How can you tell if a discussion is interesting?

Characterizing and curating conversation threads

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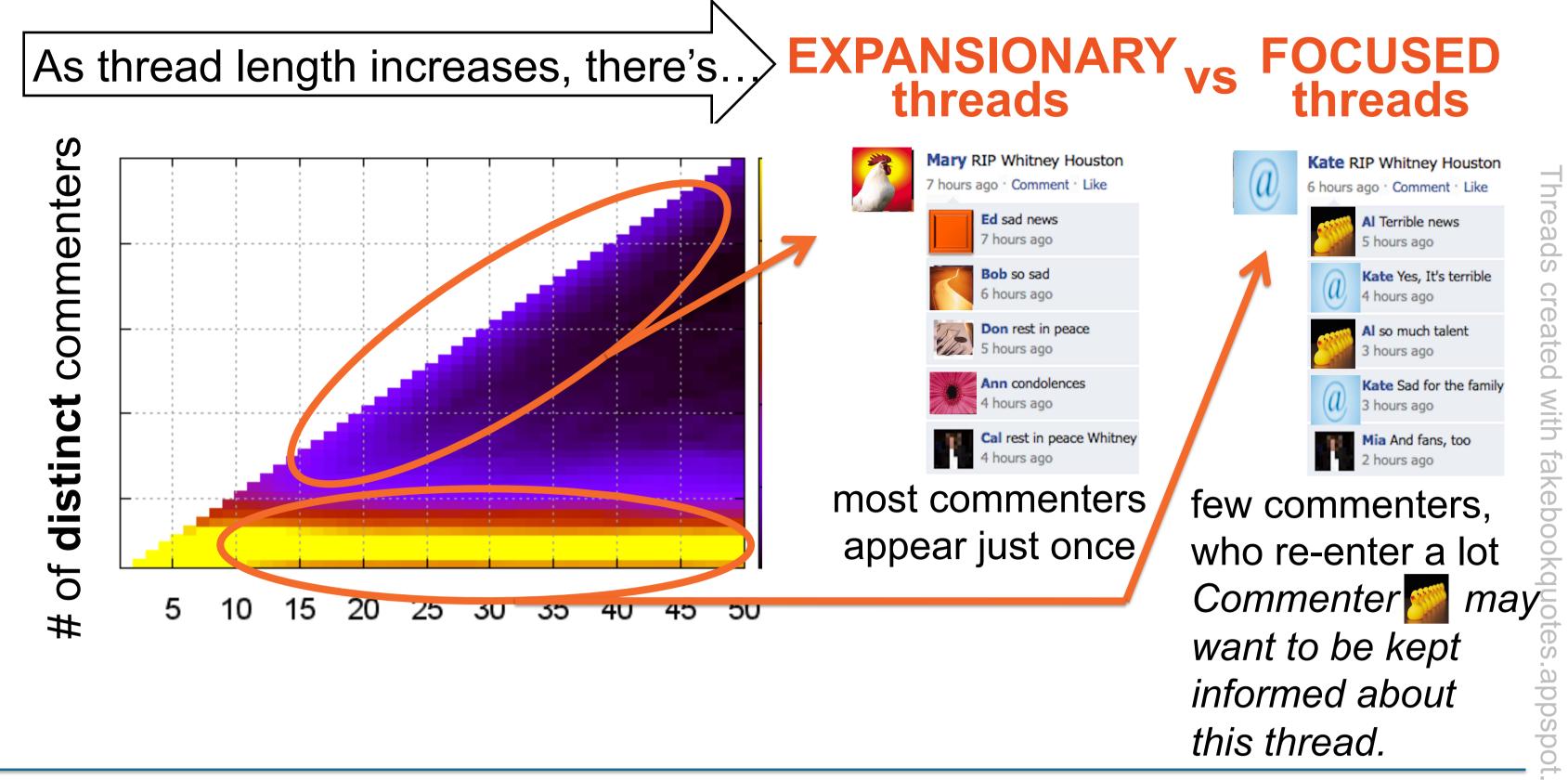
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TWO PREDICTION PROBLEMS



This heatmap reveals a key bimodality:



Aside: It inspires a new generative model.

A standard Chinese-restaurant-style model *can't* generate this bimodality.

We show by simulation that it *can* be generated by a model inspired by the theory of nonlinear urn processes.

The details:

To generate commenters c_1, c_2, \ldots for thread: start with c_0, c_1 , each with $weight\ w(c_j) = 1$. Parameters $\alpha \geq 1$ and $\beta \geq 0$.

(i) Choose commenter c_{j+1} after c_1, \ldots, c_j chosen. Choose preexisting thread participant (different from c_j) with probability proportional to the weights. Pre-existing participant $c \neq c_j$ is chosen with probability $w(c)/(\beta + \sum_{c' \neq c_j} w(c'))$, whereas a new participant c_{j+1} is introduced into the thread with probability $\beta/(\beta + \sum_{c' \neq c_j} w(c'))$.

(ii) Update weights. If participant c_{j+1} chosen in (i) is a reentrant, we define $w(c_{j+1}) \leftarrow \alpha w_j(c_{j+1})$, and leave all other weights unchanged. If c_{j+1} is new, we set $w(c_{j+1}) \leftarrow 1$ and for all other pre-existing participants $c_j \neq c_{j+1}$ we reduce

their weights via $w(c) \leftarrow w_j(c)/\alpha$.

