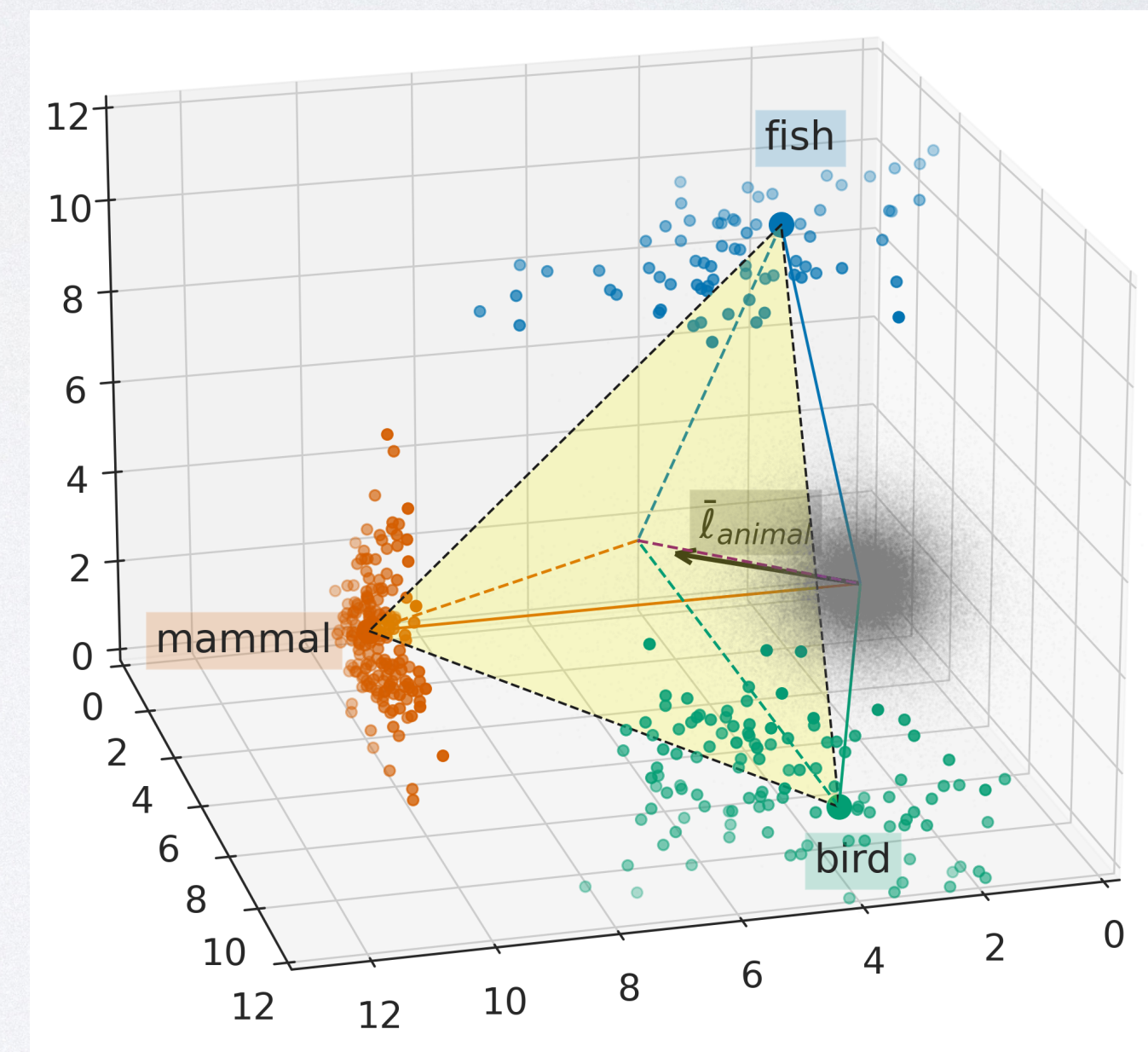
