Improving Search Through *Efficient* A/B Testing:

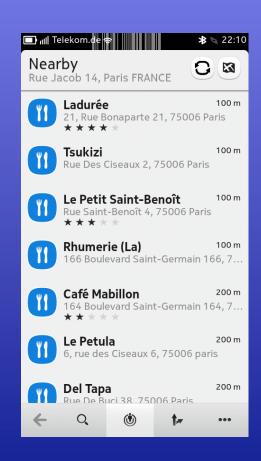
A Case Study

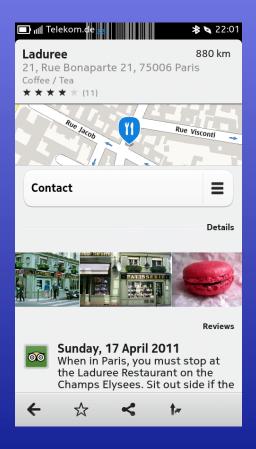
Nokia Maps "Place Discovery" Team, Berlin:

Hannes Kruppa, Steffen Bickel, Mark Waldaukat, Felix Weigel, Ross Turner, Peter Siemen

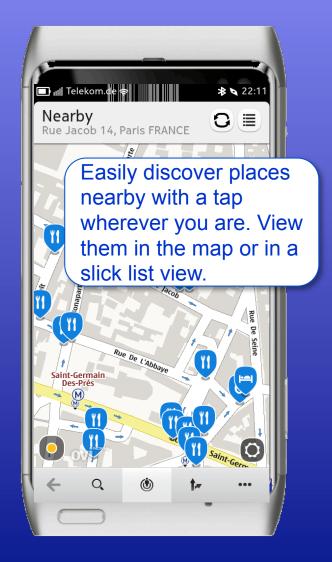


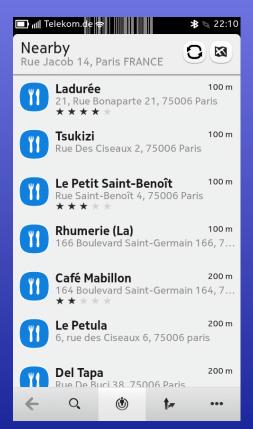


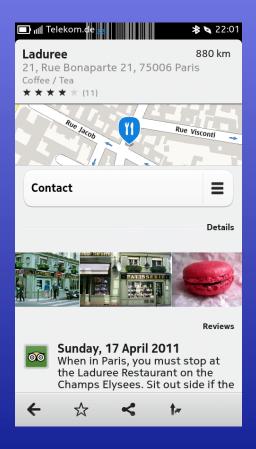




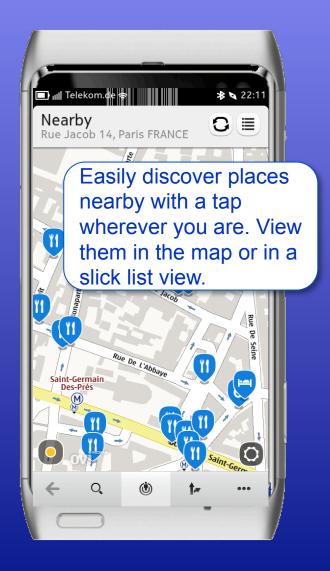


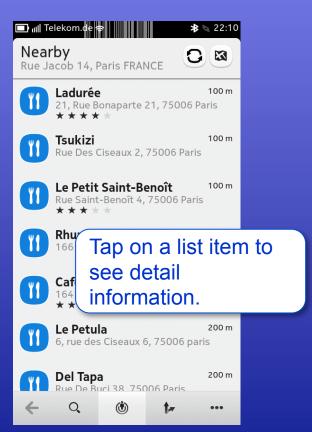


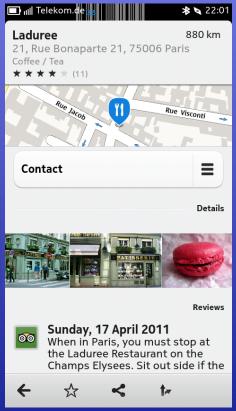


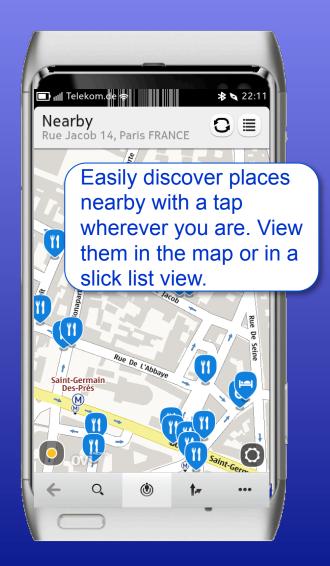


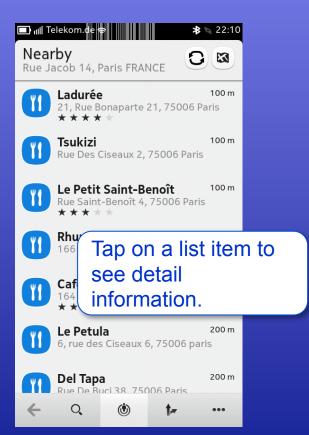


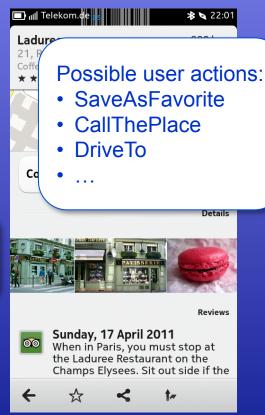












