Estimating emissions from tourism activities

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# 1 Estimating emissions from tourism activities 2 M. A. Russo<sup>1\*</sup>, H. Relvas<sup>1</sup>, C. Gama<sup>1</sup>, M. Lopes<sup>1</sup>, C. Borrego<sup>1</sup>, V. Rodrigues<sup>2</sup>, M. Robaina<sup>2</sup>, M. 3 Madaleno<sup>2</sup>, M. J. Carneiro<sup>2</sup>, C. Eusébio<sup>2</sup>, A. Monteiro<sup>1</sup> 4 <sup>1</sup>CESAM and Department of Environment and Planning, University of Aveiro, Portugal 5 <sup>2</sup> GOVCOPP and Department of Economics, Management, Industrial Engineering and Tourism, 6 University of Aveiro, Portugal 7 \*Corresponding author: michaelarusso@ua.pt | Phone: +351 234370220 | Fax: +351 234 370309 8 9 ABSTRACT 10 Data on atmospheric pollutant emissions from tourism activities was identified as a critical 11 knowledge gap. Building an emissions inventory is a standard procedure that most countries 12 perform for regulatory or research purposes. At a European level, these inventories are 13 developed using the standard Nomenclature for Reporting (NFR sectors). However, none of the 14 NFR are exclusively for tourism or explicitly include it. This paper presents a methodology to 15 estimate the emissions from main touristic activities, focusing on Portugal as a case study. The 16 emissions were distributed using tourism data as a proxy, namely the contribution of tourism to 17 characteristic industries, as well as the nights spent in tourism establishments by non-residents.

- 18 The proxy data was used to distribute emissions throughout the municipalities, using the 19 national reported emissions data as a starting point. An analysis of the spatial distribution of 20 tourism emissions was performed, highlighting that tourism has a significant impact on 21 atmospheric emissions over specific areas (up to 40.1%), and contributing to areas where air 22 pollution is already an environmental stress factor (urban centres of Porto and Lisbon). While 23 this methodological framework was developed specifically for Portugal, it may be adapted to 24 assess atmospheric pollutant emissions from tourism activities in other regions. Beyond the 25 methodology proposed and the analysis of the results, other alternative methods to estimate 26 emissions from the tourism sector are discussed and suggested.
- 27

# 28 Key words: air pollution; atmospheric emissions estimate; tourism; economic sectors

29

# 30 **1. INTRODUCTION**

Recently, tourism has been identified as one of the largest sources of externalities and responsible for the overexploitation of certain environmental resources (Jones and Munday, 2004). Therefore, increased awareness on the issue of tourism has led to it being a central discussion point in the scientific community (Becken et al., 2017; Saenz-de-Miera and Rosselló, 2014). An increase in travel and other services industries has both direct, indirect and induced environmental impacts, causing the same forms of pollution as any other industry: air emissions, noise, solid waste, or even architectural/visual pollution. While extensive research has

38 documented the significant economic impact of such service industries as tourism, little has 39 been written about their effect on environmental quality (Saenz-de-Miera and Rosselló, 2014), 40 specifically on how air pollution will affect tourists' experiences (Law and Cheung, 2007) and 41 visitors' quality of life (Eusébio and Vieira, 2013). The majority of publications related to air 42 quality impacts indicate that air pollution is closely linked to increased premature mortality and 43 hospitalization induced by a number of diseases, with the most prevalent being of respiratory 44 origin (Costa et al., 2014). Compared with residents in polluted areas, tourists are more 45 susceptible to acute effects (Zhang et al., 2015). Among the externalities related to tourism, 46 greenhouse gas (GHG) emissions have become a recurring topic of discussion in literature 47 (Becken and Simmons, 2008), which has also included global warming issues (Becken, 2002). 48 Some regions have registered an exponential growth in tourism, making them an interesting 49 case study for the link between tourism and atmospheric pollution (UNWTO, 2010). There have 50 been studies focusing on the impacts of negative environmental factors on tourism, how it 51 affects visitor perception of atmospheric pollution and its connection to an increased trip 52 dissatisfaction and reduced likelihood of visitors to return (Jarvis et al., 2016). In some cases, 53 during peak air pollution episodes, monthly visitors to certain locations could decrease by more 54 than 25 000 people, as poor air quality discourages some tourism activities (Chen et al., 2017). 55 Heavily polluted areas can also suffer from reduced visibility, which may change tourists' 56 perceptions and decrease enjoyment (Anaman and Looi, 2000; Latif et al., 2018; Law and 57 Cheung, 2007; Zhang et al., 2015).

