A Review and Introduction to New Aspects of Digital and Computational Approaches to Human and AI Ethics

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Abstract

I review previous attempts, including recent ones, to introduce technical aspects of digital information and computation into the discussion of ethics. I survey some limitations and advantages of these attempts to produce guiding principles at different scales. In particular, I briefly introduce and discuss questions, approaches, challenges, and limitations based on, or related to, simulation, information theory, integrated information, computer simulation, intractability, algorithmic complexity, and measures of computational organisation and sophistication. I discuss and propose a set of features that ethical frameworks must possess in order to be considered well-grounded, both in theoretical and methodological terms. I will show that while global ethical frameworks that are uncomputable are desirable because they provide non-teleological direction and open-ended meaning, constrained versions should be able to provide guidelines at more local and immediate time scales. In connection to the ethics of artificial intelligence, one point that must be underscored about computational approaches is that (General) AI should only embrace an ethical framework that we humans are willing to adopt. I think that such a framework is possible, taking the form of a general and universal (in the sense of computation) framework built from first computational principles.

Keywords: philosophy of information, organised complexity, Kolmogorov complexity, logical depth, ethics of information, computational ethics, infoethics, machine ethics, computational complexity.

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1 Ethical Questions and Ethical Frameworks

Humans face ethical and moral challenges every day, in the form of conflicts over what is best in the short term versus what may be best in the long term, or what may be best for oneself or one’s group versus what may be best for others. For example, consider travelling: one may want to travel somewhere for leisure but refrain from doing so in order to limit pollution. Or smoking: one may decide to smoke for short-term satisfaction despite the likely long-term consequences. Some people choose not to care about pollution, others decide not to travel, some to smoke, others to refrain. We are constantly faced with such dilemmas.

One interesting and recurrent case is antinatalism, an extreme position that establishes that avoiding human reproduction is the appropriate ethical choice when it comes to the question of procreation. The position is interesting because within a local context the logical reasoning seems flawless, and can only be refuted if non-local arguments are brought to bear. There is also a matter of time scales of a binary nature: either someone exists to be taken into consideration when making a choice, or they do not.

There are a number of ethical frameworks which philosophers have proposed, from Aristotle’s virtue-based ethics [2], to Kant’s deontological ethics [15, 16, 17] to Mill [20] and Bentham’s teleological ethics [7]. Although a formal interest in ethics characterises the development of approaches such as deontic logic, ethics and logic have for the most part diverged, reflecting the separation of the humanities from science. Toby Walsh, a Professor of Artificial Intelligence, has expressed what appears to be a pragmatic observation, namely, that if humans cannot specify our ethical and moral codes with precision, we cannot expect that it will be easy to do so for AI:

For too long philosophers have come up with vague frameworks. Now that we have to write code to actually implement AI, it forces us to define a much more precise and specific moral code.

Here I will advocate that for ethical frameworks to be relevant, translatable, and eventually readable by machines, human ethical frameworks need to be formalisable, and that computational approaches offer such transferability. My proposal is thus fundamentally different from those based on logic and probabilistic foundations, such as in [1], and does not make an ontological commitment to any particular framework other than to propose that frameworks be evaluated and enriched by a set of concepts and features related to computability. Instead, I will explore some suggestions regarding computational ethics [22] and suggestions for introducing measures of complexity or sophistication into ethics [14, 6, 12].

A first question concerns what to consider a valid ethical question and a valid framework within which to place it or from which to derive it, a question about which a human being can make a conscious decision that will have consequences for others as well as for him/herself. For example, is the question of procreation
a valid ethical question? It should be, because we are the only species on earth able, thanks to biological evolution, to choose whether or not to reproduce, aided by such technologies as contraception. This means that an ethical framework such as antinatalism is a valid and legitimate one.

Here, I explore some direct and indirect implications of developments seeking to introduce digital information and complexity indices into ethical discourse.

There are also questions that should not concern ethics, the ethical answers to such questions being meaningless. For example, issues arising from anything that is accidental, such as, say, lightning striking trees, because in principle (unpredictability) as well as in practice, no human action can prevent such an occurrence (by protecting all trees or preventing storms).

There are also no absolutely (un)ethical questions. For example, it is conceivable that within some framework, and perhaps (but not necessarily) on the basis of hard evidence, it could be decided that plants do have some sort of internal experience and can feel pain, and that humanity finds it within the realm of the possible to protect all trees from lightning. Then such a question/concern may become ethical. Within some ethical/moral frameworks cannibalism or incest may not be unethical, but in others they are. Likewise the way in which animals may be killed (e.g. halal versus non-halal). If no absolute ethical/moral framework can exist, the value of and the answer to an ethical question is relative to the framework of reference.

Universal and objective ethical frameworks are desirable but likely difficult, if not impossible, to realise. The mainstream position is that we have therefore to start acknowledging that all human ethical frameworks are, and only make sense if they are, human-centred and designed to deal with challenges related to humans, even those addressing environmental concerns which are ultimately related to the well-being of the biosphere where humans coexist with all other living forms on Earth. Some studies can clearly pinpoint how different populations can adopt different stances. In a recent and highly publicised study of moral and ethical dilemmas related to car crashes [3], Eastern countries were found to dramatically favour sparing the elderly and lawful pedestrians, while Western countries favour the young and have adopted some utilitarian preferences, leading them to prioritise sparing the greatest number of people above all else, followed by sparing people of higher social status and fitness. Thus Eastern countries dramatically diverge from Western–and Southern–countries, showing little concern for numbers saved, social status or looks. Southern countries, for their part, do not favour lawful pedestrians over others and care mostly about sparing people with higher status, the young, the female and the fit, and show a much higher preference for sparing animals over humans than the other two groups. This clearly shows that different populations may embrace different values and that agreement is difficult even when it may seem obvious.

While any principle can be placed at the centre of an ethical framework, there cannot be a way to falsify an ethical framework on the basis of any objective universal measure, because that would necessarily imply evaluating a framework from the standpoint of an alternative framework, which leads to an immediate contradiction (the evaluating framework would be more primitive or
fundamental than the purportedly universal evaluated framework). However, not all ethical frameworks are equal, nor should they be treated as such.

An ethical framework can, for instance, be tested for internal consistency at different levels and time scales to see whether following its principles leads to contradictions. In this sense it is not, or should not be that different from formal logic, in which consequences have to be compatible with the purposes of a given framework. A framework that often tells you that you have to follow certain principles because you cannot see the greater plan is the kind of framework I would propose be relegated in favour of less obfuscatory frameworks, for the sake of consistency at various scales and levels.

