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Article

The cooling effect of large-scale urban parks on surrounding area thermal comfort

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Abstract: This empirical study investigates large urban park cooling effects on the thermal comfort of occupants in the vicinity of the main central park, located in Madrid, Spain. Data were gathered during hot summer days, using mobile observations and a questionnaire. The results showed that the cooling effect of this urban park of 140 ha area at a distance of 150 m is able to reduce temperature by an average of 0.63 °C and 1.28 °C for distances of 380 m and of 665 meters from the park. Moreover, the degree of the Physiological Equivalent Temperature (PET) index at a distance of 150 meters from the park is on average 2 °C PET and 2.3 °C PET less compared to distances of 380 m and 665 m, respectively. Considering distance from the park, the correlation between occupant Perceived Thermal Comfort (PTC) and PET is inverse. That is, augmenting the distance from park increases PET, while the extent of PTC reduces accordingly. The correlation between these two factors at the nearest and furthest distances from the park is meaningful (P-value < 0/05). The results also showed that large-scale urban parks generally play a significant part in creating a cognitive state of high-perceived thermal comfort spaces for residents.

Keywords: Cooling effect; urban park; thermal comfort; physiological equivalent temperature; perceived thermal comfort

1. Introduction

Due to climate change and an increase in urbanization, urban heat is rising rapidly (Dimoudi et al., 2013). Today, heat has adversely affected urban life across the world (IPCC, 2017), including urban areas of Mediterranean climate (Sobrino et al., 2012; Sánchez-Guevara Sánchez, Núñez Peiró, and Neila González, 2017). The increase in temperature in urban areas, especially densely populated areas has given rise to the phenomenon of Urban Heat Islands (UHI) (Taha, 2017; Oke, 1982) which can threaten the health and comfort of citizens (Leal Filho et al., 2018).

Green urban spaces have been researched by numerous studies as an adaptive strategy to reduce the effect of urban heat and improve the health of citizens, by considering thermal comfort (Ling and Chiang, 2018; Wolch, Byrne, and Newell, 2014) as well as their socioeconomic role (Aram, Solgi, and Holden, 2019). Various studies have been conducted on the effects of various types of green infrastructures aimed at reducing urban heat and thermal discomfort (Demuzere et al., 2014; Zölch et al., 2016), which include different scales and forms such as small local parks (Park et al., 2017), large urban parks (Buyadi, Mohd, and Misni, 2013), urban gardens (Lottrup, Grahn, and Stigsdotter, 2013), green roofs (Imran et al., 2018), green walls (Tan, Wong, and Jusuf, 2014) and street trees (Kong et al., 2017). This study emphasizes the cooling effect of large urban parks.

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The studies conducted on large urban parks have shown that such parks have a significant effect on air temperature reduction and UHI (Wong and Yu, 2005; Sugawara et al., 2016), which is especially noticeable during the summer (Qiu et al., 2017; Lai et al., 2019). However, it is noteworthy that the temperature drop does not only take place within the park but also within surrounding areas (Cao et al., 2010). The cooling effect of urban parks, referred to as Park Cooling Island (PCI) (Vidrih and Medved, 2013), is known to affect the surrounding environment depending on the area size of the green space and quality of green coverage (Jamei et al., 2016; Xu et al., 2016). Two important indices to measure the cooling effect of urban parks are the Cooling Effect Distance (CED) and Cooling Effect Intensity (CEI) (Aram et al., 2019), which have been extensively used in various scales and climates (Feyisa, Dons, and Meilby, 2014; Lu et al., 2017; Anjos and Lopes, 2017).

Studies on the cooling effect of urban parks in different regions have shown that large scale urban parks with areas over 10 ha can have an average of 1 to 2 °C effect on surrounding areas that are an average of 350 meters away (Aram et al., 2019). In general, air temperature reductions in urban parks are typically up to 0.5-4 °C and may even cause up to 5-7 °C reduction (Yan, Wu, and Dong, 2018). Despite the cooling effect of urban parks on their surroundings, such effect is not merely dependent on the traits of the green space (Aflaki et al., 2017). The morphology of the surrounding area of parks, the sky view factor (SVF), the spatial configuration of the location, and the covered area, also affects the perception of the cooling effect of urban parks and thermal comfort (Yan, Wu, and Dong, 2018).

Numerous studies have been carried out on the cooling effect of urban parks. However, research pertaining to the extent of this effect on thermal comfort perception from psychological and physical perspectives is limited. This gap necessitates examining the cooling effect of urban parks on the aforementioned parameters at different intervals from the park to determine factors which may be applied in sustainable urban development.

2. Park cooling effect and thermal comfort perception

A set of thermal indices was investigated in two experimental (ET, RT, HOP, OP, WCI) and analytical (ITS, HIS, ET*, SET*, OUT_SET, PMV, PT, PET) groups (ASHRAE, 2010; Coccolo et al., 2016; Fanger, 1970; Walls, Parker, and Walliss, 2015). The basis of these analytical indices was the energy balance (produced and wasted human energy). The main issue of thermal indices pertains to average thermal comfort assessment and climate conditions of each area (Bruse, 2009). The result of various studies on validating other indices shows that examples such as SET and PET have a high correlation (89%) with thermal comfort in open spaces (Monteiro, 2009). Most studies in recent years have utilized SET, PMV and PET to predict comfort levels in open spaces (Potchter et al., 2018; Honjo, 2009; Roshan et al., 2019).

PET enables the comparison of the complete effect of thermal conditions pertaining to the outside environment with individual experience (Lin, Matzarakis, and Hwang, 2010). PET is one of the recommended indices in urban and regional urban planning around the world, used to predict thermal changes of urban or regional clusters (Chen and Ng, 2012). This index has shown a significant correlation to thermal comfort in various climatic conditions in open urban spaces as validation (Tseliou et al., 2010). One of the prominent influencing factors on PET condition is the T_{mrt} climate variable. The aforementioned indices provide a single image of a set of individual and climate variables and enable the comparison of comfort conditions in various environments (due to global factors) (Thorsson et al., 2007).

In studies conducted to investigate the thermal comfort of urban parks, the PET index is commonly used. In a study conducted on a warm sunny day in Beijing, China (21 August, 2 pm) (Sun et al. 2017), it was shown that the effect of the Yuan Dynasty Relics green space with 102 ha area, in close approximately to the area of Retiro Park, reduces the PET by an average of 2 °C and a maximum of 15.6 °C. Another study conducted at Zhongshan Park in Shanghai city center with an area of 21.42 ha, showed that the PCI led to thermal comfort during Shanghai's winter as well as summer and had a PET of 15-29 °C (Chen et al., 2015).

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In a study conducted by Cohen, Potchter, and Matzarakis (2012), a total of 10 urban parks of various areas (0.2 to 0.36 ha) were assessed in Tel Aviv. Although the investigated parks were smaller in terms of area compared to Retiro Park, the results illustrated that parks with richer vegetation density had greater cooling effects and thermal comfort, and could reduce temperatures by up to 3.8 °C whilst bringing PET to 18 °C during summer. Additionally, a study conducted at the central park of Cairo with an area of approximately 26.01 ha (Mahmoud, 2011) showed that parks had a significant effect in enhancing thermal comfort during summer, which, based on this effect, entailed a PET value of 22–30 °C throughout the day.

