

Tip Ranker: A M.L. Approach to Ranking Short Reviews



Enrique Cruz, Berk Kapicioglu

The Problem

Foursquare users have written over 87 million tips (short reviews) at millions of different venues and places around the world.

Whether to proactively ping a user with the best tip when he is at a given place, or to order the tips shown on a venue's page to maximize quality of the content - How do we best rank tips?

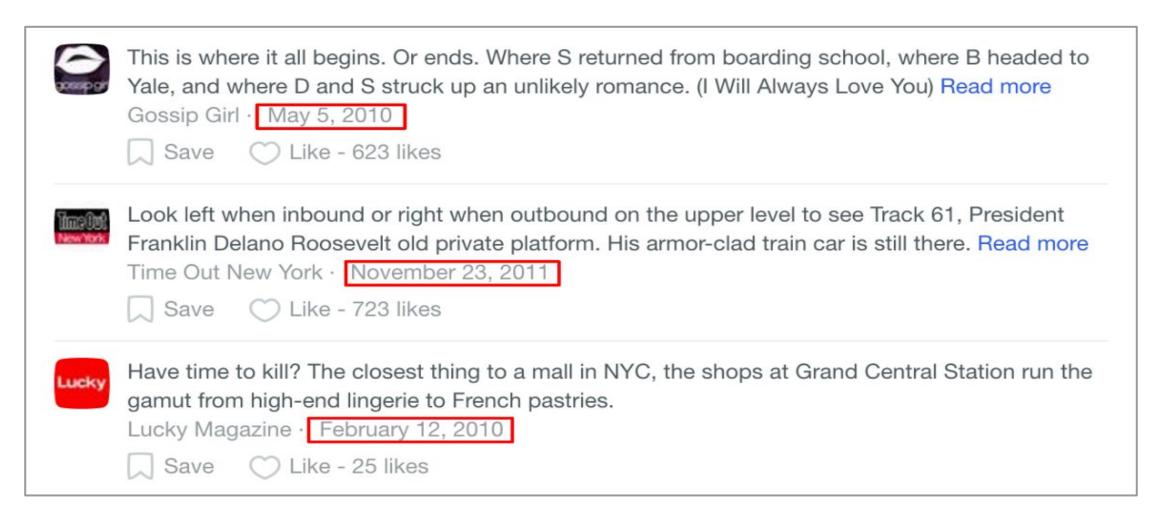
Background

- Foursquare has previously tried used a few different mechanisms to tackle tip selection.
- None of these were fully satisfactory on their own, each having its particular set of shortcomings.

POPULARITY Measurements of social interactions on tip.

Showcases relevant content that is interesting and useful to users.

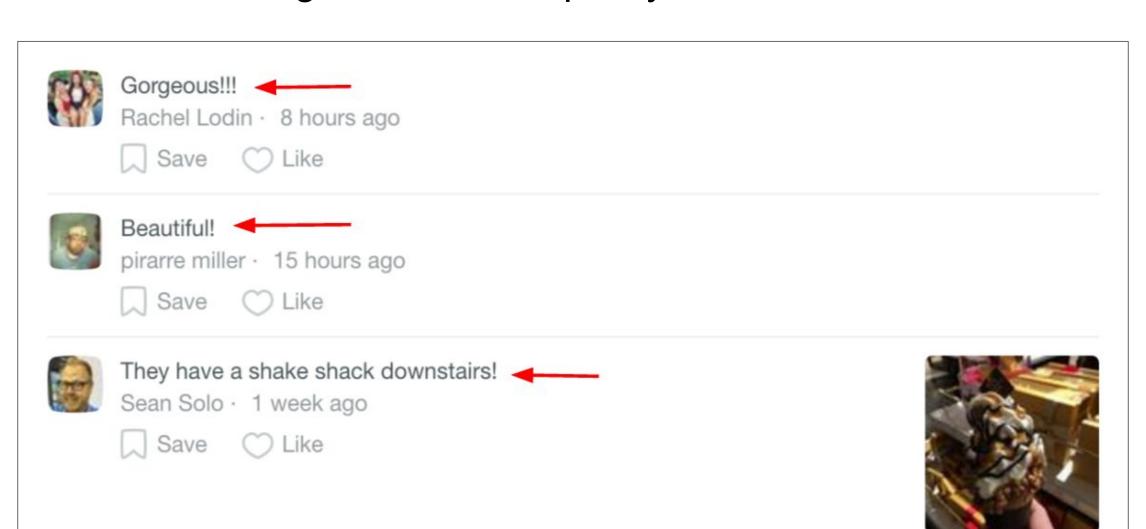
Favors old & stale content. Leads to a feedback loop prominently exposing old content.