Problem: Which Places to Show?

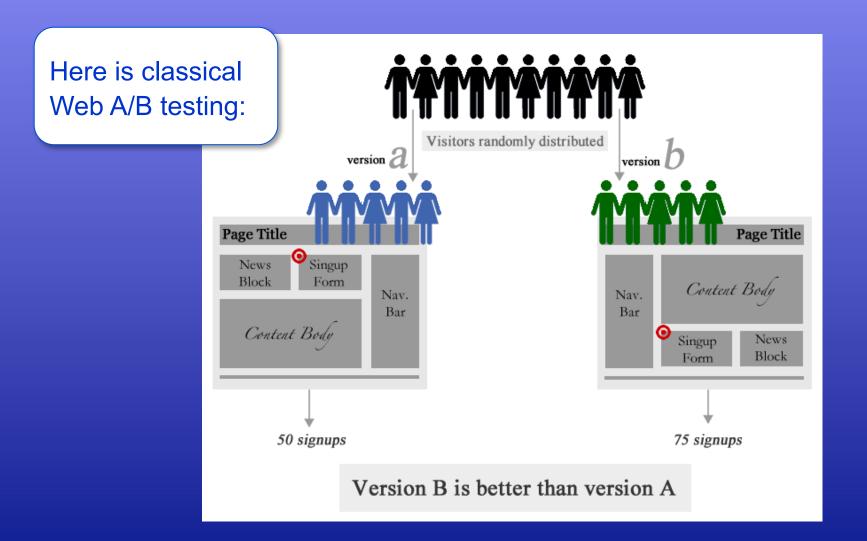
- Restaurants? Hotels? Shopping? ...
- rank by Ratings?
- Distance?
- Usage?
- Trending?

• ...





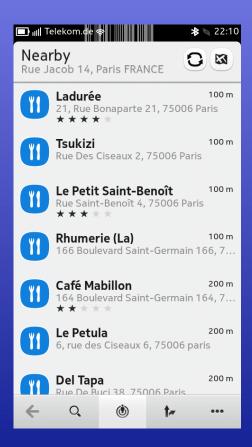
Approach: A/B-Test Different Versions!



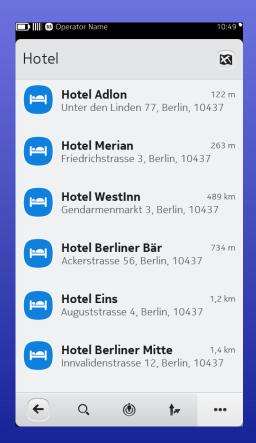


A/B-Test for *Nearby Places*

Version A: Best of Eat'n'Drink



Version B: Best of Hotels



Versions Compete for User engagement:

= Number of Actions performed on places.



There Is A Better Approach For Ranked Lists

[Joachims et al 2008]:

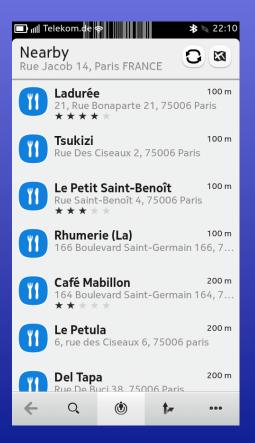
"How Does Clickthrough Data Reflect Retrieval Quality?"

