

Improved Newton Linearization for L^1 -Norm-Type Minimization with Application to Viscoplastic Fluid Solvers

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Summary

- Our challenge: nonlinear applications modelled by optimization problems with singularities in the Hessian due to a L^1 -norm type term in the objective.
- Hessian exhibits a null space upon linearization with Newton's method that is problematic for convergence.
- Applications include inverse problems with total variation regularization as well as viscoplastic flows like Earth's mantle convection.
- We analyze issues with the standard Newton linearization through numerical experiments and theoretically in an abstract setting.
- We propose an improved linearization based on a perturbation of an otherwise implicitly assumed equality constraint.
- We achieve robust and fast Newton convergence independent of the discretization. We further show advantageous theoretical properties of the new linearization.

Introduction in 1 Dimension

Minimization problem Given constants $a > 0$, $b \in \mathbb{R}$, and $0 \leq \theta \leq 1$,

$$\text{find minimizer } x^* \in \mathbb{R} \text{ of } \min_{x \in \mathbb{R}} J(x) := \frac{a}{1+\theta} |x|^{1+\theta} - bx$$

Denote $|x|_\theta := (|x|^2 + \epsilon^2)^{1/2}$. Note that the parameter θ controls the nonlinearity such that Newton's method *converges well* for $\theta = 1$; and Newton's method *converges poorly* for $\theta \rightarrow 0$. The gradient and Hessian of $J(x)$ are

$$g(x) := J'(x) = \frac{ax}{|x|_\theta^{1-\theta}} - b, \quad H(x) := J''(x) = \frac{a}{|x|_\theta^{1-\theta}} \left(1 - (1-\theta) \frac{x^2}{|x|_\theta^2} \right).$$

Standard Newton linearization One Newton step (simplified, e.g., without backtracking line search) is

$$\text{solve for } \hat{x} \text{ in } H(x) \hat{x} = -g(x), \text{ update } x \leftarrow x + \hat{x},$$

therefore, we need to solve the linearized equation

$$\frac{a}{|x|_\theta^{1-\theta}} \left(1 - (1-\theta) \frac{x^2}{|x|_\theta^2} \right) \hat{x} = - \left(\frac{ax}{|x|_\theta^{1-\theta}} - b \right) \quad (\text{standard Newton step})$$

Perturbed Newton linearization Define for $x, y \in \mathbb{R}$

$$\text{model error: } E(x, y) := \frac{x}{|x|_\theta^{1-\theta}} - y,$$

$$\text{model perturbation: } D(x, y) := x - |x|_\theta^{1-\theta} y.$$

The perturbed Newton linearization for \hat{x} and \hat{y} assumes $D(x, y) \neq 0$, then

