Interactive comment on “The onset of chaos in nonautonomous dissipative dynamical systems: A low-order ocean–model case study” by Stefano Pierini et al.

Stefano Pierini et al.

stefano.pierini@uniparthenope.it

Received and published: 15 July 2018

Anonymous Referee #1

We would like to thank the reviewer for his/her valuable comments that have helped us improve the manuscript. Please find below our responses to the reviewer’s comments.

REVIEWER: p1, line 18: not all autonomous dissipative system exhibit period doubling cascade, etc. Think of nondriven mechanical systems in the presence of friction: all motion stops, there is no asymptotic dynamics. The sentence should be reformulated.
RESPONSE: We thank the reviewer for this comment. We have added “in the presence of time-independent forcing”, now on p. 1, l. 20 of the revised ms. “[…] in autonomous dissipative systems - in the presence of time-independent forcing - include […].”

REVIEWER: Fig.1: what is the time t at which the attractors are plotted?

RESPONSE: The attractors are time-independent, but of course only after spinup. Following this useful comment we have now specified the time in the text (p. 5, l. 2): “…and it is plotted at t = T* = 400 yr, i.e. after spinup.”

REVIEWER: Eq.(5): This quantity is similar to the broadly used finite time Lyapunov exponent. Why do you think that sigma is better suited here?

RESPONSE: The metric sigma provides information that is different from that provided by the Lyapunov exponents. Why have we used sigma here? We had pointed out in this paper’s original version - and still do so - that “Pierini et al. (2016) found the quantity sigma to be a good indicator of the degree of sensitivity of the system’s evolution with respect to the initial state during the phase of convergence to the attractor.” In addition, following this useful comment, we have now stressed, furthermore, that (p. 6, l. 1-4): “The determination of the PBAs of the periodically forced system and the application of the new qualitative and quantitative diagnostic methods proposed in section 4 need an analysis of the behavior of trajectories that lie at t = 0 on a given subset Omega of phase space, as is the case when calculating sigma(X,Y) above. Thus, investigating the behavior of model trajectories as they emerge from Omega is the most unifying and distinctive feature of the present model study.”

REVIEWER: Fig.2: what is the value of T*? How do the plots change if T* changes? Why exactly this T* is taken?

C2
RESPONSE: The value of $T^*$ (400 yr) was, and still is, defined in section 2, between Eqs. (3) and (4). The plots of Fig. 2 are independent of $T^*$ provided the latter it is sufficiently greater than the spinup time, but as clearly shown by the graphs of Fig. 3, this is definitely achieved for our choice. Following this useful reviewer’s comment, this is now specified on p. 4, l. 11: “... as shown in Fig. 3 and 9 below, $T^*$ is much greater than the spinup time in all cases.”

Reviewer: Fig.3, panel c: this is a chaotic case, and the blue curve cannot be distinguished from the red?

RESPONSE: Yes, the red and blue curves are virtually coincident; this is what happens for trajectories leaving from the very restricted cold-color regions of Fig. 2d; on the other hand, this possibility is not surprising in view of the results of Pierini et al. (2016) (p.7, l. 3): “If sigma < 1, as is the case near P1, the two trajectories are virtually coincident (Fig. 3c).” Of course, these trajectories are nonetheless unstable once they have converged onto the PBA.

Reviewer: This contradicts Fig.4a where the green points are spread even if started from the vicinity of P_1.

RESPONSE: We thank the reviewer for this comment. For the response please see p. 8, l. 3-6: “In the chaotic case gamma=1.35, sigma<=1 for 43% of the points contained in theta1, while sigma>1 for the remaining points. The evolution of the former lead to the localized blue dots in Figure 4a while the evolution of the latter lead to the green dots scattered over the strange attractor. The green line of Figure 4b, giving Stheta1 computed with all the trajectories, shows the gradual spreading of the initial points with sigma>1.” Please see also p. 9, l. 7-10 (see a further response below).

