Utilization of multispectral satellite images is an excellent approach in the reconnaissance stage of gold prospecting due to its high effectiveness and low cost. This research used Landsat 8 Remote Sensing data and Geographical Information System to identify, delineate and map hydrothermal alteration zone relating to gold pattern mineralization in Prestea Huni Valley District in the Western Region of Ghana. Principle Component Analysis (PCA), three band combinations, spectral rationing and Crosta techniques are used in this study. For three band combinations, bands 5, 4, 3 and 5, 6, 7 are used to map location of hydrothermal alterations. Band ratios (7/5 and 6/7), (6/5) and (4/2) are used to identify presence of clay, ferrous and iron oxide minerals respectively. Sabin’s ratio 4/2, 6/7 and 6/5, Kaufmann ratio 7/5, 5/4 and 6/7, and 4/2, 6/7 and 5 are used to identify vegetation, outcrop and hydrothermal alterations respectively. Crosta and PCA techniques were used to suppress the interference of vegetation in delineating the alteration zones. Results revealed that the three band combinations and spectral ratioing clearly depicts hydrothermal deposit of ferrous minerals, clay and iron oxide minerals. PCA identifies presence of iron-oxide and hydroxyl minerals as bright pixels. The first three high order principal components (PC1, PC2 and PC3) of input spectral bands gave more than 98% of the spectral information. Thus, the results from satellite images an effective and efficient way of mapping hydrothermal alteration zones at regional scale.

Contribution/Originality: This is one of the very few studies which have investigated to identify hydrothermally alteration relating to gold mineralization. The study documents how Satellite data and Remote Sensing and Geographical Information System techniques can be used to delineate hydrothermal alterations at regional scale at a reduced cost.
1. INTRODUCTION

Remote Sensing (RS) has proved to be an important tool tracing mineral deposits by distinguishing processes of mineralization through spectral anomalies (Rajesh, 2004; Van der Meer et al., 2012). RS is a powerful technique in identifying hydrothermally altered rocks, structures, lineaments, oxidation products, morphology, lithological units, vegetation anomalies and other valuable information to geologists (Goetz et al., 1982; Goetz, 2009). It is broadly acknowledged that hydrothermal alteration plays an important role in mineral prospecting. The necessary dispersal of alteration zones around the limited borders of an ore body helps narrow down the exploration events into smaller targets (Masoumi et al., 2017). Hydrothermal mineralization alteration is a process which alters the mineralogy and chemistry of the host rocks resulting in creating mineral assemblages which differ according to the setting, degree and time of the alteration processes (Mia and Fujimitsu, 2012; Sadiya et al., 2014). Usually, hydrothermal alteration processes lead to the formation of clay and other silicate minerals including kaolinite, sericite, illite, quartz, chlorite, epidote, and calcite (Ranjar et al., 2004).

