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SPECIAL ISSUE

Using hyperspectral plant signatures for $CO₂$ leak detection during the 2008 ZERT $CO₂$ sequestration field experiment in Bozeman, Montana

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Abstract Hyperspectral plant signatures can be used as a short-term, as well as long-term (100-year timescale) monitoring technique to verify that CO₂ sequestration fields have not been compromised. An influx of $CO₂$ gas into the soil can stress vegetation, which causes changes in the visible to near-infrared reflectance spectral signature of the vegetation. For 29 days, beginning on July 9, 2008, pure carbon dioxide gas was released through a 100-m long horizontal injection well, at a flow rate of 300 kg day^{-1} . Spectral signatures were recorded almost daily from an unmown patch of plants over the injection with a ''FieldSpec Pro'' spectrometer by Analytical Spectral Devices, Inc. Measurements were taken both inside and outside of the $CO₂$ leak zone to normalize observations for other environmental factors affecting the plants. Four to five days after the injection began, stress was observed in the spectral signatures of

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plants within 1 m of the well. After approximately 10 days, moderate to high amounts of stress were measured out to 2.5 m from the well. This spatial distribution corresponded to areas of high $CO₂$ flux from the injection. Airborne hyperspectral imagery, acquired by Resonon, Inc. of Bozeman, MT using their hyperspectral camera, also showed the same pattern of plant stress. Spectral signatures of the plants were also compared to the $CO₂$ concentrations in the soil, which indicated that the lower limit of soil $CO₂$ needed to stress vegetation is between 4 and 8% by volume.

Keywords Geologic carbon sequestration · Hyperspectral plant signatures · Reflectance spectra · $CO₂$ leak detection \cdot Surface monitoring of carbon sequestration

Introduction

With the effects of climate change on the rise, reduction of the amount of $CO₂$ released into the atmosphere is crucial. Injection of $CO₂$ into deep underground formations, known as geologic carbon sequestration, can be an effective method of mitigation by preventing the $CO₂$ injected into underground sequestration formations from entering the atmosphere. In order for geologic sequestration to be successful, methods of verification are important to maintain the sequestered $CO₂$ and to assure the public that the sequestration operation is safe. It is therefore important to develop techniques for long-term $CO₂$ leak detection at the surfaces of the $CO₂$ sequestration fields (Pickles and Cover [2005](#page-11-0)).

Analyzing hyperspectral plant signatures over $CO₂$ sequestration fields can confirm that the sequestration fields have not been compromised. If a leak were to occur, the excess amount of $CO₂$ in the top layers of soil near the surface would stress the vegetation above the sequestration field, which can be seen as changes in their spectral signatures. $CO₂$ -induced stress has been recognized in the spectral signature of plants over volcanic $CO₂$ vents at the Latera (Bateson et al. [2008](#page-11-0)) and Long Valley (De Jong [1996](#page-11-0); Hausback et al. [1998;](#page-11-0) Martini et al. [2000\)](#page-11-0) calderas, as well as in laboratory experiments (Noomen and Skidmore [2009](#page-11-0)). Conversely, if vegetation over a sequestration field has healthy spectral signatures, it would indicate that $CO₂$ is being sequestered effectively. Because the basic requirement of this technique is just the presence of healthy vegetation over the sequestration field, hyperspectral plant signatures are particularly useful tool for monitoring sequestration fields for years to centuries (Pickles and Cover [2005](#page-11-0)).

This technique was studied during a shallow underground leak experiment held at Montana State University (MSU) in Bozeman, Montana. Pure $CO₂$ gas was released for 29 days from a 100-m long, horizontal $CO₂$ injection well, located about 1–2.5 m underground. During this time, the health of vegetation over the west end of the injection well was determined by measuring their hyperspectral reflectance signatures with a field spectrometer. In addition, airborne hyperspectral imagery was acquired from a low-flying aircraft using a hyperspectral camera system developed by Resonon, Inc. of Bozeman, MT (www.resonon.com). The spectral signatures of the plants were also correlated with variations in soil $CO₂$ concentrations measured directly over, and at 2.5, 5, 7.5, and 10 m from the well by Lewicki et al. ([2009,](#page-11-0) this volume).

