Atmospheric Measurement Techniques



Impact of Biomass Burning emission on total peroxy nitrates: fire plume identification during the BORTAS campaign

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20 Abstract

- 21 The total peroxy nitrates (Σ PNs) concentrations have been measured using a thermal dissociation
- 22 laser induced fluorescence (TD-LIF) instrument during the BORTAS campaign, which focused on
- the impact of boreal biomass burning emissions on air quality in the Northern hemisphere. The strong
- 24 correlation observed between the Σ PNs concentrations and those of the carbon monoxide (CO), a
- 25 well-known pyrogenic tracer, suggests the possible use of the Σ PNs concentrations as marker of the
- 26 biomass burning (BB) plumes. We applied both statistical and percentile methods to the ΣPNs
- 27 concentrations, comparing the percentage of BB plume selected using these methods with the
- 28 percentage evaluated applying the approaches usually used in literature. Moreover, adding the
- pressure threshold (~750 hPa) to Σ PNs, HCN and CO, as ancillary parameter, the BB plume
- 30 identification is improved. An artificial recurrent neural network (ANN) model was adapted to
- simulate the concentrations of Σ PNs and the HCN including as input parameters Σ PNs, HCN, CO
- 32 and atmospheric pressure, to verify the specific role of these input data to better identify BB plumes.





1 1. Introduction

Biomass burning (BB) events are an important source of many trace gases and particles in the 2 atmosphere [Crutzen et al., 1979; Crutzen and Andreae, 1990; Goode et al., 2000; Andreae and 3 4 Merlet, 2001,]. The signature of BB emission in a plume detected far away from the fire is usually the increase of the atmospheric concentration of pyrogenic species including carbon dioxide (CO₂), 5 carbon monoxide (CO), methane (CH₄) and a series of volatile organic compounds (VOCs), 6 accompanied by elevated concentrations of peroxyacyl nitrate (PAN) [e.g., Goode et al., 2000; Cofer 7 8 et al., 1998; Bertschi et al., 2004; Dibb et al., 2003; Lewis et al., 2007, Tereszchuk et al 2011]. In recent years, several studies have focused on boreal forest fires to quantify the influence of boreal 9 fire emissions on the Earth-Atmosphere system and subsequent impact on the climate. Boreal forest 10 fires (http://www.borealforest.org) affect mainly Siberia, Canada and Alaska and occur generally 11 12 from May to October (Lavoué et al., 2000). In the past three decades occurrence of boreal forest fires and areas burned over Canada have both increased (Gillett et al., 2004; Rinsland et al., 2007; Marlon 13 et al., 2008). The impact of boreal BB on atmospheric chemistry is triggered by the large emissions 14 of CO, NO_x (NO + NO₂), nonmethane hydrocarbons (NMHC) and other NO_y species (Alvarado et 15 16 al., 2010; Parrington et al., 2012): $NO_{y} = NO_{x} + \sum RONO_{2} + \sum RO_{2}NO_{2} + HNO_{3} + HONO + 2N_{2}O_{5}$ 17 + NO₃, where RONO₂ are total alkylnitrates (known also as Σ ANs), RO₂NO₂ are total peroxynitrates (known also as Σ PNs), HNO₃ is nitric acid, HONO is nitrous acid, N₂O₅ is dinitrogen pentoxide and 18 19 NO_3 is nitrate radical. These species can influence the formation of O_3 in the Artic and at the mid latitudes: the role of boreal BB emissions on the O₃ concentration has been studied by several authors 20 showing situations where O_3 concentrations increased and others where it was unaffected (e.g. Wofsy 21 et al., 1992; Jacob et al., 1992; Mauzerall et al., 1996; Wotawa and Trainer, 2000; Val Martin et al., 22 2006; Real et al., 2007; Leung et al., 2007, Jaffe and Wigder, 2012; Parrington et al., 2012). The 23 formation of NO_x oxidation products (Σ ANs, Σ PNs and HNO₃) plays a very important role in 24 25 controlling the ozone budget as they inhibit local O_3 formation (Leung et al., 2007). Many 26 investigations demonstrated the central role played by the peroxy acetyl nitrate (PAN), one of the 27 most common PNs, in BB plume chemistry and, specifically, in the NOx oxidation processes (Jacob 28 et al., 1992; Hudman et al., 2007). Alvarado et al. (2010) demonstrated the rapid PAN formation occurring within 1-2 hours after the emissions by a BB plume: the 40% of the NOx initially emitted 29 30 by the fires is rapidly converted into PAN and the 20% into NO₃ (particles phase). In aged plumes, 31 PAN can represent up to the 67% of the NO_y budget (Alvarado et al., 2010). 32 Analysis of chemistry of BB emissions starts with identifying a BB plume; however, previous studies have used multiple different approaches for the identification and classification of BB plumes. Many 33

studies have recognized CO as a pyrogenic species (Crutzen et al., 1979; Andreae and Merlet, 2001;





