Random Walks Along the Streets and Canals in Compact Cities: Spectral analysis, Dynamical Modularity, Information, and Statistical Mechanics

D. Volchenkov * and Ph. Blanchard

BiBoS, University Bielefeld, Postfach 100131, D-33501, Bielefeld, Germany
Phone: +49 (0)521 / 106-2972
Fax: +49 (0)521 / 106-6455
E-Mail: VOLCHENK@Physik.Uni-Bielefeld.DE

March 31, 2022

Abstract

Different models of random walks on the dual graphs of compact urban structures are considered. Analysis of access times between streets helps to detect the city modularity. The statistical mechanics approach to the ensembles of lazy random walkers is developed. The complexity of city modularity can be measured by an information-like parameter which plays the role of an individual fingerprint of Genius loci. Global structural properties of a city can be characterized by the thermodynamical parameters calculated in the random walks problem.

PACS codes: 89.75.Fb, 89.75.-k, 89.90.+n

Keywords: Random walks, complex networks, city streets and canals

1 Introduction to city networks studies

Studies of urban networks have a long history. Many researches have been devoted to the optimizations of transportation routes and power grids, to the predictions of traffic flows between the highly populated city districts, to the
investigations of habits and artefact exchanges between distanced settlements in historical eras. In most of them, relations between different components of urban structure are often measured along streets, the routes between junctions which form the nodes of an equivalent planar graph. The graph-theoretic principles had been applied in [1] in order to measure the hierarchy in regional central place systems, in [2] to the measurement of transportation networks. The use of graph-theoretic view and network analysis of spatial systems in geographic science had been reviewed in [3]. It is interesting to mention that graphs have been widely used to represent the connectivity between offices in buildings [4] and to classify various building types in [5].

In all these studies, the traffic end points and junctions had been treated as nodes, and the routes had been considered as edges of some planar graphs. Being embedded onto the geographical and economical landscapes, these planar graphs bare their multiple fingerprints. Among the main factors featuring them are the high costs for the maintenance of long-range connections and the scarce availability of physical space. Spatial networks differ from other complex networks and call for the alternative approaches to investigate them [6].

While studying the motifs and cycles in the complex networks, a comparative analysis of the original graph and of its randomized version is used. If a motif is statistically significant, it appears in the real network much frequently than in the randomized versions of the graph [7]. However, in the planar city street patterns, its randomized version is not of significance since, first, it is surely a non-planar graph due to the randomness of edge crossings and, second, the long-range connections which inevitably present in the random graphs in abundance are extremely costly in the real cities [6]. In [8], it has been proposed to compare the city street patterns with the grid-like structures that is indeed useful essentially for the regular urban development.

It had been formulated in the classical essay of T. Harold Hughes [9] that the accessibility of important city objects for the vehicle traffic and pedestrians is always the chief factor in regulating the growth and expansion of the city. A broad, simple scheme of main traffic lines gives a sense of connectedness and unity to the various parts of the city and links up country and town. In [10, 11], a significant correlation between the topological accessibility of streets and their popularity, micro-criminality, micro-economic vitality and social liveability had been established. In the traditional representation of space syntax based on relations between streets through their junctions, the accessibility or distance is associated with points or junctions.

We would like to mention that the issues of global connectivity of finite graphs and accessibility of their nodes are the classical fields of researches in graph theory. They are studied by means of certain dynamical processes defined on the graphs. In particular, in order to reach "obscure" parts of large sets and estimate the probable access times to them, random walks are often used [12]. There are a number of other processes that can be defined on a graph describing various types of diffusion of a large number of random walkers moving on the network at discrete time steps [14]. In all such processes we have deal with discrete time Markov chains studied in probability theory. Markov chains have
the property that their time evolution behavior depends only upon their current state and the state transition properties of the model.

At each time step every walker moves from its current node to one of the neighboring nodes along a randomly selected link. The metric distance between the nodes are of no matter for such a discrete time diffusion process and then the focus of study is naturally shifted from the original problem of traditional space syntax to the dual one based on relations between the streets which themselves are treated as nodes. The distance between two streets, in such a representation, is a distance in the graph-theoretic sense. The dual graphs of a geographic network comparable in its structure with other complex networks are irrelevant to neither distance, nor the physical space constraints. The relations between the traditional space syntax representation based on the relations between streets through their junctions and the dual representation that is a morphological representation of relations between junctions through their streets was studied in details in [13]. The dual city graphs had been developed and studied within the concept of space syntax [10]. The key characteristics in space syntax is that precedence is given to linear features such as streets in contrast to fixed points which approximate locations [11]. The dual city graphs have been discussed recently in [13]. In [10], they have been called the information city network.

In [17], while identifying a street over a plurality of routes on the city map, the "named-street" approach had been used, in which two different arcs of the original street network are assigned to the same street ID provided they have the same street name. The main problem of the approach is that the meaning of a street name could vary from one district or quarter to another even within the same city. For instance, the streets in Manhattan do not meet, in general, the continuity principle rather playing the role of local geographical coordinates. Being interested in the statistics of random walks, we generalized the approach used in [17] in a way to account the possible discontinuities of streets. Namely, we assign an individual street ID code to each continuous part of it even if all of them share the same street name. The dual graph is constructed by mapping edges coded with the same street ID into nodes of the dual graph, and intersections among each pair of edges in the original graph - into edges connecting the corresponding nodes of the dual graph like it has been done in [17].

In [15], the Intersection Continuity principle (ICN) different from our identification approach has been used: two edges forming the largest convex angle in a crossroad on the city map are assigned the highest continuity and therefore are coupled together acquiring the same street ID. The main problem with the ICN principle is that the streets crossing under the convex angles would artificially exchange their identifiers that is not crucial for the study of degree statistics, but makes it difficult to interpret the results on random walks and detect the dynamical modularity of the city.

It is also important to mention that the number of street IDs identified within the ICN principle usually exceeds substantially the actual number of street names in a city. In [6] [15] [25] [26] [27], the degree statistics and various centrality measures for the data set of the square mile samples of different world
cities had been investigated. However, the decision on which a square mile would provide an adequate representation of a city is always questionable.

In this paper, we use an alternative strategy investigating the spectral properties of random walks defined on the dual graphs of compact city patterns bounded by the natural geographical limitations. The reason we consider the compact urban domains is twofold. First, it allows us to avoid the problem of a "square mile" and, second, the compact urban domains had been usually developed "at once", in accordance to certain architectural principles, their partial redevelopment had been occasional and rear, so that they can be considered typical.

We have studied the Markov chains defined on the six undirected dual graphs corresponding to the different street and canal urban structures. Two of them are situated on islands: Manhattan (with an almost regular grid-like city plan) and the network of Venice canals (imprinting the joined effect of natural, political, and economical factors acted on the network during many centuries). We have also considered two cities founded shortly after Crusades and developed within the medieval fortresses: Rothenburg ob der Tauber (the medieval Bavarian city preserving its original structure from the XIII-th century) and the Bielefeld downtown (Alstadt Bielefeld) composed of two different parts: the old one founded in the XIII-th century and the modern part subjected to the partial urban redevelopment at the end of XIX-th century. To supplement the study, we have investigated the canal network of the city of Amsterdam. Although it is not actually isolated from the national canal network, it is binding to the delta of Amstel river forming a dense canal web showing a high degree of radial symmetry. The scarce of physical space is among the most important factors determining the structure of compact urban patterns. The general information on the dual graphs of compact urban street and canal patterns that we have studied is given in Tab. 1. The spectral properties of finite Markov chains defined on these dual graphs are compared with those of a model example: a hypothetical village extended along one principal street and composed of $N$ blind passes branching off it.

