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How Hard Is It to Detect Surveillance? A Formal Study of Panopticons and Their Detectability Problem

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Abstract: The Panopticon (which means “watcher of everything”) is a well-known prison structure of continuous surveillance and discipline studied by Bentham in 1785. Today, where persistent, massive scale, surveillance is immensely facilitated by new technologies, the term Panopticon vaguely characterizes institutions with a power to acquire and process, undetectably, personal information. In this paper we propose a theoretical framework for studying Panopticons and their detectability status. We show, based on the Theory of Computation, that detecting Panopticons, modelled either as a simple Turing Machine or as an Oracle Turing Machine, is an undecidable problem. Furthermore, we show that for each sufficiently expressive formal system, we can effectively construct a Turing Machine for which it is impossible to prove, within the formal system, its Panopticon status. Finally, we discuss how Panopticons can be physically detected by the heat they dissipate each time they acquire, effortlessly, information in the form of an oracle and we investigate their detectability status with respect to a more powerful computational model than classical Turing Machines, the Infinite Time Turing Machines (ITMs).

Keywords: formal methods; security; privacy; undecidability; panopticon; turing machine; oracle computations; irreversible computations; Infinite Time Turing Machines (ITMs)

1. Introduction

In 1785, the English philosopher and social theorist Jeremy Bentham (see [1]) described an unprecedented institutional punishment establishment, the Panopticon. The architecture of this establishment consisted of a circular building dominated by an “observation tower” in the center of which a single guard was continuously watching the inmates, imprisoned in cells arranged around the circular building. Moreover, the prisoners, themselves, could never be able to see the inspector, who remained for ever “invisible” to them. In the 70s, Foucault studied, deeply, Bentham’s concepts, pointing to the Panopticon as a generic power model impacting people’s everyday life. In his own words: “a figure of political technology that may and must be detached from any specific use” [2].

Nowadays, technological achievements have given rise to new, non-physical (unlike prisons), means of constant surveillance that transcend physical boundaries. This fact necessitates a revision of the Panopticon concept to reconcile it with the current technological progress and its, virtually unlimited, surveillance potential. Thus, the Panopticon behaviour is now manifested in the countless Internet services that accumulate people’s personal information, in the ubiquitous portable devices that contain massive and often...
sensitive information about their owners as well as the Internet giants to which billions of people trust their digital data.

Based on this observation, in this paper we investigate the question of detecting Panopticons, that is malicious entities whose mission is massive surveillance and covert information acquisition. Our approach follows the paradigm of the pioneering work of Cohen (see [3,4]) in computer viruses and, in general, computer malware. The breakthrough idea of Cohen was to model the vague, then, concept of a “computer virus” in a formal way so that the question of whether computer viruses can be detected algorithmically, i.e., by automated detection tools, can be posed precisely and unambiguously so that it is amenable to a formal analysis and resolution. Cohen modelled computer viruses using the most general, formal, model for an entity whose behaviour is determined through the execution of a sequence of elementary, or mechanical, steps: the Turing Machine (proposed in the pioneering work of Alan Turing in [5,6]). This idea, not only allowed a formal and, thus, unambiguous, definition of what computer viruses are but allowed Cohen to deploy the rich theoretical framework of the Theory of Computation for the analysis of whether computer viruses can be detected algorithmically or not (see, e.g., [7] for a comprehensive treatment of Turing Machines and the fundamental results of the Theory of Computation). To the best of our knowledge, our work is the first one which targets a realistic, formal, definition of the vague concept of Panopticons, as well as the formal analysis of their detectability properties based on the Turing Machine formalism and the fundamental results of the Theory of Computation. In other words, although all understand, intuitively, what a Panopticon is, i.e., a massive surveillance entity, there is no, to the best of our knowledge, formal definition of Panopticons which transforms this intuition into a formal framework for the definition and analysis of their properties. Our work is a first step towards this direction.

Therefore, along the same lines of thought as Cohen, in this paper we investigate the problem of whether Panopticons can be, algorithmically, detected. We formally model Panopticons, as Cohen did with computer viruses, as special types of Turing Machines and we analyze their detectability properties using results from the Theory of Computation. More specifically, we provide two different, but not unrealistic, Turing Machine based theoretical models of a Panopticon and show that there is no algorithm, i.e., no systematic procedure, that can detect all Panopticons that fall under these two definitions. In other words, detecting Panopticons, at least the ones that fall under these two plausible definitions, is an undecidable problem, in principle.

More specifically, the first formal model we examine studies Panopticons whose Panopticon properties are manifested through the execution of states (in the Turing Machine terminology) or actions (e.g., specific programming instructions such as reading from a disk file) that belong to a specific set of states that characterizes Panopticons of this type. This model is based on the formal computer virus model proposed by Cohen in [3,4]. In some sense, since the focal point of this model is the execution of states or actions of a particular type, the model captures the visible behaviour of the Panopticon, according to the actions it performs, and, thus, we call this model behavioural.

The second formal model focuses on the impact or consequences of the actions of the Panopticon and not the actions themselves. In particular, this model captures an essential characteristic of Panopticons, that of acquiring, rather, effortlessly information through surveillance and eavesdropping. We model this characteristic using Oracle Turing Machines. This concept was proposed by Turing in [5,6] to investigate the undecidability status of problems whenever they receive free information (in the form of an oracle) about some other undecidable problems. The Oracle, in the Panopticon context, models information acquired “for free” based on surveillance (observations) and eavesdropping actions, without requiring computational effort. This model is, in some sense, based on the information that a Panopticon deduces using “free” information and, thus, we call it deductive. Essentially, this model focuses on the semantics of a Turing Machine, i.e., outcomes of operation, while the first model focuses on the syntax, i.e., definition, of a Turing Machine. We extend a particular result from [8] (also proved in [7]) in order to show that this class of Panopticons...
can, not only cannot be detected, in general, but belong to a higher undecidability class in the hierarchy of undecidable problems which has important consequence for restricted types of Panopticons. For instance, restricted (in some sense) Panopticons of the first formal model we examine are detectable while similarly restricted types of Panopticons of the second formal model remain, still, undetectable.

Furthermore, based on a fundamental result from [9], we show that for any formal system, we can construct a Turing Machine whose Panopticon status, under the second formal model, cannot be proved within the formal system. That is, no proof can be produced by the formal system that this Turing Machine is a Panopticon and no proof that it is not a Panopticon. In other words, given any formal system, one can provide a procedure that generates a Turing Machine for which it is impossible to decide whether it is a Panopticon or not within the formal system.

In some final, rather philosophical, considerations we discuss, based on the exposition in [10], how Panopticons’ power of obtaining information instantaneously and effortlessly merely by observing, renders them detectable through the heat dissipated in their environment as a result of the irreversible action of obtaining this information as an oracle reply without performing any computational steps. Finally, we show how oracle-based Panopticons are easy to be detected using hypercomputation based on Infinite Time Turing Machines (ITTMs) based on concepts and results from [11]. Although hypercomputation is considered unrealistic, the concept can, nevertheless, provide an interesting context within the super-power Panopticons can be compared with super-power Turing Machines, studying how well super-power TMs fare against super-power Panopticons.