1 Lewis et al., 2013), but other chemical species, such as HCN and CH₃CN, have also been employed to identify a BB plume. In the assembly of the BORTAS data analysis, different procedures have 2 3 been applied. Palmer et al. (2013) identified a threshold for each of these species that would separate air masses produced during boreal BB from background air masses; they defined a BB plume when 4 the CO, hydrogen cyanide (HCN) and acetonitrile (CH₃CN) levels are higher than 148 ppb, 122 ppt 5 and 150 ppt, respectively. These thresholds correspond to the 99th percentile (~mean $\pm 3\sigma$) of the data 6 for the species measured during the flight B625, in which there was not a significant correlation 7 between the CO and the CH₃CN. In the context of the BORTAS campaign, Lewis et al. (2013) 8 9 classified the air masses in three groups: 1) background if CO < 200 ppb; 2) BB plumes if the CO >200 ppb with the presence of pyrogenic species such as furan or furfural; 3) anthropogenic plumes if 10 CO > 200 ppb with the absence of furan or furfural. Le Breton et al (2013) observed a strong 11 correlation between the HCN, CO and CH₃CN indicating the utilisation of HCN as a BB marker. 12 They identified a BB plume using a standard deviation (σ) method applied to the 1 Hz HCN 13 measurements: when the HCN concentrations were 6σ above the background for the flights in 14 analysis, they flagged the air mass as a BB plume. The 6σ threshold was chosen since it produces the 15 highest R² values for the correlation between the HCN and the CO. For other experiments different 16 methods for the BB plumes identification have been suggested: 1) Holzinger et al. (2005) used a 17 method similar to that used by Le Breton et al. (2013); their results highlighted a good correlation 18 between acetonitrile (CH₃CN) and CO and they identified BB plumes by significant peaks in the 19 CH₃CN volume mixing ratio (using a threshold of 3σ above the background level); 2) Vay et al. 20 21 (2011) identified a plume when CO > 160 ppb (the median observed CO concentration at the surface), $CH_3CN > 225$ ppt (threshold characterized by an evident enhancement in CH_3CN mixing ratios) or 22 HCN > 500 ppt (when the CH₃CN was not available); 3) Hornbrook et al. (2011) defined a plume by 23 elevated fire tracers above the local background using thresholds of $CH_3CN > 200$ ppt, HCN > 40024 25 ppt and CO > 175 ppb; 4) Tereszchuk et al. (2011) selected as BB plumes air masses when HCN > 350 ppt, assuming that background concentrations of HCN in the free troposphere range between 225 26 27 ppt and 250 ppt and, finally, 5) Alvarado et al. (2010) observed the enhancement of CO correlated with an enhancement in HCN and CH₃CN; they identified a plume as having a CO concentration at 28 least 20 ppb above the background level. 29 An aircraft campaign was conducted in Nova Scotia (Canada) with the principle purpose of evaluating 30 the BB emissions impact on tropospheric photochemistry in the Northern Hemisphere. This work is 31

32 part of the BORTAS project (quantifying the impact of BOReal forest fires on Tropospheric oxidants

- 33 over the Atlantic using Aircraft and Satellites). This project included a field campaign using a
- research aircraft (the FAAM BAe-146) conducted in July and August 2011 during the boreal forest





- 1 fire season in Canada. Further details about this project can be found in Palmer et al. (2013) and at
- 2 http://www.geos.ed.ac.uk/research/eochem/bortas/.
- 3 In our analysis of the BORTAS aircraft measurement data, we found a good correlation between the
- 4 ∑PNs and CO, suggesting the potential of ∑PNs as pyrogenic tracer to discriminate the BB plumes.
 5 Therefore, in this work we propose two different methods to use the ∑PNs as BB tracer: a statistical
- 6 approach using the 6σ threshold and the 99th percentile of the Σ PNs calculated for the B625 flight.
- 7 We evaluated all of the methods described above to our dataset to compare the results of different
- 8 methods to identify BB plumes. We show also that, in some cases, the introduction of meteorological
- 9 parameters, which take into account of the air masses vertical position in the atmosphere,
- 10 discriminates better the origin of the air masses and, then, helps to select more precisely a BB plume.
- 11 Finally, in order to refine the method we adapted an artificial neural network model and we used it to
- simulate the Σ PNs and the HCN in two different procedures to BB plume identification: 1) using as
- 13 inputs the CO, NO, CH₃CN; 2) adding to the inputs also the pressure.
- 14

15 2. Experimental

16 **2.1 TD-LIF measurements:** NO₂, Σ PNs, Σ ANs

17 A detailed description of the BORTAS experiment can be found in Palmer et al. (2013), along with a description of the FAAM BAe-146 instrumental payload. During BORTAS campaign (Canada, 18 19 summer 2011), observations of NO₂, total Σ PNs, Σ ANs, on board the British FAAM BAe 146 research aircraft, were carried out using the TD-LIF (Thermal Dissociation - Laser Induced 20 Fluorescence) instrument developed at the University of L'Aquila (Italy) (Dari-Salisburgo et al., 21 2009; Di Carlo et al., 2013). Briefly, this technique uses direct measurements of NO₂ molecules by 22 23 Laser Induced Fluorescence (LIF), whereas Σ PNs, Σ ANs and HNO₃ are thermally dissociated into NO₂ heating the air sample at 200°C, 400 °C and 550°C, respectively (Day et al., 2002; Di Carlo et 24 25 al., 2013).

