



THE UNIVERSITY  
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# Radical Uncertainty in Subsurface Science

## How do we make decisions?

Prof. Andrew Curtis

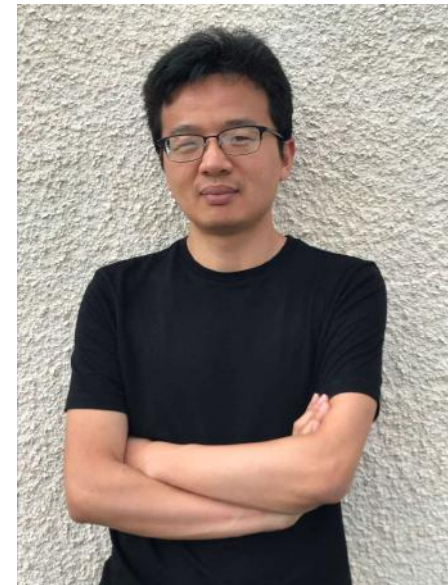


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# Radical Uncertainty in Subsurface Science

## How do we make decisions?



*Special credit to  
Richard Arnold, Xuebin Zhao and Xin Zhang + 1*

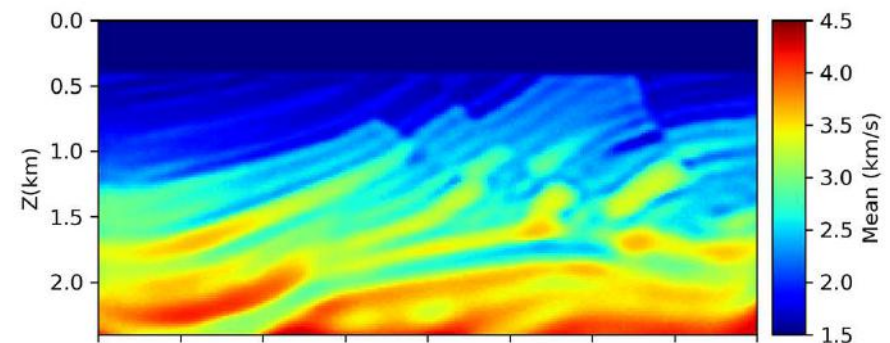
# Interrogation Theory

Seek **low-dimensional** answers by interrogating models & data

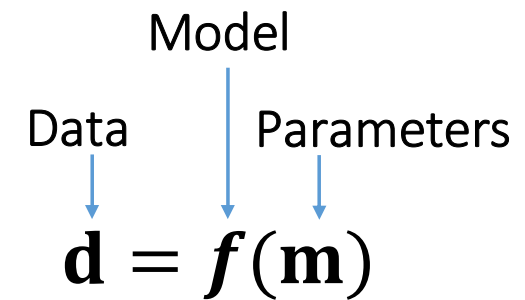
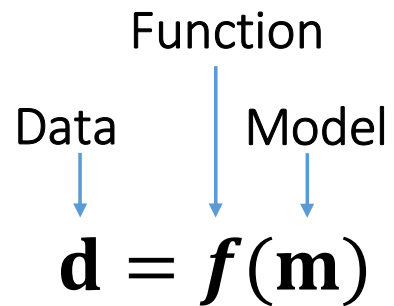
- What is the volume of a particular subsurface reservoir? →  $X \text{ km}^3$
- Has more than  $1 \text{ Kt}$  of  $\text{CO}_2$  escaped from the subsurface store? → Yes/No
- Which dynamic model best explains the *true* seismicity? → Model 1,2,3...

Answers often lie within **high dimensional spaces**

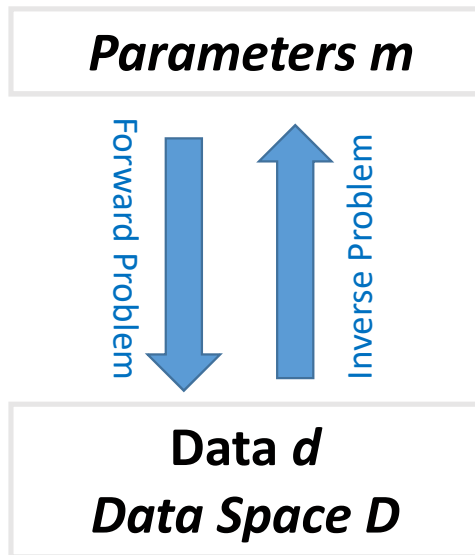
- Introduce Interrogation Theory
- 2D seismic imaging example
- 3D & 4D interrogation is now feasible
- *Radical* uncertainty: How do we make decisions?



# Inverse Theory and Interrogation Theory



# Inverse Theory and Interrogation Theory



Inverse Theory

Model

Data Parameters

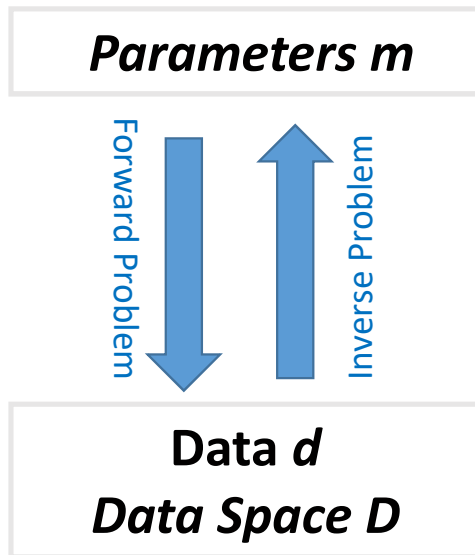
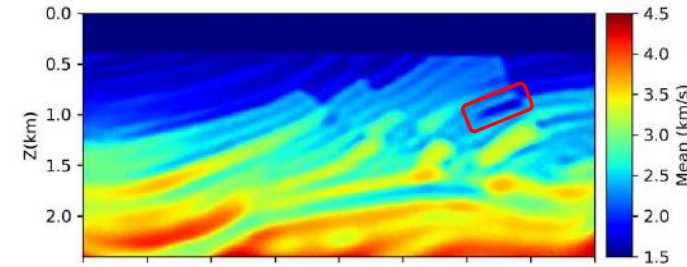
$$\mathbf{d} = \mathbf{f}(\mathbf{m})$$

→ Find parameter values that fit observed data best

The parameter values are definitely wrong!

# Inverse Theory and Interrogation Theory

Has more than 1 Kt of CO<sub>2</sub> escaped from the store?



Inverse Theory

→ Bayes' theorem

$$p(\mathbf{m} | \mathbf{d}_{obs}) = \frac{p(\mathbf{d}_{obs} | \mathbf{m}) p(\mathbf{m})}{p(\mathbf{d}_{obs})}$$

likelihood

Prior pdf

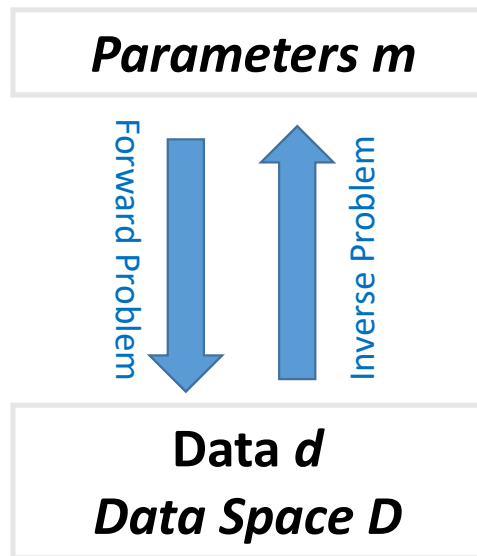
evidence

Posterior probability distribution function (pdf)

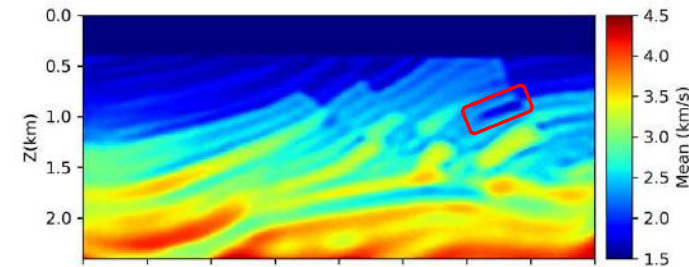
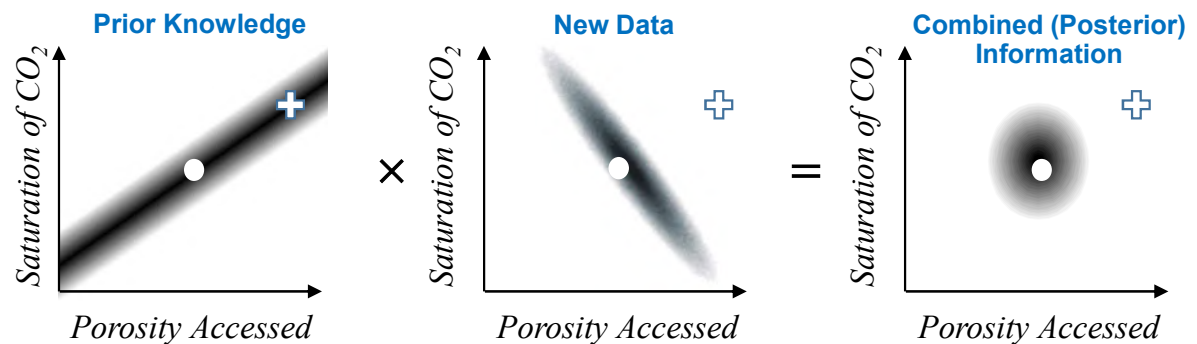
→ Probability that  $\mathbf{m}$  is true – given model assumptions & observed data  $\mathbf{d}_{obs}$

# Inverse Theory and Interrogation Theory

Has more than 1 Kt of CO<sub>2</sub> escaped from the store?



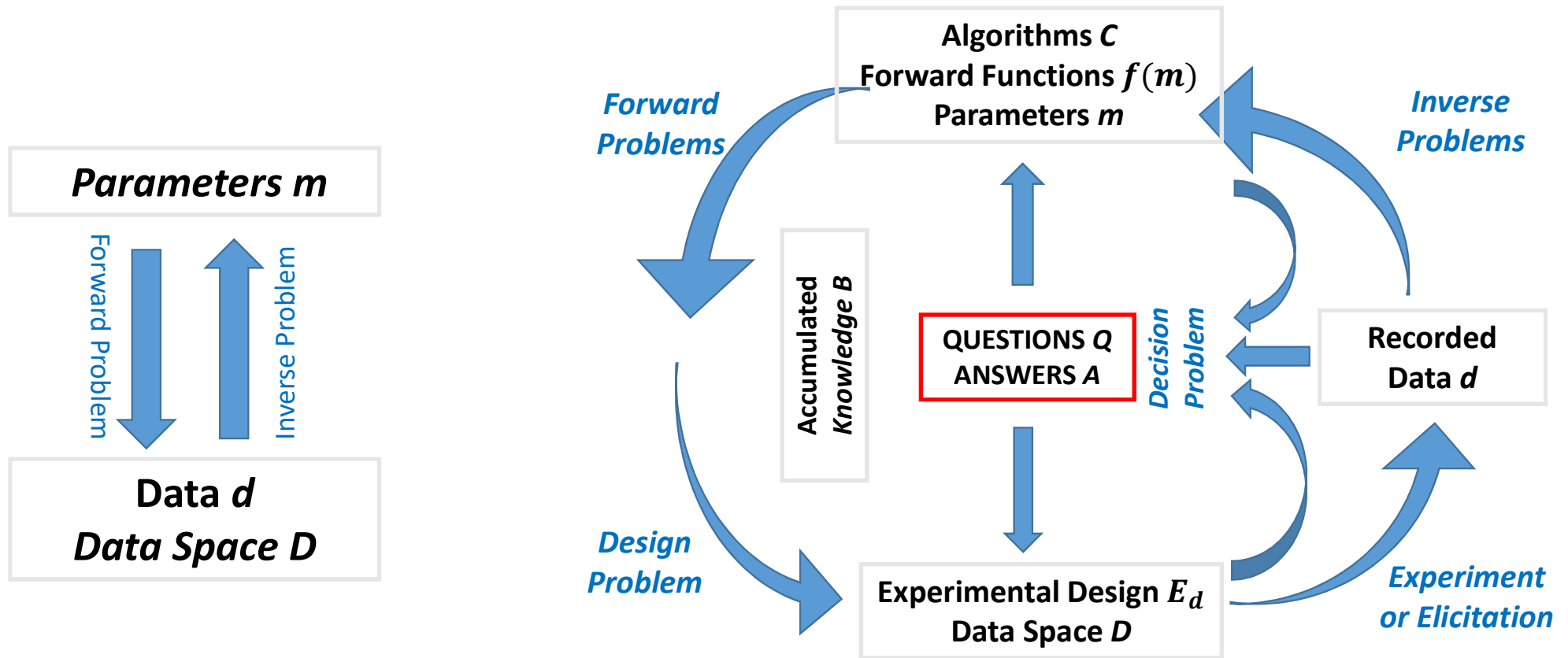
→ Bayes' theorem



Given our question, which parameter values should we evaluate?