A square mile of the New York city grid and a square mile pattern of Venice streets array have been discussed recently in [15, 25, 26, 27], however, up to our knowledge, the canal patterns have never been subjected to the network analysis. The navigation efficiency in the Manhattan streets has been studied in [16].

It is worth to mention the importance of implemented street identification principle for the conclusion on the degree statistics of dual city graphs. The comparative investigations of different street patterns performed in [6, 15] implementing the ICN principle reveal scale-free degree distributions for the vertices of dual graphs. However, in [17] it had been reported that under the street-name approach the dual graphs exhibit the small-world character, but scale-free degree statistics can hardly be recognized. The results on the probability degree statistics for the dual graphs of compact urban patterns analyzed in accordance to the street identification principle that we have described above are compatible with that of [17]. Compact city patterns do not provide us with
sufficient data to conclude on the universality of degree statistics. It is remarkable that the probability degree distributions for the dual graphs corresponding to the compact city patterns are broad and have a clearly expressed maximum and a long right tail. The presence of a noticeable maximum in the probability degree distributions indicates that the structures of compact urban patterns are usually close to a regular one and that there is the most probable number of junctions an average street has in a given city. The long right tails of distributions correspond to the highly connected nodes of dual graphs, just a few "broadways", embankments, and belt roads crossing many more streets than an average street in the city. To give an example, in Fig. 1, we have displayed the log-log plot of fractions of streets $n(s)$ via the number of their junctions $s$ in Manhattan. These number are shown by points and the solid line is for the relevant cumulative distribution $P_c(s) = \sum_{s'=s}^{\infty} n(s')$ [7].

2 Random walks on the dual city graphs: the description of models and the sketch of results

We consider a connected graph $G = (V, E)$ with $|V| = N$ nodes and $|E| = m$ undirected edges specified by its adjacency matrix $A$ such that $A_{ij} = 1$ if the node $i$ is connected to $j$ and $A_{ij} = 0$ otherwise. A random walk starting at a node $v_0 \in V$ and traversing a sequence of random nodes $\{v_t\}$ as $t = 0, 1, \ldots$ is a Markov chain characterized by the matrix of transition probabilities $T = DA$, in which $D$ is the diagonal matrix of inverse vertex degrees, $D_{ij} = k_i^{-1} \delta_{ij}$, where $k_i = \deg(i)$, the degree of vertex $i$ in graph $G$ (the number of junctions a street shares on the city plan). The transition matrix $T$ meets the normalization condition, $\sum_j T_{ij} = 1$. We denote by $\pi^t \in \mathbb{R}^N$ the distribution of $v_t$ in the Markov chain, $\pi^t_i = \Pr(v_t = i)$. Then the rule of the walk can be expressed by the simple equation

$$\pi^{t+1} = T^t \pi^t$$

(1)

where $T^t$ is the transposed transition matrix. In the present paper, we discuss only the undirected dual city graphs with symmetric adjacency matrices, $A_{ij} = A_{ji}$. Nowadays, certain driving directions are specified for the streets by the traffic regulation polices, so that the relevant transportation lines appears to be directed. We do not consider them, limiting our present study only to the network of pedestrian access. The distribution of a current node in the random walk defined by (1) on the undirected graphs after $t$ steps tends to a well defined stationary distribution $\pi_i = k_i/2m$ (which is uniform if if the graph is regular) that is a left eigenvector of the transition matrix $T$, belonging to the largest eigenvalue 1 [12,18]. The spectrum of problem (1) is contained in the interval $[-1, 1]$.

In Sec. 3, we calculate and analyze the expected number of steps a random walker starting from node $i$ makes before node $j$ is visited (the access time) for all pairs of streets (canals) in the compact urban patterns. The properties of access times can be used in order to estimate the accessibility of certain streets
and districts by random walkers starting from the rest of city. In particular, the access times allows one to introduce the equivalence classes of nodes and to obtain a well-defined ordering of these equivalence classes, independent of any reference node. The nodes in the lowest class are difficult to reach but easy to get out of, while the nodes in the highest class are easy to reach but difficult to leave. We also discuss the random target access times and the distributions of mean access times in the compact cities. The latter characteristics can be used in order to detect ghettos (the groups of dynamically isolated nodes) and estimate the accessibility of certain districts from the streets located in other parts of the city.

The problem of random walks (1) defined on finite graphs can be related to a diffusion process which describes the dynamics of a large number of random walkers. The associated Laplace operator is more convenient, since its spectrum is positive. Indeed, the eigenvalues in (1) can be negative, so that the spectral moments \( \sum \lambda_i^{-n} \) could oscillate strongly for large networks. In contrast to it, the eigenvalues of Laplacian are positive that allows one to study its spectral properties by the powerful methods of statistical mechanics, in which various functions \( f(\lambda_i) \) defined on the spectrum \( \{\lambda_i\} \) are considered.

The diffusion process is defined by the expectation number of random walkers \( n \in \mathbb{R}^N \) and described by the equation

\[
\dot{n} = L n,
\]

in which the scaled Laplacian \( L \) (symmetric) is defined by

\[
L = 1 - D^{-1/2}TD^{1/2}
\]

where \( T \) is the unit matrix. Let us note that \( L \) is related to the transition matrix \( T \) in (1) by \( T = D^{1/2}(1 - L)D^{-1/2} \). All eigenvalues of \( L \) belong to the interval \( \lambda_\alpha \in [0, 2] \), the eigenvectors are normalized, \( |n^{(\alpha)}| = 1 \), and orthogonal, \( \langle n^{(\alpha)}\rangle n^{(\beta)} = \delta_{\alpha\beta} \).

The linear equation (2) supplied with an initial condition \( n_0 \) is solved by the function \( n^t = Q \exp(-tJ)(Q^{-1}n_0) \), in which \( J \) is a block diagonal matrix (the Jordan canonical form of \( L \)), \( Q \) is the the transformation matrix corresponding to the Jordan form. Given the eigenvalue \( \lambda_\alpha \) with the multiplicity \( m_\alpha > 1 \), the relevant contribution to \( n^t \) is given by

\[
\exp(-t\lambda_\alpha) \sum_{k=0}^{m_\alpha} \left( \sum_{l=0}^{m_\alpha-1} c_{l+1} \frac{t^l}{l!} \right) u_k
\]

where \( u_k \) is the \( k \)-th vector of the \( m_\alpha \)-dimensional eigenspace of degenerate eigenvalue \( \lambda_\alpha \) and \( c_l \) is the \( l \)-th component of the transformed vector of initial conditions, \( c = Q^{-1}n_0 \). It follows from (4) that in the presence of degenerate modes with \( m_\alpha \gg 1 \) the relaxation process can be delayed significantly. Moreover, if \( \lambda_\alpha \leq 1 \) the expectation numbers of walkers in a certain city modules would substantially increase in a lapse of time.
The analysis of spectral properties of Laplacian operator allows for detection of fine-scale dynamical modularity of city which cannot be seen from the transition matrix $T$ and provide us a tool for the estimation of entire city stability with respect to occasional cuts of certain transportation lines breaking it up into dynamically isolated components. In Sec. 4, we consider the modes of diffusion process defined on the dual city graphs and compute the coefficients of linear correlations between densities of random walkers that flow along the edges of dual city graphs in the discrete steps. Specifying the certain correlation threshold, we detect the groups of nodes traversed by the essentially correlated flows of random walkers. The structures of essentially correlated flows and their appearance are the individual characteristics of a city. We measure the quantity and extension of clusters of nodes sharing the essentially correlated flows of random walkers by an information parameter reflecting the complexity of random walk traffic.

