

Confidence in Beliefs and Rational Policy Making

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Climate Policy & Uncertainty

Climate Policy (e.g. Social Cost of Carbon)

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Equilibrium Climate Sensitivity (ECS)

Climate Policy & Uncertainty

Climate Policy (e.g. Social Cost of Carbon)



Equilibrium Climate Sensitivity (ECS)

a) Equilibrium climate sensitivity estimates ($^{\circ}\text{C}$)

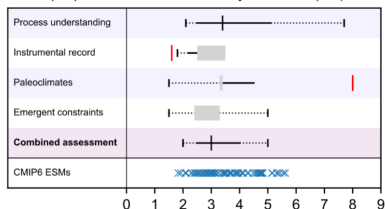


Figure: (Masson-Delmotte et al., 2021, fig TS.16b).

- ▶ Best estimate: 3°C
- ▶ Likely in range 2.5 to 4°C (high confidence)
- ▶ Virtually certain above 1.5°C (high confidence)
- ▶ Very likely below 5°C (medium confidence)

(Masson-Delmotte et al., 2021)

IPCC uncertainty language

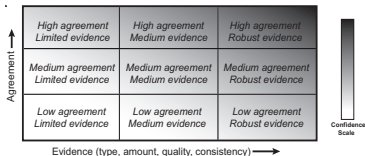
ECS is likely in range 2.5 to 4° C (high confidence)

CONFIDENCE ‘in the validity of a finding, based on the type, amount, quality, and consistency of evidence . . . and the degree of agreement.’

PROBABILITY ‘based on statistical analysis of observations or model results, or expert judgment’ (‘Likelihood’)

Table 1. Likelihood Scale

Term*	Likelihood of the Outcome
<i>Virtually certain</i>	99-100% probability
<i>Very likely</i>	90-100% probability
<i>Likely</i>	66-100% probability
<i>About as likely as not</i>	33 to 66% probability
<i>Unlikely</i>	0-33% probability
<i>Very unlikely</i>	0-10% probability
<i>Exceptionally unlikely</i>	0-1% probability



Climate Policy & Uncertainty

How to decide when there is this much uncertainty?

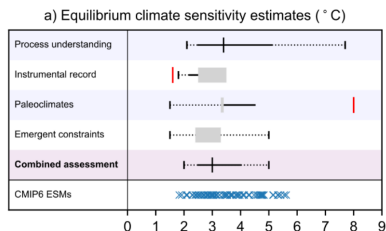


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Climate Policy & Uncertainty

Aim Overview of issues & a proposal concerning:

How to decide when there is this much uncertainty?

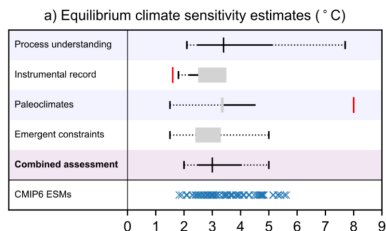


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Plan

Introduction

The standard story: Bayesianism

The Challenge of Uncertainty

Confidence and Decision

Defense

Climate policy making

Aggregation

Conclusion

Bayesianism

A Quick Introduction

What is it? Three tenets:

Belief Beliefs (and hence uncertainty) can be represented by **probabilities**.

Decision Decision Makers maximise **expected utility**.
Subjective Expected Utility (SEU)

Belief Formation Update by **conditionalisation**.

Expected utility

$$\mathbb{E}_p(u(f)) = \sum_{s \in S} u(f(s)) \cdot p(s)$$

- ▶ u : utility function, representing desires or tastes
- ▶ p : probability measure, representing beliefs

Bayesianism

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Saying something:

- ▶ Descriptive
- ▶ **Normative**
 - ▶ **guide the rationalisation, evaluation and deliberation of (tough) decisions and uncertainty**

Bayesianism

A Quick Introduction

Why? Purportedly:

1. has acceptable choice-theoretical consequences
2. is conceptually clear about the roles of different mental attitudes

Bayesianism

A Quick Introduction

Why? Purportedly:

1. has acceptable choice-theoretical consequences
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► Dutch Book Arguments, Representation Theorems

i.e. results of the following form:

Properties of preferences / choice $\Leftrightarrow \exists p, u$ s.t. choice maximises $\mathbb{E}_p(u(f))$

And the p, u are appropriately unique.

Bayesianism

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Bayesianism

A Quick Introduction

Why? Purportedly:

1. has acceptable choice-theoretical consequences
2. is conceptually clear about the roles of different mental attitudes

And also Provides framework for:

- ▶ belief update
- ▶ uncertainty reporting
- ▶ belief aggregation
- ▶ belief elicitation
- ▶ decision analysis

Bayesianism and Climate Uncertainty

The lack of firm probabilities is not a reason to give up expected value theory. You might despair and adopt some other way of coping with uncertainty That would be a mistake. Stick with expected value theory . . . and do your best with probabilities and values. (Broome, 2012)

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→ Just pick a (precise) probability and use it

The Challenge(s)

Unknown urn	Known urn
100 balls	100 balls
Each red or black	50 red, 50 black

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What do you **believe** about the next ball drawn?

Bayesianism:
same ($\frac{1}{2}$).

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Keynes:

- ▶ **Balance of evidence**: same
- ▶ **Weight of evidence**: different

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Which urn would you rather **bet on**?

Bayesianism:
indifferent.

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Ellsberg:
Known urn

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Bayesianism denies any role to (something like)
weight of evidence
in choice

Severe Uncertainty

Project

Belief Representation & Decision

Aim

To formulate and defend:

- ▶ **Normatively valid** models to give guidance for **rational** decision making under uncertainty.

Severe Uncertainty

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And Beyond

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Develop the framework to account for:

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- ▶ ...

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Belief Representation & Decision

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To formulate and defend:

- ▶ **Normatively valid** models to give guidance for **rational** decision making under uncertainty.

Desiderata

1. it has acceptable choice-theoretical consequences
2. it is conceptually clear about the roles of different mental attitudes

And of course:

3. it can fruitfully deal with real-life “severe” uncertainty situations such as those above.

Plan

Introduction

Confidence and Decision

Confidence

... and choice

Defense

Climate policy making

Aggregation

Conclusion

Proposal in a nutshell

The concept ...

to express the proper state of belief, not one number but two are requisite, the first depending on the inferred probability, the second on the amount of knowledge on which that probability is based. (Peirce, 1878, p179)

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Beyond the degree to which one endorses a particular proposition

... there is the degree to which one is confident in this endorsement.

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Beyond the degree to which one endorses a particular proposition

... there is the degree to which one is confident in this endorsement.

If the former is one's beliefs, the latter is one's

confidence in one's beliefs

Modelling confidence

Starting point Represent beliefs by

- ▶ A single probability measure (Bayesianism)

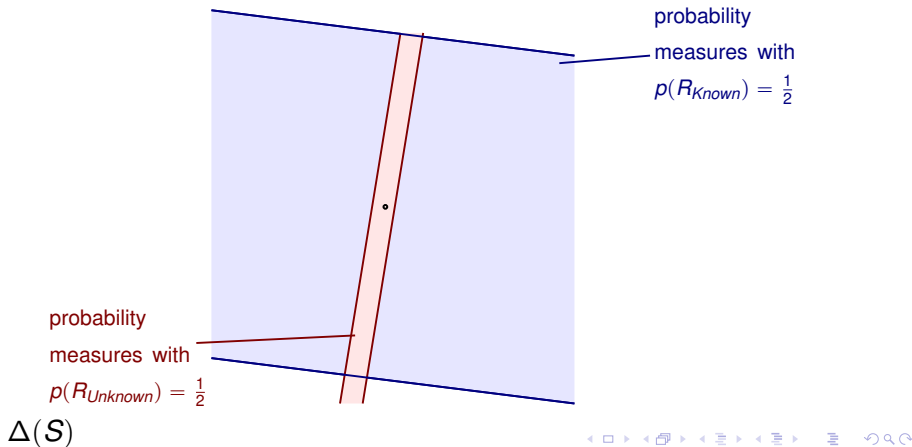
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$\Delta(S)$

Modelling confidence

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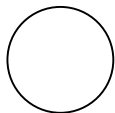
- ▶ A single probability measure (Bayesianism)
- ≡ A complete, consistent set of probability judgements (e.g. 'the probability of A is no smaller than x')



Modelling confidence

First attempt Represent beliefs by

- ▶ A set of probability measures (multiple priors, credal sets)
- ≡ A consistent set of probability judgements

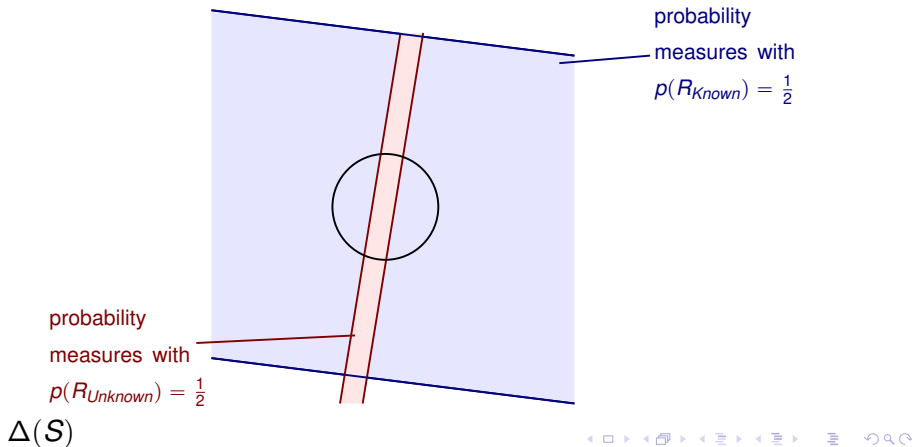


$\Delta(S)$

Modelling confidence

Interpretation

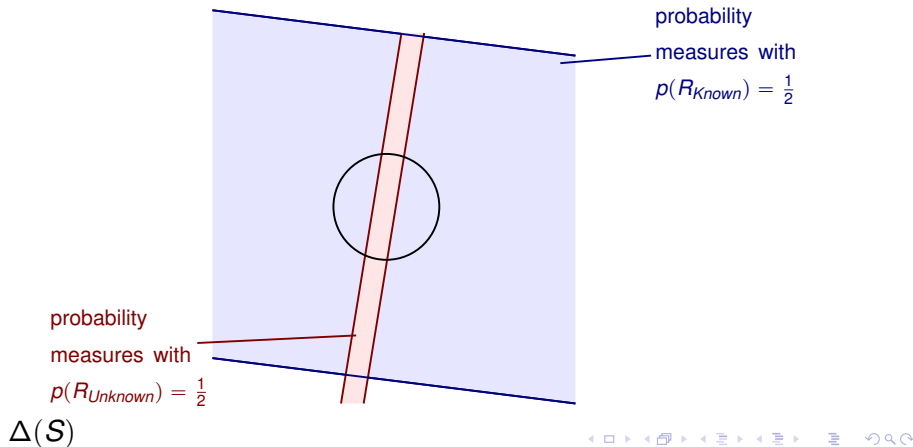
- ▶ confident that the probability of R_{Known} is 0.5
- ▶ unsure about whether the probability of $R_{Unknown}$ is 0.5



Modelling confidence

Problem (credal sets)

- ▶ confidence is represented as “binary”.
- ▶ but it comes in degrees.



