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A closure study on aerosol extinction in urban air in Hungary

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Abstract—In this study, we present our results from an investigation into the use of visibility data as a viable tool for the survey of long-term variations in air quality. We found that visibility data in general can be used to estimate atmospheric aerosol extinction coefficients, and that PM₁₀ can be successfully estimated from aerosol chemical composition. Our results indicate that PM₁₀ concentrations provide a good basis for the reconstruction of aerosol extinction coefficients. It was also shown that both derived (from visibility) and reconstructed aerosol extinction coefficients were in good accordance with each other, mainly in the case of dry aerosols. Ambient values can be determined if an adequate hygroscopic growth rate for aerosol extinction is considered. We also found that a rather precise estimation of extinction coefficient can be reached if a modified version of the widely used IMPROVE formula is applied.

Key-words: visibility, reconstruction of extinction coefficient, PM₁₀, aerosol composition

1. Introduction

Visibility (VIS) is a good, simple indicator of air quality. It is well known that VIS is inversely related to atmospheric extinction. Light extinction is controlled by the scattering and absorption of air molecules and aerosol particles and is proportional to the number concentration of molecules and particles. The number of air molecules is a function of temperature and pressure; however, its variation does not result in a significant change in VIS. Drastic decreases or increases in visibility can be attributed to variations in aerosol concentration and changes in the physical and chemical properties of the particles.

According to the Koschmieder theory (*Koschmieder, 1924*), visibility is determined by

$$VIS = \frac{\ln 0.02}{\sigma_{ext}} = \frac{3.912}{\sigma_{ext}}. \quad (1)$$

In the formula, the constant of 3.912 represents the 2% contrast threshold of daylight visual detection of objects against the sky horizon, and σ_{ext} is the total extinction of solar radiation at 550 nm wavelength.

The scattering and absorption efficiencies of particles are functions of aerosol chemical composition and particle size. Sulfates, nitrates, and organics generally contribute to scattering, while elemental and organic carbon compounds are mainly responsible for absorption (e.g., *Malm et al., 1994; Tao et al., 2012*). Particles participate most in light extinction when their size is in the optically active size range (0.1–1.0 μm). On the other hand, the water vapor content of air (specifically relative humidity) has a significant influence on ambient light extinction as a consequence of water soluble and hygroscopic compounds in the aerosol. According to previous studies, hygroscopic growth can cause the scattering coefficient of dry particles to be doubled or greater as a result of increases in relative humidity (e.g., *Horvath, 1992; Seinfeld and Pandis, 1998*). Consequently, variations in aerosol concentration (through their size and chemical composition) control changes in dry extinction/visibility; however, in ambient air, water content can also play a major role in light attenuation and visibility impairment (e.g., *Jung et al., 2009; Cheng et al., 2011*).

High concentrations of air pollutants are a prevalent cause of air quality impairment in both cities and remote areas. Visibility can vary within a wide range, from a few meters to a few hundred kilometers (*Horvath, 1995; Singh and Dey, 2012*), and can easily become a critically important parameter in the everyday functioning of cities, because low visibility can obstruct surface and aerial traffic, and thereby, unfavorably impact businesses, public safety, and even tourism.

Despite the significant influence of water vapor, visibility (extinction) data provide an efficient and inexpensive tool for the study of long-term variations in air quality and may be utilized as a proxy for the concentration of aerosols and trace gases (*Singh and Dey, 2012*). For several decades, an empirical formula for calculating the light extinction coefficient as a function of chemical species in the PM_{2.5} particulate matter has been used by the Interagency Monitoring of Protected Visual Environments (IMPROVE) network (*Hand et al., 2011; Malm et al., 2013*). The IMPROVE program is a cooperative measurement effort among the U.S. Environmental Protection Agency (EPA), federal land management agencies, and state agencies (*Hand et al., 2011*). The IMPROVE network has collected air quality data since 1988. The main goals of this program are to monitor real-time visibility and aerosol conditions in 156 mandatory Class I areas throughout the United States to identify aerosol species and their emission sources that are responsible for anthropogenic visibility impairment and to study and document long-term trends in air quality and visibility.

The empirical formula used by the IMPROVE network is based on the relationship between light extinction and aerosol chemical composition. The light extinction coefficients of an external mixture of aerosols can be estimated by assuming a linear combination of mass concentrations (M_j) and the corresponding extinction efficiencies (α_j) of different aerosol species (*Hand et al., 2011*):

$$\sigma_{ext} = \sum \alpha_j \cdot M_j. \quad (2)$$

To account for the hygroscopic effect, extinction efficiencies are multiplied by a humidification factor that is computed by assuming a size distribution and a composition-dependent hygroscopic growth factor. The IMPROVE formula (see later Eq. (7)) is used to reconstruct σ_{ext} (corresponding to 550 nm wavelength) based on measurements of aerosol composition (ammonium sulfate, ammonium nitrate, particulate organic matter, light absorbing carbon, soil, and coarse mass) and a Rayleigh scattering term (*Hand and Malm, 2007*). The units of aerosol extinction coefficient and Rayleigh scattering are Mm^{-1} , mass concentrations are given in $\mu g m^{-3}$, and mass extinction (scattering and absorption) efficiencies have units of $m^2 g^{-1}$.

The empirical IMPROVE formula has been applied extensively for different environments, from regional background to urban areas. An important issue is whether this algorithm, which is designed for background air, can be applied in cities. *Cabada et al. (2004)* found that by applying this formula during the Pittsburgh Air Quality Study, the scattering coefficient could be reproduced based on bulk PM_{2.5} composition with relative success. Furthermore, in a number of studies, the extinction coefficient or visibility of megacities in China was reconstructed based on this formula. *Cao et al. (2012)* found that in

Xi'an, China, the model underestimated the measured extinction coefficient and that ammonium sulfate was the largest contributor. *Cheng et al.* (2011) in Jinan, *Pan et al.*, (2009) at a rural site near Beijing and *Tao et al.* (2012) in the Guangzhou urban area all concluded that the IMPROVE formula can provide realistic estimates of the real atmospheric extinction in cases where the relative humidity (RH) was less than 70% (*Tao et al.*, 2012). Under high RH, due to the hygroscopicity of the particles, the absorbed water plays a much greater role in limiting visibility. The study by *Singh and Day* (2012), conducted in the megacity of Delhi, India, resulted in similar conclusions. They found that below 80% RH, aerosols contribute ~90% to the observed visibility degradation, but that above 80% RH, the aerosol relative contribution decreases rapidly due to the strong impact of hygroscopicity. Visibility is most sensitive to water-soluble particles and soot in all seasons.

