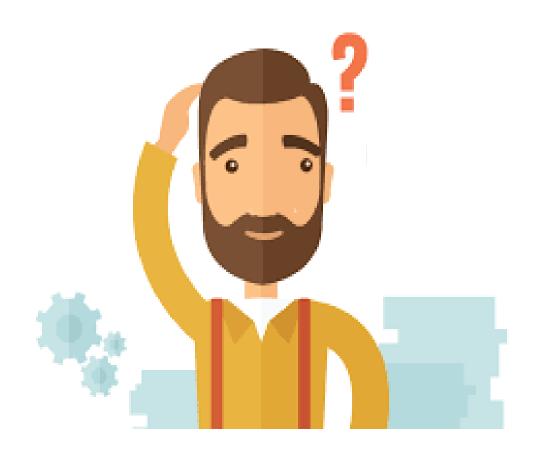


Identifying Miscategorized Products

M. Hidayath Ansari hidayata@ Applied Scientist II | Browse

Subhadeep Chakraborty sdeep@ Sr SDE | Browse









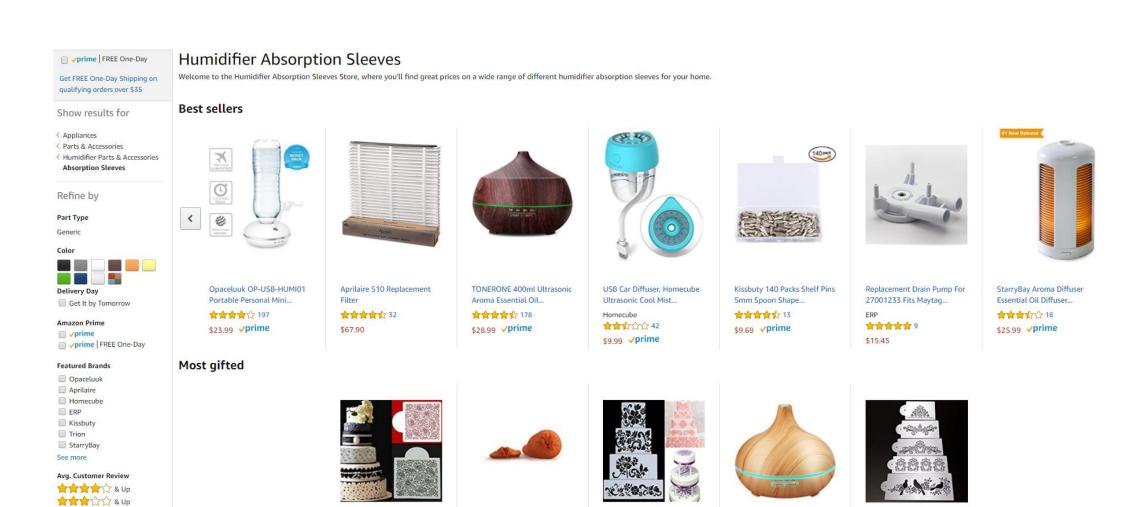




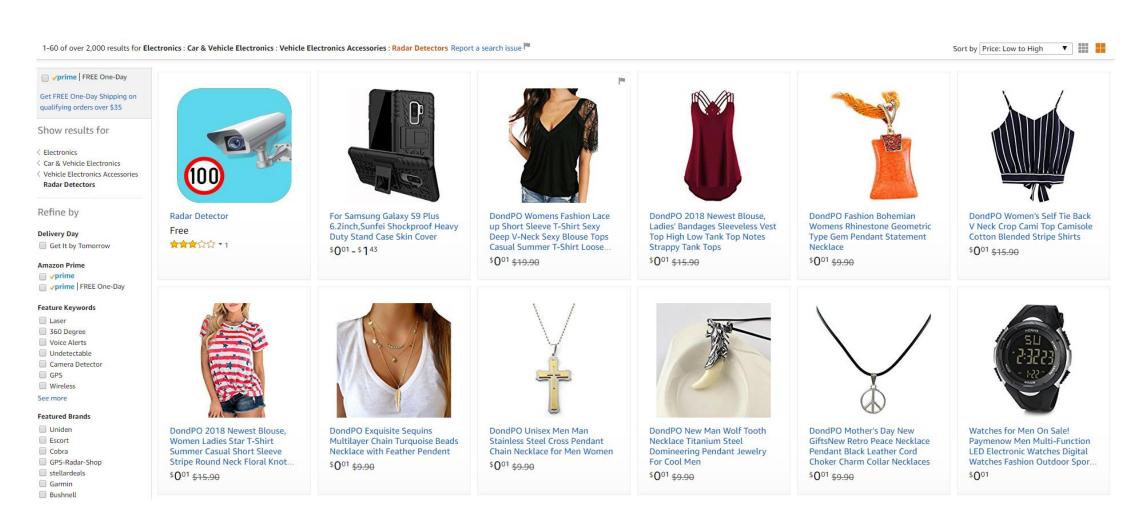




Customer Experience – Category Landing Page



Customer Experience: Sort by Price



"Radar Detectors"

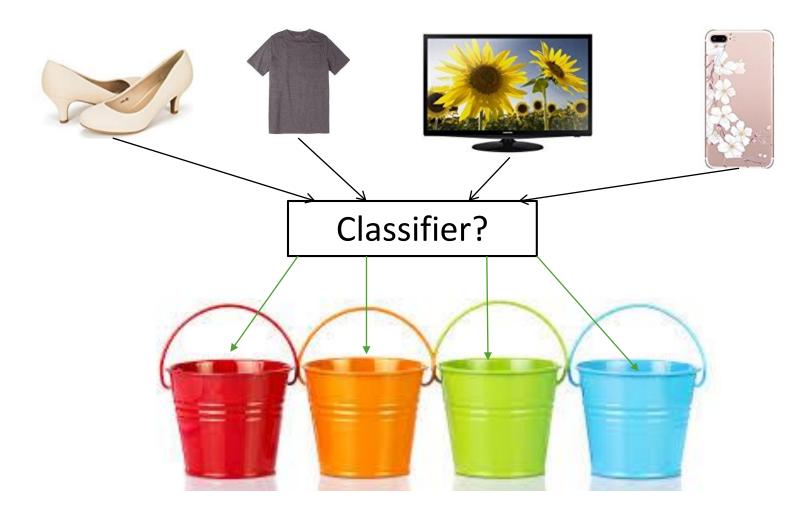
Product Discovery

