Optimal Design of Composite Granular Protectors

Fernando Fraternali [a]; Mason A. Porter [b]; Chiara Daraio [c]

[a] Department of Civil Engineering, University of Salerno, Fisciano, Italy [b] Oxford Centre for Industrial and Applied Mathematics, Mathematical Institute, University of Oxford, United Kingdom [c] Graduate Aerospace Laboratories (GALCIT) and Department of Applied Physics, California Institute of Technology, Pasadena, California, USA

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Fernando Fraternali,[1] Mason A. Porter,[2] and Chiara Daraio[3]

[1]*Department of Civil Engineering, University of Salerno, Fisciano, Italy*
[2]*Oxford Centre for Industrial and Applied Mathematics, Mathematical Institute,*
University of Oxford, United Kingdom
[3]*Graduate Aerospace Laboratories (GALCIT) and Department of Applied Physics,*
California Institute of Technology, Pasadena, California, USA

We employ an evolutionary algorithm to investigate the optimal design of composite protectors using one-dimensional granular chains composed of beads of various sizes, masses, and stiffnesses. We define a fitness function using the maximum force transmitted from the protector to a "wall" that represents the body to be protected and accordingly optimize the topology (arrangement), size, and material of the chain. We obtain optimally randomized granular protectors characterized by high-energy equipartition and the transformation of incident waves into interacting solitary pulses. We consistently observe that the pulses traveling to the wall combine to form an extended (long-wavelength), small-amplitude pulse.

1. INTRODUCTION

One-dimensional (1D) lattices (chains) of particles interacting according to nonlinear potentials have been receiving increasing attention in the scientific community because of their special wave dynamics, which allows energy transport through solitary waves [1–7]. In the case of granular systems, particle interactions are strongly nonlinear because of nonlinear contact interactions between particles and tensionless behavior [6–10]. As a result, granular lattices can support traveling compacton-like solitary waves [6]. The evolution of nonlinear particle systems toward energy equipartition (or thermalization), predicted by statistical mechanics [11], is also particularly interesting. Stable or transient energy transport through coherent modes (solitary waves, breathers, etc.) can develop [2, 12–18], and eventual thermalization might not occur at all, as investigated in great detail for the Fermi-Pasta-Ulam problem and related nonlinear lattice systems [1].

By designing protectors or containers optimally, the strongly nonlinear dynamics of granular systems can be exploited to produce fast decomposition of an external impulse into trains of solitary waves, energy trapping, shock disintegration, and more [10, 19–25]. Furthermore, it has been emphasized that using a suitable randomization of the granular system—involving, for example, the variation of particle sizes, masses, and materials—one might induce nonuniformity in the steady states in the velocity profiles; the appearance of negative velocities; marked thermalization; wave-amplitude decay; and anomalous features of wave propagation through interfaces between particles differing in masses, sizes, and/or mechanical properties [6, 20, 23, 24, 26].

When dealing with the optimal design of granular protectors, one can optimize features such as particle distribution, connectivity, size, and material through either discrete or continuous approaches. (These ideas are known, respectively, as shape, topology, size, and material optimization.) Discrete approaches introduce suitable background structures in which the material densities of predefined connections are subject to optimization [27–31]. Continuous models instead use homogenization theory, as they examine design domains with perforated composite microstructures [27, 32–36]. There are numerous possible choices for optimization methodology. Well-established gradient-based optimization techniques include mathematical programming, optimality criteria (that is, finding suitable mathematical conditions that define an optimally designed structure) [37, 38], and sequential approximate optimization [39–41]. Available methods that are not based on gradients include simulated annealing [42, 43], biological growth [44, 45], and genetic and evolutionary algorithms [46–50].

Evolutionary Algorithms (EAs) provide a family of optimization methods inspired by Darwin's theory of evolution. They search for the best "phenotype" in a given population of candidate solutions by applying selection mechanisms and genetic operators similar to the intermingling of chromosomes in cell reproduction and replication. First, one evaluates each

element (individual) of the population in terms of a quantitative fitness, which represents the feature that discriminates between phenotypes. One then mates the individuals using a recombination operator. Finally, one mutates a given percentage of the individuals, thereby creating a new population. One then repeats these steps cyclically until some termination criterion is reached. The individual with the best fitness in the final population provides a guess of the global optimum of the fitness function. EAs are a natural fit for optimization problems in granular systems, as in such problems one can easily identify the system particles (i.e., the beads) with cells and their geometrical and mechanical properties (including radii, mass densities, elastic moduli, material types, and so on) with the corresponding genes. Furthermore, EAs require little knowledge of the search environment, can escape from local optima (in order to achieve a better global optimum), and are well-suited to problems with large and complex solution spaces [49] (such as those arising from the optimization of strongly nonlinear dynamical systems).

The present work exploits EAs for the optimal design of composite granular protectors. We identify the fitness function with the force F_{out} transmitted from the protector to a "wall" that represents the body to be protected. We compute the performance of the candidate solutions under given impact loadings through a Runge-Kutta time-discretization of Hamilton's equations of motion. We adopt a Hertz-type model of interactions between adjacent beads for computational reasons, as a very large number of simulations are required by the optimization process. We also ignore dissipative effects, in accord with the standard models in the literature [6]. We note, however, that including relevant dissipative effects [51, 52] such as friction, plasticity, large deformations, and so on, using (for example) time-stepping techniques of non-smooth contact dynamics [53] or molecular dynamics [54], would not change the above EA framework. Dissipation is expected to enhance the effectiveness of the protector by further reducing the force amplitude transmitted at the wall, as shown (for example) in the experimental results reported below.

In this paper, we investigate several optimization problems. We focus, in particular, on topology, size, and material optimization of 1D composite granular chains. We compare the dynamics of the optimized systems we obtain with those of granular protectors and special granular systems (sonic vacua) available in the literature [6, 19, 22, 23]. We show that the use of EAs offers a dramatic advantage in the design of granular protectors, leading to a significant decrease of the transmitted force. This EA-driven optimal design generates suitable topology, size, and material randomization by combining effects of wave disintegration and reflection at the interfaces between different particles. A general feature we observe in the optimized protectors is the transformation of incident waves into a collection of interacting reflected and transmitted solitary pulses, which in particular form an extended (long-wavelength), small-amplitude wave that travels to the wall. We also find that optimization randomizes these systems (adding to their disorder) and produces a marked

thermalization. We constantly observe (in the absence of forced symmetry constraints) the appearance of soft/light beads near the wall, hard/heavy beads near the end impacted by the striker, and alternating hard and soft beads in the central section of the optimized chains. The observed "shock mitigation" behavior allows one to think of granular protectors in a new way—as tunable kinetic systems rather than purely as a means to add dissipation. Consequently, they offer the exciting possibility of creating much more effective energy transformation and shielding devices.

The remainder of the paper is organized as follows. In Section 2, we formulate our mechanical and numerical models. We then show how to optimize granular protectors in Section 3. We consider, in turn, topology optimization, size optimization, periodic sequences of optimized cells, and material optimization. As an extended example, we investigate the optimization of a container proposed by Hong [23] with both impulsive and shock-type loading. Finally, we summarize our results in Section 4.

2. MECHANICAL AND NUMERICAL MODELING

Consider a non-dissipative chain of N granular particles described by the Hamiltonian [6]

$$H = \sum_{i=1}^{N} \left(\frac{1}{2} \frac{p_i^2}{m_i} + V_i(q_i - q_{i+1}) - W_i(q_i) \right), \quad (1)$$

where m_i, q_i, and p_i, respectively, denote the mass, the displacement from the "packed" configuration (particles touching each other without deformation), and the momentum of the ith particle; V_i is the potential of the interaction force between particles i and $i + 1$; and W_i is the potential of the external forces acting on the ith particle (including gravity, static precompression, etc.). We introduce an $(N + 1)$th particle in order to model a wall that constrains the chain. In so doing, we assume that $p_{N+1} = q_{N+1} = 0$ during the motion. The Hamiltonian Eq. (1) yields a system of $2N$ first-order differential equations describing the motion of the system:

$$\dot{p}_i = -\frac{\partial H}{\partial q_i}, \quad \dot{q}_i = \frac{\partial H}{\partial p_i}, \quad i = 1, \cdots, N, \quad (2)$$

to be solved with the initial conditions $p_i(t = 0) = p_i^{(0)}$, $q_i(t = 0) = q_i^{(0)}$, where $t \in [0, \bar{t}]$ denotes the time variable, $\bar{t}$ indicates the final observation instant, and a dot over a variable denotes its derivative with respect to time.

Assuming that stresses remain within the elastic threshold and that particle contact areas and velocities are sufficiently small, we introduce tensionless, Hertz-type power-law interaction potentials [6]

$$V_i (q_i - q_{i+1}) = \frac{1}{n_i + 1} \alpha_i [(q_i - q_{i+1})^+]^{n_i+1}, \quad (3)$$

where α_i and, n_i are coefficients depending on material properties and particle geometry, and $(\cdot)^+$ denotes the positive part of $(\cdot)$. Most of the examples that we examine in this paper are spherical grains, for which Hertz's law implies

$$\alpha_i = \frac{4\mathcal{E}_i\mathcal{E}_{i+1}\sqrt{\frac{r_ir_{i+1}}{r_i+r_{i+1}}}}{3\mathcal{E}_{i+1}(1-\nu_i^2)+3\mathcal{E}_i(1-\nu_{i+1}^2)}, \quad n_i = \frac{3}{2}, \quad (4)$$

where r_i, $\mathcal{E}_i$, and ν_i denote, respectively, the radius, elastic (Young) modulus, and Poisson ratio of particle i. In the case of the granular container investigated by Hong [23], we instead use the values shown in Table 1 for α_i and n_i.

Additionally, let

$$T = \sum_{i=1}^{N} \frac{1}{2}\frac{p_i^2}{m_i}, \quad V = \sum_{i=1}^{N}[V_i(q_i - q_{i+1}) - W_i(q_i)], \quad (5)$$

and $E = H = T + V$, where T denotes the system's kinetic energy, V denotes the potential energy, and E denotes the total energy. We also introduce the local energies

$$E_i = \frac{1}{2}\frac{p_i^2}{m_i} + \frac{1}{2}[V_i(q_{i-1} - q_i) + V_i(q_i - q_{i+1})] \quad (6)$$

at each site (bead), and the *energy correlation function* (which is slightly different from that introduced in Ref. [12])

$$C(t,0) = \frac{c(t)}{c(0)}, \quad (7)$$

where

$$c(t) = \frac{1}{N}\left\langle \sum_{i=1}^{N} E_i^2(t) \right\rangle - \left\langle \frac{1}{N}\sum_{i=1}^{N} E_i(t) \right\rangle^2, \quad (8)$$

$\langle\cdot\rangle$ denotes the average over time (from 0 to the current time), and $C(t,0)$ indicates how the energy is transferred between the different beads. Observe that $C(t,0) = 0$ corresponds to energy equipartition.

