

# Research Content Recommender

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# Bersin by Deloitte.

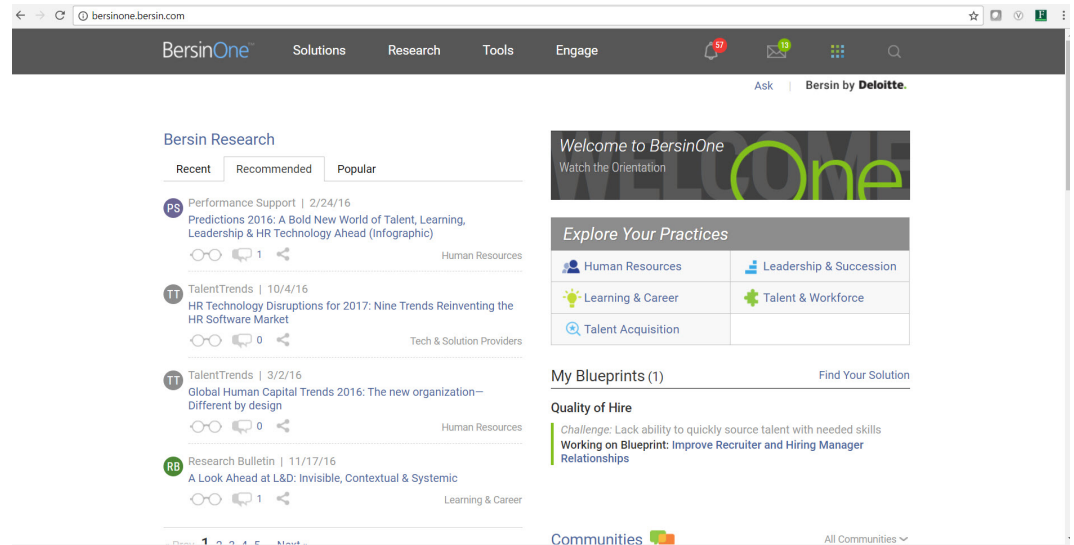
Online research and content service for HR professionals

Provides articles, decision support tools, and online community to help hire, retain, reward best people.

- ~ 45,000 registered users
- ~ 3,000 research items to download

**Recommend content tailored to the user's tastes & increase user engagement and retention**

- Currently recommend most visited content (~ 2 % accuracy)



**GOAL: Harness User Preferences and User Activity to predict favorable content.**

# Data & User Activity

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10 years of User activity data and Content metadata:

- Downloads (~25,000 users)
- Favorites (~1,800 users)
- User's choice of Subject Areas and professional profiles
- Content metadata

Explicit preference measures are minimal.

**Extract as much signal as possible from **implicit** preferences**

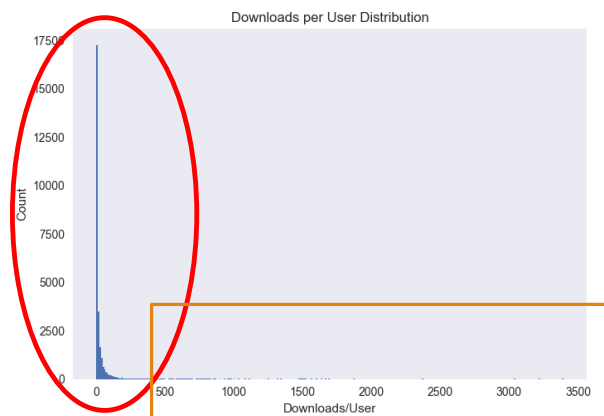
Focus on:

- Download history
- Recommendation User behavior
- Basic User/Item profile features

