

1 Proceedings

2 Source apportionment of atmospheric deposition species in an 3 agricultural Brazilian region using Positive Matrix Factoriza- 4 tion[†]

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11 † Lavras, Minas Gerais, Brazil. July 2021.

12 **Abstract:** We investigated the influence of natural and anthropogenic sources on bulk atmospheric
13 deposition chemistry, from November 2017 until October 2019, in a Brazilian agricultural area. The
14 pH mean value was 5.99 (5.52 – 8.46) and most deposition samples (~98%) were alkaline (pH > 5.60).
15 We identified Ca²⁺ as the predominant specie accounting for 33% of the total ionic species distribu-
16 tion, and as the mainly precursors of atmospheric acidity neutralization (Neutralization Factor =
17 6.63). PMF analysis resulted in four factors, which demonstrated the influence of anthropogenic and
18 natural sources, such as fertilizer application and production, marine intrusion / biomass burning
19 and biogenic emissions, and revealed the importance of atmospheric neutralization processes.

20 **Keywords:** Atmospheric Deposition; Source apportionment; Agricultural region; PMF; Brazil

21 Pereira, J. N.; Fornaro, A.; Vieira-
22 Filho, M. Source apportionment of
23 atmospheric deposition species in an
24 agricultural Brazilian region using
25 Positive Matrix Factorization. *Pro-*
26 *ceedings* **2021**, *65*, x.
27 <https://doi.org/10.3390/xxxxx>

Received: date

Accepted: date

Published: date

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29 tral with regard to jurisdictional
30 claims in published maps and insti-
31 tutional affiliations.



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1. Introduction

Because of rapid economic development, increased energy consumption, anthro-
pogenic activities and industrialization in the last centuries, was inevitable the atmo-
spheric accumulation of several gases and aerosols [1]. These mentioned pollutants are
removed from the atmosphere to Earth's surface mainly by bulk deposition processes,
which encompasses wet (in-cloud and below-cloud scavenging processes) and dry depo-
sition (scavenging processes in precipitation absence) [2]. Therefore, owing to atmo-
spheric deposition, nutrients and pollutants may entry into terrestrial and aquatic ecosys-
tems causing significant impacts, such as eutrophication and soil acidity [3].

Atmospheric deposition chemistry has been widely studied in many areas world-
wide, since provide useful information that feeds several receptor models to identify an-
thropogenic and natural source apportionment [4–7]. Broadly, natural origin is relation to marine salt and soil dust,
while anthropogenic origins are mainly fossil fuel combustion, industrial processes and agricultural production [8].

1 Moreda-Piñeiro et al. [9] assessed rainwater samples in Spain and reported that Cl^- , Na^+ and Mg^{2+} were linked to sea-
2 salt origin, while SO_4^{2-} and Ca^{2+} were released from a terrestrial crustal source. Also in this study, NH_4^+ and NO_3^- were
3 mainly attributed to agricultural activity.

4 Statistical techniques, such as Factor Analysis (FA), Principal Component Analysis (PCA), Chemical Mass Bal-
5 ance (CMB) and Positive Matrix Factorization (PMF), have been employed as receptor models for identification of the
6 main pollutants sources [10–12]. However, PMF is commonly applied to airborne particles and atmospheric gases
7 [13,14] and scarcely to bulk atmospheric deposition, due to complexity of its process [15].

8 Considering the absence of an established network for routinely monitoring wet and dry deposition in Brazil
9 [16], a study was carried out in the Southern Region of Minas Gerais, which is an important economic region responsible
10 for 21.8% of agricultural commodities (mainly from coffee producing regions), and also accounts for about 12% of the
11 state's Gross Domestic Product [17]. In this sense, we investigated the influence of natural and anthropogenic sources
12 on bulk atmospheric deposition chemistry in the Lavras city, South of Minas Gerais, using PMF.

