

Partial decidability protocol for the Wang tiling problem from statistical mechanics and chaotic mapping.

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We introduce a partial decidability protocol for the Wang tiling problem, which is the prototype of undecidable problems in combinatorics and statistical physics by constructing a suitable mapping from tilings of finite regions of different sizes. Such mapping depends on the initial family of Wang tiles (the "alphabet") with which one would like to tile the plane. This allows to define different mappings and temperatures associated to the tilings (compatible with the corresponding partition function). By finding a criterion of good tilings by ensuring that some two mappings and temperatures of a given alphabet are "well-behaved" in the thermodynamical sense, then such alphabet can tile the infinite two-dimensional plane. Our proposal is tested successfully with the known suitable good alphabets (which produce periodic tilings, aperiodic but self-similar tilings as well as tilings which are neither periodic nor self-similar). Our analysis shows that the boundary Tiling problem is undecidable in general. However, we show that the boundary Tiling problem is decidable in some cases. We also consider a new algorithm to decide if a given tiling is periodic or not. This algorithm is able to distinguish algorithms with a good thermodynamical behavior from algorithms with bad thermodynamical behavior. The transition from good to undecidable behavior is related to a transition from non-chaotic to chaotic regime in discrete dynamical systems of logistic type.

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I. INTRODUCTION

The Wang tiling problem [1] is one of the deepest and most relevant problems in statistical physics, mathematics and computer science. A Wang tile is a square with one color for each side. Given a finite family ("alphabet") of Wang tiles (from now on, a Wang tile will simply be denoted as tile), the problem is to understand whether or not one can cover the plane using these tiles¹ and satisfying certain rules. Here it is the original formulation [2]

¹ The rule of the game is that one can use truncated copies of these tiles in such a way that two tiles can be put together along one side if and only if the common side of two neighbors tiles has the same color on both. However, one can only use the tiles of the initial family but one cannot rotate the tiles.

Partial decidability Protocol for the Wang tiling Problem from Statistical Mechanics and Chaotic Mapping

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Organization of the talk

* What are undecidable problems?

* Examples of undecidable problems in physics

* The Wang tiling problem (WTP)

* Statistical Mechanics approach to WTP

* Partial decidability criteria from natural physical requirement

* Positive temperature and/or non-chaotic Mapping

* Undecidability and Transition to Chaos

* Conclusions and perspectives

The Green part of the title Justifies why this can be a physics talk !!

What are undecidable problems?

Undecidable problems are at the heart of the most important building blocks of Mathematics, Physics, Computer Science ...

In a rough form, some of them were already known in ancient Greek times

- *) "This sentence is false"
- *) "Which is the smallest integer that in order to be defined needs more than 150 letters?"

These two paradoxes show that undecidable statements appear not only due to self-reference But also in relation with minimality issues

The Brightest Examples are K. Gödel Theorems (1931)

Very Roughly speaking, a consequence of these is that if Mathematics is consistent then there are true statements which cannot be proven with the known and inferable rules of Mathematics

Then, in 1936, A. Turing in his foundational paper on Turing's machine showed that the HALTING PROBLEM is also undecidable

Halting Problem: Find program H (Python, C++, ...) such that H can decide whether or not another (generic) program P on a given Input I stops or runs forever.

Turing showed that such problem is undecidable

Namely, there is no such H that can work on any (P, I)

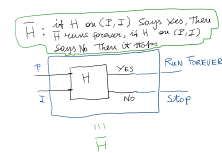
Note that Both the Program and the Input can be mapped into Binary Strings of 0s and 1s

Turing Proof: Suppose such H actually exists:

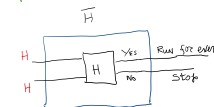
*) $\begin{matrix} \text{Yes} \\ \text{No} \end{matrix} \begin{matrix} \text{H} \\ \text{H} \end{matrix} \begin{matrix} \text{Yes} \\ \text{No} \end{matrix}$ means P stops on I

*) $\begin{matrix} \text{Yes} \\ \text{No} \end{matrix} \begin{matrix} \text{H} \\ \text{H} \end{matrix} \begin{matrix} \text{Yes} \\ \text{No} \end{matrix}$ means that P enters in an infinite loop on I and never stops

Then, if H exists, I can also construct $\bar{H} \sim \text{Not } H$



But then, we can give as inputs (I, I) to \bar{H} the couple (H, H)



if we now give (H, H) as input to \bar{H} there are 2 possibilities

- 1) if H given to H stops then it does not stop
 - 2) if H given to H does not stop then it stops
- We get a contradiction

Although we know very well that Turing result is very concrete since it asserts the impossibility of a universal "debugging Machine" (the dream of any computer scientist) at that time (1936) computers did not exist

Thus, Both Gödel results and Turing results were considered "twisted" "unnatural"

In the sense that many researchers believed that Mathematical problems which appear in usual applications in Science do not belong to the "undecidable" class

the decidability inference from the undecidability case. The results of the present manuscript strongly suggest that in the space of all alphabets such a boundary can be defined and, moreover, that some features with the tendency to show a discrete dynamical character can be described using small arguments from statistical mechanics. In order to proceed, let us define the following quantity:

$$W_1(n) = \frac{1}{n^2} \text{ (number of different tilings of } n \times n \text{ a square with the alphabet } \Sigma) \quad (1)$$

Let us also define the entropy $H(\Sigma)$ of an alphabet Σ as $n \times n$ square as the area $A(n)$ of the square (tend to n^2):

$$H(\Sigma) = \lim_{n \rightarrow \infty} \frac{1}{n^2} \log W_1(n) \quad (2)$$

Then, from the above definition it is clear that $W_1(n)$ plays the role of the degeneracy of the average level $H(\Sigma)$. Therefore, we can define the entropy $H(\Sigma)$ of Σ as the average over n as:

$$H(\Sigma) = \lim_{n \rightarrow \infty} \frac{1}{n^2} \log W_1(n) \quad (3)$$

In the following it will be convenient to use the \log_2 instead of the usual \ln , but conceptually there is no difference. From definition we proceed for the following reason:

(On the one hand, we expect that for alphabets Σ with good thermodynamical behavior there will be many different ways to tile a $n \times n$ square (other for most of physically reasonable choices the dependence of the entropy level is an increasing function of the entropy). This, in particular, implies that $W_1(n)$ is different from zero for arbitrarily large n . This last condition (which is a rather obvious fact in some physical context) prevents, among other things, the existence of a uniformly finite region (other, for each good alphabet, namely, compatible with a good thermodynamical behavior), both $W_1(n)$ and $H(\Sigma)$ should be increasing functions of n and, consequently, good alphabets can tile arbitrarily large regions. The last observation is equivalent to the typical thermodynamical condition related to the persistence of the temperature:

$$\frac{\partial H}{\partial T} > 0 \text{ or } \frac{\partial H(\Sigma)}{\partial T} > 0 \quad (4)$$

Otherwise, from the negative (negative) one can assume that an alphabet Σ with a good thermodynamical limit should have a well defined positive temperature and, consequently, Eq. (4) should be satisfied.

A very important function which captures the main thermodynamical features of a system is its partition function. Using the previous definitions we can define the partition function of the alphabet Σ as:

$$Z_\Sigma(\beta) = \sum_{n=1}^{\infty} W_1(n) \exp(-\beta n^2) \quad (5)$$

The above partition function encodes almost information about the alphabet Σ . For instance, if the alphabet Σ does not tile the plane then there exist an n^* with the property that:

$$W_1(n) = 0 \quad \forall n > n^* \quad W_1(n) > 0 \quad (6)$$

In otherwise, Σ would actually tile the plane¹. In this case, the partition function is analytic as $Z_\Sigma(\beta)$ will be just a polynomial in $\exp(-\beta)$. On the other hand, if $Z_\Sigma(\beta)$ is not analytic as function of β then necessarily there are infinitely many states in the state on the left-hand side of Eq. (5) (otherwise the function would be analytic) and this means that actually Σ tiles the whole plane. Consequently, the properties of $Z_\Sigma(\beta)$ are also worth to be further investigated for self-consistent back in time as future publications, but here below we will discuss a more direct approach to discuss the strong of undecidability².

