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2 The second law of thermodynamics and multifractal
3 distribution functions: Bin counting, pair correlations, and
4 the Kaplan–Yorke conjecture

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11 **Abstract**

12 We explore and compare numerical methods for the determination of multifractal dimensions for a dou-
13 bly-thermostatted harmonic oscillator. The equations of motion are continuous and time-reversible. *At*
14 equilibrium the distribution is a four-dimensional Gaussian, so that all the dimension calculations can
15 be carried out analytically. *Away from* equilibrium the distribution is a surprisingly isotropic multifractal
16 strange attractor, with the various fractal dimensionalities in the range $1 < D < 4$. The attractor is relatively
17 homogeneous, with projected two-dimensional information and correlation dimensions which are nearly
18 independent of direction. Our data indicate that the Kaplan–Yorke conjecture (for the information dimen-
19 sion) fails in the full four-dimensional phase space. We also find no plausible extension of this conjecture to
20 the *projected* fractal dimensions of the oscillator. The projected growth rate associated with the largest
21 Lyapunov exponent is *negative* in the one-dimensional coordinate space.

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24 Keywords: Fractal dimensions; Nonlinear dynamics; Kaplan–Yorke dimension

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26 1. Introduction

27 In 1983, Shuichi Nosé discovered a deterministic and time-reversible thermostatted dynamics
28 capable of imposing a time-averaged kinetic temperature $\langle T \rangle$ on selected degrees of freedom
29 [1,2]. His dynamics was both time-reversible and deterministic, but could nevertheless be used
30 to model irreversible behavior. The most useful form of his dynamics has been called “Nosé–Hoo-
31 ver dynamics”, after the studies of a thermostatted harmonic oscillator inspired by Nosé’s work
32 [3,4].

33 In 1987 related studies of the Galton Board [5] and Galton Staircase [6] models showed that
34 *nonequilibrium* stationary states generated with time-reversible motion equations generate multi-
35 fractal phase-space distributions [5–8]. A typical sample, from the Galton Board studies, is shown
36 in Fig. 1. A dynamical analysis, through the Lyapunov spectrum, shows how symmetry breaking,
37 through dynamical instability, results in trajectories obeying the Second Law of Thermodynamics.
38 The motion *forward* in time is more stable (smaller Lyapunov exponents) than is the reversed mo-
39 tion *backward* in time. The resulting fractal distributions thus provide a simple resolution of the
40 Loschmidt paradox, which contrasts the one-way Second Law of Thermodynamics with the
41 either-way nature of time-reversible microscopic dynamics [6,9].

42 Here we investigate a prototypical nonequilibrium problem which generates a multifractal
43 strange attractor in its (four-dimensional) phase space. The adjective “Multifractal” signifies that
44 the apparent dimensionality of the attractor (the number of attractor points lying within a dis-
45 tance r is proportional to r^D) varies from point to point, making it possible to define whole fam-
46 ilies of fractal dimensions. Measures proportional to different powers of the phase-space
47 probability density emphasize different regions of the attractor, giving rise to different character-
48 istic overall dimensionalities. The dimensionalities of the nonequilibrium fractal distributions are

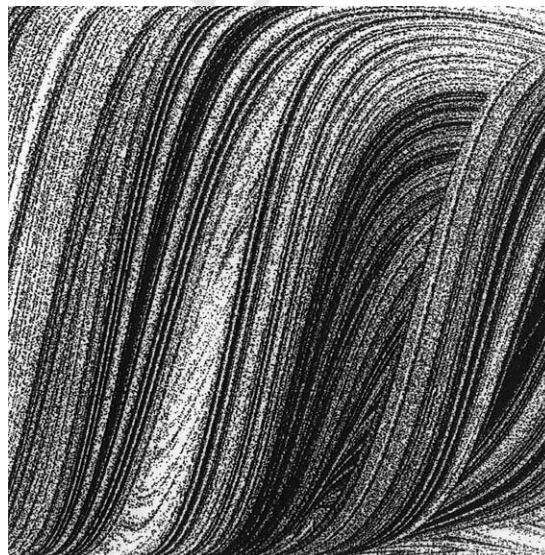


Fig. 1. Phase plane for the Galton Board problem. The 10^6 points shown here represent successive collisions of a point particle in an infinite periodic array of hard-disk scatterers. For details see Ref. [5].

49 all less than those of the corresponding equilibrium distribution, typically by an amount propor-
50 tional to the rate of entropy production [9–13].

51 Much more recently, numerical studies of the many-body thermal conductivity for the ϕ^4 po-
52 tential model—a crystal with harmonic interactions and quartic tethers of particles to sites—re-
53 vealed that the nonequilibrium dimensionality loss can be quite large. The calculated
54 dimensionality losses for the ϕ^4 model systems agreed nicely with predictions based on simple
55 ideas from chaotic dynamics and irreversible thermodynamics [14–17]. Posch and Hoover esti-
56 mated the dimensionality reduction in the subspace of the purely Hamiltonian degrees of freedom
57 in ϕ^4 systems with a few hundred degrees of freedom. This work, based on a large-system exten-
58 sion of the Kaplan–Yorke conjecture, quantified the dimensionality reduction that can occur in
59 Hamiltonian phase spaces far from equilibrium. We were motivated to test these same ideas
60 for the small system studied here.

61 The realization that multifractal distribution functions are commonplace in nonequilibrium
62 systems has led to the creation and exploration of many simple models. Among these, the dou-
63 bly-thermostatted harmonic oscillator has unique properties. Unlike maps and hard-particle mod-
64 els, the oscillator trajectory is smooth everywhere, free of any singularities. Near equilibrium the
65 distribution is ergodic and analytic, tracing out a four-dimensional Gaussian probability density
66 in the full phase space. Away from equilibrium the oscillator control equations can be designed to
67 obey either of the thermodynamic identities,

$$\langle TdS_{\text{ext}}/dt \rangle \equiv \langle -dQ/dt \rangle$$

70 or

$$\langle dS_{\text{ext}}/dt \rangle \equiv \langle -(1/T)dQ/dt \rangle,$$

73 where Q is the heat transferred to the oscillator at the temperature T and S_{ext} is the external en-
74 tropy. Both T and S_{ext} vary with time. The angular brackets indicate long-time averages. These
75 heat-transfer relations follow directly from the chosen thermostatted equations of motion, as
76 we see in detail in Section 3. The second of them leads directly to Clausius’ form of the Second
77 Law of Thermodynamics [18]:

$$\langle dS_{\text{ext}}/dt \rangle \equiv \langle -(1/T)dQ/dt \rangle > 0,$$

80 where the time average has to be taken over many successive realizations of a cyclic irreversible
81 process.

82 It is difficult to conceive of a simpler set of deterministic time-reversible flow equations which
83 still exhibits all the qualitative features of more complicated many-body stationary states. We dis-
84 cuss and apply algorithms for determining the multifractal dimensions of the nonequilibrium
85 oscillator’s strange attractor, both in the full four-dimensional phase-space and in its various sub-
86 spaces. The various multifractal dimensions characterizing nonequilibrium attractors vary
87 smoothly from the full phase-space dimension down to unity as the departure from equilibrium
88 is increased.