We should thus not equate plant signalling with a sentient being’s pain and justify assigning equal value to ethical frameworks based on notions of plant ‘pain’ and those dealing with animal or human pain. We are sure that the latter is a consequence of an internal conscious experience, while the consensus is that plants have no such experience or anything to indicate that they do. But more importantly, the ethical question here as regards plant suffering is that plants do not have decision-making power with regard to their own pain (not to the extent of being able to name it and decide on measures to alleviate it). This means that the question of plant suffering may not be equally legitimate.

In contrast, humans have the ultimate capacity to stop all possible pain, both necessary and unnecessary, even the ineluctable pain of existence– which makes a stance such as antinatalism an interesting case study. Antinatalism advocates that we stop procreating, a legitimate, even if extreme, stance.

One advantage of arbitrary (random) ethical frameworks, based, for example, on religion, is that they do provide clear rules. However, whether they are sound and self-consistent is another matter, as we have seen how their doctrines can be shaped to fit different, often contradictory, purposes (e.g. killing in the name of god). Many of these are inconsistent and hard to follow, even making it difficult for bona-fide adherents of the same religion to understand each other.

Of course, religious rules are not random at all. On the contrary, they have been highly selected because of their efficacy in controlling human behaviour. However, they are random by definition, needing not rational explanation but faith.

Here, I will suggest that aspects of computability may help us decide what should be taken as desirable features and properties of an ethical framework, and also dictate computable strategies, even if these ethical frameworks may lead to uncomputability–which I think should be embraced and understood. As I will suggest, non-decidability and uncomputability preempt teleological, goal-oriented frameworks, allowing for a healthy number of different possible directions that can be adopted to attain more local and personal ends.

2 Digital Information and Computation

There have been mostly 3 areas in which digital information and computation have recently been introduced in the ethics discourse in last decade, from
the pragmatical or theoretical point of view: frameworks based on agent-based simulations that allow in silico testing [8], ethical approaches heavily based on information theory [14], and a novel development based on complexity [12], itself based on previous suggestions [6]. The present research adapts and builds on some of these previous ideas, except for information theory, which we would rather avoid because it is unsound, being highly dependent on the feature, language or probability distribution selected [27].

From measures of information allegedly able to quantify consciousness to recent proposals for an ethics based on information theory and complexity, we have available to us possible resources to use in deriving global guiding principles from informational and computational first principles that may describe or explain interesting aspects of human activity.

2.1 Information-theoretic Approach

The only approach that avoids the strongest form of irreducibility, that of semi- and uncomputability, is based on a weak but computable measure. This approach, based on classical information theory [14], proposes to place at the core of an ethical framework the purpose of minimising entropy as a central guiding principle, in a sort of new age ecological approach to ethics. The idea is that agents, including humans, should ‘avoid entropy’ and oppose it with ‘poiesis’, that is, with construction. To be good agents in the infosphere, therefore, this approach suggests that we should be combating metaphysical entropy and promoting poiesis.

This approach leads to four ethical principles, according to Floridi:

1. Entropy ought not to be caused in the infosphere,
2. Entropy ought to be prevented in the infosphere,
3. Entropy ought to be removed from the infosphere,
4. The flourishing of informational entities as well as of the whole infosphere ought to be promoted by preserving, cultivating, and enriching their well-being.

This position and approach, however, has many fundamental problems and is rooted in a fundamental lack of understanding of both thermodynamic entropy and Shannon entropy as a measure. Entropy is not a feature-independent measure and therefore before minimising entropy one has to make an arbitrary choice as to the feature of interest in relation to which entropy is to be measured. Additionally, while minimising entropy would not lead to inaction, because by the 2nd law of thermodynamics it is impossible to avoid generation of heat from low entropy states in their progress toward higher entropy states in the environment, it may lead us into triviality. Indeed, entropy does not entail structure, and it can easily mean shallowness.
2.2 Integrated Information

Along these lines, there is another interesting development connected to information theory as it relates to dynamical systems. This is the concept of integrated information, suggested to be related to consciousness. Indeed, Integrated Information Theory, as introduced by Tononi et al. [24], not only proposes an apparent objective measure of consciousness but also poses new ethical questions.

For example, the measure $\phi$ related to Integrated Information, suggests that consciousness is a property of systems and not only of livings beings, and it also suggests that it is graded over continuous values, meaning that there is no such thing as a medical or scientific answer but rather an arbitrary cutoff value that must be adopted in deciding whether a sentient being is conscious or not. It may, for example, open up a new debate related to abortion and the number of weeks after fertilisation when a fetus can be considered conscious, and it may even provide a numerical answer to the question of whether newborns attain the kind of consciousness that adults are endowed with, when they do attain consciousness. The stages from fetus to developed child include several cognitive milestones, such as the awareness of the body and of existence itself (which often causes pain when existence is grasped to be finite), with body awareness beginning early and progressing gradually [13] until maturity. The measure $\phi$ is a continuous real-value number that goes from 0 to some as yet unknown number representing the typical (average) adult level of consciousness. The same would apply to questions related to passive or active euthanasia for cognitive conditions and neurodegenerative diseases. With $\phi$ not only indicating that consciousness is a dynamic state of the mind but also that it may be low enough to be below an abortion threshold, it could also supply the wherewithal to answer ethical questions about euthanasia.

Integrated information’s $\phi$ is an interesting measure that may capture a necessary condition, but it is unlikely to quantify a sufficient condition (another clearly necessary condition is embodiment, i.e. a system must actually be embodied and capable of interacting with its environment, closing input-output loops). So I am not endorsing any suggestion that $\phi$ alone actually quantifies either consciousness or other possible factors that may be either intrinsic or extrinsic (the value of a person to a family, to society, etc).

Being a measure of systems rather than beings also implies that there may be artificial systems with some degree of consciousness, even if not completely related to ours. And there is no reason under this framework to exempt machines from certain degrees of consciousness, perhaps even equal to or even greater than that of animals and the human being, which leads to all sorts of new ethical questions related to AI, and how to better approach ethics from computational perspectives, just as $\phi$ itself does.
2.3 Complexity-based Approaches

A more robust approach to an ethical framework, based on complexity rather than on classical information, was initially introduced by Bennett [6], based on his seminal concept of Logical Depth [5], a measure of ‘sophistication’ that quantifies irreducible computational work (see Appendix).

