

Big Data is Not About the Data!

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 - **Innovative analytics:** enormously better than off-the-shelf

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

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 - (Technically correct, & politically much easier)

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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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 - 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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For more information

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