



10 mins about

PROCESS

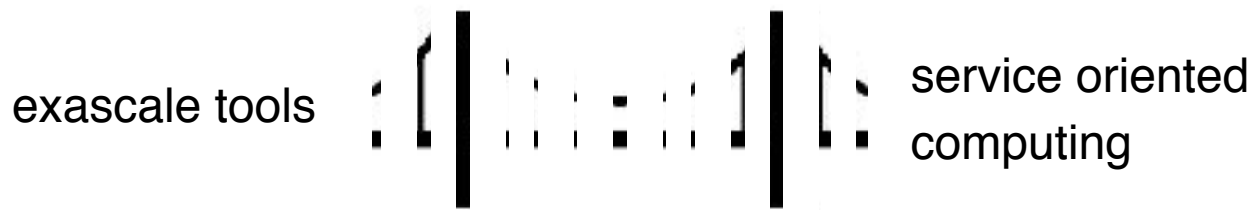
PROviding Computing
solutions for ExaScale
challengeS



*PROCESS aims at creating a shared architectural framework to provide a **user-friendly** service platform for exascale computation and very large datasets analysis (PB).*



Motivation: a mid-level infrastructure is missing

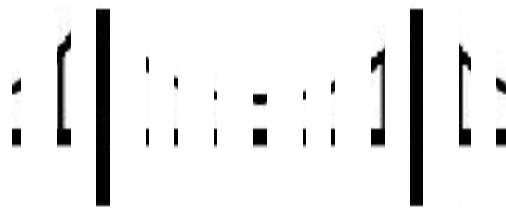


Scientific communities face:

- prevented data relocation due to dataset sizes
- need for tailored “**expert only**” solutions to access HPC

Motivation: a mid-level infrastructure is missing

exascale tools



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SUPERMUC JOB COMMAND

```
#!/bin/bash
#
#@ energy_policy_tag = my_tag
#
#@ job_type = MPICH
#@ class = large
#@ node = 1000 #@ island_count=2,4
#@ total_tasks=16000
### other example
###@ tasks_per_node = 16
###@ island_count = 1,18
#@ wall_clock_limit = 1:20:30
#@ job_name = mytest
#@ network.MPI = sn_all_not_shared.us
#@ initialdir = $(home)/mydir
#@ output = job$(jobid).out
#@ error = job$(jobid).err
#@ notification=always
#@ notify_user=erika.mustermann@xyz.de
#@ queue
. /etc/profile
. /etc/profile.d/modules.sh
#setup of environment
module unload mpi.ibm
module load mpi.intel
36Best Practice Guide SuperMUC v1.0
export OMP_NUM_THREADS=2
#optional:
#module load mpi_pinning/hybrid_blocked
mpixec -n 16000 ./myprog.exe
```

Objective: building a bridge

exascale tools



service oriented
computing

Objective: building a bridge



Such infrastructure will:

- absorb the **needs** in terms of **data** and **computing** intensive science
- soften the learning curve for scientific communities using the service

Service prototypes: representing user needs

5 different use cases:

