Computing the noncomputable

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Abstract

We explore in the framework of Quantum Computation the notion of computability, which holds a central position in Mathematics and Theoretical Computer Science. A quantum algorithm that exploits the quantum adiabatic processes is considered for the Hilbert’s tenth problem, which is equivalent to the Turing halting problem and known to be mathematically noncomputable. Generalised quantum algorithms are also considered for some other mathematical noncomputables in the same and of different noncomputability classes. The key element of all these algorithms is the measurability of both the values of some physical observables and of the quantum-mechanical probability distribution for these values. It is argued that computability, and thus the limits of Mathematics, ought to be determined not solely by Mathematics itself but also by physical principles.

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1 Introduction

This article is a brief introduction to an effort to compute the otherwise mathematical noncomputable. The new ingredients here for this effort are those supplied by physical principles and lied outside the domain of Mathematics where the limits of computability are set.

In the next section we will summarise the important concept of Turing machines and their relationship to the mathematical recursive functions. They will set the scene for the notion of computability as delimited by the Church-Turing hypothesis. We then review the noncomputable results for the Turing halting problem and the equivalent Hilbert’s tenth problem, which is directly accessible for some quantum-mechanical exploration in a later section. We also introduce some other noncomputable problems and the Chaitin’s Ω numbers before moving on to a discussion of Quantum Mechanics and Quantum Computation in Secs. 3 and 4.

In Sec. 3, we emphasise the concept of coherent states and the quantum adiabatic theorem in order to introduce the model of Quantum Computation by adiabatic processes in Sec. 4. With this computation model we then propose a quantum algorithm for Hilbert’s tenth problem in Sec. 5, and discuss some of its finer points in the following Sec. 6. To explore the limits of this quantum algorithm we also consider some generalisation for other noncomputables in Sec. 7.

Before concluding with some remarks in the final section, we present a reformulation of Hilbert’s tenth problem, as inspired by our quantum algorithm, in terms of an infinitely coupled set of differential equations in Sec. 8.

2 Effective Computability and Noncomputability

2.1 Turing machines

The concept of computation is at the heart of Mathematics and hard sciences but it was only made precise by Alan Turing relatively recently, at the beginning of the twentieth century [20]. With the introduction of theoretical and mathematically well-defined machines, Turing was able to capture the essence of computation processes and algorithms. There are a few other models of computation [25] but they all, except possibly the quantum model to be considered later on, can be shown to be equivalent to that given by Turing machines.

Turing machines are equipped with an infinite one-dimensional tape over which a head can move in discrete single steps backward and forward. What symbol is already on the tape directly beneath the head and what state the head is in will specify the new symbol (including the blank symbol) to be written, if at all, over the old one and the new state of
the head and the direction in which the head moves next. There needs to be only a finite number of such symbols, similar to the finiteness of the number of letters in the English alphabet. The number of available states for the head is also finite. Everything about the machine is actually finite despite the appearance of the infinite tape – of which only a finite, but unbounded, portion will ever be used. The infiniteness requirement for the tape is only there to enable the accommodation of unbounded but finite sequences of “words” and “sentences” which can be meaningfully made out from the symbol alphabets.

Initially, a finite portion of an otherwise blank tape is prepared with some suitably encoded input for the Turing machine. Each Turing machine can do only one specific task as completely defined by the transition function \( T \):

\[
\langle \text{new symbol, new state, next movement} \rangle = T(\text{present symbol, present state}).
\] (1)

Once the task is done, the machine goes into a special halt state and stops with the output written out on the tape ready to be inspected. Transition functions are but representations for computer programs implementing algorithms in general; so we will use interchangeably the names Turing machines and programs.

Such deceptively simple machines are so powerful that they can in fact capture the notion of computation by algorithms as we intuitively know it. Turing went further to introduce his universal machines, which are not restricted to do just a single computational task but can be “reprogrammed” to do many different ones. The key conceptual leap that enables Turing universal machines is the recognition that the descriptions of Turing machines, given by the transition functions, are not at all different to and can be encoded in the same way as their very own inputs. The input to a Turing machine and the description of the Turing machine itself can be put together to form a new, single input to yet another machine, a Turing universal machine. This latter machine only needs to emulate the encoded machine, which can be any well-formed Turing machine, and thus acquires the status of being universal. Universal machines are not unique and one can be encoded into the form of an input for another as well.

The ability for a machine to act on its own kind enables us to investigate the capability and limitations of Turing machines using no other instruments other than the machines themselves. It is precisely this self-referential property that Gödel exploited to embed statements about arithmetics in statements of arithmetics in his famous Incompleteness Theorem [29]. The embedding in the Theorem is the same as the encoding of Turing machines into input forms acceptable for universal machines and is achieved by converting the finite description of a Turing machine into a unique non-negative integer which can then be expressed in binary or decimal or any other convenient notation. The conversion is possible as we are only dealing here with machines having a finite number of states, a finite number of symbols in its alphabet, and only a finite number of movements for their heads. Multi-tape Turing machines have been introduced and can simplify certain computation; so can non-deterministic Turing machines whose next computation step at a given instant can be chosen, with certain probabilities, among a finite number of possibilities. Nevertheless, these generalisations cannot compute what a traditional Turing machine cannot do: all of the Turing machines have the same power in terms of computability [25].
Viewed from this perspective, (universal) Turing machines are just functions from non-negative integers (encoding the input) to non-negative integers (encoding the output). But simple arguments can show that the class of functions realised by Turing machines cannot be the same as the whole class of functions from the set of natural numbers to the same set. On the one hand, the number of Turing machines is only countably infinite because each machine can be mapped into a unique integer. On the other hand, the whole class of functions from natural numbers to natural numbers can be shown by the Cantor diagonal arguments [35] to be uncountably infinite (in fact, this class of functions has the same cardinality as the set of reals). A pictorial and heuristic way to visualise this fact is where each mapping from natural numbers to themselves is represented by a real-valued angle between the two semi-infinite half-lines, one of which contains the domain of the function, the other the range. There are as “many” functions as there are angles between the two lines.

The set of functions that are captured by and identified with Turing machines has been identified as that of so-called partial recursive functions and is briefly reviewed in the next section. From here on, we also call Turing machines interchangeably with programs and partial recursive functions.

2.2 Recursive functions

We begin with the set of basic primitive recursive functions, where all the variables are in the set of non-negative integer \( \mathbb{N} \),

1. The successor function: \( s(x) = x + 1 \);
2. The zero function: \( z(x) = 0 \);
3. The projections functions: \( p_i^{(n)}(x_1, \ldots, x_n) = x_i, 1 \leq i \leq n \).

These are simple and intuitive functions.

From this basic set, other primitive recursive functions can be built using a finite number of the following operations

1. Composition: \( f(x_1, \ldots, x_k) = h(g_1(x_1, \ldots, x_k), \ldots, g_n(x_1, \ldots, x_k)) \); where \( h \) is an \( n \)-variable function and the \( n \) functions \( g_i \) are \( k \)-variable.

2. Primitive recursion: Let \( g \) and \( h \) be functions with \( n \) and \( n+2 \) variables, respectively. The \( n+1 \)-variable function \( f \) can then be defined as

\[
\begin{align*}
 f(x_1, \ldots, x_n, 0) &= g(x_1, \ldots, x_n); \\
 f(x_1, \ldots, x_n, y + 1) &= h(x_1, \ldots, x_n, y, f(x_1, \ldots, x_n, y)).
\end{align*}
\]

(2)

It is easy to see that the primitive recursive functions so constructed are total in the sense that they are defined for all the non-negative integers.
The set of primitive recursive functions is quite large but still inadequate and not large enough to encompass all the functions that are admissible and computable by the Turing machines. For this, we have to introduce yet another necessary operation, in addition to the two above, the unbounded \( \mu \)-minimisation which is defined as:

\[
f(x_1, \ldots, x_n) = \text{the least } y \text{ such that } p(x_1, \ldots, x_n, y) = 0; \\
\equiv \mu y[p(x_1, \ldots, x_n, y)];
\]  

(3)

where \( p(x_1, \ldots, x_n, y) \) is a total function for all the non-negative integral values of the variables. The general, mechanical method to find \( y \) that is applicable to all total functions \( p \) is to step \( y \) through 0, 1, 2, \( \ldots \) until \( p = 0 \) is satisfied. (For a particular function \( p \) one might be able to find \( y \) in a different and quicker way, but the mechanical method is applicable in general without any privy knowledge about characteristics of \( p \), as long as it is total.)