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27 2.2 Ancillary measurements: O₃, CO, VOCs

Table 1 reports a number of the compounds measured on-board the BAe-146 aircraft during the BORTAS campaign, specifically those used in this analysis. The instrument used to measure O₃ was a commercial UV absorption system Model 49C (Thermo Environmental Corp.) (Wilson and Birks, 2006). CO was measured using the vacuum ultraviolet (VUV) resonance fluorescence technique. This type of CO instrument was applied to aircraft measurement by Gerbig et al. (1999). NO was measured using a single-channel chemiluminescence instrument manufactured by Air Quality Design, Inc.





- 1 (Colorado, USA, http://www.airqualitydesign.com/) (Lee et al. 2009; Reidmiller et al. 2010).
- 2 Concentrations of volatile organic compounds (VOCs) ranging from C5 to C12 were measured by an
- 3 automatic gas chromatograph system equipped with a mass spectrometer GC-MS (Purvis et al. 2013).
- 4 C2-C7 hydrocarbons and C2-C5 oxygenated volatile organic compounds (o-VOCs) including
- 5 alcohols, aldehydes, ketones and ethers were obtained by the University of York (UK) using whole
- 6 air sampling (WAS) coupled to an automatic gas chromatograph system equipped with a mass
- 7 spectrometer and a flame ionization detector (GC-MS/FID) (Hopkins et al. 2003). Measurements of
- 8 a suite of volatile organic compounds: CH₃CN, C₃H₆O, C₅H₈, MVK+MACR, C₄H₈O, C₆H₆, C₇H₈,
- 9 $C_{10}H_{16}$ were made by proton transfer reaction mass spectrometry (PTR MS) (Murphy et al. 2010).
- 10 Finally, a chemical ionisation mass spectrometer (CIMS) was used for measuring hydrogen cyanide
- 11 (HCN) from biomass burning events (Le Breton et al., 2013).
- 12

13 **Table 1**. List of FAAM BAe-146 instrumental payload and observed compounds used in the analysis

14 in this paper. A description of the FAAM BAe-146 instrumental payload, with accuracy and detection

Species	Method	Reference
СО	VUV resonance/fluorescence	Gerbig et al. (1999)
O ₃	UV absorption	Wilson and Birks (2006)
NO_2 , $\Sigma RO_2 NO_2$, $\Sigma RONO2$,	TD-LIF	Dari-Salisburgo et al. (2009); Di Carlo et al. (2013)
NOy		Di Callo et al. (2013)
NO	AQD chemiluminescence	Lee et al. 2009; Reidmiller et al. 2010
C ₅ -C ₁₂ VOCs	GC-MS	Purvis et al. (2013)
C ₂ –C ₇ NMHCs, acetone	WAS-GC	Hopkins et al. (2003)
CH ₃ OH		
HCN	CIMS	Le Breton et al. (2013)

15 limit, is reported in Palmer et al. (2013).





CH ₃ CN, C ₃ H ₆ O, C ₅ H ₈ ,	PTR-MS	Murphy et al. (2010)
MVK+MACR, C ₄ H ₈ O,		
$C_6H_6, C_7H_8, C_{10}H_{16}$		

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2 **3. Plume identification**

Figure 1 shows the time series of the species investigated in this paper (Σ PNs, acetonitrile, HCN, CO 3 and furfural) for one of the flights analysed (B623), illustrated as an example. The **SPNs** trends shows 4 exactly the same structures present on CO and CH₃CN, with the exception of two plumes (in the 5 temporal intervals of $(8.2 - 8.4) \times 10^4$ and $(8.6 - 8.8) \times 10^4$ seconds After Midnight (A.M.)) that are 6 extensively analysed in the next sections. Moreover, the furfural shows similar trend for the first part 7 of the flight; since CO, acetonitrile, HCN and Furfural are a known BB tracers (Lewis et al. 2013; 8 Palmer et al. 2013; Le Breton et al., 2013), the significant correlation between the Σ PNs and those 9 species suggests that the Σ PNs also originated at the same boreal biomass burning source and that it 10 can be used as additional chemical species for the identification of the BB plumes. 11 12



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Figure 1. Time series of Σ PNs, CH₃CN, Furfural, CO and pressure during the flight B623, shown as example.





- 1 As described in the introduction, recently different compounds (CO, HCN or CH₃CN) and different
- 2 concentrations thresholds have been introduced as tracers to establish if the air mass monitored is
- 3 affected by forest fire emissions or not. These methods are summarized in Table 2.
- 4 **Table 2**. Methods to identify BB plumes.