Equations (2) can be solved numerically using a standard fourth-order Runge-Kutta integration scheme (as discussed in, for example, Ref. [6]) with a time integration step of

$$\Delta t = k\left[\min_{i=1,\dots,N}\left\{\frac{r_i}{c_i}\right\}\right], \quad (9)$$

where c_i is the sound speed in the material for the ith particle and $k \in (0, 1]$ is a scaling factor. Equation (9) gives a time-integration step of about 2×10^{-8} s for $k = 0.1$ and 1 mm stainless steel bead chains (see Table 2), ensuring relative errors lower than 10^{-8} in the total energy conservation for times up to few thousand μs.

We now assume that the configuration of the granular system is described by a collection of design variables or "genes" (which can include particle radii, mass densities, elastic moduli, material types, etc.)

$$x = \{x_i\}_{i=1,\dots,M}, \quad (10)$$

subject to simple bounds of the form

$$x \in X = [x_1^{lb}, x_1^{ub}] \times \dots \times [x_M^{lb}, x_M^{ub}]. \quad (11)$$

One can always assume that $x_i^{lb} = 0$ and $x_i^{ub} = 1$ for all $i \in \{1, \dots, M\}$ through suitable rescaling of design variables.

Given an assigned x, a numerical simulation of the system dynamics under a prescribed impulse or shock loading gives the protection performance (fitness) $f = \|F_{out}\|_{L^\infty}$ of the corresponding design configuration. Here, F_{out} denotes the force profile transmitted from the system to the wall, and $\|F_{out}\|_{L^\infty}$ denotes its norm with respect to the Sobolev space $L^\infty([0, \bar{t}])$ [55]. The optimal design configuration x_{opt} can then be identified with the solution of the multivariate optimization problem,

$$\min_{x \in X} f(x), \quad (12)$$

which is expected to be influenced by multiple local optima. The problem (12) can be conveniently solved via EAs (see, for example, Refs. [46–50, 56]) through the cyclic iterative procedure illustrated in Figure 1. In the present paper, we will use the Breeder Genetic Algorithm (BGA) presented in Ref. [57]. BGAs, in contrast to other EAs (in which the selection is fully stochastic), selects only from among the $T_R\%$ best elements of the current population of N_I individuals (where $T_R\%$ denotes the so-called *truncation rate*) to be recombined and mutated (mimicking animal breeding). This feature makes BGAs more efficient than standard EAs for performing optimization in large search spaces [58, 59].

3. OPTIMIZATION OF GRANULAR PROTECTORS

We deal in this section with topology, size, and material optimization of 1D composite granular protectors subject to impulsive and shock-type loadings. We employ formula (9) with $k = 0.1$ for time discretization and always assume that genes are continuous variables ranging over [0, 1] with a population size of 50 individuals. We use an initial, randomly-chosen truncation rate (T_R) equal to 15%, employ Extended Intermediate Recombination (EIR) [59], and consider mutation rates in the interval [10%, 50%]. (We use the value 10% for size optimization, which has genuinely continuous genes, and 50% in all of the other examples, which instead model discrete design variables using continuous genes.) EIR generates offspring along the line defined by the parents in the search space and allows one to also
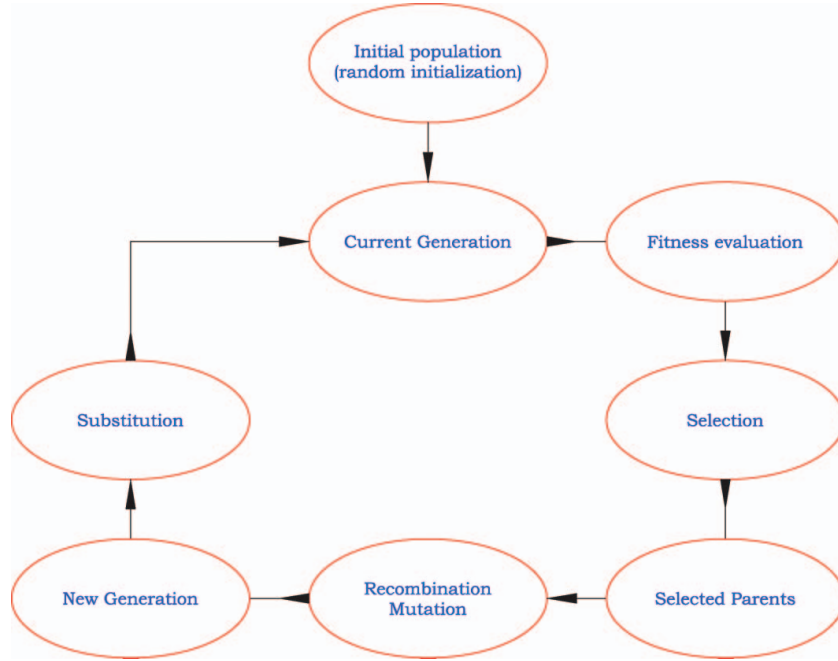
FIG. 1.   Diagrammatic representation of an evolutionary algorithm.

create offspring outside of the segment joining the parents. See [57–59] for further technical details of the employed BGA.

The examples of Sections 3.1 and 3.2 consider chains of stainless steel beads, whereas those in Section 3.4 examine a composite chain composed of polytetrafluoroethylene (PTFE) and stainless steel beads. We show the material properties of these beads in Table 2. The final example, discussed in Section 3.5, considers a long composite chain—the protector recently investigated by Hong [23])—with the material properties shown in Table 1. We studied protectors with one end in unilateral contact with a rigid wall (simulating the body to be protected) and the other end free. In most cases, we assumed that the free end was impacted by a striker; in the final example, we assumed that it was loaded by a prescribed force. We focused our attention on the short-term dynamics of the protector over an observation time slightly larger than that necessary to transmit the input actions to the wall.

TABLE 1

Material properties (mass m, contact coefficient α, and contact exponent n) of the granular container investigated by Hong [23] (in abstract units, as discussed in the main text)

Label	m	α	n
mat1	2.0	5657	1.0
mat2	1.0	5657	2.0
mat3	0.3	5657	1.5
mat4	0.1	5657	1.5

3.1.   Topology Optimization

Nesterenko used the moniker *sonic vacua* to describe unprecompressed (or weakly precompressed) granular chains because the sound speed is zero or very small in such systems [6]. He studied the behavior of two adjacent monodisperse sonic vacua (2SV), characterized by a sharp variation in bead size (i.e., a *stepped* 2SV), under the impact of a striker. He observed two remarkable phenomena: disintegration of the incident pulse into a solitary wave train when it passes from the sub-chain with larger radius to the one with smaller radius; and a partial reflection in the opposite case (see also Ref. [26]). Here we examine topology optimization of a stepped 2SV in order to determine the particle arrangement that minimizes F_{out} under a given impact event. Figure 2 shows different force-time histories in a 2SV hit by a striker at the sub-chain with larger radius. The system is composed of 20 large beads of radius $r = r_L = 3.95$ mm, 20 small beads of radius $r = r_S = 2.375$ mm, and the striker (particle number 1), which has radius $r = r_L$ and initial velocity $v = 1$ m/s (see § 1.6.10 of [6]). The plots in Figure 2 show the force F_{in} at the contact between the striker and the first bead, the

TABLE 2

Material properties (mass density ρ, elastic modulus $\mathcal{E}$, and Poisson ratio ν) of stainless steel and PTFE beads [20]

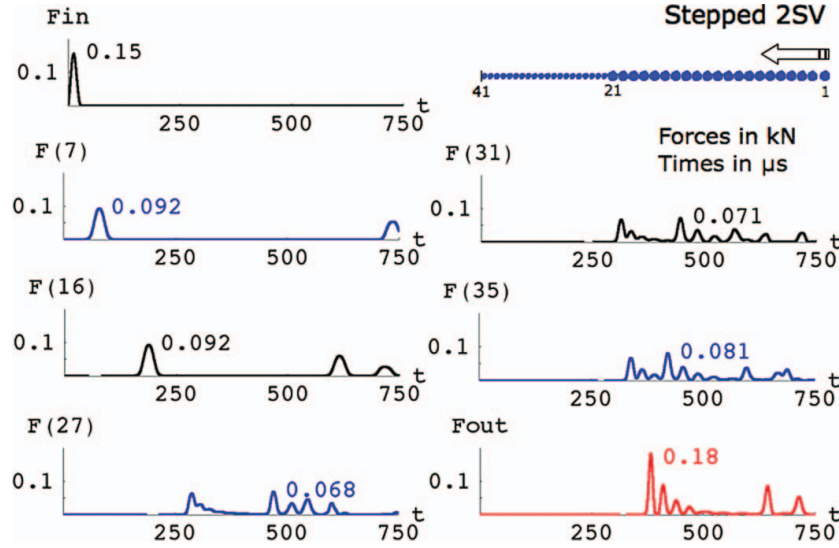|	ρ (kg/m^3)	$\mathcal{E}$ (GPa)	ν
Stainless steel	8000	193.00	0.30
PTFE	2200	1.46	0.46

FIG. 2. Force vs. time plots for the stepped two sonic vacua (2SV). The striker impacts the end with larger-radius beads.

force $F(i)$ that denotes the mean of the contact force between particles i and $i-1$ and that between $i+1$ and i, and the force F_{out} recorded at the wall. The observation time is 750 μs. All of the beads are made of stainless steel (see the material properties in Table 2). The $F(i)$ plots for $i > 21$ in Figure 2 clearly illustrate the aforementioned pulse disintegration phenomenon. One can also see that the fitness $f = \|F_{out}\|_{L^\infty}$ of the 2SV is equal to 0.18 kN.