All these operations acting, in a finite number of steps, on the basic functions define the set of partial recursive functions which contains the set of primitive recursive functions as a proper subset. The name partial recursive signifies the fact that there is no guarantee that a (least) \( y \) can always be found to satisfy the condition \( p(x_1, \ldots, x_n, y) = 0 \) for given \( x_1, \ldots, x_n \) in the (unbounded) \( \mu \)-minimisation. If such a \( y \) cannot be found, \( f \) in (3) is undefined at that point. The problem to establish whether \( f \) is defined or not at a general point \( (x_1, \ldots, x_n) \) in its arguments, i.e. whether such a \( y \) can be found, is a decision problem which could have a binary answer, yes or no. However, this particular problem is a mathematically undecidable problem, in the sense that there exists no general algorithm that can always yield the required answer.

We define the set of \( \mu \)-recursive functions being the subset of partial recursive functions where there always exists a decision procedure for the \( \mu \)-minimisation operation, if needed, that is, there always exists a value for \( y \). This set is bigger than the set of primitive recursive functions. And it can be proved that a function is \( \mu \)-recursive if and only if there exists a Turing machine which can compute it and then halts, i.e., if and only if it is Turing computable.

The Church-Turing thesis stipulates that all the functions that are effectively computable are in fact Turing computable, and vice versa, thus restricting the intuitive and informal notion of computability to the well-defined mechanical operations of Turing machines. This identification is a thesis and not a theorem, not even a conjecture, because it can never be proven right in the mathematical sense. Nevertheless, it can be shown to be wrong if a counter-example can be found in which the computation steps are clearly and acceptably identified. Since it is testable, the thesis does not have the status of an axiom in Mathematics either.

We dispute this thesis by showing in a later section that there exist computable functions, computable by executing well-defined quantum mechanical procedures in a finite manner, that are not Turing computable.

The evaluation of partial recursive functions, even though well defined (through the three operations of composition, primitive recursion and \( \mu \)-minimisation upon basic functions and their resultants), may require an infinite number of steps – in contrast to the
μ-recursive functions whose evaluation requires only a finite number of steps. That is, partial recursive functions can be implemented on a Turing machine with well-defined execution steps but the machine could never halt in some cases.

One example of non-halting is the simplest program to find a number that is not the sum of four square numbers [32]. All that the program can do is to step through the natural numbers one-by-one and test each of these numbers by direct substitution of square numbers in increasing order of magnitudes. When a number can be found which is not the sum of four squares, the program prints out that number and halts. But this is a false hope. Constructed in this way, such a program never halts because there exists Lagrange’s Theorem which confirms that all numbers can in fact be expressed as the sum of four squares!

Another example can be seen through the famous but unproven Goldbach’s conjecture that every even number greater than 2 can be written as a sum of two primes. The simplest program would be one which steps through the even numbers one by one, and for each of these even numbers tests for all prime numbers (less than the even numbers). If an even number is found which is not a sum of two primes, the program prints out that number and halts. In other words, the program halts if and only if the Goldbach’s conjecture is false. So far, no counter-example to Goldbach’s conjecture has been found this way. However, we cannot rely on such a program simply because it may never halt.

There are many other important conjectures and problems in Mathematics which can be resolved if we somehow have a way to tell whether simple Turing machines/programs, directly constructed similar to those above, will halt or not. Thus is the all-important Turing halting problem: given a Turing machine, the question is whether there is a general algorithm which is able to determine if the machine would halt with a specified input. As is well-known in the framework of classical computability, this halting problem is as undecidable as the question whether a partial recursive function is defined with a given argument.

2.3 The Turing halting problem

The halting problem for Turing machines is a manifestation of undecidability: a Turing computation is equivalent to the evaluation of a partial recursive function, which is only defined for a subset of the integers. As this domain is classically undecidable, one cannot always tell in advance whether the Turing machine will halt (that is, whether the input is in the domain of the partial recursive function) or not (when the input is not in the domain).

A version of the proof of the unsolvability of the halting problem based on the Cantor diagonal argument goes as follows. The proof is by contradiction with the assumption of the existence of a computable halting function \( h(p, i) \) which has two integer arguments - \( p \) is the Gödel encoded integer number for the algorithm and \( i \) is its (encoded) integer input:

\[
h(p, i) = \begin{cases} 
0 & \text{if } p \text{ halts on input } i \\
1 & \text{if } p \text{ does not}
\end{cases}
\]
One can then construct a program $r(n)$ having one integer argument $n$ in such a way that it calls the function $h(n, n)$ as a subroutine and

$$
\begin{cases}
    r(n) \text{ halts if } h(n, n) = 1 \\
    r(n) \text{ loops infinitely (i.e., never stops) otherwise.}
\end{cases}
$$

The application of the halting function $h$ on the program $r$ and input $n$ results in

$$h(r, n) = \begin{cases} 
    0 & \text{if } h(n, n) = 1 \\
    1 & \text{if } h(n, n) = 0
\end{cases} \tag{5}$$

A contradiction is clearly manifest once we put $n = r$ in the last equation above.

The construction of such a program $r$ is transparently possible, unless the existence of a computable $h$ is wrongly assumed. Thus the contradiction discounts the assumption that there is a classically algorithmic way to determine whether any arbitrarily given program with arbitrary input will halt or not.

However, this contradiction argument might be avoided if we distinguish and separate the two classes of quantum and classical algorithms. A quantum function $qh(p, i)$, similar to eq. (4), can conceivably exist to determine whether any classical program $p$ will halt on any classical input $i$ or not. The contradiction in eq. (5) would be avoided if the quantum halting $qh$ cannot take as an argument the modified program $r$, which is now of quantum character because it now has quantum $qh$ as a subroutine. This will be the case if $qh$ can only accept integers while quantum algorithms, with proper definitions, cannot in general be themselves encoded as integers. In fact, the no-cloning theorem [42] of quantum mechanics does restrict the type of operations available to quantum algorithms.

In essence, the way we will break the self-referential reasoning here by the differentiation between quantum and classical algorithms is similar to the way Bertrand Russell resolved the set theory paradox (to do with “The set of all sets which are not members of themselves”) by the introduction of classes as distinct from sets. (For other lines of arguments, see [38, 37, 8] for example.)

To investigate the decidability of the Turing halting problem in the framework of quantum computability, we will need to isolate the point which causes the classical undecidability. With this aim we turn to the Diophantine equations in the next section.

### 2.4 Relation to Diophantine equations and Hilbert’s tenth problem

Identities between polynomials with integer coefficients in several unknowns over the natural numbers have been studied for some time in mathematics under the name of Diophantine equations. At the turn of the last century, David Hilbert listed, as challenges for the new century, 23 important problems among which the problem number ten is the only decision problem and could be rephrased as:

*Given any polynomial equation with any number of unknowns and with integer coefficients (that is, any Diophantine equation): To devise a universal process*
according to which it can be determined by a finite number of operations whether
the equation has integer solutions.

There are many important and interesting mathematical conjectures which can be
proved or disproved depending on whether some corresponding Diophantine equations
have an integer solution or not. These are, for instance, Goldbach’s conjecture which we
already mentioned, the Riemann hypothesis about the positions in the complex plane of
the zeroes of the zeta function, the four-colour conjecture for planar maps, Fermat’s last
theorem, etc ... Thus, we can appreciate the importance of having a general method for all
Diophantine equations, instead of considering each of them individually on its own merits.

However, there are only few special cases of Diophantine equations that are solvable.
These include linear (first-degree) equations in the unknowns, of which the existence and
absence of solutions can be inferred from the Euclid’s algorithm. Also solvable are second-
degree equations with only two unknowns, that is, in quadratic forms. But Hilbert asked
for a single general and finite decision procedure that is applicable for any Diophantine
equations. Little was it known that this problem is ultimately connected to the seemingly
unrelated notions of computation and Turing machines which were to be introduced some
40 years later.

