Class slides for Thursday, October 8: Social fads, part 2

Matthew J. Salganik

COS 597E/SOC 555 Limits to prediction Fall 2020, Princeton University

Can Cascades be Predicted?

Justin Cheng Stanford University jcccf@cs.stanford.edu Lada A. Adamic Facebook ladamic@fb.com P. Alex Dow Facebook adow@fb.com

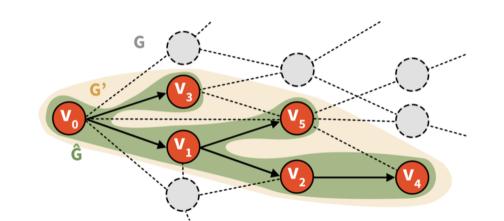
Jon Kleinberg Cornell University kleinber@cs.cornell.edu Jure Leskovec Stanford University jure@cs.stanford.edu ▶ cascade growth prediction problem: given a cascade that currently has size k, will it grow beyond the median size f(k)?

cascade growth prediction problem: given a cascade that currently has size k, will

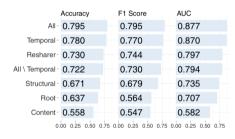
it grow beyond the median size f(k)? In this setting, this is equivalent in this setting to: given that a cascade has size k will it grow beyond 2k?

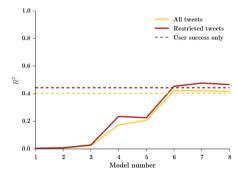
- ▶ cascade growth prediction problem: given a cascade that currently has size k, will it grow beyond the median size f(k)? In this setting, this is equivalent in this setting to: given that a cascade has size k will it grow beyond 2k?
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 prediction problems that are balanced
 - creates one prediction problem for each *k*
 - it closely approximates real tasked needing to be solved for managing viral content

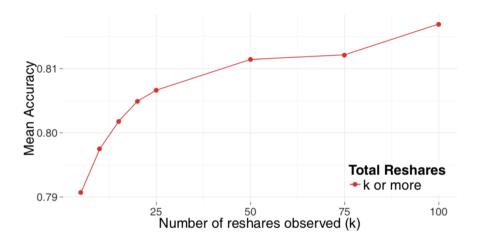


	Accuracy	F1 Score	AUC
All-	0.795	0.795	0.877
Temporal-	0.780	0.770	0.870
Resharer-	0.730	0.744	0.797
All \ Temporal -	0.722	0.730	0.794
Structural-	0.671	0.679	0.735
Root-	0.637	0.564	0.707
Content-	0.558	0.547	0.582
0.00 0.25 0.50 0.75 0.00 0.25 0.50 0.75 0.00 0.25 0.50 0.75			

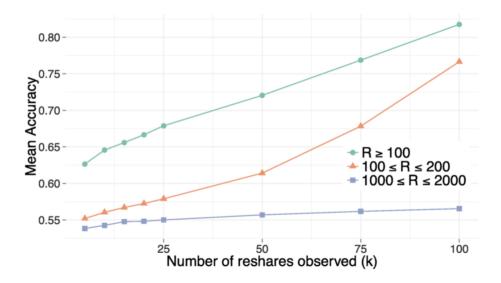




One (set) of pretty basic features does about as well as lots of features. Does this imply that we are close to a limit?



How could we apply this same idea to other domains?



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Can Cascades be Predicted?

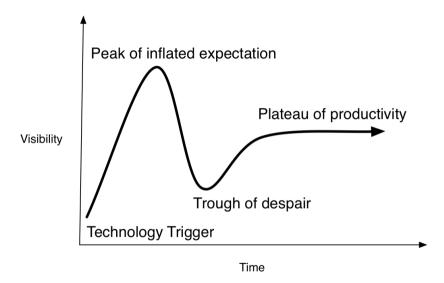
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Predicting consumer behavior with Web search

Sharad Goel¹, Jake M. Hofman¹, Sébastien Lahaie¹, David M. Pennock¹, and Duncan J. Watts¹

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Detecting influenza epidemics using search engine query data