89 The remainder of this paper is organized as follows: in Section 2 the *equilibrium* isothermal
90 oscillator is considered. Its dynamics generates a four-dimensional Gaussian distribution in phase
91 space. This special case makes it possible to test our numerical algorithms against known analytic
92 results. A *nonequilibrium* version of this oscillator is described in Section 3, with the connection to

4 *Wm.G. Hoover et al. / Communications in Nonlinear Science and Numerical Simulation xxx (2005) xxx–xxx*

93 irreversible thermodynamics discussed in Section 4. Detailed numerical results for this nonequilibrium case are given in Section 5. The extension of the equilibrium case considered there uses a simple hyperbolic tangent relation linking the thermodynamic temperature to the oscillator coordinate. Section 6 explores the applicability of Kaplan and Yorke’s ideas to multifractal dimensions in both the full phase space and its subspaces. Section 7 is devoted to the conclusions gleaned from this work.

99 2. Gaussian phase-space distributions

100 At equilibrium a simple model for the phase-space distribution is a many-dimensional Gaussian, with the probability density for a typical phase-space variable x given by the normalized Gaussian function

$$\sqrt{2\pi}g(x) \equiv e^{-x^2/2}.$$

105 The *four-dimensional* version of this distribution, with phase-space variables (q, p, ζ, ξ) , is

$$f(q, p, \zeta, \xi) \equiv g(q)g(p)g(\zeta)g(\xi).$$

108 It can be generated by the long-time-average trajectory from the set of four ordinary differential equations describing a doubly-thermostatted oscillator. For simplicity we write the equations here in the most basic possible form, with each of the several arbitrary parameters set equal to unity:

$$\dot{q} = p; \quad \dot{p} = -q - \zeta p - \xi p^3;$$

$$\dot{\zeta} = p^2 - 1; \quad \dot{\xi} = p^4 - 3p^2.$$

115 Nonequilibrium generalizations are discussed in the following Sections. For additional examples, see Ref. [10]. Here, the oscillator coordinate is q . The momentum is p . The two control variables, or “friction coefficients”, are ζ and ξ . They control, respectively, the second and fourth moments of the momentum distribution, $\langle p^2 \rangle$ and $\langle p^4 \rangle$. These four ordinary differential equations generate the full four-dimensional Gaussian distribution in (q, p, ζ, ξ) space:

$$(2\pi)^2 f(q, p, \zeta, \xi) = \exp \left[-\frac{1}{2} (q^2 + p^2 + \zeta^2 + \xi^2) \right].$$

122 The normalization constant $(2\pi)^2$ follows from the four-dimensional definite integral:

$$\int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f(q, p, \zeta, \xi) dq dp d\zeta d\xi \equiv 1.$$

125 **Fig. 2** indicates how the smooth four-dimensional Gaussian distribution develops from the entirely one-dimensional trajectory. It is a projection of the motion into the two-dimensional (q, p) subspace. Each view of the projection consists of 200,000 separate points, with successive points separated in time by 0.1, 1.0, 10.0, and 100.0. (The correlation time for the system is of the order of the unconstrained oscillator period, 2π , as is indicated by the Lyapunov spectrum discussed in Section 5 below). In the following Section we detail the modifications necessary to treat nonequilibrium systems with a variable temperature, $T = T(q)$.

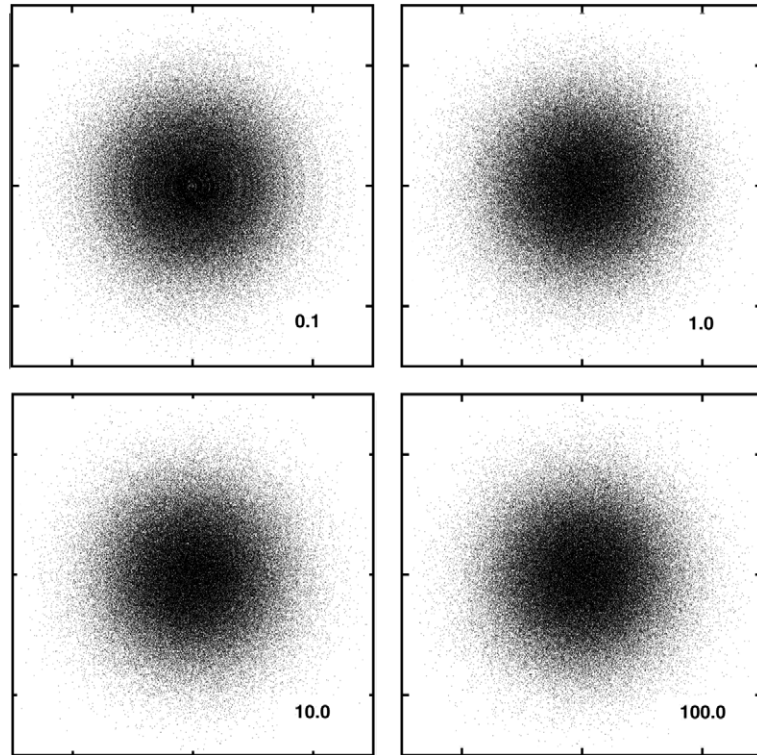


Fig. 2. Development of the $\{q,p\}$ distribution function at equilibrium, where the final distribution is a Gaussian, $\propto \exp[-\frac{1}{2}(q^2 + p^2)]$. Each of the four plots shows 200,000 points. The separations between successive points are 0.1, 1.0, 10.0, 100.0. The lack of any significant difference between the last two sets of points suggests (in agreement with the Lyapunov spectrum) that correlations are lost after a time of order 10–100. The abscissa and ordinate ranges are both ± 4 .

132 3. Nonequilibrium equations of motion

133 Now let us include explicitly all of the arbitrary parameters, still for a single harmonic oscillator
 134 with coordinate q and momentum p . Then the nonequilibrium equations of motion we study here
 135 have the form:

$$\begin{aligned} \dot{q} &= p/m; & \dot{p} &= -\kappa q - \zeta p - \xi p(p^2/mkT_0); \\ \dot{\zeta} &= [(p^2/mkT_0) - (T/T_0)]/\tau^2; \\ \dot{\xi} &= [(p^2/mkT_0)^2 - 3(p^2/mkT_0)(T/T_0)]/\tau^2. \end{aligned}$$

138 Here the coordinate-dependent temperature $T = T(q)$ is distinguished from the constant value
 139 T_0 . We choose a particular form for the coordinate dependence of T so that the reduced temper-
 140 ature T/T_0 varies from $1 - \epsilon$ (as $q/h \rightarrow -\infty$) to $1 + \epsilon$ (as $q/h \rightarrow +\infty$):

$$T/T_0 = T(q)/T_0 = 1 + \epsilon \tanh(q/h).$$

6 *Wm.G. Hoover et al. / Communications in Nonlinear Science and Numerical Simulation xxx (2005) xxx–xxx*