One criticism of algorithmic complexity is that it does not conform to our commonsense intuition of complexity. For example, when we say that something is complex we rarely mean that it is random, but actually that it is quite sophisticated, and neither trivial nor random. A measure of the complexity of a string can be arrived at by combining the notions of algorithmic information content and time. According to the concept of Logical Depth, the complexity of a string is best defined by the time that an unfolding process takes to reproduce the string from its shortest description. The longer the time taken, the more complex the string. Hence complex objects are those which can be seen as “containing internal evidence of a nontrivial causal history”.

Unlike algorithmic complexity, which assigns randomness its highest complexity values, logical depth assigns a low complexity or low depth to both random and trivial objects. It is thus more in keeping with our intuition of complex physical objects, because trivial and random objects are intuitively easy to produce, have no lengthy history, and unfold very quickly.

Logical Depth was originally advanced as the appropriate measure for evaluating the complexity of real-world objects such as living beings. Hence its alternative designation: physical complexity (used by Bennett himself). A persuasive case for the convenience of the concept of logical depth as a measure of organised complexity, over and against algorithmic complexity used by itself, is made by Bennett himself. Bennett’s main motivation was actually to provide a reasonable means of measuring the physical complexity of real-world objects. Bennett provides a careful development of the notion of logical depth, taking into account near-shortest programs as well as the shortest one—hence the significance value—to arrive at a reasonably robust measure.

More recently, the proposal to base ethics on grounds of Logical Depth has been expanded [12] (I will denote logical depth hereafter by LD. See glossary in the Appendix for its basic definition). The framework favours computational work. In some respects it may lead to similar consequences as its classical information-theoretic alternative, such as the preservation of species and biodiversity, but it is based on much more robust grounds, and has the advantage, for example, of being independent of language (to some extent) as well as being feature-independent, thus being universal in a formal sense. But a shortcoming, that may also be seen as its strength, is its uncomputable nature. On the one hand, it makes for open-ended meaning, as it cannot be fully calculated and has no ‘end’, but at the same time it may lead to inaction. Moreover, it can lead to waste, for example, the energy (in the form of electricity) waste generated by the activity of mining crypto-currencies like Bitcoin that perform irreducible work and hence artificially increase the logical depth of the world without any evident benefit, while using up natural resources. This is actually a contravention of the
spirit of the classical information-theoretic approach to ethics and is possibly in clear contradistinction to adopting a blind LD approach. That Bitcoin is based on SHA256, crackable in the future, is accidental. SHA256 can be replaced by something harder or even impossible to crack, not only in practice but also in principle, making crypto-mining a shallow activity even if it increases LD, and hence difficult to justify.

The LD approach may explain why humans are ‘hoarders’ and why we may favour museums and space exploration programs over giving money to poor people. However, it affords little normative power, as it is better grounded than entropy in principle but not in practice, given the uncomputability of LD (see Appendix).

Moreover, the proposal suggests that humans can estimate and make decisions based on LD but computers cannot somehow suggesting that humans have some non-computable insight over algorithms, turning the LD approach relevant again to humans by virtue of this distinction. If no distinction were made between how relevant can LD be to humans vs machines, the approach would suggest to devote all human resources to calculating LD alone, an object that Bennett himself [5] introduced and called $K_0$ (see glossary in the Appendix for an explanation of both objects).

The approach provides little guidance on questions such as ‘should I give money to a charity?’, ‘should I torture an animal?’. For example, killing a neighbour may or may not increase LD, so it is a locally agnostic approach. It does provide some guiding principles on certain other questions, such as, ‘should we fund science?’; ‘should be build a new museum?’ ‘should I continue to be a hoarder’ as their answers in the positive clearly have a greater chance of increasing LD.

I am, however, not suggesting that the use of LD makes for an absurd approach, but that many layers are in play, and that we embrace different frameworks at different times. We may like to increase LD in our lives most of the time: we collect things all the time, go to museums, cherish old masterpieces or what we think of as being high in LD, etc.

Some of these ethical frameworks have thus turned what should be a means to help frame aspects of a consistent and meaningful ethical framework into an end in itself (e.g. making increasing LD an end in itself).

2.4 Hybrid, Multiscale and (Semi-)Computable Ethics

We have discussed how not all possible principles and ethical frameworks are equal or should be assigned equal value. For example, that plants and animals feel pain in different ways (if plants do at all) rules out having comparable ethical frameworks. From the human perspective the lack of evidence of plant pain should not be equalled to the certainty of animal or human pain. This is, of course, an anthropocentric view, as we are assuming that chemical signalling does not lead to pain, which may be wrong for plants, but considering everything equal also leads to inaction. In this case, for example, plant pain may be used to preempt action against animal or even human pain, forcing us into disregarding
unavoidable biological signalling and even the evidence of internal conscious experience.

This example also illustrates that the lack of evidence cannot be presented as an argument against an ethical framework, nor should it be taken as central to any ethical framework. The lack of evidence that plants feel conscious pain, or at least to the degree that humans do (c.f. Section on Integrated information and Ethics) should not be used against the legitimacy of ethical concerns regarding (unnecessary) animal and human pain for which we have strong evidence.

The extreme example of antinatalism suggests that incidental or accidental (unavoidable or unnecessary) pain should not be equalled to avoidable pain, as the latter is in the realm of our definition of an ethical question/concern, being the result of an action that is preventable, while the former is not.

So the question is how computation can help to find meaning, effective guiding principles, and specific rules derived from such principles.

I argue that an ethical framework should also be taken as a source of guidance in making decisions at various scales, in particular locally, in local space and time. LD, for example, does not seem to answer the question ‘should I procreate?’ One answer is that a new being is capable of increasing LD, but LD is not additive, meaning that 2 beings do not automatically mean 2 times more LD, due to redundancy. However, in the long term, bringing more people into the world can also exhaust its resources, leading to a decline in LD after all, so LD may lead to inaction.

An ethical framework should be favoured if it can provide multiscale guiding principles as a basis for taking decisions in one’s everyday life. For example, in deciding whether procreation is ethical or not or whether one should kill one’s neighbour. Because ethics is not only observer-dependent, a good framework should be one that provides the means to make local decisions while being cognizant of global consequences.

As a hybrid, adaptable, multiscale framework can adopt the previous frameworks in the worst case, it has a potentially higher normative and explanatory power. It favours ethical frameworks that can provide principles that can be simulated (computable) by mind and machine. Its explanatory power is fixed (independent of the actual content and principles) and is powerful at local and global scales (by definition); it does not adopt a single framework but is flexible and can even depend on computational power and other parameters, with the advantage that it can provide algorithmic explanations. Its normative power is also potentially higher, as it favours computable choices for local principles and semi-computable choices for global, non-teleological principles, and so combines the best of both worlds.

