



## A New Large-Scale Sentinel-2 Benchmark Archive and A Three-Branch CNN

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## Outline



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- 3 BigEarthNet: Large-Scale Sentinel-2 Benchmark Archive

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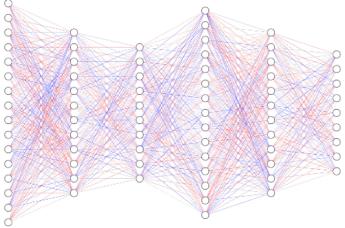


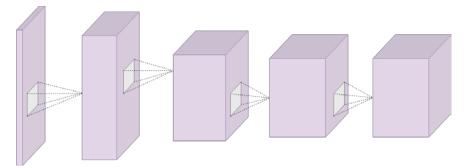
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#### Introduction



- One of the most challenging and emerging applications in remote sensing (RS) is related to the accurate description of RS images present in the archives.
- Recent advances in deep learning have attracted great attention RS due to high capability of deep networks, e.g.,
  - Convolutional Neural Network;
  - Recurrent Neural Networks;
  - Generative Adversarial Networks.





✓ To train such networks, very large training sets are needed with a high number of annotated images.



## **Existing Benchmark Archives in RS**



Archive Name	Image Type	Annotation Type	Number of Images
UC Merced	Aerial RGB	Single Label	2100
		Multi-Label	2100
WHU-RS19	Aerial RGB	Single Label	1,005
RSSCN7	Aerial RGB	Single Label	2,800
SIRI-WHU	Aerial RGB	Single Label	2,400
AID	Aerial RGB	Single Label	10,000
NWPU-RESISC45	Aerial RGB	Single Label	31,500
RSI-CB	Aerial RGB	Single Label	36,707
EuroSat	Satellite Multispectral	Single Label	27,000
PatternNet	Aerial RGB	Single Label	30,400

**Problem:** Publicly available RS image archives contain only a small number of annotated images and a large-scale benchmark archive does not yet exist.



#### **State of the Art Solutions**



 ✓ Use of deep learning models pre-trained on large scale computer vision archives (e.g., ImageNet)



**Problem:** Differences on the characteristics of images between computer vision and RS.



### **Limitations on Existing Archives in RS**



✓ Existing RS archives contain images annotated by single high-level category labels.



farmland

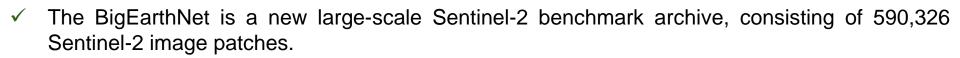


non-irrigated arable land, vineyards, pastures, land principally occupied by agriculture

**Problem:** RS images generally contain multiple classes associated to different landcover class labels (i.e., multi-labels).

 Most of the benchmark archives contain Aerial images that include only RGB image bands.

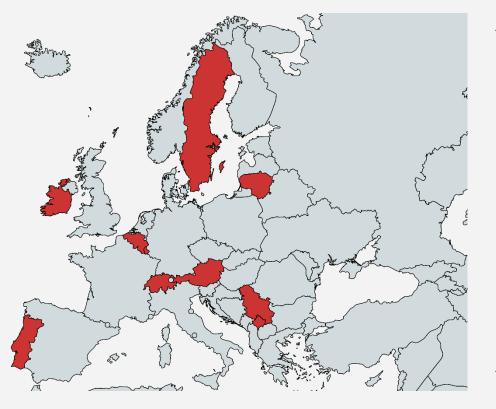








✓ To construct the BigEarthNet, 125 Sentinel-2 tiles (associated to cloud cover percentage less than 1%) acquired between June 2017 and May 2018 were selected.