Eventually, the Hilbert’s tenth problem was finally shown to be undecidable in 1970
through a crucial step by Matiyasevich [27, 9]: The Hilbert’s tenth problem could be
solved if and only if the Turing halting problem could also be solved. The two are simply
equivalent. Consequently, as we have a proof that the Turing’s is not solvable, the Hilbert’s
tenth problem is noncomputable/undecidable in the most general sense if one accepts, as
almost everyone does, the Church-Turing thesis of computability. One would thus have
to be content with the fact that individual Diophantine equations need to be considered
separately, and may or may not be solved with a different approach anew each time.

For a precise discussion and the history of the negative result, see [27, 9]; for a semi-
popular account, see [5]. We briefly sketch below some key arguments leading to the
nondecidable result.

It is recognised that Turing machines as devices mapping inputs to outputs are equiva-
 lent to partial recursive functions, whose domain are restricted to proper subsets of \( \mathbb{N} \).
Without loss of generality, we can always restrict the computation of Turing machines
to that of numeric partial recursive functions – because for non-numeric algorithms there
always exist mappings into the computation of some corresponding partial recursive func-
tions. The question whether a Turing machine would halt or not upon some particular
input is now equivalent to the question whether the corresponding argument value is in the
domain of the corresponding function or not. When a machine cannot halt with a given
input is when the partial recursive function corresponding to that machine is not defined
at that particular argument.

Now, there is a universal representation, known as the Kleene normal form, for any
partial recursive function \( f \)

\[
 f(x_1, \cdots, x_n) = \psi(\mu y[\tau(g, x_1, \cdots, x_n, y)]), \tag{6}
\]
where \( \psi \) and \( \tau \) are fixed primitive recursive functions, independent of the particular function \( f \) on the left hand side. The information about \( f \) on the right hand side of (6) is encoded in the first argument of the function \( \tau \) as the Gödel number \( g \) of \( f \). (We recall that the class of partial recursive functions is countably infinite and each function can be systematically given a unique identification number according to some Gödel numbering scheme.) It is the \( \mu \)-minimisation operation in (6) that could potentially turn a total primitive recursive function, which is defined everywhere, into a partial recursive function. Thus, given \( (x_1, \ldots, x_n) \) if there exists a (least) \( y \) that for some integer \( g \) solves \( \tau(g, x_1, \ldots, x_n, y) = 0 \) then the Turing machine, which corresponds to \( g \), would halt upon the input corresponding to \( (x_1, \ldots, x_n) \). If no such \( y \) exists, the Turing machine in question would not halt.

Gödel, as quoted in [33], has shown that the \( \mu \)-minimising operation (3) can always be represented as some arithmetic statement with a set of identities between multi-variate polynomials over the integers, together with a finite number of the existential quantifiers, \( \exists \), and bounded versions of the universal quantifiers, \( \forall \). (The boundedness comes from the requirement of some least number that satisfies a \( \mu \)-minimisation as defined in (3).) Note that quantifiers hold a crucial role in arithmetic. Restricted arithmetics with only the addition operation or the multiplication operation, but not both, are in fact complete: any of their statements can be decided to be provable or not within the framework. On the other hand, we have the famous Gödel’s Incompleteness Theorem for ‘ordinary’ arithmetic which has both addition and multiplication operations. Normally we would think that multiplication is just a compact way to express long and repeated additions, and might thus be puzzled over the difference in completeness. But it is the quantifiers that make such a huge difference.

Arithmetic statements for solutions of Diophantine equations (i.e. the Hilbert’s tenth problem) can only involve existential quantifiers, but the elimination of bounded universal quantifiers in the \( \mu \)-minimisation necessitates the appearance of exponentiation of variables. A famous example of variable exponentiation is the equation of Fermat’s last theorem:

\[
(x + 1)^{u+3} + (y + 1)^{u+3} + (z + 1)^{u+3} = 0,
\]

which has no non-negative integer solutions for the unknowns \( x, y, z \) and \( u \).

After various attempts by many people, it was finally shown that exponentiation is indeed Diophantine – that is, variable exponents can be eliminated to result in only polynomial Diophantine equations. This proves the equivalence between the two decision problems and implies that the Hilbert’s tenth is not computable in the Turing scheme of computation.

There are many other noncomputable problems, some are also equivalent to the Turing halting problem, and others belong to altogether different non-computability classes.

### 2.5 Some other noncomputables

Similarly based on the Turing machines, many other problems can be shown to be equivalent to the Turing halting problem, and thus noncomputable. We can name in this class [25]
the tiling problem in a plane, Post’s correspondence problem, Thue’s word problem, Wang’s domino problem, etc. . . For example, the tiling problem asks for a decision procedure to see if any given finite number of sets of tiles (but each set has an infinite number of the same tiles available) can tile the first quadrant of the plane or not, with a specified tile at the origin at the lower left corner and some adjacency rules for the tiling.

One could also ask questions related to and generalising the Hilbert’s tenth problem such as [9]: Is there a single algorithm for testing whether the number of solutions for any Diophantine equation is finite, or is infinite, or equal to one, or is even?

Besides the above equivalent to the Turing halting problem, there are yet others which belong to different classes in the non-computability hierarchy. The most famous of all must be the Gödel’s Incompleteness Theorem. It establishes that any finitely axiomatic, consistent mathematical system sufficiently complex to embrace Arithmetic must be incomplete – that is, there exist some statements whose truth cannot be confirmed or denied from within the system, resulting in undecidability. (Interestingly, it is precisely the statement about consistency of the system that is neither provable nor deniable.) We can see that this undecidability is more than that of Turing machines, whose applicable mathematical statements can only have bounded universal quantifiers as opposed to arbitrarily unbounded universal quantifiers of Arithmetic at large.

Chaitin, approaching from the perspectives of Algorithmic Information Theory [6, 7], has shown that there exist many unprovable statements in Arithmetic simply because they have irreducible algorithmic contents, measurable in bits, that are more than the complexity, also measurable in bits, of the finite set of axioms and inference rules of the system. Infinitely irreducible algorithmic complexity is randomness which exists even in pure mathematics. And more frustratingly, we can never prove randomness since we could only ever deal with finite axiomatic complexity. Chaitin, to illustrate the point, has introduced the number Ω as the halting probability for a random program, with some random input, being emulated by a particular Turing machine. Ω is an average measure over all programs run on the universal Turing machine. This number has many interesting properties, and has been generalised to a quantum version [38], but we only mention here the linkage between this number and polynomial Diophantine equations. When expressed in binary, the value of the k-th bit of Ω is 0 or 1 depending on whether some Diophantine equation, corresponding to the Turing machine in consideration,

\[ C(k, N, x_1, \ldots, x_K) = 0, \quad (8) \]

for a given \( k > 0 \), has solutions in non-negative integers \((x_1, \ldots, x_K)\) for finitely or infinitely many values of the parameter \( N > 0 \).

From this Diophantine representation we can see that the noncomputability of Ω is “more” than the noncomputability of the Turing halting problem. Even if, somehow, we have an algorithm for the latter to decide the equations in (8) for each \( N \) respectively, we would still need to apply the algorithm for an infinite number of times just to get a single digit for Ω! An infinite number of times cannot be normally performed in a finite time, hence the different class of noncomputability from that of the Turing halting problem. We
will come back to the $\Omega$ number with a hypothetical algorithm in Quantum Field Theory later.

We end our brief review of relevant mathematical concepts here with a discussion of computable numbers. If we had restricted such numbers to only those that can be output by some Turing machine (which then halts), then we would have had to effectively restrict ourselves to integers and to treat irrational numbers, like $\sqrt{2}$ and $\pi$, as noncomputable. This is clearly undesirable. A better and more practical definition of computable numbers are those which we can approximate to an arbitrary degree of accuracy with (integer/rational) outputs of some Turing machines.

This excludes the Chaitin’s $\Omega$ as being computable because we cannot estimate this number to an arbitrarily high accuracy, although it is bounded from above by unity (being the maximum probability) and can be approximated only from below by some very slowly converging process [6] whose convergence rate is indeterminable.

3 Quantum Mechanics

3.1 The postulates of measurement and associated problems

Quantum Physics, including Quantum Mechanics and Quantum Field Theory, is the most successful theory that we have in Science for the description and prediction of phenomena in Nature. And so far there is not a single discrepancy between the theory and experiments.

