IMPLEMENTATION OF CONSTITUTIVE EQUATIONS

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- Few constitutive equations are available in commercial FE codes
- Implement new constitutive models in commercial codes (ABAQUS, ANSYS, MARC, ..., Code_Aster, CAST3M, ..., Zset, WARP3D)
- Implement specific treatment of implemented models

- "Numerical" definition of constitutive equations
- Methods (explicit/implicit) for numerical integration
- Consistant tangent matrox
- Example : von Mises plasticiy model
- Convergence

"Numerical" definition of constitutive equations

- In the case of a displacement based FE formulation, nodal displacements are known and deformations can be computed
- The constitutive equations are used to evaluate:

• the stress tensor $\underline{\sigma}$

the "consistent" tangent matrix

$$\underline{\underline{L}} = \frac{\partial \Delta \underline{\sigma}}{\partial \Delta \varepsilon}$$

for a given deformation increment $\Delta \underline{\varepsilon}$.

- Material behaviours are also characterized be internal variables A.
- The constitutive equations are used to update these variables.

Constitutive equations within the finite element method



: local time integration of the constitutive equations



$$\frac{d\mathbf{A}}{dt} = \dot{\mathbf{A}} = \mathbf{G}(\mathbf{A}, t)$$
$$\frac{dA_i}{dt} = \dot{A}_i = G_i(A_1, \dots, A_n, t)$$

- Time is used here to represent an imposed deformation as well as imposed external parameters depending on both time and position such as an temperature (T(x, t)).
- Evaluation of the constitutive equations = integrate the previous system from t_0 to t_1 .
- Usually: $\mathbf{A} = (\underline{\varepsilon}_{e}, \dots)$ so that

$$\underline{\sigma}(t_1) = \underline{E}(t_1) : \underline{\varepsilon}_{e}(t_1)$$

• <u>L</u> must still be computed...

Integration methods for constitutive equations

$$\mathbf{A}(t_1) = \mathbf{A}(t_0) + \dot{\mathbf{A}} (\mathbf{A}(t_0), t_0) \Delta t = \mathbf{A}(t_0) + \mathbf{G} (\mathbf{A}(t_0), t_0) \Delta t$$
$$\Delta \mathbf{A} = \mathbf{G} (\mathbf{A}(t_0), t_0) \Delta t$$

- The method is not stable ; it must not be employed
- Explicit ? : because A is evaluated at t₀ using (A(t₀), t₀)

- Numerical estimation of derivatives of Å: dG/dA, d²G/dA², d³G/dA³...
- Error estimator to control the solution

The Runge-Kutta integration method is easy to set up as it only uses the differential equation:

$$\dot{\mathbf{A}} = \mathbf{G}(\mathbf{A}, t)$$

It however has some drawbacks:

- Integration may require long CPU time
- In the case of plastic material, the plastic multiplier must be computed. This can be tricky in cases where material coefficients depend on external parameters (e.g. temperature).
- The stability of the integration scheme is not certain for softening materials.

- Estimation of $\dot{\mathbf{A}}$ at time t_{θ} between t_0 and t_1
- $t_{\theta} = t_0 + \theta \Delta t \text{ avec } 0 \le \theta \le 1$
- Two solutions :

$$\Delta \mathbf{A} = \mathbf{G}(\mathbf{A}(t_0) + \theta \Delta \mathbf{A}, t_0 + \theta \Delta t) \Delta t$$

= $\mathbf{G}(\mathbf{A}_{\theta}, t_{\theta}) \Delta t$
$$\Delta \mathbf{A} = [(1 - \theta)\mathbf{G}(\mathbf{A}(t_0), t_0) + \theta \mathbf{G}(\mathbf{A}(t_1), t_1)] \Delta t$$

= $[(1 - \theta)\mathbf{G}(\mathbf{A}_0, t_0) + \theta \mathbf{G}(\mathbf{A}_0 + \Delta \mathbf{A}, t_0 + \Delta t)] \Delta t$

- Implicit i? : △A appears on both sides of the previous equations
- Integrrate the constitutive equations = solve the previous equations
- $\theta = 0 \rightarrow \text{Euler}$
- ... in the following we will focus on the first method

