3-D visualisation of palaeoseismic trench stratigraphy and trench logging using terrestrial remote sensing and GPR – combining techniques towards an objective multiparametric interpretation

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Abstract

Two normal faults on the Island of Crete and mainland Greece were studied to create and test an innovative workflow to make palaeoseismic trench logging more objective, and visualise the sedimentary architecture within the trench wall in 3-D. This is achieved by combining classical palaeoseismic trenching techniques with multispectral approaches. A conventional trench log was firstly compared to results of iso cluster analysis of a true colour photomosaic representing the spectrum of visible light. Passive data collection disadvantages (e.g. illumination) were addressed by complementing the dataset with active near-infrared backscatter signal image from t-LiDAR measurements. The multispectral analysis shows that distinct layers can be identified and it compares well with the conventional trench log. According to this, a distinction of adjacent stratigraphic units was enabled by their particular multispectral composition signature. Based on the trench log, a 3-D-interpretation of GPR data collected on the vertical trench wall was then possible. This is highly beneficial for measuring representative layer thicknesses, displacements and geometries at depth within the trench wall. Thus, misinterpretation due to cutting effects is minimised. Sedimentary feature geometries related to earthquake magnitude can be used to improve the accuracy of seismic hazard assessments. Therefore, this manuscript combines multiparametric approaches and shows: (i) how a 3-D visualisation of palaeoseismic trench stratigraphy and logging can be accomplished by combining t-LiDAR and GRP techniques, and (ii) how a multispectral digital analysis can offer additional advantages and a higher objectivity in the interpretation of palaeoseismic and stratigraphic information. The multispectral datasets are stored allowing unbiased input for future (re-)investigations.

1 Introduction

Seismic hazard assessment is still predominantly based on the instrumental and historical catalogues of seismicity. However, these catalogues are generally too short com-
pared to the recurrence interval of particular faults (e.g. Wesnousky, 1986; Yeats and Prentice, 1996; Machette, 2000). As a result, the sample from the statistical elaboration of the historical and instrumental data is incomplete and a large number of faults would have not ruptured during the period where the historical record is considered complete (Grützner et al., 2013; Papanikolaou et al., 2015). The need for fault specific studies and the extraction of recurrence intervals from palaeoseismological trenches was then initiated in the late 1970s (Sieh, 1978; McCalpin, 2009). The goal is to extend the history of slip on a fault back many thousands of years, a time span that generally encompasses a large number of earthquake cycles (Yeats and Prentice, 1996).

Over the last few years fault specific studies and palaeoseismology have been further advanced and are now supported by new remote sensing tools that offer high spatial resolution (e.g. LiDAR) and geophysics that extend our data into the subsurface (Ground Penetration Radar (GPR), Electric Resistivity Tomography (ERT)) (Papanikolaou et al., 2015). This manuscript adds on such approaches and shows: (i) how a 3-D visualisation of palaeoseismic trench stratigraphy and logging can be accomplished by combining t-LiDAR and GRP techniques, and (ii) how a multispectral digital analysis can offer additional advantages and a higher objectivity in the interpretation.

Palaeoseismological studies are often undertaken to identify earthquake recurrence intervals and maximum credible magnitudes of prehistoric earthquakes (McCalpin, 2009). These parameters are needed for the accurate calculation of seismic hazard potential of active fault zones (Michetti et al., 2005; Reicherter et al., 2009). Evidence for palaeoearthquakes can be found within the sedimentary architecture of active faults where conditions are favourable for their preservation. Typical features caused by recurrent seismic events include: (i) progressive displacements (Keller and Rockwell, 1984), (ii) colluvial wedges, (iii) Liquefaction, and (iv) fissure fills (Reicherter et al., 2003; Kokkalas et al., 2007; McCalpin, 2009) (see Fig. 1vi). The geometry and stratigraphic position of these features allow the relative dating of recurrent surface rupturing events, whereas carbon rich material (usually within buried palaeosols) can be used to date prehistoric earthquakes and determine recurrence intervals. To access these
potential archives of seismic information expensive trenches are excavated across deformation zones. Then, the classical approach is to document stratigraphy and structure by careful logging, either on paper and/or with photographs (e.g. Wallace, 1986; McCalpin, 2009). The accuracy of the trench log is, however, dependent on the logger’s experience and ability to define mappable units; discrete deposits that are composed of similar lithology need to be distinguished from adjacent deposits.

