

**EAGE**

INCLUDING



Leading  
**GEOSCIENCES**  
in a New Era

**2022**

**EAGE**  
**ANNUAL**

**83<sup>RD</sup> CONFERENCE & EXHIBITION**

**MADRID | SPAIN**

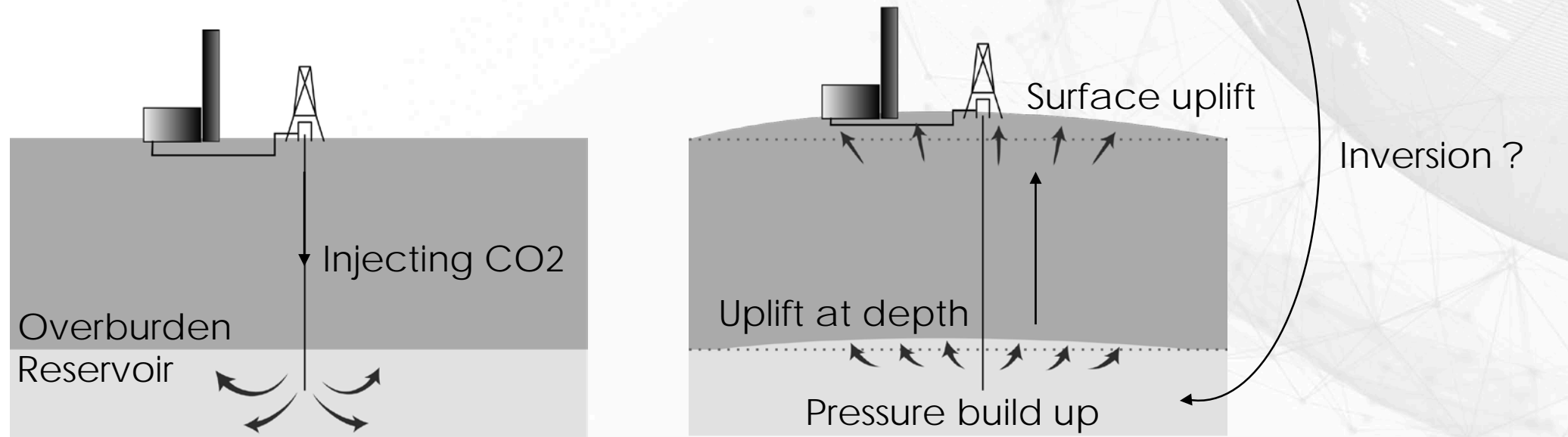
From surface deformation to pressure fields: contribution of machine learning to cost-effective CO2 injection monitoring

6-9 JUNE 2022 | [WWW.EAGEANNUAL2022.ORG](http://WWW.EAGEANNUAL2022.ORG)

# From surface deformation to pressure fields: contribution of machine learning to cost-effective CO2 injection monitoring

