## Geography Internal Assessment

How does environmental quality, determined by the severity of pollution and the availability of green spaces, differ between a tourist-catering area of Paris versus a residential area?

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Word count: 1156

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## 1 Introduction

### 1.1 Fieldwork question

How does environmental quality, determined by the severity of pollution and the availability of green spaces, differ between a touristic area of Paris and a residential area?

In order to answer this question, the fieldwork for our paper was conducted in Paris, one of the largest and most well known cities in the world. The city itself is arranged into 20 districts, known locally as *arrondissements*, are placed in a spiral in the city. It is globally known for its high number of tourists per year, equating to around 35 million in the year 2019 alone. Many of them come to visit the world-renowned Eiffel Tower, located in the VII<sup>th</sup> *arrondissement*. To add, Paris is home to 2.2 million people, who mostly live in the outer residential areas, from the XI<sup>th</sup> to the XX<sup>th</sup> *arrondissements*.

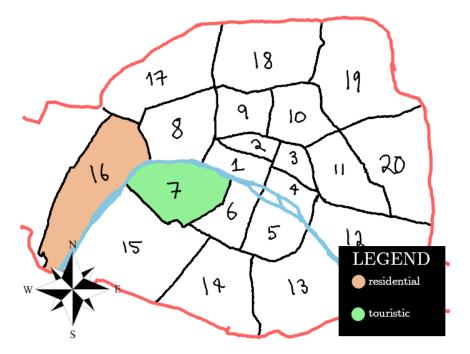


Figure 1: A map of Paris arrondissements, drawn by hand.

## 1.2 Hypotheses

1. According to the Global Development Goals of the UN, 90% of urban areas in the world had polluted air in 2016.<sup>3</sup> Paris was among countries that didn't satisfy WHO's air quality minimum<sup>4</sup> of 2018, with on average a 50% higher than normal pollution density. However, a

<sup>&</sup>lt;sup>1</sup>Statista Research Department. *Hotel arrivals in Paris 2011-2019*. Apr. 2020. URL: https://www.statista.com/statistics/468164/number-tourist-arrivals-hotels-paris.

 $<sup>^2</sup>$ CondorFerries. Latest France Tourism Statistics & Industry Trends (2020-2021). URL: https://www.condorferries.co.uk/france-tourism-statistics.

<sup>&</sup>lt;sup>3</sup> "The Sustainable Development Goals Report 2020". In: *The Sustainable Development Goals Report* (2020), p. 47. DOI: 10.18356/214e6642-en.

 $<sup>^4</sup>Ambient\ (outdoor)\ air\ pollution.$  May 2018. URL: https://www.who.int/news-room/fact-sheets/detail/ambient-(outdoor)-air-quality-and-health.

counterargument to this could be that despite a higher concentration of people, tourist areas in Paris do not suffer as much from high traffic conditions from things such as typical morning rush hours, and tourists preferably using public transport or bikes.

- 2. As there are more people moving about in residential areas, for example in cars for the morning commute or at noon for lunch, it can logically be theorized that noise pollution, which is obviously a function of the amount of cars, would be higher in these places. Today it is estimated that an average noise level of 60dB can be found in residential areas, according to the very comprehensive Bruitparif government-sponsored report. This value largely surpasses the WHO's safe level of 53dB.
- 3. The Parisian mayor, Anne Hidalgo included the improvement of environmental quality in her campaign. The mayor has promised to make so-called "green spaces" no further than 200 meters to any person,<sup>7</sup> and as such, it should be hypothesized that green spaces, which include parks, agglomerations of trees, shall be distributed evenly with no difference between residential and touristic areas. The mayor emphasized on "urban forests" places where residents and tourists alike could enjoy the company of trees while walking along the city streets.

### 2 Method

For our investigation, the topic in question is the environmental quality. We will compare the environmental quality of two areas, one meant as a residential one and one with a heavy tourist presence. To best represent these areas, we have chosen the XVIe and the VIIe. The XVIe is home to many housing complexes and fosters facilities aimed at catering to the residents whereas the VIIe sees many tourists as it is home to the famed Eiffel Tower and the Seine river, prime tourist attractions of Paris.

