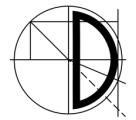




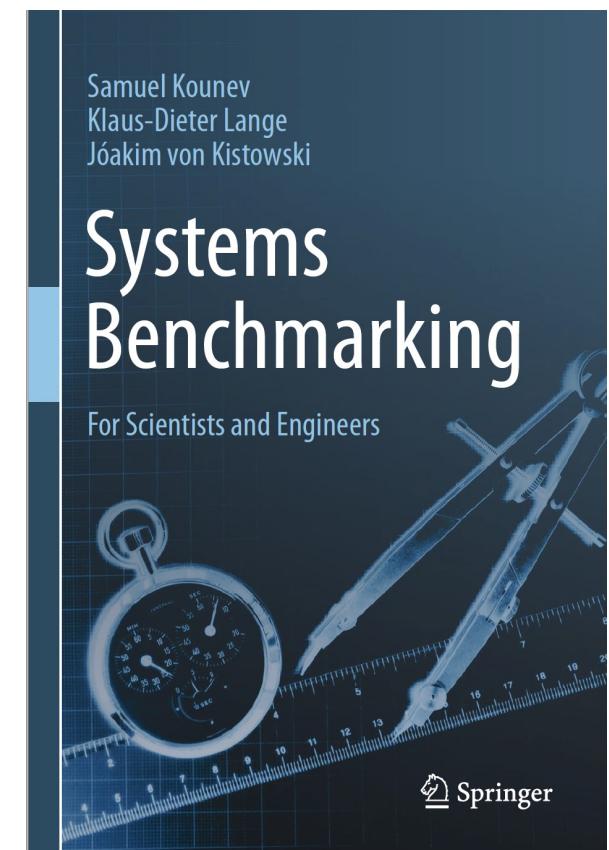
Update from Descartes Research Group

Samuel Kounov
Chair of Software Engineering



New Course on Systems Benchmarking

- Theoretical and practical foundations
- Both a textbook and a handbook on benchmarking
- Includes modern applications, case studies, and latest research based on input from over 40 benchmarking experts
- Teaching materials available on request
- Credits: Numerous (under-)graduate students (2006-2022):



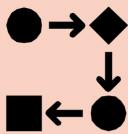
<http://benchmarking-book.com>



Software Engineering Group

Software Architecture

- Distributed Systems
- DevOps
- Modeling & Simulation
- Performance Engineering

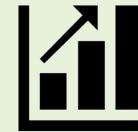


Benchmarking & Experimental Analysis

- Performance
- Energy Efficiency
- Security
- Dependability

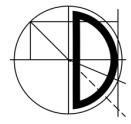
Autonomic Systems

- Self-Adaptation
- Self-Organization
- Self-Protection
- Artificial Intelligence



Predictive Data Analytics

- Statistical Modeling
- Machine Learning
- Time Series Forecasting
- Critical Event Prediction



Research Areas

Internet-of-Things / Cyber-Physical Systems



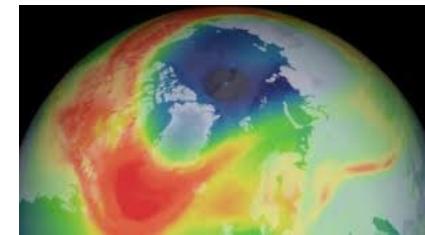
- Mobile networking and communication
- Transport / logistics
- Predictive maintenance
- Robot localization & people detection

Data Centers



- Cloud computing
- Green IT
- Security
- Resilience and robustness

Earth Observation

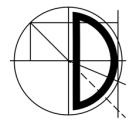


- Scientific computing
- Scalable data processing
- Satellite image analysis

Medicine and Sports Science



- AI-based decision support systems
- Telemedical and real-time AI systems
- ML-based pattern recognition for vital time series



Earth Observation (EO)

