



**RÉPUBLIQUE  
FRANÇAISE**

*Liberté  
Égalité  
Fraternité*

**CNIG**

Conseil national  
de l'information  
géolocalisée



# Intelligence artificielle & informations géolocalisées

Atelier – 8 janvier 2026

# Atelier 2

Les IA métier pour la production de données :  
état des lieux  
(Extraction, structuration et  
traitement d'images géographiques)

Jocelyn Chanussot, INRIA

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Matthieu Porte, MTE Ecolab

# AI and Remote Sensing

Jocelyn Chanussot



**Future INRIA project team:**  
**ReSeT**  
**Remote Sensing Team @ Inria**

THOTH

Modeling visual knowledge from large-scale data



# Remote sensing

**Sensing:** Observing, measuring, monitoring

**Remotely:** from a distance (close range... or from far away)

Platforms:

satellites

airplanes

UAV (drones) *and more*

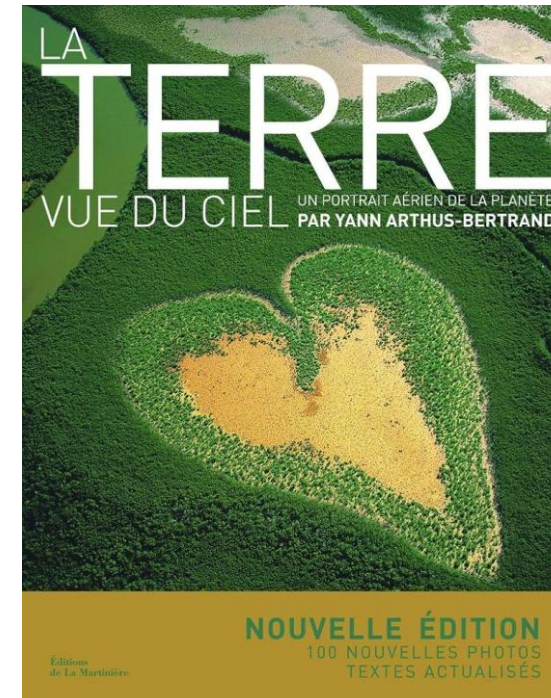
Sensors:

Optical

Hyperspectral

Radar

Lidar *and more*







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# Remote sensing



La plus vieille photographie aérienne de Nadar



1858



vue aérienne de Paris (1866)

# Remote sensing



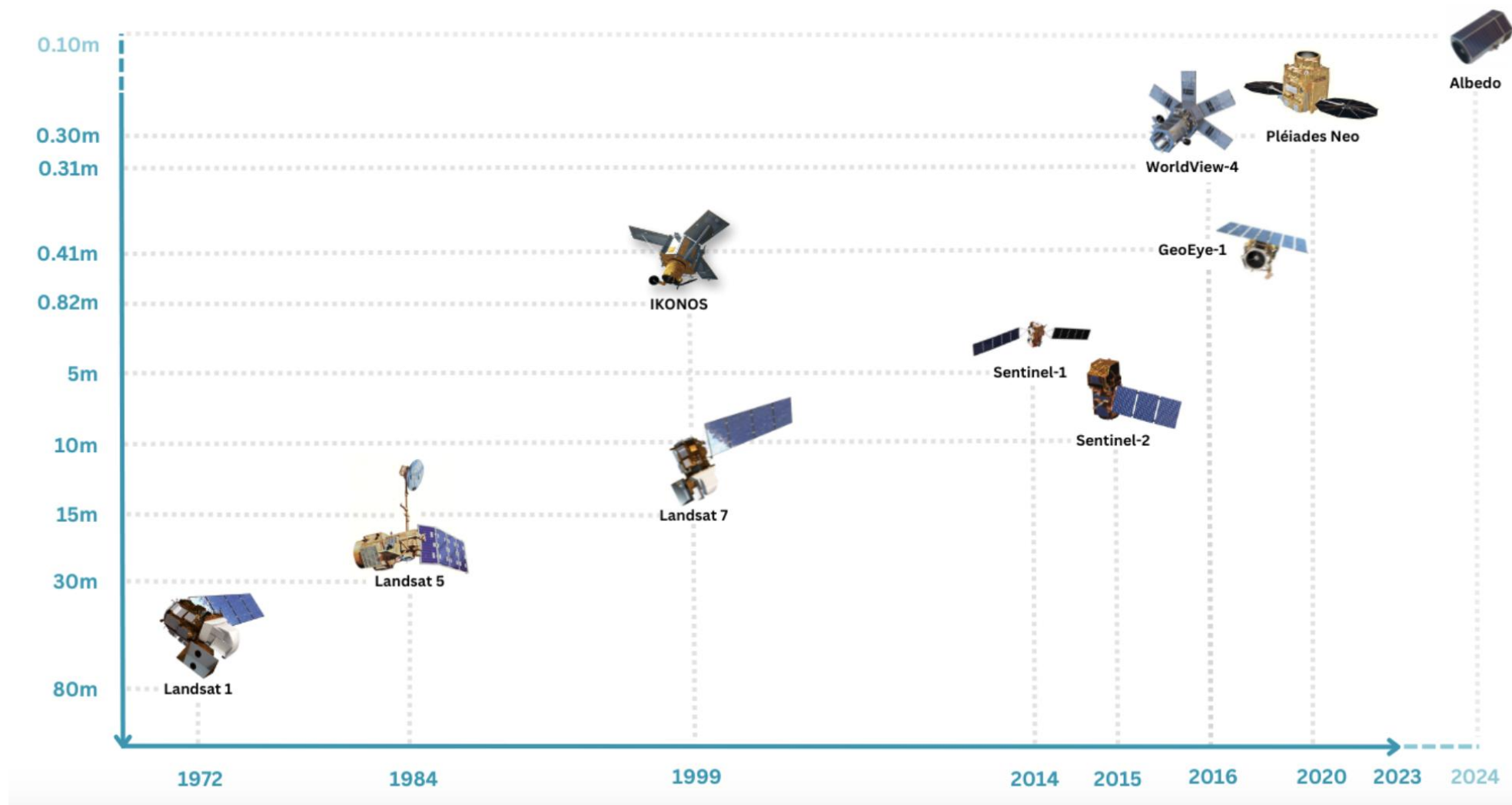
1903, système développé par Julius Neubranner



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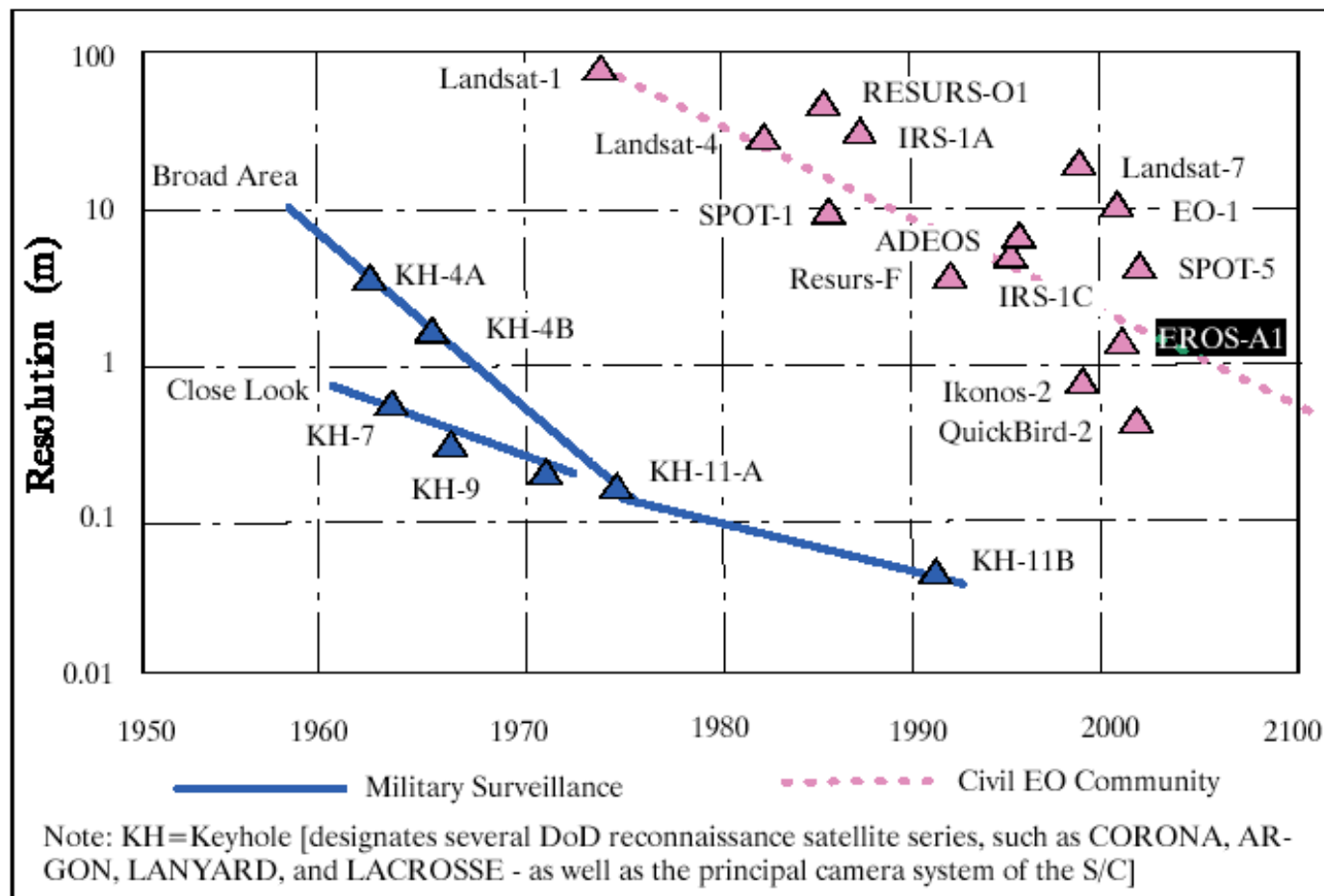




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# Remote sensing



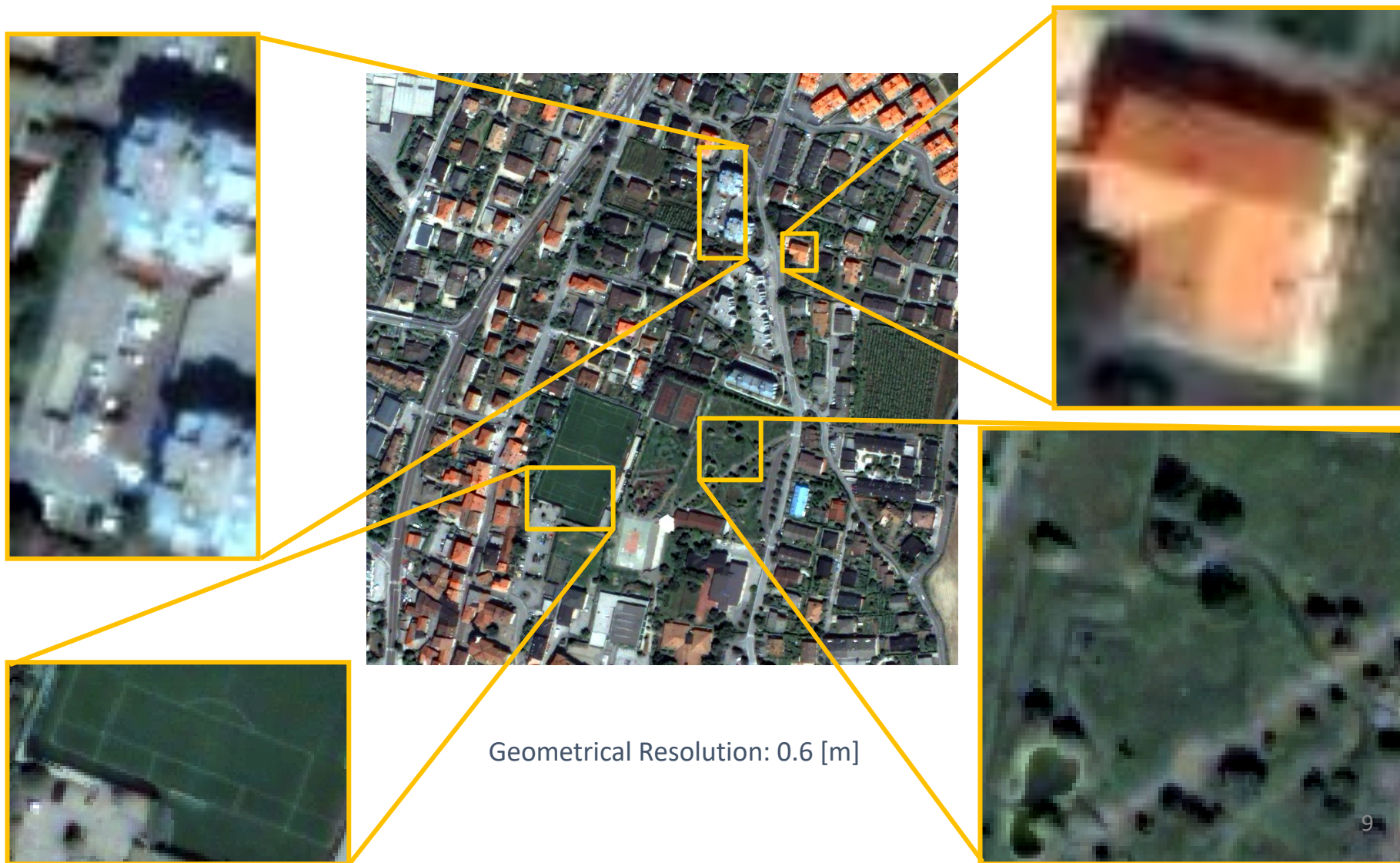




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# Remote sensing





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## Active government satellites [\[ edit \]](#)

Name	Status	Agency	Launch <a href="#">[note 1]</a>	Description
<a href="#">ALOS-2</a>	Active	<a href="#">JAXA</a>	2014	
<a href="#">ALOS-4</a>	Active	<a href="#">JAXA</a>	2024	
<a href="#">Alsat-2A and 2B</a>	Active	<a href="#">Algerian Space Agency (ASAL)</a>	2016	
<a href="#">Amazônia-1</a>	Active	<a href="#">Brazil's National Institute for Space Research (INPE)</a>	2021	
<a href="#">Aqua</a>	Active	<a href="#">NASA</a>	2002	Carries six instruments to observe interactions among the four spheres for Earth's systems: oceans, land, atmosphere, and biosphere. <sup><a href="#">[5]</a></sup>
<a href="#">ASNARO-2</a>	Active	<a href="#">JAXA</a>	2018	
<a href="#">Aura</a>	Active	<a href="#">NASA</a>	2004	Studies earth's ozone, air quality, and climate through observation of composition, chemistry, and dynamics of the atmosphere. <sup><a href="#">[6]</a></sup>
<a href="#">Badr-B</a>	Active	<a href="#">Pakistan's Space and Upper Atmosphere Research Commission (SUPARCO)</a>	2001	
<a href="#">Cartosat-1</a>	Active	<a href="#">Indian Space Research Organization (ISRO)</a>	2005	
<a href="#">Cartosat-2A and 2B</a>	Active	<a href="#">ISRO</a>	2007	





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Name ◆	Status ◆	Agency ◆	Launch <a href="#">[note 1]</a> ◆	Description ◆
<a href="#">Cartosat-2C, 2D, 2E, and 2F</a>	Active	ISRO	2016	
<a href="#">Cartosat-3</a>	Active	ISRO	2019	
<a href="#">CBERS-4</a>	Active	<a href="#">Brazil's National Institute for Space Research (INPE) and China National Space Administration (CNSA)</a>	2014	
<a href="#">Chollian 1, 2A, and 2B</a>	Active	<a href="#">KARI</a>	2010	Also known as Communication, Ocean and Meteorological Satellites (COMS). Used for communication, oceanography, and meteorological observation.
<a href="#">CloudSat</a>	Active	NASA	2006	Uses radar to measure the altitude and properties of clouds. <sup>[7]</sup>
<a href="#">COSMO-SkyMed 1 to 4</a>	Active	<a href="#">Italian Space Agency (ASI)</a>	2007	Used for defense and security assurance in Italy and other countries, seismic hazard analysis, environmental disaster monitoring, and agricultural mapping.
<a href="#">CryoSat-2</a>	Active	<a href="#">ESA</a>	2010	
<a href="#">CYGNSS</a>	Active	NASA	2016	Cyclone Global Navigation Satellite System.
<a href="#">DSCOVR</a>	Active	NASA	2015	Deep Space Climate Observatory. Designed to study the Sun-lit side of Earth from the L1 <a href="#">Lagrange point</a> . <sup>[8]</sup>
<a href="#">DubaiSat-1 and 2</a>	Active	<a href="#">Mohammed bin Rashid Space Centre (MBRSC)</a>	2009	



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Name ⇅	Status ⇅	Agency ⇅	Launch [note 1] ⇅	Description ⇅
<a href="#">EarthCARE</a>	Active	ESA and JAXA	2024	Designed to study clouds and aerosols. <sup>[9]</sup>
<a href="#">Elektro-L No. 1, 2, and 3</a>	Active	Russia's <a href="#">Roscosmos</a>	2011	
<a href="#">Fengyun 2D to 4A</a>	Active	<a href="#">China Meteorological Administration</a>	2006	Translated from Chinese, the word Fengyun means "wind cloud." Series 3 satellites are Sun-synchronous and series 2 and 4 satellites are geosynchronous.
<a href="#">Formosat-5</a>	Active	<a href="#">Taiwan's National Space Organization (NSPO)</a>	2017	
<a href="#">Gaofen-2</a>	Active	<a href="#">CNSA</a>	2014	
<a href="#">Gaofen-3</a>	Active	<a href="#">CNSA</a>	2016	
<a href="#">GOES-16 and 17</a>	Active	NASA	2016	Geostationary Operational Environmental Satellite. Collects weather observations. See also: <a href="#">List of GOES satellites</a> .
<a href="#">Gokturk-1</a>	Active	<a href="#">Turkish Ministry of National Defense</a>	2016	Used for mapping and planning, landcover survey, geology, ecosystem monitoring, disaster management, environmental control, coastal zone management, and water resources.
<a href="#">Gokturk-2</a>	Active	<a href="#">Turkish Ministry of National Defense</a>	2012	Used for mapping and planning, landcover survey, geology, ecosystem monitoring, disaster management, environmental control, coastal zone management, and water resources. <sup>[10]</sup>
<a href="#">GPM</a>	Active	<a href="#">NASA</a> and <a href="#">JAXA</a>	2014	Global Precipitation Measurement (GPM) Core Observatory. Used to study rainfall and snowfall. <sup>[11][12]</sup>



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Name	Status	Agency	Launch <a href="#">[note 1]</a>	Description
GOES-16, -17	Active	NASA	2016	monitors weather for NOAA.
GRACE-FO	Active	NASA	2018 <sup>[13][14]</sup>	Gravity and climate. The mission will track changes in global sea levels, glaciers, and ice sheets, as well as large lake and river water levels, and soil moisture. <sup>[15]</sup>
GOSAT	Active	JAXA	2009	Greenhouse Gases Observing Satellite. The first precise carbon dioxide observing satellite and precursor to OCO-2.
Himawari 8 and 9	Active	<a href="#">Japan Meteorological Agency</a>	2014	Similar to NASA's GOES satellites.
ICESat-2	Active	NASA	2018	Measures ice sheet height changes for climate change diagnoses. <sup>[16][17]</sup>
IMS-1	Active	<a href="#">ISRO</a>	2008	
ISS	Active	<a href="#">NASA</a> , <a href="#">Roscosmos</a> , <a href="#">JAXA</a> , <a href="#">ESA</a> , and <a href="#">CSA</a>	1998	The <a href="#">International Space Station (ISS)</a> has long been used as a central satellite platform for other sensors, including Earth observation sensors. For example: <a href="#">LIS</a> , <a href="#">SAGE III</a> , <a href="#">TSIS-I</a> , <a href="#">ECOSTRESS</a> , <a href="#">GEDI</a> , <a href="#">OCO-3</a> , <a href="#">Diwata-1</a> , and <a href="#">HICO</a> .
Jason-3	Active	<a href="#">NASA</a> and <a href="#">CNES</a>	2016	Radar altimeter used to monitor ocean surface height.
KhalifaSat	Active	<a href="#">MBRSC</a>	2018	Also known as DubaiSat-3.
KOMPSAT-2	Active	<a href="#">KARI</a>	2006	Korean Multi-purpose Satellite-2. Also known as Arirang-2.
KOMPSAT-3, 3A, and 5	Active	<a href="#">KARI</a>	2012	Korean Multi-purpose Satellite-3, 3A, and 5. Also known as Arirang-3, 3A, and 5.