- Resolution using the Newton-Raphson method
- Residual expressed as:

$$\mathbf{R}(\Delta \mathbf{A}) = \Delta \mathbf{A} - \mathbf{G}(\mathbf{A}(t_0) + \theta \Delta \mathbf{A}, t_0 + \theta \Delta t) \Delta t$$

$$R_i(\Delta A_1, \dots, \Delta A_n) = \Delta A_i - G_i(A_1(t_0) + \theta \Delta A_1, \dots, A_n(t_0) + \theta \Delta A_n) \Delta t$$

First order Taylor expansion at sth estimate ΔAs

$$\mathbf{R} = \mathbf{R}(\Delta \mathbf{A}_{s}) + \frac{\partial \mathbf{R}}{\partial \Delta \mathbf{A}} (\Delta \mathbf{A} - \Delta \mathbf{A}_{s}) = \mathbf{0}$$

• Calculation of the next estimate (*s* + 1):

$$\Delta \mathbf{A}_{s+1} = \Delta \mathbf{A}_s - \left(\frac{\partial \mathbf{R}}{\partial \Delta \mathbf{A}}\right)_{\Delta \mathbf{A} = \Delta \mathbf{A}_s}^{-1} \cdot \mathbf{R}(\Delta \mathbf{A}_s)$$

• $\mathbf{J} = \partial \mathbf{R} / \partial \Delta \mathbf{A}$ ($J_{ij} = \partial R_j / \partial A_j$): Jacobian matrix, $\mathbf{J}^* = \mathbf{J}^{-1}$

- The A vector often holds second order tensors
- The Voigt notation is then used
- Notation

Standard notation		Preferred notation
$\underline{\varepsilon} = \begin{pmatrix} \varepsilon_{11} \\ \varepsilon_{22} \\ \varepsilon_{33} \\ 2\varepsilon_{12} \\ 2\varepsilon_{23} \\ 2\varepsilon_{31} \end{pmatrix}$	$, \underline{\sigma} = \begin{pmatrix} \sigma_{11} \\ \sigma_{22} \\ \sigma_{33} \\ \sigma_{12} \\ \sigma_{23} \\ \sigma_{31} \end{pmatrix}$	$\underline{x} = \begin{pmatrix} x_{11} \\ x_{22} \\ x_{33} \\ \sqrt{2}x_{12} \\ \sqrt{2}x_{23} \\ \sqrt{2}x_{31} \end{pmatrix}$

Examples

Additive decomposition of deformations:

$$\underline{\varepsilon} = \underline{\varepsilon}\mathbf{e} + \underline{\varepsilon}\mathbf{p}$$

Yield surface

$$\phi = \sigma_{\rm eq} - R(p)$$

Normality

$$\underline{\dot{\varepsilon}}_{p} = \dot{p}\frac{\partial\phi}{\partial\underline{\sigma}} = \frac{3}{2}\dot{p}\frac{\underline{s}}{\sigma_{\rm eq}} = \dot{p}\underline{n}$$

Internal variables : (ε_e, p)

• Calculation of \dot{p} using the consistency condition: $\dot{\phi} = 0$

$$\dot{\phi} = \frac{\partial \phi}{\partial \underline{\sigma}} : \underline{\dot{\sigma}} + \frac{\partial \phi}{\partial p} \dot{p} = \underline{n} : \underline{\dot{\sigma}} - H\dot{p}$$

• avec
$$\underline{\sigma} = \underline{\underline{E}} : \underline{\varepsilon}_{e} = \underline{\underline{E}} : (\underline{\varepsilon} - \underline{\varepsilon}_{p}) \to \underline{\dot{\sigma}} = \underline{\underline{E}} : \underline{\dot{\varepsilon}}_{e} = \underline{\underline{E}} : (\underline{\dot{\varepsilon}} - \underline{\dot{\varepsilon}}_{p})$$

Finally one gets

$$\dot{p} = \frac{\underline{n}:\underline{\underline{E}}:\underline{\dot{e}}}{\underline{\underline{n}}:\underline{\underline{E}}:\underline{\underline{n}}+H}$$