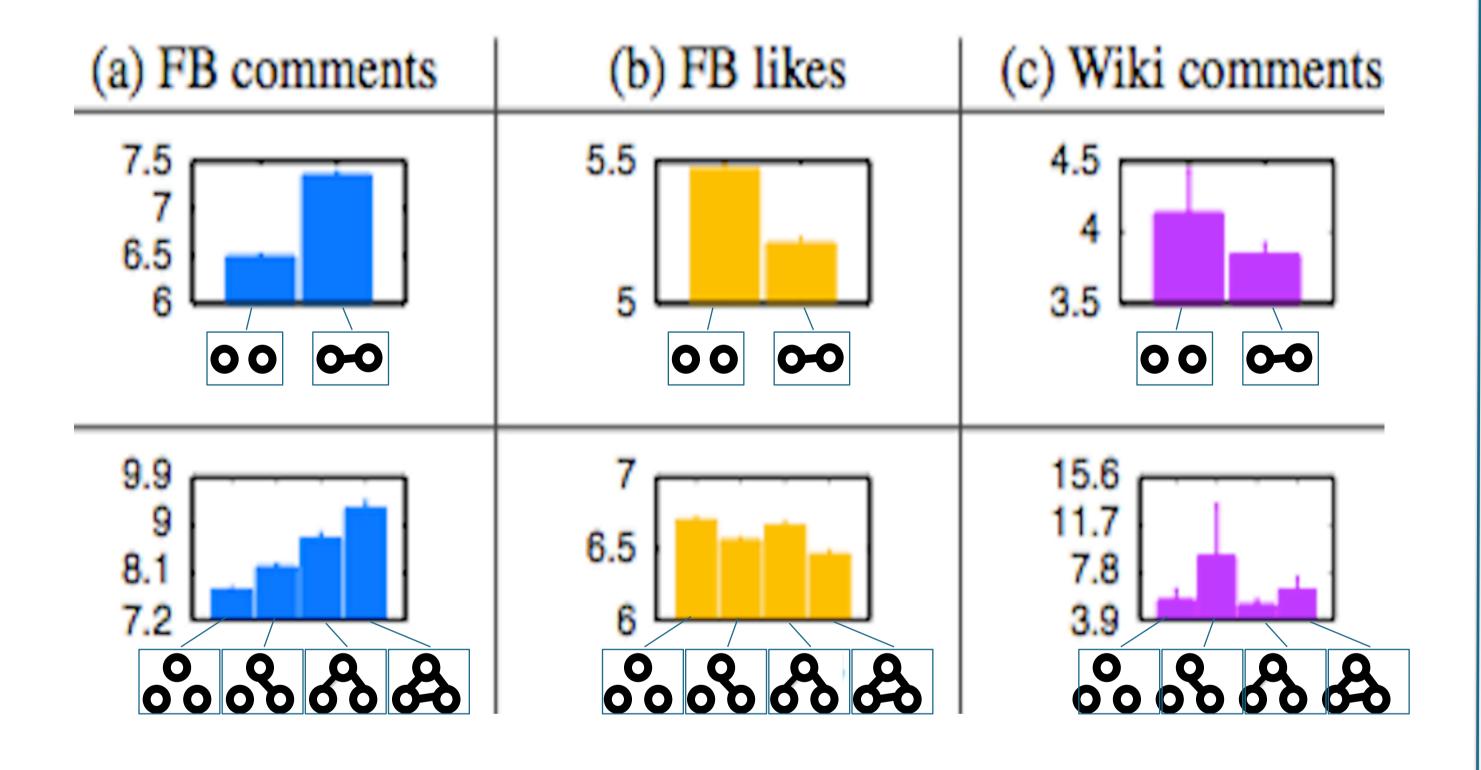
SOCIAL,

1,2,... SEQUENTIAL,

0 & TEMPORAL

features outperform textual features at predicting length and re-entry.

& Social connectedness of first participants



1,2,... Arrival sequence



Chance ID code 1 re-enters: 26.7% vs. 5.6% (For 1,0,1,0,0,... this chance is 47.5%.)

Top features for Facebook length prediction

| | Features | Area under ROC curve |
|------------------------|--|----------------------|
| eature-selection order | Text-only baseline (icon: "Tt") ("comment", "agree", "anybody", etc., selected by text regression) | .529 |
| | Temporal (time 'til 5 th comment) | .695 |
| | 1,2, +arrival sequence (# uniq in first 5) | .705 |
| | "Tt" +number of words (in 5 th comment) | .714 |
| | +temporal (time 'til 3 rd comment) | .721 |
| | "Tt" +contains "?" (in 5 th comment) | .7256 |
| ure | 1,2, +arrival seq (ID code of 5 th commenter) | .7258 |
| eat | +arrival seq (# uniq in first 4) | .7260 |

Top features for Facebook re-entry prediction:

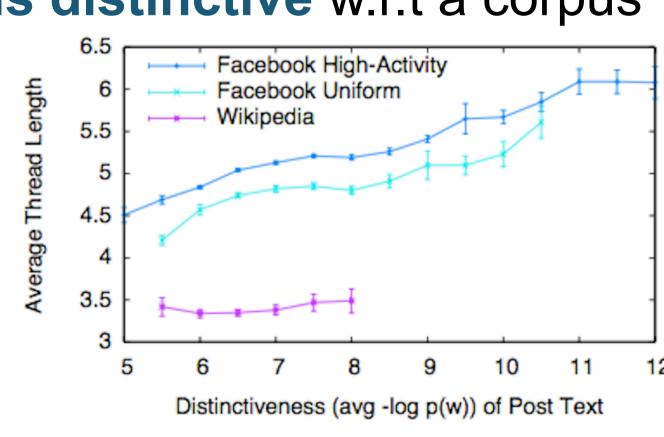
arrival sequence 1,2,...., then temporal (

Further directions: distinctiveness

Under certain conditions, thread length increases when ...

... the post's text is distinctive w.r.t a corpus

language model ...



... or the 1st commenter is distinctive w.r.t who usually comments on the poster's posts.