13 2. Material and Methods

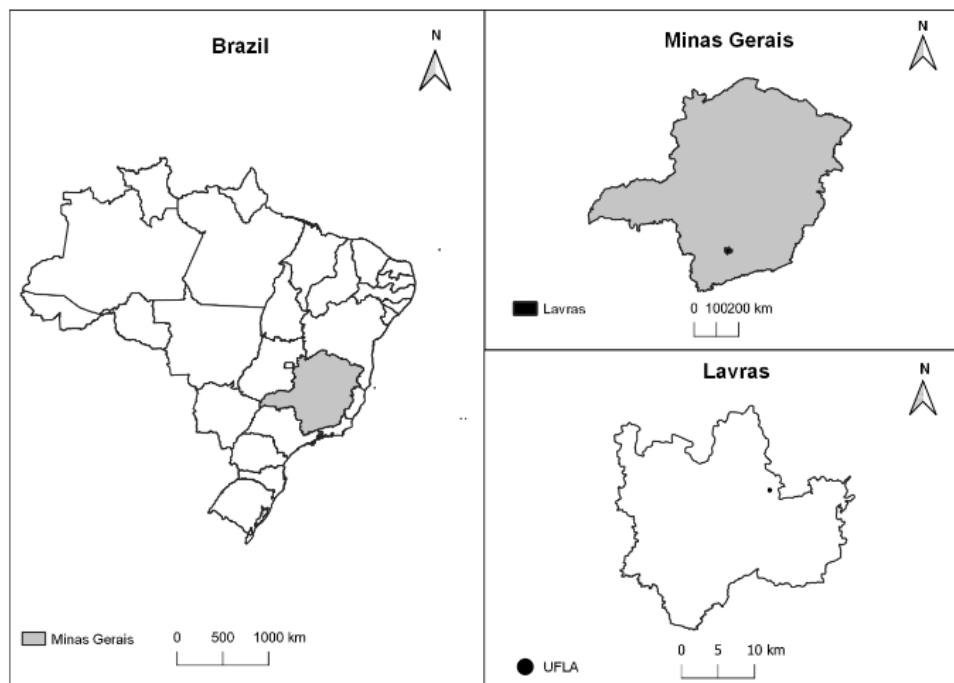
14 2.1. Study Area

15 Lavras is located in the Southern Minas Gerais state ($21^\circ 13' 45,3''\text{S}$ and $44^\circ 58' 32,4''\text{W}$), Brazil, 241 km from the
16 Atlantic Ocean (in a straight line), with an area of 564.744 km², altitude of 919 m (Fig.1) and a population of the 102,728
17 inhabitants [18]. Lavras soil is predominantly classified as Dystrophic Red-Yellow Oxisol (RYOd1) and Eutrophic Red-
18 Yellow Argisol (RYAe12). RYOd1 soils are suitable for tillage but are naturally acids [19]. Therefore, the liming process
19 is common, in which lime is applied to correct soil acidity for agricultural purposes. In this sense, Lavras generally is
20 known for its agricultural production, since has about 19% (107 km²) of its total area associated with agricultural activ-
21 ities, mainly coffee production [20].

22 Lavras also undergo influence of several anthropogenic activities including transport, farming, biomass burn-
23 ing and industrial activities such as agroindustrial, limestone mining and cement plants [17]. The vehicular fleet has
24 about 50 thousand light-duty vehicles, corresponding to 54% of automobiles and 26% of motorcycles. Moreover, vehicle
25 fleet was on average 15 years old, where 62% of passenger cars were produced before 2010 and only 14% before 1990
26 [21].

27 Regarding weather conditions, the city presents climatological (1981-2010) mean annual temperature of 20.3 °C,
28 with minimum (16.9 °C) values in July and maximum (22.8 °C) in February. The annual rainfall of 1461.8 mm is mainly

- 1 concentrated in two well defined periods: (i) rainy season from October to March (covering 85% of total rainfall); and
2 (ii) a dry one from April to September [22].



3

4 **Figure 1.** Geographical location of sampling site (UFLA) in Lavras city, Southern Minas Gerais state, Brazil

5

2.2. Sampling and Sample Analysis

6

7 A total of 65 bulk deposition samples were collected at *Universidade Federal de Lavras (UFLA)* campus in Lavras
8 city (Fig. 1), between November 2017 and October 2019. We using a handmade sampler composed of a high-density
9 polyethylene bucket (NALGON) of 10L with a collecting area of 439 cm², protected by a sun-protective PVC structure
10 and covered with a nylon mesh. In order to follow GAW's sampling procedures, we kept the sampler 1.5 m above the
11 ground and clean it with deionized water (18 MΩ) [23]. Highlighted that for dry atmospheric only samples, a 50 mL of
deionized water was added in order to analyze soluble species.