According to Quantum Mechanics, pure states of a physical system can capture the most that can be said about the system and are associated with vectors, unique up to phases, in some linear vector Hilbert space. When the system, particularly when it is a subsystem of a bigger entity, cannot be described by a pure state but is in a mixed state, the language of density matrices would be necessary for its description. Acting on the Hilbert space are linear operators, of which hermitean and unitary operators are of particular interest.

In the Schrödinger picture where the time dependency is explicitly carried by the states, the time evolution of the system is governed by the Schrödinger equation, in which the hermitean hamiltonian operators play a unique role. In general, each physical observable is associated with a hermitean operator; the hamiltonian operator, for instance, is associated with the system’s energy. The real-valued eigenvalues (which can be continuous or discrete) of these hermitean operators restrict the obtainable values under observation. Each time when the associated observable is measured, only one single value, among the eigenvalues given, is obtained. Repetitions of the measurement under identical conditions could yield different measured values each time. And the probability of getting a particular eigenvalue in a measurement is given by the square of the absolute value of the inner product between the corresponding eigenvector and the state describing the system at that instant. If the system is in a mixed state described by a density matrix, the probability is then given by the trace of the product between the density matrix and the corresponding projector associated with the eigenvector in question.
After a measurement, the state of the system is a pure state whose representing vector is the same as the eigenvector, up to a phase, corresponding to the eigenvalue obtained. Note that measurement thus is a non-unitary and irreversible operation in general, unless the system is already in the observable eigenstate. Different observables can be measured simultaneously, with the same degree of accuracy after many repetitions, only when the associated hermitean operators commute with each other.

But already seeded in the summary of quantum mechanical postulates above is a fundamental problem of inconsistency. The act of measurement, on the one hand, is a process unfolded itself in time. On the other hand, why should it not be governed by the unitary Schrödinger time-evolution operator?

Even when we extend the system to a larger system consisting of the considered system and the measuring apparatus, we still face the same Quantum Measurement Problem, under a slightly different guise. In setting up a measurement we effectively establish some correlation between the state of the measured system and the pointers of the measuring apparatus. This correlation is one-to-one for the eigenstates of the operator representing the measured observable: if the system is in the state $|e_i⟩$, $i = 1, 2$, then the pointer, initially in a neutral state $|A_0⟩$, is subsequently in the state $|A_i⟩$ respectively,

$$|e_i⟩|A_0⟩ → |e_i⟩|A_i⟩.$$ (9)

Now, according to the quantum mechanical principles of superposition and linearity, it follows that if the system is in a superposition then so is the pointer,

$$(\alpha|e_1⟩ + \beta|e_2⟩) |A_0⟩ → \alpha|e_1⟩|A_1⟩ + \beta|e_2⟩|A_2⟩.$$ (10)

The problem is that we have never been able to observe the classical pointer in an entangled state as in the right hand side of the last expression. Instead, we either get the pointer position $A_1$ with a probability proportional to $|\alpha|^2$ or $A_2$ with a probability proportional to $|\beta|^2$.

To resolve this problem one could postulate, as in the Copenhagen interpretation, some undefined separation between the quantum and the classical worlds, or modify the theory to have elements of non-linearity admitted. (This second situation is unlike mathematics in the way that the postulates/axioms of a physical theory can themselves also be subjects of investigation.) But if we believe in Quantum Mechanics as the universal and fundamental theory then we have a difficult problem at hand. Various resolution attempts ranging from the many-world interpretation to decoherence have not been deemed successful.

We shall leave the measurement problem here, and only wish to emphasise that the power of all quantum algorithms in quantum computation relies crucially on such mysterious measurement processes.

Also for later use, we now introduce the concept of measurable quantities [17]. Analogous to the concept of computable numbers discussed in Sec. 2.5, a number $w$ is deemed measurable if there exists a finite set of instructions for performing an experiment such that a technician, given an abundance of unprepared raw materials and an allowed error $\epsilon$,}

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is able to obtain a rational number within $\epsilon$ of $w$. The technician is analogous to the computer, the instructions analogous to the computer program, the "abundance of unprepared raw materials" analogous to the infinite Turing tape, initially blank.

In particular, not only the (stochastic) outcomes of an observable are obviously measurable and of interest but so are the probability distributions for these outcomes. The probabilities can be obtained within any given accuracy by increasing the number of measurement repetitions. Later, we make full use of this crucial fact that the quantum mechanical probabilities are both computable from the theory of Quantum Mechanics and measurable in principle.

3.2 Coherent states

One of the simplest and most successful problems in Quantum Mechanics is that of the (one-dimensional) Simple Harmonic Oscillator (SHO) with the hamiltonian

$$H_{\text{SHO}} = (P^2 + X^2)/2,$$

which can also be expressed as

$$H_{\text{SHO}} = a^\dagger a + \frac{1}{2}.$$  

(12)

The operators $a^\dagger$, $a$ are linearly related to the position and momentum operators, which satisfy the commutation relation

$$[P, X] = i,$$

$$X = \frac{1}{\sqrt{2}} (a + a^\dagger),$$

$$P = \frac{i}{\sqrt{2}} (a - a^\dagger).$$

(13)

The operators $a^\dagger$, $a$ satisfy different commutation relations

$$[a, a^\dagger] = 1,$$

$$[a, a] = [a^\dagger, a^\dagger] = 0.$$  

(14)

The spectrum of the number operator $N = a^\dagger a$ that appears in (12) is discrete and spans over the natural numbers. Its eigenstates are termed the number states $|n\rangle$,

$$N|n\rangle = n|n\rangle; \quad n = 0, 1, 2, \ldots$$

(15)

These eigenstates also constitute an orthonormal basis for a Fock space, a special type of Hilbert space, and can be constructed by the operators $a^\dagger$ acting on the special "vacuum" state $|0\rangle$, the lowest-eigenvalue state,

$$|n\rangle = \frac{a^n}{\sqrt{n!}}|0\rangle.$$  

(16)
from which follow the recursive relations

\[ a^\dagger |n\rangle = \sqrt{n + 1} |n + 1\rangle, \]
\[ a|n\rangle = \sqrt{n} |n - 1\rangle. \]  

(17)

These relations lead us to the names *creation* and *annihilation* operators respectively for \(a^\dagger\) and \(a\).

The number state \(|n\rangle\) can be realised in Quantum Optics as one having a definite number of \(n\) photons, all at the same frequency. But these number states are not the states of travelling optical modes generated by idealised lasers [13], which have an indefinite number of photons. For the description of these modes, we need the coherent states [41], \(|\alpha\rangle\), which are the eigenstates of \(a\) and are labeled by the complex number \(\alpha\),

\[ a|\alpha\rangle = \alpha|\alpha\rangle. \]  

(18)

With the relation to the number states,

\[ |\alpha\rangle = e^{-|\alpha|^2} \sum_{n=0}^{\infty} \frac{\alpha^n}{\sqrt{n!}} |n\rangle, \]
\[ = e^{-|\alpha|^2} e^{a^\dagger a} |0\rangle, \]  

(19)

the coherent states are not orthogonal but can still be used for spanning the Hilbert space. They have some unique and nice properties, one of which is that they are the states that optimise the amplitude-phase Heisenberg uncertainty relation. The other fact, which we will exploit later, is that they are also the ground states, i.e. the eigenstates having the lowest “energy” eigenvalues, for the family of semi-definite hamiltonians

\[ H_\alpha = (a^\dagger - \alpha^*) (a - \alpha). \]  

(20)

With the simple substitution \(\alpha_\alpha = a - \alpha\) we are back to a family of \(\alpha\)-labeled SHO hamiltonians (11), except for the additive constant, all of which have the same spectrum over the natural numbers but with different sets of eigenvectors \(|n_\alpha\rangle\).

### 3.3 The quantum adiabatic theorem

The dynamical evolution of a quantum system is governed by the hamiltonian through the Schrödinger equation. If the system is closed then the hamiltonian is time independent. If the system is subject to external influences, whose dynamics are not of direct concern to the investigation, then the hamiltonian is time dependent; and the modification in the quantum state of the system critically depends on the time \(T\) during which the change of the hamiltonian takes place. This dependency is particularly simplified when the rate of change of the external fields is very fast compared to some intrinsic time scale, whence we can apply the sudden approximation, or is very slow, whence we can appeal to the quantum adiabatic theorem.
The sudden approximation says that if the time change $T$ is sufficiently slow relative to the inverse of the average hamiltonian during that time, $\Delta \bar{H}$,

$$ T \ll \hbar/\Delta \bar{H}, \quad (21) $$

then the dynamical state of the system remains essentially unmodified.

