

# Randomization and ambiguity perception

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## MOTIVATION

- People dislike unknown probabilities—***ambiguity aversion***
  - $\langle \text{unambiguous event} \rangle \succ \langle \text{ambiguous event} \rangle$
  - Incompatible w/ SEU ©Ellsberg (1961)
  - Motivated various ambiguity models MEU, CEU, smooth ambiguity, ...

### This paper

$\langle \text{less \# of ambiguous events} \rangle \succ \langle \text{more \# of ambiguous events} \rangle$

- e.g., Machina's (2009) paradox Lab experiment: L'Haridon · Placido (2010)
  - Incompatible w/ many ambiguity models MEU, CEU, smooth ambiguity, ...

## MAIN AXIOM

$\langle \text{less \# of ambiguous events} \rangle \succ \langle \text{more \# of ambiguous events} \rangle$

### Ex ante aversion to randomization

\*informal

DM dislikes randomization *over* acts.

- **Idea:** randomization increases # of relevant ambiguous events
- Domain = { lotteries over AA acts }  $\supset$  { AA acts }
  - AA act: state  $\mapsto$  outcome distribution

## RANDOMIZATION & AMBIGUOUS EVENTS

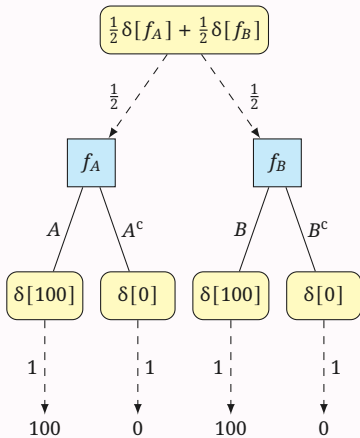
### Example

Compare the two acts & their randomization:

$$f_A = \begin{cases} 100 & \text{if } \geq t^\circ\text{C @Athens} \\ 0 & \text{else} \end{cases} \quad f_B = \begin{cases} 100 & \text{if } \geq t^\circ\text{C @Beijing} \\ 0 & \text{else} \end{cases}$$

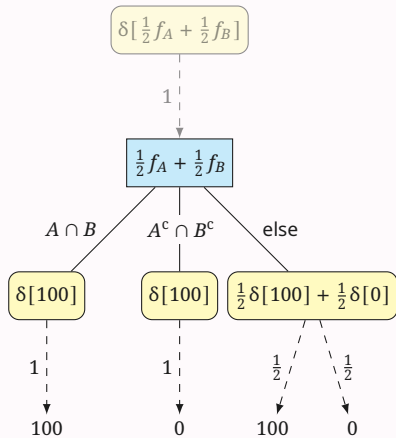
$$\frac{1}{2}\delta[f_A] + \frac{1}{2}\delta[f_B] \neq \frac{1}{2}f_A + \frac{1}{2}f_B$$

## RANDOMIZATION & AMBIGUOUS EVENTS



*ex ante randomization*

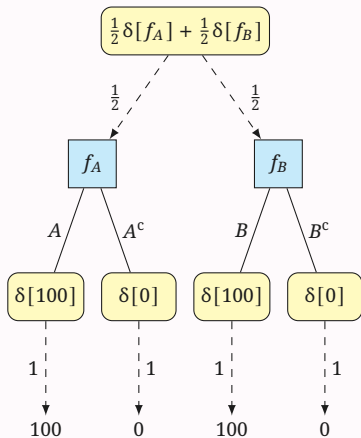
$A = [\geq t^\circ\text{C @Athens}],$



*ex post randomization*

$B = [\geq t^\circ\text{C @Beijing}]$

## RANDOMIZATION & AMBIGUOUS EVENTS



- Both  $A$  and  $B$  are relevant in the lottery
- $\delta[f_A] \succsim \delta[f_B] \implies \delta[f_A] \succsim \frac{1}{2}\delta[f_A] + \frac{1}{2}\delta[f_B]$

*DM dislikes **ex ante** randomization*

## OVERVIEW

- Explore the implications of *ex ante aversion to randomization*
  - Novel testable manifestation of  
⟨less # of ambiguous events⟩  $\succ$  ⟨more # of ambiguous events⟩
- Characterize the ***costly ambiguity perception*** model
  - Generalizes MEU— $U_{\text{MEU}}(f) = \min_{\mu \in M} \int u \circ f \, d\mu$      $M$ : ambiguity perception
  - DM optimizes ambiguity perception at a cost
- Fully identify the parameters of the model
  - Ex ante randomization is important!

