## Contents

1 Bayesian probability theory \& classical causality ..... 3
1.1 Probability space and probability distributions ..... 3
1.2 Random variables and expectation values ..... 6
1.3 A short disclaimer ..... 7
1.4 Classical causality ..... 7
2 Operational Probabilistic Theories ..... 11
2.1 Operational language \& compositional rules ..... 11
2.2 Outcomes \& probabilistic models ..... 14
2.3 Quotient theories ..... 15
2.4 Linear structure. ..... 16
2.5 Examples ..... 17
3 Spekkens' Toy Theory ..... 21
3.1 Ontic vs. epistemic - Spekkens' toy theory ..... 21
3.2 "Quantum" features in the toy theory ..... 25
3.3 The toy theory in ZX language ..... 28
3.4 Example: the Peres-Mermin proof of non-classicality ..... 29
4 Signatures of Non-Classicality ..... 31
4.1 Bell theorem \& quantum violations ..... 31
4.2 Non-signalling correlations and postquantum violations ..... 34
4.3 Generalised contextuality ..... 34
5 Resource Theories ..... 39
5.1 Mathematical framework for resource theories ..... 39
5.2 Example: Local Operations and Shared Randomness (LOSR) ..... 41

## Chapter 1

## Bayesian probability theory \& classical causality

On this account all the sciences would only be unconscious applications of the calculus of probabilities; to condemn this calculus would be to condemn science entirely.

- H. Poincaré [1]


### 1.1 Probability space and probability distributions

As a mathematical theory, probability theory requires us to define three elements: the sample space, the algebra of events, and a measure on this algebra.

The sample space is essentially a set $\Omega$ (for this school will always be considered finite), from which we will draw the events. Events are simply subsets $X \subset \Omega$, and will be associated with measurement outcomes of experiments. In particular, a single measurement outcome is a subset with a single element in the event space, $x \in \Omega$.

The algebra of events is a collection $\mathcal{X}$ of the subsets of $\Omega$ relevant for the theory. In particular, this collection needs to satisfy some properties:

1. $\emptyset \in \mathcal{X}$ and $\Omega \in \mathcal{X}$. This means that the set of "no events" and the set of "all events" are both relevant events and must be in the algebra;
2. Given $X$ and $X^{\prime}$ both in $\mathcal{X}$, the sets $X \cup X^{\prime}, X \cap X^{\prime}$, and $X / X^{\prime}$ are also in $\mathcal{X}$. This means that we can make logical combinations between events (respectively OR, AND, and NOT). As a consequence, $\bigcup_{i=1}^{N} X_{i} \in \mathcal{X}$, where $N$ is the total number of subsets in the algebra.

From now on, our events will be subsets $X \in \mathcal{X}$. This way, we are ensuring that there is a structure allowing us to make logical combinations of events without incurring any abnormalities.

Finally, a measure over this algebra is simply a map $p: \mathcal{X} \rightarrow \mathbb{R}$ such that $X \mapsto p(X) \in$ $\mathbb{R}$. A probability distribution is a measure over an algebra of events satisfying the following properties:

1. $p(X) \geq 0, \forall X \in \mathcal{X}$;
2. $p(\Omega)=1$, which is called normalisation;
3. For any collection of disjoint events $X_{1}, \ldots, X_{N}$ with $X_{i} \cap X_{j}=\emptyset, \forall i \neq j=1, \ldots, N$, then

$$
\begin{equation*}
p\left(\cup_{i=1}^{N} X_{i}\right)=\sum_{i=1}^{N} p\left(X_{i}\right) \tag{1.1}
\end{equation*}
$$

These properties, called Kolmogorov axioms, ensure many of the usual features of probabilities we will be using along the subject. For instance, you can derive from these axioms and basic set theory the following equation

$$
\begin{equation*}
p\left(X \cup X^{\prime}\right)=p(X)+p\left(X^{\prime}\right)-p\left(X \cap X^{\prime}\right), \quad \forall X, X^{\prime} \in \mathcal{X} \tag{1.2}
\end{equation*}
$$

Another relevant concept for this course is the one of conditional probability. It is meant to capture the likelihood with which an event will occur given that another event has occurred. It consists of a map $p(\cdot \mid X): \mathcal{X} \rightarrow \mathbb{R}$, defined for all $X \in \mathcal{X}$ such that $p(X)>0$, and such that

$$
\begin{equation*}
X^{\prime} \mapsto p\left(X^{\prime} \mid X\right):=\frac{p\left(X^{\prime} \cap X\right)}{p(X)} \tag{1.3}
\end{equation*}
$$

The demand that $p(X)$ is nonzero is natural since we want to condition the event $X^{\prime}$ to something that has occurred. Conditional probabilities are important because it is from them that we construct the notion of independence: two events $X, X^{\prime}$ are said to be independent or uncorrelated when

$$
\begin{equation*}
p\left(X^{\prime} \mid X\right)=p\left(X^{\prime}\right) \tag{1.4}
\end{equation*}
$$

or from Eq. 1.3 ,

$$
\begin{equation*}
p\left(X^{\prime} \cap X\right)=p\left(X^{\prime}\right) p(X) \tag{1.5}
\end{equation*}
$$

Also from Eq. 1.3, we can find that

$$
\begin{equation*}
p\left(X^{\prime} \cap X\right)=p\left(X^{\prime} \mid X\right) p(X) \tag{1.6}
\end{equation*}
$$

where we just multiplied both sides of Eq. 1.3 by $p(X)$. Therefore

$$
\begin{align*}
p\left(X \mid X^{\prime}\right) & =\frac{p\left(X \cap X^{\prime}\right)}{p\left(X^{\prime}\right)}  \tag{1.7}\\
& =\frac{p\left(X^{\prime} \cap X\right)}{p\left(X^{\prime}\right)}  \tag{1.8}\\
& =p\left(X^{\prime} \mid X\right) \frac{p(X)}{p\left(X^{\prime}\right)} \tag{1.9}
\end{align*}
$$

which is the so called Bayes' rule. It simply states that one can invert the conditioning between two variables by multiplying it by the ratio between the individual probabilities.

## Example 1 - Tossing a coin

Consider the following sample space associated with the tossing of a single coin: $\Omega=$ $\{H, T\}$, where $H$ is a shortcut for the string "the outcome of the tossing is heads", and similar for $T$ and tails. The algebra $\mathcal{X}$ of this sample space is given by

$$
\begin{equation*}
\mathcal{X}=\{\{\emptyset\},\{H\},\{T\},\{H, T\}\} \tag{1.10}
\end{equation*}
$$

Notice that it satisfies the properties of the algebra of events: it contains the empty set as well as the whole sample space, and any logical combination of subsets is in the algebra. To see that, notice for instance that

$$
\begin{equation*}
H \cup T=\{H, T\} \in \mathcal{X} ; \quad H \cap T=\emptyset \in \mathcal{X} ; \quad H / T=\{H\} \in \mathcal{X} \tag{1.11}
\end{equation*}
$$

Feel free to try with any other combination of two subsets in $\mathcal{X}$ ! Consider now the probability measure $p: \mathcal{X} \rightarrow \mathbb{R}$ such that

$$
\begin{equation*}
p(H)=p(T)=\frac{1}{2} \tag{1.12}
\end{equation*}
$$

Notice that this measure satisfies the Kolmogorov axioms: it is non-negative and normalised, so the probability of $H \cup T$ is just the sum of the individual probabilities.

## Example 2 - Tossing two coins

Consider now the sample space associated with the tossing of two distinguishable coins, i.e., always two coin outcomes are produced. The possible outcomes will be $\Omega=\{H H, H T, T H, T T\}$, where $H H$ is the shortening for "the outcome of the tossing is heads for coin 1 and heads for coin 2", and similar for the other events.

The algebra of this set will contain all possible partitions of the set, such that

$$
\begin{gather*}
\mathcal{X}=\{\{\emptyset\},\{H H\},\{H T\},\{T H\},\{T T\},\{H H, H T\},\{H H, T H\},\{H H, T T\},  \tag{1.13}\\
\{H T, T H\},\{H T, T T\},\{T H, T T\},\{H H, H T, T H\},\{H H, H T, T T\} \\
\{H H, T H, T T\},\{H T, T H, T T\},\{H H, H T, T H, T T\}\}
\end{gather*}
$$

Finally, the map

$$
\begin{equation*}
p(H H)=p(H T)=p(T H)=p(T T)=\frac{1}{4} \tag{1.14}
\end{equation*}
$$

forms a probability measure.

### 1.2 Random variables and expectation values

Although the previous definition of the sample space $\Omega$ algebra of events $\mathcal{X}$ is sufficient to construct probabilities and the most important rules to manipulate them, the events $X$ are still abstract concepts. They could be, for instance, sentences in English describing how the outcome of a measurement is perceived in a lab. It is convenient instead to treat events as numbers, which we are much more used to manipulating.

A random variable is a function $A: \Omega \rightarrow \mathbb{R}$ mapping each event $x$ to a real number $a=A(x)$, such that it has a reasonably defined inverse map. This means that $X=A^{-1}(B)$ is in the algebra of events for any upper bounded set $B \in \mathbb{R}$. This demand ensures that we can attribute probabilities to the random variables the same way as we attribute to the events themselves, such that

$$
\begin{equation*}
p(A \in B):=p\left(A^{-1}(B)\right) . \tag{1.15}
\end{equation*}
$$

Usually, we will want to estimate expectation values of some random variable. This is defined as

$$
\begin{equation*}
\langle A\rangle:=\sum_{a \in \operatorname{Im}(A)} a p(A=a), \tag{1.16}
\end{equation*}
$$

and this definition can be generalised for any function $f: \mathbb{R} \rightarrow \mathbb{R}$ over the numbers $a$. That is,

$$
\begin{equation*}
\langle f(A)\rangle:=\sum_{a \in \operatorname{Im}(A)} f(a) p(A=a) . \tag{1.17}
\end{equation*}
$$

In particular, expectation values of functions that are just of the form $f(a)=a^{m}$ for some $m \in \mathbb{N}^{*}$ receive the special name of moments, and are relevant for the majority of statistical analyses. This however goes a bit beyond the scope of this course and thus will not be commented on.

## Example 3 - Correlators

Consider again the case where a coin is tossed. Let us consider the map $A: \Omega \rightarrow \mathbb{R}$ such that

$$
\begin{equation*}
A(H)=1, \quad A(T)=-1 \tag{1.18}
\end{equation*}
$$

In this case, the expectation value of $A$ is simply

$$
\begin{equation*}
\langle A\rangle=A(H) p(H)+A(T) p(T)=p(H)-p(T) . \tag{1.19}
\end{equation*}
$$

In the case of two coin tosses, we can define a new map $A^{\prime}$ as a function of $A$, such that

$$
\begin{align*}
A^{\prime}(H H) & =A(H) \cdot A(H), & & A^{\prime}(H T)=A(H) \cdot A(T),  \tag{1.20}\\
A^{\prime}(T H) & =A(T) \cdot A(H), & & A^{\prime}(T T)=A(T) \cdot A(T) . \tag{1.21}
\end{align*}
$$

The expectation value of $A^{\prime}$ will thus be

$$
\begin{align*}
\left\langle A^{\prime}\right\rangle & =A(H) A(H) p(H H)+A(H) A(T)[p(H T)+p(T H)]+A(T) A(T) p(T T)(1.22) \\
& =p(H H)-p(H T)-p(T H)+p(T T) . \tag{1.23}
\end{align*}
$$

This quantity often receives the name of correlator, and can be ubiquitously found in the literature of Bell nonlocality.

### 1.3 A short disclaimer

In this school, we take an operationalist approach to probabilities, attaching to them some substantial meaning about experiments. In fact, probabilities in this school will not represent "how many times this particular event happens among all possible events when we repeat this same experiment to infinity". Instead, we will interpret probabilities as the degree of likelihood that a rational agent is willing to attribute to the occurrence of an event.

Despite the bad impression that this called bayesian bias might give out, it is still as realist as the frequentist one in the sense that infinite repetitions of an experiment are never possible. Therefore, you can think of the frequentist interpretation as being subjective too, since it is up to the experimenter to determine when a sufficient number of runs of the experiment have been carried out. It is crucial to emphasize that everything discussed in this school can also be easily imported into the frequentist interpretation. The subject will however not dive deep into this topic, but students are free to search for literature on the topic if they are interested.

### 1.4 Classical causality

We are now acquainted with bayesian probability theory, but how can we tell that a correlation between random variables $p\left(A_{1}, \ldots, A_{n}\right)$ is classical? One way of doing so is by assuming that there is an underlying classically causal model to the correlations.

Let's begin by defining a notion of a classical causal model. These are described by a collection of random variables and a collection of arrows between these representing when one random has a direct causal influence on another. We can graphically denote these by
directed acyclic graphs (DAGs), for example:

where here the random variables are $A, \ldots, E$, and there are five arrows representing direct causal influence.

The fact that this is a directed graph means that the connections are represented by arrows, which captures the fact that, for example, $A$ is the cause of $E$ rather than vice versa. The fact that it is acyclic means that we don't get any causal loops. For example, we don't find a situation where $A \rightarrow B \rightarrow C \rightarrow A$ which would be paradoxical as it now says that $A$ is the cause of itself. We then want the sorts of correlations that we can have between these random variables to always come from a causal connection.

Definition 1.4.1 (Reichenbach's principle) A correlation between two random variables satisfies Reichenbach's principle if (i) there is a direct causal influence from one to the other; (ii) there is some common cause that influences them both; or (iii) there may be some common future that one has conditioned on which induces the correlation. In other words, two correlated random variables $A$ and $C$ must be causally explained by one of the following DAGs:


This implies that all correlations between the random variables $A_{1}, \ldots, A_{n}$ satisfying Reichenbach's principle will have the form

$$
\begin{equation*}
p\left(A_{1}, \ldots, A_{n}\right)=\prod_{i=1}^{n} p\left(A_{i} \mid \operatorname{Pa}\left(A_{i}\right)\right) \tag{1.26}
\end{equation*}
$$

where $\mathrm{Pa}\left(A_{i}\right)$ is the set of parent nodes of $A_{i}$, i.e., the set of all variables $A_{j \neq i}$ with an arrow pointing towards $A_{i}$. Evidently, $p\left(A_{i} \mid A_{j}\right)$ are all valid probability distributions, satisfying all the necessary axioms.

Now, the framework as we have developed it so far assumes that all of the nodes are, in principle, observed. We may marginalise over one variable to find a particular distribution, but we can also condition over all of them taking particular values. However, there will be many experiments in which a particular variable is not observable. We denote these as circular nodes, for example:


In this case, the correlations between the observed random variables $A_{1}, \ldots, A_{n}$ when there are $O_{1}, \ldots, O_{m}$ unobserved ones will have the form

$$
\begin{equation*}
p\left(A_{1}, \ldots, A_{N}\right)=\sum_{o_{1}, \ldots, o_{m}} \prod_{i=1}^{n} p\left(A_{i} \mid \mathrm{Pa}\left(A_{i}\right)\right) \prod_{j=1}^{m} p\left(O_{j}=o_{j} \mid \mathrm{Pa}\left(O_{j}\right)\right) \tag{1.28}
\end{equation*}
$$

where all that is going on here is that we're imagining that there is some underlying description in which all nodes are observed and we're just marginalising over those that we have decided are unobserved. Despite these probabilities being unconditioned, we can always apply Bayes' rule (as long as the conditioners are nonzero).

## References

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## Chapter 2

## Operational Probabilistic Theories


#### Abstract

Without an interpretation probabilities are a purely mathematical concept and thus, they cannot explain anything about nature. What makes this issue difficult is that our understanding of probabilities is overloaded from everyday experience. For instance, we may say that it is very probable that it rains tomorrow or that the probability that the universe came out the way it did is 0.0000034. But what exactly do we mean by such statements on an operational level?


- D. Frauchiger [1]


### 2.1 Operational language \& compositional rules

Before introducing the formalism of Operational Probabilistic Theories (OPTs), you might be asking yourself what it means for a theory to be operational. Essentially, this framework wants to drop off any assumptions about the underlying reality of an experiment and focus exclusively on objective elements that are intrinsic to any experiment. An experiment here is understood as a chain of actions that can be implemented in order to investigate a physical system ${ }^{1}$, in particular, instructions on how to prepare the system to be investigated, which questions can you make about it, and which answers you can get for each question.

An operational language (OL) $\Theta$ is the formalisation of these concepts. It consists of a tuple $\Theta:=($ Sys, Test, Out, Event). The elements $S, A, B, C \in \operatorname{Sys}(\Theta)$ are labeling different system types: you can think of them as coins, electrons, Hilbert or Euclidian spaces, etc.

Events $\mathscr{T}, \mathscr{A}, \mathscr{B} \in \operatorname{Event}(\Theta)$ connect system types $A$ and $B$, and may have outcomes $x, a, b \in \operatorname{Out}(\Theta)$ associated to it. You can think of them as generalisations of transformations and measurements. A test $\mathrm{T}_{X}^{A \rightarrow B}$ is essentially a set of events from system-type $A$ to system-type $B$, with outcomes in the set $X$, i.e., $\mathrm{T}_{X}^{A \rightarrow B}=\left\{\mathscr{T}_{x}^{A \rightarrow B}\right\}_{x \in X}$. Diagrammatically,

[^0]the operational language is represented by boxes and wires, such that
\[

$$
\begin{equation*}
\mathscr{T}_{x}^{A \rightarrow B} \mapsto \quad \stackrel{A}{\mathscr{T}_{x}}-\quad \mathrm{T}_{X}^{A \rightarrow B} \mapsto \quad{ }^{A} \mathrm{~T}_{X}{ }^{B} \tag{2.1}
\end{equation*}
$$

\]

Notice that the diagrammatic representation is convenient because it allows us to detach the subscripts $A \rightarrow B$ from the labels of tests and events.

We will talk about the outcomes $\operatorname{Out}(\Theta)$ in a moment, but they will demand special treatment. Right now, let us see how we can combine $\operatorname{Sys}(\Theta)$, Event $(\Theta)$, and $\operatorname{Test}(\Theta)$ to describe complex scenarios.