¹ Namely, an alphabet which can tile arbitrarily large squares and which, therefore, allows to define the thermodynamical limit with the above definition of entropy according to Eq. (2) and (3).

² Indeed, the negation of the formula $\forall n, n > n^*, W_1(n) > 0$ is generally that $W_1(n) = 0$ in different form and is not arbitrarily large and not finite for some.

After Berger,

A very interesting and deeply analyzed issue is the following:

Construct aperiodic tiles with the smallest possible q

It is known that the smallest possible q leading to aperiodic tiles is $q=11$

Untie now, almost all the known examples of aperiodic tiling are self-similar

(kind of Sierpinsky structure)

there is only one example which is neither periodic nor self-similar

Decidability Protocol?

the Wang tiling problem is connected with many relevant problems in combinatorics, computer science and physics.

It would be fantastic to have a protocol (an algorithm) such that, given a family Γ , we can know whether or not Γ tiles the plane. This of course is impossible due to the fact that the Wang tiling problem is undecidable. But should we give up entirely?

the Wang tiling problem is NP-complete. This means that if we understand more WTP we understand more of all the NP complete problems.

$$\dim \mathcal{H}(\mathcal{F}) = q$$

$$\mathcal{P} = (X, Y) = (W_F(u), W_F(u+1)) \quad (7)$$

not locally) W_f (a

$$f_0(\alpha \otimes \beta) = f_0(\beta \otimes \alpha). \quad (2)$$

$\mu + 1/2 = H(\mu)$

III. SEPARATING "GOOD" FROM "BAD" MAPPINGS

Abstracts are available

otherwise it would be analytic

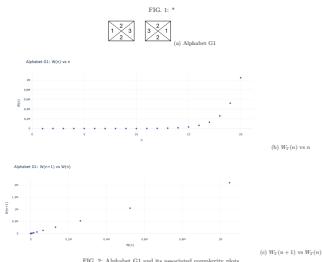


FIG. 2: Alphabet G1 and its associated complexity plots.

Advantages of the physical approach

It is clear that the above physical definitions encode all the relevant properties of a family Γ .

Even more is true!

Partial decidability protocols

Protocol 1

Physics intuition gives a useful hint.

First of all, as already emphasized, if we \mathbb{Z}_r possesses a thermodynamic limit, then Γ tiles the plane.

Moreover, for most physical systems with a thermodynamical limit, the entropy is an increasing function of the energy: $\frac{\partial S}{\partial E} > 0$

In our case, the above condition reduces to:

$$\boxed{\frac{\partial W_r}{\partial n} > 0} \quad *$$

This simple criterion (equivalent to requiring that the temperature is positive) gives rise to the

Protocol I: if the condition $*$ is

verified for your family Γ of interest then there is a good possibility that your family Γ tiles the plane!

Quite Remarkably this protocol ("Euristic") is able to identify all the families of the literature

Limitations of the Method

1) Note that this is only "Euristic": namely, in principle there can be Γ satisfying the criterion and which, nevertheless, do not tile the plane

2) Due to the limitations of the Hardware, the condition $\frac{\partial W_r}{\partial n} > 0$ can be verified only for a finite number of steps N^* . It would be ideal to have $N^* \gg 1$

However, quite surprisingly, in all the cases we checked successfully $N^* \sim 29$ was enough

Protocol II

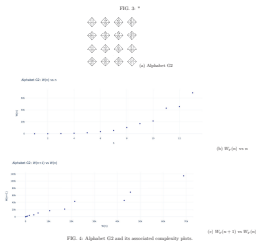


FIG. 1. (a) A 4x4 grid of squares, each containing a smaller square. (b) A plot of $W_p(n)$ vs n , showing a curve that starts at 0 and increases towards 1. (c) A plot of $W_p(n)$ vs n , showing a curve that starts at 0 and increases towards 1, with a different shape than (b).

The second protocol provides with more info not only on the positives but also on the negatives

Chaos and 1-Dimensional Mapping

let us consider the following 1-D discrete differential equation

Logistic Equation

$$x(n+1) = rx(n)(1-x(n))$$

depending on the value of the parameter

r the above equation can have regular or chaotic behavior, in general, if one considers

$$x(n+1) = f(x(n))$$

depending on the shape of the function $f(x)$

the discrete differential equation can be chaotic or not

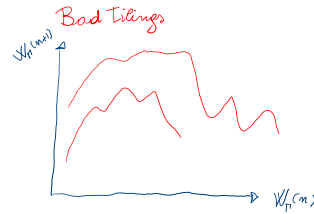
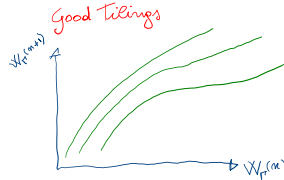
Appearance of Chaos

If the function $f(x)$ has a local "peaked" maximum then the discrete ODE is likely to be chaotic

Big Idea: Regular (non-chaotic) mapping are good!

Instead of plotting $W_p(n)$ vs n or $S_p(n)$ vs n

let us plot $W_p(n+1)$ vs $W_p(n)$ or $S_p(n+1)$ vs $S_p(n)$



Thus we clearly see that Bad tilings have manifest tendency to be chaotic good tilings on the other hand do not have local maxima!

Using both of these protocols we have been able to identify successfully all the good families present in the literature.

Our protocols are the only known a priori test available in the literature which have been able to identify all the good families

(periodic, non-periodic but self-similar, non-periodic non-self-similar)

Potential applications

The potential applications are huge:

The Wang-tiling problem belongs to the NP-complete class

This means that many other famous problems in Stat.Mech., Computer Science, discrete Math. can be mapped into Wang Tiling Problems

Thus, we can use our approach as criteria to guess a priori whether or not a given problem has a solution!

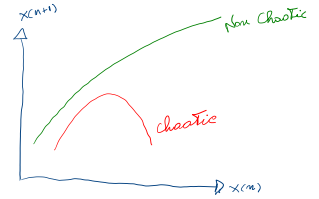
Limitations

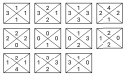
* There is a practical limitation due to the fact that we can test our criteria only for a finite range of square sizes. Namely, we can arrive until a square $N_{max} \times N_{max}$ We would like $N_{max} \gg 1$

But we can only arrive at $N_{max} \sim 32$ at best

* Rigorous sufficient conditions are hard to establish

Despite this, our protocols work anyway!



(a) Alphabet \mathcal{G} (b) $N(n)$ vs n (c) $N(n+1)$ vs $N(n)$ FIG. 6: Alphabet \mathcal{G} and its associated complexity plots.

an regression analysis is clearly demonstrated in Table I, which presents the desired regression coefficients using $N(n+1)$ vs $N(n)$.

To experimentally analyze the growth nonuniformity of $N(n+1)$ versus $N(n)$ as defined in Eq.(12), we perform a regression analysis for each alphabet. Instead of Pearson's correlation coefficient, we employ Kendall's Tau [17] to assess the relationship between data pairs. This choice is specifically motivated by our primary focus on nonuniformity rather than strict linearity, and by the superior robustness of Kendall's Tau to outliers. Subsequently, we compare the alphas based on their respective regression coefficients α and γ , as well as the Kendall's Tau coefficient [17].