#### 2.1 Study site choices

We chose 10 sites in total to conduct a bipolar survey, shown below in Figures 2, 3.

<sup>&</sup>lt;sup>5</sup>Mairie de Paris. *PLAN DE PRÉVENTION DU BRUIT DANS L'ENVIRONNEMENT 2015 > 2020*. URL: https://www.bruitparif.fr/PPBE/75056%20-%20Paris/PPBE%20Paris%202015-2020.pdf.

<sup>6&</sup>quot;Environmental Noise Guidelines for the European Region". In: (2018), p. 8. URL: https://www.euro.who.int/\_\_data/assets/pdf\_file/0009/383922/noise-guidelines-exec-sum-eng.pdf.

<sup>&</sup>lt;sup>7</sup>Anne Hidalgo. Comment Paris peut-elle être une ville encore plus végétale? URL: https://annehidalgo2020.com/question/comment-paris-peut-etre-une-ville-encore-plus-vegetale/.

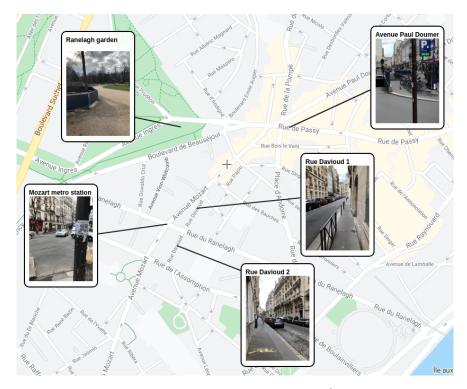


Figure 2: Map of sites chosen for the residential area, the XVI<sup>th</sup> arrondissement. Map base layer courtesy of Google Maps, pictures seen are taken on a mobile phone.

This site was chosen as, like stated before, it is a residential area. There are other residential areas in Paris, however this one was specifically chosen for reasons of convenience.<sup>8</sup>

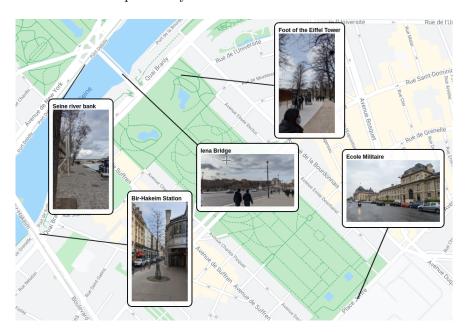


Figure 3: Map of sites chosen for the residential area, the XVI<sup>th</sup> arrondissement. Map base layer courtesy of Google Maps, pictures seen are taken on a mobile phone.

 $<sup>^8\</sup>mathrm{We}$  had 2 members of our group residing in this area.

This area in the VII<sup>th</sup> was chosen as it is a staple of tourism in France. It is home to the famed Eiffel Tower, the Ecole Militaire building and the Seine river bank.

#### 2.1.1 Our sampling method

The sites in the two areas were chosen using a method of stratified sampling. Considering the VII<sup>th</sup> does indeed contain residential sites and the XVI<sup>th</sup> touristic ones, we can discard the ones that are irrelevant to our study.

#### 2.2 Method

#### 2.2.1 Bipolar semantic survey

The sites were visited to conduct a so-called "bipolar survey", which consisted of rating the site based on multiple criteria. This process was done by selecting a single person from our group to visit one of the sites, and complete a bipolar semantic survey. They would grade several different aspects of the site (such as general cleanliness, noise, amount of cars), and take a set of 3 images of the site. Our surveys were conducted during the week of the 10th of March, 2021 in the afternoon each time. The weather was overcast and moderately cold.

Our results for the bipolar survey are shown in Appendix A.1

When tallied, the total scores for each site can be expressed as a fraction over 84, as shown in Figures 4, 5.

