

DETECTION AND CHARACTERISATION OF POLLUTANT ASSETS WITH AI AND EO TO PRIORITISE GREEN INVESTMENTS: THE GEOASSET FRAMEWORK

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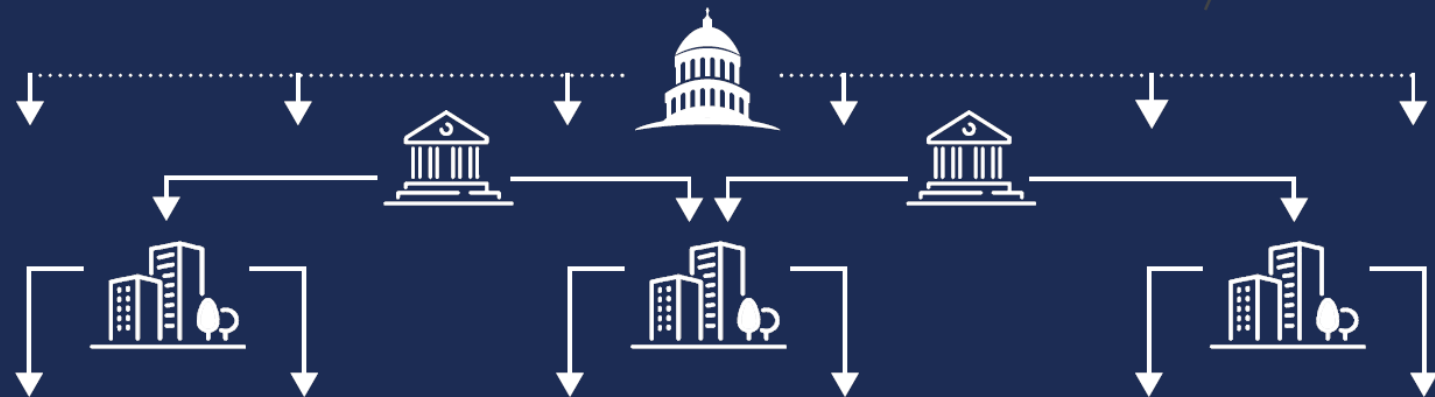
SPATIAL FINANCE AND ASSET LEVEL DATA

Spatial finance is the integration of geospatial data and analysis into financial theory and practice

GOVERNMENTS regulate and create policies across all sectors

INVESTORS own companies

COMPANIES own exposed assets



ASSETS, both built and natural, are exposed to different climate risks, impacts and opportunities



OBSERVATIONAL DATA
E.g. GHG emissions, climate hazard, air pollution

ASSET DATA
E.g. location, ownership, production type, capacity, age

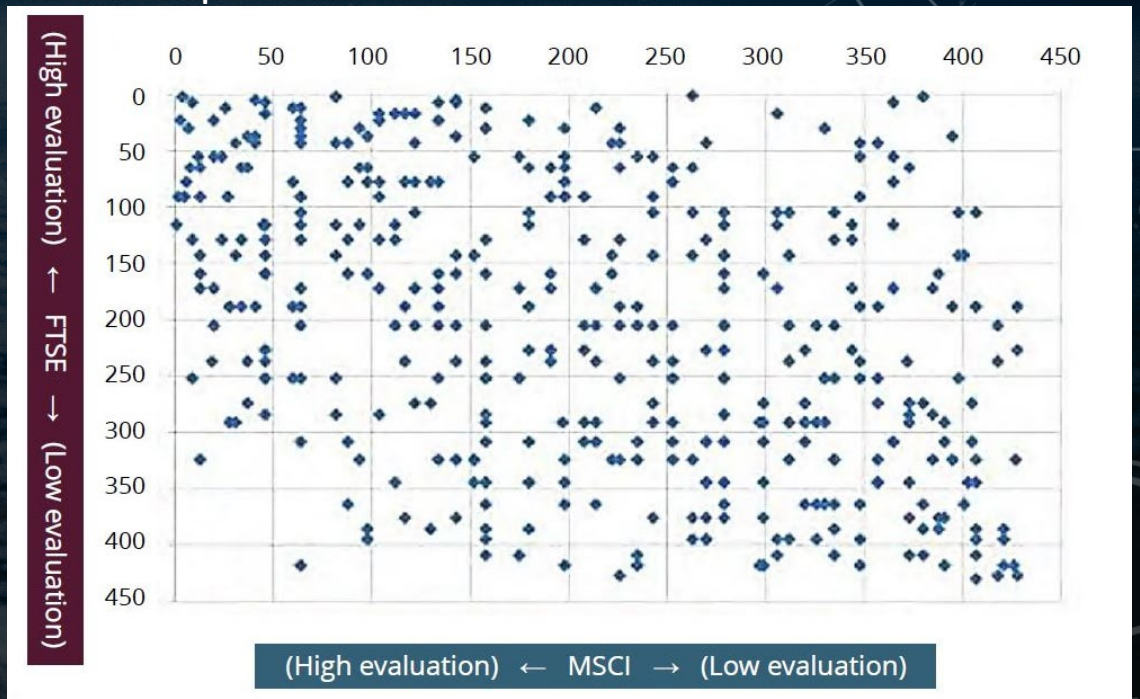
DATA CHALLENGES

Main source of data on a company's sustainability performance is its voluntarily reported information or 'disclosures'.

