



1 **Do phenomenological dynamical paleoclimate models have physical similarity**
2 **with nature?**

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8 **Abstract.** Phenomenological models may be impressive in reproducing empirical time series but this is
9 not sufficient to claim physical similarity with nature until comparison of similarity parameters is
10 performed. We illustrated such a process of diagnostics of physical similarity by comparing the
11 phenomenological dynamical paleoclimate model of Leloup and Paillard (2022) with the more physically
12 explicit Verbitsky *et al* (2018) model and established that, for the late Pleistocene, there is considerable
13 physical similarity in terms of two crucial similarity parameters: (a) the ratio of the astronomical forcing
14 amplitude to the terrestrial ice-sheet mass influx, and (b) the ratio of amplitudes of time-dependent
15 positive and negative feedbacks.

16
17 **1. Introduction**

18 In the Epilogue of his monumental book “Dynamical Paleoclimatology” Barry Saltzman (2002)
19 reflects on mathematical modeling of ultra-slow, ice-age-like paleoclimate processes and advocates for a
20 phenomenological approach “through the construction of low-order models in which the full behavior is
21 projected onto the dynamics of a reduced number of ... highly aggregated variables...” as an alternative to
22 a more explicit approach when a dynamical system is derived from more sophisticated models or, ideally,
23 directly from basic laws of physics as it has been argued by Lorenz (1970). Now, 20 years later, this
24 dispute is still far from being settled, and dynamical paleoclimate models of both a phenomenological
25 nature (e.g., Saltzman and Maasch, 1991, Paillard, 1998, Tziperman et al, 2006, Crucifix, 2013, Leloup
26 and Paillard, 2022) and of a more explicit type (e.g., Talento and Ganopolski, 2021) are widely used in
27 the field without much consideration about their physical similarity or absence thereof. In this study, the
28 author who was earlier involved in the development of phenomenological models (e.g., Saltzman and
29 Verbitsky, 1992, 1993, 1994) but recently proposed a dynamical ice-age model derived from the
30 conservation laws of viscous ice media (Verbitsky *et al*, 2018) joins this discussion with a basic question:
31 *Do phenomenological dynamical paleoclimate models have physical similarity with nature?* In fluid



32 dynamics, the concept of physical similarity is the cornerstone of any judgement built on model
33 experimentations and classical similarity parameters such as the Reynolds number, the Peclet number, the
34 Euler number, etc., describing the relative importance of different aspects of fluid flow, can be derived
35 from fundamental conservation laws. The most basic logic of the physical-similarity approach is that such
36 relative importance in the model must be the same as in nature. A migration from explicit conservation
37 equations to highly integrated dynamical systems will definitely require different similarity parameters to
38 be preserved, but this does not mean that diagnostics of physical similarity can be neglected.

39 To establish physical similarity, regardless of the mathematical modeling approach we are going to
40 pursue (phenomenological or explicit), we first need to assume that there exists a natural parent
41 dynamical system that created the time series given to us as the observations. Let us assume further that
42 this parent system is governed by n physical parameters a_i such that a dependent variable of interest, x ,
43 can be expressed as function

$$44 \quad x = \varphi(a_1, a_2, \dots, a_i, \dots, a_n) \quad (1)$$

45 If k parameters of $a_1, a_2, \dots, a_i, \dots, a_n$ are parameters with independent dimensions, then, according to π -
46 theorem (Buckingham, 1914), in the dimensionless form, the phenomenon (1) can be described by
47 $m = n - k$ adimensional similarity parameters $\Pi_1, \Pi_2, \dots, \Pi_i, \dots, \Pi_m$:

$$48 \quad \Pi = \Phi(\Pi_1, \Pi_2, \dots, \Pi_i, \dots, \Pi_m) \quad (2)$$

49 Two physical phenomena have physical similarity if both of them are described in the adimensional
50 form by the same function $\Phi(\Pi_1, \Pi_2, \dots, \Pi_i, \dots, \Pi_m)$ and have identical numerical values of similarity
51 parameters $\Pi_1, \Pi_2, \dots, \Pi_i, \dots, \Pi_m$, though numerical values of the governing parameters
52 $a_1, a_2, \dots, a_i, \dots, a_n$ may be different (e.g., Barenblatt, 2003).

