

Decision Theories for Some IPs – what dominance gains for us

Teddy Seidenfeld – CMU

I make free use of recent collaborations/discussions with:

Jay Kadane, Mark Schervish, Fabio Cozman, and Matthias Troffaes

Outline

- 1. Review of structural (not normative) assumptions for this overview.**
 - Moral hazards and dominance**
 - State-dependent utilities**
 - Normal versus extensive form decision making**
- 2. IP theory relating to de Finetti's *coherence of previsions*.**
 - Fundamental Theorem: imprecise vs. indeterminate* previsions**
 - Limitations of binary comparisons with *coherent* IP-sets**
 - Dominance and coherent previsions for unbounded variables**
- 3. IP theories related to relaxing canonical axioms of SEU theory**
 - Γ -Maximin, Maximality, and E-Admissibility***
 - On dominance and rationalizability-- D.Pearce's (1984) result.**

1. Structural assumptions used in this overview

- *Act-state independence*: no cases of “moral hazards” are considered – so strict dominance is valid.

Reminder: Consider the following binary state, two act decision problem, with outcomes ordinally (or cardinally) ranked so that more is better.

	ω_1	ω_2
Act ₁	3	1
Act ₂	4	2

Act₂ strictly dominates Act₁. Nonetheless, if

$$Prob(\omega_i | Act_i) \approx 1 \quad (i = 1, 2),$$

then dominance carries no force. A rational decision maker prefers Act₁ to Act₂.

In today's overview only decision problems without *moral hazards* are considered.

- **State-independent utility:** no cases where the value of a prize depends upon the state in which it is received.

Reminder: Once we entertain, generalized state-dependent utilities for prizes, there is maximal under-determination (= up to *mutual absolute continuity* of probability) of probability/utility pairs that represent the very same preference ranking of acts.

Matrix of m -many acts on the partition of n -many uncertain states

	ω_1	ω_2		ω_j			ω_n
Act ₁	o_{11}	o_{12}		o_{1j}			o_{1n}
Act ₂	o_{21}	o_{22}		o_{2j}			o_{2n}
Act _i	o_{i1}	o_{i2}		o_{ij}			o_{in}
Act _m	o_{m1}	o_{m2}		o_{mj}			o_{mn}

In accord with generalized SEU preference over acts, suppose

Act₁ is dispreferred to Act₂ if and only if $\sum_j P(\omega_j)U_j(o_{1j}) \leq \sum_j P(\omega_j)U_j(o_{2j})$

Choose P* mutually absolutely continuous with P and define the constants

$$c_j = P(\omega_j)/P^*(\omega_j)$$

and let

$$U^*_j(\bullet) = c_j U_j(\bullet) \quad (j = 1, \dots, n).$$

Then, trivially,

$\sum_j P(\omega_j)U_j(o_{1j}) \leq \sum_j P(\omega_j)U_j(o_{2j})$ if and only if $\sum_j P^*(\omega_j)U^*_j(o_{1j}) \leq \sum_j P^*(\omega_j)U^*_j(o_{2j})$.

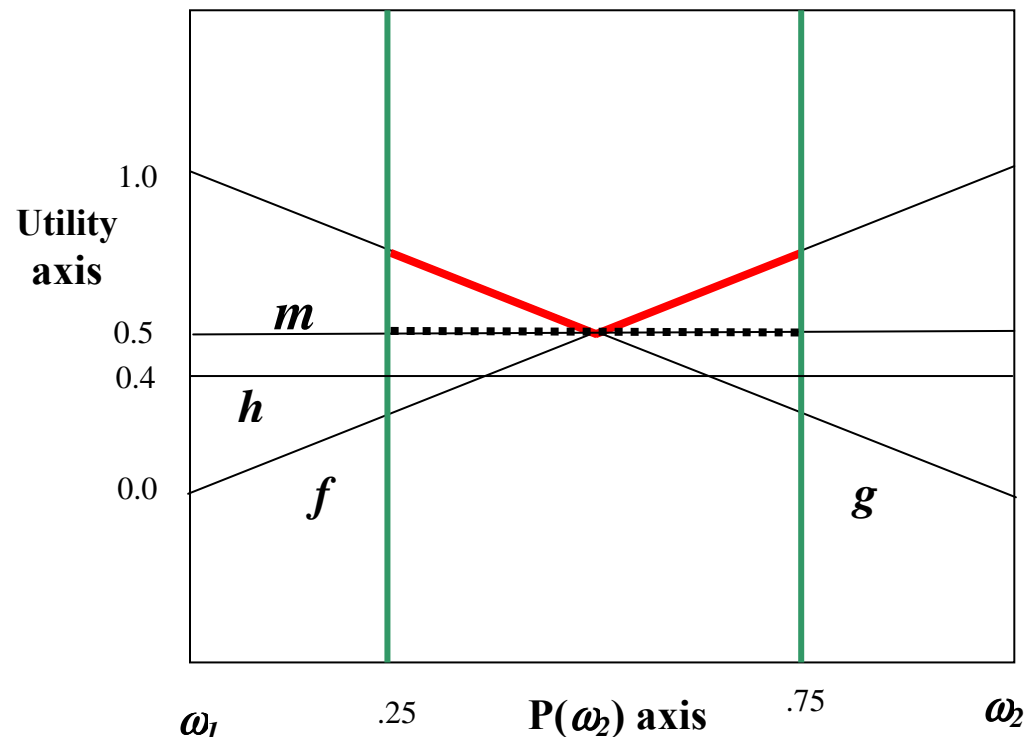
Note well: This problem arises even when one of the representations is by a state-independent utility!

- The focus here is on *normal* form (aka “strategic form”) games/decisions. This is where the decision maker can commit, in advance, to all contingency planning.

Reminder: We make no assumption that *normal* and *extensive* form decisions are equivalent, and generally they will not be equivalent for IP-decision theories.

- This issue is particularly important for so-called “dynamic Book” arguments

Dilation Example (in the spirit of Ellsberg’s Paradox)



Consider the following *normal* and *extensive* decision problems.

