Complex Urban Systems: Challenges and Integrated Solutions for the Sustainability and Resilience of Cities

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1. Introduction

Cities offer one of the most challenging test beds for any complexity-oriented modeling approach. The reason is simple: they present a multiscale structure integrating a multitude of social and technological subsystems. While being large enough to be amenable to macromodeling, at the same time, they are not large enough to be exclusively approached at that scale, thus raising ambition for detailed microstructural analysis and understanding. For these reasons, the modeling and analysis of cities sit naturally at the mesoscale, at the edge between micro and macro, and offer an ideal environment for the development of “statistical mechanics” of human interaction.

On the basis of these premises, it is paradoxically not surprising that the most authoritative and celebrated account of how cities “work,” which has informed a countless number of different approaches and analysis of all sorts, is Jane Jacobs’ book “The Death and Life of Great American Cities” [1], which is essentially an autoethnography of the...
experience of the city par excellence, New York City, that is, an approach that could not be farther away from conventional scientific standards: subjective observations about a single city through time. Despite its lack of “hard scientific” method, the book has been so influential and was celebrated to become a sort of conceptual map for all scientists aiming at building solid scientific explanations and analyses of the urban dynamics. Such influence stems from its unique capacity of summing up, through the author’s gaze, so many different, subtly related aspects of the essence of cities functioning and living. Not incidentally, one of its major insights among many is that cities thrive only if they are able to maintain their own form of highly idiosyncratic complexity. If they fall for structural oversimplification and loss of diversity, they decay and possibly eventually die. Traditional top-down planning practices have not successfully passed the urban complexity test, due to their inability to credibly address the mutability of social interactions in urban settings through their rigid schemes and their consequent tendency to micromanage environmental complexity rather than enable its generative potential [2]. Moreover, Jacobs’ lesson reminds us that the complexity of the city is not only about the manifold aspects of the urban environment and their interrelations but also about the complexity of our own mental representation of the city. Jacobs’ insights are also about what planning means for the city and about the subtle balance between self-organization and intelligent design, and here too there is enormous inspiration for readers who are accustomed to think in complexity science terms. There are, therefore, at least three different dimensions of complexity that one should keep in mind thinking of the city: the complexity of the urban environment, the complexity of human cognition about the city, and the complexity of city planning. A comprehensive approach to urban complexity should be able to encompass all three and even more so in the complex postpandemic scenario with which all cities will have to cope in the coming years.

2. Modeling Urban Complexity

Research on the complexity of urban environments has a long tradition, and although its roots are difficult to trace back, a fundamental text is Christopher Alexander’s “A City Is Not a Tree” [3]. In this short, insightful essay, later developed into a book [4], Alexander makes use of biological analogies to explore the inherent geometrical properties of urban organization, largely prefiguring the complexity science of the next two decades. As shown in his later book [5], Alexander clearly understands the relationship between the emergent macrostructures of the city and the microlevel of building construction patterns and their compositional space grammar, thus characterizing architectural building rules as subject to adaptive pressures. Alexander then goes further on [6] to identify vernacular architecture as a self-organizing system of space organization which reflects an extremely complex system of socioenvironmental cognition and constitutes an ideal bridge between the complexity science of urban environments and that of their mental representations. Finally, in the monumental 4-book series, “The Nature of Order” [7–10], Alexander arrives at an all-encompassing evolutionary synthesis of human and biological organization structures, where he finally investigates issues such as why certain human settlements have more “life” in them than others. One might think of a city’s “liveliness” in terms of an ensemble of emergent structural properties that result from the coevolution of built environments and human interaction, following a logic that closely resembles that of biological design.

It is from these bases that the literatures on shape grammars [11], space syntax [12], and their inevitable confluences [13] take off, building the premises of a computational approach to urban form and function. The marriage with the concurrently upcoming complexity science would not only be inevitable [14, 15] but necessary, with the urban dimension becoming one of the natural testing grounds for fractal [16], agent-based [17], and cellular automata modeling [18] of the emergent order properties of multiscale systems, once again naturally coalescing into a unified theory of urban complexity [19].

