Effective target arrangement in a deterministic scale-free graph

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(Received 17 March 2010; published 14 July 2010)

We study the random-walk problem on a deterministic scale-free network, in the presence of a set of static, identical targets; due to the strong inhomogeneity of the underlying structure the mean first-passage time (MFPT), meant as a measure of transport efficiency, is expected to depend sensitively on the position of targets. We consider several spatial arrangements for targets and we calculate, mainly rigorously, the related MFPT, where the average is taken over all possible starting points and over all possible paths. For all the cases studied, the MFPT asymptotically scales like $\sim N^{\theta}$, being N the volume of the substrate and θ ranging from $1-\log 2/\log 3$, for central target(s), to 1, for a single peripheral target.

DOI: 10.1103/PhysRevE.82.011118

PACS number(s): 05.40.Fb, 89.75.Hc

I. INTRODUCTION

The importance of first-passage phenomena stems from their fundamental role in stochastic processes that are prompted by a first-passage event; examples range from diffusion-limited growth, to neuron firing, and to the triggering of stock options [1]. These problems can be successfully dealt with by mapping the problem into a random-walk (RW) moving on a proper substrate and in the presence of an absorbing boundary, in such a way that first-passage properties can be expressed in terms of survival probability and mean hitting time. The relation between the first-passage probability and the probability distribution for a RW has been derived in different frameworks (continuous and discrete space and/or time), see e.g., [1-4]. In this context, the mean firstpassage time (MFPT) to a target point, possibly averaged over all possible starting points, has been widely used in order to measure the transport efficiency [5].

The case of regular [2] and fractal lattices [6,7] as substrate has been thoroughly investigated, and recently complex structures, especially (pseudo-)scale-free networks, both random and deterministic, have also attracted a lot of interest [8–10]. The reason is that such networks are able to model some features typical of many real systems and they can exhibit interesting, nonintuitive first-passage properties, whose understanding could pave the way to engineered devices (see, e.g., [11,12]). In particular, several examples were shown where the MFPT to the most connected node displays a sublinear dependence on the size N [8,13].

While previous works dealing with complex structures mainly focused on the case of one single trap placed on a central position, here we consider a special realization of deterministic scale-free structure where we study the MFPT for different arrangements of targets. The problem of the trapping of random walks by randomly distributes traps on regular infinite lattices has been investigated extensively [2,3], and also the case of finite lattices has been studied in presence of perfect multiple traps [14]. On more complex topologies, analytical approaches are mainly based on mean field like approximations [15], such as Rosenstock's one, or on asymptotic expressions [16] predicting stretched expo-

nential decays for the survival probability [17]. In particular, within these approximations, given a concentration $c \ll 1$ of traps, and a high dimensional structure, the trapping time is expected to grow linearly with the system size [18]. Our results, mainly rigorous, allow to go beyond these mean field approximations on a specific quenched topology and they evidence how deeply the position and/or the number of target(s) affect the first-passage properties. More precisely, for the cases analyzed we find that the MFPT scales like $\sim N^{\theta}$ (up to logarithmic corrections), where N is the number of nodes and θ is a proper exponent. Moreover, θ ranges from $\theta_0 = 1 - \log 2 / \log 3 \approx 0.37$ (when one single target is placed on the central node) to 1 (when the target is placed on a peripheral node), which is expected to be the upper bound [19]. Interestingly, the former sublinear scaling is recovered also when a number $N^{\log 2/\log 3}$ of targets is "centrally arranged." On the other hand, such an effective performance is by far not representative of the whole network and the MFPT averaged over the target site, denoted as $\overline{\tau}$, is found to scale linearly with N.

The paper is organized as follows: in Secs. II and III we briefly resume the topological properties of the graph under study, while in Sec. IV we analyze the average time $\bar{\tau}$. Then, we consider special arrangements of traps distinguishing between "central," in Sec. V, and "peripheral," in Sec. VI Finally, Sec. VII contains conclusions and discussions, while in Appendix we have collected the details of analytical calculations.

II. DETERMINISTIC SCALE-FREE GRAPH

The graph under study can be built recursively in such a way that at the *g*th iteration we have the graph of generation *g*, denoted as \mathcal{G}_g (see Fig. 1 for *g*=3); here we briefly describe its main topological properties, for more details we refer to [9,20–22].

Starting from the so-called root constituted by one single node, at the first iteration one adds two more nodes and connects each of them to the root; the resulting chain of length three represents the graph of generation g=1. We call \mathcal{B}_1 the set of sites introduced at the first generation and



FIG. 1. (Color online) Deterministic scale-free graph of generation g=3. According to the labeling used $B_3 = \{14, 15, 17, 18, 23, 24, 26, 27\}$. The main hub has label 1; main hub and minor hubs are depicted with darker nodes. Hubs of generation 2 have label 10 and 19; analogously, nodes 11,12,20,21 are rims of generation 1.

linked to the root. Then, at the second iteration one adds two copies of \mathcal{G}_1 , namely, \mathcal{G}'_1 and \mathcal{G}''_1 ; for each of them one can distinguish the sets \mathcal{B}'_1 and \mathcal{B}''_1 , which make up the set \mathcal{B}_2 , i.e., $\mathcal{B}_2 = \mathcal{B}'_1 \cup \mathcal{B}''_1$; each node in \mathcal{B}_2 is then directly connected to the root. Proceeding analogously, at the *g*th iteration one introduces two replica of the existing graph of generation g - 1 and connects the root with all the sites belonging to \mathcal{B}_g , henceforth called "rims."

