

# Modelling Subsurface Stratigraphy within the Taipei Basin using Markov Random Field Theory

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# Introduction

## Aims:

- Produce a random field model of the subsurface stratigraphy in the Taipei Basin.
- Assess the spatial correlation of this stratigraphy.
- Quantify the uncertainty within the geological model.

## Key terminology:

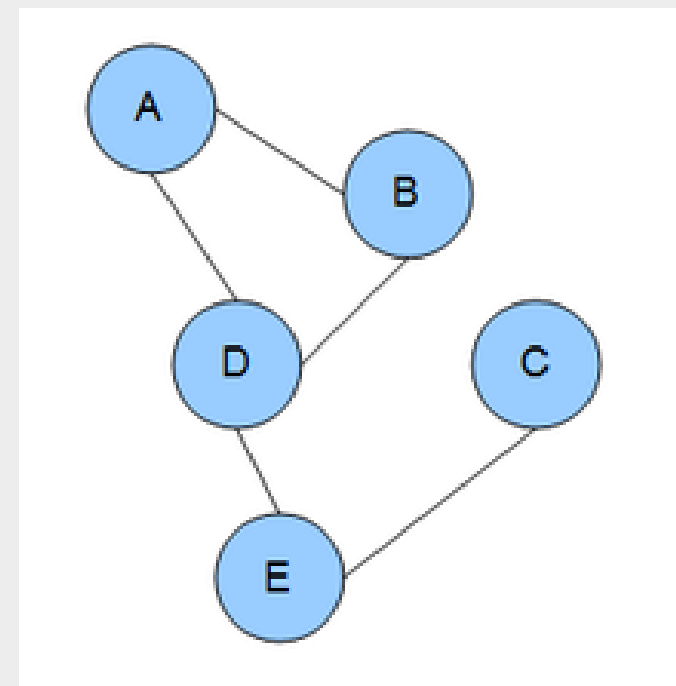
- **Spatial correlation** – refers to the tendency of neighbouring samples or locations in a spatial domain to be more similar to each other than those further away.
- **Uncertainty** – refers to the uncertainty of boundaries between different soil layers and lithological units.
- **Stochastic geological model** – a model that uses random or probabilistic methods to describe the uncertainty and variability of geological features and properties in the subsurface.

# Why is it important?

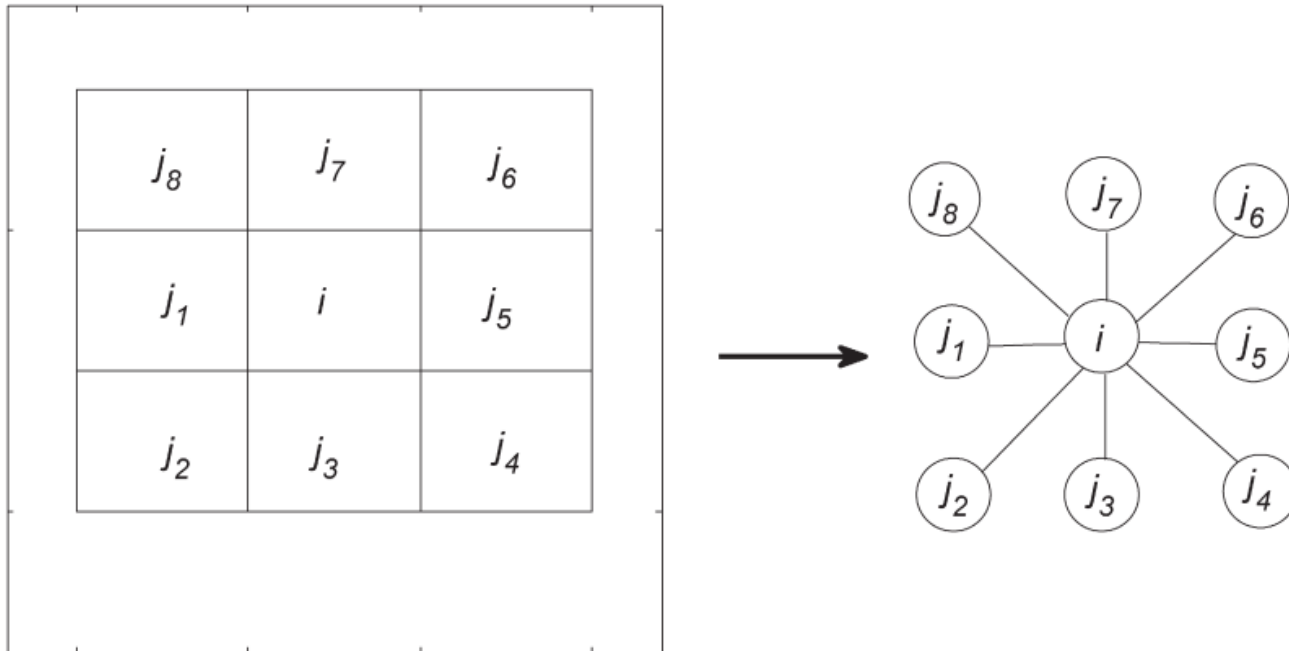
- Although there has been previous research about random field modelling, it hasn't been widely applied in real-case scenarios.
- Subsurface geological models are usually inaccurate due to **highly variable strata** and **sparse site investigation data**.
- Random field models could help **reduce uncertainty** in geological models, and therefore ground investigations as well.
- It is estimated that around 80% of problems discovered on construction projects are attributable to **unexpected ground conditions**.

# Markov random fields

- Markov Random Fields (MRFs) are graphical models used to represent **probability distributions over a set of random variables**.
- Variables are represented by **nodes** in the graph, and the **edges** between nodes represent conditional dependencies between them.
- The main idea behind MRFs is that the probability of a particular configuration of variables **depends only on the values of neighbouring variables**, and not on the values of distant or unrelated variables. (Markov property).
- This can be used to model the **spatial distribution** of different types of soil/rocks in a given area.
- The different types of soil or rocks would be represented as the variables in the model, and their spatial relationships would be represented by the edges connecting the nodes.



