

# Discretization methods for AD-GPRS

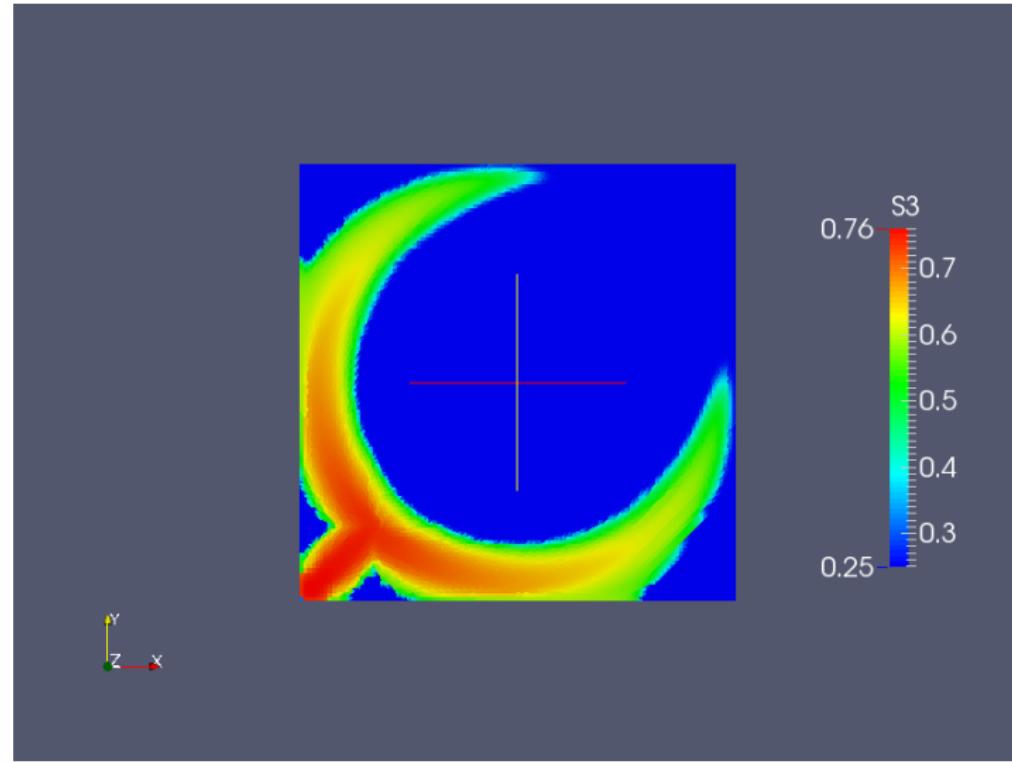
Bradley Mallison, Hamdi Tchelepi, Kirill Terekhov  
SUPRI-B meeting

April 7, 2015

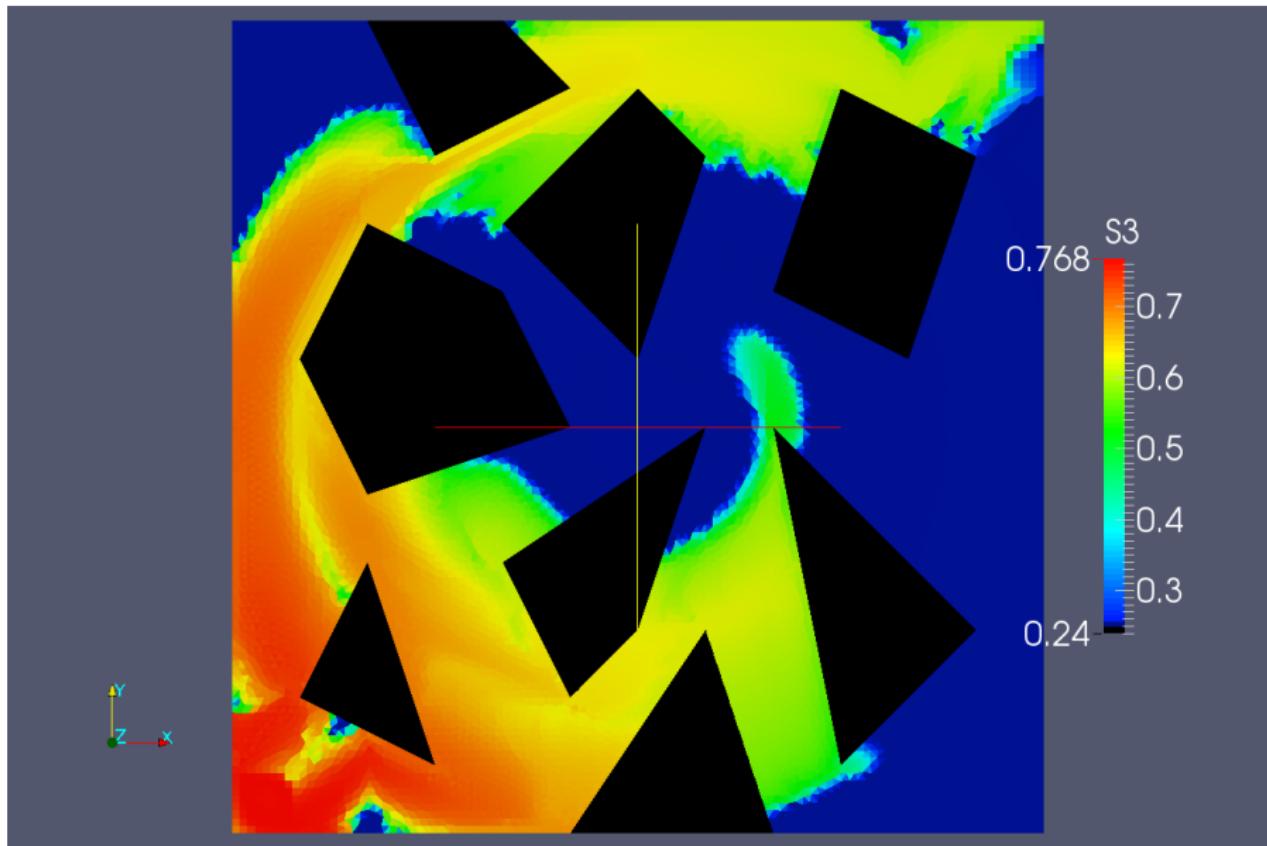
# Contents

- Nonlinear Finite Volume Schemes.
- Discretization Toolkit for AD-GPRS.
- Performance improvements for AD-GPRS.

