- 1 Near-real-time service provision during effusive crises at Etna and Stromboli:
- 2 Basis and implementation of satellite-based IR operations
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8 Abstract. Using the NEODAAS-Dundee AVHRR receiving station (Scotland), NEODAAS-Plymouth can 9 provide calibrated brightness temperature data to end users or interim users in near-real time. 10 Between 2000 and 2009 these data were used to undertake volcano hot spot detection, reporting 11 and time-average discharge rate dissemination during effusive crises at Mount Etna and Stromboli 12 (Italy). Data were passed via FTP, within an hour of image generation, to the hot spot detection 13 system maintained at Hawaii Institute of Geophysics and Planetology (HIGP, University of Hawaii at 14 Manoa, Honolulu, USA). Final product generation and quality control was completed manually at 15 HIGP once a day, so as to provide information to onsite monitoring agencies for their incorporation 16 into daily reporting duties to Italian Civil Protection. We here describe the processing and 17 dissemination chain, which was designed so as to provide timely, useable, quality-controlled and 18 relevant information for "one voice" reporting by the responsible monitoring agencies.

19

20 Introduction

21 The 1980's saw a number of studies that used Advanced Very High Resolution Radiometer (AVHRR) 22 mid-infrared (MIR) and long-wave infrared (TIR) data to detect, track and measure the spatial and 23 temporal occurrence of natural fires and anthropogenic hot spots, such as those associated with oil 24 platforms and industry (e.g., Matson and Dozier, 1981; Muirhead and Cracknell, 1984; 1985). In her 25 review, "Fire from space: Global fire evaluation using infrared remote sensing", Robinson (1991) 26 listed 14 papers that focused on such efforts using AVHRR data between 1980 and 1989, to which a 15th can be added: the study of Dozier (1980) (Table 1). As part of these efforts, the decade spanning 27 28 1985 to 1995 saw the development of a number of algorithms to detect wild fires in AVHRR, as well 29 as GOES-Imager, data. Table 2 flags the paper of Flannigan and Vonder Haar (1986) as the first

publication of an automated fire detection algorithm, where we then tabulate 11 different
 algorithms developed during following nine years, with a 12th – the "*Contextual algorithm for AVHRR fire detection*" of Flasse and Ceccato (1996) – being published in 1996.

33 Algorithms used to detect hot spots in satellite-sensor data, as developed by the fire and 34 volcanological communities, can be split into three classes depending on way in which the algorithm 35 defines a hot spot (Steffke and Harris, 2011). Fixed threshold algorithms use single or multiple test 36 and thresholds to determine whether a pixel is hot or not, assessing whether the target pixel's 37 spectral character flags it as anomalously hot. Contextual algorithms assess the pixel's spatial 38 context, using statistics from the target pixel's immediate image background to assess whether the 39 pixel brightness is significantly different from that of its surrounding pixels or not. Finally, temporal 40 algorithms assess whether the pixel brightness is significantly different from that of its proceeding 41 history, thus determining whether a pixel is thermally anomalous in a temporal sense. As a result of 42 the work reviewed in Tables 1 and 2, the fire community had defined the basis of fixed threshold and 43 contextual algorithms by 1995. Of the algorithms collated in Table 2, seven algorithms were fixed 44 threshold and five were contextual. These fire detection algorithms, and their physical basis, 45 underpinned many of the volcanic hot spot detection algorithms that followed. Importantly, the 46 concept of ΔT was established by the fire community, being used by seven of the algorithms of Table 47 2. That is, the differing sensitivities of the MIR and TIR to a sub-pixel hot spot will mean that the 48 pixel-integrated temperature for the hot-spot pixel will be higher in the MIR than in the TIR. In the 49 example given by Harris (2013), a 2 m radius volcanic vent at 950 °C is set against a 0 °C background 50 in a 1000 m AVHRR pixel, with solar-heated pixels being apparent lower on the volcanoes flanks at 51 40 °C. For this case, the MIR pixel-integrated temperature (T_{MIR}) is 11 °C, but in the TIR the pixel-52 integrated temperature (T_{TIR}) is 0.04 °C, i.e., colder than the pixels lower on the volcanoes flanks at 53 40 °C. However, if we subtract the brightness temperature in the TIR from that in the MIR we have a 54 difference ($\Delta T = T_{MIR} - T_{TIR}$) of ~10 °C. If we take the surrounding solar heated pixels at 40 °C, the 55 temperature will be approximately the same in both wavebands, so that ΔT is ~0 °C. Now, the hot 56 spot that was not resolvable using one waveband of data becomes resolvable using ΔT . That is, it 57 shows up as a value of 10 °C against a flat background of near-zero values.

58 Based on advances made by the fire community, the first automated detection algorithm for volcanic 59 hot spots, the VAST (Volcanic Anomaly SofTware) code of Higgins and Harris (1997), was introduced 60 in 1995 (Harris et al., 1995a). Written in ANSI C and made generally available through download 61 from the *Computers & Geosciences* web-site, VAST was initially tested on AVHRR data for Etna and 62 later on AVHRR data for Australian wild-fires burning around Sydney (Harris, 1996). VAST was a

63 contextual algorithm that used ΔT . Shortly thereafter, the first temporal algorithm – the Robust 64 AVHRR Techniques (RAT) algorithm - came on-line (Tramutoli, 1998). Later renamed the Robust 65 Satellite Technique (RST) the algorithm relied on an archive of MIR data to create an Absolutely Local 66 Index of Change of Environment (ALICE) (Pergola et al., 2008; 2009). ALICE provided an estimate of 67 how much a pixel brightness diverged from its normal conditions as determined from the data time 68 series, normalized for its natural variability in the time domain so as to detect temporally anomalous 69 behavior including volcanic hot spots (e.g., Di Bello et al. 2004). In 2000, the now widely-used 70 MODVOLC system became operational (Flynn et al., 2002; Wright et al., 2002). Based on a detection 71 routine that used a fixed threshold algorithm based on the ΔT principle, the normalized thermal 72 index (NTI), MODVOLC provided a simple global hot spot detection capability that required a minimal 73 number of mathematical operations (Wright et al., 2002; 2004). Thus, as of 2000, a number of 74 volcano hot spot satellite-sensor detection and reporting systems were operational, including the 75 Okmok algorithm (Dehn et al., 2000). This was developed at the Alaska Volcano observatory (AVO) 76 to aid with operational hot spot detection in AVHRR data. As part of the AVO function, over 100 77 volcanoes across Alaska, the Aleutians, Kamchatka, and the northern Kurile islands were monitored 78 in as close-to-real-time-as-possible using direct reception of AVHRR, GOES and GMS data at a 79 receiving station installed at the University of Alaska (Fairbanks) in 1990 (Dean et al., 1996; 1998). 80 We here explore the implementation and utility of an operational satellite-sensor based hot spot 81 detection and tracking system launched in 2000 and still, like the MODVOLC system, operational 82 today.

83 The Natural Environment Research Council (NERC) Earth Observation Data Acquisition and Analysis 84 Service (NEODAAS) is funded by NERC to support UK research scientists with remote sensing data 85 (http://www.neodaas.ac.uk/). The service has the capability to automatically receive, process, and 86 archive data from multiple polar-orbiting sensors, including MODIS and AVHRR, in near-real time. 87 Data are also received and processed from multiple geostationary satellites, including SEVIRI, VISSR, 88 GOES and MTSAT (Groom et al., 2006). Between 2000 and 2009, AVHRR data supplied in near-real 89 time by NEODAAS were used to communicate hot spot information during effusive crises at Mt. Etna 90 and Stromboli (Italy). We here describe this data reception, processing and communication chain.

91

92 The satellite data: reception and pre-processing

The NEODAAS service is hosted at two sites. While data reception and acquisition is provided by the
Dundee Satellite Receiving Station at the University of Dundee (NEODAAS-Dundee), data processing
is provided by the Remote Sensing Group at the Plymouth Marine Laboratory (NEODAAS-Plymouth).

