

Decision-Making During Violence

How do ordinary people make strategic decisions when facing violence?

Observational study suggests perceptions of *control* and *uncertainty* shape preferences for approach/avoid and disruptive/moderate strategies of survival [7].

		Orientation to Threat	
		Avoid	Approach
Disruptiveness	Extreme	Flee	Fight
	Moderate	Hide	Adapt

The way people perceive violence affects propensity to flee, fight, adapt to a violent environment, or hide from danger.

Why would this be true?

- Control appraisals associated with approach/avoid behavior in many settings outside violence [4, 2]
- “Unexpected” uncertainty is associated with larger behavior deviations in psych and neuroscience research [8, 6]

I manipulate perceptions about hypothetical violence in a lab-in-the-field experiment. Changing perceptions changes strategy preferences.

(Primary) Hypotheses

H1: Higher perceived *control* increases likelihood of choosing “approach” strategies (i.e. fighting, adaptation).

H2: Higher perceived *uncertainty* increases likelihood of choosing “disruptive” strategies (i.e. fighting, fleeing).

Pre-registered at: <https://osf.io/rehp3>

Study Sample

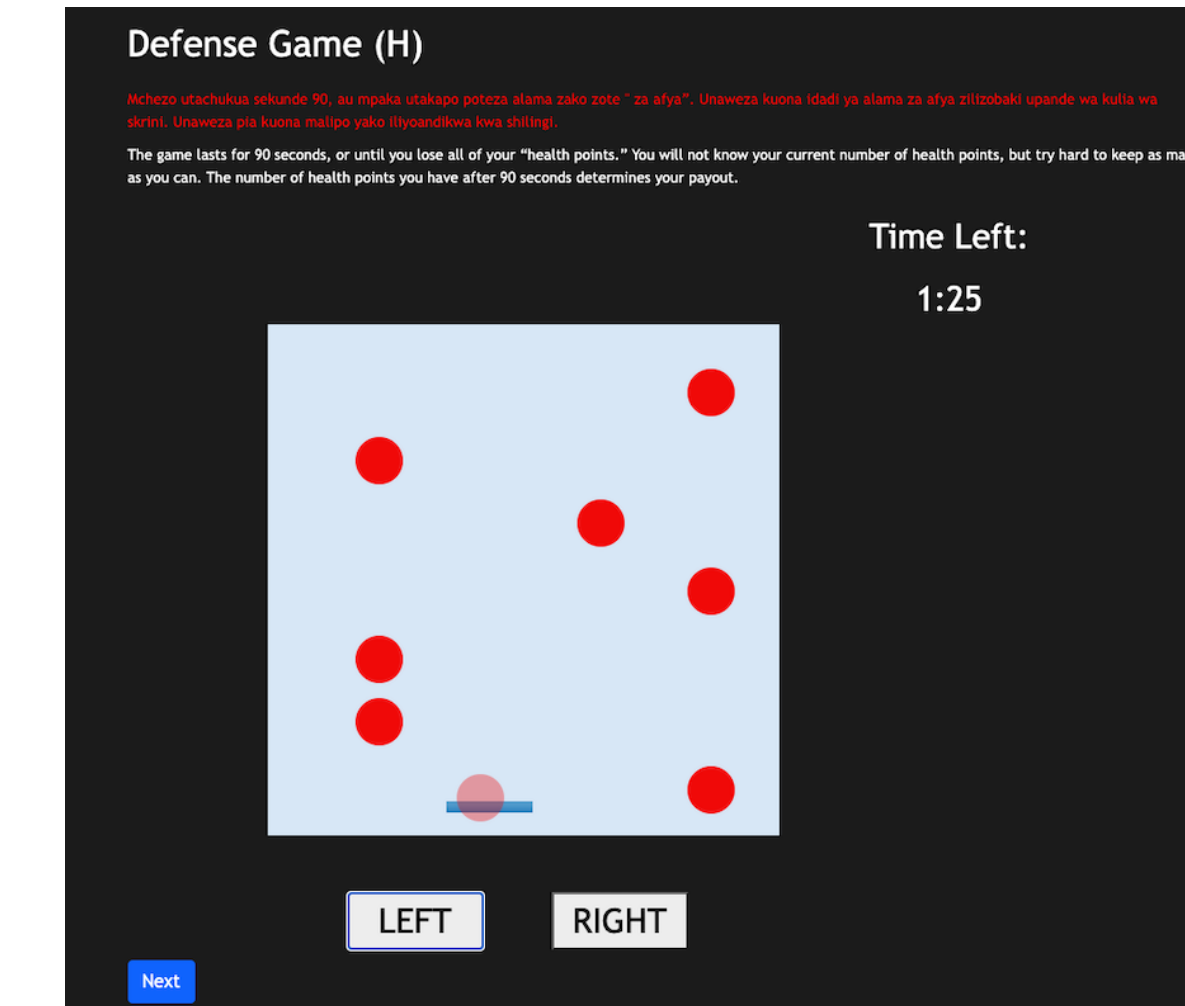
- 1,506 participants from Katoloni locality, Machakos, Kenya
- ~ 48/52 men-women split
- Median education: Secondary
- 70% involved in agriculture
- Most attend church > monthly
- 37% violence exposure (family)



