

# *Some Finitely Additive Probabilities and Decisions*

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***Part 1: Some (Purely) Finitely Additive Probabilities for Logic***

***Using analysis by M. Amer (1985a,b)***

**1.1 Structural assumptions for finitely additive probabilities.**

**A finitely additive measure space,  $\langle \Omega, \mathcal{B}, P \rangle$ ,**

**where  $P$  is a finitely additive probability (see below)**

**defined on a field of sets  $\mathcal{B}$ ,**

**with the sure-event  $\Omega$  equal to a set of disjoint and mutually exhaustive possibilities, called “states”:  $\Omega = \{\omega_i: i \in I\}$ .**

**A set  $B$  is *measurable* if  $B \in \mathcal{B}$ .**

**A numerical probability  $P$  is finitely additive [f.a.] if it satisfies:**

**(i) For each measurable set  $B$ ,  $0 \leq P(B) \leq 1$ .**

**(ii)  $P(\Omega) = 1$ .**

**(iii) If  $E$  and  $F$  are disjoint measurable sets, with  $G = E \cup F$ , then**

$$P(G) = P(E) + P(F).$$

For probability  $P$  to be countably additive is to require also either of the following two axioms, which, given  $P$  is f.a., are equivalent provided that  $\mathcal{B}$  is at least a field of sets. [See Billingsley, §2.]

(iv-a) Let  $\{A_i : i = 1, \dots\}$  be a sequence of measurable, pairwise disjoint sets, and their union,  $A$ , is measurable.

That is, assume  $A_i \cap A_j = \emptyset$  if  $i \neq j$ , and  $A = \bigcup_i A_i \in \mathcal{B}$ .

$P$  is countably additive<sub>1</sub> [c.a.<sub>1</sub>] if  $P(A) = \sum_i P(A_i)$  for each such sequence.

(iv-b) Let  $\{B_i : i = 1, \dots\}$  be a decreasing (or, respectively increasing) sequence of measurable sets, and their limit,  $B$ , is measurable.

That is,  $B_i \supseteq B_{i+1}$ , and assume  $B = \bigcap_i B_i$

(or respectively  $B_i \subseteq B_{i+1}$  and  $B = \bigcup_i B_i$  is measurable).

$P$  is countably additive<sub>2</sub> [c.a.<sub>2</sub>] if  $P(B) = \lim_i P(B_i)$  for each such sequence.

- Countable additivity<sub>2</sub> is a continuity condition for probability.

If  $P$  is finitely but not countably additive, it is *merely* finitely additive.

An extreme version of mere f.a. probability is a purely f.a. probability:

A f.a. probability  $P$  is called a purely finitely additive if, for each  $\varepsilon > 0$ , there exists a (measurable) denumerable partition of the sure event

$\Pi = \{A_i: i = 1, \dots\}$ , such that  $\sum_i P(A_i) < \varepsilon$ .

And  $P$  is called strongly f.a. if  $\sum_i P(A_i) = 0$ .

## **1.2 Probabilities for infinite free Boolean algebras.**

**Amer's (1985*b*) Theorem 5 reports the following result:**

**Proposition 1: Let  $\mathcal{A}$  be an infinite, free Boolean algebra.**

**There are no countably additive probabilities on  $\mathcal{A}$ .**

**Amer proves that each infinite, free Boolean algebra embeds mandatory discontinuities in the sense of continuity required by c.a.<sub>2</sub>.**

***Corollary:* Each f.a. probability on such a Boolean algebra is purely finitely additive. (See the appendix to these slides for details.)**

***Example:*** The Lindenbaum-Tarski algebra  $\mathcal{L}$  for sentential logic (with top  $T$  and bottom  $\perp$ ) is a countable, free Boolean algebra with the denumerable set of sentence letters  $\{S_i: i = 1, \dots\}$  serving as a set of free generators.

Let  $\varphi_i$  be either one of  $S_i$  or  $\neg S_i$ .

Regardless of which is chosen, the algebra compels:  $\bigvee \{\varphi_i\} = T$  !!

Then,  $\mathcal{L}$  supports no countably additive probability.

But  $\mathcal{L}$  supports uncountably many merely f.a. probabilities.

**Additional clarificatory remarks about this example.**

**See Theorem 5 of Amer (1985a).**

**There are  $2^{\aleph_0}$  different, 2-valued, semantic models  $M$  for  $\mathcal{L}$ .**

**A model  $M$  provides a truth value (T or F) for each sentence letter  $S_i$ .**

**Each model  $M$  provides a 2-valued probability over all of  $\mathcal{L}$ :**

**For each  $\lambda \in \mathcal{L}$ ,  $P_M(\lambda) = 1$  if  $M(\lambda) = T$ , and**

**$P_M(\lambda) = 0$  if  $M(\lambda) = F$ .**

**Consider the sequence of  $\varphi_i$  where each is F under model  $M$ .**

**Then, for  $k = 1, \dots$ ,  $P_M(\varphi_1 \vee \dots \vee \varphi_k) = 0$ , but  $P_M(\bigvee \{\varphi_i\}) = P_M(T) = 1$ , which violates c.a.2.**

**Thus, there are uncountably many (strongly) f.a. probabilities on  $\mathcal{L}$ .**

***Summary of Part 1:* Some Boolean algebras compel mere f.a. by mandating failures of continuity in the sense of c.a.2**

***Part 2:      Some Finitely Additive (Statistical) Decisions***

**or**

***How Bruno de Finetti might have channeled Abraham Wald***

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**Based on our (2019) *What Finite-Additivity Can Add to Decision Theory***

## ***Organization of Part 2 of this presentation.***

### **2.1 Three dominance principles and finitely additive expectations – in increasing strength:**

***Uniform (bounded-away) dominance***

***Simple dominance***

***Admissibility (aka Strict Dominance)***

### **2.2 Finitely additive mixed strategies and Wald's (statistical) *Loss* functions.**

- **An example involving a discontinuous, strictly proper scoring rule.**

**Some results – assuming that *Loss* is bounded below:**

**Existence of a *Minimal, Complete Class of Bayes Decisions***

**Existence of a *Minimax Strategy* and a *Worst-case prior***

**Uniform dominance of never-Bayes decisions for bounded loss  
– *generalized Rationalizability***

### **2.3 But, not all *priors* have Bayes-decisions (!)**

## 2.1 Three dominance principles, in increasing order of logical strength

Fix a partition  $\pi = \{\omega_1, \dots, \omega_n, \dots\}$ , which might be infinite.

