

Agentic AI for Earth Observation

Open concepts and architectures for the next generation of EO workflows

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Today, in 60 minutes

- 1 Why now** — the EO data deluge meets agentic AI 10 min
- 2 What “agentic” actually means in EO** — planning, tools, memory, feedback 10 min
- 3 The open-source building blocks** — data → APIs → tools → models → agents → HPC 15 min
- 4 Three walkthroughs, step by step** — forest change · solar farms · flood impact 13 min
- 5 Trust, governance & discussion** — benchmarks, failure modes, EU AI Act, Q&A 12 min

You'll leave with

- ✓ A working mental model of agentic EO systems
- ✓ A reference architecture and a map of open tools
- ✓ A critical checklist: reliability, evaluation, governance

Format notes

No live demos or exercises — static, reproducible walkthroughs instead. Slides + an annotated link list may follow by e-mail.

Questions: you can email me and I am happy to discuss specifics! karol.bot.goncalves@vsb.cz

Earth observation has a scale problem

22 TB+

of new Copernicus data, every single day

EU Space Programme

50+ PB

online at the Copernicus Data Space Ecosystem — 500,000+ registered users

Nature Sci. Data, 2026

245 PB

NASA Earth-science archive projected by 2025; NISAR alone adds ~80 TB/day

NASA JPL

50+ yrs

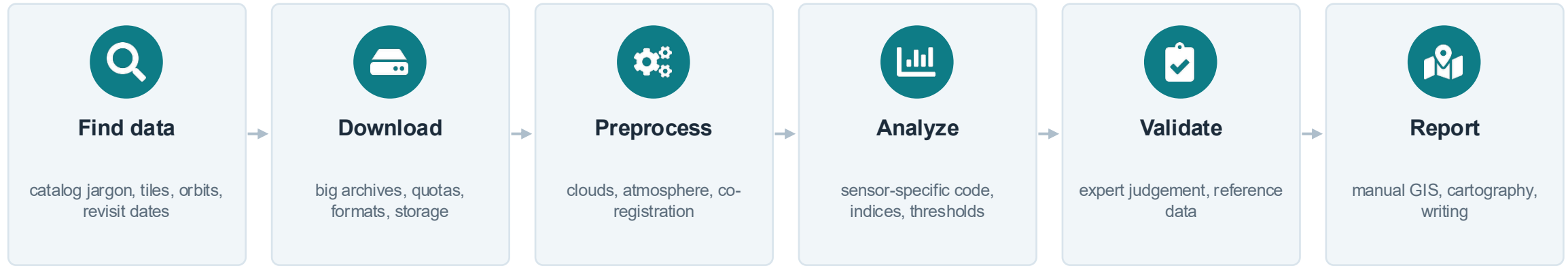
of Landsat — more than 10 million scenes and counting

USGS

Data-rich, insight-poor. The bottleneck is no longer pixels — it is expert time.

Every question still needs someone who knows sensors, formats, indices and GIS tooling. So most monitoring questions are never asked — or answered once a year instead of continuously.

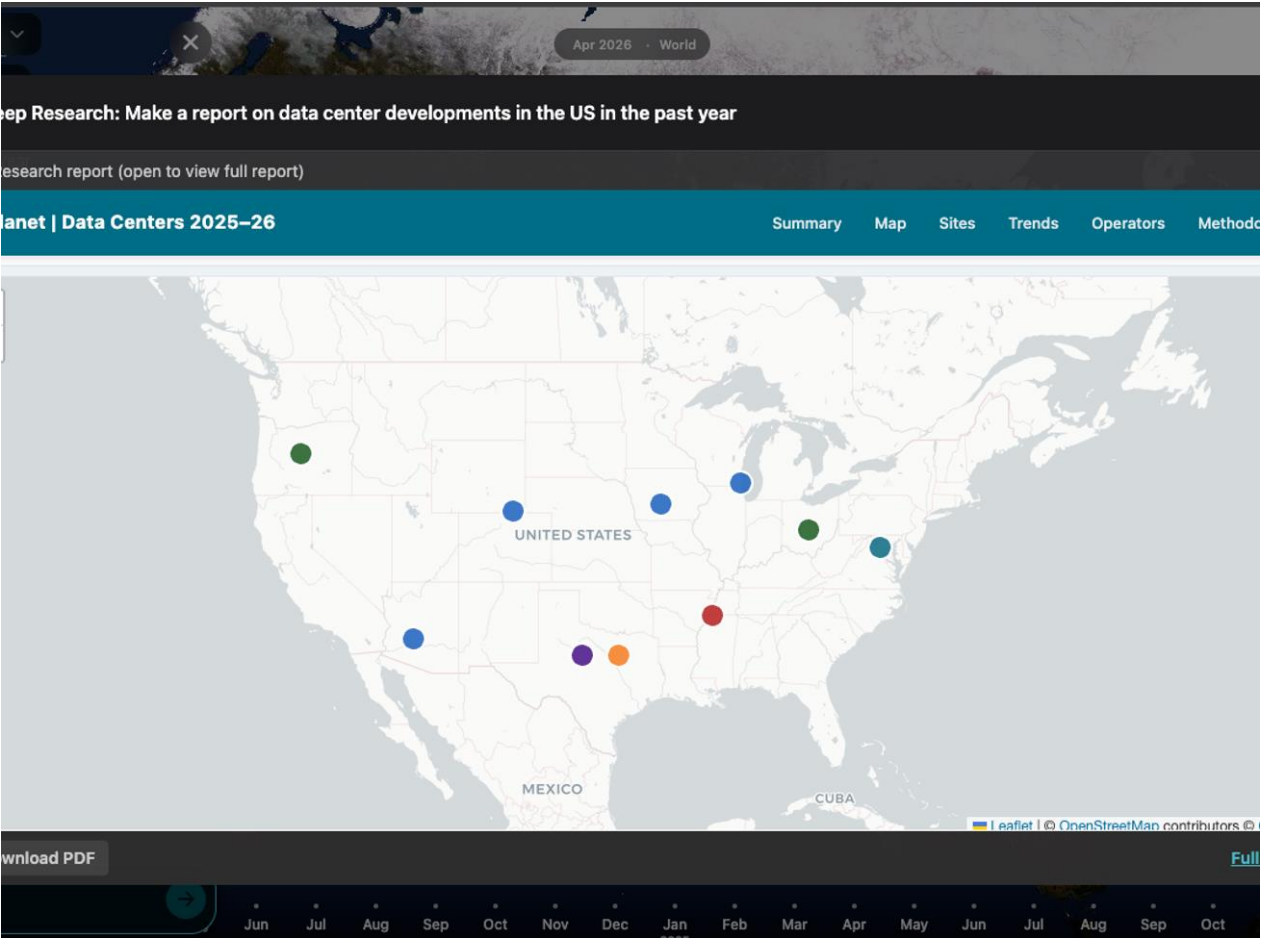
Why answering one question takes weeks



Each stage assumes a different specialism — remote sensing, data engineering, statistics, cartography. **Multiply by every region, hazard and reporting cycle → weeks of specialist effort per question.**

Agentic systems attack the orchestration, not the science: they automate the gluing between stages, while the measurements stay in verifiable tools.

