1	A spatial analysis of childhood cancer and industrial air pollution in a
2	metropolitan area of Colombia
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19	analysis. CJUP participated with obtaining and reviewing data on childhood cancer and
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21	have approved the final manuscript.

22 Abstract

23	Background: Air pollutants are considered carcinogenic to humans. In some European
24	countries, an association with childhood cancer in children has been established. This
25	phenomenon has not been addressed in Latin America, despite the spatial variability of air
26	pollutants that may limit the extrapolation of the results to other geographical areas.
27	Objective: To conduct a spatial analysis of the relationship between childhood cancer and
28	air pollution from industrial sources in a metropolitan area of Colombia.
29	Methods: Incident cases of childhood cancers were obtained from the Population-based
30	Cancer Registry of the Bucaramanga Metropolitan Area (2000-2015). Local and focused
31	cluster tests were used for the detection of spatial clusters and the Poisson multivariable
32	model was used to evaluate the combined effects of spatial variables.
33	Results: The Kulldorff's focused test found a significant spatial cluster (p=0.001) around one
34	industrial agglomerate and the multivariable model results suggests that the distance effect
35	is modified by the directional effect of the wind.
36	Conclusion: A spatial cluster of incident cases of childhood cancer occurred in the
37	municipality of Bucaramanga. Our finding supports the hypothesis that childhood cancer
38	might be related with industrial air pollution exposure in a Latin American city.
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40	Keywords: Childhood cancer, spatial analysis, air pollution
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44 Introduction

45	Childhood cancer (CC), defined here as cancer in children aged 0-14 years, is
46	considered a group of rare diseases. Nevertheless, about one new case of CC is diagnosed
47	in the world every two minutes [1] and it is the leading cause of non-accidental death for
48	this age group [2]. Worldwide, approximately 215,000 cancers are diagnosed annually in
49	children younger than 15 years [2], and in Latin America and the Caribbean 17,500 children
50	are newly diagnosed every year. It is estimated that 84% of all cases of CC in the world occur
51	in low-income and middle-income countries, where 90% of the child population live [3].
52	Knowledge regarding the etiology of CC is limited. Some authors argue that
53	environmental factors can be the cause of 85-96% of all cancers in childhood [4] [5] [6] [7].
54	In recent years, studies conducted in England, Spain and the United States found an
55	association between CC, especially leukemia and CNS tumors, and air pollution exposure [8]
56	[9] [10]. Air pollution is considered the world's largest single environmental health risk as
57	the World Health Organization (WHO, 2016) affirmed that 92% of the world population lives
58	in places where air quality levels are poor [11].
59	During 2013, the International Agency for Research on Cancer (IARC) found that
60	there is sufficient evidence about the carcinogenicity of outdoor air pollution in general and

there is sufficient evidence about the carcinogenicity of outdoor air pollution in general and fine particulate matter (PM_{2,5}) [12]. Research published in Europe and North America, have found association between CC and proximity to industrial sources that emit pollutants into the air [5][9][10][13][14]; also, since the beginning of the 20th century, geographic clusters of cancer have been identified, mainly for leukemia [15]. The mixture of pollutants varies widely in space and time, reflecting the heterogeneity of their sources, the influence of climate, atmospheric transformations, and other factors, and therefore health effects might
also differ widely in space and time and cannot be generalized.

68 In Colombia, a developing country located at the north extreme of South America, 69 there are no exact estimates of the incidence of CC since 90% of the regions of the country 70 do not have cancer registries, which limits making truthful national statistics [16]. 71 Population-based registries with validated high quality data by the IARC are located only in 72 the metropolitan areas of Cali, Manizales, Pasto and Bucaramanga [17]. In Colombia, 73 outdoor air pollution is considered one of the three most hazardous environmental 74 exposures [18] and annual mean concentrations of particulate matter in capital cities 75 exceeds the ones reported for most European and US cities [19]. Previous studies in Colombia have identified adverse effects of ambient air pollutants on respiratory and 76 cardiovascular mortality and morbidity [20,21]. 77

78 Industrial sources are important contributors of outdoor air pollution and location 79 of industrial facilities is often close to residential areas [22]. To our knowledge there are no 80 studies in Colombia or other Latin American country that evaluate the geographic location 81 of incident cases of CC with proximity to industrial sources of air pollution, despite the 82 contribution of industrial emissions to outdoor air pollution in developing countries. 83 Therefore, the aim of this study was to assess the spatial relationship between CC and air 84 pollution from industrial sources in the metropolitan area of Bucaramanga, Colombia, during 2000-2015. 85

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88 Methods

89 Study area and population

90 Colombia is the fourth largest country in South America located at the top corner of 91 the continent. The study area was the Bucaramanga Metropolitan Area (BMA) that it is 92 located in the Northeast Andean region of Colombia with an extension of approximately 93 1,479 km² covering four municipalities: Bucaramanga, Floridablanca, Giron, and 94 Piedecuesta. The study population were the children under 15 years old living in BMA; 95 according to the National Administrative Department of Statistics (DANE for its acronym in 96 Spanish) for the year 2015, the total population in BMA was 1,122,961 people and the population under 15 years was 258,097 children. 97