Site	Total score
Bir-Hakeim	51
Seine river bank	69
Foot of the Eiffel Tower	45
Iena Bridge	42
Ecole Militaire	54

Figure 4: Total Bipolar Semantic scores for the VII<sup>th</sup>

Site	Total score
Rue Davioud 1	47
Rue Davioud 2	53
Avenue Mozart	31
Avenue Paul Doumer	35
Ranelagh Garden	77

Figure 5: Total Bipolar Semantic scores for the XVI<sup>th</sup>

 $<sup>^9\</sup>mathrm{Images}$  taken for our study can be found at https://github.com/kinnounko/notes/tree/main/geography/ia/images

When plotted on a map, these totals can be shown with a gradient of colors.

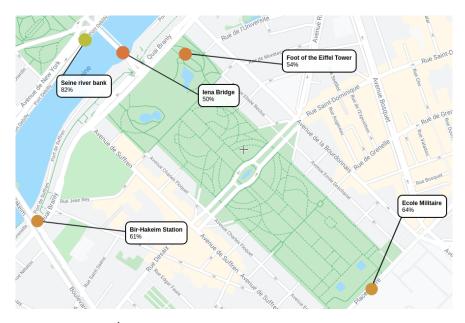


Figure 6: Each site in the VVI<sup>th</sup>, with its respective total score expressed as a percentage for clarity, and a color assigned based on the gradient in Figure 8.

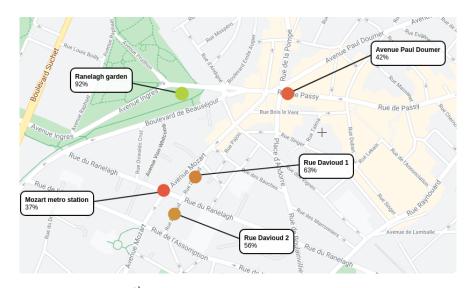


Figure 7: Each site in the XVI<sup>th</sup>, with a percentage. Color assigned with gradient in Figure 8.



Figure 8: Gradient from red (hex color #FF0000) for 0%, to green (hex color #00FF00) for 100%.

#### 2.2.2 Statistical evaluations

In order to either support or disprove our third hypothesis, it is important to study how trees are scattered in Paris, which reveals how dispersed green spaces are from one another. A statistical test called the Nearest Neighbor Index (NNI) test can provide a reasonable answer to this: if the index is similar for both sites, it can easily be said then that our hypothesis is supported. This value also reveals information about the distribution of trees: clustered, random or in a regular pattern, as seen in Figure 5 below. These values will be analysed in the later section 3 Analysis.

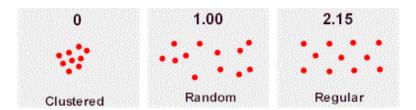


Figure 9: The NNI measures the spacial distribution of data. From a value of 0, representing a clustered pattern, to 1 (random) and to 2.15 (uniform pattern)

As individually counting trees in these areas would be a tedious task, we used the Paris OpenData platform (licensed under the permissive ODbL license<sup>10</sup>), owned by the government. The fact that this kind of database platform exists, shows a certain level of transparency on the government's part.

We used the trees dataset procured from the library, which contains the exact coordinates of trees in the city.<sup>11</sup> There are also other information such as species of the tree and such, however these are cleaned from the dataset for our purposes.

A script, written in Python was tasked to calculate the NNI value, because with a dataset of over 20,000 data points this would not be possible by hand. We go through each point, and find its nearest neighbor. In order to calculate distances between these two points, we use the Haversine Formula, as the two points would be in the form of coordinates. With this information the NNI can be calculated with the simple formula

$$Rn = \frac{2\bar{D}}{\sqrt{\frac{a}{n}}}$$

where Rn is the NNI index value,  $\bar{D}$  is the mean observed distance to the nearest neighbor, a is the area of the zone and n is the total number of data points.

