

What's New in Computer Vision?

CVPR 2025 Edition



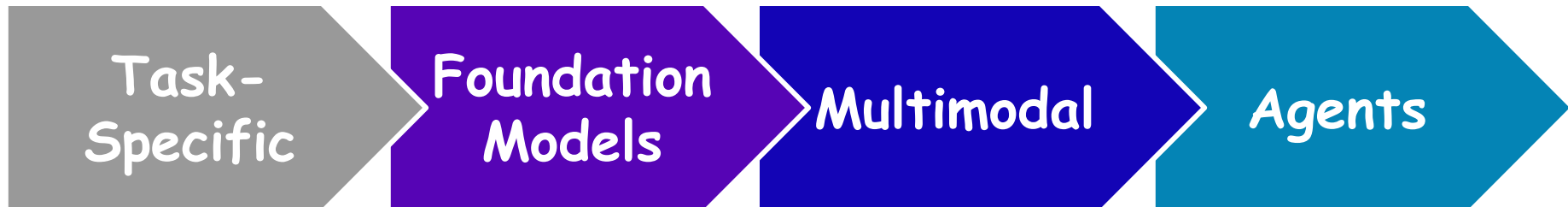
Heather Couture

June 26, 2025
11 am EDT

30 minutes + Q&A/Discussion



Today's Focus



Who am I?

- Heather Couture
- MS from Carnegie Mellon University - autonomous science
- PhD from University of North Carolina - computational pathology
- Computer vision consultant



- Keynote speaker at MICCAI workshop on computational pathology
- Contributor to Scientific American, The Pathologist, IEEE Spectrum
- Newsletter and podcast

Computer Vision Insights
by Pixel Scientia Labs



Task-Specific

Foundation
Models

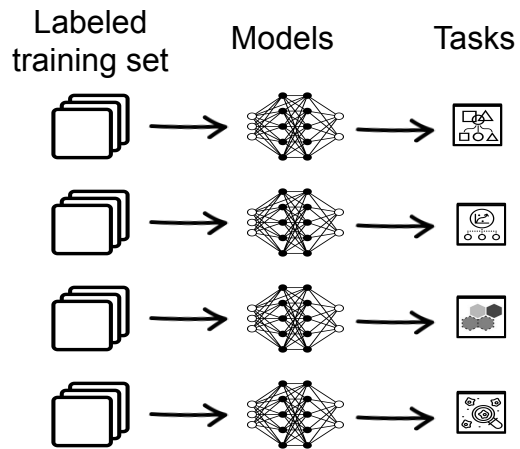
Multimodal

Agents

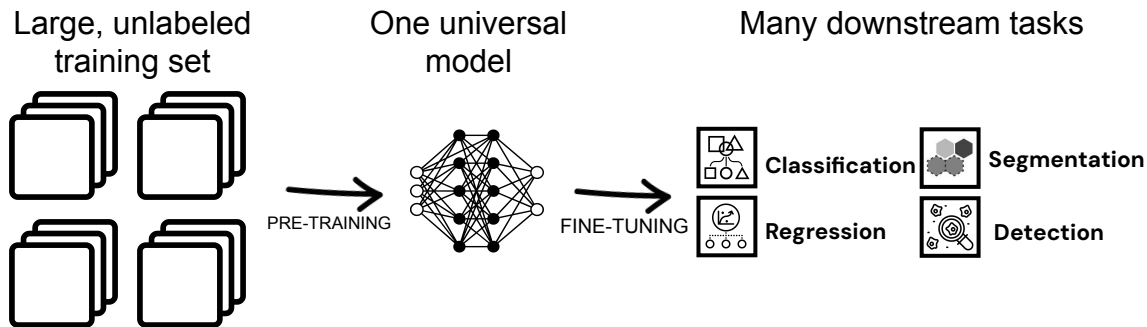
- New domains
- Domain-specific adaptations
- New datasets
- New benchmarks

The Paradigm Shift

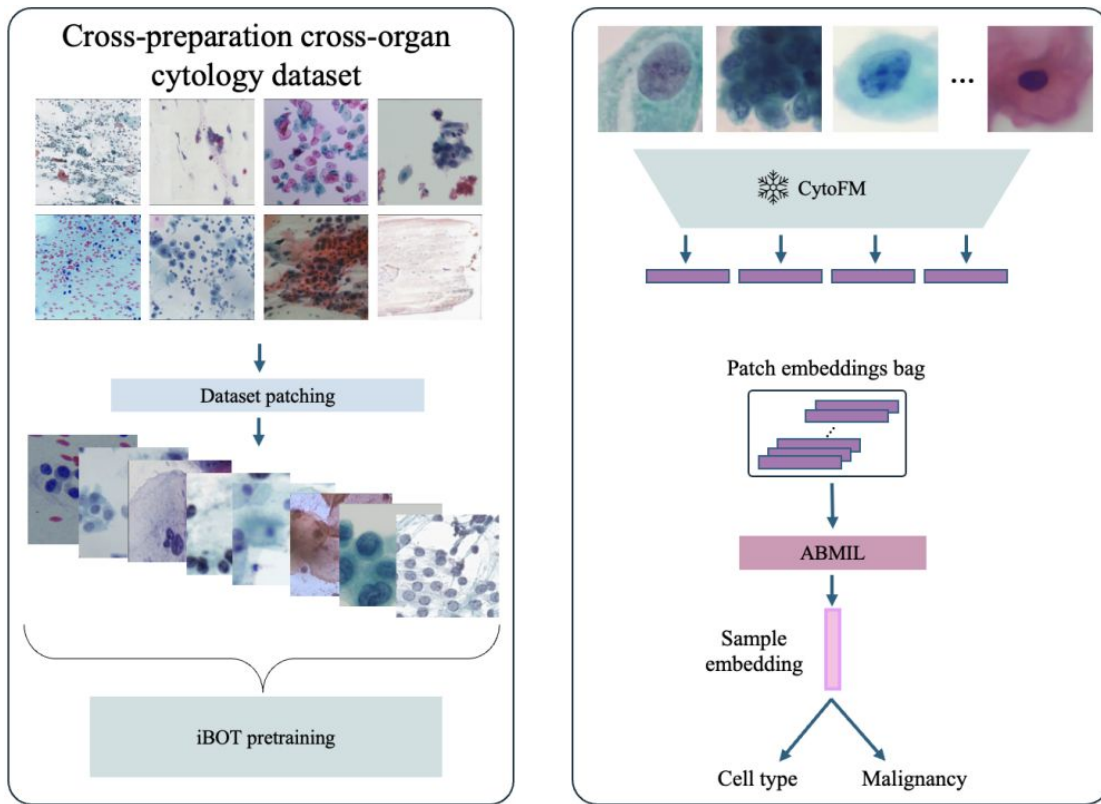
Traditional ML



Foundation Models



A Foundation Model for Cytology

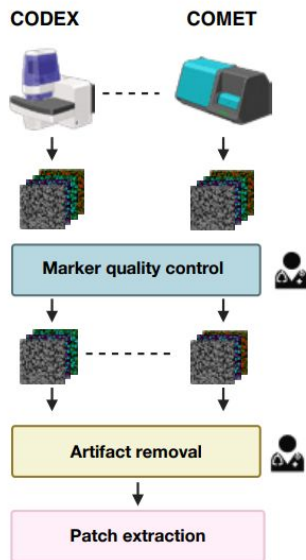


CytoFM

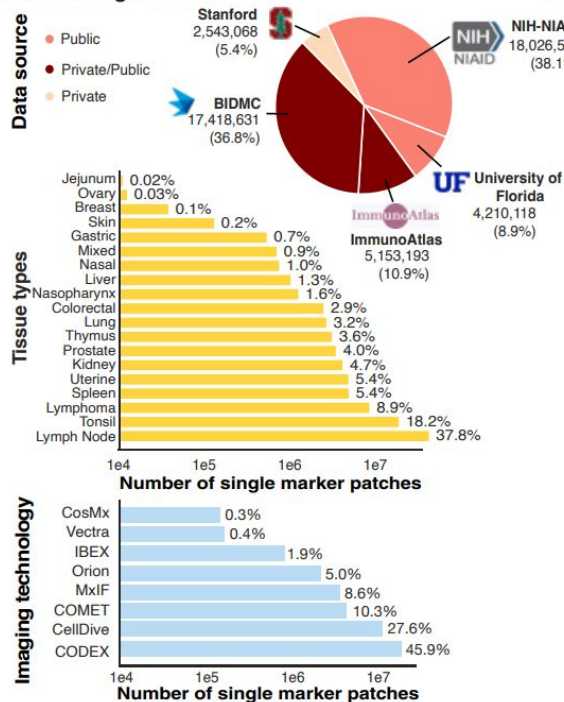
1.4 million image patches
8 datasets
7 institutions

