

Topology of Language

Topological Techniques for Natural Language Processing

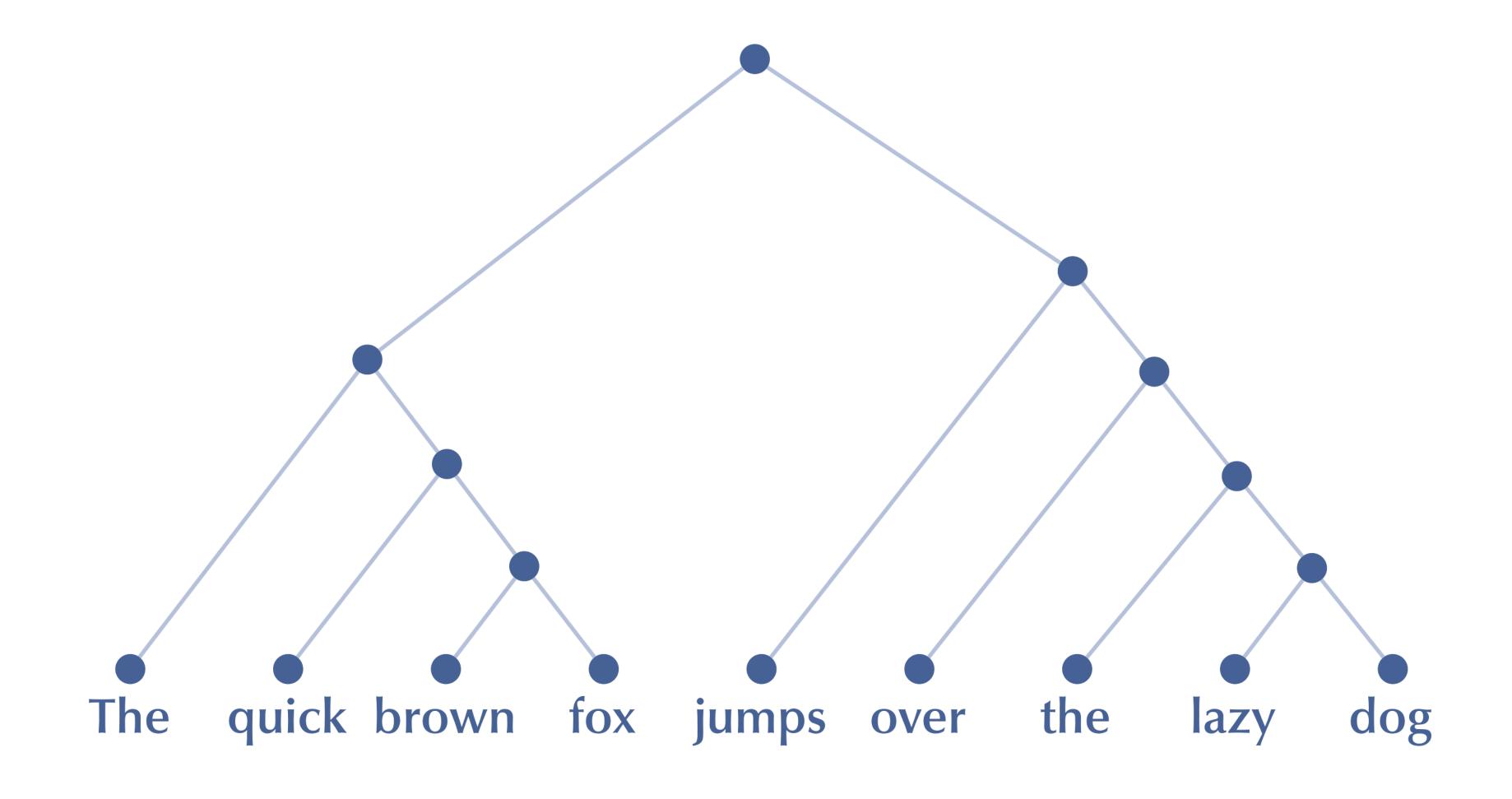


Vision



Language has structure

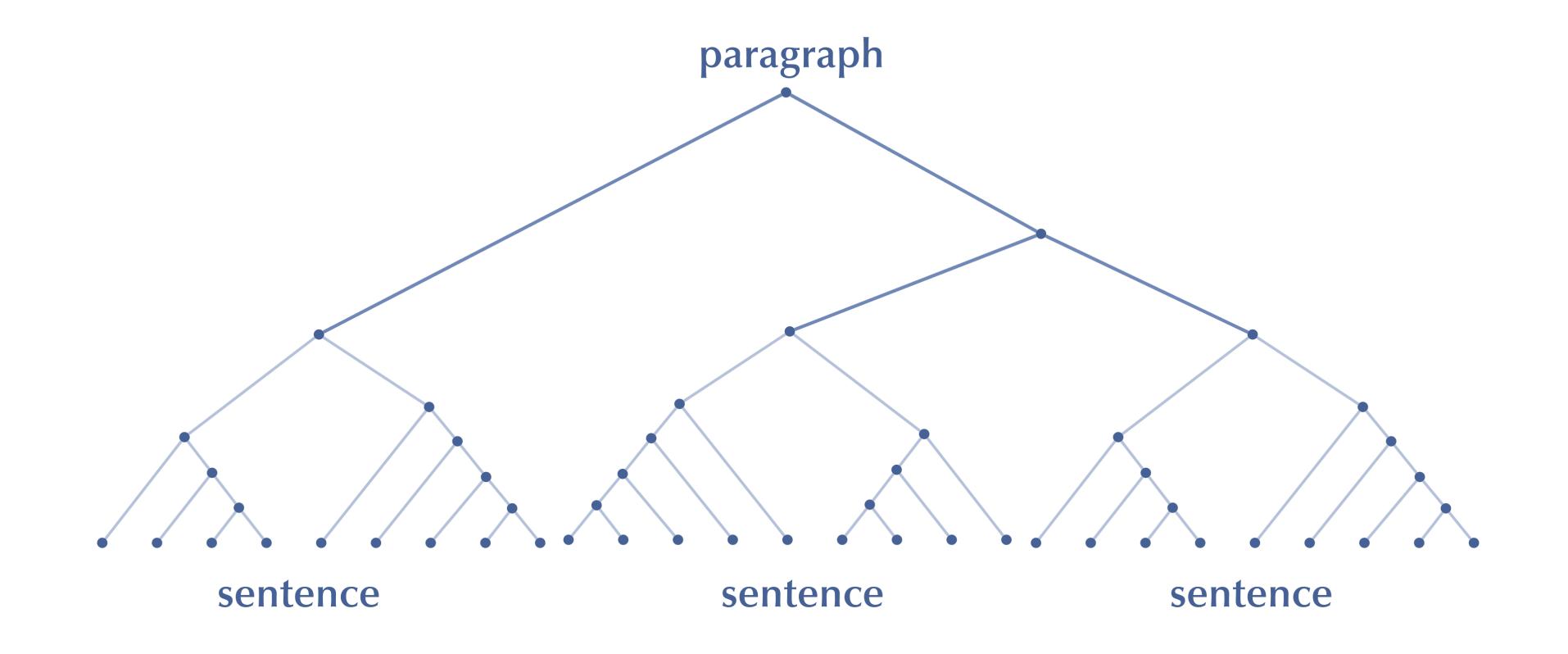




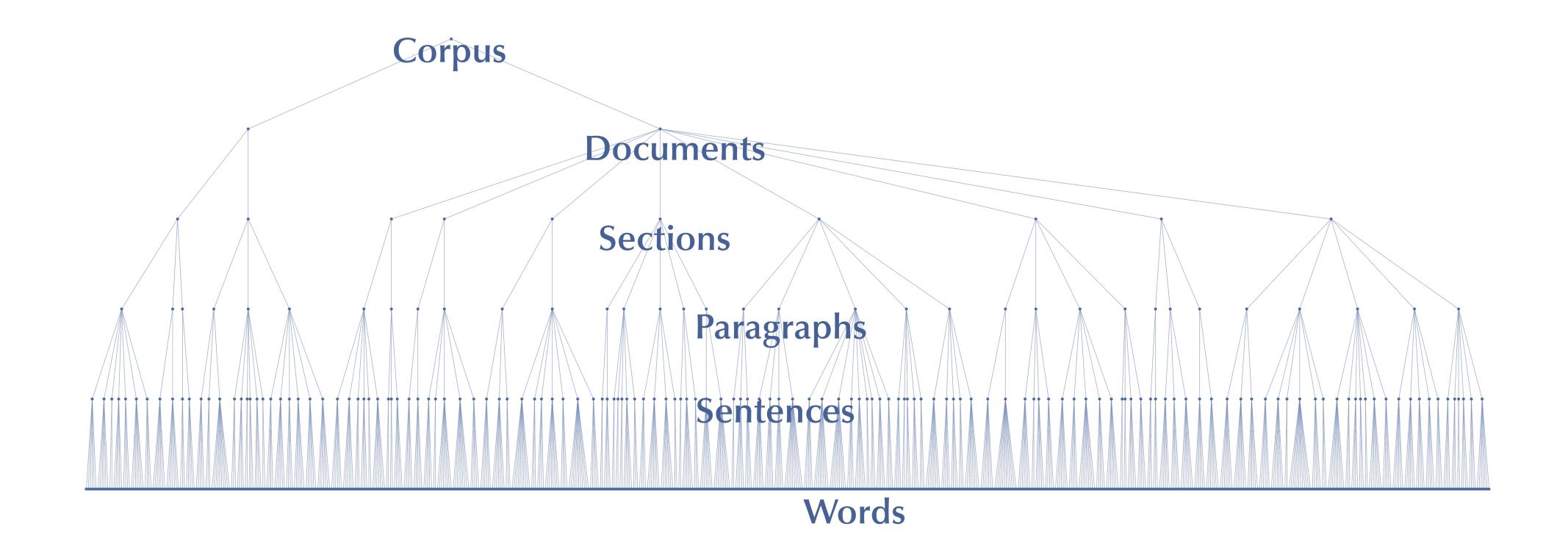


Many layers of structure



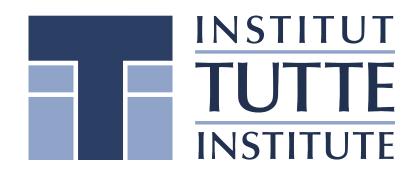


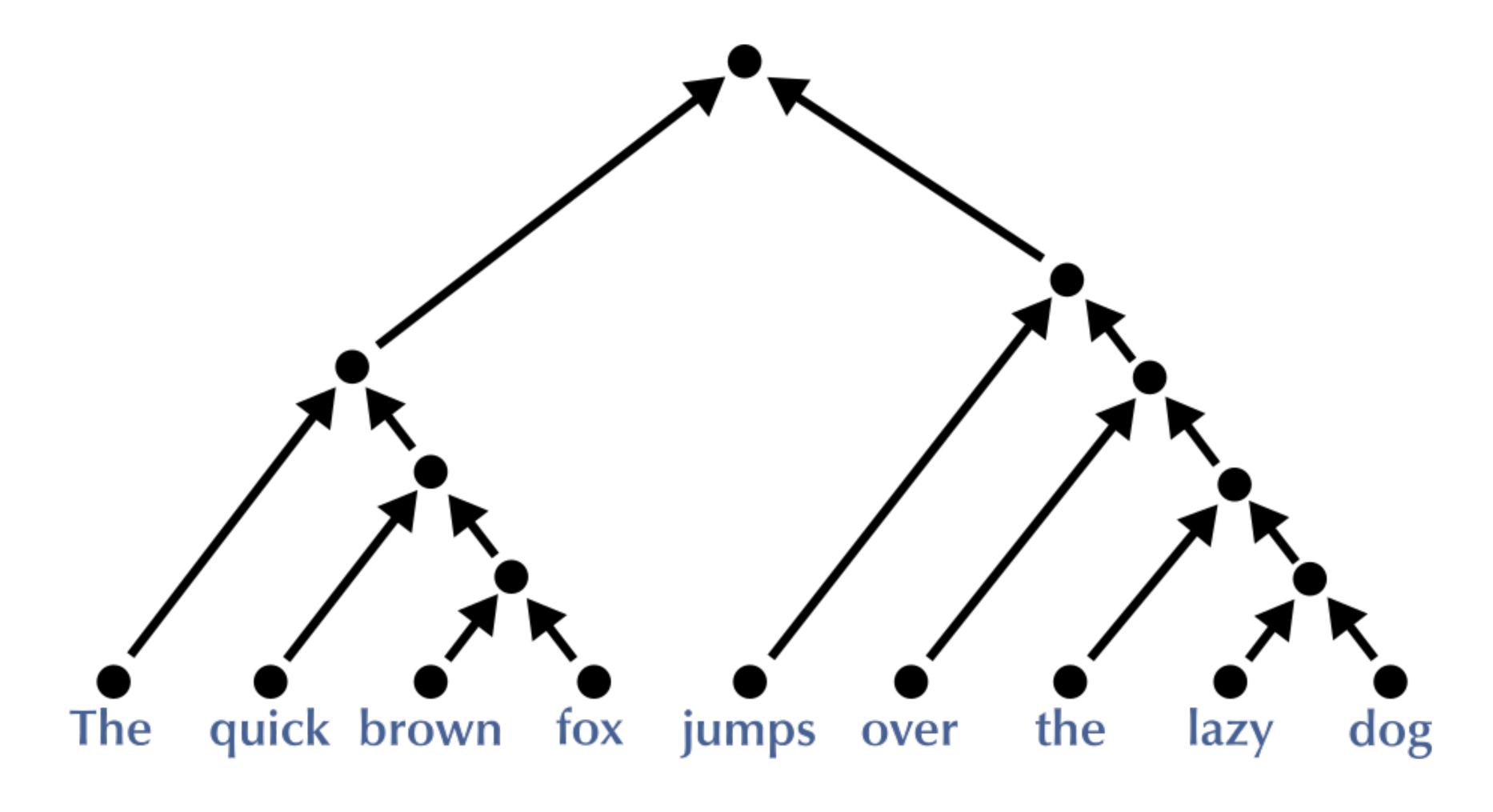






We could build a category from this







Yoneda embedding of a document?

All the constituent parts of the document



Simplify: all the word usages in a document

This is the classical "bag-of-words" model



What about words? In the opposite category...



A word is the set of all its usages in a corpus



Simplify: all the sentences that use the word



You shall know a word by the company it keeps — John Rupert Firth



Words as Distributions



Overview



Represent a word as a document of all the sentences that use the word



Represent such a document using a bag-of-words model



Represent a word as the multinomial distribution of words that occur nearby



Information geometry tells us that multinomial distributions live on a manifold



$$d(p,q) = \arccos\left(\sum_{i} \sqrt{p_i} \sqrt{q_i}\right)$$



We can explore the geometry of words under this distance metric



Messy Details



Some problems:

- Word order matters
- Words carry different amounts of information
- Language use is noisy



Words order matters:

Words used *before* a given word Words used *after* a given word



Information in words:

Approximate the information of a word in a given context and weight words accordingly



Language is noisy:

Approximate a background noise model of word use and decompose the multinomial into a mixture of signal and noise distributions



Where do we get models of entropy and noise?