System of differential equations to be solved

sur
$$\underline{e}_{e}$$
 $\underline{\dot{e}}_{e} = \underline{\dot{e}} - \dot{p}\underline{n}$
sur p $\dot{p} = \frac{\underline{n} : \underline{\underline{E}} : \underline{\dot{e}}}{\underline{n} : \underline{\underline{E}} : \underline{n} + H}$

- Attention must be paid on dependences on external parameters (temperature...)
- Ready for explicit Runge-Kutta integration !

von Mises plasticity : implicit integration

$$\begin{split} & \underline{\dot{\varepsilon}}_{\theta} = \underline{\dot{\varepsilon}} - \dot{p}\underline{n} & \to & \Delta \underline{\varepsilon}_{\theta} = \Delta \varepsilon - \Delta p\underline{n} \\ & \dot{p} = \frac{\underline{n} : \underline{\underline{E}} : \underline{\dot{\varepsilon}}}{\underline{n} : \underline{\underline{E}} : \underline{n} + H} & \to & \Delta p = \frac{\underline{n} : \underline{\underline{E}} : \Delta \varepsilon}{\underline{n} : \underline{\underline{E}} : \underline{n} + H} \quad \textcircled{2}$$

• Evaluation of
$$\underline{n}, \underline{E}, H$$
? ... at time $t_{\theta} = t_0 + \theta \Delta t$.

• Application :

$$\underline{\underline{n}} = \frac{3}{2} \frac{\underline{\underline{s}}^{\theta}}{\sigma_{eq}^{\theta}} \quad \text{avec} \quad \underline{\underline{\sigma}}^{\theta} = \underline{\underline{\underline{E}}}^{\theta} : \underline{\underline{\varepsilon}}^{\theta}_{\theta} \quad \underline{\underline{\varepsilon}}^{\theta}_{\theta} = \underline{\underline{\varepsilon}}^{0}_{\theta} + \theta \Delta \underline{\underline{\varepsilon}}_{\theta}$$
$$\underline{\underline{\underline{E}}}^{\theta} = \underline{\underline{\underline{E}}}(T^{\theta}) = \underline{\underline{\underline{E}}}(T^{0} + \theta \Delta T)$$
$$H^{\theta} = H(p^{\theta}) = H(p^{0} + \theta \Delta p)$$

The equation

$$\Delta p = \frac{\underline{n} : \underline{\underline{E}} : \Delta \varepsilon}{\underline{\underline{n}} : \underline{\underline{\underline{E}}} : \underline{\underline{n}} + H}$$

is correct but may be replaced by:

$$\phi = \sigma_{\rm eq} - R(p) = 0$$

This equation is wrong if *R* depends on an external parameter as in that case:

$$\dot{p} = \frac{\underline{n}:\underline{E}:\underline{\dot{e}}-R,_{T}T}{\underline{n}:\underline{E}:\underline{n}+H} \rightarrow \Delta p = \frac{\underline{n}:\underline{E}:\Delta\underline{e}-R,_{T}\delta T}{\underline{n}:\underline{E}:\underline{n}+H}$$

The incremental equation is then

$$\Delta p = \frac{\underline{n}^{\theta} : \underline{\underline{E}}^{\theta} : \Delta \underline{\underline{\varepsilon}} - \underline{R}^{\theta}_{,T} \Delta T}{\underline{n}^{\theta} : \underline{\underline{E}}^{\theta} : \underline{\underline{n}}^{\theta} + H^{\theta}}$$

The method must be avoided and it is better to use the yield condition $\phi = \sigma_{eq} - R(p) = 0$ • N.B.: in the following the supscript θ will be omitted for the sake of simplicity

$${R_e \over R_p} = \Delta \underline{\varepsilon}_e + \Delta p \underline{n} - \Delta \varepsilon$$