Planet turns a question into a map and a research report



What is actually visible in the beta

Surface	Status	Input	Process	Output
Map + chat	Open beta	Plain-language question	Multi-step research	Interactive map
Planet archive	May 2026	AOI + topic	Change detection	Site evidence
Public sources	Product demo	Time window	Synthesis	Shareable report
Important	MCP ≠ analytics	MCP: search/order	Beta: analysis	Check provenance



A deployed-looking interface is not proof of scientific validity

- Use it as motivation: the interaction model is real. Then ask what scene IDs, model versions, checks and uncertainty are exposed behind the polished report.

Google’s research agent grounds damage assessment in specialist models

an aerial view of a mixed landscape of buildings, primarily residential, with varying degrees of damage. Many buildings exhibit damage caused by a hurricane, including missing tiles, displaced materials, and roofs covered with blue tarps. Debris is visible in some of the buildings. A few buildings show no visible damage.



What the research demonstration chains

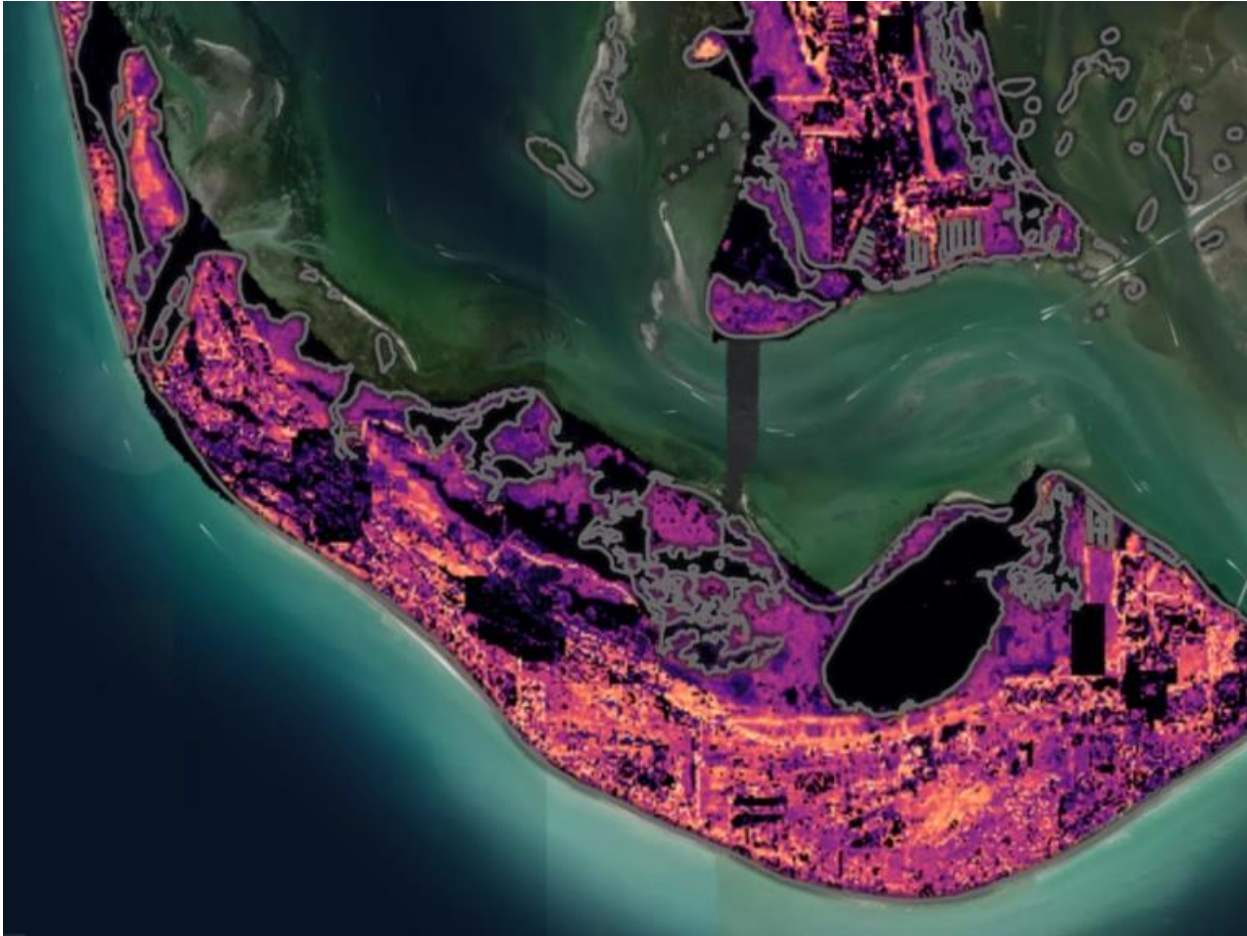
Stage	System	Evidence	Role	Maturity
Retrieve	Cloud Storage	Aerial image	Input	Demo
Detect	OWL-ViT RS	Bounding boxes	Tool	Research
Assess	Gemini / captioner	Image patches	Reason	Research
Summarise	Agent workflow	Damage narrative	Output	Edited demo



The visible boxes are tool outputs—not the language model “looking harder”

- Google states that its demonstration video was edited for brevity. This is evidence of a workable architecture, not an independent accuracy claim.

NASA Earth Copilot explores natural-language access to open Earth data



Prototype facts—not a production claim

Owner	Status	Interface	Backend	Access
NASA IMPACT	Prototype	Natural language	Azure OpenAI	Internal test
NASA VEDA	Exploration	Map + chat	Earth datasets	Researchers
Goal	Discovery	Complex questions	Cloud services	Broader access
Caveat	Evaluation	Misuse checks	Before release	No GA claim

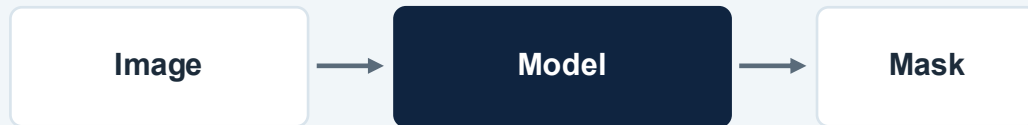


Why this matters at IT4I: open data still needs a trustworthy access layer

- Public systems cannot stop at fluent answers. They need reproducible queries, transparent data selection, access controls and reviewable evidence.