On the other hand, in the case of an infinitely slow, or adiabatic passage, if the system is initially in an eigenstate of the hamiltonian at the initial time it will, under certain conditions, pass into the eigenstate of the hamiltonian at the final time, that derives from it by continuity [28]. This is the content of the adiabatic theorem, provided the following conditions are satisfied throughout the relevant time interval:

- The instantaneous eigenvalues remain distinct;
- The first and second derivatives of the instantaneous eigenvectors with respect to time are well-defined and piece-wise continuous.

This important theorem has been exploited for a model of quantum computation involving the ground state and to be discussed below. In practice, the evolution time $T$ is finite but the more it satisfies the condition

$$ T \gg \frac{\| \Delta H \|}{g^2}, \quad (22) $$

then the higher the probability that the system remains in the instantaneous eigenstate. In the above,

$$ \| \Delta H \| \equiv \max_{0 \leq t \leq T} |\langle e(t)|H(T) - H(0)|g(t)\rangle|, \quad (23) $$

and

$$ g \equiv \min_{0 \leq t \leq T} (E_e(t) - E_g(t)), \quad (24) $$

where $|g(t)\rangle$ and $|e(t)\rangle$ are respectively the instantaneous ground state and first excited state with instantaneous eigenvalues $E_g(t)$, $E_e(t)$. The critical scale is this gap $g$.

4 Quantum Computation

The underlying laws of all physical phenomena in Nature, to the best of our knowledge, are those given by quantum physics. However, the best present day computers, as they are of classical nature, cannot even in principle simulate quantum systems efficiently. Feynman pointed out in 1982 that [16], see also [2], only quantum mechanical systems may be able to simulate other quantum systems more efficiently. Furthermore, according to Moore’s law, the exponential rate of miniaturisation of micro-electronic semiconductor devices will soon, if not already, take us to the sub-micron and nano dimensions and beyond. On
this scale, quantum physics will become more and more relevant in the design and production of computer components. Heat dissipation in irreversible computation will be yet another problem at these dimensions. Even though reversible classical computation can be implemented in principle, quantum computation, being almost reversible except the final read-out by measurement, could automatically minimise this heating problem.

The notion of effective computability, as delimited by the Church-Turing thesis “Every function which would be naturally regarded as computable can be computed by the universal Turing machine”, begs the question whether it could be extended with quantum principles. Initial efforts seem to confirm that quantum computability is no more than classical and mathematical computability [3]. However, more recent indications may prove otherwise [22, 8]. We will come back to this computability notion in a section below.

All of the above have inevitably led to the recent convergence of quantum physics, mathematics, and computing and information processing.

4.1 “Standard” Model of Quantum Computation

According to the “standard” model of quantum computation (see [31] for instance), which is a generalisation of classical digital computing, the fundamental unit of a quantum computer is the quantum bit, shortened as qubit, which is the generalisation of a binary bit. Physical implementation of a qubit could be any (measurable) two-state system: the up and down values of a quantum spin, or the two polarisation states of a photon, etc ... But unlike the binary bit, a qubit can be in a superposition state of its two states/values. Upon measurement the superposition is destroyed, revealing one of the two classical values of a qubit. The two states of a qubit, denoted by |0⟩ and |1⟩, should be unambiguously distinct by measurement and thus be orthogonal to each other.

There are three stages of operation for a quantum computer, corresponding to the input, the processing, and finally the output. The input preparation stage can be and has been carried out in laboratories for certain well-known systems. So can the output stage in which the output is read out by an act of measurement – even though quantum measurement is not that well understood, as already alluded to in Sec. (3.1). The speed of state preparation and measurement, which should be carried out in such a way as not to perturb other subsystems/qubits not being directly measured, is crucial for quantum computation. The information processing stage is the most difficult to be implemented. In principle, it is governed by unitary evolution of a set of qubits well isolated from the surroundings to avoid as much of the decoherence effects of the environment as possible. The discovery of error correcting codes for quantum computation was a pleasant surprise. Without this possibility, realisation of quantum computation would have been unthinkable as computers are inevitably and constantly subject to errors induced by either the internal interactions or the environment or both.

The power of a quantum computer firstly lies in the massive parallelism resulting directly from the superposition possibility of the quantum states. If each qubit is a superposition of two states then the measurement of such a superimposed N-qubit system could in general access $2^N$ distinguished states simultaneously. However, such quantum parallelism
is not that useful because of the probabilistic nature of the measured outcomes. (After all, many classical wave systems, like water waves, can also have superposition but cannot provide a better computation model.) The second and most important power of quantum computation is thought to have its root in quantum entanglement [4], which has no counterpart in the classical world (even though it might be expensively simulated by classical means). Quantum entanglement provides the extra dimensions in information storage and processing that distinguishes the quantum from the classical. It is the entanglement that allows us to control the massive quantum parallelism through selective interference of different computational paths to extract the information desired.

These characteristics have been exploited to reduce the computational complexity of some problems. So far there are only few quantum algorithms discovered [36, 19]; most notable is Shor’s factorisation algorithm which employs Quantum Fourier Transformation. QFT is the only known quantum algorithm that could offer an exponential increase in computational speed, due to the interference of different computation paths (as Fourier Transformation is intimately linked to interference) and to quantum entanglement. More quantum algorithms are urgently needed.

The approach above with qubits and unitary gates of so-called quantum networks has been accepted as the standard model for quantum computation. It has been argued [3] that the computability obtainable in this model is not better but is the same as classical computability.

However, it is not the only model available.

4.2 Quantum Adiabatic Computation

Among the alternative models for quantum computation is the recent proposal [15] to employ quantum adiabatic processes for computation. The idea is to encode the solution of some problem to be solved into the ground state, $|g\rangle$, of some suitable hamiltonian, $H_P$. But as it is easier to implement the hamiltonian than to obtain the ground state, we should start the computation in yet a different and readily obtainable initial ground state, $|g_I\rangle$, of some initial hamiltonian, $H_I$, then deform this hamiltonian in a time $T$ into the hamiltonian whose ground state is the desired one, through a time-dependent process,

$$H\left(\frac{t}{T}\right) = \left(1 - \frac{t}{T}\right)H_I + \frac{t}{T}H_P.$$  \hfill (25)

The adiabatic theorem of Quantum Mechanics, Sec. 3.3, stipulates that if the deformation time is sufficiently slow compared to some intrinsic time scale, the initial state will evolve into the desired ground state with high probability – the longer the time, the higher the probability.
5 Quantum algorithm for the Hilbert’s tenth problem

5.1 General approach

It suffices to consider non-negative solutions, if any, of a Diophantine equation. Let us consider a particular example

\[(x + 1)^3 + (y + 1)^3 - (z + 1)^3 + cz = 0, \quad c \in \mathbb{Z}, \quad (26)\]

with unknowns \(x, y,\) and \(z\). To find out whether this equation has any non-negative integer solution by quantum algorithms, it requires the realisation of a Fock space. Upon this Hilbert space, we construct the hamiltonian corresponding to (26)

\[H_P = (a_x^\dagger a_x + 1)^3 + (a_y^\dagger a_y + 1)^3 - (a_z^\dagger a_z + 1)^3 + c(a_x^\dagger a_x)(a_y^\dagger a_y)(a_z^\dagger a_z)^2,\]

which has a spectrum bounded from below – semidefinite, in fact. Note that the operators \(N_j = a_j^\dagger a_j\) have only non-negative integer eigenvalues \(n_j\), and that \([N_j, H_P] = 0 = [N_i, N_j]\) so these observables are simultaneously measurable. The ground state \(|g\rangle\) of the hamiltonian so constructed has the properties

\[N_j |g\rangle = n_j |g\rangle,\]

\[H_P |g\rangle = \left((n_x + 1)^3 + (n_y + 1)^3 - (n_z + 1)^3 + cn_xn_yn_z\right)^2 |g\rangle \equiv E_g |g\rangle,\]

for some \((n_x, n_y, n_z)\).

