

Big Data is Not About the Data!

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- **In each: without new analytics, the data are useless**

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Fast Company Names Crimson Hexagon Number Seven on "The 10 Most Innovative Companies in Web" List Leading Social Intelligence Firm Recognized For Revolutionary Measurement of Consumer Opinions in Social Media

Published: Wednesday, 16 Mar 2011 | 9:20 AM ET Text Size
CAMBRIDGE, Mass., Mar 16, 2011 (BUSINESS WIRE) -- Fast Company named

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2. Worldwide cause-of-death estimates for



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Bias in U.S. Social Security Administration Forecasts

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- **New customized analytics we developed:**
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 - Many other applications to different types of forecasts

Thresher: Finding Those Hiding in Plain Sight

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 - (Lots of technology, but it's behind the scenes)

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- How common is it? **27% of all Senatorial press releases!**

Modern Analytics to Improve Student Learning

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 - Intrinsic motivation: collaborative annotation in threads
 - Extrinsic motivation: automated grading of annotations & engagement (better than instructors can do on their own)
 - Novel data analytics: keep students on track, with automated personal guidance, nudges, nonadversarial grading
 - Instructors save time, stay engaged: automated student confusion reports
 - Want to try it here? see Perusall.com

Reverse Engineering Censorship in China

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Reverse Engineering China's 50c Party

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For more information

GaryKing.org

Institute for Quantitative Social Science
Harvard University