In this work, we studied and discussed the following issues:

- a. How the extinction coefficient derived from visibility data can be estimated using the PM₁₀ mass and chemical composition;
- b. How PM₁₀ can be reconstructed from the chemical composition of PM₁₀;
- c. How PM₁₀ can be modelled from visibility observations. This information can result in a retrospective estimation of PM₁₀ for periods when PM₁₀ data are not available;
- d. How the derived extinction coefficient (from VIS) corresponds to the reconstructed data;
- e. Similar to other studies, we applied the IMPROVE formula in Budapest. We aimed to clarify how it can be applied in a Central European city and how the parameters in the formula should be changed to better reproduce the measured extinction coefficient;
- f. How the hygroscopic effect should be considered to obtain a viable ambient extinction coefficient.

2. Experimental

2.1. Sampling

In this study, data from two sampling campaigns representing winter (February 2 – March 2, 2009) and summer (July 20 – August 20, 2009), aerosols are presented. In both cases, the sampling was conducted at the Marcell György Observatory of the Hungarian Meteorological Service. This site is located in the south-eastern part of Budapest, Hungary. Here, at a standard synoptic weather station, meteorological parameters, including visibility, temperature, and dew point temperature, are measured on an hourly basis. Visibility is also determined

by visual observation. In the observatory, there is also an urban background air pollution monitoring site operated by the Hungarian Air Quality Network (www.levegominoseg.hu), which provides PM_{10} mass concentration. PM_{10} is monitored using the β -gauge method (*Chueinta and Hopke, 2001*).

In addition to these routine measurements, aerosol samples were collected daily to determine their chemical composition. A two-stage multi-jet impactor was applied at a sampling rate of 20 Lmin^{-1} . The PM_1 fraction was collected on quartz filters, and PM_{1-10} was sampled on Al-foils. From these samples, the inorganic ion (sulfate, nitrate, chloride, ammonium, sodium, potassium, and magnesium) and total carbon contents were measured. Inorganic ion content was determined by ion chromatography (Dionex, 2120) with a detection limit below 10 ppb. The total carbon concentrations of the aerosol samples were measured using an Astro Model 2100 TOC analyzer. This method is based on NDIR absorption. The detection limit of these measurements was $2 \mu\text{g C}$.

In the winter campaign, the scattering coefficient of aerosol at 550 nm was also monitored using an M903 integrating nephelometer calibrated with carbon dioxide. These data were used to reconstruct the scattering coefficient (see in Section 3.3). Unfortunately, in the summer campaign, this measurement was not available due to instrument failure. In both campaigns, the daily average absorption coefficients of the aerosol samples were determined indirectly. We supposed that the majority of absorbing components (soot) can be found in the PM_1 aerosol. Using the PM_1 quartz filters, the absorbance of the samples was determined (Eq. (3)). The light transmittance of blank and exposed filters was measured by PSAP (particle soot absorption photometer) at 550 nm. Their ratio gave transmittance (T), and absorbance (A) was derived by

$$A = \log\left(\frac{1}{T}\right). \quad (3)$$

Considering the air volume (V) and surface (S) of the filter, the average absorption coefficient was estimated:

$$\sigma_a = \frac{A \times S}{V}. \quad (4)$$

This method was checked with parallel measurements. During the summer campaign, a PSAP was operated to directly monitor the absorption coefficient. Using the same PSAP, the absorption coefficient was measured both directly and indirectly. From the direct measurements, the daily average absorption coefficients are calculated and compared to the absorption coefficients resulted from the indirect method. The results of this comparison are shown in *Fig. 1*. One can conclude that the absorption coefficients determined in both ways are

linearly related, as the average difference between them was $2.16 \pm 2.39 \text{ Mm}^{-1}$. This means that the absorption coefficient was slightly overestimated (less than 10%) when it was calculated using the transmittance of the aerosol filter.

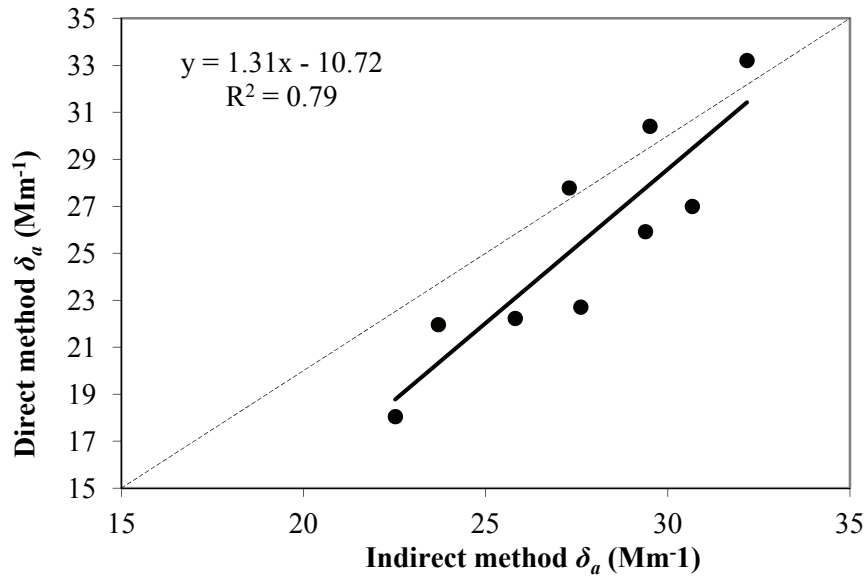


Fig. 1. Comparison of directly and indirectly measured absorption coefficients.

