



**EAGE**

EUROPEAN  
ASSOCIATION OF  
GEOLOGISTS &  
ENGINEERS

CONFERENCE & EXHIBITION



# NEAR SURFACE GEOSCIENCE '19

1<sup>st</sup> Conference on Geophysics for  
Geothermal-Energy Utilization  
and Renewable-Energy Storage

THE HAGUE, THE NETHERLANDS

**8-12 September 2019**

[WWW.NSG2019.ORG](http://WWW.NSG2019.ORG)

**#NSG2019**

# 10 years of temperature monitoring experiments using electrical resistivity tomography:

## What have we learned?

Thomas Hermans, Frédéric Nguyen, Guillaume De Schepper, Nolwenn Lesparre, & **Tanguy Robert\***

[Tanguy.Robert@uliege.be](mailto:Tanguy.Robert@uliege.be)

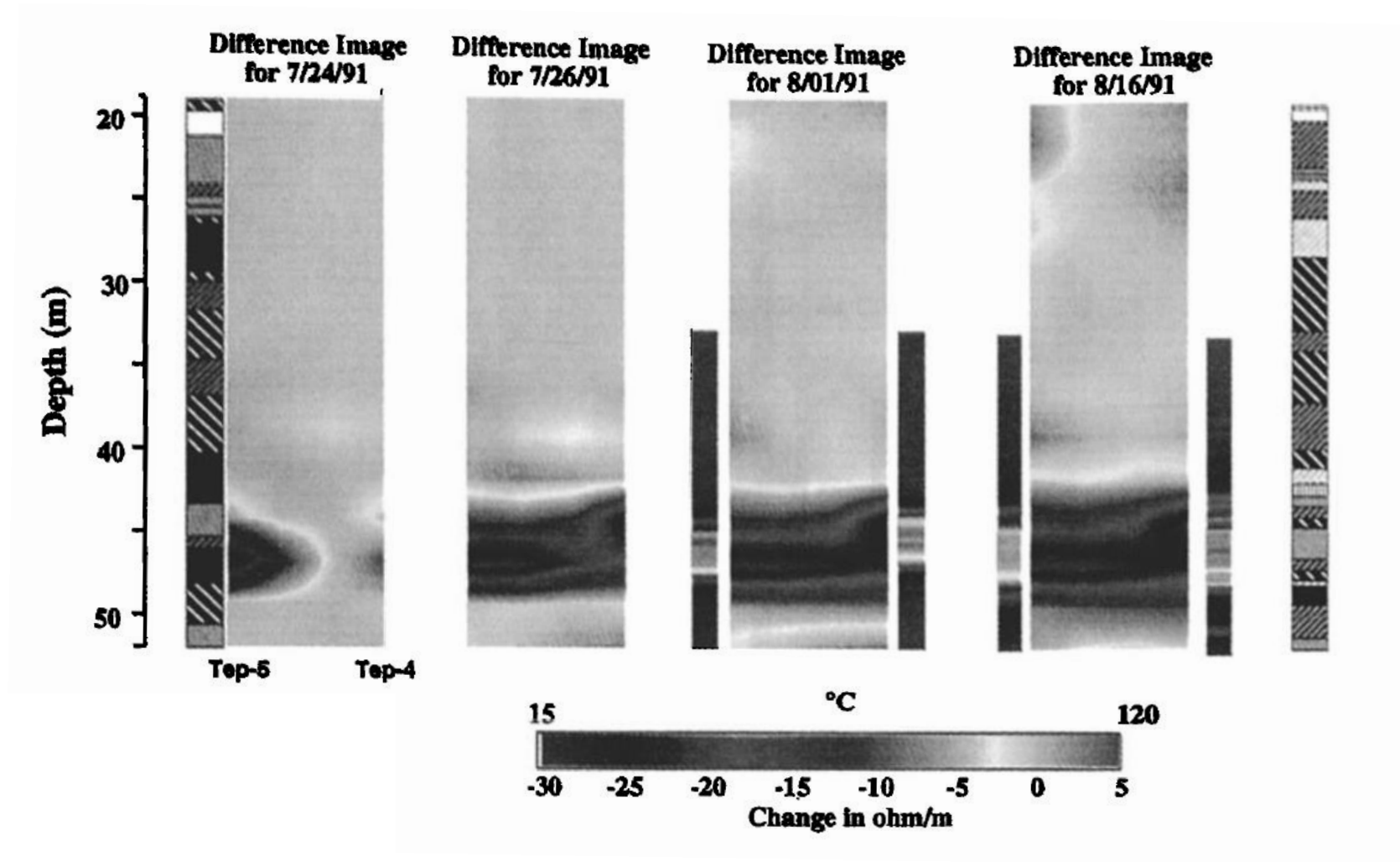
Paper 25, 11/09/2019

\*University of Liège & Fonds National de la Recherche (FNRS), Brussels, Belgium



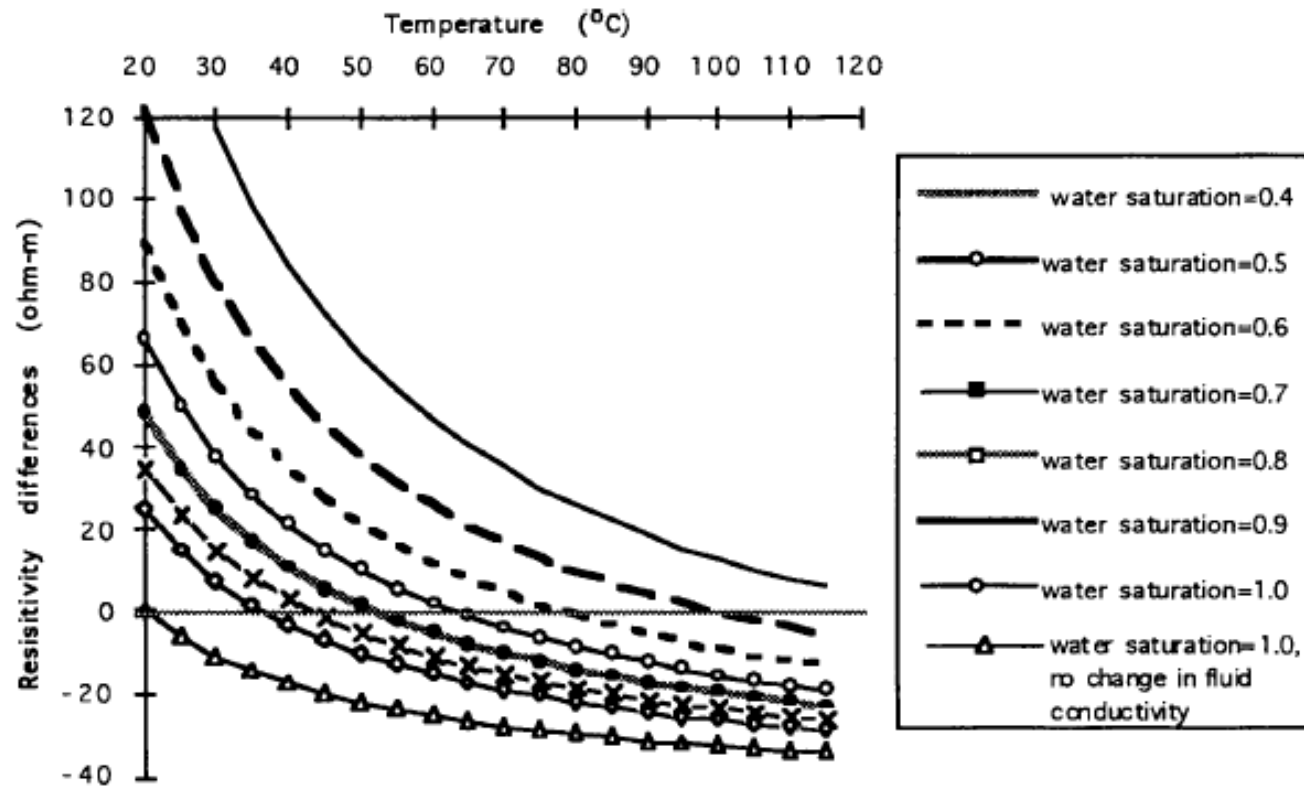
# Back to the future

Ramirez et al. 1993, WRR, 29(1), 73-87



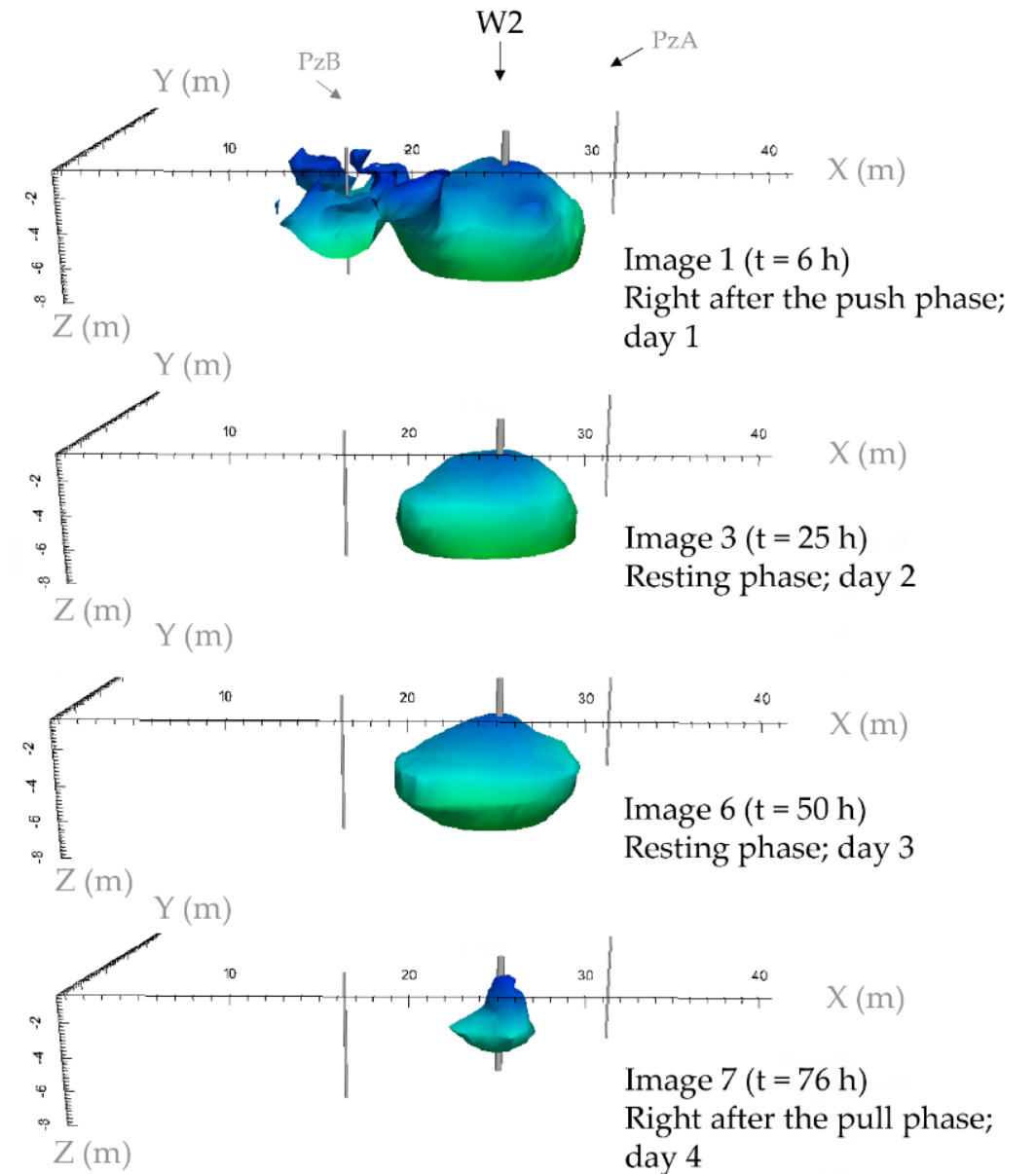
# Back to the future

Ramirez et al. 1995, JEEG, 0(1), 39-55



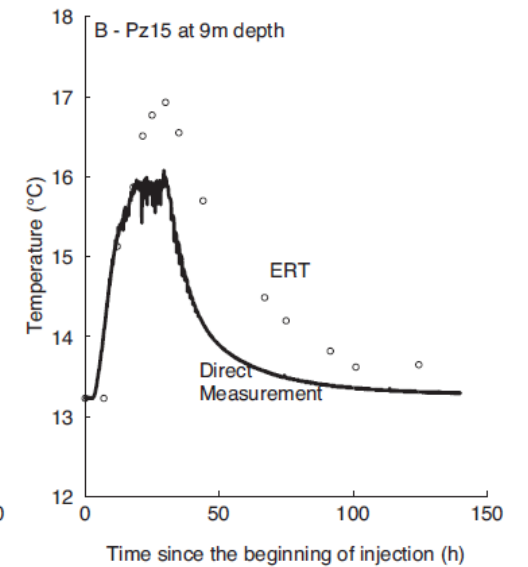
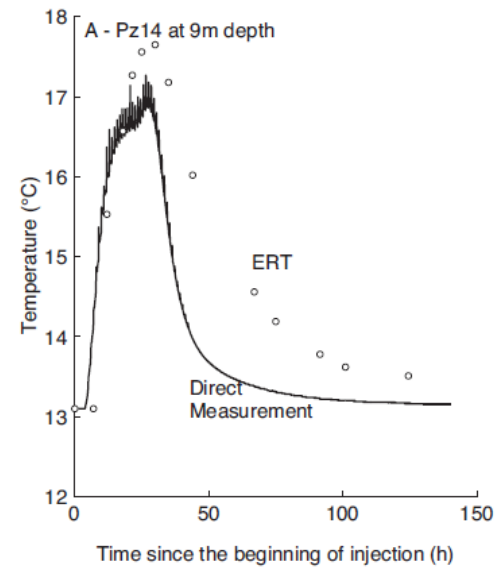
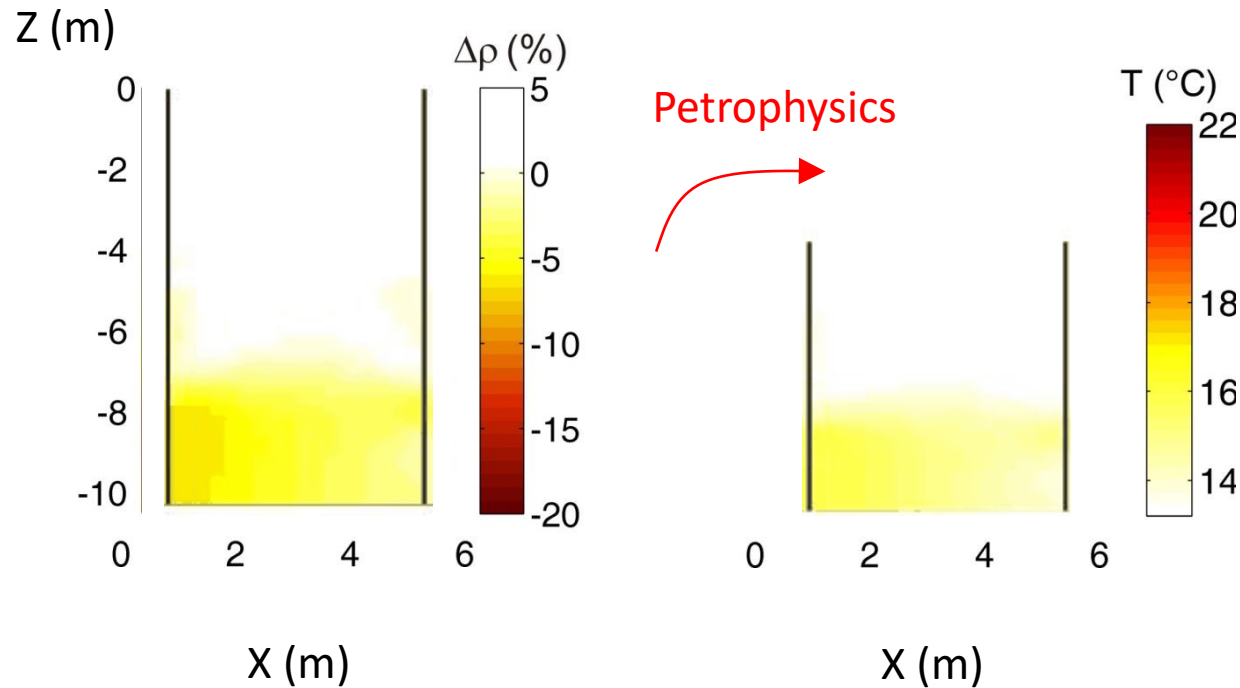
# Back to the future

Robert et al. 2019, Geosciences  
Special Issue  
Subsurface thermography ...



# Back to the future

Hermans et al. 2015, Geothermics, 53, 14-26



Petrophysics

Noise in the data

Survey design