Jean-Remi Dujardin\*  
Joonsang Park  
Tore I. Bjørnarå  
Bahman Bohloli

# Introduction



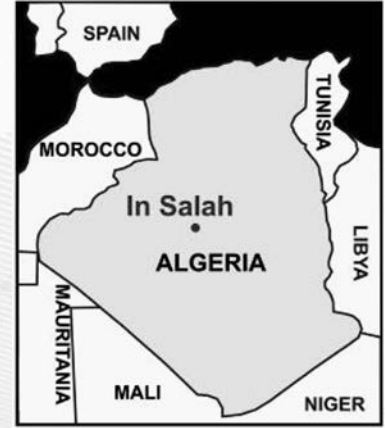
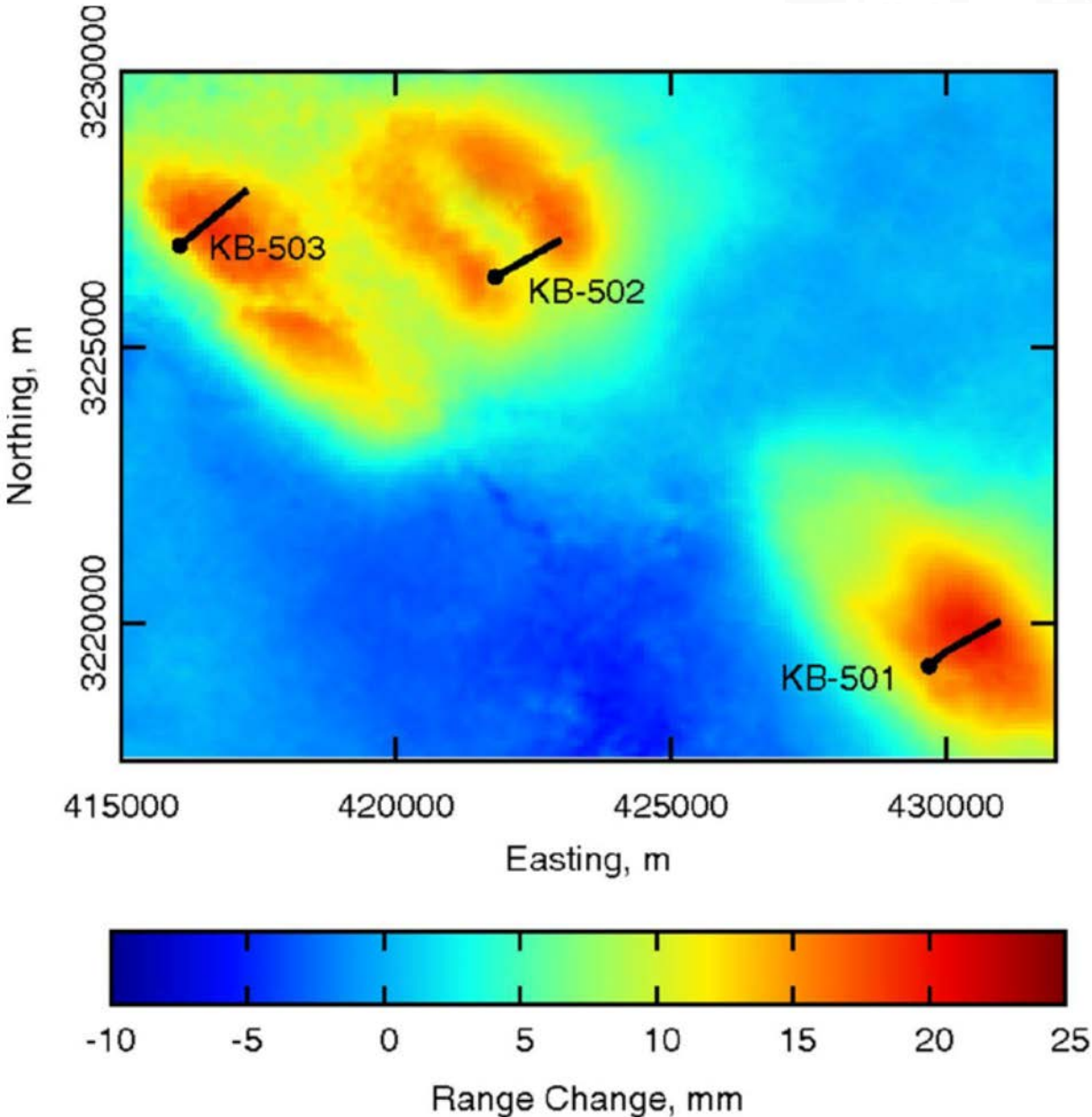
- Injecting CO2 → Pressure build up → Surface uplift
- Surface uplift → Estimation of pressure distribution ?



# Headlines

- Context on the In Salah injection site
- Measured InSAR data
- Synthetic dataset -> reference data
- Machine Learning ? Requirements, training set ...
- Results
- Conclusions

# In Salah injection site



- Production site for natural gas
- CO<sub>2</sub> injection between 2004 and 2011
- 3 injection wells
- More than 3.8 million tonnes of CO<sub>2</sub> stored
- Surface deformation used in this study measured in 2011 with InSAR data

Review of the injection history - Bohloli et al., 2018

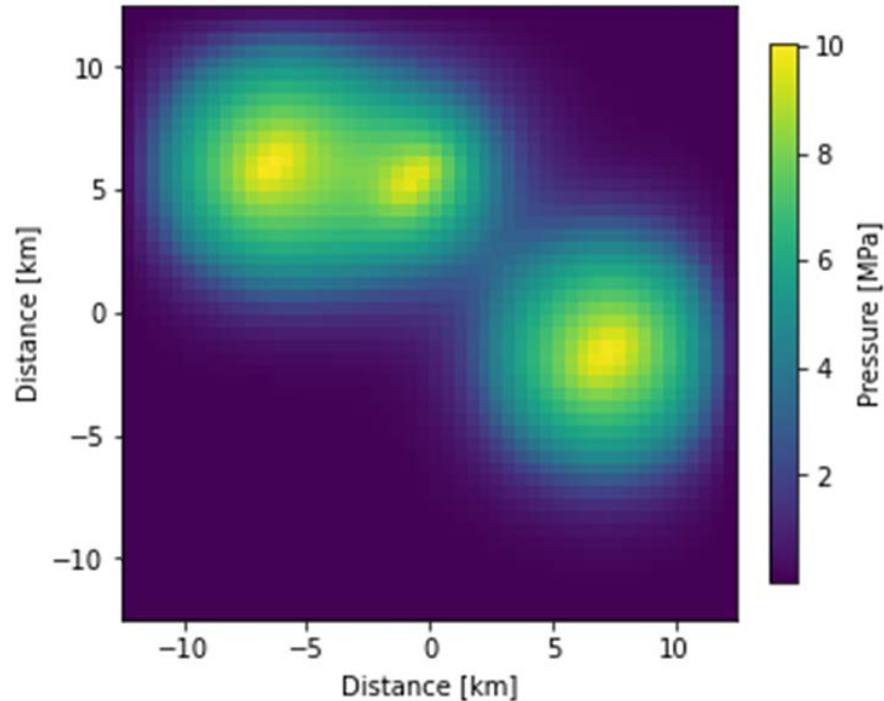
Surface uplift from InSAR data  
Figure from White et al. 2014

