

# Walkability Indicators in the Aftermath of the COVID-19 Pandemic: A Systematic Review

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**Abstract:** Walkability indicators are a pivotal method to evaluate the role of the built environment in people's decisions regarding active mobility, contributing to help promote more sustainable and equitable regions. Currently, there is a lack of literature describing the trends and findings from recent studies, and how research on this topic has responded to COVID-19. In this work, we apply the PRISMA methodology to perform a systematic literature review of studies that employed walkability indicators to uncover associations between walkability and different topics, such as physical activity, health and mobility, in the aftermath of the pandemic. Results show that the growth rate of studies in this period almost doubled compared to the three years prior to the outbreak. We explore the main subjects addressed in recent literature and found that most studies resorted to well-known indicators, predominantly to Walk Score<sup>®</sup>, while fewer works developed and applied new indicators. General findings point to a positive association between walkability and physical activity, overall health and active mobility. Moreover, in recent works, walkability's relationship with environmental metrics is not clear and there is a lack of studies linking walkability indicators to COVID-19. More than two years after the worldwide outbreak of the COVID-19 virus, our work provides an effective resource for researchers to understand recent trends in studies employing walkability indicators.

**Keywords:** walkability; indicators; sustainability; urban planning; transportation; public health; COVID-19; systematic literature review



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## 1. Introduction

Active modes of transportation have become one of the beacons of Sustainability [1–3]. In the 2030 Agenda for Sustainable Development, the United Nations has determined that countries and regions should provide for safe, affordable, accessible and sustainable transportation for their citizens [4]. Hence, among other resolutions, promoting walking as a viable means of transportation has become the focus of sustainable development policies [5–7].

Active mobility can contribute to the reduction of the number of circulating private vehicles [2]. Consequently, this can help alleviate traffic congestion, as well as lead to a reduction in air pollutant emissions, thus promoting sustainability by reducing environmental degradation [8–10]. Likewise, walking is the most democratic method of mobility, so guaranteeing access to critical services for pedestrians also guarantees greater equality in the use of these services [11]. Moreover, pedestrian friendly public spaces promote physical activity [12,13], social interactions and recreation activities [14], resulting in benefits for physical and mental health [15], and social well-being [16].

Hence, it is crucial to identify and understand the factors influencing travel decisions and walking behavior and, in particular, the features of the environment that increase walking as the choice of mobility.

In that respect, the concept of walkability refers to how friendly the built environment is for pedestrians, considering both street conditions and the proximity and distribution of key services in the surrounding area [17,18]. This concept has been studied in different

works that try to objectively measure the characteristics of the environment [19–21], generally through the application of a walkability indicator—a composite score encompassing the main features that contribute to walking [22,23]. These indicators allow assessment of the impact of the built environment on different phenomena [24]. Among other topics, researchers have studied the association between objectively measured walkability and physical activity [25–28], health [29–32], active commuting [33–35], pedestrian safety [36–38], pollution [39], and environmental well-being [5]. Hence, walkability indicators represent an efficient tool to support policy makers and urban designers in decision-making, by informing and gauging actions and initiatives [40,41].

Recently, the COVID-19 outbreak has brought more awareness to the general public about the topic of walkability [42]. Population lockdowns and social distancing have disturbed and altered mobility patterns to prevent and restrict the spread of the virus [43,44]. As a consequence, people resorted more often to active modes of transportation, such as walking and cycling [45–49]. In addition, people are spending more time in their residential area [50–52], which increases the importance given to the effect the built environment and urban design have on their mobility and overall quality of life [42]. Despite this, there is a lack of research systematizing the recent trends and findings of studies that resort to walkability indicators, and how they have been affected by the pandemic [1,49].

Therefore, the present study identifies and reviews works that employ walkability indicators and link walkability to objectively measured phenomena that are relevant to sustainable development. To fill the research gaps, our aim is to answer the following research questions:

- RQ1: What are the main subjects investigated in recent literature on walkability indicators?
- RQ2: What are the most used walkability indicators in recent literature?
- RQ3: What are the main findings from the recent literature?
- RQ4: How do these findings compare to those before the COVID-19 pandemic?
- RQ5: How do recent works on walkability indicators reflect the topic of the COVID-19 pandemic?
- RQ6: What are the main gaps identified in the reviewed literature?

We performed a systematic literature review following the PRISMA [53] methodology and conducted exploratory and bibliometric analysis of the selected works on the topic of walkability indicators in the aftermath of COVID-19. This study contributes to the literature by (1) identifying influential authors, papers, journals, and countries; (2) systematizing and synthesizing recent findings, methodologies, and results used in selected articles; (3) investigating the association between walkability indicators and the subject of the pandemic; and (4) identifying some of the limitations of recent literature.

The rest of this study proceeds as follows: Section 2 details the methodology employed in this research. In Section 3, we describe the results of the exploratory and bibliometric analysis and discuss and answer the research questions. Finally, in Section 4 we summarize and conclude the work conducted.

## 2. Methodology

This systematic literature review aims at compiling, interpreting, and evaluating the methods and findings of relevant scientific work regarding walkability indicators and the way these are associated to different phenomena in the period after the outbreak of the COVID-19 pandemic. For that purpose, we resort to the well-established systematic literature review framework Preferred Reporting Items for Systematic reviews and Meta-Analyses (PRISMA) [53]. PRISMA is a four-phase framework for identifying, screening, and validating the eligibility of relevant documents for a research topic. In Figure 1, a diagram is displayed outlining the process through which relevant works were selected for inclusion in this study.

Identification	Records identified through Web of Science database (n = 572)	→	Records limited to journal articles, English text, and the period between 2020–2022 (n = 236)
	↓		
Screening	Records for abstract screening (n = 184)	→	Records excluded: I. do not develop or implement a walkability indicator II. one duplicate paper (e = 63)
	↓		
Eligibility	Full-text screening for eligibility (n = 121)	→	Records excluded: I. Do not build a walkability indicator from objectively measured built environment characteristics II. Do not use external variables/phenomena to validate indicator III. One paper about indoor walkability (e = 57)
	↓		
Included	Studies included in the review (n = 64)		

**Figure 1.** Application of PRISMA four-phase flow diagram (n = included e = excluded). Adapted from Vong et al. [54].

### 2.1. Database Search

Web of Science (WoS) [55] was the chosen online database for literature search, as it is one of the most used and well-known databases for that purpose and for supporting bibliometric analysis, having a broad coverage of most scientific work sources for different subjects. The user-friendly interface of WoS allows to define a search query based on keys words within the article content (title, abstract and keywords). We searched for the following keywords in the documents abstract: “walkability”, “indicator”, “score”, “index”. The word “walkability” ensured that all retrieved documents addressed this topic, while “indicator”, “score” and “index” guaranteed that the documents obtained used an indicator to measure walkability.

### 2.2. Eligibility Criteria

The research timeframe was set to the period after the outbreak of the COVID-19 pandemic, comprising the years of 2020, 2021 and 2022 (until June 2022), as our purpose was to analyze the patterns in literature during and after the health crisis. To guarantee the scientific credibility of this review, only articles published in peer-reviewed journals from quartile one and two, according to Journal Citation Report (JCR) [56], were selected. Hence, conference papers, review papers, books, book chapters and research notes were removed from the analysis. Furthermore, we filtered for English-written documents only. The resultant WoS search query was the following: (AB = walkability AND AB = (score OR indicator OR index) AND DOP = (2020-01-01/2022-01-05) AND DT = (Article) AND LA = (English)).

### 2.3. Screening

After the application of these criteria, from the initial number of 572 documents, a set of 236 works were identified. Furthermore, we proceeded to screen each article's abstract to ensure that all retrieved papers were relevant for the topic under study. From this screening process, we removed 63 articles that, despite approaching the subject of walkability and using certain indicators, did not actually implement or develop a walkability indicator. Lastly, an integral screen (full-text screening) led us to remove 57 papers that did not fit the purpose of this review because: (1) they did not build a walkability indicator from objectively measured built environment characteristics; (2) they did not use external variables/phenomena to validate the built indicator; (3) one paper was measuring indoor walkability.