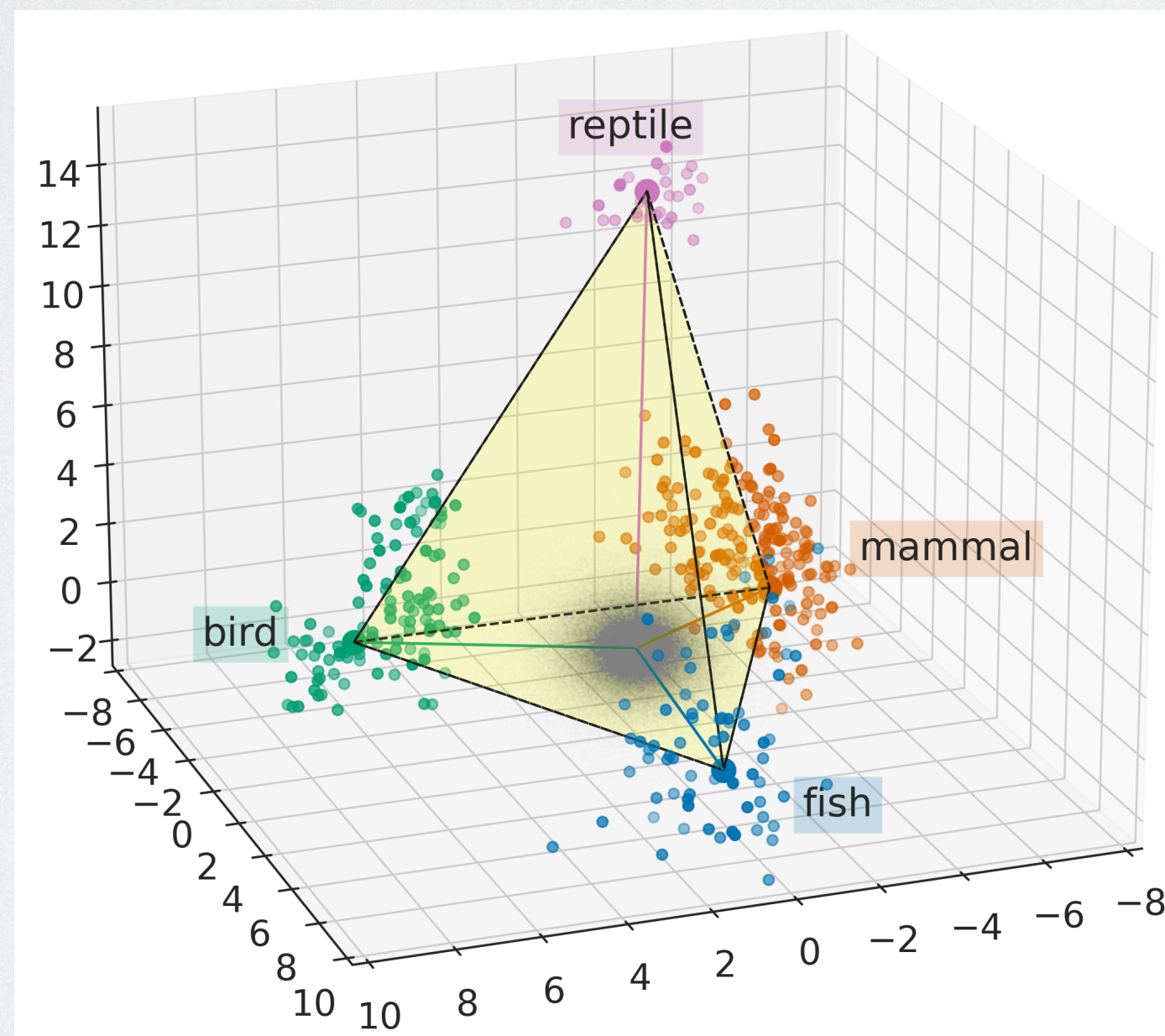


