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Explaining versus describing human decisions: Hilbert space structures in decision theory

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Abstract

Despite the impressive success of quantum structures to model long-standing human judgement and decision puzzles, the *quantum cognition research programme* still faces challenges about its explanatory power. Indeed, quantum models introduce new parameters, which may fit empirical data without necessarily explaining them. Also, one wonders whether more general non-classical structures are better equipped to model cognitive phenomena. In this paper, we provide a *realistic–operational foundation of decision processes* using a known decision-making puzzle, the *Ellsberg paradox*, as a case study. Then, we elaborate a novel representation of the Ellsberg decision situation applying standard quantum correspondence rules which map realistic–operational entities into quantum mathematical terms. This result opens the way towards an independent, foundational, rather than phenomenological, motivation for a general use of quantum Hilbert space structures in human cognition.

Keywords Quantum structures · Cognitive science · Decision theory · Ellsberg paradox · Operational realism

1 Introduction

Traditional cognitive theories systematically apply classical set-theoretic structures to model human judgements and decisions under uncertainty. This is particularly evident in theories of rational decision-making, like expected utility theory, where Bayesian, or Kolmogorovian [\(1933](#page-10-0)), models of probability directly follow from axioms on agents' preferences (von Neumann and Morgenster[n](#page-10-1) [1944](#page-10-1); Savag[e](#page-10-2) [1954](#page-10-2)).

However, several *cognitive puzzles* have been discovered in empirical tests, which provide evidence of systematic deviations from Kolmogorovian probability structures (see, e.g. Busemeyer and Bruz[a](#page-10-3) [2012\)](#page-10-3). For example, Kahneman and Tversky identified a *conjunction fallacy* in human probability judgements, namely the law of monotonicity of Kolmogorovian probability does not generally hold in this kind of judgements (Kahneman et al[.](#page-10-4) [1982](#page-10-4)). Also, in human decision-making, Tversky and Shafir proved that the law of total Kolmogorovian probability does not hold in the *disjunc-*

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tion effect (Kahneman and Tversk[y](#page-10-5) [2000\)](#page-10-5), while Allais and Ellsberg indicated that people do not always choose by maximizing an expected utility with respect to a Kolmogorovian probability measure (Ellsber[g](#page-10-6) [1961\)](#page-10-6).

As a consequence of the puzzles above, traditional theories using Kolmogorovian structures, though normatively compelling, are descriptively flawed, which led several authors to elaborate alternative proposals able to more efficiently and realistically represent human behaviour. This was the starting point of the *bounded rationality research programme*, initially proposed by Simo[n](#page-10-7) [\(1955\)](#page-10-7) and systematically applied by Kahneman et al[.](#page-10-4) [\(1982\)](#page-10-4), Kahneman and Tversk[y](#page-10-5) [\(2000\)](#page-10-5) to describe concrete judgements and decisions. Bounded rationality models give good predictions in a variety of circumstances. However, despite their simplicity and intuitive character, these models lack a unitary methodology, as well as deeper explanations, and thus provide a very fragmented picture of cognitive phenomena (Blutner and beim Grabe[n](#page-10-8) [2016](#page-10-8)).

The *quantum cognition research programme* has recently attracted the interest of the scientific community due to the superiority of quantum models over traditional and bounded rationality models to deal with the puzzles above. Quantum models were successfully applied to a variety of complex cognitive processes, including human language (Dalla Chiara

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et al[.](#page-10-9) [2006;](#page-10-9) Aert[s](#page-10-10) [2009](#page-10-10); Aerts et al[.](#page-10-11) [2013b](#page-10-11); Dalla Chiara et al[.](#page-10-12) [2015\)](#page-10-12), judgement (Busemeyer and Bruz[a](#page-10-3) [2012;](#page-10-3) Aerts et al[.](#page-10-13) [2013a](#page-10-13); Haven and Khrenniko[v](#page-10-14) [2013](#page-10-14)) and decision (Haven and Khrenniko[v](#page-10-14) [2013;](#page-10-14) Aerts et al[.](#page-10-15) [2014](#page-10-15); Aerts and Sozz[o](#page-10-16) [2016\)](#page-10-16) (see also Sozz[o](#page-10-17) [2017\)](#page-10-17). Despite these impressive results, however, quantum cognition still raises doubts regarding its explanatory power. Indeed, on the one side, quantum cognitive models introduce new parameters, which may fit experimental data, but do not necessarily explain them. On the other side, one is naturally led to wonder whether cognitive science really needs the entire mathematical formalism of quantum theory in Hilbert space or, on the contrary, non-Kolmogorovian non-Hilbertian models of probability are needed (see, e.g. Holik et al[.](#page-10-18) [2016](#page-10-18), [2017\)](#page-10-19).

In the present paper, we present binding motivations towards an independent, foundational, rather than purely phenomenological, justification of the quantum formalism in human judgement and decision-making under uncertainty. We start from the realistic and operational axiomatizations of quantum physics initiated by Jauc[h](#page-10-20) [\(1968](#page-10-20)) and Piro[n](#page-10-21) [\(1976\)](#page-10-21) in Geneva and extended by Aerts (see, e.g. Aert[s](#page-10-22) [1999](#page-10-22), [2002\)](#page-10-23) in Brussels. Efforts have been made in the second part of the last century to derive the mathematical formalism of quantum theory in Hilbert space from more intuitive and empirically justified axioms, resting on basic notions directly connected with the operations that are performed in a laboratory. Particularly, in the Brussels approach, any physical entity is expressed in terms of the basic notions of state, context, property and mutual relationships between them (*SCoP* system). The approach is*realistic*, in the sense that the state, being the result of an effective preparation procedure, describes aspects of the reality of the entity. The approach is also *operational*, in the sense that all basic notions are expressed in terms of well-defined empirical terms, like preparation and registration devices, statistics of outcomes, etc. If suitable "purely operational" axioms are imposed on a SCoP system, then the Hilbert space representation uniquely arises for the physical entity.

We believe that the above realistic–operational justification of the quantum Hilbert space formalism in physics also provides a strong motivation, if not a justification in itself, for the use of quantum Hilbert space structures in cognition. To this end, we particularize in Sect. [4](#page-6-0) to a specific decision-making situation, the *Ellsberg paradox* situation, used as a case study here, the realistic–operational foundation of cognitive entities we have recently elaborated (Aerts et al[.](#page-10-24) [2016](#page-10-24)), in which a cognitive entity is abstractly described in terms of well-defined empirical notions, i.e. state, context, property and outcome probability. Then, the stunning analogies in the realistic and operational descriptions of entities in physical and cognitive realms suggest that the same Hilbert space leading axiomatics should be used for a cognitive, e.g. decision-making, entity (Sect. [2\)](#page-1-0).

The Ellsberg paradox is reviewed in Sect. [3,](#page-3-0) where we explain the difficulties of both expected utility and bounded rationality theories, to accommodate Ellsberg preferences and the results of more general Ellsberg-like decision situations.

