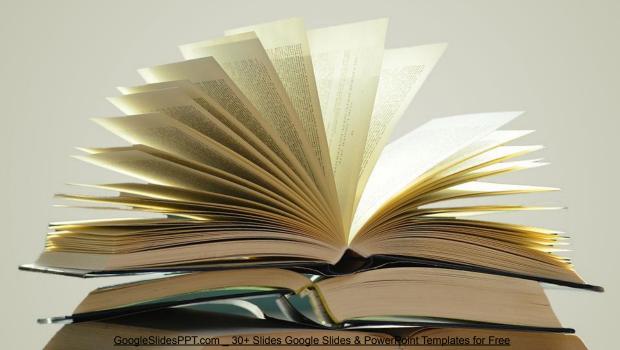
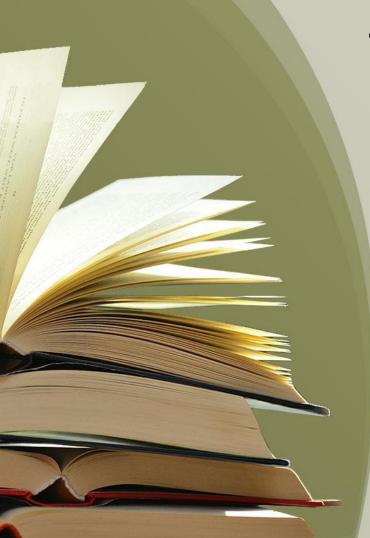
Real Relevant Reviews (R³)

Only The Most Helpful Product Reviews

W266 Section 003
Jen Darrouzet & Bradley Nott





The R³ Project



The Problem

Too many online reviews



The Data

8 Million book reviews from Amazon.com



The Approach

Helpful Vote Arrival Rates + Neural Nets



The Results

To be revealed...



The Next Steps

Real, relevant, crowd-sourced advice for buyers

R³ System















The Problem

Need an efficient method to make product reviews useful



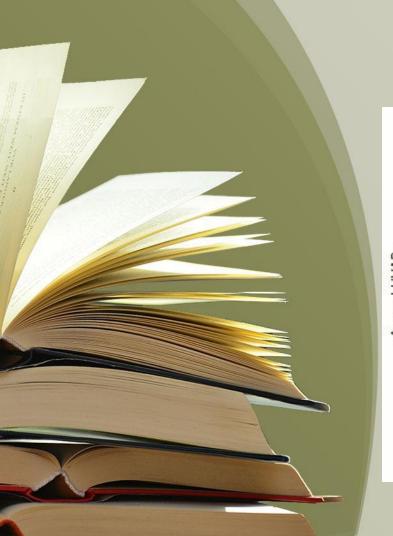
The Data

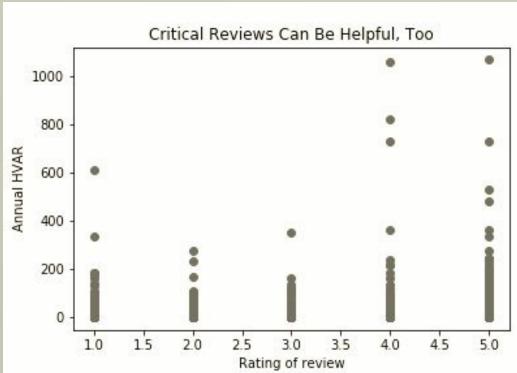
8 Million book reviews from Amazon.com (2006-2014) that contain a helpful vote tally



The Approach

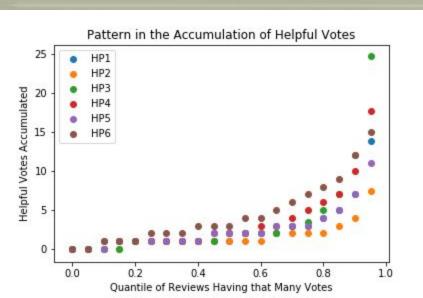
Transform the helpful vote tally into a metric that enables comparisons

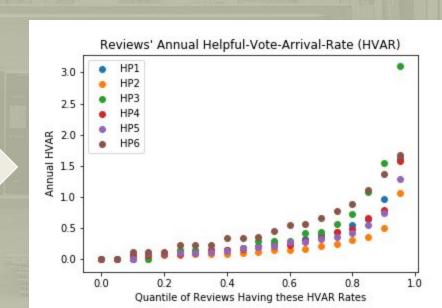




How to Train for Helpfulness

Counts? Proportions? Apples-to-apples?