98 Data Sources

99 The study used secondary retrospective data available in different regional and 100 national institutions: The Population-based Cancer Registry of the Bucaramanga Metropolitan Area (RPC-AMB for its acronym in Spanish) was the source for cancer cases 101 102 data, The DANE was the source for estimations of population and geographical data at small 103 geographical level (census sector), the Corporation for the Defense of the Bucaramanga 104 Plateau (CDMB for its acronym in Spanish) was the source for the identification and location 105 of industrial facilities emitting air pollutants to the air, and planning office of each 106 municipality was the source for the socioeconomic information at small area level.

107

108

109 Childhood cancer data

The CC data were obtained from the RPC-AMB, which is one of the four Populationbased Cancer Registry of regions of Colombia considered by the IARC as high quality. Incident confirmed cases of CC diagnosed from January 1, 2000 to December 31, 2015 were included in the study. The information obtained for CC cases were: the type of cancer, sex, date of birth, date of diagnosis, and place of residence at the time of diagnosis. In the RPC-AMB and for the present research, it is considered a resident who is verified as having been residing 6 months or more in the BMA before the first time they were diagnosed with CC.

117 **Population data and geographical data**

118 The population data for the children's population in the BMA were taken from the 119 official DANE population estimates based on the 2005 national census. The projected infant 120 population under 15 years was used as the population at risk to calculate CC rates during 121 the study period. The DANE population estimates information is available with free access 122 with a minimum level of geographic disaggregation up to block level that is added in an 123 ascending way at the level of sections, sectors, communes, urban/rural areas and 124 municipalities. For the analysis, the census sectors were used as geographical grouping unit 125 (average population of each census sector is 1857 children under 15 years old); these 126 geographical units have a unique identification number assigned by DANE and that was kept 127 as the identification number for each unit of analysis.

128 The geographic coordinates (latitude and longitude) of each centroid of the 129 geographic units (census sector) were estimated using the estimation method with 130 weighting by population location and therefore the centroids are not geographical centers

of polygons necessarily. For the analysis of proximity to sources of industrial pollution, the distance and direction between each "putative source" and the centroids of the census sectors were calculated using distance (in meters) and angle (in geodetic degrees) calculation tool of ArcGIS [®] [23]. The map of the BMA with resolution at the block and sector census level was obtained from the Geoportal cartographic data tool of DANE [24] and the spatial data were created in ArcGIS[®] using the projection of Colombia in mode Custom Azimuth Equidistant and Datum WGS 1984.

138 **Data sources of industrial sources and socioeconomic status**

The industrial facilities emitting air pollutants to the air were identified with the fixed sources inventory existing in the local environmental authority, the CDMB. There were 32 industrial facilities that were located spatially and displayed four identifiable industrial areas around and within BMA. The central point (centroid) of each identified industrial area was used as the location of the "putative source of industrial pollution" and was calculated as the geographic center of the area, georeferenced in latitude and longitude coordinates.

The socioeconomic strata (SES) is a DANE classification of the socioeconomic resources of a residential place [25] and was used as proxy of socioeconomic status. The predominant SES of each census sector was assigned according to the SES of the neighborhoods that make up each census sector. This information was obtained from the planning office of each municipality.

150

152 Data analysis

153 **Exploratory analysis**

154	Crude rates (CR) of annual and cumulative CC incidence were calculated by census
155	sector. The DANE population estimates for each geographical unit were used as population
156	denominators of the annual and accumulated rates. Standardized rates were calculated by
157	age and sex and their 95% confidence intervals, using the direct method and as a standard
158	population, the population of Colombia was used by five-year groups according to the 2005
159	DANE Census. To reduce the heterogeneity in estimating the risk of CC, the calculation of a
160	standardized morbidity ratio (SMR) of the incident CC was made using a Bayesian smoothing
161	technique. Crude rates, standardized rates and Bayesian SMR were calculated in Stata 15 $^{ m extsf{ e}}$
162	and then visualized on choropleth maps using ArcGIS 12 [®] .

163 Hypothesis tests

A cancer spatial cluster is defined as a region with a statistically significant excess in 164 165 the number of cancer cases that occurs [26]. We used the Kulldorff circular space-time 166 scanning test to identify spatial and spatio-temporal local clusters of CC. Three different hypothesis focused tests were used to identify potential spatial clusters of CC around the 167 168 four identified industrial pollution areas. Based on the spatio-temporal parameters, retrospective space-time scanning analysis was applied to identify the geographic areas and 169 170 time periods of potential clusters. The Kulldorff 's circular spatial scanning test in its focused version having as reference grid or "putative source" the coordinates of the centroid of each 171 172 industrial area identified in the BMA; this test was implemented using the SaTSCan 9.1 ® 173 software, with a Poisson probability distribution model, scanning for high rates and with a maximum cluster size of 25% the population at risk [27]. The Stone's test was used to evaluate wheather the risk of the disease is altered with the distance to the putative source; this test was implemented using the "DCluster" programming package encoded in the R software [28]. Finally, the Lawson's directional score test was used to evaluate the variation in direction of the clusters using a score statistic for a single parameter that has a Chi2 distribution with one degree of freedom [29]; the score calculation was implemented in a Microsoft Excel [®] spreadsheet.