53 As we have already mentioned, our knowledge about a parent dynamical system is available to us
54 as empirical time series. It means that one of the similarity parameters, let say Π_1 , is adimensional time $\frac{t}{\tau}$
55 (t and τ are dimensional time and a timescale, correspondingly), and all other parameters $\Pi_2, \dots, \Pi_i, \dots, \Pi_m$
56 are fixed to specific values. It means that an experimental time series can be described as

$$57 \quad \Pi = \Phi\left(\frac{t}{\tau}, \Pi_2, \dots, \Pi_i, \dots, \Pi_m\right) \quad (3)$$

58 If we created a model dynamical system such that it is governed by p physical parameters b_i

$$59 \quad x = \psi(b_1, b_2, \dots, b_i, \dots, b_p) \quad (4)$$



60 and r parameters of $b_1, b_2, \dots, b_i, \dots, b_p$ are parameters with independent dimensions, then, again,
61 according to π -theorem, in the dimensionless form, the model can be described by $q = p - r$
62 adimensional similarity parameters $\pi_1, \pi_2, \dots, \pi_i, \dots, \pi_q$:

$$63 \quad \pi = \Psi(\pi_1, \pi_2, \dots, \pi_i, \dots, \pi_q) \quad (5)$$

64 For a specific time series, and for a fixed set of parameters $\pi_2, \dots, \pi_i, \dots, \pi_q$, the model (5) can be
65 presented as

$$66 \quad \pi = \Psi\left(\frac{t}{\tau}, \pi_2, \dots, \pi_i, \dots, \pi_q\right) \quad (6)$$

67 The essence of the phenomenological approach is to fit function $\Psi\left(\frac{t}{\tau}, \pi_2, \dots, \pi_i, \dots, \pi_q\right)$ to the
68 function $\Phi\left(\frac{t}{\tau}, \Pi_2, \dots, \Pi_i, \dots, \Pi_m\right)$ under the “best” set of parameters $\pi_2, \dots, \pi_i, \dots, \pi_q$, i.e. to equate model
69 time series $\Psi\left(\frac{t}{\tau}, \pi_2, \dots, \pi_i, \dots, \pi_q\right)$ and natural, empirical, time series $\Phi\left(\frac{t}{\tau}, \Pi_2, \dots, \Pi_i, \dots, \Pi_m\right)$:

$$70 \quad \Psi\left(\frac{t}{\tau}, \pi_2, \dots, \pi_i, \dots, \pi_q\right) = \Phi\left(\frac{t}{\tau}, \Pi_2, \dots, \Pi_i, \dots, \Pi_m\right) \quad (7)$$

71 It is obvious that even if the goal (7) is achieved at every $\frac{t}{\tau}$ - point, we still cannot claim the model
72 (6) to be physically similar to “nature” (3) because we cannot compare similarity parameters
73 $\pi_2, \dots, \pi_i, \dots, \pi_q$ and $\Pi_2, \dots, \Pi_i, \dots, \Pi_m$ for one simple reason: we do not know enough about
74 $\Pi_2, \dots, \Pi_i, \dots, \Pi_m$. Even if we are reasonably confident about the physics involved in the natural
75 phenomena $\Phi\left(\frac{t}{\tau}, \Pi_2, \dots, \Pi_i, \dots, \Pi_m\right)$, and each similarity parameter π_i , representing a specific aspect of
76 the phenomena, is believed to have its counterpart Π_i in the “real” dynamical system (i.e., $q = m$), we still
77 do not know the numerical values of similarity parameters $\Pi_2, \dots, \Pi_i, \dots, \Pi_m$, and therefore we cannot
78 judge if the π_i -physics in the model is as significant as the Π_i -physics of nature. Simply speaking, merely
79 *matching a proposed phenomenological model with empirical data does not make a case for physical*
80 *similarity* because it does not provide an evidence that it happens for the right reason, the reason being the
81 similarity parameters of the right value, i.e., $\pi_i = \Pi_i$.