This comes with various challenges:

- Information is **self-disclosed** and typically compiled top-down
- ESG reporting is not mandatory, which means **not all companies report**
- ESG reporting not standardised, which means **data is not comparable**
- Annual reporting means ESG **data is outdated** once released

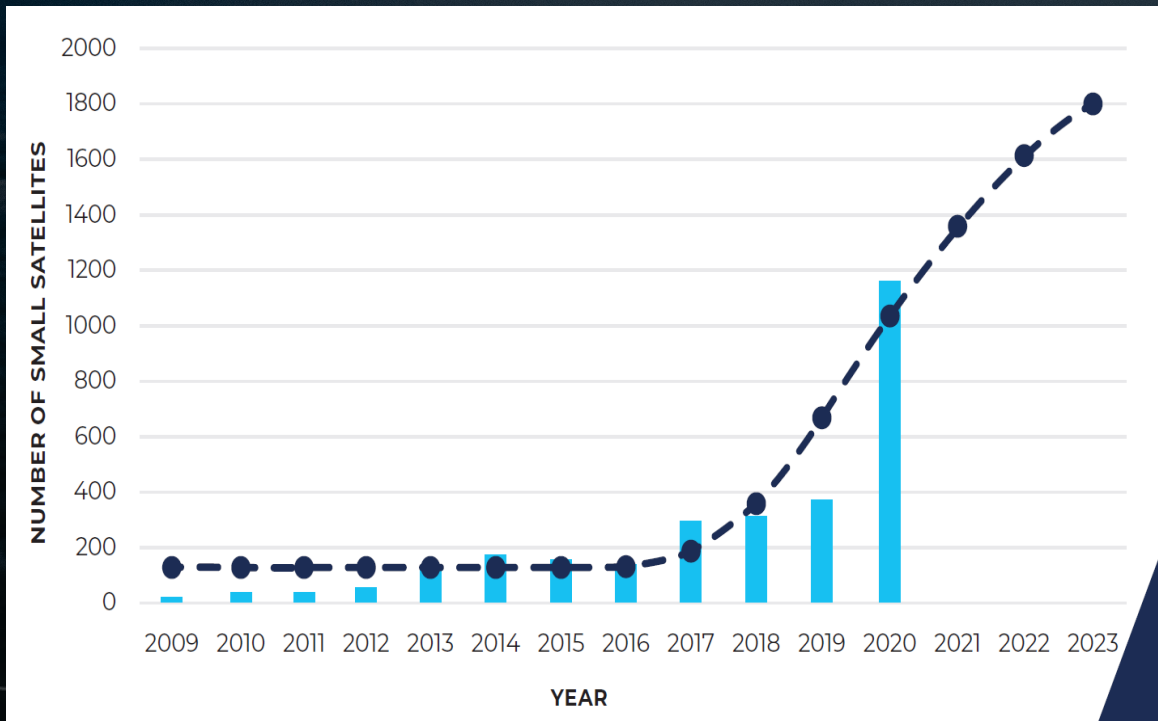
Comparison of ESG scores from FTSE and MSCI



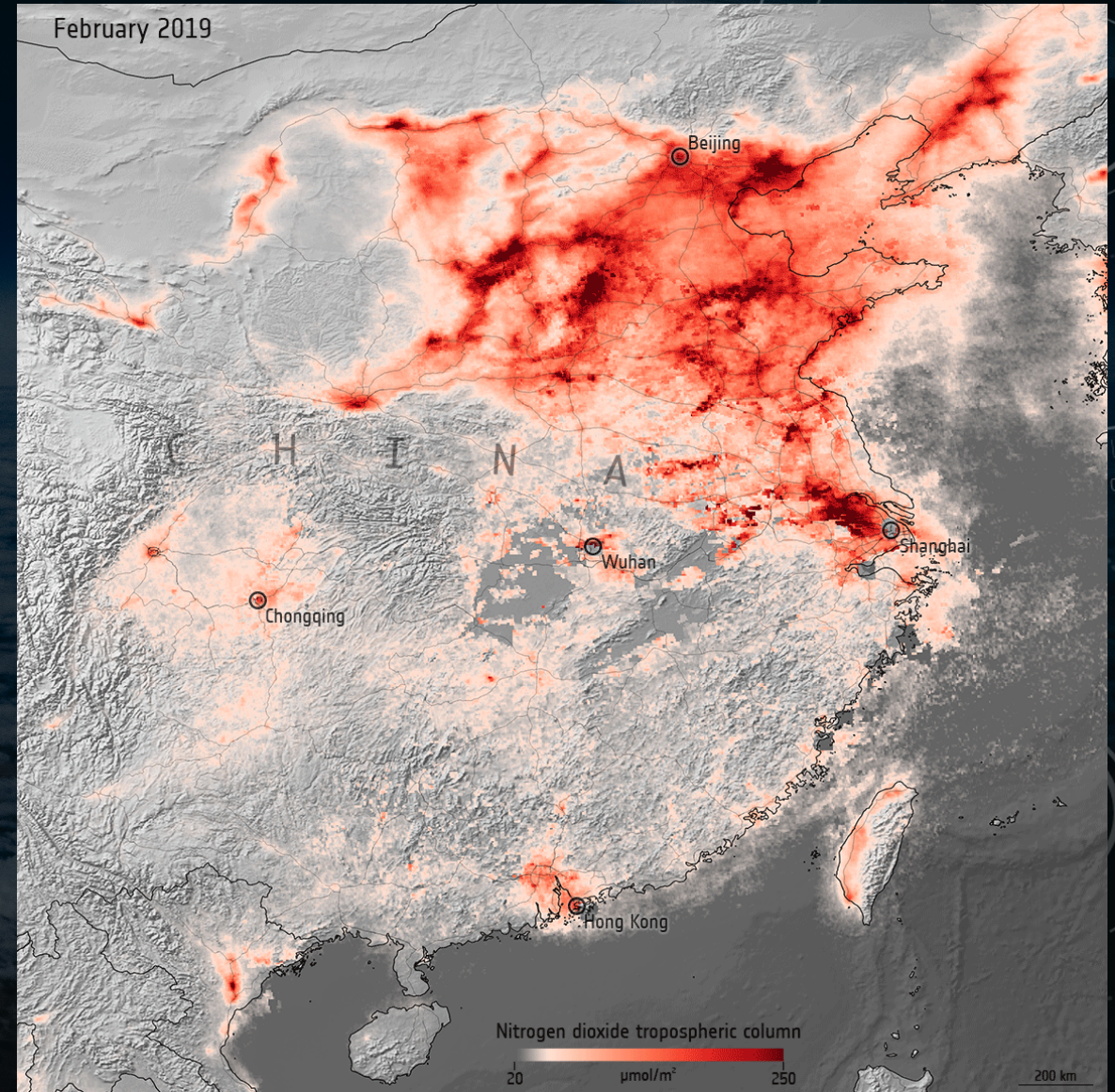
Source: CLSA/ACGA

"... this chart does not discredit ESG data or the practice of scoring ... it underscores the danger of relying on a simple final score for investment decisions"

