



1 **Integration of Remote Sense and Geographic Information Systems in Geological Faults**

2 **Detection in Crete Island, Greece.**

3 **Mohamed Elhag^{1*}, Dalal Alshamsi²**

4 ¹Department of Hydrology and Water Resources Management, Faculty of Meteorology,
5 Environment & Arid Land Agriculture, King Abdulaziz University Jeddah, 21589. Saudi Arabia.

6 ²Department of Geology, College of Science, United Arab Emirates University.

7 **Correspondence to: melhag@kau.edu.sa*

8 **Abstract**

9 Fracture systems are of great importance in the field of structural geology. Faults commonly
10 afford easy passage to groundwater and fluids such as hydrothermal fluids and magmas (mineral
11 entrapment over the years) or even contribute in earthquake hazard monitoring. For a geologist it
12 is not always easy to discern such morphotectonic structures at close range (i.e. heavy
13 overgrowth of vegetation). Both remote sensing techniques and spatial modeling (GIS) permit
14 the recognition and better understanding of the brittle tectonics in an area. This study was an
15 effort to delineate the tectonic structures (i.e. fault system) on the Crete Islands by combining
16 Sentinel-2 satellite data and spatial data. For the enhancement and better discrimination of
17 photolineaments primarily recognized on satellite imagery, a variety of enhancement techniques
18 have been applied. The evaluation of a photolineament as a potential fracture zone was based on
19 several factors; the DEM of the study area, the shaded relief, the slopes and corresponding
20 aspects, the drainage network, the geology and general observations on vegetative coverage
21 appearance. The application of these methods revealed several fracture zones, which we
22 recommend being certified by field investigations. Fault-mapping results may be used for a
23 variety of purposes. Indicative places of large concentration of groundwater are of vital
24 importance for subsequent exploitation by areas of need. Furthermore, because the well-known
25 Anatolian fault zone extends over the Northern part of Crete, the present work may provide
26 useful information for further analysis by geophysicists and seismologists.

27 **Keywords:** DEM. Fracture Detection, Morphometric, Morphotectonic, Remote Sensing,
28 Photolineaments.



29 **1. Introduction**

30 O' Leary et al. (1976) and Colwell, (1983) gave a good-informing definition of the term
31 lineament. It refers to a mappable, single or composite linear feature of a surface, whose parts are
32 aligned in a rectilinear or slightly curvilinear relationship and which differs distinctly from the
33 pattern of adjacent features and presumably reflects a subsurface phenomenon.

34 An extensive literature review on the consideration of linear features and lineaments is well
35 presented by Siegal and Gillespie (1980). Suffice to say here, that lineaments identified on aerial
36 photographs and satellite images may represent diverse topographic features (drainage lines),
37 vegetation/soil alignments, coastal lines, crests, ridges, stratigraphic contacts, fold axis (foliation)
38 and seismic zones (Lunkka 1994).

39 The importance of detecting lineaments lies in the fact that they often represent fault systems.
40 Faults indicate failure of the crust along a surface, accompanied by the relative movement of the
41 geological units from both sides of that surface (Caumon et al. 2009). Such a zone of structural
42 weakness is a major component in structural geology and may be related to a series of other
43 phenomena. It is noticeable that, in many cases in the past, such a system has been related to
44 seismic events occurring in an area. Locating and identifying the movement is of dual
45 importance to a geologist as the impacts of seismic hazards may encompass of both lives and
46 economic losses (Colwell 1983).

47 Furthermore, faults identification is of economic importance too. That is for; they commonly
48 afford easy passage to several fluids like water, hydrothermal fluids and magmas (Oelkers et al.
49 2009). Groundwater on the one hand, can be easily transferred through such channel ways,
50 sometimes over large distances and finally reaching and supplying areas of need. On the other



51 hand, mineral entrapping at some time in the past may be of advantageous exploration use (Elhag
52 and Bahrawi 2014).

53 In the case of faults detection, integration of Remote Sensing and Geographic Information
54 Systems is an issue of great interest. That is for, photolineaments on satellite images do not
55 always account for failure of the crust. They may well represent the drainage lines of the area
56 (rivers), geological boundaries of formations or even cultural features such as roads. The criteria
57 for faults investigation are discussed in several aspects (Greenbaum 1992; Schulson 2004).

