

How the News Media Activate Public Expression and Influence National Agendas¹

Gary King²

Institute for Quantitative Social Science
Harvard University

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¹Based on joint work with Benjamin Schneer and Ariel White (*Science* 2017)

²GaryKing.org

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 - **Outcome variable:** individual knowledge and opinion
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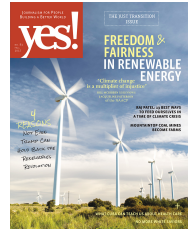
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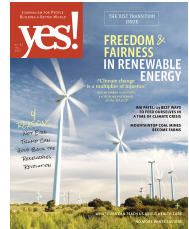
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Randomization

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Matched Pair Randomization

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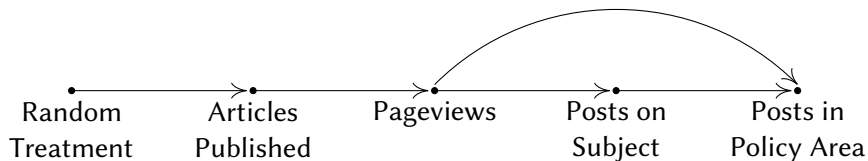
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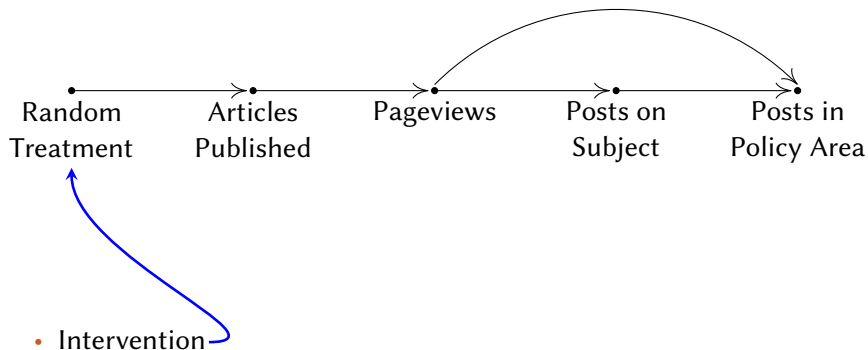
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- **(Ex post:** Automated text analysis & qualitative evidence: indistinguishable from normal publications & practices; no outlet received a single complaint)

Quantities of Interest (& observable implications)