Normal (strategic) form:

- Choose among the mixed option m and constant acts h and 0 .

m is defined as f if *heads* and g if *tails* on a flip of a *fair* coin.

- All (?) decision rules recommend m uniquely from the trio $\{0, h, m\}$ in the normal form

Extensive (sequential) form:

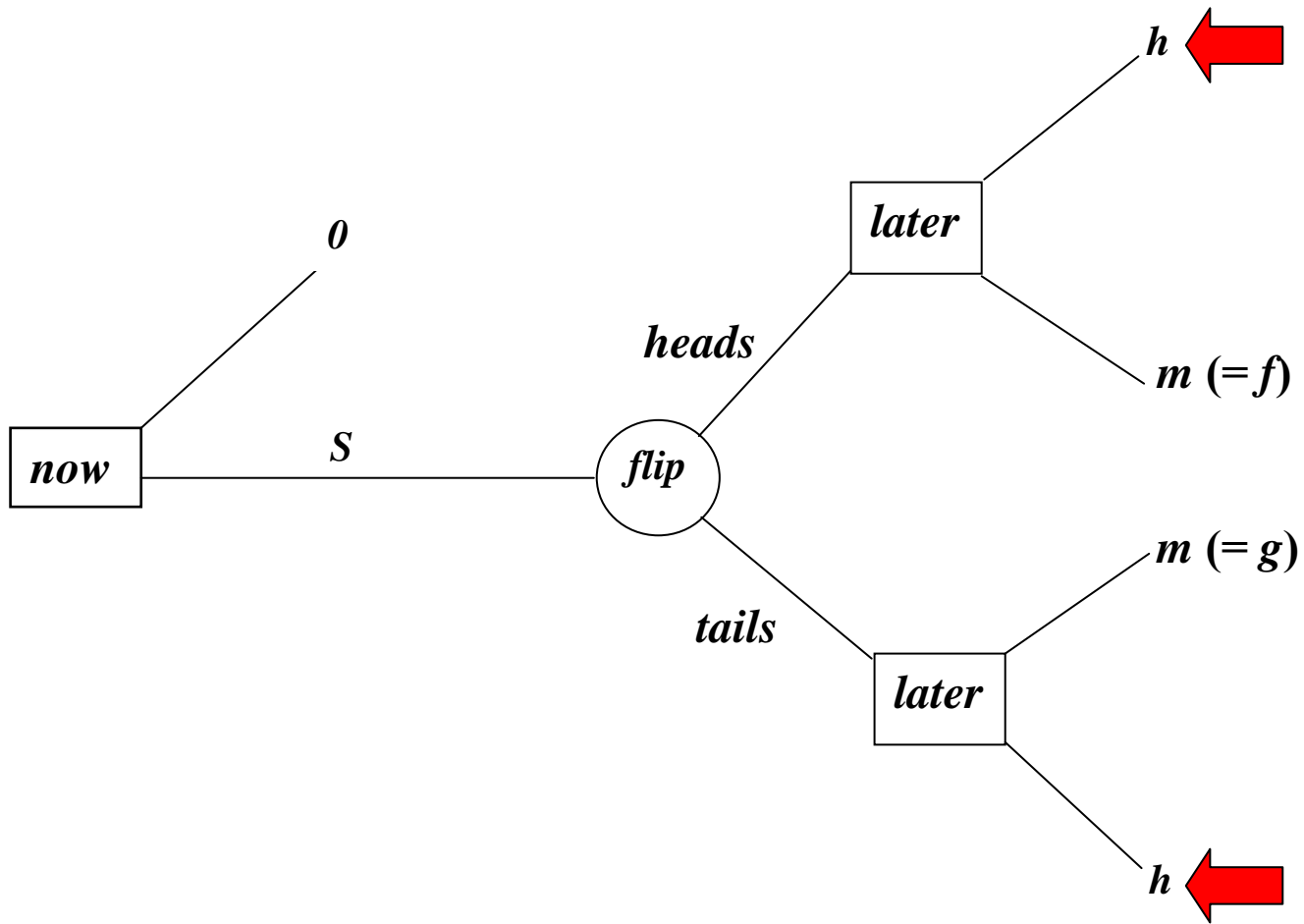
- Choose *now* between the constant act 0 and the sequential option S

With S you observe the fair coin flip and *later* choose between m and h .

But, later m is equivalent to f if heads, or equivalent to g if tails.

Many IP-decision rules will advise h in a pairwise choice either between h and f , or between h and g .

- Then, h (not m) is the outcome of the *extensive* form decision.



2. IP theory relating to de Finetti's *coherence of previsions*.

An important, and historically early application of strict dominance for IP-decision theory is based on de Finetti's criterion of *coherence of previsions*. Let us first review de Finetti's two related theorems.

Coherence – de Finetti's notion of coherence begins with
an arbitrary partition of states, $\Omega = \{\omega_i: i \in I\}$, and
an arbitrary collection of real random variables, $\chi = \{X_j: j \in J\}$, defined on Ω .

For each random variable $X \in \chi$, the rational agent has a (two-sided) *prevision* $P(X)$ which is to be interpreted as a *fair price* (both for *buying* and *selling*) as follows.

For all real $\beta > 0$, small enough so that the agent is willing to pay the possible losses,

the agent is willing to pay $\beta P(X)$ in order to buy (i.e., to receive) βX in return.

and, willing to accept $\beta P(X)$ in order to sell (i.e., to pay) βX in return.

In symbols, the agent will accept the gamble

$$\beta[X - P(X)]$$

as a change in fortune, for all sufficiently small (positive or negative) β .

The agent is required to accept all finite sums of gambles of the preceding form.

That is, for all finite n and all small, real β_1, \dots, β_n and all $X_1, \dots, X_n \in \mathcal{X}$, the agent will accept the combination of gambles

$$\sum_{i=1}^n \beta_i [X_i - P(X_i)].$$

Where β_i is positive, the agent buys β_i -units of X_i for a price of $\beta_i P(X_i)$

where it is negative, the agent sells β_i -units of X_i for a price of $\beta_i P(X_i)$.

