

Uniform Sources of Uncertainty for Subjective Probabilities and Ambiguity

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(joint with Aurélien Baillon and Peter
Wakker)

“Informal”

- Central in this work will be the recent finding of the home bias, with investors preferring domestic stocks to foreign stocks in ways beyond uncertainty or utility.
- Here decisions depend on the source of uncertainty (Fox & Tversky, 1995)



Introduction

- Since Keynes (1921) and Knight (1921), economist has argued that in most situations of interest to economics, uncertainty concern one-shot events.
- We owe to de Finetti (1931) and Ramsey (1931) that subjective probabilities can be defined for one-shot events from choice.
- Savage formalized the work of de Finetti and Ramsey providing a *subjective theory of rational choice / probability*.

Introduction

- Two main criticisms against SEU MSOffice1
 - Allais paradox

Table 1: Allais paradox

		100 Balls		
		# 1	#2 - #11	#12 - #100
	<i>A</i>	\$1M	\$1M	\$1M
	<i>B</i>	0	\$5M	\$1M
	<i>A'</i>	\$1M	\$1M	0
	<i>B'</i>	0	\$5M	0

Diapositive 4



MSoftware1 I claimed here that this violation was due to "absolute non-ambiguity pronness".

Abdellaoui Mohammed; 03/04/2007

Introduction

- Ellsberg (three-color) paradox

Table 2: Ellsberg (three-color) paradox

		60 balls		
		30 balls	Black	Yellow
		Red	Black	Yellow
	f	\$1000	0	0
	g	0	\$1000	0
	f'	\$1000	0	\$1000
	g'	0	\$1000	\$1000

Introduction

Msoffice2

□ As a reaction to these criticisms

○ Generalizations of Expected Utility

- CEU (Gilboa, 1987; Schmeidler 1989)
- MEU (Gilboa & Schmeidler 1989)
- CPT (Tversky & Kahneman 1992)

The ‘common denominator’ of these models is a biseparable preference model (Ghirardato & Marinacci 2001).

○ ‘Liberation’ of subjective probability theory from Specific Preference Functionals (Probabilistic Sophistication)

- Machina & Schmeidler (1992)
- Chew & Sagi (2006)

Diapositive 6

MSOffice2 I claimed here that we finally combined the two criticisms to end up with a tool allowing quantitative ambiguity analysis:
"Common denominator" model + "Exchangeability-based PS"

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Preliminaries

□ Savagean setup

- S denotes a state space
- X the outcome set (non-negative real numbers)
- $(E_1: x_1, \dots, E_n: x_n)$ denotes a simple act
- xEy denotes $(E:x, S-E:y)$
- \succsim is a weak order on the set of acts

Preliminaries

MSOffice3

□ We assume a biseparable representation of \succsim

○ Unknown probabilities

- For $x \geq y$, $E \subset S$:

$$V(xEy) = W(E)u(x) + (1-W(E))u(y)$$

- W is a weighting function (a capacity)
- $u: X \rightarrow \mathbb{R}$

○ Known probabilities

- $V(xpy) = w(p)u(x) + (1-w(p))u(y)$
- $w: [0,1] \rightarrow [0,1]$ is the probability weighting function

Diapositive 8

MSOffice3 I insisted here on the neutrality of using the dichotomy: "known" vs. "unknown" probabilities; to avoid the trap of the use of words like ambiguity or uncertainty. This seems simple, but i think that it is pedagogically more helpful. A reading of the introduction of Epstein & Zhang shows how the early use of the word ambiguity is problematic.

Abdellaoui Mohammed; 03/04/2007

Preliminaries

□ Equally likely relation

- $A \sim^* B$ if $xAy \sim xBy$ for some $x \succ y$.

□ Exchangeability

- A and B are exchangeable if exchanging the outcomes under events A and B in an act does not affect its preference value.

$$(xAyBf \sim xByAf)$$

- A partition is exchangeable if all of its elements are mutually exchangeable.

Preliminaries

□ Probabilistic sophistication

- PS holds if there exists a probability measure P on S such that for each act f the only relevant aspect for its preference value is the induced probability distribution.

$$(P_f = P_g) \Rightarrow (f \sim g)$$

- Consequence: The preference of an event captures everything relevant for preference evaluation.

Preliminaries

□ Probabilistic sophistication

Under PS:

- two events are exchangeable iff they have the same probability
- All events in an exchangeable partition (E_1, \dots, E_n) have probability $1/n$.

Bayesian Beliefs for Ellsberg?

□ Ellsberg two-color paradox

○ Two urns:

- Known Urn contains 50 R^k (Red from known) and 50 B^k (Black from known);
- Unknown Urn containing 100 R^u (Red from unknown) and B^u (Black from unknown).

○ Common preferences (under SEU):

- $1000R^k0 > 1000R^u0 \Rightarrow P(R^k) > P(R^u)$
- $1000B^k0 > 1000B^u0 \Rightarrow P(B^k) > P(B^u)$

Bayesian Beliefs for Ellsberg?

□ Reconciliation with Bayesian Beliefs

- Distinguishing two sources (small worlds) of uncertainty.
 - Decision makers has a general dislike of the unknown Urn (source 1) relative to the known Urn (source 2).
 - Similarly, the performance of the Dow Jones index tomorrow can be one source of uncertainty, and the performance of the Nikkei index tomorrow is another.

A source is a collection (an algebra) of events that pertain to a particular mechanism of uncertainty

Bayesian Beliefs for Ellsberg?

□ Reconciliation with Bayesian Beliefs

- The two-color paradox concerned a between source comparison
- Events R^k and B^k are exchangeable and events R^u and B^u are also exchangeable:
 - $(R^k:\$100, B^k:0) \sim (B^k:\$100, R^k:0)$
 - $(R^u:\$100, B^u:0) \sim (B^u:\$100, R^u:0)$
- The above exchangeabilities suggest that the events in question have subjective probability $1/2$ (Chew and Sagi 2006a, 2006b).

Uniform Sources

□ PS and Uniformity

- We call a source *uniform* if PS holds with respect to that source.
- Chew and Sagi (2006a, b) showed that, under some regularity conditions, a source is uniform iff the following conditions hold.
 - i. Events that are equally likely are exchangeable.
 - ii. For each pair of disjoint events, one contains a subset that is exchangeable with the other.
 - iii. For each n there exists an exchangeable n -fold partition.

