Automating Insect Detection in Videos with Metacentrum

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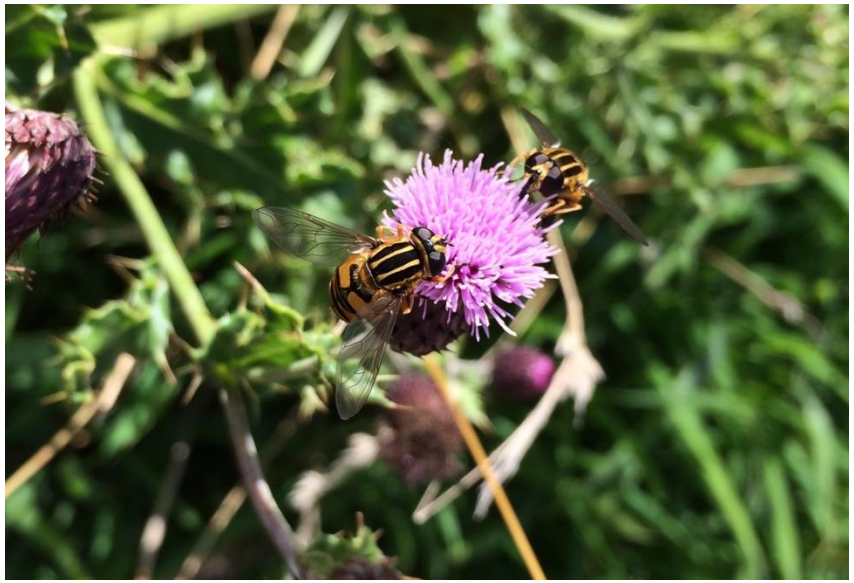

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Deployment and Testing: Jan Filip, Dominik Anýž

cesnet metacentrum

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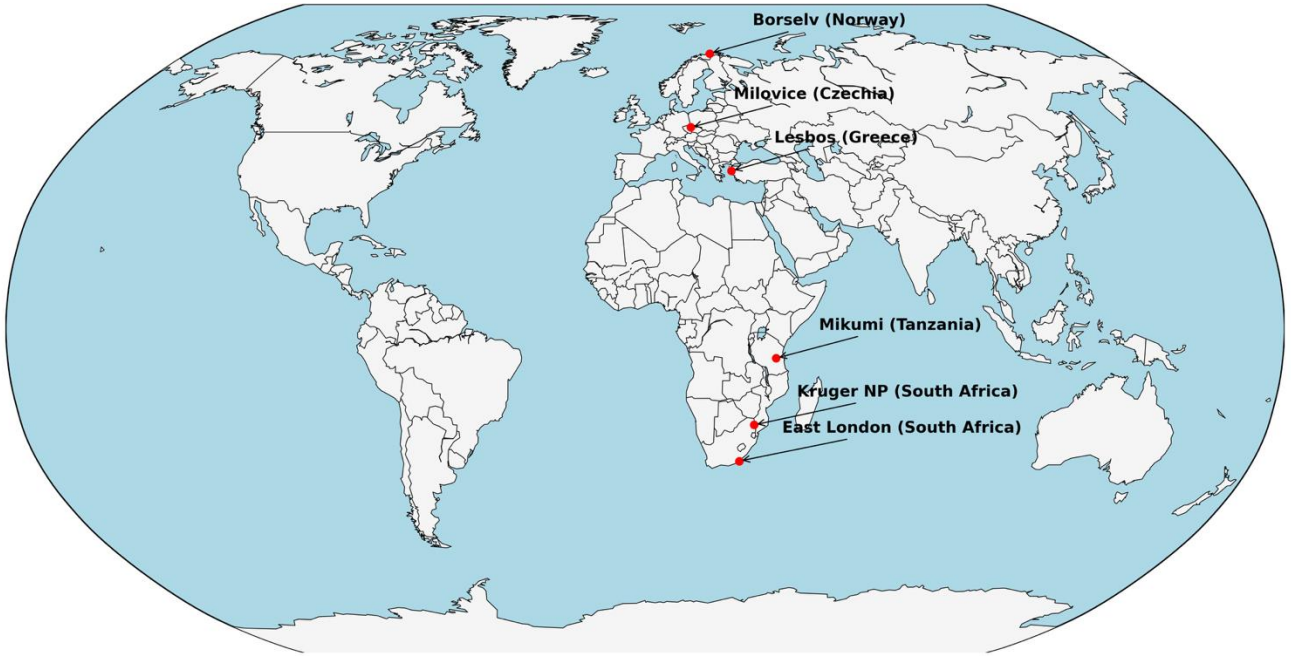
DetectFlow
Object Detection Pipeline