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Name	Status	Agency	Launch <a href="#">[note 1]</a>	Description
<a href="#">LAGEOS-1 and 2</a>	Active	NASA	1976	LAGEOS 1 launched in 1976 and LAGEOS 2 launched in 1992. Used as an orbiting benchmark for geodynamical studies. <sup>[18]</sup>
<a href="#">Landsat-7</a>	Active	NASA and USGS	1999	Images Earth's land surfaces and coastal areas with global coverage at high spatial resolution. <sup>[19]</sup>
<a href="#">Landsat-8</a>	Active	NASA and USGS	2013	Follow on to Landsat-7 with improved imager OLI and thermal sensor TIRS.
<a href="#">Landsat-9</a>	Active	NASA and USGS	2021	Follow on to Landsat-8 with OLI sensor and thermal sensor TIRS-2. Landsat-9 will extend the <a href="#">Landsat program</a> to maintain the time series of these type of data.
<a href="#">Megha-Tropiques</a>	Active	<a href="#">CNES</a> and <a href="#">ISRO</a>	2011	
<a href="#">Meteor-M No. 1 and 2</a>	Active	<a href="#">Roscosmos</a>	2009	
<a href="#">Meteosat 8</a>	Active	<a href="#">EUMETSAT</a>	2002	Also known as MSG 1.
<a href="#">MetOp A, B, and C</a>	Active	<a href="#">NASA</a> , <a href="#">ESA</a> , and <a href="#">NOAA</a>	2006	Meteorological Operational satellite. Part of the Polar Operational Environmental Satellites (POES) program.
<a href="#">Mohammed VI-A and VI-B</a>	Active	<a href="#">Arianespace</a> and <a href="#">Morocco</a>	2017	<i>See also: Vega flight VV11 and Vega flight VV13</i>
<a href="#">NigComSat-1R</a>	Active	<a href="#">NASRDA</a>	2009	DFH-4 satellite and replacement for the failed NigComSat-1
<a href="#">NigeriaSat-1 and 2</a>	Active	<a href="#">NASRDA</a>	2003	Part of the worldwide Disaster Monitoring Constellation System



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Name	Status	Agency	Launch <a href="#">[note 1]</a>	Description
<a href="#">NOAA-15, 18, and 19</a>	Active	<a href="#">NASA</a> , <a href="#">ESA</a> , and <a href="#">NOAA</a>	1998	Part of the Polar Operational Environmental Satellites (POES) program.
<a href="#">NOAA-20</a>	Active	<a href="#">NASA</a> and <a href="#">NOAA</a>	2017	Part of the Joint Polar Satellite System (JPSS) program.
<a href="#">Oceansat-2</a>	Active	<a href="#">ISRO</a>	2009	
<a href="#">OCO-2</a>	Active	<a href="#">NASA</a>	2014	Orbiting Carbon Observatory 2. Part of the A-Train. The second precise carbon dioxide observing satellite after GOSAT.
<a href="#">PACE</a>	Active	<a href="#">NASA</a>	2024	Plankton, Aerosol, Cloud, and ocean Ecosystem measures hyperspectral and polarimetric characteristics of solar radiation reflected off Earth's surface and atmosphere.
<a href="#">PakTES-1A</a>	Active	<a href="#">SUPARCO</a>	2018	
<a href="#">Paz</a>	Active	Spain's <a href="#">Instituto Nacional de Técnica Aeroespacial</a>	2018	
<a href="#">Pleiades 1A and 1B</a>	Active	<a href="#">CNES</a>	2011	
<a href="#">PRISMA</a>	Active	<a href="#">Italian Space Agency (ASI)</a>	2019	PRecursore IperSpettrale della Missione Applicativa, in English: Hyperspectral PRecursor of the Application Mission.
<a href="#">PROBA-V</a>	Active	<a href="#">ESA</a>	2013	Continues the traditional Vegetation (the "V" in PROBA-V) products that began with the SPOT satellites. <sup><a href="#">[20]</a></sup>
<a href="#">PRSS-1</a>	Active	<a href="#">SUPARCO</a>	2018	Pakistan Remote Sensing Satellite 1.
<a href="#">RCM</a>	Active	<a href="#">CSA</a>	2019	RADARSAT Constellation Mission.



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Name	Status	Agency	Launch <a href="#">[note 1]</a>	Description
<a href="#">RADARSAT-2</a>	Active	<a href="#">CSA</a>	2007	C-band synthetic aperture radar (SAR-C) satellite.
<a href="#">RASAT</a>	Active	<a href="#">TÜBITAK-UZAY</a>	2011	
<a href="#">Resourcesat-1 and 2</a>	Active	<a href="#">ISRO</a>	2003	
<a href="#">Resurs-P No.1 and 2</a>	Active	<a href="#">Roscosmos</a>	2013	
<a href="#">SAOCOM</a>	Active	<a href="#">CONAE</a>	2018	
<a href="#">SARAL</a>	Active	<a href="#">ISRO</a>	2013	
<a href="#">Sentinel-1A, B and C</a>	Active	<a href="#">ESA</a>	2014	Constellation of two, each satellite carries C-SAR sensor. Part of the <a href="#">Copernicus Programme</a> .
<a href="#">Sentinel-2A, B, and C</a>	Active	<a href="#">ESA</a>	2015	Constellation of three, each satellite carries MSI sensor for high spatial resolution imaging. Part of the Copernicus Programme.
<a href="#">Sentinel-3A and B</a>	Active	<a href="#">ESA</a>	2016	Constellation of two, each satellite carries sensors OLCI and SLSTR. Slightly coarser spatial resolution and more spectral bands than Sentinel-2. Part of the Copernicus Programme.
<a href="#">Sentinel-6A</a>	Active	<a href="#">ESA</a>	2020	Continuing the legacy of the Jason series missions, Sentinel-6/Jason-CS will extend the records of sea level (sea surface height) and provide information for operational oceanography, marine meteorology, and climate studies. <sup><a href="#">[21]</a></sup>





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Name	Status	Agency	Launch <a href="#">[note 1]</a>	Description
SMAP	Active	NASA	2015	Soil Moisture Active Passive. Measures soil moisture and its freeze/thaw state, which enhance understanding of processes that link water, energy, and carbon cycles to extend the capabilities of weather and climate models. Radar payload failed in July 2015, leaving a radiometer as the primary instrument of the mission. <sup>[22]</sup>
SORCE	Active	NASA	2003	monitors total output from the Sun for understanding of Earth's absorption of radiation energy. <sup>[16]</sup>
Suomi NPP	Active	NASA	2011	Part of the Joint Polar Satellite System (JPSS) program.
SWOT	Active	NASA	2022	
TanDEM-X	Active	DLR	2010	
Terra	Active	NASA	1999	Carries five instruments to observe the state of the atmosphere, land, and oceans, as well as their interactions with solar radiation and with one another. <sup>[23]</sup>
TerraSAR-X	Active	DLR	2007	
THEOS	Active	GISTDA	2008	Also known as Thaichote.
TIMED	Active	NASA	2001	Thermosphere, Ionosphere, Mesosphere, Energetics, and Dynamics.
VNREDSat-1A	Active	VAST	2013	The Vietnamese Natural Resources, Environment and Disaster Monitoring Satellite.
VRSS-1 and 2	Active	ABAE	2012	The Venezuelan Remote Sensing Satellite.



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Name ◆	Status ◆	Agency ◆	Launch <a href="#">[note 1]</a> ◆	Description ◆
Sentinel-5 Precursor (S5P)	Active	ESA	2017	
SMOS	Active	ESA	2009	



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Name	Status	Owner/Agency	Launch <a href="#">[note 1]</a>
Disaster Monitoring Constellation <sup>[40]</sup>	Active	DMC International Imaging	2009
EROS A and B	Active	ImageSat International	2000
Flock-1 Constellation	Active	Planet	2014
GeoEye-1	Active	DigitalGlobe (Maxar)	2008
GRUS <a href="#">[ja]</a> -1A to E <sup>[41]</sup>	Active	Axelspace <a href="#">[ja]</a> <sup>[42]</sup>	2018
ICEYE	Active	ICEYE	2018
Jilin-1 (Hyperspectral)	Active	Chang Guang Satellite Technology	2019
Jilin-1 (Optical)	Active	Chang Guang Satellite Technology	2015
NovaSAR-S1 <sup>[42]</sup>	Active	UK Space Agency and Surrey Satellite Technology	2018
PlanetScope-2 <sup>[43]</sup>	Active	Planet	2016
SkySat-1 to 3	Active	Planet	2013
SkySat-4 to 7	Active	Planet	2016
SkySat-8 to 13	Active	Planet	2017
SPOT 6 and 7	Active	EADS Astrium Azercosmos, and CNES	2012
SuperView-1 <sup>[44]</sup>	Active	Beijing Space View Technology	2018
TripleSat (UK-DMC 3) <sup>[45][46]</sup>	Active	DMC International Imaging	2015
Vivid-i 1 to 5 <sup>[47]</sup>	Active	Earth-i <sup>[48]</sup>	2018
WorldView-1	Active	DigitalGlobe (Maxar)	2007
WorldView-2 and 3	Active	DigitalGlobe (Maxar)	2009
Pleiades Neo 3	Active	Airbus Defence and Space	2021
Pleiades Neo 4	Active	Airbus Defence and Space	2021

## Conclusion:

**We have A LOT of multimodal, highly heterogeneous, data**  
**New acquisitions and archives**

# Artificial intelligence

# Artificial intelligence

Artificial Intelligence is not that big, scary thing in the future.



# Artificial intelligence

Artificial Intelligence is not that big, scary thing in the future.  
It's here with us.

# Artificial intelligence

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Fei Fei Li  
American Computer Science Professor

# Artificial intelligence

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Fei Fei Li  
American Computer Science Professor  
1976



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# Artificial intelligence

Artificial Intelligence is not that big, scary thing in the future.  
It's here with us.

Fei Fei Li  
American Computer Science Professor  
1976

2025

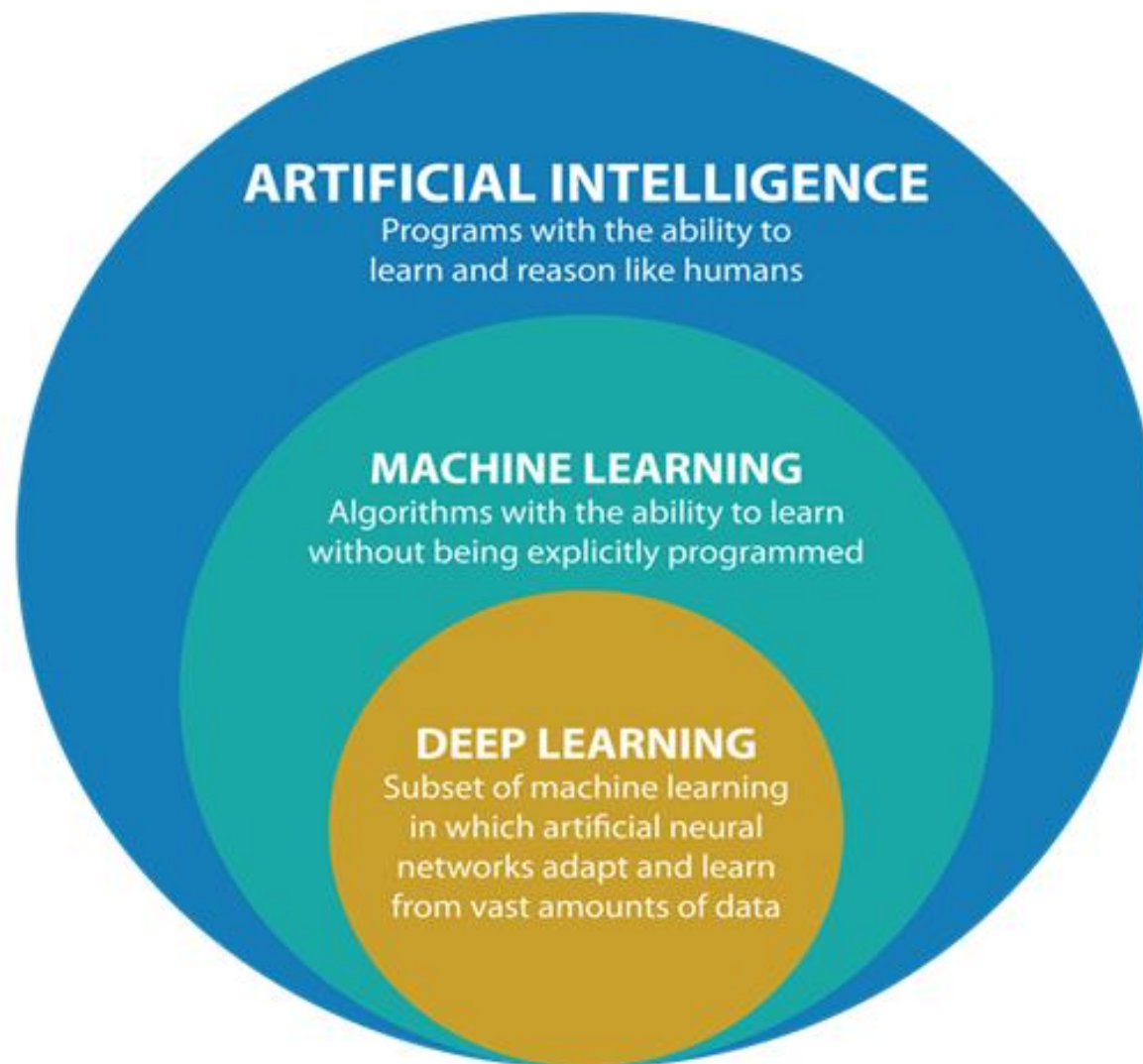




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# Artificial intelligence



## Conclusion:

We have A LOT of AI algorithms.  
High Performance Computing

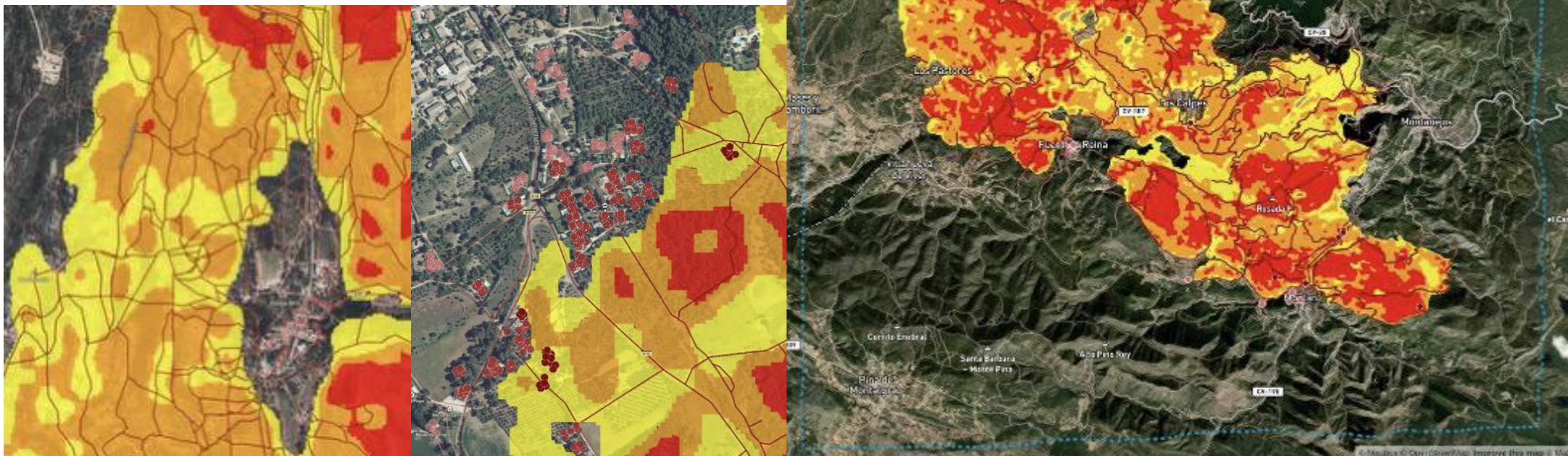




# AI and remote sensing



## Monitoring of wildfires

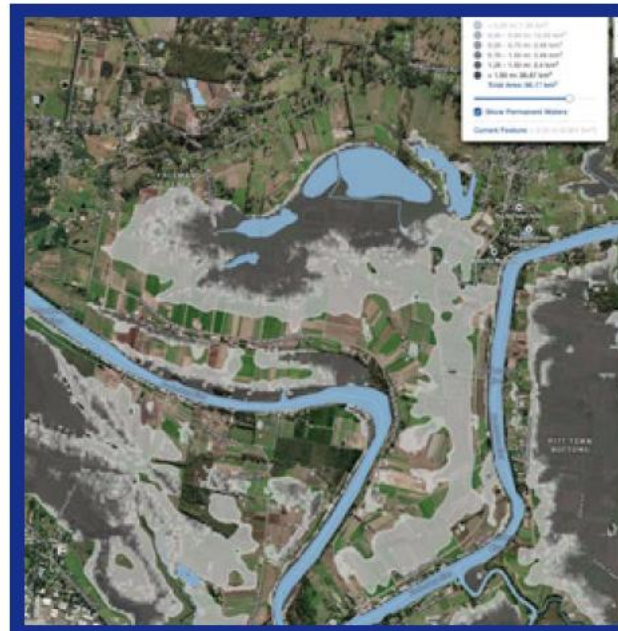




## Monitoring of floods



*Richmond (Australia) flood*

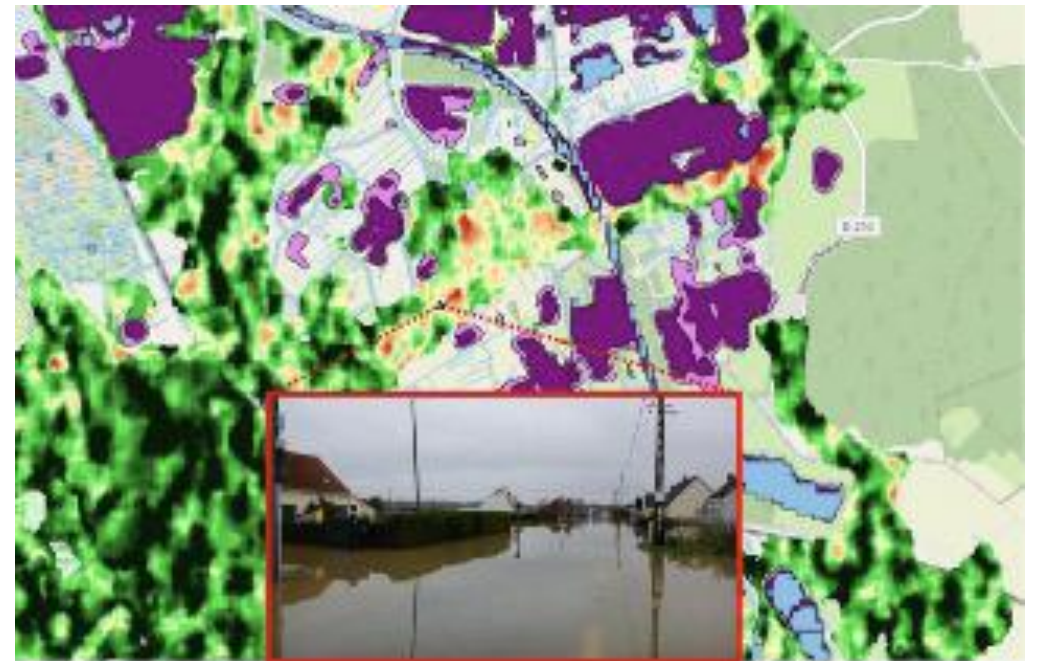


*Water depth*



*Impact on housing and roads*

## Monitoring of floods



# AI and remote sensing

**Conclusion: AI and deep learning:  
a game changer for the analysis of remote sensing data**

- A lot of data
- High performance computing
- Advanced algorithms (optimization)

**Remote sensing differs from standard computer vision:**

Not a lot of annotated data

-> importance of self-supervised techniques / few shot learning

Observations = physical quantities

-> importance of the interplay between physics and AI



A few recent papers published in IEEE TGRS (just a drop in the ocean...)