96	
97	NEODAAS-Dundee
98	During the 1970s, an AVHRR receiving station was developed at the University of Dundee. Data were
99	archived onto magnetic tapes and a quick-look image archive was maintained, and updated daily, in a
100	photographic format filed in ring-binders. During the 1990s raw data could be ordered on magnetic
101	tape, but delivery delays were of the order of weeks. However, archived AVHRR data were used to:
102	1. Test initial hot spot detection algorithms (Harris et al., 1995a; Pergola et al., 2001);
103	2. Track effusive eruptions through spatial and temporal analysis of spectral radiance (Harris et
104	al., 1995b, 1997a; Tramutoli et al., 2001);
105	3. Develop means for time-averaged discharge rate extraction (Harris et al., 1997b; Harris and
106	Neri, 2002); and
107	4. Define and understand detector response problems over high temperature targets (Setzer
108	and Verstraete, 1994; Harris et al., 1995c).
109	All data received by NEODAAS-Dundee are processed in near-real time and made available on-line.
110	These are automatically added to the online archive, whose sensor and data base coverage is
111	summarized in Table 3. Currently, AVHRR, MODIS and MSG data are received directly at NEODAAS-
112	Dundee and two products are generated:
113	Level 0: unprocessed instrument data;
114	• Level 1: geolocated unprocessed instrument data including calibration parameters.
115	For AVHRR, coverage extends from Newfoundland to Moscow and from North Africa though
116	Greenland (Figure 1a). Image frequency depends on location (Figure 1b), with up to five images a
117	day being available for Etna, although at-least two may be close to the scan edge and, although
118	useful for event detection, are potentially difficult to use quantitatively (Harris et al., 1997b).
119	

120 NEODAAS-Plymouth

During the 1990s, the Plymouth Marine Laboratory (at the time operating under the auspices of the
 Remote Sensing Data Analysis Service) provided calibration coefficients for conversion from DN to
 spectral radiance, upon agreement. Also provided were on-board blackbody temperatures,

necessary for non-linear correction of the conversion. Today, direct-broadcast data received at
Dundee are transferred over the internet to Plymouth where higher level processing is undertaken.
AVHRR data are processed into sea-surface temperature following Miller et al. (1997), and MODIS
data are processed into ocean color and atmospheric products (Shutler et al., 2005); typically 1-2
hours after reception (Groom et al., 2006). Global coverage is available from NASA and ESA sourced
data and from geostationary archives for Meteosat, MSG, IODC, GOES-East, GOES-West and MTSAT.
Data are processed to provide three further levels of product:

- Level 2: derived variables, such as vegetation indices, sea surface temperature, ocean color
 and atmospheric properties;
- Level 3: temporally or spatially binned level 2 data, such as daily, weekly or monthly
 composites for a single product or region;
- Level 4: derived from multiple measurements or models e.g., ocean front analyses.

Third party (user-specified) products can also be generated. Quick look browse of all products are made available via the internet as quickly as possible, with ftp access to products and data being made available to registered users at the same time.

139

140 Data sets for hot spot detection: The role of AVHRR

141 Across the infrared spectral region $(0.7 - 20 \,\mu\text{m})$ there are eight atmospheric windows within which 142 atmospheric transmission is greater than 90 % (Table 4). Hence, wavebands for Earth-surface 143 thermal applications need to be placed in these spectral regions if the surface emission properties 144 are to be measured from the in-orbit location (i.e., above the top of the atmosphere). Following 145 Wein's displacement law, channels placed in the MIR, between 3.44 and 4.13 μ m, and in the TIR, 146 between 8.6 and 12.2 µm, will be most sensitive to surfaces at elevated temperatures (fires and 147 active lavas) and typical Earth surface ambient temperatures, respectively (Table 4). The two 148 wavebands have thus respectively been used for measurements of the two surface types, with the 149 extreme sensitivity of the 3.44 to 4.13 µm waveband to sub-pixel hot spots meaning that it has 150 become informally termed the "fire channel" (e.g., Vermote et al., 2009).

AVHRR's channel 3 has been long known to be extremely sensitive to, and thereby capable of, detecting small, high temperature sub-pixel heat sources, such fires due to straw burning and gas flares on oil platforms (Matson and Dozier, 1981; Muirhead and Cracknell, 1984; 1985). However, AVHRR's channel 3 also has a saturation temperature of between 50 and 60 °C, so that data saturate

155 quite easily over sub-pixel fire and volcanic hot spots (Setzer and Verstraete, 1994; Harris et al., 156 1995c). Solutions have, though, been found to work around this problem and unsaturated thermal 157 data are usually available over hot spots in AVHRR's two TIR, channels 4 (10.3 – 11.3 μ m) and 5 (11.5 158 - 12.5 μm). In addition, AVHRR data represent the longest continuous meteorological satellite data 159 set, with the NEODAAS MIR and TIR archive dating back to the first launch of AVHRR on TIROS-N in 160 October 1978. If we consider the AVHRR's predecessor, the Very High Resolution Radiometer, the 161 TIR data set can be extended back to first launch on NOAA-2 in November 1972 (Cracknell, 1997). As 162 of 2015, these archives potentially provided a 43-year-long base-line data set, which for equatorial 163 targets has a nominal temporal resolution of four images per day, increasing to 10 or more towards 164 the poles due to convergence of orbits (Harris et al., 1999). Pixels increase in size from, for channel 165 3B on NOAA's -15, -16 and -17, 1.11 km and circular at nadir to 7.9 × 2.6 km (oblate ellipse) at the 166 edge of the 3000 km wide scan. Pixels will also undergo distortion, become rotated, and become 167 heavily overlapped towards the edge of the ± 55.4 ° wide swath. However these effects can be 168 assessed and corrected for (Harris, 2013). Because of AVHRR's utility and longevity, as of 2005 169 AVHRR accounted for 47 (or 39 %) of the 120 studies published within the field of satellite-sensor 170 based detection, tracking and measurement of volcanic hot spots since 1965 (Harris, 2013).

171 As Robinson (1991) pointed out, although AVHRR was designed for meteorological observations, 172 channel 3 was somewhat "serendipitously well placed" to detect hot spots. However, the utility of 1-173 km satellite-based measurements in the MIR for fire and volcano hot spot studies have led to some 174 sensors, such as MODIS and BIRD, being designed with a high gain setting channel at 3.9 μ m (that 175 saturates at temperatures of up to 400-450 K) with the fire community specifically in mind (e.g., 176 Kaufman et al., 1998; Wooster et al., 2003). In addition, 3.9 µm channels on geostationary satellites, 177 such as the Imager on GOES and SEVIRI on Meteosat, have long proved capable of tracking hot spots 178 due to fires at temporal resolutions of 15 minutes or better, in-spite of having 3-4 km pixels (e.g., 179 Prins and Menzel, 1994; Roberts and Wooster, 2008). Consequently, the high temporal resolution 180 and "fire channel" detection capability offered by sensors mounted on geostationary platforms have 181 proved to be of extreme utility for shot-lived effusive events or activity varying over time-scales of 182 10's of minutes, such as activation and deactivation of active fissure segments (e.g., Harris et al., 183 1997c; Harris and Thornber, 1999; Ganci et al., 2012). Such events may be missed, or just imaged 184 once or twice, by the polar orbiters that carry sensors with thermal capabilities which have return a 185 period of 6-12 h. As a result, today, although NOAA NESDIS (2015a) state that the objective of the 186 AVHRR instrument is to:

187 "provide radiance data for investigation of clouds, land-water boundaries, snow and ice
188 extent, ice or snow melt inception, day and night cloud distribution, temperatures of
189 radiating surfaces, and sea surface temperature;"

190 added is:

"In addition, land use applications of the AVHRR include monitoring of: food crops; volcanic
 activity; forest fires; deforestation; vegetation; snow cover; sea ice location; desert
 encroachment; icebergs; oil prospecting and geology applications. Other miscellaneous
 AVHRR applications include the monitoring of: migratory patterns of various animals; animal
 habitats; environmental effects of the Gulf War; oil spills; locust infestations; and nuclear
 accidents such as Chernobyl."

Thus, volcano monitoring has become established and recognized as part of the application of NOAA
 NESDIS data. In this regard, following NOAA NESDIS (2015b), NESDIS is currently,

"dedicated to providing timely access to global environmental data from satellites and other
sources to promote, protect and enhance the Nation's economy, security, environment and
quality of life."

In terms of compatibility with volcano monitoring, the NOAA NESDIS mission is to (NOAA NESDIS,203 2015b):

204 (i) manage operational environmental satellites,

205 (ii) operate the NOAA National Data Centers,

206 (iii) provide data and information services including Earth system monitoring,

207 (iv) perform official assessments of the environment, and

208 (v) conduct related research.

AVHRR will thus likely continue to be a robust and reliable resource for volcano hot spot monitoring for the fore-seeable future, providing a data base that has its foundations in 55 years of technological and applicative development. As of 2015, six AVHRR sensors were in orbit aboard NOAA's -14 through -19. Although "old", the first satellite in the series having been launched on 1 April 1960 carrying the Vidicon sensor (Cracknell, 1997), the series is by no means obsolete, and is constantly being upgraded,

215 "to support a complete meteorological payload plus the necessary support subsystems to
216 meet all interface and system requirements" (Robel, 2009).

217 At the same time,

218 "NOAA has tried to keep the changes to a minimum" (Robel, 2009)

so as to maintain the continuity of service and the "digital archive of data collected from the current
 generation of NOAA operational polar orbiting satellites" (Kidwell, 1998).

221 The series thus provides a reliable MIR and TIR data set adding, at minimum, four extra 1-km spatial

resolution observation data points to an ensemble-based approach that can over up to 16 "looks"

223 per day if we combine NOAA+METOP+TERRA+AQUA capabilities (Harris et al., 2015). What's more,

the service us underwritten by NOAA, providing continuity of data and a reliable resource.