Complexity-based approaches to urban issues have since then proliferated to practically every sphere of city life [20, 21]: transportation systems [22], utilities and infrastructure [23, 24], pollution [25], and crime [26], just to limit ourselves to a few examples. These approaches have allowed the development of much deeper insights into the nature and effects of structural interdependencies across urban environments [27], also deriving from the unique tension between the general nonlinear effects typical of all urban environments with local, specific factors and dynamics [28, 29]. The rapidly increasing availability of large databases and the big data revolution in social and urban sciences has further boosted this tendency, leading to a new wave of complexity-oriented urban science which is likely still in its early phase [30], and it is consequently challenged by the need of developing proper analytical methods for the extraction of reliable behavioral information [31, 32]. There is therefore room to expect that the application of complexity science to the modeling, analysis, and understanding of urban systems is a long-term scientific endeavor rather than a transitory phase and that this will have profound effects on many dimensions of science, society, and the economy.

From a complexity science perspective, cities cannot be simply viewed as structures in space but also as functional systems of flows and networks [33]. In 1961, Gilbert used a special class of networks, namely, random geometric graphs, to model the structure of spatially embedded networks and the effects of spatial constraints on the system [34]. The core idea is to consider spatially distributed nodes representing, for instance, geographic areas which are connected to each other if their distance is within some spatial scale used as a reference. This class of models is desirable for studying the structure and the function of complex systems like a city, consisting of areas connected by transportation infrastructures [35].

Nowadays, network modeling and analysis of urban ecosystems is a widely adopted framework to cope with the
complexity of cities and of their societies at different scales [36]. The analysis of the Boston underground transportation system through the lens of global and local efficiency in information flows revealed that its underlying logic of construction is, in fact, a small-world principle [37]. Complex networks have been used for geographical modeling and, by means of combined cellular models of land and behavior, it has been shown that they provide a compelling framework for growth dynamic that is consistent with large-scale regularities, such as fractality or power-law scaling relations [38]. The analysis of the dual graph representation, where roads and junctions are mapped into nodes and edges, respectively, of six urban street networks characterized by different patterns and historical roots revealed their unique connectivity patterns with respect to nongeographic systems [39]. Network science has been used for spatial analysis of the topology of Singapore inferred from human-generated data, by identifying city hubs, centers, and other elements which are essential to characterize urban interactions. Results from longitudinal analysis suggest that Singapore is rapidly developing towards the designed polycentric urban form [40]. An example of network analysis in action is shown in Figure 1.

Models and analytical tools borrowed from or inspired by complexity science are proliferating, providing convincing evidences of their application to real cities [42–46], from human mobility [47–50] and traffic congestion [51–55] to energy consumption [56], air quality [57, 58], climate [59], and health and wellbeing [60–63], as well as accessibility to important facilities like hospitals [64]. The city is seen as a huge complex system which grows and expands [65, 66] and whose spatial organization [67, 68] dynamically experiences a transition from monocentric to polycentric [69, 70].

The relevance of complexity modeling tools to understand urban ecosystems ignited an unprecedented deluge of open and crowdsourced information about the topology of the city and its fundamental constituents [71, 72], as well as about its function, directly related to the behavior of its inhabitants inferred from the data they generate, such as phone call detail records [47, 73–77], transactions [78–83], GPS trajectories [49, 84–88], and geo-tagged social media [89–101].

2.1. The Multilayer Structure of the City. Very recently, it has been suggested that a new level of complexity characterizes cities. In fact, accounting for multiplexity [102, 103] of transportation, that is, multimodality [104] (see Figure 2), and interdependencies [105, 106], that is, structural and functional relationships with other systems (see Figure 3), allows, on the one hand, to gain new insights about the functioning of a city and the complex society it hosts. On the other hand, it allows to better understand its resilience to targeted policies, such as infrastructural interventions, or to random perturbations, such as unexpected failures in transportation or energy networks, as well as catastrophic events.

Multilayer models for transportation infrastructures [104, 107, 108] group together connections from the same transportation mode, assign them to a layer, and sometimes couple layers with each other, the last one depending on whether information about time or economic cost to move across layers is available or not [109].