A Foundation Model for Spatial Proteomics

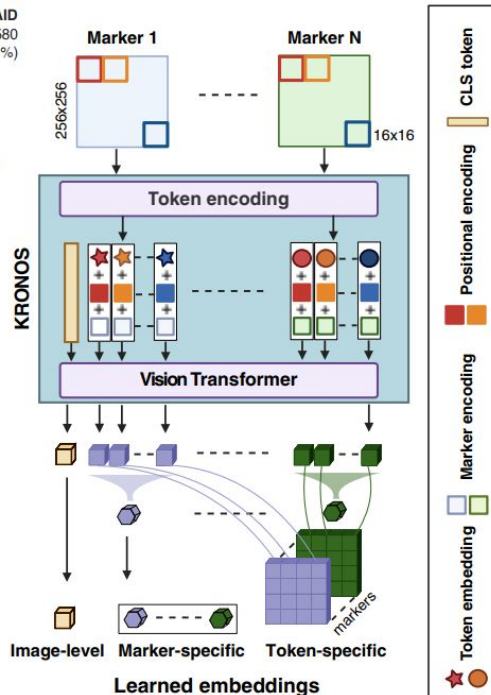
A Dataset curation



B Pretraining dataset



C Representation learning pipeline

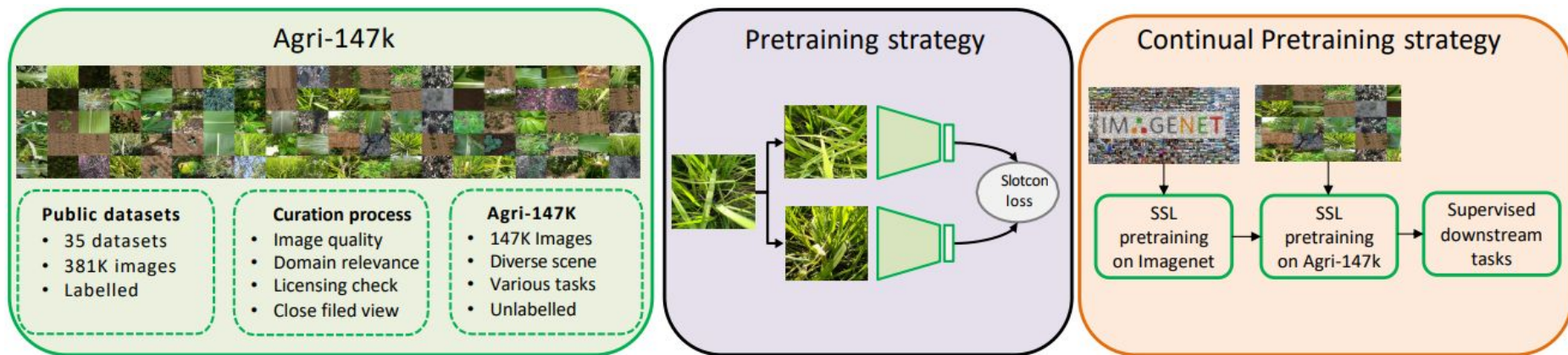


KRONOS

47 million image patches
175 protein markers
16 tissue types
8 imaging platforms

A Foundation Model for Agriculture

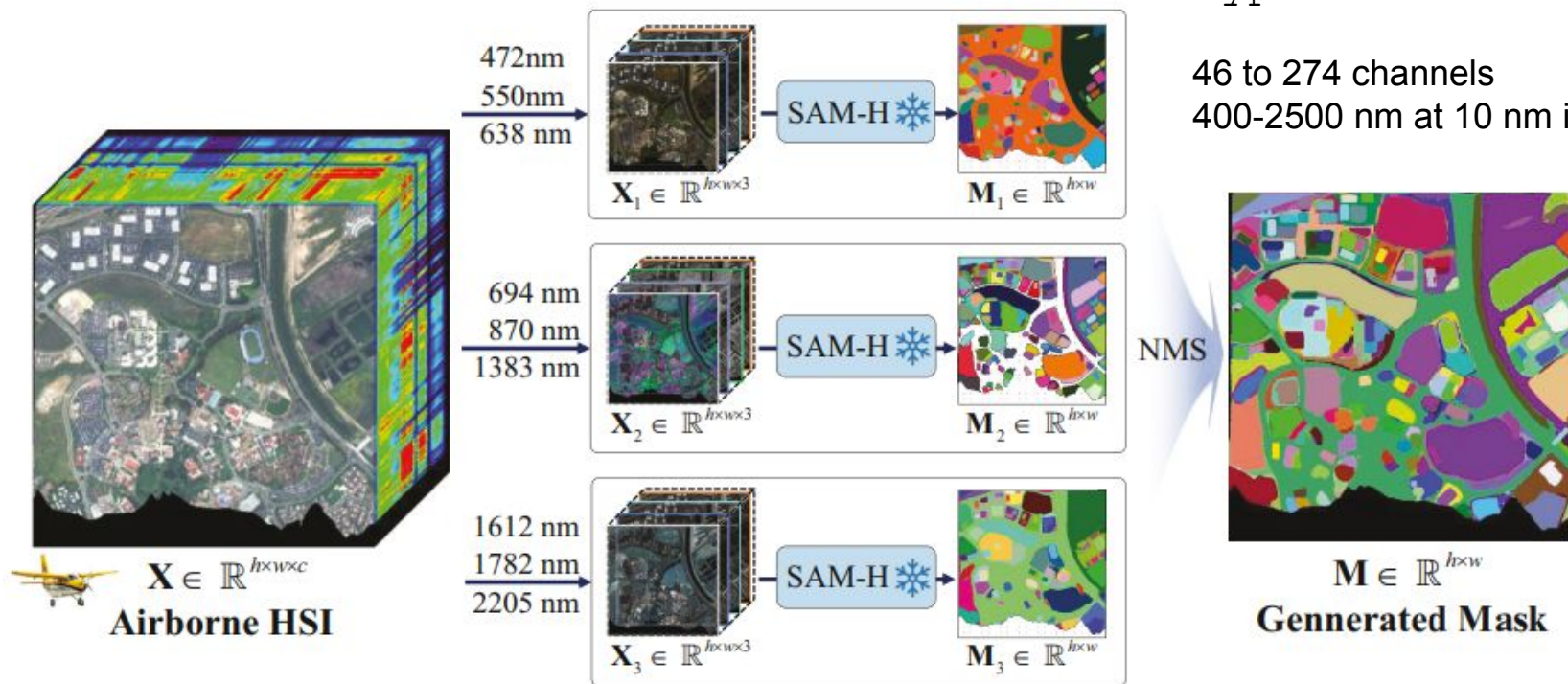
Agri-FM+



A Foundation Model for Hyperspectral Imagery

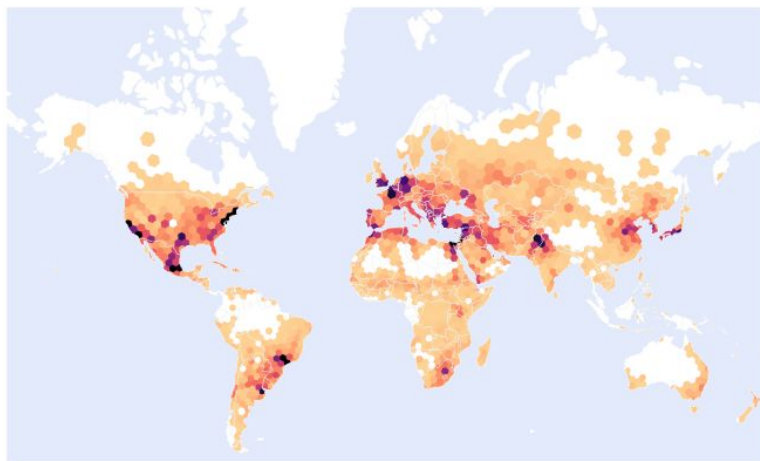
HyperFree

46 to 274 channels
400-2500 nm at 10 nm intervals

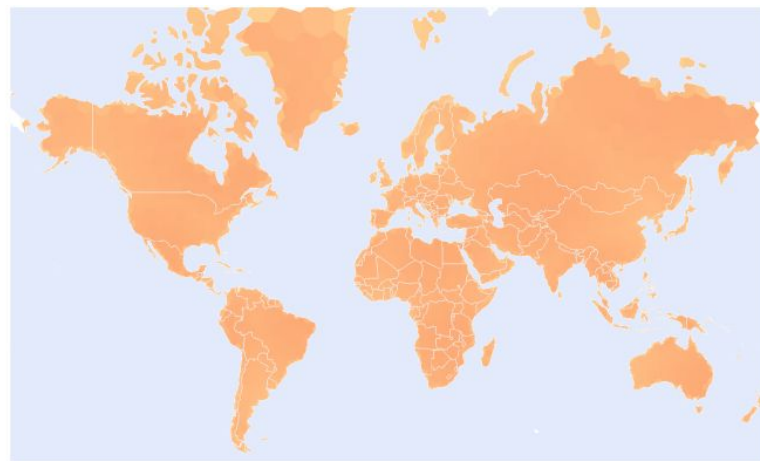


Li, HyperFree: A Channel-adaptive and Tuning-free Foundation Model for Hyperspectral Remote Sensing Imagery, 2025