## PRIMITIVES

- $\Omega$ : finite set of **states**
- $X$ : set of **outcomes**
- $\mathcal{F}$ : set of **acts**  $f: \Omega \rightarrow \Delta_s(X)$ 
  - $\Delta_s(X)$ : set of finitely supported distributions on  $X$
  - Identify each  $p \in \Delta_s(X)$  w/ the constant act  $\omega \mapsto p$
  
- $\succsim$ : DM's preference on  $\Delta_s(\mathcal{F})$

## REPRESENTATION

### Definition

A **costly ambiguity perception representation** of  $\succsim$  is  $\langle u, (\mathbb{M}, c) \rangle$  s.t.

- $u: \Delta_s(X) \rightarrow \mathbb{R}$  is surjective & mixture linear;
- $\mathbb{M}$  is a compact set of nonempty closed convex subsets of  $\Delta(\Omega)$ ;
- $c: \mathbb{M} \rightarrow \mathbb{R}_+$  is
  1. lower semicontinuous;
  2.  $\min c(\mathbb{M}) = 0$ ;
  3.  $M \subseteq M' \implies c(M) \geq c(M')$ ;
- $\succsim$  is represented by

$$U_{\text{CAP}}(P) = \max_{M \in \mathbb{M}} \left[ \int \left( \min_{\mu \in M} \int u \circ f \, d\mu \right) dP(f) - c(M) \right].$$

## SPECIAL CASES—W/O EX ANTE RANDOMIZATION

- $\mathbb{M} = \{M\} \rightsquigarrow$  maxmin expected utility ©Gilboa · Schmeidler (1989)

$$U_{\text{MEU}}(\delta[f]) = \max_{M \in \mathbb{M}} \left[ \int \left( \min_{\mu \in M} \int u \circ f \, d\mu \right) dP(f) - c(M) \right]$$

- $c \equiv 0 \rightsquigarrow$  dual-self expected utility ©Chandrasekher et al. (2022)

$$U_{\text{DSEU}}(\delta[f]) = \max_{M \in \mathbb{M}} \left[ \int \left( \min_{\mu \in M} \int u \circ f \, d\mu \right) dP(f) - c(M) \right]$$

- $\mathbb{M} = \{ \text{core}(v) \mid v \in V \} \rightsquigarrow$  optimal ambiguity attitude

$V$ : a set of convex capacities = supermodular set fn.s

©Payró et al. (2026)

$$U_{\text{OAA}}(\delta[f]) = \max_{M \in \mathbb{M}} \left[ \int \left( \min_{\mu \in M} \int u \circ f \, d\mu \right) dP(f) - c(M) \right]$$

# BASIC AXIOMS

## A1—Regularity

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$\succsim$  is nondegenerate, complete, transitive, & mixture continuous.

## A2—Monotonicity

---

“ $\succsim$  respects statewise dominance.”

## A7—Unboundedness

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“Utility has unbounded range.”

## TIMING AXIOMS

### A3—Attraction to ex post randomization

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" $\langle \text{XP randomization} \rangle \succeq \langle \text{XA randomization} \rangle$ ."

- Ex post randomization may smooth payoff variability

$$\frac{1}{2} \begin{bmatrix} 100 \\ 0 \end{bmatrix} + \frac{1}{2} \begin{bmatrix} 0 \\ 100 \end{bmatrix} = \begin{bmatrix} \frac{1}{2}\delta[100] + \frac{1}{2}\delta[0] \\ \frac{1}{2}\delta[100] + \frac{1}{2}\delta[0] \end{bmatrix}$$

### A4—Indifference to randomization timing of constant acts

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" $\langle \text{XP randomization of } f \text{ \& } p \rangle \sim \langle \text{XA randomization of } f \text{ \& } p \rangle$ ."