Uniform Sources

□ Consequence:

- For a rich uniform source, we can elicit subjective probabilities to any degree of precision using a bisection method and using condition ii (comparability).
 - For example we can partition S into two equally likely events E^1_1 and E^1_2 that then must have probability $\frac{1}{2}$.
 - Next we partition E^1_1 into two equally likely events E^2_1 and E^2_2 that must both have probability $\frac{1}{4}$, and we partition E^1_2 into E^2_3 and E^2_4 that also have probability $\frac{1}{4}$.

Uniform Sources

□ Uniformity and Ellsberg three-color paradox

- Assume that an urn contains 30 R balls, and 60 B and Y balls in unknown proportion.

- People prefer betting on R to betting on B ($P(R) > P(B)$);
- People prefer betting on [B or Y] to betting on [R or Y], which contradicts the inequality derived before.

- The (ambiguity) of the urn is not uniform

- Events have different effects and interactions in different configurations, with the weight of Y high in the presence of B but low in the absence of B.

Uniform Sources

□ Combining *biseparability* and PS

○ Unknown Probabilities

- Assume that For $x \geq y$, $E \subset S$:

$$V(xEy) = W(E)u(x) + (1-W(E))u(y)$$

- Under PS, there exists a transformation function w_S such that $W(E) = w_S[P(E)]$.

○ Known Probabilities

- $V(xpy) = w(p)u(x) + (1-w(p))u(y)$

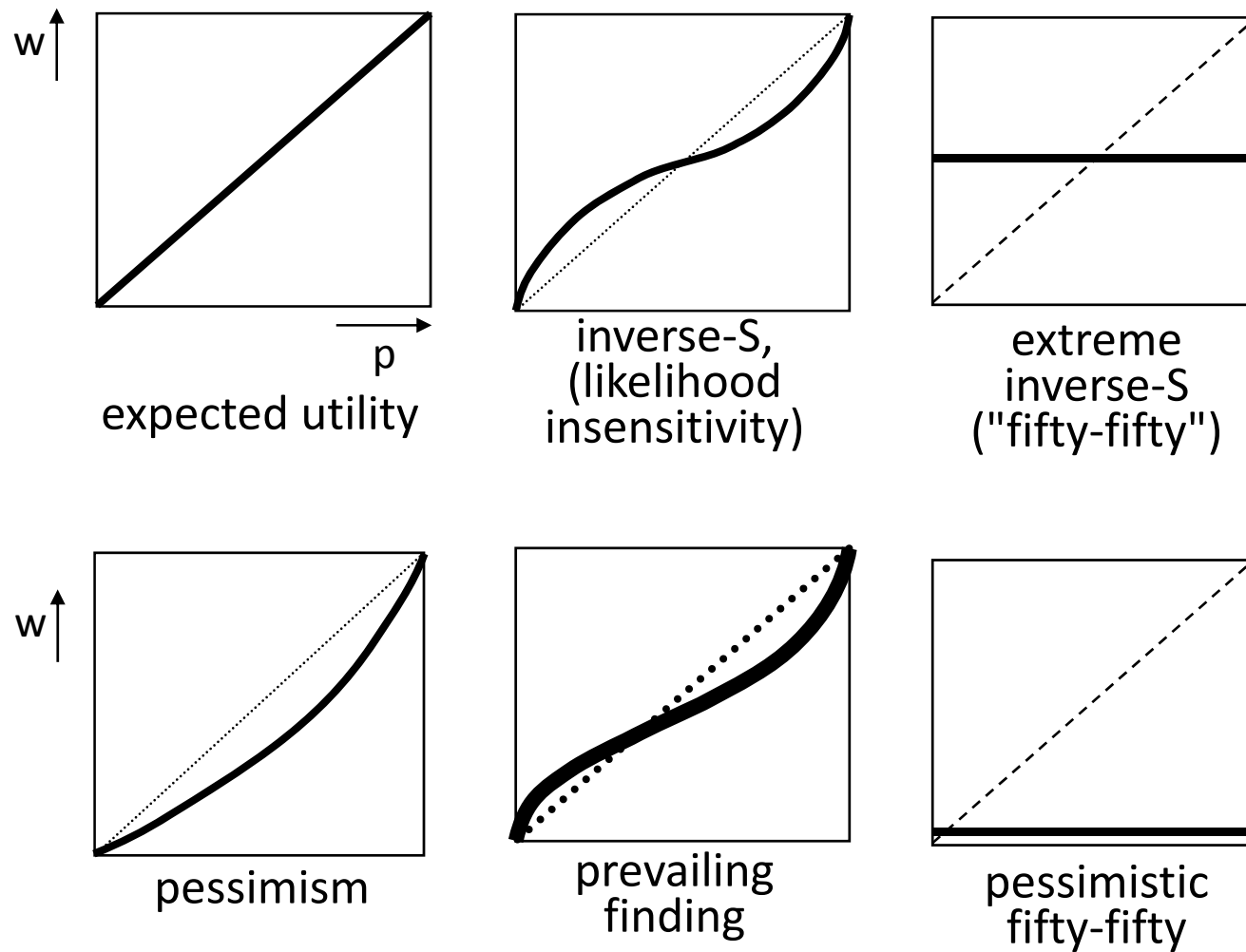
Function w , called the weighting function, is specific to the case of known probabilities.

Ambiguity Attitudes

- Quantifying attitudes towards uncertainty
 - The weighting functions w_s and w provide convenient tools for expressing various attitudes towards uncertainty and risk.
 - Consider two sources of uncertainty S_1 and S_2 and two (corresponding) exchangeable partitions $(A^1_i, i=1, \dots, n)$ and $(A^2_i, i=1, \dots, n)$. $A^j(k)$ is the cumulative union of exch. events in source j .
 - Comparison of willingness to bet on events generated by S_1 and willingness to bet on events generated by S_2 , could quantitatively analyzed through the decision weights

$$w_1(k/n) \text{ and } w_2(k/n), i= 1, \dots, n-1.$$

□ Shapes of probability transformations



Abdellaoui (2000); Bleichrodt & Pinto (2000); Gonzalez & Wu 1999; Tversky & Fox, 1997.

