State space dimensionality in short memory hidden variable theories

Alberto Montina
Perimeter Institute for Theoretical Physics, 31 Caroline Street North, Waterloo, Ontario N2L 2Y5, Canada
(Dated: October 11, 2010)

Recently we have presented a hidden variable model of measurements for a qubit where the hidden variable state space dimension is one-half the quantum state manifold dimension. The absence of a short memory (Markov) dynamics is the price paid for this dimensional reduction. The conflict between having the Markov property and achieving the dimensional reduction was proved in [A. Montina, Phys. Rev. A, 77, 022104 (2008)] using an additional hypothesis of trajectory relaxation. Here we analyze in more detail this hypothesis introducing the concept of invertible process and report a proof that makes clearer the role played by the topology of the hidden variable space. This is accomplished by requiring suitable properties of regularity of the conditional probability governing the dynamics. In the case of minimal dimension the set of continuous hidden variables is identified with an object living an N-dimensional Hilbert space, whose dynamics is described by the Schrödinger equation. A method for generating the economical non-Markovian model for the qubit is also presented.

I. INTRODUCTION

One of the most peculiar features of quantum mechanics is the exponential growth of resources required to define the quantum state $|\psi\rangle$ of a composite system. It makes the direct simulation of even a handful of particles impossible in practice. This growth is due to the fact that $|\psi\rangle$ contains the full statistical information about the probabilities of any possible event, such as the joint probability $p(s_1,...,s_N)$ of obtaining the outcomes $\{s_1,...,s_N\}$ by measuring the z axis components of $N$ 1/2-spins. The information in $\rho(s_1,...,s_N)$ grows exponentially for a given accuracy and considerably exceeds the classical information required to specify the actual measurement outcome. Since the quantum state is not a physically accessible observable, but can be statistically reconstructed only by performing many measurements on different replicas [1], it is natural to wonder if this resource excess is strictly necessary to describe the actual state of a single realization. The quantum probabilities could be reproduced by a hidden variable theory where a single system carries less information than the quantum state. In such a theory the quantum state $|\psi\rangle$ is mapped to a probability distribution on a space $X$ of hidden variables, i.e.,

$$|\psi\rangle \rightarrow \rho(X|\psi).$$

It is clear that the sampling space $X$ can have in principle a smaller dimension than the quantum state manifold. For example the space of functions on a one-dimensional domain is infinite-dimensional and any finite-dimensional Hilbert space can be embedded within it.

In accordance with recent terminology, we will refer to the actual state $X$ of a quantum system and the corresponding space as ontic state and ontological space [2], respectively, and name the dimensional reduction of the ontological space ontological shrinking. It is interesting to note that, in any known short memory (Markov) hidden variable theory, the dimension of the state space is never smaller than the quantum state manifold dimension. As an example, in the de Broglie-Bohm model the wave-function has the role of a field of physical quantities and is supplied by additional variables describing the particle positions.

The ontological shrinking has a connection with the concept of classical ”weak simulation” [3,4] in quantum information theory. In a classical ”strong simulation” of a quantum computer, the goal is to evaluate the measurement probabilities with high accuracy. This requirement is stronger than necessary, since in a real quantum computer a single run does not give the measurement probabilities and the output is a precise event. The probabilities concern the behavior of many experimental realizations. The goal of a classical ”weak simulation” is not to compute the probability weights, but the outcomes in accordance with the weights. There are examples of quantum circuits that cannot be efficiently simulated in the strong way, but whose weak simulation is nevertheless tractable [4]. In a hidden variable theory with reduced sampling space, the evaluation of the actual dynamics of a single realization would require less resources than the computation of the quantum state dynamics. Thus, the ontological shrinking could offer in a natural way an efficient method of ”weak simulation” of quantum computers.

In the recent years the possibility of a statistical representation of quantum states on a reduced sampling space was discussed by various authors [5-10]. The problem of the smallest dimension of the ontological space was posed in Ref. [8]. It was subsequently proved that the ontological dimension cannot be smaller than the quantum state manifold dimension in the case of a Markov hidden variable theory with an additional hypothesis of trajectory relaxation [8]. We will refer to this result as the noshrinking theorem. Recently we reported an example of hidden variable model of measurements for a qubit whose state space is one-dimensional, i.e., smaller than the two-dimensional Bloch sphere [10]. As a consequence of the dimensional reduction, the dynamics is not a Markov process. This counterexample makes evident
that the short memory hypothesis is strictly necessary for the proof of the theorem in Ref. [7]. In this article, we review the one-dimensional model providing a method for generating it and analyze in more detail the hypothesis of trajectory relaxation. We define the concept of \textit{invertible process} and show that it is always possible to find a sub-region of a compact ontological space where all the processes are invertible. Discarding insignificant states, we present a new version of the \textit{no-shrinking} theorem that makes clearer the role played by the topology of the hidden variable space. This is accomplished by requiring reasonable properties of regularity of the conditional probability governing the dynamics and explicitly using them in the theorem proof. By the way, it is useful to remind that the dimension is a topological property and is not defined by the space cardinality.

In section \textbf{III} we introduce the general framework of a hidden variable theory. In section \textbf{IV} the economical ontological model in Ref. [10] is reviewed. The properties of regularity and the concept of invertible process are introduced in section \textbf{V} where we also prove the no-shrinking theorem and discuss its consequences in terms of resource cost. In the same section we show that in any Markov hidden variable theory with minimal space dimension the set of continuous hidden variables can be identified with an object, living in an $N$-dimensional Hilbert space, whose dynamics is described by the Schrödinger equation. In appendix, we report a systematic construction method to generate the model in Sec. \textbf{III} starting from a particular form of the probability distribution.

\section{II. GENERAL FRAMEWORK}

In a hidden variable theory, the quantum state is translated into a classical language by replacing it with a probability distribution on a sampling space $X$ of ontic states. We assume that the ontological space is a $M$-dimensional manifold described by an $M$-tuple $\vec{x}$ of continuous variables and a possible discrete index $n$. The mapping (1) is not the most general, since the probability distribution could depend on the preparation context. For example, a 1/2-spin can be prepared in the up state by merely selecting the beam outgoing from a Stern-Gerlach apparatus or can be prepared in the state $|\uparrow\rangle + |\downarrow\rangle$ and then suitably rotated. In order to account for this possible dependence, we add suitable parameters $\eta$ that identify the preparation context (1), i.e.,

$$|\psi\rangle \rightarrow \rho(\vec{x}, n|\psi, \eta).$$

(2)

The set of ontological variables $X = (\vec{x}, n)$ contains the whole information about a single realization, thus the probability of any event is conditioned only by it. In particular, for the measurement of the trace-one projector $|\phi\rangle \langle \phi|$, it is assumed that there exists a conditional probability $P$ for the event $|\phi\rangle$ given the ontic state $X$. In general also $P$ could depend on the context of the measurement, thus we introduce additional parameters $\tau$ in the conditional probability, i.e.,

$$|\phi\rangle \rightarrow P(|\phi\rangle \vec{x}, n, \tau).$$

(3)

The probability of the event $|\phi\rangle$ given $|\psi\rangle$ has to be equal to the quantum mechanical probability,

$$\sum_n \int d^M x P(|\phi\rangle \vec{x}, n, \tau) \rho(\vec{x}, n|\psi, \eta) = |\langle \phi|\psi\rangle|^2.$$  

(4)

Finally, it is assumed that the dynamics at the ontological level is Markovian [12]. The ontic state evolves from $X_1$ at time $t_1$ to $X_2$ at time $t_2$ according to a conditional probability $K(X_2, t_2|X_1, t_1)$. $K$ satisfies the Chapman-Kolmogorov equation [12] and is a delta distribution for $t_2 = t_1$.

$$K(\vec{x}_2, n_2, t_2|\vec{x}_1, n_1, t_1) = \delta_{n_1, n_2} \delta(\vec{x}_2 - \vec{x}_1).$$