The first thing we might want to do is to perform tests in sequence over a system. This is a natural request when you think of an optical table, for instance, in which you can put several lenses in sequence on the same mode. Let us then define a sequential composition ○:

$$
\begin{align*}
\circ: \operatorname{Test}(B \rightarrow C) \times & \operatorname{Test}(A \rightarrow B) \rightarrow  \tag{2.2}\\
\mathrm{T}_{Y}^{\prime B \rightarrow C} \times \mathrm{T}_{X}^{A \rightarrow B} & \mapsto\left(\mathrm{~T}^{\prime} \circ \mathrm{T}\right)_{X \times Y}^{A \rightarrow C} . \tag{2.3}
\end{align*}
$$

Diagrammatically, this is telling that

We will also require some features from this product. For instance, it must be associative, which means that

$$
\begin{equation*}
{ }_{-\left(\mathrm{T}^{\prime} \circ \mathrm{T}\right)_{X \times Y}}^{C}{\mathrm{~T}_{Z}^{\prime \prime}}^{D}={ }^{A} \mathrm{~T}_{X}{ }^{B} \mathrm{~T}_{Y}^{\prime}{ }^{C} \mathrm{~T}_{Z}^{\prime \prime}-D={ }^{A} \mathrm{~T}_{X}{ }^{B}\left(\mathrm{~T}^{\prime \prime} \circ \mathrm{T}^{\prime}\right)_{Y \times Z}{ }^{D} . \tag{2.5}
\end{equation*}
$$

Also, for every system-type $A$, there should exist a trivial test ${ }{ }^{A \rightarrow A}$, such that

$$
\begin{equation*}
A \mathrm{~T}_{X}^{B} \square^{B}=A \mathrm{~T}^{A} \mathrm{~T}_{X}^{B}={ }^{B} \mathrm{~T}_{X}^{B} . \tag{2.6}
\end{equation*}
$$

We are ignoring the outcome set of $\mathbf{I}^{A \rightarrow A}$ because it is possible to prove that it is the singleton set $\star$, and moreover I is unique for each system-type in $\operatorname{Sys}(\Theta)$. This invites us to represent $\mathbf{I}^{A \rightarrow A}$ as simply the wire associated with system-type $A$.

Another way of composing elements of the operational language is to have things coexist. For instance, one could investigate a coin and a photon in the same experiment, or Alice and Bob performing simultaneous measurements in their respective labs. Take into consideration that pairs of systems, $(A, B)$, must belong to $\operatorname{Sys}(\Theta)$ as well as their individual counterparts, so we say that $A B \in \operatorname{Sys}(\Theta)$. We thus define a parallel composition $\otimes$ :

$$
\begin{align*}
& \otimes: \operatorname{Test}(A \rightarrow B) \times \operatorname{Test}(C \rightarrow D) \rightarrow \operatorname{Test}(A C \rightarrow B D)  \tag{2.7}\\
& \mathrm{T}_{X}^{A \rightarrow B} \times \mathrm{T}_{Y}^{\prime} C \rightarrow D \mapsto\left(\mathrm{~T} \otimes \mathrm{~T}^{\prime}\right)_{X \times Y}^{A C \rightarrow B D} . \tag{2.8}
\end{align*}
$$

In diagrams, we have

$$
\begin{array}{lll}
A & \mathrm{~T}_{X} & B  \tag{2.9}\\
C & \mathrm{~T}_{Y}^{\prime} & D
\end{array} \equiv{ }^{A C}\left(\mathrm{~T}^{\left(\mathrm{T} \otimes \mathrm{~T}^{\prime}\right)_{X \times Y}}{ }^{B} D .\right.
$$

We will also demand associativity from this product,
as well as the existence of a trivial system-type $I$, which represents not having a system and can thus be added to or ignored in any diagram:

$$
\begin{equation*}
\underline{A}{ }^{A} \mathrm{~T}_{X}{ }_{I}^{B}=\frac{I}{A{\mathrm{~T}_{X}}^{B}}={ }^{A}{\mathrm{~T}_{X}}^{B} \tag{2.11}
\end{equation*}
$$

In particular, events of the form $\mathscr{P}_{x}^{I \rightarrow A}$ receive a special interpretation of preparation events. You can understand it as a procedure describing picking a coin out of a bag and tossing it, or a crystal emitting a photon in a particular state. Similarly, there are procedures going from non-trivial system types to the trivial $I, \mathscr{O}_{y}^{A \rightarrow I}$, which we are calling observation events. We will represent such events as

$$
\begin{equation*}
\mathscr{P}_{x}^{I \rightarrow A} \mapsto \mathscr{P}_{x}-A, \quad \mathscr{O}_{y}^{A \rightarrow I} \mapsto A-\mathscr{O}_{y} . \tag{2.12}
\end{equation*}
$$

We finally require that for any tests $T_{X}^{A \rightarrow B}, R_{Y}^{B \rightarrow C}, Q_{Z}^{D \rightarrow E}, S_{W}^{E \rightarrow F}$, we have

$$
\begin{equation*}
\left(R_{Y}^{B \rightarrow C} \otimes S_{W}^{E \rightarrow F}\right) \circ\left(T_{X}^{A \rightarrow B} \otimes Q_{Z}^{D \rightarrow E}\right)=\left(R_{Y}^{B \rightarrow C} \circ T_{X}^{A \rightarrow B}\right) \otimes\left(S_{W}^{E \rightarrow F} \circ Q_{Z}^{D \rightarrow E}\right), \tag{2.13}
\end{equation*}
$$

which looks very intricate algebraically, but when put into diagrams becomes very natural,

since if you just ignore all the dashed boxes it is immediate to see that the equality holds.

### 2.2 Outcomes \& probabilistic models

Outcomes $\operatorname{Out}(\Theta)$ in the operational language represent the labels of the events in a particular test. As abstract as it might sound, you can think of a test "checking which face of a coin is upwards". The possible events correspond to the coin being with heads up, or tails up. Outcomes, therefore, are the labels $H$ and $T$ associated with each of these events.

Like systems and tests, there is a trivial outcome set $\star \in \operatorname{Out}(\Theta)$, representing the singleton set $\star:=\{*\}$. Any test containing $\star$ as the set of outcomes is therefore related to a single event and is called a deterministic test. When this is the case, we will represent them without the subscript, $\mathrm{T}_{\star}^{A \rightarrow B} \equiv \mathrm{~T}^{A \rightarrow B}$. The trivial test $\mathfrak{I}^{A \rightarrow A}$, for instance, is an example of a deterministic test.

Another possible way of accommodating outcomes is by representing them diagrammatically as special system-types. In this convention, $T_{X}^{A \rightarrow B}$ is represented by

$$
\begin{equation*}
\mathrm{T}_{X}^{A \rightarrow B} \mapsto \stackrel{A}{\mathrm{~T}^{-} \frac{X}{B}} . \tag{2.15}
\end{equation*}
$$

Notice that this convention is perfectly compatible with the properties demanded from sequential and parallel compositions. We will mostly use the first notation introduced of outcomes as subscripts, but the notation with outcomes as wires will be convenient later when demonstrating some examples.

We also defined preparations and measurements as tests going from or to the trivial system $I$. A test going from and to the trivial system, i.e., $\mathrm{P}_{X}^{I \rightarrow I} \in \operatorname{Test}(\Theta)$ is a special element of the operational language and receives the name of scalar.

A probabilistic model consists of an operational language $\Pi$ to which every scalar $P_{X}^{I \rightarrow I}$ is associated to a probability distribution $\{p(x)\}_{x \in X}$, i.e., every scalar is a function over the outcomes satisfying the Kolmogorov axioms introduced in the previous chapter. Most generally, they will have the form

$$
\begin{equation*}
\{p(x, y, z)\}_{x \in X, y \in Y, z \in Z}=\mathrm{P}_{X} \mathrm{~T}_{Y}=\mathrm{O}_{Z} \tag{2.16}
\end{equation*}
$$

In fact, all these distributions are conditional to the tests $\mathrm{P}_{X}, \mathrm{~T}_{Y}$, and $\mathrm{O}_{Z}$ being implemented as per the provided diagram. This association of scalars to probability distributions equips the operational language with all the features introduced in Chapter 1 and therefore results such as Bayes' rule can be rederived.

### 2.3 Quotient theories

Consider now two events $\mathscr{T}_{x}^{A \rightarrow B}$ and $\mathscr{T}_{y}^{\prime} A \rightarrow B \in \operatorname{Event}(\Pi)$, which are not necessarily equal or associated to the same outcome. However, it might be the case that

Whenever this happens, we say that $\mathrm{T}_{x}^{A \rightarrow B}$ is probabilistically equivalent to $\mathrm{T}_{y}^{\prime} A \rightarrow B$, and represent it as $\mathrm{T}_{x}^{A \rightarrow B} \sim \mathrm{~T}_{y}^{\prime} A \rightarrow B$. It is possible to prove that $\sim$ indeed constitutes an equivalence relation.

We can then look not only to the set Event( $\Pi$ ) but to its partition Event $(\Pi) / \sim$ in which every element is a subset of events probabilistically equivalent to each other. Such mapping is called quotienting. Notice that quotienting the events implies changes on the tests as well since they are just sets of events. In particular, some of these quotiented partitions receive special labels:

- The set $\operatorname{Test}(A \rightarrow B) / \sim$ of tests from non-trivial to non-trivial system-types is labeled as $\operatorname{Instr}(A \rightarrow B)$, the set of instruments;
- The set $\operatorname{Event}(A \rightarrow B) / \sim$ from non-trivial to non-trivial system-types is labeled $\operatorname{Transf}(A \rightarrow B)$, the set of transformations;
- The set $\operatorname{Event}(I \rightarrow A) / \sim$ from the trivial to a non-trivial system-type is labeled St $(A)$, the set of states;
- The set $\operatorname{Event}(A \rightarrow I) / \sim$ from a non-trivial to the trivial system-type is labeled Eff $(A)$, the set of effects.

A quotiented probabilistic model $(\operatorname{Sys}(\Pi), \operatorname{lnstr}(\Pi), \operatorname{Transf}(\Pi), \operatorname{St}(\Pi), \operatorname{Eff}(\Pi), \operatorname{Out}(\Pi))$ in which all scalars are combined through multiplication of real numbers constitutes an operational probabilistic theory (OPT).

A very relevant feature of an OPT is that, since it assigns physical meaning to the scalars only (in the sense that they are the only directly observable element in the theory), every scalar boils down to a prepare-and-measure scenario, no matter how complex the underlying diagram might originally be. It means that, for example,

Conversely, prepare-and-measure scenarios are the only ones for which the OPT must assign a definite probability distribution. It doesn't mean that claims about probabilities
of tests or events occurring cannot exist in an OPT, however, they do not represent an objective probability distribution, but merely a degree of belief of an agent about the presence of a particular element in the closed diagram.

### 2.4 Linear structure

We finally have tools to establish a linear structure to the OPT. Although this can be derived straightforwardly from the definition of OPTs and probability theory, we will just provide the theorem and explore its implications.

Theorem 2.4.1 (Linear structure for OPTs) Let $\Theta$ be an OPT. Then $\operatorname{Transf}(A \rightarrow B)$ can be embedded into a real vector space $\operatorname{Transf}_{\mathbb{R}}(A \rightarrow B)$, such that the two operations + (sum) and • (scalar multiplication) are well-defined and

-     + is distributive over parallel and sequential composition;
- . is compatible with parallel and sequential composition.

What the theorem means is that, given a transformation $\mathscr{T}_{i}^{A \rightarrow B}, \mathscr{T}^{\prime} B \rightarrow C$, one has

$$
\begin{align*}
& \left(\sum_{i} q_{i} \frac{A}{\mathscr{\mathscr { T }}_{i}} \underline{B}^{B}\right) \circ \underline{B}{\underline{\mathscr{T}^{\prime}}}^{C}=\sum_{i} q_{i} \frac{A}{\mathscr{T}_{i}}{ }^{B} \mathscr{\mathscr { T }}^{\prime} C  \tag{2.19}\\
& \left(\begin{array}{c}
\left(\sum_{i} q_{i}-\frac{A}{\mathscr{T}_{i}} \frac{B}{\sqrt[B]{\mathscr{T}^{\prime} C}}\right.
\end{array}\right)=\sum_{i} q_{i} \frac{A}{\mathscr{T}_{i}} \frac{B}{\mathscr{T}^{\prime} C} . \tag{2.20}
\end{align*}
$$

In particular, scalars and summations can "float" in the diagram, i.e., they can be moved in the diagram as one sees fit. They both have a clear interpretation with respect to the underlying operational language as well: summing is a coarse-graining of the underlying events so that the outcomes related to the summed-up events are being overlooked or ignored by the agent reasoning about the particular experiment. Scalar multiplication on its turn is a randomisation, in which the agent is just attributing to the particular diagram a chance of occurrence equal to the scalar it is being multiplied by.

Since we are embedding $\operatorname{Transf}(A \rightarrow B)$ into a real vector space, and ultimately any transformation can be absorbed by a state and an effect, it is very natural to read states and effects as vectors in this real vector space. In particular, states $\mathscr{P}_{x} \in \operatorname{St}(A)$ are mapped to vectors $\left.\mid \rho_{x}\right) \in \mathbb{R}^{m}$, and effects $\mathscr{A}_{y} \in \operatorname{Eff}(A)$ are mapped to vectors $\left(a_{y} \mid \in \mathbb{R}^{m *}\right.$ in the dual of $\mathbb{R}^{m}$. One can thus think of scalars as the inner product between states and effects ${ }^{2}$,

[^1]$p\left(x, y \mid \rho_{x}, a_{y}\right)=\left(a_{y} \mid \rho_{x}\right)$. Notice that coarse-graining over all possible outcomes implies
\[

$$
\begin{equation*}
\sum_{y \in Y}\left(a_{y} \mid \rho_{x}\right)=\sum_{y \in Y} p\left(x, y \mid \rho_{x}, a_{y}\right)=p\left(x \mid \rho_{x}, a_{y}\right) \tag{2.21}
\end{equation*}
$$

\]

But as argued previously, the OPT cannot say anything about an event in its underlying diagrammatic explanation if there is no outcome associated with it. It is then natural to demand that the function above does not depend on the choice of $a_{y}$. This is called causality: demanding that the outcome statistics of a scalar test, when ignoring the outcomes of the observation, does not depend on the choice of measurement. It is very important to emphasize that not all OPTs will satisfy this property, and no OPT needs to satisfy this to admit a linear structure.

Nonetheless, admitting such an assumption does yield convenient properties to the OPT. For instance, it implies the existence of a unique unit effect $\left(u \mid \in \operatorname{Eff}_{\mathbb{R}}(A)\right.$ for every system-type $A$, such that

$$
\begin{equation*}
(u \mid \rho) \leq 1, \quad \forall \mid \rho) \in \operatorname{St}_{\mathbb{R}}(\Pi) \tag{2.22}
\end{equation*}
$$

It can be shown that the converse also holds: the existence of a unique unit effect for every system-type in an OPT implies that this OPT satisfies the causality assumption. This unit effect can be understood as taking the trace in standard quantum theory.

### 2.5 Examples

## Classical bit

The classical bit consists of a subnormalised probability distribution $\{p(0), p(1)\}$. Let us assume this distribution is normalised, and call $p(0)=q$. States can straightforwardly be embedded into a 2-dimensional real vector space, such that any valid preparation has the form

$$
\begin{equation*}
\mid \rho)=\binom{q}{1-q} \tag{2.23}
\end{equation*}
$$

Effects in this case are all vectors $(a \mid$ such that $0 \leq(a \mid \rho) \leq 1$. We can represent them as per Figure 2.1. The unit effect is therefore simply $(u)=(1,1)$, and the transformations are substochastic maps, taking states $\mid \rho) \in \operatorname{St}\left(\mathbb{R}^{2}\right)$ to $\operatorname{St}\left(\mathbb{R}^{2}\right)$.

## Qubit

In quantum theory, the qubit is represented by a Hilbert space of dimension 2, with states of the form

$$
\begin{equation*}
\rho=\frac{1}{2}\left(\operatorname{Tr}\{\rho\} \mathbb{1}+\langle X\rangle \sigma_{X}+\langle Y\rangle \sigma_{Y}+\langle Z\rangle \sigma_{Z}\right), \tag{2.24}
\end{equation*}
$$

where $\mathbb{1}$ is the identity matrix, $\sigma_{X}, \sigma_{Y}$ and $\sigma_{Z}$ are the Pauli operators and $\langle X\rangle,\langle Y\rangle$ and $\langle Z\rangle$ are the expectation values of these operators.


Figure 2.1: Real vector space representation for the sets of normalised states and effects associated with a classical bit distribution.



Figure 2.2: Real vector space representation for the sets of normalised states and effects associated with a quantum bit distribution.

Each density operator of a qubit already represents an equivalence class for preparations, since for instance the maximally mixed state will not distinguish whether the preparation employed was a convex mixture of pure $Z$ states or $X$ states, despite these being two different preparation procedures. All of these are represented by the same state, $\rho=\frac{1}{2} \mathbb{1}$.

We can then easily represent $\mid \rho)$ as vectors in $\mathbb{R}^{4}$ :

$$
\mid \rho)=\left(\begin{array}{c}
\operatorname{Tr}\{\rho\}  \tag{2.25}\\
\langle X\rangle \\
\langle Y\rangle \\
\langle Z\rangle
\end{array}\right)
$$

The maximally mixed state, for instance, has the form

$$
\mid \rho)=\frac{1}{2}\left(\begin{array}{l}
1  \tag{2.26}\\
0 \\
0 \\
0
\end{array}\right)
$$

The effects in the qubit are precisely the Bloch sphere again, so the dual of $\mathrm{St}_{\mathbb{R}^{4}}\left(\mathcal{H}_{2}\right)$
is itself. Effects are thus simply given by

$$
\left(a \left\lvert\,=\left(\begin{array}{llll}
\operatorname{Tr}\{a\} & \langle X\rangle^{\prime} & \langle Y\rangle^{\prime} & \langle Z\rangle^{\prime} \tag{2.27}
\end{array}\right) .\right.\right.
$$

The unit effect is given by $(u)=(1,0,0,0)$. The standard graphical representation of the Bloch sphere is a 3 -dimensional projection of the 4 -dimensional hypersphere formed by the vectors $\mid \rho)$ satisfying $0 \leq(u \mid \rho) \leq 1$, and can be found in Figure 2.2. Transformations in the qubit are all the maps $T: \mathrm{St}_{\mathbb{R}^{4}}\left(\mathcal{H}_{2}\right) \rightarrow \mathrm{St}_{\mathbb{R}^{4}}\left(\mathcal{H}_{2}\right)$ that take valid states into valid states again, even when $T$ is applied to one part of a bipartite system.

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## Chapter 3

## Spekkens' Toy Theory

(...) Where can you find a place that will agree better with you and me? No schools, no teachers, no books! In that blessed place there is no such thing as study. Here, it is only on Saturdays that we have no school. In the Land of Toys, every day, except Sunday, is a Saturday. Vacation begins on the first of January and ends on the last day of December.(...)

- C. Collodi [1]


### 3.1 Ontic vs. epistemic - Spekkens' toy theory

In classical theory, a system is usually described by a point in the phase space in which each degree of freedom is associated with a canonical coordinate (for instance, the position and momentum $(x, p)$ of a single particle). The possible states for this system to be in are given by the trajectory that satisfies the equations of motion, once the initial conditions are specified. If one is not certain about the initial conditions but rather assigns probabilities to each pair $\left(x_{0}, p_{0}\right)$ of possible initial conditions in a region of the phase space, then the possible states for the system are all the trajectories associated with each of these initial conditions, weighted by the probability assigned to them.