Plant stress and spectral signatures

The term ''plant stress'' tends to be used in numerous ways because monitoring plant health has many applications. In a general sense, plant stress occurs when environmental conditions are unfavorable for optimal plant growth. Some of the conditions that can cause stress are drought, extreme heat or cold, insect infestation, waterlogging, bacterial diseases, oxygen depletion, nutrient deficiencies, or acidic soil (Lichtenthaler [1998\)](#page-11-0). Although excess $CO₂$ has also been shown to negatively affect plants, the exact mechanism(s) by which it harms vegetation is not known. Most likely, the $CO₂$ gas displaces oxygen to the roots of the plant, which occurs in natural gas leaks (Noomen et al. [2008\)](#page-11-0). Other possibilities are that large amounts of $CO₂$ could change the pH and redox potential of soil or alter natural microbial environments (Noomen et al. [2006\)](#page-11-0). The result of vegetation experiencing stress over long periods can be stunted growth, reduced water content, or a decrease in leaf chlorophyll concentrations. This can lead to replacement by other more tolerant plant species which

then becomes habitat modification (Pickles and Cover [2005](#page-11-0)). We used the decrease in chlorophyll concentrations as an indicator of plant stress because it is a typical response regardless of species of vegetation or cause of stress (Carter [1993\)](#page-11-0). In addition, chlorophyll can be readily estimated in the reflectance spectra of the vegetation (Carter and Knapp 2000; Hill [2004](#page-11-0); Zhang et al. [2008](#page-11-0); Moorthy et al. [2008](#page-11-0)).

When light contacts a leaf, the various wavelengths can be absorbed, reflected, or transmitted based on the leaf's chemical and physical structure. This interaction results in a distinctive spectrum of reflected light. Healthy plants tend to have relatively low reflectance in the visible (\sim 400 to 720 nm) with a peak in the green range (\sim 550 nm) and high reflectance in the near-infrared (NIR) $(\sim 700-$ 1,400 nm). The spectral signature of plants in the visible is caused by the various pigments they contain, such as carotenoid and chlorophyll compounds. Chlorophyll gives a healthy plant its green color because chlorophyll reflects light in the green and absorbs light elsewhere in the visible (Blackburn [2007\)](#page-11-0). Specifically, chlorophyll causes a strong absorption feature in healthy plant spectra from approximately 600 to 700 nm. The size and detailed shape of this absorption feature have been found to correlate with the amount of chlorophyll in a plant because a decrease in the amount chlorophyll would also decrease the plant leaf's overall absorptivity in the range of the absorption feature (Carter [1993](#page-11-0), Carter and Knapp 2001; Smith et al. [2004](#page-11-0); Noomen et al. [2006](#page-11-0); Noomen and Skidmore [2009](#page-11-0)). Therefore, if a particular plant is stressed by excess $CO₂$ from a faulty sequestration field, the chlorophyll concentrations will decrease, and the chlorophyll absorption feature of its spectral signature will change accordingly.

Even though hyperspectral signatures can be used to estimate the amount of plant stress, it is unable to distinguish between the different causes of stress. This uncertainty can be problematic when testing for $CO₂$ leaks. A sequestration field could appear unhealthy spectrally, but the stress could be caused by something unrelated to escaping $CO₂$. To reduce the number of possible false positives, it is important to normalize measurements by comparing plant spectra to identical environmental conditions outside the sequestration field. Normalization can help determine if the identified stress is seasonal (for example, high summer heat), which would affect vegetation regionally. The spatial distribution of stressed vegetation may also yield information about the pathways by which $CO₂$ migrates from depth to the surface. In addition, knowing the location of $CO₂$ pipelines and other potential leak sites would be extremely useful since the $CO₂$ stress would probably be located near by. Thus, hyperspectral plant signatures could still be used to verify that the sequestration fields have not leaked (with healthy spectral

signatures across the field) or at least focus ground-based $CO₂$ monitoring techniques on potential leak sites.