225

226 Tools used for hot spot tracking

227 AVHRR data provided by NEODAAS-Plymouth for Italy and Iceland were ingested into the Hawaii 228 Institute of Geophysics and Planetology (HIGP) hot spot tracking system. From 2000 onwards, the 229 system ingested near-real time GOES-Imager data to track hot spot activity around the Pacific Rim 230 (Harris et al., 2000a; 2001; 2002a; 2002b), and was linked to the MODVOLC tool. The system 231 produced a number of quick-look image products. These were generated on-the-fly so as to reduce 232 storage space, meaning that only raw data were saved, and then used to generate products from the 233 archive. A rolling text-file data base, containing basic locational and radiance data for each target region of interest (ROI) was also updated with each image acquisition. The same system 234 235 automatically generated an email notice if the probability of any pixel in a ROI exceeded a threshold, 236 linking the recipient to the image products for image that generated the notice (Harris et al., 2002a). 237 This initial threshold was based on a multistep, fixed threshold approach (Table 5) which operated 238 along the lines of the fire detection algorithms given in Table 2. For the NEODAAS-Plymouth AVHRR 239 data, this algorithm was used purely for issuance of email notices which the recipient used to check 240 the veracity of a "detected" hot spot. If the hot spot was valid, then the operator proceeded 241 manually by checking all images to precedent to the notice so as to ascertain the exact start time 242 (within the temporal resolution of the sensor) of the event. Operator analysis continued by checking 243 each new image until the event was over. Manual checking initially involved detecting and logging 244 anomalous pixels by eye. Later hot pixel detection and selection was guided by application of an up-245 dated version of VAST. Pixel radiance values were then converted to useable (by the field

volcanologist and hazard responder) metrics in a timely fashion, these metrics primarily being timeand Time-Averaged Discharge Rate (TADR).

248

249 Manual checking: NEODAAS-Plymouth quick-look hot spot tool

250 **NEODAAS-Plymouth** maintains quick-look hot а spot tool at 251 https://www.neodaas.ac.uk/supportedscience/etna.php (Figure 2). The tool involves an enhanced 252 AVHRR channel 3 image of Sicily (including the Aeolian Islands) and Iceland. Enlargements of both 253 the channel 3 and channel 4 images for Etna are inset into the overview image. The most recent 254 image is given, along with all images acquired during the preceding week. The current image is 255 updated as soon as new data arrive, and the archive can be browsed using forward and backward 256 buttons at the top of the tool. This simple, but effective, system allows the presence of thermal 257 anomalies to be checked by virtue of their intense radiance in channel 3 and, if sufficiently intense, in 258 channel 4 also, as in the case given in Figure 3.

As argued by Harris (2013), visual detection using time series of such images remains the most powerful and trustworthy tool for hot spot detection and tracking. That is,

261 "by examining multiple images of the same volcano target we train a powerful neural 262 network for hot spot detection. In some subtle detection cases it may even be difficult for 263 the operator to describe the detection algorithm they have sub-consciously developed, for 264 the detection is based upon the most complex of neural networks."

In the case of Figure 3, the presence of hot spots due to volcanic activity on Mt. Etna is immediately apparent to the eye. Hence, manual checking – and tools that allow such interaction – remain the most bullet-proof way to ensure that: (i) the hot spot is valid, (ii) all pixels required for complete quantitative analysis are selected, and (iii) no spurious pixels are included.

269

270 Automated checking: VAST-II

VAST's implementation is based on the concept of "*natural variation*" (Harris et al., 1995a). This was defined as the difference between a pixel's brightness and that of its surroundings, as defined by the eight pixels immediately surrounding the target pixel (Figure 4). Although over thermally homogeneous surfaces, such as the sea, natural variation should be close to zero; over thermally heterogeneous surfaces, natural variation should be highly variable (and typically greater than 10 °C).

The updated version of VAST used to guide pixel selection in NEODAAS-Plymouth AVHRR data functioned in the same way as the original version, but used the TIR brightness temperature (T_{TIR}) image, rather than the ΔT image, a modification which allowed application when MIR data was lacking. The new version of VAST also allowed user interaction so that anomalous pixels not flagged by the algorithm could be selected or false positives de-selected (Steffke and Harris, 2011).

281 A 30 \times 30 AVHRR pixel target window was centered at 37.37°N, 15.00°E, the size of which could be 282 altered by the operator. Natural variation ($\overline{\alpha}$) was then defined for all pixels in the image. An image-283 specific natural variation threshold ($\overline{\alpha}_{\text{thresh}}$) was set from the "non-volcanic" portion of the sub-image, 284 this being a 5 pixel wide strip running around the target window, the size of which could also be 285 enlarged or reduced by the operator. The value for $\overline{\omega}_{thresh}$ was set equal to the maximum $\overline{\omega}$ found 286 within the "non-volcanic" zone. Then, if ϖ for a pixel in the target window was greater than ϖ_{thresh} , it 287 was flagged as anomalous. For pixels at the center of a large anomaly $\overline{\omega}$ might be quite low due to all 288 surrounding ΔT or T_{TR} values being high. Thus, anomalous pixels located in the first run were 289 masked, the ϖ value for each pixel recalculated with hot pixels flagged in the first run excluded, and 290 the test re-run. This process was iterated until no new pixels were found (Harris et al., 1995a). 291 Location and spectral radiance vales for detected pixels – as confirmed, rejected or added by the 292 operator - were then exported to a data file and used for conversion to TADR.

293

294 Conversion to TADR

Following Wright et al. (2001), lava area obtained from satellite-sensor IR data can be converted to time-averaged discharge rate following a linear relation:

$$TADR = x A_{lava}$$
(1)

A_{lava} being the area of active lava, and x being the slope of the linear relation between TADR and A_{lava}
for the case in hand (Harris and Baloga, 2009). The satellite-data-derived variable in this relation is
A_{lava}, which we derived as follows:

301	1.	AVHRR channel 3 and channel 4 spectral radiances $R_{3\text{-}int}$ and $R_{4\text{-}int})$ were record	ded foi
302		all detected hot spot pixels (Figure 5).	
303	2.	Atmospheric and emissivity corrections were applied to all selected radiances,	i.e.,
304		In channel 3: $R_{3-int-corr} = (R_{3-int} - R_{reflec}) / \epsilon_{\lambda 3} \tau_{\lambda 3}$	(2a)
305		In channel 4: $R_{4-int-corr} = (R_{4-int} - R_{upwell}) / \epsilon_{\lambda 4} \tau_{\lambda 4}$	(2b)

306		In which ϵ and τ are the surface emissivity and atmospheric t	ransmissivity in	the two
307		wavebands centered at wavelengths $\lambda 3$ and $\lambda 4,$ respective	y. R _{reflec} is the	surface
308		reflected spectral radiance that contributes to the at-sensor p	pixel-integrated	spectral
309		radiance in channel 3, and $R_{\mbox{\tiny upwell}}$ is the atmospherically-em	nitted spectral r	adiance
310		that contributes to the at-sensor pixel-integrated radiance in	channel 4.	
311	3.	For each hot spot pixel, a background radiance (R_{back}) is sele	ected from the r	nearest,
312		non-hot spot, pixel of lowest radiance (Figure 6).		
313	4.	Because AVHRR channel 3 is usually saturated over a volc	anic (or fire) ho	ot spots
314		(Harris et al., 1995c), only one channel is useable – channel	4 (i.e., R _{4-int} and	JR _{4-back} ;
315		which when corrected for emissivity and atmospheric en	ffects are R _{4-int-}	_{corr} and
316		$R_{4-back-corr}$). Thus we are forced to apply a one-waveba	ind method (Ha	arris et
317		al.1997a,b; Harris, 2013) so that channel 4 spectral radiar	nces are used ir	ı a two
318		component mixture model to obtain the portion of the pix	el occupied by	the hot
319		source (p):		
320		$p = (R_{4\text{-int-corr}} - R_{4\text{-back-corr}}) / (R_{4\text{-hot}} - R_{4\text{-back-corr}})$		(3)
321		in which R_{4-hot} is the spectral radiance emitted by the hot (acti	ive lava) source r	resident
322		in the pixel. Due to uncertainty and real variability in the e	xact value of R_4	-hot, this
323		value has to be assumed over a range, where the spectral rad	iance equivalent	s of the
324		temperature limits 100 °C and 600 °C have been found approx	priate for active	e lava at
325		Etna, Stromboli and Iceland (Harris et al., 2007; Harris an	d Baloga, 2009;	Harris,
326		2013). This gives two values for p:		
327		(i) a maximum value (p_{max}) obtained with the low temper	rature assumptio	on (T _I),
328		(ii) a minimum value (p_{min}) obtained with the high tempe	rature assumption	on (T _h).
329	5.	Next, pixel area $(A_{\mbox{\tiny pixel}})$ is calculated as a function of scan ang	le. This is multi	plied by
330		p_{max} and p_{min} to obtain a range of estimates for the area of ac	tive lava residen:	nt in the
331		pixel (A _{lava-p}):		
332		$A_{lava-p-max} = p_{max} A_{pixel}$	(T _I = 100 °C)	(4a)
333		$A_{lava-p-min} = p_{min} A_{pixel}$	(T _h = 600 °C)	(4b)
334		Results for all hot spot pixels are summed to obtain a maximu	im and minimum	า bound
335		on the total area of active lava at the imaged eruption site $(A_{\mbox{\tiny HI}}$	ava-max and A _{lava-min}	n).
336	6.	Finally, to obtain the limits on TADR, the two lava area estimated	ates are placed i	nto two
337		empirically-derived conversion routines tailored for the case	e in hand (Harri	s et al.,
338		2010). For the period 2000-2009 at Etna the following con	versions were fo	ound to

339 produce the best-empirical fit between model output and ground truth (Harris et al.,340 2011):

341
$$TADR_{min} (m^3 s^{-1}) = 5.5 \times 10^{-6} (m s^{-1}) A_{lava-max} (m^2) (T_l = 100 °C) (5a)$$

342
$$TADR_{max} (m^3 s^{-1}) = 150 \times 10^{-6} (m s^{-1}) A_{lava-min} (m^2) (T_h = 600 °C) (5b)$$

343 For Stromboli, we obtained (Calvari et al., 2005):

344
$$TADR_{min} (m^3 s^{-1}) = 2.5 \times 10^{-6} (m s^{-1}) A_{lava-max} (m^2) (T_l = 100 °C) (6a)$$

345 $TADR_{max} (m^3 s^{-1}) = 166 \times 10^{-6} (m s^{-1}) A_{lava-min} (m^2) (T_h = 600 °C) (6b)$

346 Uncertainty

347 Two TADR values are thus output: one each for the maximum and minimum bound on the derived 348 Alava which, in turn, results from uncertainty in (or impossibility of) applying a single temperature 349 value to describe the thermal surface structure of the lava active within the pixel. As is standard, the 350 uncertainty range lies between the highest possible temperature that can realistically be assigned to 351 this source (T_b) and the lowest (T_l) . If the conversion is appropriately set and applied, output TADR 352 have been shown to span the range of field-based estimates (Calvari et al., 2005; 2010; Harris et al., 353 1997a,b; 2007; 2011; Harris, 2013), while also (on occasion) giving a smaller range of uncertainty 354 than field-based estimates made under similar, near-real time, requirements (e.g., Harris and Neri, 355 2002).