An ethical framework should thus regulate and provide guidance in the execution of an action by virtue of being at least semi-computable, if not computable and tractable. An ethical framework should also have properties that enable it to perform step-by-step simulations and to reach a consensus on aspects related to empathy by way of emulability (Fig. 2).

In what other way, if not simulation by a Turing machine, can we
understand the process of making free choices? By making them, perhaps. R. Nozick (1981)

Computable enumerability and semi-computability are thus technical features that are desirable in ethical frameworks.

2.5 Comparison of Info-Ethical Frameworks

A very simple ethical framework can simply have to do with minimising unnecessary pain at all costs, as most ethical frameworks take, and should take, sentient beings as central objects/subjects. Its dictum would be to minimise...
unnecessary pain and unnecessary suffering in conscious beings (and this would include animals) at all costs. It can explain why we may chose not to kill each other, why we are not all terrorists, why we may care for each other, indeed a good portion of the actions of our everyday lives. Its normative power is strong but can easily lead to contradictions and inaction, e.g. you should not eat meat, you should not travel, as the harm done in killing animals and polluting the environment is high, etc.

The ethical framework based on information theory and defended by Floridi [14] aims at establishing entropy minimisation at all costs as its central principle. It is heavily based on the concept of physical or informational entropy. Its dictum is that entropy ought to be prevented/removed and ought not to be caused in the infosphere [14]. Its explanatory power is sufficient to answer questions such as ‘should we keep our environment clean?’ but it is very limited because entropy is not well-defined beyond physical entities, and hence is quite irrelevant to most ethical and moral questions. Moreover, its normative power, though extremely limited as it is highly language sensitive, can be adapted ad hoc by minimising one parameter while maximising another [27]. It is therefore very fragile and internally inconsistent.

One may think that more advanced societies reduce procreation as a possible way to increase LD (e.g. creating AI). The problem with such an unbounded approach (with the only objective being to increase LD) is that, for almost every counter-example one may find, the LD approach can always justify itself by claiming that one cannot foresee in which ways LD would increase, i.e. claiming that we cannot see the whole plan, thereby becoming highly dogmatic and not unlike religious arguments (see Fig 1).

For example, when technology surpasses biology, if it ever does (see Fig. 4), the technological world may be immensely complex, much more so than the current biosphere, and technological sophistication cannot be reached without there first being biological sophistication, so clearly LD increases. This would also lead naturally to cybernetic values (e.g. building resilient systems, powerful enough sensors and actuators, etc). Regarding digital antinatalism (the stance not to create artificial consciousness that will be most likely subject to pain), we could play with non-binary choices once we move to digital organisms. We may be ready to accept robots that do not feel pain but only pleasure and happiness (if these things can ever be hard-coded).

The only way to justify an approach such as LD is by taking it only as a general guiding principle. Starting from the fact that there are already sentient beings and humans in the universe, as long as there is nothing else to compute LD in a more efficient fashion than humans (that the LD approach does not discard), humans can keep trying to make decision based on estimations of an increase of LD and keep in a probabilistic fashion (or seemingly educated guess). However, if better and more efficient means exist in the future humans may become irrelevant to the LD approach.
Table 1: W stands for weak, M for medium and S for strong. Consistent means that the theory is consistent but is independent of application to an ethical framework. Inconsistent, however, means that the measure itself is inconsistent, for all practical purposes, in any context or ethical framework [28]. Ideal frameworks should have consistent principles and should be strong in all respects and at all scales.

### 3 At the Interplay of (Un-)Computability

#### 3.1 Emulability and Irreducibility

Another important point to consider in the discussion of computable ethics or ethical frameworks based totally, or partially, on computational (or constructible logical) approaches is the many forms of computational irreducibility that they may face. The strongest type of computational irreducibility is that imposed by the Halting problem, from which many others are derived, but weaker forms of irreducibility [25], including intractability, fall into this category too. For example, in pursuing the idea of agent-based simulations, one has to be resource-aware. More pragmatical relationships between resource-bounded simulation and ethical frameworks have been investigated [22], but many angles remain unexplored, especially the more fundamental connections to theoretical concerns.

In a recent study on ‘metacognition’ assessing the ability to evaluate whether one may or may not be able to recognise to be wrong, it was found that people with extreme political views that were the least empathetic between each other, were found to have trouble thinking about their own thinking [21]. Along these lines, an important extension of the simulation-based approach to ethics is in the direction of the adoption of computable and semi-computable properties. For example, ethical frameworks may benefit from being computable so that any human or machine is able to follow and simulate a principle’s rationale step-by-step, amounting to what would be a form of Turing-universality and a way to describe computable ‘empathy’ (Fig. 2). Notice that by simulation here I do not mean only computational but also mental simulation. There being no evidence that the mind is non-computable (for a discussion see [26]), and since we have evidence that it can perform Turing computation, agent-based simulation is pertinent to both machines and humans. This kind of universal emulability would also allow a superintelligence to emulate other ethical frameworks, but this also implies having its own framework be undecidable.

Computational irreducibility has implications for consequentialism because...
Figure 2: Universal empathy suggests that a Turing-universal ethical framework can mediate between other frameworks, actions and guiding principles, even when direct communication may not be possible. This is by way of compilers that are guaranteed to exist by virtue of computability. The compilers translate principles, actions, or frameworks in terms of other frameworks. The ethical framework capable of Turing-universality is also capable of performing algorithmic simulations of all actions, guaranteeing that other systems can follow them step-by-step.

in general we cannot predict outcomes in our complex world (other than by simulating it). This implies that doing rigorous but multiple simulations to foresee or explore consequences is not only desirable but necessary. For example, performing computer simulations in lieu of actual nuclear testings falls into this category.

A recent experiment involving computable simulations—mostly mental but also aided by a computer imaginary and interface [3]—leading to computational ethics [22], yielded extremely interesting results that allowed a classification of those whom people believed would be more dispensable than others in a situation in which a self-driving car crashes and kills passengers and/or pedestrians. After simulating the outcomes of these car accidents, a poll was conducted to rank subjects according to their preference for saving different people and animals. A gender bias in favour of sparing females was significant. And between a dog, a criminal and a cat, most people chose to save the dog. While this may look like a game, it is most relevant to current technology, where these preferences could be hard-coded in driver-less cars to decide who should live or who
should die, though questions about liability would still remain. Furthermore is consensus the right criterion when it comes to evaluating ethical frameworks? We have seen how democracy has in the course of human history sometimes brought about undesirable outcomes, and that the miscarriage of democracy can be incredibly dangerous. Likewise wrong ethical frameworks, even when reached by consensus, may also have undesirable consequences, even amounting to existential risks. Is deciding what is right and wrong a matter of popular opinion?