- Classical A/B testing converges slowly for ranked lists
- Classical A/B testing often doesn't reflect actual relevance
- A/B Tests for Ranked Result Lists: Rank-Interleaving
- Use Rank-Interleaving for faster statistical significance

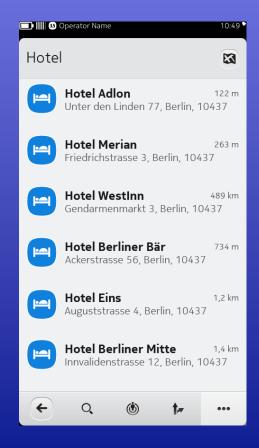


Efficient A/B Testing: Rank Interleaving

Version A: Best of Eat'n'Drink

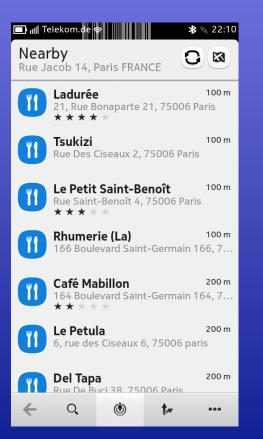


Version B: Best of Hotels

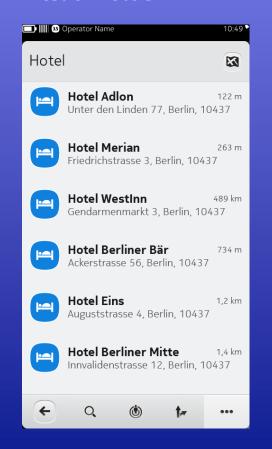


Efficient A/B Testing: Rank Interleaving

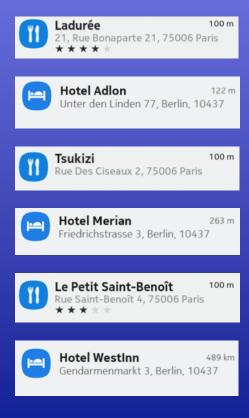
Version A: Best of Eat'n'Drink



Version B: Best of Hotels



Rank Interleaving: *Version A* + *B*





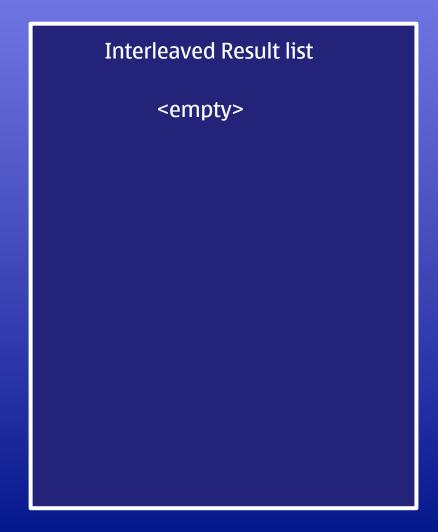
Randomized Mixing of Result Lists

Interleaved list is filled with pairs of results, one item from each version.
 Coin toss decides who comes first.