# Intro video



# Intro video



# Nonlinear Finite Volume Schemes

Next slides

Derivation of one of the schemes.

# Nonlinear Finite Volume Schemes

## Why one would like to use nonlinear scheme?

- To obtain wanted properties of the schemes:
  - ▶ Positivity of the solution.
  - ▶ Satisfy discrete maximum principle.
- To improve robustness and speed of the simulator.

## Is there anything wrong with them?

Yes!

- Proof of convergence requires coercivity.
- DMP schemes require specific nonlinear solvers.

In the following: derivation of one of the nonlinear schemes.

# Nonlinear Finite Volume Schemes

## Toy problem

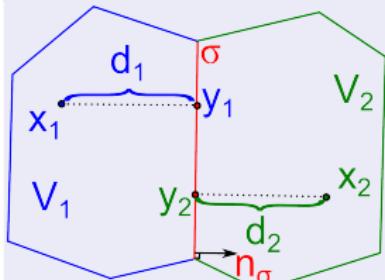
Solve for  $p$  that satisfy anisotropic diffusion equation with source  $f$ :

$$\begin{cases} -\nabla \cdot \mathbb{K} \nabla p = f & \text{in } \Omega, \\ \mathbb{K} \frac{\partial p}{\partial \vec{n}} = f_N & \text{in } \Gamma_N, \\ p = f_D & \text{in } \Gamma_D. \end{cases}$$

Where domain  $\Omega$  is decomposed into polyhedral mesh,  $\mathbb{K}$  is a symmetric positive definitive and piecewise-constant on each polyhedron of the mesh and  $\Gamma_N \cup \Gamma_D = \partial\Omega$  is a piecewise-linear boundary.

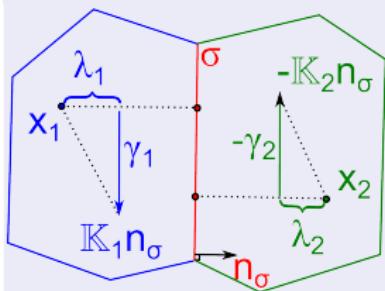
# Nonlinear Finite Volume Schemes

## Approximation of the gradient in two materials



$$\nabla_1 p = \frac{p(\mathbf{y}_1) - p_1}{d_1} \vec{n}_\sigma + \vec{g}_\sigma,$$
$$-\nabla_2 p = \frac{p(\mathbf{y}_2) - p_2}{d_2} \vec{n}_\sigma + \vec{g}_\sigma.$$

## Representation of co-normals in two materials



$$\lambda_1 = \vec{n}_\sigma \cdot \mathbb{K}_1 \vec{n}_\sigma, \quad \vec{\gamma}_1 = (\mathbb{K}_1 - \lambda_1 \mathbb{I}) \vec{n}_\sigma,$$
$$\lambda_2 = \vec{n}_\sigma \cdot \mathbb{K}_2 \vec{n}_\sigma, \quad \vec{\gamma}_2 = (\mathbb{K}_2 - \lambda_2 \mathbb{I}) \vec{n}_\sigma.$$

# Nonlinear Finite Volume Schemes

Flux continuity condition on interface  $\sigma$

$$\mathbb{K}_1 \nabla_1 p - \mathbb{K}_2 \nabla_2 p = 0$$

or

$$\lambda_1 \frac{p(\mathbf{y}_1) - p_1}{d_1} + \vec{\gamma}_1 \cdot \vec{g}_\sigma = \lambda_2 \frac{p_2 - p(\mathbf{y}_2)}{d_2} + \vec{\gamma}_2 \cdot \vec{g}_\sigma.$$

Interpolation on interface  $\sigma$

$$p(\mathbf{y}_1) = p(\mathbf{y}) + (\mathbf{y}_1 - \mathbf{y}) \cdot \vec{g}_\sigma, \quad p(\mathbf{y}_2) = p(\mathbf{y}) + (\mathbf{y}_2 - \mathbf{y}) \cdot \vec{g}_\sigma.$$

We want to get an expression for  $p(\mathbf{y})$  on  $\sigma$ .

# Nonlinear Finite Volume Schemes

Interpolation on interface  $\sigma$

$$\begin{aligned} p(\mathbf{y}) &= \frac{\lambda_1 d_2 p_1 + \lambda_2 d_1 p_2}{\lambda_1 d_2 + \lambda_2 d_1} + \frac{d_1 d_2}{\lambda_1 d_2 + \lambda_2 d_1} (\gamma_2 - \gamma_1) \cdot \vec{g}_\sigma \\ &+ \left( \mathbf{y} - \frac{\lambda_1 d_2 \mathbf{y}_1 + \lambda_2 d_1 \mathbf{y}_2}{\lambda_1 d_2 + \lambda_2 d_1} \right) \cdot \vec{g}_\sigma. \end{aligned}$$

Harmonic averaging point

$$\begin{aligned} \mathbf{y}_\sigma &= \frac{\lambda_1 d_2 \mathbf{y}_1 + \lambda_2 d_1 \mathbf{y}_2 + d_1 d_2 (\gamma_1 - \gamma_2)}{\lambda_1 d_2 + \lambda_2 d_1}, \\ p(\mathbf{y}_\sigma) &= \frac{\lambda_1 d_2 p_1 + \lambda_2 d_1 p_2}{\lambda_1 d_2 + \lambda_2 d_1}. \end{aligned}$$

# Nonlinear Finite Volume Schemes

## Harmonic averaging point

**Léo Agelas, Robert Eymard, Raphaèle Herbin:** *A nine-point finite volume scheme for the simulation of diffusion in heterogeneous media*, C. R. Acad. Sci. Paris 2009

We will proceed in derivation the decomposition of the total flux into **transverse** and **normal** parts.