Thus, a projective measurement of the energy \(E_g\) of the ground state \(|g\rangle\) will yield the answer for the decision problem: The Diophantine equation has at least one integer solution if and only if \(E_g = 0\), and has not otherwise. (If \(c = 0\) in our example, we know that \(E_g > 0\) from Fermat’s last theorem.)

If there is one unique solution then the projective measurements of the observables corresponding to the operators \(N_j\) will reveal the values of various unknowns. If there are many solutions, finitely or infinitely as in the case of the Pythagoras theorem, \(x^2 + y^2 - z^2 = 0\), the ground state \(|g\rangle\) will be a linear superposition of states of the form \(|n_x\rangle|n_y\rangle|n_z\rangle\), where \((n_x, n_y, n_z)\) are the solutions. In such a situation, the measurement may not yield all the solutions. However, finding all the solutions is not the aim of a decision procedure for this kind of problem.

Notwithstanding this, measurements of \(N_j\) of the ground state would always yield some values \((n_x, n_y, n_z)\) and a straightforward substitution would confirm if the equation has a solution or not. Thus the measurement on the ground state either of the energy, provided the zero point can be calibrated, or of the number operators will be sufficient to give the result for the decision problem.

The quantum algorithm with the ground-state oracle is thus clear:

1. Given a Diophantine equation with \(K\) unknown \(x\)’s

\[D(x_1, \cdots, x_K) = 0, \quad (27)\]
we need to simulate on some appropriate Fock space the quantum hamiltonian

\[ H_P = (D(a_1^\dagger a_1, \cdots, a_K^\dagger a_K))^2. \]  

(28)

2. If the ground state could be obtained with high probability and unambiguously verified, measurements of appropriate observables would provide the answer for our decision problem.

The key ingredients are the availability of a countably infinite number of Fock states, the ability to construct/simulate a suitable hamiltonian and to obtain and verify its ground state. As a counterpart of the semi-infinite tape of a Turing machine, the Fock space is employed here instead of the qubits of the more well-known model of quantum computation. Its advantage over the infinitely many qubits which would otherwise be required is obvious.

One way to construct any suitable hamiltonian so desired is through the technique of ref. [26]. We consider the hermitean operators, where \( j \) is the index for the unknowns of the Diophantine equation,

\[ X_j = \frac{1}{\sqrt{2}}(a_j + a_j^\dagger), \]
\[ P_j = \frac{i}{\sqrt{2}}(a_j - a_j^\dagger), \]
\[ [P_j, X_k] = i\delta_{jk}. \]  

(29)

Together with the availability of the fundamental hamiltonians

\[ X_j, P_j, (X_j^2 + P_j^2), \pm(X_k P_j + P_j X_k), \] and \((X_j^2 + P_j^2)^2\)  

(30)

one could construct the unitary time evolutions corresponding to hamiltonians of arbitrary hermitean polynomials in \( \{X_j, P_j\} \), and hence in \( \{a_j^\dagger a_j\} \), to an arbitrary degree of accuracy. These fundamental hamiltonians correspond to, for examples, translations, phase shifts, squeezers, beam splitters and Kerr nonlinearity.

With the polynomial hamiltonian constructed, we need to obtain its ground state. Any approach that allows us to access the ground state will suffice. One way is perhaps to use that of quantum annealing or cooling [21]. Another way is to employ the quantum computation method of adiabatic evolution [15].

5.2 Quantum adiabatic approach

In the adiabatic approach, one starts, for example, with a hamiltonian \( H_I \),

\[ H_I = \sum_{i=1}^{K}(a_i^\dagger a_i^*) (a_i - \alpha_i), \]  

(31)

which admits the readily achievable coherent state \(|g_I\rangle = |\alpha_1 \cdots \alpha_K\rangle\) as the ground state. Then, one forms the slowly varying hamiltonian \( \mathcal{H}(t/T) \) in (25), which interpolates in
the time interval $t \in [0, T]$ between the initial $H_I$ and $H_P$. According to the quantum adiabatic theorem, the initial ground state with certain probability will evolve into our desired ground state $|g\rangle$ up to a phase:

- **Step 1**: We first estimate roughly (and can be wrongly) the time $T_{est}$.
- **Step 2**: We use the physical (and not classical computer simulated) quantum adiabatic process which terminates after a time $T' > T_{est}$ to obtain by projective measurement a candidate state $|\cdots N_i \cdots\rangle$, $i = 1, \cdots, K$.
- **Step 3**: Then we verify whether the state $|\cdots N_i \cdots\rangle$ is the ground state by using (numerical) Quantum Mechanics in the verification steps below.
- **Step 4**: If this is not the ground state, we increase the estimated time $T_{est}$ by, say, a factor of 10 and repeat from the Step 2 above.

In contrast with the classical algorithm, the quantum algorithm above will terminate in principle as $T'$ is finite (even though it could be very long) and give us the decision result for the Hilbert’s tenth problem.

## 5.3 Verification of the ground state

- If the numbers $N_i$, $i = 1, \ldots, K$, so obtained satisfy the Diophantine equation then we have a solution.

- If not and if in a finite, sufficiently large integer neighbourhood of the numbers $N_i$ so obtained we cannot find a local minimum of the square of the Diophantine polynomial then we go next to the Step 4 in the above.

- If $(N_1, \cdots, N_K)$ is a local minimum, we include the occupation number state $|\cdots N_i \cdots\rangle$ in a truncated basis (starting from $|\cdots 0 \cdots\rangle$). By numerically studying the spectral flow of the hamiltonian $\mathcal{H}(t/T)$ in this truncated basis we can find the minimum energy gap between the lowest energy branch and the next branch, provided there are sufficient states in the truncated basis such that adding more vectors having higher energies into the basis does not change the energy gap of these two lowest branches to some given accuracy. From this numerical study we can derive the various probabilities from solving the Schrödinger equation numerically and go back to the physical adiabatic processes to confirm these probabilities for various adiabatic times for:

  - The reversed adiabatic process, starting from the eigenstate $|\cdots N_i \cdots\rangle$ and the hamiltonian $H_P$ and ending with $H_I$ to get back to the original space in which contains the state $|g_I\rangle$ and of which we know well. Since this latter space is better known, this method could help tagging and identifying the candidate $|\cdots N_i \cdots\rangle$. Note that this process is not quite reversible because of the intervening measurement of $|\cdots N_i \cdots\rangle$. Care must also be taken when there is
degeneracy for the ground state of $H_P$, for the adiabatic theorem may not be applicable to this reversed run.

- And/or the adiabatic processes starting from $H_I$ and ending with $H_P$ but with different initial coherent states $|\alpha\rangle, |\alpha'\rangle$ such that $\|\alpha - \alpha'\| \gg 1$. (This method is applicable for degeneracy cases of $H_P$ as the ground states of $H_I$’s are not degenerate for different $\alpha$’s.)

- And/or the adiabatic process starting from $H_I$ and ending with $H_P$ but with the initial coherent state $|\alpha_{\text{max}}\rangle$ which overlaps maximally with the candidate state $|\cdots N_i \cdots\rangle$. This maximal overlapping will in general enlarge the energy gap and thus reduce the time to obtain the true ground state.

The statistics so obtained for the dynamical processes agree with the numerical Quantum Mechanics if and only if the candidate state $|\cdots N_i \cdots\rangle$ is the true ground state of $H_P$.

In summary, to solve the Hilbert’s tenth problem we need both the physical adiabatic processes to obtain a candidate state and the numerical Quantum Mechanics to verify this is the ground state through the usual statistical predictions from the Schrödinger equation with a few low-lying energy states of $\mathcal{H}(t/T)$. This way, we can overcome the problem of where to truncate the infinite basis for a numerical study of Quantum Mechanics, and reconcile with the Cantor diagonal arguments which state that the problem could not be solved entirely in the framework of classical computation.

The key factor in the ground state verification is the probability distributions, which are, as mentioned in Sec 3.1, both computable in numerical Quantum Mechanics (with more and more truncated levels involved for a better and better accuracy in the estimation) and measurable in practice. However, in using the probability distributions as the verifying criteria, we have to assume that Quantum Mechanics is able to describe Nature correctly to the precision required. Note also that we have here a peculiar situation in which the computational complexity, that is, the adiabatic time, might not be known exactly before carrying out the quantum computation – although it can be estimated approximately.