# Neighbourhood system



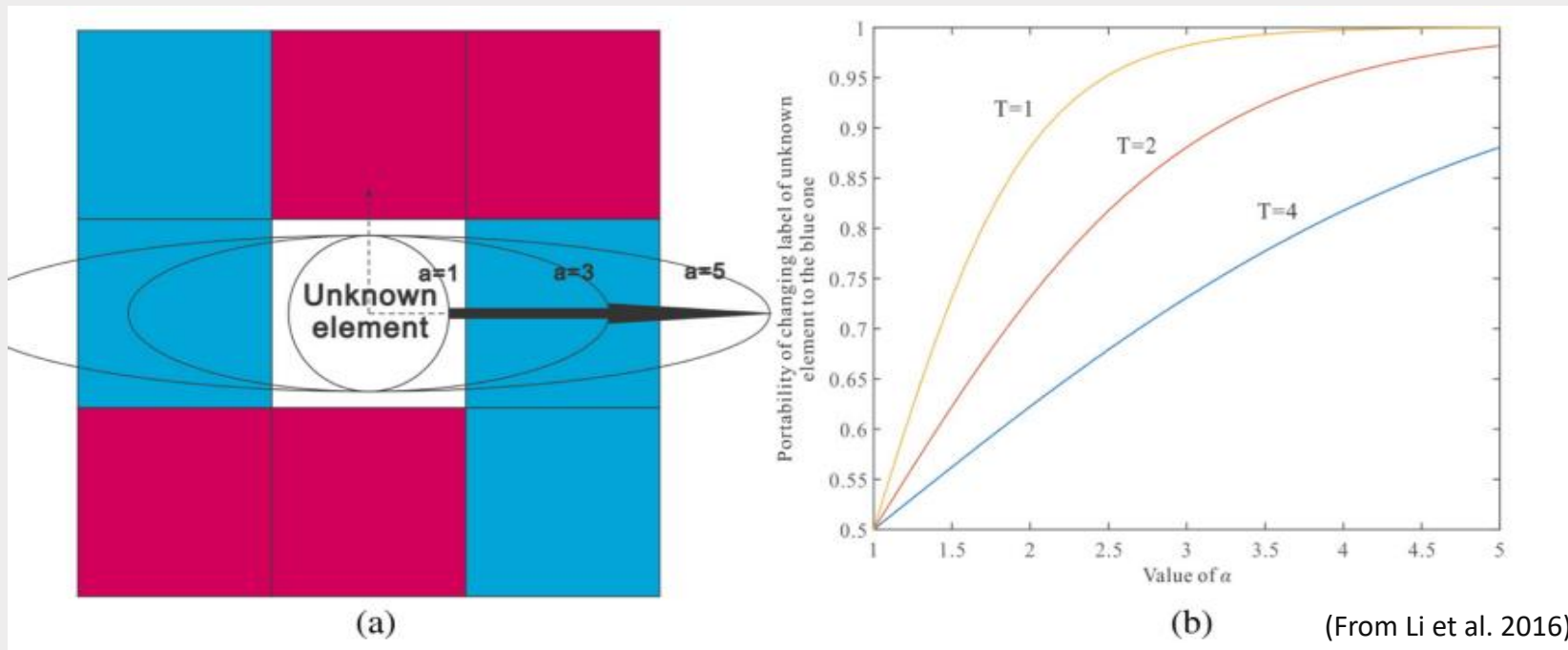
- Geological model is constructed by discretizing the geological body of interest into small square elements.
- The neighbourhood system defines the set of variables that are considered when computing the probability of a given variable, given the values of its neighbours.
- Neighbours are spatially related.
- Spatial correlation divided into 2 parameters

# Spatial Correlation

The spatial correlation in the local neighbourhood system is divided into two components:

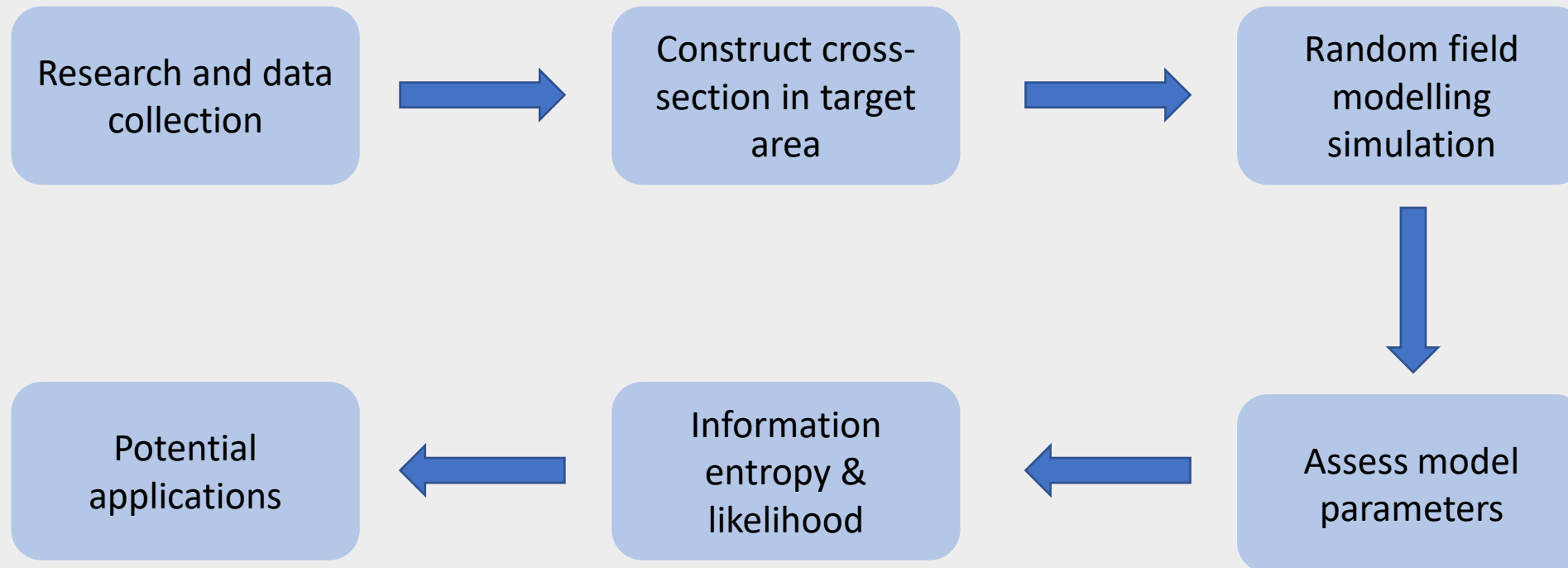
- $\psi$  – orientation information of geological formations
- $a$  – the ratio of strength of tangential correlation and normal correlation