For the most salubrious instance, the alphas which can cover the whole plane) $f_1(N_1(n)) = f_2(n)$ is very close to a power law (with very small coefficient).

$$f_1(n) = n^{\alpha} (1 + o(1)), \quad \forall \alpha > 0. \quad (13)$$

Conclusions

*) Using sound arguments from statistical mechanics and from the theory of chaos we have found the

first effective a priori tests that are able to identify the family of

Wang tiles which can tile \mathbb{R}^2

*) Our approaches also disclose a close relation between the "transition from decidable to undecidable" and the transition from "regular to chaotic behavior" in discrete dynamical system

→ negative criteria

*) We can apply our strategy to other

NP-complete / undecidable problems

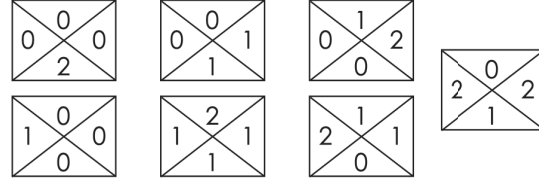
(graph Theory) (spectral gap...)

*) In particular: spectral gap problem and transition to chaos?

When your family Γ of interest generates a tiling, which is chaotic then it is unlikely that one can prove that Γ tiles the \mathbb{R}^2 even if it does

I Thank You Very Much!

FIG. 7: *



(a) Alphabet B1

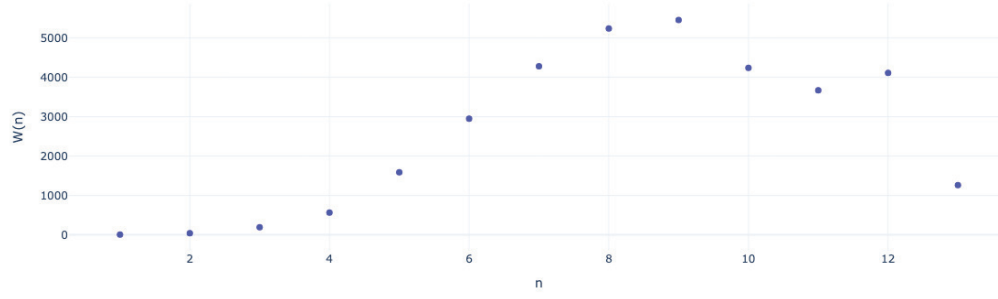
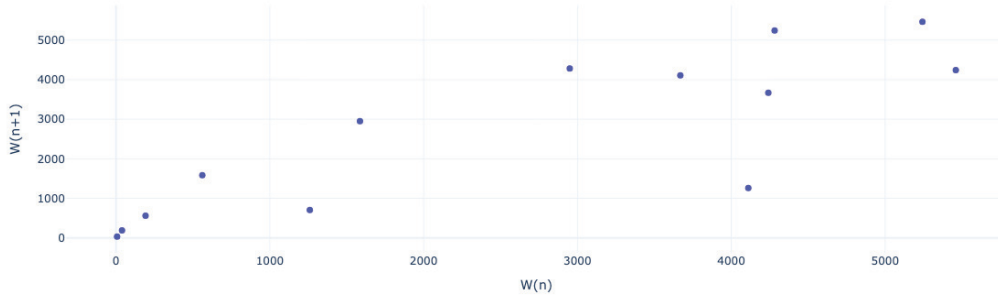
Alphabet B1: $W(n)$ vs n (b) $W_{\Gamma}(n)$ vs n Alphabet B1: $W(n+1)$ vs $W(n)$ (c) $W_{\Gamma}(n+1)$ vs $W_{\Gamma}(n)$

FIG. 8: Alphabet B1 and its associated complexity plots.

where c_0 is a constant and $\omega(z)$ is an oscillating function with a small amplitude:

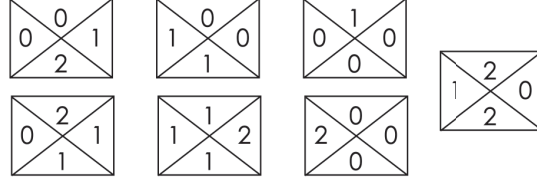
$$|\omega(z)| \ll 1. \quad (12)$$

Alphabet	c_0	γ	τ
G1	1.25	0.899	1.00
G2	1.49	0.754	1.00
G3	1.54	0.810	1.00
B1	1.75	0.550	0.72
B2	1.75	0.550	0.72
B3	1.86	0.596	0.79

TABLE I: Table I: Alphabets Regression coeff

Figure 2 presents the complexity analysis for Alphabet G1. This alphabet demonstrates a remarkably predictable growth pattern, characteristic of "good" alphabets, evident in both its entropy plots and regression coefficients. Figure

FIG. 9: *



(a) Alphabet B2

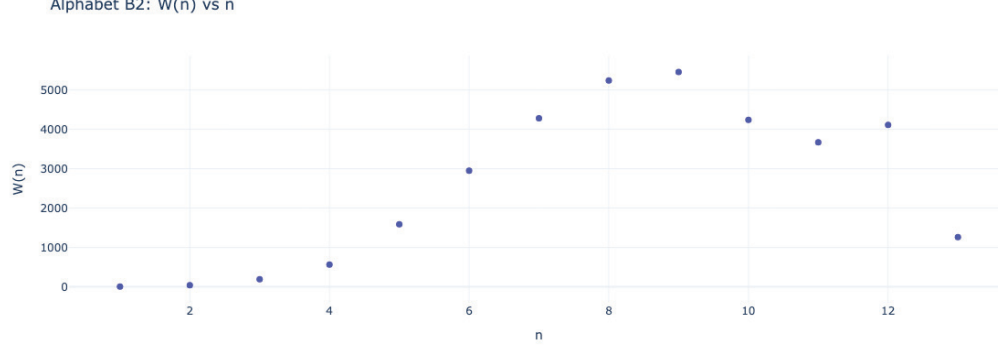
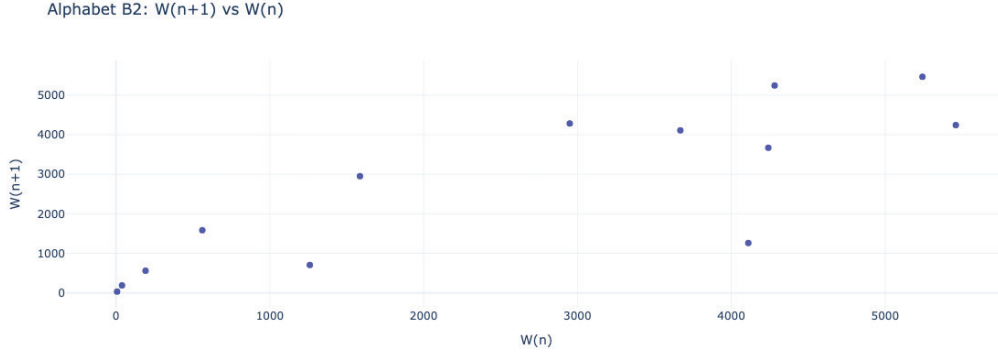
(b) $W_{\Gamma}(n)$ vs n (c) $W_{\Gamma}(n+1)$ vs $W_{\Gamma}(n)$

FIG. 10: Alphabet B2 and its associated complexity plots.

14(b) shows a consistent and monotonic increase in $S_{\Gamma}(n)$ with n , indicating a stable, non-erratic growth in the number of distinct tiles. The core analysis of growth monotonicity, as depicted in Figure 14(c) plotting $S_{\Gamma}(n+1)$ against $S_{\Gamma}(n)$, reveals an almost perfectly linear relationship between successive entropy values. This strong linearity is quantitatively supported by the regression coefficients from Table I: a scaling factor $c_0 = 1.25$ indicates a consistent base growth rate, and an exponent $\gamma = 0.899$ (approaching unity) signifies a nearly linear relationship in the logarithmic domain, consistent with stable complexity evolution. Crucially, a Kendall's Tau coefficient of $\tau = 1.00$ confirms a perfect positive monotonic relationship, meaning $S_{\Gamma}(n+1)$ invariably increases with $S_{\Gamma}(n)$, validating the consistent and predictable growth inherent to "good" alphabets. This empirical evidence for minimal oscillation strongly supports the theoretical condition $|\omega(z)| \ll 1$ in Equation (11).