This probability distribution of possible states in which your system might be in is what is called epistemic state. The name comes from the ancient Greek epistémē, knowledge, so epistemic states represent states of knowledge about a system's state. Naturally, if the probability distribution in question is a Dirac delta ${ }^{1}$, then the agent is completely sure about which is the initial state of the system and can conclude which trajectory of the phase space the system will assume when it evolves with time. This state in which your system truly is is called ontic state, from the Greek ón, being, existing. Ontic states are therefore the fundamental states your system is occupying.

[^2]

Figure 3.1: Ontic states in Spekkens' toy theory.

When we talk about quantum theory, the most widespread belief is that pure states are ontic, and mixed states are epistemic. This belief comes from the fact that as well as in classical theory, the evolution of a pure state is completely specified by initial conditions. Establishing a parallel with the previous chapter, an isolated qubit under the action of a constant magnetic field will completely specify a trajectory on the surface of the Bloch sphere, once the initial state $\left|\psi_{0}\right\rangle$ is given. Spekkens' toy theory is an attempt to defend the idea that all quantum states, both pure and mixed, are epistemic. In short, many of the apparently weird features of quantum theory, such as interference and entanglement, are totally mundane when quantum states are viewed as epistemic rather than ontic.

Consider a toy bit: a system with two degrees of freedom, $(Q, P)$, such that to each of these physical quantities it can be assigned value 0 or 1 . In this phase space, there will be 4 possible states for a system to be in: $(0,0),(0,1),(1,0)$, and $(1,1)$. In Figure 3.1, you can see a representation of this phase space as a column. We also introduce the following rule in our theory:

An agent can only be certain about the value of one degree of freedom per system, being completely ignorant about the values of any other degree of freedom.

This principle is known as knowledge balance principle, epistemic restriction or principle of classical complementarity, the last one being justified by its similarity with the quantum complementarity principle that acts over canonical pairs. What this principle is imposing is that an agent cannot know states of the form of Figure 3.1. Instead, when certain that $Q=0$, an agent should not be able to tell whether the system is in state $(0,0)$ or $(0,1)$, and similarly for certainty about other values of $Q$ or $P$. It means that the valid states are the ones given in Figure 3.2. There is a natural mapping from these valid epistemic states and pure quantum states of the qubit: the first two states represent the $|0\rangle$ and $|1\rangle$ of the Bloch sphere; the third and fourth states represent $|+\rangle$ and $|-\rangle$; the fifth and sixth represent $|+i\rangle$ and $|-i\rangle$. The last epistemic state is associated with a mixed state, in particular the maximally mixed one, $\mathbb{1} / 2$.

Transformations in the toy theory must map valid epistemic states to valid epistemic states again, therefore no transformation can provide an agent with complete certainty about the ontic state of the system. But more than that, we demand that transformations


Figure 3.2: Valid epistemic states in Spekkens' toy theory
operate on the ontic level, i.e., they change epistemic states by acting directly on the ontic state. For instance, a transformation that swaps $(0,0)$ and $(1,1)$, while maintaining $(0,1)$ and $(1,0)$ invariant, will map epistemic states to valid epistemic states, since

leaving the other epistemic states unchanged. This definition will be particularly relevant when discussing the bipartite case later.

Valid measurements provide certainty about the value of a single degree of freedom, and "mix up" any information about the other one. For instance, measuring the value $Q=0$ will just refresh any state of knowledge about the ontic state to the first epistemic state of Fig. 3.2. Measurements, however, are usually not reversible. Notice also that if for instance one measures $Q=0$ on the state

one might conclude that the ontic state of the system prior to the measurement must have been $(0,1)$. After the measurement however this might be no longer the case, due to the complementarity principle imposed over the epistemic states.

One can compose systems in the toy theory in the following manner: two systems will be in ontic states $(Q, P)_{A}$ and $(Q, P)_{B}$, respectively. The valid ontic states for the
composite systems will be one out of the 16 entries of the following table


The complementarity principle tells us that one cannot have complete knowledge about a single system, which means that one cannot be certain about both $Q_{A}$ and $P_{A}$. However, being certain about $Q_{A}$ and $P_{B}$ for instance does not posit any violation of the principle, yielding valid epistemic states. Some examples of valid epistemic states for two toy bits are given in Figure 3.3.

It is easier to understand what states are not valid for the case of two toy bits. For instance, the state

is not valid, since an agent would have maximal knowledge about the ontic state of system $B$. Another more intricate example is given by


The problem with this state is that if one agent performs a measurement and learns that $Q_{A}=1$ and $Q_{B}=0$, which is allowed by the complementarity principle, the outcome of


Figure 3.3: Some examples of valid epistemic states for two toy bits in the toy theory. Any permutations of rows and columns for these states are also valid in the theory.
the measurement could be given by

which is not allowed since again the agent would know both $Q_{B}$ and $P_{B}$. In fact, after some leveraging of possible states by considering this aspect of the principle, one can conclude that the only valid epistemic states are the ones in Figure 3.3 and permutations thereof.

## 3.2 "Quantum" features in the toy theory

There are plenty of features displayed in the toy theory that have an immediate analogy with quantum behaviours. We briefly comment on some of them here.

## Purity

Pure epistemic states are states of maximum knowledge. By the complementarity principle, this means that they are the states in which an agent is certain about exactly one outcome of the canonical pair for each system. By this definition, all but the last epistemic state in Figures 3.2 and 3.3 are pure states, and they are the only pure states in the theory.

## Convex combination

Like in quantum theory, it is possible to describe convex combinations of epistemic states in a very straightforward manner: the convex combination of two epistemic states consists of all possible ontic states inferred by both. For instance,

## Coherence

Coherence is the generalisation of the notion of superposition in quantum theory, i.e., the idea that two pure states can be combined into another pure state, as opposed to the convex combination that will always result in a mixed state (a state of non-maximum
knowledge). There are in fact four of such operations in the toy theory. For the states corresponding to $Q=0$ and $Q=1$, for instance, they yield


The same can be defined for other combinations of pure states. The parallel with quantum theory comes by interpreting the above equalities as

$$
\begin{align*}
\frac{1}{\sqrt{2}}(|0\rangle+|1\rangle)=|+\rangle ; \quad \frac{1}{\sqrt{2}}(|0\rangle-|1\rangle)=|-\rangle ;  \tag{3.10}\\
\frac{1}{\sqrt{2}}(|0\rangle+i|1\rangle)=|+i\rangle ; \quad \frac{1}{\sqrt{2}}(|0\rangle-i|1\rangle)=|-i\rangle, \tag{3.11}
\end{align*}
$$

and so on for other pure states.

## Non-orthogonality

We can say that two epistemic states are nonorthogonal if they share at least one ontic state among the possible ones in the state of knowledge they represent. For instance, the states

are all non-orthogonal, since all infer the state $(0,0)$ as a possible ontic state. On the other hand, the states

are said to be orthogonal, since the set of possible ontic states for both is disjoint. In classical theory, all pure states are orthogonal to each other, which does not hold in the toy theory.

## Entanglement

The states

and their permutations of rows and columns are all valid and pure epistemic states in the theory. However, nothing is particularly known about the individual outcomes $Q$ or $P$ of each toy bit. Instead, the maximum knowledge is on how these quantities are related between the bits, such that once you find out the value of $Q_{A}$, for instance, you should be able to infer either $Q_{B}$ or $P_{B}$ with certainty. The first state, for example, tells that the ontic state of the bit $A$ is always the same as the ontic state of the bit $B$, but it does not tell which ontic state is it. The others follow a similar reasoning.

Pure states for composite systems that tell nothing about the individual systems are called entangled states in quantum theory. In classical theories, having some maximal knowledge about a composite system always implies maximal knowledge about the individual systems as well, but the complementarity principle imposes a trade-off between local and global knowledge.

## Geometrical structure

As introduced in the previous chapter, the toy theory can also be accommodated in the framework of OPTs and therefore embedded into a real vector space. Unlike both classical and quantum theory, its state of spaces $\mathrm{St}_{\mathbb{R}}$ (ToyBit) is given by the one in Figure 3.4 .


Figure 3.4: Set of states of the real vector space representation of the toy bit OPT.

### 3.3 The toy theory in ZX language

A very interesting feature of the toy theory is that it can easily be imported into the framework of ZX calculus. For this purpose, we merely need to associate some of its components with the graphical components of ZX. The first of such components will be the cloning spider:
G
such that it maps ontic states to epistemic states in the following manner:


Furthermore, we will establish the following convention: spiders with no input and one output will be always labeled with a phase $00,01,10$, or 11 , and will represent the following epistemic states:


Finally, we introduce a Hadamard transformation

$$
\begin{equation*}
\square- \tag{3.18}
\end{equation*}
$$

that will map ontic states to ontic states of single toy bits such that $(0,1) \leftrightarrow(1,0)$, while the other ontic states remain unaltered. We also establish that any spider with Hadamards applied to all its legs is written as a red spider, i.e.,

where a spider with multiple legs is just a concatenation of nested copying spiders and their respective adjoint spiders.

With only these tools, we can see how all properties of ZX calculus follow naturally. For instance, the spider rule will be such that

where $\oplus$ is the sum module 2. Finally, all bialgebra properties of ZX calculus hold for phase 00 :


Finally, it is possible to demonstrate that the toy theory calculus is complete by employing the same completeness proofs as for standard ZX. This is a much more convenient result: now we can leverage any knowledge about ZX calculus to demonstrate things with the toy theory!

### 3.4 Example: the Peres-Mermin proof of non-classicality

The Peres-Mermin square is a classic proof of non-classicality of quantum theory. It consists of a table of measurements over a two-qubit state, with the form

| $\mathbb{1} \otimes \sigma_{Z}$ | $\sigma_{Z} \otimes \mathbb{1}$ | $\sigma_{Z} \otimes \sigma_{Z}$ |
| :---: | :---: | :---: |
| $\sigma_{X} \otimes \mathbb{1}$ | $\mathbb{1} \otimes \sigma_{X}$ | $\sigma_{X} \otimes \sigma_{X}$ |
| $-\sigma_{X} \otimes \sigma_{Z}$ | $-\sigma_{Z} \otimes \sigma_{X}$ | $\sigma_{Y} \otimes \sigma_{Y}$ |

Each row of this matrix multiplies to $\mathbb{1} \otimes \mathbb{1}$, except for the last row that results in $-\mathbb{1} \otimes \mathbb{1}$. Then, we assume that each of these observables has a definite outcome $\pm 1$ assigned to it, without giving attention to how these measurements will be implemented. This assumption is incompatible with quantum theory since it is simply impossible to replace the entries of the square by $\pm 1$ in such a way that the quantum constraints are satisfied!

The Peres-Mermin square is considered a proof of contextuality, a signature of nonclassicality that will be explored in the next chapter. Interestingly enough, Spekkens' toy theory cannot prove non-classicality in this case. That is because, by replacing the quantum measurements with the corresponding toy measurements, i.e.,

| -- | -0- | 二〇- |
| :---: | :---: | :---: |
| -0- | -0- | O- |
| $\begin{equation*} -\overline{-0} \tag{3.23} \end{equation*}$ | $-\underset{0}{-0}$ | $\begin{aligned} & -0-0-1 \\ & =0-0 \end{aligned}$ |

where each node has phase 01 . It is possible to check that every row and column of this square when concatenated sequentially, will yield merely two identities paired up, the last row inclusive. This means that all entries of the square can be assigned values $\pm 1$ with no contradiction.

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## Chapter 4

## Signatures of Non-Classicality

To suppose two things indiscernible, is to suppose the same thing under two names. And therefore to suppose that the universe could have had at first another position of time and place, than that which it actually had; and yet that all the parts of the universe should have had the same situation among themselves, as that which they actually had; such a supposition, I say, is an impossible fiction.

- G. W. F. von Leibniz [1]


### 4.1 Bell theorem \& quantum violations

Last chapter, we learned that many quantum features are displayed by Spekkens' toy theory, a classical theory with an epistemic restriction. The last section introduced the Peres-Mermin square, an example of something that quantum theory predicts but the toy theory does not. This raises the question of what is truly non-classical in quantum theory, or put in other words, what does quantum theory can explain, but the toy theory cannot?

Bell nonlocality consists of a particular feature of operational theories that can accommodate a space-time structure. Consider the following Bell experiment: two agents, the ubiquitous Alice and Bob, are placed in separate laboratories. Each of them receives a system produced by the same source, chooses a measurement to perform over their share of the system, and registers the outcome observed. It is common to label $\mathbb{X}, \mathbb{Y}$ the sets of labels for the measurement choices of Alice and Bob, respectively, and $\mathbb{A}, \mathbb{B}$ the labels for the measurement outcomes. Without loss of generality, we will consider that every measurement choice has the same set of outcome labels, i.e., every measurement in $\mathbb{X}$ has the same set of outcomes $\mathbb{A}$, and the same for Bob. The setup of a Bell experiment is represented in Figure 4.1 a.

Finally, we assume that the laboratories are space-like separated. What it means is that their light cones have no intersections during the whole Bell experiment, from the moment


Figure 4.1: (a) OPT representation of a Bell experiment; (b) space-time diagram representing Alice and Bob's laboratories when the experiment happens. Notice that the diagonal lines, representing the trajectory of light, intersect only before or after the experiment is concluded.
they choose what measurements to perform and receive their halves of the system until they have performed the measurements and registered the outcomes. This is illustrated in Figure 4.1-b.

This operational setup produces a conditional probability distribution

$$
\begin{equation*}
\mathbb{P}_{\mathbb{A} \mathbb{B} \mid \mathbb{X Y}}:=\{p(a b \mid x y)\}_{a \in \mathbb{A}, b \in \mathbb{B}, x \in \mathbb{X}, y \in \mathbb{Y}} \tag{4.1}
\end{equation*}
$$

Furthermore, the constraint that they are space-like separated (which we will call nosignaling condition) implies that coarse-graining over the outcomes of a party implies ignoring also the measurement choice this party has made, i.e.,

$$
\begin{align*}
& \sum_{a \in \mathbb{A}} p(a b \mid x y)=p(b \mid y), \quad \forall b \in \mathbb{B}, x \in \mathbb{X}, y \in \mathbb{Y}  \tag{4.2}\\
& \sum_{b \in \mathbb{B}} p(a b \mid x y)=p(a \mid x), \quad \forall a \in \mathbb{A}, x \in \mathbb{X}, y \in \mathbb{Y} \tag{4.3}
\end{align*}
$$

This is similar to the causality condition defined for an OPT.
Now consider a causal structure given by the following graph

where the square nodes represent observed random variables, and the circular node represents an unobserved or ignored variable. If we demand that the correlations in a Bell
experiment are explained by this graph, in which all nodes are classical random variables, then they must necessarily have the form

$$
\begin{equation*}
p(a b \mid x y)=\sum_{\lambda \in \Lambda} p(\lambda) p(a \mid x \lambda) p(b \mid y \lambda), \quad \forall a \in \mathbb{A}, b \in \mathbb{B}, x \in \mathbb{X}, y \in \mathbb{Y} \tag{4.5}
\end{equation*}
$$

Bell's theorem demonstrates that there are operational theories, in particular quantum theory, satisfying the no-signaling condition and that yet cannot be explained by the above classical, causal structure (which is often called locally causal). It is easier to understand this result by looking at a particular case, where $\mathbb{X}=\mathbb{Y}=\mathbb{A}=\mathbb{B}=\{0,1\}$. Mathematically, each experiment with fixed measurements $x, y \in\{0,1\}$ is exactly the same as the example of tossing two coins simultaneously, provided in Chapter 1, where we assign values 0 to $H$ and 1 to $T$. We can then compute the correlators

$$
\begin{equation*}
E_{x y}=p(00 \mid x y)-p(01 \mid x y)-p(10 \mid x y)+p(11 \mid x y), \quad \forall x, y \in\{0,1\} . \tag{4.6}
\end{equation*}
$$

If the probabilities $p(a b \mid x y)$ satisfy local causality, then there is $\Lambda$ such that they have the form of Equation 4.5 and the correlators can be rewritten as

$$
\begin{align*}
E_{x y} & =\sum_{a, b=0}^{1}(-1)^{a+b} p(a b \mid x y)  \tag{4.7}\\
& =\sum_{a, b=0}^{1}(-1)^{a+b} \sum_{\lambda \in \Lambda} p(\lambda) p(a \mid x \lambda) p(b \mid y \lambda)  \tag{4.8}\\
& =\sum_{\lambda \in \Lambda} p(\lambda)\left(\sum_{a=0}^{1}(-1)^{a} p(a \mid x \lambda)\right)\left(\sum_{b=0}^{1}(-1)^{b} p(b \mid y \lambda)\right)  \tag{4.9}\\
& =\sum_{\lambda \in \Lambda} p(\lambda) E_{x}(\lambda) E_{y}(\lambda), \tag{4.10}
\end{align*}
$$

where $E_{x}(\lambda)$ and $E_{y}(\lambda)$ are simply the expectation values of the individual coin tosses, parametrised by the choice of $\lambda$ and $x, y$.

Consider now the Clauser-Horne-Shimony-Holt (CHSH) functional for this particular scenario

$$
\begin{equation*}
I_{C H S H}\left(\mathbb{P}_{\mathbb{A B} \mid \mathbb{X Y}}\right)=E_{00}+E_{01}+E_{10}-E_{11} . \tag{4.11}
\end{equation*}
$$

If each correlator factorises, it is possible to demonstrate that

$$
\begin{equation*}
\left|E_{00}+E_{01}+E_{10}-E_{11}\right| \leq 2, \tag{4.12}
\end{equation*}
$$

which is the famous CHSH inequality.
Consider now the experiment in which Alice and Bob, each in their separate lab, receive a qubit. This pair of qubits was prepared in the state

$$
\begin{equation*}
|\Phi\rangle=\frac{1}{\sqrt{2}}\left(|0\rangle_{A}|1\rangle_{B}+|1\rangle_{A}|0\rangle_{B}\right) . \tag{4.13}
\end{equation*}
$$

If Alice chooses $x=0$, it means that she is going to perform measurement $Z$ on her qubit, registering $a=0$ if she gets outcome 0 and $a=1$ otherwise. If she chooses $x=1$, then she performs measurement $X$ over the qubit registering outcomes in a similar way. Bob measures $\frac{1}{\sqrt{2}}(X+Z)$ when he chooses $y=0$, and $\frac{1}{\sqrt{2}}(X-Z)$ when $y=1$, registering outcomes like Alice.