(5)

For a time-homogeneous process the transition probability depends only on the time difference $t_2 - t_1$,

$$K(X_2, t_2|X_1, t_1) = K(X_2, t_2 - t_1|X_1, 0) \equiv K(X_2|X_1, t_2 - t_1)$$

(6)

In order to link the quantum language with the classical one, we can label $K$ with the corresponding unitary operator $\hat{U}$. As for the state preparation and measurement, in general the conditional probability can depend on additional parameters $\chi$, the transformation context, i.e.,

$$\hat{U} \rightarrow K_{\hat{U}, \chi}(X_2|X_1).$$

(7)

Indeed an operator $\hat{U}$ can be physically implemented in different ways. For example, the spin rotation $e^{-it\hat{\sigma}_z}$ can be performed directly rotating along the $x$ axis or implementing the three-step rotation $e^{-it\hat{\sigma}_z}e^{-it\hat{\sigma}_y}e^{-it\hat{\sigma}_z}$ along the $z$ and $y$ axes. These two schemes are physically different and not necessarily described by the same conditional probability. By the way, it is useful to note that $K_{\hat{U}, \chi}(X|X)$ for $\hat{U} = I$ is not necessarily the delta distribution $\delta(X - \bar{X})$ for all the contexts. For example the identity evolution can correspond to physically performing the three step rotation $e^{-it\hat{\sigma}_z}e^{-it\hat{\sigma}_y}e^{-it\hat{\sigma}_z}$ followed by the inverse transformation $e^{it\hat{\sigma}_z}$. The overall operation is not equal to doing nothing and does not necessarily correspond to a delta peaked conditional probability.

If the quantum state $|\psi\rangle$ evolves to $\hat{U}|\psi\rangle \equiv |\bar{\psi}\rangle$, the associated probability $\rho(X|\psi, \eta)$ evolves to

$$\rho(X|\bar{\psi}, \bar{\eta}) \equiv \int dY K_{\hat{U}, \chi}(X|Y)\rho(Y|\psi, \eta).$$

(8)

Some regularity properties of $K_{\hat{U}, \chi}$ will be discussed in Sec. \textbf{IV}. Any short memory hidden variable theory has this general structure.
The events \( \vec{w} \) with \( w_z < 0 \) correspond simply to the non-occurrence of the events \( -\vec{w} \) with \( w_z > 0 \), i.e., \( P(-\vec{w}|x, n) = 1 - P(\vec{w}|x, n) \).

It is easy to prove that these probability functions fulfill the condition \((\ref{3.1})\), that is,
\[
P(\vec{w}|\varphi, 0)\sin \theta + P(\vec{w}|\theta, 1)(1-\sin \theta) = \frac{1}{2}(1+\vec{w}\cdot\vec{v}) \tag{13}
\]

As shown in Ref. \cite{10}, the conditional probabilities \( P(\vec{w}|x, 0) \) and \( P(\vec{w}|x, 1) \) are always smaller than or equal to 1, but \( P(\vec{w}|x, 1) \) is positive only if \( \theta < \theta_0 \equiv \arccos\left(\frac{2}{3}\right) \approx 53.12 \text{ degrees} \).

Thus, the model in this form works only for a set of prepared states whose Bloch vector lies inside a cone with aperture \( 2\theta_0 \), the \( z \) axis being the symmetry axis. Since the positive region has non-zero measure in the quantum state manifold, it is possible to extend the model to the whole Bloch sphere by covering the manifold with a sufficiently large number of patch regions with different coordinate systems and enriching the ontic state with a finite quantity of information. This was accomplished in Ref. \cite{10} by adding a discrete index \( m \) taking 12 possible values, labelling 12 different regions of the Bloch sphere. See the referred paper for further details.

This is a concrete example of ontological shrinking, where the ontic state space is smaller than the quantum state manifold, which is in this case the two-dimensional Bloch sphere. It is a remarkable fact that a single realization of the ontic state \( \{x, n\} \) contains less information than the quantum state. The whole information on \( |\psi\rangle \) is contained in the probability distribution \( \rho \).

The absence of a short memory description of the dynamics is the price paid for the dimensional reduction. Indeed it is impossible to associate a positive conditional probability with each unitary evolution, that is, it is impossible to satisfy the identity
\[
\rho(x, n|\hat{U}\psi) = \sum_{\hat{n}} \int d\vec{x} K_{\hat{U}}(x, n|\vec{x}, \hat{n})\rho(\vec{x}, \hat{n}|\psi) \tag{14}
\]

with the probability distribution in Eq. \((9)\), apart from the unitary evolution \( e^{i\Delta t}\hat{U} \). Indeed in the other cases the dynamical equation of \( \rho(x, n|\psi) \) is in general nonlinear. Let us consider the unitary evolution with the Pauli matrix \( \hat{\sigma}_y \) as generator. The dynamical equations of the Bloch vector are
\[
\frac{\partial x}{\partial t} = v_z, \\
\frac{\partial y}{\partial t} = 0, \\
\frac{\partial z}{\partial t} = -v_x,
\]
which correspond in spherical coordinates to
\[
\frac{\partial \varphi}{\partial t} = -\cot \theta \sin \varphi, \\
\frac{\partial \theta}{\partial t} = \cos \varphi.
\]

Let \( K(x, n|\vec{x}, \hat{n}, t) \) be the transition probability associated with the evolution in Eq. \((15)\), we have from Eq. \((14)\)
that
\[ \frac{\partial \rho(x,n,t)}{\partial t} = \sum_n \int dx \frac{\partial K(x,n|x,\bar{n},t)}{\partial t} \rho(x,\bar{n}), \] (17)
that becomes by means of Eq. [9]
\[ \sin \theta_0 \frac{\partial \delta(x-\varphi)}{\partial x} + (1 - \sin \theta) \delta_n \frac{\partial}{\partial x} + \cos \theta \frac{\partial}{\partial (n,\delta(x-\varphi))} \delta_n \frac{\partial}{\partial \theta}, \] (18)

In particular, for \( n = 0 \), and using Eq. [10],
\[ -\cos \theta \frac{\partial \delta(x-\varphi)}{\partial x} \sin \varphi + \cos \theta \delta(x-\varphi) \cos \varphi = \frac{\partial K}{\partial x}(x,\varphi,0,t) \sin \theta + \frac{\partial K}{\partial \theta}(x,\varphi,0,t,1,t)(1 - \sin \theta). \] (19)

Dividing both sides by \( \sin \theta \) and differentiating with respect to \( \theta \), we obtain that
\[ \frac{\partial}{\partial \theta} \left[ -\cot \theta \frac{\partial \delta(x-\varphi)}{\partial x} \sin \varphi + \cot \theta \delta(x-\varphi) \cos \varphi \right] = \frac{\partial}{\partial \theta} \left[ \frac{\partial K}{\partial x}(x,0|\theta,1,t) \frac{\sin \theta}{\cos \theta} \right], \] (20)

This equation is not satisfied by any function \( K \), since the left-hand side is a function of both \( \theta \) and \( \varphi \), whereas the right-hand side depends only on \( \theta \). Thus, the dynamical equation of the probability distribution \( K \) is non-Markovian.

IV. ONTOLOGICAL SHRINKING AND MARKOV PROCESSES

In this section we will prove that the ontological shrinking is in contradiction with a Markov dynamics. This purpose is achieved by means of a very reasonable hypothesis concerning the support of the conditional probability \( K(X_2|X_1,t) \). It will be introduced in the following subsection. In subsection [IV B] we will introduce the concept of invertible process and will show that it is always possible to find a sub-region of a compact ontological space where all the processes are invertible. Then we prove the no-shrinking theorem in subsection [IV C]. In subsection [IV D] it is shown that in the case of minimal ontological dimension it is possible to identify the set of continuous ontological variables with a vector of the Hilbert space whose dynamics is given by the Schrödinger equation. In the last subsection we discuss the consequences of the no-shrinking theorem in terms of resource cost.

