

A Linearization Technique for Nonlinear Parabolic Problems in Porous Media

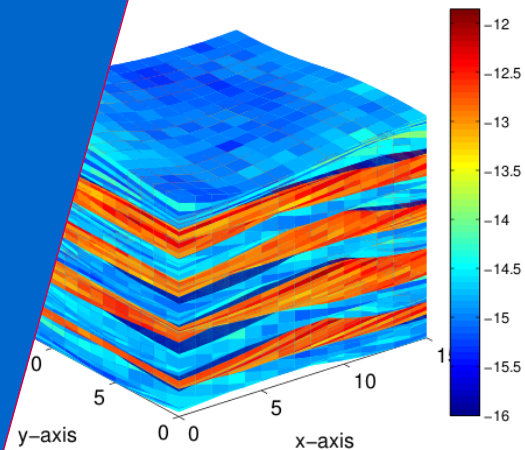
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University of Technology

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Takes initial guess x_0

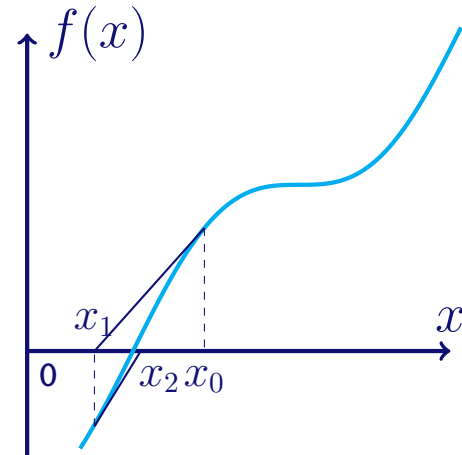
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 Updates for all $i \in \mathbb{N}$

$$x_i = x_{i-1} - \frac{f(x_{i-1})}{f'(x_{i-1})}$$

 The solution being $\lim x_i = \bar{x}$



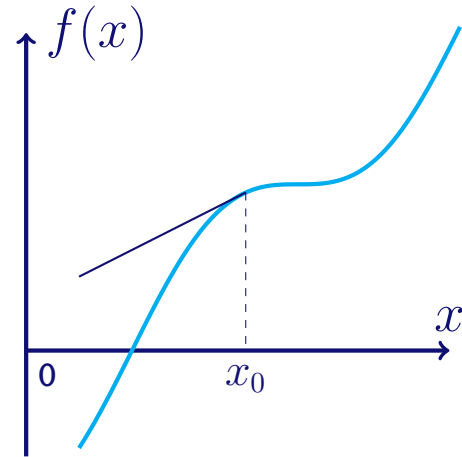
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-  The solution being $\lim x_i = \bar{x}$

However, if x_0 is not close to \bar{x} then the scheme might not converge



If instead one uses the iteration




$$Lx_i = Lx_{i-1} - f(x_{i-1})$$

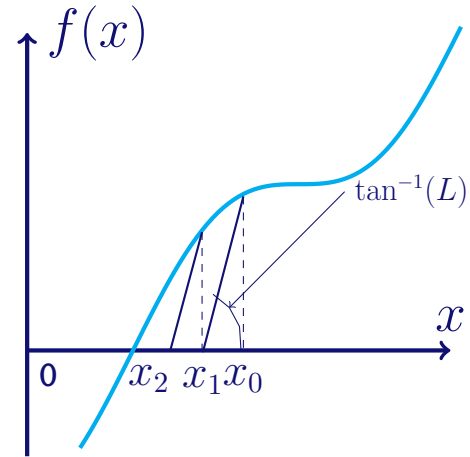
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for $L > \max_{x \in \mathbb{R}} \{f'(x)\}$, then

-  Iterations converge irrespective of initial guess
-  Errors decrease monotonically
-  However, the convergence is slower (linear)



Learning from above we propose

$$L^i x_i = L^i x_{i-1} - f(x_{i-1})$$



with $L^i = f'(x_{i-1}) + \mathfrak{M}$, $\mathfrak{M} > 0$ being a tolerance.