We ran a topology optimization of the 2SV by introducing $M = N = 40$ genes x_i related to the radius size (large or small) of the different beads. (This does not include the striker—particle number 1—which is assumed to have a large radius.) We defined the genes so that $x_i \in [0, 0.5]$ implies

$r_{i+1} = r_S$, whereas $x_i \in (0.5, 1]$ implies $r_{i+1} = r_L$. We used a penalty technique to constrain the number of particles with large and small radii to each be equal to 20; that is, we assigned a very large fitness f to (unfeasible) solutions that do not satisfy this criterion. We show the BGA-optimized system and the corresponding force-time plots in Figure 3. The optimized system has many large beads near the end of the chain that is hit by the striker (shown on the right), small beads near the wall, and an alternation of sequences of multiple consecutive large and small particles in the center of the system ($x_{opt} = \{1, 0, 1_2, 0_2,$ $1_3, 0_3, 1_5, 0_9, 1_6, 0_2, 1, 0, 1, 0_3\}$, where 1 denotes a large particle, 0 denotes a small particle, and a_b refers to b consecutive particles of type a). We obtained a stable solution (i.e., a
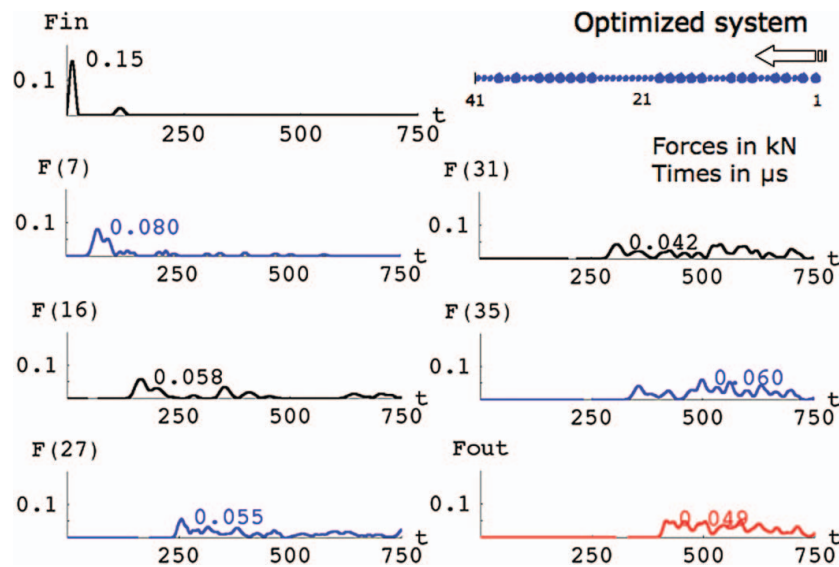

FIG. 3. Force vs. time plots in the topology-optimized system. (Compare this to the (unoptimized) stepped 2SV configuration in Figure 2.)
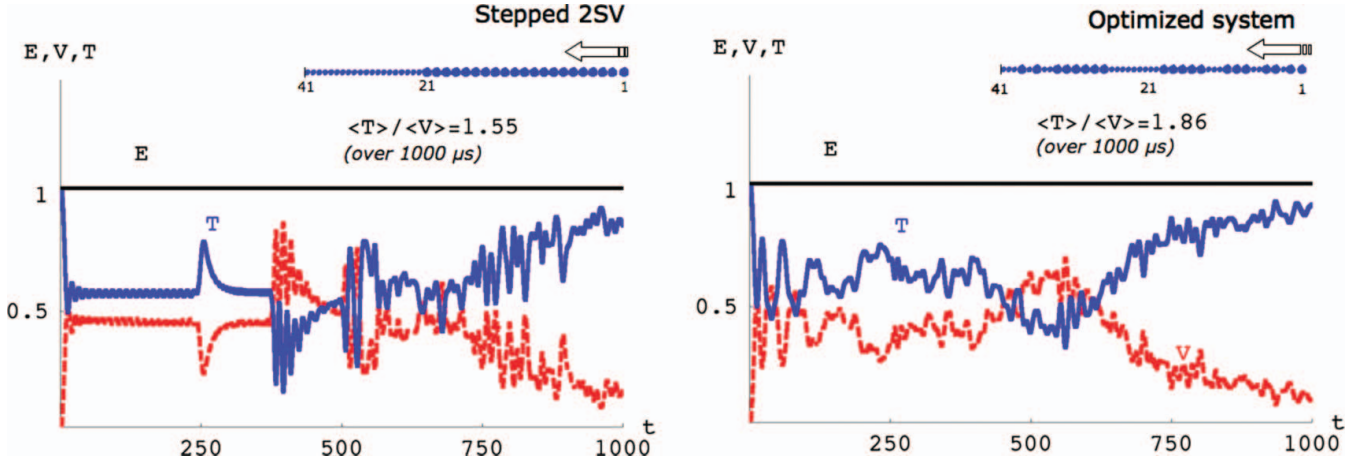
FIG. 4.   Energy vs. time plots in the stepped 2SV and in the topology-optimized system (energies are in mJ, and times are in μs).

solution with constant best fitness) after about 340 generations of the algorithm. Observe that pulse disintegration appears very early (within the first few beads) in the optimized system, so that the leading solitary wave transforms into a train of interacting, small-amplitude pulses. This configuration exhibits a fitness of about 0.049 kN, which is almost four times smaller than that of the 2SV.

We compare the energies (as a function of time) of the stepped 2SV and the optimized system in Figure 4 over a time window preceding the achievement of a loose state (in which there are no interaction forces).[1] We obtained this by restricting the energy time-histories up to the first instant $t > 0$ for which $T = 0.99E$. The kinetic energy T of the 2SV shows a marked peak when the leading wave passes from the larger sub-chain to the smaller one and valleys when the wave is reflected at the wall. The potential energy V behaves in the opposite manner because the total energy is conserved. In the optimized system, on the other hand, the valleys and peaks of T and V arise earlier during wave propagation, and the peaks of the potential energy are markedly lower than those observed in the 2SV. Denoting by $\langle T \rangle$ and $\langle V \rangle$ the time-averaged values of T and V, respectively, over a time of 1000 ms from the striker impact, we observe that in both the 2SV (which has $\langle T \rangle / \langle V \rangle \approx 1.55$) and the optimized system (which has $\langle T \rangle / \langle V \rangle \approx 1.86$), the ratio $\langle T \rangle / \langle V \rangle$ deviates from the value 1.25 predicted by the virial theorem of statistical physics [11]. Nesterenko observed similar results using randomized granular chains subject to piston-like impacts [6].

Figure 5 shows density plots of particle energies E_i for the stepped 2SV and the optimized system. In each plot, the horizontal axis indicates the particle site, the vertical axis shows the time step (we produced a plot for every five integration steps), and the shading gives the energy normalized to unity (i.e., the energy

[1]In the absence of gravity and precompression, a loose state is reached after a sufficiently long time because the granular chain is constrained only at one end. The dynamics evolve so that the interactions go to zero.

divided by its maximum value among all of the beads). One can clearly recognize the disintegration phenomenon in the stepped 2SV when the incident pulse passes from the large-bead regime to the small-bead regime. In this system, pulse reflection occurs only at the wall (and not along the body of the chain) during the first transmission. Note that when already-reflected pulses pass from the small-bead regime to the large-bead regime, they are reflected for a second time. In the optimized system, however, one observes a combination of disintegration and reflection of traveling pulses along the entire chain. One also observes the production of interacting pulses that travel in opposite directions. In particular, the pulses moving toward the wall combine to form an extended (long-wavelength), small-amplitude wave that is clearly visible in Figure 3. Figure 6 shows the time histories of the energy correlation function for the stepped 2SV and topology-optimized systems, revealing that the latter exhibits a faster and stronger thermalization (i.e., equipartition of energy) than the former.

3.2.   Size Optimization

Doney and Sen recently studied the energy absorption capabilities of 1D granular protectors consisting of "tapered" and/or "decorated" chains [21, 22]. The simplest type of tapered chain is composed of a sequence of progressively larger or progressively smaller beads, and a decorated chain is a composite chain obtained by placing interstitial small grains between the large grains in a monodisperse chain. Using analytical and numerical techniques, Doney and Sen observed marked energy absorption in highly tapered chains. For their analytical work, they employed the Hertz contact model. In their numerical studies, they utilized hydrocode simulations up to very high impact velocities (up to 1 km/s). In Figures 7 and 8, we show several numerical force recordings for, respectively, a decorated and a tapered chain impacted by strikers with different velocities. In both cases, the initial force peak is $F_{in} = 1.4$ kN. The decorated chain consists of a sequence of dimers formed by alternating
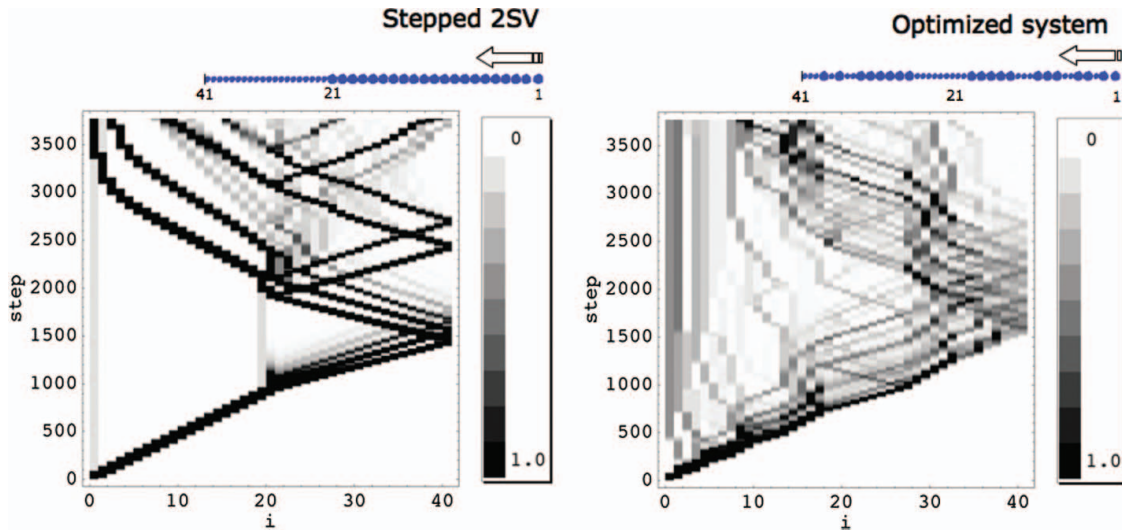
FIG. 5. Density plots of particle energies normalized to unity. Horizontal axes are labeled according to particle site and vertical axes give the time step.
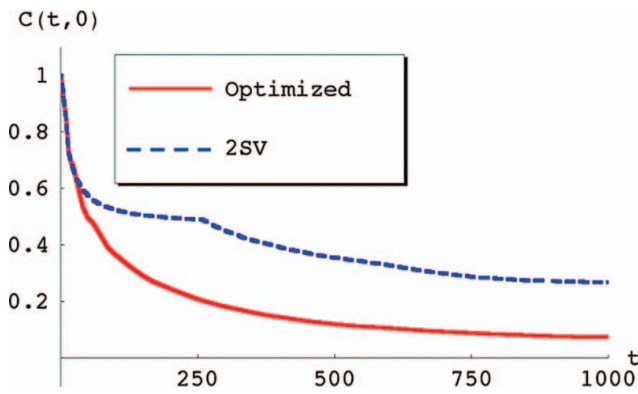

FIG. 6. Energy correlation function vs. time for the topology-optimized and stepped 2SV systems.