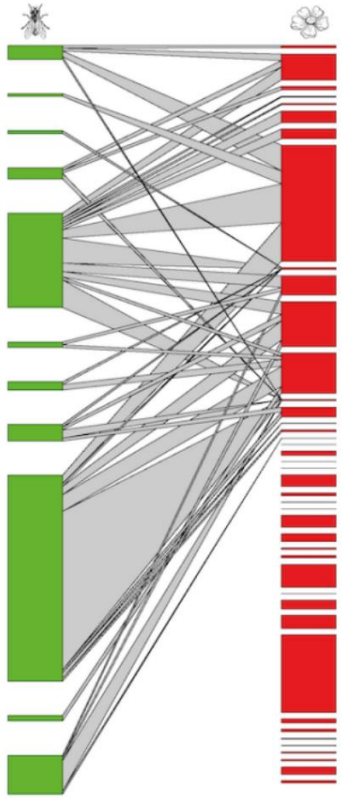
Scientific Motivation

- We study **pollination networks** (who pollinates whom, how often)
  - help quantify ecosystem interactions and specialization
- Current project: **latitudinal gradient** (Junior Star GAČR, 2021-2025)



Advertisement
a) sensory biases
b) reward motivation

Pollen Transfer
a) anthers -> pollinator
b) pollinator -> stigma



The Data Challenge

- Method: **CCTV cameras on flowers**
  - recording full 24h cycles across many flowering plant species

Advantages: Captures **diurnal** and **nocturnal** activity (infrared at night)

Ensures **statistical power** (huge recording times)

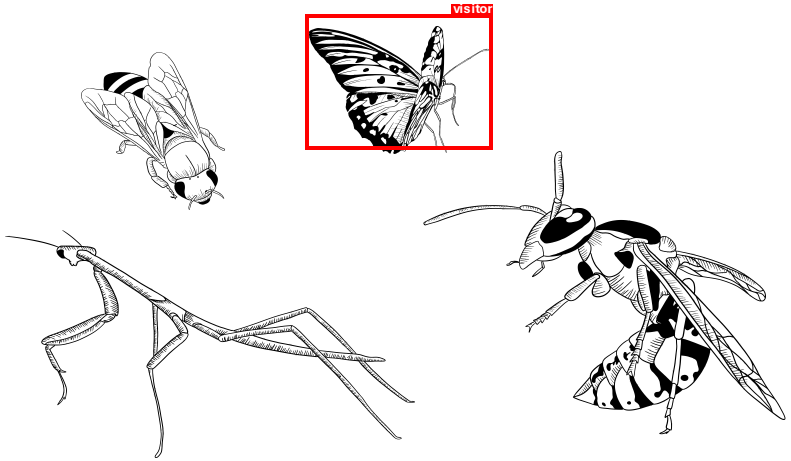
Deployable at **ecosystem scale** (30-50 flowering species per site)

Drawbacks: **10 years** of accumulated recording time

~3,500 full-day recordings = 350,000 video files = **~8 billion frames**

Video **quality varies** (day vs night, movement of flowers, small insects)

