Volcanic Ash Cloud Observation using Ground-based Ka-band Radar and Near-Infrared Lidar Ceilometer during the Eyjafjallajökull eruption

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Abstract

Active remote sensing techniques can probe volcanic ash plumes, but their sensitivity at a given distance depends upon the sensor transmitted power, wavelength and polarization capability. Building on a previous numerical study at centimeter wavelength, this work aims at i) simulating the distal ash particles polarimetric response of millimeter-wave radar and multi-wavelength optical lidar; ii) developing and applying a model-based statistical retrieval scheme using a multi-sensor approach. The microphysical-electromagnetic forward model of volcanic ash particle distribution, previously set up at microwaves, is extended to include non-spherical particle shapes, vesicular composition, silicate content and orientation phenomena for both millimeter and optical bands. Monte Carlo generation of radar and lidar signatures are driven by random variability of volcanic particle main parameters, using constraints from available data and experimental evidences. The considered case study is related to the ground-based observation of the Eyjafjallajökull (Iceland) volcanic ash plume on May 15, 2010, carried out by the Atmospheric Research Station at Mace Head (Ireland) with a 35-GHz Ka-band Doppler cloud radar and a 1064-nm ceilometer lidar. The detection and estimation of ash layer presence and composition is carried out using a Bayesian approach, which is trained by the Monte Carlo model-based dataset. Retrieval results are corroborated exploiting auxiliary data such as those from a ground-based microwave radiometer also positioned at Mace Head.

I. Introduction

Volcanoes are among most important natural sources of ash, which may influence meteo-climatological conditions on large scales modifying the Earth radiation budget [Graf et al., 2007]. Continuous monitoring of such phenomena is crucial for the initialization of ash dispersion models [Degruyter and Bonadonna, 2012]. Satellite visible-infrared radiometric observations from geostationary platforms are usually exploited for long-range

trajectory tracking and for measuring low-level eruptions [e.g., Rose et al., 2000; Corradini et al., 2011]. Ground-based microwave radars represent an important tool to detect ash clouds [Harris and Rose, 1983; Lacasse et al., 2004; Marzano et al., 2006; Gouhier and Donnadieu, 2008; Schneider and Hoblitt, 2009; Marzano et al., 2010]. The possibility of monitoring in all weather conditions at a fairly high spacetime resolution is the major advantage of using ground-based scanning weather radar systems at S, C and X band [Marzano et al., 2013]. On the other

hand, Ka band Doppler radar can provide a higher sensitivity to medium size particles [Madonna et al., 2010].

The physical-chemical properties of volcanic particles are modified during advection, and size sorting takes place due to aggregation, breaking and fallout [Sparks et al., 1997]. Multi-wavelengths lidars can be complementary systems useful to integrate the micron-sized particle measurement, especially if far from the volcanic vent [e.g., Ansmann et al., 1992]. Lidar techniques developed for aerosol particle detection and estimation can be properly adapted for the retrieval of ash clouds [Gesteiger et al., 2011; Martucci et al., 2012; Scollo et al., 2012].

Previous methodological studies investigated the possibility of using ground-based radar systems for the quantitative remote sensing of volcanic ash cloud [Marzano et al., 2006; 2010]. A volcanic ash radar retrieval (VARR) algorithm for single- and dual-polarization radar systems was proposed and applied to S-, C- and X- band weather radar data volumes [Marzano et al., 2013; Montopoli et al., 2014]. This forward and inverse model framework can be extended to include very fine particles and to ingest both Ka band radars and multi-wavelength lidars.

The paper is organized as follows. Sect. II illustrates the ground-based observations carried out at the Atmospheric Research Station (Ireland) on May 15, 2010, concerning the ash plume emitted by the Eyjafjallajökull volcano. Sect. III summarizes the results and discusses the outlooks. The Annex deals with ash microphysics, scattering and extinction models for microwave and near-infrared wavelengths.

II. VACR COMBINED RETRIEVAL CASE STUDY

In order to fully exploit the multi-sensor multi-wavelength polarimetric forward scattering model for ash cloud remote sensing purposes, we can apply a Bayesian metrics to combined radar-lidar data [Marzano et al., 2013]. This approach will be briefly illustrated before discussing the Eyjafjallajökull eruption case study with measurements taken from the Mace Head site on May 15, 2010.

