Product Categorization

A Text Classifier for Food Data



Evelyn Trautmann, Sebastian Hansmann, Sumit Sidana

Takeaway

41 Million orders last quarter

~ 19 Million products



~ 40 k active restaurants

Active in 12 countries

Looking for a Data Engineer right now



Agenda

- Introduction
- FastText
- Open Food Dataset
- Similarities
- Multilabel Classification
- Applications



Introduction

- Product catalogues are common entities in business
- Categorizing unstructured items according to catalogue
- Catalogues often contain a vast amount of classes







String Matching

Misspellings



String Matching

- Misspellings
- Word Concatenations



String Matching

- Misspellings
- Word Concatenations
- Different Contextual Meaning





Text Classifier

fastText





recent NLP developments





Model Architecture

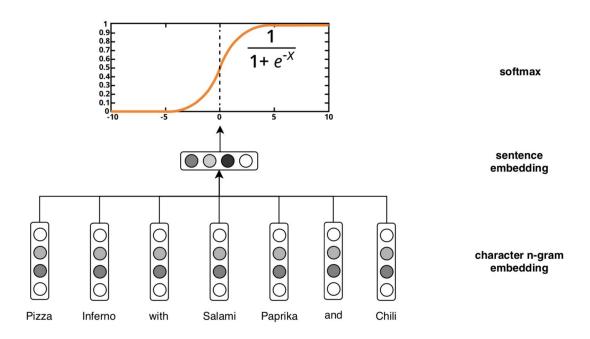
- Character n-gram Embeddings
 - CBOW
 - skip-gram
- Averaging n-gram features
- Softmax

Pizza Inferno with [Salami] Paprika and Chili

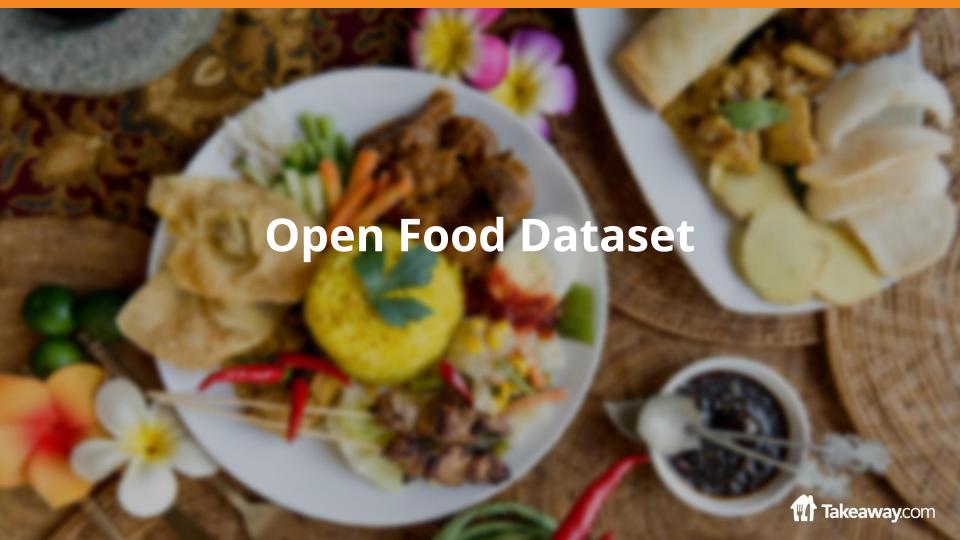
[Pizza] [Inferno] [with] Salami [Paprika] [and] [Chili]



Model Architecture







Dataset

List of categories - World

25840 categories:

	Search:		
Category	\$	Products _	• 0
Plant-based foods and beverages		108367	
Plant-based foods		92833	
Snacks		53620	
Beverages		49224	
Sweet snacks		43829	
Dairies		39904	
Cereals and potatoes		31198	
Meats		30341	
Non-Alcoholic beverages		30099	
Fruits and vegetables based foods		30067	
Fermented foods		28201	
Fermented milk products		28118	
Meals		27465	
Groceries		22671	









Problem Description

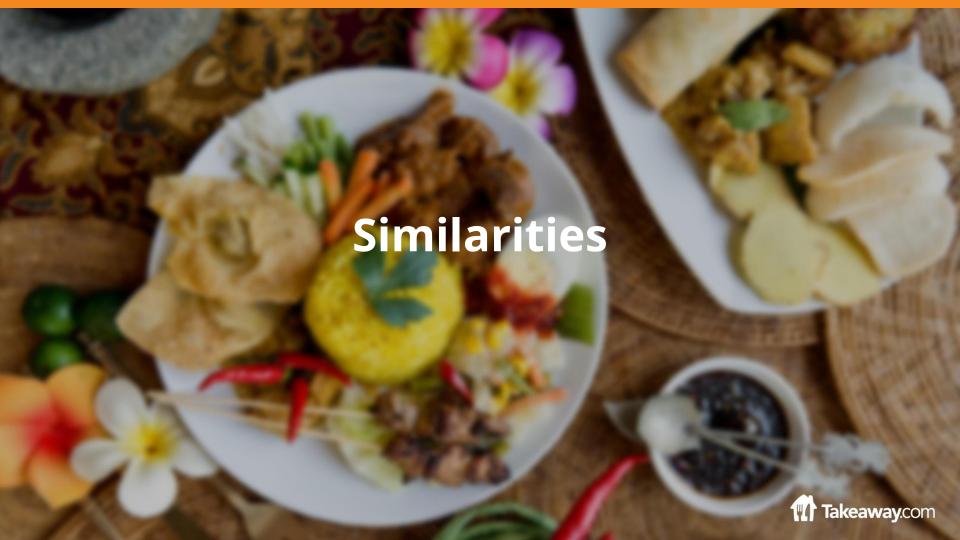
Can we assign appropriate categories to specific products given its product name, generic name and brand?