Imaging

# Petrophysics

Noise in the data

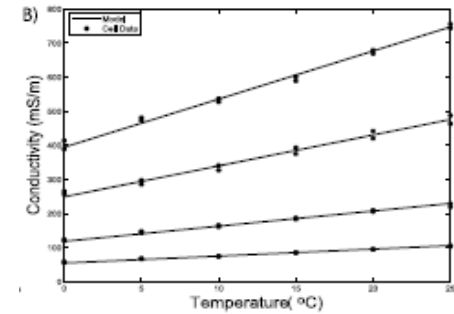
Survey design

Imaging



# Petrophysical laws exist

Hayley et al. 2007, GRL, 34, L18402

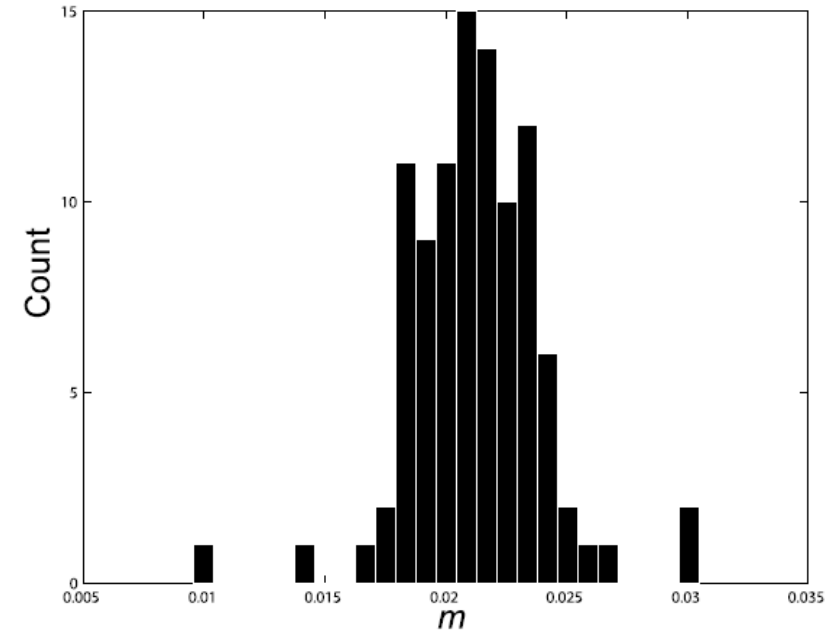


$$\sigma^f = \sigma^{f25} [m^f (T - 25) + 1]$$

With  $m^f \sim 0.02$



$\Delta T = +1$  K equals  $\Delta \sigma^f + 2\%$

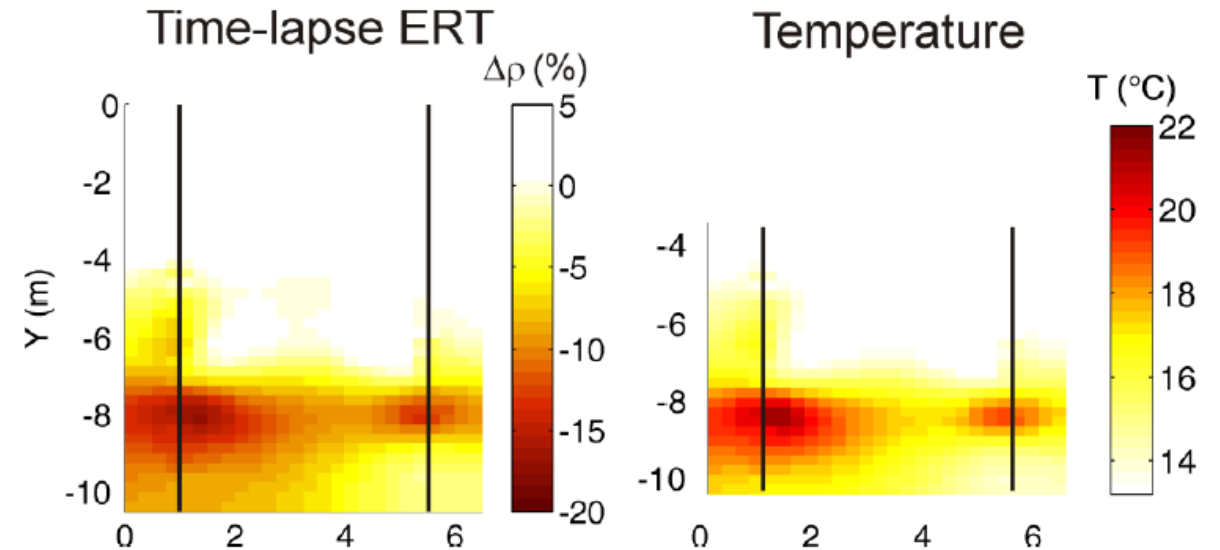
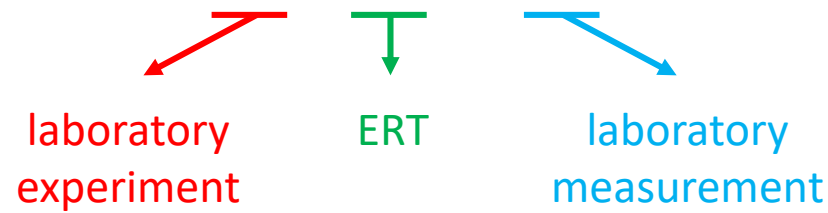


**Figure 2.** Histogram of extracted slopes from the analysis of the *Scott and Kay* [1988] study. This distribution has a mean of 0.021 and a standard deviation of 0.03.

# They are applied to retrieve T

Hermans et al. 2014, Energies, 7(8), 5083-5118

$$T_i = 25^\circ + \frac{1}{m_f} \left[ \frac{\sigma_{b;ti}}{\sigma_{b;t_0}} \frac{\sigma_{f;t_0}}{\sigma_{f;25^\circ}} - 1 \right]$$



# But strong assumptions remain

Robert et al. 2013, NSG2013, Tu S2a 10, Bochum, Germany

e.g. Hayley et al. 2007

$$\frac{\sigma_{fluid}^T}{\sigma_{fluid}^{T_{ref}}} = m_f (T - T_{ref}) + 1$$

1° absence of chemical reactions

Archie's law

$$\sigma_{bulk} = \frac{\sigma_{fluid}}{F}$$

2°  $\sigma_{surface}$  neglected

Ratio of Archie's law

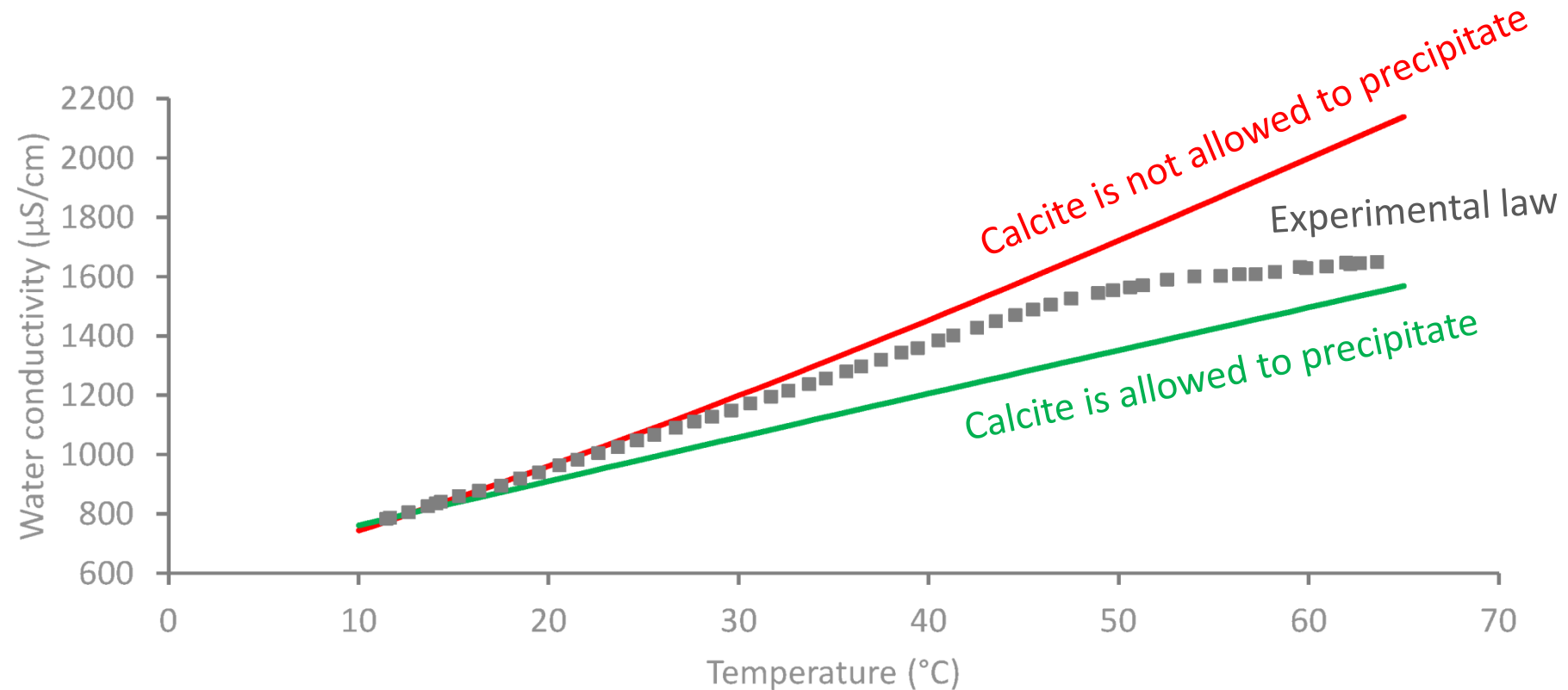
$$\frac{\sigma_{bulk;ti}}{\sigma_{bulk;t_0}} = F = \frac{\sigma_{fluid;t_0}}{\sigma_{fluid;ti}}$$

