





Implementing Sustainable Development under Deep Uncertainty

Ambiguity Aversion, Modern Bayesianism and Small Worlds

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Sustainability Policy Framework



Top-Down Mobilization Green New Deals around the World



Systems Innovation co-designed with Problem Owners

Integrated & Coordinated Interventions in economic, financial, political and social systems and along whole value chains. In systems, by means of the relations, elements are arranged in such a fashion that gives rise to a **new structure** functioning.



Working through gradual, incremental changes is not enough!

What is needed now is a **fundamental transformation** of economic, social and financial systems that will trigger exponential change in decarbonisation rates and strengthen climate resilience – IPCC report: "**rapid, far- reaching and unprecedented changes in all aspects of society**".





ATHENS UNIVERSITY OF ECONOMICS AND BUSINESS



Cluster for Sustainability Transition

Transforming Research and Innovation into Climate Action

Director: Professor Phoebe Koundouri



The Cluster on Sustainability Transition (CST)



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UN SDSN GREECE http://www.unsdsn.gr/



EIT Climate-KIC HUB GR <u>https://www.athena-</u> <u>innovation.gr/en/eit-</u> <u>climate-kic-greece-hub</u>

CLUSTER ON SUSTAINABILITY TRANSITION

Research - Innovation Acceleration Deep Demonstration - Education & Training



Projects Green-Digital-Just Recovery / Circular Economy / Climate Change Mitigation and Adaptation





4-Seas Initiative

An initiative led by the regional networks SDSN Black Sea and SDSN Mediterranean and the national networks SDSN Greece, SDSN Italy, SDSN Spain, SDSN France, SDSN Turkey and SDSN Russia

GLOBAL ROUNDTABLE FOR SUSTAINABLE SHIPPING AND PORTS

 Aims at bringing together researchers and technology developers, shipbuilders, shipowners, ports, policy makers and politicians, from across the globe, to work on technological and policy innovations, related to zero emissions shipping, to target net-zero emissions by 2050.

• Find more at: <u>http://www.unsdsn.gr/global-</u> roundtable-for-sustainable-shipping-2



Projects

Blue Growth



COASTAL H2020 European Commission Project

a unique research and innovation project

a multi-actor collaboration between entrepreneurs, administrations, stakeholders and experts in coastal and rural natural and social sciences and sciences

aims to formulate and evaluate business solutions and policy recommendations to improve coastal-rural synergy to promote rural and coastal development while preserving the environment.

Find more at: https://h2020-coastal.eu Implementation period: 2018-2022 Budget: € 5 million

Projects Water-Food-Energy Nexus Smart Agriculture & Smart Urban Water Systems





ccelerator elimple-HIC scheme 2020 **APPLY NOW! Deadline 1 March** www.climate-kic.org/ris/accelerator

Successful Greek Start-ups





Uncertainty affects Preferences and Decisions used to Estimate Total Economic Value Total Economic Value Shapes Policy Recommendations

Global climate change illustrates particularly well the importance of considering uncertainty when making a decision.

- Do we face RISK (uncertainty but known probabilities)?
- Do we face DEEP UNCERTAINTY (unknown probabilities)?
- Decision making under deep uncertainty?
- IPCC (2007) wrote:

"In most instances, objective probabilities are difficult to estimate. Furthermore, a number of climate change impacts involve health, biodiversity, and future generations, and the value of changes in these assets is difficult to capture fully in estimates of economic costs and benefits..... The literature on how to account for ambiguity in the total economic value is growing, even if there is no agreed standard."

How do agents make decisions under "deep uncertainty"?

- Our Literature Review pointed to alternative decision-making rules away from Expected Utility Maximization Rule...
- ...Instead of finding ideas to refine the theoretical underpinnings of our models and valuation methods, we ended up criticizing this literature!