The previsions are *incoherent* if there is a *uniformly negative* acceptable finite combination of gambles. That is, if there exists a sum of the form above and $\varepsilon > 0$ such that, for each $\omega \in \Omega$,

$$\sum_{i=1}^n \beta_i [X_i(\omega) - P(X_i)] < -\varepsilon.$$

Otherwise the agent's previsions are *coherent*.

Where previsions are incoherent, the book that indicates this constitutes a combination of gambles that is uniformly, strictly dominated by *not-betting* (= 0).

de Finetti's *Coherence Theorem*:

- **A set of previsions are coherent if and only if they are the expected values for the respective random variables under a (finitely additive) probability distribution over Ω .**
- **When the variables are indicator functions for events (subsets of Ω), coherent previsions are exactly those in agreement with a (finitely additive) probability. And then the $|\beta_i|$ are the stakes in winner-take-all bets, where the previsions fix betting rates, $P(X_i) : 1-P(X_i)$.**

- **de Finetti's *Fundamental Theorem of Previsions***

Suppose coherent (2-sided) previsions are given for all variables in a set χ defined with respect to Ω .

Let Y be a real-valued function defined on Ω but not in χ .

Define: $\underline{A} = \{X: X(\omega) \leq Y(\omega) \text{ and } X \text{ is in the linear span of } \chi\}$

$\bar{A} = \{X: X(\omega) \geq Y(\omega) \text{ and } X \text{ is in the linear span of } \chi\}$

Let

$$\underline{P}(Y) = \sup_{X \in \underline{A}} P(X) \quad \text{and} \quad \bar{P}(Y) = \inf_{X \in \bar{A}} P(X)$$

Then the 2-sided prevision, $P(Y)$, may be any finite number from $\underline{P}(Y)$ to $\bar{P}(Y)$ and the resulting enlarged set of previsions is coherent.

Outside this interval, the enlarged set of previsions is incoherent.

The *Fundamental Theorem* provides an early instance of IP-theory where, in I.Levi's (1980) terms relating to I.J.Good's "Black Box Theory, the 'I' stands for an imprecise prevision, rather than an indeterminate prevision.

That is,

The interval for a new prevision [$\underline{P}(Y)$ $\bar{P}(Y)$] given by the *Fundamental Theorem* constrains a new, 2-sided prevision for a variable, $Y \notin \chi$, while preserving coherence of the 2-sided previsions already assigned to $X \in \chi$.

Thus, coherence for 2-sided previsions does not require the rational agent to identify precise previsions beyond those in the linear span of the variables in the arbitrary set χ .

Specifically, the rational agent is not required by *coherence* to have determinate probabilities defined on an algebra of events, let alone on a power-set of events. It is sufficient to have probabilities defined as-needed for the arbitrary set χ , as might arise in a particular decision problem.

- See, e.g., F. Lad, 1996 for interesting applications of this result.

- *Toy Example 1.1* – good, realistic examples are given by Lad:
 $\Omega = \{1, 2, 3, 4, 5, 6\}$ the outcome of rolling an ordinary die.
 χ is the set of indicator functions for the following four events

$$\chi = \{ \{1\}, \{3,6\}, \{1,2,3\}, \{1,2,4\} \}$$

Suppose 2-sided previsions for these four events are given, and agree with the judgment that the die is “fair.”

$$P(\{1\}) = 1/6; \quad P(\{3,6\}) = 1/3; \quad P(\{1,2,3\}) = P(\{1,2,4\}) = 1/2.$$

The set of events for which a precise prevision is fixed by the 2-sided previsions for these four events is given by the *Fundamental Theorem*.

- That set does not form an algebra. Only 22 of 64 events have precise previsions.

For instance, by the Fundamental Theorem,

$$\underline{P}(\{6\}) = 0 < \bar{P}(\{6\}) = 1/3;$$

likewise

$$\underline{P}(\{4\}) = 0 < \bar{P}(\{4\}) = 1/3;$$

however,

$$P(\{4,6\}) = 1/3.$$

Moreover, the smallest algebra containing these 4 events is
the power set of all 64 events on Ω .

De Finetti's results apply also for conditional, 2-sided previsions, given an event F. These results obtain using called-off previsions of the form

$$I_F \beta[X - P(X)]$$

where I_F is the indicator function for the conditioning event F.

Then, with a proviso, coherence assures that coherent called off 2-sided previsions are (finitely additive) conditional expectations, given the conditioning event.

The restriction is that the conditioning event not be a null-event.

Otherwise, a non-Archimedean (non-real valued) theory of previsions results.

When the random variables, X_i , include indicator functions for events, the resulting coherent 2-sided previsions include conditional probabilities for these events.

- Note well: The called-off previsions correspond only to *normal form*, and not *extensive form* decisions. There is no *dynamical coherence* in de Finetti's theory – merely *static* aspects of coherence are covered by his 2 theorems. The previous observation about the non-equivalence of *normal* and extensive form IP-decisions reinforces this boundary on the scope of de Finetti's *coherence*.**
- Thus, de Finetti's theory of coherence does not require updating/learning by Bayesian conditional probabilities.**

In order to link de Finetti's *coherence* with IP-theory where, the 'I' stands for indeterminate previsions, we shift from 2-sided, to 1-sided previsions.

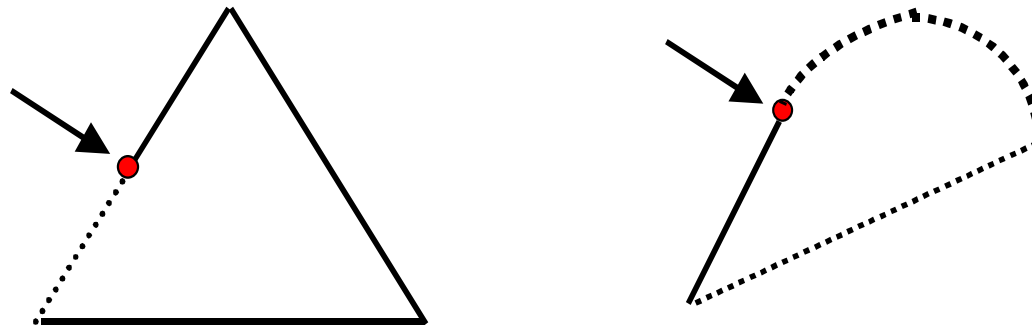
Then the decision maker is required only to provide a pair of (1-sided) previsions $\{\underline{P}(X_i), \bar{P}(X_i)\}$ for each random variable X_i in \mathcal{X} , corresponding to a largest "buy" price and smallest "sell" price for the corresponding 1-sided previsions, depending upon whether β_i is positive, or negative, respectively.

