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Abstract

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To my parents

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Nomenclature

\vec{E}_i	Spatial basis unit vectors, $i = 1, 2, 3$
\vec{e}_i	Material basis unit vectors, $i = 1, 2, 3$
X_i	Initial configuration coordinates, $i = 1, 2, 3$
\vec{X}	Position vector of a material point in the initial configuration
x_i	Current configuration coordinates, $i = 1, 2, 3$
\vec{x}	Position vector of a material point in the current configuration
\vec{r}_o	Position vector of a material point on the centerline in the initial configuration
\vec{r}	Position vector of a material point on the centerline in the current configuration
\vec{u}	Displacement vector of points on the centerline with respect to the initial configuration.
\vec{t}	Position vector of a material point on the local system of a cross-section.
ϕ	Angle between \vec{E}_i and \vec{e}_i
$\mathbf{R}(\phi)$	Rotation matrix
κ	Curvature of the beam centerline.
ϵ	Axial strain of the beam centerline
γ	Shear strain of the beam centerline
\vec{s}	Vector containing ϵ and γ but not κ
u	Displacement component along X_1
v	Displacement component along X_2
\vec{d}	Displacement vector containing u, v and ϕ
\vec{q}	Strain vector containing ϵ, γ and κ
f	Generic objective function

\vec{h}	Vector of discretized constraints
\mathcal{S}	Constrained space for \vec{x}
\mathcal{L}	Lagrangian function associated with the NLP
$\vec{\lambda}$	Vector of Lagrange multipliers associated with equality constraints
Π	Total potential energy function
U	Strain energy function
W	Potential energy of external loads
\mathcal{W}_s	Cross-section strain energy density
\vec{P}	Vector of external loads
\vec{F}_{sec}	Cross-section stress resultants
N	Section axial force
V	Section shear force
M	Section internal moment
E	Young's modulus
G	Shear modulus
I	Cross-section moment of inertia
A	Cross-section area
A_s	Effective area of the cross-section
ℓ	Element length
n_{nod}	Number of non-restrained nodes
n	Number of quadrature points in the element
w_i	Quadrature weight for the i -th quadrature point
ξ	Normalized coordinate associated with X_1
L_i	Lagrange cardinal function
\mathbf{T}	Curvature mapping matrix
Θ	Vandermonde matrix
\vec{y}	Internal field variable vector, associated with unknowns at quadrature points

Λ	Element orientation matrix
θ	Angle of element centerline in the initial configuration with \vec{E}_1
\mathcal{E}^e	Energy contribution of element e in the Lagrangian function
H	Hessian matrix
\vec{z}	Vector of all unknown state variables of the system
\mathbf{k}_s	Section stiffness matrix
K	Stiffness matrix
$\vec{\mathbf{F}}_{int}$	Internal nodal force vector
ΔS	Incremental arc-length parameter
$\vec{\sigma}_f$	Vector of active stress components on a fiber
$\vec{\epsilon}_f$	Strain components corresponding to $\vec{\sigma}_f$
N_s	Cross-section shape function matrix
φ	Shear strain distribution function along the cross-section height
C	Material tangent modulus
$\vec{d}_{l,r}$	Element nodal displacement vector for left (l) or right (r) elememtn edge
$\vec{\mathbf{P}}_{l,r}$	Element externally applied nodal force vector for left (l) or right (r) elememtn edge
σ	Cauchy stress tensor
ϵ	Strain tensor
a	Back stress tensor
q	Current uniaxial yield stress
e^{pl}	Equivalent plastic strain
λ	Plastic parameter
Φ	Yield function
σ_{eq}	Equivalent stress associated with the appropriate yield criterion
H_{kin}	Kinematic hardening modulus
H_{iso}	Isotropic hardening modulus
C, C	Fourth order constitutive tensor and its matrix representation

J_2	Second invariant of the deviatoric stress tensor
σ_y	Yield stress of the virgin material
ζ	Effective stress tensor
\mathbf{w}^d	Deviatoric part of a tensor \mathbf{w}
\mathbf{C}_c^{pl}	Consistent elastoplastic tangent modulus
$\vec{\mathbf{H}}$	Homotopy (vector) function
t	Pseudo-time or homotopy parameter
C^k	Space of functions with their first k derivatives continuous
Γ	A (connected) component, path or loop, in $\vec{\mathbf{H}}^{-1}(\vec{0})$
Ω	Feasible space for an NLP or PNLP program
I_h	Index set of equality constraints
I_g	Index set of inequality constraints
$\vec{\boldsymbol{\mu}}$	Vector of Lagrange multipliers associated with inequality constraints
\bar{I}_g	Index set of active inequality constraints
\mathcal{D}	Subdivision of an interval
$\tau(\bar{I}_g)$	Cell associated with subdivision of Ω
W^q	Subinterval associated with subdivision \mathcal{D}

Acronyms

AB2 Two-step Adams-Bashforth.

AB3 Three-step Adams-Bashforth.

DOF Degree of Freedom.

FEM Finite Element Method.

IS Initial State.

IVP Initial Value Problem.

KKT Karush-Kuhn-Tucker.

LICQ Linear Independence Constraint Qualification.

NLP Non-Linear Programming.

NPNLP Naturally Parameterized Non-Linear Programming.

PC Predictor-Corrector.

PNLP Parametric Non-Linear Programming.

SCC Strict Complementarity Condition.

TPE Total Potential Energy.

TS Target State.

WLS Weighted Least Squares.

WLSE Weighted Least Squares Extrapolation.

WLSIT Weighted Least Squares Implicit Tangent.

WLST Weighted Least Squares Tangent.

Acknowledgements

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CHAPTER 1

INTRODUCTION AND BACKGROUND

1.0.1 Notation

Unless state otherwise, boldface lowercase letters indicate second order tensor. Wherever the same letters appear with an overhead arrow, this indicates the vector notation for the tensor represented with that letter. The convention adopted herein for storing second order tensors in a vector object is the following:

$$\boldsymbol{\sigma} = \begin{bmatrix} \sigma_{11} & \sigma_{12} & \sigma_{13} \\ \sigma_{12} & \sigma_{22} & \sigma_{23} \\ \sigma_{13} & \sigma_{23} & \sigma_{33} \end{bmatrix} \rightarrow \vec{\boldsymbol{\sigma}} = \begin{bmatrix} \sigma_{11} & \sigma_{22} & \sigma_{33} & \sigma_{12} & \sigma_{23} & \sigma_{13} \end{bmatrix}^T$$

For strain tensors, components for which $i \neq j$ are multiplied by 2.

In addition, any boldface letter, lowercase or uppercase, with an overhead arrow indicates a vector object. In contrast, matrices are represented exclusively by uppercase, non-italic, boldface letters. Finally, normal font letters indicate scalar quantities.

CHAPTER 2

HYBRID BEAM ELEMENT FORMULATION

2.1 Introduction

In this chapter we present the formulation of a novel hybrid beam element which is based on NLP principles. The kinematic assumptions adopted fall under the category of geometrically exact or Simo-Reissner beam theory[1, 2, 3, 4], whereby no simplifying approximation is made with respect to the strain-displacement equations. This allows for capturing arbitrarily large displacements and rotations, as well as accounting for the effect of shear deformation at the section level.

As opposed to deriving the system equations from the Galerkin form,, we recast the problem in an NLP framework by utilizing the underlying variational structure. The total potential energy (TPE) functional is augmented with all relevant conditions that enforce the exact kinematics, by introducing a set of Lagrange multipliers that act as conjugate force quantities. The resulting modified functional is then approximated by employing a Gauss-Legendre quadrature rule, which yields the objective function to be minimized. With this particular approach, the primary variables in the element interior contributing to the elastic strain energy are the generalized strain measures of the centroid, which are the unknown quantities at the quadrature points. Displacement measures at the edge nodes of the element, namely, the translations along coordinate axes and the rotation of cross sections, are only associated with the external work. Kinematic consistency between the rotational measures of displacement and strains is enforced by using a Lagrange interpolation scheme for the curvature field, similar to the one used by Neuenhofer and Filippou [5] and Schulz & Filippou [6] for force-based elements. In this work, the points used for the interpolation coincide with the integration points of the quadrature rule used for approximating the energy functional. To avoid ill-conditioning issues of the linearized operator,

we also outline a block-elimination procedure for the linear system involving the Hessian matrix so that the reduced system includes only displacement components. The corresponding stiffness operator is well-conditioned, sparse, banded and symmetric. This renders standard FEM routines from existing codes reusable and, in conjunction with the capability for accuracy even with crude discretization, provides an attractive formulation for fast and accurate computation. Additionally, accuracy and locking free performance are guaranteed with just one element per structural member, even in the presence of arbitrarily large displacements and rotations. Accordingly, the element can also capture high curvature gradients due to plastic hinge formation, in the case of inelastic analysis.

This chapter starts with a brief outline of the beam geometric description while adhering to the derivations by Cardona & Geradin[7], followed by the hybrid NLP formulation for the element. The third section is devoted to implementation details, where special attention is given to the block elimination technique used for the reduction of the linear system involving the Hessian matrix. Finally, the chapter concludes with a set of numerical examples where the focus is the Hybrid NLP element performance with respect to coarse mesh discretizations in nonlinear analysis.

2.2 Geometrically-Exact Model

2.2.1 Beam Kinematics

Let $\{\mathbf{E}\}$ be a fixed orthonormal coordinate frame with unit vectors $\{\vec{\mathbf{E}}_1, \vec{\mathbf{E}}_2, \vec{\mathbf{E}}_3\}$ along the axes X_1, X_2, X_3 . For an initially straight beam of length ℓ with its centerline coinciding with X_1 , the undeformed configuration is completely described by:

$$\vec{\mathbf{X}}(X_1, X_2) = X_1 \vec{\mathbf{E}}_1 + X_2 \vec{\mathbf{E}}_2 = \vec{\mathbf{r}}_o(X_1) + \vec{\mathbf{t}}(X_2) \quad (2.1)$$

where $\vec{\mathbf{r}}_o$ traces the centerline in the reference configuration and $\vec{\mathbf{t}}$ locates a material point on the cross section. Assuming a rectangular cross-section of height h , without loss of generality,

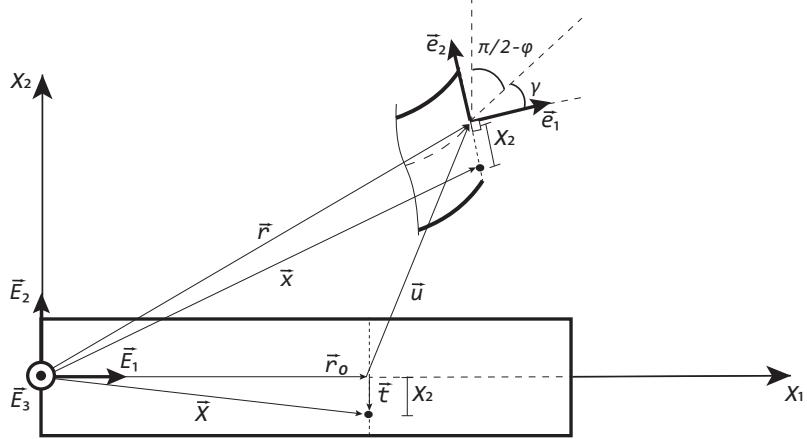


Fig. 2.1. Undeformed and deformed configurations of the beam.

we have that $X_1 \in [0, \ell]$ and $X_2 \in [-\frac{h}{2}, \frac{h}{2}]$. We should note here that X_3 is omitted as it does not explicitly come into the expressions. We nevertheless hold on to the \mathbb{R}^3 vector formalism to maintain consistency with the representation of rotation as a linear operator. After the application of a displacement field $\vec{u}(X_1)$, the deformed configuration is given by vector \vec{x} such that:

$$\vec{r} = \vec{r}_o + \vec{u} \quad (2.2)$$

$$\vec{x} = \vec{r} + X_2 \vec{e}_2 \quad (2.3)$$

where $\{\mathbf{e}\}$ is a local coordinate system attached at cross-sections that completely describes their orientation. For the unit vectors of the local base \vec{e}_1 , \vec{e}_2 , \vec{e}_3 , we let \vec{e}_1 be normal to the cross-section, but not necessarily tangent to the deformed centerline, and \vec{e}_2 (and \vec{e}_3) coincide with the cross-section principal axes of inertia, as shown in Fig. (2.1).

If \mathbf{R} is the rotation operator that rotates $\{\mathbf{E}\}$ to $\{\mathbf{e}\}$, then:

$$\vec{e}_i = \mathbf{R} \vec{E}_i, \quad i = 1, 2, 3 \quad (2.4)$$

For the plane case, the component form of \mathbf{R} reduces to:

$$\mathbf{R} = \begin{bmatrix} \cos \phi & -\sin \phi & 0 \\ \sin \phi & \cos \phi & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad (2.5)$$

Then, using Eq. (2.4), Eq. (2.3) becomes:

$$\vec{x} = \vec{r} + \mathbf{R}\vec{t} \quad (2.6)$$

For Eqs. (2.3),(2.6) to hold in this manner, we tacitly assume that cross-sections remain rigid during deformation, that is, the position of a material point on a cross-section with respect to the centroid does not change.

2.2.2 Strain Measures, Strain-Displacement Relations

Within the context of geometrically-exact formulations, it is common to adopt as a strain measure \vec{s} the difference of the position vector gradients, \vec{r} and \vec{r}_o , with respect to the local frame (e.g. see [2, 7, 8]). Denoting derivatives with respect to X_1 by $(\cdot)'$, we get:

$$\vec{s} = \mathbf{R}^T \vec{r}' - \vec{r}'_o \quad (2.7)$$

whereas the cross-section curvature is given as:

$$\kappa = \phi' \quad (2.8)$$

The axial and shear strain of the centroid, ϵ and γ respectively, can then be retained as follows:

$$\epsilon = \vec{e}_1^T \vec{s} \quad (2.9)$$

$$\gamma = \vec{e}_2^T \vec{s} \quad (2.10)$$

where $\vec{s} = \begin{bmatrix} \epsilon & \gamma & 0 \end{bmatrix}^T$, with the third component, representing the shear strain along the out-of-plane axis, being by assumption zero.

Using Eq. (2.2) and performing the differentiations in Eq. (2.7) with respect to X_1 , we get:

$$\vec{s} = \mathbf{R}^T [\vec{u}' + \vec{E}_1] - \vec{E}_1 \quad (2.11)$$

Deriving the exact differential strain-displacement relations as presented in Reissner¹ is straightforward if we solve Eq. (2.11) for \vec{u}' :

$$\vec{u}' = \mathbf{R}\vec{s} + \vec{e}_1 - \vec{E}_1 \quad (2.12)$$

If we write $\vec{u} = \begin{bmatrix} u & v & 0 \end{bmatrix}^T$, then the explicit expressions for u', v' along with Eq. (2.8) are:

$$u' = \vec{E}_1^T \vec{u}' = (1 + \epsilon) \cos \phi - \gamma \sin \phi - 1 \quad (2.13a)$$

$$v' = \vec{E}_2^T \vec{u}' = (1 + \epsilon) \sin \phi + \gamma \cos \phi \quad (2.13b)$$

$$\phi' = \kappa \quad (2.13c)$$

By introducing vectors \vec{d} and \vec{q} , such that $\vec{d} = \begin{bmatrix} u & v & \phi \end{bmatrix}^T$ and $\vec{q} = \begin{bmatrix} \epsilon & \gamma & \kappa \end{bmatrix}^T$, we can maintain the structure of Eq. (2.12) and express all fields involved in matrix notation, as follows:

$$\vec{d}' = \mathbf{R}\vec{q} + \vec{e}_1 - \vec{E}_1 \quad (2.14)$$

Strain-displacement equations as represented in Eq. (2.14) will be subsequently utilized in the formulation of the beam model.

¹See equations (15) in [1]

2.3 Nonlinear Programming Formulation

The hybrid beam element discretization is presented by utilizing the underlying variational structure of the static problem. The minimizing principle for the TPE is recast here in an NLP formulation and the strain-displacement equations in (2.14) are incorporated, representing the equality constraints of the problem.

2.3.1 NLP framework

The NLP problem adopted herein is of the following form:

$$\begin{aligned} & \text{minimize} && f(\vec{x}) \\ & \text{subject to} && \vec{h}(\vec{x}) = \vec{0} \\ & && \vec{x} \in \mathcal{S} \end{aligned} \tag{2.15}$$

The objective function $f(\vec{x})$ represents the total potential energy of the structure, whereas vector $\vec{h}(\vec{x})$ includes the active constraints of the program. The corresponding Lagrangian is:

$$\mathcal{L}(\vec{x}, \vec{\lambda}) = f(\vec{x}) + \vec{\lambda}^T \vec{h}(\vec{x}) \tag{2.16}$$

where $\vec{\lambda}$ are the Lagrange multipliers. The first-order optimality conditions are given from the following expression:

$$\nabla \mathcal{L} = \vec{0} \tag{2.17}$$

This is the standard form of a nonlinear constrained minimization program with equality constraints, as encountered in the literature (e.g. [9]).

2.3.2 Total potential energy

We consider the following decomposition of the total potential energy of a beam:

$$\Pi = U - W \quad (2.18)$$

where U and W are the stored strain energy and potential energy associated with external loading respectively. For the case of concentrated external loads, Eq. (2.18) can be expressed as:

$$\Pi = \int_0^\ell \mathcal{W}_S(X_1, \vec{q}) dX_1 - \sum_{i=1}^2 \vec{P}_i^T \vec{u}_i \quad (2.19)$$

where \mathcal{W}_S is the strain energy density of a cross section at X_1 and \vec{q} denotes the neutral axis strain vector, \vec{P}_i, \vec{u}_i the element nodal force and displacement vectors, respectively. It can be proven[10] that for strain hardening materials within the premises of small strain elastoplasticity, the exact solution to the problem renders Π an absolute minimum. Equivalently, the exact solution solves the program in Eq. (2.15) for $f = \Pi$ with \vec{h} acting as constraints. Assuming a stable work-hardening material, the corresponding Euler-Lagrange equations yield the constitutive equations of elastoplasticity (e.g. see discussion in Simo and Hughes [11]). As it will be shown in subsequent sections, since the constitutive update is carried out locally at the section fiber level.

Section stress resultants are then given by the gradient of \mathcal{W}_S :

$$\vec{F}_{sec} = \nabla_{\mathbf{q}} \mathcal{W}_S \quad (2.20)$$

with $\vec{F}_{sec} = \begin{bmatrix} N & V & M \end{bmatrix}^T$. In the case of linear elasticity, the energy density is given by

$\mathcal{W}_S = \frac{1}{2}[EA\epsilon^2 + GA_s\gamma^2 + EI\kappa^2]$ and the section resultants are:

$$\vec{\mathbf{F}}_{sec} = \begin{bmatrix} EA\epsilon \\ GA_s\gamma \\ EI\kappa \end{bmatrix}$$

where A_s is the effective shear area of the cross-section. Numerical approximation of Eq. (2.19) by an appropriate quadrature yields:

$$\Pi(\vec{\mathbf{q}}_1, \dots, \vec{\mathbf{q}}_n, \vec{\mathbf{u}}_1, \dots, \vec{\mathbf{u}}_N) = \sum_{i=1}^n w_i \mathcal{W}_{S,i} - \sum_{i=1}^2 \vec{\mathbf{P}}_i^T \vec{\mathbf{u}}_i \quad (2.21)$$

with w_i and n being the weights and the number of integration points respectively, and $\mathcal{W}_{S,i} = \mathcal{W}_S(\vec{\mathbf{q}}_i)$. Eq. (2.21) serves as the objective function of the NLP of Eq. (2.15).

2.3.3 Strain-Displacement Constraints

The constrained conditions are derived by applying the same quadrature rule on the integral form of the strain-displacement equations (2.14):

$$\vec{\mathbf{d}}(\ell) - \vec{\mathbf{d}}(0) = \int_0^\ell \vec{\mathbf{d}}' dX_1 = \int_0^\ell \mathbf{R} \vec{\mathbf{q}} + \vec{\mathbf{e}}_1 - \vec{\mathbf{E}}_1 dX_1 \quad (2.22)$$

Using Eq. (2.4), numerical approximation of the right-hand side integral leads to the element specific constraints:

$$\vec{\mathbf{h}}^A = \vec{\mathbf{d}}(\ell) - \vec{\mathbf{d}}(0) - \sum_{i=1}^n w_i \mathbf{R}_i (\vec{\mathbf{q}}_i + \vec{\mathbf{E}}_1) + \ell \vec{\mathbf{E}}_1 = \vec{\mathbf{0}} \quad (2.23)$$

The component form of Eq. (2.23) can be given as:

$$\vec{h}^A = \begin{bmatrix} u(\ell) - u(0) - \sum_{i=1}^n w_i \left[(\epsilon_i + 1) \cos \phi_i - \gamma_i \sin \phi_i \right] + \ell \\ v(\ell) - v(0) - \sum_{i=1}^n w_i \left[(\epsilon_i + 1) \sin \phi_i + \gamma_i \cos \phi_i \right] \\ \phi(\ell) - \phi(0) - \sum_{i=1}^n w_i \kappa_i \end{bmatrix} \quad (2.24)$$

Although the strain fields appear explicitly only in the weighted evaluation points of the quadrature, the rotational field ϕ is involved in both the integration points and the element edge nodes. As such, we introduce a Lagrange interpolation scheme for the curvature field in the same fashion as in [12]:

$$\kappa(\xi) = \sum_{i=1}^n L_i(\xi) \kappa_i \quad (2.25)$$

where L_i are the Lagrange cardinal functions:

$$L_i(\xi) = \frac{\prod_{j=1, j \neq i}^n (\xi - \xi_j)}{\prod_{j=1, j \neq i}^n (\xi_i - \xi_j)}, \quad \xi = \frac{X_1}{\ell}$$

Substitution of Eq. (2.25) to Eq. (2.13c) yields the expression for rotations ϕ_i :

$$\phi_i - \phi(0) = \sum_{j=1}^n \left(\int_0^{\xi_i} L_j(x) dx \right) \kappa_j = \sum_{i=1}^n T_{ij} \kappa_j \quad (2.26)$$

with derivation of T_{ij} illustrated in Fig. (2.2). In matrix form, the above equation can be directly restated as a linear equality constraint set as:

$$\vec{h}^B = \vec{\phi} - \phi(0) \vec{1} - \mathbf{T} \vec{\kappa} = \vec{0} \quad (2.27)$$

where

$$\vec{\phi} = \begin{bmatrix} \phi_1 & \phi_2 & \cdots & \phi_n \end{bmatrix}^T, \vec{1} = \begin{bmatrix} 1 & 1 & \cdots & 1 \end{bmatrix}^T \in \mathbb{R}^n$$

$$\mathbf{T} = L \begin{bmatrix} \xi_1 & \frac{\xi_1^2}{2} & \cdots & \frac{\xi_1^n}{n} \\ \vdots & \vdots & \ddots & \vdots \\ \xi_n & \frac{\xi_n^2}{2} & \cdots & \frac{\xi_n^n}{n} \end{bmatrix} \Theta^{-1}, \Theta = \begin{bmatrix} 1 & \xi_1 & \xi_1^2 & \cdots & \xi_1^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \xi_n & \xi_n^2 & \cdots & \xi_n^{n-1} \end{bmatrix}$$

We can then collect all active constraints, Eqs. (2.23),(2.27), in one vector \vec{h} :

$$\vec{h} = \begin{bmatrix} \vec{h}^A \\ \vec{h}^B \end{bmatrix} = \begin{bmatrix} \vec{0} \\ \vec{0} \end{bmatrix} \quad (2.28)$$

containing all constraints pertaining to one element. While vector \vec{h}^A of strain-displacement constraints will always contain three components for each element, the number of components in vector \vec{h}^B will depend on the chosen quadrature rule.

2.3.4 Lagrangian function

In order to derive the Lagrangian function in the form of Eq. (2.16), we introduce a vector $\vec{\lambda}$ of additional Lagrange multiplier variables and augment Eq. (2.21) with Eq. (2.28):

$$\mathcal{L} = f + \vec{\lambda}^T \vec{h} \quad (2.29)$$

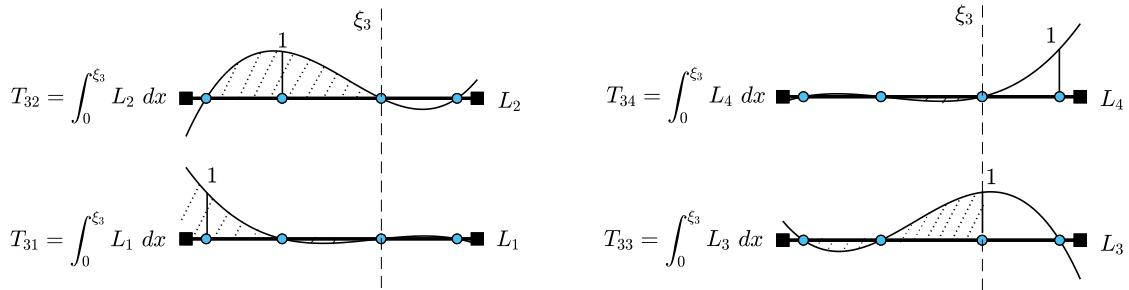


Fig. 2.2. Integration of curvature shape functions.

A more convenient form for the Lagrangian function can also be achieved by expanding f into its constituents, namely, the element stored energy and the potential energy:

$$\mathcal{L} = \sum_{e=1}^{n_{el}} U^e - W + \vec{\lambda}^T \vec{h} \quad (2.30)$$

where:

$$U^e = \sum_{i=1}^n w_i \mathcal{W}_{S,i}^e \quad , \quad W = \sum_{i=1}^N \vec{P}_i^T \vec{u}_i$$

Thereby, additional elements can be simply incorporated by adding their stored energy in the corresponding sum, and considering element constraints by augmenting vectors $\vec{\lambda}$, \vec{h} , after transforming them to the global system. Potential function W is “global” in the sense that external loads and the corresponding work-conjugate displacement degrees of freedom are directly added to the expression and are not influenced by adding nodal contributions from adjacent elements. As a result, no additional connectivity constraints at the element interfaces are needed, and Eq. (2.30) is thus a global function for the whole structure.

2.4 Implementation

The set of vectors associated with internal field variables, $\{\vec{y}_i\}_{i=1}^n$, is defined as $\vec{y}_i = \begin{bmatrix} \vec{q}_i & \phi_i \end{bmatrix}^T = \begin{bmatrix} \epsilon_i & \gamma_i & \kappa_i & \phi_i \end{bmatrix}^T$. Moreover, and according to Fig. (2.3), we designate the local element edge degrees of freedom in vector form as $\hat{\vec{d}}_l = \vec{d}(0)$ and $\hat{\vec{d}}_r = \vec{d}(\ell)$. The element displacement vector is thus $\hat{\vec{d}} = \begin{bmatrix} \hat{\vec{d}}_l^T & \hat{\vec{d}}_r^T \end{bmatrix}^T$. The corresponding global displacement vector \vec{d}_g is related to the local vector $\hat{\vec{d}}$ via the typical linear transformation Λ :

$$\hat{\vec{d}} = \begin{bmatrix} \Lambda & \mathbf{0} \\ \mathbf{0} & \Lambda \end{bmatrix} \vec{d}_g \quad , \quad \Lambda = \begin{bmatrix} \cos \theta & \sin \theta & 0 \\ -\sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad (2.31)$$

The element constraints of Eqs. (2.23),(2.27) transformed in the global system become:

$$\vec{h}_g^A = \mathbf{V}_1 \vec{d}_g - \Lambda^T \left[\sum_{i=1}^n w_i \mathbf{R}_i (\vec{q}_i + \vec{E}_1) - \ell \vec{E}_1 \right] \quad (2.32a)$$

$$\vec{h}_g^B = \vec{\phi} - \mathbf{V}_2 \vec{d}_g - \mathbf{T} \vec{\kappa} \quad (2.32b)$$

where:

$$\mathbf{V}_1 = \begin{bmatrix} -\mathbf{I} & \mathbf{I} \end{bmatrix} \quad , \quad \mathbf{V}_2 = \begin{bmatrix} \vec{0} & \vec{0} & \vec{1} & \vec{0} & \vec{0} & \vec{0} \end{bmatrix}$$

$$\mathbf{I} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad , \quad \vec{0} \in \mathbb{R}^n$$

In the equation above the superscript denoting an element is ommited for clarity. The element

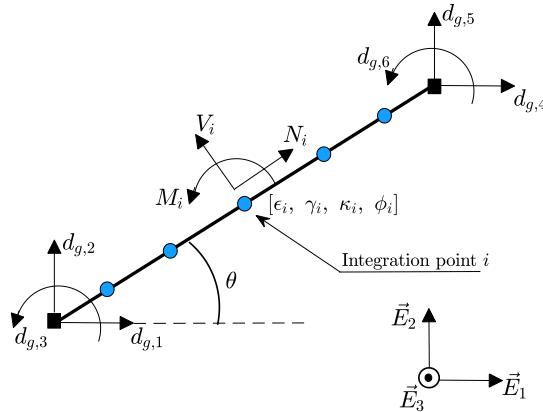


Fig. 2.3. Typical element representing one member.

contribution to the Lagrangian of the assemblage, denoted \mathcal{E}^e , is then:

$$\mathcal{E}^e(\vec{\mathbf{y}}_1, \dots, \vec{\mathbf{y}}_n, \vec{\mathbf{d}}_g, \vec{\lambda}^e) = U^e + (\vec{\lambda}^e)^T \vec{\mathbf{h}}_g^e \quad (2.33)$$

Notice that the external potential function W is not included in Eq. (2.33) as it is added directly via the work done by external forces in the global degrees of freedom and not from element contributions. The element Lagrange multipliers $\vec{\lambda}^e$ are force conjugate measures at the corresponding degrees of freedom and U^e is the sum total of cross-section strain energies. The Lagrangian of the whole system is accordingly given by:

$$\mathcal{L}(\vec{\mathbf{y}}, \vec{\mathbf{d}}_g, \vec{\lambda}) = U - W + \vec{\lambda}^T \vec{\mathbf{h}}_g \quad (2.34)$$

which is a restatement of Eq. (2.30) in the global coordinate system and $\vec{\mathbf{y}} = \begin{bmatrix} \vec{\mathbf{y}}_1^T & \vec{\mathbf{y}}_2^T & \cdots & \vec{\mathbf{y}}_m^T \end{bmatrix}^T$, with m being the total number of quadrature points in the structure.

2.4.1 Equilibrium

In what follows, it is assumed we are dealing with quantities pertaining to the total assemblage, after all element contributions have been resolved. The first-order necessary optimality condition for the Lagrangian function (see Eq. (2.17)), after the imposition of boundary conditions, yields the following relations:

$$\nabla_{\mathbf{y}_i} \mathcal{L} = \nabla_{\mathbf{y}_i} U + [\nabla_{\mathbf{y}_i} \vec{\mathbf{h}}_g] \vec{\lambda} = \vec{0} \quad (2.35a)$$

$$\nabla_{\mathbf{d}_g} \mathcal{L} = -\nabla_{\mathbf{d}_g} W + [\nabla_{\mathbf{d}_g} \vec{\mathbf{h}}_g] \vec{\lambda} = \vec{0} \quad (2.35b)$$

$$\nabla_{\lambda} \mathcal{L} = \vec{\mathbf{h}}_g = \vec{0} \quad (2.35c)$$

Eq. (2.35a) expresses the equilibrium between external and internal forces acting on the i -th cross-section for $i = 1, 2, \dots, m$, while Eq. (2.35b) ensures consistency between externally

applied loads and Lagrange multipliers. Finally, Eq. (2.35c) requires that all constraints are active when the minimum is attained. Explicit expressions for the first and second derivatives of all quantities involved are given in the Appendix. We should note that derivatives of the strain energy U are computed numerically during the cross-section state determination phase in Sec. (2.4.3).

2.4.2 Hessian matrix

The Hessian matrix of the Lagrangian function contains second order information of the system and is utilized during the solution phase. Eqs. (2.35a)-(2.35c) constitute a set of nonlinear algebraic equations and an iterative scheme is needed to solve the system. In block matrix form, the Hessian is provided as:

$$\mathbf{H} = \begin{bmatrix} \nabla_{\mathbf{y}\mathbf{y}}^2 \mathcal{L} & \nabla_{\mathbf{y}\mathbf{d}_g}^2 \mathcal{L} & \nabla_{\mathbf{y}\lambda}^2 \mathcal{L} \\ \nabla_{\mathbf{y}\mathbf{d}_g}^2 \mathcal{L}^T & \nabla_{\mathbf{d}_g\mathbf{d}_g}^2 \mathcal{L} & \nabla_{\mathbf{d}_g\lambda}^2 \mathcal{L} \\ \nabla_{\mathbf{y}\lambda}^2 \mathcal{L}^T & \nabla_{\mathbf{d}_g\lambda}^2 \mathcal{L}^T & \nabla_{\lambda\lambda}^2 \mathcal{L} \end{bmatrix} = \begin{bmatrix} \nabla_{\mathbf{y}\mathbf{y}}^2 \mathcal{L} & \mathbf{0} & \nabla_{\mathbf{y}} \vec{\mathbf{h}}_g^T \\ \mathbf{0} & \mathbf{0} & \nabla_{\mathbf{d}_g} \vec{\mathbf{h}}_g^T \\ \nabla_{\mathbf{y}} \vec{\mathbf{h}}_g & \nabla_{\mathbf{d}_g} \vec{\mathbf{h}}_g & \mathbf{0} \end{bmatrix} \quad (2.36)$$

The resulting Hessian is fairly sparse. Moreover, the block matrix $\nabla_{\mathbf{y}\mathbf{y}} U$ which contains section stiffness information is block-diagonal (see Appendix B) and its symmetry guarantees the full symmetry of \mathbf{H} .

2.4.3 Solution scheme

As mentioned in Appendix B, the nonlinear system of equations in Eq. (2.17) is solved in an incremental-iterative fashion. Linearization around the current iteration k yields:

$$\nabla \mathcal{L}^k + \mathbf{H}^k \delta \vec{\mathbf{z}}^k = \vec{\mathbf{0}} \quad (2.37)$$

where \vec{z} is the vector of all unknowns:

$$\vec{z} = \begin{bmatrix} \vec{y} \\ \vec{d}_g \\ \vec{\lambda} \end{bmatrix}, \quad \text{with} \quad \vec{y} = \begin{bmatrix} \vec{y}_1 \\ \vdots \\ \vec{y}_m \end{bmatrix}$$

Reference to the current iteration step is omitted subsequently for clarity. Solving Eq. (2.37) directly may be cumbersome in some cases, especially for problems that require denser distribution of integration points (e.g. plasticity), since the Hessian is not banded and may also be badly conditioned. In addition, implementation of continuation schemes in order to attain solutions past critical points is more involved since the state variable vector contains displacements, strains and Lagrange multiplier unknowns. A different approach is hence sought, where the initial linear system is reduced to a smaller one involving only the displacement vector \vec{d}_g .