### 2.4. Bibliometric Analysis

A bibliometric analysis is a method for exploring and analyzing scientific data. It is used to uncover trends, collaboration patterns, research terms, and the overall intellectual structure of a research domain [57]. It does so by analyzing the data at different levels, from authors, to documents, publishers, journals, countries, keywords, and citations. In this work, we conduct a bibliometric analysis to derive conclusions regarding the main developments on walkability indicators and their context in the scientific intellectual environment. We run different forms of analysis, namely co-authorship, co-occurrence, citation, and co-citation analysis. To do so, we resort to VOSviewer [58], an open-source data-visualization software used to construct bibliometric maps based on literature datasets [54].

## 3. Results and Discussion

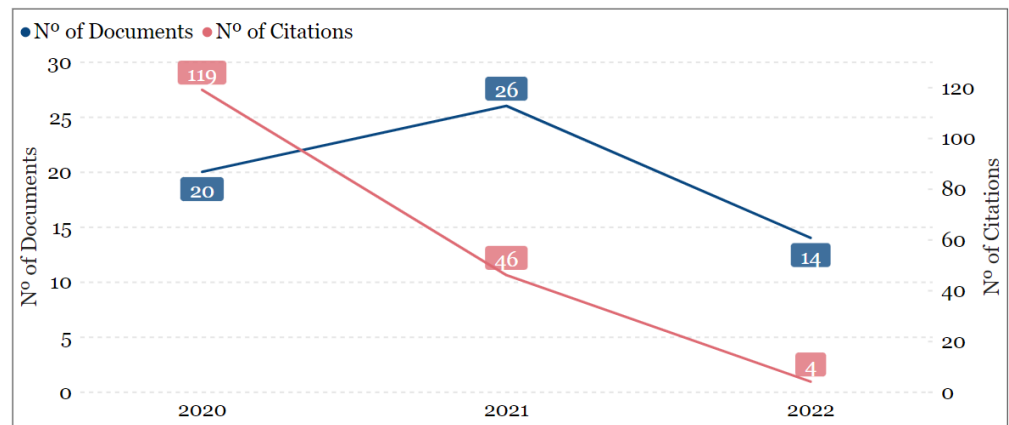
This section describes the work conducted and results obtained—first, for an exploratory analysis of the reviewed documents, in which we analyze the number of documents and citations per year, country, publisher, journal, author and document; and second, for a bibliometric analysis. The, this section answers and discusses the research questions specified in the introduction of this review.

### 3.1. Exploratory Analysis

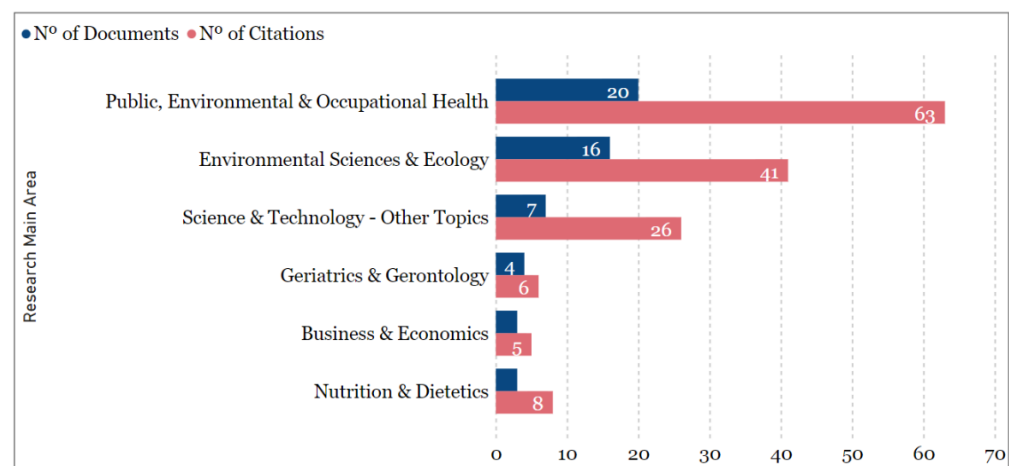
We start by analysing the yearly evolution of documents and citations during the period under study. From Figure 2, we can see that 2021 was the year with more publications, with 26, followed by 2020, with 20 and 2022, with 14. The timeframe considered for 2022 only encapsulates the first 5 months of that year, while both 2020 and 2021 have the entirety of the period evaluated. Thus, from the current trend it is likely that, by the end of 2022, this year will surpass the other two regarding the number of documents published. Naturally, the number of citations by year displays a different pattern, where 2020 stands out as the year with more citations, with 119, documents from 2021 have a total of 46 citations, and 2022 documents have 4 citations already.

Next, Figure 3 displays the number of documents and citations per WoS research area. "Public, Environmental & Occupational Health" is the area with most documents (20), followed by "Environmental Sciences & Ecology" (16), while "Business & Economics" and "Nutrition & Dietetics" (3) are the areas with less documents. Regarding citations, the pattern found is similar, except for "Nutrition & Dietetics", which is the fourth area with more citations (8), having, on average, 2.7 citations per article. Regarding this metric, "Science & Technology" is the area with more average citations per article, with 3.7 citations.

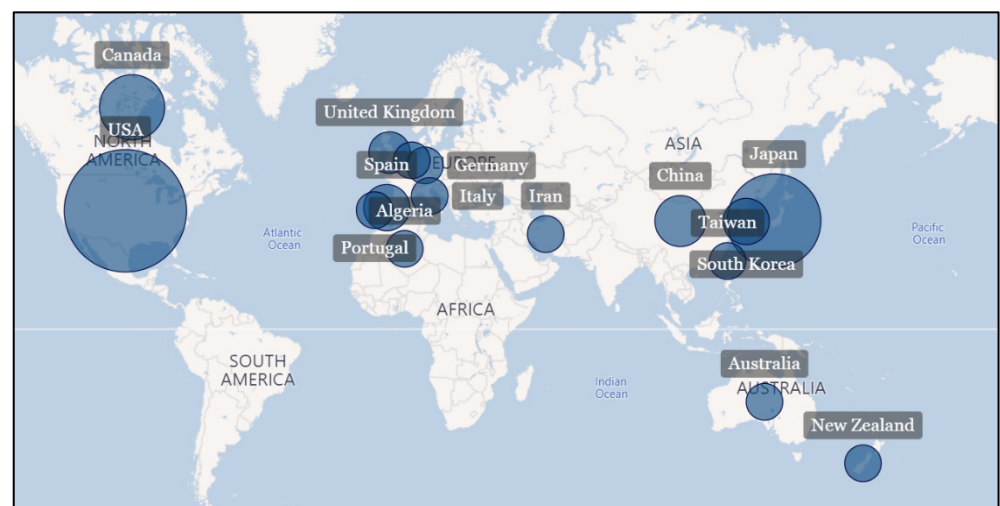
The distribution of documents per country where the indicators were implemented (Figure 4), shows that the USA is the country with more publications, with 19 documents, followed by Japan and Canada, with 13 and 7 documents, respectively. China, with 4 documents, and South Korea and Spain with 3 documents each, close the top 5. One walkability indicator was employed in Algeria, Australia, England, Germany, Iran, Italy, the Netherlands, Portugal, Scotland, Taiwan, and New Zealand.



**Figure 2.** Number of documents (blue) and number of citations (red) by year. For the year 2022, only the period until May is considered.