Palaeoseismic indicators are widely spread and their formation varies along fault strike (e.g. Bubeck et al., 2015). For this reason, geophysical surveys undertaken prior to the trenching phase have become common practice over the last decade. For instance, ground-penetrating radar (GPR) measurements have been carried out to identify optimum trenching locations (e.g. Demanet et al., 2001; Alasset and Meghraoui, 2005; Grützner et al., 2012) and many studies have shown that earthquake related structures can be identified in the shallow subsurface with geophysics (e.g. Chow et al., 2001; Reiss et al., 2003; Bubeck et al., 2015). The excavated trench is then a 2-D representation of the fault zone stratigraphy. It is assumed that the 2-D geometry of the logged sedimentary features continues along strike either side of case; without widening the trench along strike, or excavating more trenches, we must assume that the 2-D trench log is representative for this location along the fault. Trenches target predominantly palaeosols on either side of the fault, and then according to empirical relationships (Wells and Coppersmith, 1994) palaeomagnitudes can be estimated based on these co-seismic displacements. If no or only poorly expressed displaced palaeosols exist the geometry of sedimentary features within trenches is used to estimate previous earthquake displacements. As a “rule of thumb” colluvial wedge thickness equals half of the initial scarp height (e.g. Reichert, 2001; Reiss et al., 2003; McCalpin, 2009). Such information are then used as input parameters for seismic hazard assessment. Therefore, tracing the geometry of these features is essential for the most accurate seismic hazard calculations. A better visualisation can improve the definition of separate unit boundaries and features, offering better interpretations and limiting uncertainties.
In this study we demonstrate how high-resolution t-LiDAR (terrestrial light detection and ranging) measurements and photomosaics can be used to assist in the interpretation of palaeoseismological exposures; we also show how GPR can be used to visualise sedimentary structures in 3-D within the trench wall. The t-LiDAR’s backscatter signal represents material reflectance of radiation in the near-infrared wavelength, and digital photo cameras collect information of the reflectance of visible light; therefore, a quasi-multispectral inspection of the exposures is possible. Ragona et al. (2006) developed a method using imaging spectroscopy on palaeoseismic exposures with hyperspectral and normal digital cameras. As an outcome they were able to enhance the visualisation of the sedimentary layers and other features that are not obvious or even not visible to the human eye. Another study undertaken by Wiatr et al. (2015) places emphasis on the use of the monochromatic laser beam’s backscattered signal to determine varying surface conditions. Using these techniques we make experienced-based trench logging more objective. GPR undertaken on top of the trench and on the vertical trench wall is used in combination with a high-resolution digital elevation model (DEM) from t-LiDAR scanning. This allows radar facies (Neal, 2004) to be distinguished and the sedimentological architecture at depth within the trench wall to be identified. Thus, the resulting 3-D model from the GPR provides information on varying layer thicknesses and minimises misinterpretation due to cutting effects. The workflow comprising data acquisition, statistical analysis, interpretation and storage was calibrated on a road cut on the Island of Crete. We then applied this workflow on a professionally excavated trench in mainland Greece.

2 Geological setting of the study sites

The study sites are both located in Greece, which is one of the most seismically active parts of the Mediterranean (McKenzie, 1972; Le Pichon and Angelier, 1979; Papazachos et al., 2000) due to the presence of the Hellenic Arc and Trench System. Crustal extension orientated both arc-parallel and arc-perpendicular (Mariolakos and
Papanikolaou, 1981; Lyon-Caen et al., 1988) has led to the development of bedrock fault scarps throughout both mainland Greece (Stewart and Hancock, 1991; Benedetti et al., 2002) and the island of Crete (e.g. Wiatr et al., 2013). These normal faults mainly consist of footwall Mesozoic carbonates juxtaposed against hanging-wall flysch and/or post-alpine sediments. Earthquake features such as colluvial wedges (a consequence of degradation of the scarp), fissure fills and displaced strata occur within the hanging-walls of these faults and datable material may be contained within buried palaeosols (see Fig. 1vi) (McCalpin, 2009). To create those archives and preserve them over geological timescales, erosional processes must be lower than the rate of tectonic activity. These features therefore represent geological archives of palaeoearthquakes because they can record information about Holocene and Late Pleistocene earthquakes (e.g. Morey and Schuster, 1999; McCalpin, 2009). Ambraseys and Jackson (1990) estimate a maximum earthquake magnitude of $M_s 7.0$ could occur on these normal faults, which coincides with fault segment lengths of 15–30 km as determined through empirical relationships (Wells and Coppersmith, 1994).

2.1 The Sfaka Fault (NE Crete, Greece)

The island of Crete is the largest within the Greek territory and is directly adjacent to the subduction zone between Europe and Africa. The NNE–SSW trending Sfaka fault is located in northeastern Crete (Fig. 1i) and forms the easternmost segment within the Ierapetra Fault Zone which is a major tectonic line of approximately 25 km cutting through the whole island (Gaki-Papanastassiou et al., 2009). This northwest dipping normal fault is easy to recognise as it offsets smooth mountain slopes, has a steeply dipping (ca. 70°) fault scarp up to 6 m in height, and has an onshore length of approximately 5 km (Fig. 1iv). Together with the opposing Lastros fault a 2 km wide graben structure is formed.

An outcrop in the form of a road cut (located at 35°7′58.97″ N, 25°54′26.01″ E) exhibits the fault zone as a contact between footwall Mesozoic carbonates and hanging-
wall colluvium (Fig. 1v). The outcrop cuts the fault at an angle of approximately 75° from the fault strike.

2.2 The Kaparelli Fault (Gulf of Corinth, Greece)

The Kaparelli fault is located in the easternmost part of the Gulf of Corinth (see Fig. 1i) which is associated with rapid extension oriented N-S (e.g. Papanikolaou and Royden, 2007). The Kaparelli fault became well-known as it ruptured during the Corinthian Alkyonides earthquake sequence in spring 1981 (Jackson et al., 1982). Many palaeoseismological studies using various approaches have been undertaken along this ca. 20 km long south dipping normal fault. For example Benedetti et al. (2003) used 36Cl cosmic ray exposure dating to determine the history of surface rupturing events on the 4–5 m high limestone scarp of the Kaparelli fault. Their results show evidence for seismic activity $20 \pm 3$, $14.5 \pm 0.5$ and $10.5 \pm 0.5$ ka prior to the 1981 earthquake sequence. A palaeoseismological trenching study was conducted by Kokkalas et al. (2007). The authors found evidence for at least three events in the past 10 000 years: $9370 \pm 120$, $7290 \pm 140$ and $1165 \pm 105$ a. The excavations from Kokkalas et al. (2007) are still open; therefore, the already logged and interpreted structures within trench Kap-1 (Fig. 1v) is a perfect site to test the workflow developed on the Sfaka fault road cut.

