

10 mins about

# PROCESS

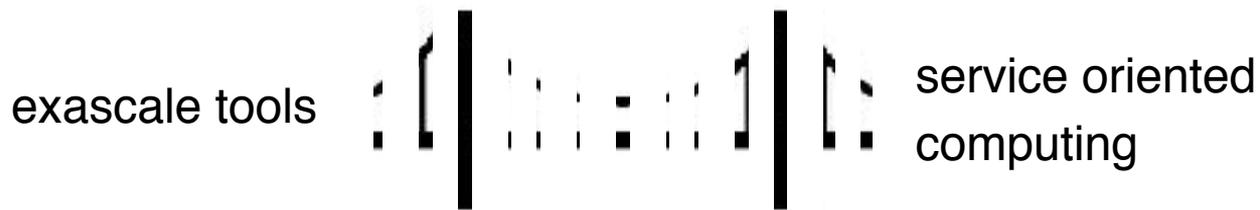
PROviding Computing  
solutions for ExaScale  
challengeS



*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

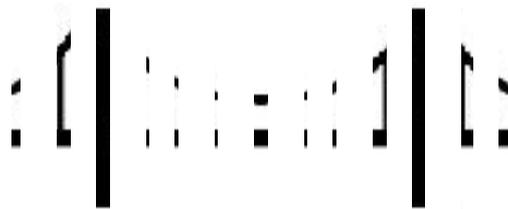


Scientific communities face:

- 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



Scientific communities face:

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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