Classification Demo

http://localhost:8889/notebooks/OpenFoodFactExample.ipynb

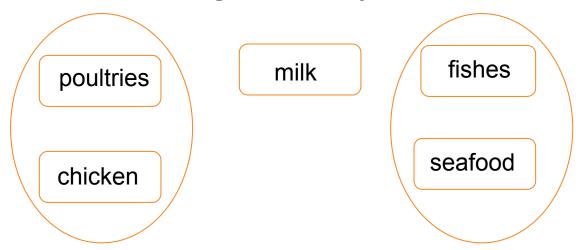
GitHub: https://github.com/metterlein/evaluate_supervised/





Similarity Matrix

- Not all classes miss-classifications are actual miss-classifications
- Some classes exhibit higher similarity



Classification Metrics wrt Similarities

In Precision calculation numerator changes

From:
$$Pr(i) = \frac{C_{ii}}{\sum_{i} C_{ij}}$$

Recall likewise

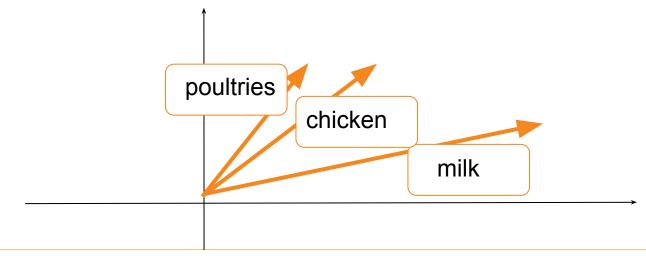
To:
$$Pr(i) = \frac{\sum_{j} C_{ij} S_{ij}}{\sum_{i} C_{ij}}$$

Weighted Truth - Prediction Pairs

Confusion Matrix:
$$C = \begin{pmatrix} C_{11} & \dots & C_{1n} \\ \dots & & \dots \\ C_{n1} & \dots & C_{nn} \end{pmatrix}$$

Determine Similarity Matrix

- Word Embeddings contain context information
- Labels appearing in similar contexts are close to each other
- Distance between word vectors





- Fasttext Default: Precision@k, Recall@k
- Only Average, not per class
- Scikit Learn: Confusion Matrix per Class (OvA)
- How can we evaluate approximately accurate predictions?
- Multilabel Confusion Matrix



Classification Report

Multilabel Confusion Matrix

Truth **Prediction** Matches snacks, snacks, snacks, confectioneries, confectioneries confectioneries, sweet-snacks candies chickens, poultries, chickens chickens, appetizers, prepared-meats poultry-meals

Increase **C[snacks,snacks]**and **C[confectioneries,confectioneries]**

Matches: Snacks, confectioneries



- Classification Report
- Multilabel Confusion Matrix

Truth	Prediction	Matches
snacks, confectioneries,	snacks, confectioneries,	snacks, confectioneries
sweet-snacks	candies	
chickens, appetizers, poultry-meals	chickens, poultries, prepared-meats	chickens

Increase

C[snacks,snacks]

and

C[confectioneries,confectioneries]

Increase

C[sweet-snacks,candies]

Miss-classified: candies

Not detected: sweet-snacks



- Classification Report
- Multilabel Confusion Matrix

Truth	Prediction	Matches
snacks, confectioneries, sweet-snacks	snacks, confectioneries, candies	snacks, confectioneries
chickens, appetizers, poultry-meals	chickens, poultries, prepared-meats	chickens

Misclassifications:

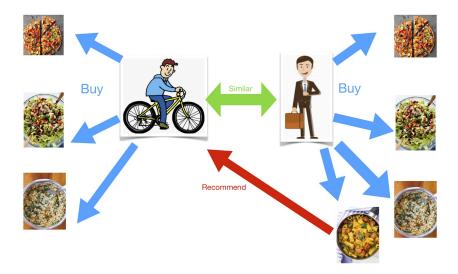
- appetizers, poultries
- appetizers,prepared-meats
- poultry-meals, poultries
- poultry-meals,prepared-meats





Recommendations

- Collaborative filtering: Make use of past order history to make recommendations and no meta categories
- Recommendations without features: Collaborative filtering (an Ok solution, but not enough!)





Recommendations with categories

- Difficult to do Dish-Based recommendations with Interaction data alone!
 - 6 M distinct dishes
 - Pizza 12 inches = Pizza 24 inches (what we have now)
 - Both need to be classified as Pizza (using categories)
 - Need a common ground (such as a broad category!)
- Collaborative filtering with meta-information for restaurant recommendations
 - Embed categories in CF model for restaurants
 - Handle Restaurant cold start (problem of recommending new restaurants)
- Customer Segmentation
 - Clustering on the basis of categories



References

- Joulin, A., Grave, E., Bojanowski, P. & Mikolov, T. (2016). Bag of Tricks for Efficient Text
 Classification (cite arxiv:1607.01759) https://arxiv.org/abs/1607.01759
- https://ai.facebook.com/blog/fasttext-blog-post-open-source-in-brief/
- Wikipedia contributors. "Open Food Facts." Wikipedia, The Free Encyclopedia. Wikipedia, The Free Encyclopedia, 13 Sep. 2019. Web. 2 Oct. 2019.
- https://github.com/metterlein/evaluate_supervised/



Thank You!

Questions?





Takeaway.com brands

Lieferando.de Lieferservice.at Lieferservice.ch Pizza.be Pizza.fr Pizza.lu Pyszne.pl Pizza.pt Thuisbezorgd.nl and Vietnammm.com