# Improving the scalability of an Universiteit Utrecht environmental modelling framework Faculty of Geosciences to allow for large-scale high-resolution geosimulations

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## Introduction

The high computational requirements for stochastic spatio-temporal modelling, and an increasing demand to run models over large areas at high resolution, e.g. in global hydrological modelling or epidemiology, require an optimal use of available, heterogeneous computing resources. Domain-specific modelling software used by environmental scientists, however, often do not provide built-in capabilities to distribute model runs over the compute nodes of a supercomputer. We propose to enhance the PCRaster model building framework with built-in capabilities to run models on various hardware platforms, resulting in hardware scalable models

### Data assimilation



Update model state when observations are available using Bayes' theorem

#### Solution scheme

• • •

```
for each period in periods:
   for each n in Monte Carlo Samples:
       for each t in period:
```

### that can be constructed by environmental modellers.



# The PCRaster modelling framework

The PCRaster modelling framework

- Is targeted at the development of spatio-temporal models
- Fast model development and execution
- Scripting environments: PCRcalc and Python
- Rich set of model building blocks for manipulating raster maps
- Framework for stochastic spatio-temporal model building
- Framework for data assimilation
- Tool for visualisation of spatio-temporal stochastic data
- Runs on Linux, Microsoft Windows and Apple OS X



evaluate Bayes' theorem

Solution framework (Python)	
<pre>def suspend(self):     self.report(self.snow, "s")</pre>	store model state at
<pre>def updateWeight(self):     sum = exp(maptotal(((obs - mod)**2)/         (2.0 * (observedStd ** 2))))     weight = exp(sum)     return weight</pre>	] calculate weight of Mo solution of Bayes' equ
<pre>def resume(self):     self.read("s")</pre>	read model state at s

#### end of period

onte Carlo sample required for ation and return to framework

start of next period

# PCRaster on shared memory systems

A binding between PCRaster and Fern provides about 50 parallel local and focal operations. Fern is a is a highly generic C++ software library for raster processing that can be tailored to the configuration of a modelling framework.



- Can be downloaded for free and is open source

# Stochastic spatio-temporal modelling

Model



### Solution scheme

for each n in Monte Carlo samples: for each t in time steps:  $\mathbf{z}_{t}^{(n)} = f(\mathbf{z}_{t-1}^{(n)}, \mathbf{i}_{t}^{(n)}, \mathbf{p}^{(n)})$ 

### Building blocks

discharge = kinematic(flowDir, precipitation, ...) result map spatial function input maps

#### Solution framework (Python)

Policy	Behaviour	
execution	kind of parallelism	Configuration options per algorithm
input no-data	how to handle no-data in input	
output no-data	how to handle no-data in output	
out-of-domain	how to test for out-of-domain values in input	
out of range	how to handle out-of-range values in algorithm result	

template<</pre> template<typename> class OutOfDomainPolicy, // <--</pre> typename InputNoDataPolicy, typename OutputNoDataPolicy, typename ExecutionPolicy, typename Value, typename Result void sqrt(

InputNoDataPolicy const& input\_no\_data\_policy,

OutputNoDataPolicy& output\_no\_data\_policy,

ExecutionPolicy& execution\_policy,

Excerpt of C++ implemenation of square root

from pcraster import \* values = readmap("raster.map") result = sqrt(values)

Value const& value,

Result& result);

Operation as used by modeller

## PCRaster on distributed memory systems

Algorithms that operate on an irregular topology, such as material

```
from pcraster import *
from pcraster.framework import *
```

```
class SnowModel(DynamicModel, MonteCarloModel):
 def __init__(self):
```

• • •	
<pre>def premcloop(self):     dem = self.readmap("dem")     self.ldd = lddcreate(dem,)</pre>	sets co and pa
<pre>def initial(self):    self.snow = scalar(0)</pre>	is run a Monte
<pre>def dynamic(self):   runoff = accuflux(self.ldd, rain)   self.report(runoff, "q")</pre>	is run f and for
<pre> def postmcloop(self):   mcpercentiles("q",percentiles,)</pre>	is run a statisti

onstant variables arameters

at t = 0 for each Carlo sample

for each Monte Carlo sample r each time step

at end calculating sampling statistics over Monte Carlo samples transport over a local drainage network, require a decomposition into fine grained sets of concurrent tasks for efficient execution. These tasks will be connected with other tasks from multiple algorithms into a task-graph, and an external HPX runtime library executes all tasks both on shared and distributed memory systems.

# Download and further information

http://www.pcraster.eu

https://github.com/geoneric/fern

http://stellar.cct.lsu.edu/

D. Karssenberg, O. Schmitz, P. Salamon, K. de Jong, and M. F. P. Bierkens. A software framework for construction of process-based stochastic spatio-temporal models and data assimilation. Environmental Modelling & Software, 25 (4):489–502, 2010. doi: 10.1016/j.envsoft.2009.10.004 M. P. de Bakker, K. de Jong, O. Schmitz, and D. Karssenberg. Design and demonstration of a data model to integrate agent-based and field-based modelling. Environmental Modelling & Software, 89:172-189, 2017. doi: 10.1016/

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