Spectral ratioing (arithmetic operations) and three band combinations of remotely sensed data can detect the above-mentioned minerals and discover buried deposits (Masoumi et al., 2017). Band ratioing have been used by different authors to examine and interpreted remotely sensed data to map hydrothermally alteration relating to minerals deposit patterns. For example, Gad and Kusky (2007) used ASTER band ratios 4/7, 4/6 and 4/10 to map the granite and metamorphic belt of the Wadi Kid area of Sinai, Egypt. Pour et al. (2017) applied qualitative image processing procedures such as colour composites, band ratios and Principal Components Analysis (PCA) on ASTER and Landsat-8 bands to map lithological and alteration minerals in poorly exposed lithologies in north-eastern Graham Land, Antarctic Peninsula. Sadiya et al. (2014) used band ratioing, PCA and crosta technique to map and detect minerals in Bwari Area Council, Abuja Nigeria. According to Sadiya et al. (2014) band ratios 3/1, 4/5 – 4/3 and 5/7 suggests the presence of ferric iron minerals, hydroxyl minerals and clay mineralization respectively. Also, band ratios 3/1, 4/5 – 4/3 and a sum of the band ratios (3/1, 4/5 – 4/3) displayed as RGB enhanced the location of mineral deposits. Applications of remotely sensed satellite data are widely used and unique in mapping of different lithologies, mineral resources, ore deposits and minerals of alteration zone in many types of deposits, such as porphyry copper deposits (Alimohammadi et al., 2015; Abuzied et al., 2016). Long term Landsat data has made possible in the advancement of using the satellite images in geological application. The development in the image processing methods which can be used to improve the satellite images and show the spectral appearances of required features on the ground (Pour and Hashim, 2014). Landsat TM data, both TM5 and TM7 are among the most utilized satellite multispectral remote sensing data in mineral investigation (Mshiu et al., 2015). The datasets are utilized to find target zones in the reconnaissance phases of investigation (example (Kruse et al., 2002; Crosta et al., 2003)). They are equipped for distinguishing the territories rich in Fe-minerals and earth ± carbonate ± sulfate minerals. Landsat 8 data can detect the altered rocks and ferrous minerals through the Operational Land Imager (OLI) part of the image due to the absorption and reflectance characteristics of these rocks which appear in this range (Abhary and Hassani, 2016). When it comes to silicate the two additional thermal bands have the ability to detect the reflectance characteristic for it in their range (Dehnavi et al., 2010). Landsat-8 OLI/TIRS data can be utilized to monitor variety of earth-based and atmospheric occurrence, including rural observing; topographical mapping; evapotranspiration; cloud location and examination; mapping heat motions from urban communities; checking air quality; checking volcanic movement; biomass consuming; mechanical warm contamination in the environment, streams and lakes; checking/following material transport in lakes and seaside locales; finding insect breeding zones and applications that will ultimately arise later as a result of global warming and climate change (Pour and Hashim, 2014). In Ghana, there has been exceptionally small research in utilizing remote sensing techniques to identify mineral deposits. One of the few carried out was (Kwang et al., 2014) they applied remote sensing(RS) and geographical information system techniques(GIS) to map gold potential areas at Birim North District in the Eastern Region of Ghana. However, a remote sensing study has barely been carried out in the...
Prestea Huni Valley district to identify hydrothermally alteration relating to gold mineralization. The aim of this study is to use remote sensing and GIS techniques to detect and map location of hydrothermal alteration relating to gold pattern mineralization in the Prestea Huni Valley district as a means of producing a mineral deposit database for Ghana. This study offers detailed understanding and effective way of prospecting for gold, which is an economic livelihood of the economy of Ghana, critical for effective planning, at the Prestea Huni Valley District of Ghana to confirm sustainable accessibility of gold for present and future generation.

2. MATERIALS AND METHODS

2.1. Study Area

The Prestea Huni Valley District lies approximately on latitude 5° 20’N and 5° 40’N and between longitude 1° 50’W and 2° 10’W. The Prestea/Huni-Valley District Assembly is one of the recently made Districts in the western area. It was cut out from the then Wassa West District Assembly in 2004 and has its central station at Bogoso, which is a mining town arranged around 33 kilometers north of Tarkwa. Prestea/Huni-Valley District Assembly imparts limits on the north west to Wassa Amenfi Central District, on the south east with Wassa East District, on the south west with Axim Municipal Assembly, on the south with Tarkwa-Nsuaem Municipal Assembly and toward the north by Wassa Amenfi East District Assembly (MLGRD, 2006). The district lies within the South Western Equatorial Zone (Oduro, 2011).

![Figure-1. Location of the study area.](image-url)
2.1.1. Geology of the Study Area

The geology of southwestern Ghana is dominated by the Birimian Supergroup, which is comprised of Proterozoic sedimentary and volcanic rocks. These units are intensely deformed by faulting and folding activities and were later intruded by various granitoid intrusions. Overlying the Birimian unit is a less deformed coarse clastic sediments of the Tarkwa group which is thought to be erosional products of the Birimian Supergroup and Belt type granitoids that were deposited in long narrow intermontane grabens, which formed as the result of localized rifting in the Birimian metavolcanic belts (Leube et al., 1990). All these formational events took place during the Eburnean orogeny which affected the whole of Ghana spanning from 2,080 to 2,240Ma (Kesse, 1985). Rocks of the Tarkwaian Group are characteristically highly magnetic and can be identified easily on aeromagnetic images (Griffis, 1998).

