IEEE GRSS HDCRS Working Group High Performance and Disruptive Computing in Remote Sensing Summer School 29 May - 1 June, 2023, Reykjavik, Iceland

Lowering the Barrier for Modern Cloud-based Geospatial Big Data Analysis by Combined Use of Innovative and Traditional Infrastructure

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Big data is large and complex data that is difficult to process

(can also be defined with many V-words)





Gooding and a second and a second and a second a	Center hilding 57198-0001 maat 99 hede hede DED AULOU			
	Amount of Data	16GB	Tererer	
	Transfer Rate Total Transfer Time 0 Days, 0	1000 Mbps Hours, and 2 Minutes, 17 Seconds		





Complex & difficult for me Simple & easy for Ruihang Xu



What is the size of the largest data you processed?



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What are the file formats of large data you process frequently?





What are the complex data you need to process?

High dimensional	Geospatial	Complex-valued raw SAR data	
Data cubes, time series	Hyperspectral	Large dimensions,extent resolution	
Large datasets in a short amount of time. Automating workflows	Heterogeneous data with different spatial and temporal resolutions,	optical data	
Multiple tiles/time series	different reference coordinate systems etc	Multi dimensional	
Timo corios	DEM	SAR	
Time series	Time series	-	
high dimensional		i ne data is simple, the volume is the challenge	
SAR and Hyperspectral	Data-Cubes	LiDAR point clouds, SAR data	
	Time Series		
Multi-modal		Time-series	
Post processing how to aggregate	time-series	Automation to make more efficient	
and get the aggregated results	Efficient access to the data and		
Computing power	scaling		

What are the challenges you encounter while processing such data?

Time	Lacking ressources and time	Computation and computing power
Computering power	computational cost	Loading to memory
Disk space	Storage and computing	Optimization, fast calcluation
Optimisation and processing time	Processing speed	Quality
computing power	Preprocessing	storage
Heteorogeneity in plate-forme, viewing angles storage processing	Efficient access and scaling	Supporting operators and packages
time		
	Unconsistant Data	Tweaking the process and waiting on computation again
Time and computational power		

Solutions require expert know-how and infrastructure

• Local and regional studies with **medium size data**

Analyses can be done faster by **parallel computing** on a workstation

• Machine learning and AI studies with medium size data

Analyses require **special processing units** (e.g., GPU/TPU) due to computational complexity

• National, continental, and global studies with **big data**

Analyses require **distributed computing** on a computing cluster due to computational complexity and/or large volume of data





Cloud computing is on-demand availability of computer system resources, especially **data storage** and **computing power**, without direct active management by the user

Cloud computing has a few distinctive features

- **On-demand self-service**: provision of computing capabilities as needed without requiring human interaction.
- **Broad network access**: availability over the Internet with standard access mechanisms for different client platforms (e.g., tablets, laptops, mobile phones).
- **Resource pooling**: dynamic assignment and reassignment of physical and virtual resources according to consumer demand.
- **Rapid elasticity**: capability to scale rapidly outward and inward proportionate to consumer demand.
- **Measured service**: accurate monitoring, control, and reporting of resource and service utilization.



Computing is moving to the Cloud, so is geocomputing

- Developments in infrastructure, both hardware and software, gave a big push to data processing and analysis capabilities.
- Scalable and affordable computing is available through:
 - Open-source systems that allow computing clusters on commodity hardware
 - Proprietary cloud-based data storage and computing services
 - More accessible research ICT infrastructure and research cloud
- Using the solutions usually requires a transition in modus operandi.
- But challenges exists in **identifying the cases** where cloud computing can play a role and in **proper selection and efficient use** of cloud computing methods, tools, and services.



What are your needs for better use of big data and cloud computing technologies?



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Suggestion One

Inspect what's going on!

Monitoring resource utilization helps to identify bottlenecks

- Geospatial workflows usually involve many software components, including yours.
- Not all software components are able to utilize available modern computing capabilities or utilize them efficiently.
- **Monitoring** of the resource utilization is the crucial first step to **understand** the situation.
- Scaling without efficient use of resources is suboptimal and costly.



Suggestion Two

Read the documentation!

Distributed computing and the Cloud are not magical tools

- Recent tools and technologies significantly lowered the barrier to use cloud computing and parallel/distributed computing capabilities for geospatial workflows.
- However, time and effort are required to get competent in using them efficiently.
- Scaling blindly is suboptimal and costly.



Suggestion Three

Use the right services!

Cloud computing comes with many ...aaSs!

