Machine Learning approach for multi-perspective volcanic eruption recognition using thermal infrared images

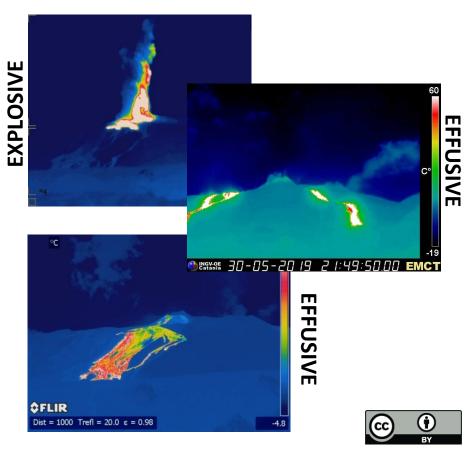


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Goal: Detection and automatic classification of the eruptive activity typology using thermal images

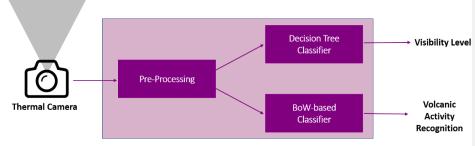
- Detecting, locating and characterizing eruptions in real-time is fundamental to monitor volcanic activity.
- Here, we present an automatic system able to discover and identify the main types of eruptive activities by exploiting infrared images acquired by the thermal cameras installed around Mount Etna volcano.
- The proposed system applied to each image of all thermal cameras over Etna in real-time, provides two outputs, namely the visibility level and the recognized activity status.
- By merging the outcomes coming from each thermal camera, the monitored phenomena can be fully described from different perspectives getting deeper information in real-time and in an automatic way.



DATA

EBT

METHOD



к 07- 12-20 19 2 I: 15:30.00 ENT

EMOT

20 19 00:00:00 00 EMO

ESR

Workflow

EMCI



Fig. 1: Location of thermal cameras on Etna. Google Earth view of summit area showing each thermal camera position and the corresponding sample image, namely ENT (yellow box), EMOT (green box), EBT(blue box), ESR (pink box) and EMCT (orange box). Table 1: Thermal cameras main features.

Thermal camera	Location	Distance from summit	Elevation s.l.m.	Field of Vlew
EMOT	La Montagnola	3 km S	2600 m	18.8° × 25°
ESR	Schiena dell'Asino	~5 km SSE	1985 m	18.8° × 25°
EMCT	Monte Cagliato	8.5 km ESE	1160 m	18.8° × 25°
EBT	Bronte	~12 km WNW	971 m	18° × 24°
ENT	Nicolosi	15 km S	730 m	18° × 24°

Fig.2: Scheme of the machine learning approach to classify volcanic eruptions using thermal images. The images acquired from each thermal camera located on Etna (Fig.1) are used as input of the proposed ML approach. After a pre-processing step consisting in manipulating the raw false-color images, a *Decision Tree* (*DT*) classifier is adopted to recognize the visibility conditions of the monitored scene and a *Bag of Words* (*BoW*)-*based classifier* is used to detect the onset of the eruptive activity and recognize the eruption type among explosive, effusive, explosive and effusive and degassing classes.