We then elaborate in Sect. [5](#page-7-0) a mathematical representation in Hilbert space of the Ellsberg paradox situation and the ambiguity aversion pattern found in empirical literature. We had already presented quantum models of various Ellsberg's thought experiments, including two-colour and three-colour urns (Aerts et al[.](#page-10-15) [2014;](#page-10-15) Aerts and Sozz[o](#page-10-16) [2016](#page-10-16); Aerts et al[.](#page-10-25) [2018a](#page-10-25), [b](#page-10-26); Sozz[o](#page-10-27) [2019\)](#page-10-27). The novelty of the mathematical representation developed here consists in the fact that it follows directly from the canonical quantum representation of the realistic–operational terms of state, context, property and outcome probability in Hilbert space, which makes the use of quantum mathematics in this kind of situations more firmly founded and generalizable to other decision situations.

We finally offer some conclusive remarks and considerations in Sect. [6,](#page-9-0) where we specify that the realistic– operational foundation of cognitive science can be in principle extended to several other judgement and decision-making situations, which constitutes a strong indication that "possible failures of Hilbert space modelling" should be searched in other cognitive domains than individual judgements and decisions.

2 Descriptive versus explanatory power of quantum structures

Traditional theories of individual judgement and decisionmaking use, often implicitly, set-theoretic structures, that closely resemble the formal operations of classical Boolean logic and Kolmogorovian probability theory (Kolmogoro[v](#page-10-0) [1933](#page-10-0)). This is specially evident in *rational decision theory*, according to which rational agents behave in such a way to maximize expected utility with respect to a Kolmogorovian probability measure and an underlying economic model (von Neumann and Morgenster[n](#page-10-1) [1944](#page-10-1); Savag[e](#page-10-2) [1954](#page-10-2)).

These theories are normatively compelling; however, the judgement and decision puzzles in Sect. [1](#page-0-0) make them descriptively problematical and suggest alternative more realistic approaches to human behaviour under uncertainty. A major research programme of this kind was initiated by Herbert Simon who put forward the *bounded rationality project* (Simo[n](#page-10-7) [1955\)](#page-10-7). Boundedly rational agents experience practical limitations in formulating and solving complex problems and in processing information. They tackle such limitations by taking mental short cuts, making subjective evaluations and putting psychological aspects above rational reasoning.

Within the bounded rationality project, one can cope with cognitive puzzles with judgement heuristics and rea-

soning biases, namely the conjunction fallacy with the *representativeness heuristics* (Kahneman et al[.](#page-10-4) [1982](#page-10-4)), the Allais paradox with *prospect theory* (Kahneman and Tversk[y](#page-10-5) [2000](#page-10-5)), the Ellsberg paradox with *cumulative prospect theory* (Kahneman and Tversk[y](#page-10-5) [2000](#page-10-5)), the disjunction effect by *uncertainty aversion* (Kahneman and Tversk[y](#page-10-5) [2000\)](#page-10-5), etc. These approaches undoubtedly provide an intuitive account of how individuals actually behave in situations of uncertainty. However, the reader recognizes at once that a rather eclectic methodology or, better, a variety of methodologies are employed to accommodate the puzzles above, and while some authors support the hypothesis of an *adaptive toolbox* to deal with these problems (Gigerenzer and Selte[n](#page-10-28) [2001\)](#page-10-28), many psychologists will find the bounded rationality research programme as unsatisfactory, and many philosophers of science will try to derive these puzzling phenomena from a universal theory able to overcome the fragmentation of existing approaches.

The quantum cognition research programme reaches both effectiveness and unitarity. Since the nineties, quantum Hilbert space models have shown impressive superiority over traditional and bounded rationality approaches in dealing with the puzzles of human cognition and attributing them to genuine quantum effects, like contextuality, emergence, entanglement, interference and superposition. On the other side, quantum models introduce new parameters which can be possibly fitted by empirical data, without, however, necessarily explaining them. Hence, the quantum cognition research programme, though phenomenologically successful, does not seem to offer a deeper understanding and/or explanation of these puzzles. In addition, it is reasonable to wonder whether one really needs the entire Hilbert space formalism to represent cognitive phenomena and should not better use more general non-Kolmogorovian representations outside physics (see, e.g. Holik et al[.](#page-10-18) [2016](#page-10-18), [2017\)](#page-10-19). In this respect, it should be noted that prospect theory already proposes non-Kolmogorovian probability models of probability of human decision (Kahneman and Tversk[y](#page-10-5) [2000\)](#page-10-5).

It is clear from the considerations above that one needs a *deeper justification* for the use of the Hilbert space formalism of quantum theory in cognition and decision and, more important, of its *necessity*. In this respect, a crucial result comes from increasing evidence that "judgements and decisions create rather than record" (Busemeyer and Bruz[a](#page-10-3) [2012\)](#page-10-3)—see. e.g. the following quotations.

"There is a growing body of evidence that supports an alternative conception according to which preferences are often constructed – not merely revealed – in the elicitation process. These constructions are contingent on the framing of the problem, the method of elicitation, and the context of the choice." Tversky and Simonso[n](#page-10-29) [\(1993\)](#page-10-29)

"…the process of choice – and in particular the act of choice – can make substantial difference to what is chosen. …, there is a particular necessity to take note of (i) chooser dependence, and (ii) menu dependence, of preference, even judged from a particular person's perspective." Se[n](#page-10-30) [\(1997\)](#page-10-30)

"…valuations are initially malleable but become 'imprinted' after the agent is called upon to make an initial decision." Ariely et al[.](#page-10-31) [\(2003\)](#page-10-31)

It is more and more acknowledged that, in any judgement or decision, a contextual interaction occurs between the situation that is the object of the evaluation and the individual who takes the decision (agent, decision-maker), which may affect the situation itself. At the end of this interaction, a result is actualized among a set of results that were only potential before the interaction (Aerts et al[.](#page-10-26) [2018b\)](#page-10-26). Hence, a judgement/decision process closely resembles a quantum measurement process, where a contextual interaction occurs between the quantum particle that is measured and the measurement apparatus, which changes the state of the quantum particle determining what Heisenberg called "transition from potential to actual".

We believe that these analogies between micro-physics and cognition are a good starting point towards a foundational justification for the use of Hilbert space quantum formalism in cognition and decision.

