Probability

April 23, 2016

1 Probability Spaces

1.1 Classical Probability Spaces

Probability theory [7, 9, 19] is defined using the notions of a sample space Ω , a space of events \mathcal{E} , and a probability measure μ . In this paper, we will only consider finite sample spaces: we therefore define a sample space Ω as an arbitrary non-empty finite set, the space of events \mathcal{E} as 2^{Ω} , the powerset of Ω , and the probability measure as a function $\mu: \mathcal{E} \to [0,1]$ such that:

- $\mu(\Omega) = 1$, and
- for a collection of pairwise disjoint events E_i , the probability measures are additive $\mu(\bigcup E_i) = \sum \mu(E_i)$.

Example of a problem on a finite sample space (Two coin experiment) Consider an experiment that tosses two coins. We have four possible outcomes that constitute the sample space $\Omega = \{HH, HT, TH, TT\}$. The event that the first coin is "heads" is $\{HH, HT\}$; the event that the two coins land on opposite sides is $\{HT, TH\}$; the event that at least one coin is tails is $\{HT, TH, TT\}$. Depending on the assumptions regarding the coins, we can define several probability measures. Here is a possible one:

```
\mu(\emptyset)
                                           \mu(\{HT, TH\}) =
    \mu(\{HH\})
                   1/3
                                            \mu(\{HT,TT\})
                                                               0
    \mu(\{HT\})
                   0
                                            \mu(\{TH, TT\}) = 2/3
    \mu(\{TH\})
               = 2/3
                                      \mu(\{HH, HT, TH\})
                                                          = 1
     \mu(\{TT\})
                = 0
                                       \mu(\{HH, HT, TT\}) = 1/3
\mu(\{HH, HT\})
                  1/3
                                       \mu(\{HH,TH,TT\})
                                                           = 1
\mu(\{HH,TH\})
                                       \mu(\{HT, TH, TT\})
\mu(\{HH,TT\}) =
                  1/3
                                  \mu(\{HH, HT, TH, TT\}) =
```

Note that the probability measure for disjoint events such as $\{HT\}$ and $\{TH\}$ do indeed add.

1.2 Quantum Probability Spaces

The mathematical framework above assumes that one has complete knowledge of the events and their relationships. But even in many classical situations, the structure of the event space is only partially known and the precise dependence of two events on each other cannot be determined with certainty. In the quantum case, this partial knowledge is compounded by the fact that not all quantum events can be observed simultaneously. Indeed, in the quantum world, there are non-commuting events which cannot even happen simultaneously. To accommodate these more complex situations, we abandon the sample space Ω and define and reason directly about events. A quantum probability space consist of just two components: a set of events \mathcal{E} and a probability measure $\mu: \mathcal{E} \to [0,1]$. We give an example before giving the formal definition.

Consider the two-qubit Hilbert space with computational basis $|0\rangle$ and $|1\rangle$ and states:

$$|+\rangle = \frac{|0\rangle + |1\rangle}{\sqrt{2}}, \qquad |-\rangle = \frac{|0\rangle - |1\rangle}{\sqrt{2}}.$$

The set of events associated with this Hilbert space consists of all projections including the empty projection $\mathbb{1} = |0\rangle\langle 0| + |1\rangle\langle 1|$:

$$\{0, |0\rangle\langle 0|, |1\rangle\langle 1|, |+\rangle\langle +|, |-\rangle\langle -|, \dots, 1\}$$

Each event is interpreted as a possible post post-measurement state of a quantum system as follows: given some arbitrary current quantum state $|\psi\rangle$ to be measured, the event $|0\rangle\langle 0|$ states that the post-measurement state will be $|0\rangle$; the event $|1\rangle\langle 1|$ states that the post-measurement state will be $|1\rangle$; the event $|+\rangle\langle +|$ states that the post-measurement state will be $|+\rangle$; the event $|-\rangle\langle -|$ states that the post-measurement state will be a linear combination of $|0\rangle$ and $|1\rangle$; and the event 0 states that the post-measurement state will be the empty state.

Irrespective of the current state $|\psi\rangle$ and irrespective of the particular experiment, the probability of event 0 will always be 0 (it is an impossible event) and the probability of event 1 will always be 1 (it is a certain event). The probabilities attached to other events will depend on the particular state in question. If the state is $|0\rangle$, the probability of event $|0\rangle\langle 0|$ is 1; the probability of event $|1\rangle\langle 1|$ is 0; the probability of event $|+\rangle\langle +|$ is $\frac{1}{2}$; and the probability of event $|-\rangle\langle -|$ is $\frac{1}{2}$. If the state is $|+\rangle$, the probability of each event $|0\rangle\langle 0|$ and $|1\rangle\langle 1|$ will be $\frac{1}{2}$; the probability of event $|+\rangle\langle +|$ is 1; and the probability of event $|-\rangle\langle -|$ is 0.