This way, at the *g*th generation the total number of nodes is $N_g = 3^g$, the total number of rims is $|\mathcal{B}_g| = 2^g$ and the root turns out to be the main hub with coordination number $2(2^g - 1)$. Apart from the main hub, one can detect "minor hubs" (see Fig. 1), whose degree distribution is given by the power law $P(k) \sim k^{-\gamma}$ with exponent $\gamma = \log 3/\log 2 \approx 1.59$ [20]; the remaining nodes follow an exponential degree distribution: $P(k) \sim (2/3)^k = e^{-\overline{\gamma}k}$, where $\overline{\gamma} = \log(3/2) \approx 0.405$ [21]. We also define the so-called hubs of generation g-n, being $n=0,1,\ldots,g-1$, which correspond to the hubs pertaining to the inner subgraphs (apart the one containing the main hub); analogously we will refer to rims of generation g-n, with $n=0,1,\ldots,g-1$ (see Fig. 1).

Finally, the average degree is $\langle k \rangle_g \equiv \sum kP(k)/N_g = 4[1 - (2/3)^g]$, namely, it approaches 4 as $g \rightarrow \infty$; on the other hand, the second moment $\langle k^2 \rangle$ is divergent [21].

III. VAN HOVE SPHERE AND AVERAGE DISTANCES

For a graph of generation g, being X(k,g) the number of nodes at a distance k from the main hub and, analogously, Y(k,g) the number of nodes at a distance k from an arbitrary rim, the following equations hold:

$$X(k,g) = X(k,g-1) + 2Y(k-1,g-1),$$
(1)

$$Y(k,g) = X(k-1,g-1) + 2Y(k,g-1).$$
 (2)

Notice that X(k,g) represents the cardinality of the borders of the generalized Van Hove sphere with radius k in the chemical distance and centered in the main hub [12]. Passing to the corresponding generating functions, namely,

$$\widetilde{X}(s,g) \equiv \sum_{k=0}^{\infty} X(k,g) s^g$$
 and $\widetilde{Y}(s,g) \equiv \sum_{k=0}^{\infty} Y(k,g) s^g$, we get

$$X(s,g) = X(s,g-1) + 2Y(s,g-1)s,$$
(3)

$$\widetilde{Y}(s,g) = \widetilde{X}(s,g-1)s + 2\widetilde{Y}(s,g-1).$$
(4)

Now, by combining Eqs. (3) and (4) we have

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$$\widetilde{X}(s,g+2) = 3\widetilde{X}(s,g+1) - 2(1-s^2)\widetilde{X}(s,g).$$
 (5)

The solution of this finite difference equation can be obtained by recalling the initial condition

$$X(k,0) = \delta_{k,0} \Longrightarrow \widetilde{X}(s,0) = 1, \tag{6}$$

$$X(k,1) = \delta_{k,0} + 2\delta_{k,1} \Longrightarrow \widetilde{X}(s,1) = 1 + 2s, \tag{7}$$

due to the fact that the root of the graph (generation zero) is just the main hub, while the first generation graph is made up by the main hub and by two rims only. Hence we get [23]

$$X(2k,g) = \left(\frac{3}{2}\right)^g 2^{3k} \sum_{l=0}^{g/2} \frac{1}{3^{2l}} \binom{l}{k} \binom{g}{2l} \left[1 - \frac{g-2l}{3(2l+1)}\right], \quad (8)$$

$$X(2k+1,g) = \left(\frac{3}{2}\right)^{g} \frac{4}{3} 2^{3k} \sum_{l=0}^{g/2} \frac{1}{3^{2l}} \binom{l}{k} \binom{g}{2l+1}, \qquad (9)$$

which provide the cardinality of the van Hove surface of even and odd radius, respectively. Interestingly, these expressions clarify the peculiar topology of the deterministic scalefree graph: due to the nonmonotonicity of X(k,g) and to the large number of surface points at a given distance, quantities obtained from the limit of a sequence of larger and larger graphs could present some anomalies [23]. For instance, X(k,g) is approximately peaked at k=g/2 and it asymptotically decays like a stretched exponential. Peculiar behaviors due to parity effects have been evidenced in a variety of networks and they have been shown to affect first-passage quantities [24,25].