Finally, the organic and elemental (light absorbing) carbon contents of the samples were differentiated. We assumed that the total carbon content of the aerosol would be composed of organic and elemental fractions and that the inorganic carbon would be negligible (see in detail in Section 3.1). The elemental fraction was estimated from the absorption coefficient, considering a mass absorption coefficient of $10 \text{ m}^2 \text{ g}^{-1}$. Additionally, organic carbon was calculated as the difference between the total carbon (TC) and elemental carbon concentrations.

2.2. Extinction coefficient and the effect of hygroscopicity

Visibility (VIS) is generally determined by the light extinction of aerosol particles and air molecules. The extinction coefficient (σ_{ext}) can be estimated by means of Koschmieder theory (see Eq. (1)) which refers to 550 nm wavelength.

As previously mentioned, due to the hygroscopicity of the particles, relative humidity also plays an important role in determining visibility. The hygroscopic effect was excluded from the data sets by using the γ -approach as described by, e.g., Zhou et al. (2001):

$$\sigma_e = 10^c \times \left(1 - \frac{RH(\%)}{100}\right)^{-\gamma}, \quad (5)$$

$$f(RH) = \sigma_{RH} / \sigma_{40\%}. \quad (6)$$

We found that hygroscopic growth of the extinction coefficient (Eq. (6)) derived from VIS data was substantially different in winter and summer, as shown in *Table 1*. In winter, the aerosol was found to be much more hygroscopic than in summer, and the increase in the particle growth rate with rising RH was considerably greater than that in summer (e.g., at 80% RH, the growth rate was twice as much in winter than summer). As a result, considering the same PM₁₀ concentrations, this difference yields a doubled extinction coefficient in winter compared with the summer values.

Table 1. Hygroscopic growth rate of extinction coefficients as a function of relative humidity

Relative humidity (%)	Winter	Summer	Winter/Summer
40	1.0	1.0	1.0
50	1.2	1.1	1.1
60	1.6	1.2	1.2
70	2.2	1.4	1.5
80	3.4	1.7	2.0
90	7.3	2.3	3.1

3. Results and discussion

3.1. Chemical composition and reconstruction of PM₁₀

The PM₁₀ mass concentration was more variable during the winter campaign. In winter and summer, the average PM₁₀ concentrations were 32 μgm^{-3} and 23 μgm^{-3} , respectively, with maximum concentrations of 107 μgm^{-3} and 54 μgm^{-3} . These average concentrations do not differ significantly from those obtained in three Austrian cities (*Gomišček et al.*, 2004), and Bologna, Italy (*Matta et al.*, 2003). In Vienna, Linz, Graz, and Bologna, the winter and summer PM₁₀ concentrations varied in the ranges of 27–39 μgm^{-3} and 17–26 μgm^{-3} , respectively (*Gomišček et al.*, 2004; *Matta et al.*, 2003). In Lens, France (*Waked et al.*, 2014), the overall PM₁₀ concentrations were lower, and in winter and summer, the values were 20 μgm^{-3} and 14 μgm^{-3} , respectively. Chemical analysis showed that in Budapest, carbonaceous compounds dominated the PM₁₀ compositions, and the mass fractions of organic compounds in winter and summer were 35% and 27%, respectively. In Bologna and Lens, the mass fraction of organic carbon in PM₁₀ was rather similar to that of Budapest (Bologna: winter 35% and summer 37% (*Matta et al.*, 2003); Lens: winter 34% and summer 27% (*Waked et al.*, 2014)). In contrast, in Budapest, elemental carbon represented 13% and 14% of PM₁₀, which was generally higher than the EC/BC data published for the other urban background sites around Europe. In

Barcelona, London North Kensington, Lugano (*Reche et al.*, 2011), Bologna (*Matta et al.*, 2003), and Lens (*Waked et al.*, 2014), the mass fractions of EC/BC in PM₁₀ were in the range of 4%–10%.

In Budapest, the inorganic compounds were 16% (winter) and 18% (summer) of the total PM₁₀ mass. These results are in accordance with earlier results obtained for PM₁₀ in Budapest (*Maenhaut et al.*, 2005), but they are significantly lower than those found in other cities. It was found that in winter and summer in Bologna, inorganic species were 53% and 41% of PM₁₀ (*Matta et al.*, 2003), while in Lens their contributions were 52% and 42% (*Waked et al.*, 2014).

During both campaigns in Budapest, all components analyzed from the filters were found dominantly in PM₁ as shown in *Table 2*. Among inorganic ions, nitrate was dominant in winter, whereas in summer, sulfate was found in highest concentrations, which is similar to results obtained in other cities. In Bologna, nitrate concentrations were almost 3 times higher than those of sulfate (*Matta et al.*, 2003), whereas in Lens, nitrate was twice as high as sulfate (*Waked et al.*, 2014). The lower nitrate concentrations in summer are the result of the temperature dependency of ammonium nitrate volatility, which was also indicated by lower nitrate fractions among the fine mode compared with the values obtained in winter. Yearly increases in sulfate concentrations during summer months are already known (e.g., *Hidy et al.*, 1978). Higher photochemical activities in summer result in higher rates of SO₂ conversion, which yield summertime maximums in sulfate concentrations. *Table 2* shows that the contributions of fine sulfate, ammonium, and total carbon to PM₁₀ were significantly higher during summer than in winter. In contrast, the other components were less accumulated in PM₁ in summer than winter. Specifically, the fraction of fine nitrate concentration decreased from 80% (winter) to 70% (summer).

Table 2. PM₁₀ aerosol composition and PM₁ mass fractions (%). The standard deviation is given in parentheses.