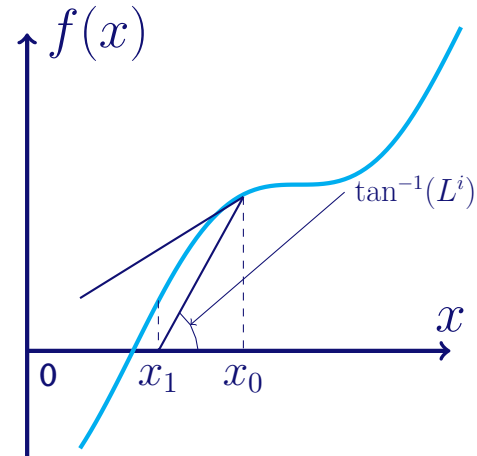
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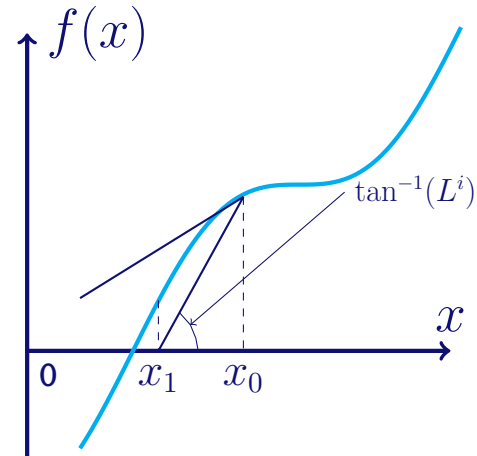
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We look for such a scheme for nonlinear PDEs in the study of porous flows



Richards Equation

$$\partial_t S_w = \nabla \cdot [k_w(S_w)(\nabla p - \rho_w \hat{g})], \quad -p = P_c(S_w)$$



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The two-phase porous media equation

$$\begin{aligned} \partial_t S_\alpha &= \nabla \cdot [k_\alpha(S_\alpha)(\nabla p_\alpha - \rho_\alpha \hat{g})], \quad \alpha \in \{o, w\} \\ S_o + S_w &= 1, \quad p_o - p_w = P_c(S_w) \end{aligned}$$



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Non-equilibrium effects: hysteresis and dynamic capillarity

$$-p \text{ or } p_o - p_w \in P_c(S_w) - \gamma(S_w) \text{sign}(\partial_t S_w) - \mathcal{T}(S_w) \partial_t S_w$$



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Domain decomposition schemes for unsaturated and two-phase cases (Seus *et al.* (2018))

Let us talk about the nonlinear advection diffusion equation

$$\partial_t b(u) + \nabla \cdot \mathbf{F}(\mathbf{x}, u) = \nabla \cdot [\mathcal{D}(\mathbf{x}, u) \nabla u] + r(\mathbf{x}, t, u)$$

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for a fixed n is an elliptic equation of the form

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For example



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L-scheme

To solve

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Approximation of the nonlinearities using the last iteration



Generally they converge if the initial guess u_n^0 is close enough to u_n

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- A **severe restriction**: for $d \geq 2$, for processes that involve large time scales or fine mesh-resolution, e.g. reservoir modelling

^aRadu *et al.* (2006)

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Theorem

If $\mathcal{B}' \geq 0$; $\partial_u \mathcal{R} \leq 0$; $\mathcal{D}, \mathbf{F}_i \in C^1(\Omega \times \mathbb{R})$; $0 < \mathcal{D}_m \leq \mathcal{D} \leq \mathcal{D}_M$ then there exists a τ_0 and L_0 (independent of meshsize) s.t. for all $\tau < \tau_0$ and $L > L_0$, L-scheme converges linearly in $H^1(\Omega)$ irrespective of the initial guess.

