Accélération de la méthode de Newton par le préconditionnement de Jacobi non linéaire

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## Nonlinear algebraic system

Model problem: Find $\mathbf{u} \in \mathbb{R}^{N}$

$$
\beta(\mathbf{u})+A \mathbf{u}=\mathbf{b}, \quad \mathbf{b} \geqslant 0
$$



Assumptions:
■ $\beta_{i}: \mathbb{R}^{+} \rightarrow \mathbb{R}^{+}$increasing and concave, $\beta_{i}^{\prime}(0) \leqslant+\infty$

- $J(\mathbf{u})=\beta^{\prime}(\mathbf{u})+A$ is M-matrix:
$J(\mathbf{u})^{-1} \geqslant 0$ and $(J(\mathbf{u}))_{i j} \leqslant 0, i \neq j$


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Objective: Efficient Newton-like method


## Outline

## Motivation

Monotone Newton Theorem versus numerical experiment

Variable switching by parametrization

Nonlinear Jacobi preconditioning

## Motivation

## Approximate solution of nonlinear PDEs

Nonlinear PDE

$$
\partial_{t} \beta(u)+L(u)=0
$$

$L$ is a linear elliptic operator.
Nonlinear discrete problem

$$
\frac{\beta\left(u^{n+1}\right)-\beta\left(u^{n}\right)}{\Delta t_{n}}+A u^{n+1}=0
$$

Discretization

- Implicit in time
- Monotone discretization of $L$ :

■ TPFA finite volumes

- $\mathbf{P}_{1}$ finite elements with mass-lumping (on an appropriate mesh)

■ Upstream weighting for convection

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Applications in Geosciences:

- Porous media equation
- Contaminant transport with adsorption
- Richards' equation


## Applications: porous media-like equations

Porous media equation

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\partial_{t} u^{1 / m}-\Delta u=0, \quad m>1
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Contaminant transport with adsorption

$$
\partial_{t} \underbrace{(u+a(u))}-\operatorname{div}(\nabla u+u \mathbf{V})=0
$$

dissolved + adsorbed conc.
Freundlich isotherm

$$
a(u)=c u^{1 / m}, \quad c>0, \quad m>1
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Limit case $m \rightarrow+\infty$

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\partial_{t} v+L(u)=0, \quad v \in \beta(u)
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- $\beta$ is maximal monotone
- Connections with obstacle problems (Brugnano \& Sestini '09)


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Limit case $m \rightarrow+\infty$

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Nonlinear solver must be robust w.r.t. to the shape of $\beta$

## Richards' equation

Richards' equation

$$
\partial_{t} s-\operatorname{div}(\lambda(s)(\nabla p-\mathbf{g}))=0, \quad s=S(p)
$$

Natural variables

- pressure $p$
- saturation $s$



Curve $s=S(p)$ reflects the macroscopic capillary effects

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Introducing Kirchhoff transform

$$
U(p)=\int_{0}^{p} \lambda(S(a)) \mathrm{d} a
$$

## Richards' equation

We obtain Richards' equation using generalized pressure

$$
\partial_{t} s-\Delta u=-\operatorname{div}(\lambda(s) \mathbf{g}), \quad s=\beta(u)
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with $\beta(u):=S\left(U^{-1}(u)\right)$


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- Using semi-implicit discretization we find

$$
\beta(\mathbf{u})+A \mathbf{u}=\mathbf{b}
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## Objectives

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Objective: Newton-like iterative method
■ efficient and robust w.r.t. to the shape of $\beta$
■ with guarantied (semi-)global convergence

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Monotone Newton Theorem versus numerical experiment

## Monotone Newton's method

Let

$$
F(\mathbf{u})=\beta(\mathbf{u})+A \mathbf{u}-\mathbf{b}
$$

Newton's method:

$$
\mathbf{u}_{k+1}=\mathbf{u}_{k}-F^{\prime}\left(\mathbf{u}_{k}\right)^{-1} F\left(\mathbf{u}_{k}\right), \quad k \geqslant 0
$$