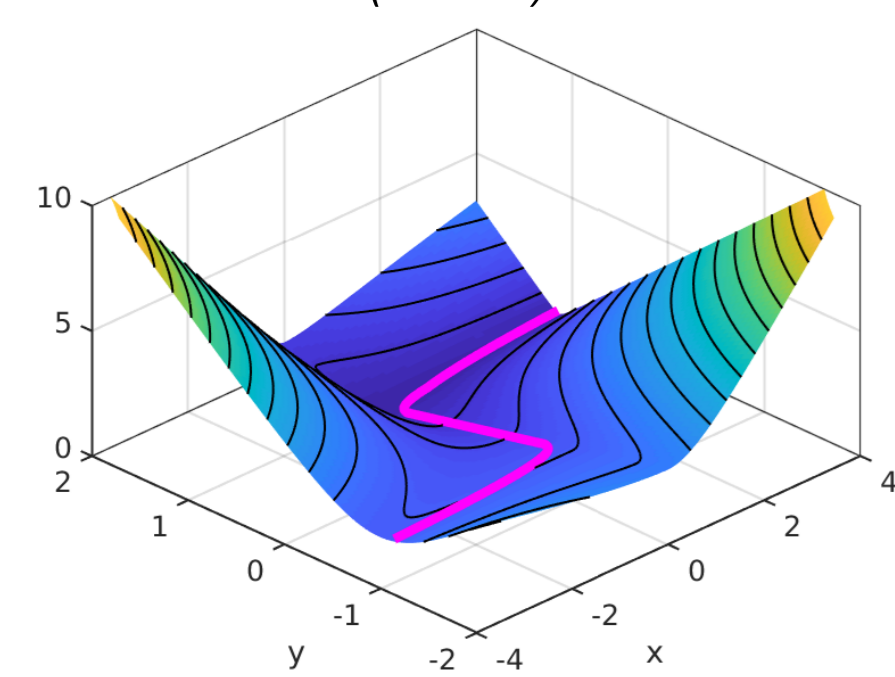
$$a\hat{y} = -(ay - b),$$

$$\left(1 - (1-\theta) \frac{xy}{|x|_\theta^{1+\theta}} \right) \hat{x} - |x|_\theta^{1-\theta} \hat{y} = - \left(x - |x|_\theta^{1-\theta} y \right).$$

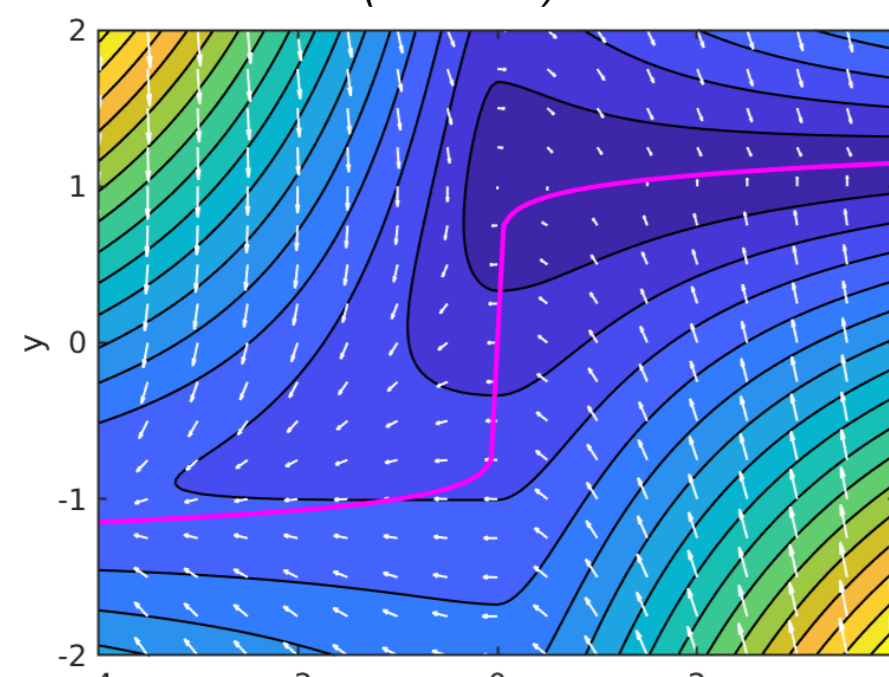
Substituting for \hat{y} gives the perturbed and reduced Newton linearization for \hat{x}

$$\frac{a}{|x|_\theta^{1-\theta}} \left(1 - (1-\theta) \frac{xy}{|x|_\theta^{1+\theta}} \right) \hat{x} = - \left(\frac{ax}{|x|_\theta^{1-\theta}} - b \right) \quad (\text{improved Newton step})$$

Magnitude of perturbed gradient (colors)



Directions of perturbed gradient (arrows)



The magenta curve in both plots represents the (absolute value of) the unperturbed gradient $g(x)$ in (x, y) -space, meaning $E(x, y) = 0$. Introducing the perturbation, $D(x, y) \neq 0$, relaxes the rapid changes of the unperturbed $g(x)$ around $x = 0$.

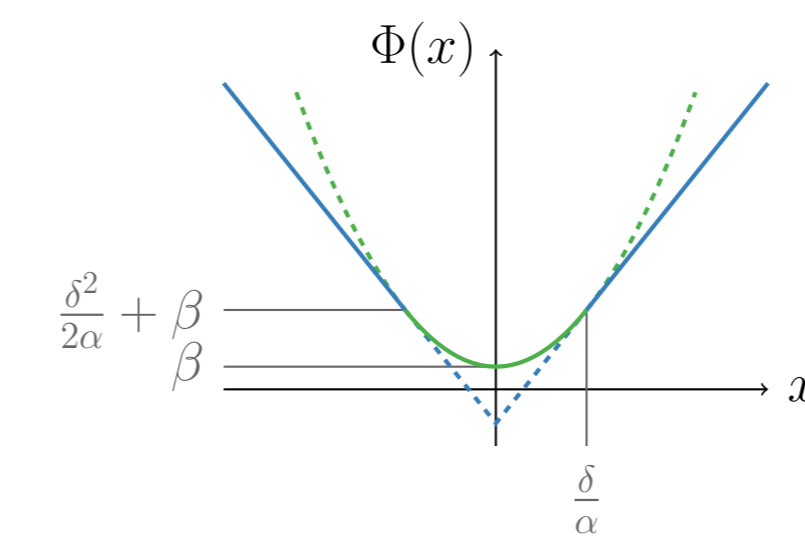
Abstract Derivation of Perturbed Newton Linearizations