 $R_p = \phi = \sigma_{\rm eq} - R(p)$

$$\mathbf{R}=(R_e,R_p)$$

• Writing using sub-blocks

$$\mathbf{J} = \begin{pmatrix} \frac{\partial \underline{R}_{e}}{\partial \Delta \underline{\varepsilon}_{e}} & \frac{\partial \underline{R}_{e}}{\partial \Delta p} \\ \hline \\ \frac{\partial R_{p}}{\partial \Delta \varepsilon_{e}} & \frac{\partial R_{p}}{\partial \Delta p} \end{pmatrix} = \begin{pmatrix} \frac{\mathsf{T4} | \mathsf{T2}}{\mathsf{T2} | \mathsf{Sc}} \end{pmatrix}$$

Calculation of the derivatives of $\underline{R}_e = \Delta \underline{\varepsilon}_e + \Delta p \underline{n}^{\theta} - \Delta \varepsilon$

$$\frac{\partial R_{e}}{\partial \Delta \underline{\varepsilon}_{e}} = \underline{1} + \Delta p \frac{\partial \underline{n}}{\partial \underline{\sigma}} : \frac{\partial \underline{\sigma}}{\partial \underline{\varepsilon}_{e}} : \frac{\partial \underline{\varepsilon}_{e}}{\partial \Delta \underline{\varepsilon}_{e}}$$
$$\frac{\partial \underline{n}}{\partial \underline{\sigma}} = \underline{N} = \frac{1}{\sigma_{eq}} \left(\frac{3}{2} \underline{\underline{J}} - \underline{\underline{n}} \otimes \underline{\underline{n}} \right)$$

where \underline{J} is such that $\underline{J} : \underline{a} = \text{deviator}(\underline{a})$

$$\frac{\partial \underline{\sigma}}{\partial \underline{\varepsilon}_{e}} = \underline{\underline{E}}$$
$$\frac{\partial \underline{\varepsilon}_{e}}{\partial \Delta \underline{\varepsilon}_{e}} = \theta \underline{\underline{1}}$$

so that:

$$\frac{\partial R_{e}}{\partial \Delta_{\underline{\varepsilon}e}} = \underline{1} + \theta \underline{\underline{N}} : \underline{\underline{E}}$$

$$\frac{\partial R_{e}}{\partial \Delta p} = \underline{n}$$

Calculation of the derivatives of $R_p = \sigma_{eq} - R(p)$

$$\frac{\partial R_{p}}{\partial \Delta \underline{\varepsilon}_{e}} = \frac{\partial \sigma_{eq}}{\partial \underline{\sigma}} : \frac{\partial \underline{\sigma}}{\partial \underline{\varepsilon}_{e}} : \frac{\partial \underline{\varepsilon}_{e}}{\partial \Delta \underline{\varepsilon}_{e}} = \boxed{\theta \underline{n} : \underline{\underline{E}}}$$
$$\boxed{\frac{\partial R_{p}}{\partial \Delta p}} = -\frac{\partial R}{\partial p} \frac{\partial p}{\partial \Delta p} = \boxed{-\theta H}$$

Tangent matrix

$$\underline{\dot{\sigma}} = \underline{\underline{L}}_p : \underline{\dot{\varepsilon}}$$

• Calculation :

and

$$\dot{\underline{\sigma}} = \underline{\underline{E}} : (\underline{\dot{\varepsilon}} - \dot{p}\underline{n})$$

$$\dot{\underline{p}} = \frac{\underline{n} : \underline{\underline{E}} : \underline{\dot{\varepsilon}}}{\underline{n} : \underline{\underline{E}} : \underline{n} + H}$$

imply that

$$\underline{\underline{L}}_{p} = \underline{\underline{E}} - \frac{(\underline{\underline{E}}:\underline{\underline{n}}) \otimes (\underline{\underline{n}}:\underline{\underline{E}})}{\underline{\underline{n}}:\underline{\underline{E}}:\underline{\underline{n}} + H}$$