# Nonlinear Finite Volume Schemes

**Normal** component of the flux

$$\frac{\lambda_1}{d_1}(p(\mathbf{y}_\sigma) - p_1) = \frac{\lambda_2}{d_2}(p_2 - p(\mathbf{y}_\sigma)) = \frac{\lambda_1 \lambda_2}{\lambda_1 d_2 + \lambda_2 d_1}(p_2 - p_1).$$

Which is exactly definition of the Two-Point Flux Approximation or TPFA.

**Transverse** direction of the flux

$$\begin{aligned}\vec{\gamma} &= \mathbb{K}_1 \vec{n}_\sigma - \frac{\lambda_1}{d_1}(\mathbf{y}_\sigma - \mathbf{x}_1) = \mathbb{K}_2 \vec{n}_\sigma - \frac{\lambda_2}{d_2}(\mathbf{x}_2 - \mathbf{y}_\sigma) \\ &= \frac{\lambda_1 \lambda_2 (\mathbf{y}_1 - \mathbf{y}_2) + \lambda_2 d_1 \vec{\gamma}_1 + \lambda_1 d_2 \vec{\gamma}_2}{\lambda_1 d_2 + \lambda_2 d_1}.\end{aligned}$$

# Nonlinear Finite Volume Schemes

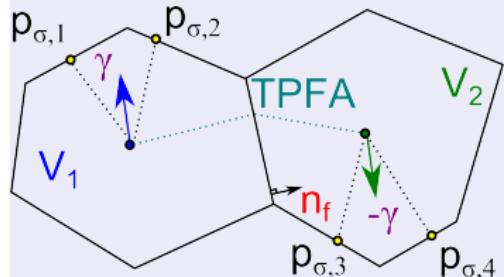
**Transverse** direction of the flux

**Lina Chang, Guangwei Yuan:** *An efficient and accurate reconstruction algorithm for the formulation of cell-centered diffusion schemes*, J. Comp. Phys., 2012

How do we approximate the **transverse** part of the flux?  
In each polyhedron  $V_1$  and  $V_2$  adjacent to the interface  $\sigma$   
we search a positive-definite basis.

# Nonlinear Finite Volume Schemes

## Transverse flux approximation



$$\vec{\gamma} \cdot \nabla_1 p = \sum_{i_1=1}^{\dim} \alpha_{i_1} (p_{\sigma,i_1} - p_1),$$
$$-\vec{\gamma} \cdot \nabla_2 p = \sum_{i_2=1}^{\dim} \beta_{i_2} (p_{\sigma,i_2} - p_2),$$
$$\forall i_1, i_2 \in [1, \dim] : \alpha_{i_1}, \beta_{i_2} \geq 0.$$

We define transverse flux  $\gamma = \vec{\gamma} \cdot \nabla p$  as a convex combination of two parts  $\gamma_1 = \vec{\gamma} \cdot \nabla_1 p$  and  $\gamma_2 = \vec{\gamma} \cdot \nabla_2 p$ .

# Nonlinear Finite Volume Schemes

## Transverse flux approximation

$$\begin{cases} \gamma = \mu_1\gamma_1 + \mu_2\gamma_2, \\ \mu_1 + \mu_2 = 1. \end{cases}$$

Choice of  $\mu_1 = \mu_2 = \frac{1}{2}$  is not very stable. We can follow two approaches of the definition of weights  $\mu_1$  and  $\mu_2$ .

## Nonlinear Two-Point Flux Approximation

**Christophe Le Potier** *Schéma volumes finis monotone pour des opérateurs de diffusion fortement anisotropes sur des maillages de triangles non structurés* C. R. Acad. Sci. Paris, 2005

# Nonlinear Finite Volume Schemes

## Nonlinear Two-Point Flux Approximation

Expanded expressions for one-sided fluxes:

$$\gamma_1 = \alpha_1(p_{\sigma,1} - p_1) + \alpha_2(p_{\sigma,2} - p_1),$$

$$\gamma_2 = \beta_3(p_2 - p_{\sigma,3}) + \beta_4(p_2 - p_{\sigma,4}).$$

Linear combination:

$$\gamma = \mu_2(\beta_3 + \beta_4)p_2 - \mu_1(\alpha_1 + \alpha_2)p_1$$

$$+ \mu_1(\alpha_1 p_{\sigma,1} + \alpha_2 p_{\sigma,2}) - \mu_2(\beta_3 p_{\sigma,3} + \beta_4 p_{\sigma,4}).$$

We define  $\mu_1$  and  $\mu_2$  so that expression in red is zero and convexity  $\mu_1 + \mu_2 = 1$  is satisfied. If we freeze  $\mu_1$  and  $\mu_2$  on each iteration then the matrix obtained is an M-matrix with nonnegative column-sum.