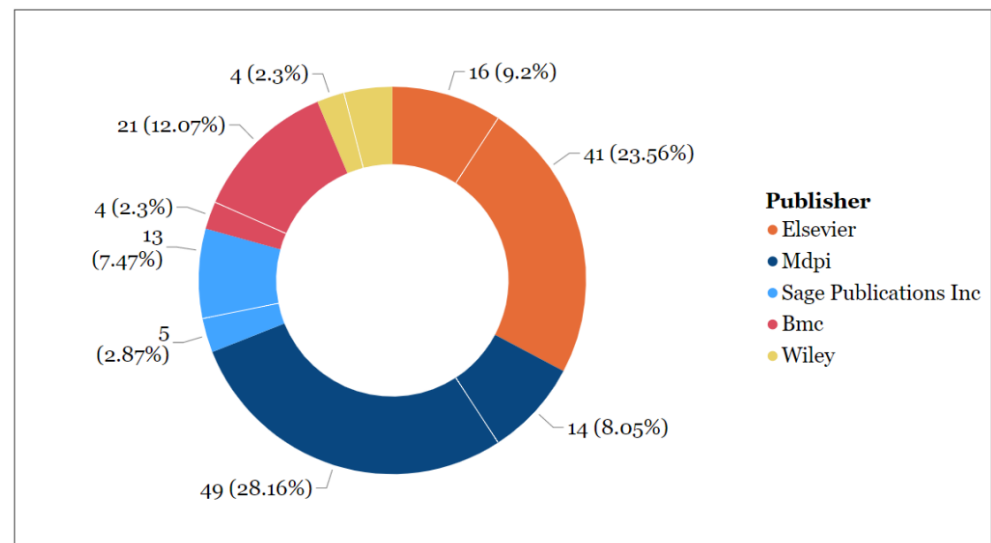
**Figure 3.** Number of documents (blue) and number of citations (red) by research area (top 5).



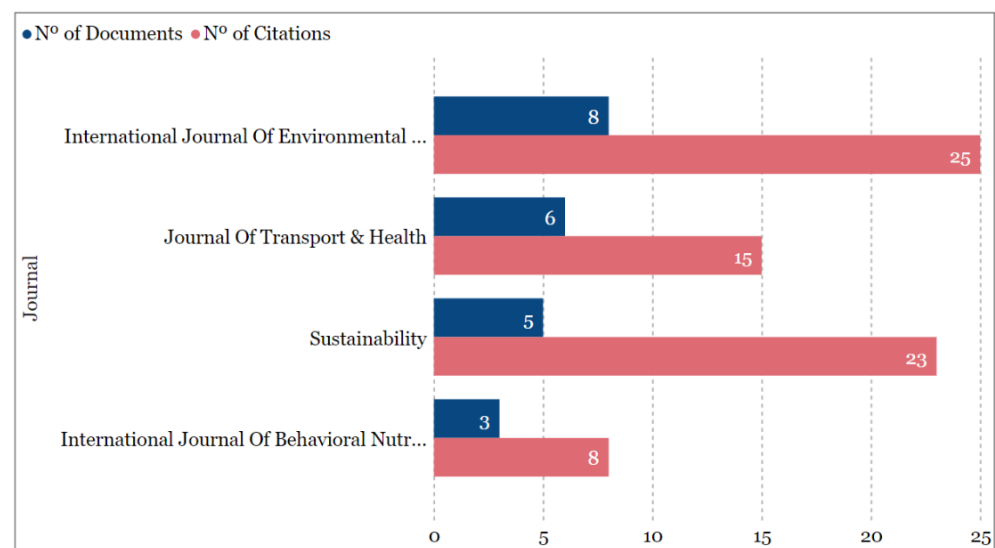
**Figure 4.** Number of documents by country of author.

Looking into the previous metrics by publisher (journal) (see Figure 5), “Elsevier” is the publisher with more documents (16) and the second one with more citations (49). “MDPI”, with 14 documents and 49 citations, is the second publisher with more articles and the one with more citations. With regards to Journals (Figure 6), the “International Journal

of Environmental Research and Public Health” is the one with more publications (8), as well as citations (25). “Sustainability” is the journal with more citations per document, with an average of 4.6.



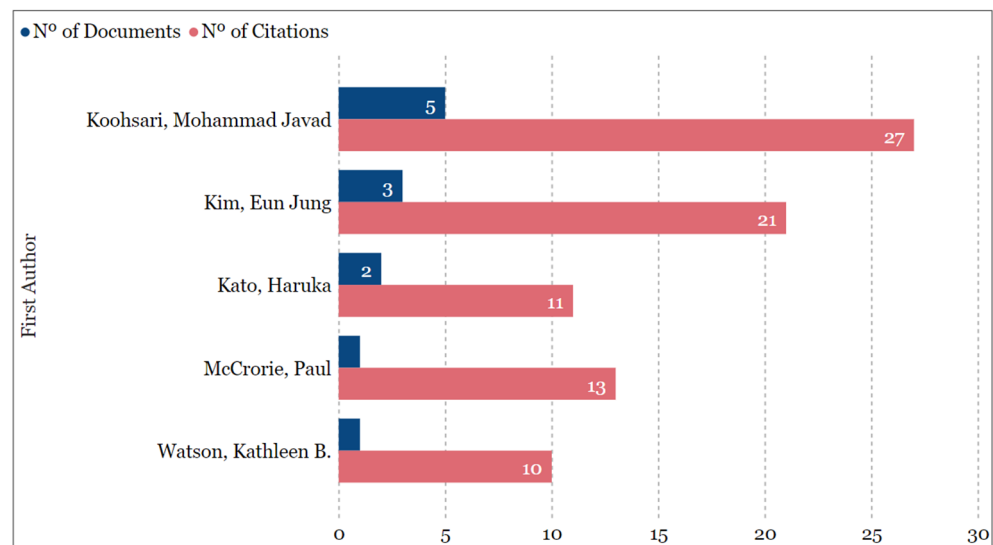
**Figure 5.** Percentage of documents (left of slice) and citations (right of slice) per publisher (top 5).



**Figure 6.** Number of documents and citations per journal (top 4).

As for the authors analysis, Figure 7 compiles the top 5 authors per number of documents and citations. Mohammad Javad Koohsari, from Wasada University in Japan, stands out, with 5 documents published between 2020 and mid-2022, and 27 citations in the same period. Paul McCrorie is the author with more citations per document (13), with only 1 document published and 13 citations.

Likewise, we analyzed the five most cited documents. “The relationship between living in urban and rural areas of Scotland and children’s physical activity and sedentary levels: a country-wide cross-sectional analysis” by McCrorie et al. [28] is the most cited paper, with 13 citations. The indicator, based on Frank et al. [59] and Macdonald et al. [60], was implemented in Scotland to measure the association between walkability and children’s physical activity. Results show that walkability was not associated with any outcome of physical activity among children. Additional information regarding the documents that make up the most cited top 5 can be found in Table A1.



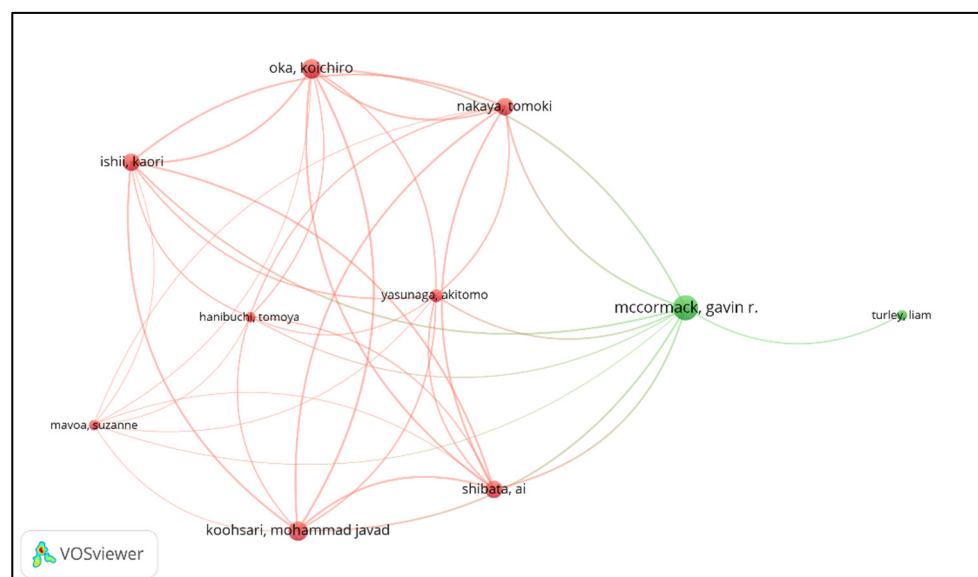
**Figure 7.** Number of documents and citations per first author (top 5 by documents).

### 3.2. Bibliometric Analysis

In this section, we perform a bibliometric analysis over the selected documents for review. We conduct co-authorship analysis, co-occurrence analysis, citation and co-citation analysis.

