

Discussion of “Inattentive Valuation and Reference-Dependent Choice” by Mike Woodford

Jennifer La'O

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The Shaw & Shaw Experiment

The Experiment

- a letter $\{T, V\}$ is flashed on the screen in 2 locations $\{H, L\}$
- example:

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Formalization of Experiment

- states of the world $x_{ij} \in X$
- nature
 - draws location $i \in \{H, L\}$ from dist (π_H, π_L)
 - draws letter $j \in \{T, V\}$ from dist. $(1/2, 1/2)$

Formalization of Experiment

- states of the world $x_{ij} \in X$
- nature
 - draws location $i \in \{H, L\}$ from dist (π_H, π_L)
 - draws letter $j \in \{T, V\}$ from dist. $(1/2, 1/2)$
- agents are asked to correctly identify the letter

$$\min \sum \mathbb{I}(k \neq j)$$

where $k \in \{T, V\}$ is agent's response

Shaw & Shaw Results

- $(\pi_H, \pi_L) = (.5, .5)$
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- $(\pi_H, \pi_L) = (.5, .5)$
 - same fraction of errors made in both locations
- $(\pi_H, \pi_L) = (.9, .1)$
 - fraction of errors different across locations
 - greater fraction of errors made in L location

Rational Inattention

Rational Inattention: General Setup (Sims)

- primitives
 - $x \in X$ states of the world
 - $\pi(x)$ prior
 - agent's payoffs (preferences)

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- primitives
 - $x \in X$ states of the world
 - $\pi(x)$ prior
 - agent's payoffs (preferences)
- agent chooses information channel s.t. constraint
 - $s \in S$ signals (perceived states)
 - $p(s|x)$ conditional probabilities

Rational Inattention: the Shaw & Shaw Experiment

- for simplicity, assume agents can perfectly observe location i
- agents observe some signal $k \in \{T, V\}$ of letter j ,
but signal has error:

$$\begin{aligned} p(ij|ij) &= 1 - e_i \\ p(ik|ij) &= e_i \quad \text{for } k \neq j \end{aligned}$$

- agent optimally chooses probability of error $\mathbf{e} = \{e_H, e_L\}$

Rational Inattention Problem

- objective

$$\min_{\mathbf{e}} \sum_i \pi_i e_i + \theta I(\mathbf{e})$$

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- mutual information

$$I(\mathbf{e}) \equiv H(x) - H(x|k)$$

is reduction in entropy (measure of uncertainty)

$$H(x) \equiv - \sum_{x \in X} p(x) \log p(x)$$

Rational Inattention Problem

- mutual info

$$I = - \underbrace{\left(\sum_i \pi_i \log \pi_i + \log \frac{1}{2} \right)}_{H(x)} - \underbrace{\sum_i \pi_i h(e_i)}_{H(x|k)}$$

- $h(e_i)$ is conditional entropy within location i

$$h(e_i) = -(1 - e_i) \log(1 - e_i) - e_i \log e_i$$

Rational Inattention Solution

- objective

$$\min_{\mathbf{e}} \sum_i \pi_i e_i + \theta \left(H(x) - \sum_i \pi_i h(e_i) \right)$$

- FOC

$$\pi_i = \theta \pi_i h'(e_i)$$

Rational Inattention Solution

- objective

$$\min_{\mathbf{e}} \sum_i \pi_i e_i + \theta \left(H(x) - \sum_i \pi_i h(e_i) \right)$$

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Rational Inattention Solution

- FOC $1 = \theta h'(e_i)$ implies

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Rational Inattention Solution

- FOC $1 = \theta h'(e_i)$ implies

$$e_i = \bar{e}, \quad \forall i$$

- solution incompatible with Shaw & Shaw results
- why? letter j is the only payoff relevant variable
 - cost/benefit same across locations
 - no need to add additional info of location

Woodford's Alternative Formulation

Alternative Formulation

- rational inattention problem

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- rational inattention problem

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- alternative problem (Woodford)

$$\min_{\mathbf{e}} \sum_i \pi_i e_i + \theta C(\mathbf{e})$$

$$C(\mathbf{e}) \equiv \max_{\pi} I(\mathbf{e}; \pi)$$

Woodford's Cost Function

- for given prior, $I(\mathbf{e})$ is *actual* reduction in entropy

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- for given prior, $I(\mathbf{e})$ is *actual* reduction in entropy
- $C(\mathbf{e})$ is *potential* reduction in entropy over any possible prior

$$C(\mathbf{e}) \equiv \max_{\pi} I(\mathbf{e}; \pi)$$

- given channel \mathbf{e} , choose prior to maximize reduction in entropy

Alternative Formulation

- rational inattention problem

$$\min_{\mathbf{e}} \sum_i \pi_i e_i + \theta I(\mathbf{e})$$

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Alternative Formulation

- rational inattention problem

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- alternative problem (Woodford)

$$\min_{\mathbf{e}} \sum_i \pi_i e_i + \theta \log \left(\sum_i \exp(-h(e)) \right)$$

Results of Woodford's formulation

- FOC

$$\pi_i = \tilde{\theta} \exp(-h(e_i)) h'(e_i)$$

RHS decreasing in e_i

- thus, e_i^* inversely related to π_i

$$e_i^* = g(\pi_i) \quad g' < 0$$

- compatible with Shaw & Shaw results!
 - in fact, Woodford calibrates θ and predictions fit quite well

Comments

Interpreting Woodford's Cost Function

- Woodford's new cost function

$$C(\mathbf{e}) \equiv \max_{\pi} I(\mathbf{e}; \pi)$$

represents greatest *potential* entropy reduction from channel \mathbf{e}

- why max? could have prior μ over priors

$$C(\mathbf{e}) = \int I(\mathbf{e}; \pi) \mu(\pi) d\pi$$

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- need: cost/benefit to differ in π_i
 - isomorphic to rational inattention problem with θ_i
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Interpreting Woodford's Cost Function

- need: cost/benefit to differ in π_i
 - isomorphic to rational inattention problem with θ_i
 - could have chosen arbitrary cost function C
- however, Woodford's formulation more elegant
 - retains properties of Shannon measure of info
 - more restrictive than just choosing θ_i or C_i
 - explains experimental anomalies (stochastic choice, focusing effects, choice-set effects, reference dependence, etc.)

Interpreting Woodford's Cost Function

$$C(\mathbf{e}) \equiv \max_{\pi} I(\mathbf{e}; \pi)$$

- possible interpretation: sequential game
 - first stage. agent chooses channel \mathbf{e}
 - second stage. evil nature chooses prior π to max $I(\mathbf{e}; \pi)$
- possible interpretation: robust control

$$\min_{\mathbf{e}} \sum \pi_i e_i + \theta \left\{ \max_{\pi} I(\mathbf{e}; \pi) \right\}$$

Exogenous Coding/Channel matters

- rational inattention \rightarrow only payoff relevant variables matter
 - underlying shocks out in the world
 - we choose our perception of them
 - depends only on: curvature of payoffs, volatility of shocks
- but exogenous information channels matter
 - newspapers, own sales, competitors' prices, word-of-mouth