OBSERVATIONAL DATA



Number of small satellites launched, historical (columns) and modelled (line)
 Source: Satellite Applications Catapult



Nitrogen dioxide emissions over China in 2019-2021
 Source: European Space Agency

SPATIAL FINANCE RELEVANCE

Spatial finance allows for a **bottom-up** understanding of **risks, opportunities and impacts**, driven by

- Neutral and objective source of information
- Data that is more timely and comparable
- Connecting financial system with real economy

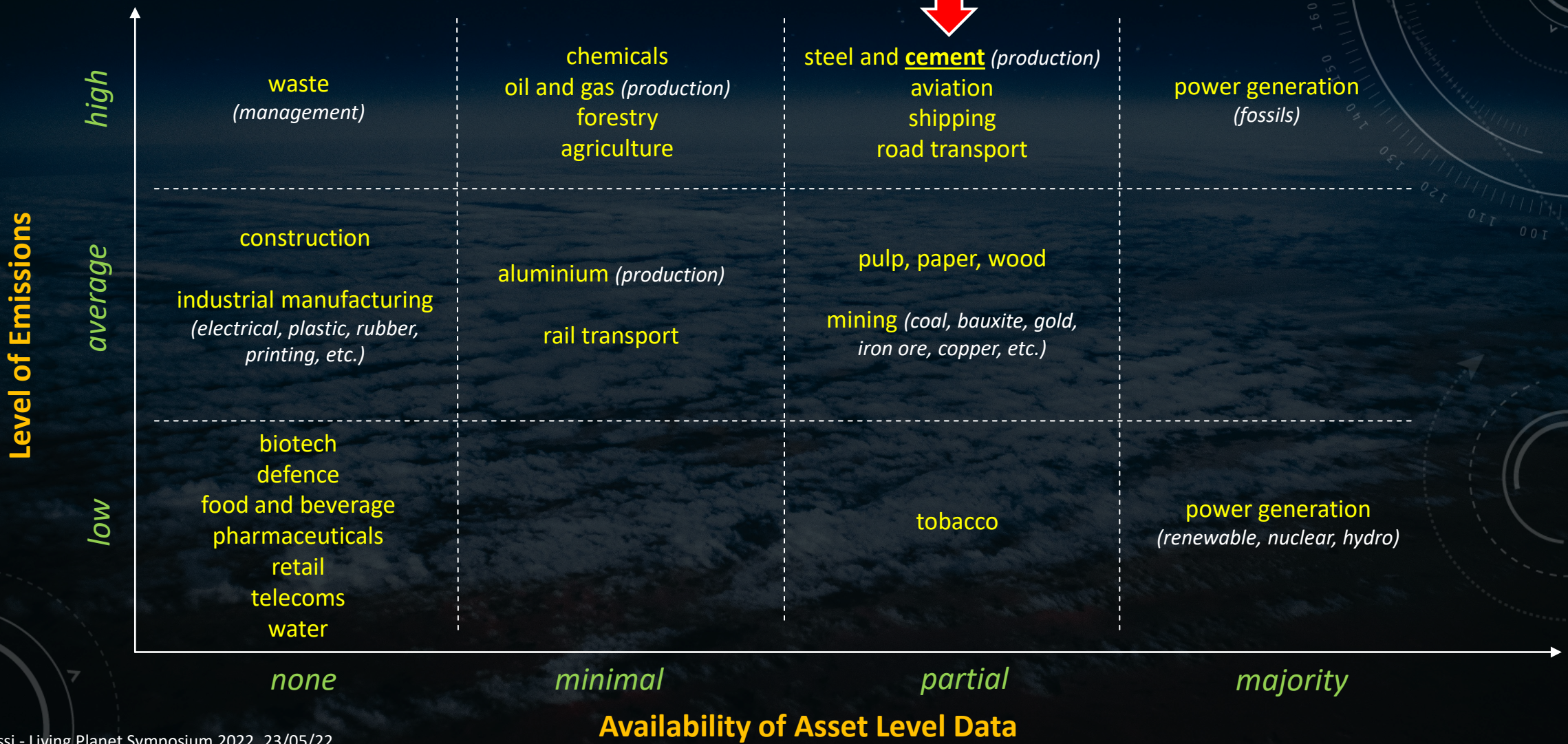
We are seeing an increasing uptake of spatial finance approaches and analytics, driven by

- Climate-related financial risk assessment and disclosure
- Nature-related financial risks and impact assessment
- Investor engagement campaigns on sectoral ESG issues