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The convergence speed is substantially less^b for $L \gg 1$ or τ small

^aPop et al. (2004)



^bList and Radu. (2016)

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- $b' \geq m \geq 0$
 - $\partial_u r \leq 0$
 - $0 < \mathcal{D}_m \leq \mathcal{D} \leq \mathcal{D}_M$
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- Translates to $\|\partial_t u\|_{L^\infty(\Omega \times (0, T])} < \infty$
- This holds for sufficiently regular domains, ICs and BCs: e.g. if $u_0 \in \mathcal{C}^2(\Omega)$

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Lemma 1.1 With $u_n^0 = u_{n-1}$ and (A1)-(A3)

$$\|u_n^i - u_n\|_{L^\infty(\Omega)} < \Lambda\tau$$

for all $i \in \mathbb{N}$

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Theorem 1.1 With $u_n^0 = u_{n-1}$ and $\mathfrak{M} > \mathfrak{M}_0 = \Lambda \max_{u \in \mathbb{R}} \{|b''| + \tau |\partial_{uu} r|\}$

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Theorem 1.1 With $u_n^0 = u_{n-1}$ and $\mathfrak{M} > \mathfrak{M}_0 = \Lambda \max_{u \in \mathbb{R}} \{|b''| + \tau |\partial_{uu}r|\}$



The \mathfrak{M} -scheme converges linearly in $H^1(\Omega) \cap L^\infty(\Omega)$ for all $\tau > 0$, $m \geq 0$ with convergence rate

$$\alpha = \sup \frac{\|u_n^i - u_n\|_\chi}{\|u_n^{i-1} - u_n\|_\chi} \leq \sqrt{\frac{2\mathfrak{M}}{2\mathfrak{M} + C_\Omega \mathcal{D}_m}}, \quad \chi \in \{H^1(\Omega), L^\infty(\Omega)\}$$

The scheme

$$L_n^i(\mathbf{x})u_n^i - \tau \nabla \cdot (\mathcal{D} \nabla u_n^i) = L_n^i(\mathbf{x})u_n^{i-1} - (b(u_n^{i-1}) - b(u_{n-1})) + \tau r(\mathbf{x}, n\tau, u_n^i)$$


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 If $m > 0$ and $\tau < \tau_0 = \frac{m}{2\mathfrak{M}}$ then the convergence rate is $\mathcal{O}(\tau)$

Time-discrete equation

$$b(u_n) - b(u_{n-1}) + \tau \nabla \cdot \mathbf{F}(\mathbf{x}, u_n) = \tau \nabla \cdot [\mathcal{D}(\mathbf{x}, u_n) \nabla u_n] + \tau r(\mathbf{x}, n\tau, u_n)$$

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$$L_n^i (u_n^i - u_{n-1}^i) - \tau \nabla \cdot (\mathcal{D}_n^{i-1} \nabla u_n^i) = -(b(u_n^{i-1}) - b(u_{n-1}^i)) + \tau [r_n^{i-1} - \nabla \cdot \mathbf{F}_n^{i-1}]$$

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Assumptions:

A4. $\|\nabla u_n\|_{L^\infty(\Omega)} \leq \Lambda_1$ for some $\Lambda_1 > 0$

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- Holds if $u_0 \in W^{2,2q}(\Omega)$, $q \in \mathbb{N}$, $2q > d$

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For $u_n^0 = u_{n-1}$, $\mathfrak{M} > \mathfrak{M}_0$ and $\tau < \tau_0$ assume (A1)-(A4)*. Then

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Richards equation in 2-D

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Take van Genuchten parameters^a: for $m = \frac{2}{3}$, $n = \frac{1}{1-m}$

$$S_w(p) = \begin{cases} \frac{1}{(1 + (-p)^n)^m} & \text{if } p < 0 \\ 1 & \text{if } p \geq 0 \end{cases}$$