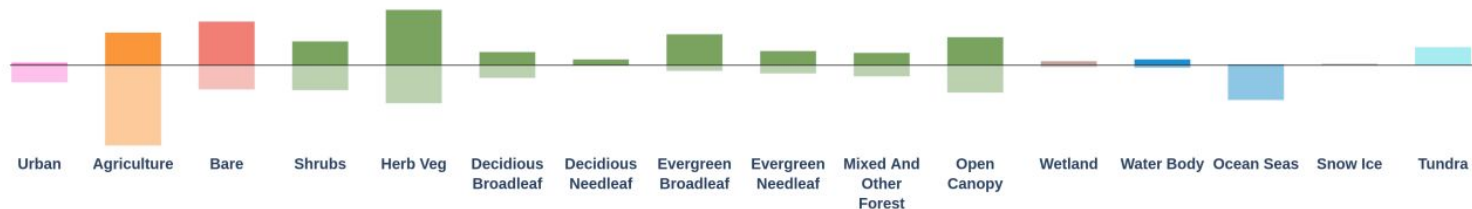
Remote Sensing Dataset: Global Distribution



(a) Spatial distribution of SSL4EO-S12 [89]

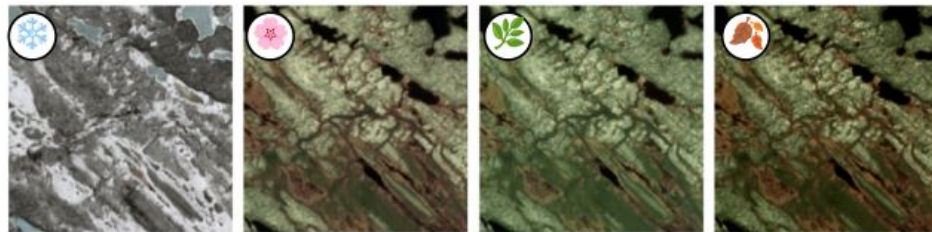
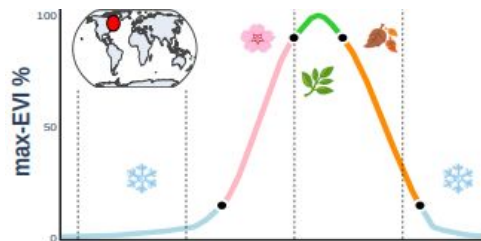


(b) Spatial distribution of SSL4Eco



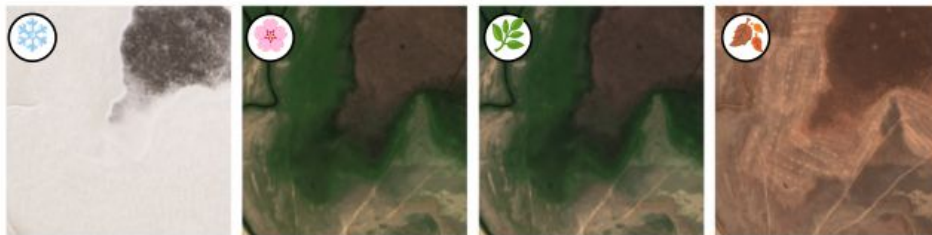
(c) Copernicus land cover [55] distribution for SSL4Eco (upwards) and SSL4EO-S12 [89] (downwards)

Remote Sensing Dataset: Seasonal Distribution

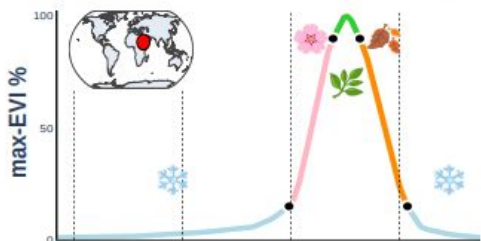


SSL4Eco

250k locations across
entire landmass



Enhanced Vegetation
Index-based seasonal
sampling

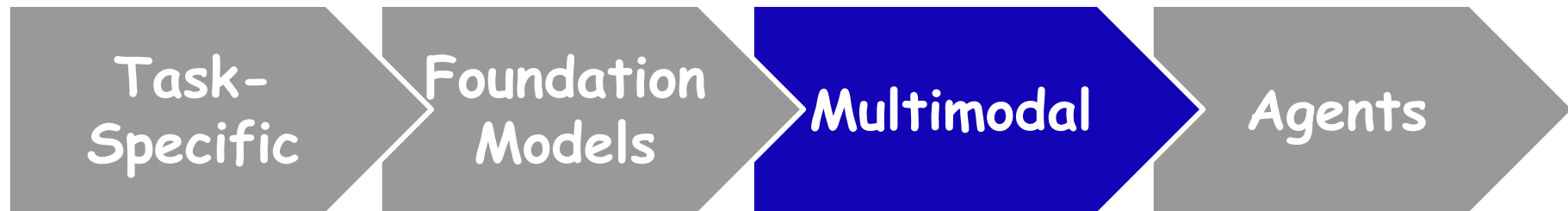


SeCo-Eco

Seasonal contrastive
learning

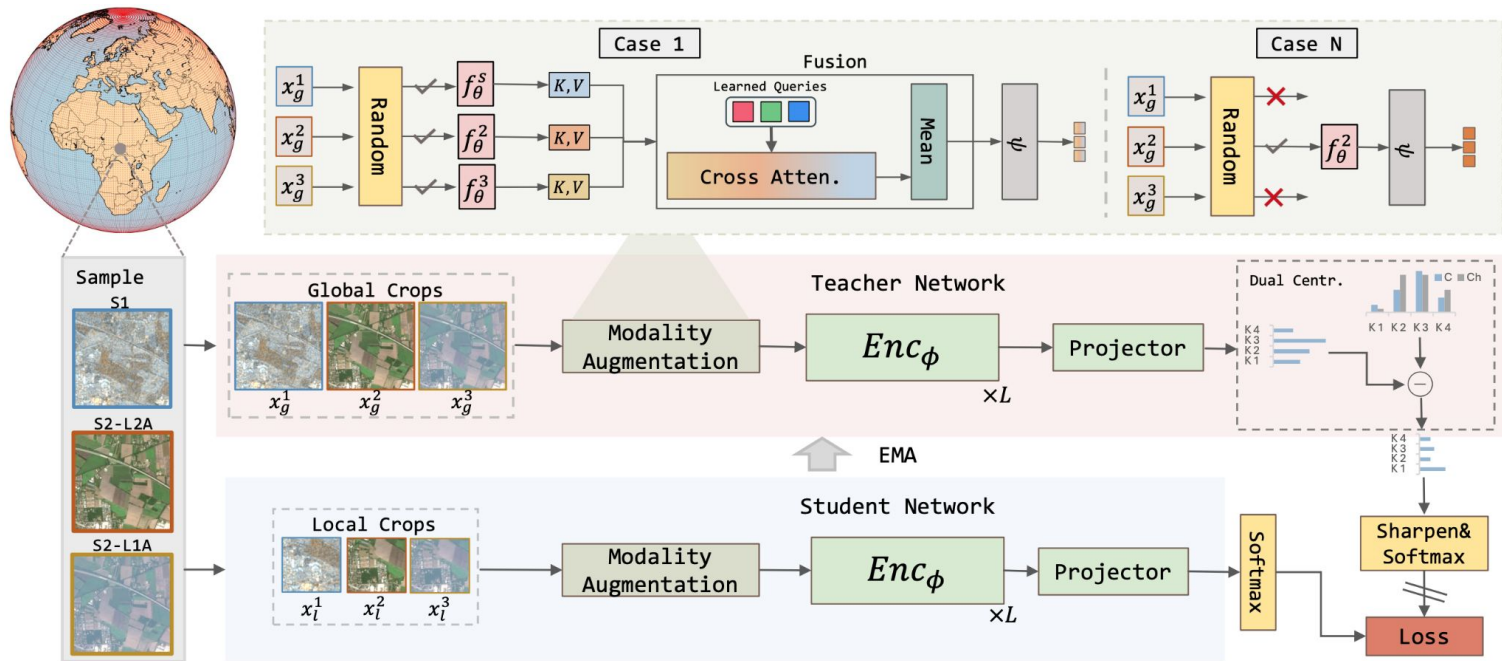
(a) EVI-based seasons

(b) Seasonal images



- Open weights and open data
- Grounding
- VLMs - language is the “glue”
- Multi-stage pretraining
- Larger inputs
- Benchmarks