58 Even though economic activities have long been related with air pollution, such as energy 59 production (Casler and Blair, 1997) or transport (Peeters et al., 2007), tourism has only recently 60 been investigated as a potential cause for these environmental issues (Saenz-de-Miera and 61 Rosselló, 2014). To date, the majority of studies have focused on translating tourism into  $CO_2$ 62 emissions as a way of quantifying its environmental impact. This has been achieved by 63 gathering data regarding energy consumption and generated waste, and then applying a  $CO_2$ 64 emissions factor to the data (Basarir and Cakir, 2016; Katircioglu et al., 2014; Ng et al., 2016; 65 Rosselló-Batle et al., 2010). For an extensive air quality analysis, detailed emissions for 66 atmospheric pollutants are required for each activity sector. Currently, there are no studies 67 where an emissions inventory was built specifically for tourism.

Nowadays, Portugal is one of the most important tourism worldwide destinations. The international recognition of Portugal as a tourism destination has increased considerably in last years. Consequently, in 2018, this country received the title of World's Leading Destination, in the World Travel Awards. In this country, tourism is one of the most important economic activities. According to the World Travel & Tourism Council (World Travel & Tourism Council, 2018) the total contribution (direct, indirect and induced effects) of travel and tourism to Gross Domestic Product (GDP) was of 17.3%. In terms of employment, 20.4% of the total

r5 employment is generated, directly and indirectly, by the tourism industry. Therefore, the main

76 objective of paper is to quantify direct emissions from tourism in each municipality in Portugal,

as a first step in developing the data required for an in-depth air quality analysis.

78 The paper is organized as follows. In section 2, the data used and methodology developed to 79 estimate emissions from tourism are detailed. In section 3, the total emission values and spatial 80 distribution of the emissions throughout the country are presented. Finally, in section 4, the

- 81 main conclusions are summarised.
- 82

# 83 2. DATA & METHODS

84

# 85 2.1 Tourism data

To estimate the impact of tourism on air quality, 2015 data from the Portuguese Tourism Satellite Account and Tourism Statistics were used as it is the most up to date data available. First, in order to analyse the direct economic relevance of tourism, the Gross Value Added (GVA) together with the GVA generated by tourism characteristic activities (GVAGT) were used (Table 1)

91

Table 1. Contribution of tourism characteristic activities to the Gross Value Added of Portugal
 2015 (INE, 2019)

Tourism characteristic activities	Total GVA (a) [€/year]	Total GVAGT (b) [€/year]	% GVAGT ((a/b)*100)
Hotels and similar	3 263 946	3 197 032	97.95
Second homes - own account	1 066 429	1 066 429	100.00
Restaurants and similar	5 281 649	2 412 898	45.68
Railway passenger transport	192 157	106 015	55.17
Road passenger transport	966 251	234 460	24.26
Water passenger transport	79 019	57 143	72.32
Air passenger transport	903 142	610 028	67.55
Passenger transport supporting services	2 729 962	68 802	2.52
Passenger transport equipment rental	856 350	320 589	37.44
Travel agencies and similar	269 744	194 205	72.00
Cultural services	434 612	146 067	33.61
Sports and recreational services	602 516	177 293	29.43
Connected activities	3 504 143	233 234	6.66
Non-specific activities	136 688 984	1 633 455	1.20
Total	156 838 904	10 457 651	6.67

In 2015, the GVA generated by tourism characteristic activities represented 6.67% of national
GVA. However, an analysis of the different tourism characteristic activities clearly reveals a
great variety in the contribution of tourism to the GVA of theses economic activities. For

98 example, in the case of tourism accommodation (hotels and similar), it is responsible for
99 97.95% of the total GVA generated, while in the case of road passenger transport, tourism only
100 contributes 24.26% to the total GVA.