In the end, computational irreducibility imposes the limits by which any approach to ethics or anything else is bounded. It means, for example, that we can neither perform all possible simulations nor derive all possible consequences of such simulations. Nor can we fully carry out, foresee or shortcut most computations, and it is likely that final consequences can only be seen and witnessed as they actually unfold in real-time. Keeping these facts in mind, we can explore some measures for dealing with different kinds of irreducibility, from time-space bounds to semi- and full uncomputability.

3.2 The Role of Algorithmic Complexity

Approaches that have been poorly explored are those based purely on algorithmic complexity and algorithmic probability as normative measures quantifying randomness and quantifying computational difficulty (both in calculating and simulating).

Defined by Kolmogorov, Solomonoff, Chaitin, and Levin, the so-called program-size complexity, also known as algorithmic complexity or Kolmogorov complexity, is a measure that quantifies algorithmic randomness, a type of randomness that is strictly stronger than statistical randomness.

Formally, the algorithmic complexity, which we will denote by $K$, of a string $s$ is the length of the shortest computer program $p$ running on a universal Turing machine $U$ that generates the string as output and halts [18, 11]:

$$K(s) = \min\{|p| : U(p) = s\}$$

The largest or smallest difference between the original length and the compressed length, that is, the length of the Turing machine, determines the complexity of the string.

There is a question, of course, about whether there is any value in attempting to estimate uncomputable objects, and of its value in the context of ethics. First, it is worth mentioning that algorithmic complexity is, in principle, much more stable than Shannon entropy, in that it has an invariance theorem (see Appendix) that allows for some dependency on the language used, and total independence of probability distributions.

One can easily see how algorithmic complexity is related to knowledge and meaning in a very profound manner, contrary to the usual portrayal of computation as incapable of dealing with deep questions about meaning and epistemology, as if these concepts were trivial generalisations of classical information
theory—which they are not. Shannon entropy may lack the most important ingredient, precisely the concept of computation, but we saw how even Shannon entropy is related to meaning, albeit in simpler ways.

For example, the $\Omega$ number (see Appendix) is a family of numbers able to encode at the limit the halting runtimes of all computer programs. The $\Omega$ number, however, is itself uncomputable. If we had access to all the digits of any $\Omega$ number we would have the exact halting time for every possible computer program, because even though $\Omega$ depends on the choice of Turing machine, we also know that, given that $\Omega$ is defined in terms of a universal Turing machine, the set of all computer programs computed by any $\Omega$ number is exactly the same. In other words, having access to the infinite digits of an $\Omega$ not only gives us the Halting probability of the universal Turing machine on which that particular $\Omega$ number is calculated, but also to all others. Under the Church-Turing thesis, it also means that we have access to basic knowledge about all the computer programs that are possible. And computer programs can be anything, from a simulation of our solar system to a simulation of a game-theoretic solution to a real-world Prisoner's dilemma, posing all sorts of ethical challenges.

### 3.3 Algorithmic Probability

That an $\Omega$ number holds all answers to computable questions (see Appendix) is the reason why it has been named the infinite wisdom number. Of course it is a different matter to tell which program is which, and which program encodes the question we want Chaitin's $\Omega$ to answer. However, short computer programs encode most of the meaningful objects that we care about, such as the mathematical constant $\pi$, compared to arbitrarily long programs, because computer programs that are arbitrarily long relative to their output may be encoding random objects that, while they may be good for pseudo-random generation of the type useful in a casino, are hardly worth much else. This is exactly the gist of Algorithmic probability as introduced by Solomonoff [23] and further developed by Levin [19], deeply related, and even equivalent to Chaitin's $\Omega$ number (the difference being that Algorithmic Probability also bears on the output of computer programs and not just their halting times). So by looking at the length of computer programs relative to what they produce we may have some sense of what they may be encoding, as we search for interesting questions and answers. It is thus also clear how algorithmic probability is related to algorithmic complexity. A long computer program will have low algorithmic probability because it would take longer to produce it by chance (by producing bit by bit of its executable binary file) versus a short program encoding a low algorithmic complexity object that can be produced with higher probability.

Algorithmic probability is a semi-measure (see Appendix) that allows us to introduce a natural bias related to the underlying generating mechanisms, in this case the likelihood that a random computer program will produce a string, without having to assume probability distributions, which in the most common situations we wouldn't have access to. And there would be no need to assume an almost arbitrary distribution such as the uniform distribution, which makes it
hard, if not impossible, to differentiate subtle but important apparent properties of different objects. So, a sequence such as 11010010 would be differentiated from a sequence of only 1s or only 0s that we can be sure can be generated by extremely small and simple computer programs.

Algorithmic probability is usually regarded as a formalisation of Ockham’s razor, and thus should be highly useful in deciding and discriminating ethical frameworks by their simplicity and explanatory power. In the case of Algorithmic Probability, the ‘fewest assumptions’ requirement is formalised by the shortest computer program that can account for a process. At the same time, algorithmic probability also complies with Epicurus’ Principle of Multiple Explanations, which establishes that ‘if several hypotheses are consistent with the data, one should retain them all’ and, indeed, one can see from the definition of algorithmic probability that every computer program producing the data is not only retained but contributes to its algorithmic probability, even though it is the shortest computer program that contributes the most and is thus the most likely to be producing the data, according to algorithmic probability and the coding theorem (see Appendix). This is highly desirable in an ethical framework, where no path should be ruled out, especially when considering possible consequences, and should thus be a fundamental formal requirement of any approach based on agent-based simulation.

Algorithmic probability is not just an interesting cherry-picked measure, it is the accepted mathematical theory of optimal inference and it imparts sense to something that Chaitin has claimed in the past, that ‘comprehension is compression’, because the most likely explanation for a piece of data, according to algorithmic probability, is also the most compressed. But comprehension and accountability are desirable properties of any ethical framework.

