

AnchorViz

Facilitating Classifier Error Discovery through Interactive Semantic Data Exploration

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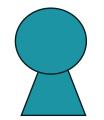
Jina Suh, Johan Verwey, Gonzalo Ramos, Steven Drucker, Patrice Simard (Microsoft Research)



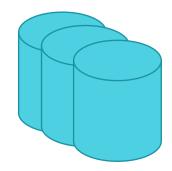


Interactive Machine Learning (iML)

Student-Teacher Metaphor for ML

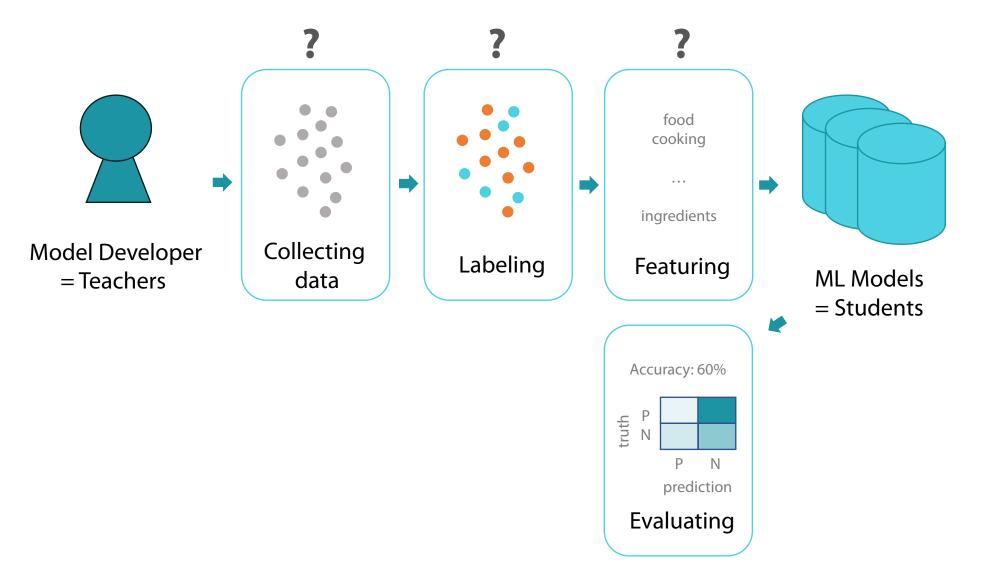


Model Developer = Teachers

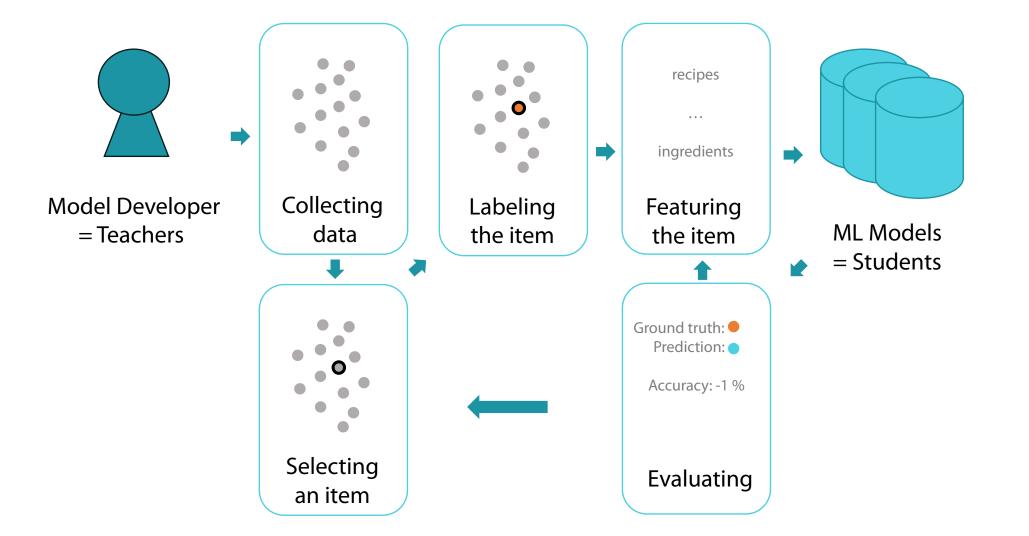


ML Models = Students

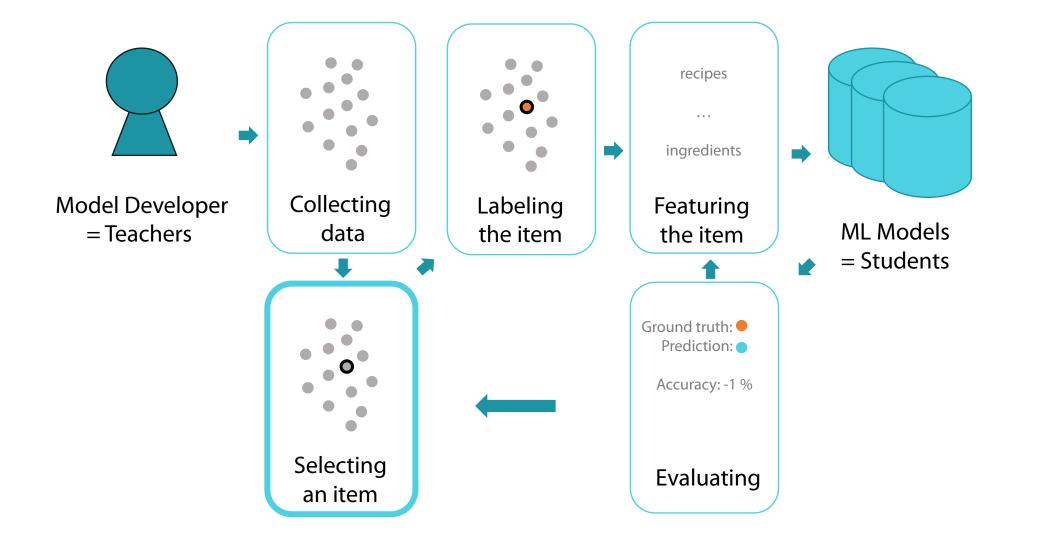
Traditional ML Paradigm



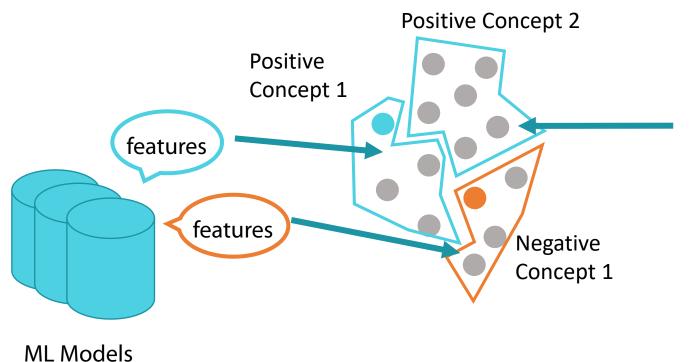
iML Paradigm



Key Challenge: Selecting Items for iML



Islands of concepts for the target class



= Students

Feature blindness error:

When there is no feature representing a concept

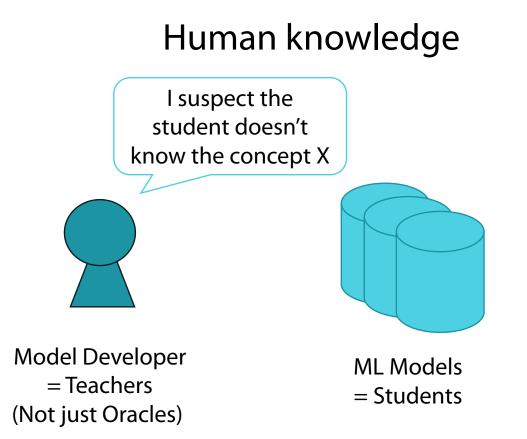
Two Typical Ways to Select Items

Active learning methods

- Uncertainty Sampling
- Uniform Sampling
- Stratified Sampling



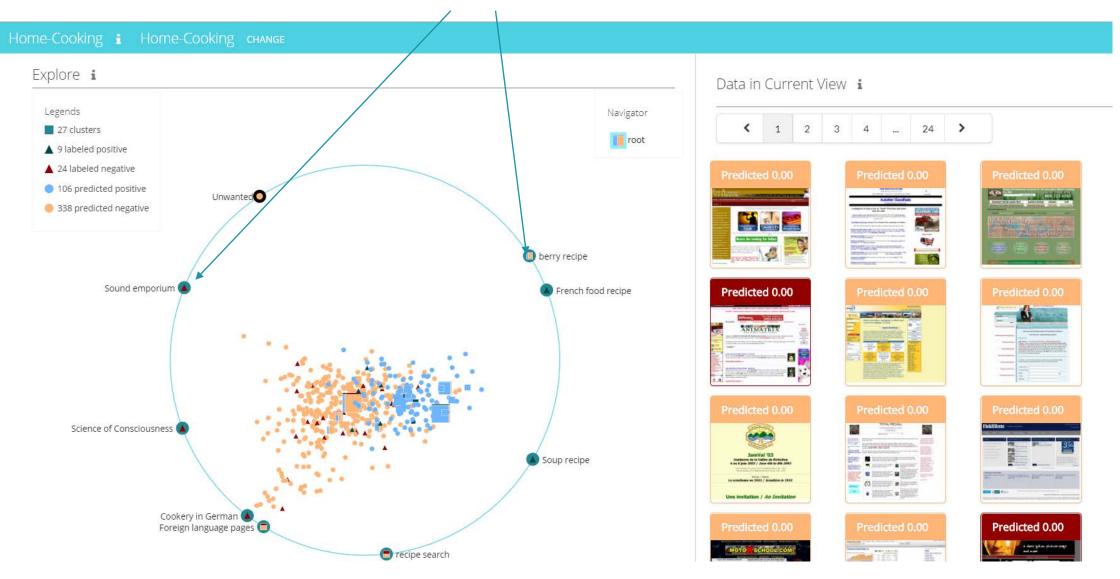
Random or ambiguous items



Can't enumerate all concepts!

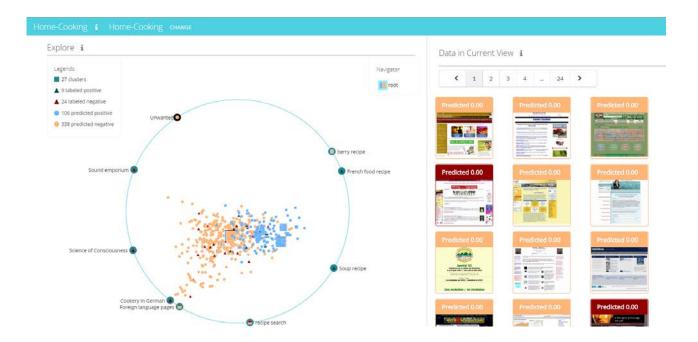
AnchorViz

anchor: the representation of a concept



Outlines

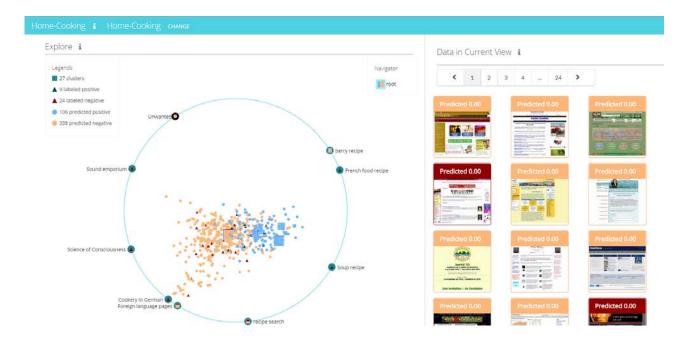
- 1. Motivating example & AnchorViz design
- 2. Evaluation & Analysis
- 3. Discussion & Future work