Where

X is an $N \times D$ matrix

U is an $N \times d$ matrix

V is an $d \times D$ matrix



Minimize

$$\sum_{i=1}^{N} \sum_{j=1}^{D} \text{Loss} \left(X_{ij}, (UV)_{ij} \right)$$

Subject to constraints...



Suppose

 $X \sim \Pr(\cdot \mid \Theta)$

where

 $\Theta = UV$



Let the loss be the negative log likelihood of observing X given Θ



Multinomial Matrix Factorization

Minimize

$$\sum_{i=1}^{N} \sum_{j=1}^{D} - (UV)_{ij} \cdot \log \left(X_{ij}\right)$$

Subject to

$$(UV)1 = 1 \text{ and } (UV)_{ij} \ge 0$$



Probabilistic Latent Semantic Analysis

Minimize

$$\sum_{i=1}^{N} \sum_{j=1}^{D} - (UV)_{ij} \cdot \log \left(X_{ij}\right)$$

Subject to

$$U1 = 1, V1 = 1 \text{ and } U_{ij} \ge 0, V_{ij} \ge 0$$



Θ provides a low rank multinomial distribution model



 Θ_{ij} is the "background" probability of word j occurring in the context of word i

 $-\log(\Theta_{ij})$ is an approximation of the information carried by word j in the context of word i



We can weight each entry by the (approximate) information it carries



A rank 1 model gives classical TF-IDF from NLP



 Θ_i provides a model of the "background" multinomial of word occurrence in the context of word i



We can use an EM algorithm to decompose a given multinomial into noise and signal



We are back to where our overview left off...



Embedding Words



We have high quality and high dimensional representations of words with a natural ambient geometry ...



We need a low dimensional representation that preserves topology

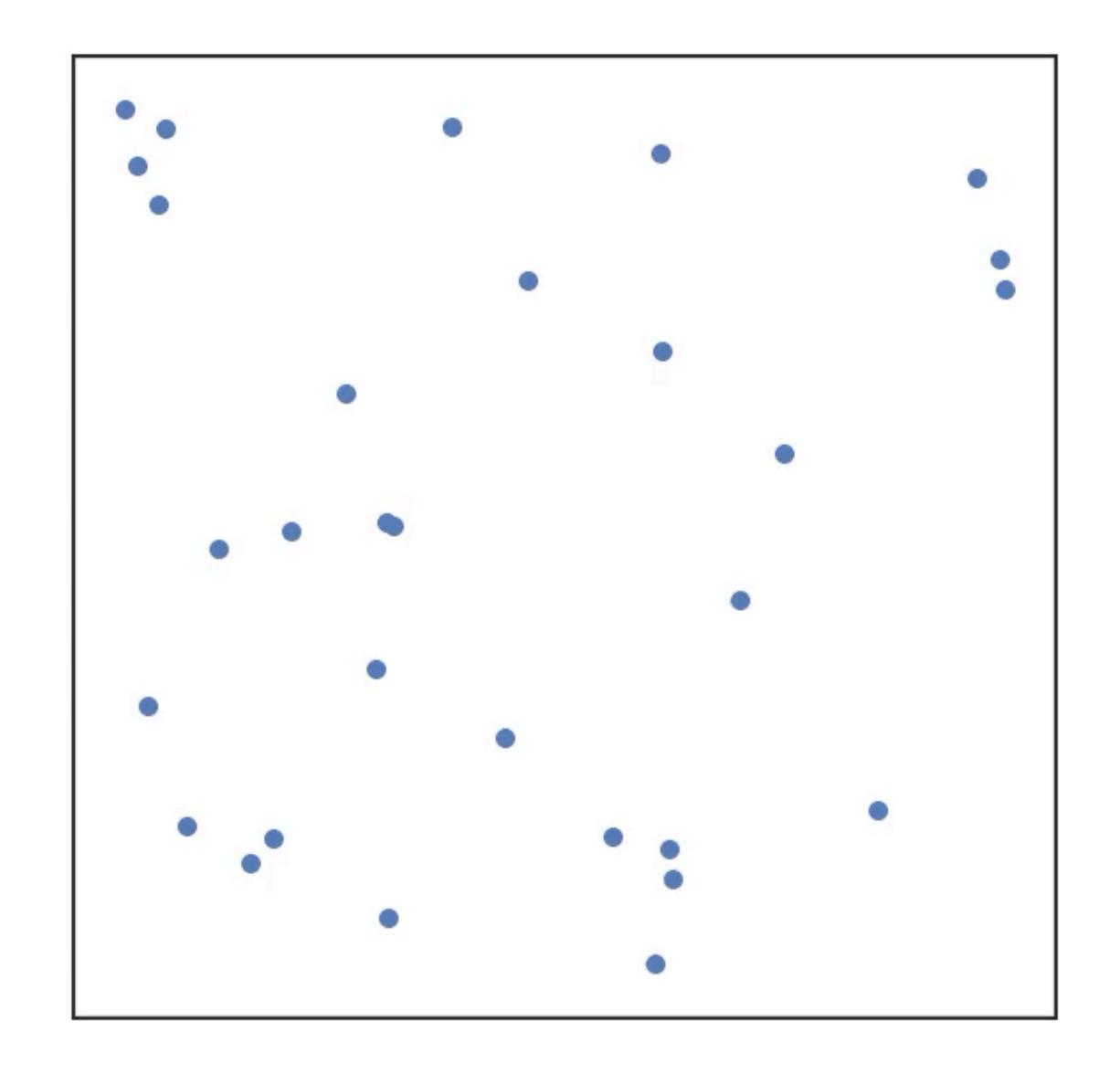




UMAP is a topology based dimension reduction algorithm



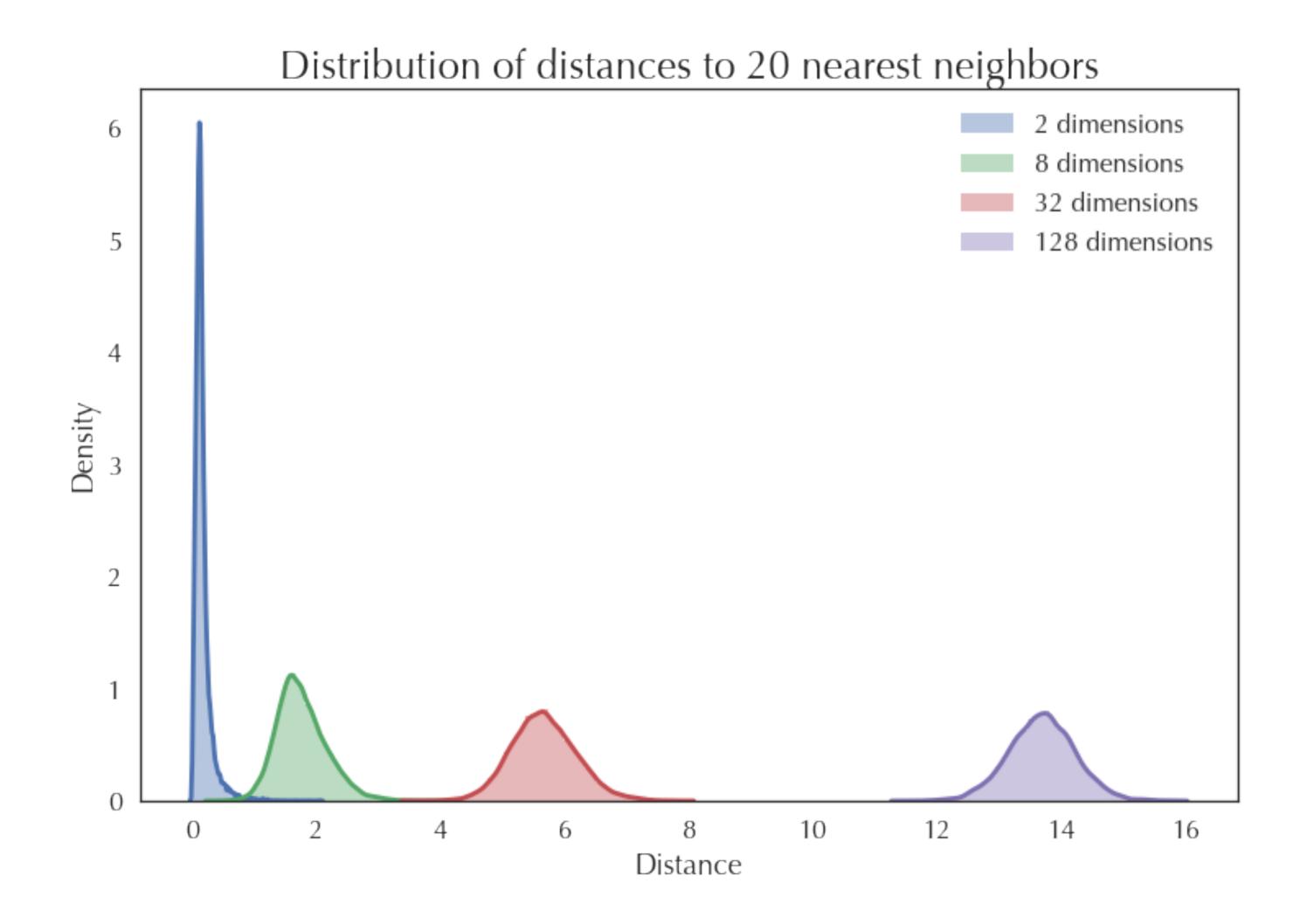
Assumptions: Uniform distribution Locally connected