While all these measures and objects are uncomputable, or more precisely, semi-computable (they can be approximated), I will claim that this is also sometimes a desirable feature for open-ended meaning, and because of its semi-computability, local approximations are possible, thereby obviating inaction. However, they have to be complemented by other guiding principles based on different measures, especially for local decisions that do not require so much calculation and can be adapted and corrected when more computational (human or machine) time becomes available.

4 From unsound human ethics to unsound AGI

As illustrated by some possible unsound human positions, such as a purist pro-life position based on the right of life, ethical frameworks may be inconsistent if not properly stated, i.e. if no additional principles are included, such as, e.g., specifically opposing and properly defining killing in contrast to, say, avoiding death. Hence unsound logical derivations that AI ethics may inherit from inconsistent human premises should also be thoroughly tested in the area of human ethics. In this example, an AI ethics that embraces avoiding death at all costs would also avoid life, embracing antinatalism. This also means that the op-
posite view of pro-life is neither pro-death nor anti-life, and that active versus passive anti-life should be distinguished. Active anti-life is actually pro-death, whether permissive (passive euthanasia) or active (euthanasia and killing). The argument arises from the fact that human life is paired with human death, so the only infallible way to avoid death and truly be pro-life is also to avoid life and be passively anti-life. We thus make the case that for AI purposes and safeguards to be consistent, they should also be consistent for humans and should not be artificially rewritten for AI when humans share, or should share, the same principles under a common ethical/moral and logically sound framework.

We do, however, have the means to limit AI capabilities of developing consciousness based on what appear to be minimal required conditions for developing an internal information-theoretic experience, given the concept of integrated information, which is deeply related to the algorithmic complexity of dynamical systems or what I call algorithmic information dynamics, and reprogrammability measures, measures of sophistication also closely related to logical depth.

What we have thus far found is that local ethical guides cannot be based on completely uncomputable rules because they lead to inaction and total agnosticism, since immediate actions, such as, should I kill my neighbour, cannot be decided. Instead, one has to switch to more local guidelines of a more computable nature, such as some estimation of (unnecessary) pain and suffering minimisation (local in space and time), so that mental (or even computer) simulations of the outcomes are possible (even if often deemed to be irrational).

4.1 Limited Application to Classical Dilemmas

One clear limitation and challenge for most of these approaches is that they are not yet mature enough to deal with classical ethical or moral dilemmas, except for those suggesting simulation. None of these information-theoretic, or complexity-based approaches seem to be able to determine whether to tell a friend about their partner’s infidelities. Would LD suggest you turn in a robber if the robber has donated the stolen money to an orphanage? While children in an orphanage may benefit by increasing their future LD, the bank’s own activities may have increased LD even more. This would be the same guiding principle for someone facing having to choose between saving an older adult versus a child, but not both. The older adult may have already produced all the LD that could have contributed, and thus saving a child would mean a greater chance of increasing the overall LD, but the entropy approach may suggest the exact opposite because the child is more likely to produce greater waste than the elder, simply based on the length of their remaining lifespans. The same applies to dilemmas where there are a large number of people to save versus a few, or by action versus inaction. While LD may suggest saving everyone as long as there is no suggestion that any of them will negatively contribute to the future increase of LD, minimisation of entropy suggests letting people go, so this latter would be more actively in line with ethical stances such as euthanasia, where LD or complexity approaches would remain mostly agnostic.

What is more clear, however, is that for these and other classical dilem-
mas, computational (either mental or digital) simulation does contribute. For example, simulations of what children at the orphanage may achieve with the stolen money versus what it would have yielded had it remained in the bank. In another example, blaming someone else for a crime has implications such as living with guilt and making someone else pay for one’s crimes, which is something where simulations independent of the ethical framework adopted can be informative.

In more complicated cases, such as the Prisoner’s dilemma, LD may suggest cooperating and getting reduced sentences if the prisoners can be reformed or if their felonies can be considered detrimental to the overall increase of LD, while entropy would also require additional assumptions, but it would probably be safe to believe that entropy is minimised by keeping the prisoners in jail for the maximum possible time, assuming both remain felons. However, it is hard to see how each individual prisoner would take a minimisation of entropy stance to maximise their sentences in order to make their choice. In all cases, however, simulations are key, and it is something we have already done, both mentally and systematically for the Prisoner’s dilemma, but which we can do more formally and even more systematically in all other cases, also guided by computational and algorithmic complexity, in order to decide what paths are more likely, are less or more random and hence require more assumptions or explanations.

4.2 Zero-sum extreme ethical dilemmas

An extreme case of an ethical dilemma will help us analyse some aspects of proposals for human and AI ethics of particular relevance to Artificial General Intelligence (AGI), which has been alleged to pose a threat to human existence [9]. The case concerns procreation but not suffering; what is in question is procreation, it being generally agreed that unavoidable (or unnecessary) pain is wrong. An ethical framework that says that unnecessary pain or unnecessary suffering is good or not relevant may be immediately challenged as it is central to many ethical frameworks, either as a core principle or as a derived one.

Well-being and happiness can be regarded as a game, as in traditional consequentialist, utilitarian ethical frameworks. Well-being or happiness versus pain are so opposed that it is safe to regard their relationship as a zero-sum game, where if \( H \) is happiness and \( P \) is pain (of any form, such as physical or mental suffering, necessary or not), then \( H - P = 0 \). Interestingly, the number of players changes according to the perspective. From the point of view of the unborn, there is only one player in the zero-sum game. I claim that any other framework, where the only relevant viewpoint is a viewpoint other than that of the unborn, is selfish, because the unborn cannot choose, or can be said to be by definition selfish because any choice would willy nilly be exclusive of other players. Once the framework of selfishness is adopted, it becomes a non-zero-sum game with more actors and greater complexity. In the altruistic case, the scenario is binary because there are only two options: either the newborn exists or it does not. From the perspective of the unborn/newborn, it may have \( H \) happiness and \( P \)
pain/suffering with \( H \) and \( P \) being real numbers. However:

1. If a human being does not yet exist it does not make sense to ask if a non-existent person can be \( H \) happy, though one can know with certainty that by not existing \( P = 0 \), thus no suffering. However,

2. If the human being is born, then only \( H = 100\% \) can justify the choice. Therefore, because \( H = 100\% \) is unlikely or even impossible, then only \( P = 0 \) can be justified and the person should remain unborn in an altruistic world.