Different perspectives can be adopted. For instance, one might assign different submodes (e.g., distinct lines of the tube) to different layers [110, 111] or group them together within the same layer encoding a unique means of transportation [104]. The two approaches, applied to the backbone of 9 different cities in Europe, from small towns to megacities, highlighted different vulnerabilities and provided a framework for testing improvements [112]. The analysis of topological pathways across layers revealed, for instance, that London’s public transportation is designed to minimize redundancies [113]. The London underground network exhibits patterns that are not observed in other systems, like social networks. The calculation of clustering, that is, the tendency to triadic closure, revealed mechanisms to avoid redundant connections, with 3-mode triangles more frequent than 2-mode triangles which, in turn, are more frequent than single-mode triangles [114]. The analysis of the interdependence between the street network and the subways of London and New York City unraveled similar mechanisms, with the underground network acting as a decentralizing force which pushes congestion towards the end of its lines. It has been found that uneven spatial distributions of accessibility might emerge if the speed of subways is increased without a systemic view of the city [115].

While transportation networks with their multiplexity and interdependency are fundamental to enhance our understanding of the city, it is similarly of utmost importance to include other systems, for example, sociotechnical and ecological ones, which are in turn shaped by urban activities, in the integrated picture required to maintain some control on city sustainability and resilience.

2.2. The Multilayer Dynamics of the City. Tightly related to the modeling of the urban backbone is the analysis of its flows. In fact, the city is a complex system consisting of geographic areas which integrate local flows of goods and people within the overall urban ecosystem. It is not surprising that modeling and understanding both individual and collective human mobility patterns [47, 49, 75, 116] play an important role in our understanding of the city: the identification and quantification of functional patterns, for example, daily mobility motifs [117], can be used to drive the development of transportation systems and enhance urban infrastructures [118].

The abundance of available urban data is already being used to better understand one of the most important urban problems: traffic congestion. Only recently the traditional assumption that people follow the minimum cost path [121] has been challenged [122], showing that routing mechanisms accounting for the complexity of the city have
huge potential in mitigating traffic [53], especially in hypothetical smart cities [123]. Stochastic theories of urban growth, validated on US and OECD empirical data, show that congestion shapes cities, revealing intriguing relationships between mobility patterns and scaling laws, such as the dependence on population size of the total number of miles driven daily, the total length of the road network, the total traffic delay, the total consumption of
gasoline, the quantity of CO₂ emitted, and the relation between area and population of cities [70]. Remarkably, those results highlight the unsustainability of cities whose transportation infrastructure mostly relies upon traffic-sensitive modes, despite polycentrism [70]. The feedback between the urban backbone and routing systems is responsible for the emergence of congestion which temporarily degrades the functionality of the city, reminiscent of the slower-is-faster effect [124] and of the Braess’s paradox [125], further supporting the hypothesis that a complexity science perspective, at the crossroads of multiple disciplines, is required. Despite its emergent nature, multilayer modeling [126] and analysis [127, 128] are enhancing our ability to anticipate congestion phenomena, at least from a theoretical perspective.

As for structure, also urban and interurban dynamics can be understood in terms of interdependent processes. This is the case, for instance, of human movements and epidemics spreading, with integrated models proposed for large-scale mobility [131] between cities, based on different sources of human-generated information, from commuting flows [132, 133] to mobile phone data [134–136] and georeferenced social media [137]. Similarly, mobile phone data [138] and credit transactions [78] have been used to model and predict human mobility within a city, showing that predictive performances improve when information about social patterns is accounted for [139–141]. Social multiplexity inferred from the intersection of popular online platforms revealed that users connected on both platforms tend to have more similar neighborhoods, as well as more

Figure 3: Interdependent urban networks. Illustration of complex relationships among different subsystems and infrastructures [119]. Connectivity patterns include feedback and feedforward paths, as well as branching topologies, whose structure and dynamics are better understood within the framework of multilayer systems [102, 103, 105, 106]. Figure reproduced with permission from [120], all rights reserved.
similar social and spatial properties on both platforms with respect to users connected on just one layer [142]. Recently, human flows with recurrent mobility patterns within the city of Medellin in Colombia have been stratified by socioeconomic classes characterizing the city, to unveil the geographic location of patches triggering the epidemic state at the critical point of the process. Remarkably, those patches depend on the social mixing between classes and mobility [129] (see Figure 4). This result is extremely relevant for policy and decision-making, which cannot overlook the knowledge of the conditions under which such types of critical regimes are expected [130].