DASNet: Dual Attentive Fully Convolutional Siamese Networks for Change Detection  
in High-Resolution Satellite

SNUNet-CD: A Densely Connected Siamese Network for Change Detection of VHR Images

LANet: Local Attention Embedding to Improve the Semantic Segmentation  
of Remote Sensing Images

TEMDnet: A Novel Deep Denoising Network for Transient Electromagnetic Signal  
With Signal-to-Image Transformation

SemiCDNet: A Semisupervised Convolutional Neural Network for Change Detection  
in High Resolution Remote-Sensing Images

SCAttNet: Semantic Segmentation Network With Spatial and Channel Attention Mechanism  
for High-Resolution Remote Sensing Images

MAP-Net: Multiple Attending Path Neural Network for Building Footprint Extraction  
From Remote Sensed Imagery

SwinSUNet: Pure Transformer Network for Remote Sensing Image Change Detection

DLA-MatchNet for Few-Shot Remote Sensing Image Scene Classification

DABNet: Deformable Contextual and Boundary-Weighted Network for Cloud Detection  
in Remote Sensing Images

SSR-NET: Spatial–Spectral Reconstruction Network for Hyperspectral and  
Multispectral Image Fusion

TransUNetCD: A Hybrid Transformer Network for Change Detection  
in Optical Remote-Sensing Images

BockNet: Blind-Block Reconstruction Network With a Guard Window  
for Hyperspectral Anomaly Detection

F3Net: Adaptive Frequency Feature Filtering Network for Multi-modal Remote Sensing  
Image Registration

EMYNet-BDD: EfficientViTB Meets Yolov8 in the Encoder–Decoder Architecture for Building  
Damage Detection Using Postevent Remote Sensing Images

Your next publication:

**Jane/JohnDoeNet:**

**a great architecture for a critical application using specific remote sensing data**



## Foundation models

Move from *John/JaneDoeNet* to

- Large scale pre-trained AI models
  - Able to address a wide array of tasks
  - Finetuneable
- 
- **A lot of data**
  - **High performance computing**
  - **Advanced algorithms (optimization)**



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# AI and remote sensing

## Foundation models: specific models for remote sensing data

### Data

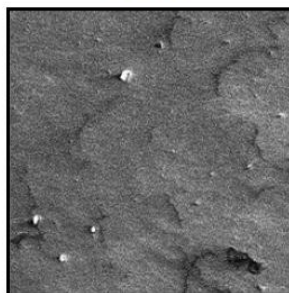
Panchromatic



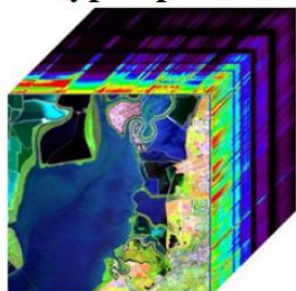
True Color



SAR



Hyperspectral



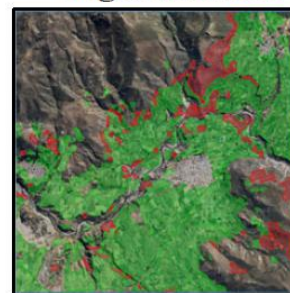
Multispectral



...

### Downstream Tasks

Segmentation



Object Detection



Classification



Change Detection



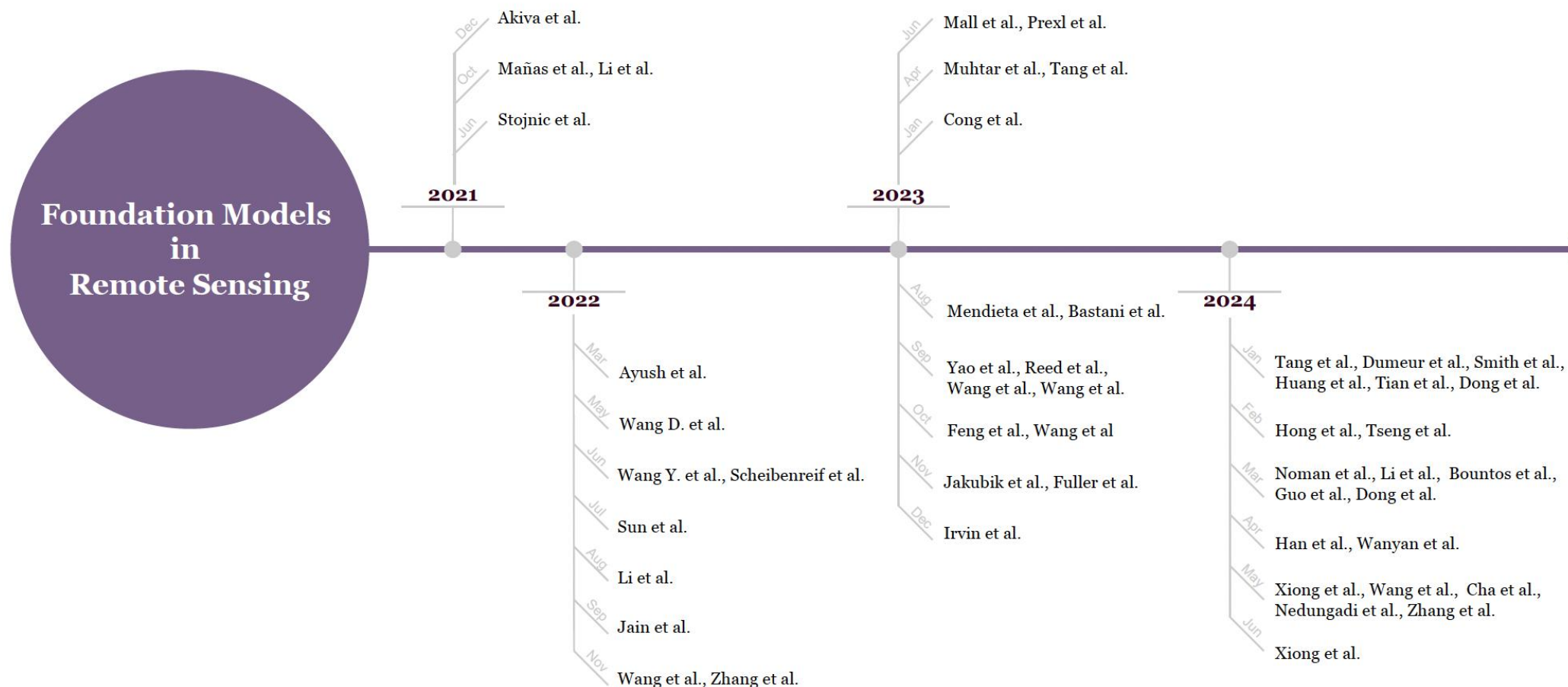


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# AI and remote sensing

## Foundation models: specific models for remote sensing data





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Foundation models:  
specific models  
for remote sensing data

Year-Month	Architecture	Model Name	Scene Classification	Semantic Segmentation	Object Detection	Change Detection
2021 Jun	ResNet-50	CMC-RSSR [59]	✓			
2021 Oct	ResNet-50	SeCo [46]	✓			✓
2021 Oct	ResNet-50	GeoKR [39]	✓	✓	✓	
2021 Dec	ResNet-34	MATTER [1]	✓	✓		✓
2022 Mar	ResNet-50	GASSL [3]	✓		✓	
2022 May	ViTAEv2-S	RSP [69]	✓	✓	✓	✓
2022 Jun	ViT-S/8	DINO-MM [73]	✓			
2022 Jun	Swin Transformer	Scheibenreif, et al. [56]	✓	✓		
2022 Jul	ViT/Swin Transformer	RingMo [61]	✓	✓	✓	✓
2022 Aug	ResNet-50	GeCo [40]	✓		✓	
2022 Sep	BYOL	RS-BYOL [32]	✓	✓		
2022 Nov	ViT-B	CSPT [87]	✓		✓	
2022 Nov	ViT	RVSA [71]	✓	✓	✓	
2023 Jan	MAE-based Framework	SatMAE [11]	✓	✓		
2023 Apr	TOV	TOV [63]	✓	✓	✓	
2023 Apr	Teacher-student Self-distillation	CMID [49]	✓	✓	✓	✓
2023 Jun	CACo	CACo [47]	✓	✓		✓
2023 Jun	ResNet-18	Ial-SimCLR [54]	✓			
2023 Aug	Teacher-Student	GFM [48]	✓			✓
2023 Aug	Swim Transformer	SatLasPretrain [4]	✓	✓		
2023 Sep	Multi-Branch	RingMo-Sense [85]	✓	✓		
2023 Sep	ViT	Scale-MAE [55]	✓	✓		
2023 Sep	CNN-Transformer	RingMo-lite [76]	✓	✓	✓	✓
2023 Sep	Multimodal SSL	DeCUR [72]	✓	✓		
2023 Oct	MSFE+MMFH	Feng et al. [21]	✓	✓	✓	✓
2023 Oct	ViT	FG-MAE [75]	✓	✓		
2023 Nov	ViT	Prithvi [33]	✓	✓		
2023 Nov	Multimodal Encoder	CROMA [22]	✓	✓		
2023 Dec	ViT	USat [31]	✓			
2024 Jan	ViT-B	Cross-Scale MAE [62]	✓	✓		
2024 Jan	Unet+Transformer	U-BARN [20]	✓	✓		
2024 Jan	Autoregressive Transformer	EarthPT [58]	✓			
2024 Jan	Teacher-Student Network	GeRSP [29]	✓	✓	✓	
2024 Jan	Dual-Branch	SwiMDiff [65]	✓			✓
2024 Jan	Generative ConvNet	SMLFR [17]		✓	✓	
2024 Feb	3D GPT	SpectralGPT [28]	✓	✓		✓
2024 Feb	MAE-based Framework	Presto [66]		✓		
2024 Mar	SatMAE	SatMAE++ [52]	✓			
2024 Mar	Joint-Embedding Predictive Architecture	SAR-JEPA [41]	✓			
2024 Mar	ViT	FoMo-Bench [5]	✓	✓	✓	
2024 Mar	Factorized Multi-Modal Spatiotemporal Encoder	SkySense [24]	✓	✓	✓	✓
2024 Mar	Multi-Modules	UPetu [18]	✓	✓		✓
2024 Apr	Swim Transformer	msGFM [25]	✓	✓		✓
2024 Apr	DINO	DINO-MC [77]	✓			✓
2024 May	OFA-Net	OFA-Net [84]	✓	✓		
2024 May	Shared Encoder, Task-Specific Decoders	MTP [70]	✓	✓	✓	✓
2024 May	ViT	BFM [7]		✓	✓	
2024 May	MP-MAE	MMEarth [51]	✓			
2024 May	ViT	CxMIM [86]	✓	✓	✓	
2024 May	HIViT	SARATR-X [37]	✓		✓	
2024 Jun	Dynamic OFA	DOFA [83]	✓			

## Foundation models: specific models for remote sensing data

Foundation models refer to **large-scale**, **pretrained models** that provide a **robust starting point** for **various downstream tasks** across different domains. These models leverage extensive datasets and advanced architectures, enabling them to capture complex patterns and features that can be fine-tuned for specific applications with minimal additional training.

## Foundation models: specific models for remote sensing data

Foundation models face the usual challenges:

- Need for high-quality and diverse training data
- Need for significant computational resources
- Need for effective domain adaptation for specific applications



## Foundation models: Methodologies

### Self-supervised learning training strategies:

The model learns part of the input data from other parts of the input data  
-> reduces the need for large ***labeled*** datasets.

- Contrastive learning:  
Learns representations by comparing different views (data augmentation) of the same data point
- Predictive coding:  
Learns to predict missing (masked) parts of the data from the observed parts

SSL methods: MoCo, SimCLR, BYOL, DINO...

# SPECTRAL EARTH: TRAINING HYPERSPECTRAL FOUNDATION MODELS AT SCALE

Nassim Ait Ali Braham, Conrad Albrecht, Julien Mairal, Jocelyn Chanussot, Yi Want, Xiao Xiang Zhu

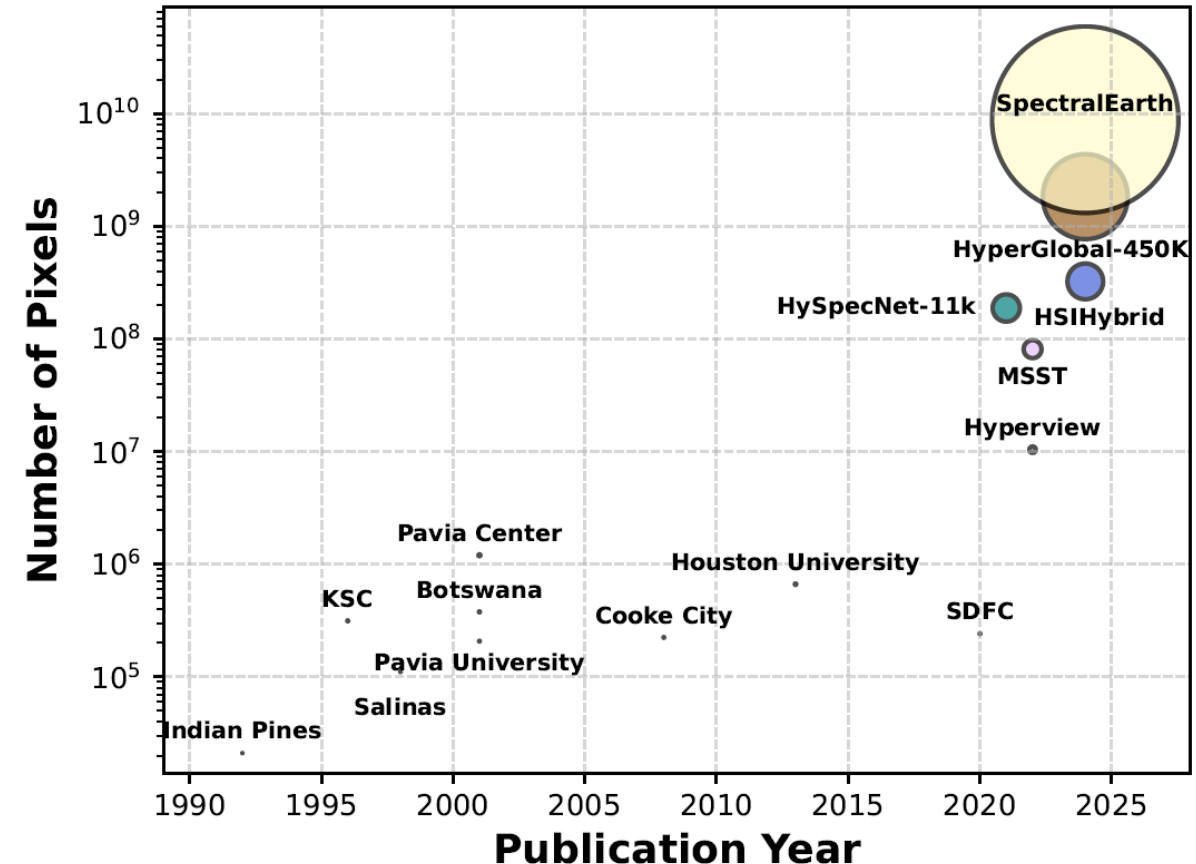
EO Data Science, Remote Sensing Technology Institute, DLR

Data Science in Earth Observation, Technical University of Munich, Germany

Univ. Grenoble Alpes, Inria, CNRS, Grenoble INP, LJK, 38000 Grenoble, France

# Motivation

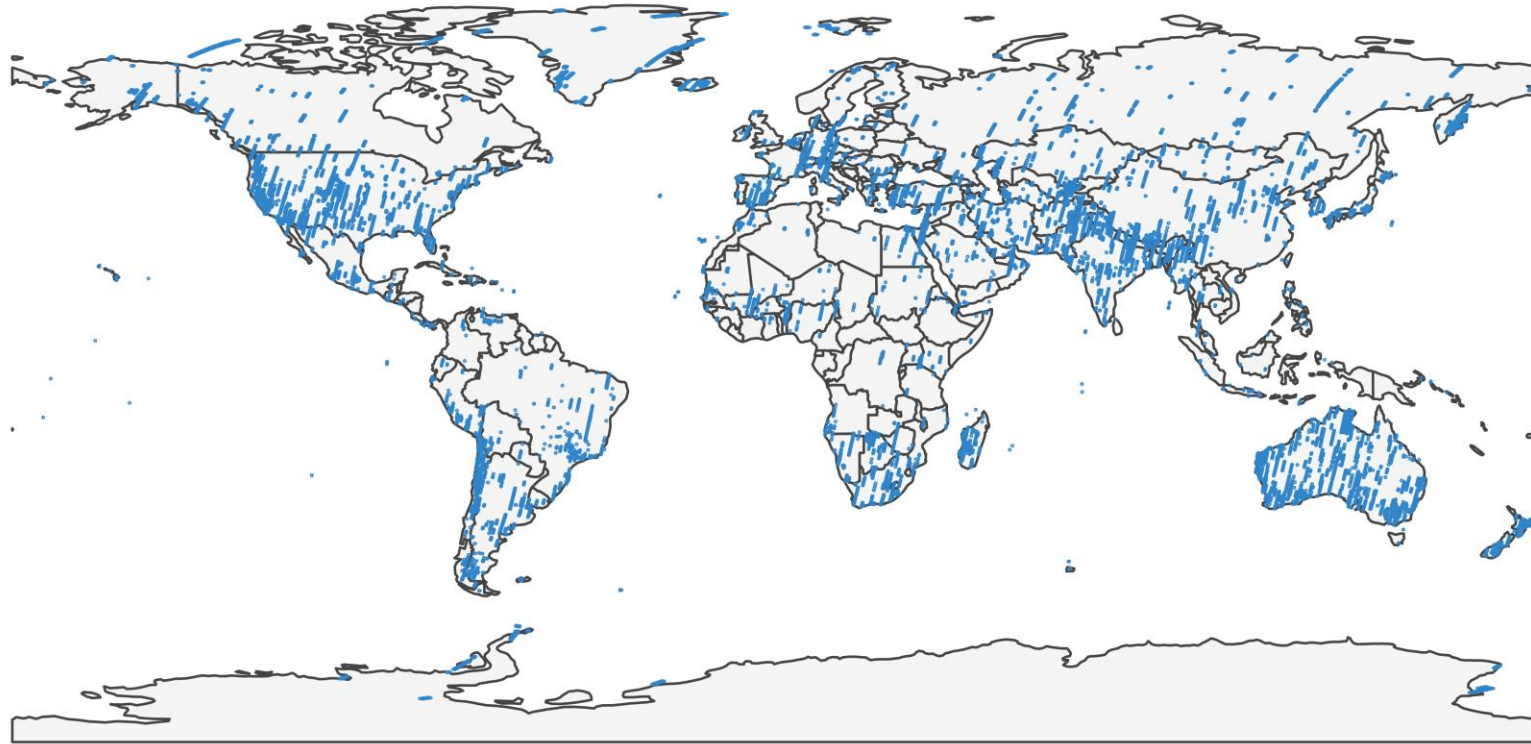
- A lot of research on foundation models for **MSI**: SatMAE, ScaleMAE, Prithvi, DOFA, SkySense, etc.
- Less research on foundation models in **HSI**
- **No suitable dataset for pre-training hyperspectral foundation models**
- **Contribution: SpectralEarth** a globally distributed dataset, pre-trained models and benchmark



