

Sensitivity Analysis

Begin with finite dimensional analysis

$$\min_{\mathbf{m} \in \mathbb{R}^M} f(\mathbf{u}(\mathbf{m}), \mathbf{m}) = \frac{1}{2} (\mathbf{B}\mathbf{u} - \mathbf{d}_-)^\top (\mathbf{B}\mathbf{u} - \mathbf{d}_-) + \frac{1}{2} \mathbf{m}^\top \mathbf{R} \mathbf{m}$$

where $\mathbf{r}(\mathbf{u}, \mathbf{m}) = 0$

$\mathbf{d} \in \mathbb{R}^D$; $\mathbf{u} \in \mathbb{R}^N$, $\mathbf{a} \in \mathbb{R}^N$; $\mathbf{B} \in \mathbb{R}^{D \times N}$; $\mathbf{R} \in \mathbb{R}^{M \times M}$

Gradient of f wrt m found from total derivative:

$$g_i = \frac{\partial f}{\partial m_i} = \frac{\partial f}{\partial m_i} + \underbrace{\frac{\partial f^\top}{\partial u} \frac{\partial u}{\partial m_i}}_{\text{sensitivity of state wrt model params}}, \quad i = 1, \dots, M$$

$$= \bullet + \xrightarrow{N}$$

$$g^\top = \frac{\partial f^\top}{\partial m} + \frac{\partial f^\top}{\partial u} \frac{\partial u}{\partial m} \quad (\text{derivatives are row vectors})$$

$$= \xrightarrow{M} + \xrightarrow{N} \frac{\partial f}{\partial u} \begin{array}{c} M \\ \downarrow \\ N \end{array} \begin{array}{c} \frac{\partial u}{\partial m} \\ \downarrow \\ \frac{\partial u}{\partial m_i} \end{array}$$

$$g = \frac{\partial f}{\partial m} + \frac{\partial u^\top}{\partial m} \frac{\partial f}{\partial u} = \begin{array}{c} \downarrow \\ M \end{array} + \begin{array}{c} \xrightarrow{N} \\ \downarrow \end{array} \begin{array}{c} \downarrow \\ N \end{array} \quad \downarrow N \times 1$$

$$\frac{\partial f}{\partial m} \equiv \mathbf{R} \mathbf{m}; \quad \frac{\partial f}{\partial u} = \mathbf{B}^\top (\mathbf{B}\mathbf{u} - \mathbf{d})$$

$$\frac{\partial u}{\partial m} = ? \quad \text{For: } \mathbf{r}(\mathbf{u}, \mathbf{m}) = 0$$

$$\frac{\partial \mathbf{r}}{\partial \mathbf{u}} \frac{\partial \mathbf{u}}{\partial \mathbf{m}} + \frac{\partial \mathbf{r}}{\partial \mathbf{m}} = 0$$

$$\boxed{A} \boxed{\frac{\partial \mathbf{u}}{\partial \mathbf{m}}} = -\boxed{C} \Rightarrow \frac{\partial \mathbf{u}}{\partial \mathbf{m}} = -\boxed{C}^{-1} \frac{\partial \mathbf{r}}{\partial \mathbf{m}} = -A(m, n)^{-1} C(m, n)$$

Special case: $A(m)u = b(n)$

$$\frac{\partial (A\mathbf{u})}{\partial m} + A \frac{\partial \mathbf{u}}{\partial m} = \frac{\partial b}{\partial m} \Rightarrow \frac{\partial \mathbf{u}}{\partial m} = A(m)^{-1} \left[\frac{\partial b}{\partial m} - A \frac{\partial \mathbf{u}}{\partial m} \right]$$

$$g^T = \frac{\partial f^T}{\partial m} + \frac{\partial f^T}{\partial u} \frac{du}{dm}$$