3 Methodology

The herein presented workflow combines palaeoseismic trenching techniques with t-LiDAR measurements to improve the accuracy of palaeoearthquake reconstruction. A multispectral analysis of t-LIDAR backscatter data and the luminescence of true colour photographs were compared to the manual trench log. A GPR survey was then conducted to obtain 3-D information of layer continuation and thickness in depth within the trench wall (Fig. 2).
3.1 Palaeoseismic trenching

A palaeoseismic trench is characterised by an often artificially produced subsurface exposure of sedimentological coseismic features. To accurately interpret these features, apparent dips and anthropogenic and/or exogenous influences must be excluded. Moreover, sketching lithological contents requires an exposure devoid of weathered and smeared parts that were caused by the excavation (McCalpin, 2009). To simplify and prove the geometrical correctness of the trench log, a reference grid of one square metre was attached to the wall. The grid’s points of intersection also act as reference points for remote sensing applications.

The trenches were conventionally logged in 1 : 10 scale in accordance with McCalpin (2009). Thereby, discrete deposits that are composed of similar lithology considering consistent texture, sorting, bedding, fabric, and colour of individual layers are mapped. Photographs of every square metre were taken and later stitched together using an automatic panorama recognising tool including a manual editor of control points and straightening functions (Autopano Giga, Kolor). It must be noted that error values are already stored within image information due to differing luminous exposures; furthermore, holes and protruding boulders create shadows that partially change the reflection characteristics of certain sedimentological features. The Sfaka road cut faces north (see Fig. 1iv and v) and is surrounded by steep slopes. In Kaparelli the eastern trench wall (see Fig. 1ii and iii) was investigated because it preserved the best stratigraphy and exhibits faulting events with clear marker horizon displacements (Kokkalas et al., 2007). To avoid most of the differing luminous exposures, the photographs were either taken in the morning when the angle of sunlight was shallow and did not shine directly onto the investigated wall (Kaparelli) or in the afternoon when the sun disappeared behind the surrounding hills (Sfaka).

The photomosaic of true colour images (RGB; red, green, blue) was converted into a grey-level image to eliminate hue and saturation information while retaining the luminance (0–255) using the rgb2gray function in MATLAB®. In a GIS the resulting im-
age was georeferenced to a custom frame in order to make it comparable to all other datasets of this study.

### 3.2 t-LiDAR measurements

t-LiDAR (terrestrial Light Detection and Ranging) is a remote sensing technique with high spatial and temporal resolution and is a very effective instrument for reconstructing morphological and geological settings and monitoring approaches. A generated coherent laser beam with little divergence by stimulated emission is reflected off surfaces and the proportionate backscattered signal is detected, forming a non-contact and non-penetrative active and stationary recording system. Thus, from measuring the two-way-travel time (TWT) of a first pulse detection sequence, 3-D surface data is acquired. The illuminated area is controlled by wavelength, beam divergence, range between sensor and target, and also by the angle of incidence (Jörg et al., 2006; Wiatr et al., 2015). In our study we used an ILRIS 3-D laser ranging system (wavelength $\lambda$ is 1500 nm) from OPTECH Inc., Ontario, Canada.

The limitations of using t-LiDAR are high humidity (e.g. Lobell and Asner, 2002) and low target reflection with cumulative distance and shallow incident angle (e.g. Höfle and Pfeifer, 2007). In order to assume constant soil moisture and to ensure the backscatter signal data quality, close range scans were done during the summer in dry conditions within a few hours. The scans were carried out almost perpendicular to the trench wall and less than 10 m from the exposure.

Other benefits of applying t-LiDAR is its flexibility, the relatively quick availability of an actual dataset, and also its high spatial resolution with information about backscatter signal each referenced in $x$, $y$, $z$ coordinates. The result is an irregular but dense point cloud representing a highly detailed digital 3-D surface model which can be easily implemented in geographical information systems (GIS) to generate accurate digital elevation models (DEM) or digital terrain models (DTM) (e.g. Wiatr et al., 2015).

For this study, the t-LiDAR scanning was undertaken at both close range and long-mid range to determine geometrical relationships between the footwall, hanging-wall,
prolongation of the scarp, and trench wall (see Fig. 2). The backscatter signal of the t-LiDAR results from the reflection of transmitted waves of near-infrared light. In other words, each measurement is usually accompanied by a surface remission value, which quantifies the intensity of the reflected laser beam. The monochromatic backscatter signal values are stored as grayscale values from 0–255. The information on the monochromatic wavelength and the detected backscattered signal in the near-infrared reflects the surface properties which are invisible to human eyes. Thus, the backscatter signal was also used for the multispectral analysis. The raw-data was cleaned from isolated points and those that do not represent the area of interest. The mathematical and geometrical alignment of the different scan windows was then carried out. For project specific demands, the datasets were translated into a custom grid. The long-mid range data were used for the overall geometrical analysis creating high-resolution DEMs with a resolution smaller than 0.1 m. Data from the close range scan were processed for statistical calculations of the backscatter signal’s spatial distribution. A detailed description on the applied workflow is given in Sect. 3.3.

3.3 Imaging spectroscopy

Visualising an array of simultaneously acquired images that record separate wavelength intervals or bands is part of multispectral analyses. A common multispectral camera employs a range of film and filter combinations to acquire photographs that record narrow spectral bands of non-imaging data. Reflectance spectra map the percentage of incident energy (e.g. sunlight) that is reflected by a material as a function of energy wavelength. Absorption of incident energy is represented by downward excursions of a curve (absorption features). Upward excursions represent superior reflectance (reflectance peaks). These features are valuable clues for recognising and distinguishing certain materials (Sabins, 1996). Multispectral imaging, or imaging spectroscopy, has been used at many different scales for remote sensing. Probably the most prominent example for macro-scale investigations is the inspection of visible and near-infrared satellite imagery for mapping and monitoring vegetation (e.g. Tucker, 1979;
De Rose et al., 2011; Nouri et al., 2014). Ragona et al. (2006) introduced an application of high-resolution field imaging spectroscopy on paleoseismic exposures using hyperspectral and common digital photo cameras. The authors conclude that imaging spectroscopy can be successfully applied to assist in the description and interpretation of palaeoseismic exposures because: (i) subtle or invisible features are displayed, (ii) quantitative analysis and comparisons of units using reflectance spectra can be undertaken, and (iii) unbiased data are stored for future access and analysis.