The analysis of critical properties provides the methodological baseline for understanding and quantifying the resilience of the city. The multiplex structure of London’s public transportation network enhances its robustness to random failures of single stations and of entire routes. Analysis of the empirical distribution of check-ins and checks-out shows that passengers travel along fastest paths in a network affected by real disruptions, offering a basis for data-driven policies to enhance the navigability of the city [109]. The spatial constraints play a crucial role for critical properties: in contrast to other systems which are not embedded in space, interdependent urban networks are rather sensitive to failures, and abrupt collapse can be driven by any small fraction of interdependent nodes [143].

Multilayer modeling is therefore fundamental for better understanding city resilience and its complex, interdependent structures and dynamics. Urban ecosystems are, in fact, the result of growing networks which are interconnected and coevolving, inducing strong correlations across layers that can alter their response to social, economic, and environmental processes [144].

3. An Integrated View of Complex Urban Systems

3.1. The Cognitive Challenges of Urban Complexity. As already remarked, urban environments are defying not only from the viewpoint of the modeling of urban environments and of their structural interdependencies but also in terms of the challenges they pose to human cognitive systems [145]. For them to be useful, we need to integrate the insights deriving from the analysis of urban dynamics into mental models that enable us to represent, enact, and assess our strategies of navigation, utilization, and governance of the urban space. Such a feature of modeling creates a fascinating parallel with literary fiction [146] and contributes to explaining why narratives have always traditionally been a key tool to conceptualize, represent, and communicate the complexity of urban environments [147]. This is true not only for urban designers and policy-makers but also for local communities and citizens alike. This is why urban cognition is a key frontier of knowledge transmission and application of our findings from research on city complexity.

It has indeed long been recognized that humans have intrinsic cognitive limits in processing information [148]. Given the increasing complexity of highly urbanized cities, such limits have now become a challenge in the interaction of citizens with the urban environment [149]. The concept linking individual experiences to the increasingly complex circumstances of urban life is that of cognitive load [150]. A person’s brain can be overloaded when there are too many inputs to reckon with or when sequences of inputs hit so fast that a new input is still being processed when the new one arrives. Under overload, individuals adapt their behavior by changing priorities and recurring to simplifying choice heuristics.

Redefining traditional tools such as maps to deal with the complexity of urban environments is a key step in this process. Recent studies on visual search strategies [151–153] have demonstrated that the time needed to find a route in a transportation network grows with the complexity of its map, with a transition in search strategies from following metro lines to the scattering of eye focus all over the map [152]. A similar transition from directional to isotropic random search has been observed for visual search of hidden objects when one increases the number of distractors [151]. The ability to manage complex “mental maps” is thus limited, and only extensive training on spatial navigation can push this limit with morphological changes in the hippocampus [154]. Therefore, transportation network structures may be too confusing, requiring to wade through too much detail to figure out whether the service is useful [155]. To measure the cognitive load associated with the visual search of a route in transportation networks, an information perspective has been proposed [111] to quantify the difficulty to navigate them. Using a measure of “information search” associated with a trip that goes from one route to another [156], it has been possible to characterize our difficulty to navigate in a public transit map to identify and measure the cognitive load. To overcome such limits, a “fractal” approach to cartography is in a sense an obvious move, but what is less obvious is the “just in time” adaptive attitude that it commands upon planners, interest groups, and citizens [157].