58 Vegetation may play a great role in photogeology for it may reveal underlying structural features
59 not easily discerned at close range. As far as fracturing is concerned, attention should be given to
60 preferential growth of vegetation along linear-curvilinear surfaces (Rodomsky 2011). The reason
61 for this, as mentioned earlier on, is that fractures often act as channel for underground water.
62 Water subsequently, increases the moisture content of soil (in relation to surrounding area) and
63 encourages in one sense, a preferential alignment of vegetation along these fracture zones
64 (Singhal and Gupta 2010). Moreover, it is not rare to identify on an image (or photograph) abrupt
65 changes in types of vegetation or even sudden disappearance of a certain plant species.
66 Particularly in the case where vegetation varies over a surface underlain by the same type of
67 bedrock, fracturing is implied (Phillips et al. 2008).

68 Faults in some cases, may even throw permeable against impermeable rocks. Water again finds
69 ways along the zone of structural weakness and reaches the Earth's surface by numerous springs
70 (along the contact of the two formations). At these points, high moisture conditions are
71 particularly favorable to intense vegetation growth (Singhal and Gupta 2010).



72 Abrupt changes in slope are commonly associated with brittle tectonics (Agliardi et al. 2001;
73 Agliardi et al. 2009). Attention should be given though to the orientation of slopes with respect
74 to the illumination source. It may have an impact on vegetation and hence confuse the
75 photointerpreter. As Singhal et al. (2010) pointed out, this factor may greatly influence the
76 survival of plants and explained that moisture content along the surface of a given slope varies.
77 That is for; the parts that are not so well exposed to sun lighting retain less moisture than the
78 more illuminated ones and hence a contrast of low-to-high density vegetation growth occurs.

79 The knowledge of underlying bedrock type is crucial for the processing of a photogeologist's
80 work. Any truncation and displacement of beds may effectively reveal fracturing (Odling et al.
81 1999). Abrupt changes in geological formations may also indicate fault zones (Stein 1991). A
82 fault may bring in contact rocks of different petrology and general characteristics that show no
83 physical or inherent association between the two formations. In other words, these formations are
84 not expected to be in contact with each other unless fracturing has occurred at some time in the
85 past (Boggs 2009).

86 Variations in vegetation discussed above, may be also due to petrological differences of
87 underlying rocks, without necessarily the presence of a fault (Rodomsky 2011). One explanation
88 for such variations is that of bedrock influencing soil composition and consequently the plant
89 species that can exist.

90 This study was an effort to delineate the brittle tectonics of Crete islands in Kolymvari area
91 (Greece) by using spatial models and digital image processing. The satellite image that was
92 initially provided in the laboratory was enhanced by a variety of methods for the better
93 discrimination of photolineaments. Certainly, not all lineaments detected, were expected to
94 represent failure of the crust (faults) and could be easily confused with roads and rivers. They



95 had to be studied with respect to real-life conditions. In particular, information related to
96 geomorphology (elevation, slopes, aspect, and drainage system) and geology of the region (types
97 of rocks, boundaries of geological formations) was integrated with the satellite data and the final
98 evaluation of a certain lineament, as a potential fault was carried out.

99 **2. Materials and Methods**

100 **2.1. Study Area**

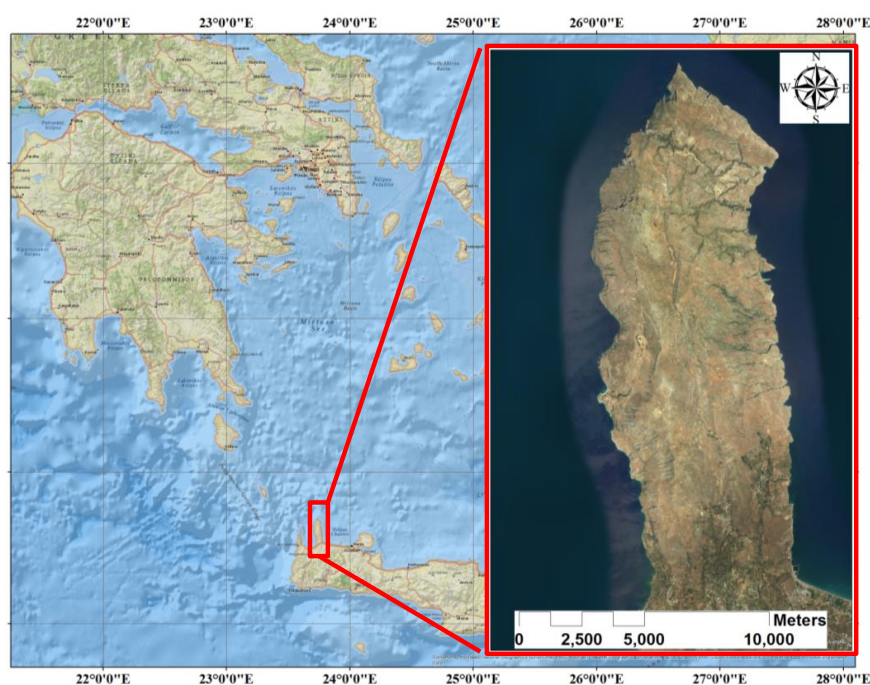
101 The hydrological basin of the study area is situated in the western part of Crete Island and is
102 referred to Municipality of Kolymvari (Figure 1). The landscape structure is fundamentally
103 mountainous, resulting in plain only near the coast. The area has a sub-humid Mediterranean
104 climate with an annual average temperature of 19.96°C (Elhag and Bahrawi 2016). The
105 watershed of Kolymvari is mainly an agricultural area where the most common cultivations are
106 olive trees, citrus trees, vineyards, and vegetables. The area has also light industrial activities
107 such as olive mills, wineries, and other agricultural factories. In the coastal zone of Keritis
108 watershed, there are many touristic units (Papafilippaki et al. 2007). However, the possible flow
109 paths were all directed to Kolymvari Stream, which is, therefore, the surface water body (Elhag
110 et al. 2017). From geotectonic point of view, it has not been easy to classify. Lack of fossils in
111 many cases and difficulties in comprehending the exact geo-processes involved in the formation
112 of the Aegean Sea during the Pliocene and Pleistocene, have been an issue under investigation
113 from many geologists over the past years.