a larger value of 'a' indicates a greater degree of correlation in the model, while a smaller value of 'a' allows for more spatial variability or heterogeneity.



(From Li et al. 2016)

# Workflow



# Data Collection

## Data used for cross section:

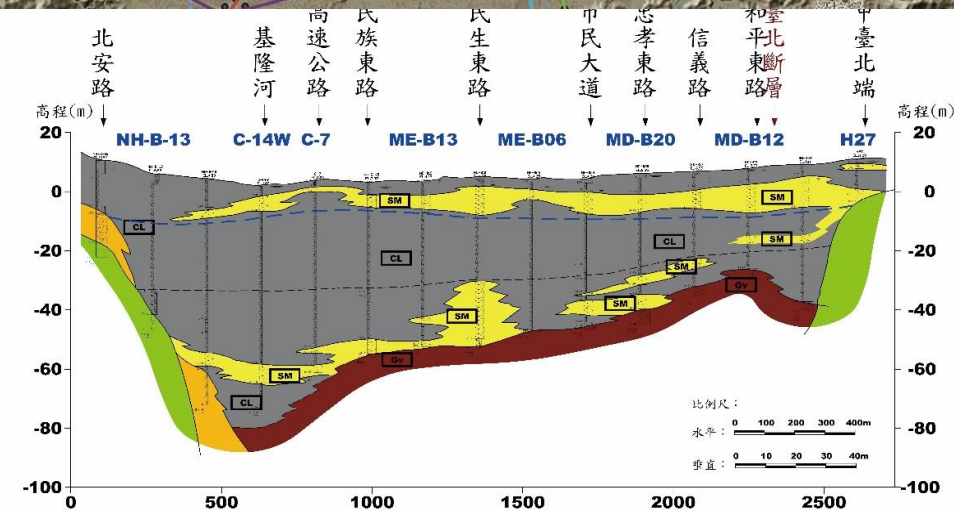
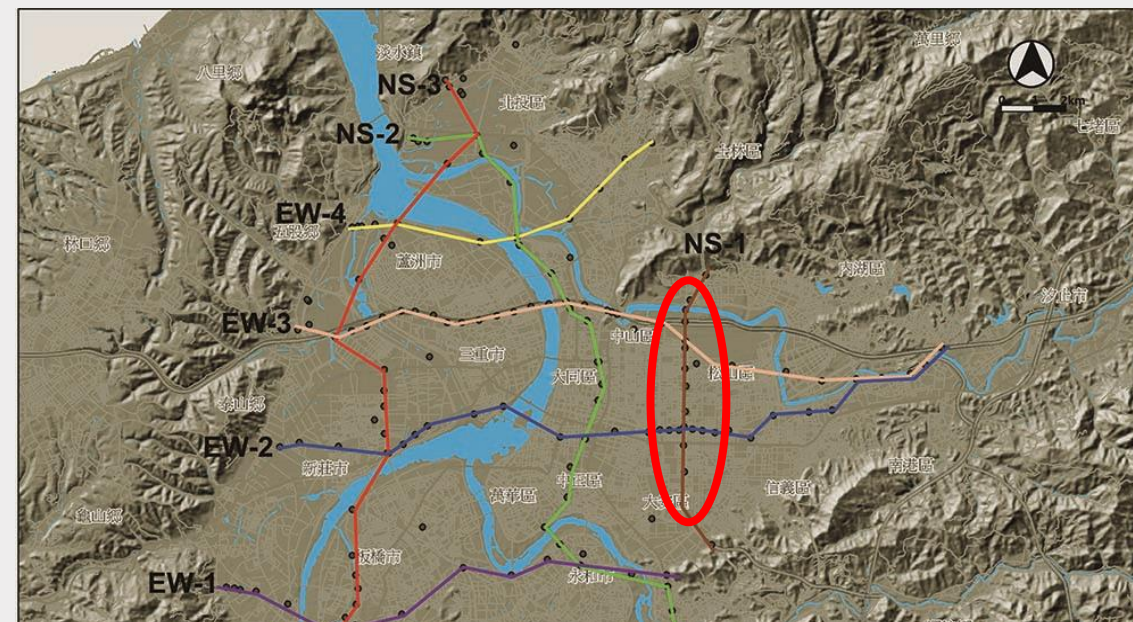
- Borehole data
- Lab test data

Borehole data and lab test results in the Taipei Basin from CGS.

A total of 22 boreholes were used.