Classifier

Decision Tree





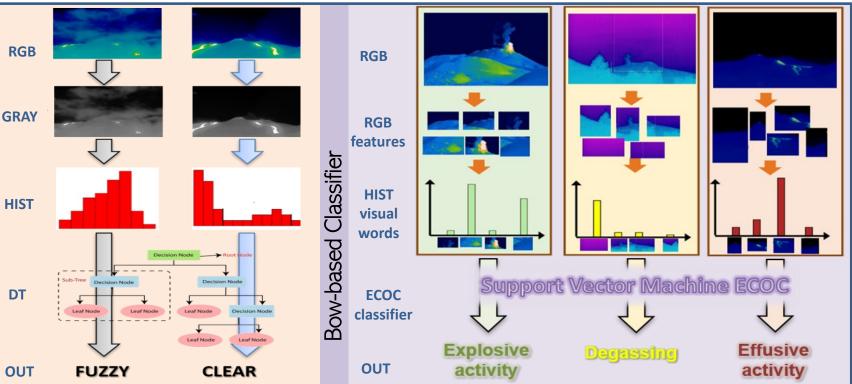


Fig. 3: Decision Tree classifier. The RGB images are converted in GRAY images and their intensity histograms are set as attributes of the DT. By opportunely setting thresholds intensity levels (i.e. DT nodes' criteria) during the training phase, DT is able to distinguish between "clear" and "fuzzy" classes of the DT. By opportunely setting thresholds intensity levels (i.e. DT nodes' criteria) during the training phase, DT is able to distinguish between "clear" and "fuzzy" classes of the DT. By opportunely setting thresholds intensity levels (i.e. DT nodes' criteria) during the training phase, DT is able to distinguish between "clear" and "fuzzy" classes of the DT. By opportunely setting thresholds intensity levels (i.e. DT nodes' criteria) during the training phase, DT is able to distinguish between "clear" and "fuzzy" classes of the DT. By opportunely setting thresholds intensity levels (i.e. DT nodes' criteria) during the training phase, DT is able to distinguish between "clear" and "fuzzy" classes of the DT. By opportunely setting thresholds intensity levels (i.e. DT nodes' criteria) during the training phase, DT is able to distinguish between "clear" and "fuzzy" classes of the DT. By opportunely setting thresholds intensity levels (i.e. DT nodes' criteria) during the training phase, DT is able to distinguish between "clear" and "fuzzy" classes of the DT. By opportunely setting thresholds intensity levels (i.e. DT nodes' criteria) during the training phase, DT is able to distinguish between "clear" and "fuzzy" classes of the DT. By opportunely setting the training phase, DT is a distinguish between "clear" and "fuzzy" classes of the DT. By opportunely setting the training phase, DT is a distinguish between "clear" and "fuzzy" classes of the DT. By opportunely setting the training phase, DT is a distinguish between "clear" and "fuzzy" classes of the DT. By opportunely setting the training phase, DT is a distinguish between "clear" and "fuzzy" classes of the DT. By opportunely the training phase, DT is a disti



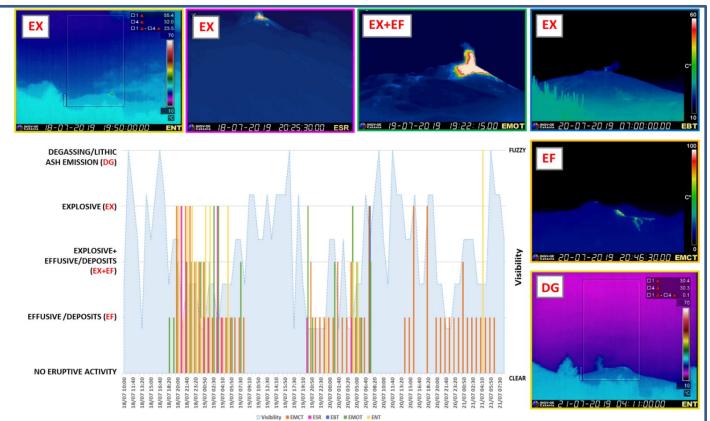




Fig.5: Output of the ML approach for the July 2019 volcanic eruption.

Volcanic activity evolution detected by the ML algorithm from the 18th to the 21st of July, 2019. "EX", "EX+EF", "EF" and "NA" stand for explosive, explosive and effusive, effusive and no activity respectively.

For each day, the retrieved activity for each thermal camera is shown as a colored bar, namely yellow (ENT), green (EMOT), blue (EBT), pink (ESR) and orange (EMCT). The visibility index is given by the average of the visibility indexes for all the cameras.



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