The limitations of multispectral approaches are, by their very nature, closely connected to the application of photomosaics and t-LiDAR measurements. We re-emphasise the influence of moisture; where present it not only causes a darkening of the sediments (reduction in reflectance), but there is also a hard-to-quantify content variation across the exposure (Ragona et al., 2006). Another error source appears due to morphological characteristics of a certain exposure, especially on surfaces that are not well prepared for palaeoseismic investigations and data collection. This means that the exposure must be flattened and cleaned to avoid changes in spectral amplitudes accompanying changes in illumination angle and distance.

To reduce errors we assume that the moisture content was similar throughout the exposure and water absorptions should not affect the correlations because the spectral change is similar along the trench wall. Furthermore, the photos and t-LiDAR scans were taken almost perpendicular to the exposure so that optimal data quality can be expected.

The workflow contains geo-referencing and snapping the high-resolution raster data from the photomosaic and t-LiDAR backscatter signal to a coherent cell size (0.001 m) in a GIS. Afterwards, an iso (iterative self-organising) cluster unsupervised classification was applied to a two-channel composition of both raster layers. Thereby, the number of classes was set to ten times the amount of included bands (photomosaic grey-level image and t-LiDAR backscatter signal image) as this provides sufficient statistics and enough cells to accurately represent a certain cluster. This type of clustering uses a process in which all samples are assigned to existing cluster centres during each iter-
ation; new means are then recalculated for every class. The actual number of classes is usually unknown; therefore, we started with 20 classes and analysed the attribute distances between sequentially merged classes with the dendrogram method (hierarchical clustering). This reduces statistical misclassifications and provides information on distinct classes. Based on the outcome, classes which are statistically closest get merged and the dataset gets reclassified. Block statistics within a \(3 \times 3\) cell environment are applied to erase noise by overwriting cell values to all of the cells in each block with the median value (Fig. 3). Moreover, resampling down to 0.02 m cells enhances visibility and allows a more general interpretation and comparison to the conventional log. This is because average gridding and sketching inaccuracy is around 2\% (McCalpin, 2009).

3.4 Ground-penetrating radar

GPR is a non-invasive and non-destructive geophysical technique that operates with high-frequency electromagnetic waves in the radio band to detect electrical discontinuities in the shallow subsurface up to approximately 50 m. Every GPR measurement contains a five-step process of: (i) generating, (ii) transmitting, (iii) propagating, (iv) reflecting, and (v) receiving electromagnetic pulses. The differing relative dielectric permittivities (\(\varepsilon_r\)) of varying materials control the transmitting velocity in relation to the speed of light (\(c = 0.2998 \text{ m ns}^{-1}\)) once the pulse is emitted from the antenna. Fractional reflections of the pulse on inhomogeneities and layer boundaries get received due to a dielectric contrast. In order to calculate depths of reflection the TWT (two way travel-time) is recorded in the order of nanoseconds. Depending on the frequency of the antenna, objects smaller than 0.1 m in diameter can be resolved. Common GPR systems perform at frequencies between 50 MHz and 1 GHz, where achievable resolution is a quarter of the wavelength. The relationship between penetration depth and spatial resolution is an inverse one; hence, a higher spatial resolution occurs at the expense of penetration depth and vice versa (Neal, 2004; Schrott and Sass, 2008).
Water is almost the only limiting parameter for the application of GPR because of its high relative dielectric permittivity. Moisture content dramatically decreases the electromagnetic wave velocity by stronger attenuation and leads to reduced penetration depths (Schrott and Sass, 2008). Soil moisture differences often severely disrupt wave energy, which makes it even more difficult to interpret reflections. Dielectric contrasts are the main features of the GPR image interpretation, since any dielectric discontinuity is detected. Thus, targets can be classified according to their geometry and reflection facies.

GPR was carried out on the vertical trench wall and on the slope surface above the trench (see Fig. 2). In order to make the GPR operationally effective, our survey provided efficient coupling of electromagnetic radiation into the ground and a sufficiently large scattered signal for detection at or above the ground surface. Furthermore, a 400 MHz antenna together with a SIR-3000 control unit from Geophysical Survey Systems Inc. (GSSI, Salem, NH, USA) was used to obtain desired resolution and noise levels. The data processing was done using the software ReflexW® (Sandmeier Scientific Software, Karlsruhe, Germany) involving the following processing sequence: remove header gain, move start time, energy decay, 1-D bandpass frequency, background removal, and average \( xy \). Reflection hyperbolas of gravels were used to estimate wave velocity. Data migration was undertaken to correct angles, because dips are usually underestimated due to a complex 3-D cone in which electromagnetic energy radiates (Neal, 2004).

Based on distinct layers in the trench log and taking into account the results of the multispectral analysis, GPR-data were then used to interpret the outcropping strata in 3-D.
4 Results

4.1 Sfaka Fault, Crete

4.1.1 Trench log

In accordance with McCalpin (2009) the trench was logged and divided into ten distinct layers. These vary in colour, matrix specifications, geometrical alignments and soil formation. As seen in Fig. 4a, two palaeosols that depict fissure fills are observed in the trench wall. These layers represent hanging-wall sediments, rather than material from the footwall. Overlying deposits rapidly filled ground cracks that occur during a rupturing event. Both palaeosols contain a combination of fine-grained and gravel sized material. Colluvial layers of gravels of different colour and component size and orientation complete the hanging wall’s architecture to its western end (see Fig. 4a, C1–C6). However, C1 is made of heavily cemented colluvial material and thus will not be further addressed. Adjacent to the bedrock fault plane towards the eastern end of the trench wall, fault gouge of approximately 1 m thickness is exposed. However, true thickness is calculated to around 0.8 m when correcting for the trench’s 75° from fault strike. The yellowish light coloured fine-grained cohesive matrix obviously differs from other sediments within the hanging wall exposure (Fig. 4b).