From $\tilde{X}(s,g)$ we can also derive the average distance from the main hub d_{H} . Let us define d_{ij} the average (chemical) distance from *i* to *j*, then we have

$$d_g = \frac{\sum_{i \in N_g} d_{iH}}{N_g},\tag{10}$$

where H is the node representing the main hub. Now, the following relation holds:

$$d_g = \frac{1}{N_g} \frac{\partial}{\partial s} \widetilde{X}(s,g) \bigg|_{s=1}.$$
 (11)

Then, from Eq. (5), we get

$$\frac{\partial}{\partial s}\tilde{X}(s,g+2) = 3\frac{\partial}{\partial s}\tilde{X}(s,g+1) + 4s\tilde{X}(s,g)$$
$$-2(1-s^2)\frac{\partial}{\partial s}\tilde{X}(s,g), \qquad (12)$$

which, for s=1, gives

$$N_{g+2}d_H(g+2) = 3N_{g+1}d_H(g+1) + 4X(1,g).$$
(13)

From Eq. (5), $\tilde{X}(1,g)=3^g=N_g$ follows, as expected, and by plugging $\tilde{X}(1,g)$ into Eq. (13), we obtain the following recursive relation:

$$9d_{g+2} = 9d_{g+1} + 4, \tag{14}$$

whose solution is

$$d_g = \frac{2}{9}(2g+1),$$
 (15)

where we used the initial condition $d_1=2/3$. The average distance to the main hub can also be expressed as a function of the volume, recalling that $g=\log N_g/\log 3$

$$d_g = \frac{2}{9} \left(2 \frac{\log N_g}{\log 3} + 1 \right) \sim \log N_g.$$
(16)

While d_g provides information about the "accessibility" of the main hub itself, the average distance \overline{d}_g calculated over all pairs of nodes, provides a more global information, being tightly connected with dynamics phenomena on graphs such as spreading, random walks and synchronization [7,26,27]. The average distance is defined as

$$\bar{d}_{g} = \frac{\sum_{i,j=1}^{N_{g}} d_{ij}}{N_{g}(N_{g} - 1)}.$$
(17)

The exact expression of \overline{d}_g for the graph under study has been obtained in [22] and is given by

$$\bar{d}_g = \frac{8N_g \log N_g}{9(N_g - 1)\log 3} \sim \log N_g, \tag{18}$$

where the last expression holds in the thermodynamic limit $N_g\!\rightarrow\!\infty.$

The consistency between the leading behaviors of d_g and \overline{d}_g , see Eqs. (16) and (18), suggests that, despite its centrality, the main hub is not, in the average, spatially closer than any other site. In fact, the maximum distance among nodes grows logarithmically with the volume, that is, $\max_{i,j=1,\ldots,N_g} d_{ij} = 2g$, and the graph is generally rather compact.

IV. TOTAL MEAN FIRST PASSAGE TIME

The average distance \overline{d}_g discussed in the previous section allows to determine an upper bound for the total mean first passage time $\overline{\tau}_g$ defined as

$$\bar{\tau}_g = \mathbb{E}(t_{ij}) \equiv \frac{1}{N_g(N_g - 1)} \sum_{i \neq j}^{N_g} \sum_{j=1}^{N_g} t_{ij},$$
(19)

where t_{ij} is the mean time taken by a simple RW on \mathcal{G}_g to first reach the site *j* starting from *i*, and the average \mathbb{E} is taken over all possible couples. As shown in [19,28], $\overline{\tau}_g$ is subject to the following, rather strict, constraints:

$$\frac{N_g}{2} \le \overline{\tau}_g \le \frac{N_g \langle k \rangle \overline{d}_g}{2} \sim N_g \log N_g, \tag{20}$$

where we used Eq. (18) for \overline{d}_{g} [22].

Such an average is characteristic of a network, in the sense that it provides an overall measure of the transport efficiency on that network, but, on the other hand, it conceals the peculiarities of the graph itself and it does not allow to see whether there exists any target position which results to be particularly effective or ineffective.

In order to deepen this point we start by considering the set $\{t_{ij}\}_{i,j=1,...,N_g}$: for an arbitrary connected couple (i, j), the analytical calculation of the mean time t_{ij} can be rather lengthy and awkward, while the whole set $\{t_{ij}\}_{i,j=1,...,N_g}$ can be efficiently calculated numerically by exploiting a method based on the so-called pseudoinverse Laplacian $\tilde{\mathbf{L}}^{\dagger}$ [11,29]. More precisely, the latter is generically defined as $\tilde{\mathbf{L}}^{\dagger} = (\mathbf{L} - \mathbf{e}\mathbf{e}^T/N)^{-1} + \mathbf{e}\mathbf{e}^T/N$, where \mathbf{L} is the Laplacian matrix [12] associated to the arbitrary connected graph of size N and \mathbf{e} denotes the unity-vector; the mean times can then be written in terms of matrix elements: $t_{ij} = \sum_{k=1}^{N} (L_{ik}^{\dagger} - L_{ij}^{\dagger} - L_{jk}^{\dagger} + L_{jj}^{\dagger}) z_k$, where z_k is the number of nearest neighbors of node k. In this way we can directly derive $\overline{\tau}_g$ from Eq. (19) and the variance of the set $\{t_{ij}\}_{i,j=1,...,N_g}$ as $\sigma^2 = \mathbb{E}(t_{ij} - \overline{\tau}_g)^2$.

