

1 **Detecting a moorland wildfire scar in the Peak District, UK, using**  
2 **Synthetic Aperture Radar (SAR) from ERS-2 and Envisat ASAR**

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# 1 **Detecting a moorland wildfire scar in the Peak District, UK, using** 2 **Synthetic Aperture Radar (SAR) from ERS-2 and Envisat ASAR**

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9 Wildfires occur annually in UK moorland environments, especially in drought  
10 years. They can be severely damaging to the ecosystem when they burn deep into  
11 the peat, killing ground-nesting birds and releasing CO<sub>2</sub> into the atmosphere.  
12 Synthetic Aperture Radar (SAR) was evaluated for detecting the 18 April 2003  
13 Bleaklow wildfire scar (7.4km<sup>2</sup>). SAR's ability to penetrate cloud is advantageous  
14 in this inherently overcast area. SAR can provide fire scar boundary information  
15 which is otherwise labour intensive to collect in the field using a Global Positioning  
16 System (GPS). This paper evaluates the potential of SAR intensity and InSAR  
17 coherence to detect a large peat moorland wildfire scar in the Peak District of  
18 northern England. A time-series of pre-fire and post-fire ERS-2 and ASAR Single  
19 Look Complex (SLC) data were pre-processed using SARscape 4.2 to produce  
20 georeferenced greyscale images. SAR intensity and InSAR coherence values were  
21 analysed against CORINE land cover classes and precipitation data. SAR intensity  
22 detected burnt peat well after a precipitation event and for previous fire events  
23 within the CORINE peat bog class. For 18 April 2003 fire event intensity increased  
24 to 0.84 dB post-fire inside the fire scar for the peat bog class. InSAR coherence  
25 peaked post-fire for moors and heathland and natural grassland classes inside the  
26 fire scar, but peat bog exposed from previous fires was less responsive. Overall,  
27 SAR was found to be effective for detecting the Bleaklow moorland wildfire scar  
28 and monitoring wildfire scar persistence in a degraded peat landscape up to 71 days  
29 later. Heavy precipitation amplified the SAR fire scar signal, with precipitation after  
30 wildfires being typical in UK moorlands. Further work is required to disentangle the  
31 effects of fire size; topography and less generalised land cover classes on SAR  
32 intensity and InSAR coherence for detecting fire scars in degraded peat moorlands.

## 33 **1. Introduction**

34 UK moorland wildfires need to be monitored as they can burn deep into the peat  
35 releasing CO<sub>2</sub> into the atmosphere, contributing to greenhouse gas emissions, and hence  
36 global warming. In addition the UK Climate Impacts Programme (UKCIP) suggest a 1-  
37 5°C increase in average summer temperatures, increasing the probability of wildfire  
38 frequency (McEvoy *et al.* 2006). Wildfires result in fire scars producing areas of exposed  
39 peat which release Dissolved Organic Carbon (DOC) into nearby streams, causing  
40 discolouration to drinking water supply and requiring costly restoration (Worrall and  
41 Evans 2009). It is important to detect wildfire scars in moorland habitats to understand  
42 their size, distribution and persistence in the landscape because this information can  
43 help inform moorland restoration projects and ecosystem changes (Anderson *et al.*  
44 2009).

45 Remotely sensed data may offer an alternative and more cost effective approach to  
46 map and monitor moorland wildfire scars at the landscape scale, potentially providing  
47 consistent information on fire size, persistence of wildfire scars, and variation in burn  
48 severity within them. However, in UK moorlands optical remote sensing is limited by

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1 frequent cloud cover (Armitage *et al.* 2007, Legg 1991). Thermal hotspot detection  
2 using MODIS (250m resolution) can miss these fires; for instance, the 0.8 km<sup>2</sup>  
3 Wainstalls fire in West Yorkshire 30 May 2011, as shown by analysis of hotspot data  
4 from the European Forest Fire Information System (EFFIS 2011).

5 The use of SAR for wildfire scar detection in a UK moorland environment has not  
6 yet been fully explored and provides a unique line of enquiry. The benefits of SAR data  
7 are that as an active system it can penetrate cloud and images can be acquired both day  
8 and night, improving temporal resolution (Rykhous and Zhong 2011). SAR has  
9 previously been used effectively for fire scar detection in boreal (Bourgeau-Chavez *et al.*  
10 *al.* 2002, 2007), savannah (Menges *et al.* 2004), Mediterranean (Gimeno *et al.* 2004)  
11 and tropical (Huang and Siegert 2004) environments however not in a UK moorland  
12 environment.

13 The Peak District National Park (PDNP) is one of the few moorland areas in the  
14 UK where wildfire scar perimeters are systematically mapped using a Global  
15 Positioning System (GPS); perimeters have been recorded by rangers since 2003. Point  
16 data of the approximate centre of the fire ground has also been collected by rangers  
17 since 1976 (CCVE 2005). The PDNP is therefore a good case study, providing essential  
18 ground truth data on fire perimeter and fire history. In other moorland landscapes,  
19 spatially-robust data on wildfire scars is lacking due to the operational difficulties of  
20 getting to locations and the lack of labour to map the wildfire scars in the field. The aim  
21 of this paper is to explore whether C-band SAR data can be used to detect a significant  
22 wildfire scar in a UK degraded moorland landscape. It uses the case study of the 18  
23 April 2003 fire, which burnt deep into the peat on the Bleaklow plateau in the PDNP.  
24 The fire scar was GPS-mapped by the PDNP rangers as 7.4 km<sup>2</sup> in area with a 28 km  
25 perimeter (figure 1). The objectives of the paper are:

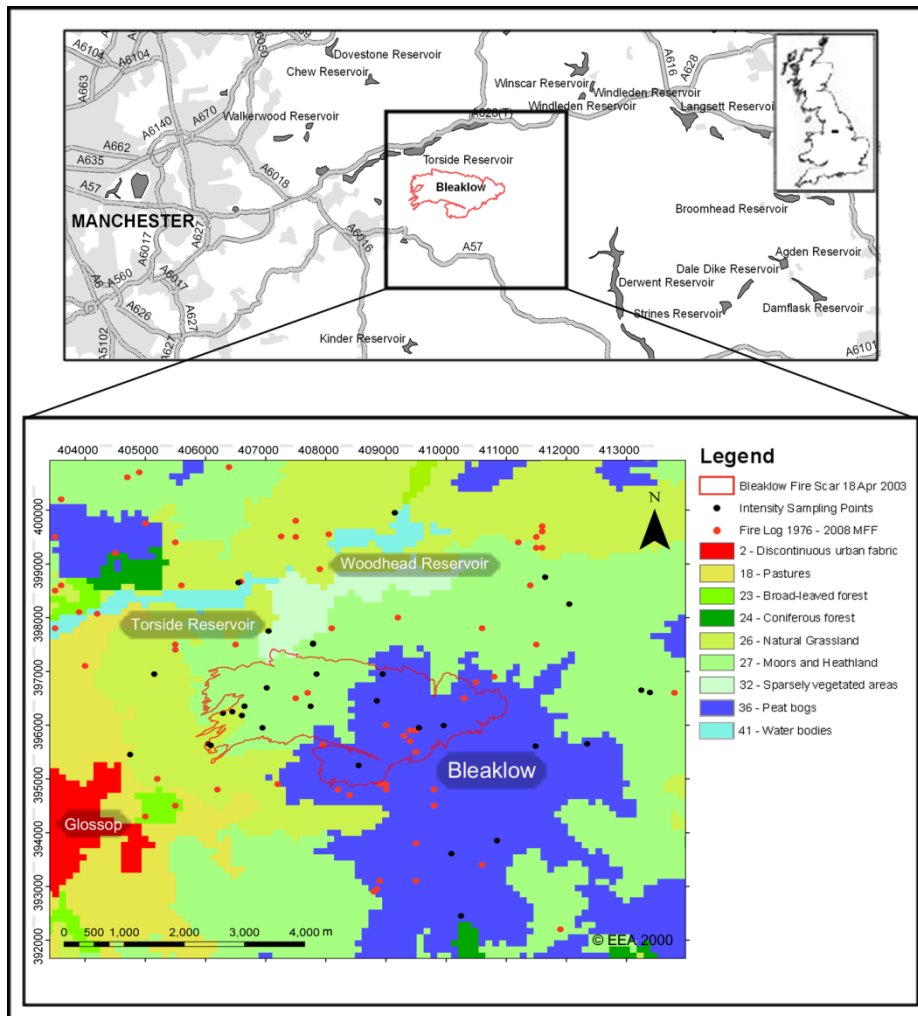
- 26 (i) To assess the temporal detectability of the Bleaklow 2003 wildfire scar  
27 using the SAR parameters of intensity and InSAR coherence for a time  
28 series of greyscale georeferenced ERS-2 and ASAR images;
- 29 (ii) To identify compare and contrast spatial patterns in SAR intensity and  
30 InSAR coherence signals inside and outside the fire scar.
- 31 (iii) To relate SAR spatial and temporal patterns to CORINE land cover classes  
32 and precipitation.

## 33 **2. Methodology**

### 34 **2.1 Study area**

35 The PDNP is one of the most degraded moorland landscapes in the UK (McEvoy *et al.*  
36 2006) and hence it is vulnerable to the effects of climate change as modelled by the UK  
37 Climate Impacts Programme (UKCIP) (McMorrow *et al.* 2009). Bleaklow is part of the  
38 Dark Peak situated between the conurbations of Manchester to the west and Sheffield to  
39 the east. It is classed as an upland moorland environment with significant peat bog  
40 coverage and only small areas of coniferous forest to the south and north-west of the  
41 71.93 km<sup>2</sup> study site (figure 1). The habitat therefore differs significantly for fire scar  
42 detection using SAR from the boreal forested areas in Alaska, Canada and Russia  
43 studied by Bourgeau-Chavez *et al.* (1997, 2002, 2007) and Kasischke *et al.* (2007). The  
44 Bleaklow plateau is of high scientific importance which is reflected in its multiple  
45 statutory designations to protect habitat and wildlife; for example the vegetation of  
46 mosses (*Sphagnum* spp.), heather (*Calluna vulgaris*) and cotton-grass (*Eriophorum*  
47 spp.) and upland birdlife such as the Golden Plover (*Pluvialis apricaria*) and Dunlin  
48 (*Calidris alpina*). *Sphagnum* spp. is essential for carbon sequestration with the  
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1 underlying peat being a major carbon store and supporting ecosystem services, which  
 2 wildfires can destroy (Evans and Lindsay 2010).  
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 6 Figure 1. CORINE land cover data for Bleaklow, Peak District National Park, northwest England, and  
 7 showing: GPS outline of the 18 April 2003 fire scar in red; fire log point data 1976 to 2008 courtesy of  
 8 Moors for the Future (MFF) as red dots; and intensity/coherence sampling points as black dots.  
 9

## 10 2.2 SAR data selection

11 A time series of ERS-2 and Advanced Synthetic Aperture Radar (ASAR) SAR images  
 12 was acquired from the Landmap Service. They were originally sourced from the  
 13 European Space Agency (ESA) as level 1b Single Look Complex (SLC) products in  
 14 slant range (table 1). Images were selected which had a small incidence angle, ranging  
 15 from 22.76° - 23.23° because Huang and Siegert (2006) found that the SAR intensity  
 16 signal of fire scars decreased by 0.1 dB for each degree increase in the incidence angle.  
 17 InSAR pairs were acquired for coherence analysis for dates before and after the fire  
 18 (table 2). ERS-2 pairs with baselines less than the Sarmap (2007) recommended 500m  
 19 threshold were selected for three of the four pairs, with 654m for the fourth pair.  
 20  
 21

1 **2.3 SAR pre-processing of intensity data**

2 SARscape 4.2 was used to pre-process the data listed in table 1 and table 2. Intensity  
3 images were produced by focusing and multilooking the data using five looks in  
4 azimuth and one look in range. Different speckle filtering algorithms were assessed for  
5 ERS-2 08/02/03 and ERS-2 28/06/03 (figure 2).

6  
7 The multitemporal Degrandi filter applied to amplitude coregistered images smoothed  
8 speckle most effectively (figure 2 c and f). The filtered images were radiometrically  
9 calibrated and geocoded to British National Grid with a pixel size of 25m. The sigma  
10 nought ( $\sigma_0$ ) radar backscatter coefficient values were obtained in dB in ENVI © band  
11 math using equation (1).

$$10 * \log_{10}(b_1) \tag{1}$$

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14  
15 The dB filtered images were degraded in ENVI to a pixel size of 100m using a nearest  
16 neighbour resampling method to match the spatial resolution of the CORINE land cover  
17 data and to preserve the original pixel values. A standard linear 2% contrast stretch was  
18 used to display all images.

