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Event-based simulation of single-photon beam splitters and Mach-Zehnder interferometers

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Abstract. – We demonstrate that networks of locally connected processing units with a primitive learning capability exhibit a behavior that is usually only attributed to quantum systems. We describe networks that simulate single-photon beam splitter and Mach-Zehnder interferometer experiments on a causal, event-by-event basis and demonstrate that the simulation results are in excellent agreement with quantum theory.

Introduction. – The objective of the research reported in this paper is to demonstrate that locally connected networks of processing units with a primitive learning capability are sufficient to simulate, event by event, the single-photon beam splitter and Mach-Zehnder interferometer experiments of Grangier *et al.* [1]. This is one of the basic experiments in quantum physics [1, 2] that has not been simulated in the event-by-event manner in which the experimental observations are actually recorded [3]. Although quantum theory gives us a recipe to compute the frequency for observing different types of events, it does not describe individual events [2, 4, 5]. Reconciling the mathematical formalism (that does not describe single events) with the experimental fact that each observation yields a definite outcome is often referred to as the quantum measurement paradox. This is a central, fundamental problem in the foundation of quantum theory [5–7].

From a computational viewpoint, quantum theory provides us with a set of rules (algorithms) to compute probability distributions [5, 8, 9]. Therefore, we may wonder what kind of algorithm(s) we need to perform an event-based simulation of the experiments mentioned above without using wave functions. Evidently, this formulation rules out any method based on the solution of the (time-dependent) Schrödinger equation. We have to step outside the framework that quantum theory provides.

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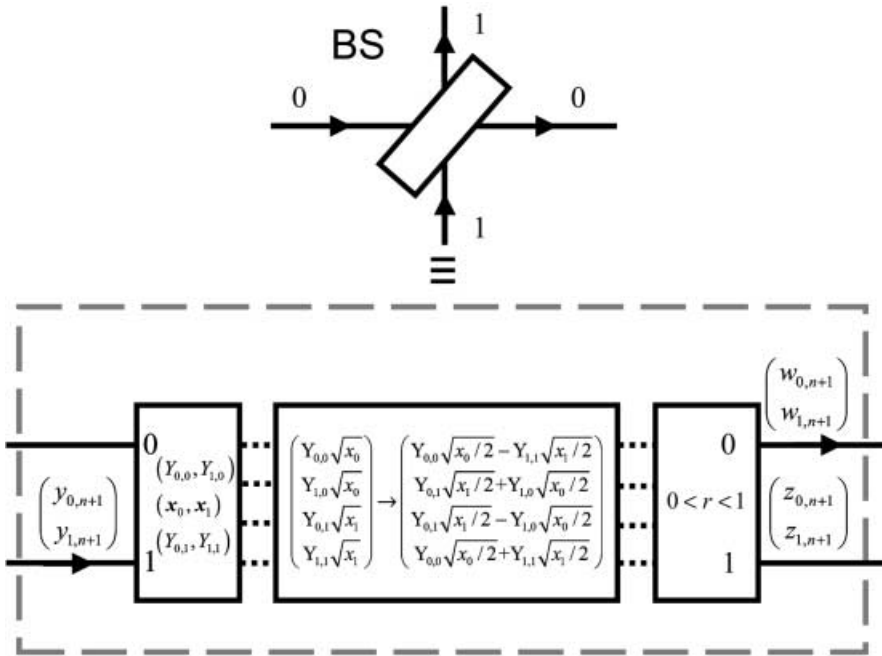


Fig. 1 – Diagram of a SLM that performs an event-by-event simulation of a single-photon beam splitter (BS). The solid lines represent the input and output channels of the BS. The dashed lines indicate the flow of data within the BS.