AMBIGUITY AVERSION, MODERN BAYESIANISM AND SMALL WORLDS

Phoebe Koundouri, Nikitas Pittis, Panagiotis Samartzis, Nikos Englezos

Outline

Ambiguity Aversion (AA): Aversion to Unknown Probabilities Mathematical Economics – Decision Theory under Subjective Uncertainty

Modern Bayesianism (MB)

Bayesian Epistemology – Bayesian Confirmation Theory (BCT)

Small Worlds

Worlds where Small Number of Propositions are required to cover all cases of Interest (Logic –BCT)

Main Result of the Paper:

AA under MB Collapses to Dynamic Inconsistency of Beliefs

Also,

In Small Worlds, MB (apart from Normatively appealing) is Descriptively plausible

Ambiguity Aversion: History

Origins of AA: Ellsberg Paradox (1961)

Interpretation of Ellsberg Paradox: One (or more) of SEU axioms Fail

Evidence of Irrationality SEU Conditions of Rationality too Demanding

New Axiomatization – Replacing SEU Axioms

Background Material for AA



Unknown Probabilities

Question: What is Probability?



Unknown Probabilities

Subjective Probabilities: Problems

Assuming that such a thing exists:

In order to be able to test whether it is consistent with Kolmogorov's axioms, it has to be measurable (in the sense of observable)





Frank Ramsey (1926) Subjective probabilities can be inferred by observing actions that reflect individuals' personal beliefs.

The degree of probability that an individual attaches to a particular outcome can be measured by finding what odds the individual would accept when betting on that outcome

To be precise, given a subjective probability p for the proposition, you will accept odds of up to p : (1 - p) on its truth

Unknown Probabilities

Ramsey's Insights were ignored until the mid of 1950s

1954: Leonard Savage's "The Foundations of Statistics"

More Appropriate Title: "The Foundations of Economics"

Kreps (1988): "The Crowning Glory of Choice Theory"

Why is it so Important?

Simultaneous Axiomatization of Subjective Probability and Expected Utility Maximization

An Economic Agent does not Need Exogenous Objective Probabilities: All that is Needed is in his Own Mind

Savage's Framework

Objects of Choice: Acts

Acts are functions from the Sample Space (source of uncertainty) to Outcome Space

Savage defines a preference relation on the set of (all) Acts

Savage Representation Theorem (Finite Outcomes Case)

A binary relation > defined on the set of Acts satisfies Savage's seven axioms,

iff there exist (i) A (subjective) probability function (defined on the sample space)

and (ii) a Utility function (defined on the outcome space)

such that for any two acts, f and g:

 $f \succeq g \text{ iff } SEU(f) \ge SEU(g)$

All in One Package: Proper Subjective Probability Cardinal Utility EU Form

Does Savage's Representation Theorem Address the Issue of Measurability of Subjective Probability?

Ask him questions of the type: Do you prefer this to that?

> No Numerical Questions are Asked



You will be able to get Numerical Degrees of Belief which are Coherent

You will be able to get Numerical Utilities which contain Cardinal Information

Profound Implications for:



Question: Where do the priors come from?

Answer: The same place Utility Function comes from!

Savage's Representation Theorem provides a Separation of Tastes from Beliefs



It also delivers three properties in one Package:

Subjective Tastes are represented by Utilities Subjective Beliefs are represented by Probabilities Subjective Probabilities and Utilities are used in conjunction with the maximization rule: You cannot have subjective probabilities and use them to maximize something else (other than expected utility)

Do Economic Agents Behave according to Savage's Axioms?

Descriptive

Should Economic Agents Behave according to Savage's Axioms?

Normative



Evidence Against SEU Behavior

Ellsberg Paradox (1961)

People are Not Probabilistically Sophisticated







The Paradox:

The Preferences

 $f_1(\cdot)$ over $f_2(\cdot)$ and $f_4(\cdot)$ over $f_3(\cdot)$

cannot be represented by an SEU Preference Function

Assume that such a representation exists:

 $f_1(\bullet) > f_2(\bullet) \iff SEU(f_1) > SEU(f_2)$ $SEU(f_1) = P(red) \times 100 + P(black) \times 0 + P(yellow) \times 0 = P(red) \times 100$ $SEU(f_2) = P(red) \times 0 + P(black) \times 100 + P(yellow) \times 0 = P(black) \times 100$ Hence, $SEU(f_1) > SEU(f_2) \iff \mathbf{P(red)} > \mathbf{P(black)}$

Similarly for the second pair:

 $f_4(\cdot) > f_3(\cdot) \iff SEU(f_4) > SEU(f_3)$ $SEU(f_3) = P(red) \times 100 + P(black) \times 0 + P(yellow) \times 100 = P(red) \times 100 + P(yellow) \times 100$ $SEU(f_4) = P(red) \times 0 + P(black) \times 100 + P(yellow) \times 100 = P(black) \times 100 + P(yellow) \times 100$ Hence, $SEU(f_4) > SEU(f_3) \iff \mathbf{P}(\mathbf{black}) > \mathbf{P}(\mathbf{red})$