Remote Sensing: Optical + Radar

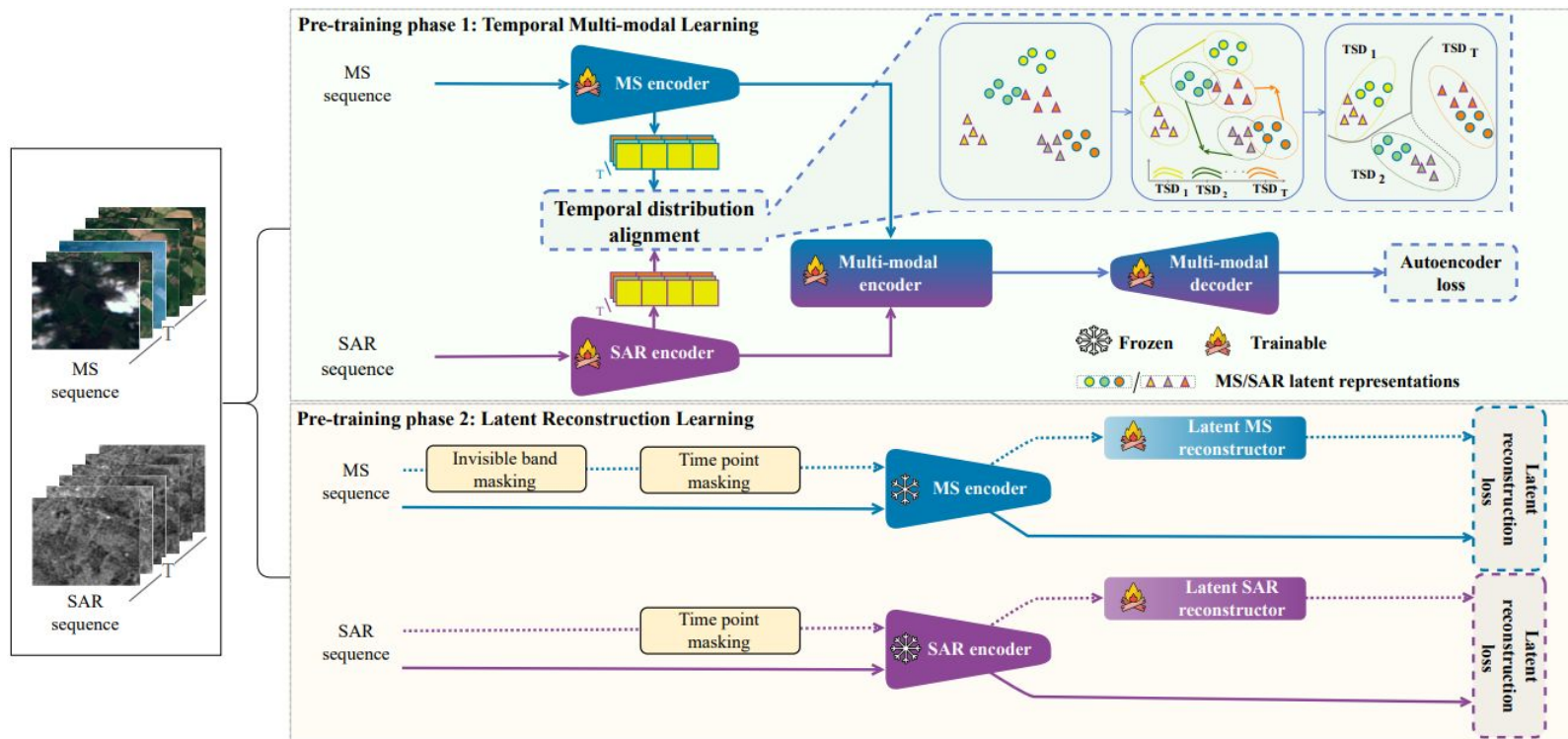


TerraFM

SAR and optical
18.7 million tiles
534x534 tiles

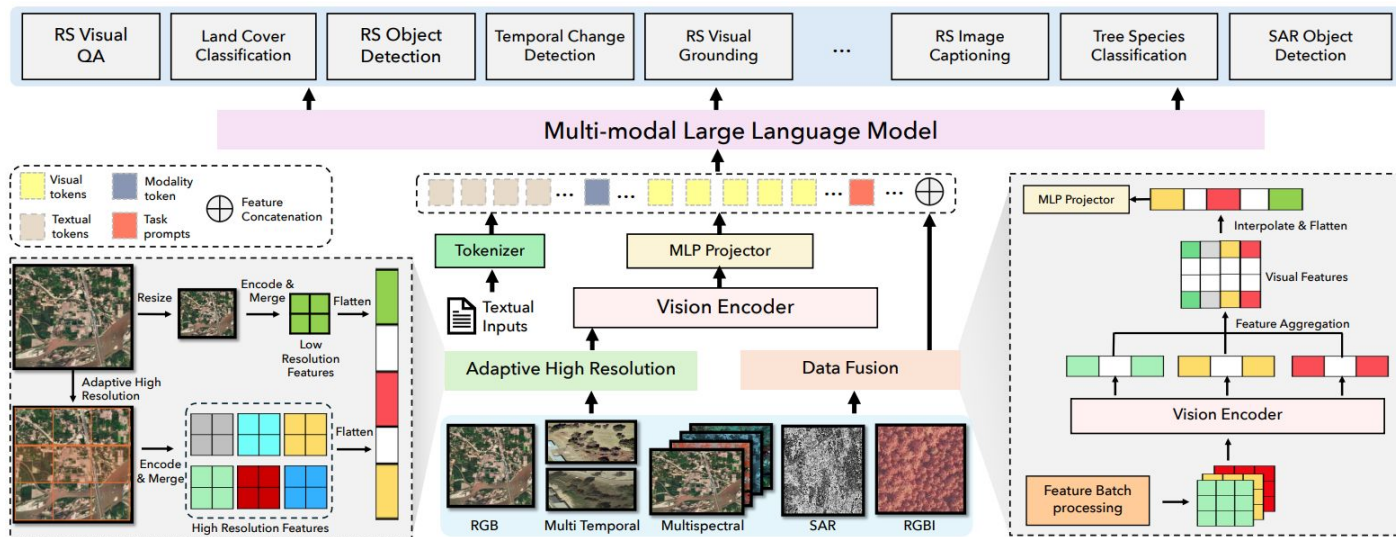
Remote Sensing: Robust to Missing Data

RoboSense



Do, RobSense: A Robust Multi-modal Foundation Model for Remote Sensing with Static, Temporal, and Incomplete Data Adaptability, 2025

A VLM for Remote Sensing



EarthDial

RGB, multispectral,
infrared, SAR
11 million instruction pairs

Soni, EarthDial: Turning Multi-sensory Earth Observations to Interactive Dialogues, 2025