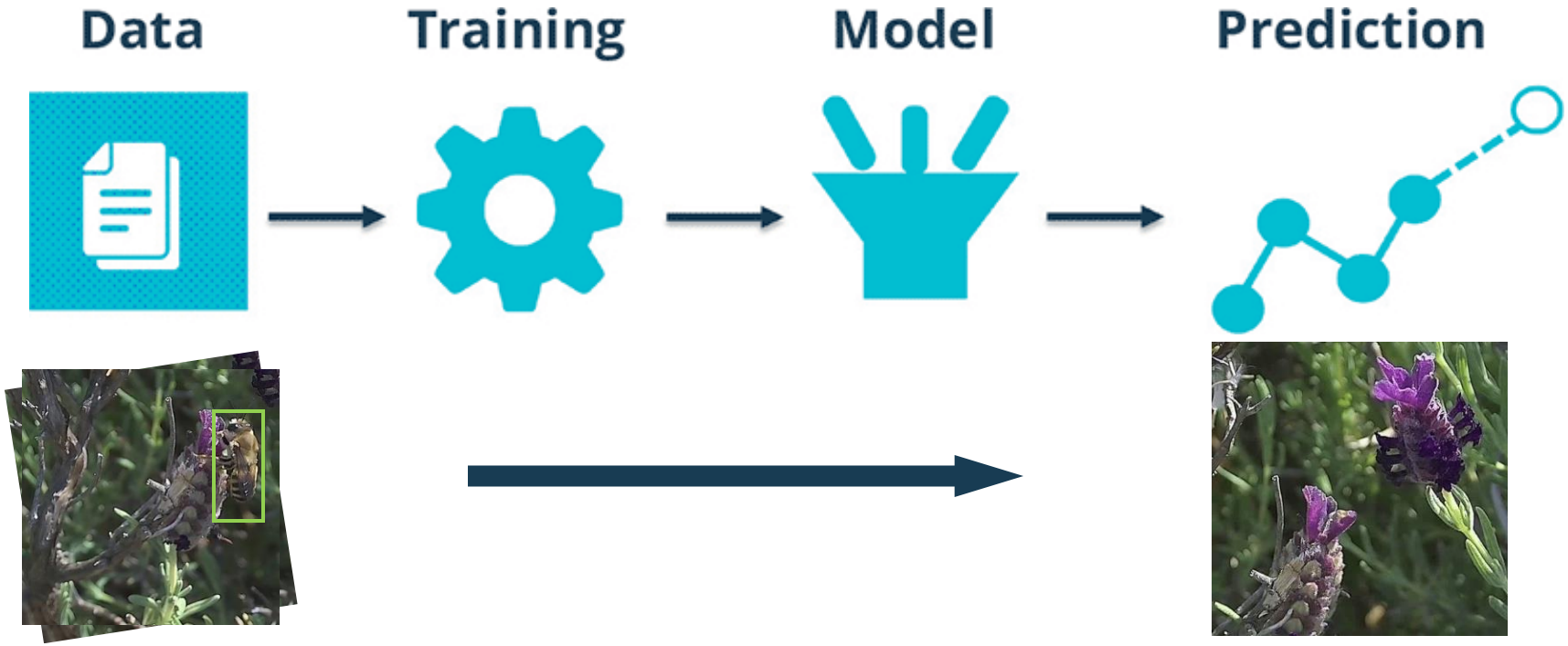
Data Processing Goals



- **Raw videos** -> detect insect visits

  locate insects relative to flowers

  extract **visit-level records** (arrival/departure)