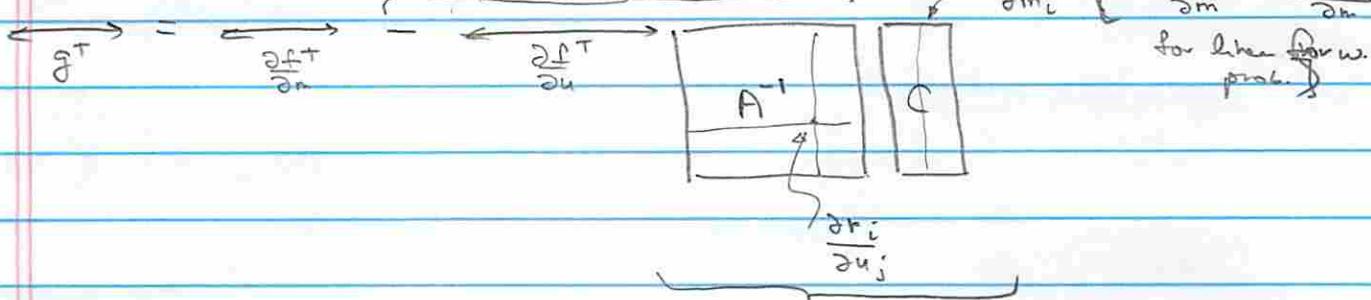
$$= \frac{\partial f^T}{\partial m} - \frac{\partial f^T}{\partial u} A^{-1} \frac{\partial u}{\partial m}$$

straightforward

Jacobian (of res wrt state); common component of Newton forward solver

p^T : adjoint variable

$$\frac{\partial r_i}{\partial m_i} \quad (= \frac{\partial b}{\partial m} - \frac{\partial f}{\partial m})$$



for linear forward prob.

• Direct sensitivity:

- 1) first compute $\frac{du}{dm_i} \rightarrow i = 1, \dots, M$
- 2) Solve $A^{-1}C$

disadvantage: M solves with A needed (not bad if

A is factored first) \Rightarrow if iterative solver used, restrict to small

advantage: same solver as Newton for Fwd prob.

- 2) then compute $\frac{\partial f^T}{\partial u} \frac{du}{dm}$ and subtract from $\frac{\partial f^T}{\partial m}$
- 3) cost: linear algebra (M inner products)

• Adjoint sensitivity:

$$1) \text{ first: } p = -A^{-T} \frac{\partial f}{\partial u} = - \boxed{A^{-T}} \downarrow \frac{\partial f}{\partial u}$$

$$\text{i.e. } A^T p = - \frac{\partial f}{\partial u} \Rightarrow \underline{\text{one solve!}} \quad (\text{but with } A^T)$$

- 2) then $C^T p$ (M inner products, just linear algebra)
- and add to $\frac{\partial f}{\partial m}$

advantage: just one solve - independent of M !

disadvantage: not exactly same operator as
Newton forward solver

In both cases,
first to need
to solve
forward
problem
 $c(u, m) \Rightarrow$
 $c(u, m)$ from
to u

Adjoint method can be derived via Lagrangian:

$$\text{define Lagrangian } \mathcal{L}(u, m, p) := f(u, m) + p^T r(u, m)$$

objective (obj. mult.) form residues

$$1) \frac{\partial \mathcal{L}}{\partial p} = 0 \Rightarrow \text{Solve } r(u, m) = 0 \quad \text{forward problem}$$

$$2) \frac{\partial \mathcal{L}}{\partial u} = 0 \Rightarrow \frac{\partial f}{\partial u}^T + p^T \left(\frac{\partial r}{\partial u} \right) = 0$$

$\hookrightarrow A$

$$\Rightarrow \text{Solve } A^T p = -\frac{\partial f}{\partial u} \quad \text{adjoint problem}$$

$$3) \frac{\partial \mathcal{L}}{\partial m} = \text{gradient} \Rightarrow \boxed{\frac{\partial f}{\partial m} + p^T \frac{\partial r}{\partial m}}$$

C

Given some m , solve forward u ^{with u & m ,} then ^{to be adjoint;} evaluate gradient with m, u, p

Lagrangian is a device for facilitating gradient computation
 \Rightarrow only partial derivatives needed; no dependence of state variable of parameters!