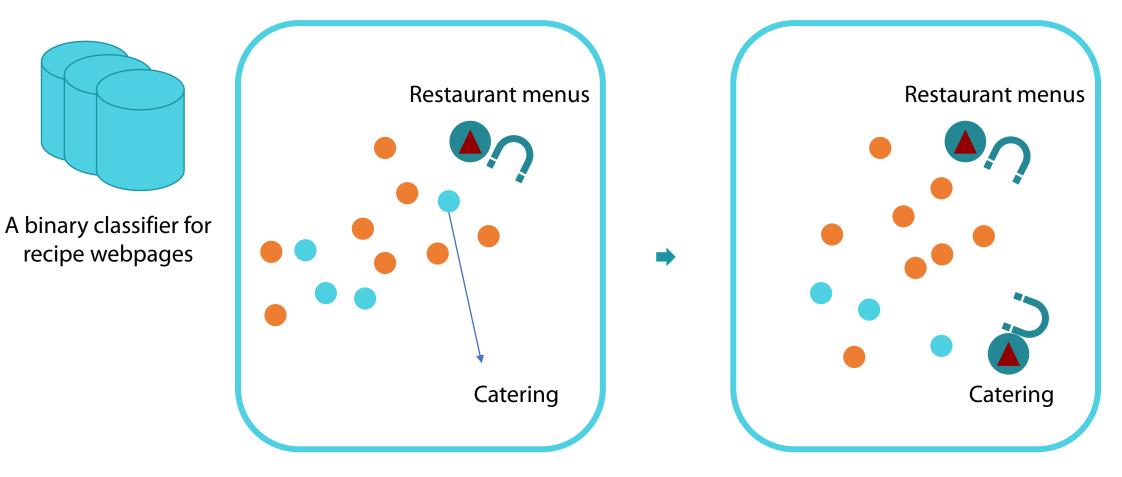
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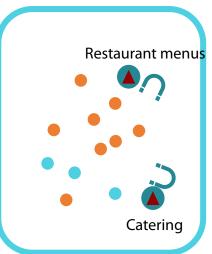


Motivating Example: anchor & nearby item discrepancy

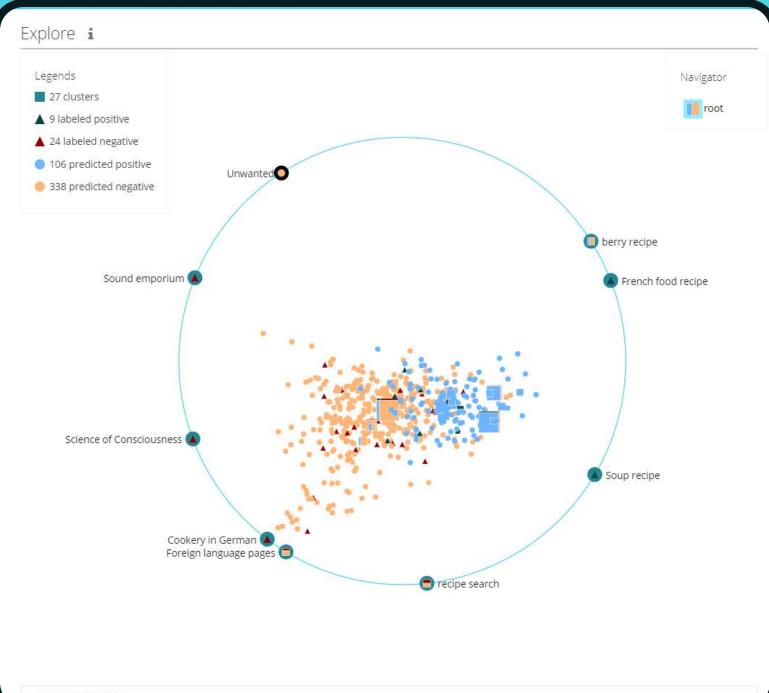


Design Objectives

- 1. Let users define concepts of the target class and unrelated classes
- 2. Spread the dataset based on concepts
- 3. Show how user-defined concepts impact the positions of items
- 4. Provide information about the model's current prediction along the user's labels
- 5. Optimize for efficient reviewing process