Methods

Field design

In a MSU agricultural field, a 100-m long, horizontal $CO₂$ injection well was installed at depth varying between approximately 1 and 2.5 m. Pure $CO₂$ gas was injected for 29 days from July 9 to August 7, 2008 at a flow rate of 300 kg day⁻¹ (Spangler et al. 2009 , this volume). Prior to the start of the injection, a 20- by 30-m patch of vegetation, centered over the west end of the horizontal injection well, was selected for the plant stress experiment. This section of the field was left unmown except for access paths and was not disturbed for the duration of the experiment (Fig. 1). Within the plant study area, 68 sites were chosen where plant spectral measurements were taken daily. Measurements started 2 days before the start of the injection until 1 week after the $CO₂$ injection was shut off (July 5 to August 11, 2008). Measurement sites ranged in position from directly over the injection well out to 10 m from the well horizontally. In addition, 24 measurement sites were located at 1-m intervals directly above the injection well (Fig. 2). This array gave plant reflection spectra both within and outside of the injected $CO₂$ leak zones so the observations could be normalized for other environmental factors affecting the plants. The plant species within the field consisted primarily of various species of short and tall grasses, alfalfa, dandelions, and a variety of clovers. Each measurement site contained random mixes of vegetation species. Table 1 contains a full list of plant types with their scientific names that were identified on July 30, 2008. This

Fig. 1 Aerial photo of field site, taken August 5, 2008 by Resonon Inc

Fig. 2 Schematic representation of plant study area. Distances are measured relative to the center of the injection well

Table 1 Plant species identified at field site

Composite family—Asteraceae
Western salsify (<i>Tragopogon dubius</i>)
Dandelion (Taraxacum officinale)
Canada thistle (Cirsium arvense)
Legume family—Leguminaceae
Alfalfa (<i>Medicago sativa</i>)
Birdsfoot trefoil (Lotus corniculatus)
Yellow blossom sweet clover (Meliotus officinalis)
Red clover (<i>Trifolium pretense</i>)
Lupine (<i>Lupinus argenteus</i>)
Grass family—Poaceae
Quackgrass (Agropyron repens)
Orchard grass (<i>Dactylis glomerata</i>)
Timothy (<i>Phleum pretense</i>)
Tall fescue (Festuca pratensis)
Kentucky bluegrass (<i>Poa pratensis</i>)
Field brome <i>(Bromus arvensis)</i>
Smooth brome (<i>Bromus inermis</i>)

list may not be complete as some species may have been dormant at the time.

Field spectrometer measurement techniques

Plant spectral signatures were measured using a ''FieldSpec Pro'' spectrometer by ASD, Inc., [\(www.asdi.com](http://www.asdi.com)). It was a

full range spectrometer that measured reflectance spectra from 350 to 2,500 nm with a 1-nm bandwidth. The spectrometer was connected to a fiber optic cable fitted with a 5° lens. The lens was held approximately 1 m above the ground from a nadir position, which resulted in the spectrometer having a field-of-view of approximately 10 cm. To reduce noise in the spectra, the spectrometer was programed to give a final spectral measurement that was the average of 100 spectra taken continuously at each site. This spectrometer was also designed to self-regulate many of the settings. To optimize this feature, the lens of the spectrometer was held over the vegetation to be measured for approximately 3 s prior to acquiring each spectral signature. This time allowed the spectrometer to self-adjust its settings before taking each measurement. Precautions were also made not to expose the lens of the spectrometer to extreme bright light or darkness. These techniques appeared to reduce noise and increased the reproducibility of the data significantly.

