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The ambiguity of simplicity in quantum and classical simulation

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ABSTRACT

A system's perceived simplicity depends on whether it is represented classically or quantumly. This is not so surprising, as classical and quantum physics are descriptive frameworks built on different assumptions that capture, emphasize, and express different properties and mechanisms. What is surprising is that, as we demonstrate, simplicity is ambiguous: the *relative* simplicity between two systems can *change sign* when moving between classical and quantum descriptions. Here, we examine the minimum required memory for simulation. We see that the notions of absolute physical simplicity at best form a partial, not a total, order. This suggests that appeals to principles of physical simplicity, via Ockham's Razor or to the "elegance" of competing theories, may be fundamentally subjective. Recent rapid progress in quantum computation and quantum simulation suggest that the ambiguity of simplicity will strongly impact statistical inference and, in particular, model selection.

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We are to admit no more causes of natural things than such as are both true and sufficient to explain their appearances.

[Isaac Newton, 1687
 Philosophiæ Naturalis Principia Mathematica,
 Book III, p. 398 [1]]

1. Introduction

Beyond his theory of gravitation, development of the calculus, and pioneering work in optics, Newton engendered a critical abstract transition that has resonated down through the centuries, guiding and even accelerating science's growth: Physics began to perceive the world as one subject to concise mathematical Laws. Above, Newton suggests that these Laws are not only a correct perception but they are also *simple*. Consequently, one should abandon the Ptolemaic epicycles for Newton's elegant $F = ma$ and $F_g \propto m_1 m_2 / r^2$.

The desire for simplicity in a theory naturally leads us to consider *simplicity as a means for comparing* alternative theories. Here, we compare the parsimony of two descriptions of stochastic processes – one classical and one quantum. Classical versus quantum comparisons have, of late, captured our attention both for reasons of principle and of experiment. *Quantum supremacy* holds that quantum systems behave in ways beyond those that can be efficiently simulated by classical computers [2]. A single cold 2D

Fermi gas supports coexistence of both quantum mechanical states at its core and classical states on its periphery [3,4]. The overriding impression is that now is an interesting time for the foundations of quantum mechanics. The following adds a new phenomenon to these debates on the balance of classical and quantum theories, as concerns the simplicity of their descriptions.

To start, we consider a Nature full of stationary stochastic processes. A theory, then, is a mathematical object capable of yielding a process' probabilities. We can straightforwardly say that one process is more random than another via comparing their temperatures or thermodynamic entropies. But how to compare them in terms of their structural simplicities? We make use of a well developed measure of simplicity in stochastic processes – the statistical complexity – a measure of internal memory [5] or the minimum required memory to simulate a process. It provides a concrete and interpretable answer to the question, which process is structurally simpler? By applying this comparison, we may order all processes from the simplest to the most complicated [6].

With recent progress in quantum computation [7–9], an interesting twist comes about if we add quantum mechanics to our modeling toolbox. Descriptions that act on a quantum substrate offer new and surprising options. For example, it was shown that a quantum mechanical description can lead to a simpler representation [10–14] and even in some cases infinitely simpler [15,16]. Recently, this quantum advantage was verified experimentally [17]. Proceeding with these methods, we discover what is most surprising: the *relative simplicity* of classical and quantum descriptions can change. Specifically, there are stochastic processes, A and B ,

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1 for which classical theory says A is simpler than B , but quan-
 2 tum mechanics says B is simpler than A . What started out as a
 3 neat classical array is upended by a new quantum simplicity or-
 4 der. This means quantizing a simple classical model may not be
 5 as simple as quantizing a more complicated classical model. As a
 6 consequence model selection is complicated by the addition of a
 7 quantum model class.

8 **2. Classical and quantum simplicity**

9 We consider stationary, ergodic processes: each a bi-infinite
 10 sequence of random variables $X_{-\infty:\infty} = \dots X_{-2}X_{-1}X_0X_1X_2\dots$
 11 where each random variable X_t takes some value x_t in a discrete
 12 alphabet set \mathcal{A} and where all probabilities $\Pr(X_t, \dots, X_{t+L})$ are
 13 time-invariant.