$$k_w(S) = \sqrt{S}(1 - (1 - S^{\frac{1}{m}})^m)^2$$

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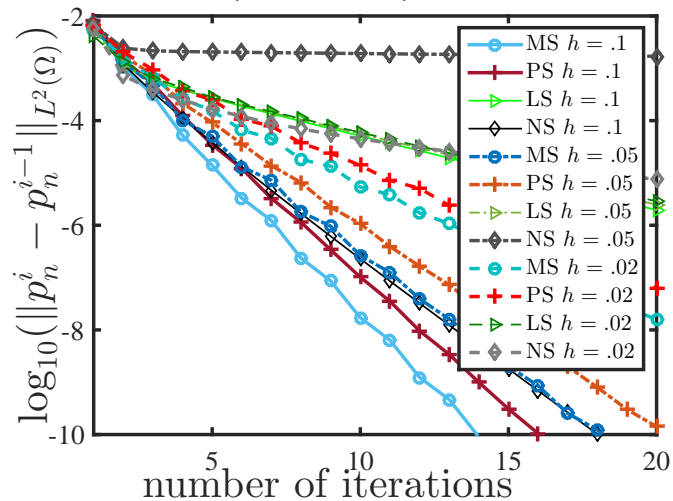
$$k_w(S) = \sqrt{S}(1 - (1 - S^{\frac{1}{m}})^m)^2$$

Assumed initial and boundary conditions with $\tilde{p}(x, y, t) = 1 - (1 + t^2)(1 + x^2 + y^2)$,

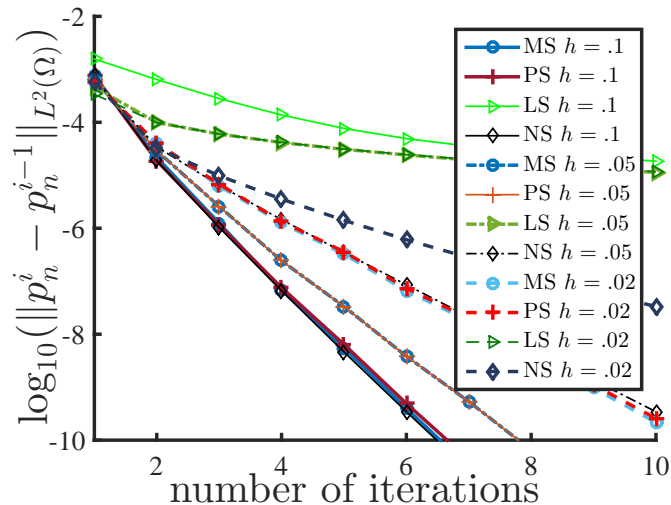
IC	$t = 0$	$p(x, y, 0) = \tilde{p}(x, y, 0)$	on Ω
BC	$x = 0$:	$p(0, y, t) = \tilde{p}(0, y, t)$,	$x = 1$: $p(1, y, t) = \tilde{p}(1, y, t)$,
	$y = 0$:	$\partial_y p = 0$,	$y = 1$: $k(S(p))\partial_y p = k(S(\tilde{p}(x, 1, t)))\partial_y \tilde{p}(x, 1, t)$.

^avan Genuchten. (1980)