	Winter		Summer	
	Concentration ($\mu\text{g m}^{-3}$)	Fraction in PM ₁ (%)	Concentration ($\mu\text{g m}^{-3}$)	Fraction in PM ₁ (%)
chloride	2.4 (4.4)	85 (15)	1.1 (1.2)	77 (17)
nitrate	7.1 (3.6)	80 (12)	2.2 (0.9)	70 (8)
sulfate	4.2 (2.8)	80 (12)	6.2 (2.7)	93 (4)
sodium	2.0 (0.9)	95 (7)	1.4 (1.3)	65 (30)
ammonium	1.6 (1.6)	82 (17)	0.9 (0.5)	94 (13)
potassium	0.6 (0.3)	96 (5)	0.7 (0.5)	77 (18)
magnesium	0.1 (0.1)	85 (14)	0.2 (0.1)	71 (10)
calcium	1.3 (0.6)	81 (14)	1.7 (0.9)	69 (13)
total carbon	12.5 (6.6)	81 (4)	6.5 (1.2)	88 (7)

The total carbon concentration was twice as much in winter than in summer, and in both seasons more than 80% of the total carbon (TC) concentration was found in PM₁ (see *Table 1*). It is supposed that fine TC is composed primarily of organic and elemental carbon, and that the contribution of inorganic carbon (carbonate) can thus be neglected (e.g., *Karanasiou et al.*, 2011). Carbonate may be present in the coarse fraction; however, its presence has not been ubiquitously confirmed. According to a European survey (*Sillanpää et al.*, 2005), among six cities (Duisburg, Prague, Amsterdam, Helsinki, Barcelona, and Athens), carbonate was detected in the coarse mode in only the two Mediterranean cities. It should be mentioned that in Barcelona and Athens, the coarse fraction of the aerosol was greater than fine, whereas in Duisburg, Prague, and Amsterdam, similarly to Budapest, the fine mode dominated. Other results obtained for Chinese cities also showed the inorganic carbon (carbonate) content of the aerosol to be rather low (*Wang et al.*, 2010). Considering these results, we neglected the contribution of inorganic carbon in both the fine and coarse size ranges. Upon further evaluation, we supposed that the total carbon of PM₁₀ was composed of organic and elemental (light absorbing) carbon and that this latter could be exclusively found in the fine fraction (PM₁).

The aerosol chemical composition was reconstructed on the basis of PM₁₀, and the inorganic and carbonaceous compounds were all considered. The chemical mass closure of the inorganic constituents was based on stoichiometry. In addition, based on the recommendations of *Stelson and Seinfeld* (1981), other alkaline (potassium) and alkaline earth metallic (calcium and magnesium) ions were included in the reconstruction of the PM₁₀ mass. Excess nitrate was assumed to be organic nitrate (e.g., *Fry et al.*, 2014), and the organic and elemental carbon mass concentrations were estimated using conversion factors of 1.4 and 1, respectively. The average chemical composition of the reconstructed aerosol is shown in *Fig. 2*. It should be mentioned that in summer, the inorganic fraction of the aerosol was composed mainly of sulfate containing compounds. In contrast, during the winter season, nitrate compounds dominated the inorganic aerosol fraction, and sulfates (generally in the form of ammonium sulfate) gave a smaller contribution.

Finally, the reconstructed and the directly measured PM₁₀ (by the β -gauge monitor described in Section 2.1) mass concentrations were compared. *Fig. 3* shows that these mass concentrations agreed relatively well in both sampling campaigns. In winter, PM₁₀ is overestimated by 3.2 $\mu\text{g m}^{-3}$, whereas in summer, the directly measured PM₁₀ is slightly lower than the reconstructed value.

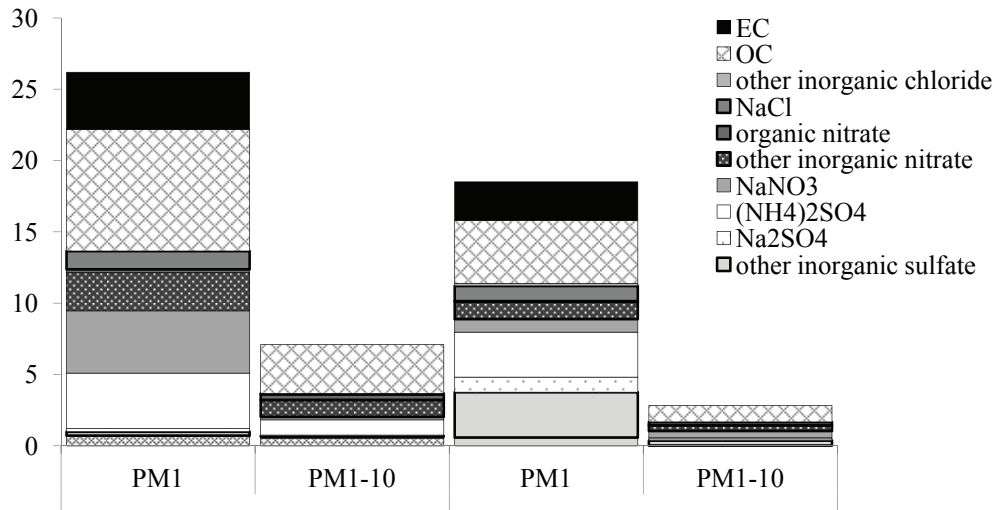


Fig. 2. Chemical mass closure of PM_{10} and PM_{1-10} in Budapest.

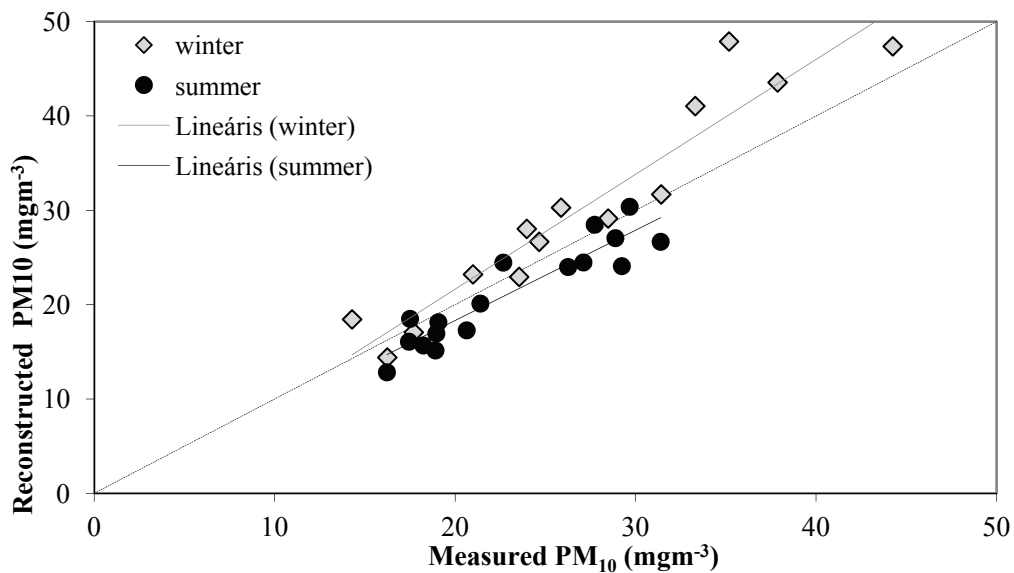


Fig. 3. Reconstructed vs. measured PM_{10} concentration.