181 Multivariable models

182 For the estimation of multivariable regression models with spatial data, we used the approach proposed by Lawson [30]. In the basic or null Poisson model, the log of the 183 184 expected cases based on the number of populations by geographic area is used as an offset. Subsequently, the spatial functions (distance, direction and their interactions) are included 185 186 in the multivariable model; to model direction, the functions of sine (longitude) and cosine 187 (latitude) were used. These multivariable generalized linear models (GLM) for CC were constructed separately for each putative source of industrial pollution. Finally, the effects 188 189 of the spatial functions were adjusted by the SES of the census sectors. The Akaike 190 information criterion (AIC) and the deviance of the potential models were used as the 191 parameters for model selection. The residuals of the models were estimated, and the spatial 192 autocorrelation of the models was evaluated using the global Moran's index. These 193 analyzes will be performed using Stata 15 [®].

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196 Missclassification bias and sensitivity analyses

A sub-sample of the CC were contacted by the RPC-AMB in order to assess the quality of the residential address at the time of diagnosis. We inquired about the address at the time of diagnosis and mobility in the two years prior to the time of diagnosis. With this information, the proportion of misclassification in the subsample was estimated to estimate the potential misclassification of the exposure in the study.

202 Sensitivity analysis was carried out for the local and focused cluster Kulldorf's test. 203 In both cases, the methodology was to execute the analysis with six circular search windows 204 with different parameters, with the following specifications: radius of 0.5 km, 1 km, and 3 205 km and 5%, 10% and 25% of population, Poisson distribution, non-overlap parameter and 206 level of significance of ≤0.05. Conducting multiple analyses using different search windows 207 may strengthen findings and provide confidence about the detected CC clusters.

208

209 Ethical considerations

210 Ethical approval for this study was obtained from the ethics committee of the 211 Industrial University of Santander. The information received from the RPC-AMB did not 212 contain identifying information about participants. The data was analyzed aggregated and 213 not at the individual level, the georeferencing corresponds to the count of cases of CC for 214 each census sector and it is not the exact georeferencing of the place of residence of the 215 cases. For the analysis of misclassification where a subsample was recontacted, the 216 recontact was managed with the caregiver of the minor through the attending physician. 217 Informed consent was obtained from all participants of this subsample.

218 **Results**

219 **Descriptive analysis of childhood cancer**

220 From 2000 to 2015, 679 children under 15 years of age with a median age of 6 years (Interquartile range [IQR] 3 - 11 years) were diagnosed with some subtype of CC in BMA and 221 222 43.89% were girls. The median annual CC was 44 cases (IQR 38,5 - 46 cases). The two 223 municipalities with the highest proportion of cases were Bucaramanga (54.05%) and 224 Floridablanca (25.77%). The most commonly diagnosed types of cancer were: 1) leukemia, 225 myeloproliferative diseases and myelodysplastic diseases (38.29%); 2) neoplasms of the central nervous system (CNS) and intracranial and intraspinal neoplasms (16.05%); and 3) 226 227 lymphomas and reticuloendothelial neoplasms (13.4%). The overall CR of incident CC for 228 BMA over the study period was 158.12 cases per million in children aged under 15 years; 229 the overall annual CR increased from a minimum of 103.53 in 2000 to a maximum of 195.62 in 2013. The overall direct sex and age standardized rate (SASR) for BMA was 158.92 cases 230 231 per million (IC 95%: 147.18 – 171.34); according to the 95% confidence interval (95% CI), 232 the overall SASR of the municipality of Bucaramanga was significantly higher than that of 233 Giron and Piedecuesta. Table 1 shows the overall crude and age-standardized incidence 234 rates by sex for AMB and each municipality.