101

# 102 **2.2 Emission data**

In the EU, the official reporting of emissions under the UNECE convention (EMEP protocol) adopted the Nomenclature for Reporting (NFR) sectors to develop emissions inventories; these represent different activities for which emissions must be estimated. Each year, member states are required to develop national emissions inventories using this system and update the data in the EMEP database (https://www.emep.int/).

108 NFR are aggregated into Gridding NFR (GNFR), which are more encompassing sectors that 109 include each NFR related to the same general activity. Currently there are 14 GNFR, and 110 various NFR for each of them in the Portuguese Inventory Report developed by the Portuguese 111 Environmental Agency, with NFR emissions detailed at a municipality level. Since only a few 112 of them are going to be detailed in this paper, for more information regarding which GNFR 113 sectors exist and what they encompass, refer to CEIP 2019.

114 For an overview of the sectors and their contributions in terms of emissions, Figure 1 displays

115 each of the reported sectors and highlights those that are directly linked with tourism and a

116 focus of this study. The emission data shown below (for 2015) includes the total emission

117 values for each of the reported sectors and their contributions to the national total per studied

118 pollutant. This work will focus on NOx (NO + NO<sub>2</sub>) and PM10, since they are two critical

119 pollutants in Portugal that regularly exceed legislated air quality limit values in the country

120 (APA, 2019). Nonetheless, a brief exposition of data regarding SO<sub>2</sub> and CO emissions (critical

121 pollutants for Aviation and Shipping, which are sectors strongly connected to tourism), is also

122 included in section 3.1 and Figure 2.



Figure 1. 2015 national emission totals by GNFR sector for NOx (black) and PM10 (grey) in kilo tonnes (kt) per year with highlights for sectors directly linked to tourism (dotted lines).

126

With a contribution of 40.0%, road transport is the largest source for NOx emissions, followed by the industrial sector with 23.9%. In each of these sectors, the emissions mostly originate from internal combustion engines or industrial combustion processes. This explains why PM10 contributions are lower for these sectors, since combustion is the main source for NOx.

Regarding PM10 emissions, the highest contributors are industry and stationary combustion, with 36.8% and 27.1%, respectively. Stationary combustion accounts for residential combustion emissions, which are a significant contributor to PM10 emissions due to cooking, heating and auxiliary engines that primarily use biomass or fossil fuels (Carvalho et al., 2009). Both aviation and shipping emissions are residual when compared to these other sectors.

None of the tourism activities is directly linked to the national reported emission sectors (NFR), so to estimate the contribution of tourism to total emissions, it was necessary to estimate the share of tourism in each NFR sector. The GNFR/NFR pairs, along with the tourism characteristic industries to which they can be related to, are identified in Table 2.

- 140
- 141

# Table 2. GNFR and NFR sector pairs associated to tourism activities

		Corresponding	
GNFR	NFR	tourism	<b>Emissions calculation</b>
		characteristic	methodology
		industry	
C_OtherStationaryComb	Commercial/	Hotels and similar	Emissions are estimated from fuel

Journal Pre-proof			
	institutional: Stationary	&	sales for each municipality (APA,
	&	Restaurants and	2017).
	Residential: Stationary	similar	
	Road transport:		
	Passenger cars	Road passenger transport	Emissions from road transport
F_RoadTransport	&		were calculated using the
	Road transport: Heavy		COPERT V model (APA, 2017).
	duty vehicles and buses		
	Railways	Railway passenger transport	Emissions estimates are calculated
I Offroad			using railway fuel consumption
			and pollutant emission factors
			(APA, 2017).
	international aviation LTO	Emissions are estimated from	
H_Aviation	&	transport	Landing/Take-off cycles (LTO)
	Domestic aviation LTO	transport	(APA, 2017).
			The STEAM shipping emissions
G_Shipping	National navigation (shipping)	See section 2.3.1	model was used to calculate
			emissions from ships (Jalkanen et
			al., 2009; Johansson et al., 2017;
			Russo et al., 2018).

Since aviation and shipping emissions cannot be distributed throughout the municipalities, the analysis of these sectors focused on national totals. While to understand how emissions from the other sectors are distributed throughout the country, the spatial comparison between total and tourism emissions is shown in the figures included in the results section.