# 3 Analysis

Not done

 $<sup>^{10}\,</sup>Open\,\,Data\,\,Commons\,\,Open\,\,Database\,\,License\,\,(ODbL)\,\,-\,\,Open\,\,Data\,\,Commons:\,\,legal\,\,tools\,\,for\,\,open\,\,data.\,\, {\tt URL:}\,\, {\tt https://opendatacommons.org/licenses/odbl/}.$ 

<sup>&</sup>lt;sup>11</sup>Les arbres. May 2021. URL: https://opendata.paris.fr/explore/dataset/les-arbres/information.

# 3.1 Environmental and cultural sustainability

 $Not\ done$ 

# 3.2 Economic sustainability

 $Not\ done$ 

# 4 Conclusion

 $Not\ done$ 

## Notes

This paper is written with the aid of the typesetting software IATEX. On a PDF viewer, the sections in the table of contents can be clicked to access that section.

All media and information related to this IA can be found on the URL https://github.com/kinnounko/notes/tree/main/geography/ia. This includes images, the IATEX source code to this paper, and the Python program used to calculate the NNI value.

## References

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Open Data Commons Open Database License (ODbL) - Open Data Commons: legal tools for open data. URL: https://opendatacommons.org/licenses/odbl/.

Les arbres. May 2021. URL: https://opendata.paris.fr/explore/dataset/les-arbres/information.

# A Appendix

#### A.1 Bipolar survey results

Shown below are our results from our bipolar semantic survey:

XVI<sup>th</sup> residential area:

	ART				RUE DAVIOUD																		
NEGATIVE	1	2		3	4		5	6	7	POSITIVE		NEGATIVE		1	2	3	4		5	6	7	POSITIVE	
Lots of vandalism			X							No vandali	sm	Lots of vand	dalism		X							No vandalis	n
Noisy				X						Calm		Noisy							X			Calm	
Crowded			X							Spacious		Crowded							X			Spacious	
Lots of cars X										No cars		Lots of cars						X				No cars	
Lots of smokers						X				No smoke	'S	Lots of smo	kers						х			No smokers	
Lots of litter	х									No litter		Lots of litte	r			×	(					No litter	
Poorly maintained				Х						Well maint	tained	Poorly mair	tained					Х				Well mainta	ned
No nature X										Lots of nat	ure	No nature										Lots of natu	e
Overfilled bins	X									Empty bin:	s	Overfilled b	ins						X			Empty bins	
No recycling bins	X									Empty bins Lots of recycling bins		No recycling	bins					X				Lots of recy	ling bins
Foul / polluted air				Х						No noticea	No noticeable odor		ted air			×	(					No noticeab	le odor
Lots of dog feces	Х									No dog fed	es	Lots of dog	feces		Х							No dog fece	5
										Total: 31/	B4											Total: 47/84	

				F	RUE DA	VOIU	D 2																	
NEGATIVE		1	2		3	4		5	6		7	POSITIVE						RANE	LAGH J	ARDIN				
Lots of van	dalism				-			х				No vandalism		NEGATIVE		1	2	3		4	5	6		7 POSITIVE
Noisy	/	/		1	/		х	/		/		Calm		Lots of vanda	lism								X	No vandalism
Crowded	′	- /		′	- /		X	- /		/		Spacious		Noisy							Х			Calm
Lots of cars							^	х				No cars		Crowded							Х			Spacious
Lots of smo								^				No smokers		Lots of cars									X	No cars
Lots of litter										х		No litter		Lots of smoke	ers								X	No smokers
										X				Lots of litter							х			No litter
Poorly maintained								Х				Well maintained		Poorly maint	ained								Х	Well maintained
No nature		Х										Lots of nature		No nature									X	Lots of nature
Overfilled bins					Х							Empty bins		Overfilled bir							Х			Empty bins
No recyclin				Х								Lots of recyclin		No recycling							Х			Lots of recycling bi
Foul / pollu								Х				No noticeable	odor	Foul / pollute									X	No noticeable odo
ots of dog feces				X								No dog feces		Lots of dog fe	eces					Х				No dog feces
												Total : 53/84												Total: 77/84
				AVEN	IUE PA	UL DC	DUMER																	
NEGATIVE		1	2		3	4		5	6		7	POSITIVE												
Lots of vand	dalism	х										No vandalism												
Noisy		Х										Calm												
Crowded		X										Spacious												
ots of cars	Х											No cars												
ots of smo	kers			X								No smokers												
Lots of litte	r				Х							No litter												
Poorly mair	ntained							Х				Well maintaine	ed											
No nature		X										Lots of nature												
Overfilled b	ins			X								Empty bins												
lo recyclin					X							Lots of recyclin												
oul / pollu	ted air				х							No noticeable	odor											
ots of dog	feces			Х								No dog feces												
												Total: 35/84												