$r = \hat{r}_L = 3.243$ mm and $r = \hat{r}_S = 0.973$ mm stainless steel beads with total length 7.78 cm (not counting the striker). The striker (3.243 mm) impacts this chain with an initial velocity of 18.88 m/s. The tapered chain, on the other hand, is composed of 20 stainless steel beads with decreasing radii, ranging from $r = r_L = 5$ mm (the striker, which impacts the chain at a speed of 7.20 m/s) to $r = r_S \approx 0.675$ mm over a length of 7.78 cm (not counting the striker). That is, the tapering ratio is $q_s = r_{i+1}/r_i = 0.9$. The fitness of the decorated chain is about 1.7 kN, whereas that of the tapered chain is equal to about 0.75 kN.

We carried out a size optimization of the above systems, introducing $M = 19$ genes x_i related to the radii of the different beads (not including the striker), which in the
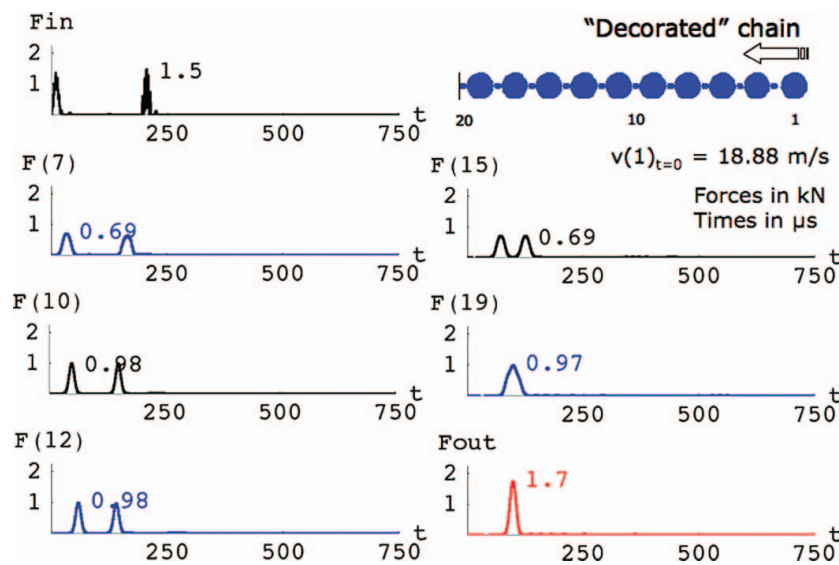

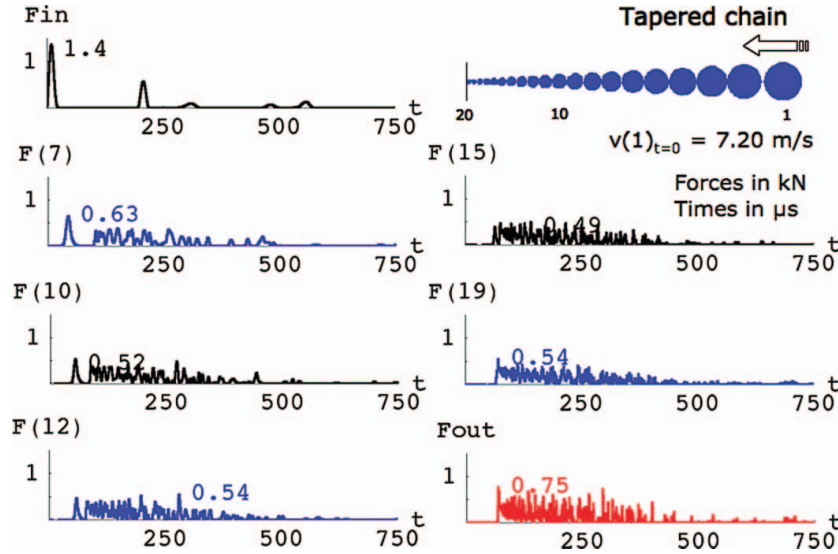FIG. 7. Force vs. time plots in a decorated chain.

FIG. 8.   Force vs. time plots in a tapered chain.

present example (in contrast to the previous one) were assumed to change continuously in the interval $[r_S, r_L]$ (with $r_{i+1} = r_S + x_i(r_L - r_S)$). We ran a BGA optimization with a striker radius always equal to 5 mm and an impact speed equal to 10 m/s. We also constrained the total length of the system to remain equal to 7.78 cm (not including the striker). We obtained constant best fitness and the solution shown in Figure 9 after about 590 generations (x_{opt} ={0.605, 0.584, 0.570, 0.677, 0.009, 0.021, 0.506, 0.988, 0.875, 0.012, 0.006, 0.469, 0.015, 0.568, 0.018, 0.007, 0.001, 0.002, 0.011}). The fitness of the size-optimized system (0.42 kN) is about 1.8 times smaller than that of the (unoptimized) tapered chain. Observe that there is simple reflection at the wall in the decorated chain (see Figure 7), significant pulse disintegration in the tapered chain (see Figure 8), and a transformation of

the leading solitary pulses into an extended (long-wavelength), small-amplitude wave in the optimized chain (Figure 9). The choice of the fitness parameters used in this study differs from the strategy employed in Ref. [22], which instead attempts to minimize the kinetic energy ratio K_{out}/K_{in} between output and input. Because of the continuum formulation of the genes in the current examples, our work encompasses all geometries considered in Ref. [22]. We compare the energy profiles of the decorated, tapered, and optimized chains in Figure 10. We observed the highest $\langle T \rangle / \langle V \rangle$ in the tapered chain ($\langle T \rangle / \langle V \rangle \approx 1.94$), resulting from the anticipated evolution of this system toward a loose state. Figure 11 shows the density plots of particle energies for the three systems under examination. As in the previous example, one can clearly observe from the plots the effects of both wave disintegration and reflection in the optimized system. The
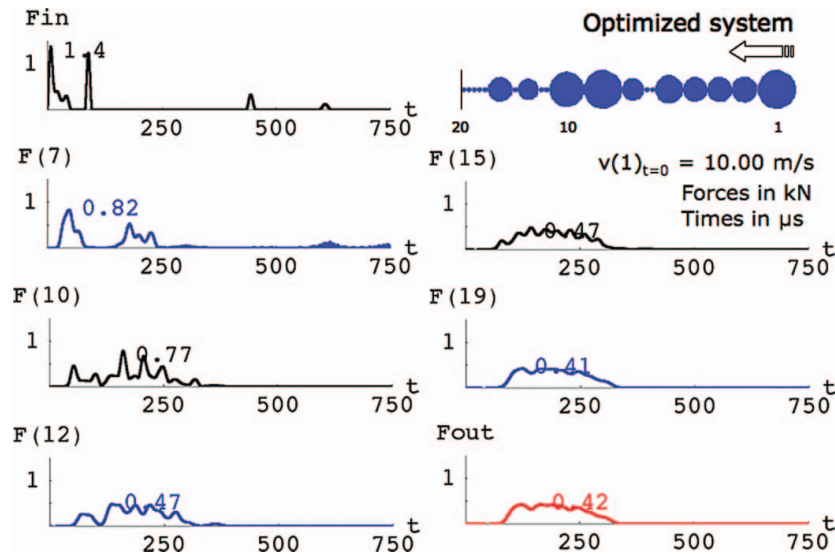

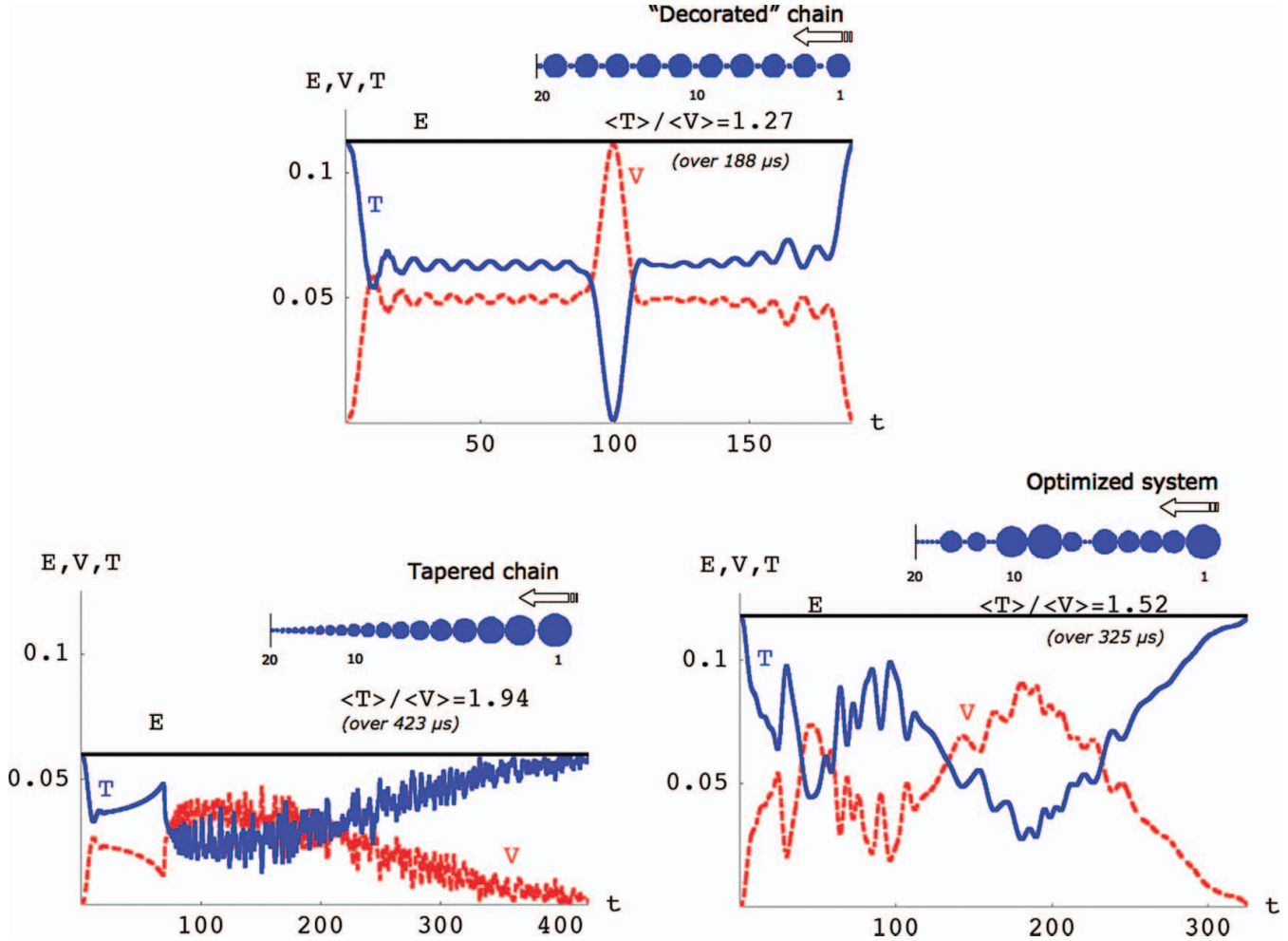FIG. 9.   Force vs. time plots in the size-optimized system.