	Methods
Holzinger et al. (2005)	$CH_3CN > 3\sigma$
Alvarado et al. (2010)	CO increase of at least 20 ppb respect to the background
	and its correlation with HCN or CH ₃ CN
Vay et al. (2011)	$CO > 160 \text{ ppb}, \text{HCN} > 500 \text{ ppt or } CH_3CN > 225 \text{ ppt}$
Hornbrook et al. (2011)	$CO > 175 \text{ ppb}, \text{HCN} > 400 \text{ ppt}, \text{CH}_3\text{CN} > 200 \text{ ppt}$
Tereszchuk et al. (2011)	HCN > 350 ppt
Lewis et al. (2013)	CO > 200 ppb and presence of furan or furfural
Palmer et al. (2013)	$CO > 148 \text{ ppb}$, $HCN > 122 \text{ ppt}$ and $CH_3CN > 150 \text{ ppt}$
Le Breton et al. (2013)	$HCN > 6\sigma$
This work	\sum PNs > 6 σ or \sum PNs > 418 ppt

5

In our analysis we tested the use of the concentration profiles of Σ PNs as possible identifier of BB 6 plumes. We applied two different methods: 1) evaluating the 99th percentile of the Σ PNs 7 concentrations measured during the background flight B625 (as done by Palmer et al., 2013) and 8 using this value as threshold to distinguish air masses emitted by fires; 2) applying the statistical 9 approach (σ method). In the first case, we calculate the 99th percentile of the 1 second Σ PNs data 10 sampled during the B625 finding a threshold of 418 ppt. In the second case, first we selected the parts 11 of each flight identifiable as background and, then, we evaluated for these data the mean and the 12 standard deviation of the Σ PNs; after that, we identified as BB plumes the parts of each flight in 13 which the difference between Σ PNs concentrations and the background level were higher than 6 14 standard deviations of the background. This threshold has been evaluated calculating the correlation 15 coefficients between the Σ PNs and the CO varying the sigma threshold between 10 σ and 3 σ for each 16 flight in which the Σ PNs showed evident and clear plumes (such as B622 and B623): we found that 17 the maximum R occurred when we selected the BB plume using the 6σ threshold. In Table 3 are 18 summarized the correlation coefficients R and the percentage of flight selected as BB plume obtained 19 20 for all the flights in analysis using both the methods just described. Moreover, we indicated the 21 percentage of flight identified as BB plume applying all the methods listed in Table 2 to our dataset.





- 2 Table 3. Correlation coefficient (R) and percentage (%) of each flight of the BORTAS campaign
- selected as BB plume using both the Σ PNs methods (6 σ and 99th percentile thresholds) and percentage
- 4 (%) of BB plume evaluated for each BORTAS flight using the methods applied in other studies.

Flight	This study (SPNs methods)			This study (\sum PNs methods)Other studies	studies		
		6σ	$\sum PNs$	> 0.418			
			99 pe	rcentile			-
	R	%	R	%		%	Comments
B619a	-	0	-0.045	13.5	Palmer et al.	0	Only CO
					Vay et al.	0	Only CO
					Hornbrook et al.	0	Only CO
					Holzinger et al.	-	No CH ₃ CN
					Alvarado et al.	1.3	Only CO
					Lewis et al.	0	No Furfural – CO < 200
					Tereszchuk et al.	-	No HCN
					Le Breton et al.	-	No HCN
B619b	-	0	-0.168	1.3	Palmer et al.	0	Only CO
					Vay et al.	0	Only CO
					Hornbrook et al.	0	Only CO
					Holzinger et al.	-	No CH ₃ CN
					Alvarado et al.	0	Only CO
					Lewis et al.	0	No Furfural – CO < 200
					Tereszchuk et al.	-	No HCN
					Le Breton et al.	-	No HCN
B620	-	0	-	0	Palmer et al.	0	Only CO
					Vay et al.	0	Only CO
					Hornbrook et al.	0	Only CO
					Holzinger et al.	-	No CH ₃ CN
					Alvarado et al.	5.1	Only CO
					Lewis et al.	0	No Furfural - CO < 200
					Tereszchuk et al.	-	No HCN
					Le Breton et al.	-	No HCN





B621a	0.953	59.1	0.951	77.6	Palmer et al.	58.9	
					Vay et al.	23.1	
					Hornbrook et al.	32.5	
					Holzinger et al.	72.3	
					Alvarado et al.	76.8	
					Lewis et al.	-	No Furfural – CO > 200
					Tereszchuk et al.	41.6	
					Le Breton et al.	47.7	
B621b	0.644	3.6	0.644	3.6	Palmer et al.	2.0	NO CH ₃ CN in the
							plume; missing data of
							HCN in the flight (only
							a plume)
					Vay et al.	0	NO CH ₃ CN in the
							plume; missing data of
							HCN in the flight (only
							a plume)
					Hornbrook et al.	0	NO CH ₃ CN in the
							plume; missing data of
							HCN in the flight (only
							a plume)
					Holzinger et al.	0.6	Missing data for the
							CH ₃ CN in the flight
					Alvarado et al.	14.4	
					Lewis et al.	-	No Furfural – CO >200
					Tereszchuk et al.	26.3	
					Le Breton et al.	32.4	Missing data (only a
							plume)
B622	0.867	51.9	0.864	55.8	Palmer et al.	52.2	
					Vay et al.	7.2	
					Hornbrook et al.	10.7	
					Holzinger et al.	50.0	
					Alvarado et al.	74.2	
					Lewis et al.	45.6	