We now formalize a quantum probability space as follows [4, 8, 18, 1, 13]. We first assume an ambient Hilbert space \mathcal{H} and define the set of events \mathcal{E} as all projections on \mathcal{H} . Each quantum state $|\psi\rangle \in \mathcal{H}\setminus\{0\}$ induces a probability measure $\mu_{\psi}: \mathcal{E} \to [0, 1]$ on the space of events defined for any event $E \in \mathcal{E}$ as follows¹:

$$\mu_{\psi}(E) = \langle \psi | E \psi \rangle \tag{1}$$

which is called the Born rule [5, 14].

Yu-Tsung says: One small thing is whether we should say

for each normalized state $|\psi\rangle...\mu_{\psi}(E) = \langle \psi|E\psi\rangle$

as in Mermin [14] and Maassen [13] or

for each state
$$|\psi\rangle \in \mathcal{H} \setminus \{0\}...\mu_{\psi}(E) = \frac{\langle \psi | E\psi \rangle}{\langle \psi | \psi \rangle}$$

as in Abramsky [1]. Personally, I prefer Abramsky's approach, because normalizing a state in DQT is messier than in CQT, and we have already considered states whose norm is not one in our previous paper [11].

Our previous examples can be verified in terms of μ_{ψ} . Because $|0\rangle\langle 0|$ and $|-\rangle\langle -|$ are non-commutative, their total probabilities are not a probability of an event

$$\mu_0(|0\rangle\langle 0|) + \mu_0(|-\rangle\langle -|) = \frac{3}{2} > 1.$$

When events are orthogonal, i.e., commutative, they can be measured together and their probability is additive:

$$\mu_0(|0\rangle\langle 0|) + \mu_0(|1\rangle\langle 1|) = \mu_0(|+\rangle\langle +|) + \mu_0(|-\rangle\langle -|) = 1 = \mu_0(1)$$
.

Similarly to the classical case, the total probability is one and the orthogonal additivity can be summarized into the following two rules:

¹Recently, people extend the domain of μ_{ψ} to all operators \mathcal{A} on \mathcal{H} and consider $\mu_{\psi}: \mathcal{A} \to \mathbb{C}$ [13, 20]. When an operator $A \in \mathcal{A}$ is Hermitian, $\mu_{\psi}(A)$ is the expectation value of A. We does not take this approach because we want to focus only on probability.

- $\mu(1) = 1$, and
- for a collection of pairwise orthogonal E_i , we have $\mu(\sum_i E_i) = \sum_i \mu(E_i)$.

These two conditions define quantum probability measures. Recall that the Born rule (1) induces a probability measure from a state. Naturally, we may ask: is every probability measure induced from a state by the Born rule? The answer is yes by Gleason's theorem when the dimension ≥ 3 [8, 15, 18].

Yu-Tsung says: There are three ways to charactize a quantum probability:

- 1. The above two conditions as Gleason did.
- 2. As a linear functional on a * algebra.
- 3. As induced by a state vector by the Born rule (1).

According to equation (37) in section 4 Quantum Probability in [13], the equalivance between 1. and 2. only hold for $d \geq 3$ because of Gleason' theorem. Therefore, it seems hard to get insight for how to handle Gleason's theorem by *-algebra formulism...

The equalivance between 2. and 3. can be found in Proposition 4.1.1 in [20]. Notice that this proposition holds for every dimension... So it is not exactly Gleason's theorem, or at least not the main part of the Gleason's theorem...

1.3 Plan

In the remainder of the paper, we consider variations of quantum probability spaces motivated by computation of numerical quantities in a world with limited resources:

- Instead of the Hilbert space \mathcal{H} (constructed over the uncountable and uncomputable complex numbers \mathbb{C}), we will consider variants constructed over finite fields [12, 11, 10].
- Instead of real-valued probability measures producing results in the uncountable and uncomputable interval [0, 1], we will consider finite set-valued probability measures [3, 17].

We will then ask if it is possible to construct variants of quantum probability spaces under these conditions. The main question is related to the definition of probability measures: is it possible to still define a probability measure as a function that depends on a single state? Specifically,

- given a state $|\psi\rangle$, is there a probability measure mapping events to probabilities that only depends on $|\psi\rangle$? In the conventional quantum probability space, the answer is yes by the Born rule [5, 14] and the map is given by: $E \mapsto \langle \psi | E \psi \rangle$.
- given a probability measure μ mapping each event E to a probability, is there a unique state ψ such that $\mu(E) = \langle \psi | E \psi \rangle$? In the conventional case, the answer is yes by Gleason's theorem [8, 15, 18].

2 All Continuous or All Discrete

Before we turn to the main part of the paper, we quickly dismiss the possibility of having one but not the other of the discrete variations. Speficially, it is impossible to maintain the Hilbert space and have a finite set-valued probability measure and it is also impossible to have a vector space constructed over a finite field with a real-valued probability measure.