The intrinsic symmetry of the graph yields degenerate values for t_{ij} which have been omitted in the plot of Fig. 2. From the plot one can notice that t_{ij} spans a wide range $\sim \mathcal{O}(N_g)$ and the variance σ^2 of the set $\{t_{ij}\}$ (see the inset of Fig. 3) is shown to grow asymptotically as the square of the graph volume. The large variability of t_{ij} mirrors the strong inhomogeneity of the graph under study; this is a further confirmation that the efficiency of first-passage processes is sensitively affected by the target position. Finally, as shown in Fig. 3 (main figure) the asymptotic behavior of $\overline{\tau}_g$ is consistent with the linear law $\overline{\tau}_g \sim 3N_g$, in agreement with Eq. (20).

V. TARGET ON CENTRAL NODES

As shown in [9], for a simple RW on \mathcal{G}_g , the mean time to first reach the main hub starting from any of the rims in \mathcal{B}_g is

$$R_g = \frac{8}{3} \left(\frac{3}{2}\right)^g - 3, \qquad (21)$$

while the mean time to first reach any rim starting from the main hub is

$$H_g = \frac{4}{3} \left(\frac{3}{2}\right)^g - 1, \qquad (22)$$

where the mean is taken over all possible paths. Such results were used to calculate the mean time to first reach the main hub $\tau_g(H)$ (where the mean is taken over all possible paths and over all possible starting points), which reads as



FIG. 2. (Color online) Mean times t_{ij} divided by the volume N_g for graphs of generations g = 4, 5, 6, as indicated by the legend; only nondegenerate data have been depicted. Notice that data are plotted in ascending order and versus a proper normalized interval accounting for the effective (size dependent) number of nondegenerate values.

We now calculate the mean hitting time $\tau_g(\mathcal{B}_g)$ at any rim, averaged over all possible starting points $\notin \mathcal{B}_g$ and over all possible paths,

$$T_g(\mathcal{B}_g) \equiv \frac{1}{N_g - |\mathcal{B}_g|} \sum_{i \notin \mathcal{B}_g} t_{i\tilde{b}}, \qquad (24)$$

where $t_{i\tilde{b}} = \min_{b \in B_g} t_{ib}$. By properly combining H_g and R_g we can write the following recursive equation:

$$\tau_{g}(\mathcal{B}_{g}) = \frac{1}{3^{g} - 2^{g}} [3^{g-1}H_{g} + (3^{g-1} - 1)\tau_{g-1} + 2\tau_{g-1}(\mathcal{B}_{g-1})(3^{g-1} - 2^{g-1})].$$
(25)

In fact, if the RW starts from a site belonging to the subgraph \mathcal{G}_{g-1} containing the main hub, then, in order to arrive to rims, it must pass through the main hub itself (in a mean time $\tau_{g-1}(H)$) from which rims are first reached in a time H_g . Conversely, if the starting point is in $\mathcal{G}'_{g-1} \cup \mathcal{G}''_{g-1}$, the MFPT to rims is just $\tau_{g-1}(\mathcal{B}_{g-1})$. The recursive relation can be solved to obtain

$$\tau_g(\mathcal{B}_g) = \frac{3^g}{9(3^g - 2^g)} \left[20 \left(\frac{3}{2}\right)^g - 8g - 19 \right] \sim \left(\frac{3}{2}\right)^g.$$
(26)

Interestingly, the leading behavior is the same as $\tau_g(H)$, even if in this case the number of targets is $2^g \sim N_g^{1/\gamma}$, moreover, $\tau_g(\mathcal{B}_g)/\tau_g(H) \sim 5/8$. It is rather intuitive at this point that when targets are placed on the main hub *and* on rims the leading behavior of the hitting time is conserved. Indeed, one has

$$\tau_{g}(H \cup \mathcal{B}_{g}) = \frac{1}{3^{g} - 2^{g} - 1} [(3^{g-1} - 1)\tau_{g-1} + \tau_{g-1}(\mathcal{B}_{g-1})(2 \cdot 3^{g-1} - 2^{g})].$$
(27)

FIG. 3. (Color online) Total mean first passage time $\overline{\tau}_g$ (main figure) and related variance σ^2 (inset) as a function of the graph volume N_g . Data points have been fitted with the power law $\sim N_g$ and $\sim N_g^2$, respectively.

10

10

10

 ${{10}^2} {N_g}$

10

10³

10

10³

10⁴

being $1-1/\gamma = \theta_0 = 1 - \log 2/\log 3 \approx 0.37$. Notice that the

asymptotic behavior of R_g , H_g and $\tau_g(H)$ is the same, namely, $(3/2)^g$. The exponent θ_0 is very small if compared with the result found for $\overline{\tau}_g$ and also with similar results

found in the literature [8,9,13,30], making the main hub a particularly effective node where to place a target. Interestingly, the same exponent θ_0 was also found in [31] for a hierarchical network featuring indeed topological properties

similar to those of the graph considered here.