Restating the system in terms of its distinct vector components \vec{y} , \vec{d}_g and $\vec{\lambda}$ gives:

$$\begin{bmatrix} \nabla_y \mathcal{L} \\ \nabla_{d_g} \mathcal{L} \\ \nabla_\lambda \mathcal{L} \end{bmatrix} + \begin{bmatrix} \nabla_{yy}^2 \mathcal{L} & \mathbf{0} & \nabla_y \vec{h}_g^T \\ \mathbf{0} & \mathbf{0} & \nabla_{d_g} \vec{h}_g^T \\ \nabla_y \vec{h}_g & \nabla_{d_g} \vec{h}_g & \mathbf{0} \end{bmatrix} \begin{bmatrix} \delta \vec{y} \\ \delta \vec{d}_g \\ \delta \vec{\lambda} \end{bmatrix} = \begin{bmatrix} \vec{0} \\ \vec{0} \\ \vec{0} \end{bmatrix} \quad (2.38)$$

Utilizing Eqs. (2.35a)-(2.35c) and setting $\delta \vec{\lambda} = \vec{\lambda}^{k+1} - \vec{\lambda}^k$, Eq. (2.38) can be rewritten as:

$$\begin{bmatrix} \nabla_y U \\ -\vec{P} \\ \vec{h}_g \end{bmatrix} + \begin{bmatrix} \nabla_{yy}^2 \mathcal{L} & \mathbf{0} & \nabla_y \vec{h}_g^T \\ \mathbf{0} & \mathbf{0} & \nabla_{d_g} \vec{h}_g^T \\ \nabla_y \vec{h}_g & \nabla_{d_g} \vec{h}_g & \mathbf{0} \end{bmatrix} \begin{bmatrix} \delta \vec{y} \\ \delta \vec{d}_g \\ \vec{\lambda}_{k+1} \end{bmatrix} = \begin{bmatrix} \vec{0} \\ \vec{0} \\ \vec{0} \end{bmatrix} \quad (2.39)$$

As mentioned previously, the gradient of the strain energy with respect to the strain vector represents the cross-sectional stress resultants. Solving the first equation in the system of Eq.

(2.39) for $\delta\vec{y}$ we get:

$$\delta\vec{y} = -\nabla_{yy}^2 \mathcal{L}^{-1} \left[\vec{F}_{sec} + \nabla_y \vec{h}_g^T \vec{\lambda}_{k+1} \right] \quad (2.40)$$

Substituting Eq. (2.40) in the third equation of Eq. (2.39) and solving for $\vec{\lambda}^{k+1}$ yields:

$$\vec{\lambda}^{k+1} = \mathbf{B}^{-1} [\vec{h}_g - \vec{b}] + \mathbf{B}^{-1} [\nabla_{d_g} \vec{h}_g] \delta \vec{d}_g \quad (2.41)$$

By substituting Eq. (2.41) into the second system equation Eq. (2.35b) we arrive at the familiar static equilibrium form of the system of equations:

$$\mathbf{K} \delta \vec{d}_g = \vec{P} - \vec{F}_{int} \quad (2.42)$$

which can now be solved for the iterative displacement vector $\delta \vec{d}_g$. Vectors \vec{b} , \vec{F}_{int} and matrices \mathbf{B} , \mathbf{K} , are given by the following explicit formulas in the global system:

$$\vec{b} = [\nabla_y \vec{h}_g] [\nabla_{yy}^2 \mathcal{L}]^{-1} \vec{F}_{sec} \quad (2.43a)$$

$$\vec{F}_{int} = [\nabla_{d_g} \vec{h}_g]^T [\mathbf{B}]^{-1} [\vec{h}_g - \vec{b}] \quad (2.43b)$$

$$\mathbf{B} = [\nabla_y \vec{h}_g] [\nabla_{yy}^2 \mathcal{L}]^{-1} [\nabla_y \vec{h}_g]^T \quad (2.43c)$$

$$\mathbf{K} = [\nabla_{d_g} \vec{h}_g]^T [\mathbf{B}]^{-1} [\nabla_{d_g} \vec{h}_g] \quad (2.43d)$$

where $\vec{b} \in \mathbb{R}^p$, $\vec{F}_{int} \in \mathbb{R}^{3N}$, N is again the number of structural nodes, $\mathbf{B} \in \mathbb{R}^{p \times p}$ with $p = m + 3n_{nel}$, and $\mathbf{K} \in \mathbb{R}^{3N \times 3N}$.

Equivalently, an assembly process can also be implemented by casting Eqs. (2.43a), (2.43c) in local form, where they can be further simplified by being expanded in terms of element

cross-section contributions:

$$\vec{\mathbf{b}}^e = \sum_{i=1}^n [\nabla_{\mathbf{y}_i} \vec{\mathbf{h}}_g] [\nabla_{\mathbf{y}_i}^2 \mathcal{L}]^{-1} \vec{\mathbf{F}}_{sec}^{(i)} \quad (2.44a)$$

$$\mathbf{B}^e = \sum_{i=1}^n [\nabla_{\mathbf{y}_i} \vec{\mathbf{h}}_g] [\nabla_{\mathbf{y}_i}^2 \mathcal{L}]^{-1} [\nabla_{\mathbf{y}_i} \vec{\mathbf{h}}_g]^T \quad (2.44b)$$

where n is the number of element quadrature points, $\vec{\mathbf{b}}^e \in \mathbb{R}^{n+3}$ and $\mathbf{B}^e \in \mathbb{R}^{(n+3) \times (n+3)}$. The element stiffness matrix $\mathbf{K}^e \in \mathbb{R}^{6 \times 6}$ and internal force vector $\vec{\mathbf{F}}_{int}^e \in \mathbb{R}^6$ are given by the same expressions as in Eqs. (2.43b), (2.43d), but with the gradients now cast in the local element form. For the assembly, the standard FEM routines can be directly employed and the resulting global stiffness operator retains all properties typically associated with it in the context of classical finite element analysis, i.e. it is a symmetric positive definite matrix, it is well-conditioned and, importantly, it is sparse and banded. Hence, in this case, the global internal force vector and the global stiffness matrix are given by the standard assembly process, designated here via operator Λ :

$$\vec{\mathbf{F}}_{int} = \prod_{e=1}^{n_{el}} \vec{\mathbf{F}}_{int}^e \quad (2.45)$$

$$\mathbf{K} = \prod_{e=1}^{n_{el}} \mathbf{K}^e \quad (2.46)$$

with

$$\vec{\mathbf{F}}_{int}^e = [\nabla_{\mathbf{d}_g} \vec{\mathbf{h}}_g^e]^T [\mathbf{B}^e]^{-1} [\vec{\mathbf{h}}_g^e - \vec{\mathbf{b}}^e] \quad (2.47)$$

$$\mathbf{K}^e = [\nabla_{\mathbf{d}_g} \vec{\mathbf{h}}_g^e]^T [\mathbf{B}^e]^{-1} [\nabla_{\mathbf{d}_g} \vec{\mathbf{h}}_g^e] \quad (2.48)$$

where $\vec{\mathbf{h}}_g^e \in \mathbb{R}^{n+3}$ denotes the *element* constraint vector given by Eq. (2.28) and $\nabla_{\mathbf{d}_g} \vec{\mathbf{h}}_g^e \in \mathbb{R}^{(n+3) \times 6}$ its gradient with respect to the global displacement degrees of freedom (DOFs) asso-

ciated with it, is given in Appendix A. It is clear that with this formulation the inversion of large global matrices for the computation of \mathbf{K} and $\vec{\mathbf{F}}_{int}$ is avoided and, instead, only inversion of the local elements flexibility matrices \mathbf{B}^e is required. Given that for highly nonlinear problems we typically have $n \in \mathbb{N}([5, 10])$ for satisfactory accuracy, this results in an element flexibility matrix of dimension $\dim(\mathbf{B}^e) \leq 13$, thus accelerating the analysis considerably.

Having written the system of equations in the form of Eq. (2.42), implementation of arc-length type schemes is now straightforward. In the present work, we adopt the algorithm proposed by Crisfield [13] whereby the additional equation supplemented to the system is:

$$\Delta \vec{\mathbf{d}}_g^T \Delta \vec{\mathbf{d}}_g = \Delta S^2 \quad (2.49)$$

with ΔS being the user-specified arc-length parameter. The incremental displacement vector $\Delta \vec{\mathbf{d}}_g$ is updated in each iteration as follows:

$$\Delta \vec{\mathbf{d}}_g^k = \Delta \vec{\mathbf{d}}_g^{k-1} + \delta \vec{\mathbf{d}}_g^k \quad (2.50)$$

where k denotes the current iteration and $\delta \vec{\mathbf{d}}_g^k$ is the vector of iterative displacements.

Solution updating

After the determination, at an arbitrary iteration k within step j , of the iterative displacement vector $\delta \vec{\mathbf{d}}_g^k$ from Eq. (2.42), the strain and Lagrange multiplier vectors in Eqs. (2.40),(2.41) have then to be updated. The detailed steps of the updating procedure are as follows:

Step j , iteration k : $\left\{ \vec{\mathbf{y}}_j^0, \quad \vec{\mathbf{d}}_{g,j}^0, \quad \vec{\lambda}_j^0, \quad \vec{\mathbf{P}}_j, \quad \Delta \vec{\mathbf{y}}_j^{k-1}, \quad \Delta \vec{\mathbf{d}}_{g,j}^{k-1} \right\}$

(1): Get section stiffnesses $\nabla_{yy}^2 \mathcal{L}^k$ from Eq. (B.3) in Appendix B and section forces $\vec{\mathbf{F}}_{sec}^k$ from section integration, Eq. (2.55).

(2): Evaluate $\vec{\mathbf{b}}^k$ and \mathbf{B}^k from Eqs. (2.43a), (2.43c) respectively (or, alternatively, $\vec{\mathbf{b}}^{e,k}$, $\mathbf{B}^{e,k}$ from Eqs. (2.44a), (2.44b)).

(3): Evaluate \mathbf{K}^k from Eq. (2.43d), (or, alternatively, from Eqs. (2.46), (2.48)).

(4): Solve Eq. (2.42) for $\delta\vec{\mathbf{d}}_g^k$.

(5): Update incremental displacement vector: $\Delta\vec{\mathbf{d}}_{g,j}^k \leftarrow \Delta\vec{\mathbf{d}}_{g,j}^{k-1} + \delta\vec{\mathbf{d}}_g^k$

(6): Update displacement vector: $\vec{\mathbf{d}}_{g,j}^k \leftarrow \vec{\mathbf{d}}_{g,j}^0 + \Delta\vec{\mathbf{d}}_g^k$

(7): Update Langrange multiplier $\vec{\lambda}_j^k$ from Eq. (2.41).

(8): Evaluate iterative strain vector $\delta\vec{\mathbf{y}}^k$ from Eq. (2.40).

(9): Update incremental strain vector: $\Delta\vec{\mathbf{y}}_j^k \leftarrow \Delta\vec{\mathbf{y}}_j^{k-1} + \delta\vec{\mathbf{y}}^k$

(10): Update total strain vector: $\vec{\mathbf{y}}_j^k \leftarrow \vec{\mathbf{y}}_j^0 + \Delta\vec{\mathbf{y}}_j^k$

(11): Evaluate $\vec{\mathbf{F}}_{int}^k$ from Eq. (2.43b) (or, alternatively, from Eqs. (2.45), (2.47)).

(12): Check $\|\vec{\mathbf{P}}_j - \vec{\mathbf{F}}_{int}^k\| \leq \text{tol} \cdot \|\vec{\mathbf{P}}_j - \vec{\mathbf{F}}_{int}^0\|$.

(13): If **FALSE**, set $k \leftarrow k + 1$ and go to (1). If **TRUE**, set $j \leftarrow j + 1$, $\vec{\mathbf{y}}_j^0 \leftarrow \vec{\mathbf{y}}_{j-1}^k$, $\vec{\mathbf{d}}_{g,j}^0 \leftarrow \vec{\mathbf{d}}_{g,j-1}^k$, $\vec{\lambda}_j^0 \leftarrow \vec{\lambda}_{j-1}^k$.

2.4.4 Cross-section state determination

During the updating scheme described in the previous section, the section stress resultants $\vec{\mathbf{F}}_{sec}$, along with the matrix $\nabla_{yy}^2 \mathcal{L}$, are required. The latter, as seen subsequently, corresponds to the generalized section stiffnesses. In the general case where inelastic behavior is considered, the stress-strain constitutive law has to be integrated at the cross-section level and the material properties have to be updated accordingly. In this work, each cross-section is discretized in n_l number of layers and the stress update is performed independently for each layer. This is equivalent to the composite midpoint rule applied along the height of the cross-section. The shear coefficient k_s is introduced as a correction factor for the simplifying aforementioned assumption, in accordance with Cowper [14].

Below we present the kinematic and constitutive relations for a cross-section object. The (multiaxial) constitutive law at the fiber level is treated in the next chapter. The update procedure is regarded as strain-driven, in the sense that the known initial state is updated given an increment in the centerline strains at a particular quadrature point. First, we define the stress and strain vectors associated with a particular fiber:

$$\vec{\sigma}_f = \begin{bmatrix} \sigma_{11} & \sigma_{12} \end{bmatrix}^T, \quad \vec{\epsilon}_f = \begin{bmatrix} \epsilon_{11} & \gamma_{12} \end{bmatrix}^T \quad (2.51)$$

In line with the plane section hypothesis, the fiber strain vector can be determined from the centerline strain vector as follows:

$$\begin{aligned} \epsilon_{11} &= \varepsilon - X_2 \kappa \\ \gamma_{12} &= \varphi(X_2) \gamma \end{aligned} \iff \vec{\epsilon}_f = \mathbf{N}_s \vec{q}, \quad \mathbf{N}_s = \begin{bmatrix} 1 & 0 & -X_2 \\ 0 & \varphi(X_2) & 0 \end{bmatrix} \quad (2.52)$$

where $\varphi(X_2)$ is an as of now unspecified function that defines a shear strain distribution along the height of the cross-section.

Thus, given $\{\vec{q}, \vec{q}^{pl}\}$, $\Delta\vec{q}$, along with a set of internal state variables (e.g. accumulated plastic strain), we can evaluate the incremental strains at the midpoint of a layer at distance X_2 from the neutral axis as follows:

$$\vec{\epsilon}_f = \mathbf{N}_s \vec{q} \quad (2.53a)$$

$$\Delta\vec{\epsilon}_f = \mathbf{N}_s \Delta\vec{q} \quad (2.53b)$$

where it is reminded that \vec{q} is the vector containing the neutral axis generalized strains and $\vec{\epsilon}_f$ is the strain vector associated with the fiber midpoint. With Eqs. (2.53a), (2.53b) and the initial state known, we can perform the stress update which will yield the updated stress vector $\vec{\sigma}_f$ and the (consistent) elastoplastic modulus for the fiber. The conjugate stress resultants associated with the assumed kinematic assumptions are derived from the element virtual work equation:

$$\vec{P}_l^T \delta \vec{d}_l + \vec{P}_r^T \delta \vec{d}_r = \int_V \vec{\sigma}_f^T \delta \vec{\epsilon}_f \, dV = \int_0^L \left[\int_A \mathbf{N}_s^T \vec{\sigma}_f dA \right]^T \vec{q} \, dX_1 \quad (2.54)$$

where we define the conjugate section stress resultants associated with \vec{q} as follows:

$$\vec{F}_{sec}^{(i)} = \int_A \mathbf{N}_s^T \vec{\sigma}_f \, dA \quad (2.55)$$

Application of the composite midpoint rule on Eq. (2.55) yields:

$$\vec{F}_{sec}^{(i)} \approx \begin{bmatrix} \sum_{j=1}^{n_l} \sigma_{11,j} \Delta A_j \\ \sum_{j=1}^{n_l} \sigma_{12,j} \Delta A_j \\ \sum_{j=1}^{n_l} X_{2,j} \sigma_{11,j} \Delta A_j \end{bmatrix} \quad (2.56)$$

where ΔA_j the area of layer j .

Let $\mathbf{C} = \partial \vec{\sigma}_f / \partial \vec{\epsilon}_f$ designate the tangent modulus at a fiber. During elastic steps, $\mathbf{C} \equiv \mathbf{C}^{el} = \text{diag}[E, G]$. In the next chapter we will see that during plastic steps the diagonal components of \mathbf{C} are, generally, also non-zero. The tangent section stiffness is derived as follows:

$$\mathbf{k}_{sec}^{(i)} = \frac{\partial \vec{F}_{sec}^{(i)}}{\partial \vec{q}} = \int_A \mathbf{N}_s^T \frac{\partial \vec{\sigma}_f}{\partial \vec{q}} \, dA = \int_A \mathbf{N}_s^T \frac{\partial \vec{\sigma}_f}{\partial \vec{\epsilon}_f} \frac{\partial \vec{\epsilon}_f}{\partial \vec{q}} \, dA = \int_A \mathbf{N}_s^T \mathbf{C} \mathbf{N}_s \, dA \quad (2.57)$$

Application of the midpoint rule on Eq. (2.57) yields the following component form for $\mathbf{k}_s^{(i)}$:

$$\mathbf{k}_{sec}^{(i)} = \begin{bmatrix} C_{11}^i & \varphi^i C_{12}^i & -X_2^i C_{11}^i \\ \varphi^i C_{21}^i & (\varphi^i)^2 C_{22}^i & -x_2^i \varphi^i C_{21}^i \\ -X_2^i C_{11}^i & -x_2^i \varphi^i C_{12}^i & (X_2^i)^2 C_{11}^i \end{bmatrix} \quad (2.58)$$

where C_{ij} are the components of the elastoplastic consistent tangent modulus of the fiber. The

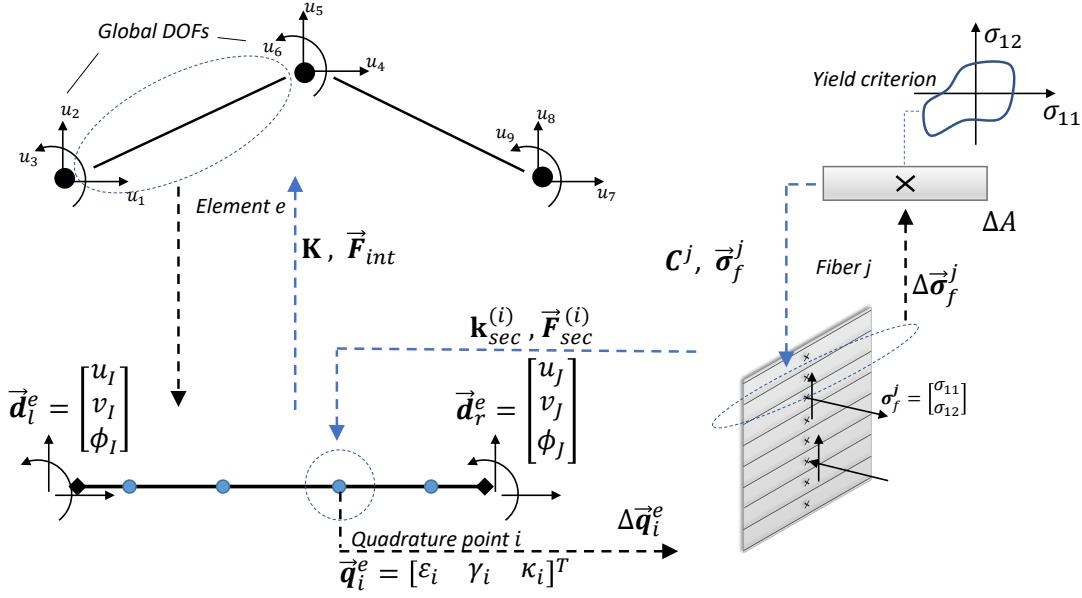


Fig. 2.4. Solution process stages. Quantities \mathbf{K}^e and $\vec{\mathbf{F}}_{int}$ stand for stiffness matrix and internal nodal force vector for element e respectively.

section stiffness is then given by the following sum:

$$\mathbf{k}_{sec} = \sum_{i=1}^{n_f} \mathbf{k}_{sec}^{(i)} \Delta A^i \quad (2.59)$$

Given an increment in nodal displacements at the global/structural level, the assembly process has to loop over all elements in order to determine the element stiffness and internal nodal forces. This procedure, illustrated in Fig. 2.4, involves analysis at three different levels: the global/element, the section state determination, and the fiber stress update. Having briefly outlined the basic theory for the hybrid element as it pertains to the first two levels mentioned above, next we introduce the elastoplastic stress update formulation, which pertains to the third level of the solution process and results in a fast return mapping algorithm.

The generalized section stiffness, $\nabla_{\mathbf{y}_i \mathbf{y}_i}^2 \mathcal{L}$, is given by the second derivatives of the Lagrangian with respect to strain vector $\vec{\mathbf{q}}_i = [\vec{\mathbf{q}}_i^T \ \phi_i]^T$:

$$\nabla_{\mathbf{y}_i \mathbf{y}_i}^2 \mathcal{L} = \nabla_{\mathbf{y}_i \mathbf{y}_i}^2 U + \nabla_{\mathbf{y}_i} ([\nabla_{\mathbf{y}_i} \vec{\mathbf{h}}_g] \vec{\lambda}) \quad (2.60)$$

where:

$$\nabla_{\mathbf{y}_i \mathbf{y}_i}^2 U = w_i \begin{bmatrix} \mathbf{k}_{sec}^{(i)} & \mathbf{0} \\ \mathbf{0} & 0 \end{bmatrix}$$

For the explicit expression of the second term in the right-hand side of Eq. (2.60), see Appendix B. It generally includes parameters associated with section rotations and Lagrange multipliers, and is greatly simplified if small displacement assumptions are used.

2.5 Numerical Examples

In this section we present the capabilities and efficiency of the proposed element in a number of well-known benchmark examples. In cases where analytical solutions are available, they are included in the comparisons. Our results are compared against the ones achieved i) with the structural FEM software OpenSees[15], using its flexibility-based beam element [5, 16] with five quadrature points and fiber section integration, and ii) with the FEM software Abaqus[17], using plane-stress modeling for the frame members. For the elastic analyses, reduced integration linear quadrilateral elements (CPS4R) were used in Abaqus to avoid shear-locking, while for the elastoplastic simulations, reduced integration quadratic quadrilaterals (CPS8R) elements were used instead. For all cases the Gauss-Legendre quadrature rule is used for approximating the integrals in Eqs. (2.19),(2.22). For the elastoplastic analyses, we assume a linear isotropic hardening law and the elastoplastic modulus is given as a percentage r of the elastic modulus E . Moreover, we assume a rigid rectangular cross-section shape for all cases, with shear coefficient $k_s = 0.870$ [14]. For all examples, the SI units corresponding to material and geometric properties are Pa and m, while for the load units Newton(N) is used. The convergence termination criterion for the iterative procedure was 10^{-7} for all cases. Shear deformation is allowed to occur in all analyses unless stated otherwise.

2.5.1 Cantilever with tip load

In this first example we demonstrate the element performance with fully nonlinear assumptions against available analytical solutions, as well as its locking-free behavior when small displacement assumptions hold.

Nonlinear analyses

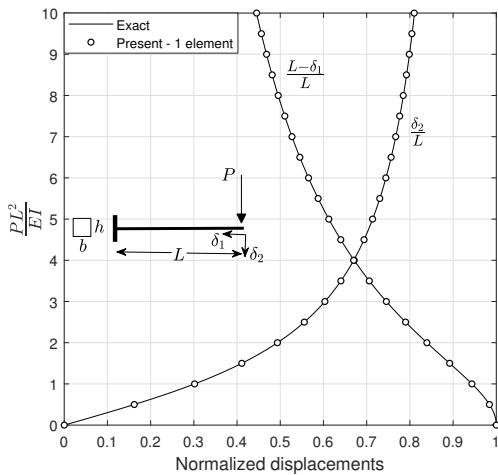
Both elastic and inelastic geometrically nonlinear cases are considered. One element and six integration points are used, with the beam assumed to be inextensible, and the results are presented in Fig. (2.5). The elastic case assumed $EI = 10$, $L = 1$ and shear-rigid response for comparison purposes with available exact equilibrium paths[18], while the inelastic case allows for shear effects to take place with $GA_s = 500$. In order to maintain consistency with the attributes $EI = 10$ and $L = 1$, we used $E = 200 \times 10^9$ for Young's modulus and prescribed the height of the cross-section to be $h = 0.02$, which is now required in order to carry out the cross-sectional stress-update. For the material nonlinearity example, the yield stress is $\sigma_y = 180 \times 10^7$ with a tangent-to-elastic modulus ratio $r = 3\%$ and results are compared against Abaqus and OpenSees. As can be seen in Fig. (2.5), in both cases remarkable accuracy is achieved with just one element. Slight differences between the Abaqus quadrilateral model and the beam models are observed when extensive plastification takes place. In Table 2.1 we also provide distinct comparisons with exact values for the elastic case. It is clear that when the normalized load is equal to 1, displacements remain small, and thus, only $n = 4$ quadrature points suffice to achieve small absolute error. At the maximum load level, however, $n = 6$ quadrature points were required for highly accurate results. For the shear-rigid case exact solutions are computed from numerical evaluation of the elliptic integrals that describe the problem. The reader can also consult Table 1 in [19] for a wide range of values reported. Additional results pertaining to the elastic shear-flexible case are also reported in Table 2.1. For this case, we used again $GA_s = 500$, $EI = 10$, $L = 1$ and compared against available analytical solutions from[20], where the normalized load level therein is held constant and equal to 1. The values correspond-

Table 2.1 Comparison of tip normalized displacements for different load levels for elastic response.

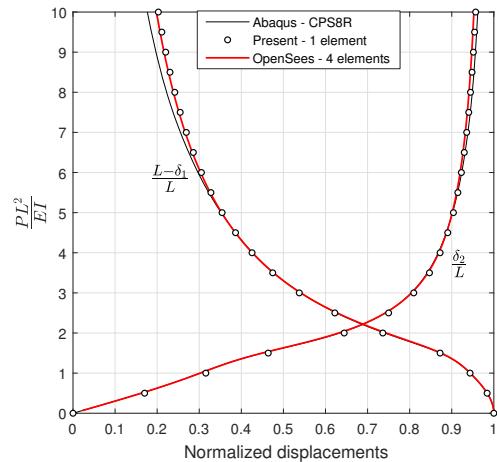
$\frac{PL^2}{EI}$	n	Shear Rigid			Shear Flexible			
		Exact	Present	Error	Exact	Present	Error	
$\frac{\delta_2}{L}$	1	4	0.3017207	0.3017206	1.12E-07	0.3178138	0.3178137	1.64E-07
$\frac{L}{L}$	10	6	0.8106090	0.8106113	2.32E-06	-	0.8536137	-
$\frac{L - \delta_1}{L}$	1	4	0.9435668	0.9435666	1.82E-07	0.9386844	0.9386842	1.75E-07
$\frac{L}{L}$	10	6	0.4450044	0.4450017	2.62E-06	-	0.4208910	-

ing to the maximum normalized load equal to 10 are also reported. Although exact solutions are not available for this load level to compare against, we can see that the relative change in the transverse and axial displacement with respect to the shear-rigid case is consistent and about 5.30% and 5.40%, respectively.

In Fig. (2.6) we see how the moment and curvature distributions change based on the beam deformations for both cases of Fig. (2.5), captured at distinct load levels. The moment values at the left fixed end are also reported on the graphs. The observed nonlinear distributions for the elastic case are due to the deformed shape of the beam being drastically different from its initial configuration. As seen, an important feature of the formulation is its ability to capture curvature localization when plastic hinges are also formed. In displacement-based conventional beam elements, a fine mesh is in constrast required in the areas where plastification is expected.



(a) Elastic analysis



(b) Inelastic analysis

Fig. 2.5. Geometrically nonlinear analyses of cantilever.

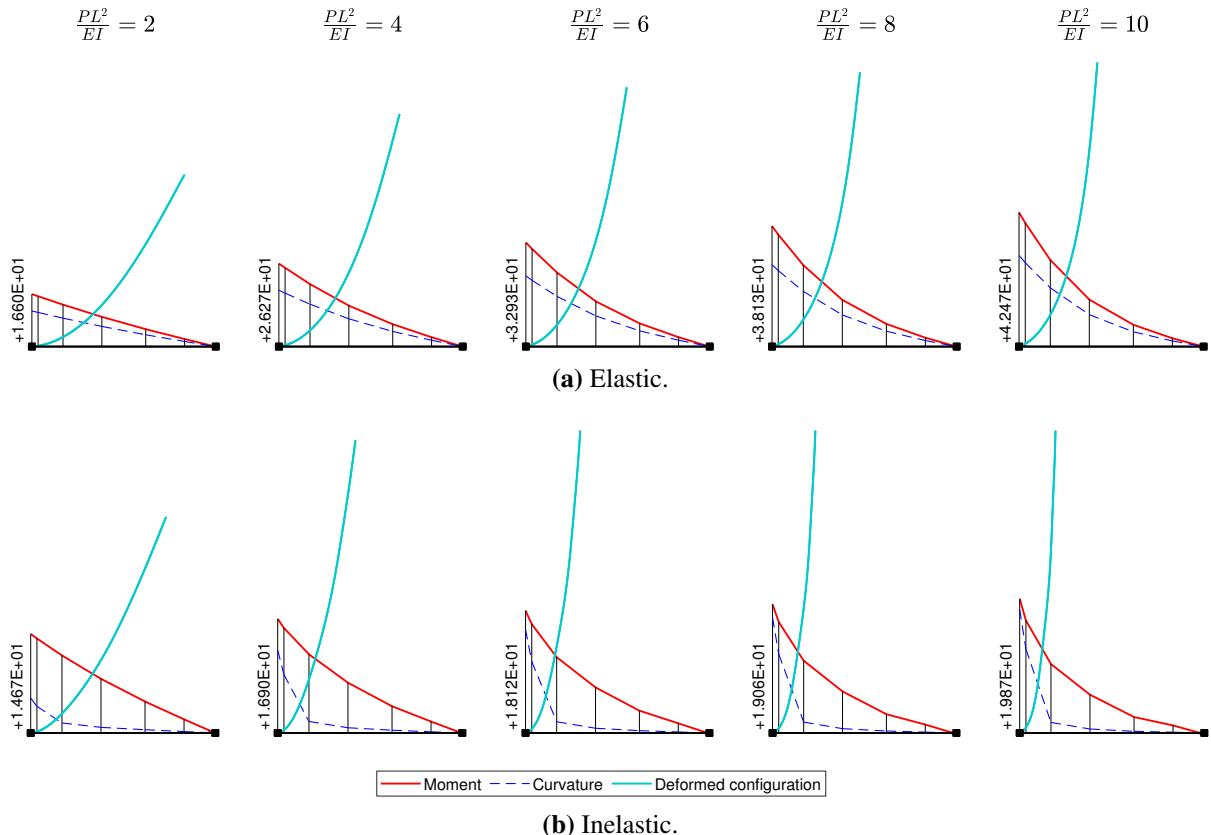


Fig. 2.6. Moment and curvature distributions for the nonlinear analyses of the cantilever - one element with $n = 6$ quadrature points.

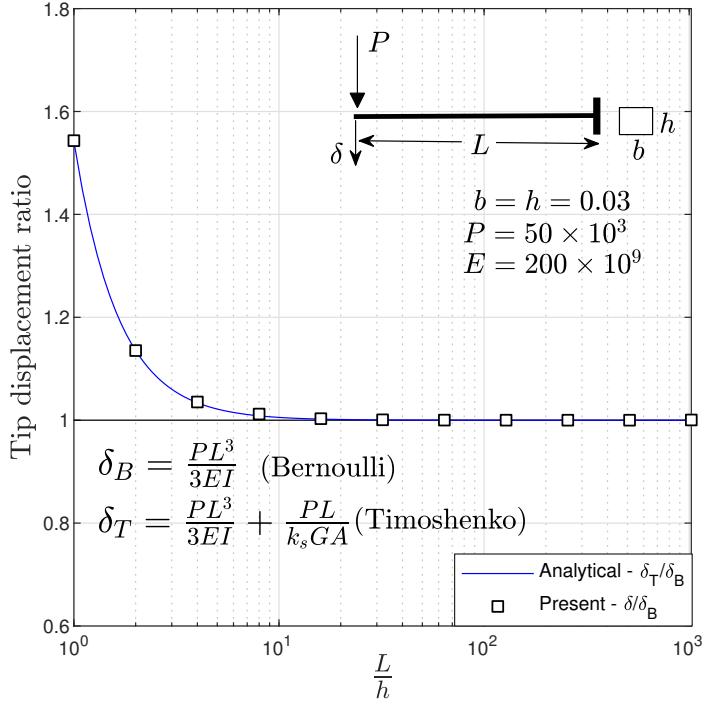


Fig. 2.7. Influence of L/h on the tip deflection and comparison with analytical solution.

Shear-locking

Shear locking may occur in shear deformable elements undergoing bending. Due to inconsistent interpolation of fields that are naturally related, spurious shear stresses develop and the element response appears to be stiffer, particularly as the beam slenderness is increased. Appropriate numerical treatments include using field consistent higher order elements or applying reduced numerical integration [21, 22, 23]. Here, we perform small displacement elastic analyses for various length-to-thickness ratios. The tip displacement ratios of δ_T/δ_B and δ/δ_B are then plotted against the length to thickness ratio L/h , where δ_B and δ_T denote the Bernoulli and Timoshenko displacements respectively. Only one element is used with two integration points. The section dimensions are kept fixed, while the length L of the beam varies. Results, along with used data are shown in Fig. (2.7), where it can be seen that the proposed element exhibits locking-free behavior.

2.5.2 Lee's frame

Lee's frame is a well-known benchmark for geometrically nonlinear capabilities of beam-column element formulations. It was first analyzed by Seng-Lip Lee et.al. in [24] where the exact solution to the problem was also provided. We first consider here the classic case, where the frame is pinned on both supports. An investigation on the effect of shear flexibility to the elastic response is also carried out by varying the length-to-thickness ratio, while maintaining large-displacement assumptions. It is demonstrated that for ratios smaller than five, the equilibrium path ceases to be similar to the shear-rigid case. Lastly, we investigate the response of the same frame when both supports are fixed. This again leads to a drastically different response, with abrupt changes in the equilibrium path. In all cases, the centerline is assumed inextensible.

Frame with pinned supports

An elastic, geometrically nonlinear analysis is first performed. The minimum number of elements that can be used for this specific example is three and a 4-point quadrature is applied for each element, whereas twelve flexibility elements are required in OpenSees to achieve comparable accuracy. For the elastoplastic analysis, again three elements are used in our simulation with eight integration points, whereas fourteen flexibility elements are now required in OpenSees for comparable accuracy. The yield stress is $\sigma_y = 1.3 \times 10^9$ and the tangent-to-elastic modulus is $r = 3.00\%$. In the elastic analysis, the frame members are also considered shear-rigid, in order to remain consistent to the initial assumptions in the original investigation[24]. Results are shown in Fig. (2.8), with the colored markers corresponding to the deformed profiles of the same color shown in Fig. (2.9).