In Sec. 5, we discuss the statistical mechanics of so called lazy random walks specified by the parameter $0 < \beta \leq 1$. In the model of lazy random walks, an agent located at a node $v$ moves to a neighboring node with probability $\beta k_v^{-1}$, but rests modeless with probability $(1 - \beta)$. We compute the well-known thermodynamical quantities (the internal energy, entropy, the free energy, and pressure) describing the macroscopical states of an ensemble of "lazy" random walkers on the dual graphs of compact urban patterns. We provided the detailed interpretation for each thermodynamical parameter in the context of random walks.

3 Access times in the compact city structures

The important characteristics of random walks defined on finite graphs is the access time $H_{ij}$, the expected number of steps before node $j$ is visited, starting from node $i$ \cite{12}. The elements of matrix $H_{ij}$ are computed for each pair of nodes $i, j$ following the formula:

$$H_{ij} = 2m \sum_{s=2}^{N} \frac{1}{1 - \mu_s} \left( \frac{\varphi_s^2}{k_j} - \frac{\varphi_s(\varphi_s)}{\sqrt{k_i k_j}} \right),$$

in which $\mu_1 = 1 > \mu_2 \geq \ldots \mu_N \geq -1$ are the eigenvalues and $\varphi_s$ are the relevant eigenvectors of the symmetrized transition matrix $D^{-1/2}TD^{1/2}$.

The access time from $i$ to $j$ may be different from the access time from $j$ to $i$, $H_{ij} \neq H_{ji}$, even in a regular graph. A deeper symmetry property of access times for undirected graphs was discovered in \cite{19},

$$H_{ij} + H_{jk} + H_{ki} = H_{ik} + H_{kj} + H_{ji}$$

for every three nodes in $G$. This property allows for the ordering of nodes in the graph with respect to their accessibility for the random walkers. It has been pointed out in \cite{12} that the nodes of any graph can be ordered so that if $i$ precedes $j$ then $H_{ij} \leq H_{ji}$. This ordering is not unique, but one can partition
the nodes by putting $i$ and $j$ in the same equivalence class if $H_{ij} = H_{ji}$ and obtain a well-defined ordering of the equivalence classes, independent of any reference node. The nodes in the lowest class are difficult to reach but easy to get out of, while the nodes in the highest class are easy to reach but difficult to leave. If a graph has a vertex-transitive automorphism group, then $H_{ij} = H_{ji}$ for all $i, j \in G$. The random target identity [18],

$$\sum_j \pi_j H_{ij} = \text{Const},$$

states that the expected number of steps (the random target access time, $\tau$) required to reach a node randomly chosen from the stationary distribution $\pi$ is a constant, independent of the starting point of the given graph $G$. The values of random target access times grow with the size of graphs and are very sensitive to their structures. Their values for the compact urban structures are given in the last column of Tab. 1.

The properties of access times can be used in order to estimate the accessibility of certain streets and districts by random walkers starting from the rest of the city. Computations of access times to the streets in the studied compact urban structures convinced us that for any given node $i$ the access times to it, $H_{ij}$, change with $j$ inferentially in comparison with their typical values, and therefore, the mean access time,

$$h_i = \frac{1}{N} \sum_{j=1}^{n} H_{ij},$$

(8)

can be considered as a good parameter for estimating the accessibility of a street by random walkers. Distributions of mean access times, $\alpha_{h}$, in the city can be considered as estimations of its connectedness. In particular, it helps to detect the ghettos, the groups of streets almost isolated (in the dynamical sense) from the rest of town. In Fig. 2, we have presented together the distributions of mean access times to the streets in Manhattan (dashed line) and Rothenburg o.d.T. (solid line). The distribution for Rothenburg o.d.T. exhibits a local maximum at relatively long access times ($\approx 300$ steps) indicating that there is a number of low accessible streets in the town. The distributions of mean access times to the canals in Amsterdam (dashed line) and Venice (solid line) is presented on Fig. 3.

The distribution of mean access times in the downtown of Bielefeld is of essential interest since it comprises of two structurally different parts (see Fig. 4.a). The part ”A” keeps its original structure (founded in XIII-XIV cs.), while the part ”B” had been subjected to the partial redevelopment in the XIX-th century (Fig. 4.a). It is important to mention that the city districts constructed in accordance to different development principles and in different historical epochs can be easily visualized on the dual graph of the city. In Fig. 4.b, we have shown the 3D-representation of the dual graph of the Bielefeld downtown. The $(x_i, y_i, z_i)$ coordinates of the $i-th$ vertex of the dual graph $G$ in three dimensional space are
given by the relevant $i-$th components of three eigenvectors $u^{(2)}$, $u^{(3)}$, and $u^{(4)}$ of the adjacency matrix $A_G$ (which does not coincide with the transition matrix $T$). These eigenvectors correspond to the second, third, and fourth largest (in absolute value) eigenvalues of $A_G$ \cite{20}. The 3D-dual graph of Bielefeld displays clearly a structural difference between "A" and "B" parts: in 3D-representation, the relevant subgraphs are located in the orthogonal planes (Fig. 4.b). Sometimes other symmetries of dual graphs can be discovered visually by using other triples of eigenvectors if the number of nodes in the graph is not too large.

In Fig. 5, we have displayed the distributions of mean access times $h$ to the streets located in the medieval part "A" starting from those located in the same part of Bielefeld downtown, from "A" to "A" (solid line). It has been computed by averaging $H_{ij}$ over $i, j \in A$ in \cite{5}. The dashed line presents the distribution of mean access times to the streets located in the modernized part "B" starting from the medieval part "A" (from "A" to "B", $i \in A$ and $j \in B$). One can see that in average it takes longer time to reach the streets located in "B" starting from "A". The similar behavior is demonstrated by the random walkers starting from "B" (see Fig. 6): in average, it requires longer time to leave a district for another one. Study of random walks defined on the dual graphs helps to detect the quasi-isolated districts of the city.

4 Dynamical modularity in the compact city structures

Dynamical modularity is a particular division of the graph into groups of nodes on which the certain modes of diffusion process are localized. It can be detected by analyzing spectral properties of the relevant dual graph that is of the spectrum of some differential operator defined on it \cite{21}. In particular, for the undirected graphs, it is convenient to consider differential elliptic self-adjoint operators with a positive spectrum. We have studied the Laplace operator defined by \cite{6}.