10

10

ພື 10

10

10⁰

0[°]10

10¹

We finally consider the case of one single target placed on a rim, i.e., on a site $b \in \mathcal{B}_g$; with respect to the original graph,



FIG. 4. (Color online) Mean hitting times at the main hub $\tau_g(H)$, at rims $\tau_g(\mathcal{B}_g)$ and at one rim $\tau_g(b)$, as a function of the number of nodes N_g for a simple RW moving on \mathcal{G}_g . Data from numerical calculations coincide with those calculated analytically [for the former two cases see Eqs. (23) and (26)] and are compared with the leading behaviors $\sim N_g^{1-1/\gamma}$ and $\sim N_g/\log N_g$ (continuous line).

this kind of perturbation breaks the symmetry of the structure itself so that the exact analytical evaluation of the mean time is rather lengthy and the calculation has been performed numerically exploiting the following expression [30]

$$\tau_g(b) = \frac{1}{N_g - 1} \sum_{\substack{i=1\\i \neq b}}^{N_g} t_{ib} = \frac{1}{N_g - 1} \sum_{\substack{i=1\\i \neq b}}^{N_g} \sum_{\substack{j=1\\i \neq b}}^{N_g} (-\mathbf{L}^{-1})_{ij},$$

where L is the Laplacian matrix of the underlying structure. In the range considered, i.e., $g \in [0,6]$, data points can be fitted by the following:

$$\tau_g(b) \sim \frac{N_g}{\log N_g},\tag{28}$$

which is consistent with the analytical result found in [19] (although the average here is taken over the flat distribution). The MFPT's $\tau_g(H)$, $\tau_g(\mathcal{B}_g)$, and $\tau_g(b)$ are shown and compared in Fig. 4

VI. TARGET ON PERIPHERAL NODES

After having analyzed particularly effective arrangements for targets, i.e., main hub and rims, we now want to focus on trapping at the farthest nodes from the main hub which are expected to be rather difficult to reach.

We notice that in a graph of generation g, the radius of the outmost van Hove surface centered on the main hub, i.e., the largest chemical distance achievable from the main hub, equals g and its cardinality is X(g,g)=g if g is even, otherwise X(g,g)=g+1; such farthest nodes correspond to hubs (g even) or rims (g odd), of subgraphs of generation 1. We call \mathcal{F}_g the set of nodes at a chemical distance g from the main hub so that $|\mathcal{F}_g|=X(g,g)$.

We will first consider the case of \mathcal{F}_g working as a boundary, that is each node in \mathcal{F}_g is occupied by a target, hence, exploiting the intrinsic symmetry of the arrangement, the problem will be solved analytically. Later we will focus on the case of one single target fixed on a node belonging to \mathcal{F}_g and we will solve the problem numerically.

A. Multitarget

The basic step to calculate the hitting time $\tau_g(\mathcal{F}_g)$ when each node in \mathcal{F}_g is occupied by a target, is to obtain H_{g-n} and R_{g-n} , which represent, respectively, the MFPT from the hub of a subgraph of generation g-n to any rim of the same subgraph, and the MFPT from any rim in a subgraph of generation g-n to the pertaining hub; for example, referring to Fig. 1, R_{3-1} is the mean time to go from node 14 to 10, while H_{3-2} is the mean time to reach 11 starting from 10 (and equivalently for similar nodes). The detailed calculations are reported in the Appendix, while here we just summarize the main passages and results.

We first consider the MFPT from the main hub to \mathcal{F}_g ; starting from the main hub the RW has first to reach the rims in a time H_g [see Eq. (22)], from \mathcal{B}_g the RW has to reach the hub of the subgraph of generation g-1 and this is found to be

$$R_{g-1} = \frac{28}{3} \left(\frac{3}{2}\right)^g - 3, \tag{29}$$

which exhibits the same asymptotic behavior of H_g and of R_g [see Eqs. (21) and (22)] [9]. Then, the RW has to "bounce" from rim to hub and from hub to rims of inner and inner subgraphs, the pertaining MFPT's read as

$$R_{g-n} = 4 \times 2^{n/2} \left[\left(\frac{3}{2}\right)^g - 1 \right] + 1 - \frac{8}{9} \left(\frac{3}{2}\right)^{g-n}$$
(30)

and

$$H_{g-n} = 4 \times 2^{n/2} \left[\left(\frac{3}{2} \right)^g - 1 \right] + 3 - \frac{8}{3} \left(\frac{3}{2} \right)^{g-n}.$$
 (31)

We notice that R_{g-n} and H_{g-n} [Eqs. (30) and (31)] display the same asymptotic behavior, namely $(3/2)^{g}2^{n/2}$, which, interestingly, is the same behavior found for $\tau_{g}(H)$ but "corrected" by a factor which depends on *n* as $2^{n/2}$,

$$R_{g-n} \sim 2^{n/2} R_g,$$
 (32)

$$H_{g-n} \sim 2^{n/2} H_g.$$
 (33)

Some remarks are in order here. Although, for a given subgraph, the distance between the hub and a rim is the same and equal to 1, the mean time taken by the RW depends significantly on how "deep" the subgraph is (i.e., how small g-n is). Such a result is somehow counterintuitive because the probability to go directly from rim to hub [1/(g-n+1)]and from hub to rim $[2^{g-n-1}/(2^{g-n}-1)]$ grows with *n*. Actually, what really matters is that if the right way is missed at the first step then it is more and more difficult for the RW to retrieve it as *n* gets larger. The factor $2^{n/2}$ is a clear indication of the fact that targets placed on inner nodes are in general rather difficult to reach.