In the sixties and seventies of the previous century, several authors wondered whether and how one can provide an independent justification for the Hilbert space formalism in quantum physics, deriving this formalism from physically justified axioms, resting on well-defined empirical notions, directly connected with the operations that are usually performed in a laboratory. One of the well-known approaches to the foundations of quantum physics is the *Geneva–Brussels realistic–operational approach*, initiated by Jauc[h](#page-10-20) [\(1968\)](#page-10-20) and Piro[n](#page-10-21) [\(1976\)](#page-10-21) in Geneva, and successively extended by Aerts in Brussels (see, e.g. Aert[s](#page-10-22) [1999](#page-10-22), [2002](#page-10-23)). This research consisted in abstractly describing any physical entity by relevant sets of states, contexts, properties and statistical connections between these notions (SCoP system). These theoretical notions are directly interpretable on physical operations on macroscopic apparatuses, such as preparation and registration devices, performed in spatio-temporal domains, such as physical laboratories. Measurements, state transformations, outcome probabilities and dynamics can then be expressed in terms of these more fundamental notions. If suitable axioms are imposed on the mathematical structures underlying a SCoP system, then the Hilbert space structure of quantum theory emerges as a unique mathematical representation, up to isomorphisms (Beltrametti and Cassinell[i](#page-10-32) [1981](#page-10-32)).

This justification provides the "fundamental architecture of quantum theory in Hilbert space".

We have recently proved that any cognitive entity Ω , e.g. a concept, a conceptual combination, a proposition, or a more complex decision-making situation, can be abstractly described by a SCoP system $(\Sigma, \mathcal{L}, \mathcal{C}, \mu, \nu)$ (Aerts et al[.](#page-10-24) [2016\)](#page-10-24), exactly like in physics. We review the essential elements of a SCoP system in cognition in the following.

- (1) The complex of experimental procedures conceived by the experimenter, the experimental setting and the cognitive effect that one wants to analyse define a *cognitive entity* Ω and are usually associated with a preparation procedure of a state of Ω.
- (2) Σ is the set of all states of Ω . A *state* p of Ω is the consequence of a preparation procedure of Ω and has a cognitive, rather than physical, nature. The state of the cognitive entity is a *state of affairs*. It indeed expresses a "reality of the cognitive entity", in the sense that, once prepared in a given state, such condition is independent of any measurement procedure and can be confronted with the different participants in an experiment, leading to outcome data and their statistics.
- (3) $\mathcal C$ is the set of all contexts of Ω . A *context e* is an element that can provoke a change of state of the cognitive entity. A special context is the one introduced by a measurement. Indeed, when the cognitive experiment starts, an interaction occurs between the measured entity Ω under study and a participant in the experiment, in which the state p of Ω generally changes, being transformed to another state *q*. This cognitive interaction is formalized by means of a context *e*.
- (4) *L* is the set of all properties of Ω. A *property a* of Ω is something Ω "has" independently of any context influencing the entity. An entity Ω in a given state p has a set of properties that are *actual* in that state, the others being *potential*. A context *e* may change the status actual/potential of a property, but cannot change the property itself.
- (5) The change function $\mu : \Sigma \times \mathcal{C} \times \Sigma \longrightarrow [0, 1]$ is such that, for every $p, q \in \Sigma$, $e \in \mathscr{C}$, $\mu(q, e, p)$ is the probability, as the large number limit of relative frequencies, that the context *e* changes the initial state *p* of Ω to the final state q .

Once recognizes at once in (1) – (5) the building blocks of the realistic–operational description of a physical entity, in the sense that in both physical and cognitive realms, a SCoP system incorporates all what is needed to study what an entity is, behaves and changes under a context. These impressive analogies indicate that the axioms generally used to justify the Hilbert space formalism of quantum physics are also appropriate to represent cognitive entities and processes.

This provides an independent foundational clue and nonphenomenological motivation, if not a justification, that the mathematics of Hilbert space should be used to represent judgement and decision phenomena.

In Sect. [4,](#page-6-0) we will provide a realistic–operational description of a specific decision-making situation, the Ellsberg paradox, setting the grounds for a quantum mathematical representation of it in Sect. [5.](#page-7-0) In the next section, we will instead summarize the serious difficulties of both traditional and bounded rationality approaches to handle such kind of decision-making situations.

3 Rational decision theory and its puzzles

Traditional theories of rational decision-making rest on the tenet that, in situations of uncertainty, individual agents choose in such a way to maximize their expected utility, or degree of satisfaction.

In 1944, von Neumann and Morgenstern presented in a seminal work the first axiomatic formulation of expected utility theory. People continuously take decisions among different options. These decisions are assumed to *reveal* underlying preferences. Then, von Neumann and Morgenstern proposed a set of "reasonable" axioms on human preferences such that, if the decisions are coherent, in the sense that they reveal axiom satisfying preferences, then the decisions are equivalent to the maximization of an expected utility functional with respect to a Kolmogorovian probability measure (von Neumann and Morgenster[n](#page-10-1) [1944\)](#page-10-1).

von Neumann and Morgenstern's formulation of expected utility theory has a major limitation, in that it only deals with the uncertainty that can be formalized by known probabilities (also referred to as *objective uncertainty*, or *risk*). On the other hand, situations frequently occur in which uncertainty cannot be formalized by known probabilities (also referred to as *subjective uncertainty*, or *ambiguity*) (Knigh[t](#page-10-33) [1921](#page-10-33)). The Bayesian approach to probability minimizes the distinction between objective and subjective uncertainty introducing the notion of *subjective probability*. Even when probabilities are not known, people may still construct their own *beliefs*, or *priors* (which may differ from one individual to another), and they maximize expected utility with respect to these priors. Indeed, Leonard Savage presented in 1954 an axiomatic formulation of expected utility theory which extends the one of von Neumann and Morgenstern to subjective uncertainty (Savag[e](#page-10-2) [1954](#page-10-2)).

We summarize in the following the essential definitions and results of Savage's expected utility theory, together with its major pitfalls. We refer to Gilboa and Marinacc[i](#page-10-34) [\(2013](#page-10-34)), Machina and Siniscalch[i](#page-10-35) [\(2014](#page-10-35)) for detailed reviews of these results.

Savage introduced a set of basic notions, including states of nature, consequences, preferences and looked for justified axioms on preferences able to provide a representation theorem in which ordering of preferences is characterized by maximization of expected utility. This procedure formally resembles the procedures used in the axiomatizations of quantum physics in Sect. [4.](#page-6-0)

Let $\mathscr S$ be the set of all (physical) *states of nature*, which we assume to be discrete and finite here, for the sake of simplicity. Let $\mathcal{P}(\mathcal{S})$ be the power set of \mathcal{S} and $\mathcal{A} \subseteq \mathcal{P}(\mathcal{S})$ be a (Boolean) σ-algebra. An element *E* ∈ *A* denotes an *event*. A *Kolmogorovian probability measure* over *A* is a function $p : \mathscr{A} \subseteq \mathscr{P}(\mathscr{S}) \longrightarrow [0, 1]$ satisfying the axioms of Kolmogoro[v](#page-10-0) [\(1933\)](#page-10-0).