82 For example, it is undisputable that astronomical forcing plays an important role in forming
83 Pleistocene climate variability. Variations of insolation (e.g., mid-July insolation at 65° N, e.g., Berger
84 and Loutre, 1991), of the amplitude ε and dominating periods of the precession and obliquity thus create



85 the similarity parameter $\Pi_2 = \frac{\varepsilon}{a}$ that is the ratio of the intensity of the astronomical forcing to the intensity
86 of the terrestrial ice-sheet mass influx (ε is measured in units of a , where a is snow accumulation minus
87 ablation). Practically all phenomenological ice-age models incorporate astronomical forcing and
88 terrestrial ice-sheet mass influx and therefore a corresponding similarity parameter π_2 can always be
89 constructed but we do not have the means to judge if $\pi_2 = \Pi_2$.

90 Thus, we have phenomenological models that occasionally may be successful in reproducing
91 experimental time series, but we cannot be confident that the similarity parameters employed are fully
92 legitimate, and, therefore, any further use of such models is rather speculative. On the other hand, most
93 comprehensive space-resolving models are still not practical for calculating climate on Pleistocene
94 timescales. In this situation, some insight could be found in the analysis of physical similarity of two
95 dynamical models, the phenomenological dynamical paleoclimate model of Leloup and Paillard (2022),
96 LP22 thereafter, that will provide us with the function Ψ and the dynamical paleoclimate model of
97 Verbitsky *et al* (2018), VCV18 thereafter, that, indeed, cannot fully represent nature but, certainly, is
98 more physically explicit and, therefore, will be nominated to provide us with the function Φ .

99 Obviously, since the time series produced by the LP22 model and by the VCV18 model are not
100 identical, the models are not physically similar in the full sense of the equation (7). We will demonstrate
101 though that the answer to the physical-similarity question is not that straightforward if our dependent
102 variable of interest x is not necessarily a time series but a time-independent attribute such as the period of
103 the system response to the astronomical forcing.

104

105 2. Method

106 The dimensional analysis and similarity properties of the VCV18 model have already been
107 comprehensively described (Verbitsky and Crucifix, 2020, Verbitsky, 2022). It was demonstrated that its
108 large-scale variability is mostly governed by two dimensionless parameters: by the ratio of the
109 astronomical forcing amplitude ε to the terrestrial ice-sheet mass influx, $\Pi_2 = \varepsilon/a$ and by the so-called V -
110 number, $\Pi_3 = V$ that is the ratio of amplitudes of time-dependent positive and negative feedbacks. For
111 example, the period P of the VCV18 system response to the astronomical forcing of period T can be
112 written as:

$$113 \frac{P}{T} = \Phi(\Pi_2, \Pi_3) \quad (8)$$



114 We will now focus on the dimensional analysis of the LP22 model. It is described by two differential
115 equations, first, for the growing ice volume,

$$116 \quad \frac{dv}{dt} = -\frac{I}{\tau_i} + \frac{1}{\tau_g} \quad (9)$$

117 and for the waning ice volume

$$118 \quad \frac{dv}{dt} = -\frac{I}{\tau_i} - \frac{v}{\tau_d} \quad (10)$$

119 Here, v and I are normalized ice volume and astronomical forcing, correspondingly, τ_i , τ_g , and τ_d are
120 dimensional timescales. Additionally, if $I < I_0$ the system switches from equation (10) to equation (9),
121 and if $I + v > V_0$, the system switches from equation (9) to equation (10).

122 First, we will convert equations (9) and (10) into a form more convenient for our purpose:

$$123 \quad \frac{dv}{dt} = \frac{1}{\tau_g} \left(-\frac{I\tau_g}{\tau_i} + 1 \right) \quad (11)$$

$$124 \quad \frac{dv}{dt} = \frac{1}{\tau_g} \left(-\frac{I\tau_g}{\tau_i} - \frac{v\tau_g}{\tau_d} \right) \quad (12)$$

125 If (without any loss of generality) we consider astronomical forcing I as a sinusoid of amplitude ε and of
126 period T , then dynamical properties like, for example, a period of the system response to the astronomical
127 forcing, P , will be fully described by 6 parameters:

$$128 \quad P = \psi \left(\frac{\varepsilon\tau_g}{\tau_i}, T, V_0, \tau_g, \frac{\tau_d}{\tau_g}, I_0 \right) \quad (13)$$