143 The mass and force constant of the oscillator are m and κ ; k is Boltzmann's constant, and τ is a
144 relaxation time. In the event that ϵ is nonzero, the oscillator is exposed to a temperature gradient.
145 For a long-time-averaged simulation (several oscillator vibrations) heat will flow against the direc-
146 tion of the temperature gradient. The resulting dissipation generates a steady stationary state of
147 the system: a strange attractor in the four-dimensional phase space. Several special cases of such
148 multifractal attractors are discussed in Ref. [10]. In the remainder of the paper we consider the
149 following special case:

$$m = 1; \quad \kappa = 1; \quad k = 1;$$

$$T_0 = 1; \quad \tau = 1; \quad \epsilon = 1; \quad h = 1.$$

152 4. Thermodynamic relations from phase-space time averages

153 The energy of the thermostatted oscillator changes in time, due to the nonHamiltonian force
154 $-\zeta p - \xi p^3$. As is usual in thermodynamics, we use dQ/dt to indicate the heat transferred *to* the
155 oscillator *from* the thermodynamic thermostats. We would expect to find that the time-averaged
156 value $\langle (1/T)dQ/dt \rangle$ is negative, reflecting the dissipation induced in the external reservoirs through
157 the coordinate-dependent temperature. At the same time, in any stationary state with no work
158 done the total time-averaged heat transfer must vanish:

$$\langle dQ/dt \rangle = \langle -\zeta p^2 - \xi p^4 \rangle = 0,$$

$$\langle (1/T)dQ/dt \rangle = \langle (-\zeta p^2 - \xi p^4)/T \rangle < 0.$$

161 For this oscillator there is no equivalence between the time-averaged rate of external entropy
162 production, $\langle dS_{\text{ext}}/dt \rangle$, and the loss of phase volume, $\langle d \ln \otimes / dt \rangle$:

$$\langle dS_{\text{ext}}/dt \rangle = \langle -(1/T)dQ/dt \rangle \simeq 0.093;$$

$$\langle d \ln \otimes / dt \rangle = -\langle d \ln f / dt \rangle \simeq 0.440.$$

165 It is easy to calculate the difference between the two:

$$(1/T)dQ/dt = -d \ln f / dt - (1/2T)d/dt(\zeta^2 + \xi^2).$$

168 Even so, we have chosen to study the present model in detail, rather than some of the many
169 alternatives considered in Ref. [10] and the alternative discussed just below, because the present
170 model has a particularly interesting fractal structure in its phase space and provides also a strin-
171 gent test of the Kaplan–Yorke conjecture discussed later.

172 A direct link with thermodynamics *can* be forged by solving the alternative set of motion
173 equations:

$$\dot{q} = p; \quad \dot{p} = -q - \zeta p - \xi(p^3/T);$$

$$\dot{\zeta} = [(p^2/T) - 1]/\tau^2; \quad \dot{\xi} = [(p^4/T^2) - 3(p^2/T)]/\tau^2,$$

$$T = T(q) = 1 + \epsilon \tanh(q/h).$$

176 In this case the heat transfer to the oscillator, divided by kT , has a time-averaged value exactly
177 matching the dissipation rate, as seen through the loss of phase volume:

$$\langle (1/kT)dQ/dt \rangle = \langle -\zeta(p^2/T) - \xi(p^4/T^2) \rangle = \langle -\zeta - 3\xi(p^2/T) \rangle = \langle d \ln \otimes / dt \rangle = -\langle d \ln f / dt \rangle.$$

180 Beyond checking that this model too has a stationary dissipative state for $\epsilon = \tau = h = 1$, similar
181 to the one studied in detail here, but less far from equilibrium (higher fractal dimensions) and
182 therefore geometrically less interesting, we have carried out only limited investigations of this
183 model. At equilibrium, for any of these models, the phase-space probability density for a fixed
184 constant T and τ is:

$$(T/\tau^2)(2\pi)^2 f(q, p, \zeta, \xi) = \exp \left[-\frac{1}{2T}(q^2 + p^2) \right] \exp \left[-\frac{\tau^2}{2}(\zeta^2 + \xi^2) \right],$$

187 The equilibrium thermodynamic identity that results is:

$$\langle (dQ/dt)/T \rangle = \langle (-\zeta - 3p^2\xi)/T \rangle = \langle -dS_{\text{ext}}/dt \rangle.$$

190 In the corresponding nonequilibrium case, where the temperature is a given function of the
191 coordinate rather than constant, $T = T(q)$, any of these models must necessarily satisfy the Second
192 Law of Thermodynamics, with f diverging on a multifractal strange attractor and with the time-
193 averaged entropy production $\langle dS_{\text{ext}}/dt \rangle$ strictly positive. Because a stationary state can be viewed
194 as many repetitions of a cyclic irreversible process the exact relation which results for the second
195 of the models is

$$\langle (dQ/dt)/T \rangle = \langle -dS_{\text{ext}}/dt \rangle < 0.$$

198 This time-averaged result is exactly Clausius' form of the Second Law of Thermodynamics [18].

199 5. Multifractal dimensions via bin counting, pair correlations, and the Kaplan–Yorke conjecture

200 In the simplest nonequilibrium case, with $(m, k, \kappa, T_0, \tau, \epsilon, h)$ all equal to unity, and

$$0 < T(q) = 1 + \tanh(q) < 2,$$

203 the stationary distribution still occupies the same four-dimensional phase space as at equilibrium,
204 but the information dimension D_I , along with the Kaplan–Yorke dimension D_{KY} , (defined by a
205 vanishing Lyapunov sum detailed later in this section) drops below the equilibrium value,

$$D_{KY} \simeq D_I = D_1 < D_{\text{eq}} = 4.$$

208 Nonequilibrium distributions for this example, projected into the six two-dimensional planes,
209 (q, p) , (q, ζ) , (q, ξ) , (p, ζ) , (p, ξ) , and (ζ, ξ) , make up Fig. 3.

210 Fig. 4 shows how the fractal distribution develops in the (ζ, ξ) plane, as the sampling time be-
211 tween successive points is increased. The figure shows 200,000 points, projected into the (ζ, ξ)
212 plane, with sampling intervals of 0.001, 0.01, 0.1, 1.0, and 10.0. The appearance of these fractal
213 distributions is qualitatively different to that of the smooth Gaussian distribution which is the
214 equilibrium solution. A variety of fractal dimensions have been defined in order to characterize
215 such nonequilibrium systems.