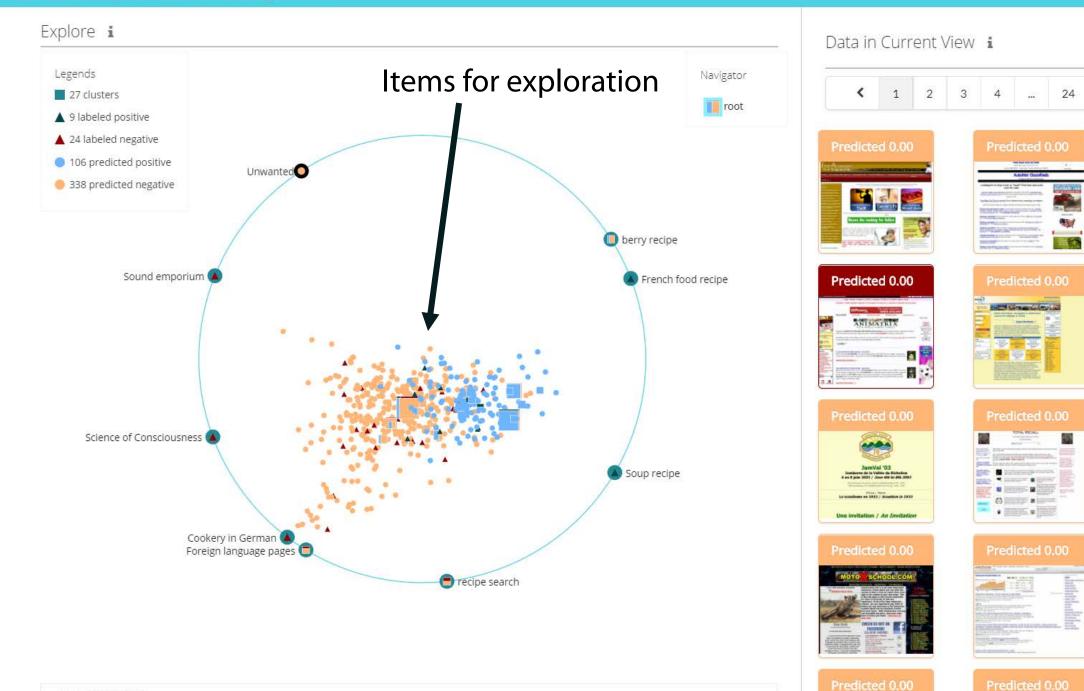
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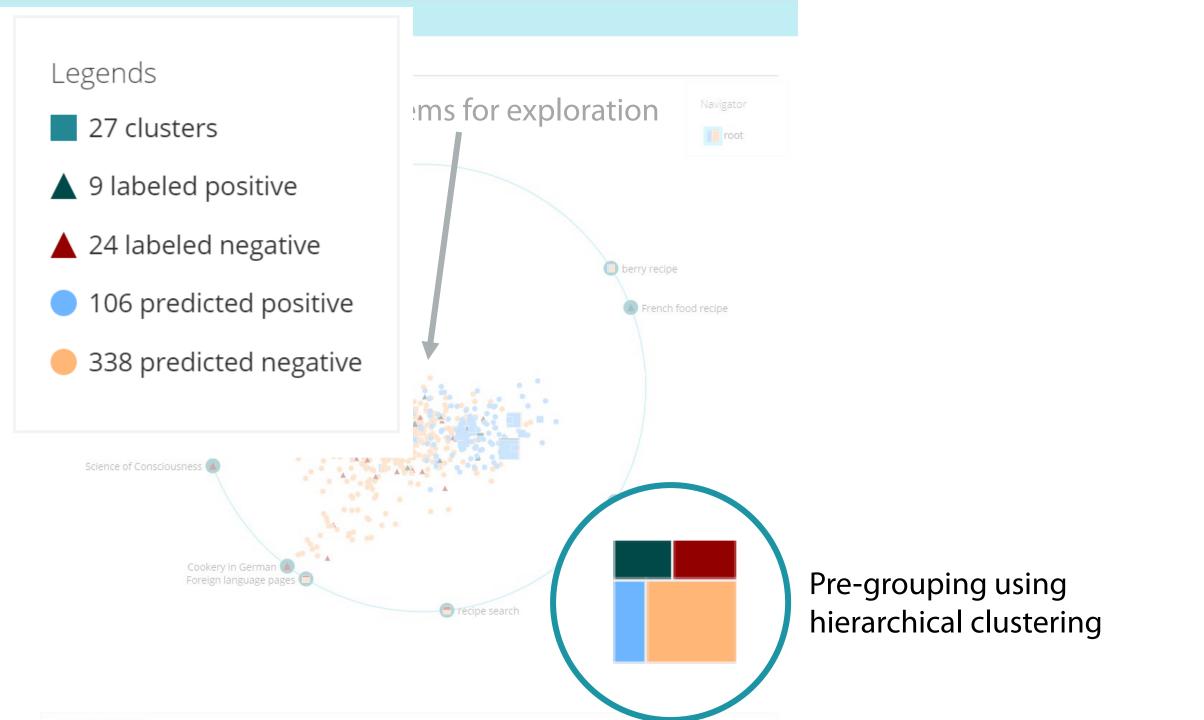
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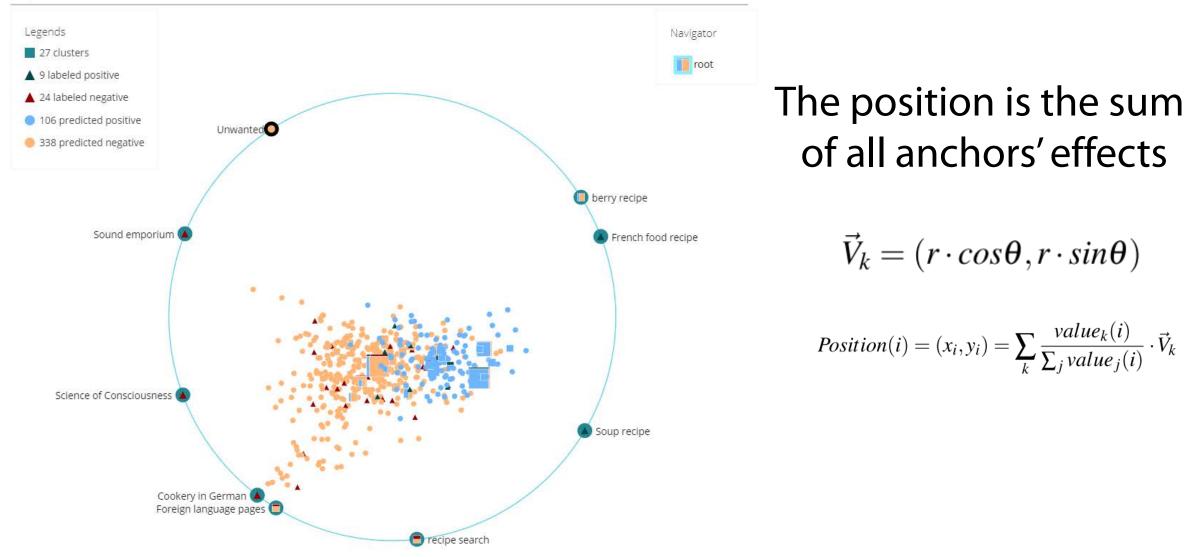
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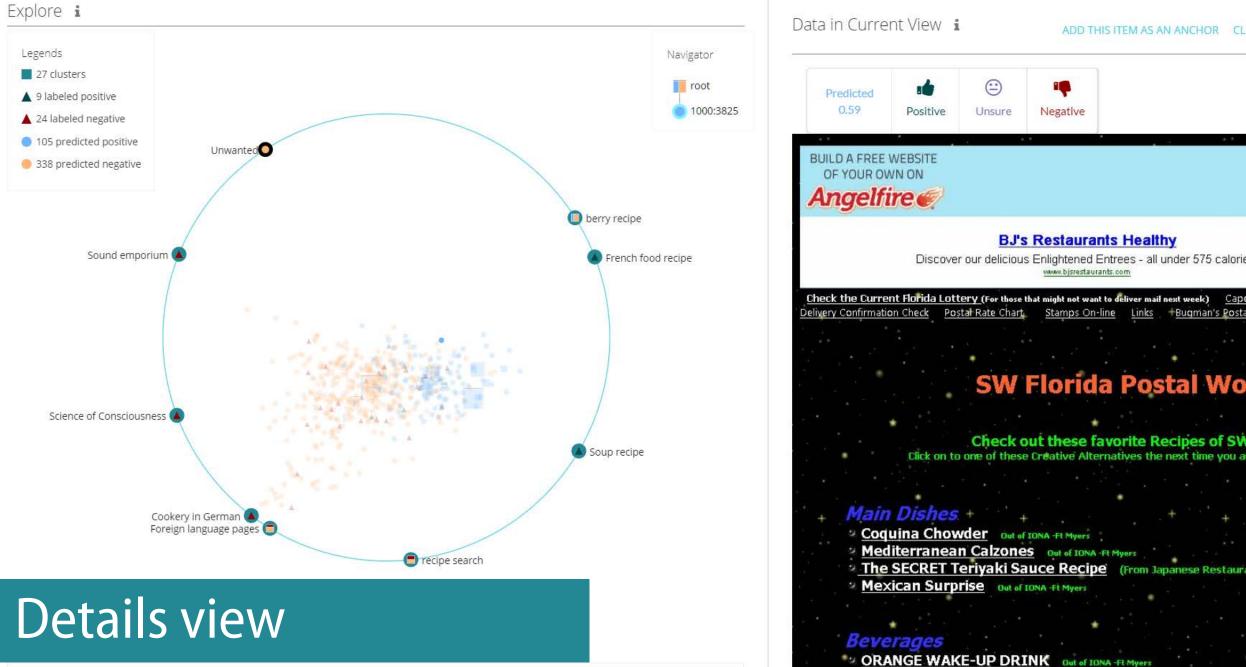
Home-Cooking 🖬 Home-Cooking Change