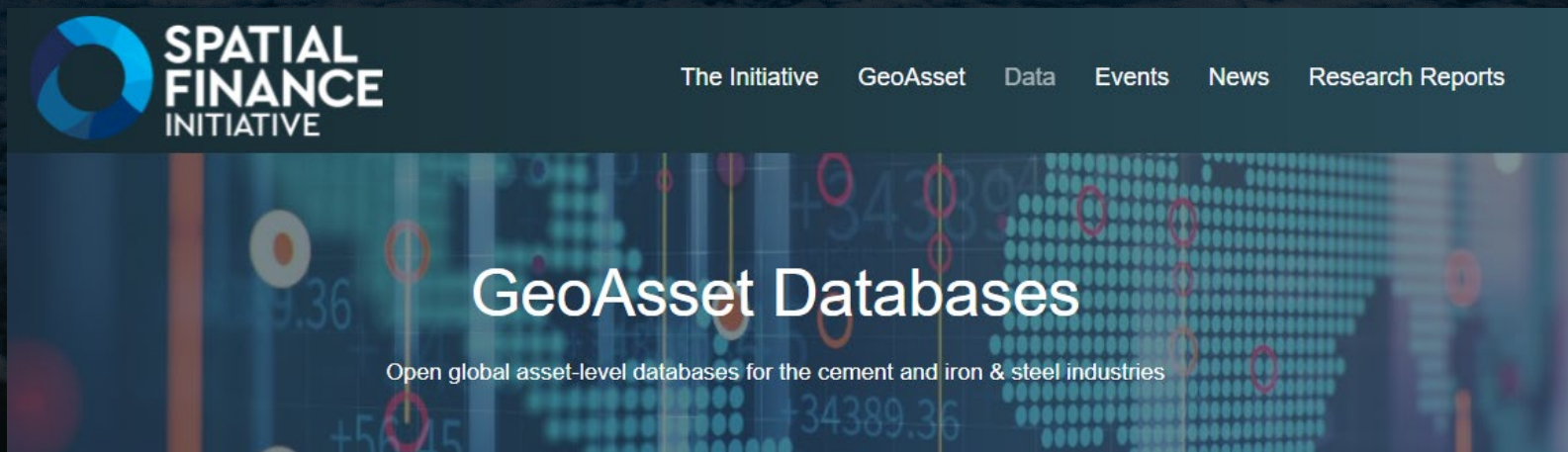
Source: European Space Agency

MISSING DATA



GEOASSET FRAMEWORK

- GeoAsset is a **collaborative** endeavour to provide accurate, comprehensive, comparable foundational data on physical assets across all major sectors, tied to financial ownership information, at low cost and as in as **open** a format as possible.
- Part of the new **UK Centre for Greening Finance & Investment**. Core partners are Oxford Sustainable Finance Programme, The Alan Turing Institute, and the Satellite Applications Catapult.



Datasets are available to download from:

<https://spatialfinanceinitiative.com/geoasset-project/data/>

BUILDING THE CEMENT DATASET: HIGH-LEVEL ARCHITECTURE



EXISTING DATASETS CURATION = TRAINING DATASET



Coordinate set 1

- Cement tag
- Incomplete

Coordinate set 2

- PDF format
- Incomplete

Coordinate verification and refinement

Ownership identification

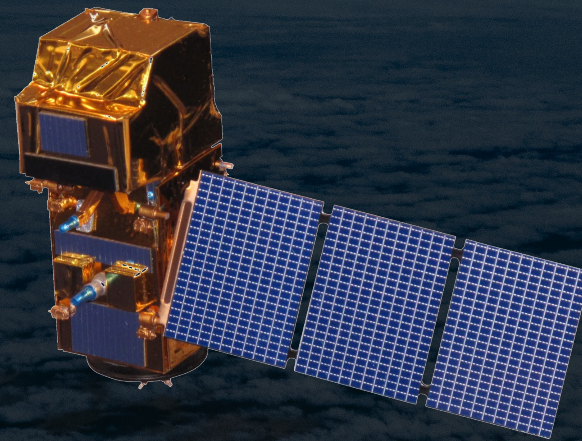
- Company disclosure of major producers
- Disclosure of **capacity** and **starting production year**
- **Ownership** verification with OpenPermid database

REMOTE SENSING DATASETS



Landsat-8 (©USGS)

Open Access, 100m (thermal)



Sentinel-2 (©ESA)

Open Access, 10m

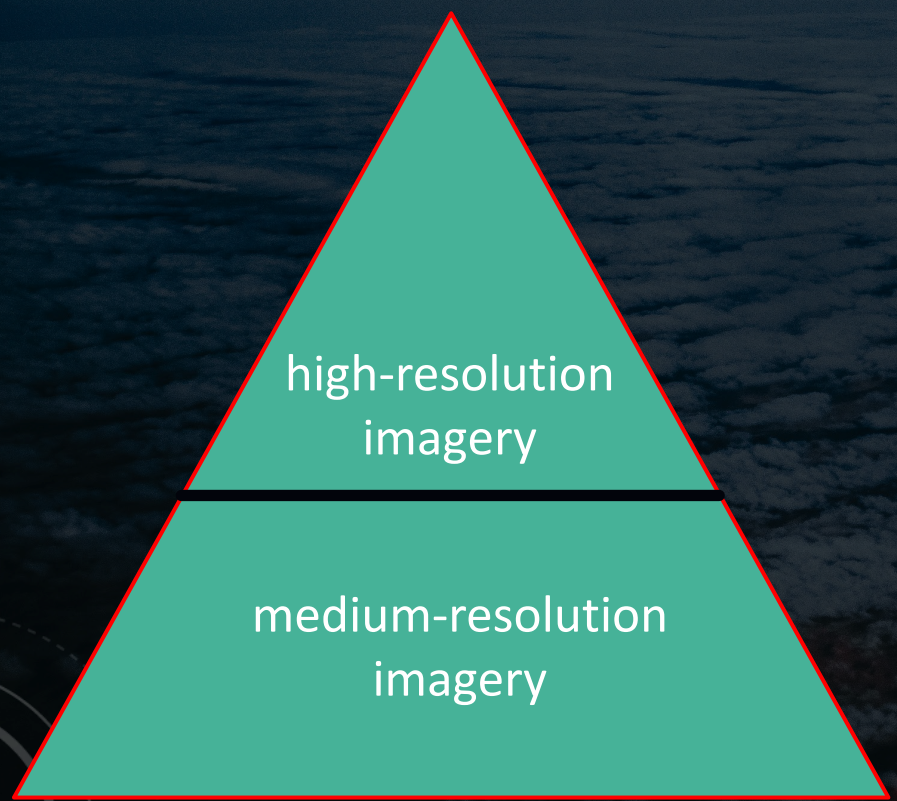


WorldView-3 (©Maxar)