We formulate the simulation of quantum processes in terms of events, messages, and units that process these events and messages. In terms of the experiments of Grangier *et al.* [1], the photon carries the message (a phase), an event is the arrival of a photon at one of the input ports of a beam splitter, and the beam splitters are the processing units. The essential feature of a processing unit is its ability to learn from the events it processes. We use a standard linear adaptive filter [10] for this purpose. The processing unit sends a message (carried by a photon) through an output port that is selected randomly according to a distribution that is determined by the current state of the processing unit. A processing unit that operates according to this principle will be referred to as a stochastic learning machine (SLM). Here the term *stochastic* does not refer to the learning process but to the method that is used to select the output channel that will carry the outgoing message. The learning process itself is deterministic. Elsewhere we describe processing units that are fully deterministic and are not based on linear adaptive filters, but can be used for exactly the same purpose and also for simulating many-body quantum phenomena [11–14].

By connecting an output channel to the input channel of another SLM we can build networks of SLMs. As the input of a network receives an event, the corresponding message is routed through the network while it is being processed. At any given time during the processing, there is only one input-output connection in the network that is actually carrying a message. The SLMs process the messages in a sequential manner and communicate with each other by message passing. There is no other form of communication between different SLMs. Although networks of SLMs can be viewed as networks that are capable of unsupervised learning, they have little in common with neural networks [15].

Beam splitter. – According to quantum theory [9], the amplitudes (b_0, b_1) of the photons in the output modes 0 and 1 of a beam splitter (see fig. 1) are given by

$$\begin{pmatrix} b_0 \\ b_1 \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} a_0 + ia_1 \\ a_1 + ia_0 \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & i \\ i & 1 \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \end{pmatrix}, \tag{1}$$

where the presence of photons in the input modes 0 or 1 is represented by the amplitudes (a_0, a_1) [1, 16–18]. Here we construct a SLM that acts as a beam splitter, not by calculating the amplitudes according to eq. (1) but by processing individual events (photons). First we describe the various components of the machine. Then we demonstrate that it acts as a beam splitter.

A schematic diagram of the SLM is shown in fig. 1. We label events by a subscript $n \geq 0$. At the $(n + 1)$ -th event, the SLM receives a message on either input channel 0 or 1, never on both channels simultaneously. If the event occurs on channel 0 (1) we call it an event of type 0 (1). Every message is a two-dimensional unit vector $\mathbf{y}_{n+1} = (y_{0,n+1}, y_{1,n+1})$. This message represents the time of flight, that is the length of the optical path of the photon when it travels from source to beam splitter, from beam splitter to beam splitter and the like. In quantum theory, this information is encoded in the phases of the complex numbers a_0 and a_1 .

The first stage (see fig. 1) of the SLM (called front-end) stores the message \mathbf{y}_{n+1} in its internal register $\mathbf{Y}_k = (Y_{0,k}, Y_{1,k})$, where $k = 0$ (1) if the event occurred on channel 0 (1). The front-end also has an internal two-dimensional vector $\mathbf{x} = (x_0, x_1)$ with the additional constraints that $x_i \geq 0$ for $i = 0, 1$ and that $x_0 + x_1 = 1$. Here and in fig. 1, we have omitted the event label n from the internal register \mathbf{Y}_k and internal vector \mathbf{x} to simplify the notation.

After receiving the $(n + 1)$ -th event on input channel $k = 0, 1$ the internal vector is updated according to the rule

$$x_{i,n+1} = \alpha x_{i,n} + (1 - \alpha)\delta_{i,k}, \tag{2}$$

where $0 < \alpha < 1$ is a parameter that controls the learning process and is discussed later. By construction $x_{i,n+1} \geq 0$ for $i = 0, 1$ and $x_{0,n+1} + x_{1,n+1} = 1$. Hence, the update rule eq. (2) preserves the constraints on the internal vector. These constraints are necessary if we want to interpret the $x_{k,n}$ as (an estimate of) the probability for the occurrence of an event of type k .