Modelling confidence

Idea Represent beliefs by

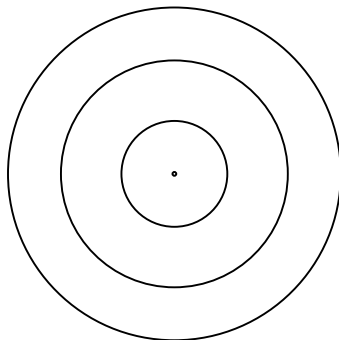
- ▶ A consistent set of probability judgements **for each confidence level**

$\Delta(S)$

Modelling confidence

Idea Represent beliefs by

- ▶ **A nested family of sets of measures**
- ≡ A consistent set of probability judgements for each confidence level

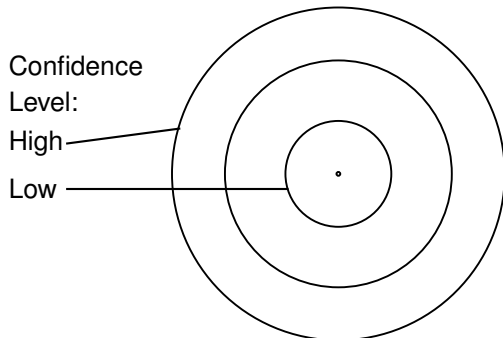


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Modelling confidence

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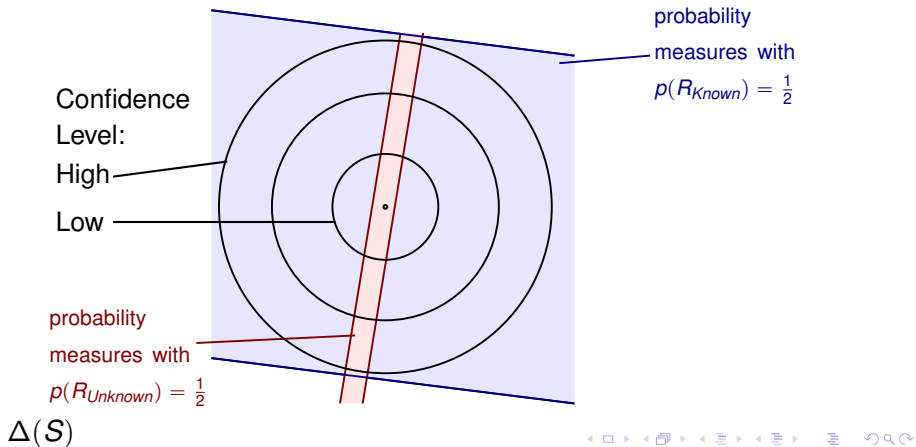


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Modelling confidence

Interpretation

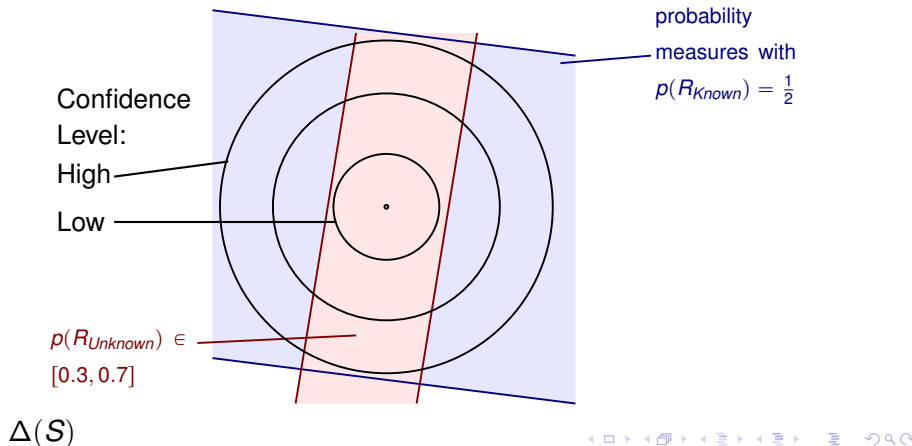
- ▶ **more confident** in $p(R_{Known}) = 0.5$ than $p(R_{Unknown}) = 0.5$



Modelling confidence

Interpretation

- ▶ **more confident** in $p(R_{Unknown}) \in [0.3, 0.7]$ than $p(R_{Unknown}) = 0.5$



Modelling confidence

Definition

A **confidence ranking** Ξ is a nested family of closed subsets of $\Delta(S)$.

Centered: it contains a singleton set (Bayesian with confidence).

Convex: every $C \in \Xi$ is convex.

Continuous: for every $C \in \Xi$, $C = \overline{\bigcup_{C' \subsetneq C} C'} = \bigcap_{C' \supsetneq C} C'$.

Modelling confidence

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- ▶ “Logic of confidence”:

Lifting High confidence in judgement \Rightarrow held at lower confidence levels; not vice versa.

IPs At each confidence level, standard logic of credal sets.

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IPs At each confidence level, standard logic of credal sets.

- ▶ Confidence-imprecision tradeoff:

- ▶ Confidence level \uparrow , imprecision \uparrow .

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Equivalent representations

O : (ordered) set of confidence levels.

Conf. ranking $c : O \rightarrow 2^\Delta$

Implausibility fn $\iota : \Delta \rightarrow O$

Conf. in judg. $conf : 2^\Delta \rightarrow O$

$$\iota(p) = \min \{o : p \in c(o)\}$$

$$conf(\mathcal{P}) = \max \{o : c(o) \subseteq \mathcal{P}\}$$

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Also

- ▶ $O \rightarrow$ sets of desirable gambles, coherent lower previsions, belief functions etc.

Confidence and Decision

There appear to be many significant decisions where confidence in beliefs do, and should, play a role.

The action which follows upon an opinion depends as much upon the amount of confidence in that opinion as it does upon the favorableness of the opinion itself. (Knight, 1921, p226-227)

Confidence and Decision

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But what role?

Proposal in a nutshell

... and the intuition

- ▶ would we like decisions about climate change policy to be taken on the basis of “best hunch” estimates?

Proposal in a nutshell

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- ▶ would we like decisions about climate change policy to be taken on the basis of “best hunch” estimates?
- ▶ and what about wagers between us?

Proposal in a nutshell

... and the intuition

Maxim

The higher the stakes involved in a decision, the more confidence is needed in a belief for it to play a role in the decision.

The role of confidence in choice

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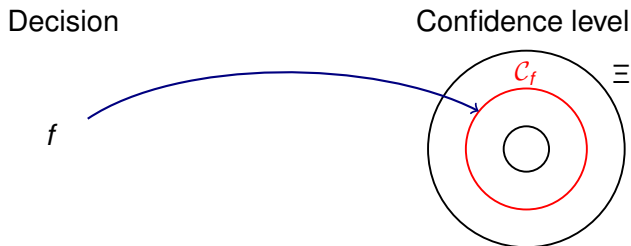
Decision

f

The role of confidence in choice

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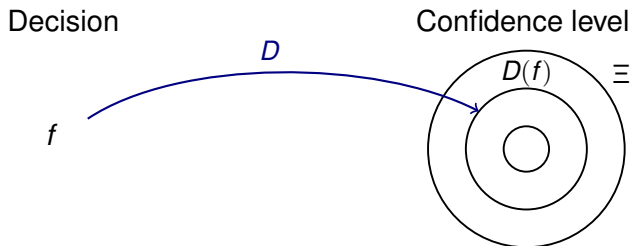
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The role of confidence in choice

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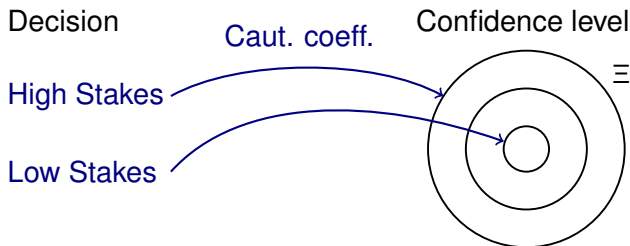


A **cautiousness coefficient** for a confidence ranking Ξ is a surjective function $D : \mathcal{A} \rightarrow \Xi$

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A **cautiousness coefficient** for a confidence ranking Ξ is a surjective function $D : \mathcal{A} \rightarrow \Xi$

- ▶ the higher the stakes, the larger $D(f)$

Decision

The Confidence Framework

Ingredients:

- ▶ utility function u
- ▶ confidence ranking Ξ
- ▶ cautiousness coefficient D
- ▶ decision rule I

General form:

preferences concerning f are a function of

$$u(f(s)) \text{ and } D(f)$$

according to decision rule I

Confidence and choice

Examples

Decision rules using a (closed) set of probabilities:

- ▶ unanimity rule

$$f \leq g \quad \text{iff} \quad \mathbb{E}_p(u(f)) \leq \mathbb{E}_p(u(g)) \quad \text{for all } p \in \max\{D(f), D(g)\}$$

- ▶ maxmin expected utility

$$f \leq g \quad \text{iff} \quad \min_{p \in D(f)} \mathbb{E}_p(u(f)) \leq \min_{p \in D(g)} \mathbb{E}_p(u(g))$$

- ▶ Hurwicz or α -maxmin rule

$$\alpha \min_{p \in D(f)} \mathbb{E}_p(u(f)) + (1 - \alpha) \max_{p \in D(g)} \mathbb{E}_p(u(f))$$

- ▶ etc.

▶ On stakes

Confidence and chance

Example: Careful preferences

- ▶ “Maxmin EU” decision rule

Confidence and chance

Example: Careful preferences

Choose to maximise:

$$\min_{p \in D(f)} \mathbb{E}_p(u(f))$$

Confidence and chance

Example: Careful preferences

Choose to maximise:

$$\min_{p \in D(f)} \mathbb{E}_p(u(f))$$

NB:

- ▶ $D(f)$ is a belief function \Rightarrow convex capacity
monotonicity
 \implies Choquet EU \equiv maxmin EU (over core)

Confidence and chance

Example: Careful preferences

Choose to maximise:

$$\min_{p \in D(f)} \mathbb{E}_p(u(f))$$

Under such a rule:

- ▶ for higher stakes, one is effectively only relying on beliefs in which one has sufficient confidence.
- ▶ behaviour is as “pessimistic” as one’s confidence: the more confident in appropriate beliefs or the lower the stakes, the less pessimistic.

Confidence and chance

Example: Careful preferences

Choose to maximise:

$$\min_{p \in D(f)} \mathbb{E}_p(u(f))$$

Conclusion This yields the following advice:

Choose boldly if one has sufficient confidence; choose cautiously if not.

Confidence and chance

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Comparison Few “non-EU” rules correspond so closely to plausible maxims of this sort.

Comparison This rule is not as extreme as maxmin EU.

Ellsberg Can accommodate Ellsberg behaviour in the same way as the “standard maxmin EU rule”.

Plan

Introduction

Confidence and Decision

Defense

- Choice

- Beliefs and desires

Climate policy making

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Desiderata

1. it has acceptable choice-theoretical consequences
2. it is conceptually clear about the roles of different mental attitudes

And of course:

- ✓ it can fruitfully deal with “severe uncertainty” situations such as those above.

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▶ Dutch Books

▶ Representation Theorems

▶ Neither in detail

Dynamic choice-theoretic arguments: see Hill (2020) for a rebuttal.