19  
20 **2.4 SAR pre-processing of coherence data**

21 Interferograms were generated for the ERS-2 pairs in table 2, followed by interferogram  
22 flattening to split the low frequency phase from the high frequency (differential) phase.  
23 The flattened interferogram was filtered to reduce phase noise and interferometric  
24 coherence, which was also produced during this pre-processing step. Phase unwrapping  
25 was then applied to the flattened interferogram using a region growing algorithm with a  
26 coherence threshold set at the typical value of 0.20 to avoid unwrapping islands. A  
27 phase-to-displacement conversion was applied to the unwrapped phase and was  
28 geocoded into British National Grid along with the coherence image. Coherence ranges  
29 between 0 – 1 and is: *'The function of systemic spatial de-correlation, the additive*  
30 *noise, and the scene decorrelation that takes place between the two acquisitions'*  
31 (Sarmap 2007, p.144).

32  
33 1 = High coherence (temporal correlation, no change on the ground)

34  
35 0 = No coherence (no correlation, temporal decorrelation, significant change on  
36 the ground over time)

1 Table 1. SAR images used for the intensity analysis. The fire occurred on 18 April 2003 (108 JD). A.  
 2 Ascending Pass D. Descending Pass  
 3

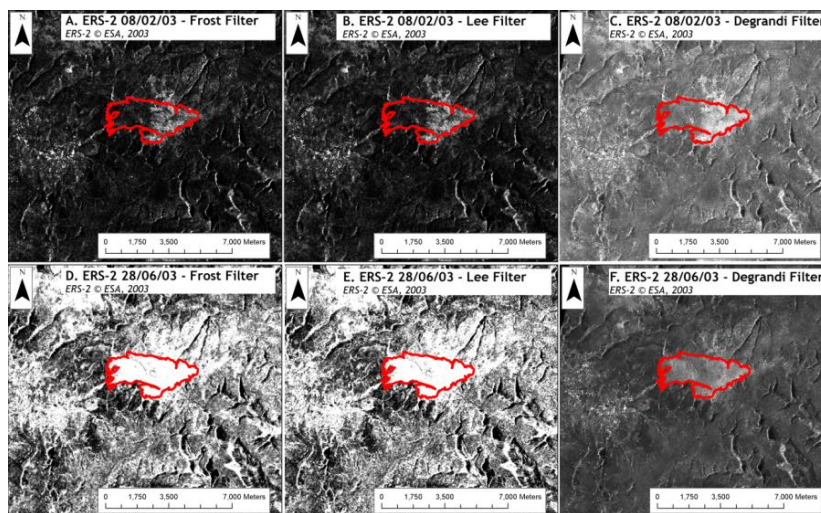
SAR Data/ Mode/ Swath	Acquisition Date/Time (ddmmyyyy hr:mm)	Time relative to fire (JD Julian day)	Incidence Angle (IA)	Az pixel spacing (m)	Rg pixel spacing (m)	Ground Range (GR) (m)	Pass Type
ERS-2	08/02/2003 11:01	-69 days (39 JD)	23.23°	3.97	7.90	20.26	D
ERS-2	15/03/2003 11:01	-34 days (74 JD)	23.23°	3.97	7.90	20.26	D
ASAR IM I2	22/03/2003 21:37	-27 days (81 JD)	22.82°	4.04	7.80	20.00	A
ASAR AP I2 HH	03/04/2003 10:36	-15 days (93 JD)	22.76°	4.04	7.80	20.00	D
ASAR AP I2 VV	03/04/2003 10:36	-15 days (93 JD)	22.76°	4.04	7.80	20.00	D
ERS-2	19/04/2003	+1 day (109 JD)	23.05°	3.97	7.90	20.26	D
ERS-2	24/05/2003 11:01	+36 days (144 JD)	23.21°	3.97	7.90	20.26	D
ERS-2	28/06/2003 11:01	+71 days (179 JD)	23.28°	3.97	7.90	19.75	D

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Table 2. InSAR pairs used for the coherence analysis.

ERS-2	Acquisition Date	Orbit	Track	Baseline (m)	Description
Pair 1	08/02/2003 15/03/2003	40801/41302	366	134	Pre-fire
Pair 2	15/03/2003 19/04/2003	41302/41803	349	349	Pre/post-fire
Pair 3	19/04/2003 24/05/2003	41803/42304	366	147	Post-fire
Pair 4	24/05/2003 28/06/2003	42304/42805	366	654	Post-fire

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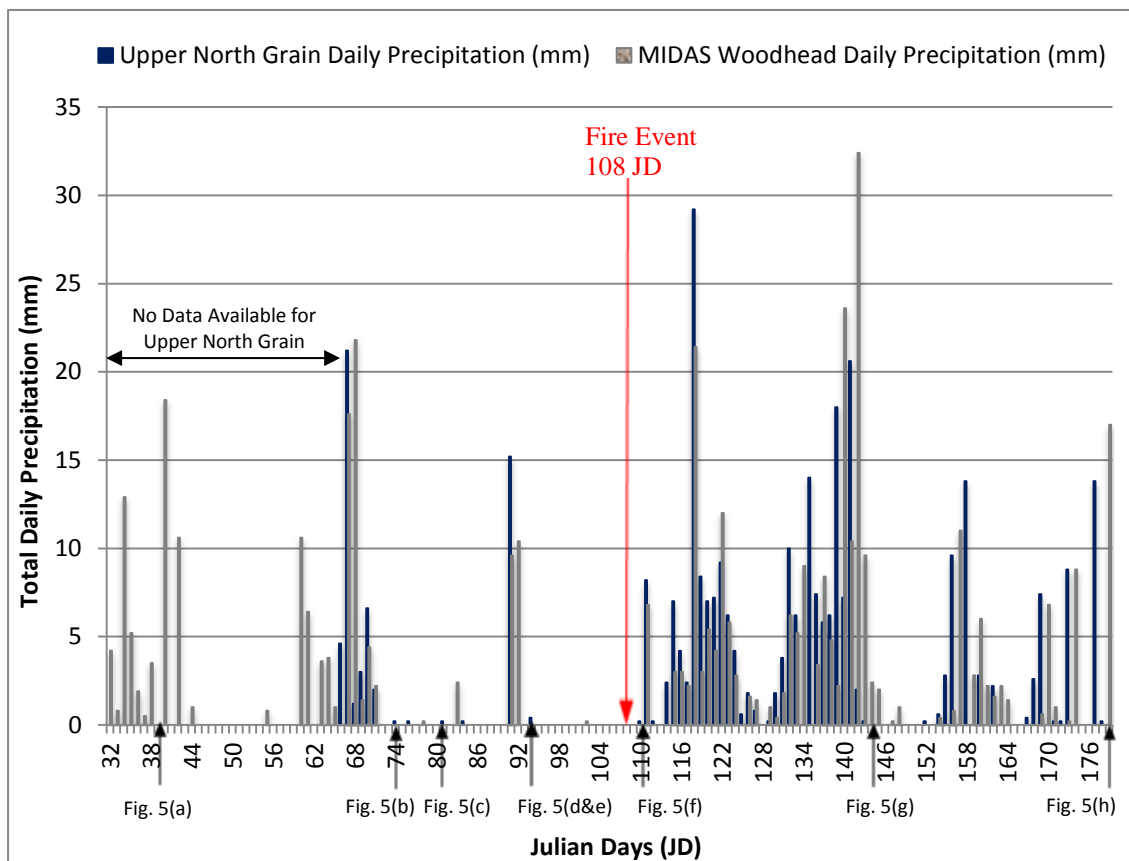


