

Dynamic Online Profiling of Users' Content Consumption Preferences

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What Is Outbrain?



Distribution Partners





PAGE VIEWS

PER MONTH



200 BILLION RECOMMENDATIONS

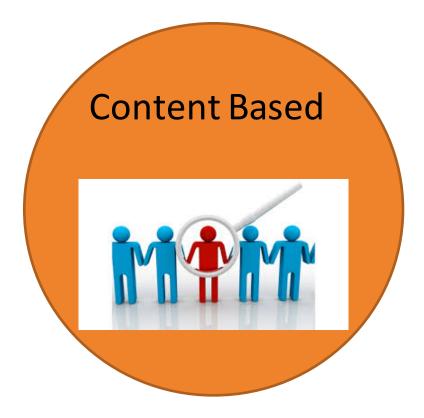
SERVED PER MONTH

The Lighthouse

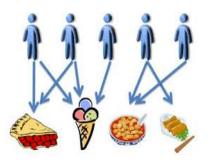
Help people discover content they can trust to be interesting, relevant, and timely for them

Personalization

We use 2 approaches:



Collaborative Filtering



This talk

User profiles lifecycle

- <u>Extract</u> predictive features from documents
- <u>Construct</u> user profiles incrementally
- <u>Match</u> content with user profile

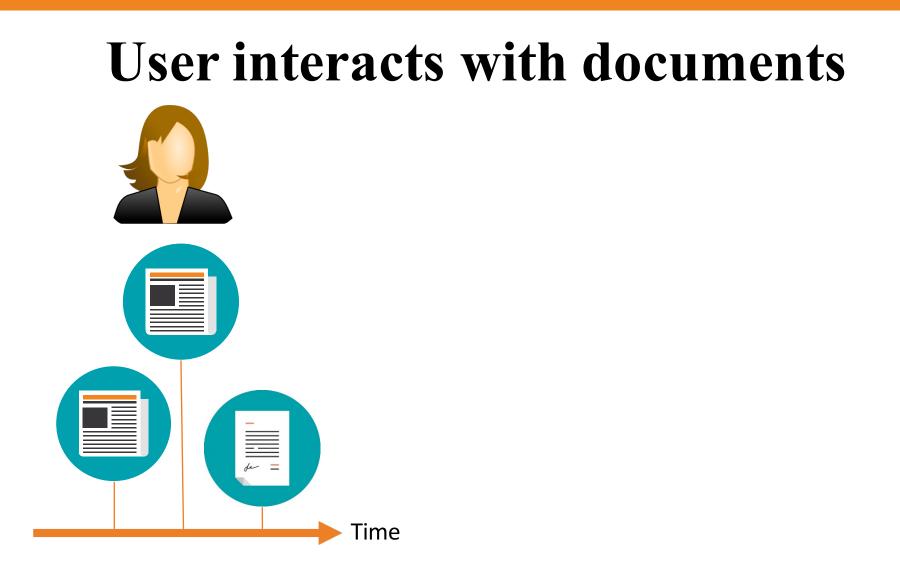


Constructing the profile

Interesting algorithmic questions due to:

- Content understanding challenges
- Online updates
- Dynamic interests
- Scale

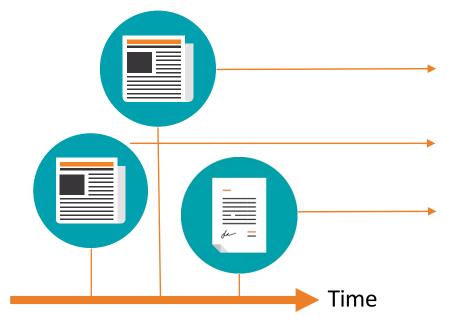




Interaction types:

Page Views, Clicks, Shares, Likes, Video play/pause, ...

Extract rich features from documents



- Sites
- NLP features: categories, topics, entities
- Cartesian
 products of
 features

NLP: What is a document **about**?