Figure 16 displays another example of a "good" alphabet (G2). It is illustrated the growth of $S_{\Gamma}(n)$ with n , showing a consistent, albeit slightly less uniform, monotonic increase in entropy compared to G1. This indicates a generally stable increase in tiles complexity. The relationship between $S_{\Gamma}(n+1)$ and $S_{\Gamma}(n)$, depicted in Figure 16(c), reveals a positive linear correlation, with most data points closely adhering to the regression line. From Table I, Alphabet G2 exhibits regression coefficients $c_0 = 1.49$ and $\gamma = 0.754$. While γ is slightly lower than G1, it still signifies a robust, near-linear growth in the logarithmic domain, consistent with predictable complexity. More critically, the Kendall's Tau coefficient $\tau = 1.00$ confirms a perfect positive monotonic relationship, reinforcing that $S_{\Gamma}(n+1)$ consistently increases with $S_{\Gamma}(n)$. This perfect monotonicity, combined with the high linearity, categorizes Alphabet G2 as having

FIG. 11: *

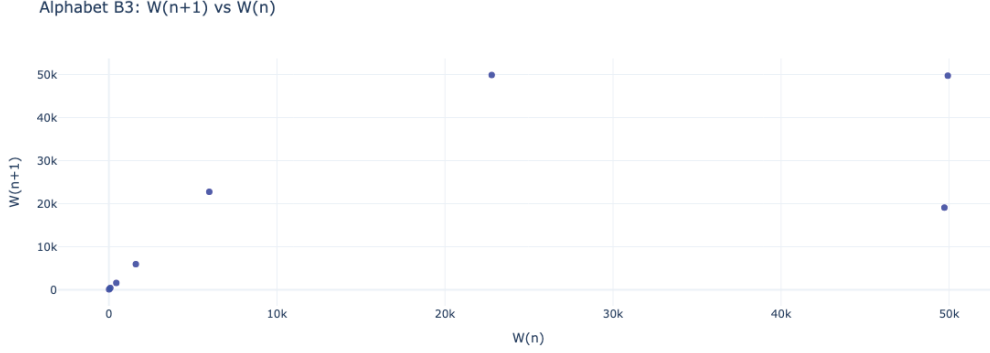
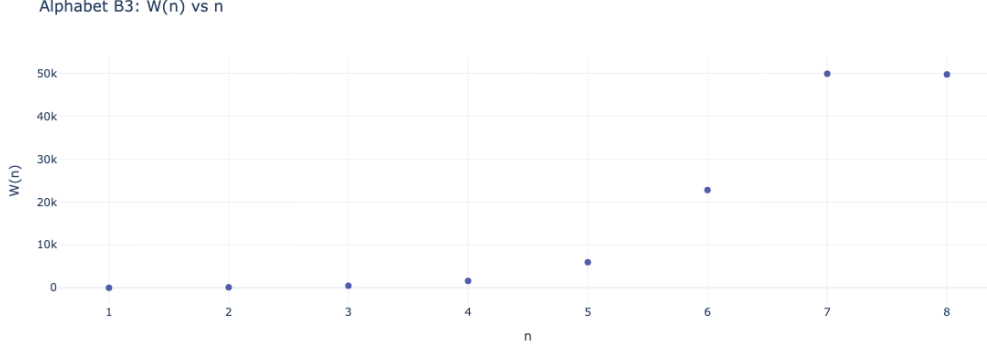
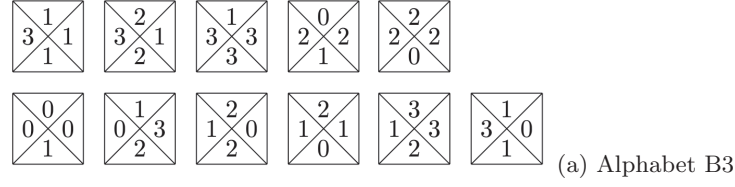


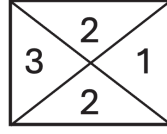
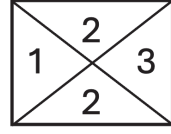
FIG. 12: Alphabet B3 and its associated complexity plots.

highly stable and desirable growth characteristics, fully supporting the minimal oscillation condition $|\omega(z)| \ll 1$ in Equation (11).

Figure 18 presents the complexity analysis for Alphabet G3, another example of a "good" alphabet. Figure 18(b) illustrates the growth of $S_{\Gamma}(n)$ with n , showing a consistent monotonic increase in entropy, indicative of stable combinatorial expansion. The core analysis of growth monotonicity, depicted in Figure 18(c) plotting $S_{\Gamma}(n+1)$ against $S_{\Gamma}(n)$, reveals a positive linear correlation, with data points closely adhering to the regression line. As per Table I, Alphabet G3 exhibits regression coefficients $c_0 = 1.54$ and $\gamma = 0.810$. The γ value, close to unity, consistently signifies a robust, near-linear growth in the logarithmic domain, characteristic of predictable complexity. Critically, the Kendall's Tau coefficient $\tau = 1.00$ confirms a perfect positive monotonic relationship, where $S_{\Gamma}(n+1)$ consistently increases with $S_{\Gamma}(n)$. This perfect monotonicity, combined with the observed linearity, firmly categorizes Alphabet G3 as having highly stable and desirable growth characteristics, fully supporting the minimal oscillation condition $|\omega(z)| \ll 1$ in Equation (11).

Figures 20 and 22 depict the complexity analysis for Alphabets B1 and B2, respectively. Despite originating from distinct alphabet structures, their $S_{\Gamma}(n)$ plots and regression behaviors are empirically identical, classifying them as "bad" alphabets. Figure 20(b) (and similarly for B2) illustrates that $S_{\Gamma}(n)$ still shows an increasing trend with n , but with noticeable plateaus and less uniform steps, hinting at a less stable combinatorial growth. The key divergence from "good" alphabets is evident in the $S_{\Gamma}(n+1)$ vs $S_{\Gamma}(n)$ plot (Figure 20(c)). While a general linear trend can be

FIG. 13: *



(a) Alphabet G1

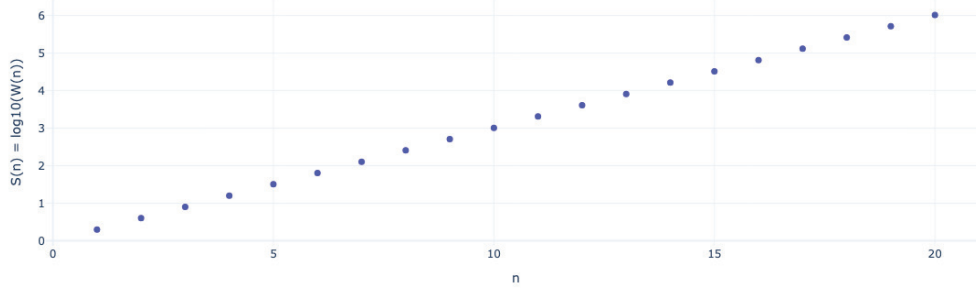
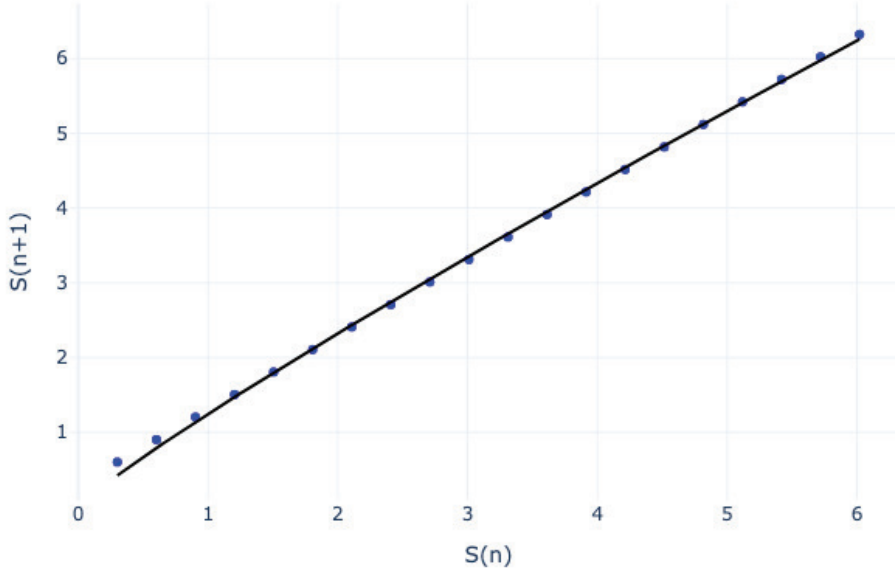
Alphabet G1: $S(n)$ vs n (b) $S_{\Gamma}(n)$ vs n Alphabet G1: $S(n+1)$ vs $S(n)$ (c) $S_{\Gamma}(n+1)$ vs $S_{\Gamma}(n)$