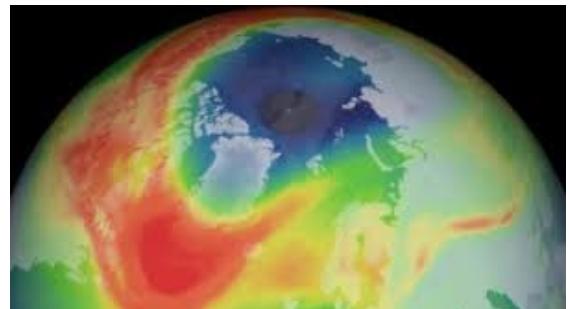
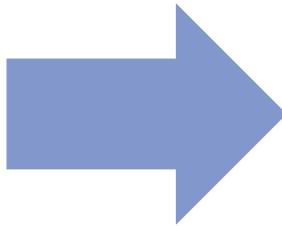
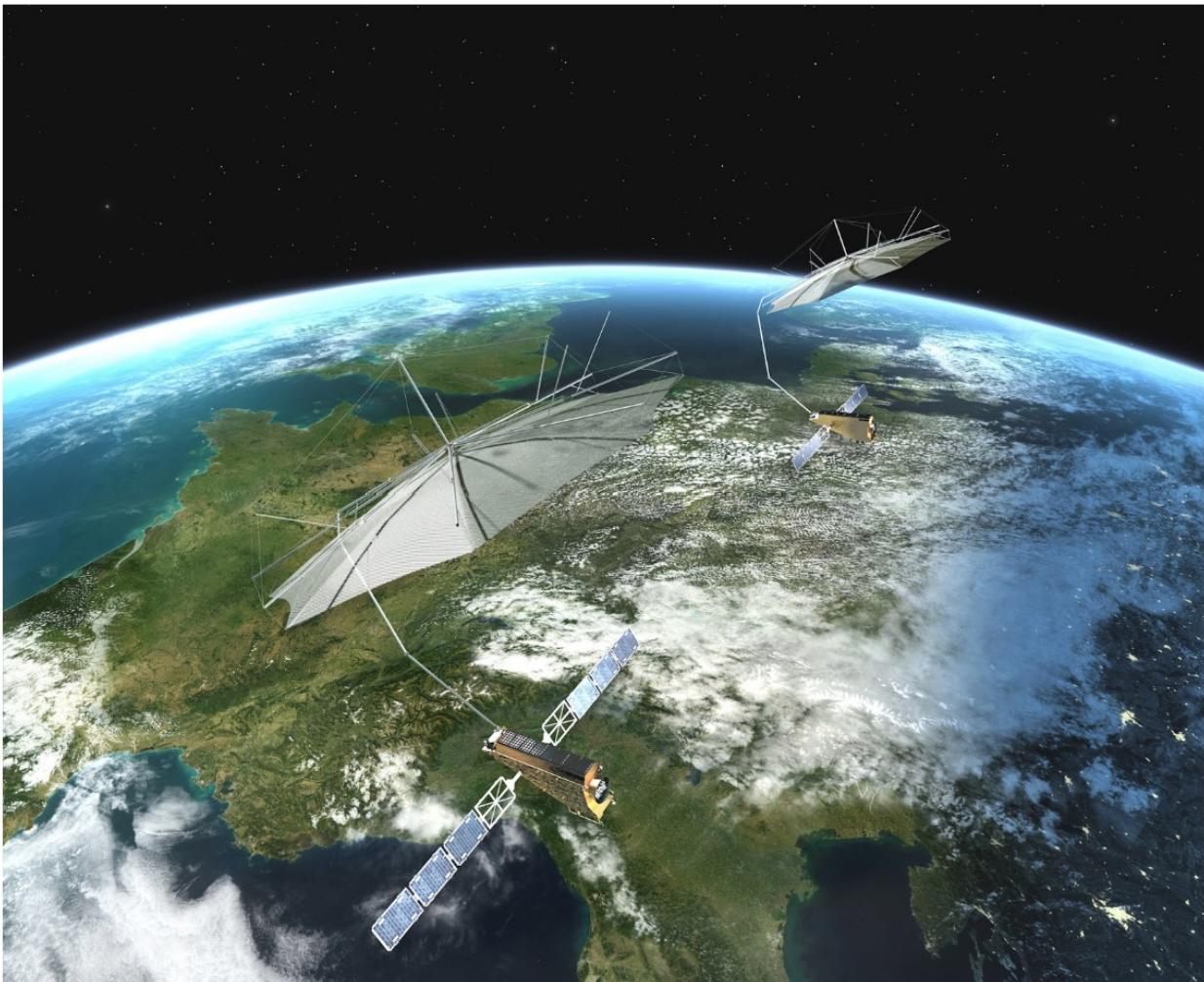
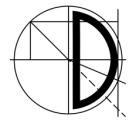
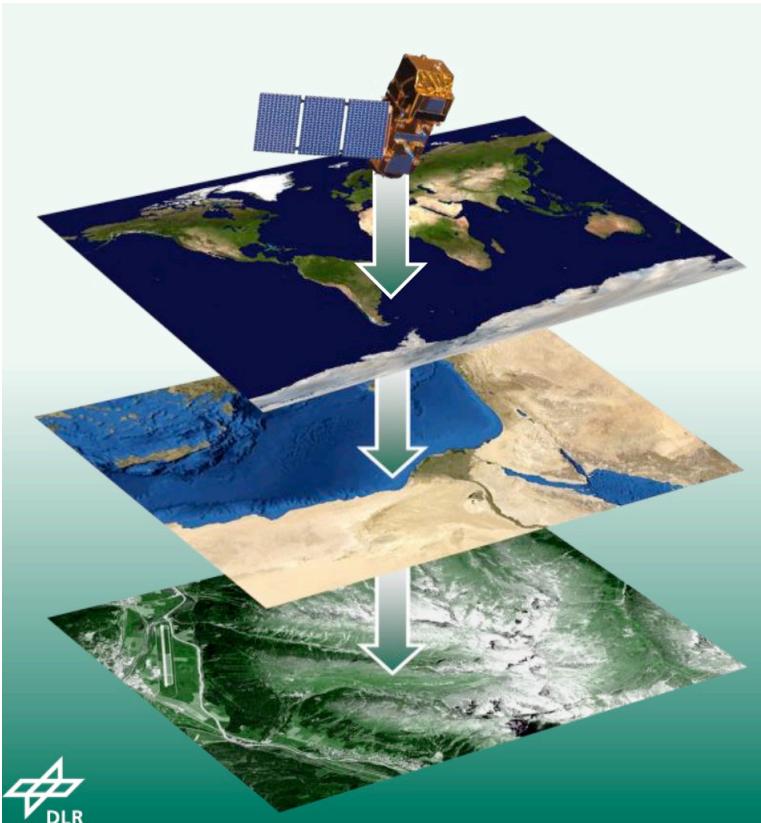