# The synthetic dataset

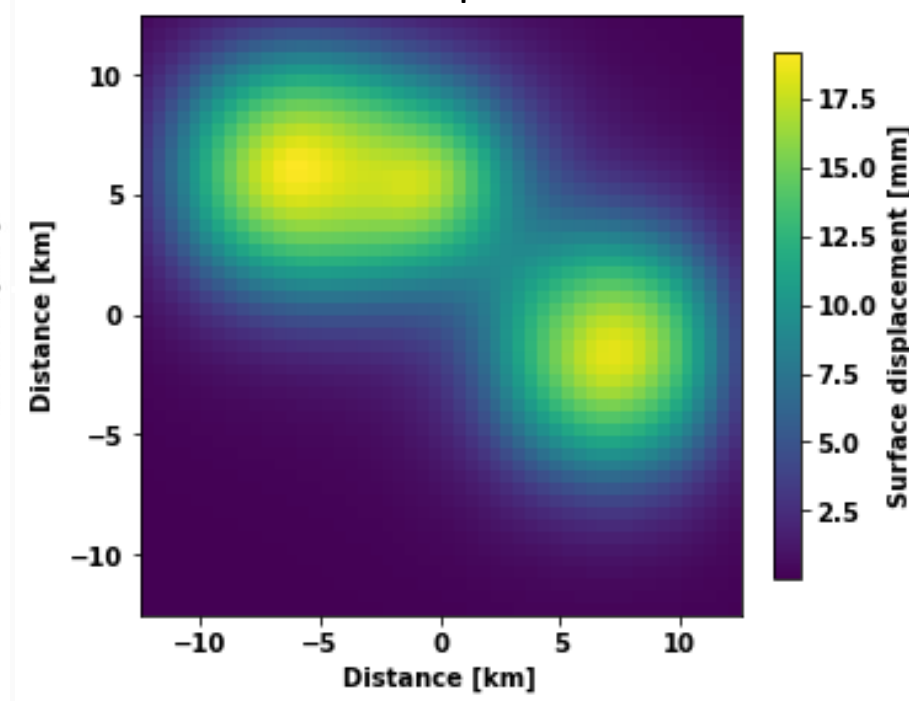
Layer	Thickness[m]	Young's modulus[GPa]	Poisson's ratio [-]	Remark
1	900	3	0.25	Shallow aquifer (Cretaceous)
2	750	5	0.30	Cap rock (Visean mudstone)
3	130	2	0.30	Lower cap rock
4	20	20	0.25	Tight sandstone
5	20	9	0.15	Reservoir
6	$\infty$	15	0.30	Devonian (underburden)

Layering and material properties for In Salah inspired synthetic model (after Bjørnarå et al., 2018)

Pressure distribution



Surface Displacement



Extension : 25km  
Bin size : 250 m

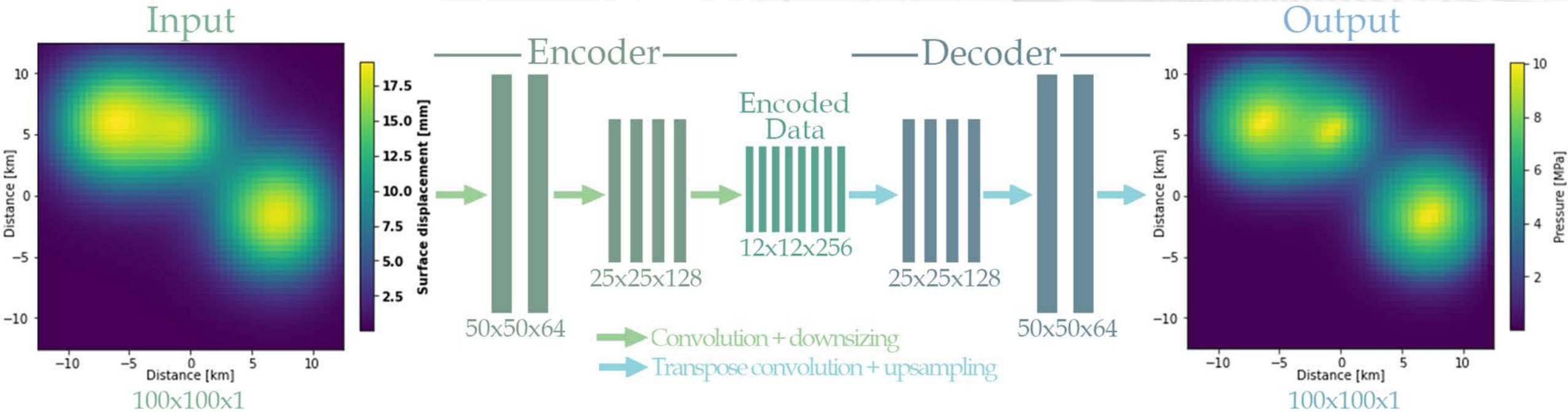
After Park et al., 2021  
Inspired after Bjørnarå et al., 2018