By computing the correlators $E_{x y}=\langle\Phi| A_{x} \otimes B_{y}|\Phi\rangle$, one can get to the value

$$
\begin{equation*}
\left|E_{00}+E_{01}+E_{10}-E_{11}\right|=2 \sqrt{2}>2 . \tag{4.14}
\end{equation*}
$$

Because assuming that the correlations satisfy local causality implies inequality 4.12 , violating the inequality implies that the correlations do not admit of a locally causal explanation. Quantum theory is therefore deemed as non-local in the sense of Bell.

### 4.2 Non-signalling correlations and postquantum violations

The violation obtained in the previous example is in fact the best violation one can get with quantum theory for the CHSH inequality. However, if one considers all possible correlations $\mathbb{P}_{\mathbb{A} \mathbb{B} \mid X Y}$ satisfying the no-signaling condition for the Bell scenario we investigated, one can conclude that

$$
\begin{align*}
\left|E_{00}+E_{01}+E_{10}-E_{11}\right| & \leq \sum_{x, y=0}^{1}\left|E_{x y}\right|  \tag{4.15}\\
& \leq 4 \tag{4.16}
\end{align*}
$$

This means that the optimal quantum violation, which we will refer to as the Tsirelson bound, is not tight: other probabilistic models out there are compatible with relativistic assumptions, and yet cannot be explained by quantum theory. This is illustrated in Figure 4.2

Exploring these postquantum theories is more than a creative exercise. It has proven to be an invaluable tool for exploring what features of quantum theory are inherently quantum, in the sense that they cannot be explained by classical theory while at the same time ruling out other possible posquantum theories.

### 4.3 Generalised contextuality

A more general notion of non-classicality that is not present in the toy model is generalised contextuality, or simply contextuality from now on. Differently from Bell nonlocality, assessments of contextuality do not demand a space-time structure and thus can be present in many more operational theories. The object of study of contextuality is often a prepare-and-measure scenario, something that was briefly discussed in Chapter 2. Because any


Figure 4.2: Representation of classical (normal line), Tsirelson (dashed line), and nonsignaling (doubled line) bounds for the 2-measurements-2-outcomes Bell scenario. Quantum correlations (thick circle) never violate the Tsirelson bound, but other non-signaling correlations can.
operational scenario yielding probabilities can be reduced to a prepare-and-measure scenario, contextuality can also be assessed in a vast range of experiments.

To study prepare-and-measure scenarios, one must only know the preparations, observations, outcomes, and statistics in the operational language. We will call $\mathcal{P}$ the set of preparations, $\mathcal{M}$ the set of measurements, $K$ the set of outcomes, and $p$ the shortcut for $\{p(k \mid M, P)\}_{k \in K, M \in \mathcal{M}, P \in \mathcal{P}}$ the set of correlations derived from this scenario. As before, we can then represent the operational scenario by the tuple ( $\mathcal{P}, \mathcal{M}, K, p$ ).

As introduced in chapter 2, we want to quotient this operational language, so that any information that cannot be captured by preparing and measuring is ignored. We will keep using the symbol $\sim$ to tell that two preparations or measurement outcomes are operationally equivalent.

As in the Bell scenario, we want to supplement this operational scenario with some information about the causal structure of the experiment, a guess of what is happening to the system in the process of preparing and measuring it. We will call this extra information an ontological model. In the spirit of the toy theory, an ontological model will be composed of a space $\Lambda$ containing all possible ontic states $\lambda$ for the system. Each preparation will induce a state of knowledge about the system, i.e., an epistemic state $\{\mu(\lambda \mid P)\}_{P \in \mathcal{P}, \lambda \in \Lambda}$, and each measurement is associated to a response function mapping ontic states to probabilities associated to each outcome, $\{\xi(k \mid M, \lambda)\}_{k \in K, M \in \mathcal{M}, \lambda \in \Lambda}$. Finally, we want that this ontological model explains all statistics in the operational scenario,

$$
\begin{equation*}
p(k \mid M, P)=\int_{\lambda \in \Lambda} \xi(k \mid M, \lambda) \mu(\lambda \mid P) d \lambda, \quad \forall k \in K, M \in \mathcal{M}, P \in \mathcal{P} . \tag{4.17}
\end{equation*}
$$

However, if there is information about preparations and measurement outcomes that cannot be captured by our experiment, there is no reason to include them in the ontological model. This is motivated by Leibniz's quotation at the beginning of this chapter: if two
elements of a theory are indistinguishable, an explanation for that theory that needs to distinguish them is not the best explanation possible. What this means for our ontological model is that whenever two preparation procedures $P, P^{\prime}$ are equivalent, the epistemic states induced by them should be equal, and similar for measurement outcomes. This assumption is what we call noncontextuality:

$$
\begin{gather*}
P \sim P^{\prime} \Rightarrow \mu(\lambda \mid P)=\mu\left(\lambda \mid P^{\prime}\right), \quad \forall \lambda \in \Lambda, P, P^{\prime} \in \mathcal{P}  \tag{4.18}\\
{[k \mid M] \sim\left[k^{\prime} \mid M^{\prime}\right] \Rightarrow \xi(k \mid M, \lambda)=\xi\left(k^{\prime} \mid M^{\prime}, \lambda\right), \quad \forall \lambda \in \Lambda k, k^{\prime} \in K, M, M^{\prime} \in \mathcal{M} .} \tag{4.19}
\end{gather*}
$$

It is the incompatibility with such an assumption that constitutes proof of contextuality. A theory is, therefore, contextual whenever it contains a prepare-and-measure scenario incompatible with the assumption of noncontextuality.

The reason why the toy model does not exhibit contextuality is that it is a noncontextual ontological model itself. It satisfies the principle of noncontextuality by construction, for instance, when the convex combination is defined by mapping three different combinations of epistemic states to the same maximally mixed state.

In fact, the geometric representation of states and effects in theories such as the toy model or the classical bit always ends up in simplices, i.e., generalisations of a triangle in multiple dimensions. Therefore, simplicial theories are often called strictly classical. In this sense, it has been shown that assessing contextuality for an operational theory is equivalent to assessing whether its geometric representation in the real vector space admits of a simplex embedding, i.e., a linear mapping from its states and effects into states and effects of a simplicial theory.

Formalising this concept, let $\Theta=\left(\mathrm{St}_{\mathbb{R}^{m}}, \mathrm{Eff}_{\mathbb{R}^{m}}\right.$, Sys $)$ be the geometric representation of a quotiented operational theory in a real vector space. We say that $\Theta$ admits of a simplex embedding if there exists $\Delta_{d}$ a simplex in a (not necessarily the same) real vector space $\mathbb{R}^{n}$, and linear maps $\iota, \kappa: \mathbb{R}^{m} \rightarrow \mathbb{R}^{n}$ such that

- $\iota\left(\mathrm{St}_{\mathbb{R}^{m}}\right) \subseteq \Delta_{d} ;$
- $\kappa\left(\operatorname{Eff}_{\mathbb{R}^{m}}\right) \subseteq \Delta_{d}^{*}$;
- The inner products are preserved by $\iota$ and $\kappa$.

This means that the quotiented operational theory $\Theta$ is a subtheory of the simplicial one, and therefore its statistics can be simulated by a strictly classical theory. This shows the strength of contextuality as a notion of non-classicality: every scenario that admits of a noncontextual ontological model is classically explainable in the sense that the whole setup can be simulated by a strictly classical quotiented theory. Quantum theory is not such a theory: the sets of states and effects of a qubit, and even small subsets of it, do not admit of a simplex embedding.

To prove it, consider the following six preparations:

$$
\begin{gather*}
\sigma_{0}=\left(\begin{array}{cc}
1 & 0 \\
0 & 0
\end{array}\right) ; \quad \sigma_{1}=\left(\begin{array}{cc}
0 & 0 \\
0 & 1
\end{array}\right) ; \quad \sigma_{2}=\frac{1}{4}\left(\begin{array}{cc}
1 & \sqrt{3} \\
\sqrt{3} & 3
\end{array}\right) ;  \tag{4.20}\\
\sigma_{3}=\frac{1}{4}\left(\begin{array}{cc}
3 & -\sqrt{3} \\
-\sqrt{3} & 1
\end{array}\right) ; \quad \sigma_{4}=\frac{1}{4}\left(\begin{array}{cc}
1 & -\sqrt{3} \\
-\sqrt{3} & 3
\end{array}\right) ; \quad \sigma_{5}=\frac{1}{4}\left(\begin{array}{cc}
3 & \sqrt{3} \\
\sqrt{3} & 1
\end{array}\right), \tag{4.21}
\end{gather*}
$$

with measurement outcomes being represented by the same matrices. Among the statistics for this experiment, we have that

$$
\begin{gather*}
p(1 \mid 0)=p(0 \mid 1)=\operatorname{Tr}\left\{\left(\begin{array}{cc}
1 & 0 \\
0 & 0
\end{array}\right)\left(\begin{array}{cc}
0 & 0 \\
0 & 1
\end{array}\right)\right\}=0 ;  \tag{4.22}\\
p(3 \mid 2)=p(2 \mid 3)=\operatorname{Tr}\left\{\frac{1}{4}\left(\begin{array}{cc}
1 & \sqrt{3} \\
\sqrt{3} & 3
\end{array}\right) \frac{1}{4}\left(\begin{array}{cc}
3 & -\sqrt{3} \\
-\sqrt{3} & 1
\end{array}\right)\right\}=0 ;  \tag{4.23}\\
p(5 \mid 4)=p(4 \mid 5)=\operatorname{Tr}\left\{\frac{1}{4}\left(\begin{array}{cc}
1 & -\sqrt{3} \\
-\sqrt{3} & 3
\end{array}\right) \frac{1}{4}\left(\begin{array}{cc}
3 & \sqrt{3} \\
\sqrt{3} & 1
\end{array}\right)\right\}=0 \tag{4.24}
\end{gather*}
$$

and also that

$$
\begin{align*}
\frac{1}{2} \mathbb{1} & =\frac{1}{2}\left(\sigma_{0}+\sigma_{1}\right)  \tag{4.25}\\
& =\frac{1}{2}\left(\sigma_{2}+\sigma_{3}\right)  \tag{4.26}\\
& =\frac{1}{2}\left(\sigma_{4}+\sigma_{5}\right)  \tag{4.27}\\
& =\frac{1}{3}\left(\sigma_{0}+\sigma_{2}+\sigma 4\right)  \tag{4.28}\\
& =\frac{1}{3}\left(\sigma_{1}+\sigma_{3}+\sigma_{5}\right) \tag{4.29}
\end{align*}
$$

and the same for the respective measurement outcomes.
Equations 4.25 will impose constraints on how the $\mu(\lambda \mid P)$ from the ontological model relate to each other, as well as the $\xi(k \mid \lambda)$. The statistics will impose constraints on how the epistemic states and response functions relate to each other, resulting in a long but simple system of equations that will only admit the trivial solution: all $\mu(\lambda \mid P)$ and $\xi(k \mid \lambda)$ must be null for all values of $\lambda$, which means that there is no ontological model compatible with this scenario.

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## Chapter 5

## Resource Theories

The guiding philosophy in the pragmatic tradition is that understanding a phenomenon means being able to make use of it. Physical phenomena are studied in order to better leverage certain resources.
-B. Coecke, T. Fritz, R. W. Spekkens [1]

### 5.1 Mathematical framework for resource theories

The concept of resource is borrowed from the economical concept of scarcity. A certain state of things is more or less valuable according to the easiness with which it can be extracted, obtained, or implemented. It is convenient thus to consider a theory that can keep track of how valuable a given state or experimental protocol is under certain physical restrictions. Structurally, a resource theory is a commutative ordered monoid. This can be defined as follows:

Definition 5.1.1 (Commutative ordered monoids) Let $\mathcal{A}$ be a set of resources equipped with a binary operation + (representing the situation in which one holds two resources simultaneously), a distinguish element e (a free resource), called identity, and an ordering relation $\geq$ (telling how valuable a resource is compared to another). Then $\mathcal{A}$ is said to be a commutative ordered monoid if, for any $a, a^{\prime}, a^{\prime \prime} \in \mathcal{A}$, we have

- if $a \geq a^{\prime}$ and $a^{\prime} \geq a^{\prime \prime}$, then $a \geq a^{\prime \prime}$;
- if $a \geq a^{\prime}$ and $a^{\prime} \geq a$, then $a=a^{\prime}$;
- $a+\left(a^{\prime}+a^{\prime \prime}\right)=\left(a+a^{\prime}\right)+a^{\prime \prime}$ and $a+a^{\prime}=a^{\prime}+a$;
- $a+e=a$;
- if $a \geq a^{\prime}$, then $a+a^{\prime \prime} \geq a^{\prime}+a^{\prime \prime}$.

In physical terms, one should read the symbol $\geq$ as "is convertible to". Whenever $\rho \geq \sigma$, it means that, in the set of operations and descriptions allowed by the restriction imposed over the experimental scenario, there will be ways of transforming $\rho$ into $\sigma$. If, otherwise, $\rho \nsupseteq \sigma$, it must be read as " $\rho$ is not convertible to $\sigma$ ", meaning that there is no way of performing this transformation with the knowledge one has access to in the experimental scenario. Notice that not all resource theories need to satisfy the second property, i.e., there can be resource theories in which $(a \geq b) \wedge(b \geq a) \nRightarrow a=b$. We call such structures preordered monoids, and they represent situations in which resources can be converted into one another albeit being different.

It is thus convenient to define a resource theory in terms of free operations and free states:

Definition 5.1.2 (Resource theory) Let $(\operatorname{Sys}(\Theta), \operatorname{St}(\Theta), \operatorname{Transf}(\Theta), \operatorname{Eff}(\Theta), \operatorname{Out}(\Theta))$ be a quotiented operational theory. Let $\mathcal{F} \subseteq \operatorname{St}(\Theta)$ be a subset of states that are not resourceful, and $\mathcal{O}(A \rightarrow B) \subseteq \operatorname{Transf}(A \rightarrow B)$, for all $A, B \in \operatorname{Sys}(\Theta)$ be a subset of transformations that cannot create resources. Then, the tuple $\mathcal{R}=(\mathcal{F}, \mathcal{O})$ is a resource theory if

- $\mathrm{I} \in \mathcal{O}(A), \forall A \in \operatorname{Sys}(\Theta)$, where I is the trivial transformation;
- $\mathrm{T} \in \mathcal{O}(A \rightarrow B)$ and $\mathrm{T}^{\prime} \in \mathcal{O}(B \rightarrow C) \Longrightarrow \mathrm{T}^{\prime} \circ \mathrm{T} \in \mathcal{O}(A \rightarrow C), \forall A, B, C \in \operatorname{Sys}(\Theta)$.

The set $\mathcal{F}$ is called the set of free states, while the set $\mathcal{O}$ is called the set of free operations. The demands for constructing a resource theory are perfectly reasonable: to do nothing is always a free operation, and a sequence of free operations must be free as well. A corollary of this definition is often highlighted due to its interpretational convenience:

Definition 5.1.3 (Golden rule of resource theories) Let $\mathcal{R}=(\mathcal{F}, \mathcal{O})$ be a resource theory for the operational theory $\Theta$. If $\mathrm{T} \in \mathcal{O}(A \rightarrow B)$ and $\rho \in \mathcal{F}(A)$, then $\mathrm{T} \circ \rho \in \mathcal{F}(B)$.

The interpretation is as simple as it seems: performing a free operation over a free state will necessarily lead to a free state. In other words, free operations cannot convert free states into resource states.

Resource theories are always constructed operationally, taking into account what are the central phenomena to be studied to define the set of free operations or of free states (usually, one starts by defining just one of the sets, and the other is obtained from the Golden Rule). The quantum resource theory of entanglement, for example, starts from the assumption that for bipartite scenarios, operations performed locally over each of the parties and classical communication between them are always allowed. The set of free states, as a consequence, is restricted to separable states.

One of the most useful features of a resource theory is the possibility of identifying monotones associated with the resource property. Monotones are functions capable of witnessing or quantifying the convertibility between states and are mathematically described as homomorphisms between the resource theory $\mathcal{A}$ and $\mathbb{R}_{\geq 0}$, where $\mathbb{R}_{\geq 0}$ is the set of nonnegative real numbers.

Definition 5.1.4 (Homomorphism) Let $\mathcal{A}$ and $\mathcal{A}^{\prime}$ be commutative ordered monoids. An ordered map $f: \mathcal{A} \rightarrow \mathcal{A}^{\prime}$ is an homomorphism if, for every $a, a^{\prime} \in \mathcal{A}$, we have

- $a \geq a^{\prime} \Rightarrow f(a) \geq f\left(a^{\prime}\right)$;
- $f\left(a+a^{\prime}\right)=f(a)+f\left(a^{\prime}\right)$;
- $f(0)=0$.

Since $\mathbb{R}_{\geq 0}$ is also a commutative ordered monoid with respect to addition, we want to search for a functional $f$ capable of quantifying the convertibility between elements of $\mathcal{A}$ with real numbers.

Catalytic convertibility is a common concept in many resource theories and is borrowed from the concept of catalysis in Chemistry. Mathematically, a resource theory equipped with catalytic convertibility can be defined by a non-cancelative commutative ordered monoid.

Definition 5.1.5 (Non-cancelative commutative ordered monoid) Let $x, y, z \in \mathcal{A}$ be elements of a commutative ordered monoid. $\mathcal{A}$ is said to be non-cancelative if

$$
\begin{equation*}
x+z \geq y+z \nRightarrow x \geq y . \tag{5.1}
\end{equation*}
$$

In resource-theoretic terms, it means that $x$ is not convertible in $y$ by itself, but in the presence of $z$, this process is allowed. The following example, due to Fritz, illustrates the idea: the conversion of wood+nails to table is not allowed, but the conversion wood+nails+hammer to table+hammer is possible. The state $z$ is called the catalyst of this conversion. A resource theory that is non-cancellative can be turned into a cancellative one by redefining its ordering relation, such that for any $x, y, z \in \mathcal{A}$,

$$
\begin{equation*}
x+z \geq y+z \Longrightarrow x \succeq y \tag{5.2}
\end{equation*}
$$

This relation can be read as " $x$ is catalytic convertible into $y$ ", and a resource theory which is cancellative becomes an abelian ordered group ${ }^{1}$. Resource theories of this type are the ones that allow for borrowing resources, since the concept of a debt resource $-x$ is included in an abelian ordered group.

### 5.2 Example: Local Operations and Shared Randomness (LOSR)

We are interested in constructing a resource theory to quantify the non-classicality of common-cause scenarios. Common-cause scenarios consist of two parties, Alice and Bob,

[^3]sitting in separate labs, who have no direct way to influence one another. They might, however, have some systems in their lab which could have been interacting with one another at some point in the past when Alice and Bob met up with one another. It is these systems that are the common cause that can lead, for example, to correlations between what they observe in their labs. With this in mind, given that we are trying to understand nonclassicality, it is natural to divide the things that Alice and Bob can do into free and nonfree by saying that the transformations they can do freely are those that rely only on a classical common cause - that is, some shared randomness - and the things that they can do non-freely are those that rely on a quantum common cause - that is, some shared entangled state.