Inverse Theory

# Inverse Theory and Interrogation Theory



Inverse Theory

Interrogation Theory

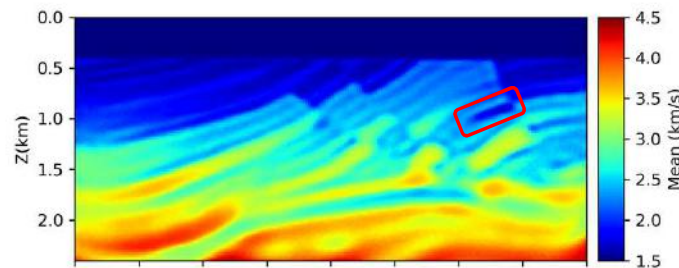
*Arnold & Curtis (2018)  
Geophys. J. Int.*



## Synthetic Example

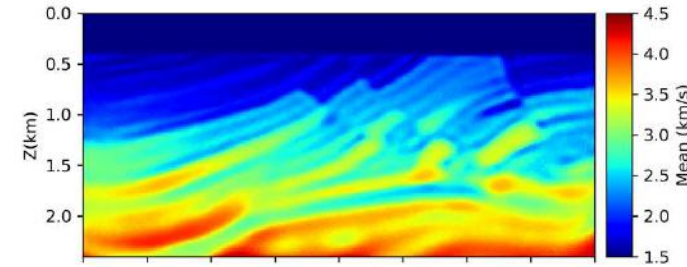
**Decision:** If reservoir volume  $> V$ , consider it as a CO<sub>2</sub> storage asset

Q: What is the volume of this reservoir?



# What is the size of a reservoir?

- **Parametrization:**  $\theta_m$  = seismic velocity  $V(\mathbf{x})$
- **Data:** Seismic full waveforms (data  $y_d$ , design  $d$ )
- **Modelling:** Finite difference modelling (model  $m$ :  $\theta_m \rightarrow y_d$ )
- Define **target** function  $t = T(\theta_m|m, Q)$ :  
transforms any velocity structure  $\theta_m$  into corresponding reservoir volume
- Define Utility:  $U(a|t) = -(a - t)^2$  <sup>②</sup> Defines the net benefits of any answer



*Arnold & Curtis (2018), Geophys. J. Int.*  
*Zhang & Curtis (2022), Geophys. J. Int.*

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- **Optimal Answer** is then:

$$a^* = \sum_{m \in \mathbb{M}} \int_{\theta_m} T(\theta_m | m) p(\theta_m, m | y_d, d) d\theta_m$$

*Arnold & Curtis (2018), Geophys. J. Int.*

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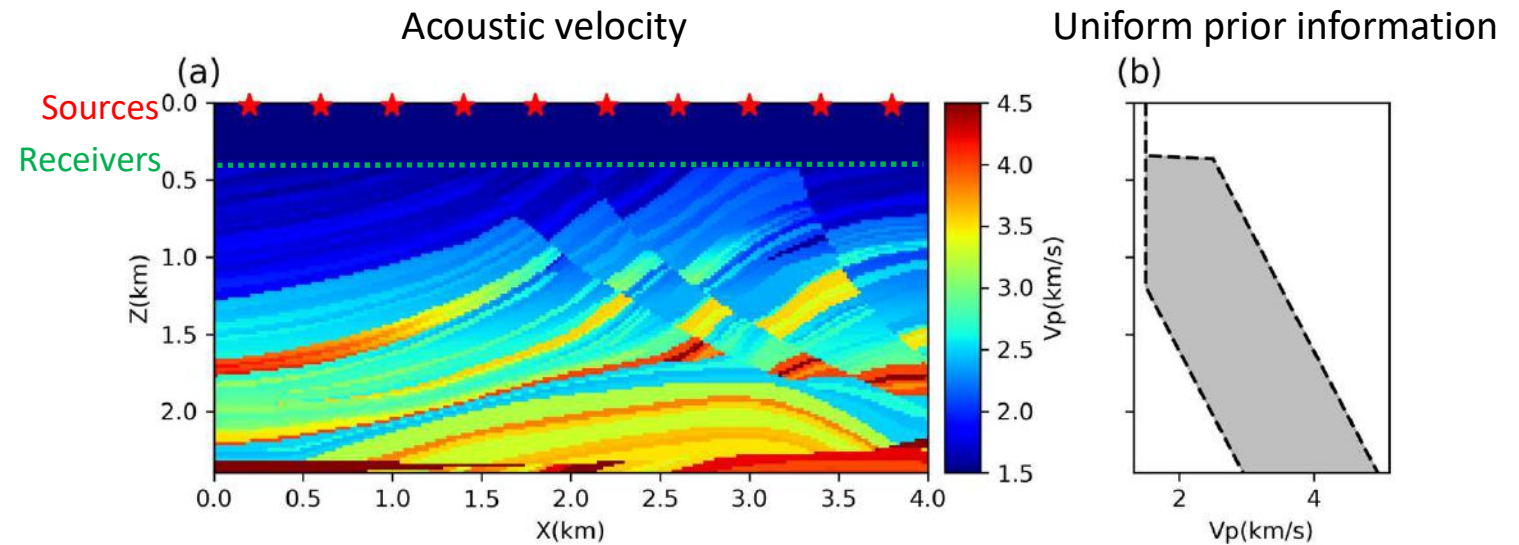
*Average over all possible models*      Target      Bayesian Posterior pdf

*Arnold & Curtis (2018), Geophys. J. Int.*

*Zhang & Curtis (2022), Geophys. J. Int.*

# Variational full-waveform inversion

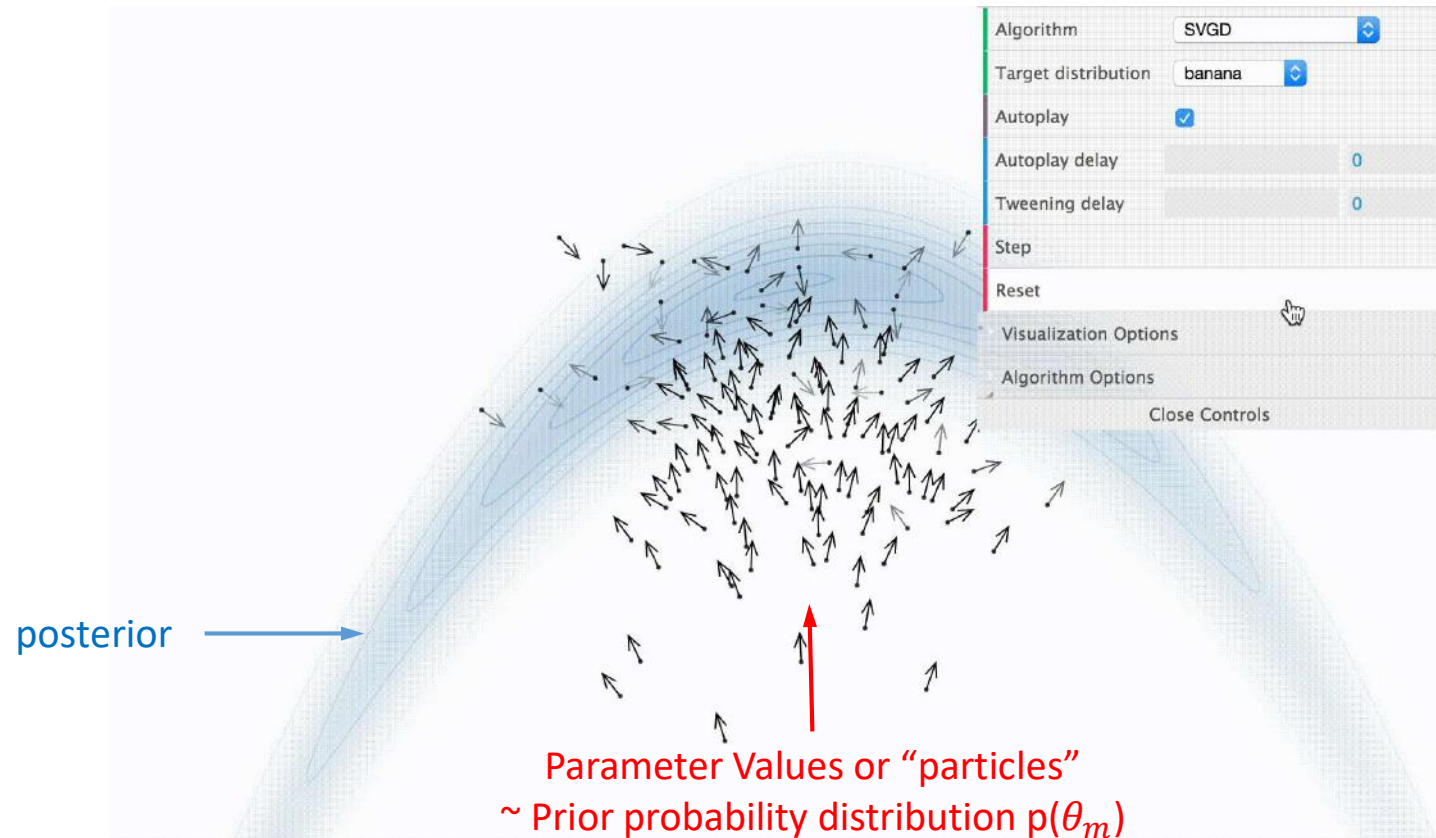
- Acoustic Marmousi model
- 10 shots, 200 receivers
- 200\*120 free parameters
- Uniform prior
- Data simulated using 4 Hz Ricker wavelet