					Tereszchuk et al.	10.4	
					Le Breton et al.	51.1	
B623	0.804	41.2	0.776	42.6	Palmer et al.	76.9	No HCN
					Vay et al.	73.6	No HCN
					Hornbrook et al.	70.5	No HCN
					Holzinger et al.	80.5	
					Alvarado et al.	85.6	
					Lewis et al.	71.0	
					Tereszchuk et al.	-	No HCN
					Le Breton et al.	-	No HCN
B626a	-	0	-	0	Palmer et al.	0	No HCN
					Vay et al.	0	No HCN
					Hornbrook et al.	0	No HCN
					Holzinger et al.	0.6	
					Alvarado et al.	2.0	
					Lewis et al.	0	No Furfural – CO < 200
					Tereszchuk et al.	-	No HCN
					Le Breton et al.	-	No HCN
B626b	0.755	8.1	0.259	0.4	Palmer et al.	8.6	No CH ₃ CN
					Vay et al.	5.6	No CH ₃ CN
					Hornbrook et al.	3.7	No CH ₃ CN
					Holzinger et al.	-	No CH ₃ CN
					Alvarado et al.	8.5	
					Lewis et al.	1.9	
					Tereszchuk et al.	15.1	
					Le Breton et al.	19.3	
B627	-	0	0.073	11.0	Palmer et al.	0	No HCN
					Vay et al.	0	No HCN
					Hornbrook et al.	0	No HCN
					Holzinger et al.	0.3	
					Alvarado et al.	3.1	
					Lewis et al.	0	No Furfural – CO < 200
					Tereszchuk et al.	-	No HCN





					Le Breton et al.	-	No HCN
B628a	-	0	-	0	Palmer et al.	0	
					Vay et al.	0	
					Hornbrook et al.	0	
					Holzinger et al.	0.3	
					Alvarado et al.	21.0	
					Lewis et al.	0	No Furfural – CO < 200
					Tereszchuk et al.	0	
					Le Breton et al.	40.1	
B628b	0.477	11.9	0.540	10.6	Palmer et al.	2.5	No HCN
					Vay et al.	0	No HCN
					Hornbrook et al.	0	No HCN
					Holzinger et al.	6.3	
					Alvarado et al.	61.1	
					Lewis et al.	0	No Furfural – CO < 200
					Tereszchuk et al.	-	No HCN
					Le Breton et al.	-	No HCN
B629	-0.036	6.4	-0.010	5.7	Palmer et al.	3.9	No HCN
					Vay et al.	1.1	No HCN
					Hornbrook et al.	0	No HCN
					Holzinger et al.	7.6	
					Alvarado et al.	6.2	
					Lewis et al.	0	No Furfural – CO < 200
					Tereszchuk et al.	-	No HCN
					Le Breton et al.	-	No HCN
B630	0.007	2.0	0.005	2.6	Palmer et al.	12.0	No HCN
					Vay et al.	5.1	No HCN
					Hornbrook et al.	2.4	No HCN
					Holzinger et al.	26.2	
					Alvarado et al.	42.9	
					Lewis et al.	-	No Furfural $-$ peak with CO > 200
					Tereszchuk et al.	-	No HCN





Le Breton et al. _ No HCN 1 2 In order to make the analysis as accurate as possible, the double flights were separated and applied all the methods in each part of the flights taking into account of different Canadian regions and 3 4 different hours of the day during which we were flying. The Σ PNs methods give, in most cases, results in agreement with each of the methods that use CO, HCN and CH₃CN (see table 3) and allow 5 6 to discriminate 3 flights (B621a, B622 and B623) in which a significant percentage (higher than the 40% with R ranging between 0.804 and 0.953) of the data have been collected sampling air masses 7 emitted by biomass burning. Moreover, these results are also in agreement with what is found 8 applying the approaches suggested in other studies. We found three flights (B621b, B626b and 9 10 B628b) in which the majority of the methods show that a small percentage (less than 10% with R 11 ranging between 0.477 and 0.755) of data can be identified as BB plumes (see table 3). In six flights (B619a, B619b, B620, B626a, B627 and B628a) the ΣPNs 6σ method suggests, in agreement with 12 13 the majority of the other techniques, a dominance of background air masses sampled; the 99th percentile method applied to the Σ PNs for the flights B619a, B619b and B627 show a small fraction 14 15 of flight selected as BB. Moreover, the R values between the Σ PNs and the CO in those points is so low (or even negative) to suggest that these parts of the flights are not ascribable as BB plume. The 16 17 flights B629 and B630 present a small percentage of the flight identified as BB plume, using the Σ PNs methods (confirmed also by other methods), but the low (or negative) R values suggest a 18 19 different origin for these plumes. Summarizing, we classified three flights (highlighted in grey in Table 3) with evidence of significant BB plumes sampled, three with a small percentage of BB plumes 20 and 8 flights with a dominance of background air masses sampled. Le Breton et al. (2013) evaluated 21 the percentage of flights ascribable as BB plumes using the 6σ method applied to their HCN 22 23 measurements and the R² between the CO and the HCN in the BB plumes for five flights. Our results, with the percentage of flight identified as BB plume using their method, are in agreement with those 24 that they presented, with the exception of the B621 flight. In this case, the difference could be due to 25 the fact that we split the double flights into two datasets and also because for the B621b there are 26 some missing data for the HCN and only one big plume. Moreover, the R^2 between HCN and CO 27 28 range between 0.46 (B622) and 0.83 (B621) (Le Breton et al., 2013). Figure 2 shows the scatter plot between the Σ PNs and the CO for the flights B621a, B622 and B623: 29 30 the red circles are the BB plume selected by the 6σ of the Σ PNs technique and the green circles the