14 *How is their degree of randomness quantified?* Information the-
 15 ory [18] measures the uncertainty in a single observation X_0 via
 16 the Shannon entropy: $H[X_0] = -\sum_{x \in \mathcal{A}} \Pr(x) \log_2 \Pr(x)$ and the ir-
 17 reducible uncertainty per observation via the entropy rate [19]:
 18 $h_\mu = \lim_{L \rightarrow \infty} H[X_{0:L}]/L$. If we interpret the left half $X_{-\infty:0} =$
 19 $\dots X_{-2}X_{-1}$ as the “past” and the right half $X_{0:\infty} = X_0X_1X_2\dots$
 20 as the “future”, we see that the entropy rate is the average
 21 uncertainty in the next observable given the entire past: $h_\mu =$
 22 $H[X_0|X_{-\infty:0}]$. Thus, as we take into account past correlations, the
 23 naive uncertainty $H[X_0]$ reduces to h_μ .

24 *How reducible is our uncertainty in the future $X_{0:\infty}$ knowing the past*
 25 $X_{-\infty:0}$? The answer is given by the mutual information between
 26 the past and the future – the excess entropy [20]: $\mathbf{E} = I[X_{-\infty:0} :$
 27 $X_{0:\infty}]$. With h_μ and \mathbf{E} , we measure randomness and predictability,
 28 respectively.

29 Let’s say we want to simulate a given process. To do this we
 30 write a computer code that follows an algorithm and allocate the
 31 memory the algorithm needs. For a given process computational
 32 mechanics [21] identifies the optimal algorithm – the process’
 33 ϵ -machine. This is a unifilar hidden Markov model [22] that uses
 34 only the minimum required memory for simulation. We view a
 35 process’ ϵ -machine as the “theory” of a process in that it speci-
 36 fies a mechanism that exactly simulates a process’ behaviors. In
 37 this way, computational mechanics supplements \mathbf{E} and h_μ with a
 38 measure of structure – the minimum required amount memory to
 39 simulate the given process.

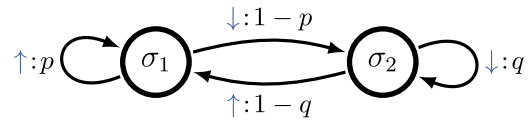
40 The ϵ -machine consists of causal states $\sigma \in \mathcal{S}$ defined by an
 41 equivalence relation \sim that groups histories, say $x_{-\infty:t}$ and $x_{-\infty:t'}$,
 42 that lead to the same future predictions $\Pr(X_{t:\infty}|\cdot)$: $x_{-\infty:t} \sim$
 43 $x_{-\infty:t'} \iff \Pr(X_{t:\infty}|x_{-\infty:t}) = \Pr(X_{t:\infty}|x_{-\infty:t'})$. From this, one con-
 44 cludes that a process’ ϵ -machine is, in a well defined sense, its
 45 simplest predictive theory.

46 Translating this notion of simplicity into a measurable quantity,
 47 we ask: *What is the minimum memory necessary to implement opti-
 48 mal prediction?* The answer is the historical information stored in
 49 the ϵ -machine. Quantitatively, this is the Shannon entropy of the
 50 causal-state stationary distribution $\{\pi_\sigma\}$, the statistical complexity:

51
$$C_\mu = H[\mathcal{S}] = -\sum_{\sigma \in \mathcal{S}} \pi_\sigma \log_2 \pi_\sigma, \quad (1)$$

52 It is well known that the excess entropy is a lower-bound on
 53 this structural measure: $\mathbf{E} \leq C_\mu$. In fact, this relation is only rarely an
 54 equality [23]. And so, while \mathbf{E} quantifies the amount to which
 55 a process is subject to explanation by its ϵ -machine “theory”, this
 56 simplest theory is typically larger, informationally speaking (C_μ),
 57 than the predictability benefit it confers. That said, the ϵ -machine
 58 is the best (simplest) theory. Thus, we use C_μ to define our no-
 59 tion of classical simplicity. It provides an interpretable ordering of
 60 processes – process A is simpler than process B when $C_\mu^A < C_\mu^B$.

61 We may also consider the recently proposed quantum-machine
 62 representation of processes [10–12]. The quantum-machine con-



67 **Fig. 1.** The ϵ -machine for the nearest-neighbor Ising spin chain has two causal states
 68 σ_1 and σ_2 . If the last observed spin x_0 is up ($s_0 = +1$) the current state is σ_1 and
 69 if it’s down ($s_0 = -1$) is σ_2 . If the current state is σ_1 , with probability p the next
 70 spin observed is up and, if the current state is σ_2 , with probability q the next spin
 71 observed is down.