While the cooling effect of urban parks has generally been recognized as a strategy for mitigating urban heat and providing thermal comfort (Bartesaghi Koc, Osmond, and Peters, 2018; Taleghani, 2018), the extent of perceived thermal comfort stemming from the cooling effect of urban parks is difficult to predict, necessitating investigations of people's attitudinal and bodily experiences concerning physical thermal conditions at various intervals in the surrounding areas of these parks (Lenzholzer, Klemm, and Vasilikou, 2018; Nikolopoulou, Baker, and Steemers, 2001). The degree of perceived thermal comfort by an occupant which is caused by the cooling effect of urban parks is dependent on factors such as individual behavioral and psychological traits, in addition to distance and cooling effect intensity (Klemm, Heusinkveld, Lenzholzer, Jacobs, et al., 2015; Knez and Thorsson, 2006; Nagashima, Tokizawa, and Marui, 2018). Individual demographic traits include age, gender, and physical characteristics such as height and weight of occupants (Thorsson, Lindqvist, and Lindqvist, 2004; Pezzoli et al., 2012). Psychological traits include the individual's experience of being in the environment and their thermal expectations and tolerance (Nasir, Ahmad, and Ahmed, 2012; Lin, 2009). Behavioral aspects include the extent of being covered (in terms of clothes) and the type of activities conducted by the individual (Nikolopoulou and Lykoudis, 2006).

A common method to determine the extent of thermal comfort experienced by occupants is the use of surveys (Ng and Cheng, 2012). Essentially, by using such survey methods as open, semi-open or multiple choice questions (Li, Zhang, and Zhao, 2016), individuals can be asked about their thermal comfort experiences (Yang et al., 2017). In order to measure human thermal comfort perception level, cognitive mapping can be utilized in conjunction with questionnaires. Employing this method and asking residents to identify places where they feel more comfortable from a thermal point of view provided a more comprehensible image of the residents' thermal comfort level of understanding (Klemm, Heusinkveld, Lenzholzer, and van Hove, 2015; Lenzholzer, 2008). In recent years, utilizing questionnaires and asking direct questions about the perceived comfort of citizens alongside the micro-climatic perceptions have yielded valuable results (Xu et al., 2017).

The purpose of this study is to investigate the cooling effect of the large urban park and its effect on the thermal comfort of occupants in the areas around the park. This study focuses on the Perceived Thermal Comfort (PTC) and Physiological Equivalent Temperature (PET) of locations, at a defined distance from the park and similar in terms of urban aspects and influential factors such as floor coverings, enclosures, street canyon and vegetation, whilst responding to the following research questions:

- What is the extent of the large urban park cooling effect on thermal comfort based on PET?
- What is the extent of the large urban park cooling effect on occupants' PTC?
- What is the relationship between the measured PTC and the measured occupant PET?

3. Materials and Methods

In this study, using a combination of methods such as landscape architecture, micrometeorology, and social geography, the effect of a large-scale urban park on thermal comfort was investigated.

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3.1. Location and selection of the sites

This research was located in Madrid (40 ° 25'08 "N; 3 ° 41'31" O), the capital of Spain with a population of 3,223,334 and population density of 5265.91/km², and a Mediterranean climate (Csa: dry-summer subtropical) according to the Köppen classification (Kottek et al., 2006). The Retiro Park has an area of approximately 140 ha and a 123 km perimeter. It is the largest and oldest park in the center of Madrid with rich plant diversity. One of the prominent issues of this study was the method of selecting distances from the Retiro Park. Although studies have not been conducted on the average cooling effect of Retiro Park, an updated UHI map of Madrid was presented in 2015 by Núñez Peiró et al. (2017), where the effect of Retiro Park was made evident on the basis of the temperature color spectrum. Essentially, the UHI map of Madrid from 26 July 2015 represents the effect of Retiro Park in counteracting to the effect of UHI in the northern, eastern and southern areas of the park (Román, Gómez, and de Luxán, 2017).

In order to accurately assess the cooling effect of Retiro Park on its surrounding areas, it was necessary to select places on a micro scale with common features that are located at various distances from the park. The northern area of Retiro Park due to its well-organized form has more regular urban type compared to other areas (east and south). Based on the temperature zones of the updated thermal map of Madrid's urban heat islands in 2015 (ABIO, 2018), three street crossings (intersections) located in the Salamanca and Recoletos neighborhoods in the north of Retiro Park along the Lagasca Street, were chosen. These sites were located at 150 meters, 380 meters and 665 meters from the park at Lagasca intersection with the Conde de Aranda, Jorge Juan, and Hermosilla streets, giving zones of different temperature ranges (yellow, orange and red) (Fig 1). These types of selected local intersections represent most of Retiro Park's northern area intersections (Fig 2). Based on this map (Núñez Peiró et al., 2018; ABIO, 2018), each of the three intersection A, B and C had a temperature difference of about 0.8 °C.

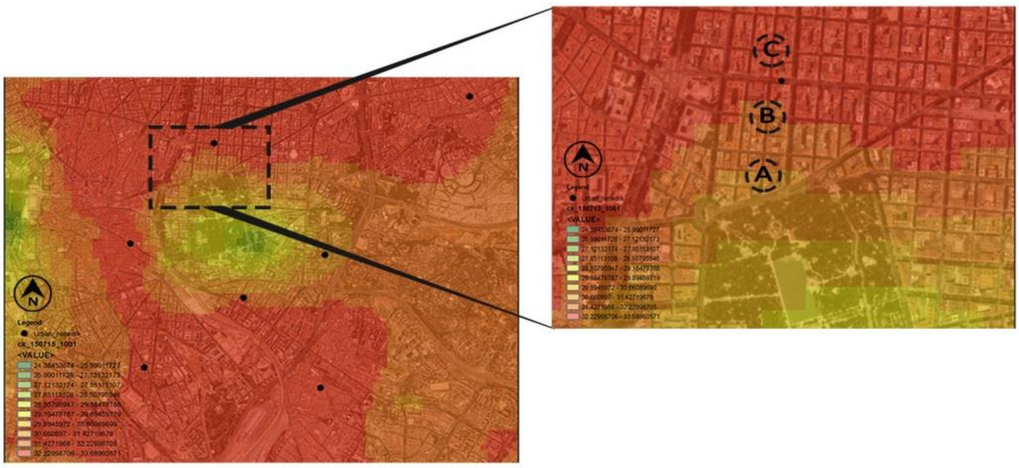

Figure 1. Madrid's UHI map on 26 July 2015 (Núñez Peiró et al., 2018; ABIO, 2018) and the selected northern region of the Retiro Park. Intersection A is in the yellow zone, intersection B is in the orange zone and intersection C is in the red zone, close to one of the UHIs of Madrid.

All three assessed intersections were selected based on a common geometric configuration. The considered geometric configuration included an enclosure, sky view factor (SVF), and street height and width (Fig 3, Table 1). The considered material structure included the ground surface coverage material, pavement material, and building facade material which were similar in all three regions; the ground surface material in all three regions was asphalt; the pavement material was gray tiles; and the building material is mostly red-colored bricks and bright cement (Fig 3). The presence of green spaces in street canyon was highly significant in terms of micro-climatic issues (Morakinyo et

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al., 2017). In this regard, vegetation was also considered as well as the aforementioned parameters (Table 1), so the selected regions were also similar in terms of this parameter (Fig 3).