Reviewer: p7, l4: Fig.4a is taken at t=400, but the caption says t=300. What is the
correct statement?
RESPONSE: The correct statement in both cases is \( t = 300 \text{ y} \). Thank you.

REVIEWER: Fig.4a: I do not see the red dots mentioned in line 5 of page 7.
RESPONSE: As explained in the discussion of Fig. 4a, “...The entropy of the periodic case \( \gamma = 1.1 \), characterized initially by \( \sigma < 1 \), oscillates between 0 and 1, with the final evolution limited to virtually a single cell over the limit cycle; ...”. You can recognize the small cluster at \((\Psi_1,\Psi_2)\approx(7,12)\), but we understand that this is not easy to be noticed. So, following this useful comment, now we have added a circle in the new Fig. 4a and have mentioned it in the text (p. 8, l. 2): “... with the final evolution limited to virtually a single cell over the limit cycle; the latter cell is enclosed in the red circle of Figure 4a.”

REVIEWER: Fig.4b,d: I wonder if the use of this entropy is useful if it gives different values for the same chaotic attractor even after 400 time units. We know that \( S \) should asymptotically converge to a constant (exactly since the driving is constant). What if we take more initial points in the small square used in Fig.4a?
RESPONSE: We thank the reviewer for this comment. For our response please read p. 9, l. 7-10: “Finally, it is worth stressing that, since the forcing is constant, the range of variability of the entropy in the chaotic case with \( \sigma > 1 \) must tend to zero as the number of points tends to infinity. This tendency is clearly illustrated by the green line of Figure 4d. On the other hand, the range of variability of \( S_{\theta_1} \) in the chaotic case (Figure 4b) is still quite large after 400 yr because, as pointed out above, the number of points with \( \sigma > 1 \) contained in \( \theta_1 \) is relatively small.”
REVIEWER: p.8, lines 2-5: you speak about a "sensitive dependence on initial data". I agree, there is some dependence but this is certainly different from the traditional sensitivity since the letter holds on the chaotic attractors. Yours is there for a periodic motion and disappears, as you say in line 5, on the attractor. It might be useful to use a different terminology here.

RESPONSE: We thank the reviewer for this comment. Now we have used the expression “sensitive phase dependence on initial data” when referring to the non-chaotic case with sigma>1. Besides, we have also modified the first paragraph of section 3.2 in order to better characterize this specific form of sensitivity (p. 9, l. 12-20): “We conclude the analysis of the autonomous system by discussing an apparent paradox. We have just seen that, in regions of Gamma where sigma> 1, the trajectories for gamma = 1.1 exhibit sensitive phase dependence on initial data, as shown, for instance, by Fig. 3b, by the red dots in Fig. 4c and by the red curve in Fig. 4d. Sensitive dependence on initial data is usually associated with chaotic dynamics, but in this case the dynamics is periodic. This paradox is resolved by noting that such sensitivity concerns only the phase of the periodic trajectories, as already noticed in the previous subsection and, in addition, it occurs only if the initial data lie outside the attractor, e.g, elsewhere on Gamma; on the attractor, this phase sensitivity disappears, as we will show below. On the contrary, in the chaotic case gamma = 1.35, the sensitivity to initial data for trajectories with sigma > 1 always holds, off the attractor as well as on it, in excellent agreement with the chaotic character of the dynamics in the latter case.”

In any case, in the same section we have clearly explained the fundamental difference between this sensitivity and that associated with chaotic dynamics, so we believe there can be no misunderstanding.

REVIEWER: p9, line17: what is the meaning of "spans the attractor roughly six times"? 
RESPONSE: We have now improved the explanation (p. 10, l. 2-3): “… indeed, these
points that start from $t = 300$ yr, evolve anticlockwise around the attractor, covering it roughly six times during the interval $4T\Delta = 100$ yr that separates the first snapshot from the last one.