The eigenvalues of Laplace operator \cite{3} along with the continuous approximations of their densities for the dual graphs of compact urban patterns are shown in Figs. 7-9. The similarity of spectra allows us to divide the studied compact urban patterns into three categories: i) the medieval cities: Bielefeld (Fig. 7) and Rothenburg o.d.T.; ii) the canal patterns: Venice (Fig. 8) and Amsterdam; iii) the spectra with a highly degenerated eigenmode ($\lambda = 1$): Manhattan (Fig. 9).

It is important to note that the densities of eigenvalues for the compact urban patterns differ dramatically from those computed for the classical random graphs of Erdős-Rényi model \cite{22} \cite{23}, deviate form the semicircular law, and the densities found for the scale-free random tree-like graphs in \cite{24}. In \cite{28}, the density of eigenvalues for the Internet graph on the Autonomous Systems (AS) level had been computed. The Internet spectrum on the AS level is broadly distributed with two symmetric maxima and similar to the eigenvalue density
of random scale-free networks. In contrast to them, the spectral density for the compact city samples are bell shaped and tend to turn into a sharp peak localized at $\lambda = 1$ due to the highly degenerate 1-mode which would score a valuable fraction of eigenvalues (48% for Manhattan). 1-modes come in part from pairs of streets of minimal connectivities branching of either ”broadways” or belt roads. These structures are overrepresented in the compact city patterns.

The slowest modes of diffusion process (2) allow one to detect the city modules characterized by the individual accessibility properties. Indeed, these random walks do not describe the properties of an actual city traffic, since the approach does not concern the lengths of streets referring uniquely to the topological properties. However, the modularity plays an important role in any local dynamical process taking place in the city including its traffic conditions. The primary feature of the diffusion process in the compact urban patterns is the flow between the dominant pair of city modules: the ”broadways” and the relatively isolated streets remote from the primary roads.

Due to the proper normalization, the components of the eigenvectors $n^{(\alpha)}$ play the role of the Participation Ratios (PR) which quantify the effective numbers of nodes participating in a given eigenvector with a significant weight. This characteristic has been used in [28] and by other authors to describe the modularity of complex networks. However, under the abundance of a highly degenerate mode PR is not a well defined quantity (since the different vectors in the eigensubspace corresponding to the degenerate mode would obviously have different PR).

As time advances the distribution of random walkers approaches a steady state $n_i^\infty \propto k_i$, in which all diffusion currents are balanced. It corresponds to the principal eigenvector $n^{(1)}_i$ related to the smallest eigenvalue of $L$. The relaxation processes toward the steady state are described by the remaining eigenvectors $n^{(\alpha)}_i$, $\alpha > 1$, with the characteristic decay times $\tau^{(\alpha)}$, such that $\exp(-1/\tau^{(\alpha)}) = \lambda^{\alpha}_\alpha$. The second smallest eigenvalue of the scaled Laplacian (3) is related to the graph diameter, $\text{diam}(G) \leq -\ln(N - 1)/\ln(\lambda_2)$, the maximum distance between any two vertices in the graph. It is also related to the Fiedler vector [29] describing the algebraic connectivity of the graph. Namely, let us consider the components of the eigenvector $n^{(2)}_i$ corresponding to the second smallest eigenvalue $\lambda_2$ for the scaled Laplacian (3) defined on a connected graph $G(V,E)$. Define $V_1 = \{ v \in V : n^{(2)}_v < 0 \}$ and $V_2 = \{ v \in V : n^{(2)}_v \geq 0 \}$, then the subgraphs induced by $V_1$ and $V_2$ are connected.

In general, each nodal domain on which the components of the eigenvector $n^{(\alpha)}$ does not change sign refers to a coherent flow (characterized by its decay time $\tau^{(\alpha)}$) of random walkers toward the domain of alternative sign. The nodal domains participate in the different eigenmodes as one degree of freedom, and therefore their total number is important for detecting the dynamical modularity of city networks. It is known from [30] that the eigenvector $n^{(\alpha)}$ can have at most $\alpha + m\alpha - 1$ strong nodal domains (the maximal connected induced subgraphs, on which the components of eigenvectors have a definite sign) where $m\alpha$ is the multiplicity of the eigenvalue $\lambda_\alpha$, but not less than 2 strong nodal domains.
(α > 1) [31]. However, the actual number of nodal domains can be much smaller than the bound obtained in [31]. In the case of degenerate eigenvalues, the situation becomes even more difficult because this number may vary considerably depending upon which vector from the $m_\alpha$-dimensional eigenspace of degenerate eigenvalue $\lambda_\alpha$ is chosen. A fragment of nodal matrix for the Chelsea village (Fig. 10.a) in the Manhattan island is shown on Fig. 10.b: the components of all eigenvectors localized on the nodal domains displayed in white (black) have always the positive (negative) sign.

We investigate the statistics of components of eigenvectors $n^{(\alpha)}$ localized on a given street, $\{n_i^{(\alpha)}\}_\alpha$. To uncover a fine modular structure of compact city patterns, we have computed the linear correlation coefficients between the lists of eigenvector components for all pairs of streets in a city,

$$C_{ij} = \frac{\text{Cov}\left(\{n_i^{(\alpha)}\}_\alpha, \{n_j^{(\alpha)}\}_\alpha\right)}{\sqrt{\text{Var}\left(\{n_i^{(\alpha)}\}_\alpha\right) \text{Var}\left(\{n_j^{(\alpha)}\}_\alpha\right)}}$$

(9)

The linear correlation measures how well a linear function explains the relationship between two data sets. The correlation is positive if an increase in the eigenvectors components localized on a given street corresponds to an increase in those related to other street, and negative when an increase in one corresponds to a decrease in the other.

The study of correlations between the components of eigenvectors allows us for a precise recovering of all dynamical modules in a city, street by street. By tuning the sensitivity threshold $\kappa > 0$, one can detect the groups of streets characterized by significant pairwise correlations $C_{ij} \geq \kappa$. In case of a degenerate eigenmode, the whole bunch of streets of minimal accessibility joins a correlation cluster at once. Investigating the size of clusters characterized by the significant pairwise correlations between the eigenvectors components, we have found that it is very sensitive to the threshold value $\kappa$. Pairs of correlated streets can be detected for $\kappa \leq \kappa_c$, below the critical value $\kappa_c$ individual for each city. For instance, the strongest correlations in the Manhattan island ($\kappa \geq 0.16$) are observed between the Bowery st. (Chinatown) and the 6-th Avenue (Greenwich Village), and between the Lafayette st. and Mott st. (Chinatown). By reducing $\kappa$ just by 1%, we immediately get many new correlated pairs and [South FDR Dr., Allen st.], [Greene st.(Soho), Ave. D], [Crosby st. (Little Italy), Ave. D], [Centre st, Elizabeth st.], [Allen st., 6-th Ave.], [Lafayette st., Ave. B], [Broadway, West st.] among the others. In most cases the streets in correlated pairs have the same driving directions although the driving direction data has not been initially used in the adjacency matrices. Reducing the value of sensitivity threshold, one can detect the triples of correlated streets and more structurally complicated dynamical modules. Then the correlated clusters merge exhibiting a sharp phase transition into a giant connected component for essentially small $\kappa$.