Representation Theorems

Preliminaries

A typical framework

S non-empty finite set of states

$\Delta(S)$ set of probability measures on S

C set of consequences

\mathcal{A} set of acts (functions $S \rightarrow C$)

\preceq preference relation on \mathcal{A}

Representation Theorems

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Notation:

- ▶ $f_\alpha g$: shorthand for $\alpha f + (1 - \alpha)g$.

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Special case (Anscombe-Aumann framework)

$C =$ set of lotteries (probability distributions with finite support) over a set X of prizes.

Careful preferences

Axioms

Expected utility (Anscombe and Aumann):

For all $f, g, h \in \mathcal{A}$, $\alpha \in (0, 1)$:

Non triviality and weak order \leq is non-trivial, reflexive, transitive and complete.

Independence $f \leq g$ iff $f_\alpha h \leq g_\alpha h$.

Continuity $\{\alpha \in [0, 1] \mid f_\alpha h \leq g\}$ and $\{\alpha \in [0, 1] \mid f_\alpha h \geq g\}$ are closed.

Monotonicity if $f(s) \leq g(s)$ for all $s \in S$, then $f \leq g$.

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Standard maxmin EU model (Gilboa-Schmeidler):

For all $f, g, h \in \mathcal{A}$, $c \in \mathcal{C}$, $\alpha \in (0, 1)$:

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C-Independence $f \leq g$ iff $f_\alpha c \leq g_\alpha c$.

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Careful preferences

Axioms

Confidence-based careful preference model:

For all $f, g, h \in \mathcal{A}$, $c, d \in \mathcal{C}$, $\alpha \in (0, 1)$:

Non triviality and weak order \leq is non-trivial, reflexive, transitive and complete.

S-Independence (i) if $f \geq c$, then $f_\alpha d \geq c_\alpha d$ whenever the stakes are lower.
(ii) if $f \leq c$, then $f_\alpha d \leq c_\alpha d$ whenever the stakes are higher.

Continuity $\{\alpha \in [0, 1] \mid f_\alpha h \leq g\}$ and $\{\alpha \in [0, 1] \mid f_\alpha h \geq g\}$ are closed.

Monotonicity if $f(s) \leq g(s)$ for all $s \in S$, then $f \leq g$.

Uncertainty Aversion For all $f, g \in \mathcal{A}$, $\alpha \in (0, 1)$, if $f \sim g$ then $f_\alpha g \geq f$.

Careful preferences

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- S-Independence (i) if $f \geq c$, then $f_\alpha d \geq c_\alpha d$ whenever the stakes are lower.
- (ii) if $f \leq c$, then $f_\alpha d \leq c_\alpha d$ whenever the stakes are higher.

Continuity $\{\alpha \in [0, 1] \mid f_\alpha h \leq g\}$ and $\{\alpha \in [0, 1] \mid f_\alpha h \geq g\}$ are closed.

Monotonicity applies to acts of the same stakes

Uncertainty Aversion applies to acts of the same stakes

Careful preferences

Representation theorem

Theorem

\leq satisfies axioms above

\Leftrightarrow there exist $u : X \rightarrow \mathbb{R}$, Ξ and $D : \mathcal{A} \rightarrow \Xi$ such that, for all $f, g \in \mathcal{A}$,
 $f \leq g$ iff

$$\min_{p \in D(f)} \mathbb{E}_p u(f(s)) \leq \min_{p \in D(g)} \mathbb{E}_p u(f(s))$$

Furthermore u is unique up to positive affine transformation, and Ξ and D are unique.

► Enough

Incomplete preferences

Axioms

Expected utility (Anscombe and Aumann):

For all $f, g, h \in \mathcal{A}$, $\alpha \in (0, 1)$:

Non triviality and reflexivity \leq is non-trivial and reflexive.

Completeness $f \leq g$ or $f \geq g$.

Transitivity if $f \leq g$ and $g \leq h$, then $f \leq h$.

Independence $f \leq g$ iff $f_\alpha h \leq g_\alpha h$.

Continuity $\{(\alpha, \beta) \in [0, 1]^2 \mid f_\alpha h \leq g_\beta h\}$ is closed.

Monotonicity if $f(s) \leq g(s)$ for all $s \in S$, then $f \leq g$.

Incomplete preferences

Axioms

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Incomplete preferences

Axioms

Standard unanimity model (Bewley):

For all $f, g, h \in \mathcal{A}$, $\alpha \in (0, 1)$:

Non triviality and reflexivity \leq is non-trivial and reflexive.

Completeness $f \leq g$ or $f \geq g$ whenever f, g are constant acts.

Transitivity if $f \leq g$ and $g \leq h$, then $f \leq h$.

Independence $f \leq g$ iff $f_\alpha h \leq g_\alpha h$.

Continuity $\{(\alpha, \beta) \in [0, 1]^2 \mid f_\alpha h \leq g_\beta h\}$ is closed.

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Incomplete preferences

Axioms

Confidence-based incomplete preference model:

For all $f, g, h \in \mathcal{A}$, $\alpha \in (0, 1)$:

Non triviality and reflexivity \preceq is non-trivial and reflexive.

Completeness $f \preceq g$ or $f \succeq g$ whenever f, g are constant acts.

Transitivity if $f \preceq g$ and $g \preceq h$, then $f \preceq h$.

Independence $f \preceq g$ iff $f_\alpha h \preceq g_\alpha h$.

Continuity $\{(\alpha, \beta) \in [0, 1]^2 \mid f_\alpha h \preceq g_\beta h\}$ is closed.

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Monotonicity if $f(s) \leq g(s)$ for all $s \in S$, then $f \leq g$.

Consistency when the stakes decrease, one cannot suspend (determinate) preferences.

Incomplete preferences

Representation theorem

Theorem

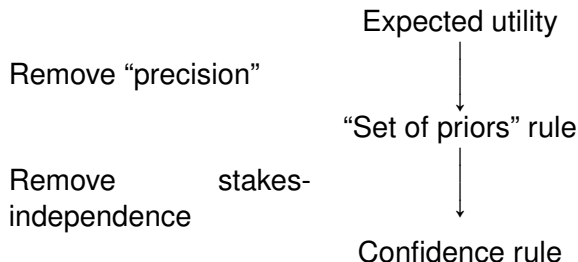
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$$\sum_{s \in S} \mathbb{E}_p(u(f)) \leq \sum_{s \in S} \mathbb{E}_p(u(g)) \quad \forall p \in D((f, g))$$

Furthermore u is unique up to positive affine transformation, and Ξ and D are unique.

General moral



Conclusion

The basic issue between the confidence family and fixed set of priors (aka imprecise probability) models:

- ▶ is stakes-independence a rationality constraint?

Dutch Books

The standard argument (approximately):

Conditions on bets \Leftrightarrow betting quotients are probabilities

Where:

Bet on A with stakes S yields $\in S$ if A and $\in 0$ if not A .

Betting quotient $q(A)$ value such that you are indifferent between buying and selling the bet at stakes S for $\in q(A)S$.

Dutch Books

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In prevision language:

RV X with stakes S_X E.g. $S_X = \max_{\omega \in \Omega} |X(\omega)|$

$\frac{P(X)}{S_X}$ for lower prevision \underline{P}

Dutch Books

The standard argument (approximately):

Conditions on bets \Leftrightarrow betting quotients are probabilities

Assumptions:

For any bets on events A with stakes S :

- ▶ $\epsilon q(A)S$ is the price at which you are indifferent between buying and selling the bet
- ▶ $\epsilon q(A)S$ is the buying / selling price for all stakes S

Buy-sell coincidence $\underline{P}(X) = -\underline{P}(-X)$

Stakes Independence $\frac{P(\frac{S}{S_X}X)}{S} = \frac{P(\frac{T}{S_X}X)}{T}$

Dutch Books

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Dutch Books

Removing it gives:

Conditions on bets \Leftrightarrow buying / selling prices are minimal / maximal probabilities of **fixed set of priors**

Assumptions:

For any bets on events A with stakes S :

- ▶ you have a buying price $\in \underline{q}_S(A)S$ and a selling price $\in \overline{q}_S(A)S$
- ▶ $\in q(A)S$ is the buying / selling price for all stakes S

Stakes Independence $\frac{P(\frac{S}{S_X}X)}{S} = \frac{P(\frac{T}{S_X}X)}{T}$

Dutch Books

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- ▶ $\in q(A)S$ is the buying / selling price for all stakes S

Stakes Independence $\frac{P(\frac{S}{S'}X)}{S} = \frac{P(\frac{T}{S'}X)}{T}$

Dutch Books

Which gives:

Conditions on bets \Leftrightarrow buying / selling prices are minimal / maximal probabilities of a **confidence ranking**

Assumptions:

For any bets on events A with stakes S :

- ▶ you have a buying price $\in \underline{q}_S(A)S$ and a selling price $\in \overline{q}_S(A)S$
- ▶ quotients $\underline{q}_S(A)$, $\overline{q}_S(A)$ may depend on stakes
- ▶ whatever price you accept for high stakes, you will accept for lower stakes

Stakes Dependence $T \leq S \Rightarrow \frac{P(\frac{S}{S_X} X)}{S} \leq \frac{P(\frac{T}{S_X} X)}{T}$

Stakes Independence, in the language of previsions

Coherent lower previsions:

Constant additivity $\underline{P}(X + c1) = \underline{P}(X) + c$

Positive homogeneity $\underline{P}(\alpha X) = \alpha \underline{P}(X)$

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Fact (Grant and Polak, 2013)

Constant additivity \Leftrightarrow Constant absolute uncertainty aversion

Positive homogeneity \Leftrightarrow Constant relative uncertainty aversion

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- ▶ Both lack normative justification

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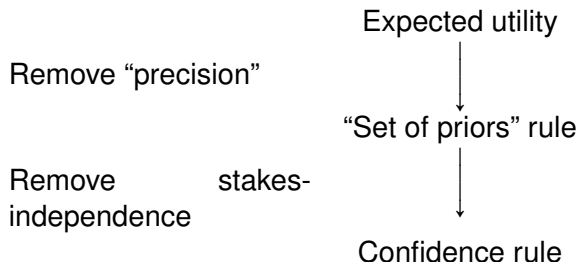
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- ▶ Neither hold descriptively (Baillon and Placido, 2019; Abdellaoui et al., 2024)
 - ▶ Both lack normative justification
- Confidence decision maxim justifies potential violation of both

General moral



Conclusion

The basic issue between the confidence family and fixed set of priors models:

- ▶ is stakes-independence a rationality constraint?

Desiderata

- ✓ it has acceptable choice-theoretical consequences
- 2. it is conceptually clear about the roles of different mental attitudes

And of course:

- ✓ it can fruitfully deal with “severe uncertainty” situations such as those above.

Separation of beliefs and desires: Maxmin EU

$$(\min_{p \in \mathcal{C}} \mathbb{E}_p u(f(s)))$$

Elements of the model:

- ▶ Utility function
- ▶ Set of priors

Separation of beliefs and desires: Maxmin EU

$$(\min_{p \in \mathcal{C}} \mathbb{E}_p u(f(s)))$$

Elements of the model:

- ▶ Utility function = **Desires over outcomes**
- ▶ Set of priors = **Beliefs?**

Separation of beliefs and desires: Maxmin EU

$$(\min_{p \in C} \mathbb{E}_p u(f(s)))$$

Elements of the model:

- ▶ Utility function = Desires over outcomes
- ▶ Set of priors = **Beliefs?**

There is a natural comparison of attitude to uncertainty

- ▶ DM 1 is more averse to uncertainty than DM 2 if $\forall f$, constant acts c , $f \succeq_1 c \Rightarrow f \succeq_2 c$.