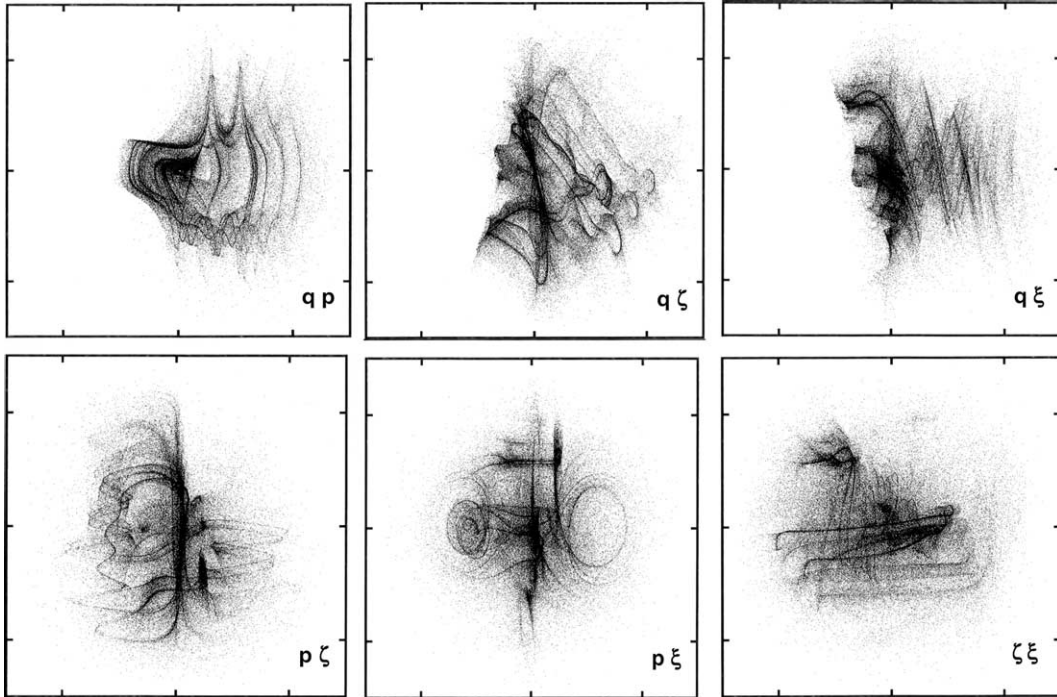


Fig. 3. Projections (200,000 points each) of the nonequilibrium $qp\zeta\xi$ oscillator dynamics into six phase-space subspaces: $\{qp, q\zeta, q\xi, p\zeta, p\xi, \zeta\xi\}$. The time separation between successive points is 10.0. The abscissa and ordinate ranges are both ± 6 for these projections.

216 The multifractal dimensions $D_0, D_1, D_2, D_3, \dots$ can all be computed from the moments (or mea-
217 sures) associated with phase-space boxes or “bins”. For the j th bin, the various measures are $\{\mu_p\}$:

$$\mu_p(j) \equiv N_j^p / \sum_k N_k^p,$$

220 where the $\{N_k\}$ are the number of observed points (proportional to the probability) in the k th bin.
221 If the bins have a characteristic size Δ then the multifractal dimensions correspond to the limiting
222 slopes of the plots of $\langle \ln \mu_p \rangle$ versus $\ln \Delta$. For ergodic systems a sufficiently long trajectory eventu-
223 ally reaches *all* bins. In such a case the measure μ_0 is uniform. For other values of p the measure is
224 concentrated in a characteristic part of the attractor. Fig. 5 shows the variation of the measures’
225 above-average-probability “cores” for the nonequilibrium oscillator. Because the measures μ_0 and
226 μ_3 are relatively slow to converge, in our numerical work we concentrate on the information and
227 correlation dimensions derived from μ_1 and μ_2 .

228 Chhabra and Jensen [19] developed an equivalent, but more direct approach to the determina-
229 tion of the multifractal dimensions. They showed that the various multifractal dimensions $\{D_q\}$
230 were given by the simple set of small-bin-size limits ($\Delta \rightarrow 0$):

$$D_q = \left(\sum \mu_q \ln \mu_q \right) / \ln \Delta.$$

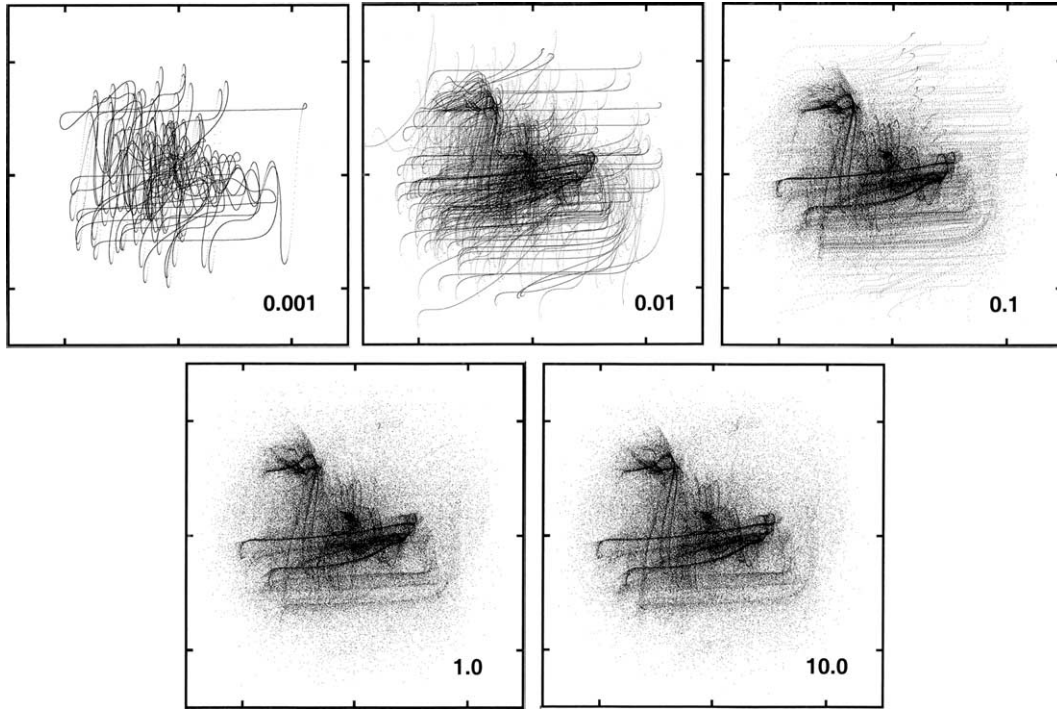


Fig. 4. Development of the nonequilibrium $\{\zeta, \xi\}$ distribution function. Each of the five plots shows 200,000 points. The separations between successive points are 0.001, 0.01, 0.1, 1.0, and 10.0. The lack of any significant difference between the last two sets of points suggests (in agreement with the Lyapunov spectrum) that correlations are lost after a time of order 10–100. The abscissa and ordinate ranges are both ± 6 for these plots.

233 It is also possible to define generalized dimensions by using two *different* measures, q_1 and q_2 :

$$D_{(q_1, q_2)} = \left(\sum \mu_{q_1} \ln \mu_{q_2} \right) / \ln \Delta.$$

236 The usual $f(\alpha)$ relation linking fractal dimension f to singularity strength α uses $q_1 = q_2$ for
237 $f(q) \equiv D_{(q, q)}$ and $q_1 = 1$ for $\alpha(q) \equiv D_{(1, q)}$ [19].

238 With presentday computers it is inconvenient to consider a four-dimensional grid with a sub-
239 stantially higher resolution than

$$128 \times 128 \times 128 \times 128 = 2^{28} = 268,435,456$$

242 phase-space bins. Both storage capacity, as well as the need to generate an average of *several*
243 points per bin (with successive points separated by 10^3 or 10^4 timesteps to avoid serial correlation)
244 combine to make this four-dimensional problem a severe computational challenge. We carried out
245 this stage of refinement by dividing up the grid data among storage files. The three-dimensional
246 subspaces are simpler to treat. A resolution of 512 bins in each of three directions requires arrays
247 half the size (2^{27} bins) of the four-dimensional ones considered here.