A binary feature $W = \{\text{not_w}, \text{is_w}\}$ has a **vector representation** \bar{l}_w if it is a linear representation for W with an associated *magnitude*.

Result 2. Polytope Representations of Categorical Concepts

Result 2. Polytope Representations of Categorical Concepts

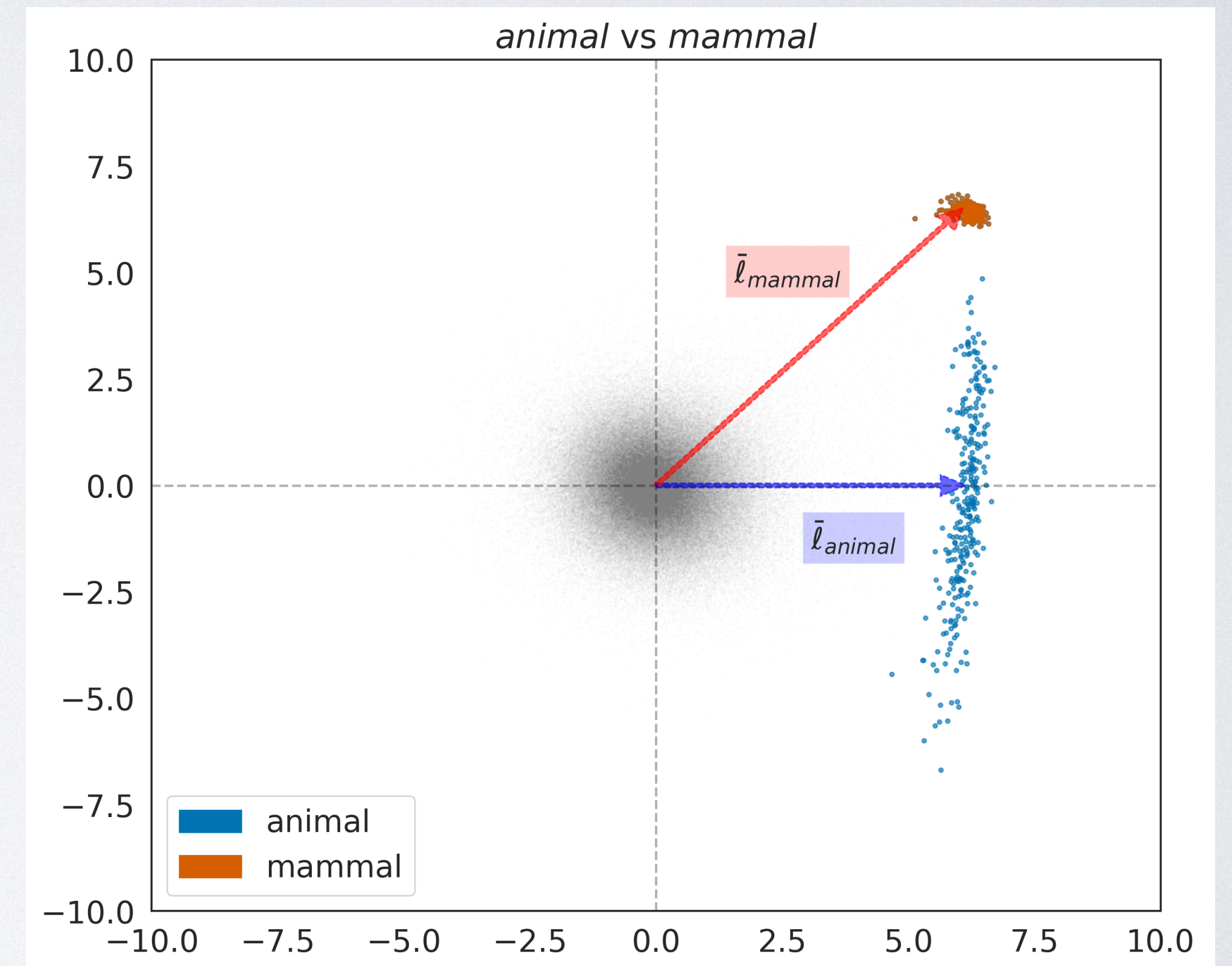
The **polytope representation** of a categorical concept $\{w_0, \dots, w_{k-1}\}$ is the convex hull of vector representations $\bar{l}_{w_0}, \dots, \bar{l}_{w_{k-1}}$ for each element.



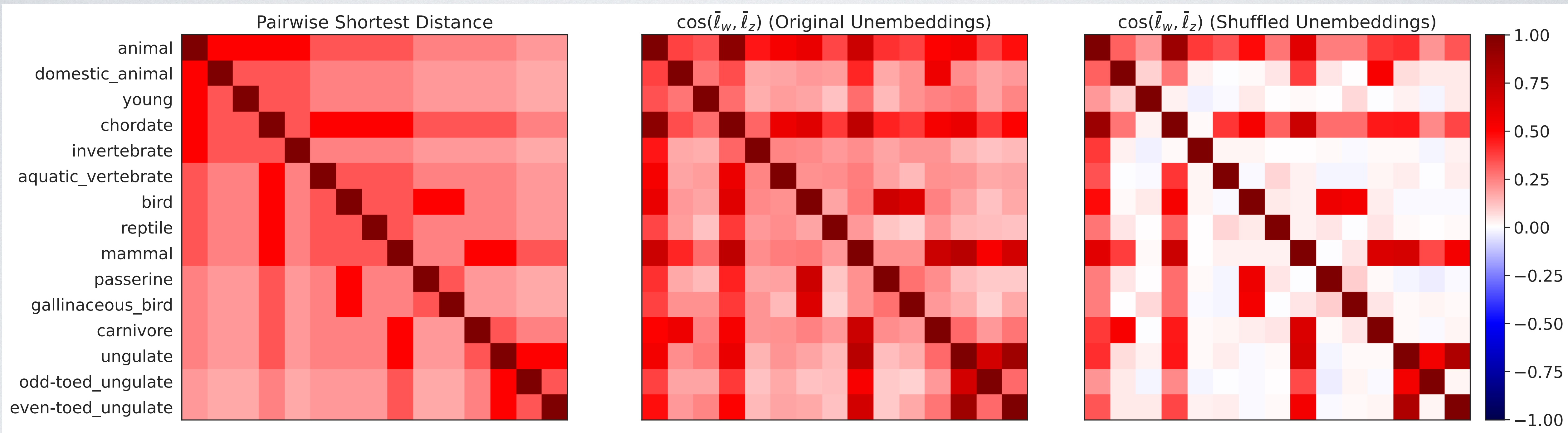
Result 3. Hierarchical Semantics Are Encoded as Orthogonality