3.2. Visibility and extinction coefficient (VIS , $\sigma_{e,VIS}$)

Temporal variations in PM_{10} , visibility, and RH are presented in Figs. 4a (winter) and 4b (summer). The relationship among the parameters is clear. Low visibility coincides with high PM_{10} concentrations and/or high RH; conversely, high visibility occurs when PM_{10} and RH are low. As an example, in Fig. 4a, one can follow the development of a winter air pollution episode beginning on February 18 and finishing on February 25. Parallel to a general increase in PM_{10} (occasionally exceeding $100 \mu\text{gm}^{-3}$), visibility decreased (average $VIS = 7 \text{ km}$),

which was disrupted by a change in RH. In summer, aerosol aging processes also influence variations in visibility. *Bäumer et al.* (2008) demonstrated that when the prevailing air mass undergoes an aging process, and as a result, a significant decrease in VIS is observed, an increase in PM₁₀ can be detected.

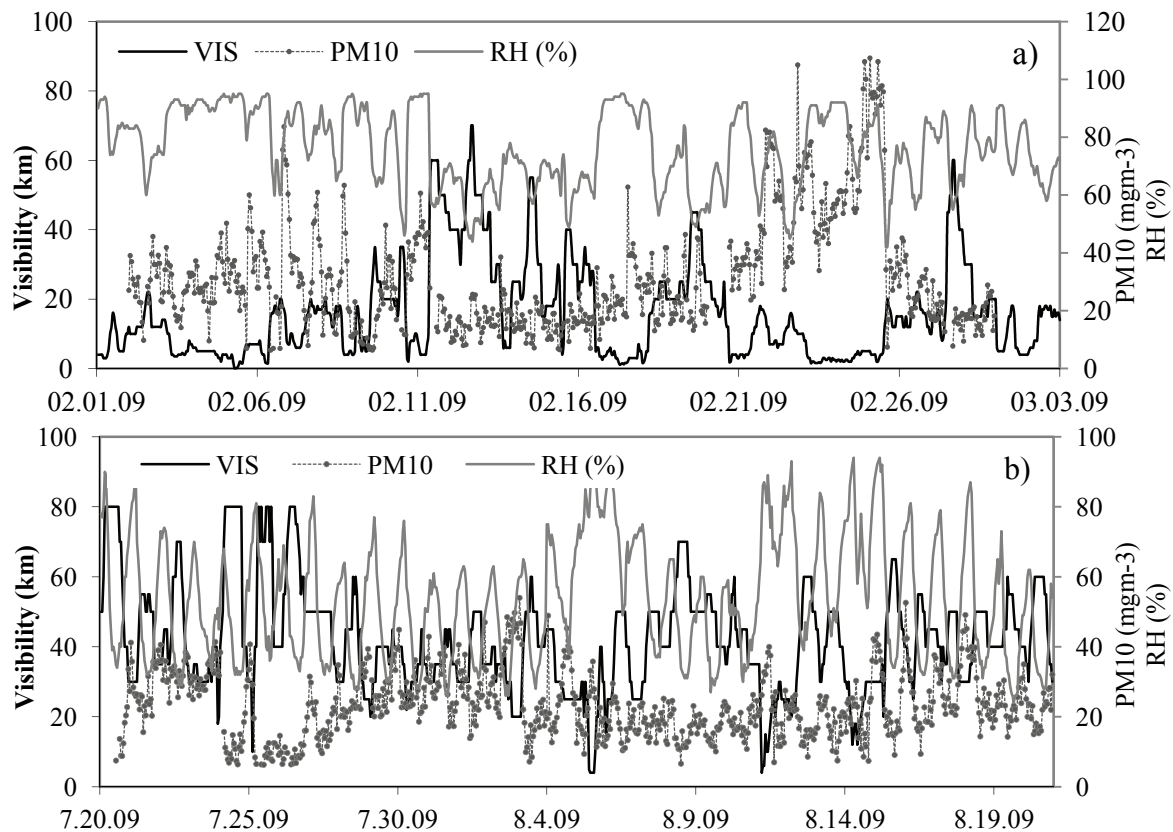


Fig. 4. Temporal variation in VIS, PM₁₀ and RH during winter (a) and summer (b) campaigns.

In *Figs. 5a* and *5b*, temporal variations in the ambient and dry extinction coefficients are presented. The ambient extinction coefficients were derived from VIS (Eq. (1)), whereas the dry extinction data referring to 40% RH was obtained by means of the γ -approach (see Eqs. (5) and (6) in Section 2.2). In winter and summer, the average ambient extinction coefficients were 550 Mm⁻¹ and 103 Mm⁻¹, and the dry average values were 126 Mm⁻¹ and 87 Mm⁻¹, respectively. The difference between the ambient and dry extinction coefficients is attributed to the hygroscopic behavior of the aerosol. The effect of hygroscopicity on the aerosol extinction was particularly important in winter (see *Fig. 5a* and *Table 1*). The significant variation in hygroscopic growth rate is assumed to be the result of seasonal changes in PM₁₀ chemical composition.