• Classic "Bag of words" approach is too high-dimensional

• We want to encode document meaning in a more compact space



Reduced dimensionality

- Categories
 - *Supervised* classification relative to a fixed set of categories
- Topics
 - Unsupervised probabilistic model
 - topic = distribution over words
 - doc = distribution over topics
- Most *relevant* Named entities (person, organization, location)



Illustration: topics



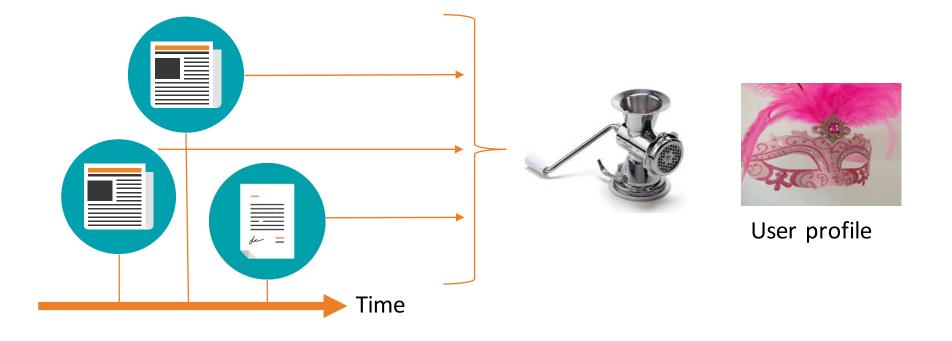


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Aggregate features into a profile



Learning to Aggregate: requirements

• Incremental

• Dynamic

• Low space-complexity



Learning to Aggregate: solutions

• Streaming algorithms:

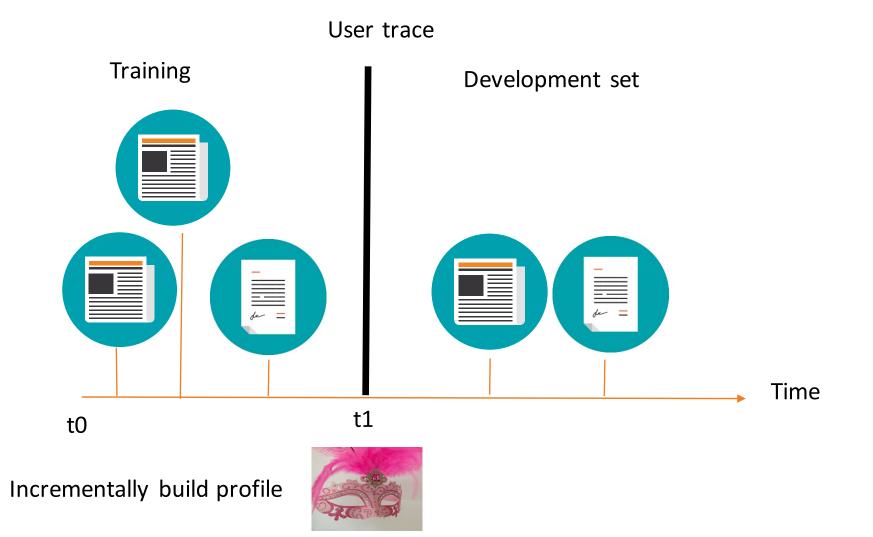
"Lossy" counting over streams: count items but *selectively* forget/decay

• Sketching algorithms:

Compact hashing schemes with partial overlap and redundancy



Offline Evaluation





Offline evaluation: difficulties

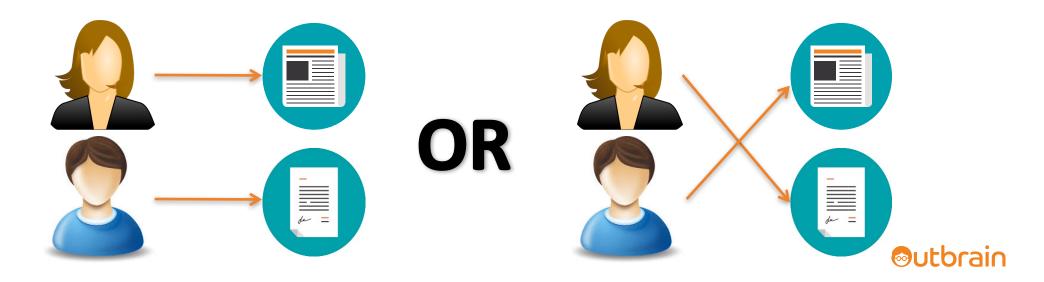
• Positive interactions are observed but negative examples are not

• Heavy bias due to publishers' editorial staff



Offline prediction: Solution

- Pick a pair of users and a pair of docs from their development set
- Does the model prefer actual pairing or counterfactual one?



Conclusion

- Content based profiling is extremely powerful
- Interesting algorithmic challenges in building dynamic online profiles
- Interesting software architecture challenges for a scalable implementation





Thank you rani at outbrain dot com, y @RaniNelken

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