FIG. 5. (Color online) Mean hitting times at peripheral nodes $\tau_g(\mathcal{F}_g)$ (\bullet) and $\tau_g(f)$ (\blacktriangle), as a function of the number of nodes N_g for a simple RW moving on \mathcal{G}_g . Data from numerical calculations coincide with those calculated analytically from Eq. (26) and are compared with the leading behavior $\sim N_g^{1-1/(2\gamma)}$ and $\sim N_g$ (continuous line), according to Eqs. (38) and (39), respectively.

By merging the previous results one can find the average time t_{g-n}^H to reach a node in \mathcal{F}_g , starting from the hub of any subgraph of generation g-n

$$t_{g-n}^{H}(\mathcal{F}_{g}) = \sum_{l=g-n}^{g-2} \left(R_{g-l} + H_{g-l} \right)$$
(34)

or from any rim of generation g-n

$$t_{g-n}^{R}(\mathcal{F}_{g}) = \sum_{l=g-n}^{g-2} R_{g-l} + \sum_{l=g-n-1}^{g-2} H_{g-l}.$$
 (35)

These expressions are used in Appendix to get the average time $\tau_g(\mathcal{F}_g)$ to reach an arbitrary node $f \in \mathcal{F}_g$, where the average is meant over all possible starting points $\notin \mathcal{F}_g$ and over all possible paths, namely,

$$\tau_g(\mathcal{F}_g) \equiv \frac{1}{N_g - |\mathcal{F}_g|} \sum_{i \notin \mathcal{F}_g} t_{if}.$$
 (36)

Hence, we finally get

$$\tau_{g}(\mathcal{F}_{g}) = \frac{2^{g+1}}{45(3^{g} - 2^{g/2})} \left[4 \left(\frac{3}{2}\right)^{2g} (45 \times 2^{g/2} - 97) - \left(\frac{3}{2}\right)^{g} (45 \times 2^{g/2+2} - 25g - 527) - 144 \right]$$
$$\sim \left(\frac{3}{2}\right)^{g} 2^{g/2} \sim N_{g}^{1-1/2\gamma}, \qquad (37)$$

where the last line contains the asymptotic behavior valid in the limit $N_g \rightarrow \infty$ (see Fig. 5). Although the number of targets is $\sim \log N_g$, the mean time to first reach any of them grows much faster than the mean time to reach the main hub $\tau_g(H) \sim (3/2)^g$, but still slower than the average hitting time $\overline{\tau}_g \sim 3^g$. Indeed, recalling $1 - \frac{1}{2\gamma} = 1 - \frac{1}{2\log 3} \approx 0.6845$, we have that

$$\tau_g(\mathcal{F}_g) \sim \overline{\tau}_g^{1-1/2\gamma} \sim \tau_g(H)^{1+1/2(\gamma-1)},\tag{38}$$

where $1 + 1/(2\gamma - 2) \approx 1.85$.

B. Single target

We now turn to the case of one single target placed on a node $f \in \mathcal{F}_g$, which has been investigated numerically according to

$$\tau_g(f) = \frac{1}{N_g - 1} \sum_{\substack{i=1\\i \neq f}}^{N_g} t_{if} = \frac{1}{N_g - 1} \sum_{\substack{i=1\\i \neq f}}^{N_g} \sum_{\substack{j=1\\i \neq f}}^{N_g} (-\mathbf{L}^{-1})_{ij}.$$

Results found for several generations are shown in Fig. 5. By means of fitting procedures we find that when the size is large, the mean hitting time $\tau_g(f)$ scales like

$$\tau_g(f) \sim N_g^{\theta_f},\tag{39}$$

where the exponent θ_f is consistent with the value 1, which, following [19], suggests a transient behavior for f.

VII. CONCLUSIONS

In this paper we studied the mean first-passage problem for a random walk on a deterministic scale-free structure, focusing the attention on different arrangement of targets. First of all, we highlighted that, due to the high inhomogeneity of the structure, the mean first-passage time (MFPT) t_{ij} from a node *i* to *j* sensitively depends on the couple (i,j)chosen, spanning a range $\sim \mathcal{O}(N_g)$, being N_g the size of the structure.

The case of a single target placed on the main hub turns out to be strategic as it gives rise to a slow sublinear scaling $\sim N_g^{\theta_0}$, being $\theta_0 = 1 - 1/\gamma = 1 - \log 2/\log 3 \approx 0.37$. Interestingly, the same scaling is obtained in the case that the whole set of rims \mathcal{B}_g , with cardinality $\sim \mathcal{O}(N_g^{1/\gamma})$, is assumed as absorbing boundary. On the other hand, when only one single node $b \in \mathcal{B}_g$ works as a target we get a faster growth $\sim N_g/\log N_g$, where the logarithmic correction accounts for the large coordination number of rims (see [19]).