114 **2.2. Remote Sensing Data**

115 The goal of digital image processing was to improve such spectral responses and generate
116 images more interpretable than the original ones, where photolineaments would be better



117 discerned. The digital image of Sentinel-2A (Tile Number: T34SGE, with 0% cloud cover)
118 dating 12/03/2018 and covering the pilot area, has been subject to several enhancement
119 techniques. There were two basic types of distortions that had to be reckoned with, prior to any
120 of the enhancement methods; radiometric distortions and geometric distortions.



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Figure 1. Location of the study area

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2.2.1. Lineaments Evaluation by RS means

124

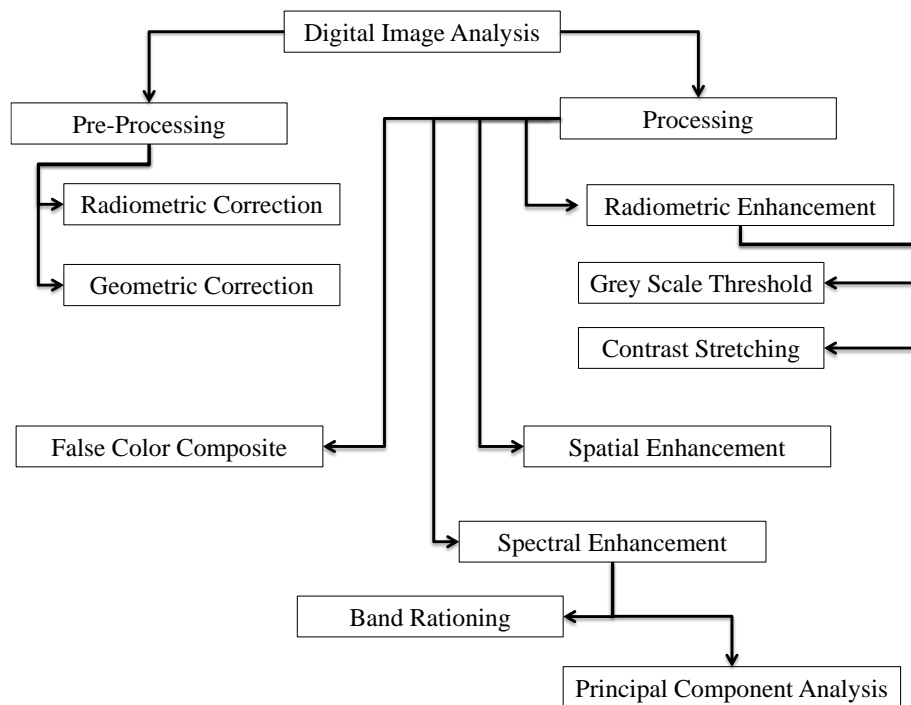
A variety of enhancement methods have been applied to digital raw data in order to improve the
125 spectral characteristics of objects and emphasize the photolineaments of the area under
126 investigation. The original bands have been contrast-stretched for DNs occupy more gray levels
127 than before, in accordance with their frequency of occurrence (histogram equalize stretching).

128

According to Oskoei and Huosheng (2010), the Isotropic Laplacian, non-directional convolution



129 filtering as well as edge-enhancement technique, have been applied to Sentinel-2 (for vegetation
130 and geological features are particularly responsive in these two wavelength bands). Principal
131 Component Analysis aimed to compress information in fewer bands than the original ones,
132 uncorrelated to each other. Images generated by rationing, accounted for the real spectral
133 characteristics and physical properties of objects (topography and illumination effects on the
134 brightness values of pixels, have been eliminated). Schematic flow chart of the implemented
135 procedure in digital image analysis is illustrated in Figure 2.