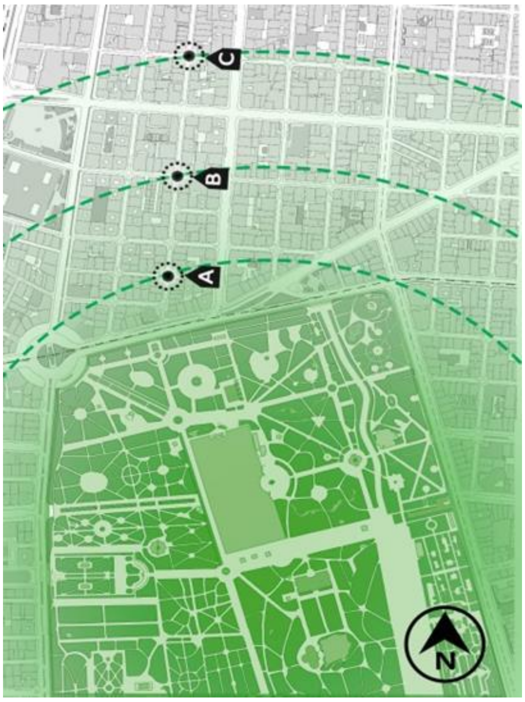

Figure 2. The northern region map of Retiro Park and the location of the selected sites. Point A is located 150 meters away from the park; Point B is located 380 meters away from the park; and Point C is located 665 m away from the park.

Essentially, there were two priorities in selecting the intersections. Firstly, the points were selected according to Madrid's UHI map (Fig 1) where the distances were of different temperature ranges, and secondly, locations were selected that shared the greatest similarity in terms of physical and structural traits. In this vein, three sites A, B and C were selected to extract data, in accordance with the goals of this study.

Table 1. Geometric configuration of the intersections and street trees properties in three investigated intersections.

	Street Canyons					**Street trees**			
Part	Width N_S (M)	Width W_E (M)	H/W Ratio N_S (-)	**H/W Ratio W_E (-)**	**SVF [a]**	**Number of trees**	**Mean Tree Height(m)**	**Mean Crown Diameter (m)**	**Crown Shape [b]**
A	15	12	1.23	1.54	0. 17	26	12.74	5	Cone
B	15	15	1.28	1.33	0. 2	30	12.34	6	Cone
C	15	15	1.2	1.23	0. 2	25	12.66	5	Cone

[a] Calculated by Ray Man 1.2 [b] Crown Shape Classification by Park et al. (2017).

Figure 3. Street view of the three intersections (A, B and C) as well as fisheye shots.

3.2. Microclimate parameter measurements

Micro-climatic measurements were conducted during 6 days in summer, on 22 June, 10 and 24 of July, 10 and 24 August, and 10 September 2018. Data extraction started from 22 June, and was spaced at roughly every two weeks, targeting sunny and calm days (no clouds) up to the end of summer (10 September). Ta and relative humidity (RH) were collected during the 6 days, were measured by mobile microclimate stations (HOBO MX2301A Temperature/RH Data Logger, manufactured by Onset Computer Corporation, MA, USA) with precision T_a; ± 0.2 °C and RH; ± 2.5% between 10:40 and 12:10 CEST. The data was collected at one minute intervals and the weather data collection duration at each location was 10 minutes (Table 2).

Table 2. Intersection mean values for air temperature (T_a), relative humidity (RH) and Wind Speed (WS) on the all measurement days (10:40–12:10 CET).

Date	Part	Time	Mean T_a, °C	Mean RH, %	Mean WS,m/s
22 Jun	**C**	10:40-10:50	29.26	29.1	1.43
	B	11:00-11:10	28.87	29.27	1.15
	A	11:30-11:40	27.81	32.04	0.80
10 July	**C**	11:15-11:25	29.8	33.96	1.43
	B	11:42-11:52	29.7	22.1	1.16

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	A	12:00-12:10	29.4	29.71	1.36
24 July	**C**	10:40-10:50	29.45	21.71	1.36
	B	11:10-11:20	28.1	30.25	1.43
	A	11:30-11:40	27.74	30.56	1.43
10 Aug	**C**	10:40-10:50	25.12	18.16	1.43
	B	11:00-11:10	23.77	41.05	1.72
	A	11:20-11:30	23.22	43.45	2.47
24 Aug	**C**	10:50-11:00	26.63	38.81	2.35
	B	11:15-11:25	26.2	38.13	1.13
	A	11:35-11:45	26.05	37	1.29
10 Sept	**C**	10:45-10:55	23.41	56.71	1.5
	B	11:15-11:25	23.13	52.45	1.36
	A	11:35-11:45	21.73	51.15	1.91

The devices were placed 1.5 meters from the ground and covered by a radiation shield. The Wind Speed (WS) was determined by a Proster Digital Anemometer MS6252a placed at the same distance. The type of ground coverings, walls, and types of and distance to vegetation that affect the temperature conditions were checked according to field studies; high-resolution images were taken by a Canon Eos 600D, 5184 x 3456 pixels digital camera. Arial images were taken from Google Earth, 2018-2019. Fisheye images were taken at three intersections using a fisheye (Sigma 8mm circular) lens. Additionally, the weather data of Retiro Park were collected via the AEMET (Agencia Estatal de Meteorologia) fixed station (AEMET, 2018) located inside the park. These parameters were previously identified to show the effect of urban parks on thermal comfort (Sun et al., 2017).

3.3. The questionnaire survey

A random semi-structured survey was conducted on 145 pedestrians (nA: 49, nB: 45, nC: 51) in three considered locations (workplace and residential). The interviewees comprised different gender and age groups (with the exception of children), and were active at different levels. Table 3 shows detailed information on the number and characteristics of individuals, on different days and intersections.

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Table 3. The proportional percentages questionnaire data in the three investigated intersections on all the measured days.

Variable	Categories	Percentage (%)		
		A	B	C
Age	13-21	2.0	2.2	11.8
	22-30	26.5	20.0	13.7
	31-45	40.8	26.7	31.4
	46-60	16.3	24.4	29.4
	61-85	14.3	26.7	13.7
Gender	Men	67.3	73.3	51
	Women	32.7	26.7	49
Q1	Very low (1)	0	0	0
	Low (2)	10.2	6.7	9.8
	Medium (3)	28.6	57.8	43.1
	High (4)	42.9	31.1	35.3
	Very high (5)	18.4	4.4	11.8
Q2	Very low (1)	5.9	4.4	0
	Low (2)	23.5	31.1	18.4
	Medium (3)	33.3	26.7	22.4
	High (4)	27.5	28.9	38.8
	Very high (5)	9.8	8.9	20.4
Q3	Very low (1)	4.1	2.2	13.7
	Low (2)	8.2	20.0	25.5
	Medium (3)	28.6	31.1	27.5
	High (4)	42.9	31.1	25.5
	Very high (5)	16.3	15.6	7.8
Q4	Very low (1)	2.0	0	3.9
	Low (2)	10.2	15.6	25.5
	Medium (3)	30.6	60.0	35.3
	High (4)	42.9	13.3	23.5
	Very high (5)	2.0	.0	3.9

Data extraction from the surveys took place during the period of determining micro-climatic parameters at three intersections A, B and C (more and less than 10 minutes). Questions were divided into three sections.

