

What to do about Biases in Survey Research

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- Gary King and Jonathan Wand. “Comparing Incomparable Survey Responses: Evaluating and Selecting Anchoring Vignettes,” *Political Analysis*, 15, 1 (Winter, 2007): 46–66.

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- Gary King; Christopher J.L. Murray; Joshua A. Salomon; and Ajay Tandon. “Enhancing the Validity and Cross-cultural Comparability of Measurement in Survey Research,” *American Political Science Review*, Vol. 98, No. 1 (February, 2004): 191–207.

- Gary King and Jonathan Wand. “Comparing Incomparable Survey Responses: Evaluating and Selecting Anchoring Vignettes,” *Political Analysis*, 15, 1 (Winter, 2007): 46–66.
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- Papers, FAQ, examples, software, conferences, videos:
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Examples of the Problem

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Presidential Approval: the longest public opinion time series

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- On 9/10/2001, **55%** of Americans approved of the way George W. Bush was “handling his job as president”.
- The next day — which the president spent in hiding — **90%** approved.
- Was this massive opinion change, or was the same question interpreted differently?

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- The facts of the case seemed clear
- Did he do it? Whites: 62% say “yes”. Blacks: 14% say “yes”.
- Did black and white Americans have genuinely opposing views about whether Simpson committed murder, or did the two groups interpret the same survey question differently?

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- Is the young woman less healthy or are the two interpreting the same question differently?
- In some countries, responses to this survey question correlate negatively with objective measures of health status (Sen, 2002).

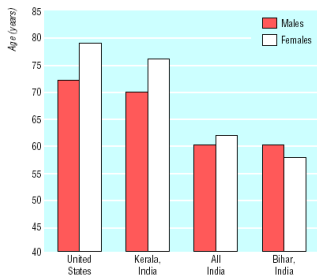


Fig 1 Life expectancy among males and females in India compared with United States, mid-1990s^{2, 3}

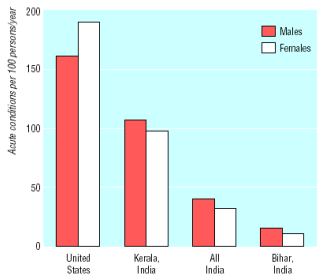


Fig 2 Incidence of reported morbidity in India, mid-1970s, compared with United States, mid-1980s^{6, 9}

Anchoring Vignettes & Self-Assessments: Political Efficacy (about voting)

How much say [does 'name' / do you] have in getting the government to address issues that interest [him / her / you]?

(a) Unlimited say, (b) A lot of say, (c) Some say, (d) Little say, (e) No say at all

Anchoring Vignettes & Self-Assessments: Political Efficacy (about voting)

- “[Alison] lacks clean drinking water. She and her neighbors are supporting an opposition candidate in the forthcoming elections that has promised to address the issue. It appears that so many people in her area feel the same way that the opposition candidate will defeat the incumbent representative.”

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- “[Jane] lacks clean drinking water because the government is pursuing an industrial development plan. In the campaign for an upcoming election, an opposition party has promised to address the issue, but she feels it would be futile to vote for the opposition since the government is certain to win.”

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- “[Jane] lacks clean drinking water because the government is pursuing an industrial development plan. In the campaign for an upcoming election, an opposition party has promised to address the issue, but she feels it would be futile to vote for the opposition since the government is certain to win.”
- “[Moses] lacks clean drinking water. He would like to change this, but he can’t vote, and feels that no one in the government cares about this issue. So he suffers in silence, hoping something will be done in the future.”

How much say [does 'name' / do you] have in getting the government to address issues that interest [him / her / you]?

(a) Unlimited say, (b) A lot of say, (c) Some say, (d) Little say, (e) No say at all

Does R_1 or R_2 have More Political Efficacy?



- The only reason for vignette assessments to change over respondents is DIF
- Assumption holds because investigator creates the anchors (Alison, Jane, Moses)
- Our simple (nonparametric) method works this way.

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- (Our parametric method doesn't)

Comparing China and Mexico

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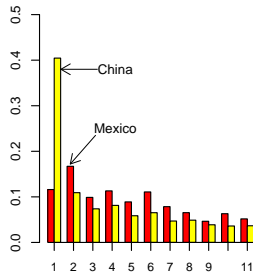
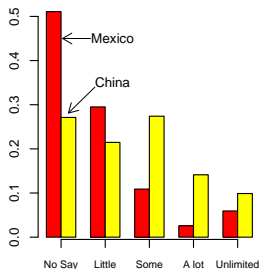
Opposition leader Vicente Fox elected President.
71-year rule of PRI party ends.
Peaceful transition of power begins.