# Machine Learning requirements

- Validation set – previous slide
- Training set
- ML network/architecture

# Machine Learning network



- "Translate" an Image to an image -> encoder-decoder
- Images -> convolutional layers

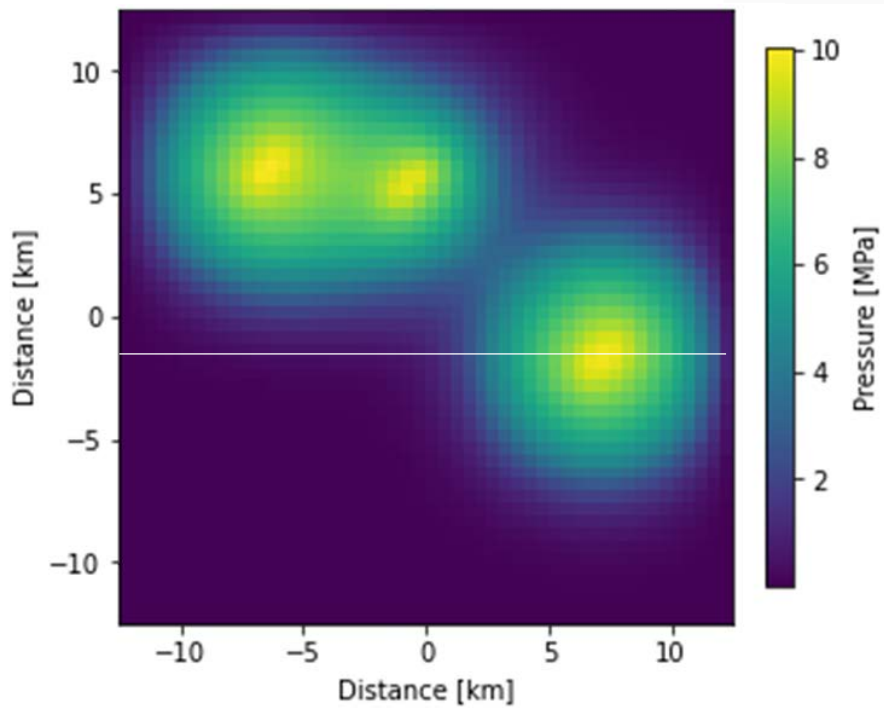
**Technical information**  
Activation functions : ReLU  
For last layer: Linear  
Loss function : mse