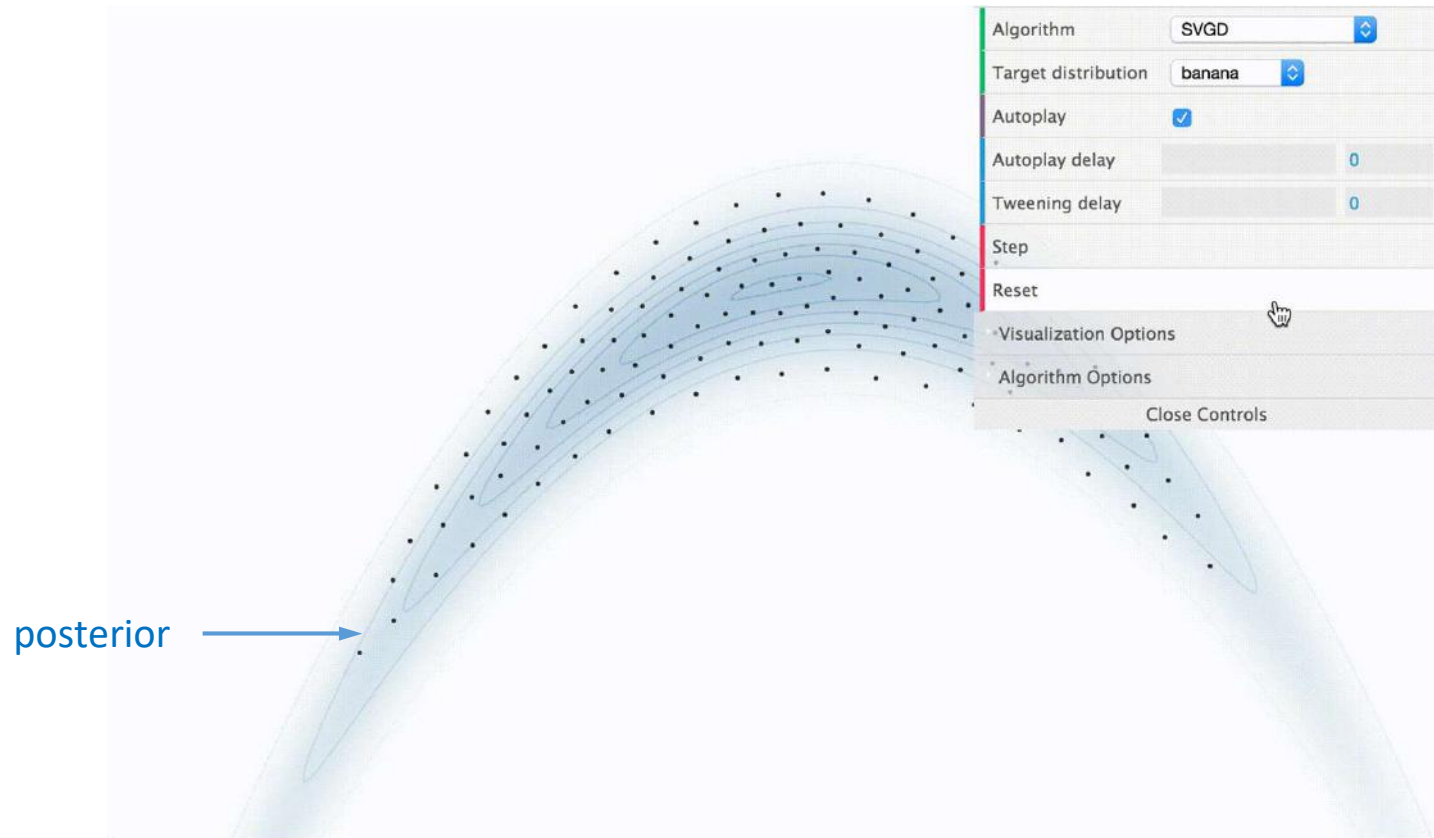
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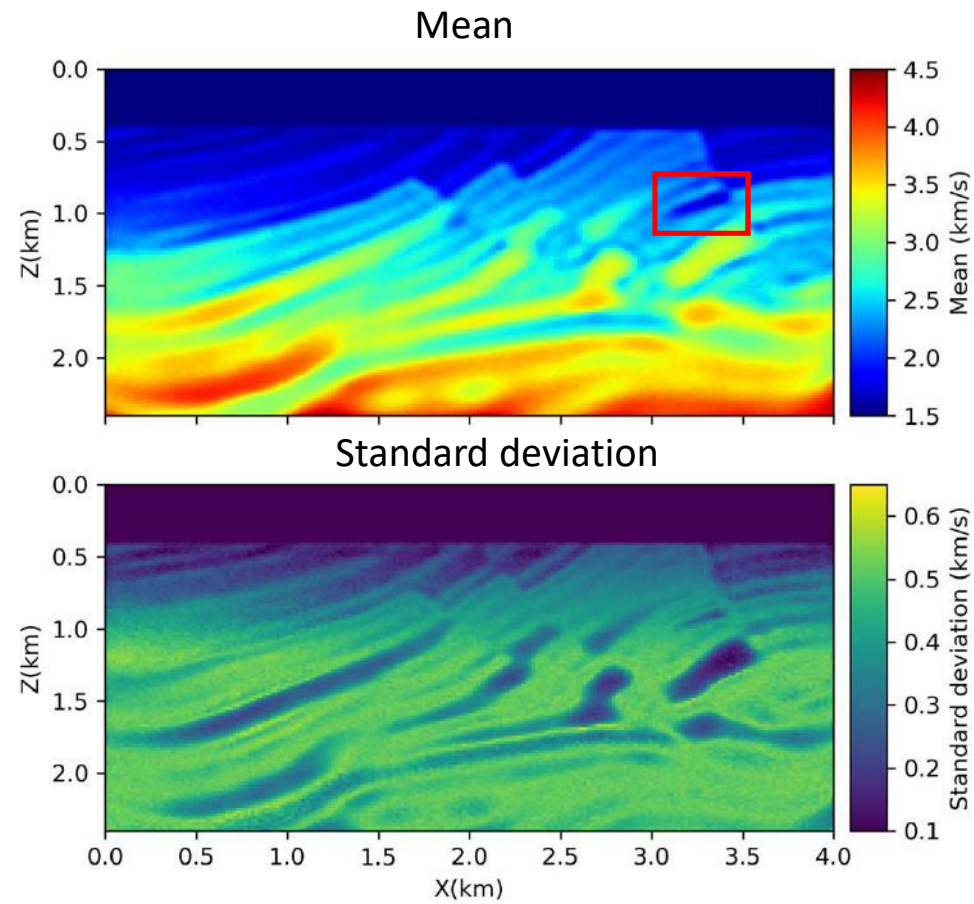
# Stein variational gradient descent (SVGD)



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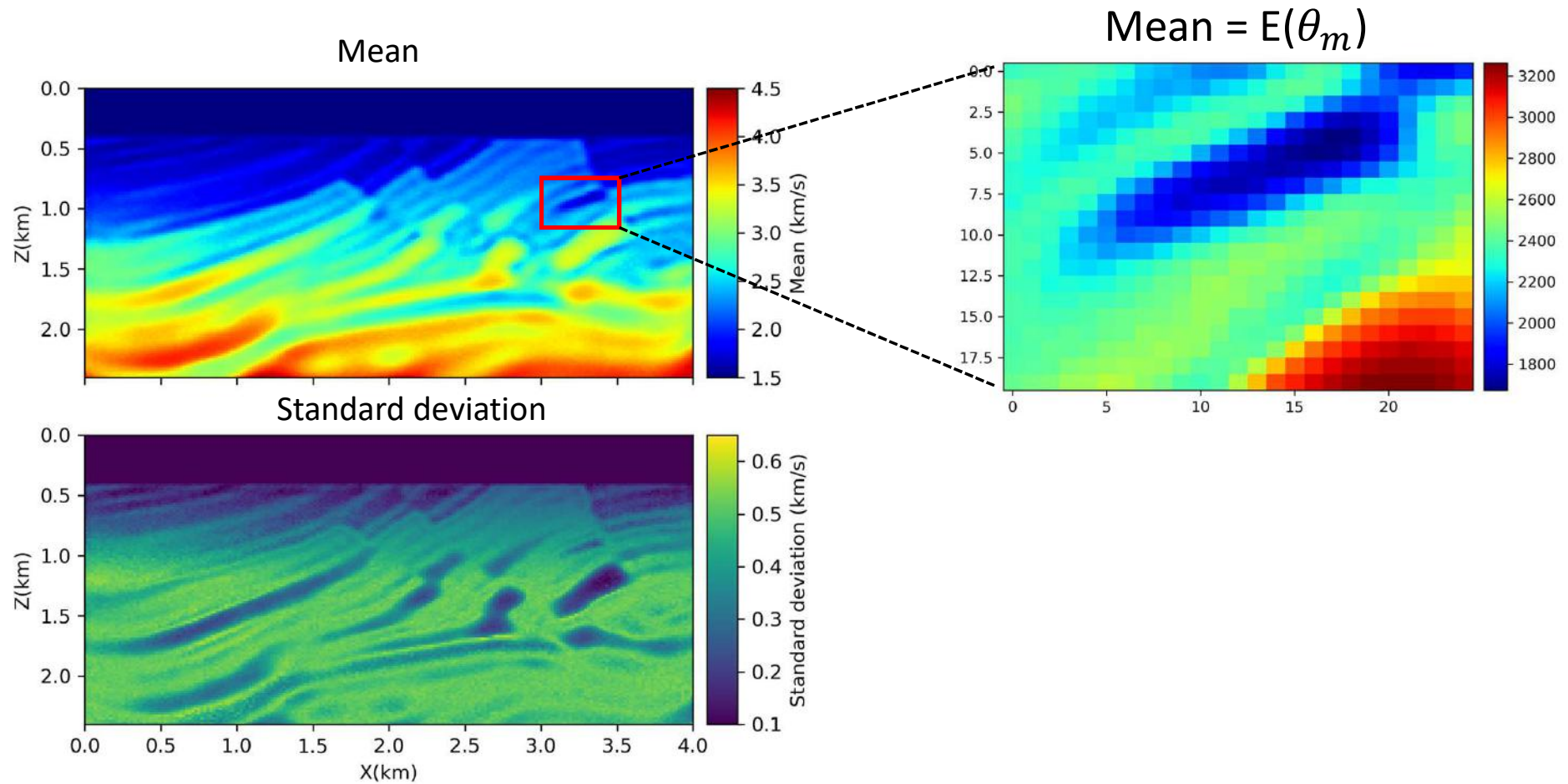