<https://doi.org/10.48550/arXiv.2408.08447>

# SpectralEarth: A large-scale HSI dataset

- Based on *EnMAP* imagery
- 30m resolution, 202 bands
- ~**538,974** patches, 128x128 pixels.
  - ~**415,153** unique locations
  - ~**73,000** locations with > 1 timestamp
  - Sampled from **11,636** tiles
- ~**3.3 TB** of data
- Mostly cloud free



*Geographical distribution of SpectralEarth*



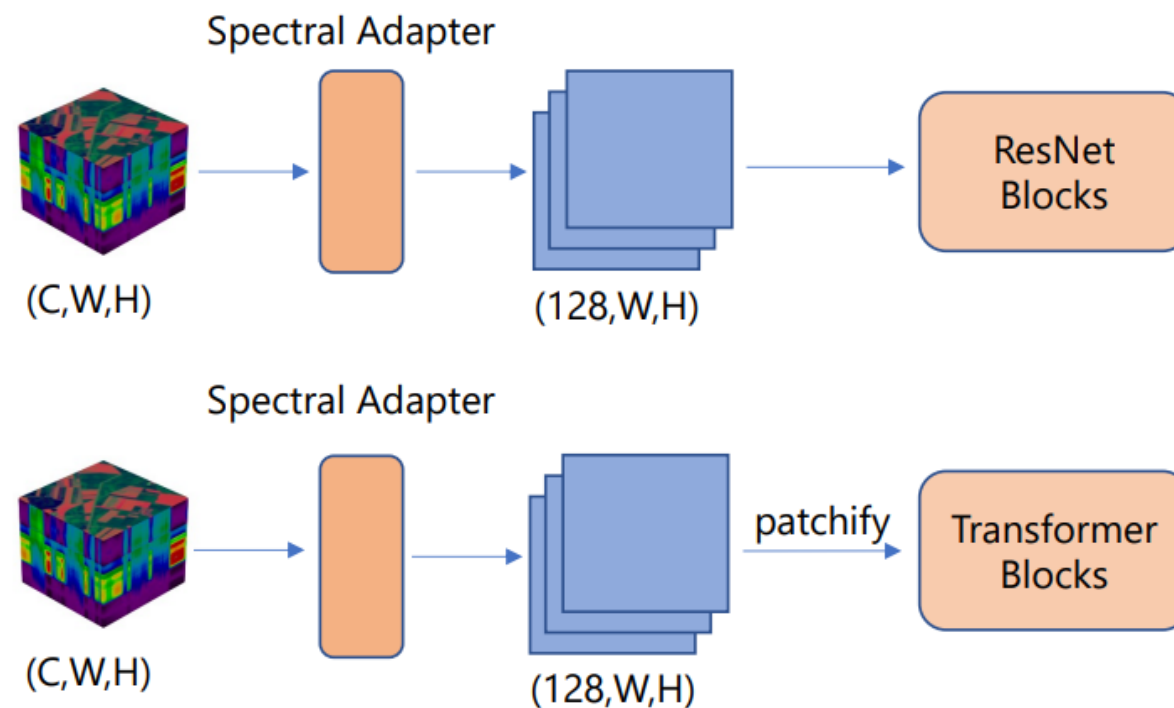
# Models

- **Network Architectures**

- Simple variation of classical CNN and Vision Transformer architectures
- 1D convolutions to extract spectral features
- Models ranging from **22M** to **1.1B** parameters

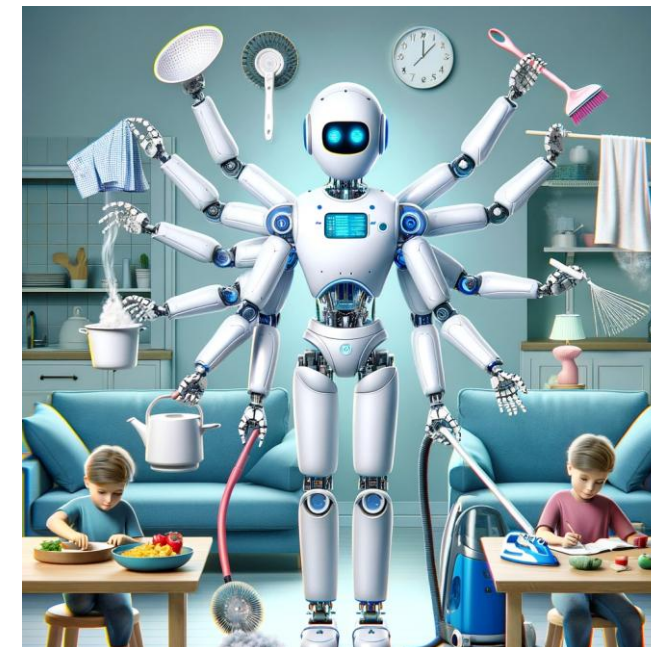
- **3 SSL Algorithms**

- **> 10** pre-trained models



*Backbone architectures*

# AI and Remote Sensing





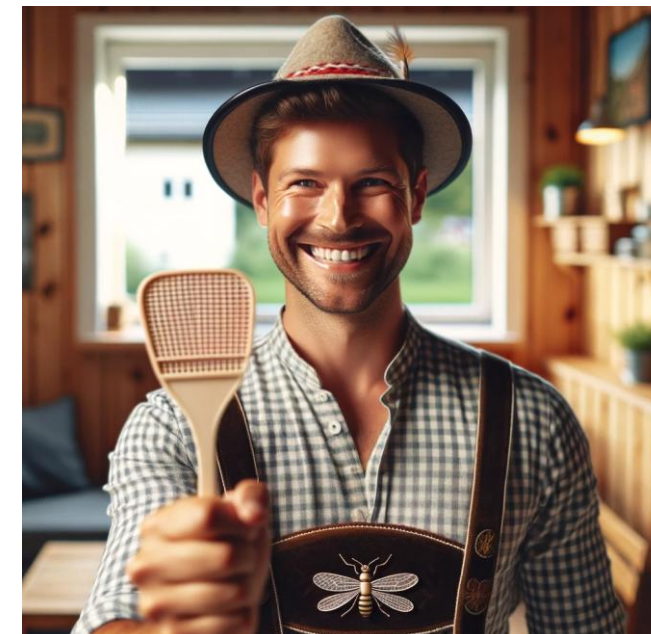
# AI and Remote Sensing



# AI and Remote Sensing

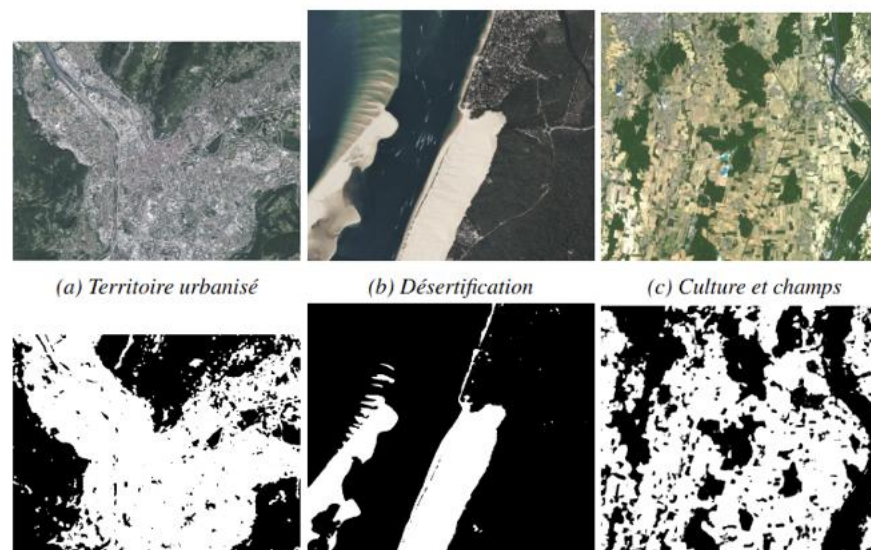


# AI and Remote Sensing





# AI and Remote Sensing



# AI and Remote Sensing

Next challenges ?

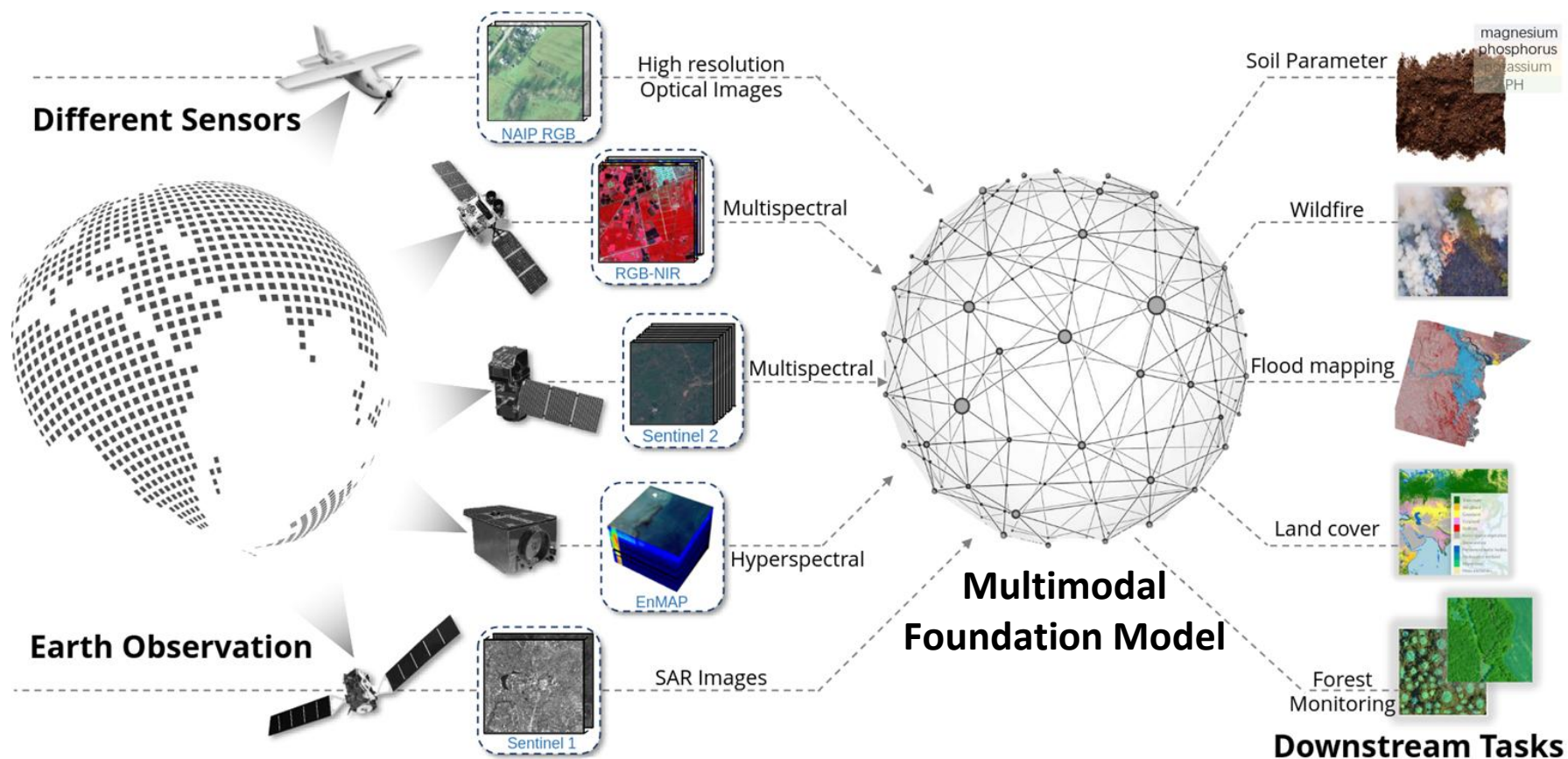


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Remote Sensing Team

# AI and Remote Sensing

## Future directions: multimodal data integration



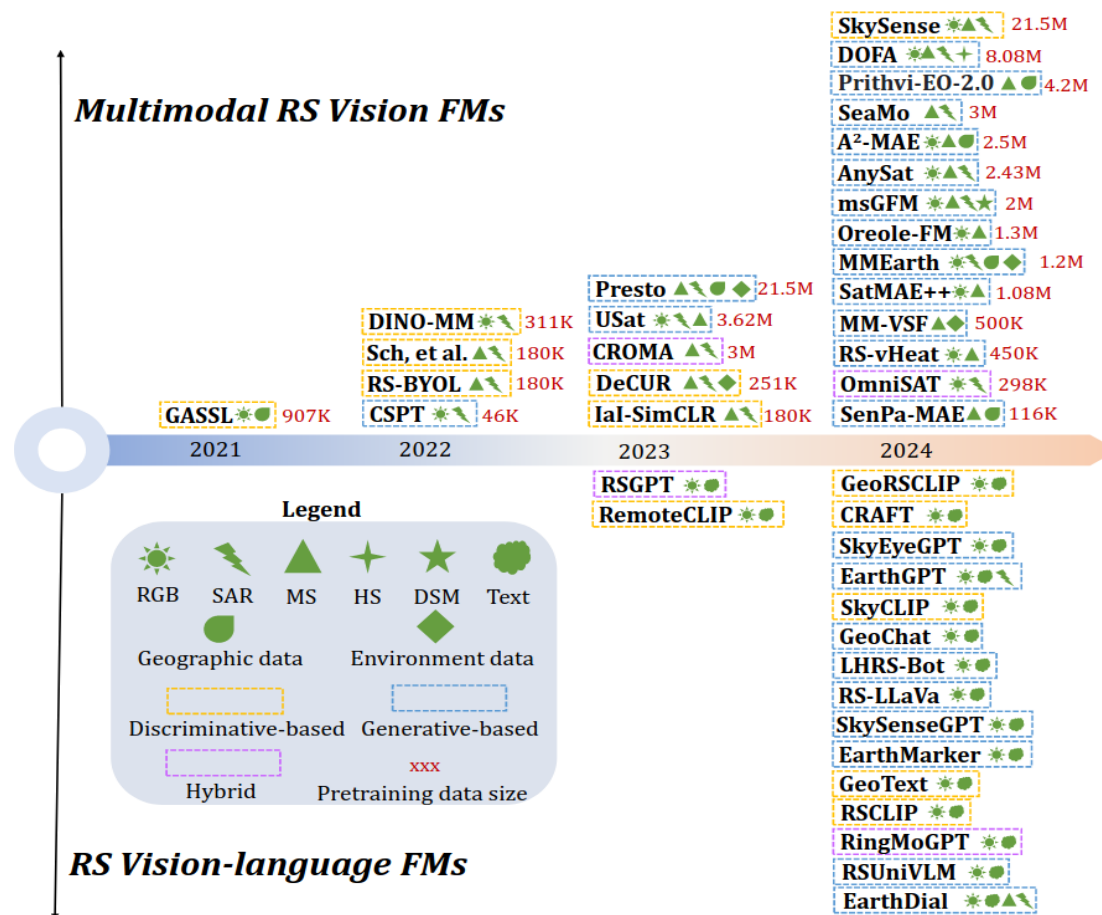




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# AI and Remote Sensing

## Future directions: multimodal data integration



## Conclusion and next challenges:

Multi-scale / multi-modal / multi-temporal integration

Foundation models

Lightweight / frugal AI

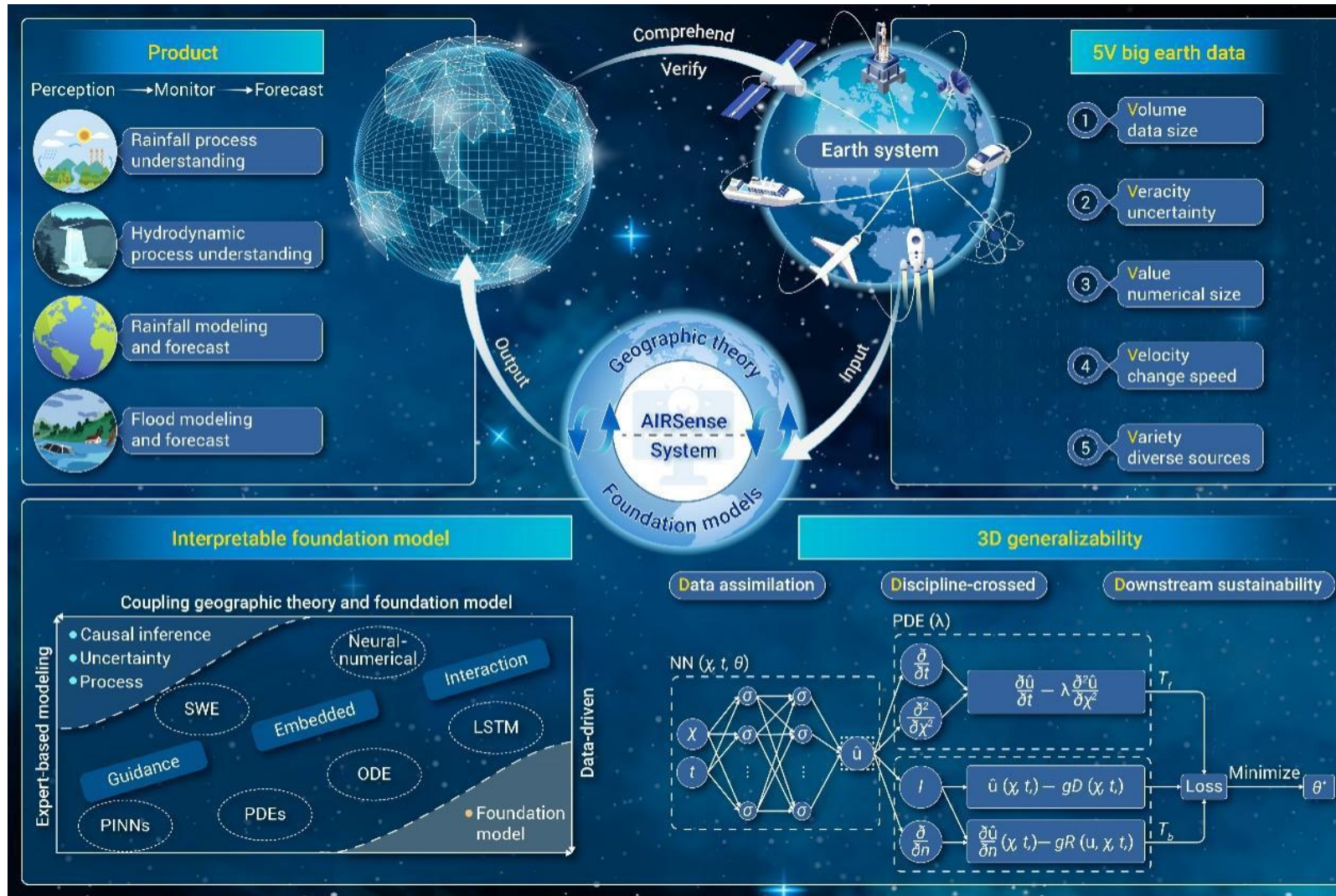
Explainable AI

Physical modeling



**Thank you for your attention.**

# Future directions: interpretable foundation models



Discovering *what* interactions drive the model's predictions?