An *Act* is a function from  $\pi$  to a set of outcomes  $O$ .

Assume that outcomes may be compared by preference, at least within each  $\omega$ .

	$\omega_1$	$\omega_2$	$\omega_3$	...	$\omega_n$	...
<i>Act</i> <sub>1</sub>	$o_{1,1}$	$o_{1,2}$	$o_{1,3}$	...	$o_{1,n}$	...
<i>Act</i> <sub>2</sub>	$o_{2,1}$	$o_{2,2}$	$o_{2,3}$	...	$o_{2,n}$	...

### Uniform dominance:

For each  $\omega_i$  in  $\pi$ , outcome  $o_{2,i}$  is strictly preferred to  $o_{1,i}$  by at least  $\varepsilon > 0$ .

### Simple dominance:

For each  $\omega_i$  in  $\pi$ , outcome  $o_{2,i}$  is strictly preferred to outcome  $o_{1,i}$ .

### Admissibility (Wald, 1950) – Strict dominance (Shimony, 1955):

For each $\omega_i$	$o_{2,i}$ is weakly preferred to $o_{1,i}$
and for some $\omega_j$	$o_{2,j}$ is strictly preferred to $o_{1,j}$ .

Then, by *dominance* applied with partition in  $\pi$ ,: *Act*<sub>2</sub> is *strictly preferred* to *Act*<sub>1</sub>.

de Finetti (1974):

A class  $\{X\}$  of real-valued variables defined on a privileged partition of *states*,  $\Omega$ .

Let  $P$  be a (f.a) probability on  $\Omega$ .

Denote by  $\mathcal{E}_P(X)$  the (f.a.) expected value of variable  $X$  with respect to  $\Omega$ .

*Preference* between pairs of variables based on finitely additive expectation:

2. obeys *Uniform Dominance* in  $\Omega$

3. but may fail *Simple Dominance* in  $\Omega$ .

*Example<sub>1</sub>* – Let  $\Omega$  be countably infinite  $\Omega = \{\omega_1, \omega_2, \dots\}$ .

Consider variables  $X(\omega_n) = -1/n$ , and the constant  $Z(\omega_n) = 0$ .

Let  $P$  be a (strongly) finitely additive probability  $P(\{\omega\}) = 0$ .

Then  $\mathcal{E}_P(X) = 0 = \mathcal{E}_P(Z)$ , so indifference between  $X$  and  $Z$ .

But  $Z$  simply dominates  $X$ .

	$\omega_1$	$\omega_2$	$\omega_3$	...	$\omega_n$	...
$X$	-1	-1/2	-1/3	...	-1/n	...
$Z$	0	0	0	...	0	...

***Finitely additive mixed strategies: Making lemonade from lemons.***

**Example<sub>2</sub>:** Decision making under certainty:  $\Omega = \{\omega\}$ .

Consider the half-open interval of constant rewards,  $\mathcal{X} = \{X: 0 \leq X < 1\}$ .

Each *pure strategy*  $X$  is (uniformly) dominated.

Likewise, each countably additive *mixed strategy*  $P^\sigma$  over  $\mathcal{X}$  has expectation  $< 1$ .

But let be  $P$  a f.a. *mixed strategy* over  $\mathcal{X}$  where, for each  $\varepsilon > 0$ ,  $P[X > 1-\varepsilon] = 1$ .

- Then,  $\mathcal{E}_P(X) = 1$ .

In f.a. jargon,  $P$  agglutinates  $X$  at the (missing) value 1.

## ***2.2 Elementary Statistical Decision Theory in the fashion of A.Wald.***

- **An agent has a set  $\mathcal{A}$  of available (*pure strategy*) actions, and there is uncertainty over a set  $\Theta$  of *parameters* or *states of Nature*.  $\Theta$  forms a privileged partition.**
- **The agent suffers *loss*  $L(\theta; a)$  if she chooses  $a$  and  $\theta$  is the *state of Nature*.**
- **Sometimes the agent is allowed to choose action  $a$  using a probability measure (a mixed strategy)  $\delta$  over  $\mathcal{A}$ , and (when there are no data) we replace *loss*  $L(\theta; \cdot)$  by the *risk*  $R(\theta; \delta) = \int_{\mathcal{A}} L(\theta; a) \delta(da)$ .**

***Aside:*** As usual, the probability measure  $\delta_a(A) = \mathbf{I}_A(a)$  for every  $A \subseteq \mathcal{A}$  is equivalent to the pure strategy  $a$ .

**The agent wants to choose  $\delta$  to minimize *Risk*: respect *dominance* in  $\Theta$ .**

**A.Wald (1950): Respect *Admissibility* for *Risk* in  $\Theta$ .**

***Example 3a Brier Score for two complementary events.***

$\Omega = \{B, B^c\}^2$  where  $B$  is also the indicator function  $\mathbf{I}_B$  for some event  $B$ .