CONTRADICTION: No (Proper) Subjective Probability Exists. Ellsberg-type Choices cannot be represented by a Coherent-Probability based SEU Rule

	30 Balls	60 Balls	
	Red	Black	Yellow
$f_1(\cdot)$	\$100	\$ 0	\$ 0
$f_2(\cdot)$	\$0	\$100	\$0
$f_{3}(\cdot)$	\$100	\$0	\$100
$f_4(\cdot)$	\$0	\$100	\$100

Response of the Literature to Ellsberg-type Behavior



Response of the Literature to Ellsberg-type Behavior

Which of S1-S7 is Violated by Ellsberg Behavior?



S2: The Sure-thing Principle S3: Strong Comparative Probability

STP: If two subjective acts offer the same prize over some event,...

...then replacing it with any other prize will not change the ranking of the acts $f_3(\cdot) 100 \$0 \$100 $f_4(\cdot) 0 \$100 \$100 $f_1(\cdot) \text{ over } f_2(\cdot) \text{ and } f_4(\cdot) \text{ over } f_3(\cdot)$

	30 Balls	60 Balls	
	Red	Black	Yellow
$f_1(\cdot)$	\$100	\$ 0	\$ 0
$f_2(\cdot)$	\$0	\$100	<u>\$0</u>
$f_{3}(\cdot)$	\$100	\$ 0	\$100
$f_4(\cdot)$	\$0	\$100	\$100

A stronger version of Savage's weak Comparative Probability Axiom (Machina and Schmeidler 1992) allows qualitative probabilistic comparisons between events.

Violation of STP

Response of the Literature to Ellsberg-type Behavior

Schmeidler (1989): Non-additive probabilities (Capacities) - Choquet Integration

Gilboa and Schmeidler (1989): Multiple Probabilities - Maxmin Expected Utility

(i) For each act f compute all of its expected utilities for all $\mu_{\theta}(\cdot)$.

(ii) Find the minimum expected utility for act f

(iii) Compare all acts in terms of their minima. Choose the one with the maximum minimum

The Maxmin Expected Utility decision rule suggests that the decision maker can be characterized by a utility function and a set of prior probabilities, such that the chosen act maximizes the minimal expected utility, where the minimum is taken over the priors in the set

Criticism: Too Pessimistic

Source of Ambiguity – Ambiguity Aversion



Source of Ambiguity – Ambiguity Aversion

Source of Ambiguity:

At the moment of Formation of his Beliefs (Prior Beliefs), DM is faced with Asymmetric Information about Objective Probabilities (Chances)

 $\mathcal{F} = \{\{R, B, Y\}, \emptyset, \{R, B\}, \{R, Y\}, \{B, Y\}, \{R\}, \{B\}, \{Y\}\}\}$

He has Specific Information (Chances) about Some events

QUESTIONS

general characteristics of the chance set up: all H, E-relevant to H, entailment relations H-E

 I_B

Is- actual outcomes or info on objective chances

Is- will be utilized via conditionalization on the Prior

Should we allow Specific Information to Affect the Formation of Prior Beliefs?

Should we allow only Background Information to Affect the Formation of Prior Beliefs?

Background vs. Specific Information as Direct Determinants of Prior Beliefs



Background vs. Specific Information as Direct Determinants of Prior Beliefs



Modern vs. Classical: Comparisons

Modern Bayesianism is adopted (almost without exception) from Philosophers of Science

It is a standard practice for Philosophy-of-Probability papers to start with a "reasonable initial Credence function" (before receiving any evidence)



Rudolph Carnap argues strongly in favor of MB His monumental work on inductive logic (Carnap 1950) is based on the concept of hypothetical or counterfactual initial credence function that can be ascribed to the agent, before the collection of any evidence.

X's momentary inclination (to believe) at time t



X's permanent disposition to believe

Modern vs. Classical: Comparisons



Main Result of the Paper:

In the context of MB, Ambiguity Aversion collapses to Dynamic Inconsistency of Beliefs



soon as I earn 8%

One month later... with 8% return being a fact

Schmeidler (1989) uses the following coin example, which aims at conveying the same message with Ellsberg's paradox



The two coins are about to be tossed and X has the option to bet either on A-related events or B-related events



X is forming his prior...