The second section comprised four questions about the degree of thermal comfort perception during the interview. The questions were short and designed as five options (very high = 5 to very low = 1). Research questions concerning the level of individual thermal comfort included: *[Q1] How much thermal comfort do you feel? (not too hot, not too cold); [Q2] How warm do you feel? [Q3] What is the extent of thermal comfort perceived through Retiro Park?; [Q4] How much heat can you tolerate in this location?*

The third section included open personal questions such as gender, height, weight, level of activity, and type of clothing.

Essentially, the questionnaire questions were designed so that in a short time period (10 minutes), pedestrians in the neighborhood in different age and gender groups could be interviewed to extract information about their level of perceived thermal comfort (Fig 4). The total score from the questions is presented as an indicator of the Perceived Thermal Comfort (PTC). However, the inquiry of PTC was ambiguous for occupants so, this question was addressed by asking How comfortable do you feel in terms of temperature (not too hot not too cold)? Also, for the second inquiry, the question that was directly asked was, How much heat do you feel? so that data relevant to this question was considered in reverse form to determine the perceived thermal comfort index (very high=1 to very

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low=5). It should be noted that in order to accurately derive the research data, prior to the start of summer on 22 May, 18 experimental micro-climatic data and questionnaires (6 questionnaires at each intersection) were compiled (10 minutes at each intersection) in order to address the issues and queries of questionnaire gaps and to identify bio-climatic notions amidst the main questionnaire compilation stage.

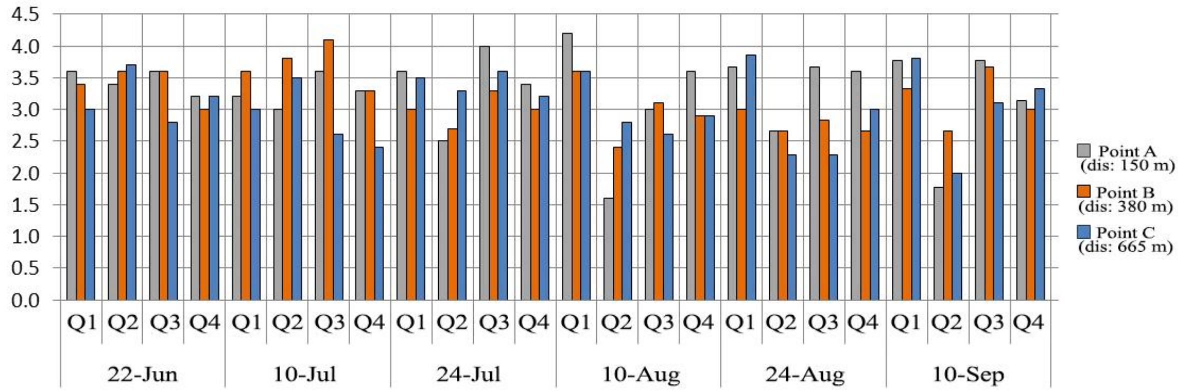

Figure 4. The average frequency of questionnaire data (Questions 1-4) at three intersections. Intersection A is gray, Intersection B is orange, and Intersection C is blue.

3.4. Analyzing the thermal comfort parameters

3.4.1. Physiological parameters

In order to analyze thermal comfort and derive PET, the relevant parameters should be calculated. The calculation of parameters for PET assessment includes clo, level of activity, SVF, and T_{mrt}. In this study, Ray Man 1.2 (Fröhlich, Gangwisch, and Matzarakis, 2019; Matzarakis, Rutz, and Mayer, 2007) software was used to derive SVF, T_{mrt} and PET. The calculation of clo, which pertains to the extent of being covered for an individual, is based on its computational indices (Auliciems and Szokolay, 2007). Given that each clothing item has its own index, the clo for each person was based on the collected data. The greatest extent of clothing coverage for participants was a T-shirt with trousers (34%), shirt with trousers (25.4%) and the least was shorts with a t-shirt (22.6 %), which were of clo 61, 65 and 40, respectively. Determination of activity levels was based on individual activity in the environment. Every activity had an indicator, walking 115 W/m2, sitting 60 W/m2, standing 70 W/m2, and fast walking 220 W/m2 (ASHRAE Standard 55, 2013) (Table 4).

Table 4. Clothing and activity level of responders in the three investigated intersections.

Variable	Percentage (%)																	
	Summer Days																	
	22 Jun			10 July			24 July			10 Aug			24 Aug			10 Sep		
	A1	A2	A3	A1	A2	A3	A1	A2	A3	A1	A2	A3	A1	A2	A3	A1	A2	A3
_Clothing																		
Shirt and Normal Pants- 65 clo	40	40	33.3	44.4	33.3	25	30	33.3	10	0	20	0	0	16.6	28.5	22.2	50	30
Tshirt and normal pants_61 clo	20	20	33.3	22.2	22.2	37.5	30	44.4	50	30	40	50	33.3	16.6	57.1	33.3	33.3	50
shirt and Shorts-45 clo	0	0	0	11.1	0	0	10	0	0	10	10	0	16.6	16.6	0	11.1	0	10

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Tshirt and Shorts (or skirt)-40 clo	**40**	**20**	**16.6**	**11.1**	**33.3**	**0**	**20**	**22.2**	**20**	**50**	**10**	**40**	**50**	**33.3**	**14.3**	**0**	**16.7**	**10**
Dress-35 clo	**0**	**0**	**0**	**11.1**	**11.1**	**12.5**	**10**	**0**	**10**	**10**	**10**	**0**	**0**	**0**	**0**	**0**	**0**	**0**
Suiet-90 clo	**0**	**0**	**16.6**	**0**	**0**	**12.5**	**0**	**0**	**10**	**0**	**10**	**10**	**0**	**16.6**	**0**	**11.1**	**0**	**0**
Others	**0**	**20**	**0**	**0**	**0**	**12.5**	**0**	**0**	**0**	**10**	**0**	**0**	**0**	**0**	**0**	**22.2**	**0**	**0**
_Activity																		
Wlking- 115 w	**60**	**10**	**66.6**	**55.5**	**77.8**	**50**	**40**	**44.4**	**60**	**30**	**50**	**80**	**50**	**66.6**	**57.1**	**55.5**	**83.3**	**40**
Standing-70 w	**20**	**0**	**16.6**	**44.4**	**22.2**	**12.5**	**50**	**44.4**	**20**	**40**	**30**	**10**	**16.6**	**33.3**	**14.3**	**11.1**	**16.7**	**20**
Sitting- 60 w	**20**	**0**	**16.6**	**0**	**0**	**37.5**	**10**	**0**	**20**	**30**	**20**	**10**	**33.3**	**0**	**28.5**	**33.3**	**0**	**40**
Jogging-220 w	**0**	**0**	**0**	**0**	**0**	**0**	**0**	**11.1**	**0**	**0**	**0**	**0**	**0**	**0**	**0**	**0**	**0**	**0**

SVF calculation is conducted by importing fisheye images in the RayMan software, and 3D simulation of spatial geometry and environment vegetation in the Obstacle section of the software. The SVF index is one of the effective parameters in deriving the thermal comfort of the environment (Matzarakis, Rutz, and Mayer, 2010).