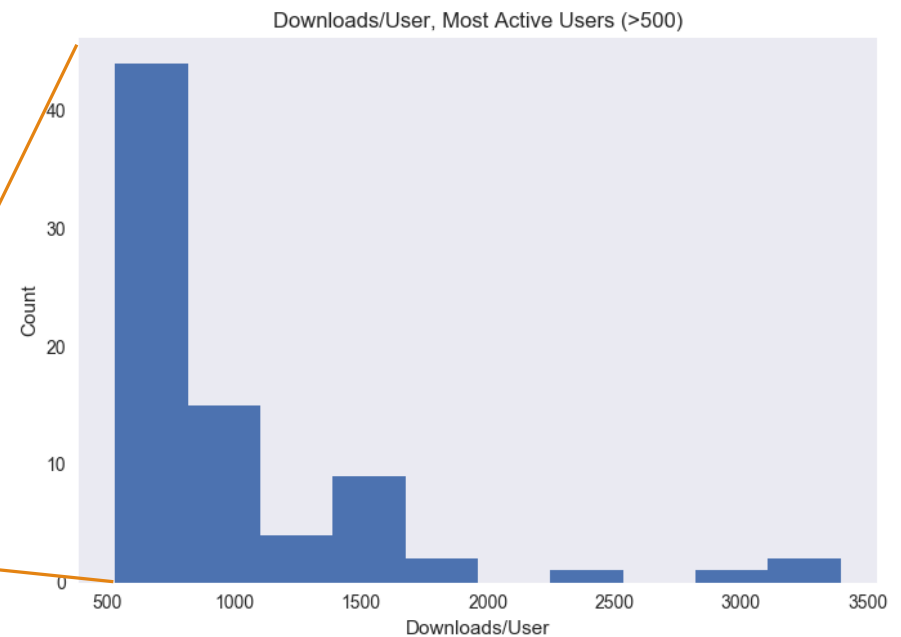


# Data & User Activity

A big **cold-start** and **sparsity** problem

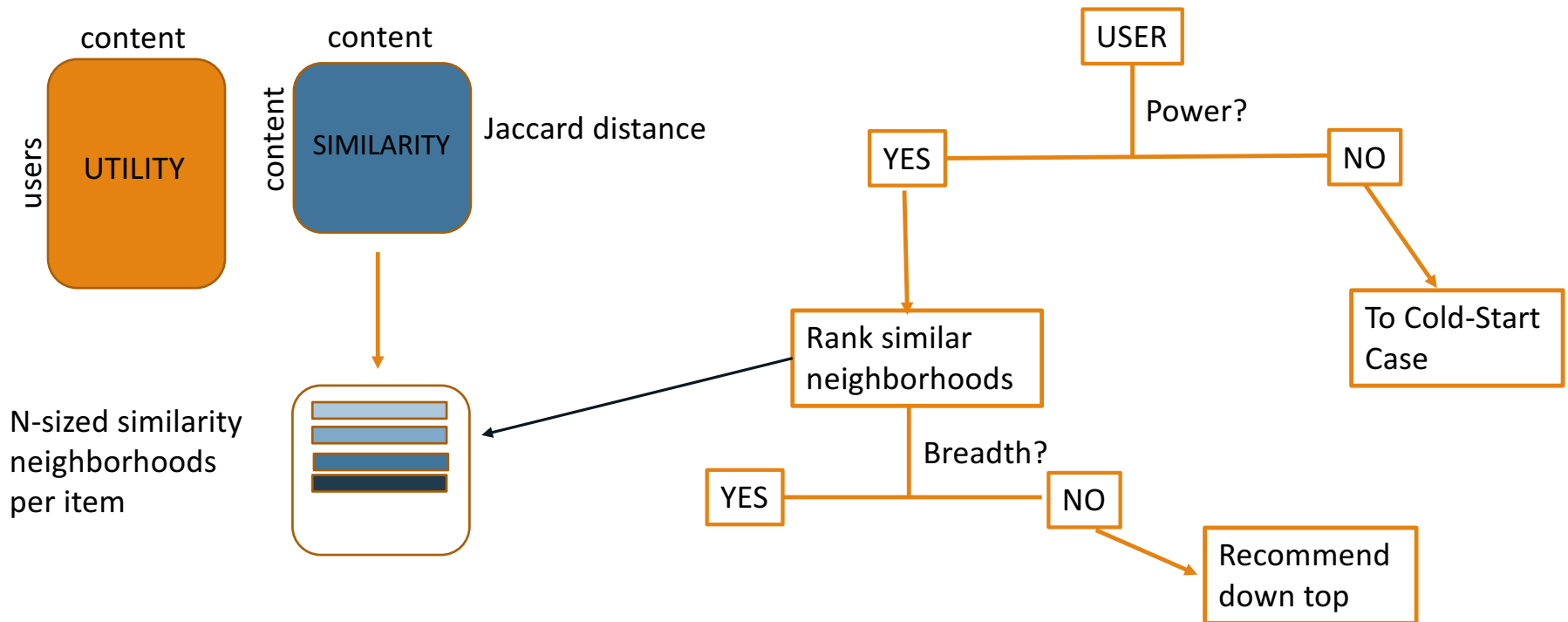


**Vast majority of Users download  
very few documents (1 or 2)**



**Download activity of top 500 users**

# Collaborative Filtering



# Collaborative Filtering

Parameters:

- Neighborhood size N
- Breadth vs. Depth
- Number of Recommendations

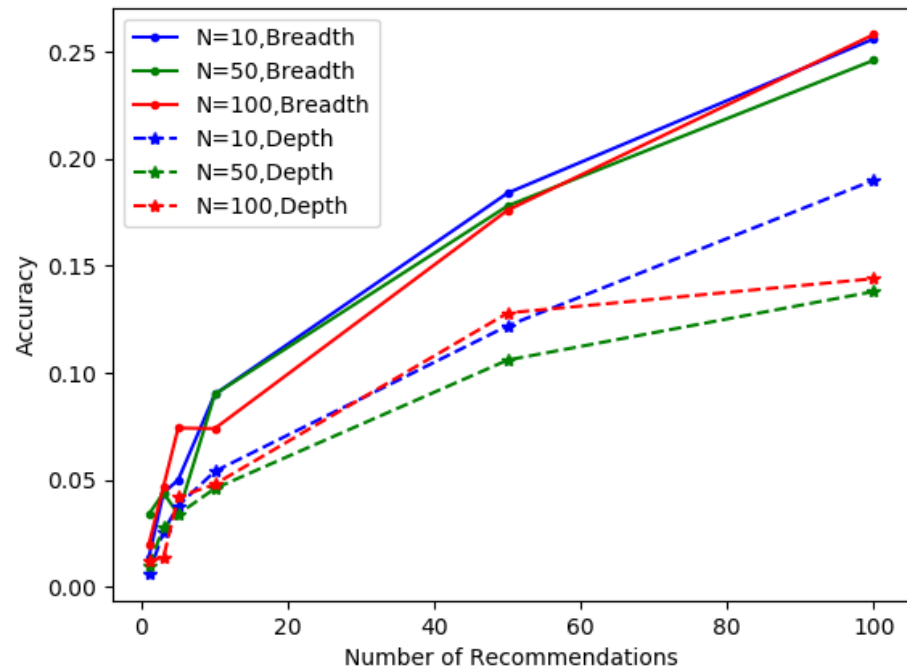
Validation:

- Recommender relies on User behavior
- **Leave-one-out**

**Recommendation improves when emphasizing breadth over depth**

**Use recommender as 1st step to generate initial set, then narrow further with content/profile information**

**Accuracy = Is item in recommended set?**



# Conclusions & Recommendations

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Users prefer breadth to depth in recommendations

Present A/B testing framework to evaluate recommender performance vs. current baseline

Future:

- More in-depth feature work
- Clustering of users/Items to reduce dimensionality of problem
- Other ML techniques to further narrow down from initial set

Cold-start problem: similarity with active users and nudge

Feature work along with cleaning of the data

Investigate time effect: interest drift, trends

Recommendations on data cleaning/management