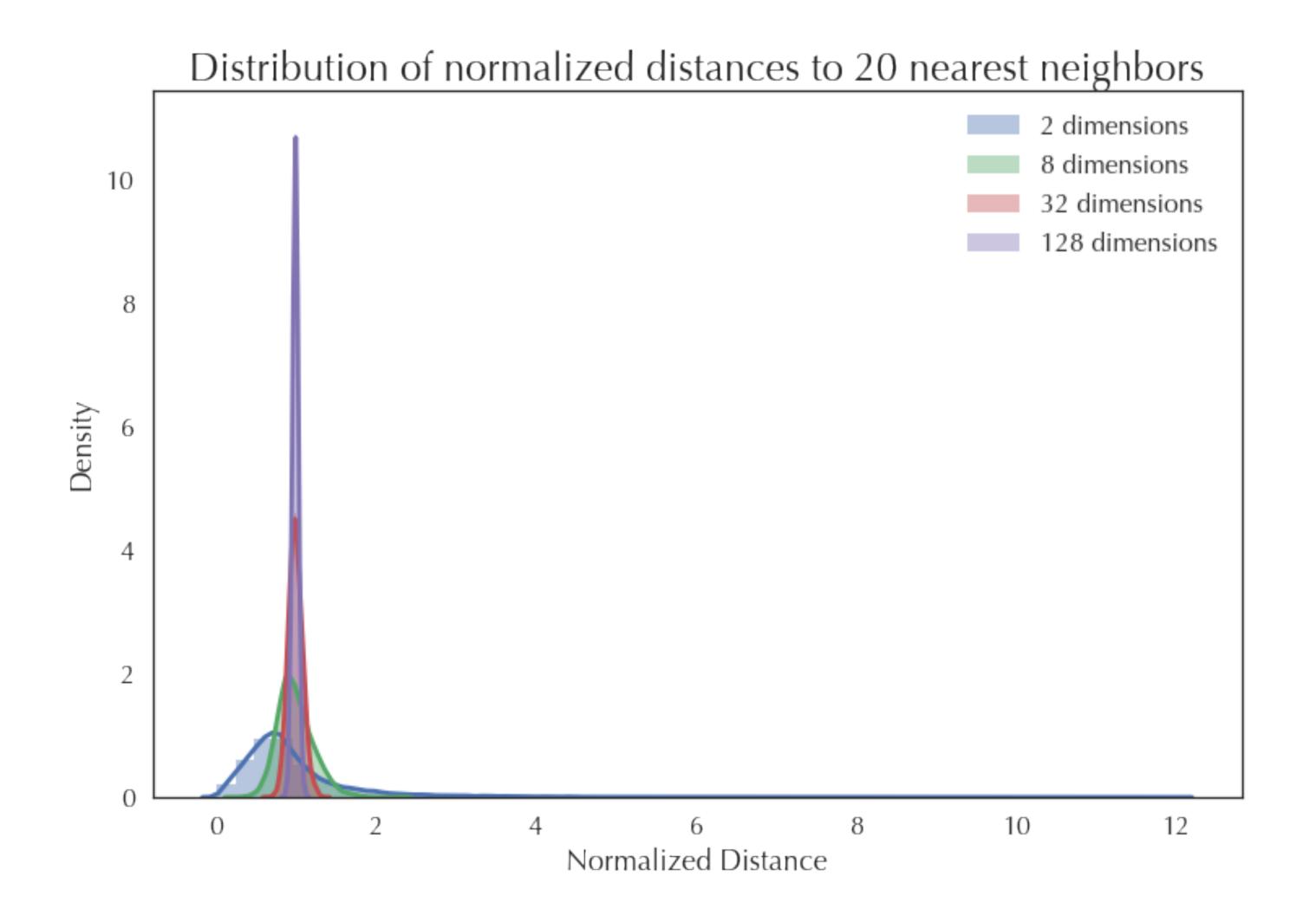


Assumptions: Uniform distribution Locally connected



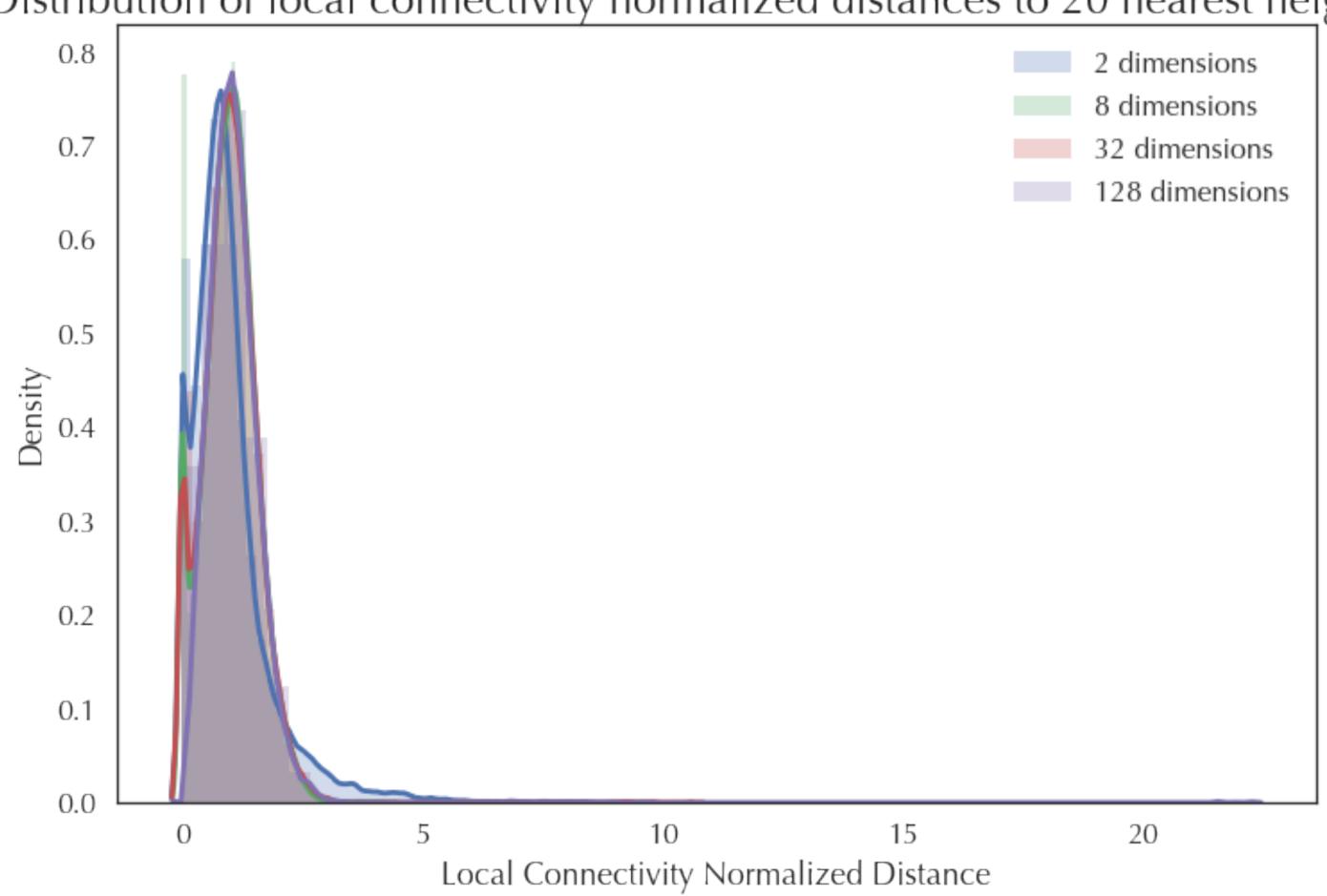














We now have a filtered simplicial complex



Optimize a low dimensional representation via cross-entropy



$$\sum_{a} \mu(a) \log \left(\frac{\mu(a)}{\nu(a)}\right) + (1 - \mu(a)) \log \left(\frac{1 - \mu(a)}{1 - \nu(a)}\right)$$