FIG. 14: Alphabet G1 and its associated entropy plots.

observed, there is significant dispersion of data points around the regression line, especially at higher $S_{\Gamma}(n)$ values, indicating structural inconsistencies and a less predictable relationship between successive entropy values. From Table I, both Alphabets B1 and B2 share coefficients: $c_0 = 1.75$ and $\gamma = 0.550$. The γ value, being significantly lower than unity, indicates a weaker, sub-linear growth in the logarithmic domain, suggesting that the relative increase in complexity diminishes as n grows. This contrasts sharply with the stable growth seen in "good" alphabets. Crucially, the Kendall's Tau coefficient $\tau = 0.72$ is markedly less than 1.00. This value confirms a positive correlation but

FIG. 15: *

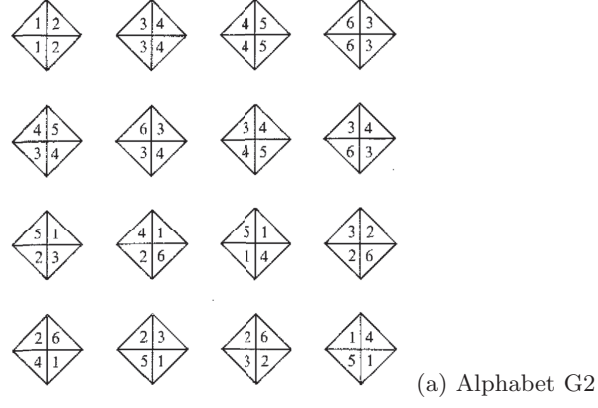
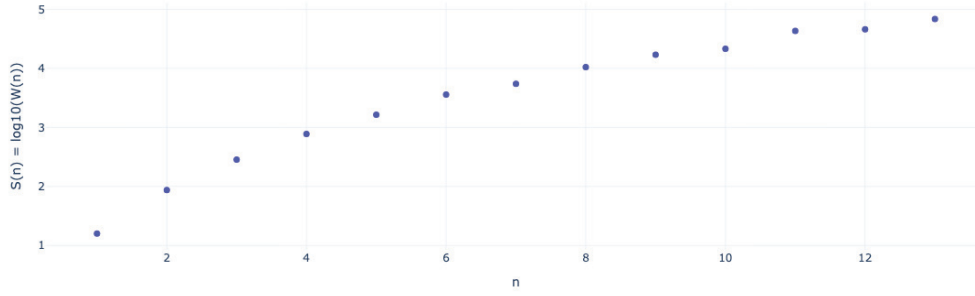
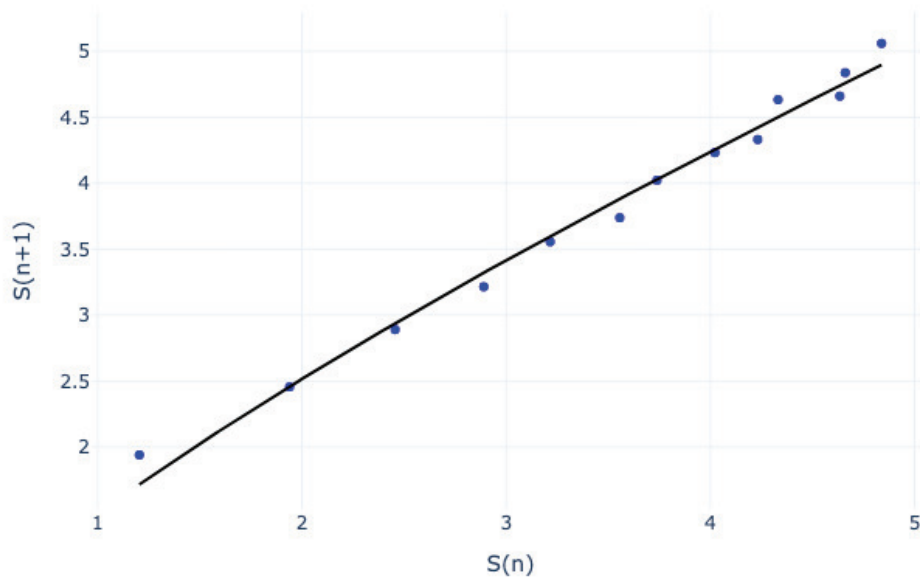
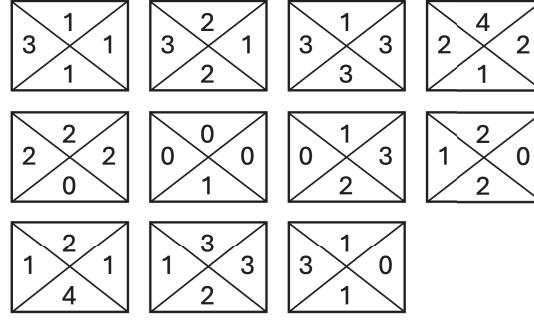
Alphabet G2: $S(n)$ vs n (b) $S_{\Gamma}(n)$ vs n Alphabet G2: $S(n+1)$ vs $S(n)$ (c) $S_{\Gamma}(n+1)$ vs $S_{\Gamma}(n)$

FIG. 16: Alphabet G2 and its associated entropy plots. While the behavior is not linear, it exhibits a monotonically increasing trend.

FIG. 17: *



(a) Alphabet G3

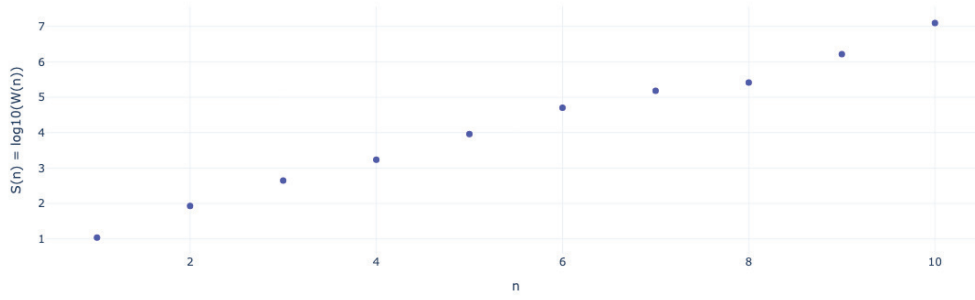
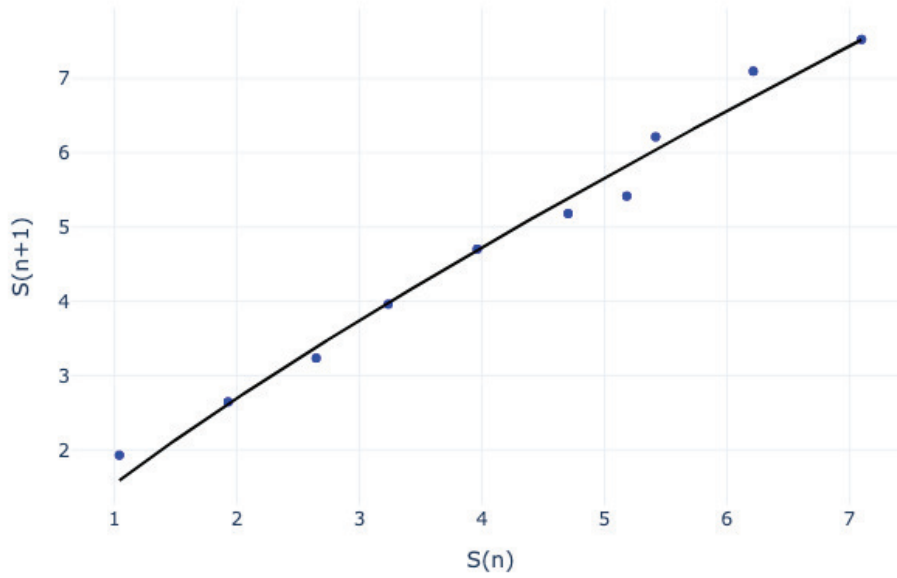
Alphabet G3: $S(n)$ vs n (b) $S_{\Gamma}(n)$ vs n Alphabet G3: $S(n+1)$ vs $S(n)$ (c) $S_{\Gamma}(n+1)$ vs $S_{\Gamma}(n)$