The solution of eq. (2) reads

$$\mathbf{x}_n = \alpha^n \mathbf{x}_0 + (1 - \alpha) \sum_{i=0}^{n-1} \alpha^{n-1-i} \mathbf{v}_{i+1}, \tag{3}$$

where $\mathbf{x}_n = (x_{0,n}, x_{1,n})$, and \mathbf{x}_0 denotes the initial value of the internal vector. The input events are represented by the vectors $\mathbf{v}_{n+1} = (1, 0)^T$ or $\mathbf{v}_{n+1} = (0, 1)^T$ if the $(n + 1)$ -th event occurred on channel 0 or 1, respectively. Let p_0 ($p_1 = 1 - p_0$) be the probability for generating statistically independent events of type 0 (1) and let $\langle \mathbf{x} \rangle_n = (\mathbf{x}_1 + \dots + \mathbf{x}_n)/n$ denote the Cesaro mean of the sequence $\{\mathbf{x}_1, \dots, \mathbf{x}_n\}$. Then, if $0 < \alpha < 1$, we have $\lim_{n \rightarrow \infty} \langle \mathbf{x} \rangle_n = (p_0, 1 - p_0)^T$. Therefore, the front-end “learns” the probabilities for events 0 and 1 by processing these events in a sequential manner.

The second stage of the SLM in fig. 1 takes as input the values stored in the registers $\mathbf{Y}_0, \mathbf{Y}_1, \mathbf{x}$ and transforms this data according to the rule

$$\frac{1}{\sqrt{2}} \begin{pmatrix} Y_{0,0}\sqrt{x_0} - Y_{1,1}\sqrt{x_1} \\ Y_{0,1}\sqrt{x_1} + Y_{1,0}\sqrt{x_0} \\ Y_{0,1}\sqrt{x_1} - Y_{1,0}\sqrt{x_0} \\ Y_{0,0}\sqrt{x_0} + Y_{1,1}\sqrt{x_1} \end{pmatrix} \longleftarrow \begin{pmatrix} Y_{0,0}\sqrt{x_0} \\ Y_{1,0}\sqrt{x_0} \\ Y_{0,1}\sqrt{x_1} \\ Y_{1,1}\sqrt{x_1} \end{pmatrix}, \tag{4}$$

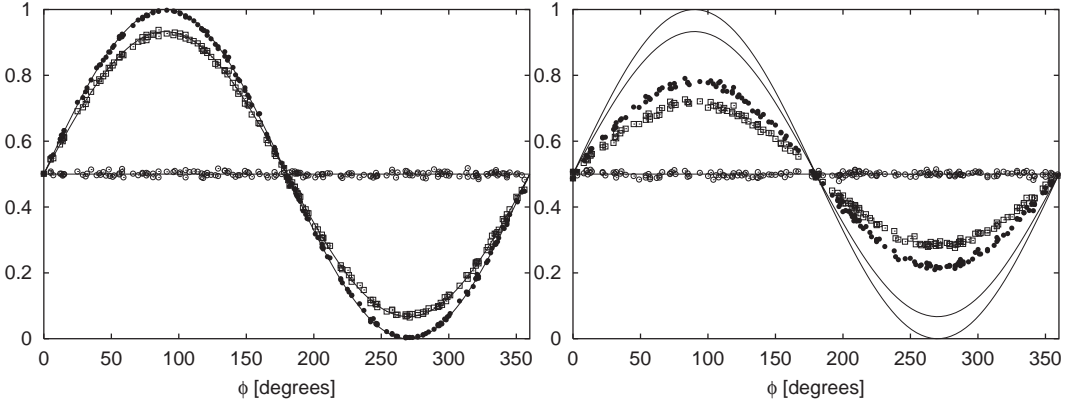


Fig. 2 – Simulation results for the beam splitter shown in fig. 1. Input channel 0 receives $(\cos \psi_0, \sin \psi_0)$ with probability p_0 . Input channel 1 receives $(\cos \psi_1, \sin \psi_1)$ with probability $p_1 = 1 - p_0$. Each data point represents 10000 events ($N_0 + N_1 = 10000$). After each set of 10000 events, two uniform random numbers in the range $[0, 360]$ are used to choose the angles ψ_0 and ψ_1 . Markers give the simulation results for the normalized intensity $N_0/(N_0 + N_1)$ in output channel 0 as a function of $\phi = \psi_0 - \psi_1$. Open circles: $p_0 = 1$; bullets: $p_0 = 0.5$; open squares: $p_0 = 0.25$. Lines represent the results of quantum theory (see eq. (6)). Left: $\alpha = 0.98$; right: $\alpha = 0.25$.