# Nonlinear Finite Volume Schemes

## Nonlinear Multi-Point Approximation

**Enrico Bertolazzi, Gianmarco Manzini:** A second-order maximum principle preserving finite volume method for steady convection-diffusion problems, SIAM J. Numer. Anal., 2005

The weights are defined as:

## Nonlinear Multi-Point Approximation

$$\mu_1 = \frac{|\gamma_2|}{|\gamma_1| + |\gamma_2|}, \quad \mu_2 = \frac{|\gamma_1|}{|\gamma_1| + |\gamma_2|}.$$

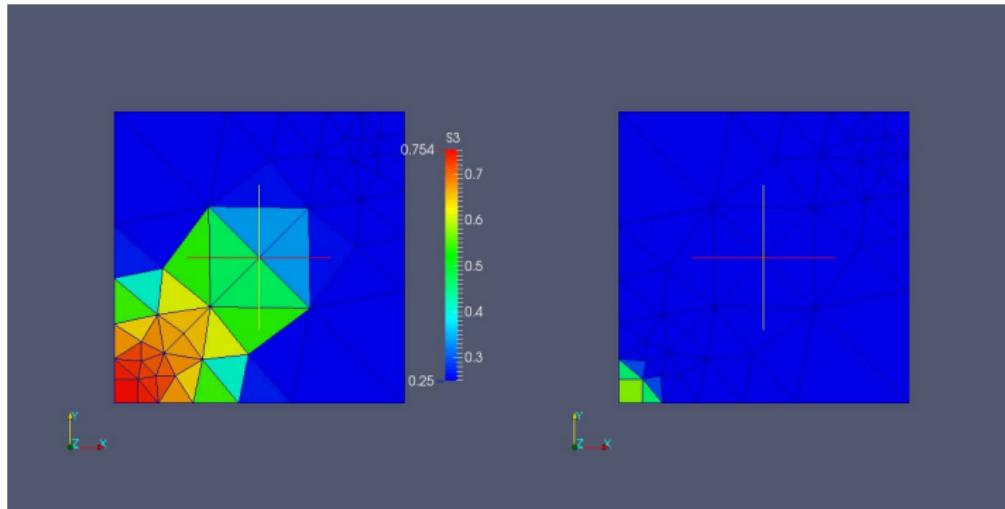
# Nonlinear Finite Volume Schemes

## Nonlinear Multi-Point Approximation

$$\begin{aligned}\gamma &= \begin{cases} \gamma_1\mu_1 = \gamma_2\mu_2, & \gamma_1\gamma_2 > 0, \\ 0, & \gamma_1\gamma_2 < 0. \end{cases} \\ &= \frac{|\gamma_1||\gamma_2|}{|\gamma_1| + |\gamma_2|}(\text{sign}(\gamma_1) + \text{sign}(\gamma_2)).\end{aligned}$$

If we freeze  $\mu_1$  and  $\mu_2$  and will use  $\gamma = \gamma_1\mu_1$  to define flux on  $V_1$  and  $\gamma = \gamma_2\mu_2$  on  $V_2$  then on each iteration the matrix obtained will be an M-matrix with nonnegative row-sum.

# Nonlinear Finite Volume Schemes



Why a new formulation for the scheme was derived?

# Nonlinear Finite Volume Schemes

## Convergence

**Jérôme Droniou, Christophe Le Potier:**

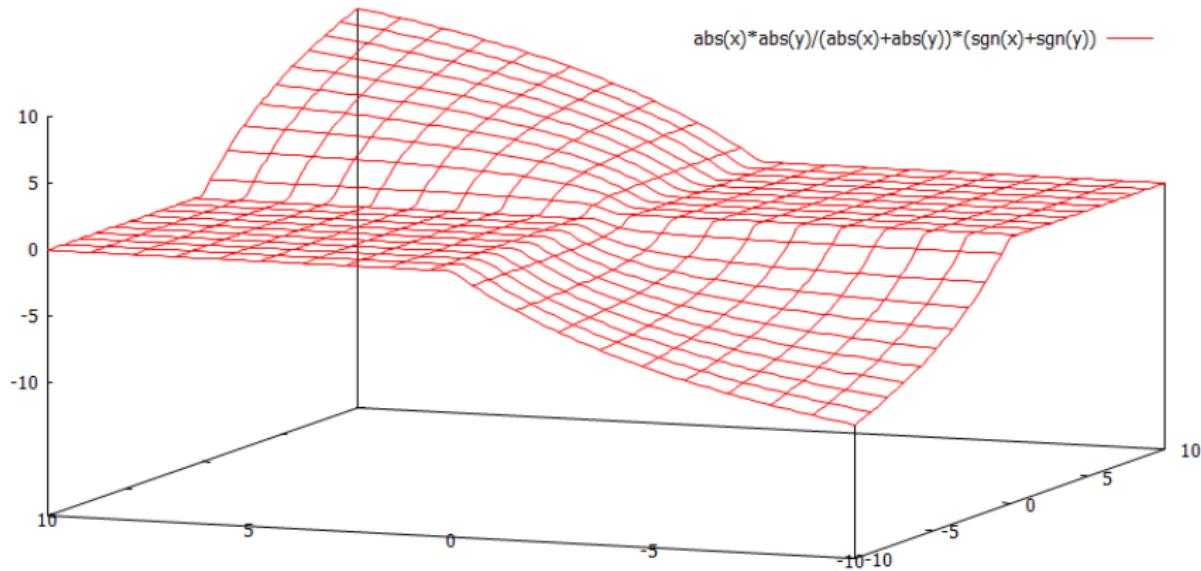
*Construction and convergence study of schemes preserving the elliptic local maximum principle, SIAM J. Numer. Anal., 2011*

## Nonlinear solver

Since methods are convergent under coercivity assumption we will use the Newton method with the full Jacobian.