Several procedures were used to eliminate illumination and atmospheric factors that would affect the data. A ''Spectralon'' white ''reflectance plate'' provided the solar reference spectrum used to calculate reflectance spectra from the raw collected spectra. A white plate reference spectrum was measured before collecting any plant signatures and re-measured it after every few spectral readings. In addition, spectral readings were only acquired under cloud-free skies with minimal haze. Each day, data were collected in the same sequence as close to solar noon as possible, typically between 9:30 am and 1:30 pm. This approach gave a consistent solar intensity and angle at each measurement site throughout the experiment.

Classification of hyperspectral plant signatures

To aid in analyzing patterns of plant stress, the hyperspectral plant spectra were classified using the software program ENVI 4.5 [\(www.ittvis.com](http://www.ittvis.com)). As a first step, we created a computer program to compile the individual spectra recorded by the FieldSpec Pro spectrometer into data cubes, thereby creating artificial images of the plant study area. The artificial images show the discrete spectral measurements taken on a given day with their true relative orientations. The images have a 0.5-m pixel size. Each pixel contains either the spectra measured by the FieldSpec Pro at that site or were assigned a spectrum with zero reflectance and were ignored in subsequent analysis. These artificial images were created for each day of the experiment. They are oriented with the east edge of the plant study area at the top of the artificial image with the $CO₂$ injection well running along the center of the image (Fig. 3).

Fig. 3 Example artificial image created from data collected from FieldSpec Pro Spectrometer. Black area represents the plant study area. White pixels are measurement sites. The east edge of the plant study area is the top of the image

Each artificial image of the field was processed with an ENVI classification tool called Spectral Angle Mapper (SAM). This method classifies measured spectra to reference endmember spectra based on how closely the observed and reference spectra compare (Kruse et al. [1990](#page-11-0)). This technique has been found to be relatively insensitive to illumination effects caused by atmospheric conditions or differences in leaf angle relative to the spectrometer lens line of sight (Sohn and Rebello [2002\)](#page-11-0).

The SAM classification utilized more of the spectral signature than other simple indices of plant health, which may only compare reflectance at select wavelengths. The supervised SAM classification was based on spectral signatures from 350 to 800 nm. The spectra were divided into three classes: Healthy Vegetation, Moderate to Highly Stressed Vegetation, and Extremely Stressed Vegetation. A total of nine spectra were chosen as endmembers for the SAM classification, with four endmembers as ''Healthy Vegetation'', two as ''Moderately to Highly Stressed Vegetation'', and three as ''Extremely Stressed Vegetation''. Endmembers were based on the overall shape of the spectra, variety of vegetation, quality of spectral signal,

Fig. 4 Graph of representative spectra for each endmember class used to classify artificial images

and on the depth of their chlorophyll absorption feature from approximately 600 to 700 nm. The depth of this absorption feature was estimated by subtracting the maximum reflectance, between 525 and 575 nm, from the minimum reflectance, between 675 and 700 nm. A relatively large absorption depth was considered to be ''Healthy Vegetation,'' whereas an absorption depth very close to zero (either positive or negative) was ''Moderate to Highly Stressed Vegetation'' and a large negative absorption value was ''Extremely Stressed Vegetation''. While measuring the depth of this chlorophyll absorption feature gives an estimation of stress level, it was not used as a final classifier because we wanted to utilize the entire spectral signature. Figure 4 shows representative endmember spectra for each class.

Aerial hyperspectral imagery

On August 5, 2008 (after 27 days of $CO₂$ injection), a hyperspectral image was acquired of the entire injection well from a low-flying aircraft. The spectrometer, developed by Resonon, Inc. of Bozeman, MT, had a spectral range of 400–880 nm with a wavelength resolution of approximately 6 nm. ENVI's supervised classification tool with Spectral Angle Mapper was used to analyze the image over the spectrometer's full spectral range. Reference endmember spectra were determined on the basis of spectral shape, particularly the depth of the chlorophyll absorption feature, as well as on plant species. The image was classified into five categories: High stress, Moderate stress, Low or Seasonal Stress, Healthy Vegetation (grasses), and Healthy Vegetation (herbaceous legumes). Figure 5 is the hyperspectral imagery shown in true color.