Generalizations of de Finetti's *Coherence Theorem* for 1-sided previsions have been given by many researchers. There are variations, e.g., that use only closed intervals of previsions, and those with mixed boundaries for their IP sets of probabilities.

Since these generalizations of de Finetti's *coherence* all rely on binary comparisons between gambles, identifying those which are favorable versus others, the resulting IP sets of probabilities that may be distinguished from one another are convex, with relatively simple boundaries – where extreme points of the convex set are also exposed.

- The common technique uses one or another *Separating Hyperplane Theorem*.

**Here are two convex sets with extreme points that are not exposed points.
(Dotted segments are open boundaries.)**



Generalizations of de Finetti's *coherence* that use 1-sided previsions, those that persist in using binary comparisons exclusively to determine admissible options, in principle cannot distinguish such sets from rivals that differ on their boundaries.

- In higher dimensions, the dimension of the boundary may be large too!**

Before concluding this selected review of aspects of de Finetti's *coherence* as that relates to decision making with IP sets, consider one more link between *strict dominance* and *coherence*, a link that pertains to de Finetti's insistence on allowing the *coherence* of (merely) finitely additive previsions.

For those who know my interests in finite additivity, it will come as a mild surprise that what comes next has little relation to non-conglomerability. It is *coherent* that for each h_i in a denumerable partition $\{h_1, h_2, \dots\}$, the conditional previsions satisfy:

$$c_1 \leq P(X | h_i) \leq c_2 \text{ yet } P(X) > c_2.$$

There is no contradiction with coherence, which requires that the decision maker responds to finitely many (called-off) previsions only at one time.

For what does come next, begin with an arbitrary set of real-valued, random variables, $\{X\}$, each defined on a common space, which we take as a σ -field of sets. It is enough to use the power set of a countable state space, $\Omega = \{\omega_i: i = 1, 2, \dots\}$. So there are no issues of non-measurable sets.

For each variable, X , the decision maker has a prevision, $P(X)$.

- When X is bounded, coherence entails that $P(X)$ is real-valued.**
- But when X is unbounded, $P(X)$ *may* be infinite, negative or positive.**

- When the prevision for X is a real-value, it is 2-sided, with the real-value payoff $c_X[X - P(X)]$, where c_X is a real number, either positive or negative. That is, the decision maker is committed to using $P(X)$ as the “fair price” when buying or selling the $|c_X|$ -multiples of the random quantity X .
- When the prevision for X is infinite-positive, i.e., when X has a value to the decision maker greater than any finite amount, then for each real constant k and for each $c_X > 0$, we require that the decision maker accepts (i.e., is committed to “buy”) a 1-sided payoff $c_X(X - k)$.
- Likewise, when the prevision for X is infinite-negative, with value less than any finite amount, then for each real constant k and for $c_X < 0$, we require the decision maker accepts (i.e., is committed to “sell”) a 1-sided payoff $c_X(X - k)$.
- In accord with deFinetti’s theory, the decision maker is required to accept a *finite sum* of such real-valued payoffs across (finitely many) variables.

Definition (as per de Finetti):

Previsions are coherent if there is no *finite* selection of non-zero constants, c_X , with the sum of the payoffs *uniformly* dominated by 0.

The previsions are incoherent otherwise.

The theory of coherent previsions is not confined to bounded random variables, as deFinetti notes (1974, Sections 3.12.4, 6.5.4-6.5.9).

However, in such cases, as previsions are not necessarily real-valued, they may induce a non-Archimedean ordering of random variables, as illustrated below with the St. Petersburg lottery.

Aside: The von Neumann-Morgenstern Archimedean axiom requires that whenever there is a chain of strict preferences among 3 lotteries, $X \prec Y \prec Z$, there are compound lotteries, with $0 < \alpha, \beta < 1$, satisfying:

$$\alpha X \oplus (1-\alpha)Z \prec Y \prec \beta X \oplus (1-\beta)Z.$$

***Define* The (weak) order \ll over random variables according to their coherent previsions by:**

***Definition:* $X \ll Y$ if and only if $P(Y - X) > 0$,**

with $X \equiv Y$ if and only if $P(Y - X) = P(X - Y) = 0$.

- Under rather general conditions, illustrated below, the class of coherent weak-orders « with coherent previsions for unbounded random variables must distinguish by strict preference « among a finite set of *equivalent* (\approx) random variables.

Two (real-valued) random variables are *equivalent* if they have the same probability distributions over *all* sets of outcomes.

- This situation is problematic for Expected Utility theory then, as strict preference is NOT a function of probability distributions over utilities!
It is a challenge to any IP decision theory for unbounded variables that seeks to represent coherent preference as a function of sets of expectations

There are two cases worth distinguishing in this analysis:

- **Non-Archimedean Previsions**
 - **Discontinuous Previsions.**
-
- ***Non-Archimedean Previsions:*** Let Z be an unbounded variable so that for each coherent prevision, and for each real number $r > 0$, either $P(Z) > r$ or $P(Z) < -r$. Then, coherent prevision for Z is infinite. An example of this is the familiar St. Petersburg variable, Z , where with probability 2^{-n} , $Z = 2^n$. Flip a fair coin until it lands heads first at flip n , when Z equals 2^n .
 - **A non-Archimedean ordering of random variables results,**

The von Neumann-Morgenstern Archimedean axiom requires that whenever there is a chain of strict preferences among 3 lotteries, $X \prec Y \prec Z$, there are compound lotteries, with real-valued $0 < \alpha, \beta < 1$, satisfying:

$$\alpha X \oplus (1-\alpha)Z \prec Y \prec \beta X \oplus (1-\beta)Z.$$

When $X = 1$, $Y = 2$, and $Z = St.P$, there is no real-valued $0 < \alpha < 1$ satisfying

$$\alpha X \oplus (1-\alpha)Z \prec Y = 2$$

- *Discontinuous, coherent previsions.*

Let $\{X_n\}$ ($n = 1, \dots$) be a sequence of non-negative random variables that converges (pointwise, *from below*) to the random variable X .

