

The first-order finite volume (FV) is the default option in many standard reservoir simulators, both commercial and open-source.

- + It is robust.
- + Easy implementation.
- Suffers from numerical diffusion.
- Incorrect computations of the front position, components concentrations, water breakthrough, etc.

To reduce the numerical diffusion and increase the accuracy, there are mainly two options:

- ▶ to refine the grid,
- ▶ to increase the order of the numerical method.

Content of the presentation

1. Second-order method with linear programming reconstruction
2. Explore the method's capabilities in a realistic setting.
 - ▶ accuracy
 - ▶ verification in the absence of "true" solution
3. Run WAG and CO2 injection scenarios on the realistic test cases:
 - ▶ a medium-sized realistic reservoir with an unstructured corner point grid
 - ▶ an openly available Norne field model

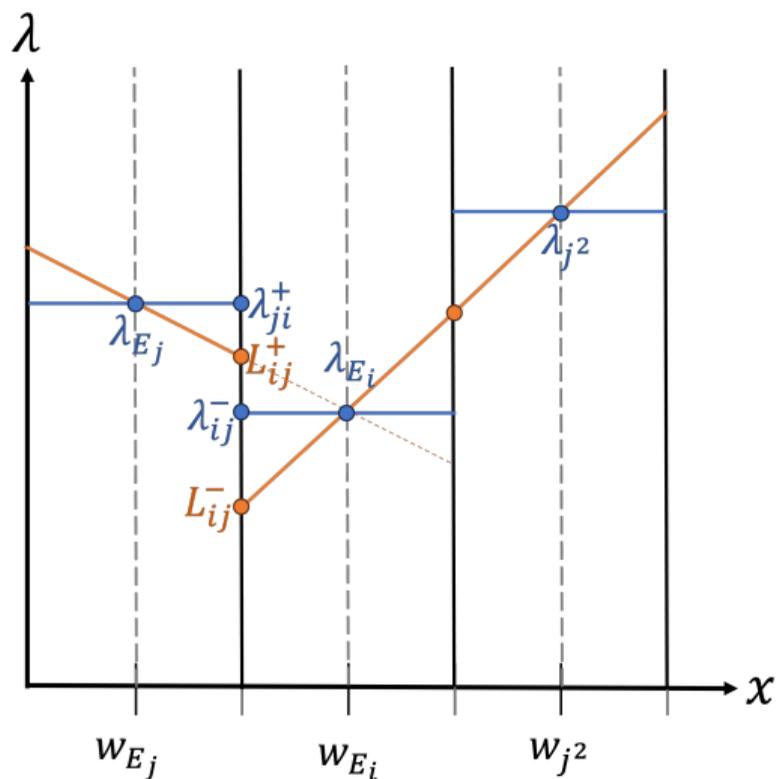
Build and run

The models and the build instructions are available in the repository <https://github.com/kvashchuka/second-order-opm-tests>.

To run a test case with the second-order method, you need to enable certain flags:

```
./*path_to_the_build_folder_of_opm-simulators*/bin/flow CASE_NAME --enable-higher-order=1 --enable-local-reconstruction=1 --reconstruction-scheme-id=3 --only-reconstruction-for-solvent-or-polymer=false
```

First- vs Second-Order FV Method



First-order FV method

$$\lambda_w^{ij} = \begin{cases} \lambda_{ij}^-, & \text{if } (\nabla p_w^{n+1} - \rho_w \mathbf{g}) \cdot \mathbf{n} \geq 0, \\ \lambda_{ij}^+, & \text{otherwise,} \end{cases}$$

Second-order FV method

$$\lambda_w^{ij} = \begin{cases} L_{ij}^-, & \text{if } (\nabla p_w^{n+1} - \rho_w \mathbf{g}) \cdot \mathbf{n} \geq 0, \\ L_{ij}^+, & \text{otherwise,} \end{cases}$$

where the linear reconstruction function has to satisfy the following requirements:

$$L_{E_i}(x) := \lambda_{E_i} + \nabla L_{E_i} \cdot (x - \mathbf{w}_{E_i}),$$

$$L_{E_i}(\mathbf{w}_{E_j}) = \lambda_{E_j}, \forall (E_i, E_j) \in \partial E_i.$$

Second-order method with Linear Programming reconstruction

We want to minimize the total gaps between the reconstructed values and the cell-averaged values at all neighboring cells:

$$\delta(L) := \sum_{\forall(E_i, E_j) \in \partial E_i} |\lambda_{E_j} - L_{E_i}(\mathbf{w}_{E_j})|. \quad (1)$$