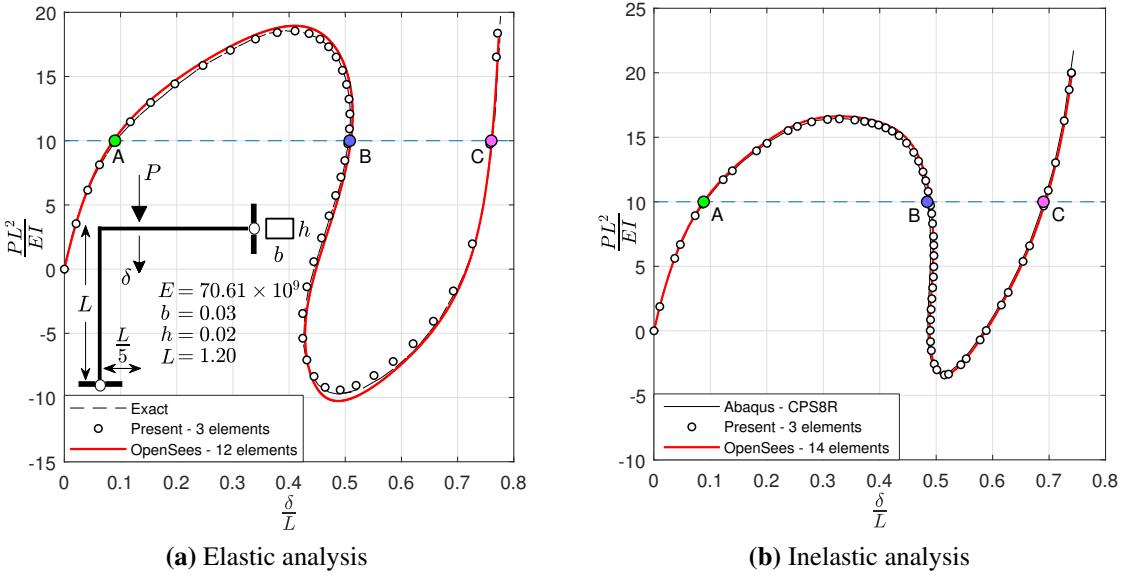


Fig. 2.8. Geometrically nonlinear analyses of Lee's frame with pinned supports.

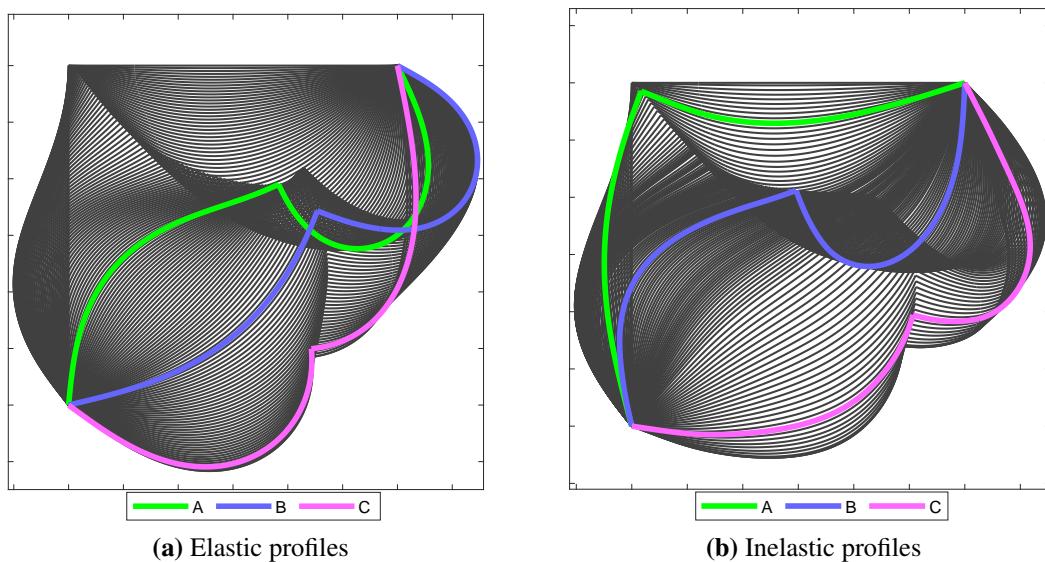


Fig. 2.9. Deformation profiles for Lee's frame with pinned supports.

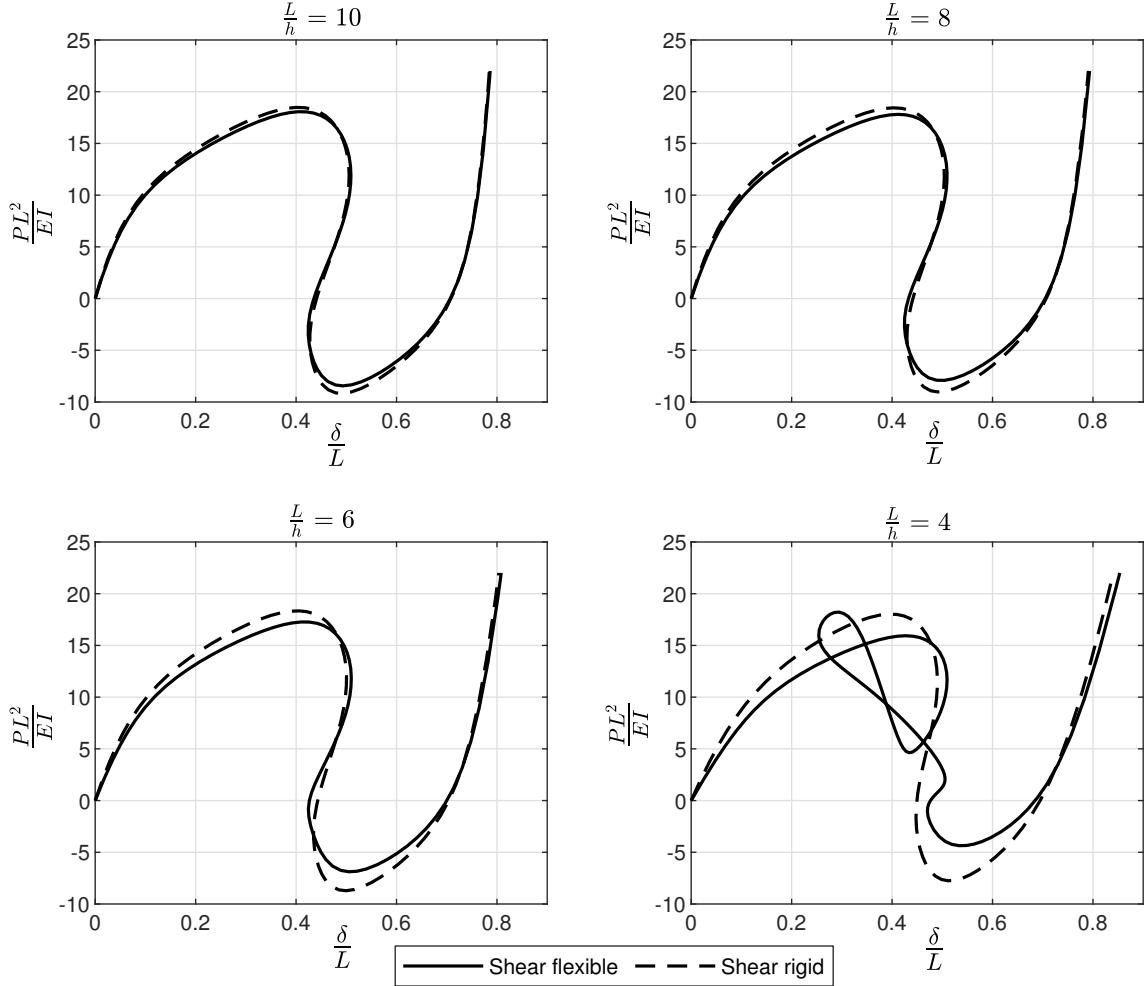


Fig. 2.10. Effect of shear flexibility on Lee's frame response.

In Fig. (2.10), results related to shear flexibility effects on the frame's response are shown. As previously mentioned, in [24] the members are assumed to be shear rigid, effectively adhering to the Bernoulli assumptions. In this example, we relax this assumption and examine how the slenderness decrease affects the structural response. As seen in the figure, four different length-to-thickness ratios are examined and the equilibrium path gradually deviates from the shear-rigid case. For ratio $L/h = 4$, the equilibrium path is drastically different and the structure, after entering the snap-back region, follows a rather complex equilibrium branch. Note that around $\delta/L = 0.5$, the structure exits its current branch and continues on a similar path as in the previous cases.

Frame with fixed supports

This variation of the benchmark problem results in a response that exhibits a dramatic snap-back, with abrupt changes in the stiffness sign along the equilibrium path for the nonlinear elastic problem. Seven integration points are used here per element and, again, one element per member, to achieve the same accuracy as with the Abaqus plane stress model, which terminated prematurely. In contrast, twenty elements are necessary in OpenSees for comparable results. The geometric and material properties are the same as in Sec. (2.5.2) (see Fig. (2.8)). For the inelastic analysis, five integration points are sufficient, since the response is now fairly smooth overall, in comparison to the elastic case. All members are shear flexible in these examples, however the contribution of shear deformation to the overall responses is negligible for both elastic and inelastic cases. The response and deformation profiles for both analyses are shown in Figs. (2.12),(2.11) respectively.

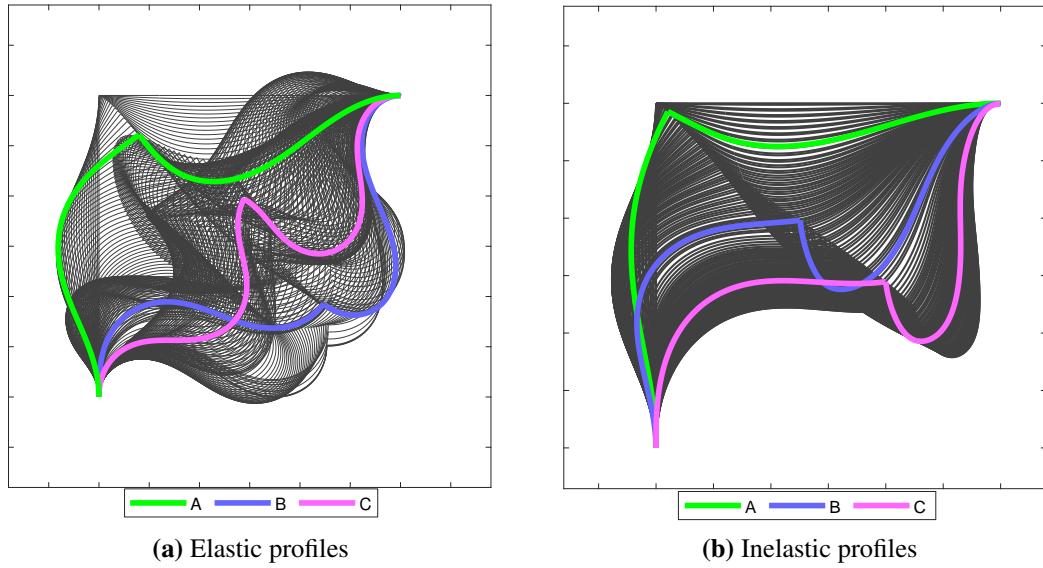


Fig. 2.11. Deformation profiles for Lee's frame with fixed supports.

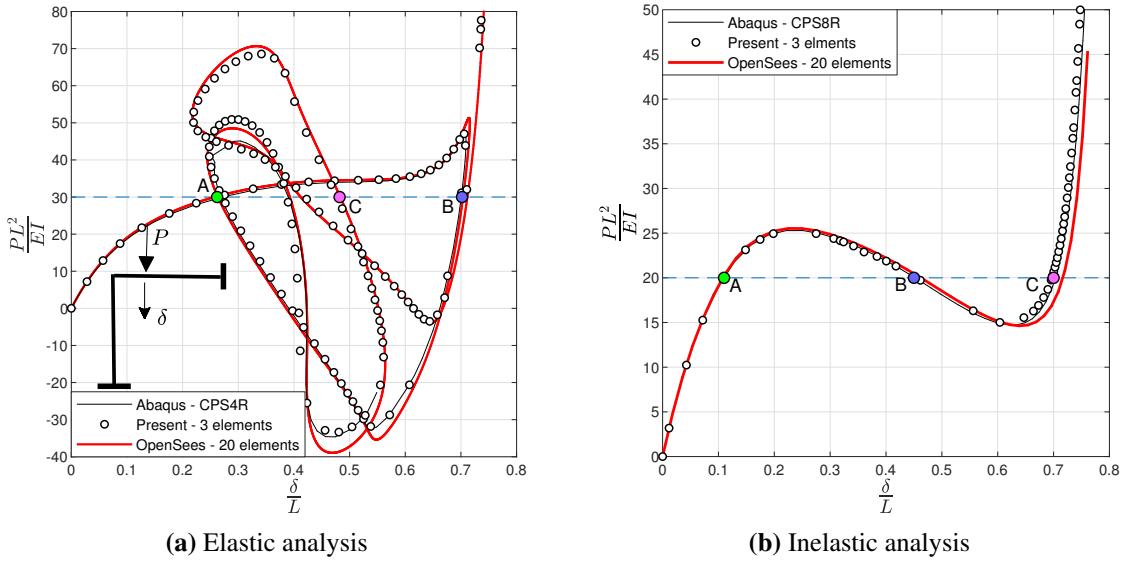


Fig. 2.12. Geometrically nonlinear analyses of Lee's frame with fixed supports.

2.5.3 Two-Storey frame

In this example a two-story frame with one bay is analyzed, first studied regarding its plastic limit load in [25]. It was also later used as a benchmark example by Santos [26], for the performance of a nonlinear beam finite element based on complementary energy principles. The example is slightly modified here, however the rectangular cross-section dimensions of the structural members and their length L have been chosen so their the slenderness remains in accordance with [26]. The yield stress is $\sigma_y = 0.9 \times 10^6$, and the tangent-to-elastic modulus ratio is $r = 3\%$. The full frame characteristics and discretization are shown in Fig. (2.13).

The minimum number of hybrid elements for this example is eight and the number of integration points per element for the elastic and plastic analyses are seven and ten, respectively. For the OpenSees analyses, 36 and 48 elements were used in total for the elastic and inelastic analyses. In both cases, however, the OpenSees analyses stopped prematurely. Results are demonstrated in Fig. (2.14), showing excellent agreement with Abaqus results, while the relevant deformation profiles are provided in Fig. (2.15).

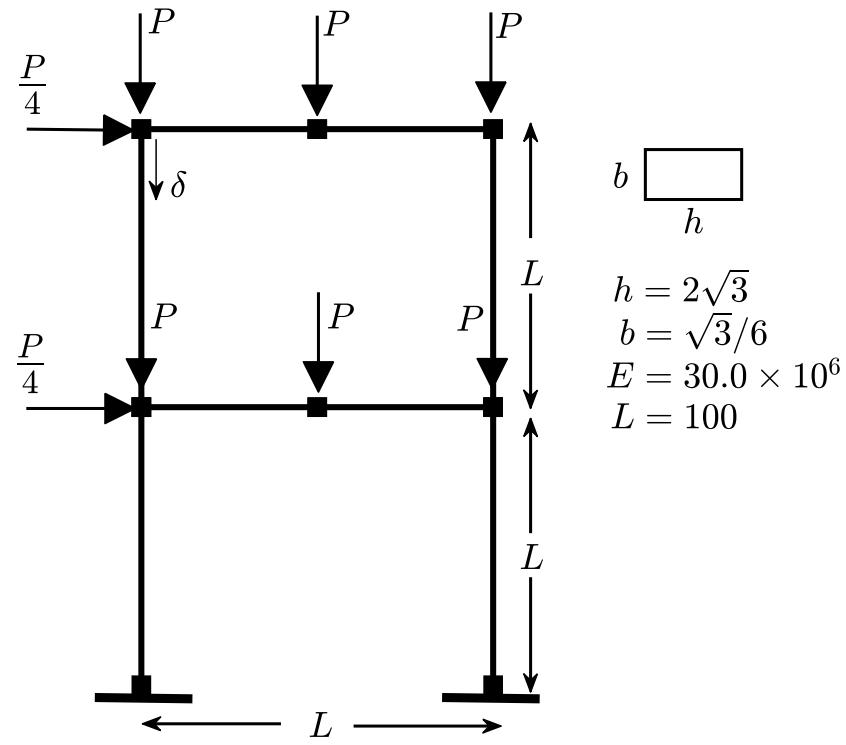


Fig. 2.13. Two-storey portal frame discretization.

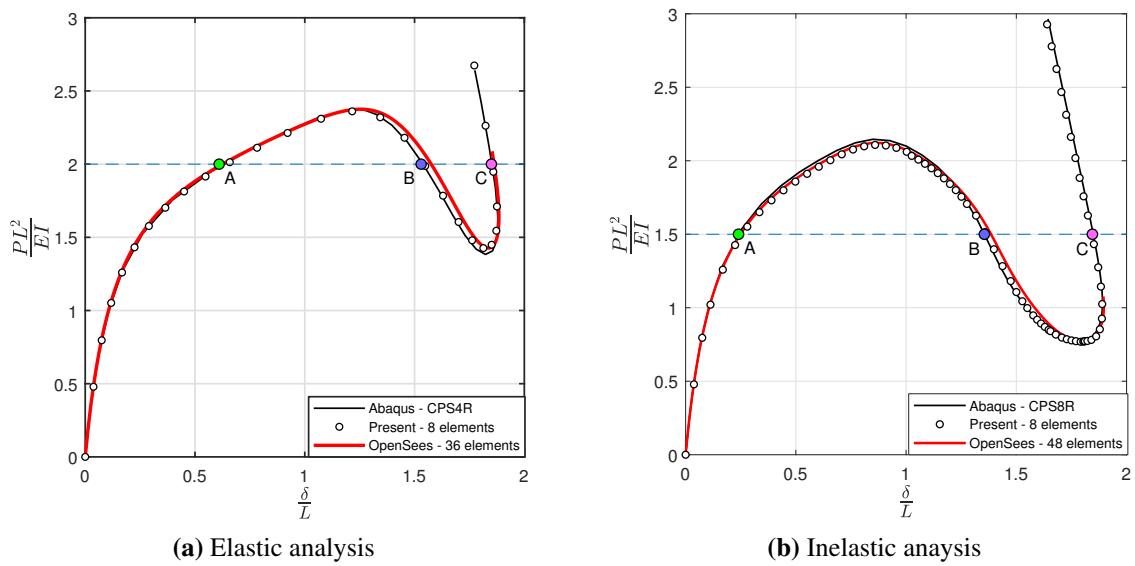


Fig. 2.14. Geometrically nonlinear analyses of the two-storey frame.

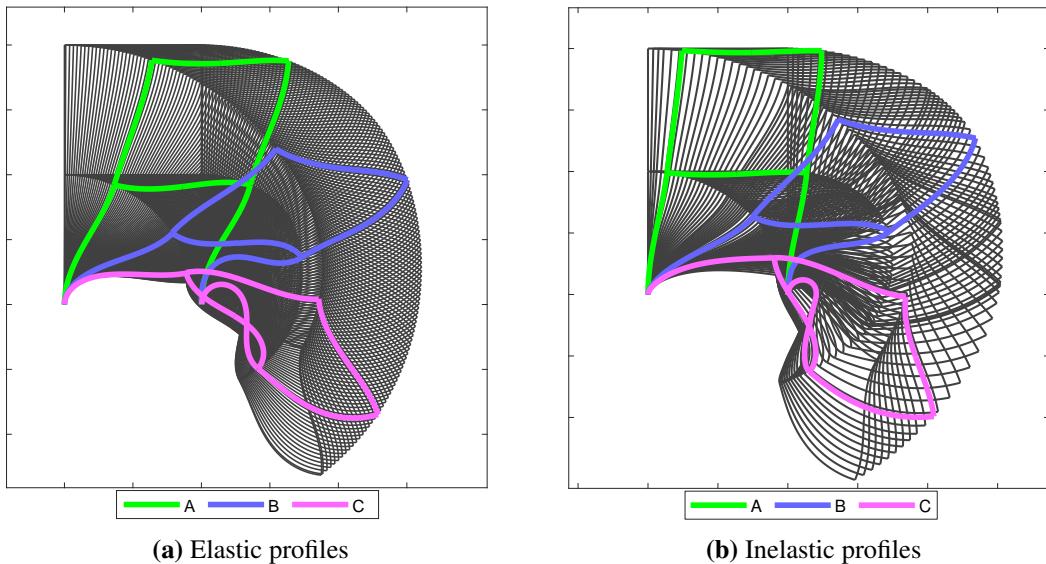


Fig. 2.15. Deformation profiles for two-storey frame.

CHAPTER 3

MATERIAL CONSTITUTIVE MODEL

3.1 Introduction

In this chapter we introduce an algorithm for shear-axial-flexure interaction suited for fibre beam elements of any kind (displacement-based or otherwise). The coupling is taken into account by considering a multiaxial constitutive law at the level of cross-section fibers. Thus, stress resultant interaction as well as the consistent tangent stiffness of a cross-section are derived during the state determination phase. Integration of the elastoplastic rate equations on each fiber is carried out using the fully backward (implicit) Euler method, which as is known, results in the so-called closest point projection algorithm. In addition, the constrained stress state of a fiber is exploited in order to devise a return-mapping scheme that avoids incorporating stress components that are not energetically active. The appropriate mapping on the constrained space is inspirited by the work of Simo[27] and appropriately adapted for fibre discretized elements. This leads to a stress update algorithm that is significantly faster while also requiring less memory requirements for storage of non-active tensor components, compared to formulations that utilize a general three dimensional algorithm[28, 29, 30, 31, 32]. In the latter case, stress update is a nested procedure within an outer Newton iteration which is necessitated in order to enforce the constraint for the transverse stress components, $\sigma_{22} = \sigma_{33} = 0$. This is because during the trial phase of a plastic step, the transverse components become non-zero. Furthermore, a static condensation of the three-dimensional consistent tangent is required to enforce both the local Newton scheme and to derive the consistent tangent modulus pertaining to the fiber stress state. The algorithm introduced in this chapter bypasses the need for additional outer iterations, thereby resulting in a much faster stress update procedure.

For the exposition we adopted the *J2* constitutive model which is based on the von Mises

yield criterion, which is suited for modelling the elastoplastic response of metals. Linear kinematic and general isotropic hardening are incorporated in the formulation. Moreover, infinitesimal strains and a rate-independent associative plasticity framework are the core assumptions that underlie the proposed algorithm. This chapter is subdivided in three main sections: first, we briefly outline the general (rate) form of the three-dimensional elastoplastic equations. In the second section we present the rate constitutive model as it pertains to the fiber stress state and derive expressions for the continuous elastoplastic moduli. In the third section we outline the implementation details, which are based on application of the fully implicit Euler scheme. In addition we derive the expression for the consistent tangent modulus at a fiber, which is required to ensure quadratic rates of convergence for the global Newton method. Finally, we conclude this chapter with a section dedicated to the assessment of the proposed algorithm. Accuracy is demonstrated through iso-error maps for both perfect and hardening plasticity. Furthermore, we compare the performance in terms of total plastic step iterations required for i) the present formulation, ii) the general, three-dimensional and iii) the plane stress return mapping algorithm developed in [27]. Numerical examples pertaining to ultimate collapse load, cyclic loading, elastoplastic buckling and a pushover analysis of a steel frame are also included to highlight the capabilities of the Hybrid NLP element equipped with the multiaxial stress-update algorithm proposed herein.

3.2 Three-dimensional Constitutive Model

Below we outline the general rate form of rate-independent elastoplasticity with combined isotropic and linear kinematic hardening. Infinitesimal strains and associative plasticity are assumed, while the yield criterion is purposefully unspecified at this stage in order to maintain generality.

Let σ , ϵ be the second order symmetric stress and strain tensors. We introduce the internal hardening variables α and q which represent the second order back-stress tensor and the equivalent uniaxial yield stress respectively. The former is associated with kinematic hardening

while the latter models isotropic hardening. Finally, consider the decomposition of the strain tensor into elastic and plastic parts as follows:

$$\boldsymbol{\epsilon} = \boldsymbol{\epsilon}^{el} + \boldsymbol{\epsilon}^{pl} \quad (3.1)$$

Then the rate form of the constitutive equations for the assumptions mentioned above is the following:

$$\dot{\boldsymbol{\epsilon}} = \dot{\boldsymbol{\epsilon}}^{el} + \dot{\boldsymbol{\epsilon}}^{pl} \quad (\text{Strain decomposition}) \quad (3.2a)$$

$$\dot{\boldsymbol{\sigma}} = \mathbb{C}^{el} [\dot{\boldsymbol{\epsilon}} - \dot{\boldsymbol{\epsilon}}^{pl}] \quad (\text{Elastic constitutive law}) \quad (3.2b)$$

$$\dot{\boldsymbol{\epsilon}}^{pl} = \dot{\lambda} \frac{\partial \Phi}{\partial \boldsymbol{\sigma}} \quad (\text{Flow rule}) \quad (3.2c)$$

$$\dot{\boldsymbol{\alpha}} = -\dot{\lambda} H_{kin} \frac{\partial \Phi}{\partial \boldsymbol{\alpha}} \quad (\text{Kinematic hardening law}) \quad (3.2d)$$

$$\dot{q} = \frac{\partial q}{\partial e^{pl}} \dot{e}^{pl} \quad (\text{Isotropic hardening law}) \quad (3.2e)$$

$$\dot{e}^{pl} = \dot{\lambda} \quad (\text{Equivalent plastic strain}) \quad (3.2f)$$

$$\Phi(\boldsymbol{\sigma}, \boldsymbol{\alpha}, e^{pl}) = \sigma_{eq}(\boldsymbol{\sigma}, \boldsymbol{\alpha}) - q(e^{pl}) \leq 0 \quad (\text{Yield criterion}) \quad (3.2g)$$

In the above system of differential algebraic equations, λ is the plastic parameter, H_{kin} is the kinematic hardening modulus, \mathbb{C}^{el} is the elastic moduli, a fourth order tensor, of the material and σ_{eq} is the equivalent stress, which is determined by the yield criterion used. Furthermore, in the case of linear isotropic hardening with modulus H_{iso} , the corresponding harening law (Eq. 3.2e) becomes $\dot{q} = H_{iso} \dot{e}^{pl}$, where e^{pl} is the equivalent plastic strain. Finally, with the inequality in Eq. (3.2g) we define a feasible stress space. Stress points stictly satisfying the inequality, $\Phi < 0$ represent an elastic state while those that cause plastic flow render Φ zero.

The so-called Karush-Kuhn-Tucker loading/unloading conditions derived from the underly-

ing constrained variational problem[33] can be stated as follows:

$$\Phi \leq 0, \quad \dot{\lambda} \geq 0, \quad \dot{\lambda}\Phi = 0 \quad (3.3)$$

From these conditions we can determine the current stress state:

- if $\dot{\lambda} > 0$, then $\dot{\lambda}\Phi = 0$ implies $\Phi = 0$ which means the state is plastic
- if $\Phi < 0$, then $\dot{\lambda}\Phi = 0$ implies $\dot{\lambda} = 0$ and from (3.2c) we get that the state is elastic.

Since $\Phi = 0$, $\dot{\lambda} > 0$ when plastic loading persists, the third equation in (3.3) leads to the so-called plastic consistency condition:

$$\dot{\lambda}\dot{\Phi} = 0 \quad (3.4)$$

3.3 The J_2 formulation for fiber stress state

3.3.1 Mapping on the constrained stress space

We begin by defining the admissible stress space for a planar beam fiber, \mathcal{S} :

$$\mathcal{S} = \{\vec{\sigma} \in \mathbb{R}^6 \mid \sigma_{22} = \sigma_{33} = \sigma_{23} = \sigma_{13} = 0\}$$

The von Mises yield criterion is stated as follows:

$$\Phi(J_2, e^{pl}) = \sqrt{3J_2} - q(e^{pl}) \leq 0 \quad (3.5)$$

where J_2 is the second invariant of the deviatoric stress tensor:

$$J_2 = \frac{1}{2}\boldsymbol{\sigma}^d : \boldsymbol{\sigma}^d \quad (3.6)$$

The deviatoric part of any tensor is given by $()^d = () - \frac{1}{3}\mathbf{I} \cdot \text{trace}[()]$, with \mathbf{I} being the identity

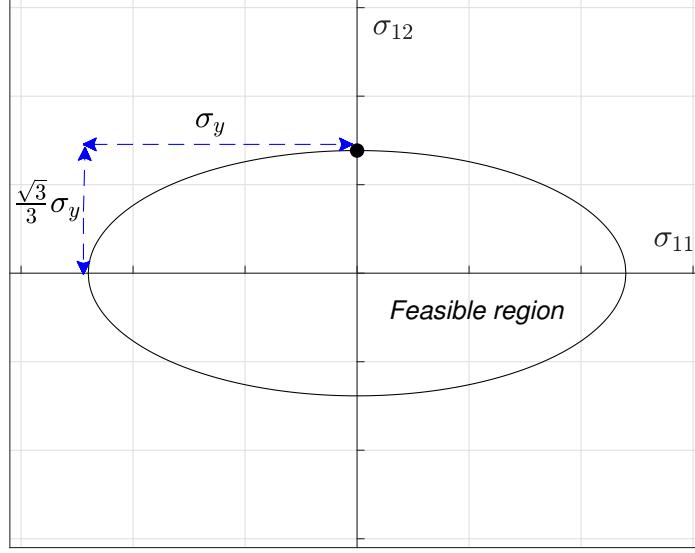


Fig. 3.1. Feasible space and yield locus for the J_2 model. At the boundary of the ellipse, $\sqrt{3J_2} = q$.

tensor and symbol $\mathfrak{z} : \zeta$ designates the contraction operator: $\boldsymbol{\sigma}^d : \boldsymbol{\sigma}^d = \sum_{i,j} \sigma_{ij}^d \sigma_{ij}^d$.

In the von Mises J_2 framework, the hydrostatic pressure cannot cause plastic flow. In other words, plastic deformation is due to the deviatoric components of the stress tensor and volume changes are caused only by elastic deformations. This implies that $\text{trace}[\boldsymbol{\epsilon}^{pl}] = 0$ which, coupled with Eq. (3.2d), also leads to $\text{trace}[\boldsymbol{\alpha}^{pl}] = 0$. This means that the back stress tensor in Eq. (3.2d) deviatoric.

In active stress space \mathcal{S} the yield function is $\Phi = \sqrt{\sigma_{11}^2 + 3\sigma_{12}^2} - q$ describes an ellipse, shown in 3.1, with semi-major and semi-minor axes σ_y , $\sqrt{3}/3\sigma_y$, respectively.

When kinematic hardening is included, we introduce the effective stress tensor and its deviatoric part in order to facilitate notational simplicity:

$$\zeta = \boldsymbol{\sigma} - \boldsymbol{\alpha}, \quad \zeta^d = \boldsymbol{\sigma}^d - \boldsymbol{\alpha}^d \quad (3.7)$$

We now define the following vectors associated with the active tensor components that ex-

pliely enter subsequent derivations:

$$\vec{\alpha}_f = \begin{bmatrix} \alpha_{11} \\ \alpha_{12} \end{bmatrix}, \quad \vec{\alpha}_f^d = \begin{bmatrix} \alpha_{11}^d \\ \alpha_{12}^d \end{bmatrix}, \quad \vec{\epsilon}_f^{pl} = \begin{bmatrix} \epsilon_{11}^{pl} \\ \gamma_{12}^{pl} \end{bmatrix},$$

$$\vec{\sigma}_f^d = \begin{bmatrix} \sigma_{11}^d \\ \sigma_{12}^d \end{bmatrix}, \quad \vec{\zeta}_f = \begin{bmatrix} \zeta_{11} \\ \zeta_{12} \end{bmatrix}, \quad \vec{\zeta}_f^d = \begin{bmatrix} \zeta_{11}^d \\ \zeta_{12}^d \end{bmatrix}$$

The vector representation of (3.6) in terms of the effective stress is:

$$J_2 = \frac{1}{2}(\vec{\zeta}^d)^T \mathbf{J} \vec{\zeta}^d \quad (3.8)$$

where matrix $\mathbf{J} = \text{diag}[1, 1, 1, 2, 2, 2]$ is introduced to account for the symmetry of ζ^d .

We now seek a mapping from the deviatoric stress space onto the fiber stress space S of the deviatoric effective stress vector $\vec{\zeta}^d$ that preserves the inner product and, thus, the second deviatoric invariant J_2 . The non-zero components of $\vec{\zeta}^d$ are $\zeta_{11}^d, \zeta_{22}^d, \zeta_{33}^d, \zeta_{12}^d, \zeta_{21}^d$ and thus we can establish the following mappings between the active components of $\vec{\zeta}^d$ and $\vec{\zeta}_f$:

$$\vec{\zeta}^d = \mathbf{Q} \vec{\zeta}_f, \quad \vec{\zeta}_f^d = \mathbf{S} \vec{\zeta}_f \quad (3.9)$$

where the mapping matrices above are given by:

$$\mathbf{Q} = \begin{bmatrix} \frac{2}{3} & -\frac{1}{3} & -\frac{1}{3} & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix}^T, \quad \mathbf{S} = \begin{bmatrix} \frac{2}{3} & 0 \\ 0 & 1 \end{bmatrix} \quad (3.10)$$

We can now express J_2 purely in terms of $\vec{\zeta}_f$:

$$J_2 = \frac{1}{2}(\vec{\zeta}^d)^T \mathbf{J} \vec{\zeta}^d = \frac{1}{2} \vec{\zeta}_f^T \mathbf{Q}^T \mathbf{J} \mathbf{Q} \vec{\zeta}_f = \frac{1}{2} \vec{\zeta}_f^T \mathbf{V} \vec{\zeta}_f = \frac{1}{2} \|\vec{\zeta}_f\|_{\mathbf{V}}^2 \quad (3.11)$$

where $\mathbf{V} = \text{diag}[\frac{2}{3}, 2]$ and the notation used for the inner product induced by matrix \mathbf{A} between

vectors \vec{w} , \vec{v} is $(\vec{w}, \vec{v})_A = \vec{w}^T \mathbf{A} \vec{v}$, which leads to the notation adopted in Eq. (3.11).

With these derivations at hand, system (3.2) can be recast in the constrained fiber stress state as follows:

$$\dot{\vec{\epsilon}}_f = \dot{\vec{\epsilon}}_f^{el} + \dot{\vec{\epsilon}}_f^{pl} \quad (\text{Strain decomposition}) \quad (3.12a)$$

$$\dot{\vec{\sigma}}_f = \mathbf{C}^{el} \left[\dot{\vec{\epsilon}}_f - \dot{\vec{\epsilon}}_f^{pl} \right] \quad (\text{Elastic constitutive law}) \quad (3.12b)$$

$$\dot{\vec{\epsilon}}_f^{pl} = \dot{\lambda} \vec{n} \quad (\text{Flow rule}) \quad (3.12c)$$

$$\dot{\vec{\alpha}}_f = \dot{\lambda} H_{kin} \mathbf{V}^{-1} \vec{r} \quad (\text{Kinematic hardening law}) \quad (3.12d)$$

$$\dot{q}_f = \dot{\lambda} \frac{\partial q_f}{\partial e_f^{pl}} \quad (\text{Isotropic hardening law}) \quad (3.12e)$$

$$\Phi(\vec{\zeta}_f, e^{pl}) = \sqrt{\frac{3}{2}} \|\vec{\zeta}_f\|_{\mathbf{V}} - q_f(e_f^{pl}) \leq 0 \quad (\text{Yield criterion}) \quad (3.12f)$$

Vector \vec{n} is expressed as follows:

$$\vec{n} = \frac{\partial \Phi}{\partial \sigma_f} = -\frac{\partial \Phi}{\partial \alpha_f} = \sqrt{\frac{3}{2}} \frac{\mathbf{V} \vec{\zeta}_f}{\|\vec{\zeta}_f\|_{\mathbf{V}}} \quad (3.13)$$

Equations (3.2e), (3.2f) remain the same in the current system. In addition, note that for the J_2 model, the left-hand side of eq. (3.2d) is a deviatoric tensor whereas in (3.12d) it has been mapped to fiber stress space \mathcal{S} .

