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Applying spectral analysis for identification of alteration zones in north Saveh area, Iran

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An extensive series of volcanic rocks are exposed in the north of Saveh city, Iran, which consist of phyllic, argillic and propylitic hydrothermal alteration types. For the purpose of the investigation, a FieldSpec3[®] spectroradiometer was used to measure the spectral response of the mineral content of these rocks. The spectral analyses of reflectance curve by The Spectral Geologist (TSG) software could discriminate kaolinite and montmorillonite (argillic), illite, muscovite, phengite and paragonite (phyllic), hornblende and chlorite, siderite (propylitic), hematite and goethite from the gossans. It also detected gypsum of hydrothermal alteration zones. The Advanced Spaceborne Thermal Emission and Reflectance Radiometer (ASTER) image, which was used for mapping the hydrothermal alteration minerals, contains the Visible and Near Infrared (VNIR) wavelengths between 0.52 μm and 0.86 μm , Short Wave Infrared (SWIR) wavelengths between 1.6 μm and 2.43 μm and Thermal Infrared (TIR) wavelengths between 8.125 μm and 11.65 μm with 15, 30 and 90m spatial resolutions, respectively. For calibration of the ASTER images, the extracted spectra of different rocks and minerals were used for atmospheric and radiometric corrections. Mixture tuned matched filtering (MTMF) and Spectral Angle Mapper (SAM) were applied on ASTER data to map the hydrothermal alteration of minerals. The use of the spectroradiometry techniques in conjunction with other data exhibits the ability of these new methods for non-destructive and rapid identification of mineral types for more detailed investigation. The results show that the area has undergone different levels of hydrothermal alteration, so much so that phyllic, argillic and propylitic types of hydrothermal alteration are present in the study area. This may point to high potential and promising zones for the exploration of porphyry mineralisation.

Keywords: spectroscopy, ASTER image, alteration zones, Saveh-Iran, Urumieh–Dokhtar Arc

Introduction

Identification of the behaviour of the material at various wavelengths using reflective sensing data can be conducive to recognising and distinguishing types of occurrences in remote sensing data.¹ Minerals show

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absorption features in the visible near infrared (VNIR) and shortwave infrared (SWIR) regions, which help to distinguish hydrothermal alteration zones.^{2,3} Infrared spectroscopy is extensively used for the recognition of hydrothermal alteration minerals.^{4,5} Rangzan *et al.*^{6,7} studied the spectral properties of the igneous and metamorphic rocks of the Hamedan region in Iran using a FieldSpec3[®] spectroradiometer. In their work, they studied the spectral behaviour of the alumino-silicate minerals as well as the spectral classification of the rocks containing these minerals. In another study, Pourkaseb *et al.*⁸ used a spectroradiometer and extracted the spectra of the rock units of Firoozabad salt diapir of the Zagros mountain of southeast Iran. Karami *et al.*⁹ created a GIS database and interactive maps with spectral data sharing to show the possibility of the contribution of referenced spectral data in producing geo-information. Many authors have applied remote sensing techniques for mineral exploration.^{10–16}

Azizi *et al.*¹⁷ applied ASTER SWIR data to identify hydrothermal alteration minerals from the Zanjan area, Iran. Honarmand *et al.*¹⁸ applied ASTER/Enhanced Thematic Mapper (ETM)⁺ images for the alteration mapping of Kerman, Iran. Amer¹⁹ used ASTER images and utilised Spectral Information Divergence (SID) and Spectral Angle Mapper (SAM) classification methods to detect altered minerals related to gold mineralisation in Egypt. The significance of the Mixture Tuned Matched Filtering (MTMF) method in enhancing the detection of altered minerals has already been revealed in many studies.^{20,21} Our study area is located in a part of the Urumieh–Dokhtar magmatic arc. It seems that the study area has been tectonically active, as revealed by the numerous faults, folds and fractures (Figure 1). Rock outcrops in the study area are dacitic lava, dacitic tuff, andesite, basaltic andesite, trachyandesite, rhyodacite, granite, diorite, granodiorite, gabbro-diorite, tonalite and quartz diorite, which are exposed in north Saveh, Iran.²² Moreover, the interaction of hot aqueous fluids with the country rocks produce hydrothermal mineral deposits.^{23,24} Lowell and Guilbert²⁵ concluded that potassic, phyllic, argillic and propylitic zones are associated with porphyry copper deposits. Since the same situation may prevail in vast portions of Iran, spectroradiometric techniques were used for rapid investigation in the study area. Spectroradiometer and The Spectral Geologist (TSG) are standard and effective tools for mineral exploration. Additionally, the created spectra are used for various

corrections of satellite images and accuracy assessment of classifications. Also, remote sensing can be used to detect and map hydrothermally altered rocks.