### 6 Discussion of finer points

#### 6.1 Difference from classical algorithms

We do not look for the zeroes of the polynomial, $D(x_1, \cdots, x_K)$, which may not exist, but instead search for the absolute minimum of its square which always exists,

$$0 \leq \min(D(x_1, \cdots, x_K))^2 \leq (D(0, \cdots, 0))^2,$$

and is finite because $\lim_{x \to \infty} (D(x_1, \cdots, x_K))^2$ diverges.

While it is equally hard to find either the zeroes or the absolute minimum in classical computation, we have converted the problem to the realisation of the ground state of a quantum hamiltonian and there is no known quantum principle against such an act.
In fact, there is no known physical principles against it. Let us consider the three laws of thermodynamics concerning energy conservation, entropy of closed systems and the unattainability of absolute zero temperature. The energy involved in our algorithm is finite, being the ground state energy of some hamiltonian. The entropy increase which ultimately connects to decoherence effects is a technical problem for all quantum computation in general.

It may appear that even the quantum process can only explore a finite domain in a finite time and is thus no better than a classical machine in terms of computability. But there is a crucial difference.

In a classical search, even if the global minimum is encountered, it cannot generally be proved that it is the global minimum (unless it is a zero of the Diophantine equation). Armed only with mathematical logic, we would still have to compare it with all other numbers from the infinite domain yet to come, but we obviously can never complete this comparison in finite time – thus, mathematical noncomputability.

In the quantum case, the global minimum is encoded in the ground state. Then, by energetic tagging, the global minimum can be found in finite time and confirmed, if it is the ground state that is obtained at the end of the computation. And the ground state may be identified and/or verified by physical principles. These principles are over and above the mathematics which govern the logic of a classical machine and help differentiate the quantum from the classical. Quantum mechanics could “explore” an infinite domain, but only in the sense that it can select, among an infinite number of states, one single state (or a subspace in case of degeneracy) to be identified as the ground state of some given hamiltonian (which is bounded from below). This “sorting” can be done because of physical principles which are not available to mathematical computability.

6.2 Difference from the standard model of Quantum Computation

Our proposal is in contrast to the claim in [3] that quantum Turing machines compute exactly the same class of functions as do Turing machines, albeit perhaps more efficiently. We can only offer here some speculations about this apparent discrepancy. The quantum Turing machine approach is a direct generalisation of that of the classical Turing machines but with qubits and some universal set of one-qubit and two-qubit unitary gates to build up, step by step, dimensionally larger, but still dimensionally finite unitary operations. This universal set is chosen on its ability to evaluate any desirable classical logic function. Our approach, on the other hand, is from the start based on infinite-dimension hamiltonians acting on some Fock space and also based on the special properties and unique status of their ground states. The unitary operations are then followed as the Schrödinger time evolutions. Even at the hamiltonian level higher orders of the operators $a$ and $a^\dagger$, i.e. not just two-body but many-body interactions in a sense, are already present. This proliferation, which is even more pronounced at the level of the time-evolution operators, together with the infinite dimensionality and the unique energetic status of the vacuum could be the
reasons behind the ability to compute, in a finite number of steps, what the dimensionally
finite unitary operators of the standard quantum Turing computation cannot do in a finite
number of steps. Note that it was the general hamiltonian computation that was discussed
by Benioff and Feynman [2, 16] in the conception days of quantum computation.

6.3 Possible pitfalls

Our quantum algorithm is based on the key ingredients of:

- Our ability to physically implement certain hamiltonians having infinite numbers of
  energy levels;
- Our ability to physically obtain and verify some appropriate ground state;
- The exactitude, to the level required, of the theory of Quantum Mechanics in de-
  scribing and predicting physical processes.

If any of these is not achievable or approximable with controllable accuracy, the quantum
algorithm simply fails and further modifications may or may not work.

For example, we could implement the algorithm with Quantum Optical apparati, in
which a beam of quantum light is the physical system on which final measurements are per-
formed and the number of photons is the quantity measured. The hamiltonians could then
be physically simulated by various components of mirrors, beam splitters, Kerr-nonlinear
media (with appropriate efficiency), etc ... We should differentiate the relative concepts
of energy involved in this case; a final beam state having one single photon, say, could
correspond to a higher energy eigenstate of $H_P$ than that of a state having more photons!
Only in the final act of measuring photon numbers, the more-photon state would transfer
more energy in the measuring device than the less-photon state.

A fundamental problem [12] is that the hamiltonians which we need to be simulated
in the optical apparati are only effective hamiltonians in that their descriptions are only
valid for certain range of number of photons. When there are too many photons, a mirror,
for example, may respond in a different way from when only a few photons impinge on
it, or the mirror simply melts. That is, other more fundamental processes/hamiltonians
different than the desirable effective hamiltonians would take over beyond certain limit in
photon numbers. Thus it seems that we cannot have available an unbounded number of
levels for the quantum algorithm.

This situation is not unlike that of the required unboundedness of the Turing tape. In
practice, we can only have a finite Turing tape/memory/register; and when the register is
overflowed we would need to extend it.

Similar to this, we would have to be content with a finite range of applicability for our
simulated hamiltonians. But we should also know the limitation of this applicability range
and be able to tell when in a quantum computation an overflow has occurred – that is, when
the range of validity is breached. We could then use new materials with extended range of
(photon number) applicability or modify the Diophantine equations until the appropriate
ground state is verifiably obtained. A simple way to reduce the number of photons involved would be a shifting the unknowns in the Diophantine equations by some integer amounts, \( x_i \rightarrow x_i + n_i \), as in the example of (7).

This way of patching results in some approximation for our algorithm. The important thing is that the approximation is controllable because, within some given accuracy, only a finite number of low-lying energy levels of \( H_P \) can influence the statistics for obtaining the ground state. In other words, the probability distribution, necessary for the verification of the ground state, is still measurable to within any given accuracy with appropriate increase in the physically available number of low-lying levels.

7 Implications for other decision problems and Gödel’s Incompleteness Theorem

Generalised noncomputability and undecidability set the boundary for computation carried out by mechanical (including quantum mechanical) processes, and in doing so help us to understand much better what can be so computed. With this in mind, we mention here the results of some considerations [23] about some modified version of the Hilbert’s tenth problem and the computation of Chaitin’s Ω number in Sec. 2.5. These problems are all inter-related through questions about existence of solutions of Diophantine equations.

As pointed out in Sec. 2.5, we can also ask different questions [9] whether some Diophantine equation has a finite number of non-negative integer solutions (including the case in which it has no solution) or an infinite number, or whether the number of solutions is even, etc ... In general, we cannot tell the degree of degeneracy of the ground state; but with some modifications, the quantum algorithm above for the Hilbert’s tenth problem can be generalised to tackle this new class of questions. The possibility of such a generalisation confirms the mathematical equivalence between the Hilbert’s tenth problem and this new class.

Notwithstanding this, the situation is different with the evaluation of the Chaitin’s Ω number. We have to appeal to the hypothetical ability to physically construct hamiltonians involving a countably infinite number of distinct pairs of creation and annihilation operators. (Recall that for the Hilbert’s tenth problem and its equivalence, we only need as many pairs as the number of unknowns in a Diophantine equation.) In other words, we would have to possess the ability to create or simulate hamiltonians in Quantum Field Theory. That is, if we stick to the example of quantum computation with Quantum Optics, we would have to have individual control over infinitely many different optical modes, each with a distinct frequency. Clearly, the situation is worse than before and is even more hypothetical.

The difficulty we encounter for the Ω number is nothing but another manifestation of the different classes in the hierarchy of classical noncomputability. In Sec. 2.5 we pointed out that, on the one hand, the evaluation of this number is as classically noncomputable as the completeness of Arithmetic and that, on the other hand, Gödel’s Incompleteness is
“more” noncomputable than the Hilbert’s tenth problem.

Our decidability study so far only deals with the property of being Diophantine, which does not cover the property of being arithmetic in general (which could involve unbounded universal quantifiers). As such, our consideration has no direct consequences on Gödel’s Incompleteness theorem. However, it is conceivable that Gödel’s theorem may lose its restrictive power once the concept of mathematical proof is suitably generalised with quantum principles.