Definition: Say that previsions are continuous if $\lim_n P(X_n) = P(X)$.

- **When random variables are bounded, as is well known (deFinetti), coherent previsions are continuous *if and only if* they are the expectations of some countably additive probability.**
- **But when random variables are unbounded, they may fail to be continuous, even when probabilities are countably additive.**

Previsions that differ from their expected values:

Let Z be an unbounded, discrete random variable ($Z = 1, 2, \dots$).

Example: Z is Geometric(p): $\text{Prob}(Z = n) = p(1-p)^{n-1}$ ($n = 1, 2, \dots$).

- **Coherence assures that $P(Z) \geq E[Z]$ the expected value of Z .**

Let the prevision for Z , $P(Z)$, be greater than its expectation, $E[Z]$, where for the Geometric(p), $E[Z] = p^{-1}$. So, then we can write $P(Z) = E[Z] + b$, with $b > 0$.

- Let ' b ' denote the *boost* that the prevision receives in excess of its expectation.

Possibly, b is $+\infty$ in the example, though assume not in what follows.

BUT $P(Z) = E[Z] + b$ ($b \geq 0$) is a coherent prevision!

If b is finite, the prevision is a 2-sided coherent prevision in deFinetti's sense, since there can be no *sure loss* when this prevision is combined with coherent previsions for bounded variables.

- But this prevision is not *continuous* when $b > 0$.

For each n ($n = 1, 2, \dots$) define the bounded random variable Z_n by

$$Z_n = 0 \text{ for states when } Z > n, \text{ and } Z_n = Z \text{ otherwise.}$$

So, $P(Z) = E(Z)$, since there can be no "boost" for bounded variables.

The sequence $\{Z_n\}$ converges from below to Z .

Of course, since the probability is countably additive, $\lim_n E[Z_n] = E[Z]$.

However, $P(Z) = E[Z] + b$.

- So, when $b > 0$, previsions are *not continuous*.

Strict preference among equivalent random variables.

Theorem 1: Let X be a Geometric(p) random variable. Assume the prevision for X , $P(X)$, is finite but greater than its expectation ($E[X] = p^{-1}$), then there exist three equivalent random variables, W_1 , W_2 , and W_3 , each with finite prevision, such that no two have the same prevision.

***Example:* The following illustrates the theorem for the special case of fair-coin flipping, e.g., $p = 1/2$, $\text{Prob}(X = n) = 2^{-n}$, $n = 1, 2, \dots$, with $E[X] = 2$.**

Let $\{B, B^c\}$ be the outcome of an independent flip of another fair coin, so that $\text{Prob}(B, X = n) = 2^{-(n+1)}$ for $n = 1, 2, \dots$.

Define W_1 and W_2 by:

	ω_1	ω_2	ω_n
B	$X = 1$	$X = 2$		$X = n$	
	$W_1 = 2$	$W_1 = 3$		$W_1 = n+1$	
	$W_2 = 1$	$W_2 = 1$		$W_2 = 1$	
B^c	$X = 1$	$X = 2$		$X = n$	
	$W_1 = 1$	$W_1 = 1$		$W_1 = 1$	
	$W_2 = 2$	$W_2 = 3$		$W_2 = n+1$	

That is:

$$W_1(B, X = n) = n+1; \quad W_1(B^c, X = n) = 1 \quad (n = 1, 2, \dots)$$

and $W_2(B^c, X = n) = n+1; \quad W_2(B, X = n) = 1 \quad (n = 1, 2, \dots).$

- Obviously, W_1 and W_2 are equivalent.
- Each has a Geometric($1/2$) distribution;
- Hence, $X \approx W_1 \approx W_2$.
- However, $W_1 + W_2 - X = 2$.

Thus, $P(W_1 - X) + P(W_2 - X) = 0$ *if and only if*
 $P(W_1) = P(W_2) = P(X) = 2$,

when the prevision for a Geometric($1/2$) variable is its expectation, and previsions are continuous, with no *boost*: $b = 0$.

Non-Archimedean preferences for generalized St. Petersburg variables.

We turn next to non-Archimedean previsions that result from random variables whose coherent prevision is mandated to be infinite, as with the St. Petersburg lottery.

Consider the class of Geometric(p_m) distributions, $p_m = 1 - 2^{-m}$, $m = 2, 3, \dots$.

For each member of this class, we define a generalized St. Petersburg gamble, Z_m , and construct a set of 2^{m-1} equivalent random variables $X_1 \approx X_2 \approx \dots \approx X_{2^{m-1}}$ ($\approx Z_m$), so that coherent previsions, though infinite, obey:

Theorem 2:
$$\sum_{i=1}^{2^{m-1}} [P(X_i) - P(Z_m)] = 2^{m-1}. \quad (*)$$

That is, though each of X_i ($i = 1, \dots, 2^{m-1}$) and Z_m are pairwise equivalent variables, the coherent prevision of their differences cannot all be 0.

Example: We illustrate this result for a special case, the Geometric($\frac{1}{2}$). The construction parallels the illustration of *Theorem 1*.

The St. Petersburg gamble

- Let states ω_n ($n = 1, \dots$) carry the Geometric($\frac{1}{2}$) distribution,

$$\text{Prob}(\omega_n) = 2^{-n} \quad (n = 1, 2, \dots).$$

- Let $\{B, B^c\}$ be another fair-coin flip, independent of the states, ω_n .