## Data input for modelling:

1. Surface data – elevation and distance
2. Soil types
3. Layer boundaries
4. Location of boreholes

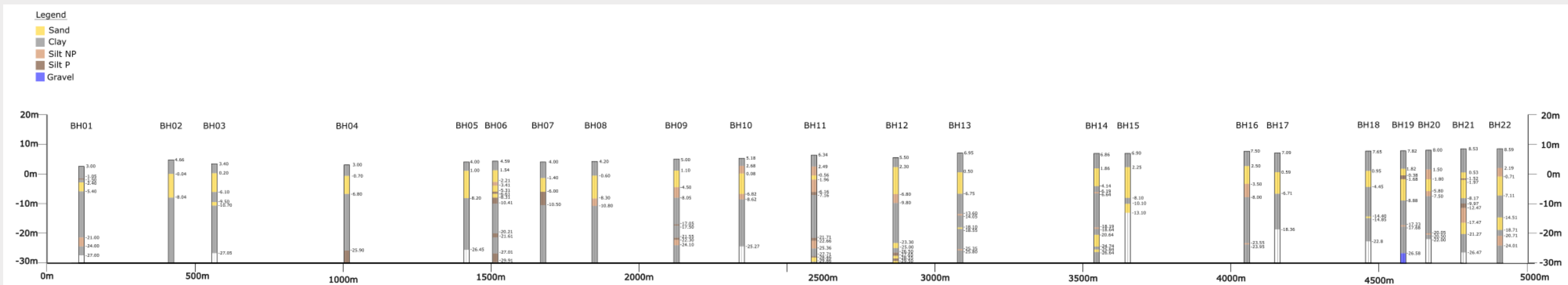


圖例:

- 粉砂質粘土
- 粉砂質砂(夾礫石)
- 崩積層
- 景美層
- 第三紀岩盤
- 推估最大海漫面
- 海進起始界面

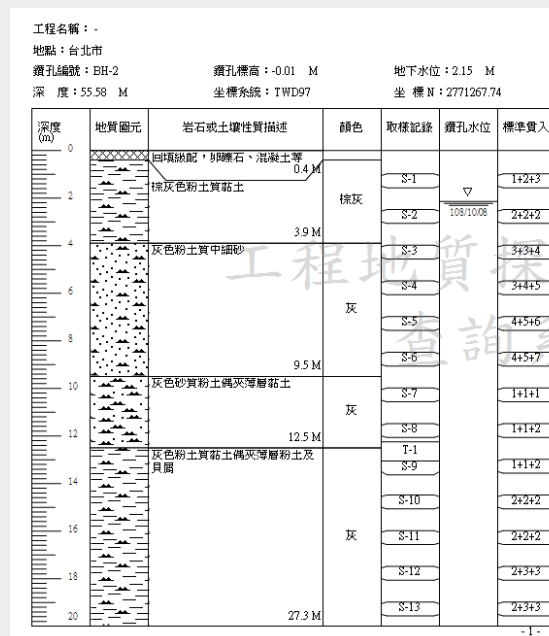


# Cross-section borehole data



## Difficulties:

- Not all borehole data matches lab test data
- Boreholes aren't uniformly distributed
- Made ground
- Lenses and non-continuous layers



# Criteria for combining layers

## Soil layer is the same above and below:

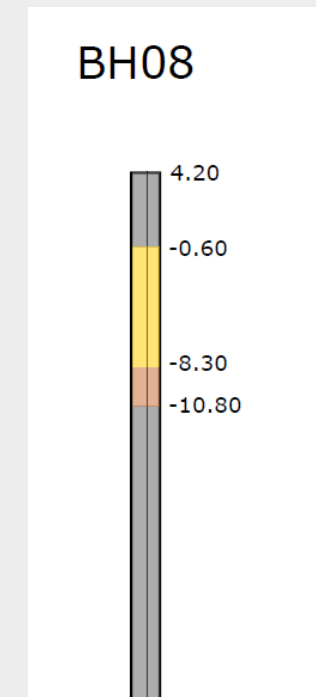
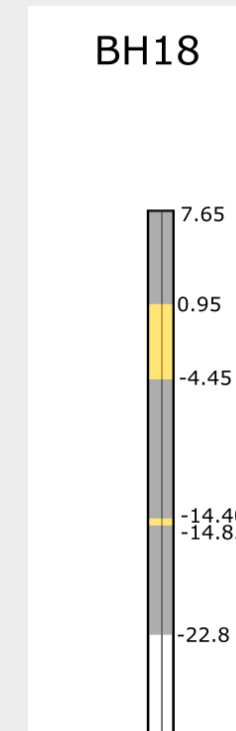
- <2m thick and the material is the same above and below, then combine with surrounding soil type.

## Soil layer is different above and below:

- Silt with plasticity – clay
- Silt without plasticity – sand

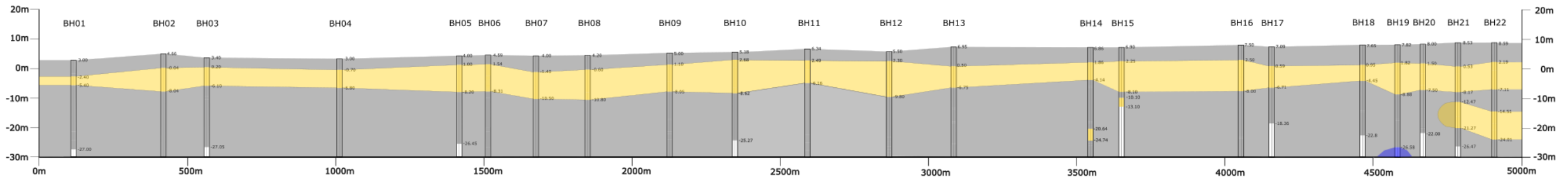
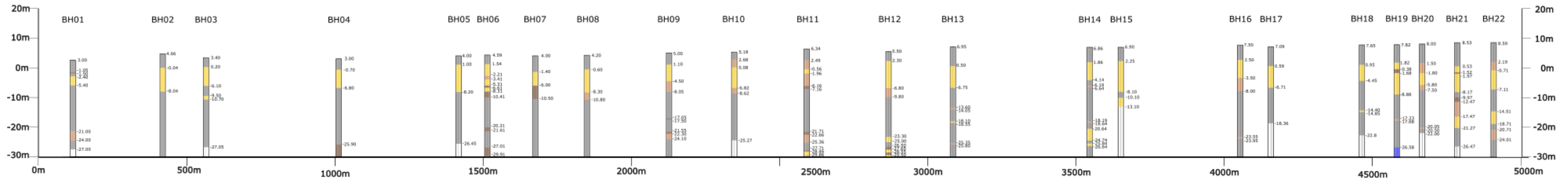
\*If the silt without plasticity has a higher percentage of clay than sand then change to clay. (BH1)