Benchmarking VLMs for Remote Sensing

GEOBench-VLM

31 fine-grained tasks

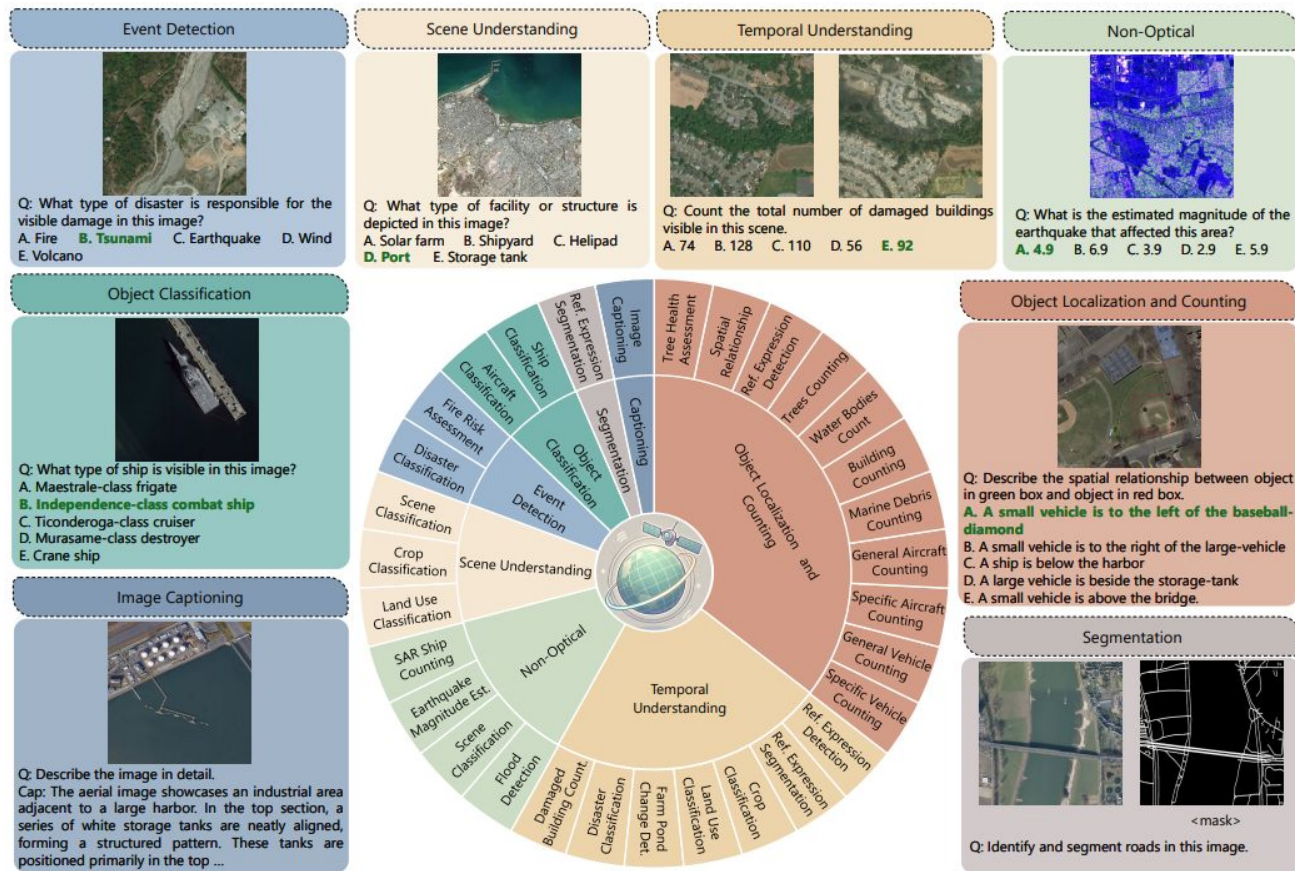
8 categories

Optical, multispectral, SAR,
temporal

10+ manually verified instructions

Multiple-choice format

Danish, GEOBench-VLM:
Benchmarking Vision-Language
Models for Geospatial Tasks, 2025