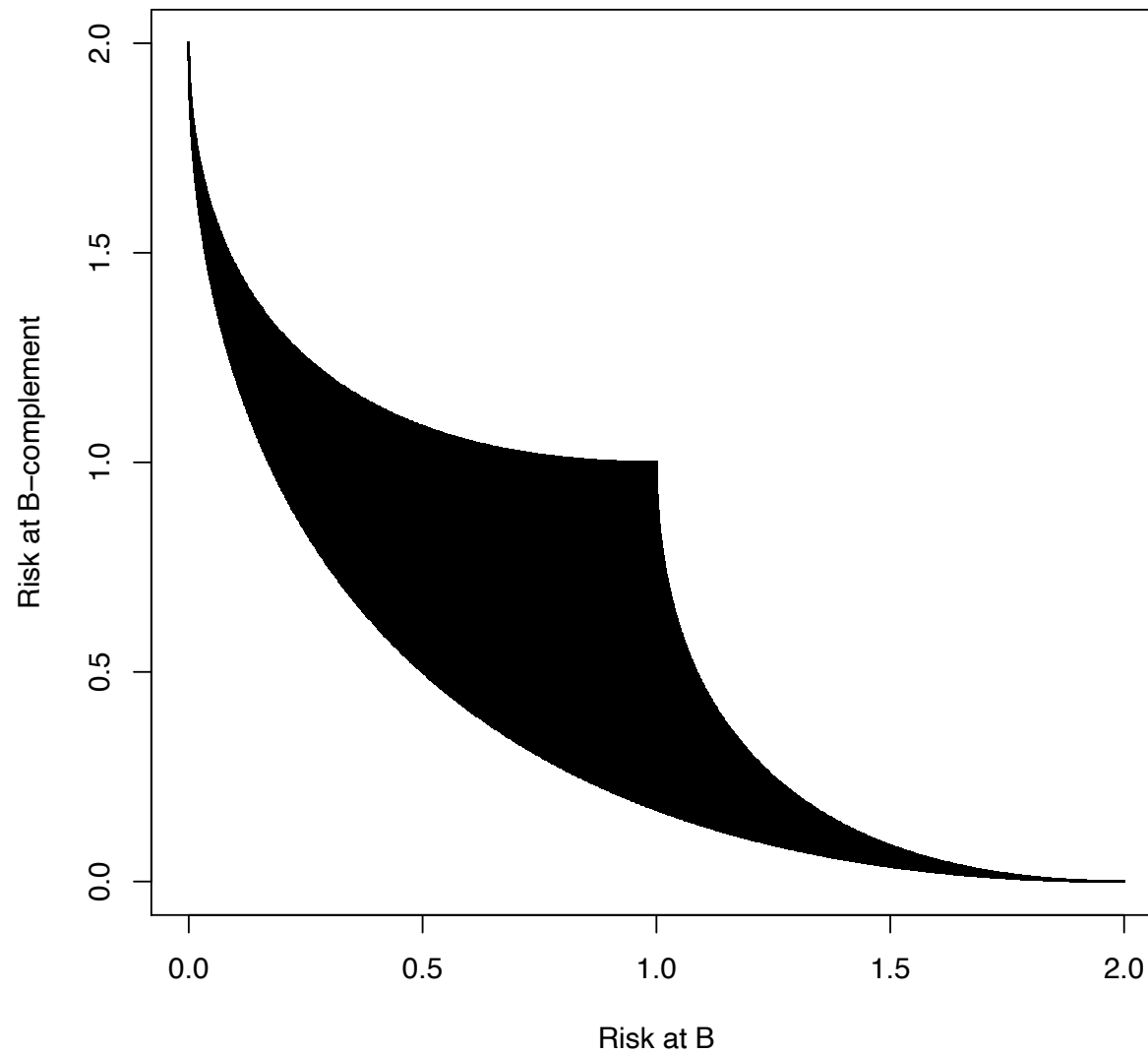
$\mathcal{A} = [0, 1]^2$ . There are no data.

$$L(\theta; (a_1, a_2)) = (\mathbf{I}_B - a_1)^2 + (\mathbf{I}_{B^c} - a_2)^2$$

- The only *admissible* actions are  $\{(a_1, a_2): a_1 + a_2 = 1\}$ , which correspond to the lower boundary of the Risk set – see next slide.
- Brier Score is a *strictly proper* scoring rule.

The Bayesian agent minimizes expected score uniquely by announcing her degrees of belief for  $(B, B^c)$ :  $a_1 = \text{Prob}(B)$  and  $a_2 = \text{Prob}(B^c)$

### Risks of Pure Strategies



**Example 3b: A discontinuous Brier Score.**

$\Omega = \{B, B^c\}^2$  where  $B$  is also the indicator function  $I_B$  for some event  $B$ .

$\mathcal{A} = [0, 1]^2$ . Again, there are no data.

$$\begin{aligned}
 L(\theta; (a_1, a_2)) &= (I_B - a_1)^2 + (I_{B^c} - a_2)^2 \\
 &\quad ( I_{[0, .5]}(a_1) + I_{(.5, 1]}(a_2) ) \quad \text{if } \theta = B \\
 &\quad + (1/2) \times \\
 &\quad ( I_{(.5, 1]}(a_1) + I_{[0, .5]}(a_2) ) \quad \text{if } \theta = B^c
 \end{aligned}$$