The Prestea concession falls within the southern part of the Ashanti greenstone belt which host world class gold mineral deposits; and it’s characterized by a large number of gold deposits including active and mines like Anglogold Ashanti, Golden Star Resources at Prestea and Bogosso and Gold Fields at Tarkwa and Damang (Kuma et al., 2010). The geology of the Prestea concession is grouped into three main rock-structural units which are separated by a westward steeply dipping fault. These structural assemblages are; the Tarkwaian litho-structural units, the tectonic structural units made up of shared graphitic sediments and volcanic flows and undeformed sedimentary units of the Kumasi basin. Both rocks and Mineralization are structurally controlled and conforms to the regional NNE to NE tectonic corridors which affected southwestern Ghana (Griffis et al., 2002). The style of gold mineralization occurs in variable forms within the belt which include gold associated with quartz veins and disseminated types, gold bearing pebble conglomerate which is typical of the Tarkwian group, alluvial and oxide deposits (Kuma et al., 2010).

The eastern part of the concession is made up of the Tarkwaian litho-structural units which comprise of sandstones, pebbly sandstone, and conglomerates. The quartz pebble conglomerates are by far the most important member of the group because of its gold mineralizing potential. The deformational activities have resulted in the formation of another litho-structural units which overlies the Tarkwaian sediments called the tectonic breccia. These are formed in areas which share close contact with the Kumasi sedimentary basin. Various forms of alteration patterns are used to separate one volcanic unit from the other. Weakly altered mafic volcanic rocks are characterized by distal chlorite/calcite alterations patterns while strongly altered mafic volcanic rocks which are generally located at the main Reef Fault are characterized by silica/sericite/Fe-Mg carbonates alteration patterns. Finally, within the tectonic assemblage, very distinctly, the volcanic lenses are seen to intercalate with the sheared graphitic sedimentary horizons which represents sheared and brecciated sequences of siltstone, mudstone and greywacke units affected by pervasive graphitic alteration (Griffis et al., 2002). The primary features within these units are generally overprinted and destroyed by deformational activities even though primary beds are locally preserved. All the major lithologies are illustrated in Figure 2.
2.2. Data

The RS image used is Landsat 8. It was downloaded from Google Earth Engine Platform (https://code.earthengine.google.com/2854ac10f2cc8be21ebdb5018a86d69d). This image (cloud free) was captured on 15 of February 2016 according to Google Earth Engine Platform. Geological Map was obtained from the Geological Survey Department at a scale of 1:1000000.
2.3. Method

2.3.1. Image Processing Techniques

Spectral characteristics of the study area in term of mapping hydrothermal alteration zones were presented using Remote Sensing techniques. Sub-setting and stacking image processing were performed on the Landsat 8 image to obtain the boundary image of the study area. Image enhancement method such as colour composite, band rationing and principal component analysis were executed on the image of the study area to map location of hydrothermal alteration relating to gold mineralization in the study area.

2.3.2. Three Band Combinations

Three band combinations (false colour composite) is useful for mineral and rock discrimination and infer possible alterations centered in colour intensity disparities. A composite with the visible bands of the spectrum that correspond to red, green and blue is called a true colour composite. When a composite is created with non-visible bands it is called a false colour composite image. Different false colour composites were generated using bands 5, 4, 3 and 5, 6, 7 for the purpose of mapping location of alteration mapping.

2.3.3. Spectral Ratioing (Band Ratioing)

Spectral ratioing or band ratioing is a multispectral image processing method that involves the arithmetic division of one spectral band by another. This arithmetic division results in the ratio of spectral reflectance measured in the one spectral band to the spectral reflectance measured in another spectral band. Spectral ratioing transforms the data thereby minimizing the effects of such environmental conditions (Jensen and Lulla, 1987). ArcGIS software was used to performed band ratios using various bands in Landsat 8 for enhancement of spectral signatures for alteration minerals. Clay minerals alterations were highlighted using the Landsat 8 band ratios 7/5 and 6/7. Landsat 8 band ratio 4/2 was used to discriminate alteration patterns for iron minerals. Landsat 8 band ratio 6/5 was used to highlights alteration patterns for ferrous minerals. Colour combination of band ratios 4/2, 6/7 and 5 were used for the detection of hydrothermal zones. For instance Sabins (1999) and Kaufmann (1988) used band ratios for discriminating between altered rocks, vegetation and rock types which yielded significant results.