- Infrastructure as a Service (laaS)
 - Provider supplies the infrastructure that needs to be set up by the user. e.g. <u>Amazon AWS</u>, <u>Microsoft Azure</u>, <u>Google Cloud</u>, <u>ESA DIASs</u>, National Research Clouds
- Platform as a Service (PaaS)
 - Provider supplies the platform that allows the user to deploy applications (scripts). e.g. <u>Google Earth Engine</u>, <u>Microsoft Planetary Computer</u>, <u>Google Colab</u>
- Software as a Service (SaaS)
- Function as a service (FaaS)
- Data Processing as a service (DPaaS)
- Data as a service (DaaS)



There are also many cloud service providers!





Coogle Cloud

- Common features
 - Virtual machines
 - Cloud storage
 - Open-source software
 - Open datasets

- Azure Machine Learning Platform Cloud-based environment to train, deploy, automate, manage ML models
- Azure Data Science Virtual Machines Geo AI Data Science VM with ArcGIS
- EMR Cloud-native Big Data Platform EC2 + S3 clusters without provisioning, with OSS (Hadoop, Spark, etc.)
- Google Compute Engine

Cloud TPU (eg. ResNet-50, 90 ep.: 8 V100 GPU: 216 min, Cloud TPU V2: 7.9 min)

• BigQuery

• Different features

BigQuery ML: create and execute ML models using standard SQL BigQuery GIS: analyze and visualize geodata by using standard SQL

Google Earth Engine is a gamechanger for geospatial computing

Combination of a multi-petabyte catalog of EO imagery and geospatial datasets with planetary-scale analysis capabilities available for free*.

https://earthengine.google.com





openEO develops an open API to connect R, Python, JavaScript and other clients to big Earth observation cloud back-ends in a simple and unified way.

https://openeo.org

Geocomputing on local cloud can be efficient and cost effective

- ITC Geospatial Computing Platform provides GPU-enabled general purpose (8 vCPU, 32 GB RAM) and big data (72 vCPU, 768 GB RAM) units with large storage, analysis ready data, ready-to-use interactive and desktop software (1500+ packages), and shared workspaces.
- Currently serves 550+ registered users.
- Provided **250,000+ hours** of computing since 2021.
- Already returned 16+ times the investment costs.
- Monthly cost is < 200 Euro.

https://crib.utwente.nl



Geospatial Computing Platform is designed to serve the needs of the user community*

- Designed for the primary activities:
 - Self learning
 - Exploratory research
 - Education
- Designed based on the primary criteria:
 - Highly available
 - Ready to use
 - User friendly
 - GPU enabled
 - Distributed-computing friendly
 - Low-cost

24/7, no queue Pre-installed software Interactive user interface GPU for each user Computing cluster Feasible investment



* Girgin, S. (2020) Big Geodata at ITC: Status Quo and Roadmap

Illustration by Storyset

We utilized innovative solutions to develop a platform fulfilling the criteria



NVIDIA Jetson AGX Xavier Cluster

8-core CPU NVIDIA Carmel ARMv8.2, 2.26GHz 512-core GPU

Volta architecture with 64 Tensor Cores

32GB memory 256-bit LPDDR4x, 2133MHz, 137GB/s

32GB internal storage

DL and CV Accelerators

Gigabit Ethernet

i <u>https://elinux.org/Jetson_AGX_Xavier</u>



JupyterHub on Docker Swarm

Interactive Jupyter Notebooks, Desktop Applications

Flexible Multiple computing units, multiple language kernels

Scalable Can be deployed with container technology

One can start small...



... but it is important to have an expansion policy

- Grow only if there is **meaningful demand**.
- Upgrade first, then expand.
- **Repurpose idle resources** to make them available for common use.
- Select low-cost, good-performance solutions.
- Use **refurbished** equipment if possible.



Today we have an efficient shared infrastructure

• **16 x** GPU-enabled General Purpose Computing Units

8-CPU NVIDIA Carmel ARMv8.2 @ 2.26GHz, 32 GB RAM, 512-core Volta GPU, 64 Tensor Cores

• **6 x** General Purpose Computing Units

8-CPU Intel Core i7-7700 @ 3.60GHz (max. 4.20GHz), 32 GB RAM

• 3 x GPU-enabled Big Data Computing Units

72-CPU Intel Xeon E5-2695 v4 @ 2.10GHz (max. 3.30GHz), 768 GB RAM, NVIDIA RTX A4000 GPU

• **1 x** GPU-enabled Big Data Computing Unit with Fast Storage

32-CPU Intel Xeon E5-2640 v3 @ 2.60GHz (max. 3.40GHz), 768 GB RAM, NVIDIA RTX A4000 GPU, 22 TB RAID 20+2