where we have omitted the event label $(n + 1)$ to simplify the notation. Using two complex numbers instead of four real numbers, eq. (4) can also be written as

$$\frac{1}{\sqrt{2}} \begin{pmatrix} Y_{0,0}\sqrt{x_0} - Y_{1,1}\sqrt{x_1} + i(Y_{0,1}\sqrt{x_1} + Y_{1,0}\sqrt{x_0}) \\ Y_{0,1}\sqrt{x_1} - Y_{1,0}\sqrt{x_0} + i(Y_{0,0}\sqrt{x_0} + Y_{1,1}\sqrt{x_1}) \end{pmatrix} \leftarrow \begin{pmatrix} Y_{0,0}\sqrt{x_0} + iY_{1,0}\sqrt{x_0} \\ Y_{0,1}\sqrt{x_1} + iY_{1,1}\sqrt{x_1} \end{pmatrix}. \quad (5)$$

By identifying a_0 with $Y_{0,0}\sqrt{x_0} + iY_{1,0}\sqrt{x_0}$ and a_1 with $Y_{0,1}\sqrt{x_1} + iY_{1,1}\sqrt{x_1}$, it is clear that the transformation eq. (5) plays the role of the matrix-vector multiplication in eq. (1).

The third stage (see fig. 1) of the SLM (called back-end) responds to the input event by sending a message $\mathbf{w}_{n+1} = (Y_{0,0}\sqrt{x_0} - Y_{1,1}\sqrt{x_1}, Y_{0,1}\sqrt{x_1} + Y_{1,0}\sqrt{x_0})/\sqrt{2}$ through output channel 0 if $w_{0,n+1}^2 + w_{1,n+1}^2 < r$, where $0 < r < 1$ is a uniform random number.

Otherwise, since $\mathbf{w}_{n+1}^2 + \mathbf{z}_{n+1}^2 = 1$, the back-end sends the message $\mathbf{z}_{n+1} = (Y_{0,1}\sqrt{x_1} - Y_{1,0}\sqrt{x_0}, Y_{0,0}\sqrt{x_0} + Y_{1,1}\sqrt{x_1})/\sqrt{2}$ through output channel 1.

In fig. 2 we present results of event-based simulations using the SLM depicted in fig. 1. Before the simulation starts we set $\mathbf{x}_0 = (x_{0,0}, x_{1,0}) = (r, 1 - r)$, where r is a uniform random number in the interval $[0, 1]$. We also use uniform random numbers to generate two two-dimensional unit vectors to initialize the registers \mathbf{Y}_0 and \mathbf{Y}_1 . The simulation results of the left and right panel were obtained for $\alpha = 0.98$ and $\alpha = 0.25$, respectively. Input channel 0 receives $(\cos \psi_0, \sin \psi_0)$ with probability p_0 . Input channel 1 receives $(\cos \psi_1, \sin \psi_1)$ with probability $p_1 = 1 - p_0$. In the quantum theory setting, this corresponds to the input amplitudes $a_0 = \sqrt{p_0}e^{i\psi_0}$ and $a_1 = \sqrt{1 - p_0}e^{i\psi_1}$. The data in fig. 2 show the normalized intensity $N_0/(N_0 + N_1)$, where N_0 (N_1) denotes the number of events of type 0 (1), as a function of $\phi = \psi_0 - \psi_1$. Each data point in fig. 2 represents a simulation of 10000 events ($N_0 + N_1 = 10000$). For each set of 10000 events, two uniform random numbers in the range $[0, 360]$ determine the two angles ψ_0 and ψ_1 . From fig. 2, it is clear that the SLM-based beam splitter reproduces the probability distributions

$$|b_0|^2 = \frac{1 + 2\sqrt{p_0(1 - p_0)}\sin(\psi_0 - \psi_1)}{2}, \quad |b_1|^2 = \frac{1 - 2\sqrt{p_0(1 - p_0)}\sin(\psi_0 - \psi_1)}{2}, \quad (6)$$

as obtained from eq. (1).