Plenty of political efficacy

China: How much say do you have in getting the government to address issues that interest you?



Nonparametric Estimates of Political Efficacy



- The left graph is a histogram of the observed categorical self-assessments.
- The right graph is a histogram of C , our nonparametric DIF-corrected estimate of the same distribution.

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 - (c) The scale being tapped is perceived as unidimensional.
3. In other words: we allow response-category DIF but assume stem question equivalence.

Ties and Inconsistencies Produce *Ranges*

	Survey	1:	2:	3:	4:	5:	
Example	Responses	$y < z_1$	$y = z_1$	$z_1 < y < z_2$	$y = z_2$	$y > z_2$	C

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Analyzing the DIF-Free Variable: More Efficiencies

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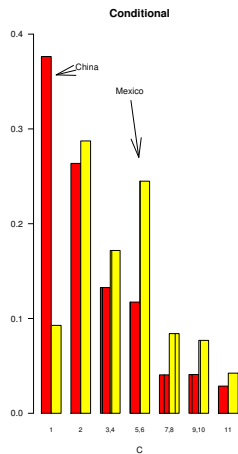
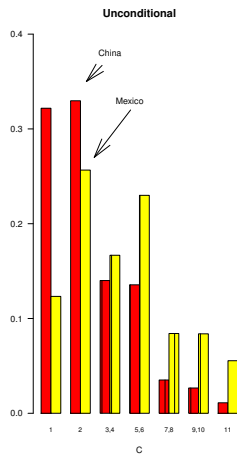
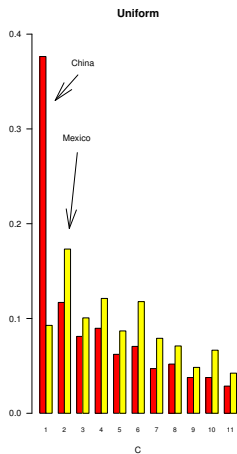
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- We define a new method (censored ordered probit), a direct extension of ordinal probit allowing for ranges of responses
- Useful for vignettes; also useful for survey questions that allow ranges of responses

Improved Efficiency in Practice



Optimally Choosing Vignettes

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- **Ultimate Goal:** Define categories with vignettes to learn about a continuous unobserved variable (health, efficacy).

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- Only question remaining: How do we calculate entropy when C is not a scalar?

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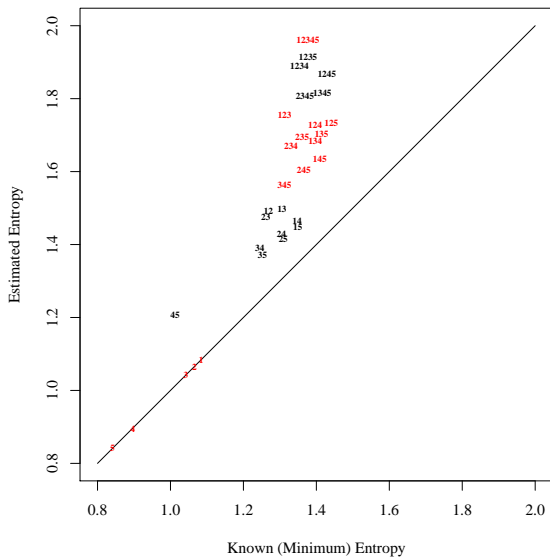
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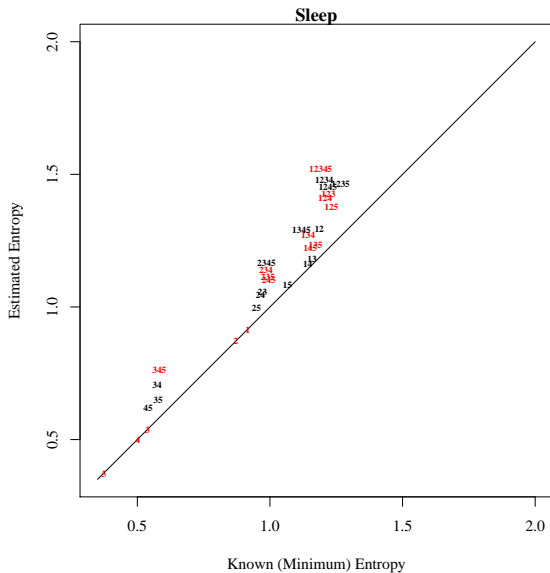
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- Result is easy to use: one measure indicating information in survey question and vignettes