3° F remains constant

# But strong assumptions remain

Robert et al. 2013, NSG2013, Tu S2a 10, Bochum, Germany

1° absence of chemical reactions



Petrophysics

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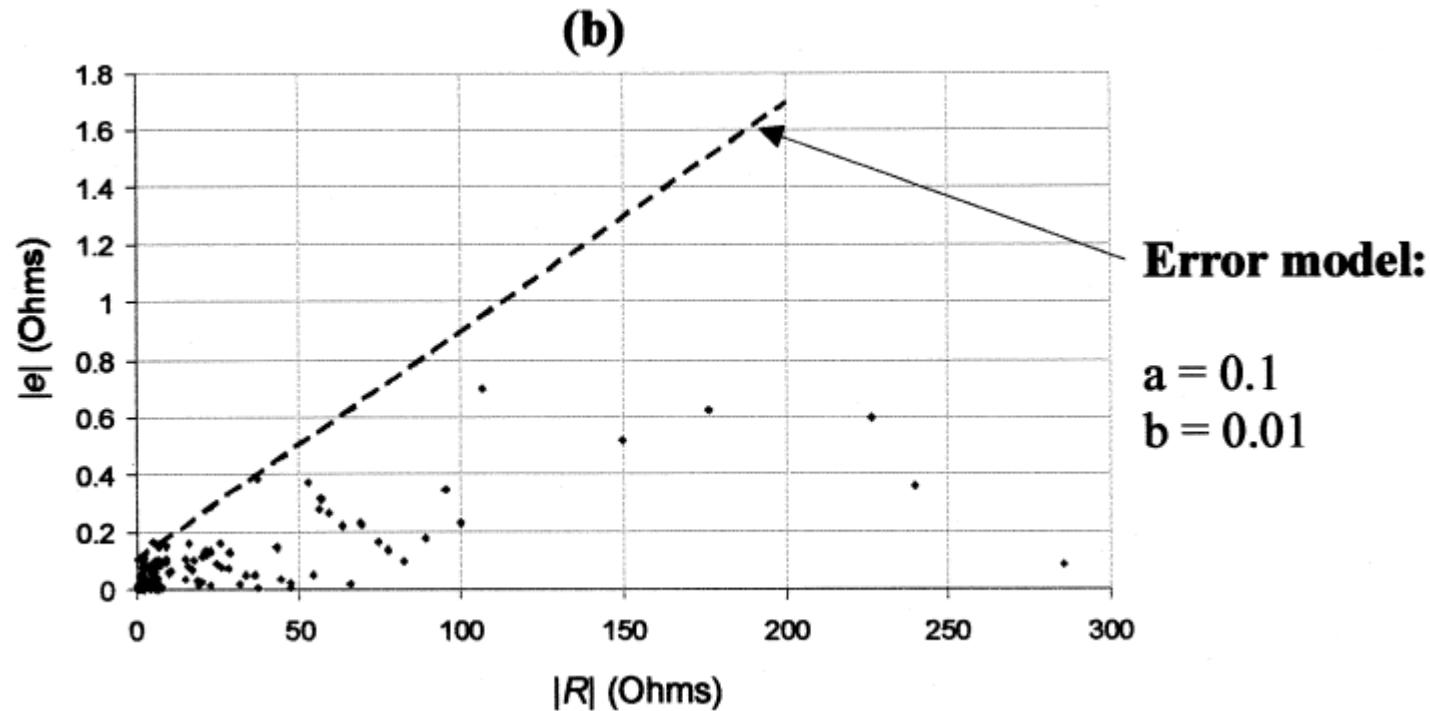
Imaging

# Authors use reciprocals to estimate noise

Slater et al. 2000, JAG, 44(2), 85-102

Static error model

$$|e| = a + b|R|$$

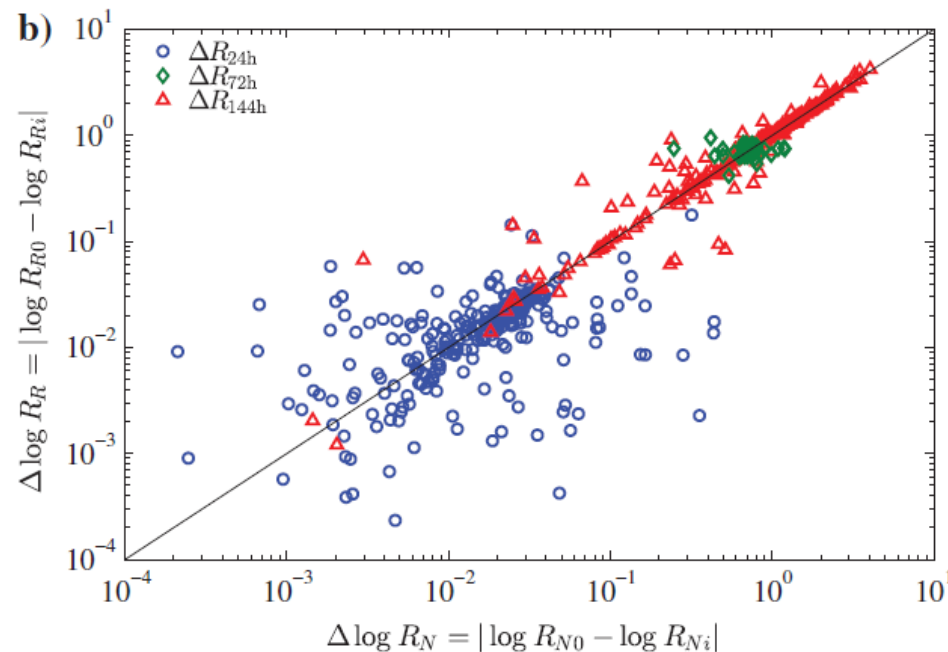


# Authors use reciprocals to estimate noise

Lesparre et al. 2017, Geophysics, 82(6), E325-E333

Time-lapse error model

$$|\Delta \log R_N - \Delta \log R_R| = \frac{a}{R} + b$$



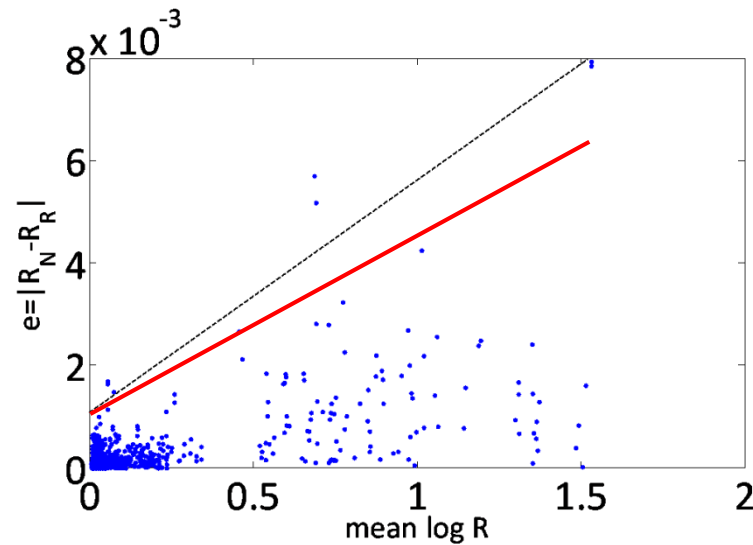
# Noise estimation is crucial

Robert 2012 (unpublished PhD thesis, Liège, Belgium)

$$e = a + bR$$

with  
 $a = 1 \text{ m}\Omega$  and  $b = 2\%$

what if  
 $a = 1 \text{ m}\Omega$  and  $b = 1.5\%$  ?

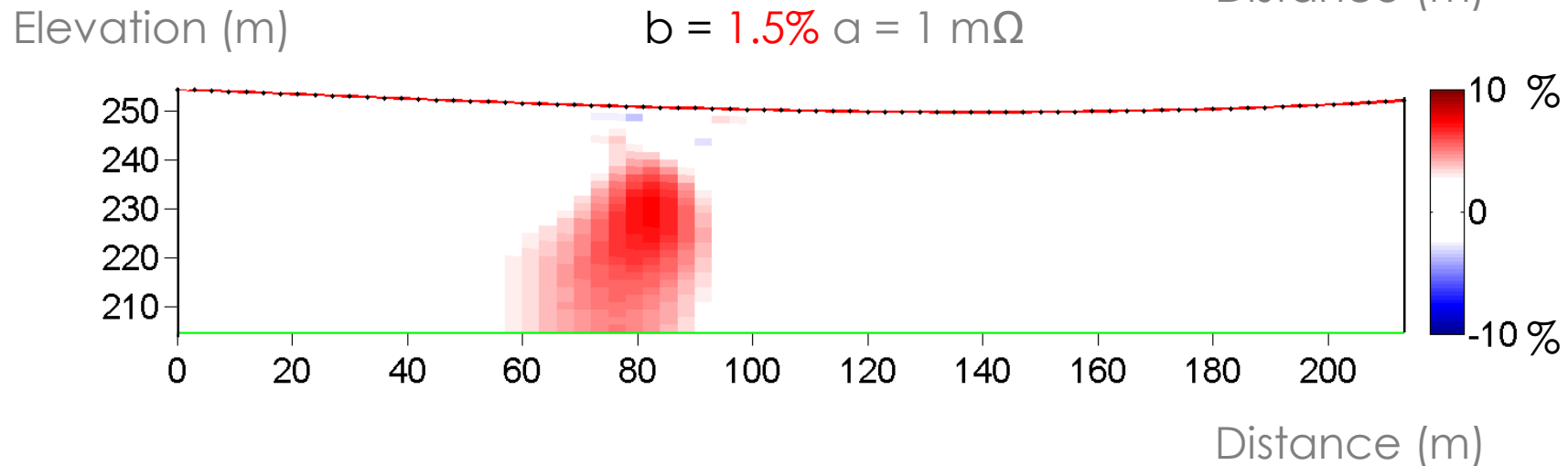
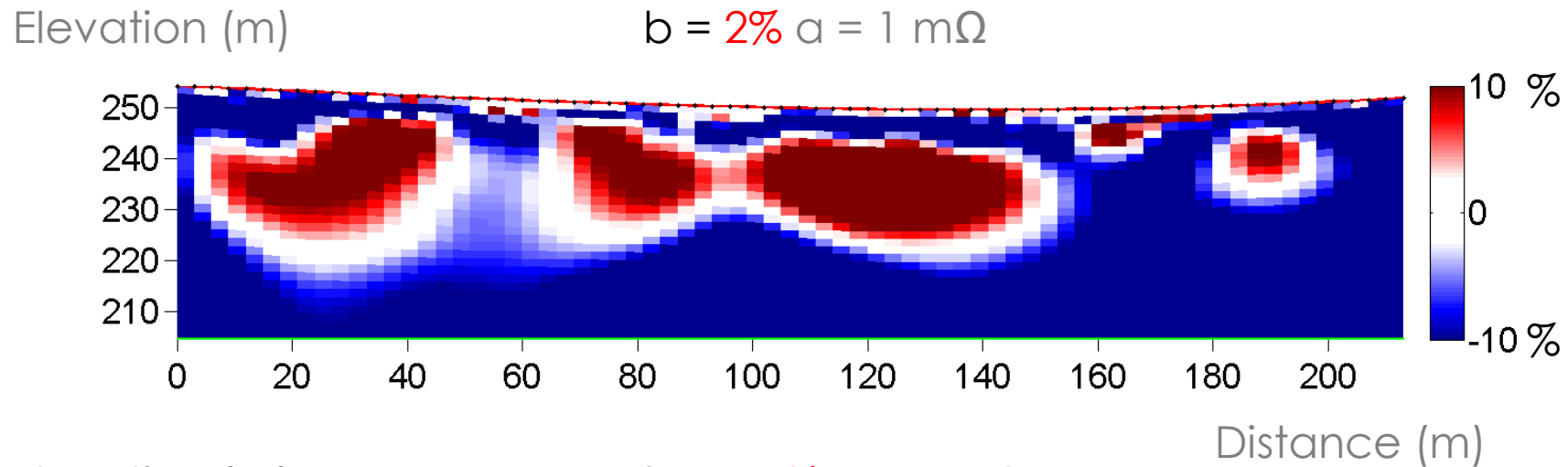




# Let the data speak...

Robert 2012 (unpublished PhD thesis, Liège, Belgium)

standalone  
inversions

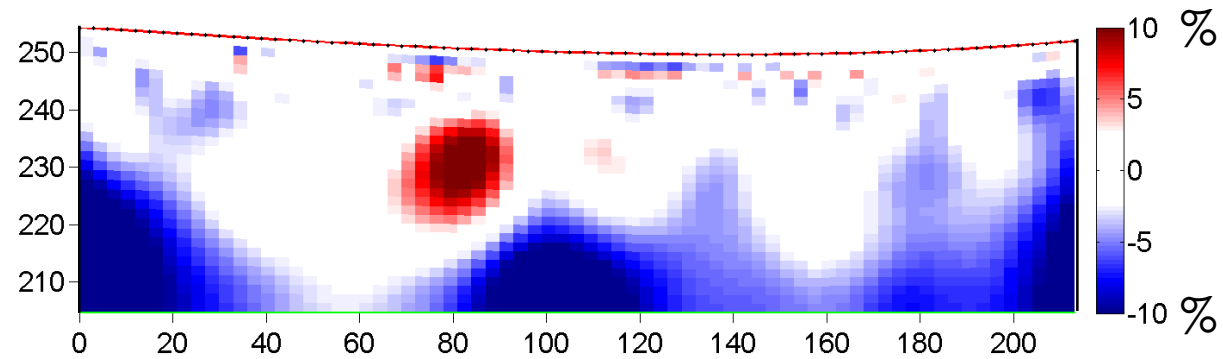


# ... but pay attention not to model noise

Robert 2012 (unpublished PhD thesis, Liège, Belgium)

Elevation (m)

higher noise assumption

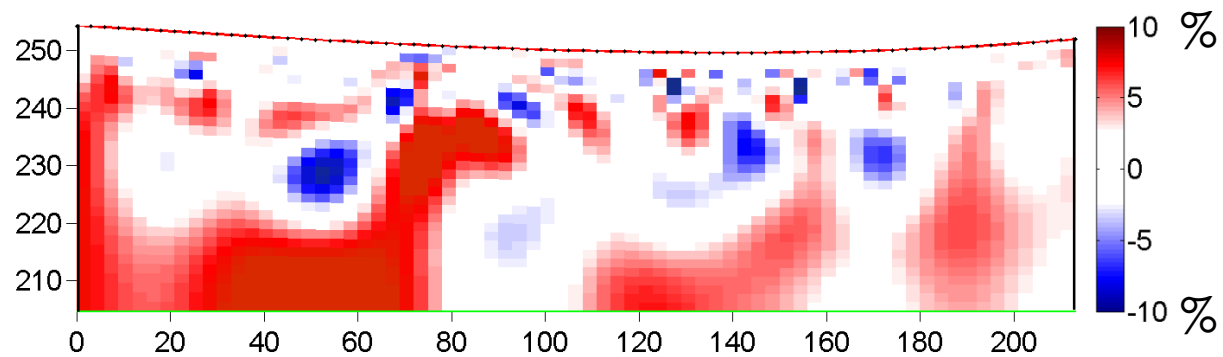


data difference  
inversions

Distance (m)