The Mace Head Atmospheric Research Station supersite over the west of Ireland in Carna, County Galway, is an example of integrated measuring site [Martucci et al., 2012]. Ground-based remote sensing of cloud microphysics is performed using Ka-band Doppler cloud radar (MIRA36), a Jenoptik CHM15K lidar-ceilometer at 1024 nm, and a RPG-HATPRO multi-channel microwave radiometer combined with the synergistic analysis scheme. The radar MIRA36 radar is a monostatic magnetron-based pulsed Ka-Band Doppler radar. Linearly polarized signal at 35.5 GHz is transmitted, while co- and cross polarized signals are received simultaneously to detect Doppler spectra of the reflectivity and linear depolarization ratio. The radar is also equipped with a 3-D scanning unit even though it is usually zenith pointing with a vertical resolution up to 15 m. Note that CHM15K data are available only up to 8000 m and its sensitivity is lower than multi-wavelength research lidar [e.g., Madonna et al., 2010]. However, our purpose is to show the potential and flexibility of the combined inversion methodology.

II.A Combined radar-lidar retrieval algorithm

Similarly to the volcanic ash radar retrieval (VARR) approach [Marzano et al., 2006], the Volcanic Ash Combined Retrieval (VACR) utilizes 2 steps: i) ash classification; ii) ash parameter estimation. Both steps are trained by the HAPESS forward polarimetric model, where particle distributions, density, and permittivity parameters are supposed to be constrained random variables within a Monte Carlo approach (see Annex for details and symbols).

Within the VACR technique, ash classification is performed by means of *M*aximum *A P*osteriori Probability (MAP) estimation criterion. The probability density function (PDF) of each ash class (*ci*), condi-

tioned to the measurement vector \mathbf{x}_m , can be expressed by Bayes' theorem (Marzano et al., 2006). The MAP estimation of ash class c_i , corresponds to the maximization with respect to c of the posterior PDF $p(c_i | \mathbf{x}_m)$. Under the assumption of multivariate Gaussian PDFs, the previous maximization reduces to minimizing a quadratic *distance* $d(\mathbf{x}, c_i)$ with respect to c_i :

$$d(\mathbf{x}_m, e_i) = \begin{cases} (\mathbf{x}_m - \mathbf{m}_{xi})^T \mathbf{C}_{xi}^{-1} (\mathbf{x}_m - \mathbf{m}_{xi}) + \\ + \ln[\det(\mathbf{C}_{xi})] - 2\ln[p(e_i)] \end{cases}$$
(1)

where "T" is the matrix transpose, whereas m_{xi} and C_{xi} are the mean vector and covariance matrix, respectively, of the combined simulated vector x of the class c_i . The a priori probability $p(c_i)$ can be used to weight the different classes on the basis of ancillary information and/or data. Using HAPESS dataset, a regressive model can be used as a function of the class c to estimate both ash concentration C_p and numberweighted mean diameter D_{np} of the class c_i . Using a log-linearized parametric model, it holds:

$$\begin{cases} \ln C_p^{(c)} = m_{\ln c}^{(c)} + \mathbf{C}_{\ln x c} \mathbf{C}_{\ln x c}^{-1} (\ln \mathbf{x}_m - \mathbf{m}_{\ln x}^{(c)}) \\ \ln D_{np}^{(c)} = m_{\ln D}^{(c)} + \mathbf{C}_{\ln x D} \mathbf{C}_{\ln x c}^{-1} (\ln \mathbf{x}_m - \mathbf{m}_{\ln x}^{(c)}) \end{cases}$$

where m_{InC} and m_{InD} are the log-value averages of C_p and D_{np} , C_{InxC} and C_{InxX} are the log-value cross-covariance and auto-covariance matrices, whereas lnx_m and m_{Inx} are log-value measurement vector and its mean vector. In (2) the regression coefficient matrices are obtained assuming a zero-mean random noise due to instrumental and forward modeling uncertainties. As already done elsewhere for VARR [Marzano et al., 2010, 2012], a numerical analysis of the VACR retrieval uncertainty has been carried out using a synthetic simulated dataset in presence of noise. Using 4 combined observables (as later on in

sect. 4.2), results show that: i) the particle correct classification, using (1), has an average probability higher than 95%; ii) the regression root mean square error, using (2), is less than 25% with respect to the same parameter mean value. Of course, these scores degrade if the noise increases due to instrumental causes or data pre-processing errors (e.g., path attenuation correction).

II.B Available data and VACR results

The Eyjafjöll stratovolcano is located under the Eyjafjallajökull small glacier within the Icelandic east volcanic zone. The eruptions in 2010 lasted several weeks [Guðmundsson et al., 2010; Marzano et al., 2010]. Geostationary satellite observations from SEVIRI, combined with other sources, indicate that the Eyjafjallajökull volcanic cloud covered much of Ireland on May 15, 2010 [Guðmundsson et al., 2012]. The ash cloud was indeed detected at the Mace Head site as documented in [O'Dowd et al., 2012].

