# RETRIEVAL OF MINERAL ABUNDANCES OF THE DELTA REGION IN EBERSWALDE, MARS

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#### ABSTRACT:

Eberswalde Crater, a hotspot of Mars exploration, possesses an unambiguous hydrological system. However, little research has been performed on the large-scale mineral abundances retrieval in this region. Hence, we employed hyperspectral unmixing technology to quantitatively retrieve mineral abundances of the delta region in Eberswalde. In this paper, the single-scattering albedos were calculated by the Hapke bidirectional reflectance function from Compact Reconnaissance Imaging Spectrometer for Mars (CRISM) data (FRT000060DD) and CRISM spectral library respectively, and a sparse unmixing algorithm was adopted to quantitatively retrieve mineral abundances. The abundance maps show that there are six kinds of minerals (pyroxene, olivine, plagioclase, siderite, diaspore, and tremolite). By comparing minerals spectra obtained from images with corresponding spectra in spectral library, we found the similar trend in both curves. Besides, the mineral abundance maps derived in this study agree well spatially with CRISM parameter maps. From the perspective of mineralogy, the instability of pyroxene and olivine indicates the area in which they distribute is close to provenance, and the original provenance is ultrabasic rock(e.g. peridotite) and basic rock(e.g. gabbro), respectively. And minerals, existing in the area of alluvial fan, also distribute in the outside of alluvial fan, which might be caused by fluid transportation.

## 1. INTRODUCTION

The fan delta in Eberswalde crater has been interpreted as strong evidence for sustained liquid water on early Mars(Malin and Edgett 2003; Mangold et al., 2012; Pondrelli et al., 2014). Previous study mainly focus on the morphology, stratigraphy, and geology evolution of this place(Lewis & Oded 2006; Mangold et al., 2012; Pondrelli et al., 2008). Although these research provide clues of their geology history, mineral identification and abundance retrieval can further provide unique constraints on their formation environment(Liu et al., 2016). Poulet et al (2014) modeled the mineral abundances by average spectrum of Compact Reconnaissance Imaging Spectrometer for Mars (CRISM) 121 pixels over the Eberswalde deltaic deposits. However, little research has been performed on the large-scale minerals retrieval in this region. Spectral unmixing aims at extracting the spectrally pure constituent materials and their respective fractional abundances in each pixel of a hyperspectral image scene(Bioucas-Dias et al., 2012). Hence, we employed a spectral unmixing technology to quantitatively retrieve mineral abundances of the delta region in Eberswalde.

There are several efforts have been made to attempt quantitative mineral abundances based on visible near infrared (VNIR) imaging spectroscopy on Mars. In the VNIR, the observed signal of reflected light from a particulate surface is a non-linear combination of the spectral properties of the minerals (Goudge et al., 2015). The widely used non-linear mixing models in planetary spectra study are Hapke radiative transfer model(Hapke 1981) and the geometric optics model of Shkuratov (Shkuratov et al., 1999). Both the Hapke and Shkuratov models have been used to retrieve mineral abundances in VNIR spectra of Mars(Liu et al., 2016; Poulet et al., 2014).

In this study, we used Hapke radiative transfer model to convert CRISM I/F to single-scattering albedos, so as to perform linear unmixing to Martain minerals spectra. It's challenging to find pure spectral signatures(endmembers) in practice(Iordache et al., 2011). Iordache et al (2011) proposed a sparse unmixing algorithm which sidesteps endmember extraction process, it aims at finding the optimal subset of signatures in a spectral library that can best model each mixed pixel in the scene. Therefore, we adopted a sparse unmixing algorithm to quantitatively retrieve mineral abundances by using image obtained by CRISM together with CRISM spectral library.

### 2. STUDY AREA AND DATA SETS

#### 2.1 Study Area

Eberswalde Crater centered at 24°S, 33°W, is a partially buried impact crater located north–northeast of the Holden crater. Its diameter is about 65.3km. Geomorphology provide unambiguous evidence that water was present and stable here during some part of its history (Pondrelli et al., 2014). It possesses a fan delta, fluvial channels. According to stratigraphy of cross cutting relations, Pondrelli et al (2008) interpreted there are five different lobes, and each lobe corresponds to a different stage of fan delta evolution. It was one of the final four proposed landing sites for the Mars rover Mars Science Laboratory (MSL) mission. Study on minerals and their abundances can further enhance our knowledge on their deposition process.



Figure 1. Eberswalde Crater.(a) Overview of Eberswalde Crater, which is a partially buried impact crater. Image is a mosaic of CTX images, overlain on the THEMIS 100 m/pixel global daytime infrared mosaic. CRISM FRT000060DD footprint is shown outlined in red. (b) CRISM image FRT000060DD (R:2.529µm, G:1.506µm, B: 1.080µm). North is up in both images.