As before, partition each ω_n into two equi-probable cells using $\{B, B^c\}$ and define three (equivalent) random variables, X , Y , and Z , as follows:

	ω_1	ω_2	ω_n
B	$Z = 2$	$Z = 4$		$Z = 2^n$	
	$X_1 = 4$	$X_1 = 8$		$X_1 = 2^{n+1}$	
	$X_2 = 2$	$X_2 = 2$		$X_2 = 2$	
B^c	$Z = 2$	$Z = 4$		$Z = 2^n$	
	$X_1 = 2$	$X_1 = 2$		$X_1 = 2$	
	$X_2 = 4$	$X_2 = 8$		$X_2 = 2^{n+1}$	

- the St. Petersburg random variable: $Z(\omega_n) = 2^n$, independent of B .
- the random variable $X_1(\omega_n \cap B) = 2^{n+1}$, and $X_1(\omega_n \cap B^c) = 2$.
- the random variable $X_2(\omega_n \cap B) = 2$, and $X_2(\omega_n \cap B^c) = 2^{n+1}$.

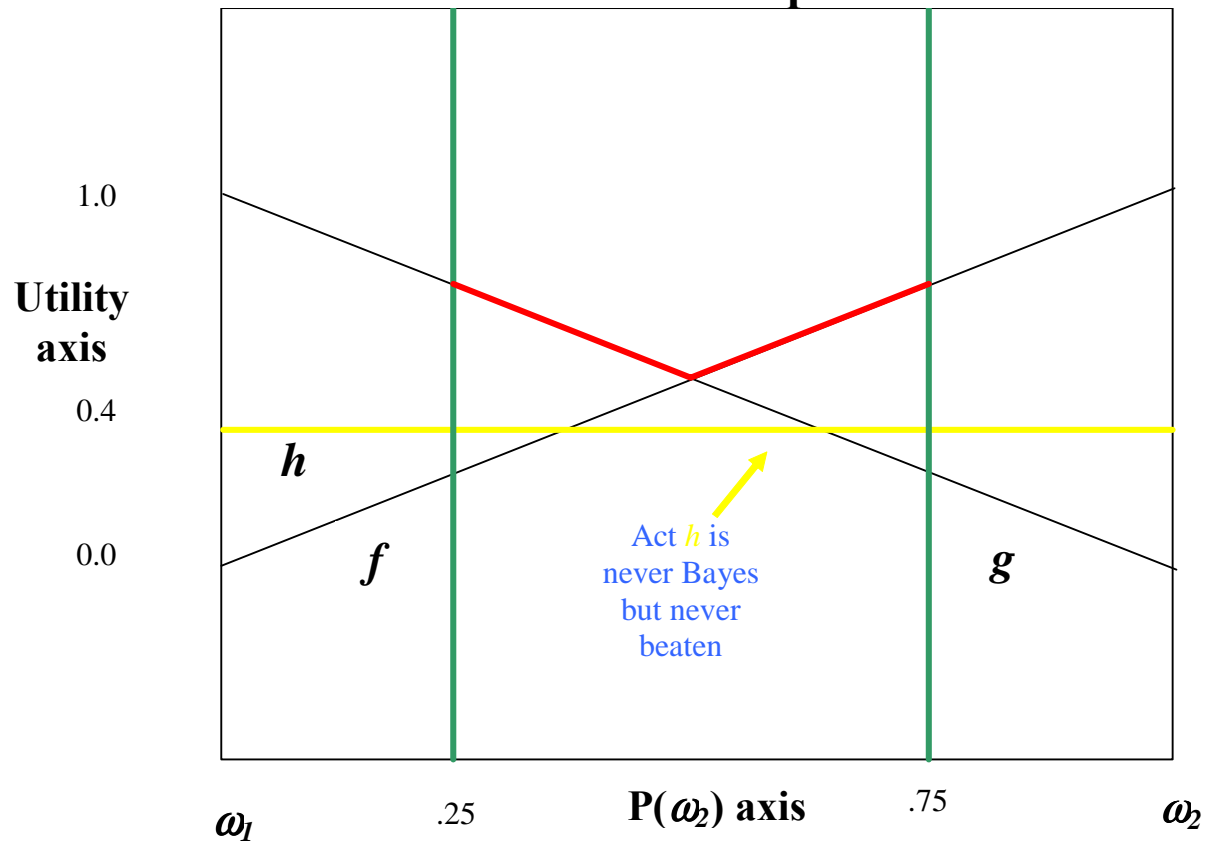
Though $X_1 \approx X_2 \approx Z$,
if previsions are coherent, $P(X_1 + X_2 - 2Z) = 2$.

This is in contradiction with the hypothesis that the prevision of the difference between equivalent random variables is 0, as that requires

$$P(X_1 - Z) + P(X_2 - Z) = P(X_1 + X_2 - 2Z) = 0.$$

- **Thus, de Finetti's *coherence* of previsions, respect for uniform, strict dominance over a (countable) partition, precludes preference according to SEU, when discontinuous or non-Archimedean previsions obtain and equivalent variables cannot be given equal previsions.**

**Return to question the relation between IP-decision theory and simple dominance
Reconsider this decision problem.**



**Only $\{f,g\}$ are Bayes-admissible from the triple $\{f,g,h\}$;
however, all pairs are Bayes-admissible in pairwise choices.
Levi calls h second worst in the triple $\{f,g,h\}$.**

Contrast three *coherent* decision rules for extending Expected Utility [EU] theory when probability – but not cardinal utility – is indeterminate.

The decision problems involve (bounded) sets of lotteries, where the outcomes have well-defined cardinal utility but where the (act-independent) states are uncertain, represented by a *convex* set of probabilities \mathcal{P} .