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9 Figure 2. (a-f) Comparison of filtering methods for ERS-2 of the Bleaklow fire scar for 08/02/03 (pre-fire,  
 10 top row) and ERS-2 28/06/03 (post-fire, bottom row): Frost (a,d), Lee (b,e), multitemporal Degrandi (c,f).  
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2 **2.5 Land Cover and Wetness Factors**

3 ArcGIS 10 was used to extract SAR intensity and InSAR coherence values using a  
4 stratified random sampling technique. The sample consisted of 15 sample points in total  
5 inside the fire scar and 15 sample points in total outside the fire scar for each of the  
6 three CORINE land cover classes: peat bog, moors and heathland, and natural grassland  
7 (figure 1). The average of 5 sample points at a spatial resolution of 100m per pixel for  
8 each land cover type was adequate for capturing the spatial accuracy without losing the  
9 spatial precision of the data for land cover classes both inside and outside the fire scar.

10 To provide an indirect indicator of surface wetness, total daily precipitation was  
11 collected from the Upper North Grain (UNG) weather station, located 4 km southeast of  
12 the fire scar (figure 3). The UNG precipitation data starts on JD 66. However, the SAR  
13 time-series begins at JD 39, so precipitation data from the Met Office Integrated Data  
14 Archive System (MIDAS) for Woodhead Reservoir (WR) weather station is provided  
15 for the first SAR image of the time-series (figure 3). UNG was used as the primary  
16 precipitation dataset for the remaining SAR images. There is high spatial and temporal  
17 variability of precipitation events, as shown in figure 3; magnitude and duration varies  
18 between the two sites but the temporal pattern is consistent.  
19



20  
21 Figure 3. Total daily precipitation (mm) from 01/02/03 – 28/06/03 at Woodhead, 4km north of the fire  
22 scar data provided by MIDAS hosted by The British Atmospheric Data Centre (BADC) and total daily  
23 precipitation (mm) from 07/03/03 – 28/06/03 at Upper North Grain, 4km southwest of the fire scar.  
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1  
2 **3. Results and discussion**

3 **3.1 SAR Intensity**

4 Figure 4 shows how average SAR intensity values for CORINE land cover classes  
5 inside and outside the Bleaklow fire scar change over time. Sites outside the fire scar  
6 were included to establish the baseline variability due to other factors such as vegetation  
7 seasonality and heterogeneity. Sites of the same land cover classes inside and outside  
8 the fire scar were compared to assess the background effect of land cover on the fire  
9 signal.

10  
11 *3.1.1 Pre-fire intensity analysis*

12 Figure 5 shows a time series of SAR intensity images before the fire (figures 5a-e)  
13 and after it (figures 5f-h). The CORINE class ‘peat bog’ represents exposed, badly  
14 eroded peat created by earlier fire events. Inside the fire scar, peat bog is located at the  
15 eastern end. Peat bog inside the fire scar had the overall highest pre-fire intensity (figure  
16 4); with an average intensity value of 0.49 dB for the ASAR Alternating Polarisation  
17 (AP) VV and HH images (figures 5d-e). The peat bog is spatially detected in the pre-fire  
18 images as greater backscatter of the radar signal, visualised by bright pixels (figure 5a-  
19 e). This spatial pattern is less distinct in the ASAR IM image (figure 5c). The peat bog  
20 class is spatially heterogeneous across the study site; inside the fire scar, its pre-fire  
21 SAR intensity signal in figure 5a-b is 6-7 dB higher than outside. The reason for this  
22 could be increases in soil moisture as a result of precipitation, which is detected inside  
23 the fire scar due to the eroded nature of the peat bog class in this area, caused by  
24 previous fire events (figure 1, red dots). Similar results have been reported in boreal  
25 forested areas in Canada by Bourgeau-Chavez *et al.* (2002) During JD 72 – JD 90, little  
26 precipitation occurred, followed by one notable precipitation event on JD 91 with  
27 15.2mm of precipitation (figure 3). SAR intensity increased in response to this event  
28 for all land cover types, irrespective of whether the land cover class was inside or  
29 outside the future fire scar (JD 108); differences between classes of the same type were  
30 reduced to 1 – 2 dB (figure 4).

31 After the precipitation event on JD 91 and before the fire event, there was a two  
32 week dry period which is expressed as an overall downward trend in SAR intensity  
33 signal for all land cover classes. This agrees with Gimeno *et al.* (2004), who found that  
34 scenes taken during the dry season exhibit lower intensity values (figure 4).