Separation of beliefs and desires: Maxmin EU

$$(\min_{p \in \mathcal{C}} \mathbb{E}_p u(f(s)))$$

Elements of the model:

- ▶ Utility function = Desires over outcomes
- ▶ Set of priors = **Beliefs?**

For DMs with the same u , DM 1 is more averse to uncertainty iff $\mathcal{C}_1 \supseteq \mathcal{C}_2$. (Ghirardato and Marinacci, 2002)

- ▶ DM 1 is more averse to uncertainty than DM 2 if $\forall f$, constant acts c , $f \geq_1 c \Rightarrow f \geq_2 c$.

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Policy example

Policy maker 1's set of priors, $\mathcal{C}_1 \supseteq \mathcal{C}_2$, Policy maker 2's set.

- ▶ Does 2 have further information / beliefs?
- ▶ Or is he just less cautious?

Separation of beliefs and desires: Maximin EU

$$(\min_{p \in \mathcal{C}} \mathbb{E}_p u(f(s)))$$

Elements of the model:

- ▶ Utility function = Desires over outcomes
- ▶ Set of priors = Beliefs?

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Policy example

Policy maker 1's set of priors, $\mathcal{C}_1 \supseteq \mathcal{C}_2$, Policy maker 2's set.

- ▶ Does 2 have further information / beliefs?
- ▶ Or is he just less cautious?

Maximin EU can't decide the question ... or even properly represent the possibilities.

Separation of beliefs and desires: Confidence

Elements of the model:

- ▶ Utility function
- ▶ Confidence ranking
- ▶ Cautiousness coefficient
- ▶ Decision rule

Separation of beliefs and desires: Confidence

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- ▶ Utility function = **Desires over outcomes**
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Separation of beliefs and desires: Confidence

Elements of the model:

- ▶ Utility function = Desires over outcomes
- ▶ Confidence ranking = Beliefs and confidence in beliefs
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- ▶ Decision rule

Separation of beliefs and desires: Confidence

Elements of the model:

- ▶ Utility function = Desires over outcomes
- ▶ Confidence ranking = Beliefs and confidence in beliefs
- ▶ Cautiousness coefficient = **Attitude to choosing on the basis of limited confidence**
- ▶ Decision rule

Separation of beliefs and desires: Confidence

Elements of the model:

- ▶ Utility function = Desires over outcomes
- ▶ Confidence ranking = Beliefs and confidence in beliefs
- ▶ Cautiousness coefficient = Attitude to choosing on the basis of limited confidence
- ▶ Decision rule: Maxmin EU

And there is a natural comparison of attitude to uncertainty

- ▶ DM 1 is more averse to uncertainty than DM 2 if $\forall f$, constant c ,
 $f \geq_1 c \Rightarrow f \geq_2 c$.

Separation of beliefs and desires: Confidence

Elements of the model:

- ▶ Utility function = Desires over outcomes
- ▶ Confidence ranking = Beliefs and confidence in beliefs
- ▶ Cautiousness coefficient = Attitude to choosing on the basis of limited confidence
- ▶ Decision rule: Maxmin EU

And there is a natural comparison of attitude to uncertainty that corresponds precisely to differences in the cautiousness coefficient.

For DMs with the same u and Ξ

- ▶ 1 is more averse to uncertainty

$\Leftrightarrow D_1(f) \supseteq D_2(f)$ for all acts f .

Separation of beliefs and desires: Confidence

Elements of the model:

- ▶ Utility function = Desires over outcomes
- ▶ Confidence ranking = Beliefs and confidence in beliefs
- ▶ Cautiousness coefficient = Attitude to choosing on the basis of limited confidence
- ▶ Decision rule: Unanimity

There is a natural comparison of decisiveness

- ▶ DM 1 is more indecisive than DM 2 if $\forall f, g \ f \succeq_1 g \Rightarrow f \succeq_2 g$.

Separation of beliefs and desires: Confidence

Elements of the model:

- ▶ Utility function = Desires over outcomes
- ▶ Confidence ranking = Beliefs and confidence in beliefs
- ▶ Cautiousness coefficient = Attitude to choosing on the basis of limited confidence
- ▶ Decision rule: Unanimity

There is a natural comparison of decisiveness
that corresponds precisely to differences in the cautiousness
coefficient

For DMs with the same u and Ξ

1 is more indecisive

$\Leftrightarrow D_1((f, g)) \supseteq D_2((f, g))$ for all pairs f and g .

Separation of beliefs and desires

Elements of the model:

- ▶ Utility function = Desires over outcomes
- ▶ Confidence ranking = Beliefs and confidence in beliefs
- ▶ Cautiousness coefficient = Attitude to choosing on the basis of limited confidence
- ▶ Decision rule

Conclusion There is a clean separation between beliefs and desires (attitudes to outcomes and to choosing in the absence of confidence).

Comparison Maxmin EU, unanimity preferences, as well as many other “non-EU” models of decision making, do not exhibit such a separation.

In summary

Belief representation & Decision

1. it has acceptable choice-theoretical consequences
2. it involve a neat separation of beliefs and tastes

And of course:

3. it can fruitfully deal with “severe uncertainty” situations such as those above.

In summary

Belief representation & Decision

1. it has acceptable choice-theoretical consequences
2. it involve a neat separation of beliefs and tastes

And of course:

3. it can fruitfully deal with “severe uncertainty” situations such as those above.

Moreover

4. it is involves an ordinal notion of confidence (useful wrt tractability)
 - It is the only model (I am aware of) with 2, 3 and 4.

▶ Ambiguity models

▶ Uncertainty Attitudes

▶ Climate policy

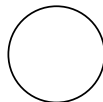
Ambiguity models

Decision models can be categorised by their 'belief' component

Model	Belief repr
Maxmin EU & al	Set of priors

Maxmin-EU

$$\min_{p \in \mathcal{C}} \mathbb{E}_p(u(f))$$



Ambiguity models

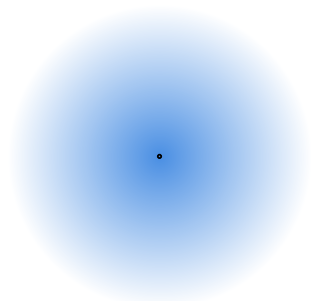
Decision models can be categorised by their 'belief' component

Model	Belief reprn
Maxmin EU & al	Set of priors
Multiplier / Variational	Function on prob. space

Multiplier / Variational

$$\min_{p \in \Delta} (\mathbb{E}_p(u(f)) + c(p))$$

Δ



Ambiguity models

Decision models can be categorised by their 'belief' component

Model	Belief repr
Maxmin EU & al	Set of priors
Multiplier / Variational	Function on prob. space

Multiplier / Variational

$$\min_{p \in \Delta} (\mathbb{E}_p(u(f)) + c(p))$$

- ? normative plausibility
- X belief-taste separation
- X ordinal at the 2nd-order level

Ambiguity models

Decision models can be categorised by their 'belief' component

Model	Belief reprn
Maxmin EU & al	Set of priors
Multiplier / Variational Smooth	Function on prob. space 2nd-order prob.

Smooth $\mathbb{E}_{\mu} \phi(\mathbb{E}_p(u(f)))$

Δ

Ambiguity models

Decision models can be categorised by their 'belief' component

Model	Belief repr
Maxmin EU & al	Set of priors
Multiplier / Variational Smooth	Function on prob. space 2nd-order prob.

Smooth $\mathbb{E}_{\mu} \phi(\mathbb{E}_p(u(f)))$

- $\times?$ normative plausibility
- $\checkmark?$ belief-taste separation
- \times ordinal at the 2nd-order level

Ambiguity models

Decision models can be categorised by their 'belief' component

Model	Belief repr
Maxmin EU & al	Set of priors
Confidence	Confidence Ranking
Multiplier / Variational	Function on prob. space
Smooth	2nd-order prob.

Uncertainty attitudes

- ▶ Utility function : Desires
- ▶ Confidence ranking : Beliefs and confidence in beliefs
- ▶ Cautiousness coefficient : Uncertainty Attitudes
- ▶ **Decision rule**

Uncertainty attitudes

- ▶ Utility function : Desires
- ▶ Confidence ranking : Beliefs and confidence in beliefs
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- ▶ **Decision rule**

Example: α -maxmin EU

$$\alpha \min_{p \in D(f)} \mathbb{E}_p(u(f)) + (1 - \alpha) \max_{p \in D(f)} \mathbb{E}_p(u(f))$$

Uncertainty attitudes

- ▶ Utility function : Desires
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$$\alpha \min_{p \in D(f)} \mathbb{E}_p(u(f)) + (1 - \alpha) \max_{p \in D(f)} \mathbb{E}_p(u(f))$$

- ▶ α : attitude to uncertainty / imprecision.

Uncertainty attitudes

- ▶ Utility function : Desires
- ▶ Confidence ranking : Beliefs and confidence in beliefs
- ▶ Cautiousness coefficient : Uncertainty Attitudes
- ▶ **Decision rule : Uncertainty Attitudes**

Example: α -maxmin EU

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- ▶ α : attitude to uncertainty / imprecision.
 - ▶ Prudence; concern for robustness in decision
 - ▶ Value assigned to the “Pragmatic”

Uncertainty attitudes

- ▶ Utility function : Desires
- ▶ Confidence ranking : Beliefs and confidence in beliefs
- ▶ Cautiousness coefficient : Uncertainty Attitudes
- ▶ Decision rule : Uncertainty Attitudes

Example: α -maxmin EU

$$\alpha \min_{p \in D(f)} \mathbb{E}_p(u(f)) + (1 - \alpha) \max_{p \in D(f)} \mathbb{E}_p(u(f))$$

- ▶ α : attitude to uncertainty / imprecision.
 - ▶ Prudence; concern for robustness in decision
 - ▶ Value assigned to the “Pragmatic”
- ▶ Cautiousness coefficient:
 - ▶ How willing to risk big decisions on fragile beliefs
 - ▶ Value assigned to the “Epistemic”

Uncertainty attitudes

- ▶ Utility function : Desires
- ▶ Confidence ranking : Beliefs and confidence in beliefs
- ▶ Cautiousness coefficient : Uncertainty Attitudes
- ▶ Decision rule : Uncertainty Attitudes

Moral

Decision under (severe) uncertainty involves Uncertainty Attitudes:

- ▶ Value judgements
- ▶ Multi-dimensional

Plan

Introduction

Confidence and Decision

Defense

Climate policy making

Aggregation

Conclusion

Back to: Climate Policy & Uncertainty

Aim Overview of issues & a proposal concerning:

How to decide when there is this much uncertainty?

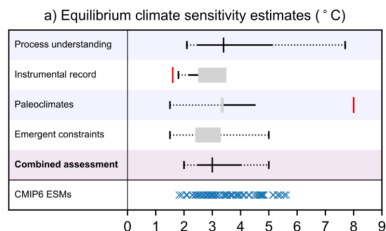


Figure: (Masson-Delmotte et al., 2021, fig TS.16b).