Minimization problem Given dimensions $d \in \mathbb{N}$ and $n \in \{1, d, d \times d\}$, domain $\Omega \subseteq \mathbb{R}^d$, and a linear and bounded functional $F : L^2(\Omega)^n \rightarrow \mathbb{R}$,

$$\text{find minimizer } U^* : \Omega \rightarrow \mathbb{R}^n \text{ of } \min_{U \in \mathbb{R}^n} J(U) := \int_{\Omega} \Phi(U) - F(U)$$

using a generalization of the *Huber loss*:

$$\Phi(U) := \begin{cases} \frac{1}{2} \alpha |U|^2 + \beta, & \alpha |U| \leq \delta, \\ \delta |U| - \frac{1}{2} \frac{\delta^2}{\alpha} + \beta, & \text{otherwise,} \end{cases}$$



with parameters $\alpha, \delta > 0$ and $\beta \in \mathbb{R}$.

Standard Newton linearization The gradient and Hessian are the 1st- and 2nd-order variations of J :

$$g(U) \hat{U} := \int_{\Omega} \left(\chi \alpha + (1-\chi) \frac{\delta}{|U|} \right) \langle U, \hat{U} \rangle - F(\hat{U}), \quad \chi := \begin{cases} 1, & \alpha |U| \leq \delta, \\ 0, & \text{o.w.}, \end{cases}$$

$$(H(U) \hat{U}, \hat{U}) := \int_{\Omega} \left(\chi \alpha + (1-\chi) \frac{\delta}{|U|} \left(\mathbf{I} - \frac{U \otimes U}{|U|^2} \right) \right) \langle \hat{U}, \hat{U} \rangle.$$

The standard Newton linearization requires to solve for Newton step \hat{U} in

$$\int_{\Omega} \left(\chi \alpha + (1-\chi) \frac{\delta}{|U|} \left(\mathbf{I} - \frac{U \otimes U}{|U|^2} \right) \right) \langle \hat{U}, \hat{U} \rangle = -g(U) \hat{U}$$

The outer product term is computationally challenging because:

- Coefficient $\left(\mathbf{I} - \frac{U \otimes U}{|U|^2} \right)$ in Hessian represents an **orthogonal projector**.
- Hessian has a **zero eigenvalue** associated to eigenvector $(1-\chi)U$.

Perturbed Newton linearization Define for $U, S \in L^2(\Omega)^n$

$$\text{model error: } E(U, S) := \frac{U}{|U|} - S,$$

$$\text{model perturbation: } D(U, S) := U - |U|S.$$

We augment the previous gradient by a model perturbation, $D(U, S) \neq 0$,

$$g(U, S) \hat{U} := \int_{\Omega} \langle \chi \alpha U + (1-\chi) \delta S, \hat{U} \rangle - F(\hat{U}),$$

$$(D(U, S), \hat{S}) := \int_{\Omega} \langle U - |U|S, \hat{S} \rangle,$$

to get the **perturbed Newton linearization** for step (\hat{U}, \hat{S})

$$\int_{\Omega} \langle \chi \alpha \hat{U} + (1-\chi) \delta \hat{S}, \hat{U} \rangle = -g(U, S) \hat{U},$$

$$\int_{\Omega} \left\langle \left(\mathbf{I} - \frac{U \otimes S}{|U|} \right) \hat{U} - |U| \hat{S}, \hat{S} \right\rangle = -(D(U, S), \hat{S}).$$

Here, the dual step has the explicit expression

$$\hat{S} = \frac{U}{|U|} - S + \frac{1}{|U|} \left(\mathbf{I} - \frac{U \otimes S}{|U|} \right) \hat{U}.$$