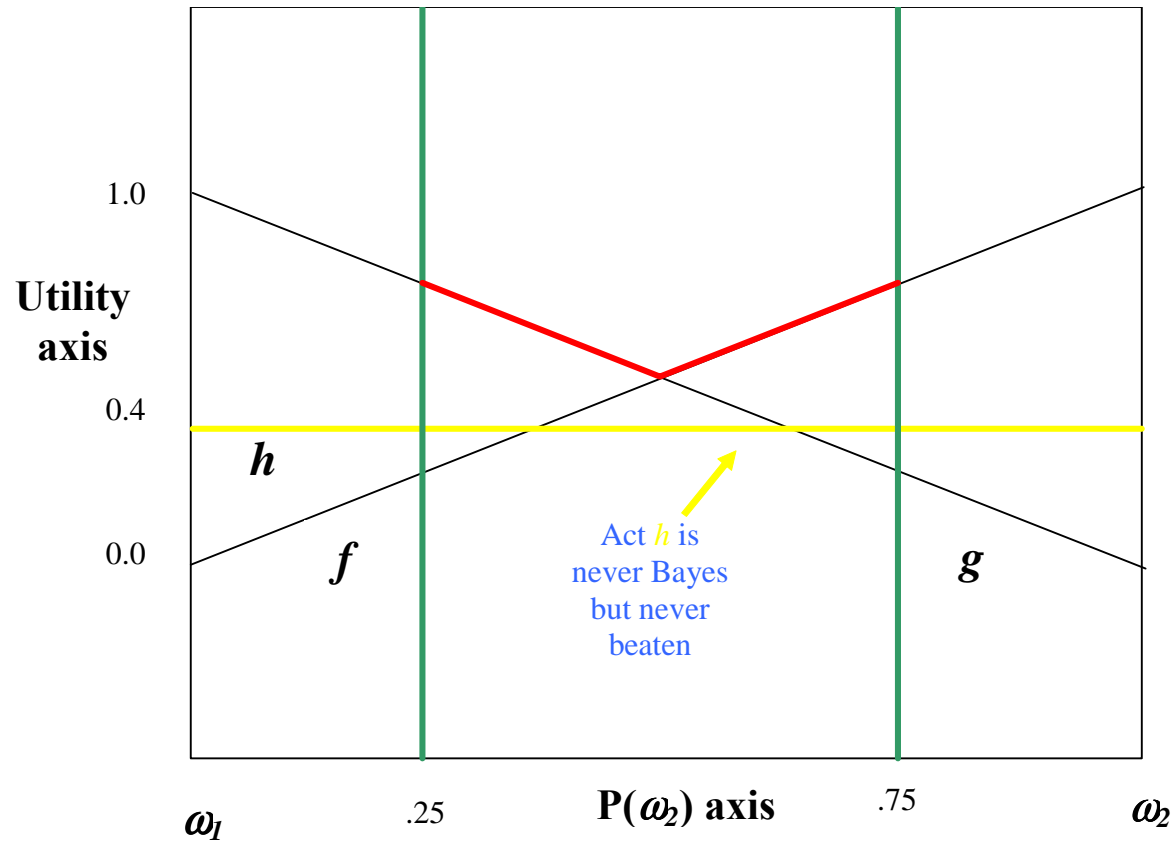
- ***G-Maximin* (Gilboa-Schmeidler) – maximize minimum expectations over \mathcal{P} .**
- ***Maximality* (Walley) – admissible choices are undominated in expectations over \mathcal{P} by any single alternative choice.**
- ***E-admissibility* (Levi/Savage) – admissible choices have Bayes' models, i.e., they maximize *EU* for some probability in the (convex) set \mathcal{P} .**

Each rule has *EU* Theory as a special case when probability is determinate, i.e., when \mathcal{P} is comprised by a single probability distribution.

And each rule is *coherent* in the sense that sure loss (*Book*) is not possible.

The three rules are chosen to reflect the following progression, where each rule relaxes more of the ordering assumption than does its predecessors:

- ***G-Maximin* produces a (real-valued) ordering of options; hence, defined by binary comparisons – but it fails *Independence*.**
- ***Maximality* does not generate an ordering of options; however, it is given by binary comparisons.**
- ***E-admissibility* does not generate an ordering, nor is it given by binary comparisons.**



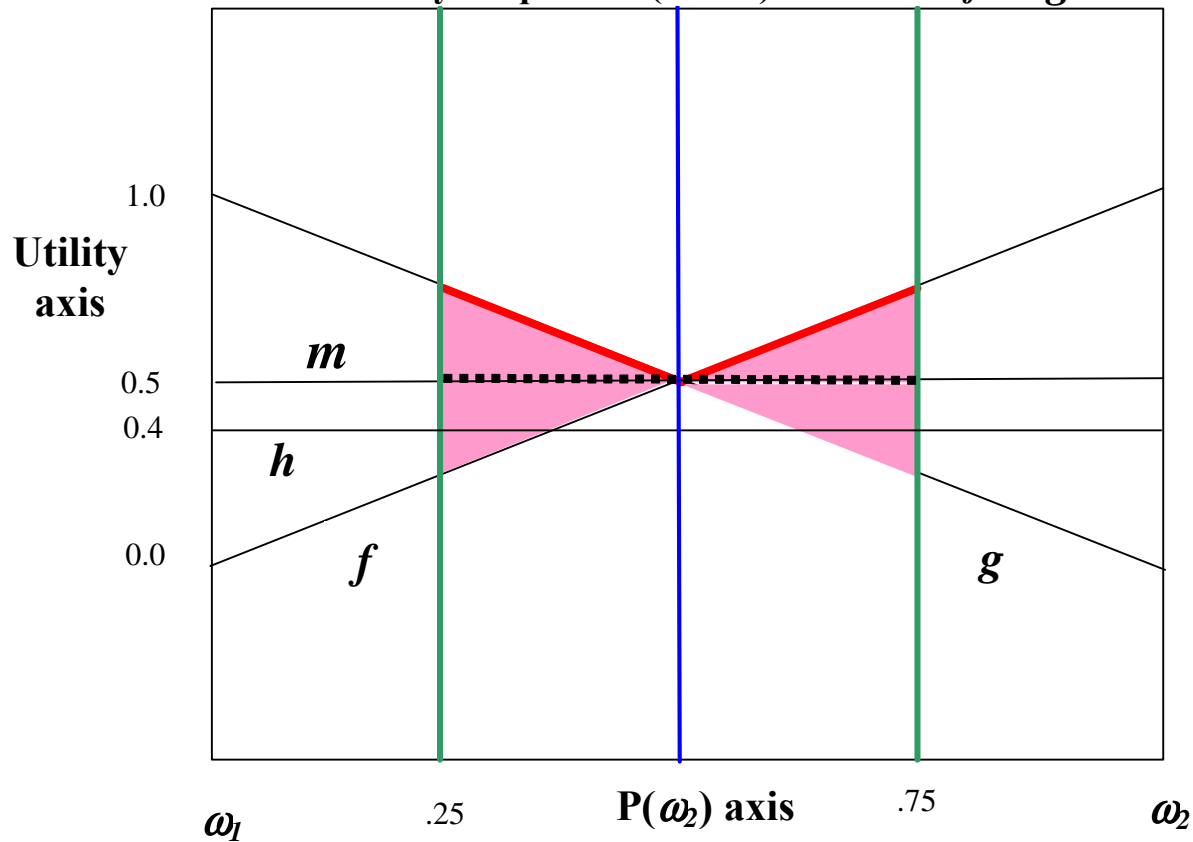
- **The Γ -Maximin solution is $\{h\}$.**
- **The E -admissible solution set is $\{f, g\}$.**
- **And Maximality finds all three options admissible, $\{f, g, h\}$.**