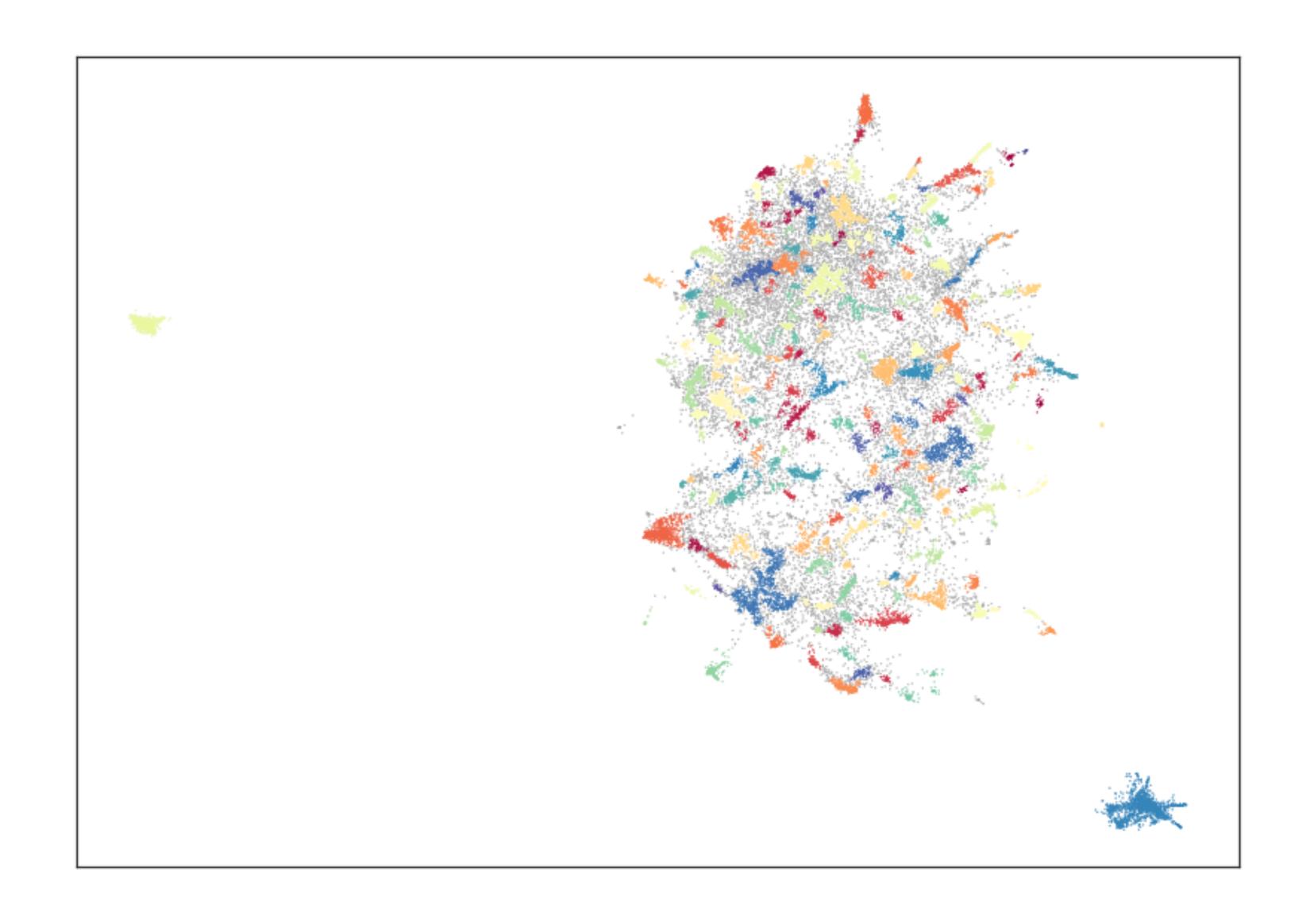


$$d(p,q) = \arccos\left(\sum_{i} \sqrt{p_i} \sqrt{q_i}\right)$$

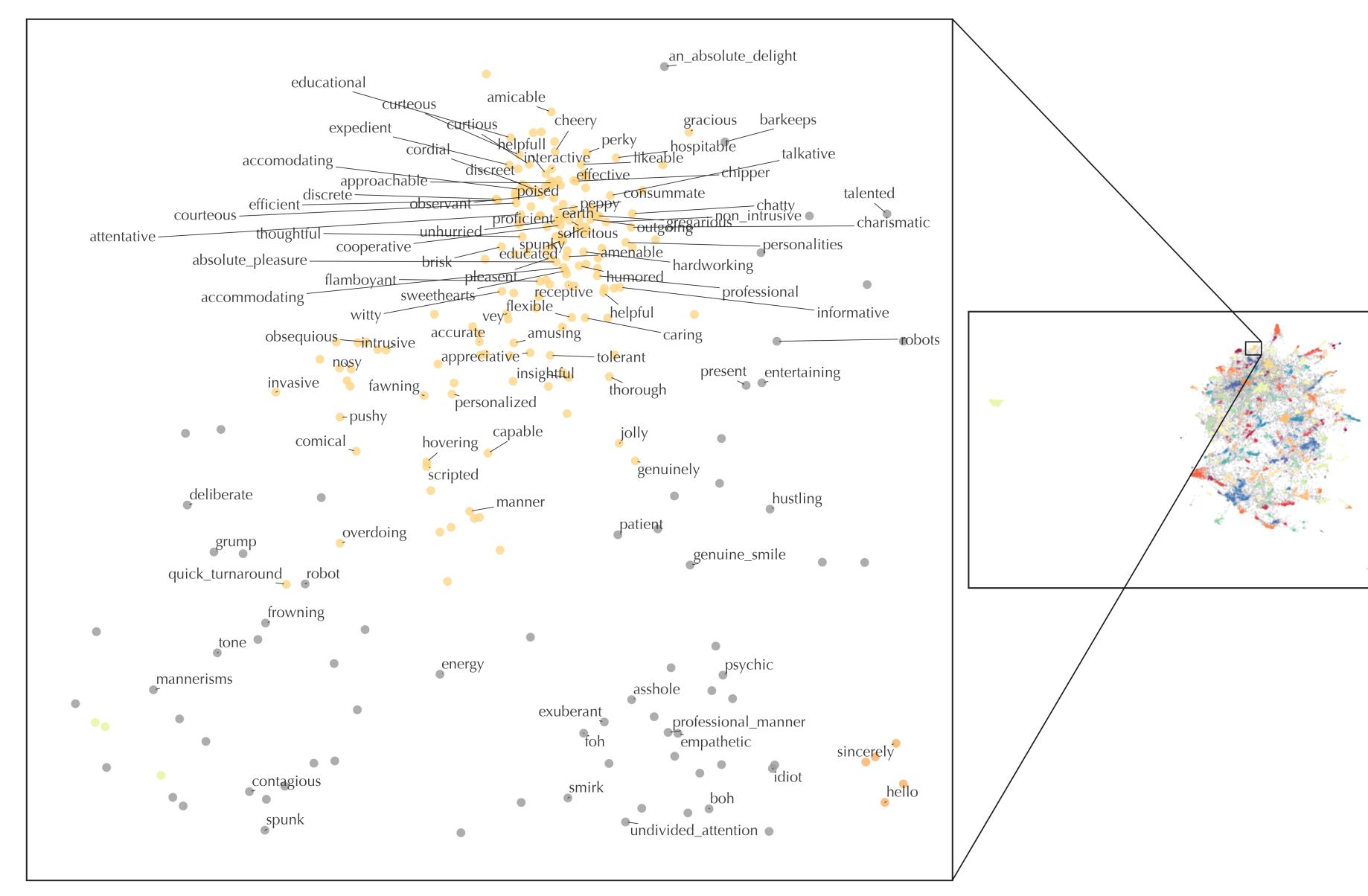


Embedding Yelp Reviews

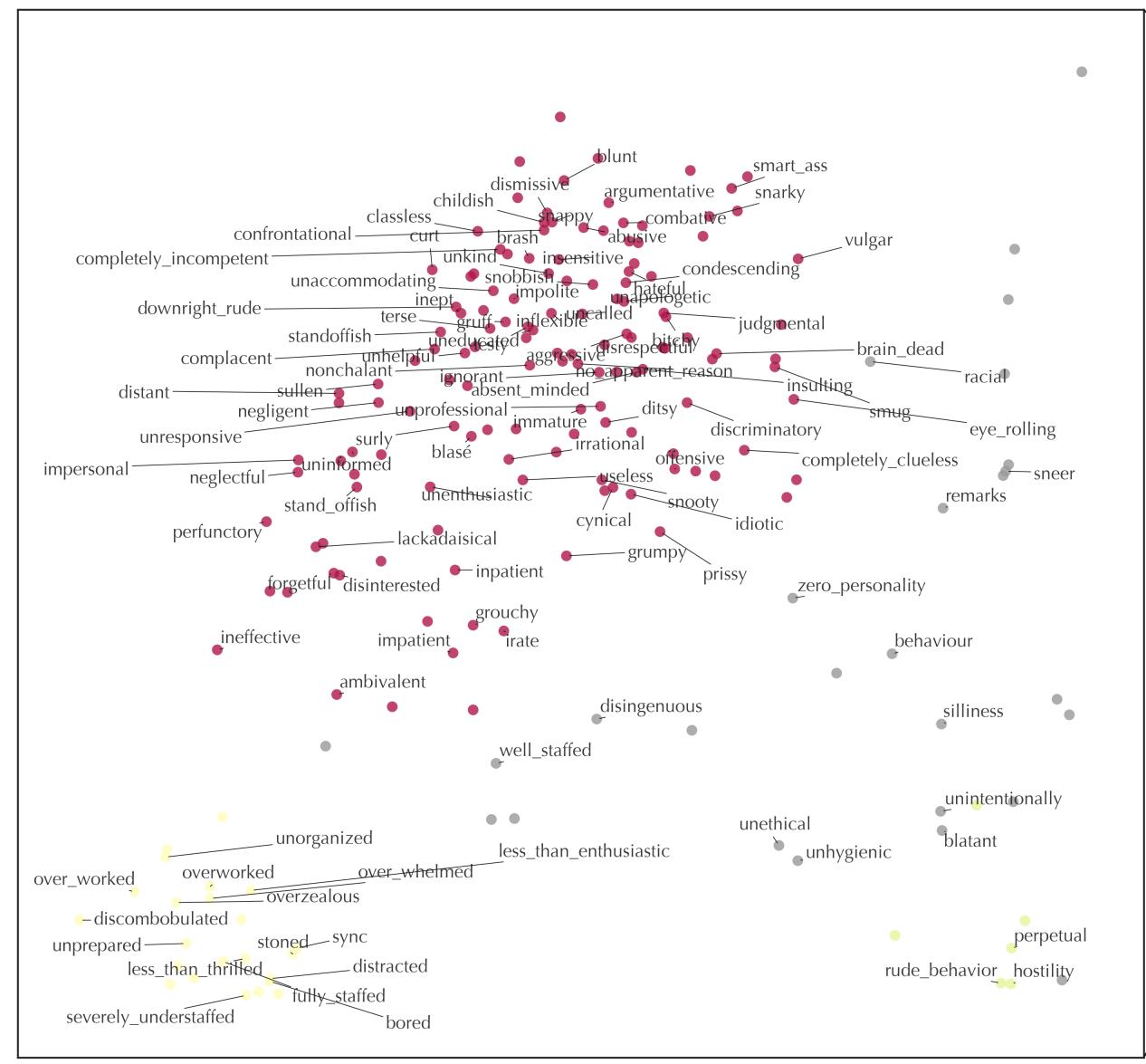


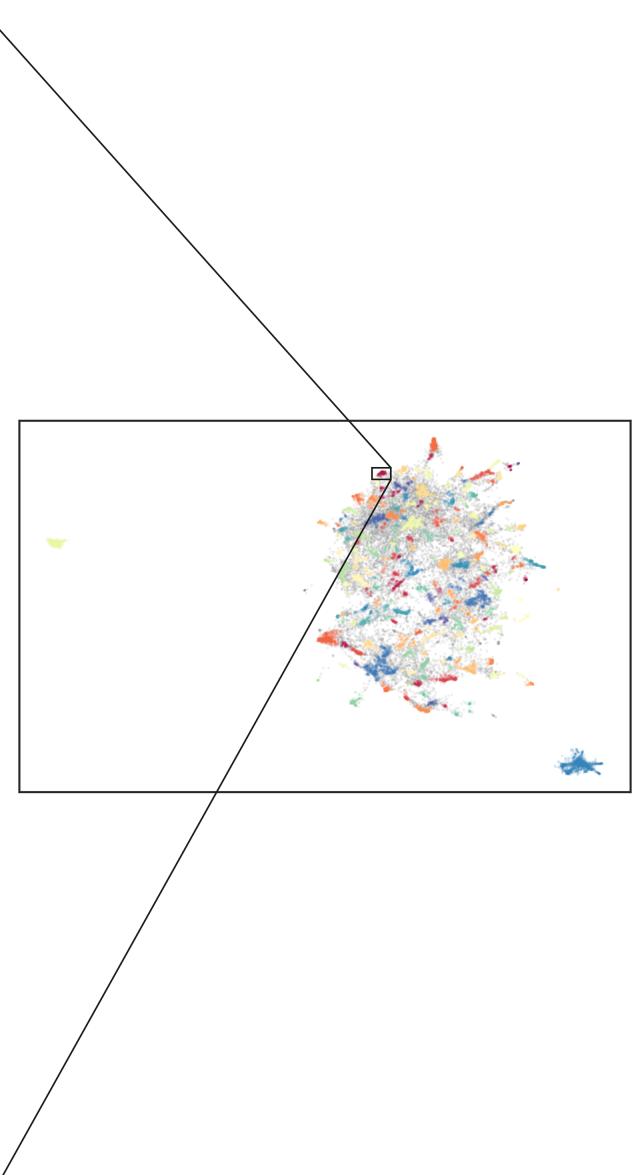




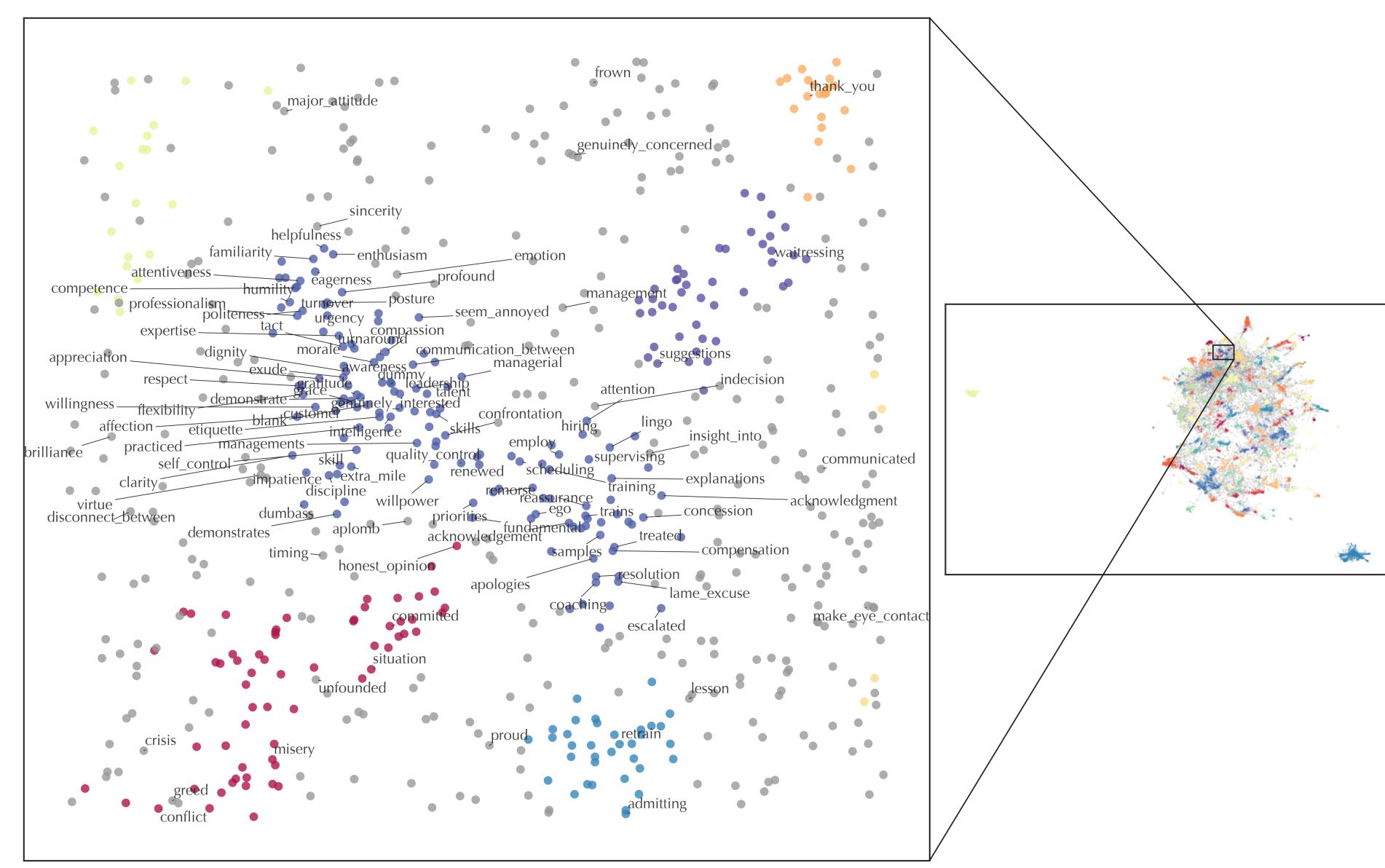




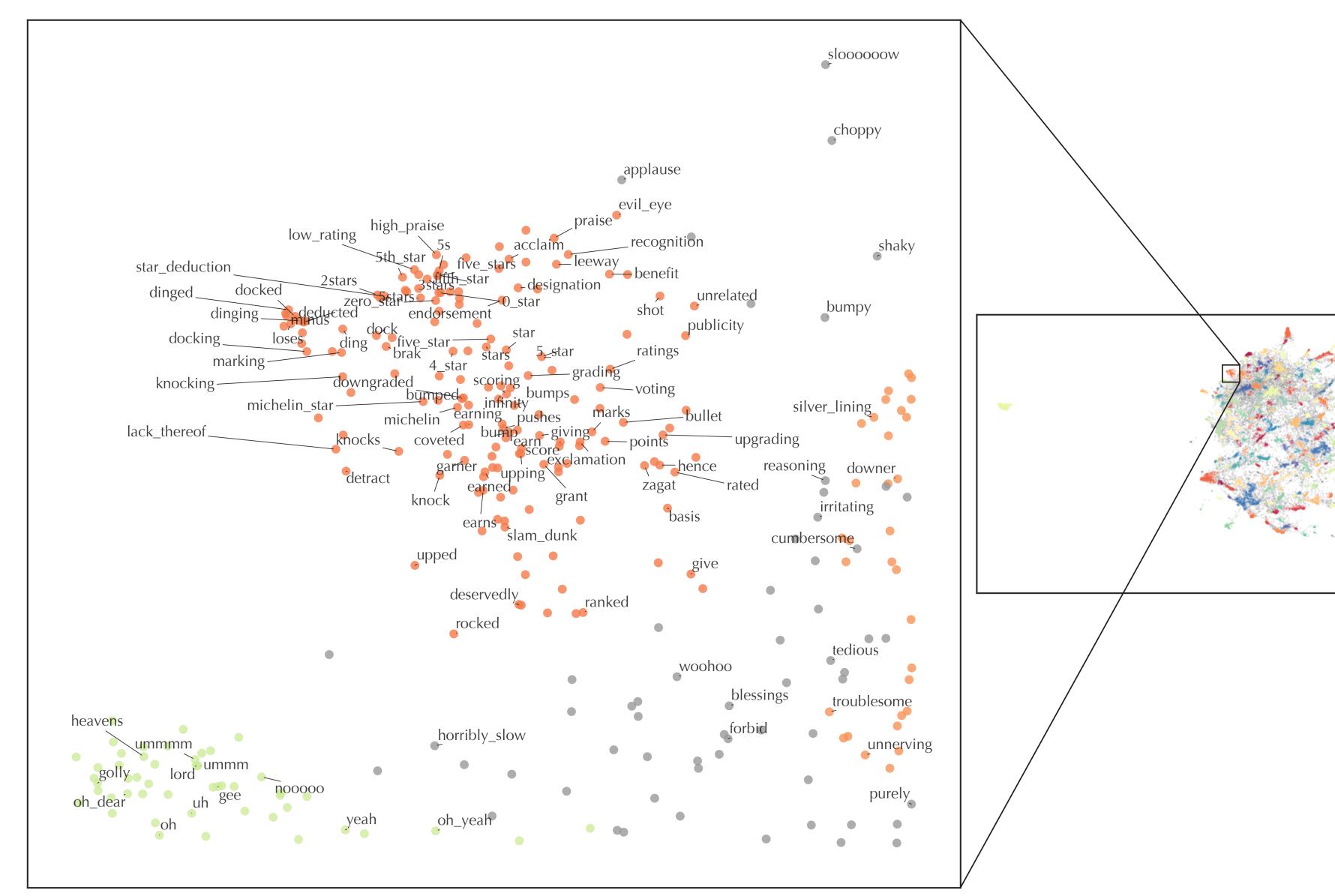




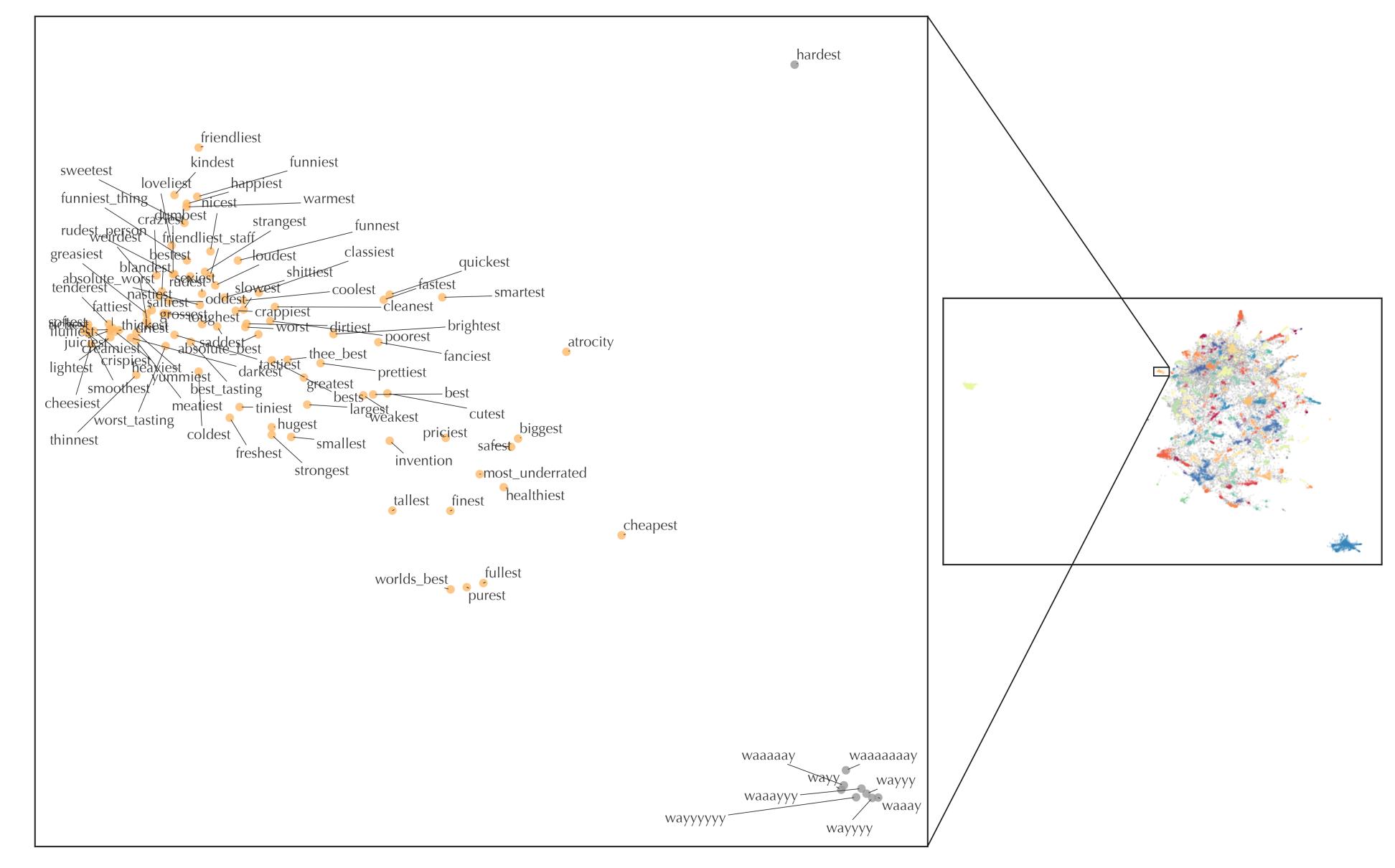




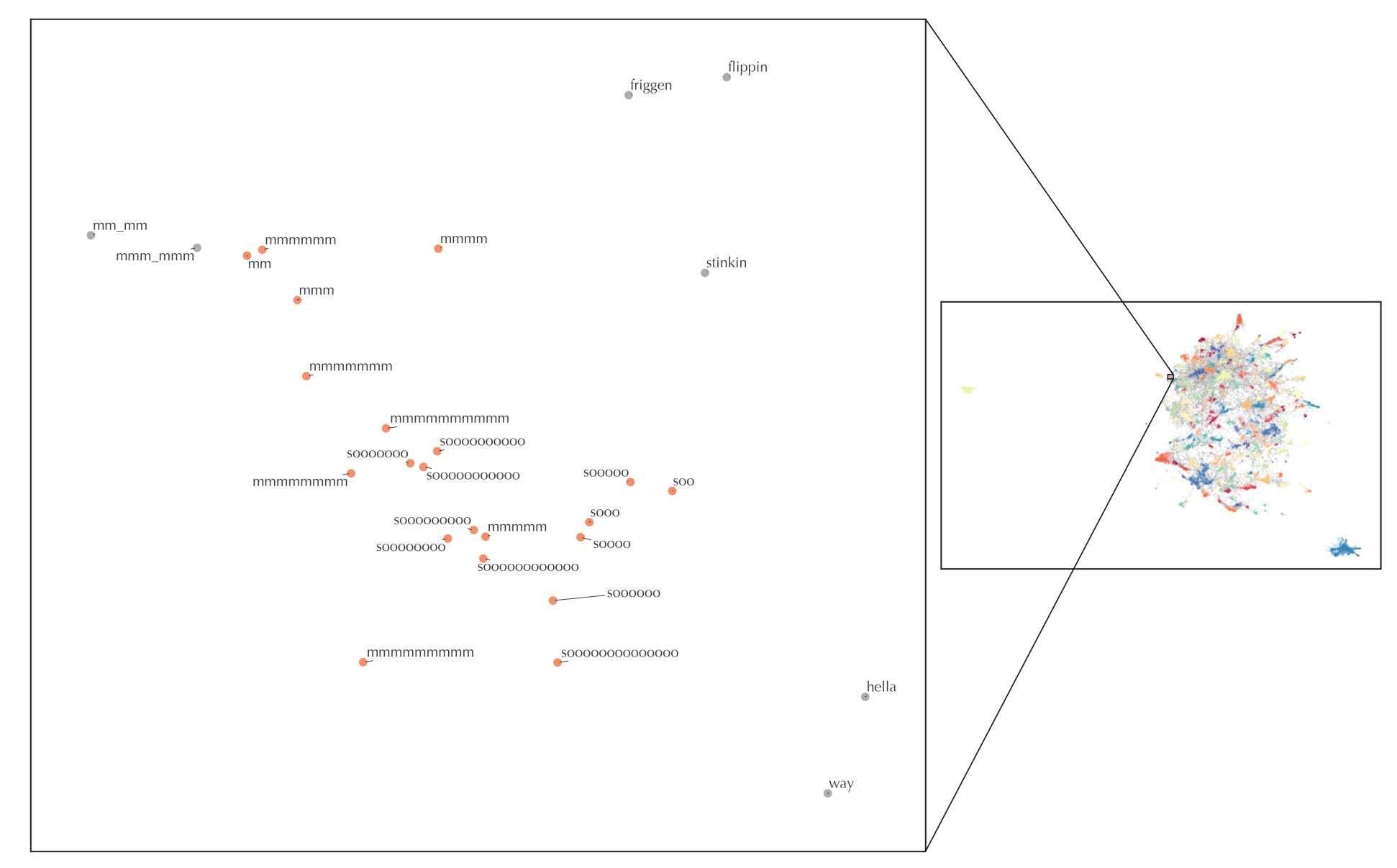




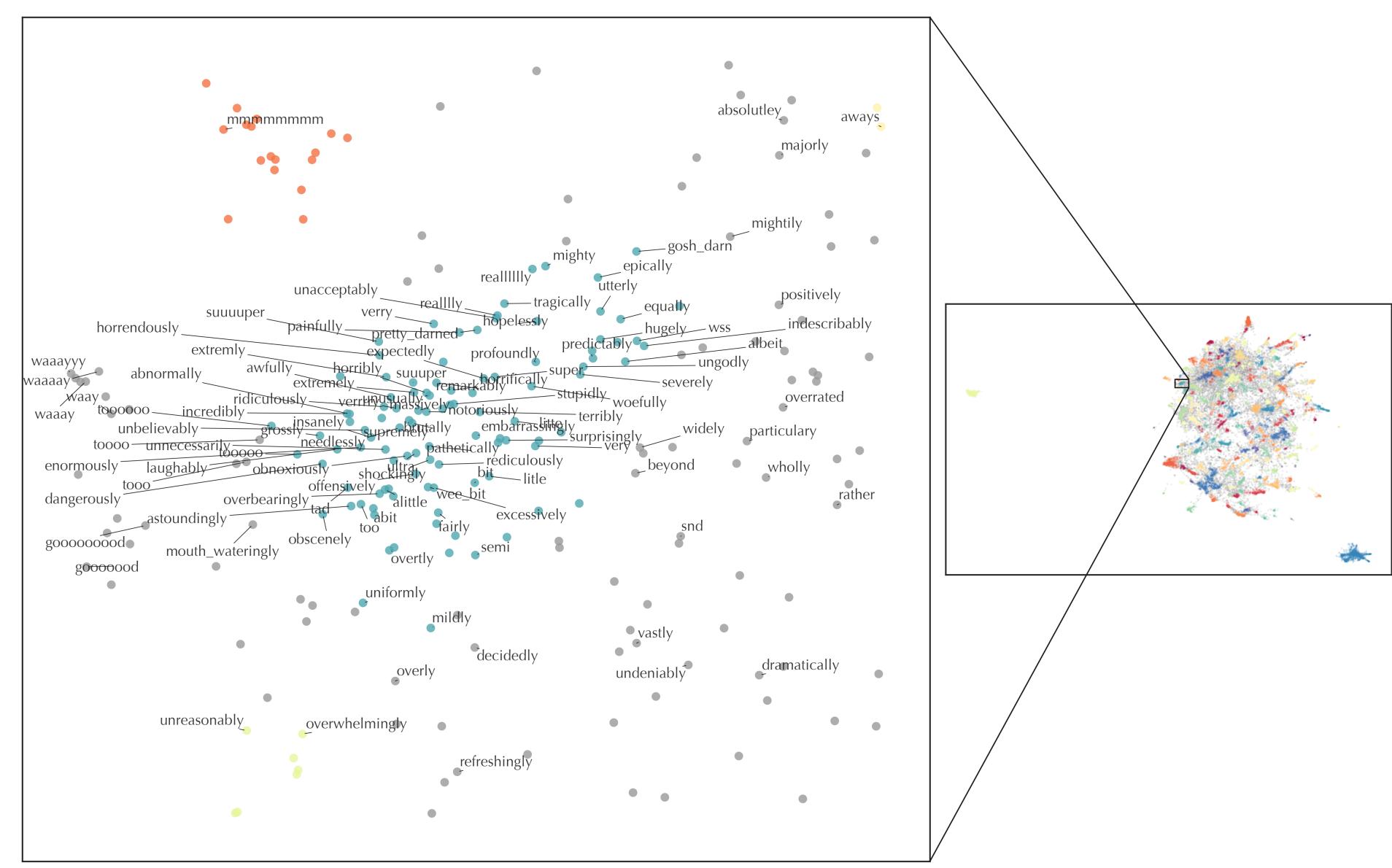




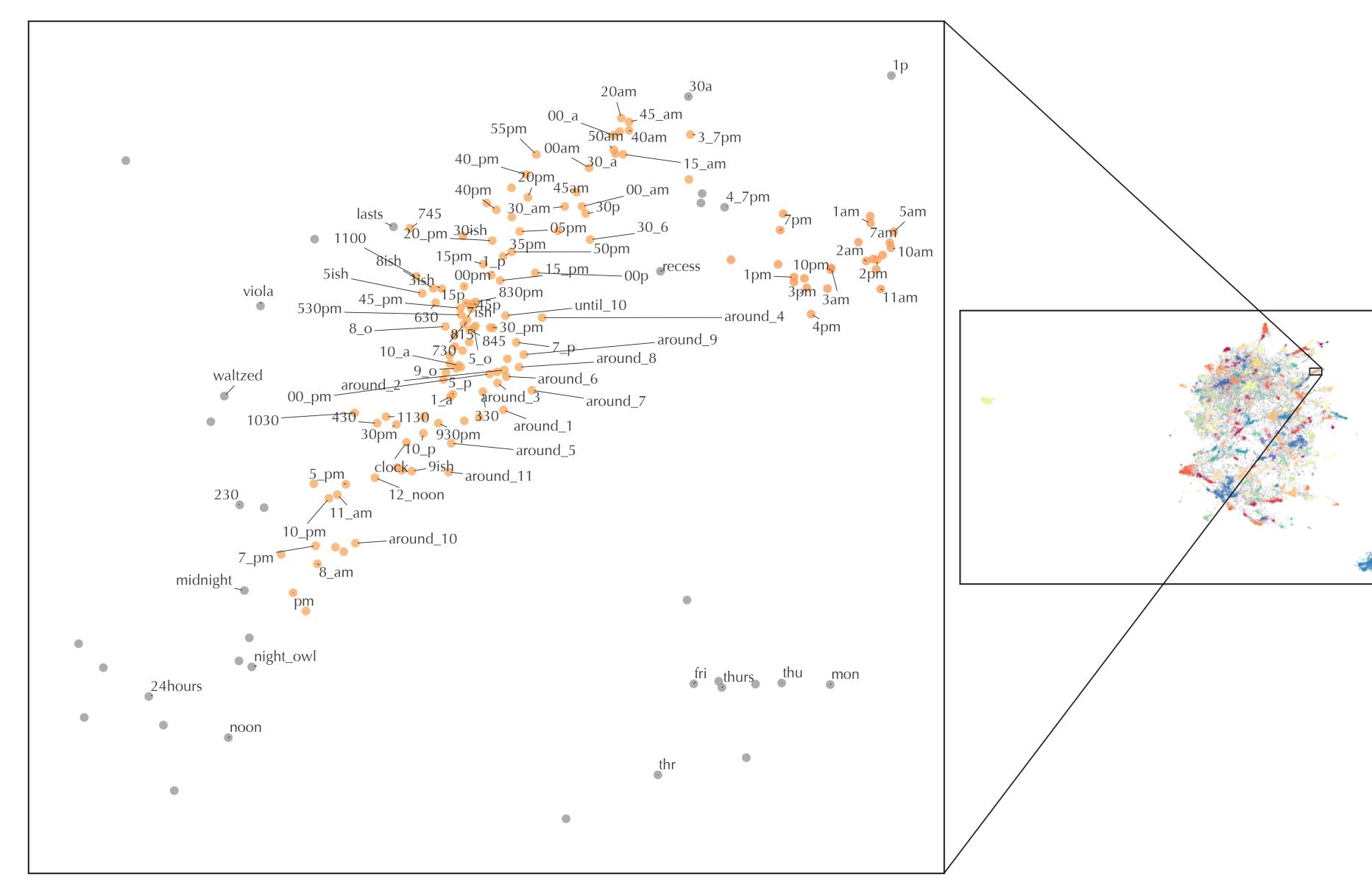




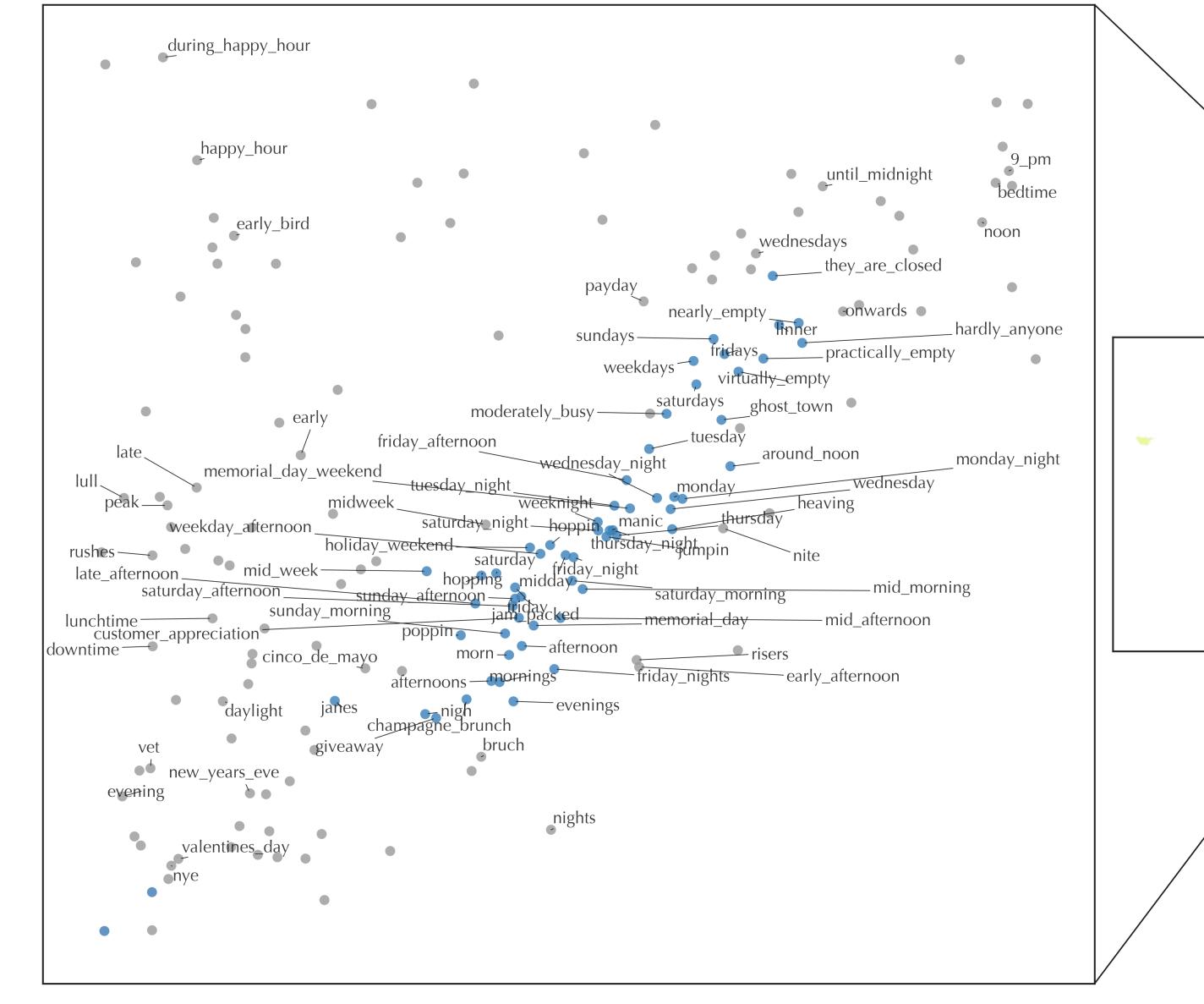