Consider the specific example of a resource state: a bipartite quantum state rho shared by Alice and Bob


In this chapter, we will read diagrams from bottom to top for a change. Single wires will always represent classical systems, i.e., sets of random variables $\mathbb{A}, \mathbb{X}$, etc, while double wires will represent quantum systems (Hilbert spaces).

In this resource theory, we demand that they can freely convert $\sigma$ into any other bipartite state $\rho$ by performing local operations, i.e., complete-positive trace-preserving (CPTP) maps $\mathcal{E}^{A}, \mathcal{E}^{B}$ on their shares of the system, and by sharing some source of classical randomness $\{p(i)\}_{i \in I}$. These processes take the form


It is easy to show that these operations will satisfy the transitivity and reflexibility of the order relation $\geq$, i.e., $(\rho \geq \sigma) \wedge(\sigma \geq \chi) \Longrightarrow \rho \geq \xi$ and $\rho \geq \rho$. For the first one, it suffices to show that sequentially composing two of the above processes is again a free operation. The fact that the trivial process $\mathbb{1} \otimes \mathbb{1}$ belongs to the set of free operations proves the second property.

In particular, the type of states that can be created freely are separable states. That is because if you take maps $\mathcal{E}^{A}: \star \rightarrow \mathcal{H}_{A}$ and $\mathcal{E}^{B}: \star \rightarrow \mathcal{H}_{B}$, these are simply preparation
procedures of quantum states, and

has the form of a free operation. Also, discarding is always a free operation. These two facts put together mean that any resource can be converted to a separable state, since one can always freely discard whatever resource is available and then freely create a separable state.

Notice however that this is one particular example of a resource. We can think of more general cases, such as bipartite stochastic maps. In our resource theory, we might want to say that they are free when they admit of a quantum common-cause explanation, in the spirit of a quantum Bell scenario, i.e.,


We can then ask ourselves what it means to perform local operations and have shared randomness in this scenario. These will be local stochastic maps acting on the inputs $\mathbb{X}$ and $\mathbb{Y}$ and outputs $\mathbb{A}$ and $\mathbb{B}$, such that


It turns out that for this set of free operations, free resources are the classical-common cause Bell scenarios, i.e.,

So by changing the notion of what is a free resource, we can use LOSR operations to quantify when something does not admit of a Bell classical common-cause explanation! In fact, this framework can be applied to virtually any process with classical, quantum, or even more general inputs/outputs. Take for instance the case with quantum Einstein-Podolsky-Rosen scenarios. They consist of a common cause for both Alice and Bob, with the difference that now Bob always receives a quantum system and never performs any measurement over it. The relevant objects in this scenario are called assemblages: sets of subnormalised quantum states labeled by Alices inputs and outputs,

$$
\begin{equation*}
\Sigma_{\mathbb{A} \mid \mathbb{X}}:=\left\{\sigma_{a \mid x}\right\}_{a \in \mathbb{A}, x \in \mathbb{X}} \tag{5.9}
\end{equation*}
$$

If we consider the resources to be quantumly realisable assemblages, i.e., each element of the assemblage has the form

then the LOSR operations will have the form

i.e., some local operations on the inputs and outputs of Alice and quantum channels on Bob's state, all conditioned to a shared probability distribution. It is possible to verify that the free resources, in this case, are the classical common-cause assemblages, i.e.,


We see therefore how this type-independent resource theory of LOSR allows us to quantify a myriad of interesting non-classicality scenarios without having to build a whole
resource theory from scratch. One only needs to specify what processes in the theory are the resourceful ones, have a well-established notion of local operations and shared randomness for them, and identify when are the resources freely achievable. This framework can explore other common-cause scenarios beyond quantum theory - one just has to add postquantum system types and specify how LOSR processes will look for the sorts of scenarios under investigation.

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From Probabilities to Quantum Theory


Probabilistic Logic

Quantum Theory
Replace Prob. in [0, I] by
"Amplitudes" in $\mathbb{C}$
Intend of Bits $\{0,1\}$ or Prot. Bits $\{p|0\rangle+(1-p)|1\rangle \mid$ P $[0,1\}\}$ Have Quantum bits $\{a|0\rangle+b|1\rangle \mid a, b \in \mathbb{C}\}$ anbits Example: $|0\rangle+|1\rangle,|0\rangle-|1\rangle, \frac{1}{3} i|0\rangle+e^{i \frac{2 \pi}{3}}|1\rangle$

$$
\text { Note: } 107:=\left(\begin{array}{l}
1 \\
0 \\
0
\end{array}\right)
$$

$$
\|>:=\binom{0}{1}
$$

Other way to think of it:
Go from $\{0,1\}$ to $\mathbb{C}^{\{0,1\}}:=\{f:\{0,1\} \rightarrow \mathbb{C}\}$
For In ultiple bits:
Go from $\{0,1\}^{n}$ to $\mathbb{C}^{\{0,1\}^{n}} \cong \mathbb{C}^{2^{n}}$

Specitiying a Classical state in $\{0,1\}^{n}$
we need $n$ bits: oIl $(n=3)$
To specify a $Q$. State in $\mathbb{C}^{2^{n}}$ need $2^{n}$ numbers

$$
\begin{aligned}
& a_{0}|000\rangle+a_{1}|001\rangle+a_{2}|010\rangle+a_{3}|011\rangle+a_{4}|100\rangle \\
& +\ldots+a_{8}|111\rangle \quad\left(n=3,2^{n}=8\right)
\end{aligned}
$$

Normalisation
Just like How Prob. Distributions are normalised:
Eon $\sum_{\lambda} P_{\lambda}|\lambda\rangle \Rightarrow \sum_{\lambda} P_{\lambda}=1$
So $\mathbb{Q}$. States have normalisation:
Eon $\quad \vec{v}, \vec{n} \in \mathbb{C}^{k} \quad \vec{v}=\left(v_{1}, v_{2}, \ldots, v_{k}\right)$

$$
\stackrel{\rightharpoonup}{w}=\left(w_{1}, w_{2}, \ldots, w_{k}\right)
$$

$\langle\vec{v}, \vec{w}\rangle=\sum_{j} \bar{v}_{j} w_{j}$ where $\bar{v}_{j}$ is Complex Conjugate
Inner Product

$$
(\overline{a+b i})=a-b i
$$

$$
\overline{(a+b i)}\left(a+b_{i}\right)=a^{2}+a b i-b a i-b^{2} i^{2}=a^{2}+b^{2}
$$

norm: $|z|:=\sqrt{z z} \quad\left|a+b_{i}\right|=\sqrt{a^{2}+b^{2}}$

$$
\begin{aligned}
& \langle\vec{v}, \vec{v}\rangle=\sum_{j} \bar{v}_{j} v_{j}=\sum_{j}\left|v_{j}\right|^{2} \\
& \|\vec{v}\|=\sqrt{\langle\vec{v}, \vec{v}\rangle}=\sqrt{\sum_{j}\left|v_{j}\right|^{2}}
\end{aligned}
$$

$\stackrel{\rightharpoonup}{v} \in \mathbb{C}^{k}$ normalised when $\|\vec{v}\|=1$

Transformations


Quantum: 2 Rules

1. Linear $\quad f: \mathbb{C}^{2^{n}} \rightarrow \mathbb{C}^{2^{n}}$

$$
\begin{aligned}
f(\vec{v}+\vec{w}) & =f(\vec{v})+f(\vec{w}) \\
f(\lambda \vec{v}) & =\lambda f(\vec{v}) f
\end{aligned}
$$

Hence: Matrices $M$ of size $2^{n} \times 2^{n}$
2. Preserve Normalisation:

$$
\|\vec{v}\|=1 \Rightarrow\|f(\vec{v})\|=1
$$

$\Longrightarrow$ Matrices are Unitary

$$
\begin{aligned}
\langle M \vec{v}, M \vec{w}\rangle & =\langle\vec{v}, \vec{w}\rangle \\
O R \quad M M^{+} & =I D \quad\left(M^{+}\right)_{i j}=\overline{M_{j i}} \\
M M^{+} & =I D
\end{aligned}
$$

Examples

$$
I O=\left(\begin{array}{ll}
1 & 0 \\
0 & 1
\end{array}\right) \quad N O T=\left(\begin{array}{ll}
0 & 1 \\
1 & 0
\end{array}\right)
$$

$$
S=\left(\begin{array}{c}
1 \\
0 \\
0
\end{array}\right) \quad\left(\begin{array}{c}
\text { Because } S^{T}=\left(\begin{array}{cc}
1 & 0 \\
0
\end{array}\right) \text { and } \\
S S^{t}=\left(\begin{array}{cc}
1 & 0 \\
0 & i .-i
\end{array}\right)=\left(\begin{array}{ll}
1 & 0 \\
0 & 1
\end{array}\right)
\end{array}\right.
$$

$H=\frac{1}{\sqrt{2}}\left(\begin{array}{cc}1 & 1 \\ 1 & -1\end{array}\right) \quad H^{2}=I D \quad$ Hadamard

$$
\text { SNOT }=\left(\begin{array}{llll}
1 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{array}\right)
$$

NON -Examples
$\left(\begin{array}{ll}1 & 0 \\ 0 & 0\end{array}\right)$ Because $\binom{1}{00}^{2}=\left(\begin{array}{ll}1 & 0 \\ 00\end{array}\right) \neq\left(\begin{array}{ll}1 & 0 \\ 0 & 1\end{array}\right)$
The operation $\begin{aligned} & |0\rangle \mapsto|0\rangle \\ & |1\rangle \mapsto|0\rangle\end{aligned}$ is $\binom{11}{00}$
Note $\binom{1}{0} 1->=\frac{1}{\sqrt{2}}\binom{11}{00}\binom{1}{-1}=\binom{0}{0}$
So $\|\mapsto\|=1, B_{n} t\left\|\left.\binom{11}{00} \right\rvert\, \rightarrow\right\|=0$
$Q$ Circuits


Tensor Product Let $V$ and $W$ be $v$.spaces $w /$ Bases $\left\{\mid v_{i}>\right\}_{i=1}^{n}$ and $\left\{\left|w_{j}\right\rangle\right\}_{j=1}^{m}$
Then $V \otimes W$ is $v . s p a c e$ w/ Basis $\left\{\left|v_{i}\right\rangle \theta\left|w_{j}\right\rangle\right\}_{i, j=1}^{m, m}$ $\operatorname{dim} V=n \operatorname{dim} W=m$

$$
\Rightarrow \operatorname{dim}(V \oplus W)=n \cdot m
$$

$$
\begin{array}{lr}
V=W=\mathbb{C}^{2} & \\
\left|v_{0}\right\rangle=\left|w_{0}\right\rangle=|0\rangle & \text { Then Basis of } \sigma^{2} \otimes \sigma^{2} \text { is } \\
\left|V_{1}\right\rangle=\left|w_{1}\right\rangle=|1\rangle & \{|0\rangle \theta|0\rangle,|1\rangle|1|,|1\rangle \theta|0\rangle,|1\rangle|1\rangle\}
\end{array}
$$

SPecial Notation: 100$\rangle:=|0\rangle \otimes|0\rangle$
If $A: \mathbb{C}^{k} \rightarrow \mathbb{C}^{k}$ are Matrices, Then

$$
E x:\left(\begin{array}{ll}
1 & 0 \\
0 & 1
\end{array}\right) \otimes\left(\begin{array}{ll}
0 & 1 \\
10
\end{array}\right)=\left(\begin{array}{ll}
0 & 0 \\
0 & 0 \\
0 & 0 \\
0 & 0 \\
0 & 0 \\
0
\end{array}\right)
$$

Lemma: If $u_{1}, u_{2}$ unitary, Then $u_{1}$ on also unitary
So $W=$ means $\begin{aligned} & u: \mathbb{C}^{2} \rightarrow \mathbb{C}^{2} \\ & V: \mathbb{C}^{2} \theta \sigma^{2} \rightarrow \mathbb{C}^{2} \theta \mathbb{C}^{2}\end{aligned}$ and

$$
\begin{aligned}
W & =V \cdot(u \otimes I D) \\
\text { NOTE: } \quad-\sqrt{u_{1}}-\sqrt{u_{2}}- & =\sqrt{u_{2}}-\sqrt{u_{1}} \\
\left(I 00 u_{2}\right) \cdot\left(u_{1} \otimes I D\right) & =\left(u_{1} \otimes I D\right) \cdot\left(I D \otimes u_{2}\right)
\end{aligned}
$$

$$
\begin{aligned}
& B: \mathbb{C}^{l} \rightarrow \mathbb{C}^{l} \quad A \oplus B: \mathbb{C}^{k} \otimes \mathbb{C}^{l} \rightarrow \mathbb{C}^{k} \otimes \mathbb{a}^{l} \text { via } \\
& (A \otimes B)|v\rangle \otimes|w\rangle=(A|v\rangle) \otimes(B|w\rangle) \\
& \text { If } A=\left(\begin{array}{ccc}
a_{1} & & \\
\vdots & a_{k<} \\
a_{k 1} & \ddots & \vdots \\
a_{k k}
\end{array}\right) \text {. Then } A Q B=\left(\begin{array}{cccc}
a_{1}, B & \cdots & a_{k}, B \\
\vdots & \ddots & \ddots & \vdots \\
a_{k}, B & \cdots & a_{k k}
\end{array}\right)
\end{aligned}
$$

States\& Effects
-(u)- has 1 input \& I output
States have 0 inputs $\Psi-\in \mathbb{T}^{2}$
(They Just "are")

$$
\begin{equation*}
=\text { New state } \tag{四}
\end{equation*}
$$

Tensor Prod: : (4) $\otimes\langle\Phi|-=|4|$
So

$$
\begin{aligned}
\langle\phi|-u_{1}- & =\left(u_{1} \theta u_{2}\right)(|\psi\rangle \otimes|\varphi\rangle) \\
\left\langle u_{2}-\left(u_{2}-\right.\right. & =\left(u_{1}|\psi\rangle\right) \otimes\left(u_{2}|p\rangle\right)
\end{aligned}
$$

Effects have 0 outputs

$e: \mathbb{C}^{2} \rightarrow \mathbb{C} \quad$ It takes a state and Produces a number $e|\psi\rangle$.
Ex: for state ip we hove effect - $\varphi$ given by $\langle 4-P\rangle\langle\varphi \mid \psi\rangle \leftarrow$ Inner Product

A Basis $\left|\varphi_{1}\right\rangle, \ldots,\left|\varphi_{k}\right\rangle F_{\text {arms }}$ a measurement $\left.\left\{\phi_{i}\right\rangle\right\}_{i} w /$ Probabilities $P\left(i||\psi\rangle)=\left|\left\langle\varphi_{i} \mid \psi\right\rangle\right|^{2}\right.$ Observing outcome:

Check $\sum_{i} P(i|14\rangle)=1 \quad$ For instance, suppose
Then $\left.\sum_{i} p(i \| \psi\rangle\right)=\sum_{i}|\langle i \mid \psi\rangle|^{2}$

$$
\left|\psi_{i}\right\rangle=|i\rangle=\left(\begin{array}{l}
0 \\
0 \\
0 \\
0 \\
0 \\
0
\end{array}\right)
$$

$$
=\sum_{i}\left|\psi_{i}\right|^{2}=\|\psi\|^{2}=1
$$

After observing outcome i,
state $|\psi\rangle$ has collapsed to $\left|\rho_{i}\right\rangle$

$$
|\psi\rangle \leadsto|\varphi\rangle
$$

Measuring state again will Then always give i

So: Measuring Destroys \&. information
But: measuring is the only way to get information out of the system

Quantum in a nutshell: SCum
States are normalized vectors in complex V. space

AKA

$$
|4\rangle \in \mathbb{C}^{n} \quad \||\psi\rangle \mid \|^{2}=\langle | x,\left.\right|^{2}=1
$$

'Hilbert
Compound systems are Made by Tensor Product

$$
\mathbb{4}, \sqrt{P} \in \mathbb{C}^{2} \Rightarrow \mathbb{\mathbb { F }} \in \mathbb{C}^{2} \otimes \mathbb{C}^{2} \equiv \mathbb{C}^{4}
$$

Unitaries are the allowed transformations

$$
u: \mathbb{U}^{m} \rightarrow \mathbb{C}^{n} \quad u u^{+}=u u^{+}=I D
$$

Measurements are how you get information out of a Q. System
Get possible outcome if on input w/ Prob. $\left|\left\langle y \mid \varphi_{i}\right\rangle\right|^{R}$

Also known as The "Born rule"


Quantum Computation


Hence, outcome is a bitstring following Some prow. Dist.
Goal: Find problems we can "solve" w/
high Prob. using a "small" Q. Circuit
"solve":= we yet some bitting we can
posterocess into an answer

Problem: n-qubit computations are $2^{n} \times 2^{n}$ matrices
$\Rightarrow$ Hard to work with
Solution: Diagrams
Instead of $\left(\begin{array}{lll}1 & 0 \\ 0 & 0 & 1 \\ 0 & 1\end{array}\right)\left(\begin{array}{cccc}1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 10 & 1\end{array}\right)\left(\begin{array}{ll}1 & 0 \\ 0 & 0 \\ 0 & 1\end{array}\right)=\left(\begin{array}{ll}1 & 1 \\ 0 & 1 \\ 10 & 1\end{array}\right)$
Have $-0.2=X$

Tensor Networks

$$
\begin{aligned}
& i[\bar{A}]^{i}\left[(B]^{k}\right. \\
& \left\langle\operatorname{Hin}^{j}\right. \\
& (B A)_{i}^{k}=\sum_{j} A_{i}^{j} B_{j}^{k} \quad(A|\psi\rangle)^{i}=\sum_{1} A_{i}^{i} \psi^{i} \\
& i \Delta A] \quad(A \otimes B)_{j k}^{i l}=A_{i}^{i} B_{k}^{l} \\
& \text { K } \ell
\end{aligned}
$$

More generally:
has matrix elms:

$$
T_{a, b, c}^{k_{j}}=\sum_{j, k} f_{a}^{i} g_{i j}^{x q} h_{b c k}^{j} \longleftarrow \text { tensor network }
$$