136

137

Figure 2. Workflow of Digital Image Analysis

138 **Grey level threshold and Contrast stretching**, Sentinel-2, B5 has turned out to be the most
139 appropriate one, for in this near infrared channel, land areas are highly reflected whereas water
140 areas are highly absorbed (Elhag 2017). In 16-bit computer encoding, a digital image can be



141 displayed over a dynamic range of 65536 gray levels (Lillesand et al. 2014). In many cases
142 however, only a small portion of these 65536 levels that are available by devices is utilized. The
143 radiometric ranges (minimum and maximum values of DN values) in each of the six (6) utilized
144 bands (Sentinel-2 B2-B7) of the original dataset are corrected.

145 **Spatial Enhancement**, the concept is based on a moving window (the so-called Kernel window)
146 that in short, contains an array of weighting factors, moves successively over all the pixels of a
147 single black and white image (individual band) and ascribes new weighted DN-values as a result
148 of the weighted original ones (Gupta 2017). A high pass filter, in order to emphasize high
149 frequency features or else local spatial detail (Aiazzi et al. 2002). The filter size was 5*5 and was
150 added back to the fourth band of the original band so that low frequency brightness information
151 was not totally lost.

152 **Principal Component Analysis (PCA)** is a unique mathematical transformation, designated to
153 reduce such redundancy in multispectral data (Kaarna et al. 2000). The idea is to compress all the
154 information contained in the n-original bands, into fewer than N-new channels/components
155 (Lillesand et al. 2014).

156 The high interband correlations (very close to 1) suggested that the best way to proceed with this
157 work was by applying a Principal Component Analysis (Table 1). No pair of bands presented
158 covariance = 0 i.e. no pair of bands were completely independent to each other. Moreover, as
159 covariance > 0, data appeared to be positively correlated (Elhag 2016).

160 **Rationing**, it involves the division as well as more complex functions (additions, subtractions,
161 multiplications) between the DNs of two single bands (Wu et al. 2008). The technique is
162 indicative for both preserving the spectral reflectance characteristics of surficial matter and



163 masking brightness variations derived from illumination conditions and topographic effects
164 (Soulakellis et al. 2006).

165 **Table 1. Interband correlations.**

	Band 2	Band 3	Band 4	Band 5	Band 6	Band 7
Band 2	1	0,950535	0,927525	0,244727	0,6878	0,767209418
Band 3	0,950535	1	0,974041	0,406527	0,7474	0,815756429
Band 4	0,927525	0,974041	1	0,418655	0,8131	0,879577936
Band 5	0,244727	0,406527	0,418655	1	0,607	0,468485418
Band 6	0,687812	0,7474	0,813061	0,607007	1	0,956571297
Band 7	0,767209	0,815756	0,879578	0,468485	0,9566	1

166

167 **False Color Composite (FCC)**, one band at a time can be displayed in each of the three color-
168 guns; generating false - color composite images (objects do not appear in their natural colors)
169 that are undoubtedly more interpretable to human's eye than black and white images. For the
170 present work several combinations have been tried between original bands, stretched ones, the
171 principal component products as well as the ratio images (Pohl and Van Genderen 1998). The
172 scope was to produce a FCC image where Photolineaments would be discriminated to a
173 satisfactory degree for their further digitizing on screen. Table 2 summaries the used band
174 combinations and its uses.

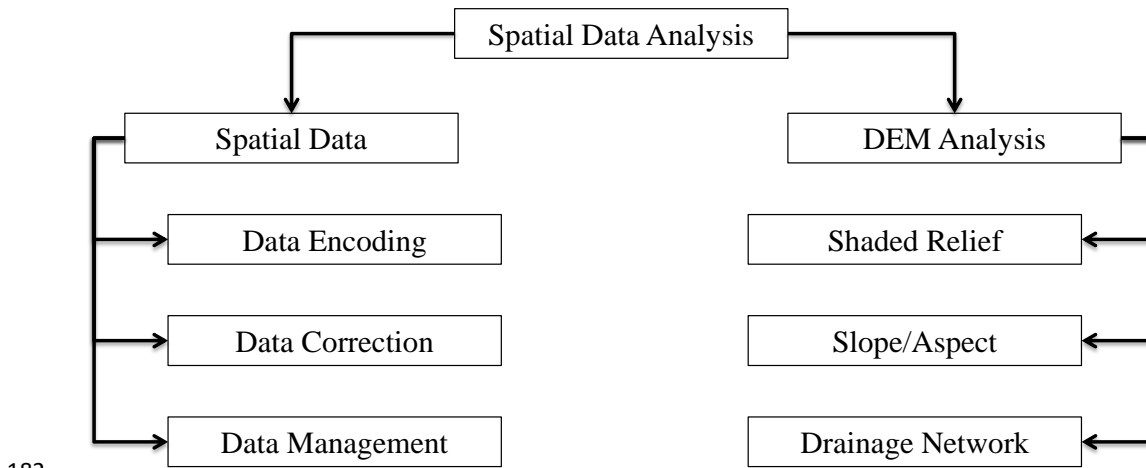
175 **2.2.2. Lineaments Evaluation by GIS means**