356 The range of uncertainty on the satellite-derived TADR is quite large (of the order of ± 60 %); but no 357 current near-real time (even post-mortem) lava flux rate measurement method is without its 358 problems, and all can have large error of ~50 % (Harris et al., 2007). Even the careful, but time-359 consuming, post-emplacement volumetric measurements made of Etna's 1991-1993 lava flow field 360 Stevens et al. (1997) gave (4 years after the event ended) a mean output rate with an uncertainty of 361 12.5%. Uncertainty and error on TADT measurements is an old problem, which still needs to be 362 addressed. The key challenge remains: how do we make regular, and trusted, TADR measurements 363 in near-real time with uncertainties that are less than 10%? That being said, satellite-derived TADR 364 estimates with 50-60% uncertainty have been proved adequate for tracking effusive events, 365 providing time-series sensitive to real changes in output rate (e.g., Harris et al., 1997a,b; 2000b; 366 2010), so as to provide output of use in a monitoring and modeling role (e.g., Bonaccorso et al., 2015; 367 Vicari et al., 2009; 2011).

In the case of Stromboli, validated satellite-derived TADR were particularly useful, inspire of large
 uncertainty, due to (Harris et al., 2005; Calvari et al., 2007; Lodato et al., 2007):

370 (i) difficulty and danger of access to the active lava flow field, which was emplaced on a
371 steep slope receiving rock fall and hot grain flow from the collapsing lava flow fronts;

372 (ii) the braided nature of the channel and tube system, and multiplicity of sources;

- (iii) a lava emplacement mechanism that eroded the loose substrate on which the flows
 were emplaced so that lava units became embedded in the surrounding terrain so that
 thickness measurements were difficult to make; the location of the flow base being
 impossible to obtain;
- (iv) lava flow instability, where flows were constantly collapsing into the sea, so that
 emplaced lava units never remained in place for more than a few hours so that postmortem field-based area (A) and thickness (h) measurements were impossible Meaning
 that TADR derivation from Ah/t, t being duration of unit emplacement were likewise
 impossible on a regular, operational, basis.

382

383 TADR tracking using NEODAAS data during effusive crises at Etna and Stromboli

Beginning in 2000, through collaborative agreement with NEODAAS, calibrated, geo-referenced AVHRR channel 3, 4 and 5 brightness temperature data were made available to HIGP for tiles covering Sicily and Iceland (Figure 7). Data were placed on the NEODAAS-Plymouth ftp site upon generation of the brightness temperature image product. A routine running at HIGP checked the ftp site every minute, downloading any new data found. In this way a mirror archive for these two volcanically active zones was built as part of the HIGP hot spot monitoring system.

Because no algorithm can be 100% trusted to select all anomalous pixels 100% of the time (Steffke and Harris, 2011), and so as to be sure that all images used were free of cloud contamination, we preferred a manual approach. Also, because only around five images were received each day, the task of image checking and processing was manageable by a human operator. The task was more daunting when using the 96 GOES images received per day by the system for active targets in the GOES footprint; but manageable by an individual charged with maintaining the system. Such a philosophy ensured product quality control and accountability.

As described by Moxey et al (2003), operators at HIGP would check the archive at 07:00 am (Hawaiian Standard Time) every morning. All new data would be checked, manually, for cloud cover and presence of a hot spot. The radiances of any hot pixels and their cold background were logged and used to convert to time-averaged discharge rate. For a day's worth of AVHRR data, this process would typically take less than 30 minutes. New vales were appended to a summary table which also 402 contained comments on image quality (e.g., Table 6). This was sent to a controlled email distribution 403 list that involved the main monitoring actors, which included INGV-Catania for eruptions on Etna 404 (Bonaccorso et al., 2015), as well as University of Florence for operations on Stromboli. A hot spot 405 summary table was also maintained (e.g., Table 7). These Tables were appended to a daily up-date 406 email giving, in a standard and consistent format, details regarding the number of images checked, 407 their quality and any TADRs that could be derived (e.g., Table 8).

During Etna's 2001 and 2003 eruptions, as well as during Stromboli's 2002-2003 and 2007 eruptions, TADRs were checked against field measurements to allow an assessment of their reliability (e.g., Figure 8). Results indicated that the empirical conversion applied, that is the x value used in Equation (1), was valid and gave results that agreed with ground-based values (see, for example, Calvari et al. 2005; 2010). In addition, the empirical equation of Calvari and Pinkerton (1998) that relates flow length (L) to TADR,

414
$$L = 10^{3.11} \times TADR^{0.47}$$
 (7)

was applied. For the maximum TADR obtained by AVHRR during Etna's 2001 south flank eruption (30 m³ s⁻¹), this gave a length of 6370 m and compared with a final length of 6.5 km. During the 2001 eruption, this assessment was particularly important because flows were advancing towards the towns of Nicolosi and Belpasso (Figure 9), which lay just a few kilometers further down slope from the flow front (Bonaccorso et al., 2015) and accounted for a total population of 6,959 plus 23,606 for the two population centers, respectively (ISTAT, 2008).

Less useful were the 1 km hot spot location maps, because the location of the lava flows was already well-known by ground observers. The maps, though, were posted on the HIGP hot spots web-site (http://goes.higp.hawaii.edu/goes/etna/) as soon as data had been received from NEODAAS-Plymouth in case of need. Three of these location maps (in raw, labeled and DEM-merged format) are given in Figure 9, and allow – in spite of the 1 km precision – hot spot location to be assessed in relation to vulnerable locations.

427

428 Stromboli: Operations case-study

All eruptions at Etna during the period 2000-2009 were tracked as part of the NEODAAS-HIGP collaboration, and the resulting data base for Etna is given in Harris et al. (2011). No effusive eruption occurred in Iceland over the time period of operations, although two major effusive crises occurred on Stromboli during 2002-2003 and 2007. During the 2002-2003 crisis we were invited, by

the Italian Civil Protection Department (DPC), to provide methodologies for timely delivery of TADR
(Harris et al., 2003). The site conditions, as already discussed, meant that implementation of groundbased methodologies to deliver TADR were difficult-to-impossible. As a result, by May 2003, after
five months of effusive activity only one or two field-based measurements of TADR were available.
DPC, however, recognized the value of the near-real time TADR metric in tracking the eruption, and
so were eager to develop a means of up-dating TADR time-series on at-least a daily basis.

439 Initially, a routine to obtain TADR from thermal camera data obtained during the routine helicopter 440 over-flights made at 09:00 (local time) each morning was developed (Harris et al., 2005). Results 441 were delivered, by 10:00 each morning, to DPC (Harris et al., 2003). In parallel with this, TADR were 442 derived from all available (AVHRR and MODIS) satellite resources and added to the daily report, so as 443 to allow an assessment of whether lava output was stable, increasing or decreasing. During 31 May 444 to 16 June 2003 these were given as power-point presentations each evening at 18:00, during the 445 daily DPC briefing held at the COA (Centro Operativo Avanzato) - the on-site operations center 446 (Bertolaso et al., 2009).

Thereafter, the same operation protocol was followed. That is, TADR were prepared on a daily-basis from satellite-flown sensors and then delivered in time for inclusion in daily reporting to DPC. The same protocol was followed during the 2007 crises and involved delivery of TADR reports to members of the Scientific Synthesis Group (SSG). This group was set up under the auspices of DPC (Barberi et al., 2009),

452 "to evaluate the scientific aspects of a volcanic emergency, recommend improvements of the
453 surveillance and provide advise and suggestions to the Civil Protection on the appropriate
454 urgent actions to be undertaken to mitigate risks."

455 The SSG appointed for the 2007 eruption of Stromboli included groups involved in operational 456 monitoring at Stromboli – mainly from INGV or the University of Florence (UNIFI). These groups 457 were also charged with scientific reporting before, during and after the effusive crisis (Bertolaso et 458 al., 2009). We thus passed information to INGV and UNIFI so as to ensure appropriate injection into 459 the response protocol as given in the flow chart of Bertolaso et al. (2009). Based in this experience, and other interactions with local volcano monitoring agencies around the Pacific Rim (Harris et al., 460 461 2002a), we developed the internal protocol charted in Figure 10 for information communication; this 462 preventing crossing of communication lines and protocols set-in place by the responding agencies.