The constraints are the following monotonicity conditions:

$$\min\{\lambda_{E_i}, \lambda_{E_j}\} \leq L_{E_i}(\mathbf{w}_{E_j}) \leq \max\{\lambda_{E_i}, \lambda_{E_j}\}, \quad \forall(E_i, E_j) \in \partial E_i. \quad (2)$$

Second-order method with Linear Programming reconstruction

We solve the following LP problem:

$$\begin{aligned} \max \quad & \sum_{\forall (E_i, E_j) \in \partial E_i} \text{sgn}(v_{E_j})(\mathbf{w}_{E_j} - \mathbf{w}_{E_i}) \cdot \nabla L_{E_i} \\ \text{subject to} \quad & v_{E_j}^- \leq (\mathbf{w}_{E_j} - \mathbf{w}_{E_i}) \cdot \nabla L_{E_i} \leq v_{E_j}^+, \end{aligned} \quad (3)$$

where

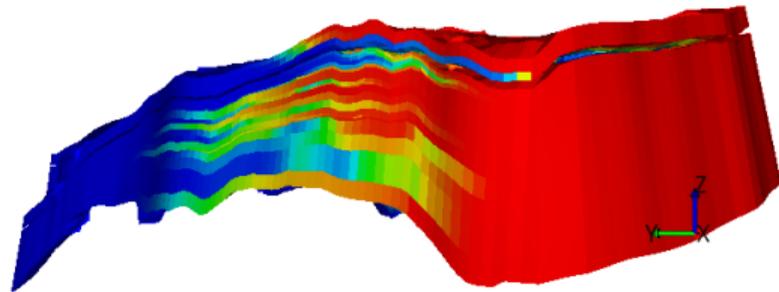
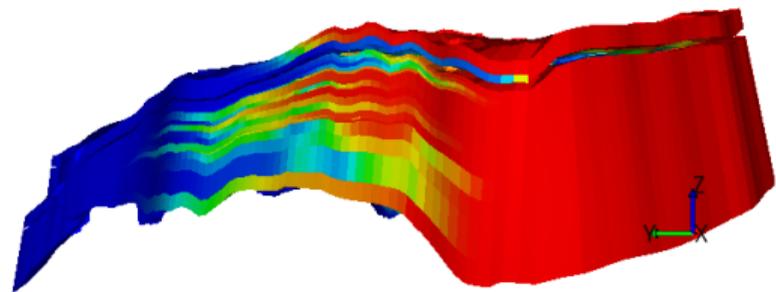
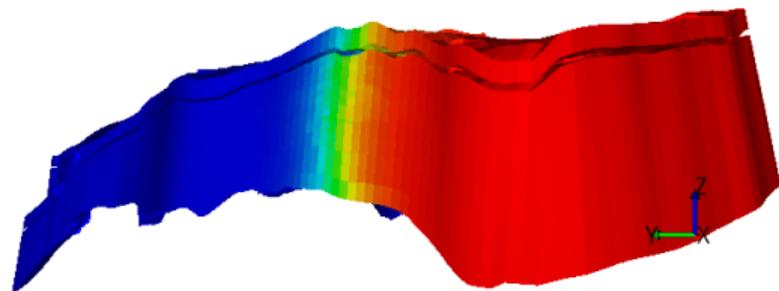
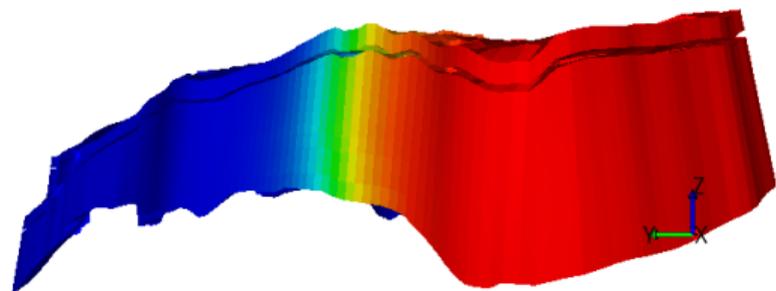
$$\begin{aligned} v_{E_j}^- &= \min\{0, \lambda_{E_i} - \lambda_{E_j}\}, \\ v_{E_j}^+ &= \max\{0, \lambda_{E_i} - \lambda_{E_j}\}. \end{aligned} \quad (4)$$