Therefore, substitution leads to the **perturbed and reduced Newton linearization** for \hat{U} , where only the outer product term of the Hessian has changed compared to the standard linearization:

$$\int_{\Omega} \left(\chi \alpha + (1-\chi) \frac{\delta}{|U|} \left(\mathbf{I} - \frac{U \otimes S}{|U|} \right) \right) \langle \hat{U}, \hat{U} \rangle = -g(U) \hat{U}$$

Improvements gained from perturbation

- Perturbation results in **model error dependent regularization**:

$$\mathbf{I} - \frac{U \otimes S}{|U|} = \mathbf{I} - \frac{U \otimes U}{|U|^2} + \frac{U \otimes E(U, S)}{|U|}.$$

- Perturbed linearization acts as a **nonlinear preconditioner** far from the solution while enabling fast Newton convergence close to solution, since

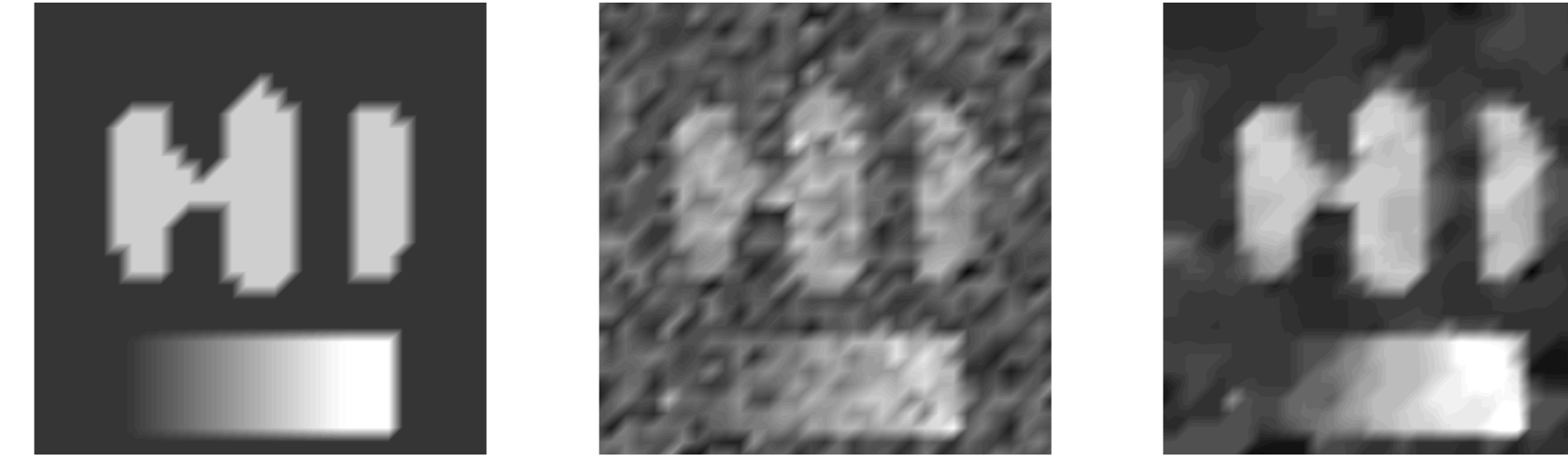
$$\mathbf{I} - \frac{U \otimes S}{|U|} \rightarrow \mathbf{I} - \frac{U \otimes U}{|U|^2} \text{ as } E(U, S) \rightarrow 0.$$

- The explicit expression of \hat{S} enables a **simple and computationally cheap** update of the dual variable S after computing step \hat{U} :

$$S \leftarrow S + \hat{S} = \frac{U}{|U|} + \frac{1}{|U|} \left(\mathbf{I} - \frac{U \otimes S}{|U|} \right) \hat{U}.$$

Image Restoration with Total Variation (TV)

True image Noisy image Restored image



A true image is generated (left), to which Gaussian noise with standard deviation $\sigma = 0.2$ is added (middle). The noisy image serves as input data to the image restoration problem with the result shown right.