• 1 x Multi-GPU Computing Unit

32-CPU AMD Ryzen Threadripper PRO 3955WX @ 3.9GHz, 160 GB RAM, 4 x NVIDIA RTX A4000 GPU

• 1 x Platform Server

72-CPU Intel Xeon E5-2695 v4 @ 2.10GHz (max. 3.30GHz), 512 GB RAM, 240 TB RAID1 (ZFS)

• 2 x Application Servers

12-CPU Intel Xeon E5-2420 v2 @ 2.20GHz (max. 2.7GHz), 192 GB RAM, 48 TB RAID1 (ZFS)

• 2 x Storage Servers

Synology, 60 TB

The platform aims to enable access to a full hardware and software stack for big geodata processing

High-level Applications					
High-level Frameworks Spatial Extension Hive, Mahout, GeoTrellis, MLlib, SpatialHadoop, GeoS		Spatial Extensions tialHadoop, GeoSpark, PostGIS,		Runtime/Compiler Frameworks tvm, Mace,	
	Distribute Casandra, Dyna	d Databases mo, Ignite, HBase,		Geos, JTS, S2	High-level Machine Learning Frameworks Keras, Gluon,
Distributed Computing Frameworks Hadoop, Spark, Storm, Kafka,		Databases PostgreSQL	Machine Learning Frameworks TensorFlow, PyTorch, Caffe2, MXNet,		
Distributed Processing Frameworks YARN, Mesos, Borg,		MongoDB	High-level Libraries NumPy, CuPy, Dask, GDAL,		
Distributed File Systems HDFS, EOS,		File Systems	Low-level Libraries BLAS, MKL, OpenCL, Vulkan, Metal, CUDA,		



Resource sharing is at the core of the platform

- Accessible through a **web browser** (No software installation or VPN are required)
- **No registration** is required (Login with the University credentials)
- Each user has an individual and isolated working environment
- Each user has access to all available* **unit resources**, including **GPU**
- Each user has access to all available* cluster resources
- **Replicated storage** with minimum two copies (Hardware failure protection, ZFS)
- **Distributed storage** for big data processing (HDFS)
- Automatically **balances workload** among the units

* Resource availability depends on resource usage of other active users.

It provides features to simply research activities

- Interactive notebook, terminal and remote desktop access are available
- Multiple interactive languages are supported (Python, R, Julia, Octave, Go, ...)
- Up-to-date and optimized software packages are ready to use (No setup required)
- Users can install additional packages (e.g., Python, R packages)
- Distributed computing clusters are **ready to use** (Dask, Apache Spark)
- Public assets are shared by all users (e.g., OSM, NL 0.5m DTM, TOP10-1000, ...)
- Shared workspaces allow assets to be shared by selected users
- Access can be granted to **external users**
- User support is available
- Provided and maintained at no cost (i.e., free PaaS)

Hundreds of software packages are available ready to use



Desktop applications are also available

- QGIS
- GRASS GIS
- SAGA GIS
- **OTB**
- ENVI*
- SARScape*
- Metashape*
- SNAP
- ILWIS 3*
- ILWIS 4*

- VS Code
- PyCharm
- R Studio
- Netlogo
- GNU Octave
- MATLAB*
- Glueviz
- Orange Data Mining



- * Windows application, available only on Intel units through emulation.
- * Licensed software, available only on Intel units through license server.

Prominent geospatial technologies are integrated to the platform







GeoServer

Open source server for sharing geospatial data

Open source platform for publishing spatial data

MapServer



MariaDB Open source relational database



GeoNode Open source geospatial content management system



Dataverse Open source research data repository software



Gitea Open source lightweight code hosting solution



Open Data Kit

Open source platform to collect data quickly, accurately, offline, and at scale

24/7 user support is available through the Support Center

Support Center Home	Knowledgebase	Open a New Ticket	Check Ticket S	Status
Search our knowledge	Search		Open a New Ticket	
Welcome to	the CRIB	Support Ce	enter!	Check Ticket Status
In order to streamline support requests and better serve you, we utilize a support ticket system. Every support request is assigned a unique ticket number which you can use to track the progress and responses online. For your reference we provide complete archives and history of all your support requests. Quick Access • <u>Report a Problem</u> • <u>Shared Workspace Request</u> • <u>Course Workspace Registration with Canvas Integration</u> • <u>External Account Request</u> • <u>Account Removal Request</u>			er which you e we provide	How can I access to the platform? Is it secure?
				How can I use the platform? Which programming languages are supported on the platform?
			Which libraries and packages are supported by the platform?	
Account Transfer Re Software Request Dataset Request	iquest			1 .