Thus, each rule gives a different set of admissible options in this problem.

Create a convex option space by allowing mixed strategies.

Expected Utility for the (Bayes) mixed options $\alpha f \oplus (1-\alpha)g$ is in pink; they maximize EU at $p(\omega_2) = .5$ (blue)

The Bayes equalizer (mixed) act is $m = .5f \oplus .5g$



- The *F-Maximin* solution is the EU-equalizer $\{m\}$.
- The *E-admissible* and *Maximally admissible* options are the same set of Bayes solutions (pink).

The agreement of the 3 decision rules on Bayes solutions is no accident as:

- ***Walley* (Theorem 3.9.5, 1990) establishes that when the option set is *convex* and the (convex) set of probabilities \mathcal{P} is *closed*, then *E-admissibility* and *Maximality* give the same solution sets:**

Their admissible sets are precisely the Bayes-admissible options.

- **And then it also follows that the *Γ -Maximin* admissible acts are a (proper) subset of the Bayes-admissible options.**

Under these conditions, pairwise comparisons of acts suffice to determine the set of Bayes-admissible choices.

D.Pearce (1984), reports a related result which is important for understanding the underlying connection between *dominance* and *Bayes-admissibility*.

***Theorem* (Pearce, 1984): In a decision problem under uncertainty,**

- with finitely many states and finitely generated option set O ,**
- with utility of outcomes determinate,**

if an option $o \in O$ fails to be Bayes-admissible,

then o is uniformly, strictly dominated by a finite mixture from O .

Aside: This result can be extended to infinite decision problems.

In this sense, incoherent choices suffer deFinetti's penalty – being uniformly strictly dominated by a mixed option – within the decision at hand – and not merely for previsions, which are specialized decisions.

**In accord with Pearce's Theorem, in the example above,
the mixed act $m = .5f \oplus .5g$ strictly dominates h .**

I conclude by illustrating how this insight relates to coherent choice in the setting of games, which is the domain where Pearce directs his analysis,

Game 1: *Iterative elimination of dominated strategies.*

Consider the following 2×3 game, with utility payoffs (*row, column*).

	<i>L</i>	<i>M</i>	<i>R</i>
<i>T</i>	(2,1)	(1,4)	(0,3)
<i>B</i>	(1,8)	(0,2)	(1,3)

There is no strict dominance between Row-player's two options.

But from Column-player's perspective, *R* is *never* a best reply.

For example, the mixed strategy $[.25L \oplus .75M]$ strictly dominates *R*.

So, if Row-player models Column player as one who plays Bayes-admissible options, from Row-player's perspective, state *R* may be eliminated. Then *T* dominates *B*. And if Column-player models Row-player also as playing Bayes-admissible options (and models Column-player likewise), then Column-player may eliminate state *B*. This results in the choices (*T, M*), which also is the unique Nash equilibrium of this game.

Game 2: An undominated option that is not part of a Nash equilibrium

Consider the following $2 \times 2 \times 4$, 3-player game: where player-1 chooses a Row, player-2 chooses a Column, and player 3 chooses a Matrix $\{A, B, C, D\}$.

The numbers are the utility payoffs to player 3.

	<i>L</i>	<i>R</i>
<i>T</i>	9	0
<i>B</i>	0	0

A

	<i>L</i>	<i>R</i>
<i>T</i>	0	9
<i>B</i>	9	0

B

	<i>L</i>	<i>R</i>
<i>T</i>	0	0
<i>B</i>	0	9

C

	<i>L</i>	<i>R</i>
<i>T</i>	6	0
<i>B</i>	0	6

D

Matrix *D* is undominated by any mixture of $\{A, B, C\}$; however, it is not part of a Nash equilibrium. This is because all of *D*'s Bayes-models live off the surface of independence between Row “states” and Column “states,” which is where all Nash-equilibria strategies must reside! *D* is not a *best reply* to a probability on $R \times C$ that makes *row-states* independent of *column-states*. Why does rationality in games require such independence? (It is not a matter of act/state independence.)

- When we apply IP-decision making to non-cooperative games, then *coherence* – avoiding dominated options – does not result in an endorsement of Nash’s criterion of equilibria! What results is the theory of *Rationalizable* strategies.

Summary

- 1. Review of structural (not normative) assumptions for this overview.**
 - Moral hazards and dominance**
 - State-dependent utilities**
 - Normal versus extensive form decision making**
- 2. IP theory relating to de Finetti's *coherence of previsions*.**
 - Fundamental Theorem: imprecise vs. indeterminate* previsions**
 - Limitations of binary comparisons with *coherent* IP-sets**
 - Dominance and coherent previsions for unbounded variables**
- 3. IP theories related to relaxing canonical axioms of SEU theory**
 - Γ -Maximin, Maximality, and E-Admissibility***
 - On dominance and rationalizability-- D.Pearce's (1984) result.**