Minimization problem Given blurry and noisy image data, $d : (0, 1)^2 \rightarrow \mathbb{R}$, the blurring operator, $B : L^2(\Omega) \rightarrow L^2(\Omega)$, $\Omega = (0, 1)^2$, and the total variation operator

$$\Phi(\nabla u) = \begin{cases} \frac{1}{2\epsilon} |\nabla u|^2, & |\nabla u| \leq \epsilon, \\ |\nabla u| - \frac{\epsilon}{2}, & \text{otherwise,} \end{cases}$$

with Huber parameters $\alpha = 1/\epsilon$, $\delta = 1$, $\beta = 0$, where $0 < \epsilon \ll 1$; consider

$$\text{find minimizer } u^* : (0, 1)^2 \rightarrow \mathbb{R} \text{ of } \min_u J(u) := \frac{1}{2} \|Bu - d\|_{L^2}^2 + \gamma \Phi(\nabla u)$$

Note: The regularization with $\Phi(\nabla u)$ has the advantage of preserving edges in the restored image u^* . The parameters $\epsilon > 0$ and $\gamma > 0$ regularize the TV operator and the image restoration problem, respectively.

Standard Newton linearization The 1st- and 2nd-order variations of $\Phi(\nabla u)$ are

$$\delta_u[\Phi(\nabla u)](\hat{u}) = \int_{\Omega} \left(\frac{\chi}{\epsilon} + \frac{(1-\chi)}{|\nabla u|} \right) \nabla u \cdot \nabla \hat{u},$$

$$\delta_u \delta_u[\Phi(\nabla u)](\hat{u})(\hat{u}) = \int_{\Omega} \left(\frac{\chi}{\epsilon} + \frac{(1-\chi)}{|\nabla u|} \left(\mathbf{I} - \frac{\nabla u \otimes \nabla u}{|\nabla u|^2} \right) \right) \nabla \hat{u} \cdot \nabla \hat{u}.$$

The standard Newton linearization requires to solve for Newton step \hat{u} in

$$-\nabla \cdot \left[\gamma \left(\frac{\chi}{\epsilon} + \frac{(1-\chi)}{|\nabla u|} \left(\mathbf{I} - \frac{\nabla u \otimes \nabla u}{|\nabla u|^2} \right) \right) \nabla \hat{u} \right] + B^* B \hat{u} = -\gamma \delta_u[\Phi(\nabla u)] - B^*(Bu - d)$$

The left-hand side constitutes a Poisson operator with an anisotropic 2nd-order tensor coefficient.

Perturbed Newton linearization Define for $u \in L^2(\Omega)$, $S \in L^\infty(\Omega)$, $\|S\|_{L^\infty} \leq 1$