Database Request

The platform is maintained and software are updated regularly

- On-demand and bi-monthly regular rolling updates.
- **Similar** working environment for ARM64 and Intel x86-64 units.
- Automated shared workspaces for the **departments**.
- Automated shared workspaces for **courses**.
- Automated notification to newcomers.
- Daily storage snapshots for 7 days.
- **Bi-daily** check for malicious threads .
- **Continuous** resource and performance monitoring.







The user community of the platform includes hundreds of researchers, students, and alumni

- Operational since January 2021
- 555* registered users (max. ~865)
- 106* shared workspaces for projects and courses
- 15-50* concurrent users at a time
- **250,000+*** hours of multi-core/GPU computation
- **460+** support tickets* closed (excluding support by e-mail)
- Overall **positive feedback** from a wide-range of use cases 4.61/5.00 according to the <u>user survey</u>
- Central ICT built a similar platform for **university-wide use** Co-developed by CRIB



Illustration by Storyset

* As of 29 May 2023

Existing and potential use cases cover a wide range of activities

- Education
 - Computation platform for **courses** Shared course workspaces can be created easily with <u>Canvas integration</u>
- Research
 - M.Sc. / Ph.D. thesis studies
 - Collaborative (big) data analysis and visualization
 - Strengthen **project proposals** Small projects can use the platform and reduce their budget needs
- Capacity Development
 - Self-learning

e.g., cloud computing, distributed computing, GPU computing, ML, ...

• Computation platform for **activities** e.g., training workshops, hackathons, ...

Lessons learned: Mixed computing resources work nicely

- Innovative = "unproven" solution
 - ✓ Serving quite well (8 core, 32 GB RAM, GPU, in cluster formation)
 - ✓ High performance/cost ratio (≈ 3 EUR/h VM, pays off in 300 h)
 - ✓ Low energy consumption (max. 30W/unit)
- Support from the departments / staff
 - ✓ Donation (servers, storage servers, GPUs)
 - ✓ Donation + shared upgrade (big data unit)
 - X Sharing (Jetsons)
- Low-cost solutions
 - \checkmark Second-hand equipment

Lessons learned: Rolling updates do not cause major trouble

- Rolling update = State of the art
 - ✓ 950+ Python, 550+ R packages (Statistical, Spatial, EO, ML, AI)
 - ✓ QGIS, GRASS, SAGA, ILWIS, SNAP, Octave, NetLogo, MATLAB, ENVI, ...
- State of the art = Difficult to maintain

 - \land Stability*
- Docker virtualization / orchestration
 - 😂 Works quite well (multi-architecture)
 - 🔄 Image rebuilds take a lot of time (7+ days)
- ✓ Support center (460+ closed tickets)
- Additional services





Lessons learned: Building a community is challenging

- Computing Platform
 - ✓ In use for MSc/PhD studies, courses, projects, and trainings.
 - ✓ Overall user satisfaction is quite positive.
 - \triangle Reaching potential users is not easy.
 - \triangle Getting useful feedback is difficult.

ID↑	Name	Responses
1	anonymous	STOP SENDING ME MAILS STOP SENDING ME MAILSSTOP SENDING ME MAILSSTOP SENDING ME MAILSSTOP SENDING ME MAILSSTOP SENDING ME MAILSSTOP SENDING ME MAILSSTOP SENDING ME MAILSSTOP SENDING ME MAILSSTOP SENDING ME MAILSSTOP SENDING ME MAILSSTOP SENDING ME MAILSSTOP SENDING ME MAILSSTOP SENDING ME MAILSSTOP SENDING ME MAILSSTOP SEN

- CRIB
 - \triangle Collaboration exists, but can be improved.
 - ✓ Automated e-mail to newcomers and newcomer meetings help.



4.37 Average Rating

Lessons learned: User stories are difficult to collect

- Best way to convince people is to show examples
- Best examples are in-house from "familiar" faces
- They are not easy to collect



absquatulate (v):

to leave without saying goodbye

A few suggestions for newcomers

- Ensure familiarity with the cloud computing technology through short talks and lectures.
- Improve know-how by participating tool- and technology-specific training
- **Try and use** the infrastructure and platforms available for free or through partner organizations.
- Follow a hybrid approach (local + cloud) to maximize the benefits.
- Ask for technical and scientific support for better implementation and integration of the technology.
- Ask for guidance for the planning of future activities.
- Share your knowledge and good practices with your colleagues (e.g., for cost-effective and efficient use).

Following best practices facilitates moving to the Cloud



Source: Best practice for using cloud in research (Hong et al., 2018)

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