Geological setting

The Saveh area is located in the central part of the Urumieh–Dokhtar magmatic arc between 49° 59' 59" and 50° 36' 29" eastern longitudes and 35° 04' 11" and 35° 15' 54" northern latitudes. As mentioned above, it seems that the study area has been tectonically active, which is revealed by the numerous faults, folds and fractures (Figure 1). Rock outcrops in the study area are dacitic lava, dacitic tuff, andesite, basaltic andesite, trachyandesite, rhyodacite, granite, diorite, granodiorite, gabbro-diorite, tonalite and quartz diorite, which are exposed in north Saveh, Iran. The present-day geotectonic framework of Iran is strongly influenced by two major orogenic events, one in the Early Mid-Mesozoic time reflecting the closure of the Paleo-Tethys Ocean, and the second in the Early Cenozoic to Recent, reflecting the closure of the Neo-Tethys Ocean.^{26,27} The closure of Neo-Tethys and the consequent collisional tectonic regime resulted in the formation of the Zagros Structural Belt.²⁸ It consists of three major tectonic units with an NW–SE trend in Iran, namely: (1) Zagros fold and thrust belt, (2) Sanandaj–Sirjan Metamorphic Zone and (3) Urumieh–Dokhtar magmatic arc (UDMA).²⁹ The UDMA is considered as a tectonically active zone dominated by intermediate-silicic volcanoes as well as plutonic rocks composed mainly of quartz diorite and granodiorite.^{30–34} Magmatism in UDMA at first had tholeiitic nature, then changed to calc-alkaline and alkaline.³⁵ A wide variety of porphyry Cu ± Mo mineralisations have been reported and investigated along the UDMA.³⁶ In the UDMA, there are at least two giant porphyry deposits, namely Sarcheshmeh [1200 million tons (Mt) at 0.7% Cu and 0.03% Mo]³⁷ and Sungun (500 Mt at 0.7% Cu and 0.01% Mo),³⁸ one large porphyry deposit (Meiduk: 170 Mt at 0.86% Cu, 0.007% Mo, 82 ppb Au and 1.8 ppm Ag),³⁹ and at least three medium-sized porphyry deposits. Nevertheless, the fact is that compared with southern parts, the central parts of UDMA are not well understood.^{40,41} The Saveh area is located in the central part of the UDMA (Figure 1). As shown in Figure 1b, the oldest rocks cropping out in the study area include Eocene volcanogenic rock sequences, i.e. dacitic lava and tuff, with andesite and basaltic

Figure 1. (a) The location map of the study area shown on ASTER image; (b) geological map at the scale of 1:100,000, adapted from a Geological Survey of Iran geological maps.²²

andesite, trachyandesite and rhyodacite.³⁵ It is intruded by Oligo-Miocene intrusions which consist mainly of granite, diorite, granodiorite, gabbro-diorite, tonalite and quartz diorite. The displacement along WNW-ESE trend

faults has affected granite, diorite, granodiorite, gabbro-diorite, tonalite and quartz diorite units.⁴² These rocks are overlaid by a sequence of Miocene gypsiferous marl having interlayers of sandstone and conglomerate that

are covered by younger clastic rocks.²² Therefore, the present work aims to utilise new technology based on the integration of spectroscopy and remote sensing analyses to have a wider view of the types of mineralisation zones that have been created by igneous activities. It is noteworthy that according to the knowledge of the authors, this integrated methodology has not been used to study alteration zones.

Research methodology

In this paper, the ASTER and ETM⁺ images of 6 October 2003 and 25 July 2000, respectively, were used. The images were transferred from the USGS website (<http://earthexplorer.usgs.gov/>). The ETM⁺ image was used for geometric correction and the total RMS error of less than 2 m was calculated. The ETM⁺ consists of eight spectral bands, that is six bands of data in the VNIR and SWIR spectral regions at a resolution of 30 m, the panchromatic band and a thermal band with a spatial resolution of 60 m and 15 m. Image-to-image geometric correction is performed by finding similar ground control points (GCPs) on two images. These points are selected in fixed areas with appropriate distribution and numbers. Moving points from their place in the image without geometric correction to their place in the image with geometric correction is the general pattern of this type of correction, which can be done in the ENVI software environment with high accuracy. As mentioned before, there are 14 spectral bands in ASTER data which consist of three bands of VNIR, six bands of SWIR and five bands of Thermal Infrared (TIR) with 15, 30 and 90 m spatial resolution, respectively. The coordinate of the images is between 49° 59' 59" and 50° 36' 29" eastern longitudes and 35° 04' 11" and 35° 15' 54" northern latitudes.

