Scoring Dimensions of User Reviews

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Motivation

Problem

- Ratings are not sufficient
- Too many reviews

Solution

- Summarize reviews by identifying and scoring features.
 - Eg: Features for restaurants food ,quality, service, ambience
- Support search based on preferences
 - Eg: Student prefers worthiness opposed to ambience
- Predict the rating of the restaurant for that user

Uniqueness

- Crowd based feature identification
- Determining opinion strength instead of opinion orientation

Visualization

Salient features about Green:



place: Opinion Count - 308 | Sentiment - 6.92

Algorithm - Feature Scoring

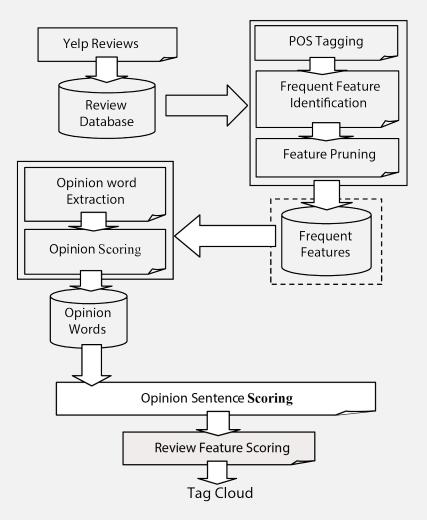
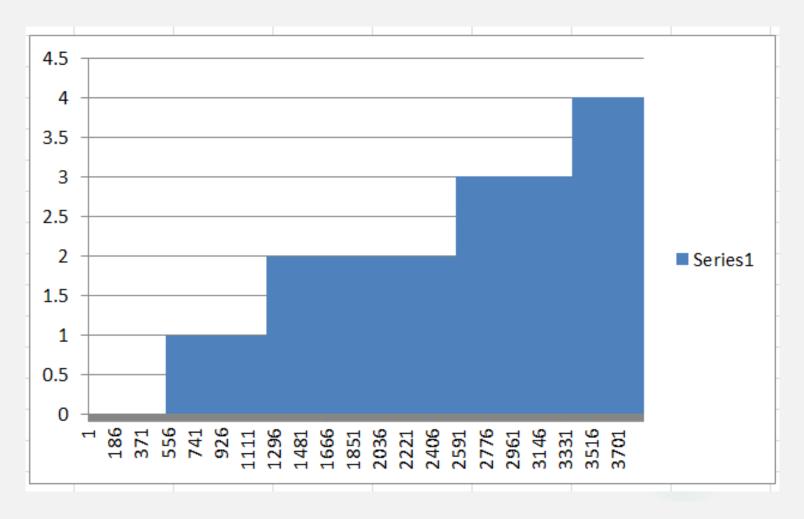


Figure: Feature -based opinion summari zation

Data

- Yelp Dataset from the greater Phoenix, AZ metropolitan area including:
 - 15,585 businesses
 - 70,817 users
 - 335,022 reviews
- Total Size 353 MB JSON file
- Total Restaurants 5556
- Top 50 Restaurants with total 26,746 reviews
- 535 Average Reviews Per Restaurant

Results Sentiment Dictionary



Comparison



Search Products

Sony Cyber-shot DSC-W370 14.1 MP Digital Camera (Silver)

Overview - Online stores - Nearby stores - Reviews - Technical specifications - Similar items - Accessories



\$140 online, \$170 nearby





5 stars

Reviews

Summary - Based on 159 reviews

3 stars

What people are saying

pictures

"We use the product to take quickly photos."