Amazon Browse: Enabling customers' discovery experience



Without having to weed through irrelevant products

Finding Product Miscategorizations



BUT ...

"Which of these doesn't belong?"

"Which of these doesn't belong?"









































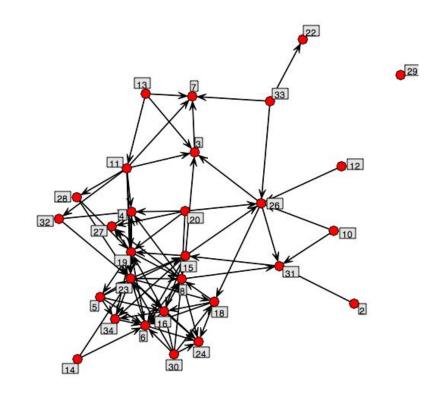


Unsupervised/Lightly supervised methods

PageRank

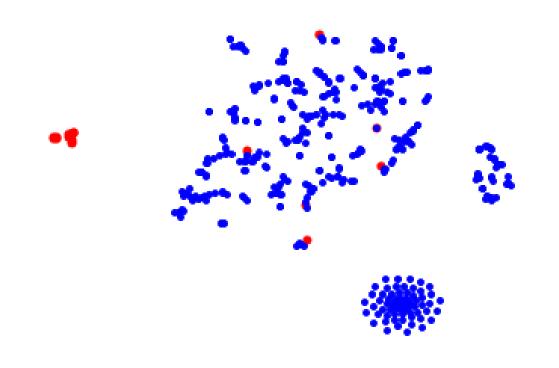
 Finds the most "central"/"important nodes

Conversely ...