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- 236
- 237

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	Crude	e rate						
Desien	per millio	n children	Age-standardized rate per million children under 15 years old					
Region	under 15	years old						
-	Female	Male	Female	95% CI	Male	95% CI		
АМВ	141.72	173.85	142.35	(126.60 –	174.76	(157.64 -		
				159.43)		193.23)		
Bucaramanga	156.44	203.30	157.01	(133.24 –	203.97	(177.31 -		
				183.79)		233.51)		
Floridablanca	140.89	190.69	142.32	(111.33 –	193.29	(157.41 -		
				179.24)		234.87)		
Giron	110.23	103.31	110.33	(77.25 –	103.29	(71.94 –		
				152.76)		143.66)		
Piedecuesta	120.27	110.30	120.67	(83.06 –	109.80	(75.09 –		
				169.46		155.04)		

239 Table 1. Overall incidence rates for BMA and its municipalities by sex, 2000-2015

240

BMA is divided into 151 census sectors, in 121 (80.13%) there was at least one case of CC during 2000-2015. The overall CR in the census sector varied from a maximum of 866.35 to a minimum of 36.72 cases per million, with a median of 177.04 (IQR: 103.1 - 259.7). The SASR were between 33.78 (95% CI: 4.09 - 125.89) and 795.67 cases per million (95% CI: 20.14 - 4650.31), with a median of 179.18 (IQR: 103.2 - 263.6). A census sector located in the northwest of the municipality of Bucaramanga, had the overall SASR significantly higher than the overall SASR of BMA. The Bayesian SMR in the census sectors ranged from 0.63 to 1.49 (median: 1.02, IQR: 0.87 - 1.14). Figure 1 show the choropleth maps of the SASR of CC
for BMA over the study period.

250 Figures 1. Age and sex standardized rates of childhood cancer in the urban area of the

- 251 municipalities of the BMA during 2000 to 2015.
- 252 Descriptive analysis of industrial sources

253 The BMA inventory included a total of 38 fixed industrial sources, 78.95% were 254 constituted before the year 2000, 7.89% in 2004 and 13.16% between 2007 and 2009. Four main areas of industrial facilities agglomerations were identified, to which the centroid and 255 256 the coordinates were calculated; these four industrial areas and their centroids were used 257 as a source of exposure for subsequent analyzes. The first agglomerate was referenced as industrial area A with latitude 7.138530546 and longitude -73.13197221 and located in the 258 259 census sector 68001100000000306. The second referenced as industrial area B with 260 latitude of 7.095961661 and longitude -73.16810226 and located in the census sector 261 68307100000000001. The third area is the denominated C with latitude 7.081772454 and 262 longitude -73.15087541 and located in the census sector 680011000000001196. The fourth 263 industrial area, the D has latitude 7.06152597 and longitude -73.09061454, located in the census sector 68276100000000009 (Fig 2). 264

265 Hypothesis tests

266 Of the total of cases, 598 (88.07%) had data for georeferencing. The Kulldorff circular 267 scan local test identified a spatio-temporal cluster of statistically significant of CC cases 268 (p<0.001), with a radius of 1.64 km and estimated relative risk of 1.79 in the municipality of 269 Bucaramanga; furthermore, it identified a space-time cluster of CC cases covering 18 census

270	sectors in the municipality of Bucaramanga, the high-risk period was the year 2009 and
271	estimated relative risk of 6.21 (p:0.02). Around the industrial agglomerate A, the focused
272	cluster test of Kulldorff found a significant cluster (p:0.001) with a radius of 2.65 km and
273	estimated relative risk of 1.41 in the municipality of Bucaramanga, being coincident with
274	the cluster identified by the local test. No significant clusters were found around the other
275	three points of industrial area with this test. Stone's test did not detect significant clusters
276	for the points of industrial agglomerates. Lawson's directional score was significant from
277	industrial areas A, B and C. Table 2 summarizes the results of the three focused tests located
278	for the four industrial areas.

279

280 Table 2. Results of focused hypothesis tests for clusters of childhood cancer in proximity

281 to industrial areas in BMA, 2000-2015

Industrial areas	Kulldorf's circular spatial		Stone's test for		Lawson's directional	
	scan te	st	distance o	decline	score	test
			effe	ct		
	Statistic	p-	Statistic	p-	Statistic	p-
		value		value		value
Α	12.27	<0.001	2.44	0.09	6.82	0.01
В	No clusters	found	1.02	0.89	13.17	<0.001
С	0.002	0.86	1.45	0.25	6.93	0.01
D	2.11	0.08	1.19	0.29	0.94	0.33

282

284 Multivariable Analysis

The Moran coefficient was 0.059 (p < 0.001) for the total number of CC events, which means that there was a low spatial autocorrelation among the census sectors. The distribution of the SMR as outcome variable of the multivariable model (mean: 1.09, variance: 1.00) showed to be appropriated for the proposed Poisson model.

289 The results of the multivariable Poisson models support a consistent significant 290 inverse effect of the distance to the industrial area A on the SMR (Table 3). The model that 291 evaluated only the directional effect had a significant result for the positive effect of the 292 longitude, suggesting a dominant effect toward the East from the industrial area A. The 293 model that evaluates the combined effect of distance and direction suggested a significant 294 effect by decreasing the distance and directional towards the East from A. When the 295 interaction terms between the distance and the address parameters were added to the 296 model and adjusted by the SES, the model improved the deviance suggesting the best model 297 quality. In this model the inverse effect of the distance to industrial area A remained 298 significant, and the statistical significance of the interaction terms suggest that the distance 299 effect is modified by the directional effect (that is, the smaller the distance the stronger is 300 the directional effect), the coefficients and signs of the interaction terms suggest that the 301 predominant directionality is towards the SouthEast direction from A (Table 3).