2. Definitions and Notation

Since our theoretical model for the Panopticon considers it as a computational entity which is what actually happens in practice in active massive surveillance based on information technologies, we briefly state the relevant definitions and notation from the computation theory that will be used in the subsequent sections (see [7]).

**Definition 1 (Turing Machine(s)).** A Turing machine is a septuple

\[ M = (Q, \Sigma, \Gamma, \delta, q_0, B, F) \]

where \( Q \) is a finite set of normal operation states, \( \Gamma \) is a finite set called the tape alphabet, where \( \Gamma \) contains a special symbol \( B \) that represents a blank, \( \Sigma \) is a subset of \( \Gamma - \{B\} \) called the input alphabet, \( \delta \) is a partial function from \( Q \times \Gamma \) to \( Q \times \Gamma \times \{L, R\} \) called the transition function, \( q_0 \in Q \) is a distinguished state called the start state, \( F \subset Q \) is a set of final states.

Notation-wise, given \( M \) we denote by \( \langle M \rangle \) its code, i.e., an encoding of its description elements as stated in Definition 1 using any fixed alphabet, usually the alphabet \( \{0, 1\} \) (binary system). The details can be found in, e.g., [7,12] but they are inessential for our arguments.

One of the main outcomes of Turing’s pioneering work [5] was that there exist problems that Turing Machine(s) cannot solve. The first, such, problem was the, so called, Halting problem (see, also, [13] for an excellent historic account):

**The Halting Problem**

**Input:** A string \( x = \langle M, w \rangle \) which is actually the encoding (description) of a Turing machine \( \langle M \rangle \) and its input \( w \).

**Output:** If the input Turing machine \( M \) halts on \( w \), output True. Otherwise, output False.

The language corresponding to the Halting problem is \( L_u = \{ \langle M, w \rangle \mid w \in L(M) \} \).

In other words, the language \( L_u \) contains all possible Turing machine-input pair encodings \( \langle M, w \rangle \) such that \( w \) is accepted by \( M \). This is why \( L_u \) is also called universal language since the problem of deciding whether a given Turing machine \( M \) accepts a given input \( w \).
is equivalent to deciding whether \( \langle M, w \rangle \in L_u \). The language \( L_u \) was the first language proved to be non-recursive or undecidable by Turing in his seminal work [5].

In order to discuss Panopticons, we need an important variant of Turing Machine(s), called oracle Turing Machines. Such a machine has a special tape on which it can write queries to which they obtain the answer instantaneously in one step, no matter what query it is. This type of Turing Machines was, first, discussed, briefly, by Turing himself in [6] under the name \( O \)-machine. Post developed further this concept in a series of papers [14–16] and his collaboration with Kleene in [17] resulted to the definition that is used today in the Theory of Computation.

Below, we give a formal definition of an Oracle Turing Machine:

**Definition 2 (Oracle Turing Machine).** Let \( A \) be a language, \( A \subseteq \Sigma^* \). A Turing machine with oracle \( A \) is a single-tape Turing machine with three special states \( q_? \), \( q_y \) and \( q_n \). The special state \( q_? \) is used to ask whether a string is in the set \( A \). When the Turing machine enters state \( q_? \) it requests an answer to the question: “Is the string of non-blank symbols to the right of the tape head in \( A \)?” The answer is provided by having the state of the Turing machine change on the next move to one of the states \( q_y \) or \( q_n \). The computation proceeds normally until the next time \( q_? \) is reached, at which point the Turing machine requests another answer from the oracle.

With respect to notation, we denote by \( M^A \) the Turing machine \( M \) with oracle \( A \). Also, a set (language) \( L \) is recursive with respect to \( A \) if \( L = L(M^A) \) for some Turing machine \( M^A \) that always halts while two oracle sets (languages) are called equivalent if each of them is recursive in the other (see [7]).

### 3. The Panopticon Detection Problem and Our Approach

Formal proofs about the impossibility of detecting malicious computation-based entities already exist for a long time for a very important category of such entities, the computer viruses or malware in general.

In Cohen’s pioneering work (see [3,4]) a natural, formal, definition of a virus is provided based on Turing Machine(s). Specifically, Cohen defined a virus to be a Turing machine that simply injects its transition function into other Turing Machine(s)’ transition functions (see Definition 1) replicating, thus, itself indefinitely. Then, he proves that \( L_u \) reduces to the problem of deciding whether a given Turing Machine behaves in this way proving that detecting viruses is an undecidable problem.

Following Cohen’s paradigm, we will propose two definitions that precisely describe the operation of a computation-based Panopticon system. In the context of the first model, a Panopticon is a Turing machine that when executed will demonstrate a specific, recognizable, behaviour particular to Panopticons manifested by the execution of a sequence of actions. For instance, it will publish secret information about an entity, it will download information illegally etc., that is it will take actions that are reflected by actually reaching, during its operation, particular states in a set \( Q_{\text{pan}} \) which contains malware behaviour states.

However, an objection to this rationale can be posed on the grounds that one can scan the transition function (i.e., program code) of a Turing Machine and if a malware state is detected on the right-hand side of \( \delta \) then the Turing Machine can be declared as malware or Panopticon in our case. That is, one can propose that the mere existence of malware states, but not their actual execution, is sufficient to declare a particular Turing Machine as malware or Panopticon. There, is, however a plausible reply to this objection. First of all, the set \( Q_{\text{pan}} \) may be unknown and, thus, no particular state of the Turing Machine can be identified as belonging in this (unknown) set of states. In other words, it may not be known which subset of the states of a Turing Machine description actually belong in \( Q_{\text{pan}} \) until they are invoked and manifest, themselves, as malicious actions through their execution and actual negative impact on the Turing Machine’s environment. On the other hand, let us assume that \( Q_{\text{pan}} \) is known. Then one can now, indeed, locate a Panopticon state (i.e., a state in the known, set \( Q_{\text{pan}} \)) in the right-hand side of \( \delta \) of a Turing Machine, if it exists.
However, this Turing Machine is not known whether it will ever operate as a Panopticon and perform the threatening actions represented by the states in $Q_{\text{pan}}$. The argument of simply locating states in $Q_{\text{pan}}$ to detect Panopticons is similar to an argument of declaring a Turing machine as halting simply because on the right-hand side of its transition function it contains a halting state. To be characterized as “halting”, the Turing Machine must, actually, invoke a halting state.