Infinite dimensions : direct sensitivity

Take model problem:

$$\min_m \mathcal{J}(u(m), m) := \frac{1}{2} \int_{\Omega} (b(x) u - d)^2 dx + \frac{\alpha}{2} \int_{\Omega} \nabla m \cdot \nabla m dx$$

$$\text{where } -\nabla \cdot (m \nabla u) + v \cdot \nabla u = 0 \quad \text{in } \Omega$$

$$u = 0 \quad \text{on } \Gamma$$

$$u \in H_0^1(\Omega); \quad m \in H^1(\Omega) \quad (\text{from regularization term})$$

First step: BVP \Rightarrow weak form

$$\int_{\Omega} p \left[-\nabla \cdot (m \nabla u) - f \right] dx = 0 \quad \forall p \in H^0$$

Green's identity:

$$\Rightarrow \int_{\Omega} [m \nabla u \cdot \nabla p - pf] dx = \int_{\Gamma} p m \nabla u \cdot n ds = 0 \quad \forall p \in H^0$$

$$\Rightarrow \int_{\Omega} [m \nabla u \cdot \nabla p - pf] dx = 0 \quad \forall p \in H^0$$

Replace m with family of variations:

$$m \rightarrow m + \varepsilon \hat{m}; \quad \hat{m} \in H^1$$

this leads to $u \rightarrow u + \varepsilon \hat{u} \quad \hat{u} \in H^1$

$$\Rightarrow \int_{\Omega} [(m + \varepsilon \hat{m}) (\nabla u + \varepsilon \nabla \hat{u}) \cdot \nabla p + pu \cdot \nabla (u + \varepsilon \hat{u}) - pf] dx = 0 \quad \forall p \in H^0$$

Now take derivative wrt ε : $\frac{d}{d\varepsilon} \Big|_{\varepsilon=0}$

$$\frac{d}{d\varepsilon} \Big|_{\varepsilon=0} = \int_{\Omega} [\hat{m} \nabla u \cdot \nabla p + m \nabla \hat{u} \cdot \nabla p + pu \cdot \nabla \hat{u}] dx = 0 \quad \forall p \in H^0$$

weak form of direct sensitivity equation

Solve for \hat{u} given \hat{m} and u

$$\text{Strong form: } \int_{\Omega} p \left[-\nabla \cdot (\hat{m} \nabla \hat{u}) - \nabla \cdot (m \nabla \hat{u}) + vu \cdot \nabla \hat{u} \right] dx$$

$$+ \int_{\Gamma} (p \hat{m} \nabla \hat{u} \cdot n + p m \nabla \hat{u} \cdot n) ds = 0 \quad \forall p \in H^0$$

$$\Rightarrow \boxed{-\nabla \cdot (\hat{m} \nabla \hat{u}) + vu \cdot \nabla \hat{u} = -\nabla \cdot (m \nabla \hat{u}) \text{ in } \Omega}$$

strong
form of
sensitivity equation

$$\hat{u} = 0 \text{ on } \Gamma$$

Do same for objective function:

$$\mathcal{J}(u + \varepsilon \hat{u}, m + \varepsilon \hat{m}) = \frac{1}{2} \int_{\Omega} (b(x)(u + \varepsilon \hat{u}) - d)^2 dx$$

$$+ \frac{\alpha}{2} \int_{\Omega} (\nabla m + \varepsilon \nabla \hat{m}) \cdot (\nabla m + \varepsilon \nabla \hat{m}) dx$$

$$\left. \frac{d\mathcal{J}}{d\varepsilon} \right|_{\varepsilon=0} = \int_{\Omega} (b(x)(u + \varepsilon \hat{u}) - d) b(x) \hat{u} dx + \alpha \int_{\Omega} \nabla \hat{m} \cdot \nabla m dx$$

