

# Analytic Jacobian Richards Equation

We want to solve

$$\frac{\partial \theta}{\partial t} - \nabla \cdot [D_H(\theta) \nabla \theta] = f$$

form of a general non-linear diffusion eqn.

Discretization with backward Euler

$$\underline{\Gamma} = \theta^{n+1} - \theta^n - \Delta t \underline{D} * \left[ \left\{ \underline{H} \underline{D}_H(\theta^{n+1}) \right\}_f \underline{G} \theta^{n+1} \right] - \Delta t f_s = \underline{0}$$

where  $\underline{D}_H(\theta)$  is vector of  $D_H$  at cell centers

and  $\left\{ \underline{H} \underline{D}_H(\theta^{n+1}) \right\}_f$  is a  $N_f$  by  $N_f$  matrix

of arithmetic averages on the faces

$\underline{\theta}^{n+1} \equiv \underline{\theta}$  unknown we find using Newton-Raphson

$n$  = superscript for time step

$k$  = superscript for N-R iteration

Linearize residual as before at  $\bar{\theta}$

$$\begin{aligned}\underline{L}_{\bar{\theta}} \underline{r} &= \underline{r}(\bar{\theta}) + \nabla_{\bar{\theta}} \underline{r} \cdot \underline{\Delta\theta} & \underline{\Delta\theta} &= \epsilon \hat{\underline{\theta}} \\ &= \underline{r}(\bar{\theta}) + \epsilon \nabla_{\bar{\theta}} \underline{r} \cdot \hat{\underline{\theta}} & & \uparrow \\ & & & \text{dir. of update}\end{aligned}$$

Directional derivative of  $\underline{r}$  at  $\underline{h}$  in direction  $\hat{\underline{\theta}}$

$$\nabla_{\hat{\underline{\theta}}} \underline{r} \cdot \hat{\underline{\theta}} = \left. \frac{d}{d\epsilon} \underline{r}(\bar{\theta} + \epsilon \hat{\underline{h}}) \right|_{\epsilon=0}$$

Simple example: scalar-valued vector function

$$f(\underline{x}) = x + y^2 \quad \nabla f = \begin{pmatrix} 1 \\ 2y \end{pmatrix}$$

Given some direction  $\hat{\underline{x}} = \begin{pmatrix} \hat{x} \\ \hat{y} \end{pmatrix}$  and location  $\bar{\underline{x}} = \begin{pmatrix} \bar{x} \\ \bar{y} \end{pmatrix}$

$$\text{a) } \nabla f|_{\bar{\underline{x}}} \cdot \hat{\underline{x}} = (1 \quad 2\bar{y}) \cdot \begin{pmatrix} \hat{x} \\ \hat{y} \end{pmatrix} = \underline{\hat{x} + 2\bar{y}\hat{y}}$$

$$\text{b) } \left. \frac{d}{d\epsilon} f(\bar{\underline{x}} + \epsilon \hat{\underline{x}}) \right|_{\epsilon=0} = \left. \frac{d}{d\epsilon} \bar{x} + \epsilon \hat{x} + (\bar{y} + \epsilon \hat{y})^2 \right|_{\epsilon=0}$$

$$\begin{aligned}&= \left. \frac{d}{d\epsilon} \bar{x} + \epsilon \hat{x} + \bar{y}^2 + 2\epsilon \bar{y} \hat{y} + \epsilon^2 \hat{y}^2 \right|_{\epsilon=0} \\ &= \hat{x} + 2\bar{y}\hat{y} + 2\epsilon \hat{y}^2 \Big|_{\epsilon=0} \\ &= \underline{\hat{x} + 2\bar{y}\hat{y}}\end{aligned}$$

## Directional derivative of residual

To linearize our discrete system we need directional derivative of a vector-valued vector function.