2.3.4. Principal Component Analysis (PCA)

PCA is a multivariate statistical strategy that examines the Eigenvector loadings of a multispectral satellite data to determine the unique spectral reaction of distinctive minerals and rocks contained in the image called Crosta technique (Crosta et al., 2003; Mia and Fujimitsu, 2012). PCA is normally utilized for distinguishing and mapping the dispersion of alteration in metallogenic regions (Nouri et al., 2012). The panchromatic image is histogram coordinated to the first PC. It then substitutes the selected component and a reverse PC transform takes the merged dataset back into the original multispectral feature space (Chavez et al., 1991). By applying PCA a new data set with lesser variables is formed (Lillesand et al., 2000). PC1 is generally the weighted average of all data and represents albedo and topographic effects found in the scene (Ranjbar et al., 2004). Sabins (1999) effectively adopted an approach where PCA of band sets 1347 and 1547 were processed individually to determine the occurrence of iron-oxide and hydroxyl-rich minerals. The method guarantees the undesirable signals from materials such as vegetation are isolated from geological/ mineralogical target materials and collect them in a separate principal component (PC). Crosta strategy does not require much consideration on pretreatments to suppress signals from vegetation since the undesirable signals will be collected in other PCs, different from the PC conveying target signals (Mshiu et al., 2015). It is by and large acknowledged that the first three high order principal components (PC1, PC2 and PC3) of input spectral bands gave more than 99% of the spectral information, while lower order components usually contain low signal-to-noise ratios (Amer et al., 2010).
3. RESULTS

3.1. Colour Composite

Using false colour composite, hydrothermal zones with mineral spectral signatures near infrared (IR) to mid-infrared can be detected using ETM+ (Jensen, 2005). In Figure 3, red represents vegetation, black represents water and the greyish colour represents rock or soil. Vegetated areas appear in red, outcrops in light blue and water in black and hydrothermal altered rocks can be identified as blue Figure 4.

![False colour composite Bands 5, 4, 3, in the study area.](image-url)
3.2. Band Ratio

The band ratio 4/2 reveals areas that contain iron minerals and can be seen in bright tones in Figure 5. Iron oxide minerals have high absorption in TM band 2 and higher reflectance in TM band 4. Iron-stained hydrothermally rocks are shown by areas having higher Digital Numbers (DNs) as shown in the TM 4/2 ratio image Figure 5. The areas with high DN values are shown in bright tones and they relate to altered rocks. Band 6/5 ratio Figure 6 depicts areas with ferrous minerals due to the high reflectance of these minerals in this ratio (Gupta, 2003). Areas with high DN values are the bright tones in Figure 6 and depicts ferrous minerals alteration and hence correlate with altered rocks. Band ratio 7/5 detects clay mineralization in the study area. Areas with high DN values indicated by bright tones represent clay minerals alterations which correlate to altered rocks as shown in Figure 7. The clay minerals may be illite, montmorillonite or kaolinite. All these features have a high reflectance on band 6 and low reflectance in band 7 of the Landsat 8 image.

Figure 4. False colour composite Bands 5, 6, 7 in the study area.
Band ratio 6/7 in Figure 8 shows altered rocks on the image with brighter pixels due to the reflectance features for the altered rocks in band 6 and absorption features in band 7 on Landsat 8 image. This ratio is considered the best ratio for mapping altered rocks (Ali and Pour, 2014). The combination of band ratios 4/2, 6/7 and 5 allows the recognition of outcrops as rose and brown, altered rocks as red areas, vegetation in light blue and water as black Figure 9. Kaufmann band ratio combinations using bands 7, 4, 3 and 5 in Landsat TM is equivalent to bands 7, 5, 4 and 6 in Landsat 8. Kaufmann band ratio was determined and it shows a good combination for geological determination. The Sabins band ratio 7/4, 6/3, 5/7 was also tested and it revealed a good geological purpose. From the final image Figure 10, black indicates water bodies, dark green indicates vegetation, light green indicates clay rich rocks and red indicates mineral rocks-iron oxides (Sabins, 1999; Abubakar et al., 2014). The combination of band ratio 7/5, 5/4, 6/7 gives an image which where red represents minerals containing iron, green represents vegetated zones and blue represents OH/H2O-, SO4- or CO bearing minerals (Kaufmann, 1988; Abubakar et al., 2014) displayed in Figure 11.
Figure 6. Landsat 8 band ratio 6/5 image of the study area.
Figure 7. Landsat 8 band ratio (7/5) image of the study area.
Figure-8. Landsat 8 band ratio (6/7) image of the study area.
Figure 9. Colour composite of landsat 8 band ratio in RGB sequence: 4/2, 6/7 showing vegetation, outcrop and altered rocks in the study area.
Figure 10. Sabin's ratio image (4/2, 6/7, 6/5) showing vegetation, outcrop and hydrothermal alterations in the study area.
3.3. Principal Component Analysis and Crosta Technique