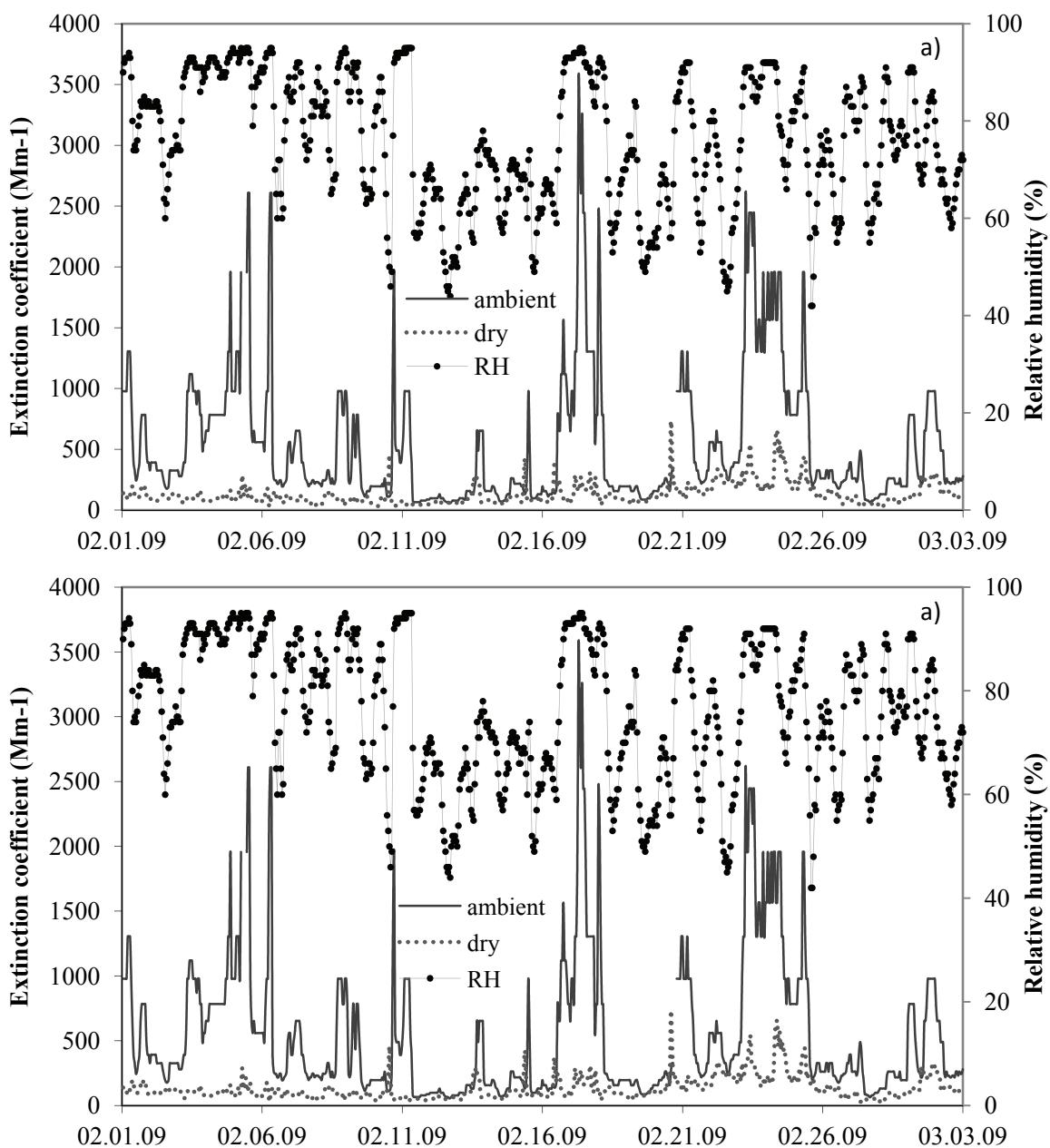


Fig. 5. Temporal variation in the ambient and dry extinction coefficients during winter (a) and summer (b) campaigns in 2009.

The relationship between PM_{10} and dry extinction coefficient is presented in Fig. 6. In both seasons, dry extinction coefficient varied similarly as a function of PM_{10} . Based on a linear regression analysis of the combined data sets, the dry mass extinction efficiency was $2.2 \text{ m}^2\text{g}^{-1}$, with a correlation coefficient of 0.52. This value is in accordance with the mass extinction efficiencies found in typical continental air. According to Nemuc *et al.* (2013), the mass extinction efficiencies of PM_{10} are typically in the range of 2.2 and $2.7 \text{ m}^2\text{g}^{-1}$, which was further confirmed by the value of $2.6 \text{ m}^2\text{g}^{-1}$ that was obtained at the Hyytiälä Forestry Field Station in central Finland (Virkkula *et al.*, 2011).

Moreover, in urban air, the mass extinction coefficients of PM₁₀ do not differ significantly from those obtained for background air. For reference, *Kim (2015)* and *Jung et al. (2009)* reported mass extinction efficiencies of 2.7 m²g⁻¹ and 2.5 m²g⁻¹ for Seoul, Korea, and Beijing, China, respectively.

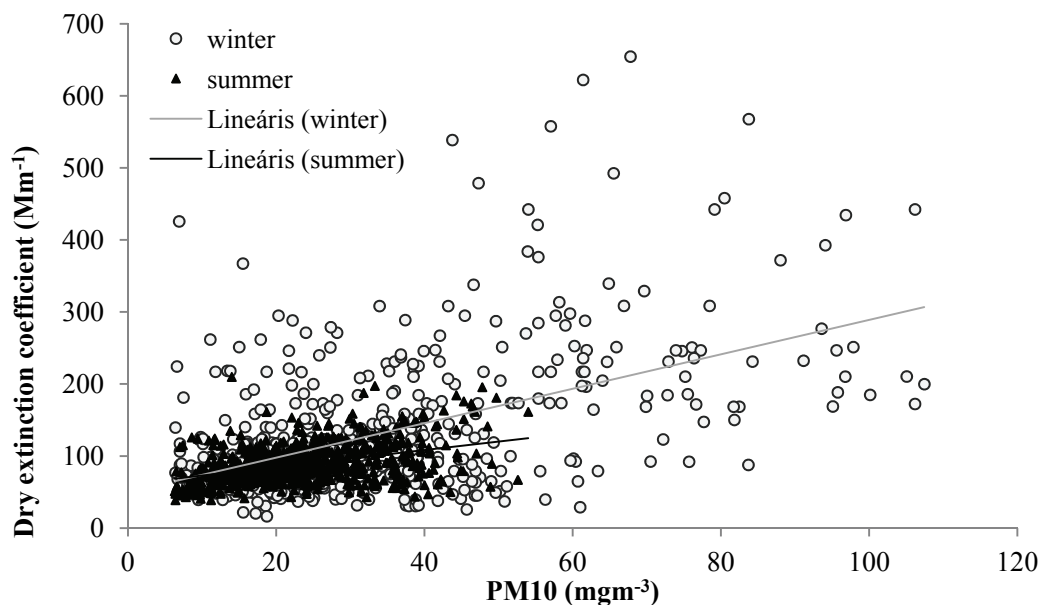


Fig. 6. PM₁₀ vs. dry extinction coefficient during the winter and summer campaigns. The relationship between the parameters is statistically significant (at p=99.9%).