Elevation (m)

lower noise assumption



Distance (m)

Petrophysics

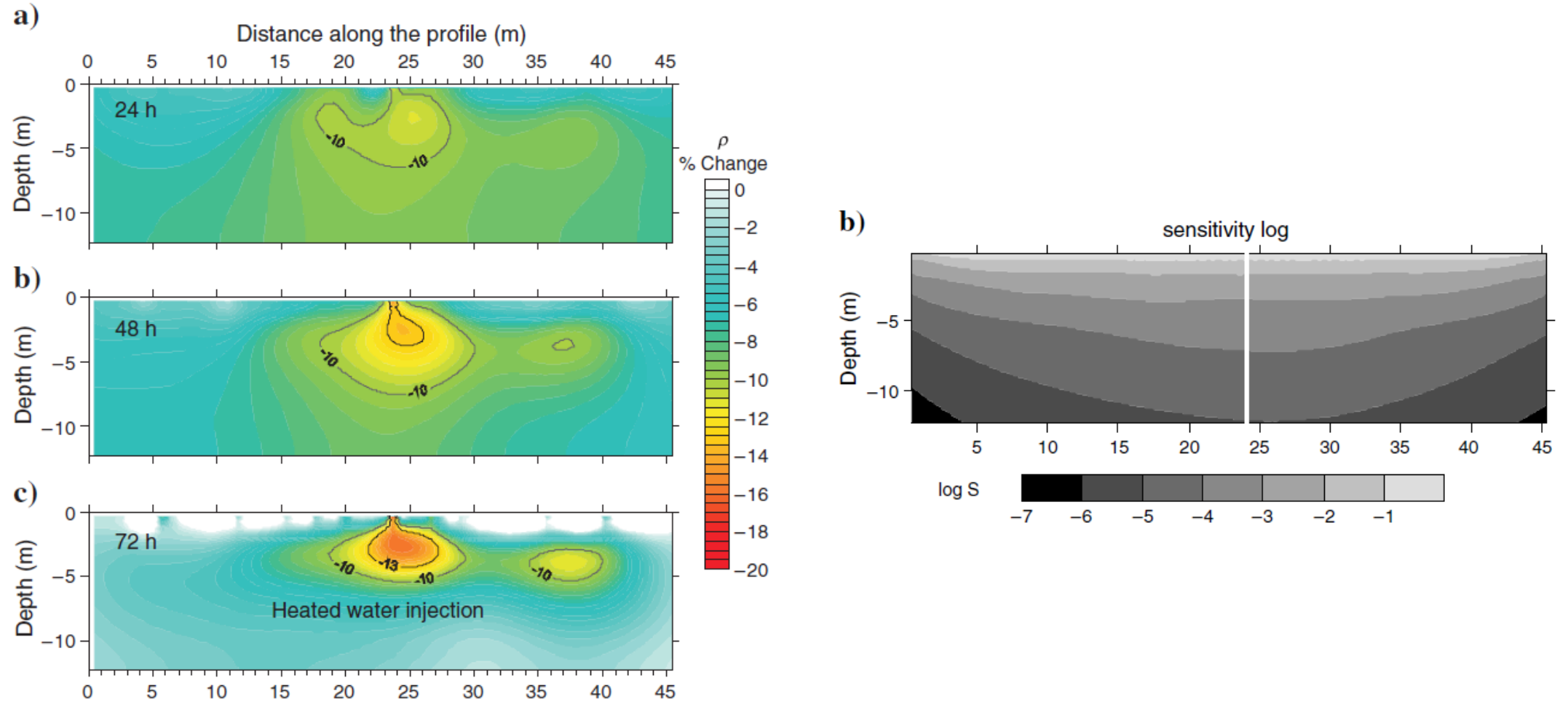
Noise in the data

Survey design

Imaging

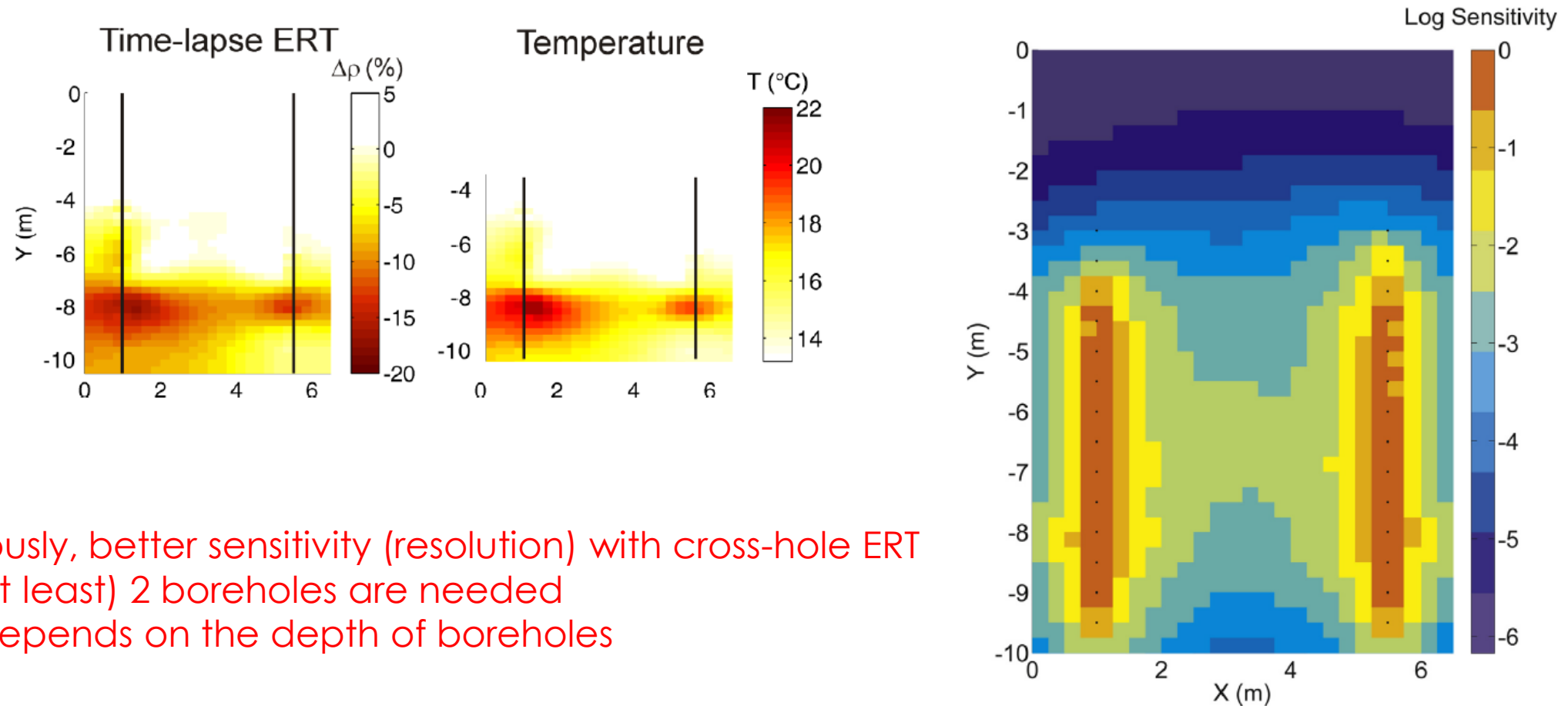
# Surface vs. cross-hole

Hermans et al. 2012, Geophysics, 77(1), B11-B21



# Surface vs. cross-hole

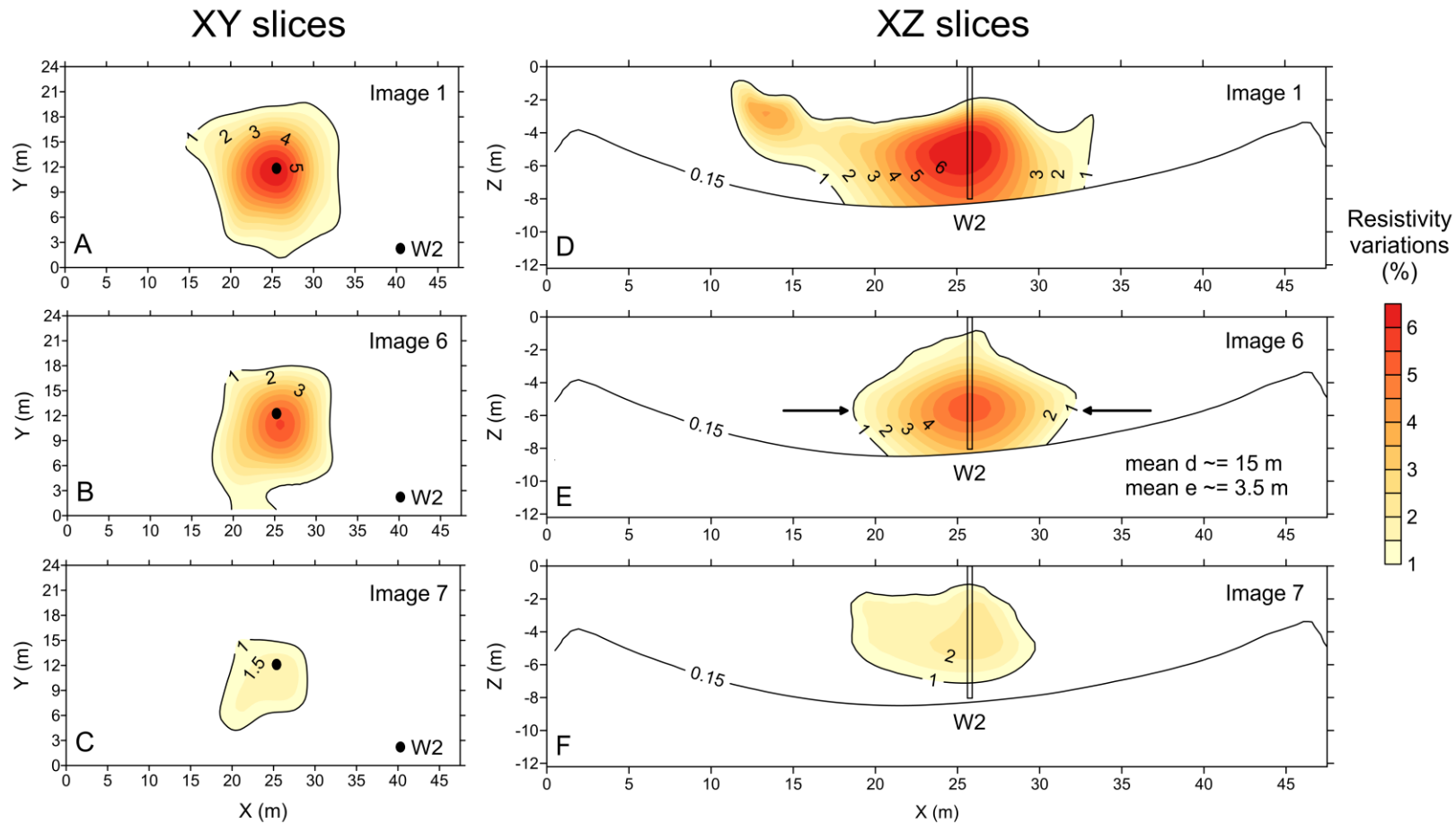
Hermans et al. 2015, Geothermics, 53, 14-26



Obviously, better sensitivity (resolution) with cross-hole ERT  
But (at least) 2 boreholes are needed  
DOI depends on the depth of boreholes