- For $t = 0.5$, $\mathfrak{M} = 10$, $L = 1$

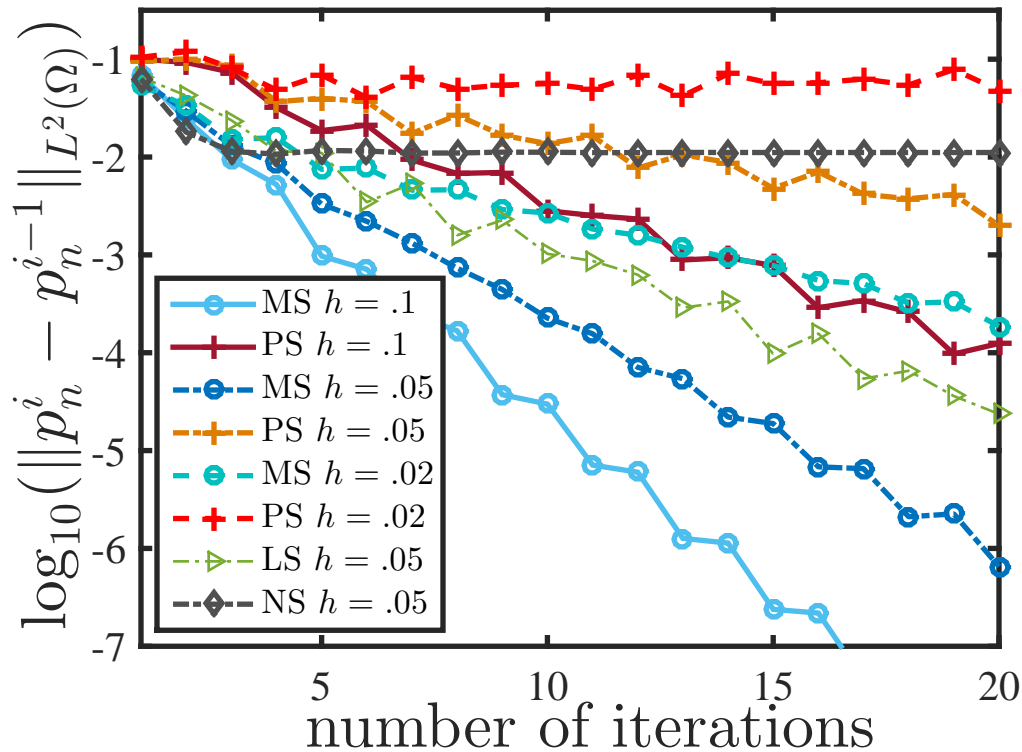


$\tau = 0.01$



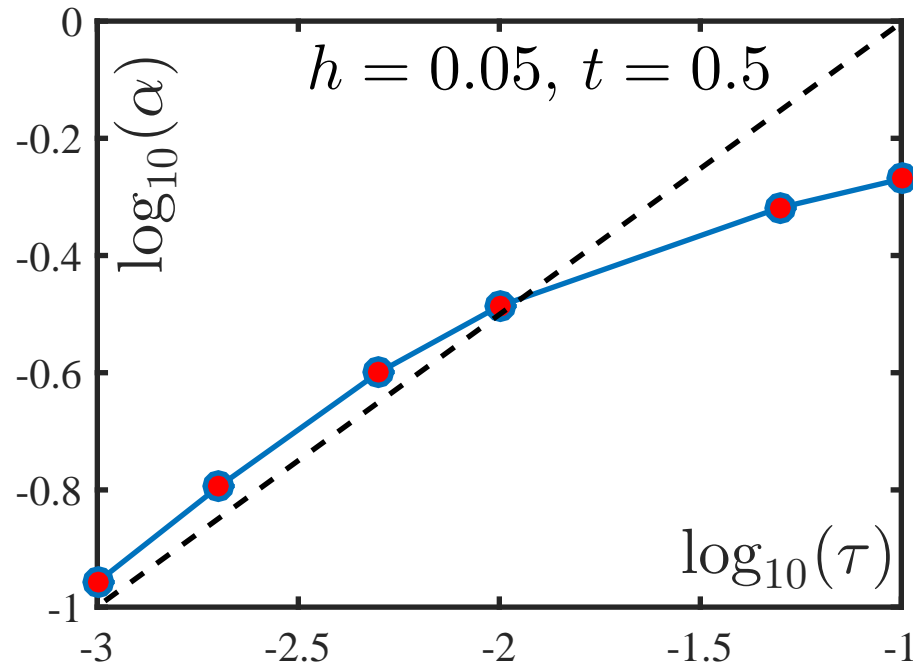
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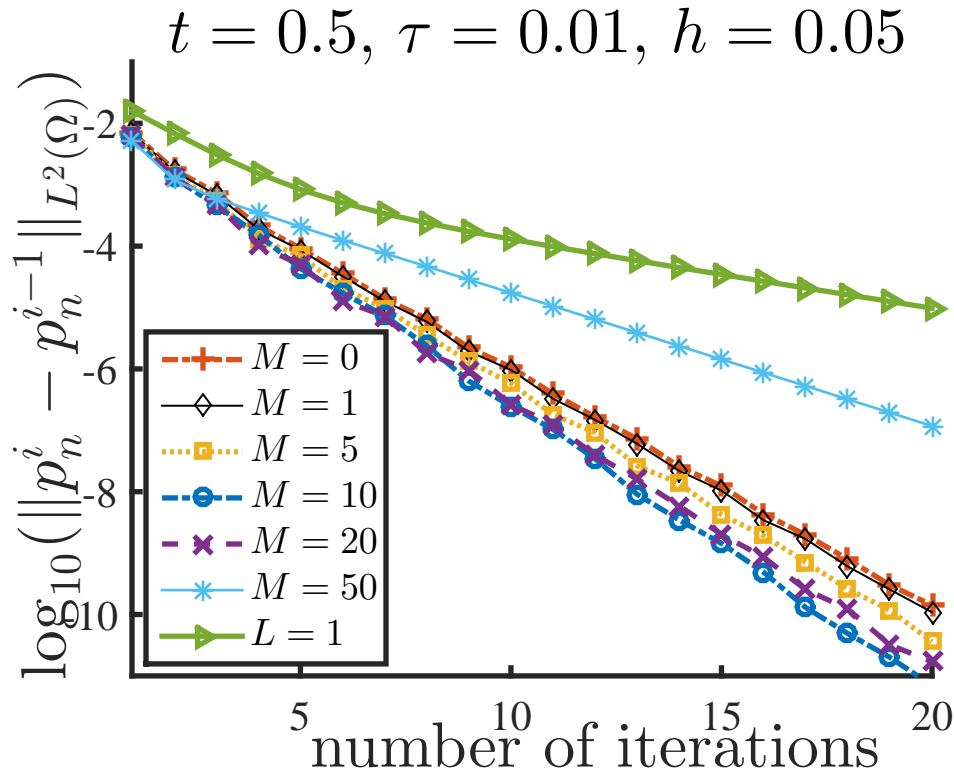