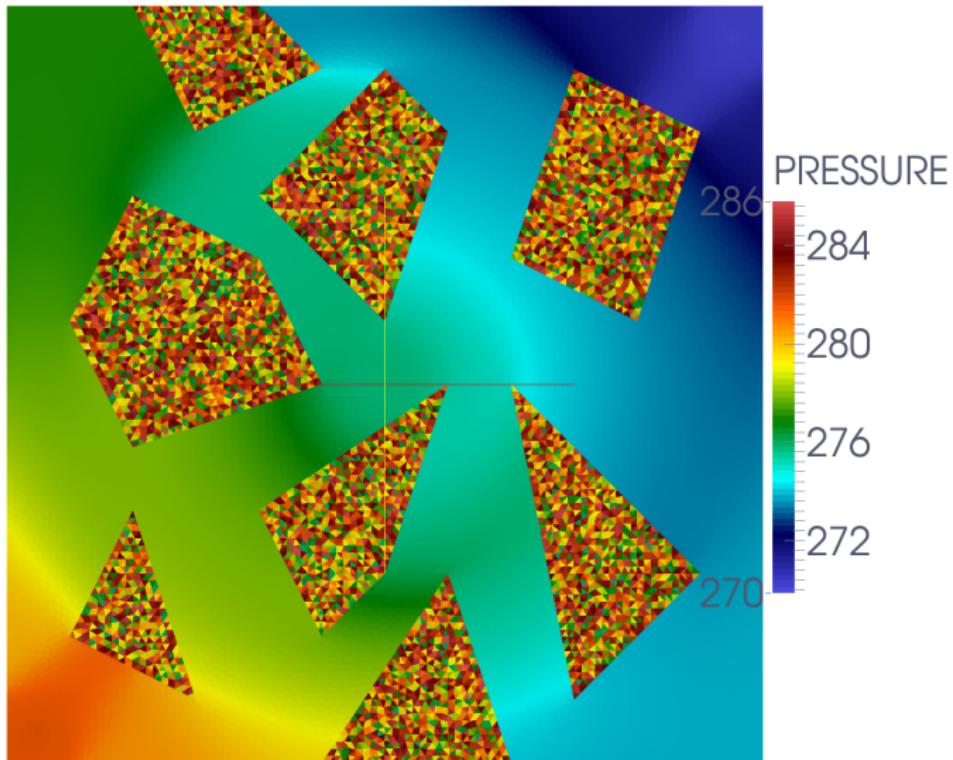
- NTPFA has a convex flux function in positive region.
- NMPFA has an upwind-style plateau in the flux and may not converge with pure newton method.

# Nonlinear Finite Volume Schemes



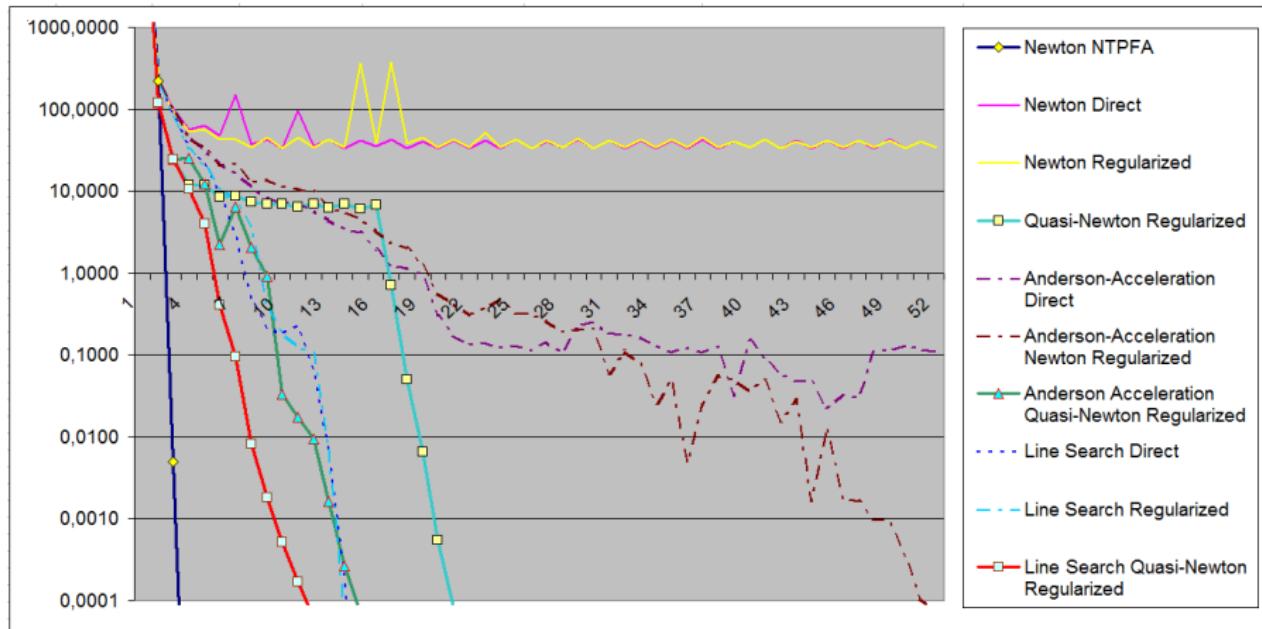
Shape of the **transverse** flux function. Presence of two plateau regions result in unconstrained behavior of transverse half-fluxes in these regions.

# Nonlinear Finite Volume Schemes



Convergence of the NMPFA method on a simple problem.

# Nonlinear Finite Volume Schemes



NTPFA: 3 iterations of Newton method

NMPFA: 11 iterations of Quasi-Newton + Line Search or  
14 iterations of Quasi-Newton + Anderson Acceleration.

# Discretization Toolkit

Next slides

Discretization toolkit for AD-GPRS.

# Discretization Toolkit

## Purposes

- Produce an input for AD-GPRS:
  - ▶ produce discretization for a general polyhedral mesh;
  - ▶ discretize a fracture network;
  - ▶ intersect wells with mesh geometry, compute well indices;
  - ▶ prepare geometry, initial and boundary condition for geomechanics (under work).
- Works for me as a sandbox for various methods.