8 Back to Mathematics

While the proposal is about some quantum processes to be implemented physically, it illustrates the surprisingly important rôle of Physics in the study of computability. This is an unusual state of affairs when Physics, which has its roots in the physical world out there, could perhaps help in setting the limits of Mathematics.

Inspired by this connection between the two, we present next a reformulation of the Hilbert’s tenth problem. The reformulation is made possible since physical theories in general, and Quantum Mechanics in particular, have enjoyed the support and rigour of mathematical languages. We wish to stress here that, in spite of the inspiration, the connection is established entirely in the domain of Mathematics; we need not appeal to some real physical processes as we do with the proposed quantum algorithms above.

Mathematically, all we need to do is to sort out the instantaneous ground state \( |g\rangle \) among the infinitely many eigenvectors of \( \mathcal{H}(S) \) in (25); but this is a hard task. The trick we will use [24], as inspired by quantum adiabatic processes, is to tag the state \( |g\rangle \) by some other known state \( |g_I\rangle \) which is the ground state of some other operator \( H_I \) and can be smoothly connected to \( |g\rangle \) through some continuous parameter \( s \in [0,1] \). To that end, we consider the interpolating operator (25), rewritten as

\[
\mathcal{H}(s) = H_I + f(s)(H_P - H_I),
\]

\[\equiv H_I + f(s)W, \tag{33}\]

which has an eigenproblem at each instant \( s \),

\[
[\mathcal{H}(s) - E_q(s)]|E_q(s)\rangle = 0, \quad q = 0, 1, \cdots \tag{34}\]

with the subscript ordering of the sizes of the eigenvalues, \( |E_0(1)\rangle \equiv |g\rangle \), and \( f(s) \) not necessarily linear but some continuous and monotonically increasing function in \( [0,1] \)

\[
f(0) = 0; \quad f(1) = 1. \tag{35}\]

Clearly, \( E_0(0) = E_I \) and \( E_0(1) = E_g \). It turns out that for the linkage \( E_0(s) \) to connect a ground state to another ground state we require that

\[
[H_P, H_I] \neq 0. \tag{36}\]
The details are presented in [24] to arrive at
\[
\frac{d}{ds}|E_q\rangle = f'(s) \sum_{l \neq q}^{\infty} \frac{\langle E_l|W|E_q\rangle}{E_q - E_l}|E_l\rangle, \tag{37}
\]
\[
\frac{d}{ds}E_q(s) = f'(s)\langle E_q(s)|W|E_q(s)\rangle. \tag{38}
\]
Equations (37) and (38) form the set of infinitely coupled differential equations providing the tagging linkage as desired.

Analytical and numerical methods can now be employed to investigate the unknown ground state of \( H_P \) from the constructively known spectrum of \( H_I \) as the initial conditions. In this reformulation, the Diophantine equation in consideration has at least one integer solution if and only if
\[
\lim_{s \to 1} E_0(s) = 0. \tag{39}
\]
The limiting process is necessary here since \( H_P \), i.e. \( H(1) \), will have a degenerate spectrum because of certain symmetry (\( H_P \) commutes with \( a_i^\dagger a_i \)).

The equations above are infinitely coupled and cannot be solved explicitly in general. But we are only interested in certain information about the ground state. And since the influence on the ground state by states having larger and larger indices diminishes more and more thanks to the denominators in (37) (once no degeneracy is assured), this information may be derived, numerically or otherwise, with some truncation to a finite number of states involved. This may not work for all the Diophantine equations.

While the ground-state outcome for our differential equations might or might not be computable, it should be noted that there are instances [1, 34] where very simple differential equations, such as the wave equations, could have noncomputable outcomes because of ill-posed initial conditions.

With care we can slightly modify the derivation for (37,38) to come up with similar equations even when there is some degeneracy in \([0, 1]\). But for the condition (39) to be the indicator for the existence of solutions of the Diophantine equation, simple topological consideration only requires that the initial ground state \( |E_0(0)\rangle = |g_I\rangle \) is not degenerate and that this state does not cross with any other state in the open interval \( s \in (0, 1) \). With the freedom of choice for \( H_I \) satisfying (36), we should be able to eliminate any symmetry in the open interval \( s \in (0, 1) \) for \( \mathcal{H}(s) \) in order to have a stronger condition of a totally avoided crossing.

9 Concluding remarks

We have emphasised the issue of computability in principle, not on that of computational complexity. This attempt of broadening the concept of effective computability, taking into account the quantum mechanical principles, has been argued to be able in principle to decide some of the classical undecidables, the Hilbert’s tenth problem and thus the Turing
halting problem in this instance. If the quantum algorithm is realisable, and we don’t have any evidence of fundamental nature to the contrary, the Church-Turing thesis for effective computability should be modified accordingly.

On the other hand, if for any reasons the algorithm is not implementable in principle then it would be an example of information being limited by physics, rather than by logical arguments alone. That is, there might be some fundamental physical principles, not those of practicality, which prohibit the implementation. For example, a precision of, say, $10^{-20}$ would be required but that is not allowed in principle, and not just because of present-day limitation on technology. Or, there might not be enough physical resources (ultimately limited by the total energy and the lifetime of the universe) to satisfy the execution of the quantum algorithm. (In this case it is likely that the Turing program under consideration, even if it eventually halts upon some input, would take a running time longer than the lifetime of the universe.) In either case, the whole exercise is still very interesting as the unsolvability of those problems and the limit of mathematics itself are also dictated by physical principles and/or resources.

That some generalisation of the notion of computation could help solve the previous undecidability/noncomputability has been recognised before in mathematics. But this has not been believed possible until now simply because of the non-recognition of quantum physics as the missing ingredient. Our quantum algorithm could in fact be regarded as an infinite search through the integers in a finite amount of time, the type of search required to solve the Turing halting problem.

It should be emphasised again here that not only the values of some observables are measured but so are the probability distributions of these values. These measurable probability distributions are also to be compared against those computable from the theory of Quantum Mechanics. This helps identify the ground state, and thus the decision result, and is the pivotal element of our algorithm to separate quantum computability from Turing computability.

In doing so, we have to assume that Nature is describable by Quantum Mechanics correctly at least to the precision required. If this is not the case, testing a known Diophantine equation might yield some evidence for the failure of Quantum Mechanics as a theory of Nature. On the other hand, if there are any physical principles (or limitation of ultimate physical resources) which forbid the physical implementation then the limits of computability would be set also by Physics, and not by Mathematics solely.

To understand quantum computability and its limits, we have also considered, with mixed results, some problems generalised from the Hilbert’s tenth and the evaluation of Chaitin’s $\Omega$ number. Also in need of further investigation for our algorithm are the effects of errors in the implementation of hamiltonians, and of decoherence, and of measurement on the final outcome of quantum computation.

Our study is an illustration of “Information is physical”. Very recently there also exist in the literature some efforts to explore mathematical computability in the framework of Physics [37, 14, 8].

Inspired by Quantum Mechanics, we have then reformulated the question of solution existence of a Diophantine equation into the question of certain properties contained in
an infinitely coupled set of differential equations. In words, we encode the answer of the
former question into the smallest eigenvalue and corresponding eigenvector of a hermitean
operator bounded from below. And to find these eigen-properties we next deform the
operator continuously to another operator whose spectrum is known. Once the deformation
is also expressible in the form of a set of coupled differential equations, we could now start
from the constructive knowns as a handle to study the desired unknowns.

Note that the reformulation is entirely based on mathematics. If a general mathematical
method could be found to extract the required information from the differential equations
for any given Diophantine equation then one would have the solution to the Hilbert’s tenth
problem itself. This may be unlikely but not be as contradictory as it seems –because
the unsolvability of the Hilbert’s tenth problem is only established in the framework of
integer arithmetic and in Turing computability, not necessarily in Mathematics in general.
Tarski [39] has shown that the question about the existence of real solutions of polynomials
over the reals is, in fact, decidable.

It seems appropriate to end here with a quotation from the man whose famous Incompleteness result has often been misquoted as spelling the end for computability. In Gödel’s own words [18]: “... On the other hand, on the basis of what has been proved so far, it remains possible that there may exist (and even be empirically discoverable) a theorem-proving machine which in fact is equivalent to mathematical intuition, but cannot be proved to be so, nor even be proved to yield only correct theorems of finitary number theory.” Perhaps, quantum computation is that possibility.

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