$$\text{model error: } E(u, S) := \frac{\nabla u}{|\nabla u|} - S,$$

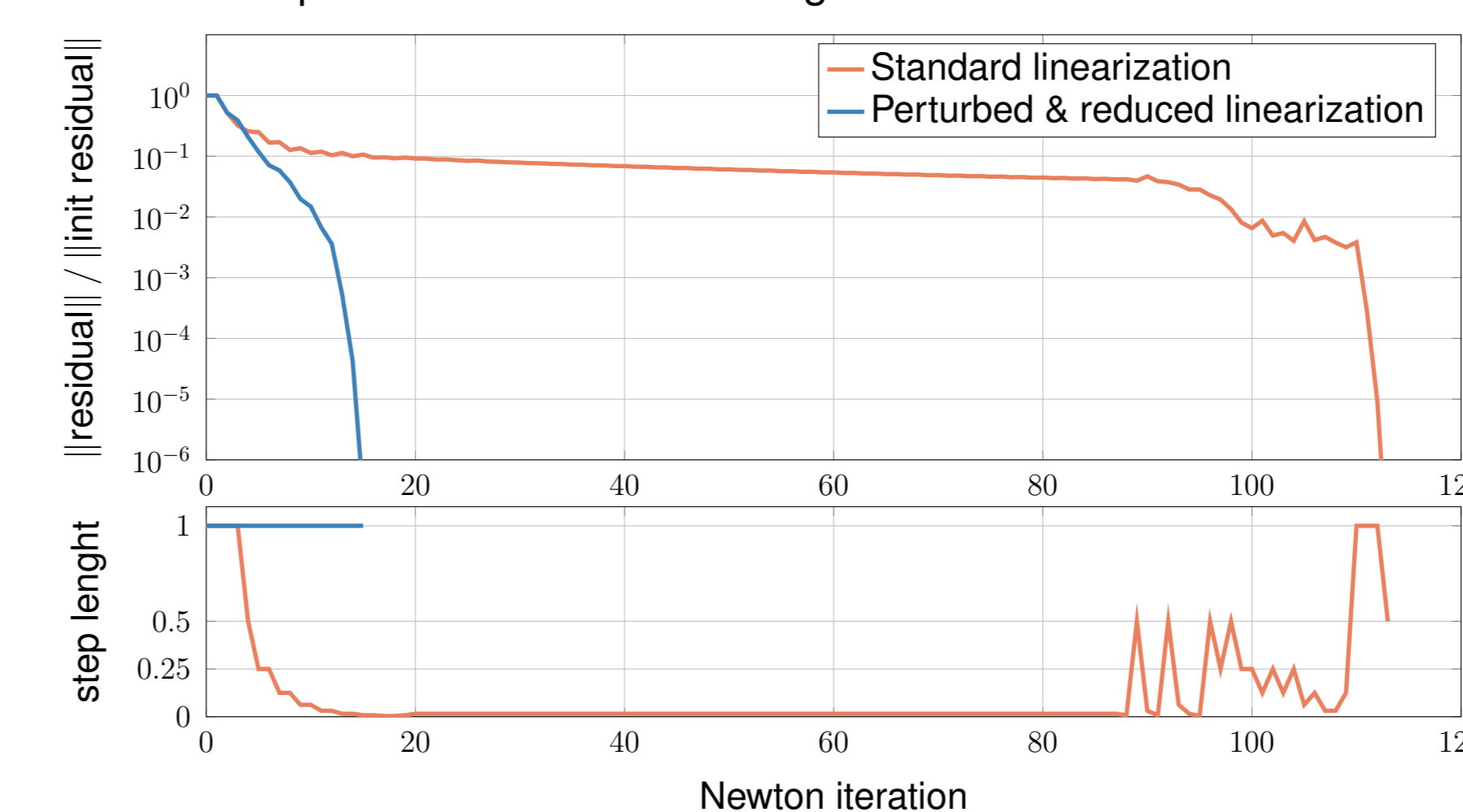
$$\text{model perturbation: } D(u, S) := \nabla u - |\nabla u|S.$$

The perturbed and reduced Newton linearization for \hat{u} is the system

$$-\nabla \cdot \left[\gamma \left(\frac{\chi}{\epsilon} + \frac{(1-\chi)}{|\nabla u|} \left(\mathbf{I} - \frac{\nabla u \otimes S}{|\nabla u|} \right) \right) \nabla \hat{u} \right] + B^* B \hat{u} = -\gamma \delta_u[\Phi(\nabla u)] - B^*(Bu - d)$$

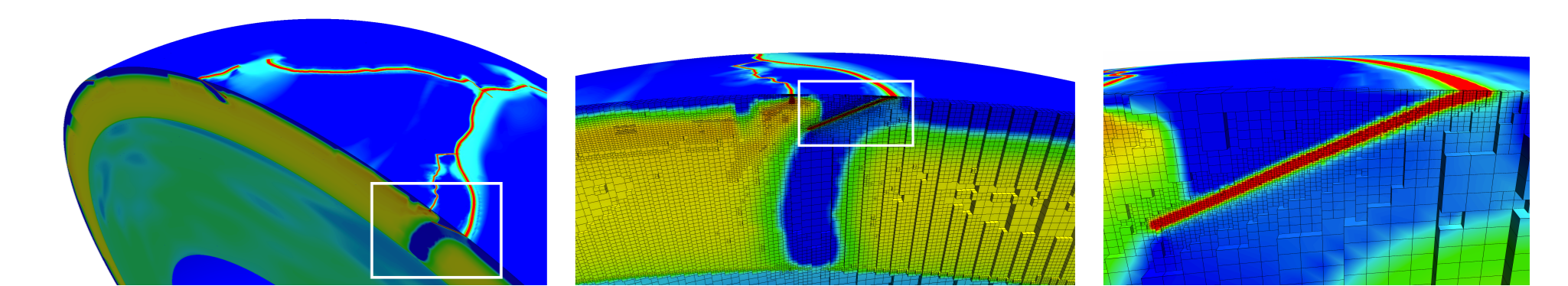
Numerical experiments Comparison of Newton convergence with standard and perturbed linearizations using parameters $\epsilon = 0.01$, $\gamma = 0.002$, and added Gaussian noise to the true image with standard deviation $\sigma = 0.2$.