3.3.2 Continuous tangent modulus

The continuous elastoplastic modulus \mathbf{C}^{ep} is found by enforcing the plastic consistency condition (3.4) during a plastic step:

$$\vec{n}^T \dot{\vec{\zeta}}_f - \dot{\lambda} \frac{\partial q_f}{\partial e_f^{pl}} = 0 \quad (3.14)$$

The rate form for the effective stress is given by combining Eqs. (3.12b), (3.12d) with $\dot{\zeta}_f = \dot{\vec{\sigma}}_f - \dot{\vec{\alpha}}_f$:

$$\dot{\zeta}_f = \mathbf{C}^{el}\dot{\epsilon}_f - \dot{\lambda}\mathbf{Z}\vec{n} \quad (3.15)$$

where $\mathbf{Z} = \mathbf{C}^{el} + H_{kin}\mathbf{V}^{-1}$. Substituting (3.15) into (3.14) and solving for $\dot{\lambda}$ we get:

$$\dot{\lambda} = \frac{\vec{n}^T \mathbf{C}^{el} \dot{\vec{\epsilon}}_f}{\vec{n}^T \mathbf{Z} \vec{n} + \frac{\partial q_f}{\partial e_f^{pl}}} \quad (3.16)$$

Finally, by making use of (3.16), (3.12c) and (3.12b), we arrive at the expression for the continuous elastoplastic modulus:

$$\mathbf{C}^{pl} = \mathbf{C}^{el} - \frac{\vec{m}\vec{m}^T}{\vec{n}^T \mathbf{Z} \vec{n} + \frac{\partial q_f}{\partial e_f^{pl}}} \quad (3.17)$$

where $\vec{m} = \mathbf{C}^{el}\vec{n}$.

3.4 Implicit integration of rate equations

We employ the fully implicit backward Euler method to discretize the rate constitutive equations in system (3.12) in the pseudotime interval $[t_n, t_{n+1}]$. We regard the incremental/iterative process as strain driven, in that at the start of increment n , the dependent state variables for each fiber, $S_n = \{\vec{\epsilon}_f, \vec{\epsilon}_f^{pl}, \vec{\alpha}_f, e_f^{pl}\}_n$, are known and stored. We also know the increment in the centerline strain vector \vec{d} , $\Delta\vec{d}$. The incremental strain vector for the current fiber is then determined using eq. (2.53b):

$$\Delta\vec{\epsilon}_f = \mathbf{N}_s \Delta\vec{q}$$

Given the incremental strain vector $\Delta\vec{\epsilon}_f$, we need to update the state variables for t_{n+1} : $S_n \rightarrow S_{n+1}$. In what follows we omit the subscript denoting the vector pertains to the fiber state:

$$S_n \equiv \{\vec{\epsilon}_n, \vec{\epsilon}_n^{pl}, \vec{\alpha}_n, e_n^{pl}\}, \quad S_{n+1} \equiv \{\vec{\epsilon}_{n+1}, \vec{\epsilon}_{n+1}^{pl}, \vec{\alpha}_{n+1}, e_{n+1}^{pl}\},$$

The update of the fiber total strain vector $\vec{\epsilon}$ is trivial: $\vec{\epsilon}_{n+1} = \vec{\epsilon}_n + \Delta\vec{\epsilon}$. For the remaining

state variables the rate system (3.12) is discretized as follows:

$$\vec{\sigma}_{n+1} = \mathbf{C}^{el} [\vec{\epsilon}_{n+1} - \vec{\epsilon}_{n+1}^{pl}] \quad (3.18a)$$

$$\vec{\epsilon}_{n+1}^{pl} = \vec{\epsilon}_n^{pl} + \dot{\lambda} \vec{n}_{n+1} \quad (3.18b)$$

$$\vec{\alpha}_{n+1} = \vec{\alpha}_n + \lambda H_{kin} \mathbf{V}^{-1} \vec{n}_{n+1} \quad (3.18c)$$

$$q_{n+1} = q_n + \lambda \frac{\partial q_{n+1}}{\partial e_{n+1}^{pl}} \quad (3.18d)$$

$$\Phi(\vec{\zeta}_{n+1}, e_{n+1}^{pl}) = \sqrt{\frac{3}{2}} \|\vec{\zeta}_{n+1}\|_V - q_{n+1}(e_{n+1}^{pl}) \leq 0 \quad (3.18e)$$

and

$$\vec{\zeta}_{n+1} = \vec{\sigma}_{n+1} - \vec{\alpha}_{n+1} \quad (3.19)$$

3.4.1 Elastic predictor - Plastic Corrector algorithm

It is convenient to recast system (3.18) purely in terms of stress by utilizing (3.19),(3.18a),(3.18c):

Stress rate system	Discretized system
$\dot{\vec{\zeta}} = \mathbf{C}^{el} \dot{\vec{\epsilon}} - \dot{\lambda} \mathbf{Z} \vec{n}$ (3.20a)	$\vec{\zeta}_{n+1} = \vec{\zeta}_n + \mathbf{C}^{el} \Delta \vec{\epsilon}_{n+1} - \lambda \mathbf{Z} \vec{n}_{n+1}$ (3.21a)
$\dot{q} = \dot{\lambda} \frac{\partial q}{\partial e^{pl}}$ (3.20b)	$q_{n+1} = q_n + \lambda \frac{\partial q_{n+1}}{\partial e_{n+1}^{pl}}$ (3.21b)

We need not differentiate between initial and final value for the plastic parameter λ since it is initialized to zero at the beginning of every step: $\lambda_n = 0$. The solution of (3.21) is based on a two-step procedure. The first phase is called trial elastic step, where we assume that the increment $\Delta \vec{\epsilon}$ is elastic and all state variables are updated by setting $\lambda = 0$: $S_n \rightarrow S_{n+1}^{TR}$. If the trial stresses satisfy the yield condition $\Phi_{n+1}^{TR} < 0$, then the step was indeed elastic and we can set $S_{n+1}^{TR} \rightarrow S_{n+1}$. If the yield criterion is violated, $\Phi_{n+1}^{TR} > 0$, then the step is plastic and we need to apply (plastic) corrections to the state variables so that, upon convergence, we have achieved $\Phi_{n+1} = 0$. This procedure is referred to as Elastic Predictor - Plastic Corrector algorithm and can be traced back to the work of Wilkins[34]. It has enjoyed widespread application in solving

problems in elastoplasticity[35, 36, 37, 38, 39, 40] and the basic relations for each step are summarized below.

Elastic Predictor	Plastic Corrector
$\vec{\zeta}_{n+1}^{TR} = \mathbf{C}^{el} [\vec{\epsilon}_{n+1} - \vec{\epsilon}_n^{pl}] - \vec{\alpha}_n$	$\vec{\zeta}_{n+1} = \vec{\zeta}_{n+1}^{TR} - \lambda \mathbf{Z} \vec{n}_{n+1}$ (3.23a)
$q_{n+1}^{TR} = q_n$	$\xrightarrow{\Phi_{n+1}^{TR} > 0} q_{n+1} = q_{n+1}^{TR} + \lambda \frac{\partial q_{n+1}}{\partial \epsilon_{n+1}^{pl}}$ (3.23b)
$\Phi_{n+1}^{TR} = f(\vec{\zeta}_{n+1}^{TR}, q_{n+1}^{TR})$	$\Phi_{n+1} = 0$ (3.23c)

The Plastic Corrector step is a system of 4 nonlinear algebraic equations with 4 unknowns, $\{\zeta_{11,n+1}, \zeta_{12,n+1}, q_{n+1}, \lambda\}$, and generally a local Newton method is required to determined the unknowns. It should be stressed here that, in multiaxial constitutive algorithms that do not make use of the mapping as developed here, this local process is nested within an upper level Newton procedure that enforces the zero transverse stress condition.

Substituting Eqs. (3.23a),(3.23b) into (3.23c) furnishes a nonlinear scalar equation that needs to be solved for λ . Once λ is computed, then the fiber state can be updated $S_{n+1}^{TR} \rightarrow S_{n+1}$. Refer also to Box 1 for a detailed summary of the plastic updates. Detailed expressions for the local Newton method on $\Phi_{n+1}(\lambda) = 0$ are provided in Appendix C.

3.4.2 Consistent Tangent Modulus

The tangent modulus consistent with the numerical discretization used on system (3.12) is necessary to restore the quadratic rates of convergence for the global Newton method[41]. That is:

$$\mathbf{C}_c^{pl} = \frac{d\vec{\sigma}_{n+1}}{d\vec{\epsilon}_{n+1}} \quad (3.24)$$

We take the differentials of (3.18a), (3.18c) and (3.18e), using the fact that $d\vec{\zeta}_{n+1} = d\vec{\sigma}_{n+1} - d\vec{\alpha}_{n+1}$ from (3.19) and arrive at the following system:

$$\begin{bmatrix} \mathbf{C}^{el} + \lambda \Psi_{n+1} & -\lambda \Psi_{n+1} \\ -\lambda \Psi_{n+1} & H_{kin}^{-1} \mathbf{V} + \lambda \Psi_{n+1} \end{bmatrix} \begin{bmatrix} d\vec{\sigma}_{n+1} \\ d\vec{\alpha}_{n+1} \end{bmatrix} = \begin{bmatrix} d\vec{\epsilon}_{n+1} - d\lambda \vec{n}_{n+1} \\ d\lambda \vec{n}_{n+1} \end{bmatrix} \quad (3.25a)$$

$$\vec{n}_{n+1}^T [d\vec{\sigma}_{n+1} - d\vec{\alpha}_{n+1}] = d\lambda \frac{\partial q_{n+1}}{\partial e_{n+1}^{pl}} \quad (3.25b)$$

where Ψ is given as follows:

$$\Psi = \frac{\partial \vec{n}}{\partial \vec{\zeta}} = \frac{\partial^2 f}{\partial \vec{\zeta}^2} = \frac{1}{\|\vec{\zeta}\|_V} [\mathbf{V} - \vec{n} \vec{n}^T] \quad (3.26)$$

Upon solving (3.25a) for $d\vec{\sigma}_{n+1}$, $d\vec{\alpha}_{n+1}$ we substitute them into (3.25b) and solve for $d\lambda$:

$$d\lambda = \frac{\vec{n}_{n+1}^T [\Xi_{11} + \Xi_{12}^T] d\vec{\epsilon}_{n+1}}{\vec{n}_{n+1}^T \mathbf{T} \vec{n}_{n+1} + \frac{\partial q_{n+1}}{\partial e_{n+1}^{pl}}} \quad (3.27)$$

Finally, substituting $d\lambda$ into (3.18a) furnishes the formula for the consistent tangent modulus:

$$\mathbf{C}_c^{pl} = \Xi_{11} - \frac{\vec{N}_{n+1} \vec{N}_{n+1}^T}{1 + \delta_{n+1}} \quad (3.28)$$

Details regarding all newly defined vectors and matrices involved in the derivation of \mathbf{C}_c^{pl} is given in Appendix D. In Box 1 we have summarised the return mapping algorithm for a fiber.

3.4.3 Consistent Section Stiffness

Once the stress update on a particular fiber has been carried out, its contribution to the stress resultant vector and the section stiffness are accounted for via a midpoint integration rule along the section height (see Eqs. (2.56), (2.58)).

Box 1 : Summary of Elastic Predictor - Plastic Corrector algorithm.

- **Start of step - Elastic Predictor**

<ul style="list-style-type: none"> ■ $\vec{\epsilon}_{n+1}^{pl,TR} = \vec{\epsilon}_n^{pl}$ ■ $\vec{\sigma}_{n+1}^{TR} = \mathbf{C}^{el} [\vec{\epsilon}_{n+1} - \vec{\epsilon}_n^{pl}]$ ■ $\vec{\alpha}_{n+1}^{TR} = \vec{\alpha}_n$ 	<ul style="list-style-type: none"> ■ $e_{n+1}^{pl,TR} = e_n^{pl}$ ■ $q_{n+1}^{TR} = q_n$ ■ $f_{n+1}^{TR} = \Phi(\vec{\zeta}_{n+1}^{TR}, q_{n+1}^{TR})$
---	--

- IF $\Phi_{n+1}^{TR} < 0 \rightarrow$ **Elastic Step**

<ul style="list-style-type: none"> ■ $\vec{\epsilon}_{n+1}^{pl} = \vec{\epsilon}_{n+1}^{pl,TR}$ ■ $\vec{\sigma}_{n+1} = \vec{\sigma}_{n+1}^{TR}$ ■ $\vec{\alpha}_{n+1} = \vec{\alpha}_{n+1}^{TR}$ 	<ul style="list-style-type: none"> ■ $e_{n+1}^{pl} = e_{n+1}^{pl,TR}$ ■ $q_{n+1} = q_{n+1}^{TR}$ ■ $\mathbf{C} = \mathbf{C}^{el}$
---	---

- ELSE $f_{n+1}^{TR} > 0 \rightarrow$ **Plastic Corrector**

1. Iteratively solve (3.4.1) for λ : $\lambda^{j+1} = \lambda^j - \frac{\Phi_{n+1}^j}{(\frac{d\Phi_{n+1}}{d\lambda})^j}$, with $\lambda^0 = 0$, $\Phi_{n+1}^0 = \Phi_{n+1}^{TR}$. Loop termination: $|\Phi_{n+1}^{j+1}| \leq e_{tol}$, where e_{tol} small.

2. Update state variables:

<ul style="list-style-type: none"> ■ $\vec{\zeta}_{n+1} = \Omega(\lambda) \vec{\zeta}_{n+1}^{TR}$ ■ $\vec{\epsilon}_{n+1}^{pl} = \vec{\epsilon}_n^{pl} + \lambda \vec{n}_{n+1}$ ■ $\vec{\alpha}_{n+1} = \vec{\alpha}_n + \lambda H_{kin} \mathbf{V}^{-1} \vec{n}_{n+1}$ 	<ul style="list-style-type: none"> ■ $\vec{\sigma}_{n+1} = \vec{\zeta}_{n+1} + \vec{\alpha}_{n+1}$ ■ $e_{n+1}^{pl} = e_n^{pl} + \lambda$ ■ $q_{n+1} = q_n + \lambda \frac{\partial q_{n+1}}{\partial e_{n+1}^{pl}}$
---	---

3. Get consistent tangent modulus

$$\mathbf{C} = \mathbf{C}_c^{pl} = \Xi_{11} - \frac{\vec{N}_{n+1} \vec{N}_{n+1}^T}{1 + \delta_{n+1}}$$

See Appendix C for $\Omega(\lambda)$.

Notice that while in the present formulation $C_{21} = C_{12}$, this will not be the case if a generalized integration rule is used for the rate equations. In such a scenario, if enforcement of the plastic consistency does not correspond to the interior time $t_a \in [t_n, t_{n+1}]$ chosen, that is $\Phi_a = 0$, then the consistent tangent will not be symmetric. For more details on this refer to [42].

3.5 Numerical Examples

3.5.1 Accuracy of the proposed algorithm

Iso-error maps are commonly used to test the accuracy of return-mapping algorithms in elasto-plasticity[43]. The contours are generated by imposing a family of stress or strain prescribed loadings from a selected point on the yield surface which cause further plastification (no unloading). For each load incrementation, the error is computed as the relative difference between the stress predicted by the proposed algorithm and “exact” stress. The exact stress is typically found by imposing the same increment in a number of sequential subincrements, n_s . A convergence analysis is required to ensure that subincrementation indeed converges to a stress point. In the present study it was found that for $n_s \geq 500$ the stress components had achieved convergence with respect to the first 7 decimal points. Therefore, we use $n_s = 500$. The error plotted is given by:

$$err = \frac{\|\vec{\sigma}^* - \vec{\sigma}\|}{\|\vec{\sigma}^*\|} \cdot 100 \quad (3.29)$$

where $\vec{\sigma}^*$ is the exact stress vector and $\vec{\sigma}$ the one predicted by the proposed scheme by solving in one step. The three points on the yield surface that are selected and shown in Fig. 3.2 correspond to an initial state of pure shear (A), mixed state where $\sigma_{11,0} = \sigma_{12,0}$ (B) and a pure uniaxial stress state (C). For each of these initial states the family of strain increments are imposed along the directions shown, with range up to $\Delta\|\vec{\epsilon}\| = 5\epsilon_y$, where ϵ_y is the yield strain in tension/compression. The material data used are: $E = 2.1 \times 10^5$, $\sigma_y = 2.4 \times 10^2$ and Poisson’s ratio $\nu = 0.3$. For the hardening cases, linear isotropic and kinematic moduli were prescribed as follows: $H_{iso} = H_{kin} = 10^4$.

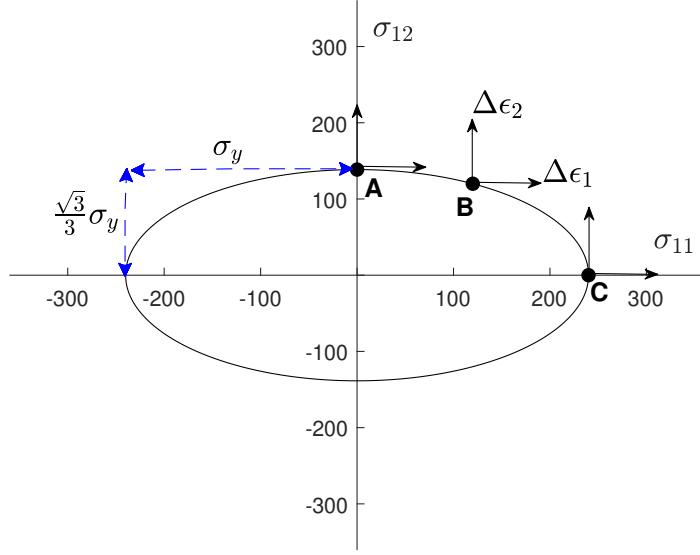


Fig. 3.2. Points of interest on yield surface and direction of load increments.

The parameters associated with the horizontal and vertical axes, a_1 , a_2 , are defined in the following way: $a_1 = \Delta\vec{\epsilon}_1/\epsilon_y$, $a_2 = \Delta\vec{\epsilon}_2/\epsilon_y$. As can be seen from figures 3.3, 3.4, the relative error remains small even for quite large increments in the strain vector. Note also that for along the semi-major and semi-minor axes in the cases of pure uniaxial state and pure shear respectively, the return mapping gives the exact stress even without subincrementation. The same is true for point B when the strain increments are along a direction around 68° with respect to a_1 axis.

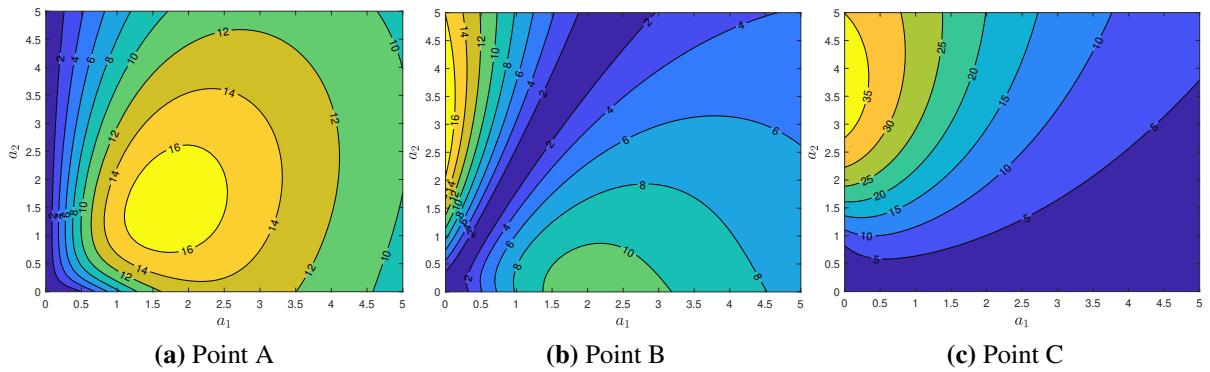


Fig. 3.3. Iso-error maps for a perfectly plastic material.

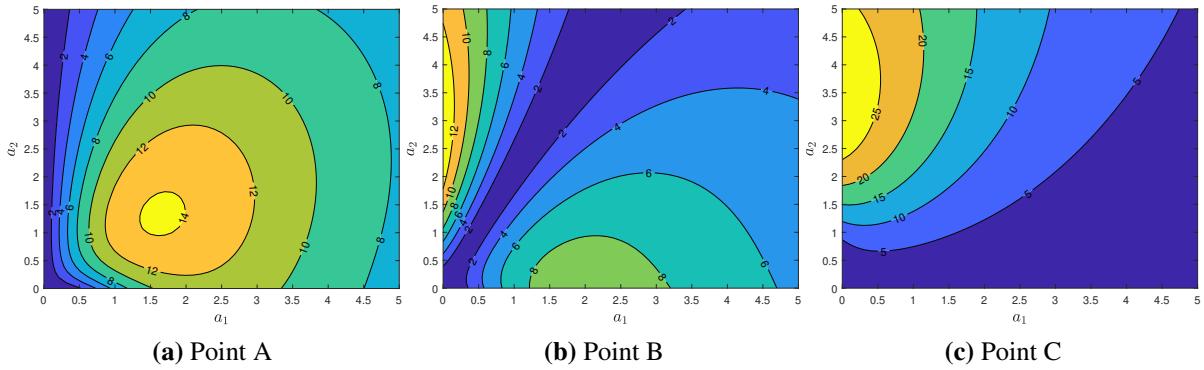


Fig. 3.4. Iso-error maps for combined linear isotropic and kinematic hardening.

3.5.2 Performance against non-monotonic strain histories

In this benchmark we investigate the performance of the proposed algorithm against the three dimensional/axisymmetric procedure typically implemented in beam elements when shear-flexure interaction is taken into account using a multiaxial law at the fiber level. We also compare it with the plane stress projection algorithm by Simo[41], which we followed in our derivations. In the former approach, the return mapping using an implicit integration results in the so-called radial return, which doesn't require iterations in the case of J_2 with perfect plasticity or linear hardening[34]. In contrast, the latter requires iterations since the topology of the stress space is not as simple as the von Mises cylinder anymore. We show below the drastic performance boost when using the present algorithm. As already state earlier, this is mainly due to the fact that we avoid an additional Newton procedure to enforce the zero transverse stress condition. The results for each scenario are shown in the tables below. A perfectly plastic material was assumed and both the average execution time $t(\text{sec})$ and the total number of iterations, N_i are considered as performance metrics. The *additional* iterations during purely elastic steps are also included in parentheses for the 3D and Simo's plane stress algorithm. Moreover, for each scenario, four different incrementations of the strain history are employed, shown in Fig. 3.5, and the effect of this scaling on the performance is highlighted. Parameter a determines the increment "length": $\alpha = \|\Delta\vec{\epsilon}\|/\epsilon_y$. Material data are the same as in the previous example.

As can be seen from the Tables 3.1-3.4, the proposed scheme offers significant speed-up

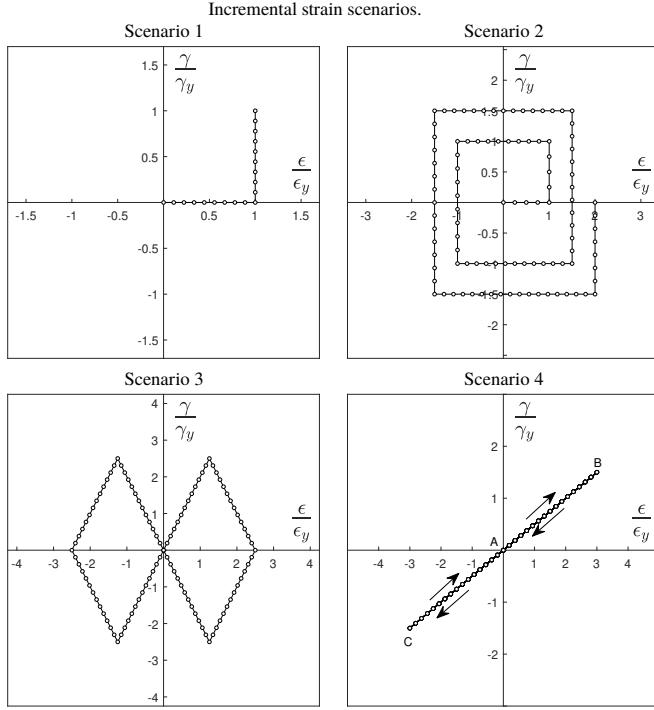


Fig. 3.5. The four different strain histories imposed, with the unstresses configuration as initial point.

compared to the other two methods. It should be highlighted that the 3D algorithm outperforms the one proposed by Simo because a perfectly plastic J_2 was used. In general, it too would require internal iterations to determine λ , in addition to the outer level Newton procedure to enforce the stress constraints. For Scenario 4, shown in Table 3.4, both the 3D and plane stress algorithms did [440, 188, 84, 62] elastic iterations corresponding to values of parameter α : [0.10, 0.25, 0.40, 0.50].

3.5.3 Plastic collapse of a doubly clamped beam

In this example we investigate the collapse load of a doubly clamped beam, shown in fig. 3.6, for two slenderness ratios, L/h : 20 and 4, where h is the cross-section height. We conduct analysis using both uncoupled and coupled, multiaxial, constitutive models at the fiber level and compare result, for each case, with the lower bound collapse load as predicted by limit analysis.

In addition, we showcase the influence of numerical quadrature in predicting the analytical

Table 3.1 Results for Scenario 1, perfect plasticity.

Algorithm	Exact $(\sigma_{11}, \sigma_{12})$	$\alpha = 0.10$		$\alpha = 0.25$		$\alpha = 0.40$		$\alpha = 0.50$	
		t^a	N_i	t	N_i	t	N_i	t	N_i
Proposed	(1.617, 1.023)	0.57	29	0.44	13	0.40	8	0.38	5
3D		1.40	46(20)	0.83	20(8)	0.65	11(4)	0.56	6(2)
Simo		1.64	165(19)	1.13	88(7)	0.96	64(3)	0.83	55(1)
Error(%) ^b			1.32		3.00		5.16		8.05

^a Measured in $\times 10^{-3}$ sec.^b Relative error in converged stress per increment scenario.**Table 3.2** Results for Scenario 2, perfect plasticity.

Algorithm	Exact $(\sigma_{11}, \sigma_{12})$	$\alpha = 0.10$		$\alpha = 0.25$		$\alpha = 0.40$		$\alpha = 0.50$	
		t^a	N_i	t	N_i	t	N_i	t	N_i
Proposed	(1.173, 1.208)	0.33	524	0.16	234	0.10	144	0.08	108
3D		1.24	849(97)	0.61	394(40)	0.31	196(17)	0.23	143(11)
Simo		2.09	2791(96)	1.17	1363(39)	0.57	790(15)	0.42	596(10)
Error(%)			2.32		5.00		10.10		13.35

^a Numbers in execution time columns are measured in $\times 10^{-2}$ s.**Table 3.3** Results for Scenario 3, perfect plasticity.

Algorithm	Exact $(\sigma_{11}, \sigma_{12})$	$\alpha = 0.10$		$\alpha = 0.25$		$\alpha = 0.40$		$\alpha = 0.50$	
		t^a	N_i	t	N_i	t	N_i	t	N_i
Proposed	(1.628, 1.018)	0.35	546	0.16	237	0.11	147	0.09	126
3D		1.15	920(115)	0.61	405(50)	0.31	194(22)	0.29	167(16)
Simo		2.20	2711(114)	1.42	1307(50)	0.55	789(20)	0.48	685(14)
Error(%)			1.15		2.51		5.40		5.80

^a Numbers in execution time columns are measured in $\times 10^{-2}$ s.**Table 3.4** Results for Scenario 4, five cycles, perfect plasticity. One cycle: A-B-C-A.

Algorithm	Exact $(\sigma_{11}, \sigma_{12})$	$\alpha = 0.10$		$\alpha = 0.25$		$\alpha = 0.40$		$\alpha = 0.50$	
		t^a	N_i	t	N_i	t	N_i	t	N_i
Proposed	(2.196, 0.559)	0.92	1500	0.44	678	0.22	392	0.19	345
3D		3.50	2500	1.61	1130	0.77	490	0.69	445
Simo		5.20	7497	2.93	4519	1.60	2352	1.38	2122
Error(%)			0.19		0.40		0.84		0.91

^a Numbers in execution time columns are measured in $\times 10^{-2}$ s.

solution in the slender case. More specifically, we compare Gauss-Legendre and Gauss-Lobatto quadratures. The quadrature rule used is crucial for both flexibility-based beam elements and the hybrid one presented and used herein, since, in both formulations, accuracy relies more on the precision of integration rather than mesh density. Lastly, each cross-section is discretized into 15 layers.

The shear strain distribution function, $k_q\varphi(X_2)$, is assumed to be parabolic and defined as follows:

$$\varphi(X_2) = (1 - 4 \frac{X_2^2}{h^2}) \quad (3.30)$$

where k_q is a parameter introduced in order to establish shear strain energy equivalence between assumed and exact distributions. For that matter, it is convenient to consider the uniform shear strain-average force hypothesis coupled with the corresponding shear coefficient as a representation of the exact one, assuming elastic behavior:

$$\mathcal{U}_{\text{shear}}^{\text{exact}} = k_s \mathcal{U}_{\text{shear}}^{\text{unif}} = \mathcal{U}_{\text{shear}}^{\text{quad}} \iff \frac{1}{2} k_s G A \gamma^2 = \frac{1}{2} G k_q^2 \left(\int_A \varphi(X_2)^2 dA \right) \gamma^2 \quad (3.31)$$

where k_s is the shear coefficient for uniform shear strain assumption. For rectangular sections with $\nu = 0.3$, $k_s = 0.886$ and solving eq. 3.31 for k_q we get:

$$k_q = \sqrt{\frac{k_s A}{\int_A \varphi(X_2)^2 dA}} \quad (3.32)$$

A first observation from fig 3.7 is that Gauss-Legendre quadrature tends to overestimate the capacity load in both edges of the slenderness spectrum and . Gauss-Lobatto quadrature, in contrast, is more accurate and converges to the “exact” capacity from “below”. In addition, it provides accuracy with just 5 quadrature points whereas even with 9 points, Gauss-Legendre still is not as accurate and converges from “above”. The results are in agreement with the ones reported by Papachristidis et al.[28] in a similar investigation but using a flexibility element.

When it comes to the influence of slenderness in the plastic capacity, the collapse load for

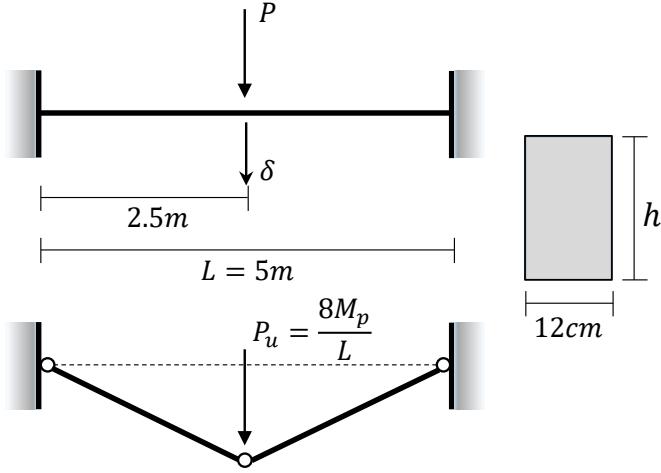


Fig. 3.6. Geometry, initial and final configuration of clamped beam, along with cross-section dimensions.

the L/h case predicted by the multiaxial constitutive law is lower than the one based on limit analysis. Interestingly, the uncoupled constitutive law results in a purely flexural collapse mode, thus yielding a ultimate load coinciding with the theoretical from limit analysis. When the beam is slender, both coupled and uncoupled constitutive laws result, expectably, in a flexure-dominant mode. The results are summarized in Table 3.5 and Fig. 3.7. In Fig. 3.8 the stress distributions for σ_{11} , σ_{12} and the von Mises stress, $\sigma_{VM} = \sqrt{3J_2}$ are presented at three different load levels indicate by point A, B and C (see Fig. 3.7b). The plastic zone propagation is also shown. We observe that the lower capacity as captured by the multiaxial model in the $L/h = 4$ case, which Fig. 3.8 pertains to, is due to shear plastification of the -shrinking- elastic core.

Table 3.5 Collapse moments and loads for clamped beam.

Collapse load, $P_u(kN)^a$					
L/h	$M_p(kN \cdot m)$	Limit Analysis	Uncoupled	Coupled	
20	$3.75 \cdot 10^2$	$6.00 \cdot 10^2$	$\approx 6.00 \cdot 10^2$	$\approx 6.00 \cdot 10^2$	
4	$9.37 \cdot 10^3$	$1.50 \cdot 10^4$	$\approx 1.50 \cdot 10^4$	$\approx 1.43 \cdot 10^4$	

Material data: $E = 200 \text{ GPa}$, $\sigma_y = 200 \text{ MPa}$.

^a For coupled and uncoupled loads, 5 Gauss-Lobatto points are assumed.

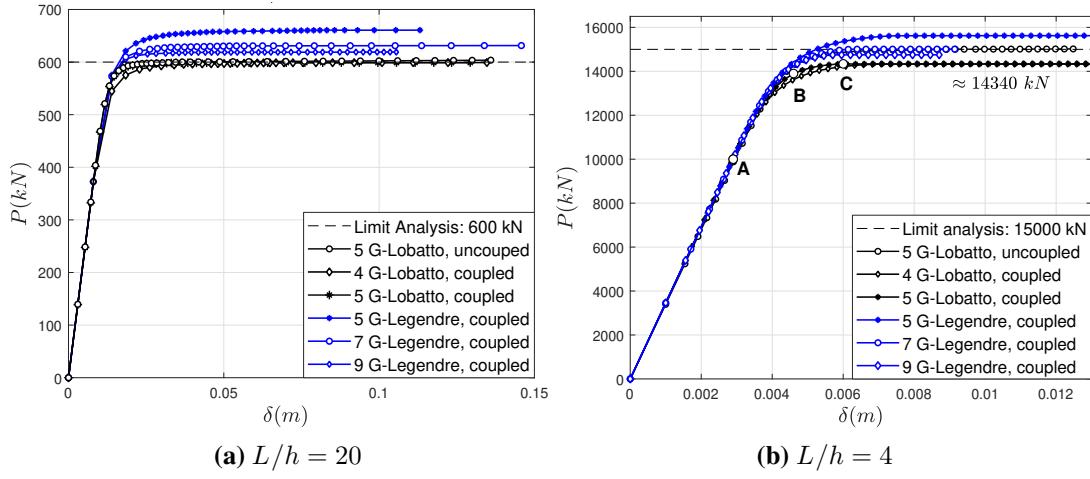


Fig. 3.7. Load-deflection plots for different slenderness ratios. Influence of quadrature type and number of points.