The unknown vector x is the gradient of the linear reconstruction

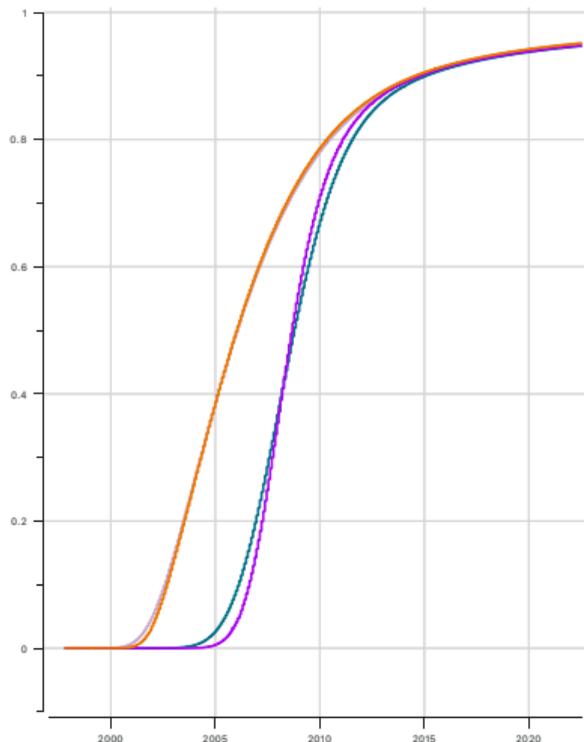
$$x = [\nabla L_{E_i}^x, \nabla L_{E_i}^y, \nabla L_{E_i}^z]^T.$$

We use an all-inequality simplex method to solve the LP.

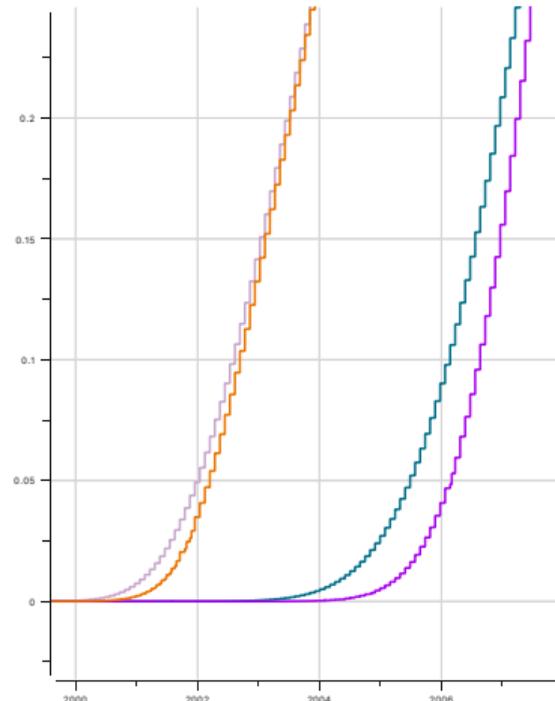
Norne: homogeneous (top) and heterogeneous (bottom)



Norne: homogeneous and heterogeneous

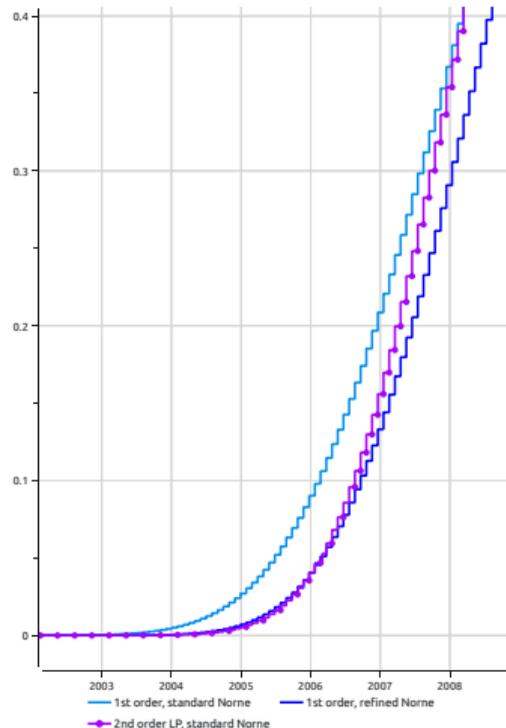
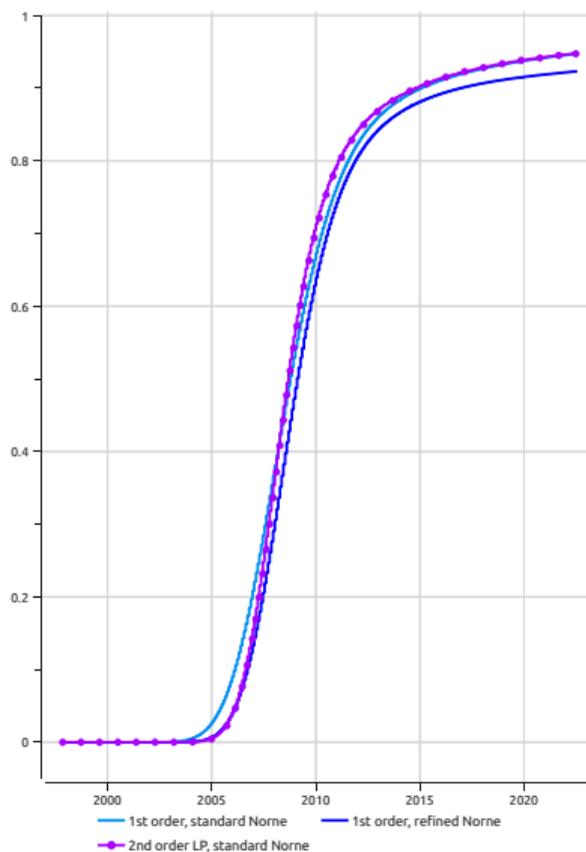