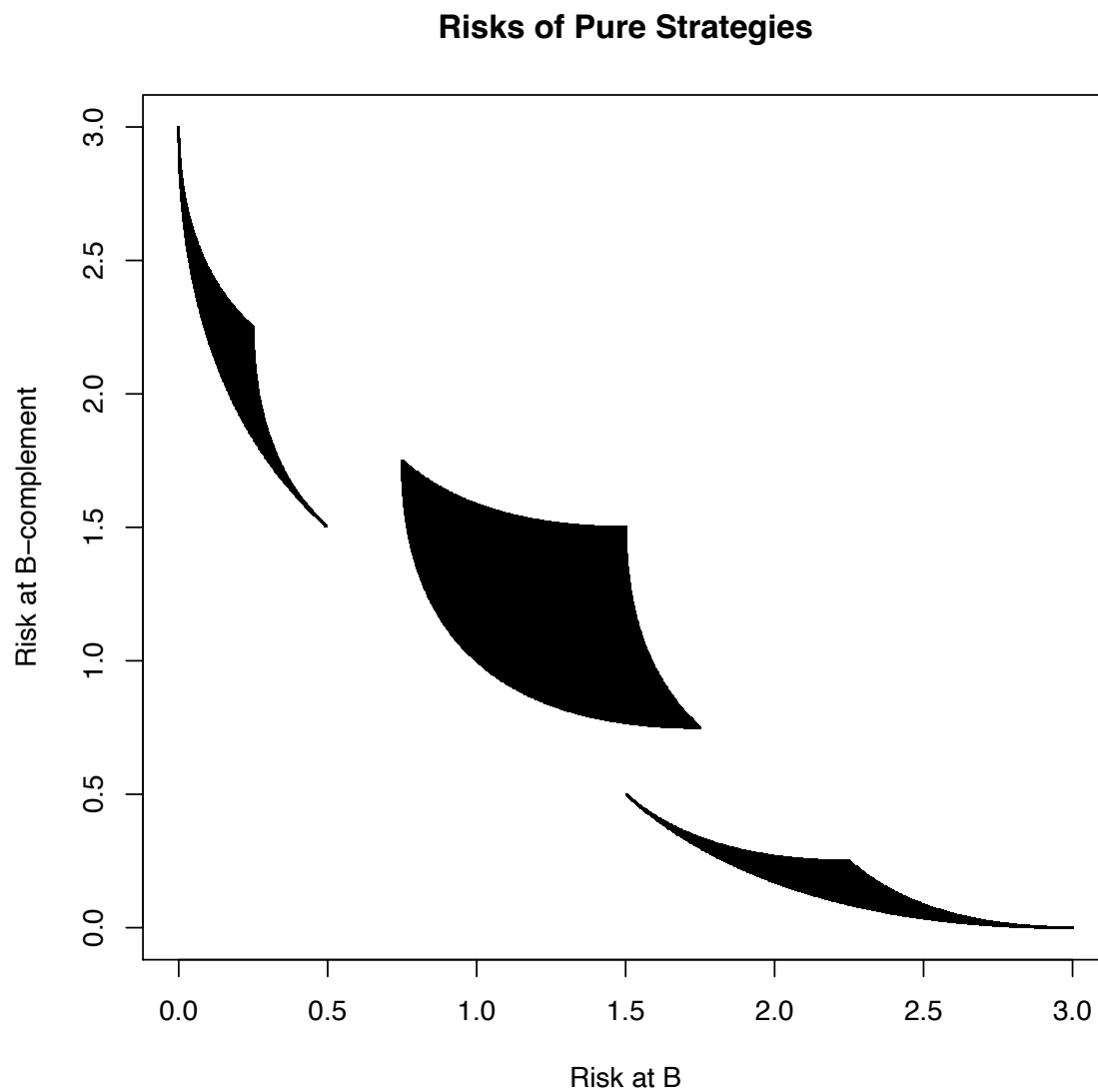
This *Loss* carries an added penalty when the forecast is on the wrong side of  $1/2$ .

- The only *admissible* actions are  $\{(a_1, a_2): a_1 + a_2 = 1\}$ .
- This discontinuous Brier Score is a *strictly proper* scoring rule.

The Bayesian agent (uniquely) minimizes expected score by announcing her degrees of belief for  $(B, B^c)$ :  $a_1 = \text{Prob}(B)$  and  $a_2 = \text{Prob}(B^c)$

*but ...*

$L(\theta; (a_1, a_2))$  is a point in a two-dimensional set  $[0, 3]^2$ .



**Recall:** The *admissible* options are on the lower boundary.

The shaded risk set has the properties that for pairs  $(p, 1-p)$ :

*Top*                      From  $(0, 3)$  down to but not including  $(.5, 1.5)$  are the points on the lower boundary, which correspond to  $0 \leq p < .5$ .

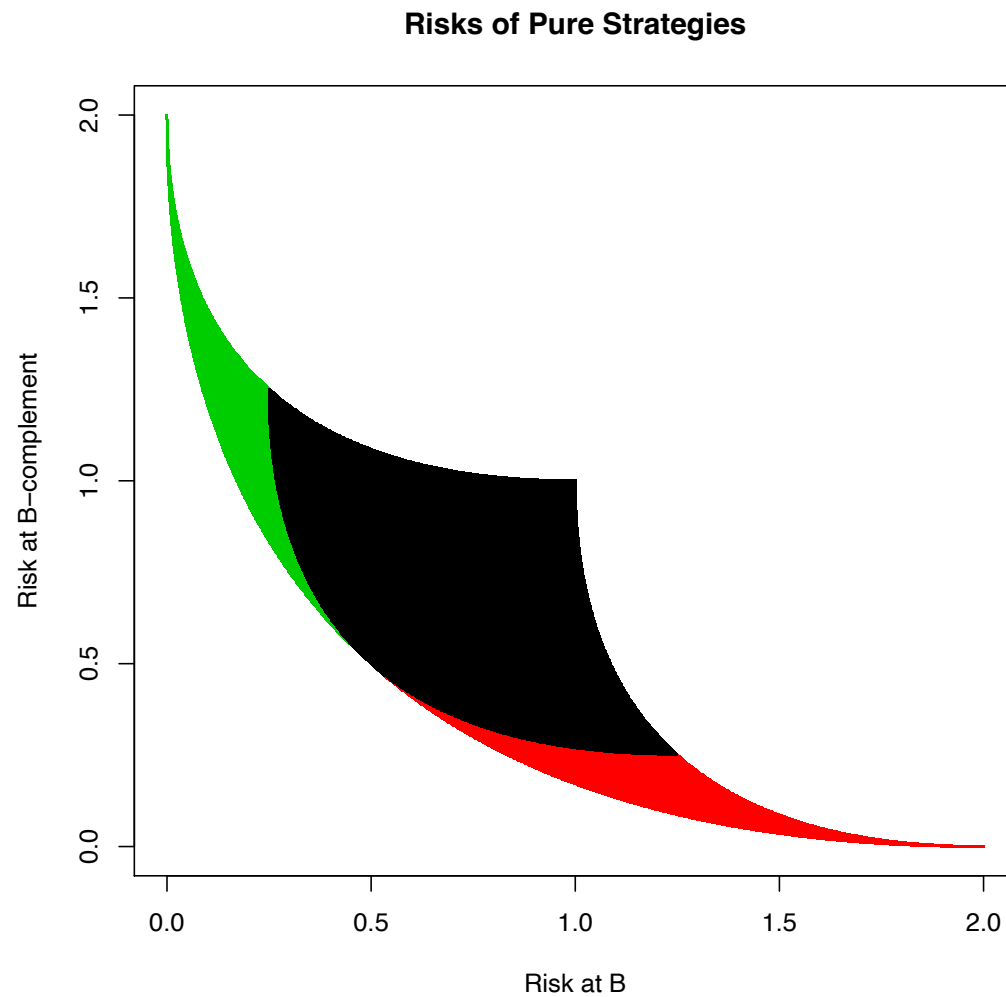
*Middle*                      In the middle section, only the point  $(1, 1)$  is on the lower boundary, corresponding to  $p = .5$ .