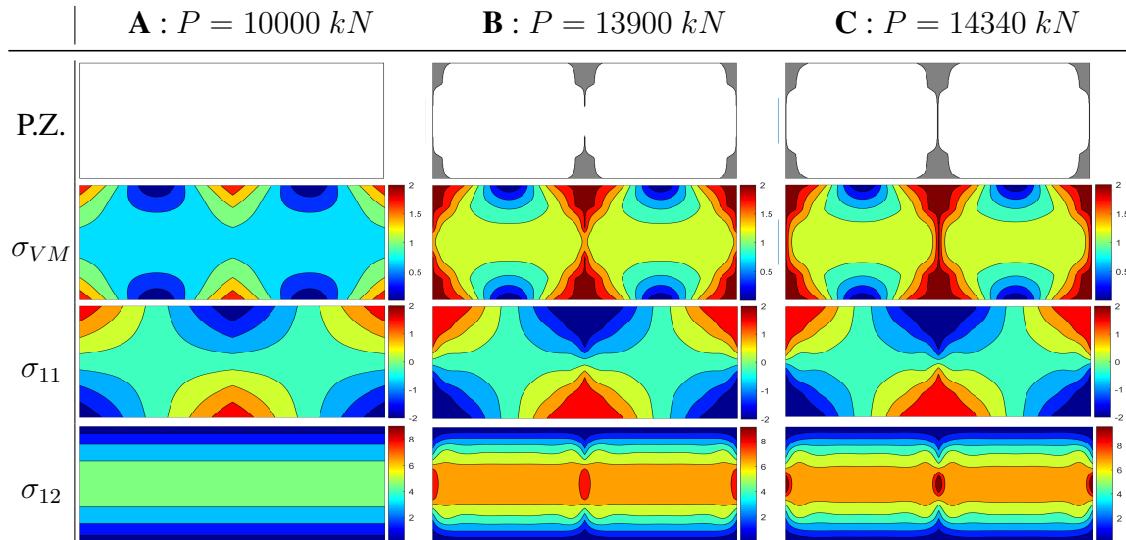


Fig. 3.8. Case $L/h = 4$. Spread of Plastic Zone (P.Z.) and stress distribution for von Mises, axial and shear stress at three different load levels.

3.5.4 Cyclic loading of shear link

In this example we are examining the response of a metallic shear link (Fig.3.9a) with a symmetric wide flange cross-section under displacement-controlled cyclic loading. The specimen we are concerned with was part of an experimental investigation by Hjelmstad & Popov[44] and later was numerically analyzed by Saritas & Filippou[29] using a mixed beam element formulation with shear-flexure interaction at the fiber level. In our analysys the specimen was modelled using one hybrid element with 5 Gauss-Lobatto points and 31 layers at each section with five layers on each flange.

The specimen length is $L = 28$ in. and the cross-section type is W 18 × 40 with basic dimensions specified as follows: total height $h = 17.88$ in., flange width $b = 5.985$ in., web and flange thickness, $t_w = 0.314$ in. and $t_F = 0.521$ in. respectively. The material properties are summarized in Table 3.6, while the displacement history, in accordance with Hjelmstad & Popov's experiment, is shown in Fig.3.9b The plastic modulus corresponding to linear kinematic hardening was determined from uniaxial data to be $H_{kin} = 102$ kips/in². A nonlinear hardening law was specified for the isotropic part, which is given as follows:

$$q(e^{pl}) = \sigma_y + C(1 - \exp(-\rho e^{pl})) \quad (3.33)$$

where $C = \sigma_u - \sigma_y$, σ_u , σ_y the ultimate and yield stresses respectively and ϵ_u , ϵ_y the corresponding strains. Parameter ρ is specified as $\rho = 20$ in this analysis. The shear strain distribution is again assumed to be quadratic, where $k_s = 0.466$ and, using eq. (3.32), $k_q = 1.33$.

As can be seen from Fig.3.10, during the first cycle we have satisfactory agreement between

Table 3.6 Material properties for Example 3.5.4.

Section	σ_y (kips/in ²)	σ_u (kips/in ²)	ϵ_y (%)	ϵ_u (%)	E (kips/in ²)	H_{kin} (kips/in ²)
W 18 × 40						
Web	39.5	60.1	1.8	22	28,300	102
Flange	35.0	58.5	1.4	24	28,000	102

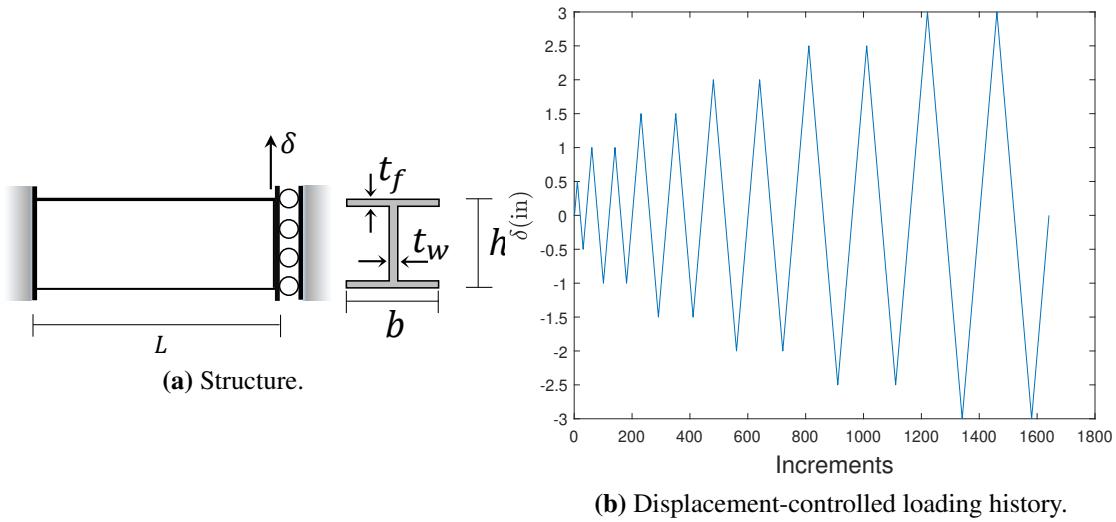


Fig. 3.9. Geometry of the specimen and displacement-controlled loading history.

the present analysis and the experimental result, while for all cycles after the third, we have almost excellent match between all cyclic histories shown. For the second and third cycles, with peak displacement at 1 inch, the present analysis exhibits a more abrupt yielding. This happens because most of the core becomes plastified within a smaller range of increments. While the hardening rules adopted might affect this behavior, it is the assumed shear strain distribution that plays a central role in transitioning from the elastic to plastic regime. Formulations that also consider shear stress distribution functions along the cross-section, section warping or even more accurate shear strain distribution models can definitely improve the approximation. We can also observe that both numerical models show identical accuracy during the unloading steps and a noticeable discrepancy with the experimental analysis. As explained in [29], this is due to improper restraint of the specimen during the experiment.

3.5.5 Elastoplastic post-buckling behavior of two-beam structure

This problem has been previously analyzed in [45, 46]. The structure, shown in Fig.3.11a, enters its inelastic regime immediately after the buckling load is reached, whereby point B starts moving inwards. In Fig.3.11b we compare the results we get using just 3 hybrid elements and 5 Gauss-Lobatto points in each element with Argyris et al.[45]. In addition, we include the inelastic

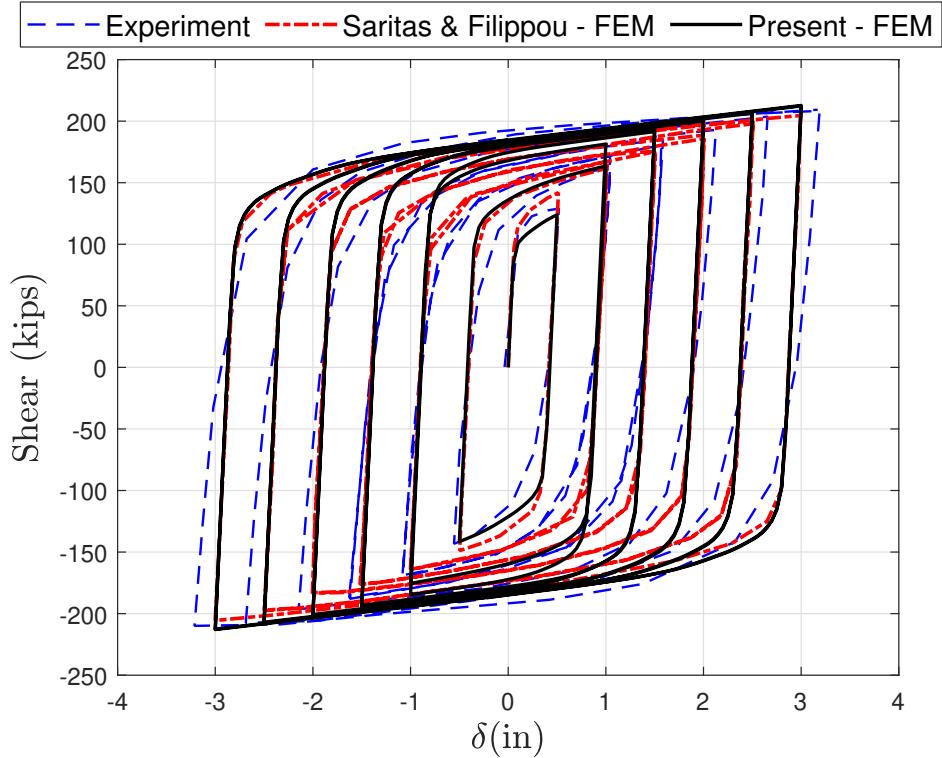
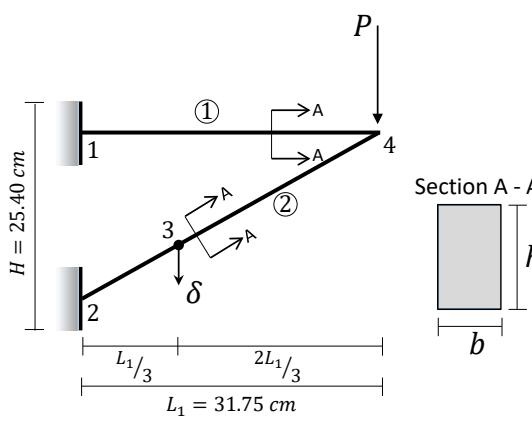


Fig. 3.10. Experimental and numerical results for the specimen.

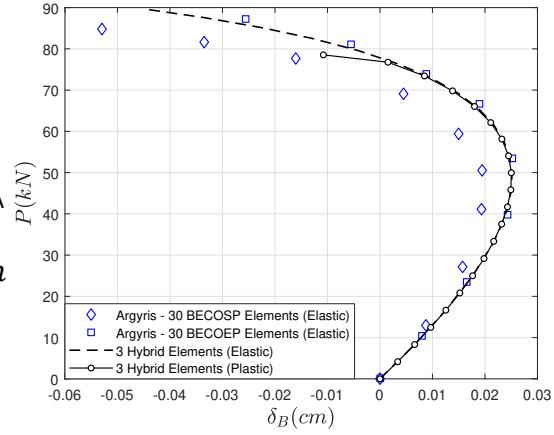
Table 3.7 Member geometric and material data.

Member	$E(\frac{\text{kN}}{\text{cm}^2})$	$H_{kin}(\frac{\text{kN}}{\text{cm}^2})$	$\sigma_y(\frac{\text{kN}}{\text{cm}^2})$	$h_s(\text{cm})$	$h_t(\text{cm})$	$b(\text{cm})$
(1)	6.86×10^3	2.84×10^2	26.20	1.91	8.00	2.54
(2)	6.86×10^3	2.84×10^2	33.10	1.91	8.00	2.54

analysis, which demonstrates that plastification starts immediately after then instability. The material and geometric data are shown in Table 3.7. Again, we highlight the impact of a coupled multiaxial law by investigating a variation of this problem. For all cases, inelastic response with kinematic hardening is assumed. The convergence rates in residual norm and energy norms for the case of thick members is also shown in Tables 3.8, 3.9. As can be seen from Fig.3.12b for the inelastic analysis of the thick member variation, the equilibrium paths corresponding to a coupled and uncoupled constitutive law tend to diverge shortly after the start of plastic deformation.



(a) Structure.



(b) Load vs deflection at point B.

Fig. 3.11. Geometry and loading of two-beam structure. Comparison of load vs deflection curves for different cases.

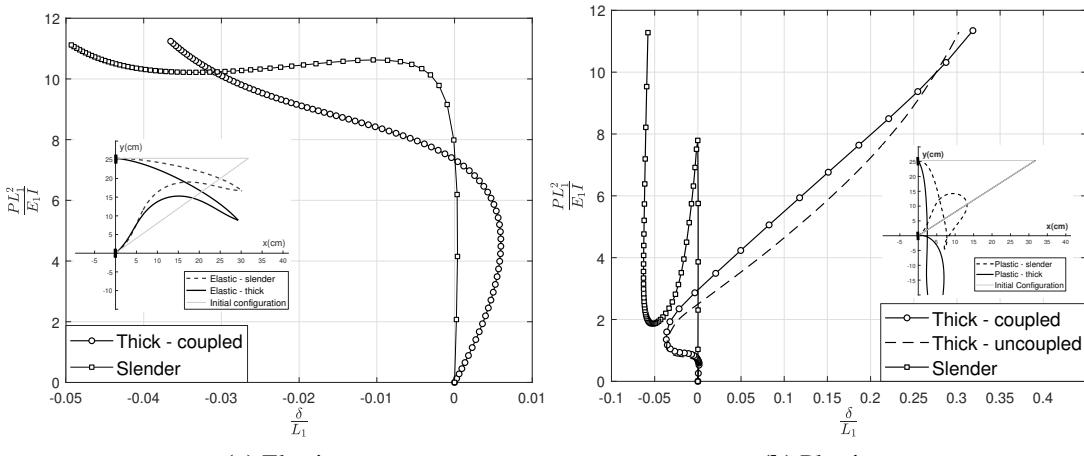


Fig. 3.12. Influence of slenderness and constitutive model on the response for elastic and elasto-plastic behavior.

Table 3.8 Convergence in residual norm during steps 5, 15, 25 35. $\epsilon_{tol} = 10^{-10}$.

iters.	5	15	25	35
1	$1.000000 \cdot 10^0$	$1.000000 \cdot 10^0$	$1.000000 \cdot 10^0$	$1.000000 \cdot 10^0$
2	$1.251499 \cdot 10^{-2}$	$1.988783 \cdot 10^{-2}$	$1.737845 \cdot 10^{-3}$	$6.187773 \cdot 10^{-2}$
3	$2.132711 \cdot 10^{-5}$	$1.066255 \cdot 10^{-5}$	$1.824164 \cdot 10^{-6}$	$4.214041 \cdot 10^{-3}$
4	$9.483080 \cdot 10^{-10}$	$4.882520 \cdot 10^{-10}$	$2.229120 \cdot 10^{-11}$	$6.111545 \cdot 10^{-5}$
5	$5.150000 \cdot 10^{-14}$	$7.790000 \cdot 10^{-13}$	$4.800000 \cdot 10^{-14}$	$1.218231 \cdot 10^{-8}$
6				$7.110000 \cdot 10^{-14}$

Table 3.9 Convergence in energy norm during steps 5, 15, 25, 35. $\epsilon_{tol} = 10^{-10}$.

iters.	5	15	25	35
1	$1.000000 \cdot 10^0$	$1.000000 \cdot 10^0$	$1.000000 \cdot 10^0$	$1.000000 \cdot 10^0$
2	$3.050574 \cdot 10^{-4}$	$1.065944 \cdot 10^{-4}$	$2.618542 \cdot 10^{-2}$	$1.188951 \cdot 10^{-2}$
3	$2.235712 \cdot 10^{-5}$	$5.373434 \cdot 10^{-5}$	$4.974768 \cdot 10^{-5}$	$1.495507 \cdot 10^{-4}$
4	$1.532842 \cdot 10^{-8}$	$2.585263 \cdot 10^{-9}$	$1.551550 \cdot 10^{-7}$	$1.706384 \cdot 10^{-5}$
5	$6.283000 \cdot 10^{-12}$	$1.962000 \cdot 10^{-12}$	$1.687900 \cdot 10^{-11}$	$3.948037 \cdot 10^{-7}$
7				$5.157400 \cdot 10^{-11}$

The full post-buckling elastoplastic equilibrium curve for the slender model is also provided for comparison. In Fig.3.12a the post-buckling response of both slender and thick models are illustrated. In the former, both uniaxial and multiaxial constitutive models yield identical results, assuming an infinite yield stress.

As we reduce the slenderness, slight differences arise compared to a purely uniaxial behavior at the fiber. This is mainly due to the assumed strain distribution, which is user defined and not accounted in the purely flexural model.

3.5.6 Three story frame with shear links

The three-story four-bay steel frame, first analyzed in [47], is shown in Fig. 3.13a, where the diagonal braces are ignored. Gravity loads are derived from the material and geometric properties of the elements and, for simplicity, are assumed to be acting on frame joints only. The values for each joint gravity load are shown in Table 3.10.

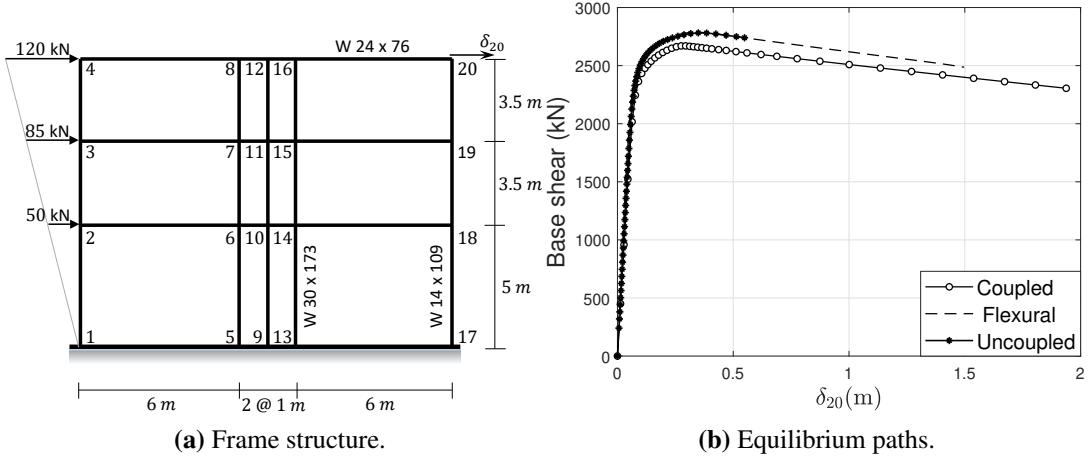


Fig. 3.13. Geometry of the steel frame and 3.13b response curves for three different models.

Table 3.10 Vertical gravity loads at nodes.

Nodes	2,18	3,19	4,20	6,14	7,15	8,16	10	11	12
$\downarrow P_i$ (kN)	240	220	145	340	300	190	255	215	120

A static pushover analysis is then conducted by with respect to the superimposed horizontal loads on nodes 2,3, and 4. For the discretization we used one hybrid element per member, 27 elements in total, and five Gauss-Lobatto quadrature points per element. All members have the same yield stress and Young's modulus: $\sigma_y = 290$ MPa, $E = 210$ GPa, and the Poisson's ratio is set to $\nu = 0.3$. Perfect plasticity is assumed and both material and geometric nonlinearity options are activated for the analysis. In Fig. 3.13b we show the results we get for 3 different modelling options regarding the constitutive law at the fiber level. Both the purely flexural model and the uncoupled elastoplastic model, whereby shear and axial states on fibers are uniaxial and do not interact, yield the same response curve and collapse load. In contrast, the coupled multiaxial model predicts, again, a lower plastic capacity load.

CHAPTER 4

NATURALLY PARAMETERIZED NLP FRAMEWORK

4.1 Introduction

This chapter presents a parametric nonlinear programming (PNLP) framework that extends the problem solving capabilities of the Hybrid NLP beam element beyond analyses concerned with mechanical loading. Such cases can be contact problems or structural sensitivity analysis. A strong feature of this framework lies in the utilization of homotopy continuation principles in order to establish conditions of global convergence to at least one solution. With the term “solution” we broadly refer to any critical point for the objective function, which in the present context is the TPE, with respect to the system variables.

Homotopy continuation, which forms the backbone of the PNLP framework, has been the subject of extensive studies in the field of applied mathematics[48, 49, 50, 51, 52, 53, 54, 55, 56, 57, 58, 59] as a technique for “globalizing” root finding algorithms. That is, even if the initial guess is not in the neighborhood of a root, homotopy theory guarantees convergence to at least one root, given certain regularity conditions hold. These conditions in essence establish the existence of a smooth path connecting the initial guess and that root. When applied to finding a root of a system of equations $F(x) = 0$, this can be done by introducing an “easy” problem whose a solution is easy to find, say $x - x_0 =$, and then construct a homotopy function $H(x, t) = tF(x) + (1 - t)(x - x_0)$, which represents a system of n equations with $n + 1$ unknowns. The system $H(x, t) = 0$ is sequentially solved for a discretization of $t \in [0, 1]$, starting at $t_0 = 0$, where the solution is known (“initial guess”). When $t = 1$ we have arrived at a root of the initial problem $F(x) = 0$. The process of varying t from 0 to 1 and solving the homotopy function is called deformation (not to be confused with mechanical deformation) and construction of such a homotopy function is called artificial homotopy. It is clear that the same principles can

be applied if system $F(x) = 0$ is undetermined, with n equations, $n + 1$ unknowns and a solution is available or can be found easily. In a nonlinear optimization context, application of the same principles have led to the formulation of global optimization algorithms[60, 61, 62, 63, 64, 65, 66, 67]. In addition, this framework highlights the firm theoretical underpinnings of various incremental methods that have been developed independently over the past decades for the numerical treatment of engineering problems[13, 68, 69, 70, 71, 72, 73, 74, 75, 76, 77, 78, 79, 80, 81, 82]. Application of this framework for approximating roots of systems of equations is commonly referred to as numerical continuation, whereas in the context of finding optimizers it is termed parametric optimization or PNLP.

In structural mechanics, differential equilibrium equations can be reformulated as an optimization problem due to the existence of a variational structure, as discussed in Chapter 2. Thus, the PNLP framework provides a suitable and versatile engine for generating global solutions to engineering problems using the Hybrid NLP-based element, since it allows for the incorporation of constraint specifications directly at the minimization statement. It is shown that, in the presence of inequality constraints, the global path that connects the initial guess to a critical point is continuous but piecewise differentiable[60, 62, 61]. Finally, utilizing the underlying homotopy arguments, we establish the conditions under which the path we follow is unique and convergent to at least one critical point. Non-uniqueness of the homotopy path implies the existence of bifurcation points. Because the derivations do not make use of artificial functions in order to introduce initial guesses, we termed the formulation as Naturally Parameterized NLP (NPNLP). While this has the downside of requiring a known solution of the system in order to initiate the homotopy deformation process, in structural mechanics such states can often be found easily (e.g. initial configuration).

This chapter is divided into 5 sections. In the first section we give an overview of homotopy continuation as applied in (globally) solving systems of nonlinear equations and in parametric optimization. In the second section formulate the NPNLP framework for the Hybrid element and in section three we outline the conditions that both the TPE and the element constraints

need to satisfy in order to guarantee global convergence following a unique path. A brief note on bifurcating paths is also included but the topic is not explored further herein. Finally, we demonstrate the capabilities of the framework through a set of problems from the mechanics and applied mathematics literature.

4.2 Homotopy Continuation

4.2.1 Main idea

The application of numerical approximation methods in virtually all engineering and scientific problems leads, generally, to an algebraic system of equations, \vec{F} , to be solved for the system unknowns, \vec{x} . This system, of which we seek one or more roots, is expressed in the following form:

$$\vec{F}(\vec{x}) = \vec{0} \quad (4.1)$$

where for simplicity we assume that $\vec{x} \in \mathbb{R}^n$ and $\vec{F} : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is a sufficiently smooth vector function. In the derivations that follow, boldface symbols denote vectors while bracketed bold-face symbols denote matrices. Given an initial estimate, \vec{x}_0 , one can use the Newton method to iteratively find a root:

$$\vec{x}_{i+1} = \vec{x}_i - \mathbf{F}_{\mathbf{x}, i}^{-1} \vec{F}_i, \quad i = 0, 1, \dots \quad (4.2)$$

This approach will yield a solution if i) \vec{x}_0 is already in the vicinity of a root and ii) if the Jacobian $\mathbf{F}_{\mathbf{x}}$ is not singular for any of the iterates $\{\vec{x}\}_i$. The main idea of the homotopy continuation is to overcome the difficulty that singularity points pose by imbedding problem (4.1) in a higher dimensional space where both \vec{x} and parameter t are considered independent variables. This is done by pairing function \vec{F} with a simpler problem, $\vec{G} : \mathbb{R}^n \rightarrow \mathbb{R}^n$ which has a trivial solution $\vec{G}(\vec{x}_0) = \vec{0}$, using a homotopy function, $\vec{H}(\vec{x}, t)$. For $\vec{H} : \mathbb{R}^n \times \mathbb{R} \rightarrow \mathbb{R}^n$ we require that i) $\vec{H} = \vec{0}$ for all (\vec{x}, t) and ii) $\vec{H}(\vec{x}_0, 0) = \vec{G}(\vec{x}_0)$ and $\vec{H}(\vec{x}_1, 1) = \vec{F}(\vec{x}_1) = \vec{0}$. Under certain regularity conditions, the Jacobian of \vec{H} , $D\vec{H}$, is of full rank even at points where $\mathbf{F}_{\mathbf{x}}$ is singular and, by the implicit function theorem, there exists a unique differentiable path $\Gamma : \mathbb{R} \rightarrow \mathbb{R}^n \times \mathbb{R}$,

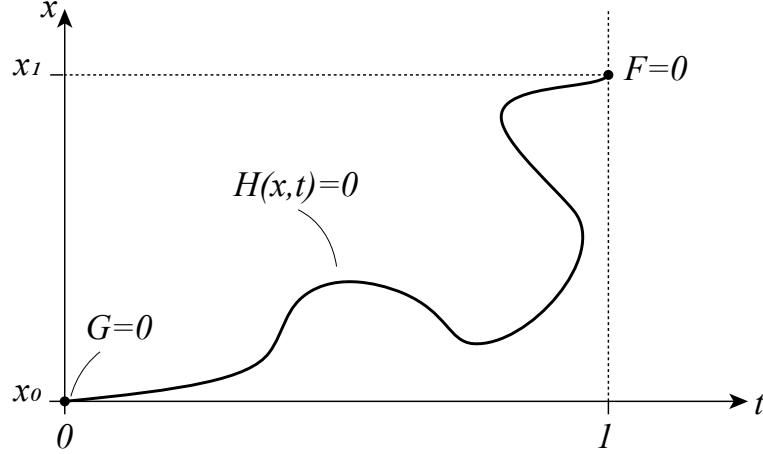


Fig. 4.1. A homotopy path connecting the trivial solution of $G = 0$ to a root of $F = 0$ for a single scalar equation dependent on x .

with $\Gamma(s) = (\vec{x}(s), t(s))$, $s \in \mathbb{R}$ and $\vec{H}(\Gamma) = \vec{0}$. This path connects the solution of the trivial problem, \vec{x}_0 , at $t_0 = 0$ with at least one root of \vec{F} , \vec{x}_1 , when $t = 1$ and can be tracked using numerical continuation[48]. The process of varying parameter t from $0 \rightarrow 1$ is referred to as deformation from the simple problem $\vec{G} = \vec{0}$ to the initial one, $\vec{F} = \vec{0}$, and such a path is depicted in figure (2.1).

The path Γ depends on how \vec{H} is defined, which in turn is determined by the choice for \vec{G} . We distinguish two general cases of homotopy imbeddings: artificial and natural.

One typically resorts to artificial homotopies when a simple solution to problem (4.1) is generally not available. The homotopy parameter t and the simpler problem \vec{G} are then simply devices that initialize the deformation process. In this case, only the final point $(\vec{x}_1, 1)$ is of interest and not the path Γ . A commonly used form of artificial homotopy is the convex homotopy:

$$\vec{H}(\vec{x}, t) = t\vec{F}(\vec{x}) + (1 - t)\vec{G}(\vec{x}) = \vec{0} \quad (4.3)$$

Among the variety of convex homotopy functions that have been developed, the ones most frequently used in implementations are:

- Probability one homotopy: $\vec{G}(\vec{x}) = \vec{x} - \vec{x}_0$ [52, 83]
- Newton homotopy: $\vec{G}(\vec{x}) = \vec{F}(\vec{x}) - \vec{F}(\vec{x}_0)$ [50, 84]

- Affine homotopy: $\vec{G}(\vec{x}) = \mathbf{F}_{x_0}(\vec{x} - \vec{x}_0)$ [85, 59]

On the other hand, systems whose response depends on variations of a parameter, t , that has physical meaning can be recast as natural homotopies as follows:

$$\vec{R}(\vec{x}) = 0 \xrightarrow{\text{parameterization}} \vec{R}(\vec{x}, t) = 0 \quad (4.4)$$

Frequently, in such cases an initial state of the system, \vec{R}_0 , is easily found and the deformation process starts from there. The process ends when one or more target states are found, which, by convention, correspond to $t = 1$. In contrast with artificial imbeddings, the path Γ generated by Eq. (4.4) is also of interest because all $(\vec{x}, t) \in \Gamma$ represent states of the system. As an example, the residual balance equations in nonlinear mechanics, $\vec{R}(\vec{u}) = \vec{0}$, can be cast as a natural homotopy, $\vec{R}(\vec{u}, t) = \vec{0}$, where \vec{u} is the displacement vector and λ represents the load intensity of extern loads.

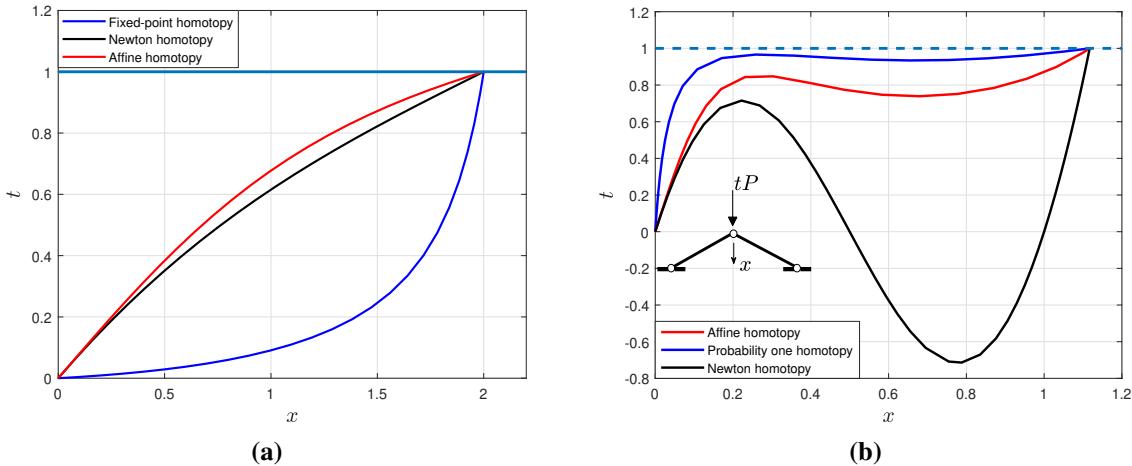


Fig. 4.2. Homotopy paths for (a) $f(x) = x^3 - 6x^2 + 21x - 26 = 0$, with starting point $x_0 = 0$ and convergence at root $x_1 = 2$ when $t = 1$ and (b) truss, with initial state the undeformed configuration($x_0 = 0$) and target load level $P = 0.03$.

A simple demonstration is provided in figure 4.2, where the three artificial homotopies mentioned above are used in two problems. In the first problem (Fig. 4.2a), we find a root of a cubic polynomial, while in the second (Fig. 4.2b) we solve the one degree of freedom truss for target state corresponding to load $P = 0.03$. Details for this example are taken from [86]. We see from the truss case that while, in principle, artificial imbeddings can be used for systems that can be naturally parameterized, the intermediate solution points on the path do not correspond to actual equilibrium configurations.

4.2.2 Regularity and existence of homotopy paths

From the discussion so far it is evident that the study of mappings $\vec{H} : \mathbb{R}^{n+1} \rightarrow \mathbb{R}^n$ is of primary importance. Consider such a mapping, $\vec{H}(\vec{v})$, which is at least of C^k continuity, $k \geq 2$, with $\vec{v} \in \mathbb{R}^{n+1}$ and t being its $(n+1)$ th component. For any given $\vec{w} \in \mathbb{R}^n$, the equation $\vec{H}(\vec{v}) = \vec{w}$ represents a system of n equations with $n+1$ unknowns and we designate the subset of \mathbb{R}^{n+1} satisfying this system as follows:

$$\vec{H}^{-1}(\vec{w}) = \{\vec{v} \in \mathbb{R}^{n+1} \mid \vec{H}(\vec{v}) = \vec{w}\} \subseteq \mathbb{R}^{n+1}$$

Definition 4.2.1 (Regular values of \vec{H}). *An element $\vec{w} \in \mathbb{R}^n$ is a regular value of $\vec{H} : \mathbb{R}^{n+1} \rightarrow \mathbb{R}^n$ if its Jacobian, $\mathbf{D}\vec{H}$, has rank n for all $\vec{v} \in \vec{H}^{-1}(\vec{w})$.*

Definition 4.2.2 (Regular points of \vec{H}). *An element $\vec{v} \in \mathbb{R}^{n+1}$ is a regular point of $\vec{H} : \mathbb{R}^{n+1} \rightarrow \mathbb{R}^n$ if its Jacobian at that point is of rank n .*

The components of the $n \times (n+1)$ Jacobian are given by $\mathbf{D}\vec{H}_{ij} = \partial H_i / \partial v_j$. It follows that if \vec{w} is a regular value for \vec{H} , then all $\vec{v} \in \vec{H}^{-1}(\vec{w})$ are regular points. The following lemma[48], stated here without proof, plays a constitutive role in homotopy continuation theory:

Lemma 4.2.1. *Let $\vec{H} : \mathbb{R}^{n+1} \rightarrow \mathbb{R}^n$ be a C^k map, $k \geq 2$, and having $\vec{0}$ as a regular value. Then, the set $\vec{H}^{-1}(\vec{0})$ is a C^k 1-dimensional manifold in \mathbb{R}^{n+1} , comprised of non-intersecting*

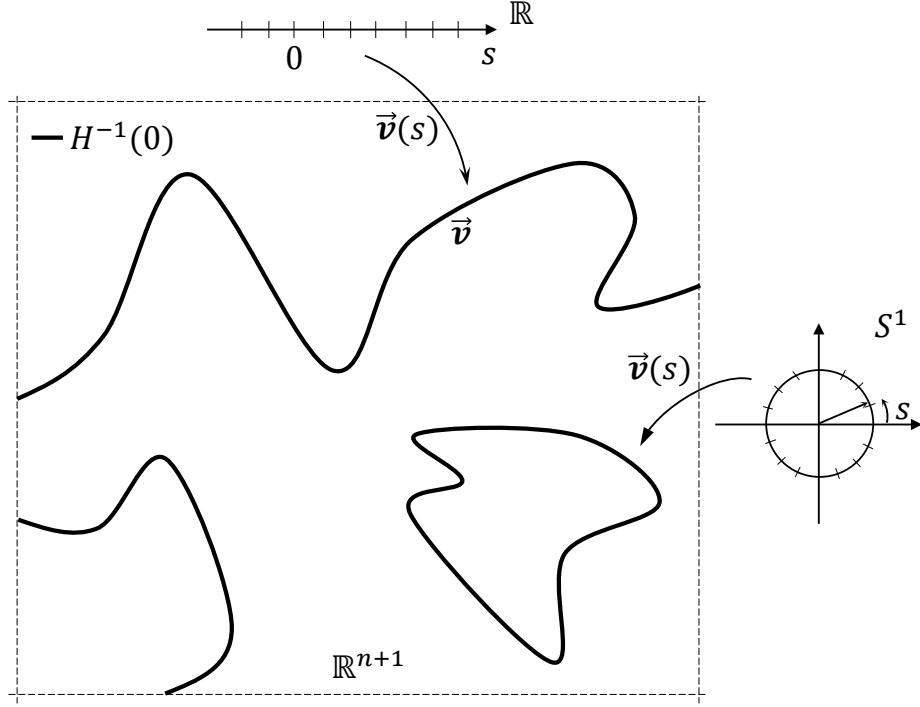


Fig. 4.3. Components of $H^{-1}(0)$ in \mathbb{R}^{n+1} .

components that are either i) homeomorphic with the unit circle S^1 (loops) or ii) homeomorphic with the real line \mathbb{R} (paths).