# What is the size of a reservoir?

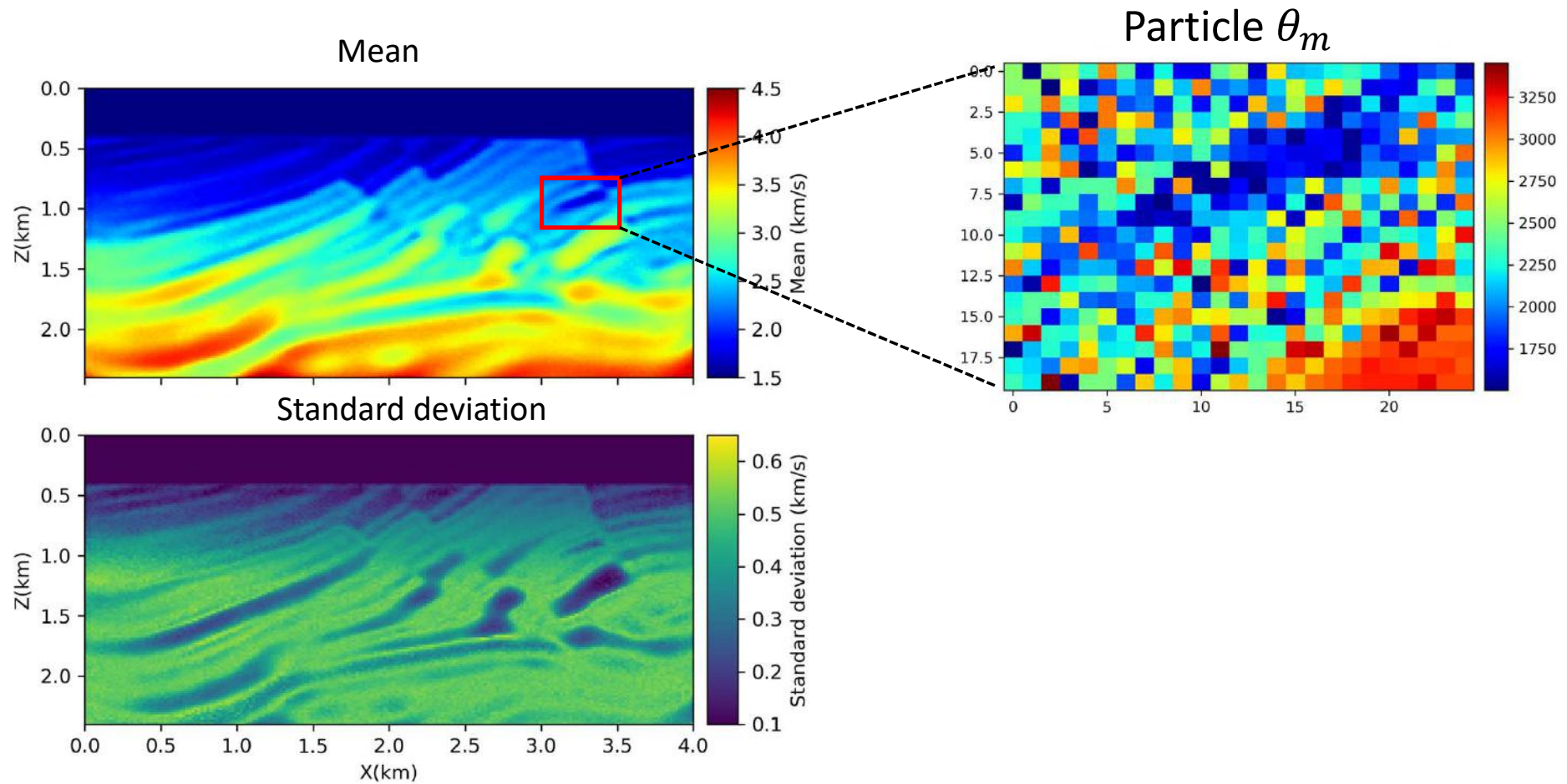




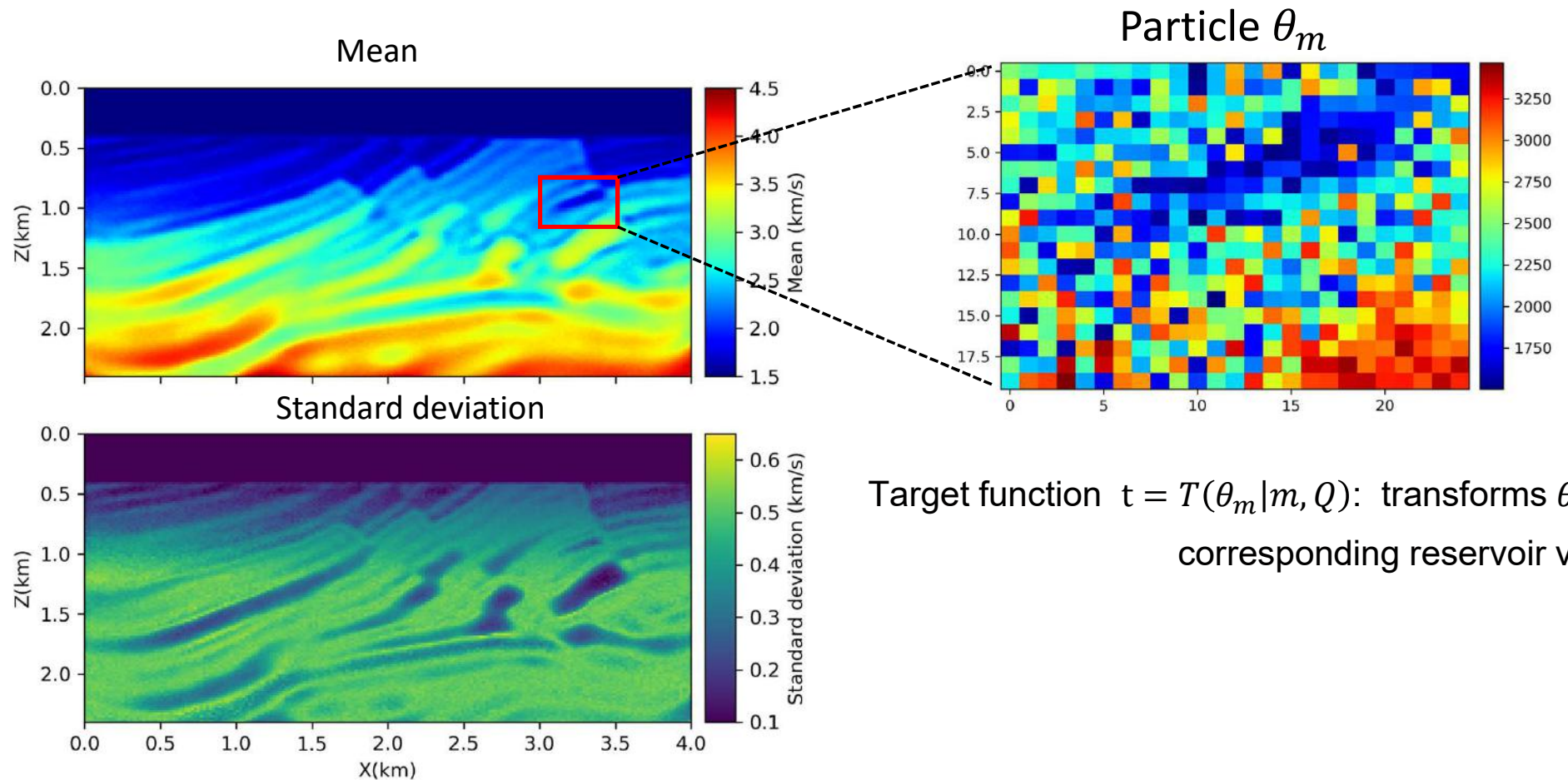
# What is the size of a reservoir?



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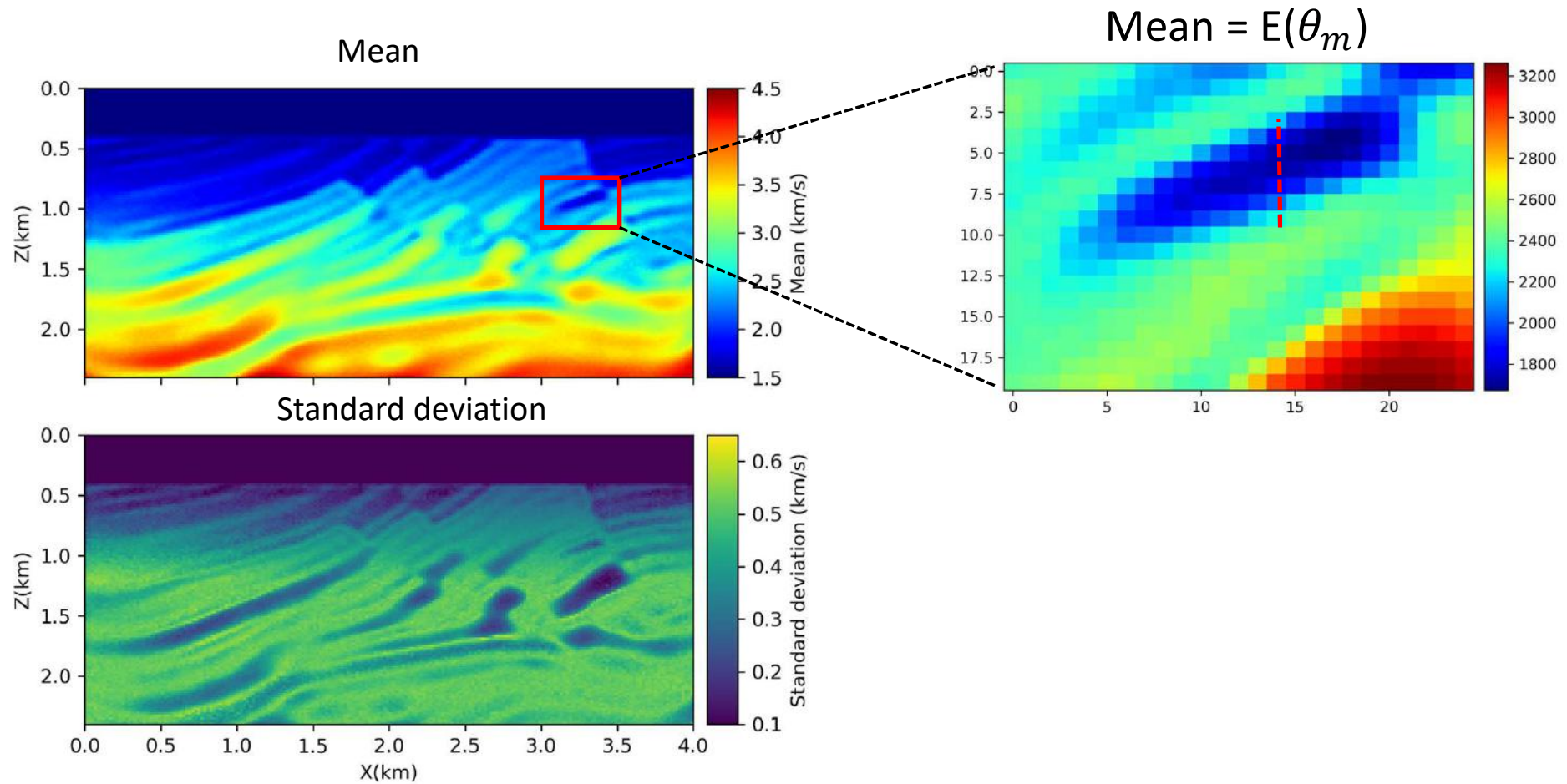
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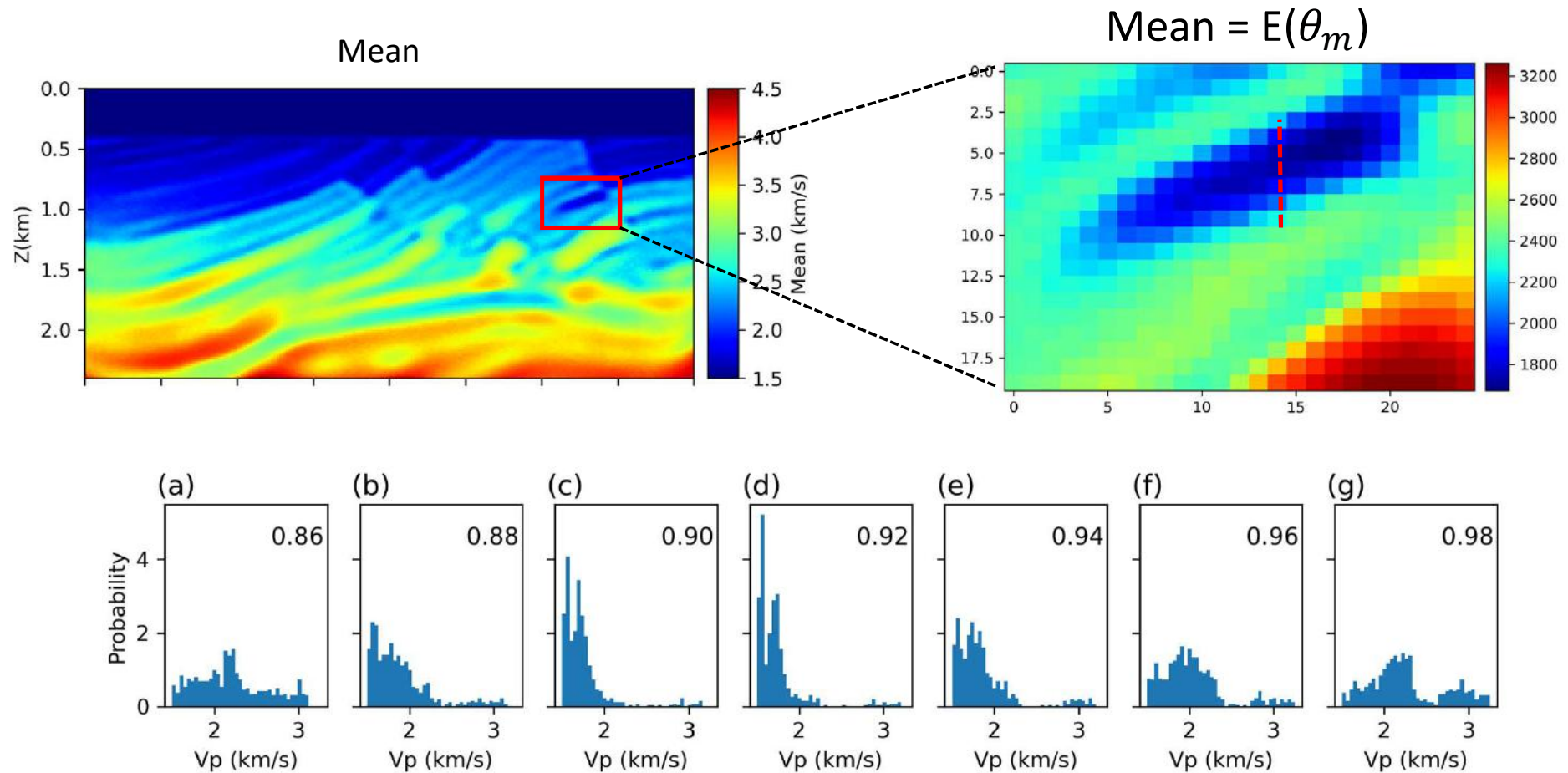
Target function  $t = T(\theta_m|m, Q)$ : transforms  $\theta_m$  into corresponding reservoir volume