The PET index is considered a vital indicator in studying thermal comfort, and has been used in numerous studies as a reliable indicator in deriving the thermal comfort of the external environment (Potchter et al., 2018). T_{mrt} and PET were derived based on extracted micro-climatic data (Table 5), and SVF data derived via the Ray Man software (Table 1).

Table 5. Values of the T_{mrt}(°C) and average values for PET (°C) in Part A,B and C.

Date	Part	Mean PET, °C	T_{mrt}, °C
22 Jun	C	35.3	48.4
	B	35.8	48.7
	A	33.7	48
10 Jul	C	36.3	49.4
	B	36.5	49.3
	A	35.8	49.3
24 Jul	C	35.7	49
	B	33.9	47.7
	A	33.5	47.6
10 Aug	C	28.9	43.8
	B	26.8	43.1
	A	24.3	42.7
24 Aug	C	29.1	44.4
	B	28.6	44.5
	A	26.9	44.3
10 Sep	C	26.1	40.6
	B	26.2	41
	A	23.1	39.8

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3.4.2. Psychological parameters

SPSS software was utilized, according to the statistical data obtained from the questionnaire to examine the significance of the data in three intersections, and also to examine the significance of the relationships between various indices with PTC (Maria Raquel, Montalto, and Palmer, 2016; Yen et al., 2017). In order to examine cognitive maps more accurately, AramMMA software was used (Aram, Solgi, Higueras García, et al., 2019). In this program, all the cognitive maps are overlapped, and then according to the color spectrum, it determines which locations have the most point of reference (Fig 5). Each color represents the number of pointing a location (for example red means one time or purple means 12 times). In order to convert the qualitative data of cognitive maps into quantitative data, inside the AramMMA software, the Retiro area was rated, so that 100 percent of the score was dedicated to the maps that noted parks, while zero percent was considered for those that did not mention to the park. By using the cognitive map analyses, the importance of park cooling effect on resident psychological thermal comfort in different parts of the Retiro Park was represented.

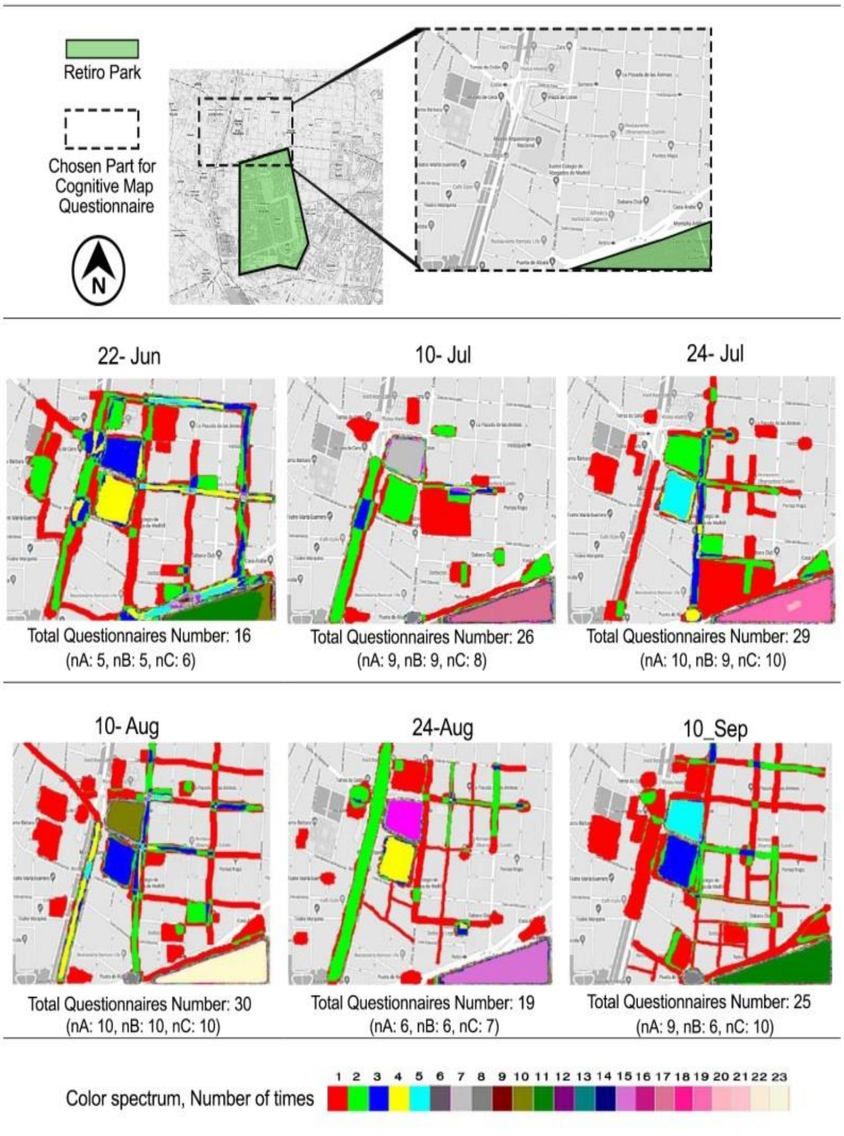

Figure 5. Cognitive map analyses using the AramMMA software (Aram, Solgi, Higueras García, et al., 2019) during six summer days of 2018.

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4. Results

Data was gathered during six days in summer, on 22 June, 10 and 24 July, 10 and 24 August and 10 September at three different intersections at different distances from the northern area of Retiro Park (Table 2). The temperature data extracted at points A, B and C were close to the average park temperature (mid temperature taken by the AEMET weather station (AEMET, 2018) on the data extraction days) (Table 6). Based on Fig. 6, the temperature range for points A, B and C (gray rectangle) is between the maximum and minimum extracted park temperature (linear range). It is noteworthy that among the three areas, the average temperature of point A was closer to the average temperature of the park (mid temperature) (Table 6) compared to the other points. AEMET data showed that the lowest and highest temperatures pertaining to the data extraction days in the Retiro Park was related to the 4:50-6:00 and 13:40-14:40 CEST timeframes.