*Bottom*                      From (but not including)  $(1.5, .5)$  to  $(3, 0)$  are the points on the lower boundary, which correspond to  $.5 < p \leq 1$ .

So, points in the middle section (other than  $(1,1)$  ) are *inadmissible* though some are not dominated by  $(1,1)$ .

But those are dominated too, but only by other *inadmissible* options.

**The discontinuous (strictly proper) Brier Score carves up the continuous Brier Score.**



**Some decision-theory results in the fashion of Wald (1950)**

***Definitions:***

**Call a subclass  $C \subseteq \mathcal{A}$  of available decisions Complete**

**if for each decision  $\delta \notin C$  there is  $\delta_0 \in C$  where  $\delta_0$  *dominates*  $\delta$   
in the sense of *admissibility*.**

**Call a subclass  $C \subseteq \mathcal{A}$  of available decisions Minimally Complete**

**if  $C$  is complete and no proper subset of  $C$  is Complete.**

- **If there exists a *Minimally Complete* class it consists of the *admissible* decisions.**
- **In *Example 3<sub>b</sub>* (discontinuous Brier), there is no *Minimally Complete* class.**

**And using countably additive mixed strategies does not help this way.**

**BUT, augment the decision space by allowing (merely) f.a. mixed strategies.**

**Then, these f.a. mixed strategies fill in the missing lower boundary for *Risk*.**

**For example, consider f.a. mixed strategies  $\delta_1$  and  $\delta_2$  with the features that**

$$\forall \varepsilon > 0, \quad \delta_1\{a_1: .5 - \varepsilon < a_1 < .5\} = 1$$

**and**  $\delta_2\{a_1: .5 < a_1 < .5 + \varepsilon\} = 1,$

**and**  $\text{where } a_2 = 1 - a_1.$

**Then**  $R(\theta; \delta_1) = (.5, 1.5) \quad \text{and} \quad R(\theta; \delta_2) = (1.5, .5)$

***Aside:* As  $R(\theta; (.5, .5)) = (1, 1)$ , the 3 risk points  $R(\theta; \delta_1)$ ,  $R(\theta; (.5, .5))$  and  $R(\theta; \delta_2)$  are colinear.**

Some results – assuming that *Loss* is bounded below:

Strategies are (f.a.) *mixed strategies*.

Strategy  $\delta$  is admissible if there is no strategy  $\delta'$  such that

$$(\forall \theta) \quad R(\theta; \delta) \geq R(\theta; \delta') \quad \text{and} \quad (\exists \theta) \quad R(\theta; \delta) > R(\theta; \delta').$$

Strategy  $\delta^0$  is Bayes with respect to a f.a. “prior” probability  $\lambda$  on  $\Theta$  if

$$\int_{\Theta} R(\theta; \delta^0) \lambda(d\theta) = \inf_{\delta} \int_{\Theta} R(\theta; \delta) \lambda(d\theta).$$

Strategy  $\delta^*$  is minimax provided that

$$\sup_{\theta} R(\theta; \delta^*) = \inf_{\delta} \sup_{\theta} R(\theta; \delta)$$

Denote the Bayes-risk for  $\delta$  wrt “prior”  $\lambda$  by  $r(\lambda, \delta) = \int_{\Theta} R(\theta; \delta) \lambda(d\theta)$ .

A “prior”  $\lambda^*$  on  $\Theta$  is *least favorable* provided that

$$\inf_{\delta} r(\lambda^*, \delta) = \sup_{\lambda} \inf_{\delta} r(\lambda; \delta).$$

- Assume that the Loss function is bounded below, that decision rules are mixed strategies, and that “prior” probabilities are finitely additive.

***Theorem<sub>1</sub>:*** The decision rules whose risks form the lower boundary of the risk function constitute a minimal complete class of admissible rules.

Each admissible rule is a Bayes rule.

***Theorem<sub>2</sub>:*** There exists a minimax decision rule and a corresponding least-favorable prior.

Each minimax rule is Bayes *wrt* each least-favorable prior.

***Theorem<sub>3</sub> (Rationalizability for infinite games):***

Assume that the loss function is bounded above and below.

Suppose that  $\delta_0$  is a decision rule that is not Bayes for any prior, i.e.,  $\delta_0$  is not *E-admissible* against the vacuous prior.