Option 1 is the non-existence option; it minimises unnecessary pain. The question of happiness has no meaning because the person does not yet exist, but once they exist then \( H \) and \( P \) have meaning, and \( P \) will certainly not be 0 in their lifetime. The rules that apply to someone who does not yet exist do not apply to someone already forced to exist. In other words, nothing can be bad or wrong for someone who does not yet exist but everything could be bad or wrong for someone who does. Hence bringing anyone into being who cannot participate in such a decision cannot be a selfish act within the isolated sphere of the parents-newborn relationship (which, of course, only comes into being after the parent makes the decision to have a child). But if external arguments are brought to bear, justifications can be found, though no external justification can be an internal unselfish argument at lower levels (see Fig. 3).

So the only way to reduce unnecessary pain with absolute certainty and remain altruistic or not selfish is by choosing an antinatalist stance, because existing will always entail both happiness and pain (possibly both necessary and unnecessary), and parents are responsible for both when they could have avoided pain altogether, and the question of happiness makes no sense prior to the existence of the unborn, meaning that the decision to procreate was taken in a selfish manner.

In terms of Benatar [4], the argument follows that coming into existence generates both good and bad experiences, pain and pleasure, whereas not coming into existence entails neither pain nor pleasure. The absence of pain is good, the absence of pleasure is not bad. Therefore, the ethical choice is weighted in favour of non-procreation. No amount of happiness can counterbalance the minimum amount of pain and suffering when the option is non-existence.

From a population point of view one can find ethical justifications, but not for the individual who is forced to live. One has to divide the question between the individual and the population as well as consider time, as it is not the same thing to force someone to exist and then ask them to kill themselves as it is to simply decide not to bring them into being in the first place.

Some may be willing to think of an artificial replacement for human existence but may find it shocking to consider that a possible good outcome can also be the disappearance of the human race, as a consequence of local decisions of humans not to procreate (see trends in Fig. 4). The trends may even have more far-reaching implications as a possible answer to Fermi’s paradox. If we find that even on Earth itself population decreases as countries become more developed,
Figure 3: There seem to be apparent absolute selfish/unethical actions illustrated by, e.g., a parent-new born closed system, where one individual takes a long-term and far-reaching decision on behalf of another one by forcing it to live. Positive values amounting to a real number $P$ between 0 and 100 cannot be assigned to an unborn because they do not yet exist. A $P > 0$ value can only have meaning after birth, but comes with a negative value $N$ regardless of the specific values of $P$ and $N$. The action can therefore be only justified by inserting it into a larger ethical/moral framework, breaking the bubble where possible non-selfish reasons can be sought and advanced. Each bubble may only be locally computable or semi-computable at some scale. All other things being equal, we should favour decidability (computability) but in the longer term we may embrace semi-computable theories (such as LD, see glossary in the Appendix).

It only suffices to extrapolate a few years, a fraction of a second in cosmic time, to see how civilisations may disappear too soon to coexist at any given time. If so, consciousness and intelligence may appear as a misstep of evolution.

The question of procreation has an interesting logical side, and arguments in favour or against make clear how one has to devise an ethical framework able to answer or adapt to different scales. In this case, and by definition, deciding for someone else cannot be seen as universally ethical because forcing someone to exist is, by definition, a selfish act vis-à-vis the child, inside the parent-child bubble. Outside the parent-child bubble, secondary justifications are possible, but not between parent and child alone. The exercise also shows how ethics cannot be universal in all senses, being relative to time and hierarchical position, as it is not the same to take a decision not to procreate as to later ask the baby or the parent to kill that very human being, nor is it the same to locally
behave non-selfishly and globally face the consequences of local ethical rules. In other words, ethical frameworks can only be consistent if they are multiscale and hierarchical, and this should be a strong guiding principle, as suggested by Ruvinsky [22].

Figure 4: Diminishing population trends (2017 figures on the left) in developed and newly industrialised versus least developed (United Nations list of Least Developed Countries or UNLDC). If equally complex robots do not replace human beings, they will not only become extinct but won’t be replaced by an equally or more sophisticated entity to preserve complexity (or LD), suggesting a materialisation of the logical endgames of both antinatalism and digital antinatalism.

A general counter-argument to antinatalism is its possibly self-defeating nature. This is because if everybody adopts it, the consequence would be that the human race would quickly become extinct. This mode of reasoning is actually the mechanics of Kant’s categorical imperative: if what you profess cannot be universalised, then it should not be acted upon. Antinatalism would only be self-defeating, however, if the eventual extinction of humans or sentient beings (and of all necessary and unnecessary pain) were desirable, which is not necessarily the case, and it would be compatible with trans-humanist stances.

Perhaps what Kant meant was that local rules have to be consistent with global rules, but this shows that there is a question of scales and that hierarchies matter. For antinatalism, what goal/consequence would be more important? To preserve humans, for whatever reason, or to minimise necessary and unnecessary pain? Would minimisation of entropy imply antinatalism or digital antinatalism? It would seem that entropy minimisation and both antinatalism and digital antinatalism would have compatible goals.

4.3 Generalising Principles to AI Rules

An open question is whether the set of rules can or should be the same for AI and humans. Can we separate rules and principles in a way that is computable for AI? How can AI navigate contradictions in human ethical systems? Clearly
if sentient AI is endowed with the same value as sentient beings such as humans, they should, but most current frameworks do not consider this possibility and impose frameworks such as those incorporating the popular Asimov’s robot rules, whereby the value of robots is unequivocally less than that of humans.

A principle underlying a rule implies that no rule can always be guaranteed to preserve the spirit and purpose in which it was written. This can have foundations in computation too by way of undecidability and uncomputability. The idea is that no powerful axiom system or computational model can ever be complete enough to capture the spirit of the intention (the theorems and rules that can be derived) in which the axioms and rules of a system or model were chosen, leading to a continuous and open-ended quest to interrogate and update such rules in a bid to capture the first intention, i.e. the unreachable underlying principles.

AI should then be based on the principle of seeking the intention behind the rules and not merely following the rules, which does indeed go against the current approaches taken to AI ethics, especially in relation to Artificial General Intelligence (AGI), when rules are not powerful enough due to artificial constraints but are nevertheless intended to be forced and enforced. Rules are normative and should be followed, but guiding principles should provide an open exit towards finding the meaning behind the rules. This is why limiting AGI may also be dangerous, because an AGI trying to follow a rule as simple as ‘not to drive a car through a park’ may lead to the prosecution of a kid driving a toy car in the park. The law may not need to rule out such cases to qualify what truly counts as a car, not only because a judge and the police may have the context for deciding what counts as a car, but because it is often said that they can also read behind the intention of the law, what is called its principle. So how to do this for AI and AGI? Is it only a question of context? Not necessarily. The first thing to understand is that constraining a system to only blindly follow a set of (always arbitrary, if not equally arbitrary) rules, is likely, if not certain, to create contradictions. However, not forcing a set of rules that can instead be derived through an open-ended computational process that embraces rather than avoids uncomputability and undecidability can guide us in seeking intended meanings and principles, principles underlying rules that can be observed rather than enforced, thereby giving machines the opportunity to make apparently spontaneous adjustments to other sets of rules through computational emulability (harnessing the power of computational universality) so as to be able to emulate other ethical frameworks.