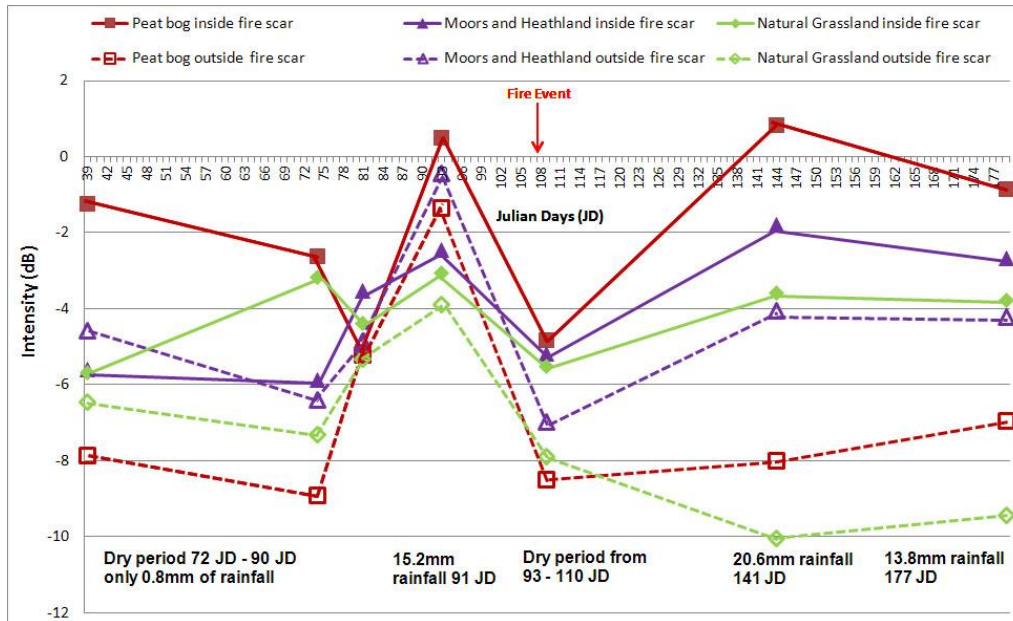


Figure 4. Average intensity values (dB) inside and outside the Bleaklow fire scar for CORINE land cover classes. NB. 93 JD provides the average intensity for AP HH/VV shown in fig.5d and fig.5e.

### 3.1.2 Post-fire intensity analysis

The Bleaklow fire occurred on JD 108, with the first post-fire image acquired on JD 109, two days before the end of the dry period (figure 3). The intensity values for all CORINE land cover classes decreased by at least 3 dB during the dry period (JD 93 – JD 110). The SAR response to the fire, however, was not significantly different from the previous dry period (JD 72 – JD 90), so it is difficult to establish whether SAR is responding primarily to dry conditions, the fire event itself, or to a combination of both (figure 4). However, one day post-fire, figure 5f(iii) shows a new spatial pattern within the fire scar; for the first time, a general brightening is seen in the western part of the fire scar within the former moors and heathland class, in response to the combustion of the above ground biomass.

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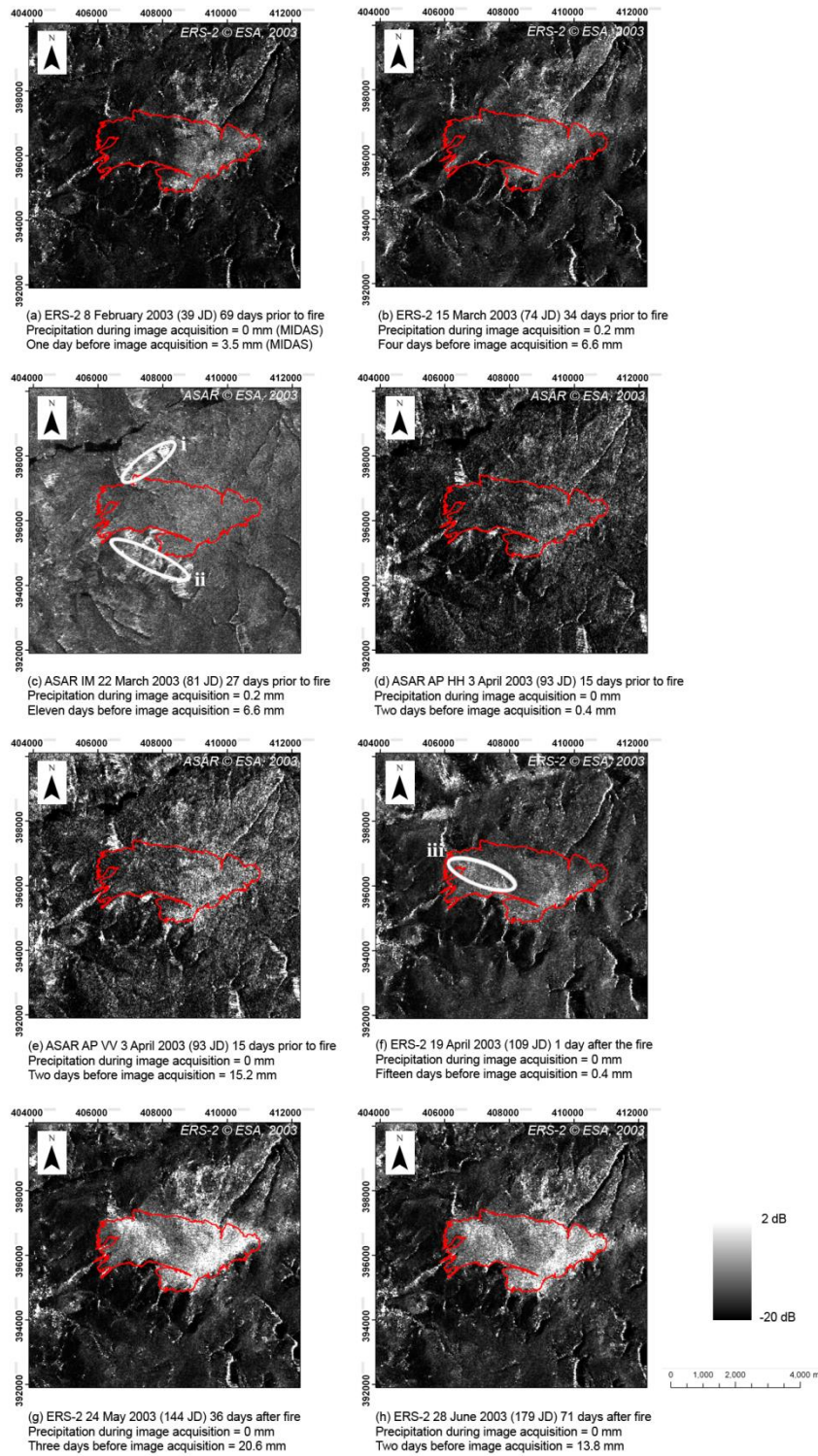


Figure 5. Time series of SAR intensity images for the Bleaklow 18 April 2003 wildfire. Key spatial patterns are annotated (i / ii) ascending pass topographic features and (iii) removal of moors and heathland vegetation after the fire. N.B. Precipitation values for (a) MIDAS (b-h) Upper North Grain.

