

Good News and Bad news: Representation Theorem and Applications

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- Instrument: Monotonicity of the Likelihood Ratio Property (MLRP) for conditional densities .

- Theory and Application to the Moral Hazard Model

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- Application to the Persuasion game (model studied more deeper by Malin)

Theory and Application to Moral Hazard Problem

Representation theorem

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- Let $f(x/\theta)$, the conditional density (or probability mass) function on X when $\tilde{\theta}$ takes the particular value θ .

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- Let $f(x/\theta)$, the conditional density (or probability mass) function on X when $\tilde{\theta}$ takes the particular value θ .
- We define the relation "more favorable than" such that:
a signal x is more favorable than y if for every nondegenerate prior distribution G for θ , the posterior distribution $G(. / x)$ dominates the posterior distribution $G(. / y)$ in the sense of strict first-order stochastic dominance.

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- The Bayes' rule in odds form for x is such that:

$$\frac{g(\bar{\theta}/x)}{g(\theta/x)} = \frac{f(x/\bar{\theta}) g(\bar{\theta})}{f(x/\theta) g(\theta)} \quad (1)$$

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- We use this bayes' rule also for y .

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- Then, it follows that:

$$\frac{f(x/\bar{\theta})}{f(x/\theta)} > \frac{f(y/\bar{\theta})}{f(y/\theta)} \quad (2)$$

Theory and Application to Moral Hazard Problem

Representation theorem

Theorem

x is more favorable than y if and only if for every $\theta < \bar{\theta}$,

$$f(x/\bar{\theta})f(y/\theta) - f(x/\theta)f(y/\bar{\theta}) > 0.$$

Theorem

The family of densities $\{f(./\theta)\}$ has the strict MLRP iff $x > y$ implies that x is more favorable than y , i.e., that the posterior distribution $G(. / x)$ dominates the posterior distribution $G(. / y)$ in the sense of strict first-order stochastic dominance .

- When comparing two conditional functions, the MLRP establishes that a comparatively high outcome is always more likely under one distribution among the pair. This implies the FOSD is both necessary and sufficient on domain-conditioned distribution functions.

- Note that two signals are called **comparable** if they are equivalent (the posterior beliefs about $\tilde{\theta}$ are identical) or if one is more favorable than the other.

Theorem

Let X be general and suppose that any two signals in X are comparable. Then, there exists a function $H : X \rightarrow \mathbb{R}$ such that $H(\tilde{x})$ is a sufficient statistic for \tilde{x} and such that the densities of $H(\tilde{x})$ have the strict MLRP.

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- This sufficient statistic captures all information that is relevant to guessing the values of the unobservable parameters.

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Theorem

Let \tilde{x} be a random variable whose densities have the strict MLRP. For any two intervals $[a, b]$ and $[c, d]$ with $a \geq c$ and $b \geq d$, where at least one inequality is strict, the signal $\{\tilde{x} \in [a, b]\}$ is more favorable than $\{\tilde{x} \in [c, d]\}$.

Theory and Application to Moral Hazard Problem

Moral Hazard

- Let θ an agent's effort influencing the profit of a venture π as well as the random state of nature $\tilde{\alpha}$ such that: $\tilde{\pi} = \pi(\tilde{\alpha}, \theta)$. This effort always improved profits ($\partial\pi/\partial\theta > 0$), but there are diminishing returns to effort ($\partial^2\pi/\partial\theta^2 < 0$).

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- Let $V(x, \theta) = U(x) - \theta$, the agent's payoffs that is an increasing function of his wealth x and a decreasing function of effort θ .

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- Let $V(x, \theta) = U(x) - \theta$, the agent's payoffs that is an increasing function of his wealth x and a decreasing function of effort θ .
- We assume that the agent is risk averse, i.e., $U'' < 0$.
- The principal's payoff $G(x)$ is such that $G' > 0$ and $G'' \leq 0$.
- $s(\cdot)$ is the sharing rule depending only on π ; $\tilde{\alpha}$ and θ are unobservable by the principal.

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- Holmstrom (79) showed that the optimal sharing rule must satisfy the following relationship for some θ^* , b , and c :

$$\frac{G'(\pi - s(\pi))}{U'(s(\pi))} = b + c \frac{f_{\theta}(\pi/\theta^*)}{f(\pi/\theta^*)}$$

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- Greater profits are evidence of greater efforts by the agent, then, the sharing rule should slope upwards to provide the correct incentives.

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- The salesman's payoff is his commission, which is some increasing function of q . It is assumed that F is bounded, increasing, concave, and differentiable, and that $F'(0) = +\infty$.

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- If the buyer purchases q units of the commodity at price p , his payoff is $\tilde{\theta}F(q) - pq$.
- The salesman's payoff is his commission, which is some increasing function of q . It is assumed that F is bounded, increasing, concave, and differentiable, and that $F'(0) = +\infty$.
- Let the salesman have N pieces of data about his product, represented by $\tilde{x} = (\tilde{x}_1, \dots, \tilde{x}_N)$.

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- The purchasing strategy $b(S)$ depend on the salesman report.

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- The salesman report $r(x)$ is such that $x \in r(x)$.
- The purchase decision is q .
- The purchasing strategy $b(S)$ depend on the salesman report.
- Concept of the sequential equilibrium such that the buyer interprets the reports he receives and then concludes that $\tilde{x} \in C(s)$.

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- For every $x \in \mathbb{R}^N$, $r(x)$ solves $\max_S b(S)$, subject to $x \in S$.
- For every S in the range of r , $c(S) = r^{-1}(S)$.

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- A reporting strategy r is called a strategy of *full disclosure* if r together with any optimal response (b, c) satisfies $b(r(x)) \equiv b(\{x\})$.

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- Consequently, r is a strategy of full disclosure if $E[\tilde{\theta}|r(\tilde{x}) = x] \equiv E[\tilde{\theta}|r(\tilde{x}) = r(x)]$.

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Theorem

At every sequential equilibrium of the sales encounter game, the salesman uses a strategy of full disclosure.

The persuasion game

Generalization

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Theorem

The modified sales encounter game has a sequential equilibrium in which the salesman always reports the k most favorable observations.