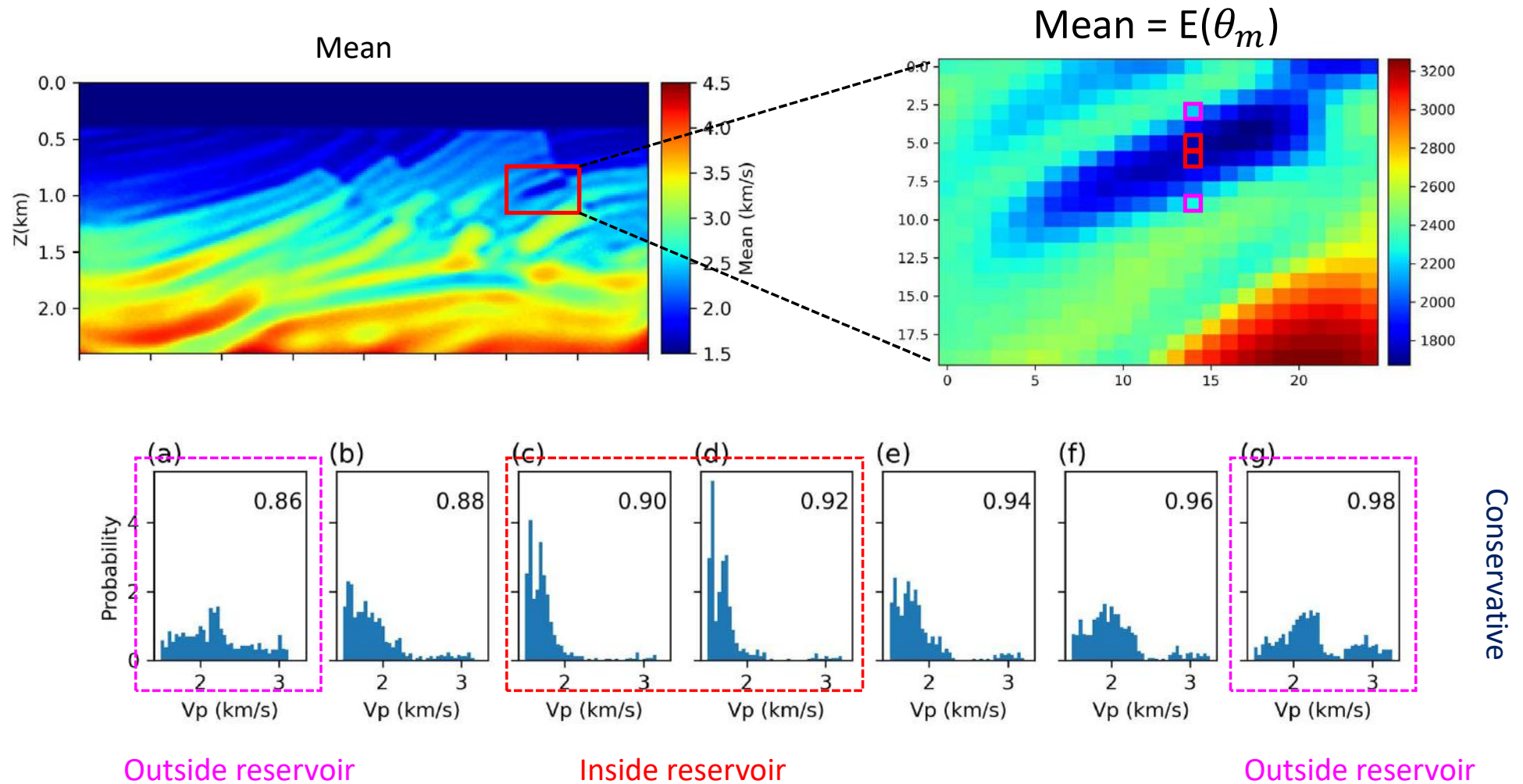
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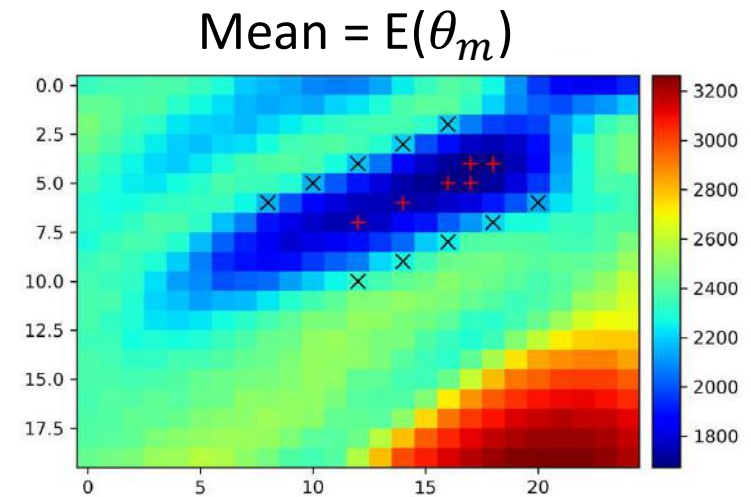
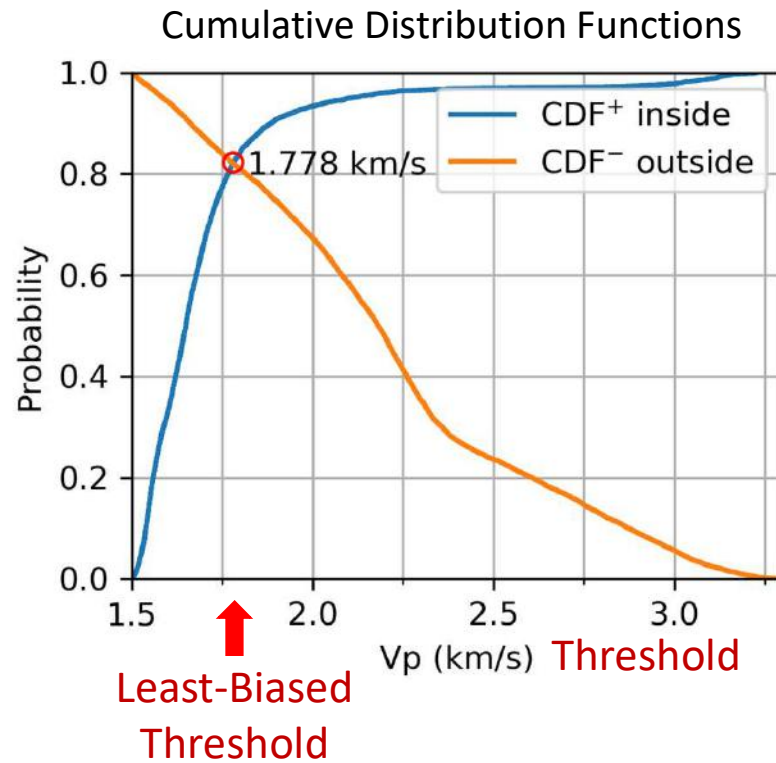
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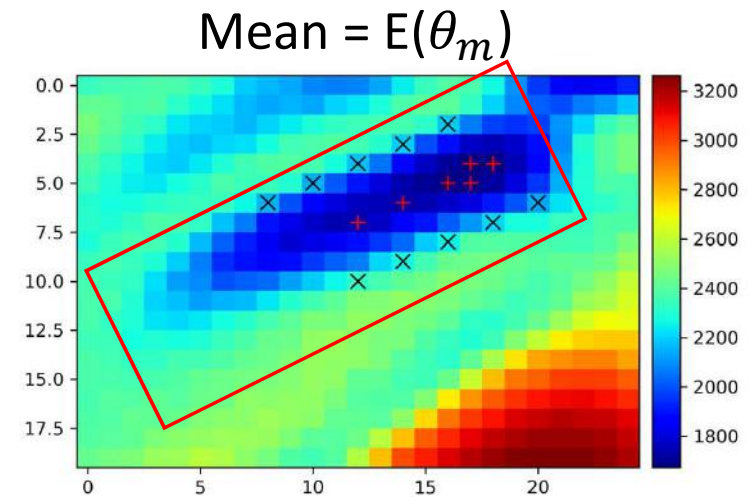
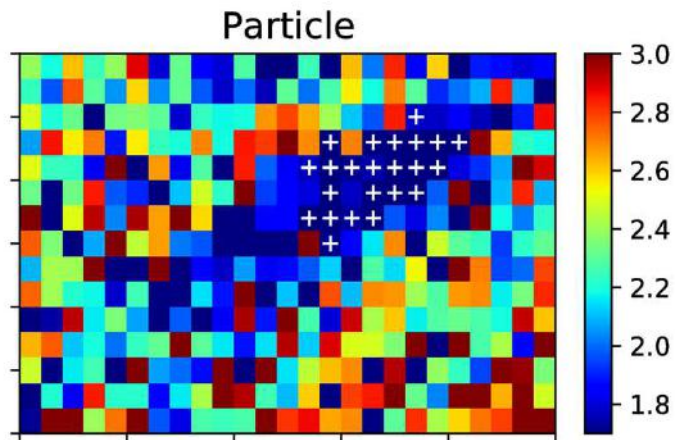
*Zhang & Curtis (2022), Geophys. J. Int.*  
*Zhao et al., (2022), J. Geophys. Res.*

$$a^* = \sum_{m \in \mathbb{M}} \int_{\theta_m} T(\theta_m | m) p(\theta_m, m | y_d, d) d\theta_m$$

$T$  = size of the largest continuous body  
 in the model with velocity < 1.778 km/s



# What is the size of a reservoir?



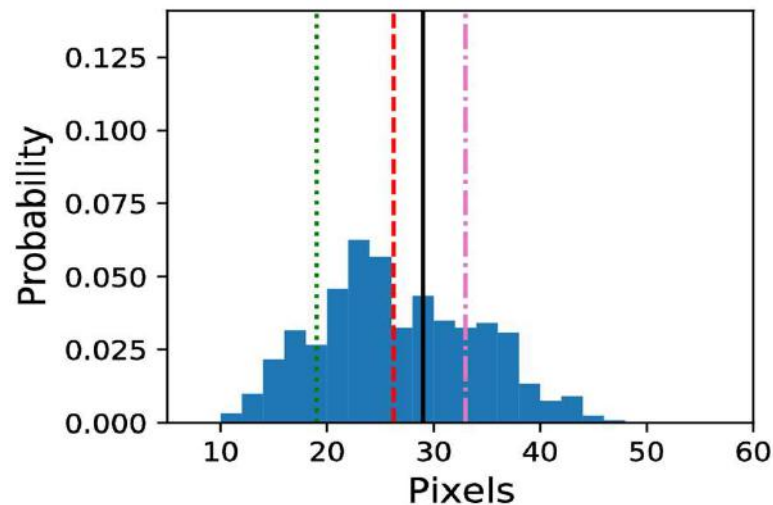
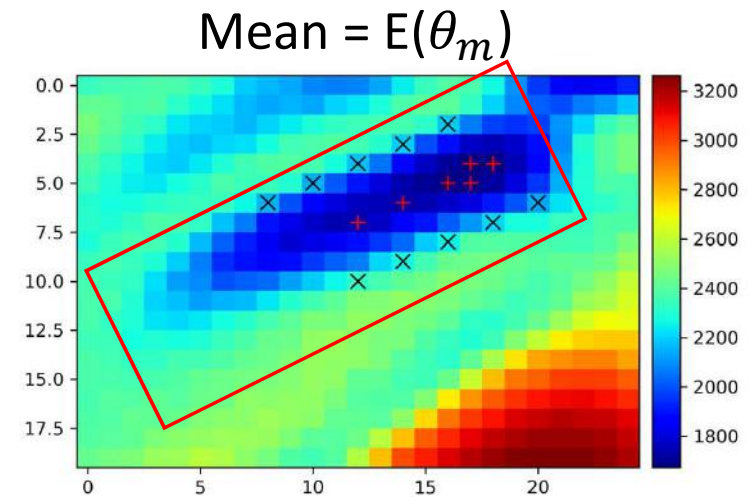
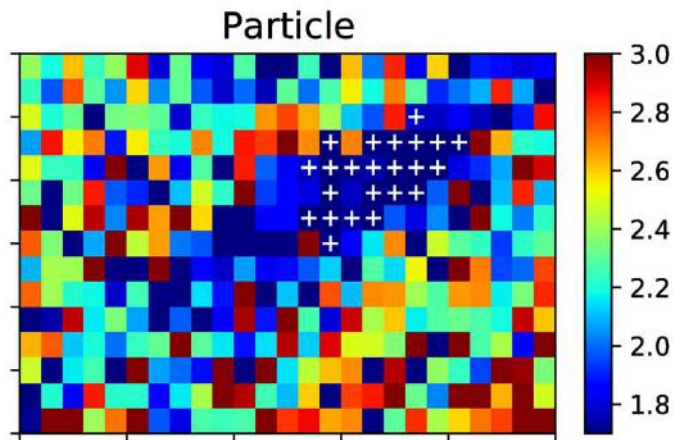
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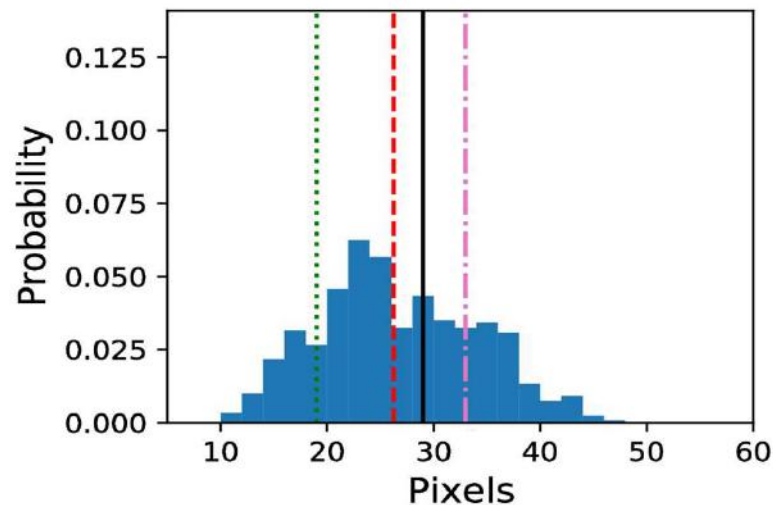
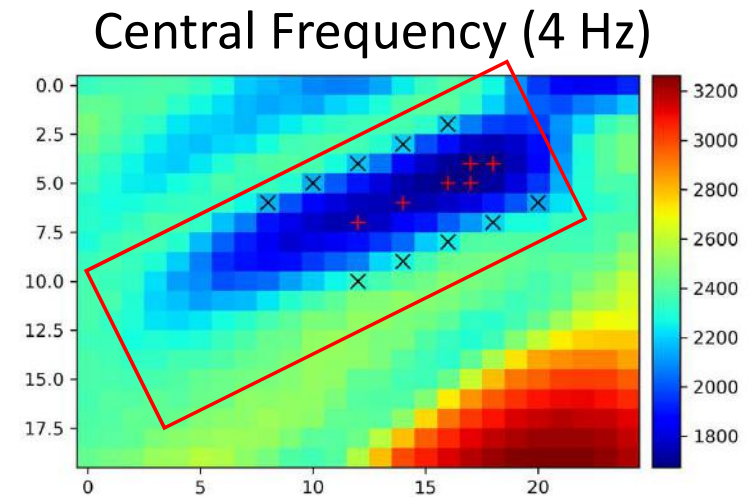
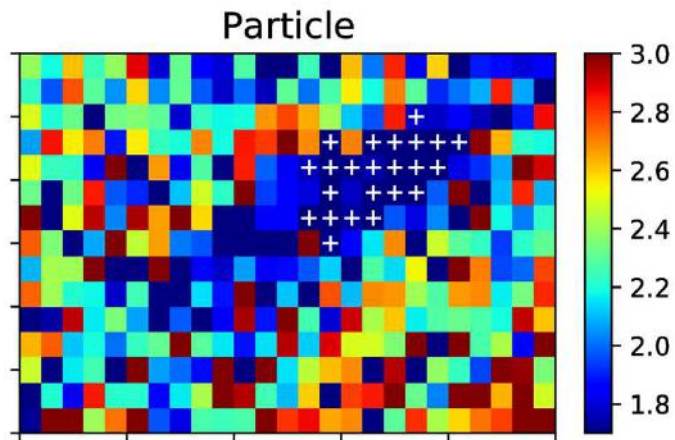