- 1st order, homogeneous Norne model
- 1st order, heterogeneous Norne model
- 2nd order LP, homogeneous Norne model
- 2nd order LP, heterogeneous Norne model



- 1st order, homogeneous Norne model
- 1st order, heterogeneous Norne model
- 2nd order LP, homogeneous Norne model
- 2nd order LP, heterogeneous Norne model

Norne: verification with refined model



CO₂ injection on Norne

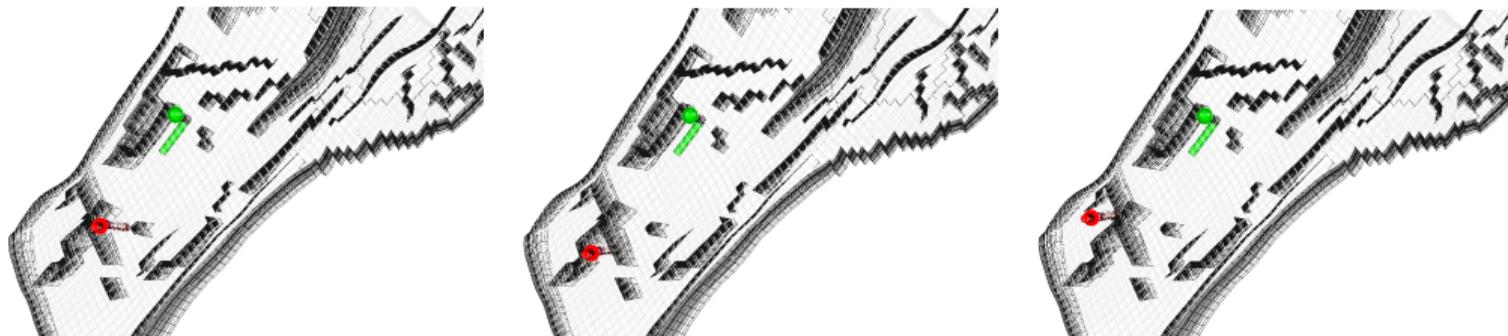


Figure: Positions of the wells in the Norne CO₂ injection scenarios: left for the wells in the same compartment, middle - wells are separated by a fault, right - injection well in the corner.

CO₂ injection on Norne

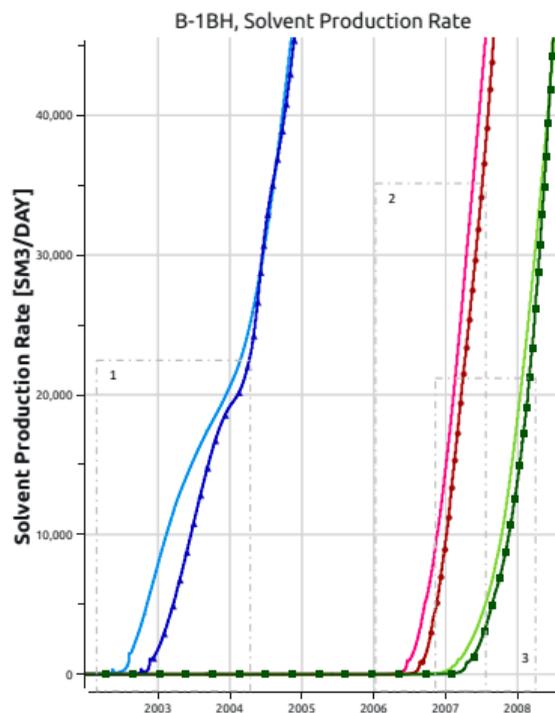


Figure: Solvent production rate for the three scenarios of solvent injection on Norne.