- Ex post mixing of  $p$  preserves the pattern of payoff variability in  $f$

$$\frac{1}{2}\delta \left[ \begin{bmatrix} 100 \\ 0 \end{bmatrix} \right] + \frac{1}{2}\delta[0] \sim \begin{bmatrix} \frac{1}{2}\delta[100] + \frac{1}{2}\delta[0] \\ \delta[0] \end{bmatrix}$$

## EX ANTE RANDOMIZATION AXIOMS

### A5—Ex ante aversion to randomization

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For each  $(\lambda, (P, Q)) \in [0, 1] \times \Delta_s(\mathcal{F})^2$ ,

$$P \succsim Q \implies P \succsim \lambda P + (1 - \lambda)Q.$$

- Ex ante randomization increases # of relevant ambiguous events

## EX ANTE RANDOMIZATION AXIOMS

### A6—Independence of constant acts

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For each  $(\lambda, (P, Q), (p, q)) \in [0, 1] \times \Delta_s(\mathcal{F})^2 \times \Delta_s(X)^2$ ,

$$\lambda P + (1 - \lambda)\delta[p] \succeq \lambda Q + (1 - \lambda)\delta[p]$$

$$\implies \lambda P + (1 - \lambda)\delta[q] \succeq \lambda Q + (1 - \lambda)\delta[q].$$

- **Idea:** constant act terms do not affect ambiguity

## CHARACTERIZATION

### Representation theorem

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$\succsim$  satisfies A1–A7  $\iff \succsim$  has a CAP representation.

- A1—*Regularity*
- A2—*Monotonicity*
- A3—*Attraction to ex post randomization*
- A4—*Indifference to randomization timing of constant acts*
- A5—*Ex ante aversion to randomization*
- A6—*Independence of constant acts*
- A7—*Unboundedness*

## CHARACTERIZATION

### Corollary

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$\succsim$  satisfies A1–A4 + *independence*  $\iff \succsim$  has an MEU representation.

- A1—*Regularity*
- A2—*Monotonicity*
- A3—*Attraction to ex post randomization*
- A4—*Indifference to randomization timing of constant acts*
- A5—*Ex ante aversion to randomization*
- A6—*Independence of constant acts*
- A7—*Unboundedness*

# UNIQUENESS

## Definition

A CAP representation  $\langle u, (\mathbb{M}, c) \rangle$  is **convex** if  $\mathbb{M}$  and  $c$  are convex.

## Identification theorem

Every CAP preference  $\succsim$  has convex rep.s  $\langle u, (\mathbb{M}^*, c^*) \rangle$  &  $\langle u, (\mathbb{M}^{**}, c^{**}) \rangle$  s.t. for each convex rep.  $\langle u, (\mathbb{M}, c) \rangle$  of  $\succsim$ ,

1.  $\mathbb{M}^* \subseteq \mathbb{M} \subseteq \mathbb{M}^{**}$ ;      2.  $c = c^{**}|_{\mathbb{M}}$ .

1.  $\exists$  smallest & largest feasible sets      2. Cost is essentially unique

## SMALLEST/LARGEST REPRESENTATION

- $(u, \mathbb{M}^*)$  represents the *expected utility core*: smallest feasible set

$$P \succ^* Q \iff$$

$$\int \left( \min_{\mu \in M} \int u \circ f \, d\mu \right) dP(f) \geq \int \left( \min_{\mu \in M} \int u \circ f \, d\mu \right) dQ(f) \quad \forall M \in \mathbb{M}^*$$

$\succ^*$ : the largest subrelation of  $\succ$  satisfying *independence*

cf. Cerreia-Vioglio et al. (2015), Ke · Zhang (2020), ...

- $(\mathbb{M}^{**}, c^{**})$  is explicitly written as: largest cost structure

$$\mathbb{M}^{**} = \langle \text{the } \supseteq\text{-upward closure of } \mathbb{M}^* \rangle = \text{dom } c_{\succ, u}^*$$

$$c^{**}(M) = c_{\succ, u}^*(M) \equiv \sup \left\{ \int \left( \min_{\mu \in M} \int u \circ f \, d\mu \right) dP(f) - u(\bar{P}) \mid P \sim \delta[\bar{P}] \right\}$$

cf. Maccheroni et al. (2006), de Oliveira et al. (2017)

## IMPORTANCE OF EX ANTE RANDOMIZATION

### Example

Let  $\Omega = \{0, 1\}$  and  $X = \mathbb{R}$ . Identify  $\mu \in [0, 1]$  w/ the prob. of  $\omega = 1$ .

