



Information Seeking and Responding Networks in Physical Gatherings: A Case Study of Academic Conferences in Twitter

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Jonathan Deber @jdeber · Apr 28 3D printed teddy bears with a felt printer. They're like normal teddy bears, but CADier. And cooler. #cadierisawordright? #chi2014

★ 47 ★ ···

ITS 2014 @ITS_2014 · Apr 28

@hilfeheiko Sounds interesting. Care to comment or expound? #chi2014

• ti + ···



View conversation



Wendy Moncur @wendymoncur · Apr 28

Fitz at **#chi2014** - regular FitBit users often put their devices through the wash & have to buy new ones. Washable wearables the next step?

★ 17 2 ★ 4 ···



Compatibility versus natural interactions? How much does your interaction resemble the action you want to do. **#chi2014**

• ti * ···



Ben Kirman @benki · Apr 28

The **#chi2014** robot has a human slave following it around. Did they not learn from the robots paper last year? dl.acm.org/citation.cfm?i...



CHI2014 @chi2014 · Apr 28

@barneyc excellent question about #CHI2014 proceedings and mobile devices. Anyone out there have a clever way to solve that?

View conversation



ti 🛧

Is there a substantial difference between immersion and flow? Is flow just the max attainable level of immersion. #chi2014

...

★ 17 ★ 2 ···



Dr Simone Stumpf @DrSimoneStumpf · Apr 28 "What does this mean for HCI?" Theoretical understanding of Making culture is all fine but maybe also look to supporting practice? #chi2014

◆ tì ★ …

Tia Shelley @TiaTalksHCl · Apr 28



▲.

When moving to a mobile device, how do controls change experience of gameplay? #chi2014

◆ 好1 ★ …



Challenges

- Participants may seek and disseminate information while rushing to keep up with the event.
- Participants often find themselves exposed to new places, new people, and new topics.

Research Questions

• RQ1: How do we meaningfully capture the distinctive information needs on Twitter during academic conferences?

Research Questions

- RQ2: Can communications of the information needs be inferred from users' prior tweeting patterns and network positions?
 - Who are the users that tend to post certain types of information needs?
 - Can we identify potential responders to these information needs?

Data Collection

- 190k tweets posted in 66 CS conferences over five years (2009-2013).
- ~12k information-seeking tweets (tweets that contain at least one "?" [Paul 2011]).

Information Seeking Annotation & Labels

- We developed a category scheme (*Fleiss Kappa = 0.58*) of the information seeking tweets based on previous studies [Morris 2011 & Efron 2010].
- We obtain annotations of the type labels for ~2k information-seeking tweets through Amazon Mechanical Turk.
- We trained n-gram multi-class SVM classifiers (*accuracy = .721*) to automatically label the rest of the information-seeking corpus.

Coordinate Events/Ask Favors	3.2%	Getting set to head to LA for #siggraph2012 with a stable of #makerbot replicators. Anyone wanna join me there?
Express Opinions	19.7%	"Since we couldn't do either qual or quant research, we decided on mixed methods." huh?! #chi2012
Promote Information	32.7%	<i>Can't make it to #FAST13 this week? Consider attending via live stream instead: <u>https://t.co/</u> <u>mNyA4IGe</u></i>
Request Information	44.4%	I wish I was at #CHI2010. I wish I got to listen to Genevieve Bell. Does anyone know if I can listen to the keynote online?

RQ2A:

Who are the users that tend to post certain types of information needs?







Express Opinions

Promote Information

Request Information











Less likely to give information pointers (via posting URLs).

SIGKDD retweet network (Size of the nodes are proportional to the closeness centrality)



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Promoters

More likely to give information pointers, be mentioned, or be retweeted.

SIGKDD retweet network (Size of the nodes are proportional to the out-degrees)

Opinionators

Less likely to give information pointers, but more likely to be mentioned before.

SIGKDD retweet network (Size of the nodes are proportional to the out-degrees)

RQ2B:

Can we identify potential responders to the information needs?









Link Prediction Task



Link Prediction Task



Link Prediction Task Content-based Similarity



- User Similarity
 - Similarity between users' tweets in the past.

Link Prediction Task Content-based Similarity



- Text Similarity
 - Similarity between question and users' tweets in the past.

- Baseline
 - The time difference between the question's posted time and the potential respondent's last seen activity.
- Network Proximity
- Content Similarity
 - User Similarity, Text Similarity, LIWC Similarity
- Balanced dataset

Feature Sets	RF	ADA	Bagging	SVM
Baseline	0.73	0.774	0.77	0.669
Baseline+Network	0.824	0.774	0.834	0.689
Baseline+Content	0.784	0.783	0.81	0.688
Combined	0.843	0.783	0.859	0.714



Similarity between question and users past tweets is more important!

Summary

- We provided the first large-scale empirical study on the information seeking and responding networks in physical gatherings, using Twitter in Academic conferences as a case study.
- We profiled the four prototypical information seekers in the conferences.
 - Opinionators, Coordinators, Promoters, Requesters.
- With a set of similarity measures, we are able to predict the replying action to the questions with high AUC.

Potential Applications & Implications

- For the Individuals:
 - We can route the less noted questions to the probable respondents.
- For the communities:
 - Event organizers could consider constructing question leader list, so that people with various interests could selectively join the discussions more quickly.
 - Information Seeking and Responding dynamics relates to the sustainability of the community.

Thank you! Any Questions?