A. Conditional probabilities and associated unitary operators

For the sake of simplicity, from now on we will omit without loss of generality the discrete index \( n \) in the definition of the ontic state and assume that the ontological space is a differentiable manifold whose points are identified by the vector \( \vec{x} \). When the local Euclidean structure is not required, we will use the more generic symbol \( X \) to indicate the ontic state.

In order that a transition probability describing a Markov process makes physical sense, it has to satisfy some conditions of regularity. For our purpose, it is sufficient to require a very weak condition. As a reasonable hypothesis, we assume that there exists a subset of the support of the conditional probability \( K(X_2|X_1,t) \) that changes smoothly with respect to the condition \( X_1 \) and the evolution time \( t \). Let us refine this statement in a more precise way.

Property 1: given a time-homogeneous process \( X_1 \to X_2 \) with non-zero probability \( K(\vec{x}_2|\vec{x}_1,t) \), there exist an \( M \times M \) matrix \( \hat{A} \) and a vector \( \vec{a} \) such that \( K(\vec{x}_2+\lambda \delta \vec{x}_1+\hat{A} \delta t|\vec{x}_1, t+\delta t) \neq 0 \) for any infinitesimal variation \( \delta \vec{x}_1 \) and \( \delta t \). The matrix \( \hat{A} \) and the vector \( \vec{a} \) are functions of \( t \) and the process \( X_1 \to X_2 \).

It is important to note that this property is fulfilled for the very large class of Markov processes that involve drift, diffusion and jumps (they are discussed for example in Ref. [12]). Furthermore, the matrix \( \hat{A} \) and the vector \( \vec{a} \) are not necessarily unique. For example in the case of a stochastic process any choice of \( \hat{A} \) and \( \vec{a} \) is suitable, since the conditional probability is a multidimensional smooth function whose support is the whole manifold. For a pure deterministic process both \( \hat{A} \) and \( \vec{a} \) are unique. The latter gives the drift velocity of the ontic state. In the case of pure jumps with finite transition probability, one choice is \( \vec{a} = 0 \) and is unique if the jump distance cannot be arbitrarily small. One can imagine more complicated cases that involve diffusion in some direction, drift in the other ones and jumps, however also in these situations the stated property is fulfilled for some \( \hat{A} \) and \( \vec{a} \).

Property 1 strictly depends on the topology of the ontological space since it involves its local Euclidean structure. Using this structure in the proof of the no-shrinking theorem is fundamental because the space dimension is not a property defined merely by the cardinality of the space.

The quantum unitary evolution is a continuous process and corresponds to a trajectory in a Lie group manifold with the Hamiltonian as generator. For a time-homogeneous process, the unitary operator of the evolution has the form \( e^{-it\hat{H}} \), where \( \hat{H} \) is the transformation generator and \( t \) is the evolution time. In general, it is not possible to directly implement every generator of the \( su(N) \) algebra by means of a purely time-homogeneous process. In practice, only a small set of evolutions (building blocks) can be directly generated by a physical process. The other evolutions are obtained by suitably concatenating the building blocks. Given any generator \( \hat{G} \), it will be possible in principle to experimentally implement a unitary evolution \( \hat{U}_1 \), an physically attainable time-homogeneous process \( e^{-it\hat{H}} \) and another unitary evolution \( \hat{U}_2 \) such that
\[ \hat{U}_2 e^{-it\hat{H}} \hat{U}_1 = e^{-it\hat{G}}. \] (21)
The operators $\hat{U}_i$ can be constructed by suitably concatenating physically attainable processes. This allows us to associate with any unitary operator $e^{-it\hat{G}}$ a conditional probability $K(X_2|X_1, t)$ satisfying Property 1. Note that $t$ in $K$ does not correspond in general to the evolution time of a purely time-homogeneous process. It is the evolution time of only a part of the overall process involving also the transformations $\hat{U}_i$. This implies that for $t = 0$ the conditional probability $K(X_2|X_1, t)$ is not necessarily a delta distribution. Furthermore, we can add a shift time $t_0$ to $t$ and absorb the extra-term $e^{-it_0\hat{H}}$ in Eq. (21) into the operators $\hat{U}_1$ and $\hat{U}_2$. Thus, it is neither necessary to require that $t$ is a positive quantity, as in the case of stochastic Markov processes where the propagation kernels are elements of a semi-group. For our purpose, it sufficient that $K(X_2|X_1, t)$ is defined in a neighborhood of $t = 0$.

Let $\{G_i\}$ with $i = 1, ..., D$ be a set of $D \equiv N^2 - 1$ generators of the Lie algebra. The unitary operators $e^{-itG_i}$ are associated with the conditional probabilities $K_i(X_2|X_1, t)$.

Any unitary evolution $\hat{U}$ in a region around the identity can be constructed in the following way,

$$\hat{U}(t_1, ..., t_D) = e^{-it_1G_1}...e^{-it_DG_D} = \prod_{i=1}^{D} e^{-it_iG_i},$$

(22)

where the variables $t_i$ parametrize the $SU(N)$ manifold. Let the linear operator

$$\rho(X) \rightarrow \int dY K_i(X|Y, t)\rho(Y)$$

be denoted by $K_i(t)$. The overall evolution $\hat{U}(t)$ is associated with the conditional probability

$$K(X_2|X_1, t_1, ..., t_D) = \left[ \prod_{i=1}^{D} K_i(t_i) \right] (X_2|X_1),$$

(23)

the product order is such that the sum index grows from left to right.

It is easy to prove a general property of the conditional probability $K(X_2|X_1, t)$.

**Property 2**: Suppose that $K(\tilde{x}_2|\tilde{x}_1, t)$ is different from zero for some process $X_1 \rightarrow X_2$, then there exists a $M \times D$ matrix $\tilde{\eta}$ such that, for any small variation of the time parameters

$$\delta t = \begin{pmatrix} \delta t_1 \\ \delta t_2 \\ \vdots \\ \delta t_D \end{pmatrix},$$

the conditional probability $K(\tilde{x}_2 + \tilde{\eta}\delta t|\tilde{x}_1, t + \delta t)$ is different from zero. In the following, $\tilde{\eta}$ will be called the shift matrix of $K(X_2|X_1, t)$. As with the matrix $\tilde{\lambda}$ and the vector $\tilde{\alpha}$, it is a function of $t$ and the process $X_1 \rightarrow X_2$.

Proof: first, we consider the concatenation $K_{12}$ of two conditional probabilities, that is,

$$K_{12}(X_2|X_1, t_1, t_2) \equiv \int dZ K_1(X_2|Z, t_1)K_2(Z|X_1, t_2).$$

(24)

Let $X_1 \rightarrow X_2$ be a process with non-zero transition probability $K_{12}(X_2|X_1, t_1, t_2)$. There exists a state $Z$ such that $K_1(X_2|Z, t_1)$ and $K_2(Z|X_1, t_2)$ are different from zero. Property 1 implies that there exist two vectors $\tilde{\alpha}_1$ and $\tilde{\alpha}_2$ and a matrix $\tilde{\lambda}_1$ such that

$$K_1(\tilde{y} + \tilde{\lambda}_1\tilde{\alpha}_2\delta t_2 + \tilde{\alpha}_1\delta t_1|\tilde{z} + \tilde{\alpha}_2\delta t_2, t_1 + \delta t_1) \neq 0,$$

$$K_2(\tilde{z} + \tilde{\alpha}_2\delta t_2|\tilde{z}, t_2 + \delta t_2) \neq 0.$$

Thus, the conditional probability $K_{12}(\tilde{x}_2 + \tilde{\eta}\delta t|\tilde{x}_1, t_1 + \delta t_1, t_2 + \delta t_2)$ is different from zero, where the columns of the $M \times 2$ matrix $\tilde{\eta}$ are $\tilde{\alpha}_1$ and $\tilde{\lambda}_1\tilde{\alpha}_2$. The property can be proved by induction for any concatenation of processes $K_i$. □

There is an important direct consequence of this property.