4.1.2 Imaging spectroscopy

The greyscale photomosaic stores visual impressions in a way similar to the human eye and represents a weighted sum value of luminance within the range of visible light per pixel. Luminance at 1500 nm detected by t-LiDAR significantly differs in some parts of the trench wall (Fig. 4b and c). As shown in Table 1, the light fault gouge material is highly reflective in both photomosaic and HRDBSM.

The homogeneous silty layer contains only a view voids due to excavation works that influence reflectance value range. Resultant colorimetric shift expressed by the 2-
component composition almost solely depicts the highest value ranges for this part of the trench wall (Fig. 4d). In contrast, the cemented colluvium to the west is highly irregular in the sense of reflectance. Both photomosaic and HRDBSM show a heterogeneous greyscale value distribution that is even more embodied by high-grade contrasts in the two-channel composition. Similar observations occur for larger boulders that protrude out of the trench wall (Fig. 4a–d).

Colluvial layers C2–C5 are distinctively different in their reflectance characteristics. Where transition between both units is indeed visible in the photomosaic, a sharp contrast in reflectance characteristics of near-infrared is recognisable. Moreover, the named colluvial deposits do not only appear as a conglomerate of diffuse values but show evidence of alignments. An upward oriented structure of approximately 0.5 m thickness is obvious in HRDBSM and false colour composition. The structure follows a lineament of displacement within the colluvial strata.

Figure 5 visualises percentages of seven classes, estimated from the unsupervised classification on individual identified layers within the trench log. Either the majority of a certain layer is fulfilled by one single class or by a certain composition of two or three classes. Where Table 1 shows the dominance of high values within the fault gouge layer, the illustration of unsupervised classification proves this layer to be almost completely (70 %) represented by one single class (7). Although class 4 covers almost the same value range as class 7, fault gouge exposure is only covered by 8 % by class 4 (see Fig. 5a).

In the unsupervised classification, the fault gouge is the only layer in this trench wall where the majority is covered by one single class. Palaeosol I and palaeosol II have a similar ratio of effecting classes but class 6 is not present in the palaeosol II signature, allowing them to be differentiated. By visualising the spatial arrangement of influencing classes the differentiation between these two layers is even better (Fig. 4e). While Fig. 5b only shows percentage significance of class ratios per layer, the spatial distribution promotes the reconstruction of a certain layer. The accumulation of class 5
especially in the lower part of palaeosol I is obviously different from any other cluster in palaeosol II, although quantitative statistics conclude a similar composition of classes. Except for C6, which is well represented to around 80% by class 1 (31%) and 2 (52%), and C1, which appears as a unsorted conglomerate of classified responses, the remaining colluvial lithologies appear with similar ratios, especially classes 1, 2, and 5. In a quantitative way no distinction can be recognised. Also, large scale clustering of classes within the layers is absent. However, arrangements, especially of class 7, are obvious and coincide with coarse-grained gravels within the colluvium. Within C3 a micro-cluster of approximately 25 pixels are arranged along a slightly bent line dipping about 50° towards the footwall. A similar arrangement of class 7 with an even smaller cluster (3 x 3 pixels) and wider spread is indicated in C5 dipping 15° towards the footwall. Furthermore, the surrounding matrix is slightly more expressed by class 5 in C5, whereas C3 has subjectively no preferred matrix content (Fig. 4). Alterations are expected to decrease with increasing depth. Dependent on rock composition and mean annual precipitation, the formation of new minerals is commonly related to depth from surface. C4 does not show any spectroscopical attribute except for a complete absence of class 6 and low range greyscale values (see Fig. 5a). Clasts or large boulders protruding out of the trench wall are represented by intermediate value range class 5 on top and wide value range class 1 at the bottom (Fig. 4c).

4.1.3 GPR

Using the trench log and multispectral information enables radar facies to be distinguished. Figure 6 confirms the distinction of individual layers by comparison of reflected electromagnetic signal intensity. Reflections of visible and near-infrared light within certain zones that fit with trace increment and dimensions of the GPR system (30 cm x 2 cm) were sampled and correlated with the radar’s first arrival. As the vertical resolution is a quarter of the wavelength λ (here: 30–40 cm), we averaged reflection amplitudes for 9 cm into depth per trace.
A good correlation between backscattered signals of both passive and active methods is obvious in some parts. A significant contrast in all three datasets is traced by the abrupt transition from fault gouge to palaeosol I (see Figs. 4c and 6). Where reflections of visible and near-infrared light are intense on the surface of the fault gouge exposure, they rapidly decrease in signal strength on the palaeosol surface. The opposite reflectance behaviour is observed for radar reflections in the very shallow subsurface; the first lithological transition is characterised by the change of low to moderate reflection amplitudes in the fine-grained homogeneous fault gouge to higher reflection intensities from heterogeneous palaeosol I.

Moderate reflectance with intermediate variance designates the exposure of palaeosol I. A slightly decreasing trend is obvious within this section just before an abrupt rise in both visible and near-infrared light reflection values. This changeover is not obvious from GPR mean values. However, value range given by standard deviation reach wider than the in previous section. Moreover, there is little distinction between individual colluvial deposits from GPR reflection amplitudes.