Then, let *X* be the set of all *consequences*, whose elements we assume to denote *monetary payoffs*, hence real numbers, here, for the sake of simplicity. In Savage's formulation, a function $f : \mathscr{S} \longrightarrow \mathscr{X}$ denotes an *act*. Let \mathscr{F} be the set of all acts. Let us endow *F* with a *weak preference relation* \succsim , that is, a complete and transitive relation over the Cartesian product $\mathscr{F} \times \mathscr{F}$. In \succsim , the relations \succ and \sim denote *strong preference* and *indifference*, respectively, that is, we write $f > g$ whenever an individual strictly prefers act *f* to act *g* and *f* ∼ *g* whenever the individual is indifferent between *f* and *g*.

Next, let $u : \mathcal{X} \longrightarrow \mathbb{R}$ be a *utility function* over \mathcal{X} . This function typically expresses the decision-maker's taste; hence, it is assumed to be strictly increasing and continuous, with additional technical constraints related to the specification of the decision-maker's attitude towards risk.

The mathematical definitions above can be simplified by introducing a set $\{E_1, \ldots, E_n\}$ of mutually exclusive and exhaustive elementary events, where $E_i = \{s_i \in \mathcal{S}\}\,$, $i \in$ $\{1, \ldots, n\}$, which thus form a partition of \mathscr{S} . For every $i \in$ $\{1, \ldots, n\}$, let x_i be the utility associated by the act f to the event E_i . Then, f can be equivalently expressed by the 2ntuple $f = (E_1, x_1; \ldots; E_n, x_n)$, meaning that the individual will get the outcome x_1 if the event E_1 occurs (i.e. the state of nature s_1 realizes), ..., the outcome x_n if the event E_n occurs (i.e. the state of nature s_n realizes).

Finally, we denote by $W(f) = \sum_{i=1}^{n} p(E_i)u(x_i)$ the *expected utility* associated with the act *f* with respect to the Kolmogorovian probability measure *p*.

In his representation theorem, Savage proved that, if the algebraic structure (\mathscr{F}, \succsim) satisfies a number of "reasonable" axioms^{[1](#page-4-0)} then, for every $f, g \in \mathcal{F}$, a unique Kolmogorovian probability measure *p* and a unique (up to positive affine

transformations) utility function u exist such that f is preferred to *g*, i.e. $f \succeq g$, if and only if the expected utility of *f* is greater than the expected utility of *g*, i.e. $W(f) \geq W(g)$. For every $i \in \{1, \ldots, n\}$, the utility value $u(x_i)$ depends on the decision-maker's risk preferences, while $p(E_i)$ is interpreted as the subjective probability, expressing the individual's belief that the event *Ei* occurs (Savag[e](#page-10-2) [1954\)](#page-10-2).

Savage's result is both compelling at a *normative level* and testable at a *descriptive level*. Indeed:

- (i) if the axioms are intuitively reasonable and decisionmakers agree with them, then they *must* all behave as if they were maximizing an expected utility with respect to a single subjective probability distribution satisfying Kolmogorov's axioms;
- (ii) the axioms suggest to design decision-making experiments to test the validity of expected utility theory, hence of the axioms themselves, in real life situations.

Because of (i), Savage's expected utility formulation is generally accepted to prescribe "how rational agents should choose". However, on the one side, the theory offers very little about where beliefs come from and how they should be calculated, and on the other side, regarding (ii), decisionmaking experiments have systematically found deviations from that rational behaviour in concrete situations.

In particular, Daniel Ellsberg proved in 1961 in a number of thought experiments that decision-makers generally prefer acts with known (or objective) probabilities over acts with unknown (or subjective) probabilities (Ellsber[g](#page-10-6) [1961](#page-10-6)). We analyse here the famous *Ellsberg three-colour example* as a paradigmatic example to show that (i) traditional decision theories do not work, (ii) bounded rationality approaches are not sufficiently explanatory, (iii) quantum structures are needed.

Consider one urn with 30 red balls and 60 balls that are either yellow or black, the latter in unknown proportion. One ball will be drawn at random from the urn. Then, free of charge, a person is asked to bet on pairs of the acts f_1 , f_2 , f_3 and *f*⁴ in Table [1.](#page-4-1) Ellsberg suggested that, when asked to rank these acts, most individuals will prefer f_1 over f_2 and f_4 over *f*3. Indeed, *f*¹ and *f*⁴ are *unambiguous acts*, in the sense that they are associated with events over known probabilities the events "a red ball is drawn" and "a yellow or black ball

¹ One of the axioms is the famous*sure-thing principle*, which is violated in the Ellsberg paradox. The other axioms are: ordinal event independence, comparative probability, non-degeneracy, small event continuity and dominance, and have a technical nature (Savag[e](#page-10-2) [1954](#page-10-2)). However, these axioms are not relevant to the present purposes, and hence, we will not dwell on them, for the sake of brevity.

Fig. 1 A sample of the questionnaire related to the decision-making experiment on the Ellsberg three-colour example: choice between acts *f*¹ and f_2 in Table [1](#page-4-1)

EXPERIMENT 1A

The urn in Fig. 1A contains 90 balls of three different colors: red, yellow and black. You know that there are 30 red balls and that the sum of the black balls and the yellow balls is 60 (in unknown proportion). A ball will be drawn at random within the urn. You can choose between two situations.

Fig. 1A. The urn.

Situation A. If the ball is red, you receive \$100 If the ball is yellow or black, you receive \$0

Situation B. If the ball is black, you receive \$100 If the ball is red or yellow, you receive \$0

Which situation do you prefer?

By "preference" we mean "the choice you would take if this situation happened to you in real life". You are expected to choose one of the situations, i.e. "not choosing is no option".

I prefer situation B

is drawn" are associated with objective probabilities 1/3 and 2/3, respectively. On the contrary, *f*² and *f*³ are *ambiguous acts*, in the sense that they are associated with events over unknown probabilities—the events "a yellow ball is drawn" and "a black ball is drawn" are both associated with a probability ranging from 0 to 2/3. This attitude of decision-makers to prefer "probabilized over non-probabilized uncertainty" has been known as *ambiguity aversion* since Ellsberg studies (Ellsber[g](#page-10-6) [1961\)](#page-10-6).

Several experiments on Ellsberg urns decisions, but also on financial, insurance and medical decisions, have confirmed the *Ellsberg preferences* $f_1 \succ f_2$ and $f_4 \succ f_3$, thus indicating that ambiguity aversion is a good candidate to explain concrete decisions in this case, and only Slovic and Tversky found *ambiguity seeking patterns*(see, e.g. Machina and Siniscalch[i](#page-10-35) [2014](#page-10-35) for a review of experimental studies).

In Aerts et al[.](#page-10-25) [\(2018a\)](#page-10-25), we tested various human decision puzzles, including the Ellsberg three-colour example. We asked 200 people, chosen among colleagues and friends, to fill a questionnaire in which they had to choose between various options. People had on average a basic knowledge of probability theory, but no specific training in decision theory.