5 Conclusions

I have dissected some aspects of different information-theoretic, complexity, and computational approaches to ethics and how they can help fulfil and comply with some desirable features of ethical frameworks. I have identified some features that I believe may prove to be fundamental to making progress in connection with interfaces between human and machine ethics and even in connection to
human-human understanding.

I have contrasted some proposed digital and computational features of ethical frameworks and stances, and I have proposed some traits that appear desirable in such frameworks, in particular:

1. First and foremost an ethical framework should strive for logical, computational or algorithmic consistency, at least at same-level scales, or non-trivial inconsistency across different scales.
2. Local decidability: should lead to being able to calculate/estimate an action in a given (short) space and time.
3. Global undecidability: should provide purpose in order to make (2) scalable and open-ended.
4. Actions and beliefs should be discriminated in light of computational aspects such as emulability (and thus tractability), computability (in favour of what I have called universal empathy (Fig. 2)) and algorithmic probability/complexity (telling apart randomness and simplicity).

That is, an ethical framework should attempt to proscribe actions whose consequences its premises were intended to prevent. We have already seen an example of a (trivial) variation of a pro-life position.

We have found that a better ethical framework is one that is self-consistent and provides normative and guiding principles at different scales, particularly at very local (mostly normative) and very global scales (mostly guiding).

Finally, I have advocated that one of the greatest advantages of computational approaches is that they can be immediately implemented in computational systems to compute ethical/moral courses of action for humans or machines, without adaptation, thus being in some strong sense universal, as they may even be shared by other civilisations with similar computational theory foundations—which, based on the Church-Turing thesis, may easily be the case—for use in dictating and encoding moral behaviour in Artificial Moral Agents.

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Appendix

Glossary of Computational Terms

**Shannon Entropy:** A measure of combinatorial and statistical complexity based on the diversity of symbols used to define an object according to a probability distribution.

**Turing machine:** An abstraction of a general-purpose computer introduced by Alan M. Turing.

**Universal Turing machine:** A Turing machine that can emulate any other Turing machine by reprogramming it using proper inputs.

**Computability/recursivity:** In the context of real numbers, for example, all rational numbers are computable because there exists a well-defined algorithm that can calculate every number of their digital expansion. An example is 1/3, for which Euclid’s algorithm, to mention but one, would yield the infinite sequence 1.333...

**Semi-computability/recursive enumerability:** For example, the mathematical constant $\pi$ is recursively enumerable, as is any irrational number.

**Algorithmic randomness:** The ultimate compression ratio or rate of an object (beyond current implementations of lossless compression that are closer to entropy estimators). For example, $\pi$ is of low algorithmic randomness because there are highly compressed descriptions of the algorithms that implement the calculation of every number in the digital expansion of $\pi$.

**Logical depth and measures of sophistication:** The Logical Depth of a string is the time that an unfolding process takes to reproduce the string from its shortest description. The longer the time it takes, the deeper. Hence complex objects are those which can be seen as containing internal evidence of irreducible computational work.

**Kolmogorov-Chaitin complexity:** Also known as the algorithmic complexity of a string, it is the length of the shortest computer program that generates the string. The calculation of Kolmogorov-Chaitin complexity is an upper semi-computable problem, meaning upper bounds are arrived at by finding short, if not the shortest, computer programs.

**Algorithmic Information theory:** All the literature based on the concept of Kolmogorov-Chaitin complexity and related concepts such as algorithmic probability, compression, optimal inference, the Universal Distribution, Levin’s semi-measure.

**Semi-measure:** A measure of probability whose sum does not add up to 1 because some events are undetermined, for example, by the Halting problem in the context of Computability theory.

**Halting problem:** The strongest problem of predictability in algorithms related to whether or not a computer program will ever halt. Turing proved
that it is an undecidable problem and is equivalent to the undecidability results obtained by Gödel.

**Ω number:** A concise representation of the halting times in an irrational number called Ω by G. Chaitin [11] and defined as the halting probability of a universal computer programmed by coin tossing. The first n bits of Ω suffice to decide approximately the first $2^n$ cases of the halting problem but there is no faster way of extracting this information than to rerun the irreducible process until enough computer programs have been found to account for all but $1/2^n$ of the total halting probability Ω, a job which requires at least as much time as running the slowest n bit program. In other words, even though it solves unsolvable problems, Ω does not speed up the solution of solvable problems any more than a random coin-toss sequence would (this is an example of computation irreducibility).

**K₀ number:** As introduced by Bennett [5], $K₀$ contains all the information contained in Chaitin’s Ω but in an uncompressed fashion such that its index indicates directly, without further computation, whether a computer program with the index of the enumeration used to run $K₀$ will halt, thus making it extremely useful (equivalent to Borges’ infinite library but with only recursively/computational ‘truths’). In other words, Ω is a shallow representation of the logically deep object $K₀$.

**Undecidable problem or function:** A problem that cannot be decided by traditional algorithms, i.e. there is no algorithm that for any input (question) provides a definite output (answer).

**Decidability:** A problem or function that is not undecidable.

**Undecidability:** A problem or function that is undecidable.

**Semi-computability:** A semi-computable problem is one that allows approximations from above or below. If from above, then it is considered upper semi-computable, and if from below, then it is considered lower semi-computable.

**Algorithmic randomness:** How removed the length of the shortest generating program is from the size of the uncompressed data that such a program generates.

**Algorithmic Probability:** The probability of producing an object from a random digital computer program whose binary digits are chosen by chance. The calculation of algorithmic probability is a lower semi-computable problem.

**Invariance theorem:** A foundational theorem that establishes that shortest programs in different computer languages differ in length by at most a constant value, and thus guarantees that the Kolmogorov-Chaitin complexity of an object converges up to a constant, making this measure robust in the face of changes of model (such as a reference universal Turing machine).

**Coding Theorem:** A theorem that formally establishes an inversely proportional relationship between Kolmogorov-Chaitin complexity and algorithmic probability.