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304

305 Table 3. Poisson models assessing distance and directional effects of childhood cancer

306 rates around the industrial area A, 2000-2015

	Veriable			Industrial a	area A		
Model type	Variable	Coefficient	SE	p-value	Deviance	AIC	BIC
Distance decline	Distance	-0.00003	0.0000	<0.001	224.16	4.11	-517.41
Directional effect	Latitude	-0.137	0.070	0.853	228.51	4.15	-508.05
	Longitude	0.306	0.941	0.001			
Distance Decline	Distance	-0.00003	0.0000	<0.001	209.69	4.04	-521.86
plus Directional	Latitude	-0.124	0.077	0.109			
Effects	Longitude	0.324	0.933	0.001			
Distance Decline	Distance	-0.0001	0.0000	0.01	198.01	3.99	-523.51
plus Directional	Latitude	-0.092	0.109	0.4			
Effects with	Longitude	-0.109	0.157	0.488			
interaction terms	Distance*Latitude	-0.00003	0.0000	0.422			
	Distance*Longitude	0.00008	0.0000	0.001			
Distance Decline	Distance	-0.0001	0.0000	0.004	169.38	4.03	-464.35
plus Directional	Latitude	0.104	0.132	0.430			
Effects with	Longitude	-0.080	0.169	0.635			
interaction terms	Distance*Latitude	-0.0001	0.0000	0.052			
adjusted by SES	Distance*Longitude	0.00007	0.0000	0.01			
	SES	0.048	0.042	0.250			

307 Sine=Longitude

308 Cosine=Latitude

309

310	In the model diagnosis, standardized Pearson, Deviance and Ascombe residuals fulfill the
311	assumption of normal and mean distribution close to zero. Finally, the Moran coefficient
312	for the residuals was 0.74 (p < 0.001), which means that there is a low spatial autocorrelation
313	of the residuals of the model.
314	For the industrial areas B, C and D the multivariable models adjusted by ESE that
315	incorporated the terms of interaction between distance and directionality were also those
316	that had the best criteria to be the best model and in no case the models had significant
317	results for the effect of distance or directional effect.
318	
319	Sensitivity analysis
320	Misclassification analysis
321	A total of 40 CC cases were contacted. The median age was 6 years (IQR 2 - 9) and
322	86.84% were residents in neighborhoods of SES 1 to 3. Of the total, four (10%) had changed
323	their homes in the year prior to diagnosis. Of the four cases with mobility in the last year
324	
	prior to diagnosis, two had changed their housing within the same municipality, one
325	prior to diagnosis, two had changed their housing within the same municipality, one between the municipalities of the BMA and one to a municipality outside the BMA. A total
325 326	
	between the municipalities of the BMA and one to a municipality outside the BMA. A total
326	between the municipalities of the BMA and one to a municipality outside the BMA. A total of 22 (55%) children resided in the same home since birth.

330 windows executed in the 1km circular window; 12 census sectors were identified that

331	extend to 26 in the 3km window and that were equivalent to the analysis with a population
332	threshold at risk of 5% and 25%, respectively. Figure 2 shows in red census sectors that were
333	detected repeatedly within 5 clusters, in yellow those detected within 3 clusters and in
334	green those detected within 2 clusters.
335	
336	Figure 2. Census sectors detected within the results of the six analyzes of focused
337	childhood cancer clusters through different search windows, Bucaramanga, Colombia,
338	2000-2015
339	
340	In the focused Kulldorf's tests, taking as reference the industrial area A, in the six
341	search windows a spatial cluster was identified, with two census sectors that were always
342	present in the six analyzes executed, 68001100000000306 (Sector that corresponds to the
343	area industrial A) and 680011000000000305. In the 0.5km circular window, two census
344	sectors were identified that extend to 41 in the 3km window and that were equivalent to
345	the analysis with a population threshold at risk of 5% and 25%, respectively.
346	

347 **Discussion**

The incidence rates of CC in the BMA between 2000 and 2015 are higher than the estimated world rates for the same period of time. The cumulative SASR 2000-2015 in the BMA was 1.5 times higher than the world rate reported by the World Cancer Report in 2014 that was 100 CC cases per million [31]. However, the cumulative SASR in the BMA was similar to the cumulative incidence rate of CC reported by the Population-based Cancer

Registry of Cali during 2002-2011 [32]. The cancer registry of Cali is the oldest in Latin America and has international recognition for its high quality methods and data [33][34], therefore, the similarities in the incidence of CC and the frequency of presentation of the different types represent not only the quality and maturity of the RPC-AMB but also the consistency of having higher CC rates in different cities in Colombia.