Ambiguity Attitudes

- Tractable quantifications of ambiguity attitudes:
 - The use of probability transformation functions (resulting from the combination of biseparability and PS) allows for easy and intuitive within subject comparisons of attitudes towards uncertainty.
 - Comparative concepts can be defined, with one weighting function being more convex or more inverse-S shaped than another one.

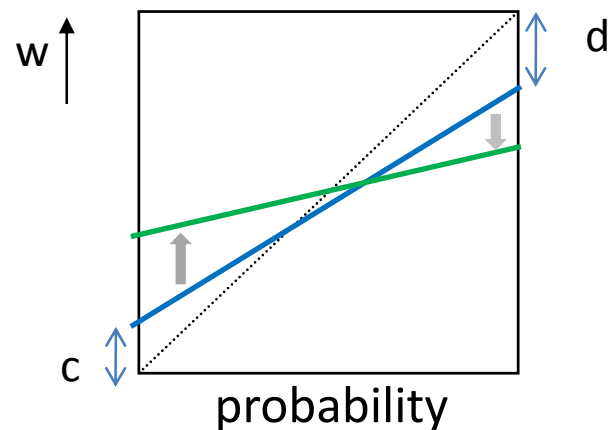
Ambiguity Attitudes

- For empirical purposes, we consider two simple indexes for likelihood insensitivity and pessimism.
 - Assume that
$$w(p) = c + sp$$
on the open interval $(0,1)$ with c the intercept and s the slope.
 - Let $d = 1 - c - s$ be the distance from 1 of the line at $p = 1$ (dual intercept).

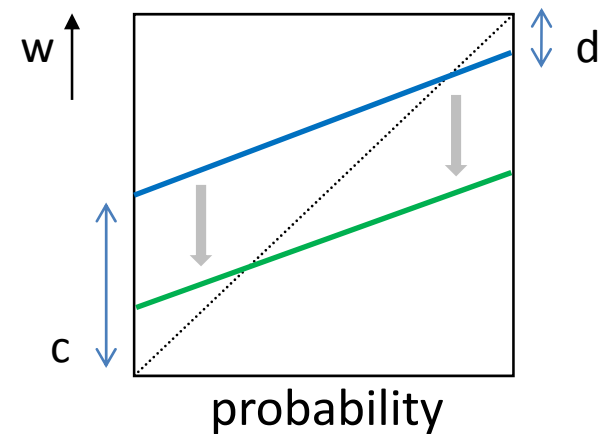
Ambiguity Attitudes

□ Likelihood Sensitivity and Pessimism

- Index of Likelihood insensitivity: $a = c + d$,
- Index of pessimism: $b = d - c$.



Likelihood (in-)Sensitivity



Pessimism

Ambiguity Attitudes

□ Examples

Quantitative indexes of pessimism and likelihood insensitivity

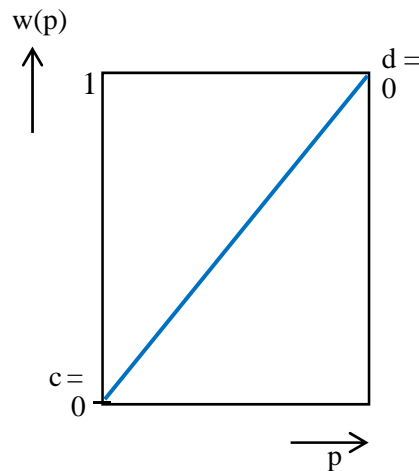


Fig.1.
Insensitivity
index a: 0;
pessimism
index b: 0.

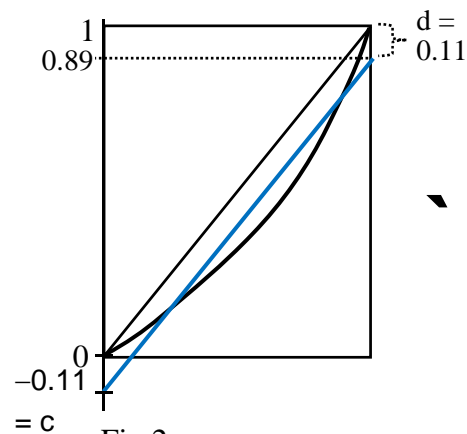


Fig.2.
Insensitivity
index a: 0;
pessimism index
b: 0.22.

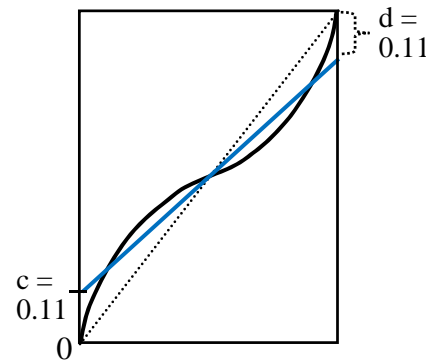


Fig.3.
Insensitivity
index a: 0.22;
pessimism index
b: 0.

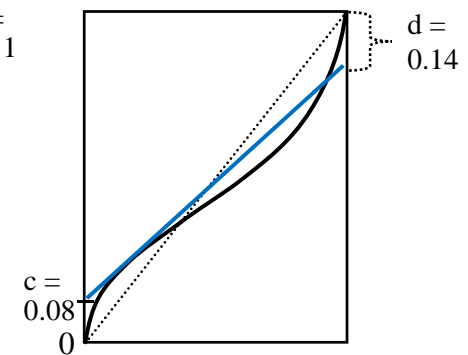


Fig.4.
Insensitivity
index a: 0.22;
pessimism
index b: 0.06.

Ambiguity Attitudes

□ Ellsberg two-color paradox continued

○ Let us remember that

$$P(R^k) = P(B^k) = P(R^u) = P(B^u) = \frac{1}{2}.$$

○ For the events R^k and B^k , the weight is $w_k(1/2) = 0.4$, and for events R^u and B^u , the weight $w_u(1/2) = 0.3$.

○ Different weighting explains the different preferences.