Explore i



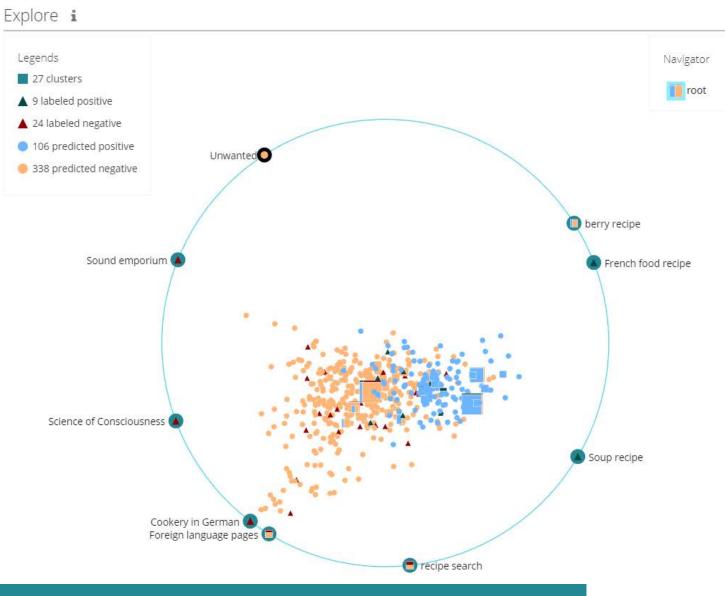
Visualization Layout

1 Sector Annaly Contraction Accession Statements



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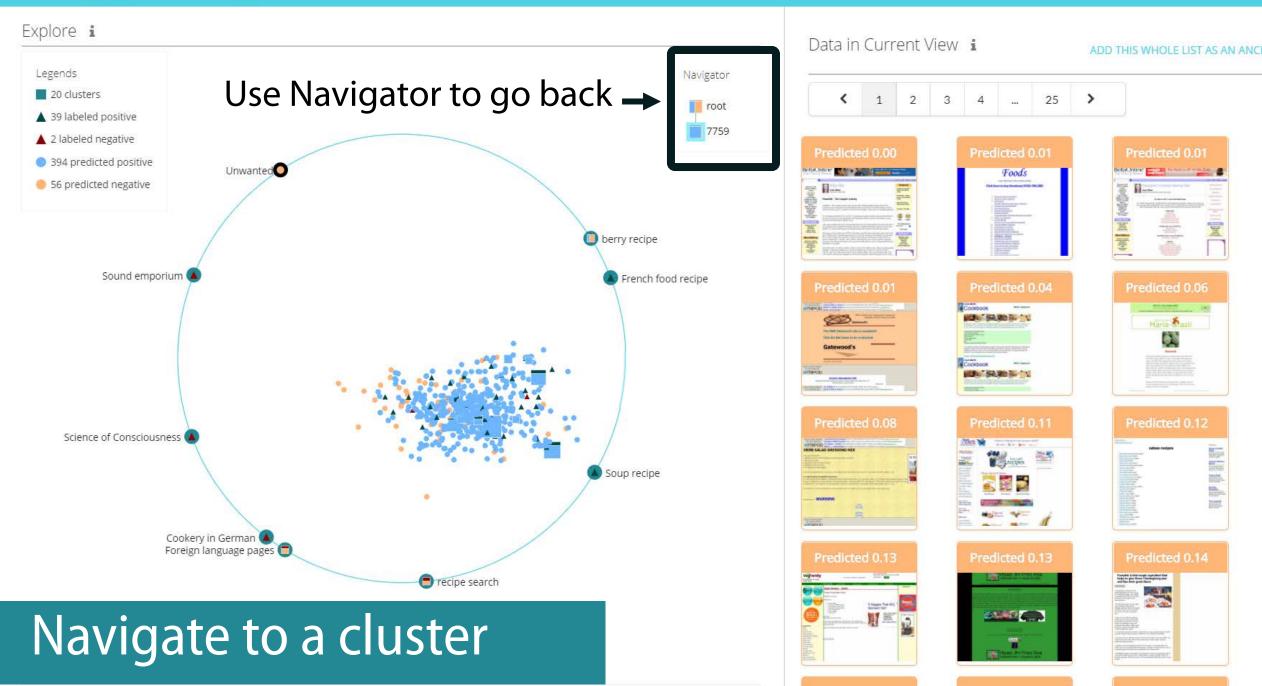
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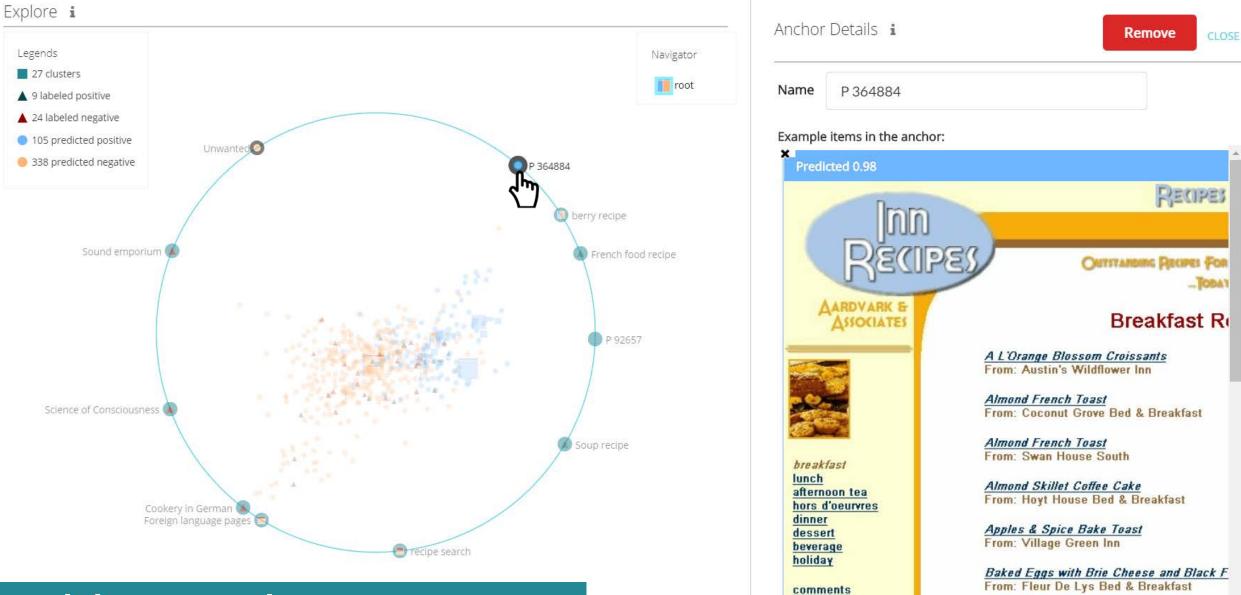