- ▶ Best estimate: 3°C
- ▶ Likely in range 2.5 to 4°C (high confidence)
- ▶ Virtually certain above 1.5°C (high confidence)
- ▶ Very likely below 5°C (medium confidence)

(Masson-Delmotte et al., 2021)

Climate Policy, Uncertainty & Confidence

IPCC assessments:

- ▶ translate directly into confidence rankings
- ▶ are difficult to connect to other uncertainty representations

▶ Reporting and modelling

▶ Decision

▶ Conclusion

Climate Policy, Uncertainty & Confidence

Working with Uncertainty

The confidence framework:

Climate Policy, Uncertainty & Confidence

Working with Uncertainty

The confidence framework:

- ▶ provides a (hitherto absent) “logic of confidence”

e.g. ECS likely in 2.5–4°C (high conf.) *implies* that
ECS *unlikely* greater than 6°C (high conf.)

e.g. the ECS statements are consistent

Climate Policy, Uncertainty & Confidence

Working with Uncertainty

The confidence framework:

- ▶ provides a (hitherto absent) “logic of confidence”
- ▶ provides the tools for systematic propagation to “downstream” modelling

e.g. standard imprecise prob. tools *at each level*

and “lifting” rules: if held with high confidence then held with (at least) medium confidence

Climate Policy, Uncertainty & Confidence

Working with Uncertainty

The confidence framework:

- ▶ provides a (hitherto absent) “logic of confidence”
- ▶ provides the tools for systematic propagation to “downstream” modelling
- ▶ provides recommendations for future reporting practices

e.g. report confidence in a likelihood assessment

not confidence before likelihood assessment

Climate Policy, Uncertainty & Confidence

Decision

The confidence framework:

Climate Policy, Uncertainty & Confidence

Decision

The confidence framework:

- ▶ shows how IPCC reports can be used to guide decision making

- ▶ Belief / value separation:
 - ▶ policy makers choose the appropriate confidence level for the decision at hand
 - ▶ they use the assessments made with that much confidence to decide
 - ▶ they select the appropriate 'pragmatic' uncertainty attitude (robustness etc.)

Climate Policy, Uncertainty & Confidence

Decision

The confidence framework:

- ▶ shows how IPCC reports can be used to guide decision making
 - ▶ is a pragmatic vindication of IPCC practices
-
- ▶ Solid normative credentials for the confidence model
- justification of the IPCC framework

Climate Policy, Uncertainty & Confidence

Decision

The confidence framework:

- ▶ shows how IPCC reports can be used to guide decision making
- ▶ is a pragmatic vindication of IPCC practices
- ▶ provides practical recommendations for the future use of the language

e.g. policy maker implication in the choice of confidence level to report

e.g. reporting at multiple levels

Confidence representations

Given a set of (IPCC) confidence statements:

→ present the nested family of distributions satisfying them.

▸ Confidence calibration

▸ Conclusion

Confidence representations

Given a set of (IPCC) confidence statements:

→ present the nested family of distributions satisfying them.

Web tool does this:

- ▶ uses the logic just discussed
- ▶ detects inconsistencies
- ▶ traces p-boxes for each confidence level

Also:

- ▶ Elicitation of confidence in beliefs

`http://confidence.hec.fr/app/`

Plan

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Confidence Aggregation

Expertise

Conclusion

Probabilistic belief aggregation

i (honest, well-intentioned) experts; 0 group.

Only differ in beliefs (same u^i).

Probabilistic belief aggregation

i (honest, well-intentioned) experts; 0 group.

Only differ in beliefs (same u^i).

Linear opinion pooling For all events E :

$$p^0(E) = \sum_{i=1}^n w^i p^i(E)$$

for weights w^i .



Pareto $f \geq^i g$ for all $i \Rightarrow f \geq^0 g$

- ▶ Respect (issue-level) consensus

(Mongin, 1995)

Challenges

Example

Challenges

Example

Probability certain interest rate rise has limited effect on:

	Labour	Real estate	Both
Laura	0.9	0.1	0.09
Ray	0.1	0.9	0.09
Lin. pool.	$0.1 + 0.8w^L$	$0.9 - 0.8w^L$	0.09

$$p^0(E) = w^L p^L(E) + (1 - w^L) p^R(E)$$

Challenges

Example

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$$p^0(E) = w^L p^L(E) + (1 - w^L) p^R(E)$$

Spurious Unanimity

- ▶ Why respect spurious consensus?

(Mongin, 2016; Mongin and Pivato, 2020; Dietrich, 2021; Bommier et al., 2021)

Challenges

Example

Probability certain interest rate rise has limited effect on:

	Labour	Real estate	Both
Laura	0.9	0.1	0.09
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$$p^0(E) = w^L p^L(E) + (1 - w^L) p^R(E)$$

Spurious Unanimity

- ▶ Why respect spurious consensus?

Diverse (intra-agent) expertise

- ▶ Why is Laura's judgement on Labour treated the same as her judgement on Real-estate?

(Genest and Zidek, 1986; French, 1985)

Challenges

Summary

Desiderata A belief aggregation procedure that:

1. respects the right consensus(es)
 - ▶ avoiding spurious unanimities
2. can do justice to varying expertise

Proposal

Insights

Issue-level
consensus

Spuriousness

Desideratum A belief aggregation procedure that:

1. respects the right consensus(es)
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Spuriousness



Evidence,
reasons,
information

$p^i(E)$ does not exhaust the elements of belief states pertaining to event E relevant for aggregation . . .

Proposal

Insights

Issue-level
consensus

Spuriousness



Evidence,
reasons,
information



Confidence

$p^i(E)$ does not exhaust the elements of belief states pertaining to event E relevant for aggregation . . .

Confidence in beliefs: Hill, 2013; ?; ?; Bradley, 2017

but also Klibanoff et al., 2005a; Maccheroni et al., 2006a; Hansen and Sargent, 2008; Chateauneuf and Faro, 2009a

Proposal

Insights

Issue-level
consensus

Spuriousness



Evidence,
reasons,
information



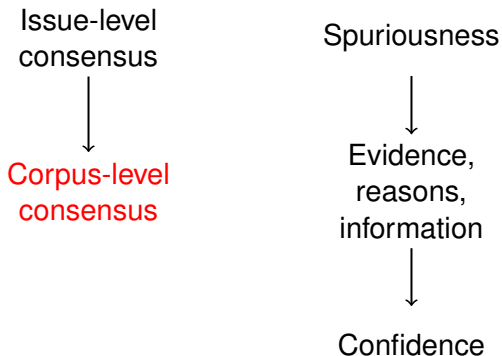
Confidence

Desideratum A belief aggregation procedure that:

1. respects the right consensus(es)
 - ▶ avoiding spurious unanimities

Proposal

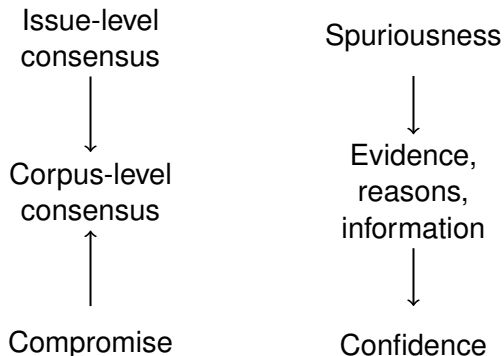
Insights



Corpus: (coherent) set of probability judgements \equiv set of priors.

Proposal

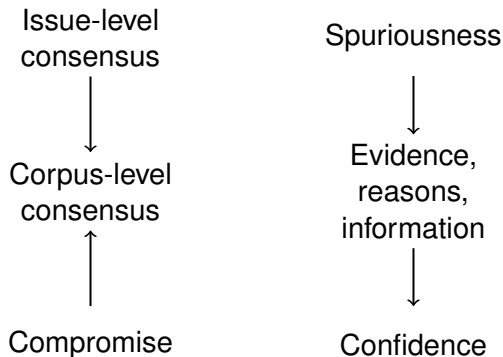
Insights



Corpus-level consensus: Everyone is willing to 'leave off the table' or compromise any potential disagreement.

Proposal

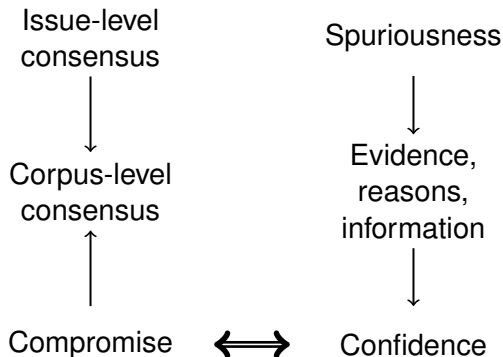
Insights



What compromises are agents willing to make?

Proposal

Insights



Confidence and Compromise

The more confident an individual is in a belief, the less willing she is to compromise on it.

In aggregation:

Respect corpus-level consensus

where

The more confident an individual is in a belief,
the less willing she is to compromise on it.

Aim

- ▶ Develop such an aggregation rule
- ▶ Spurious unanimity, expertise . . . and beyond.

Plan

Introduction

Confidence and Decision

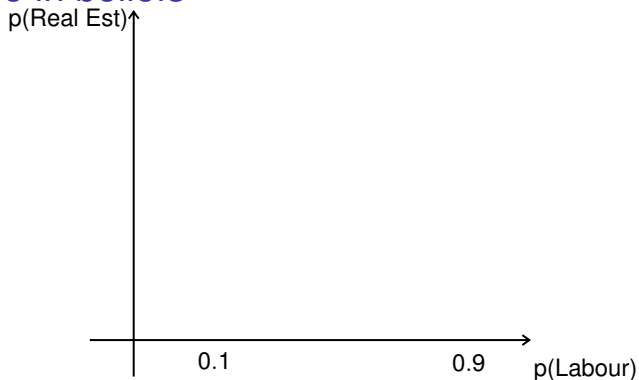
Defense

Climate policy making

Aggregation

Conclusion

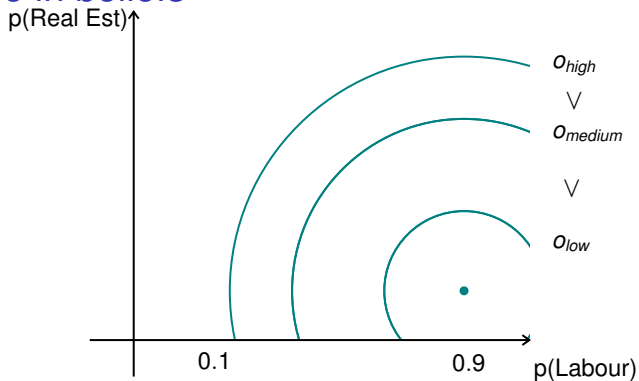
Confidence in beliefs



Preliminaries

- Δ probability measures (over states Ω)
- $(O, >)$ confidence levels

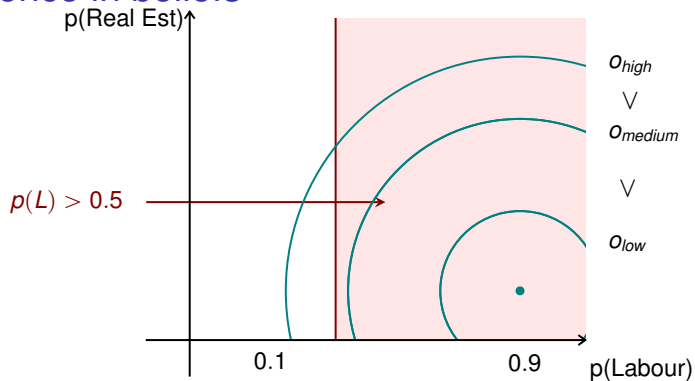
Confidence in beliefs



Confidence ranking: Increasing $c^i : O \rightarrow 2^\Delta \setminus \emptyset$.