Then there is decision rule  $\delta_1$  and  $\varepsilon > 0$  such that

$$(\forall \theta) \quad R(\theta; \delta_0) > R(\theta; \delta_1) + \varepsilon.$$

That is, then there is a rival  $\delta_1$  that *uniformly dominates*  $\delta_0$  in Risk.

### 3. But not all *priors* have Bayes-decisions (!)

One of the challenges associated with (merely) f.a. expectations is that the order of integration matters – *Fubini's Theorem* has restricted validity.

So, even though the *Risk* function has a closed (lower) boundary composed of Bayes decisions, it does *not* follow that for an arbitrary “prior”  $\lambda$  on  $\Theta$  there is a Bayes decision  $\delta^0$  wrt  $\lambda$ , where

$$\int_{\Theta} R(\theta; \delta^0) \lambda(d\theta) = \inf_{\delta} \int_{\Theta} R(\theta; \delta) \lambda(d\theta).$$

***Example<sub>4</sub>:***

**Parameter space,  $\Theta = (0, 1)$**

**$\mathcal{A}$  is the set of all non-empty open subintervals of  $(0, 1)$ .**

**That is,  $\mathcal{A} = \{ (x, y): 0 \leq x < y \leq 1 \}$ .**

**Denote by  $Len[(x, y)] = y - x$ , the length of interval  $a$ .**

**The *Loss* function reflects a goal of *anti-estimation* for  $\theta$ :**

$$L(\theta; a) = I_a(\theta)/Len[a] + (1-Len[a])/10$$

**Consider a (strongly) f.a. “prior”  $\lambda^\#$  on  $\Theta$  where, for each  $y > 0$ ,**

$$\lambda^\#\{\theta: 0 < \theta < y\} = 1.$$

**In f.a. jargon,  $\lambda^\#$  agglutinates its mass at the (missing)  $\theta = 0$ .**

The *Bayes risk* with respect to  $\lambda^\#$  satisfies, for each  $n = 1, 2, \dots$ ,

$$\begin{aligned}
 r(\lambda^\#; (1/n, 1)) &= \int_{\Theta} R(\theta; (1/n, 1)) \lambda^\#(d\theta) \\
 &= \int_{\Theta} [I_a(\theta)/\text{Len}[a] + (1-\text{Len}[a])/10] \lambda^\#(d\theta) \\
 &= \int_{\Theta} [I_{(1/n, 1)}(\theta) / (n-1)/n] + (1/n)/10] \lambda^\#(d\theta) \\
 &= 0 + 1/10n = 1/10n.
 \end{aligned}$$

Hence,  $\inf_{\delta} r(\lambda^\#; \delta) = 0$ .

- But, there is no decision rule  $\delta^\#$  with *Bayes risk*  $r(\lambda^\#; \delta^\#) = 0$ .

To see this, note that, by indirect reasoning:

If  $r(\lambda^\#; \delta) = 0$ , then – from the 2<sup>nd</sup> term in the *Loss* function –  $\mathcal{E}_{\lambda^\#}[\text{Len}(\delta)] = 1$ .

But then, because of the order of integration, with the “prior”  $\lambda^\#(d\theta)$  integration on the outside – from the 1<sup>st</sup> term in the *Risk* function –

$$\int_{\Theta} (I_{\delta}(\theta)/\text{Len}[\delta]) \lambda^\#(d\theta) > 0.$$

## ***SUMMARY – Part 2***

**We have reviewed how the use of some merely f.a. mixed strategies convert the failure of simple dominance – a *lemon*, into the closure of the lower-boundary for a (bounded-below) statistical *Loss function*, understood in the fashion of A. Wald – *lemonade*!**

**It follows that there exist:**

- **a *Minimal Complete Class of Admissible* decisions, each of which is Bayes with respect to some (f.a.) “prior”;**
  - **a Minimax rule and Worst-case “prior” for which the Minimax is Bayes;**
- and**
- **a generalized *Rationalizability result* where each never-Bayes decision is uniformly dominated by some alternative (mixed strategy) decision.**

**BUT – not every “prior” has its Bayes rule.**

## Some References

- Amer, M. A. (1985a) Classification of Boolean algebras of logic and probabilities defined on them by classical models. *Zeitschr f. math. Logik und Grundlagen d. Math.* 31: 509-515.
- Amer, M.A. (1985b) Extension of relatively  $\sigma$ -additive probabilities on Boolean algebras of logic. *J. Symbol Logic* 50: 589-596.
- Billingsley, P. (1986) *Probability and Measure* (2<sup>nd</sup> ed.). John Wiley & Sons: New York
- de Finetti, B. (1974) *Theory of Probability* (vol. 1). John Wiley & Sons: New York.
- Schervish, M., Seidenfeld, T., Stern, R. and Kadane, J. (2019) What finite-additivity can add to decision theory. *Stat. Methods and Applications* 29: 237-263.
- Wald, A. (1950) *Statistical Decision Functions*. John Wiley & Sons: New York.

## Appendix

- **Corollary:** Let  $\mathcal{A}$  be an infinite, free Boolean algebra. Each finitely additive probability  $P$  on  $\mathcal{A}$  is purely finitely additive.

We establish the result for a subalgebra  $\mathcal{A}_\Gamma \subseteq \mathcal{A}$  generated by  $\Gamma = \{\gamma_1, \dots\}$ , a denumerable subset of generators of  $\mathcal{A}$ . Without loss of generality, we use the Lindenbaum-Tarski algebra  $\mathcal{L}$  of sentential logic for this subalgebra. That is, up to isomorphism,  $\mathcal{L}$  is the free Boolean algebra with countably many generators. Next, we summarize relevant details of  $\mathcal{L}$ .