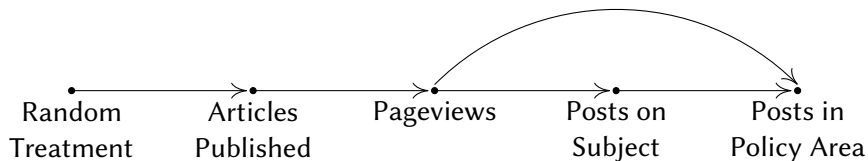
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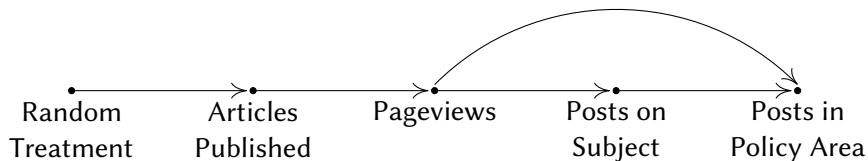
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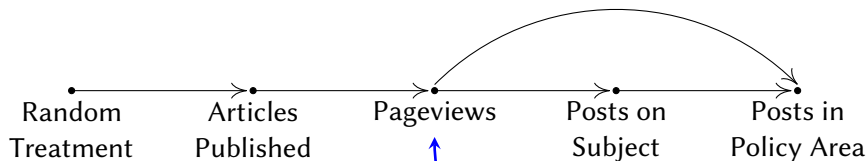
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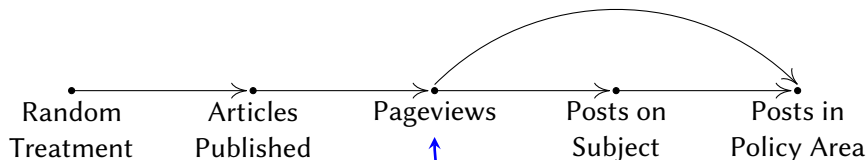
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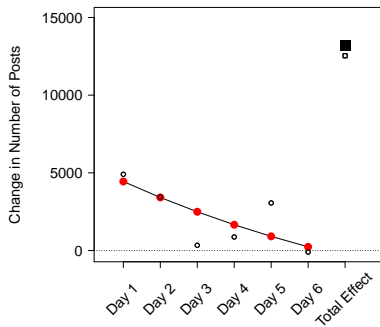
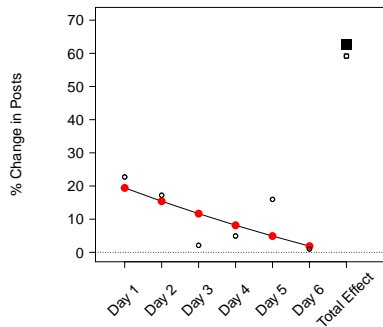
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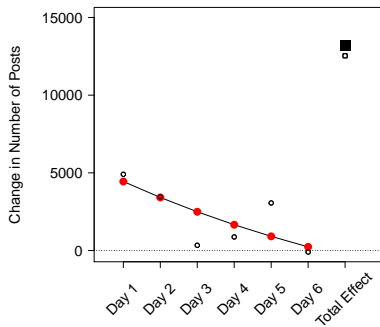
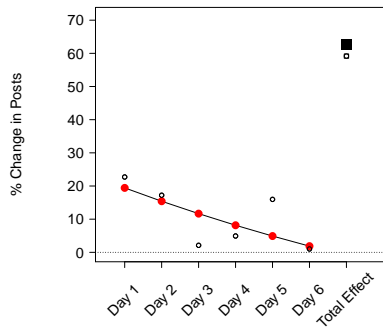
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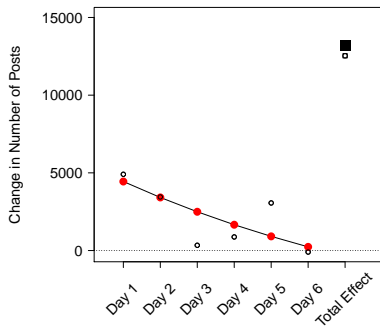
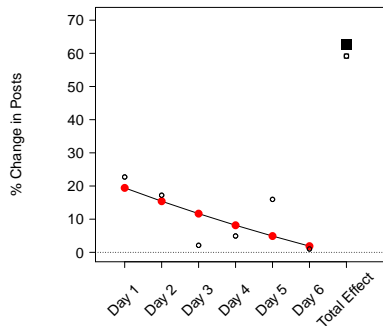