Frechet derivative of \mathcal{J} at m in direction of \hat{m}

[Infinite dimensional version of $\hat{g}^T \hat{m}$]
weak form of "gradient"

$$\Rightarrow \underbrace{\int_{\Omega} (bu - d)b\hat{u} dx}_{\sim \frac{\partial f}{\partial u} \frac{du}{dm}} + \alpha \underbrace{\int_{\Omega} \nabla \hat{m} \cdot \nabla m dx}_{\sim \frac{\partial f}{\partial m}} = \text{gradient } \hat{t}^m \in H^1$$

Adjoint method in infinite dimensions

weak form of
adjoint eqn

$$\int_{\Omega} \hat{m} \nabla \hat{u} \cdot \nabla p + p v \cdot \nabla \hat{u}] dx = - \int_{\Omega} (bu - d)b\hat{u} dx \quad \forall \hat{u} \in H^1$$

$$\sim p^T A \frac{du}{dm}$$

Solve for p

Then gradient. (given u, p, \hat{m}):

$$\int_{\Omega} \hat{m} \nabla u \cdot \nabla p dx + \alpha \int_{\Omega} \nabla \hat{m} \cdot \nabla m dx = \text{grad. } \hat{t}^m \in H^1$$

Frechet
derivative

Strong form of adjoint eqn:

$$\Rightarrow -\nabla \cdot (n \nabla p) - \nabla \cdot (p \vec{v}) = -b(bu - d) \quad \text{in } \Omega$$

$$p = 0 \quad \text{on } \Gamma$$

Strong form of gradient:

$$-\alpha \Delta m + -\nabla u \cdot \nabla p \in \Omega; \quad \alpha \frac{\partial m}{\partial n} = 0 \quad \text{on } \Gamma$$

Lagrangian approach to adjoint method:

$$\begin{aligned} \mathcal{L}(u, p, m) = & \frac{1}{2} \int_{\Omega} (bu - d)^2 dx + \frac{\alpha}{2} \int \nabla m \cdot \nabla m dx \\ u \in H_0^1; p \in H_0^1; m \in H^1 & \\ & + \underbrace{\int_{\Omega} [m \nabla u \cdot \nabla p + p v \cdot \nabla u - pf] dx}_{\text{weak form of flux prob}} \end{aligned}$$

Forward eqn: $\int_p \mathcal{L} \Rightarrow$

$$\begin{aligned} \text{Weak form of } \{ \text{f}, \text{f} \}: \int_{\Omega} [m \nabla u \cdot \nabla \hat{p} + \hat{p} v \cdot \nabla u - \hat{p} f] dx & \Rightarrow \forall \hat{p} \in H_0^1 \\ \Rightarrow \int_{\Omega} \hat{p} [-\nabla \cdot (m \nabla u) + v \cdot \nabla u - f] dx & = 0 \quad \forall \hat{p} \in H_0^1 \\ \Rightarrow -\nabla \cdot (m \nabla u) + v \cdot \nabla u - f & = 0 \quad \text{in } \Omega, \hat{p} = 0 \text{ on } \Gamma \end{aligned}$$

Adjoint
equation

$$\int_u \mathcal{L} \Rightarrow \int_{\Omega} (bu - d) b \hat{u} dx + \int_{\Omega} [m \nabla u \cdot \nabla p + p v \cdot \nabla u] dx = 0 \quad \forall u \in H_0^1$$

weak form of adjoint equation

$$\begin{aligned} \Rightarrow \int_{\Omega} [b(bu - d) - \nabla \cdot (m \nabla p) - \nabla \cdot (pv)] dx & = 0 \quad \forall u \in H_0^1 \\ \Rightarrow -\nabla \cdot (m \nabla p) - \nabla \cdot (pv) & = -b(bu - d) \quad \text{in } \Omega \\ p & = 0 \quad \text{on } \Gamma \end{aligned}$$