$$\nabla \underline{r} |_{\underline{\theta}} \hat{\underline{\theta}} = \frac{d}{d\varepsilon} \underline{r}(\underline{\theta} + \varepsilon \hat{\underline{\theta}}) \Big|_{\varepsilon=0}$$

$$\underline{r}(\underline{\theta}) = \underline{\theta} - \underline{\theta}^n - \Delta t \underline{D} * \left[ \left\{ \underline{M} \underline{D}_H(\underline{\theta}) \right\}_f \underline{G} \underline{\theta} \right] - \Delta t \underline{f}_s = \underline{0}$$

$$\text{where } \underline{\theta} = \underline{\theta}^{n+1} \quad \nabla$$

$$\begin{aligned} \frac{d}{d\varepsilon} \underline{r}(\underline{\theta} + \varepsilon \hat{\underline{\theta}}) \Big|_{\varepsilon=0} &= \frac{d}{d\varepsilon} \underline{\theta} + \varepsilon \hat{\underline{\theta}} - \underline{\theta}^n - \Delta t \underline{D} * \left[ \left\{ \underline{M} \underline{D}_H(\underline{\theta} + \varepsilon \hat{\underline{\theta}}) \right\}_f \underline{G} (\underline{\theta} + \varepsilon \hat{\underline{\theta}}) \right] \Big|_{\varepsilon=0} \\ &= \hat{\underline{\theta}} - \Delta t \underline{D} * \frac{d}{d\varepsilon} \left[ \left\{ \underline{M} \underline{D}_H(\underline{\theta} + \varepsilon \hat{\underline{\theta}}) \right\}_f \underline{G} (\underline{\theta} + \varepsilon \hat{\underline{\theta}}) \right] \Big|_{\varepsilon=0} \end{aligned}$$

$$= \hat{\underline{\theta}} - \Delta t \underline{D} \left[ \underbrace{\frac{d}{d\varepsilon} \left\{ \underline{M} \underline{D}_H(\underline{\theta} + \varepsilon \hat{\underline{\theta}}) \right\}_f \underline{G} (\underline{\theta} + \varepsilon \hat{\underline{\theta}})} + \left\{ \underline{M} \underline{D}_H(\underline{\theta} + \varepsilon \hat{\underline{\theta}}) \right\}_f \underline{G} \hat{\underline{\theta}} \right]$$

$$\begin{aligned} \frac{d}{d\varepsilon} \left\{ \underline{M} \underline{D}_H(\underline{\theta} + \varepsilon \hat{\underline{\theta}}) \right\}_f &= \left\{ \underline{M} \frac{d}{d\varepsilon} \underline{D}_H(\underline{\theta} + \varepsilon \hat{\underline{\theta}}) \right\}_f \\ &= \left\{ \underline{M} \frac{d \underline{D}_H(\underline{\theta} + \varepsilon \hat{\underline{\theta}})}{d \underline{\theta}} \frac{d}{d\varepsilon} (\underline{\theta} + \varepsilon \hat{\underline{\theta}}) \right\}_f \\ &= \left\{ \underline{M} \frac{d \underline{D}_H(\underline{\theta} + \varepsilon \hat{\underline{\theta}})}{d \underline{\theta}} \hat{\underline{\theta}} \right\}_f \end{aligned}$$

substitute and evaluate at  $\epsilon=0$

$$= \hat{\underline{\theta}} - \Delta t \underline{\underline{D}} * \left[ \left\{ \underline{\underline{H}} \frac{dD_H}{d\theta}(\bar{\theta}) \hat{\underline{\theta}} \right\}_f \underline{\underline{G}} \bar{\theta} + \underbrace{\left\{ \underline{\underline{H}} \underline{\underline{D}}_H(\bar{\theta}) \right\}_f \underline{\underline{G}} \hat{\underline{\theta}}}_{\underline{\underline{Kd}}(\bar{\theta})} \right]$$