Commercial, 1.5m

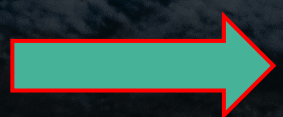
MACRO-LOCALISATION VS MICRO-LOCALISATION

The asset localisation can be solved in a two-steps approach: a macro-localisation, that focuses on the broader area around the site, and a site-localisation, that focuses on the asset localisation



→ Test phase on the final approach ←

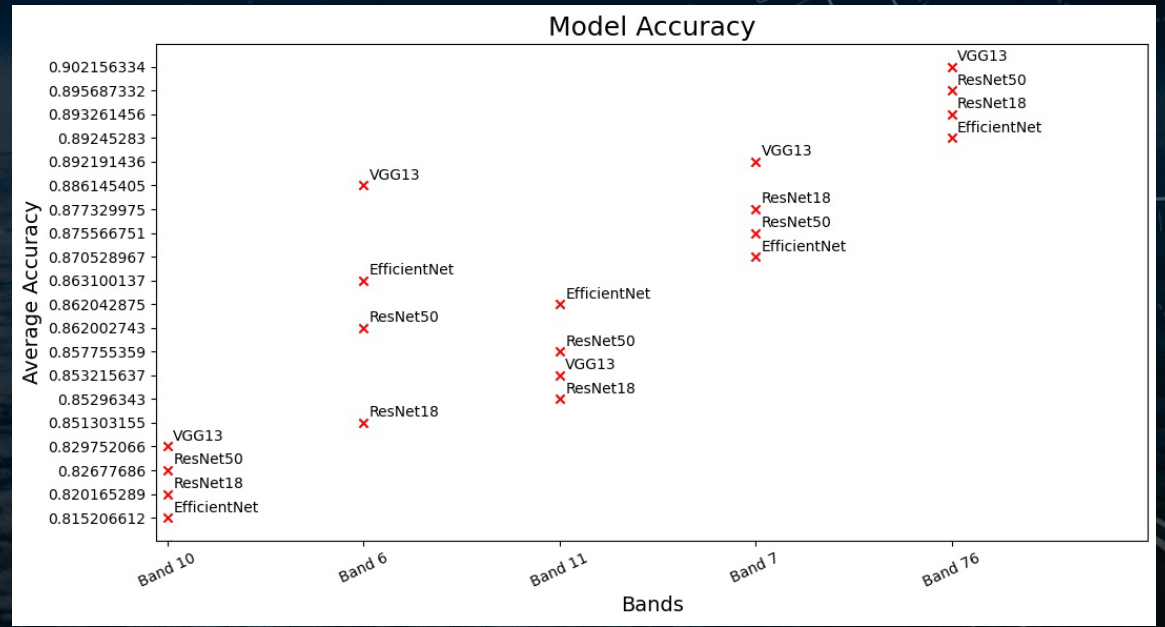
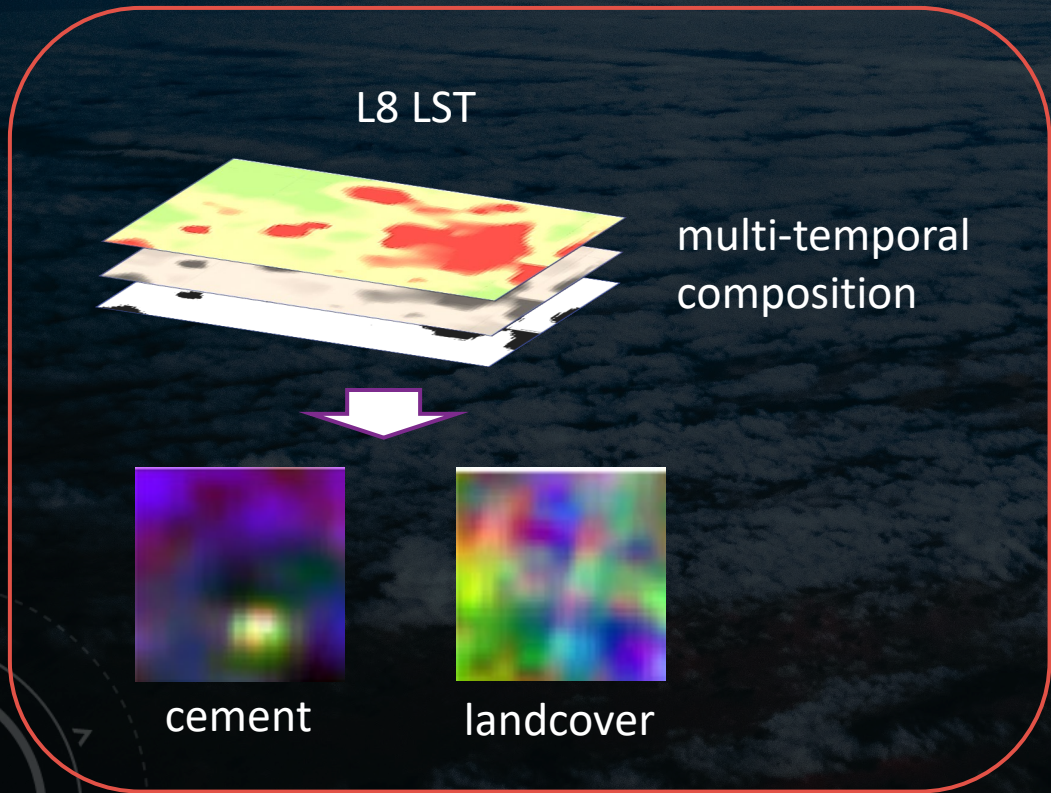
Computer vision/collective intelligence exploration from commercial imagery (Maxar)



Spectral/CV exploration from open access imagery (Landsat-8/Sentinel-2)