248 Because (according to the Kaplan–Yorke conjecture discussed below) the nonequilibrium infor-
249 mation dimension is less than three for the special case chosen here: $D_1 \simeq D_{KY} = 2.80 < 3$, we

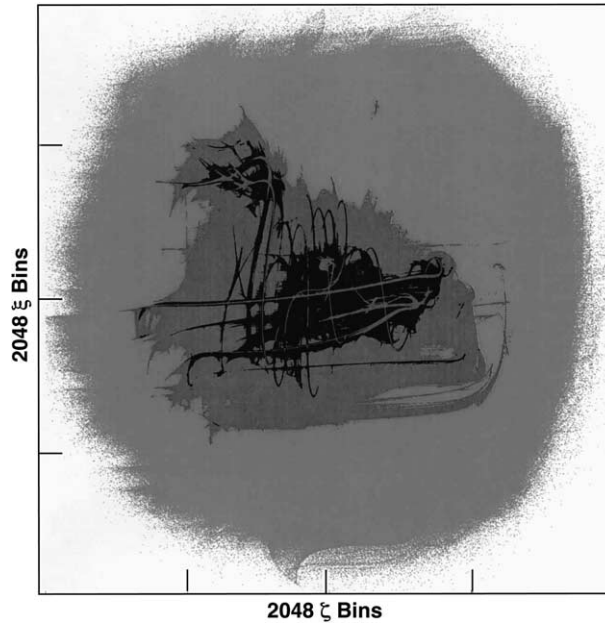


Fig. 5. Successively smaller attractor cores are shown here, projected into $\{\zeta, \xi\}$ space. They correspond, in order of decreasing size, to the “bin-counting”, “information”, “correlation”, and “three-point” dimensions. The “cores” correspond to those bins (on a 2048×2048 grid) with above average probability, where the bin probabilities are proportional to $N_k^{0.1}$, N_k^1 , N_k^2 , and N_k^3 for the three cases. One billion data points, separated by 100 timestep intervals, were used in constructing this figure.

250 expect to glean significant information by comparing nonequilibrium and equilibrium studies car-
251 ried out in the various subspaces:

$$\begin{aligned} &\{(q, p, \zeta), (q, p, \xi), (q, \zeta, \xi), (p, \zeta, \xi)\}, \\ &\{(q, p), (q, \zeta), (q, \xi), (p, \zeta), (p, \xi), (\zeta, \xi)\}, \\ &\{(q), (p), (\zeta), (\xi)\}. \end{aligned}$$

254 In the full four-dimensional phase space, the information dimension $D_I = D_1$ can be estimated
255 independently of bin counting. Kaplan and Yorke conjectured that $D_{KY} \simeq D_I$ can be estimated
256 from the Lyapunov spectrum. D_{KY} corresponds to the number of Lyapunov exponents (starting
257 with the largest) for which the sum $\sum \lambda_i$ vanishes. Typically the sum has to be linearly
258 interpolated.

259 In the equilibrium case ϵ vanishes and the temperature is constant. If we also choose the ther-
260 mostat relaxation time τ to be unity, the Lyapunov spectrum (based on 10^9 timesteps of length
261 0.001 each) is:

$$\{\langle \lambda \rangle\}_{\text{eq}} = \{+0.066, +0.000, -0.000, -0.066\}.$$

264 The spectrum shows the time-reversal symmetry associated with equilibrium. Because the sum
265 of the (time-averaged) exponents vanishes, the various partial sums (corresponding to subspace
266 growth rates) are never negative:

$$\left\{ \sum \langle \lambda_{\text{eq}} \rangle \right\} = \{+0.066, +0.066, +0.066, +0.000\}.$$

269 In the simplest nonequilibrium (far-from-equilibrium) case we choose to study here we set
270 $T = 1 + \tanh(q)$. This combination generates a multifractal with an information dimension be-
271 tween 2 and 3 in the four-dimensional phase space. The Lyapunov exponents are:

$$\{\langle \lambda \rangle\}_{\text{neq}} = \{0.072_6, 0.0000, -0.091_2, -0.411_0\}.$$

274 It is important to emphasize that though the four ordinary differential equations generating the
275 flow, as well as the sixteen additional equations describing its sensitivity to perturbations, are all
276 perfectly symmetric in the time, the time-symmetry of the solution is broken by instability, with
277 the Lyapunov exponents no longer occurring in symmetric pairs.

278 This symmetry breaking has been analyzed in considerable detail for similar systems [20]. It re-
279 flects the fact that a flow proceeding forward in time is less unstable (negative Lyapunov sum)
280 than is the time-reversed flow (positive Lyapunov sum) which would violate the Second Law of
281 Thermodynamics. The nonequilibrium Lyapunov sum of all four Lyapunov exponents is neces-
282 sarily negative, for stability. In the particular case considered here, the one-, two-, three- and
283 four-exponent sums are

$$\left\{ \sum \langle \lambda_{\text{neq}} \rangle \right\} = \{0.073, 0.073, -0.019, -0.430\}.$$

286 Linear interpolation, between the two-exponent sum, 0.072_6 , and the three-exponent one,
287 -0.018_6 , gives the Kaplan–Yorke estimate for the information dimension,
288 $D_{KY} = 2 + \frac{73}{90} \simeq 2.80$. This is a dimensionality reduction of 1.20 below the equilibrium dimension-
289 ality of 4.00. In the large-system work described in Ref. [17] dimensionality reductions (from the
290 Lyapunov spectrum, through D_{KY}) as large as 34 were observed (in a 578-dimensional phase
291 space).

292 The information dimension, $D_1 = D_I \simeq D_{KY}$, can also be evaluated, with an uncertainty of or-
293 der one percent, by simple bin-counting. See Fig. 6. Using 10^9 points separated in time by $1000dt$
294 gives the considerably lower estimate $D_I = 2.56$. The entropy-binsize plot, spanning the range
295 from 8^4 bins to 128^4 bins, gives an excellent straight line, $S_1 \propto \ln(\Delta)$. The data show that the Kap-
296 lan–Yorke conjecture is simply wrong for this four-dimensional attractor. Evidently the rapid
297 rotation rates of the Lyapunov vectors are responsible for the 10% discrepancy, $D_{KY} \simeq 1.1D_I$.

298 The correlation dimension D_2 can also be estimated independently of bin counting, but without
299 such high accuracy. A logarithmic plot of the number of *pairs* of points lying within a distance r of
300 one another increases as $D_2 \ln r$ provided that r is not too large and that the sampling time between
301 successive points is enough for correlations to decay (for a short sampling time the dimensionality
302 of the distribution would be one-dimensional, corresponding to a trajectory). The equilibrium
303 case can be used to test these ideas, for all the dimensions are precisely equal to four. In Fig. 6
304 we show the variations of all the one- and two-point entropies,

$$S_p \equiv \langle -\ln(\mu_p) \rangle \equiv \left\langle \sum -\mu_p \ln \mu_p \right\rangle.$$

307 The calculations shown are all based on 10^9 points, separated by 1000 timesteps of $dt = 0.001$
308 each. The dimensionalities from these data all lie within half a percent of the correct values
309 (1, 2, 3, 4).

