

# Big Data is Not About the Data!

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Abt Associates 9/28/2017

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World Health Organization



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

CAMBRIDGE, Mass., Mar 16, 2011 (BUSINESS WIRE) -- Fast Company named



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  - Many other applications to different types of forecasts

# Thresher.io: Finding Those Hiding in Plain Sight

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自由



# Thresher.io: Finding Those Hiding in Plain Sight

Example Substitution 1:

自由          “Freedom”

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**CENSORED**

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Coding documents into categories

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

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# Reverse Engineering Censorship in China

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  - (Technically correct, & politically much easier)

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For more information

[GaryKing.org](http://GaryKing.org)

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