FIG. 10. Energy vs. time plots in the decorated, tapered, and size-optimized chains (energies in J, times in μs).

profiles of the energy correlation function shown in Figure 12 indicate that the size-optimized and tapered chains both evolve toward thermalization and that the decay of $C(t, 0)$ is slightly faster in the former system. We show in the next section that a similar behavior can also be induced in a long dimeric system (i.e., in a long decorated chain) by introducing suitable alterations of the periodic particle arrangement (i.e., by introducing another form of disorder into the system).

3.3. Periodic Sequences of Optimized Cells

We now consider a periodic sequence of the 19-particle size-optimized cell in the decorated chain shown in Figure 9. As discussed above, the single optimized cell that we obtained can be viewed as a disordered configuration, so it is interesting to investigate the effects on the wave dynamics of periodically repeating such a structure to obtain a "quasi-disordered" configuration. As can be seen in Figure 13, a reasonably localized wave structure does develop as long as there are enough periods, just as with periodic arrangements of simpler cells [25, 60].

However, the original wave disintegrates and emits a significant number of secondary pulses, so that a stable coherent structure does not form.

Using long-wavelength asymptotics, one can obtain a nonlinear partial differential equation (PDE) description of the decorated chain in the continuum limit [6, 25, 60]. This PDE has known compact solitary wave solutions, which we illustrate in the top panels of Figure 14. However, adding even a small number of impurities to the system can change things completely (introducing some pulse disintegration, pulse reflection, and thermalization), although one still obtains a basically localized pulse. The impurities that we consider consist of particles of radius 5 mm and mass 2.31 g, so that they are much larger and heavier than the other beads in the chain. Throughout the region of the chain that has impurities, we place one of them every 19 particles, giving a cell length that is the same as that in the periodically repeated size-optimized chain of Figure 9. The second through fourth rows of Figure 14 contain regions of different lengths that contain these periodic impurities. In each case, the last impurity is placed before bead 1000. The first
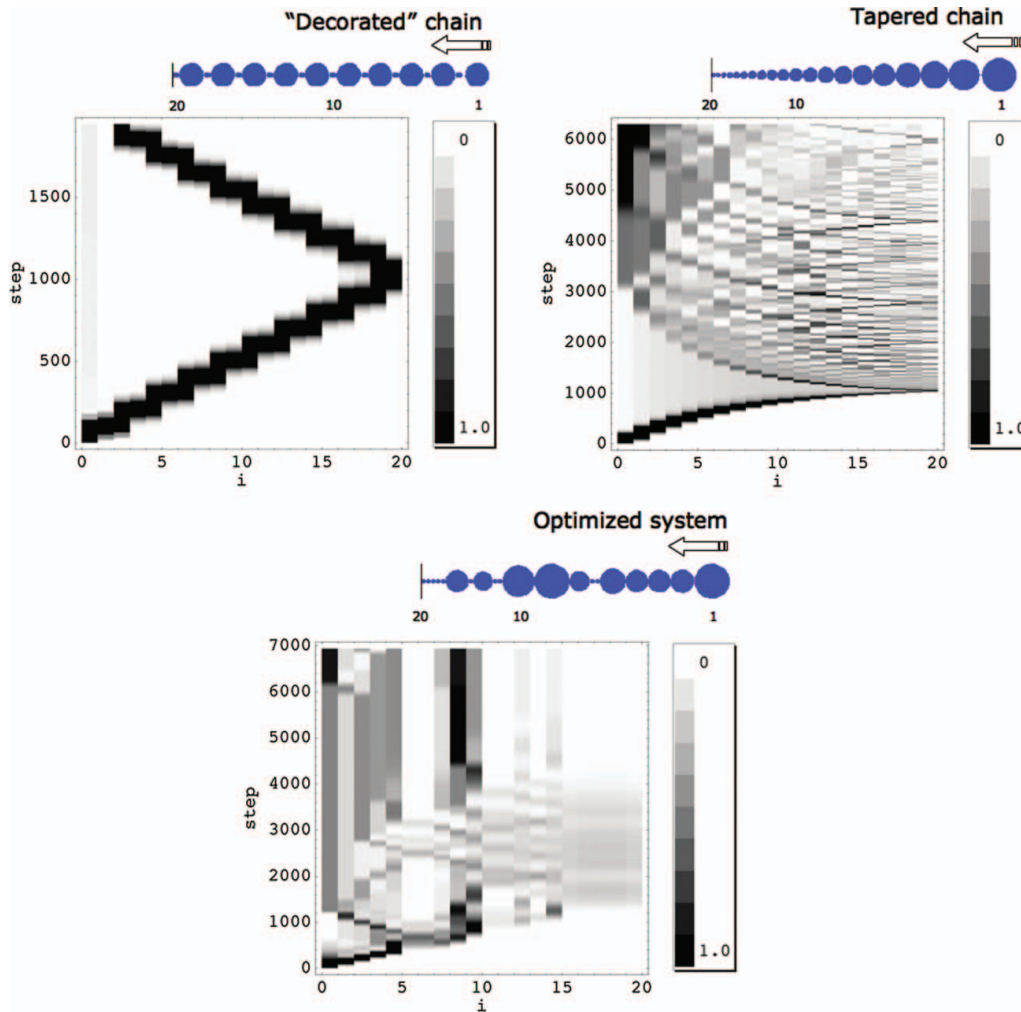
FIG. 11. Density plots of particle energies normalized to unity. Horizontal axes are labeled according to particle site and vertical axes give the time step.

impurity is in particle 501 in the second row of Figure 14. The third row of the figure is for a chain with impurities every 19 beads starting from bead 801, and the bottom row is for a chain with impurities every 19 beads starting from bead 951 (so that there are three impurities in total—at beads 951, 970, and 989—in this last example). As shown in these plots, the insertion of such heavy impurities leads to partial reflections of the wave, some thermalization, significant pulse disintegration, and even a bit of trapping (see the bottom left panel). Also observe in the bottom row that we have induced delays in the wave reflection. By tuning the material properties carefully, one can perhaps optimize the properties of such wave trapping so that they can be exploited in applications. Moreover, these numerical experiments also illustrate the complicated series of secondary pulse emission and partial wave reflections that occur in the periodic sequence of optimized/randomized cells.

The study of soliton-like pulses in perturbations of uniform Hertzian chains was reported earlier in works that suggest the possibility of using them as systems to detect buried impurities via the analysis of back-scattered signals [61, 62]. Our results

show that similar phenomena can also be observed in "quasi-disordered" systems; as shown in Figure 14, the masses and positions of the defects in the chain detectably shift the reflection and the radiation.
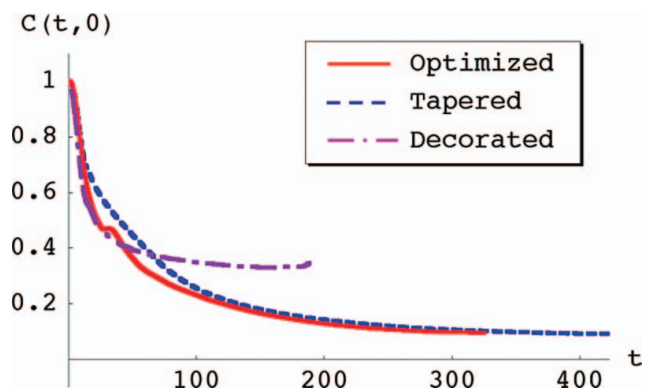

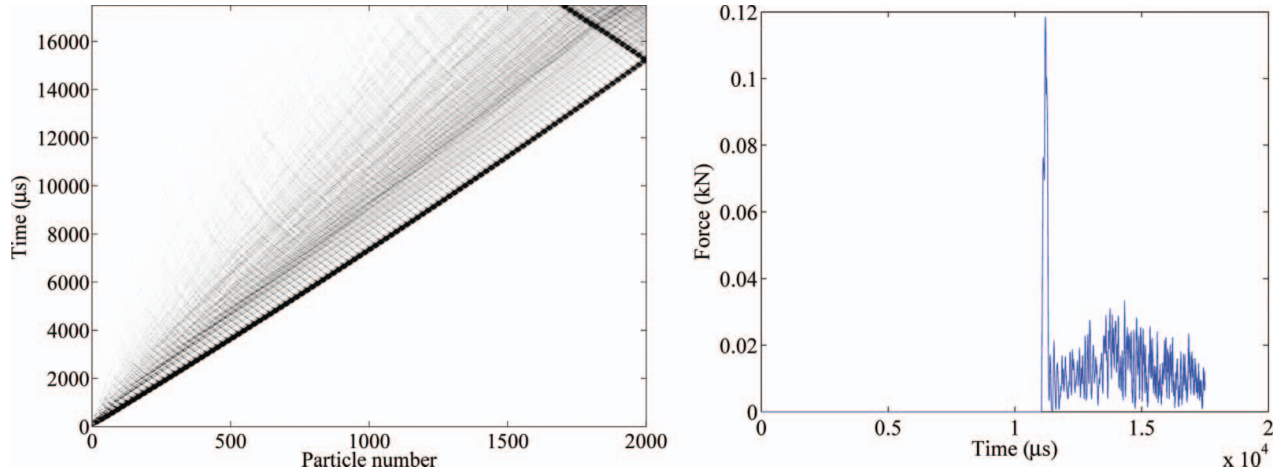FIG. 12. Energy correlation function vs. time for the examined systems.