Now, based on the discussion above, we define formally the behavioural Panopticons:

**Definition 3** (Behavioural Panopticons). A Behavioural Panopticon is an octuple

$$M = (Q, Q_{\text{pan}}, \Sigma, \Gamma, \delta, q_0, B, F)$$

where $Q$ is a finite set of normal operation states, $\Gamma$ is a finite set called the tape alphabet, where $\Gamma$ contains a special symbol $B$ that represents a blank, $\Sigma$ is a subset of $\Gamma - \{B\}$ called the input alphabet, $\delta$ is a partial function from $Q \times \Gamma$ to $Q \times \Gamma \times \{L, R\}$ called the transition function, $q_0 \in Q$ is a distinguished state called the start state, $F \subset Q$ is a set of final states, and $Q_{\text{pan}} \subset Q$, $Q_{\text{pan}} \cap F = \emptyset$, is a distinguished set of states linked to Panopticon behaviour. We assume that transitions from states in $Q_{\text{pan}}$ do not change the Turing Machine’s tape contents, i.e., they are purely interactions with the external environment of the Turing Machine and can affect only the environment.

This is much like Cohen’s definition of a virus since it characterizes Panopticons according to their displayed or manifested behaviour. We stress the word interactions in order to preclude situations where a false alarm is raised for “Panopticons” which merely list actions that are characteristic of Panopticon behaviour without ever actually invoking them during their operation. Instead, they operate normally without any actions taking place that manifest Panopticon behaviour.

Beyond displayed behaviour, Panopticons can be reasonably assumed to also possess deductive powers, not directly visible or measurable. In other words, one type of such Panopticons may operate by gathering or computing totally new information, distinct from the information already known to it. We model this type of Panopticon with the language $S'_{1}$ defined later in this Section. Moreover, another type of Panopticons can take advantage of easily acquired, or even stolen, freely provided (in some sense) information. In other words, based on information the Panopticon acquires for free by eavesdropping or data breaching, for instance, it deduces further information, perhaps expending some computational effort this time. We model the characteristic Panopticon action, i.e., observation, surveillance and illicit information gathering, using oracle Turing Machines, where the freely acquired or eavesdropped information for particular targets is modelled by the oracle set of the machine. Based on this information, the Turing machine deduces, through its normal computation steps, further information about its targets. This type of Panopticon is modelled with the language $S'_{2}$ defined later in this Section. Below, we describe both types of Panopticons, the ones based on $S'_{1}$ and the ones based on $S'_{2}$ since their common characteristic is the deduction of new information from already known information.

**Definition 4.** (Deductive Panopticons) A Deductive Panopticon is a Turing Machine that either by itself (language $S'_{1}$) or based on observed or stolen information acquired without expending any computational effort (language $S'_{2}$), deduces (perhaps with computational effort) further information about entities.

In the definition above, the Panopticon operating by itself, i.e., without oracles (language $S'_{1}$), is weaker (as we will show in what follows) than the one with oracles (language $S'_{2}$) since the latter is allowed to obtain free advice or information, in the form of an oracle.
Naturally, many other deductive Panopticon definitions would be reasonable or realistic. Our main motivation behind the ones stated above was a balance of theoretical simplicity and plausibility in order to spark interest on the study on formal properties of Panopticons as well as the difficulty of detecting them algorithmically.

Based on the two Panopticon definitions we gave above, we can define the corresponding Panopticon detection problems. The aim of a Panopticon detection algorithm or Turing machine, is to take as input the encoding of another Turing machine and decide whether it is Panopticon or not based on the formal definition.

The Deductive Panopticon Detection Problem 1

**Input:** A description of a Turing machine (program).
**Output:** If the input Turing machine operates like a Panopticon according to Definition 3 output True. Otherwise, output False.

More formally, if by $L_b$ we denote the language consisting of Turing machine encodings $<M>$ which are Panopticons according to Definition 3, then we want to decide $L_b$, i.e., to design a Turing machine that, given $<M>$, decides whether $<M>$ belongs in $L_b$ or not.

The Deductive Panopticon Detection Problem 2

**Input:** A description of a Turing machine (program).
**Output:** If the input Turing machine operates like a Panopticon according to Definition 4 output True. Otherwise, output False.

Essentially, the deductive Panopticon detection problem 2 asks to decide the languages $S_1'$ and $S_2'$.

Our approach is different for each of the two Panopticon models we propose since they are of a different nature, i.e syntactic (for the behavioural model) vs. semantic (for the deductive model). For the behavioural model, we provide a simple adaptation of Cohen’s pioneering formal model of a virus and prove a Panopticon detection impossibility result much like Cohen’s result for virus detection. For the deductive model, we follow a completely different approach using Oracle Turing Machines and a technique that can be applied to prove undecidability results for this type of machines.

More specifically, in Chapter 8 of [7] a technique from [8] is presented that establishes a hierarchy of undecidable problems for Oracle Turing Machines. In particular, the technique targets the oracle set $S_1 = \{<M> | L(M) = \emptyset \}$, with $<M>$ denoting the encoding of Turing machine $M$. Then, the sets $S_{1+1} = \{<M> | L(M^{S_1}) = \emptyset \}$ can be, recursively, defined and the following can be proved (see [7,8]):

**Theorem 1.** The membership problem for TM’s without oracles is equivalent to $S_1$ (i.e., $L_u$ is equivalent to $S_1$).

**Theorem 2.** The problem of deciding whether $L(M) = \Sigma^*$ is equivalent to $S_2 = \{<M> | L(M^{S_1}) = \emptyset \}$.

Our first contribution is to propose a plausible Panopticon model which incorporates the information deduction characteristic of a Panopticon (see Definition 4). Information deduction takes place whenever the Turing machine under scrutiny produces a completely new information set given a set of fixed, finitely many, already known information sets. This set models the information that the Panopticon already knows through surveillance and observation, without (usually) expending considerable effort since it, merely, intercepts or eavesdrops information.

More formally, let $N_i = \{L_{i1}, L_{i2}, \ldots, L_{ik}\}$ be a set of recursively enumerable languages, for some fixed integer $k \geq 1$, such that $\emptyset \notin N_i$ for all $i$. Also, let $M_{i1}, M_{i2}, \ldots, M_{ik}$ the Turing Machine(s) that, correspondingly, accept these languages. These Turing Machine(s) and their corresponding languages model the fixed information sets already known to the Panopticon. We, also, say that a set is disjoint from a collection of sets
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Theorem 3. The language \( L_b \) is undecidable.

Proof. Let \(<M,w>\) be an instance of the Halting problem. We will show how we can decide whether \(<M,w>\) belongs in \( L_b \) or not using a hypothetical decision procedure (Turing machine) for the language \( L_b \). In other words, we will show that \( L_b \) is recursive in \( L_b \).

Given \(<M,w>\), a set \( Q_{\text{pan}} \) of states, \( Q_{\text{pan}} \cap Q = \emptyset \), where each state is related with actions that manifest Panoptic behaviour, is embedded into \( M \). We design a Turing machine \( M_{\text{con}} \) that it gets \( M \) as input and modifies the transition function (see Definition 1) of \( M \) so as when a final state is reached (i.e., a state in the set \( F \) of \( M \)) a transition takes place that essentially starts the execution of the actions related to some state of \( Q_{\text{pan}} \). In a sense, \( M_{\text{con}} \) produces now a new Turing machine \( M' \) containing the actions of \( M \) followed by actions (any of them) corresponding to the states in \( Q_{\text{pan}} \). Now, \( M' \) is given as input the input of \( M \), i.e., \( w \), and operates as described above.