ASTER image pre-processing

To convert the digital number (DN) data of ASTER images into reflectance data, various processing methods such as cross talk, FLAASH atmospheric correction and layer stacking were carried out using ENVI software.^{43,44}

ASTER image processing

To improve geological detail of the study area, various types of band ratios^{45–50} and False Colour Composite (FCC) combinations were used.^{51,52}

Mixture Tuned Matched Filtering (MTMF) method Boardman *et al.*²⁰ used the MTMF technique to sub-pixel level for mapping target minerals from ASTER data. The MTMF algorithm uses the spectra extracted from the image or selected from a spectral library to compare them to the reference or field sample spectra in terms of mapping target minerals. The main steps in the MTMF algorithm are Minimum Noise Fraction (MNF) and Matched Filtering (MF). The MNF is necessary as input data for the MTMF algorithm. It means that MNF recognises the noise via the assessment from the covariance matrix and decreases the dimension of data via the range of standard principal components.⁵³ MF is used for abundance assessment and mixture tuning for identifying unacceptable or false-positive pixels.²⁰ MF and infeasibility are the results of this algorithm to evaluate the mapping process selected the pixels with low infeasibility and high MF in a two-dimensional space.²⁰ The infeasibility image is used to reduce the number of false positives. False positives are sometimes found when using MF. Pixels having high infeasibility are expected to be MF false positives. Correctly mapped pixels will have an MF score more than the background distribution around zero and a low infeasibility value. The infeasibility amounts are in noise sigma units that vary in the DN scale with an MF score.

Spectral Angle Mapper (SAM) method

The SAM classification method is done by calculating the spectral similarity between the image spectrum and a reference spectrum and permits the rapid mapping of various minerals.^{54,55} Similar to the MTMF method, in the SAM algorithm, the reference spectra are taken directly from the image or extracted from laboratory or field measurements. Calculating the angle among the two spectra and treating them as vectors in *n*-dimensional space, measures the spectral resemblance is the basis of the SAM method. In this method, high similarity occurs when the angle between the two spectra is small and high angles depicts low similarity.⁵⁶ This algorithm is an easy, fast and very powerful classification method for mapping the spectral likeness of image spectra to reference spectra.⁵⁷

In this paper, the FCC, MTMF and SAM methods are employed to map the phyllic, argillic and propylitic mineral zones. The FCC method creates an image with an unusual colour to enhance a target. Furthermore, at this stage, a second field visit was conducted for sample collections

from the areas enhanced by MTMF and SAM algorithms for further spectral analysis with a FieldSpec3® and TSG.

Spectral measurement

FieldSpec3® spectrometer

The spectra of samples were measured using the FieldSpec3® at the Shahid Chamran University of Ahvaz, Iran. The FieldSpec3® spectrometer can measure the reflection of the target. This optical device is specifically applied for field and laboratory measurements which determine the spectral response of the targets in the VNIR and SWIR. The spectral range is 350–2500 nm with a collection time of 0.1 s for each spectrum. Rock outcrops in the study area which are andesite, basaltic, basaltic andesite, granite, granodiorite and tuff were measured using the FieldSpec3®.

The Spectral Geologist® (TSG)

TSG software (version 7.1.0.073 July 2017)⁵⁸ developed by CSIRO, Australia, is the industry-standard tool for geological analysis of spectral reflectance data of minerals, rocks and soils with the help of the Spectral Assistant™ (TSA) subroutine. The benefits of using the TSA software are: it uses a fast subset selection routine, decides on the number of materials in the mixture, the concurrent assessment of the background and the mixture constituents, and the use of an estimated within-class covariance matrix to improve the classification accuracy. The current TSA SWIR library includes 60 classes that are kaoline (well-crystallised kaolinite, poorly crystallised kaolinite, dickite and nacrite); mica group (illite, muscovite, phengite and paragonite); other AIOH minerals (montmorillonite, prehnite, pyrophyllite, topaz); three varieties of chlorite (magnesium, iron and intermediate); amphiboles (tremolite, actinolite, hornblende and riebeckite); a variety of other MgOH and FeOH bearing minerals (biotite, phlogopite, nontronite, talc); sulphates (jarosite, gypsum and three varieties of alunite); carbonates (calcite, dolomite, ankerite, magnesite and siderite); epidote and opaline silica.