129 If we choose period T as a parameter with independent dimensions, then, according to π -theorem:

$$130 \quad \frac{P}{T} = \Psi \left(\frac{\varepsilon\tau_g}{\tau_i}, V_0, \frac{T}{\tau_g}, \frac{\tau_d}{\tau_g}, I_0 \right) \quad (14)$$

131 The parameter I_0 is settled as a constant, $I_0 = 0$ (LP22), and the period P is not very sensitive to the
132 choice of $\frac{\tau_d}{\tau_g}$ as long as $\tau_g > \tau_d$, and therefore without loss of much of physical content we can set $\frac{\tau_d}{\tau_g} = 0$

133 (instant disintegration of an ice sheet) such that:

$$134 \quad \frac{P}{T} = \Psi \left(\frac{\varepsilon\tau_g}{\tau_i}, V_0, \frac{T}{\tau_g} \right) \quad (15)$$



135 Finally, we notice that the period P is largely defined not by individual values of $V_0, \frac{T}{\tau_g}$ but by their ratio

136 $V_0 \frac{\tau_g}{T}$ and therefore:

$$137 \quad \frac{P}{T} = \Psi(\pi_2, \pi_3) \quad (16)$$

138 where $\pi_2 = \frac{\varepsilon \tau_g}{\tau_i}$, $\pi_3 = V_0 \frac{\tau_g}{T}$

139 Now, we are ready to discuss physical similarity. Both VCV18 and LP22 models are fairly
 140 successful in reproducing late-Pleistocene ~100-kyr-period variability, meaning that the first necessary
 141 condition of physical similarity is satisfied:

$$142 \quad \Psi(\pi_2, \pi_3) = \Phi(\Pi_2, \Pi_3) \quad (17)$$

143 The physical meaning of $\pi_2 = \frac{\varepsilon \tau_g}{\tau_i}$ is the ratio of the astronomical forcing amplitude $\frac{\varepsilon}{\tau_i}$ to the
 144 terrestrial ice-sheet mass influx $\frac{1}{\tau_g}$, i.e. it is identical to the physical meaning of Π_2 . Numerically, for the
 145 Late Pleistocene, in LP22, $\pi_2 \sim 3$, and, in VCV18, $\Pi_2 \sim 2$, this may be considered being reasonably close.
 146 Accordingly, we may conclude that models VCV18 and LP22 *are physically similar in terms of the*
 147 *similarity parameter that is the ratio of the astronomical forcing amplitude to the terrestrial ice-sheet*
 148 *mass influx.*

149 Positive and negative feedbacks are not explicitly presented in equation (9) and the similarity
 150 parameter $\pi_3 = V_0 \frac{\tau_g}{T}$ is the only implicit outcome of the interplay between positive and negative
 151 feedbacks. Yet, we may suggest with confidence that the ratio of positive-to-negative feedbacks in (9) is
 152 implicitly set to be $V = 1$, because this is the only possibility that may provide a linear growth of ice
 153 volume in (9). Thus (17) can be re-written as

$$154 \quad \Psi[\pi_2, \pi_3(V)] = \Phi(\Pi_2, \Pi_3) \quad (18)$$

156 Here $\pi_3(V)$ means that π_3 is a function of V , i.e.:

$$157 \quad \pi_3 = V_0 \frac{\tau_g}{T}, \text{ if } V = 1 \quad (19)$$

158 and equation (18) can be re-written as



159 $\Psi'(\pi_2, V) = \Phi(\Pi_2, V)$ (20)

160 In VCV18, the numerical value of $\Pi_3 = V = 0.75$ for the late Pleistocene, and, as we just discussed, in
161 LP22, $V = 1$, i.e. close to the VCV18 value. We may therefore conclude that models VCV18 and LP22
162 *are physically similar in terms of the similarity parameter that is the ratio of amplitudes of time-*
163 *dependent positive and negative feedbacks.* This result is remarkable because, in fact, the entire LP22
164 model can be described by the following powerful statement: If positive and negative feedbacks of the
165 global ice-climate system are in balance ($V = 1$), the period of the system response is approximately
166 equal to $V_0 \tau_g$.