Consider  $\succsim_1$  and  $\succsim_2$  represented by  $\langle u, (\mathbb{M}_i, c_i) \rangle$ :

$$u(\delta[x]) = x, \quad \mathbb{M}_1 = \{\{\frac{1}{2}\}\}, \quad \mathbb{M}_2 = \{[0, \frac{1}{2}], [\frac{1}{2}, 1]\}, \quad c_1 = c_2 \equiv 0.$$

While  $\succsim_1$  is neutral,  $\succsim_2$  is averse to ex ante randomization.

Nevertheless,  $\succsim_1$  and  $\succsim_2$  coincide on the set of acts:

$$\begin{aligned} U_2(\delta[h]) &= \left[ \frac{1}{2}h(0) + \frac{1}{2}(h(0) \wedge h(1)) \right] \vee \left[ \frac{1}{2}(h(0) \wedge h(1)) + \frac{1}{2}h(1) \right] \\ &= \frac{1}{2}(h(0) \wedge h(1)) + \frac{1}{2}(h(0) \vee h(1)) = U_1(\delta[h]) \quad \forall h \in \mathcal{F}. \quad \diamond \end{aligned}$$

## COMPARATIVE AMBIGUITY AVERSION

### Definition

cf. Ghirardato · Marinacci (2002)

$\succsim_1$  is *more ambiguity-averse than*  $\succsim_2$  if

$$\delta[p] \succsim_2 P \implies \delta[p] \succsim_1 P \quad \forall (P, p) \in \Delta_S(\mathcal{F}) \times \Delta_S(X).$$

### Proposition

The following are equivalent.

- $\succsim_1$  is more ambiguity-averse than  $\succsim_2$ .
- $u_1 \approx u_2$ ,  $M_1^{**} \subseteq M_2^{**}$ , &  $c_1^{**} \geq c_2^{**}|_{M_1^{**}}$ .

- Cost function describes ambiguity attitudes

## COMPARATIVE AVERSION TO ADDITIONAL AMBIGUITY

### Definition

$\succsim_1$  is *more averse to additional ambiguity than*  $\succsim_2$  if

$$\begin{aligned} & \lambda\delta[p] + (1 - \lambda)Q \succsim_2 \lambda P + (1 - \lambda)Q \\ \implies & \lambda\delta[p] + (1 - \lambda)Q \succsim_1 \lambda P + (1 - \lambda)Q \end{aligned}$$

for each  $(\lambda, (P, Q), p) \in [0, 1] \times \Delta_s(\mathcal{F})^2 \times \Delta_s(X)$ .

- $\implies (\not\Leftarrow) \succsim_1$  is more ambiguity-averse than  $\succsim_2$

## COMPARATIVE AVERSION TO ADDITIONAL AMBIGUITY

- Define  $\mathcal{C}_i: \Delta_s(\mathcal{F}) \rightrightarrows \mathbb{M}_i$  by

$$\mathcal{C}_i(P) = \arg \max_{M \in \mathbb{M}_i} \left[ \int \left( \min_{\mu \in M} \int u_i \circ f \, d\mu \right) dP(f) - c_i(M) \right]$$

### Proposition

---

*The following are equivalent.*

- $\succsim_1$  is more averse to additional ambiguity than  $\succsim_2$ .
- $u_1 \approx u_2$  &

$$\forall P \in \Delta_s(\mathcal{F}), M_1 \in \mathcal{C}_1(P) \implies \exists M_2 \in \mathcal{C}_2(P): M_2 \subseteq M_1.$$

- “DM 2 always reduces ambiguity perception more than DM 1”

## SUMMARY

- DM who dislikes increases in ambiguity is studied
- DM is averse to randomization over acts

$$P \succsim Q \implies P \succsim \lambda P + (1 - \lambda)Q$$

- DM behaves as if optimizing her ambiguity perception at a cost

$$U_{\text{CAP}}(P) = \max_{M \in \mathcal{M}} \left[ \int \left( \min_{\mu \in M} \int u \circ f \, d\mu \right) dP(f) - c(M) \right]$$

- Parameters are uniquely identified & given behavioral meanings
- In the paper ... ,

➔ characterization of several special cases

DSEU, OAA, ...

➔ more on comparatives

absolute notion, xA rand., ...