Result 3. Hierarchical Semantics Are Represented As Orthogonality

- (a) $\bar{l}_w \perp \bar{l}_z - \bar{l}_w$ for $z < w$ (e.g., $\bar{l}_{animal} \perp \bar{l}_{mammal} - \bar{l}_{animal}$)
- (b) $\bar{l}_w \perp \bar{l}_{z_1} - \bar{l}_{z_0}$ for $Z \in_R \{z_0, z_1\} < W \in_R \{\text{not_}w, \text{is_}w\}$
- (c) $\bar{l}_{w_1} - \bar{l}_{w_0} \perp \bar{l}_{z_1} - \bar{l}_{z_0}$ for $Z \in_R \{z_0, z_1\} < W \in_R \{w_0, w_1\}$
- (d) $\bar{l}_{w_1} - \bar{l}_{w_0} \perp \bar{l}_{w_2} - \bar{l}_{w_1}$ for $w_2 < w_1 < w_0$



Result 3'. Cosine Similarities Between Vector Representations Capture Their Semantic Relations

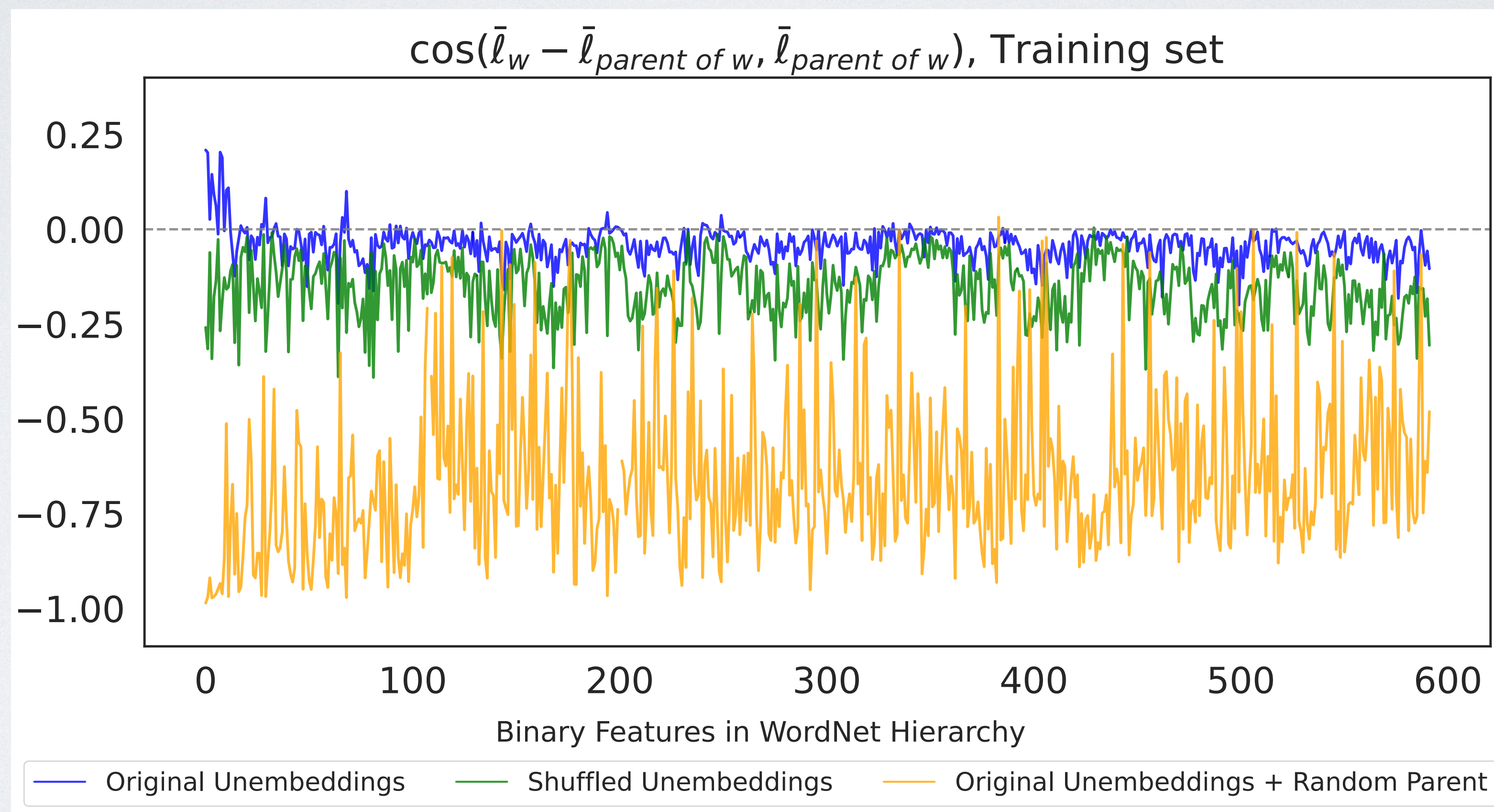


Shortest-path distances between features in WordNet hierarchy

Cosine similarities between vector representations

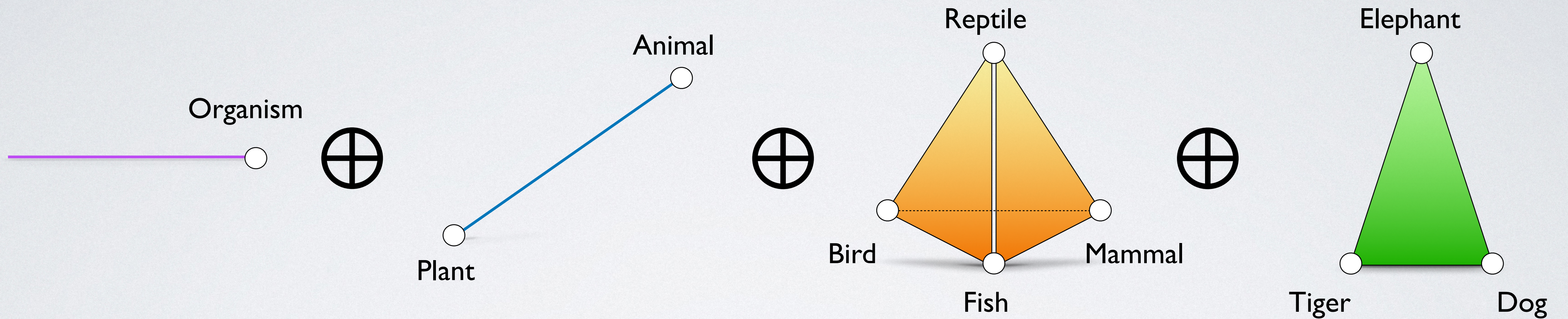
Cosine similarities between *shuffled* vector representations

Result 3'’. Validating Hierarchical Orthogonality on the Full WordNet Hierarchy

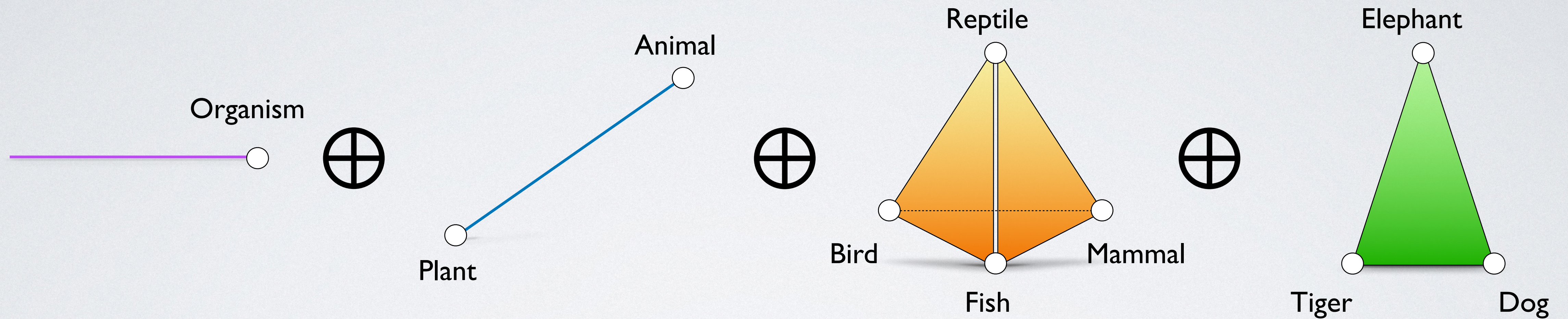


*How is semantic meaning encoded in
the representation spaces of LLMs?*

How is semantic meaning encoded in the representation spaces of LLMs?