Image Sources: DLR EOC

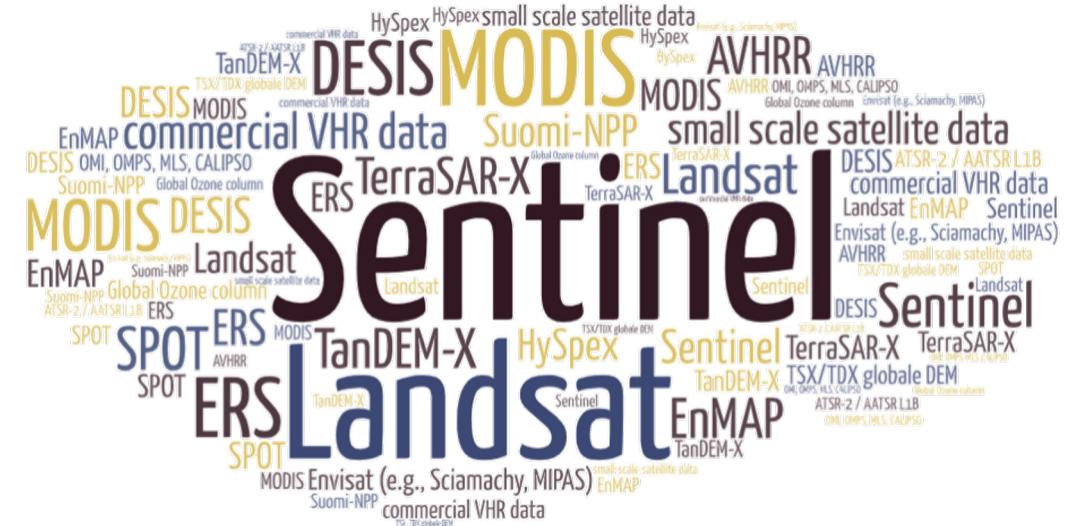


Earth Observation Today

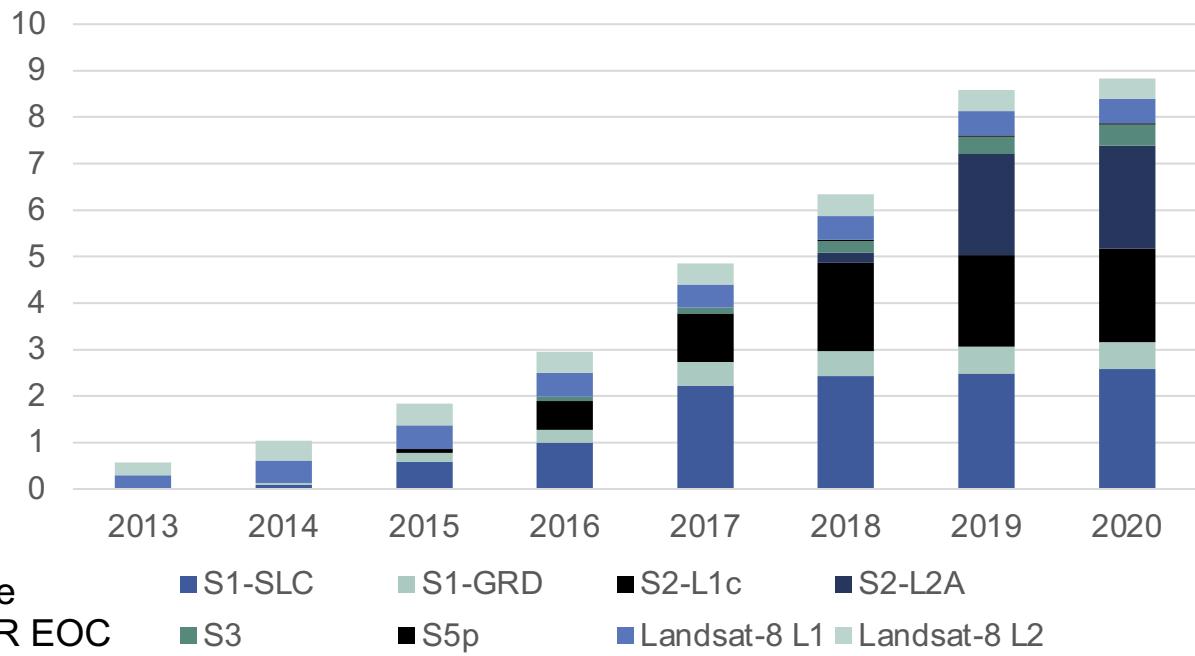
Continuous monitoring and quantification of global environmental change at all scales

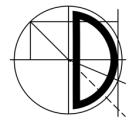


Data & Image Sources: DLR EOC



New EO data per year in Petabyte





Example: DLR Global SnowPack Proc. Workflow

Data input

