Big Data is Not About the Data!

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(Talk at "Text and Social Analytics Summit,", 6/6/2013)

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9. $>$ 90% of all data ever created was created last year

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

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crimson hexagon ⇝ 

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Published: Wednesday, 16 Mar 2011 | 9:20 AM ET Text Size

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    - More data isn't helpful! Novel analytics needed.

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

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   - Results:
     - Uncensored: criticism of the government
     - Censored: attempts at collective action

For more information

GaryKing.org