Let  $\mathbf{L}$  be the first order sentential language with denumerably many proposition letters,  $p$ , which are the atomic propositions of  $\mathbf{L}$ ,  $\mathcal{P} = \{p_1, p_2, \dots\}$ . For convenience, let ' $\&$ ', ' $\vee$ ' and ' $\neg$ ' be the sentential operators in  $\mathbf{L}$ , whose semantics are respectively the usual truth functions 'and', 'or', and 'not'. Let **WFF** be the denumerable set of well formed formulas in  $\mathbf{L}$ , which is the syntactic, recursive closure of the atomic propositions under the sentential operators.

Let  $\equiv$  denote (semantic) logical equivalence, an equivalence relation between pairs of well formed formulas in  $\mathbf{L}$ .

For  $s \in \mathbf{WFF}$ , let  $\bar{s}$  be the equivalence class of its logically equivalent well formed formulas.

$\mathcal{L}$  is the Lindenbaum-Tarski algebra over  $\mathbf{WFF}/\equiv$ .

$\mathcal{L}$  is a countable Boolean algebra, where, for  $s, t \in \mathbf{WFF}$

the algebraic join  $\bar{s} \vee \bar{t} = \overline{s \vee t}$ ,

the algebraic meet  $\bar{s} \wedge \bar{t} = \overline{s \wedge t}$

the algebraic complement  $\bar{s}' = \overline{\neg s}$ .

For convenience, denote  $\mathbf{T}$  = equivalence class of tautologies,

and  $\perp$  = equivalence class of contradictions.

Define the (transitive) partial order  $\leq$  on  $\mathcal{L}$  by  $\bar{s} \leq \bar{t}$  if  $s$  (semantically) entails  $t$ .

Note that  $\leq$  is a strict partial order; that is, if  $\bar{s} \leq \bar{t}$  and  $\bar{t} \leq \bar{s}$  then  $\bar{s} = \bar{t}$ .

Denote by  $\bar{s} < \bar{t}$  the asymmetric, transitive relation,  $\bar{s} \leq \bar{t}$  and  $\bar{s} \neq \bar{t}$ .

$\mathbf{T}$  is the maximal element and  $\perp$  is the minimal element of this strict partial order.

That is,  $\perp < \mathbf{T}$  and if  $\perp \neq \bar{s} \neq \mathbf{T}$  then  $\perp < \bar{s} < \mathbf{T}$ .

When neither  $\bar{s} \leq \bar{t}$  nor  $\bar{t} \leq \bar{s}$ , say that  $\bar{s}$  and  $\bar{t}$  are *independent*.

Observe that  $\mathcal{L}$  is atomless: That is, consider  $s \in \mathbf{WFF}$  where  $\bar{s} \neq \perp$ . Let  $t = s \ \& \ p$ , where 'p' does not appear among the atomic propositions in  $s$ . Then  $\perp < \bar{t} < \bar{s}$ . So  $\bar{s}$  is not an atom of  $\mathcal{L}$ .

$\mathcal{L}$  is (up to isomorphism) the countable, atomless Boolean algebra. Because  $\mathcal{L}$  is a countable Boolean algebra, it is not a Boolean  $\sigma$ -algebra. (See Sikorski [1969, p. 66, (E)].) So we have to be careful about the existence of infinitary joins and infinitary meets within  $\mathcal{L}$ . That is, an infinitary join is the least upper bound under  $\leq$  and the infinitary meet is the greatest lower bound under  $\leq$  of a (countable) set of elements of the Boolean algebra. These need not exist in  $\mathcal{L}$ .

For  $\bar{S} = \{\bar{s}_i \in \mathcal{L} : i = 1, 2, \dots\}$ , say that  $\bar{t} \in \mathcal{L}$  is *the infinitary join* of  $\bar{S}$ , written  $\bar{t} = \bigvee \bar{S}$ , provided that,

for each  $\bar{s}_i \in \bar{S}$ ,  $\bar{s}_i \leq \bar{t}$ , and

if also there exists  $\bar{t}' \in \mathcal{L}$  where, for each  $\bar{s}_i \in \bar{S}$ ,  $\bar{s}_i \leq \bar{t}'$ , then  $\bar{t} \leq \bar{t}'$ .

The infinitary meet of  $\bar{S}$  is defined similarly.

A (finitely additive) probability  $P$  on  $\mathcal{L}$  satisfies:

- (i)  $0 \leq P(\bar{s}) \leq 1$
- (ii)  $P(\mathbf{T}) = 1, P(\perp) = 0$
- (iii)  $P(\bar{s} \vee \bar{t}) = P(\bar{s}) + P(\bar{t})$  whenever  $\bar{s} \wedge \bar{t} = \perp$ .

*Definition:*  $P$  is countably additive<sub>1</sub> on  $\mathcal{L}$  provided that, whenever  $\bar{S} = \{\bar{s}_i \in \mathcal{L}: i = 1, 2, \dots\}$  is a denumerable partition, i.e., satisfying

- (i)  $\bar{s}_i \wedge \bar{s}_j = \perp$  whenever  $i \neq j$ , and
- (ii) where the infinitary join  $\bar{t} = \bigvee \bar{S}$  exists,

then  $P(\bar{t}) = \sum_i P(\bar{s}_i)$ .

**Proof of the Corollary:** Let  $P$  be a finitely additive probability on  $\mathcal{L}$ . Let  $\varepsilon > 0$ . We show there exists a denumerable partition  $\Psi = \{\psi_1, \psi_2, \dots\}$  in  $\mathcal{L}$  with  $\sum_i P(\psi_i) < \varepsilon$ .

Let  $\Gamma = \{\gamma_1, \dots\}$  be the set of the sentential generators of  $\mathcal{L}$ : the set of (equivalence classes of the) atomic propositions.

Choose integer  $k$  that satisfies,  $(1+\varepsilon)/\varepsilon < 2^k$ ; equivalently,  $\frac{1/2^k}{1-1/2^k} < \varepsilon$ .