Experimental Method

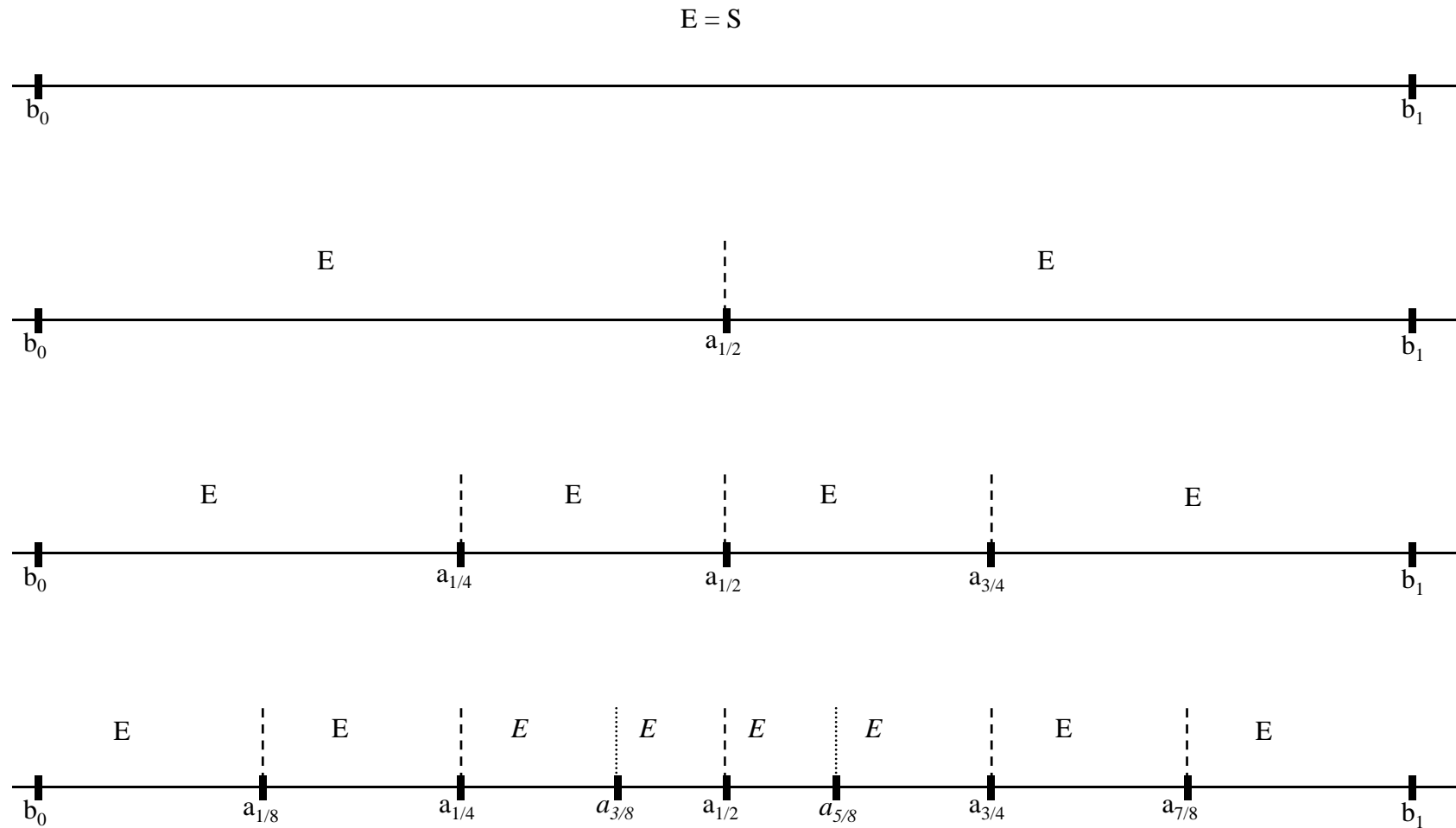
□ Participants:

- 62 students from the ENSAM-Paris. They were mathematically sophisticated and well acquainted with probability theory, but had no training in economics or decision theory.

□ Stimuli; unknown probabilities

- Three sources of uncertainty:
 - I. French Stock Index (CAC40) [how much it would change in a given day]
 - II. Temperature in Paris [on May 31, 2005]
 - III. Temperature in the capital of a randomly drawn remote country [on May 31, 2005].

□ Decomposition of the universal event



The italicized numbers and events in the bottom row were not elicited.

Experimental Method

- Stimuli; known probabilities
 - Utility $u(\cdot)$ and probability weighting function $w(\cdot)$ were elicited through certainty equivalents.

- Procedure and Motivation
 - Each participant was interviewed individually (95 minutes with a break of few minutes).
 - All participants received a flat payment of 20 €. For 31 subjects, real incentives were implemented through the random lottery incentive system (higher prize: 1000 €) in addition to flat payment.