• The tangent matrix cannot be defined for viscoplastic materials

• "Consistent" tangent matrix

$$\underline{\underline{L}}_{c} = \frac{\partial \Delta \underline{\sigma}}{\partial \Delta \underline{\varepsilon}}$$

$$\Delta \underline{\sigma} = \underline{\underline{E}} : (\Delta : \underline{\varepsilon} - \Delta p \underline{\underline{n}})$$

calculation

$$\begin{split} \delta \Delta \underline{\sigma} &= \underline{\underline{E}} : (\delta \Delta \underline{\varepsilon} - \delta \Delta p \underline{n} - \Delta p \delta \underline{n}) \\ \underline{\underline{E}} : (\delta \Delta \underline{\varepsilon} - \delta \Delta p \underline{n}) = \underline{\underline{L}}_p : \delta \Delta \underline{\varepsilon} \\ \Delta p \delta \underline{\underline{n}} = \Delta p \frac{\partial \underline{n}}{\partial \underline{\sigma}} : \frac{\partial \underline{\sigma}}{\partial \Delta \underline{\varepsilon}} : \delta \Delta \underline{\varepsilon} = \Delta p \underline{\underline{N}} : \underline{\underline{E}} : \delta \Delta \underline{\varepsilon} + \dots \\ \text{so that} \\ \underline{\underline{L}}_c &\approx \underline{\underline{L}}_p - \Delta p \underline{\underline{E}} : \underline{\underline{N}} : \underline{\underline{E}} + \underline{O} (\Delta p^2) \end{split}$$

• The consistent tangent matrix can be defined for viscoplastic materials

• It is generally possible to write:

$$\begin{array}{lll} \mathbf{A} &=& (\underline{\varepsilon}_{e}, \mathbf{a}) \\ \mathbf{R} &=& (\mathbf{R}_{e}, \mathbf{R}_{a}) \\ \mathbf{R}_{e} &=& \Delta \underline{\varepsilon}_{e} + \Delta \underline{\varepsilon}_{\mathrm{irr}} - \Delta \underline{\varepsilon} \end{array}$$

- Influence of a variation of Δ_E on the variations of integrated variables (<u>ε</u>e, a) ?
- At the solution point R must remain equal to 0

$$\delta \mathbf{R} = \mathbf{0} = \delta \begin{pmatrix} \Delta \underline{\varepsilon}_{\theta} + \Delta \underline{\varepsilon}_{inr} \\ \mathbf{R}_{a} \end{pmatrix} - \delta \begin{pmatrix} \Delta \underline{\varepsilon} \\ \mathbf{0} \end{pmatrix}$$
$$\delta \mathbf{R} = \frac{\partial \mathbf{R}}{\partial \mathbf{A}} - \begin{pmatrix} \delta \Delta \underline{\varepsilon} \\ \mathbf{0} \end{pmatrix} = \mathbf{J} \cdot \delta \Delta \mathbf{A} - \begin{pmatrix} \delta \Delta \underline{\varepsilon} \\ \mathbf{0} \end{pmatrix}$$

• Therefore

$$\delta \Delta A = \mathbf{J}^{-1} \cdot \begin{pmatrix} \delta \Delta \underline{\varepsilon} \\ \mathbf{0} \end{pmatrix}$$

• \mathbf{J}^{-1} can be expressed using sub-blocks

$$\mathbf{J}^{-1} = \mathbf{J}^{\star} = \begin{pmatrix} \mathbf{J}_{ee}^{\star} & \mathbf{J}_{ea}^{\star} \\ \mathbf{J}_{ae}^{\star} & \mathbf{J}_{aa}^{\star} \end{pmatrix},$$

so that

$$\delta \Delta \underline{\varepsilon}_{e} = \mathbf{J}_{ee}^{\star} . \delta \Delta \underline{\varepsilon}$$

• Using Hooke's law:

$$\underline{\sigma}(t_1) = \underline{\sigma}(t_0) + \Delta \underline{\sigma} = \underline{\underline{E}}(t_1) : \underline{\varepsilon}_{\boldsymbol{\theta}}(t_1) = \underline{\underline{E}}(t_1) : (\underline{\varepsilon}_{\boldsymbol{\theta}}(t_0) + \Delta \underline{\varepsilon}_{\boldsymbol{\theta}})$$

therefore

$$\delta\Delta\underline{\sigma} = \underline{\underline{E}}(t_1) : \delta\Delta\underline{\varepsilon}_{e} = \underline{\underline{E}}(t_1) : \underline{\underline{J}}_{ee}^{\star} : \delta\Delta\varepsilon$$