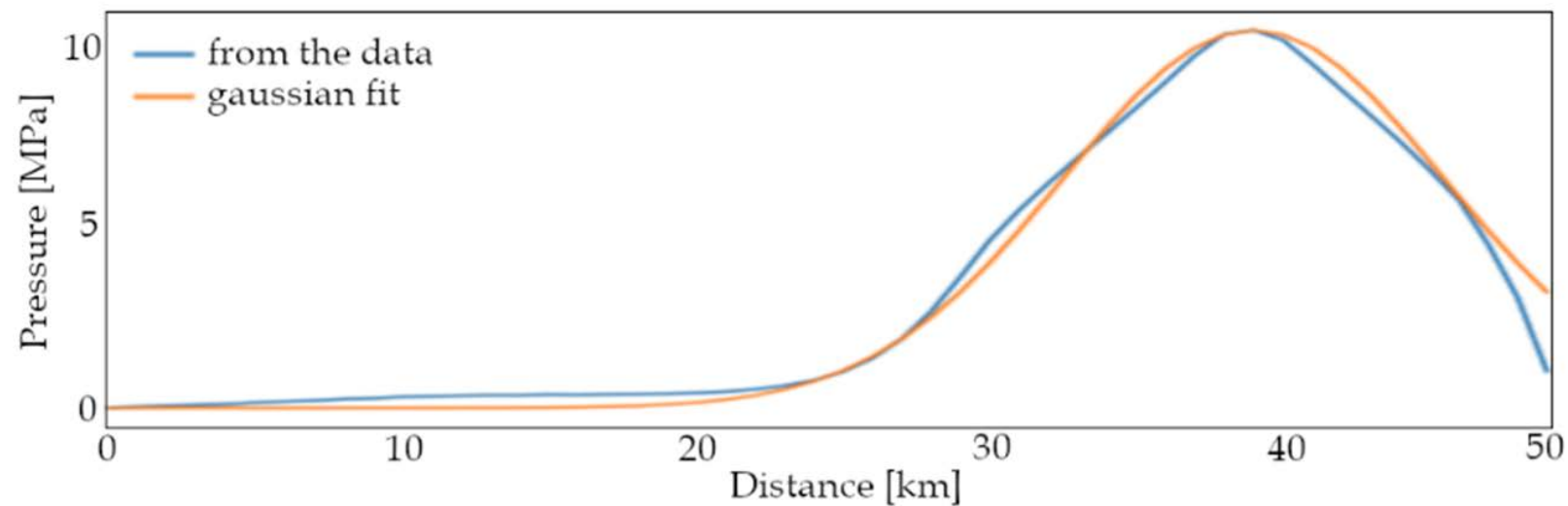
# Training set

- Pressure distribution
  - Randomness
- Corresponding surface displacement (forward modelling)
  - Generalized Geertsma solution – Park et al., 2021
  - Considered a tabular model – needs the thickness, Young's modulus and Poisson's ratio of each layer

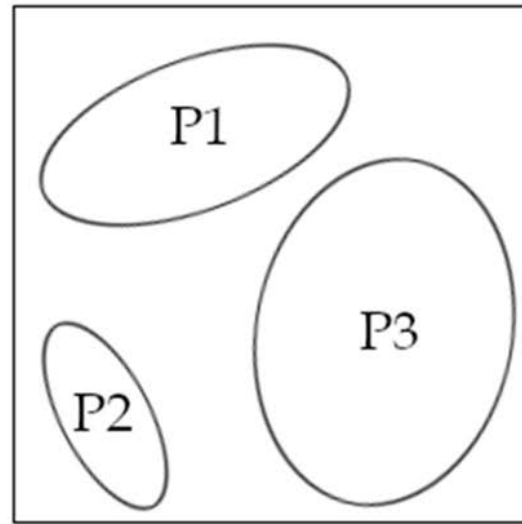
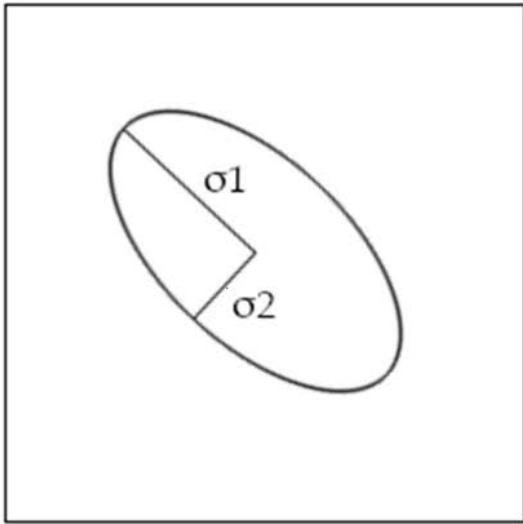
# Training set – pressure maps



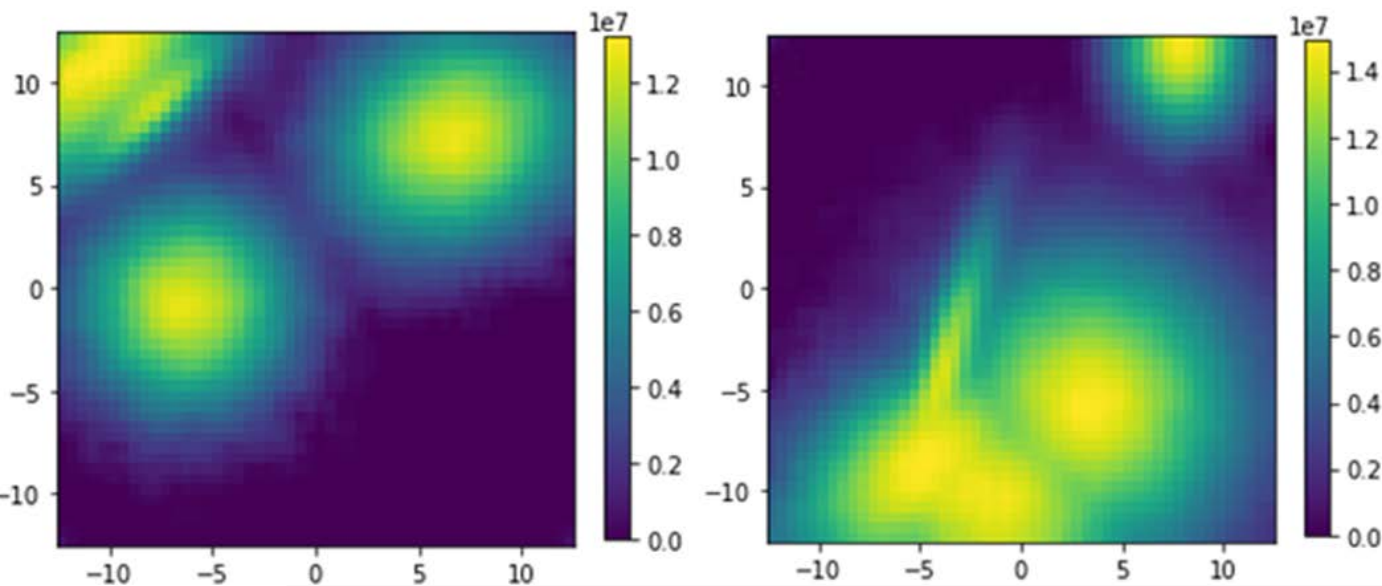
- Criterion – randomness and “realistic”