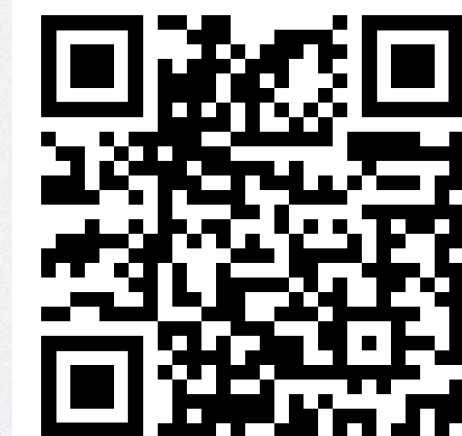
How is semantic meaning encoded in the representation spaces of LLMs?



The Geometry of Categorical and Hierarchical Concepts in Large Language Models

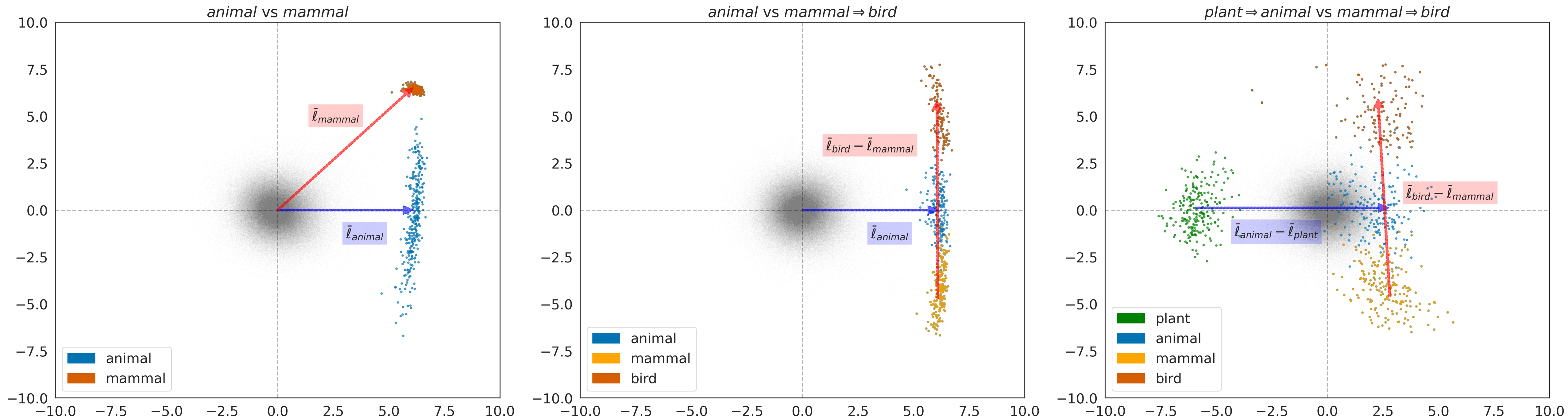
Kiho Park, Yo Joong Choe, Yibo Jiang, Victor Veitch