463

464 Conclusion

NEODAAS-Plymouth has extensive expertise in processing satellite data for sea surface temperature and ocean color, as well as near-real time processing of Earth observation data for volcano monitoring. The primary focus is on processing of data from polar orbiting (medium spatial resolution) sensors (~1 km). Product provision focuses on the needs of the UK academic research community, NEODAAS being funded by NERC to support UK research efforts. However, support can be provided to non-UK applications through agreement.

471 In the case followed here, through memorandum of understanding, calibrated, geo-referenced 472 brightness temperature images covering Etna, Stromboli and Iceland were provided in near-real-time 473 for ingestion into the HIGP hot spots system. For the July-August 2001 eruption of Etna, as well as 474 Stromboli's 2007 eruption, event onset was alerted to by the automated detection algorithm linked 475 to a hot spot email notice [see Harris et al (2002a) for notice format]. Veracity of the detection was 476 checked through immediate communication with institutes formally responsible for monitoring Etna 477 and Stromboli who, in turned worked with Civil Protection and were mandated to provide 478 information during eruptive crises. Products were prepared and delivered on a daily basis by 18:00 479 (local – Italian – time). This meant that results when ready for use in daily reporting meetings with 480 Civil Protection, where Bonaccorso et al. (2005) described on-site responses and integration of 481 delivered TADR data into situation appraisals. In this way, dissemination of sensitive data was 482 controlled, allowing information to flow to the source charged with managing the crisis, thereby 483 allowing communication through a single, relevant voice. This followed the dissemination philosophy 484 given in Figure 10. The communication design was also set up so that the recipient, rather than 485 receiving raw data, received fully-processed data that was immediately useable. This was the guiding 486 philosophy of the system: to work with users to provide products, and develop a system, which fitted 487 the user needs, communication protocols and reporting requirements (McArdell, 2002).

489	Acronyms used	d (with reference or URL where appropriate)
490	ALICE:	Absolutely Local Index of Change of Environment (Pergola et al., 2008; 2009)
491	ATSR:	Along Track Scanning Radiometer
492	AQUA:	Part of NASA's A-train satellite series
493	AVHRR:	Advanced Very High Resolution Radiometer
494	AVO:	Alaska Volcano Observatory
495	BIRD:	Bispectral Infra-Red Detection
496	DN:	Digital Number
497	DPC	Italian Department of Civil Protection
498	ESA:	European Space Agency
499	EUMETSAT:	www.eumetsat.int
500	GOES:	Geostationary Operational Environmental Satellite
501	GMS:	Geostationary Meteorological Satellite
502	HIGP:	Hawaii Institute of Geophysics and Planetology
503	INGV:	Istituto Nazionale di Geofisica e Vulcanologia
504	IODC:	Indian Ocean Data Coverage (Meteosat-7)
505	IR:	InfraRed: 0.7 – 20 μm
506	METOP:	component of the overall EUMETSAT Polar System
507	MIR:	Mid-InfraRed: 3.0 – 5.0 μm
508	MODIS:	Moderate-Resolution Imaging Spectroradiometer
509	MODTRAN:	MODerate resolution atmospheric TRANSmission
510	MODVOLC:	HIGP MODIS Thermal Alert System (http://modis.higp.hawaii.edu/)
511	MSG:	Meteosat Second Generation
512	MTSAT:	Multifunctional Transport Satellite
513	NASA:	National Aeronautics and Space Administration
514	NEODAAS:	Earth Observation Data Acquisition and Analysis Service
515	NERC:	Natural Environment Research Council
516	NESDIS:	National Environmental Satellite, Data, & Information Service
517	NIR:	Near InfraRed: 0.7 – 1.1 μm
518	NOAA:	National Oceanic and Atmospheric Administration
519	NTI:	Normalized Thermal Index (Wright et al., 2002)
520	RST:	Robust Satellite Technique (Pergola et al., 2008; 2009)
521	SEVIRI:	Spinning Enhanced Visible and InfraRed Imager
522	SSG:	Scientific Synthesis Group (Betaloso et al., 2009)
523	UNIFI:	University of Florence
524	TADR:	Time-Averaged Discharge Rate (Harris et al., 2007)
525	TERRA:	NASA's "flagship Earth observing satellite" (http://terra.nasa.gov/)

- 526 TIR: Longwave InfraRed: $5.0 20 \,\mu\text{m}$
- 527 UAF: University of Alaska, Fairbanks
- 528 UK: United Kingdom
- 529 VAST: Volcanic Anomaly SofTware (Higgins and Harris, 1997)
- 530 VISSR: Visible and Infrared Spin Scan Radiometer
- 531 USA (or US): United States of America
- 532

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760 Figure Captions

Figure 1. (a) coverage of the NEODAAS-Dundee station, and (b) frequency of pass by geographicalposition within the mask.

Figure 2. NEODAAS-Plymouth hot spot browser for Etna and Stromboli. The example is from 10
March 2013, and hot spots are apparent at Mt. Etna's summit due to effusive and fountaining activity
at the SE Crater. In this single-band black-and white rendition, black is cold and white is hot.

766 Figure 3. (a) AVHRR channel 4 sub-image of Sicily and Calabria obtained at 00:46Z on 29 May 2001. 767 Sub-image is 320 × 300 pixels, or ~350 × 330 km, in size. Lighter tones indicate higher pixel-768 integrated spectral radiances. The Aeolian islands of Alicudi, Filicudi, Salina, Lipari, Vulcano, Panarea 769 and Stromboli are labeled using the first letter of each islands name. The oil refinery at, and cape of, 770 Milazzo are located using the red circle (the oil refinery registers a hot spot in the AVHRR MIR and TIR 771 bands). Yellow circle contains an "apparent" hot spot due to the presence of a lake (Biviere di 772 Lentini) which appears relatively warm by night (against the cool land background) and relatively cool 773 by day (against the warm, solar-heated, land background). Note that the real hot spot on Etna is 774 somewhat crisper (more cleanly defined) than the fuzzier "apparent" hot spot of the lake, and two 775 other lake-related "apparent" hot spots can be seen to the SW of Etna. Mount Etna is located using 776 the red box, and is magnified top right. In this nighttime image, the sea (like the lake) is relatively 777 warm (lighter tones) compared with the land (darker tones), and Etna is apparent as a cold, circular, 778 feature (due to its elevation, surface temperatures decrease with height causing the volcano to 779 appear as a cold zone). The hot spot at Etna's summit is centered in the yellow box and magnified 780 lower right. The hot spot is due to pixels containing active lava, and is obvious as a group of hot 781 (white) pixels against a cold (black) background. (b) AVHRR channel 4 sub-image of Sicily and 782 Calabria obtained at 00:36Z on 30 May 2001 (from Electronic Supplement 7 of Harris, 2013).

Figure 4. Schematic summarizing implementation of the VAST hot spot detection algorithm (from
Higgins and Harris (1997, Fig. 3): with permission from Elsevier). The algorithm used three equations
a executes five steps.

Figure 5. Pixel grids of AVHRR channel 4 brightness temperature centered on Mt. Etna's summit hot spot for (a) the 29 May, and (b) the 30 May 2001 sub-images of Figure 3. Values are brightness temperature plus 30 °C multiplied by 100. Area of the anomaly is marked by the red line, with a potential extra "anomalous" pixel in the 30 May being highlighted in yellow. Inclusion of this pixel increases the TADR estimate from $1.7 - 2.1 \text{ m}^3 \text{ s}^{-1}$ to $2.0 - 2.3 \text{ m}^3 \text{ s}^{-1}$ (from Electronic Supplement 7 of Harris, 2013).

Figure 6. Pixel grids of AVHRR channel 4 brightness temperature centered on Mt. Etna's summit hot spot for (a) the 29 May, and (b) the 30 May 2001 sub-images of Figure 3. Values defined within the hot spot are given in red, and "cold" background pixels are given in blue; these are linked to the "hot" pixels for which they are used to characterize the background spectral radiance in the mixture model using the blue, arrowed, line (from Electronic Supplement 7 of Harris, 2013).

Figure 7. AVHRR quick look products for (a) the Mt. Etna tile and (b) the Iceland tile; with (c) AVHRR quick look during eruptive activity at Mt Etna in 2001 (AVHRR channel 4 brightness temperature, 22 July 2001 15:48 GMT) and (d) MODIS quick look during eruptive activity during 2002 (MODIS top-ofatmosphere true-colour, 28 Oct. 2002 12:15 GMT).

Figure 8. TADR checks completed during Etna's 2001 eruption. Field observations were provided by
Sonia Calvari (INGV-CT) based on channel dimensions and lava flow velocity upon derivation, field
mapping estimates were based on the change in volume of the flow field over 24 hour increments.
Lines simply link data points and may not be representative of the actual trend between each linkedpoint.