# ${\rm VII^{th}}$ touristic area:

				BIR HA	KEIM	METRO	STATION																	
NEGATIVE		1	2	,	3	4		5	6	7	POSITIVE							ECOLE 1	/ILITAIRE					
Lots of vand	aliem	-	-		3		,	Х			No vandal	icm	NEGATIVE		1	2		3	4	5		6	7 POSITIVE	
Noisy	alism	х						^			Calm	ISITI	Lots of var	dalism			х						No vandalis	m
		^											Noisy						х				Calm	
Crowded							Х				Spacious		Crowded							)	(		Spacious	
ots of cars					Х						No cars		Lots of car	s			Х						No cars	
Lots of smok							Х				No smoke	rs	Lots of sm	okers			Х						No smokers	
ots of litter				Х							No litter		Lots of litte	er				х					No litter	
Poorly main	tained			X							Well main		Poorly mai	ntained					х				Well mainta	ained
No nature				Х							Lots of nat	ture	No nature						х				Lots of natu	ıre
Overfilled bi	ins				X						Empty bin	s	Overfilled	bins						)	(		Empty bins	
No recycling	bins						Х				Lots of rec	ycling bins	No recyclin	ng bins					х				Lots of recy	
oul / pollut	ed air						Х				No noticea	able odor	Foul / polls					х					No noticeal	
ots of dog f	feces							х			No dog fee	es	Lots of do						х				No dog fece	
											Total: 51/8												Total: 54/84	
					OF TH	IE EIFFE	L TOWER											PONT	D'IÉNA					
NEGATIVE		1	2	2	3	4		5	6	7	POSITIVE		NEGATIVE		1	2		3	4	5		6	7 POSITIVE	
Lots of vanc	Х										No vandal	ism	Lots of var	ndalism								х	No vandalis	m
Noisy					Х						Calm		Noisy						Х				Calm	
Crowded		Х									Spacious		Crowded	Х									Spacious	
ots of cars					Х						No cars		Lots of car	s	Х								No cars	
Lots of smol	kers			Х							No smoke	rs	Lots of sm	okers	Х								No smokers	
Lots of litter				Х							No litter		Lots of litt	er			Х						No litter	
Poorly main	tained							X			Well main	tained	Poorly ma	intained						>	<		Well mainta	ined
No nature								Х			Lots of nat	ture	No nature		Х								Lots of natu	ıre
Overfilled bi	ins						X				Empty bin	s	Overfilled	bins			Х						Empty bins	
No recycling	bins			Х							Lots of rec	ycling bins	No recycli	nį X									Lots of recy	cling bir
Foul / pollut	ed air							X			No noticea	able odor	Foul / poll	uted air					X				No noticeal	ole odor
Lots of dog t	feces			Х							No dog fee	ces	Lots of do	g feces						>	(		No dog fece	s
											Total: 45/8	34											Total: 42/84	ŀ
						OF THE S																		
NEGATIVE		1	2	2	3	4	,	5	6	7	POSITIVE													
Lots of vand	lalism							Х			No vandali	ism												
Noisy									X		Calm													
Crowded								Х			Spacious													
ots of cars									Х		No cars													
ots of smol								Х			No smoke	rs												
ots of litter					Х						No litter													
oorly main	tained						Х				Well maint													
No nature					Х						Lots of nat													
Overfilled bi								Х			Empty bin													
No recycling							Х				Lots of rec													
Foul / pollut									Х		No noticea													
Lots of dog t	feces							х			No dog fed	es												
											Total: 69/8	14												