GMTED2010
DEM

MOD10A1

MYD10A1

ESA DUE
GlobSnow

1) Pre-Processing

Resampling to MODIS Tiles

DEM for each MODIS Tile

Classification of Layer 'NDSI_Snow_Cover'

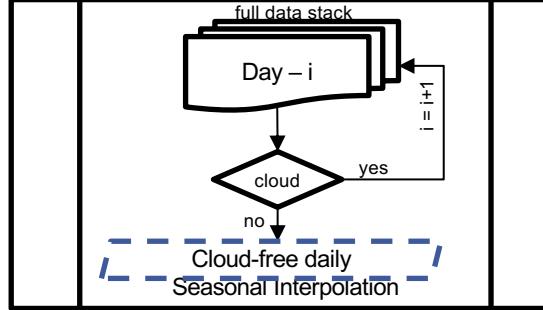
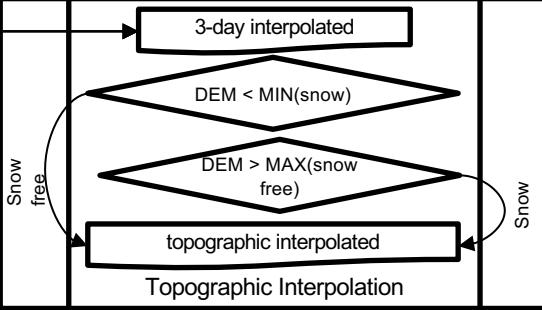
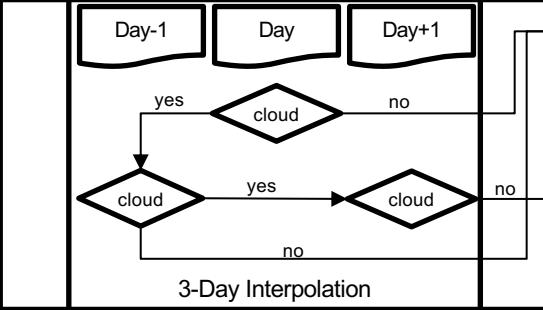
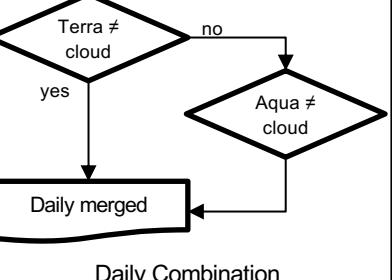
Global SnowPack Array
Stacked for hydrolog. year ± 1 day