Results: Subjective Probabilities

- Uniformity confirmed for 5 out of 6 cases

Figure 1. Probability distributions for CAC40

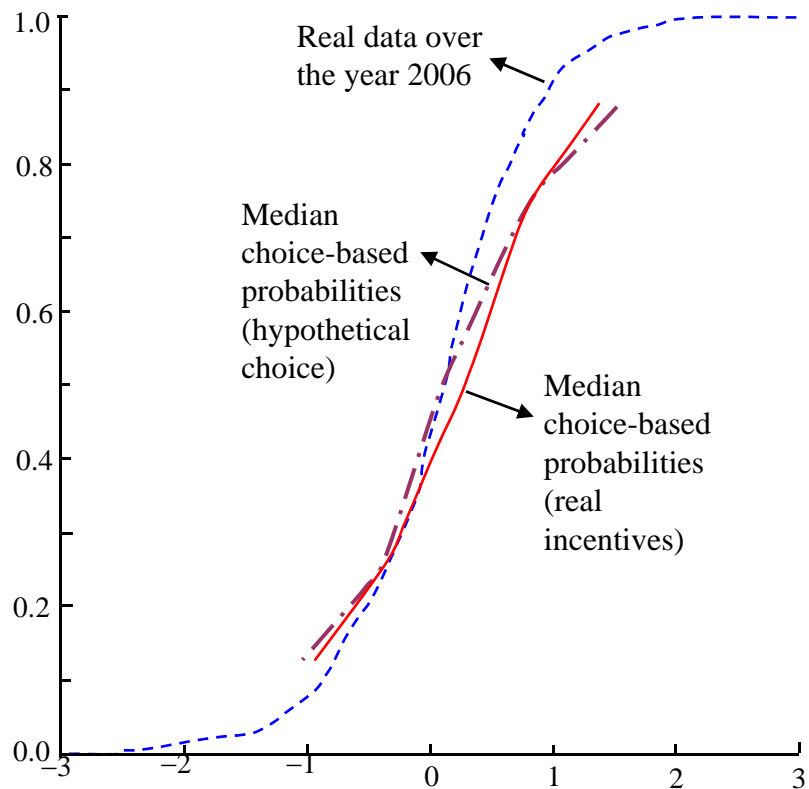
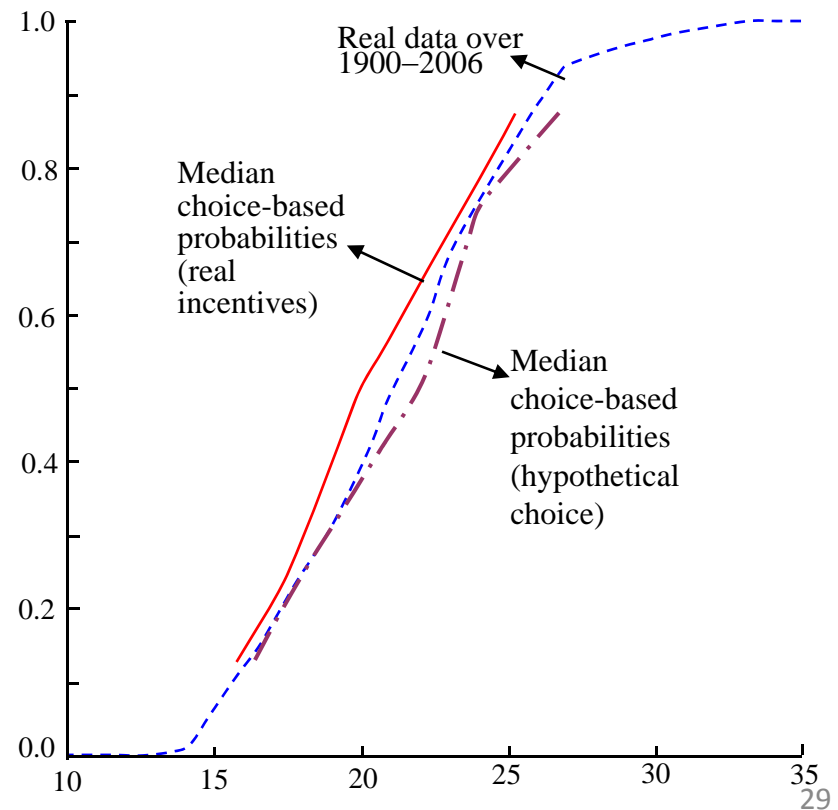


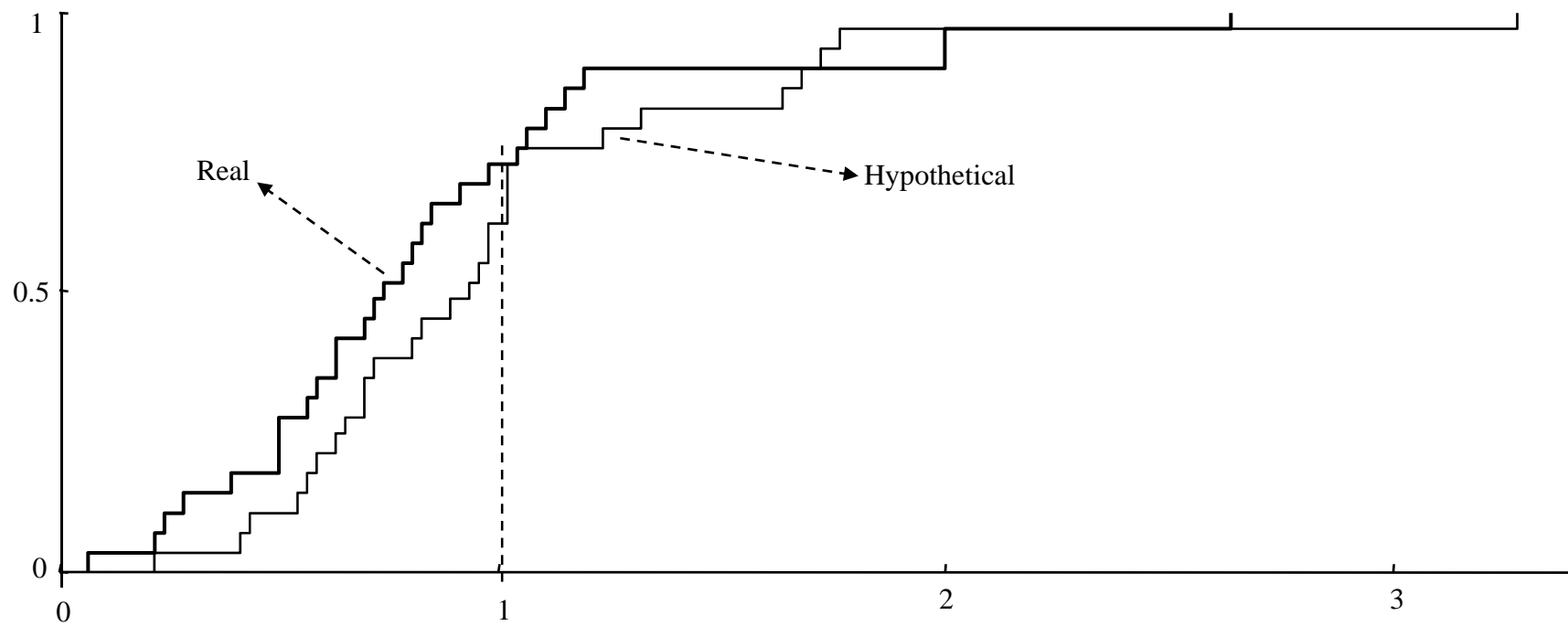
Figure 2. Probability distributions for Paris temperature