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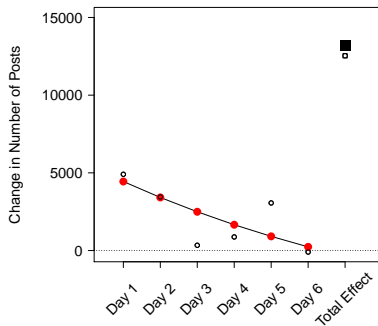
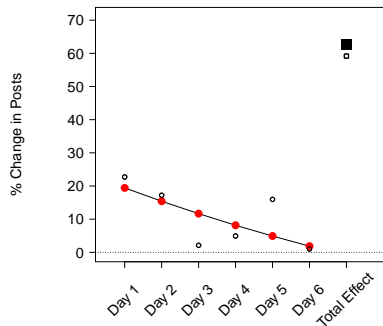
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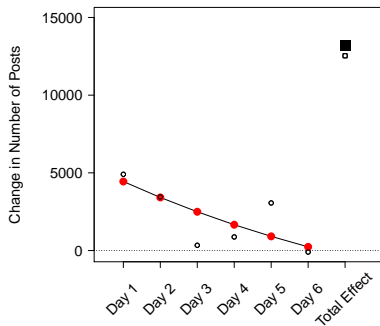
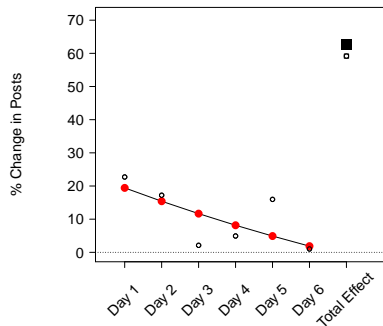
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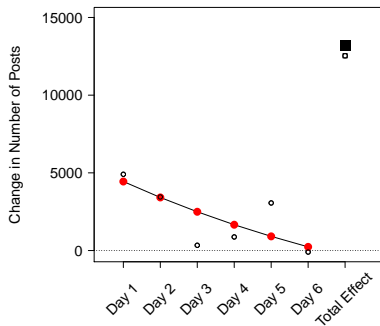
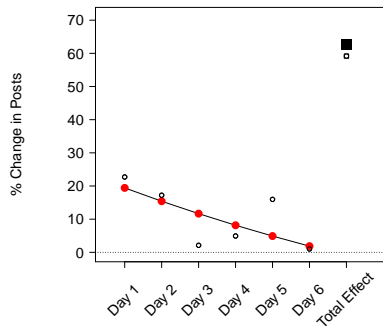
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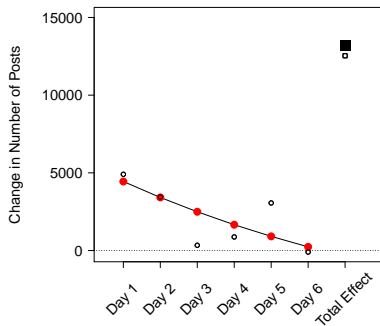
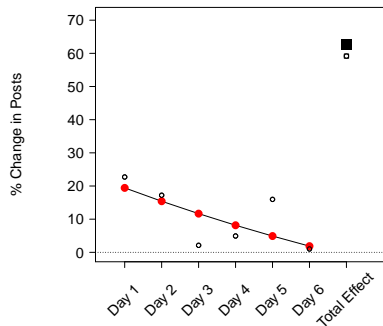
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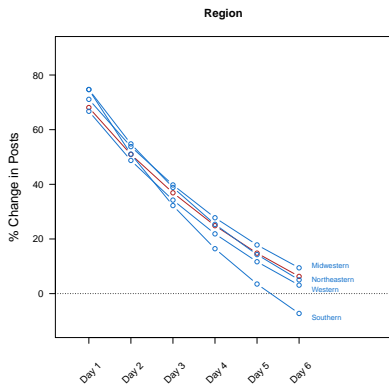
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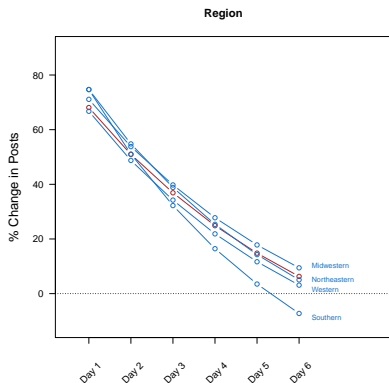
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Causal Effect: Indistinguishable Across Subgroups

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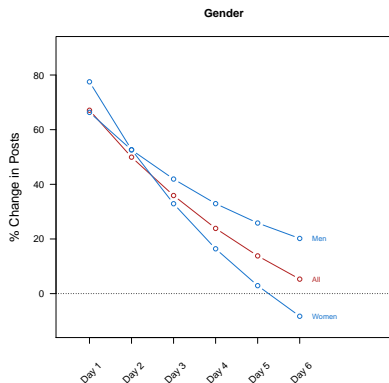