Figure 9. AVHRR channel 4 image of 24 July 2001 (00:58 GMT) showing two hot spots cut by a (cold) plume due to phreatomagmatic activity (Taddeucci et al., 2002; 2004) at the Piano del Lago cone at 2500 m (dark is cold; bright is hot). The northern hot spot is due to lava flows from fissures in the summit zone; southern hot spot is due the active lava flow (LSF1) advancing towards Nicolosi and Belpasso. (a) Georeferenced channel 4 image, and (b) fitted to a town and road grid. In addition, the best images were fitted to a DEM of Etna (c). Fit given in (c) is the image for 25 July 2001 at 00:54 GMT.

Figure 10. Response model and flow of data/information through the hot spot response system implemented at HIGP. Example agencies are given on the basis of the Etna 2001 and 2002, as well as the Stromboli 2002-2003 and 2007, experiences. A qualitative assessment of the time delay for data or information provision is indicated at each step. Dashed line is a communication "wall", where we link into the base of the local communication protocol, where the local communication protocol linked to here is a crude summary of that detailed by Bertolaso et al. (2009) for Stromboli's 2007 effusive crisis.

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(A) ALGORITHM STEPS

Step 1

Create ΔT image (= pixel area within the black box).

Step 2

Define volcano sub-image (= all pixels within the grey box).

Step 3

Apply equations A & B to all pixels in the nonvolcanic zone (= all pixels within the black box but outside the grey box) on a pixel by pixel basis. Store the maximum ΔT_{diff} . Natural variation (ΔT_{nat}) is set equal to maximum ΔT_{diff}

Step 4

Apply equations A, B & C to all pixels in the volcano sub-image on a pixel by pixel basis.

Step 5

Iterate step 4, excluding pixels flagged as "hot" in a previous iteration from equation A, until no new "hot" pixels are found.



 $n_n =$ number of pixels in the non-volcanic zone $n_v =$ number of pixels in the volcanic sub-image

(C) EQUATIONS

Equation A

$$\Delta T_{neigh} = \frac{\sum_{i=1}^{8} \Delta T_i}{8}$$

in which ΔT_{i} = ΔT for neighbourhood pixel i, where i = 1 to 8

Equation B

 $\Delta T_{\text{diff}} = \Delta T_x - \Delta T_{\text{neigh}}$ in which $\Delta T_x = \Delta T$ for pixel x

Equation C

If $\Delta T_{diff} > \Delta T_{nat}$ then pixel is "hot" else pixel is "cold"

KEY

Current target pixel in the non-volcanic zone with temperature ΔT_i , where i = 1 to n_n

Non-volcanic zone pixel with maximum ΔT_{diff} (= ΔT_{nat})

Current target pixel in the volcano sub-image with temperature ΔT_i , where i = 1 to n_v

(a) 29 May 2	001,0046Z
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30	27	23	24	28
85	45	40	38	64
30	33	46	43	32
05	50	82	81	40
32	41	60	53	34
24	80	31	73	32
29	31	35	33	30
95	52	29	14	26

(b) 30 May 2001,0036Z

32	33	32	31	32
77	20	14	74	40
40	49	49	42	36
64	22	36	06	27
40	47	47	41	36
56	62	90	41	67
29	28	29	32	34
52	09	24	79	87

(a) 29 May 2001,0046Z

30	27	23	24	28
85	45	40 N	38	64
30	33	46*	43	32
05	50	82	81	40
32	41	60	53	34
24	80	31	73 🛌	32
29	31	35	33	30
92	52	29	14	26

(b) 30 May 2001,0036Z









(c)





Study	Details	Reference
Dozier (1980)	Specifications of algorithms to estimate the size and temperature of sub-pixel hot spots using two bands of infrared (AVHRR) data (i.e., definition of the "dual- band method"). Atmospheric correction methods also considered.	NOAA Technical Memorandum, NOAA- 81021710, Washington, DC
Dozier (1981)	Ditto	<i>Remote Sensing of Environment</i> , 11 , 221-229.
Matson & Dozier (1981)	The Dozier (1980; 1981) algorithm applied to subpixel hot spots associated with oil flares in the Persian Gulf. Industrial hot spots around Detroit identified.	Photogrammetric Engineering & Remote Sensing, 47 , 1311-1318.
Wan (1985)	Simulation of smoke interference with fire signal reception using multiple scattering radiative transfer model linked to model of AVHRR response.	<i>PhD Dissertation</i> , University of Santa Barbara (CA).
Matson et al. (1984)	Case study of LAC fire imagery described for various sites; fire sightings in western U.S. compared to hot spots appearing in nighttime (2 am) HRPT images	<i>NOAA Technical Report</i> , NESDIS 7 , Washington, DC.
Muirhead & Cracknell (1984)	Rectification accuracy tested by comparing hot spot locations on rectified LAC (MIR) images containing gas flare locations of known location and associated with North Sea drilling rigs.	International Journal of Remote Sensing, 5 , 199- 212.
Muirhead & Cracknell (1985)	Hot spots counted on three rectified LAC MIR images of U.K. to assess straw burning and extent of compliance with bans on burning on certain days.	International Journal of Remote Sensing, 6 , 827- 833.
Malingreau et al. (1985)	Hot spot chronology and NDVI studied of Borneo and East Kalimantan during immense fires of 1983.	Ambio, 14, 314-315.
Malingreau (1984)	Ditto	8 th International Symposium on Remote Sensing of Environment, held in Paris (France), 1-4 October 1984. Ann Arbor: Environmental Research Institute of Michigan.
Flannigan (1985)	Fire reports from severe fire outbreak in Alberta compared to fires detected by AVHRR. Dual-band algorithm used to estimate fire size and temperature. Cloud screening applied to reject cloud contaminated pixels	<i>MSc. Thesis</i> , Colorado State University, Fort Collins (Colorado).
Flannigan & Vonder Haar (1986)	Ditto	Canadian Journal of Forest Research, 16 , 975- 982

Table 1. Studies, and brief details, of AVHRR studies of fire as reviewed by Robinson (1991) [modified from Table 4 of Robinson (1991)].

Matson & Holben (1987)	Hot spots and vegetation studied on one LAC image for a $3 \times 6^{\circ}$ box over Manaus, Brazil. Dual-band algorithm applied.	International Journal of Remote Sensing, 8 , 509- 516.			
Malingreau & Tucker (1987)	Fire points in Southern Amazon Basin studied on a daily basis over two years in conjunction with studies of NDVI. Inference drawn about penetration of settlement into remote areas.	Proceedings of IGARSS '87 held in Ann Arbor, Michigan, 18-21 May 1987. IEEE 87CH2434-9 (New York: IEEE), pp. 484-489.			
Pereira (1988)	Fire counts and analysis of smoke trajectories with estimates of areas burned and mass combusted based on Brazilian HRPT data of Amazonia. Landsat TM compared to AVHRR.	INPE-4503-tdl/325, Inst. Nactional de Pesquisas Espacias, 12.201 Sao Jose dos Campos, SP, Brazil.			
Setzer et al. (1988)	Ditto	INPE-4534-RPE/565, Inst. Nactional de Pesquisas Espacias, 12.201 Sao Jose dos Campos, SP, Brazil.			
 Acronyms: HRPT: High Resolution Picture Transmission (direct read-out of AVHRR data to ground stations). GAC: Global Area Coverage. LAC: Local Area Coverage. NDVI: Normalized Difference Vegetation Index [see Cracknell (1997) for full definition of each] 					

Table 2 Automated fire detection algorithms published in the peer-reviewed literature between 1985 and 1996. Algorithms are listed in chronological order of publication [modified from Harris (2014)]. For more background on algorithm heritage, detail on test set up and execution for each case, see Electronic Supplement 8 of Harris (2013)[§].

Study	Algorithm Type	Tests Executed	Data Type (Application Region)
Flannigan & Vonder Haar (1986)	Contextual	ESTIMATE the mean T_{MIR} and T_{TIR} for cloud-free background pixels from the eight pixels in a 9 \times 9 pixel box centered on the target pixel, then: $T_{MIR-t} > mean T_{MIR-b}$ $T_{TIR-t} > mean T_{TIR-b}$ $\Delta T > 8 K (nighttime);$ $\Delta T > 10 K (daytime)$	AVHRR (Forest fires – Canada)
Kaufman et al. (1990) ⁽¹⁾	Fixed (Generic)	$\begin{split} T_{\text{MIR}} &= 316 \text{ K} \\ \Delta T &> 10 \text{ K} \\ T_{\text{TIR}} &> 250 \text{ K} \qquad \text{(cloud test)} \end{split}$	AVHRR (Fires – Brazil)
Lee and Tag (1990)	Contextual	$T_{TIR} < 263 \text{ K} (cloud test)$ Then; ESTIMATE T_{TIR-b} using the mean of the four side pixels in a 9 × 9 pixel box centered on the target pixel; SELECT a threshold hot component (fire) temperature (T_{fire}) for a two component mixture model and use this, with T_b , to estimate the size of the fire required to yield T_{TIR-i} ; USE T_{fire} , T_{TIR-b} and fire size to estimate the corresponding pixel- integrated temperature in the MIR (T_{thresh}) $T_{MIR} < T_{thresh}$	AVHRR (Wild fires – Yellowstone) (Gas flares – Persian Gulf) (Structure fires – California)
Setzer and Pereira (1991) ⁽²⁾	Fixed (Generic)	T _{MIR} > 319 K PLUS: manual detection of smoke	AVHRR (Fires – Brazil)
Brustet et al. (1991)	Fixed (Specific)	MIR and TIR thresholds set manually on a case-by-case basis using frequency distributions and T_{TIR} versus T_{MIR} scatter plots.	AVHRR (Wild fires – West Africa)
Kennedy et al. (1994)	Fixed (Generic)	$\begin{array}{l} T_{MIR} > 320 \ K \\ \Delta T > 15 \ K \\ T_{TIR} > 250 \ K \\ and/or \ R_{NIR} < 16 \ \% \end{array} \ (cloud \ test) \end{array}$	AVHRR (Wild fires – West Africa)
Langaas (1993)	Contextual	Create frequency distribution of DN_{MIR} ; DN_{MIR} with frequency of 50 = DN_{thresh} IF DN < DN_{thresh} AND $T_{fire}^* > 470$ K THEN	AVHRR (Wild fires – West Africa)