Results and discussion

ASTER processing

In this paper, ASTER data spectroradiometry techniques and conventional methods such as MTMF and SAM are

employed for detection and mapping of minerals in the phyllic, argillic and propylitic zones in the northern parts of Saveh, Iran.

FCC of bands 4, 6 and 8 [band 4 (1.6–1.7 μm), band 6 (2.225–2.245 μm) and band 8 (2.295–2.365 μm)] is a good band combination for the enhancement of phyllic, argillic and propylitic alteration zones due to the presence of absorption features at band 6 (Figure 2). Important minerals of phyllic and argillic alteration zones (AIOH minerals such as kaolinite, muscovite, montmorillonite and illite) have a strong reflection in band 4, while, notable minerals of propylitic alteration zones (MgOH minerals such as chlorite and epidote) show high reflection in bands 5 and 6.⁵⁹ By using the FCC468 image, alteration zones could be identified (Figure 3). In this FCC, the pink colour is related to argillic and phyllic zones, and green indicates propylitic zone.^{51,60}

To get an overview of the area, a 468-colour combination was used, then a field visit was made in order to collect samples for spectroscopy. Based on a 468 FCC image of the area, field sampling sites were selected. The first sampling was performed on 11 February 2017 at two sites (Figure 4). The collected samples showed

Figure 2. Alteration zones absorptions.

Figure 3. ASTER FCC468 image of the study area. In this image, phyllic and argillic alteration zones are pink and green colour shows a propylitic alteration zone.

Figure 4. The first field sampling in the study area.

Figure 5. The second sampling points in the study area.

phyllic and argillic alteration. Samples for spectroscopy were collected from these alterations and from chlorite porphyry andesite rocks in the area. Other collected samples were porphyry andesite. These samples spectrum was used by the MTFM method to identify alterations in the region. The second sampling was performed on 21 April 2017 to evaluate the classification results (Figure 5). This sampling showed that rock outcrops in the study area are andesite, basaltic, basaltic andesite, granite, granodiorite and tuff.

Spectral analysis

Spectral curves created by FieldSpec3® underwent rigorous analyses using TSG software. Figure 6 presents the flowchart of this study. In this flowchart, continuum removal normalises the studied spectra, therefore, the values will be set between 0 and 1 for easier comparison. The continuum is a convex hull fitted to a curve between two endpoints of it. The software could identify mineral groups such as the mica group (illite, muscovite, phengite and paragonite), kaolinite [well-crystallised kaolinite (wx), poorly crystallised kaolinite (px)], amphibole and chlorides (Fe-chlorites and Fe-Mg chlorites) (Figure 7).

Illite $\{K_{2-x}Al_4(Si_{6+x}Al_{2-x})O_{20}(OH)_4\}$, phengite $\{K_2(Al,Mg,Fe)_4(Si_{6+x}Al_{2-x})O_{20}(OH)_4\}$, muscovite $\{K_2Al_4(Si_6Al_2)O_{20}(OH)_4\}$ and paragonite minerals $\{Na_2Al_4(Si_6Al_2)O_{20}(OH)_4\}$ (TSG, version 7.1) of the phyllic zone show absorptions with respect to the elements in their composition.^{61,62} Major distinguishing features of illite are absorptions near 1409 nm, 1904 nm, 2203 nm and 2341 nm. The spectrum of phengite absorption features developed near 1414 nm, 1911 nm, 2205 nm and 2348 nm (Figure 7a). Muscovite features are similar to those of illite, but with less well-developed water bands. Distinctive features are sharp absorptions near 1412 nm, 1914 nm, 2201 nm, 2334 nm (Figure 7b). The paragonite spectrum is similar to muscovite spectra, with absorption features developed near 1410 nm, 1914 nm, 2205 nm and 2345 nm (Figure 7c).

The kaolinite $\{Al_2Si_2O_5(OH)_4\}$ (TSG, version 7.1) spectrum has absorptions developed near 1400 nm, 1912 nm, 2170–2210 nm, 2320 nm and 2334–2380 nm (Figure 7d).