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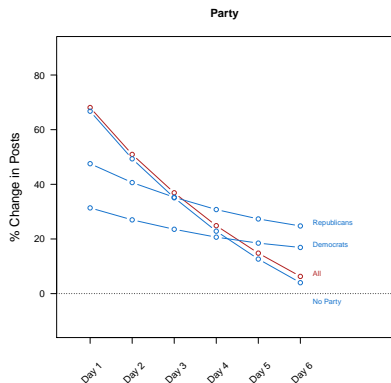
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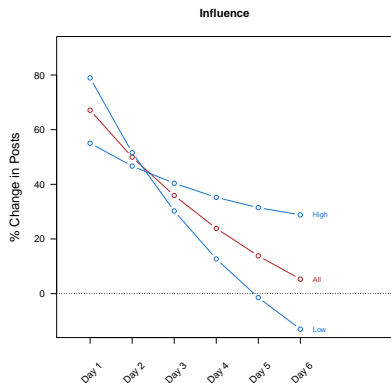
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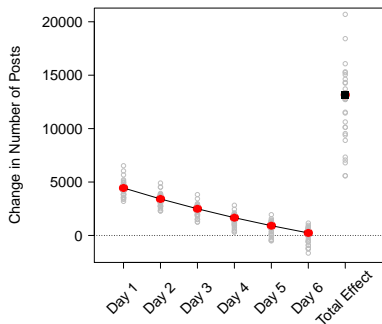


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Causal Heterogeneity: Leave-One-Outlet-Out

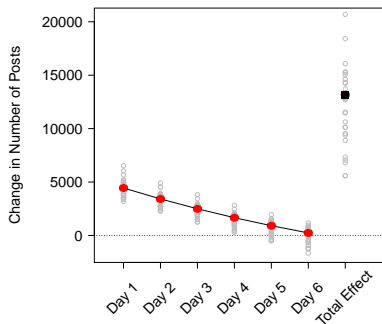
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Jackknife Estimation on Policy Area Effects



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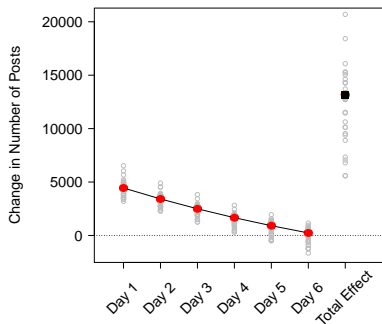
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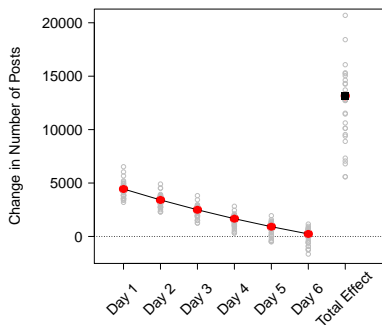
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- **Results:** no dominant outlet; high heterogeneity

Introduction

Research Design

Results

Supporting Analyses

Implications

High Experimental Compliance

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- # Articles published by pack in policy area