# Training set – pressure maps

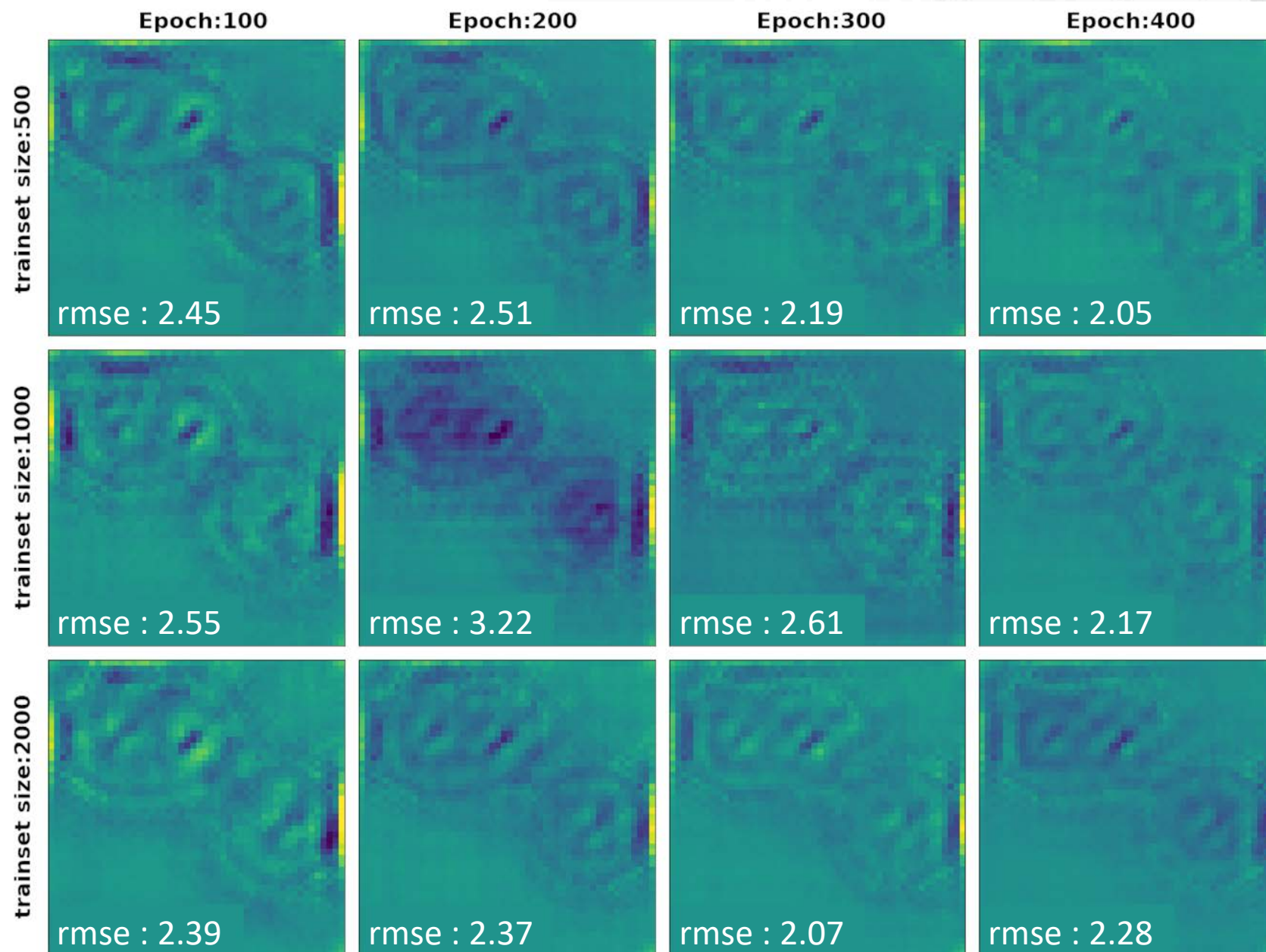


- 1 Patch defined by :
  - pressure patch width along 2 directions ( $\sigma_1$  and  $\sigma_2$ )
  - center
  - rotation angle
- Multiple patches (2 to 5) creates a pressure distribution