FIG. 13. (Left) Density plot (colored by force, with larger values given by darker shading) of the optimized chain in Figure 9. (Right) Force (in kN) vs. time plot for particle 1500 of this chain.

It would be interesting to extend this type of discussion by considering increasingly disordered configurations, such as systems with quasiperiodic arrangements of cells (with various lengths and component particles) rather than periodic ones. Some preliminary research in this direction (using, for example, arrangements that follow Fibonacci sequences) has appeared recently in the literature in order to study Anderson localization in atomic chains [63, 64]. It would similarly be interesting to consider systems with random arrangements of cells.

3.4. Material Optimization

In Ref. [19], Daraio et al. investigated the optimization of a composite granular protector consisting of 22 stainless steel beads and 10 PTFE beads (see Table 2) with a uniform radius of 2.38 mm. The authors examined different design solutions, based on material distribution, using both numerical and laboratory experiments. The protector they considered was impacted by an Al_2O_3 striker (0.47 g) with initial velocity of 0.44 m/s and was initially precompressed by a static force of 2.38 N. Using piezosensors embedded in selected particles, they obtained laboratory measurements of force versus time profiles in several beads and compared them against numerical predictions. Here we report analogous experiments to confirm the results that we obtained from optimization. We used a four-garolite-rod stand as the holder for the beads and assembled sensors as described in Refs. [25, 60]. We selected a steel particle (0.45 g) as the striker and recorded the traveling signal with a TKTDS 2024 oscilloscope (Tektronix, Inc.). The sensors (PiezoSystems, Inc.) were calibrated by conservation of linear momentum. The 1.86 N static precompression included the preloading of the topmost particle with about 190 g of symmetrically suspended masses.

Figure 15 shows numerical and experimental force recordings in a soft-hard-soft configuration (see Figure 2 of Ref. [19]) with sequences of five PTFE beads at both the top and bot-tom of the chain. This system had the minimum value of F_{out} of all of the configurations considered in [19]. (A different, hard-soft-hard-soft-hard configuration minimized F_{out}/F_{in} ratio, as shown in Figure 3 of Ref. [19].) Observe the good qualitative agreement between numerical and experimental results over the initial phase of the pulse propagation. The dynamics of the experiments and numerics subsequently deviate from each other, as the laboratory tests are affected by dissipative effects that are not included in the numerics. Dissipation can arise from friction, inelastic collisions, viscous drag, etc. The experimental traveling wave is thus progressively damped as it travels through the chain, resulting in an even better protector.

We carried out a material optimization by introducing 32 genes x_i related to the material identification of the individual beads (not including the striker). They are defined such that $x_i \in [0, 0.5]$ implies that the $(i + 1)$th bead is made of PTFE, whereas $x_i \in (0.5, 1]$ implies that the same bead is instead made of stainless steel. (We assumed that the striker was always made of steel.) We constrained the total number of PTFE beads to be equal to 10 through a penalty technique. We also introduced an additional gene (so that the total number of genes M is equal to 33) related to the intensity of the preloading, allowing the static precompression force F_0 to vary continuously within the interval [0, 2.38] N (i.e., $F_0 = 2.38x_{33}$ N). The optimized system, obtained after about 110 BGA generations, is shown in Figure 16 together with the corresponding numerical and experimental force plots ($x_{opt} = \{1_{11}, 0, 1_2, 0, 1, 0_2, 1, 0_2, 1_7, 0_4\}$, where 1 denotes a steel particle and 0 denotes a PTFE particle). As in the previous examples, observe that the material-optimized system contains soft beads near the wall, hard beads near the end impacted by the striker, and alternating hard and soft beads in the central section of the chain. We computed the optimal precompression force to be about 1.86 N. Both the numerical and the experimental force plots of Figuer 16 show that the leading solitary wave first decomposes into a
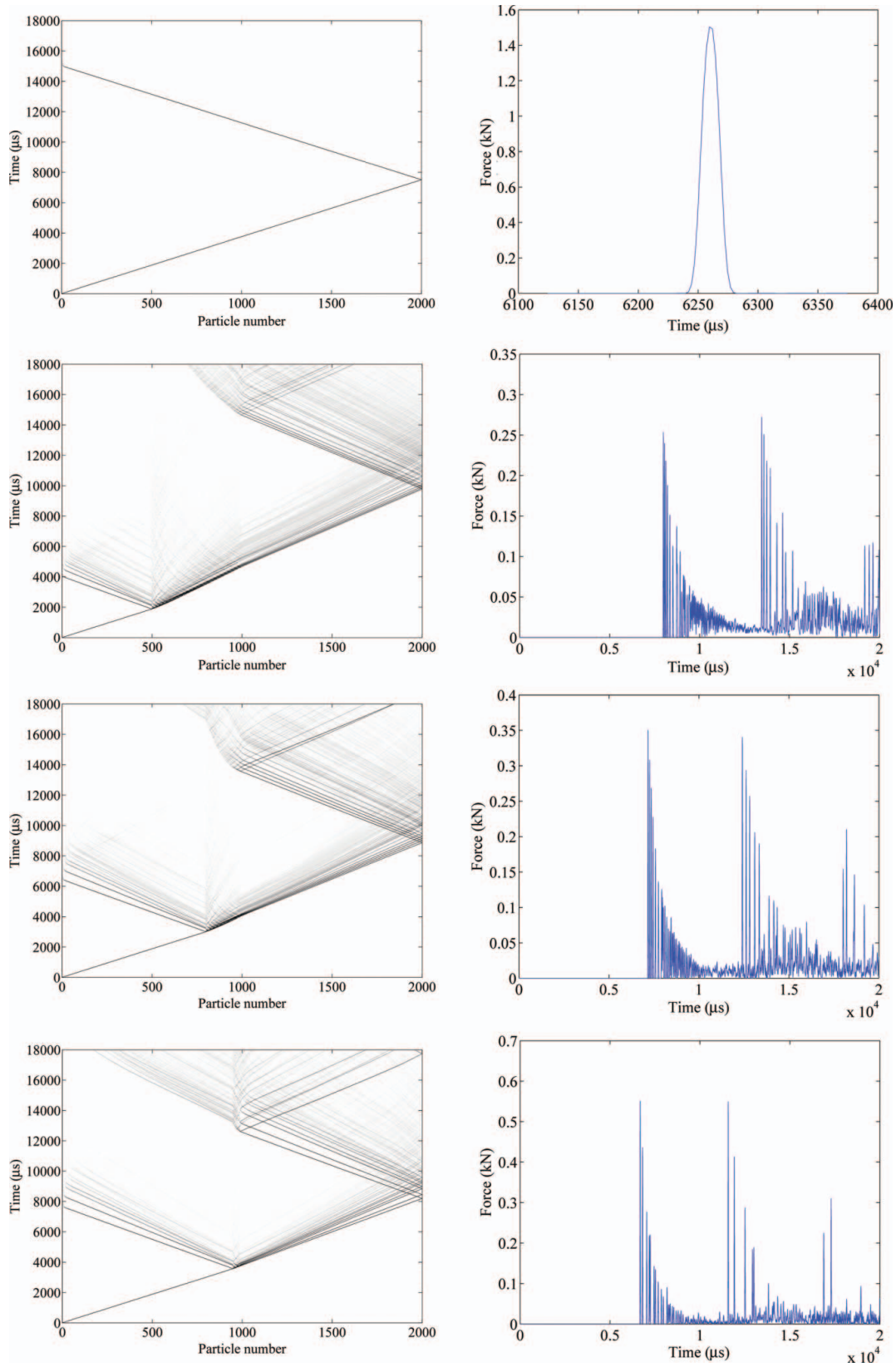
FIG. 14.   (Left) Density plots (colored by force, with larger values given by darker shading) for (top row) decorated chain of Figure 7 and for decorated chains with a single impurity in each 19-particle cell between beads 501 and 1000 (second row), 801 and 1000 (third row), and 951 and 1000 (bottom row). (Right) Force (in kN) vs. time plots for particle 1500 in each of these chains.
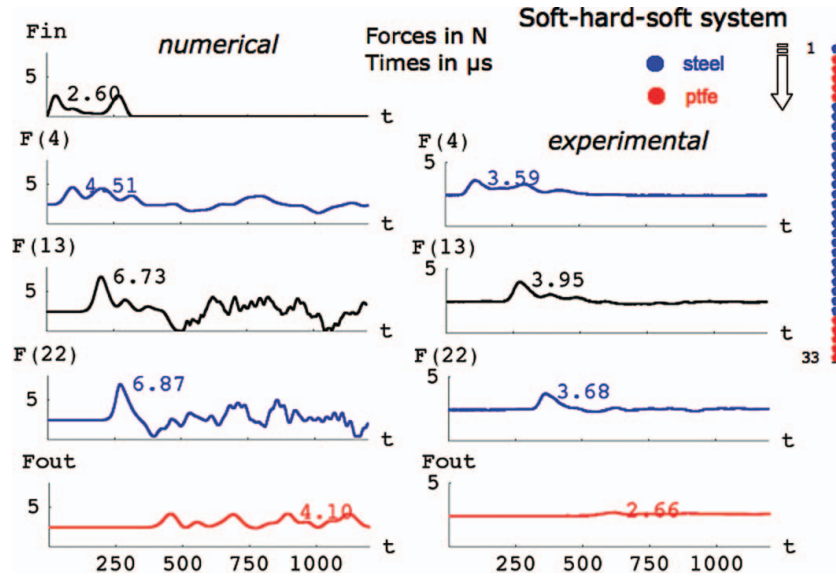
FIG. 15.   Force vs. time plots in a soft-hard-soft granular chain. (The applied precompression was added to the force profiles experimentally recorded through piezosensors.)
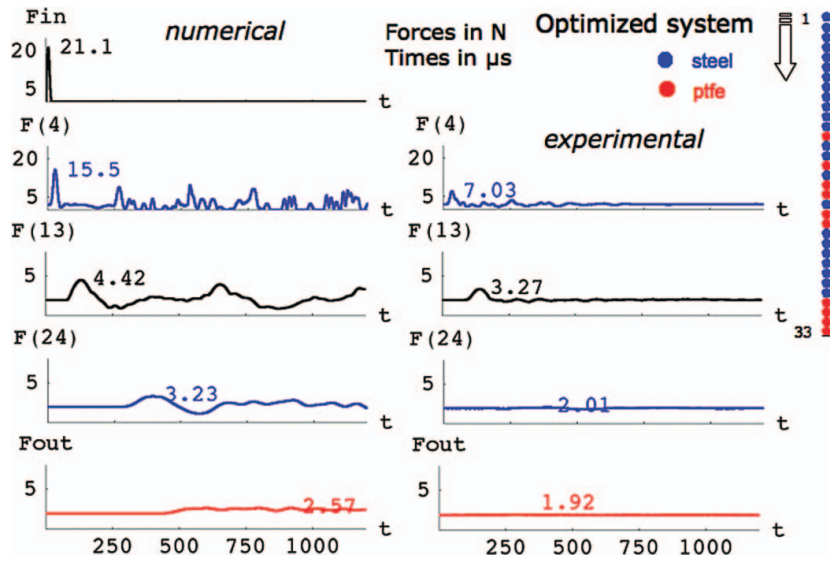