(Hill, 2013; ?; Manski, 2013; Bradley, 2017)

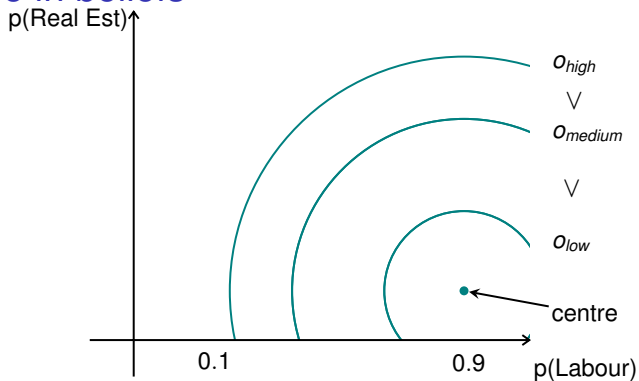
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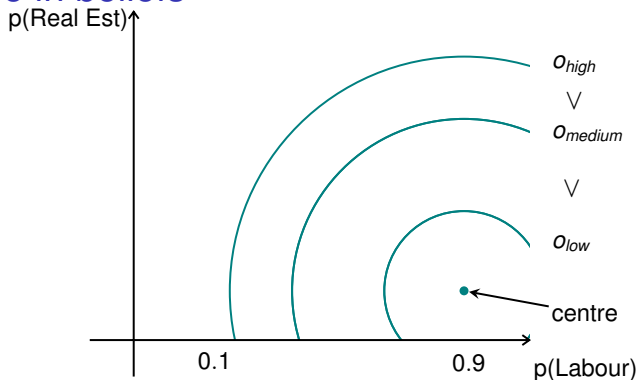
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Confidence in beliefs



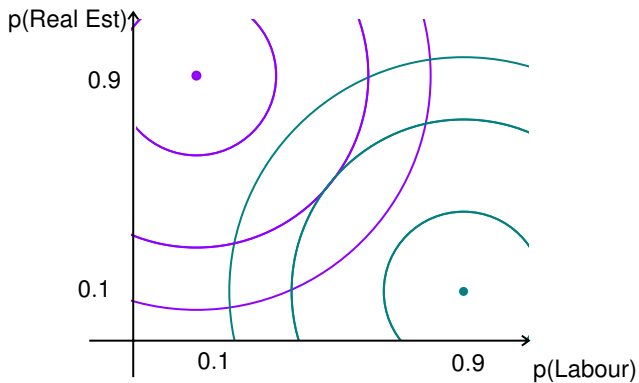
Confidence ranking: Increasing $c^i : \mathcal{O} \rightarrow 2^\Delta \setminus \emptyset$.

Implausibility fn $\iota : \Delta \rightarrow \mathcal{O} \cup \emptyset$

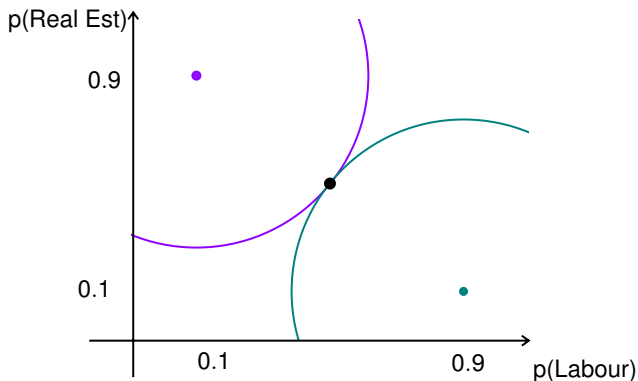
(Hill, 2013; ?; Manski, 2013; Bradley, 2017)^a

^aReduced form for Klibanoff et al., 2005a; Maccheroni et al., 2006a; Hansen and Sargent, 2008; Chateauneuf and Faro, 2009a ...

Consensus



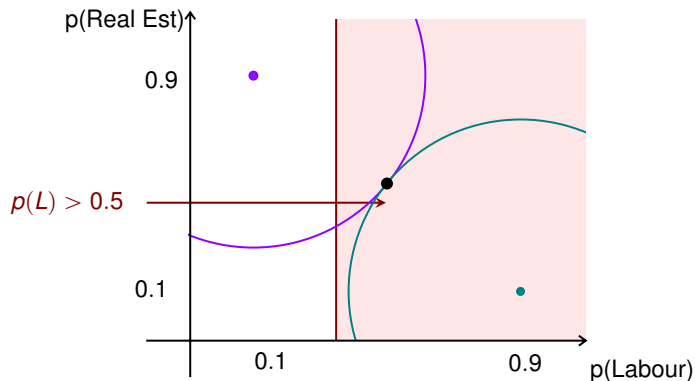
Consensus



Consensus: a coherent set of probability judgements rejected by no-one (at the relevant confidence levels).

$$\bigcap_i c^i(o)$$

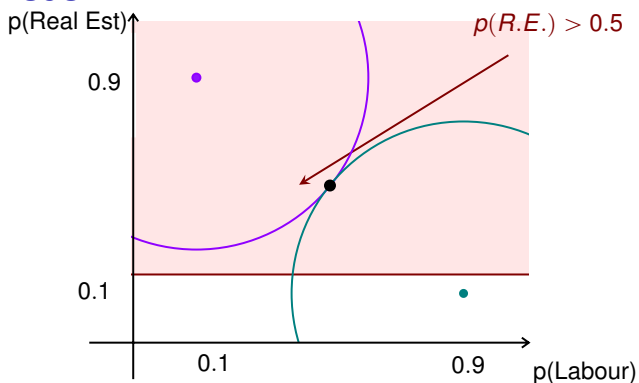
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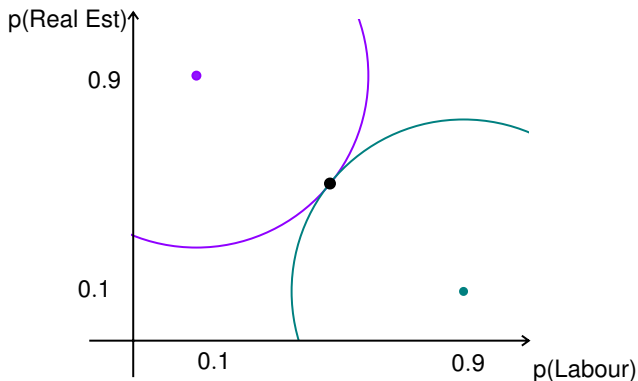
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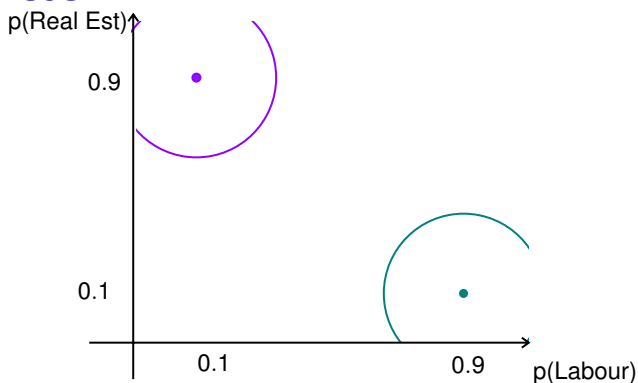
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Maxim If more confident in a belief, less willing to compromise it.

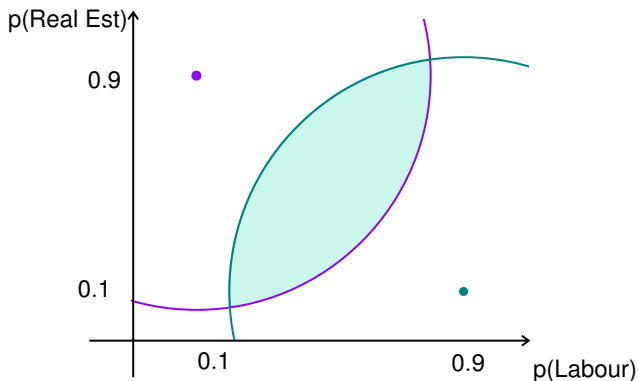
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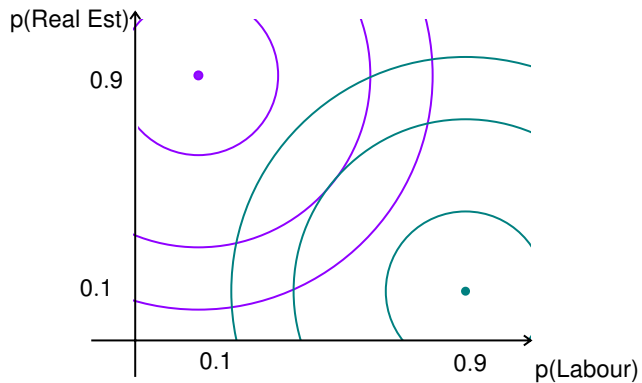
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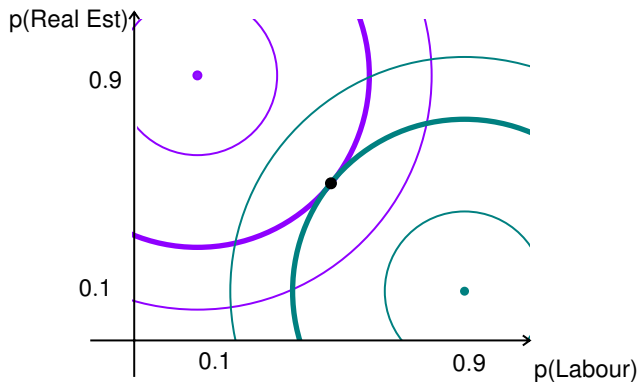
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Confidence aggregation

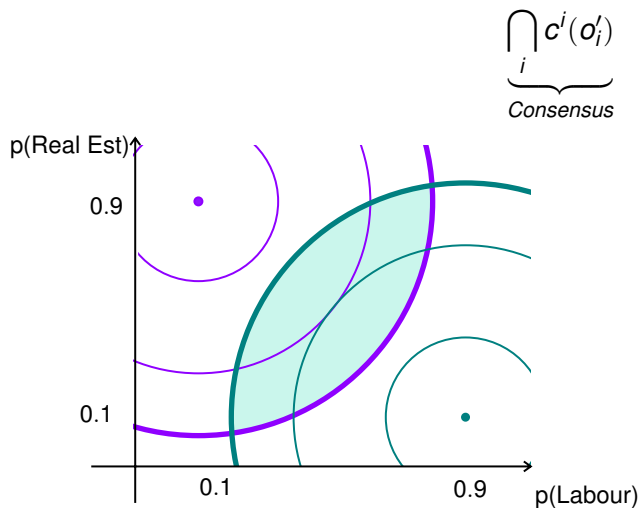


Confidence aggregation

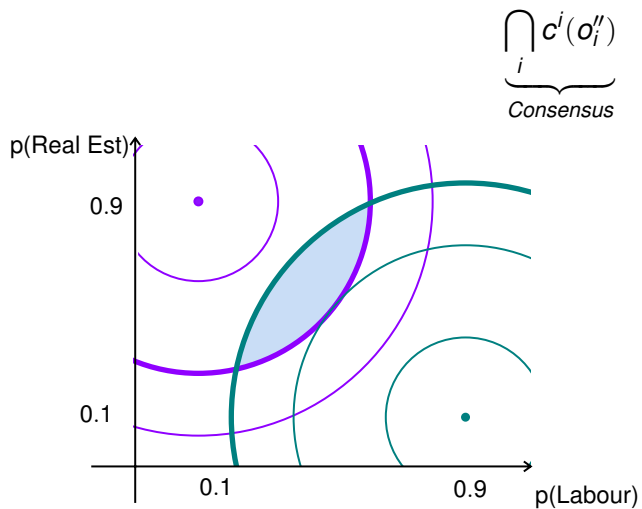
$$\underbrace{\bigcap_i c^j(o_i)}_{\text{Consensus}}$$