31 background part of each flight







1 **Figure 2.** Scatter plot between the \sum PNs and the CO for the B621a, B622 and B623 flights (from the 2 top to the bottom, respectively). The red points indicate the data selected as BB plume using the 6σ

3 method to the \sum PNs measurements, the green points indicate the background air masses.





- 1 The \sum PNs show a very good correlation with the CO, in presence of BB plumes: the correlation
- 2 coefficients R, in fact, vary between 0.804 and 0.953. The data of the B623 flight (Figure 2) show an 3 interesting, well distinct double trends between the Σ PNs and the CO; the time series (Figure 1)
- 4 shows that the CO and the CH₃CN present two plumes (between 8.2×10^4 8.4×10^4 seconds A.M. and
- 5 between 8.6 $\times 10^4$ -8.8 $\times 10^4$ seconds A.M.) at pressure (P) higher than ~750 hPa (corresponding to an
- 6 altitude of ~2000 m a.s.l.): at these plumes do not correspond an increase of the Σ PNs nor of the
- 7 Furfural. In order to explain the double trends in the Σ PNs vs CO scatter plot of the flight B623 (see
- 8 figure 2), we analysed the all dataset of this flight as function of the pressure, finding that the two
- 9 trends can be distinguished using a pressure threshold of 750 hPa (Figure 3), or equivalently an
- 10 altitude threshold of ~2000 m a.s.l. (not shown).



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14 The fact that the two distinct trends of the flight B623 are separable by the 6σ method of the Σ PNs and by pressure, suggests that the air masses sampled at different altitude are originated by different 15 16 sources, and the smaller correlation (R = 0.31) between Σ PNs and CO for pressure above 750 hPa, supports the hypothesis that they are not impacted by BB emission. To prove this thesis, we evaluated 17 18 the lagrangian back-trajectories (Hysplit model, Draxler et al., 1999) selecting the starting point at different altitudes along the flight trajectory and running the model up to 200 hours back. Panel c) in 19 20 Figure 4 shows the black-trajectories of air masses sampled at pressure higher than 750 hPa (black points) or lower than 750 hPa (grey points): a different origin is clearly identifiable. In particular, air 21 masses collected at lower altitudes (P > 750 hPa) come from the northern region of the Canada and 22 the altitude of provenience increases the further back the trajectory goes and decreases in the last 23 24 three days of simulation (Figure 4d); on the contrary the air masses corresponding to P < 750 hPa 14





1 originated in the South-West region of the Canada (e.g., Alberta, Saskatchewan,...) and in the North-West states of the U.S.A. (e.g., Montana, Idaho, Wyoming, Washington, Oregon,...). Figure 4a 2 represents the position of the air masses along the back-trajectories during the days between the July 3 12th and July 20th (corresponding respectively to -200 hours back and 0 hours back in the back-4 trajectories model setup). In the same way, Figure 4b shows the fire spots recorded by the FLAMBE 5 (The Fire Locating And Monitoring of Burning Emissions, Reid et al., 2009) archives as function of 6 the day in which the biomass burning occurred (between the July 11th and July 20th). The Figure 4 7 reveals not only that air masses collected for P > 750 hPa come from a different region of the North 8 9 America respect to those collected for P < 750 hPa, but also that the regions of interest for air masses 10 at P > 750 hPa are, most likely, not impacted by fire emissions, at least for the days in which the air masses passed over those areas. On the contrary, the plumes (identified as BB plumes) measured at 11 P < 750 hPa (i.e. altitude higher than 2000 m a.s.l.) have originated from regions strongly influenced 12 by biomass burning fires (see panel b of figure 4) in the central-west coast of the North America) in 13 the days of July in which the air masses flew above those territories. The back-trajectories analysis, 14 therefore, confirm the existence of two different layers of air masses sampled during the B623 flight 15 16 distinguishable by the pressure (i.e. the altitude).