12

13 Samples were collected at intervals of about 7 days and after each collect were divided into two parts. In a
14 fraction we measure pH using pHmeter (AKSO AK model 151), calibrated with buffer solutions (pH 4.0 and 7.0). The
15 other sample aliquot was filtered with a 0.22 μm diameter membrane (Millex), stored in pre-cleaned polyethylene bot-
16 tles kept at -18 °C prior to ion chromatography (IC) analysis (Metrohm model 851) with anionic column (Metrosep
17 ASupp5 - 250 mm x 4 mm) and cationic column (Metrosep C2 150 - 150 x 4 mm). Analytical quantification was per-
18 formed using an external calibration curve from the standards concentrations for each ion. The species quantified by IC
were calcium (Ca²⁺), ammonium (NH₄⁺), magnesium (Mg²⁺), sodium (Na⁺), potassium (K⁺), nitrate (NO₃⁻), chloride (Cl⁻)

1), sulfate (SO_4^{2-}), fluoride (F^-), formate (CHO_2^-), acetate ($\text{C}_2\text{H}_3\text{O}_2^-$) and oxalate ($\text{C}_2\text{O}_4^{2-}$), all presenting detection limits (DL)
 2 lower than $0.8 \mu\text{molL}^{-1}$. Also, blank samples analysis were carried out through the sampling campaign.

3 2.3. Data Analysis

4 Data processing was performed by programming in R language, specifically the functions contained in the
 5 stats and ggplot2 packages [24,25].

6 2.3.1. Samples Validation

7 The internal consistency of the entire data set was analyzed through Cooperative Programme for Monitoring
 8 and Evaluation of Long-Range Transmission of Air Pollutants in Europe (EMEP) guidelines for rainwater samples val-
 9 idation [23]. This method is based in the equations 1, 2 and 3:

$$10 \quad IS = \sum_{cations} C_i + \sum_{anions} C_i (1)$$

$$11 \quad ID = \sum_{cations} C_i - \sum_{anions} C_i (2)$$

$$12 \quad IB = (ID/IS) * 10^2 (3)$$

13 Where C_i is the concentration of ion type i in a specific sample, expressed in μeqL^{-1} . IS is the sum of all ion
 14 concentrations and ID is the difference between the sum of the cation concentrations and the sum of the anion concen-
 15 trations. Both IS and ID are expressed in μeqL^{-1} . The ion balance, IB , expresses the difference, ID , in percent of the sum
 16 of all concentrations, IS . Apart from the equations, the samples pH values were also considered in this validation pro-
 17 cess, a critical value of 5.5 was used, moreover specific details are described elsewhere [23,26].

18 2.3.2. Non-measure Species Estimate

19 We estimate bicarbonate (HCO_3^-) concentrations from theoretical relationship between pH and HCO_3^- (Eq. 4), and
 20 for carbonate (CO_3^{2-}) we considered the same concentration of Ca^{2+} (WMO, 2004), since no direct method was applied
 21 for measuring these species.

$$22 \quad [\text{HCO}_3^-] = 5.1/10^{6-pH} (4)$$

23 2.3.3. Volume Weighted Mean

24 The volume weighted mean (VWM) was expressed through the relationship between the sum of the concentra-
 25 tions product of each species (X_i) found in the n samples by the respective volume (V_i) and the sum of all the samples
 26 volumes according to the equation 5 [27]. For dry atmospheric deposition samples, in which we added 50 mL of deion-
 27 ized water due to precipitation absence, we considered that volume for calculations.

$$28 \quad VWM = \frac{\sum_{i=1}^n (X_i * V_i)}{\sum_{i=1}^n V_i} (5)$$

29 2.3.4. Neutralization Factor

1 The neutralization factor (NF) or index was calculated to assess the role of crustal components (Ca^{2+} and Mg^{2+}) and
 2 NH_4^+ on the neutralization of rainwater acidity due to the presence of NO_3^- and SO_4^{2-} , using the equation 6 adapted
 3 from Huang et al. and Qiao et al. [28,29]:

$$4 \quad NF_X = \frac{[X]}{[\text{NO}_3^-] + [\text{SO}_4^{2-}]} \quad (6)$$

5 Where X is the cation of interest and all the concentrations were used in μeqL^{-1} .

6 2.3.5. Positive Matrix Factorization (PMF)

7 In order to identify sources of pollutants and their relative contributions we using the software Positive Matrix
 8 Factorization (PMF, version 5.0) made available by United States Environmental Protection Agency, which consists of
 9 a mathematical receptor model [30].