# Discretization Toolkit

## Includes

- A dialog-style console application that produces input for AD-GPRS.
- Converter of ASCII output of AD-GPRS to grid data.
- Several models for testing purposes:
  - ▶ Anisotropic diffusion problem.
  - ▶ Black oil problem.
- Compute error norms on different grids.
- Sample grid generators and converters.

# Discretization Toolkit

## Includes

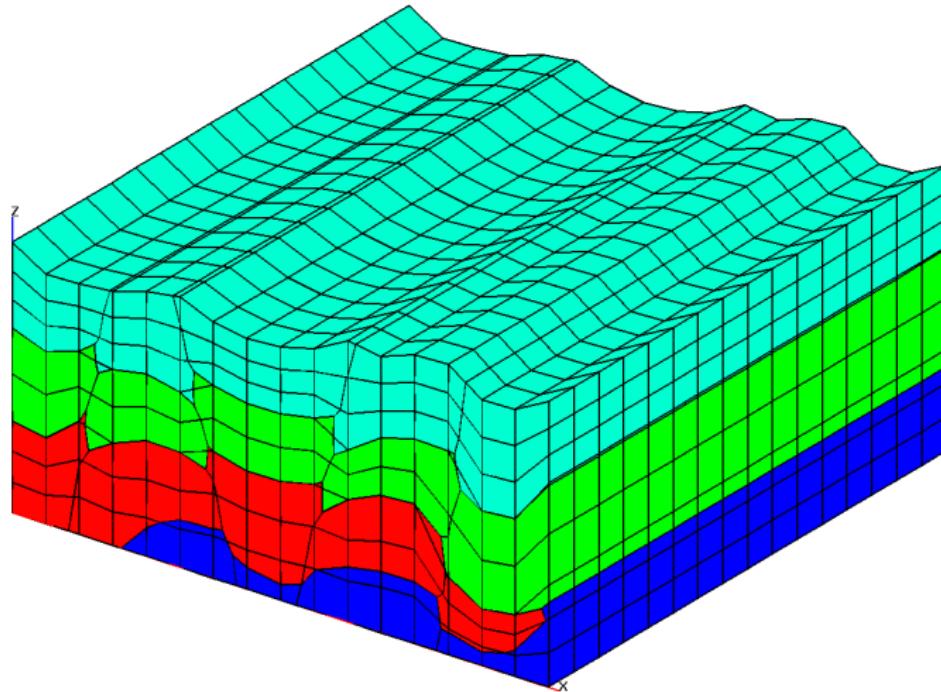
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  - ▶ Black oil problem.
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- Sample grid generators and converters.

# Discretization Toolkit

## Workflow for AD-GPRS input

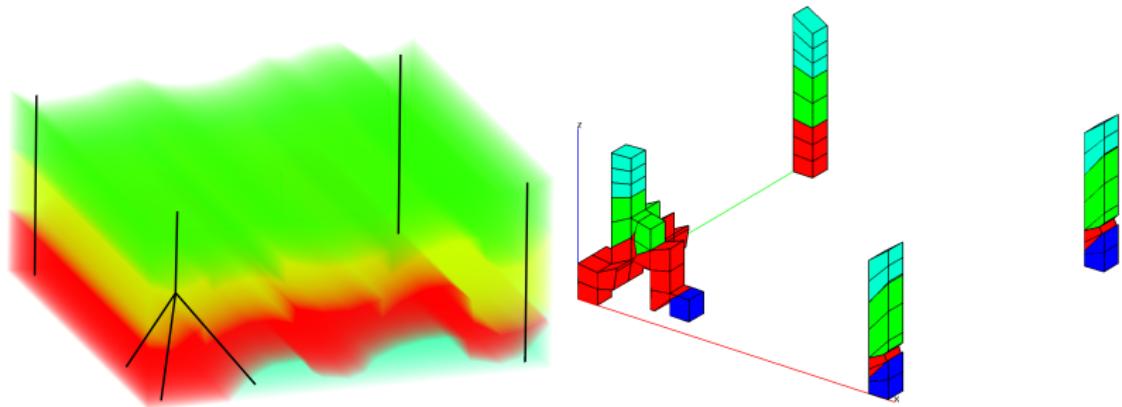
- Define properties such as porosity or permeability.
- Define apertures on the fracture faces.
- Select one of the available schemes:
  - ▶ TPFA, MPFA-O, MPFA-L, MPFA-G, MPFA-A, MPFA-B
  - ▶ NTPFA-A, NTPFA-B, NTPFA-C
  - ▶ NMPFA-A, NMPFA-B
  - ▶ fracture treatment: full scheme, reduce to two-point inside fracture, reduce to two-point between matrix and fracture.
- Provide inclinometry of the well.
- Input initial and boundary conditions of geomechanics, not well tested yet. :(

# Discretization Toolkit



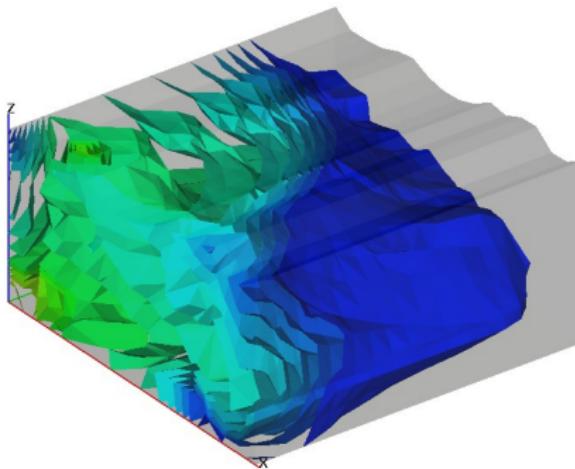
Imagine we were given a mesh like this and were asked to calculate something.