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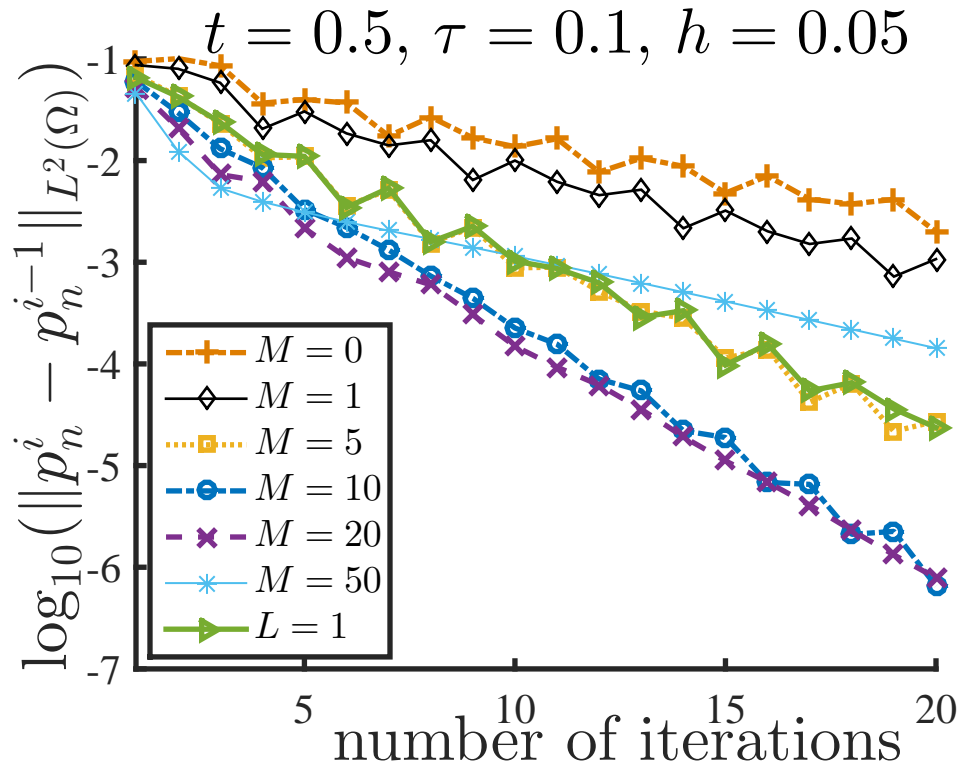


$\tau = 0.1$

- For $t = .5$, $h = 0.05$, $\mathfrak{M} = 10$







see

Mitra, K. & Pop, I. S. (2018). A modified L-scheme for nonlinear parabolic equations. *Computers & Mathematics With Applications*.