Participants were provided with a questionnaire similar to the one in Fig. [1,](#page-5-0) in which they had to choose between acts *f*¹ and f_2 and then between acts f_3 and f_4 in Table [1.](#page-4-1) Overall, 125 participants preferred acts *f*¹ and *f*4, 38 preferred acts f_1 and f_3 , 6 preferred acts f_2 and f_3 , and 31 preferred acts *f*² and *f*4. This means that 163 participants over 200 preferred act *f*¹ over act *f*2, which entails a *preference weight* of 0.815. Also, 156 participants over 200 preferred act *f*⁴ over act *f*3, which entails a preference weight of 0.780. The inversion rate is 0.655, a pattern that agrees with the Ellsberg preferences found in the literature and significantly indicates the presence of ambiguity aversion.

Preferences of decision-makers who are sensitive to ambiguity, that is, are ambiguity averse or ambiguity seeking, cannot be explained within Savage's expected utility theory, because they violate the sure-thing principle, according to which, preferences should be independent of the common outcome. In the specific case of the three-colour example, preferences should not depend on whether the common event "a yellow ball is drawn" pays off \$0 or \$100. More technically, Savage's expected utility theory predicts *consistency of preferences*, namely f_1 is preferred to f_2 if and only if

*f*³ is preferred to *f*4. A simple calculation shows that this is impossible within a traditional expected utility framework. Indeed, if we denote by \tilde{p}_R , \tilde{p}_Y and \tilde{p}_B the probability that a red ball, a yellow ball, a black ball, respectively, are drawn (with $\tilde{p}_R = 1/3 = 1 - (\tilde{p}_Y + \tilde{p}_B)$), then the expected utilities $W(f_i)$, $i = 1, 2, 3, 4$, are such that $W(f_1) > W(f_2)$ if and only if $(\tilde{p}_R - \tilde{p}_B)(u(100) - u(0)) > 0$ if and only if $W(f_3) > W(f_4)$. We can equivalently say that no assignment of Kolmogorovian probabilities \tilde{p}_R , \tilde{p}_Y and \tilde{p}_B reproduces a preference with $W(f_1) > W(f_2)$ and $W(f_4) > W(f_3)$, whence the *Ellsberg paradox*.

Several extensions of Savage's expected utility theory have been put forward in order to accommodate the Ellsberg paradox (see, e.g. the reviews in Gilboa and Marinacc[i](#page-10-34) [2013](#page-10-34) and Machina and Siniscalch[i](#page-10-35) [2014\)](#page-10-35). One of the major proposals is Tversky and Kahneman's cumulative prospect theory, mentioned in Sect. [2](#page-1-0) and elaborated within a bounded rationality research programme (Kahneman and Tversk[y](#page-10-5) [2000](#page-10-5)). In particular, to reproduce Ellsberg preferences, Tversky and Kahneman replaced (i) the utility function *u* by a *scaling function u'* reflecting the subjective value of the outcome utility and (ii) the subjective probability measure *p* by a non-additive measure p' satisfying the mathematical properties of a *capacity*. As we have mentioned in Sect. [2,](#page-1-0) such bounded rationality models, though descriptively interesting and easily interpretable intuitively, provide a too fragmented view of decision theory; hence, they are not able to provide a unitary and adequate explanatory framework to understand the deep aspects of decision processes. In addition, cumulative prospect theory, as well as other major non-expected utility models, fails to reproduce the empirical results of a recently elaborated variant of the Ellsberg paradox, the *Machina paradox* (Aerts et al[.](#page-10-25) [2018a;](#page-10-25) Machin[a](#page-10-36) [2009](#page-10-36); L'Haridon and Placid[o](#page-10-37) [2010\)](#page-10-37).

An innovative aspect of descriptive, like bounded rationality, approaches, is the representation of subjective probabilities by more general, possibly non-Kolmogorovian, mathematical structures. This is crucial towards a more satisfactory framework for human decision-making that goes beyond Savage's expected utility, as we will see in Sect. [5.](#page-7-0)

In the next section, we intend to elaborate a realistic– operational description of a decision-making situation, using the Ellsberg three-colour example as a case study. We will demonstrate that, once the Ellsberg paradox situation is formulated in terms of states, contexts, properties and transition probabilities, then the application of the mathematical formalism of quantum theory directly follows from the canonical representation of these realistic–operational terms in Hilbert space.

4 A realistic–operational description of a decision-making situation

In this section, we specify the realistic–operational description of cognitive entities in Sect. [2](#page-1-0) to the decision-making situation presented in the Ellsberg three-colour example (Aerts et al[.](#page-10-24) [2016](#page-10-24)). In it, we explicitly distinguish physical from cognitive, in this case, decision-making, entities. Analogously, we distinguish physical from cognitive states of nature, though one can intuitively see that some cognitive states are mapped into the corresponding physical states.

In the Ellsberg three-colour example, the cognitive, i.e. decision-making, entity Ω_{DM} is the urn with 30 red balls and 60 yellow or black balls in unknown proportion. This is what the individual reads in a questionnaire, interacts with and takes a decision on.

The cognitive entity Ω_{DM} is associated with a defined set Σ_{DM} of states.^{[2](#page-6-1)} A state $p \in \Sigma_{DM}$ has a cognitive nature and incorporates aspects of ambiguity. A context *e* does not pertain to Ω_{DM} but can interact with it. Let \mathcal{C}_{DM} be the set of all contexts of Ω*DM* . The interaction of Ω*DM* with a context $e \in \mathcal{C}_{DM}$ may determine a change of the state of Ω*DM* from *p* to a different state *q*. The probability of such a state transition will be denoted by $\mu(q, e, p)$, where μ : $\Sigma_{DM} \times \mathcal{C}_{DM} \times \Sigma_{DM} \longrightarrow [0, 1]$. We might complete the realistic–operational description of Ω*DM* defining a set \mathscr{L}_{DM} of properties and an actuality relation connecting properties and states. However, they are not needed in the Ellsberg three-colour scenario; hence, we omit specifying these notions here, for the sake of brevity, though they may be relevant in more general decision situations.

Let us now introduce a colour context $e_C \in \mathcal{C}_{DM}$, which is the context associated with a drawing of a ball from the urn. As a result of the drawing, we have three possible outcomes, *R*, *Y* and *B*, corresponding to the colours of the balls, red, yellow and black, respectively. The outcomes *R*, *Y* and *B* are the eigenvalues of e_C and are, respectively, associated with the final states, or eigenstates, p_R , p_Y and p_B of the cognitive entity Ω_{DM} . These eigenstates are such that $\mu(p_i, e_C, p_i)$ = $1, i \in \{R, Y, B\}.$

The colour context e_C introduces three mutually exclusive and exhaustive elementary events $E_i = (e_C, \{i\})$, $i \in \{R, Y, B\}$, which are such that the subjective probability that the event E_i occurs when the cognitive entity

² Some authors identify the notion of "state" with the notion of "belief state" of the individual participating in the cognitive experiment, e.g. taking the decision (see, e.g. Busemeyer and Bruz[a](#page-10-3) [2012;](#page-10-3) Blutner and beim Grabe[n](#page-10-8) [2016](#page-10-8); Haven and Khrenniko[v](#page-10-14) [2013](#page-10-14)). We instead neatly distinguish states from measurements here. A state is defined by a preparation procedure of the cognitive entity under investigation. The participant in the experiment acts as a (measurement) context that interacts with the cognitive entity and changes its state.