167 Importantly, the postulated $V = 1$ of LP22 is not constrained by any specific nature of positive or
168 negative feedbacks. Likewise, the V -number in VCV18 is a conglomerate similarity parameter
169 (Verbitsky, 2022) and thus it is not connected to a specific feedback's physics. This observation makes
170 our similarity finding even more general.

171 It is also interesting that, since $\Phi(2, 0.75) \sim 2$, for $T = 41$ kyr, i.e., VCV18 makes period doubling
172 of the obliquity-period forcing, and since $\Psi(3, 1) = \Phi(2, 0.75)$, we may conclude that LP22 makes
173 obliquity-period doubling as well.

174 We have to emphasize however, that the V -number similarity between LP22 and VCV18 models
175 can be observed only for the late Pleistocene period. The early Pleistocene variability is reproduced in the
176 VCV18 model by $V \rightarrow 0$ (reduced impact of positive feedbacks) while LP22 has the assumption of $V = 1$
177 unchanged.

178 This newly discovered physical similarity of two very different models is not intuitive and adds
179 credibility to both of them. For the LP22 model, it is a demonstration that the model is physically viable
180 in terms of two critical similarity parameters, and its most questionable saw-tooth variability may be
181 supported by some physical reasoning. For VCV18, which itself does not demonstrate exceptional
182 performance in reproducing empirical time series, physical similarity with the model most celebrated for
183 matching empirical data, it is an encouraging sign also. Nevertheless, even though the analysis revealed
184 more physical similarity between the two models than one may expect, the VCV18 model is not true
185 nature, and therefore such a comparison is by no means a final verdict on the LP22 model. Yet it
186 illustrates the importance of the physical-similarity analysis.

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188



189 **3. Conclusions**

190 Are phenomenological dynamical paleoclimate models physically similar to nature? We
191 demonstrated that, though they may be remarkably accurate in reproducing empirical time series, this is
192 not sufficient to claim physical similarity with nature until comparison of similarity parameters is
193 performed. We illustrated such a process of diagnostics of physical similarity by comparing the LP22
194 phenomenological dynamical paleoclimate model with the more explicit VCV18 model and established
195 that, for the late Pleistocene, there is physical similarity in terms of two crucial similarity parameters,
196 namely, the ratio of the astronomical forcing amplitude to the terrestrial ice-sheet mass influx and the
197 ratio of amplitudes of time-dependent positive and negative feedbacks. We will not, indeed, have a
198 definite answer until we learn more about “true” similarity parameters $\Pi_2, \dots, \Pi_i, \dots, \Pi_m$.

199 But how can we learn about natural similarity parameters $\Pi_2, \dots, \Pi_i, \dots, \Pi_m$? We think that Saltzman
200 (2002) was right when he proposed that “the essential slow physics is to be sought in the low-order
201 models.” We suggest here that this “essential slow physics” may be an inventory of candidate similarity
202 parameters $\pi_2, \dots, \pi_i, \dots, \pi_q$. Like, for example, the Reynolds number compares inertial and viscous
203 forces, or the Peclet number speaks to the significance of advective heat transfer relative to heat diffusion,
204 each of these candidate similarity parameters must describe relationships that are critical for dynamical
205 paleoclimatology. At this point, we can confidently recommend into this inventory two ratios that, we
206 believe, largely define Pleistocene climate – the ratio of intensities of orbital and terrestrial forcings and
207 the ratio of intensities of system’s positive and negative feedbacks (the V-number). Since our proposed
208 similarity parameters are so general, it is difficult to imagine other similarity parameters that would not
209 fall into these two categories, but the final say on this belongs, indeed, to the super-models. From here,
210 we deviate from Saltzman’s (2002) proposal and instead of following his idea that more explicit models
211 should be tuned to satisfy a best phenomenological model, we propose to use available super-models for
212 research and evaluation of natural similarity parameters $\Pi_2, \dots, \Pi_i, \dots, \Pi_m$. This will hopefully close a gap
213 between phenomenological and physical models.

214

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