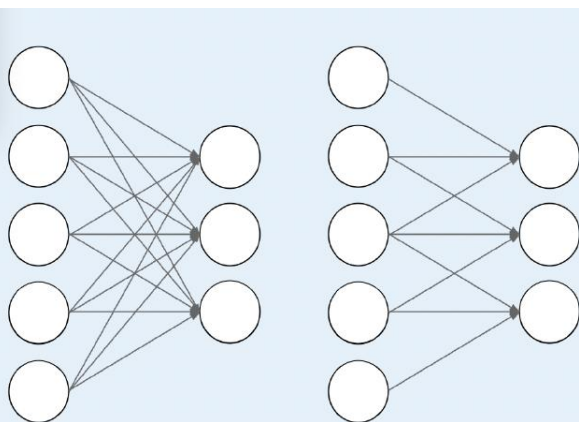
Verifying *why* certain features are instrumental in driving model's decision-making process?

Assessing *how* the effectiveness of decisions is validated by real-world data?

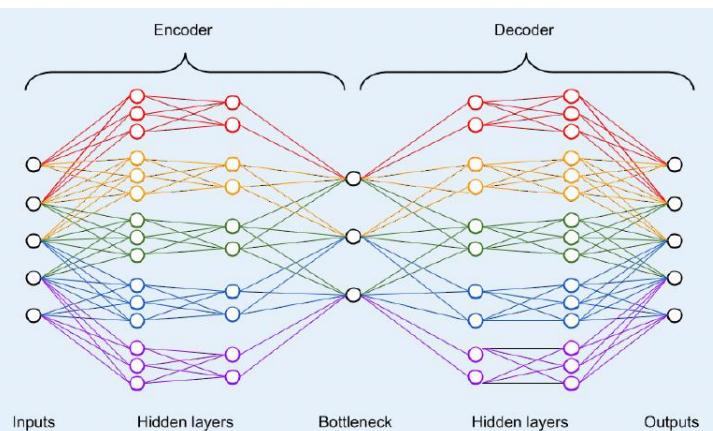


### (iii) areas of research

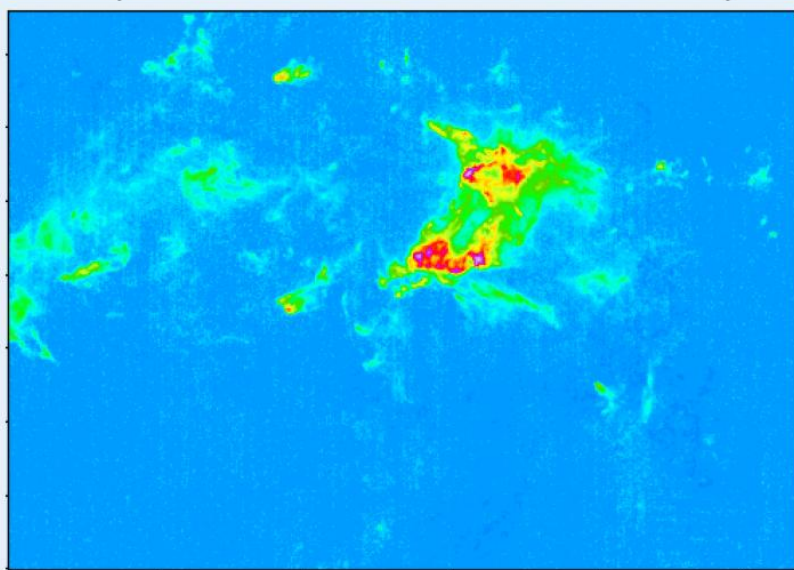
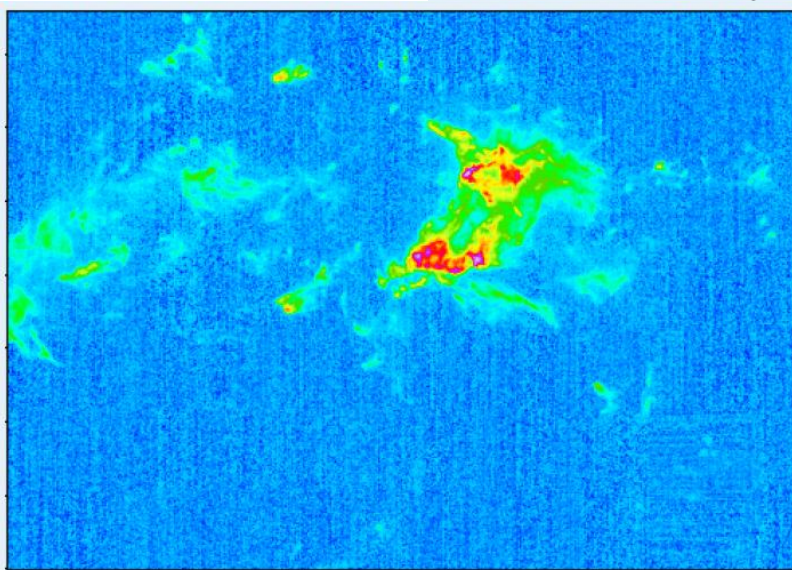
#### Applications in radioastronomy (A&A)



Dense and locally dense layers



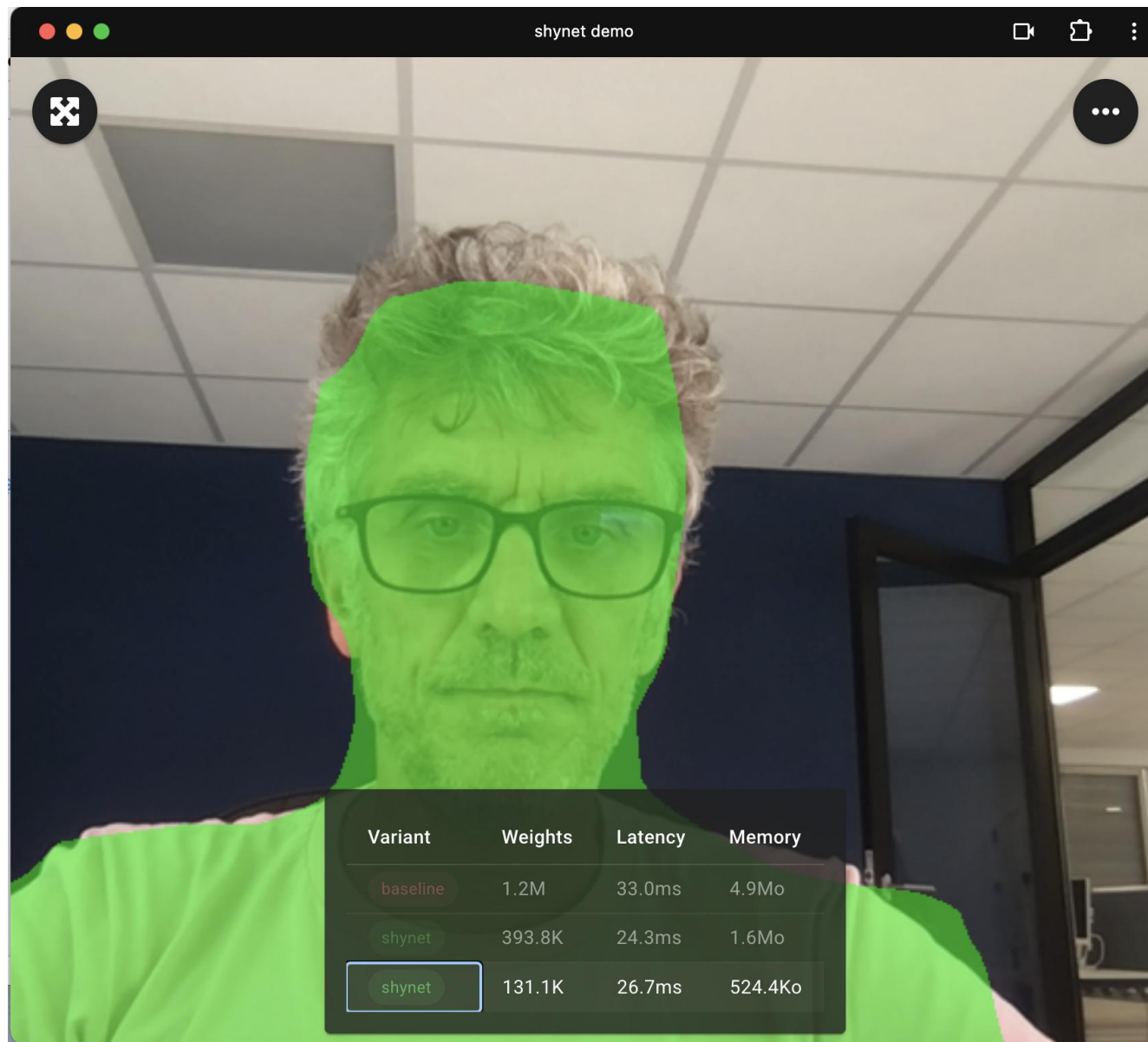
Example of locally dense autoencoder





### (iii) areas of research

shynet demo



Variant	Weights	Latency	Memory
baseline	1.2M	33.0ms	4.9Mo
shynet	393.8K	24.3ms	1.6Mo
shynet	131.1K	26.7ms	524.4Ko



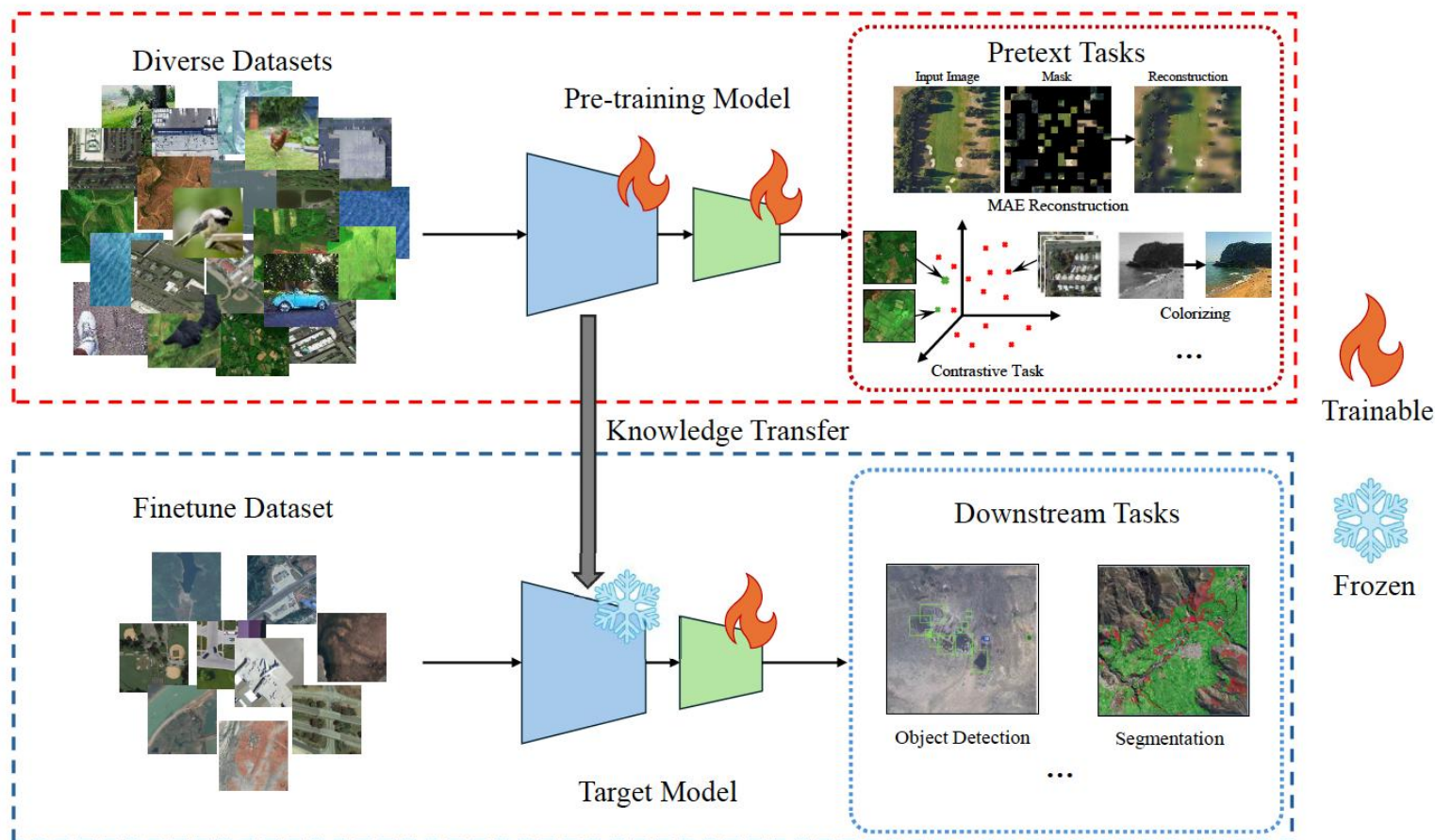
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# AI and remote sensing

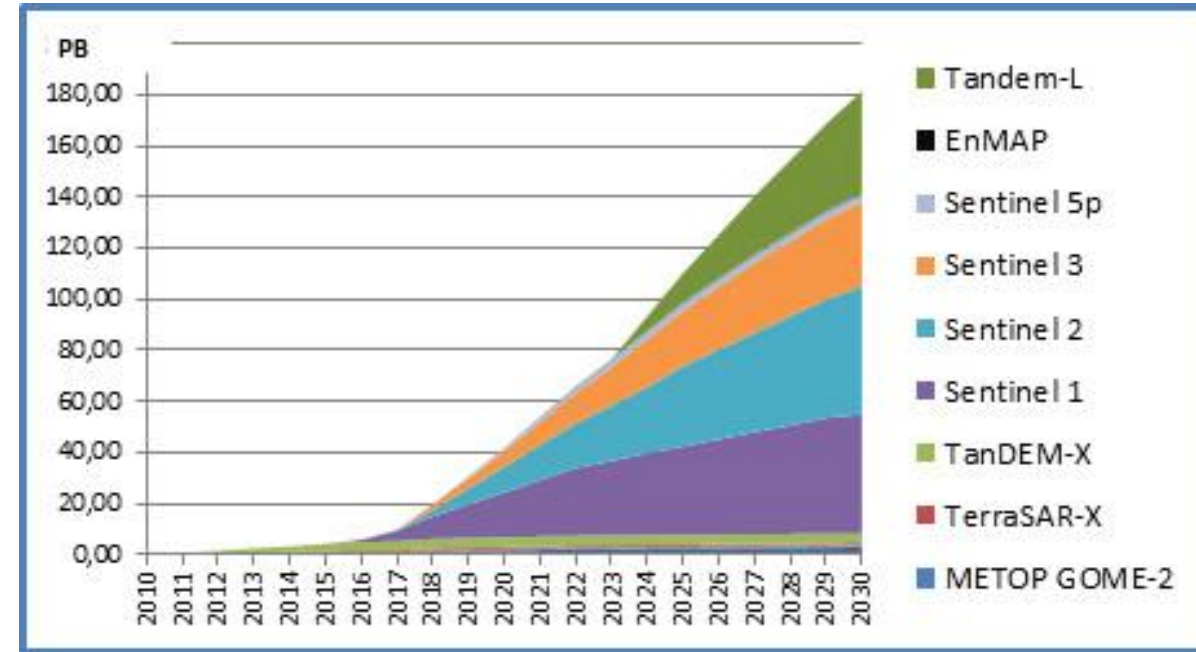
## Foundation models: Methodologies

### Self-supervised learning training strategies:



# Labeled vs. Unlabeled Data in EO

- Deep Learning requires annotated data
- **Labeled data is rare**
  - Costly to obtain
  - Tedious annotation process
- **Unlabeled data is abundant**
  - Satellite archives with Petabytes of data



How to exploit unlabeled data for deep learning  
with hyperspectral imagery?



Self-Supervised Learning

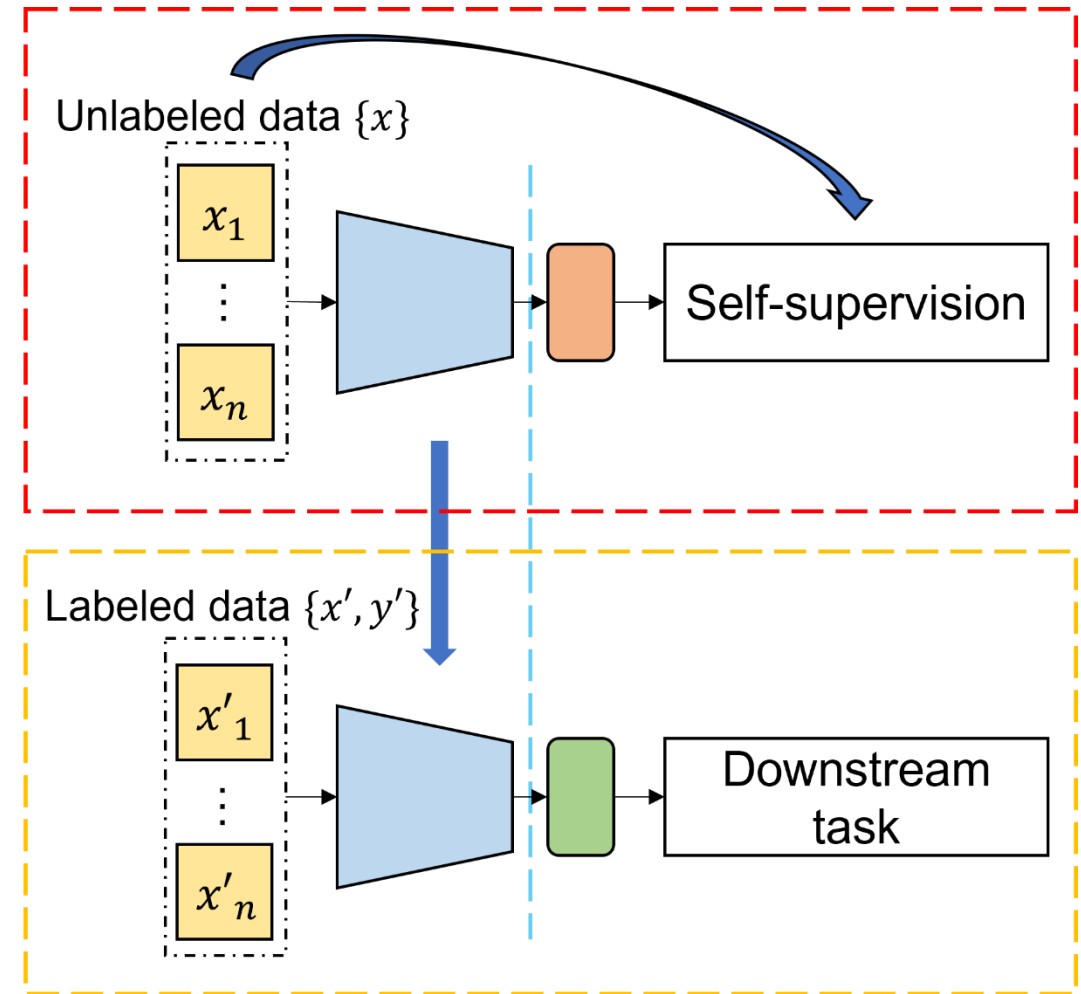
# Self-Supervised Learning

- **Goal**

- Obtain *training feedback* from the data itself
- Learn representations in a self-supervised fashion
  - no human annotation

- **Why?**

- A pre-trained model can be transferred to downstream tasks
- Improve **accuracy** and **label efficiency**



Overview of Self Supervised Learning\*

\*Wang, Yi, et al. "SSL4EO-S12: A large-scale multimodal, multitemporal dataset for self-supervised learning in Earth observation [Software and Data Sets]." IEEE Geoscience and Remote Sensing Magazine 11.3 (2023): 98-106.



