# 2D vs. 3D

Robert et al. 2019, Geosciences

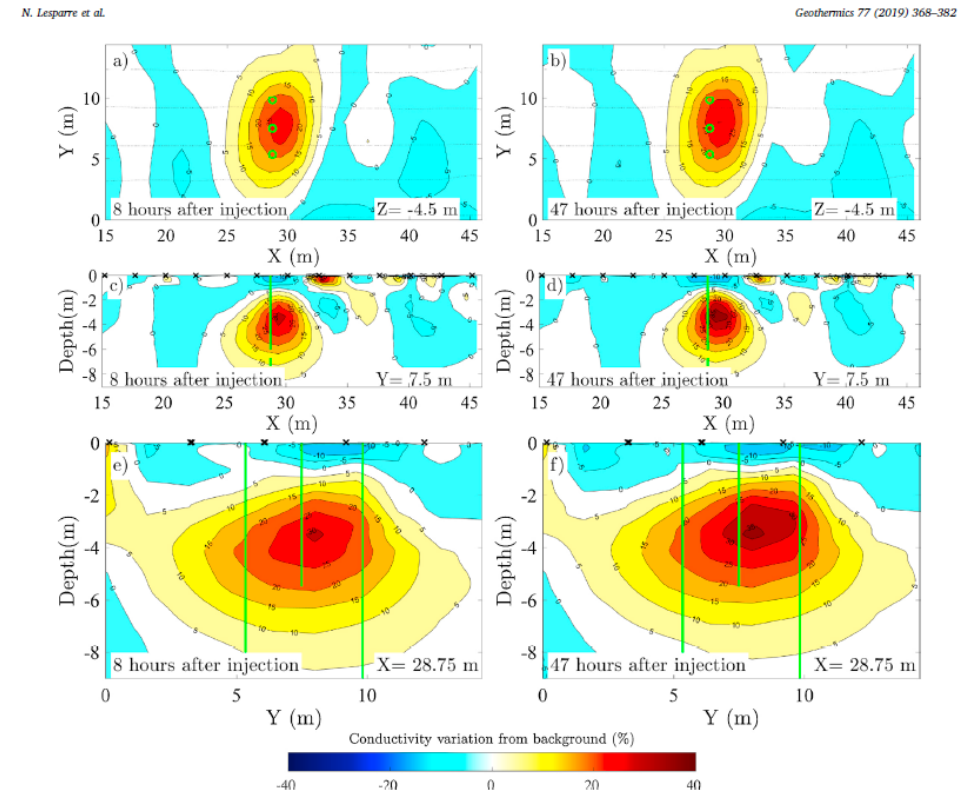
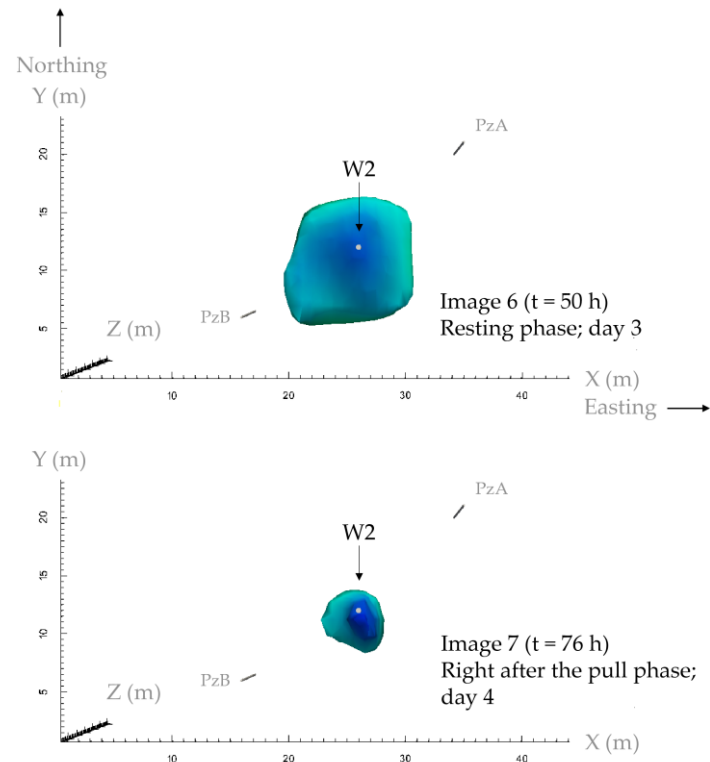


# 3D is a powerful visual tool

e.g. to estimate the thermal affected zone (TAZ)

Robert et al. 2019,  
Geosciences

Lesparre et al. 2019,  
Geothermics, 77, 368-382



Petrophysics

Noise in the data

Survey design

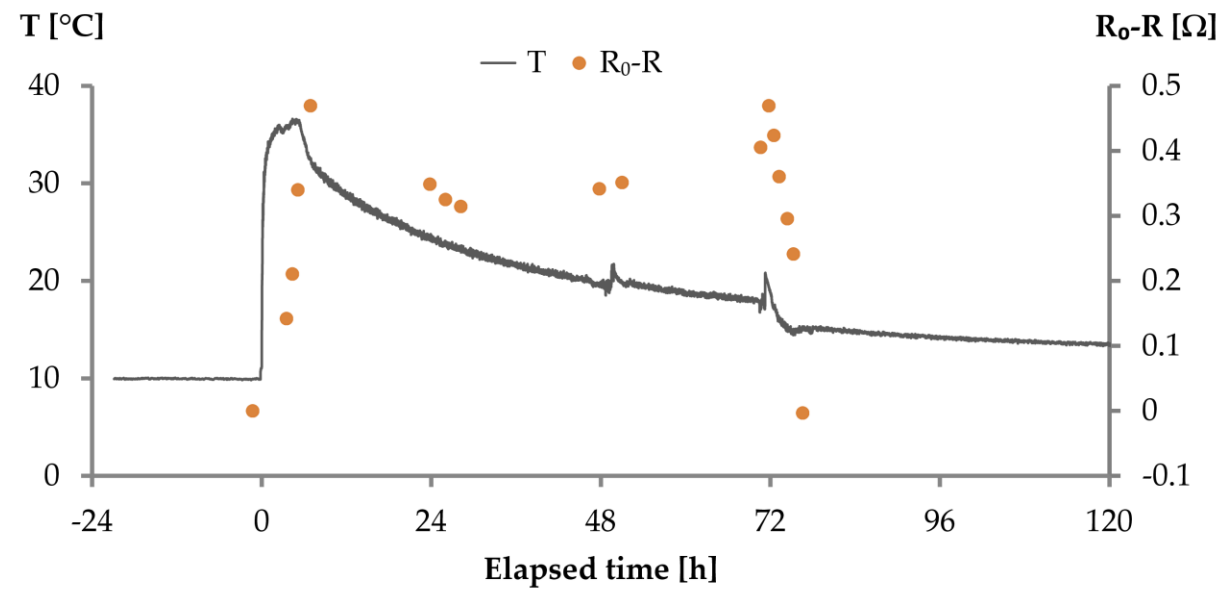
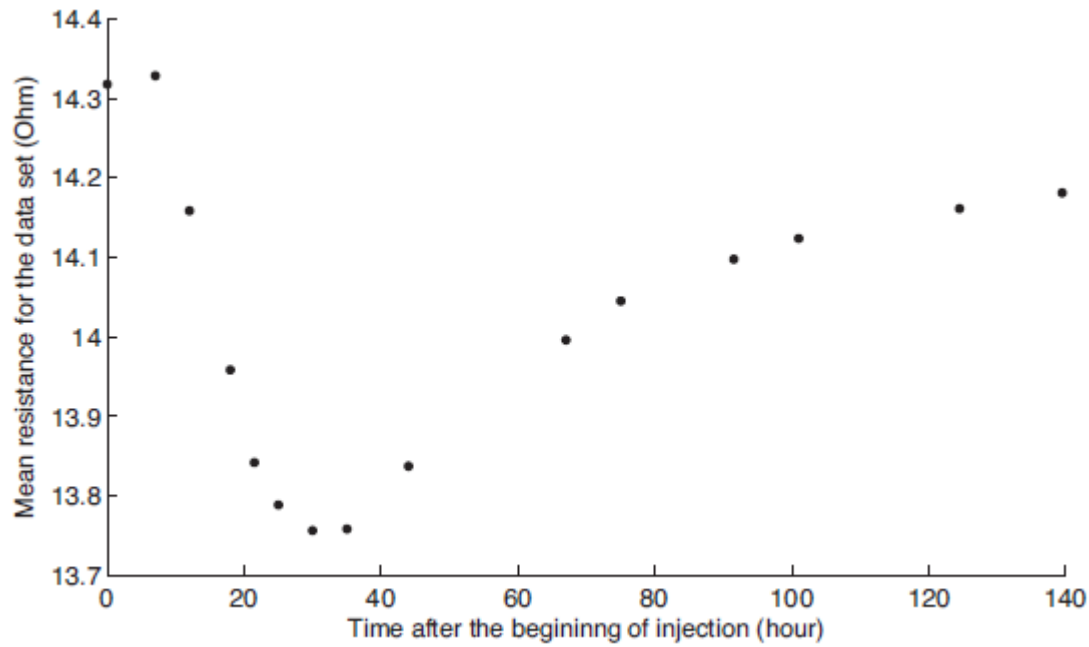
**Imaging**



# First, check if your data contains information about temperature variations

Hermans et al. 2012,  
Geophysics, 77(1), B11-B21

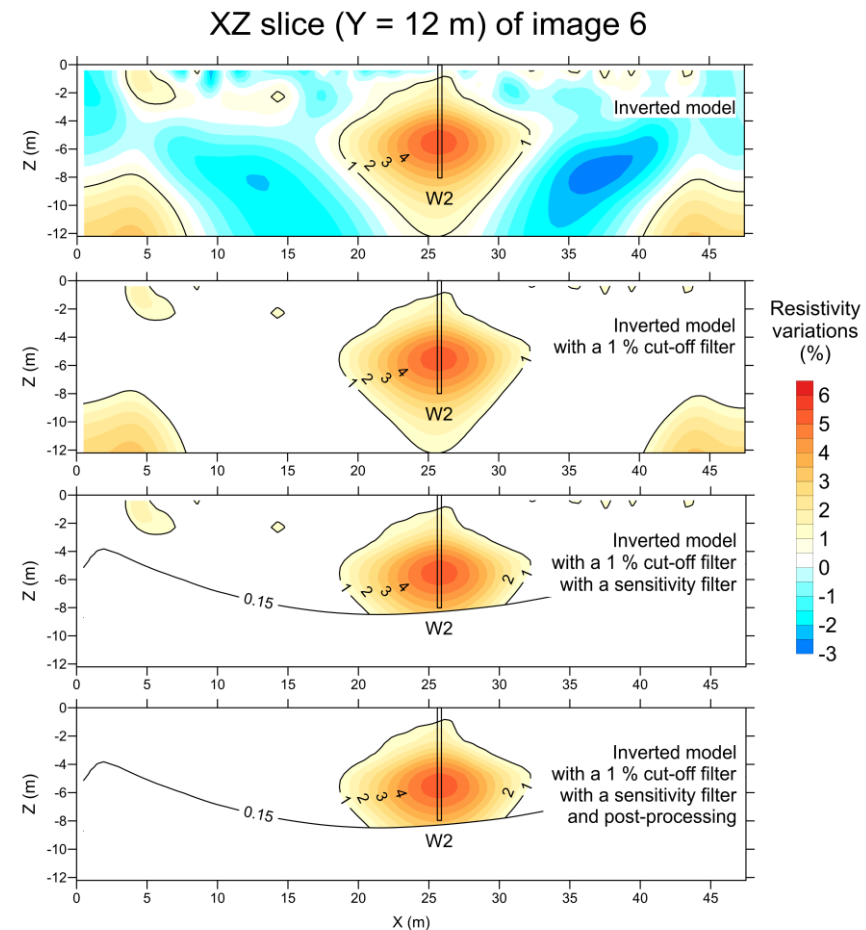
Robert et al. 2019,  
Geosciences



# Choose an appropriate technique

Standalone inversion, data-difference inversion, active time-constrained inversion (Karaoulis et al. 2011, JAG, 73, 25-34), BEL, and many more...