For  $j = 1, 2, \dots$ , define successive (disjoint) blocks  $b_j$  containing  $j \times k$  many generators from  $\Gamma$ .

That is,  $b_j = \{\gamma_{\frac{k(j-1)}{2}+1}, \dots, \gamma_{\frac{k(j+1)}{2}}\}$ .

Specifically,  $b_1 = \{\gamma_1, \dots, \gamma_k\}$ ,  $b_2 = \{\gamma_{k+1}, \dots, \gamma_{3k}\}$ ,  $b_3 = \{\gamma_{3k+1}, \dots, \gamma_{6k}\}$ , etc.

The set of blocks partitions the set of generators in  $\Gamma$ .

Each block,  $b_j$ , generates  $2^{j \times k}$  many Boolean elements  $\beta_m^j$  ( $m = 1, \dots, 2^{j \times k}$ ) of  $\mathcal{A}_\Gamma$  of the form

$$\beta_m^j = \delta_{\frac{k(j-1)j}{2}+1} \wedge \dots \wedge \delta_{\frac{k(j+1)j}{2}}$$

where  $\delta_i = \gamma_i$  or  $\delta_i = \gamma_i'$ .

Note that, since the algebra  $\mathcal{A}_\Gamma$  is free, each  $\beta_m^j$  satisfies:  $\perp < \beta_m^j < \mathbf{T}$ .

Trivially, for each block  $b_j$ , if  $\beta_m^j \neq \beta_n^j$  then  $\beta_m^j \wedge \beta_n^j = \perp$ .

Equally evident, for each block  $b_j$ ,  $\mathbf{T} = \bigvee \{\beta_m^j : m = 1, \dots, 2^{j \times k}\}$ .

Because the generators are independent, the Boolean elements  $\beta_m^j, \beta_n^k$  from different blocks  $b_j$  and  $b_k$  are also independent, i.e., neither  $\beta_m^j \leq \beta_n^k$  nor  $\beta_n^k \leq \beta_m^j$ .

As  $P$  is finitely additive, then for each block  $b_j$  ( $j = 1, 2, \dots$ ), there exists (at least) one Boolean element  $\beta_m^j$  with  $P(\beta_m^j) \leq 1/2^{j \times k}$ . For ease of notation, denote this element of  $\mathcal{A}_\Gamma$  as  $\beta_j$ .

Define elements  $\psi_j$  of  $\mathcal{A}_\Gamma$  as follows:

$$\text{for } j = 1, \psi_1 = \beta_1; \text{ and for } j \geq 2, \psi_j = \beta_j \wedge \psi_1' \wedge \dots \wedge \psi_{j-1}'$$

and let  $\Psi = \{\psi_j : j = 1, \dots\}$ .

*Claim:*  $\Psi$  is a partition:

- (i)  $\psi_j \wedge \psi_k = \mathbf{0}$  whenever  $j \neq k$ . (So, also  $\Psi$  is an anti-chain.)
- (ii)  $\mathbf{T} = \bigvee \Psi$

*Proof:* (i) Immediate from the definition of the  $\psi_j$ .

(ii) Trivially,  $\psi_j \leq \mathbf{T}$ . Next we show  $\mathbf{T}$  is the least upper bound for  $\Psi$ .

By a simple induction, for each  $n = 1, 2, \dots$ ,  $\psi_1 \vee \dots \vee \psi_n = \beta_1 \vee \dots \vee \beta_n$ . So, for each  $n$ ,  $\psi_1 \vee \dots \vee \psi_n$  and  $\beta_1 \vee \dots \vee \beta_n$  share the same upper bounds in  $\mathcal{L}$ . Argue indirectly. Let  $\bar{s} < \mathbf{T}$  and suppose  $\bar{s}$  is an upper bound for  $\Psi$ . Then  $(\beta_1 \vee \dots \vee \beta_{k-1})$  entails  $\bar{s}$ . Let  $\gamma_k$  be an atomic proposition not appearing in  $s$ . So  $\beta_k \not\leq \bar{s}$ , i.e., there is a truth assignment where  $\mathbf{t}(\beta_k) = \mathbf{T}$  and  $\mathbf{t}(\bar{s}) = \mathbf{F}$ . Since the atomic propositions have independent truth assignments, there is a semantic model where, also,  $\mathbf{t}(\beta_1 \vee \dots \vee \beta_{k-1}) = \mathbf{F}$  and  $\mathbf{t}(\bar{s}) = \mathbf{F}$ . (If not, i.e., if  $\mathbf{t}(\beta_1 \vee \dots \vee \beta_{k-1}) = \mathbf{F}$  entails  $\mathbf{t}(s) = \mathbf{T}$ , then  $(\beta_1 \vee \dots \vee \beta_{k-1})'$  entails  $\bar{s}$ . And then, as  $(\beta_1 \vee \dots \vee \beta_{k-1})$  entails  $\bar{s}$ ,  $\mathbf{T} \leq \bar{s}$ .) Thus,  $(\beta_1 \vee \dots \vee \beta_k) \not\leq \bar{s}$ . Therefore,  $\mathbf{T} = \bigvee \Psi$ , and the *claim* is verified.

It is evident that  $P(\psi_j) \leq P(\beta_j)$ . So,  $P(\psi_j) \leq 1/2^{j \times k}$ .

Then  $\sum_j P(\psi_j) \leq \sum_j 1/2^{j \times k} = \frac{1/2^k}{1 - 1/2^k} < \varepsilon$ , which establishes that  $P$  is purely finitely additive. ♦Corollary

**Note:** What drives this result is the fact that  $\mathbf{T} = \bigvee \Psi$ . Were  $\mathcal{A}$  a  $\sigma$ -algebra then  $\bigvee \Psi = (\psi_1 \vee \dots \vee \psi_n \vee \dots) < \mathbf{T}$ .