Model Training

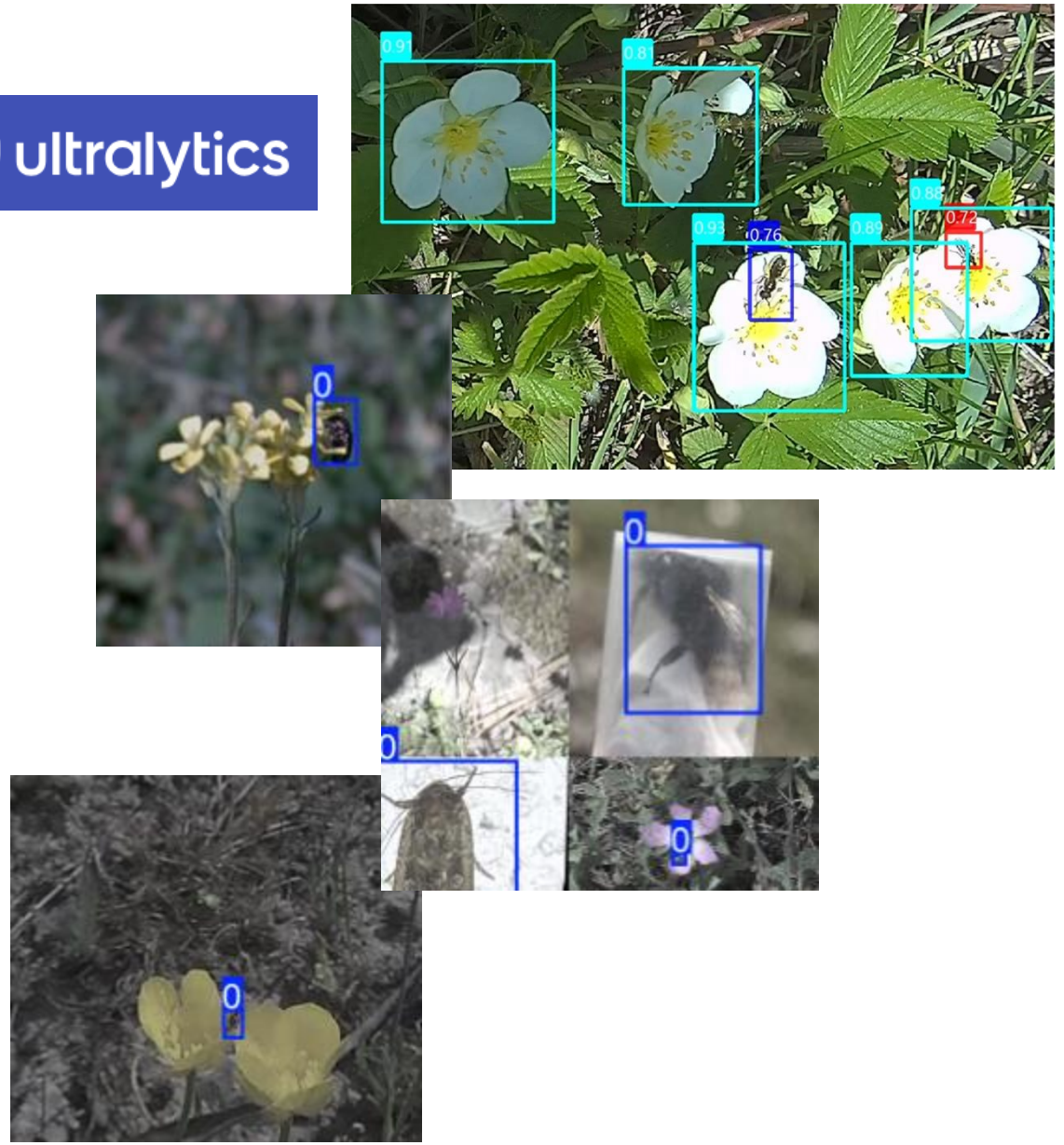
- Chosen algorithm: **YOLO** (Ultralytics)

- **Training data**:
  - Manually annotated frames
  - Semi-automated labeling
  - Enriched with internet insect images

- Strategy: pretrain „**insect expert**“

  -> fine-tune on flower videos

- **Flower** x **insect** model



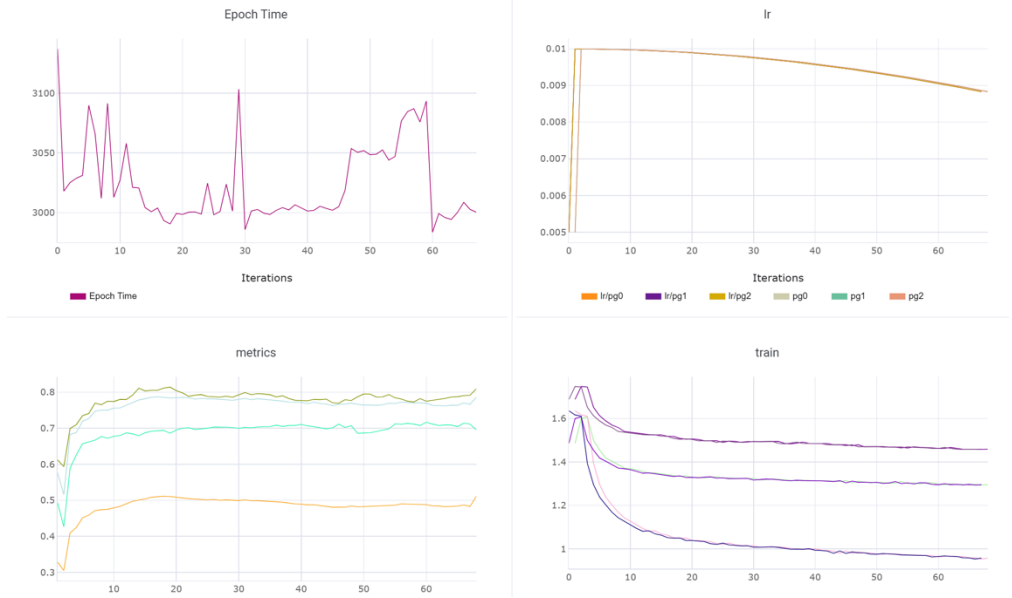
Why MetaCentrum?