\*If the silt with plasticity has a very low plasticity index (<5%) change to sand. (BH7)



# Cross-section for modelling

## Legend



## Assumptions for modelling:

- Made ground area is clay
- Clay extends to 30m in shallower boreholes

# Simulation setup

Realisations: **1000**

Number of elements in x direction: **500**

Number of elements in y direction: **80**

Model length in x direction: **5000**

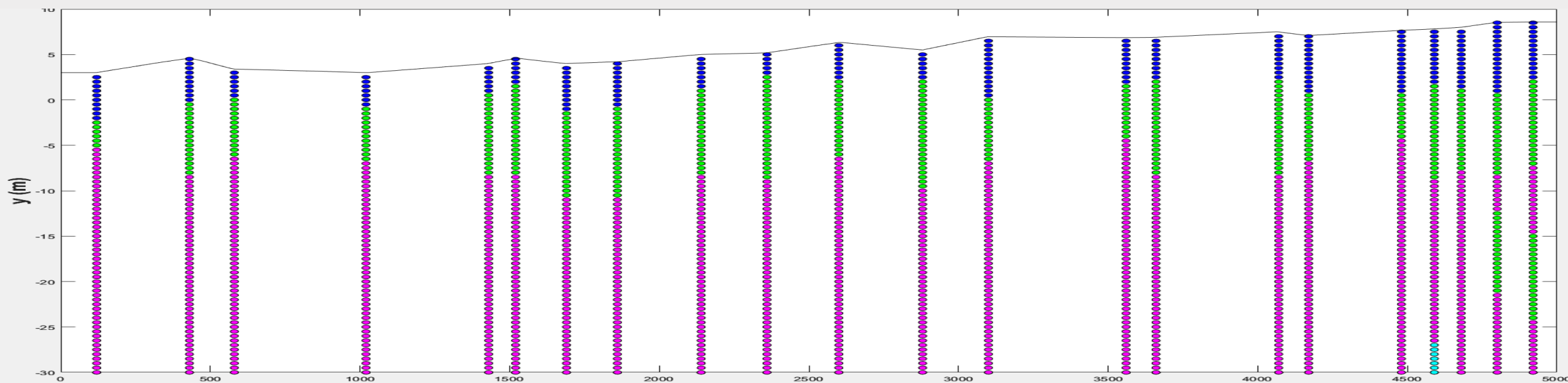
Model length in y direction: **40**



0.5m

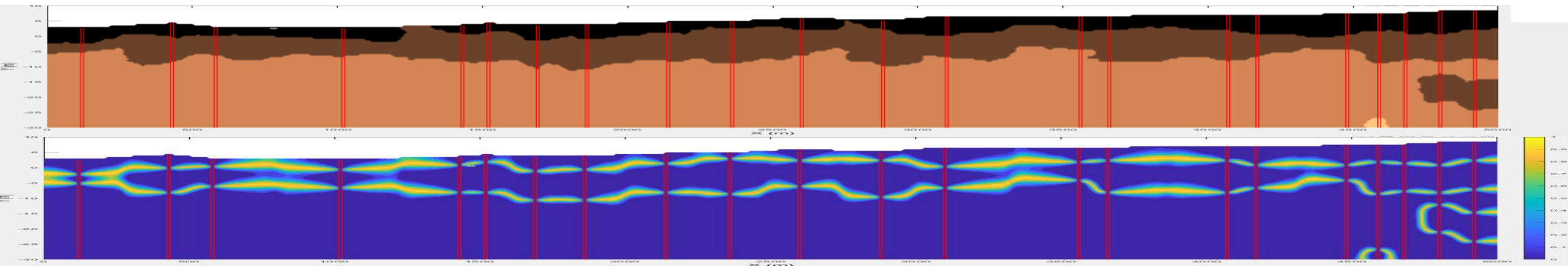
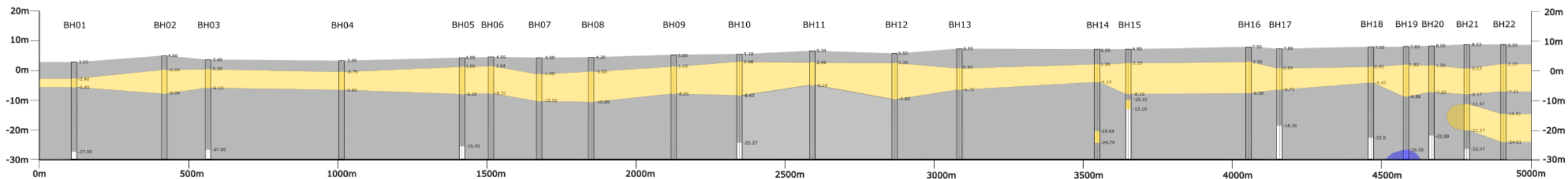