Chuvieco and Martin (1994)	Fixed (Generic)	$T_{MIR} > 317$ K (day) $T_{TIR} > 295$ K (night) (applied only within forest mask to reject false detections due to solar heated soil, which could approach saturation).	AVHRR (Forest fires - Spain)
Justice and Dowty (1994)	Contextual (3)	$\begin{array}{l} T_{MIR-t} > 316 \ K \\ T_{TIR-t} > 290 \ K \\ T_{TIR-t} < T_{MIR-t} \\ NOW, \ Target \ pixel \ \Delta T > \Delta T \\ mean \ from \ the \ background, \ plus \\ two \ times \ the \ standard \ deviation \\ of \ the \ \Delta T \ for \ the \ background \\ pixels \ \dots \ \dots \ or \ 3 \ K \ (whichever \ is \ greater). \end{array}$	Algorithm developed at NASA/Goddard Space Flight Centre
Prins and Menzel (1994)	Contextual	Estimate the mean and standard deviation in T_{MIR} and T_{TIR} for all cloud-free pixels across a 150 km \times 150 km sector. This defines the background values for each band (T_{MIR-b} and T_{TIR-b}); NOW: $\Delta T_t > \text{mean } \Delta T_b$ $T_{MIR-t} - T_{MIR-B} > 1.5 \sigma(T_{MIR-B})$ $T_{MIR-t} > 300 \text{ K and } T_{TIR-t} > 295 \text{ K}$ $T_{TIR-t} - T_{TIR-b} > 1 \text{ K}$ $T_{MIR-t} - T_{MIR-b} > 5 \text{ K}$ $T_{fire}^{\text{irre}} > 400 \text{ K}$	GOES-VAS (Burning – S. America)
Arino and Melinotte (1995)	Fixed (Generic)	$\begin{array}{ll} T_{MIR} > 320 \ K & (saturation \ test); \\ T_{MIR} > T_{TIR} + 15; \\ T_{TIR} > 245 \ K & (cloud \ test); \\ R_{VIS} < 25 \ \% & (reflection \ test); \\ R_{VIS} - R_{NIR} > 1 \ \% \ (sunglint \ test). \end{array}$	AVHRR (Fires – Africa)
Franca et al. (1995)	Fixed (Generic)	$\begin{array}{l} T_{MIR} > 320 \ K \\ \Delta T > 15 \ K \\ T_{TIR} > 287 \ K \\ 0 \leq T_{10 \mu m} \ \text{-} \ T_{12 \mu m} \geq 5 \ K \\ R_{VIS} < 9 \ \% \end{array}$	AVHRR (Wild fires – West Africa)
Flasse and Ceccato (1996)	Contextual (3)	$\begin{split} T_{MIR} &> 311 \text{ K} \\ \Delta T &> 8 \text{ K} \\ T_{MIR} - [\text{mean } T_{MIR\text{-}b} - 2\sigma] &> 3 \text{ K} \\ \Delta T &> [\text{mean } \Delta T_b - 2\sigma] \end{split}$	

pixel is anomalous

[§] http://www.cambridge.org/us/academic/subjects/earth-and-environmental-science/remote-sensing-and-gis/thermal-remote-sensing-active-volcanoes-users-manual

(1) The algorithm was also applied for fire detection in AVHRR data for West Africa by Kennedy et al. (1994).

(2) Another algorithms was published with a similar basis, but using DN criteria (fire if DN < 10 or 8, i.e., if

DN are close to AVHRR saturation), by Pereira and Setzer (1993).

(3) Algorithm uses 9×9 pixel box centered on background cleaned of potential fire pixels; box expanded up to limit of 21×21 pixels until at least 25 % of pixels are non-fire.

*estimated using the dual-band method of Dozier (1981).

 T_{MIR} = Mid-infrared pixel-integrated temperature; T_{TIR} = Thermal-infrared pixel-integrated temperature;

 $T_{10\mu m}$ = Pixel-integrated temperature at 10 μ m; $T_{12\mu m}$ = Pixel-integrated temperature at 12 μ m;

 T_{MIR-t} = Target pixel MIR brightness temperature; T_{MIR-b} = Background pixel MIR brightness temperature; T_{TIR-t} = Target pixel TIR brightness temperature; T_{TIR-b} = Background pixel TIR brightness temperature;

 $\Delta T = T_{MIR} - T_{TIR}; T_{fire} = \text{sub-pixel fire temperature}; T_{thresh} = \text{threshold temperature}; \sigma = \text{standard deviation} \\ R_{NIR} = \text{Near-infrared reflection}; R_{VIS} = \text{Visible reflection}; DN = \text{Digital Number (subscripts as per temperatures)}$

Table 3. Data archiving at NEODAAS-Dundee began in 1978 with the launch of TIROS-N. Today the archive includes data from 10 sensors flown on polar orbiter and geostationary platforms. While a collation of all AVHRR archive data spanning effusive eruptions at Etna are given in Harris et al. (2011); those for Krafla (Iceland) are given in Harris et al. (2000a).

Satellite - sensor	Archive temporal coverage	Source
NPP VIIRS	July 2012 - present	NEODAAS-Dundee
Aqua/Terra MODIS	Apr. 2000 - present	NEODAAS-Dundee & NASA
NOAA/METOP AVHRR	Nov. 1978 - present	NEODAAS-Dundee
Nimbus-7 CZCS	Aug. 1979 - Jun 1986	NEODAAS-Dundee
OrbView-2 SeaWiFS	Sept. 1997 - Dec 2010	NEODAAS-Dundee
Envisat MERIS	2004 - 2012	ESA
Geostationary satellites ¹	2001-present	NEODAAS-Dundee, ESA & NOAA

¹SEVIRI, VISSR, GOES and MTSAT

Table 4. Locations of, and average transmissivities, $\tau(\lambda)$, across, the seven main atmospheric windows in the NIR, MIR and TIR (adapted from Harris, 2013). Values obtained using MODTRAN applied using a 1976 US Standard atmosphere with a vertical path from sea-level to space (zenith = 180 °, observer height = 100 km), a CO₂ mixing ratio of 380 ppm·v. In the final column is the range of temperatures (T_{peak}) which have their peak of spectral exitance (λ_m) in the given waveband following re-arrangement of Wien's displacement law, i.e., $T_{peak} = 2898 \ \mu m \ K / \lambda_m$.

Window	Waveband	Width	Average	Max	Location of	T _{peak}
Location	(µm)	(µm)	τ(λ)	$\tau(\lambda)$	Max $\tau(\lambda)$	(°C)
					(µm)	
NIR	0.7 to 0.89	0.19	0.90	0.93	0.89	3900 to 3000
NIR	1.0 to 1.1	0.1	0.94	0.95	1.07	2600 to 2300
SWIR	1.18 to 1.31	0.13	0.94	0.96	1.25	2200 to 1900
SWIR	1.51 to 1.76	0.25	0.96	0.97	1.68	1650 to 1400
SWIR	2.03 to 2.36	0.33	0.96	0.98	2.14	1150 to 950
MIR	3.44 to 4.13	0.69	0.94	0.97	3.96	570 to 430
TIR	8.6 to 12.2	3.6	0.92	0.96	10.11	64 to -35

Table 5. Tests and thresholds used by Harris et al. (2001a) to estimate of hot spot probability. Values output by the algorithm are scaled to 0 to 1, where negative values are mapped to the range 0-0.5 (i.e. 0-50 % probability) and positive values are mapped to the range 0.51-1 (i.e. 51-100 % probability).