Let us assume that there exists a Turing machine \( M_b \) that decides \( L_b \) and suppose that \( M_b \) answers that \( M' \in L_b \) when \( M' \) is given as input to \( M_b \). Since a state in \( Q_{\text{pan}} \) was finally activated, as \( M_b \) decided, this implies that \( M \) halted on \( w \) since \( M' \) initially simulated \( M \) on \( w \). Then we are certain that \( M \) halts on \( w \).

Assume, now, that \( M_b \) decides that \( M' \) is not a Panoptic, i.e., \( M' \notin L_b \). Then a state in \( Q_{\text{pan}} \) was never reached, which implies that no halting state is reached by \( M \) on \( w \) since
The Halting Problem for Turing Machine(s) without oracles, i.e., \( L \). Thus, \( M \) and, thus, we have shown that \( L_u \) is recursive in \( L_b \).

However, there is a catch that invalidates this reasoning: if \( M \) itself can exhibit the Panopticon behaviour, i.e., it already contains \( Q_{\text{pan}} \) and it can reach a state in \( Q_{\text{pan}} \) before reaching a final state. This means that Panopticon behaviour can be manifested without ever \( M \) reaching a final state that would lead \( M \) to invoke a Panopticon state in \( Q_{\text{pan}} \), by its construction. This situation can be solved if we create a new set of dummy ("harmless" or "no-operation") "Panopticon" states \( Q'_{\text{pan}} \) which contains a new state for each of the states in \( Q_{\text{pan}} \). Then we replace the states from \( Q_{\text{pan}} \) that appear in the transition function of \( M \) with the corresponding states in \( Q'_{\text{pan}} \). This new version of \( M \) is given to \( M_{\text{con}} \) to produce \( M' \) as mentioned previously. Actually, this transformation removes from a potential Panopticon the actions that if executed would manifest a Panopticon. We stress, again, the fact the mere existence of Panopticon actions is not considered Panopticon behaviour.

With this last transformation, \( M' \) is a Panopticon if and only if \( M \) halts on \( w \) and, thus, \( L_u \) is recursive in \( L_b \). □

5. Impossibility of Detecting Deductive Panopticons

In the following two theorems, we prove the undecidability of \( S'_1 \) and \( S'_2 \). Although their undecidability follows, directly, from Rice’s Theorem (see [7,18]), the proofs we give below provide more insightful information as they place \( S'_2 \) in a higher undecidability level than \( S'_1 \) (see, also, the discussion about the Analytical Hierarchies and the Analytical Hierarchies in Section 7.3).

We, first, prove the undecidability of \( S'_1 \), i.e., the impossibility of deciding for a given Turing machine (its encoding, to be precise) whether it accepts a language disjoint from a given, fixed, finite set of languages. In other words, it is impossible to detect Turing Machine(s) that decide, perhaps with effort, new information sets given some known ones.

Theorem 4. The Halting Problem for Turing Machine(s) without oracles, i.e., \( L_{u} \), is equivalent to \( S'_1 \).

Proof. We first prove that given an oracle for \( S'_1 \) we can recognize \( L_u \). We construct a Turing machine \( M_{S'_1} \) such that given \( \langle M, w \rangle \) constructs a Turing machine \( M' \) which operates as follows. It ignores its input and simulates, internally, \( M \) on \( w \). \( M' \) accepts its input if \( M \) accepts \( w \) which means that \( L(M') = \Sigma^* \) otherwise, i.e., if \( M \) does not accept \( w \) then \( M' \) does not accept its input and \( L(M') = \emptyset \). If the answer is no, then \( M_{S'_1} \) asks the oracle whether \( < M' > \in S'_1 \). If the answer is yes, i.e., \( L(M') = \emptyset \), then \( M \) does not accept \( w \). We, thus, can recognize \( L_u \).

Now, we show that given an oracle for \( L_u \) we can recognize \( S'_1 \). We will construct a Turing machine \( M' \) such that, given \( M \), it constructs another Turing machine \( M' \) that operates as follows. \( M' \) ignores its own input initially, and uses a generator of triples \( (i, j, l) \), \( 1 \leq l \leq k+1 \), for simulating the \( i \)th Turing machine, \( M_i \), with \( M_{k+1} = M \), on the \( i \)th string lexicographically constructed for \( j \) steps (the Turing machine \( M_i \) runs on the \( i \)th string for \( j \) steps i.e., it applies the transition function \( j \) times). The triples are generated in increasing order of the sum \( n = i + j + l \) of their components and for triples of equal component sum, in increasing \( i \), then in increasing \( j \) (if the \( i \) components are equal), and finally in increasing \( l \) (if the \( i \) and \( j \) components are equal). Each time one of the Turing Machine(s) \( M_1, M_2, \ldots, M_k \) accepts a particular input, this input is recorded on \( M' \)'s second tape. Each time \( M_{k+1} \) accepts an input (i.e., \( M \) accepts an input), it is also recorded on \( M' \)'s second tape separately from the inputs accepted by \( M_1, M_2, \ldots, M_k \). Then, \( M' \) checks (using the recorded inputs stored on its second tape) whether this \( M_{k+1} \) input, or one accepted previously by \( M_{k+1} \), has been accepted by one of the \( M_1, M_2, \ldots, M_k \). If no, the process continues. If yes, \( M' \) stops the simulation and accepts its own input that was initially ignored. Thus, \( < M > \in S'_1 \) if \( L(M') = \emptyset \) since this means that \( L(M) \) is disjoint from
would be infinite. By definition, all valid computations conclude in a halting state (see [7,8] for details). In other words, the oracle replied that \( M \) accepts its input if and only if there is a string \( x \) that operates in the Panopticon.

Theorem 5. The problem of deciding whether \( L(M) = \Sigma^* \) is equivalent to \( S'_2 \).

Proof. We first show that deciding whether \( L(M) = \Sigma^* \) is recursive in \( S'_2 \). We construct a Turing machine \( \hat{M}^{S'_1} \) that takes as input a Turing machine \( M \) and constructs from it a Turing machine \( M^{S'_1} \), that is a Turing machine with oracle set \( S'_1 \) that operates in the following way. It enumerates strings \( x \) over the alphabet \( \Sigma \) and for each such string it uses oracle \( S'_1 \) in order to decide whether \( M \) accepts \( x \). This can be accomplished by constructing \( M' \) which ignores its input and simulates \( M \) on \( x \). If \( M \) accepts \( x \) then \( M' \) accepts its input which means that \( L(M') = \Sigma^* \) while \( L(M') = \emptyset \) if \( M \) does not accept any \( x \). Then, \( \hat{M}^{S'_1} \) asks the oracle whether \( M' \in S'_1 \). If the answer is yes, which means that \( M \) does not accept any \( x \), then \( \hat{M}^{S'_1} \) accepts its input.