The appearance of correlated $k$-tuples and their total numbers at given $\kappa$ are by no means random and encode the important information on the city
connectedness. The complexity of dynamical modularity can be measured by a quantity of information encoded by the number of various $k$-tuples of essentially correlated nodes at the different values of correlation threshold $\kappa$ in the following way. Since the probability that a connected correlated $k$-tuple appears in a random labelled graph of $N$ nodes is

$$p_k(N) = \binom{N}{k}^{-1},$$

its appearance corresponds to some quantity of information $I = -p_k(N) \log_2 p_k(N)$. For a mesh of values $\kappa > 0$, we had counted the total numbers of all correlated $k$-tuples appeared in each city, $N_k(\kappa)$, and then found the total amounts of information $\mathcal{I}(\kappa)$ encoded by the dynamical modularity in each studied city pattern,

$$\mathcal{I}(\kappa) = - \sum_{k\text{-tuples}} N_k p_k (|V|) \log_2 p_k (|V|). \tag{10}$$

It is worth to mention that in the absence of correlations as well as in the case when the all fluxes of random walkers through nodes of the dual city graph are correlated, no information can be encoded, and therefore $\mathcal{I} = 0$. The results on the comparative information analysis of dynamical modularity of the compact urban structures are displayed in Figs. 11-13.

In general, the quantity of information encoded by the dynamical modularity of ancient cities exceeds that in the modern or redeveloped ones since a number of correlated clusters of diverse sizes could appear in the graphs with a less regular structure. Information (via the threshold of correlations, $\kappa$) calculated for Rothenburg and Bielefeld (subjected to a partial redevelopment) are almost equal (Fig. 11). Both medieval cities contain a number of streets characterized by the relatively highly correlated traffic (they are the belt roads encircling the cities along their fortress walls), while the traffic along the streets close to the city centers appears to be less correlated, although all streets of low accessibility join the correlation cluster at once at some level $\kappa$, then the information on the dynamical modularity turns to zero.

Information profiles obtained for the Venice and Amsterdam canal networks look similar (see Fig. 12). However, it seems that information in a message on possible correlations of gondoliers traffic in Venice could be much more valuable than a cruise schedule along the Amstel embankments. The triggering between different information states displayed on Fig. 12 corresponds to the merging of diverse correlated clusters into the bigger correlated modules. The information peaks located close to $\kappa = 0$ indicate a phase transition to a giant correlated component which cover most of the nodes in the networks. Nevertheless, not all canals join the giant correlated module (since the amount of information is not zero as $\kappa \to 0$, but, on the contrary, turns to be high). It happens because of the presence of an anti-correlated module in the canal network, in which the components of Laplacian eigenvectors are negatively correlated, $C_{ij} < 0$.

The scale of the graph shown in Fig. 13 is incompatible with those of Fig. 11 and Fig. 12 since the dynamical modularity in the Manhattan island (one of
5 Statistical mechanics of lazy random walks in the compact city structures

A number of different "measures" quantifying the various properties of complex networks has been proposed in so far in a wide range of studies in order to distinguish the groups of nodes and enlighten the relations between them. Some measures can be computed directly from the graph adjacency matrix: likelihood \[36\], assortativity \[37\], clustering \[38\], degree centrality \[33\], \[39\], \[40\], betweenness centrality \[41\], link value \[42\], structural similarity \[43\], distance (counting the number of paths between vertices) \[44\]. Other measures (concerned with the networks embedded into Euclidean space) involve the lengths of links or the true Euclidean distances between nodes: closeness centrality \[33\], \[43\], straightness centrality \[44\], \[45\], expansion \[41\], information centrality and graph efficiency \[44\], \[46\]. A good summary on the several centrality measures can be found in \[27\] and in \[47\] for the Internet related measures. The list of available measures is still far from being complete, the new measures appearing together with any forthcoming network model. It is also worth to add some spectral measures (concerned the eigenvalues of graph adjacency matrix): subgraph centralization \[48\], subgraph centrality \[44\], \[41\], \[49\], network bipartivity \[49\] and many others.

In the present section, we study the statistics of flows of lazy random walkers roaming in the compact city patterns. The random walkers have no mass and do not interact with each other, so that they do not neither to contribute to the energy nor to the momentum transfer. Nevertheless, their flows have the nontrivial thermodynamical properties induced by the complex topology of streets and canals they flow along. The obvious advantage of statistical mechanics is that the mathematical objects we introduce and all relations between various statistical quantities are well known in the framework of thermodynamic formalism.

In the following, we use the inverse temperature parameter \( \beta > 0 \) which can be considered either as an effective time scale in the problem (the number of streets a walker passes in one time step) or as the laziness parameter defined in Sec. 2. Since the eigenvalues \( \lambda_\alpha \) of scaled Laplacian operator \( \mathcal{L} \) are positive and bounded, one can define for them three well known spectral functions:

1. the heat kernel (the partition function),

\[
K(\beta) = \text{Tr} \exp(-\beta \mathcal{L}) = \sum_{\alpha=1}^{N} e^{-\beta \lambda_\alpha}, \tag{11}
\]

converging as \( \beta \geq 0 \), for \( N \to \infty \),
2. the spectral zeta function (the spectral moments)

\[ \zeta(s) = \sum_{\alpha=1}^{N} \lambda_{\alpha}^{-s}, \]  

(12)

3. and the spectral density,

\[ \rho(\lambda) = \sum_{\alpha=1}^{N} \delta(\lambda - \lambda_{\alpha}). \]  

(13)

These spectral functions are related to each other by the Laplace transformation,

\[ K(\beta) = \int_{0}^{\infty} d\lambda \ e^{-\beta \lambda} \rho(\lambda), \]  

(14)

and by the Mellin transformation (up to the \( \Gamma \)-function) [50],

\[ \zeta(s) = \frac{1}{\Gamma(s)} \int_{0}^{\infty} \beta^{s-1} K(\beta) d\beta. \]  

(15)

Furthermore,

\[ \zeta(s) = \int_{-\infty}^{\infty} \rho(\lambda) \lambda^{-s} d\lambda. \]  

(16)

Let us note that since \( 0 < \lambda_{\alpha} < 2 \), we immediately obtain bounds for the spectral moments,

\[ \zeta(-n) < N \cdot 2^n, \quad \zeta(n) > N \cdot 2^{-n}. \]

The partition function \( K(\beta) \) meets the conditions of Bernstein’s theorem [32]:

\[ (-1)^n K^{(n)}(\beta) \geq 0, \beta > 0, n \geq 1, \]

and therefore defines a unique non-negative measure \( d\mu(\beta) = K(\beta)d\beta \) on \( \mathbb{R}_{+} \), providing the solutions for the Hamburger and Stiltjes moments problems, [50].

Concerning the list of “measures” given above, it is worth to note that, in principle, any of spectral functions \( \{11,13\} \) at a given temperature \( \beta^{-1} \) can be used as a measure of some quantities relevant to a certain lazy random walk model.