$$\begin{aligned} \text{Gradient expression: } \int_m \mathcal{L} \Rightarrow & \alpha \int_{\Omega} \nabla m \cdot \nabla \hat{m} dx + \int_{\Omega} \hat{m} \nabla u \cdot \nabla p dx \equiv \text{grad } \forall \hat{m} \in H^1 \\ & \underbrace{\quad \quad \quad}_{\text{weak form of grad}} \end{aligned}$$

$$\begin{aligned} \Rightarrow -\alpha \Delta m + \nabla u \cdot \nabla p & \quad \text{in } \Omega \\ \frac{\partial m}{\partial \vec{n}} & = 0 \quad \text{on } \Gamma \end{aligned} \quad \left. \right\} \text{Strong form of gradient}$$

Summary of infinite dimensions:

Direct:

State: find $u \in H_0^1$ s.t. $\int_{\Omega} [m \nabla u \cdot \nabla p + p v \cdot \nabla u - pf] dx = 0 \quad \forall p \in H_0^1$

Sensitivity eqn: Find $\hat{u} \in H_0^1$ s.t. $\int_{\Omega} [\hat{m} \nabla u \cdot \nabla p + m \nabla \hat{u} \cdot \nabla p + p v \cdot \nabla \hat{u}] dx = 0 \quad \forall p \in H_0^1$

Gradient for $\hat{m} \in H^1$: $\int_{\Omega} (bu-d) b \hat{u} dx + \alpha \int_{\Omega} \nabla \hat{m} \cdot \nabla m dx$ (Erreicht den Wert von \hat{m} in direction of \hat{m})

Adjoint

State $\underset{\text{find } u \in H_0^1}{\int_{\Omega}} [m \nabla u \cdot \nabla p + p v \cdot \nabla u - pf] dx = 0 \quad \forall p \in H_0^1$

Adjoint: $\int_{\Omega} [m \nabla \hat{u} \cdot \nabla p + p v \cdot \nabla \hat{u} + (bu-d) b \hat{u}] dx = 0 \quad \forall \hat{u} \in H_0^1$
Find $p \in H_0^1$ s.t.

Gradient $\hat{m} \in H^1$: $\int_{\Omega} \hat{m} \nabla u \cdot \nabla p dx + \alpha \int_{\Omega} \nabla \hat{m} \cdot \nabla m dx$

Hessian

$$(B_{u-d})^T (B_{u-d}) + m^T R_m$$

$$\min_{M \in \mathbb{R}^{n \times n}} f(u(m), m) \quad \text{s.t. } r(u(m), m) = 0$$

$$\mathcal{L}^G(u, m, p) := f(u, m) + p^T r(u, m)$$

$$\frac{\partial \mathcal{L}^G}{\partial p} = 0 \Rightarrow r(u, m) = 0 \quad \text{for } w$$

$$\frac{\partial \mathcal{L}^G}{\partial u} = 0 \Rightarrow (\frac{\partial f}{\partial u})^T + p^T \underbrace{(\frac{\partial r}{\partial u})}_{\hookrightarrow A} = 0 \Rightarrow A^T p = -\frac{\partial f}{\partial u} \quad \text{adjoint}$$

$$\frac{\partial \mathcal{L}^G}{\partial m} = 0 \Rightarrow \frac{\partial f}{\partial m}^T + p^T \underbrace{(\frac{\partial r}{\partial m})}_{\hookrightarrow C} = 0 \Rightarrow C^T p + \frac{\partial f}{\partial m} = \text{gradient}$$