1 Table 3. Spatial differences in intensity response for land cover classes during the post-fire wet period

Wet Period: Land Cover Class	Inside Fire Scar	Outside Fire Scar	Intensity Difference dB
144 JD: Peat bog	0.84	-8.00	7.16
144 JD: Moors and Heathland	-1.79	-4.06	2.27
144 JD: Natural Grassland	-3.59	-10.04	6.45
179 JD: Peat bog	-0.86	-6.96	6.10
179 JD: Moors and Heathland	-2.68	-4.21	1.53
179 JD: Natural Grassland	-3.78	-9.43	5.65

2  
3 The intense Bleaklow fire was followed by a very wet post-fire period from JD 111  
4 onwards (figure 3). Differences in post-fire response of the intensity signal for land  
5 cover classes inside the fire scar were compared to the same land cover classes outside  
6 the fire scar to assess sensitivity of fire scar persistence to land cover type. Differences  
7 were most evident on JD 144 and JD 179 when precipitation had occurred (table 3).  
8 Significant precipitation events were recorded at the adjacent weather stations e.g. UNG  
9 20.6mm on JD 141 and at Woodhead – 23.6mm on JD 140 and 32.4mm on JD 142  
10 (figure 3). All three land cover classes increased in intensity inside the fire scar. This is  
11 most pronounced for the peat bog class, with intensity values 7.16 dB higher inside the  
12 fire scar compared to those outside on JD 144, and 6.10 dB higher inside the fire scar  
13 compared to outside on JD 179. During this wet period, the pre-existing exposed peat  
14 bog exhibited the highest intensity values at 0.84 dB (JD 144) and -0.86 dB (JD 179).  
15 We believe the most likely explanation is that the intense heat from the fire event  
16 caused the underlying and previously exposed peat to char and become hydrophobic  
17 (Atanassova and Doerr 2011). The hydrophobic nature of the peat provides ground  
18 conditions for shallow depression pools of water to form (around 10cm – 30cm in size)  
19 after several significant precipitation events, which then causes radar backscatter to  
20 increase due to the dielectric properties of water (figure 4 and figure 5g-h).

21 Natural grassland shows a similar pattern to peat bog, with SAR intensity inside the  
22 fire scar 6.45 dB higher on JD 144 and 5.65 dB higher on JD 179. We suggest that the  
23 increased intensity for natural grassland is explained by combustion of the grassland  
24 biomass, which then made the underlying newly exposed peat more prone to surface  
25 wetness. Outside the fire scar, natural grassland is the class with the lowest SAR  
26 intensity on both post-fire dates; -10.04 dB on JD 144 and -9.43 dB on JD 179. This low  
27 backscatter return could be due to fine grassland vegetation creating a specular  
28 scattering effect on the short wavelength of the C-band data, as the grassland would  
29 appear smooth and uniform so reducing the return of the SAR signal producing a  
30 similar dark response to the reservoirs (figure 5g-h). The intensity difference due to fire  
31 is not as pronounced for moors and heathland, as can be seen in figure 4.

### 32 3.1.3 Discussion of intensity results

34 The time-series of ERS-2 and ASAR data illustrates that SAR images spatially detect  
35 the Bleaklow fire scar in a moorland habitat. This result also suggests that SAR should  
36 also be investigated as a surrogate for spatially-distributed information on surface  
37 wetness in moorland habitats; it could be a helpful tool (in the long-term) for  
38 understanding soil moisture for such a spatially heterogeneous environment.

39 The combined effect of a fire followed by precipitation events resulted in the area  
40 of the fire scar being clearly detected. Figure 5g-h shows a strong intensity signal  
41 (bright pixels) within the perimeters of the fire scar following the precipitation events.  
42 Bourgeau-Chavez *et al.* (1997, 2002, 2007) reported similar results for boreal burned  
43 forests such as in Alaska where fire scars post-burn were easily detected due to

1 increases in soil moisture caused by snowmelt and rainfall months after the fire event.  
2 Their linear regression results show ERS-2 backscatter positively correlated to soil  
3 moisture for light, moderate and severe burns (Bourgeau-Chavez *et al.* 2007). This  
4 positive correlation between SAR intensity and soil moisture has also been used for  
5 applications such as monitoring ground conditions of wetlands in southern Florida  
6 (Kasischke *et al.* 2003) and can be assumed to apply to moorland environments in the  
7 UK based on the temporal results of this Bleaklow case study. For the current study, it  
8 is not known whether snow melt as well as precipitation would have contributed to any  
9 variation in soil moisture, as is identified in Kasischke *et al.* (2007). It is unlikely during  
10 the fire event period in April, but could be a contributing factor for peaks in SAR  
11 intensity for the badly eroded peat bog class inside the fire scar (bare peat) during the  
12 winter and early spring period (figure 5a-b). A wet post-fire period is typical in UK  
13 moorlands, indeed Fire and Rescue Services find it difficult to extinguish deep-seated  
14 smouldering peat fires such as this, and it is normal for them to be extinguished in due  
15 course by precipitation.

16 To summarise, it is evident from figure 5 that a large wildfire event which is  
17 followed (as is typical) by heavy precipitation in UK moorlands, can produce a fire scar  
18 which is detectable spatially on the SAR scenes. Furthermore, it could be detected with  
19 regular SAR images as much as 71 days after the fire event, which is a benefit of using  
20 radar data as compared to optical or thermal data. Table 3 shows that the persistence of  
21 the wildfire scar in the landscape is strongest for the peat bog and grassland classes.  
22 Potentially this technique could save time and money in monitoring the impact of  
23 wildfire events and is capable of providing results several months after the event.  
24

### 25 **3.2 SAR Coherence**

26 SAR coherence was used as another parameter to detect the Bleaklow fire scar. Four  
27 pairs of ERS-2 InSAR coherence was generated (table 2 and figure 6). SAR coherence  
28 was low for the first InSAR pair, which means that there was significant change on the  
29 ground pre-fire for all three land cover classes and could be due to phenological change  
30 during this 35 day period. It ranged from 0.14 – 0.24, depending on the CORINE land  
31 cover class (figure 7). There was an unexpected slight increase in coherence for the  
32 second InSAR pair, shortly before and immediately after the fire. The exception was,  
33 the burnt natural grassland inside the fire scar, which remained constant at 0.19 (figure  
34 7). It had been expected that combustion of biomass would have decreased coherence  
35 inside the fire scar for moors and heathland and natural grassland classes.