Transferring these kind of ideas from specialists to nonspecialists is difficult and calls for a profound revision of individual mindsets, not to speak of mental models, and especially so if such knowledge has to become useful and applicable in specific problem-solving contexts [158]. This is an especially burning issue in view of the deliberation processes that support public decision-making in modern democracies. The appeal to oversimplified, and outright incorrect, solutions to complex urban challenges may be especially tempting for politicians and policy-makers when citizens are basically unable to grasp the subtleties of urban policy dilemmas and their implications for their own interests [159]. This can only be balanced by integrating such knowledge into the practical experience and local capability building processes of citizens [160]. But this “pedagogical” change cannot happen without a substantial redefinition of the professional culture of planners and urban experts of all sorts [161].

The ultimate sense of the challenges posed by applying a complexity science perspective to urban systems is that a city is a complex system emerging from the collective behavior of individuals who are themselves complex systems. This
multiscale complexity feeds back in terms of cognitive constructs into the minds of the very individuals that make it exist [162]. Multiscale complexity is common to all large-scale social phenomena, from markets to States to online communities. However, the fact that the city is the socio-spatial context that mediates most human experiences makes it especially sensitive and to some extent fundamental to most other forms of social phenomena. Not incidentally, as noted by Portugali [162], Prigogine in his Nobel lecture indicated the city as the natural metaphor of social complexity. In a sense, then, the city’s capacity of functioning as a complex system also depends on how the mental representations of the city, elaborated by its inhabitants, influence their own choices and behaviors and how this in turn reflects into certain spatial and behavioral patterns at the macro-scale. It is therefore fundamental to think of the city in terms of systems ecology and to develop new ecologically informed models of sociospatial cognition that mediate between the subjective and objective levels of urban experience. In a sense, then, the city’s capacity of functioning as a complex system also depends on how the mental representations of the city, elaborated by its inhabitants, influence their own choices and behaviors and how this in turn reflects into certain spatial and behavioral patterns at the macro-scale. It is therefore necessary that these goals become a relevant item in the urban policy agenda.

3.2. Planning and Complexity: Adaptive Urban Policy-Making. From the point of view of planning theory, the main takeaway of a couple of decades’ experience of dealing with cities from a complexity-focused perspective is the necessity of a radical redefinition of planning practices as an interplay between institutionally driven design and urban self-organization principles [170]. The stakes are high. Planning practices are mainly justified to secure fair and inclusive access to urban resources. At the same time, cities may look very different when seen from different vantage points corresponding to different levels of benefit and privilege [171]. The consequent danger is that the management of urban complexity may also function as a convenient smokescreen to decline political responsibility and accommodate certain vested interests at the expense of others [172], irrespective of the declared intentions. However, to the contrary, it is also possible that seriously accounting for the self-organization dimension of urban processes into planning practices may function as an enabler of community initiative and active participation [173]. In essence, taking advantage of the lesson of complexity in planning means creating the context for decentralized,
collaborative action at least as much as prescribing centralized, top-down forms of planning [174], thus laying the premises for a reflexive approach to planning [175].

The real issue from the planning perspective is therefore how to empower citizens and local communities to play an active role in this cocreative process and to be able to conceptualize, promote, and assess collective action in the pursuit of common interests [176] and in the exercise of their right to the city [177]. One possible solution is rediscovering rituals of collective pleasure in public spaces as a foundation for a common social intentionality [178], in the spirit of a repurposed collective action aimed at social change along the lines of Gramscian thinking, consequently turning the urban environment into a “playable” public domain [179]. Many collective practices of public art in the public domain are exploring possibilities and breaking new ground in this direction, by engaging local communities to take an active, propositional attitude towards urban complexity in terms of shared agency and not of passive delegation to planners and high-level stakeholders [180]. In this perspective, we look with great interest at gamification-based approaches insofar as they are intended not as domesticating formats for passive engagement and manipulative conditioning but rather as a smart deployment of collective resources and talent [181, 182] and of self-empowering behavioral change [183], which appeals to expressive rather than to instrumental motives [184]. In this sense, gamification has shown promise as a surprisingly practical and flexible tool to pursue complex technical and sociopolitical collective goals [185], due to two main characteristics: its intrinsic narrative potential and appeal, which as already remarked may effectively function in conveying detailed contextual information in complex urban environments [186], and its endogenous scoring metrics that, while providing direct feedback as to the efficacy of certain individual or collective actions, also build motivation and engagement [187].