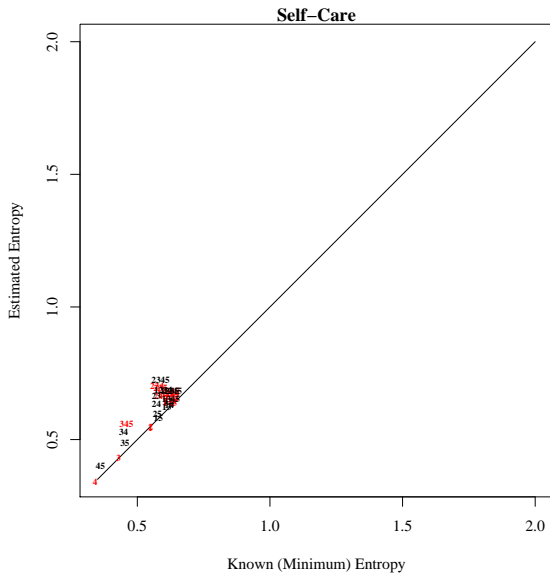
Political Efficacy (Mex & China)



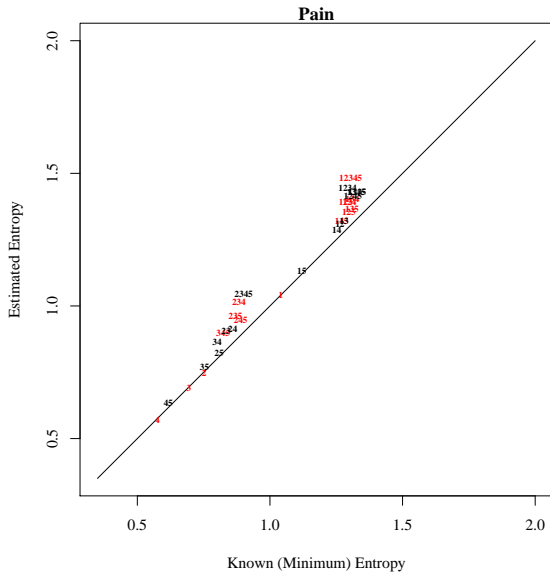
One vignette can be better than three: Sleep (China)



Some vignette sets are uninformative: Self-Care (China)

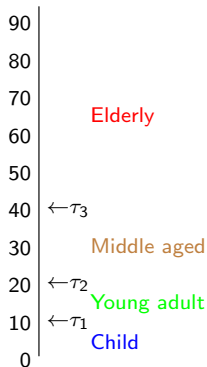


Some covariates are unhelpful: Pain (China)

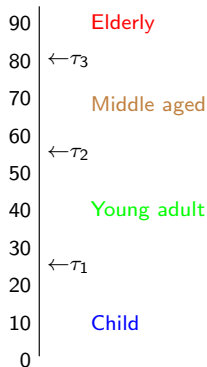


Categorizing Years of Age

Respondent 1



Respondent 2



- If thresholds vary, categorical answers are meaningless.
- Our parametric model works by estimating the thresholds.
- Vignettes provide identifying information for the τ 's.

Self-Assessments v. Medical Tests

Self-Assessment:

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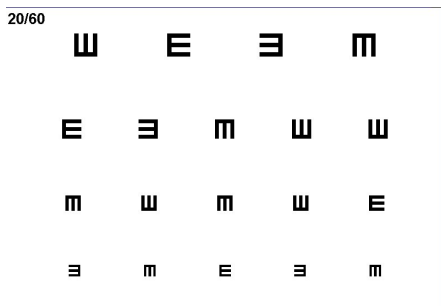
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The Snellen Eye Chart Test:



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- Chopit reproduces the same result as the medical test (though on different scale)

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<http://GKing.Harvard.edu/vign>

Includes:

- Academic papers
- Anchoring vignette examples by researchers in many fields,
- Frequently asked questions,
- Videos
- Conferences
- Statistical software