**Lemma 1**: A process $X_1 \rightarrow X_2$ with non-zero probability $K(X_2|X_1, \tilde{0})$ is associated with a $D_x$-dimensional manifold of unitary transformations $\hat{U}(\delta t)$, where $D_x \geq D - M$, $M$ being the ontological space dimension. The manifold is identified by the $M$ equations

$$\sum_{j=1}^{D} \eta_{i,j}\delta t_j = 0, \text{ with } i = 1, ..., M,$$

(25)

where $\tilde{\eta}$ is the shift matrix of $K(X_2|X_1, \tilde{0})$. In particular, $D_x = D - N_l, N_l(\leq M)$ being the number of independent equations in the constraints (25).

Proof: because of Property 2, there exists a $M \times D$ matrix $\tilde{\eta}$ such that $K(\tilde{x}_2 + \tilde{\eta}\delta t|\tilde{x}_1, \delta t)$ is different from zero, i.e. $\tilde{x}_1 \rightarrow \tilde{x}_2 + \tilde{\eta}\delta t$ is a process associated with the unitary evolutions $\hat{U}(\delta t)$. In particular the submanifold of unitary evolutions with $\tilde{\eta}\delta t = 0$ is associated with the process $X_1 \rightarrow X_2$. Its dimension $D_x$ is equal to $D - N_l$, $N_l$ being the number of independent equations in $\tilde{\eta}\delta t = 0$. Since $N_l \leq M$, we have $D_x \geq D - M$.

The actual value of $D_x$ depends on the number $N_l$ of independent constraints in the vectorial equation $\tilde{\eta}\delta t = 0$. For example, $D_x$ is equal to $N^2 - 1$ for $\tilde{\eta} = 0$, which is the case if $K_i(\tilde{y}|\tilde{x}, t)$ are $M$-dimensional diffusive processes.

### B. Set S(X) and its symmetry property

In an ontological model the quantum state is associated with a probability distribution $\rho(X|\psi, \eta)$ according to the mapping (2). It is useful to introduce the following definition of the set $S(X)$.

**Definition 1**: a vector $|\psi\rangle$ of the Hilbert space is in
\( \mathcal{S}(X) \) if and only if there exists a context \( \eta \) such that the probability \( \rho(X|\psi, \eta) \) is different from zero for the state \( X \).

In other words, the set \( \mathcal{S}(X) \) contains every quantum state that is compatible with the occurrence of the ontic state \( X \). As discussed in Ref. [3], the set \( \mathcal{S} \) cannot lose vectors along its evolution. More precisely, if \( X \xrightarrow{U_2} Y \) is a non-zero probability process associated with the unitary evolution \( U \), then \( U\mathcal{S}(X) \subseteq \mathcal{S}(Y) \). This is a direct consequence of the definition. Indeed, if \( |\psi\rangle \) is in \( \mathcal{S}(X) \), then there exists a probability distribution associated with \( |\psi\rangle \) such that \( X \) is in its support. Since \( X \rightarrow Y \) is a non-zero probability process, then \( Y \) is in the support of a probability distribution associated with the evolved quantum state \( U|\psi\rangle \), that is, \( |\psi\rangle \in \mathcal{S}(X) \Rightarrow U|\psi\rangle \in \mathcal{S}(Y) \). The opposite implication \( |\psi\rangle \in \mathcal{S}(X) \Leftarrow U|\psi\rangle \in \mathcal{S}(Y) \) is not trivially satisfied. Thus, the set \( \mathcal{S} \) cannot lose vectors, but in principle it could grow acquiring vectors.

The opposite implication can be deduced by assuming that each process is invertible.

**Definition 2:** a non-zero probability process \( X_1 \xrightarrow{U_2} X_2 \) is said to be invertible if there exists a unitary operator \( U_2 \) such that \( X_2 \xrightarrow{U_2} X_1 \) is a non-zero probability process.

**Property 3:** every process is invertible.

Note that the operator \( U_2 \) associated with the inverse process is not required to be necessarily the inverse of \( U_1 \). Property 3 is very reasonable and is satisfied by any known hidden variable theory. Indeed a state connected to other states by means of a non-invertible process would be only transient and could be safely eliminated by the theory. The fact that the transient states are insignificant is made clearer if it is assumed that the ontological space is compact.

**Proposition:** if the ontological space is compact, then the processes in any series \( X_1 \xrightarrow{U_1} X_2 \xrightarrow{U_2} X_3 \xrightarrow{U_3} \ldots \) become closer and closer to being invertible.

Proof: a metric ontological space is compact if it is closed and bounded. Let \( B_R(X) \) be \( M \)-dimensional balls with radius \( R \) and center \( X \). Suppose that for any \( R \) there exists a series \( X_1 \xrightarrow{U_1} X_2 \xrightarrow{U_2} X_3 \xrightarrow{U_3} \ldots \) with an infinite number of processes that take a state \( X \) away from its ball \( B_R(X) \). Then there exists a subseries \( Y_1 \rightarrow Y_2 \rightarrow Y_3 \rightarrow \ldots \) where \( Y_m \notin B_R(Y_n) \) for every \( m > n \), but this is impossible because the space is bounded. Along the series, the processes connecting an element to the following ones become in fact closer and closer to being invertible. \( \square \)

Discarding insignificant transient states and taking for granted Property 3, we can prove the second lemma.

**Lemma 2:** Assuming Property 3, if \( X_1 \xrightarrow{U_1} X_2 \) is a non-zero probability process, then \( U_1S(X_1) = S(X_2) \).

Proof: the process \( X_1 \xrightarrow{U_1} X_2 \) is allowed, thus
\[ U_1S(X_1) \subseteq S(X_2). \] (26)

Since there exists an operator \( \hat{U}_2 \) such that \( X_2 \xrightarrow{\hat{U}_2} X_1 \) is a non-zero probability process, then we have also that
\[ \hat{U}_2S(X_2) \subseteq S(X_1). \] (27)

These two relations imply that
\[ \hat{U}_2S(X_2) \subseteq S(X_1), \] (28)

where \( \hat{U} = U_1\hat{U}_2 \). By iteration we obtain that
\[ \hat{U}^nS(X_2) \subseteq S(X_2), \] (29)

for any integer \( n \). For a finite dimensional Hilbert space, it is always possible to find an integer \( n \) such that \( \hat{U}^n \) is very close to the inverse operator \( \hat{U}^{-1} \). Thus, \( \hat{U}^{-1}S(X_2) \subseteq S(X_2) \), that is,
\[ \hat{U}S(X_2) \supseteq S(X_2). \] (30)

From inclusions (28,30) we have that
\[ \hat{U}S(X_2) = S(X_2). \] (31)

Applying the operator \( \hat{U}_1 \) to both sides of inclusion (27) and using Eq (31), the inclusion
\[ S(X_2) \subseteq \hat{U}_1S(X_1) \] (32)

is obtained. This relation and inclusion (26) imply that
\[ \hat{U}_1S(X_1) = S(X_2) \] (33)

and the lemma is proved. \( \square \)

Note that, for symmetry reasons, a similar equation holds also for the inverse process, that is,
\[ \hat{U}_2S(X_2) = S(X_1). \] (34)

In Ref. [3] we proved the following property for \( \mathcal{S}(X) \).

**Lemma 3:** the set \( \mathcal{S}(X) \) cannot contain every vector of the Hilbert space. Equivalently, the set \( \mathcal{S}(X) \) is not invariant with respect to the group \( SU(N) \).