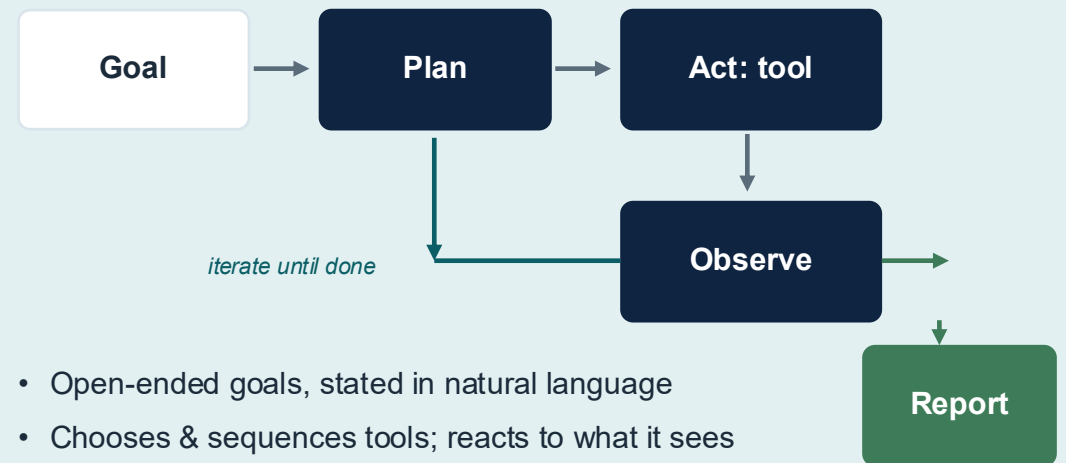
From one-shot models to agentic systems

A one-shot ML model



- One fixed task: cloud mask, crop map, ship detector...
- No context, no follow-up questions, no goal
- You still build the whole pipeline around it

An agentic system



- Open-ended goals, stated in natural language
- Chooses & sequences tools; reacts to what it sees
- Stops when the goal is met — or asks for help

Agentic AI = an LLM-driven system that plans multi-step work, calls tools, keeps state, and checks its own progress toward a goal.

Four capabilities — and one mental shift



Planning

Decomposes “monitor deforestation” into concrete, ordered, checkable steps.



Tool use

Calls catalogues, code and models — instead of guessing what pixels say.



Memory

Keeps the AOI, intermediate products and decisions across steps and sessions.



Feedback

Checks outputs, retries, re-plans — or escalates to a human.

The mental shift: the LLM is the orchestrator — the science happens in the tools it calls.

In EO terms: the model that writes the plan should never be the instrument that measures the world. Language models reason about the workflow; rasters are measured by deterministic, testable code.

How an agent decomposes a real EO question



“Has deforestation increased in region X since 2023 — and by how much?”

1 Resolve the AOI

geocoder · OSM boundaries

2 Find usable imagery

STAC search: S2 L2A, cloud < 20%

3 Build pre/post composites

openEO / xarray median

4 Detect change

dNDVI + threshold + min-area

5 Quantify

zonal statistics, areas per year

6 Report

map + numbers + caveats + provenance

Same skeleton for floods, urban growth, solar farms — only steps 4–5 change. The agent's job is choosing and wiring these steps for each new question.

An agent is not a script — and not just a model

	Scripted pipeline	One-shot model	Agentic system
Interface	code & parameters	an API call	natural language
Scope	one workflow, fixed	one task, fixed	open-ended goals
On failure	crashes; needs a developer	fails silently	retries, re-plans, asks
Determinism	high	high	stochastic — runs differ
Auditability	read the code	model card	must be engineered: logs

Agents trade determinism for flexibility. That trade is precisely what the governance section at the end is about.

Division of labour: what the LLM must never do



Let the LLM...

- Interpret the user's intent and constraints
- Plan and order the tool calls
- Write small snippets of analysis code
- Draft the report — from tool outputs only



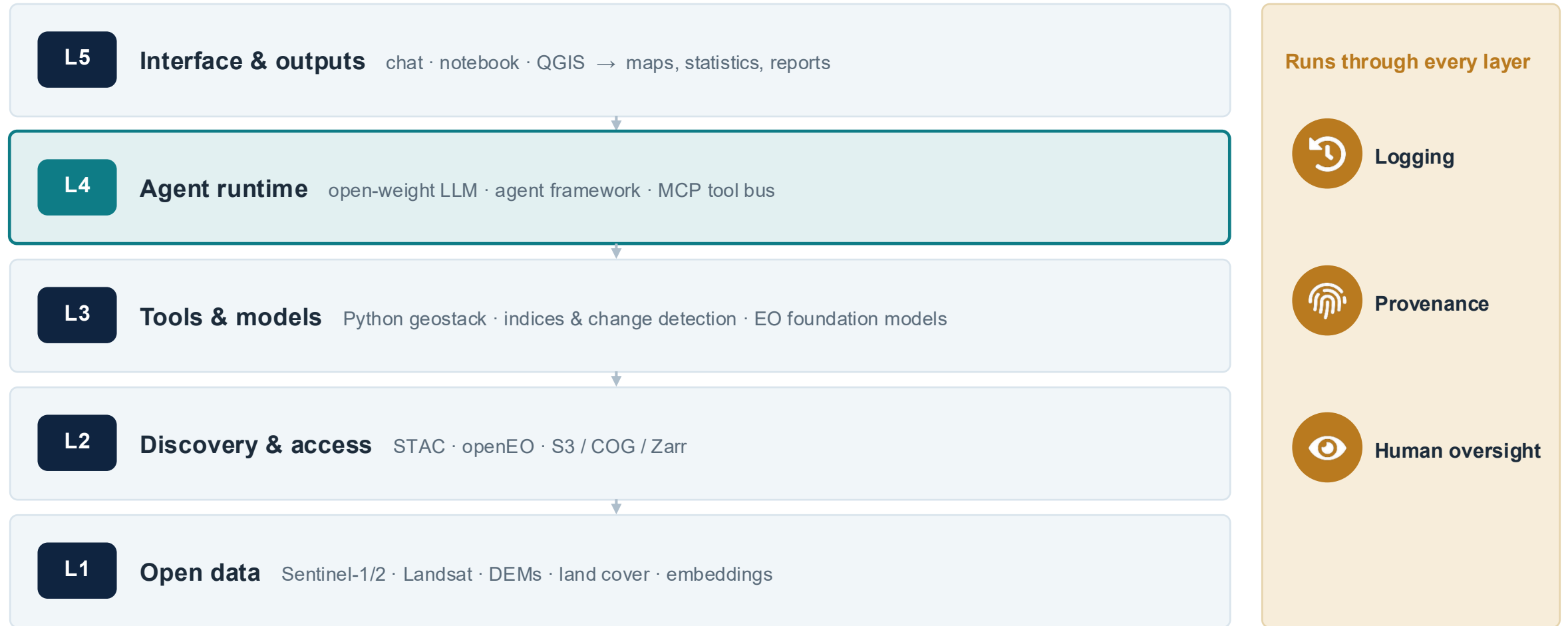
Never let the LLM...

- Read pixel values “by eye” from an image
- Count objects in a scene
- Produce numbers from memory
- Decide alone what is “true” on the ground

Evidence: GPT-4V on EO imagery — “good at captioning, bad at counting” (VLEO-Bench, CVPR Workshops 2024)

Rule № 1 of agentic EO: every number in the report must be traceable to a tool output.

