

Big Data is Not About the Data!

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

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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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 - Other applications to insurance industry, public health, etc.

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自由

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自由 “Freedom”

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自由

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CENSORED

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自由

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自由
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“Eye field”

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Following Conversations that Hide in Plain Sight

Example Substitution 1:

自由
目田

“Freedom”

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“Eye field” (nonsensical)

Following Conversations that Hide in Plain Sight

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自由
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和谐

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自由
目田

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For more information

GaryKing.org