CO₂ injection on Norne: zoom-in

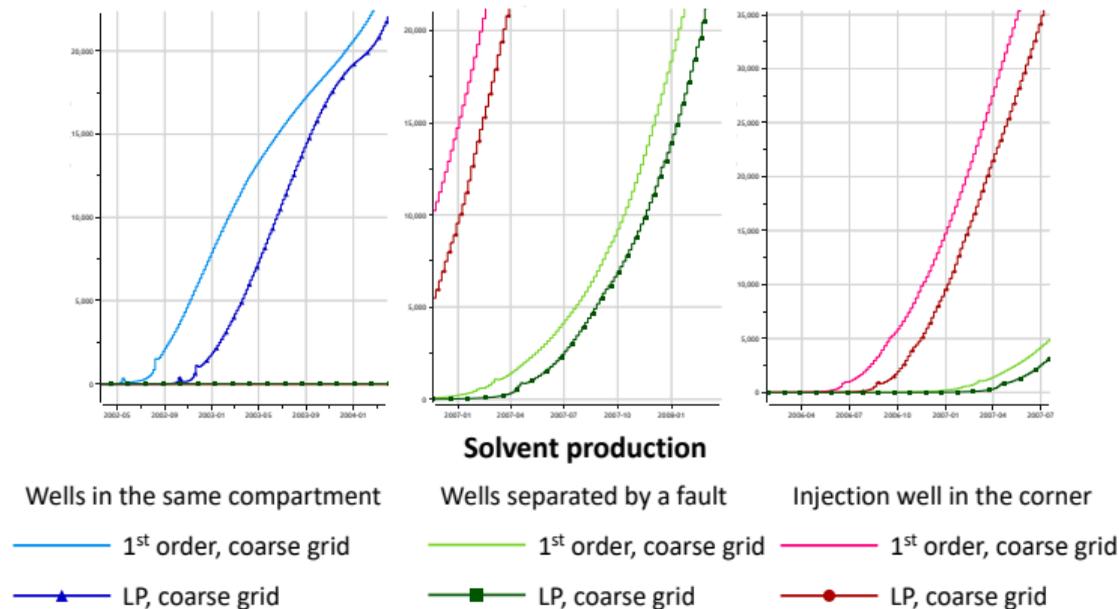