Thus, \( \hat{M}^{S'_1} \) accepts its own input if and only if there is a string \( x \) not accepted by \( M \). Consequently,

\[
L(\hat{M}^{S'_1}) = \begin{cases} \emptyset, & \text{if } L(M) = \Sigma^* \\ \Sigma^*, & \text{otherwise.} \end{cases}
\]

Now \( \hat{M}^{S'_2} \) asks its oracle \( S'_2 \) whether \( \hat{M}^{S'_1} \in S'_2 \), i.e., whether \( L(\hat{M}^{S'_1}) \) is disjoint from all sets in \( N_2 \). If the answer is yes, then \( L(\hat{M}^{S'_1}) = \emptyset \) and consequently, \( L(M) = \Sigma^* \). If the answer is no, on the other hand, then \( L(\hat{M}^{S'_1}) = \Sigma^* \) and, thus, \( L(M) \neq \Sigma^* \). Thus, deciding whether \( L(M) = \Sigma^* \) is recursive in \( S'_2 \).

We now turn to showing that \( S'_2 \) is recursive in the problem of whether \( L(M) = \Sigma^* \). In other words, if by \( L \) we denote the codes of the Turing Machine(s) which accept all their inputs, then we will prove that there exists a Turing machine \( \hat{M}''^{L'} \), i.e., a Turing machine with oracle set \( L' \), that recognizes \( S'_2 \).

Given a Turing machine \( M^{S'_1} \), we define the notion of a valid computation of \( M^{S'_1} \) using oracle \( S'_1 \) in a way similar to the notion defined in [7,8]. A valid computation is a sequence of Turing Machine step descriptions, called Instantaneous Descriptions or ID, such that the next one follows from the current one after a computational (not oracle query) step, according to the internal operation details (i.e., transition function or program) of the Turing machine. Roughly, an ID describes fully the status of a Turing Machine computation at each time step, containing information such as tape contents, head position, and current state.

However, if a query step is taken, i.e., the Turing machine \( M^{S'_1} \) enters state \( q_a \), and the next state is \( q_1 \), this means that \( M^{S'_1} \) submitted a query to the oracle \( S'_1 \) with respect to whether some given Turing machine, say \( T \), belongs to the set \( S'_1 \), receiving the answer \( \emptyset \). In other words, the oracle replied that \( < T > \notin S'_1 \), or, equivalently, \( L(T) \) is not disjoint from all sets in \( N_1 \). As evidence for the correctness of this reply from the oracle, we substitute the query step with a valid computation of the ordinary (i.e., with no oracle) Turing machine \( T \) that shows that a particular string from a language in \( N_1 \) is, also, accepted by \( T \). If, however, after \( q_1 \) the state \( q_4 \) follows, no computation is inserted. Intuitively, such a computation would be infinite. By definition, all valid computations conclude in a halting, i.e., acceptance state (see [7,8] for details).

We, now, describe the operation of \( \hat{M}''^{L'} \) with \( < M^{S'_1} > \) as input. Given \( M^{S'_1} \), \( \hat{M}''^{L'} \) constructs a Turing machine \( \hat{M} \) which accepts all computations of \( M^{S'_1} \) which show that they are not a Panopticon. We call these computations non-Panopticon computations and they are of two disjoint types: (i) invalid computations, i.e., computations which contain invalid successions of IDs, and (ii) unsuccessful computations, i.e., computations which, although not invalid, they demonstrate that \( M^{S'_1} \) is not a Panopticon.
Given a formal system, Theorem 7. Let \( T \) be a Turing machine over \( \Sigma \) and let \( M \) be a set of corresponding Turing Machines. We will reduce the problem of deciding whether \( M \) exists in \( F \), to the problem of deciding whether \( T \) halts on a given input string.

Proof. We will reduce the problem of deciding whether \( T \) halts on a given input string to the problem of deciding whether \( M \) is a Panopticon.

Based on Definition 4 of a Panopticon (see, also, Sections 3 and 5 for the formal details of the definition), Theorem 6 and the Turing Machine \( M_F \), we now prove the following:

Theorem 6. Given a formal system \( F \), we can construct a Turing Machine \( M_F \) for which no proof exists in \( F \) that it either halts or does not halt on a particular input.

Proof. We will reduce the problem of deciding whether \( M_F \) halts on a given input string to the problem of deciding whether \( M \) is a Panopticon.

Based on Definition 4 of a Panopticon (see, also, Sections 3 and 5 for the formal details of the definition), Theorem 6 and the Turing Machine \( M_F \), we now prove the following:

Theorem 7. Let \( F \) be a consistent formal system. Then we can construct a Turing Machine for which there is no proof in \( F \) that it behaves as a Panopticon and no proof that it does not behave as a Panopticon.
corresponding Turing Machine that recognizes it. Also, note that \( L \) is disjoint from the set \( \{L_1, L_2\} \) and, thus, the Turing Machines that accept this language are Panopticons.

Given these elements as well as the Turing Machine \( M_F \) and an input \( w \) on which we want to decide whether \( M_F \) halts or not, we proceed as follows. We construct a Turing Machine \( M_T \) which, given \( w \) as input, simulates \( M \) on \( w \) and \( M_F \) on \( w \), alternating between them. Then \( M_T \) accepts if either of them accepts at some step of the simulation process.

We, now, observe that \( L(M_T) = L(M) \), if \( M_F \) does not halt on \( w \) while \( L(M_T) = \Sigma^* \) if \( M_F \) halts on \( w \). Then, according to the definition of a deductive Panopticon, \( M_T \) is a Panopticon if and only if \( M_F \) does not halt on \( w \).

But, now, if a proof existed in the formal system \( F \) that \( M \), either is a Panopticon or it is not a Panopticon then the same proof could be used to prove that \( M_F \) either halts or does not halt, correspondingly, contradicting, thus, Theorem 6.

7. Philosophical Remarks on Panopticon Detectability

In this final section, we consider the problem of detecting oracle-based Panopticons through their physical properties and by Hypercomputation computational models, focusing on the Infinite Time Turing Machine (ITTM) model.

With respect to the physical detection, we discuss how the action of receiving information “for free” through the oracle reply, forces the Panopticon to perform an irreversible operation which, necessarily, dissipates heat to the environment, manifesting the Panopticon’s presence. A reversible operation of a Turing Machine implies that one can trace or rewind the computation back to the beginning, from any step of the evolution of computation or, equivalently, if the Turing Machine is found at a particular state, then there is only a unique predecessor state from which this state may have resulted. We base our discussion on the theory of the thermodynamics of computation and the fact that irreversible computational operations, necessarily, consume energy and dissipate heat to the environment (see [10,19]).

Moving to hypercomputation, we consider the power of Infinite Time Turing Machines with respect to detecting oracle-based Panopticons. We conclude that since the hypercomputation decidability powers transcend the Arithmetical Hierarchy and reach the first level of the Analytical Hierarchy (see [11]), simple oracle-based Panopticons are detectable by ITTM. Then we consider the question of defining ITTM-based Panopticons that evade the detectability power of ITTM.