A reference architecture — every layer has open options



Keep this picture in mind — the three walkthroughs will light it up, layer by layer.

Layer 1 — the open data is already there



Sentinel-2 (optical)

10 m, 5-day revisit, 13 bands. Vegetation, land cover, water, change — the workhorse.



Sentinel-1 (radar)

Sees through clouds and darkness. Floods, ice, ships, ground deformation — weather-proof monitoring.



Landsat

30 m since 1972 — 10M+ scenes. The long time machine for trends and baselines.



EO as embeddings — new

AlphaEarth Satellite Embeddings (annual, 10 m, open in GEE/AWS) · Major TOM (ESA Φ-lab, on Hugging Face): search the planet by similarity.

All of it free and global. Copernicus data is full, free and open by regulation — a strategic asset for building agentic systems in Europe.

Layer 2 — discovery & access have standardised

STAC — a common JSON language for “what imagery exists, where, when”. One search client works across providers — exactly what a tool-calling agent needs.

STAC v1.1

OGC Community Standard 2025

openEO: OGC 2026

ONE SEARCH, ANY CATALOGUE — PYSTAC-CLIENT

```
from pystac_client import Client
cat = Client.open("https://stac.dataspace.copernicus.eu/v1")
items = cat.search(
    collections=["sentinel-2-l2a"],
    intersects=aoi_geojson,
    datetime="2026-06-01/2026-08-31",
    query={"eo:cloud_cover": {"lt": 20}},
).item_collection()  # → scenes, as JSON
```

Copernicus Data Space (CDSE)

free Sentinel archive: STAC + openEO + Sentinel Hub APIs + S3; cloud-native Zarr (EOPF) rolling out now

Microsoft Planetary Computer

free STAC API over 100+ open collections (the classic catalogue — distinct from the paid “Pro”)

Earth Search (AWS / Element 84)

no-auth STAC for Sentinel-2/-1 COGs and Landsat on AWS open data

NASA Earthdata CMR-STAC

all of NASA's Earth-science holdings, exposed as STAC catalogues

Layer 3a — the analysis toolbox agents call

The Python geostack

pystac-client

odc-stac

xarray

rioxarray

Dask

GeoPandas

GDAL 3.12

samgeo

spyindex

...all mature, maintained, scriptable — i.e. callable as agent tools.

The classic recipe (80% of monitoring):

search → *composite* → *index* → *difference* → *threshold* → *vectorise*
→ *summarise*

Spectral indices — simple, robust, explainable

NDVI

$(\text{NIR} - \text{Red}) / (\text{NIR} + \text{Red})$
vegetation vigour

dNBR

$\text{NBR}_{\text{pre}} - \text{NBR}_{\text{post}}$
burn severity

NDWI

$(\text{Green} - \text{NIR}) / (\text{Green} + \text{NIR})$
open water

CHANGE DETECTION IN FIVE LINES — XARRAY

```
cube = odc.stac.load(items, bands=["red", "nir"])
ndvi = (cube.nir - cube.red) / (cube.nir + cube.red)
change = ndvi.sel(time="2026-07").median("time")
        - ndvi.sel(time="2023-07").median("time")
loss = (change < -0.25) & (baseline > 0.6)
```

```
# deterministic, testable, unit-checkable —
# this is what the agent calls, not what it is
```

Layer 3b — open geospatial foundation models

Prithvi-EO-2.0

NASA × IBM · 300M/600M params · Apache-2.0

Multi-temporal ViT on NASA HLS data. Floods, burn scars, crop mapping. A variant has even flown in orbit (2026).

TerraMind 1.0

ESA × IBM · Apache-2.0

Any-to-any multimodal: S-1, S-2, DEM, land cover, text. ESA-reported best-in-class in evaluations at up to 10× less compute.

Clay v1.5

Nonprofit · Apache-2.0

General-purpose Earth embeddings for search, similarity and light-weight classification heads.

SAM 2 + samgeo

Meta · Apache-2.0 (SAM 3: custom licence)

Promptable segmentation for imagery: fields, roofs, solar panels — click or text-prompt, get polygons.

Fine-tune with TerraTorch / TorchGeo (community-run). LoRA-scale budgets, not training from scratch. And remember: these models are tools the agent calls — not the agent.

Layer 4 — an agent brain you can self-host

Open-weight LLMs with tool calling (Aug 2026)

- **Qwen 3.x** · Apache-2.0 — strong agentic focus, all sizes
- **gpt-oss-120b** · Apache-2.0 — sized for a single 80-GB GPU
- **Mistral 3** · Apache-2.0 — EU-based provider
- **DeepSeek V3/V4** · MIT — frontier-scale open weights
- **Llama 4** · community licence — open weights, some terms
- **Gemma 4** · Apache-2.0 (new in 2026) — small & efficient

Orchestration (open source)

LangGraph 1.0 · **smolagents** · **Pydantic AI** · **CrewAI**

(AutoGen lives on as MS Agent Framework / AG2 — don't cite it by its old name)