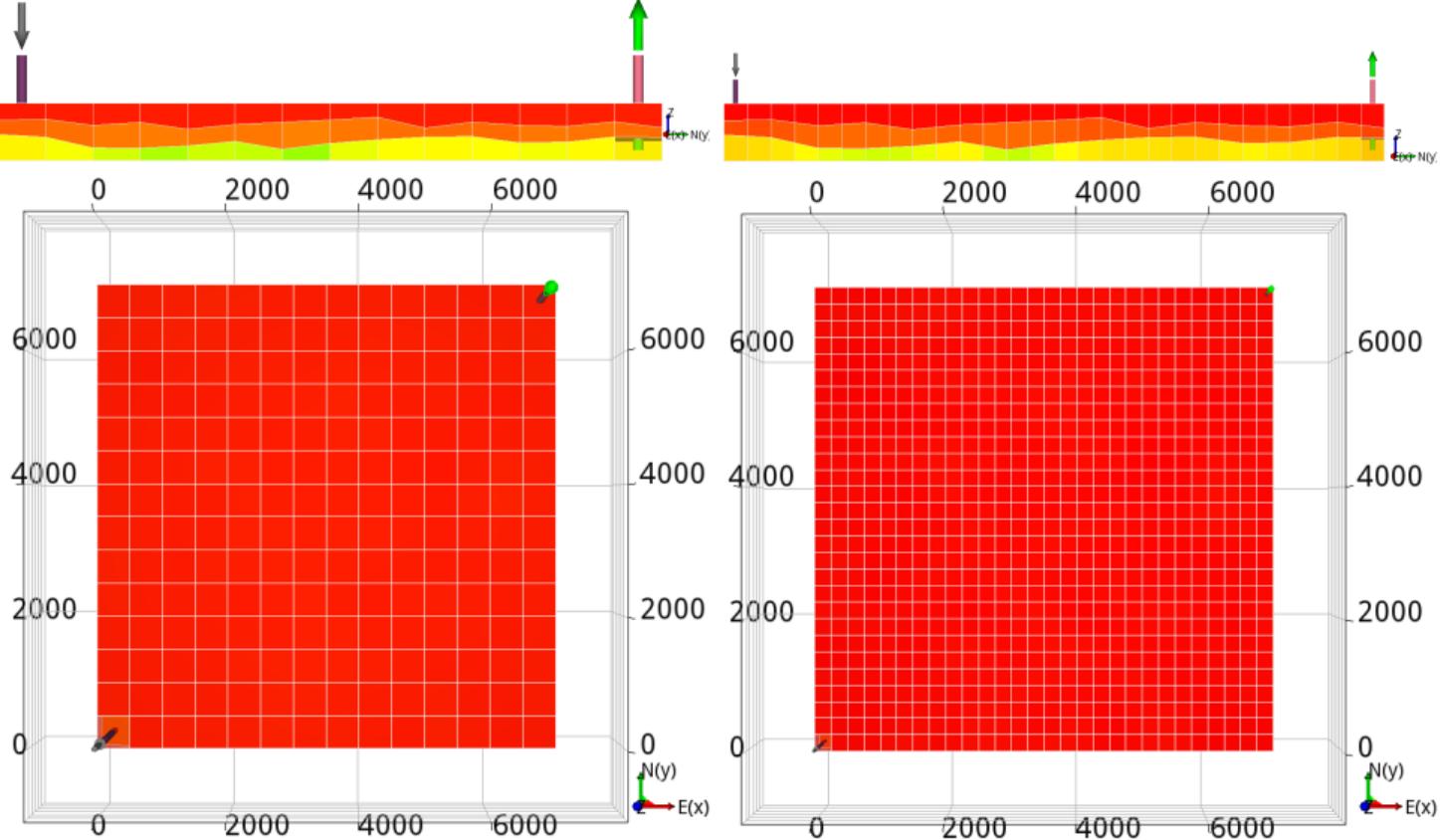
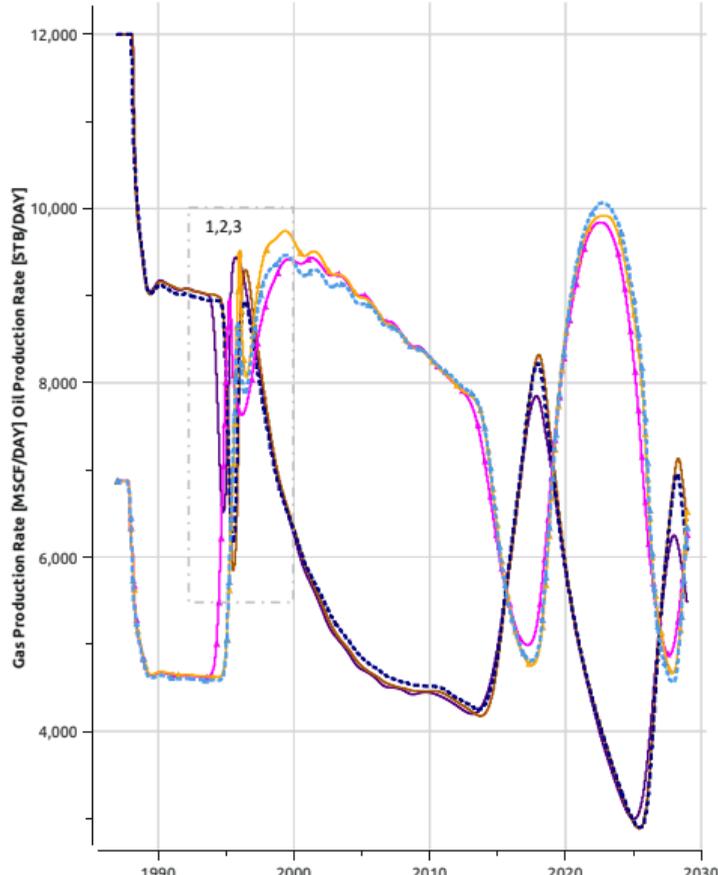