12 *Wm.G. Hoover et al. / Communications in Nonlinear Science and Numerical Simulation xxx (2005) xxx-xxx*

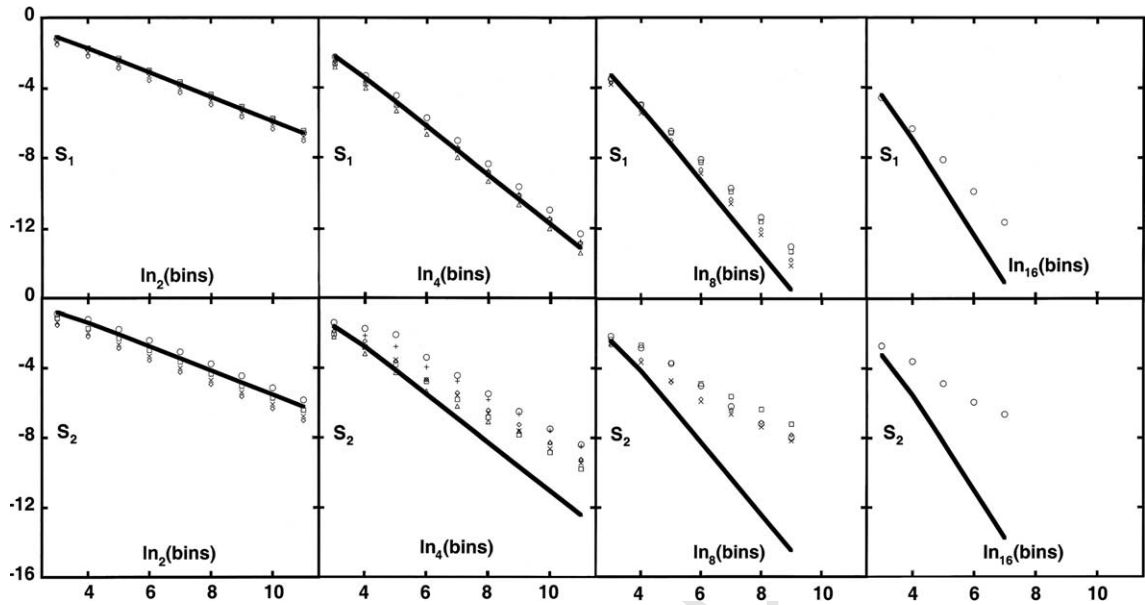


Fig. 6. Dependence of the one-, two-, three- and four-dimensional entropies on the number of bins. The equilibrium (lines) and nonequilibrium (discrete plotting symbols) entropies S_1 and S_2 are compared by using measures proportional to N_k^1 and N_k^2 . One billion points, separated by 1000dt were used to generate these data. The minimum and maximum limits for each of the four variables were ± 6 . In one, two, three, and four dimensions we used up to 2^{11} , 2^{22} , 2^{27} , and 2^{28} bins, respectively. The logarithms of the numbers of bins are logarithms with bases (2, 4, 8, 16) in (1, 2, 3, 4) dimensions.

310 The dimensionalities in the nonequilibrium case (See again Fig. 6) reveal some interesting dif-
 311 ferences. We noticed that the entropy corresponding to the *three*-point measure μ_3 varies in a *non*-
 312 *monotonic* way with the bin size Δ . A simple four-bin example for such a variation can be based on
 313 the following bin occupancy numbers:

$$\{1, 1, 2, 0\} \rightarrow \{\mu_3\} = \{0.1, 0.1, 0.8, 0.0\}.$$

316 The entropy S associated with this measure is

$$-\sum \mu_3 \ln \mu_3 = 0.2303 + 0.2303 + 0.1785 + 0.000 = 0.639.$$

319 Combining the data into pairs (corresponding to coarsening the grid) leads to a (counterintu-
 320 itive) *increase* in the entropy:

$$\{2, 2\} \rightarrow \{\mu_3\} = \{0.5, 0.5\} \rightarrow S = 0.693.$$

323 Figs. 7 and 8 summarize the correlation-dimension data for both the equilibrium and nonequi-
 324 librium data sets in all 15 of the various subspaces. The equilibrium data (Fig. 7) show that the
 325 number of pairs of points in the full four-dimensional phase space and with
 326 $\sqrt{q^2 + p^2 + \zeta^2 + \xi^2} < R_{qp\zeta\xi}$ varies as R^4 . The four sets of three-dimensional data, corresponding to

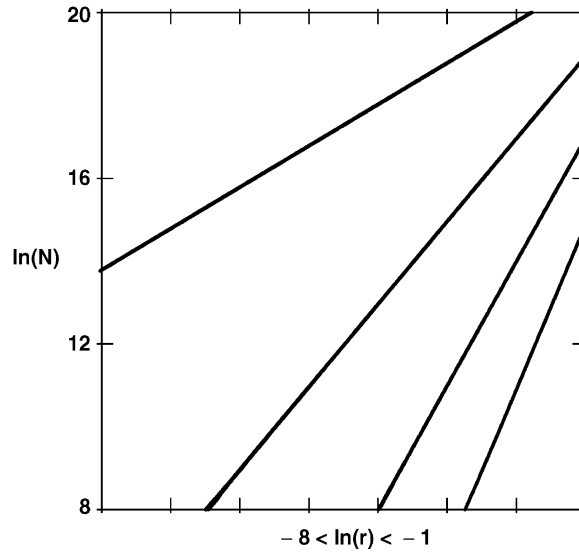


Fig. 7. Dependence of the fifteen equilibrium pair correlations on the bin size. 100,000 data points, separated by intervals of $1000dt$, generated the 4,999,950,000 pairs of points contributing to this plot. The slopes of the straight-line portions of the fifteen curves are accurately 1.00, 2.00, 3.00, and 4.00, corresponding to the dimensionalities of the corresponding Gaussian functions. The four one-dimensional sets of data are indistinguishable within the width of the plotting line, as are also the six two-dimensional data sets, and the four three-dimensional data sets.

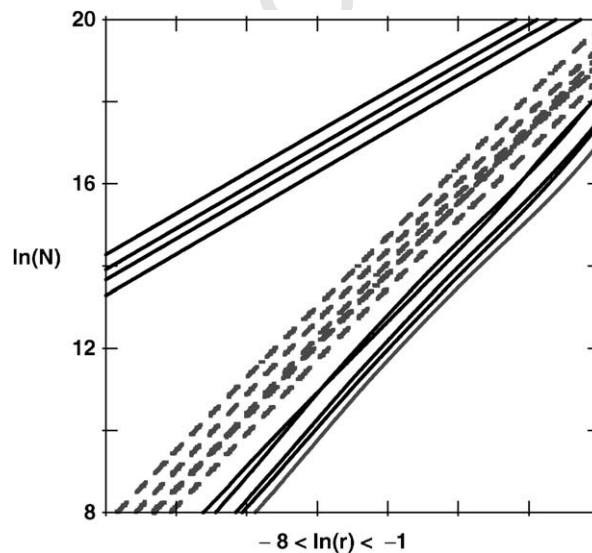


Fig. 8. Dependence of the fifteen nonequilibrium pair correlations on the bin size. 100,000 data points, separated by intervals of $1000dt$, generated the 4,999,950,000 pairs of points contributing to this plot. The slopes of these plots correspond (beginning with the one-dimensional data at the top of the figure) to correlation dimensions of 1.00, 1.7₂, 1.8₅, and $\simeq 2$ for the 1, 2, 3, and 4-dimensional subspaces, respectively, with no significant difference between the various subspaces that have the same dimensionality.