#### 2.2 Data Sets

This paper intent to study minerals of delta region. Therefore we emphasize CRISM hyperspectral data and CRISM spectral library, which help to distinguish minerals and retrieve their abundances.

#### 2.2.1 CRISM data

CRISM is a VNIR imaging spectrometer onboard the Mars Reconnaissance Orbiter (MRO) that covers the wavelength range of 0.36~ $3.94\mu$ m. In this paper, we used one targeted mode observation, which has a full spatial resolution (FRT) of 18 m/pixel, and a spectral sampling of 6.5 nm/channel(Murchie et al., 2007).

In this work, we downloaded the CRISM I/F data from Planetary Data System (PDS). CRISM data processed to I/F through the ratio of the measured sensor radiance and a solar irradiance divided by  $\pi$  (McGuire et al., 2009a). Images were photometrically corrected by dividing each spectrum by the cosine of the incidence angle. Then they were atmospherically corrected using the "Volcano Scan" method(McGuire et al., 2009a). Although Volcano Scan technique does not remove aerosol contributions to spectra completely, it corrects the main atmospheric absorption caused by CO<sub>2</sub> (Mcguire et al., 2009b). We applied filter to remove both spatial and spectra noise in images. Finally, we made geometric correction by projection. All above operations were performed by CRISM Analysis Toolkit. Because minerals primary spectral feature are within 1~2.6µm, and avoid surface thermal emission at wavelengths >2.6µm, hence, the bands befor 1µm and after 2.6µm were not used.

#### 2.2.2 CRISM Spectral Library

The CRISM spectral library is a collection of laboratory spectra of Mars-analog materials supplied by the CRISM Team. The library was used for pure endmembers in spectral unmixing. It contains 2463 spectral analyses of 1228 specimens, all measured under desiccating conditions to remove atmospheric H2O contamination and provide a better proxy to current desiccated Martian conditions(Lin et al., 2015). We chose the naturally occurring minerals and retained valid minerals spectra, 768 samples in total. All spectra were resampled at the wavelength of the CRISM using linear interpolation.

#### 3. METHODS

#### 3.1 Single-Scattering Albedo Retrieval

Single-Scattering Albedo(SSA) is the ratio of scattered light to total extincted light. Hapke bidirectional reflectance distribution function relates the reflectance of a mixture to a linear combination of the SSA of its constituent endmembers(Hapke, 1981). It enables us to perform linear mixture analysis of the observed spectra(Goudge et al., 2015; Lin and Zhang 2017; Liu et al., 2016). The function is described as the following equation(Hapke, 1993):

$$r(i,e,g) = \frac{\omega}{4} \frac{\mu_0}{(\mu_0 + \mu)} [(1 + B(g))P(g) + H(\mu_0)H(\mu) - 1]$$
(1)

$$H(\mu) = \frac{1+2\mu}{1+2\mu\sqrt{1-\omega}}$$
(2)

where r(i,e,g) is the radiance factor, equivalent to CRISM I/F; *i*, *e*, and *g* are incidence, emergence, and phase angles, respectively; *w* is the average single particle scattering albedo; B(g) is the backscattering function at the phase angle g; P(g) is the surface phase function; H is the Chandrasekhar integral function associated with the observation geometry.

In this work, we referred to (Goudge et al., 2015), assumed that there is no opposition effect, B(g)=0; and also assumed that P(g)=1, which implies that the surface scatters isotropically. Because the opposition effect is strongest for phase angles  $<15^{\circ}$ (Mustard & Pieters, 1989), and our data are at phase angles  $\ge 30^{\circ}$ . The effect of anisotropic scattering on abundance estimates for binary and ternary mixtures is about 5–10% when measured at phase angles of 15–120°(Mustard & Pieters, 1989). We can calculate analytic solution for SSA based on equation (1), (2) and B(g)=0, P(g)=1. Therefor, we converted reflectance of CRISM images and spectral library to SSA, respectively.