In this regard, the pandemic crisis and the consequent necessity to rely upon contact tracing apps to mitigate the spreading of the contagion can be regarded as a missed opportunity. In countries whose local culture is characterized by a strong civic sense and on the emphasis on collective responsibility and duty as well as by a mature stage of the digital transition, such as many Far-Eastern societies [188, 189], contract tracing apps have been massively adopted and have functioned well despite the inevitable concerns on privacy [190], keeping the level of contagion and human losses remarkably low [191] if compared to that of more individualistic societies [192] where contract tracing has been implemented relatively late and dedicated apps have been adopted by a minority [192]. Even in a situation of high personal risk, reliance on responsibility and fear as incentives to app adoption has not proven effective in mobilizing citizens to behave prosocially, despite a generalized declared willingness to download contact tracing apps across most countries [193]. On the other hand, the apps themselves were merely designed, following a purely functional logic, as carriers of information and control devices in a moment in which people felt a sudden, heavy burden from the almost complete disruption of their previous social life. Maybe considering social incentives to communication and exchange that were put particularly under stress during the pandemic crisis could have been a key to a stronger motivation to adopt the app by choosing suitable, motivationally salient cues. For instance, people could be shown how the probability of lifting up social restrictions would be affected by the rate of adoption of the app and consequently how that probability could change for every extra thousand people adopting it, inviting them to contribute by adopting the app themselves and convincing their acquaintances to do the same. Or, alternatively, in terms of appeal to responsibility, people could receive a constantly updated estimate of the probability to be infectious given their history of social contacts, so as to discourage risky behavior through appeal to regret [194].

4. Outlooks

Most applications of complexity science to urban issues tend to focus, for understandable reasons, on specific dimensions of the urban environment. Clearly, studying the structure and the dynamical evolution of urban utilities networks, transportation systems, or resource flows is already more than enough to challenge our modeling abilities. However, if we want to fully acknowledge the implications of the previous analysis and in particular the necessity of integrating the three dimensions of modeling, cognition, and governance, we need to look at urban environments not from a sector-specific but from a system-wide perspective that cuts across different sectors and dimensions of the urban fabric. Such an integrated framework to understand cities through data-powered tools sits at the edge between many different disciplines such as statistical physics, social sciences, economics, digital health and wellbeing, and engineering, to name only a few ones. The state-of-the-art applications reviewed above provide us with a promising benchmark by characterizing urban complexity in terms of multiplex networks and possibly point to the multiplex city as a computationally appropriate and conceptually scalable [195] representation of urban complexity [196], which will allow the development and deployment of new urban governance strategies, as well as the redesign of old ones [54].

Urban environments clearly raise specific health issues, which are becoming of increasing relevance once we are reminded that, according to recent estimations, up to 70% of the world’s population will be living in urban areas by 2050 [197], and even if in the postpandemic scenario this trend could be subject to changes in the medium-long term, cities will inevitably keep on playing a pivotal role in future economies and societies [198]. This extraordinary scale of urbanization is affecting both the environmental impact of human activities and the nature and scope of human health and wellbeing issues by posing new adaptive challenges [199]. The Ottawa global health milestones agreement Charter puts the concept of the promotion of healthy practices and lifestyles at the center of a socially and financially sustainable approach to public health [200], where, as formalized by the salutogenesis [215] paradigm [201], the
focus shifts from the causes and reinforcing conditions of diseases to those that favor and preserve health. The crucial step that enables individuals to successfully adapt to the stressful conditions of the urban environments of socioeconomically advanced societies is to enhance their capacity to cope with such critical factors by improving resilience at all social scales [202]. The urban dimension of salutogenesis clearly becomes all the more central in the postpandemic scenario and will likely leave a deep trace in future urban policies [203].