In order to apply the VACR algorithm, we have considered ground-based observations of the Eyjafjalla-jökull volcanic ash plume on May 15, 2010 at Mace Head, carried out with the MIRA36 Doppler cloud radar and CHM15K lidar ceilometer. The Mace Head combined radar-lidar system provided both measured Z_{hhm} and L_{drm} at Ka-band from MIRA36 and β_{hhm} at NIR from CHM15K. Note that, being the lidar ceilometer system at single wavelength, α_{hh} is derived after inverting measured β_{hhm} data through an inversion algorithm assuming a proper lidar ratio at λ =1064 nm [Ferguson and Stephens, 1983; Martucci et al., 2012].

In order to apply VACR to Mace Head data, the combined measurement vector is set to $x_m = [Z_{hhm}(36\text{GHz}) \quad L_{drm}(36\text{GHz}) \quad \beta_{hhm}(1024\mu\text{m})]^T$ which can be reduced to radar-only and lidar-only special cases (see Annex). Based on a priori information, the number of particle classes can be optimized by merging sub-classes and selecting

two low concentration classes (VC and SC) and three orientations (TO.2, OO, and PO). Being far from the volcano vent, we do not expect lapilli and the coarse particles over Ireland, so that only 2 dispersed classes (VA, FA) are considered here. In order to consider clouds located above the freezing level, we have assumed a mixed-phased refractive index with a balanced mixture of ash and ice (VAm and FAm). Finally, in a cold region, pure ice crystals (IC) and dry snow (DS) may be expected and their microphysical modeling are derived from Marzano et al. (2007). Spherical ash particles are also included for VA, FA, CA considering both ash and mixed-phase particles. This implies that the number of VACR classes has been set to 18.

The Eyjafjallajökull volcanic ash plume over Mace Head, discussed here, is related to the observation of May 15, 2010 from 20:00 till 24:00 UTC. Measurements of MIRA36 and CHM15K have been aligned in time and spatially averaged in order to deal with the different sensor specifications. Their time series have been resampled to every 30 s, whereas in the vertical direction both measurements have been averaged at 30-m resolution.

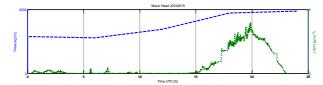


Figure 1: Time evolution of liquid water path (in kg/m²), derived from HATPRO microwave radiometer, and freezing level during May 15, 2010.

The discrimination of ash clouds with respect to water clouds is still an open issue [Martucci et al., 2012]. The detection of a water cloud is aided by the liquid water path (LWP) estimate derived from collocated HATPRO microwave radiometer measurements. Fig. 1 shows the time series of LWP, together with

the the freezing level height estimated from the temperature profile derived from HATPRO. Note that LWP is increasing up to 4 kg/m² from 15:00 till 22:00 being almost negligible after then, while the freezing level is below 2 km. Colocated radar-lidar measurements are shown in Fig. 2 in terms of 24-hour profile time series of copular reflectivity Z_{hh} and L_{dr} from MIRA36 and backscatter coefficient β_{hh} from CHM15K.

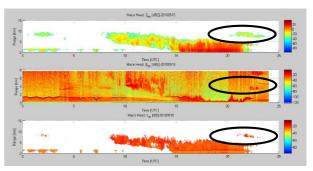


Figure 2: Range-time section of copolar reflectivity $Z_{hh}(36GHz)$ and linear depolarization ratio $L_{dr}(36GHz)$ from MIRA36 (upper and lower panel) and copolar backscatter coefficient $\beta_{hh}(1064nm)$ from CHM15K (middle panel) on May 15, 2010. The black circles indicate the ash signature.

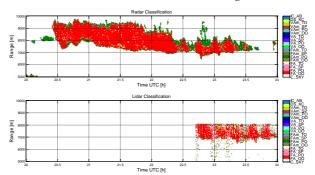


Figure 3: (Upper panel) Classification of ash cloud observed at Mace Head May 15 2010, from 20:00 to 24:00 UTC, using only the radar observables (Zhih and Ldr). (Lower panel) Same as upper panel, but using both radar and lidar observables (Zhih, Ldr, βhih, ακhh), when available. Note lidar-ceilometer data are not available above 8000 m.

The signature before 20:00 below 5 km can be attributed to an incoming altostratus which is followed

by a stratus clouds with some rainfall. Between 20:00 and 22:00 the lower stratus is coexisting with a high cloud, as showed by the radar above 6 km, which is not detected by CHM15K due to the stratus extinction. After 22:00, when the LWP gets almost negligible due to the dissipation of the lower stratus (see Fig. 1), the lidar-ceilometer signal shows a peak at some 8 km. Correspondingly, the radar signal shows a feature around 8 km, a bit weaker than before 22:00. In summary, the ash cloud lidar-radar combined signature can be clearly detected around 6.5-10 km between 20:00 and 24:00, as indicated by black ellipses in Fig.2.

