NORWEGIAN CCS RESEARCH CENTRE

Utilizing compressive sensing techniques to reduce geophysical monitoring costs at CO₂ injection sites

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Aims

- Investigate sparse geophysical monitoring techniques that allow assessments of site conformance to be undertaken
- Make use of compressive sensing algorithms to enhance sparse seismic data
- Use analytical solutions to determine likely range of CO₂ migration and set bounds for monitoring
- Assess how plume modelling can guide the use of compressive sensing in data acquisition and interpretation
- Use models of the Smeaheia storage site to test these approaches





Why Smeaheia?

- Potential storage site for Northern Lights full-chain CCS project
- Expected capacity of > 100 million tonnes
- Excellent scale-up potential
- Offshore Norway, adjacent to Troll Field, so local knowledge and infastructure
- Significant data availability and international interest in site





Smeaheia

- Potential storage in Jurassic Sognefjord Formation
- 1200m 1500 m depth with high porosity and permeability
- Barosita distoiguftionsNamedvægiðeityell 32/4-1
 pleonvingsthievSegatedjærldipasthorveckvingplug
 hheadstere Froentations





Analytical solutions

- Represent a buoyant fluid spreading radially beneath a sealing caprock
- The solutions require several assumptions
 - no capillary pressure, no relative permeability effects, no viscosity differences
- Compute the radius, r(t), and height, h(r,t), of a spreading CO2 layer as a function of time, t
- A release with a constant flux, Q







Assessing end members

 Analytical solutions and data from well/core plug analysis are used to define the extent of the monitoring program





Smeaheia reservoir model





• Then, generate multiple realisations of reservoir model from property distributions



Smeaheia model construction





Smeaheia modelling

- CO₂ trapped under topography of seal
- Multiple realisations of forward flow
- Here, distribution is fairly radial
- Seiondistinot (CO₂i hgyessafotan array contrespondencie jectionaepropertmeability zone between Upper and Lower Sogefjord Formation





Seismic modelling

- Baseline and repeat data generated.
- Difference data allows extent of plume to be mapped
- Attribute analysis possible on synthetic data
- Site conformance can be assessed along series of 2D profiles





Sparse monitoring - seismic







Sparse monitoring – OBS system







Sparse monitoring – reduce receivers







Sparse monitoring – random distributions







Compressive sensing

- Compressive Sensing (CS) is a signal processing technique that exploits sparsity inherent within a signal to fully recover that signal, using fewer measurements than are required by the Nyquist-Shannon sampling theorem
- We investigate the potential of CS to reduce the cost of monitoring a CCS site using a fixed Ocean Bottom Seismometer (OBS) array
- There are three basic requirements for compressive sensing to be successful (Herrmann et al., 2011): a random sampling scheme, an appropriate transform domain in which the complete signal has a sparse representation and a sparsitypromoting recovery algorithm





Compressive sensing

- Signal is removed through limited receiver stations – pre-stack interpolation
- OBS system used:
 - Improved repeatability.
 - Shear wave information, useful for lithology and fracture characterisation. Shear wave data also has the potential to discriminate fluid and pressure effects.
 - Wide azimuths and longer offsets. Long offsets are desirable for seismic inversion studies.
 - Improved multiple attenuation.





Compressive sensing

- Assess different sampling schemes:
 - (a) original data;

NCOS

- (b) data reconstructed from uniformly random sample;
- (c) data reconstructed from jittered under sampling;
- (d) data reconstructed from random piecewise random sampling

• Demonstrates the effect of gaps (indicated by arrows) in a random sampling scheme







Quantify difference between datasets

NRMS is the summation of the difference between two traces (a_t and b_t) in a specified time window divided by the average RMS amplitude of the two input traces:

•
$$NRMS = \frac{200 \times RMS(a_t - b_t)}{RMS(a_t) + RMS(b_t)}$$

• Predictability is the summed squared cross-correlation of two traces (ϕ_{ab}) in a time window, divided by the summed product of the trace auto-correlations $(\phi_{aa} \& \phi_{bb})$.

•
$$PRED = \frac{100 \times \sum \phi_{ab}(\tau) \times \sum \phi_{ab}(\tau)}{\sum \phi_{aa}(\tau) \times \sum \phi_{bb}(\tau)}$$

Quantitative assessment

- The top third of the seismic section shows the lowest repeatability (%NRMS ~120% and %predictability of c. 60%)
- At the level of the CO₂ plume, repeatability is generally very good (%NRMS <20% and %predictability of >90%).





Conclusions

- Compressive sensing techniques are suitable for pre-stack interpolation of monitoring data
- The high amplitude seismic response, following CO₂ fluid substitution, enhances capability
- Measures of repeatability are high
- Testing on real data is underway





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Industry-driven innovation for fast-track CCS deployment





Conformance - synthetic example

 Use of data for conformance assessment – injected CO₂ must stay within regulatory bounds











Conformance: History matching

- Incorporate data measured during CO₂ injection to update model realizations
 - Ensemble-based data assimilation methods





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Survey characteristics

Varying time of acquisition

 $t_{survey} = \{300, 600, 900, 1200, 1500\}$ days