#1- Exascale learning of medical image data

#2 - Square Kilometre Array SKA

#3 - Supporting innovation based on global disaster risk data

#4 - Ancillary pricing for airline revenue management

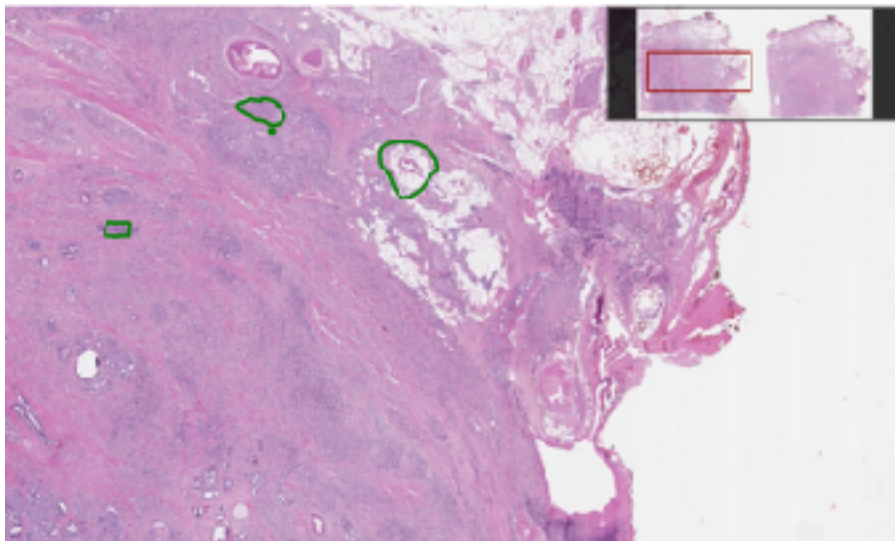
#5 - Agro-Copernicus

Medical
Imaging



#1

Exascale learning on medical images



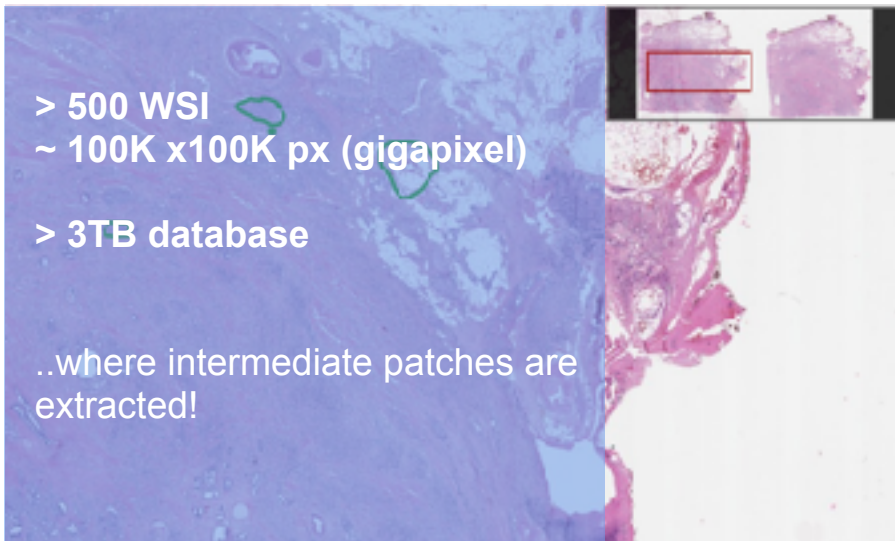
#1

Exascale learning on medical images

> 500 WSI
~ 100K x100K px (gigapixel)

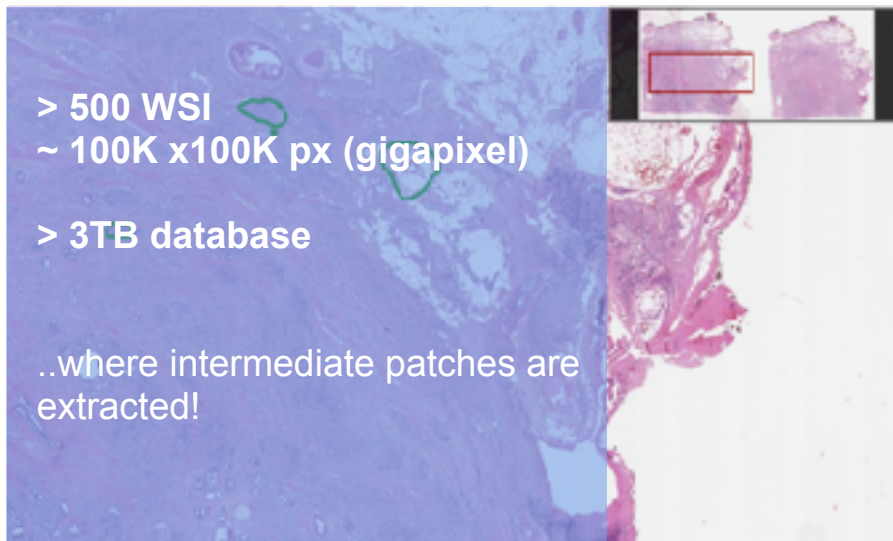
> 3TB database

..where intermediate patches are extracted!



#1

Exascale learning on medical images

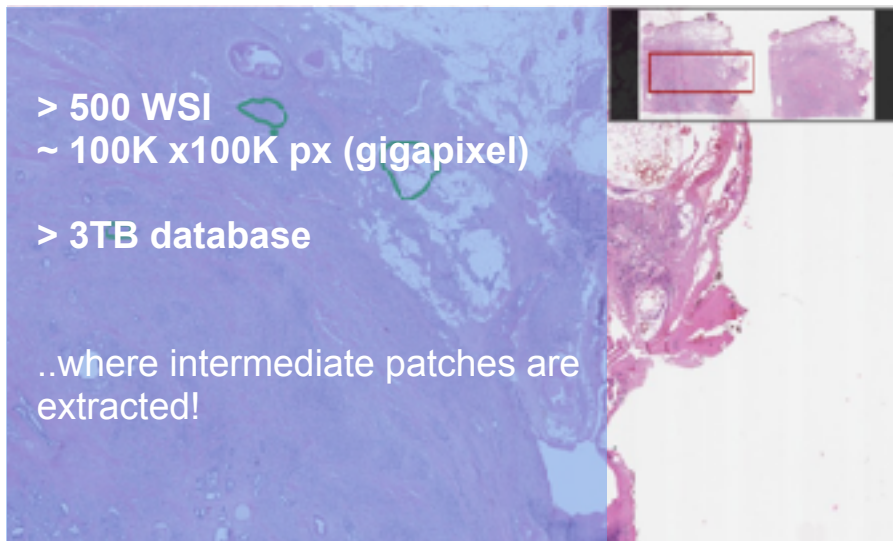


Possible applications:

- content-based search

#1

Exascale learning on medical images

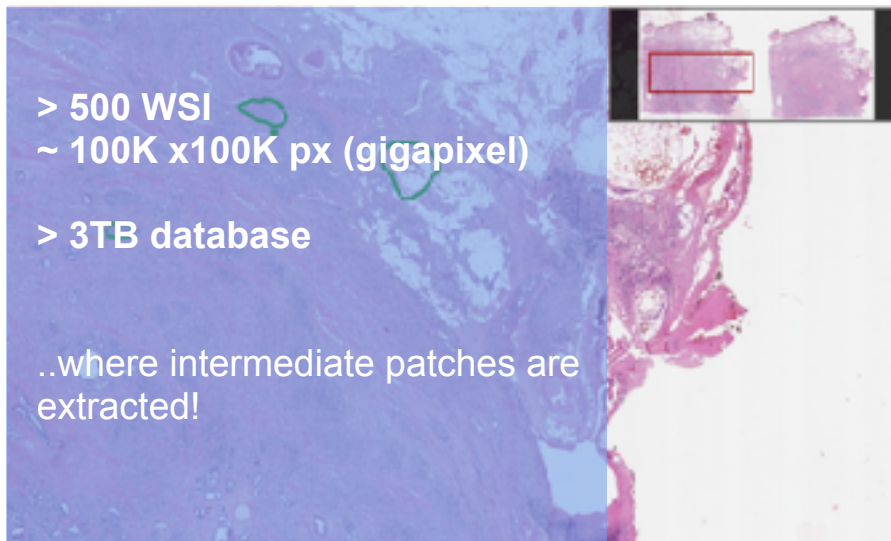


Possible applications:

- content-based search
- patients clustering

#1

Exascale learning on medical images

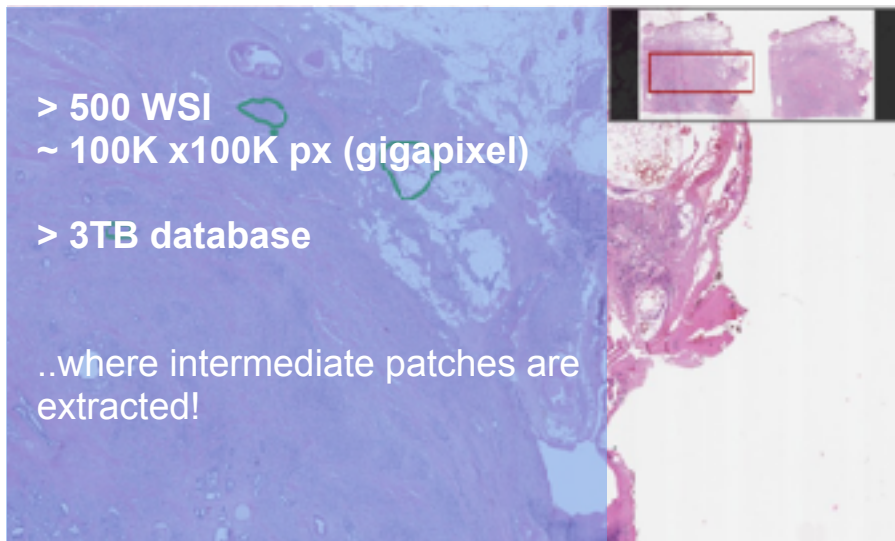


Possible applications:

- content-based search
- patients clustering
- aided ROI annotation

#1

Exascale learning on medical images



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- patients clustering
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Brainstorming other suggestions?

Use case #1 - Roadmap Plan:



0. development in Sierre, port to LRZ

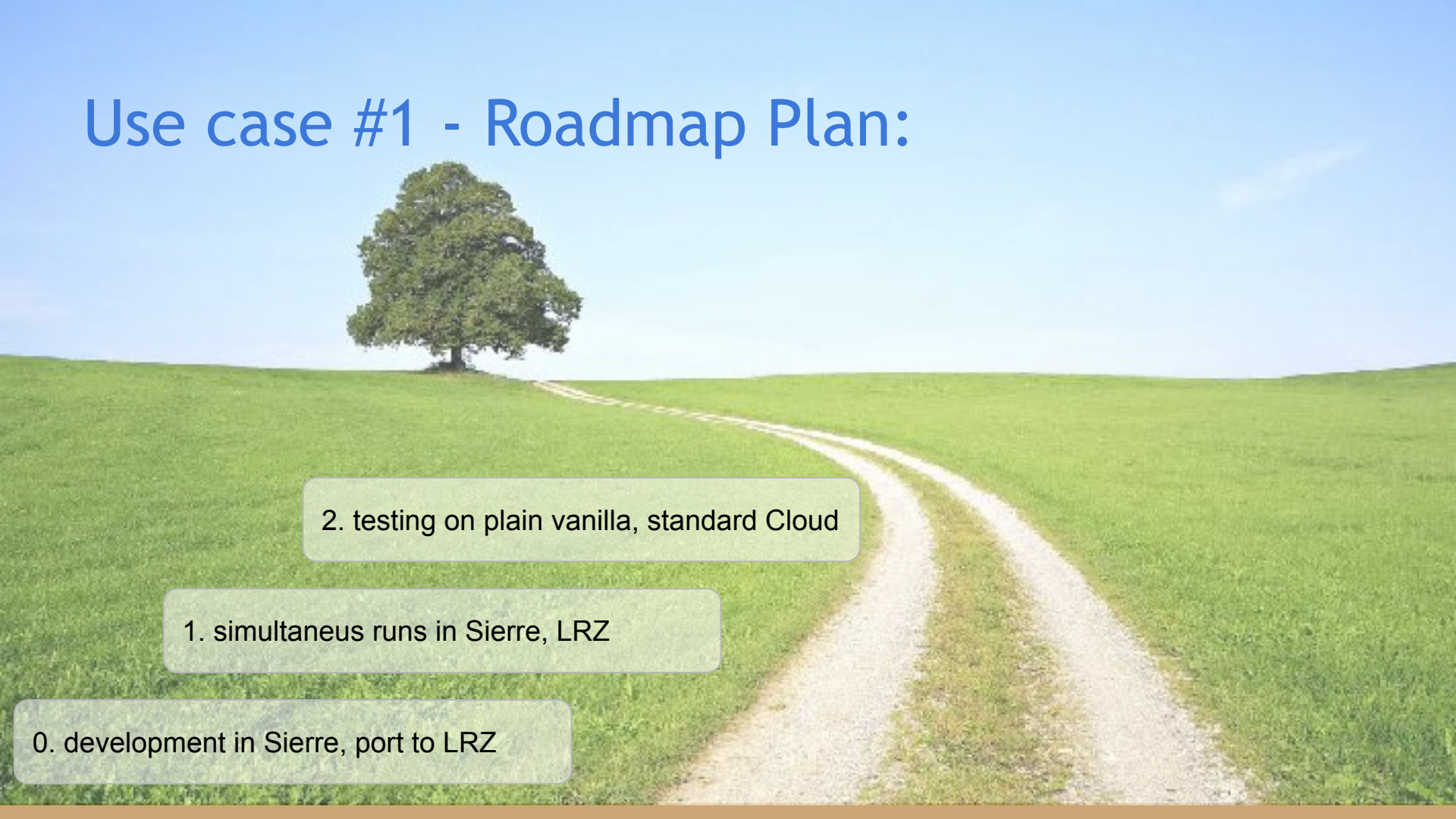
Use case #1 - Roadmap Plan:

A scenic landscape featuring a dirt road that curves from the bottom right towards a large, leafy tree on a grassy hill in the distance. The sky is clear and blue.

1. simultaneous runs in Sierre, LRZ

0. development in Sierre, port to LRZ

Use case #1 - Roadmap Plan:



2. testing on plain vanilla, standard Cloud

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3. distribute

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4. evaluation from not e-infrastructure veterans

0. development in Sierre, port to LRZ

Use case #1 - Roadmap Plan:

5. Success!

4. evaluation from not e-infrastructure veterans

3. distribute

2. testing on plain vanilla, standard Cloud

1. simultaneous runs in Sierre, LRZ

0. development in Sierre, port to LRZ

Use case #1 - Tech details:

Starting dataset: **Camelyon 17** (testing still in download, 500 WSI, slide, patient and lesion annotations),
Camelyon 16 (400 WSI, lesion annotations)
Further datasets: Pubmed, TUPAC16, TCGA, SKIPOGH

Patch-extraction module based on the Camelyon17 challenge best results (Lunit).
Intermediate computation results storage in hierarchical format h5ds
Basic set of Deep Learning Models in development
Monitoring of validation accuracy and loss during training

Official Data infrastructure: tbd (meeting in Amsterdam early February)
Key Performance Indicators: Data volume, performance increase

Tools used so far: Python 2.7, Tensorflow 1.4.0, Keras 2.1.2
Computing infrastructure: CPU for patch extraction, GPU #2 on desuto for network training

Questions?

Thanks!