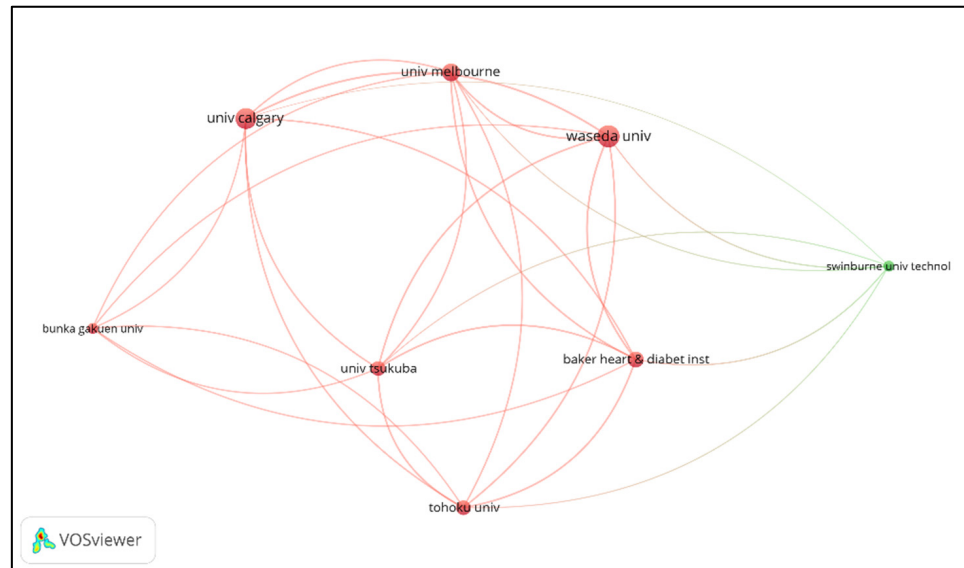
#### 3.2.1. Co-Authorship Analysis

Co-authorship analysis provides information regarding study collaboration between entities, namely authors, organizations, and countries. Figure 8 displays a network map of the full counting of co-authorship, considering a minimum of 3 documents per author. A total of 10 authors were identified, resulting in two clusters of collaborations. The red cluster (left) presents a more complex network, comprising mostly authors from Japanese institutions. The second cluster, in green (right), only contains two authors, both from Canadian institutions. The link between the two clusters is established by Gavin R. McCormack, who has worked with all authors from the red cluster. The collaborations expressed in this map implemented walkability indicators generally to measure the effect of walkability on health and physical activity.



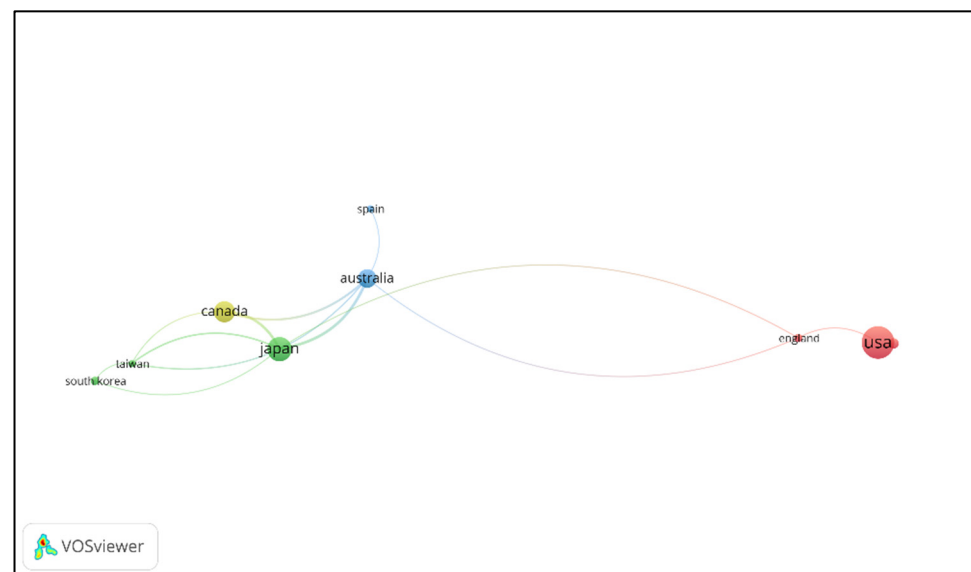
**Figure 8.** Network map of author co-authorship analysis.

With regards to organizational co-authorship (Figure 9), again two clusters were identified. The red one (left) comprises 4 Japanese schools, 2 Australian and 1 Canadian, all of which have developed walkability indicators studies on health and physical activity. The green cluster (right) contains only one institution from Australia. The study focus of documents from this affiliation is the relationship between walkability and urban greenness.



**Figure 9.** Network map of organization co-authorship analysis.

Countries' co-authorship analysis results in 4 clusters (Figure 10). With exceptions, countries are grouped by language and geographical proximity. One cluster, in green, comprises South Korea, Taiwan and Japan. Japan is the main bridge between these and other clusters, namely the yellow one, which considers only Canada, the blue one containing Australia and Spain, as well as the red one, with the USA, England, and China. The relationship between this last cluster and the others is established through England.

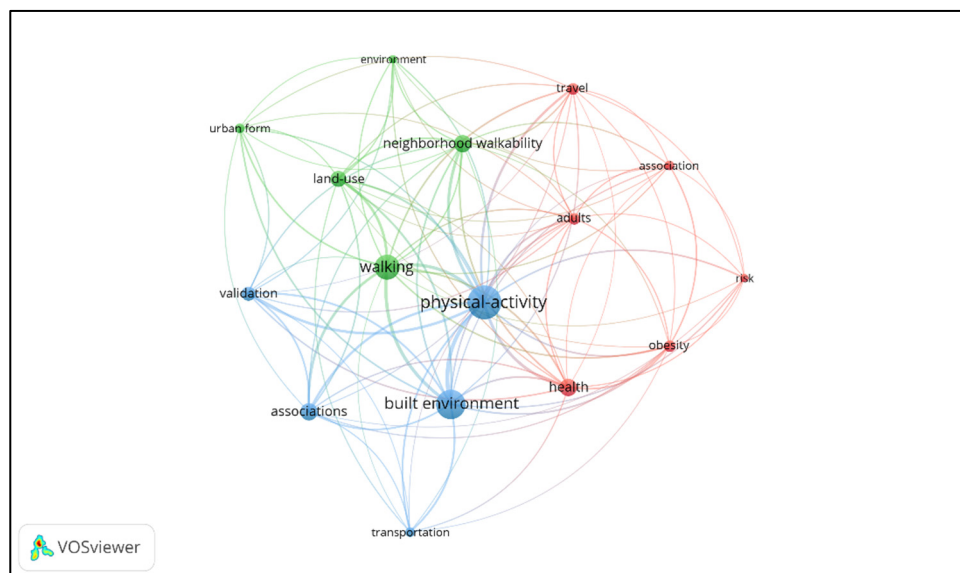


**Figure 10.** Network map of country co-authorship analysis.



### 3.2.2. Co-Occurrence Analysis

Co-occurrence analysis highlights the most common words in the documents' text and their occurrence together. Figure 11 displays a network map of the fractional counting of author keyword co-occurrence, considering a minimum of 3 documents per keyword. The size of the nodes indicates the weight of the keywords, the connections between nodes represent the co-occurrence of keywords, and the thickness of the links and the distance between two nodes represent the frequency of the co-occurrence.



**Figure 11.** Network map of keyword co-occurrence analysis.

Three clusters were identified: in red keywords are related to the subject of health, the node for the word “health” being the one with the biggest size. The cluster in green related mostly with environmental factors, while the blue cluster seems to concern physical activity and its association with walkability and the built environment. Unsurprisingly, physical activity, built environment and walking are the most important keywords identified in terms of occurrence and generally appear together. This is indicative of the topic of most papers, which focus on the importance of the built environment (measured through a walkability indicator) in promoting walking and physical activity.