Two Phase Equation: The \mathfrak{M} -scheme given as

$$\begin{aligned} -(S_{w,n}^i - S_{w,n-1}) &= \tau \nabla \cdot [k_o(1 - S_{w,n}^{i-1})(\nabla p_{o,n}^i - \rho_o \hat{g})] \\ (S_{w,n}^i - S_{w,n-1}) &= \tau \nabla \cdot [k_w(S_{w,n}^{i-1})(\nabla p_{w,n}^i - \rho_w \hat{g})] \\ p_{o,n}^i - p_{w,n}^i &= P_c(S_{w,n}^{i-1}) - L_n^i(S_{w,n}^i - S_{w,n-1}^i) \end{aligned}$$

with $L_n^i := -P_c'(S_{w,n}^{i-1}) + \mathfrak{M}\tau$

Theorem 3.1 With $(p_{o,n}^0, p_{w,n}^0) = (p_{o,n-1}, p_{w,n-1})$ define

$$e_n^i = \|p_{w,n}^i - p_{w,n}\|_{H^1(\Omega)} + \|p_{o,n}^i - p_{o,n}\|_{H^1(\Omega)} + \|S_{w,n}^i - S_{w,n}\|_{L^2(\Omega)}.$$

Assume for $i \in \mathbb{N}$, $p_n \in W^{1,\infty}(\Omega)$ and $\|S_n^i - S_n\|_{L^\infty(\Omega)} < \Lambda\tau$ for some $\Lambda > 0$. Then $e_n^i \rightarrow 0$ as $i \rightarrow \infty$ for τ small enough and \mathfrak{M} large enough. Moreover, if $P_c'(S) < 0$ and $P_c \in C^2(\mathbb{R})$ then for small enough τ

$$\frac{e_n^i}{e_n^{i-1}} = \mathcal{O}(\sqrt{\tau}).$$

Capillary Hysteresis (play-type) and Dynamic Capillarity

Richards equation: $\partial_t S_w = \nabla \cdot [k(S_w)(\nabla p - \rho_w \hat{g})]$

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This can be simplified to

$$\partial_t S_w = \mathcal{F}(S_w, p) := \frac{1}{\mathcal{T}(S_w)} \begin{cases} P_c(S_w) - \gamma(S_w) + p & \text{if } p < P_c(S_w) - \gamma(S_w) \\ 0 & \text{if } p \in [P_c - \gamma, P_c + \gamma](S_w) \\ P_c(S_w) + \gamma(S_w) + p & \text{if } p > P_c(S_w) + \gamma(S_w) \end{cases}$$

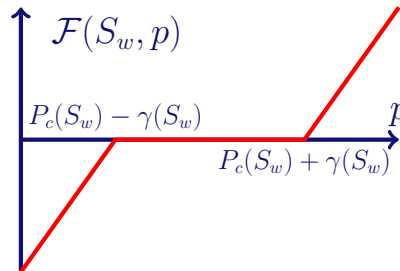
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Time-discrete version

$$S\text{-equation: } S_{w,n} = S_{w,n-1} + \tau \mathcal{F}(S_{w,n}, p_n)$$

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Solution strategy: \mathfrak{M} -scheme

$$\text{Update: } S_{w,n}^i = S_{w,n-1} + \tau \mathcal{F}(S_{w,n}^{i-1}, p_n^{i-1})$$

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Theorem 4.1 For small enough τ , large enough \mathfrak{M} , $p_n \in W^{1,\infty}(\Omega)$, there exists a $\alpha = \mathcal{O}(\tau/\mathcal{T})$ such that

$$\|S_{w,n}^i - S_{w,n}\|_{W^{1,\infty}} + \|p_n^i - p_n\|_{W^{1,\infty}} \leq \alpha [\|S_{w,n}^{i-1} - S_{w,n}\|_{W^{1,\infty}} + \|p_n^{i-1} - p_n\|_{W^{1,\infty}}]$$