The ensemble of random walkers on the graph \( G(V,E) \) can be characterized by the following \textit{macroscopical} quantities: 1) the internal (averaged) energy, \( \bar{E} = -\partial_{\beta} \ln K(\beta) \); 2) the entropy, \( S = \ln K(\beta) + \beta \bar{E} \); 3) the free energy, \( F = \beta^{-1} \ln K(\beta) \), and 4) the pressure, \( P = \beta^{-1} \partial_{|V|} \ln K(\beta) \). In the last equality, we have considered the differential with respect to the graph size as \( d|V| = \rho(\lambda)d\lambda \). The thermodynamics of compact urban city patterns is presented on the Figs. 14-22 and in Tab. 2. The collected data give an insight into the various aspects of \textit{averaged} streets (canals) accessibility in a given city.
Due to the complicated topology of streets and canals, the flows of random walkers exhibit the spectral properties similar to that of a thermodynamical system characterized by a nontrivial internal energy (Figs. 14-16). It grows with temperature (the inverse parameter of lazy random walks) $\beta^{-1}$ (albeit still bounded in the interval $[0, 2]$). As usual, the absolute value of internal energy relevant to a given city cannot be precisely measured, but we can measure its difference in any temperature interval. In principle, the slopes of internal energy curves are steeper (the internal energy grows faster with $\beta^{-1}$) in the modern cities with a quite regular grid-like structures, in which any street has a relatively good accessibility.

In thermodynamics, entropy is an extensive state function that accounts for the effects of irreversibility in thermodynamic systems. It describes the number of the possible microscopic configurations of the system. In the problem of finite random walks, its value quantifies the diversity of flows which can be detected in the city at a given temperature and grows with temperature. In Figs. 17-19, we have displayed the entropy curves via $\beta^{-1}$. It is important to mention that the improve of street accessibility causes a decrease of entropy growth rates for large temperatures (small $\beta$). The entropy of random walker flows as well as its growth rate in Manhattan is ever less than in any other compact city structure we studied.

Due to the numerous junctions and the highly entangled meshes of city street and canals, the random walkers loss their ways and it takes long time for them to cross a city roaming randomly along the streets. This can be interpreted as an effect of a slight negative pressure involving the random walkers into the city. It is obvious that the strength of pressure should vary from one district to another within a city and can be different for the different modes of diffusion process which have no rigorous bind to the city administrative units being localized on the certain groups of streets and canals. We have computed the pressure spectra $P(\lambda)$ forcing the flows of random walkers with eigenmodes $\lambda$ into the compact city structures (see Figs. 20-22). Generally speaking, the more junctions a city has, the stronger is the drag force: despite Venice has more canals than in Amsterdam, the number of junctions between them is less than in Amsterdam, and therefore the negative pressure in the latter city is stronger (Fig. 20). The numbers of streets in Rothenburg and in the downtown of Bielefeld are equal, but there are more crossroads in Bielefeld than in Rothenburg (Fig. 21).

One can see that pressure profiles have maxima close to $\lambda = 1$ (correspondent to a minimal drug force) for Amsterdam, Venice, and Manhattan. It calls for the idea of a "transparency corridor", i.e. a sequence of streets and junctions (on which the relevant eigenmodes are localized) along which the city can be crossed at minimal time. In Manhattan, the pressure profile is almost zero at $\lambda = 1$. The multiply degenerated eigenmode $\lambda = 1$ appears in Manhattan due to numerous junctions of low accessible streets to Broadway and FDR Drive. In Venice, the minimal pressure is achieved on the Grand Canal and Giudecca Canal, and it is due to Het IJ and Amstel river in Amsterdam.

In Tab. 2, we have sketched the figures for the internal energy, entropy, the free energy, and heat capacity for all studied cities at $\beta = 1$ (for the usual
random walks model), in purpose of comparison. We also gave the location of maxima for the heat capacity profiles.

6 Discussion and Conclusion

We have studied the finite Markov chain processes defined on the dual graphs of compact urban patterns. The traffic of random walkers, indeed, do not describe the actual traffic conditions in the city patterns that we have analyzed, but concerns their topological properties giving a sense to the notions of accessibility and modularity. Our approach can be readily used in order to investigate connectedness and efficiency of transportation lines in different complex networks.

The methods developed in graph theory and in probability theory give us a detailed picture of local and global properties of city structures. Dynamical modularity representing the localization of dynamical modes on certain streets and canals cannot be detected neither from the adjacency matrix of a graph nor from the transition probability matrix. We have studied the random walks of a large number of massless particles introduced in the dual graphs of compact cities and investigated their properties by means of statistical mechanics. The spectrum of Laplace operator is broadly distributed for the compact city patterns and has a bell shape, in general. However, it turns into a sharp peak if there is a number of low accessible streets branching of a prime street in a city.

To detect the dynamical modularity precisely, we have computed the linear correlation coefficients between the lists of eigenvector components for all pairs of streets (canals) in the cities. Tuning the sensitivity threshold of correlations, one can detect the various modules of essentially correlated streets. The complexity of dynamical modularity can be measured by the information quantity. The information versus correlation profile is an individual city fingerprint.

The statistical mechanics description of random walks in the compact cities provides us with a rigorous definition of the unique measure and all statistical moments which have a definite relation to the entire topological properties of complex networks. Due to a complicated topology of city streets and canals, the flows of random walkers acquire nontrivial thermodynamical characteristics which can be implemented in studies of interactions between different city modules and of city stability with respect to occasional accessibility problems that would be of vital importance for the city traffic conditions in urgency.

The obvious advantage of the method is that the information on the role of a given street or canal in the entire city network can be interpreted by the very end, and all city modules can be named street by street.

The method can be generalized for the complex networks which can be described by directed or weighted graphs. For instance, one can consider a graph of city plan in which the weights of edges are the actual lengths of streets. In the case of weighted adjacency matrix, the spectrum of the relevant Laplace operator acquires the pairs of complex conjugated eigenvalues. Then the components of eigenvectors are also complex, and the coefficients of linear correlations
should be computed for the real and imaginary parts of spectrum separately. Other approach could be used while studying directed graphs for which the adjacency matrix is not symmetric. For the random walks defined on the directed graphs, the probability that a random walker enters a node is not equal to the probability it leaves the node. The ensembles of random walkers introduced on the directed graphs can be described by Nelson stochastic mechanics \cite{51,52}. The stationary configurations of random walkers in such models can be studied by means of biorthogonal decomposition \cite{53}.

7 Acknowledgements

One of the authors (D.V.) is grateful to the Alexander von Humboldt Foundation (Germany), to the Bielefeld-Bonn Stochastic Research Center (BiBoS, Bielefeld, Germany), to the DFG-International Graduate School "Stochastic and real-world problems", to the Centre de Physique Theorique (CPT, C.N.R.S. Contract CEA - EURATOM N006385) for their financial support and hospitality during the preparation of the present paper.

We thank R. Lima, B. Cessac, and G. Zaslavsky for fruitful discussions and multiple comments on our work.