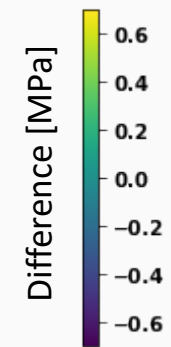




# Training the model and results



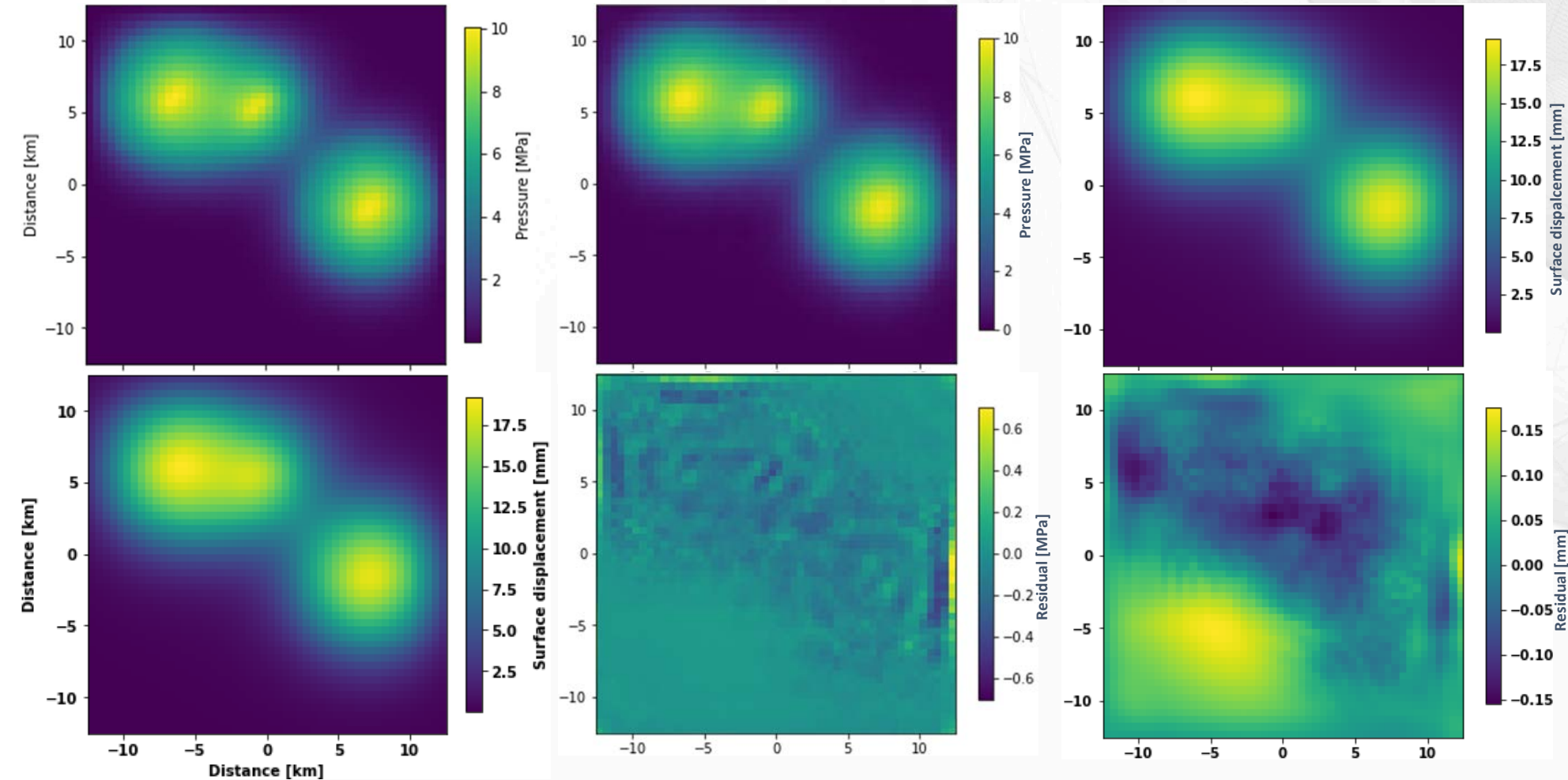
- Size of training set
- Epoch number





# Training the model and results

Epoch 300  
training size 2000



Proof

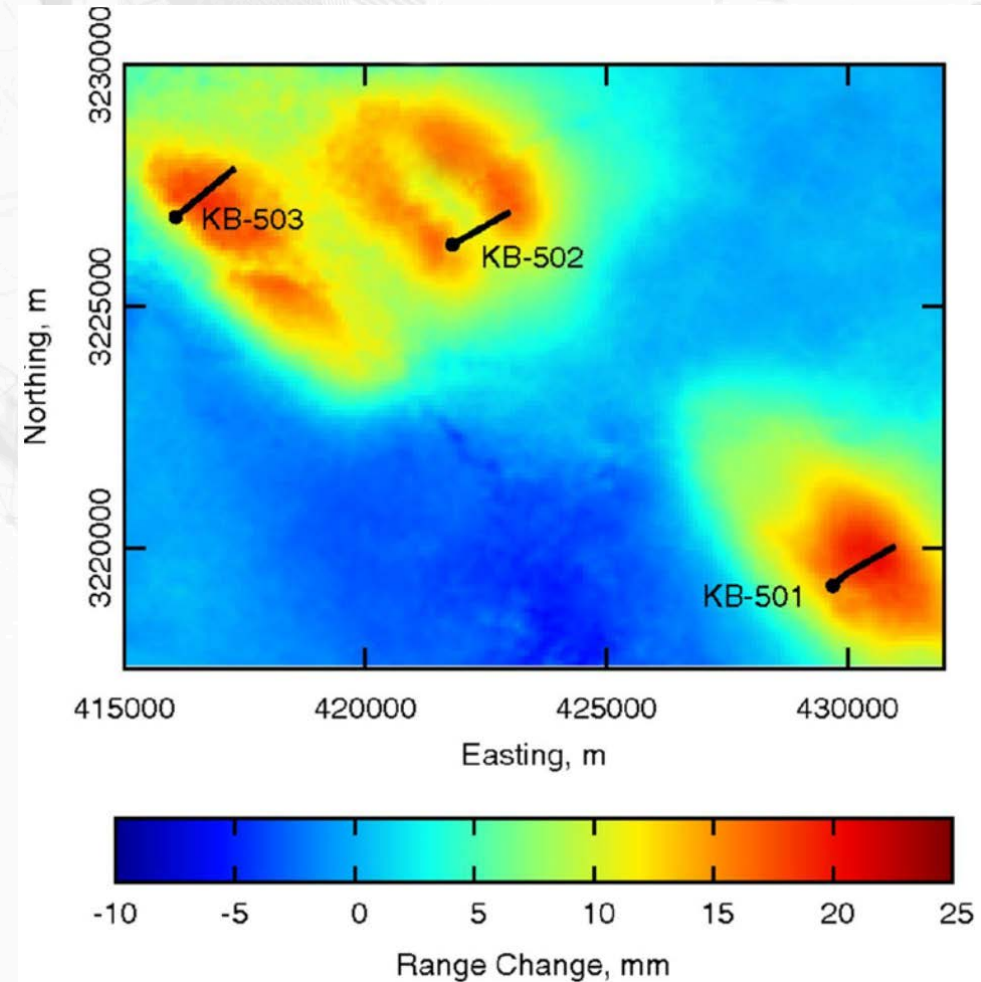
Predicted P

Surface uplift from prediction

# Conclusions

Pressure distribution throughout a reservoir can be obtained based on ground displacement at surface using Machine Learning, with some conditions:

- Simplified surface uplift measurements
  - Tabular geological model
  - Good geological model is required (Poisson's ratio, Young's modulus)
- How to do better ?
    - What accuracy do we want to obtain ?
    - Improving the generation of random pressure distribution
  - Application ?
    - Time lapse monitoring



# Thank you

## Aknowledgement

The authors thank the project partners of the In Salah Gas Joint Venture (BP, Statoil, and Sonatrach) for use of data made available as part of the In Salah Joint Industry Project (2006–2013).

The study is financially supported by SENSE (Assuring integrity of CO<sub>2</sub> storage sites through ground surface monitoring, Climit Demo Project No. 299664). SENSE has been subsidized through ACT (EC Project no. 691712) by Gassnova, Norway, United Kingdom Department for Business, Energy and Industrial Strategy, Forschungszentrum Jülich GMBH, Projektträger Jülich, Germany, The French Agency for the Environment and Energy Management, The United States Department of Energy, State Research Agency, Spain, with additional support from Equinor and Quad Geometrics.