Theorem (Monotone Newton Theorem (Baluev '52; Ortega \& Rheinboldt '70))
Let $\mathbf{u}_{0}$ satisfy $F\left(\mathbf{u}_{0}\right) \leqslant 0$, then

- $\mathbf{u}_{k}$ converges to the unique solution $\mathbf{u}_{*}$
- $\mathbf{u}_{k} \leqslant \mathbf{u}_{k+1} \leqslant \mathbf{u}_{\star}$ for all $k \geqslant 0$


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Main ingredients:

- $F$ is concave (or convex)
- $F^{\prime}(\mathbf{u})$ is an M-matrix

Illustration ( $\mathrm{N}=1$ )

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Using nested iterations concavity (convexity) assumption can be removed
■ Piece-wise linear systems: Brugnano \& Casulli '09
■ Systems with diagonal nonlinearities: Casulli \& Zanolli '12

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The method is semi-globally convergent. Is it efficient?

## 1D numerical experiment

Porous media equation on $(0,1) \times(0, T)$

$$
\partial_{t} \beta(u)-\partial_{x x}^{2} u=0, \quad \beta(u)=u^{1 / m}
$$

with Neumann boundary conditions
■ Inflow at $x=0:-\partial_{x} u(0, t)=q>0$

- No-flow at $x=1$
- Almost "dry" initial condition: $\beta(u(x, 0))=10^{-10}$


Solution profile at different time steps

## Performance assessment: $u$ - and $v$-formulations

Original $u$-formulation:
Alternative $v$-formulation:

$$
\beta(\mathbf{u})+A \mathbf{u}-\mathbf{b}=0 \quad \mathbf{v}+A \beta^{-1}(\mathbf{v})-\mathbf{b}=0
$$

Different values of $m>1$ in $\beta(u)=u^{1 / m}$


■ Dashed: Original formulation is inefficient, manly because $\beta^{\prime}(0)=+\infty$.

- Solid: Alternative formulation is more efficient, but concavity is lost:
note that $(A)_{i i}(A)_{i j} \leqslant 0, i \neq j$


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Different values of $m>1$ in $\beta(u)=u^{1 / m}$


- Performance of both formulations depends on $m$
- Can we find some even more efficient primary variable?


## Variable switching by parametrization

## Adaptive choice of the variable



- Switching between $v$ and $u$ may be a good idea
- Well known for Richards' equation


## Efficiency of variable switching

- $v$-formulation: $\partial_{t} v-\triangle \beta^{-1}(v)=0$

■ variable switching: PDE?


- Variable switching: is more efficient and is robust w.r.t. $m$
- Drawback: implementation using if/else conditions


## Graph parametrization

Parametrization of the graph $v=\beta(u)$ :
Let $\bar{u}, \bar{v}: \mathbb{R}^{+} \rightarrow \mathbb{R}^{+}$such that

$$
\bar{v}(\tau)=\beta(\bar{u}(\tau)) \quad \forall \tau \in \mathbb{R}^{+}
$$



PDE in terms of the new variable $\tau$

$$
\partial_{t} \bar{v}(\tau)-\Delta \bar{u}(\tau)=0
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Variable switching:

$$
\max \left(\bar{v}^{\prime}(\tau), \bar{u}^{\prime}(\tau)\right)=1
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- Multi-valued closure $v \in \beta(u)$ is Ok



## Estimates (B. \& Cancès '17)

Define $F_{\tau}(\boldsymbol{\tau})=\bar{v}(\boldsymbol{\tau})+A \bar{u}(\boldsymbol{\tau})-b$
Estimates on $F_{\tau}^{\prime}(\boldsymbol{\tau})$

$$
\left\|F_{\tau}^{\prime}(\boldsymbol{\tau})\right\|,\left\|F_{\tau}^{\prime}(\boldsymbol{\tau})\right\|^{-1}<C
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uniformly w.r.t. $\tau$ and the shape of $\beta$.