4 stars

"Impressive panoramic feature." features

"It also record better and focus better on sunny days." zoom/lens

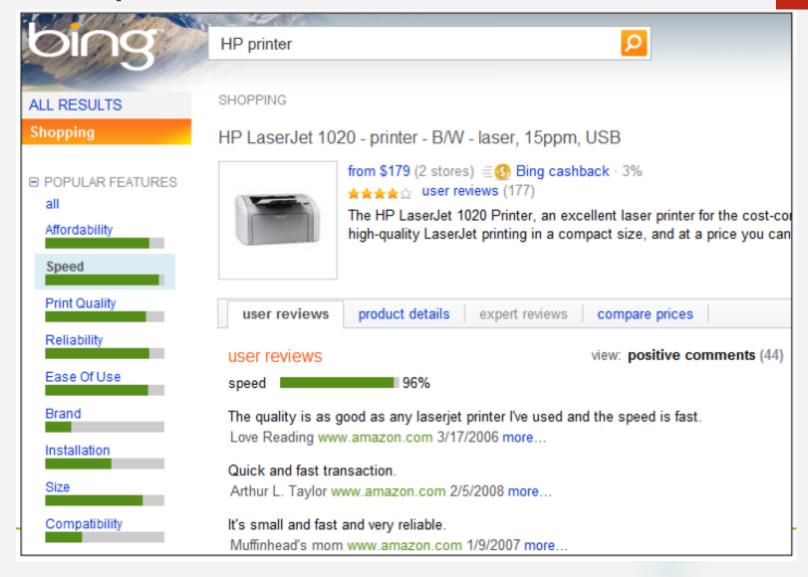
"It has the slightest grip but it's sufficient." design

"Video zoom is choppy." video

"Even better, the battery lasts long." battery life

"I Love the Sony's 3" screen which I really wanted." screen

Comparison



Algorithm - Rating Prediction

- Latent Dirichlet Allocation (LDA)
 - Discover hidden dimensions
 - Dimensions are not correlated with ratings
 - Supervised LDA combines both ratings and review
- Matrix Factorization Models
 - Alternating Least Squares
 - Stochastic Gradient Descent
 - Bias-Stochastic Gradient Descent
- Hidden Factors and Topics (HFT)
 - Associate latent rating dimensions with latent review topics
 - Use Discover features in reviews

$$f(\mathcal{T}|\Theta, \Phi, \kappa, z) = \sum_{r_{u,i} \in \mathcal{T}} \underbrace{\left(rec(u,i) - r_{u,i}\right)^2}_{\text{rating error}} - \mu \underbrace{l(\mathcal{T}|\theta, \phi, z)}_{\text{corpus likelihood}}.$$

Data

For Top 50 restaurants

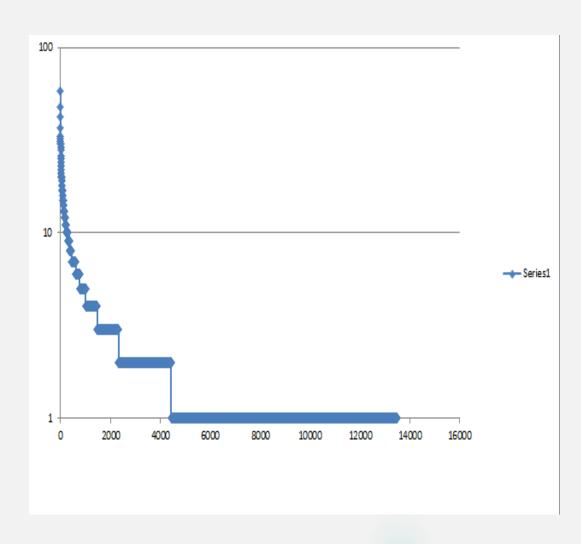
- 13,499 total users
- 26,746 user ratings

Subset

- 294 user with 10 or more reviews
- 4,437 user ratings

Prediction

14,700 user ratings



Prediction Results

Method	Number of Latent Factors	Training RMSE	Validation RMSE
ALS	20	0.220959	1.3471
ALS	40	0.0963635	1.6252
SGD	20	0.423545	1.3556
SGD	40	0.3834435	1.4567
Bias-SGD	20	0.402323	1.1233
Bias-SGD	40	0.39897	1.0933

Conclusion

Value

- Business owners
 - Evaluate performance
 - Compare with competitors
- Users
 - Reduce time and effort to find business as per personal preferences

Future Potential

- Implicit Feature Identification
- Sentiment Analysis of Comparative Opinions