Proof by contradiction: suppose that \( \mathcal{S}(X) \) contains every vector of the Hilbert space, then it contains in particular also two orthogonal vectors. This means that there exist two overlapping distributions associated with two orthogonal quantum states. But this is impossible because two orthogonal states can be perfectly discriminated by a measurement [3, 11]. Indeed the probability of obtaining \( |\psi\rangle \) given \( \langle\psi|\psi\rangle \) is
\[ \int dX P(\psi|X, \tau)\rho(X|\psi, \eta) = |\langle\psi|\psi\rangle|^2 = 1. \] (35)

This implies that \( P(\psi|X, \tau) \) is equal to 1 in the support of \( \rho(X|\psi, \eta) \). However the probability of \( |\psi\rangle \) given an orthogonal state \( |\psi_\perp\rangle \) is
\[ \int dX P(\psi|X, \tau)\rho(X|\psi_\perp, \bar{\eta}) = |\langle\psi|\psi_\perp\rangle|^2 = 0. \] (36)
This implies that \( \rho(X|\psi,\eta) \) cannot be different from zero if \( X \) is in the support of \( \rho(X|\psi,\eta) \), where \( P(\psi|X,\tau) = 1 \). □

Finally, we enunciate the last lemma.

**Lemma 4:** Let \( G \) be a Lie subgroup of the group \( SU(N) \) acting on a \( N \)-dimensional Hilbert space \( \mathcal{H} \) and \( S \) be a set of vectors in \( \mathcal{H} \). If the manifold dimension of \( G \) is larger than \((N - 1)^2\) and \( S \) is invariant with respect to \( G \), then \( S \) contains every vector of the Hilbert space, that is, \( S \) is invariant with respect to \( SU(N) \).

Proof: any compact Lie group and their linear representations on \( \mathbb{C}^N \) are well-known. One can check that, for \( N \neq 4 \), the proper Lie subgroup of \( SU(N) \) with largest manifold dimension is \( SU(N - 1) \times U(1) \). Its dimension is equal to \((N - 1)^2\). Thus, if the dimension of \( G \) is larger than \((N - 1)^2\), then \( G \) is in fact \( SU(N) \) and \( S \) contains every vector. In the special case \( N = 4 \), the symplectic group \( Sp(2) \) is the subgroup of \( SU(4) \) with largest dimension. A set of generators in the representation space \( \mathbb{C}^4 \) is

\[
\hat{s}_i^{(1)}, \hat{s}_i^{(2)}, \hat{s}_i^{(3)}, \hat{s}_i^{(3)}.
\]

with \( i = 1, 2, 3 \), \( \hat{s}_i^{(k)} \) being two sets of Pauli matrices acting on the tensor space \( \mathbb{C}^2 \otimes \mathbb{C}^2 = \mathbb{C}^4 \). It is easy to check that these generators form the basis of a Lie algebra with dimension equal to 10, which is larger than \((N - 1)^2 = 9\). However also in this case if the set \( S \) is invariant with respect to \( Sp(2) \), then it contains every vector of the Hilbert space. Indeed, it is possible to show that the only orbit of \( Sp(2) \) is the whole Hilbert space, the orbit of a subgroup being the set of states connected by means of some subgroup element. It is sufficient to prove that any vector is connected to \( \uparrow \uparrow \uparrow \) and \( \downarrow \downarrow \downarrow \)

A generic vector \( |\psi\rangle \) has the form

\[
|\psi\rangle = \cos \theta |\uparrow\rangle |\phi_1 > + \sin \theta |\downarrow\rangle |\phi_2 > ,
\]

where \( |\phi_1\rangle \) and \( |\phi_2\rangle \) are two-dimensional vectors. This vector is connected through the generators \( \hat{s}_i^{(1)} \) to a vector with the form

\[
|\tilde{\psi}\rangle = \cos \theta |\uparrow\rangle |\uparrow\rangle + \sin \theta |\downarrow\rangle |\tilde{\phi}_2 > .
\]

Through the unitary operator \( e^{i \frac{\beta}{2} \sigma_3^{(1)} (\sigma_1^{(2)} + \sigma_2^{(2)})} \) and a suitable choice of \( \beta \), it is possible to connect \( |\tilde{\psi}\rangle \) to

\[
|\psi\rangle = \cos \theta |\uparrow\rangle |\uparrow\rangle + e^{i \beta} \sin \theta |\downarrow\rangle |\downarrow\rangle .
\]

This last vector is connect to \( |\uparrow\rangle |\uparrow\rangle \) through the generators \( \hat{s}_i^{(1)} \) and \( \hat{s}_i^{(2)} \), \( \hat{s}_i^{(1)} \) being the raising/lowering operators. □

**C. No-shrinking theorem**

At this point we have sufficient tools to prove the no-shrinking theorem. We will show that if the ontological space dimension would be smaller than the quantum state manifold dimension, then the set \( S(X) \) would be invariant with respect to the group \( SU(N) \), in contradiction with lemma 3.

**Theorem:** If the dynamics in a Markov ontological theory satisfies Property 1 and all the processes are invertible, then the ontological space dimension \( M \) is not smaller than the quantum state manifold dimension, that is, \( M \geq 2N - 2 \), \( N \) being the Hilbert space dimension.

Proof by contradiction: suppose that \( M < 2N - 2 \). Let us consider the unitary evolution \( \hat{U}(\vec{t}) \) defined in Eq. (22) and the associated conditional probability \( K \) given by Eq. (23). We have from lemma 1 that a process \( X \to Y \) with non-zero probability \( K(\vec{X}|\vec{X},\vec{0}) \) is associated with a \( D_s \)-dimensional manifold of unitary evolutions, say \( U \), with

\[
D_s \geq N^2 - 1 - M > N^2 - 1 - 2N + 2 = (N - 1)^2 .
\]

The manifold contains the identity. As a consequence, the set \( S(X) \) evolves, for any unitary evolution in \( U \), to the same \( S(Y) \). By means of lemma 2 we have that

\[
\hat{U}S(X) = S(Y), \quad \forall \hat{U} \in U.
\]

Since \( 1 \in U \),

\[
S(X) = S(Y).
\]

Thus,

\[
\hat{U}S(X) = S(X), \quad \forall \hat{U} \in U,
\]

that is, \( S(X) \) is symmetric with respect to a group whose generators are algebraically generated by \( D_s \) generators. The manifold dimension of this group is equal to or larger than \( D_s > (N - 1)^2 \). Because of lemma 4, \( S(X) \) is invariant with respect to \( SU(N) \), but this is in contradiction with lemma 3. We conclude that the ontological space dimension \( M \) cannot be smaller than the quantum state manifold dimension \( 2N - 2 \). □

**D. Minimal dimension and Schrödinger equation**

It is possible to show that in the case of minimal ontological dimension the set of continuous variables can be identified with a vector living in the Hilbert space, whose dynamics is described by the Schrödinger equation. It does not necessarily coincide with the quantum state, but moves rigidly with respect to it.

Suppose that the ontological dimension is minimal, that is, \( M = 2(N - 1) \). From lemmas 1-4 we have that the set \( S(X) \) has to be symmetric with respect to a subgroup of \( SU(N) \) with manifold dimension equal to
$N^2 - 1 - M = (N-1)^2$. The subgroup is $SU(N-1) \times U(1)$. Thus, the set has a symmetry axis $|\phi\rangle$, which is a vector in the Hilbert space and identifies the orientation of $S$. Since each ontic state $X$ is associated with a set $S(X)$, we have the mapping

$$X \to |\phi\rangle. \quad (45)$$

Because of Lemma 2, in a process associated with the unitary evolution $\hat{U}(t)$, the set $S$ rotates rigidly to $\hat{U}(t)S$. In particular, its symmetry axis satisfies the Schrödinger equation

$$i \frac{\partial |\phi\rangle}{\partial t} = \hat{H}(t)|\phi\rangle, \quad (46)$$

where $\hat{H}(t)$ is the Hamiltonian that generates $\hat{U}(t)$.