#### 3.2 Sparse Unmixing Algorithm

Sparse unmixing is an active research area in hyperspectral unmixing in recent years(Giampouras et al., 2015; Iordache et al., 2011, 2012, 2014a; Iordache et al., 2014b; Shi et al., 2014; Tang et al., 2015), which aims to find the optimal subset of signatures in a spectral library that can best model hyperspectral data. A sparsity regularizer is commonly imposed to promote the number of selected signatures as small as possible. This sparse unmixing problem can be writen as equation(3)(Iordache et al., 2014a):

$$\min_{X} \|AX - Y\|_{F} + \lambda \|X\|_{2,1} + \kappa_{R+}(X)$$
(3)

where *Y* is the measured spectra of the pixels (*L* bands, *N* pixels),  $Y \in \mathbb{R}^{L \times N}$ ; *X* is the abundance matrix corresponding to reflectance spectral library *A*(*L* bands, *M* signatures),  $A \in \mathbb{R}^{L \times M}$ ,  $X \in \mathbb{R}^{M \times N}$ ;  $||X||_{2,1} = \sum_{k=1}^{M} ||x^k||_2$  to denote the  $\ell_{2,1}$  norm, this term imposes sparsity among the endmembers simultaneously for all pixels;  $\kappa_{R+}(X) = \sum_{j=1}^{n} \kappa_{R+}(x)$ ,  $x_i$  represents the *i*-th column, this term is a indicator function( $\kappa_{R+}(x_i)$  is zero if  $x_i$ belongs to the nonnegative orthant, otherwise  $\kappa_{R+}(x_i)$  is infinity).

We adopted this algorithm, the main steps in this work are the following ones : (1) To infer the subspace in which the CRISM data lives by using the hyperspectral signal subspace identification by minimum error (HySime) algorithm(Bioucas-Dias & Nascimento, 2008).(2) All the mineral spectra of the spectral library were projected onto the subspace and the

projection errors for each mineral spectra were calculated.(3) The spectra with smaller projection error were then retained as the endmember which truly contributed to the observed mixtures. (4) A sparse unmixing algorithm (Iordache et al., 2014a) was applied to retrieve the minerals abundance of the delta region in Eberswalde.

#### 4. RESULTS

#### 4.1 Spectral Unmixing Results

The abundance maps show that there are five kinds of primary minerals (pyroxene, olivine, plagioclase, siderite, diaspore) and one kind of alteration minerals (tremolite). As shown in Figure 2, pyroxene is mainly distributed in the front of alluvial fan. Olivine is distributed in the central area, shown as east-west distribution. Plagioclase is mainly distributed in the western delta region and a small crater edge in the northeastern corner. Siderite, with relatively less content, is mainly distributed along the edge of alluvial fan. Diaspore is mainly distributed along the river valley and the relatively enriched diaspore in the northeastern corner was probably caused by transportation of liquid water in the past. Tremolite is distributed in the northeastern corner of the delta region.



Figure 2. Abundance maps of minerals(R: Abundance map, G: 1.506µm, B: 1.080µm). The minerals are in red areas.

## 4.2 Validation of Results

We used the ratioing technique to remove residual atmospheric or instrument-related spectral features. Spectra from high abundance areas were extracted and ratioed against areas with spectrally bland material at the same column. By comparing ratioed spectra with corresponding spectra in CRISM library, we found the similar trend in both curves. The features of diaspore(Figure 3c) are not obvious. It's likely that the presence of dust partly masks mineral features. In order to validate the spatial distribution of minerals, scatterplots of CRISM index values(BD1300, LCPINDEX, OLINDEX)(Viviano-Beck et al., 2014) versus derived major mineral abundance were produced to evaluate the correlation between them. Comparisons of spectral index map and abundance map are shown in Figure 4. The mineral abundance maps derived in this study agree well spatially with CRISM parameter maps. Particularly, the pyroxene abundance map nicely matches the index map of LCPINDEX, and the scatterplot also shows strong correlation. There are minor discrepancies between olivine abundance map and OLINDEX map in Figure 4

(red circle areas), a possible explanation is that olivine index may overestimates the distribution of olivine when pyroxene is present(Viviano-Beck et al., 2014).







 $(\mathbf{g})$  Abundance map of olivine

(h) Index map of olivine

(i) Scatterplot of olivine

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Figure 4. Comparison of spectral index map and abundance map. The red circle ares in (g), (h) are minor discrepancies between olivine abundance map and OLINDEX map.

#### 5. CONCLUSION

In this paper, we first used Hapke model to convert CRISM I/F to single-scattering albedos(SSA), then adopted a sparse unmixing algorithm to quantitatively retrieve mineral abundances of the delta region in Eberswalde. Six kinds of minerals have been found, including pyroxene, olivine, plagioclase, siderite, diaspore and tremolite. From the perspective of mineralogy, the instability of pyroxene and olivine indicate the area in which they distribute are close to provenance, and the original provenance are ultrabasic rock(e.g. peridotite) and basic rock(e.g. gabbro), respectively. Tremolite may be formed in contact metamorphism, and the magma provides high temperature containing Si fluid, carbonate surrounding rock provides Ca. Although the unmixing results are encouraging, we have to take such problems into consideration: (1) uncertainty from end-members selection, (2) uncertainty from singlescattering albedo inversion model.

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