Serving on your hardware

vLLM — the de-facto standard server

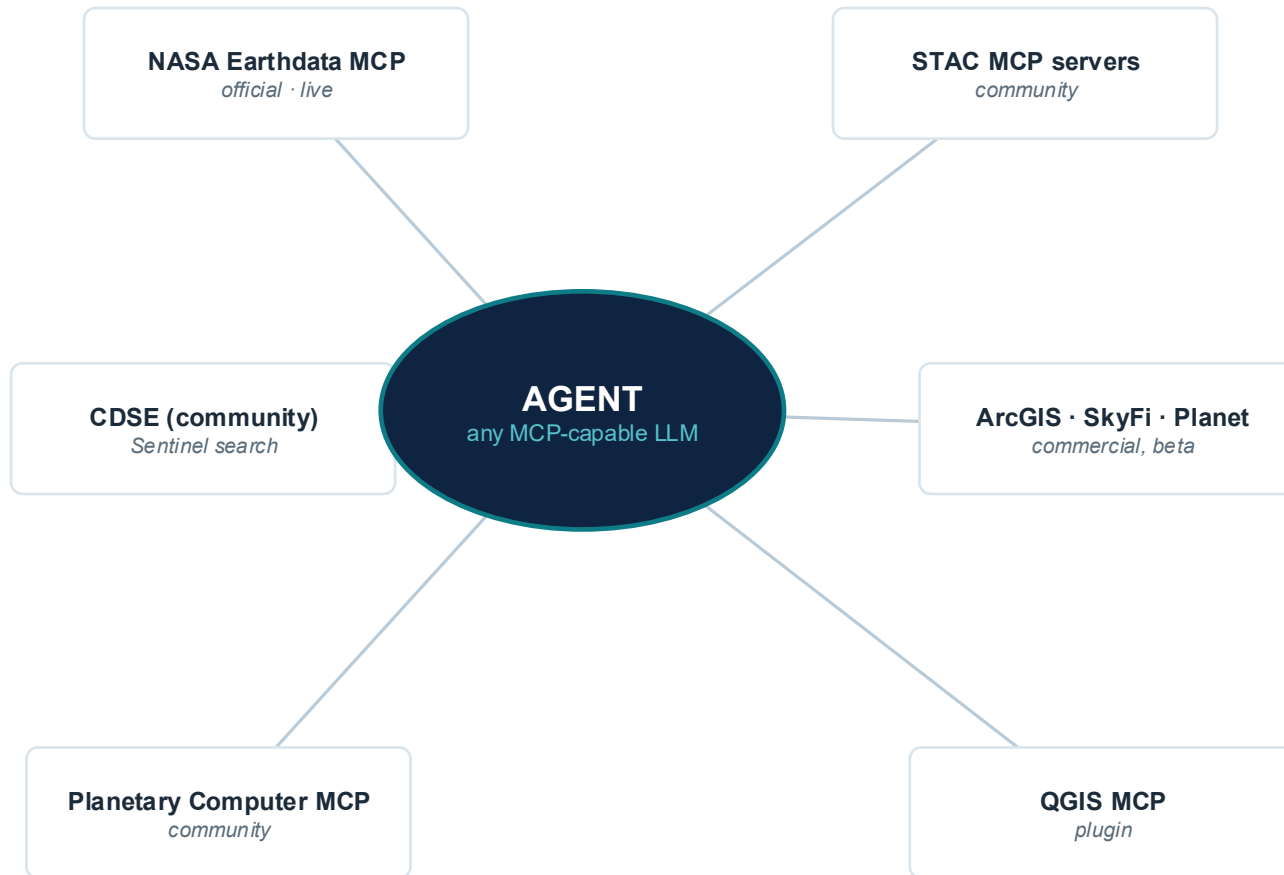
SGLang — big MoE models, multi-node

Ollama / llama.cpp — laptop-scale prototyping

(HF TGI is in maintenance mode — avoid for new work)

One GPU node $\approx$ a private, per-token-free agent brain — your prompts and your data never leave your infrastructure.

MCP — the “USB-C port” between agents and EO tools



Model Context Protocol

- An open standard for connecting agents to tools & data (born at Anthropic, Nov 2024)
- Since Dec 2025 governed by the Linux Foundation's Agentic AI Foundation
- Write a tool definition once — any MCP-capable model can use it
- **77+ geospatial MCP servers already catalogued (Sparkgeo, Aug 2026)**

A TOOL, AS THE AGENT SEES IT

```
{ "name": "search_imagery",
  "description": "Find scenes in a STAC catalogue",
  "inputSchema": { "properties": {
    "bbox": { "type": "array",
    "datetime": { "type": "string",
    "max_cloud": { "type": "number" } } } }
```

Shortcuts — high-level tooling that exists today

geoai (opengeos) *MIT*

Detection/segmentation pipelines, ~20 RS foundation models, agent-powered interactive mapping, QGIS plugin.

samgeo *MIT*

SAM 1/2/3 on georeferenced imagery — text-promptable segmentation straight to GeoPackage.

leafmap *MIT*

Interactive maps, STAC/COG browsing and results display in notebooks.

QGIS agents *research-grade*

IntelliGeo (U. Twente) · GIS Copilot (Penn State): natural language → desktop GIS workflows.

Curated maps of the field *open*

“Agentic AI in RS” survey + reading list (Jan 2026) · Sparkgeo geo-MCP index.

Maturity, honestly *caveat*

Research-grade and moving fast: pin versions, expect rough edges, read the issues.

Net effect: a capable agentic-EO prototype is days of work now — not months.

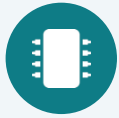
Where HPC fits — and how you actually get it

Three jobs for a supercomputer



Serve open-weight LLMs privately

vLLM on GPU nodes: data governance + no per-token costs



Fine-tune EO foundation models

TerraTorch across multi-GPU nodes, at LoRA-scale budgets



Scale the analysis itself

Dask over national or continental AOs, petabyte archives

The access ladder — free for research

1

IT4I Open Access — 3 calls/year for Czech research: Karolina GPUs, Barбора NG — and the Czech share of LUMI

2

EuroHPC JU calls — EU-level, peer-reviewed — incl. a dedicated “AI & Data-Intensive Applications” track

3

AI Factories — LAIF — LUMI AI Factory (Czechia is a partner via IT4I/VSB-TUO): Playground access in ~2 days, Fast Lane ≤ 50k GPU-h, free for academia & SMEs. Plus CZAI — the Czech AI Factory (launched 2026, “KarolAIa” system incoming)

The LAIF Hub Ostrava is open at IT4I

Three walkthroughs, step by step

Static, reproducible narratives — no live calls, no magic. What an open agentic EO system would actually do.



A · Forest change

optical time series, change detection,
trend



B · New solar farms

object detection + verification against
OSM



C · Flood impact

SAR through the clouds + land-cover
overlay

How to read the next slides: every step names a real open tool · every figure is an illustrative schematic · every number is invented for teaching (and marked so). *The pattern is real; the pixels are not.*

The request, the plan, the calls



“How much forest was cleared in NW Rondônia between July 2023 and July 2026? Map it, give me area per year — and tell me how sure you are.”

THE AGENT'S PLAN

- 1 Resolve AOI to a polygon *OSM / gazetteer*
- 2 Find scenes: S2 L2A, July windows, cloud < 20% *pystac-client* → *CDSE*
- 3 Median composites for 2023 & 2026 *odc-stac + xarray*
- 4 dNDVI < -0.25 on forest baseline → patches ≥ 1 ha *numpy + GDAL*
- 5 Areas per year, map, report with caveats *GeoPandas + matplotlib*

AGENT TRACE (ABRIDGED — VALUES ILLUSTRATIVE)

```

PLAN 5 steps → approved (auto: read-only tools)
CALL stac.search(s2-l2a, bbox=[-64.3,-9.9,-63.1,-9.0],
  2023-06/2023-08, cloud<20)
  ↳ 96 items · 19 usable
CALL stac.search(..., 2026-06/2026-08, cloud<20)
  ↳ 118 items · 22 usable
CALL composite(median) × 2 ↳ 2 cloud-free mosaics
CALL ndvi_diff(thr=-0.25, min_patch=1 ha)
  ↳ 214 patches · 186 km² → loss_polygons.gpkg
CHECK cloud-gap coverage = 4.2% of AOI
  ↳ recorded as uncertainty in the report
CALL zonal_stats(by=year) → drafting report...
  
```

Note the CHECK step: a good agent budgets for its own uncertainty — it was planned, not improvised.

Real Landsat evidence: repeat observation makes change measurable



2005 · Machadinho d'Oeste, Rondônia

The expanding 'fishbone' road-and-clearing pattern is visible in the real Landsat record. This image is evidence of land-cover change—not the pseudo-demo's invented 2023–26 result.