Fig. 5 Aerial hyperspectral imagery of field site shown in true color. Imagery was acquired after 27 days of the $CO₂$ injection

Results

By comparing changes in shape of the hyperspectral plant signatures, stress could be identified in some of the plants close to the well after 4–5 days of $CO₂$ injection (Fig. [6](#page-6-0)). After 14 days, the spatial extent of plant stress reached a maximum with two discrete, identifiable zones of plant stress. One zone was located over the injection well, along the eastern edge of the vegetation study area. It extended out to at least 2.5 m on the north side of the injection well and out to approximately 1 m on the south side. The second zone of plant stress was in the center of the vegetation study area approximately 20 m from the east edge. It was also located over the injection well and extended out to at least 2.5 m in all directions.

Fig. 6 Spectral signatures throughout experiment of vegetation 0.5-m north of injection well, at the eastern edge of plant study area. Note the changes in absorption features between 550 and 750 nm

Because hyperspectral plant signatures are very sensitive to environmental changes, this method indicated plant stress using before changes in vegetation were visible to the eye. Large decreases in chlorophyll can make vegetation appear more yellow or brown (Fig. 7). However, changes in the spectral signature can indicate chlorophyll decreases even when there are only small deceases in chlorophyll and the vegetation is still green. In addition, the amount of stress measured during the experiment seemed to be species dependent. The tall grasses and alfalfa in the field reacted quickly, whereas dandelion and some naturally short grasses appeared to be more resistant to stress.

Spectral measurements were collected for 1 week past the end of the injection. It was unclear whether the extremely stressed vegetation began to recover during this time. It was likely that the prolonged exposure to the high amount of $CO₂$ applied enough stress to irreparably damage the vegetation. However, new vegetation, specifically orchard grass and dandelions, began to grow anew in zones of extreme plant stress, which may suggest that certain species of vegetation are more capable of growing under $CO₂$ stress.

Classification results: field spectrometer artificial images

The classification from the artificial images of the field also shows two locations of plant stress by the end of the $CO₂$ injection: (1) on the east edge and (2) in the center of the plant study area. Figure [8a](#page-7-0)–f is the classification images at the start of the injection, at 10, 14, 16, and 23 days of injection, and 1 day following the end of the injection, respectively. According to the classifications, plants began to experience moderate to high amounts of stress after

Fig. 7 Photos of vegetation 0.5 m north of injection well, at east edge of plant study area. Top photo was taken before injection (July 8, 2008). Bottom photo taken 4 days after end of injection (August 11, 2008)

approximately 6 days of the injection. The size and intensity of the plant stress zones increased throughout the duration of the injection. Near the end of the experiment (early August), the entire field appears to be stressed to some degree (Fig. [8f](#page-7-0)), which was due to natural changes in the grasses lifecycle and/or by extreme weather (for example, excessive heat or the hail storm on July 22, 2008). Despite seeing this trend of increasing stress across the field, the pattern of two zones of extreme stress is still identifiable. This is an additional benefit for $CO₂$ sequestration monitoring, because it shows that stress from an influx of $CO₂$ can be distinguished from seasonal or other environmental stressors.