176 Basically, not all photolineaments detected using Remote Sensing techniques are represented as
177 fracturing (Karnieli et al. 1996). Cultural features such as roads had to be readily recognizable on
178 satellite data to avoid their confusion with faults and lead to false photointerpretation results.
179 Schematic flow chart of the implemented procedure in spatial image analysis is illustrated in
180 Figure 3.



181 **Table 2. False Color Combinations**

	Band Combination	Uses
FCC1	B4, B3, B2	Vegetation outcrop in false color composite image appeared red. In cases of preferential vegetation appearance along linear/ curve linear features, potential underlying fracturing.
FCC2	PC1, B4, B2	Geology was not easily discriminated whereas drainage network appeared partially enhanced (bluish tones).
FCC3	PC1, B6, B5	NDVI's contribution in the green color-gun was reflected by the very bright green pixels. No other noteworthy remark could be made. Topography was totally lost.
FCC4	PC1, B4, PC2	Vegetation appeared in bluish tones. The drainage lines in the northern part of the study area were well detected.
FCC5	PC1, B5convolved, B3	No clear discrimination of the geology of the area was achieved.
FCC6	PC1, B5convolved, B7stretched	Less certain discrimination of the geological boundaries/geology of the area. No actual discrimination was made.
FCC7	PC1, PC2, PC3	no prediction of the color characterizing each feature was easily made. The overthrust in the northern part was readily differentiated from the rest of the study area.
FCC8	PC1, PC2, B3	The use of two PCs did not give noteworthy results. Bright green pixels represented vegetative coverage, whereas dark tones of green delineated to a limited extend the drainage network. The elongated feature was not enhanced in this product.
FCC9	PC1, B5convolved, PC3	Topography was better expressed than in the previous case (PC1 in combination with PC2). Several lineaments were readily seen (for non-directional filtering of B5 component).
FCC10	B7enhanced, B5convolved, B3	No particular enhancement was observed compared to the previous products.
FCC11	PC1, B5convolved, B5	Several photolineaments were enhanced, topography was well expressed while stratigraphic contacts were not well delineated.
FCC12	PC1, B5convolved, B3	Partial enhancement of drainage network and preferential alignments of vegetation (greenish tones).
FCC13	PC1, B6, B3	Topographic sense was lost. Some vegetation alignments and abrupt changes in green tonal variations.
FCC14	PC1, B5, B3	topography was very well expressed whereas objects appeared in very natural colors and tones. The drainage network was well delineated, and several lineaments were detected. The same difficulty met in previous combinations in discriminating between the diverse geological formations.



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Figure 3. Workflow of Spatial Image Analysis

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3. Results and Discussion

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3.1. RS Lineaments Evaluation

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To transform the original data onto the new PC axes, transformation coefficients (eigenvalues, eigenvectors) should be obtained (Johnson and Wichern 2002). Eigenvalues are presented in Table 3 and describe the variation within the dataset. The % of total scene variance explained by each PC is given by:

190

$$(\text{eigenvalue for PC}_x * 100) / (\text{sum of all eigenvalues})$$

191

Table 3. Eigenvalues - % of total variance

PC	Eigenvalue	% of Variance	Cumulative% of Variance
PC1	900.7121015	82.27270992	82.27270992
PC2	124.3086818	11.354585	93.62729492
PC3	58.20089834	5.316177743	98.94347266
PC4	7.348228513	0.671200789	99.61467345
PC5	3.339351294	0.305022526	99.91969598
PC6	0.879159159	0.080304024	100
SUM	1094.788421		



192 Where $x : 1,2,3 \dots 6$

193 PC1 contains the largest amount of total scene-variance (82, 27%) and hence, is the most
194 correlated component with the original bands. PC2 on the other hand, accounts for a smaller
195 amount of the remaining information (11,35%), PC3 for 5,3% and so on. The three first PCs
196 account for 98, 94% of total scene-variance. As Nikolakopoulos et al. (2008), pointed out, noise
197 is suppressed to the less correlated extracted PCs (Lukáš et al. 2006). Hence, the rest of the
198 principal components (PC4, PC5, PC6) have been ignored.