➔ Machina's (2009) paradoxes: reflection & 50–51 examples

## RELATED WORKS

### Choice of ambiguity perception w/o ex ante randomization

- Chandrasekher · Frick · Iijima · Le Yaouanq (2022) ← NO COST
- Payró · Takeoka · Xia (2026) ← CHOICE OF CAPACITIES

### Ambiguity aversion & ex ante randomization

- Saito (2015), Ke · Zhang (2020)
- Their DM believes even ex ante randomization reduces ambiguity

$$U_{\text{DMEU}}(P) = \min_{M \in \mathcal{M}} \int \left( \min_{\mu \in M} \int u \circ f \, d\mu \right) dP(f)$$

### Models consistent w/ Machina's examples

- Siniscalchi (2009), Dillenberger · Segal (2015), He (2021), ...

## SUMMARY

- DM who dislikes increases in ambiguity is studied
- DM is averse to randomization over acts

$$P \succsim Q \implies P \succsim \lambda P + (1 - \lambda)Q$$

- DM behaves as if optimizing her ambiguity perception at a cost

$$U_{\text{CAP}}(P) = \max_{M \in \mathcal{M}} \left[ \int \left( \min_{\mu \in M} \int u \circ f \, d\mu \right) dP(f) - c(M) \right]$$

- Parameters are uniquely identified & given behavioral meaning

## APPENDIX

**A** SPECIAL CASES

**B** MORE ON COMPARATIVES

**C** REFLECTION EXAMPLE

**D** 50-51 EXAMPLE

## DUAL-SELF EXPECTED UTILITY

### A8—Strong independence of constant acts

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For each  $(\lambda, (P, Q), p) \in [0, 1] \times \Delta_s(\mathcal{F})^2 \times \Delta_s(X)$ ,

$$P \succsim Q \iff \lambda P + (1 - \lambda)\delta[p] \succsim \lambda Q + (1 - \lambda)\delta[p].$$

### Corollary

---

$\succsim$  satisfies A1–A7 + A8  $\iff \succsim$  has a DSEU representation.

## OPTIMAL AMBIGUITY ATTITUDE

### A9—Indifference to randomization timing of comonotonic acts

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For each  $((\kappa, \lambda), P, (f, g)) \in [0, 1]^2 \times \Delta_s(\mathcal{F}) \times \mathcal{F}^2$ ,

if  $\delta[f(\omega)] \succeq \delta[f(\omega')]$  is equivalent to  $\delta[g(\omega)] \succeq \delta[g(\omega')]$ , then

$$\kappa\delta[\lambda f + (1 - \lambda)g] + (1 - \kappa)P \sim \kappa[\lambda\delta[f] + (1 - \lambda)\delta[g]] + (1 - \kappa)P.$$

### Corollary

---

$\succsim$  satisfies A1–A7 + A9  $\iff \succsim$  has an OAA representation. 

# APPENDIX

**A** SPECIAL CASES

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## ABSOLUTE AMBIGUITY AVERSION

### Definition

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$\succsim$  is *absolutely ambiguity-averse* if

$\succsim$  is more ambiguity-averse than an SEU preference.

### Proposition

---

A CAP  $\succsim$  w/  $\langle u, (\mathbb{M}, c) \rangle$  is absolutely ambiguity-averse  $\iff \bigcap \mathbb{M} \neq \emptyset$ .

- $\bigcap \mathbb{M}$  is independent of the choice of rep. (even w/o convexity)

## COMPARATIVE AVERSION OF CAP AGAINST MEU

### Proposition

---

Let  $\langle (u, (\mathbb{M}, c)) \rangle$  be a CAP rep. of  $\succsim$ , and  $(v, L)$  an MEU rep. of  $\succcurlyeq$ .

The following are equivalent.

- a.  $u \approx v$  &  $\cap \mathbb{M} \supseteq L$ .
- b.  $\succsim$  is more ambiguity-averse than  $\succcurlyeq$ .
- c.  $\succsim$  is more averse to additional ambiguity than  $\succcurlyeq$ .