Results from the two-step VACR are presented in Fig. 3, where the VACR classification is obtained through (1) by using radar and radar-lidar observables. Using (2), Fig. 4 shows the VACR estimate of ash concentration by employing only radar observables, only lidar observables, and combined radar-lidar observables.

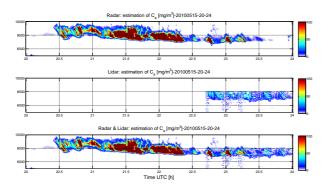


Figure 4: Estimation of ash concentration in the zoomed ash cloud region in Fig. 3, using radar (upper panel), lidar-ceilometer (middle panel) and combined lidar-radar observables, when available (lower panel).

Fig. 3 and 4 indicate that the prevailing ash classes are fine ash FA-OO (about 77%) with FAm-PO (16%) and some FA-PO, FA-TO, FA-SP (less than 4%) and FAm-OO, FAm-TO, VAm-TO (less than 3%). This means that oblate ash particles are coexisting with

heterogeneous nucleation of ice crystals. Lidar signatures are sensitive to micron-sized particles and reveal the presence of mixed very fine ash VAm which are mainly detected at the upper edges of the ash cloud (notably after 22 UTC in Fig. 4).

From Fig. 4 it is interesting to note that estimated ash concentration in the middle of ash cloud can reach values of 100 mg/m³ and its vertical profile is far from being uniform. By vertically integrating the VACR-based estimates of ash concentration of Fig. 4, we can derive the ash cloud columnar content shown in Fig. 5. Similarly to Fig. 4, Fig. 6 shows the mean diameter of ash particles retrieved by VACR using (2). The combination of radar and lidar data extends the capability of each to detect ash particles between 20:00 and 24:00.



Figure 5: Columnar ash concentration derived from VACR algorithm by vertically integrating ash concentration derived from Fig. 4.

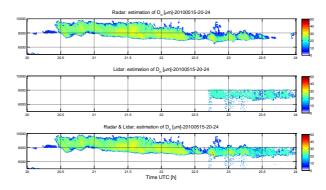


Figure 6: As in Fig. 4, but for particle mean diameter D_n .

Estimated effective diameters goes from 5 microns at the ash cloud edge and in regions where the cloud layer is thin (mainly detected by lidar) up to 40 microns in the cloud inner core (mainly detected by Kaband radar).

III. CONCLUSIONS

Microwave radars and multi-wavelength lidars are complementary instruments, providing a complementary view with respect to the satellite segment. This work has shown how dual-polarization groundbased weather radars and lidars can be merged for volcanic ash cloud dynamical monitoring and quantitative retrieval of ash category, concentration, and effective diameter. The expected accuracy of VACR algorithm is conditioned by the microphysical assumptions. The Eyjafjallajökull volcanic ash plume over the Mace Head site on May 15, 2010 has been used for testing the VACR methodology using a Kaband radar and NIR lidar-ceilometer. Results confirm the potential of the combined approach highlighting interesting features of the retrieved ash cloud in terms of concentration and mean diameters. Future work shall be focused on in situ data for a systematic characterization of the VACR absolute error estimates.

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REFERENCES

[Ansmann et al. 2012] Ansmann, A., et al.. Combined Raman elastic-backscatter lidar for vertical profiling of moisture, aerosol extinction, backscatter and lidar ratio, Appl. Phys. B., 55, 18–28.

[Corradini et al., 2011] Corradini, S., et al. (2011). Volcanic Ash Cloud Properties: Comparison Between MODIS Satellite Retrievals and FALL3D Transport Model. IEEE GRSL, 8, 248-252. [Degruyter and Bonadonna, 2012] Degruyter, W., and Bonadonna, C. (2012). Improving on mass flow rate estimates of volcanic eruptions, Geophysical Research Letters, 39, L16308.

[Gasteiger et al., 2011] Gasteiger, J., et al. (2011). Volcanic ash from Iceland over Munich: mass concentration retrieved from ground-based remote sensing measurements, ACP, 11, 2209-2223.

[Gouhier and Donnadieu, 2008] Gouhier, M., et al. (2008). Mass estimations of ejecta from Strombolian explosions by inversion of Doppler radar measurements. JGR, vol. 113, pp. B10 202–B10 219.

[Graf et al., 2007] Graf, H. F., et al. (2007). Volcanic effects on climate: revisiting the mechanisms, Atm. Chem. Phys., 7, 4503–4511.

[Guðmundsson et al., 2010] Guðmundsson, M. T., et al. (2010). Eruptions of Eyjafjallajökull Volcano, Iceland. EOS 91, 190–191.

[Harris and Rose, 1993] Harris, D. M, and Rose W. I. (1983). Estimating particle sizes, concentrations, and total mass of ash in volcanic clouds using weather radar, J. Geophys. Res., 88, C15, pp. 10969–10983.