We refer to the collection of paths and loops in $\vec{H}^{-1}(\vec{0})$ as (connected) components. Figure 4.3 depicts a collection of such components for illustrative purposes. If $\vec{0}$ is a regular value of \vec{H} , by the implicit function theorem, a path(loop) Γ can be described as a C^k map $\Gamma : \mathbb{R}(\text{or } S^1) \rightarrow \mathbb{R}^{n+1}$ such that $\vec{v} = \vec{v}(s)$. The tangent to that curve is given by $\dot{\vec{v}} = d\vec{v}/ds$ and is always a non-zero vector(cf. [48, 85]).

Consider now the case of $\vec{H}(\vec{v})$ where the $n + 1$ -th component of \vec{v} can be regarded as a special parameter of interest, designated as t . We are interested in the zero points of \vec{H} :

$$\vec{H}(\vec{x}, t) = 0 \quad (4.5)$$

where $\vec{x} \in \mathbb{R}^n$, $t \in \mathbb{R}$. Any given initial point $(\vec{x}_0, 0)$ that satisfies Eq. (4.5) sets us on a

component of $\mathbf{H}^{-1}(\vec{\mathbf{0}})$ which we will designate as Γ_0 and define as follows:

$$\Gamma_0 = \{(\vec{x}, t) \mid \vec{x} \in \mathbb{R}^n, t \in \mathbb{R}, (\vec{x}_0, 0) \text{ in } \Gamma_0, \vec{\mathbf{H}}(\vec{x}_0, 0) = \vec{\mathbf{0}}\} \quad (4.6)$$

In most applications we want to ensure the initial point \vec{x}_0 sets us on a component Γ_0 which is a path, not a loop. Under the smoothness and regularity assumptions, the condition which guarantees that is for the $n \times n$ submatrix $\mathbf{D}\mathbf{H} = \partial \vec{\mathbf{H}} / \partial \vec{x}$ to be non-singular at \vec{x}_0 [85]. More conditions are generally required to guarantee that the path will reach $t = 1$ at least once and depend on the particular form of $\vec{\mathbf{H}}$. The interested reader can refer to [48], which provide proofs of existence and convergence for various types of homotopies. In addition, it is customary to formulate homotopies with $t \in [0, 1]$. In our formulation we adhere to the convention that a solution is reached when $t = 1$. However, as is the case in many practical applications, the restriction of t to lie in the unit interval is relaxed and the parameter is allowed to take values outside of that range. The expectation is that more solutions will be reached by crossing the level $t = 1$ multiple times. However, it should be noted that homotopy theory guarantees the (global) convergence to one solution $(\vec{x}_1, 1)$. While certain homotopy imbeddings can find multiple solutions when tracking numerically a single component path[50, 87, 88], it is difficult to establish conditions that guarantee convergence to many or all solutions in general[88].

4.2.3 Parametric nonlinear programming

Our derivations follow closely the works of Kojima & Hirabayashi[60], Gfrerer et al.[63], Gfrerer, Wacker et al.[61], Guddat et al.[62]. Consider the following one-parameter optimization program:

$$\mathbf{P}(t) : \min\{f(\vec{x}, t) \mid \vec{x} \in \Omega(t), t \in \mathbb{R}\} \quad (4.7)$$

with

$$\Omega(t) = \{\vec{x} \in \mathbb{R}^n \mid h_i(\vec{x}, t) = 0, i \in I_h, g_j(\vec{x}, t) \leq 0, j \in I_g\} \quad (4.8)$$

where we assume that functions $f : \mathbb{R}^{n+1} \rightarrow \mathbb{R}$, $h_i : \mathbb{R}^{n+1} \rightarrow \mathbb{R}$ and $g_j : \mathbb{R}^{n+1} \rightarrow \mathbb{R}$ are considered to be C^k , $k \geq 3$, $I_h = \{1, \dots, l\}$, $I_g = \{1, \dots, p\}$ and $l \ll n$, $p \ll n$. Let $\mathcal{L}(\vec{x}, \vec{\lambda}, \vec{\mu}, t)$ be the corresponding Lagrangian function of problem $\mathbf{P}(t)$:

$$\mathcal{L}(\vec{x}, \vec{\lambda}, \vec{\mu}, t) = f(\vec{x}, t) + \sum_{i \in I_h} \lambda_i h_i(\vec{x}, t) + \sum_{j \in I_g} \mu_j g_j(\vec{x}, t) \quad (4.9)$$

where $\vec{\lambda} \in \mathbb{R}^l$ and $\vec{\mu} \in \mathbb{R}^p$ are vectors of Lagrange multipliers, with $\vec{\lambda} = [\lambda_1 \dots \lambda_l]^T$, $\vec{\mu} = [\mu_1 \dots \mu_p]^T$. Additional specifications need to be imposed on set $\Omega(t)$ in order to derive a set of equations which will determine the solution set. Two conditions which we base our exposition on are the Linear Independence Constraint Qualification (LICQ) and the Strict Complementarity Condition (SCC). Let the index set of the active inequality constraints be defined as follows:

$$I_g^a(p) = \{j \mid g_j(\vec{x}, t) = 0, j \in I_g\} \quad (4.10)$$

Definition 4.2.3 (LICQ). At any $\vec{x} \in \Omega(t)$, if the gradients of the active constraints $\{\nabla h_i, \nabla g_j\}$, $i \in I_h$, $j \in I_g^a$, is a set of linearly independent vectors, then the linear independence constraint qualification holds at \vec{x} .

Definition 4.2.4 (SCC). At any $\vec{x} \in \Omega(t)$, if the Lagrange multipliers of the active set of inequality constraints are strictly positive, $\mu_j > 0 \forall j \in I_g^a$, then the strict complementarity condition holds at \vec{x} .

Under the LICQ, SSC and the smoothness assumptions, if \vec{x} is a minimum(maximum) point for $\mathbf{P}(t)$, then there exist unique $\vec{\lambda}$, $\vec{\mu}$ such that the point $(\vec{x}, \vec{\lambda}, \vec{\mu}, t)$ satisfies the so-called Karush-Kuhn-Tucker (KKT) system:

$$\mathbf{KKT}(t) : \quad \nabla_{\vec{x}} \mathcal{L}(\vec{x}, \boldsymbol{\lambda}, \boldsymbol{\mu}, t) = \vec{0} \quad (4.11a)$$

$$h_i(\vec{x}, t) = 0, \quad i \in I_h \quad (4.11b)$$

$$\mu_j g_j(\vec{x}, t) = 0 \quad (4.11c)$$

$$\mu_j \geq 0, g_j(\vec{x}, t) \leq 0, \quad j \in I_g \quad (4.11d)$$

The primary objective of PNLP is to examine the local and global behavior of points that satisfy the $\mathbf{KKT}(t)$ system (4.11). It is convenient to recast the KKT system above into a form that contains only equalities and, thus, homotopy continuation principles can be applied.

Let T be a subdivision of the augmented Lagrangian space $\mathbb{R}^n \times \mathbb{R}^l \times \mathbb{R}^p \times \mathbb{R}$ into cells $\tau(\bar{I}_g)$, where $\tau(\bar{I}_g) = \mathbb{R}^n \times \mathbb{R}^l \times \{\mu \in \mathbb{R}^p \mid \mu_j \geq 0, j \in \bar{I}_g, \mu_j \leq 0, j \in I_g \setminus \bar{I}_g\} \times \mathbb{R}$ for $\bar{I}_g \subset I_g$ and $I_g \setminus \bar{I}_g = \{j \in I_g \mid j \notin \bar{I}_g\}$. That is, the space is partitioned in cells with each cell identified by a particular active index set. Using an active index set strategy, Gfrerer et al[61] showed that problem $\mathbf{P}(t)$ and, consequently, $\mathbf{KKT}(t)$ can be replaced by an equivalent reduced system, which contains only (active) equality constraints[62, 61]:

$$\mathbf{P}^\tau(t) : \quad \min\{f(\vec{x}, t) \mid h_i(\vec{x}, t) = 0, i \in I_h, g_j(\vec{x}, t) = 0, j \in \bar{I}_g, \vec{x} \in \mathbb{R}^n, t \in \mathbb{R}\} \quad (4.12)$$

with the corresponding Lagrangian being:

$$\mathcal{L}^\tau(\vec{x}, \vec{\lambda}, \vec{\mu}, t) = f(\vec{x}, t) + \sum_{i \in I_h} \lambda_i h_i(\vec{x}, t) + \sum_{j \in \bar{I}_g} \mu_j g_j(\vec{x}, t) \quad (4.13)$$

Setting $\vec{z} = (\vec{x}, \vec{\lambda}, \vec{\mu}) \in \mathbb{R}^{n+l+|\bar{I}_g|}$, where $|\bar{I}_g| \leq p$ is the cardinality of \bar{I}_g , the reduced KKT

system for program **P** is:

$$\mathbf{KKT}^\tau(t) : \vec{\mathbf{H}}_{KKT}^\tau(\vec{\mathbf{z}}, t) = \begin{bmatrix} \nabla_{\vec{\mathbf{x}}} \mathcal{L}^\tau(\vec{\mathbf{z}}, t) \\ -h_i(\vec{\mathbf{x}}, t), i \in I_h \\ -g_j(\vec{\mathbf{x}}, t), j \in \bar{I}_g \end{bmatrix} = \begin{bmatrix} \vec{0} \\ 0 \\ 0 \end{bmatrix} \quad (4.14)$$

where $\vec{\mathbf{H}}_{KKT}^\tau(\vec{\mathbf{z}}, t) : \mathbb{R}^{n+l+|\bar{I}_g|+1} \rightarrow \mathbb{R}^{n+l+|\bar{I}_g|}$, that is, in each cell $\vec{\mathbf{H}}_{KKT}^\tau$ is of the form 4.5. This reformulation presupposes that i) LICQ and SCC hold throughout and ii) that $\vec{0}$ is a regular value of the reduced system $\vec{\mathbf{H}}_{KKT}^\tau$ on each cell $\tau(\bar{I}_g)$, which in turn leads to the following corollaries[60]:

1. The set $\cup_\tau [\vec{\mathbf{H}}_{KKT}^\tau]^{-1}(\vec{0})$ is comprised of piecewise continuously differentiable paths(loops) on $\mathbb{R}^{n+l+p+1}$.
2. Each path(or loop) is C^{k-1} in each $\tau(\bar{I}_g) \neq \emptyset$.
3. If a path(or loop) crosses a cell boundary between cells $\tau(\bar{I}_g^1)$ and $\tau(\bar{I}_g^2)$, then sets \bar{I}_g^1 and \bar{I}_g^2 differ by only one index and the tangent vector to the path at the intersection is not tangent to the boundary.

In addition, utilizing condition 3 above, there exists[61] a unique subdivision $\mathcal{D} = \bigcup_{q=1}^Q W^q$ of an interval $[\alpha, \beta]$ on the real line, such that:

- $[\alpha, \beta] = \bigcup_{q=1}^Q W^q$, $W^q = [s_q, s_{q+1}]$, and $s_1 = \alpha, s_Q = \beta$, $q = 1, \dots, Q$.
- $\bar{I}_g(s) = I^q$ where $I^q = I_g^a(s)$, for all s in (s_q, s_{q+1}) .
- $I^q \neq I^{q+1}$.
- By the implicit function theorem, there exist a C^{k-1} function $\vec{\phi}(s) = (\vec{\mathbf{z}}(s), t(s)) : W^q \rightarrow \mathbb{R}^{n+l+|\bar{I}_g|+1}$.
- $\vec{\phi}(s)$ is the solution set for $\mathbf{KKT}^\tau(t)$ for all $s \in W^q$.

In addition, if we require that $\|\mathrm{d}\vec{z}/\mathrm{d}s, \mathrm{d}t/\mathrm{d}s\| = 1$, then s represents the arc-length of the path(or loop). By repeating the process for $q = 1, \dots, Q$, a C^{k-1} piecewise continuous component Γ_0 can be traced numerically using appropriate continuation techniques. At a particular solution point (\vec{z}^*, t^*) , the values of inactive constraints $g_j(\vec{x}^*, t^*)$, $j \in I_g \setminus I^q$ and the Lagrange multipliers of the active constraints, μ_j , $j \in I^q$, are checked to ensure that complementarity slackness holds (eq. (4.11d)). Since adjacent cells differ in their respective active index sets by only one index, violation of the complementarity slackness condition indicates that the continuation entered cell $\tau(I^{q+1})$ and there is a unique index, j_k , such that:

$$g_{j_k}(\vec{x}^*, t^*) \geq 0, \quad j_k \in I_g \setminus I^q \quad \rightarrow \quad I^{q+1} = I^q \cup \{j_k\}, \quad g_{j_k} = 0$$

OR

$$\mu_{j_k} \leq 0, \quad j_k \in I^q \quad \rightarrow \quad I^{q+1} = I^q \setminus \{j_k\}, \quad \mu_{j_k} = 0$$

The collection of paths and loops of stationary points satisfying (4.11) is depicted in Fig. 4.4. For further details on the relevant proofs, refer to Gfrerer et al.[61] and the textbook by Guddat, Vasquez & Jongen[62].

4.2.4 Global optimization

Consider the following NLP:

$$\mathbf{P} : \quad \min\{f(\vec{x}) \mid \vec{x} \in \Omega\} \quad (4.15a)$$

$$\Omega = \{\vec{x} \in \mathbb{R}^n \mid h_i(\vec{x}) = 0, \quad i \in I_h, \quad g_j(\vec{x}) \leq 0, \quad j \in I_g\} \quad (4.15b)$$

where the same assumptions hold for f, h_i and g_j as in PNLP problem (4.7). Conventional numerical tools based in mathematical analysis guarantee the convergence to a (local) minimizer \vec{x}^* of \mathbf{P} provided that the initial guess is within the domain of convergence for the

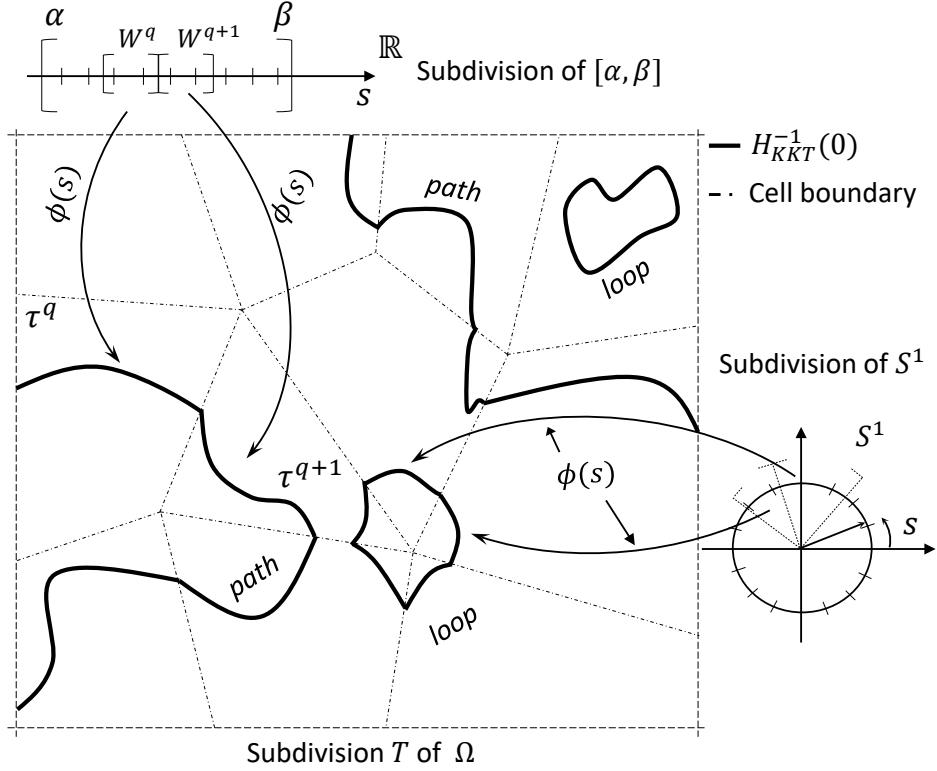


Fig. 4.4. Piecewise continuously differentiable paths and loops of $H_{KKT}^{-1}(0)$ in $\mathbb{R}^{n+l+p+1}$.

particular method employed. By utilizing Homotopy principles, we can convert \mathbf{P} into a PNLP using an appropriate imbedding. A numerical algorithm with global convergence properties can then be constructed, given that the conditions discussed previously hold. Consider the following one-parameter PNLP:

$$\mathcal{P}(t) : \quad \min\{\mathcal{F}(\vec{x}, t) \mid \vec{x} \in \Omega(t), t \in [0, 1]\} \quad (4.16a)$$

$$\Omega(t) = \{\vec{x} \in \mathbb{R}^n \mid \mathcal{H}_i(\vec{x}, t) = 0, i \in I_h, \mathcal{G}_j(\vec{x}, t) \leq 0, j \in I_g\} \quad (4.16b)$$

For $\mathcal{P}(t)$ to be a homotopy imbedding of \mathbf{P} , we require that:

1. NLP $\vec{\mathcal{P}}(0)$ possesses a trivial solution.
2. $\vec{\mathcal{P}}(1) = \mathbf{P}$

If \mathbf{P} can be naturally parameterized, then $\mathcal{P}(t) = \mathbf{P}(t)$, $\mathcal{F}(\vec{x}, t) = f(\vec{x}, t)$ and a stationary point, not necessarily unique, of $\mathbf{P}(0)$ is usually easy to compute. Artificial imbeddings can also be

employed when \mathbf{P} cannot be naturally parameterized. In such cases, a trivial stationary point for $\mathcal{P}(0)$ is easy to compute by construction. Below we list two artificial imbeddings that are frequently used in practice:

- $\mathcal{F}(\vec{x}, t) = tf(\vec{x}) + \frac{1}{2}(1-t)\|\vec{x} - \vec{x}_0\|^2$
- $\mathcal{F}(\vec{x}, t) = tf(\vec{x}) + \frac{1}{2}(1-t)\|f(\vec{x}) - f(\vec{x}_0)\|^2$

4.3 Naturally parameterized NLP framework for mechanics

We will now formulate a naturally parameterized nonlinear programming (NPNLP) framework for solid and structural mechanics applications and establish the conditions that guarantee the global convergence to a solution, using homotopy and PNLP principles discussed previously. The parameterization is considered natural because we assume that i) underlying physical quantities are parameterizable and ii) a solution for the initial state, not necessarily unique, is relatively easy to obtain. This solution can then be used as the starting point of the NPNLP deformation process.

Consider the hybrid NLP element program involving the discretized TPE (2.21) and the discretized (equality) constraints (2.28):

$$\mathbf{P} : \quad \min\{\Pi(\vec{d}) \mid \vec{d} \in \Omega\} \quad (4.17a)$$

$$\Omega = \{\vec{d} \in \mathbb{R}^n \mid h_i(\vec{d}) = 0, i = 1, \dots\} \quad (4.17b)$$

In the derivations that follow we assume a set of inequality constraints is also introduced. Furthermore, to simplify the exposition, we assume a cell subdivision of the parameter domain and work only with the set of inequality constraints that are active, \vec{g}_a . Since this allows us to deal only with equality constraints, we can consider vector \vec{h} to include both the standard hybrid element constraints (2.32a),(2.32b) and \vec{g}_a . Consequently, the corresponding vector of Lagrange

multipliers, $\vec{\lambda}$, will be augmented with the non-zero components of vector $\vec{\mu}$, as defined in the previous section. Under these assumptions, the general form of the naturally parameterized NLP (NPNLP) program can be stated as follows:

$$\mathcal{P}(t) : \min\{\Pi(\vec{d}, t) \mid \vec{d} \in \Omega(t), t \in \mathbb{R}\} \quad (4.18a)$$

$$\Omega(t) = \{\vec{d} \in \mathbb{R}^n \mid h_i(\vec{d}, t) = 0, i = 1, \dots, l\} \quad (4.18b)$$

where in \vec{d} we have collected all state variables associated with the structural system (e.g. strains, displacements etc.). The Lagrangian function of $\mathcal{P}(t)$, counterpart to (2.30), is:

$$\mathcal{L}(\vec{d}, \vec{\lambda}, t) = \Pi(\vec{d}, t) + \vec{h}^T \vec{\lambda} \quad (4.19)$$

with $\vec{h}(\vec{d}, t) = [\vec{h}^A(\vec{d})^T \ \vec{h}^B(\vec{d})^T \ g_a(\vec{d}, t)^T]^T$. The corresponding parameterized KKT system for a particular t , considering again $\vec{z} = [\vec{d}^T \ \vec{\lambda}^T]^T, \vec{z} \in \mathbb{R}^{n+l}$:

$$\vec{H}(\vec{z}, t) = \begin{bmatrix} \nabla_{\vec{d}} \mathcal{L}(\vec{z}, t) \\ \vec{h}(\vec{d}, t) \end{bmatrix} = \vec{0} \quad (4.20)$$

Since $\vec{H} : \mathbb{R}^{n+l+1} \rightarrow \mathbb{R}^{n+l}$, we can apply homotopy continuation principles to system 4.20 in order to numerically track a component (path or loop) in the current cell. We define the set of equilibrium configurations as follows:

$$\mathcal{E} = \vec{H}^{-1}(\vec{0}) = \{(\vec{z}, t) \in \mathbb{R}^{n+l+1} \mid \vec{H}(\vec{z}, t) = \vec{0}\} \quad (4.21)$$

It is implicitly assumed that, for NPNLP $\mathcal{P}(t)$ to be a natural homotopy, in addition to conditions 1 and 2 mentioned in the preceding section, stationary points that correspond to values of $t \neq 1$ should represent actual (intermediate) states of our structural system. If an

artificial imbedding was used in place of (4.18a) instead, then only solutions corresponding to $t = 1$ could be regarded as physical equilibrium states. In that case, during the numerical continuation, the homotopy path need not be tracked with strict tolerance, since intermediate solutions are of no physical significance. For the NPNLP framework however, all solutions along the path are equilibrium configurations and numerical accuracy is required through the whole process. Therefore, we will refer to solutions \vec{d} of (4.20) as equilibrium configurations or states. We will not be concerned with their characterization (stable, unstable, neutral). Moreover, initial and target values, Y_0 , Y_1 respectively, of the parameterized quantity $Y = Y(t)$ are prescribed such that $Y(0) = Y_0$ and $Y(1) = Y_1$. If a trivial equilibrium state is not apparent for $t = 0$, then $\mathcal{P}(0)$ can be solved as an NLP so that an approximate equilibrium configuration \vec{d}_0 is found. With that known, the initial configuration of the system is completely determined and the deformation process of the NPNLP can begin. As an example, consider the case where the parameterized quantity is the load vector \vec{P} such that the TPE is $\Pi(\vec{d}, t) = U_s(\vec{d}) - t\vec{P}^T \vec{u}$, with \vec{u} being the global nodal displacement vector. If we consider $Y(t) = t\vec{P}^T \vec{u}$, then it is clear that for $Y(0) = 0$, a trivial equilibrium state of $\mathcal{P}(0)$ is the unstressed configuration, $\vec{d} = \vec{0}$. In contrast, if only the k -th component of \vec{P} , P_k , is parameterized such that $Y(t) = tP_k$, then an equilibrium configuration for $\mathcal{P}(0)$ will need to be found through an analysis of the structure loaded with $\vec{P}_0 = [P_1 \ P_2 \ \dots \ P_{k-1} \ 0 \ P_{k+1} \ \dots \ P_n]^T$.

4.3.1 Existence and uniqueness of NPNLP path

We examine now under what conditions a particular component $\Gamma_0 \subseteq \mathcal{E}$ emanating from \vec{d}_0 converges to an equilibrium state of \mathbf{P} at least once. The $(n + l) \times (n + l + 1)$ Jacobian of the KKT system (4.20), considering variations in λ as well, is:

$$\mathbf{D}\mathbf{H} = [\mathbf{H}_z \ \frac{d\vec{H}}{dt}] \quad (4.22)$$

where \mathbf{H}_z is the Lagrangian matrix defined by:

$$\mathbf{H}_z = \begin{bmatrix} \mathbf{H} & \nabla \mathbf{h} \\ \nabla \mathbf{h}^T & \mathbf{0} \end{bmatrix} \quad (4.23)$$

and $\mathbf{H} = \nabla_d^2 \mathcal{L}$ is the Hessian matrix of the Lagrangian function (4.19). In the absence of constraints, $\mathbf{H}_z = \mathbf{H} = \mathbf{K}$, where \mathbf{K} is the global stiffness matrix. The following conditions guarantee that a path Γ_0 emanating from the couple $(\mathbf{d}_0, 0)$ (initial “guess”) converges to a solution of $\mathbf{P} = \mathcal{P}(1)$ at least once if the following conditions hold[61]:

1. The TPE function, Π , and the constraints, h_i , should be C^k and C^{k-1} function respectively, $k \geq 2$.
2. Zero is a regular value of the KKT system (4.20) in each cell.
3. The couple $(\vec{\mathbf{d}}_0, 0)$ is a unique equilibrium state for $\mathcal{P}(0)$.
4. \mathcal{E} is a compact set for $t \in [0, 1]$ and for the cell subdivision.
5. $\mathcal{P}(t)$ satisfies the LICQ in all cells corresponding to $t \in [0, 1]$.

Condition 3 is the most restrictive as far as natural homotopies are concerned, since it is quite often that $\mathcal{P}(0)$ possesses multiple stationary points. This may necessitate to perform additional analyses using different solutions of $\mathcal{P}(0)$ as starting points, if these points lie on different paths on \mathcal{E} . In addition, convergence to additional solutions of $\mathcal{P}(1)$, when t is allowed to take values greater than one, might subsequently lead to an accumulation point. This suggests that as $t \rightarrow \infty$, $\vec{\mathbf{d}} \rightarrow \vec{\mathbf{d}}^*$ and the path is therefore not compact.

Condition 5 can be relaxed so that we assume LICQ holds for t outside $[0, 1]$ with the expectation of finding additional equilibrium states for $\mathcal{P}(1)$. However, this assumption does not guarantee convergence beyond the first stationary point. Moreover, convergence to the first equilibrium point can be guaranteed even with the weaker Mangasarian-Fromovitz constraint qualification, as is proven in[60, 61]. In that case, the Lagrange multiplier vectors at stationary

points are not unique and the set \mathcal{E} has different structure. Its projection on $\vec{d} - t$ space however is the same as the one with LICQ assumed.

Regularity condition 2 implies that no secondary branches (bifurcations) occur along a particular path Γ_0 , as this would imply there exist points where $\text{Rank}(\mathbf{DH}) < n+l$. Let the tangent space of the current active constraints at point (\vec{d}, t) be defined as follows:

$$\mathcal{T} = \{\vec{\xi} \in \mathbb{R}^n \mid [\nabla_d \vec{h}]^T \vec{\xi} = \vec{0}\} \quad (4.24)$$

Following Poore[65], Tiahrt& Poore[67] we assume that SCC holds for all (\vec{d}, t) in the interior of a cell that also satisfy Eq. (4.20). Then, $\vec{0}$ being a regular value for \vec{H} in any cell $\tau = \tau(I^q)$, $q = 1, \dots, Q$, implies the following conditions:

A. Loss of LIQC - In the case, we require that:

- $\text{Dim}(\text{Span}(\{\nabla_d h_1 \dots \nabla_d h_l\})) \geq l - 1$.
- The matrix \mathbf{J}_h , has full rank: $\text{Rank}(\mathbf{J}_h) = l$, where $\mathbf{J}_h = \begin{bmatrix} \nabla_d \vec{h}^T & \frac{d\vec{h}}{dt} \end{bmatrix}$

B. Singularity of Hessian \mathbf{H} on \mathcal{T} - In this case, we require that:

- $\text{Rank}(\nabla_d \vec{h}^T \mathbf{H} \nabla_d \vec{h}) \geq l - 1$.
- $\text{Proj}_{\mathcal{T}}(\nabla_d(\frac{dL}{d\lambda})) \notin \text{Range}(\nabla_d \vec{h}^T \mathbf{H} \nabla_d \vec{h})$.
where $\nabla_d \vec{h}^T$ is the basis of \mathcal{T} in terms of the (active) constraint gradients and $\text{Proj}_{\mathcal{T}}(\vec{w}) = \nabla_d \vec{h} (\nabla_d \vec{h}^T \nabla_d \vec{h})^{-1} \nabla_d \vec{h}^T \vec{w}$ is the orthogonal projection of a vector $\vec{w} \in \mathbb{R}^n$ on the tangent space \mathcal{T} .

C. Loss of LIQC and Singularity of \mathbf{H} on \mathcal{T} - In this case, we require that:

- $\text{Dim}(\text{Span}(\{\nabla_d h_1 \dots \nabla_d h_l\})) \geq l - 1$.
- $\text{Rank}(\mathbf{J}_h) = l$.
- $\text{Rank}(\nabla_d \vec{h}^T \mathbf{H} \nabla_d \vec{h}) \geq l - 1$.

- $\text{Rank}(\mathbf{H}_z) \geq n + l - 1$.

The second condition in A and the second condition in B essentially guarantee that, in the event of violation of LICQ or singularity of the Hessian in the tangent space, for the last column of $D\mathbf{H}$ we will have $\frac{d\vec{H}}{dt} \notin \text{Range}(\mathbf{H}_z)$ and \mathcal{E} remains a one-dimensional manifold.

Under conditions 1-5 and the restrictions A,B,C implied by the regularity assumption, the component Γ_0 is a path which emanates from the initial configuration \mathbf{d}_0 with $t = 0$ and converges in finite steps in at least one solution of problem \mathbf{P} when $t = 1$. Moreover, the only singular points on a path in the interior of a cell are fold points. When the path crosses a cell boundary, the SCC is violated. This is essentially a bifurcating behavior , where two curves intersect at that boundary. The branch switching is carried out by enforcing the complementarity slackness condition so that we maintain a direction that leads to feasible points. Numerical continuation can now be utilized to track Γ_0 , using an appropriate parameterization and an active set strategy.

4.4 Implementation

We now present a practical and robust algorithm for the implicit path continuation for the active set KKT system which is suited for structural mechanics applications using the NPNLP framework. It follows closely the methodology of the secant length algorithm[89] but is modified by employing a technique found in Crisfield[13] and Batoz & Dhatt [71]. This is done so that the algorithm takes into account the frequently well structured form of the Lagrangian matrix \mathbf{H}_z while at the same time remaining agnostic with respect to the parameterization. This allows to bypass the expensive computation of the pseudo-inverse that is required in the secant length scheme at the start of every step and instead use the homotopy parameter t to enforce the constraint exactly.

Let $\Delta(\cdot)_i$ denote the cumulative changes of a quantity up to iteration i and $\delta(\cdot)_i$ the iterative

changes in iteration i , such that:

$$\Delta(\cdot)_i = \Delta(\cdot)_{i-1} + \delta(\cdot)_i \quad (4.25)$$

where $\Delta(\cdot)_0 = 0$. The secant length constraint, $\|(\vec{z} - \vec{z}_N, t - t_N)\|_2^2 = \Delta s^2$, leads to the following iterative form:

$$\Delta \vec{z}_i^T \Delta \vec{z}_i + \Delta t_i^2 = \Delta s^2, \quad i = 1, \dots, i_{max} \quad (4.26)$$

In the derivations that follow, it is assumed we are at iteration i and drop the subscript from the iterative change vectors, $\delta(\cdot)$. Using Eq. (4.25) for \vec{z} , t , and substituting into Eq. (4.26) yields the following iterative constraint:

$$\Delta \vec{z}_{i-1}^T \Delta \vec{z}_{i-1} + \delta \vec{z}^T \delta \vec{z} + 2\Delta \vec{z}_{i-1} \delta \vec{z} + \Delta t_{i-1}^2 + \delta t^2 + 2\Delta t_{i-1} \delta t = \Delta s^2 \quad (4.27)$$

From Eq. (4.20), we require $\vec{H}_i = \vec{H}(\vec{z}_{i-1} + \delta \vec{z}, t_{i-1} + \delta t) = \vec{0}$ which, after using Taylor expansion, leads to the following expression:

$$\vec{H}_{i-1} + [\mathbf{H}_z]_{i-1} \delta \vec{z} + \frac{d\vec{H}}{dt} \Big|_i \delta t = \vec{0} \quad (4.28)$$

We note here that all quantities in iteration $i - 1$ are known and that $\|\vec{H}_{i-1}\|_2 \neq 0$. Solving Eq. (4.28) for $\delta \vec{z}$ leads to:

$$\delta \vec{z} = \delta \vec{z}_1 + \delta t \delta \vec{z}_2 \quad (4.29)$$

where

$$\delta \vec{z}_1 = -[\mathbf{H}_z]_{i-1}^{-1} \vec{H}_{i-1}, \quad \text{and} \quad \delta \vec{z}_2 = -[\mathbf{H}_z]_{i-1}^{-1} \frac{d\vec{H}}{dt} \Big|_i \quad (4.30)$$

Substitution of Eq. (4.29) in the iterative constraint (4.27) leads to the following quadratic

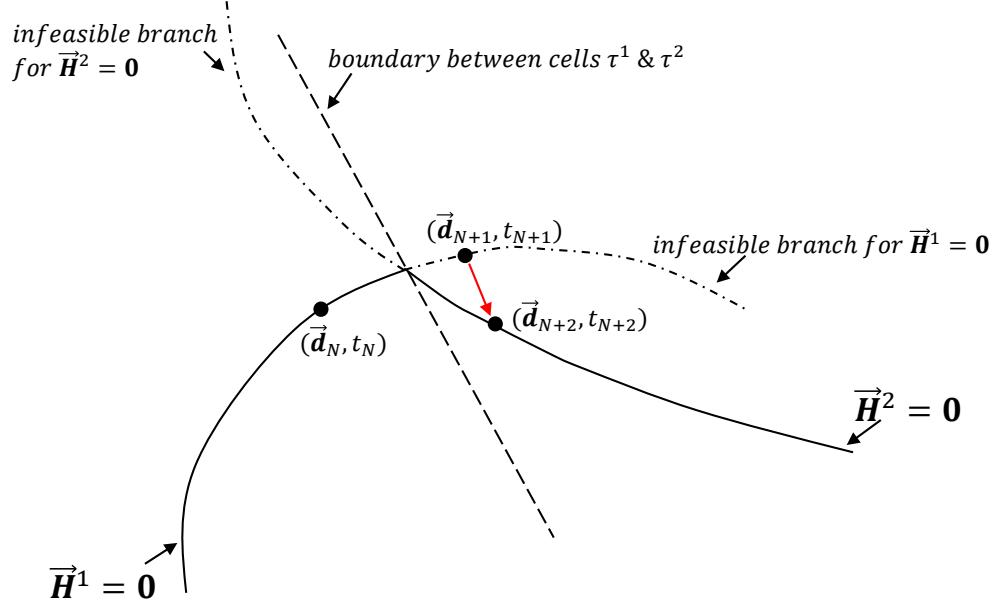


Fig. 4.5. Transition from cell τ^1 to τ^2 during numerical continuation. Projection of $\vec{H}(\vec{z}, t) = \vec{0}$ on $\vec{d} - t$ space.

equation to be solved for δt :

$$\alpha_1 \delta t^2 + \alpha_2 \delta t + \alpha_3 = 0 \quad (4.31)$$

Coefficients α_1 , α_2 and α_3 are determined as follows:

- $\alpha_1 = \delta \vec{z}_2^T \delta \vec{z}_2 + 1$
- $\alpha_2 = 2(\delta \vec{z}_1^T \delta \vec{z}_2 + \Delta \vec{z}_{i-1}^T \delta \vec{z}_2 + \Delta t_{i-1})$
- $\alpha_3 = \Delta \vec{z}_{i-1}^T \Delta \vec{z}_{i-1} + \Delta t_{i-1}^2 + \delta \vec{z}_1^T \delta \vec{z}_1 + 2\Delta \vec{z}_{i-1}^T \delta \vec{z}_1 - \Delta s^2$

The root is chosen so that a forward direction of tracking persists. An intuitive criterion proposed by Crisfield[13] suggests selecting the root of Eq. (4.31) that leads to the smallest inner product $\Delta \vec{z}_i^T \Delta \vec{z}_{i-1}$, which can be regarded as the iterative form of the continuity condition.