- Two comparatives coincide when the benchmark is MEU
  - They are different else

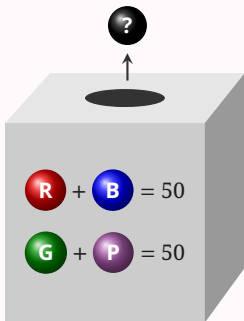
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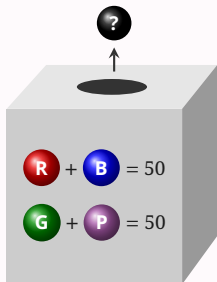


	R	B	G	P
$f_1$	100	200	100	0
$f_2$	100	100	200	0
$f_3$	0	200	100	100
$f_4$	0	100	200	100

- SEU + symmetry  $\implies f_1 \sim f_2 \sim f_3 \sim f_4$
- Typical pattern:  $f_1 \prec f_2$  &  $f_3 \succ f_4$        $\langle \star \rangle$  ©L'Haridon · Placido (2010)
- $\langle \star \rangle$  is inconsistent w/ *uncertainty aversion* 😞

©Baillon et al. (2011)  
\*under natural assumptions

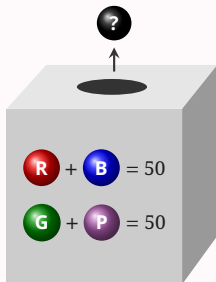
## REFLECTION EXAMPLE VS. *UNCERTAINTY AVERSION*



	R	B	G	P
$f_1$	100	200	100	0
$f_2$	100	100	200	0
$f_3$	0	200	100	100
$f_4$	0	100	200	100

	R	B	G	P		R	B	G	P
$f_1$	0	100	100	0	+	100	100	0	0
$f_2$	0	0	200	0	+	100	100	0	0
$f_3$	0	200	0	0	+	0	0	100	100
$f_4$	0	100	100	0	+	0	0	100	100

## REFLECTION EXAMPLE VS. *UNCERTAINTY AVERSION*



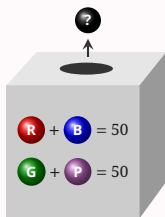
	R	B	G	P
$f_1$	100	200	100	0
$f_2$	100	100	200	0
$\hat{f}_3$	100	300	0	0
$\hat{f}_4$	100	200	100	0

- By *uncertainty aversion*,

$$f_2 \sim f_3 \sim \hat{f}_3 \implies \hat{f}_4 = f_1 = \frac{1}{2}f_2 + \frac{1}{2}\hat{f}_3 \succ f_2 \sim \hat{f}_3$$

$\leadsto$  the opposite of  $\langle \star \rangle$  —  $f_1 \prec f_2$  &  $f_3 \succ f_4$

## REFLECTION EXAMPLE VS. CAP



	R	B	G	P
$f_1$	100	200	100	0
$f_2$	100	100	200	0
$f_3$	0	200	100	100
$f_4$	0	100	200	100

$$\langle \star \rangle \quad \begin{aligned} f_1 &\prec f_2 \\ f_3 &\succ f_4 \end{aligned}$$

- Identify each prior w/  $(\mu_B, \mu_G) \in [0, \frac{1}{2}]^2 \rightsquigarrow (\mu_R, \mu_P) = (\frac{1}{2} - \mu_B, \frac{1}{2} - \mu_G)$
- $\mathbb{M} = \{ M(\beta, \gamma) \subseteq \Delta(\Omega) \mid (\beta, \gamma) \in [0, 1]^2 \}$
- $c: [0, 1]^2 \rightarrow \mathbb{R}_+$ : LSC,  $c(1, 1) = 0$ , strictly decreasing, symmetric

$$U_{\text{CAP}}(f_1) \leq U_{\text{CAP}}(f_2) = U_{\text{CAP}}(f_3) \geq U_{\text{CAP}}(f_4)$$

*strict inequality for "not too large" c*



## APPENDIX

A SPECIAL CASES

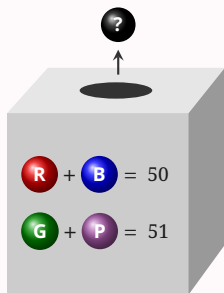
B MORE ON COMPARATIVES

C REFLECTION EXAMPLE

**D 50-51 EXAMPLE**

## 50-51 EXAMPLE

©MACHINA (2009)



	R	B	G	P
$f_5$	202	202	101	101
$f_6$	202	101	202	101
$f_7$	303	202	101	0
$f_8$	303	101	202	0