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## RS in Grenoble

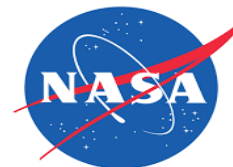
### 2023 Global Ranking of Academic Subjects

Remote Sensing 100 Institutions SHANGHAI RANKING			
World Rank	Institution	Country/Region	
1	 Wuhan University		
2	 Xidian University		
3	 University of Maryland, College Park		
4	 California Institute of Technology		
5	 Beijing Normal University		
6	 Technical University of Munich		
7	 University of Electronic Science and Technology of China		
8	 Université Grenoble Alpes		

<https://www.shanghairanking.com/>

East Coast (Goddard)

West Coast (JPL)



Inria



# RS in Grenoble

## 2024 Global Ranking of Academic Subjects

1



Wuhan University




2




Xidian University




3




Technical University of Munich




4




University of Electronic Science and Technology of China



5



Northwestern Polytechnical University



6



Beijing Normal University




7



Sun Yat-sen University




8




China University of Geosciences




9




University of Maryland, College Park



10



California Institute of Technology



11



Université Grenoble Alpes



















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## RS in Grenoble

### 2025 Global Ranking of Academic Subjects

1	Wuhan University China	
2	Xidian University China	
3	China University of Geosciences China	
4	Northwestern Polytechnical University China	
5	University of Electronic Science and Technology of China China	
6	Sun Yat-sen University China	
7	Technical University of Munich Germany	
8	Beijing Normal University China	

9	Peking University China	
10	Tongji University China	
11	The University of Hong Kong Hong Kong, China	
12	University of Extremadura Spain	
13	Beijing Institute of Technology China	
14	Université Grenoble Alpes France	



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# LA TRANSFORMATION IA DE L'IGN

**NICOLAS GONTHIER (DSI/SIMV)**  
*CHEF DE PROJET INNOVATION*

# Suivi d'occupation des sols par IA



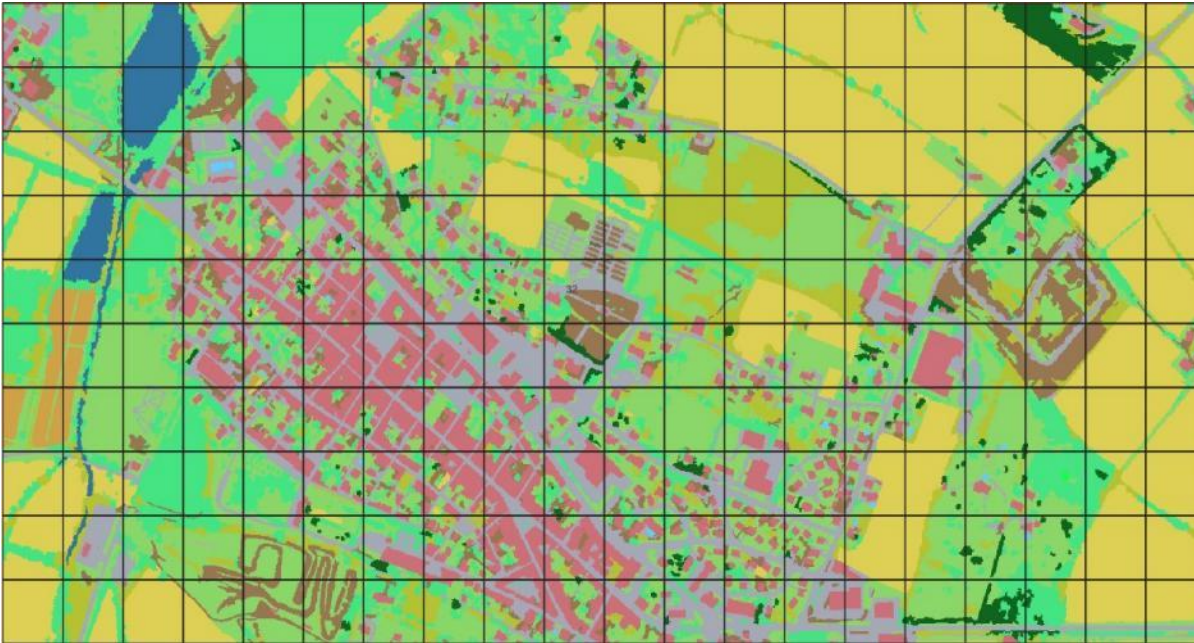
16 classes de  
Couverture du Sol

- Bâtiment
- Zone imperméable
- Zone perméable
- Piscine
- Serre
- Sol nu
- Surface eau
- Neige
- Conifère
- Feuillu
- Coupe
- Broussaille
- Pelouse
- Culture
- Terre labourée
- Vigne
- Autre

- Systèmes IA alliés de choix pour la cartographie
- Valeur concentrée sur le ***passage à l'échelle***



# Production de données d'apprentissage



~ 2500 km<sup>2</sup> d'annotations

=> Applicable sur le territoire national > 500 000 km<sup>2</sup>

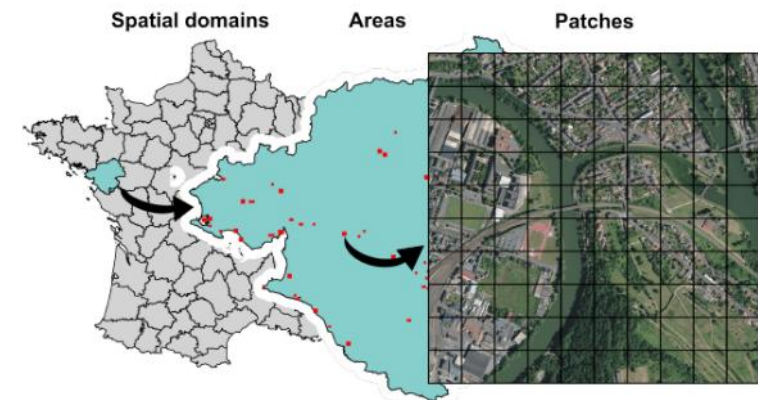
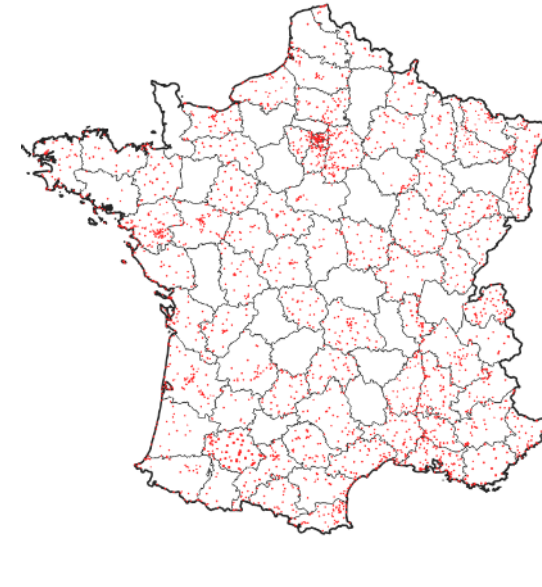


Fig. 2: Spatial domains, areas and patches.



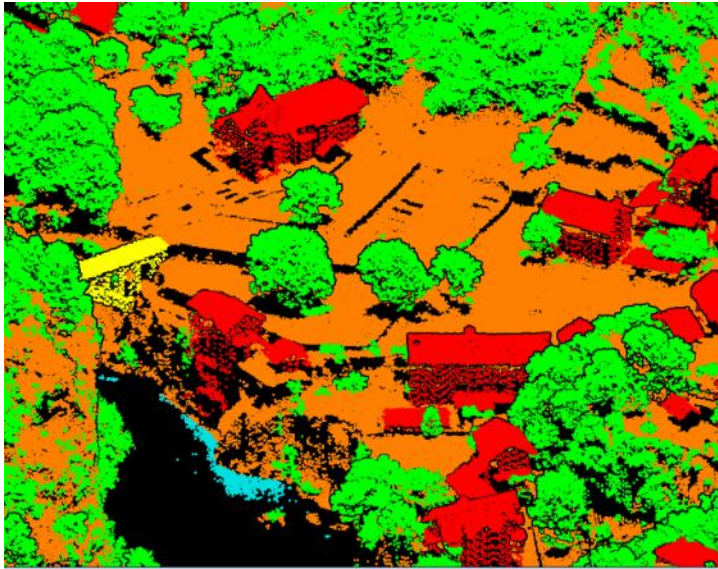
# Usages de l'occupation des sols



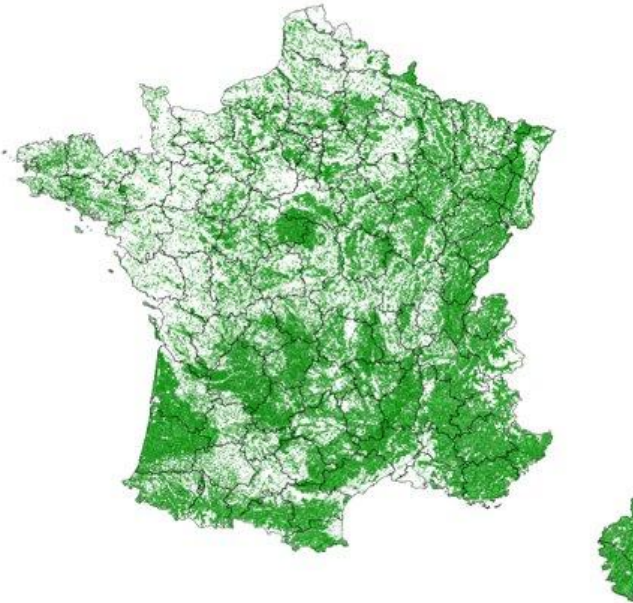


# Généralité de la démarche

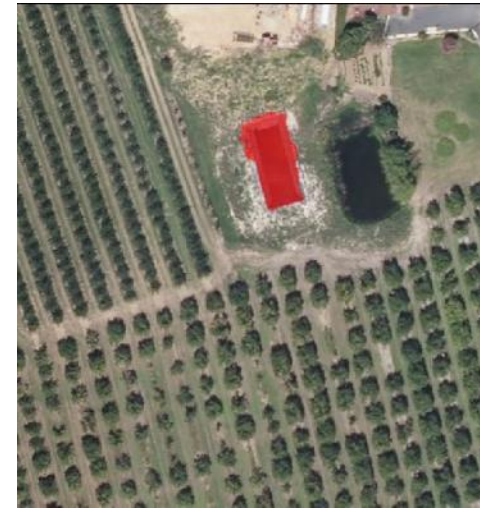
LIDAR HD



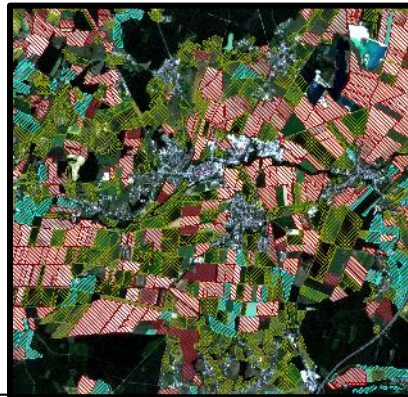
Forêt



Détection de changements

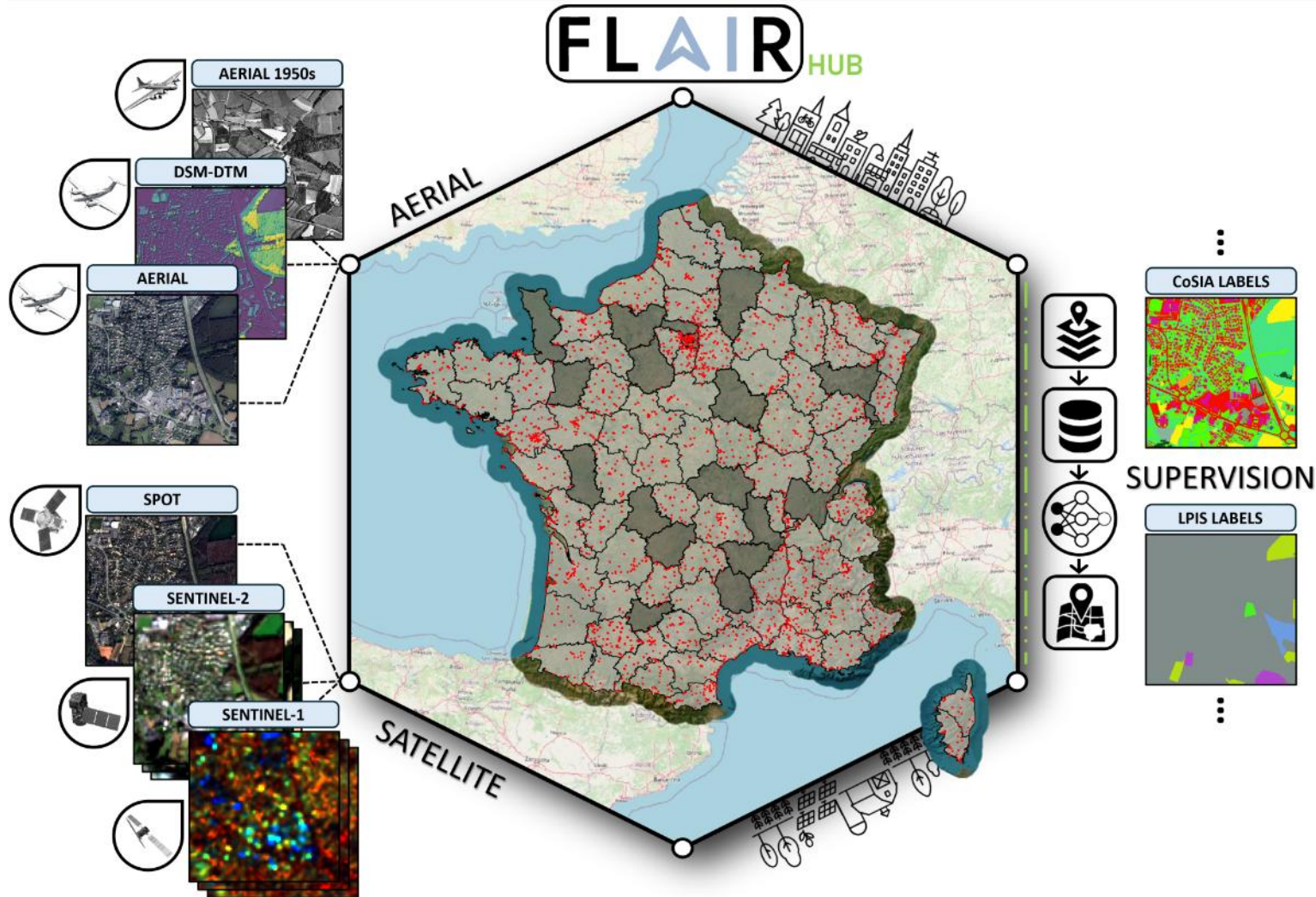


Cartographie  
agricole



Richesse technique liée à la  
diversité des thématiques et  
des structures de données





## ANIMER DES ÉCOSYSTÈMES IA

FLAIR : défis scientifiques et techniques, ouverture données (apprentissage & évaluation) & modèles.

Peu de grands jeux de données massifs, diversifiés, de qualité. L'un des plus grands jeux de données OCS THR ouverts.

2500 km<sup>2</sup>, 63 Md pixels, 19 classes d'occupation des sols et 23 de type de culture

Amélioration des productions IGN

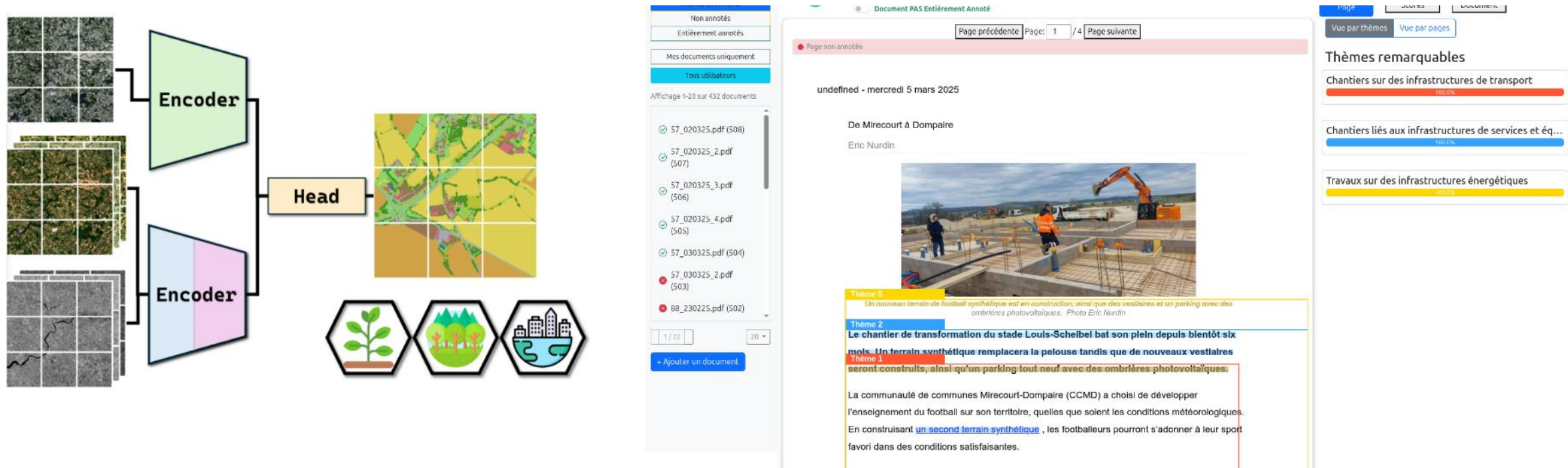
Réutilisations académiques & industrielles

Modèles pré-entraînés : réduction de coûts et de barrières à l'entrée

Atténuer la concentration dans le secteur IA

# Se maintenir à l'état de l'art et être proactif

- Modèles de Fondation multimodaux pour les données de télédétection
- Grands Modèles de Langue
- Feuille de route technologique sur la détection de changements



- Maintenir dialogue entre recherche et équipes opérationnels
- Faire communauté pour arriver à une veille pertinente



## Ressources humaines & organisation

- Développement expertise IA. 30 experts IA + Recherche.
- **Compétition forte** pour les talents IA. Atouts : **sens des missions**; open source ; crédibilité scientifique et technique ; **contributions à grande échelle**.
- Importance d'équipe **pluridisciplinaires** : itérations fréquentes **nécessitant experts IA + métiers + géomaticiens + photo-interprète + dev**
- **Projets exigeants** : intenses en ressources humaines et de calcul
- Contraintes projets innovation + co-dépendance données/modèle/code.