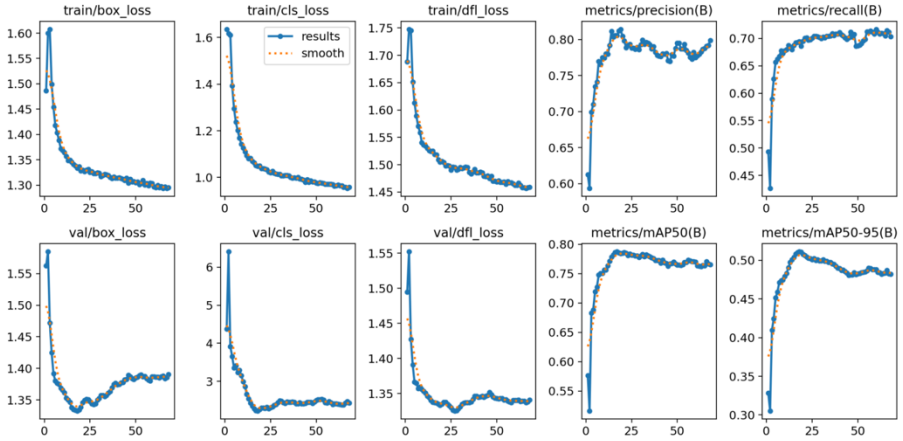
- Needed **HPC resources** for:
  - Large training runs (GPUs)
  - Huge inference workload (CPUs)
  - Long-term S3 storage of TBs of video
  - Jupyter Notebooks with processing resources
- MetaCentrum **enabled** scaling to billions of frames

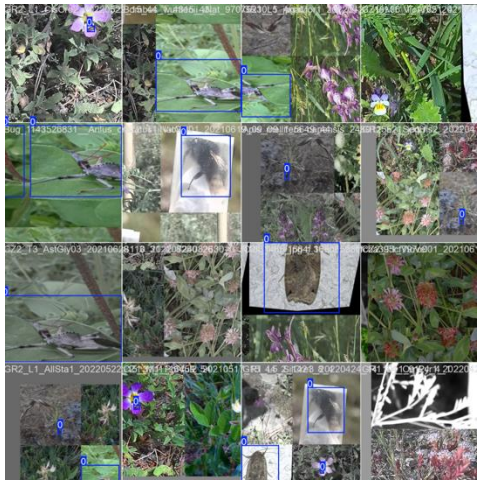
Jupyter -> OnDemand -> PBS jobs -> GPUs, CPUs, Singularity

Training on GPUs

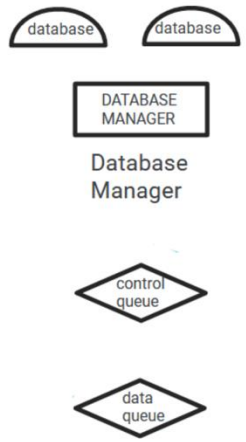
- Used GPUs nodes with high **VRAM (~64 GB)** (**adan**, **galdor** – gpu_long)
- Avoided AI-specific cluster (queues too long)
- Built and ran models in Singularity **containers**
- Solved CUDA/package version issues by pinning dependencies