The mapping (45) is surjective, but is not necessarily injective. However, since the number of continuous ontological variables is just sufficient to label $|\phi\rangle$ up to a global phase, an additional discrete index is sufficient to make the mapping bijective,

$$X \leftrightarrow (|\phi\rangle, n). \quad (47)$$

Thus, the ontic state in a minimal Markov theory is identified with a vector in the Hilbert space and a possible additional discrete index. The dynamics of the vector is given by the Schrödinger equation. $|\phi\rangle$ is equal to the quantum state if the set $S(X)$ contains only one state that coincides with the symmetry axis. This is the case in a wave-pilot theory such as the de Broglie-Bohm mechanics.

It is interesting to note that requirements (46, 47) are fulfilled by the Kochen-Specker model [13], where the ontic state is identified by a Bloch vector having the same dynamical equation of the quantum state.

### E. Resource cost with round off error

We have proved that in any hidden variable theory with short memory dynamics the dimension of the state space cannot be smaller than the quantum state manifold dimension $2N - 2$. There is a relation between dimension and resource cost required to identify an ontic state. By resource cost we mean the quantity of information required to identify an ontological state. Obviously the information carried by an ontic state is infinite, since the ontological space is continuous. Thus, the resource cost as a function of the dimension makes sense only in presence of a fixed round off error of the continuous variables.

One can find a scaling law between dimension and resource cost in the following way. Suppose that the ontic state is identified by $M$ continuous variables $x_i$, which are defined in a finite interval. With a suitable rescaling, we can assume that they run in the interval between 0 and 1. The probability $P(E|x)$ of an event $E$ is conditioned by the ontic state $x$. It is assumed that $P(E|x)$ is a smooth function of $x$, but the discussion could be extended to the case of a finite number of discontinuities. Let $g_i$ be the mean magnitude of $|\partial_{x_i} P(E|x)|$.

Let each continuous variable be discretized by $n_i$ points. This introduces a round off error $\Delta E$ in $P$ that scales as $\sqrt{\sum_{i=1}^{M} (g_i/n_i)^2}$ for sufficiently large values of $n_i$. The number of bits required to identify the ontic state on the lattice is proportional to the information

$$I = \sum_{i=1}^{M} \log n_i. \quad (48)$$

For a fixed $I$, using the Lagrange multiplier method we find that the optimal choice of $n_i$ that gives the smallest error is

$$n_i = g_i e^{\bar{x}} / \bar{g}, \quad (49)$$

where $\bar{g} \equiv (\prod_i g_i)^{1/M}$ is the log-average of $g_i$. Thus, we have the scaling law

$$\Delta E \sim \bar{g} M^{\frac{1}{2}} e^{-\frac{\bar{x}}{\Delta E}}. \quad (50)$$

The error exponentially decreases with $I$ at a rate inversely proportional to $M$. For a fixed error, the information scales as

$$I \sim M \log \frac{M^{\frac{1}{2}} \bar{g}}{\Delta E}. \quad (51)$$

The no-shrinking theorem states that $M$ is not smaller than the quantum state manifold dimension. Thus, in a composite system, $I$ grows at least exponentially in the number of parts (for example the number of qubits in a quantum computer).

It is possible to estimate a lower bound for $\bar{g}$. Since the probability of an event goes from 0 to 1 and $x_i$ rambles about the interval $[0:1]$, it is reasonable to assume that $g_i$ cannot be smaller than a value around 1, that is,

$$\bar{g} \gtrsim 1.$$

Thus the linear growth of $I$ with respect $M$ cannot be mitigated by an exponential decrease of $\bar{g}$.

### V. CONCLUSIONS

We have shown that in any hidden variable theory with a short memory dynamics the ontological space dimension cannot be smaller than the quantum state manifold dimension. Thus, like the quantum state, the ontic state necessarily carries for a given accuracy an amount of information that grows exponentially in the number of subsystems. In comparison with Ref. [2], we have provided a better justification of the relaxation hypothesis, showing that in a compact ontological space it is possible to find a trajectory along which the system unavoidably
converges towards a region where all the processes are invertible. Furthermore we have presented a hidden variable model of measurement for a qubit whose ontological space is one-dimensional, which is one-half the dimension of the Bloch sphere. The corresponding dynamics is not Markovian, in accordance with the no-shrinking theorem.

This model provides a counterexample making evident that the hypotheses of short memory is strictly necessary to prove the no-shrinking theorem. By dropping it, we have shown that a single realization can carry less information than the quantum state. More drastically, we could drop the causality hypothesis. Indeed the non-causality is implied also by the Bell theorem and the Lorentz invariance of the hidden variable theory. Thus, there are two signs that point to the same direction, that is, the rejection of causality at the ontological level. The possibility of an ontological shrinking for a general N-dimensional Hilbert space in a theory without a Markov dynamics is an open question whose answer could provide a deeper understanding of the computational complexity in quantum mechanics.

I wish to thank J. Wallman and R. W. Spekkens for the careful reading of the manuscript and useful suggestions. Research at Perimeter Institute for Theoretical Physics is supported in part by the Government of Canada through NSERC and by the Province of Ontario through MRI.

Appendix A: Deriving the economical model for a qubit

In this appendix we will present a systematic method to generate the model reported in Sec. III. We will consider the class of probability distributions with the form

\[ \rho(x, n|\vec{v}) = r(n|\vec{v})\delta(x - f_n(\vec{v})) , \]  

(A1)

where the pair \((x \in \mathbb{R}, n \in \{0, 1\})\) is the ontic state and \(\vec{v} \equiv \langle \psi, \delta|\psi \rangle\) is the Bloch vector corresponding to the quantum state \(|\psi\rangle\). The quantities \(f_n(\vec{v})\) are real functions. The normalization of the distribution gives

\[ r(0|\vec{v}) + r(1|\vec{v}) = 1. \]  

(A2)

When a system is prepared in \(\vec{v}\), at the ontological level there is a probability \(r(0|\vec{v})\) \([r(1|\vec{v})]\) that the discrete index takes the value \(n = 0\) \((n = 1)\). Correspondingly, the continuous variable takes with certainty the value \(f_0(\vec{v})\) \([f_1(\vec{v})]\). This class is the simplest one that fulfills some requirements. As noted in Subsec. IV.B, two orthogonal states cannot have overlapping probability distributions, implying that the support of \(\rho(x, n|\vec{v})\) cannot be the whole ontological space. This rules out the class of smooth analytical distributions and leads us to consider the probability distributions with zero-measure support as simplest case. In particular, the distributions with a two-point support are parametrized with 3 real parameters \([\text{the positions } f_n(\vec{v})\text{ of the points and the relative probability weight } r(0|\vec{v}) - r(1|\vec{v})]\), which are sufficient in order to cover the two-dimensional Bloch sphere.

For our purposes, it is convenient to define the variables \(x_n \equiv f_n(\vec{v})\) and to use them to parametrize the quantum state. Thus, the probability distribution becomes

\[ \rho(x, n|x_0, x_1) = r(n|x_0, x_1)\delta(x - x_n). \]  

(A3)

Let \(|\phi\rangle\) be the event of some projective measurement. As for the preparation state, we introduce the Bloch vector \(\vec{w} \equiv \langle \phi, \hat{d}|\phi \rangle\) to label the event \(|\phi\rangle\). \(P_n(\vec{w}|x)\) is the conditional probability of the event \(\vec{w}\) given the ontic state \((x, n)\). Equation (4) becomes

\[ \sum_{n=0}^{1} \int dx P_n(\vec{w}|x) \rho(x, n|x) = \frac{1 + \vec{w} \cdot \vec{v}(x)}{2} = S, \]  

(A4)

where \(\vec{v}(x)\) gives the Bloch vector \(\vec{v}\) as a function of the parameters \((x_0, x_1) \equiv x\). The quantity \(1/2(1 + \vec{w} \cdot \vec{v})\) is the Born probability \(|\langle \phi|\psi \rangle|^2\) in terms of the Bloch vectors.