# What is the size of a reservoir?



- True
- - - Optimal answer from Interrogation Theory
- ... Answer obtained by interpreting mean structure
- . - Answer obtained by interpreting median structure

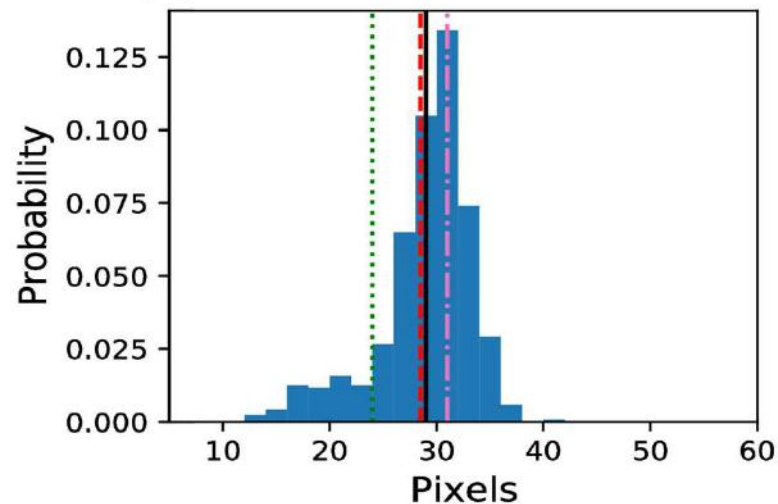
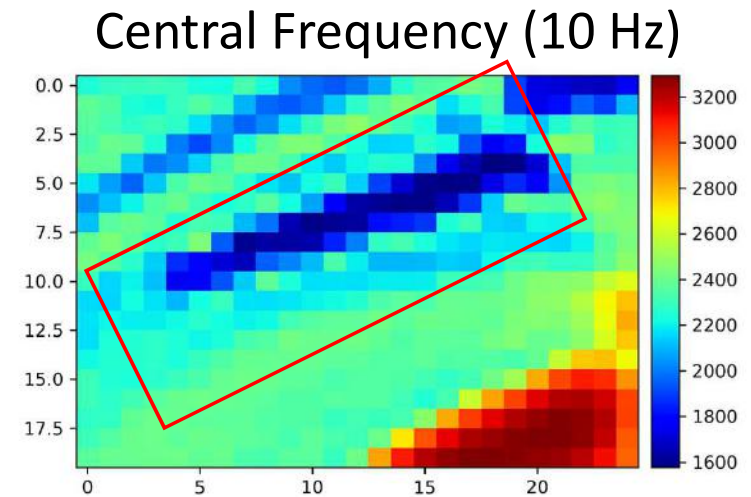
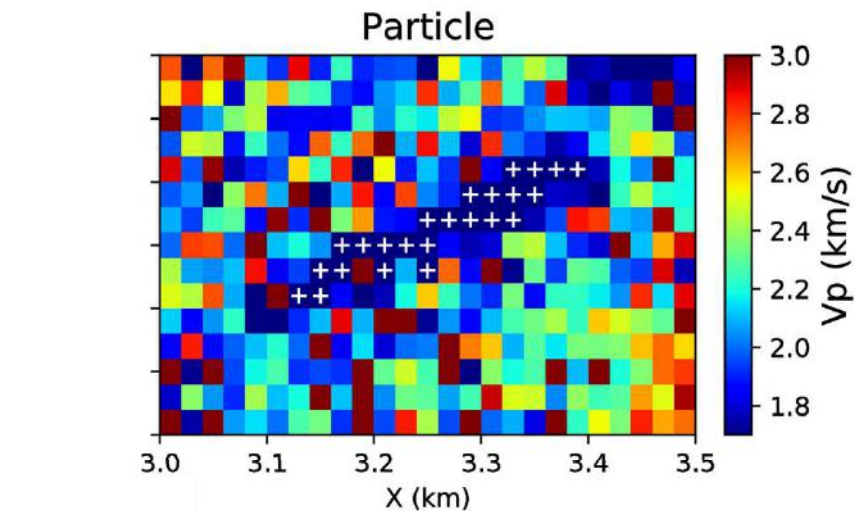
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Increase the frequency...

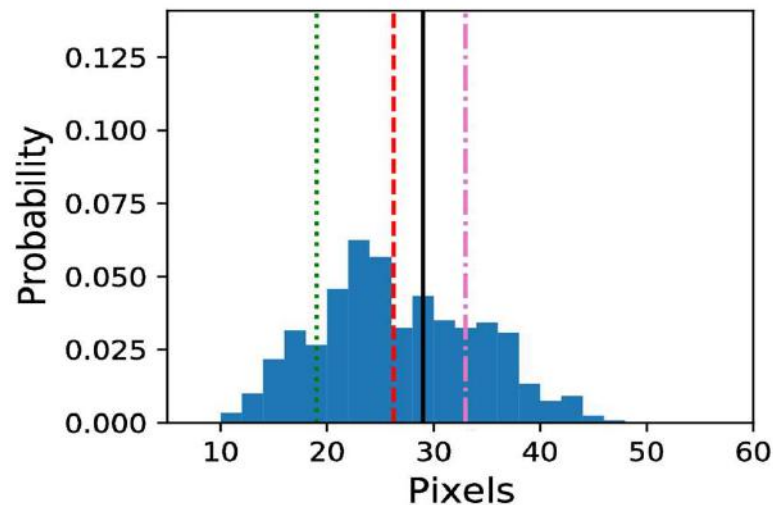
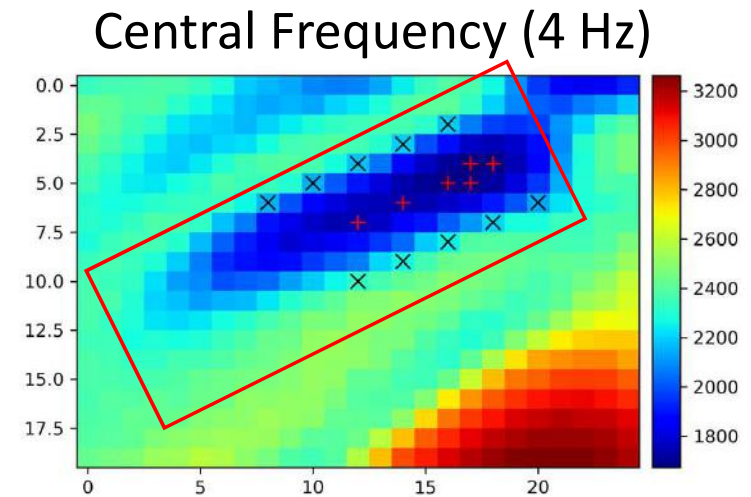
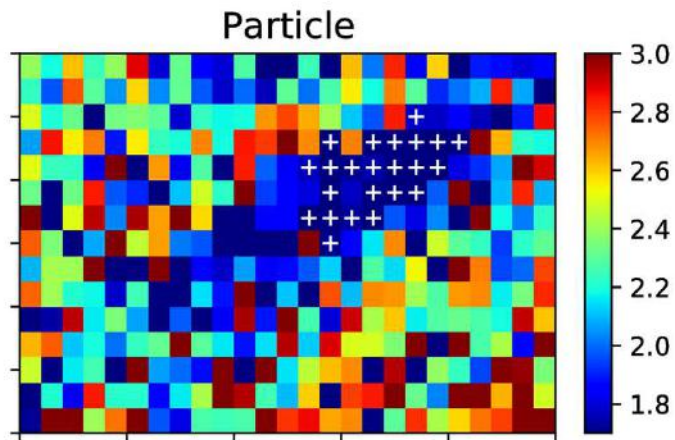
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Interrogation Theory provides least biased estimates from real data and allows Value of Information estimates for specific Questions