FIG. 16.   Force vs. time plots in the material-optimized system. (The applied precompression was added to the force profiles experimentally recorded through piezosensors.)
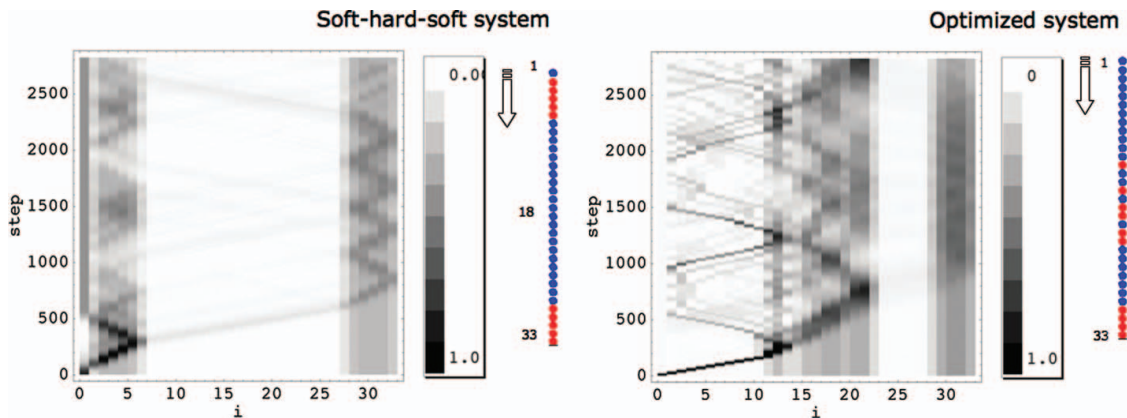

FIG. 17.   Density plots of particle energies normalized to unity. Horizontal axes are labeled according to particle site and vertical axes give the time step.
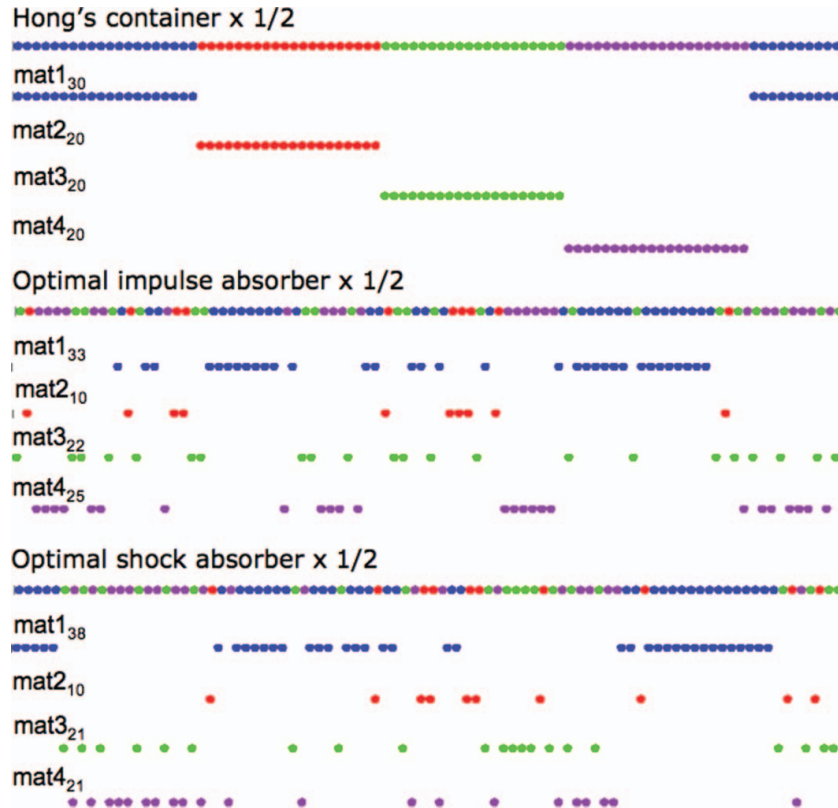
FIG. 18.   Hong's container and optimized composite long chains (showing half of the chain; the other half is obtained by reflection about the right end of the depicted half).
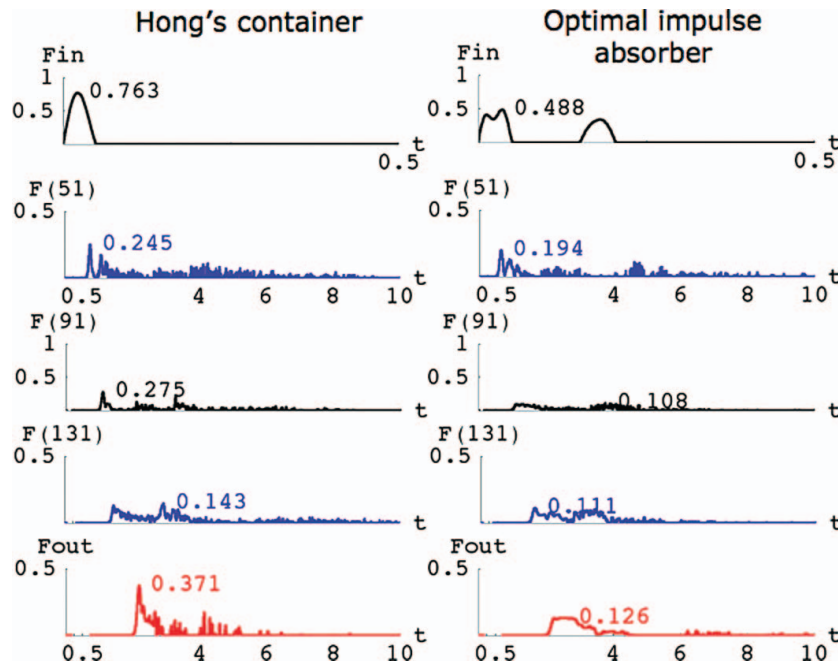

FIG. 19.   Force vs. time plots in a long composite chain subject to impulsive loading (force values divided by 1000).
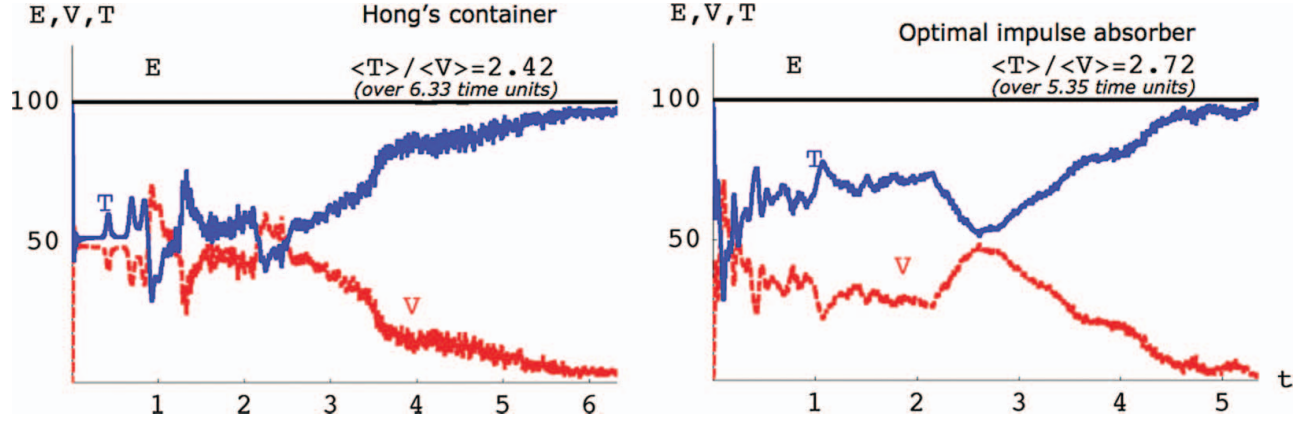
FIG. 20. Energy vs. time plots in Hong's container and the optimized chain under impulsive loading.

train of small pulses and subsequently mutates into an extended (long-wavelength), small-amplitude wave. The density plots of Figure 17 illustrate the mechanisms of wave disintegration and reflection characterizing the dynamics of these systems. The experimental results presented in this article and in Ref. [19] confirm the enhanced performance of the BGA-optimized system (minimum F_{out}), as compared to all of the other examined protectors.

3.5. Long Composite Protector

In a recent paper [23], Hong investigated a long 1D composite granular protector (or "energy container") consisting of nine 20-bead sections. The beads in this chain were made of four different materials with varying particle mass m, contact stiffness α, and contact exponent n (see the parameter values in Table 1). Hong used abstract units, introducing factors of 10^{-5} m, 2.36×10^{-5} kg, and 1.0102×10^{-3} s to convert the adopted units of length, mass, and time, respectively, into real units. The terminal and central sections of the protector are composed of a (linear) material ("material 1") characterized by a contact exponent $n = 1$ and mass $m = 2$. The remaining inner sections are composed of beads (made of nonlinear materials) that have

different masses and contact exponents greater than 1 (materials 2, 3, and 4). This is used to simulate sharp contact surfaces and rough materials such as sand (see Figure 18). The container has a reflection symmetry about its center and the initial distance between particle centers of mass is uniformly equal to 200 along its body.