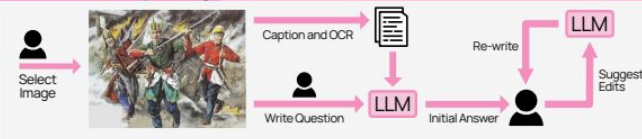
A VLM with Open Weights and Open Data

PixMo

Captions



AskModelAnything



Pointing

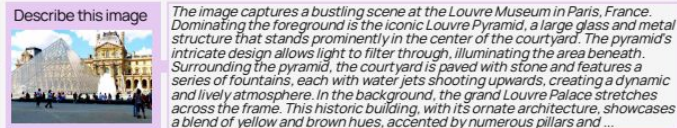


Synthetic



Molmo

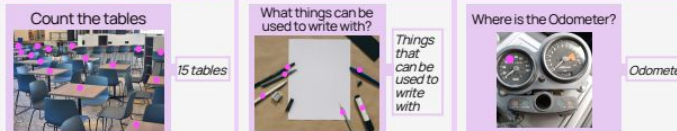
Fine-grained Understanding



User Interaction



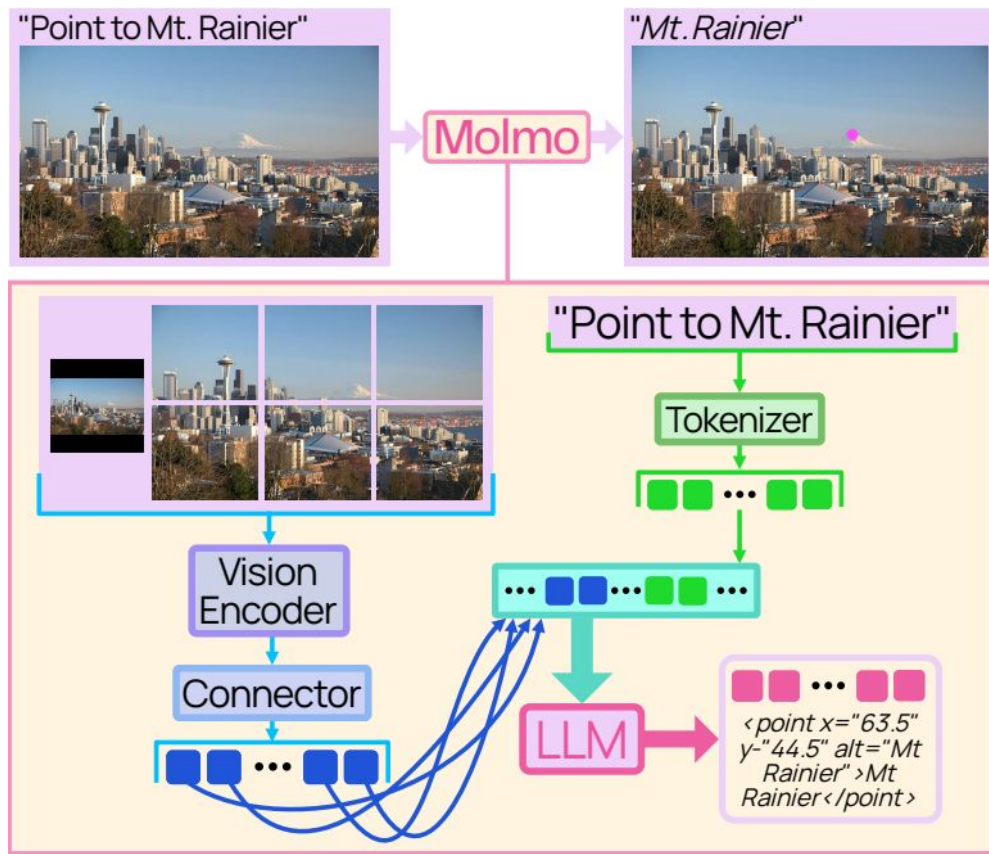
Pointing and Counting



Visual Skills

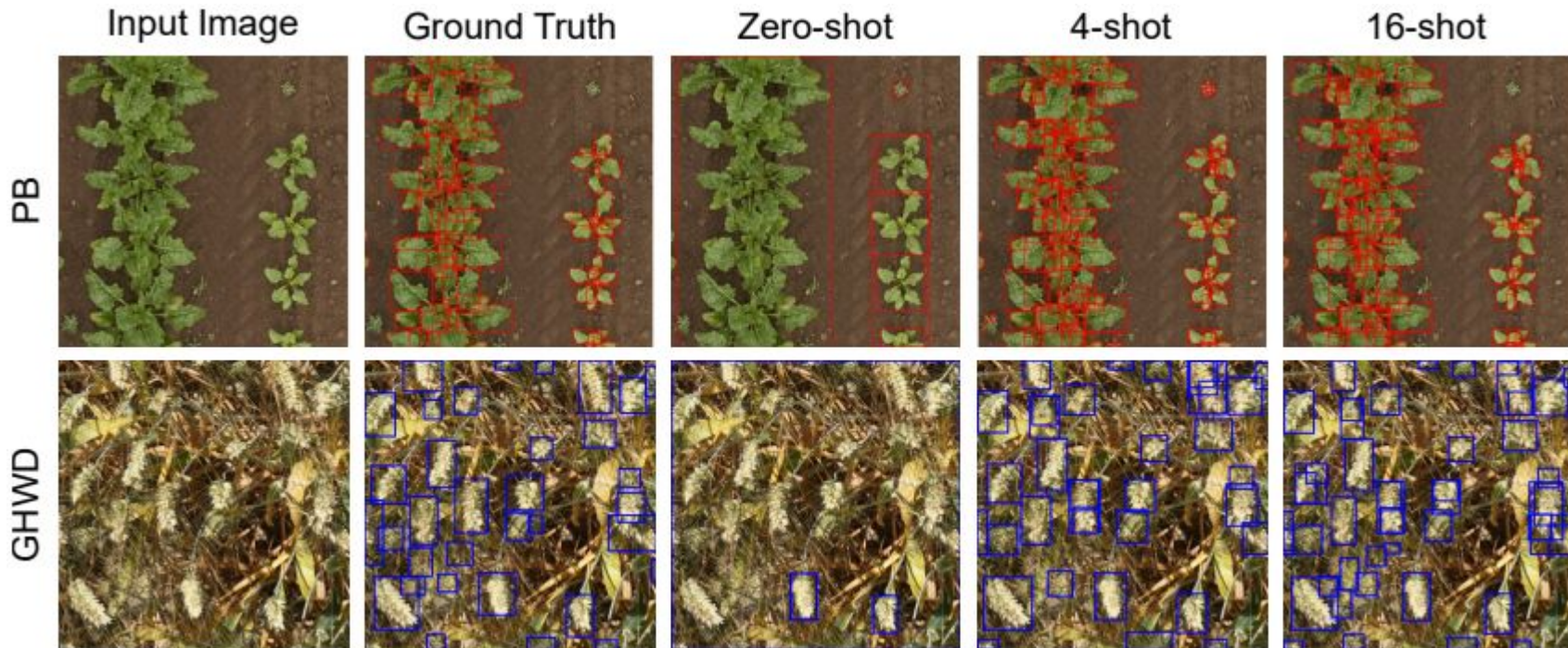


A VLM with Open Weights and Open Data + Grounding



Grounding for Agriculture

Prompt: wheat spike head



Remote Sensing VLM with Grounding

GeoPixel

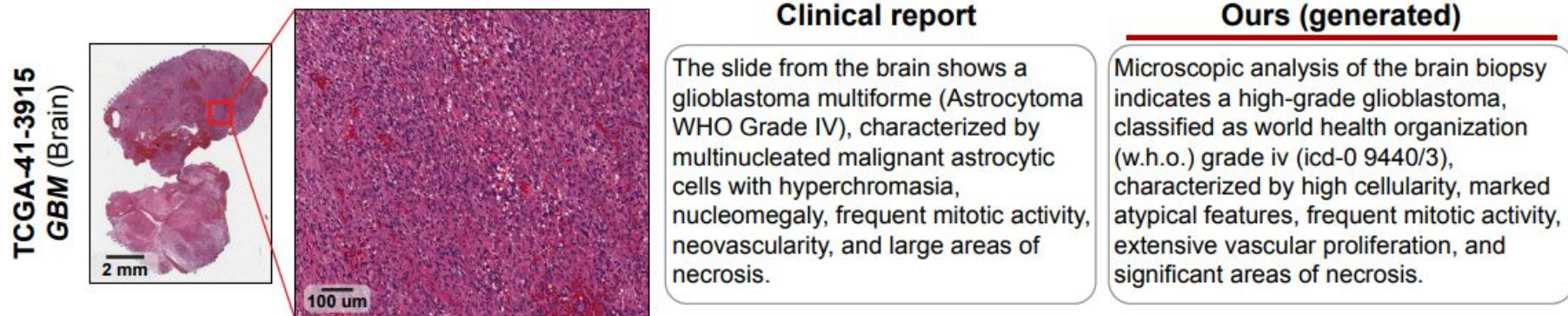
Can you give a thorough description of this image, including interleaved segmentation masks to highlight key objects?



The image is an aerial view of a sports complex featuring a running track and a swimming pool. The track has multiple lanes, and the field is marked for various sports. The surrounding area includes roads, buildings, and green spaces. The **ground track field** at the center is a large athletic venue with a multi lane running track encircling a grassy field marked for various sports. The **swimming pool** at the top is large, rectangular, and filled with a blue substance, likely water. **Numerous small vehicles** can be seen at different locations. A **soccer field** within the sports complex features a rectangular pitch with goals at each end, surrounded by a running track. The image gives a static and undisturbed appearance as no people or moving objects are visible.



A VLM for Whole Slide Images - Pathology Reports



Ding, Multimodal Whole Slide Foundation Model for Pathology, 2025

T I T A N

335k WSIs with 423k synthetic captions

8192 x 8192 tiles

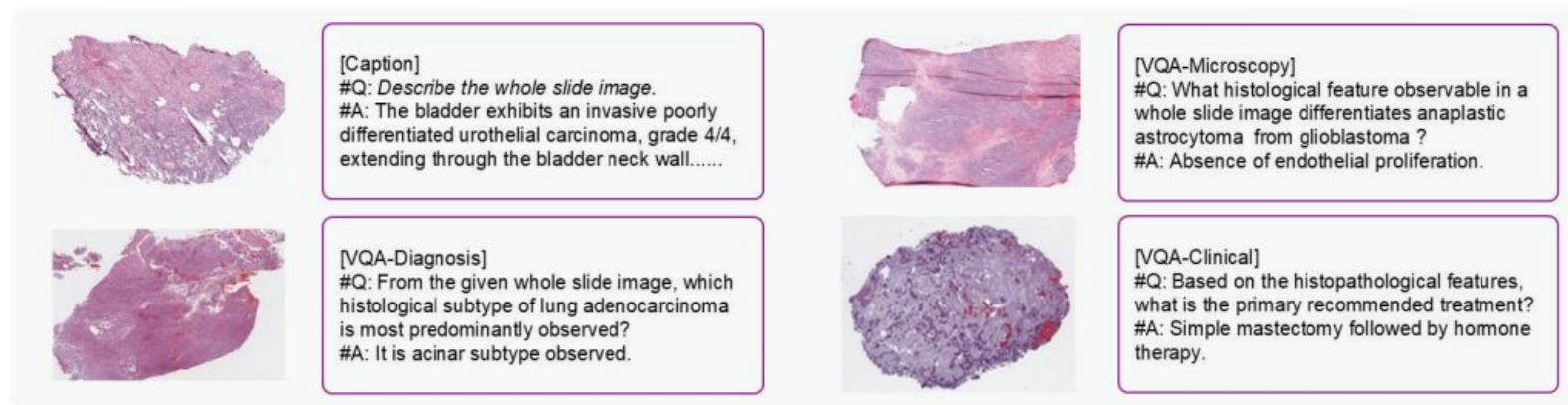
Three-stage training:

- SSL on patches

- Vision-language alignment

- Alignment with pathology reports

A VLM for Whole Slide Images - VQA



Chen, SlideChat: A Large Vision-Language Assistant for Whole-Slide Pathology Image Understanding, 2025

SlideChat

4k slide captions

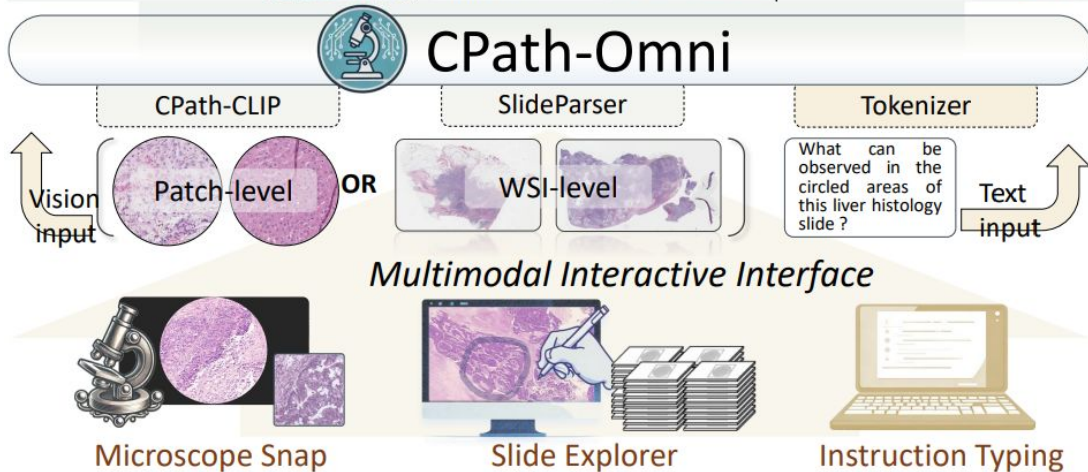
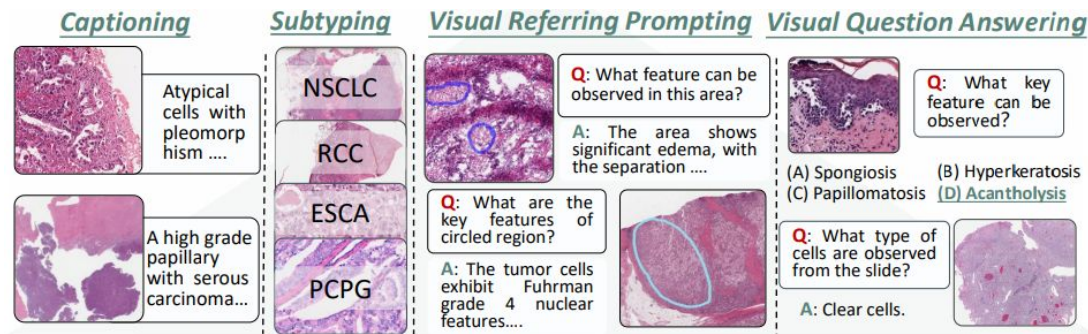
176k question-answer pairs

Two-stage training:

Cross-domain alignment

Visual instruction learning

A VLM for Whole Slide Images - Multiple Use Cases



CPath-Omni

Virchow2 (DINOv2-based) + CLIP

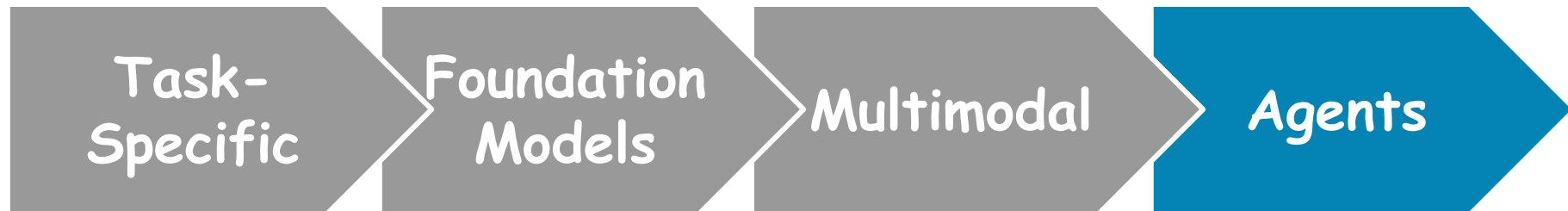
Four-stage training:

Vision-language alignment

Patches: VQA, classification, captioning

Whole slide pathology reports

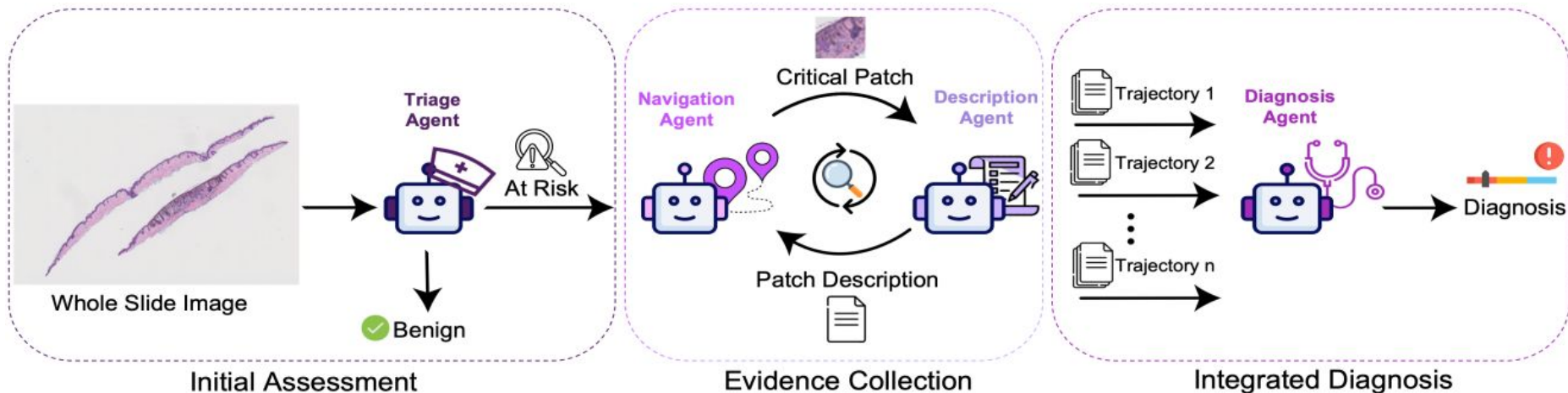
Slide and patch training



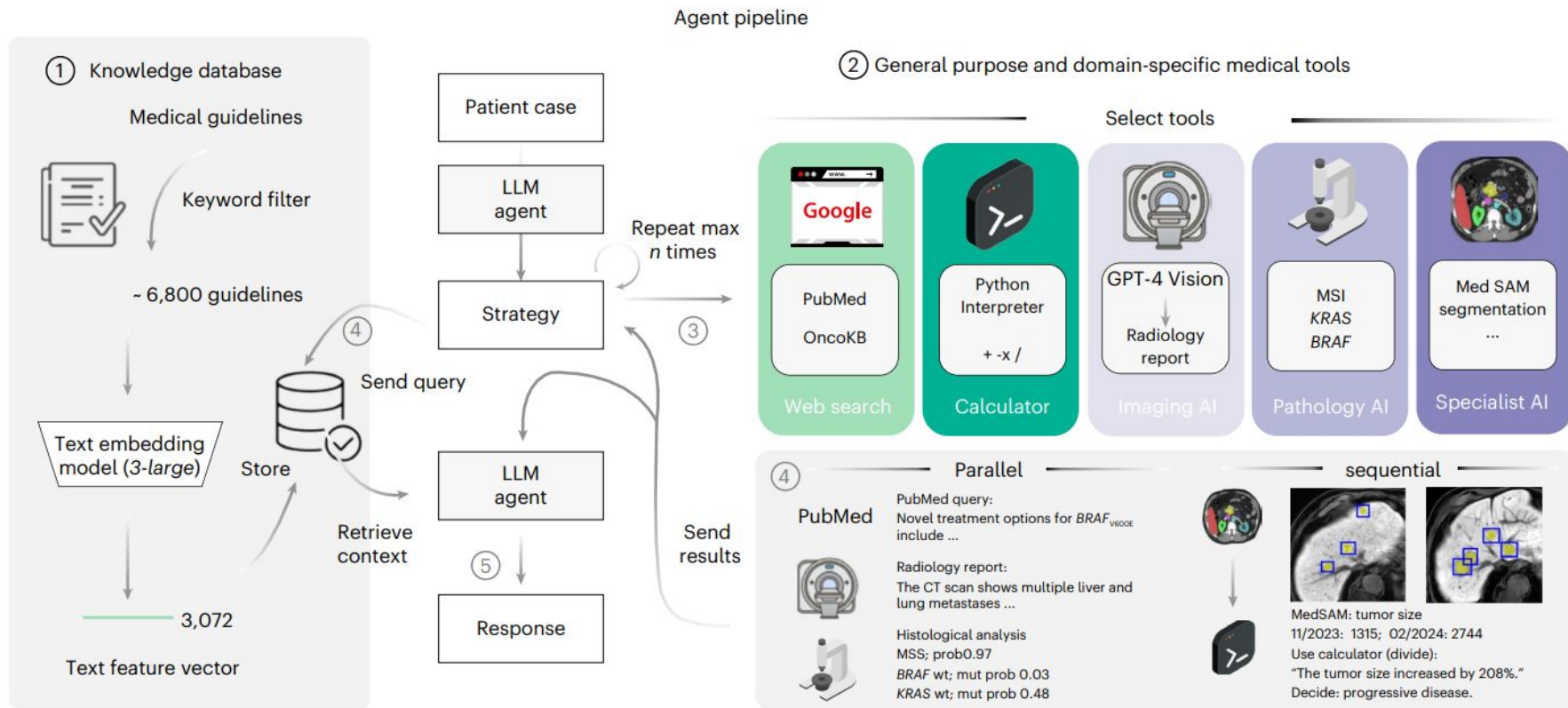
- Multi-agent systems
- Specialization
- Orchestration
- Mimic how human experts work