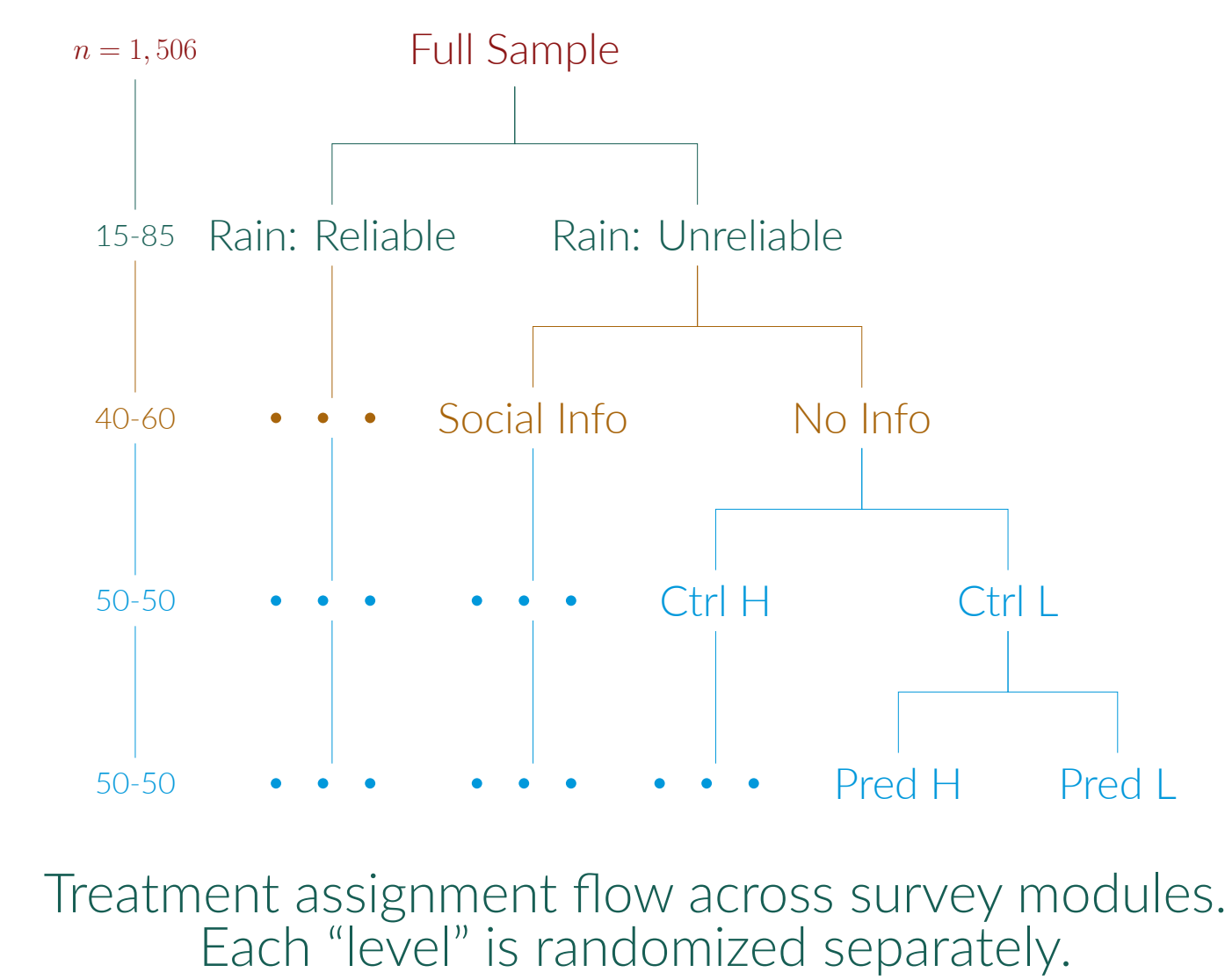
Machakos County

Lab-in-the-Field Setup

- Implemented by Busara Center for Behavioral Economics, with investigators from U. Capetown, Columbia, U. Dar Es Salaam, Harvard, Makerere U., MIT, U. Nairobi, Uganda Christian U.
- “Omnibus” design: 10 modules mostly-random order
- Other modules study: personality traits, savings and investment decisions, gambling, climate resilience, trust in mobile money etc.
- H1 and H2 treatments separately randomized (2x2 between subjects), QoI is marginal component effect of each treatment
- ≤ 20 participants in ~100 sessions do tasks on touchscreen tablets
- Lab sessions last up to 120 minutes, 329 Ksh. avg. incentive payout
- Perception treatment embedded in cash-incentivized game →



Manipulating Perceptions of Violence



Treatment assignment flow across survey modules. Each “level” is randomized separately.