Although X does not know the exact values of $P(D_H^B)$ and $P(D_T^B)$ he feels more willing to bet on D_H^A than D_H^B and on D_T^A than D_T^B

It is easy to show that such a *P* is non-additive. Indeed,

$$D^B_H \cap D^B_T = \emptyset$$

 $D_H^B \cup D_T^B = \Omega$

and

where Ω is the relevant sample space, namely

$$\Omega = \{H_A H_B, H_A T_B, T_A H_B, T_A T_B\}$$

Assuming that *P* is additive,

$$1 = P(D_{H}^{B} \cup D_{T}^{B}) = P(D_{H}^{B}) + P(D_{T}^{B}) < 1$$

which is a contradiction.

What causes the violation of the additivity property in P?

It is the fact that the agent allowed Specific Information (Information about the Chances of A) to affect directly her probabilistic beliefs, instead of utilizing (as she should) the specific information indirectly by conditionalization

...but in order to be able to conditionalize on the specific information : "the objective probability of Heads in A is 0.5"...

...he needs a (pre-existing) vehicle...

 I_S



It is treated as contingent, NOT actual information (being on par with any other conceivable piece of information)

Is does not enjoy any special status

What was Agent's Background Information at t=0, that is at the beginning of his epistemic life, when no specific information was available?

For simplicity, we assume that concerning coin A, the agent knows with certainty that only one of the following three hypotheses is true:

 $\mathcal{H}_1^A = \{Coin \ A \ is \ fair\}$

 $\mathcal{H}_2^A = \{Coin A favors H (0.6 - 0.4)\}$

 $\mathcal{H}_3^A = \{Coin \ A \ favors \ T \ (0.6 - 0.4)\}$

Similarly, for Coin B:

 $\mathcal{H}_1^B = \{Coin \ B \ is \ fair\}$

 $\mathcal{H}_2^B = \{Coin \ B \ favors \ H \ (0.6 - 0.4)\}$

 $\mathcal{H}_3^B = \{Coin \ B \ favors \ T \ (0.6 - 0.4)\}$

Let $\mathbf{H}_{A} = \{\mathcal{H}_{1}^{A}, \mathcal{H}_{2}^{A}, \mathcal{H}_{3}^{A}\}$ and $\mathbf{H}_{B} = \{\mathcal{H}_{1}^{B}, \mathcal{H}_{2}^{B}, \mathcal{H}_{3}^{B}\}$ Partition of Event Space Partition of Event Space



Back to Schmeidler's events:

 $D_H^A = \{ \text{coin A comes up heads} \} = \{ H_A H_B, H_A T_B \}$

 $D_T^A = \{ \text{coin A comes up tails} \} = \{ T_A H_B, T_A T_B \}$

 $D_H^B = \{ \text{coin } B \text{ comes up heads} \} = \{ H_A H_B, T_A H_B \}$

 $D_T^B = \{ \text{coin } \mathsf{B} \text{ comes up tails} \} = \{ H_A T_B, T_A T_B \}$

Using the Law of Total Probability, the Agent's Prior Probabilities are:

All this
activity...
$$P_{0}(D_{H}^{4}) = \sum_{i=1}^{3} P_{0}(D_{H}^{4} | \mathcal{H}_{i}^{4})P_{0}(\mathcal{H}_{i}^{4}) =$$
$$= 0.5 \times \frac{1}{3} + 0.6 \times \frac{1}{3} + 0.4 \times \frac{1}{3} = 0.5$$
$$\lim_{i \to \infty} 1 + 0.6 \times \frac{1}{3} + 0.4 \times \frac{1}{3} = 0.5$$
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at t=1

Bayesian Conditionalization $P_{new}(\mathcal{H}_1^A) = P_0(\mathcal{H}_1^A \mid < \mathcal{H}_1^A \text{ is true } >) = 1$ $P_{new}(\mathcal{H}_2^A) = P_0(\mathcal{H}_2^A \mid < \mathcal{H}_1^A \text{ is true } >) = 0$ $P_{new}(\mathcal{H}_3^A) = P_0(\mathcal{H}_3^A \mid < \mathcal{H}_1^A \text{ is true } >) = 0$