[Lacasse et al., 2004] Lacasse, C., et al. (2004). Weather radar observations of the Hekla 2000 eruption cloud, Iceland, Bull. Vol., 66, 5, 457–473.

[Madonna et al., 2010] Madonna, F., et al. (2010). Observation of nonspherical ultragiant aerosol using a microwave radar, Geophys. Res. Lett., 37, L21814, doi:10.1029/2010GL044999.

[Marticci et al. 2012] Martucci, G., et al. (2012). Impact of volcanic ash plume aerosol on cloud microphysics. Atm. Environment, 48, 205-218.

[Marzano et al. 2006] Marzano, F. S., et al. (2006). Volcanic ash cloud retrieval by ground-based microwave weather radar. IEEE TGRS, 44, 3235-3246.

[Marzano et al. 2010] Marzano, F. S., et al. (2010). Monitoring sub-glacial volcanic eruption using C-band radar imagery. IEEE TGRS, 48, 1, 403-414.

[Montopoli et al. 2014] Montopoli, M., et al. (2014) Interpretation of observed microwave signatures from ground dual polarization radar and space multi-frequency radiometer for the 2011 Grímsvötn volcanic eruption (2014), Atm. Meas. Techniques, 7 (2), pp. 537-552. doi: 10.5194/amt-7-537-2014.

[Rose et al. 2000] Rose, W. I., et al. (2000). Integrating retrievals of volcanic cloud characteristics from satellite remote sensors: A summary, Phil. Trans. R. Soc. A, vol. 358, 1770, 1585–1606.

[Schneider and Hoblitt, 2012] Schneider, D. J., Hoblitt, R. P. (2012). Doppler weather radar observations of the 2009 eruption of Redoubt Volcano, Alaska, J. of Vol. and Geoth. Res. v. 259, p. 133-144. [Scollo et al. 2012] Scollo, S., al. (2012). Monitoring Etna volcanic plumes using a scanning LiDAR. Bull Volcanol, DOI 10.1007/s00445-012-0669-y.

[Sparks et al. 1997] Sparks, R. S. J., et al. (1997). Volcanic plumes, Wiley, New York (USA).

ANNEX

VOLCANIC ASH CLOUD OBSERVATION USING GROUND-BASED KA-BAND RADAR AND NEAR-INFRARED LIDAR CEILOMETER DURING THE EYJAFJALLAJÖKULL ERUPTION

Marzano, F. S., et al.

Volcanic ash plumes are formed during explosive volcanic eruptions. After advection over several thousands of kilometers, volcanic ash particles are highly fragmented, dispersed and aged with micronsized sorting. This Annex describes the ash microphysical modeling and the simulated radar and lidar signatures.

A.1. VOLCANIC ASH MICROPHYSICAL MODELING

In this work all particles are assumed to be spheroids, i.e. ellipsoids with circular symmetry. This shape is, on the one hand, quite general and, on the other hand, simplifies the numerical treatment of scattering. Similarly to what described in Marzano et al. (2012), we set up a microphysical model of ash particle distributions. This means to specify: a) particle size distribution (PSD); b) particle ellipsoidal axial ratio (AR); c) particle orientation distribution (POD); d) particle density and dielectric constant. Ash PSDs are typically modeled through a normalized Gamma or Weibull size distribution [e.g., Marzano et al., 2006; 2010]. In case of a multimode size distribution, it is always possible to suppose more than one analytical PSD, characterized by different mean sizes and total number of particles. In this study, we have adopted the scaled Gamma (SG) PSD as a general model for all particles. If D_e is the diameter of a spherical volume-equivalent particle, the scaled-Gamma PSD N_p [mm⁻¹m⁻³] for a generic class of ash particles p can be written as:

$$N_p(D_e) = N_{np} \left(\frac{D_e}{D_{np}} \right)^{\mu_p} e^{-\Lambda_{np} \left(\frac{D_e}{D_{np}} \right)}$$
 (1)

where the "intercept" parameter N_{np} and the "slope" parameter Λ_{np} in a logarithmic plane are related to the "shape" parameter μ_p and to the particle density ρ_p through:

$$N_{np} = 10^6 \frac{6C_p \Lambda_{np}^{\mu_p + 4}}{\pi \rho_p D_{np}^4 \Gamma(\mu_p + 4)}; \quad \Lambda_{np} = \mu_p + 1$$
 (2)

with C_p the mass concentration [g/m³] and D_{np} [mm] the number-weighted mean diameter. PSD in (1) is completely specified by the three parameters μ_p , D_{np} and C_p (with a particle density ρ_p between 1 and 2.5 g/cm³). If particles are sphere equivalent, their mass is $m_p = \rho_p(\pi/6)D_e^3$.