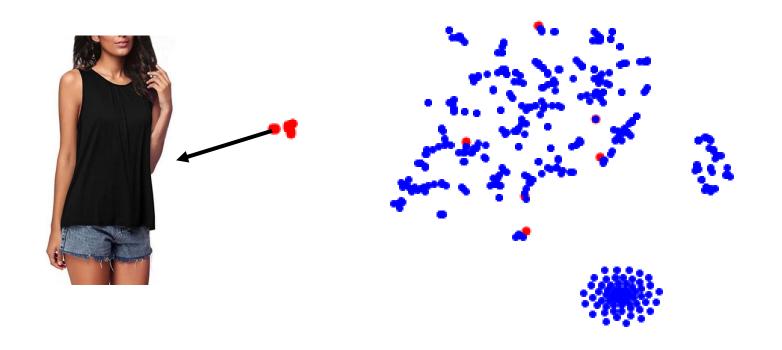




Dishwashers (dimensionally reduced)

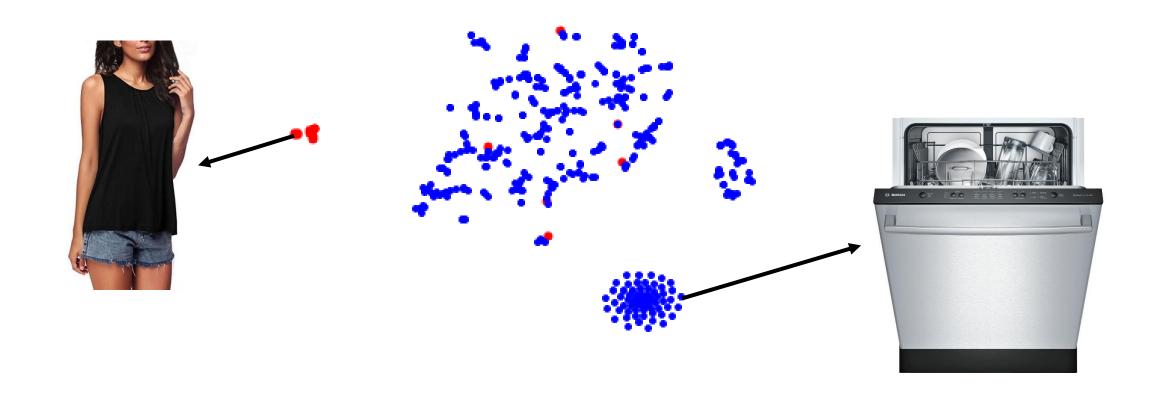


Dishwashers (dimensionally reduced)

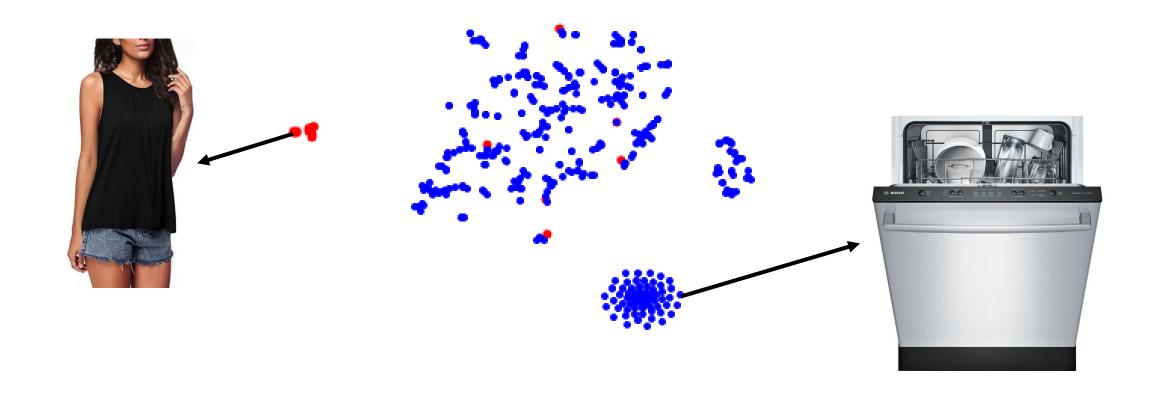


Dishwashers (dimensionally reduced)

amazon confidential 18



Dishwashers (dimensionally reduced)