147

# 148 **2.3 Tourism emissions estimation**

In this section, the methodology applied to each of the sectors is detailed. The objective was to link a tourism indicator with the emissions from each GNFR to estimate tourism emissions in each municipality, using the national reported emissions data as a starting point. Note that the same methodology was used to estimate values for both NOx and PM10 emissions.

153 To calculate the overall tourism indicator, the total GVA for relevant tourism characteristic

- 154 industries (listed in Column 3 of Table 2) and the corresponding GVAGT were used to obtain
- 155 the percentage of tourism in each sector (cross referencing the data from Table 1 and Table 2).
- As already stated, the data used was for the year 2015, for both emissions and tourism activitydata.

158

# 159 2.3.1 National percentage of Tourism in each GNFR

160 In this section, the methodology for the emissions estimate calculation is detailed. For the 161 stationary combustion, railway transport and aviation emissions, the methodology was

	Journal Pre-proof
162	straightforward. Using Eq. 1, the percentage of the corresponding NFR (Table 1) was calculated
163	directly.
164	
	$\% \text{TOUR}_{\text{GNFR}} = \frac{\text{Reported Emission}_{\text{NFR}}}{\text{Reported Emission}_{\text{GNFR}}} * 100 * \% \text{GVAGT}_{\text{GVA}} $ (Eq. 1)
165 166 167 168 169 170	<ul> <li>where,</li> <li>%TOUR<sub>GNFR</sub> – is the percentage of tourism in GNFR</li> <li>Reported Emission<sub>NFR</sub>- is the emissions reported for the NFR</li> <li>Reported Emission<sub>GNFR</sub>- is the emissions reported for the GNFR</li> <li>%GVAGT<sub>GVA</sub>- is the percentage of tourism in characteristic industries' GVA</li> </ul>
171	For road transport and shipping emissions, additional steps were needed according to the
172	available data for each of these sectors.
173	First, since passenger transport is divided into two separate NFR, namely Passenger cars and
174	Heavy duty vehicles and buses, in Eq. 1, the Reported $Emission_{NFR}$ variable needs to be the
175	sum of passenger cars and buses. To separate heavy duty vehicles and buses, the number of each
176	vehicle class, the average pollutant emission factor per kilometre and the distance travelled for
177	each class, were used to calculate their respective emissions (Eq. 2). The ratio of heavy duty
178	vehicles to buses was found by comparing those values to the reported national total (truck
179	emissions were calculated using the same method as buses).
180	
	$\%Bus_{NFR} = \frac{\sum n^{\circ} Buses * D_{travelled bus} * EF_{bus}}{Total Bus_{emiss} + Total Heavy Truck_{emiss}} * 100 $ (Eq. 2)
181 182 183	<ul> <li>where,</li> <li>%Bus<sub>NFR</sub> – is the percentage of bus emissions in NFR</li> <li>n<sup>9</sup> Buson – is the number of buson of each vahiale class</li> </ul>
185 184	<ul> <li>n<sup>e</sup> Buses - is the number of buses of each vehicle class</li> <li>D_translad has - is the average distance travelled per bus vehicle class</li> </ul>
185	• $EF_{bus}$ – is the emission factor for each bus vehicle class
186	• Total Bus <sub>emiss</sub> – is the total bus emission value
187	<ul> <li>Total Heavy Truck<sub>emiss</sub> – is the total heavy truck emission value</li> </ul>

• Total Heavy Truck<sub>emiss</sub> – is the total heavy truck emission value

189 For shipping emissions, two datasets from the STEAM model (Jalkanen et al., 2009) were used. 190 One is the result of a simulation considering all ships as emission sources. The other is a 191 simulation for ships that were considered as entirely dedicated to tourism, cruise ships. 192 Therefore, instead of using an estimate from the GVAGT data (Table 1), cruise ship traffic 193 emissions in an area up to 100 km from the Portuguese coast were compared to total shipping 194 emissions in the same area.