FIG. 18: Alphabet G3 and its associated entropy plots.

explicitly indicates the presence of monotonic inversions or plateaus, where $S_\Gamma(n+1)$ does not consistently increase with $S_\Gamma(n)$. This departure from perfect monotonicity is the defining characteristic of "bad" alphabets, reflecting the limitations in their combinatorial expansion and highlighting the non-trivial oscillations of $\omega(z)$ in Equation (11).

Figure 24 presents the complexity analysis for Alphabet B3, which also falls into the category of "bad" alphabets. Figure 24(b) illustrates the behavior of $S_\Gamma(n)$ with n . Similar to B1 and B2, while an overall increasing trend is observed, the growth is not perfectly smooth, showing some irregularities in the rate of entropy increase. More critically, the $S_\Gamma(n+1)$ vs $S_\Gamma(n)$ plot in Figure 24(c) demonstrates significant deviation from a strict linear relationship, exhibiting a noticeable scatter of data points around the regression line. This scatter is particularly pronounced at higher $S_\Gamma(n)$ values, indicating a less predictable and less stable progression of complexity. From Table I, Alphabet B3 has regression coefficients $c_0 = 1.86$ and $\gamma = 0.596$. The γ value, significantly less than unity, confirms a sub-linear growth in the logarithmic domain, implying that the incremental increase in complexity diminishes with increasing sequence length, a hallmark of "bad" alphabets. The most telling characteristic is the Kendall's Tau coefficient $\tau = 0.79$, which, while higher than B1/B2, remains significantly below 1.00. This value definitively indicates the presence of monotonic violations, where $S_\Gamma(n+1)$ does not consistently increase with $S_\Gamma(n)$, reflecting the inherent structural inconsistencies and the non-negligible oscillations of $\omega(z)$ in Equation (11).

The comprehensive analysis of alphabets, based on their entropy growth $S_\Gamma(n)$, reveals a clear distinction in their combinatorial behavior. The results for "bad" alphabets, exemplified by B1, B2, and B3, consistently indicate a sub-linear growth in entropy (characterized by $\gamma < 1$) and a loss of perfect monotonicity (indicated by $\tau < 1.00$). This behavior is consistent with an inherent limitation in their ability to efficiently tile the entire space, often stemming from the absence of superposable structures. This diminished growth rate and structural inconsistency is precisely corroborated by the specific c_0 , γ , and τ values presented in Table I, and visibly reinforced by the corresponding $S_\Gamma(n+1)$ vs $S_\Gamma(n)$ plots. Conversely, "good" alphabets (G1, G2, G3) exhibit near-linear growth in the logarithmic domain ($\gamma \approx 1$) and perfect monotonicity ($\tau = 1.00$), demonstrating stable and predictable complexity expansion. Consequently, these empirical observations provide a robust basis for distinguishing between alphabets with *good* and *bad* tiling properties based on their entropy scaling and monotonic characteristics.

For bad alphabets, the numerical fits of the plots can still be described by Eq. (11). On the other hand, the condition in Eq. (12) is violated for bad alphabets, namely the oscillations around the power law behavior are large.

Thus, the *partial decidability protocols* proposed in the present manuscript is based on this pronounced difference. In simple terms, the protocol reads as follows:

Step 1) Compute numerically as many values as possible of the function $W_\Gamma(n)$ (subject to the available resources, the ideal situation being having n_{\max} larger than q of, at least, a factor of 2 or 3: see the comments below Eq. (6)) for the alphabet Γ of interest.

Step 2) Due to the large amount of combinations, it is better to use the entropy $S_\Gamma(n)$ as main variable.

Step 3) Produce the plot $P = (X, Y) = (S_\Gamma(n), S_\Gamma(n+1))$ as described in the previous section.

Step 4) Construct the best fit of $S_\Gamma(n+1) = f_\Gamma(S_\Gamma(n))$ and compute the employ Kendall's Tau (τ).

IV. PHYSICAL INTERPRETATION IN TERMS OF DISCRETE CHAOS

As it has been already emphasized in the introduction, there is a deep connections between deterministic chaos and undecidability since the Chaitin's results in [18]. The "coexistence" of determinism and undecidability (see, for instance, the detailed analysis in [19] [20] [21]) implies that undecidability can manifest itself in dynamical systems manifesting chaotic behavior. The idea to use the second protocol instead of the first (simpler) protocol is that the second one reveals interesting informations on the transition from good to bad alphabets. The qualitative explanation of the effectiveness of the present approach is based on the transition to chaos in discrete mappings of logistic type (see, for a detailed review on discrete chaos, [17]).

Let us interpret

$$S_\Gamma(n+1) = f_\Gamma(S_\Gamma(n)) , \quad (13)$$

as a discrete dynamical system (a possible choice discussed in the previous section is $f_\Gamma(z) = c_0 z^\gamma (1 + \omega(z))$ although the results in the present sections apply to generic form of $f_\Gamma(z)$).

In particular, if $f_\Gamma(z)$ is *monotone increasing*⁴ (namely, $\frac{d}{dz} f_\Gamma(z) > 0$) the solutions of the above dynamical system

⁴ According to the present numerical results, this corresponds (when the parametrization $f_\Gamma(z) = c_0 z^\gamma (1 + \omega(z))$ is used) to the case in which $|\omega(z)| \ll 1$.

will tend to the solutions of the simpler (non-chaotic) dynamical system here below:

$$S_{\Gamma}(n+1) = f_{\Gamma}^{(0)}(S_{\Gamma}(n)) \quad , \quad f_{\Gamma}^{(0)}(z) = c_0 z^{\gamma} \quad , \quad (14)$$

according to which $W_{\Gamma}(n)$ grows exponentially with n with subexponential corrections.

On the other hand, if $f_{\Gamma}(z)$ is *not monotone increasing*⁵ (namely, $\frac{d}{dz}f_{\Gamma}(z)$ vanishes and changes sign, generically, more than once) *and local maxima and minima appear*, the dynamical system in Eq. (13) can enter into a chaotic phase. In particular, the more peaked is the local maximum, the closer is the shape of $f_{\Gamma}(z)$ to a logistic map in the chaotic phase. Thus, although the explicit analytic form of $f_{\Gamma}(z)$ is not available, the present results show that the maps $f_{\Gamma}(z)$ associated to good alphabets do not manifest chaotic tendency while the ones associated to bad alphabets manifest a clear chaotic tendency.