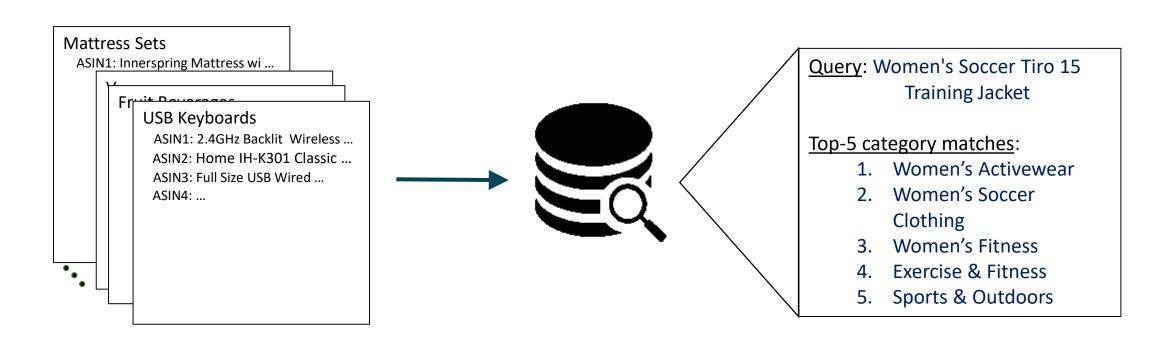
Still, Outrank has a high false positive rate

Dishwashers (dimensionally reduced)

Moonesinghe, H. D. K., and Pang-Ning Tan. "Outrank: a graph-based outlier detection framework using random walk." *International Journal on Artificial Intelligence Tools* 17.01 (2008): 19-36.

Miscategorizations – Using Search

Idea: Use a search index for product categories to find miscategorized products

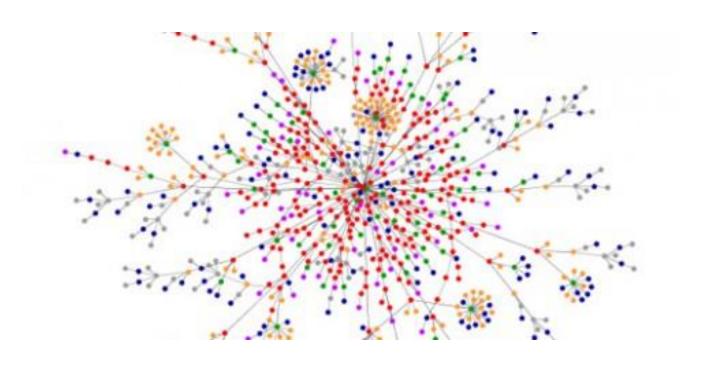


Singular outliers – Results & Impact

Marketplace	Assignments removed	Audited miscat Precision	Irrelevancy Reduction	OPS impact through irrelevancy reduction
US	2,319,105	91.8%	24% → 9%	\$15,925,972
UK	2,168,596	97.2%	20% → 14%	\$4,886,300

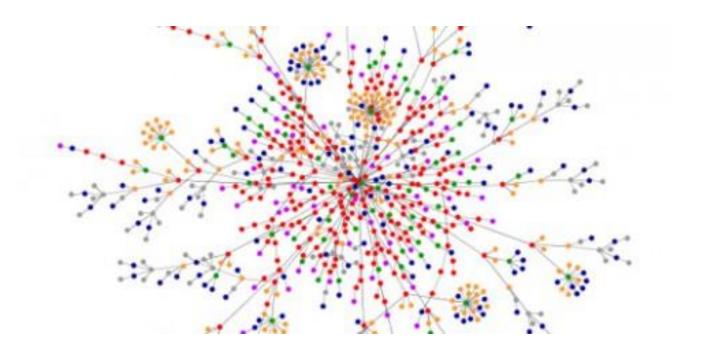
Results for bestseller ASINs

Clump outliers – Using expert input



Clump outliers – Using expert input

- 1. Select "exemplars"
- 2. Get judgments
- 3. Propagate judgments



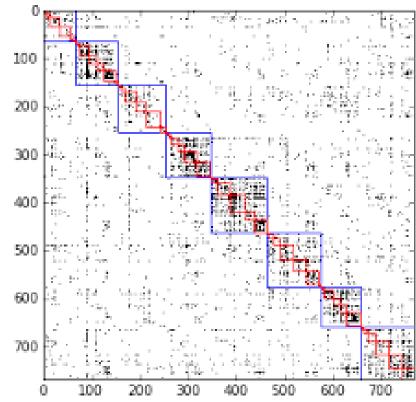
Clump outliers – Affinity Matrix

Constructing a graph adjacency/ASIN affinity matrix:

- 1. Vectorize ASINs using averaged GloVe
- 2. Find the 250 nearest neighbors of each ASIN
- 3. Construct matrix using (Euclidean) distances to nearest neighbors
- 4. Apply Gaussian kernel to each non-zero entry:

$$A_{ij} = \exp(\frac{-dist(\mathbf{x}_i, \mathbf{x}_j)}{2\sigma^2})$$

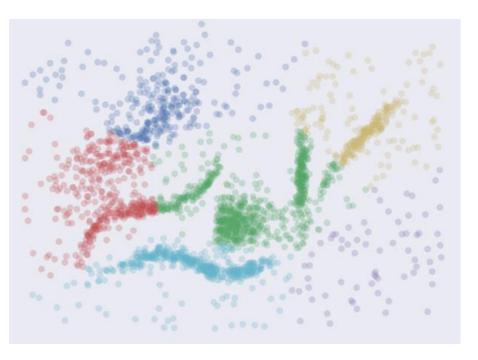
5. Symmetrize matrix



Clump outliers – Exemplar Selection

Choosing a set of "representative" ASINs:

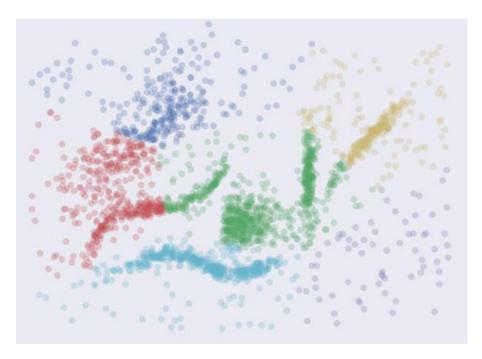
- 1. Spectral clustering using affinity matrix
- 2. Compute the "degree vector"
- 3. Select ASIN with highest "degree" in each cluster



Clump outliers – Exemplar Selection

Choosing a set of "representative" ASINs:

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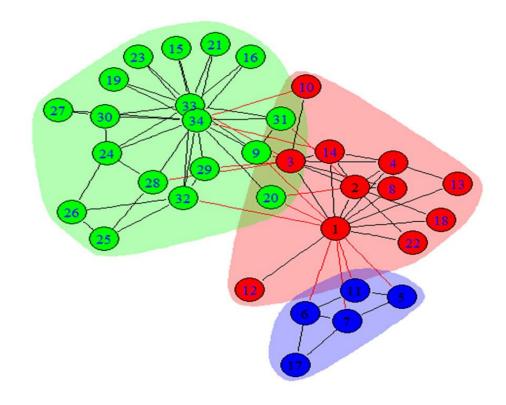


Obtain judgments for selected ASINs

Clump outliers – Judgment Propagation

Propagating expert judgments:

- 1. Construct $S = D^{-\frac{1}{2}}AD^{-\frac{1}{2}}$
- 2. Iterate $F(t + 1) = \alpha SF(t) + (1 \alpha)Y$ until convergence (for α between 0 and 1)
- 3. Obtain labels for ASINs from the limit of F(t) by applying thresholds



Clump outliers – Results

Dataset	Total ASINs	Negative exemplars	Negative Propagations	Miscat Precision	Miscat Recall
TVs	19,345	20/50	5,117	98%	77%
Storage Benches	2,123	19/20	1,647	100% (est.)	-
Radar Detectors	15,225	13/19	9,427	100% (est.)	-
Cooktops	22,266	40/50	19,263	100% (est.)	-
Dart Cases & Wallets	28,992	98/100	27,151	100% (est.)	-
50 UK categories	1,748,936	2,372/5000	818,176	95.8% (audit)	-

Towards a Pristine Catalog

Questions?