SWE to Snow Cover

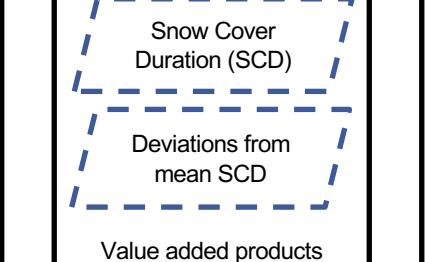
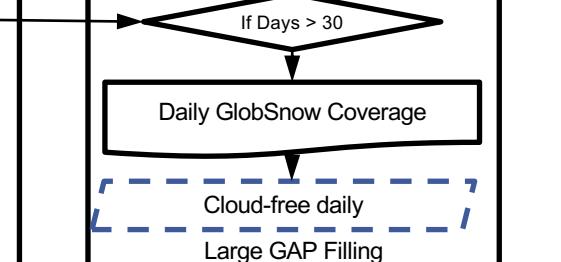
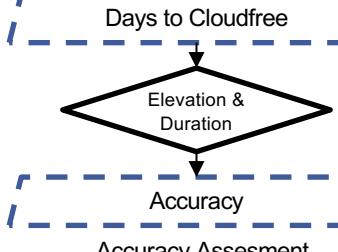
Projection to MODIS Tile