1984 → 2005: what the pixels establish

- Repeat observations expose spatial change
- The agent still needs a declared AOI, comparable dates and cloud handling
- Area estimates require a reproducible classification—not visual counting

Detection needs a verification design



“Which solar farms have been built within 30 km of Ostrava since 2022? Give me a list I can actually verify.”

THE AGENT'S PLAN

- 1 **Build a 30-km ring around the city** *GeoPandas buffer*
- 2 **Composites for 2022 and 2026** *STAC + odc-stac*
- 3 **Screen candidates: spectral + geometric signature** *raster ops*
- 4 **Segment & measure each candidate** *samgeo (SAM)*
- 5 **Cross-check OSM; rank by confidence; queue reviews** *Overpass API*

AGENT TRACE (ABRIDGED — VALUES ILLUSTRATIVE)

```
CALL buffer(city_boundary, 30 km) ↳ ring AOI
CALL stac.search + composite for 2022, 2026
CALL screen(low_NDVI + blue-grey + regular grid)
  ↳ 31 raw candidates
CALL sam_segment(31 candidates)
  ↳ 17 panel-like · 14 rejected by shape
CALL osm_check(power=plant, source=solar)
  ↳ 7 matched · 5 high-score unmatched
  ↳ 5 rejected (greenhouses, gravel pads)
PLAN output = ranked list + human review queue

# values illustrative
```

Teaching point: the agent plans its own cross-checks — detection without a verification design is just guessing at scale.

A real open workflow already maps photovoltaic farms from Sentinel-2



ESA APEx example · openEO process

Data	Execution	Training	Output	Licence
Sentinel-2 L2A	openEO	OSM polygons	PV mask	CC BY 4.0
Time series	Cloud masking	Open data	GeoTIFF	Public
Feature extraction	Process graph	Pixel models	netCDF	Reproducible
Known limit	Validated ≤20×20 km	Proof of concept	Artifacts remain	Read docs



Open data + open process ≠ automatic generalisation

- This is the kind of callable tool an agent can orchestrate. Its documented scale limit and known artifacts must travel into the final report.

When the agent must know physics



“Estimate the flooded area along the Odra during the September 2024 event — and how much farmland was affected.”

Real context: Storm Boris, Sep 2024 — rapid-mapping teams did this manually

THE AGENT'S PLAN

- 1 Define event window + river-reach AOI *gazetteer + hydro data*
- 2 Sentinel-1 GRD pairs, pre & post event *pystac-client → CDSE*
- 3 Backscatter drop → water mask *thresholding (Otsu)*
- 4 Subtract permanent water *JRC Global Surface Water*
- 5 Overlay land cover & settlements → stats + report *ESA WorldCover, GHSL*

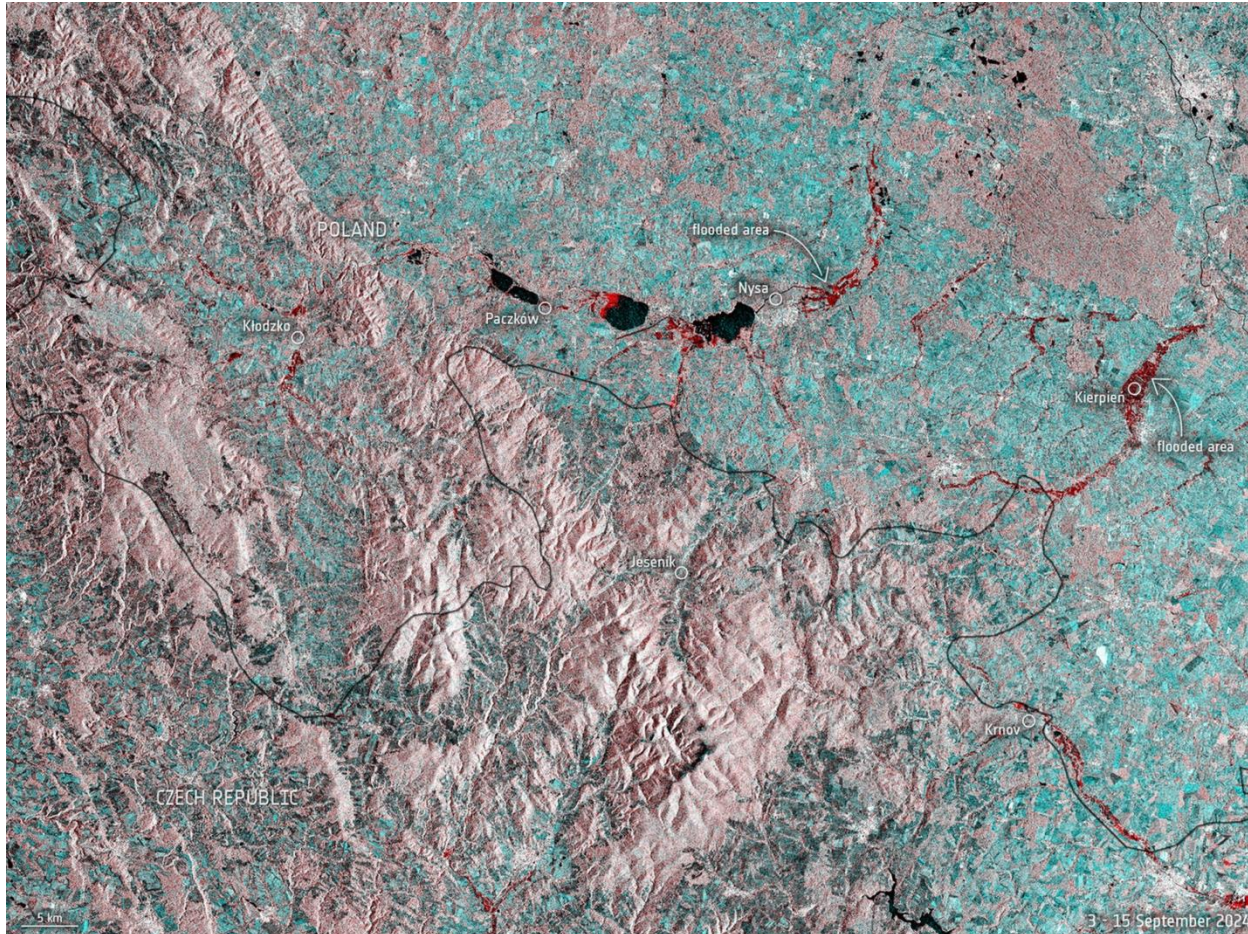
AGENT TRACE (ABRIDGED — VALUES ILLUSTRATIVE)

```

PLAN optical unusable (storm clouds) → use SAR
CALL stac.search(sentinel-1-grd, pre: 2024-09-08,
post: 2024-09-16..20, same orbit)
  ↳ 3 pre + 4 post scenes, VV+VH
CALL water_mask(backscatter_drop, otsu)
CALL subtract(jrc_permanent_water)
  ↳ flood_extent.gpkg
CALL overlay(worldcover, ghsl_settlements)
CHECK urban SAR caveat: flooded streets under-
detected between buildings → flag in report
  
```

The agent must know SAR beats optical in a storm. That knowledge lives in tool descriptions and system prompts — engineered, not hoped for.