Firstly, the Principal Component Analysis was applied to the six Landsat 8 bands (2, 3, 5, 6, and 7). The technique was executed on two sets of band combination; 2456 and 2567 to map iron oxide and hydroxyl minerals respectively. The PCA eigenvector values are shown in tables below. PCA values for all bands is displayed in table 1. PC1 contains the highest amount (75.9%) of the total variance of the six bands and it gives information mainly on albedo and topography, it decreases till it reaches (0.1%) for PC 7. PC2 represent 14.6% of the total data variance and PC3 contain 8.0%. The first three high order principal components (PC1, PC2 and PC3) of input spectral bands gave more than 98% of the spectral information.

In Table 2 in Table 1, PC1 corresponds to the albedo image with 66.0% of data variance. PC2 indicates very high vegetated areas and are the dark pixels with a percentage of variance 21.5%. PC3 describes the contrast.
between SWIR bands and visible/NIR bands. PC4 show negative eigenvector loading for band 7 (-0.75926) and positive eigenvector loading for band 6 (0.16040). Hence, hydroxyl bearing minerals are mapped as dark grey to black in the PC4 as shown in Figure 13. Hydroxyl-bearing minerals have reflectance features of 1.55-1.75 μm (band 6) and absorption of 2.1-2.4μm (band 7). PC4 highlights hydroxyl bearing minerals as dark grey to black pixels as shown in Figure 13. Clay minerals have absorption in band 7 and reflection in band 6.

PC1 corresponds to the albedo image with 66.9% of variance data Table 3. PC2 with 21.4% of variance represents the vegetated areas as darker pixels. PC3 represent the contrast of SWIR band between the visible and NIR bands. PC4 shows negative eigenvector loadings for band 2(-0.54278) and positive eigenvector loadings for band 4 (0.785618). Iron oxides minerals have absorption features and low reflectance of 0.45-0.51 μm in band 2 and reflectance features and low absorption of 0.64-0.67 μm in band 4. These two bands have higher loadings in PC analysis, and the pixels with more abundance of iron oxides minerals can be identified as bright pixels in PC4 image Figure 12.

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**Table 1. Principal component analysis of bands (2,3,4, 5, 6, 7 ).**

| Bands | PC1     | PC2     | PC3     | PC4     | PC5     | PC6     |
|-------|---------|---------|---------|---------|---------|---------|
| Band2 | 0.439021| -0.06792| 0.415927| -0.74209| 0.095784| -0.2642 |
| Band3 | 0.459153| -0.10998| 0.210331| 0.079218| -0.38578| 0.76009 |
| Band4 | 0.463945| 0.000401| -0.01311| 0.407753| -0.52491| -0.5855 |
| Band5 | 0.20495 | -0.95197| 0.186454| 0.098211| -0.01643| -0.0841 |
| Band6 | 0.357922| -0.27326| -0.85131| -0.24668| 0.089355| 0.05059 |
| Band7 | 0.459437| -0.03581| 0.151983| 0.454173| 0.747137| 0.0071  |
| STD   | 2.1337  | 0.9345  | 0.69146 | 0.2522  | 0.1568  | 0.08744 |
| Percentage Explained | 75.8788 | 14.55592 | 7.968513 | 1.059668 | 0.409718 | 0.127429 |

**Table 2. Principal component analysis of bands 2, 5, 6, 7 for hydroxyl minerals.**

| Bands | PC1     | PC2     | PC3     | PC4     |
|-------|---------|---------|---------|---------|
| Band2 | 0.56626 | 0.105688| -0.52386| 0.627491|
| Band5 | 0.3034  | 0.925537| -0.21742| -0.06357|
| Band6 | 0.48254 | 0.33564 | 0.785086| 0.160409|
| Band7 | 0.59534 | 0.084565| -0.24888| -0.75926|
| STD   | 1.62476 | 0.926625| 0.669874| 0.229781|
| Percentage Explained | 65.99595 | 21.46587 | 11.21828 | 1.319983 |