REVIEWER: Fig.10: What is $T$ and $T^*$ here? How do the results depend on their choice?

RESPONSE: Please see p. 15, l. 16-17 for the response to this comment: “... here $T^* = 400$ yr is again the maximum integration time, and $-T \leq \tau \leq T$, while $T = 50$ yr; once more, the following results are independent of $T$, provided it is sufficiently larger than the typical time scale of the phenomenon.”

REVIEWER: Section 5.2: The system treated here is governed by a real pullback attractor. The time-dependene shown in Fig. 17. is a general one but does not possess any drift. Since systems with drift are important in undertsanding e.g. climate change, the question arises: would $\theta_\Gamma$ be a useful indicator also for this type of systems?

RESPONSE: We thank the reviewer for this comment. First of all we have more clearly stated that the new diagnostic method proposed in this study can be applied to a wide class of aperiodically forced systems. Moreover, we have now stressed that the method can find application to the specific, and very interesting case of systems possessing a drift (p. 21, l. 20-21): “For example, this diagnostic method can be applied to study the onset of chaos in systems that possess a drift mimicking global warming and other climate change scenarios (as done, for instance, in Drótos et al., 2015).”

REVIEWER: p21, line 7: "In this paper, we studied the transition from nonchaotic to chaotic PBA in a nonautonomous system whose autonomous limit is is not chaotic. I
am lost: is not the attractor of Fig. 1c chaotic?

RESPONSE: But in fact our analysis of section 4 refers to the case gamma=1.1, whose attractor is that of Fig. 1b. This comment is very useful, however, because, to avoid confusion, we followed up on it by specifying that “. . . chaos is induced by the periodic forcing” (p. 22, l. 3).

REVIEWER: p25, line 17: The sentence "the union of quasiperiodic orbits that . . . may exhibit a power spectrum that contains a multitude of local variations reminescent of those exhibited by power spectra of the chaotic orbit" is not convincing. Is the attractor quasiperiodic or chaotic? The blue dots in Fig. A2 do not clarify the situation either. Therefore, here at least I recommend to show this attractor also as the result of a long time series (after an appropriate removal of transients).

RESPONSE: This paragraph was not sufficiently clear and it has been revised. We thank the reviewer for pointing out the need for further clarification. In particular, Fig. A3 has been added; this figure illustrates better, and in a more standard way, the distinction between quasi-periodic and chaotic orbits. Please see the new discussion on p. 26, l. 11-24: “By allowing the quasiperiodic trajectories that emanate from D2 to evolve up to t = 2000, one obtains the set of blue points shown in Fig. A2. Somewhat surprisingly, this set does not form a closed curve: each blue dot in Fig. A2 corresponds actually to the state at t = 2000 in the phase plane of a quasi-periodic orbit. One such orbit is represented in blue in Fig. A1(a), after removal of the transient dynamics. Each blue dot in Fig. A2 corresponds to a different quasiperiodic orbit, whose frequency characteristics may slightly change from one blue dot to another. All these quasiperiodic orbits share, however, a spectral signature that resembles the one shown by the blue curve in Fig. A1(b). To illustrate further the distinction between quasiperiodic and chaotic orbits, the return maps for the minima of the x(t)-variable have been computed. As is well-known (e.g., Strogatz, 2018), if the return map contains just one
point, the solution is periodic in time, with all minima having the exact same value, and the period of the oscillation can be estimated by calculating the time interval between two consecutive minima. If the return map contains continuous-looking curves that fill up with more and more points as the length of the orbit increases, the solution is quasi-periodic, while the presence of folds and self-similarity in the return map provides strong evidence for chaotic solutions. For the blue and red trajectories of Fig. A1(a), we plot the corresponding return maps in Figs. A3(a) and (b), respectively. The two plots clearly discriminate between the quasiperiodic nature of the former and the chaotic one of the latter solution.”

Fig. 1. Figure 4 (modified)
Fig. 2. Figure A3 (new)