and eventually use filters



Petrophysics

Noise in the data

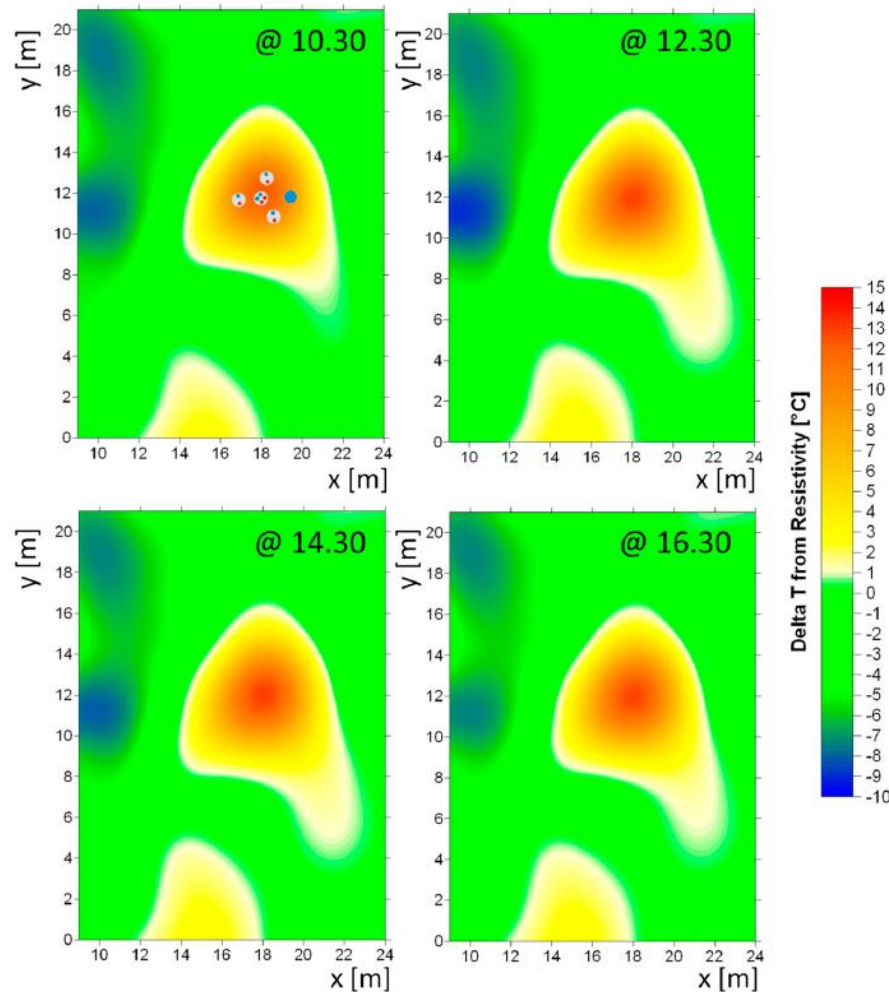
Survey design

Imaging

Applications in real-world studies

# Borehole thermal energy storage BTES

Comina et al. 2019, Geosciences



Monitoring TAZ

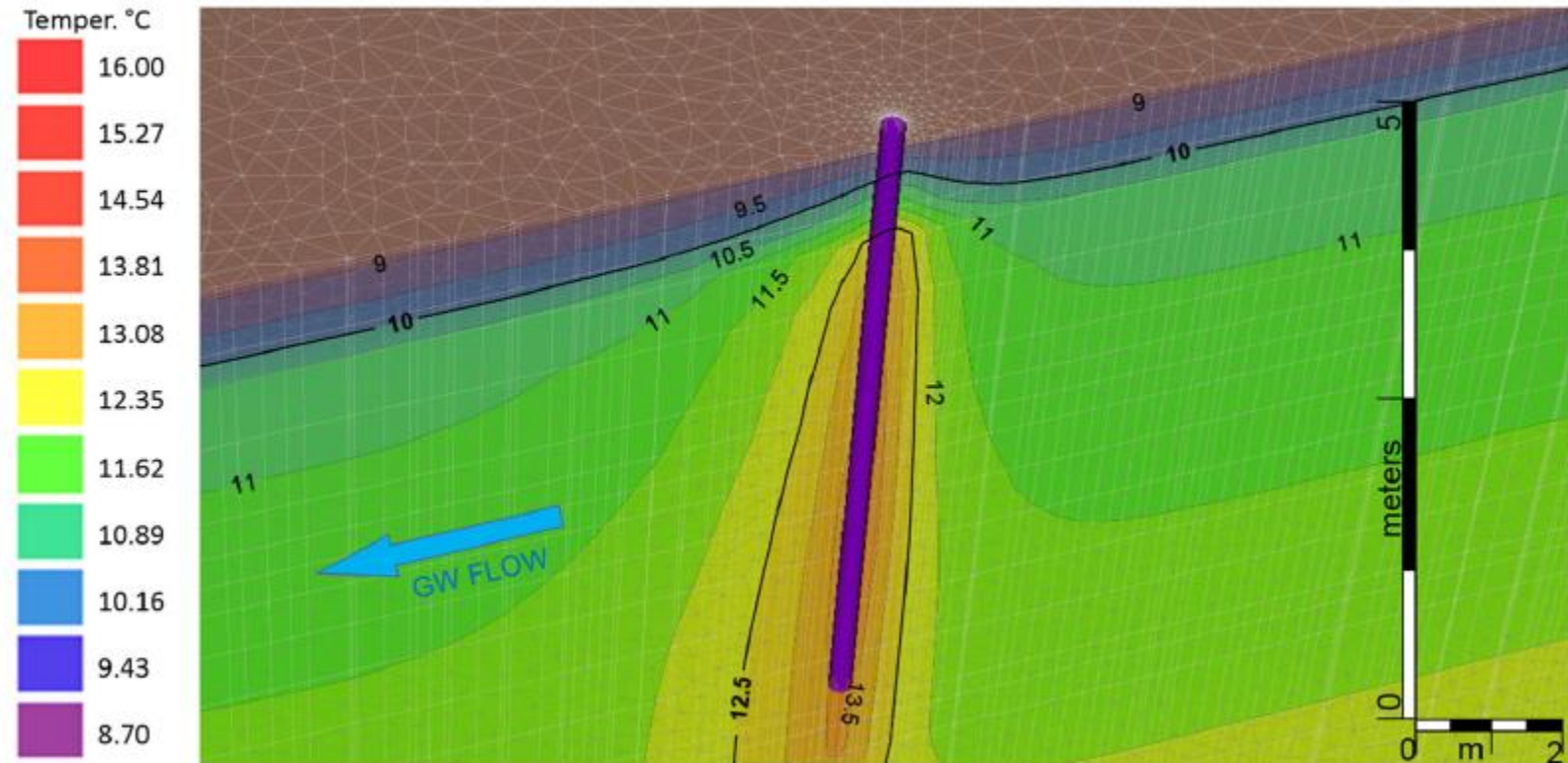
# Borehole thermal energy storage BTES

Cultrera et al. 2018, HJ, 26(3), 837-851

Hydrogeol J (2018) 26:837–851

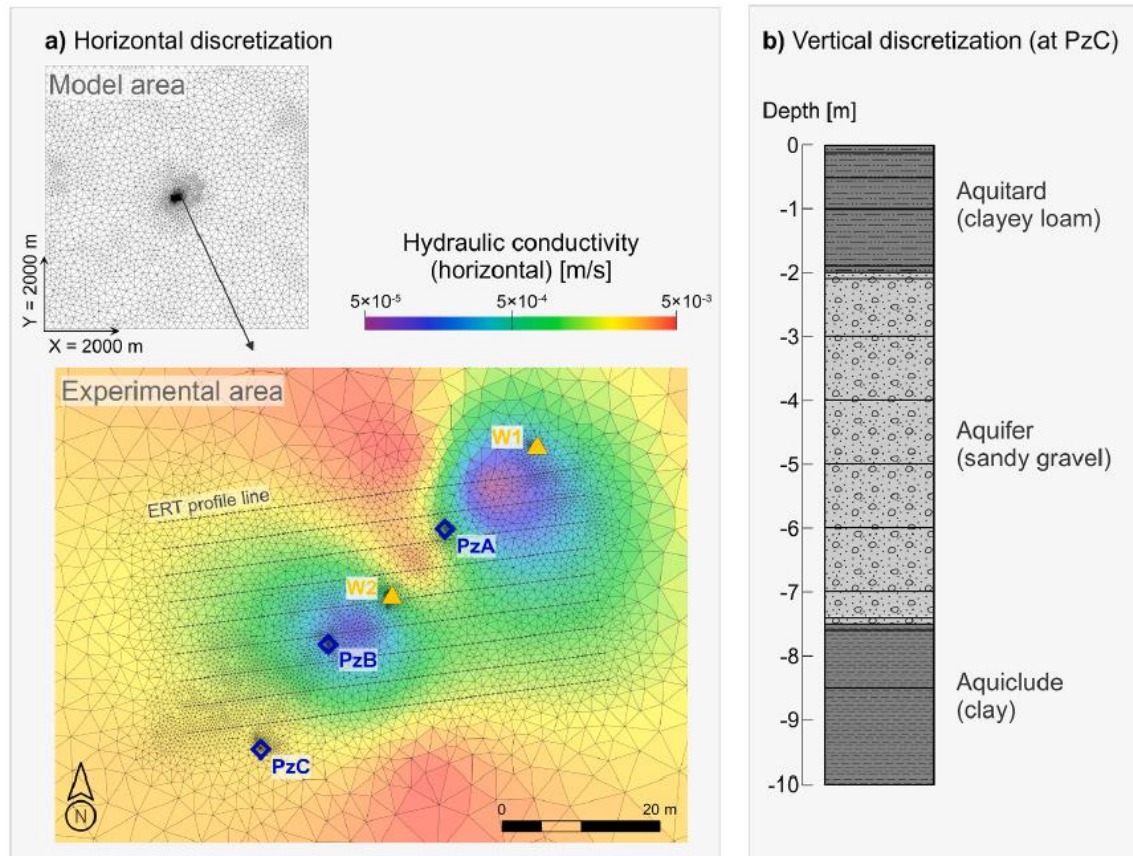
847

Fig. 11 Heat transfer due to the groundwater flow advection and the conduction processes after 5 days of heat power running



# Aquifer thermal energy storage ATES

De Schepper et al. 2019, Applied Energy, 242, 534-546



**Table 4**

Energy recovery rate values ( $\eta$ , in %) for the 77 DSM simulated scenarios (7 storage periods  $\Delta t$ , 11 temperature differences induced  $\Delta T$ ).

$\eta$ [%]	$\Delta T$ [K]											
	-4	3	6	11	35	40	45	50	55	60	65	
$\Delta t$ [h]	0.25	87	87	87	87	71	71	70	68	67	66	65
	1	87	87	87	87	71	71	69	68	67	66	65
	6	87	87	87	87	71	71	70	68	67	66	65
	12	86	87	87	87	71	70	69	68	66	65	64
	24	84	84	84	83	67	67	65	64	63	61	61
	48	84	83	83	84	67	66	65	63	62	60	59
	72	78	78	78	78	62	60	58	57	55	54	53



# Aquifer thermal energy storage ATES

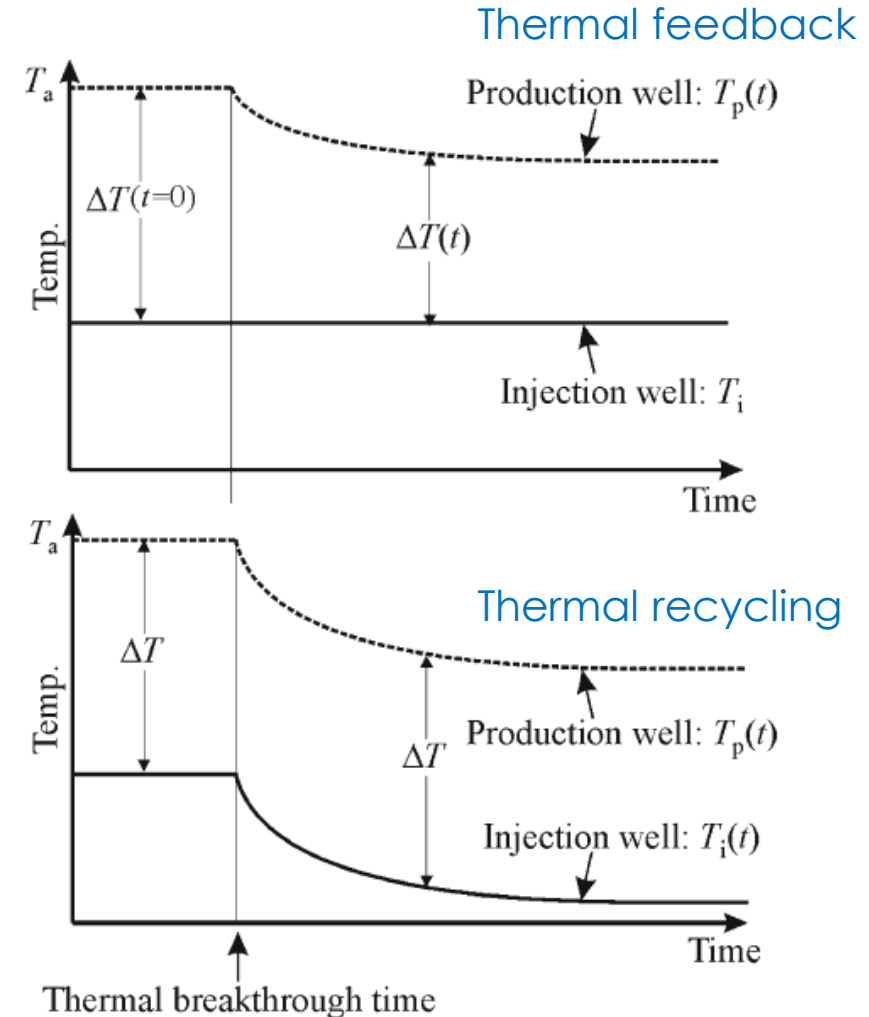
Robert et al. In preparation

Idea:  
Repeat ATES  
& recovery cycles

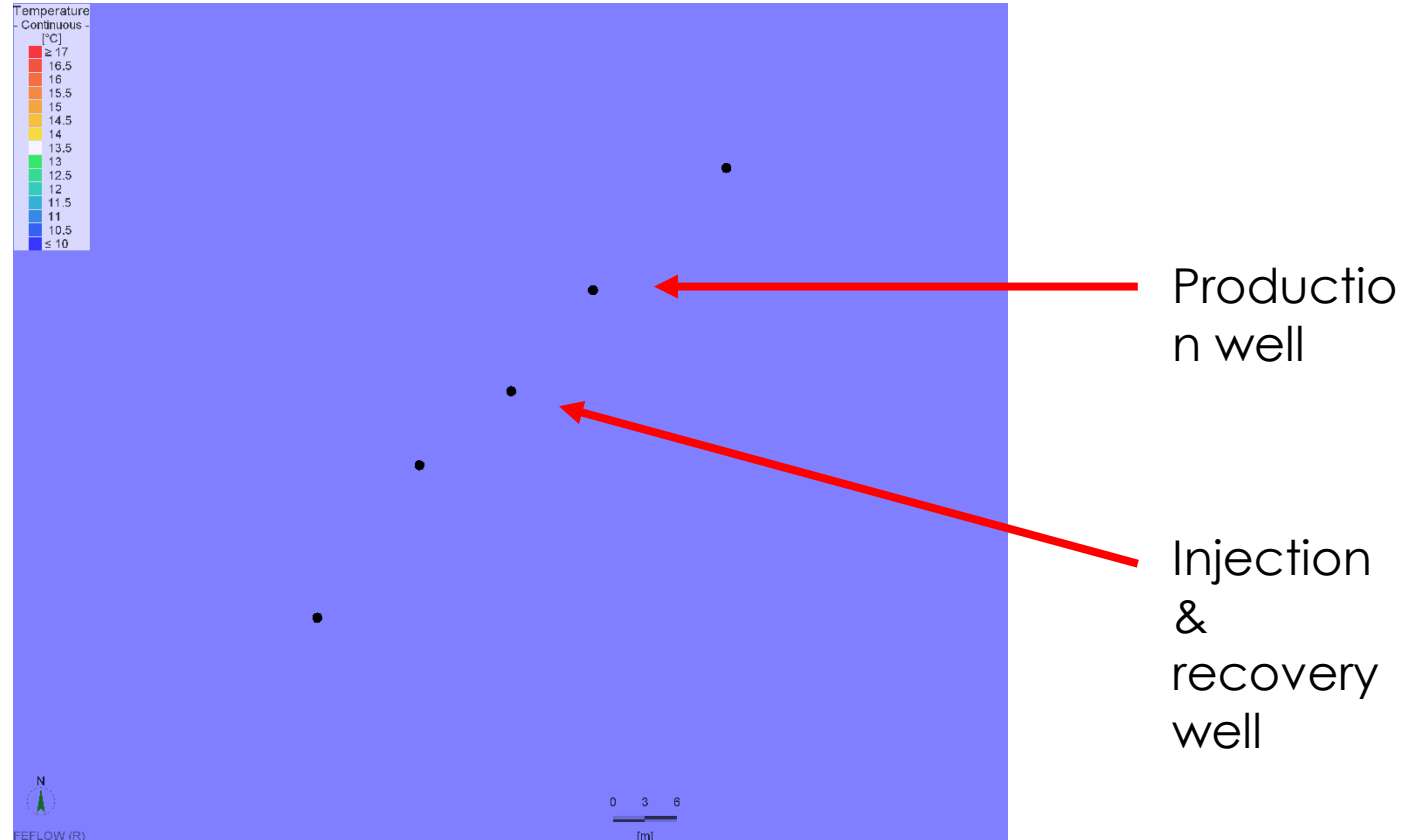
This time,  
with thermal recycling



Milnes & Perrochet (2013)