String Diagram motTo: only connectivity matters

$Z X$ - Diagrams
a specific type of tensor network
$z$-spiders
$n\{\underbrace{\alpha}_{i}\}_{m}:=\underbrace{1000\rangle}_{m} \underbrace{\langle 00.01}_{n}+e^{i \alpha}|111\rangle\langle 11-1|$

$$
\leftrightarrow 2^{n}\{\underbrace{\left(\begin{array}{llll}
1 & & & \\
& & \ddots & \\
& & 0 & \\
& & e^{i \alpha}
\end{array}\right)}_{2^{m}}
$$

i.e $\quad \int \begin{aligned} & |0 \cdots 0\rangle \longmapsto|0 \cdots 0\rangle \\ & |1 \cdots|\rangle \longmapsto i \alpha|1 . .1\rangle\end{aligned} \quad$ Note: in general not unitary
i.e a tensor $Z[\alpha]_{i,-i n}^{i \cdots j_{m}}:= \begin{cases}1 & \text { if } i_{i}=\cdots=i_{n}=j=\cdots j_{m}=0 \\ e^{i \alpha} \text { if } i_{1}=\cdots=i_{n}=j=\cdots j_{m}=1 \\ 0 & \text { else }\end{cases}$ $X$-Spiders Recall $1+\rangle=\frac{1}{\sqrt{2}}\binom{1}{1} \quad 1->=\frac{1}{\sqrt{2}}\binom{1}{-1}$

$$
\cdots:=1++\cdots+\lambda++\ldots+1+e^{i \alpha} \mid-\ldots-x-\ldots-1
$$

i.e same as $z$-spider, butwl $\{1+\rangle, 1->\}$ basis instead of $\{|0\rangle,|-\rangle\}$ basis

$$
\begin{aligned}
& \text { Ex Examples } \\
& \sigma^{\text {Noinputs }}=|0\rangle+|1\rangle=\binom{1}{0}+\binom{0}{1}=\binom{1}{1}=\sqrt{2}|+\rangle \\
& \sigma^{\pi}=|0\rangle+e^{i \pi}|1\rangle=\binom{1}{0}-\binom{0}{1}=\binom{1}{-1}=\sqrt{2}|-\rangle \\
& \alpha=1+\rangle+|-\rangle=\frac{1}{\sqrt{2}}\left[\binom{1}{1}+\binom{1}{-1}\right]=\frac{1}{\sqrt{2}}\binom{2}{0}=\sqrt{2}|0\rangle \\
& \frac{\pi}{\theta}=|+\rangle-|-\rangle=\cdots=\sqrt{2}|1\rangle \\
& -\alpha=\left\lvert\, 0 \times\langle 0|+e^{i \alpha}|1\rangle\langle 1|=\left(\begin{array}{ll}
1 & 0 \\
0 & 0
\end{array}\right)+e^{i \alpha}\left(\begin{array}{ll}
0 & 0 \\
0 & 1
\end{array}\right)=\left(\begin{array}{cc}
1 & 0 \\
0 & e^{i \alpha}
\end{array}\right)\right.
\end{aligned}
$$

is Unitary!
means "proportional to"
-\{ acts as a "Copy" for $z$-basis states $\{10\rangle, 11\rangle\}$


$$
\text { composing } Z X \text {-Diagrams }
$$

Vertical = Tensor Product


Horizontal = matrix mult

$$
\begin{aligned}
\frac{a}{a b e} & =(I D \otimes \times O R)_{0}(\operatorname{copVOID}) \\
& =\sum_{c} Z[0]_{a}^{b c} \times[0]_{c d}^{e}
\end{aligned}
$$

is unitary


This is because spiders are symmetric tensors

$$
\alpha=-\alpha=\beta^{\alpha}
$$

In general: Can treat $Z x$-Diagrams as undirected graphs

Universality
IHm (Ever decomposition) For ar single-qubit unitary $U$, $\exists$ angles $\alpha, \beta, \gamma, \theta$ s.t:


Ex The Hadamard $H=\frac{1}{\sqrt{2}}\left(\begin{array}{cc}1 & 1 \\ 1 & -1\end{array}\right)$

$$
\begin{array}{ll}
H|0\rangle=|t\rangle & H|1\rangle=1-\rangle \\
H|+\rangle=|0\rangle & H|-\rangle=|1\rangle
\end{array}
$$

basis transform between $Z$-basis $\{|0\rangle,|1\rangle\}$ and $x$-basis $\{1+>, 1-\rangle\}$
abbreviation
Tho: Any Linear map from $\mathbb{C}^{2^{n}}$ to $\mathbb{c}^{2^{m}}$ can be written as a $Z X$-Diagram

$$
\begin{aligned}
& Z X \text { - Calculus } \\
& \text { z-spiders } \\
& n\{\underbrace{\alpha}_{i} \overbrace{m}:=\underbrace{1000\rangle}_{m} \underbrace{\langle 00.01}_{n}+e^{i \alpha}|11-1\rangle<11-1 \mid \\
& X-\text { Spiders Recall } 1+\rangle=\frac{1}{\sqrt{2}}\binom{1}{1} 1 \rightarrow=\frac{1}{\sqrt{2}}\binom{1}{-1} \\
& \because:=1++\cdots+>++\ldots+1+e^{i \alpha} \mid-\ldots-x-\cdots-1
\end{aligned}
$$

Tho: Any Linear map from $\mathbb{C}^{2^{n}}$ to $\mathbb{C}^{2^{m}}$ can be written as a $Z X$-Diagram But: Can also reason diagrammatically

2X diagrams have "extreme" OCM.
They are invariant under:

- Stumping spriter-legs:

$$
I_{\alpha} x=\exists_{a}=x^{a} E=\cdots
$$

- Changing Direction


$$
\left(I \otimes \times[\beta]_{2}^{\prime}\right)\left(Z[\alpha]_{2}^{2} \in I\right)=\left(Z[\alpha]_{3}^{1} \otimes I\right)\left(I \otimes I \otimes X[\beta]_{1}^{2}\right)
$$

$\Rightarrow$ they can be treated as undirected graphs (w lists of inputs outputs)
eg.


In addition, we have rewrite rules

$$
\text { We call this the } Z X \text {-calculus }
$$

(0) "WIRE" RuLES

$$
\begin{aligned}
& \text { RULES } \\
& -0-=-\infty=\sigma^{I D} \\
& C_{|00\rangle+|11\rangle}: \alpha=\sigma \quad=0=0
\end{aligned}
$$

(1) SPIDER-Fusion

$$
\begin{aligned}
& \text { (Q): } \alpha+\beta \text { : } \\
& -\alpha \beta=0+\beta \\
& \underset{0}{\alpha-\alpha}=0-0
\end{aligned}
$$

$$
\begin{aligned}
& \text { Recall: } \quad x O R \approx כ \\
& |0\rangle \approx \pi \quad \underset{\pi}{a \pi} \approx|a\rangle \quad a \in\{0,1\} \\
& \sigma_{\pi}^{2}=\sigma_{\pi}^{\pi} 0_{\pi}^{\pi}=a_{0}^{2 \pi}=a \\
& \int_{6 \pi}^{a \pi}=(a+b) \pi=(a \otimes b) \pi \\
& \begin{aligned}
\text { copy } & =-\infty \\
|x\rangle H|x|\rangle & \text { associativity }
\end{aligned}
\end{aligned}
$$

(2) $\pi$-rule:


$$
E x:-0^{\pi} 0^{\alpha}-0^{-\alpha} \quad \pi 0^{\pi} \approx 0^{-\pi}=0^{\pi}
$$

(3) Colour Change:
works because

$$
\begin{aligned}
& H|0\rangle=|+\rangle \\
& -H|1\rangle=|-\rangle
\end{aligned}
$$

Ex:

$$
\begin{aligned}
& -D-_{O}^{\pi} D-=-\pi \\
& H Z H=H X H
\end{aligned}
$$

$$
\sigma-0=0
$$

$$
\pi_{0}^{\pi}=\frac{\pi}{2}
$$

Note: $-D-D-\stackrel{10}{=}-D-0-0-\stackrel{C}{=}-$
So: $-D-D C=$

$$
\begin{aligned}
& \cdots=10 \times 01-11><11=\left(\begin{array}{cc}
1 & 0 \\
0 & -1
\end{array}\right) \text {. } \\
& \frac{\pi}{0}=\left(\begin{array}{ll}
0 & 1 \\
1 & 0
\end{array}\right) \\
& \text { Pauli } \\
& \text { we call this } \\
& \text { the Pauli } \\
& Z:=Z[\pi] \\
& 0_{0}^{\pi}=\pi_{0}^{\pi} \text { NOT date } \\
& \text { Nणाlol }=11\rangle \quad a \pi \quad \pi \quad(a \otimes 1) \pi
\end{aligned}
$$

(4) Strong complementarity


Special cases:

$A l_{\text {so: }}$

Discarding: $n=0 \Rightarrow m: \infty-0 \approx \underset{-0}{-0}$
"Applying a function, XOR Discard
"Applying a function,
Then throwing away the
output, is the same as
$-0=\langle 01+\langle 1|$, so $\langle 0\rangle-0=1$

$$
\text { (1) } 0=1
$$

throwing away the inputs"
bialgefra $m=2, n=2 \Rightarrow 70-\alpha \approx=-\infty$


Rewriting examples
The (Complementarity)

$$
\begin{aligned}
& \stackrel{2 x}{\approx} \rightarrow 0
\end{aligned}
$$

$$
\begin{aligned}
& \stackrel{s c}{\approx} 0-0 \stackrel{c p}{\approx} \rho_{0}^{0} 0 \stackrel{s p}{=}-0 \& \stackrel{w}{=} 0 \\
& \text { (COPY XOR }
\end{aligned}
$$

corrollary:


$$
\begin{aligned}
& \text { CNOT }|a, b\rangle=|a, a \oplus b\rangle \\
& \text { So }-a, ~ \text { id }
\end{aligned}
$$

$S_{v}(-0)^{-1}=-0$

Ex 3CNOT = SWAP

$$
\stackrel{s}{\approx}-900-8) \stackrel{s p}{=}-8\left(8 \stackrel{x}{\approx}-9 x_{0}^{0-}=x\right.
$$

Note: SNOTs are important:
Thin: any n-qubit unitary can be written as a Circuit of CNOT, $Z[\alpha]$ and $X[\alpha]$ gates
So: CNot is the "Only" 2-qubit interaction we need Bell state

- is an example of a Product state

No connection
We also have entangled states
Bell state: $C=100>+11 \Delta$ The "cup"
Suppose $C=\frac{\Psi \mid-}{\infty}$ for some $|\psi\rangle,|\varphi\rangle$

so everything would disconnect

$$
\begin{array}{ll}
\text { Note } & \text { is known as the } \\
(C A P \otimes I D) O(I D \otimes C U P) & =I D
\end{array} \quad \text { Yanking equation }
$$

Teleportation

1. Alice \& bob start with a Shared Bell state

2. Then they may move far a part
3. Alice Picks a Q. State 14$\rangle$ She wants to send robot

A
B
4. Alice performs sNot \& had on her states

5. Then she measures both her qubits in $\{|0\rangle,|1\rangle\}$ basis, getting outcomes $a, b \in\{0,1\}$

6. She communicates $a, 6$ to bob, who performs a $-\frac{6 \pi}{0}$ a correction

7. Now Bob's state is Alice's former state:


Conclusion: using Entanglement
\& classical communication, Alice can send Q.info to Bob Needed! No FTL comm.

No Cloning theorem
We can "clone", ie. copy, Classical states

$$
\left.x-Q^{2}=\frac{x}{x} \right\rvert\,
$$

Is there some process $-\Delta 5$ that clones arbitrary $Q$. states?

Def: A map $\Delta$ is a cloning process when
(1) $\mathbb{\|}-\sqrt{\Delta}==\mathbb{\Psi}$
12) $-\sqrt{\Delta} x=-\sqrt{\Delta})=$


Th m: No Cloning Process exists
Pf: suppose it id. Then $\subset \frac{(3)}{=}(\sqrt{\Delta})$
$\stackrel{(2)}{=} \Delta x \stackrel{\alpha M}{\Delta x}=\triangle$


$$
\Rightarrow-=\rightarrow \mathbb{L}
$$

Completeness

We have 5 types of rewrites:


$$
-0-=-=0
$$

$$
\delta_{\alpha}^{\alpha} \beta=-\alpha^{\alpha+\beta} \quad \hat{\chi}^{\alpha} \beta=\nu^{\alpha \alpha+\beta}
$$

$$
\therefore \square^{\alpha}+=a_{0}^{\alpha}
$$

$$
\alpha=z
$$

Q: How Much can we Prove w/ this?

Def: we say a set of rules is complete When two diagrams representing the same linear map can always be rewritten into each other by the rules
The: The rules above are complete for diagrams with all phases in set
$\left\{0, \pi, \frac{\pi}{2},-\frac{\pi}{2}\right\}$ The (lifford fragment
Th: it is not complete over all Phases, but adding one additional ryle makes it complote

where $\left.\alpha^{\prime}=f_{1}(\alpha, \beta, \gamma)\right\}$ complicated

$$
\left.\begin{array}{l}
\beta^{\prime}=b^{\prime}(\alpha, \beta, \gamma) \\
\gamma^{\prime}=b_{3}(\alpha, \beta, \gamma)
\end{array}\right\} \text { functions }
$$

C not Circuits o P base $F_{\text {re e }} 2 X$ diagrams

CIRCuits MADE Just out of $I=-\frac{1}{-1}$

Prop Any Not circuit is equal to a phase free $2 x$-diagram.


Q: What about the converse?
Tho: (Unitary) phase free $2 x$-dags $\leadsto$ Coot circuits.

$$
\begin{aligned}
& \text { Parities } \\
& \qquad \begin{array}{c}
\text { Coo }|x, y\rangle
\end{array}||x, x \otimes y\rangle \\
& \text { Coot }|x, y\rangle \mapsto\left|f_{1}(x, y), f_{2}(x, y)\right\rangle \quad \text { when }\left\{\begin{array}{l}
f_{1}(x, y)=x \\
f_{2}(x, y)=x \oplus y .
\end{array}\right.
\end{aligned}
$$

Def $A$ function of the form $f\left(x_{1}, \ldots, x_{n}\right)=x_{i}, \ldots \oplus x_{i_{k}}$ is called a parity map.
Def The field $\mathbb{F}_{2}$ has elements $\{0,1\}$ where:

$$
x \cdot y:=x \wedge y \quad x+y=x \oplus y \quad(\text { ie. } x+y \bmod 2)
$$

Sometimes we call some $x \in \mathbb{F}_{2}$ a parity.

$$
\begin{aligned}
\operatorname{par}(\vec{b})=\sum_{i}^{b_{i}} b_{i} \mathbb{F}_{2} & \operatorname{par}(\vec{b})=0 \text { means } \vec{b} \text { has } \\
& \operatorname{por}(\vec{b})=1 \text { meas odd } \# .
\end{aligned}
$$

Parities for subsets of bits:

$$
\left(\begin{array}{llll}
1 & 0 & \mid & 1
\end{array}\right)\left(\begin{array}{l}
b_{1} \\
b_{2} \\
b_{3} \\
b_{4}
\end{array}\right)=b_{1} \oplus b_{3} \oplus b_{4}
$$

Multiple parities at once:

$$
\underbrace{\left(\begin{array}{llll}
1 & 0 & 1 & 1 \\
0 & 1 & 1 & 0 \\
1 & 0 & 0 & 1 \\
0 & 0 & 0 & 1
\end{array}\right)}_{i}\left(\begin{array}{l}
b_{1} \\
b_{2} \\
b_{3} \\
b_{4}
\end{array}\right)=\left(\begin{array}{c}
b_{1} \oplus b_{3} \oplus b_{4} \\
b_{2} \oplus b_{3} \\
b_{1} b_{4} \\
b_{4}
\end{array}\right)
$$

parity matrix.
The The action of a CNoT circuit on basis elements is defined by an invertible parity matrix:

$$
C\left|b_{1}, \ldots, b_{n}\right\rangle=\left|c_{1}, \ldots, c_{n}\right\rangle
$$

where $P\left(\begin{array}{c}b_{1} \\ \vdots \\ b_{n}\end{array}\right)=\left(\begin{array}{c}c_{1} \\ \vdots \\ c_{n}\end{array}\right) . \quad P P^{-1}=I D$
Note: $P$ is $n \times n$, so not exponentially Large

EX:


$$
\begin{aligned}
& C\left|x_{1}, x_{2}, x_{3}\right\rangle=\left\lvert\, \begin{array}{l}
\left.x_{1}, x_{1} \otimes x_{2} \otimes x_{3}, x_{1} \otimes x_{2}\right\rangle \\
\underbrace{\left(\begin{array}{lll}
1 & 0 & 0 \\
1 & 1 & 1 \\
1 & 1 & 0
\end{array}\right)}_{P}\left(\begin{array}{l}
x_{1} \\
x_{2} \\
x_{3}
\end{array}\right)=\left(\begin{array}{l}
x_{1} \\
x_{1} \otimes x_{2} \otimes x_{3} \\
x_{1} \otimes x_{2}
\end{array}\right)
\end{array}\right.
\end{aligned}
$$

Special case: Single CNOT.

$$
\underbrace{-0}_{\text {More astrally: }}|x, y\rangle \mapsto|x, x \not 0 y\rangle \quad \underbrace{\left(\begin{array}{ll}
1 & 0 \\
1 & 1
\end{array}\right)}_{p}\binom{x}{y}=\binom{x}{x 0 y}
$$

More generally:


$$
\begin{aligned}
& E^{j} A=A^{\prime} \\
& \uparrow \\
& \text { mon } j=\text { row } j \text { +row } i \\
& A E^{j i}=A^{\prime} \\
& \uparrow \\
& \quad \text { col } j=\operatorname{col} i+\operatorname{cod} j
\end{aligned}
$$

Suppose $P E^{i \cdot j \ldots \ldots E^{i \cdot j k}=I \text {, }, ~(1)}$


Algorithm: CNOT-SyNTH:

* Start ul Parity matrix P, empty cire. C.
* Do Gauss -Jordan reduction of columns of $P$.
- Whenever an elem. col operation $E^{j l}$ is applied, append $C_{n o} T^{j i}$ to $C$.
* $C$ now implements $P$.

$$
\begin{aligned}
& \text { Parity maps in } z x \\
& 0-\infty=\underbrace{}_{x_{\theta y}}
\end{aligned}
$$

More general parity maps:

$$
\begin{aligned}
& P=\left(\begin{array}{lll}
1 & 1 & 0 \\
1 & 0 & 1 \\
0 & 0 & 1
\end{array}\right)\left(\begin{array}{l}
x_{1} \\
x_{2} \\
x_{3}
\end{array}\right)=\left(\begin{array}{c}
x_{1} \otimes x_{2} \\
x_{1} \otimes x_{3} \\
x_{3}
\end{array}\right) \\
& x_{1}=0
\end{aligned}
$$

Ex


LEM 4.2.3


Ex

$$
\left(\begin{array}{lll}
1 & 1 & 0 \\
1 & 0 & 1 \\
0 & 0 & 1
\end{array}\right) \xrightarrow{c_{2}=c_{2}+c_{1}}\left(\begin{array}{lll}
1 & 0 & 0 \\
1 & 1 & 1 \\
0 & 0 & 1
\end{array}\right) \xrightarrow{c_{3}=c_{3}+c_{2}}\left(\begin{array}{lll}
1 & 0 & 0 \\
1 & 1 & 0 \\
0 & 0 & 1
\end{array}\right) \xrightarrow{c_{1}=c_{1}+c_{2}}\left(\begin{array}{lll}
1 & 0 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{array}\right)
$$



Def A spider is called
*on input spider if it is con. to on input * an output spider. output

* an interior spider otherwise.