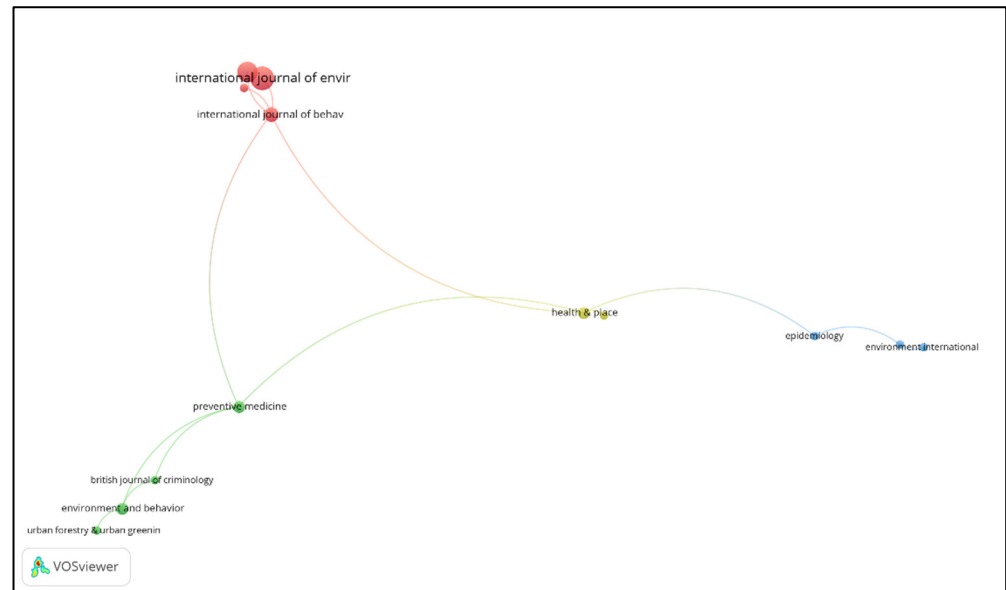
### 3.2.3. Citation Analysis

Next, we perform a citation analysis. This analysis inspects the frequency of citations of authors, organizations, and countries, in which a link between two nodes means a node cited the other.

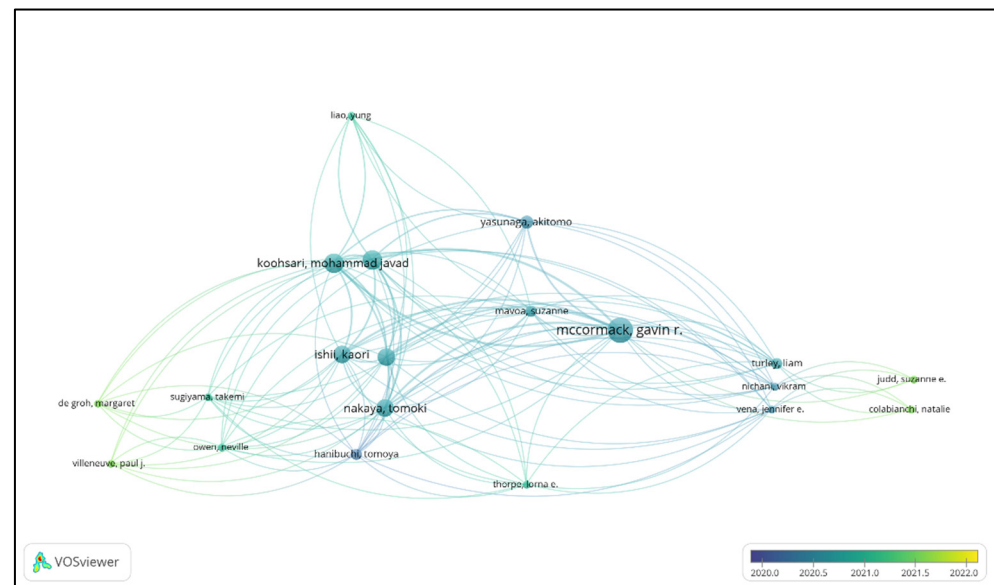
We evaluate source (journal) citations, with minimum number of documents per source set to one. The resulting network map can be seen in Figure 12. A total of 4 clusters were identified. In the blue cluster (bottom right) sources include works regarding relationship between walkability and risk of death. The cluster in yellow (comprising “Health & Place” and “Scientific Reports”) works as the bridge node between all other clusters. Documents published there investigate the relationship between walkability and sedentary behaviors. In the red cluster (top left), documents are generally focused on the correlation between active commuting and walkability. Lastly, the green cluster comprises sources with studies mostly focused on the relationship between walkability and physical activity, or walkability and crime patterns.

For author citation analysis, we selected only authors that had at least 2 documents published and one citation. As well, we added time as a dimension of analysis. Figure 13 displays the map obtained. We can see that, as previously observed, Gavin R. McCormack and Mohammad Javad Koohsari are the two first authors with the most influence in recent

walkability indicator literature and bridge the work from more recent authors (yellow colored on the edges of the network).



**Figure 12.** Network map of source citation analysis.

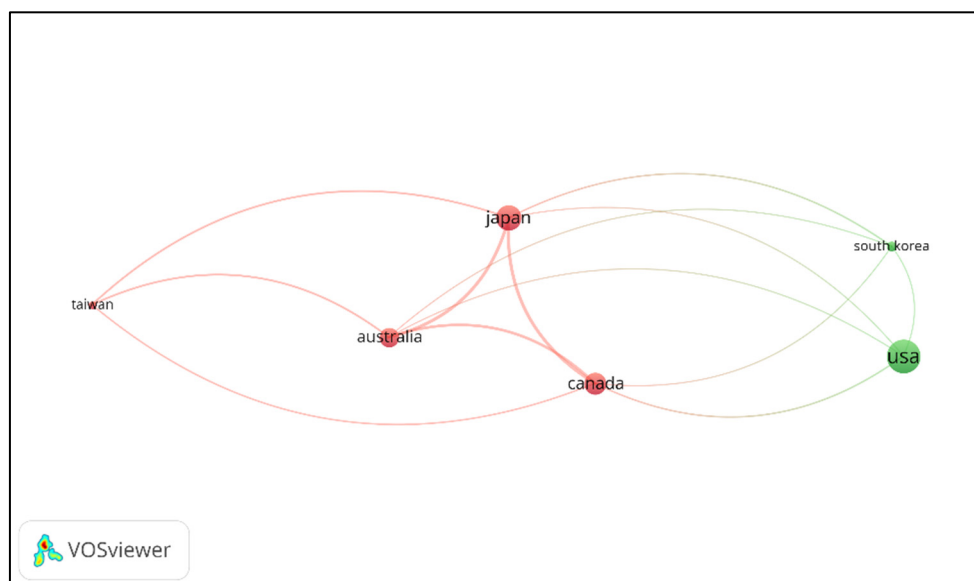


**Figure 13.** Network map of author citation analysis.

Country citation analysis revealed two main clusters (see Figure 14). On the left, red cluster countries' works relate, in general, to the correlation between walkability and health, active commuting and the environment. On the right, the green cluster is composed by the USA and South Korea. Studies from these countries also cover the relationship between walkability and socio-demographic phenomena, like crime, income, ethnicity differences, or real estate prices.