199 Eigenvector matrix (Table 4) provides 'us with the loadings or else relative contributions, of each
200 of the bands to each of the PCs.

201 **Table 4. Relative contribution of each band to each extracted PC.**

	PC1	PC2	PC3	PC4	PC5	PC6
Band-2	0.256745	-0.43474	-0.45772	0.548567	-0.439	-0.205610634
Band-3	0.179081	-0.19287	-0.33306	-0.06943	0.1385	0.892072198
Band-4	0.314162	-0.29929	-0.39863	-0.407	0.5738	-0.397391284
Band-5	0.259551	0.80314	-0.51111	-0.0477	-0.147	-0.050194624
Band-6	0.761122	0.122543	0.471784	0.346882	0.2476	0.038391621
Band-7	0.395779	-0.1555	0.19195	-0.63718	-0.613	0.004267441

202

203 PC1 has only positive loadings. That means that in the output image no particular feature or
204 structures are expected to be enhanced. It seems that it is a little bit more correlated to Sentinel-2,
205 B7 in comparison with the rest of the bands. PC1 however, bears many similarities to all six
206 bands and can be used as an image of high quality on its own. It is a weighted product of more or
207 less all input images and reflects in a very good way the albedo and topography of the area.

208 PC2 is a contrast between the high positive loading of Sentinel-2, B5 and the negative loading of
209 Sentinel-2, B2. Hence, PC2 as a single black and white image emphasizes the different sets of



210 features that are sensitive to these bands. More particularly, the spectral responses of the
211 vegetation biomass that is present in the scene (Sentinel-2, B5) contrasted to the soil-vegetation
212 differentiations (Sentinel-2, B2).

213 As far as PC3 is concerned, the output image is of less significance, as no clear discrimination is
214 available. Information related to the correlation between the bands and the extracted PC's are
215 also given in Table 5. Factor loadings R, were computed in Excel by using the following
216 formula:

$$217 \quad R(xp) = \text{eigenvector}(xp) * \text{eigenvalue}(p) \% / \text{variance}(x)^{1/2}$$

218 where x refers to the xth channel and p to the pth component (Johnson and Wichern 2002).

219 **Table 5. Factor loadings**

	PC1	PC2	PC3
Band-2	0.778556	-0.48975	-0.35283
Band-3	0.841781	-0.33679	-0.39796
Band-4	0.891965	-0.31568	-0.2877
Band-5	0.623355	0.716576	-0.31203
Band-6	0.985093	0.058921	0.155217
Band-7	0.968296	-0.14133	0.119375

220

221 The correlation coefficient R extends in a range of -1 to 1. This is nothing more than unit-
222 normalization so that loadings are between -1 and 1 (Johnson and Wichern 2002). The closer the
223 coefficient is to -1 or 1, the more significant is the contribution of a channel to the corresponding
224 PC. Similarly, a loading close to zero is an indication of practically no contribution of a band to
225 the corresponding PC. Thus, PC1 is highly correlated to all bands but mostly to Sentinel-2, B6
226 and Sentinel-2, B7. PC2 on the other hand, is highly correlated to Sentinel-2, B5 and may be
227 used instead of the latter for a color-composite image.



228 For vegetation is highly reflected in near infrared and low reflected in visible red, (Sentinel-2,
229 B4)/ (Sentinel-2, B5) ratio generated an image where vegetation appeared in dark tones. In an
230 analogous way, due to the comparatively high reflectance of vegetation in visible green to the
231 lower reflectance in mid infrared, vegetation in (Sentinel-2, B3)/ (Sentinel-2, B6) image
232 appeared in light tones. In both ratios, topography has been eliminated and the area gave the
233 impression of being flat. The two images however, as well as the NDVI product can be safely
234 viewed together for the detection of vegetation biomass. Positive NDVI values, correlate with
235 vegetation while null values correspond to rocks and soil (Johnson and Wichern 2002).

236 (Sentinel-2, B7) + (Sentinel-2, B4) ratio, was based on an idea to enhance stratigraphic contacts.
237 That is for Sentinel-2, B7 is important in geology for the discrimination of geologic rock type
238 whilst one of Sentinel-2, B4's purposes is the detection of geological boundaries where no
239 particular enhancement was noticeable though.

240 Efforts to use others than Sentinel-2, B2 band in the ratio applied resulted in negative pixel
241 values (black pixels within the study area) and hence, could not be used for further analysis. In
242 comparison with the rest of the ratios, $((\text{Sentinel-2, B2})^2 + (\text{Sentinel-2, B5})^2)^{1/2}$ product
243 delineated in a better way vegetation amount while several lineaments were emphasized.