Figure: The subplots zoom into the times of solvent arrival for each scenario: (1) Wells in the same compartment; (2) Wells separated by a fault; (3) Injection well in the corner.

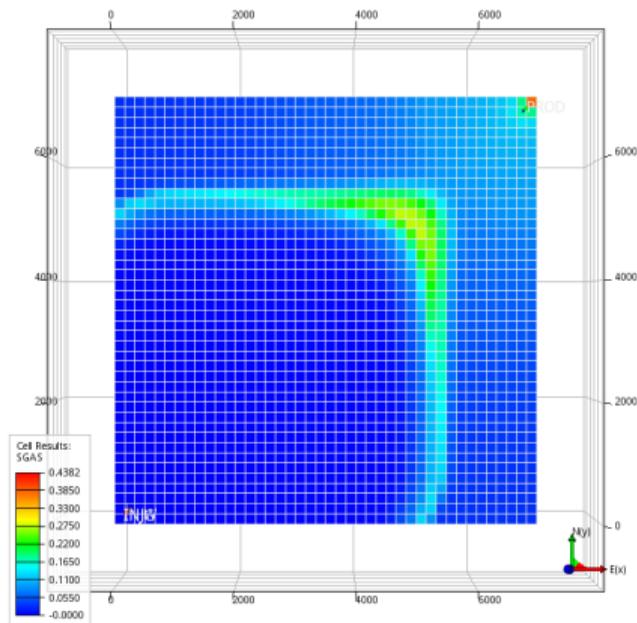
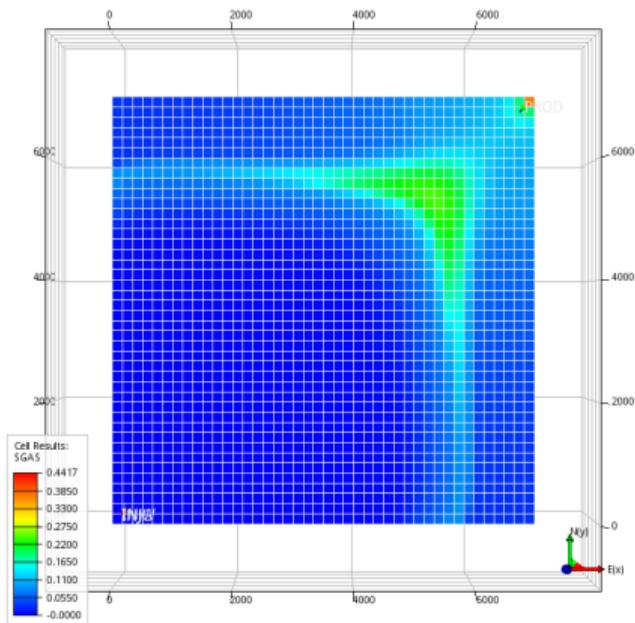
A medium-sized realistic reservoir with an unstructured corner point grid



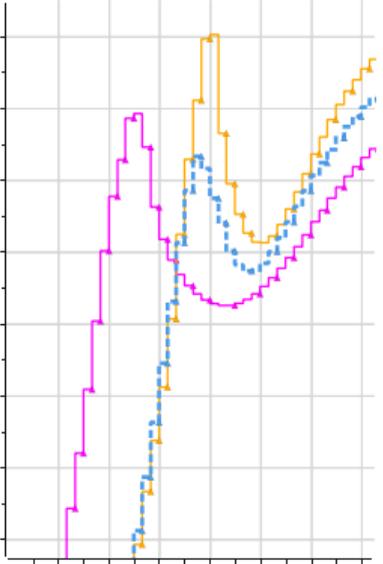
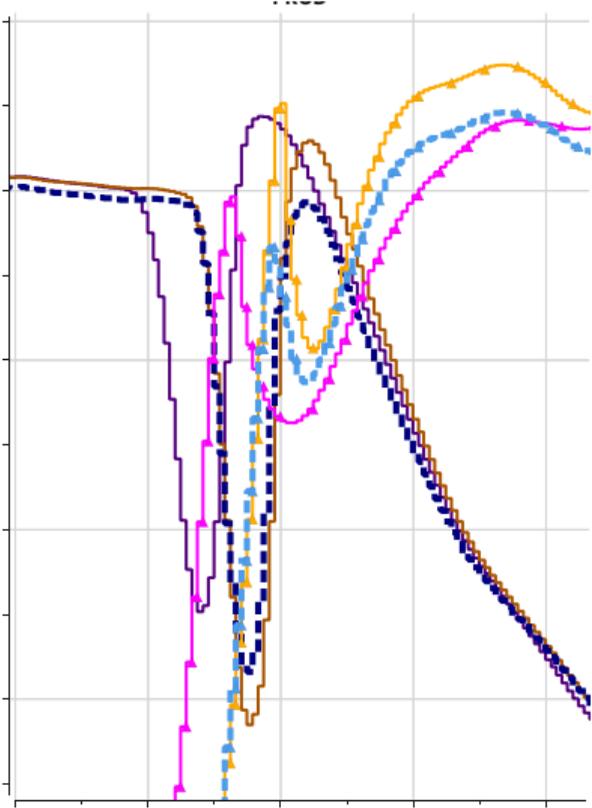
Oil and Gas production rate