Role of the control parameter α . – From eq. (2), we see that α controls the speed of learning. Furthermore, it is evident that the difference between a constant input to a SLM and the learned value of its internal variable cannot be smaller than $1 - \alpha$. In other words, α also limits the precision with which the internal variable can represent a sequence of constant input values. On the other hand, the number of events N has to balance the rate at which the SLM can forget a learned input value. The smaller $1 - \alpha$ is, the larger the number of events has to be for the SLM to adapt to changes in the input data. Simulations (results not shown) show that the error decreases as $1/\sqrt{N}$, as expected from elementary statistics.

In general, for a fixed number of events, decreasing α leads to an increase of systematic errors. For instance, if $\alpha = 1/4$, the maximum normalized intensity (at $\phi = 90$) in output channel 0 is about 0.8, as shown in fig. 2 (right). This is easy to understand: In our event-by-event approach, interference is the result of learning by the SLMs. If α is close to unity, the SLM learns slowly but accurately. If α decreases, the SLM can adapt faster to changes in the input data but it also forgets faster. For very small α , \mathbf{x}_n is always close to $(1, 0)^T$ or $(0, 1)^T$ and it becomes impossible to mimic quantum interference. Thus, α determines the visibility [1] of the interference effects. In our simulation approach, interference is the result of learning (which requires some form of memory).

Mach-Zehnder interferometer. – In quantum physics [9], single-photon experiments with one beam splitter provide direct evidence for the particle-like behavior of photons [1, 5]. The wave mechanical character appears when one performs single-particle interference experiments. We now describe a SLM network that displays the same interference patterns as those observed in single-photon Mach-Zehnder interferometer experiments [1]. This also proves that the messages generated by the SLMs preserve the phase information that is essential if the system is to exhibit quantum mechanical behavior.

The schematic layout of the SLM network is shown in fig. 3. Not surprisingly, it is exactly the same as that of a real Mach-Zehnder interferometer. The SLM network described before

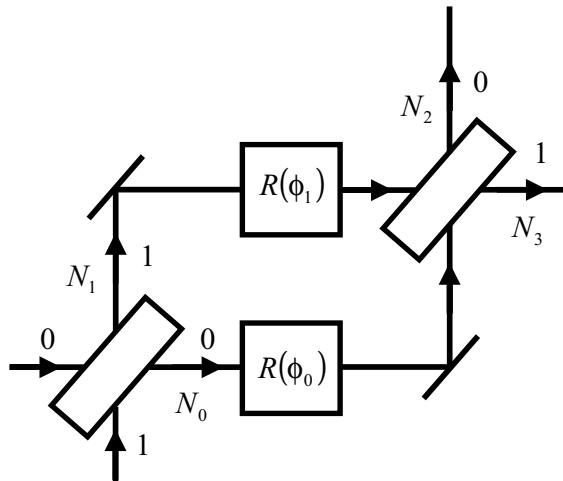


Fig. 3 – Diagram of a SLM network that simulates a single-photon Mach-Zehnder interferometer on an event-by-event basis [14]. The SLM network consists of two BS devices (see fig. 1) and two passive devices $R(\phi_0)$ and $R(\phi_1)$ that perform plane rotations by ϕ_0 and ϕ_1 , respectively. The number of events N_i in channel $i = 0, \dots, 3$ corresponds to the probability of finding a photon on the corresponding arm of the interferometer.