Navigate to a cluster

Data in Current View 🔒

<	1	2	3	4		24	>	
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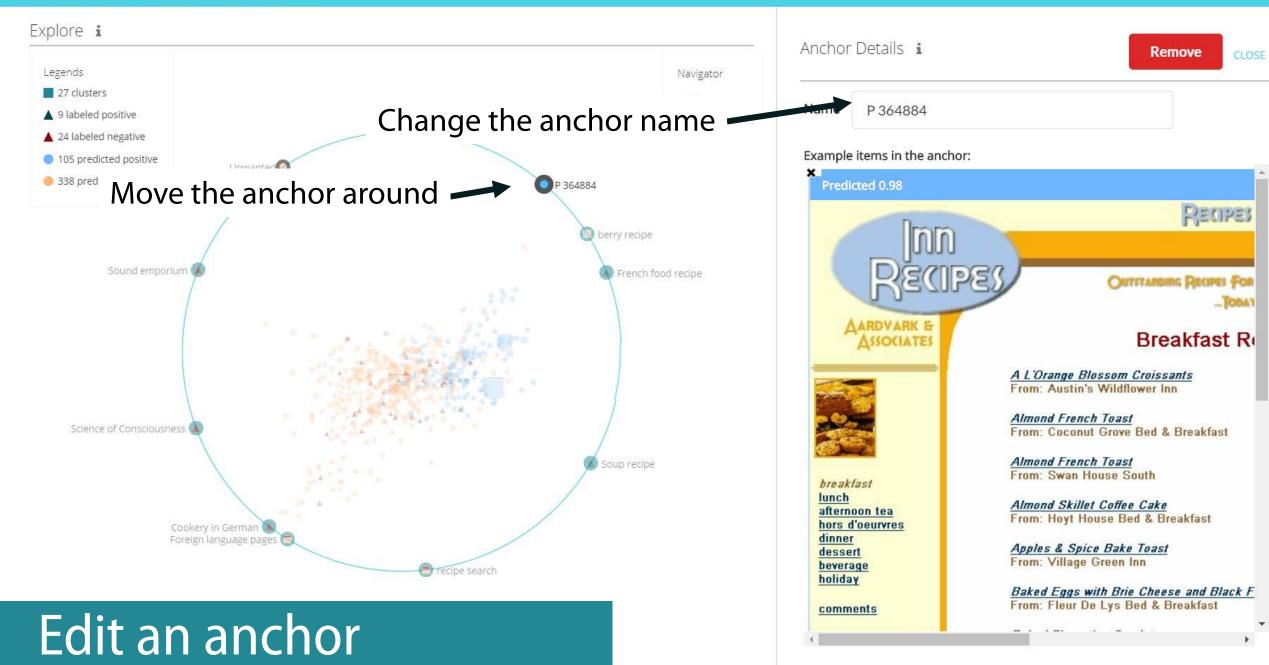




Add an anchor

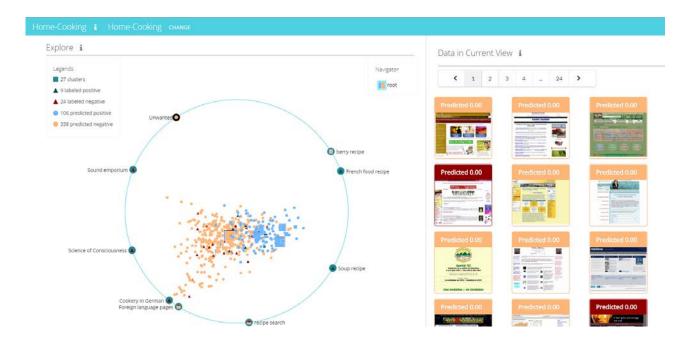
Anenor Repository

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Outlines

- 1. Motivating example & AnchorViz design
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Instructions

- (1) find items where the classifier is making a mistake
- (2) find a set of items that are diverse from each other
- (3) try to understand the dataset and classifier performance in the process



Build the model using an iML tool

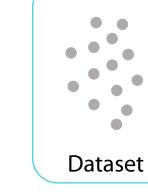


16 Participants 7 Females & 9 Males Distributing nearly equally across 4 levels of ML experiences

A binary classifier for cooking webpages (12 existing features)