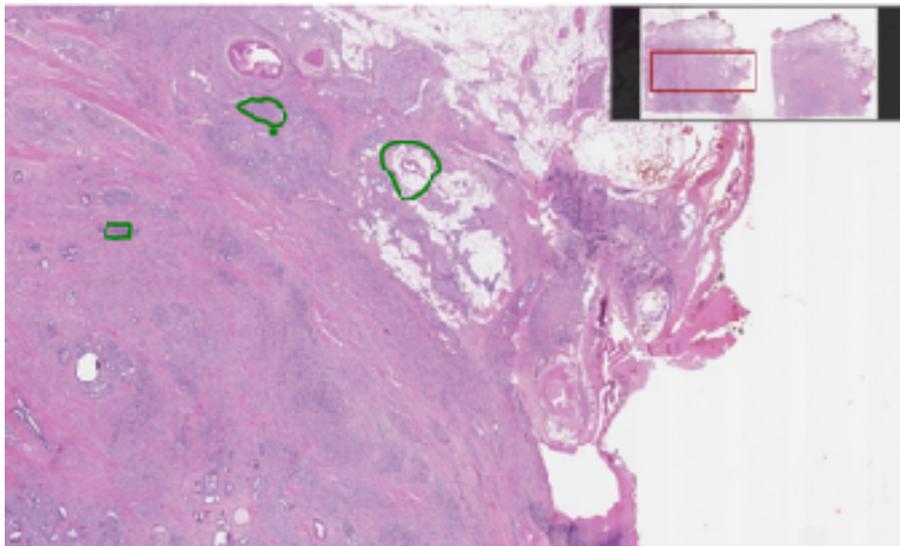
#5 - Agro-Copernicus

Medical  
Imaging



# #1

## Exascale learning on medical images



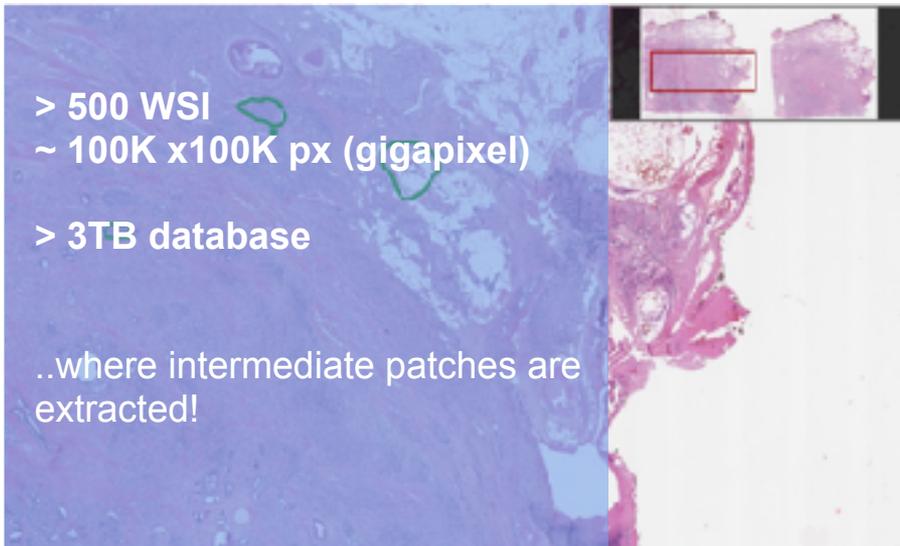
# #1

## Exascale learning on medical images

> 500 WSI  
~ 100K x100K px (gigapixel)

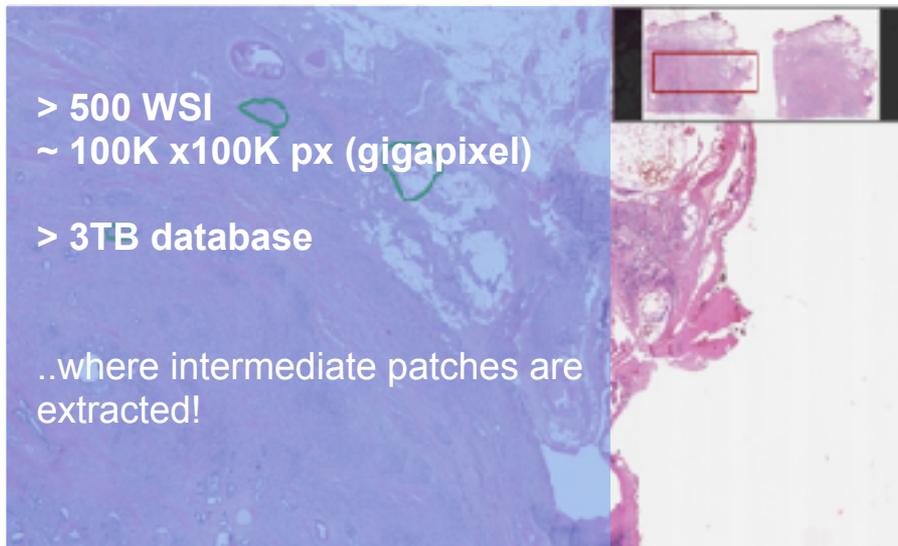
> 3TB database

..where intermediate patches are extracted!



# #1

## Exascale learning on medical images

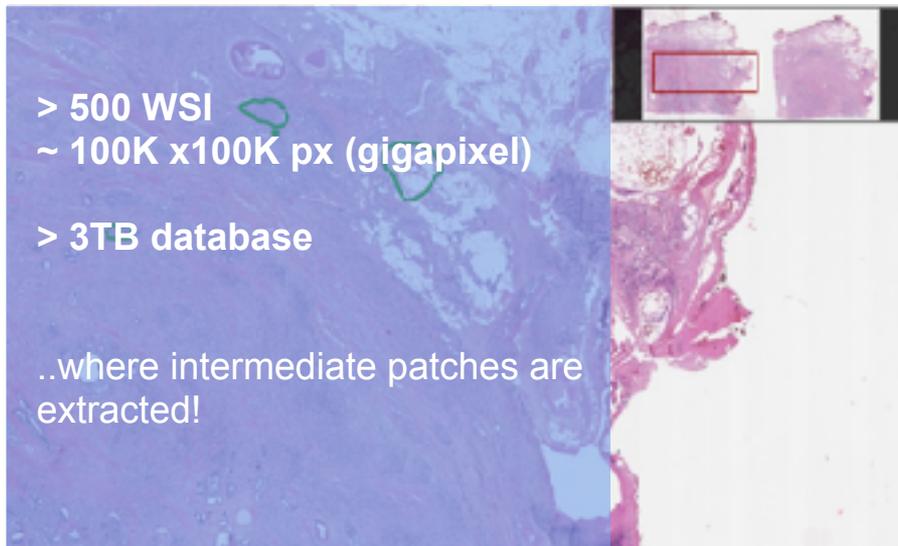


Possible applications:

- content-based search

# #1

## Exascale learning on medical images

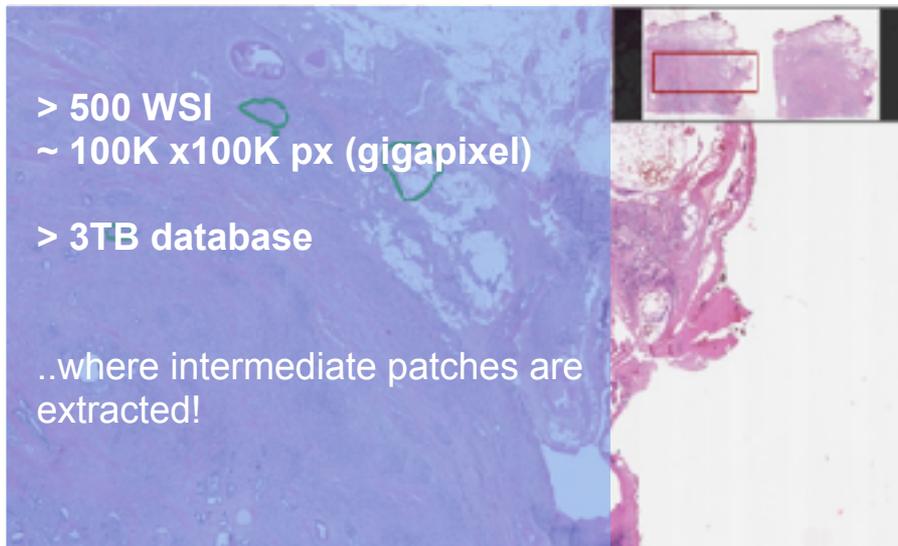


Possible applications:

- content-based search
- patients clustering

# #1

## Exascale learning on medical images

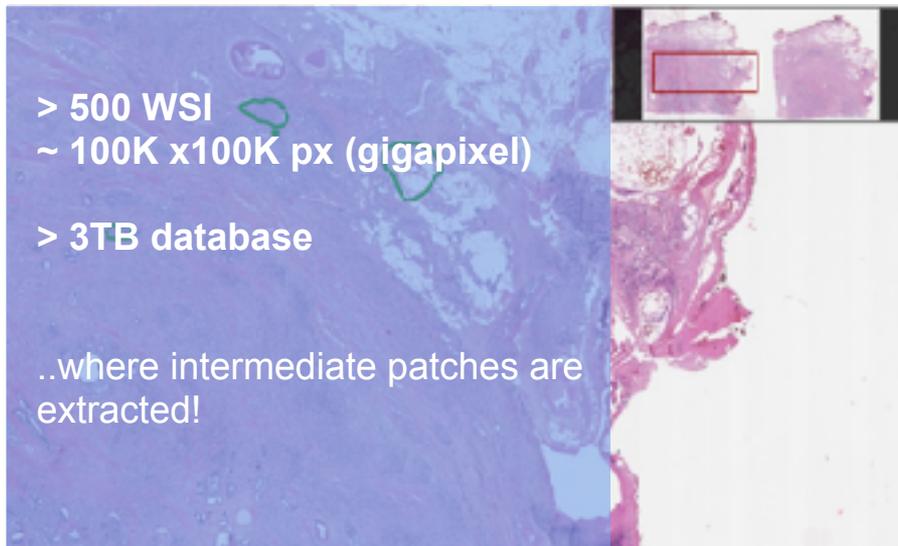


Possible applications:

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- aided ROI annotation

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## Exascale learning on medical images



Possible applications:

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- patients clustering
- aided ROI annotation