Results: Utility

□ Power Utility

Figure 3. Cumulative distribution of powers



Results: Ambiguity for U. Sources

□ Method for Measuring Ambiguity Attitude

- Certainty equivalents were measured for gambles on events. Knowing utility, we could calculate $w_S[P(E)]$ for events E , and then knowing $P(E)$, infer w_S .
- Consider a source S_1 and the corresponding exchangeable partitions $(A_i, i=1, \dots, n)$.
- Assume that $A_i(k)$ is the union of k exchangeable events
- Assume that
$$CE \sim (A_i(k): \$100, S - A_i(k): 0).$$
- We have
$$u(CE) = w_1[P(A_i(k))]u(\$100) + (1 - w_1[P(A_i(k))])u(0)$$

□ Overall Results

Figure 4. Average probability transformations for real payment

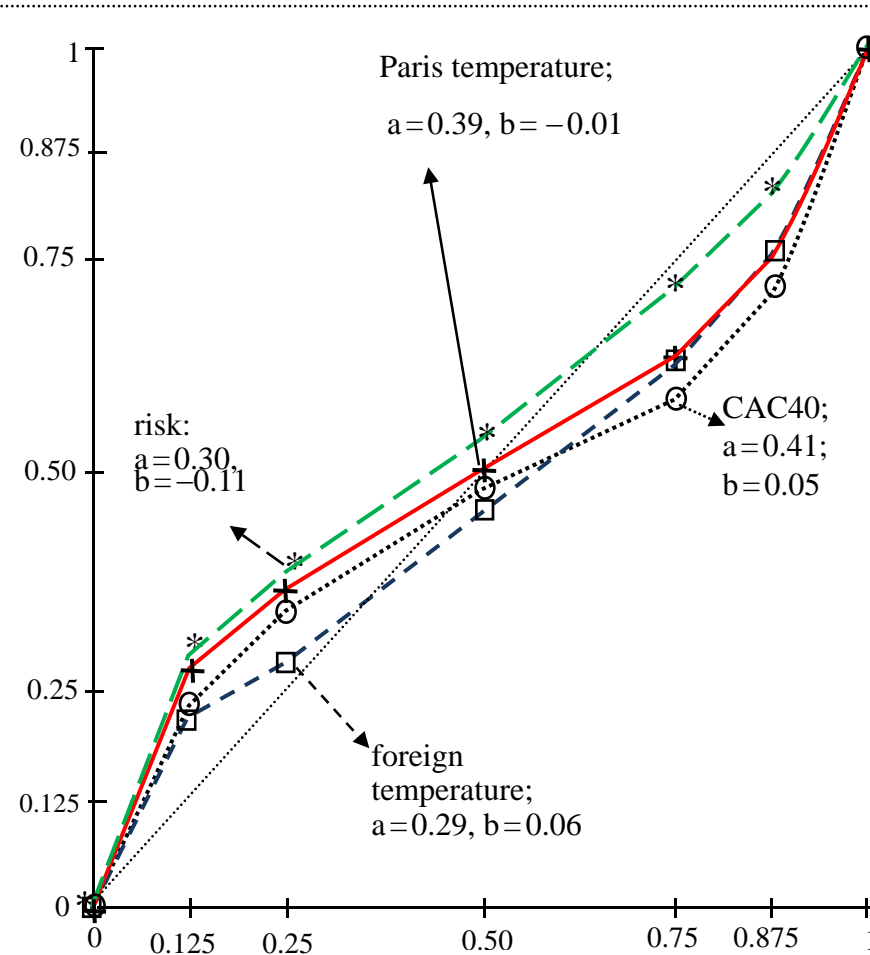


Fig. a. Raw data and linear interpolation.