195

188

### 196 2.3.2 Spatial distribution of tourism emissions

197 To allocate tourism emissions to each municipality throughout the country, each GNFR was 198 treated differently according to the available proxy data. 199 First, since stationary combustion is closely linked with lodging establishments, restaurants and 200 similar commercial businesses, the spatial distribution factor used was the nights spent by non-201 residents in lodging establishments (hotels and similar). This corresponds to an indicator that 202 provides information on how many tourists are in each municipality, which is the equivalent of 203 a percentage of tourism in each municipality for this sector. 204

> TOUR Emiss<sub>mun stat comb</sub> = Emissions<sub>GNFR</sub> \* % TOUR<sub>GNFR</sub> \* % TOUR<sub>mun</sub> (Eq. 3)

205 where.

- TOUR Emiss<sub>mun stat comb</sub> is the stationary combustion tourism emission in the 206 207 municipality 208
  - Emissions<sub>GNFR</sub> is the national emissions for GNFR sector
  - %TOUR<sub>GNFR</sub> is the percentage of tourism in GNFR
  - %TOUR<sub>mun</sub>- is the percentage of tourism in the municipality •
- 210 211 212

209

213 Second, as there is no data with higher detail to differentiate each of the municipalities 214 regarding road transport, a flat percentage was applied. This assumption has its limitations, 215 since the number of tourists and the type of transportation used vary for each municipality. 216 Whenever possible, proxy data with higher detail should be used for this type of disaggregation, 217 for example, data regarding rental car and taxi services or a description of the car fleet in each 218 municipality.

219

$$TOUR Emiss_{mun road} = Emissions_{GNFR mun} * \% TOUR_{GNFR}$$
(Eq. 4)

220 where. 221 TOUR  $Emiss_{mun \ stat \ comb}$  – is the road transport tourism emission in the municipality 222 Emissions<sub>GNFR mun</sub> – is the GNFR emissions for the municipality 223 %TOUR<sub>GNFR</sub> – is the percentage of tourism in road transport GNFR 224 225 Finally, for off-road emissions, using the same equation as road transport, the %TOUR<sub>GNFR</sub> for 226 rail passengers was used to calculate the rail emissions in each region. If the %TOUR<sub>mun</sub> is 0% 227 or if the municipality does not have rail infrastructures, the emissions in this municipality are 228 zero. 229 230 **3. TOURISM EMISSIONS** 

231 In this section, emissions for C\_OtherStationaryComb, F\_RoadTransport and I\_Offroad are 232 compared in terms of the contribution of tourism to each of these sectors. A brief analysis of

total values is presented first, and then the spatial distribution of both total and tourismemissions throughout the municipalities in Portugal is analysed and discussed.

235

# 236 **3.1 Total emissions**

Figure 2 shows total and tourism emission values for the three studied sectors,
C\_OtherStationaryComb, F\_RoadTransport, I\_Offroad, H\_Aviation and G\_Shipping for NOx,

239 PM10, CO and SO<sub>2</sub>.

- 240
- 241



Figure 2. Total and tourism emissions for C\_OtherStationaryComb, F\_RoadTransport and
I\_Offroad, H\_Aviation and G\_Shipping in tonnes per year (NOx, PM10, CO and SO<sub>2</sub>).

245 As seen in section 2.2, F\_RoadTransport is the largest contributor to total NOx emissions of the 246 studied sectors ( $\approx 67.1$  kt), while C OtherStationaryComb is responsible for the highest PM10 247 emission total ( $\approx$  15.6 kt), with tourism having a non-negligible contribution to both of them. 248 For the studied sectors, especially for the aviation sector, tourism represents 67.6% of activities, 249 which is reflected in the emission values of this sector. Relevant sectors for NOx emissions are 250 C\_OtherStationaryComb (20.6%) and F\_RoadTransport (13.1%). For PM10, other than 251 aviation, there is a significant contribution of tourism characteristic industries to emissions in 252 the F RoadTransport sector, accounting for 15.1% of total emissions. In the remaining sector 253 and emission pairs, the contributions only range from 1.1% to 5%. Most of the tourism CO 254 emissions are from H\_Aviation, followed by F\_RoadTransport and C\_OtherStationaryComb, 255 however, overall values (except aviation) of this pollutant are low. Similarly, SO<sub>2</sub> emissions are 256 almost entirely due to G Shipping, which is expected since this sector is the main source of 257 sulphur emissions. Nonetheless, there is still some noticeable contribution to these emissions 258 from C OtherStationaryComb, while others are also quite low.