Hessian? Use Lagrange idea again

directional derivative: $g^T \tilde{m}$
(in direction \tilde{m})

$$\mathcal{L}^H(u, m, p; \tilde{u}, \tilde{m}, \tilde{p}) := \tilde{m}^T \underbrace{[C^T p + \frac{\partial f}{\partial m}]}_g + \tilde{u}^T \left(\frac{\partial f}{\partial u} + A^T p \right) + \tilde{p}^T r$$

$$\frac{\partial \mathcal{L}^H}{\partial p} = A \tilde{u} + C \tilde{m} = 0 \quad (\text{incremental forward eqn})$$

$$\frac{\partial \mathcal{L}^H}{\partial u} = \frac{\partial}{\partial u} \left(\tilde{m}^T C^T p \right) + \frac{\partial}{\partial u} \left(\tilde{m}^T \frac{\partial f}{\partial m} \right) + \frac{\partial}{\partial u} \left(\tilde{u}^T \frac{\partial f}{\partial u} \right) + \frac{\partial}{\partial u} \left(\tilde{u}^T A^T p \right) + A^T \tilde{p}$$

$$\Rightarrow A^T \tilde{p} = - \underbrace{\left(L_{mu}^T \tilde{m} + L_{uu}^T \tilde{u} \right)}_{L_{mu}} \quad \text{incremental adjoint}$$

$$\begin{aligned} \frac{\partial \mathcal{L}^H}{\partial m} &\Rightarrow H \tilde{m} = \frac{\partial}{\partial m} \left[\tilde{m}^T C^T p \right] + \frac{\partial^2 f}{\partial m \partial m} \tilde{m} + \frac{\partial}{\partial m} \left(\tilde{u}^T \frac{\partial f}{\partial u} \right) + \frac{\partial}{\partial m} \left(\tilde{u}^T A^T p \right) + C^T \tilde{p} \\ &= \underbrace{L_{mm}^T \tilde{m}}_{L_{mm}} + \underbrace{L_{mu}^T \tilde{u}}_{L_{mu}} + C^T \tilde{p} \end{aligned}$$

Given m , solve for $w \Rightarrow u$; solve adjoint $\Rightarrow p$, evaluate gradient

Given m, u, p , and \tilde{m} , solve incr. forw $\Rightarrow \tilde{u}$; solve incr. adj $\Rightarrow \tilde{p}$

\Rightarrow evaluate $H \tilde{m}$

$$\text{Note: } \hat{u} = -A^{-1}C\hat{m}$$

$$\hat{p} = -A^{-T}L_{uu}\hat{m} - A^{-T}L_{mu}\hat{u} = -A^{-T}L_{uu}\hat{m} + A^{-T}L_{uu}A^{-1}C^T\hat{m}$$

$$H\hat{m} = L_{mm}\hat{m} - L_{mu}(A^{-1}C)^T\hat{m} - C^TA^{-T}L_{um}\hat{m} + \underline{C^TA^{-T}L_{uu}A^{-1}C\hat{m}}$$

$$= \underbrace{\left[C^TA^{-T}L_{uu}A^{-1}C + L_{mm} - L_{mu}A^{-1}C - C^TA^{-T}L_{um} \right] \hat{m}}_H$$

$$L_{uu} = \underbrace{\frac{\partial^2 f}{\partial u^2}}_{B^T B} + \frac{\partial}{\partial u}(A^T p)$$

$$L_{mm} = R + \frac{\partial}{\partial m}(C^T p)$$

$$L_{mu} = L_{um}^T = \underbrace{\frac{\partial}{\partial m}\left(\frac{\partial f}{\partial u}\right)}_{=0} + \frac{\partial}{\partial m}(A^T p)$$

Gauss-Newton approximation of Hessian:

$$p \approx 0 \quad [\text{motivation: adj: } A^T p = -B^T(Bu - d)]$$

≈ 0 for low noise / good model

$$\Rightarrow L_{uu} = B^T B ; L_{mm} = R ; L_{mu} = L_{um}^T = 0$$

$$\Rightarrow H^{GN} = C^T A^{-T} B^T B A^{-1} C + R$$

Infinite dimensions

$$\begin{aligned} \mathcal{L}(u, p, m) := & \frac{1}{2} \int_{\Omega} (bu - d)^2 dx + \frac{\alpha}{2} \int_{\Omega} \nabla m \cdot \nabla m dx \\ & + \int_{\Omega} [m \nabla u \cdot \nabla p + p v \cdot \nabla u - pf] dx \end{aligned}$$