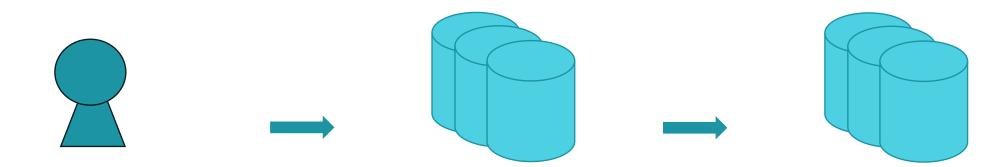
training set but ~75% on the test set

100% accuracy on the



4,000 webpages 400 labeled (50% positive)

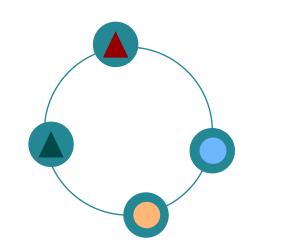
Study Procedure



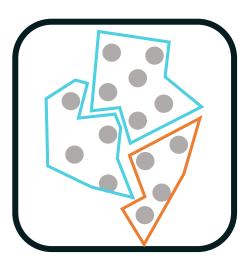
Introduction Videoconferencing Starting audio & screen recording ML concepts review **Training session** A binary classifier for travel webpages to introduce AnchorViz and practice

Real study session A binary classifier for cooking webpages

Analysis





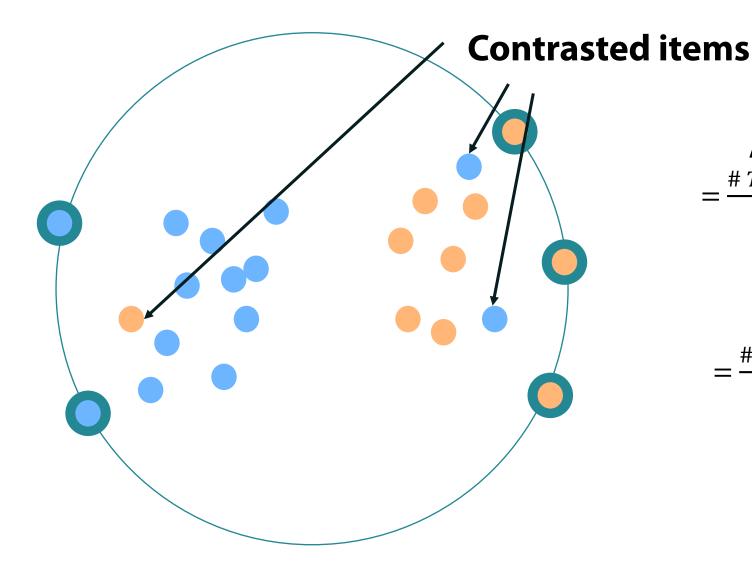






Qualitative coding on user behaviors

Anchor Effectiveness



Anchor Error Precision (AEP)

True Errors in the Contrasted Items

Items in the Contrasted Items

Anchor Error Recall (AER) # True Errors in the Contrasted Items

True Errors

Anchor Effectiveness Results

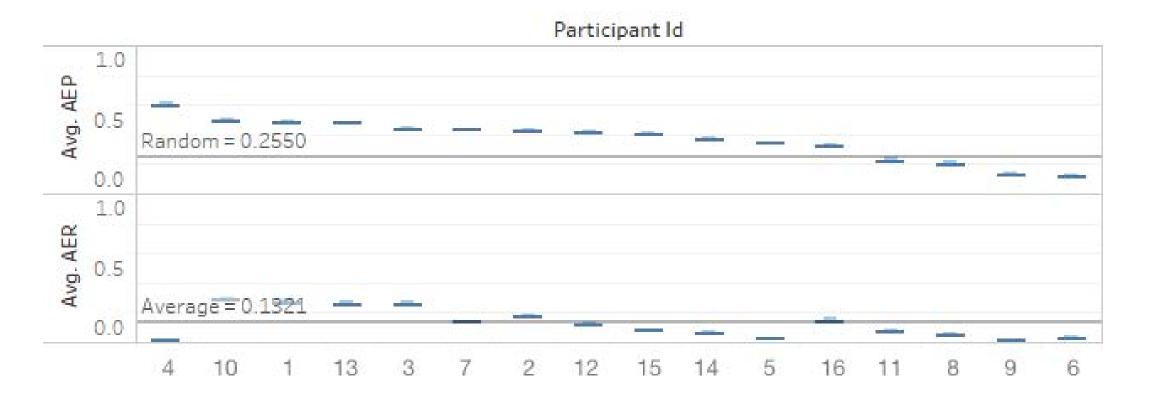
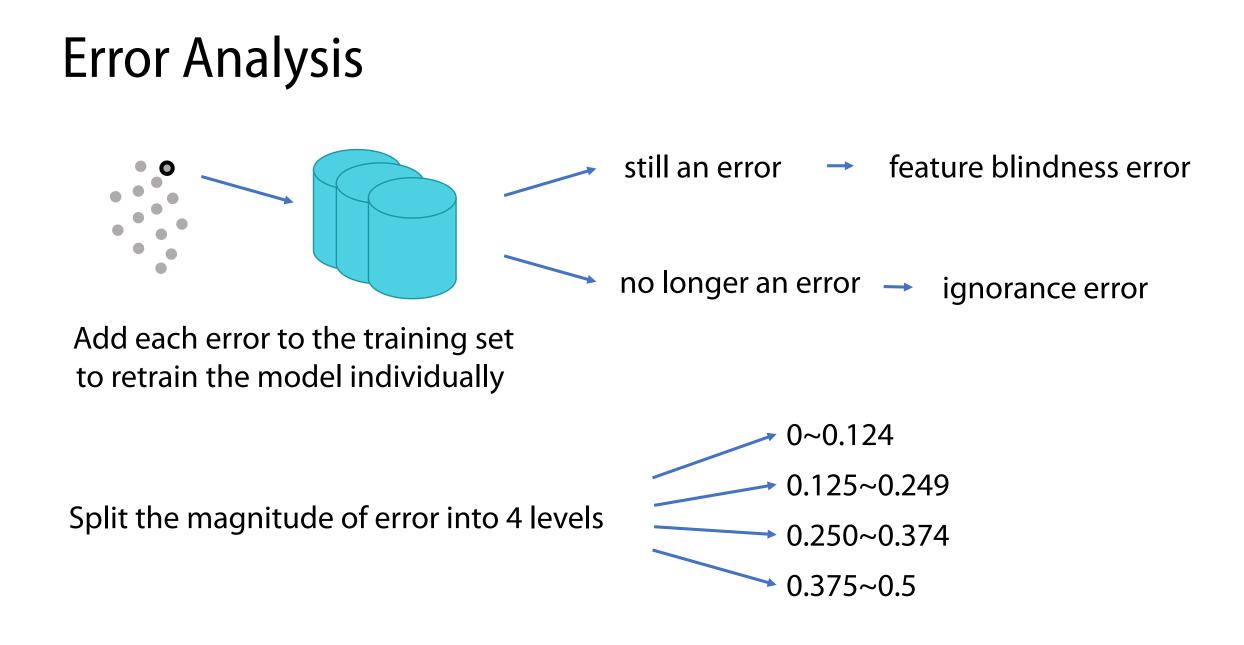


Figure 5. Average anchor effectiveness metrics (AEP/AER) for each participant. Most participants have better AEP than random.