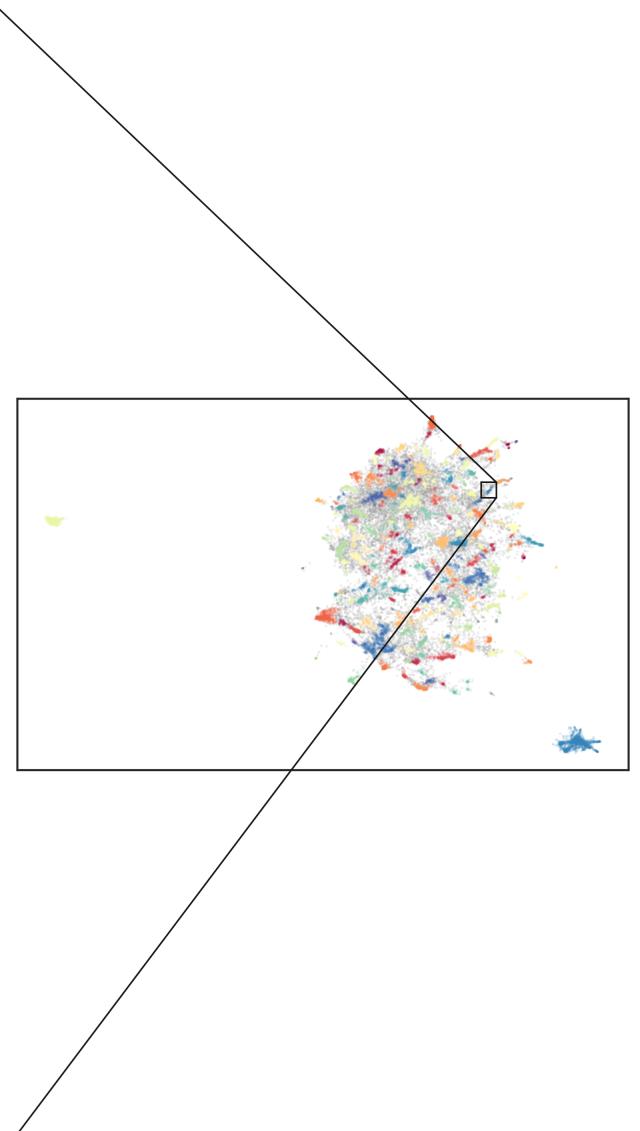




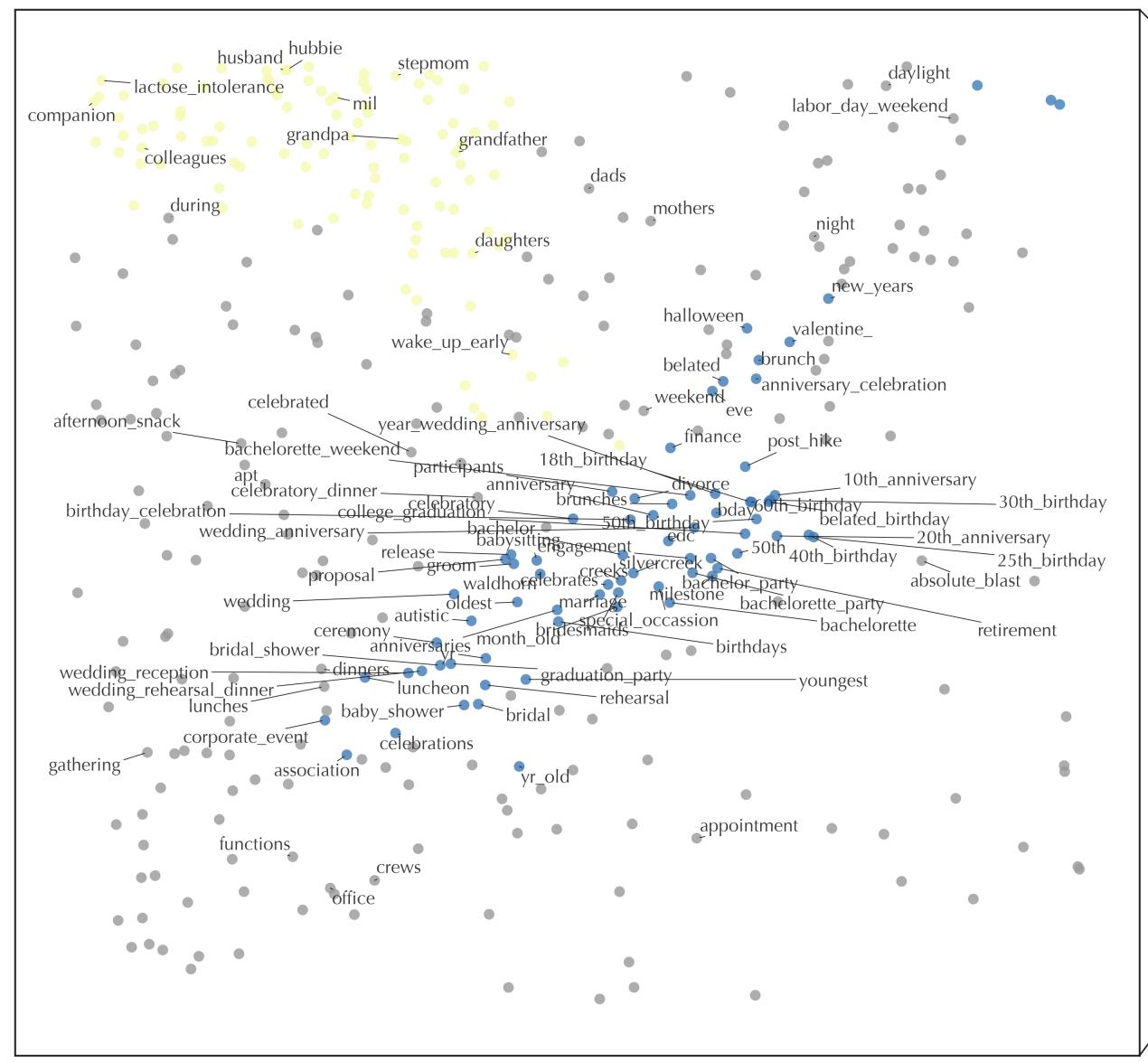


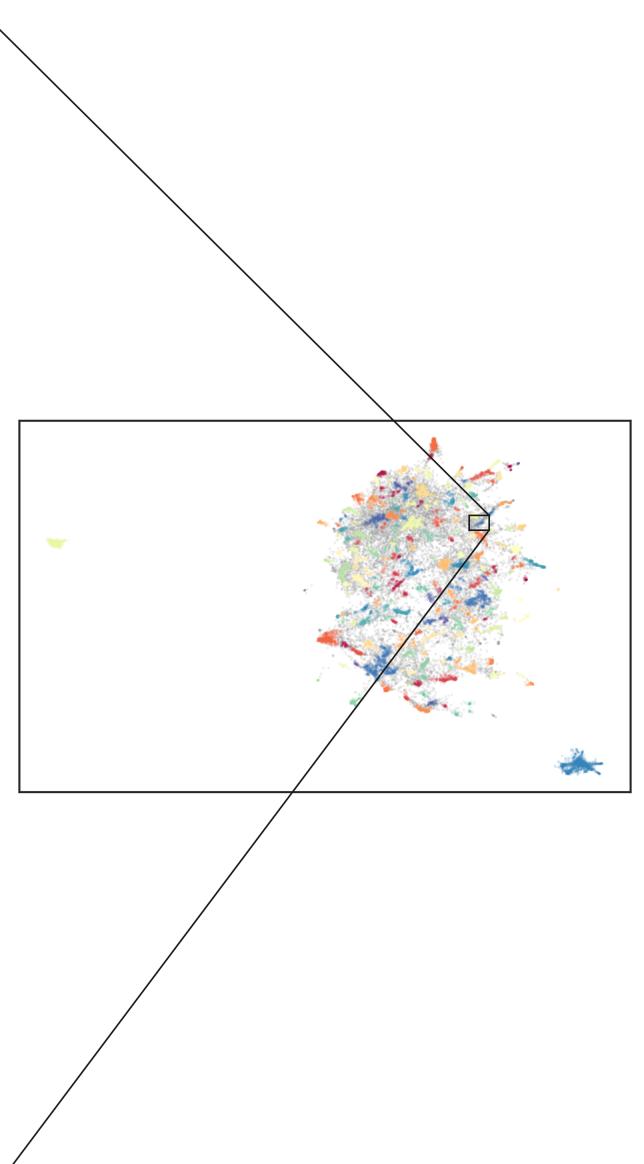


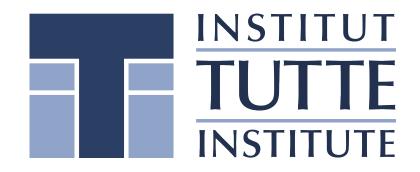


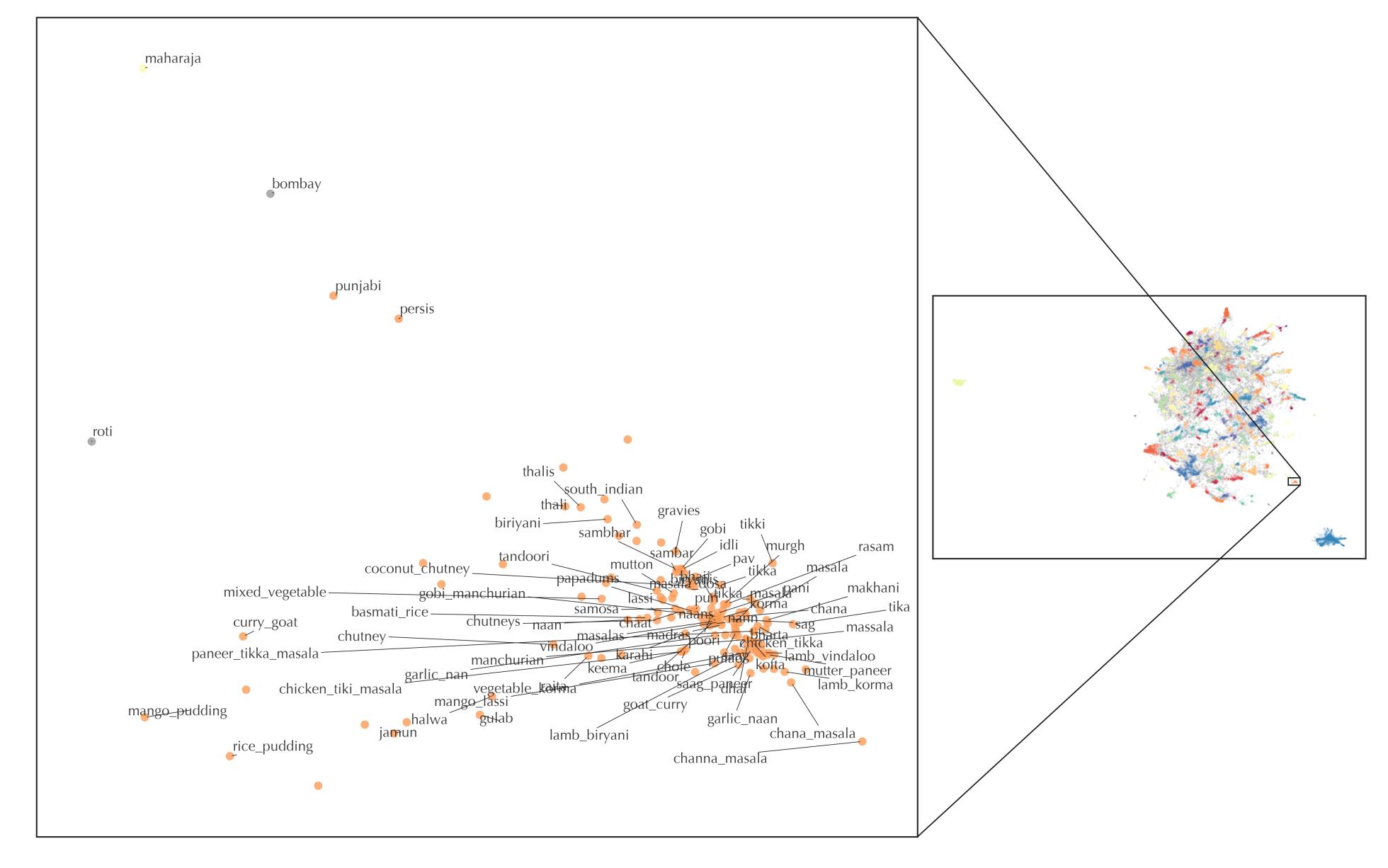




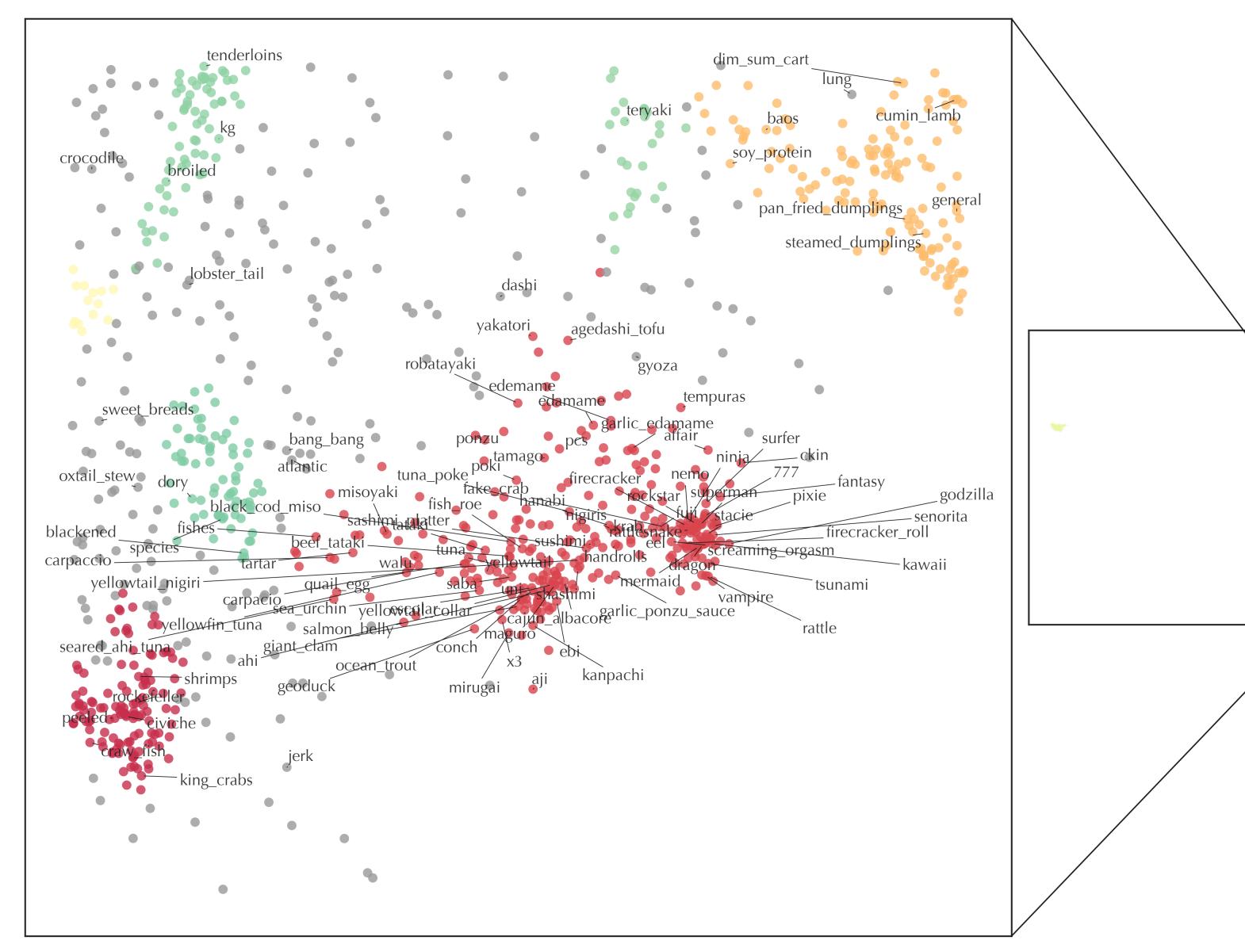




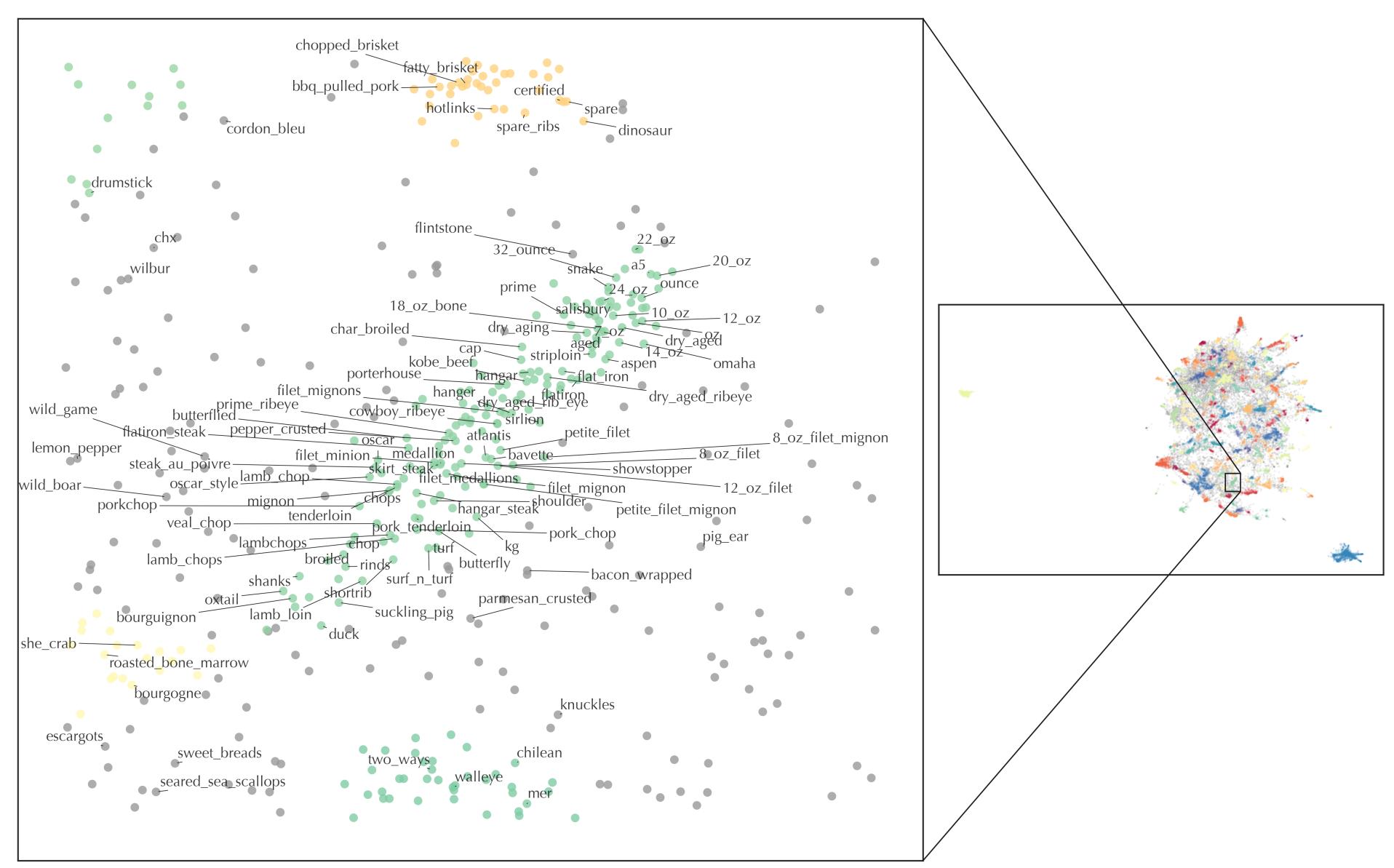




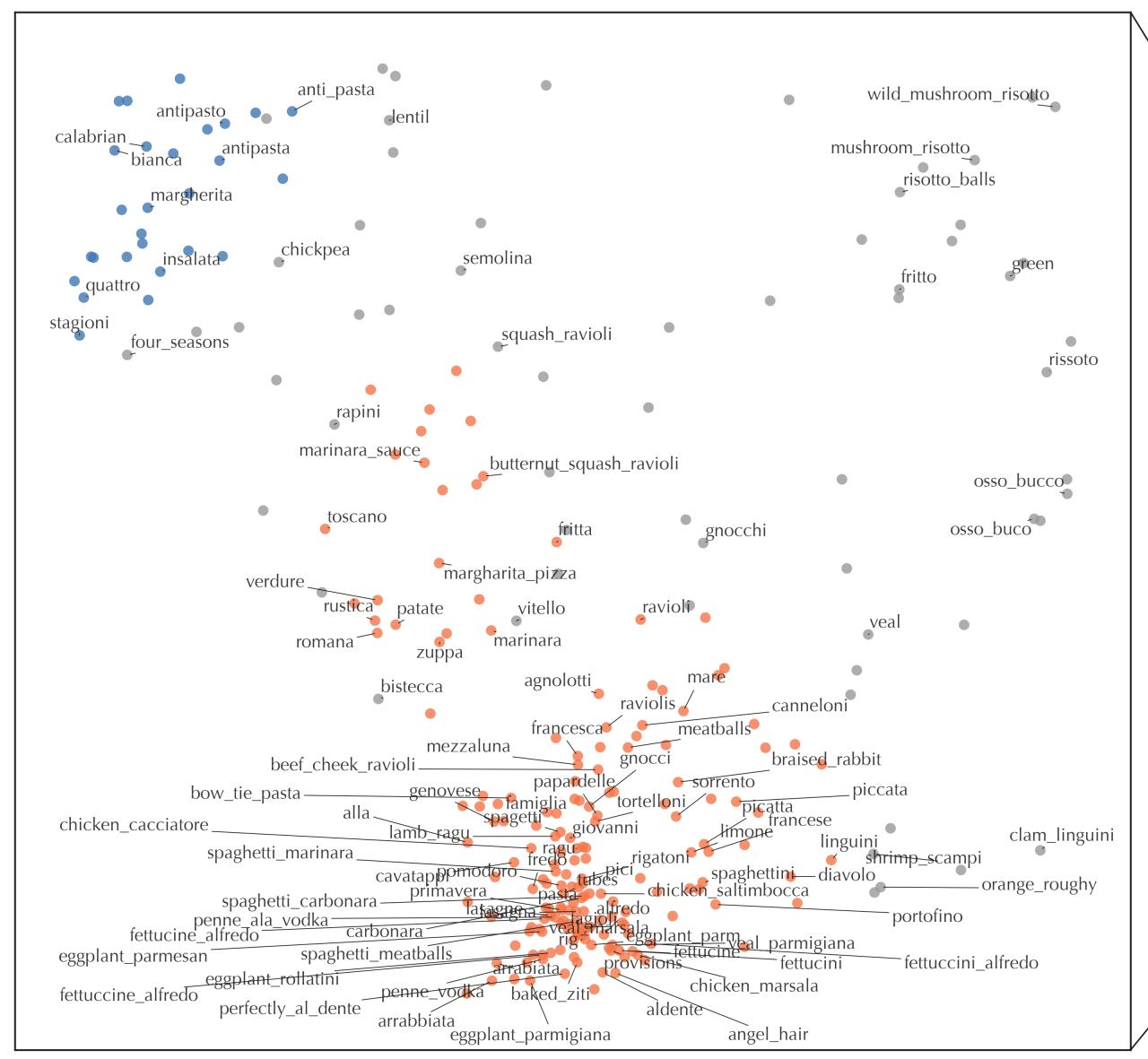


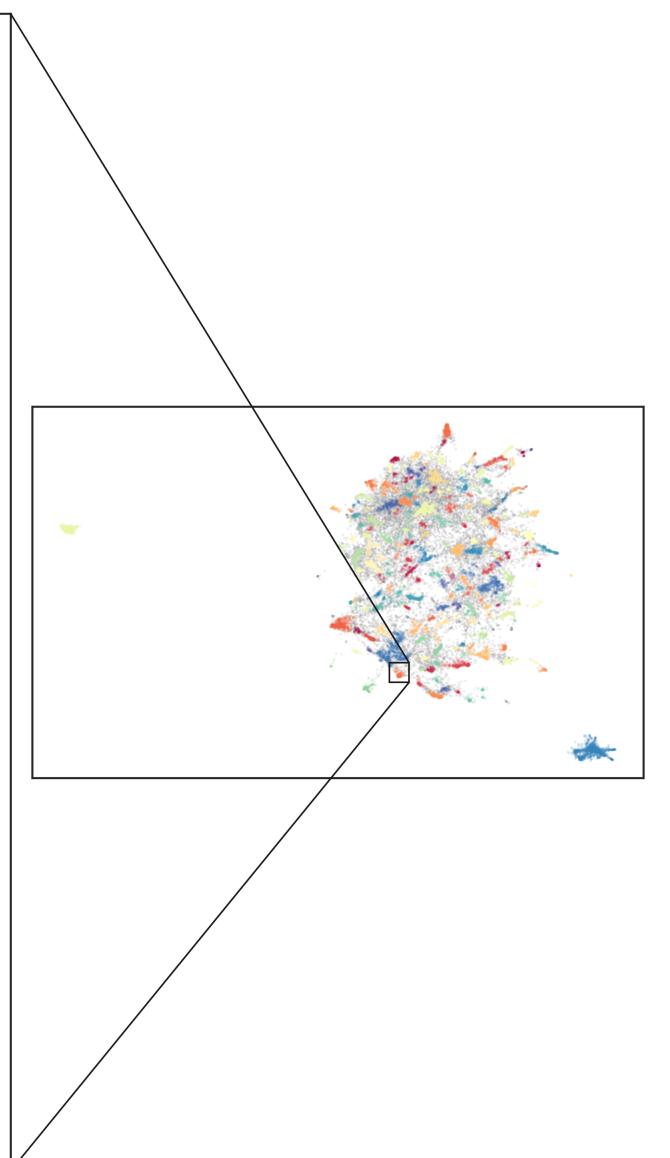




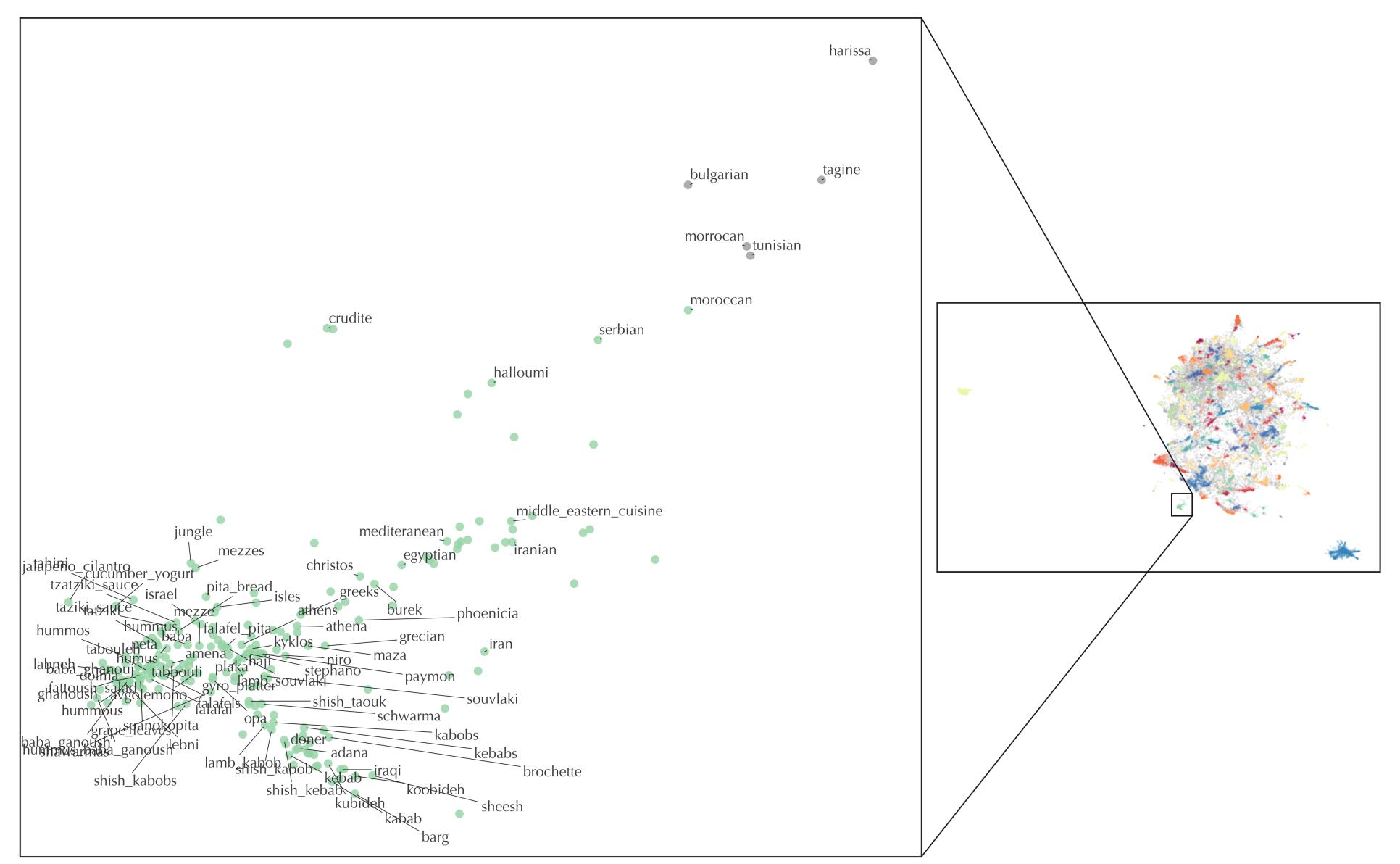




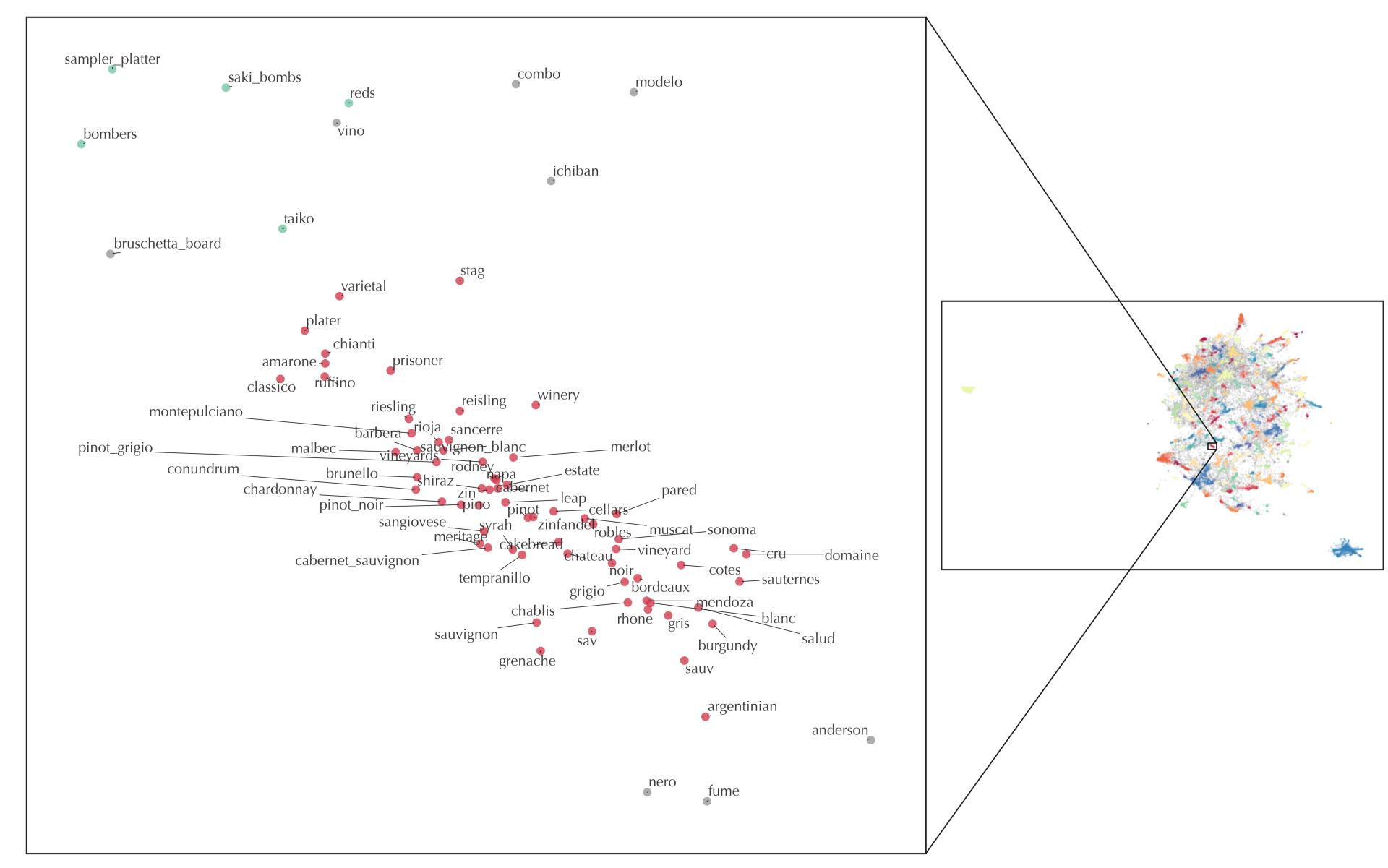




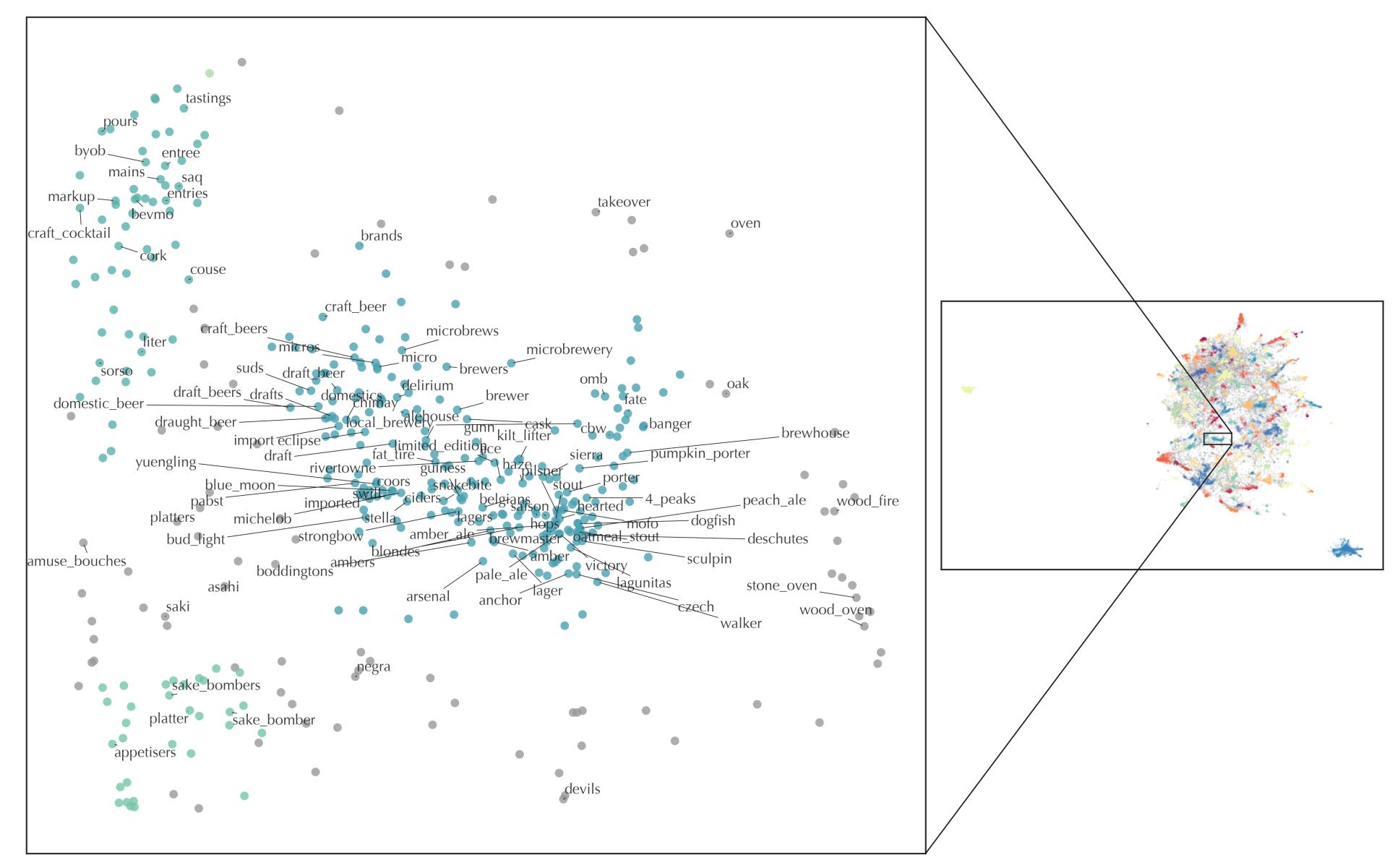














