



Assessment of the potential of Earth observation data and deep convolutional neural networks to improve the estimation and forecast of the solar power production in France

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Context and overview



Not all PV are measured equally

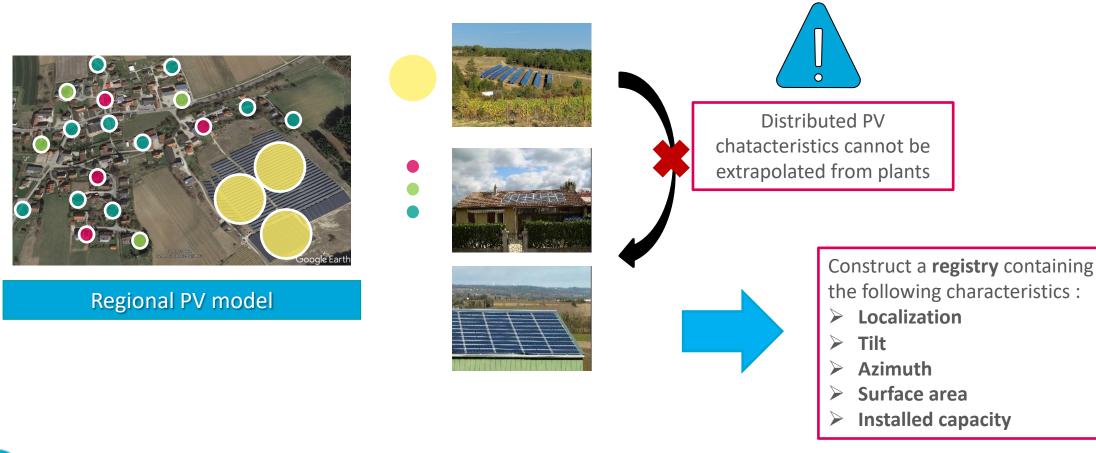


Plants ($\geq 1 MWp$) Mesurable : RTF can access real time Photovoltaic energy measurements of the production Increasing installed \succ capacity Decentralized < 1 MWp) Ever greater impact on the grid Not observable \triangleright Distributed Small plants decentralized Т Increasing uncertainty for load and supply balancing _ _ _ **Observable** : RTE can estimate the PV production



Bridging the gap leveraging a regional PV model and a registry

We can fit a **regional PV Model** to get an accurate estimation of the decentralized PV However, **we need statistics on the characteristics of PV systems**





Leveraging deep learning and EO data to build the registry

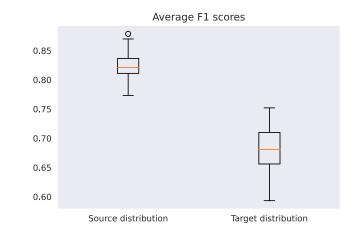
Previous studies (Yu et al, 2018, Mayer et al, 2022) proved that **deep learning** and **Earth observation data** can be leveraged to **construct large scale PV registries**

	824 53 47 62 41 23	Id	Loc.	Tilt	Azim	Inst. Cap.	Area
		0	45°, 3°	30°	-10°	3 kWp	12 sqm
		1	44°, 3°	45°	-30°	2,9 kWp	10 sqm
	64 52 56	2	46°, 3°	10°	0°	30 kWp	100 sqm
	62 54 42 40 42 40						
1. Detection and	2. Filtering and characteristics	Outcome : automated PV registry					
segmentation of PV installations	extraction using topological and LiDAR data						
Image classification and	Automatic extraction of	 Aggregated and standardized data Expected to be the largest registers with this level of 					
image segmentation	the characteristics						
		registry with this level of					

detail

Scientific questions

- Performance of current models are worse on France than on their source data. Issue: sensitivity to domain shift (which is widespread in machine learning)
- We need to be able to **certify the method**:
 - 1. We need to <u>understand how domain shift affects a model's</u> <u>accuracy.</u>
 - 2. We need to <u>assess and investigate the accuracy</u> over the whole deployment area.