3.3. Reconstruction of extinction coefficient

The reconstruction of extinction coefficients is based on the well-known relationship between light extinction and aerosol composition. Sulfate, nitrate, and organic constituents have greater importance in scattering, whereas elemental carbon is more responsible for light absorption. Based on the concentration of different compounds and their mass extinction efficiencies, the extinction coefficient can be estimated (see Eq. (2)) (*Ouimette and Flagan, 1982; Mészáros, 1999, Seinfeld and Pandis, 1998*). This method has been well applied for a long time by the IMPROVE program (*DeBell et al., 2006*) using the following equation:

$$\sigma_e \approx 3 \times f \times [(NH_4)_2SO_4] + 3 \times f \times [NaNO_3] + 4 \times [OC] + 10 \times [EC] + 1 \times [groundbased] + 0.6 \times [coarse] + 10 \times (Rayleigh). \quad (7)$$

To reconstruct the ambient extinction coefficient, the dry extinction coefficient was first estimated. In the estimation, we used aerosol chemical composition and optical data, which were independently monitored from visibility. In winter, aerosol scattering and absorption coefficients were measured along with visibility by integrating a nephelometer and PSAP (see

details in Section 2.1). The dry aerosol scattering coefficient was measured by the nephelometer and calculated using the gamma approach (Eq. (5)) that is used when estimating visibility. To estimate dry aerosol scattering, the main aerosol components including ammonium sulfate, ammonium, sodium, and other nitrates as well as organic compounds were taken into account. Multiple regression analysis was applied to determine mass scattering efficiencies. In addition to scattering, aerosol absorption was considered, and EC was calculated from the absorption coefficients of the aerosol daily samples (see details in Section 2.1.). Finally, Rayleigh scattering, which is a function of air temperature and pressure, was determined. We found that the reconstructed dry extinction coefficient can be calculated with the following equation:

$$\sigma_{dry} \approx 2.3 \times [(NH_4)_2SO_4] + 1.7 \times [NaNO_3 + KNO_3 + Ca(NO_3)_2 + Mg(NO_3)_2] + 1.5 \times [OC] + 10 \times [EC] + 12 (Rayleigh). \quad (8)$$

When comparing the mass scattering/absorption efficiencies used in the IMPROVE network to our results, some similarities and differences were noticed. Our equation refers to PM₁₀, whereas in the IMPROVE network, PM_{2.5} and coarse particles are considered separately. The mass scattering efficiencies of ammonium sulfate and nitrate salts were rather similar ($\approx 2 \text{ m}^2\text{g}^{-1}$) to those identified by the IMPROVE network ($3 \text{ m}^2\text{g}^{-1}$), although our obtained value was 30% smaller. In contrast to the inorganic species, the difference in mass scattering efficiencies for the organics is quite high (this study: $1.5 \text{ m}^2\text{g}^{-1}$; IMPROVE: $4 \text{ m}^2\text{g}^{-1}$). One possible explanation for the smaller values could be the difference between PM₁₀ and PM_{2.5}, not differentiated in this study.

In addition, for the sake of simplicity, we modeled the dry extinction coefficient on the basis of only sulfate and nitrate ion as well as, organic and elemental carbon concentrations. In this case, the reconstruction equation was

$$\sigma_{dry,anions} = 4.3 \times [SO_4^{2-}] + 1.3 \times [NO_3^-] + 1.5 \times [OC] + 10 \times [EC] + 12 (Rayleigh). \quad (9)$$

Comparing our models (Eqs. (8) and (9)) to the IMPROVE model (Eq. (7)), we concluded that all three models gave similar results for the reconstruction of the dry extinction coefficient. In *Table 3* the relationships between reconstructed and “observed” (calculated from visibility) extinction coefficients are shown. These relationships are characterized by linear regression equations and correlation coefficients. On the basis of these parameters, one can conclude that dry extinction data can be almost equally reconstructed by all three models. In other words the IMPROVE model – which is constructed for background aerosol – also provides sufficiently good estimation of dry extinction coefficient even in urban air, in Budapest. This is in agreement with other studies which

indicated similar conclusions in megacities of Beijing and Delhi: in relatively dry atmosphere, this model provided realistic estimation of the ambient extinction coefficient (*Tao et al.*, 2012; *Sing and Day*, 2012). It has to note, that our models (mainly Eq. (9)) require less input data than the IMPROVE model; and for this reason they can be more easily applied for the available data sets.

Table 3. Relationship between reconstructed (three models) and “observed” extinction coefficients

	Linear regression equation	Correlation coefficient
Dry aerosol		
Eq. (8)	$y = 0.65x + 10.4$	0.577
Eq. (9)	$y = 0.64x + 20.5$	0.619
IMPROVE	$y = 0.76x + 11.0$	0.562
Ambient aerosol		
Eq. (8)	$y = 0.96x - 27.9$	0.926
Eq. (9)	$y = 0.97x - 16.7$	0.935
IMPROVE	$y = 0.24x + 53.7$	0.779

Note: x and y are the extinction coefficient from VIS and the reconstructed (by models) extinction coefficients, respectively.