72 sists of a set $\{|\eta_k(L)\rangle\}$ of pure signal states that are in one-to-one
 73 correspondence with the classical causal states $\sigma_k \in \mathcal{S}$. Each signal
 74 state $|\eta_k(L)\rangle$ encodes the set of length- L words that may follow σ_k ,
 75 as well as each corresponding conditional probability. Fixing L , we
 76 construct quantum states:

77
$$|\eta_j(L)\rangle \equiv \sum_{w^L \in \mathcal{A}^L} \sum_{\sigma_k \in \mathcal{S}} \sqrt{\Pr(w^L, \sigma_k | \sigma_j)} |w^L\rangle |\sigma_k\rangle, \quad (2)$$

78 where w^L denotes a length- L word and $\Pr(w^L, \sigma_k | \sigma_j) = \Pr(X_{0:L} =$
 79 $w^L, \mathcal{S}_L = \sigma_k | \mathcal{S}_0 = \sigma_j)$. The resulting Hilbert space is the product
 80 $\mathcal{H}_w \otimes \mathcal{H}_\sigma$. Factor space \mathcal{H}_σ is of size $|\mathcal{S}|$, the number of classical
 81 causal states, with basis elements $|\sigma_k\rangle$. Factor space \mathcal{H}_w is of size
 82 $|\mathcal{A}|^L$, with basis elements $|w^L\rangle = |x_0\rangle \cdots |x_{L-1}\rangle$.

83 The quantum measure of memory is the von Neumann entropy
 84 of the stationary state:

85
$$C_q = -\text{Tr}(\rho \log \rho), \quad (3)$$

86 where $\rho = \sum_i \pi_i |\eta_i\rangle \langle \eta_i|$. This quantum analog of memory is gener-
 87 ally less than the classical: $C_q \leq C_\mu$. Also, due to the Holevo
 88 bound [10,24], $\mathbf{E} \leq C_q$. Though rare in process space, the classi-
 89 cal and quantum informational sizes are equal exactly when both
 90 models are “maximally simple”: $\mathbf{E} = C_q = C_\mu$.

91 **3. Ising chain simplicity**

92 The Ising spin-chain Hamiltonian is given by:

93
$$H = -\sum_{\langle i,j \rangle} (J s_i s_j + b s_i), \quad (4)$$

94 where s_i , the spin at site i , takes values $\{-1, +1\}$, J is the nearest-
 95 neighbor spin coupling constant, and b is the strength of the ext-
 96 ternal magnetic field.

97 In equilibrium the bi-infinite chain of spin random variables de-
 98 fines a stationary stochastic process which has been analyzed using
 99 computational mechanics [25]. Importantly, spins obey a condi-
 100 tional independence: $\Pr(X_{0:\infty} | x_{-\infty:0}) = \Pr(X_{0:\infty} | x_0)$. That is, the
 101 “future” spins (right half) depend not on the entire past (left half)
 102 but only on the most recent spin x_0 . The conclusion (see Supp. Ma-
 103 terials) is that the two-state Markov chain process is minimally
 104 represented by the ϵ -machine in Fig. 1. Using Eq. (1), the statis-
 105 tical complexity is directly calculated as a function of p and q .
 106 Fig. 2 shows that C_μ is a monotonically increasing function of tem-
 107 perature T : $1 - C_\mu \propto T^{-2}$ at high T . In particular, for the three
 108 processes chosen at temperatures $T_\alpha < T_\gamma < T_\delta$, $C_\mu^\alpha < C_\mu^\gamma < C_\mu^\delta$.

109 Consider now the quantum representation of these spin config-
 110 urations. Each causal state is mapped to a pure quantum state that
 111 resides in a spin one-half space [13]:

112
$$|\sigma_1\rangle = \sqrt{p} |\uparrow\rangle + \sqrt{1-p} |\downarrow\rangle$$

 113
$$|\sigma_2\rangle = \sqrt{1-q} |\uparrow\rangle + \sqrt{q} |\downarrow\rangle. \quad (5)$$

114 (We use a more compact spin up/down notation, rather than
 115 the quantum machine notation of Eq. (2).) Intuitively, the quan-
 116 tum overlap accounts for the fact that the conditional predictions

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