arXiv:2406.01506



Appendix

Result 3. Hierarchical Semantics Are Represented As Orthogonality

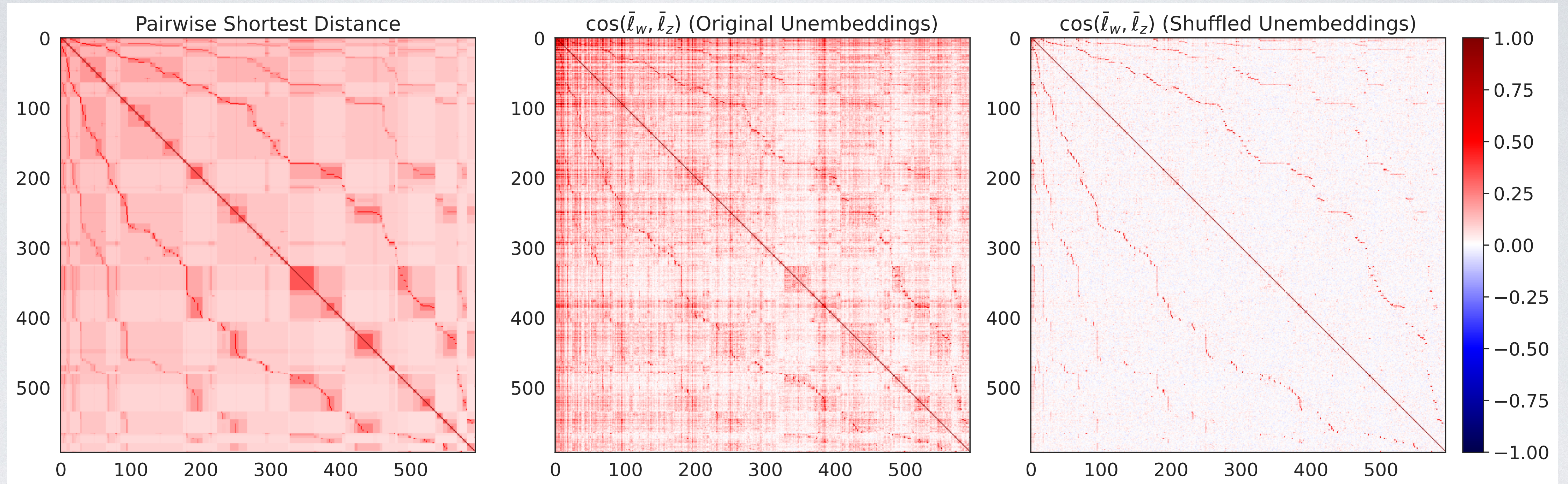


(a) $\bar{l}_w \perp \bar{l}_z - \bar{l}_w$ for $z < w$
(e.g., $\bar{l}_{animal} \perp \bar{l}_{mammal} - \bar{l}_{animal}$)

(b) $\bar{l}_w \perp \bar{l}_{z_1} - \bar{l}_{z_0}$ for
 $Z \in_R \{z_0, z_1\} < W \in_R \{\text{not_}w, \text{is_}w\}$

(c) $\bar{l}_{w_1} - \bar{l}_{w_0} \perp \bar{l}_{z_1} - \bar{l}_{z_0}$ for
 $Z \in_R \{z_0, z_1\} < W \in_R \{w_0, w_1\}$

Result 3'. Cosine Similarities Between Vector Representations Capture Their Semantic Relations