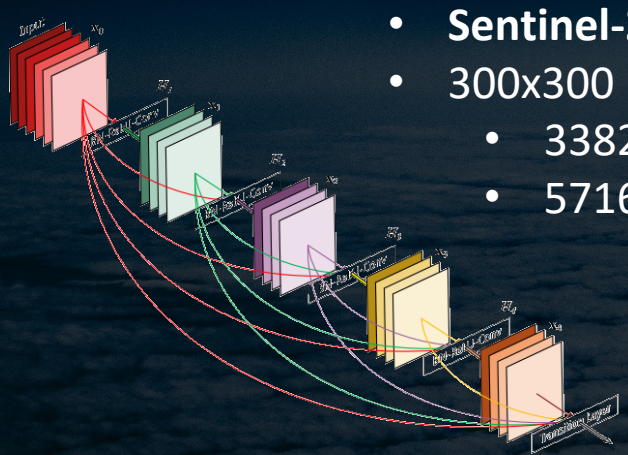
PHYSICS-AWARE AI: LANDSAT-8

- Exploration of Landsat-8 thermal infrared / SWIR bands / iron-oxide ratio



- 4 DL architectures tested
- Best cement plant detection results with **band ratio**
- Still under research, noy yet deployed

MACRO-LOCALISATION: DENSENET



- Sentinel-2 RGB band only
- 300x300 pixels chip extraction (3x3km)
 - 3382 cement chips
 - 5716 landcover chips

- Model deployment over **China** (9.3M sqkm)
- Deployment within ASTRAEA **EarthAI** platform
- Every chip (>400k chips) is scored with the S2 RGB model (0-1: probability of cement plant)

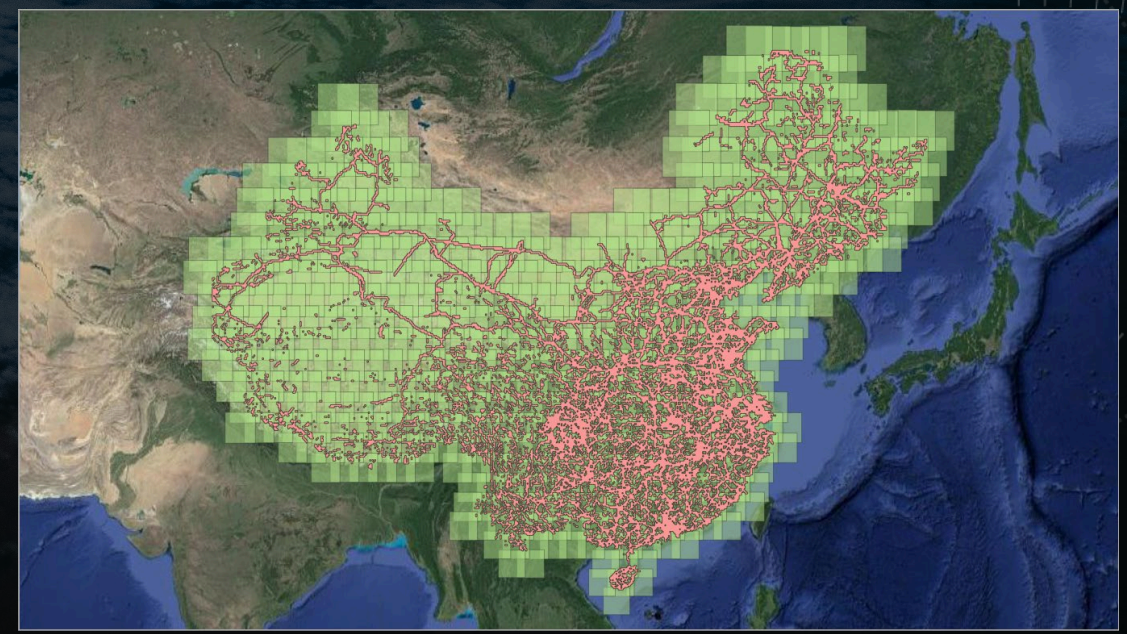
- Several architectures considered, best results* with **DenseNet161** *recall=coverage of actually positive sample

Overall Statistics:

Accuracy: 0.9331703341483293
Population: 1227

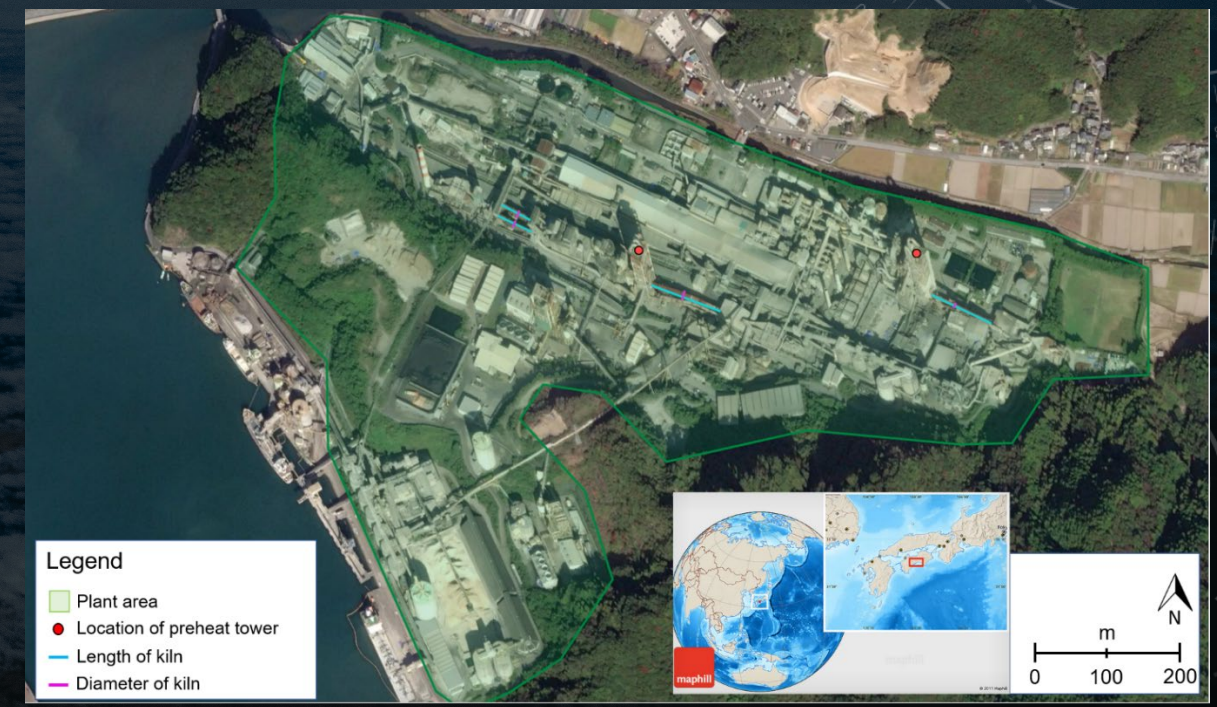
Class Statistics:

	label	TP	FP	FN	Population	Precision	Recall
0	cement	414	69	9	423	0.857143	0.978723
1	landcover	648	10	66	714	0.984802	0.907563
2	steel	83	3	7	90	0.965116	0.922222



MICRO-LOCALISATION: COLLECTIVE INTELLIGENCE

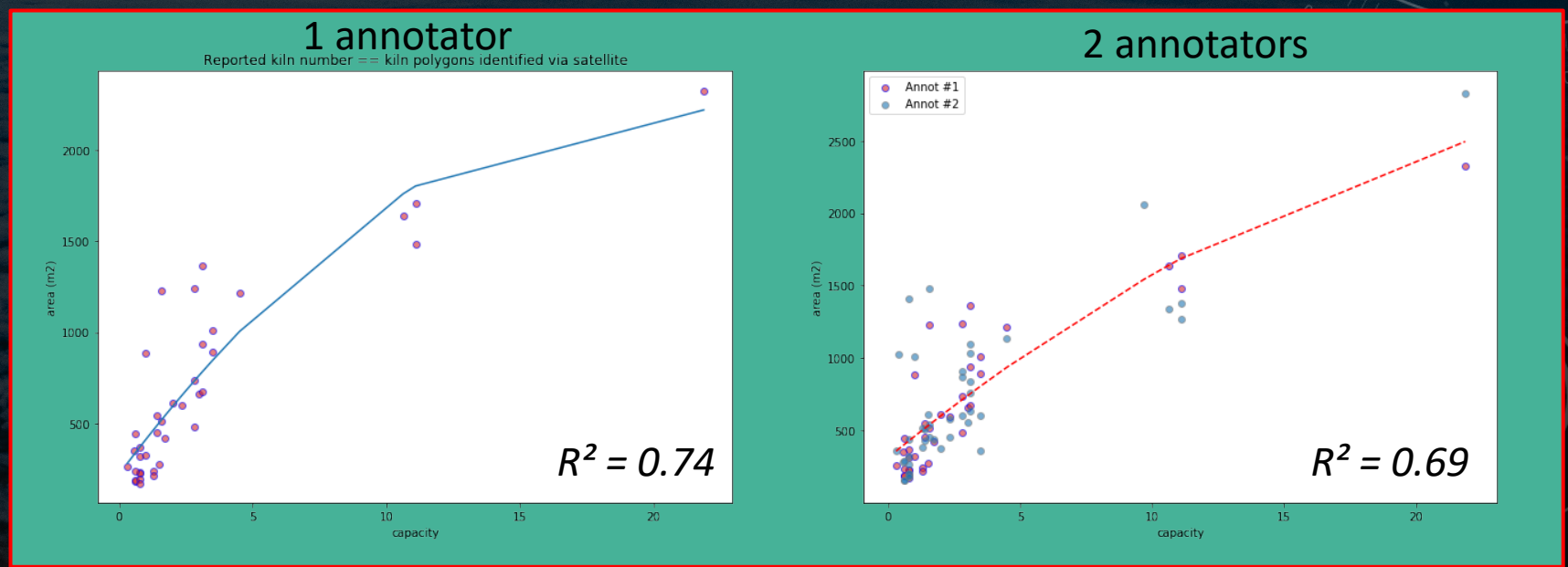
- Multiple labellers on two stages: first **Google Earth Pro**, and then Maxar's **SecureWatch** for unclear imagery