To reconstruct the ambient extinction coefficient, the hygroscopic growth of the extinction coefficient obtained from visibility data was calculated as:

$$\sigma_{ambient} = f \cdot \sigma_{dry} \quad (10)$$

Then, the inorganic (sulfate and nitrate salts (Eq.8) or ions (Eq.9)) and organic (OC) parts in our models were multiplied by these growth factors (*f*). The adequacy of the models for the reconstruction of ambient extinction coefficients is shown in *Table 3*. In each season, the adequate hygroscopic growth rates were considered. We can conclude that both approaches are suitable for the reconstruction of the extinction coefficient; however, they slightly underestimate the ambient extinction coefficient when compared to visibility data.

Figs. 7a and *7b* show the agreement of ambient extinction in more detail. We can conclude that the temporal variation in the ambient extinction coefficient is well represented by both models, but the actual extent of σ is generally underestimated by the model equations.

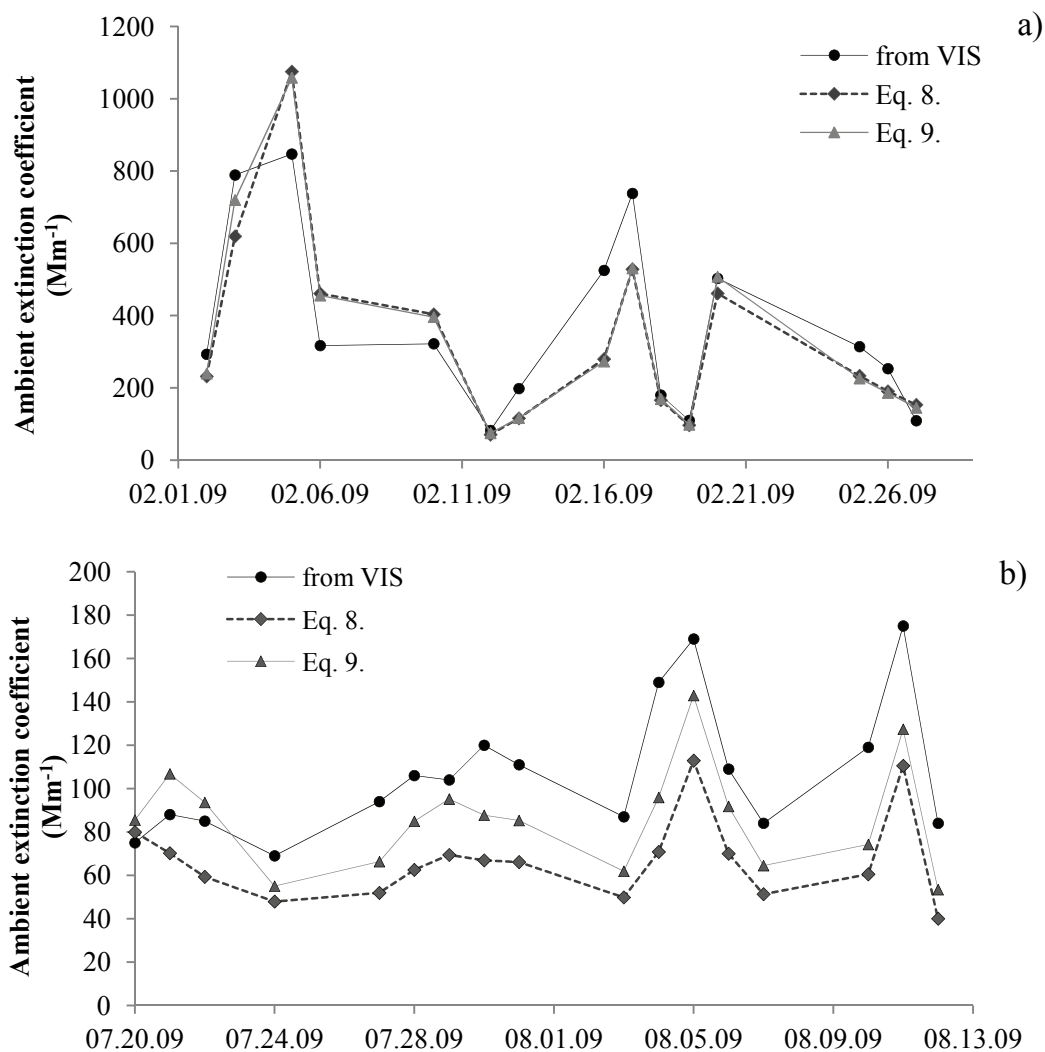


Fig. 7. Temporal variation in the ambient extinction coefficients.

Finally, we compared the three model results obtained for ambient air. We found that the results substantially differ if IMPROVE or our models are applied (see *Table 3*). The main reason for the discrepancy between modeled dry and ambient extinction coefficients can arise from the different consideration of aerosol hygroscopicity. In the case of IMPROVE, model hygroscopic growth is supposed only for ammonium sulfate and sodium nitrate content, while in our models the hygroscopicity of organic species is also involved. It supported by the well-known fact that an important part of organic compounds is hydrophilic, even water soluble. As a result, our models provide more realistic estimation of the ambient extinction coefficient compared to the IMPROVE model. This draws the attention to the significance of aerosol hygroscopicity, concerning both the inorganic and organic contents of the particles. The application of proper hygroscopic growth rate (which is a function of the chemical composition; and has significant seasonality) plays very important role in the reconstruction of ambient extinction coefficient.

4. Conclusions

Based on our results, we can conclude that visibility (extinction) data provide an efficient and inexpensive tool for the survey of long-term variations in air quality and may be utilized as a proxy for PM₁₀ concentration. We have found that visibility data can generally be used to estimate atmospheric aerosol extinction coefficients, even in a retrospective manner. The closure experiment has shown that PM₁₀ can be successfully estimated by chemical composition and that the PM₁₀ concentration can be estimated from visibility/extinction coefficient data (using the Koschmieder theory). Our results indicate that the PM₁₀ concentration (measured or modeled from the chemical composition) provides a good basis for the reconstruction of aerosol extinction coefficients. It is also shown that the derived (from VIS) and reconstructed (from PM₁₀ or the aerosol chemical composition) aerosol extinction coefficients are in good accordance with each other, mainly in the case of dry aerosols. Ambient values can be estimated if the adequate hygroscopic growth rate for the aerosol extinction is considered. We have also found that a rather precise estimation of extinction coefficient can be reached if a modified version of the widely used IMPROVE formula is applied.

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