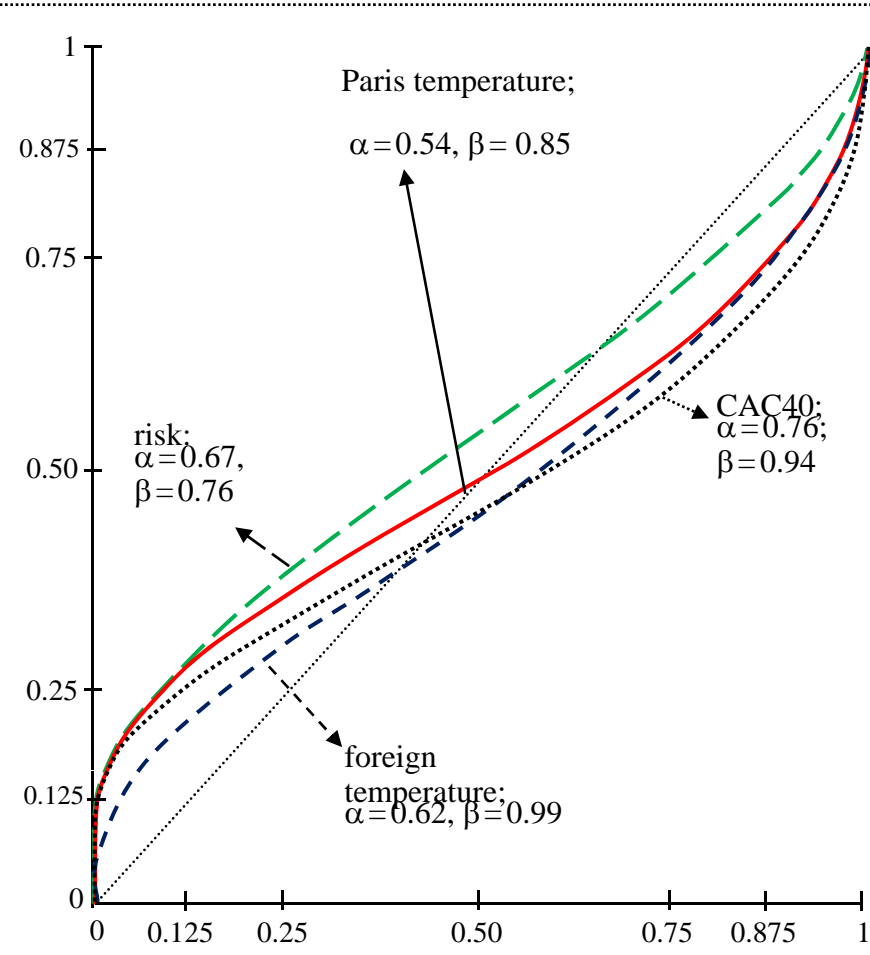
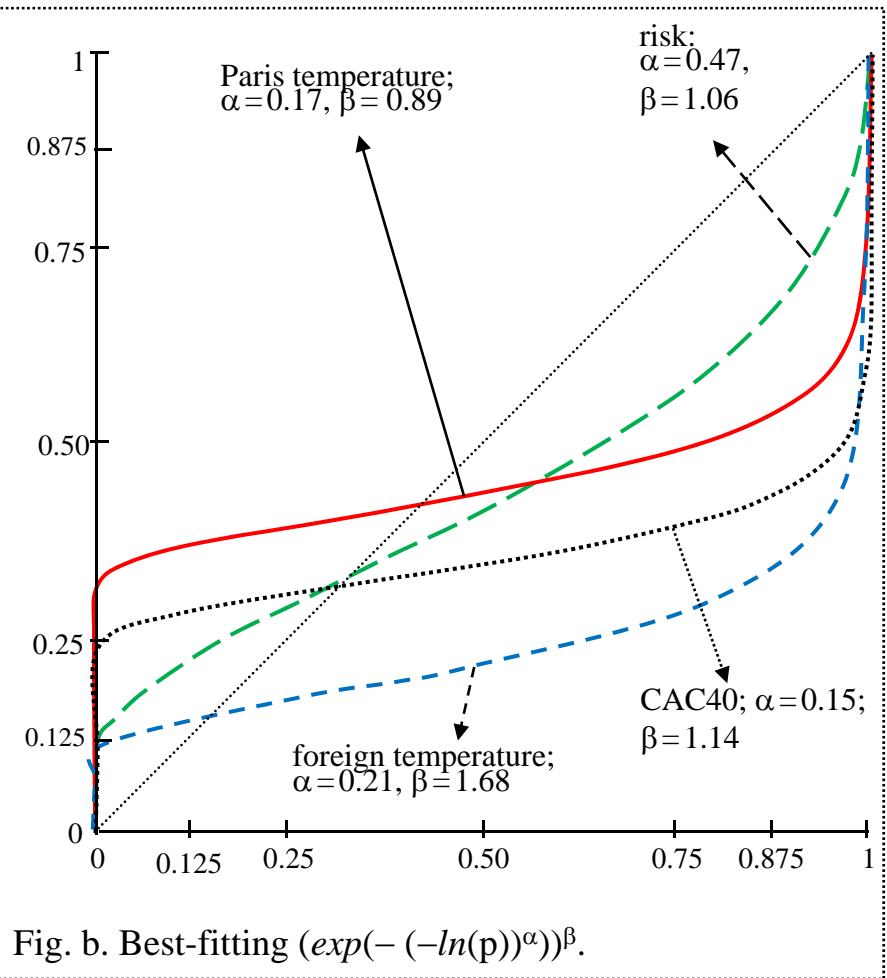
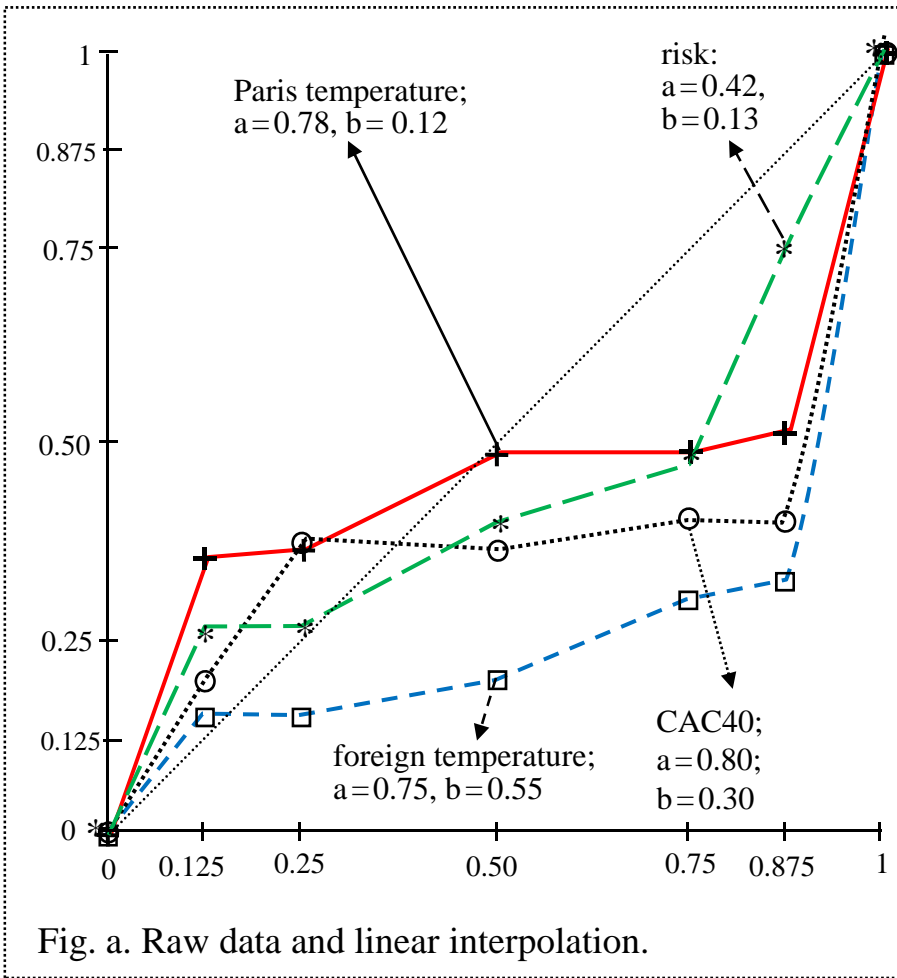


Fig. b. Best-fitting $(\exp(-(-\ln(p))^\alpha))^\beta$.