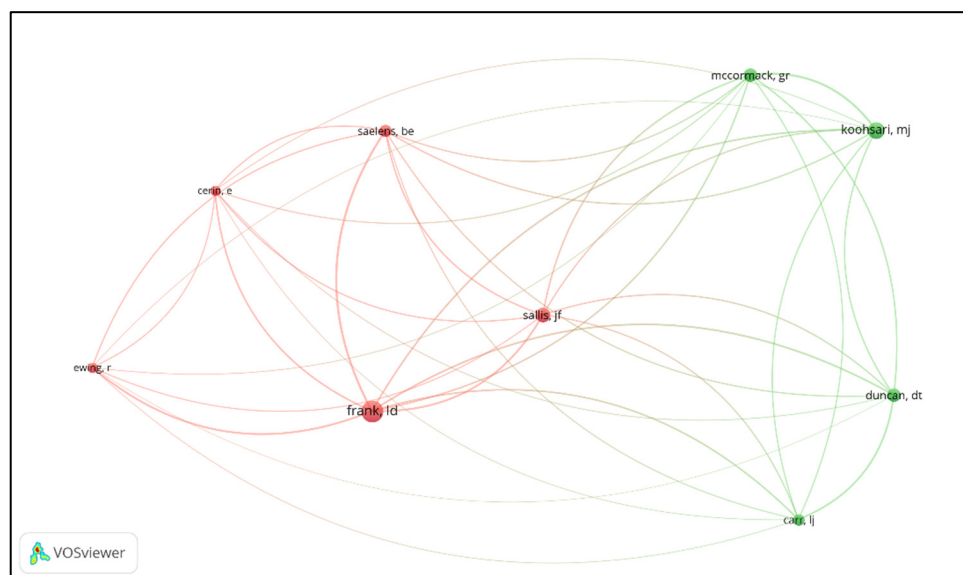
Furthermore, we perform a co-citation analysis, in which the number of documents that are cited together by two nodes is evaluated, and a link between two nodes represents one document that was cited by the two. This type of analysis allows to access what are the most influential works and authors for the latest developments on a topic. We filtered the

analysis for documents that received 10 or more citations and authors that received 20 or more citations.

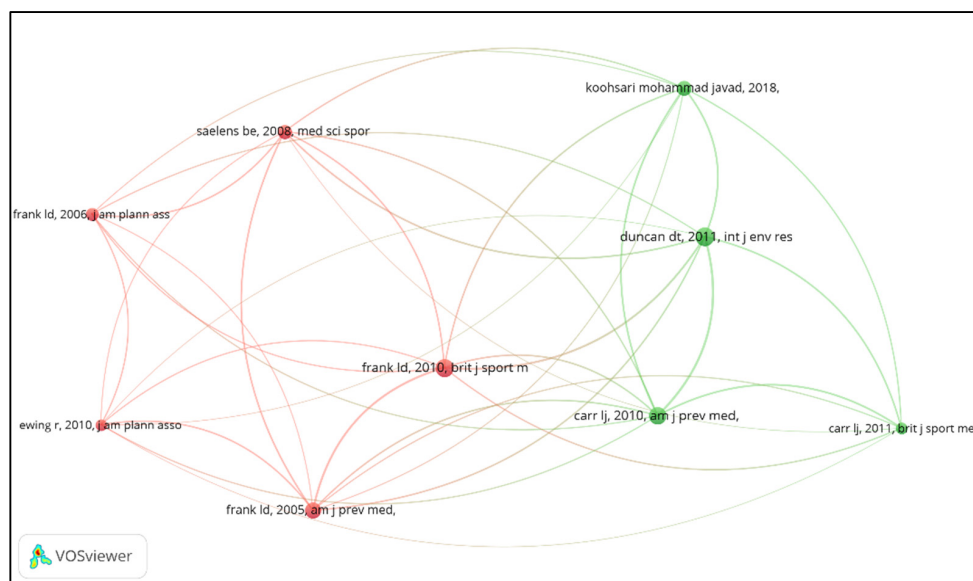


**Figure 14.** Network map of country citation analysis.

From Figure 15, the most influential author is Frank D. Lawrence, with a total of 3 papers cited, released between 2005 and 2010. In terms of documents (Figure 16), the node with the greatest size, and thus the most popular among recent walkability indicator studies, is Duncan et al.'s work [61], which validates the Walk Score<sup>®</sup> [62] for estimating neighborhood walkability in 4 USA cities.



**Figure 15.** Network map of author co-citation analysis.



**Figure 16.** Network map of document co-citation analysis.

### 3.3. Research Questions

#### 3.3.1. RQ1: What Are the Main Subjects Investigated in Recent Literature on Walkability Indicators?

After investigating and evaluating the recent literature on walkability indicators, based on the research issues being addressed we divided the documents into the 4 main categories: Public Health, Transportation, Socio-economic and Demographics, and Environment. On the subject of Public Health, most works study the association between neighborhood walkability and physical activity [26–28,63–66]. Another topic of general interest for Public Health is people’s physical health, manifested through obesity, overweight and frailty, and generally calculated using the Body Mass Index (BMI) or Waist Circumference (WC) [67–71]. Other studies focus on the correlation between walkability and prevalence of diseases or health complications, such as cancer [72,73], hypertension [30,74], cardiovascular problems [29,75,76], diabetes [32] and dementia [31]. Other topics addressed under this subject include sedentarism [77–79] and overall well-being [80].

On the subject of Transportation, most works evaluate the relationship between walkability and walking as a means of active transportation, and can be divided into general active transportation [34,81–83], school commuting [35,84,85], and pedestrian comfort and safety [38,86]. One work also addresses the role of walkability in influencing travel choices for people with disabilities [87].

Regarding Socio-economic and Demographics, topics of interest vary. Walkability has been associated with criminality, in the form of street crimes [37,88,89] and property incidents [40], housing availability and price [90,91], neighborhood residents’ social capital [92] and economic differences [93], personality traits [94], incidence of care insurance needs [95], as well as membership costs and service availability in fitness centers [96].

Lastly, studies under the Environment topic associated walkability with neighborhood greenness [97], ecological footprint [5] and thermal comfort [98].

#### 3.3.2. RQ2: What Are the Most Used Walkability Indicators in Recent Literature?

Most studies analyzed resort to well-established indicators from the literature on walkability, either using already known values of an indicator, for areas where previous work with that indicator was developed, or employing the indicator’s formulas and methodologies, sometimes with small adjustments, to calculate walkability for new areas of analysis.

In this context, Walk Score® (WS) [62] is the most commonly used indicator in the recent literature on walkability indicators. WS is a patented system available publicly in a

website (<https://www.walkscore.com/>, accessed on 15 July 2022) and it is used to calculate the number of nearby walkable destinations or amenities [61]. It uses data retrieved from sources such as Google, Open Street Map, or the U.S. Census, and, for each address, analyses different routes to nearby amenities via a geographically based algorithm. The system considers 13 categories of amenities, namely grocery stores, coffee shops, restaurants, bars, movie theaters, schools, parks, libraries, book stores, fitness centers, drug stores, hardware stores, clothing and music stores [61]. For each address, it awards points based on the distance to the amenities in each category. It gives maximum points for amenities within a 5-minute walk (0.25 miles or 400 meters) and uses a decay function to give points to more distant amenities, with no points given after a 30-minute walk [62]. Furthermore, it also considers street friendliness metrics, such as block length and intersection density. Then, equally weighted category points are summed and normalized to yield a score from 0–100.

Currently, WS website provides walkability values for North America (USA and Canada) only, but due to its easy reproducibility it has been employed and validated by researchers in different contexts [61,99–103]. In this review, we found a total of 30 studies (47% of the sample) resorting to WS. In terms of countries where the WS was employed, 10 studies were done in the USA [29,38,67,73,86,88–90,93,104], 8 in Japan [78,79,81,92,94,95,105,106], 3 in South Korea [34,82,91], 3 in Canada [65,66,107], and 2 in Australia [31,97], while the following countries have only one study that used WS: China [30], Germany [108], Italy [68], Spain [96], and Taiwan [109]. Regarding the most common research topics, in recent years WS was employed more commonly in Public Health [29–31,66–68,73,78,79,105,107,109,110] and Socio-economic and Demographics [88–96,106], whereas it was used 4 times under the Transport subject [32,81,82,105], and one study focused on the Environment [97]. The only study covering the association between walkability and COVID-19 infections also used WS [104]. Additionally, 2 other studies used Street Smart Walk Score [40,74], a variation of the original WS that performs a distance decay and destination weighting calibration.

The second most used walkability indicator, with a total of 9 works (15% of the sample), is Frank et al.'s Walkability Index (WI) [59]. The WI was initially proposed to link objectively measured physical activity with urban form. The indicator's formula consists of a weighted sum of z-scores for the variables land-use mix (evenness of distribution of square footage of residential, commercial, and office development), net residential density (number of residential units per residential acre), and street connectivity (number of intersections per square kilometer). The WI gives importance to the variable land-use mix, by multiplying its z-score by 2 in the indicator's equation.