References

[1] J.D. Nystuen, M.F. Dacey, A Graph Theory Interpretation of Nodal Regions, Papers and Proceedings of the Regional Science Association, 7 29 (1961).

[2] K.J. Kansky Structure of Transportation Networks: Relationships Between Network Geometry and Regional Characteristics, Res. Paper 84, Dept. Geograph., University of Chicago, Chicago, IL (1963).

[3] P. Hagget, R.J. Chorley, Network analysis in Geography, Edward Arnold, London (1969).

[4] L. March, J.P. Steadman, The Geometry of Environment, RIBA Publ., London (1971).

[5] J.P. Steadman, Architectural Morphology: Introduction to the Geometry of Building Plans, Pion Press, London (1983).

[6] A. Cardillo, S. Scellato, V. Latora, S. Porta, Phys. Rev. E 73, 066107 (2006).

[7] M.E.J. Newman. The structure and function of complex networks. SIAM Review, 45 167 (2003).

[8] J. Buhl, J. Gautrais, R.V. Solé, P. Kuntz, S. Valverde, J.L. Deneubourg, G. Thereulaz, Eur. Phys. Journ. B 42, 123 (2004).
[9] T. Harold Hughes, *The Principles to be Observed in Designing and Laying out Towns Treated from the Architectural Standpoint*, Journ. of the Royal Inst. of British Architects, Dec. 7 (1912), 65; Dec. 21 (1912), 125.

[10] B. Hillier, J. Hanson, *The Social Logic of Space*, Cambridge University Press, Cambridge, UK (1984).

[11] B. Hillier, *Space is the Machine: A Configurational Theory of Architecture*, Cambridge University Press, Cambridge, UK (1996).

[12] L. Lovász, *Random Walks On Graphs: A Survey*, Bolyai Society Mathematical Studies, 2: "Combinatorics, Paul Erdős is Eighty", Keszthely (Hungary), p. 1-46 (1993).

[13] M. Batty, *A New Theory of Space Syntax*, Working Parer Seies, Centre for Advanced Spatial Analysis, University College of London (2004).

[14] S. Bilke, C. Peterson, *Phys. Rev. E* 64, 036106 (2001);

[15] S. Porta, P. Crucitti, V. Latora, *Physica A* 369, 853 (2006).

[16] M. Rosvall, A. Trusina, P. Minnhagen, K. Sneppen, *Phys. Rev. Lett.* 94, 028701 (2005).

[17] B. Jiang, C. Claramunt, ”Topological analysis of urban street networks”, *Environmental Planning B* 31, 151-162 (2004).

[18] L. Lovász; P. Winkler, Mixing of random walks and other diffusions on a graph. *Surveys in combinatorics, 1995 (Stirling)*, 119-154, London Math. Soc. Lecture Note Ser., 218, Cambridge Univ. Press.

[19] D.Copersmith, P. Tetali, P.Winkler, *SIAM J. Discr. Math.* 6, 363 (1993).

[20] The 3d-drawing of networks with respect to thier eigenvectors is used in the standard routines of Maple and Matlab environments.

[21] Y. Colin de Verdière, *Spectres de Graphes*, Cours Spécialisés 4, Société Mathématique de France (1998).

[22] I.J. Farkas, I. Derényi, A.-L. Barabási, T. Vicsek, *Phys. Rev. E* 64, 026704 (2001).

[23] I. Farkas, I. Derényi, H. Jeong, Z. Neda, Z.N. Oltvai, E. Ravasz, A. Schubert, A.-L. Barabási, T. Vicsek, *Physic A* 314, 25 (2002).

[24] S.N. Dorogovtsev, A.V. Goltsev, J.F.F. Mendes, A.N. Samukhin, *Phys. Rev. E* 68, 046109 (2003).

[25] S. Scellato, A. Gardillo, V. Latora, S. Porta, *Eur. Phys. Journ. B* 50, 221 (2006).

[26] P. Crucitti, , V. Latora, S. Porta, *Phys. Rev. E* 73, 036125 (2006).
[27] P. Crucitti, V. Latora, S. Porta, *Chaos* **16**, 015113 (2006).

[28] K. A. Eriksen, I. Simonsen, S. Maslov, K. Sneppen, *Phys. Rev. Lett.* **90**, Issue 14, id. 148701 (2003).

[29] M. Fiedler, *Chechoslovak Math. J.* **23**, 298 (1973); *Chechoslovak Math. J.* **25**, 146 (1975).

[30] E.B. Davis, G. M.L. Gladwell, J. Leydold, P.F. Stadler, *Lin. Alg. Appl.* **336**, 51-60 (2001).

[31] T. Biyikoğlu, W. Hordijk, J. Leydold, T. Pisanski, P.F. Stadler, *Lin. Alg. Appl.* **390**, pp. 155-174 (2004).

[32] W. Feller, *An Introduction to Probability Theory and Applications*, Vol. 2, NY-London-Sydney: John Wiley&Sons, Inc. XVIII (1966).

[33] S. Wasserman, K. Faust, *Social Network Analysis*, Cambridge Univ. Press, Cambridge, England (1994).

[34] E. Estrada, J.A. Rodríguez-Velázquez, *Phys. Rev. E* **71**, 056103 (2005).

[35] E. A. Leicht, P. Holme, M. E. J. Newman, *Phys. Rev. E* **73**, 026120 (2006).

[36] L. Li, D. Alderson, W. Willinger, J. Doyle, *A first principles approach to understanding the Internet router-level topology*, In ACM SIGCOMM (2004).

[37] M. E. J. Newman, *Phys. Rev. Lett.* **89**, 208702 (2002).

[38] M. E. J. Newman, *Phys. Rev. E* **68**, 026121 (2003).

[39] J. Nieminen, *Scand. J. Psychol.* **15**, 322 (1974).

[40] L.C. Freeman, *Soc. Networks* **1**, 215 (1979).

[41] H. Tangmunarunkit, J. Doyle, R. Govindan, S. Jamin, W. Willinger, S. Shenker, *Does AS size determine AS degree?*, In ACM Computer Comm. Review, Oct. (2001).

[42] M. Girvan, M. E. J. Newman, *PNAS* **99** (12), 7821 (2002).

[43] L.G. Sabidussi, *Psychometrika* **31**, 581 (1966).

[44] I. Vragovic, E. Louis, A. Diaz-Guilera, *Phys. Rev. E* **71**, 036122 (2005).

[45] V. Latora, M. Marchiori, *Phys. Rev. Lett.* **87**, 198701 (2001).

[46] V. Latora, M. Marchiori, *Phys. Rev. E* **71**, 015103 (R) (2005).

[47] P. Mahadevan, D. Krioukov, K. Fall, A. Vahdat, *A Basis for Systematic Analysis of Network Topologies*, In ACM SIGCOMM 2006.
[48] D. Cvetkovic, P. Rowlinson, S. Simic, *Eigenspaces of Graphs*, Cambridge Univ. Press, Cambridge, UK (1997).

[49] E. Estrada, J.A. Rodríguez-Velázquez, *Phys. Rev. E* 72, 046105 (2005).