10m



Clay above and below sand layer is separated

# Simulation results using all boreholes



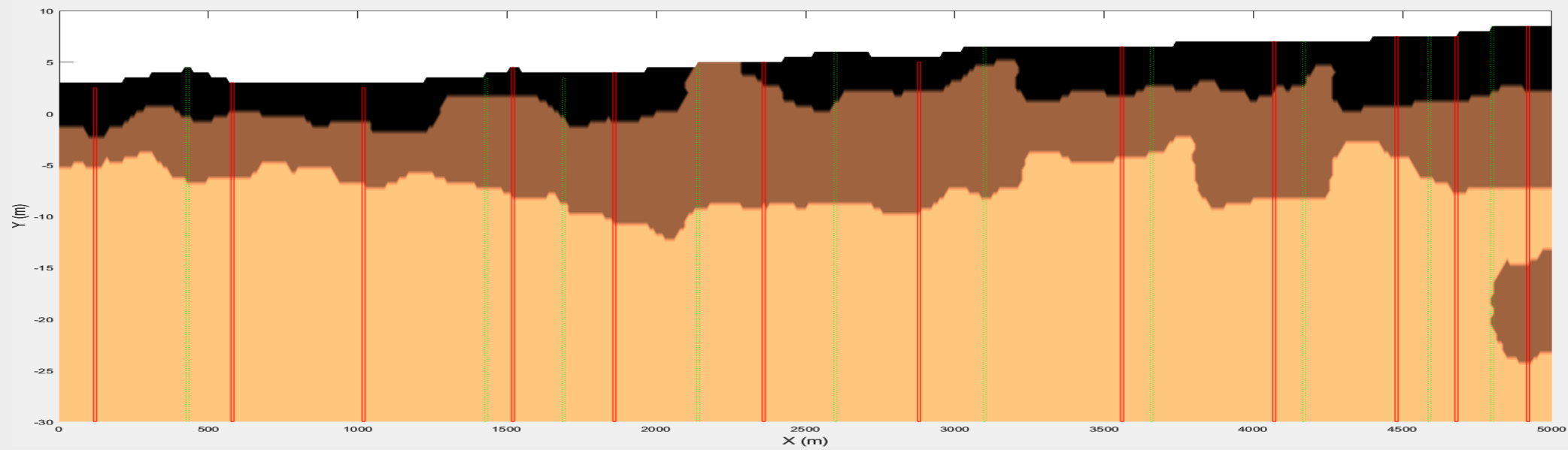
$A=6 \longrightarrow A'=120$

# Observational and conditional boreholes

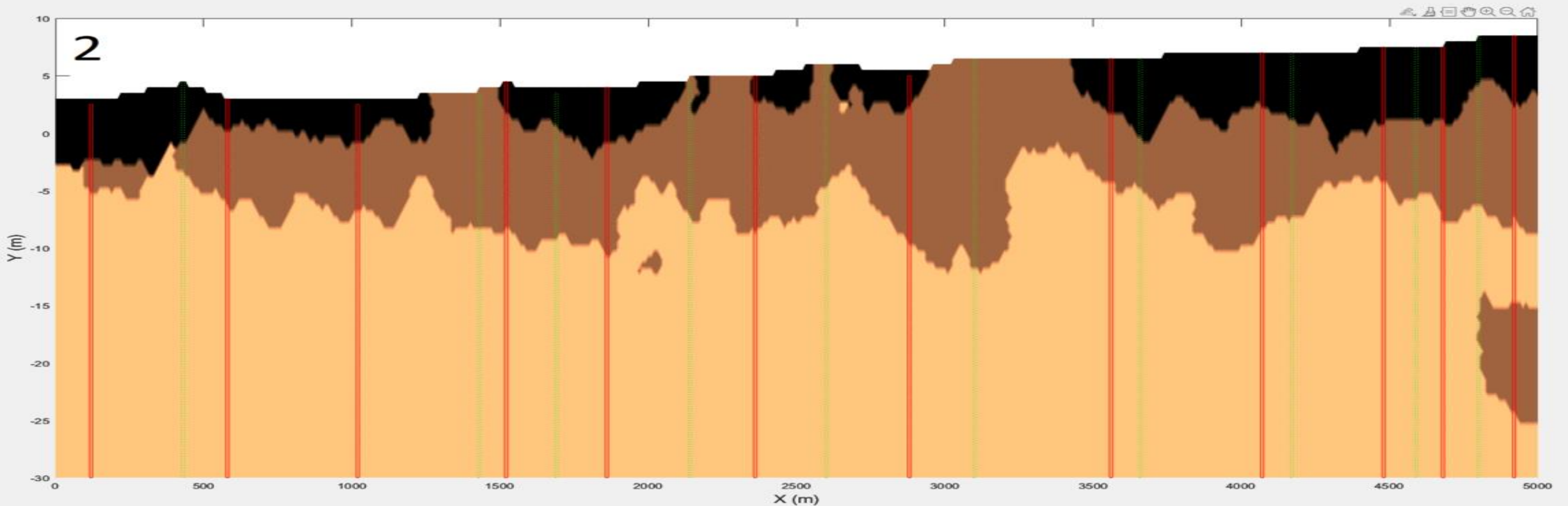
Red lines – conditional boreholes, used to make the model

Green lines – observational boreholes, used for comparison (not simulated)