Classification results: aerial hyperspectral imagery

Figure [9](#page-8-0) is the SAM classification of the hyperspectral imagery, which was acquired after 27 days of $CO₂$

Fig. 8 Classifications of artificial image using ENVI Spectral Angle Mapper algorithms. Green Healthy vegetation, orange moderately to highly stressed vegetation, red extremely stressed vegetation. a Before injection (July 9, 2008); b after 10 days (July 19, 2008); c after 14 days (July 23, 2008); d After 16 days (July 25, 2008); e 23 days (August 1, 2008); f 1 day after end of the 29-day injection (August 8, 2008)

injection (August 5, 2008). The classification figure (Fig. [9](#page-8-0)) also shows the two same zones of plant stress, each measuring approximately 2 m in diameter. Similarsized zones of plant stress are also found along the length of the entire injection well. These other zones were in mown grass that was approximately 15 cm tall. This demonstrates how this monitoring technique was capable of detecting $CO₂$ leaks over areas with different types of vegetation. One zone of plant stress is difficult to distinguish at the north end of the injection well (Fig. [9](#page-8-0)). It appears similar to zones of low plant stress caused by frequent foot traffic. However, its location directly over the injection well suggests that the plant stress is connected to the $CO₂$ injection. There are also areas that have been classified as high plant stress that were actually caused by disturbances from equipment from other participants of this experiment that can be seen in the true color aerial image (Fig. 8a). Comparing the locations of plant stress with locations of potential leak site (pipelines, for example) and artificial structures helped indentify the various causes of plant stress.

In addition to identifying zones of plant stress, different types of vegetation could be distinguished in the image. Because the tall grasses of the plant study area were left unmown, there were dried seeds on top of the grass stalks. Although that was normal for those species of grass during late summer, it made the unmown area appear slightly stressed. This effect occurs because the aerial spectrometer looks vertically onto the vegetation, seeing mainly the dried seed pods. Outside of the plant study area, the vegetation had been mown prior to the start of the experiment and fresh and healthy vegetation was growing in. This image also demonstrates the effect of plant structure on the hyperspectral signature. Across the MSU agricultural site, the vegetation is typically some variety of either grass or herbaceous legume (Alfalfa, clover, or bird's foot trefoil). Even if both categories of vegetation are considered healthy, their spectral signatures may be distinct because of the differences in their physical leaf structures, specifically vertical blades of grass compared to more horizontal wide leaves. These differences in spectral signatures were identified using ENVI's classification algorithms. It is also

Fig. 9 Classification of Aerial Hyperspectral Imagery acquired by Resonon Inc. Classification was performed using ENVI Spectral Angle Mapper algorithms. White boxes show locations of areas of apparent plant stress caused by equipment from outside experiments. The red circle indicates location of the subtle fifth zone of plant stress caused by the injected $CO₂$

important to note that these effects from leaf structure are mostly pronounced outside the chlorophyll absorption feature and do not significantly affect the ability to detect stress from the spectral signatures.

Associated observations

Soil $CO₂$ fluxes

Soil $CO₂$ fluxes were measured using the accumulation chamber method on a grid repeatedly on a daily basis during the injection by Lewicki et al. ([2009,](#page-11-0) this volume). Figure 10 shows an example of a map of log soil $CO₂$ flux, interpolated based on measurements made on August 1, 2008 (23 days of $CO₂$ injection). The location of the plant study area is shown for reference. Five separate zones of relatively high $CO₂$ flux can be seen, which are approximately 2–3 orders of magnitude above background levels. They are roughly circular to elliptical in shape ranging from approximately 5 to 15 m in diameter. Within the boundaries of the plant study area, there are two zones of high $CO₂$ flux, located in the center of the plant study area and at the eastern edge.

Fig. 10 Map of log soil $CO₂$ flux, interpolated based on measurements made at the black dots [modified from Lewicki et al. [2009](#page-11-0), (this volume)]. The blue rectangle represents the $CO₂$ injection well. The red rectangle marks the plant study area. White squares are soil $CO₂$ concentration probes. Note the two zones of high $CO₂$ flux at the east edge and in the center of the plant study area

Soil CO₂ fluxes were also measured at 1-m intervals along the surface trace of the injection well on July 14, 2008 (5 days of $CO₂$ injection) by Lewicki et al. ([2009,](#page-11-0) this volume). Figure [11](#page-9-0) shows the same five zones of high flux along the injection well, two of which were located within the plant stress area. One high $CO₂$ flux zone, located in the center of the plant study area, is approximately 6 m in diameter and has a maximum flux of approximately 3,700 g m^{-2} day⁻¹. The other maximum, at 6,500 g m⁻² day⁻¹, is located 2 m northeast of the eastern edge of the plant study area, while $CO₂$ flux is about 700 g m⁻² day⁻¹ at the edge of the plant study area.