Is it of note that the total emissions from shipping in Figure 2 are higher than the national reported data from Figure 1. This is due to the national totals only accounting for national maritime navigation, yielding significantly lower results than the methodology applied in the STEAM model, which provides more accurate results (as explained in Russo et al., 2018).

263

# 264 **3.2 Spatial analysis**

The spatial distribution obtained with the described methodology of total and tourism emissions can provide valuable insight into possible hotspots present in the country, and where future strategies regarding tourism characteristic activities can be most effective in reducing their air pollutant emissions. As previously explained, considering the available data, the spatial distribution focused on the stationary combustion, road transport and offroad emissions sector.

Figure 4 shows the C\_OtherStationaryComb sector emissions (total and tourism) for NOx and PM10, in each municipality. Additionally, the percentage of tourism in each municipality calculated above is also shown. The stationary combustion GNFR is divided into three primary NFR, namely residential, commercial/institutional and agriculture/forestry/fishing stationary emissions. As these emissions are closely linked with population, it was expected that their distribution be mainly throughout coastal areas and in some of the more populated inland cities.

C\_OtherStationaryComb

% Tourism



# Total Tourism NOx



Figure 3. Total and tourism C\_OtherStationaryComb emissions for each municipality in tonne 278 279 per year (NOx above, PM10 below), and percentage of tourism in each municipality.

281 Higher values of tourism emissions are in coastal cities and major urban areas, with most of the 282 inland regions in the country having very low tourism or no available data to be allocated to the 283 municipalities. The spatial distribution of tourism emissions also reflects the contribution of 284 commercial/institutional combustion to total GNFR emissions. This is due to the type of 285 combustion related to emission sources in tourism activities (higher influence of services and 286 restaurants) in this NFR having a higher contribution regarding NOx emissions compared to 287 PM10 to overall emissions. Lisbon shows up as the largest hotspot for PM emissions with 1132 288 tonnes of total PM10 emitted each year, contrasted by 72 tonnes due to tourism activities, which 289 accounts for 6.3% of the total value. Regarding NOx, total emission values in Lisbon are the 290 highest, with a contribution of tourism to total NOx emissions is 24.1% (89 tonnes from tourism 291 compared to 369 tonnes total). This is to be expected because it is the most populated city in the 292 country. However, in terms of percentage of tourism, the municipality with the highest 293 contribution of tourism to total emissions is Albufeira in the southern coast (40.1% for NOx and 294 35.8% for PM10). Tourism contributes directly to this sector linking to the commercial and 295 institutional stationary emissions, which includes restaurants, hotels and similar establishments. 296 Figure 4 shows the spatial distribution for the F RoadTransport sector. Road transport is 297 divided into various types of vehicles according to their utility, such as passenger transport, 298 services and heavy vehicles. The connection of this sector to tourism is related to the number of 299 passengers transported, and the most critical pollutant for this sector is NOx.



(NOx above, PM10 below).

304 In this case, the focus on coastal areas is even more evident, as larger city centres and urbanized 305 areas with a large amount of traffic are mostly near the coast. A few more hotspots for these 306 emissions are noticeable, mainly in and around the largest cities, such as Porto in the north 307 (1079 tonnes NOx), Leiria in the centre (1011 tonnes NOx) and the Lisbon metropolitan area 308 (2311 tonnes of NOx for Lisbon alone). As described in the methodology, a flat percentage is 309 applied in each municipality, therefore the percentage of tourism is always the same (14.4% for 310 NOx and 16.7% for PM10). 311 After calculating tourism emissions, major metropolitan areas are still an emission hotspot for 312 both pollutants, especially near Porto (156 tonnes of NOx) and Lisbon (334 tonnes of NOx), for

both pollutants. Contrary to stationary combustion emissions, the distribution of tourism
emissions is not entirely focused on coastal cities (although they are still emission hotspots),
with some inland municipalities still reaching over 100 tonnes of NOx emitted per year.