10 In general, a receptor model aims to solve the chemical mass balance between the species measured concentrations
 11 and the sources profiles [31]. In this sense, the PMF model decomposes a matrix of sample data into one matrix of factor
 12 contributions (G) and other of factor profiles (F). Then, for a data matrix X of n samples by m chemical species, the
 13 model will identify the number of factors p, the species profile f_k of each factor k, and the amount of mass g_k contributed
 14 by each factor k to each individual sample, according to Eq. 7:

$$15 \quad X_{ij} = \sum_{k=1}^p g_{ik} f_{kj} + e_{ij} \quad (7)$$

16 Where e_{ij} is the residual, i and j are the rows and columns of X.

17 The factor contributions and profiles are derived by the PMF model minimizing the Q function (Eq.8), with non-
 18 negativity constraints, i.e. $g_{ik} > 0$ and $f_{kj} > 0$ [32,33].

$$19 \quad Q = \sum_{i=1}^n \sum_{j=1}^m \left[\frac{X_{ij} - \sum_{k=1}^p g_{ik} f_{kj}}{u_{ij}} \right]^2 \quad (8)$$

20 Where u_{ij} are variable uncertainties, which allowing each data value to be individually weighted. These variable uncer-
 21 tainties (u_{ij}) for data with values below and above the DL were calculated according to equations 9 and 10, respectively
 22 [31,34].

$$23 \quad u_{ij} = 5 \overline{DL}_{ij} / 6 \quad (9)$$

$$24 \quad u_{ij} = \sqrt{(\text{Error Fraction} * \text{Concentração}_{ij})^2 + (0.5 * DL_{ij})^2} \quad (10)$$

25 Where the error fraction was estimated at 10% [35].

26 The quality of data was assessing based on the signal to noise ratio (S/N), which characterize species as strong
 27 when $S/N > 2.0$ and as weak in the cases that S/N ranged from 0.2 to 2.0. The species with $S/N < 0.2$ or values below DL
 28 greater than 50% were categorized as bad in quality and were excluded from the PMF analysis [36].

3. Results and Discussion

3.1. Samples Validation

The pH values for the whole data set (n = 65) ranged from 5.34 to 8.46 with average of 5.89. Given the samples alkaline behavior, quality assurance and quality control (QA/QC) were verified applying the criteria from EMEP. Following the QA/QC, among the 65 atmospheric deposition events, 7 were invalid, 3 valid and 55 valid and flagged (Fig.2a). Only seven invalid samples were rejected and, thereby, we considered a new data set (n = 58) in the following discussion.

In order to verify the electroneutrality principle, a sum of anions *versus* cations plot was depicted in Fig.2b, in units of μeqL^{-1} . In general, we observed a good and statistical significant linear fit ($R^2 = 0.80$ and $p\text{-value} < 0.01$) for the data set. In addition, the sum of cations was greater than the sum of anions, since the fitted line slope (value of 0.42 – Fig.2b) was lower than the unity. The anions deficit was reasonable in samples with pH values ranging from 5.5 to 6.0 [26] and which could have been attributed to the lack of measurements of carbonate and bicarbonate concentration [37–39]. In order to identify the carbonate deficit, an estimative of these species was carried out (WMO, 2004) and a new sum of cations *versus* anions plot was constructed (Fig.2c). It is noteworthy that the new angular coefficient presented a value of 1.09, which was closer to 1, with determination coefficient of 96% ($p\text{-value} < 0.01$). The bicarbonate estimation produced a poor linear adjustment (slope = 0.32 and $R^2 = 0.04$ with $p\text{-value} 0.07$) and therefore we considered only the presence of carbonate in the bulk atmospheric deposition samples.

