The Next Big [Social Science] Thing

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Big Data

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  ...with examples from my research

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- Unrelated substantive problems, same analytics solution:

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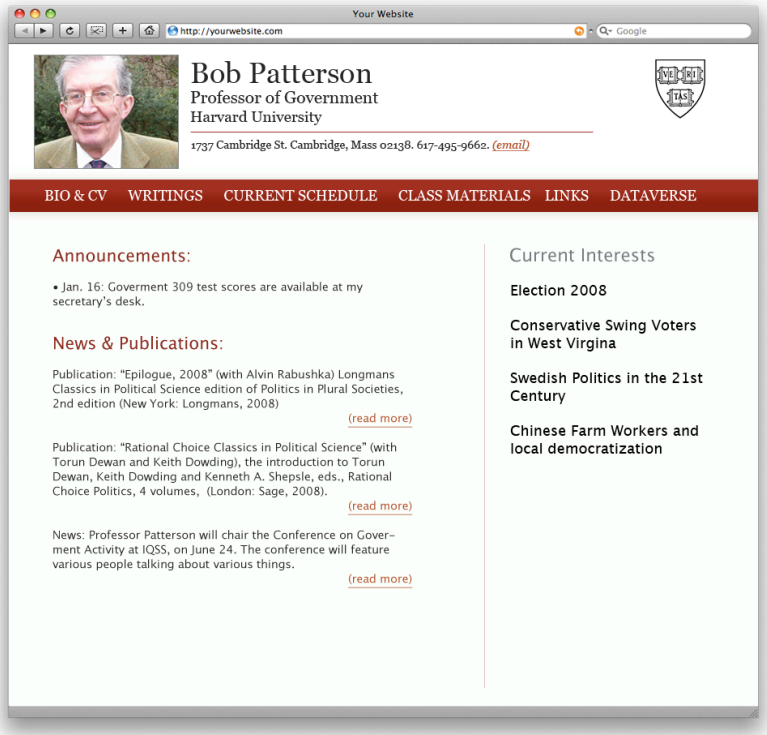
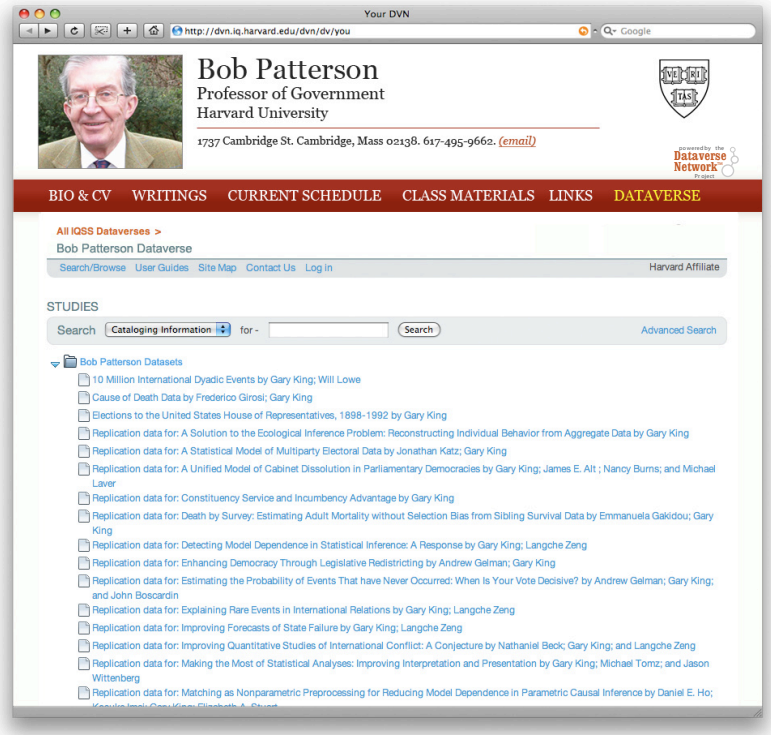
Author Dataverse