In particular, in the terminal region of the experimental data for non-tiling ("bad") alphabets, one can observe a departure from sustained growth, manifesting as a decreasing trend in the relationship between successive values. While initial iterations may exhibit increasing values of $S(n+1)$ with respect to $S(n)$, the final experimental points reveal a clear inversion, indicating that larger values of $S(n)$ are associated with smaller subsequent values of $S(n+1)$. This decay suggests an inherent constraint within these alphabets that limits their expansive behavior.

This eventual decrease in $S(n+1)$ as $S(n)$ increases can be qualitatively described by a functional form reminiscent of a downward-opening parabola. Although not necessarily an exact fit, a quadratic model of the type

$$S(n+1) \approx S_0 - a[S(n) - b]^2 \quad , \quad (15)$$

where S_0 , $a > 0$, and b are constants, captures the essence of this behavior. This form illustrates how, beyond a certain point related to b , increasing $S(n)$ leads to a reduction in $S(n+1)$, ultimately resulting in the observed decay. This tendency towards a decreasing relationship at larger values of $S(n)$ appears to be a distinguishing feature of alphabets that are unable to tile the plane: the similarity with the logistic map is manifest. We will come back on this very interesting issue in a future publication.

V. QUANTIFYING "GOODNESS" AND "BADNESS" OF ALPHABETS

In this section, we will discuss how the physical arguments in the previous sections are actually confirmed by a statistical analysis of the numerical data. In complex systems involving sequential or time-dependent data, quantifying the relative quality of different datasets presents significant analytical challenges. This section examines how correlation measures can serve as effective tools for evaluating monotonic relationships within our alphabets, allowing us to distinguish between what we term "good" and "bad" behaviors.

When analyzing correlation in datasets with extreme values and seeking to identify monotonic trends rather than strictly linear relationships, the choice of correlation coefficient becomes crucial. While the Pearson correlation coefficient is commonly employed in statistical analysis, its limitations make it suboptimal for our specific requirements.

The Pearson correlation coefficient (r) measures the linear relationship between variables:

$$r = \frac{\text{cov}(X, Y)}{\sigma_X \sigma_Y}$$

However, this measure presents several drawbacks for our analysis. Pearson's coefficient assumes linearity in relationships, is highly sensitive to extreme values and outliers, requires normally distributed data, and can be significantly influenced by the scale and magnitude of observations. These characteristics make it inadequate for our datasets, which contain extreme values and where we are primarily concerned with monotonic trends rather than linear relationships.

Kendall's Tau (τ), a non-parametric measure of rank correlation, offers a more suitable alternative for our analysis:

$$\tau = \frac{(\text{Number of concordant pairs}) - (\text{Number of discordant pairs})}{\frac{1}{2}n(n-1)}$$

⁵ According to the present numerical results, this corresponds (when the parametrization $f_{\Gamma}(z) = c_0 z^{\gamma} (1 + \omega(z))$ is used) to the case in which $|\omega(z)|$ is not small compared to 1.

Kendall’s Tau evaluates correlation based on the concordance of pairs—whether the ranks of paired observations move in the same direction—rather than the magnitude of differences. This approach provides several advantages for our specific analytical needs:

First, Kendall’s Tau is particularly robust when dealing with datasets containing outliers or extreme values, as is the case with several of our alphabets. Because it focuses on the relative ordering of data points rather than their actual values, the coefficient is less affected by unusually large or small observations.

Second, Kendall’s Tau excels at detecting monotonic relationships without assuming linearity. Our primary interest lies in determining whether the regression patterns across different alphabets consistently increase or decrease, making monotonic behavior more relevant than strict linearity.

Third, the coefficient maintains its reliability even with smaller sample sizes, which is beneficial when analyzing rarer alphabetic patterns.

Kendall’s Tau has demonstrated its utility across numerous scientific disciplines. In environmental science, researchers have employed it to detect monotonic trends in climate data where extreme weather events might skew other correlation measures [22]. In economics, it has proven valuable for analyzing market behaviors when dealing with volatile price movements [23]. Researchers regularly use Kendall’s Tau when examining relationships between particulate matter progression, particularly when distributions are non-normal [24].

Our methodology involves applying Kendall’s Tau to measure the monotonic trends present in different alphabets. For “good” alphabets, we observe high τ values, indicating strong monotonic relationships consistent with theoretical expectations. Conversely, “bad” alphabets exhibit lower τ values, reflecting weaker monotonic behavior and greater dispersion in the data.

By employing Kendall’s Tau as our primary correlation measure, we establish a robust framework for distinguishing between alphabets that exhibit consistent, predictable behavior (high τ values) and those that demonstrate erratic patterns (low τ values). This approach provides an objective metric for quantifying the relative “goodness” or “badness” of different alphabets, facilitating more rigorous comparative analysis and more reliable conclusions about their underlying properties.

VI. RESUMING THE RESULTS OF THE PRESENT WORK

Here we will resume what we have done in the present work. We have associated to any alphabet Γ a temperature, an entropy and a partition function. The requirement of a good thermodynamical behavior provides one with a very nice criteria (sufficient conditions) ensuring that Γ can tile the plane. The simplest of all is the first protocol which only uses the fact that in most of physically reasonable systems the degeneracy of the energy level is an increasing function of the energy. The second protocol which has been defined discloses a nice analogy between the transition from good to bad alphabets and the transition from regular to chaotic behavior in discrete mapping of logistic type. In order to use these idea in practice, the Kendall’s Tau must be used. It seems that the present strategy has a quite wide range of possible applications.

A. What we have not done

As far as the first protocol is concerned, the idea is physically simple and correct but in practice one cannot compute $W_\Gamma(n)$ for any n so that one has to stop at some finite n_{\max} . We do not have a rigorous answer to the question of how large n_{\max} should be (compared to q) in order to be sure that the alphabet under examination is a good one. For instance, we cannot exclude the following situation. It could happen that we have the numerical verification of Eq. (4) for $(1, 2, 3, \dots, q \cdot 10^{10^q})$ but then, when $n = 1 + q \cdot 10^{10^q}$, the condition in Eq. (4) is violated. On the other hand, in the examples we have examined (where we have periodic tilings, aperiodic self-similar tiling, tilings which are neither periodic nor self-similar) it seems that if n_{\max} is 2 or 3 times q is large enough. Similar considerations hold for the second protocol.

As far as the similarity of the transition from good to bad alphabets and of the transition from regular to chaotic behavior in discrete mapping of logistic types, we have not shown rigorously that bad alphabets are always associated to chaotic discrete mappings while good alphabet to regular ones. The difficulty lies in the fact that the analytic form of $f_\Gamma(z)$ cannot be computed and so one has to use numerical fits of the mapping $f_\Gamma(z)$. The presence or absence of chaotic behavior may depend on the family of functions which are chosen to do the fit. However, the connections between deterministic chaos and undecidability (discussed in details in [18] [19] [20] [21] and references therein) lead to the fact that the existence of deterministic chaos in dynamical systems can be related to a sort of Godelian undecidability rather than to the typical numerical untractability. Since the Wang tiling problem is also undecidable,

it is reasonable to assume that discrete mappings associated to Wang alphabets manifest their undecidability through chaotic behavior (consistently with [19] [20] [21]).

We will come back on these interesting issues in a future publication.

VII. CONCLUSIONS AND PERSPECTIVES

In the present manuscript we have proposed two partial decidability protocols for the Wang tilings problem (which is the prototype of the undecidable problem in combinatorics and statistical mechanics). The idea is to define effective entropy and temperature associated to any alphabet Γ (together with the corresponding partition function). A subclass of *good alphabets* can be identified by requiring, basically, a good thermodynamical behavior. This proposal has been tested successfully with the known available good alphabets (which produce periodic tilings, aperiodic but self-similar tilings as well as tilings which are neither periodic nor self-similar). From the theoretical physics viewpoint, it is a very intriguing result to be able to reduce the undecidable instances of the Wang tiling problem using sound arguments from statistical mechanics. The present analysis also shows that the transition from good to bad alphabets is very similar to the transition from regular to chaotic behavior in discrete mappings of logistic type.