# Discretization Toolkit



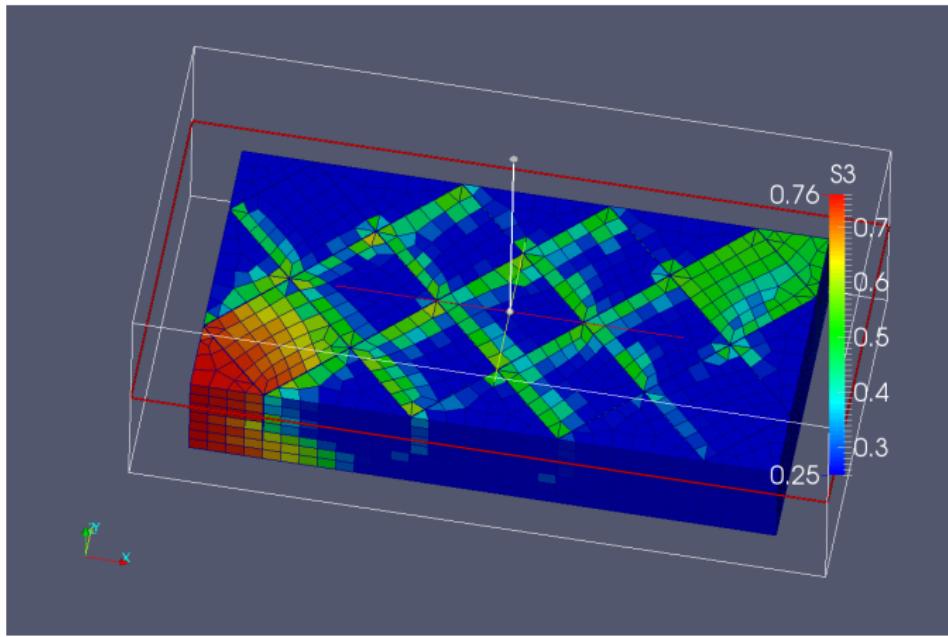
Wells definition and their intersection with mesh.

# Discretization Toolkit



We feed input to AD-GPRS and run the black oil model.  
Convert ASCII-formated results into gmv format and make  
the video with GMV and ffmpeg.

# Discretization Toolkit



The variant with fractures thanks to Timur Garipov.  
Convert ASCII-formated results into vtk format and make  
the video in paraview.

# Performance

## Next slides

Performance improvements and comparison of the performance of the schemes in AD-GPRS.

# Performance

## Operations with AD-variables

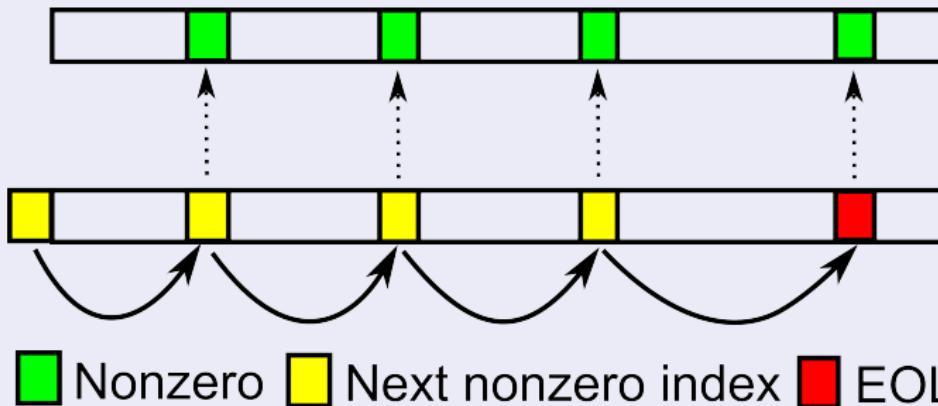
Results in the problem of calculation of linear combination of sparse vectors:

x	
*	
y	
=	
z	

AD-GPRS analyses sparsity of sparse vectors to be combined and then uses either implicit binary tree representation or dense representation. (by Rami Younis)

# Performance

Dense linked list as in  $\text{ILU}(\tau)$  preconditioners

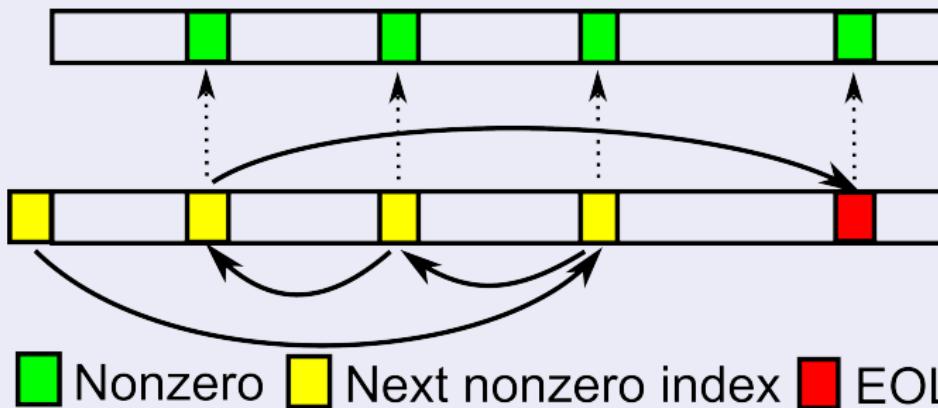


## Pros and cons

- Pros: No need to test for sparsity.
- Cons: Have to rewind and search correct position to keep list sorted.

# Performance

Dense linked list as for Schur complement



## Pros and cons

- Pros: Insertion order is arbitrary.
- Cons: Produces unsorted result, not a big problem.