Experiments

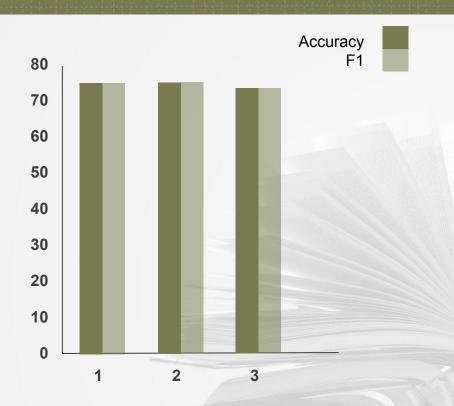
- Data is noisy and the signal is rare
- What is "means" to be helpful varies
- Difficult to be confident about signal
- Smaller models have a hard time with variations in review length



The Results

Classifier Performance

Predicting helpfulness of a review, given review text and 1-5★ rating



Modeling Notes

None of the models fit the noisy data well. Sorting by probability of helpfulness yields good top N. **Asking shoppers if we got it right could help, too.**

1 tf.idf & LogReg Super fast; unlimited vocab was

1M+

2 Bi-LSTM Insanely slow (used alt

CuDNNLSTM); 200 tokens

3 BERT Medium speed; 200 tokens; 30K

vocab; extra benefit of NSP

BERT's added benefit

If BERT is substantially same, get relevance for "free"



Query Relevance

with BERT's Next Sentence Prediction (NSP)

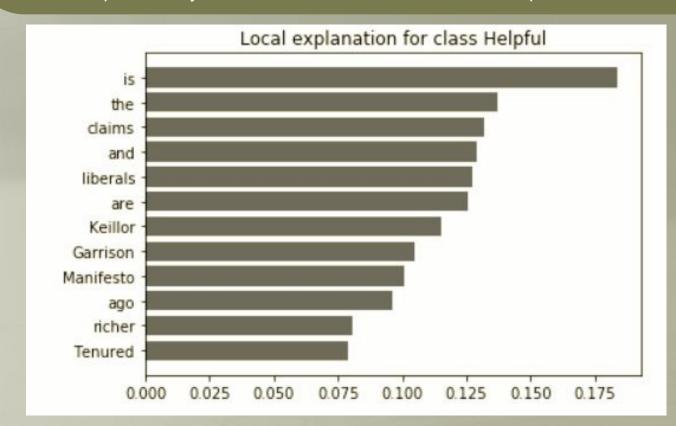
Query: Is Water for Elephants appropriate for a teenager headed off to veterinary school?

4★: ...Jacob's parents died just as he was about to write his final exams in veterinary school at Cornell University....because of his vet training, he ends up hired on to care for the horses and other animals...I read it quickly and found it very sweet and entertaining; recommended for teens and adults... Caveat: there are descriptions of mistreatment and abuse of animals in this book. If that is unpalatable to you, you should skip this book...

2★: Jacob, now an old man, begins the book by going back into his youth to the time **he was attending vet school**. ... He winds up with a circus and becomes the circus vet, although, **he rarely spent any time with or working on the animals**. **He just stood and watched animals be slaughtered and then felt bad** that he did not do anything...

Helpfulness indicator

LIME helped identify that reviews with claims were >80% helpful



Claims

Not just for nonfiction reviews, either

BEST Someone with an impressive array of murderous skills is systematically assassinating right-wing political figures one after another. The FBI and the CIA, who suspect that the killer comes from somewhere in their ranks, join forces to try to track down the assassin. An imprisoned former CIA traitor, who claims to know the identity of the assassin, wants to trade that information

Next steps

Tune R³ for identified model weaknesses (e.g. length)

- ☐ Customer tolerance for review length may be product-category dependent
 - □ **□** Book lovers embrace long-form text
 - Beauty product shoppers could prefer short text/user-submitted photos

Evaluate R³'s ranked reviews using IR metrics like Mean Average Precision (MAP)

☐ Could not automatically measure precision/recall as dataset did not have relevance labels

Optimize R³ for shopper happiness (both **useful and fast**)

- Once we can reliably deliver a TOP N HELPFUL + RELEVANT review list, SPEED is paramount
 - ☐ Can we process new reviews immediately upon submission? (Important for new releases.)
 - ☐ What is the fastest BERT model to compare most helpful reviews to a shopper query?
 - Confident we can pre-calculate a BERT representation of each incoming review such that shoppers' query-time comparisons are fast...but how fast is the question.