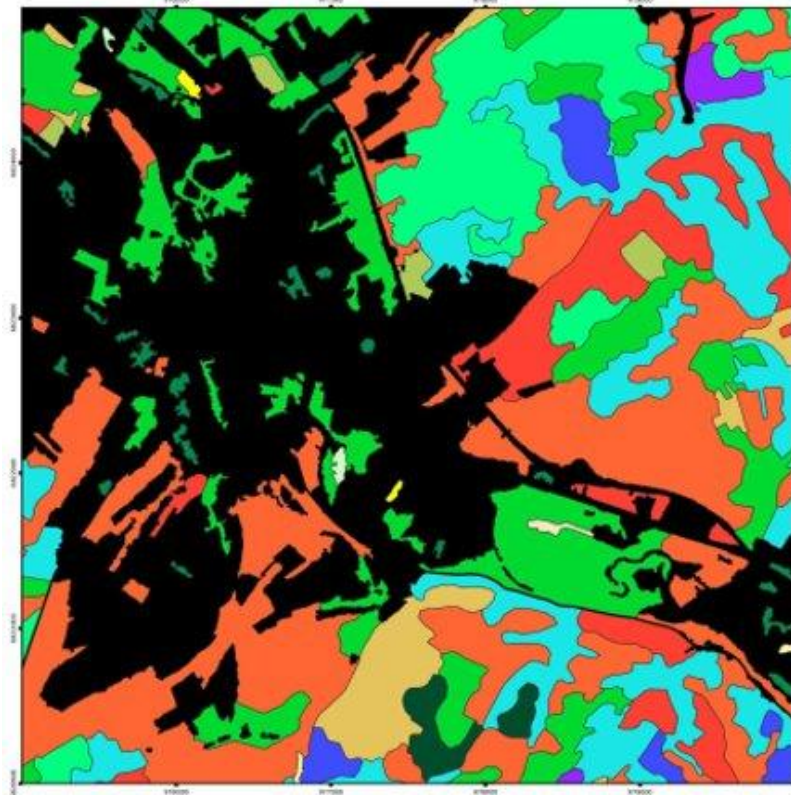
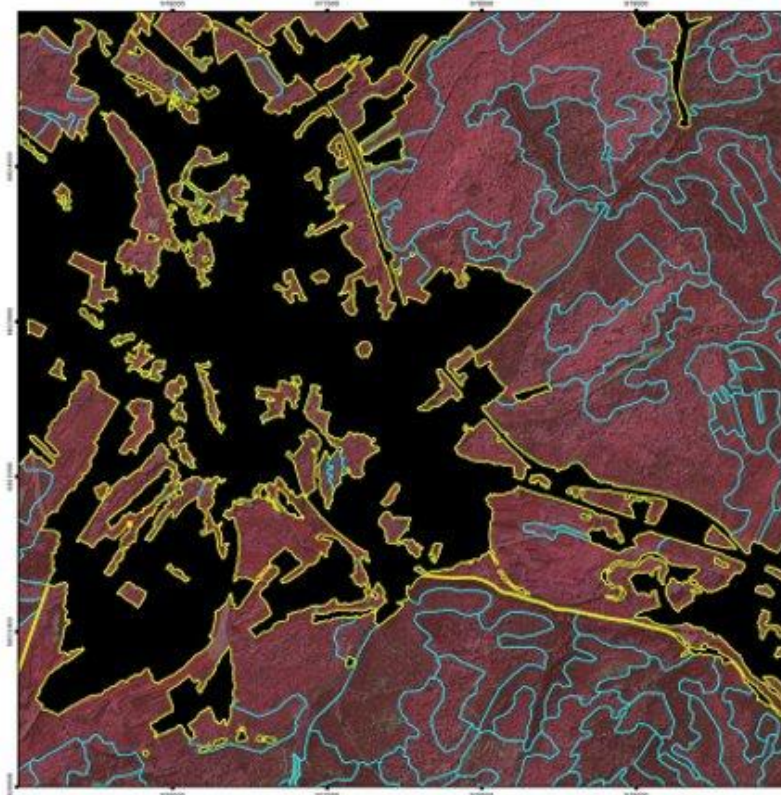
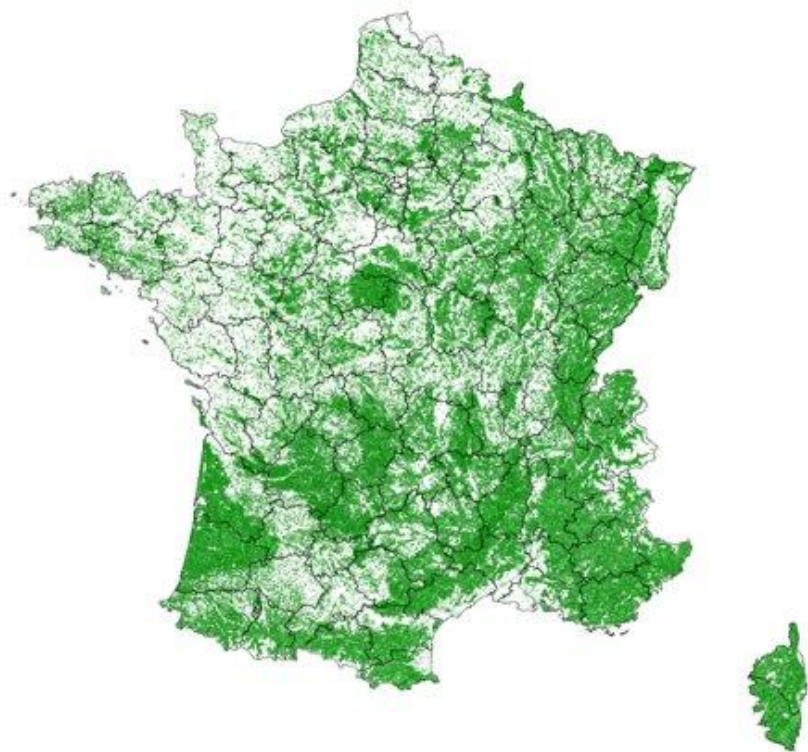




# Appui dans la sphère publique

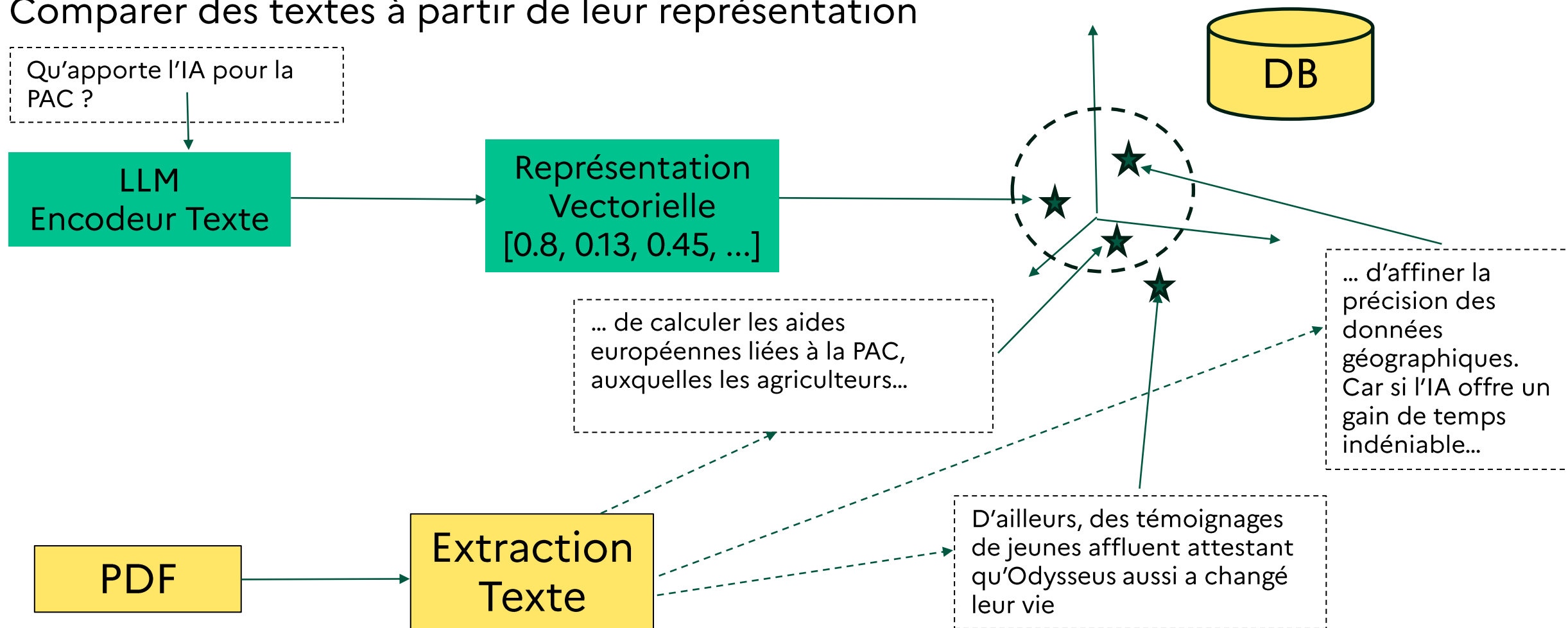
- Plusieurs années d'expérience sur l'IA / apprentissage profond (à partir de 2018)
  - Recul sur une dizaine de projets industrialisés
  - Démarche structurée tôt : feuille de route dédiée à partir de 2021
  - Nombreuses conditions pour le succès des projets IA → **Nécessite l'alignement de plusieurs unités** et implication encadrement supérieur
  - Pas de *quick win* : gains réels, mais issus d'efforts soutenus, **Approche empirique & pragmatique.**
- **Masse critique** : permet présence sur plusieurs thématiques + sur tout le spectre recherche <-> production

# Cartographie forestière – Segmentation d'essences à partir d'images aériennes – BD Forêt V3



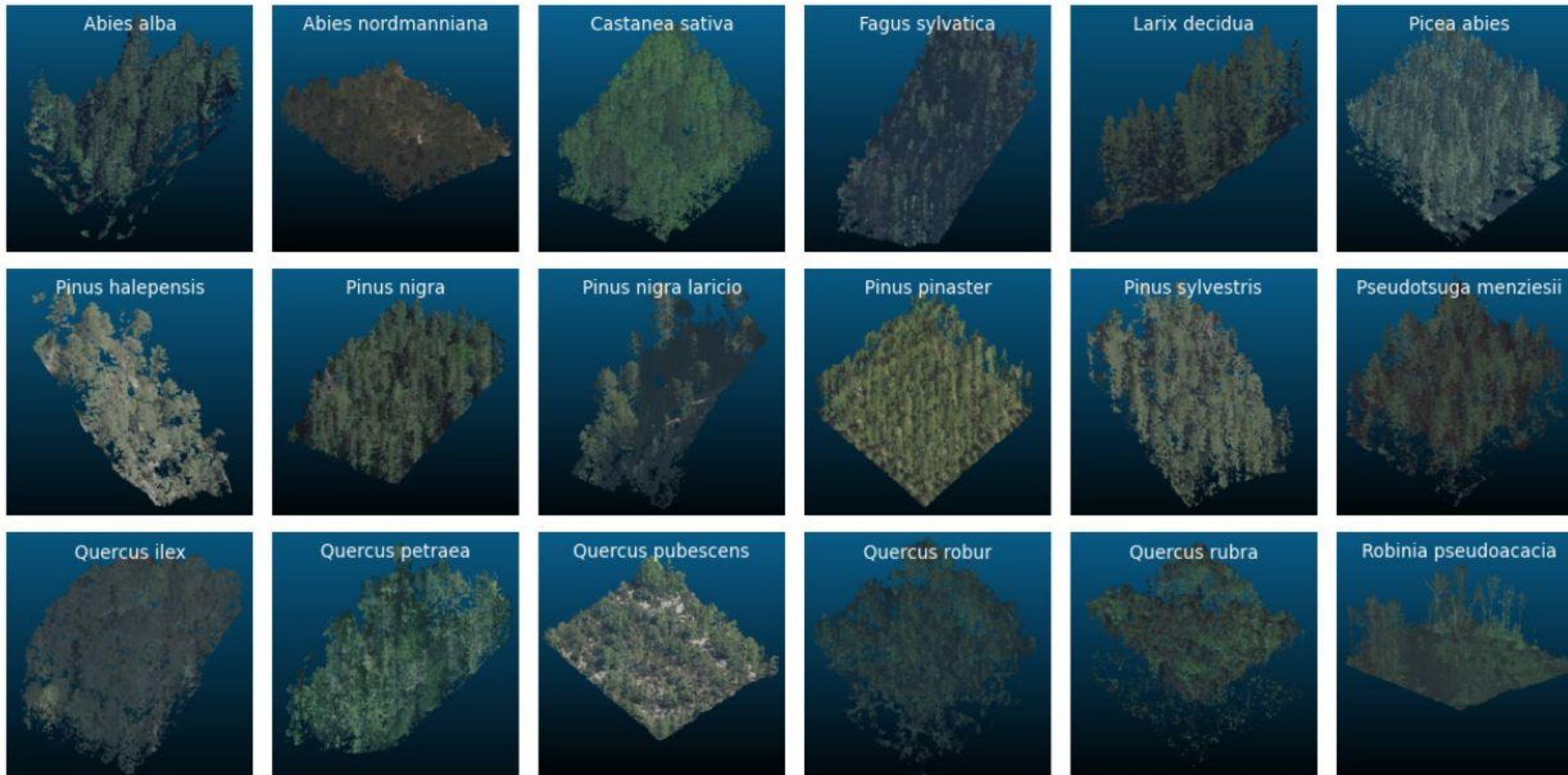
# Extraction d'informations textuelles avec des LLM

Comparer des textes à partir de leur représentation





# Classification de nuages de points de forêt



Plus grand dataset de  
classification de nuages  
de points :

LIDAR ALS 40 pts/m<sup>2</sup>

18 essences

33900 ha

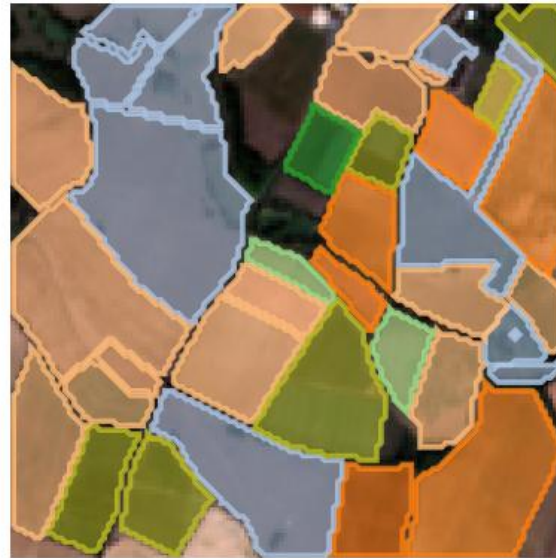
23.7M .Arbres

[Gaydon WACV 2024]

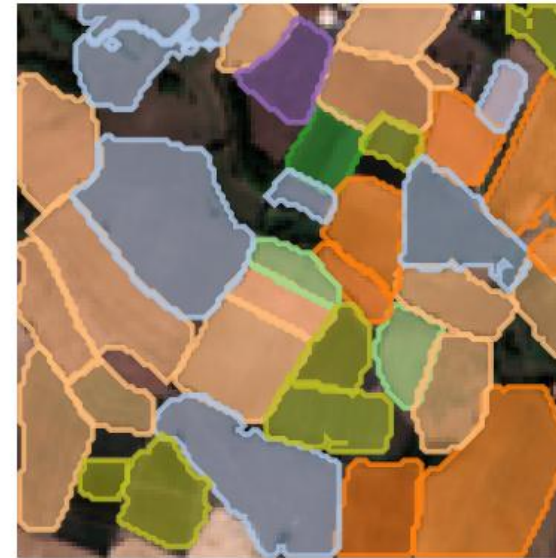
# Agriculture & analyse de séries temporelles optiques



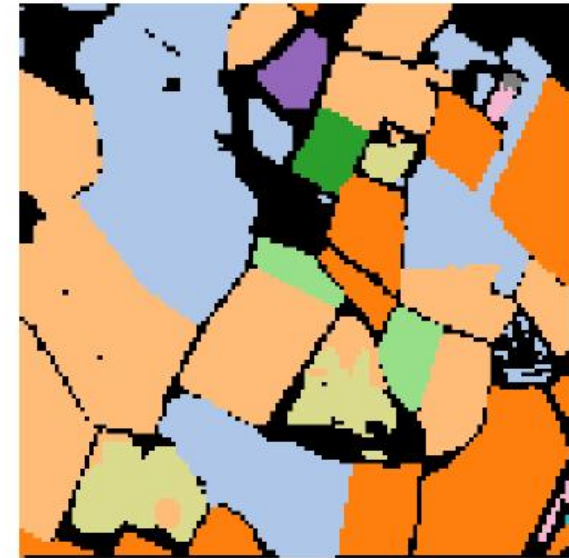
(a) Image from the sequence.



(b) Panoptic annotation.



(c) Panoptic segmentation.

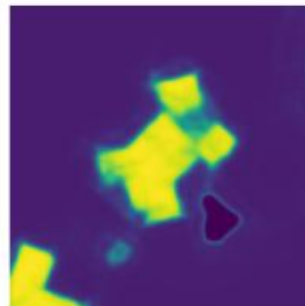


(d) Semantic segmentation.

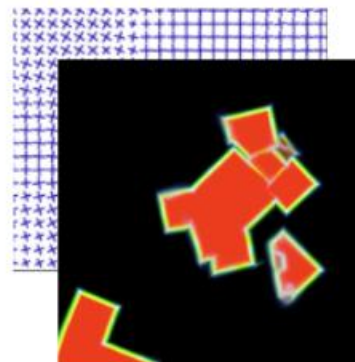


# Détection et vectorisation du bati Frame Field learning

DeepLabv3+



Framefield



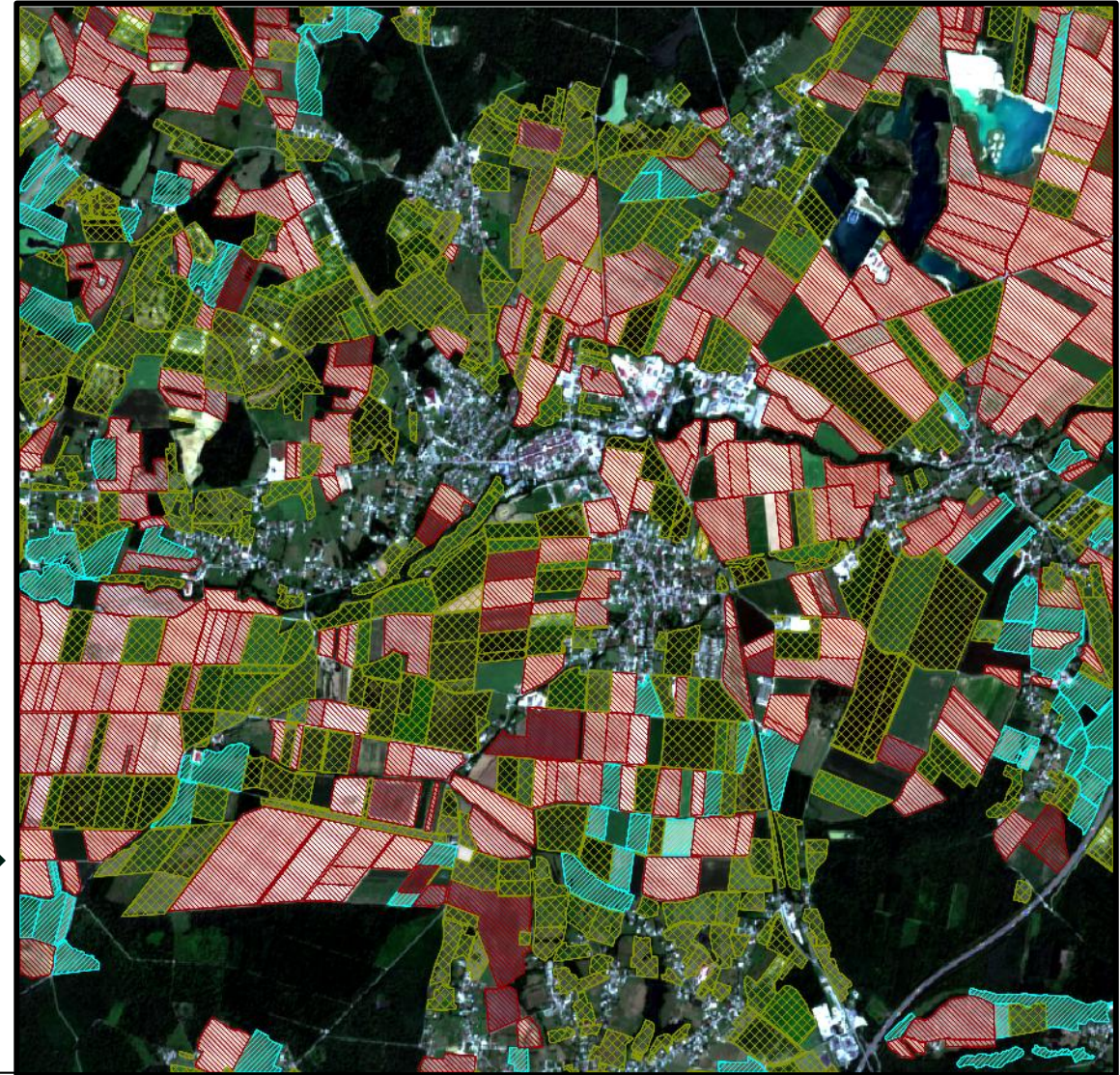
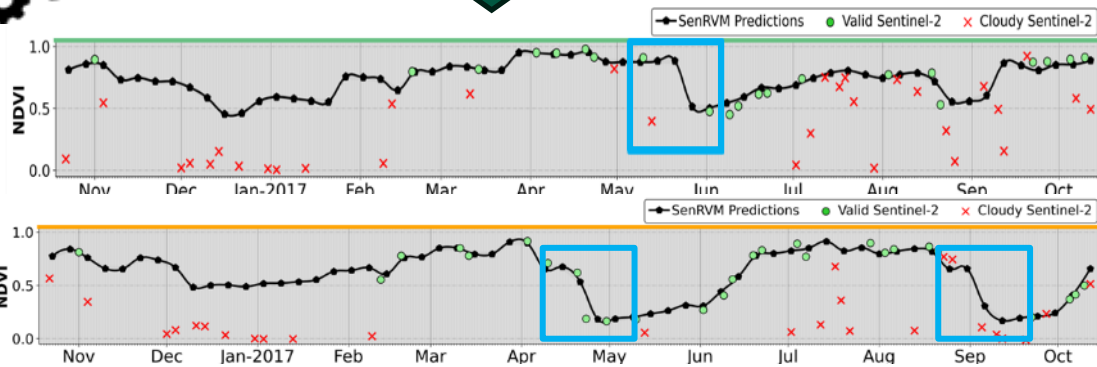