Asset annotations

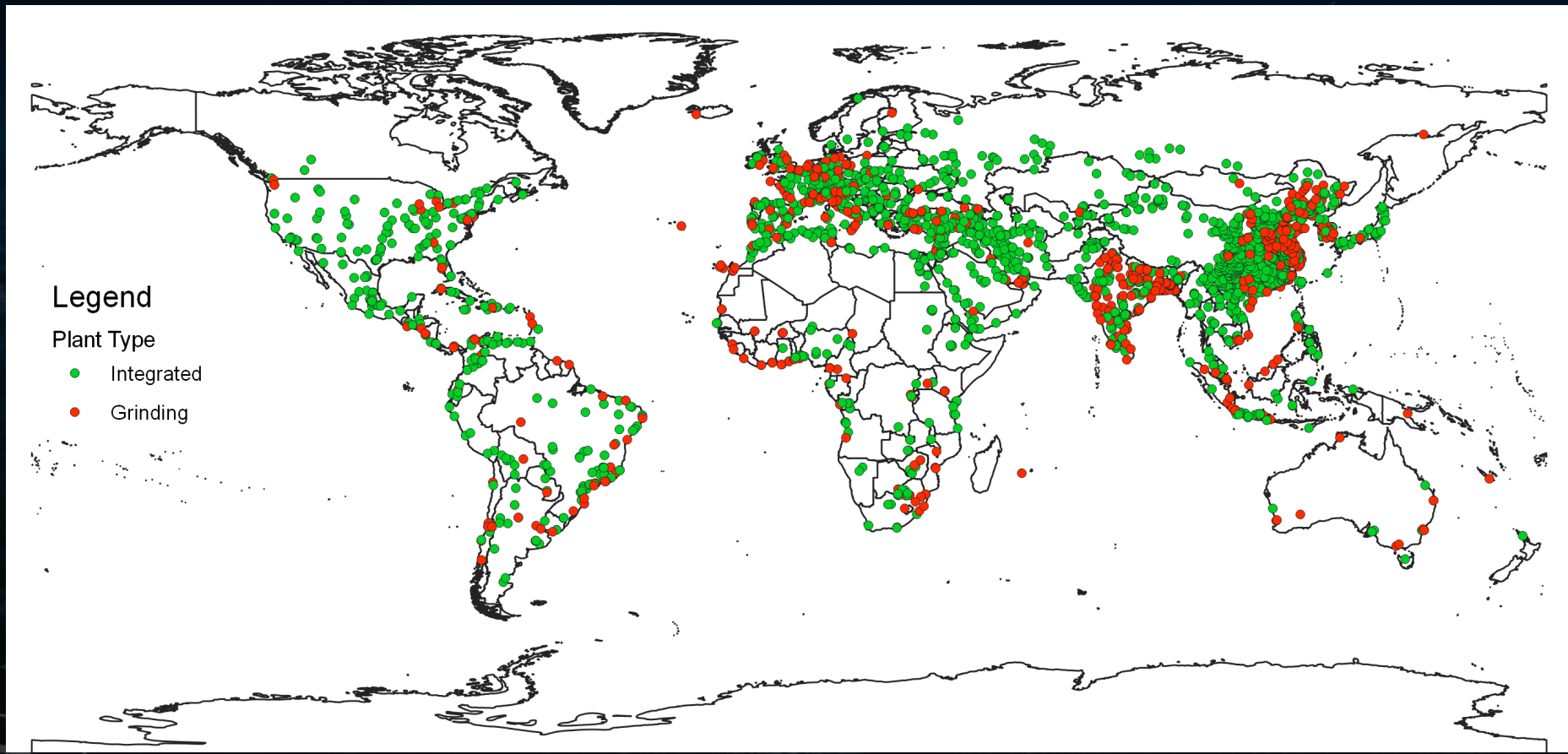
MANUAL CORRELATION KILN VS CAPACITY

- Kiln dimensions are hypothesized to give an indication on the production capacity. To investigate this relationship, the reported plant cement capacity is compared with kiln dimensions manually annotated from EO



- Positive **correlation** between reported capacity and the area of kilns from manual annotation
- Uncertainty given by discrepant number of kilns in the database vs annotated ones and geometrical distortions in the imagery

FINAL DATASET

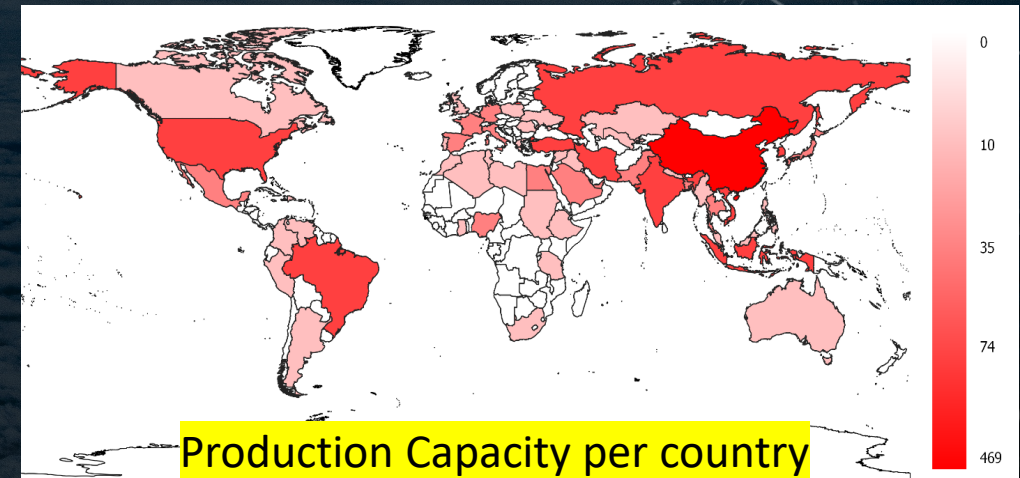


- Plant location
- Plant type
- Production type
- Plant capacity
- Year production first started
- Detailed ownership

3,117 cement production assets accounting for ~90% of global cement production capacity

DATASET VALIDATION

- Expert knowledge workshops for an accurate methodology development and crowdsourcing instruction
- 88.9% of total cement capacity captured (5.5Mt / 6.2Mt)
 - Continent statistics within 5% error
 - Near complete coverage in EU and North America
 - Overestimation in India/Russia (it might be correct!)
 - Not all plants captured in China (-25%)
- Extensive manual validation of the Global Cement Database (GCD)
 - GeoAsset Database captures more assets globally



Dataset	Integrated		Grinding		Unknown		Total	
	N	Capacity	N	Capacity	N	Capacity	N	Capacity
GCD (Reported)	2,330	3,934.4	673	537.2			3,003	4,471.6
GCD (Validated)	1,796	3,301.1	518	478.5			2,314	3,779.6
GeoAsset Database	2,243	4,351.1	738	940.4	136	221.0	3,117	5,512.6

CONCLUSIONS

- Successful demonstration of **AI4EO** frameworks
- Geoasset datasets are an important resource to adequately assess **environment-related risk** and **impact exposure** and the diffusion of these risks and impacts through the financial system
- Reference: McCarten, M., Bayaraa, M., Caldecott, B., Christiaen, C., Foster, P., Hickey, C., Kampmann, D., Layman, C., Rossi, C., Scott, K., Tang, K., Tkachenko, N., and Yoken, D. 2022. Global Database of Cement Production Assets. Scientific Data, *under review*

thank you for your attention!

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