Control Manipulation

- Participants assigned to higher/lower game difficulty
- Randomizes ability to keep in-game “lives,” worth real money (≈ a loaf of bread)
- Successful manipulation:
 - 44.6 Ksh. inter-group difference in payout ($p < .001$)
 - 29pp difference in perceived control over outcome ($p < .001$)
 - 95% power for 7pp. effect

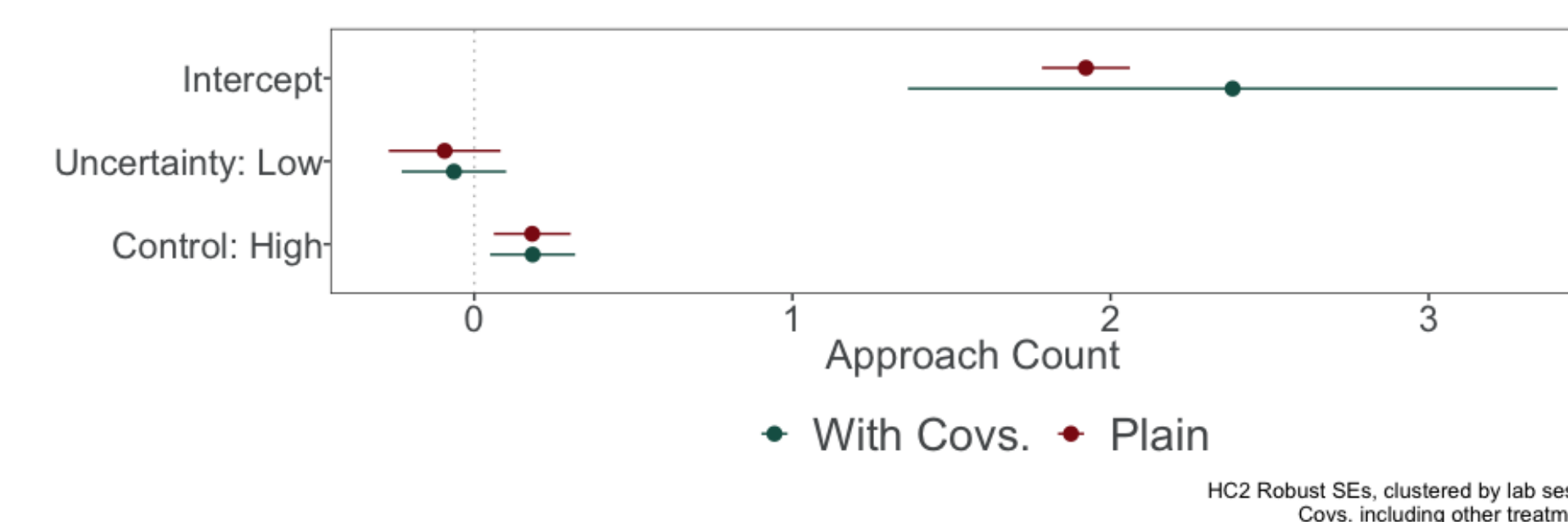
Uncertainty Manipulation(s)

- Participants see more/less information about game
- Successful manipulation for high difficulty group ($p = .006$), high variance in other group
- BUT** other omnibus modules manipulate uncertainty about other concepts
 - Future reliability of livelihood
 - Riskiness of small gambles
 - ≥ 89% power for these effects