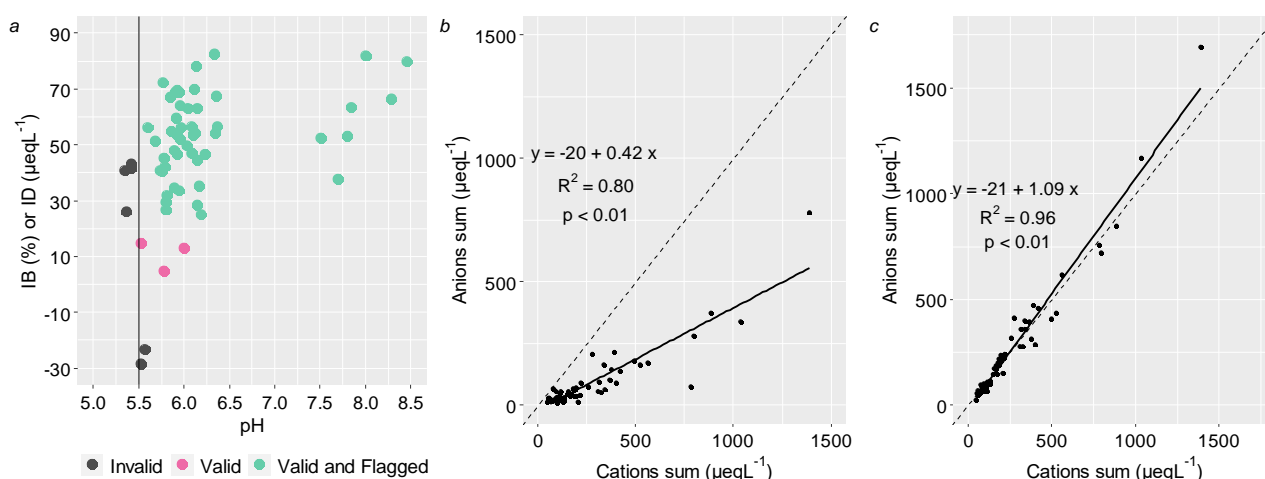
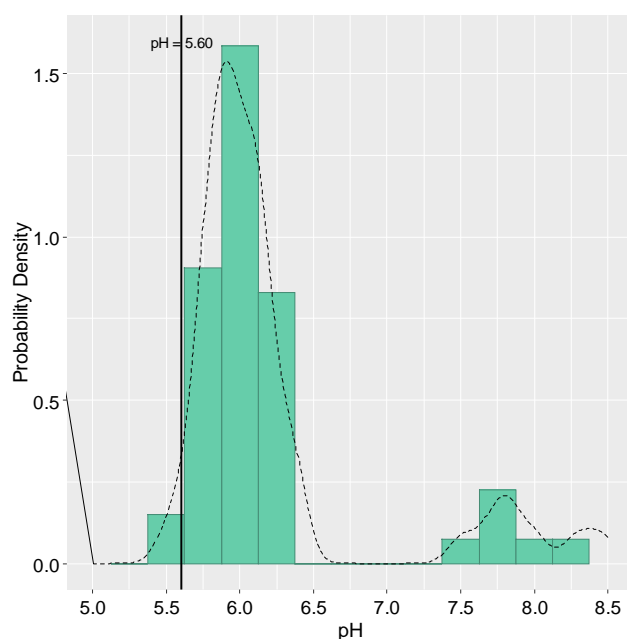


Figure 2. Ionic balance in bulk atmospheric deposition of Lavras city: (a) samples validation criteria from EMEP (n = 66) and the reference line represent a critical pH (5.5) value; (b) anions sum *vs.* cations sum (n = 58) and the dashed line represent the ideal slope of 1; (c) anions sum *vs.* cations sum considering the carbonate estimate (n = 58) and the dashed line represent the ideal slope of 1. Sampling period was between November 2017 and October 2019

3.2. pH Variation

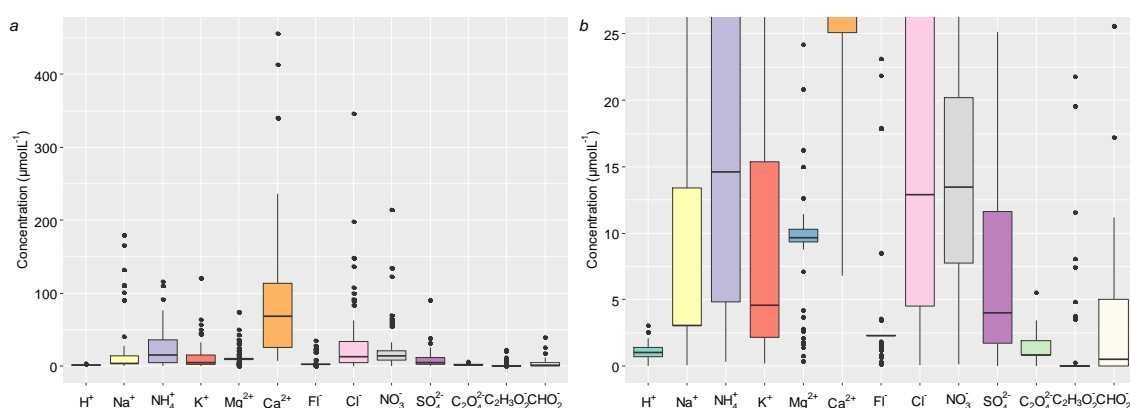
1 The mean pH value for validated samples ($n = 58$) was 5.99 (5.52–8.46) (Fig.3). Once 5.60 represents the pH
2 value resulting from the equilibrium between atmospheric CO_2 and pure water and is also used as a limit for acid rain
3 [40–42], our results suggests inputs of alkaline species into the atmosphere (98% of samples with $\text{pH} > 5.60$). This be-
4 havior is in agreement with other studies carried out in Brazil, such as those performed in Campo Bom, Rio Grande do
5 Sul [35] and in Juiz de Fora, Minas Gerais [43], in which the authors associated high pH values to inputs of crustal
6 aerosols, containing large fraction of carbonate and bicarbonate, in the atmosphere deposition.