36 The first of the post-fire InSAR pairs (pair 3 JD 109 – JD 144) (figure 6c) showed a  
37 noticeable increase in coherence for all three CORINE land cover classes inside the fire  
38 scar. This was especially true for the burnt moors and heathland class where the  
39 coherence value increased by 0.18, from 0.29 – 0.47 between the second and third pairs.  
40 A similar 0.17 increase was found for burnt natural grassland, from 0.19 – 0.36. In  
41 comparison, the coherence of burnt peat bog did also increase between the second and  
42 third coherence pair, but only by 0.07, which is less than half the coherence response of  
43 burnt moors and heathland and natural grassland. Visually, the post-fire increase in  
44 coherence can be seen by the marked relative brightness of the moors and heathland and  
45 natural grassland pixels on the west side of the fire scar compared to equivalent areas of  
46 moors and heathland and natural grassland outside the fire scar (figure 6c). This  
47 suggests that it was the biomass combustion within the fire scar keeping land cover  
48 constant whereas unburned areas changed most likely due to phenology.  
49

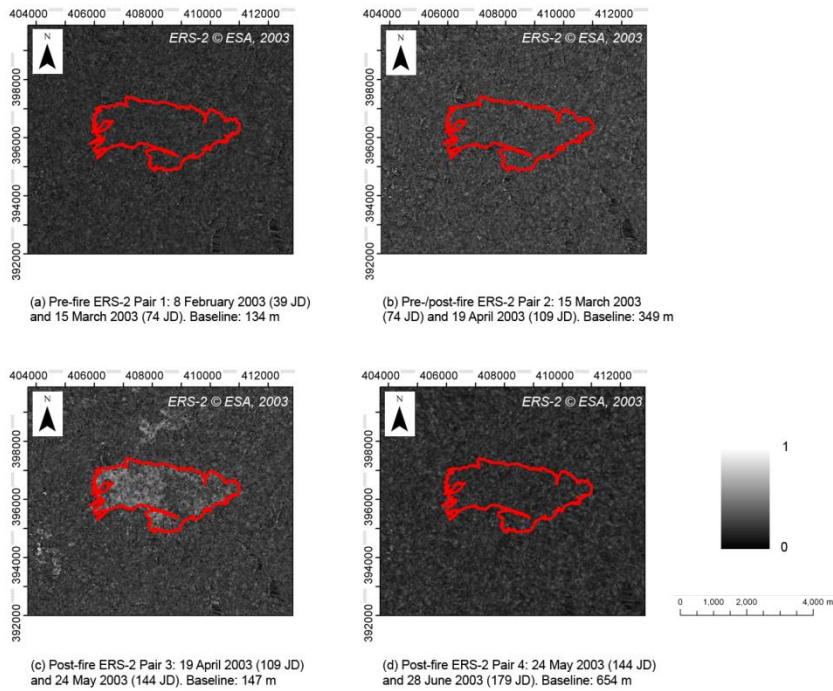


Figure 6. InSAR coherence images for Bleaklow. Each produced from a pair of SAR images (Table 2)

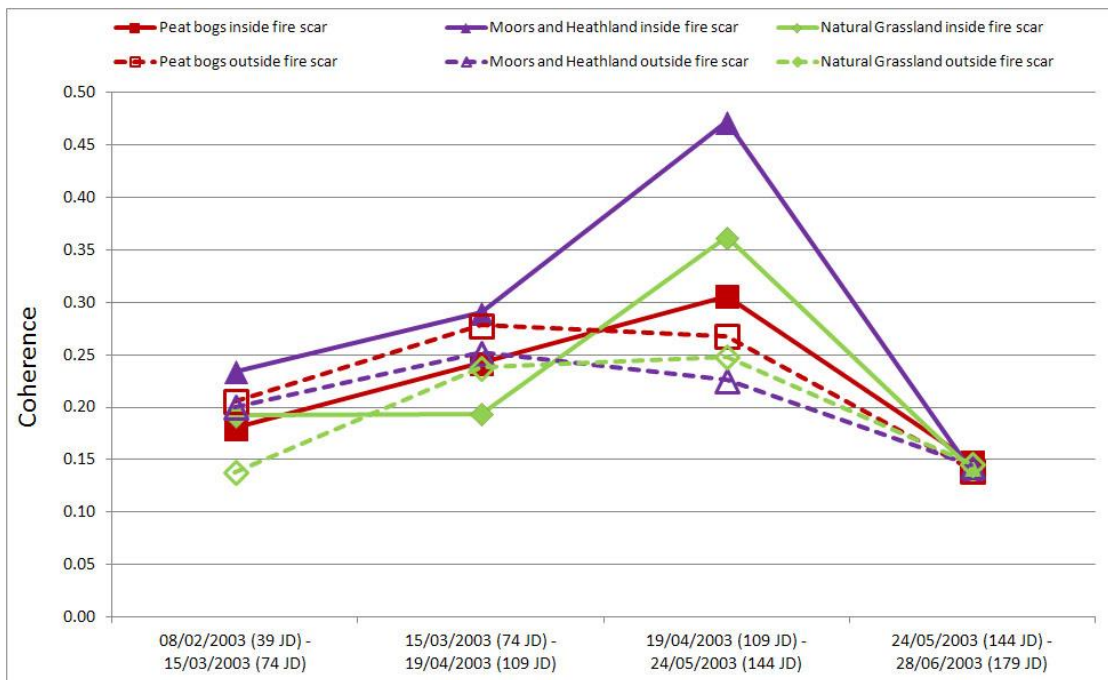


Figure 7. Trend in average coherence of three CORINE land cover classes, inside and outside the Bleaklow fire scar. Fire date 18 April 2003 (JD 108).

For each land cover in turn, the aim was to test whether *temporal* variability in coherence (between the four coherence pair dates) was greater than their *spatial* variability (within the coherence sample points). A series of null hypotheses ( $H_0$ ) were set-up that there was no significant difference in the variance between the four InSAR coherence pairs compared to the variance within the pairs. The Analysis of Variance (ANOVA) test was run separately for the three land cover types, both inside the fire scar

1 and outside the fire scar (6 ANOVA runs in total, table 4). The F statistic compares the  
 2 between group variance (between ERS-2 InSAR pairs over time for a given land cover  
 3 class) and the within group variance (spatially within the five coherence sample points  
 4 for a given land cover class) (table 4). The variation within the five coherence sample  
 5 points for each land cover class is not significant. However the variation between ERS-  
 6 2 InSAR pairs for a given land cover class was significant. The land cover classes  
 7 exhibiting the greatest between group variance were the moors and heathland class  
 8 inside the fire scar, with an F statistic of 48.30; natural grassland inside the fire scar  
 9 with an F statistic of 23.88 and peat bog outside of the fire scar with an F statistic of  
 10 7.75. The F statistic for all classes exceeded the F critical value of 3.24, confirming that  
 11 there is a significant difference between the ERS-2 InSAR coherence pairs at the 5%  
 12 significance level, so the null hypothesis can be rejected. The P-value indicates that  
 13 there would be a very low probability ( $< 0.05$ ) that the calculated F value had been  
 14 obtained by chance (random error) alone. The ANOVA test provides evidence that there  
 15 is a significant difference temporally between InSAR coherence pairs for all land cover  
 16 classes, but especially for natural grassland and moors and heathland *inside* the fire scar.