When peripheral positions for traps are considered, sublinear scaling are achievable provided that the number of targets is large enough; in particular, when the absorbing boundary is given by the whole set of peripheral nodes \mathcal{F}_g with cardinality $\log N_g/\log 3$ we have that the mean hitting time scale like $\sim N_g^{1-1/(2\gamma)}$. Conversely, when one single node $f \in \mathcal{F}_g$ works as a target our fits suggest a linear growth $\sim N_g$.

Therefore, the transport efficiency sensitively depends on the number and on the position of targets, moreover, the existence of very "close" couples, such as main hub and rims, is evened off by rather "distant" couples, giving rise to a mean hitting time $\overline{\tau}_g$, averaged over all possible pairs, which scales linearly with N_g . In the highly inhomogeneous topology of the network, the average behavior arises from very peculiar and different contributions. A key point in this framework appears to be the rather strange behavior of the Van Hove spheres, which makes the exploration of larger portions of the graph through a sequence a subtle task. In particular, the behavior of trapping times evidenced by our results is consistent with the transience of the nodes [19]. Interestingly, the general properties of the graph, which can be defined strictly only in the thermodynamic limit, could present some peculiar behavior on these highly inhomogeneous structures. A rigorous investigation of a possible connection between the rate of growth of the van Hove surface centered at a node i and the features of the mean-first passage times to i as well as (local) recurrence/transience properties are still open problems and they would provide interesting insights in the study of inhomogeneous topologies [23].

APPENDIX: MEAN FPT ON \mathcal{F}_{g}

In this appendix we report the detailed calculation of the mean time taken by a simple RW on a graph \mathcal{G}_g and started from the hub of a subgraph of generation *n*, to first reach any rim in the same subgraph, which is denoted as H_n , and the MFPT from any rim in a subgraph of generation *n* to the pertaining hub, denoted as R_n . Such times will then be merged in order to obtain the mean time $\tau_g(\mathcal{F}_g)$ (averaged over all possible starting points and over all possible paths) to first reach any node belonging to the set \mathcal{F}_g and meant to be occupied by a target.

Let us start from the highest generation, i.e., n=g. As for H_g , this has already been calculated in [9] [see Eq. (22)]. Now, from rims of generation g, in order to reach \mathcal{F}_g , we first need to pass through the hub of generation g-1 and we can write

$$R_{g-1} = \frac{1}{g} \left[\left(t_g^H + 1 + R_g \right) + 1 + \sum_{l=1}^{g-2} \left(1 + t_l^H + R_g \right) \right].$$
(A1)

In fact, starting from any rim of generation g, a RW has g options: it goes to the main hub and then returns to rims in a time H_g , either it goes directly to the hub of generation g - 1, either it reaches a hub of generation g-l from which it return to rims in an average time H_l . By plugging Eqs. (22) and (21) in Eq. (A1) we get

$$R_{g-1} = \frac{28}{3} \left(\frac{3}{2}\right)^g - 3.$$
 (A2)

Now, from the hub of generation g-1 the RW needs to reach the pertaining rims and this is first accomplished in a mean time given by

$$H_{g-1} = \frac{1}{2(2^{g-1} - 1)} \left[2^{(g-1)-1} + \sum_{l=1}^{(g-1)-2} (1 + R_l + H_{g-1}) + (1 + R_{g-1} + H_{g-1}) \right],$$
(A3)

in fact, the RW can jump directly in any of the $2(2^{(g-1)-1})$ rims of generation (g-1)-1, either it can reach rims of g – 1 from which it needs to return to the hub of generation

(g-1) in a time R_{g-1} [Eq. (A1)], either it can jump to a rim of the subgraph of generation g-l from which it can return to the hub in R_l .

As can be deduced from the previous equations, H_{g-n} and R_{g-n} are deeply connected with each other, such an interplay allows to build up a system of recursive equations. We can in fact write the following general equations

$$H_{g-n} = \frac{1}{2(2^{g-n} - 1)} \left[2^{(g-n)-1} + \sum_{l=1}^{(g-n-2)} (1 + R_l + H_{g-n}) + (1 + R_{g-n} + H_{g-n}) \right].$$
(A4)

and

$$R_{g-n} = \frac{1}{g-n} \left[(H_{g-n} + 1 + R_{g-n}) + 1 + \sum_{l=1}^{g-2} (1 + H_l + R_{g-n}) \right].$$
(A5)

Now, by plugging in Eqs. (A5) and (A4) the expressions Eqs. (21) and (22), with some algebra we get

$$R_{g-n} = H_{g-n+1} + 2 + 4 \left[\left(\frac{3}{2}\right)^{g-n} - 1 \right],$$
 (A6)

$$H_{g-n} = 2R_{g-n+1} + 1 + \frac{8}{9} \left(\frac{3}{2}\right)^{g-n}.$$
 (A7)

By properly combining these equations we have

$$R_{g-n} = R_{g-n+2} - 1 + \frac{28}{9} \left(\frac{3}{2}\right)^{g-n}$$
(A8)

and using Eq. (21) we have

$$R_{g-n} = 4\left(\frac{3}{2}\right)^g 2^{n/2} - (2^{4+n/2} - 1) - \frac{8}{9}\left(\frac{3}{2}\right)^{g-n}.$$
 (A9)