Montmorillonite $\{(Ca,Na)_{0.67}Al_4(Si,Al)_8O_{20}(OH)_4 \cdot nH_2O\}$ (TSG, version 7.1) contains absorption peaks near 2206 nm, 2340 nm, 1411 nm and 1908 nm. Because of intense water absorption, both of the

Figure 6. Flowchart of the present study.

1411 nm and 1908 nm features show an asymmetric curve (Figure 7e).

Hornblende $\{(Ca,Na,K)_{2-3}(Mg,Fe,Al)_5Si_6(Si,Al)_2O_{22}(OH,F)_2\}$ displays features near 1405 nm, 1910 nm, 2180–2250 nm, 2320 nm and 2380 nm (Figure 7f).

Chlorite $\{(Mg,Al,Fe)_{12}[(Si,Al)_8O_{20}](OH)_{16}\}$ (TSG, version 7.1) shows two major absorption features at 2180 nm and 2250–2260 nm for Fe-chlorite or 2330–2355 nm for Mg-chlorite. Other important features are those near 1409 nm and 1912 nm (Figure 7g). Figures 8 and 9 show several rock samples and their spectra and Table 1 present common alteration minerals and their series of spectral absorption that can be measured with spectro-radiometry techniques.

In summary, absorption features of the mica group (illite, muscovite, phengite and paragonite), kaolins [well-crystallised kaolinite (wx), poorly crystallised kaolinite (px)], amphibole and chlorites minerals (Fe, Fe-Mg chlorites) may depict the presence of H_2O and OH molecules which may indicate changes due to hydrothermal alteration.

After sampling and measuring the spectra, the distributions of kaolinite and montmorillonite (argillic), illite,

muscovite, phengite and paragonite (phyllitic), hornblende, and chlorite (propylitic) were mapped using SAM and MTMF algorithms.

The concentration of the phyllic, argillic and propylitic zones is enhanced by the SAM and MTMF algorithms. However, when the phyllic zone is enhanced by the SAM algorithm, it appears to be more extensive than when the MTMF algorithm is used, whereas the argillic zone depicted by the MTMF algorithm appears to be larger than that indicated by the SAM algorithm. However, the results of the propylitic zone distribution are similar in SAM and MTMF algorithms. Figures 10 and 11 show the results of the detection of phyllic, argillic and propylitic alteration zones by the SAM and MTMF methods.

The alteration zones recognised by the SAM and MTMF methods were examined in the field. In terms of examining the zones detected by the algorithms, another field sampling was carried out. The sampling showed that each zone detected by the algorithms was correct. An extensive range of igneous rock types (diorite to granodiorite and granite) are spatially associated with porphyry deposits as suggested by Richards⁶³ in this study area. In the Saveh

area, the phyllic alteration zone mostly occurs along granodiorite and granite igneous rocks, and in some areas, it is also visible on the tuff and andesitic lava, whereas the argillic and propylitic alteration zones are often associated

with the tuff and andesitic lava and in some areas the argillic and propylitic alteration zones also visible on the granodiorite rocks. The results of the two methods are very similar but the MTMF method is a little more appropriate.

Figure 8. (a) Granite rock sample, (b) the spectrum of muscovite-illite, obtained from granite rock sample, (c) basaltic rock sample and (d) the spectrum of chlorite Fe obtained from a basaltic rock sample. In (b) and (d) the vertical axis contains the reflectance values and the horizontal axis contains wavelength values (nm).

Figure 9. (a) Andesitic rock sample, (b) the spectrum of chlorite Fe, Mg obtained from an andesitic rock sample, (c) a kaolinite rock sample and (d) the spectrum of kaolinite obtained from a kaolinite rock sample. In (b) and (d) the vertical axis contains the reflectance values and the horizontal axis contains wavelength values (nm).

Table 1. Common alteration minerals measured with spectroradiometric technique in the study area and their range of spectral absorption.