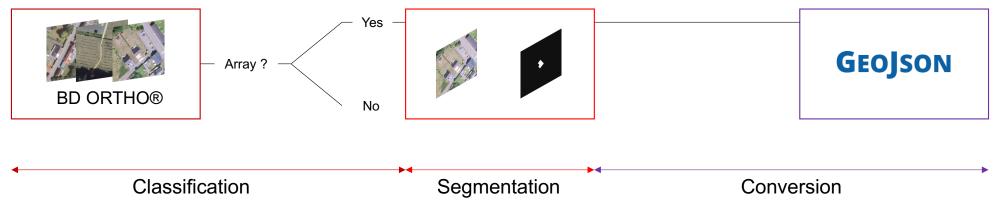
Performance drop from one region to the other (metric : F1 score)



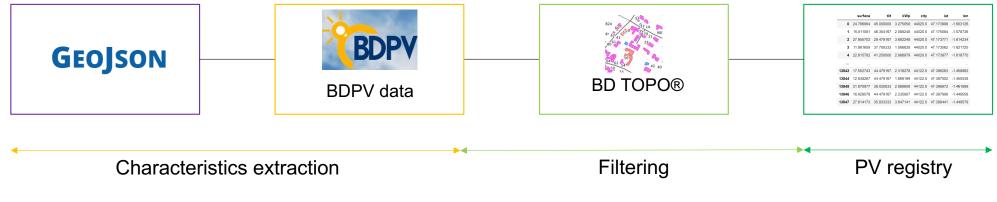




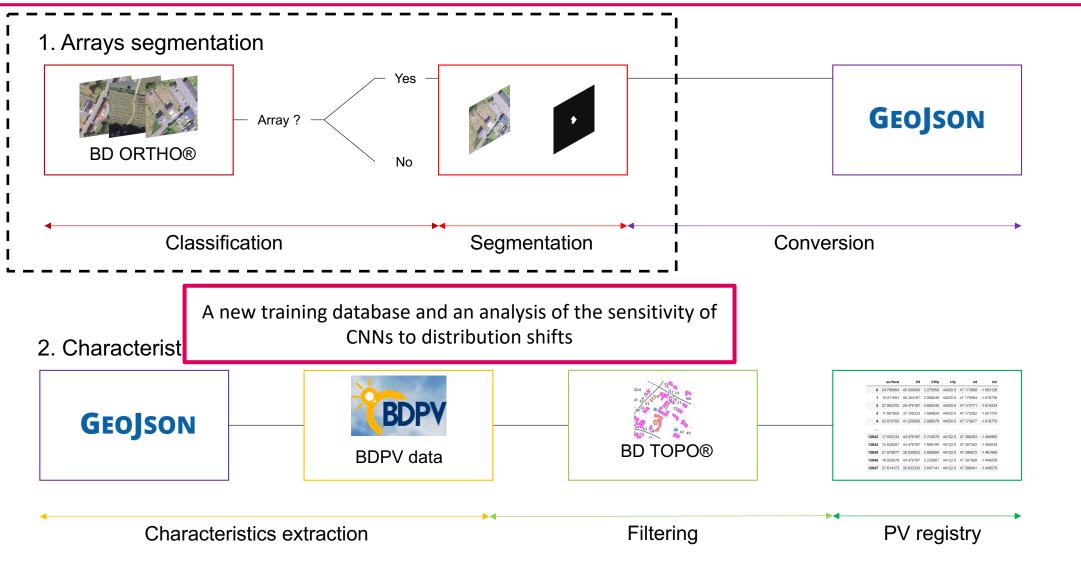
1. Arrays segmentation



2. Characteristics extraction and filtering

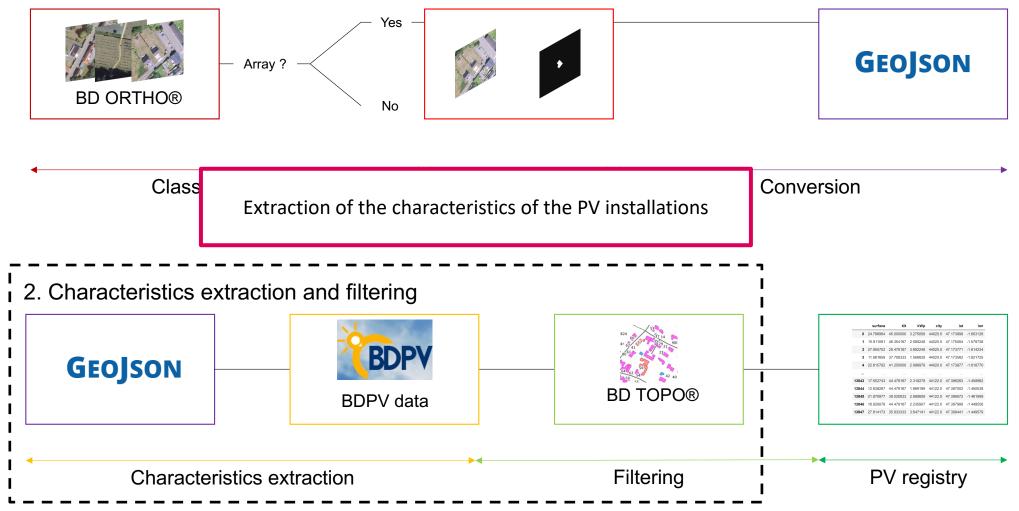






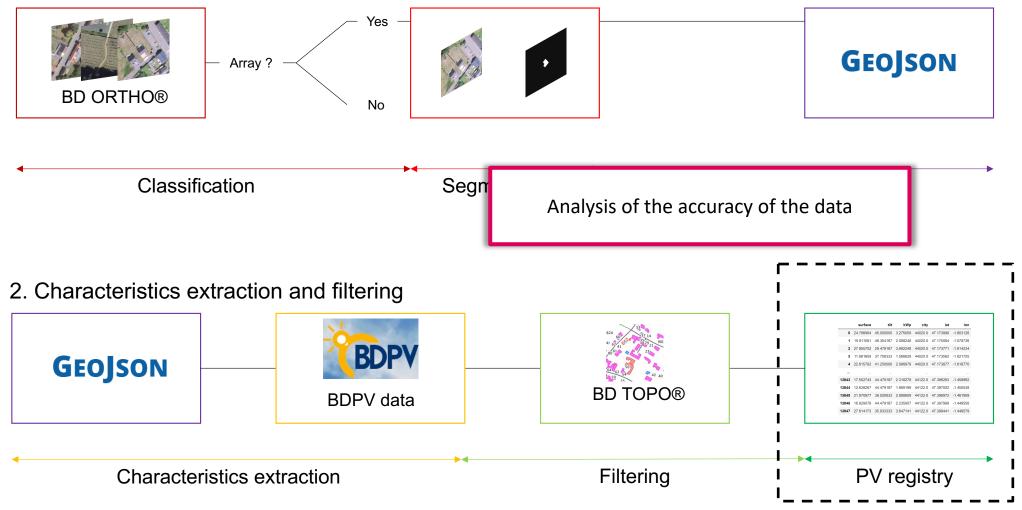


1. Arrays segmentation





1. Arrays segmentation



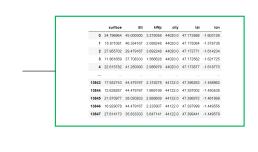








Large scale assessement of the accuracy of the registry



PV registry

- Problem: standard metrics (accuracy, F1, IoU,...) have no meaning from the application standpoint
- **Proposed approach : re-aggregate** the characteristics and compare them with the smallest scale available, the national registry of installations (*Registre national d'installations RNI*)