358 The findings of this study suggest the presence of a CC cluster in proximity to the 359 industrial area denominated A. These results were consistent across the descriptive, 360 hypothesis testing and multivariable analysis findings, and remained statistically significant 361 in the same area when different scan windows were used in the sensitivity analysis. The 362 Lawson's directional score and in the Poisson model that evaluated distance and 363 directionality indicate that there is a spatial presentation of the cases that is mostly 364 concentrated at southeast of industrial area A, finding that might be explained by the 365 potential dispersion of air pollutants as the direction of the cluster is pointing towards the 366 predominant direction of wind in the area [29].

The Stone's test was the test that had no statistically significant findings for the industrial area A; this negative result might be explained by the fact that the Stone's test is executed under the assumption that there is a decreasing trend in risk symmetrically around the radius as the distance to the source increases, and as shown in the multivariable analysis and sensitivity analysis (Fig 2) the risk is not symmetric but with direction towards southeast from A [35]. The statistically significant results in the industrial areas B and C for the directional score of the Lawson test seem to be isolated findings as they were not consistent

with the multivariable analysis, where the interaction between distance and direction wasevaluated as suggested by Lawson [36].

The spatial association found in this research should not be taken as a causal relationship between the industrial emissions of the industrial area A and the incident cases of the CC cluster in this area of the municipality of Bucaramanga. In this case, the relationship between exposure and outcome was modeled with respect to distance to the source and directionality of the wind, but the exposure to the pollutants emitted were measured. Likewise, the retrospective evaluation of the exhibition is difficult since it usually involves different changes during the study period [37].

383 When a cluster of a disease is detected the Centers for Disease Control and 384 Prevention (CDC) proposes a four-stage approach, where the final stage includes proposing a formal etiological investigation [38] and it is the proposal that is suggested as the next 385 386 step to this study. The increase of cases of IQ in this area can be due to multiple factors such 387 as the proximity to exit routes of the city with high flow of heavy duty vehicles of diesel 388 engines, which the exhaust of this type of engine is classified within the group 1 of 389 carcinogens by IARC, which has been found associated with CC [39]. As also for the 390 permanent transit of this type of automotive in this area, due to the loading and unloading 391 of merchandise in the industrial sources that are in this industrial agglomerate. Likewise, in 392 this area of industrial agglomerate the exhibition may not be related to air emissions but to 393 the occupational exposure of parents who live near the workplace; there are several authors who have found a relationship between childhood cancer and maternal or paternal 394 395 occupation [40] [41] [42] [43] [44].

Consistency with previous studies

Different studies in Europe have found an association between exposure to air pollution and CC. Steffen et al. carried out a multicenter case control study in France and found an association between dwellings neighbouring a petrol station or a repair garage during childhood and the risk of childhood leukemia (OR 4.0, 95% Cl 1.5 to 10.3) [45]. During 2009, a study with the same aim was also conducted in France and the findings supported the results and the associations established by Steffen [13].

403 In Spain several studies have been conducted assessing the association between 404 spatial location of industrial pollution sources and CC. During 2015, Ramis et al. published a 405 case-control study with spatial analysis, which included cases of leukemia, cancer of the 406 CNS and non-Hodgkin's lymphoma from five regions of Spain diagnosed from 1996 to 2011, 407 and reported absence of any statistically significant cluster [46]. Ortega-García and cols., 408 analyzed the spatial distribution of all incident cases of CC diagnosed in Spain during 1998 409 to 2015, using focused cluster tests and found a possible association between proximity to 410 certain industries and CC [10]. Previous population case-control studies conducted in Spain 411 by same research team reported an increased risk of Leukemia in children (OR: 1.31, 95% 412 CI: 1.03-1.67) [5], and renal tumors (OR: 1.97, 95% CI: 1.13-3, 42) [47] associated with living at a distance less than 2.5 km from industrial facilities; increased risk of neuroblastomas 413 414 associated with industrial facilities at 1 km (OR: 2.52; 95% CI: 1.20-5.30) and 2 km (OR: 1.99; 415 95% CI: 1.17-3.37) [48]; and increased risk of bone tumors in children under 15 years of age 416 associated with living less than 3 km from industrial sources (OR: 2.33; 95% CI: 1.17-4.63)

417 [49]. They found no excess risk for retinoblastoma, liver tumors, soft tissue sarcomas and418 germ cell tumors and living close to industrial areas [50].

419 Strengths and limitations

The main strength of this study is the quality of the CC data and the period of 16 420 421 years included for the analysis. The cancer registries are considered the best source of 422 information to measure cancer burden in a given time and region [15] [51], the RPC-AMB is 423 considered high quality by IARC due to the integrity, punctuality and accuracy of the data 424 provided, which translates into highly standardized and reliable data. Therefore, the 425 classification of the disease is considered adequate for the cases included in the analyzes 426 performed. In addition, it can be considered that the results of this research are not affected by the mobility of the population, since the misclassification analysis conducted did not 427 428 found that the CC had a high proportion of migration and 55% of the cases in BAM lived in 429 the same residence where they received the diagnosis since birth.