7
8 **Figure. 3** Histogram of the pH values of the 58 bulk atmospheric deposition samples in Lavras city, from November 2017 until
9 October 2019. The vertical line refers to critical value for rainwater acidity classification ($\text{pH} = 5.60$) and the dotted line represents the
10 probability density curve

11 3.3. Ionic Composition

12 Concentration of major ions in μmolL^{-1} is illustrated in Fig 4. Among all cations, Ca^{2+} was the specie with greater
13 variability ($6.82 - 455 \mu\text{molL}^{-1}$), following by Na^{2+} ($0.03 - 179 \mu\text{molL}^{-1}$). Regarding anionic species, Cl^- was the dominant
14 compound with the largest variability ($0.06 - 346 \mu\text{molL}^{-1}$), following by NO_3^- ($0.11 - 213 \mu\text{molL}^{-1}$). The organic anions
15 (acetate, formate and oxalate) along with F^- and H^+ presented the smallest variability, since that all these ions had con-
16 centrations below than $50 \mu\text{molL}^{-1}$.



1
 2 **Figure 4** Box and whisker plots for the ions identified and quantified in bulk atmospheric deposition in Lavras, from November
 3 2017 until October 2019. Horizontal lines in the box represent the 25th, 50th and 75th percentile values. (a) Depicts all the ionic
 4 concentrations; and (b) magnifies the species with concentrations between 0 and 25 µmolL⁻¹

5 VWM concentrations were more amenable for comparisons and were calculated for all measured ions. For the
 6 whole sampling campaign, the ions profile in VWM (molar unit) may be described in the following order: Ca²⁺ (45.7) >
 7 Cl⁻ (19.1) > Na⁺ (16.6) > NH₄⁺ (14.4) > Mg²⁺ (12.8) > NO₃⁻ (9.46) > K⁺ (5.48) > F⁻ (4.00) > SO₄²⁻ (3.88) > HCO₂⁻ (1.92) > C₂H₃O₂⁻
 8 (1.41) > C₂O₄²⁻ (1.26) > H⁺ (0.77) µmolL⁻¹. We identified Ca²⁺ as the most predominant specie accounting for 33%, which
 9 coupled with Na⁺ and NH₄⁺ represents 56% of the total ionic species distribution. Similar patterns were observed in
 10 Mbita, East Africa, which also is characterized as a tropical agricultural area [44]. It is also valuable mentioning that Cl⁻
 11 was the second largest ion, contributing with 13% of the total VWM concentration.

12 3.4. Neutralization Factor

13 We calculated the NF index in order to assess the influence of alkaline species in atmospheric samples. The
 14 NF_{Ca²⁺} ranged between 0.99 and 29.6 while NF_{Mg²⁺} and NF_{NH₄⁺} varied from 0.05 to 11.0 and from 0.01 to 7.14, respectively.
 15 Considering the mean values for all study period, the most important neutralizing agent of sulfuric and nitric acids was
 16 Ca²⁺ (NF = 6.63), followed by Mg²⁺ (NF =1.54) and NH₄⁺ (NF = 1.11), which may be considered negligible. The role of
 17 Ca²⁺ as a dominant neutralization cation is reasonable since the calcium presented the highest VWM for the entire period
 18 studied. On the other hand, Mg²⁺ contributed more than NH₄⁺ for samples neutralization although NH₄⁺ has presented
 19 higher VWM concentration, suggesting that this neutralization proxy is better suited for assessing the major chemical
 20 species interactions in rich alkaline atmosphere such as in Lavras city. Highlighted that Ca²⁺ precursor, as calcium car-
 21 bonate, was sufficient to neutralize the major atmospheric acids, since most samples (n = 57 ~ 98%) showed NF higher
 22 than neutrality (NF = 1). This pattern is in agreement with the pH values found.