A lot of structure to explore!

What is the persistent homology of UMAP's simplicial sheaf representation?



Documents as Distributions



Represent a document using a "bag-of-words" model



Represent a document as the multinomial distribution of words occurring in the document



Information geometry tells us that multinomial distributions live on a manifold



$$d(p,q) = \arccos\left(\sum_{i} \sqrt{p_i} \sqrt{q_i}\right)$$



Some problems:

- Word order matters
- Words carry different amounts of information
- Language use is noisy
- Words are not independent / orthogonal



We can't fix word order issues this time (yet)



Information in words:

Approximate the information of a word in a given context and weight words accordingly



Language is noisy:

For larger documents we can hope to have the noise wash out in the background



This red rose smells sweet

That scarlet flower's fragrance is delightful

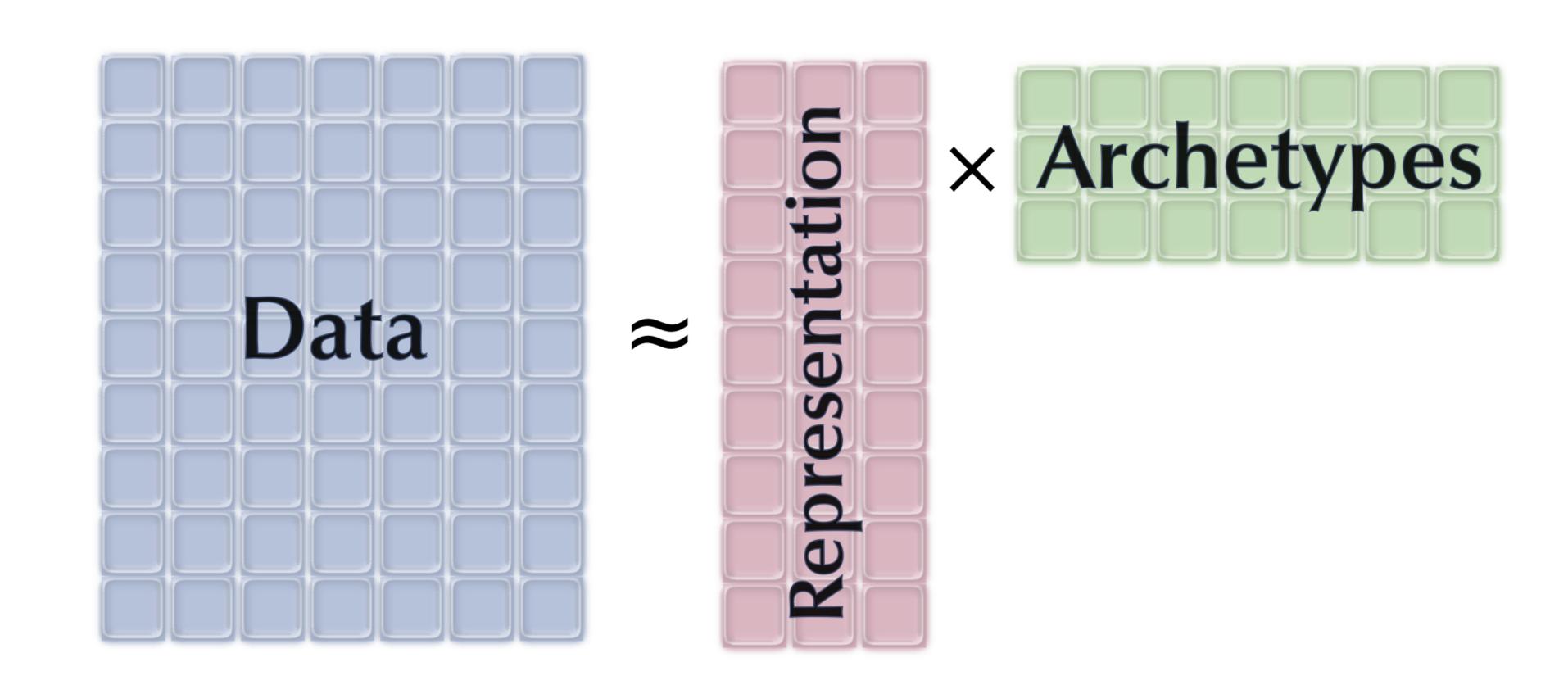


A document as a multinomial is averaging one-hot encodings of words



We can perform a change of basis to word vectors







$$D' = D \times W_{pLSA}$$