Finally, Figure 5 shows off-road emissions, which include agriculture, forestry and fishing activities (vehicles and machinery emissions), and railways. The former has no direct contribution to tourism, although it does have activities that can be indirectly related to tourism, while the later can be directly linked to tourism using data regarding transported passengers and their activities.



326 Although still prominent, the spatial distribution of these emissions is less focused on coastal 327 areas, with many inland cities having high NOx emission values. Generally, the vehicles and 328 machinery used in this sector are powered by internal combustion engines and therefore, are 329 similar to road transport. The hotspot for this sector is Matosinhos in the north with 20 tonnes of 330 NOx (1238 tonnes total) and 0.5 tonnes of PM10 (33 tonnes total) emitted due to tourism per 331 year. The total emissions value is in part due to the presence of the Port of Leixões Logistics 332 Platform and associated railway infrastructure. Here the emissions for tourism are overestimated 333 due to other high-emission sources; however, it is still a prominent region for tourism activities. 334 Railway activities have a low contribution to off-road emissions since most of the trains in 335

Portugal are electric, which is why the largest contribution of this sector to atmospheric

336 337

### 338 3.3 Time variation

339 In this section, the time variation of each of the studied sectors and tourism in Portuguese 340 municipalities is investigated. This information is fundamental when using emissions resulting 341 from the methodology suggested in this study, as it focuses on distributing annual emission 342 values for each of the pollutant which has no intrinsic time variation.

pollution could be from indirect impacts related to energy production.

343 Specific Portuguese time profiles used for the GNFR sectors (SNAP 2, 7 and 8) were collected 344 (Menut et al., 2013) and compared with time proxy data related to tourism activity (based on the 345 average of nights spent by non-residents in Portuguese municipalities). Figure 6 shows the 346 studied GNFR sectors and the tourism activity monthly profiles.





Figure 6. Specific Portuguese monthly profiles used for the GNFR studied sectors (top) and
tourism data activity monthly profiles (bottom).

As indicated in the figure, the time profiles are very distinct and none of the emission sectors reflect the temporal evolution of tourism. For example, there is a significant variation from winter to summer in the C\_OtherStationaryComb sector, while road and offroad emissions present almost no variation throughout the year. When using these emissions for air quality simulations, or whenever emissions are input data, these specific time profiles for the tourism sector should be taken into account.

357

# 358 4. SUMMARY AND CONCLUSIONS

359 In order to evaluate the contribution of the tourism sector on the atmospheric pollutants, a 360 methodology to estimate emissions from tourism activities is proposed, using Portugal as a case 361 study. The NFR sectors, recommended for emissions inventories reporting at EU level, were 362 used, in particular the ones that have a direct link to tourism: road and off-road transport, 363 stationary combustion, aviation and shipping activities. The Gross Added Value for 364 characteristic tourism industries was used as proxy data to estimate the contribution of tourism 365 to each economic activity (and corresponding NFR sector). Then, using a specific methodology 366 to each sector the total emissions and their distribution throughout the municipalities in the 367 country was achieved. The analysis of the total emissions suggests that tourism activity is 368 responsible for maximums of 67.6% (both NOx and PM10 for aviation), followed by 20.6% 369 (for NOx in the stationary combustion sector) and 15.1% (for PM10 in the transport sector) of 370 total emissions. The analysis of the spatial distribution of tourism emissions highlighted that 371 tourism has a significant impact on atmospheric emissions over specific areas (up to 40.1%) and 372 contributing to areas where air pollution is already an environmental stress factor (urban centres 373 of Porto and Lisbon). While this methodological framework was developed specifically for 374 Portugal (including the time variations shown, which are specifically for Portugal), the case 375 study may be relevant for many other areas in Europe.

376	
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383	
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- Tourism emissions has a maximum contribution of 67.6% (in the aviation sector)
- Spatial distribution shows significant impact on coastal regions
- Tourism adds to areas where pollution is already an environmental stress factor
- The methodological framework presented is easily applied to other countries

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# **Declaration of interests**

 $\boxtimes$  The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

The authors declare the following financial interests/personal relationships which may be considered as potential competing interests:

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