Figure 4. Panel a): temporal position (express as day from the 12th July to the 20th of July) along the 200 hours back-trajectories. Panel b): spot of the biomass burning fires derived from the FLAMBE archive as function of the day in which they occurred (from 11th July to 20th July – we selected the dataset at the 13.00 UTC and the 23.00 UTC). Panel c): distinction of the back trajectories position selecting the starting point along the flight trajectories for $P \ge 750$ hPa. Panel d): altitude along the back trajectories.

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Finally, we evaluated the qualitative age of the air masses sampled during the B623 flight: using the ratio between the NO_x (sum of the TD-LIF NO₂ and the chemiluminescence NO measurements) and the total NO_y (measured by the TD-LIF and the calculated as sum of NO_x and its oxidation products), we observed that the air masses sampled at P < 750 hPa are younger respect those sampled at higher pressure confirming that the air masses identifiable by the pressure are distinct not only for their sources (and provenience), but also for the age.

14 Considering what has been highlighted analysing the double trends in the CO vs Σ PNs plot of the B623, we examined in depth the correlation between the CO and the Σ PNs for all the flights and we 15 16 found that also other flights show similar trends: in fact, B621a and the B622 show two different trends that can be distinguished using a pressure threshold of 700 hPa (figure not shown). Moreover, 17 we did the same analysis also for the HCN: we found that for the B621a, B622 the scatter plot (plotting 18 19 only the data selected as BB plume using the Le Breton et al. (2013) method) between the HCN and the CO shows different trends that can be distinct by the pressure (Figure 5, panel a and b related to 20 the B621a and B622 flight, respectively). This analysis suggest that, even though the identification 21 of a BB plume done using different chemical species threshold is largely and successfully employed 22 (Le Breton et al., 2013; Lewis et al., 2013; Palmer et al., 2013; O'Shea et al. 2013), there are some 23 episodes in which the introduction of other parameters (such as the pressure or the altitude, which are 24 25 indicative of the vertical position of the air masses and, therefore, may be particularly significant for 26 observations on-board aircrafts) gives interesting information and allow to discriminate better and 27 with more details the plumes. The back-trajectories evaluation also gives an important indication of 28 the origin of the air masses and completes the analysis.







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Figure 5. Scatter plot between the HCN and the CO as function of the pressure for the flights B621a
(at the top) and B622 (at the bottom).

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6 **4. Model results**

7 In order to refine the methods to discriminate the BB plumes and better understand the origin of the air masses analysed in the BORTAS campaign, an artificial neural network (ANN) model, recently 8 developed for O_3 and PM studies (Biancofiore et al., 2015a; Biancofiore et al., 2015b), has been 9 adapted and used with BORTAS dataset. The ANN is capable of simulating non-linear relationships 10 (Lonbladd et al. 1992) and for this reason is ideal for investigating the relationship between $\sum PNs$ 11 (and HCN) and chemical and physical parameters. In this work, a recurrent architecture was used 12 (Elman, 1990) that provides a multi-step memory. We simulated the concentrations of the Σ PNs for 13 the B623 flight and of the Σ PNs and the HCN for the B622. During both these flights the scatter plots 14 between the ∑PNs and the CO or the HCN and the CO show different trends that we distinguished 15 16 by the pressure and that are indicative of air masses having different origins (coming from regions





1 interested by fires or not), as explained in Section 3. We carried out two different simulations: a) we used as input for the neural network only the O₃, CO, NO and CH₃CN (case A); b) we used as input 2 3 the O₃, CO, NO, CH₃CN and the pressure (case B). In this way, we evaluated if and how the simulations of the Σ PNs and the HCN change by adding a physical parameter to the inputs, therefore 4 taking into account the position of the air masses (that is the altitude of the aircraft during the flights). 5 Figure 6 shows the results of the case A simulations for the flight B622: the scatter plot between the 6 simulated HCN (panel a) and Σ PNs (panel b) and the measured CO as function of the pressure does 7 not present distinct trends as evident in the measured data. Similar results have been found for the 8 9 case A simulations (panel c in Figure 6) of the flight B623: the net separation of two trends in the \sum PNs vs CO scatter plot of the measured data (Figure 2) is not reproduced by the model. On the 10 contrary, adding the pressure to the inputs (case B) the simulations improve significantly for both the 11 flights and both the species; Figure 7 shows the ANN simulations in the case B. It is evident that the 12 different trends, identifiable by the pressure using the measured data (Figure 3 and Figure 5), are well 13 reproduced and become evident despite what occurred in the case A (Figure 6). Moreover, it is clear 14 also that the correlation coefficients between the ΣPNs measured and the ΣPNs simulated increase 15 16 adding the pressure to the inputs (these results are highlighted in grey in Table 4): in fact, R improves from 0.94 (case A) to 0.95 (case B) for the B622 flight and, even more, from 0.77 (case A) to 0.94 17 (case B) for the flight B623, in which the two trends between the $\sum PNs$ and the CO is significantly 18 more evident respect to the B622 flight. Similar results can be found for the HCN: the correlation 19 coefficient R, in fact, increases from 0.86 (case A) to 0.92 (case B). In addition to the correlation 20 21 coefficient R, we estimated the model performances using three more typical indices (Biancofiore et al., 2015): the fractional bias (FB), the normalized mean squared error (NMSE) and the factor of 2 22 (FA2). The results are summarized in Table 4. The FB is calculated as the ratio between the difference 23 of the mean observed and the mean modelled concentrations and their mean; it indicates the systemic 24 25 errors that can entail a bias between the modelled and the measured data and can range between -2 26 (overestimation) and +2 (underestimation): an ideal model has an FB index of 0. The NMSE is the 27 result of the ratio between the mean of the squared difference between the observed and the modelled 28 concentrations and the sum of their mean; it gives information about the total errors of the models and the best value is 0 (as the NMSE decreases approaching 0, the model performance increases). 29 Finally, the FA2 is the percentage of the simulated data for which the ratio between the observed and 30 the modelled concentrations is included between 0.5 and 2; the ideal model has an FA2 of 1 and the 31 32 worst results occur if FA2 is equal to 0. The Σ PNs simulation for the flight B622 shows that R, NMSE and the FA2 improve adding the pressure to the inputs (case B); on the contrary, the FB results slightly 33 34 worst indicating that the model tends to overestimate the concentrations. The HCN simulation for the