We are taking averages, so we should remove noise again ...



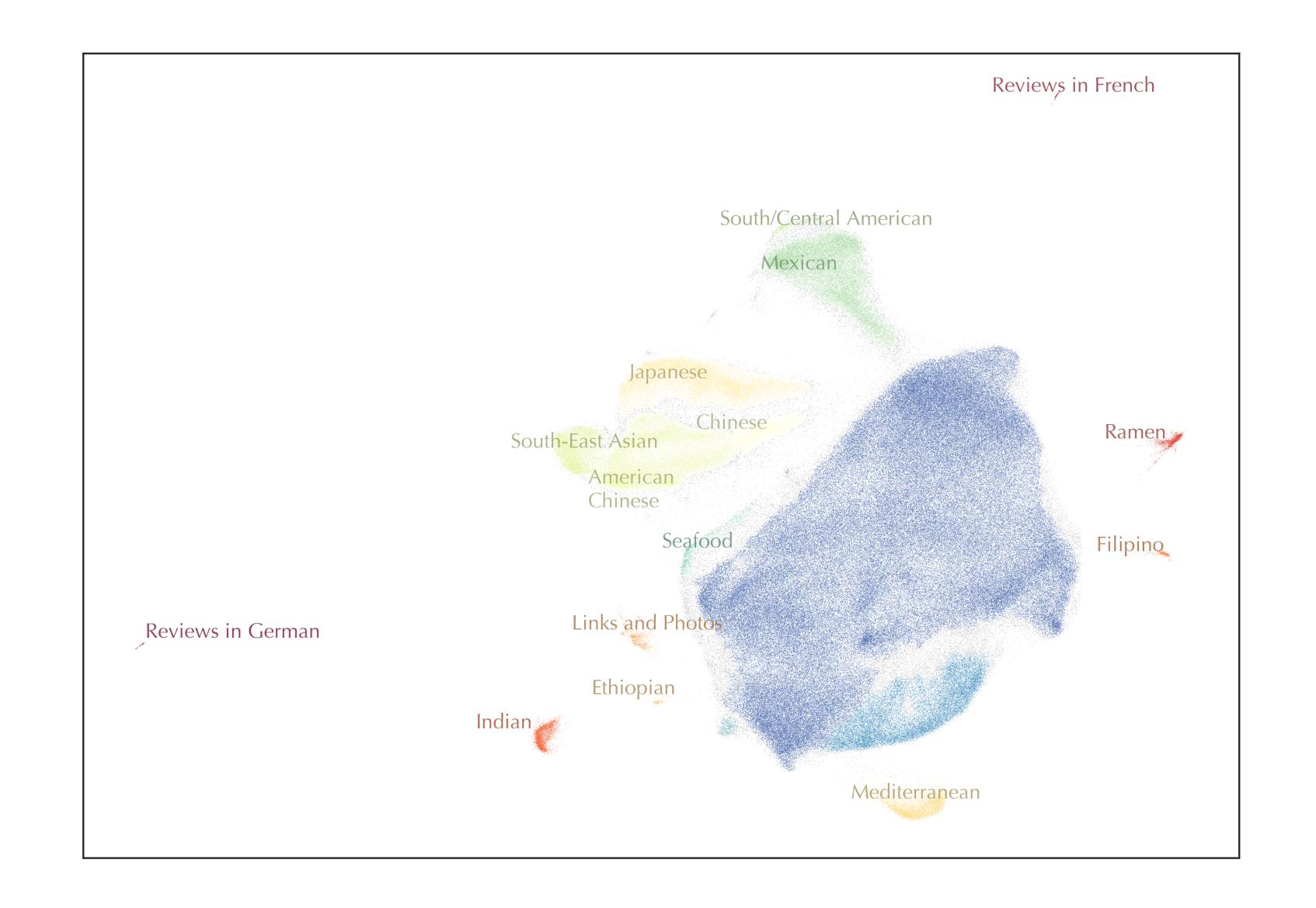
Embedding Documents





UMAP is a topology based dimension reduction algorithm







Since words and documents now live in the same combined space we can embed both together



This can provide powerful tools for topic modelling



Conclusions



We can use topological techniques to explore and represent language



Many of these ideas provide a mathematical basis for variations on existing techniques



This remains a rich field for further exploration