As is the case with the Secant Length parameterization proposed by Menzel & Schwetlick[89], the algorithm presented here bypasses the need to factorize an augmented $(n+l+1) \times (n+l+1)$ Jacobian by requiring exact satisfaction of the secant length constraint but also avoids the costly numerical computation of the pseudo-inverse. This is done by exploiting the decomposition of

DH in its abstract form (4.22) using the technique of Batoz & Dhatt[71]. However, the form proposed here is more general since the algorithm can handle any parameterization. Moreover, the vector $\frac{d\vec{H}}{dt}$ is not generally constant and depends on both the specific parameterization as well as the current active set. The algorithm is also suited for general parametric optimization problems where such a decomposition of the Jacobian is amenable.

In the case where the new solution point (\vec{z}_{N+1}, t_{N+1}) violates the complementarity slackness, then a new active set has to be determined, which will lead to a modified $\vec{H} = \vec{0}$. Let \vec{H}^1 be the homotopy associated with cell $\tau^1 = \tau(I^1)$, and \vec{H}^2 the homotopy with the updated active set on cell $\tau^2 = \tau(I^2)$. Then, point (\vec{d}_{N+1}, t_{N+1}) , which corresponds to (\vec{z}_{N+1}, t_{N+1}) , can be used as an initial estimation once the Lagrange multiplier vector has been adjusted according to the current active index set I^2 , and use the condition $\vec{H}^2 = \vec{0}$ to perform corrections back to the new branch. The infeasible point at step $N + 1$ can then be discarded from the solution set. Alternatively, other approaches can be implemented, such as interpolation in order to locate the corner point or simply repeat the step from step N but with reduced step-length so that any solutions off the feasible branches do not lead to significant violation in any constraint. This is illustrated in Fig. 4.5 for steps N , $N + 1$ and $N + 2$, where the path is projected on the $\vec{d} - t$ space.

4.5 Numerical Examples

4.5.1 Example 1 - Global Optimization

In the first application we are concerned with the global optimization of a 2-D function. In particular, we explore the landscape using PNLP and homotopy principles to generate components that lead to one or more stationary points. Second order optimality principles can be used then to characterize these points. The function we choose is a variation of the Himmelblau function, for which Kuno & Seader[90] provided seven stationary points. The function is:

$$f(x_1, x_2) = 1/2x_1^4 + 1/2x_2^4 + x_1^2x_2 + x_1x_2^2 - 11x_1^2 - 7x_2^2 + 13x_1 + 9x_2 \quad (4.32)$$

with $-5 \leq x_1, x_2 \leq 5$. We use two natural imbeddings:

- $f_1(x_1, x_2, t) = 1/2x_1^4 + 1/2x_2^4 + x_1^2x_2 + x_1x_2^2 - 11x_1^2 - 7x_2^2 + 13x_1 + g(t)x_2$
- $f_2(x_1, x_2, t) = 1/2x_1^4 + 1/2x_2^4 + x_1^2x_2 + x_1x_2^2 - 11x_1^2 - 7x_2^2 + 13tx_1 + 9tx_2$

In both cases, the stationary points of Eq. (4.32) and $f_i(x_1, x_2, 1)$, $i = 1, 2$, coincide. Function $g(t)$ in f_1 is defined in such a way so that $g(1) = 9$ and $g(0) = 50$. This results in a linear function $g(\lambda) = -41\lambda + 50$. We chose $g(0) = 50$ so that a (starting) stationary point of $f_1(x_1, x_2, 0)$ is not close enough to a stationary solution of Eq. (4.32). Moreover, we let t get values larger than 1 with the hope that more than one stationary points is found per component followed.

The first order optimality condition requires $\vec{H}_i(x_1, x_2, t) = \nabla_{\mathbf{x}} f_i = 0$, $i = 1, 2$, for any variation in λ :

$$\vec{H}_1(x_1, x_2, t) = \begin{bmatrix} 2x_1^3 + 2x_1x_2 + x_2^2 - 22x_1 + 13 \\ 2x_2^3 + x_1^2 + 2x_1x_2 - 41t + 50 \end{bmatrix}, \quad \vec{H}_2(x_1, x_2, t) = \begin{bmatrix} 2x_1^3 + 2x_1x_2 + x_2^2 - 22x_1 + 13 \\ 2x_2^3 + x_1^2 + 2x_1x_2 + 9t \end{bmatrix} \quad (4.33)$$

The 2×3 Jacobians associated with \vec{H}_i are given, in block form, as $\mathbf{D}\mathbf{H}_i = [[\mathbf{H}_{\mathbf{x}}]_i \frac{d\vec{H}}{dt}]$, with:

$$[\mathbf{H}_{\mathbf{x}}]_1 = [\mathbf{H}_{\mathbf{x}}]_2 = \begin{bmatrix} 6x_1^2 + 2x_2 - 22 & 2x_1 + 2x_2 \\ 2x_1 + 2x_2 & 6x_2^2 + 2x_1 \end{bmatrix}, \frac{d\vec{H}}{dt} \Big|_1 = \begin{bmatrix} 0 \\ -41 \end{bmatrix}, \frac{d\vec{H}}{dt} \Big|_2 = \begin{bmatrix} 0 \\ 9 \end{bmatrix} \quad (4.34)$$

Two paths are generated, Γ_1 and Γ_2 , using imbedding f_1 , emanating from points $\mathbf{P}_1 = (-4.3676, -4.3868)$ and $\mathbf{P}_2 = (3.2589, -3.5165)$. These points are stationary points of the initial program $f_1(x_1, x_2, 0)$ and they are found using an artificial Newton homotopy. The initial points that were then used as starting points in the algorithm described in Sec. 4.4 were $(\mathbf{P}_1, 0)$, $(\mathbf{P}_2, 0)$ for Γ_1 and Γ_2 respectively. The step-length used was $\Delta s = 0.4$ and t_{max} was set to 5. For the second imbedding, an obvious stationary solution of $f_2(x_1, x_2, 0)$ is $\mathbf{P}_3 = (0, 0)$. We used the point $(\mathbf{P}_3, 0)$ to generate path Γ_3 with $\Delta s = 0.4$ and $t_{max} = 12$, which we track in both directions. As can

Table 4.1 Stationary solutions of Eq. (4.32).

Kuno & Seader[90]	Γ_1	Γ_2	Γ_3
x_1	x_2		
-4.255	-3.845	✓	✓
-3.454	-1.144	✓	✓
-3.369	2.514	✓	✓
0.987	1.801		✓
0.713	0.856	✓	✓
0.793	-2.814		✓
3.252	-2.712	✓	✓

be seen from Table (4.1), path Γ_1 finds three stationary points, Γ_2 finds four and Γ_3 finds all seven stationary points. In figure (4.6a) these components are depicted in the \mathbb{R}^3 space and in figure (4.6b) we show their projections on \mathbb{R}^2 , interpreting them as the trajectory followed on the landscape of $f(x_1, x_2)$.

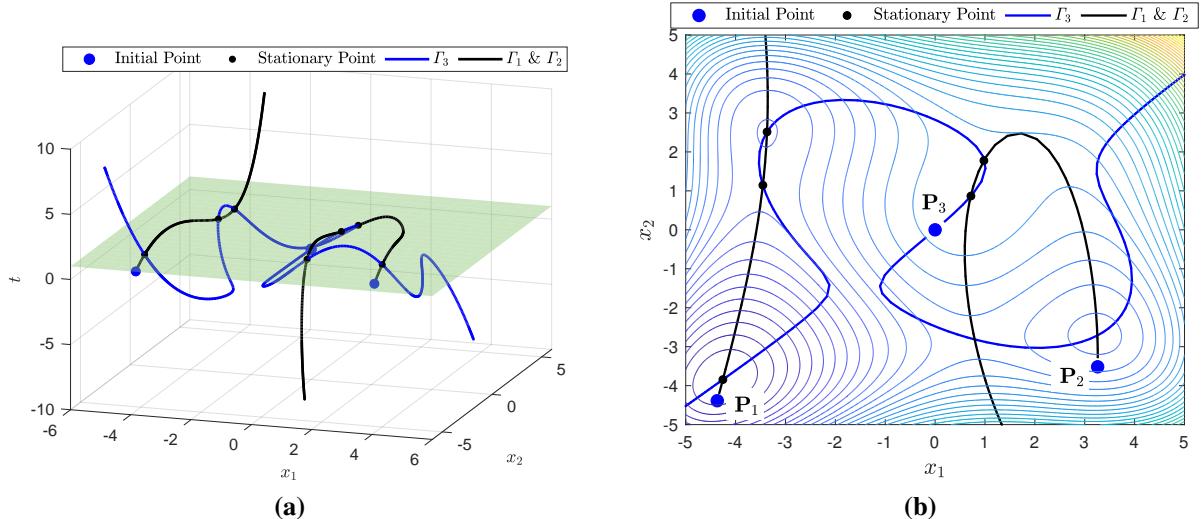


Fig. 4.6. (a) Paths Γ_1 , Γ_2 and Γ_3 in \mathbb{R}^3 , (b) interpretations of paths as trajectories on the target TPE function (4.32).

4.5.2 Example 2 - Global Sensitivity

In this application we investigate the global sensitivity of a two degree of freedom truss structure with respect to the axial stiffness of one member. The example, which has appeared in different forms in various investigations[91, 92], is adopted from [75] and modified in accordance with Fig. ???. Assuming linear elastic constitutive law, the system TPE is as follows:

$$\Pi(u_1, u_2) = \frac{1}{2} \frac{E_1 A_1}{L} \left(\sqrt{(L - u_1)^2 + (h + u_2)^2} - \sqrt{L^2 + h^2} \right)^2 + \frac{1}{2} \frac{E_2 A_2}{L} u_2^2 - P u_1$$

with the Young's moduli being $E_1 = E_2 = 30$, cross-sectional area for member 1 is $A_1 = 20$, $L = 5$, $h = 2$ and external load $P = 100$. To facilitate the exposition, we first parameterize with respect to external load in a linear fashion, $P(t) = 100t$, and solve for three different values of A_2 : $A_2 = \{4A_1, A_1/3, A_1/5\}$. We will then conduct a parametric study with A_2 being the “natural” parameter, while the external load will remain fixed at $P = 100$. The equilibrium paths from the initial analyses, which correspond to the three distinct values of A_2 , will be used as a reference for the global sensitivity of the structure to (large) variations of A_2 , starting from the system with $A_2 = 4A_1$ and setting as Target System(TS) the one with $A_2 = A_1/3$. The case where $A_2 = A_1/5$ was included to highlight the fact that below a certain threshold in terms of A_2 , the system no longer possesses multiple solutions for $P = 100$. The load parameterized NPNLP is:

$$\mathcal{P}_P(t) : \min\{\Pi(u_1, u_2, t) \mid u_1, u_2 \in \mathbb{R}, t \in \mathbb{R}\} \quad (4.35)$$

The first order optimality condition requires that $\vec{\mathbf{H}}(u_1, u_2, t) = \nabla_{\mathbf{u}}\Pi = \vec{0}$ for all t . Again, the resulting Jacobian **DH** is of dimension 2×3 , where $\mathbf{DH} = [\mathbf{H}_u \ \mathbf{H}_t]$, where \mathbf{H}_t indicates differentiation of $\vec{\mathbf{H}}$ with respect to t . In the absence of constraints, the Lagrangian matrix \mathbf{H}_u corresponds to the stiffness matrix of the structure while $\mathbf{H}_t = \nabla_{tu}\Pi = [-P \ 0]^T$. For each distinct A_2 , $\mathcal{P}_P(0)$ possesses a trivial solution, namely, the unstressed configuration:

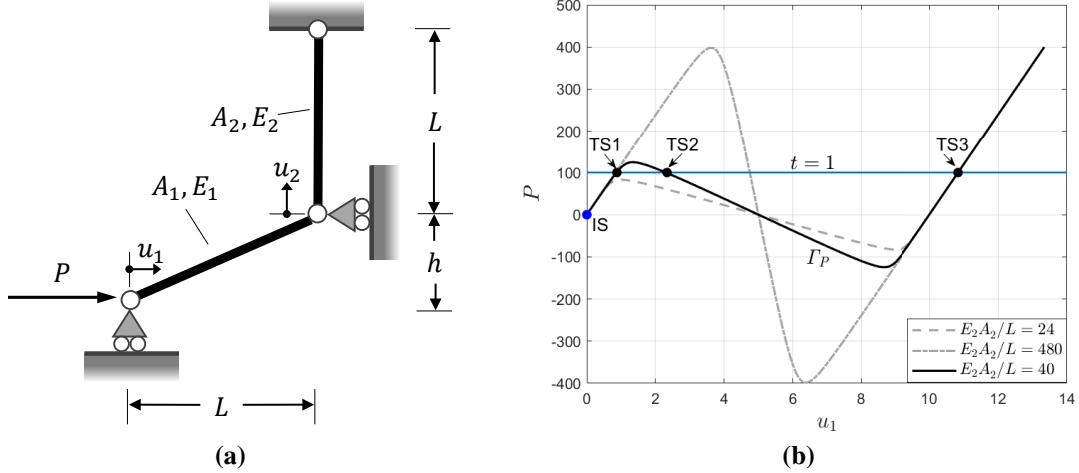


Fig. 4.7. (a) Truss structure, (b) equilibrium paths for $A_2 = \{4A_1, A_1/3, A_1/5\}$ projected on the $u_1 - P$ plane.

$(u_1, u_2)_0 = (0, 0)$. Starting from this state and applying the continuation algorithm described in Sec. 4.4 with $t_{max} = 4$, we generate the three equilibrium paths depicted in Fig. 4.7b. We now proceed to investigate the global sensitivity of the system with respect to A_2 . We define a nominal value for A_2 , $\bar{A}_2 = A_1$, and induce variations by introducing the parameterization $A_2(t) = (-11/3t + 4)\bar{A}_2$. In this way, the Target States (TS) have been chosen to correspond to $A_2(t = 1) = A_1/3$ and are also depicted in Fig. (4.7b). The resulting NPNLP is:

$$\mathcal{P}_{A_2}(t) : \quad \min\{\Pi(u_1, u_2, t) \mid u_1, u_2 \in \mathbb{R}, t \in \mathbb{R}\} \quad (4.36)$$

Again, the natural homotopy equation $\vec{H} = \vec{0}$ results from the first order conditions, with \mathbf{H}_u acting as the structure stiffness, however, in this case we have for \mathbf{H}_t :

$$\mathbf{H}_t = -\frac{11}{3} \frac{E_2 \bar{A}_2}{L} \begin{bmatrix} 0 \\ u_2 \end{bmatrix} \quad (4.37)$$

The initial NLP $\mathcal{P}_{A_2}(0)$, which corresponds to $A_2(0) = 4\bar{A}_2 = 4A_1$, has three stationary points for $P = 100$. We use two of them as Initial States (IS) to start the continuation process for $\mathcal{P}_{A_2}(t)$:

Table 4.2 Initial States for paths Γ_1 , Γ_2 and Target States reached.

	u_1	u_2	Γ_1	Γ_2
IS1	0.836	0.010	×	
IS2	10.836	-0.006		×
TS1	0.867	0.310	✓	
TS2	2.302	2.507	✓	
TS3	10.836	-0.060		✓

$$1. \text{ IS1} : (u_1, u_2) = (0.836, 0.010)$$

$$2. \text{ IS2} : (u_1, u_2) = (10.836, -0.006)$$

The paths generated from IS1 and IS2, shown in Fig. 4.8a, are Γ_1 and Γ_2 respectively. As can be seen from this figure and Table 4.2, the target NLP $\mathcal{P}_{A_2}(1)$ has three stationary points with Γ_1 passing through two of them (TS1 and TS2), while Γ_2 reaches only one, TS3. This can also be illustrated by projecting Γ_1 and Γ_2 on the target TPE function, as can be seen in Fig. 4.8b, where we have also included the equilibrium path, Γ_P , associated with the load parameterized target system discussed earlier, $\mathcal{P}_P(t)$ with $A_2 = A_1/3$. Clearly, the points of intersection between this equilibrium path and paths Γ_1 and Γ_2 are precisely the points TS1, TS2 and TS3. Moreover, the persistence of vertical orientation of paths beyond a certain point hints to the existence of an accumulation point. As can be seen in Fig. 4.8a, for Γ_1 , we have $|t| \rightarrow \infty$ and $(u_1, u_2) \rightarrow (4.950, 0.078)$ and, if traced in the opposite direction, $(u_1, u_2) \rightarrow (0.835, 0.001)$ which are indications of accumulation points. That is, the initial point IS1 is already an accumulation point. Physically, this means that after a certain value of axial stiffness for the vertical member, u_1 is only due to deformation of the inclined truss member. Similarly, Γ_2 also has an accumulation point if traced backwards. These accumulation points can also be illustrated in Fig. 4.8a, where the equilibrium paths converge before TS1 and after $u_1 \approx 9.2$. In addition, we see from Figs. 4.7b, 4.8a that when axial stiffness $E_2 A_2 / L$ becomes even lower, then there exists only one solution for $P = 100$.

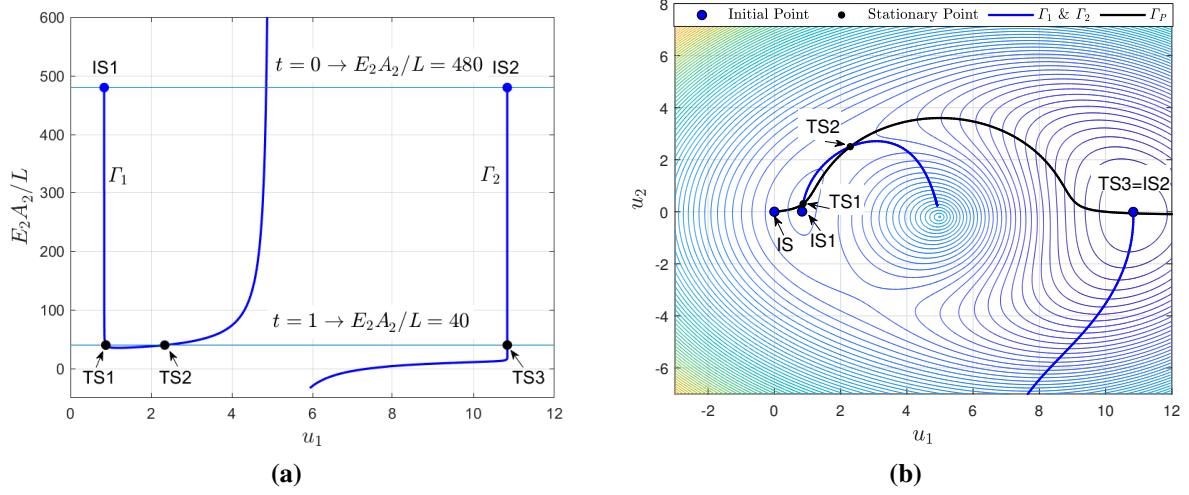


Fig. 4.8. (a) projection of components Γ_1 & Γ_2 on the $u_1 - E_2 A_2(t)/L$ plane, (b) paths Γ_1 , Γ_2 and Γ_p projected on the TPE of the target NLP $P_{A_2}(1)$.

4.5.3 Example 3 - Frictionless contact

In this example we consider a two-dimensional structure comprised of two bars, shown in Fig. 4.9, and investigate the response in the presence of a frictionless rigid surface described by function g in terms of the displacements u_1, u_2 of node 3. This configuration has previously been investigated under stability considerations[93, 94]. Here, we assume the following nonlinear elastic constitutive law:

$$\sigma_{PK2} = \frac{1}{2}E\ln(1 + \epsilon_G) \quad (4.38)$$

where E is the Young's modulus, σ_{PK2} is the second Piola-Kirchhoff stress and ϵ_G is the work conjugate Green's strain. It is equivalent to the constitutive law adopted in the relevant example in [94] when Poisson's ratio is 0.5. An initial imperfection is assumed by prescribing a small horizontal load of magnitude 0.01 and subsequently the response is parameterized with respect to the vertical load P .

The structure TPE is then:

$$\Pi(u_1, u_2, t) = U_1 + U_2 - 0.01u_1 - tPu_2 \quad (4.39)$$

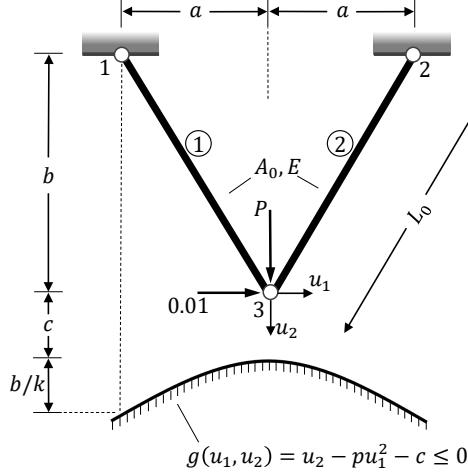


Fig. 4.9. Initial geometry for Example 3.

where $U_i = EA_0/4\ln(1 + \delta_i/L_0)\delta_i$ is the strain energy of bar i and δ_i its elongation, with A_0 , L_0 being the initial cross-section area and length respectively. The elongations are determined as follows:

- $\delta_1 = \sqrt{(b + u_2)^2 + (a + u_1)^2} - L_0$
- $\delta_2 = \sqrt{(b + u_2)^2 + (a - u_1)^2} - L_0$

and $L_0 = \sqrt{a^2 + b^2}$. The constraint surface is defined as:

$$g(u_1, u_2) = u_2 - pu_1^2 - c \leq 0 \quad (4.40)$$

where $p = b/(ka^2)$, $k \in (0, \infty)$, is a parameter controlling the curvature of g . For $k \rightarrow \infty$, the contact surface becomes horizontal at vertical distance c from node 3, while letting $k \rightarrow 0$ effectively makes g inactive for any (u_1, u_2) . The material and geometric properties were chosen as follows: $A_0 = 1$, $E = 10$, $a = 1$, $b = 3$, $c = 1$ and the target state is defined when $P = 10$. The NPNLP problem is then expressed as follows:

$$\mathcal{P}_P(t) : \min\{\Pi(u_1, u_2, t) \mid u_1, u_2 \in \Omega, t \in \mathbb{R}\} \quad (4.41a)$$

$$\Omega = \{u_1, u_2 \in \mathbb{R} \mid g(u_1, u_2) \leq 0, \} \quad (4.41b)$$

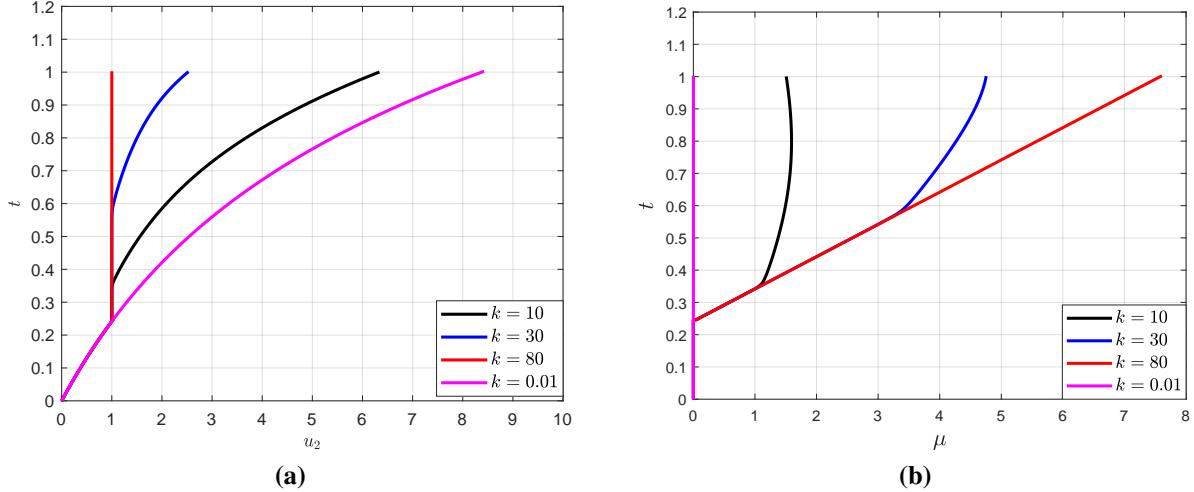


Fig. 4.10. Projection of path (a) on $u_2 - t$ plane and (b) $\mu - t$ plane.

The Lagrangian function for this problem, $\Omega \times \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$, is defined as $\mathcal{L}(u_1, u_2, \mu, t) = \Pi(u_1, u_2, t) + \mu g(u_1, u_2)$, with μ being the Lagrange multiplier for g . The initial point provided to the continuation algorithm is $\mathbf{P}_0 = (0.0317, 0)$, which is the unique stationary point for $\mathcal{P}_P(0)$ and can be sufficiently approximated considering linear response. At that point the constraint g is inactive and μ is set to zero. For values $u_2 \gtrsim 1$, the active set changes, which is indicated by the presence of a non-smooth point on the path. Constraint g becomes active and μ is nonzero. The projected paths on the $u_2 - t$ and $\mu - t$ planes for four levels of k are depicted in Fig. 4.10, with the case $k = 0.01$ corresponding to an effectively flat contact surface. The transition from the first active index set $I^1 = \{\emptyset\}$ to the second, $I^2 = \{1\}$ happens $t \approx 0.25$. The configuration profiles at four distinct levels of t , shown in Fig. 4.11, correspond to a surface g determined with $k = 30$.

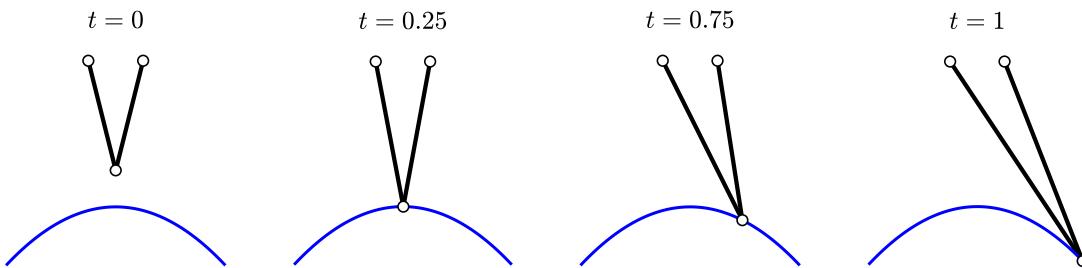


Fig. 4.11. Configuration profiles during deformation.

CHAPTER 5

WEIGHTED LEAST SQUARES PREDICTOR FOR NUMERICAL CONTINUATION

5.1 Introduction

In this chapter, an efficient prediction scheme for numerical continuation of systems of the form $\vec{F} = \vec{0}$, where $\vec{F} : \mathbb{R}^{n+1} \rightarrow \mathbb{R}^n$, is developed. The formulation does not require second order information, yields predictor directions consistent with the local geometry of the solution set and remains linear in complexity irrespective of the Jacobian structure. It can be implemented in any variant of path-tracking algorithms and uses stabilized local weighted least squares to fit a local polynomial on the last $k \geq 2$ solution points, with the weighting function defined so that solution points closer to the current one are assigned higher weight. This leads to a versatile scheme where the order of approximation is decoupled from the points supplanted and can be easily adjusted and updated, if further criteria are enforced. Moreover, the prediction can be based either on extrapolation from the fitted polynomial or its tangent, which is trivial to compute. Since the solution set of such systems is a curve, the parameterization best suited for the proposed algorithm is the arc-length of the curve.

The chapter is structured as follows: in the first section we provide a summary numerical continuation. In the second section, the formulation of the Weighted Least Squares (WLS) prediction scheme is introduced, followed by a section dealing with implementation details. The chapter concludes with a section dedicated to numerical tests. We demonstrate that predictions based on the proposed scheme are competitive compared to more expensive approaches utilizing the tangent predictor in terms of total steps, function evaluations and linear solves. Furthermore, it is shown that they can facilitate the continuation of the tracking process even in cases where conventional prediction schemes lead to premature termination.

5.2 Numerical continuation

We focus our attention to undetermined nonlinear systems of the form¹:

$$\vec{F}(\vec{y}) = \vec{0} \quad (5.1)$$

where $\vec{F} : \mathbb{R}^{n+1} \rightarrow \mathbb{R}^n$ is a sufficiently smooth map. Numerical continuation attempts to solve these systems, assuming that the regularity assumptions hold. We have already introduced the necessary theoretical concepts pertaining to such systems in the previous chapter. In what follows, we assume that \vec{F} is a sufficiently smooth map and $\vec{0}$ is a regular value for system (5.1). Under these assumptions, the solution set \mathcal{S} is a one-dimensional manifold in \mathbb{R}^{n+1} consisting of connected components homeomorphic to the real line (paths) and the unit circle (loops) and with the same smoothness as \vec{F} . For \mathcal{S} we have:

$$\mathcal{S} = \{\vec{y} \in \mathbb{R}^{n+1} \mid \vec{F}(\vec{y}) = \vec{0}\}, \quad \text{and} \quad \text{rank}(\mathbf{J}) = n \quad \forall \vec{y} \in \mathcal{S} \quad (5.2)$$

where \mathbf{J} is the Jacobian of \vec{F} . Then, by the implicit function theorem there exists $\vec{\phi}(s) : \mathbb{R} \rightarrow \mathbb{R}^{n+1}$ such that $\vec{F}(\vec{\phi}(s)) = \vec{0}$. In other words, paths or loops in \mathcal{S} implicitly defined by 5.1 can be parameterized with respect to s and, thus, $\vec{y}(s) \equiv \vec{\phi}(s)$. Differentiating 5.1 with respect to s , we arrive at the following Initial Value Problem (IVP):

$$\mathbf{J}(\vec{y})\dot{\vec{y}} = 0 \quad (5.3a)$$

$$\vec{y}(0) = \vec{y}_0, \quad \vec{F}(\vec{y}_0) = \vec{0} \quad (5.3b)$$

where (\cdot) denotes differentiation with respect to s and $\dot{\vec{y}}$ is the tangent vector to the solution curve at $\vec{y}(s)$.

The implicit function theorem forms the basis of numerical continuation[48, 95, 49, 57, 85,

¹In this chapter, we use $\vec{y} \in \mathbb{R}^{n+1}$ to denote any solution vector to an undetermined system $\vec{F} = \vec{0}$ and should not be confused with the internal field vector \vec{y}_i introduced in Chapter 2, Sec. 2.4.

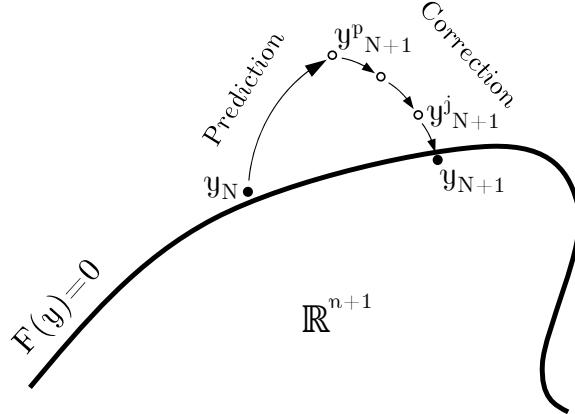


Fig. 5.1. Advancing from step N to step $N + 1$ using a PC approach.

[96, 50], which aim to track solution sets in \mathcal{S} . The idea is to generate a sequence of points $\{\vec{y}_1, \vec{y}_2, \dots\}$, with $\vec{F}(\vec{y}_i) \approx \vec{0}$ by solving 5.3 in a step-by-step fashion using a Predictor-Corrector (PC) scheme. The orientation of tracking is chosen by assigning a sign to the tangent vector and the positive orientation is characterized by condition $\det([\frac{J}{\dot{y}}]) > 0$. If s is assumed to be the arc-length of the solution curve, then $\dot{\vec{y}}(s)$ is a unit tangent vector and 5.3 is supplemented by condition $\|\dot{\vec{y}}(s)\|_2 = 1$, where $\|\cdot\|_2$ denotes the Euclidian norm.

Different initial points \vec{y}_0 might lead to a different path(loop) to be tracked and each point generated by PC methods is considered to be in the proximity of the solution curve. As shown in 5.1, to advance from the current step N^2 with known solution \vec{y}_N to step $N + 1$, a prediction for \vec{y}_{N+1} is first made, \vec{y}_{N+1}^p , followed by a correction phase whereby the initial prediction is iteratively improved, usually by employing a Newton type iteration to enforce condition 5.1. Satisfaction of this condition is usually tested using an appropriate norm and a prescribed tolerance.

Assuming that the path is parameterized by s so that $\vec{y} = \vec{y}(s)$, these two phases can be succinctly stated as follows:

1. Prediction: $\vec{y}_{N+1}^p = \vec{G}(s_N + \Delta s)$
2. Correction: $\vec{y}_{N+1}^{j+1} = \vec{y}_{N+1}^j + \mathbf{B}^j \vec{F}(\vec{x}_{N+1}^j), j = 0, 1, \dots, j_{max}, \vec{y}_{N+1}^0 = \vec{y}_{N+1}^p$

²The use of N to designate the step counter in this chapter should not be confused with the same symbol used in Chapter 2 to represent the axial stress resultant in a cross-section.