Error Analysis Results

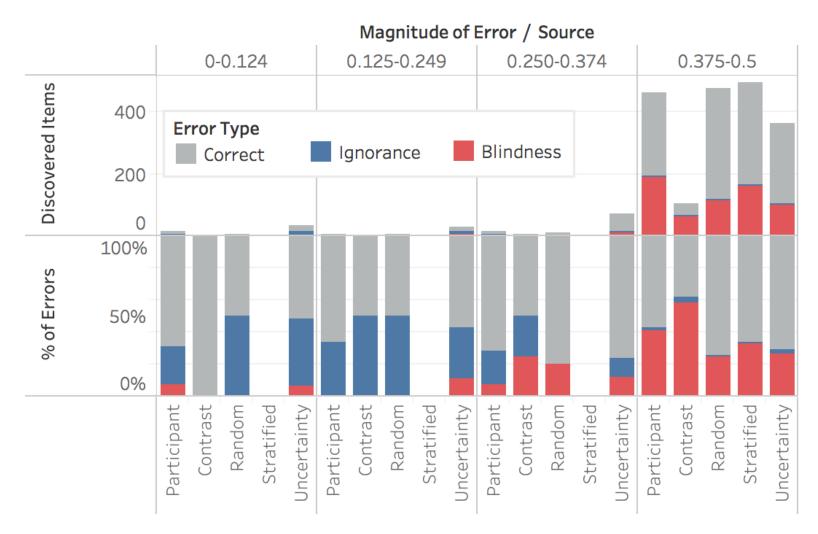
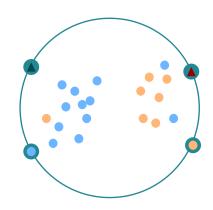


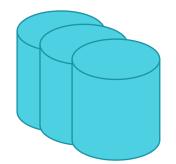
Figure 3. Distribution of discovered items (top) and the magnitude of errors (bottom) across participants and algorithmic samplers. The items are skewed towards high magnitude errors, but items discovered by participants and contrasted items higher chance of getting high magnitude items than algorithmic samplers.

Qualitative Coding on User Behaviors

Reasons for creating anchors



Exploration strategies



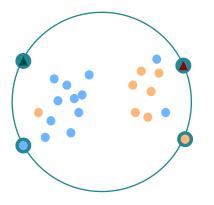
Insights about the classifier

Qualitative Coding Results

Reasons for creating anchors

- Define positive & negative concepts (recipe vs. science)
- Capture potential issues (foreign language pages)
- Validate hypotheses (suspecting and verifying the issues of index pages)

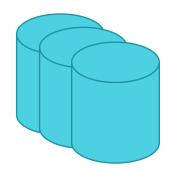
Qualitative Coding Results



Exploration strategies

- Leverage visual encoding (discrepancy in colors and positions)
- Play with the placement of anchors (changing topology to find patterns)
- Refine anchors to make them more effective (change an "cucumber" anchor to "vegetables")

Qualitative Coding Results

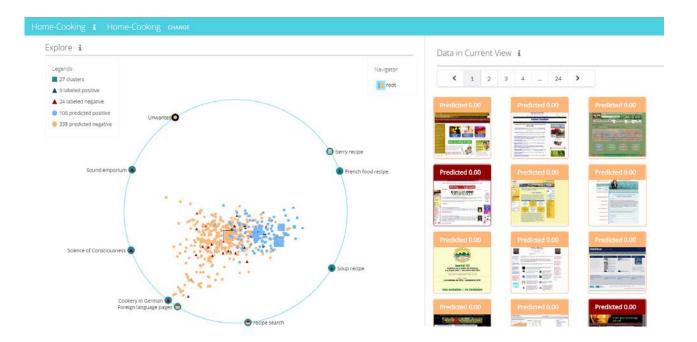


- Recall problem, not precision problem
- Clusters of errors
- Multiple meanings of a word can be a problem

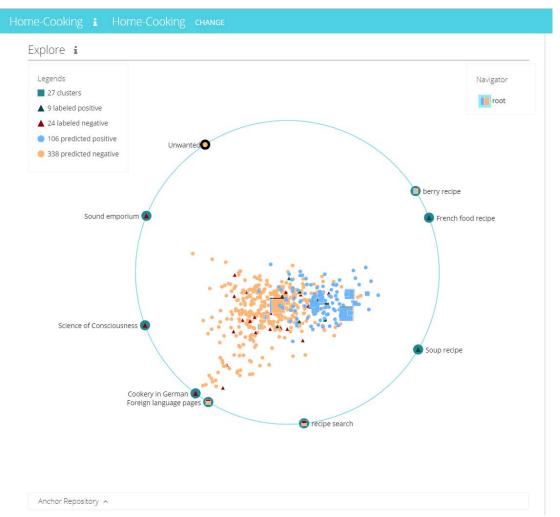
Insights about the classifier

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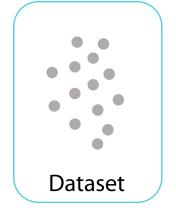
AnchorViz is not merely an error discovery tool



- Concept externalization & storage
- Concept evolution
 - Similar to structured labeling (Kulesza et al. 2014)

Featuring

Future Work



More ways to define & manipulate anchors

Integration with a complete iML loop

More evaluations with various setups

Acknowledgements





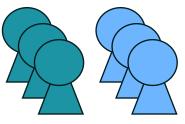


Microsoft[®] Research









And you!



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