Finally

$$\underline{\underline{L}}_{c} = \underline{\underline{E}}(t_{1}) : \underline{\underline{J}}_{ee}^{\star}$$

 In cases where <u>E</u> depends on an integrated variable (e.g. d =damage) the calculation is more difficult

$$\Delta \underline{\sigma} = \frac{\partial \underline{\underline{E}}}{\partial d} \delta \Delta d : \underline{\varepsilon}_{e} + \underline{\underline{E}}(t_{1}) : \Delta \underline{\varepsilon}_{e}$$

$$\underline{\underline{L}}_{c} = \frac{\partial \underline{\underline{E}}}{\partial d} : (\underline{\varepsilon}_{e} \otimes \mathbf{J}_{de}^{\star}) + \underline{\underline{E}} : \underline{\underline{J}}_{ee}^{\star}$$

Explicit	Implicit
easy to implement	difficult to implement
time consuming	efficient
<u> </u>	calculation of $\underline{\underline{L}}_c$

• \underline{L}_c can be evaluated using a perturbation method

$$L_{ijkl}^{c} = \frac{\sigma_{ij}(\Delta \underline{\varepsilon} + \delta \underline{\varepsilon} \underline{\mu}^{kl}) - \sigma_{ij}(\Delta \underline{\varepsilon})}{\delta \varepsilon}$$

• The tensor μ^{kl} is such that:

$$\mu_{ij}^{kl} = \delta_{ik} \delta_{lj}$$

Viscoplastity

- To introduce rate dependency (viscoplastity) the above equations need to be slightly modified
- Additive decomposition of deformations:

$$\underline{\varepsilon} = \underline{\varepsilon}_{e} + \underline{\varepsilon}_{p}$$

Yield surface

$$\phi = \sigma_{\rm eq} - R(p)$$

Normality

$$\underline{\dot{\varepsilon}}_{\rho} = \dot{\rho} \frac{\partial \phi}{\partial \underline{\sigma}}$$

- ϕ can now be positive
- p is given by a specific material dependent law

$$\dot{p} = \mathcal{F}(\phi)$$

$$\mathcal{F}$$
 is such that $\mathcal{F}(x) = 0$ if $x \leq 0$

For instance (Norton low)

$$\dot{p} = \left\langle \frac{\phi}{K} \right\rangle^n$$

Differential equations to be integrated are therefore

$$\frac{\dot{\underline{e}}_{e}}{\dot{p}} = \frac{\dot{\underline{e}} - \dot{p}\underline{n}}{\left\langle \frac{\phi}{\kappa} \right\rangle^{n}}$$

- Runge-Kutta integration is straightforward
- For the implicit integration

$$\begin{array}{rcl} \underline{R}_{e} & = & \Delta \underline{\varepsilon}_{e} + \Delta p \underline{n} - \Delta \underline{\varepsilon} \\ R_{p} & = & \Delta p - \mathcal{F}(\phi) \Delta t \end{array}$$

• Evaluation of the Jacobian matrix: only the derivatives of R_p have to be re-computed with $\mathcal{F}' = \partial \mathcal{F} / \partial \phi$

$$\frac{\partial R_{p}}{\partial \Delta \underline{\varepsilon}_{e}} = -\Delta t \frac{\partial \mathcal{F}}{\partial \phi} \frac{\partial \phi}{\partial \underline{\sigma}} : \frac{\partial \underline{\sigma}}{\partial \underline{\varepsilon}_{e}} : \frac{\partial \underline{\varepsilon}_{e}}{\partial \Delta \underline{\varepsilon}_{e}} = -\theta \Delta t \mathcal{F}' \underline{n} : \underline{\underline{E}}$$
$$\frac{\partial R_{p}}{\partial \Delta p} = -\Delta t \frac{\partial \mathcal{F}}{\partial \phi} \frac{\partial \phi}{\partial R} \frac{\partial R}{\partial p} \frac{\partial p}{\partial \Delta p} = -\theta \Delta t \mathcal{F}' H$$