**Table 3. Principal component analysis of bands 2,4,5,6 for iron oxide minerals.**

| Bands | PC1     | PC2     | PC3     | PC4     |
|-------|---------|---------|---------|---------|
| Band2 | 0.55069 | -0.10921| 0.624665| -0.54278|
| Band4 | 0.59818 | -0.0656 | 0.14382 | 0.785618|
| Band5 | 0.3125  | -0.91599| 0.220559| 0.121081|
| Band6 | 0.49119 | -0.38044| -0.73516| -0.27118|
| STD   | 1.63631 | 0.924687| 0.647693| 0.218926|
| Percentage Explained | 66.9379 | 21.37616 | 10.48771 | 1.198209 |
Figure-12. Showing Principal Component image from Landsat 8 data of iron oxide bearing minerals (PC4) as bright pixels.
Figure 13. Showing Principal Component image from Landsat 8 data for hydroxyl minerals (PC4) as bright pixels.
Figure 14. Colour composite of Principal Component 1,2,3. Green-yellow colours represent granite outcrops, vegetation in green-blue, water as light blue and metasediments as blue.

Crostata technique was used to produce an image showing pixels with concentration in both iron oxide and hydroxyl minerals by merging iron oxide and hydroxyl bearing minerals images. Colour composite of the images corresponding to hydroxyl bearing minerals, the sum of both hydroxyl and iron oxide minerals, and iron oxide minerals are presented in RGB to permit the identification of alteration zones. Bright reddish to yellow agree with areas argillaceous than iron stained; bright cyan to bluish zones are more iron stained than argilized (Loughlin, 1991; Da Cunha Frutuoso, 2015) as shown in Figure 15.
3. DISCUSSIONS

Three band combinations, spectral ratioing and PCA were used in order to spectrally improve the reaction of iron-oxide, hydroxyl and clay minerals which signify to hydrothermal zones in Prestea Huni Valley district as shown in Figure 13 and 15. Iron oxides and clay minerals associated to hydrothermal alteration have been located using this approach. Band ratio technique applied on Landsat-8 data demonstrated to be acceptable for highlighting iron oxides and clay minerals at Bogoso and Prestea as shown in Figure 5,6,7 and 8. Band ratio 4/2 detects iron oxides mainly related to the rock types in Prestea and Bogoso areas, and also iron oxides below vegetation in the...
study area. These minerals may be hematite, jarosite and geothite alterations and are formed by the oxidation of sulphide minerals (pyrite, chalcopyrite) within the rock units in the area. The presence of these mineral alterations as shown in Figure 5 correlates with the presence of pyrites and chalcopyrite within the rock units in the area. These alteration minerals may have spatial relation with gold mineralization within the area. Areas with these alterations falls within the Bogoso concession where gold is mined as observed during the ground verification exercise and also with the local geology of the area. Further evidence is revealed within the northern part of the study area where Bogoso mine is located, which is characterized by gently dipping mineralized quartz vein stockworks and disseminated auriferous pyrite and arsenopyrite forming broad zones in the vicinity of the faulting zones. Smaller zones of disseminated sulfide mineralization coincide with areas where bedding and/or foliation in the adjacent Birimian sedimentary rocks is strongly discordant with the adjacent mineralized fault (Allibone et al., 2002).

Allibone added that, lower grade sulfide mineralization, commonly containing 2 to 10 g/t gold, is generally associated with disseminated arsenopyrite and/or pyrite in the graphitic mylonite fault rocks and the undeformed adjacent wall rocks (example (Leube et al., 1990; Milési et al., 1991)). These style of mineralization within the area also conforms to the two dominant types of shear zone-related mineralization which have been described from deposits hosted by the Birimian sedimentary rocks in Ghana. However, the principal association of gold within the Prestea and Bogoso gold districts are with arsenopyrite and arsenian pyrite. His studies depict that higher grade mineralization commonly containing >10 g/t free gold is generally hosted by quartz lodes localized in brittle fractures along controlling faults (examples (Appiah, 1991; Allibone et al., 2002)). According to Allibone, these higher grade lenses contain conspicuous arsenopyrite but completely absent in unmineralized lenses and this characteristic mineralization feature is very common in the prestea district than the Bogoso area.