Using Eq. (A3), Equation (A4) becomes after integration

\[ P_0(\vec{w}|x_0) r(0|x_0, x_1) + P_1(\vec{w}|x_1) r(1|x_0, x_1) = S. \]  

(A5)

Note that the conditional probability \(P_n(\vec{w}|x_n)\) in this equation depends only on the variable \(x_n\). Thus, in order to find it, we could consider the four pairs \((x_0, x_1), (x_0, y_1), (y_0, x_1)\) and \((y_0, y_1)\). Correspondingly we have four equations with the four unknown functions \(P_0(\vec{w}|x_0), P_0(\vec{w}|y_0), P_1(\vec{w}|x_1)\) and \(P_1(\vec{w}|y_1)\). These equations are the row elements of the vector equation

\[ \hat{R}\vec{P} = \vec{S}, \]  

(A6)

where

\[ \hat{R} \equiv \begin{pmatrix} r(0|x_0, x_1) & 0 & r(1|x_0, x_1) & 0 \\ r(0|x_0, y_1) & 0 & 0 & r(1|x_0, y_1) \\ r(0|y_0, x_1) & r(1|y_0, x_1) & 0 & 0 \\ r(0|y_0, y_1) & 0 & 0 & r(1|y_0, y_1) \end{pmatrix}, \]  

(A7)

and

\[ \vec{P} \equiv \begin{pmatrix} P_0(\vec{w}|x_0) \\ P_0(\vec{w}|y_0) \\ P_1(\vec{w}|x_1) \\ P_1(\vec{w}|y_1) \end{pmatrix}, \]  

(A8)

If \(\hat{R}\) would be invertible, \(P_n(\vec{w}|x_n)\) would be a linear function of \(\vec{w}\), but this is impossible. It can be proved by contradiction. Suppose that the conditional probabilities are linear functions of \(\vec{w}\), that is, \(P_n(\vec{w}|x_n) = C_n(x_n) + d_n(x_n) \cdot \vec{w}\). They have to be non-negative and smaller than or equal to 1. It is clear that \(P_n(\vec{w}|x_n)\) are equal to 1
at most for only one vector $\vec{w}$, but this is in contradiction with the fact that the conditional probabilities have to be equal to 1 in the one-dimensional manifold of vectors $\vec{w}$ such that $f_n(\vec{w}) = x_n$. Thus, the equation
\[ \det \mathbf{R}(x_0, x_1, y_0, y_1) = 0 \] (A10)
has to be satisfied for any $x_n$ and $y_n$. Its solution is found by differentiating in $y_0$ and $y_1$,
\[ \frac{\partial^2}{\partial y_0 \partial y_1} \det \mathbf{R}(x_0, x_1, y_0, y_1) \bigg|_{y_0 = x_0, y_1 = x_1} = 0, \] (A11)
that is,
\[ r(0|x_0, x_1)^2 r(1|x_0, x_1)^2 \partial x_0 \partial x_1 \log \frac{r(1|x_0, x_1)}{r(0|x_0, x_1)} = 0. \] (A12)
The functions $r(n|x)$ cannot be identically equal to zero, thus the only acceptable solution that satisfies also the constraint (A2) is
\[ r(0|x_0, x_1) = \frac{k_1(x_1)}{k_0(x_0) + k_1(x_1)} \] (A13)
\[ r(1|x_0, x_1) = \frac{k_0(x_0)}{k_0(x_0) + k_1(x_1)}, \] (A14)
where $k_i(x_i)$ are positive functions of the variable $x_i$. It is easy to check that this is also a solution of Eq. (A11). Thus, the fact that the conditional probabilities $P_n(\vec{w}|x_n)$ cannot be linear in $\vec{w}$ allows us to find a constraint for the probabilities $r(n|x)$.

Since the determinant of $\mathbf{R}$ is equal to zero, there is a constraint also for the $\mathbf{S}$ because of Eq. (A6). Let the row vector $\vec{u}^T$ be the left eigenvector of $\mathbf{R}$ with eigenvalue 0, then we have from Eq. (A6) that
\[ \vec{u}^T \cdot \mathbf{S} = 0. \] (A15)
By means of Eqs. (A13A14), after a bit of calculations we find that the left eigenvector of $\mathbf{R}$ with zero eigenvalue is
\[ \vec{u} = \left( \begin{array}{c} k_0(x_0) \ \k_0(x_0)^{-1} + k_1(x_1)^{-1} \\ -k_0(x_0)^{-1} - k_1(x_1)^{-1} \\ -k_0(y_0)^{-1} - k_1(y_1)^{-1} \\ k_0(y_0)^{-1} + k_1(y_1)^{-1} \end{array} \right). \] (A16)
From Eqs. (A9A15A16) we obtain by differentiation the condition
\[ \frac{\partial y_0 y_1}{\partial x_0} \left( \frac{\vec{u}^T \cdot \mathbf{S}}{y_0 = x_0} \right) \bigg|_{y_0 = x_0, y_1 = x_1} = 0 \] (A17)
that is satisfied if
\[ \vec{v}(x_0, x_1) = \left( \begin{array}{c} \frac{1}{k_0(x_0)} + \frac{1}{k_1(x_1)} \\ \frac{1}{k_0(x_0)} + \frac{1}{k_1(x_1)} \\ \frac{1}{k_0(x_0)} + \frac{1}{k_1(x_1)} \\ \frac{1}{k_0(x_0)} + \frac{1}{k_1(x_1)} \end{array} \right)^{-1} [\vec{g}_0(x_0) + \vec{g}_1(x_1)], \] (A18)
where $\vec{g}_0(x_0)$ and $\vec{g}_1(x_1)$ are generic vectorial functions. This is the inverse of the equation $\vec{x} = \vec{f}(\vec{v})$. The functions $g_n(x_n)$ and $k_n(x_n)$ are constrained by the equation $\vec{v}^2 = 1$, that is,
\[ [\vec{g}_0(x_0) + \vec{g}_1(x_1)]^2 = \left( \frac{1}{k_0(x_0)} + \frac{1}{k_1(x_1)} \right)^2. \] (A19)
We will return to it later on.

From Eqs. (A14A18) and the identity $S = [1 + \vec{u} \cdot \vec{v}(\vec{x})]/2$ we find that
\[ \sum_{n=0}^{1} \left\{ P_n(\vec{w}|x_n) - \frac{1}{2} k_n^{-1}(x_n) - \frac{1}{2} \vec{u} \cdot \vec{g}_n \right\} = 0. \] (A20)
Note that each term of the summation depends only on one of the variables $x_0$ and $x_1$. Thus, the conditional probabilities have the form
\[ P_0(\vec{w}|x_0) = k_0(x_0) \left[ \frac{1}{2} \vec{u} \cdot \vec{g}_0(x_0) + H(\vec{w}) \right] + \frac{1}{2}; \] (A21)
\[ P_1(\vec{w}|x_1) = k_1(x_1) \left[ \frac{1}{2} \vec{u} \cdot \vec{g}_1(x_1) - H(\vec{w}) \right] + \frac{1}{2}, \] (A22)
where $H(\vec{w})$ is an additional function independent of $\vec{x}$. Since the probability $P_0(\vec{w}|x_0)$ has to be equal to 1 for $\vec{w} = \vec{v}$, we find that
\[ H(\vec{v}) = \frac{1}{2} \left[ \frac{1}{k_0(x_0)} - \vec{v} \cdot \vec{g}_0(x_0) \right]. \] (A23)
where $x_0 = f_0(\vec{v})$. Similarly, from the condition $P_0(\vec{v}|x_0) = 1$ we obtain the equation
\[ H(\vec{v}) = -\frac{1}{2} \left[ \frac{1}{k_1(x_1)} - \vec{v} \cdot \vec{g}_1(x_1) \right]. \] (A24)
This last equation can be derived by Eqs. (A18A23) and the constraint $\vec{v}^2 = 1$. It is interesting to note that the model we are constructing works only for a subset of preparation states $\vec{v}$, as shown in Sec. III, thus the function $H(\vec{w})$ is not necessarily given by Eqs. (A23A24) if $\vec{w}$ is outside that subset.