Although these principles are today widely recognized, finding viable approaches to their implementation in urban environments is not easy task. However, a clever use of the incentive systems connected to digital participation may provide, as discussed above, an innovative platform to motivate people to pursue healthy habits and lifestyles while promoting other public interest goals at the same time. To address this issue, it is crucial to devise data-driven solutions for the promotion of innovative practices of urban health and wellbeing to promote a salutogenic approach to urban space through proactive access to, and use of, the varied mix of wellbeing-enhancing assets in the public domain. Such promotion may be accomplished by means of gamified participatory practices which deeply engage people in the active pursuit of integrated salutogenic goals [204, 205].

Strategies to encourage citizens to pursue health promotion goals by actively engaging with their urban environment include the following:

(i) Fostering inclusiveness and relationship building instead of social stratification and segregation [206]
(ii) Promoting cultural participation by arousing curiosity about, and involvement with, urban cultural heritage and poorly known landmarks [207]
(iii) Creating new opportunities to adopt healthy habits, including increased mobility, while reducing depletion of natural capital [208, 209]

To understand to what extent an integrated approach that takes culture as a driver of urban change may become a powerful basis for a systemic view of urban functioning and change which invites citizens to be proactive in the pursuit of their own quality of life and wellbeing, two aspects of special importance are sociability and cultural experience. Sociability is an issue of increasing relevance in the public health agenda, as loneliness is now recognized as a serious public health problem [210]. Cultural experience is increasingly connected to health and wellbeing issues by a rapidly growing literature [211, 212]. The eudaimonic [216] approach to wellbeing provides a conceptual framework encompassing all three dimensions within the overarching salutogenic framework, as it postulates that a harmonic integration of different spheres of human existence best fits human sociopsychological development [213] and promotes health more effectively than narrower approaches [214]. In an eudaimonic perspective, the urban environment becomes an extremely rich and potentially stimulating playground on multiple dimensions: as a natural context for urban mobility, as an elective space of sociability, and as a theater of cultural experiences. Breathing new life into the postpandemic urban fabric will be not only a problem of technical efficiency and provision of adequate services tailored to the new social demands and needs but also an exercise of collective sense-making. In the forthcoming scenario, it will therefore be crucial to directly involve citizens in collective problem-solving processes to codesign the postpandemic city, both to build upon collective intelligence and to rebuild social cohesion after a long, critical period of isolation and social alienation and as a powerful driver of widespread adoption of prosocial practices, attitudes, and new habits without which the cities of the future might become less and less livable. This is in particular true in view of the strong call that urban life in postpandemic cities will make for widespread citizens’ responsibility and attention to safety concerns and public health norms compliance. Taking this forced start of a new urban cycle as an opportunity to bring urban sociability at the core of urban policies through the active contribution of citizens may lead to new, emergent forms of social cooperation whose collective benefits may be directly measured and, consequently, may further reinforce individual and social motivations to pursue a more sustainable, cohesive route to urban development. It is unlikely that the new challenges will be effectively tackled through business-as-usual policy approaches [2], and there will be the need to refresh what looked like the “default” conviction only one year ago, namely, that cities are the undisputed, trend-setting centers of gravity of social, economic, and cultural life. With the increasing diffusion of remote smart working, the choice to live in a city will be less instrumental and will have to be motivated by the intrinsic richness of opportunity of the urban environment as to the promotion of socially rewarding encounters and mind-opening, stimulating experiences. In the postpandemic scenario, sociability habits and patterns in the urban space will have to be rebuilt almost from scratch, and new demands for social connection and meaningful interaction will inevitably arise. Likewise, the postpandemic city will have to rethink and repurpose its public space as a theater of collective expression and inspiration. We need to be ready to tackle these challenges through a clear vision of the multilayered complexity of urban dynamics, with a strong focus on the livability and liveliness of urban environments. However important, these dimensions of sense-making have been marginally dealt with by complexity models and analyses. The time has come to integrate them into a full-fledged approach to urban systems.

**Data Availability**

No data were used to support this study.

**Conflicts of Interest**

The authors declare that they have no conflicts of interest.

**Authors’ Contributions**

All authors equally contributed to this work.
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