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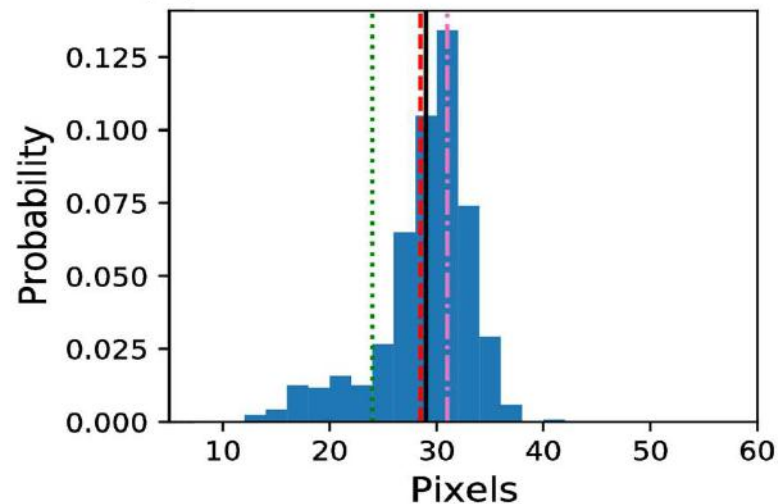
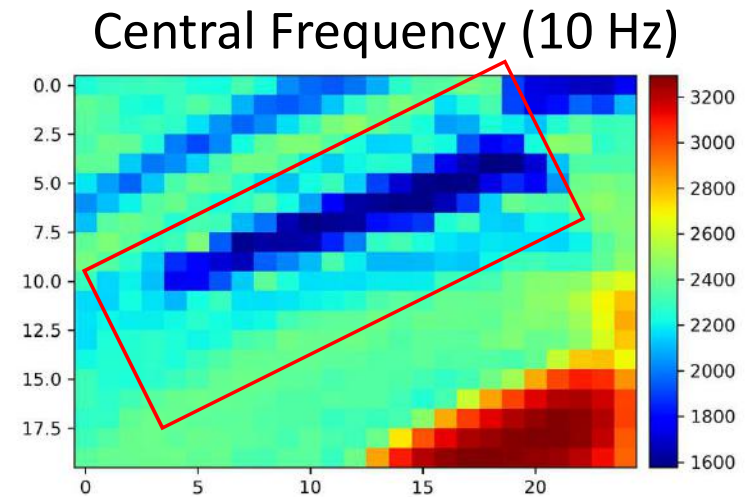
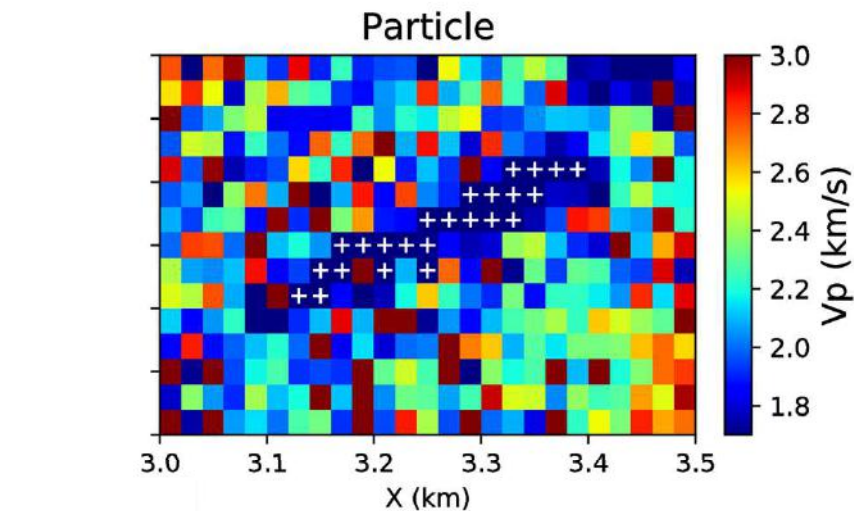
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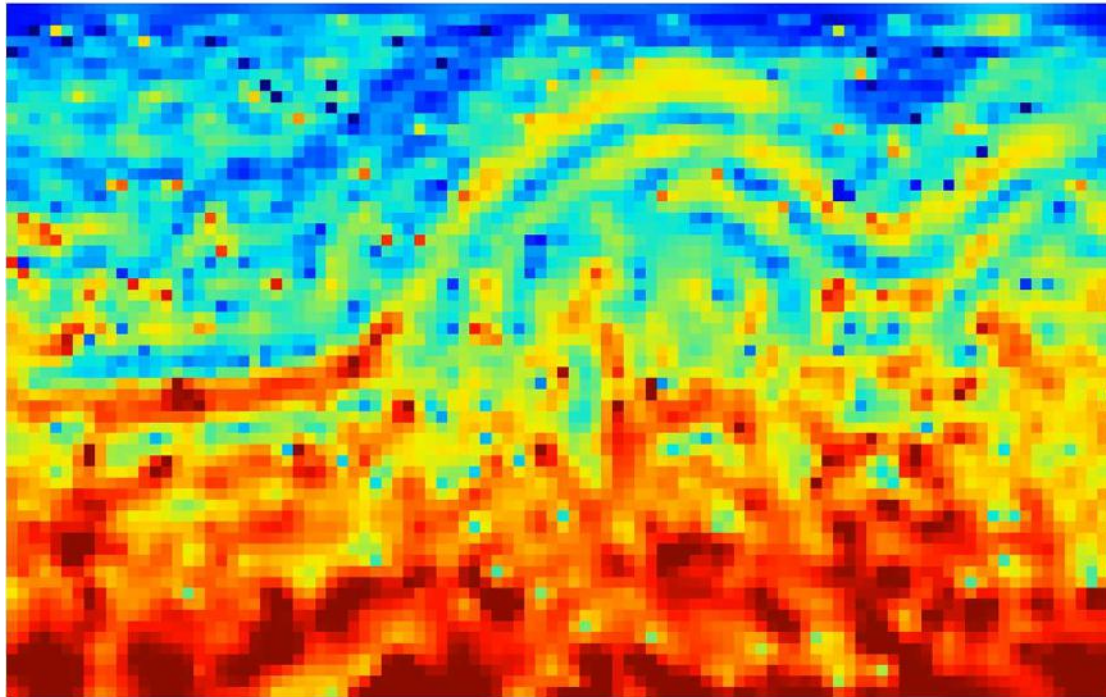
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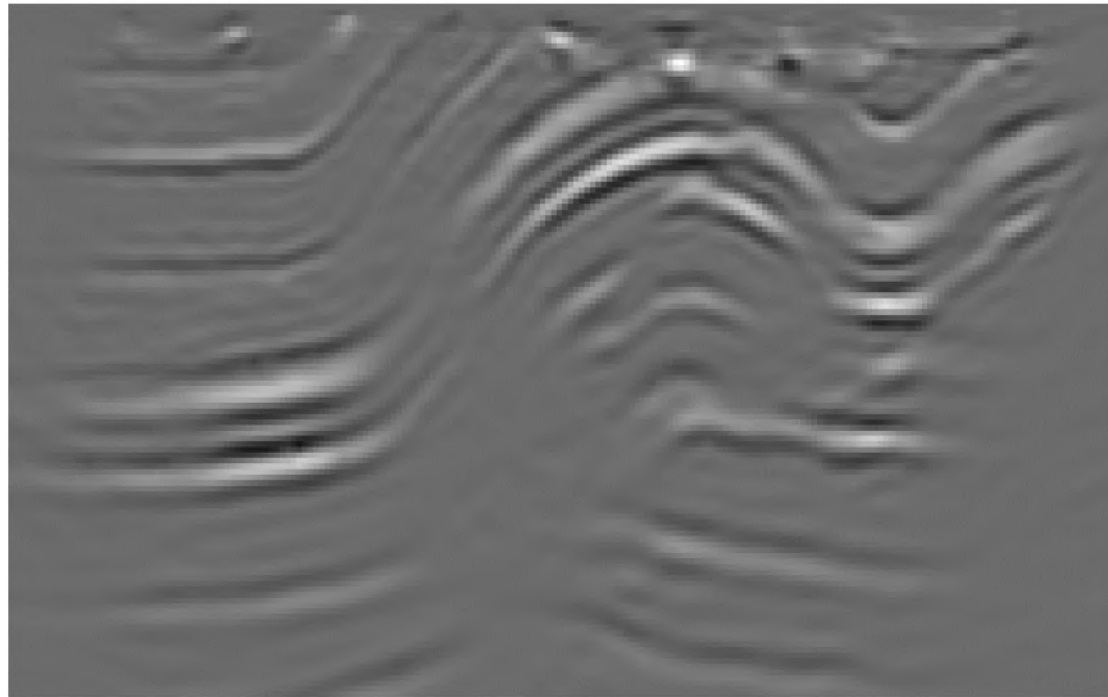
# Alternative Target Functions

Movie of all particles



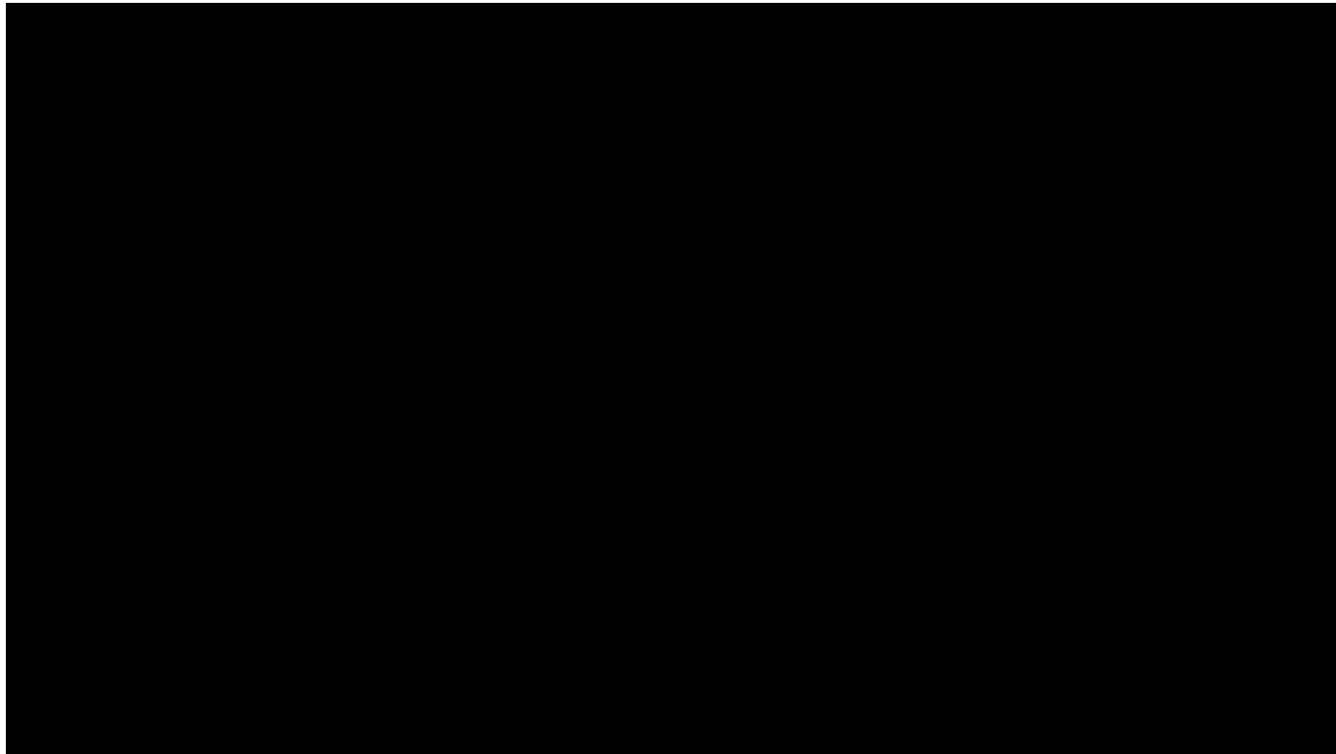
# Alternative Target Functions

Movie of migration images of all particles



# Alternative Target Functions

Interpret migration movie



$$a^* = \sum_{m \in \mathbb{M}} \int_{\theta_m} T(\theta_m | m) p(\theta_m, m | y_d, d) d\theta_m$$



# Interim Discussion: Extensions to 3D & 4D

- Need models that represent 3D-world  $p(\mathbf{m} | \mathbf{d})$
- Previous movies were 2D slices from a 3D test model
  - Required 3D Bayesian *variational* FWI
  - Only algorithm that may work: stochastic Stein variational gradient descent (sSVGD)  
See: **Zhang et al., 2022**: “3D Bayesian variational Full Waveform Inversion” (*arXiv / Geophys. J. Int., in press*)
- 3D Interrogation using FWI is computationally feasible, for small data sets
- Currently working on 4D / repeated-seismic-survey example
  - For energy transition applications, data sets will be small / sparse / focussed
  - Compared to cost of acquisition, £-cost of computation remains ‘small’ but **CO<sub>2</sub>**-cost increasing

# Interrogation Theory

- What is the volume of a particular subsurface reservoir? →  $X \text{ km}^3$
- Has more than  $1 \text{ Kt}$  of  $\text{CO}_2$  escaped from the subsurface store? → *Yes/No*
- Which model best explains the *true* distribution of seismicity? → *Model 1,2,3...*

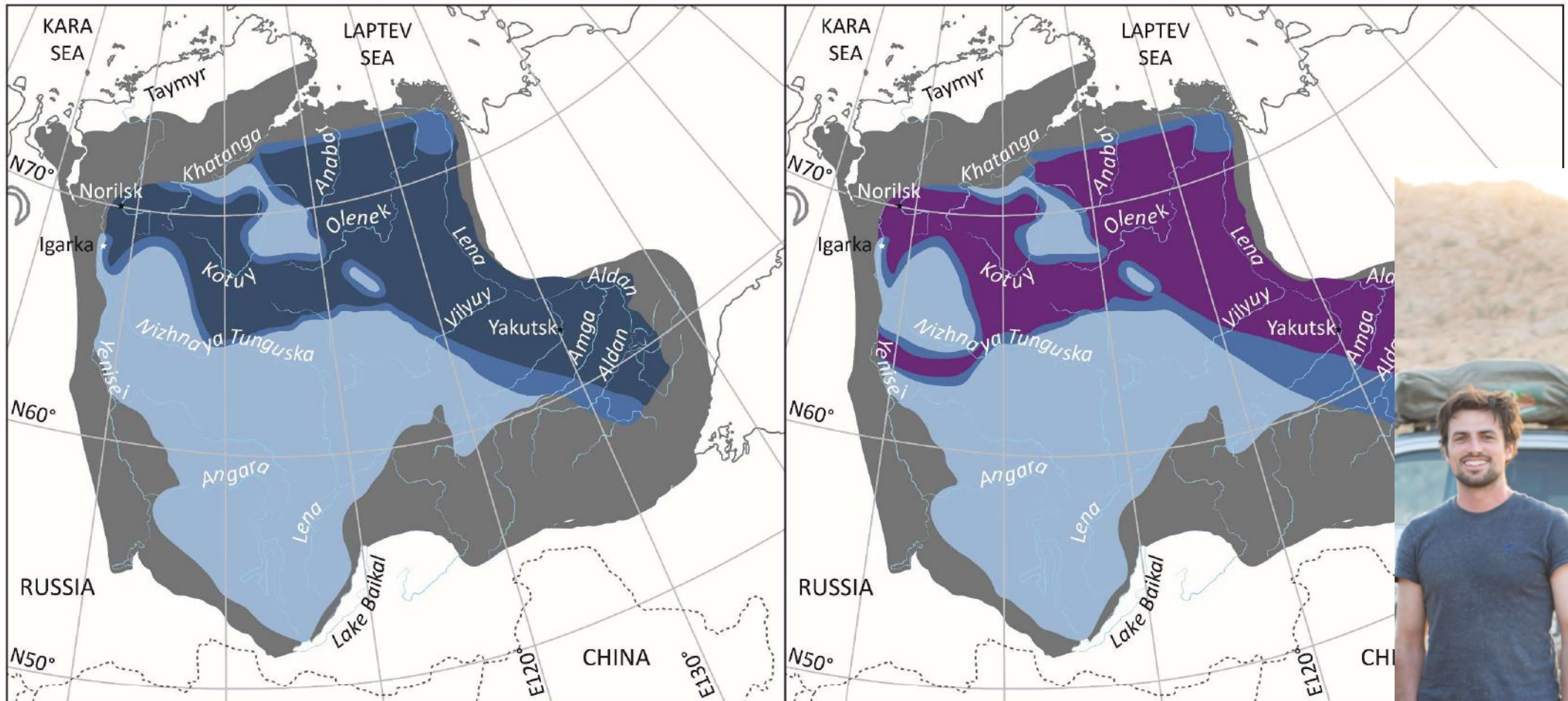
In principle, all are answerable using similar quantitative models and methods

➔ However, questions that affect the world more broadly may be different in nature...