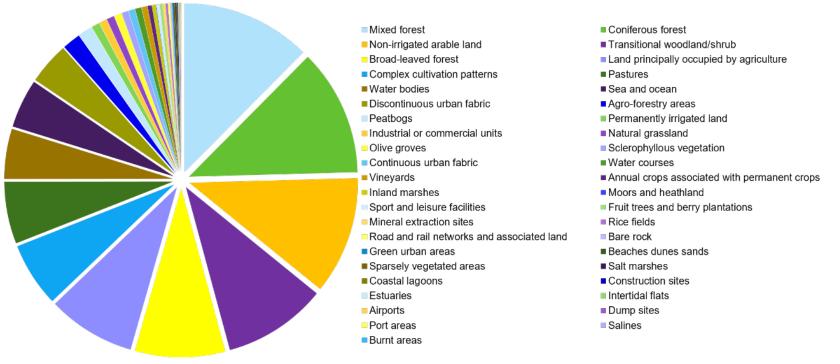
 Considered tiles are distributed over the 10 countries of Europe:

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- Austria
- Belgium
- Finland
- Ireland
- Kosovo
- Lithuania
- Luxembourg
- Portugal
- Serbia
- Switzerland
- All the tiles were atmospherically corrected.



- Selected tiles were divided into 590,326 non-overlapping image patches, each of which has size of 120x120 pixels in 10 meter resolution.
- Each image patch is associated with one or more land-cover class labels provided from the CORINE Land Cover (CLC) database of the year 2018 (CLC 2018).
- ✓ It is produced with assistance from the European Environment Agency's Eionet network.





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✓ The number of labels associated with each image patch varies between 1 and 12, whereas 95% of patches have at most 5 multi-labels.



Continuous urban fabric, Green urban areas



Non-irrigated arable land, Fruit trees and berry plantations, Pastures



, Coniferous forest, D Mixed forest, Water bodies, Transitional woodland/shrub.



Discontinuous urban fabric, Construction sites, Green urban areas

Images acquired in different seasons are considered.

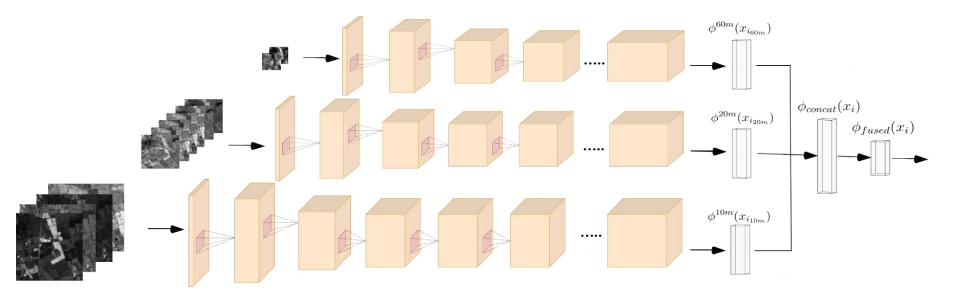
Seasons	Number of Image Patches		
Autumn	154,943		
Winter	117,156		
Spring	189,276		
Summer	128,951		



### **Three Branch CNN (TB-CNN)**



- ✓ TB-CNN includes three different convolutional branches specifically designed for different spatial resolutions of Sentinel-2 bands.
- ✓ Each branch acts as a feature extractor for different resolutions

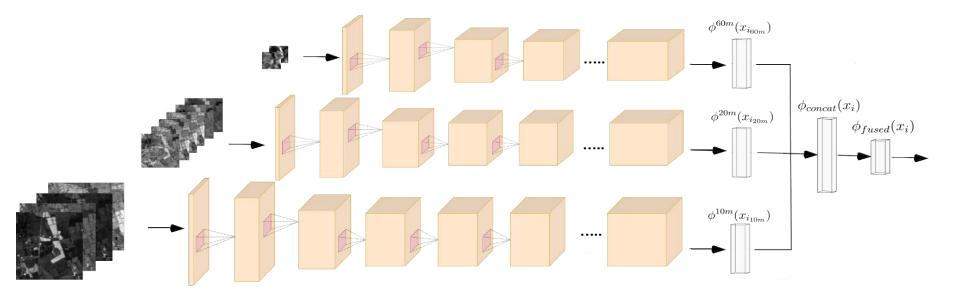




### **Three Branch CNN (TB-CNN)**



- ✓ For the first and second branches developed for 60m and 20m resolutions, 2x2 filters and 3x3 filters are used, respectively, throughout the layers.
- ✓ 5x5 filters for initial layers and 3x3 filters for deeper layers are used for the last branch, which accepts 10m resolution bands.