Def A phes-fer $2 x$-diagram is in parity normal form

- every 2 spider is conn. to exactly 1 input
- every X ... -.. output
- no wires between spiders of the same type - no parallel wires


Thin: A $z x$-Diagram in parity NF can berewritten into a CNOT circuit

Missing piece: From arbitrary phase-ftee ZX-diagram to Parity NF
Completeness

We have 5 types of rewrites:

Q: How much can we Prove w/ this?

Def: we say a set of rules is complete When two diagrams representing the same line dr map can always be rewritten into each other by the rules
Tho: The rules above are complete for diagrams with all phases in set $\left\{0, \pi, \frac{\pi}{2},-\frac{\pi}{2}\right\}$ The (lifford fragment

Tho: it is not complete over all Phases, but adding one additional ryle makes it complete

$$
\left.\begin{array}{rl}
\alpha \beta^{\beta} \gamma & \approx \alpha^{\alpha} \beta^{\prime} \gamma^{\prime} \\
\text { where } \alpha^{\prime} & =f_{1}(\alpha, \beta, \gamma) \\
\beta^{\prime} & =b_{3}(\alpha, \beta, \gamma) \\
\gamma^{\prime} & =\beta_{3}(\alpha, \beta, \gamma)
\end{array}\right\} \text { complicated }
$$

Clifford diagrams and circuits
Def A $2 x$-diagram is Clifford when it is mode of Clifford spiders

$$
\mathbb{O}_{k \pi / 2} \sum_{k=0}
$$

Def Clifford circuits are circuits made from:

Ex Some common Clifford gates:

$$
\begin{aligned}
& Z=-0-\frac{0}{\pi} \frac{-0}{2} / 2 \\
& S^{+}=-\frac{0}{-0 / 2}=-\frac{0}{\pi / 2}-\frac{0}{2} \\
& x=-\infty=-\infty-\pi-\pi \\
& \sqrt{x}=-\frac{\pi}{\pi / 2}=-\square-0-a- \\
& \mathrm{Cl}=\frac{\mathrm{C}}{\mathrm{C}}={ }_{-0}^{-0}
\end{aligned}
$$

Ex Some non-Clifford gates:

$$
\begin{aligned}
& T=-\frac{0}{1 / 4} \quad\left(S=T^{2}\right) \\
& T O F=\frac{F}{\sigma}=\cdots \\
& |x, y, z\rangle \mapsto|x, y,(x, y) \propto z\rangle
\end{aligned}
$$

Def $A$ Clifford state is a state $|\varphi\rangle=C|0 \ldots .0\rangle$ For a Clifford circuit $C$.

Q: Why care about Cliffords?

* Contains useful states, e.g.

* Quantum error correction
* Eff. classics simulation (som)
* Rich rewrite theory (now!)
* (litfordst any other single qubit gate
= approximately universal dateset

Notation:


Hadamond edge
Def $A \quad Z$ diagram is graph like if:

1. all spiders are 2 spider
2. all edges btw spiers are tladamed edges
3. no parallel. edges or self-loops
4. every input/ouput is connected to a spider.

Peso: Every $2 x$-diagram is equal to a graph-like one.
 o use $\square \square-)^{2 L}$ - to con cal extra $H^{\prime}$ s.
2. Use (Sf) to elim $m n-H$ edge: $: O_{\alpha}^{-} \alpha_{B}^{\prime}=O_{\alpha+\beta}^{O}$
3. For parallel $H$-edges:

For self-lops: $Q_{\alpha} \stackrel{s p}{=} R^{\alpha}$
4. Use — $\stackrel{i d}{=}$-0 if necessary.

$$
\begin{aligned}
& 0.9-=-69^{l} . \\
& : 0-0-\frac{10}{=} 30-0-1=0-0
\end{aligned}
$$

Ex $-0=0-$

Def A graph-like diagram is called a graph state if: $\left\{\begin{array}{l}\text { - no inputs } \\ \text { - no } \\ \text { - } \\ \text { no } \\ \text { physios }\end{array}\right.$ spiles $\quad$ interior $=$ only connected to

Ex:


Some states are almost graph states, l.g.

$$
\begin{aligned}
& C=\underset{b-\infty}{\infty}
\end{aligned}
$$

Def A graph state with local Clifford (GSLC) is a state of the form $\left(U_{1} \otimes . \otimes U_{n}\right) \mid G>$ for some graph state $|G\rangle$ and 1 -quit Clifford gates $U_{i}$.

THy Any Clifford state is equal to a GSLC.
Weill need some new tools to prove this!

First, note that for GSLCs, the graph con be deceiving!

Local complementation

Def Let $G=(V, E)$ be a graph ad $u \in V$.
The Local complementation of $G$ about $u$ is a new graph $G * u=\left(V, E^{\prime}\right)$ where

$$
\forall v, \omega \in \underset{\substack{\uparrow \\ \text { neighbourhood }}}{N_{G}(u) .} \quad(v, \omega) \in E^{\prime} \Leftrightarrow(v, \omega) \notin E .
$$

Ex

2.


$$
G+u=\prod_{v}^{v_{1}} \prod_{v_{3}}^{u}
$$

3. 


$P_{\text {RoD }}$


Graphically:

$$
\begin{aligned}
& \text { caph. }
\end{aligned}
$$

Q: Why is this the same as local comp?
$A$ : Because o on $\approx b \sigma$


Proving $\angle C$ :
in Exercises saw base case:


Prove nest by induction

So why do we care a bout $\angle c^{2}$ ?
it helps us remove spiders!
PRop


Pf



$$
\begin{aligned}
& \xrightarrow[0]{\pi / 2 \pi / 2}=0-0-\square
\end{aligned}
$$

Pivoting.
Consider the (sc) rule:

$$
\therefore 0 . \quad=-2 \div
$$

add some context:


$$
\begin{aligned}
& \infty=\mathbb{O} \\
& >=\mathbb{x}
\end{aligned}
$$

always deletes spiders!
Now, (ec) both sides to slim $X$ spiders:

deletes 2 cad. phase-free spiders.

Generalisation:

Pivot rule:

deletes 2 adjacent Pauli spiders.

Q: What if they shave neighbours?


General pivot:


Or, as I prefer to think about it, use the simpler rule, but allow boundary sp's to match twice:

$\frac{\text { Rewrite }}{1}$ strategy (Clifford-simp.)

1. convert to a graph-like diagram
2. apply $L c^{\prime}+P^{\prime}$ as long a possible.
3. remove isolated $\{0, \pi\}$-spider.

Prop 1 Clifford-simp terminates for cry $2 x$-diag and removes all interior:
$* \pm \frac{\pi}{2}$ spiders

* pars of connected $\{0, \pi \xi$-spiders

Recall: $m=\left[i n\right.$ is a $2^{n} \times 2^{m}$ matrix. $m=n=0 \Rightarrow 2^{0} \times 2^{0}=\mid \times 1$ matrix (a scalar)

Def A scalar $2 x$-diagram is a $2 x$-diag w/ no inputs and no outputs.

Cor (to Prop) There exists a terminating rewrite strategy that removes all spiders from a scalar Clifford diagram.

Pf First apply Clifford-sinp. Then the only spiders left are 0 and $\stackrel{\pi}{O}$. For these:

$$
0 \rightarrow 2 \cdot: \begin{gathered}
\vdots \\
\vdots
\end{gathered} \quad 0 \rightarrow 0
$$

Q: What's left?
A: the scalar factor
$D_{0} \rightarrow \lambda_{i} D_{1} \rightarrow \lambda_{2} \cdot D_{2} \rightarrow \ldots \rightarrow \lambda_{n} \vdots ; \vdots=\lambda_{n} \in \mathbb{C}$

Application 1 (Efficient) strong simulation of
Clifford circuits.
Problem For a circuit $C$, compute:
(*) $\left.\operatorname{Prob}\left(x_{1} \ldots x_{n}| | \psi\right\rangle\right)$ where $|\psi\rangle=C|0 \ldots 0\rangle$.
... or more generally, for $k \leqslant n$, compute the marginal probability:
(**) $\left.\left.\operatorname{Prob}\left(x_{1} \cdot x_{k}| | \psi\right\rangle\right)=\sum_{x_{k+1} \cdot x_{n}}^{1} \operatorname{Prob}\left(x_{1} \cdot x_{n}| | \psi\right\rangle\right)$
Born rule

$$
\begin{aligned}
& \left.\operatorname{Prob}\left(x_{1} . x_{n}| | \psi_{\rangle}\right):=\left|\left\langle x_{1}, x_{n}\right| C\right| 0.0\right\rangle\left.\right|^{2}
\end{aligned}
$$

$$
\begin{aligned}
& \left(\sum_{x} \underset{\rightarrow}{x \pi} \underset{x}{x \pi}=2 \cdot \sum_{x}|x\rangle\langle x|=2 I\right) \\
& \text { (2x-diagrom for }(x+m)
\end{aligned}
$$

Algorithm: For a circuit $C$ :

1. Let $D$ be the $2 X$-diagram of $\left.\operatorname{Prob}\left(x_{1} \ldots x_{k}|C| 0 \ldots\right\rangle\right)$.
2. Apply (lifford-simp to get a number.

Prop 1 Algorithm terminates in polynomial time lin the \# of quibits \& gates of C5.
Pf Assume basic diagram operations (add/remose spidr/wive) take constant time. If $C$ has $n$ quits \& $k$ gates, $D$ has at most $S:=2 \cdot(2 n+2 k)=4(n+k)$ spiders. Than:

- Each rewrite removes 1 or 2 spiders, so there one at most $4(n+k)$ steps.
- Each step adds/removes at most $(4(n+k))^{2}$ edges, So Algorithm 1 performs $O\left((n+k)^{3}\right)$ basic graph operations.

Rem this is not optimal. A good choice of $L C^{\prime}$ and $P^{\prime}$ steps actually tales $O\left(n^{2} k\right)$ time. $\Rightarrow$ if $k \gg n$, this makes a big difference!

IDEA:

2. Apply LC'a P' from left-to-right:

$\Rightarrow$ each step involves at most $O(n)$ spiders (hence $O\left(n_{n}{ }^{2}\right)$ wires)

Def $A$ graph-like $Z X$-diagram is in $A P$-form if all interior spiders:

- have phase $\in 0, \pi$
- are only connected to boundary spiders

Application: completeness (see Exercises)

$A=\{\vec{x} \mid A \vec{x}=\vec{b}\}$ is an affine subspace of $\mathbb{F}_{2}^{n}$.
$:=a$ solution to a set of linear eqns, egg:

$$
A=\left\{\begin{array}{l}
x_{1} \\
x_{2} \\
x_{4}
\end{array} \left\lvert\, \begin{array}{l}
x_{1} \oplus x_{2}=0 \\
x_{2} \oplus x_{3} \oplus x_{4}=1
\end{array}\right.\right\} \Longleftrightarrow \underbrace{x_{3}}_{-}
$$

$\phi$ is a phase polynomial

$$
\begin{aligned}
& \left\langle\left. x t^{\pi / 2}=e^{i \pi\left(\frac{1}{2} x\right)} \right\rvert\, x\right\rangle \\
& \langle x)^{-\pi / 2}=e^{i \pi\left(-\frac{1}{2} x\right)}|x\rangle
\end{aligned}
$$

- phase polynomial

$$
\begin{aligned}
& 13
\end{aligned}
$$

$$
\begin{aligned}
& U|\vec{x}\rangle=e^{i \pi \phi}|\vec{x}\rangle \text { where } \phi=\frac{1}{2} x_{1}-\frac{i}{2} x_{3}+x_{2}+x_{3} x_{4}
\end{aligned}
$$

Def A $2 X$－diagram is in graph－state wal local Clifford（GSLC）form if it has
＊all 2 spiders，fused as much as possible
＊ale spidss are connected to exactly 1 input（possibly via a l－qubit Clifford unitary）
$A P \rightarrow$ GSLC：


$$
\begin{aligned}
& 2 \text { cases: } \quad \text { C } \quad \text { CASE 2: }
\end{aligned}
$$

$$
\begin{aligned}
& \text { Clifford diagram } \rightarrow \text { AP- form } \longrightarrow \text { GSLC } \\
& \text { * only internal } \\
& \text { Spidus ane 新: } \\
& \text { spiders. }
\end{aligned}
$$

Application：Efficient synthesis of Clifford circuits．

Algorithm 2 (Clifford re-synthesis)

1. For $a_{n}^{n-q u b i t}$ Clifford circuit $C$, translate to $Z X$-diagram $D$.
2. Compute GSLC form.
3. Write as:

4. infuse $C 2$ gates $=\begin{aligned} & -0- \\ & -1 \\ & -1\end{aligned}$

5. Colour-change:

6. extract parity map cos cots $-\mathrm{O}\left(n^{2}\right)$


Prop Any Clifford circuit can be written wi at most $O\left(n^{2}\right)$ gates!

C NOT + phase Circuits


Q: What happens when we add phase gates?

$$
Z[\alpha]::|x\rangle \mapsto e^{i \alpha \cdot x}|x\rangle
$$



$$
\left.\left.\begin{array}{rl}
\left|x_{1} x_{2} x_{3}\right\rangle & \mapsto
\end{array}\left|x_{1}, x_{2} \oplus x_{3}, x_{3}\right\rangle\right), ~ \stackrel{\mapsto}{i \alpha \cdot\left(x_{2} \oplus x_{3}\right)}\left|x_{1}, x_{2} \otimes x_{3}, x_{3}\right\rangle\right)
$$

Prop Any CNOT+phase circuit describes a unitary of the form:

$$
U::|\vec{x}\rangle \mapsto e^{i \phi(\vec{x})}|L \vec{x}\rangle
$$

phase polynomial parity matrix.
From the example above: $L=\left(\begin{array}{lll}1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1\end{array}\right)$ and

$$
\phi\left(x_{1}, x_{2}, x_{3}\right)=(\alpha+\underset{\text { phase -folding }}{\gamma}) \cdot\left(x_{2} \oplus x_{3}\right)+\beta \cdot\left(x_{1} \oplus x_{2}\right)+\theta \cdot x_{2}
$$

Q: can we re-synthesise a circuit for $(L, \phi)$ ?
For $L$, we have:


To get $\phi$, we need to place 2 -phases on wires labelled:

$$
x_{2} \oplus x_{3}, x_{1} \oplus x_{2} \text {, and } x_{2}
$$

Only $x_{1} \oplus x_{2}$ is missing, so lets (temporarily) create it:


Phase polynomials, graphically (aka. phase gadgets)

Ex

$1-\operatorname{legged}:$
—( $::|x\rangle \longmapsto\left\{\begin{array}{ll}1 & \text { if } x=0 \\ e^{i \alpha} & \text { if } x=1\end{array}=e^{i \alpha \cdot x}\right.$
$K$-legged phase gadget:

$$
\sqrt{2}^{(k-1)} \frac{\partial-\alpha)}{\vdots} \quad\left(x_{1} \cdot x_{k}>\mapsto e^{i \alpha \cdot\left(x_{1} \oplus \cdot \oplus x_{k}\right)}\right.
$$

In a diagonal unitary:


Q: What happens when there is phase folding?


A: Gadget fusion!


Algorithm: CNOT+ phase optimisation.

1. unfuse phases and treat as outputs.
2. Compute PNF of phose-free part.
3. perform gadget fusion (* and other phase-poly reductions!)
?? $\rightarrow$ 4. extract a CNOT + phase circuit.
There are choices for step 4.
Naive approach: "CNOT ladders"


Prop


Pf


Cor


Näive extraction: 1. unfuse a phase gadget 2 replace using Cor 1.
2. reped until no phase gadgets
3. Synthesise CNOT circuit from phace-free diag.

* Lots of wasted C Not gates! e.g.


Vs.


Better extraction

1. write an "extended biadjacency matrix"
2. identify a set of $k$ linearly independent rows
3. reduce each row to a unit vector with column ops.
4. "extract phases" and repeat.


$$
\text { Gadgets }\left\{\left(\begin{array}{ccc}
1 & 1 & 0 \\
0 & 1 & 0 \\
0 & 1 & 1 \\
\hdashline-1 & 0 & -1 \\
0 & 1 & 1 \\
0 & 0 & 1
\end{array}\right) \xrightarrow{c_{2}=c_{2}+c_{1}}\left(\begin{array}{lll}
1 & 0 & 0 \\
0 & 1 & 0 \\
0 & 1 & 1 \\
1 & 1 & 0 \\
0 & 1 & 1 \\
0 & 0 & 1
\end{array}\right) \xrightarrow{c_{2}=c_{2}+c_{3}}\left(\begin{array}{ccc}
1 & 0 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1 \\
\hdashline 1 & 1 & 0 \\
0 & 0 & 1 \\
0 & 1 & 1
\end{array}\right)\right.
$$



High-level gates

We've seen 2 kinds of phase polynomials:
"Multilinear" form, e.g. ${ }_{-0}^{-\pi / 2}::|x, y\rangle \mapsto e^{i \pi \cdot\left(\frac{1}{2} x+x \cdot y\right)}|x, y\rangle$
$\begin{gathered}\text { "XOR"/ form, eng. } \\ \text { Phase -gadget } \\ -0-0 \pi / 2 \\ -0 \pi / 2\end{gathered}:|x, y\rangle \mapsto e^{i \pi \cdot(x \oplus y+y)}$

These two forms are related:

$$
\begin{aligned}
& x \oplus y=x+y-2 x y \quad(x, y \in\{0,1 \xi) \\
& -2 x y=x \oplus y-x-y \\
& \Rightarrow x y=\frac{1}{2}(x+y-x \oplus y)
\end{aligned}
$$

$$
\begin{aligned}
\text { —员 : }:|x y\rangle & \mapsto e^{i \pi \cdot(x y)}|x y\rangle \\
& =e^{i \pi\left(\frac{1}{2} x+\frac{1}{2} y-\frac{1}{2} x \otimes y\right)}|x y\rangle
\end{aligned}
$$



Some gates are easy to write in multilinear form.


$$
\begin{aligned}
C C Z\left|x_{1} x_{2} x_{3}\right\rangle & = \begin{cases}\left|x_{1} x_{2} x_{3}\right\rangle & \text { if } x_{1} x_{2}=0 \\
\left|x_{1} x_{2}\right\rangle \otimes \otimes \underbrace{Z\left|x_{3}\right\rangle}_{(-1)^{x_{3}}\left|x_{3}\right\rangle} \text { if } x_{i} x_{2}=1\end{cases} \\
& =(-1)^{x_{1} x_{2} x_{3}}\left|x_{1} x_{2} x_{3}\right\rangle=e^{i \pi \cdot x_{1} x_{2} x_{3}}\left|x_{1} x_{2} x_{3}\right\rangle
\end{aligned}
$$

$$
\begin{aligned}
x_{1}\left(x_{2} x_{3}\right) & =\frac{1}{2} x_{1}\left(x_{2}+x_{3}-x_{2} \oplus x_{3}\right) \\
& =\frac{1}{2}\left(x_{1} x_{2}+x_{1} x_{3}-x_{1}\left(x_{2} \oplus x_{3}\right)\right) \\
& =\frac{1}{4}\left(x_{1}+x_{2}-x_{1} \oplus x_{2}+x_{1}+x_{3}-x_{1} \oplus x_{3}-x_{1}-x_{2} \oplus x_{3}+x_{1} \oplus x_{2} \oplus x_{3}\right) \\
& =\frac{1}{4}\left(x_{1}+x_{2}+x_{3}-x_{1} \oplus x_{2}-x_{1} \oplus x_{3}-x_{2} \oplus x_{3}+x_{1} \oplus x_{2} \oplus x_{3}\right)
\end{aligned}
$$



Translation of $C C 2$ into xoR form is a special case of discrete Fourier transform.