# Pratiques agricoles

Récolte, fauche, pâture



DL regression modèle





# Comment être générique entre capteurs ?

Les modèles actuels ne sont pas assez génériques, c'est pourquoi il est nécessaire de les fine-tuner sur des plus petits datasets cibles



Model IA IGN ORTHO (RVBI) -->  
IGN ORTHO (Aerial Imagery  
20cm RGB+IR)



Model IA Aerial IGN ORTHO  
(RGBI) --> PNEO (Satellite Imagery  
30 cm RGB+IR)



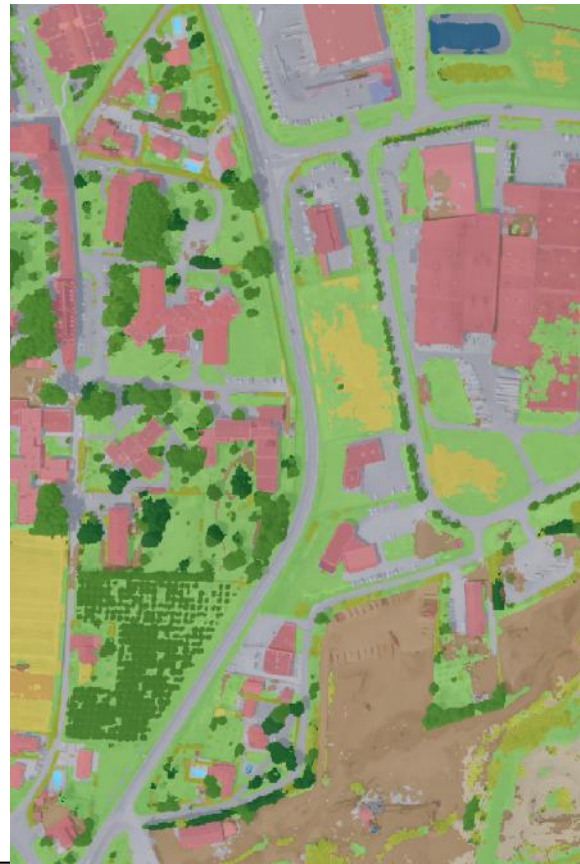


# INNOVATION & AMÉLIORATION DES MODÈLES

Orthoimage RGB



U-Net-like, conv.



Vision Transformers





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# MERCI POUR VOTRE ATTENTION





**MINISTÈRES  
TRANSITION ÉCOLOGIQUE  
AMÉNAGEMENT DU TERRITOIRE  
TRANSPORTS  
VILLE ET LOGEMENT**

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**Commissariat Général au  
Développement Durable**



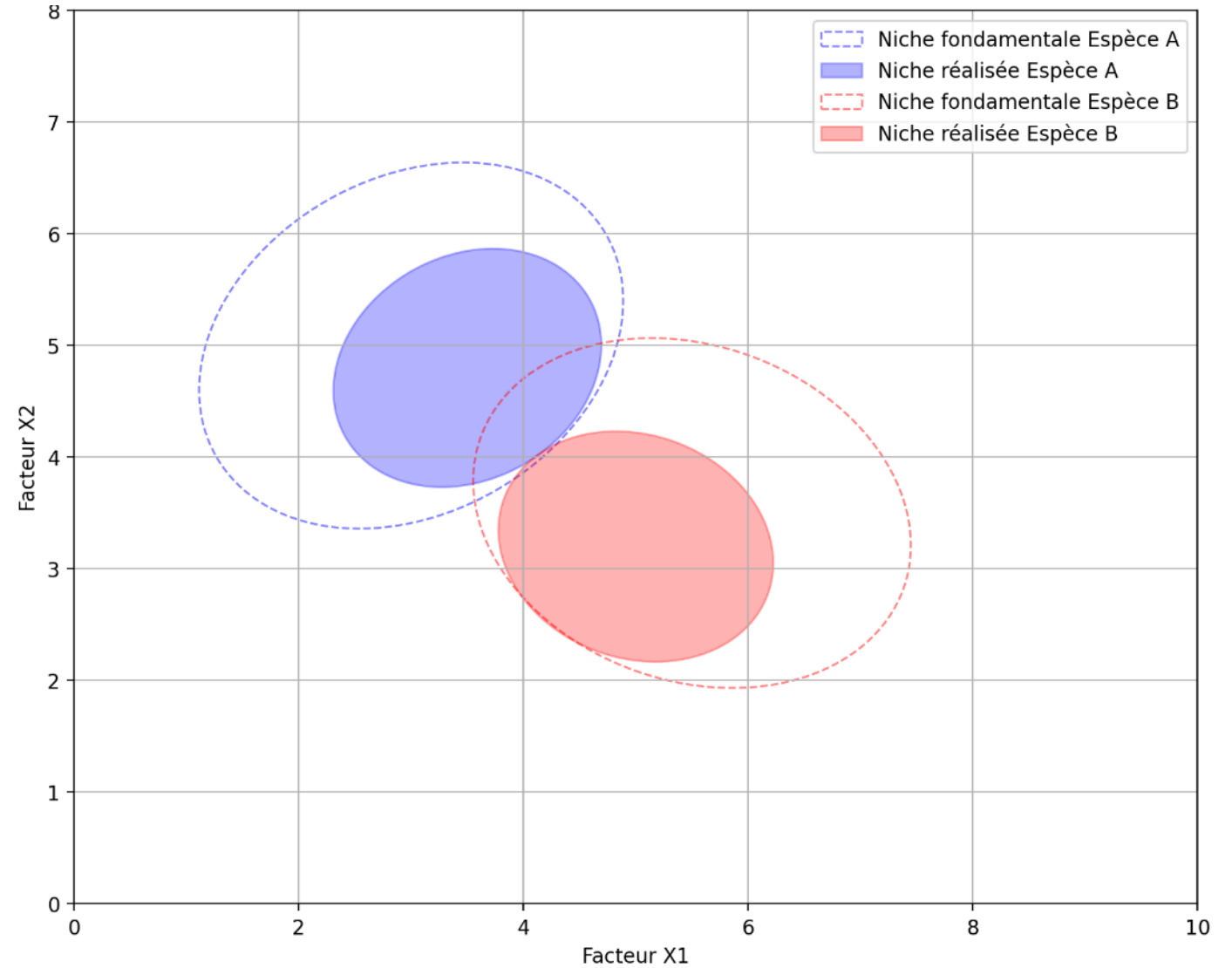
# IA métiers et production de données

**Matthieu PORTE (Ecolab du CGDD)**

# ~ 10 ans d'historique sur l'IA « métier »

## Où en est-on ?

# Niche écologique



# Déploiement très diversifié de l'IA pour l'info géo

- . Diversité des contextes : secteurs, échelles
- . Diversité des tâches & techniques
- . Diversité des fonctions endossées par les systèmes IA

# Exemple : cartographie nationale de l'occupation des sols (IGN)

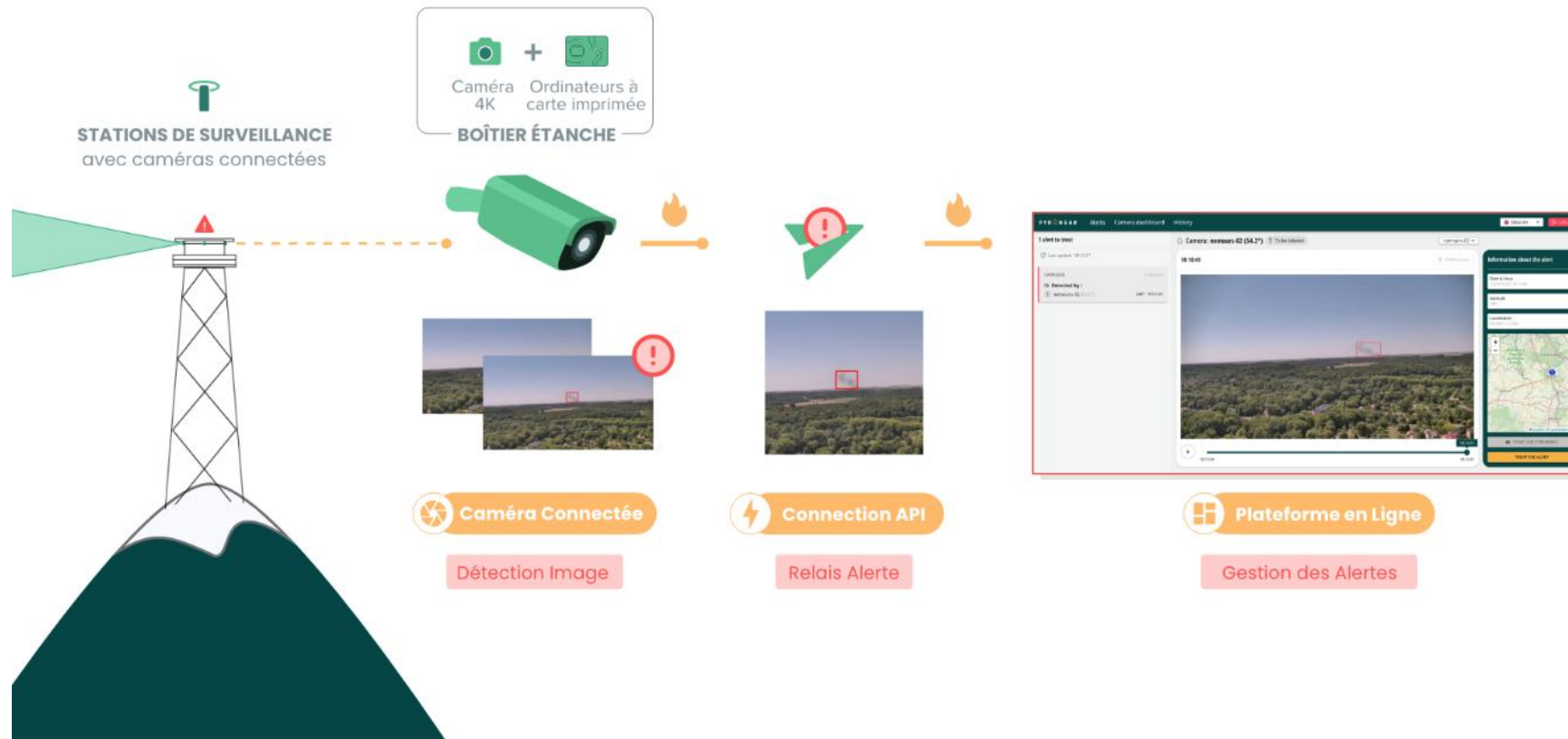




# Exemple : détection précoce de feux de forêts (Pyronear)

## — VUE D'ENSEMBLE DE LA SOLUTION —

CAS D'USAGE POUR DES POMPIERS EN FRANCE



## Exemple : prévision ressources en eau (Climate Data Hub CVDL)



## Exemple : détection et catégorisation de déchets (Metz)



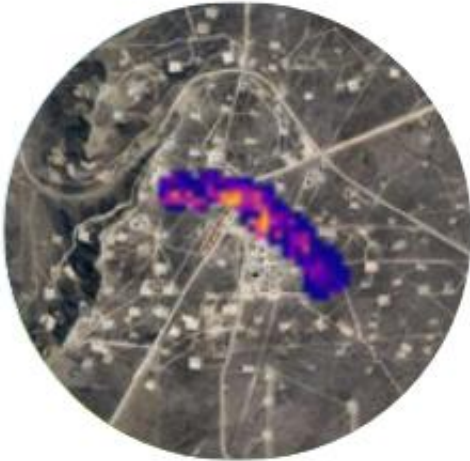


# Exemple : détection de panaches de méthane (UNEP / IMEO)

**Figure 1**  
 The first two images show plumes detected by NASA's EMIT in Chubut, Argentina, on 13 and 21 November 2024. No further emissions have been detected at the location since 8 January 2025.



2024-11-13



2024-11-21



2025-01-08



# Déploiement très diversifié de l'IA pour l'info géo

- Secteurs : climat, météo, biodiversité, aménagement, transports, déchets, risques, agriculture...
- Tâches : détection d'objets, segmentation sémantique, prédiction...
- Données : images aériennes, satellites, 2D, 3D, images naturelles, données structurées...
- Toutes échelles du local au mondial, et lien entre échelles
- Fonctions : massifier production de données, complétude données manquantes, prédire/simuler, détecter des anomalies, priorisation de contrôles...

# Déploiement très diversifié de l'IA pour l'info géo

- Population des systèmes IA a pris très largement place pour la production d'info géo : modèles spécialisés, mais grande généralité de l'intégration de l'IA

# Ressources

- . Données : instrumentation & capteurs, labellisation...
- . Compétences multiples
- . Capacités de calcul

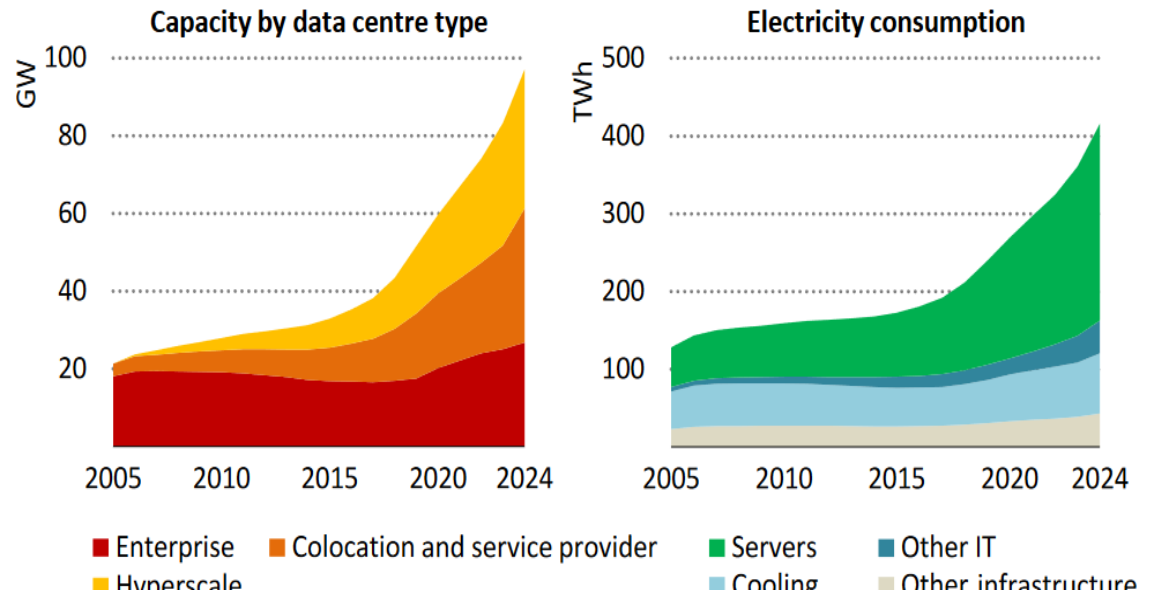
# Etudes sur les impacts environnementaux des centres de données à l'échelle mondiale et prospection

## En 2025

- Data centers = 1,5% de consommation mondiale en électricité, soit **415 TWh**
- Croissance de **12% par an** depuis 5 ans

## Prospection pour 2030

- Centres de données = 3% de consommation mondiale en électricité, soit **945 TWh**
- Croissance de **30% par an** (adoption de l'IA)



Consommation totale d'électricité des centres de données par type d'équipement et par type de centre de données, 2005- 2024



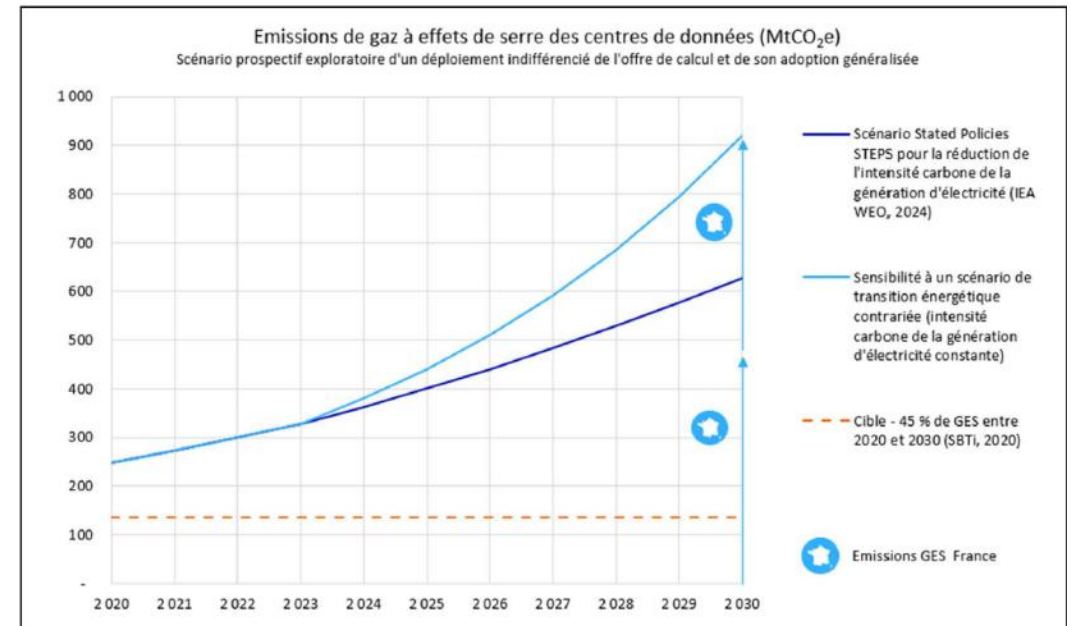
# Etudes sur les impacts environnementaux des centres de données à l'échelle mondiale et prospection

La filière centres de données justifie les infrastructures fossiles

- Relance massive des infrastructures fossiles (gaz)
- Annulation de fermeture programmée de centrales à charbon

Et l'IA dans tout ça ?

- En 2025, 15% des usages en centre de données
- Prédiction pour 2030 : **55%**



# Limites & Dynamiques en cours

- Multimodalité
- Frugalité
- Réplicabilité & capitalisation