12 conditional boreholes, 10 observational boreholes



# Spatial parameter calibration

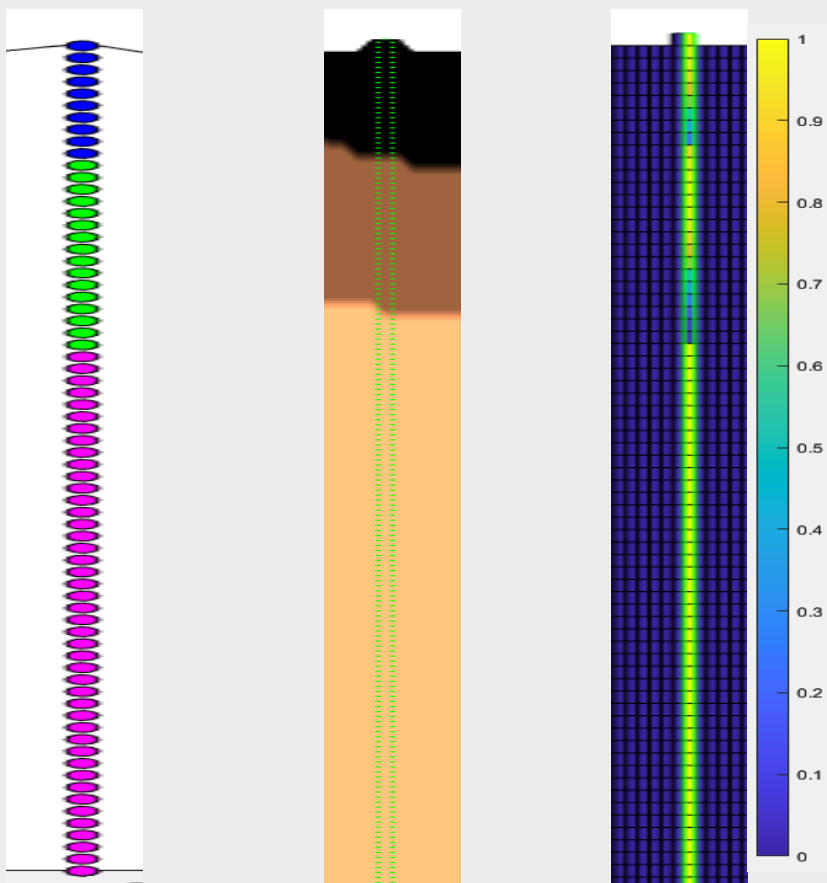


# Maximum likelihood estimation

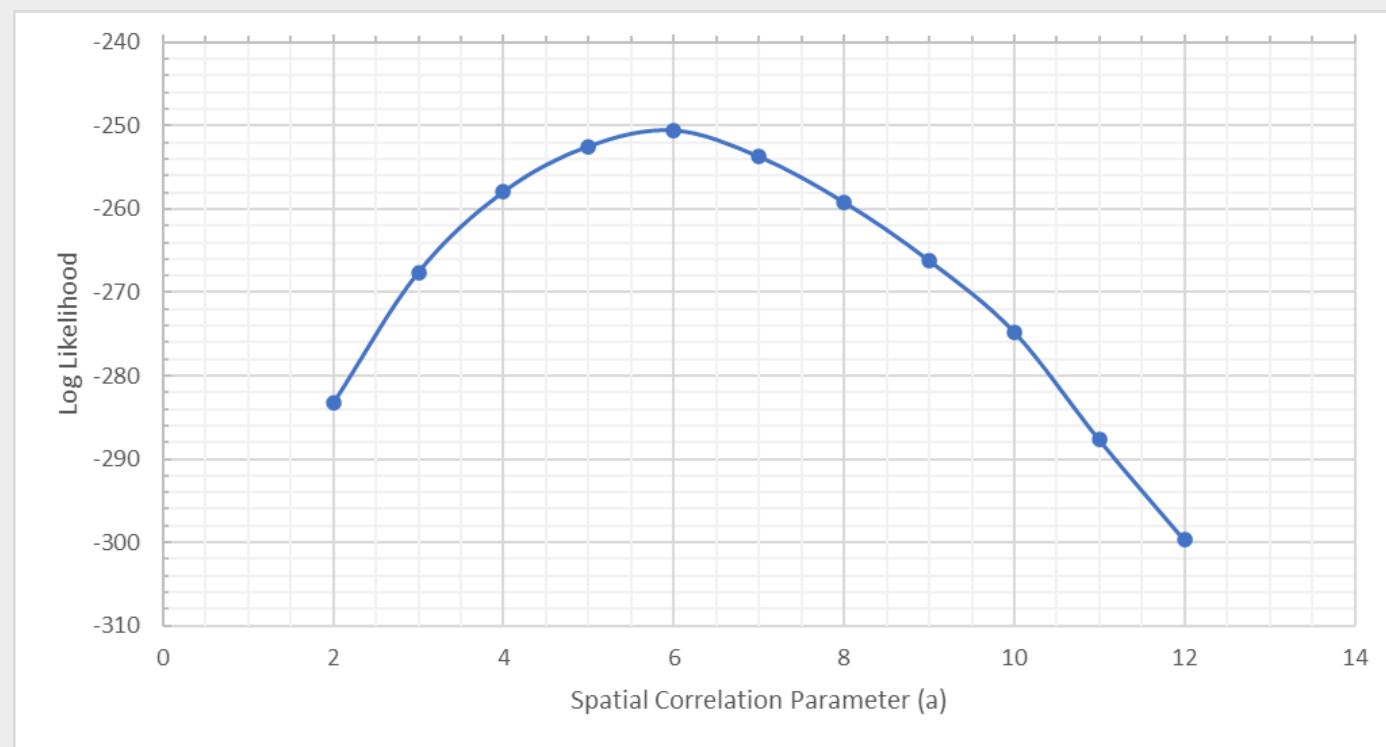
- Determines best model parameters that fit given data
- Lower likelihood = more uncertainty