As previously stated, the HRDBSM shows an unrecognised feature in the middle of the trench exposure. A change is proven by a drastic drop in reflections from the GPR signal approximately 3 m from the fault plane. At the same position there is also a minor photomosaic and HRDBSM value decline. Thus, a conspicuous progression similar to a Gaussian bell shape curve in the middle of a dataset is obvious.

Layer C1 is not individually considered since the coupling of the antenna on heavily weathered cemented material with rugged surface relief was not sufficient. However, other transitions recognised in trench log and imaging spectroscopy can be traced in GPR images. This then leads to a 3-D model of coseismic features within the hanging-wall (Fig. 7). Seven out of the ten (boulders are not included as an individual layer) previously mapped units plus the limestone fault plane to the West and the adjacent loose material to the East can be traced at depth using GPR.

The 3-D interpretation from GPR images visualises the continuation of distinct layers observed from multispectral analysis into depth. The limestone fault plane and fault
gouge clearly differ in GPR images. Also, the cemented colluvium C1 is characterised by continuous and high amplitude reflections. Coarse grained components within other colluvial layers are represented as signal scattering hyperbolae within a homogeneous matrix facies. However, a distinction between C4, C5 and, C6 could not be done with these data. The two palaeosols differ in the recorded intensity of the reflected electromagnetic waves. Where palaeosol I is characterised by high amplitude reflections, palaeosol II contains only minor reflection hyperbolae caused by small clasts within the homogeneous matrix.

4.2 The Kaparelli Fault, Gulf of Corinth

The description of the Kaparelli fault trench follows the lithological designations of Kokkalas et al. (2007). The hanging-wall and footwall of the Kaparelli fault are clearly separated by a 70–80° south dipping fault zone. This zone is characterised by a chaotic assemblage of sheared deposits and material from surrounding or overlying units that has fallen into cracks and fissures. The footwall consists of multi-coloured pebbly-cobbly gravel deposits with a wide range of coarse-grained sub-angular to well-rounded clasts in a silty cemented matrix. The hanging-wall block comprises thick deposits of sandy silt (loess deposits) with many steeply dipping fissure fills, some cutting the entire trench wall and others only partly. The fissure thicknesses ranges from around 10 cm to over 80 cm and are filled with sub-angular to rounded gravel deposits in a silty matrix (Fig. 8a).

The manually sketched trench log, calibrated using the results from Kokkalas et al. (2007), correlates well with the results from imaging spectroscopy (Fig. 8a). Coarse grained parts of the exposure to the northern end exhibit a widespread range of greyscale values in both photomosaic and HRDBSM. Due to a grain size in the order of tens of centimetres and the resulting rough relief, shadows are generated in such a way that significantly influences the colour texture of the photomosaic and the backscattered signal. However, a distinct transition to a silty-sand unit, which prior to this study was described as the fault zone of the 1981 rupture event (Kokkalas et al., 2007).
2007), is obvious. Few and much smaller clasts in this unit (diameter is about 1 cm, < 15%) and a homogeneous matrix have led to a uniform display in the false colour image. This composition of concurrent greyscale values in the photomosaic and the HRDBSM occurs three times in constant offsets along the trench exposure. Pure silt underlies the silty sand. A fissure fill structure of pebbly gravel, dipping about 70° to the south, separates the two blocks of silty-sand and silt layers. Again, a rougher relief leads to a large range of backscattered signal values from both active and passive systems. However, sharp delimitations of juxtaposed lithological units based on their spectroscopic appearance are clear and discernible. A buried soil horizon and a colluvial wedge resulting from the 1981 surface rupturing event (Kokkalas et al., 2007) are visible and clearly textured by a certain composition of greyscale values.

In Fig. 8b, a three-dimensional reconstruction of the trench wall shows that exposed structures do not only occur on the surface but are also traceable into the hanging-wall. Using layer differentiation from imaging spectroscopy helps to recognise certain radar facies even when there are only subtle distinctions. Major components of the trench wall are identified in individual GPR images. Their three-dimensional extension information is assembled by interpolating between multiple overlaying GPR images. Hence, information on continuation into depth as well as the varying thicknesses of individual layers is gathered. For instance, the colluvial wedge has only a minor variation in its thickness to 2 m penetration. The estimated average for this unit is 0.6 m.

5 Discussion

Trenching investigations have been one of the established methods in palaeoseismic research for the last decades. However, the outcome is highly dependent on the ability of the trench logger to define mappable units and the influence of sunlight. Furthermore, producing an accurate log and interpretation requires experience and excellent sketching skills. This process can be enhanced using the outcome of a numerically and...
a multispectral view of the palaeoseismic exposure, which allows quantitative information to be assigned to mapped units within the trench wall.

There are some significant disadvantages of passive data collection imaging techniques. These are mainly due to differing angles of illumination because the trench exposure is not a perfectly even surface at all scales; at larger scales surface undulations dramatically increase. Thus, the lightest parts in the photomosaic, visualised for the Sfaka roadcut as class 7 with an average value of 221, mainly represent a high matrix luminance and the top (bright) sides of boulders and clasts. Rectification and parallax effects yield an additional error in the order of a few centimetres. High-resolution 3-D images and the near infrared backscatter signal from t-LiDAR provide information on the physical properties of materials. Colour, matrix specifications, geometrical alignments and soil formation features influence the t-LiDAR backscatter signal. A multispectral approach, using unsupervised clustering on both spectra supports the results from the trench log and complements the findings. Thereby, a distinct layer signature given by particular compositions of effecting classes allows adjacent stratigraphic units to be differentiated. Some areas within the multispectral image lack evidence for distinct spectroscopical characteristics. However, these areas can still be defined when they are adjacent to areas with static characteristics; the boundary between two areas is clearly defined as long as one area can be classified using the unsupervised clustering. Therefore, a spectroscopically inconspicuous and completely heterogeneous area surrounded by regions with static characteristics is still sufficiently confined. Within a given error range due to manual gridding on the trench wall, georectification and blending pixels of the photomosaic data, the results show many resemblances to the manually drawn trench log.