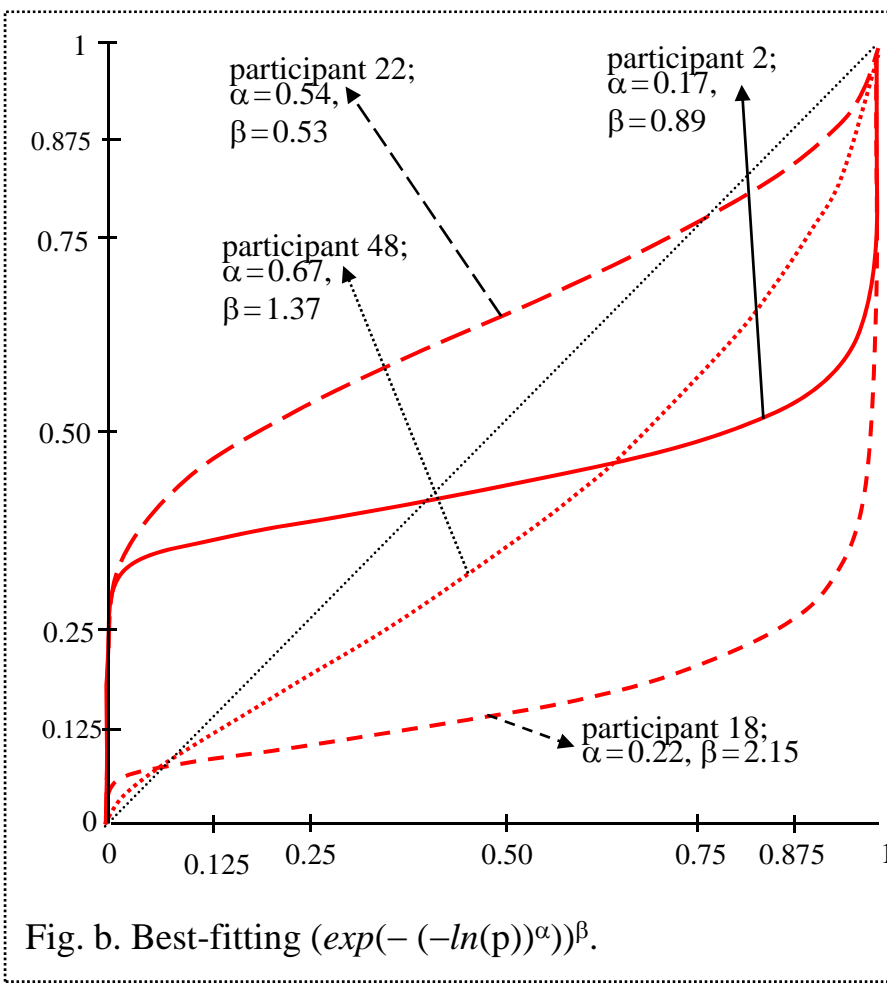
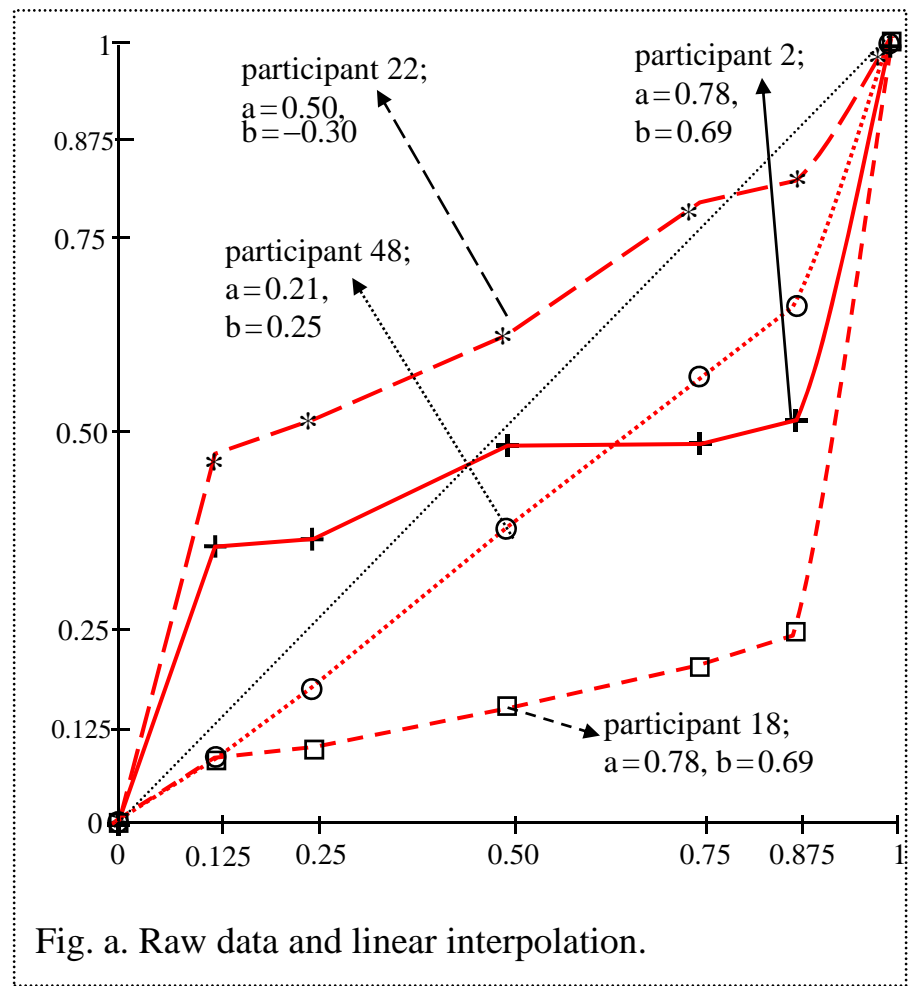
□ Results at individual level

Figure 8.3. Probability transformations for participant 2



□ Results at individual level

Figure 5. Probability transformations for Paris temperature and 4 participants



Concluding Remarks

- We have demonstrated that a biseparable representation of preferences combined with PS can be used to analyze ambiguity for uniform sources of uncertainty.
- The Ellsberg two-color example was reconciled with consistent subjective probabilities.
- We introduced a new method for deriving subjective probabilities and demonstrated its validity (good calibrations were achieved).
- We demonstrated the feasibility of complete quantifications of ambiguity attitudes.