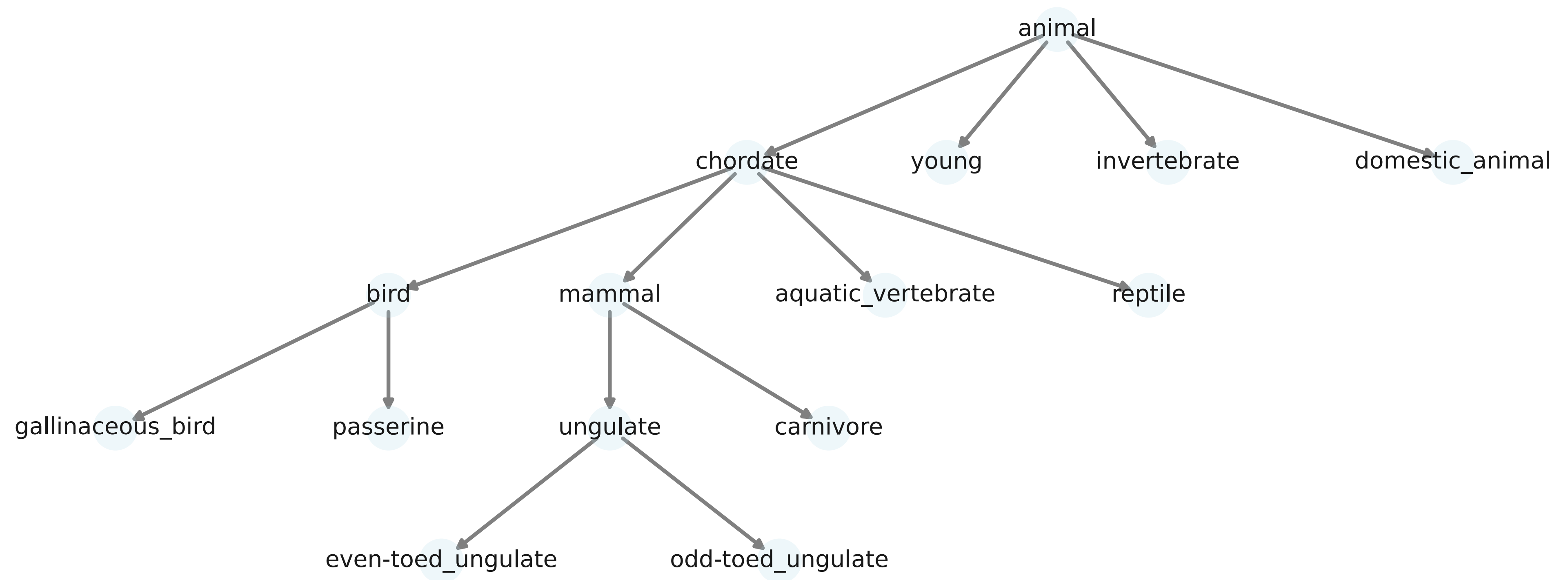


Shortest-path distances between features in WordNet hierarchy

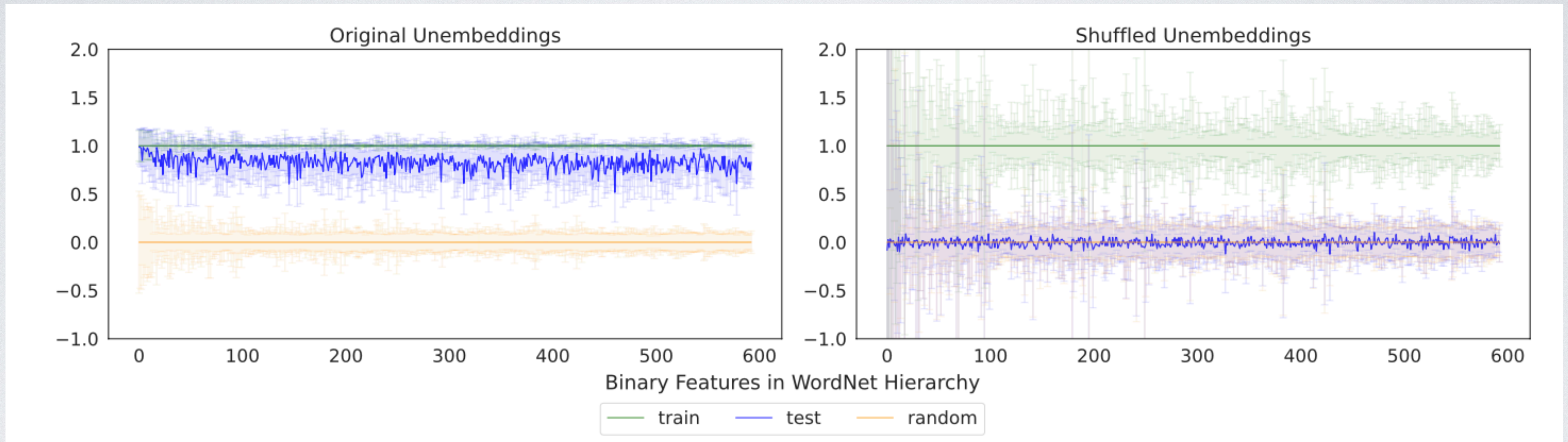
Cosine similarities between vector representations

Cosine similarities between *shuffled* vector representations

A Subgraph of the WordNet Hierarchy



Existence of Vector Representations



Theoretical Prediction: $(\bar{l}_w)^\top g(y) / \|\bar{l}_w\|^2 = 1$ for any $y \in \mathcal{Y}(w)$.

Proof Sketch for the “Magnitude Theorem”

Theorem: Linear Representations have Magnitude

If ℓ_W is a linear representation of binary feature W then there is some $b_W > 0$ such that

$$\bar{\ell}_W^\top g(y) = \begin{cases} b_W & \text{if } y \in \mathcal{Y}(W) \\ 0 & \text{if } y \notin \mathcal{Y}(W) \end{cases}$$

Proof Sketch

Adding $\bar{\ell}_{\text{animal}}$ shouldn't change probability of “dog” vs “cat”.

Softmax gives:

$$\log \frac{P(\text{“dog”} \mid \ell(x) + \alpha \bar{\ell}_{\text{animal}})}{P(\text{“cat”} \mid \ell(x) + \alpha \bar{\ell}_{\text{animal}})} = \ell(x)^\top (g_{\text{dog}} - g_{\text{cat}}) + \alpha \bar{\ell}_{\text{animal}}^\top (g_{\text{dog}} - g_{\text{cat}}).$$

This expression is free of α only if $\bar{\ell}_{\text{animal}}^\top g_{\text{dog}} = \bar{\ell}_{\text{animal}}^\top g_{\text{cat}}$.

The End