[50] M. Tierz, E. Elizalde, *Heat Kernel-Zeta Function Relationship Coming From The Classical Moment Problem*, [arXiv:math-ph/0112050](http://arxiv.org/abs/math-ph/0112050) (2001).

[51] E. Nelson, *Dynamical Theories of Brownian Motion*, Princeton Univ. Press (1967).

[52] Ph. Blanchard, Ph. Combe, W. Zheng, *Mathematical and Physical Aspects of Stochastic Mechanis*, in Lecture Notes in Physics 281, Springer-Verlag.

[53] N. Aubry, R. Guyonnet, R. Lima, *Journ. Stat. Phys.*, 64, 683 (1991).
Table 1: Some features of dual graphs of the studied city patterns

|                          | |V|   | |E|   | diam(G) | τ    |
|--------------------------|---|---|---|---|---|---|---|
| Rothenburg ob.d.T.       | 50| 115|13 | 545|
| Bielefeld downtown       | 50| 142|14 | 551|
| Amsterdam canals         | 57| 200|11 | 849|
| Venice canals            | 96| 196|14 |1550|
| Manhattan                | 355|3543|17 |4557|

**Caption for Table 1:** Some features of dual graphs of compact city patterns: |V| is the number of streets, |E| the number of junctions (crossroads), the graph diameter diam(G) is the maximal graph-theoretical distance between any two vertices of the dual graph. In the last column, the the random target access time τ, the expected number of steps required to hit a node randomly chosen in the city from the stationary distribution π. The random target access time is independent of the starting point.
Table 2: Thermodynamical properties of ensembles of random walkers in the compact cities at $\beta = 1$ (the usual random walks).

| City                  | $E(\beta = 1)$ | $S(\beta = 1)$ | $F(\beta = 1)$ | $C_{\text{max}}$ | $\beta_{C_{\text{max}}}$ |
|-----------------------|-----------------|-----------------|-----------------|------------------|---------------------------|
| Rothenburg ob.d.T.    | 0.946604        | 3.885109        | 2.93850         | 1.60468          | 11.1                      |
| Bielefeld downtown    | 0.957324        | 3.89042         | 2.93309         | 2.04195          | 9.90                      |
| Venice canals         | 0.95082         | 4.53962         | 3.58880         | 1.92003          | 27.4                      |
| Amsterdam canals      | 0.898409        | 3.88836         | 2.98996         | 1.3353; 1.1094   | 8.6; 48.75                |
| Manhattan             | 0.990083        | 5.86713         | 4.87705         | 5.87368          | 13.65                     |

Caption for Table 2: $\bar{E}$ is the internal energy, $S$ is the entropy, $F$ is the free energy, $C_{\text{max}}$ is the maximal values of heat capacity, and $\beta_{C_{\text{max}}}$ is the points they are achieved.
Figure 1: The logarithm of fractions of streets $\ln n(s)$ via the logarithm of number of their junctions $\ln s$ in Manhattan (shown by points). The solid line represents for the logarithm of the relevant cumulative distribution $\ln P_c(s)$ \cite{7}. 


Figure 2: The distributions of mean access times $h$ to the streets in Manhattan (dashed line) and Rothenburg ob der Tauber (solid line). The distribution $\alpha_h$ for Rothenburg o.d.T. exhibits a local maximum at relatively long access times (300 steps) indicating the presence of a number of low accessible streets in the town.
Figure 3: The distributions of mean access times $h$ to the canals in Amsterdam (dashed line) and Venice (solid line).
Figure 4: The city map of Bielefeld downtown (a) is presented together with its 3D representations of the dual graph (b). The "A"-part keeps its original structure (founded in XIII-XIV cs.); the part "B" which had been redeveloped in the XIX-th century. The \((x_i, y_i, z_i)\) coordinates of the \(i\)–th vertex of the dual graph in three dimensional space are given by the relevant \(i\)–th components of three eigenvectors \(u^{(2)}, u^{(3)}, u^{(4)}\) of the adjacency matrix \(A\) of the dual graph.
Figure 5: The distributions of mean access times $h$ to the streets located in the medieval part "A" starting from those located in the same part of Bielefeld downtown, from "A" to "A" (solid line). The dashed line presents the distribution of mean access times to the streets located in the modernized part "B" starting from the medieval part "A" (from "A" to "B"). In average, it takes longer time to reach the streets located in "B" starting from "A".
Figure 6: The distributions of mean access times $h$ to the streets located in the "B" part starting from "B" (solid line). The dashed line presents the distribution of mean access times to the street located in the "A" part starting from "B".
Figure 7: The eigenvalues of the Laplacian defined on the dual graph of Bielefeld together with the continuous approximation of its density.
Figure 8: The eigenvalues of the Laplacian defined on the dual graph of Amsterdam canal network together with the continuous approximation of its density.
Figure 9: The eigenvalues of the Laplacian defined on the dual graph of Manhattan together with the continuous approximation of its density.
Figure 10: a. Plan of the Chelsea village in Manhattan and the matrix plot of its nodal domains (b). All eigenvectors localized on the nodal domains displayed in white (black) have always the positive (negative) sign.
Figure 11: Quantity of information (bits) encoded by the dynamical modularities of Rothenburg and Bielefeld via the threshold of essential correlations, $0 \leq \kappa \leq 1$. 

$I(\kappa)$, "bits"
Figure 12: Quantity of information (bits) encoded by the dynamical modularities of the Venice and Amsterdam canal networks via the threshold of essential correlations, $0 \leq \kappa \leq 1$. 
Figure 13: Quantity of information (bits) encoded by the dynamical modularities of Manhattan via the threshold of essential correlations $0 \leq \kappa \leq 1$. 
Figure 14: The grows of "internal energy" in the models of lazy random walks with "temperature" $\beta^{-1}$ (the inverse parameter of lazy random walks) for the canal networks of Amsterdam and Venice.
Figure 15: The growth of "internal energy" in the models of lazy random walks with "temperature" $\beta^{-1}$ (the inverse parameter of lazy random walks) for the medieval cities, Rothenburg o.d.T. and Bielefeld.
Figure 16: The grows of "internal energy" in the models of lazy random walks with "temperature" $\beta^{-1}$ (the inverse parameter of lazy random walks) for Manhattan and the theoretical example of a "village" extended along the only principal street.
Figure 17: The entropy curves via the inverse parameter of lazy random walks, $\beta^{-1}$, for the canal networks of Venice and Amsterdam
Figure 18: The entropy curves via the inverse parameter of lazy random walks, $\beta^{-1}$, for Bielefeld and Rothenburg o.d.T.
Figure 19: The entropy curves via the inverse parameter of lazy random walks, $\beta^{-1}$, for Bielefeld, Rothenburg, and Manhattan.
Figure 20: The comparison of pressure spectra $P(\lambda)$ acting on the flows of random walkers with eigenmodes $\lambda$ into Amsterdam and Venice.
Figure 21: The comparison of pressure spectra $P(\lambda)$ acting on the flows of random walkers with eigenmodes $\lambda$ into Bielefeld and Rothenburg.
Figure 22: The comparison of pressure spectra $P(\lambda)$ acting on the flows of random walkers with eigenmodes $\lambda$ into Manhattan and Venice.