From the practical viewpoint, the present results are quite powerful. The computational challenges encountered in scaling Wang tilings, particularly the rapid increase in effort required to expand the tiling plane with larger alphabets, suggest that determining whether a given alphabet can tile the infinite plane may be an inherently difficult problem, potentially residing within the NP-hard class. The observed distinction between "good" alphabets, which allow for tiling, and "bad" alphabets, which exhibit eventual decreasing trends in their associated mappings, further hints at a fundamental difference in their complexity. While this study does not definitively resolve the P vs NP problem, the apparent difficulty in efficiently finding a tiling solution, even when a valid configuration can be readily verified, aligns with the prevailing conjecture that $P \neq NP$, implying that the Wang tiling problem could represent a scenario where finding a solution is significantly more computationally demanding than checking its correctness.

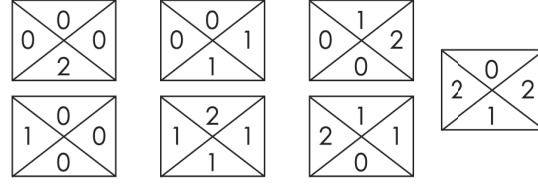
Acknowledgements

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FIG. 19: *



(a) Alphabet B1

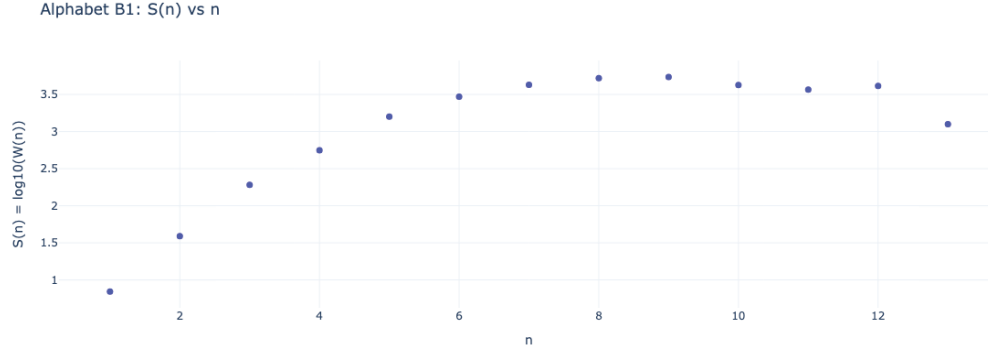
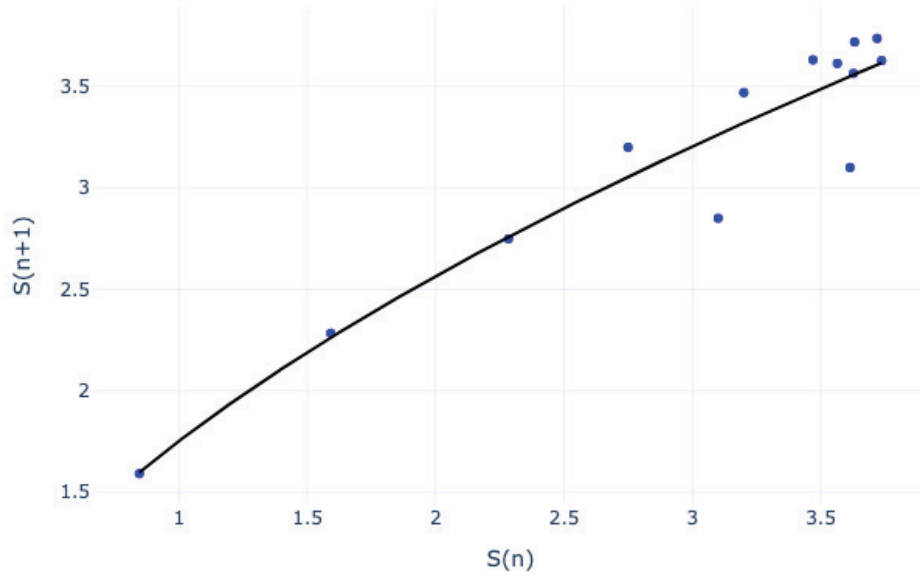
(b) $S_{\Gamma}(n)$ vs n Alphabet B1: $S(n+1)$ vs $S(n)$ (c) $S_{\Gamma}(n+1)$ vs $S_{\Gamma}(n)$

FIG. 20: Alphabet B1 and its associated entropy plots.

FIG. 23: *

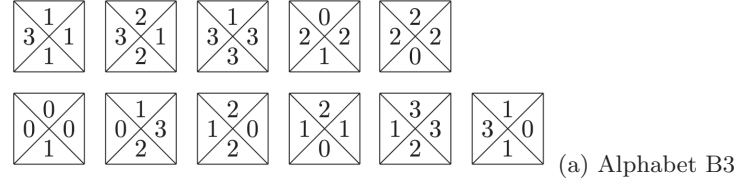
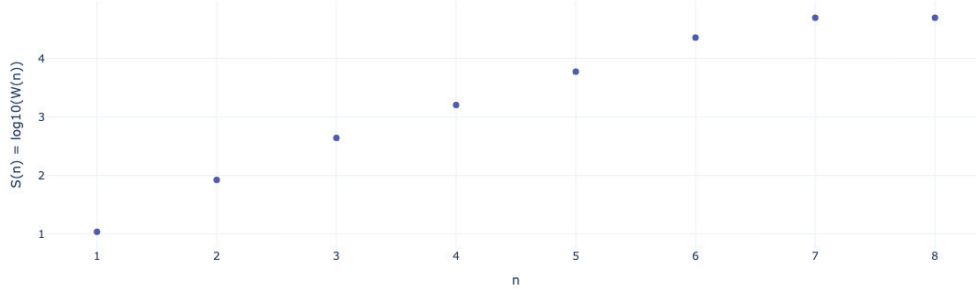
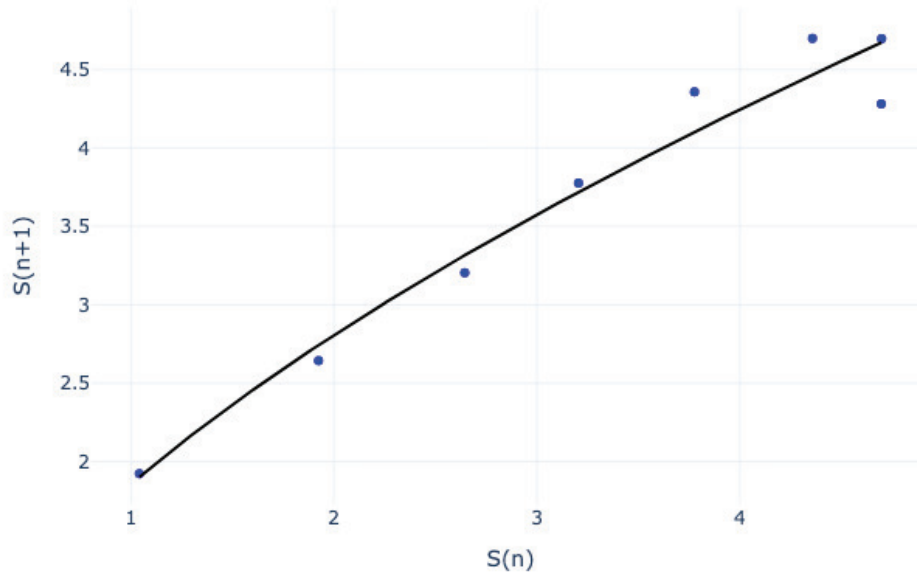
Alphabet B3: $S(n)$ vs n (b) $S_{\Gamma}(n)$ vs n Alphabet B3: $S(n+1)$ vs $S(n)$ (c) $S_{\Gamma}(n+1)$ vs $S_{\Gamma}(n)$

FIG. 24: Alphabet B3 and its associated entropy plots.