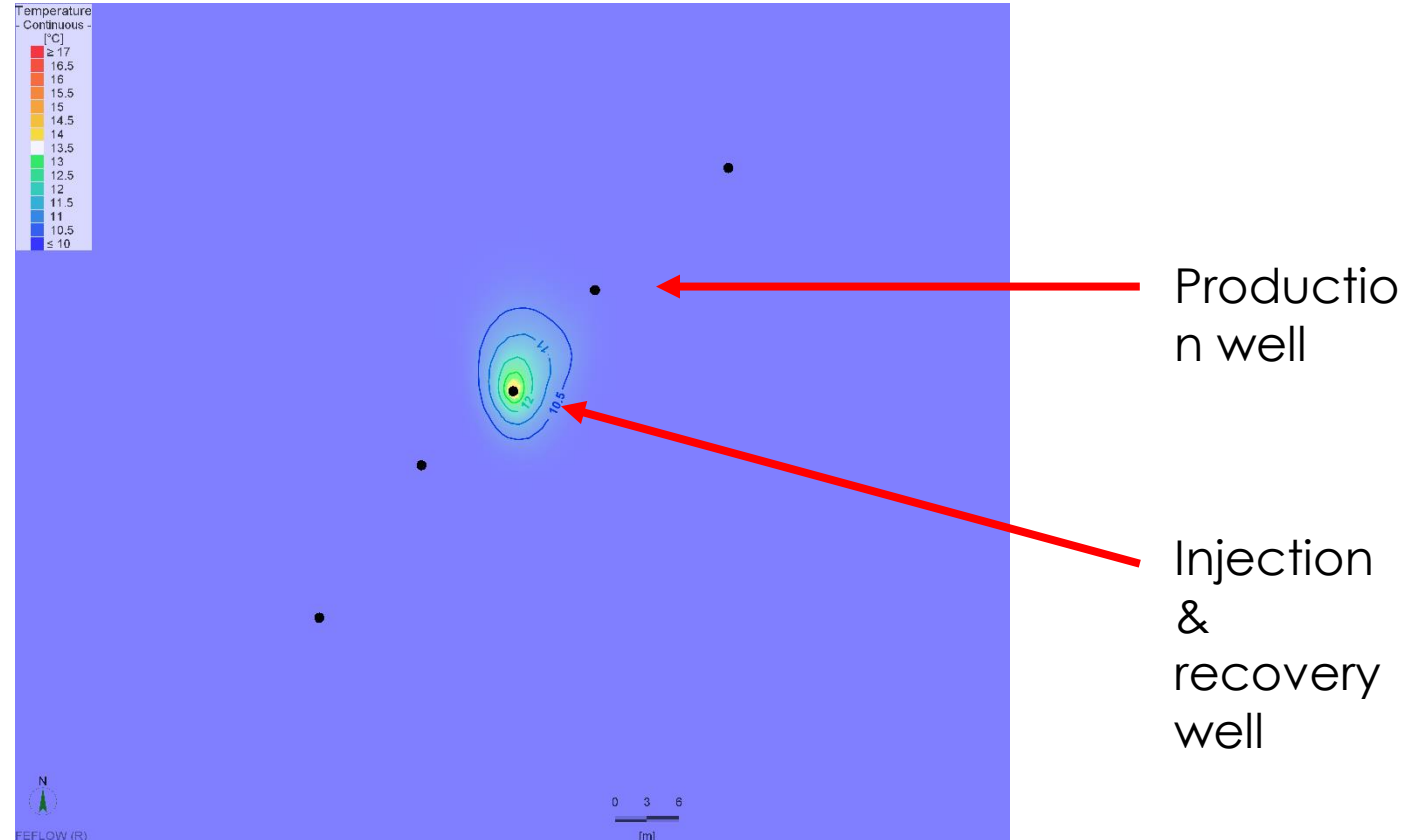
# Thermal recycling helps containing the heat plume in the area of interest



Example of LT-ATES ( $\Delta T = 6K$ ) – Day 0

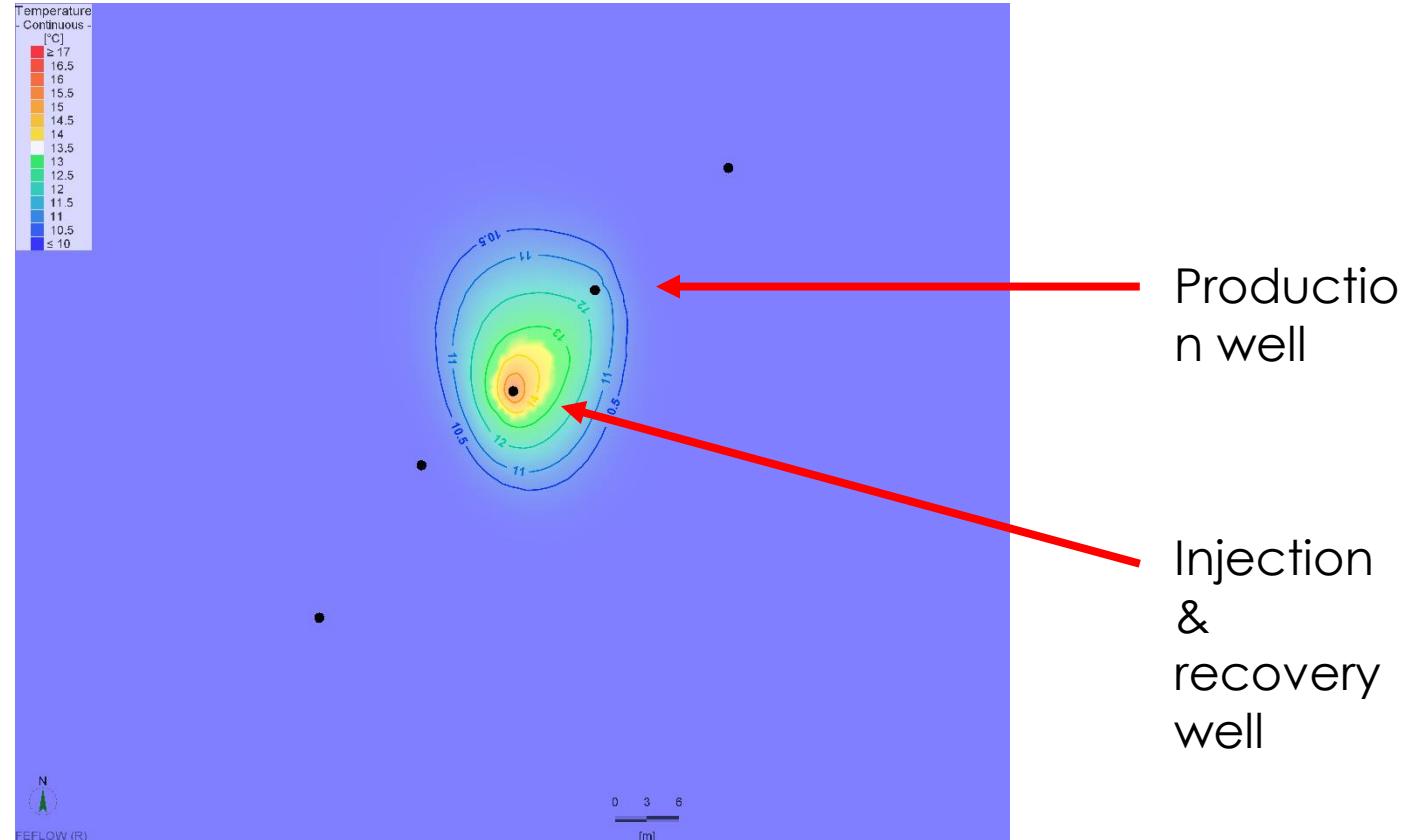


# Thermal recycling helps containing the heat plume in the area of interest



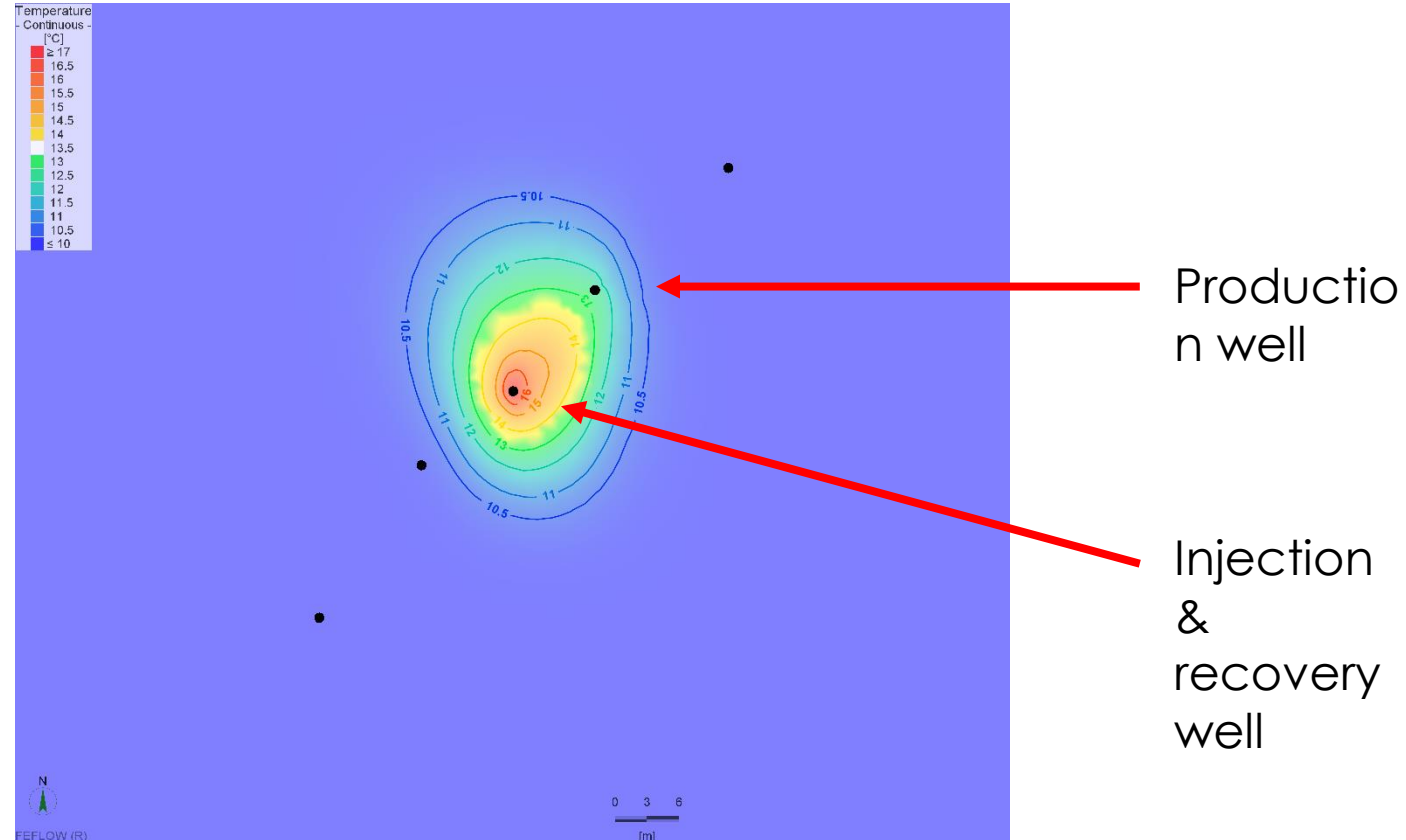
Example of LT-ATES ( $\Delta T = 6K$ ) – Day 1

# Thermal recycling helps containing the heat plume in the area of interest



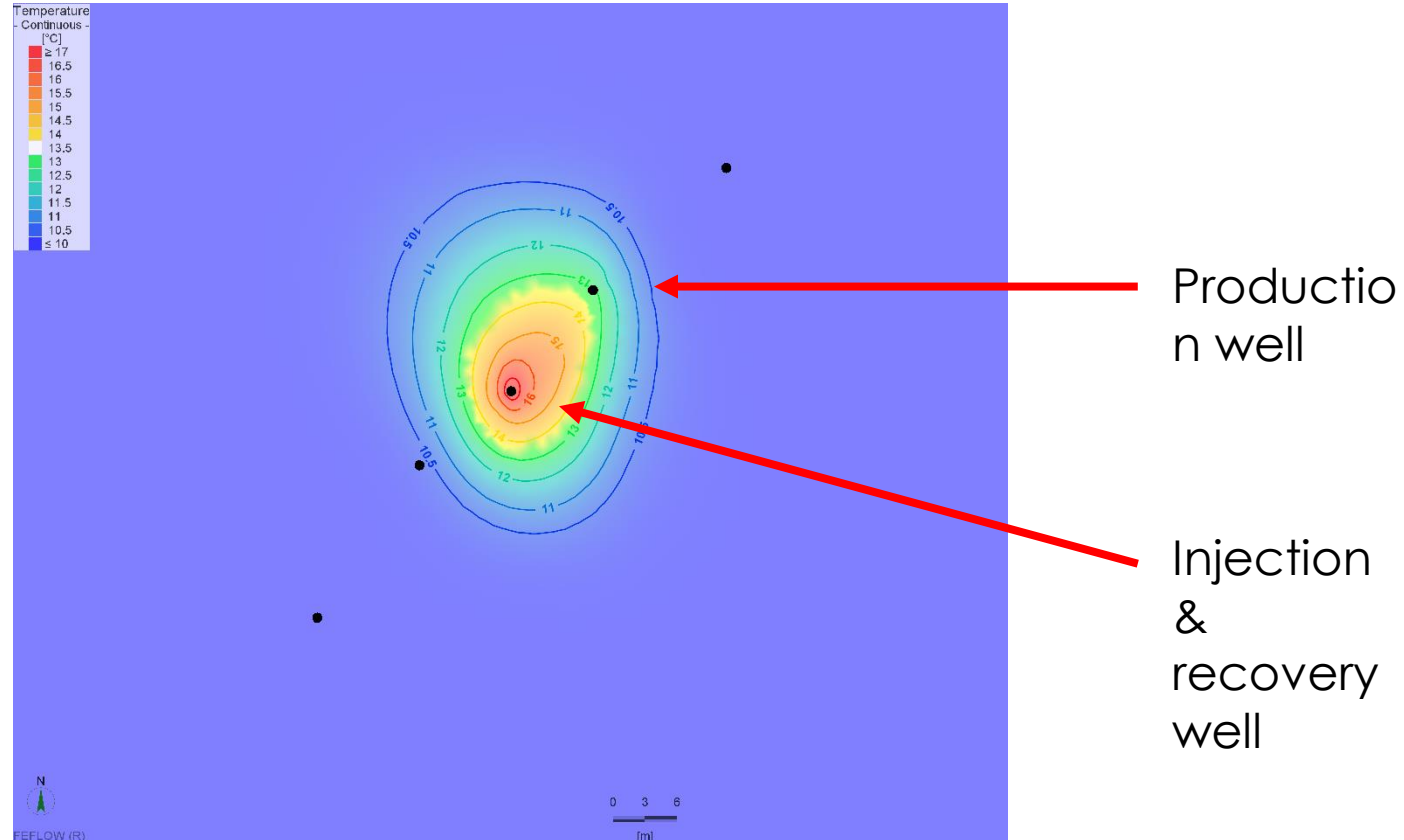
Example of LT-ATES ( $\Delta T = 6K$ ) – Day 7