Sentinel-1 saw Storm Boris through cloud and rain



3 Sep

pre-event Sentinel-1

15 Sep

post-event Sentinel-1

RED

areas affected by flooding

WHAT THE REAL IMAGE PROVES

sensor choice is part of the reasoning

- Sentinel-1 radar observes through cloud and rain
- Multi-temporal comparison reveals change
- Flood extent still needs thresholds, permanent-water masks and validation

So — what could go wrong? That's the last section. →

Reality check: agents still fail EO benchmarks

33%

of 140 real Earth-science questions answered correctly by state-of-the-art agents writing Earth-Engine code — and >58% of code runs failed.

arXiv 2504.12110 · 2025

>84%

of numerical answers fell outside acceptable tolerance — across ALL frontier models — on 403 executable Earth-system tasks.

TerraBench · 2026

60.8%

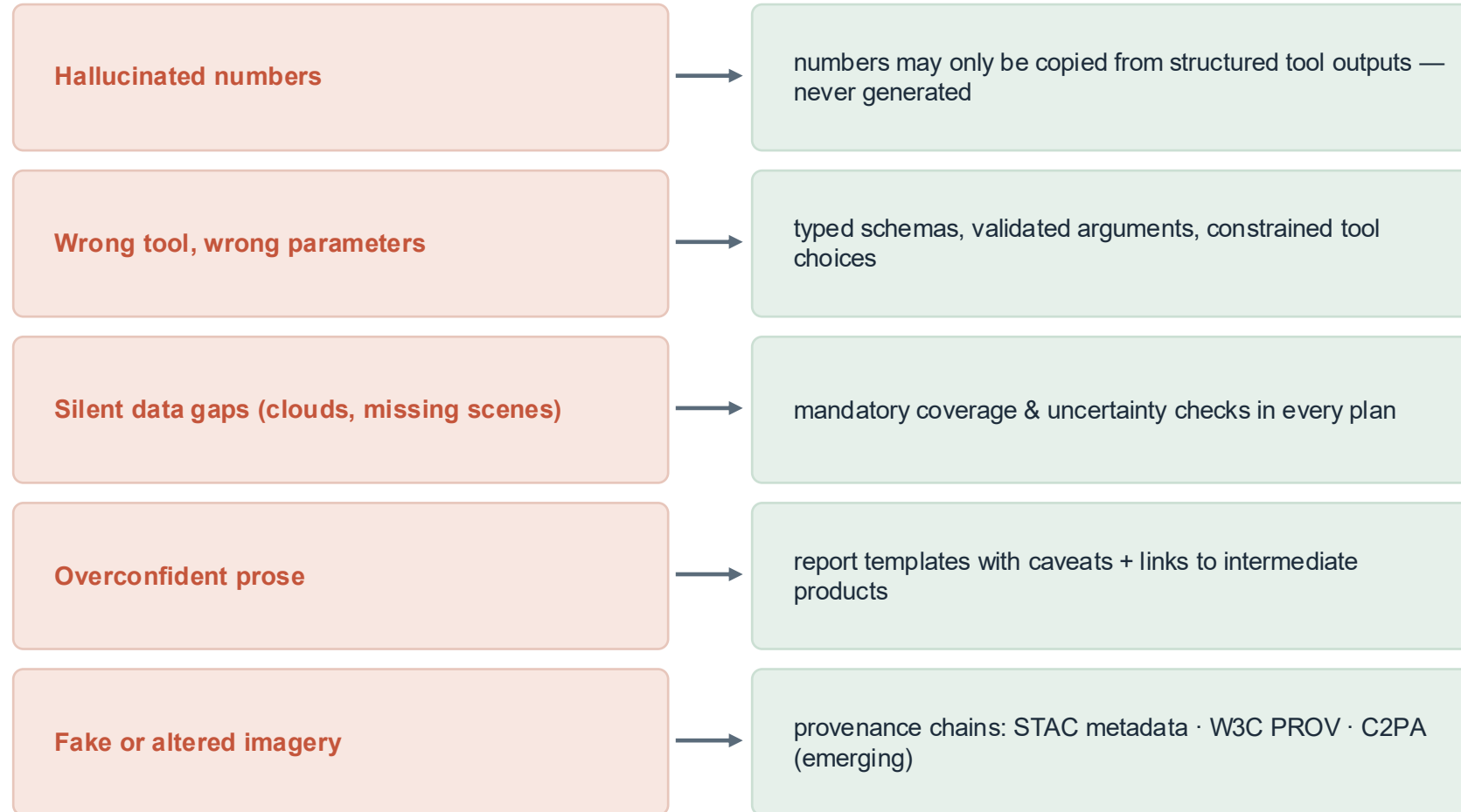
best model on a production-style environmental-API benchmark. Notably: open-weight models led the cost–accuracy frontier.

GeoNatureAgent · 2026

Demos ≠ deployments. Step-level audits (ThinkGeo, CVPR 2026) show the bottlenecks are tool selection and numeric grounding — not eloquence.

Open, domain-specific evaluation is the research frontier — and it is wide open.

Failure modes → engineering countermeasures

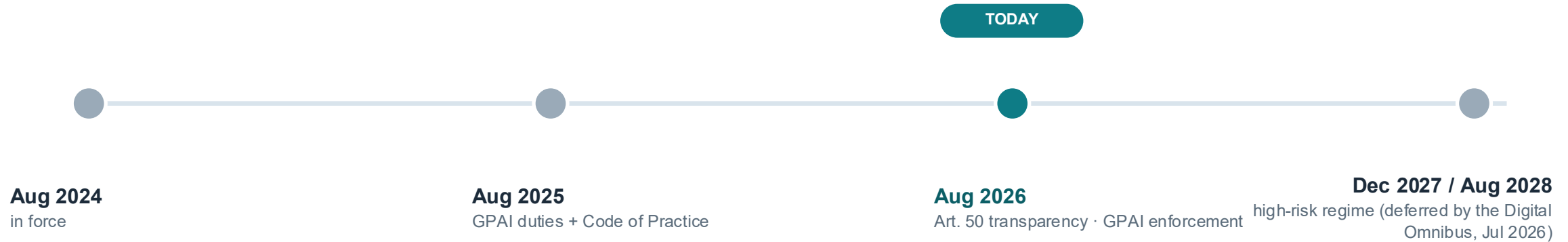


Cautionary tale — July 2026

Google added generative imagery to Google Earth. Within hours, researchers produced fake scenes of sensitive sites. The feature was pulled in a day, pending “stronger guardrails”.

In agentic EO, provenance is not optional.