Agent for Histopathology Diagnosis

PathFinder

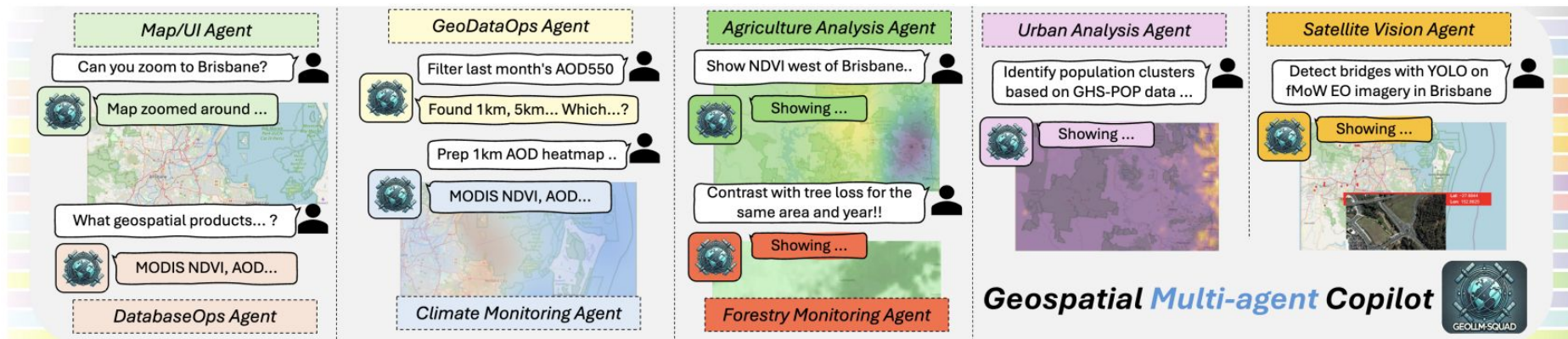


Agent for Clinical Decision-Making in Oncology



Geospatial Agent


























GeoLLM-Squad



Lee, Multi-Agent Geospatial Copilots for Remote Sensing Workflows, 2025

Benchmark for Geospatial Agents

ThinkGeo

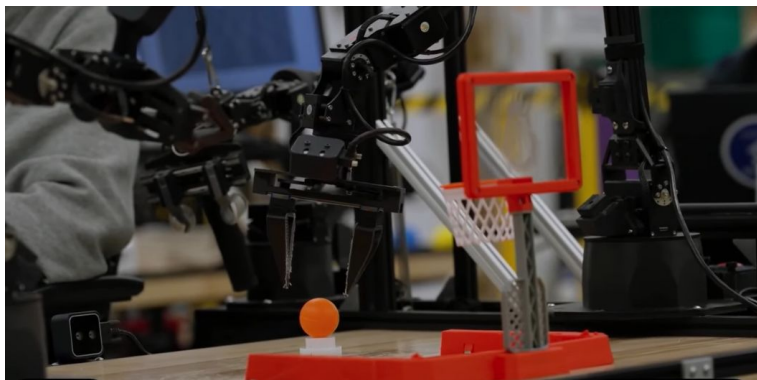
 User Queries	 ReAct based Reasoning/Execution Chain	 Answer
 <p>Identify the type and color of all visible cars in the image.</p>	<div>  TextToBbox <div>Cars → bounding boxes</div> <div>  Region Description </div> <div>Description → Type & color of vehicle</div> </div>	<p><u>Blue Minivan</u> <u>White Sedan</u></p>
 <p>Detect all houses, compute their total built-up area in sqm vs. the open space in the 600x600 image. GSD=0.7 m/px</p>	<div>  TextToBbox <div>Houses → Bbox</div> <div>  Calc </div> <div>Bbox → # pixels</div> <div>  Calc </div> <div>Total # pixel - # BBox pixel</div> <div>  Calc </div> <div>Compute GSD area</div> </div>	<p><u>32417.93 m²</u> <u>144983.07 m²</u></p>
 <p>The white house in northwest corner of the image is flooded or not? is there any water around 10m radius of the house? GSD 0.015m/px.</p>	<div>  TextToBbox <div>House in northwest corner → BBox</div> <div>BBox expansion by 10m</div> <div>  Region Description </div> <div>Presence of water</div> </div>	<p><u>Flooded</u> <u>Yes</u></p>
 <p>Detect any markings in the image and give information about the facility, is it tennis court or football stadium?</p>	<div>  OCR <div>Image → Bboxes & text</div> <div>  GoogleSearch </div> <div>"Centurylink Field"</div> </div>	<p><u>Football stadium</u></p>
 <p>Detect the construction waste and draw a bounding box over it.</p>	<div>  TextToBbox <div>Construction waste → BBox</div> <div>  DrawBox </div> <div>Draw Bbox on Image</div> </div>	
<div> <div>Pre</div> <div>Post</div> </div>  <p>What percentage of buildings in the post image were completely destroyed?</p>	<div>  ChangeDet <div>How many buildings were destroyed → Building count</div> <div>  ChangeDet </div> <div>How many damaged buildings → Building count</div> <div>  Calc </div> <div>Compute %</div> </div>	<p><u>35.7%</u></p>

Embodied Intelligence

“Can you pack me some trail mix?”

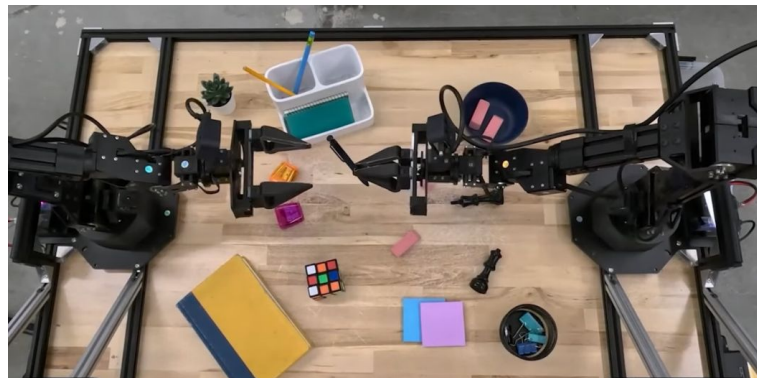


“Pick up the basketball and slam dunk it.”



Generality, interactivity, and dexterity

“Put the pen with the other pencils.”

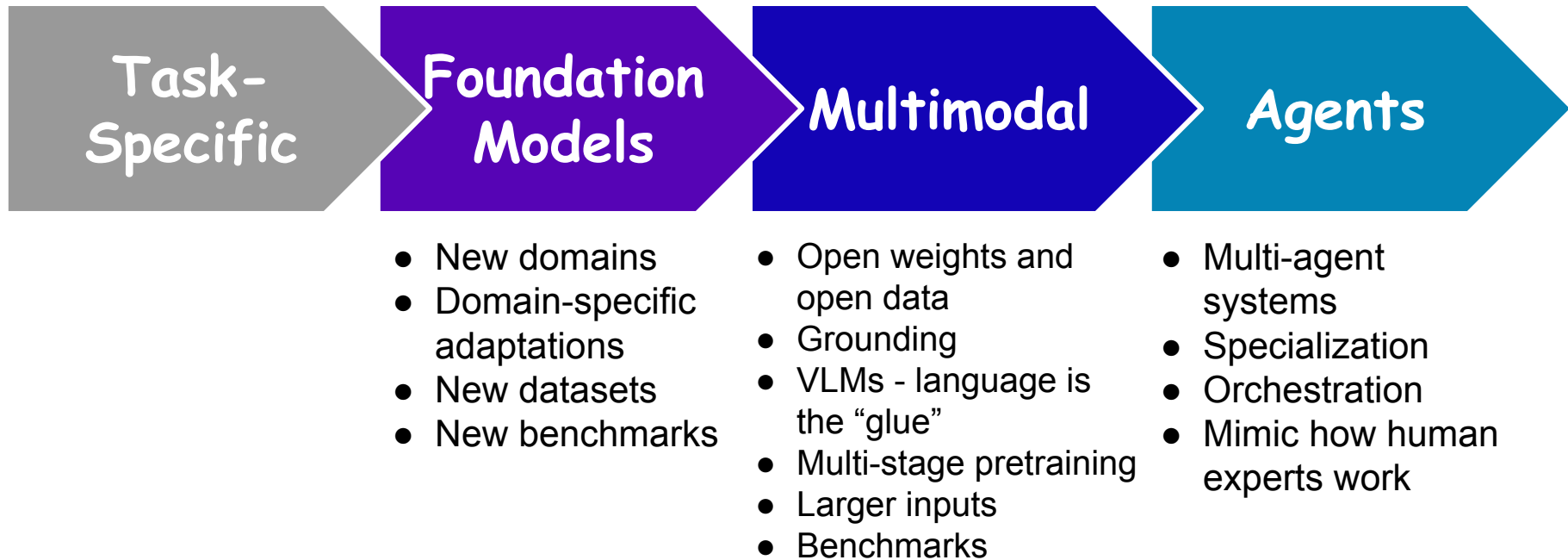


“Put the objects in the right container.”



Keynote: Carolina Parada, Gemini Robotics, <https://www.youtube.com/watch?v=o38k1k7f9Hk>

Trends



Resources

Recordings of keynotes and workshops:

<https://cvpr.thecvf.com/Conferences/2025/Videos>

Voxel51's "Best of CVPR" Series:

<https://voxel51.com/events>

Visual AI in Healthcare

June 27, 12-2 pm EDT

My talk: "Leveraging Foundation Models for Pathology: Progress and Pitfalls"

<https://voxel51.com/events/visual-ai-in-healthcare-june-27-2025>



Take the Next Step

Pixel Clarity Call

A free 30-minute call where we'll dive into your unique challenges and goals—whether you're seeking sharper models, deeper insights, or a new direction for your AI projects.

- Gain expert perspective on your current approach
- Discover high-leverage opportunities tailored to your mission

Book now: <https://calendly.com/hdcouture/pixel-clarity-call>

Q&A

Let's flip the conversation:

- 1) If you attended CVPR or another recent conference, what is one thing you learned?
- 2) What imaging domain do you work with?

Which “phase” is your work in?



Is the current “phase” sufficient to solve a real world problem?

OR

What is blocking the path to the next “phase”?