The *number of ash classes* with respect to their average size is set to 5 as follows: i) very fine ash (VA) with mean equivalent diameters uniformly distributed between $2^{-3}-2^{3} \mu m$ (0.125-8 μm); ii) fine ash (FA) between 23-26 µm (8-64 µm); iii) coarse ash (CA) between 26-29 μm (64-512 μm); iv) small lapilli (SL) between $2^9-2^{12} \mu m$ (0.512-4.096 mm); and finally v) large lapilli (LL) between 212-215 µm (4.096-32.768 mm). With respect to previous studies [Marzano et al., 2012], we have added the class of large lapilli and very fine ash to take into account ballistic fallout as well as dispersed small particles in order to deal with a fairly general observation scenario (closer to and farther from the volcanic vent). Moreover, uniform distribution around the mean size has been assumed instead of Gaussian probability.

Each diameter class may be subdivided with respect to other main parameters, e.g. the ash concentration and orientation angle [Marzano et al., 2012]. The following sub-sets have been introduced:

 4 classes of ash concentration, i.e.: very small (VC), small (SC), moderate (MC), and intense (IC) concentrations. These sub-sets are eventually compacted into 1 class uniformly going from 10-6 g/m³ up to 10 g/m³;

- ii) 5 classes of particle orientation with Gaussian POD characterized by a mean canting angle θ_m , i.e.: tumbling with θ_m =30° (TO.1), tumbling with θ_m =45° (TO.2), tumbling with θ_m =60° (TO.3), oblate θ_m =0° (OO); prolate θ_m =90° (PO). These 5 sub-sets are reducible to 3;
- iii) 2 classes of AR models, i.e.: ratio basalticandesitic (RB) and ratio rhyolitic (RR), even though we have here selected only the RB case;
- iv) 3 classes of VA, FA, CA spherical particles (SP) with all 5 concentrations.

Particle dielectric constant models have been derived from literature; in particular, their dependence on SiO₂ weight WsiO₂ and relative humidity fraction (RH) is derived from the available data [e.g., Adams et al., 1996; Shettle and Fenn, 1971]. Ash classes can exhibit a mixed-phased permittivity due to liquid and ice water in the atmosphere. Particle classes with a mixed refractive index have been also introduced [Marzano et al., 2010]. In such a case, we have limited our attention to particles smaller than lapilli, so that 3 new mixed-phase classes have been labeled by VAm, FAm, CAm for all 5 concentrations. Liquid and ice water particle classes (e.g., cloud droplets, ice crystals, snowflakes) can be added to generalize the scenario. Considering all combinations, the nominal number of classes is 233 (5·4·5·2+15+15+3) which can be effectively reduced to 20 (5·1·3·1+3+2), as we will discuss later on in this work.

A.2. VOLCANIC ASH PARTICLE SCATTERING AND EXTINCTION

The conditions allowing for the Rayleigh scattering regime, depending on particle refraction index and the ratio between the particle diameter and the incident wavelength λ , are not always satisfied by the characteristics of the problem under discussion. Simulated polarimetric radar observables for spheroidal particles can be obtained by using the T-matrix numerical method [Mishchenko et al., 1996], implemented through the

Hydrometeor Ash Particle Ensemble Scattering Simulator (HAPESS) [Marzano et al., 2012].

In order to simulate the optical polarimetric signatures of volcanic particles we have included within HAPESS the geometrical optical limits which cannot be evaluated by T-matrix code due to numerical convergence problems [Min et al, 2003; Gasteiger et al., 2011]. In the next sub-sections we will introduce observables related to backscattering amplitude since those related to forward scattering phase are beyond the scopes of this paper [see Marzano et al., 2012].

Polarimetric radar observables

Using horizontal (h) and vertical (v) polarization states, copolar (Z_{hh} , Z_v) and crosspolar (Z_{vh} , Z_{vh}) radar reflectivity factors can be expressed in terms of ash PSD. If λ is the wavelength associated to a given frequency f, K_p is the ash particle radar dielectric factor and θ is the particle canting angle, S_{xy} are the backscattering components at x (x=h,v for the receiving mode) and y (y=h,v for the transmitting mode) polarization of the complex scattering matrix S, we can define:

$$Z_{xy}(f) = \frac{\lambda^4}{\pi^5 |K_p|^2} \int_0^{\pi} \int_0^{\infty} 8\pi^2 |S_{xy}^{(b)}(D_e, \phi, f)|^2 N_p(D_e)$$
$$p_p(\theta) dD_e \sin \theta d\theta = \frac{\lambda^4}{\pi^5 |K_p|^2} \langle 4\pi S_{xy}^{(b)}(D_e, \theta) \rangle$$
(3)

In (3) angle brackets stands for average over PSD N_p and POD p_p , assuming a uniform azimuthally symmetric POD [Marzano et al, 2012]. Note that: i) $Z_{hv}=Z_{vh}$ for cross-polarization reciprocity; ii) the canting angle θ is defined in the polarization plane with respect to its vertical polarization unit-vector; iii) if Z_{xy} is expressed in [mm⁶m⁻³], then $Z_{xy}=10\log_{10}(Z_{xy})$ is conventionally expressed in dBZ.