244 Edge enhancement methods resulted in the enhancement of individual bands (black and white
245 images) to a satisfactory degree. However, due to the low capability of human's eye to discern
246 the slight spectral responses of objects in gray scaling, the need for multispectral imagery
247 (display of more than one bands at a time) resulted in the creation of False Color Composite
248 products (FCC).

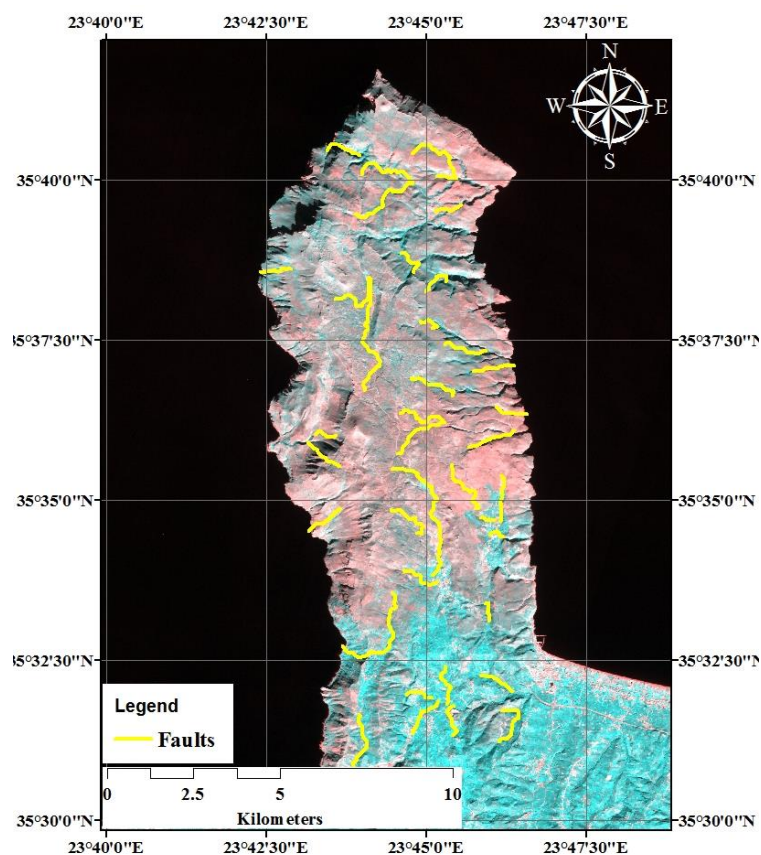


249 A variety of RGB combinations have been tested. FCC14 however, appeared to be the most
250 appropriate one for photolineaments discrimination. PC1 accounted for topography (essential
251 element in photointerpretation) whereas R5 reflected the spectral characteristics of vegetation
252 amount (for Sentinel-2, B5 component of ratio R5). Unfortunately, despite the contribution of
253 Sentinel-2, B7 in the same ratio, stratigraphic contacts were not easily detected except for FCC14
254 for the area under investigation.

255 **3.2. GIS Lineaments Evaluation**

256 The height values were divided in 15 classes and then coded in colors in order to better highlight
257 the relief. The artificial illumination of DEM was from a Northwest direction (315°) and from
258 45° altitude. From statistical point of view, almost one third ($1/3$) of the area is characterized by
259 slopes less than 8.7° whilst only 4,8% accounts for slopes $>43^\circ$. As far as the orientation of
260 slopes is concerned, very few surfaces have a northern-northeastern tendency. The scenery
261 denotes all kinds of aspect of slopes.

262 The geology of the area under investigation with the potential tectonics as resulted from the
263 integration of the enhanced satellite data and the spatial information. The corresponding area
264 show that most of the faults detected on FCC 14 (Figure 4) have been already mapped whereas
265 several more apply for field cross-checking. The integration of Remote Sensing and Geographic
266 Information Systems has proven to be a reliable. method of fault mapping to a satisfactory
267 degree. Only at one side of the mainstream, smaller tributaries develop, which is a very good
268 example of such a valley is presented while the relative displacement of beds at the sides of the
269 streamline, safely enhances the suspicion of underlying fracturing.



270

271

Figure 4. Fault detection using False Color Composition FCC14

272 Further extended fault zone recognized in the study area. It is the same feature observed in all
273 cases of digital image enhancement. The streamline following it, implies underlying fracturing.
274 A part of this zone has been already mapped. It is clear though, that the zone extends over a
275 larger area than the one found and mapped during field investigations.