Maximises log-likelihood function to estimate parameters

Borehole data – Simulated data – Likelihood

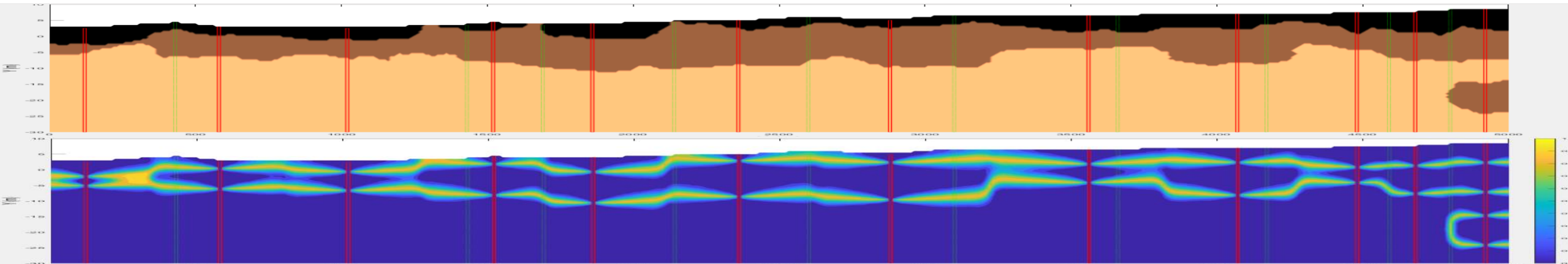
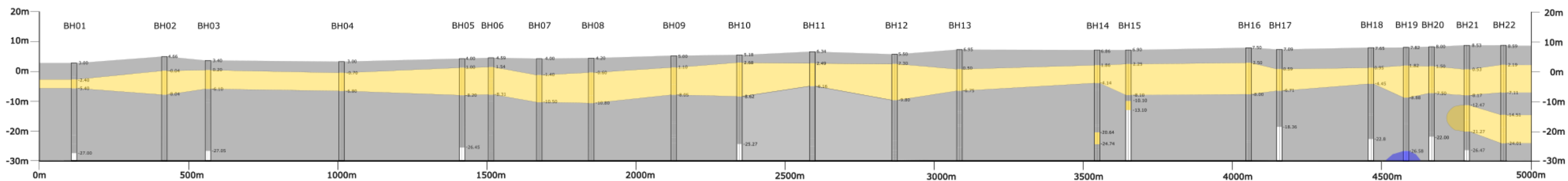


Graph of likelihood vs a value



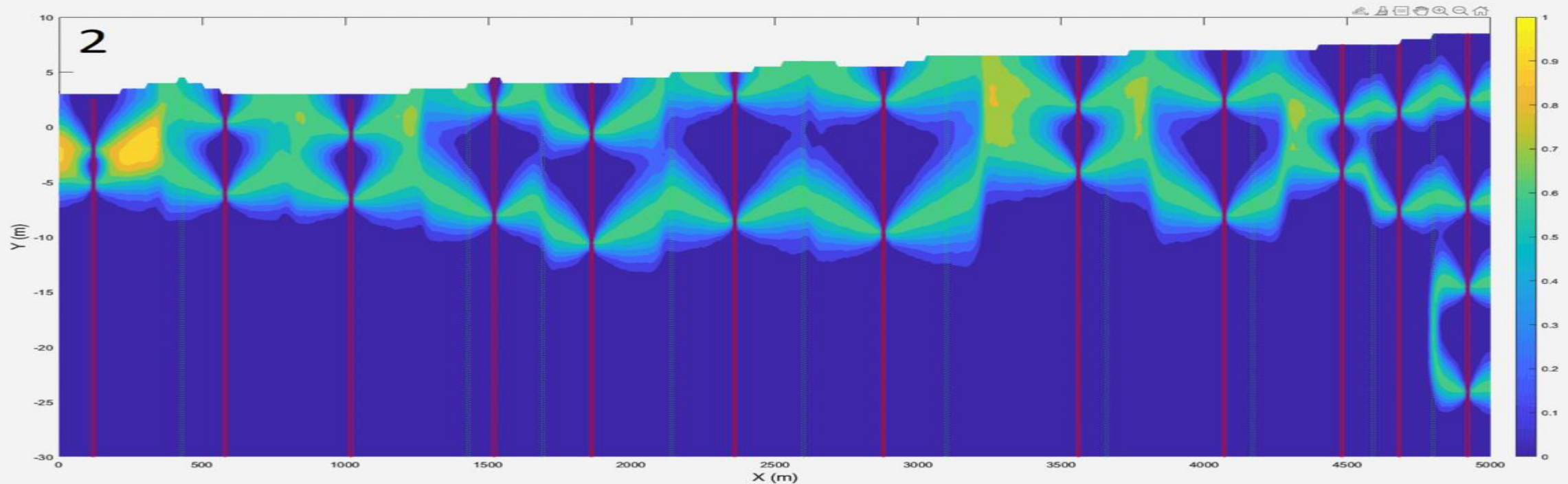


# Comparison between cross section and modelling result



# Information Entropy

- To quantify the uncertainty of this stochastic geological model, the concept of information entropy is adopted.
- The entropy is 0 when no uncertainty exists (E.g. only 1 possible lithological unit).
- Entropy is highest when all lithological units are equally probable.



# More things to consider?

- Number of realisations
- Scale of model
- Does the random field represent possible realisations
- Grid/element size
- Neighbourhood size
- Randomness of model
- Number and spacing of boreholes
- Different spatial correlation for different lithologies
- Computational cost

# Conclusion

- Preliminary results show that the best spatial correlation parameter ( $a$ ) is about 6 ( $a'=120$ ) for the north-south section chosen within the Taipei basin.
- Maximum Likelihood can be useful for finding the most optimal  $a$  value in stochastic geological models.
- However, there are many things that need to be considered for this type of modelling.
- Overall Markov random field modelling can be a useful tool to quantify uncertainty in geological models.

# Future Work

- Difference in spatial parameters for different lithologies
- Anisotropy of spatial correlation
- Correlating spatial parameter with deposition environment or different zones with the Taipei Basin
- 3D Modelling

Thanks for listening!

# Simulation Steps

## **Pre-process stage:**

- First the geological body is first discretized by a suitable mesh scheme.
- Followed by processing the geometric information of the meshed plot and constructing neighbourhood systems.
- Then, all known data are assigned to the corresponding elements.
- Finally, values for parameters  $\psi$  and  $a$  are selected.

## **Stochastic simulation stage:**

- The first step is aimed at creating an initial configuration by filling all the blank elements with lithological units.
- A practical way is sampling the geo-material type of elements which are the neighbours of those elements with known geo-material type, and then spread to the whole domain.
- Following this stochastic simulation process, a series of realisations are generated, which can be considered as samples of simulated subsurface lithological profile.

## **Post-process stage:**

- In the post-process stage, based on the simulated realisations, the uncertainty of the predicted lithological profile is assessed by the information entropy.