Soil $CO₂$ concentrations

Soil CO₂ concentrations were measured at 30 cm depth at distances of 0, 2.5, 5, 7.5 and 10 m northwest of the injection well (Fig. 9) using non-dispersive infrared gas analyzers every 1 s and averaged over half-hour periods (Lewicki et al. [2009,](#page-11-0) this volume) (see Figs. [2,](#page-3-0) 10 for locations). Figure 12 shows time series of the soil $CO₂$ concentrations. Over the injection well, soil $CO₂$ concentrations rose to approximately $13-14\%$ CO₂ by volume. The plants over the injection well also began to show stress after 4 days. $CO₂$ concentrations measured 2.5 m from the well ranged from 6 to 10%, averaging approximately 8% $CO₂$ by volume. Nearby plants showed stress after 14 days.

At 5 m from the well, soil $CO₂$ concentrations rose to a maximum of 3–4%. At 7.5 and 10 m from the injection well, soil $CO₂$ concentrations did not rise significantly during the injection, staying close to $1-2\%$ CO₂ by volume. None of the plants at 5, 7.5, or 10 m showed significant signs of stress. From these observations, the lower limit of the $CO₂$ concentration needed in the soil to begin to stress vegetation could be inferred. Because plants at 2.5 m were stressed, but those at 5 m were not, the lower limit of soil CO2 concentration can be estimated to be between 4 and 8% by volume, which was approximately 3–4 times greater than background levels. Bateson et al. [\(2008](#page-11-0)) performed a leak detection experiment over natural volcanic $CO₂$ vents in Italy by monitoring plant health using various remote sensing techniques. The minimum soil $CO₂$ concentration of their detected gas vents was 5.6% which falls within the range determined in this experiment. In a greenhouse

experiment on maize plants, Noomen and Skidmore ([2009\)](#page-11-0) found a correlation between high percentages of $CO₂$ in soil and decreases in chlorophyll content and plant growth. However, 50% $CO₂$ was the minimum $CO₂$ concentration to give statistically different results.

Discussion

Everywhere high $CO₂$ flux was measured, it was associated with a corresponding zone of plant stress. This result shows the potential of using hyperspectral plant signatures for $CO₂$ leak detection. Both the field spectrometer and the airborne hyperspectral imager measured the same two discrete zones of plant stress, on the eastern edge and in the center of the plant study area. These zones coincided with areas of high measured $CO₂$ flux. The airborne

Fig. 13 Plot of Normalized Difference Water Index (NDWI) values. Vegetation is located along east edge of plant study area at various distances from the injection well. Vegetation in the $CO₂$ leak zone is plotted in *red*. Background values are in *blue*. NDWI = $(\rho 860 \rho$ 1240)/(ρ 860 + ρ 1230), ρ = reflectance at given wavelength in nm

hyperspectral imagery also shows three additional zones of plant stress outside the plant study area, whose locations match the zones of high $CO₂$ flux measured along the entire injection well (Fig. [9\)](#page-8-0). The high similarity between the distinctive distributions of extremely stressed vegetation and high $CO₂$ flux indicated that injected $CO₂$ is the cause of the plant stress.