Version A

- 1. alpha
- 2. beta
- 3. gamma
- 4. delta
- 5. epsilon

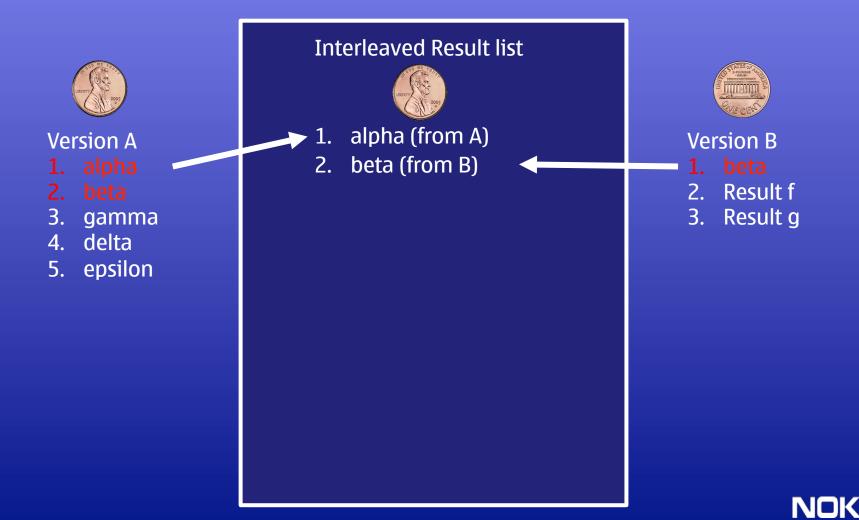




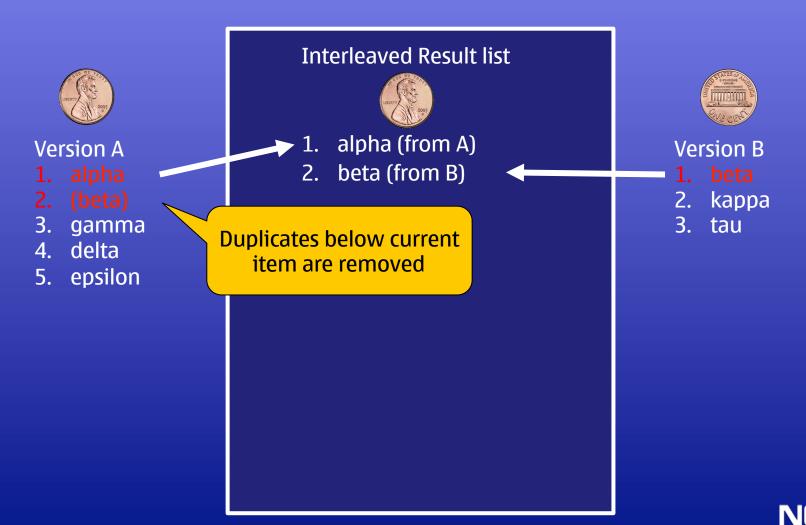
- 1. beta
- 2. kappa
- 3. tau



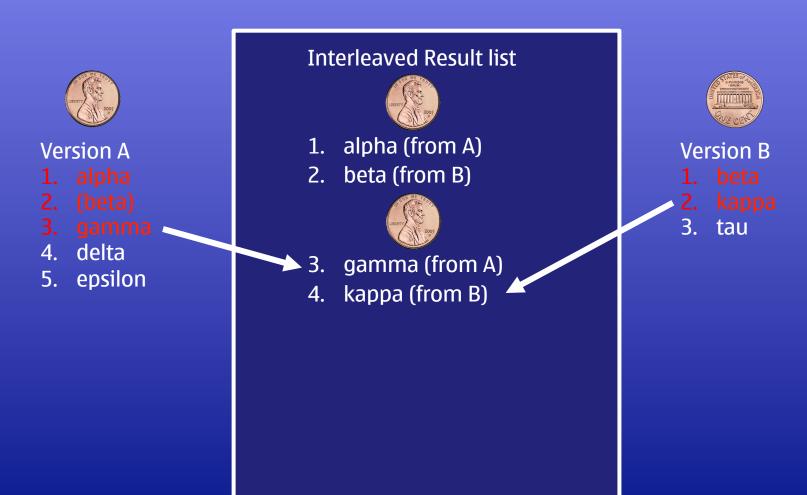
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Interleaved list is filled with pairs of results, one item from each version.
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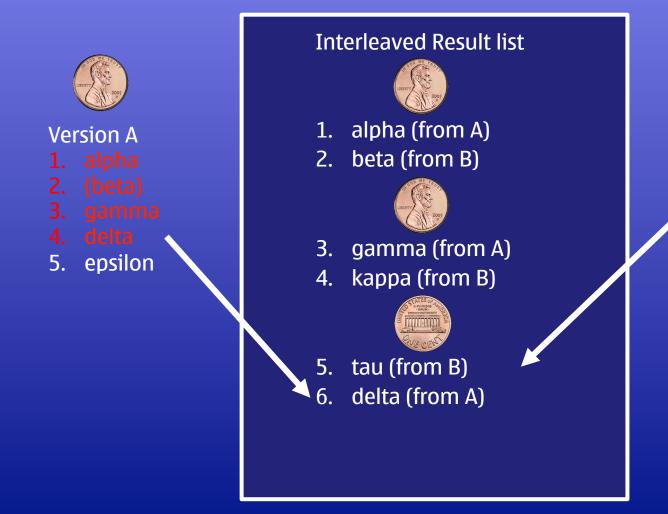


Interleaved list is filled with pairs of results, one item from each version.
 Coin toss decides who comes first.