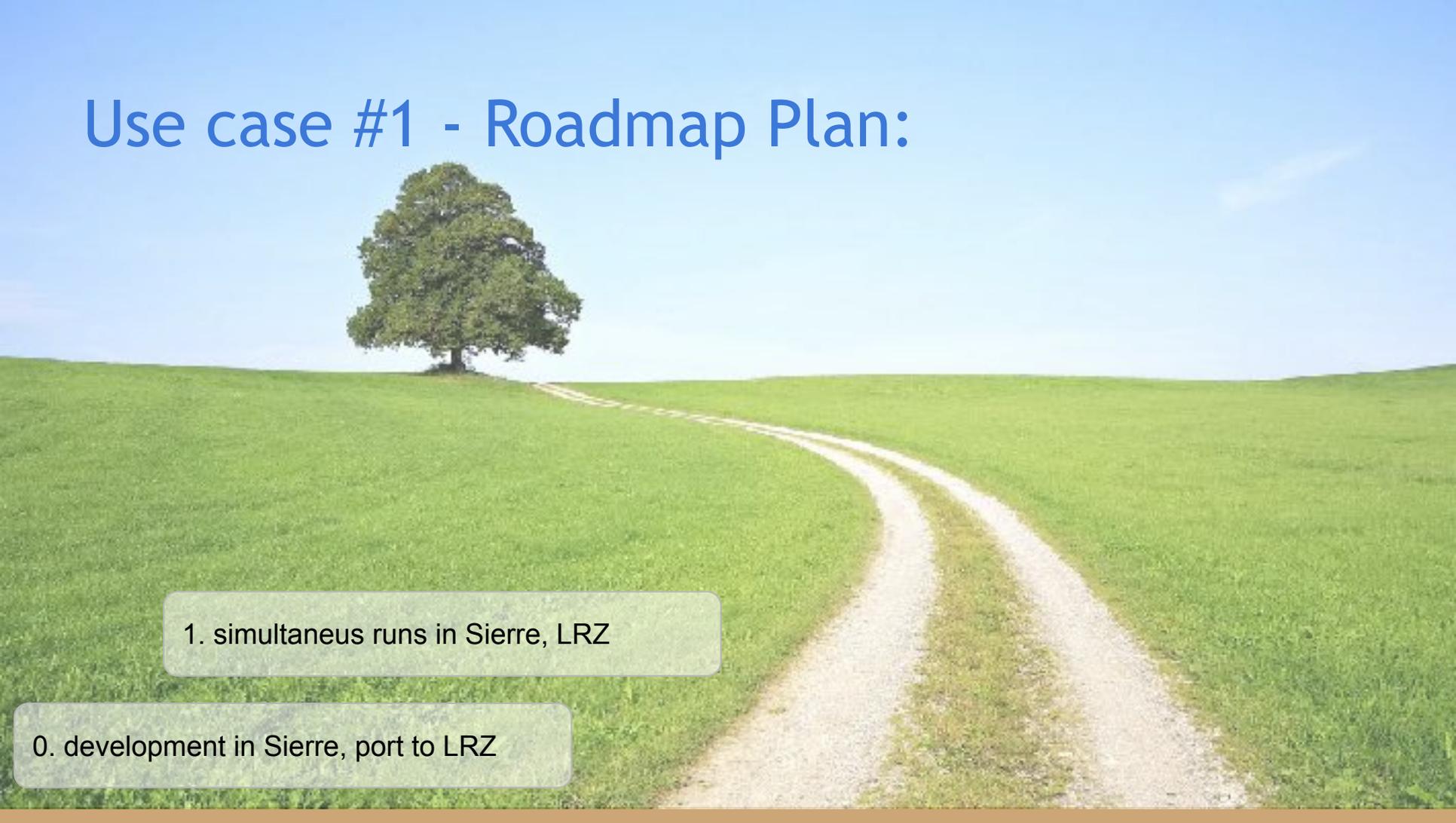
Brainstorming . . . . other suggestions?