276 Several more examples that enforced the belief of underlying failure as was initially suspected
277 during photointerpretation, and undoubtedly the drainage network has proven to be an
278 indispensable tool in the evaluation of photolineaments.



279 The final FCC14 that was used for photolineaments detection was composed of PC1, R5 and
280 Sentinel-2, B3 in Red, Green and Blue respectively. R5 on the other hand was a complex
281 function between Sentinel-2, B5 and Sentinel-2, B7. Sentinel-2, B5 is widely used for vegetation
282 amount investigations whilst Sentinel-2, B7 is very common in the field of Geology. Hence, the
283 variant tones of green in FCC14 would normally account for vegetation and/or geology. Since
284 the map of geology of the area was already in hand, in digital form no confusion was made. In
285 some cases, vegetation alignment could safely reveal underlying structural weakness whereas in
286 other cases, truncation and displacement of beds were sufficient for the evaluation of the
287 lineament concerned. Particular attention was paid when a particular lineament connected to a
288 geological boundary.

289 In order to evaluate the photolineaments detected on FCC14 as indicators of underlying
290 structural weakness, certain phenomena accompanying fracture zones had to be reckoned with.
291 Spatial models (GIS) enabled such estimations. Abrupt changes of slope and aspect, streamline
292 sudden bends and straight segments of streams were the criteria used in general terms.
293 Vegetation alignments and drag effects as were identified during photointerpretation, also
294 applied for a few more lineaments to be recognized. The results of faults detection on FCC14 by
295 Remote Sensing and GIS means whereas the differences between the results derived during this
296 work and conventional fault-mapping (during fieldwork) can be noticed.

297 **4. Conclusions**

298 Beyond a reasonable doubt, the integration of Remote Sensing and Geographic Information
299 Systems (GIS) in the case of fault detection in designated study area gave satisfactory results. A
300 variety of enhancement techniques resulted in the discrimination of several photolineaments. Not
301 all of them related to fracturing. Cultural features such as roads were immediately extracted so



302 that no confusion would be made with faults. The same was true for drainage lines and
303 geological boundaries, which in many cases were strongly emphasized and could easily result in
304 false identification.

305 The best combination has proven to be for the first Principal Component (PC1 in red color-gun,
306 accounting for 82.27% of total scene variance), ratio product $R5 = ((\text{Sentinel-2, B5})^2 + (\text{Sentinel-}$
307 $2, B7)^2)^{1/2}$, and one of the original bands (Sentinel-2, B3). For Sentinel-2, B5's contribution in
308 R5, the spectral characteristics of vegetation amount were emphasized (vegetation is particularly
309 responsive in Sentinel-2, B5). Hence, the preferential growth of vegetation along linear features,
310 suggested in many cases underlying fracturing. Despite Sentinel-2, B7's contribution in the same
311 ratio, the spectral responses of geological formations were not highlighted except for very few
312 cases of apparent displacement of beds. PC1 on the other hand accounted for the topographic
313 information that was lost due to ratio's contribution.

314 Spatial modeling (DEM and products) was crucial for the evaluation of photolineaments detected
315 on FCC14 (PC1, R5, in Sentinel-2, B3). The Digital Elevation Model of the area under
316 investigation and the derived maps of slope and aspect enabled several times to imply for
317 fracturing (wherever abrupt changes of slope and aspect were observed at the sides of a
318 particular photolineament).

319 The drainage network has proven to be particularly helpful and informative on the underlying
320 structures. Rift valleys and displacement of rivers' routes were common phenomena within the
321 scene of observation. In many cases parallel vegetation alignments differing from the vegetation
322 outcrop of the surrounding area (abrupt tonal differences) further assured us about possible
323 failure of the crust. Of course, the analysis was not based on the number of criteria satisfied each



324 time. A fault associated with more criteria than another does not necessarily make it a fracture
325 zone of more significance.

326 In comparison with the conventional methods of mapping (field investigations) most of the faults
327 recognized by automated watershed delineation, have already been mapped during field
328 investigations. Several of the zones found, were not discerned on the FCC14. That was due to the
329 following parameters that someone needs to bear always in mind before proceeding with
330 analogous to the present researches.

331 Faults mapping by using satellite data integrated with spatial information (GIS), may lead to
332 quite noteworthy results. Experience is crucial for an accurate photointerpretation. In this case
333 study, most of the faults that have been mapped in the past during fieldwork investigations were
334 also identified on the satellite image. Some of the photolineaments discerned but were not
335 readily seen on FCC14. That should be for all factors explained earlier on. Lastly, for all
336 photolineaments that were identified on the satellite image and raised suspicion of failure of the
337 crust, in- situ data collection for verification of the results is strongly recommended, in order to
338 be total aware of the tectonics of the area.

339

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