- Machina's conjecture:  $f_5 \succ f_6$  &  $f_7 \prec f_8$



- $f_6$  &  $f_8$  have slight "objective advantage"
- $f_5$  is not ambiguous, while  $f_7$  is ambiguous
- If you said  $f_6 \succ f_5$ , imagine 100 : 101 instead of 50 : 51... 🤔

## 50-51 EXAMPLE VS. DSEU

- Identify each prior  $\mu$  w/  $(\mu_B, \mu_G)$   $\rightsquigarrow (\mu_R, \mu_P) = (\frac{50}{101} - \mu_B, \frac{51}{101} - \mu_G)$
- For  $\cup M \subseteq [0, \frac{50}{101}] \times [0, \frac{51}{101}]$ ,

$$U_{\text{DSEU}}(f_5) = 151$$

$$U_{\text{DSEU}}(f_6) = 151 + 101 \max_{M \in \mathbb{M}} \min_{\mu \in M} (-\mu_B + \mu_G)$$

$$U_{\text{DSEU}}(f_7) = 150 + 101 \max_{M \in \mathbb{M}} \min_{\mu \in M} (-\mu_B + \mu_G)$$

$$U_{\text{DSEU}}(f_8) = 150 + 202 \max_{M \in \mathbb{M}} \min_{\mu \in M} (-\mu_B + \mu_G)$$

$$U_{\text{DSEU}}(f_5) - U_{\text{DSEU}}(f_6) = U_{\text{DSEU}}(f_7) - U_{\text{DSEU}}(f_8)$$

## 50-51 EXAMPLE VS. CAP

- $\mathbb{M} = \{ M(\beta, \gamma) \mid (\beta, \gamma) \in [0, 1]^2 \}$

$$M(\beta, \gamma) = \left[ (1 - \beta) \left\{ \frac{25}{101} \right\} + \beta \left[ 0, \frac{50}{101} \right] \right] \times \left[ (1 - \gamma) \left\{ \frac{25.5}{101} \right\} + \gamma \left[ 0, \frac{51}{101} \right] \right]$$

- $c: [0, 1]^2 \rightarrow \mathbb{R}_+$ : LSC,  $c(1, 1) = 0$ , & strictly decreasing
- For each  $\theta \in \mathbb{R}_{++}$ , the preference w/  $(\mathbb{M}, \theta c)$  satisfies

$$U_{\text{CAP}}(f_5) = 151 \qquad U_{\text{CAP}}(f_6) = 151.5 - \phi_\theta\left(\frac{1}{2}\right)$$

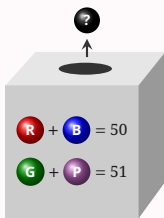
$$U_{\text{CAP}}(f_7) = 150.5 - \phi_\theta\left(\frac{1}{2}\right) \qquad U_{\text{CAP}}(f_8) = 151 - \phi_\theta(1)$$

$$\phi_\theta(t) = \min_{(\beta, \gamma) \in [0, 1]^2} [(50\beta + 51\gamma)t + \theta c(\beta, \gamma)] \qquad (\theta, t) \mapsto \phi_\theta(t) \text{ is concave}$$

## 50-51 EXAMPLE VS. CAP

- $U_{\text{CAP}}(f_5) - U_{\text{CAP}}(f_6) = -0.5 + \phi_{\theta}(\frac{1}{2})$  by concavity of  $\phi_{\theta}$
- $U_{\text{CAP}}(f_7) - U_{\text{CAP}}(f_8) = -0.5 + (\phi_{\theta}(1) - \phi_{\theta}(\frac{1}{2})) \leq U_{\text{CAP}}(f_5) - U_{\text{CAP}}(f_6)$
- $\exists(\underline{\theta}, \bar{\theta}) \in \mathbb{R}_{++}^2$  s.t.

$$\underline{\theta} < \bar{\theta}, \quad \theta > \underline{\theta} \iff f_5 \succ f_6, \quad \theta < \bar{\theta} \iff f_7 \prec f_8 \quad \Leftrightarrow$$



	R	B	G	P
$f_5$	202	202	101	101
$f_6$	202	101	202	101
$f_7$	303	202	101	0
$f_8$	303	101	202	0

$$\langle \spadesuit \rangle \quad \begin{aligned} f_5 &\succ f_6 \\ f_7 &\prec f_8 \end{aligned}$$