Another important strength to note is the use of multiple analytical approaches and tests used to evaluate the existence and characteristics of a potential CC cluster. The combination of descriptive, hypothesis testing, and multivariable analysis improved the quality of the analysis and gave consistency to the results.

The limitation of the pre-established shapes that the Kulldorff's tests used in this study have been discussed in the literature [52][53][54][55]. These circular clusters in many cases generate that areas of the detected spatial cluster correspond to a false positive, since those areas with low risks are not a genuine part of the cluster [53][55]. To overcome this

limitation this study includes other hypothesis test and multivariable analysis that include
direction testing and their interaction with distance and results of the existence of a cluster
identified by scanning tests remained consistent with additional information related to the
direction of the cluster.

442 The correct classification of the exposure is a limitation and it is considered that to 443 a certain extent all the studies evaluating the effects of exposure to air pollution inevitably 444 have a degree of misclassification of exposure, since it is not feasible to conduct population-445 based studies in which each person uses a personal measuring device for long time [56]. In 446 this study, the magnitude of the exposure was probably determined not only by the 447 distance to the industrial areas identified in the BMA, but also for quantity or air pollutant 448 emissions and the predominant wind direction in the industrial areas. Although the 449 predominant wind direction was taken into account for directional tests, the variability in 450 time of the wind direction was not included. Other important limitation of this study is that 451 exposure was based on proximity to industrial facilities only and did not include any 452 characterization or measurement of the pollutants emitted to the air by the industrial 453 sources. Therefore, proximity to industrial facilities was used in this study as proxy of 454 industrial air pollution exposure.