Probability step (label)	Test	Description
1 (Prob 1)	Brightness	Pixel albedo (A1) is compared with an albedo threshold (τ 1) to determine the probability that the pixel is cloud-covered: Bright pixels (i.e., A1 > τ 1) are given a positive weighting dependent on the magnitude of the A1 - τ 1 difference. Non-bright values (i.e., A1 < τ 1) are given a negative weighting dependent on the A1 - τ 1 difference.
2 (Prob 2)	Thermal difference	Pixel ΔT radiance is compared with its background (as characterized by the mean of neighboring pixels in a 25 × 25 pixel box centered on the target pixel = $\Delta TBck$). This provides a measure of the thermal difference between a pixel and its immediate background. If $\Delta T > \Delta TBck$, then the pixel is given a positive weighting dependent on the magnitude of the ΔT - $\Delta TBck$. Difference. If $\Delta T < \Delta TBck$ then the pixel is given a negative weighting dependent on the magnitude of the ΔT 1 - $\Delta TBck$ difference.
3 (Prob 2-4)	Thermal anomaly	Pixel ΔT is compared with a ΔT threshold of 10 °C (ΔT thresh). This gives a measure of thermally anomalous activity. If $\Delta T > \Delta T$ thresh, then the pixel is given a positive weighting dependent on the magnitude of the ΔT - ΔT thresh difference. If $\Delta T < \Delta T$ thresh, then the pixel is given a negative weighting dependent on the magnitude of the ΔT - ΔT thresh difference.
4 (Prob 4)	Cold test	Pixel TIR brightness temperature (TIR) is compared with a TIR temperature threshold (τ TIR), which corresponds to a pixel brightness temperature of -23°C. This gives a measure of how cold the pixel is, and thereby the probability that cold cloud is present. If TIR > τ TIR, then the pixel is given a positive weighting dependent on the magnitude of the TIR – τ TIR difference. If TIR < TIR, then the pixel is given a negative weighting dependent on the magnitude of TIR – τ TIR difference.
Total probability (PROB)	PROB = Prob 1 × Prob 2 × Prob 3 × Prob 4	Total probability is the product of the 4 sub-probabilities (Prob 1 to 4) and is therefore designed to provide a quantitative assessment of whether a pixel (1) contains cloud, (2) is thermally different from its background, and (3) is thermally anomalous.

Date	Time	TADR	TADR	Notes
	(GMT)	Max $(m^3 s^{-1})$	$Min (m^3 s^{-1})$	
16 July 2001	00:47	2 pixels @ SEC		
16 July 2001				Scan Edge
16 July 2001				Cloud
10 July 2001	00.27	 16 minula @ SEC		Cioud
17 July 2001	00:37	to pixels @ SEC		Scop Edge
17 July 2001	12.13	8 nivels @ SEC		Wrap round
17 July 2001	16:07	0 pixels @ 5LC		Cloud
18 July 2001	02:08			Cloud
18 July 2001	12:10	16.1	8.8	1st w/2100m hot spot
19 July 2001	01:50	10.1	4.4	
19 July 2001	11:59	21.8	11.9	
19 July 2001	13:40			Scan Edge
21 July 2001	01:29			Wrap Round
21 July 2001	11:39			Scan Edge
21 July 2001	13:19	27.7	13.7	
21 July 2001	16:18	28.1	12.2	
22 July 2001	01:19	30.0	12.0	
22 July 2001	11:28			Scan Edge
22 July 2001	13:09	19.3	9.1	
22 July 2001	15:55	22.0	9.2	
23 July 2001	01:09	31.8	12.6	
23 July 2001	12:58			Wrap Round
23 July 2001	15:32	20.7	8.9	
24 July 2001	00:58	24.9	10.0	
24 July 2001	12.48	0.0	2.0	Wran Round
24 July 2001	00:47	23.3	9.6	wiap Round
25 July 2001	12.37	23.5	9.0	Cloud
25 July 2001	16:25			Cloud
26 July 2001	00:37			Cloud
26 July 2001	02:17			Scan Edge
26 July 2001	12:27			Cloud
27 July 2001	00:27	10.8	4.5	
27 July 2001	02:07			Scan Edge
27 July 2001	12:16			Cloud
28 July 2001	01:56	12.2	5.1	
28 July 2001	12:06			Cloud
29 July 2001	01:46	14.2	5.9	
29 July 2001	11:55			Cloud
29 July 2001	13:36			Cloud
29 July 2001	10:32			Cloud
30 July 2001	01:30	19.1	8.0	Cloud
30 July 2001	13.26	26.2	13.4	Cloud
30 July 2001	16:09	26.2	12.5	
31 July 2001	01:25	20.0	8.3	
31 July 2001	11:34			Cloud
31 July 2001	13:20			Cloud
31 July 2001	15:46			Cloud
1 August 2001	01:15			Plume
1 August 2001	05:19			Plume & Scan Edge
1 August 2001	11:24			Scan Edge
1 August 2001	13:05			Plume
2 August 2001	01:04			Plume
2 August 2001	12:54			Plume
2 August 2001	16:39			Plume
3 August 2001	00:54	3.1	7.7	~
3 August 2001	12:44			Cloud
3 August 2001	10:10			Cloud
4 August 2001	00:43			Soon Eda-
+ August 2001	02.24			Scan Euge

Table 6. AVHRR-derived TADR log built during Etna's 2001 flank eruption.

4 August 2001	12:33			Plume
5 August 2001	00:33	2.5	5.8	
5 August 2001	02:13			Scan Edge
5 August 2001	05:26			Scan Edge
5 August 2001	12:23	6.4	13.6	
6 August 2001	00:23	1.8	4.1	
6 August 2001	02:03			Scan Edge
6 August 2001	12:12			Cloud
6 August 2001	16:46			Scan Edge
7 August 2001	01:52	1.0	2.4	
7 August 2001	12:02	3.7	8.7	
7 August 2001	13:42			Scan Edge
7 August 2001	16:23			Cloud
8 August 2001	00:43			Minor plume to SE
8 August 2001	11:51			Cloud: no plume
8 August 2001	13:32			Cloud: no plume
8 August 2001	16:00			Cloud: no plume
9 August 2001	01:31	1.2	2.8	
9 August 2001	11:14	0.3	0.6	
9 August 2001	13:21			Cloud
9 August 2001	15:36			Cloud

Date	Ground-based observations (INGV-Catania)	AVHRR-based observations
July 1-4	No glow observed (last paroxysm from SEC was	1-3 pixel, low magnitude hot spots @ SEC
	~23:00 (6/29) - ~21:00 (6/30)	
July 4	11:00: lava effusion begins @ SEC	8 pixel, high magnitude hot spot @ SEC
	22:30: Increase in Strombolian activity	
	00:45: Activity declining (07/5)	
July 7-17	Continuous lava flow effusion from SEC	AVHRR hot spot steadily increases in size and
		magnitude:
		7/07 = 2 pixels
		7/11 = 3 pixels
		7/15 = 6 pixels
July 9	Episode of mild strombolian activity @ SEC (early)	AVHRR = cloud covered
July 13	Paroxysm @ SEC (early am)	01:19 GMT image shows 3 pixel, high magnitude hot spot at SEC
July 17	Paroxysms @ SEC in early am, in the following hours fissures open at the S. base of SEC and extend S. These feed strombolian & lava flow activity	00:37 GMT image shows 16 pixel, high magnitude anomaly @ SEC with plume extending ENE. 12:13 GMT image shows high magnitude hot
		spot between SEC and Montagnola
July 17	Second fissure becomes active on N flank of Montagnola. Flows begin to extend away from eruptive fissure. Over the following days these flow extended S to threaten the ski area and Sapienza complex	
July 18	~02:00 (local time) 2100 m vent begins activity. Over following days flows from this vent extend S towards Nicolosi	12:03 GMT image shows first dual hot spot, with hot spot #1 between SEC and Montagnola, & #2 below Montagnola (2100 m vent)
July 18	Main vent on N flank of Montagnola opens up: becomes source of major ash emissions	
July 20	Flow active from NE extending fissure segment, flows extend into Valle del Leone	Activity apparent as third hot spot to NE of SEC especially on July 22-23 images

Table 7. Activity time-line and AVHRR observations for Etna during July 1 – 18, 2001.

Table 8. Email notices sent out between 19 March and 9 April during Stromboli's 2007 eruption.

To: Distribution list Re: Stromboli: 19-28 March 2007

Complete or partial cloud cover meant that no AVHRR-based TADR estimates were possible during 18-25 March (although the anomaly was observed through or between clouds on 10 occasions).

Since 26 March cloud conditions have improved allowing TADR estimates from 5 of the 14 passes over the last 3 days. These give the following TADRs:

26 March 07 – 09:54 UT – 0.9 – 1.4 m³/s 26 March 07 – 12:21 UT – 0.7 – 1.6 m³/s 27 March 07 – 21:00 UT – 0.7 – 1.7 m³/s 28 March 07 – 12:01 UT – 0.7 – 1.6 m³/s 28 March 07 – 20:37 UT – 0.5 – 1.4 m³/s

To: Distribution list Re: Stromboli: 28 March – 3 April 2007

Complete or partial cloud cover meant that no AVHRR-based TADR estimates were possible during 29 & 30 March.

Since the afternoon of 31 March cloud conditions have improved allowing TADR estimates from 4 passes spanning 31 March – 2 April. These give the following TADRs:

31 March 07 – 21:08 UT – 0.5 – 1.3 m³/s 01 April 07 – 01:24 UT – 0.4 – 0.9 m³/s 01 April 07 – 20:45 UT – 0.9 – 2.2 m³/s 02 April 07 – 12:50 UT – 0.6 – 1.3 m3/s

As of 3 April, cloud conditions had deteriorated again such that all 5 of today's images were cloud covered.

To: Distribution list Re: Stromboli: 3-9 April 2007

Although we've looked at data from 34 passes during 3-9 April, cloud (sometimes localized but just sitting right over Stromboli) has meant that we have had no data suitable for TADR calculation since 2 April.

We did see a band 3 hot spot on 5 occasions in this period, but they all appeared cloud contaminated and there was no convincing band 4 anomaly.