Mineral type	OH & H ₂ O = 1400 nm, H ₂ O = 1900 nm	AlOH = 2190– 2220 nm	MgOH = 2310– 2360 nm	CaCO ₃ = 2350 nm	K = 630– 740 nm	Fe = 900 nm	Na = 580 nm
Muscovite	*	*	—	—	*	—	—
Illite	*	*	—	—	*	—	—
Phengite	*	*	*	—	*	*	—
Paragonite	*	*	—	—	—	—	*
Kaolinite	*	*	—	—	—	—	—
Montmorillonite	*	*	—	*	—	—	*
Hornblende	*	*	—	*	*	*	*
Chlorite	*	*	*	—	—	*	—

*Present, — not present

The spectroradiometric technique's results

Samples have been collected from the common areas of each zone obtained by both algorithms. The sampling points are displayed in Figure 5. Then, to verify the results of the SAM and MTMF methods, each rock sample was

examined with a FieldSpec3[®] spectrometer and with TSG software.

The TSA of the TSG software, having important minerals among the 60 classes in the current library has an extraordinary ability to detect minerals. When the spectrum

enters the TSG software, the software compares these spectra based on their absorption points (depth, position, width, slope etc.) with its library spectra and, therefore, it can identify the desired mineral. The similarity matching is based on calculating the correlation between two spectra. Figure 12 shows the percentage of identified mineral groups of the study area using TSG software which is evidence of the TSG's ability to detect minerals. The mica group (illite, muscovite, phengite and paragonite) have their main absorption at about 1400nm and 2200nm (Figure 7a). Kaolinites [well-crystallised kaolinite (wx), poorly crystallised kaolinite (px), dickite and nacrite] have "doublet" absorption features at around 1400nm and 2170nm. Unlike the mica group, all the kaolinites have the same chemical formula $[Al_4Si_4O_{10}(OH)_8]$ (Figure 7b). The two small absorption features at about 2250nm and 2350nm taken together are diagnostic of chlorite features (Table 1) and the general chemical formula for chlorite is $(Fe; Mg; Al)_{12}(Si; Al)_8O_{20}(OH)_{16}(Fe; Mg; Al)_{12}$. As the Mg number increases, the positions of both of these features tend to decrease. The TSA software routine distinguishes the mineral types according to the shapes

of these absorption features. To determine the type of minerals, all the samples spectra were entered into the TSG software and the results are displayed in Table 2.

The areas detected by the MTMF method were examined and sampling was performed on those areas. Then the spectrum of rock samples was obtained using a spectrometer and entered into the TSG software in order to detect rock samples correctly. The results showed that TSG is well able to detect minerals and this software confirms the results of the MTMF method. The results also showed that the MTMF is capable of detecting the studied minerals (Table 2).

Based on the spectroscopy tools (FieldSpec3® and TSG software), prominent minerals such as illite, muscovite, phengite, paragonite, kaolinite, montmorillonite, hornblende, chlorite, gypsum, siderite, hematite and goethite are found in the study area. As previously stated, the TSG searches the similarities between the imported samples and the spectral library and it detects the percentage of minerals presented in the studied samples. Based on this, the percentage of minerals is displayed in different colours. According to Figure 12, it can be seen that the

Table 2. The identified minerals of the study area by TSG software.