IMPORTANT: All these are prior probabilities already formed at t=0

The New Probabilities for Coin-B Hypotheses:

 $P_{new}(\mathcal{H}_1^B) = P_0(\mathcal{H}_1^B \mid \langle \mathcal{H}_1^A \text{ is true } \rangle) = P_0(\mathcal{H}_1^B) = \frac{1}{3}$ $P_{new}(\mathcal{H}_2^B) = P_0(\mathcal{H}_2^B \mid \langle \mathcal{H}_1^A \text{ is true } \rangle) = P_0(\mathcal{H}_2^B) = \frac{1}{3}$ $P_{new}(\mathcal{H}_3^B) = P_0(\mathcal{H}_3^B \mid \langle \mathcal{H}_1^A \text{ is true } \rangle) = P_0(\mathcal{H}_3^B) = \frac{1}{3}$

The Posterior Probabilities of Schmeidler's Events are: (once the specific information about Coin A has been allowed to play its role)



For the event D_H^B :

 $P_{new}(D_H^B) = P_0(D_H^B \mid \langle \mathcal{H}_1^A \text{ is true } \rangle) =$ $\sum_{i=1}^3 P_0(D_H^B \mid \mathcal{H}_i^B, \langle \mathcal{H}_1^A \text{ is true } \rangle) \times P_0(\mathcal{H}_i^B \mid \langle \mathcal{H}_1^A \text{ is true } \rangle)$

Since \mathcal{H}_i^B and $\langle \mathcal{H}_1^A$ is true > are Independent Propositions

$$P_{new}(D_H^B) = P_0(D_H^B \mid \langle \mathcal{H}_1^A \text{ is true } \rangle) =$$

$$\sum_{i=1}^3 P_0(D_H^B \mid \mathcal{H}_i^B) \times P_0(\mathcal{H}_i^B) = 0.5$$

Similarly, $P_{new}(D_T^B) = 0.5$

X has formed his posterior

Let us check whether it violates additivity (a-la Schmeidler)



AA under MB is Equivalent to Dynamic Inconsistency of Beliefs



Modern vs. Classical Bayesianism



Modern Bayesianism and Small Worlds

ANSWER: In Small Worlds, YES

Small vs. Large Worlds

A Small World is a set of propositions whose number of elements is small

In a small world a Bayesian (modern or classical) can assign probabilities to all propositions of interest

> Apart from Small, the World must be Non-Evolving

Modern Bayesianism and Small Worlds

Small and Non-Evolving Worlds

Moreover, apart from small, worlds must be Non-Evolving

Assume that at some point in time t>0, you come up with a new hypothesis H-new

Motivated by H-new you acquire Evidence E-new

You attempt to Update your Beliefs based on E-new

As a good Bayesian (Modern or Classical) you consult your Prior

SURPRISE: Your Prior has not assigned probabilities to H-new, Enew Propositions (because back then they were unconceived)

Small vs. Large Worlds

- Binmore (2009): "Only in a small world, in which you can always look before you leap, it is possible to consider everything that might be relevant to the decisions you take."
- Indeed the "look before you leap" proverb is attributed to Savage who used it as antithetical to "cross that bridge when you come to it" that referred to the so called "large worlds".
- It is worth mentioning that Savage himself made quite clear that his own conception of subjective probability together with its axiomatization is relevant only for small worlds.
- This is because Savage's framework is essentially static in the sense that it does not allow for the so-called "concept formation", that is the formation of a new hypothesis or a new idea sometime in the future.
- Savage himself acknowledged the fact that the static nature of his theory makes it inapplicable in the case of large evolving worlds by referring to such an extension as "ridiculous" and "preposterous"

Conclusions

Modern Bayesianism Dissolves the Problem of Ambiguity Aversion

Under MB, AA collapses to Dynamic Inconsistency (Non-negotiable Irrationality)

MB assumes that in forming his subjective priors, the agent makes use of Background Information only

Specific Information (e.g. information about chances) is allowed to affect beliefs only through conditionalization (by means of the pre-existing prior)

Bayesianism in general (Modern or Classical) is Plausible only in Small Worlds

A third Option that relaxes BC: The Evolving Probability Model: Change your probability (in the light of new evidence) anyway you want as long as you respect coherence