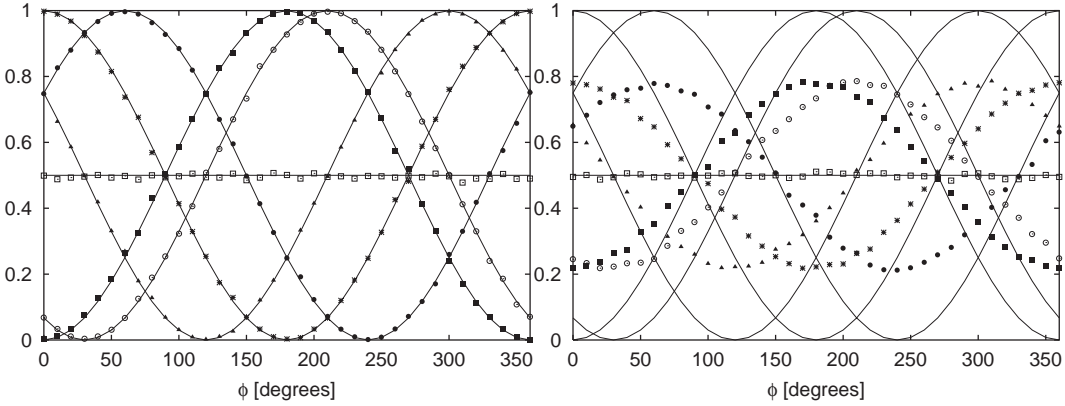


Fig. 4 – Simulation results for the SLM-network shown in fig. 3. Input channel 0 receives $(\cos \psi_0, \sin \psi_0)$ with probability one. A uniform random number in the range $[0, 360]$ is used to choose the angle ψ_0 . Input channel 1 receives no events. Each data point represents 10000 events ($N_0 + N_1 = N_2 + N_3 = 10000$). Initially, the rotation angle $\phi_0 = 0$ and after each set of 10000 events, ϕ_0 is increased by 10° . Markers give the simulation results for the normalized intensities as a function of $\phi = \phi_0 - \phi_1$. Open squares: $N_0/(N_0 + N_1)$; solid squares: $N_2/(N_2 + N_3)$ for $\phi_1 = 0$; open circles: $N_2/(N_2 + N_3)$ for $\phi_1 = 30^\circ$; bullets: $N_2/(N_2 + N_3)$ for $\phi_1 = 240^\circ$; asterisks: $N_3/(N_2 + N_3)$ for $\phi_1 = 0$; solid triangles: $N_3/(N_2 + N_3)$ for $\phi_1 = 300^\circ$. Lines represent the results of quantum theory (see eq. (8)). Left: $\alpha = 0.98$; right: $\alpha = 0.25$, see the section on the role of α .

is used for the two beam splitters in fig. 3. The phase shift is taken care of by a passive device that performs a plane rotation. Clearly, there is a one-to-one mapping from each physical component in the interferometer to a processing unit in the SLM network. The processing units in the SLM network only communicate with each other through the messages (photons) that propagate through the network.

According to quantum theory [9], the amplitudes (b_0, b_1) of the photons in the output modes 0 (N_2) and 1 (N_3) of the Mach-Zehnder interferometer are given by [1, 16–18]

$$\begin{pmatrix} b_0 \\ b_1 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 & i \\ i & 1 \end{pmatrix} \begin{pmatrix} e^{i\phi_0} & 0 \\ 0 & e^{i\phi_1} \end{pmatrix} \begin{pmatrix} 1 & i \\ i & 1 \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \end{pmatrix}. \quad (7)$$

Note that in a quantum mechanical setting it is impossible to simultaneously measure $(N_0/(N_0 + N_1), N_1/(N_0 + N_1))$ and $(N_2/(N_0 + N_1), N_3/(N_0 + N_1))$: Photon detectors operate by absorbing photons. In our event-based simulation there is no such problem.

In fig. 4 we present a few typical simulation results for the Mach-Zehnder interferometer built from SLMs. We assume that input channel 0 receives $(\cos \psi_0, \sin \psi_0)$ with probability one and that input channel 1 receives no events. This corresponds to $(a_0, a_1) = (\cos \psi_0 + i \sin \psi_0, 0)$. We use a uniform random number to determine ψ_0 . The data points are the simulation results for the normalized intensity $N_i/(N_0 + N_1)$ for $i = 0, 2, 3$ as a function of $\phi = \phi_0 - \phi_1$. Lines represent the corresponding results of quantum theory [9]. From fig. 4 (left) it is clear that the event-based processing by the SLM network reproduces the probability distribution

$$|b_0|^2 = \sin^2 \left(\frac{\phi_0 - \phi_1}{2} \right), \quad |b_1|^2 = \cos^2 \left(\frac{\phi_0 - \phi_1}{2} \right), \quad (8)$$

as obtained from eq. (7). The results of fig. 4 (right) for $\alpha = 1/4$ corroborate the analysis of the role of α , presented earlier.