2.1 Hilbert Space with Finite Set-Valued Probability Measure

In order to simplify the discussion, we will focus on $\mathcal{L}_2 = \{\text{impossible}, \text{possible}\}\$ where impossible and possible means $\{0\}$ and $\{0,1\}$ respectively while our result can be easily extend to other values with the form $\{\{0\}, \{0, \frac{1}{n}\}, \dots, (\frac{n-1}{n}, 1]\}$.

Because \mathcal{L}_2 is a partition of [0,1] [2, 21, 16, 6], there is a natural surjective map $\iota : [0,1] \to \mathcal{L}_2$ defined by $\iota(0) = \text{impossible}$ and $\iota(x) = \text{possible}$ for $x \neq 0$. By applying ι , we can define the corresponding Born rule $\bar{\mu}$ by $\bar{\mu}_{\psi}(E) = \iota(\mu_{\psi}(E))$, and the corresponding probability measure $\hat{\mu}$ should satisfy:

- $\hat{\mu}(1) = \text{possible}$, and
- for a collection of pairwise orthogonal E_i , we have $\hat{\mu}(\sum_i E_i) = \bigvee_i \hat{\mu}(E_i)$, where $x \vee y = \text{impossible}$ if and only if x = y = impossible.

However, there is a \mathscr{L}_2 -valued probability measure

$$\hat{\mu}_1(E) = \begin{cases} \text{impossible} & \text{, if } E = |+\rangle\langle +|; \\ \bar{\mu}_0(E) & \text{, otherwise.} \end{cases}$$

such that $\hat{\mu}_1 \neq \bar{\mu}_{\psi}$ for all mixed state $|\psi\rangle$.

2.2 Discrete Vector Space with Real-Valued Probability Measure

References

- [1] Samson Abramsky. Big toy models: Representing physical systems as Chu spaces. CoRR, abs/0910.2393, 2009.
- [2] M. Artin. Algebra. Prentice Hall, 1991.
- [3] Zvi Artstein. Set-valued measures. Transactions of the American Mathematical Society, 165:103–125, 1972.
- [4] Garrett Birkhoff and John Von Neumann. The logic of quantum mechanics. *Annals of mathematics*, pages 823–843, 1936.
- [5] Max Born. Zur quantenmechanik der stoßvorgänge (1926). In *Die Deutungen der Quantentheorie*, pages 48–52. Springer, 1984.
- [6] David S. Dummit and Richard M. Foote. Abstract Algebra. John Wiley & Sons, Inc., 2004.
- [7] William G. Faris. Appendix: Probability in quantum mechanics. In *The infamous boundary : seven decades of controversy in quantum physics*. Boston : Birkhauser, 1995.
- [8] Andrew Gleason. Measures on the closed subspaces of a hilbert space. *Indiana Univ. Math. J.*, 6:885–893, 1957.
- [9] R.L. Graham, D.E. Knuth, and O. Patashnik. Concrete Mathematics: A Foundation for Computer Science. Addison-Wesley, 1994.
- [10] Andrew J Hanson, Gerardo Ortiz, Amr Sabry, and Yu-Tsung Tai. Geometry of discrete quantum computing. *Journal of Physics A: Mathematical and Theoretical*, 46(18):185301, 2013.
- [11] Andrew J Hanson, Gerardo Ortiz, Amr Sabry, and Yu-Tsung Tai. Discrete quantum theories. *Journal of Physics A: Mathematical and Theoretical*, 47(11):115305, 2014.
- [12] Andrew J Hanson, Gerardo Ortiz, Amr Sabry, and Yu-Tsung Tai. Corrigendum: Geometry of discrete quantum computing. *Journal of Physics A: Mathematical and Theoretical*, 49(3):039501, 2015.
- [13] Hans Maassen. Quantum probability and quantum information theory. In *Quantum information*, computation and cryptography, pages 65–108. Springer, 2010.

- [14] N. D. Mermin. Quantum Computer Science. Cambridge University Press, 2007.
- [15] A. Peres. Quantum Theory: Concepts and Methods. Fundamental Theories of Physics. Springer, 1995.
- [16] Paul Walton Purdom and Cynthia A Brown. *The analysis of algorithms*. Oxford University Press, USA, 1995.
- [17] Madan L Puri and Dan A Ralescu. Strong law of large numbers with respect to a set-valued probability measure. *The Annals of Probability*, pages 1051–1054, 1983.
- [18] Michael Redhead. Incompleteness, Nonlocality, and Realism: A Prolegomenon to the Philosophy of Quantum Mechanics. Oxford University Press, 1987.
- [19] V.K. Rohatgi and A.K.M.E. Saleh. *An Introduction to Probability and Statistics*. Wiley Series in Probability and Statistics. Wiley, 2011.
- [20] Jan Swart. Introduction to quantum probability. Lecture Notes, 2013.
- [21] Wolfgang Wechler. Universal algebra for computer scientists, volume 25 of eatcs monographs on theoretical computer science, 1992.