Hong analyzed the behavior of this container (and variations thereof) by considering its dynamics after the impact of a striker of mass $m = 100$ traveling with speed $v = 10$. He ran numerical simulations of the chain dynamics, employing a Hertz-type contact model and introducing lateral constraints through additional beads consisting of very heavy grains ($m = 100$). He found a universal power-law scaling for how long it took the energy to leak from the protector to the lateral sides. That is, the energy remaining in the protector is given by $E_R = At^{-\gamma}$, where A is a constant that depends on the protector construction, t is the time, and γ is a constant (that Hong estimated in Ref. [23] to be about 0.7055) independent of the protector construction.

We carried out a joint topology-material optimization of Hong's container (shown in the top panel of Figure 18), introducing $M = 90$ genes x_i that characterize the material identification of each individual bead. The conditions $x_i \in [0, 0.25]$, $x_i \in$
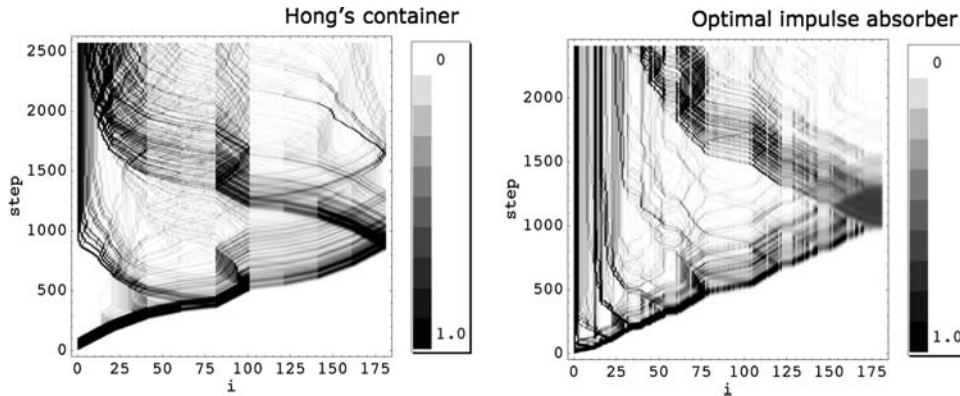

FIG. 21. Density plots of particle energies normalized to unity. Horizontal axes are labeled according to particle site and vertical axes give the time step.
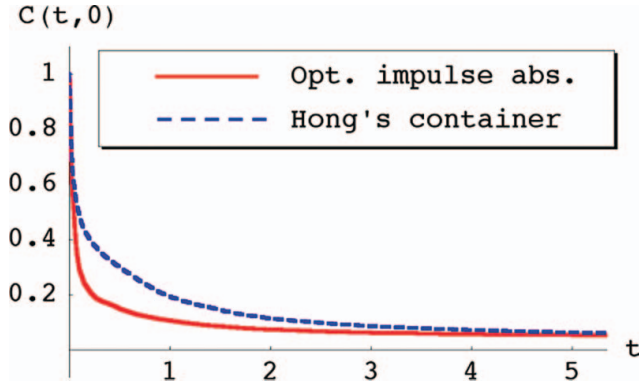
FIG. 22.   Energy correlation function vs. time for the examined systems.

(0.25, 0.5], $x_i \in$ (0.5, 0.75], and $x_i \in$ (0.75, 1] respectively imply that the ith bead of one half of the protector is composed of material 1, 2, 3, and 4. We enforced the symmetry with respect to the center of the chain by suitably relating the material identification numbers of the 90 grains in the second half to those of the grains in the first half. We did not enforce any constraints on the numbers of beads of the different materials. Due to the symmetry constraint, the optimized protector is not allowed to have different constructions near the impacted and the constrained ends, in contrast to the protectors we examined in the previous sections.

3.5.1.   Impulsive Loading

In our first optimization procedure, we considered the impact of an $m = 2$ striker (material 1) traveling with speed $v = 10$. We show the corresponding optimized protector, which we obtained after about 400 BGA generations, in Figure 18. This optimal impulse absorber has nonlinear beads near its extremities and in its center and sequences of linear beads in its remaining sec-

tions (a few of them are also near the center of the half chain). We show the corresponding force plots and energy profiles (as well as the ones for Hong's container) in Figures 19 and 20, respectively. Observe that the optimized scheme transmits to the wall a maximum force (126) that is about three times smaller than that transmitted by the basic scheme (371). As in the previous examples, the initial pulse is progressively disintegrated, reflected, and transformed into an extended wave within the optimized system (as confirmed also by the density plots of Figure 21). Hong's container instead shows wave reflection only when the incident wave crosses the central section of the system. The time histories of the energy correlation function, depicted in Figure 22, highlight the fact that the optimized system spreads out energy (i.e., thermalizes) on a faster time scale than Hong's container.

3.5.2.   Shock-Type Loading

We also carried out an optimization procedure using a shock-type force profile with constant intensity $F = 1000$ on the first bead (composed of material 1), external to the container, for a time equal to 0.25. We show the corresponding optimal container, which we obtained after about 200 BGA generations, in Figure 18. The force-time plots of this optimal shock absorber and those of the basic Hong configuration, with the shock-type loading, are shown in Figure 23. Observe that the basic protector transmits to the wall a peak force (1300 units) larger than the force of the applied shock, whereas the optimized system is able to reduce the shock's force at the wall to a peak of 770 units (a 23% reduction). Note additionally that the optimal shock absorber is characterized by heavy, linear (material 1) grains near the extremities (see Figure 18). This is likely due to the symmetry constraint discussed above. Figure 23 also shows that the input shock gets weakened when traveling along the optimal
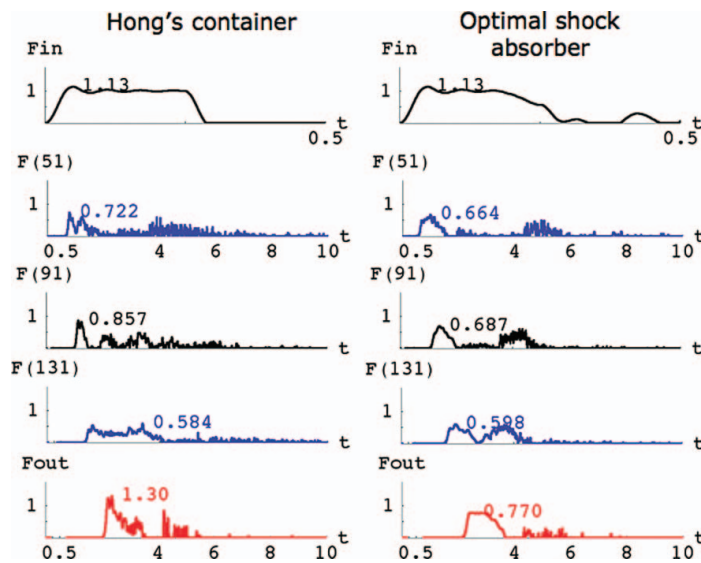

FIG. 23.   Force vs. time plots in a long composite chain subject to shock-type loading (forces values divided by 1000).
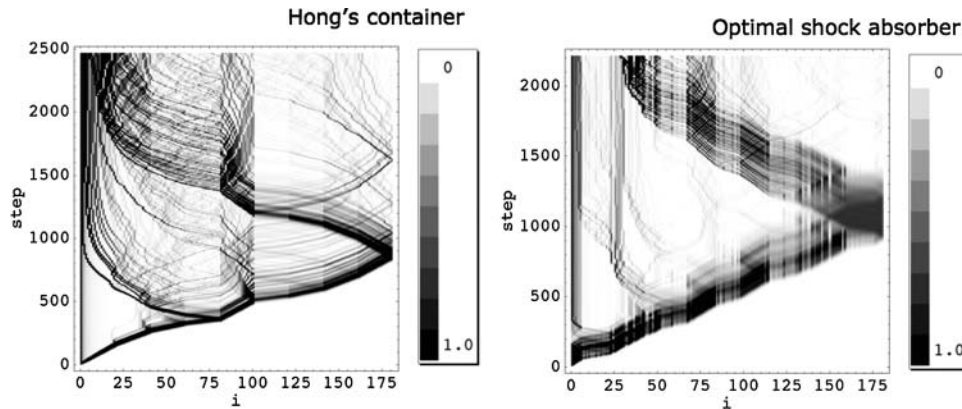
FIG. 24. Density plots of particle energies normalized to unity. Horizontal axes are labeled according to particle site and vertical axes give the time step.

protector. We show the density plots of particle energies in these two systems in Figure 24.

4. CONCLUSIONS

In summary, we used an evolutionary algorithm to investigate the optimal design of composite granular protectors using one-dimensional chains of beads composed of materials of various size, masses, and stiffnesses. Identifying the maximum force F_{out} transmitted from the protector to a "wall" that represents the body to be protected as a fitness function, we optimized the topology (arrangement), size, and material of the beads in the chain in order to minimize F_{out}. We considered several examples that were investigated recently in the literature, including stepped two sonic vacua, tapered chains, decorated chains, and a recent configuration due to Hong.

The optimization procedure, driven by a Breeder Genetic Algorithm, produced (optimally) randomized/disordered systems, along which the incident waves were disintegrated and reflected, exhibiting marked thermalization. Additionally, the solitary pulses traveling to the wall combined to form extended (long-wavelength), small-amplitude waves. In the absence of enforced bilateral symmetry, the optimal configurations had soft/light beads near the wall, hard/heavy beads near the loaded end, and alternating hard/heavy and soft/light beads in the remaining part of the chain. In the presence of bilateral symmetry, we instead obtained an optimal configuration that had light, non-linear beads toward the ends in the case of impulsive loading and one that had heavy, linear beads toward the ends in the case of shock-type loading.

The present research paves the way for many interesting developments, as our approach can be generalized to numerous situations. First, the techniques we employed can be applied to more intricate experimental configurations—including two-dimensional systems, three-dimensional systems, systems composed of ensembles of particles with non-spherical geometries, and even layered materials. Second, one can incorporate additional physical effects, such as dissipation and more complicated contact mechanics. Third, one can generalize the methods themselves by, for example, adopting continuous optimization techniques such as the material distribution method [27] or formulating multiple-scale approaches that involve scale-dependent interaction forces.

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