### **Applications on BigEarthNet**







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**Query Image** 







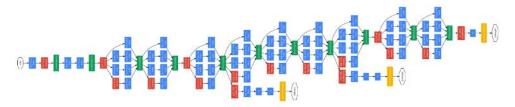
**Retrieved Images** 

Olive groves Land principally occupied by agriculture **Broad-leaved forest** Transitional woodland/shrub Water bodies

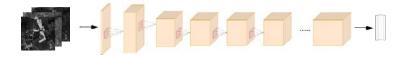
### **Design of Experiments**



- $\checkmark$  We have compared the results with those obtained by:
  - Fine-tuning the last layer of Inception-v2 pre-trained on ImageNet.



• Standard CNN architecture trained on only RGB bands



• Standart CNN architecture trained on all spectral bands







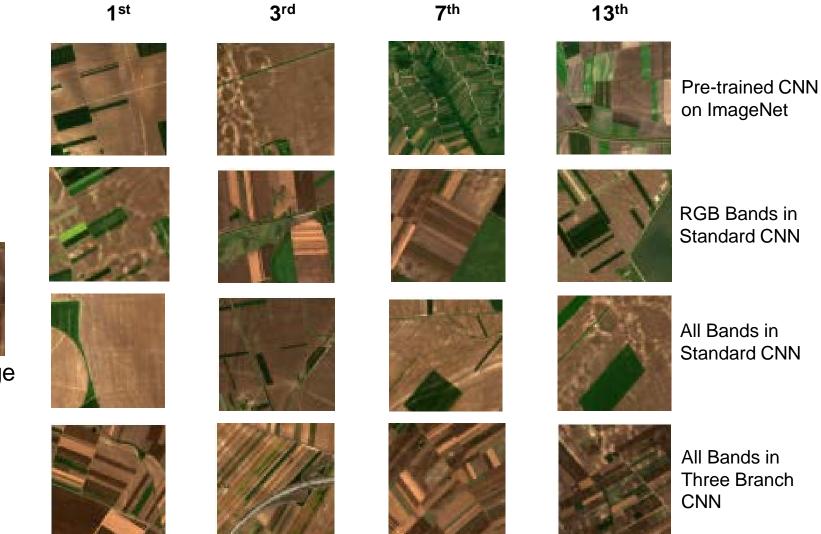
Methods	Recall	F <sub>1</sub> Score	F <sub>2</sub> Score
Pre-trained CNN on ImageNet*	40.75 %	0.4171	0.4085
RGB Bands in Standard CNN	54.72 %	0.5543	0.5451
All Bands in Standard CNN	57.20 %	0.6083	0.5812
All Bands in Three Branch CNN	70.62 %	0.6519	0.6763

\* We apply fine-tuning to the pre-trained Inception-v2 architecture.



#### **Results of Content Based Image Retrieval**

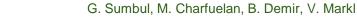






Query Image







**Results of Content Based Image Retrieval** 

#### Query Image



Non-irrigated arable land

Pre-trained CNN on ImageNet



Industrial or commercial units, Non-irrigated arable land

RGB Bands in Standard CNN



Non-irrigated arable land, Pastures, Water bodies All Bands in Standard CNN



Discontinuos urban fabric, Non-irrigated arable land All Bands in Three Branch CNN

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Non-irrigated arable land



#### **Conclusion and Future Developments**



- ✓ We have introduced a large-scale benchmark archive that consists of 590,326 Sentinel-2 image patches annotated by multi-labels, for RS image understanding.
- BigEarthNet will make a significant advancement for the use of deep learning in RS by overcoming current limitations of the existing archives.
- ✓ We plan to regularly enrich the BigEarthNet by increasing the number of annotated Sentinel-2 images.
- ✓ We are currently working on designing and implementing a scalable architecture for massive processing and analysis of images in the BigEarthNet.







# http://bigearth.net/

#### We would like to thank to all RSiM and DIMA group members!



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