7.1. Detectability of Panopticons by Their Thermal Emissions

In the preceding sections, we proved that, under a plausible formal definition of Panopticons as Oracle Turing Machines, detecting Panopticons by “normal” Turing Machines, i.e., Turing Machines without oracles, fully realizable and algorithmic in operation, is an undecidable problem. However, as we discuss in this section, it may still be possible to detect Panopticons or, more generally, Turing Machines deploying Oracles, based on non-algorithmic approaches, namely the theory of the thermodynamics of computation.

It is a well-known fact that any (non-oracle) Turing Machine can be transformed so as to operate, in a physical implementation, in a way that does not dissipate energy, i.e., it can be transformed so as to operate reversibly (see [19]). Reversibility in computation is important due to the fact that only operations that are irreversible necessarily consume energy and dissipate heat to their environment. Such irreversible operations include, for instance, the logical AND of two bits or resetting to a default state (e.g., to ‘0’) a 1-bit memory. However, as we stated above, in principle any Turing Machine, which normally operates in an irreversible way, can be transformed into an equivalent, as far as the computation is concerned, reversible Turing Machine (see [19] for the formal details of the transformation).

On the other hand, Oracle Turing Machines and Panopticons as we defined them in this paper, for that matter, include a step which cannot be performed reversibly. This step results from the consultation of the oracle when the Oracle Turing Machine enters the oracle query state. The oracle consultation cannot be replaced, in general, by the computation of a Turing
Machine that always halts and supplies the correct answer to the oracle query since the oracle may correspond to an undecidable problem such as the Halting Problem, as it is in our case. Moreover, this step is irreversible, by its nature since to a particular oracle query response there may correspond more than one (perhaps infinitely many) strings for which the response is correct. Thus, it is not possible to infer, uniquely, the query string, given the query response. More precisely, the oracle reply, *Yes* or *No*, is stored in a 1-bit memory as a 1 or 0 respectively. Since there is no connection with the query to which the reply was given, the memory can be considered as containing a *random* or *unknown* bit just before the reply is given to the next query. However, setting a previously unknown bit to either 0 or 1 is an irreversible operation since given the set value, it is impossible to infer the previous one. According to the thermodynamics of information, this operation consumes energy equal to \( k_B T \ln 2 \), where \( k_B \) is Boltzmann’s constant and \( T \) the environment temperature, and, thus, dissipates heat to the Panopticon’s environment (see, e.g., [10]). Therefore, this oracle consultation step, which is unavoidably irreversible, dissipates heat to the environment of the Panopticon and, thus, provides evidence of its existence and operation (in the form of increasing temperature in its environment), even though detecting Panopticons algorithmically, i.e., by non-Oracle Turing Machines, is an undecidable problem.

In some sense, an Oracle Turing Machine can be viewed as a type of Maxwell’s Demon. The reader is urged to consult the excellent collection of seminal papers in [20] on its history as well as impact on philosophy, information theory, and thermodynamics of computation. Similarly to Maxwell’s Demon, the Oracle Turing Machine involves itself in a cyclic process of (i) normal computation followed by an *observation*, in the form of submitting a query string to its oracle, which results to an instantaneous “observation” result, i.e., the reply of the oracle, and (ii) Storing the result of the observation (i.e., query reply, “YES” or “No”) on an 1-bit memory. As in the resolution of the tantalizing operation of Maxwell’s Demon which appears to perform useful work without consuming energy, apparently violating the *Second Thermodynamics Law*. In the final resolution of the “paradox”, after many failed attempts, Bennett showed (see [10,19]) that it was not the observation the process that should, necessarily, consume energy (it can be performed reversibly) but the operation of resetting of the 1-bit memory from an unknown value, to a specific value that reflects the result of the observation, 1 or 0 (i.e., “YES” or “NO”, as oracle replies in our case). Thus, *information theory* was linked, tightly, to *thermodynamics*. Likewise, the Panopticon operating as an Oracle Turing Machine does not betray its presence through the process of “observation”, i.e., invoking the oracle, but by manipulating the 1-bit memory that holds the oracle reply.

In summary, it appears that the very strength of Panopticons (defined as Oracle Turing Machines), which is their ability to “observe” everything and acquire knowledge without computational effort, can be their very weakness. This is due to the fact that this ability relies on irreversible operations (i.e., setting to either 0 or 1 an 1-bit memory which holds a *random* value that holds the result of the observation, or, query to the oracle) which dissipate energy and emit heat to the Panopticon’s environment revealing, in this way, the Panopticon’s existence and operation.

### 7.2. Infinite Time Turing Machine Hypercomputation Model

The Oracle Turing Machines (see Definition 2), or O-machines as defined by Turing, is an instance of a general class of theoretical computation models named *Hypercomputation* or *Super-Turing* (see [21,22]). Within the context of all these models, problems such as the Halting Problem can be solved, which cannot by solved by ordinary Turing Machines and, thus, are not confined by the Church-Turing Thesis.

In this section we focus on *Infinite Time Turing Machine* computational model or ITTM for short. A machine in this model operates much like an ordinary Turing Machine in the sense of Definition 1, as follows: each computational step is executed, in succession, after the previous one through the application of the transition function based on the contents of each tape cell and the current machine state. In this way, each configuration of...
the machine at computation step $t + 1$ results from the configuration of the machine at step $t$ according to the tape contents and the actions of the transition function, as it is the case for ordinary Turing Machines under Definition 1. The critical difference is that in contrast with Turing Machines which are required to produce the correct computation result after a finite number of steps (otherwise the computation proceeds indefinitely without producing any result) the machines in the ITTM model are allowed to compute and produce results using an infinity of steps. To avoid some important technicalities (see [11] for the details) with respect to how a result can be obtained within an infinite number of steps, the theory of ordinals is employed to define specific computation limit steps upon which results are obtained from a machine in the ITTM model. Before we continue, we should remark that this computational model has received harsh criticism (see, for instance, [13,23,24]). However, this model can be, nevertheless, proved to be a promising theoretical tool for studying the limits of Panopticism, at least when Panopticons are based on some formal theoretical computational model.

Ordinals can be used as “labels” of the order of the elements of a well-ordered set, i.e., a set in which every pair of its elements is comparable. The labeling is as follows: the smallest element of the set is labeled 0, the one coming after the smallest element is labeled 1, the next element 2 etc. The size of the set is equal to the least ordinal that does not constitute a label for a set element. This size is the order type of the set.

An ordinal can be, inductively, defined as the set of the preceding ordinals. For instance, the finite ordinal 20 is the order type (or size) of the ordinals that precede it, i.e., the finite ordinals 0 up to 19 which is the immediate predecessor of 20. Thus, the ordinal 20 is identified as the set $\{0, 1, \ldots, 19\}$. Conversely, any well-ordered set $S$ such as for each ordinal $a \in S$ and any $\beta < a$, it also holds that $\beta \in S$, can be identified with an ordinal.

Except the finite ordinals, which can be identified with the elements of the set $\mathbb{N}$ of natural numbers, there are also infinite ordinals, the smallest one being $\omega$, the order type of the natural numbers (which are the finite ordinals). The ordinal $\omega$, thus, can be identified with the set $\mathbb{N}$ of natural numbers.