14 *Wm.G. Hoover et al. / Communications in Nonlinear Science and Numerical Simulation xxx (2005) xxx-xxx*

$$\begin{aligned} \sqrt{q^2 + p^2 + \zeta^2} &< R_{qp\zeta}; & \sqrt{q^2 + p^2 + \xi^2} &< R_{qp\xi}; \\ \sqrt{q^2 + \zeta^2 + \xi^2} &< R_{q\zeta\xi}; & \sqrt{p^2 + \zeta^2 + \xi^2} &< R_{p\zeta\xi}, \end{aligned}$$

329 the six sets of two-dimensional data, and four sets of one-dimensional data are all assembled in
330 Table 1.

331 The nonequilibrium correlation dimensions are all less than 2. But the full phase-space corre-
332 lation dimension from pairs of points, 1.8 ± 0.05 , does not agree very well with the D_2 estimates
333 from bin-counting. These bin counting results for D_2 are relatively unreliable. Unlike the informa-
334 tion dimension, the correlation dimension is sensitive to the number of bins used in the analysis.
335 Strong dimensionality reduction persists in all the subspaces through the six two-dimensional
336 examples. Fig. 9 shows histograms for the one-dimensional spaces, $\ln N_k(q)$ and $\ln N_k(p)$. Though
337 the appearance of these histograms certainly suggests the possibility of a fractal dimension less
338 than unity, we found no significant deviation from $D = 1.00$ for either of them.

Table 1
Equilibrium and nonequilibrium correlation dimensions from cumulative number of pairs

| Space | Equilibrium D_2 | Nonequilibrium D_2 |
|--|--------------------------------------|--------------------------------------|
| (q, p, ζ, ξ) | (4.01) | (1.81) |
| $(q, p, \zeta), (q, p, \xi), (q, \zeta, \xi), (p, \zeta, \xi)$ | (2.98, 2.99, 2.99, 3.00) | (1.93, 1.89, 1.86, 1.90) |
| $(q, p), (q, \zeta), (q, \xi), (p, \zeta), (p, \xi), (\zeta, \xi)$ | (1.99, 1.99, 1.99, 1.99, 1.99, 1.99) | (1.73, 1.73, 1.75, 1.75, 1.69, 1.77) |
| $(q), (p), (\zeta), (\xi)$ | (1.00, 1.00, 1.00, 1.00) | (0.98, 0.92, 0.98, 0.98) |

The first column indicates the space in which the distances between all pairs were determined. The second and third columns are the equilibrium and nonequilibrium pair dimensions D_2 . The data are based on 50,000 points with a sampling interval of 10,000 timesteps between successive points.

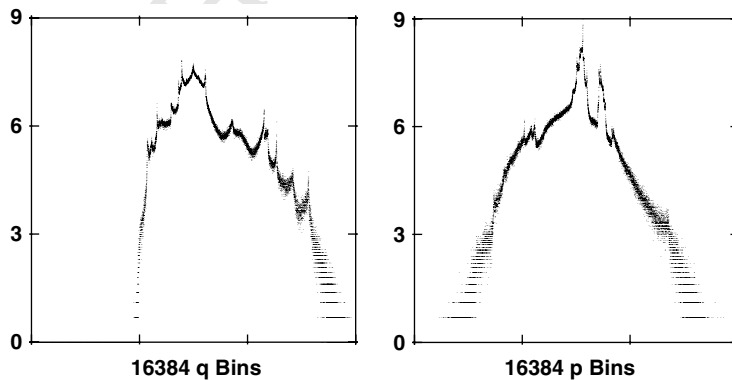


Fig. 9. Logarithms of the probability densities for the nonequilibrium coordinate q and the momentum p . 10^9 points, separated by 1000 timesteps, were used. Bin counting, in the two one-dimensional spaces, suggests information and correlation dimensions in these subspaces quite close to unity. See also the corresponding curve in Fig. 8. The 16,384 bins illustrated here span the range ± 6 for both q and p .

Table 2

The equilibrium Lyapunov exponents $\langle \lambda \rangle$ and their time-averaged projections $\langle \langle \delta_q^2, \delta_p^2, \delta_z^2, \delta_\xi^2 \rangle \rangle$ (top) and the projections weighted with the instantaneous values of the corresponding Lyapunov exponents (bottom)

| $\langle \lambda \rangle$ | $\langle \delta_q^2 \rangle$ | $\langle \delta_p^2 \rangle$ | $\langle \delta_z^2 \rangle$ | $\langle \delta_\xi^2 \rangle$ |
|---------------------------|--------------------------------------|--------------------------------------|--------------------------------------|--|
| +0.068 | 0.141 | 0.190 | 0.298 | 0.371 |
| +0.002 | 0.306 | 0.210 | 0.228 | 0.256 |
| -0.002 | 0.177 | 0.310 | 0.266 | 0.248 |
| -0.068 | 0.377 | 0.291 | 0.208 | 0.124 |
| $\langle \lambda \rangle$ | $\langle \lambda \delta_q^2 \rangle$ | $\langle \lambda \delta_p^2 \rangle$ | $\langle \lambda \delta_z^2 \rangle$ | $\langle \lambda \delta_\xi^2 \rangle$ |
| +0.068 | -0.002 | +0.014 | +0.013 | +0.043 |
| +0.002 | +0.004 | -0.011 | +0.008 | +0.002 |
| -0.002 | -0.002 | +0.002 | -0.004 | +0.001 |
| -0.068 | -0.040 | -0.013 | -0.011 | -0.004 |

Note that the row sums are unity at the top, and $\langle \lambda \rangle$ at the bottom. These data apply to the *equilibrium* oscillator, with $\epsilon = 0$ and correspond to 10^9 timesteps of 0.001 each. These data are all time averages.

339 6. Kaplan–Yorke Conjecture for Subspaces

340 Kaplan and Yorke’s conjecture has a strong intuitive basis. It is certainly “obvious” that the
 341 dimension of a strange attractor is the same as the dimensionality of an object which neither
 342 grows nor shrinks over time. We have seen that fluctuations in the Lyapunov vectors’ directions
 343 can lead to ten percent errors in the estimate. Nevertheless, it is tempting (if not irresistible) to
 344 apply the Kaplan–Yorke idea in subspaces of the full phase space, in order to estimate the pro-
 345 jected information dimensions there. We set about to do this.