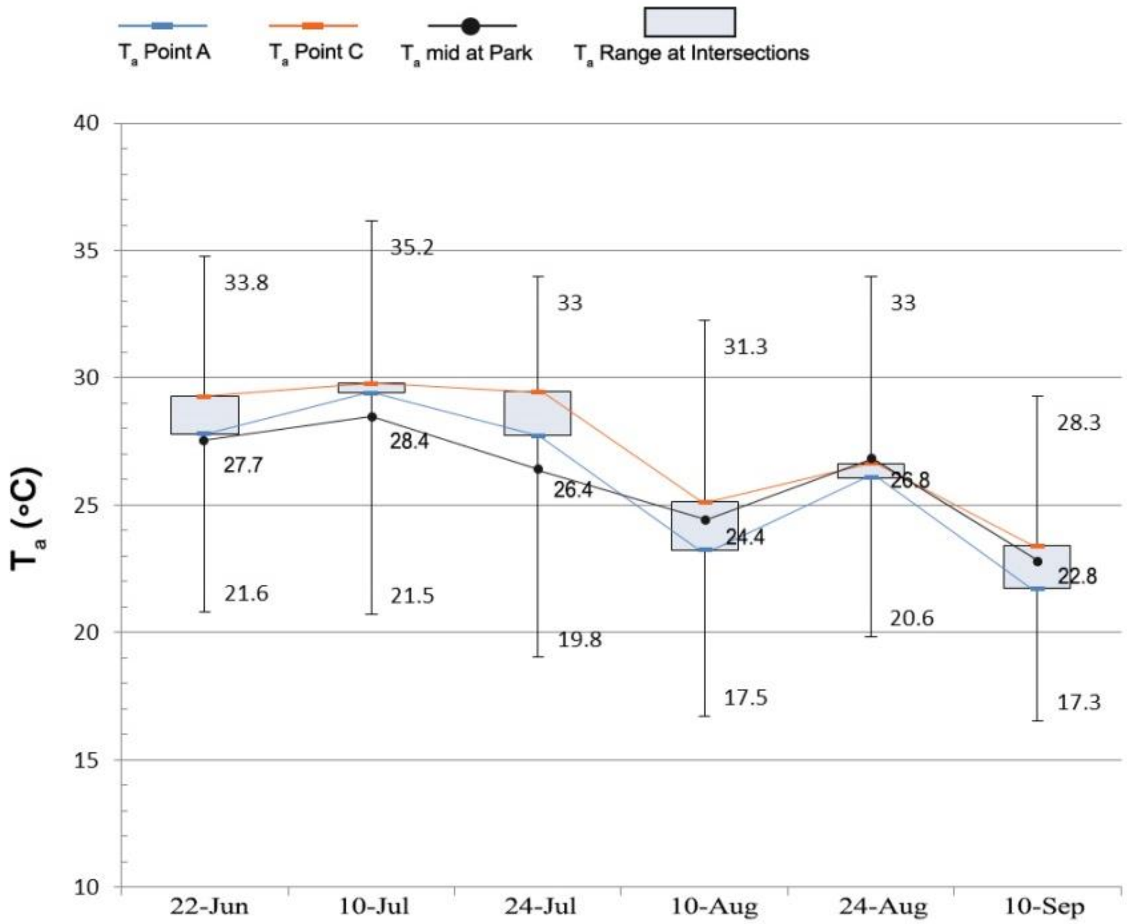

Figure 6. The diagram shows the temperature range of the three intersections A, B & C measured and the temperature range of the Rétiro Park during the 6 days of summer 2018 in Madrid.
Gray rectangles: temperature tange at three intersections (bottom side: T_a A, top Side: T_a C).
Linear range: the temperature range of the park is based on AEMET (bottom line: T_a mid of park, top line: T_a Max of park.

Table 6. Mean values for air temperature (T_a), in the three intersection (10:40–12:10 CET) and values for air temperature (T_a) relative humidity (RH) and wind velocity (W) in the Retiro Park on all the measurement days.

Date	T_a[a] (°C)			Retiro Park[b] T_a (°C)			Time of T_a in Retiro Park[b]		HR % Park[b]	Wind in Park[b]
	Part A	Part B	Part C	Min	Mid	Max	Min	Max		

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22.06.2018	27.81	28.87	29.26	21.6	27.7	33.8	04:50	14:40	**22.95**	**1.7**
10.07.2018	29.4	29.7	29.8	21.5	28.4	35.2	06:00	13:50	22.95	2.2
24.07.2018	27.74	28.1	29.45	19.8	26.4	33.0	05:00	13:50	22.94	1.9
10.08.2018	23.22	23.77	25.12	17.5	24.4	31.3	05:40	13:40	36.8	2.2
24.08.2018	26.05	26.2	26.63	20.6	26.8	33.0	05:30	14:20	19.25	1.4
10.09.2018	21.73	23.13	23.41	17.3	22.8	28.3	05:20	13:45	33.55	1.9

[a] Field measurement, [b] AEMET data.

The mean T_a for points A, B and C are 25.99 °C, 26.62 °C, and 27.27 °C, respectively. The T_a difference over the measurement days between distance of 150 m (A) and 380 m (B) from the Park was about 0.63 °C, between B and C (distance of 665 m) was about 0.65 °C and for the difference between A and C, the temperature difference was about 1.28 ° C. According to the results, as the distance from the park increases, the temperature and its difference with areas closer to the park will be greater.

4.1. Assessment of the park's cooling effecting on thermal comfort indices

In this section, PET data and questionnaire results were used to assess the cooling effect of Retiro Park on thermal comfort. PET data are derived based on the indices standard and the perceived thermal effect is obtained based on the average questionnaire and cognitive maps data. By using both these data, the extent of thermal comfort originating from the Retiro Park cooling effect at distances A, B and C can be deduced as physiological and psychological points of view.

4.1.1. Comparison of the park cooling effect significance on PET at different distances from the park

In a more accurate analysis using SPSS software, one way Anova was used to analyze the relationship between distances A, B and C from the park and the relevant PET differences. Upon every variance analysis and in the case of significant mean difference (the significant level of P-value less than 0.05), Tukey test, which is a series of post-hoc tests, were used to accurately determine which of the average of variable has a significant difference (Barnett, Neter, and Wasserman, 2006). Essentially, by using this test, the thermal comfort relationship was deduced based on the PET index which was obtained using the Ray Man software on distances A, B and C.

Table 7. Anova PET analyses in the three investigated selected points.

	Sum of Squares	df	Mean Square	F	P-value
Between Groups	151.731	2	75.865	3.645	0.029
Within Groups	2955.708	142	20.815		
Total	3107.439	144			

The results of the Anova test (Table 7) shows that concerning the PET index, there is lower significant error rate (0.05) and 95% confidence level of a significant difference between average PET data in each of the A, B and C points (P-value =0.029). Therefore, Tukey's post-hoc tests were used to determine whether there is a significant difference between each of the A, B and C intersections.

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Table 8. Tukey PET analyses in the three investigated intersections (A, B and C).

(I) Part	(J) Part	Mean Difference (I-J)	Std. Error	P-value	95% Confidence Interval		Part	N	Subset for alpha = 0.05	
					Lower Bound	Upper Bound			1	2
A	**B**	**-1.99624**	**.94199**	**.090**	**-4.2274**	.2349	A	49	29.3327	
	C	-2.28303*	.91265	.036	-4.4447	-.1214	B	45	31.3289	31.3289
B	A	1.99624	.94199	.090	-.2349	4.2274	C	51		31.6157
	C	-.28680	.93311	.949	-2.4969	1.9233				
C	A	2.28303*	.91265	.036	.1214	4.4447	Sig.		.084	.949
	B	.28680	.93311	.949	-1.9233	2.4969				

*. The mean difference is significant at the 0.05 level.