Repeating the same calculations for H_{g-n} we obtain similar results, namely:

$$H_{g-n} = 4\left(\frac{3}{2}\right)^g 2^{n/2} - (2^{4+n/2} - 3) - \frac{8}{3}\left(\frac{3}{2}\right)^{g-n}.$$
 (A10)

Following analogous operations for *n* odd, we reach the same result provided that *n* is replaced by n+1. We notice that, for a given generation *g*, the calculation of $\tau_g(\mathcal{F})$ involves either R_{g-2n} or R_{g-2n+1} occur according to whether *g* is even or odd. Hence, in the following, without loss of generality, we focus on the case of g-n.

From H_{g-n} and R_{g-n} it is now possible to derive a closed form expression for the trapping time at \mathcal{F}_g . We now use the partial results found so far to finally calculate $\tau_g(\mathcal{F}_g)$. First of all, let us calculate the first passage time $t_g^H(\mathcal{F}_g)$ on any site in \mathcal{F}_g for a RW started from the main hub; when g is even, \mathcal{F}_g is made up of nodes which correspond to hubs of generation 1, vice versa, when g is odd \mathcal{F}_g is made up of nodes which correspond to rims of generation 1. Let us now focus on g even; we then reach the following expression:

$$t_{g}^{H}(\mathcal{F}_{g}) = \sum_{l=0}^{g-2} R_{g-l} + H_{g-l}, \qquad (A11)$$

which can be straightforwardly generalized to the case of a RW started on a hub of generation g-n

$$t_{g-n}^{H}(\mathcal{F}_{g}) = \sum_{l=g-n}^{g-2} (R_{g-l} + H_{g-l}), \qquad (A12)$$

from which we get

$$t_{g-n}^{H}(\mathcal{F}_{g}) = 8(2^{g/2} - 2^{n/2}) \left[\left(\frac{3}{2}\right)^{g} - 1 \right] - \frac{32}{5} \left[\left(\frac{3}{2}\right)^{g-n} - 1 \right] + 2(g-n).$$
(A13)

Let us now calculate the mean time $t_g^R(\mathcal{F}_g)$ taken by a RW started from a rim of generation g-n to first reach any node in \mathcal{F}_g ; the following relation holds

$$t_{g-n}^{R}(\mathcal{F}_{g})\sum_{l=g-n}^{g-2}R_{g-l} + \sum_{l=g-n-1}^{g-2}H_{g-l}, \qquad (A14)$$

which is just a truncated version of sum [Eq. (34)], so that we get

$$t_{g-n}^{R}(\mathcal{F}_{g}) = 4(2 \times 2^{g/2} - 3 \times 2^{n/2}) \left[\left(\frac{3}{2}\right)^{g} - 1 \right] - \frac{56}{15} \left(\frac{3}{2}\right)^{g-n} + 2(g-n) + \frac{17}{5}.$$
 (A15)

In order to simplify the notation we introduce T_n which represents the mean time taken by the RW to first reach a site f in \mathcal{F}_{ρ} and started from a node at a chemical distance n from

f. Since we assumed *g* to be even, when *n* is even (odd) the starting point is a hub (rim). By exploiting the intrinsic symmetry of the graph, T_n can be written as the sum of two contributes, according to whether the starting node belongs to the subgraph \mathcal{G}_{g-1} (and therefore the RW has to pass through the main hub before reaching the target) or to $\mathcal{G}'_{g-1} \cup \mathcal{G}''_{g-1}$ (and therefore the RW has to pass through \mathcal{B}_g); for subgraphs of smaller generation it holds analogously. The first contribute reads as

$$\sum_{\substack{n=1\\n \text{ odd}}}^{g-1} 2^{n-1/2} (3^{g-n} - 1) (\tau_{g-n}^H + T_{g-n+1}) + T_{g-n+1}, \quad (A16)$$

in fact the RW has first to reach the main hub of the pertaining subgraph in a mean time τ_{g-n} , from which it reaches \mathcal{F}_g in a time T_{g-n+1} . The factor $2^{n-1/2}$ accounts for the number of subgraphs of generation g-n present. The second contribute reads as

$$4 \sum_{\substack{n=2\\n \text{ even}}} 2^{n-2/2} (3^{g-n} - 2^{g-n}) (\tau_{g-n}^R + T_{g-n+1}) + 2^{g-n} T_{g-n+1},$$
(A17)

where we considered RWs starting from nodes such that in order to reach \mathcal{F}_g has first to reach the rims of the pertaining subgraphs or \mathcal{B}_g itself. We sum together the previous expressions to obtain

$$\tau_g(\mathcal{F}_g) = \frac{2^{g+1}}{45(3^g - 2^{g/2})} \left[4\left(\frac{3}{2}\right)^{2g} (45 \times 2^{g/2} - 97) - \left(\frac{3}{2}\right)^g (45 \times 2^{g/2+2} - 25g - 527) - 144 \right],$$

for g odd analogous results are expected.

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