- 1 flight B622 has better results for all the indices with the exception, also in this case, of the FB which
- 2 is lightly greater in the case B than in the case A. Finally, the R, FB and FA2 of the Σ PNs modelled
- 3 for the flight B623 show a significant improvement in the case B respect the case A, but the NMSE
- 4 presents a worsening suggesting the possibility of an increase in the systemic or random errors.
- 5
- 6 Table 4. Indices to evaluate the Model performances. R is the correlation coefficient, FB is the
- 7 fractional bias, NMSE is the normalized mean squared error and FA2 the factor of 2. See the text for
- 8 a description of these indices

	Simulations	R	FB	NMSE	FA2
$\sum PNs$ (B622)	Case A	0.94	-0.017	0.124	0.755
	Case B	0.95	-0.023	0.099	0.763
HCN (B622)	Case A	0.86	0.009	0.228	0.946
	Case B	0.92	0.014	0.136	0.959
$\sum PNs (B623)$	Case A	0.77	0.014	0.028	0.771
	Case B	0.94	-8.8e-4	0.079	0.867

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Figure 6. Simulations results for the case A (only O₃, CO, NO and CH₃CN as inputs): scatter plot
between the simulated HCN (panel a)), the simulated ∑PNs (panel b)) and the CO for the B622 and
between the simulated ∑PNs and CO for the B622 (panel a)) and the B623 (panel b)).





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Figure 7. Simulations results for the case B (O₃, CO, NO, CH₃CN and pressure as inputs): scatter
plot between the HCN (panel a)) simulated and the CO for the B622 and between ∑PNs and CO for
the B623 (panel b)).





1 Conclusions

The measured \sum PNs during the BORTAS aircraft campaign show a good correlation (R varying 2 between ~0.80 and ~0.95) with CO for different flights. We used Σ PNs as BB tracers applying two 3 different methods to select a BB plumes: 1) we defined a threshold of 6σ of Σ PNs concentration for 4 each flights and all the data that are 6σ times higher than the background level have been selected as 5 BB plume; 2) evaluating the 99th percentile of the Σ PNs measurements done during the B625 flight 6 (considered a background flight not affected by BB plumes). Moreover, we applied several methods 7 present in the literature to our dataset to compare with the proposed Σ PNs methods and we calculated 8 9 the percentage of flight classifiable as BB plume; we found that all the methods identified four flights with evident and significant percentage of BB plumes intercepted. Therefore for most of the flights 10 the Σ PNs methods can be an alternative to identify BB. Moreover, we found that in some flights the 11 scatter plot between the Σ PNs and the CO (or the HCN and the CO) shows different slopes 12 identifiable by a pressure threshold, suggesting different regimes between these species that is 13 14 different air masses. The dependency of these slopes by the pressure suggested that the air masses sampled could be spatially (vertically) different and that their origins could be different. Analysing 15 16 the back-trajectories, as expected, we found that the air masses were not all originated by biomass burning and that those at lower pressure (i.e. higher altitude) reach Nova Scotia from the clean 17 northern region of the Canada. In order to refine the method we adapted an ANN model simulating 18 the Σ PNs and the HCN in two different ways: 1) using only the CO, NO and CH₃CN as inputs; 2) 19 adding also the pressure to the inputs. We found that the model results improve in the second case. 20 21 In conclusion, we suggest that the Σ PNs can be used as BB tracer for the identification of BB plumes; moreover, as consequence of the Σ PNs vs CO analysis and of the model results, we suggest that the 22 use of a meteorological parameter (such as pressure) could allow a better discrimination of the air 23 masses origin and a more selective method for the selection of BB plumes. 24

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