$\underline{\underline{G}} \bar{\theta}$  = known vector =  $\underline{\underline{GU}}$  diagonal matrix

$$\left\{ \underline{\underline{H}} \frac{dD_H}{d\theta}(\bar{\theta}) \hat{\underline{\theta}} \right\}_f \underline{\underline{GU}} = \underline{\underline{GU}} \left\{ \underline{\underline{H}} \frac{dD_H}{d\theta}(\bar{\theta}) \hat{\underline{\theta}} \right\}_f \quad \text{switch}$$

$$\left\{ \underline{\underline{H}} \frac{dD_H}{d\theta}(\bar{\theta}) \hat{\underline{\theta}} \right\}_f = \underbrace{\left\{ \underline{\underline{H}} \frac{dD_H}{d\theta}(\bar{\theta}) \right\}_f}_{\underline{\underline{dKd}}(\bar{\theta})} \hat{\underline{\theta}} = \underline{\underline{dKd}}(\bar{\theta}) \hat{\underline{\theta}}$$

substituting

$$\frac{d}{d\epsilon} \underline{\underline{r}}(\bar{\theta} + \epsilon \hat{\underline{\theta}}) \Big|_{\epsilon=0} = \hat{\underline{\theta}} - \Delta t \underline{\underline{D}} * \left[ \underline{\underline{GU}} * \underline{\underline{dKd}} \hat{\underline{\theta}} + \underline{\underline{dK}} * \underline{\underline{G}} \hat{\underline{\theta}} \right]$$

$$= \underbrace{\left( \underline{\underline{I}} - \Delta t \underline{\underline{D}} \left[ \underline{\underline{GU}} \underline{\underline{dKd}} + \underline{\underline{dK}} \underline{\underline{G}} \right] \right)}_{\underline{\underline{J}}(\bar{\theta})} \hat{\underline{\theta}}$$

$$\underline{\underline{J}}(\bar{\theta}) = \underline{\underline{I}} - \Delta t \underline{\underline{D}} \left[ \underline{\underline{GU}} \underline{\underline{dKd}} + \underline{\underline{dK}} \underline{\underline{G}} \right]$$

to evaluate  $dKd$  we need to differentiate  $D_H(\theta) \nabla$

$$\frac{dD_H}{d\theta} = \frac{k_s h_b}{\Delta \theta^2} \frac{1+2\lambda}{\lambda^2} s(\theta)^{1+\frac{1}{\lambda}}$$

$$s = \frac{\theta - \theta_r}{\Delta \theta}, \quad \Delta \theta = \theta_s - \theta_r$$

Implementation pitfalls:

Three versions of unknown vectors  $\underline{\theta}$

$$\underline{\Gamma}(\underline{a}, \underline{\theta}^n) \quad \underline{\underline{\Gamma}}(\underline{a})$$

$$\underline{\Delta\theta}^k = -\underline{\underline{J}}(\underline{\theta}^k) \setminus \underline{\Gamma}(\underline{\theta}^k, \underline{\theta}^n)$$

$\underline{\theta}^n$  is solution at last time step

$\underline{u}^k$  is current iterate of N-R

$$\underline{u}^k \rightarrow \underline{u} = \underline{u}^{n+1} \quad \text{as } k \rightarrow \infty$$

Note  $\underline{u}^k$  is initialized with old solu.  $\underline{u}^0 = \underline{u}^n$

common error is to confuse  $\underline{u}^n$  and  $\underline{u}^k$

## General outline

for  $n = 1 : N$      % time-stepping loop

$\theta$ <sub>old</sub> =  $\theta$      % save old soln      $\theta$ <sub>old</sub> =  $\theta$  <sup>n</sup>

while  $\|r\| > \text{tol} \parallel \|d\theta\| > \text{tol} \parallel k < k_{\max}$

$$d\theta = -\underline{J}(\underline{\theta}) \setminus \underline{r}(\underline{\theta}, \underline{\theta}_{\text{old}})$$

$$\underline{\theta} = \underline{\theta} + d\theta$$

end

end