RECENCY Time that has passed since the tip was created.

Keeps the content fresh and showcases our vibrant and active community.

Offers no guarantees of quality or relevance.

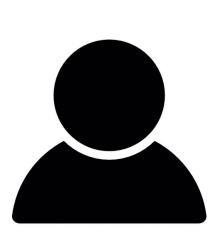


Feature Components



LANGUAGE IDENTIFICATION

Prefer tips written in the user's language

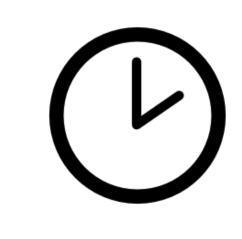


AUTHOR TRUST

User statistics (i.e. tenure and historical popularity of tips)



POPULARITY Measure of social interactions



RECENCY

Time elapsed since tip creation



META-CONTENT RICHNESS

Presence of photo, tastes and number of tokens

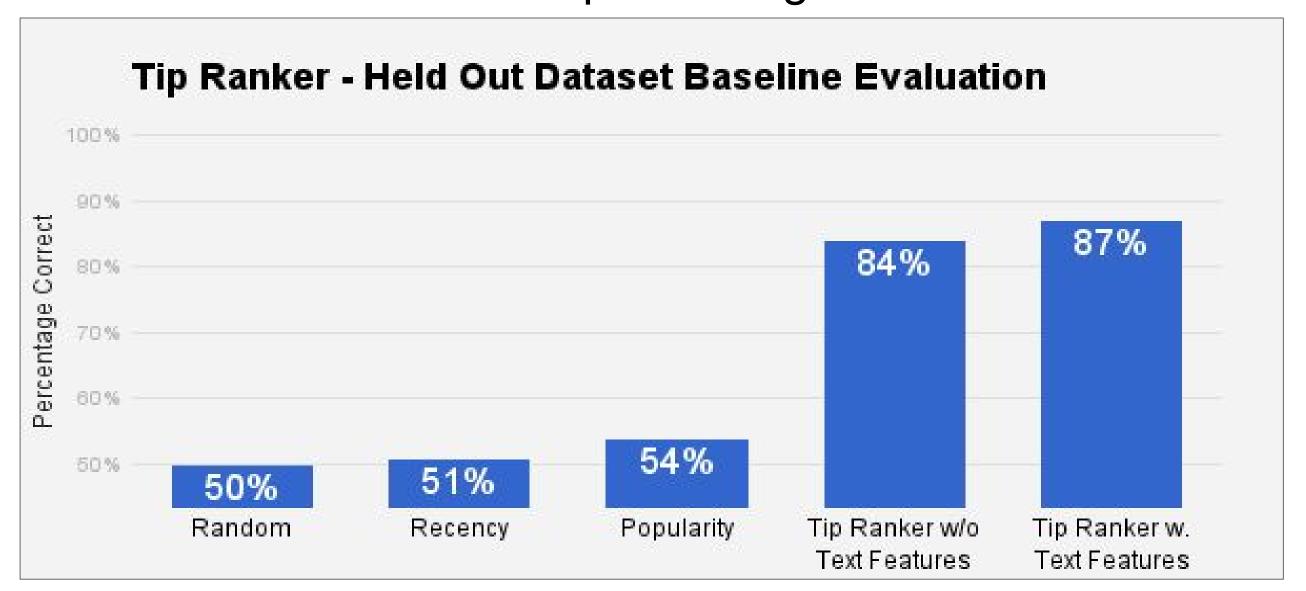


GLOBAL QUALITY

Statistical classifiers trained to identify traits like sentiment and spam

Training & Evaluation

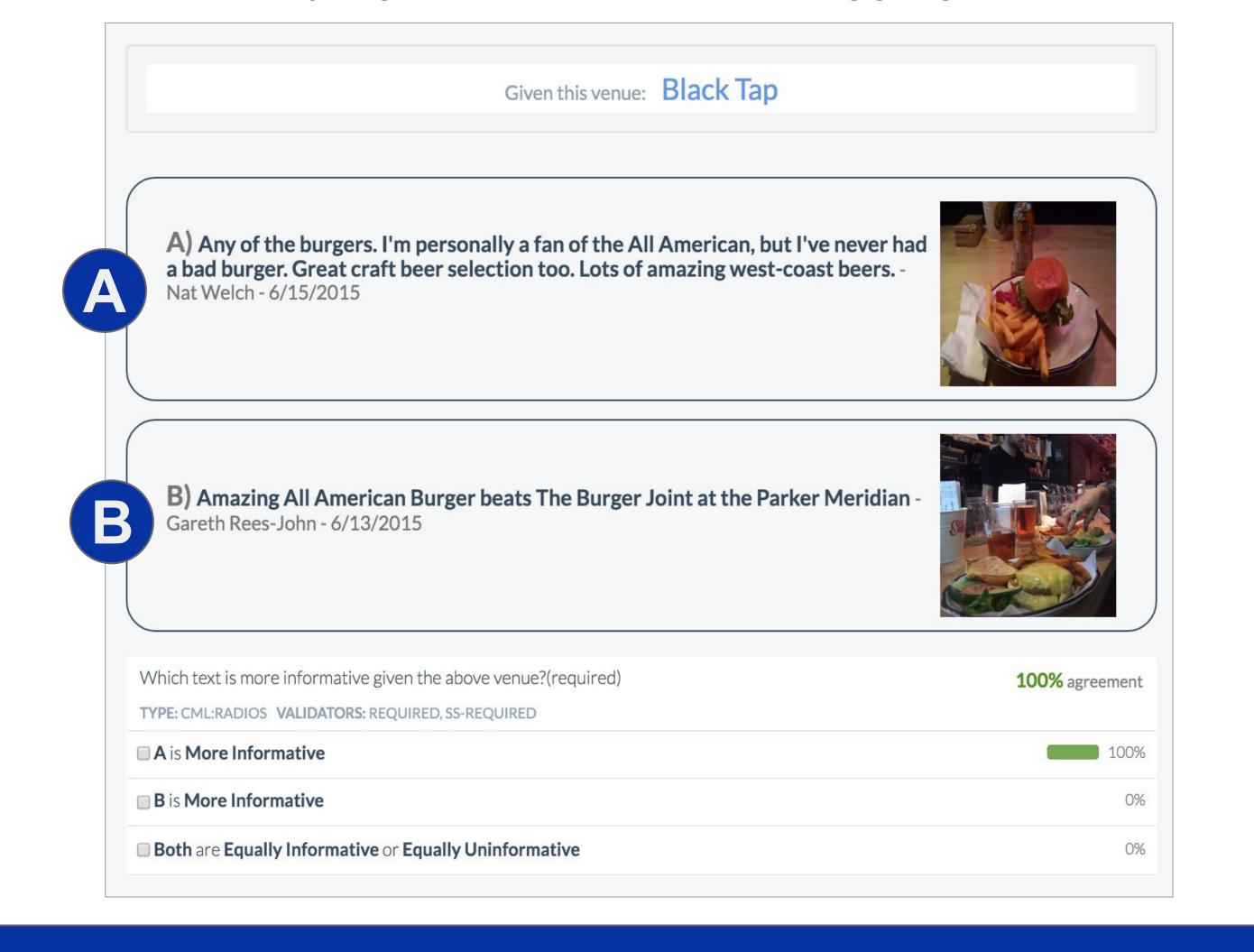
- Supervised learning algorithm: Support Vector Machine.
- Optimization metric: Minimizing the number of misordered pairs of tips with respect to labels.
- Evaluation: Measured the model's performance against a held out dataset and compared it against baseline metrics.



Data & Labels



- 1. Select the top 10K venues by user views
- 2. From each of these randomly sample 100 pairs of tips
- 3. Language filter and de-dupe \rightarrow 50K pairs total
- 4. Collect labels via crowdsourcing
 - a. Show tip pairs in context and ask judge for ordering
- **b.** Verify judge trust with honeypot
- c. Collect judgements per pair and aggregate



A/B Testing & Production Applications



1 1 2 2 % more pings sent **1 2.5** % increase in global SEO referral traffic