- The RNI contains the aggregated installed capacity and the number of installations for installations that have an installed capacity lower than 36 kWp at the scale of the cities.

It allows us to **satisfy certifiability constraints:**

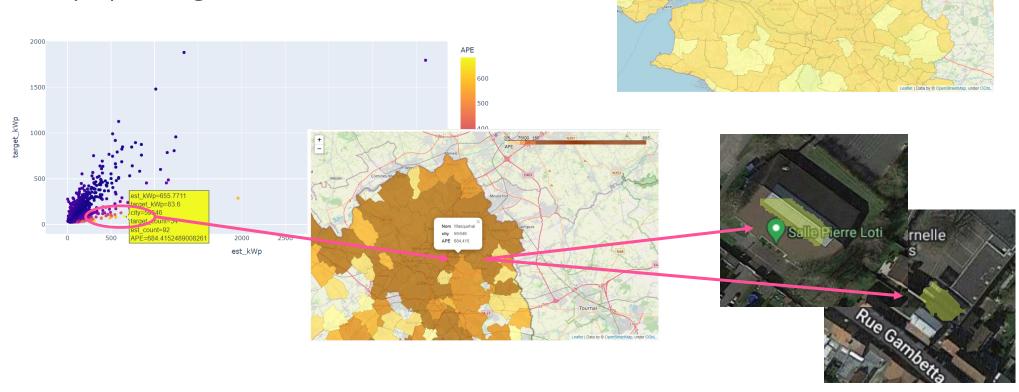
- 1. Control the accuracy over the whole territory *RNI is available everywhere in France*
- 2. Outputs investigation

Tools can be developped to visualize the results



Large scale assessement of the accuracy of the registry

- « End user » accuracy example: average percentage error of the installed capacity, per city
- → Identify « pathological cases »









Thank you ! Questions ?

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