 Ω_{DM} is in the state *p* is given by the transition probability $\mu(p_i, e_C, p) \in [0, 1], i \in \{R, Y, B\}.$

Then, in analogy with Savage's expected utility theory, we can introduce monetary payoffs $x \in \mathcal{X} \subseteq \mathcal{R}$, utility functions $u : \mathcal{X} \longrightarrow \mathbb{R}$ and acts taking the form $f =$ $(E_R, x_R; E_Y, x_Y, E_B, x_B)$, mapping the event E_i into the payoff x_i , $i \in \{R, Y, B\}$. In particular, the acts f_1 , f_2 , f_3 and *f*⁴ in Table [1,](#page-4-1) Sect. [3](#page-3-0) are defined in the way above.

Let us now come to the operational–realistic description of a decision-making process in the Ellsberg three-colour situation. Suppose that, in the absence of any context, the entity Ω_{DM} is in the initial state *p*₀. This state corresponds to a preparation of the cognitive entity Ω*DM* and can be set by the information on the corresponding physical entity. For example, it is reasonable to assume that p_0 is such that, for every $i \in \{R, Y, B\}, \mu(p_i, e_C, p_0) = 1/3$, because of the indifference principle. Whenever an individual is asked to rank f_1 and f_2 , the individual's attitude towards ambiguity, e.g. ambiguity aversion, can be described as a new context $e_1 \in \mathcal{C}_{DM}$ acting on Ω_{DM} in the initial state p_0 and changing p_0 into a new state p_1 , characterized by a new probability distribution $\mu(p_i, e_C, p_1)$, $i \in \{R, Y, B\}$. Similarly, whenever the individual is asked to rank f_3 and f_4 , the individual's attitude towards ambiguity, e.g. ambiguity aversion, can be described as a new context $e_2 \in \mathcal{C}_{DM}$ acting on Ω_{DM} in the initial state p_0 and changing p_0 into a new state *p*2, characterized by a new probability distribution $\mu(p_i, e_C, p_2)$, $i \in \{R, Y, B\}$. The cognitive states p_1 and *p*2, and their ensuing probability distributions, are responsible of the *inversion of preferences* which occur in the Ellsberg paradox situation.

Finally, a decision process between acts f_1 and f_2 can be operationally described as a measurement context *e*¹² ∈ \mathcal{C}_{DM} acting on the entity Ω_{DM} in the ambiguity-averse state *p*1, with possible outcomes "yes" and "no". Similarly, a decision process between acts f_3 and f_4 can be described as a measurement context $e_{34} \in \mathcal{C}_{DM}$ acting on the entity Ω_{DM} in the ambiguity-averse state p_2 , with possible outcomes "yes" and "no". These contexts give rise of the statistics of outcomes in a decision-making test on the Ellsberg threecolour urn.

Now, the considerations in Sect. [2](#page-1-0) naturally indicate to represent states, contexts, properties and outcome probabilities of Ω_{DM} by using the canonical quantum representation of states, contexts, properties and outcome probabilities in Hilbert space. In particular, subjective probabilities will be represented using the Born rule of quantum probability. This is what we intend to show in the next section where the realistic–operational terms defined here will be canonically represented using quantum mathematical terms.

5 A novel quantum representation of the Ellsberg paradox

In this section, we elaborate a new quantum representation of the three-colour example straightly following the canonical Hilbert space representation of the realistic and operational notions in Sect. [3.](#page-3-0) This representation generalizes and strengthens those in Aerts et al[.](#page-10-15) [\(2014](#page-10-15)), Aerts and Sozz[o](#page-10-16) [\(2016](#page-10-16)), Aerts et al[.](#page-10-26) [\(2018b\)](#page-10-26), Aerts et al[.](#page-10-25) [\(2018a\)](#page-10-25).

5.1 Quantum representation of basic notions

The cognitive entity Ω_{DM} is associated with a Hilbert space H . Since the three-colour example involves three mutually exclusive and exhaustive elementary events, *H* can be chosen to be isomorphic to the complex Hilbert space \mathbb{C}^3 of ordered triples of complex numbers. Let $\{(1, 0, 0), (0, 1, 0), (0, 0, 1)\}\$ be the canonical orthonormal basis of \mathbb{C}^3 .

A state p_v of the entity Ω_{DM} is represented by the unit vector $|v\rangle \in \mathcal{H}$, $|||v\rangle|| = \sqrt{\langle v|v\rangle} = 1$.

The elementary event E_i is represented by the onedimensional orthogonal projection operator $P_i = |i\rangle\langle i|, i \in$ ${R, Y, B}$, where we choose $|R\rangle = (1, 0, 0), |Y\rangle = (0, 1, 0)$ and $|B\rangle = (0, 0, 1)$. The colour context e_C is then represented by the spectral family $\{P_R, P_Y, P_B\}$.

In the canonical basis of \mathbb{C}^3 , the unit vector $|v\rangle$ can be written as:

$$
|v\rangle = \rho_R e^{i\theta_R} |R\rangle + \rho_Y e^{i\theta_Y} |Y\rangle + \rho_B e^{i\theta_B} |B\rangle
$$

= $(\rho_R e^{i\theta_R}, \rho_Y e^{i\theta_Y}, \rho_B e^{i\theta_B})$ (1)

where ρ_R , ρ_Y and ρ_B are non-negative numbers such that $\rho_R^2 + \rho_Y^2 + \rho_B^2 = 1$, and θ_R , θ_Y and θ_B are real numbers.

We use the Born rule to represent subjective probabilities. Then, the subjective probability that the elementary event E_i , $i \in \{R, Y, B\}$, occurs when the entity Ω_{DM} is in the state p_v is given by:

$$
\mu_v(E_i) = \langle v | P_i | v \rangle = |\langle i | v \rangle|^2 = \rho_i^2 \tag{2}
$$

In addition, the subjective probability that the event *E*, represented by the orthogonal projection operator P_E , occurs when the cognitive entity Ω_{DM} is in the state p_v is $\mu_v(E)$ = $\langle v|P_E|v\rangle = ||P_E|v\rangle||^2$. Finally, for every state p_v represented by the unit vector $|v\rangle$, the subjective probability measure

$$
\mu_v: \mathcal{L}(\mathbb{C}^3) \longrightarrow [0, 1] \tag{3}
$$

is a quantum probability measure over the lattice $\mathscr{L}(\mathbb{C}^3)$ of all orthogonal projection operators on the Hilbert space \mathbb{C}^3 .