# Thermal recycling helps containing the heat plume in the area of interest



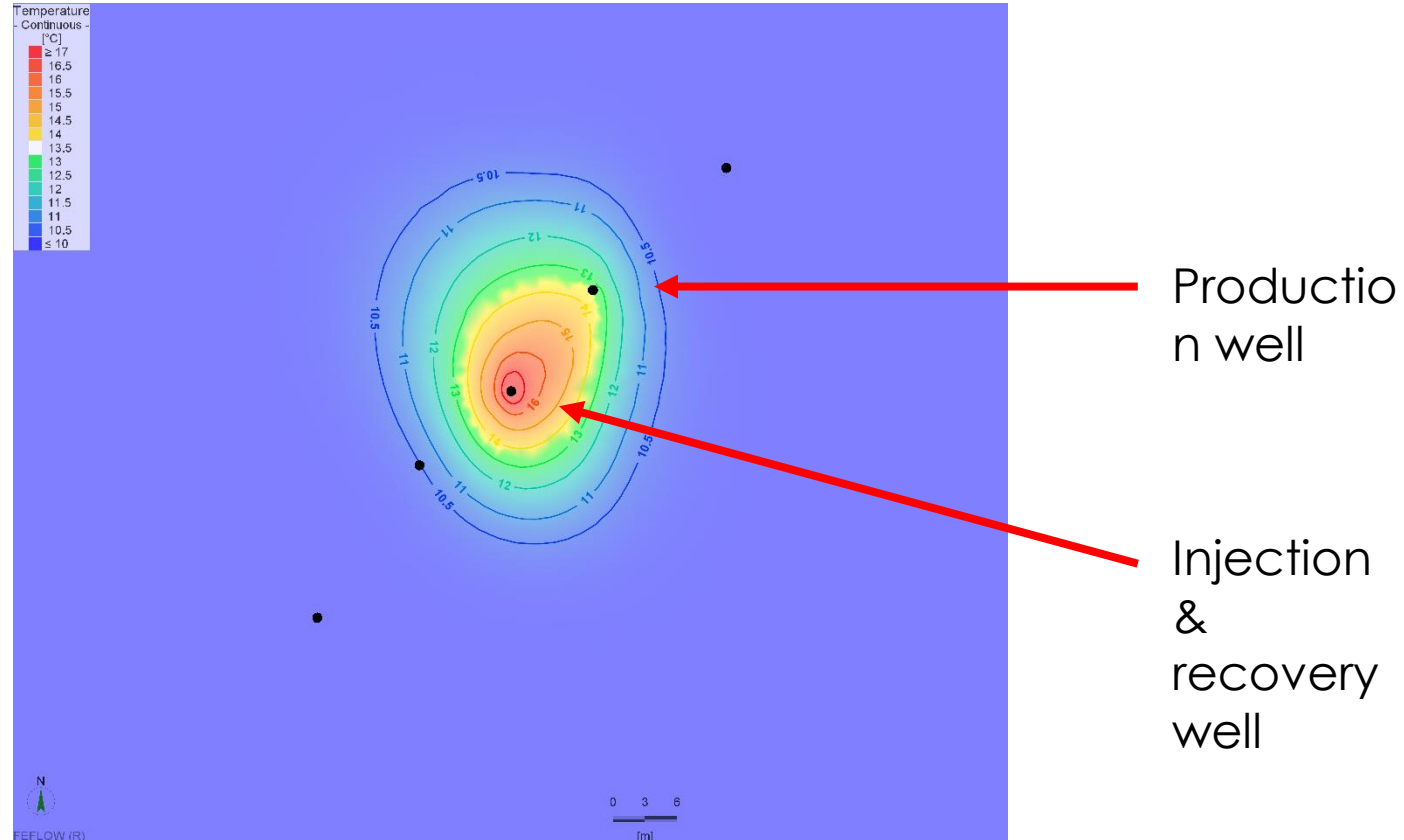
Example of LT-ATES ( $\Delta T = 6K$ ) – Day 14

# Thermal recycling helps containing the heat plume in the area of interest



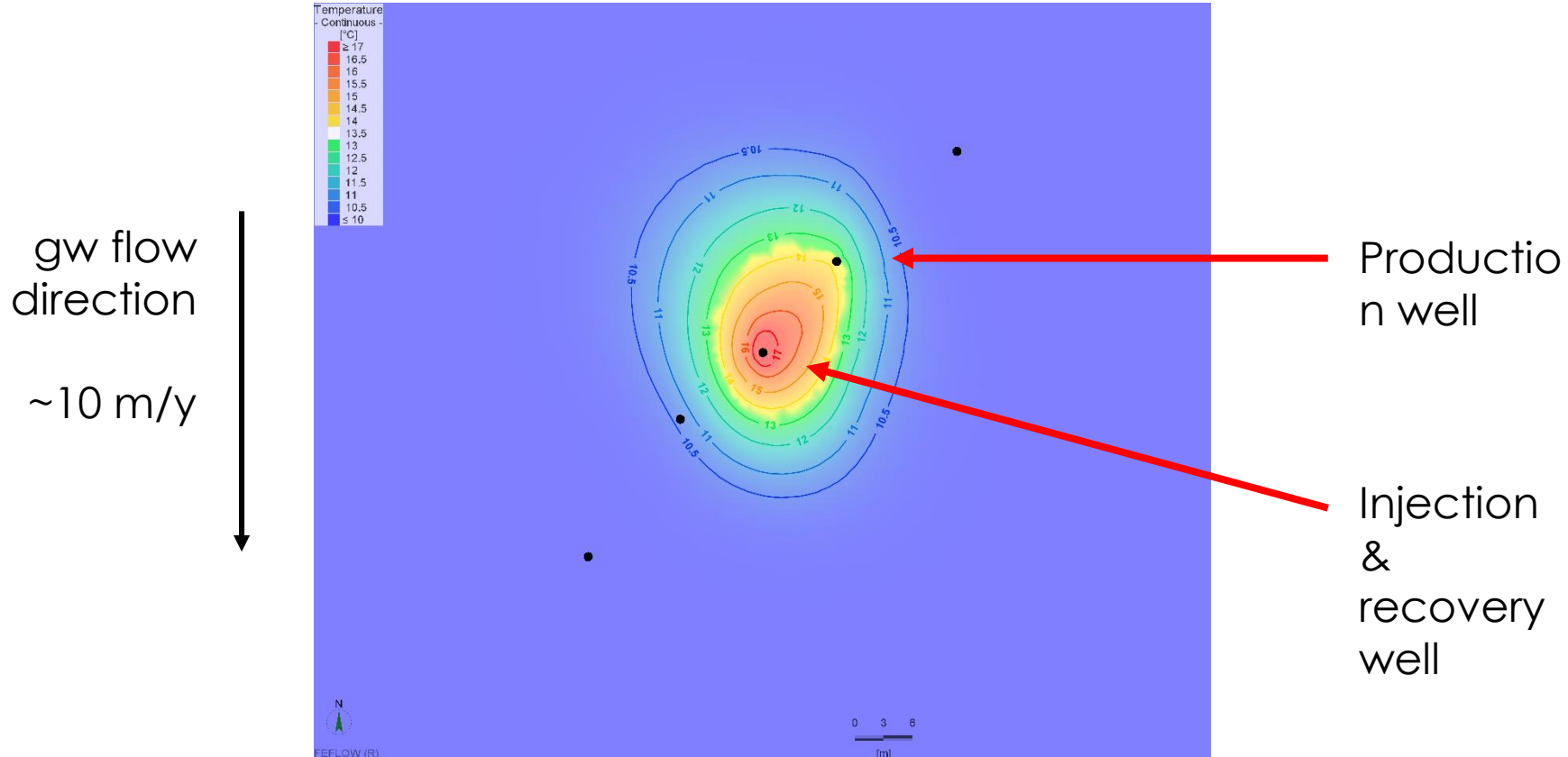
Example of LT-ATES ( $\Delta T = 6K$ ) – Day 21

# Thermal recycling helps containing the heat plume in the area of interest



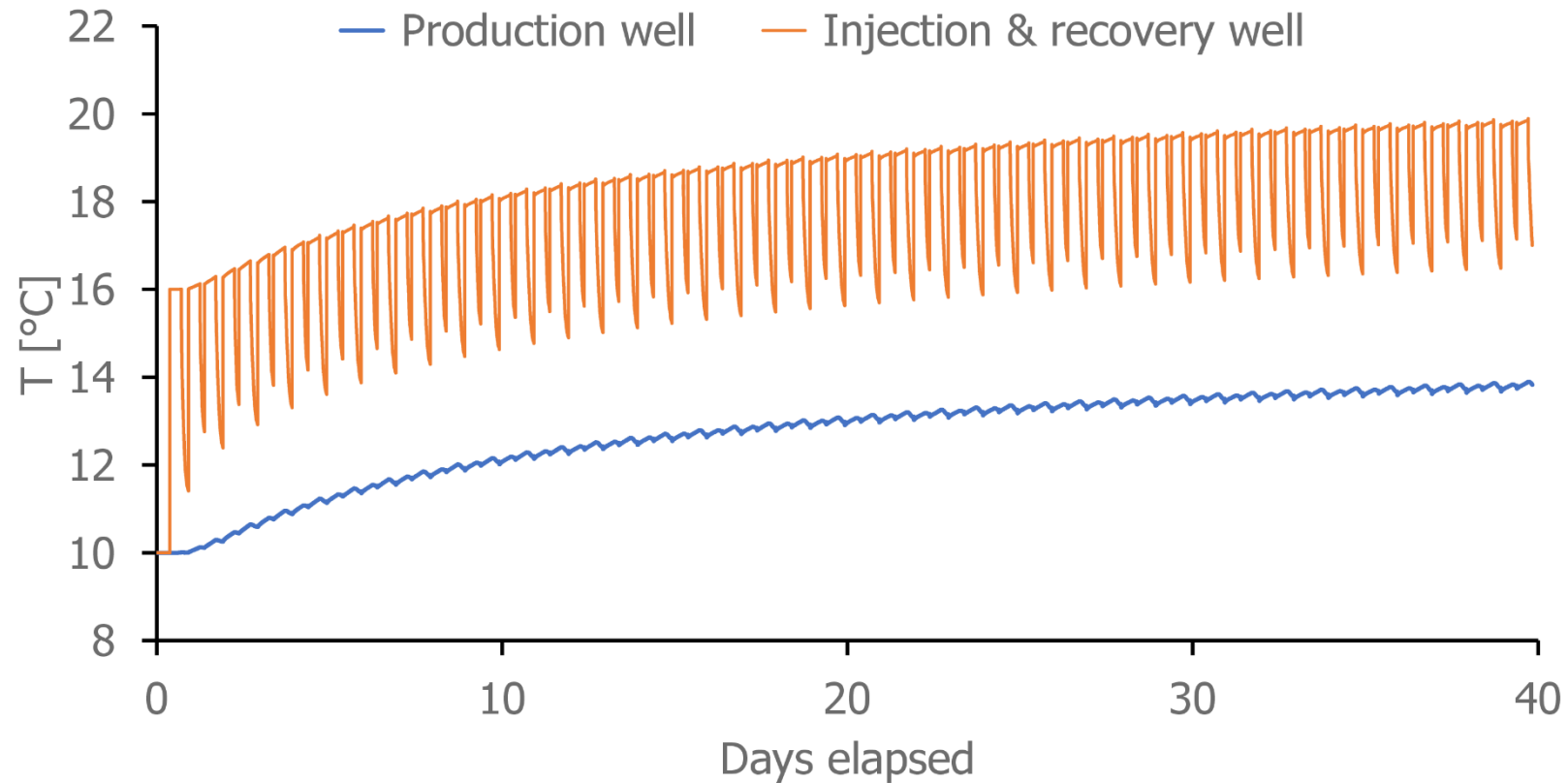
Example of LT-ATES ( $\Delta T = 6K$ ) – Day 28

# Thermal recycling helps containing the heat plume in the area of interest



Example of LT-ATES ( $\Delta T = 6K$ ) – Day 35

For LT-ATES ( $\Delta T = 6\text{K}$ ),  
this strategy works



Heat is stored during off-peak periods and recovered during peak periods (of heat demand)

Heat is then either stored or recovered for heating applications

The key message  
is to focus on the survey design  
and the collection of a good data set

From then, anything is possible