Interleaved list is filled with pairs of results, one item from each version.
 Coin toss decides who comes first.





- 1. beta
- 2. kappa
- 3. tau

Interleaved list is filled with pairs of results, one item from each version.
 Coin toss decides who comes first.



Version A

- 1. alpha
- 2. (beta)
- gamma
- delta
- 5. epsilon

Leftover results are appended but clicks are not counted

Interleaved Result list



- 1. alpha (from A)
- 2. beta (from B)



- 3. gamma (from A)
- 4. kappa (from B)



- 5. tau (from B)
- 6. delta (from A)
- 7. epsilon (from A, extra)



- 1. beta
- kappa
- 3. tau



Interleaved list is filled with pairs of results, one item from each version.
 Coin toss decides who comes first.



Version A

- 1. alpha
- 2. (beta)
- 3. gamma
- delta
- epsilon

Final list shown to user

- 1. alpha (from A)
- 2. beta (from B)
- 3. gamma (from A)
- 4. kappa (from B)
- 5. tau (from B)
- 6. delta (from A)
- 7. epsilon (from A, extra)



- beta
- 2. kappa
- tau



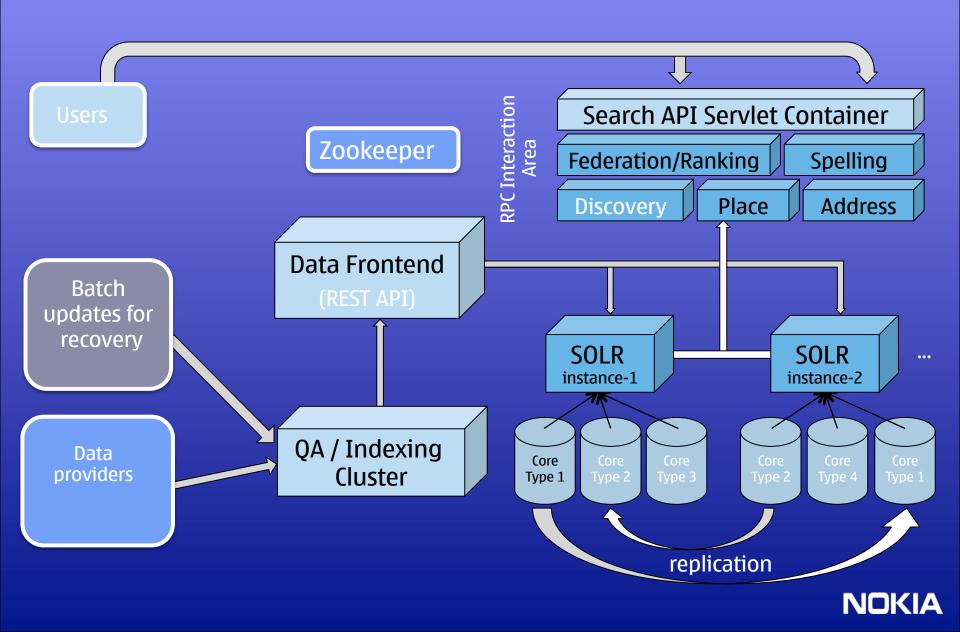
Declaring A Winner

- Statistical Significance Test
- Input (after hadoop-based log-processing...)
 - Number of clicks on version A
 - Number of clicks on version B
- G-Test:
 - improved version of Pearson's Chi-squared test.
 - G > 6.635 corresponds to 99% confidence level
- Null hypothesis:
 - Frequency of counts is equally distributed over both versions.
- Test statistic:

$$G = 2\sum_{i \in \{A,B\}} [\text{counts i}] \ln \left(\frac{[\text{counts i}]}{[\text{total counts/2}]} \right)$$



Managing Multiple Versions



Managing Multiple Versions

Every incoming query is replicated and routed to Versions A and B Each Version is implemented as specific type of SOLR ?SS query We deploy more than 2 versions to production and upda switch between them using zookeeper rec Result mixing of A and B is implemented in a processing layer above SOLR replication