The results of this studied grouping test are presented in Table 8. According to this table, it can be stated that there is a significant relationship between the mean PET index at points A and C but the mean PET index between point B with points A and C does not entail a significant difference. The mean PET for A, B and C is 29.3 °C, 31.3 °C and 31.6 °C respectively. Essentially, it can be stated that the PET index difference among intersections A and C is explicit and significant (P-value =0.036). The mean of this index in A has a difference of 2.3 °C with C, whilst the difference between A with B is about 2 °C which is not considered significant in terms of the Tukey test (P-value= 0.09).

4.1.2. Comparison of the significance of the questionnaire dataset PTC at difference distances from the park

In order to compare the significance of the PTC at difference distances from the park, the mean dataset extracted from the questionnaire for each section was derived as the PTC. The questions asked by citizens via the questionnaire was about how they felt about the ambient temperature and the effect of Retiro Park on such feelings in the form of four questions (Table 3).

Based on the results of the Anova Statistical test (Table 9) and a lower significance error level (p-value<0.05), there is 95% confidence pertaining to a significant difference between mean thermal comfort index (PTC) in each of the A, B and C regions (p-value=0.032). Thus, Tukey's post-hoc test was used to determine the significant difference between each of the intersections.

Table 9. Anova PTC analyses in the three investigated selected points.

	Sum of Squares	df	Mean Square	F	P-value
Between Groups	2.318	2	1.159	3.517	0.032
Within Groups	46.809	142	.330		
Total	49.128	144			

Since the rating of questions is from very low to very high in the form of numbers 1 to 5, Table 10 shows that the average of the total questions during the 6 days for A, B and C is approximately 3.3., 2.9 and 2.8, such that area A is of the highest share indicating more thermal comfort experienced by people in this area. In interpreting the aforementioned table, it can be stated that there is a significant difference between the mean PTC index at points A and C (P-value <0.05), but the mean PTC at point B with points A and C have no significant difference. Based on this relationship, it is also clear that with an increase in distance from the green area, the perceived comfort experienced by citizens is reduced, thus there is an inverse correlation between increased distance and PTC.

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Table 10. Tukey PTC analyses in the three investigated intersections (A, B and C).

(I) Part	(J) Part	Mean Difference (I-J)	Std. Error	P-value	95% Confidence Interval Lower Bound	95% Confidence Interval Upper Bound	Part	N	Subset for alpha = 0.05: 1	Subset for alpha = 0.05: 2
A	**B**	**.35417**	**.14270**	**.062**	**-.0165**	.7248	C	51	2.8417	
	C	.45000*	.14270	.017	.0793	.8207	B	45	2.9375	2.9375
B	A	-.35417	.14270	.062	-.7248	.0165	A	49		3.2917
	C	.09583	.14270	.783	-.2748	.4665				
C	A	-.45000*	.14270	.017	-.8207	-.0793	Sig.		.783	.062
	B	-.09583	.14270	.783	-.4665	.2748				

*. The mean difference is significant at the 0.05 level.

4.1.3. Comparing the Cognitive Maps significance at difference distances from the park

The meaningfulness of the cognitive maps data with the distance from the park was evaluated using the Anova test. The analysis of this section was performed by using a AramMMA software for analyzing cognitive maps. In this analysis, the maps pointing to the Retiro park were distinguished from maps that did not mention the park. In fact, 100 percent of the score was dedicated to the maps that noted parks, while zero percent was considered for those that did not mention to the park. The result of the Anova test (Table 11) illustrated that there is an acceptable agreement (95%) between the correlation of cognitive maps and the distance from the park (P-value<0.05).

Table 11. Anova Cognitive Maps analyses in the three investigated selected points.

	Sum of Squares	df	Mean Square	F	P-value
Between Groups	36309.751	2	18154.876	8.948	.000
Within Groups	288104.042	142	2028.902		
Total	324413.793	144			

Owing to the agreement, in each of the A, B, C regions Tukey test was used separately and the results are as follow. As illustrated (Table 12), the 87.7%, 60%, and 50.9% of the respondents in the A, B, and C areas, respectively referred to the park. As expected, the highest level of cognitive mapping from the park is in the vicinity of the park area (150 to 380 meters). Nevertheless, in the C region, with 665 meters distance from the park, more than half of the respondents claimed that the site is comfortable thermally thus prefer to spend more time there. The difference between A and C was also meaningful (P-value = 0.000). Actually, the good agreement of this relationship depicts the impact of the Retiro park (as a place with a high level of thermal comfort) on the resident's mental images and psychological dimensions.

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Table 12. Tukey Cognitive Maps analyses in the three investigated intersections (A, B and C).

(I) Part	(J) Part	Mean Difference (I-J)	Std. Error	P-value	95% Confidence Interval Lower Bound	95% Confidence Interval Upper Bound	Part	N	Subset for alpha = 0.05: 1	Subset for alpha = 0.05: 2
A	**B**	**27.75510***	**9.30015**	**.009**	**5.7273**	49.7829	C	51	50.9804	
	C	36.77471*	9.01047	.000	15.4330	58.1164	B	45	60.0000	
B	A	-27.75510*	9.30015	.009	-49.7829	-5.7273	A	49		87.7551
	C	9.01961	9.21244	.591	-12.8005	30.8397				
C	A	-36.77471*	9.01047	.000	-58.1164	-15.4330	Sig.		.589	1.000
	B	-9.01961	9.21244	.591	-30.8397	12.8005				

*. The mean difference is significant at the 0.05 level.

4.1.4. Comparison of the significance of the questionnaire dataset PTC with PET

For a closer assessment of the effect of Retiro Park on thermal comfort, the PET index and PTC by citizens were compared on the basis of the total data extracted from the questionnaire. Essentially, this PET analysis which is a standard index for perceived thermal comfort is compared with the personal opinion of people about their perceptions concerning the thermal comfort of the environment. SPSS software was used to assess the correlation test as well as other statistical tests.

Based on the results of the Pearson correlation test (Table 13), the relationship between the PET index with thermal comfort in the selected points of A and C is significant and is of inverse correlation (P-value <0.05). Due to the higher correlation coefficient, the relationship between the PET index and the PTC in area A (P-Value=0.004) is more significant compared to area C (P-value=0.006). Moreover, the significance of PET index relationship with perceived thermal comfort in area B (P-value=0.061) is not accepted (P-value> 0.05). Based on this analysis, it can be stated that the correlation intensity of these two indicators is significant in area A.

Table 13. Pearson Correlation analyses between PET and PTC.

Part			PET
A	**PTC**	Pearson Correlation	-.404*
		Sig. (2-tailed)	.004
		N	49
B	PTC	Pearson Correlation	-.281
		Sig. (2-tailed)	.061
		N	45
C	PTC	Pearson Correlation	-.379*
		Sig. (2-tailed)	.006
		N	51

4.2. Relations between thermal comfort indices (physiological and psychological)

Regarding the results, by increasing the distance from the park at intersections B and C, despite physical and structural similarities to intersection the A, there is an increase in air temperature (T_a), and consequently the extent of PET and T_{mrt} also changes accordingly.