The results of the imaging spectroscopy verified the lithology of the trench wall and the resulting image from the unsupervised classification serves as a calibration factor for GPR measurements. Due to the GPR’s resolution being about 0.1 m, the calibration is necessary to recognise and interpret minor differences in sedimentological compositions. This method allows more accurate calculations of geometric layer thicknesses to
be made, which are needed to correlate the amount of vertical offset caused by a specific surface rupturing event (e.g. Reiss et al., 2003). The quality of the 3-D GPR image and its interpretation depends on well-structured data acquisition and processing, as well as on the experience of the operator. The coupling of the antenna to the surface is decreased on bumpy surfaces, which leads to lower quality data. Moreover, the reference grid on the surface poses a source for stumbling. However, the grid is needed to fuse the geophysical data with remotely collected data and to locate the GPR images in three-dimensional space. An alternative to a grid made of string is colour spray to mark locations for orientation; but, these would have a significant impact on the results from imaging spectroscopy. When the survey is accurately planned and organised good results can be obtained which allow a 3-D interpretation of sedimentary features to between 2 and 3 m depth within the trench wall.

The biggest disadvantage of the presented workflow is by far the effect of sediment moisture content on reflectance, both in the multispectral analysis and GPR survey. For the multispectral analysis there is not only darkening of the sediments, which leads to an overall reduction of reflectance, but significant partial absorptions at wavelengths near 1.4 and 1.9 µm is also common (Lobell and Asner, 2002; Ragona et al., 2006). Moreover, water content in a given medium leads to distortion effects and high attenuations of electromagnetic waves (Neal, 2004; Schrott and Sass, 2008). However, for conditions when the moisture content is similar throughout the trench wall, water absorptions should not affect the correlations because reflectance along the wall should be affected uniformly. Ragona et al. (2006) have shown that identifying stratigraphy with samples that maintain high amounts of their original moisture content is possible; however, we reiterate the authors’ suggestion to consider necessary approaches to minimise changing reflectance.

Other potential error sources using this technique are dependent on the characteristics of the individual trenching sites and the equipment used. Some sites are hard to access because of steepness, height and/or width of the excavation. Extremely steep or narrow trenches make the installation of the scanning equipment difficult. Exposure
heights exceeding usual body heights generate problems for the GPR survey; these can be overcome using ropes and wooden tools to ensure good coupling. Scaffolding usually consists of metal which may lead to interferences in the GPR image. If the trench wall is not properly prepared in terms of cleaning, or the embedded sediments produce a rough surface because of coarser grain sizes, spectral amplitudes will change because of varying illumination and incident angles. Therefore, the spectroscopic interpretation must take these accompanying effects into account. Moreover, extremely complex sedimentological architectures may cause complicated multi-pathing effects on the radar waves. The presented workflow has basic requirements concerning computing capacities; the collected high-resolution data from conventional photo cameras, t-LiDAR scanning and GPR measurements engage substantial disk space and random access memory.

One major benefit from this workflow is the storage and future use of the raw data. The majority of paleoseismic trenches are designed to be closed after field investigations are completed. This means that not only is there no future access to these exposures, but the sedimentological environment of the excavated site is also destroyed. If a trench is left open after field investigations, the trench walls will get degraded and altered by weathering effects. t-LiDAR and GPR measurements provide and store information on the visual appearance of the trench and the reflection properties of different electromagnetic wavebands. The reflectance spectrum at each pixel of an image provides unbiased compositional information. This saved data can always be used for future (re-)analyses.

6 Conclusions

Identifying and mapping individual lithological units along a palaeoseismological exposure in accordance with colour and matrix specifications as well as sedimentary structures and soil formations are core competencies of palaeoseismic trenching studies. However, the accuracy and quality of the log and interpretation is highly dependent on
the experience of the trench logger, and is thus subjectively influenced. Hence, (minor) differences in lithological description from expert to expert are expected, especially if one logger has access to no more than a photomosaic. In order to prove whether conventional trench logging methods used to map coseismic features in a palaeoseismic trench wall can be objectively enhanced, we created an accurate digital version of the exposure and its physical properties. This was done by combining routine logging with vertical GPR measurements and imaging spectroscopic approaches from normalised photomosaics and high resolution t-LiDAR backscatter models. Both the studied palaeoseismic exposures, on Crete and in mainland Greece, exhibit sedimentary structures whose constituent parts and shape are essential information for a palaeoseismic reconstruction.

After the conventional trench logging was completed, t-LIDAR scans were undertaken at close range. The near-infrared backscattered signal was combined with a luminance bearing photomosaic of the same trench wall. Statistical and classification techniques reproduce an objective digital copy of a palaeoseismic trench log. In order to define distinct units, four options to characterise and differentiate individual layers by imaging spectroscopy can be registered:

- Significant dominance of a certain class within a distinct layer.
- Certain composition with spatial clustering.
- Certain composition with certain arrangements.
- Distinct borders between individual layers although one or both are not determined by applied statistics.

Subtle or invisible features are enhanced and become part of a quantitative analysis, and comparisons of units using their reflectance on certain wavelengths (see also Ragona et al., 2006) can be carried out. Our results show that based on distinct layers in the trench log, in combination with the outcome of imaging spectroscopy, a 3-D-interpretation of GPR-data carried out vertically on the trench wall is possible. Hence,
the spatial extent of palaeoseismic features can be traced within the trench wall. The resulting 3-D model from the GPR provides information on representative layer thicknesses, displacements, and geometries. This is highly beneficial since it minimises misinterpretation due to cutting effects.