Confidence aggregation

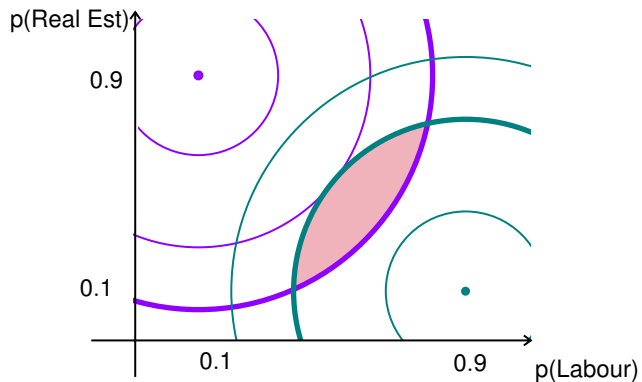


Confidence aggregation



Confidence aggregation

$$\underbrace{\bigcap_i c^i(o_i''')}_{\text{Consensus}}$$



Confidence aggregation

$$\underbrace{\bigcap_i c^j(o_i)}_{\text{Consensus}}$$

$\otimes : O^n \rightarrow O$: confidence level aggregator.

- ▶ $\otimes o$: group confidence in consensus judgements in o
- ▶ monotonic

Confidence aggregation

$$\underbrace{\bigcap_i c^j(o_i)}_{\text{Consensus}}$$

$\otimes : O^n \rightarrow O$: confidence level aggregator.

- ▶ $\otimes o$: group confidence in consensus judgements in o
- ▶ monotonic

E.g.

Maximum agg. $\otimes o = \max \{o_i\}$

Minimum agg. $\otimes o = \min \{o_i\}$

Average agg. $\otimes o = \sum \frac{1}{n} o_i + \chi$

Confidence aggregation

$$F_{\otimes}(c^1, \dots, c^n)(o) = \bigcup_{o: \otimes o \leq o} \underbrace{\bigcap_i c^i(o_i)}_{\text{Consensus}}$$

for o with $\bigcap_i c^i(o_i) \neq \emptyset$

Group judgement: held in all consensuses with that level of confidence.

Confidence aggregation

Consensus-preserving confidence aggregation:

$$F_{\otimes}(c^1, \dots, c^n)(o) = \bigcup_{o: \otimes o \leq o} \bigcap_i c^i(o_i)$$

The more individual confidence there is in a consensus judgement, the more confidence the group has in it.

Confidence aggregation

Consensus-preserving confidence aggregation:

$$F_{\otimes}(c^1, \dots, c^n)(o) = \bigcup_{o: \otimes o \leq o} \bigcap_i c^i(o_i)$$

Equivalently:

$$F_{\otimes}(l^1, \dots, l^n)(p) = \otimes(l^1(p), \dots, l^n(p))$$

So

$$\text{Centre}_{F_{\otimes}(c^1, \dots, c^n)} = \arg \min_{p \in \Delta} \otimes(l^1(p), \dots, l^n(p))$$

$$\stackrel{\text{avge}}{=} \otimes \arg \min_{p \in \Delta} \sum_{i=1}^n l^i(p)$$

Plan

Introduction

Confidence and Decision

Defense

Climate policy making

Aggregation

Conclusion

Pooling & Confidence

Probabilities p^i

Pooling & Confidence

Probabilities p^i : *stipulate* confidence rankings centred on p^i

$$c^i(o) = \left\{ q \in \Delta : w^i \rho(q, p^i) \leq o \right\}$$

ρ : distance^a

E.g.

Euclidean $\rho(q, p) = \sum_{\omega \in \Omega} (q(\omega) - p(\omega))^2$

Relative Entr. $\rho(q, p) = R(q \| p)$

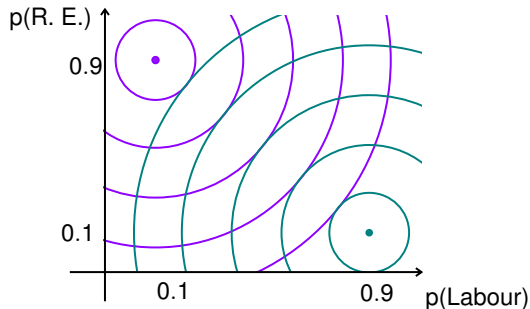
Reverse Rel. Entr. $\rho(q, p) = R(p \| q)$

^alower semicts; $\rho(q, p) = 0 \Leftrightarrow p = q$.

Pooling & Confidence

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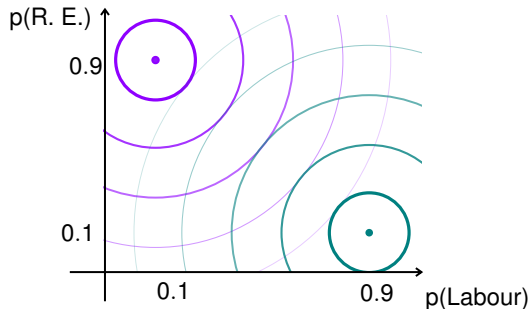


ρ : Euclidean

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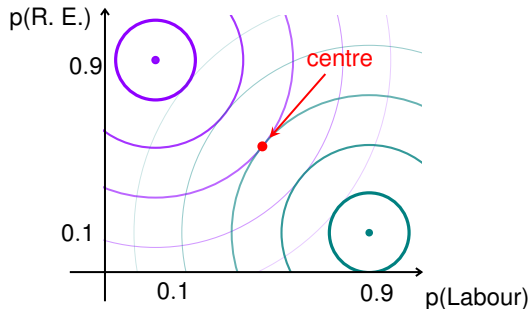
$$w^L = w^R$$

ρ : Euclidean

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Confidence aggregation

$$w^L = w^R$$

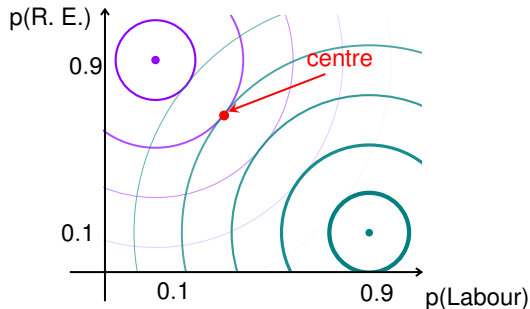
ρ : Euclidean

Average \otimes

Pooling & Confidence

Probabilities p^i : stipulate confidence rankings centred on p^i

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Confidence aggregation

$$w^L < w^R$$

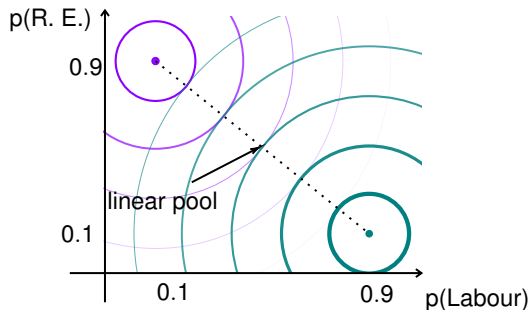
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Confidence aggregation

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Pooling & Confidence

Probabilities p^i : *stipulate* confidence rankings centred on p^i

$$c^i(o) = \left\{ q \in \Delta : w^i \rho(q, p^i) \leq o \right\}$$

Theorem

Centre of confidence aggregation = result of pooling rule

<i>Generating distance</i>	<i>Pooling rule</i>
<i>Euclidean</i>	<i>Linear</i>
<i>Relative Entropy</i>	<i>Geometric</i>
<i>Reverse Rel. Entr.</i>	<i>Linear</i>

with weights $\frac{w^i}{\sum_{i=1}^n w^i}$.

Pooling & Confidence

Probabilities p^i : *stipulate* confidence rankings centred on p^i

$$c^i(o) = \{q \in \Delta : w^i \rho(q, p^i) \leq o\}$$

Theorem

Centre of confidence aggregation = result of pooling rule

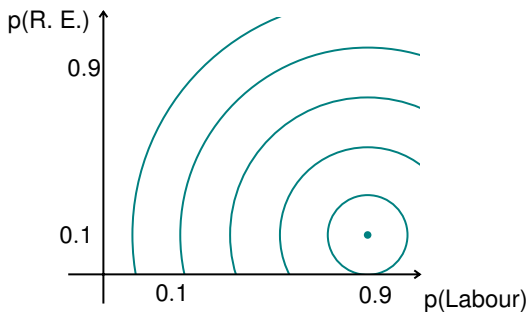
Moral Linear pooling =

- ▶ special case of confidence aggregation
- ▶ **corresponding to assumptions on individuals' confidence.**

Expertise

Euclidean generated confidence ranking:

$$c^L(o) = \left\{ q \in \Delta : w^L \sum_{s \in S'} (q(s) - p^L(s))^2 \leq o \right\}$$



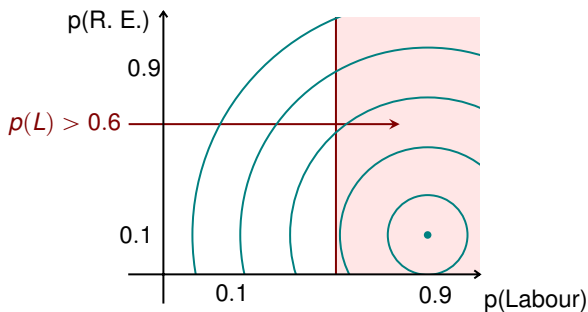
Expertise

Euclidean generated confidence ranking:

$$c^L(o) = \left\{ q \in \Delta : w^L \sum_{s \in S'} (q(s) - p^L(s))^2 \leq o \right\}$$

Fact Euclidean generated distance assumes

- ▶ divergence from p^j on Labour \equiv divergence on R. E.



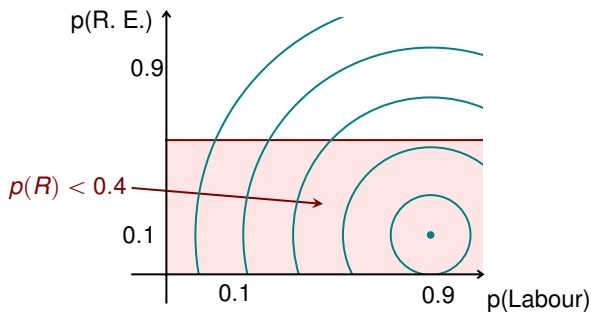
Expertise

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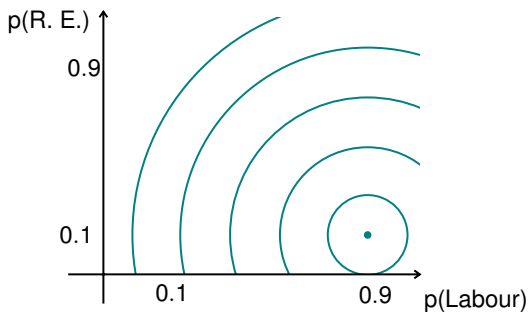
Expertise

Euclidean generated confidence ranking:

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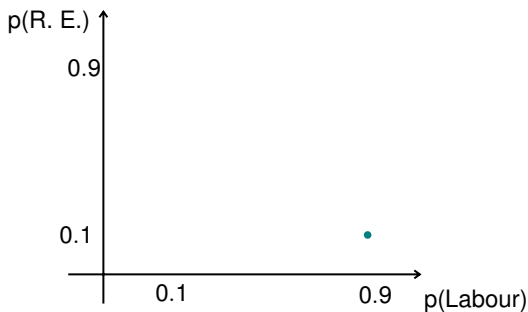
Fact Euclidean generated distance assumes

- ▶ divergence from p^j on Labour \equiv divergence on R. E.
- same confidence on all issues!