1 Similar phenomena have been observed in a number of previous studies in Brazil [35,45] and worldwide [46–
2 50]. In contrast, some studies have been reporting the predominance of NH_4^+ [51,52] and Mg^{2+} [6] in the atmospheric
3 neutralization process.

4 3.5. Source Identification Based on PMF Results

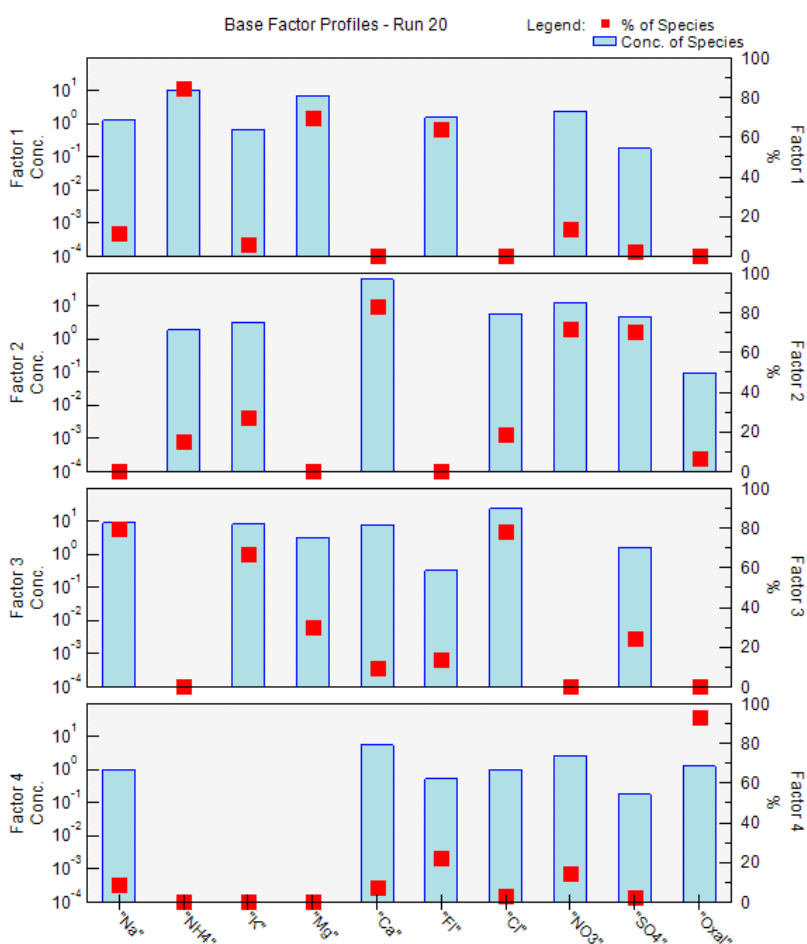
5 PMF is adopted to quantitatively analyze the emission sources of the measured ionic species. According to
6 S/N ratio, all species were categorized as strong, except the acetate that was classified as weak (S/N = 1.5). However,
7 due to large amount of data below of DL (> 50%), acetate and formate were classified as bad and excluded from the
8 analysis. The PMF model was run 20 times with a seed of 76 random starting points and a number of factors from 3 to
9 6 were assessed to get the optimal number of factors. The stability and reliability of the output were checked according
10 to following parameters: Q value, parameters of linear regression between predicted and observed concentrations, fre-
11 quency distributions of the scaled residuals, G-space plots and the physical meaning of the factor profiles. Based on the
12 evaluation of these criteria, we identify that four factors were physically reasonable and could represent the major
13 sources in the study region (Fig. 5).

14 The first factor identified presented high loading values for NH_4^+ (85%), Mg^{2+} (70%) and F^- (64%). The presence
15 of F^- is associated with HF gas emission, mostly from the productions of steel, electrolytic aluminum and phosphorus
16 fertilizers [53,54]. Mg^{2+} is originated naturally from the marine aerosols and/or crustal particles (Zhou et al. 2019), and
17 its anthropogenic sources can be associated to cement industries and fertilizers production. Regarding NH_4^+ , can be
18 directly attributed to an input of NH_3 in the atmosphere, mainly due to farming and nitrogen fertilizers production
19 and application [52]. This profile corroborates with the agricultural background of the study region, where an average
20 nitrogen and phosphor synthetic fertilizers application rate was estimated in 194 kg ha^{-1} [56]. In addition, there are sev-
21 eral agroindustrials inside the county air basin. Thus, this factor was categorized as *fertilizer application and production*.