Comparison of Newton convergence with different linearizations



Viscoplastic Stokes Flow

Motivated by rheology of Earth's mantle convection with plate tectonics:



Effective viscosity field (colors) and viscoplastic dynamic weakening of lithosphere in hinge zone (right zoom). (Visualization by L. Alisic)

Minimization problem Given viscosity $\mu > 0$, yield strength $\tau_{\text{yield}} > 0$, the 2nd invariant of the strain rate tensor, $\dot{\epsilon}_u := \frac{1}{\sqrt{2}} |\nabla_s u|$ with $\nabla_s u := \frac{1}{2} (\nabla u + \nabla u^T)$, and

$$\Phi(2\dot{\epsilon}_u) = \begin{cases} \frac{\mu}{2} (2\dot{\epsilon}_u)^2 + \frac{1}{2} \frac{\tau_{\text{yield}}^2}{\mu}, & \mu(2\dot{\epsilon}_u) \leq \tau_{\text{yield}}, \\ \tau_{\text{yield}}(2\dot{\epsilon}_u), & \text{otherwise,} \end{cases}$$

with Huber parameters $\alpha = \mu$, $\delta = \tau_{\text{yield}}$, $\beta = \frac{1}{2} \frac{\delta^2}{\alpha} = \frac{1}{2} \frac{\tau_{\text{yield}}^2}{\mu}$; consider

$$\text{find minimizer } u^* : (0, 1)^3 \rightarrow \mathbb{R}^3 \text{ of } \min_u J(u) := \int_{\Omega} \Phi(u) - f \cdot u \text{ s.t. } \nabla \cdot u = 0$$

Standard Newton linearization Solve for Newton step (\hat{u}, \hat{p}) in

$$-\nabla \cdot \left[2 \left(\chi \mu + (1-\chi) \frac{\tau_{\text{yield}}}{\sqrt{2} |\nabla_s u|} \left(\mathbf{I} - \frac{\nabla_s u \otimes \nabla_s u}{|\nabla_s u|^2} \right) \right) \nabla_s \hat{u} \right] + \nabla \hat{p} = -r_{\text{mom}}$$

$$-\nabla \cdot \hat{u} = -r_{\text{mass}}$$

The residuals of the nonlinear Stokes momentum and mass equations appear on the right-hand side. What plays the role of viscosity in the Newton step is an anisotropic 4th-order tensor.

Perturbed Newton linearization Define for $u \in H^1(\Omega)^d$ and $S \in L^\infty(\Omega)^{d \times d}$, $\|S\|_{L^\infty} \leq 1$

$$\text{model error: } E(u, S) := \frac{\nabla_s u}{|\nabla_s u|} - S,$$

$$\text{model perturbation: } D(u, S) := \nabla_s u - |\nabla_s u|S.$$

The perturbed and reduced Newton linearization of the momentum equation is

$$-\nabla \cdot \left[2 \left(\chi \mu + (1-\chi) \frac{\tau_{\text{yield}}}{\sqrt{2} |\nabla_s u|} \left(\mathbf{I} - \frac{\nabla_s u \otimes S}{|\nabla_s u|} \right) \right) \nabla_s \hat{u} \right] + \nabla \hat{p} = -r_{\text{mom}}$$

Numerical experiments Comparison of Newton convergence with standard and perturbed linearizations using a model problem with a viscosity that incorporates low-viscosity plumes in a high-viscosity background medium,

$$\mu(x) := (\mu_{\text{max}} - \mu_{\text{min}}) \chi_n(x) + \mu_{\text{min}}, \quad x \in (0, 1)^3,$$

where $\chi_n(x) \in [1, 2]$ are C^∞ indicator functions, which accumulate n plumes via products of Gaussians.

Yielding volume	Mesh level ℓ	Standard Newton		Perturbed Newton			
		It. Newton	#backtr. It. GMRES	It. Newton	#backtr. It. GMRES		
~45%	4	33	20	1469	10	0	379
~45%	5	36	25	2255	12	0	664
~45%	6	57	49	4255	13	0	876
~65%	4	29	21	1559	18	10	965
~65%	5	37	26	2464	17	9	1245
~65%	6	48	39	3892	20	9	1707
~90%	4	35	25	1505	19	11	872
~90%	5	40	32	2147	21	11	1267
~90%	6	32	21	2312	23	11	1811

References

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