Expertise

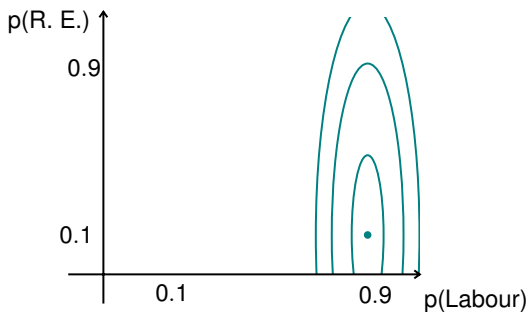
$$c^L(o) = \left\{ q \in \Delta : \begin{array}{l} w_L^L(q(L) - p^L(L))^2 \\ + w_R^L(q(R) - p^L(R))^2 \leq o \end{array} \right\}$$



Expertise

$$c^L(o) = \left\{ q \in \Delta : w_L^L (q(L) - p^L(L))^2 + w_R^L (q(R) - p^L(R))^2 \leq 0 \right\}$$

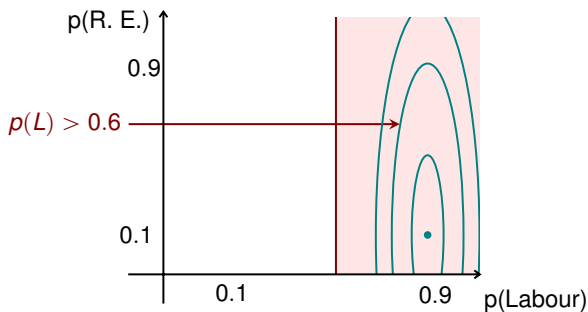
$$w_L^L > w_R^L$$



Expertise

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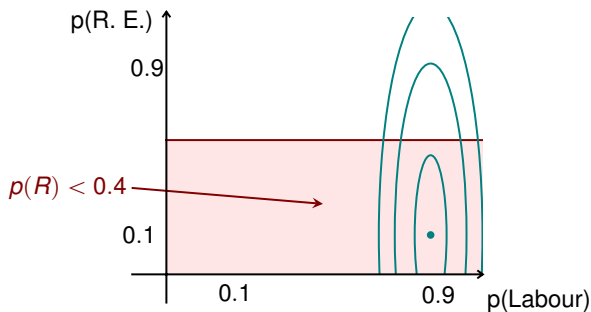
Fact $w_L^L > w_R^L$: more confident in Labour judgements



Expertise

$$c^L(o) = \left\{ q \in \Delta : w_L^L (q(L) - p^L(L))^2 + w_R^L (q(R) - p^L(R))^2 \leq o \right\}$$

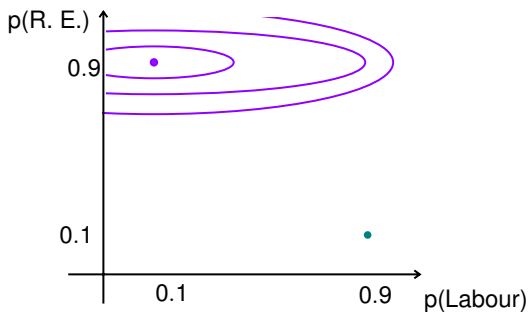
Fact $w_L^L > w_R^L$: more confident in Labour judgements



Expertise

$$c^R(o) = \left\{ q \in \Delta : w_L^R (q(L) - p^L(L))^2 + w_R^R (q(R) - p^L(R))^2 \leq o \right\}$$

$w_L^R < w_R^R$: more confident in Real-Estate judgements



Expertise

Confidence: rich enough to capture diverse expertise.

In general: (w^i, d, p) -generated confidence ranking:

$$c^i(o) = \left\{ q \in \Delta : \sum_{j=1}^m w_j^i d(q|_{\mathcal{P}_j}, p|_{\mathcal{P}_j}) \leq o \right\}$$

where:

\mathcal{P}_j Issues: partitions of Ω

w^i vector of weights

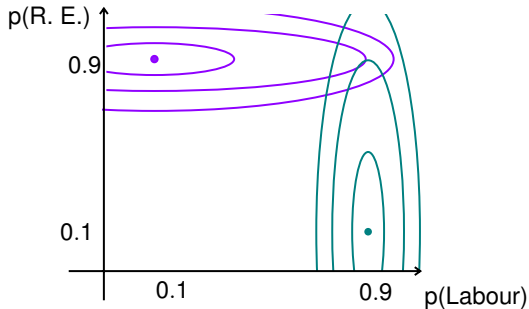
d distance, for each partition

Often, can rewrite e.g.

$$c^i(o) = \{ q \in \Delta : (q - p^i)^T D^i (q - p^i) \leq o \}$$

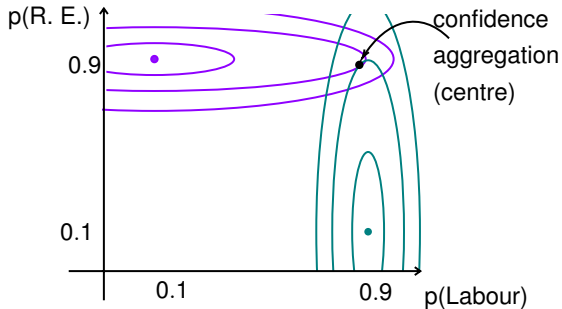
Expertise in Aggregation

Confidence: rich enough to capture diverse expertise.



Expertise in Aggregation

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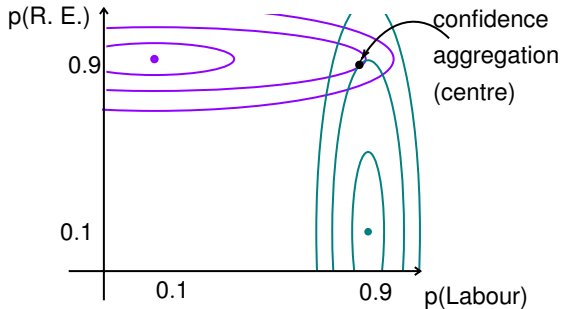


Expertise in Aggregation

Confidence: rich enough to capture diverse expertise.

Confidence aggregation:

- ▶ Does justice to varying expertise (**Desideratum 2**)

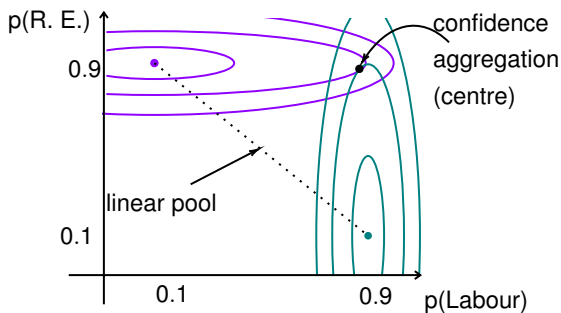


Expertise in Aggregation

Confidence: rich enough to capture diverse expertise.

Confidence aggregation:

- ▶ Does justice to varying expertise (**Desideratum 2**)
- ▶ Does not necessarily respect spurious unanimities (**Desideratum 1**)



Aggregation and Expertise

More generally:

Theorem

Confidence aggregation (w^i, d, p^i) -generated confidence rankings:

$$Centre =_{p \in \Delta} \sum_{i=1}^n \sum_{j=1}^m w_j^i d(p|_{\mathcal{P}_j}, p^i|_{\mathcal{P}_j})$$

When d convex and the issues sufficiently rich: single probability.

- ▶ Copes with within-person expertise diversity & spurious unanimity
- ▶ Some cases: $A_{q \leq r} \sum_{i=L,R} (q - p^i)^T D^i (q - p^i)$
- ▶ Always non-empty (even when issue-dependency)

Aggregation and Expertise

Generates a new probability aggregation rule:

Expert-sensitive pooling

$$F_{\mathcal{P}_1, \dots, \mathcal{P}_m}^d(p^1, \dots, p^n) =_{p \in \Delta} \sum_{i=1}^n \sum_{j=1}^m w_j^i d(p|_{\mathcal{P}_j}, p^i|_{\mathcal{P}_j})$$

for convex d and rich $\{\mathcal{P}_j\}$.

- ▶ Copes with within-person expertise diversity & spurious unanimity
- ▶ Tractable cases
- ▶ Well-defined

Aggregation and Expertise

Generates a new probability aggregation rule:

Expert-sensitive pooling

$$F_{\mathcal{P}_1, \dots, \mathcal{P}_m}^d(p^1, \dots, p^n) =_{p \in \Delta} \sum_{i=1}^n \sum_{j=1}^m w_j^i d(p|_{\mathcal{P}_j}, p^i|_{\mathcal{P}_j})$$

for convex d and rich $\{\mathcal{P}_j\}$.

- ▶ Copes with **within-person expertise diversity** & spurious unanimity
- ▶ Tractable cases
- ▶ **Well-defined**
- ▶ **Resolves a long-standing challenge (Genest and Zidek, 1986; French, 1985).**

Conclusion

Confidence in beliefs has a role in:

Rational Decision

- ▶ Formal model of confidence in beliefs and decision framework
- ▶ Account with attractive conceptual and choice-theoretic properties
- ▶ Belief-value separation
- ▶ Consequences for high-stakes decision making

The higher the stakes involved in a decision, the more confidence is needed in a belief for it to play a role in the decision.

Thank you.

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Further details (all available on the website above):

- ▶ Confidence in Beliefs and Rational Decision Making, *Economics and Philosophy*, 32, 2019.
- ▶ Confidence and Decision, *Games and Economic Behavior*, 82, 2013.
- ▶ Incomplete Preferences and Confidence, *Journal of Mathematical Economics*, 65, 2016.
- ▶ Climate Change Assessments: Confidence, Probability and Decision, *Philosophy of Science*, 84, 2017 (with R. Bradley, C. Helgeson).
- ▶ Combining probability with qualitative degree-of-certainty metrics in assessment, *Climatic Change* 149, 2018 (with R. Bradley, C. Helgeson)
- ▶ Confidence in belief, weight of evidence and uncertainty reporting. *Proceedings of Machine Learning Research*, 103, 2019.
- ▶ Updating Confidence in Beliefs, *Journal of Economic Theory*, 2022.
- ▶ Confidence, consensus and aggregation, *HEC Working Paper*, 2024.

Confidence Elicitation Web Tool <http://confidence.hec.fr/app/>

Confidence in Beliefs and Rational Policy Making

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