# Performance

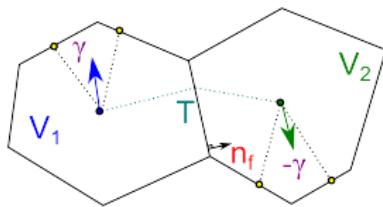
## Unaddressed issues

- AD-expression  $\frac{x}{\sqrt{x*x+\varepsilon}}$  will result in an internal temporary representation of linear combination:  
$$\frac{1}{\sqrt{x^2+\varepsilon}}\delta x + \frac{x^2}{2(x^2+\varepsilon)^{\frac{3}{2}}}\delta x + \frac{x^2}{2(x^2+\varepsilon)^{\frac{3}{2}}}\delta x.$$
- Moreover if we multiply an expression by a constant each entry in an internal representation have to be multiplied.

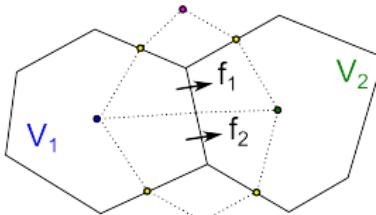
Both issues greatly impact performance of nonlinear schemes.

# Performance

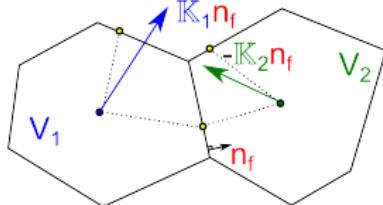
NTPFA-A



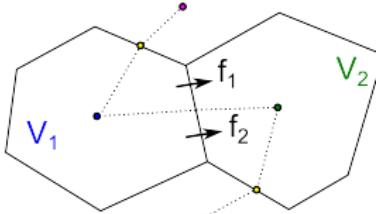
MPFA-O



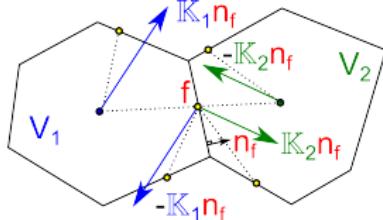
NTPFA-B



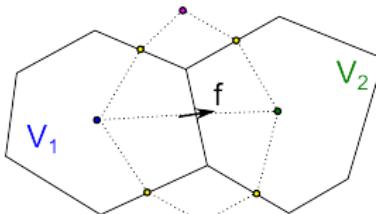
MPFA-L



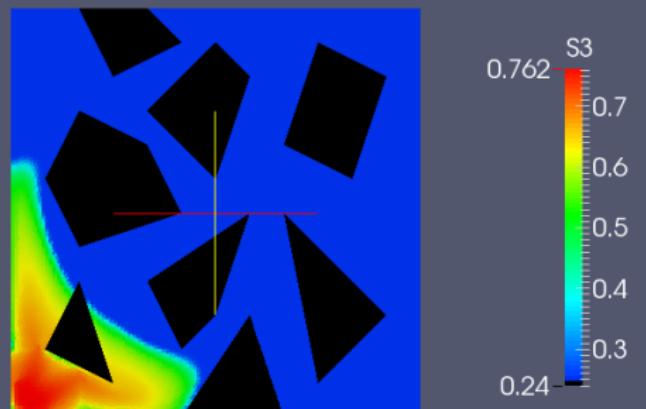
NTPFA-C



MPFA-G



# Performance



Several steps of the problem for comparison with anisotropy ratio  $\frac{10}{100}$ .

# Performance

Total time, seconds:

Method

Approach	Nonlinear iterations	Linear iterations
TPFA	180	4941
MPFA-O	157	2790
MFPA-G	188+20	9475+7021
MPFA-L	180+20	12764+4053
NTPFA-A	159	4743
NTPFA-B	158	5170
NTPFA-C	170	3449

# Performance

Total time, seconds:

Method

Approach	SAXPY+IBTREE	Ordered LL	Unordered LL
TPFA	109.35	95.73	102.46
MPFA-O	158.90	134.07	133.01
MFPA-G	525.78	355.22	341.62
MPFA-L	435.36	306.96	273.46
NTPFA-A	160.12	123.98	129.64
NTPFA-B	162.44	133.00	136.92
NTPFA-C	162.04	126.73	127.24

# Performance

Discretization, seconds:

Method

Approach	SAXPY+IBTREE	Ordered LL	Unordered LL
TPFA	15.33	15.26	15.6
MPFA-O	32.66	26.35	26.51
MFPA-G	31.90	20.78	21.28
MPFA-L	22.77	16.8	17.08
NTPFA-A	24.30	17.46	18.00
NTPFA-B	22.58	17.1	17.40
NTPFA-C	41.75	26.53	26.33

# Performance

Linear solver GMRES\_CPR0 with tolerance  $10^{-6}$ ,  
seconds:

Method

Approach	SAXPY+IBTREE	Ordered LL	Unordered LL
TPFA	68.42	56.8	61.85
MPFA-O	99.63	83.36	80.39
MFPA-G	440.06	242.36	228.71
MPFA-L	383.56	267.64	232.51
NTPFA-A	111.60	84.26	87.92
NTPFA-B	115.7	93.3	96.18
NTPFA-C	93.53	75.62	74.73

# Future considerations

- Nonlinear schemes
  - ▶ Is it possible to make DMP scheme as fast as others?
  - ▶ How to achieve unconditional coercivity and keep method as finite-volume cell-centered?
- Discretization toolkit
  - ▶ Mimic star-delta transformation on fracture joints for all the schemes.
  - ▶ Improve input methods with an interpreter.
  - ▶ Finish geomechanics input.
  - ▶ In case of demand provide interactive graphical user interface.

# Thank you for attention!