Jeremy Ginsberg¹, Matthew H. Mohebbi¹, Rajan S. Patel¹, Lynnette Brammer², Mark S. Smolinski¹ & Larry Brilliant¹

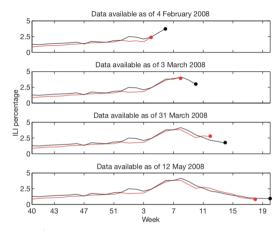


Figure 3 | ILI percentages estimated by our model (black) and provided by the CDC (red) in the mid-Atlantic region, showing data available at four points in the 2007-2008 influenza season. During week 5 we detected a sharply increasing ILI percentage in the mid-Atlantic region; similarly, on 3 March our model indicated that the peak ILI percentage had been reached during week 8, with sharp declines in weeks 9 and 10. Both results were later confirmed by CDC ILI data.

Predicting the Present with Google Trends

HYUNYOUNG CHOI and HAL VARIAN

Google, Inc., California, USA

The Parable of Google Flu: Traps in Big Data Analysis

David Lazer, 1,2* Ryan Kennedy, 1,3,4 Gary King, 3 Alessandro Vespignani 3,5,6

Google

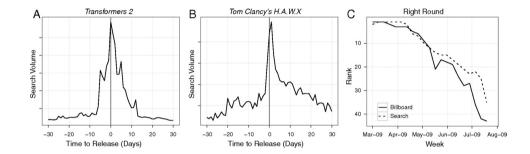
Thank you for stopping by.

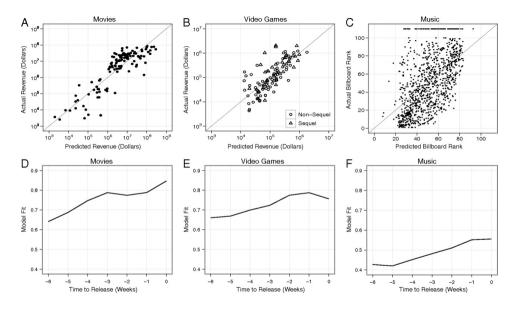
Google Flu Trends and Google Dengue Trends are no longer publishing current estimates of Flu and Dengue fever based on search patterns. The historic estimates produced by Google Flu Trends and Google Dengue Trends are available below. It is still early days for nowcasting and similar tools for understanding the spread of diseases like flu and dengue – we're excited to see what comes next. Academic research groups interested in working with us should fill out this form.

Sincerely,

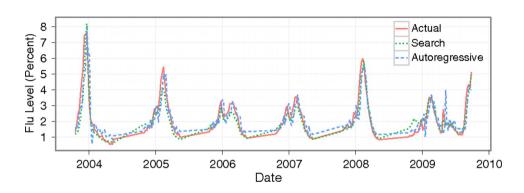
The Google Flu and Dengue Trends Team.

Nice history: https://www.theatlantic.com/technology/archive/2014/03/in-defense-of-google-flu-trends/359688/



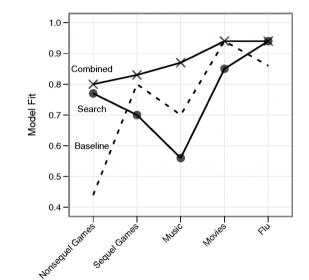


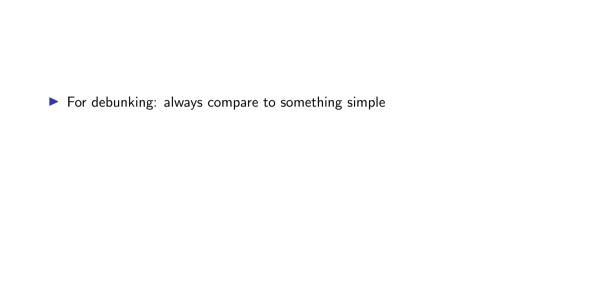
Note that measure of predictive performance is correlation. (same as Ginsburg)



correlation between prediction and truth

- ▶ simple auto-regressive model: 0.86
- ► Google Flu Trends: 0.94





- ► For debunking: always compare to something simple
- Predicting the present is not the same as predicting the future

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