Gas wave



Oil and Gas production rate: zoom-in



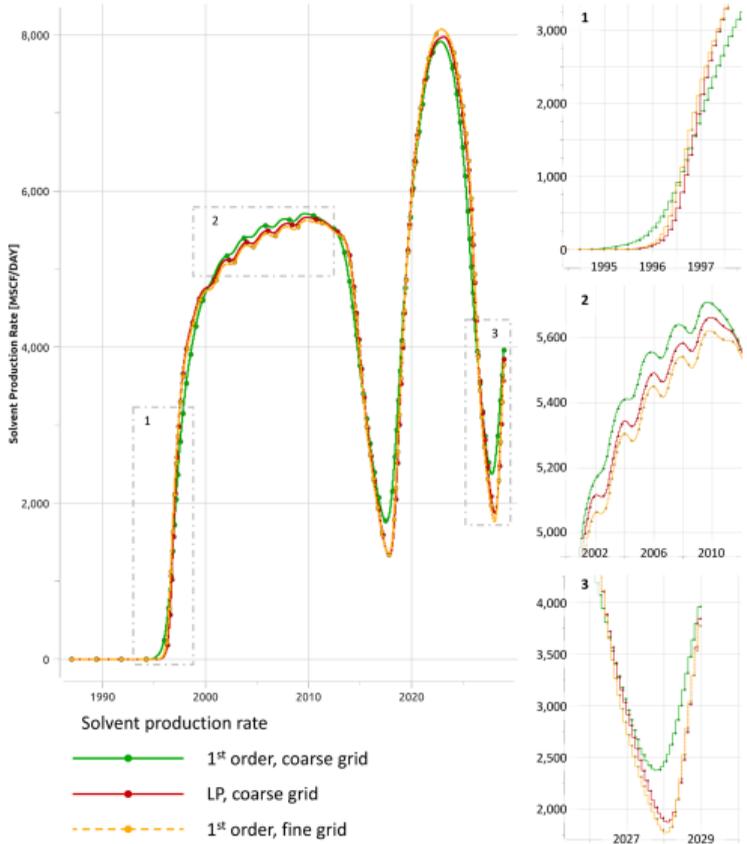
Oil production rate

Gas production rate

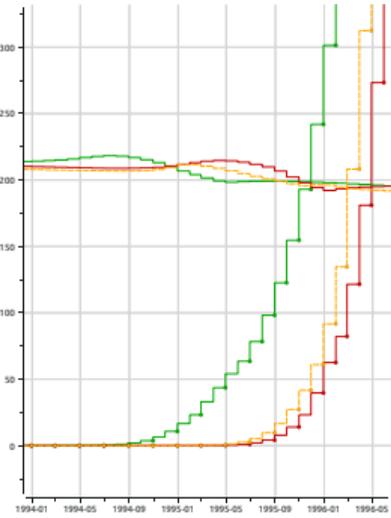
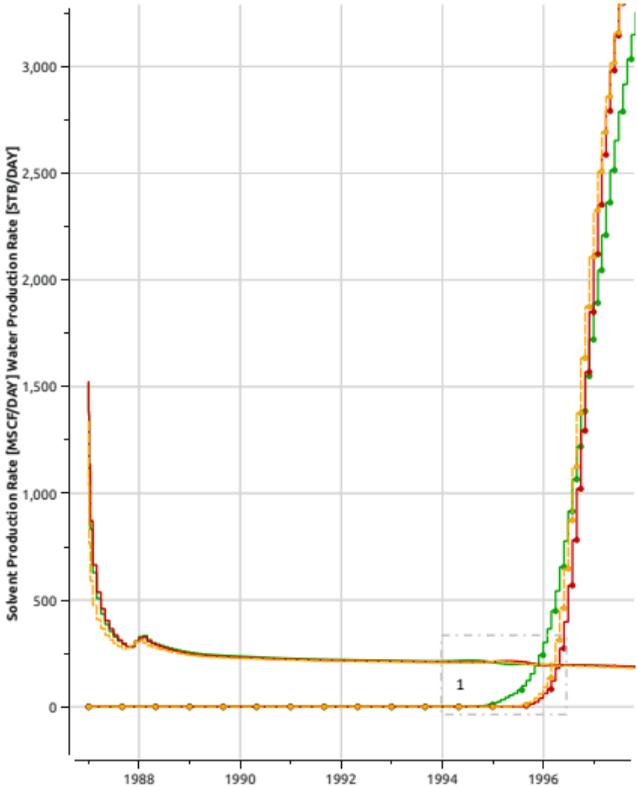
- 1st order, coarse grid
- LP, coarse grid
- - - 1st order, fine grid

- ▲ 1st order, coarse grid
- ▲ LP, coarse grid
- - -▲ 1st order, fine grid

Solvent production rate



Production rates of water and solvent during the whole simulation on the right and zoom in on the left



1

- Solvent production rate
 - 1st order, coarse grid
 - LP, coarse grid
 - 1st order, fine grid
- Water production rate
 - 1st order, coarse grid
 - LP, coarse grid
 - 1st order, fine grid

Conclusions

- ▶ Showed that second-order method improves accuracy in front positioning and reduces smearing.
- ▶ Complexity of the reservoir can overshadow the effects gained by using a higher-order computational method.
- ▶ Verified the results with the first-order method on the refined grid, both for the medium-sized reservoir and the Norne test case.

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