In the terminology of ordinals, if an ITTM reaches a halting state within a number of steps equal to a finite ordinal, then the halting procedure is the one followed by an ordinary Turing Machine. However, whereas the computation may be impossible for an ordinary Turing Machine, which implies that for some input(s) it fails to halt within a finite number of steps, as it happens for the Halting Problem of any undecidable problem, for machine in ITTM the computation is considered to step on all finite ordinals, until it reaches the first infinite ordinal $\omega$ which is the first infinite halting step. Thus, after all finite ordinals, comes the first infinite ordinal $\omega$. Next, in succession, come the infinite ordinals $\omega + 1$, $\omega + 2$ etc. until we reach $\omega + \omega$ or $\omega \cdot 2$. Then come $\omega \cdot 2 + 1$, $\omega \cdot 2 + 2$ etc. until we reach $\omega \cdot 3$ which is $\omega + \omega + \omega$. In general, the infinite ordinals are written as $\omega \cdot n + m$, with $n, m$ natural numbers (or finite ordinals). With this set of ordinals we associate the ordinal $\omega^\omega$. Likewise, some ordinals which come next are $\omega \cdot 4$, and then later on $\omega \cdot 5$. Analogously, we have ordinals such as $\omega^3$, $\omega^4$, $\omega^5$, $\omega^\omega$, $\omega^{\omega^\omega}$ and at some point the ordinal $\epsilon_0$. This can be continued infinitely, enumerating the countable ordinals until we reach the smallest uncountable ordinal $\omega_1$. Loosely speaking, much as we say an ordinary Turing Machine falls in an infinite loop and, thus, does not terminate if the number of steps is infinite, or we may say $\omega$ in the language of ordinals, we may, likewise, say that a machine in the ITTM model falls in an infinite loop if the computation steps over all countable ordinals reaches $\omega_1$.

The key element in the computation of a machine of the ITTM model is to define its state at steps that go over the countable set of the limit ordinals such as $\omega$, which is the first infinite ordinal and, thus, the first “infinite step” of a machine operation after the finite steps (ordinals) which are the natural numbers $\mathbb{N}$ and which form the numbers of steps that a computation of an ordinary Turing Machine is allowed to take.

Moreover, the most important thing in the operation and halting process of a machine of the ITTM is to define a type of limit for an infinite computation to which the computation will be said to converge providing the correct computation result. Thus, we need to define
the machine’s state or configuration at limit steps \( \omega, \omega \cdot 2, \omega \cdot 3 \) etc., i.e., all limit ordinals of the hypercomputation. As described in [11], at limit states, the machine tape head is rewound to the first cell and the machine is set at a distinguished state, the limit state much like the halting state of an ordinary Turing Machine. At this point, the limit value of all tape cells is taken according to the following rule: if the value appearing in a cell has converged, i.e., they have reached the value 0 or 1 before the limit step we examine, then the cell keeps this value at the limit step. If convergence has not taken place before the limit step, that is if the values of the cell were alternating between 0 and 1, unboundedly often, then the cell is set at the limit value 1. Equivalently, the cell’s limit value is set to the limit sup of the cell values appearing before the limit step we examine.

This discussion (see [11]) describes, precisely, the configuration of a machine of the ITTM model at all limit steps \( \beta \), i.e., all limit infinite ordinals, the first of which is \( \omega \). Between limit steps, i.e., \( \beta + 1, \beta + 2, \ldots \) before reaching the next limit step \( \beta + \omega \), the computation proceeds normally, as in ordinary Turing Machines. If at any computation step between limit steps the machine reaches the halting state, the computation stops the the cell value is the value already existing there. As long as the halting state is not reached, the computation proceeds indefinitely, over the limit ordinals and the steps (natural numbers) between them.

7.3. Detecting Deductive Panopticons with Ittm Computation Model

Based on the above, it is easy to see that the Halting Problem is decidable by a machine of the ITTM class. Given an ordinary Turing Machine \( M \) and an input \( w \) to it, we provide \( < M, w > \) to a Universal Machine of the class ITTM. This universal machine writes 0 at a designated output cell, meaning “NO”, i.e. the Turing Machine has not halted yet. Then it simulates \( M \) on \( w \) if at any step the Turing machine halts, the universal machine writes “YES” on the output cell and halts, signifying that \( M \) halted on \( w \). Otherwise, the first limit step \( \omega \) will be reached, the limit value 0 will be computed on the output cell (actually, the same as the initial value - no alternation between values took place), and the universal machine will halt.

**Definition 5.** The \( i \)-th level of the arithmetical hierarchy contains 3 classes \( \Sigma_i^0, \Pi_i^0, \text{ and } \Delta_i^0 \). The class \( \Sigma_i^0 \) is the set of languages defined as

\[
L = \{ x \in \{0, 1\}^* : \exists y_1 \forall y_2 \exists y_3 \cdots Q y_i. R(x, y_1, \cdots, y_i) = 1 \}
\]

for some total Turing machine \( R \), where \( Q \) is the quantifier \( \forall \) when \( i \) is even, or \( \exists \) when \( i \) is odd. The class \( \Pi_i^0 \) is the complement of \( \Sigma_i^0 \), and \( \Delta_i^0 = \Sigma_i^0 \cap \Pi_i^0 \).

Thus, the arithmetical hierarchy includes sets definable by first order logic or arithmetic statements. If we allow quantification over sets instead of single elements, we obtain the analytical hierarchy with analogous definition and a notation in which the superscript 0 replaced by 1 (see [25]).

In particular the first level of the arithmetical hierarchy corresponds to the classes \( \Sigma_1^0 = \text{RE}, \Pi_1^0 = \text{coRE} \), and \( \Delta_1^0 \) the set of decidable languages \( \text{RE} \cap \text{coRE} \).

We can study the set of Turing Machine computable functions within the context of oracles, as we discussed in Section 3. That is, we can consider Oracle Turing Machines with oracle set the empty set \( \emptyset \). We can iterate this process much like it is done with the sets \( S_j \) defined in Section 3. It is easy to see that O-machines at a specific iteration level \( i \) cannot solve the Halting Problem for themselves but only for lower levels, acting as oracles with the oracle set \( \emptyset \). Thus, as \( i \) increases, we obtain increasingly more powerful classes of Turing Machines.

The iteration step for obtaining the Halting Problem for a class of Oracle Turing Machines \( i \) with oracle set \( X \) in general \( X \) is the empty set in our discussion but it can be any set enumerable by an Oracle Turing Machine is written \( X' \) which is known as the jump operator. In this notation, for \( X = \emptyset \), the Halting Problem (and all Recursively Enumerable
functions) can be seen to be recursive in $\emptyset'$, the Halting Problem for Oracle Turing Machines with oracle set $\emptyset'$ is recursive in $\emptyset''$, etc.

Based on the jump operator, the following can be proved:

1. **The representation theorem:** For any relation $R$, $R$ is in the arithmetical hierarchy iff $R$ is definable in elementary arithmetic.