The derived differential reflectivity Z_{dr} , which is the ratio of reflectivity at the two polarization states, and the linear depolarization ratio L_{dr} are defined as:

$$Z_{dr}(f) = \frac{Z_{hh}(f)}{Z_{vv}(f)}; \quad L_{dr}(f) = \frac{Z_{hv}(f)}{Z_{hh}(f)}$$
 (4)

where Z_{dr} and L_{dr} are unitless and usually expressed in dB. The specific power attenuation α_{th} (α_{vv}) at polarization h (v) can be obtained from:

$$\alpha_{xy}(f) = 2\lambda \operatorname{Im} \left[\langle 4\pi S_{xy}^{(f)}(D_e, \phi, f) \rangle \right]$$
 (5)

The apex "f" of S indicates its forward components. Note that, if α_{xy} is in [km⁻¹], the above quantities are usually defined in dB within the radar community $\alpha_{XY}=4.343\alpha_{xy}$.

Note that, if the ellipsoidal axial ratio shows an intrinsic variability, the particle non-sphericity can significantly affect the differential reflectivity Z_{dr} . The L_{dr} is not dependent on the particle concentration and it is governed by the inverse of reflectivity-weighted axial ratio within the Rayleigh scattering. The L_{dr} signature will then increase with increasing canting angle and axial ratio variability.

Polarimetric lidar observables

The lidar backscattering coefficients β_{hh} , β_{vv} and β_{vh} at horizontal (h) and vertical (v) polarization states can be expressed similarly to the corresponding radar coefficients. Indeed, from (4) their definition can be written as:

$$\beta_{xy}(\lambda) = \int_0^{\pi} \int_0^{\infty} 4\pi \left| S_{xy}^{(b)}(D_e, \phi, \lambda) \right|^2 N_p(D_e)$$

$$p_p(\phi) dD_e \sin \phi d\phi = \langle 4\pi S_{xy}^{(b)}(D_e, \phi) \rangle$$
(6)

where x=h,v again stands for the receiving mode and y=h,v for the transmitting mode polarization. Note that β_{xy} is usually expressed in [km⁻¹sr⁻¹] so that $\beta_{xy}=10\log_{10}(\beta_{xy})$ is here conventionally expressed in dB β (in analogy to dBZ for Z_{xy}).

The same applies to the specific attenuation or extinction coefficient α_{xy} , which similarly to (5) is expressed in [km⁻¹] and is defined as:

$$\alpha_{XY}(\lambda) = 2\lambda \operatorname{Im} \left\{ 4\pi S_{XY}^{(f)}(D_e, \phi, \lambda) > \right\}$$
 (7)

Analogously to (7), if α_{xy} is in [km⁻¹], α_{xy} =4.343 α_{xy} is conventionally expressed in dB. Similarly to Z_{dr} and L_{dr} in (4), the unitless lidar linear co-polarization ratio δ_{co} and cross-polarization ratio δ_{cr} can be defined by:

$$\delta_{co}(\lambda) = \frac{\beta_{vv}(\lambda) - \beta_{hh}(\lambda)}{\beta_{vv}(\lambda) + \beta_{hh}(\lambda)} ; \delta_{cr}(\lambda) = \frac{\beta_{vh}(\lambda)}{\beta_{hh}(\lambda)}$$
(8)

Radar and lidar polarimetric signatures

In order to discuss combined polarimetric signatures, we have considered existing radar and lidar instrument setups, e.g. zenith-pointing Kaband radars and colocated lidars operating at 355 nm, 532 nm and 1024 nm wavelength [e.g., Madonna et al., 2010]. HAPESS outputs have been validated for volcanic ash with numerical simulations available in literature [e.g. Wiegner et al., 2009].

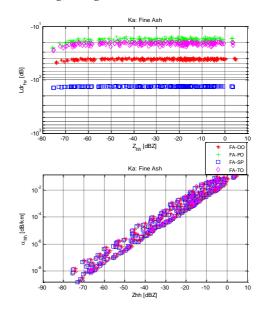


Figure 1a: Correlation between zenith Ka-band polarimetric radar observables in terms of copolar reflectivity Zhh (36GHz), linear depolarization ratio Ldr (36GHz), and specific attenuation αhh (36GHz) for FA size class, using all ash concentrations classes (VC, SC, MC and IC) and for different particle orientations (OO, PO, SP and TO.2) with basaltic axial ratio model.