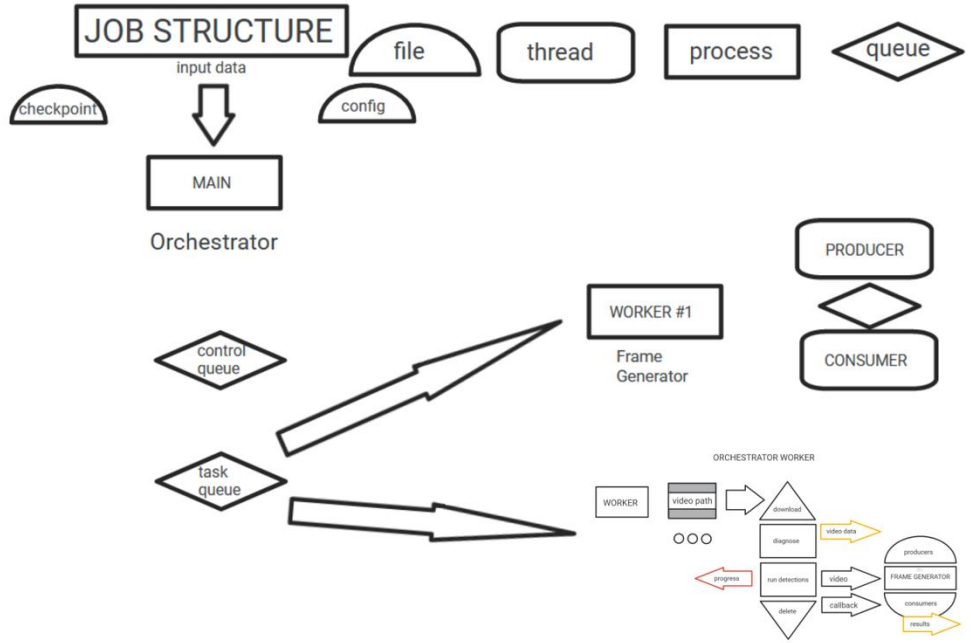




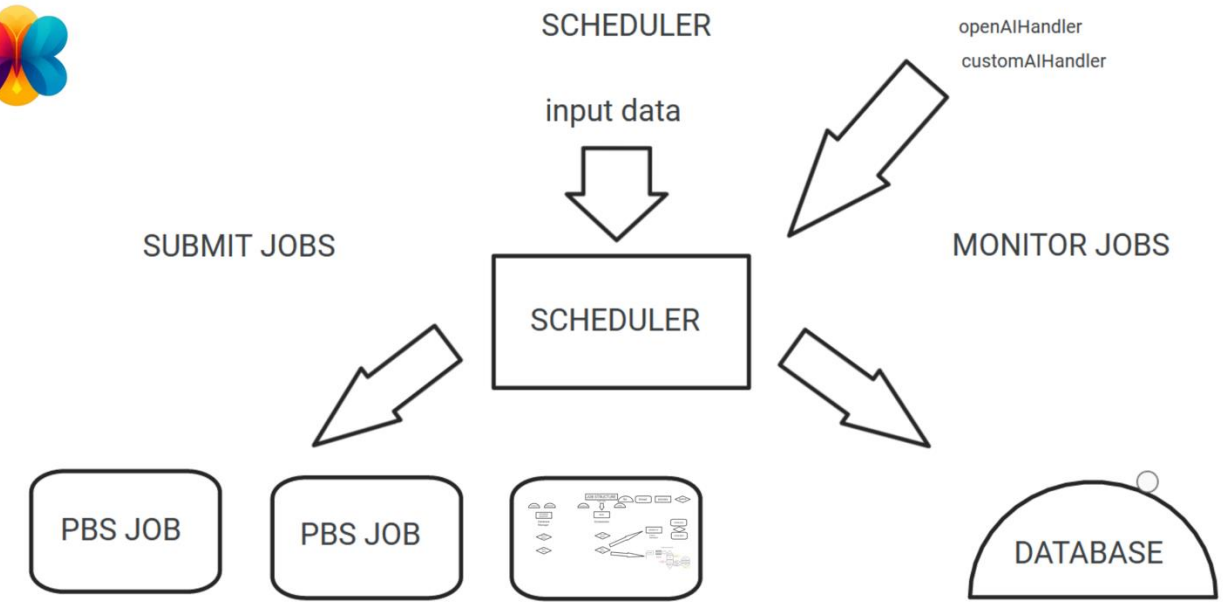
Inference on CPUs



- **PBS** job design:
  - Each job = batch of tens of vic
  - Dozens of jobs in parallel -> h