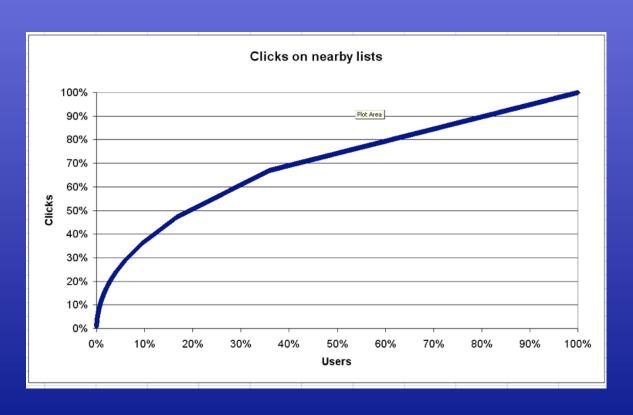
Caveat 1: Randomization

- don't confuse users with changing results, i.e.: provide a consistent user experience
- Solution:
 - Random generator is seeded with USER-ID for each query.
 - Each user gets his personal random generator.



Caveat 2: Healthy Click Data

- we are relying on the integrity of transmitted user actions
- sensitive to log contamination (unidentified QA, spam)
- user-clicks plot:



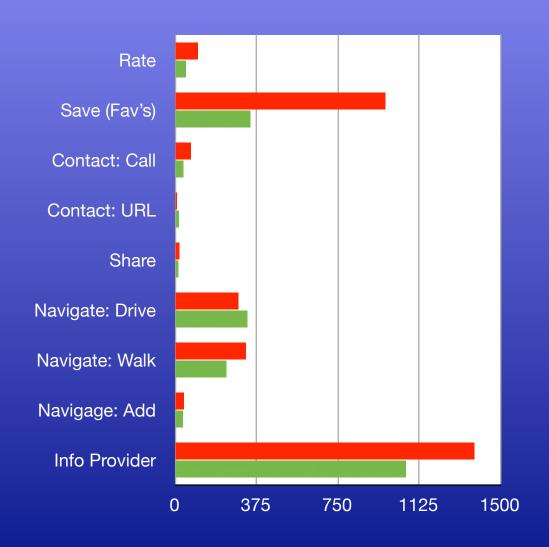


Caveat 3: A/B Clicks vs. Coverage

- Coverage = non-empty responses (in percent)
- For example
 - A/B interleaving of eat&drink vs. eat&drink + going out
 - difference is not significant
 - But coverage different, percentage of responses with POIs nearby:
 - 60% eat&drink
 - 62% eat&drink + going out
- Higher coverage decides in case there is no statistical difference

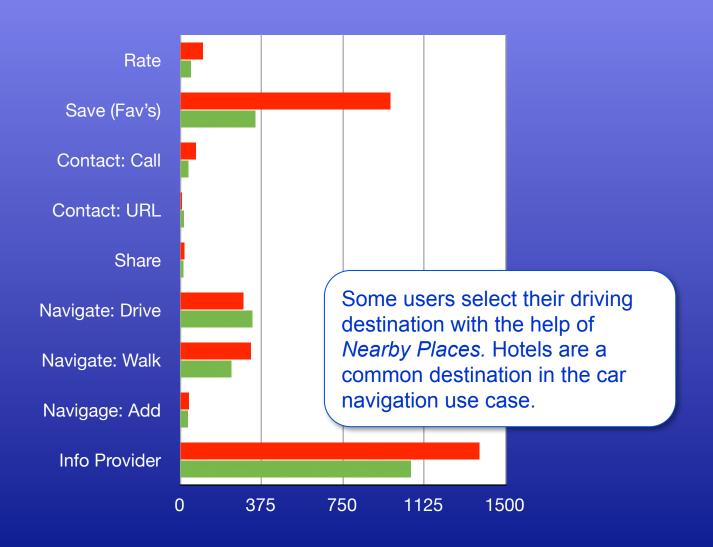


Case Study: Eat'n'Drink versus Hotels: Not the User Behaviour we had expected!





Case Study: Eat'n'Drink versus Hotels: Not the User Behaviour we had expected!





Summary

- use A/B Rank Interleaving to optimize result relevance
- Rank Interleaving is easy to implement. Works.
- in a distributed search architecture manage your A/B test configurations conveniently using Zookeeper
- harness your hadoop/search analytics stack for A/B test evaluations
- don't make assumptions about your users!



Thanks!

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