EU AI Act: what applies now — and what to do meanwhile



No agent-specific rules (yet)

- Agents inherit duties indirectly: via the GPAI model that powers them, via transparency (Art. 50) — and later via the high-risk regime
- Public-service decision support (floods, compliance...) will likely be high-risk when Annex III applies (Dec 2027)
- Neither the GPAI Code of Practice nor the guidelines mention agents — soft law and standards will move first

Design to the spirit, today

- Human oversight (Art. 14): review queues before publication
- Automatic logging (Art. 12): keep the full agent trace
- Disclosure (Art. 50): say when users talk to a machine
- Provenance: every product traceable to scenes + code

For public-sector EO decision support: build high-risk-grade discipline before it binds.

Further reading: OECD “Agentic AI Landscape” (2026) · IAPS “AI Agent Governance: A Field Guide” · International AI Safety Report 2026

An open research agenda — where you come in



Evaluation

EO-agent benchmarks, error taxonomies, uncertainty scoring — the field's biggest gap and fastest publication route.



Reliability engineering

Constrained tool use, self-verification steps, uncertainty propagation through multi-step pipelines.



Domain agents

Fine-tuned open models for EO tool use: an 8B model matched much larger ones after fine-tuning (2025).



Trustworthy deployment

Human-in-the-loop patterns, provenance chains, AI-Act-ready logging — with real public-sector pilots.

Support exists: compute & consulting via LAIF, CZAI and IT4I Open Access · community at the ESA × BIFOLD “Agentic AI for EO” workshop (Berlin, 19–21 Oct 2026) · open reading lists to start from

Four takeaways

1

Agentic = a loop, not a model — planning + tools + memory + verification around an LLM orchestrator. The LLM never does the science alone.

2

The open stack is complete — Copernicus data → STAC/openEO → Python tools + open foundation models → MCP → open-weight LLMs on EuroHPC.

3

Reliability is the frontier — impressive demos, sobering benchmarks. Verification and evaluation are where research wins.

4

Govern by design — human oversight, automatic logs, provenance — from day one. Regulation will catch up to you.

Let's discuss

Q1 Which recurring EO question in your work would you hand to an agent first?

Q2 What evidence would it take for you to trust an agent-made map in an official report?

Q3 Which decisions must stay human — even if the agent is usually right?

After the session

Slides + an annotated link list by e-mail. The session was recorded.

Contact

Karol Bot Gonçalves
IT4Innovations National
Supercomputing Center

Thank you.

Selected sources — systems, data & tools

Planet — agentic EO

[planet.com/pulse: “How Agentic AI Transforms EO...”](#) (21 May 2026) · [Planet × Anthropic](#) (6 Mar 2025) · [Planet MCP server beta](#) (12 Jan 2026)

Google

[Geospatial Reasoning](#) ([research.google](#), Apr 2025) · [AlphaEarth Foundations / Satellite Embeddings](#) ([DeepMind](#), Jul 2025) · [Earth AI](#) (Oct 2025)

NASA × Microsoft

[Earth Copilot](#) (Nov 2024) · [Earth Copilot 2.0 coverage](#) (Nov 2025) · [Planetary Computer](#) (free STAC catalogue)

ESA

[Φ-lab DA4DTE assistant](#) (Mar 2025) · [TerraMind with IBM](#) (Apr 2025, Apache-2.0) · [EarthCODE](#) · [ESA × BIFOLD](#) [Agentic-EO workshop](#), Berlin (Oct 2026)

Standards

[STAC 1.1 — OGC Community Standard](#) (Oct 2025) · [openEO — OGC Community Standard](#) (May 2026) · [MCP — Linux Foundation / AEIF](#) (Dec 2025)

Open data & APIs

[dataspace.copernicus.eu](#) (STAC · openEO · S3; EOPF Zarr) · [planetarycomputer.microsoft.com](#) · [earth-search.aws.element84.com](#) · [cmr.earthdata.nasa.gov/stac](#)

Foundation models

[Prithvi-EO-2.0](#) ([huggingface.co/ibm-nasa-geospatial](#)) · [TerraMind](#) ([ibm-esa-geospatial](#)) · [Clay](#) ([clay-foundation](#)) · [SAM/samgeo](#)

GeoAI tooling

[opengeos: geoai](#) · [samgeo](#) · [leafmap](#) ([opengeoai.org](#)) · [TerraTorch & TorchGeo](#) ([github.com/torchgeo](#)) · [IntelliGeo](#) · [GIS Copilot](#) (QGIS plugins)

Agent stack

[LangGraph](#) · [smolagents](#) · [Pydantic AI](#) · [vLLM](#) · [SGLang](#) · [Ollama](#) · [geospatial MCP index: github.com/sparkgeo/geo-mcp-servers](#)

Data-volume stats

[eu-space.europa.eu](#) (22 TB/day) · [Nature Sci. Data](#) (CDSE, Feb 2026) · [NASA JPL](#) (245 PB, NISAR) · [USGS EROS](#) (Landsat archive)

Full clickable link list ships with the slides.

Selected sources — research & governance

Surveys & lineage

“Agentic AI in Remote Sensing” survey — arXiv 2601.01891 (Jan 2026, WACV-W) · LLM-Geo / Autonomous GIS — arXiv 2305.06453 · GIS Copilot — arXiv 2411.03205

Benchmarks

Towards LLM Agents for EO — arXiv 2504.12110 (33% / >58%) · TerraBench — arXiv 2606.13148 (>84% numeric misses) · GeoNatureAgent — arXiv 2606.12821 (60.8%; open-weight Pareto) · ThinkGeo — arXiv 2505.23752 (CVPR 2026) · Earth-Agent/Earth-Bench — arXiv 2509.23141 (ICLR 2026)

VLM limits

“Good at captioning, bad at counting” (VLEO-Bench) — arXiv 2401.17600 (CVPR-W 2024)

EU AI Act

Regulation (EU) 2024/1689 · GPAI Code of Practice (10 Jul 2025) · Digital Omnibus on AI — Reg. (EU) 2026/1744 (OJ 24 Jul 2026): Annex III → 2 Dec 2027, Annex I → 2 Aug 2028 · artificialintelligenceact.eu (Art. 12, 14, 50)

Agent governance

OECD “The Agentic AI Landscape” (Feb 2026) · IAPS “AI Agent Governance: A Field Guide” (arXiv 2505.21808) · International AI Safety Report 2026 · Chan et al., “Visibility into AI Agents” (FAccT 2024)

EO quality & provenance

QA4EO (CEOS) uncertainty framework · W3C PROV-O · C2PA Content Credentials (emerging for imagery) · conformal prediction for EO — Sci. Rep. 2024

HPC access

it4i.cz/en/for-users/open-access-competition · eurohpc-ju.europa.eu (AI & Data-Intensive Access; AI Factories access modes) · lumi-ai-factory.eu (LAIF · Hub Ostrava)

Status of facts: verified August 2026. Re-check dates and versions before reuse.