Prop For any function $\phi: \mathbb{F}_{2}^{n} \rightarrow \mathbb{R}$,

$$
\phi(\vec{x})=-\frac{1}{2^{n-1}} \sum_{\vec{y}}^{1} \tilde{\alpha}_{\vec{y}}(\widetilde{\vec{x} \cdot \vec{y})}
$$

where $\tilde{\alpha}_{\vec{y}}=\frac{-1}{2^{n-1}} \sum_{\vec{z}}(-1)^{\vec{y} \cdot \vec{z}} \cdot \phi(\vec{z})$ are the Fourier coefficients.
In the $c c z$ case, taking the Fourier xform of $\phi(\vec{x})= \begin{cases}1 & \text { if } x_{1} x_{2} x_{3}=1 \\ \varnothing & 0 . w_{2}\end{cases}$ gives: $\left\{\begin{array}{l}\tilde{\alpha}_{100}=\tilde{\alpha}_{010}=\tilde{\alpha}_{001}=\frac{1}{4} \\ \tilde{\alpha}_{10}=\tilde{\alpha}_{101}=\tilde{\alpha}_{011}=-\frac{1}{4} \\ \tilde{\alpha}_{11}=\frac{1}{4} .\end{array}\right.$

This gives a general strategy for synthesising classical oracles $f:\{0,1\}^{n} \rightarrow\{0,1\}$

1. Let $\left.u_{f} \mid \vec{x}, y\right):=|\vec{x}, f(\vec{x}) \oplus y\rangle$

$$
\begin{aligned}
& \bar{\vdots}=\sqrt{u_{F}}=\sqrt{D_{F}} \\
& D_{f}|\vec{x}, y\rangle:=e^{i \pi \cdot \phi}|\vec{x}, y\rangle \text { where } \phi(\vec{x}, y)=f(\vec{x}) \cdot y
\end{aligned}
$$

2. Compute Fourier coeffs of $\phi$.
3. Synthesise $D_{f}$ as CNOT+ Phase circuit.
Path sums

We know now todeal wi cliftonds
\& $w /$ CNOT + Phase
Combination: CNOT + Phase + Hadamard
$z[\alpha]$ has $s:=Z\left[\frac{\pi}{2}\right]$
as special case
$\Rightarrow$ This is a universal dateset
$\Rightarrow$ Do not expect efficient rewriting
One approach: Path sums

ancilla in $H+>$ state
(NOT: $:\left|x_{1}, x_{2}\right\rangle \mapsto\left|x_{1}, x_{1} \otimes x_{2}\right\rangle$
$Z[\alpha] \because|x\rangle \mapsto e^{i \alpha}|x\rangle$ updating phase polynomial

$$
\rightarrow-::|x\rangle \mapsto \underset{Y \leftarrow \text { Path }}{ }(-1)^{x \cdot y}|y\rangle
$$

$\longleftarrow$ Path variable
$E x$ :

$$
\begin{aligned}
& \text { Step: }\left|x_{1}, x_{2}\right\rangle \\
& \stackrel{1}{\Rightarrow}\rangle e^{i \alpha x_{1}\left|x_{1}, x_{2}\right\rangle} \\
& \stackrel{l}{\Longrightarrow}\rangle \sum_{y} e^{i \alpha x_{1}}(-1)^{y \cdot x_{1}}\left|x_{1}, y\right\rangle \\
& \stackrel{3}{\Longrightarrow}>\sum_{y} e^{i \alpha x_{1}+\beta y}(-1)^{y \cdot x_{1}}\left|x_{1}, y\right\rangle \\
& \stackrel{4}{=}>\sum_{y} e^{i \alpha x_{1}+\beta y}(-1)^{y \cdot x_{1}}\left|x_{1}, x_{1}, \oplus y\right\rangle \\
& \stackrel{5}{=}\rangle \sum_{y, z} e^{i \alpha x_{1}+\beta y}(-1)^{y \cdot x_{1}+\left(x_{1}, \theta y\right) \cdot z}\left|x_{1}, z\right\rangle
\end{aligned}
$$

Hadumards create new Paths, "branches", these interfere via Phases

Classical simulation method:
Just sum all the branches
cost: $O\left(n \cdot k \cdot 2^{h}\right)$ \# Hadamards
\#qubits $\pi$ \#gates
Power of Q. computation = Hadamards?
cost: $O\left(n \cdot k \cdot 2^{c}\right)$ \#CNOTs

$$
\underbrace{\alpha}_{0}=\frac{1}{2} \cdots+\frac{1}{2} e^{i \alpha}-0_{0}^{\pi} \Rightarrow \text { get Clifford circuits }
$$

cost: $O\left(n^{2} \cdot k \cdot 2^{t}\right)$ \#non-clittord phases

Pauli Gadgets
Clifford + Phase is a universal family.
Q: Can we move all the non-Clifford phases out?

$H$ gates:

$$
\begin{aligned}
& \rightarrow-\ddot{6}=-\ddot{\sigma}
\end{aligned}
$$

$$
\begin{aligned}
& S \text { gates: }-\frac{\pi}{\pi / 2}=-\ddot{O}-\dot{O}
\end{aligned}
$$

Prop For $\vec{P}=P_{1} \otimes \ldots \otimes P_{n}$ with $P_{i} \in\{I, x, y, z\}$ the map:

$-\left[P_{n}\right]$
is unitary. It is called the Pauli gadget $\vec{P}(\alpha)$.


for Cliff. unitarres $C_{i}$. Since phase gadgets are unitary, so is $\vec{P}(\alpha)$. E
Tum Any Clifford+ phase circuit con be written as:

$$
C=-\underbrace{\substack{\vec{p}_{k}\left(\alpha_{k}\right) \\ \text { Clifford }}}_{\substack{\left.\hat{p_{p}(\alpha)}\right]}}
$$

Pf (Idea). Show Pauli gadgets commute past all Clifford gates,

- Move phases out of $C$, one at a time.

Prop (Pauli gadget fusion.)

$$
\bar{\square}[\vec{P}(\alpha)][\vec{P}(\beta) \vdots=\square
$$

Pf


PRop For Pauli $\vec{P}, \vec{Q}$ if $\vec{P} \vec{Q}=\overrightarrow{Q P}$, then $\vec{P}(\alpha) \vec{Q}(\beta)=\vec{Q}(\beta) \vec{P}(\alpha)$.
PE Exercise/ bock (Hint: it's complementarity!)

Algorithm Pauli "phase folding".

1. Compute Pauli gadget form of a cireint.
2. Commute PG's and combine phases where possible.
3. Merge PG's with Clifford phases no the Clifford part.
4. Repeat until no more reductions.
5. Extract circuit.*

* like with CNotaphase, there are may options.

Measurement-based quantum computing (MBQC)
$:=Q C$ where measurements make up most of the computation.

The "code" of MBQC is a measurement pattern: $=$ measurement choices + dassical control (feed-foruard)

Several models:

- (gate teleportation)
- one-way model *
- hypergraph MBAC
- fault tolerant QC
- lattice surgery
-topological FTQC

One-way model of MBQC (Raussendorf/Briegel 2001)

graph state single-qubit measurements with feed-forwarel


Single-Qubit measure ments:
$X$-meosurement: $\left\{-{ }^{k \pi}\right\}_{k=0,1}$




Similarly, yz-plane measwrements: $\{\overbrace{-\infty}^{\alpha+k \pi}\}_{k=0,1}$



Feed -forward: $\alpha^{t_{3}}=\alpha^{\prime}(a, b, c) \leftarrow f_{n}$ of (earlier)

$$
\begin{aligned}
& \alpha^{\prime}=\alpha(a, b, c) \leftarrow \text { measurement outcomes. } \\
& \beta^{\prime}=\beta^{\prime}(a, b, c) \\
& \text { (a.k.a. signals) }
\end{aligned}
$$

Def A measurement pattern for the one-way model consists of a sequence of instructions:

* $N_{j}:=0{ }^{j}$
* $E_{j k}:=\frac{{ }^{j}-0-}{k}$
* $M_{j}^{\alpha}:=\{\overbrace{-}^{\alpha+\xi \pi}\}_{s_{j} \in 0,1}$

$$
\begin{aligned}
& \text { * } M_{j}^{y z, \alpha}:=\left\{\stackrel{j}{\alpha+s j \pi}^{{ }_{s j \in 0,1}}\right. \\
& +M_{j}^{x 2 \alpha_{:}}=\left\{-_{0}^{-\pi / 2} \text { si }\right\}_{s j e 0,1}
\end{aligned}
$$

$\pm Z_{j}^{b}==_{-}^{b r}, X_{j}^{b}={ }_{-}^{b_{\pi}} \quad$ perform Pauli corrections, where

$$
\text { * } b=b\left(s_{k_{1}}, s_{k_{2}}, \ldots\right)^{\prime}
$$

$$
P:=N_{1} ; N_{2} ; N_{3} ; N_{4} ; E_{12} ; E_{34} ; E_{13} ; E_{24}
$$



$$
Q:=N_{1} ; N_{2} ; N_{3} ; E_{12} ; E_{231} \cdot M_{1}^{\frac{\pi}{4}} ; \overbrace{M_{2}^{\frac{\pi}{2}} ; X_{3}^{s_{2}}}^{\text {fend forward }}
$$



Def $A$ measurement pattern is:

* runnable if all angles / corrections are frs of past measurement outcomes.

$$
\begin{aligned}
& m_{j}^{\alpha} ; \cdots \cdots ; z_{k}^{s_{j}} \\
& z_{k}^{s_{j}} ; \ldots, \ldots ; m_{j}^{\alpha}
\end{aligned}
$$

* deterministic if all choices of measurement outcomes give the same map (up to scalars)

Q: runnable? deterministic $X$

$$
\begin{aligned}
& s_{1}=0 \Rightarrow *=\frac{\pi / 4}{\pi / 2} \\
& s_{1}=1 \Rightarrow *=0=\frac{5 \pi / 4}{\pi / 2} \\
& Q^{\prime}:=N_{1} ; N_{2} ; N_{3} ; E_{n_{2}} ; E_{23} ; M_{1}^{T_{1}} ; X_{2}^{s_{1}} ; Z_{3}^{s_{1}} ; X_{3}^{s_{2}}
\end{aligned}
$$

$$
\begin{aligned}
& Q^{\prime}=Q^{\prime \prime}:=N_{1} ; N_{2} ; N_{3} ; E_{n} ; E_{23} ; m_{1}^{\frac{\pi}{4}} ; M_{2}^{(-1)^{4} \frac{\pi}{2}} ; Z_{3}^{s_{1}} ; x_{3}^{s_{2}}
\end{aligned}
$$

$Q^{\prime \prime}$ : runnable? deterministir?

$$
s_{1}, s_{2} \in 0,1 \Rightarrow * *={ }_{0}^{\pi / 4} 0
$$

Question Can I alway "push" errors forward in time?

Answer: It depends on the graph.

there is no time ordering for quits $\{1,2,3\}$ that works.

Cluster State
(: = graph state shopped
like a square lattice)

Q: how can we classify which graph states "work"?
IDEA: 1. Fix a time -ordering $<:\left\{\begin{array}{l}\text { past }(u):=\{v \mid v<u\} \\ \text { future }(u):=\{v \mid u<v\}\end{array}\right.$
2. push errors from $u$ into future (v) (without massing up part (u))


Equivalently, think about "firing" a spider with $D O==$


The game: for each $u$, find a set $g(u)$ that is: (i) in the future of $u$
(ii) connected to $u$ an odd number of times
(iii) Connected to the past of $u$ an even number of times


Def An open graph is a graph $G$ with a set of inputs $I_{G} \subseteq V_{G}$ and outputs $O_{G} \subseteq V_{G}$.

graph-like $2 X$-diag

open graph $G$ 7

Def $A_{n}$ open graph has generalised flow (glow) if there exists a partial order $\leqslant$ on $V_{G}$ and a function $g: V_{G} \backslash O_{G} \rightarrow P\left(V_{G} I_{6}\right)$ such that $\forall u$ :
(i) $g(u) \subseteq$ future (u)
(ii) $g(u)$ corrects to $u$ an odd \# of times
(iii) $\forall v \in v_{G} D_{G}$ if $v \neq u, v \notin$ future( $u$ ) then $g(u)$ connects to $v$ an even \# of times.
THM (Determinism) For any graph-like $Z X$-diagram $D$ with gflow, there exists a runnable, deterministic pattern $P$ that implements it.
$\Rightarrow$ There are at least 2 ways that a $Z X$-diagram can be "run" on $c$ quantum computer:

1. If it can be transformed into a circuit.

2. If it has glow (hence can be implemented in $M B Q C$ ).


Now: $2 \Rightarrow 1$. (circuit extraction)

Algorithm (circuit Extraction)

1. unfuse gates as mach as possible:

frontier
2. use CNOTs to do row operations until we get an "extractible" spider (= unit-vector row)


$$
{ }_{2}\left(\begin{array}{cccc}
0 & b & c & d \\
1 & 1 & 0 & 0 \\
1 & 1 & 1 & 0 \\
0 & 1 & 1 & 0
\end{array}\right) \xrightarrow{R_{2}=R_{2}+R_{1}}\left(\begin{array}{llll}
a & b & c & d \\
1 & 1 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 1 & 1 & 0
\end{array}\right) \underbrace{}_{\text {extractible }}
$$

3. Repeat 1.2 until nothing is left of the frontier.


Tum If a $Z$-diagram has glow, Circuit ExTRACTion terminates with a quantum circuit.
De Step 1 never adds spiders to the left of the Frontier, so s.t.s. Step 2 always removes a spieler.

Take a maximal non-output $u$, w.r.t $L$. Then $g(u) \leq$ future (u) must be all outputs:

By gflowr, the only node connected on odd \# of times to $g(a)$ is $u$.

$$
g(\omega\}(E|E| O \mid E] \ldots .
$$

If we add all the rows to a single row, than we get:

$$
g^{(u)}\left\{\left(\frac{00100 \cdots 00}{}\right) \leftarrow v \in g(u)\right.
$$

So, doing cots ctrl'ed on a single $v \in g(u)$ to all other $v^{\prime} \in g(u)$ gives:
$\therefore u$ becomes extractible

$$
-\therefore=-\frac{\vdots}{}
$$

Extract a make $u$ an output. The result still has gfiow and there is one Fewer spider left of the frontier.

Quantum error correction works by encoding some logical quits into a space of (more) physical quits.


Q: Why?
A: Because some errors can be detected and/or corrected using quantum measurements without destroying the logical state.
Ex. The GHZ code:

$$
\sqrt{E}[=-6
$$

$$
\begin{aligned}
& |\overline{0}\rangle:=|000\rangle,|\bar{I}\rangle:=|111\rangle
\end{aligned}
$$

more Generrlly: $|\bar{\psi}\rangle:=E|\psi\rangle$.

Suppose I masure 22I:

$$
\begin{aligned}
& \operatorname{Prob}(1||\bar{\psi}\rangle)=\langle\psi| P_{k}|\bar{\psi}\rangle \\
&=4 \theta^{\pi}|\psi\rangle\langle\psi| 0^{\pi} \mid \psi \\
& \approx \Psi \mid \psi \theta^{\pi}=0 \\
& \Rightarrow \operatorname{Prob}(0|\mid \bar{\psi})=1
\end{aligned}
$$

Also:
$\Rightarrow$ measuring $22 I$ does nat disturb $|\bar{\psi}\rangle$.
But


$$
\operatorname{Prob}(0|(X \otimes I \otimes I)| \Psi\rangle)=
$$



$$
\Rightarrow \operatorname{Prcb}(1)(x \otimes I \otimes L)|\Psi\rangle)=1
$$

So a 221 measurement con detect the error $X \otimes L \otimes I$.

THe The GHZ code can detect (and correct) any error in the set $\{X I I, I X I, I I X\}$.
bit-flip errors

Better codes correct more errors (eng. "phase flips" like 2II..., multequbit errors, etc.)

Q: How can I compute with encoded states.

A: FTQC!

SCHEME: LATTICE SURGERY.
IDEA: - Use a grid of quits

- encode 1 quit as a "patch"

- implement operations to:
- prepare $1 Q$ states: $\langle\Psi| E=\langle\Psi$
- "Z-split" patches:

where:


$$
\begin{aligned}
& \text { - "X-split" patches } \\
& \square=-5 \times= \\
& \text { - "Z-mege" }
\end{aligned}
$$



- Pauli measwrements

Encoded computation:



[^0]:    ${ }^{1}$ Notice that in this framework, the existence of a physical system is assumed to be true. Anti-realist interpretations are thus beyond the scope of the framework.

[^1]:    ${ }^{2}$ This comes from a representation theorem attesting that one can map vectors in the dual space of a vector space into the space itself, so the inner product is in fact taken between the states and the representations of the effects in the same vector space.

[^2]:    ${ }^{1}$ A Dirac delta is a probability distribution that assigns nonzero probability to a single point and null probability to all others.

[^3]:    ${ }^{1}$ i.e., a commutative ordered monoid that has, for every $a \in \mathcal{A}$, an element $a^{-1} \in \mathcal{A}$ such that $a a^{-1}=$ $a^{-1} a=e$.