The data pertaining to the perceived thermal comfort (PTC) acquired from the questionnaire is rated from 1 to 5 (1=very low to 5=very high) and at points A, B and C during 6 days was on average approximately 3.3, 2.9 and 2.8, respectively (Table 10). At point A, residents experienced higher thermal comfort compared to the other points. Essentially, the level of PTC at this location is medium to high according to citizens opinions (medium=3), whereas for B and C, the average perceived thermal comfort for the data extraction days was lower than medium. Furthermore, the results show

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that the extent of PTC at intersection A was higher in all the days (Fig. 7). It is noteworthy that based on statistical data, the perceived thermal comfort is correlated to PET and is of inverse correlation, such that its graph behavior is inversely correlated with the PET graph (Fig 7, 8). For example, on 10th August, where the PET value is at a minimum at point A (PET 24.3 °C), the relevant perceived thermal comfort is at a maximum (3.6) or on 10th July, at point C, the PET was 36.3 °C which was higher compared to other points and days which is inversed on the perceived thermal comfort graph and the least value of this index was at point C with a value of 2.4.

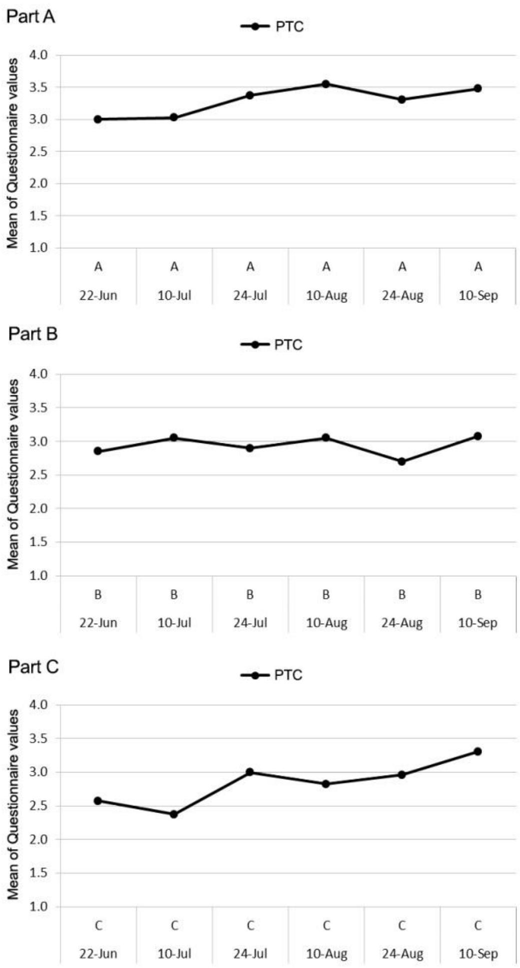

Figure 7. The average data for questionnaire data titled PTC at each three intersections A, B and C during 6 days.

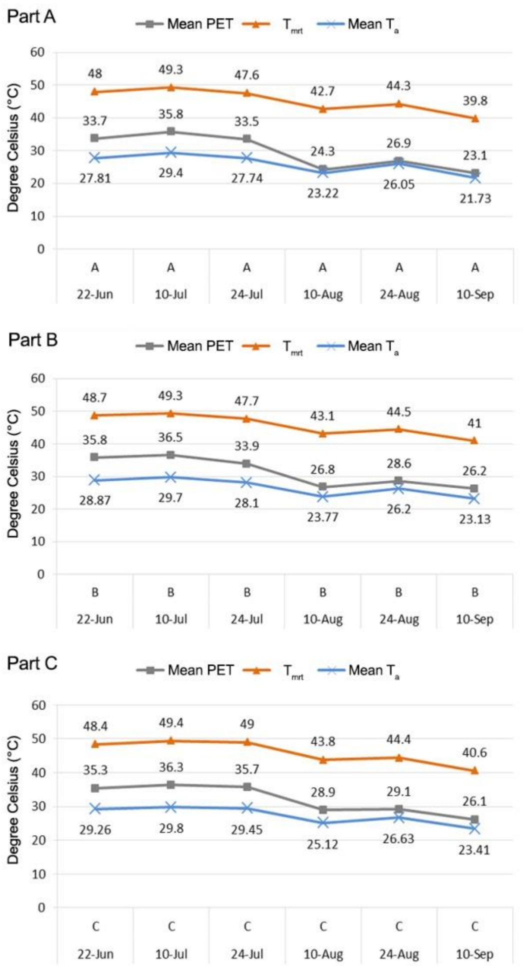

Figure 8. The relationship among mean PET (Gray), mean T_a (Blue), and T_{mrt} (Orange), at each three intersections A, B and C during 6 days.

In addition to the similar behaviors of PET, T_{mrt}, T_a and PTC parameters at each intersection during the 6 days of assessment (Fig 7, 8), the credibility of the data and calculations conducted in this study are approved. In this regard, the role of large urban parks is tangible in providing thermal comfort for citizens and it is necessary to pay more attention at various levels of urban planning and sustainable development.

5. Conclusions

This study examines the potential of large urban parks in providing thermal comfort for citizens living within the park perimeter. The results extracted amidst 6 hot summer days in Madrid shows that large urban parks exhibit a cooling effect. Considering the significance of the mean T_a difference between the distance of 150 m and 665 m from the park (P-value <0/05), as well as the lower temperature of about 1.28 °C pertaining to the distance closest to the park compared to distances further away from the park (under equal conditions), in addition to the insignificant mean T_a difference at the distance of 380 m compared to 150 m and 665 m (P-value> 0/05), we found that the cooling effect of the large urban park at distances close to the park (under 380 meters) has a significant role in temperature reduction.

Accordingly, the level of PET will increase as the distance from the park increases, and residents will perceive less thermal comfort compared to distances closer to the park. The degree of PET index at a distance of 150 meters from the park is on average 2 °C PET and 2.3 °C PET less compared to distances of 380 meters and 665 meters respectively. The PTC of citizens was acquired based on the average obtained data from the questionnaire which showed that people in the vicinity of the park experience more thermal comfort. For the other two regions of the park which were more distant, less thermal comfort was experienced (less than average).

The results of the cognitive maps analyses demonstrated that large parks play a significant role in thermal comfort improvement affecting people's mental map. For people, such parks are a space where they feel more comfortable, thereby spending more time to enjoy the desirable temperature. Although the resultant mental maps were closer to residents in the vicinity of parks, the results demonstrated that at long distances to the park (665 meters) these parks still have a psychological dimension in offering thermal comfort (more than 50%).

PET and PTC as the main variables of this research are inversely correlated such that when the distance from the park is increased, the PET is increased and thermal comfort is decreased. The correlation between the two indices were significant in the nearest and furthest distance from the park (P-value <0/05) and the highest correlation coefficient of the two indices pertains to the distance closest to the park (150 meters). Essentially, this indicates the high level of perceived thermal comfort from the citizens' point of view as well as the PET compared to distant locations.

As a result, the cooling effect of a large urban park is a suitable solution to improve people thermal comfort (as physiological and psychological perspectives) either within the park area or at the vicinity of the park. The results of this study clarify that in areas close to the park, such effect on thermal comfort is more perceptible. Thus, in order to enhance the cooling effect of large urban parks, it is necessary to implement urban design and planning solutions (qualitative and quantitative) to provide more favorable conditions for the lives of citizens.

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