Compatibility with the standard three-colour situation entails $\rho_R^2 = \frac{1}{3}$; hence, it follows from Eq. [\(1\)](#page-7-1) that

$$
|v\rangle = \left(\frac{1}{\sqrt{3}}e^{i\theta_R}, \rho_Y e^{i\theta_Y}, \rho_B e^{i\theta_B}\right)
$$

$$
= \left(\frac{1}{\sqrt{3}}e^{i\theta_R}, \rho_Y e^{i\theta_Y}, \sqrt{\frac{2}{3} - \rho_Y^2}e^{i\theta_B}\right)
$$
(4)

where the last equality is obtained from $\sqrt{\langle v|v\rangle} = 1$.

Special states are the state with no black balls represented by

$$
|v_{RY}\rangle = \left(\frac{1}{\sqrt{3}}e^{i\theta_R}, \sqrt{\frac{2}{3}}e^{i\theta_Y}, 0\right) \tag{5}
$$

and the state with no yellow balls represented by

$$
|v_{RB}\rangle = \left(\frac{1}{\sqrt{3}}e^{i\theta_R}, 0, \sqrt{\frac{2}{3}}e^{i\theta_B}\right)
$$
 (6)

The acts f_1 , f_2 , f_3 and f_4 are represented by the self-adjoint operators

$$
\hat{F}_1 = u(100)P_R + u(0)P_Y + u(0)P_B \tag{7}
$$

$$
\hat{F}_2 = u(0)P_R + u(0)P_Y + u(100)P_B \tag{8}
$$

$$
\hat{F}_3 = u(100)P_R + u(100)P_Y + u(0)P_B \tag{9}
$$

$$
\hat{F}_4 = u(0)P_R + u(100)P_Y + u(100)P_B \tag{10}
$$

respectively. The utility function $u(\cdot)$ is not given by the theory, but it is revealed in a decision test by concrete choices, in analogy with standard procedures in the literature.

The expected utility $W_v(f_i)$, $i \in \{1, 2, 3, 4\}$, in a generic state p_v of the entity Ω_{DM} , is

$$
W_v(f_1) = \langle v|\hat{F}_1|v\rangle = \frac{1}{3}u(100) + \frac{2}{3}u(0)
$$
 (11)

$$
W_v(f_2) = \langle v | \hat{F}_2 | v \rangle = \left(\frac{1}{3} + \rho_Y^2\right) u(0) + \left(\frac{2}{3} - \rho_Y^2\right) u(100)
$$
\n(12)

$$
W_v(f_3) = \langle v | \hat{F}_3 | v \rangle = \left(\frac{1}{3} + \rho_Y^2\right) u(100) + \left(\frac{2}{3} - \rho_Y^2\right) u(0)
$$
\n(13)

$$
W_v(f_4) = \langle v | \hat{F}_4 | v \rangle = \frac{1}{3}u(0) + \frac{2}{3}u(100)
$$
 (14)

As we can see, the expected utilities $W_v(f_1)$ and $W_v(f_4)$ do not depend on the cognitive state p_v of the entity Ω_{DM} , in agreement with the fact that f_1 and f_4 are unambiguous acts. On the contrary, the expected utilities $W_v(f_2)$ and $W_v(f_3)$ do depend on the cognitive state p_v , in agreement with the fact that f_2 and f_3 are ambiguous acts. This also agrees with our assumption that cognitive states provide information on ambiguity.

5.2 Reproducing Ellsberg preferences with ambiguity-averse states

Let us now suppose that, in the absence of any context, the cognitive entity Ω_{DM} is in the initial state p_0 . The principle of indifference (see Sect. [4\)](#page-6-0) then suggests that p_0 is the state represented by the unit vector

$$
|v_0\rangle = \frac{1}{\sqrt{3}}(1, 1, 1) \tag{15}
$$

leading to uniform probabilities of drawing a red, yellow and black ball. The ambiguity attitude contexts e_1 and e_2 will determine a change of state of the entity Ω_{DM} , depending on individual preferences towards ambiguity. For example, two ambiguity seeking states p_{w_1} and p_{w_2} will be such that the following inequalities hold

$$
W_{w_1}(f_2) > \frac{1}{3}u(100) + \frac{2}{3}u(0)
$$
 (16)

$$
W_{w_2}(f_3) > \frac{1}{3}u(0) + \frac{2}{3}u(100) \tag{17}
$$

We will instead explicitly determine two ambiguity-averse states p_{w_1} and p_{w_2} which reproduce Ellsberg preferences, that is,

$$
W_{w_1}(f_1) > W_{w_1}(f_2)
$$
\n(18)

$$
W_{w_2}(f_4) > W_{w_2}(f_3)
$$
\n(19)

Two general cognitive states p_{w_1} and p_{w_2} are represented by the unit vectors

$$
|w_1\rangle = \left(\frac{1}{\sqrt{3}}e^{i\theta_R}, \rho_Y e^{i\theta_Y}, \sqrt{\frac{2}{3} - \rho_y^2}e^{i\theta_B}\right)
$$
 (20)

$$
|w_2\rangle = \left(\frac{1}{\sqrt{3}}e^{i\phi_R}, \tau_Y e^{i\phi_Y}, \sqrt{\frac{2}{3} - \tau_y^2}e^{i\phi_B}\right)
$$
 (21)

respectively. For the sake of simplicity, let us look for states with simple phases, namely $\theta_R = \phi_R = 0$ and θ_Y , θ_B , ϕ_Y , $\phi_B = 0$, π . In particular, one can show that, for every $\alpha > \frac{1}{\sqrt{2}}$ $\frac{1}{3}$, the unit vectors

$$
|w_1\rangle = \left(\frac{1}{\sqrt{3}}, \alpha, -\sqrt{\frac{2}{3} - \alpha^2}\right)
$$
 (22)

$$
|w_2\rangle = \left(\frac{1}{\sqrt{3}}, -\sqrt{\frac{2}{3} - \alpha^2}, \alpha\right)
$$
 (23)

² Springer

reproduce Ellsberg preferences. However, the ambiguityaverse states p_{w_1} and p_{w_2} are not generally orthogonal, unless $\alpha = \pm 0.7887$. Let us choose the positive sign, so that the orthonormal vectors

$$
|w_1\rangle = (0.5774, 0.7887, -0.2113)
$$
 (24)

$$
|w_2\rangle = (0.5774, -0.2113, 0.7887)
$$
\n⁽²⁵⁾

reproduce Ellsberg preferences in the three-colour example within a quantum mathematical representation.