17 Figure 7 shows that for all land cover classes outside the fire scar the InSAR  
 18 coherence remained low. The ANOVA test also shows much lower F statistics for land  
 19 cover classes located outside the fire scar from 3.52 – 7.75. This result suggests that  
 20 there was change occurring on the ground within these unburned land cover classes,  
 21 which could be related to phenological changes of vegetation especially for the natural  
 22 grassland and moors and heathland classes. The removal of vegetation by the fire  
 23 reduced this background phenological signal within the fire scar. Pre-fire land cover is  
 24 therefore a key factor affecting the coherence response; pre-existing exposed peat from  
 25 earlier fires, which burnt again in the April 2003 fire, showed a smaller increase in  
 26 coherence; that is, it exhibited more change post-fire compared to newly burnt  
 27 vegetation.  
 28  
 29  
 30

Table 4. Analysis of Variance Test (ANOVA) to compare the four ERS-2 coherence pairs

<b>ANOVA Test</b>	Sum	df	F stat	P-value	F critical
Source of variation for coherence pairs: BG = Between Groups of four ERS-2 pairs for a given land cover type (temporal); WG = Within Groups coherence sample points for a given land cover type (spatial)	of Sq				
<i>BG: Peat bog inside fire scar</i>	0.07	3	4.74	0.015	3.24
<i>WG: Peat bog inside fire scar</i>	0.08	16			
<i>BG: Moors &amp; heathland inside fire scar</i>	0.30	3	48.30	$3.046 \times 10^{-8}$	3.24
<i>WG: Moors &amp; heathland inside fire scar</i>	0.03	16			
<i>BG: Natural grassland inside fire scar</i>	0.14	3	23.88	$3.773 \times 10^{-6}$	3.24
<i>WG: Natural grassland inside fire scar</i>	0.03	16			
<i>BG: Peat bog outside fire scar</i>	0.06	3	7.75	0.002	3.24
<i>WG: Peat bog outside fire scar</i>	0.04	16			
<i>BG: Moors &amp; heathland outside fire scar</i>	0.03	3	4.09	0.025	3.24
<i>WG: Moors &amp; heathland outside fire scar</i>	0.04	16			
<i>BG: Natural grassland outside fire scar</i>	0.05	3	3.52	0.039	3.24
<i>WG: Natural grassland outside fire scar</i>	0.08	16			

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1 Exceeding the critical baseline for InSAR pre-processing is the most likely cause of  
2 the overall downward trend in coherence for all CORINE land cover classes for the  
3 fourth InSAR pair, as the initial baseline is 654m, which is higher than the maximum  
4 recommended baseline of 500m (figure 6d). In addition, there is likely to have been  
5 much change on the ground, such as phenological change of vegetation outside the fire  
6 scar and also the continuation of a reseedling programme on the east side of the fire scar  
7 which could contribute to the decrease in coherence. If the fourth InSAR pair had not  
8 been affected by exceeding the critical baseline, the expected result would be for  
9 coherence to remain high within the fire scar, as was the case for the third InSAR pair.

#### 10 11 **4. Conclusion**

12 This paper has shown that in a degraded moorland environment, SAR intensity can be  
13 used as a tool to spatially detect a large wildfire scar, especially where peat is left  
14 eroded by previous fire events and remains unvegetated. The ERS-2 and ASAR images  
15 detected bare peat from previous fire occurrence in the pre-18 April 2003 images where  
16 it dominates the CORINE peat bog class inside the future fire scar area. Pre-fire  
17 precipitation events further enhanced this pre-existing bare peat as a bright area on the  
18 SAR intensity images by increasing soil moisture.

19 Post-fire, from JD 108, soil moisture repellency is likely to increase due to the  
20 exposure of peat to intense heat (Atanassova and Doerr 2011). Depression pools of  
21 water on the burnt peat surface form as a result of intense precipitation events and  
22 enhance the SAR backscattering effect. Such precipitation events are common after a  
23 fire event in a moorland environment.

24 Detectability of the fire scar using SAR intensity varied with the pre-fire land  
25 cover. The CORINE peat bog class was the most responsive, with SAR intensity values  
26 being 6 – 7 dB higher on newly re-burnt peat bog relative to peat bog outside the fire  
27 scar. Pre-existing land cover (explained here partly by fire history) combined with  
28 precipitation are therefore key factors jointly enhancing fire scar detectability on SAR  
29 intensity images. InSAR coherence provides another SAR parameter for aiding the  
30 detection of moorland fire scars. The most marked change was an increase in coherence  
31 across all three classes inside the newly-created fire scar between JD 109 and 144.  
32 Coherence increased in this post-fire period most markedly for burnt moors and  
33 heathland and natural grassland classes, responding to the loss of biomass which  
34 occurred on JD 108. A complementary decrease in coherence for the same classes in  
35 unburned areas is likely due to phenological change.

36 The importance of SAR parameters and pre-processing method should be  
37 acknowledged. For example, the recommendations are: to use images with a small  
38 incidence angle, apply a multi-temporal filtering algorithm such as Degrandi; and  
39 choose an InSAR pair for coherence analysis which has a low baseline (less than 500m)  
40 (Sarmap, 2007). Equally, understanding environmental variables such as precipitation,  
41 fire history and land cover is critical in interpreting the SAR signal both for SAR  
42 intensity and InSAR coherence.

43 This study has also shown how a wildfire in a degraded peat moorland environment  
44 can detect a fire scar that persists in the landscape for months after the fire event, or  
45 even years when fire history of the area is considered. In this analysis, SAR intensity  
46 data still detected the Bleaklow 18 April 2003 wildfire scar 71 days after the fire event.  
47 This result is encouraging and a unique finding for moorland environments. It suggests  
48 that SAR is a potential tool for assessing the extent and location of moorland wildfire  
49 scars and their long-term recovery at the landscape scale, where cloud can be an issue  
50 for optical data and thermal data only provides a short-term signal.

1 Future studies are likely to investigate the use of SAR intensity and coherence for  
2 detecting fire scars of different sizes, precipitation history, topography and less  
3 generalised land cover classes in the PDNP in order to disentangle controls on the SAR  
4 signal. A spatial analysis approach involving SAR image classification of both SAR C-  
5 band and L-band data may be explored for automating the detection of wildfire scars for  
6 degraded peat moorland environments such as this.

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