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Corollaries:

- Control of $\operatorname{cond}\left(F_{\tau}^{\prime}\right)$
- Justified stopping criterion:

$$
\left\|F_{\tau}(\boldsymbol{\tau})\right\|<\epsilon \Rightarrow\left\|\boldsymbol{\tau}-\boldsymbol{\tau}_{\star}\right\|<C \epsilon \Rightarrow\left\{\begin{array}{l}
\left\|\bar{v}(\boldsymbol{\tau})-v_{\star}\right\|<C \epsilon \\
\left\|\bar{u}(\boldsymbol{\tau})-u_{\star}\right\|<C \epsilon
\end{array}\right.
$$

# Nonlinear Jacobi preconditioning 

## Recap on various formulations




$$
\begin{array}{lll}
u-\text { formulation : } & \beta(\mathbf{u})+A \mathbf{u}-\mathbf{b} & =0 \\
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- $u$-formulations: catastrophic performance, but convergence theorem
- $\tau$-formulations: excellent performance, but no convergence theorem


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- $u$-formulations: catastrophic performance, but convergence theorem
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Can we have both performance and convergence result?

## Nonlinear Jacobi preconditioner

Nonlinear Jacobi method:

- Separate diagonal and off-diagonal terms

$$
\underbrace{\beta(\mathbf{u})+\operatorname{diag}(A) \mathbf{u}}_{f(\mathbf{u})}+\underbrace{(A-\operatorname{diag}(A)) \mathbf{u}}_{B \mathbf{u}}=\mathbf{b}
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■ Use fixed-point iterations

$$
\mathbf{u}_{k+1}=g\left(\mathbf{b}-B \mathbf{u}_{k}\right), \quad g=f^{-1}
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Our idea: Use Jacobi method as preconditioner not as a solver

- Left preconditioned method: apply Newton to

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with $\boldsymbol{\xi}=f(\mathbf{u})$

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with $\boldsymbol{\xi}=f(\mathbf{u})$
Preconditioned methods satisfy MNT: note that $B \leqslant 0$.

## Efficiency of the preconditioned methods

Left-preconditioned:
Right-preconditioned:

$$
\mathbf{u}-g(\mathbf{b}-A \mathbf{u})=0
$$

$$
\boldsymbol{\xi}+A g(\boldsymbol{\xi})-\mathbf{b}=0
$$



■ Left and right preconditioned methods beat $\tau$ - formulation!

## CPU time efficiency

Preconditioned methods have to evaluate $g=f^{-1}$ :

- At each Newton's iteration one solves $N$ uncoupled equations

How expensive is that?



Relative error versus CPU time for different grid sizes:
$\tau-$ formulation $=$ dashed lines
preconditioned method $=$ solid lines
■ Efficient for all except very small problems ( $N \gtrsim 400$ ) because less linear solves

## Conclusion

Nonlinear Jacobi preconditioning

- accelerates convergence of Newton's method,
- while preserving monotone convergence





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## Extensions and perspectives

Inexact preconditioning (B. ' $20+\epsilon$ )

Non diagonal nonlinearities and non monotone discretizations (with R. Masson): two-phase flow, heterogeneous media, etc, ...

- Works well with parametrization

■ Ongoing work on Jacobi

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## Newton's method for scalar concave problem

Newton's method for

$$
f(u)=0, \quad u \in \mathbb{R}
$$

- $f$ concave and increasing



## Heterogeneous toy problem

Heterogeneous model PDE

$$
\partial_{t} \beta(u, \mathbf{x})-\triangle u=0
$$

Piece-wise constant $\beta(\cdot, \mathbf{x})$

- $\left.\beta(p, x)\right|_{\Omega_{i}}=\beta_{i}(p), \quad i=1,2$


Multiple variable switching

via simultaneous parametrization of $\beta_{1}(u)$ and $\beta_{2}(u)$