346 See Tables 2 and 3 for the Lyapunov vector projections. In work with the many-body ϕ^4 model
 347 [16] we used the exact relation that the instantaneous subspace growth rate corresponding to the
 348 set of Lyapunov vectors $\{\lambda_i\}$ in the full phase space, is given by the weighted sum, $\sum \lambda_i \cos^2(\theta_i)$,
 349 where $\cos(\theta_i)$ is the projection of the phase-space vector δ_i into the subspace. We estimated the

Table 3

The *nonequilibrium* Lyapunov exponents $\langle \lambda \rangle$ and their time-averaged projections $\langle \langle \delta_q^2, \delta_p^2, \delta_z^2, \delta_\xi^2 \rangle \rangle$ (top) and the projections weighted with the instantaneous values of the corresponding Lyapunov exponents (bottom)

| $\langle \lambda \rangle$ | $\langle \delta_q^2 \rangle$ | $\langle \delta_p^2 \rangle$ | $\langle \delta_z^2 \rangle$ | $\langle \delta_\xi^2 \rangle$ |
|---------------------------|--------------------------------------|--------------------------------------|--------------------------------------|--|
| +0.072 | 0.098 | 0.112 | 0.257 | 0.533 |
| +0.000 | 0.227 | 0.181 | 0.308 | 0.284 |
| -0.091 | 0.293 | 0.363 | 0.237 | 0.107 |
| -0.410 | 0.382 | 0.343 | 0.198 | 0.077 |
| $\langle \lambda \rangle$ | $\langle \lambda \delta_q^2 \rangle$ | $\langle \lambda \delta_p^2 \rangle$ | $\langle \lambda \delta_z^2 \rangle$ | $\langle \lambda \delta_\xi^2 \rangle$ |
| +0.072 | -0.013 | -0.001 | +0.028 | +0.058 |
| +0.000 | +0.005 | -0.007 | +0.018 | -0.015 |
| -0.091 | -0.010 | -0.047 | -0.023 | -0.012 |
| -0.410 | -0.168 | -0.137 | -0.088 | -0.018 |

Note that the row sums are unity at the top, and $\langle \lambda \rangle$ at the bottom. All these data apply to the *nonequilibrium* oscillator, with $\epsilon = \tau = h = 1$ and correspond to 10^9 timesteps of 0.001 each. These data are all time averages.

350 information dimension of the subspace attractor from the sum $\sum \cos^2(\theta_i)$ of weights required for
351 the projected growth rate to vanish. This idea is correct in the event that the orientations of the
352 vectors are random, with $\cos^2(\theta_i)$ equal to the inverse number of vectors. In the many-body sys-
353 tems studied in Ref. [16] it was observed that the projections became increasingly uniform as the
354 system size increased. A four-dimensional phase space is a demanding test of this idea. The failure
355 here of the Kaplan–Yorke conjecture in the full phase space was unexpected.

356 The failure of Kaplan–Yorke in four dimensions led us to try to apply an idea like theirs in low-
357 er-dimensional subspaces of the full space, where convergence is enhanced. Unfortunately, the
358 data in Table 2 indicate that this approach fails completely for the thermostatted oscillator.
359 The projections of the vectors vary considerably about the random value, 0.25, with minimum/
360 maximum values of 0.12/0.38. The projected growth rates contain a surprise (which we found with
361 other oscillator models as well as with some few-body subspace projections of the many-body ϕ^4
362 dynamics). *The largest most positive Lyapunov exponent can have a negative time-averaged projec-*
363 *tion in some subspace directions.* Consider, for example, our oscillator problem projected into the
364 coordinate q subspace. Bin counting results show that the information and correlation dimensions
365 in q space are not significantly different to 1.00. But the instantaneous value of λ_1 , where the time
366 average $\langle \lambda_1 \rangle$, includes, at each instant, multiplication by its corresponding unit vector δ_1 has, on
367 the average, a *negative* projection in q space. The data in Tables 2 and 3 show that there is no
368 consistent way to obtain accurate information dimensions in the various subspaces. We conclude
369 that at best the Kaplan–Yorke procedure can work well in subspaces only in high-dimensional
370 systems. Provided that a long trajectory could be replaced by several thousand shorter ones
371 (and this could be checked numerically) presentday computers *might* be able to characterize
372 attractors in a six-dimensional space (with 10^{12} bins). There is no foreseeable chance that these
373 ideas can be checked in many-dimensional phase spaces for which bin counting is, and always will
374 be, impossibly difficult.

375 The present work shows that the Kaplan–Yorke conjecture is flawed in the full phase space.
376 The information dimension of the full phase-space attractor, as estimated by bin counting, is
377 2.56. The Kaplan–Yorke prediction is considerably, and significantly, higher, 2.80. Results for
378 the correlation dimension are inconclusive. The bin-counting value in the full phase space is
379 1.55, but with an uncertainty of ± 0.2 . The dimension estimated from pair enumerations is 1.81.

380 7. Conclusion

381 The deterministic, continuous, dissipative, doubly-thermostatted oscillator problem is a useful
382 prototype for understanding multifractal distribution functions far from equilibrium. It lies near
383 the borderline for presentday computational feasibility. A very similar oscillator model leads ex-
384 actly to Clausius' version of the Second Law of Thermodynamics,

$$\langle -dS_{\text{ext}}/dt \rangle = \langle (1/T)dQ/dt \rangle < 0.$$

387 This inequality is an automatic consequence of Nosé–Hoover mechanics, where the specified
388 reservoir temperatures are constants of the motion. In the case that temperature varies Clausius'
389 inequality is satisfied with the definition:

Wm.G. Hoover et al. / *Communications in Nonlinear Science and Numerical Simulation xxx (2005) xxx–xxx* 17

$$\zeta_{\text{NH}} \equiv [(p^2/mkT) - 1]/\tau^2,$$

392 but is *not* automatically satisfied with the alternative:

$$\zeta_{\text{NH}} \equiv [(p^2 - mkT)/mkT_0]/\tau^2.$$

395 The equilibrium case, with its four-dimensional Gaussian distribution, can be used to evaluate
396 the accuracy of algorithms. Our results indicate that the correlation dimension, which is consid-
397 erably simpler to evaluate than the bin-counting dimensions, is a good character of fractals,
398 with the numerical pair-counting and bin-counting versions of D_2 not inconsistent with one an-
399 other. For this model all six of the two-dimensional projections of the attractor had similar Kap-
400 lan–Yorke dimensions and similar correlation dimensions. This finding suggests a rapid rotation
401 in phase space, tending to make the attractor relatively isotropic and homogeneous, even for a
402 few-dimensional phase space. The model has also a particularly interesting feature, a contracting
403 time-averaged projection of the four-dimensional dynamics into the one-dimensional coordinate
404 space. This indicates that there is no simple analog of the Kaplan–Yorke conjecture for subspaces,
405 at least for the present model system. The finding that the Kaplan–Yorke conjecture is inaccurate
406 in the full phase space was a major surprise to us.

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18 *Wm.G. Hoover et al. / Communications in Nonlinear Science and Numerical Simulation xxx (2005) xxx-xxx*

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