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Conclusions

459	The results of this study suggest the presence of a spatial cluster of incident cases of CC
460	during 2000-2015 in proximity to an industrial area in the municipality of Bucaramanga.
461	Given the ecological design of this research this finding cannot be assumed as a causal
462	relationship, but results of this study generate new research hypotheses to be evaluated in
463	further research in order to contribute to have a better understanding of the environmental
464	exposures related to the high incidence of CC in Bucaramanga and Colombia.
465	
466	Declaration of interest: none
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468	Funding source: The Administrative Department of Science, Technology and Innovation of
469	Colombia (Colciencias) Contract No. 759-2017.
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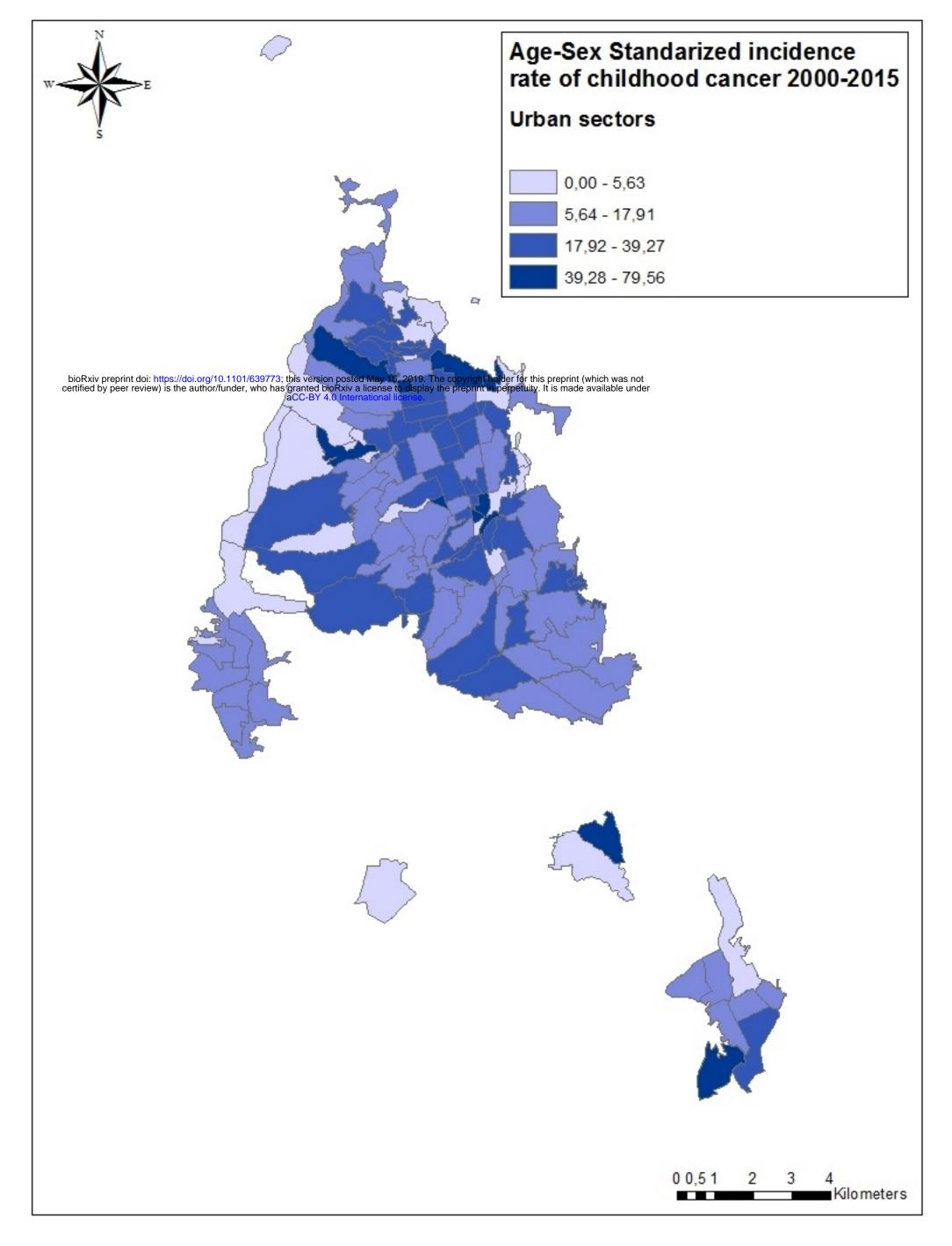
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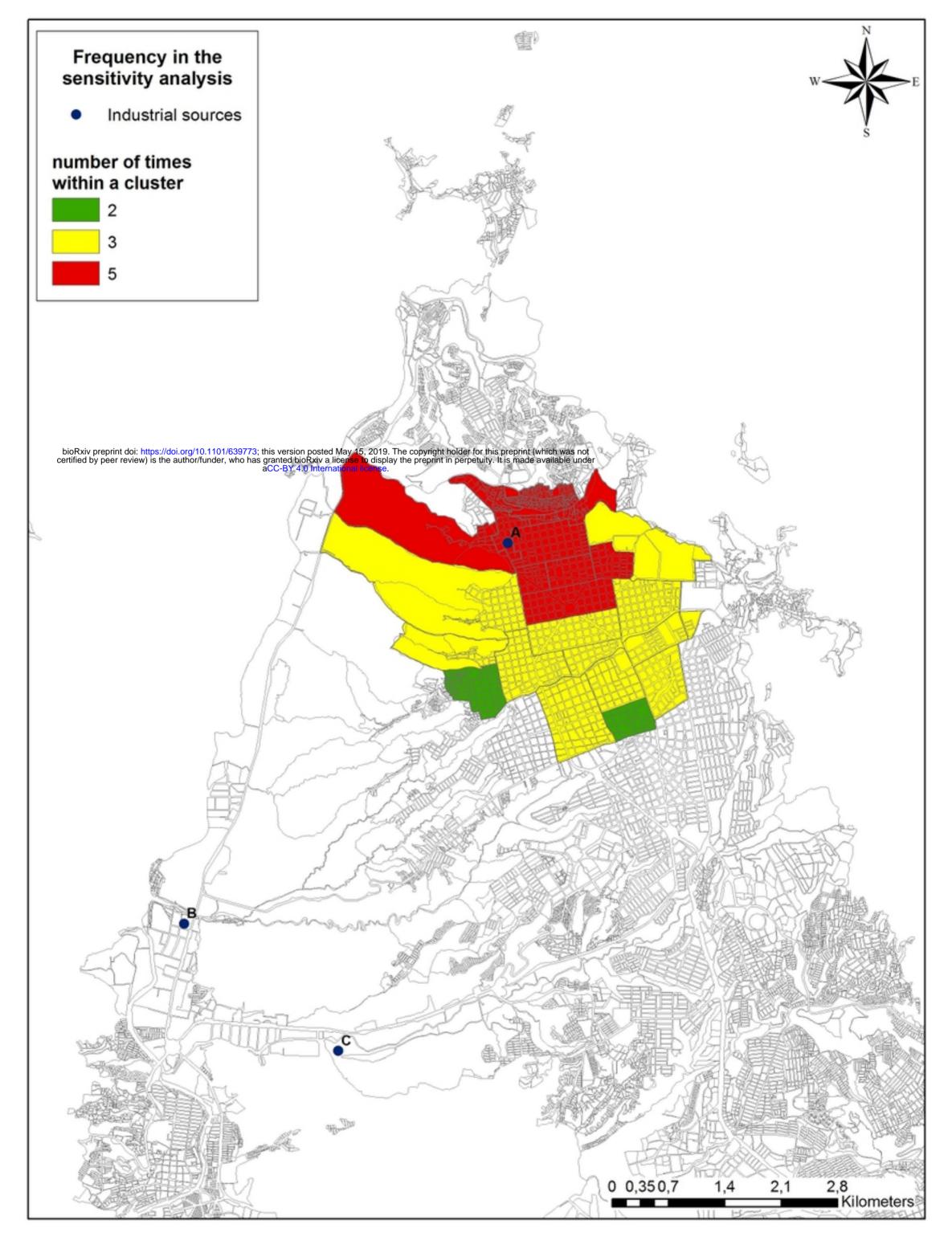
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Figure



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