# Use case #1 - Roadmap Plan:



0. development in Sierre, port to LRZ

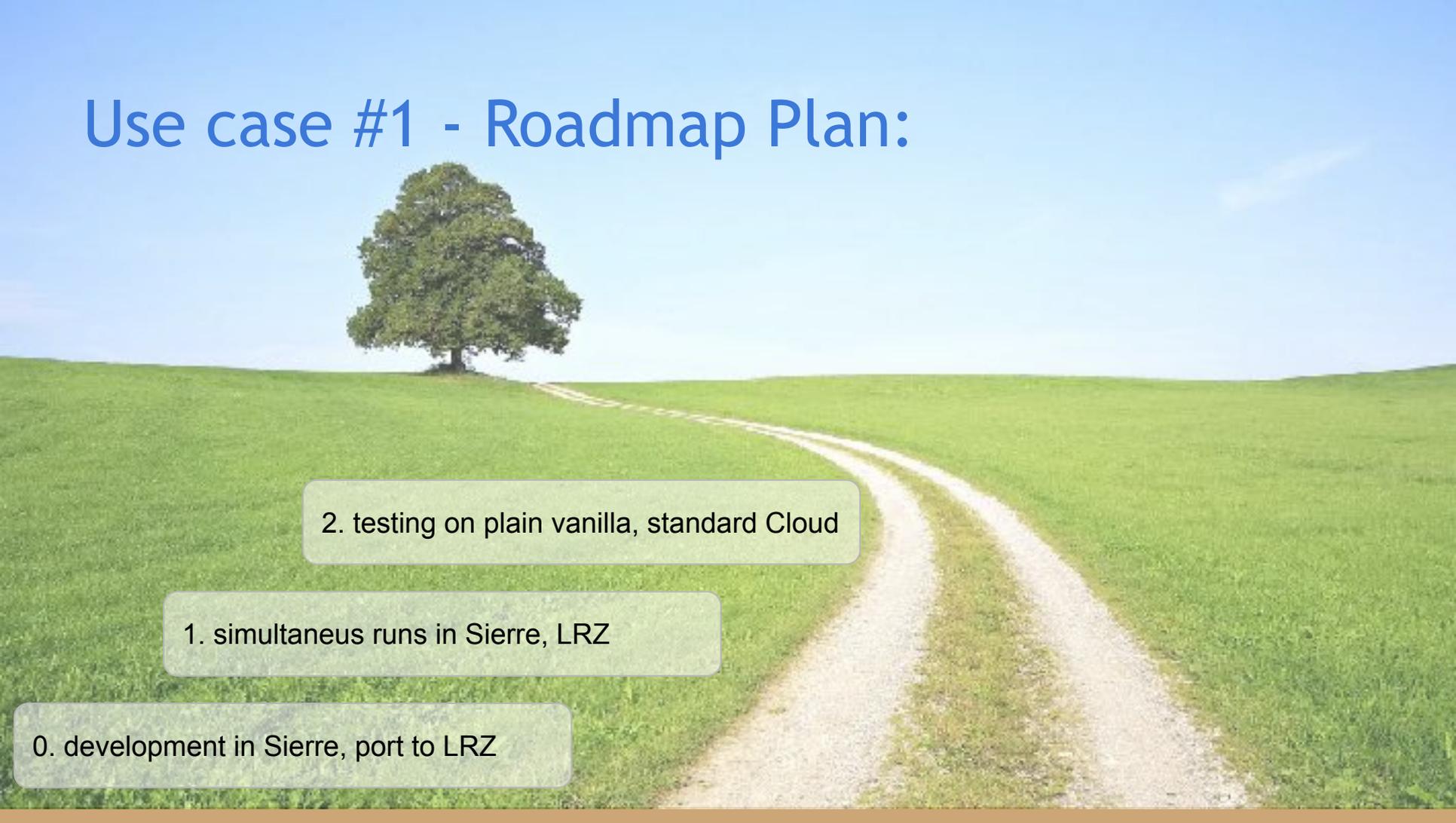
# Use case #1 - Roadmap Plan:

A scenic landscape featuring a dirt road that curves through a vibrant green field. In the distance, a large, full-canopied tree stands on a slight rise. The sky is a clear, bright blue with a few wispy clouds. The overall atmosphere is peaceful and open.

1. simultaneous runs in Sierre, LRZ

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, solitary tree on a grassy hill in the distance. The sky is clear and blue, and the field is a vibrant green.