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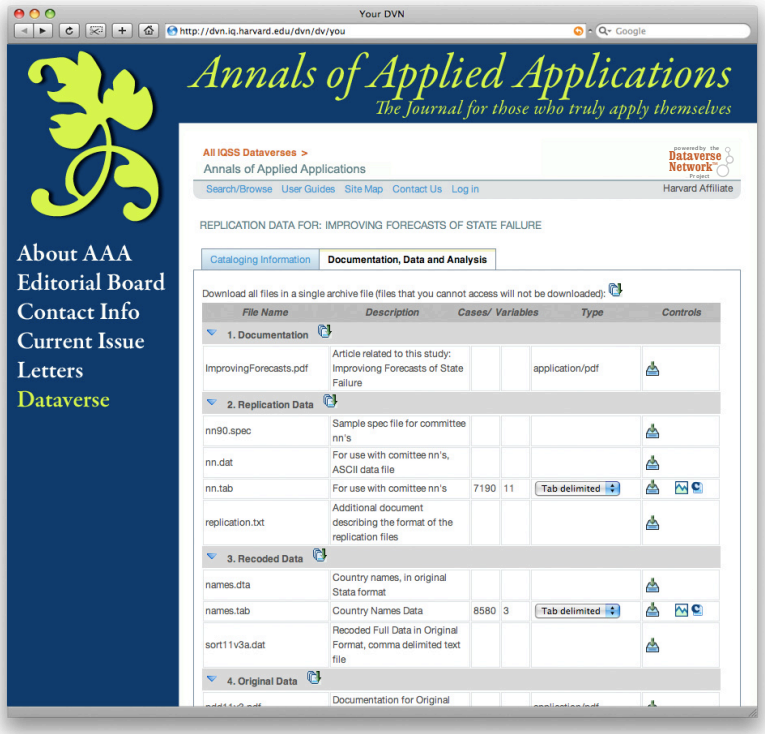
Journal Dataverse



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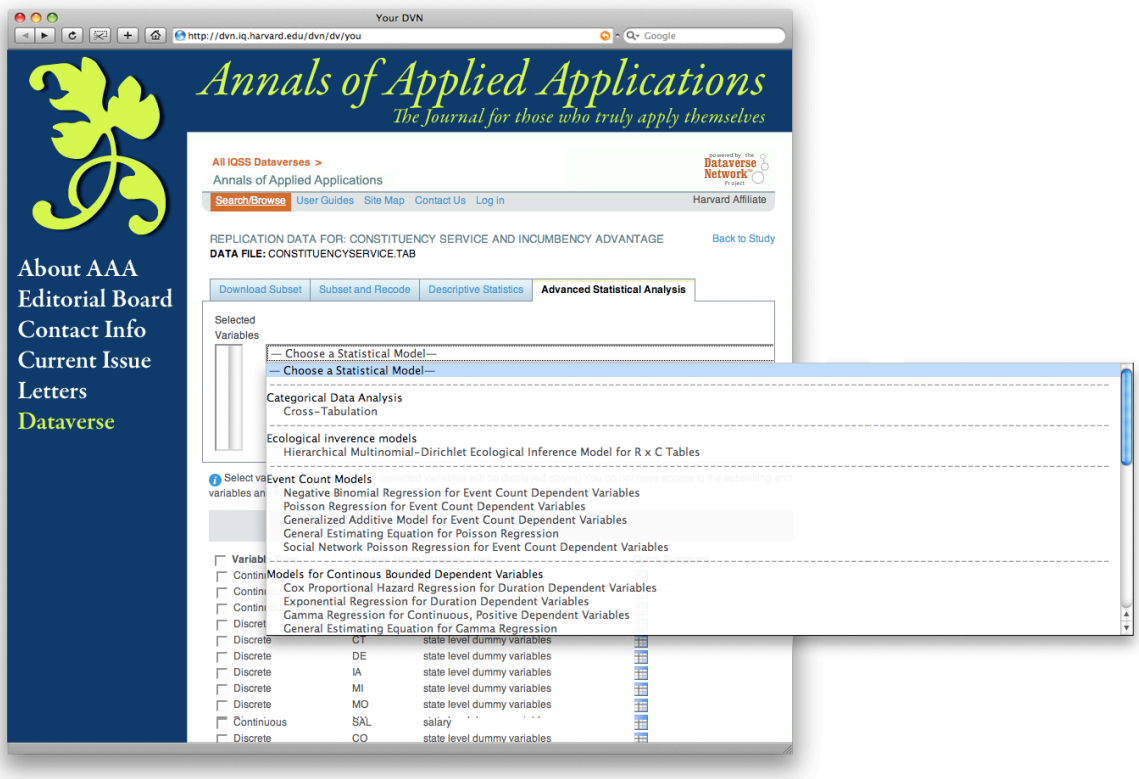

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For more information

GaryKing.org

Institute for Quantitative Social Science
Harvard University