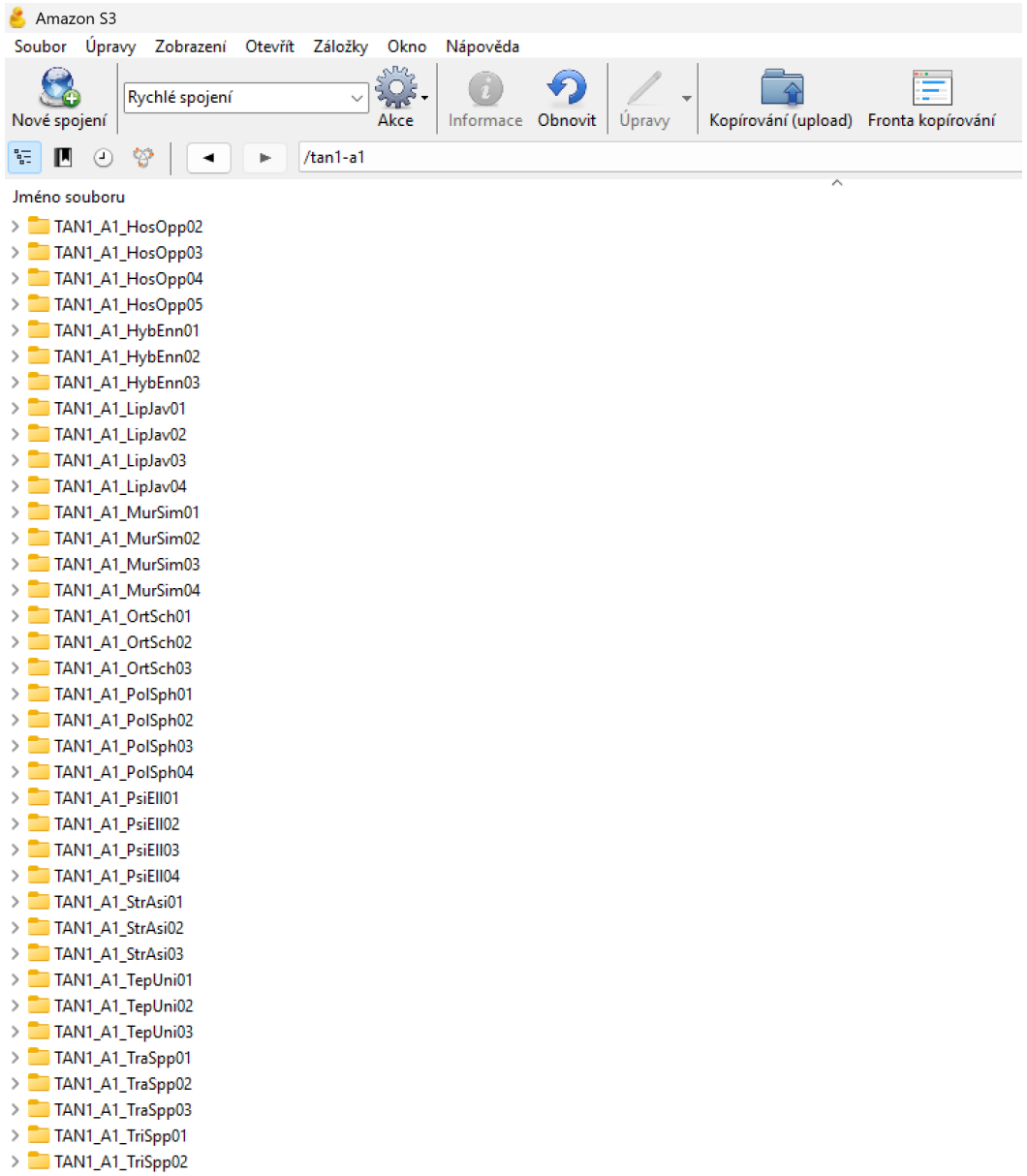


- Jobs restartable to handle failure
- All large-scale inference ran on C **only**:
  - GPU queues too slow
  - CPU jobs easy to parallelize