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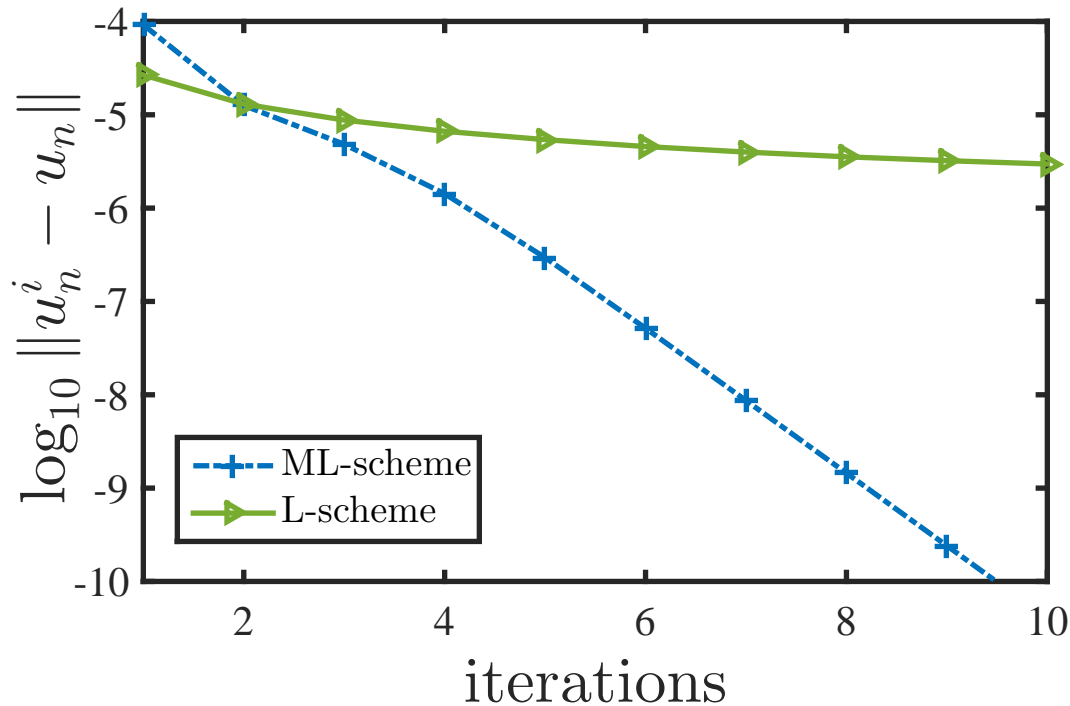
$$\text{Update: } S_{w,n}^i = S_{w,n-1} + \tau \mathcal{F}(S_{w,n}^{i-1}, p_n^{i-1})$$

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$$\text{With } L_n^i := \partial_p \mathcal{F}(S_{w,n}^i, p_n^{i-1}) + \mathfrak{M} \tau$$

Theorem 4.2 If $\mathcal{T} > 0$ then there exists $\hat{\tau} > 0$ independent of \mathcal{T} such that for $\tau < \hat{\tau}$ and large enough \mathfrak{M} , $(S_{w,n}^i, p_n^i)$ converges in $H^1(\Omega)$

- For^a $\mathcal{T} = .1$, $L = 100$, $\mathfrak{M} = 1$, $h = .1$, $\tau = .001$, $t = 10$







\mathfrak{M} -scheme $\alpha \approx .15$



^avan Duijn, Mitra and Pop. (2018)










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-  Using $u_n^0 = u_{n-1}$ and local estimations one can have a scheme that has the following properties

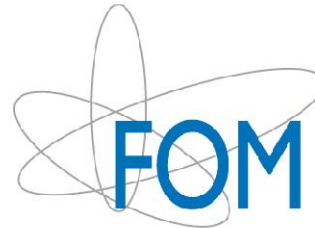
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and Thank You for listening