One difficulty of this leak detection method is in the ability to recognize stress caused by excess $CO₂$ independent of other environmental stressors. However, the spatial distribution of relative stress across a sequestration field can help to identify the possible causes for the identified plant stress. In this experiment, two zones of plant stress were located that were focused and sharply delineated from the rest of the field in circular patterns that were centered over the leaks in the injection well. This pattern would not be expected for stress caused by weather, infestation, or poor soil which would act on a regional scale. In addition, the size of plant stress zones remained consistent during changing environmental conditions. For instance, several days of consistent rainfall did not appear to alleviate stressed vegetation, indicating that dehydration was not the cause of the plant stress. The signal of extreme plant stress was also distinct enough not to be clouded by seasonal stress at the end of the summer, showing that seasonal stress was not the only stressor present. The link between injected $CO₂$ and zones of plant stress became evident after ruling out these other environmental possibilities for stress.

It can also be possible to use the spectral signature of vegetation to eliminate drought or prolonged exposure to high heat as possible stressors. Leaf water content strongly influences hyperspectral plant signatures outside the visible spectra (Carter [1991](#page-11-0)). Because this experiment took place during the end of a hot summer, there was a possibility that the vegetation was showing stress because they were simply drying out. To investigate this possibility, the Normalized Difference Water Index (NDWI) was calculated over the course of the injection for vegetation at various measurement sites across the east edge of the plant study area. NDWI values are calculated by the difference of the reflectance at 860 and 1,240 nm divided by their sum. NDWI has been found to correlate with water content of vegetation (Gao [1995](#page-11-0)). Figure 13 is a graph of NDWI for vegetation on the east edge of the plant study area located 0.5-, 1-, 2.5-, 7.5-m north, as well as 10-m south, of the injection well. Measurements from vegetation at 7.5 and 10 m from the injection well were outside the $CO₂$ leaks and can be considered as background values. Figure 13 shows an overall decrease in water content for all vegetation across the field, suggesting that high summer heat has affected the water content of all the plants. However, the water content in vegetation within 2.5 m of the injection well decreased at a faster rate than background vegetation. This result indicates an additional stress, besides high heat, was acting on these plants because the temperature and precipitation were consistent between all areas of the field. In this experiment, the additional stress can be linked to the high $CO₂$ flux within 2.5 m of the injection well because it is the only different factor known between the vegetation. However, in general, this connection cannot be made since the $CO₂$ flux is usually unknown. Normalizing for water content can at least eliminate some stressors and reduce the chance of false positives for $CO₂$ leaks. It is possible that other stressors could be eliminated similarly. Therefore, it is important to acquire hyperspectral data over an entire region and not just over pipelines and other potential leak sites of sequestration fields.

Conclusions and recommendations

 $CO₂$ leaks through the soil could successfully be identified by hyperspectral signatures to determine the health of overlying vegetation. The plant spectra began to show signs of stress within 4 days of the $CO₂$ injection. The lower limit of soil $CO₂$ concentration needed to induce this stress can be estimated to be between 4 and 8% CO₂ by volume. The spatial extent of plant stress, as determined by spectrometers in the field and from the air, match the locations of high measured $CO₂$ flux from the injection.

There are also many additional benefits to this monitoring technique, such as a quick response time. Vegetation is very sensitive to stress, so a suspected leak could be identified within days. Also, the actual spectral measurements can be collected and processed rapidly by acquiring

airborne hyperspectral imagery, which will have no impact on the sequestration field itself. More importantly, this monitoring technique can be utilized long term, even on 100-year timescales, as it would require little to no maintenance of the sequestration field. The only requirement is simply the presence of vegetation. With these benefits, analyzing hyperspectral plant signatures for stress is an efficient method to verify that sequestration fields are successful in retaining injected $CO₂$.

Future work will focus on monitoring $CO₂$ leaks with various instruments, particularly on airborne and/or satellite platforms. Work will also go toward determining the specific mechanisms that cause plant stress due to changes in soil ecosystems from a $CO₂$ leak. In addition, the species dependence of the plant stress caused by elevated $CO₂$ soil concentrations is another topic to be investigated.

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