Control and Uncertainty Perceptions Affect Strategy Preferences

Control :: Approach

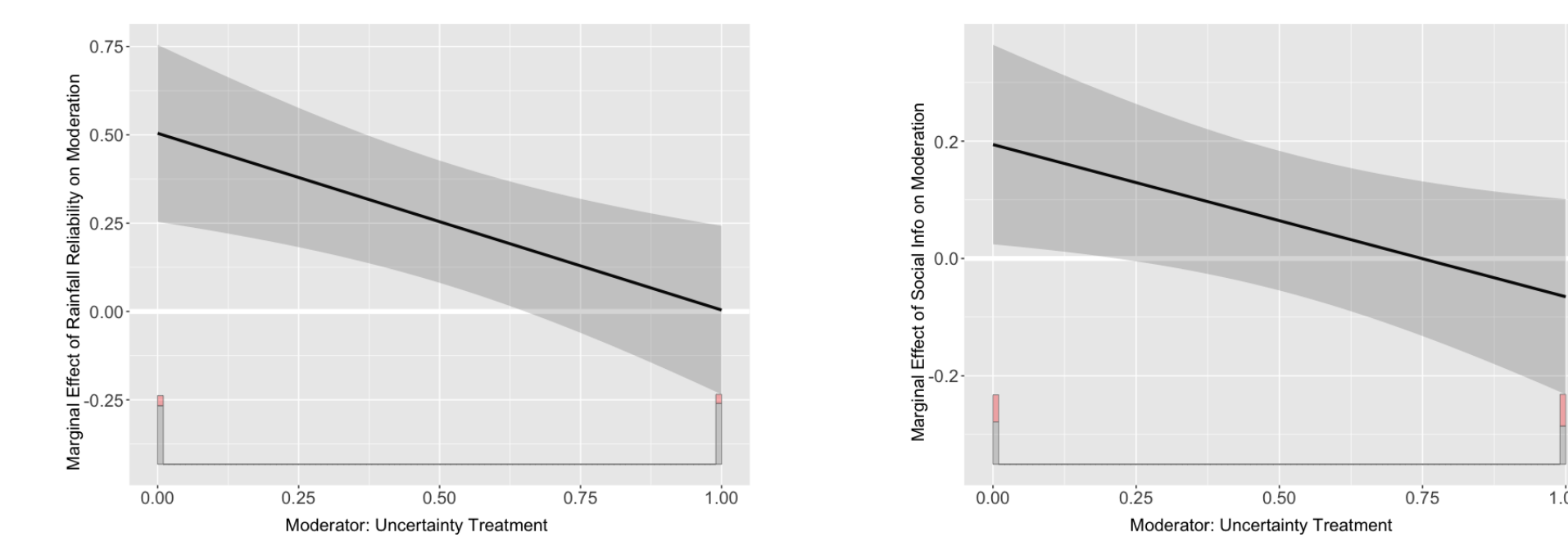
- High control perception increases preference for approach strategies by **7.5pp** in an pre-registered index of four decisions
- Results robust to clustering at session level
- Big signal, considering treatment intensity



Uncertainty* :: Disruptiveness

*with combined treatments

- Effect of main uncertainty treatment is small, insignificant (**3.8pp**, $p = .443$)
- When uncertainty treatment is aligned with rainfall un-reliability, gambling uncertainty treatments, effects are surprisingly large (**35.5pp**, $p = .005$; **17.7pp**, $p = .032$)



Interactions [3]: Rainfall × Ambiguity, Social Uncertainty × Ambiguity

Interference & Incidental Treatments

Substantive Finding

- Very incidental perceptions (even holdovers from other modules) *may* affect strategy preferences during hypothetical violence
- Difference driven by *interpretation not information*: Hypothetical violence description held constant across all treatment conditions

Experimental Methods Implications

- In studies with multiple treatments, previous, seemingly un-related treatments can spill-over. Important to:
 - Know how they relate to *D* in your study
 - Check that treatment statuses are not correlated
 - Consider controlling for them?
- Beware of especially exciting treatments [1]!



A study participant wins a soccer bet

Discussion

- “Shared” studies are increasingly common
- Many independently randomize, randomize order, ignore other modules
- This procedure shouldn’t bias estimates, but you might *mis-characterize the treatment*
- It’s also a missed opportunity for more precision [5]

References

- [1] Laura Barasa. Hitting the jackpot: how to curb youth gambling. - MIT Gov/Lab, Cambridge, 2023.
- [2] Nico H. Frijda. *The Laws of Emotion*. Psychology Press, Hove, Sep 2017.
- [3] Jens Hainmueller, Jonathan Mummolo, and Yiqing Xu. How much should we trust estimates from multiplicative interaction models? simple tools to improve empirical practice. *Political Analysis*, 27(2):163–192, 2019.
- [4] Jennifer S. Lerner and Dacher Keltner. Beyond valence: Toward a model of emotion-specific influences on judgement and choice. *Cognition and Emotion*, 14(4):473–93, 2000.
- [5] Winston Lin. Agnostic notes on regression adjustments to experimental data: Reexamining Freedman’s critique. *The Annals of Applied Statistics*, 7(1):295 – 318, 2013.
- [6] Katja Mehlhorn, Ben Newell, Peter Todd, Michael Lee, Kate Morgan, Victoria Braithwaite, Daniel Hausmann, Klaus Fiedler, and Cleotilde Gonzalez. Unpacking the exploration–exploitation tradeoff: A synthesis of human and animal literatures. *Decision*, 2(3):191–215, 2015.
- [7] Aidan Milliff. Making sense and making choices: How civilians choose survival strategies during violence. Working paper, Stanford University, Stanford, CA, 2023.
- [8] Angela J. Yu and Peter Dayan. Uncertainty, neuromodulation, and attention. *Neuron*, 46(4):681–692, 2005.

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