Rock samples	TSG results	Rock samples	TSG results
Acidic tuff	Muscovite, illite, goethite	Dacitic	Chlorite Fe, paragonite
Acidic tuff	Muscovite, illite, montmorillonite, goethite	Dacitic	Paragonite
Acidic tuff	Muscovite, illite, montmorillonite, goethite	Ferric sample	Siderite, goethite
Acidic tuff	Muscovite, goethite	Ferric sample	Siderite, kaolinite (Px), goethite
Acidic tuff	Muscovite, illite, goethite	Granite	Muscovite
Acidic tuff	Muscovite, illite, kaolinite (Px)	Granite	Illite, chlorite-Mg, goethite
Andesitic	Muscovite, goethite	Granite	Muscovite, illite, montmorillonite
Andesitic	Siderite, montmorillonite	Granite	Muscovite, goethite
Andesitic	Chlorite-Mg, goethite	Granite	Muscovite, illite, siderite, goethite
Andesitic	Chlorite-Mg, goethite	Granite	Muscovite, illite
Andesitic	Muscovite, montmorillonite, goethite	Granite	Siderite, montmorillonite, goethite
Andesitic	Siderite, montmorillonite	Granite	Siderite, montmorillonite, goethite
Andesitic	Siderite, montmorillonite, goethite	Granite	Illite, kaolinite (Px), goethite
Andesitic	Siderite, montmorillonite	Granite	Siderite, montmorillonite
Andesitic-basaltic	Chlorite Fe	Granite	Paragonite, chlorite Fe-Mg
Andesitic-basaltic	Illite, montmorillonite, goethite	Hornblende andesitic	Hornblende, kaolinite (Px), goethite
Andesitic-basaltic	Chlorite Fe-Mg	Hornblende andesitic	Hornblende, epidote, goethite
Andesitic-basaltic	Chlorite Fe-Mg, muscovite	Hornblende andesitic	Hornblende, chlorite-Mg, goethite
Andesitic-basaltic	Chlorite Fe	Kaolinite sample	Muscovite, kaolinite (Wx)
Andesitic-basaltic	Chlorite Fe, paragonite	Kaolinite sample	Kaolinite (Px)
Andesitic-basaltic	Phengite, siderite, goethite	Kaolinite sample	Muscovite, gypsum
Andesitic-basaltic	Siderite, muscovite, goethite	Kaolinite sample	Kaolinite (Px), goethite
Andesitic-basaltic	Chlorite Fe-Mg, montmorillonite	Kaolinite sample	Kaolinite (Px), hematite
Basaltic	Chlorite Fe-Mg, muscovite	Kaolinite sample	Kaolinite (Wx), goethite
Basaltic	Chlorite Fe, paragonite	Kaolinite sample	Kaolinite (Wx)
Basaltic	Chlorite Fe, paragonite	Kaolinite sample	Kaolinite (Wx), goethite
Basaltic	Siderite, muscovite, goethite	Kaolinite sample	Kaolinite (Px)
Basaltic	Phengite, goethite, siderite	Kaolinite sample	Kaolinite (Wx), goethite
Basaltic	Siderite, montmorillonite, goethite	Pyroclastic sample	Chlorite Fe, goethite
Basaltic	Chlorite Fe-Mg, kaolinite (Px), goethite

Figure 12. Percentage of identified mineral groups of the study area using TSG software.

percentage of muscovite and illite minerals was higher than other minerals. Also, mineral groups are well detected in the samples by TSG.

Detection of the types of the mica group, kaolinite and chlorite minerals is one of the most important features using TSG software. The iron minerals of goethite, hematite and siderite could be recognised by TSG. The identification of each mineral depends on the amount (depth) of the absorption and wavelength position it imposes on the spectral curve. Because the TSG procedure for mineral identification is based on the spectral curve matching of the minerals and that of the software spectral data bank, therefore, the advantage of this software is that it also can indicate any amount of concentration of the minerals.

Conclusions

The vast and complex geological setting in the Iranian Saveh area needs cutting-edge technology to be implemented for the exploration and development of new

promising mineral potential zones. Therefore, the present study presents the combined use of spectroradiometry measurements and remote sensing data processing. In order to classify the satellite image, the procedure is that the ground samples are spectroscopically determined, then their spectrum is used as a reference spectrum for the implementation of the SAM and MTMF methods based on using ASTER satellite images. The results exhibit that the simultaneous use of these technologies could enhance fast and accurate zones of mineral potential. Different kinds of minerals such as illite, muscovite, phengite, paragonite, kaolinite, montmorillonite, hornblende, chlorite, gypsum, hematite, siderite and goethite have been identified for further detailed geochemical analysis. Based on field spectral data and their assimilation with ASTER satellite data, calibration and classification were carried out to make a clear distribution of minerals in the study area as has been shown in the present text. Because the TSG procedure for mineral identification is based on the spectral curve matching of the minerals and that of the software spectral data bank, the advantage of this software is that it also can indicate

any amount of concentration of the minerals. Finally, it is concluded that the area has undergone a different level of hydrothermal alteration so much so that phyllic, argillic and propylitic types of minerals are present in the study area. It should be noted that for the first time the study area has been investigated from a different view angle, i.e. in addition to spectral data bank creation and utilisation of specialised TSG software, a synoptic satellite view was used to see the whole complex simultaneously and analyse the relationship of the different rocks. It is vital to note that porphyry copper systems are centred within a distinct shell of hydrothermally altered rocks including sodic-calcic, potassic, chlorite-sericite, sericitic and finally near-surface advanced argillic, e.g. Sillitoe.⁶⁴ Notably, in most cases, the erosion may lead to the surficial exposure of all alterations. Although there are many key factors in the exploration of porphyry deposits, the fact is that the alterations are one of the most vital signals in the exploration of porphyry copper systems. Accordingly, the findings of the present work could be complementary to the future exploratory works in the area and maybe other porphyry-bearing belts.

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