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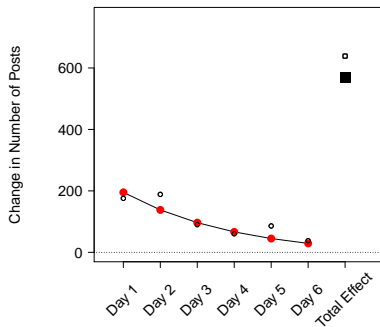
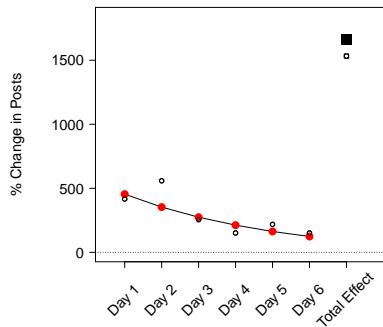
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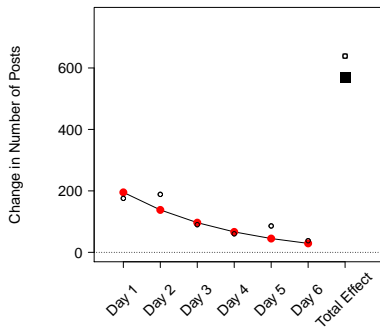
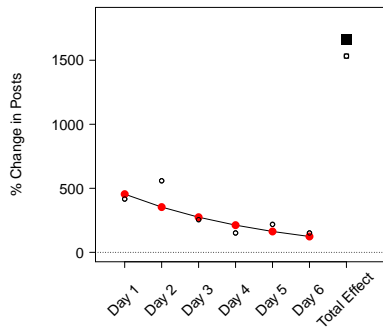
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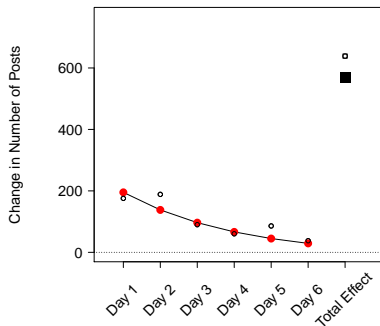
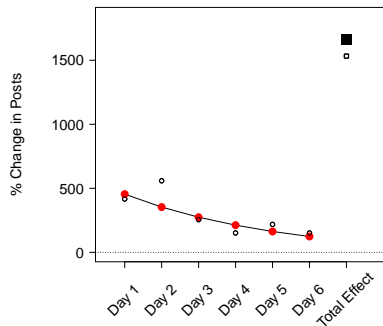


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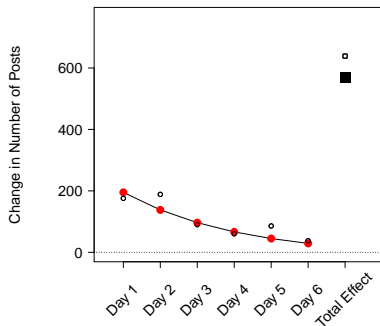
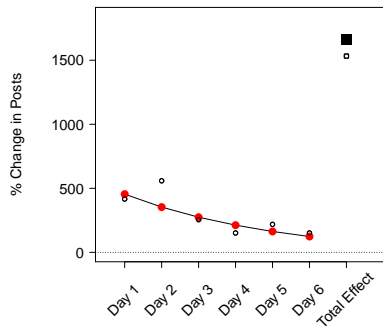
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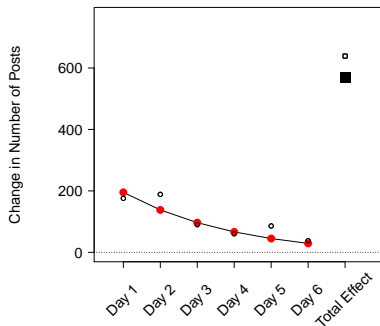
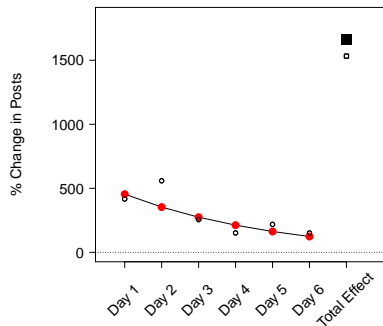
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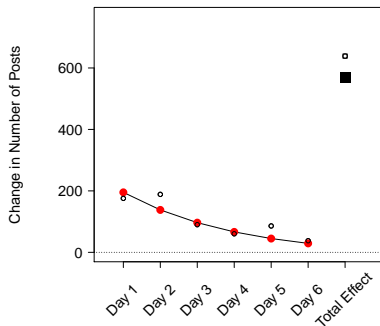
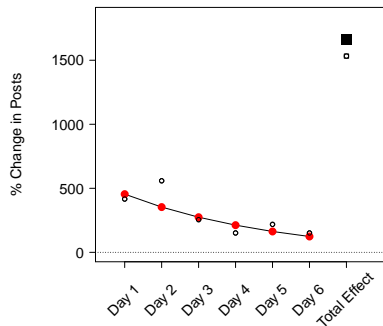
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Other Supporting Analyses

Other Supporting Analyses

- More Results

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 - Opinion change: 2.3% change in direction of article opinion

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For more information:
GaryKing.org/media

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