Since its initial development, the WI has been employed and validated in other studies [25,111]. From the documents analyzed in this review, the WI was employed 3 times in Japan [5,112,113], 2 in the USA [64,87], and once in China [76], New Zealand [84], Portugal [32], Spain [26] and Scotland [26]. As for the research areas where the WI is used, 5 documents are under the subject of Public Health [24,26,30,64,76], 3 concern Transportation [84,87,112], one comes under Socio-economic and Demographics [113] and one under the Environment category [5].

Besides the WS and the WI, other indicators from the literature that were employed in the works reviewed are: Neighborhood Environment Walkability Scale (NEWS) [13], applied in Canada, China and Japan under the Public Health category [27,70,80,114,115]; the Canadian Active Living Environments (Can-ALE) [116], applied in Canada for Public Health [117] and Transportation [35]; and the Built Environment and Health Neighborhood Walkability Index [118] and National Walkability Index [119], both applied in the USA under the subject of Public Health [83,117].

Fewer works developed and implemented their own indicators. The new indicators that were identified during the research are: the Dutch Walkability Index [120] and the Japanese Walkability Index [112], both falling under the Transportation category; the Walkability Index for London [63], used to measure the relationship between green spaces and physical activity; and the School Walkability Index [121] applied to understand walkability associations with active commuting to school (ACS).

### 3.3.3. RQ3: What Are the Main Findings from Recent Literature?

To better understand the results of the studies reviewed, we went through the main findings by subject category. On Public Health, most findings indicate a positive or neutral impact of walkability in contributing to people's physical activity and health. Studies that associate walkability indicators with the level of physical activity of a community tend to agree that there are built environment characteristics that benefit walking and sports practicing, and thus, that walkability is positively related with moderate-to-vigorous physical activity [26,27,42,83,115]. On the contrary, some works did not find any link between the two [28,66] and others reported a positive association between walkability and sedentary behavior [78,79]. Likewise, on the subtopic of physical health, most studies have found walkability to be negatively linked with being overweight and obesity [67,69–72,105], while fewer studies did not find any association between walkability and physical health [29,107]. In general, walkability was reported as being negatively related with diseases, chronic health conditions and mortality [30,73–75,109,122] and positively associated with mental wellbeing [80]. This positive association between walkability and health does not seem to be significant for older adults (more than 65 years old) [122].

On the subject of Transportation, recent research points to a positive association between walkability and walking or active commuting [34,35,82,83,112,120,121], although with differences at the level of the built environment characteristics that contribute to active commuting, depending on the level of urbanization of a region and age group. For example, in children commuting intersection density, number of four-way intersections, and residential density were positively related to ACS in urban participants, but negatively in rural participants [85]. In Italy, urban children attended schools located in neighborhoods with a higher Walk Score<sup>®</sup> than rural children [68]. Furthermore, associations with minutes of non-discretionary walking and walkability were stronger in rural areas, and in middle-aged (36–49 years) compared to young (18–35 years old) and older adults (50–65 years old) [120]. On the contrary, a study found that US adults living in more walkable areas report more transportation and leisure walking, especially among urban areas [83], and another in Japan [112] notes a stronger association between walkability and step count in older adults (65 years old or older). A study also found that higher levels of walkability are positively associated with transit use of individuals with disabilities and that enhancing neighborhood walkability has a much larger effect on individuals with disabilities than on other groups. [87].

Regarding the Socio-economic and Demographics topic, a variety of results were reported: there was not an association between walkability and social capital [92]; neighborhoods with high income tend to have lower walkability [93]; areas with more subsidized housing have less walkability [90]; one study notes that housing prices are positively correlated with walkability in areas with low housing prices, whereas no significant association was observed in areas with high housing prices [91], while another points to a general positive correlation between walkability and real estate prices [123]; fitness centers are less expensive and have wide opening hours in areas with higher walkability [96]; walkability has a strong positive linear effect on robbery rates and predatory violence, particularly in low-income neighborhoods [37,88,89]; walkability did not affect long-term care insurance service needs [95]; and people tend to be more extrovert in neighborhoods with higher walkability [94].

Lastly, on the topic of the Environment, higher levels of walkability are related with neutral thermal sensation, while a warm thermal sensation is negatively related with walkability [98]. In Sydney, a study found that walkability and greenness (presence of natural areas) were negatively related [97]. Furthermore, walkability negatively affects the Ecological Footprint to Biocapacity (EF/BC) ratio in urban residential areas with a lack of urban infrastructure, but it can have a strong positive effect on the EF/BC ratio in the more urbanized areas [5].

#### 3.3.4. RQ4: How Do These Findings Compare to Those before the COVID-19 Pandemic?

To compare recent and previous findings, we analyzed literature review studies released before 2020. In this regard, the work by Wang and Yang, 2019 [22] is one of the most complete on this topic. The authors analyzed the period between 2008 to mid-2018, retrieving a total of 136 documents on the topic of walkability indicators. In 2016, a total of 21 papers were retrieved and 22 were identified in 2017. By mid-2018, seven works had already been published, and the authors expected that the total number for that year would resemble the previous two. Overall, the number of documents on these topics presents a continual growth pattern since 2008, which had only four documents. These values are in line with what we display in Figure 2, in which 2020 had 20 documents, 2021 had 26 and, by mid-2022, 14 documents have already been published—supporting the hypothesis that there is a continual increase in popularity of the subject of walkability. The work of Wang and Yang, 2019 [22] mentions that the American Journal of Preventive Medicine and the International Journal of Behavioral Nutrition and Physical Activity were the two prominent journals. This last one is also present in the top 5 journals for the post pandemic outbreak period. Just as in the years analyzed in this work, the USA is the country with more documents, and the WS and WI the two most used indicators. J.F. Sallis is the author who made the greatest contribution on research on walkability, with a total of 18 published papers and 14 co-authorships from 2008 to 2018. The author was also found to play an important role in recent literature, as per Figure 15 co-citation analysis, but it is Frank D. Lawrence who is the researcher with the most influence in recent works on the topic. In terms of main findings, the review study by Wang and Yang, 2019 [22] states that studies did not reach a clear agreement regarding whether higher neighborhood walkability leads to more physical activity, although some studies proved that there is a positive relation between the two. This is contrary to the overall findings of the current study, in which there is a general agreement regarding the positive impact of walkability on physical activity. Despite this, the authors point that high-income countries demonstrate that there is a positive relationship between walkability and physical activity among adults, which is particularly strong in older adults, as we have also seen in RQ3. Regarding the impact of walkability in health, in general previous studies also agree that built environment attributes that lead to high walkability are associated with lower levels of obesity and BMI. Regarding the impact of walkability factors in the environment, the authors mention one work by James, Hart, and Laden (2015), who demonstrated that neighbourhood walkability was positively correlated with modeled concentrations of the air pollutant PM2.5.

Another literature review study from 2019 addressed the topic of walkability factors associated with older adults' walking behaviors [21]. A total of 70 studies were analyzed, and most of those articles were published in North America (45.7%). Again, the WS was found to be the most used indicator across the literature. The authors also found a strong relationship between the walking behaviors of older adults and neighborhood environment characteristics, and that walkability, as well as urbanization, are strongly associated with total and transport walking among older adults.

In both reviews it is mentioned that, just like in recent works, data for walking and health metrics are mostly objectively based on accelerometry and/or subjectively based on self-reported physical activity from questionnaire results.

#### 3.3.5. RQ5: How Do Recent Works on Walkability Indicators Reflect the Topic of the COVID-19 Pandemic?

The COVID-19 pandemic brought more awareness to the general public about the topic of walkability [42]. People that commuted, and for whom private transportation was not an option and public transportation could be a possible vehicle of infection [48,124,125], resorted more often to active modes of transportation, like walking and cycling [45–49]. Hence, it would be expected that this would prompt a reaction on the scientific community to analyze, on one hand, the impact of the pandemic in choices of mobility and, on the other hand, the link between urban design and COVID-19 infection rates, the latter evaluated

using walkability indicators. Nevertheless, from the final 64 works analyzed, only one study by Lima et al. [104] used the WS to measure the extent to which walkability was related with the spread of COVID-19. The study focused on the early stages of the pandemic in 93 United States counties, concluding that walkability and population density were positively correlated with COVID-19 infection rate. When extending the works analyzed to incorporate studies from the pre-screening phase, we found another work relating walkability and infection rate, by Wang et al. [126]. In this case, while the study does not employ a walkability indicator, it measures walkability through different aspects of the built environment, and it concluded, in opposition to Lima et al. [104], that walkability, biking services and higher socio-economic status were associated with lower infection during the peak of the COVID-19 pandemic in Massachusetts, Texas.