Daily GlobSnow Coverage

2) Processing

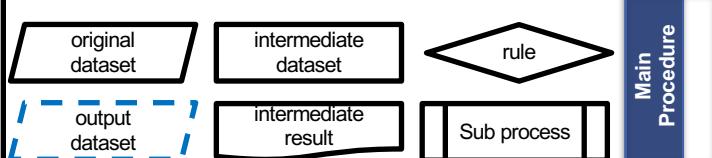


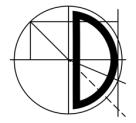
3) Post-Processing



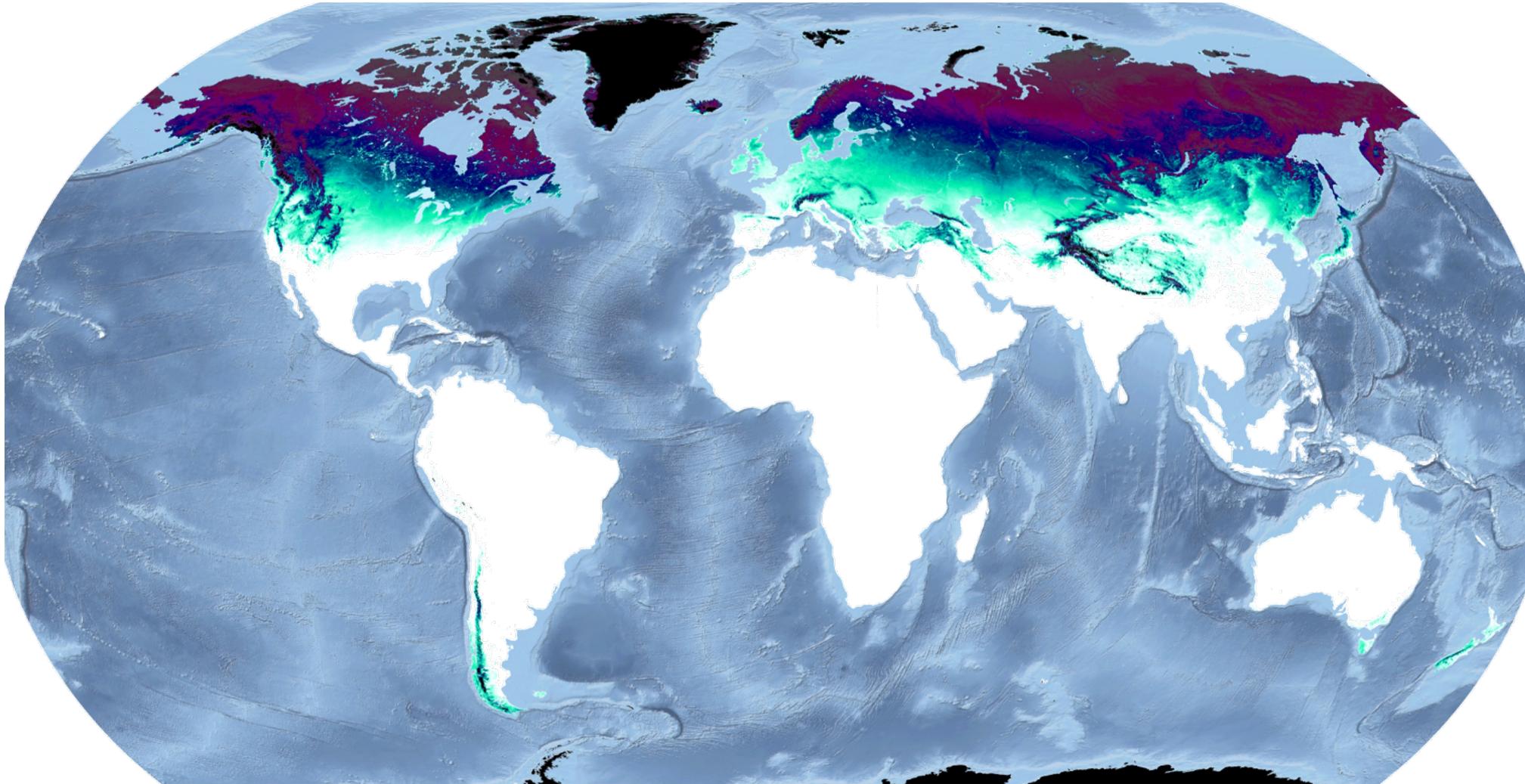
Global SnowPack

Legend





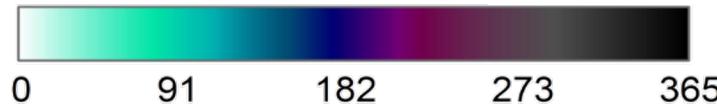
Example Result: DLR Global SnowPack

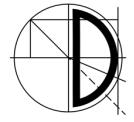


0 2.500 5.000 km

Projection: Winkel

Mean Snow Cover Duration 2000-2020





HPDA terrabyte project @ DLR



Earth Observation Center



Leibniz-Rechenzentrum
der Bayerischen Akademie der Wissenschaften



- Pilot project -
World Settlement Footprint WSF

```
import openeo
```

```
# First, we connect to the back-end and authenticate ourselves via Basic authentication.  
con = openeo.connect("https://earthengine.openeo.org")  
con.authenticate_basic("group11", "test123")
```

```
# Now that we are connected, we can initialize our datacube object with the area around Vienna  
# and the time range of interest using Sentinel 1 data.
```

```
datacube = con.load_collection("COPERNICUS/S1_GRD",  
                               spatial_extent={"west": 16.06, "south": 48.06, "east": 16.65, "north": 48.65},  
                               temporal_extent=[["2017-03-01", "2017-06-01"],  
                               bands=["VV"]])
```

```
# Since we are creating a monthly RGB composite, we need three (R, G and B) separated time ranges.  
# Therefore, we split the datacube into three datacubes by filtering temporal for March, April and May.  
march = datacube.filter_temporal("2017-03-01", "2017-04-01")  
april = datacube.filter_temporal("2017-04-01", "2017-05-01")  
may = datacube.filter_temporal("2017-05-01", "2017-06-01")
```

```
# Now that we split it into the correct time range, we have to aggregate the timeseries values.  
# Therefore, we make use of the Python client function 'mean_time', which reduces the time dimension.
```

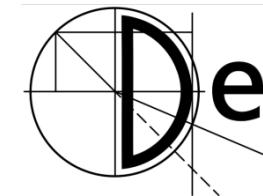


```
# With the last process we have finished the datacube definition and can create and start the job.
```

```
job = datacube.send_job()  
job.start_and_wait().download_results()
```



Fragen?



Descartes
research

<http://se.informatik.uni-wuerzburg.de>

<http://descartes.tools>