22 The second factor, named *atmospheric neutralization processes*, was represented by Ca^{2+} (83%), NO_3^- (71%) and
23 SO_4^{2-} (70%). These species are involved in neutralizing atmospheric processes, since that CaCO_3 is an alkaline specie that
24 neutralize the major atmospheric acids (H_2SO_4 and HNO_3). Highlighted that cement manufacturing and limestone min-
25 ing were considered the two major sources of Ca^{2+} to the atmosphere in the study area. Although NO_3^- and SO_4^{2-} have
26 been related to the large emissions of SO_2 and NO_x from the combustion of fossil fuel in large urban areas [39], researches
27 carried out in agricultural areas in California, United States, associated these species with agricultural soil emissions
28 [57], which is reasonable with our study.

1 Factor 3, was loaded with Na⁺ (80%), K⁺ (67%) and Cl⁻ (78%). These species generally originate from sea-salt,
 2 crustal aerosols and biomass burning [58]. In the present study, the Cl⁻/Na⁺ ratio was 1.15, which was similar to oceanic
 3 ratio (Cl⁻/Na⁺ = 1.17), indicating marine sources intrusion. However, the K⁺/Na⁺ ratio was 0.33, which was higher than
 4 oceanic ratio (K⁺/Na⁺ = 0.022), suggesting a large excess of K⁺ and, therefore, suggesting an additional source of potas-
 5 sium compounds in Lavras city, as biomass burning. Because of the ambiguity of sources in this factor, was character-
 6 ized as *marine intrusion / biomass burning*.

7 Factor 4 was categorized as *biogenic emissions* due to high loaded for oxalate (93%). Organic acids in the atmos-
 8 phere in large urban areas originate from motor vehicle emissions, fossil fuel combustion and photochemical reactions.
 9 However, in rural areas biogenic emissions, such as vegetation release and the burning of biomass, are more important
 10 sources of organic acids [59].



11
 12 **Figure. 5** Source profiles extracted by EPA PMF 5.0 model from bulk deposition samples collected in Lavras, in the period
 13 between November 2017 and October 2019

14 **4. Conclusions**

1 We assessed bulk atmospheric deposition in a Brazilian agricultural region and observed that pH mean was
2 5.99 and most deposition samples (~98%) were alkaline (pH > 5.60). In addition, the most important neutralizing agent
3 of sulfuric and nitric acids was Ca²⁺ precursors (NF = 6.63). This process was corroborated by the higher abundance of
4 Ca²⁺ (33%) among all ions measured. The PMF analysis resulted in four factors, which demonstrated the influence of
5 anthropogenic and natural sources, such as fertilizer application and production, marine intrusion / biomass burning
6 and biogenic emissions, and revealed the importance of atmospheric neutralization processes.

7 The atmospheric deposition systemic analysis allowed to monitor and evaluate the chemical transformation
8 processes, reaction routes and the identification of the polluting sources. From that, it is possible to develop emission
9 reduction strategies, as effectively done in the previous decades for acid rain. Given this perspective, our findings are
10 useful to understand the majority atmospheric species sources in the Southern Minas Gerais, Brazil.

11
12 **Author Contributions:** Conceptualization, M.V.F., J.N.P and A.F.; methodology, M.V.F., J.N.P and
13 A.F.; software, J.N.P; data curation, M.V.F., J.N.P.; writing—original draft preparation, M.V.F.,
14 J.N.P; writing—review and editing, M.V.F., J.N.P, A.F.; supervision, M.V.F.; project administration,
15 M.V.F.; funding acquisition, M.V.F. and A.F.. All authors have read and agreed to the published
16 version of the manuscript.

17 **Institutional Review Board Statement:** Not applicable.

18 **Informed Consent Statement:** Not applicable.

19 **Acknowledgments:** The authors thank to Coordenação de Aperfeiçoamento de Pessoal de Nível
20 Superior (CAPES) and Fundação de Amparo à Pesquisa do Estado de Minas Gerais (Fapemig) for
21 the Graduate and Undergraduate Scholarships. Special thanks go to “Laboratório de Processos At-
22 mosféricos da Universidade de São Paulo (LAPAt-IAG-USP)” for the facilities and equipments used
23 in this study.

24 **Conflicts of Interest:** The authors declare no conflict of interest.
25

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