Discussion. – We have proposed a new procedure to construct algorithms that can be used to simulate quantum processes without solving the Schrödinger equation. We have shown that single-particle quantum interference can be simulated on an event-by-event basis using local and causal processes, without using concepts such as wave fields or particle-wave duality. Our results suggest that we may have discovered a procedure to simulate quantum phenomena using causal, local, and event-based processes. There is a one-to-one correspondence between the parts of the processing units and network and the physical parts of the experimental setup. Only simple geometry is used to construct the simulation algorithm. In this sense, the simulation approach we propose satisfies Einstein’s criteria of realism and causality [5]. Our approach is not an extension of quantum theory in any sense and is not a proposal for another interpretation of quantum mechanics. The probability distributions of quantum theory are generated by a particle-like, causal learning process, and not vice versa [7].

REFERENCES

- [1] GRANGIER P., ROGER R. and ASPECT A., *Europhys. Lett.*, **1** (1986) 173.
- [2] TONOMURA A., *The Quantum World Unveiled by Electron Waves* (World Scientific, Singapore) 1998.
- [3] In this paper we disregard limitations of real experiments such as detector efficiency, imperfection of the source, and the like.
- [4] BALLENTINE L. E., *Quantum Mechanics: A Modern Development* (World Scientific, Singapore) 2003.
- [5] HOME D., *Conceptual Foundations of Quantum Physics* (Plenum Press, New York) 1997.
- [6] FEYNMAN R. P., LEIGHTON R. B. and SANDS M., *The Feynman Lectures on Physics*, Vol. **3** (Addison-Wesley, Reading, Mass.) 1996.
- [7] PENROSE R., *The Emperor’s New Mind* (Oxford University Press, Oxford) 1990.
- [8] VAN KAMPEN N. G., *Physica A*, **153** (1988) 97.
- [9] We make a distinction between quantum theory and quantum physics. We use the term *quantum theory* when we refer to the mathematical formalism, *i.e.*, the postulates of quantum mechanics (with or without the wave function collapse postulate) [4] and the rules (algorithms) to compute the wave function. The term *quantum physics* is used for microscopic, experimentally observable phenomena that do not find an explanation within the mathematical framework of classical mechanics.
- [10] HAYKIN S., *Adaptive Filter Theory* (Prentice Hall, New Jersey) 1986.
- [11] DE RAEDT K., DE RAEDT H. and MICHIELSEN K., *Deterministic event-based simulation of quantum interference*, arXiv:quant-ph/0409213.
- [12] DE RAEDT H., DE RAEDT K. and MICHIELSEN K., *New method to simulate quantum interference using deterministic processes and application to event-based simulation of quantum computation*, to be published in *J. Phys. Soc. Jpn.*
- [13] MICHIELSEN K., DE RAEDT K. and DE RAEDT H., *Simulation of Quantum Computation: A deterministic event-based approach*, to be published in *J. Comp. Theor. Nanosci.*
- [14] An interactive program that performs the event-based simulations of a beam splitter, one Mach-Zehnder interferometer, and two chained Mach-Zehnder interferometers can be found at <http://www.compphys.net/dlm>.
- [15] HAYKIN S., *Neural Networks* (Prentice Hall, New Jersey) 1999.
- [16] BAYM G., *Lectures on Quantum Mechanics* (W.A. Benjamin, Reading, Mass.) 1974.
- [17] RARITY J. G. and TAPSTER P. R., *Philos. Trans. R. Soc. London, Ser. A*, **355** (1997) 2267.
- [18] NIELSEN M. and CHUANG I., *Quantum Computation and Quantum Information* (Cambridge University Press, Cambridge) 2000.