5.3 Modelling empirical data in Hilbert space

The final step of the quantum representation of the threecolour example consists in modelling the experimental data in Sect. [3.](#page-3-0) To this aim, we describe the decision between acts f_1 and f_2 by a measurement context e_{12} and represent the latter by the spectral family $\{M, 1 - M\}$, where the orthogonal projection operator *M* projects onto the onedimensional subspace generated by the unit vector $|m\rangle =$ $\left(\frac{1}{\sqrt{2}}\right)$ $\frac{1}{3}$, $\rho_Y e^{i\theta_Y}$, $\rho_B e^{i\theta_B}$, where $\rho_Y^2 + \rho_B^2 = \frac{2}{3}$. The onedimensional projection operator *M* is then given by:

$$
M = |m\rangle\langle m| = \begin{pmatrix} \frac{1}{3} & \frac{1}{\sqrt{3}} \rho_Y e^{-i\theta_Y} & \frac{1}{\sqrt{3}} \rho_B e^{-i\theta_B} \\ \frac{1}{\sqrt{3}} \rho_Y e^{i\theta_Y} & \rho_Y^2 & \rho_Y \rho_B e^{i(\theta_Y - \theta_B)} \\ \frac{1}{\sqrt{3}} \rho_B e^{i\theta_B} & \rho_Y \rho_B e^{-i(\theta_Y - \theta_B)} & \rho_B^2 \end{pmatrix}
$$
(26)

Analogously, we describe the decision between acts *f*³ and *f*⁴ by a measurement context *e*³⁴ and represent the latter by the spectral family $\{N, \mathbb{1} - N\}$, where the orthogonal projection operator *N* projects onto the one-dimensional subspace generated by the unit vector $|n\rangle = \left(\frac{1}{\sqrt{n}}\right)$ $\overline{3}$, $\tau_Y e^{i\phi_Y}$, $\tau_B e^{i\phi_B}$), where $\tau_Y^2 + \tau_B^2 = \frac{2}{3}$. The one-dimensional projection operator *N* is then given by:

$$
N = |n\rangle\langle n| = \begin{pmatrix} \frac{1}{3} & \frac{1}{\sqrt{3}}\tau_Y e^{-i\phi_Y} & \frac{1}{\sqrt{3}}\tau_B e^{-i\phi_B} \\ \frac{1}{\sqrt{3}}\tau_Y e^{i\phi_Y} & \tau_Y^2 & \tau_Y \tau_B e^{i(\phi_Y - \phi_B)} \\ \frac{1}{\sqrt{3}}\tau_B e^{i\phi_B} & \tau_Y \tau_B e^{-i(\phi_Y - \phi_B)} & \tau_B^2 \end{pmatrix}
$$
(27)

We refer to the unit vectors $|v_0\rangle = \frac{1}{\sqrt{2}}$ $\frac{1}{3}(1, 1, 1)$ and $|w_1\rangle$ and $|w_2\rangle$ in Eqs. [\(24\)](#page-9-1) and [\(25\)](#page-9-1). It follows that the conditions

$$
\langle m|m\rangle = 1\tag{28}
$$

$$
\langle n|n\rangle = 1\tag{29}
$$

$$
\langle w_1 | M | w_1 \rangle = 0.815 \tag{30}
$$

$$
\langle w_2|N|w_2\rangle = 0.780\tag{31}
$$

$$
\langle v_0 | M | v_0 \rangle = 0.500 \tag{32}
$$

$$
\langle v_0|N|v_0\rangle = 0.500\tag{33}
$$

must be satisfied by the real parameters ρ_Y , ρ_B , θ_Y , θ_B , τ_Y , τ_B , ϕ_Y and ϕ_B . Equations [\(28\)](#page-9-2) and [\(29\)](#page-9-2) are determined by normalization conditions, while Eqs. [\(30\)](#page-9-2) and [\(31\)](#page-9-2) are determined by empirical data. Finally, Eqs. (32) and (33) are determined by the fact that decision-makers who are not sensitive to ambiguity should overall be indifferent between *f*¹ and *f*2, as well as between *f*³ and *f*4. Hence, on average, half respondents are expected to prefer $f_1(f_3)$ and the other half f_2 (f_4). To simplify the analysis, let us set $\theta_Y = \phi_B = 0$. Hence, we are left with a system of 6 equations in 6 unknown variables whose solution is

$$
\begin{cases}\n\rho_Y = 0.6853 \\
\rho_B = 0.4438 \\
\theta_B = 105.07^\circ \\
\tau_Y = 0.4785 \\
\tau_B = 0.6616 \\
\phi_Y = 102.87^\circ\n\end{cases} (34)
$$

Equivalently, we get

$$
|m\rangle = (0.5773, 0.6853, 0.4438e^{i105.07^{\circ}})
$$
\n(35)

$$
|n\rangle = (0.5773, 0.4785e^{i102.87^\circ}, 0.6616)
$$
 (36)

Thus, the orthogonal projection operators in Eqs. [\(26\)](#page-9-3) and [\(27\)](#page-9-4) reproducing the experimental data in Sect. [3](#page-3-0) are

$$
M = \begin{pmatrix} 0.333 & 0.396 & 0.256e^{-i105.07^{\circ}} \\ 0.396 & 0.470 & 0.304e^{-i105.07^{\circ}} \\ 0.256e^{i105.07^{\circ}} & 0.304e^{i105.07^{\circ}} & 0.197 \end{pmatrix}
$$

\n
$$
N = \begin{pmatrix} 0.333 & 0.276e^{-i102.87^{\circ}} & 0.382 \\ 0.276e^{i102.87^{\circ}} & 0.229 & 0.317e^{i102.87^{\circ}} \\ 0.382 & 0.317e^{-i102.87^{\circ}} & 0.438 \end{pmatrix}
$$
(38)

The construction of a quantum model for the data on the Ellsberg three-colour experiment in Sect. [3](#page-3-0) is thus completed. As we can see, the quantum model naturally arises from the canonical Hilbert space representation of the realistic– operational terms in Sect. [4.](#page-6-0)

6 Conclusions

Despite its phenomenological success to deal with classically problematical cognitive puzzles, the quantum cognition research programme still poses challenging questions regarding its explanatory power and necessity.

In this paper, we specialized to the Ellsberg paradox decision situation a realistic–operational foundation which we have recently extended from physics to cognition. Then, we applied to the Ellsberg three-colour example the canonical quantum representation of realistic–operational terms in Hilbert space.

This result on the Ellsberg paradox situation is paradigmatic, in the sense that one can follow the same strategy to generally claim that the mathematical representation of human judgements and decision-making in Hilbert space has now an independent motivation of a foundational, rather than phenomenological, nature.

To conclude, we agree that quantum theory in Hilbert space is not the ultimate theory in cognition—recent results on sequential measurements and order effects seem to con-firm this conclusion (see, e[.](#page-10-24)g. Aerts et al. [2016](#page-10-24)). However, we also believe that there are strong theoretical motivations, in addition to its empirical success and unitary explanation, to continue using Hilbert space structures in cognition.

Compliance with ethical standards

Conflict of interest The author has no conflict of interest to declare.

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