2. testing on plain vanilla, standard Cloud

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

1. simultaneous runs in Sierre, LRZ

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3. distribute

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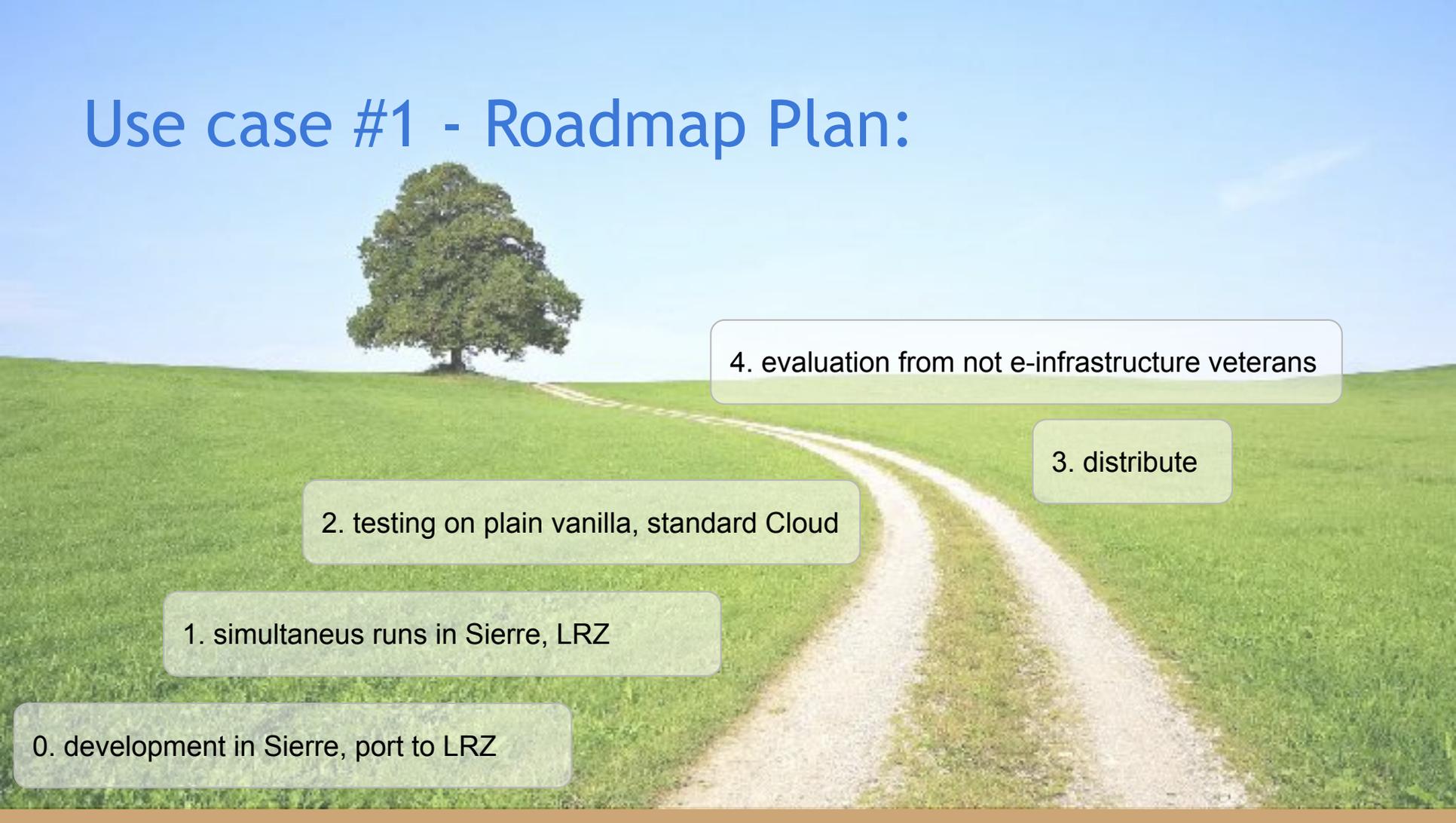
1. simultaneous runs in Sierre, LRZ

0. development in Sierre, port to LRZ

2. testing on plain vanilla, standard Cloud

3. distribute

4. evaluation from not e-infrastructure veterans



# 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!