

A Pipeline for Characterizing Fashion Items' Online Popularity

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Introduction & Dataset

Background

• The fashion retail industry is interested in using novel sources of data to forecast item popularity. As a preliminary step, it could be useful to explore how to generate data that best characterize an item's online presence.

Objective

• Identify features that characterize an apparel's online popularity and explore clustering relationships among items.

Data Set

- Provided by Chain of Demand firm which uses Al to track and predict sales demand for fashion brands
- 2 spreadsheets from brands *Uniqlo* and *Esprit* containing item-level features
- 2382 Uniqlo items; 188 Esprit items

Product.Name <fctr></fctr>	Category_1 <fctr></fctr>	Category_2 <fctr></fctr>	Color <fctr></fctr>	Material <fctr></fctr>
AIRism SLEEVELESS TOP 422267	women	sport utility wear	GRAY	59% Nylon, 31% Cupro, 10% Spandex
HEATTECH SLEEVELESS TOP 418327	women	innerwear & loungewear	OFF WHITE	38% Polyester, 32% Acrylic, 21% Rayo
CORDUROY MINI SKIRT 418882	women	bottoms	YELLOW	99% Cotton, 1% Spandex
FLEECE SET (LONG SLEEVE) 421705	women	tops	BROWN	Tops: 100% Polyester/ Rib: 90% Cotto
HEATTECH WARM LINED PANTS 420360	women	bottoms	NAVY	Shell: 90% Polyester, 10% Spandex/ L

Table 1: Five Random Samples From Uniqlo Dataset

Data Processing

Search Term Generation

- The goal: from a row in the dataset (e.g. Table 1), automatically extract search terms similar to how people would actually search this item online
- Examples: {sleeveless top; heattech innerwear; off white loungewear; polyester sleeveless top} are all valid search terms people would choose to query this item online
 - Turned all text into
- lowercase • Removed all non alphabetic

characters

- N-Gram Tokenization • Improved token
- quality using part-ofspeech tagging
- Tokens from product name & categories are grouped as the base set

Compose Search Terms

- Tokens from color & materials are grouped as the adjective set
- Scraped each term's search popularity on Google
- Multiple time series described one item

Time Series Scraping

• Search term = adjective + base

Data Cleaning Tokenization

"HEATTECH "heattech SLEEVELESS TOP_{sleeveless} top" 418247"

{"heattech", "heattech sleeveless", "sleeveless top", "top", "innerwear",

"loungewear"; "polyester"}

{"heattech loungewear"; "off white sleeveless top", "polyester top"}

Figure 1: Data Processing Workflow with Example

Figure 2: Google Trends Time Series

Time Series Generation

- Used pytrends, an unofficial API for Google Trends, to download historical interest over time for each search term.
- Weekly timestamps from Jan. 2016 to Dec. 2019
- Score from 0 to 100 is assigned to indicate each term's current search frequency compared to its historical high
- Resulting numbers not comparable across search terms, but meaningful within each term over time
- Maxed out on Google's rate limit for API calls; only generated time series for 100 Uniqlo items

Feature Extraction & Clustering

Feature Aggregation

- For each item, we have several time series
- All term's trend scores are averaged for each item before feature extraction
- Some features encompass multiple values and are thus broken into different features

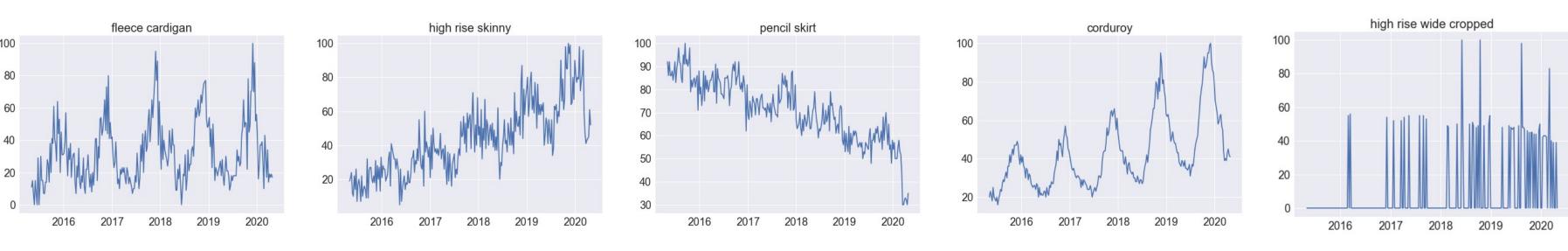


Figure 3: Example Time Series Trends We Wish to Capture

Time Series Feature Extraction

- Want to capture cyclical, rising, dropping, or booming trends
- Features extracted using the TSFEL (Time Series Feature Extraction Library) package that provides three domains:
- Spectral: describe the spectrum by representing the series with combinations of sinusoids
- Statistical: describe the observed values' marginal distribution
- o **Temporal**: describe the time features of the native time series
- Focus on features like autocorrelation, entropy, percentile, slope, and model coefficients.

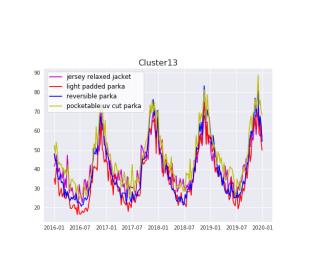
Results

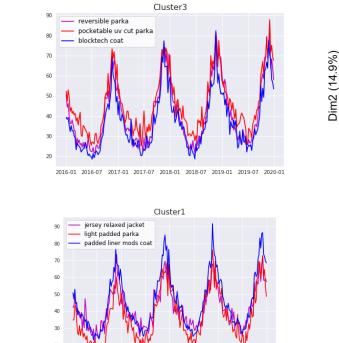
Clustering

- Clustering done using K-means Algorithm
- Use 3 and 4 clusters

Improvements

- Trends of clusters not very clear
- Data points in sample too similar
 - API limitation on downloadable data
- Search term categories too heavily weighted





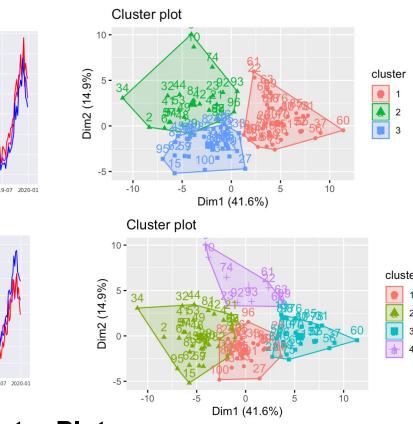


Figure 4: Example Cluster Plots

Social Media

Summary

- Google: Monthly search volume from Google Ads
- Monthly search volume's range is to big (ex. 1K-10K,10K-100K)
- Instagram: Attempted to input search terms into a keyword/hashtag tracking service
 - Tracking service not applicable for high volume of keywords

References

- RWeka Library in R for tokenization
- Natural Language Toolkit library nltk for part of speech tagging Pytrends for scraping Google Trends data
- TSFEL (Time Series Feature Extraction Library) Package for time series feature extraction
- Base R K-means Algorithm for clustering
- Factoextra Package in R for visualization of clustering
- Data was provided by Chain of Demand, a firm which uses AI and big data to track and predict sales demand for fashion brands