Reconstructing the palaeoseismological history of both trench exposures is not an integral part of this paper, but this research has shown that recognising individual event layers can be improved using multispectral viewing and 3-D visualisation of GPR images. This method can therefore contribute to the accuracy of seismic hazard assessment.

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References


Table 1. Median greyscale values of photomosaic, high-resolution digital backscatter model (HRDBSM) and 2-component composition per stratigraphic unit from the trench log. The composition is the result of allocation of both, photomosaic and HRDBSM in equal parts, to visualise certainties and their variation within given zones. Error is given by single standard deviation.

<table>
<thead>
<tr>
<th>Layer</th>
<th>Photomosaic</th>
<th>HRDBSM</th>
<th>Composition</th>
</tr>
</thead>
<tbody>
<tr>
<td>recent topsoil</td>
<td>132 ± 21</td>
<td>197 ± 12</td>
<td>138 ± 23</td>
</tr>
<tr>
<td>fault gouge</td>
<td>221 ± 19</td>
<td>239 ± 12</td>
<td>224 ± 23</td>
</tr>
<tr>
<td>palaeosol I</td>
<td>165 ± 19</td>
<td>170 ± 13</td>
<td>87 ± 23</td>
</tr>
<tr>
<td>palaeosol II</td>
<td>156 ± 19</td>
<td>192 ± 10</td>
<td>128 ± 21</td>
</tr>
<tr>
<td>C6</td>
<td>131 ± 19</td>
<td>178 ± 10</td>
<td>99 ± 21</td>
</tr>
<tr>
<td>C5</td>
<td>152 ± 21</td>
<td>199 ± 11</td>
<td>142 ± 21</td>
</tr>
<tr>
<td>C4</td>
<td>143 ± 22</td>
<td>198 ± 10</td>
<td>140 ± 21</td>
</tr>
<tr>
<td>C3</td>
<td>171 ± 19</td>
<td>200 ± 12</td>
<td>144 ± 24</td>
</tr>
<tr>
<td>C2</td>
<td>147 ± 22</td>
<td>186 ± 13</td>
<td>116 ± 25</td>
</tr>
<tr>
<td>C1</td>
<td>144 ± 18</td>
<td>199 ± 15</td>
<td>144 ± 30</td>
</tr>
<tr>
<td>Boulder</td>
<td>168 ± 26</td>
<td>177 ± 15</td>
<td>102 ± 26</td>
</tr>
</tbody>
</table>
Figure 1. Guide to the study area. (i) Map of Greece showing simplified large-scale tectonic structures (CG, Corinthian Gulf; CF, Cephalonia fault; NAF, North Anatolian fault; NAT, North Anatolian Trough; black lines with barbs show active thrusts; black lines with marks show active faults) (after Kokkalas and Koukouvelas, 2005; Papanikolaou and Royden, 2007). Black crosses highlight study areas. (ii) Satellite image (Landsat 8, 2015) of the easternmost Gulf of Corinth. The Kaparelli fault is shown in red and the cross marks the position of the paleoseismological trench of Kokkalas et al. (2007). (iii) View of the Kaparelli trench. (iv) Satellite image (Landsat 8, 2015) of the study area at the Sfaka fault (red) in northeastern Crete; the cross shows the position of the road cut along strike. (v) View of the Sfaka road cut. (vi) Sketch of a typical post-glacial normal fault showing bedrock juxtaposed against Quaternary sediments which contain structures caused by recurrent earthquakes (modified after Reicherter et al., 2003).
Figure 2. A simplified model of investigated parts on footwall, scarp, hanging-wall, and trench. Dashed lines show the different workspaces: (i) overall workspace for a long-mid range t-LiDAR scan to retrieve the geometric relation of investigated components, (ii) area of operations (log, photo, t-LiDAR, GPR) on the trench wall, (iii) workspace for GPR measurements (black arrows) on top of the colluvium.
Figure 3. An illustration on how spatial information gets down sampled. The median value of surrounding cells provides the new cell \((x, y)\) value \((\lambda)\).
Figure 4. Compilation of analytical input and outcome. (a) Trench log produced in the field and corrected with (b) photomosaic in the office. (c) High-resolution digital backscatter model (HRDBSM) from t-LiDAR measurements. (d) Two-channel composition from (b) and (c). Note, green and red are 100% different. (PM = photomosaic). (e) Visualisation of spatial distribution of seven classes from the unsupervised classification. White dashed lines indicate coinciding arrangements and some influence from daylight.
Figure 5. Statistical analysis of two-channel composition image. (a) Seven distinct classes were estimated from the unsupervised classification. (b) Histograms of representative classes per identified layer. Either the majority of a mapped layer is filled up by one class (e.g. fault gouge) or by a certain composition of 2 or 3 classes (e.g. Palaeosol II and C6).
Figure 6. Varying reflectance of electromagnetic waves along the trench wall. Transitions between individual layers are depicted by drastic changing shapes of reflectance spectra. The error bar is given by the standard deviation of each sample.
Figure 7. Three-dimensional reconstruction of differing layers within the outcrop from GPR image interpretation. Partial reflection of radar waves on layer contacts leads to significant backscatter signals at depths down to approximately 3 m.
Figure 8. Results from the trenching site of Kokkalas et al. (2007). (a) 2-channel composition from multispectral approach. Red and green are 100% different whereas yellow colouring represents intermediate correspondence of both channels. The trench log (black lines) fits with the multispectral cluster of a certain composition. (b) Three-dimensional reconstruction of the trench exposure. Recorded thickness of the colluvial wedge from 1981 is about 0.6 m.