This section illustrates examples of ash sub-classes, limited here for brevity to FA signatures in terms of Ka-band and near infrared (NIR) polarimetric observables (which are used for the combined system introduced later on). Figure 1a shows how Zhh(36GHz) correlates to linear depolarization $L_{dr}(36GHz)$ and specific power attenuation ahh(36GHz). Note that the particle orientation has an impact on L_{dr}(36GHz), with the OO (mainly horizontal with respect to the ground) providing values of about -40 dB. Specific attenuation is pretty low for FA (lower than 0.15 dB/km), but larger of 1 dB/km for larger particles. In figure 1a we have also added the behavior of spherical particles (SP) where we expect theoretically Z_{hv} =0 and L_{dr} =0.

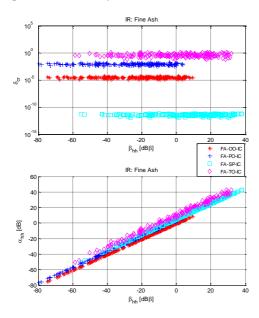


Figure 1b: Same as in the Figure 1a panels, but for the zenith lidar copolar backscatter coefficient β_{hh} (1064nm), linear crosspolarization ratio δ_{cr} (1064nm), and extinction coefficient α_{hh} (1064nm) with relative humidity ranging between 30 and 70%.

In Figure 1b panels show the correlation between zenith-pointing NIR backscatter coefficient $\beta_{hh}(\lambda)$, linear cross-polarization ratio $\delta_{cr}(\lambda)$, and extinction coefficient $a_{hh}(\lambda)$ for fine ash. Note that $\beta_{hh}(1064$ nm)

ranges from -80 to -10 dB β for FA-PO and -40 to 40 dB β for FA-SP, whereas the sensitivity to orientation is no more evident for the larger particles. The extinction α_{hh} (1064nm) ranges from -80 to 40 dB/km for FA, whereas α_{hh} (1064nm) from -110 to -10dB/km for larger particles. It is worth noting that if $\beta_{hh}(\lambda)$ increases with increasing ash concentration C_p , it decreases with increasing particle mean diameter D_n , a behavior which is explained by the optical limit regime of scattering taking place at NIR for particles larger than few microns [Wiegner et al., 2009; Min et al., 2003]. On the opposite, the radar reflectivity $Z_{hh}(f)$ increases with the increase of both C_p and D_n .

REFERENCES

[Adam et al. 1996] Adams, R., et al. (1996). Measurements of the complex dielectric constant of volcanic ash from 4 to 19 GHz. J. Geophys. Res., vol. 101, pp. 8175–8185.

[Ferguson and Stephens 1983] Ferguson, J. A., and Stephens, D. H. (1983). Algorithm for inverting lidar returns. Appl. Opt., 22, 3673-3675.

[Gudmundsson et al. 2012] Gudmundsson, M. T., et al. (2012). Ash generation and distribution from the April-May 2010 eruption of Eyjafjallajo kull, Iceland, NG Scientific report, 2 : 572 | DOI: 10.1038/srep00572.

[Marzano et al. 2012] Marzano, F. S., et al. (2012). Synthetic Signatures of Volcanic Ash Cloud Particles from X-band Dual-Polarization radar, IEEE Trans. Geosci. Rem. Sens., 50, 193-211.

[Marzano et al. 2013] Marzano, F. S., et al. (2013). Inside Volcanic clouds: Remote Sensing of Ash Plumes Using Microwave Weather Radars. Bullettin Am. Met. Soc., 1567-1586, DOI: 10.1175/BAMS-D-11-00160.1.

[Min et al. 2003] Min, M., et al. (2003). Scattering and absorption cross sections for randomly oriented spheroids of arbitrary size, J. Quant. Spectr. Rad. Transfer, 79–80, 939–951.

[Mishchenko et al., 2002] Mishchenko, M. I., et al. (2002). Scattering, absorption and emission of light by small particles, Cambridge University Press, Cambridge (USA).

[O'Dowd et al. 2012] O'Dowd, C., et al. (2012). The Eyjafjallajökull ash plume- Part I: Physical, chemical and optical characteristics., Atmos. Env., 48, 129-142. [Shettle and Fenn 1979] Shettle, E. P., and Fenn R. W. (1979). Models for the Aerosol of the Lower Atmospere and the Effect of Humidity Variation on their Optical Properties. AFGL-TR-79-0214 Environmental research papers no.076.

[Wiegner et al. 2009] Wiegner, M., et al. (2009). Numerical simulations of optical properties of Saharan dust aerosols with emphasis on lidar applications, Tellus, 61B, 180-194.