## Narrative

LOWER CAMBRIAN (ca. 529 - 514.5 Ma)

MIDDLE CAMBRIAN (ca. 514.5 - 504.5 Ma)



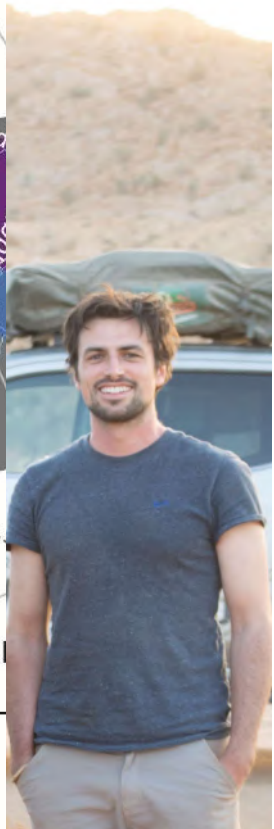
Palaeofacies (Sukhov et al., 2021):

- Siberian craton
- Saliniferous facies
- Transitional facies
- Open marine carbonate facies
- Organic-rich carbonate and shale

**Siberia in the Cambrian:**

Two theories explain the distribution of oxygen: **Theory A & Theory B**

**Everyone please choose which theory they think is most likely to be true.**



**Theory A**   or   **Theory B**

Which is most likely to be true?

**Theory A**      or      **Theory B**

Which is more **emotionally** satisfying,  
**fulfilling** or appealing, or gives more  
**positive feelings** of any kind?  
Which one do you **like**?

**Narrative**

Raise a hand if the **same hand** was both  
more likely to be true,  
and most emotionally satisfying

*i.e., if you squeezed the same hand twice*

# Interrogation Theory

- What is the volume of a particular subsurface reservoir? →  $X \text{ km}^3$
- Has more than *1 Kt* of  $\text{CO}_2$  escaped from the subsurface store? → *Yes/No*
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# Interrogation Theory

- What is the volume of a particular subsurface reservoir? →  $X \text{ km}^3$
- **Should we buy a license to store  $\text{CO}_2$ ?** → *Yes/No*
- **Should a business enter a new sector: storing Hydrogen or  $\text{CO}_2$ ?** → *Yes/No*

The questions lie outside of 'model world' – they lie inextricably in the real world

- Will there **be** a Hydrogen storage market?
- What will future interest, tax and discount rates be over next 10 years?
- How rapidly can/will society's reliance on hydrocarbons abate?
- Will the most experienced staff, key investors, or voters be inspired, or concerned?
- **How will my decision change the world itself?**



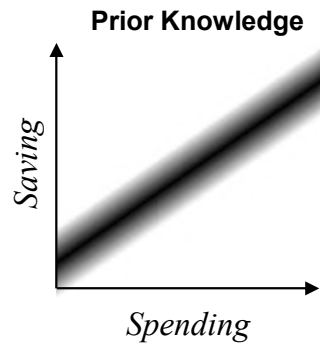
## → Escape from Model World (Erica Thompson, 2022)

- What is the volume of a particular subsurface reservoir? →  $X \text{ km}^3$
- **Should we buy a license to store CO<sub>2</sub>?** → *Yes/No*
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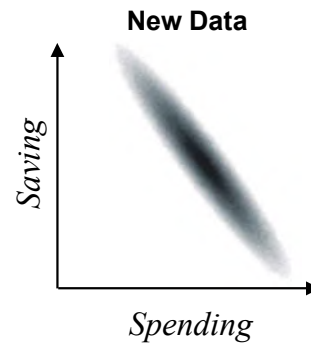
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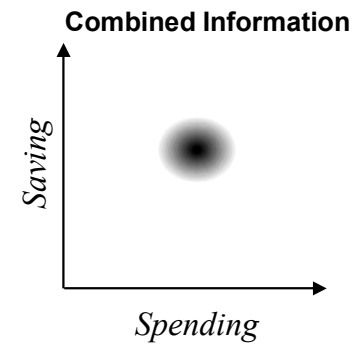
Bayes' Theorem



$\times$



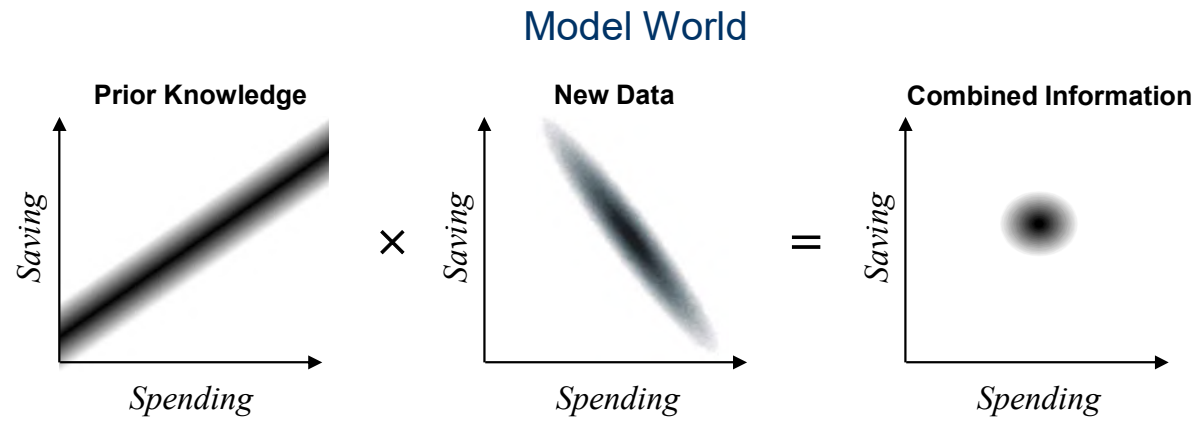
$=$



Real World

Credit Card  
Interest Rates  
↓ 28% APR

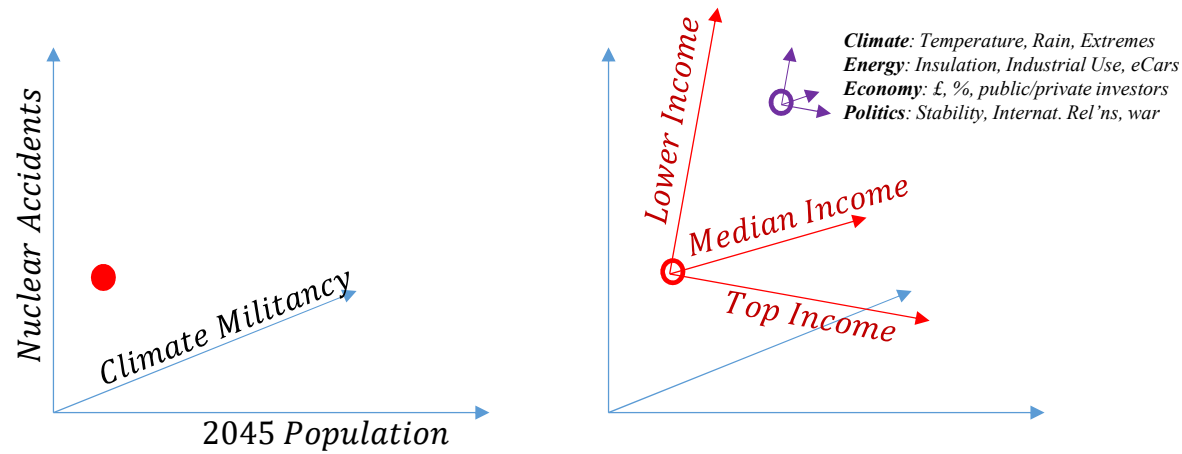
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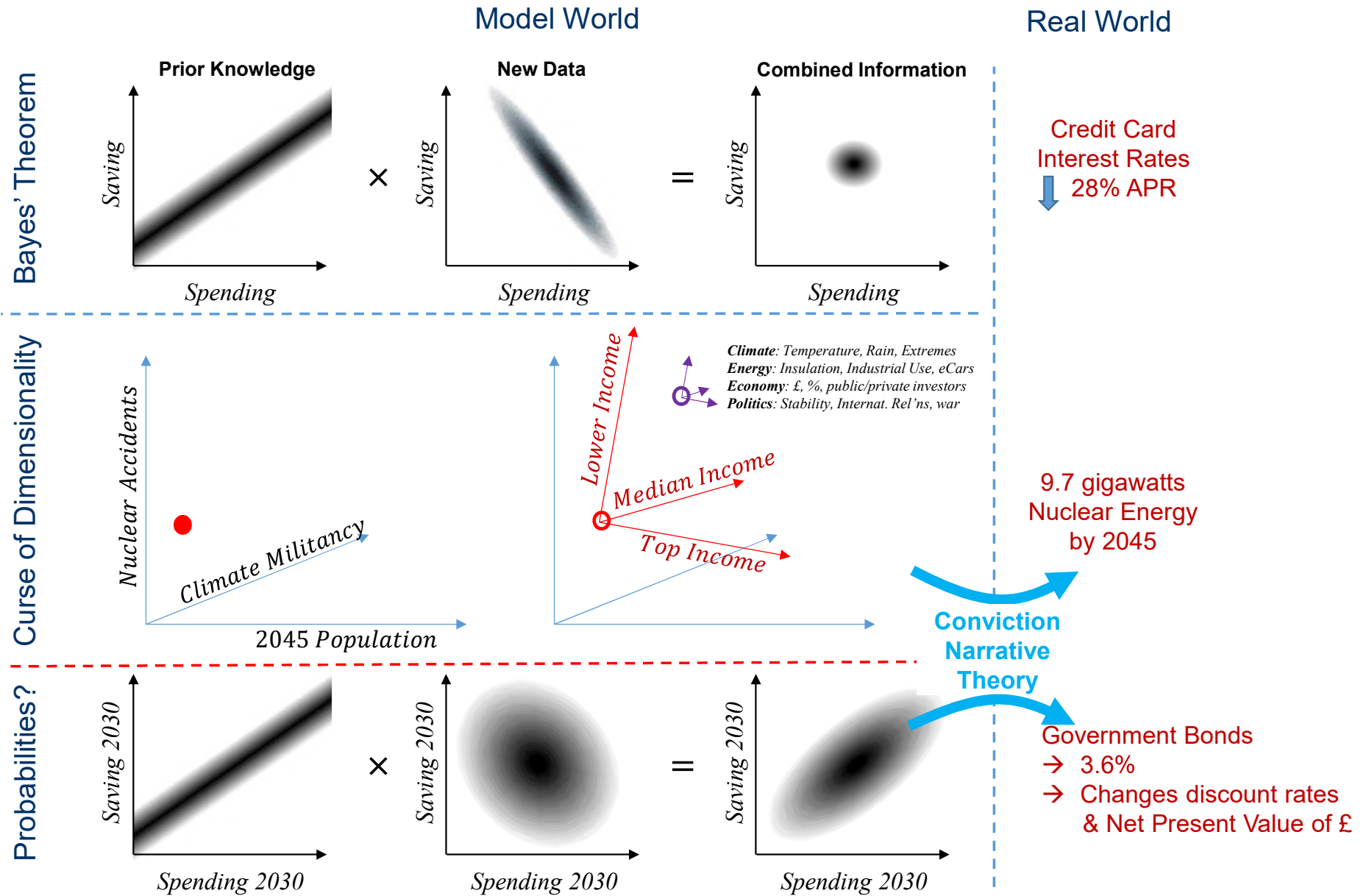
Real World

Credit Card  
Interest Rates  
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Curse of Dimensionality



9.7 gigawatts  
Nuclear Energy  
by 2045



*Johnson et al., (2022)*  
*Psychologists, Decision Theorists,*  
*Management Scientists*

# Conviction Narrative Theory

Humans don't explore and make optimal decisions

- Under radical uncertainty, people tend to explore few options (even only 1).  
**They make decisions based on *narratives*.**
- **Narratives:** mental representations that summarise relevant causal, temporal, analogical or valence information. (In short: *stories*)

A narrative is selected which *explains past/present* data, is then used to *imagine possible futures*; **emotional** reactions to the narrative provide *affective evaluation*.

→ Choose narrative that produces *conviction to take sustained action*.

“Emotions lead to biases” – Kahnemann & Tversky + many – “rational thought is evidence based”

➔ **“Emotional responses encode long-term stores of rationality”**

# Conviction Narrative Theory

*Johnson et al., (2022)*  
*Psychologists, Decision Theorists,*  
*Management Scientists*

## Matters arising:

- Conviction Narrative Theory describes what decision-makers **do**, not necessarily what they **should** do.
- This is a very different way to view '*rationality*' in decision making.
- What people do as a result may be rational in some cases – but how often?
- *What else could they do? (How did we do?)*
- If that is how decision-makers make decisions, **what should we do?**
- There is a need to develop new theory and methods that **span both worlds**



# Thank you!

All of our papers are available at: <https://blogs.ed.ac.uk/curtis/publications>

Arnold & Curtis (2018):	<i>"Interrogation Theory"</i> , Geophys. J. Int.
Zhang & Curtis (2022):	<i>"Interrogating probabilistic inversion results for subsurface structural information"</i> , Geophys. J. Int.
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