2. **The strong hierarchy theorem:**
   - $R \in \Delta_{n+1}^0$ iff $R$ is recursive in $\emptyset^{(n)}$ (i.e., the $n$th jump of the empty set).
   - $R \in \Sigma_{n+1}^0$ iff $R$ is recursively enumerable in $\emptyset^{(n)}$.
   - $R \in \Pi_{n+1}^0$ iff $R$ is recursively enumerable in $\emptyset^{(n)}$.

The Arithmetical and Analytical Hierarchies, separated by the jump operator (appearing as 0 instead of $\emptyset$) are shown in Figure 1.

![Figure 1. The Arithmetical and the Analytical Hierarchies. (a) The Arithmetical Hierarchy; (b) The Analytical Hierarchy.](image)

With respect to the power of machines of the ITTM class, in [11] it is shown that the class of the ITTM-decidable sets lie between $\Sigma_1^0 \cup \Pi_1^0$ and $\Delta_1^0$ and extends beyond the arithmetical hierarchy into the analytical hierarchy of sets. The ITTM-decidable sets, thus, certainly include $L_u$ (the Halting Problem), which is placed low in the arithmetic hierarchy in $\Sigma_0^0$ (actually all sets in $\Sigma_1^0$ many-one reduce to $L_u$) and the language consisting of encodings of Turing Machines who accept $\Sigma^*$, which is placed, also, low (but higher than $L_u$) in the arithmetical hierarchy in $\Pi_2^0 - \Pi_1^0$. 
Let us, now, examine our two deductive Panopticon detection problems in the context of the ITTM computation model. We have defined two such problems, one represented by the set \( S'_1 \) and one by the set \( S'_2 \). With respect to \( S'_1 \), we proved in Theorem 4 that the problem of deciding this language is equivalent to \( L_u \), which represents the Halting Problem. In view of the discussion above, a machine from the ITTM model decides \( S'_1 \) since it decides \( L_u \), i.e., solves the Halting Problem, in \( \omega \) steps.

With respect to the second Panopticon problem, i.e., deciding \( S'_2 \), we proved in Theorem 5 that this problem is equivalent to the problem of deciding the problem of whether, for a given Turing Machine \( M \), \( L(M) = \Sigma^* \). This problem belongs to the \( \Pi^0_2 - \Pi^0_1 \) level of the arithmetical hierarchy (see [25]) and, thus, deciding \( S'_2 \), i.e., solving the second Panopticon detection problem, is also in the \( \Pi^0_2 - \Pi^0_1 \) level.

8. Discussion and Directions for Future Research

Theorems 3–5 in Section 3 show that, even for Panopticons with the simple behaviours described in Definitions 3 and 4, it is impossible, in principle, to detect them. Potential Panopticons, naturally, can have any imaginable, complex, behaviour but then the problem of detecting them may become harder compared to our definitions.

Comparing, now, Theorems 3–5, Theorem 3 examines the detection of Panopticons based on the execution of specific visible or detectable actions, i.e., on a behavioural level, such as connecting to a server and sending eavesdropped information or sending an email to the unlawful recipient. Theorems 4 and 5 examine Panopticon detection not based on their visible behaviour but from what languages they may recognize, without having any visible clue of behaviour or actions, only their descriptions as Turing Machine(s) (i.e., programs or systems). These theorems, that is, examine the detection of Panopticons at a “metabehavioural” level.

With respect to the difference between Theorems 4 and 5, we first observe that \( L_u \) is recursively enumerable but not recursive while the \( \{ < M > | L(M) = \Sigma^* \} \) language is not recursively enumerable (see, e.g., [7]). Although they are, both, not recursive (i.e., not decidable), their “undecidabilities” are of different levels, with the \( \{ < M > | L(M) = \Sigma^* \} \) language considered “more difficult” than \( L_u \) in restricted types of Turing Machine(s) (Panopticons). For example, the \( L_u \) language is decidable for Context-free Grammars (i.e., for Turing Machine(s) modelling Context-free Grammars) while the \( \{ < M > | L(M) = \Sigma^* \} \) language is still undecidable. Also, for regular expressions, the problem of deciding \( L_u \) is solvable efficiently (i.e., by polynomial time algorithms) while the \( \{ < M > | L(M) = \Sigma^* \} \) language has been shown, almost certainly, to require exponential time (in the length of the given regular expression) to solve (see, e.g., [7]). Therefore, a similar decidability complexity status is expected from \( S'_1 \) (deductive Panopticons without external advice) and \( S'_2 \) (deductive Panopticons with external advice in the form of an oracle) since they are equivalent to the languages \( L_u \) and \( \{ < M > | L(M) = \Sigma^* \} \) respectively. That is, when we consider more restricted definitions of Panopticons that render the detection problem decidable, then deciding which Panopticons belong in \( S'_1 \) is expected to be easier than deciding which Panopticons belong in \( S'_2 \).

Finally, Theorem 7 shows that for any formal system \( F \), we can, effectively, exhibit a particular Turing Machine for which there is no proof in \( F \), that it is either a Panopticon or it is not a Panopticon, emphasizing the difficulty of recognizing Panopticons by formal means.

As a next step, it is possible to investigate the status of the Panopticon detection problem under other definitions, either targeting the behaviour (i.e., specific actions) of the Panopticon or its information deducing capabilities (e.g., recognizing languages with specific closure properties or properties describable in some formal system such as second order logic). Our team plans to pursue further Panopticon definitions in order to investigate their detection status, especially for the decidable (and, thus, more practical) cases of suitably constrained Panopticons.
Moreover, in view of the considerations discussed in Section 7.1, it would be interesting to pursue, further, the computational undetectability of Panopticons in the form of oracle Turing Machines with their, apparent, physical detectability. This detectability is possible through the energy dissipation that takes place each time the oracle is consulted and the result is provided without any computational effort which, if it was expended by the Panopticon Turing Machine, it could be transformed into a reversible sequence of steps, so that no energy dissipation takes place which could betray the Panopticon. It appears that the mere Panopticon action, i.e., “knowing” without “toiling”, deprives the Panopticon of any potential to hide its presence.

Finally, it would be interesting to study Panopticon capabilities in the context of other “unrealistic” models such as the Infinite Time Turing Machines (ITTM) which we discussed in Section 7.2. We remarked that a Turing Machine of the ITTM class decides the Panopticon detection problem under both definitions that we gave in this paper, although the two definitions lead to problems belonging in different levels of the Arithmetic Hierarchy. Actually, since the power of such a Turing Machine steps well into the first level of the Analytical Hierarchy, it would be interesting to provide (as natural as possible) Panopticon definitions that fall in higher levels. Also, the possibility of an ITTM Panopticon can be investigated in order to define its potential behaviour, its properties and power.

In conclusion, we feel that the formal study of the power and limitations of massive surveillance establishments and mechanisms of today’s as well as of the future Information Society can be, significantly, benefited from fundamental concepts and deep results of computability and computational complexity theory. We hope that our work will be one step towards this direction.

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