$$\int_p \mathcal{L}^G \rightarrow \text{form: } \int_{\Omega} [m \nabla u \cdot \nabla \hat{p} + \hat{p} v \cdot \nabla u - \hat{p} f] dx = 0 \quad \forall \hat{p} \in H_0^1$$

$$\text{Suz } \mathcal{L}^G \rightarrow \text{adj: } \int_{\Omega} (bu - d) b \hat{u} dx + \int_{\Omega} [m \nabla \hat{u} \cdot \nabla p + p v \cdot \nabla \hat{u}] dx = 0 \quad \forall \hat{u} \in H_0^1$$

$$\delta_m \mathcal{L}^G \rightarrow \text{"grad": } \underbrace{\alpha \int_{\Omega} \nabla m \cdot \nabla \hat{m} dx}_{\text{(Fréchet derivative in dir } \hat{m})} + \int_{\Omega} \hat{m} \nabla u \cdot \nabla p dx \quad \text{grad } \hat{m} \in H^1$$

$$\begin{aligned} \mathcal{L}^H(u, p, m; \tilde{u}, \tilde{p}, \tilde{m}) := & \underbrace{\alpha \int_{\Omega} \nabla m \cdot \nabla \tilde{m} dx}_{\text{F. r. w}} + \int_{\Omega} \tilde{m} \nabla u \cdot \nabla p dx \\ & + \int_{\Omega} [m \nabla u \cdot \nabla \tilde{p} + \tilde{p} v \cdot \nabla u - \tilde{p} f] dx + \int_{\Omega} (bu - d) b \tilde{u} dx + \underbrace{\int_{\Omega} [m \nabla \tilde{u} \cdot \nabla p + p v \cdot \nabla \tilde{u}]}_{\text{adj}} \end{aligned}$$

$$\begin{aligned} \text{Unr. adj: } \int_u \mathcal{L}^H = & \int_{\Omega} [\tilde{m} \nabla \hat{u} \cdot \nabla p + m \nabla \hat{u} \cdot \nabla \tilde{p} + \tilde{p} v \cdot \nabla \hat{u} + (b \hat{u})(b \tilde{u})] dx = 0 \quad \forall \hat{u} \in H_0^1 \\ \Rightarrow & -\nabla \cdot (\tilde{m} \nabla p) - \nabla \cdot (m \nabla \tilde{p}) - \nabla \cdot (\tilde{p} v) + b b \tilde{u} = 0 \quad \text{in } \Omega \\ & \tilde{p} = 0 \quad \text{on } \Gamma \end{aligned}$$

$$\begin{aligned} \text{incr. form: } \int_p \mathcal{L}^H = & \int_{\Omega} [\tilde{m} \nabla u \cdot \nabla \hat{p} + m \nabla \tilde{u} \cdot \nabla \hat{p} + \hat{p} v \cdot \nabla \tilde{u}] dx = 0 \quad \forall \hat{p} \in H_0^1 \\ \Rightarrow & -\nabla \cdot (m \nabla \tilde{u}) + v \cdot \nabla \tilde{u} = \nabla \cdot (\tilde{m} \nabla u) \quad \text{in } \Omega \\ & \tilde{u} = 0 \quad \text{on } \Gamma \end{aligned}$$

$$\begin{aligned} \text{Hessian action: } \int_m \mathcal{L}^H = & \int_{\Omega} [\alpha \nabla \hat{m} \cdot \nabla \tilde{m} + \hat{m} \nabla u \cdot \nabla \tilde{m} + \hat{m} \nabla \tilde{u} \cdot \nabla p] dx \equiv 2 \tilde{m} \quad \forall m \in H^1 \\ & \text{in dim of } \tilde{m} \end{aligned}$$

$$\Rightarrow -\nabla \cdot (\alpha \nabla \tilde{m}) + \nabla u \cdot \nabla \tilde{m} + \nabla \tilde{u} \cdot \nabla p \quad \text{in } \Omega \quad \text{on } \Gamma$$