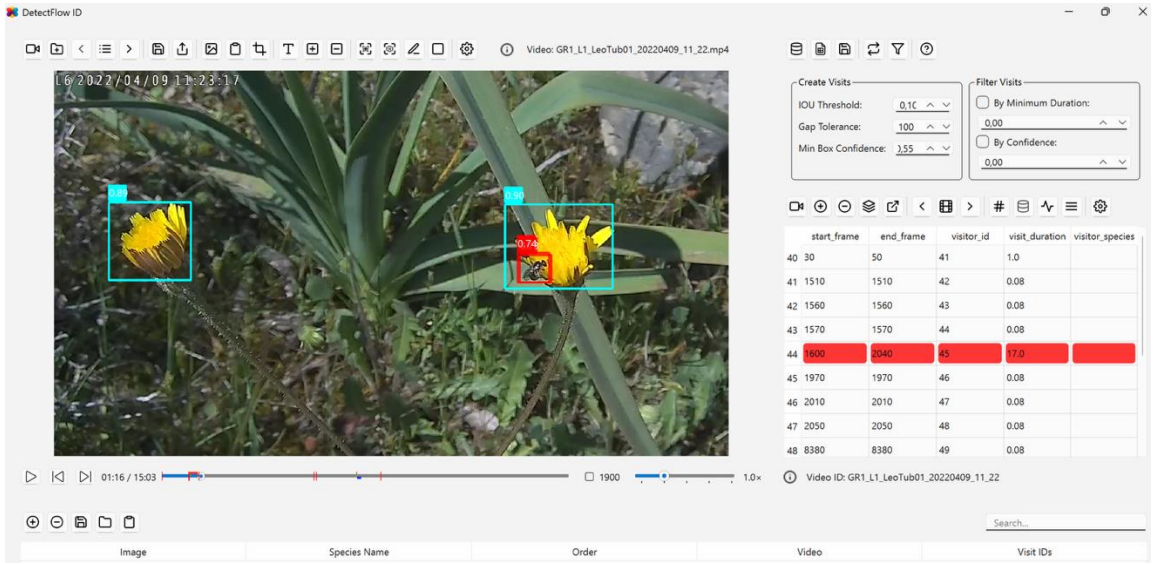
Storage & Data Handling

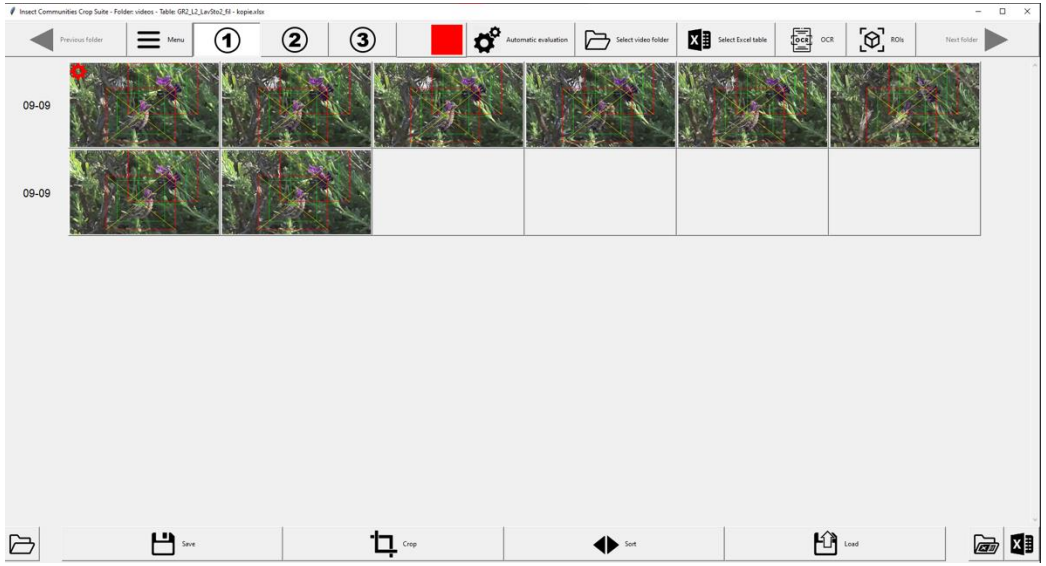
- **S3 storage**: long-term archive, easy upload
- Local cluster storage only for temporary files
- Essential for handling terabytes efficiently
- Perfect protection through **credentials**
- **Shareable** in virtual organizations, work groups



Outcomes

- HPC saved years of manual work -> project otherwise impossible
- Processing of **~8 billion frames** feasible
- Produced **visit-level dataset** from raw video
- Custom GUI app allows refining detections into final data





Lessons Learned

- **Containerization** critical for reproducibility
- **S3 storage** indispensable for large video workflows
- **PBS job parallelization** allowed scaling
- **Standardised** sampling crucial for easier AI learning

Acknowledgements

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- Thanks to the whole team and worlwide collaborators!

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Open to work ☺