The result of the Band ratio7/5 produced an acceptable index for highlighting of clay minerals. Clay minerals indicate the presence of silicate rocks which easily weather to release their clay components due to its low resistance to weathering. The clay minerals in this studies are suspected to be illite, montmorillonite or kaolinite. All these features have a high reflectance on band 6 and low reflectance in band 7 of the Landsat 8 image. Clay minerals was not adequately mapped in some portions due to the presence of thick vegetation within the study area. Sedimentary rocks, such as mudstone, shale, claystone and litharenite sandstone, contain large amounts of detrital clays such as montmorillonite, illite and kaolinite; these can be mistakenly be mapped as hydrothermally altered clay minerals. According to Knepper (1989;2010) the TM band ratio 7/5 highlights carbonate minerals (calcite and dolomite), Fe-OH bearing minerals (epidote) and Mg-OH bearing phyllosilicate minerals (chlorite, talc and serpentine). In this opinion, the area with bright tone in Figure 7 may not only be clay minerals but also carbonate minerals, chlorites, epidote, talc and serpentine. This conforms to the findings of (Allibone et al., 2002) on the Bogoso concession where gold is largely confined to the graphitic mylonite, which defines the fault zones, and imbricated slices of carbonate-altered rocks.

From principal component analysis ; PC1 gave positive eigenvalue loadings which makes distinguishing materials with spectral signature very difficult (Jensen, 2005). However, PC1 (Fig 6a) contains spectral information showing the iron-oxide bearing minerals in the study area which is significantly more than the percentage distinguished by utilizing band ratio of 4/2. This is due to the fact that PCA rejects the impact of vegetation signature interference (Ruiz-Armenta and Prol-Ledesma, 1998). From Figure 12 it can be seen that the iron-oxide materials formed little bunches within the study area. The style of gold mineralization within the area is structurally controlled and the graphitic mylonite and wall rock styles are similar to the sulphide mineralization at the Ashanti mine and other prominent mines within the belt such as Golden Star at Prestea (Oberthür et al., 1998). Both the PC4 Figure 13 and band ratio Figure 7 demonstrates a huge concentration of hydroxyl bearing minerals likely due to sericitisation during the alteration process. Indeed, in spite of the fact that the various band ratio colour composite in RGB shows the spatial distribution of hydrothermal alterations Figure 9,10 and 11, it
overstated the degree of mineralization by representing unmineralized rock-outcrop and/or uncovered surface as hydrothermal alteration zones. This may be due to the impact of vegetation obstructions and the inability for some band ratio combination to distinguished between lithologies having same spectral signatures with the actual mineral deposits and also weathering of minerals to various part of the study area. Figure 15 shows hydrothermal alteration on satellite image conforming with Botwe et al. (2018) findings within the study area. Hydrothermal deposits are formed when hot aqueous solutions containing minerals are deposited in weak zones and fissures. These fluids alters the materials they meet on their way and changes their color, chemical and mineralogical compositions. The fluids finally deposes its content and the minerals precipitates to form metaliferous deposits when are found in economic quantities. Mineral deposits from hydrothermal processes form economic reserves when they are concentrated in veins, and other cavities in lode forms or disseminated. Minerals such as quartz, pyrite, arsenopyrite are commonly associated with metaliferous minerals in lodes and they show alterations signatures in areas and have direct associations with mineralization. Same as the one acquired from the geology of the study area.

5. CONCLUSIONS AND RECOMMENDATIONS

The principal focus of this study was to conduct an investigation utilizing Landsat 8 and Remote sensing and GIS strategies to map various alteration in Prestea Huni valley District. Utilizing various alteration mapping techniques on the Landsat image in the study area, we found that the colour composite and band ratio approaches demonstrated their proficiencies to characterize the areas of clay mineral alterations, iron oxides minerals, vegetation and hydrothermal alteration. PCA presented the Fe oxides and hydroxyl altered minerals in the study area clearly. Consequently, it is reasonably clear that all these approaches are quite effective to map hydrothermal alteration using Landsat-8 OLI images and would be a very useful tool form exploration activities in the study area. These alterations correspond spatially with major gold mining sites in the study area and the signatures revealed during the studies conforms to the current signatures employed by the active mines in the area and also the general gold bearing zones within the Birimian sedimentary units of Ghana. The relationship demonstrated between the hydrothermal alteration map and major gold mining sites give an indication that, this method is suitable for the mapping of target areas in mineral exploration. However, the outcomes from the RS strategies of Landsat 8 image demonstrated the capacity of the data in mapping hydrothermal alteration zones at regional scale.

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