At this point we have almost everything, the last step is to find the functions $k_n(x_n)$ and $\vec{g}_n(x_n)$ that solve Eq. (A19). Differentiating this equation with respect to $x_0$ and $x_1$, we have that
\[ \frac{\partial \vec{g}_0(x_0)}{\partial x_0} \cdot \frac{\partial \vec{g}_1(x_1)}{\partial x_1} + \left( \frac{\partial}{\partial x_0} \frac{1}{k_0(x_0)} \right) \left( \frac{\partial}{\partial x_1} \frac{1}{k_1(x_1)} \right) = 0, \] (A25)
that is, the Minkowski inner product between the two four-vectors
\[ \alpha(x_0) \equiv \left( \vec{g}_0(x_0) + \frac{\vec{v}_0}{k_0(x_0)} + \chi_0 \right) \cdot \beta(x_1) \equiv \left( \vec{g}_1(x_1) + \frac{\vec{v}_1}{k_1(x_1)} + \chi_1 \right). \] (A26)
Thus they have to be equal to a constant \( r \), where the right-hand sides depend only on \( x \).\( \vec{g} \) could set \( \alpha \) possibilities: (a) one of the two four-vectors \( \vec{g}_0, \vec{g}_1 \) spans a one-dimensional vectorial subspace and the other one lives in the orthogonal subspace; (b) the two vectors span orthogonal two-dimensional spaces. In the former case no solution exists such that \( \vec{g}(x_0, x_1) = 0 \), \( 0 \leq x_0 \leq 2\pi \), \( 0 < x_1 < \pi \). The most general solution up to rotations, transformation \( \vec{g}_0 \to \vec{g}_0 + \vec{t}, \vec{g}_1 \to \vec{g}_1 - \vec{t}, \) for a generic vector \( \vec{t} \). This means that the constant vectors \( \vec{g}_0, \vec{g}_1 \) are redundant and for example one could set \( \vec{g}_1 = 0 \).

In order to solve Eq. (A27), we have to consider two possibilities: (a) one of the two four-vectors \( \alpha, \beta \) spans a one-dimensional vectorial subspace and the other one lives in the orthogonal subspace; (b) the two vectors span orthogonal two-dimensional spaces. In the former case no solution exists such that \( \vec{g}(x_0, x_1) = 0 \), \( 0 \leq x_0 \leq 2\pi \), \( 0 < x_1 < \pi \). The most general solution up to rotations, transformation \( \vec{g}_0 \to \vec{g}_0 + \vec{t}, \vec{g}_1 \to \vec{g}_1 - \vec{t}, \) for a generic vector \( \vec{t} \). This means that the constant vectors \( \vec{g}_0, \vec{g}_1 \) are redundant and for example one could set \( \vec{g}_1 = 0 \).

In order to solve Eq. (A27), we have to consider two possibilities: (a) one of the two four-vectors \( \alpha, \beta \) spans a one-dimensional vectorial subspace and the other one lives in the orthogonal subspace; (b) the two vectors span orthogonal two-dimensional spaces. In the former case no solution exists such that \( \vec{g}(x_0, x_1) = 0 \), \( 0 \leq x_0 \leq 2\pi \), \( 0 < x_1 < \pi \). The most general solution up to rotations, transformation \( \vec{g}_0 \to \vec{g}_0 + \vec{t}, \vec{g}_1 \to \vec{g}_1 - \vec{t}, \) for a generic vector \( \vec{t} \). This means that the constant vectors \( \vec{g}_0, \vec{g}_1 \) are redundant and for example one could set \( \vec{g}_1 = 0 \).

In order to solve Eq. (A27), we have to consider two possibilities: (a) one of the two four-vectors \( \alpha, \beta \) spans a one-dimensional vectorial subspace and the other one lives in the orthogonal subspace; (b) the two vectors span orthogonal two-dimensional spaces. In the former case no solution exists such that \( \vec{g}(x_0, x_1) = 0 \), \( 0 \leq x_0 \leq 2\pi \), \( 0 < x_1 < \pi \). The most general solution up to rotations, transformation \( \vec{g}_0 \to \vec{g}_0 + \vec{t}, \vec{g}_1 \to \vec{g}_1 - \vec{t}, \) for a generic vector \( \vec{t} \). This means that the constant vectors \( \vec{g}_0, \vec{g}_1 \) are redundant and for example one could set \( \vec{g}_1 = 0 \).

In order to solve Eq. (A27), we have to consider two possibilities: (a) one of the two four-vectors \( \alpha, \beta \) spans a one-dimensional vectorial subspace and the other one lives in the orthogonal subspace; (b) the two vectors span orthogonal two-dimensional spaces. In the former case no solution exists such that \( \vec{g}(x_0, x_1) = 0 \), \( 0 \leq x_0 \leq 2\pi \), \( 0 < x_1 < \pi \). The most general solution up to rotations, transformation \( \vec{g}_0 \to \vec{g}_0 + \vec{t}, \vec{g}_1 \to \vec{g}_1 - \vec{t}, \) for a generic vector \( \vec{t} \). This means that the constant vectors \( \vec{g}_0, \vec{g}_1 \) are redundant and for example one could set \( \vec{g}_1 = 0 \).

In order to solve Eq. (A27), we have to consider two possibilities: (a) one of the two four-vectors \( \alpha, \beta \) spans a one-dimensional vectorial subspace and the other one lives in the orthogonal subspace; (b) the two vectors span orthogonal two-dimensional spaces. In the former case no solution exists such that \( \vec{g}(x_0, x_1) = 0 \), \( 0 \leq x_0 \leq 2\pi \), \( 0 < x_1 < \pi \). The most general solution up to rotations, transformation \( \vec{g}_0 \to \vec{g}_0 + \vec{t}, \vec{g}_1 \to \vec{g}_1 - \vec{t}, \) for a generic vector \( \vec{t} \). This means that the constant vectors \( \vec{g}_0, \vec{g}_1 \) are redundant and for example one could set \( \vec{g}_1 = 0 \).
\[ P_1(\vec{w}|x) = 1 + \frac{(w_x \cos \theta_0 - 1) + w_z \cos x \sin \theta_0 + \Delta \sin x}{2(1 - s \sin x)}. \] (A40)

The model in Sec. III is obtained for \( \theta_0 = \pi/2 \) and \( s = 1 \).

[1] W. K. Wootters, W. H. Zurek, Nature 299, 802 (1982).
[2] R. W. Spekkens, Phys. Rev. A 75, 032110 (2007).
[3] R. Jozsa, A. Miyake, Proc. R. Soc. A 464, 3089 (2008).
[4] M. Van den Nest, Quant. Inf. Comp. 10 0258 (2010); M. Van den Nest, arXiv:0911.1624
[5] L. Hardy, Stud. Hist. Phil. Sci. B 35, 267 (2004).
[6] A. Montina, Phys. Rev. Lett. 97, 180401 (2006); A. Montina, J. Phys.: Conf. Ser. 67, 012050 (2007).
[7] A. Montina, Phys. Rev. A 77, 022104 (2008).
[8] B. Dakić, M. Suvakov, T. Paterek, and C. Brukner, Phys. Rev. Lett. 101, 190402 (2008).
[9] E. F. Galvão, Phys. Rev. A 80, 022106 (2009).
[10] A. Montina, arXiv:1002.3139
[11] R. W. Spekkens, Phys. Rev. A 71, 052108 (2005).
[12] C. W. Gardiner, ”Handbook of Stochastic Methods” (Springer, Berlin, 2004).
[13] S. Kochen and E. P. Specker, J. Math. Mech. 17, 59 (1967).
[14] S. G. Krantz, ”Geometric Function Theory: Explorations in Complex Analysis” (Birkhäuser, Boston, 2006).