Alternatively, the impact of COVID-19 on walkability indicator studies can also be measured indirectly, based on the amount of work developed prior and post pandemic. Because people were and still are spending more time at their residential area [50,51], they are giving more importance to how the built environment and urban design affects their mobility and overall quality of life [42]. Consequently, this can also result in a raising interest from investigators. We ran our initial search query in WoS with the timeframe filter removed. The yearly pattern of documents published reveals a continual growth of publications, from 2013 to 2022. While it is true that this pattern precedes the pandemic, from 2017 to 2019 there was a 17% growth on the number of works related with walkability indicators, whereas from 2019 to 2021 that growth rate doubled to 33%. Although there could be many reasons motivating this increased interest in walkability, indirect effects from the COVID-19 pandemic are certainly possible causes.

### 3.3.6. RQ6: What Are the Main Gaps Identified in the Reviewed Literature?

From the previous research questions (RQ4), there is a lack of studies linking walkability indicators to COVID-19, while existing research is contradictory. As well, the relationship between walkability indicators and environmental metrics is not thoroughly analyzed. From the three works that studied this topic, only one focused on the environmental impact of walkability. It concludes that the relationship between walkability and EF/BC ratio is negative in urban residential areas with a lack of urban infrastructure, but it can have a strong positive effect on more urbanized areas [5]. Hence, there is a lot of space for further research on this topic. Studies could address, for example, the relationship between walkability and pollution, using NO<sub>2</sub> as the variable of analysis, since it has been shown to be positively related with automobile traffic [52].

Moreover, most works rely on traditional methods to obtain the data for the phenomena being associated with walkability indicators. Across the different subjects of analysis, data collection for physical activity, health condition or travelling patterns is mostly done through the application of questionnaires and surveys [26,27,30,42,63,71,77,79,81–83,92,107,109,120,122], or pedometers/accelerometers [26,28,66,112]. The collection of this data is costly, not easily updated and difficult to scale [127]. This can lead to limitations when evaluating walkability associations in large urban environments. Moreover, the evaluation data used in recent literature was, in general, collected prior to the COVID-19 outbreak, since the data collection processes mentioned before are time consuming and difficult to reproduce for new population samples. From the literature analyzed, only five studies rely on data collected between 2020 and 2021 [68,76,97,104,115]. Some of the studies rely on longitudinal data collections methods, that require years of research, data treatment and analysis. For instance Lang et al. [67] resort to the *Reasons for Geographic And Racial Differences in Stroke* study, Martins et al. [70] use the *Nagoya Longitudinal* study and Matsumoto et al. [106] resort to the *Keeping Active across Generations Uniting the Youth and the Aged* study. All other studies rely on data collected between 2010 and 2019, for example [37,65,73,79,80,93,112]. This is a limiting issue for a thorough analysis of the pandemic direct impact on walkability, as the data collected on most studies does not reflect any COVID-19 influence.



To face this issue, big data, and information and communication technologies emerge as advantageous tools for the quantification and collection of relevant data on mobility, health and sustainability, helping to face challenges in the provision of public services and infrastructure [128]. For example, the widespread use of mobile devices and installation of environmental sensors can allow for the collection of large-scale data on human mobility behaviors [129] and air quality [130], respectively. These data can then provide a basis for investigating the associations between walkability indicators and human phenomena, allowing for faster data collection and update, which is especially crucial in critical times like the period under analysis.

#### 4. Conclusions, Limitations and Future Work

We performed a systematic literature review of works that employed walkability indicators to uncover associations between walkability and different phenomena, in the aftermath of the COVID-19 pandemic. We resorted to the PRISMA methodology to search, evaluate, and select, these studies. Then, exploratory and bibliometric analysis were conducted. Studies on walkability indicators have increased in the aftermath of the pandemic, almost doubling the growth rate of the three years prior to the outbreak of the virus. Although studies on this topic were performed across the globe, North America (USA and Canada) and Japan are the prominent regions. The bibliometric analysis revealed that Gavin R. McCormack and Mohammad Javad Koohsari were the most influential researchers regarding works done in recent years, whereas Frank D. Lawrence is the overall most cited author. We found that the main subjects addressed in the recent literature are Public Health (52%), Transportation (23%), Socio-economic and Demographics (20%), and the Environment (5%). Most studies resort to well-known indicators, predominantly to Walk Score<sup>®</sup> (47%), but also to Frank's Walkability Index (15%), while other works develop and apply new indicators (15%).

Findings point to a positive association between walkability and physical activity and fitness, and a negative correlation with health problems. Moreover, walkability is positively associated with active mobility and real estate prices. Only one document linked walkability indicators with COVID-19, concluding that walkability and population density were positively correlated with COVID-19 infection rate. On the Environment subject, the literature is also scarce and unclear. In general, most works rely on traditional methods to obtain the dataset for the phenomena being associated with walkability indicators. As such, there is room for the use of more modern and scalable methods of data collection, such as mobile phone GPS and environmental sensors. Furthermore, post COVID-19 works are closely aligned with previous ones, in terms of topics of research, methods of gathering evaluation data, as well as results and conclusions.

As limitations, we must note that the period covered by this literature review is not long (only two years and a half), and so it may not yet capture the long-term impact that the COVID-19 pandemic had and continues having in people's life. It is also the case that studies published after the outbreak of the pandemic might have been completed before the outbreak. To mitigate this issue, when possible, we checked if the submission date of the documents was, at minimum, during 2020. We must also mention that despite being judicious during the process of document selection, we cannot entirely assure that no relevant work has been missed.

For the future, we believe that studies aiming at evaluating the impact of walkability factors on physical activity, transportation, and overall mobility, should resort to more modern data collection methods that rely on GPS, mobile and sensor data, which tend to be highly scalable and efficient.

Our work should help researchers to understand the trends in studies employing walkability indicators, and we hope it can support and foment further research in this area.

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## Appendix A

**Table A1.** Five most cited papers from Web of Science during the period from 2020 to 2022 (May), as of June 2022.

Title	Authors	Country	Indicator	Subject	Main Findings	Citations
The relationship between living in urban and rural areas of Scotland and children's physical activity and sedentary levels: a country-wide cross-sectional analysis	McCrorie et al. [28]	Scotland	Variation of WI [59,60]	Children physical activity	Walkability was not associated with any outcome of physical activity.	13
Neighborhood Walkability and Housing Prices: A Correlation Study	Kim et al. [91]	South Korea	WS [62]	Real estate price	Housing prices positively correlated with the walkability score in areas with low housing prices, whereas no significant association was observed in areas with high housing prices.	12
Walking-friendly built environments and objectively measured physical function in older adults	Koohsari et al. [105]	Japan	WS [62]	Elderly Physical function	Walkability factors significantly associated with better physical function performance in elderly man.	11
Associations between the National Walkability Index and walking among US Adults-National Health Interview Survey, 2015	Watson et al. [83]	USA	NWI [119]	Active mobility	Walkability was associated with a higher likelihood of walking, especially for transportation.	10

Table A1. Cont.

Title	Authors	Country	Indicator	Subject	Main Findings	Citations
Different neighborhood walkability indexes for active commuting to school are necessary for urban and rural children and adolescents	Molina-García et al. [85]	Spain	z Residential density + z Distance to school +z No. traffic lanes + z Intersection density + z Positive Streetscape characteristic	Active commuting to school (ACS)	Intersection density, number of four-way intersections, and residential density were positively related to ACS in urban participants, but negatively in rural participants.	8
Effect of Walkability on Urban Sustainability in the Osaka Metropolitan Fringe Area	Kato [5]	Japan	WI [59]	Ecological Footprint to Biocapacity (EF/BC) ratio	Walkability negatively affects the EF/BC ratio in the sprawl neighborhood cluster, and has a strong positive effect on the EF/BC ratio in the old new-town neighborhood cluster.	8

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