Both \vec{G} and \mathbf{B} depend on the predictor and corrector algorithms used. For example, if the Euler predictor is used, then $\vec{G} = \vec{y}(s)$ and $\vec{G}(s_N + \Delta s) \approx \vec{y}_N + \Delta s \dot{\vec{y}}_N$ around \vec{y}_N . In addition, if $\mathbf{B} = \mathbf{J}^T(\mathbf{J}\mathbf{J}^T)^{-1}$, then the correction phase arrives at the least squares solution for \vec{y}_{N+1} , with \mathbf{B} being the Moore-Penrose inverse of \mathbf{J} . In general, the determining characteristics for a PC algorithm are:

- Choice of parameterization
- Type of predictor
- Type of corrector
- Determination and control of step-length Δs

While the most computationally costly part in PC methods is the correction phase, efficient step-length adjustment schemes in conjunction with fast and reliable prediction phases can also contribute significantly to the performance of the PC algorithm.

Predictors can be categorized in three families:

1. Predictors based on functional extrapolation
2. Predictors based on derivative information of \vec{F}
3. A combination of 1 and 2

As an example of type 1 prediction, a Lagrange interpolating polynomial, $L(s)$, can be used to extrapolate from a set of the k most recent solution points $\{\vec{y}_{N-k+1}, \dots, \vec{y}_N\}$ such that $\vec{y}_{N+1}^p = L(s_N + \Delta s)$ [95, 97]. The Euler predictor using the tangent $\dot{\vec{y}}_N$ as a direction is the simplest case of the second type. Finally, a case of type 3 predictors are those derived from interpolations that utilize derivative information, such as the Hermite interpolating polynomials. In 5.2 the two generic cases of predictors are illustrated.

In general, predictors based on extrapolation of an interpolating function tend to be sensitive to variations in accuracy of the computed set $\{\vec{y}_i\}_{i=1}^N$ and although they are computationally inexpensive, the quality of prediction worsens as the step-length becomes larger. For that reason, predictors of order higher than two are rarely used[98, 99]. On the other side, type 2 or

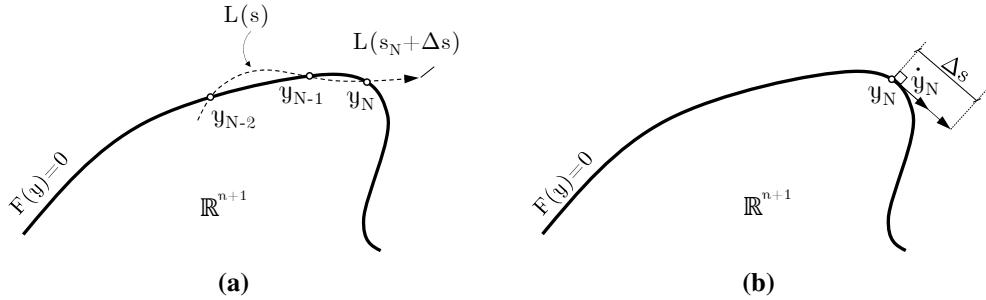


Fig. 5.2. Two types of predictors: (a) interpolatory and (b) Euler predictor.

3 predictors[100, 101, 102, 103, 104, 97, 105], tend to yield better approximations for larger step-lengths but are considerably more computationally expensive. Tangent vectors need to be extracted from the nullspace of \mathbf{J} at the beginning of each step and the determination of higher order derivatives require additional linear systems to be solved. Alternatively, finite difference schemes can be employed to approximate higher order quantities, but these too require additional functional evaluations. An alternative way to categorize predictors is based on whether they rely solely on information at \vec{y}_N or incorporate previous steps as well[106]. Since the predictor developed in this chapter is by construction a multistep formulation, the categorization presented here will help highlight type 1 and type 2 predictor variants that arise from the proposed WLS scheme.

5.3 Weighted least squares predictor

In this section we formulate the Weighted Least Squares (WLS) predictor. Let $k, m \in \mathbb{Z}^+, k > m$, be parameters that designate the last k steps, including the current step, N , and the degree of polynomial space, $\mathcal{P}_m(\mathbb{R}_{\geq 0}, \mathbb{R})$, respectively. We assume convergence has been achieved at step N and all relevant quantities for the last k steps are known. For ease of notation, we restrict our presentation to a particular component of $\vec{y} = [y_1 \cdots y_l \cdots y_{n+1}]^T$, y_l , and omit the index from derivations. Therefore, in what follows, quantities $y_N, y_{N-1}\mathbb{R}$ are to be understood as the l -th components of $\vec{y} \in \mathbb{R}^{n+1}$ at steps N and $N - 1$ respectively unless explicitly stated otherwise.

Let $\bar{y}(s) \in \mathcal{P}_m$ be the fitted polynomial at the last k points $\{y_i\}_{i=N-k+1}^N$ using a WLS approach, where $y_i = y(s_i)$, such that:

$$\bar{y}(s) = b_0 + b_1 s + \cdots + b_m s^m = \vec{p}^T \vec{b} \quad (5.4)$$

where $\vec{p} = [1 \ s \ \dots \ s^m]^T$ and $\vec{b} = [b_0 \ b_1 \ \dots \ b_m]^T$ are the polynomial basis and coefficient vectors respectively. The “energy” function associated with the WLS fitting in this case is:

$$J = \sum_{\substack{j= \\ N-k+1}}^N w(s_j) \left[\bar{y}(s_j) - y_j \right]^2 \quad (5.5)$$

where $w(s)$ is the weight function, unspecified for now. Minimization of J with respect to \vec{b} yields the so-called normal equations for the least squares solution:

$$\mathbf{A} \vec{b} = \vec{f} \quad (5.6)$$

with 5.4 substituted in 5.5. The moment matrix \mathbf{A} and “force” vector \vec{f} are given by:

$$\mathbf{A} = \sum_{\substack{j= \\ N-k+1}}^N w_j \vec{p}(s_j) \vec{p}(s_j)^T \quad (5.7a)$$

$$\vec{f} = \sum_{\substack{j= \\ N-k+1}}^N w_j y_j \vec{p}(s_j) \quad (5.7b)$$

Since the path is parameterized by the arc-length s , which is taken here as the independent parameter of the WLS fitting, a stabilization measure is required for the case that s becomes excessively large. In such a scenario, the moment matrix \mathbf{A} becomes ill-conditioned. To deal with this issue, we transform the polynomial basis and work instead in $\mathcal{P}_m([-1, 1], \mathbb{R})$. That is:

$$\bar{y}(s) = \vec{c}^T \vec{\rho} \quad (5.8)$$

where now $\vec{\rho} = [\rho_0 \ \rho_1 \ \dots \ \rho_m]$ is the coefficient vector for the transformed basis and \vec{c} is:

$$\vec{c} = \left[1 \ \frac{s - s_r}{h} \ \left(\frac{s - s_r}{h} \right)^2 \ \dots \ \left(\frac{s - s_r}{h} \right)^m \right]^T \quad (5.9)$$

with $s_r = (s_N + s_{N-k+1})/2$, $h = (s_N - s_{N-k+1})/2$ and $s \in [s_{N-k+1}, s_N]$. The transformation between the basis vectors \vec{p} and \vec{c} is carried out by operator Φ :

$$\vec{c} = \Phi \vec{p} \quad (5.10)$$

where

$$\Phi = \begin{bmatrix} 1 & 0 & \dots & 0 & 0 \\ -\frac{s_r}{h} & \frac{1}{h} & \dots & 0 & 0 \\ \vdots & \vdots & \ddots & 0 & 0 \\ \frac{(-s_r)^m}{h^m} & \binom{m}{m-1} \frac{(-s_r)^{m-1}}{h^m} & \dots & -\binom{m}{1} \frac{s_r}{h^m} & \frac{1}{h^m} \end{bmatrix} \quad (5.11)$$

Similarly, for the coefficient vectors we have:

$$\vec{b} = \Phi^T \vec{\rho} \quad (5.12)$$

Using 5.8 and $\vec{p} = \Phi^{-1} \vec{c}$ from 5.10, the normal equations in the trasnformed basis are:

$$\mathbf{A}_c \vec{\rho} = \vec{f}_c \quad (5.13)$$

where \mathbf{A}_c , \vec{f}_c are determined from 5.7 with $\vec{p}(s_j) \rightarrow \vec{c}(s_j)$. Using 5.4, 5.12, and that $\vec{\rho} = \mathbf{A}^{-1} \vec{f}_c$ from 5.13, the stabilized weighted least squares fit for component y in the original space is given by:

$$\bar{y}(s) = \vec{p}(s)^T \vec{b} \quad (5.14)$$

where

$$\vec{b} = \Phi^T \mathbf{A}_c^{-1} \vec{f}_c \quad (5.15)$$

Notice that \mathbf{A}_c is common for all components y_l , $l = 1, \dots, n+1$ in \vec{y} and only the forcing the “force” vector \vec{f}_c has to be evaluated for each component. That is, only the coefficient vector is component dependent: $\vec{b} \rightarrow \vec{b}(y_{N-k+1}, \dots, y_N)$.

5.3.1 Matrix formulation of WLS approximation

We now derive the matrix relations of the WLS process, which incorporate all components. This can be represented succinctly by collecting all the involved quantities corresponding to different components of \vec{y} into matrices. The local approximation to solution curve $\vec{y}(s)$ is as follows:

$$\vec{y} \approx \bar{y}(s) = \begin{bmatrix} \bar{y}_1(s) \\ \bar{y}_2(s) \\ \vdots \\ \bar{y}_{n+1}(s) \end{bmatrix}$$

which holds for $s \in [s_{N-k+1}, s_N]$ and all solution points corresponding to that interval³. Let \mathbf{P} be defined as follows:

$$\mathbf{P} = \begin{bmatrix} \vec{p}(s_{N-k+1}) & \dots & \vec{p}(s_N) \end{bmatrix} = \begin{bmatrix} 1 & \dots & 1 \\ s_{N-k+1} & \dots & s_N \\ \vdots & \ddots & \vdots \\ s_{N-k+1}^m & \dots & s_N^m \end{bmatrix} \quad (5.16)$$

with its counterpart in the transformed basis given by

$$\mathcal{C} = \Phi \mathbf{P} \quad (5.17)$$

where \mathcal{C} is defined in accordance with \mathbf{P} : $\mathcal{C} = \begin{bmatrix} \vec{c}(s_{N-k+1}) & \dots & \vec{c}(s_N) \end{bmatrix}$. We also define $\mathbf{W} = \text{diag}(w_{N-k+1}, \dots, w_N)$. From 5.14, 5.7b and 5.10, the coefficient vector for component l

³Note that $\bar{y}(s)$ is a vector. It denotes an approximation to vector \vec{y} . The overhead arrow is omitted in order to avoid excessive use of notation.

in $\bar{\vec{y}}$, \vec{b}_l , can be expressed in the following way:

$$\vec{b}_l = \Phi^T \mathbf{A}_c^{-1} \mathcal{C} \mathbf{W} \begin{bmatrix} y_{N-k+1,l} \\ y_{N-k+2,l} \\ \vdots \\ y_{N,l} \end{bmatrix}_{(k \times 1)} \quad (5.18)$$

By extension, a concise expression for $\bar{\vec{y}}$, whose component form is 5.14, is then:

$$\bar{\vec{y}} = \vec{p}(s)^T \mathcal{B} \quad (5.19)$$

where

$$\mathcal{B} = \Phi^T \mathbf{A}_c^{-1} \mathcal{C} \mathbf{W} \mathbf{Y} \quad (5.20)$$

and \mathbf{Y} is a $k \times (n + 1)$ matrix whose l -th column contains the values of the l -th component of \vec{y} at the last k points supplied to the algorithm (see Eq. (5.18)):

$$\mathbf{Y} = \begin{bmatrix} y_{N-k+1,1} & y_{N-k+1,2} & \dots & y_{N-k+1,n+1} \\ y_{N-k+2,1} & y_{N-k+2,2} & \dots & y_{N-k+2,n+1} \\ \vdots & \vdots & \ddots & \vdots \\ y_{N,1} & y_{N,2} & \dots & y_{N,n+1} \end{bmatrix} \quad (5.21)$$

We observe from 5.19 (or 5.14) that derivatives up to order m can be computed directly by simple differentiation of the polynomial basis vector $\vec{p}(s)$:

Component form	Vector form
$\dot{y} = \dot{\vec{p}}(s)^T \vec{b}$,	$\dot{\vec{y}} = \dot{\vec{p}}(s)^T \mathcal{B}$
$\ddot{y} = \ddot{\vec{p}}(s)^T \vec{b}$,	$\ddot{\vec{y}} = \ddot{\vec{p}}(s)^T \mathcal{B}$
\vdots	\vdots

with

$$\begin{aligned}\dot{\vec{p}}(s) &= \begin{bmatrix} 0 & 1 & 2s & \dots & ms^{m-1} \end{bmatrix} \\ \ddot{\vec{p}}(s) &= \begin{bmatrix} 0 & 0 & 2 & \dots & m(m-1)s^{m-2} \end{bmatrix} \\ &\vdots\end{aligned}$$

5.3.2 WLS prediction variants

Given an increment in the arc-length Δs , a straightforward prediction scheme is given by extrapolating from the fitted curve, $\bar{\vec{y}}$ at s_N :

$$\vec{y}^p = \bar{\vec{y}}(s_N + \widetilde{\Delta s}) \quad (5.22)$$

where $\widetilde{\Delta s}$ is introduced to account for the fact that s is not actually the arc-length of the solution curve $\vec{y}(s)$ and use of Δs can sometimes lead to unpredictable behavior. A stabilization procedure for is presented in the next part of this section.

We will refer to prediction (5.22) as Weighted Least Squares Extrapolation (WLSE) predictor. A second approach is to derive the prediction direction from first order information of the local WLS curve, which is trivial to retrieve. We propose the following general formula:

$$\vec{y}^p = \vec{y}_N + \Delta s \vec{t}_z \quad (5.23)$$

where \vec{t}_z is a unit vector determined as follows:

$$\vec{t}_z = \frac{\dot{\vec{y}}(s_N + z\widetilde{\Delta s})}{\|\dot{\vec{y}}(s_N + z\widetilde{\Delta s})\|_2} \quad (5.24)$$

with $\dot{\vec{y}}(s) = \dot{\vec{p}}(s)^T \mathbf{B}$ and $z \in [0, 1]$. For $z \neq 0$, the direction is derived by extracting the tangent of the fitted curve ahead of the current solution point, whereas for $z = 0$ the normalized tangent

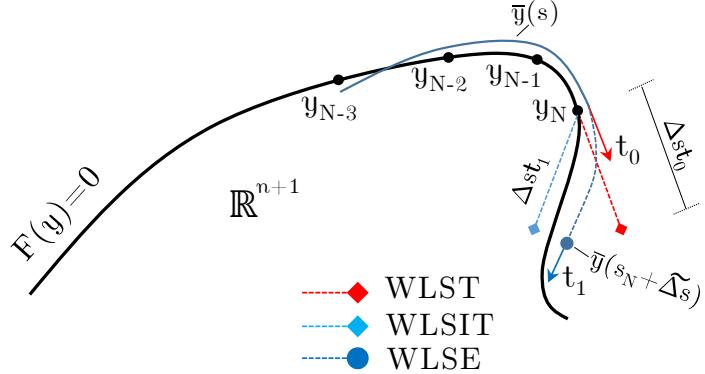


Fig. 5.3. Geometric interpretation of predictors (i)WLSE (blue), (ii) WLSIT (light blue, $z = 1$), and (iii) WLST (red, $z = 0$).

of $\bar{y}(s_N)$ is used instead and this yields a WLS approximation to the standard Euler predictor. Normalization is required because the $\dot{\bar{y}}$ is not a unit vector. When $z > 0$, will refer to this prediction scheme as the Weighted Least Squares Implicit Tangent (WLSIT) predictor. In the case that $z = 0$, we term it Weighted Least Squares Tangent (WLST) predictor. Notice that, in any case, WLSIT involves extrapolation and, therefore, the steplength adjustment procedure for $\widetilde{\Delta s}$ is required here as well. For illustrative purposes, the predictors WLSE, WLST and WLSIT with $z = 1$ are depicted in 5.3.

5.3.3 Arc-length stabilization

Given a step-length Δs , the WLSE predictor estimates the next solution point using 5.22. To determine $\widetilde{\Delta s}$ given Δs , we impose condition 5.25:

$$\|\vec{y}^p - \vec{y}_N\|_2^2 - \Delta s^2 = 0 \quad (5.25)$$

where in place of \vec{y}^p (see 5.22) we use its first order approximation:

$$\vec{y}^p \approx \bar{y}(s_N) + \dot{\bar{y}}(s_N)\widetilde{\Delta s} \quad (5.26)$$

Let $\vec{y}_0 = \vec{y}_N$, $\bar{y}_0 = \bar{y}(s_N) = \vec{p}(s_N)^T \mathcal{B}$ and $\dot{\bar{y}}_0 = \dot{\bar{y}}(s_N) = \dot{\vec{p}}(s_N)^T \mathcal{B}$. Then, substituting 5.26 into 5.25 yields the following quadratic equation to be solved for $\widetilde{\Delta s}$:

$$\|\dot{\bar{y}}_0\|_2^2 (\widetilde{\Delta s})^2 + 2(\bar{y}_0 - \bar{y}_0)^T \dot{\bar{y}}_0 \widetilde{\Delta s} + \|\bar{y}_0 - \bar{y}_0\|_2^2 - \Delta s^2 = 0 \quad (5.27)$$

This leads to two roots, of which the smallest positive one is chosen:

$$\widetilde{\Delta s}_{1,2} = \frac{-2(\bar{y}_0 - \bar{y}_0)^T \dot{\bar{y}}_0 \pm \sqrt{\Delta}}{2\|\dot{\bar{y}}_0\|_2^2} \quad (5.28)$$

where the discriminant is $\Delta = 4\|\bar{y}_0 - \bar{y}_0\|_{\mathbf{N}}^2 - 4\|\dot{\bar{y}}_0\|_2^2(\|\bar{y}_0 - \bar{y}_0\|_2^2 - \Delta s^2)$, with $\mathbf{N} = \dot{\bar{y}}_0 \dot{\bar{y}}_0^T$.

Condition $\Delta > 0$ leads to the following requirement for the user-defined step-length Δs if the WLSE predictor is used:

$$\Delta s > \sqrt{\|\bar{y}_0 - \bar{y}_0\|_2^2 - \frac{\|\bar{y}_0 - \bar{y}_0\|_{\mathbf{N}}^2}{\|\dot{\bar{y}}_0\|_2^2}} \quad (5.29)$$

5.3.4 The weight function

We now discuss the weighting function $w(s)$. The reasoning for introducing a weighting in the least squares fitting is as follows: since with the “exact” tangent predictor the direction is determined by the neighborhood of the current solution \vec{y}_N in the differential sense, we need to assign different weights to the preceding $k - 1$ solution points we provide for the least squares fitting. A reasonable option for $w(s)$, which is depicted in Fig. ??, is a generalization of the unit ramp function:

$$w(s) = \alpha + \frac{s - s_{N-k+1}}{s_N - s_{N-k+1}}(1 - \alpha), \quad \alpha \in [0, 1] \quad (5.30)$$

By choosing $\alpha = 0$, the unit ramp function is obtained, while for $\alpha = 1$, all solution points are assigned unit weight, $w(s) = 1$, with $s \in [s_{N-k+1}, s_N]$. With this weighting plan, it is ensured points closer to the current solution are assigned higher importance, with \vec{y}_N having the highest

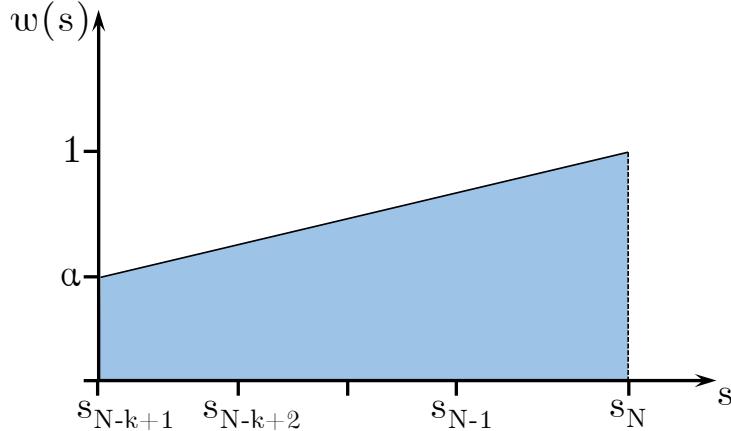


Fig. 5.4. Weight function $w(s)$ of the WLS predictor.

weight by construction. If the unit ramp is used, it is crucial that the implementation ensures there are enough points for the least squares algorithm to commence. This is because the point that corresponds to step $N - k + 1$ is now assigned a zero weight and, thus, it doesn't contribute in the moment matrices \mathbf{A} , \mathbf{A}_c (see Eqs. (5.5), (5.7a)). This issue arises when the minimum points for the desired degree of approximation are supplied.

5.4 Implementation

5.4.1 Algorithm

We outline the basic steps in the implementation of the WLS predictor in pseudocode format. We assume all quantities at the current step N are known.

As can be seen from 1, the WLS process requires the polynomial degree m we use for the fitting, the number of solution points k we take into account starting from the current one at step N , along with the curve lengths at those points. We also need to specify values for parameter α (see 5.30) and z , in case the WLSIT predictor is selected. Finally, a step-length, Δs , also needs to be provided. The most expensive action is the computation of the coefficient matrix \mathcal{B} . More specifically, the number of floating point operations is commensurate to the term $2(m + 1)k(n + 1)$ or, simply, $2(m + 1)kn$, where n is the size of \vec{y} after ignoring one component. In other words, it is of $\mathcal{O}(n)$ (time) complexity. Coupled with a degree control

Algorithm 1: Weighted Least Squares Predictor pseudocode

Input: $m, k, \{y_i\}_{i=N-k+1}^N, \{s_i\}_{i=N-k+1}^N, \Delta s, \alpha, z$
Determine Φ, \mathbf{Y} and initialize $\mathbf{A}_c, \mathbf{W}, \mathcal{C}$

for $i = k, k-1, \dots, 1$ **do**

- $s_i \leftarrow s_{N-i+1}$
- Get shifted basis at s_i , $\vec{c}(s_i)$, and add to i -th column in \mathcal{C}
- Determine weight, w_i , and add it to component (i, i) in \mathbf{W}
- $\mathbf{A}_c \leftarrow \mathbf{A}_c + w_i \vec{c}(s_i) \vec{c}(s_i)^T$

end for

$\mathcal{B} \leftarrow \Phi^T \mathbf{A}_c^{-1} \mathcal{C} \mathbf{W} \mathbf{Y}$

Determine $\widetilde{\Delta s}$ *// Step-length stabilization*

// WLSE

$s \leftarrow s_N + \widetilde{\Delta s}$

Get polynomial basis at s , $\mathbf{p}(s)$

$\vec{y}_{N+1}^p \leftarrow \vec{p}(s)^T \mathcal{B}$

// WLSIT

$s \leftarrow s_N + z \widetilde{\Delta s}$

Get first derivative of polynomial basis at s , $\dot{\vec{p}}(s)$

$\vec{v} \leftarrow \dot{\vec{p}}(s)^T \mathcal{B}$

$\vec{t}_z \leftarrow \vec{v} / \|\vec{v}\|_2$

$\vec{y}_{N+1}^p \leftarrow \vec{y}_N + \Delta s \vec{t}_z$

return \vec{y}_{N+1}^p

check, whereby m can be adjusted based on local properties of the path tracked so far, we expect high values for the coefficient only in highly nonlinear regions.

5.4.2 Arc-Length measurement

Prediction schemes that rely on some form of extrapolation require measurement of the current curve length, s , as the solution process progresses. The most direct and easy way to compute the curve length at discrete points is by summation of lengths of the line segments connecting them. After convergence is achieved at step N , the length of the segment between \vec{y}_{N-1} and \vec{y}_N is approximated as follows:

$$s_N = s_{N-1} + Ds \quad (5.31)$$

where

$$Ds = \|\vec{y}_N - \vec{y}_{N-1}\|_2 \quad (5.32)$$

If 5.1 is the finite-dimensional approximation of some operator equation, then it is recommended[107] that we account for the fact that, as is customary, the first n components are members of some function space (e.g. a Banach space) and, therefore, 5.32 should in some sense approximate the underlying L^2 norm:

$$Ds = \|\vec{y}_N - \vec{y}_{N-1}\|_{L^2} = \sqrt{h_m^d \sum_{i=1}^n (y_{i,N} - y_{i,N-1})^2 + (y_{n+1,N} - y_{n+1,N-1})^2} \quad (5.33)$$

where the $n+1$ component in \vec{y} is considered as an additional parameter and h_m, d are the mesh spacing and the spatial dimension of the problem respectively.

5.4.3 Discussion

It is clear that, in general, $\bar{y}(s_j) \neq \vec{y}_j$ for $j = N - k + 1, \dots, N$, which is expected since this is not, necessarily, an interpolatory approximation. The interpolatory condition can be retained if the relation between the number of solution points supplied, k , and the degree of the polynomial space, m , is $k = m + 1$. Furthermore, the degree of approximation is uncoupled from the number of points supplied, while in conventional interpolatory schemes this is not the case. We suspect this is one of the reasons some researchers claim higher order predictors based on extrapolation generally behave poorer than the secant predictor[98, 99]. In addition, the WLS predictor does not rely on derivative information of \vec{F} and, therefore, its performance in terms of computational cost is independent of the structure of the Jacobian. This is particularly important for problems without properties like sparsity or bandedness to exploit numerically. For the WLSIT variant, we instead use the local WLS approximation to extract derivatives when and if necessary. For step-length, Δs , of reasonable magnitude, the local WLS fitting is able to give a better approximation of the local geometry, especially closer to sudden changes

in curvature of the solution curve. This is precisely due to the fact that interpolation is not enforced. Furthermore, since extraction of the actual tangent vector from the nullspace of the Jacobian is not required, WLS predictions can couple well with correctors that also bypass Jacobian evaluations, such as nonlinear conjugate gradient corrections[106]. However, in that case, secondary branches due to bifurcation will have to be tracked using a (global) perturbation technique. Since differentiation of the WLS curve is trivial, the tangent vectors to the solution curve at solution points can be approximated by the normalized tangents to the local WLS curve.

Another attractive feature of the proposed formulation is that m and k can be adjusted during the continuation process. This provides additional options as far as calibration of the algorithm near turning points is concerned. That is, instead of reducing the step-length, m or k can instead be increased in order to represent the local curvature better. In contrast, when the curvature is small, switching to a linear basis ($m = 1$) is usually better. Furthermore, for step-lengths not excessively large, the tangent and curvature on the local WLS curve can be determined analytically and treated as approximations to those of the actual solution curve, provided the polynomial basis is of adequate order. In addition, parameter z involved in the WLSIT variant can also be varied depending on the local geometry. It has been observed that, if only \vec{y}_N for the points supplied to WLSIT predictor is located past a turning point, using $z = 1$, or even $z > 1$ yields much better predictions compared to both WLSE and WLST. Conversely, if \vec{y}_{N-k+1} lies before a turning point and the rest are located beyond it, then WLST gives better predictions and in most cases, using tangents at $s < s_N$ yield even better ones (case $z < 0$).

Since at least 2 points need to be passed to the WLS predictor, the first 2 steps of the continuation are carried out using the tangent predictor or an approximation of it. In general, at any step there need to be at least k solution points available to pass to the WLS predictor. Moreover, prediction schemes based on extrapolation tend to be sensitive to measurement of the arc-length of the solution curve (see 5.31) and the approximation quality of solution points[108, 109]. Although this can affect the prediction quality of the WLSE variant for larger prescribed step-lengths, the WLSIT is generally more stable due to normalization imposed on the tangent

to the local WLS curve.

5.5 Numerical Examples

We now illustrate the performance of the proposed predictor formulation with five examples.

We use both WLS variants described in 5.3, where we also showcase results for both the WLST and WLSIT predictors ($z = 0$ and $z = 1$ respectively). For each WLS predictor we also include a selection of cases with regards to parameters (m, k) , the order of polynomial basis and the number of points passed to the algorithm. For comparison purposes, we also include results using i) Euler predictor with the exact tangent vector and ii) two-step and three-step Adams-Bashforth predictors[108]. For reference, these predictors are given by the following formulas⁴:

- Euler predictor: $y^p = y_N + \Delta s \dot{y}_N$
- Two-step Adams-Bashforth (AB2): $y^p = y_N + \Delta s \frac{1}{2} (3\dot{y}_N - \dot{y}_{N-1})$
- Three-step Adams-Bashforth (AB3): $y^p = y_N + \Delta s \frac{1}{12} (23\dot{y}_N - 16\dot{y}_{N-1} + 5\dot{y}_{N-2})$

While step-length Δs control and adjustment is an important aspect of numerical continuation, its influence on the overall performance is not considered here and in each problem a constant step-length is prescribed. The results are listed in tables, where we use the following abbreviations:

- Ns Number of points computed on the solution curve (steps)
- Nf Number of function evaluations
- NI Number of solved linear systems
- Ni Maximum number of iterations that occurred in a step

For each problem and for each prediction scheme, we perform two analyses: one with step-length that is considered small and one that is large. The specific values for Δs considered small or large are problem-dependent. In all problems, however, we considered the $n + 1$ -th variable

⁴Expressions for a one-dimensional problem.

to be the varying “homotopy” parameter, which we designated t . In that sense, the system in 5.1 can be recast as follows⁵

$$\vec{\mathbf{F}}(\vec{x}, t) = \vec{0} \quad (5.34)$$

and its Jacobian, considered to have maximal rank at all points, is $\mathbf{J} = [\begin{bmatrix} \mathbf{F}_x & \mathbf{F}_t \end{bmatrix}]$. Data such as tolerance ϵ_{tol} , initial value for t , t_0 , and its maximum prescribed value, t_{max} are included in the table caption for each problem.

5.5.1 Numerical Example 1

This example is taken from Kubíček[110], where the dependence of a system solution to a varying parameter is explored. The nonlinear system in question is $\vec{\mathbf{F}}(\vec{x}, t) = \vec{0}$, $F : \mathbb{R}^4 \times [0, 1] \rightarrow \mathbb{R}^4$, with $\vec{\mathbf{F}} = [f_1 \ f_2 \ f_3 \ f_4]^T$ and:

$$\begin{aligned} f_1 &= t(1 - x_3)\exp\left(\frac{10x_1}{1 + 10x_1/\gamma}\right) - x_3 \\ f_2 &= tB(1 - x_3)\exp\left(\frac{10x_1}{1 + 10x_1/\gamma}\right) + \beta_1\theta_{c1} - 10(1 + \beta_1)x_1 \\ f_3 &= x_3 - x_4 + t(1 - x_4)\exp\left(\frac{10x_2}{1 + 10x_2/\gamma}\right) \\ f_4 &= 10x_1 - 10(1 + \beta_2)x_2 + tB(1 - x_4)\exp\left(\frac{10x_2}{1 + 10x_2/\gamma}\right) + \beta_2\theta_{c2} \end{aligned}$$

and values for parameters γ , B , β_1 , β_2 , θ_{c1} , θ_{c2} are taken from the aforementioned investigation. The continuation process is initialized from point $x_0 = (0, 0, 0, 0)$ with corresponding $t_0 = 0$. Parameter α in the weight function (see 5.30) is taken equal to 0.2. The correction algorithm used here is the one proposed in Chapter 4, Sec. 4.4. The results for this problem are listed in 5.1.

⁵Again, the use of vector $\vec{x} \in \mathbb{R}^n$ should not be confused with the position vector of the neutral axis in the deformed configuration, as was the case in Chapter 2.

Table 5.1 Results for Example 1, $\epsilon_{tol} = 10^{-6}$, $t_0 = 0$, $t_{max} = 1$.

Predictor	(m, k)	$\Delta s = 0.05$				$\Delta s = 0.2$			
		Ns	Nf	Nl	Ni	Ns	Nf	Nl	Ni
Euler	-	99	413	434	3	-	-	-	-
WLSE	(2, 4)	99	428	460	3	21	177	270	30
	(3, 5)	99	424	450	3	29	180	242	14
WLST	(2, 4)	99	408	420	3	-	-	-	-
	(3, 5)	99	408	418	3	-	-	-	-
WLSIT	(2, 4)	99	429	462	3	17	135	202	30
	(3, 5)	99	428	458	3	-	-	-	-
AB2	-	99	403	414	3	-	-	-	-
AB3	-	99	404	416	3	-	-	-	-

5.5.2 Numerical Example 2

In this example we use a Newton homotopy imbedding to find all stationary points of the Him-melblau function:

$$f(x) = (x_1^2 + x_2 - 11)^2 + (x_1 + x_2^2 - 7)^2 \quad (5.35)$$

with $-5 \leq x_i \leq 5$, $i = 1, 2$. At a stationary point the gradient of f is zero:

$$\nabla f(x) = \begin{bmatrix} 2x_1^3 + 2x_1x_2 - 21x_1 + x_2^2 - 7 \\ x_1^2 + 2x_1x_2 + 2x_2^3 - 13x_2 - 11 \end{bmatrix} = 0 \quad (5.36)$$

and numerical continuation is used on the following system:

$$\vec{F}(x, t) = \nabla f(x) - (1-t)\nabla f_0 \quad (5.37)$$

where ∇f_0 is the value of the gradient at the initial point provided, x_0 , which we take to be $x_0 = (4.81, -4.81)$. In artificial imbeddings such as the Newton homotopy 5.37, a root for the target system is found each time t attains the value 1, assuming no backtracking happens. Eq. 5.37 locates 9 roots for 5.36 in the prescribed domain, provided in 5.3, and we examine the how the listed prediction schemes performed in finding all of them. The results are listed in 5.2,

Table 5.2 Results for Example 2, $\epsilon_{tol} = 10^{-6}$, $t_0 = 0$, $t_{max} = 1.2$.

Predictor	(m, k)	$\Delta s = 0.1, \alpha = 0.1$				$\Delta s = 1.0, \alpha = 0.1$				$\Delta s = 1, 0, \alpha = 0.5$			
		Ns	Nf	Nl	Ni	Ns	Nf	Nl	Ni	Ns	Nf	Nl	Ni
Euler	-	304	1210	1209	2	-	-	-	-	-	-	-	-
WLSE	(2, 3)	304	1207	1201	2	-	-	-	-	-	-	-	-
	(2, 4)	304	1211	1207	2	49	323	451	30	-	-	-	-
	(2, 5)	304	1211	1207	2	-	-	-	-	-	-	-	-
	(3, 5)	304	1142	1067	2	32	157	185	4	32	156	183	4
WLST	(2, 3)	304	1209	1205	2	-	-	-	-	-	-	-	-
	(2, 4)	304	1209	1203	2	49	325	455	30	65	436	613	30
	(2, 5)	304	1213	1209	2	-	-	-	-	-	-	-	-
	(3, 5)	304	1212	1207	2	32	181	233	30	-	-	-	-
WLSIT	(2, 3)	304	1209	1205	2	-	-	-	-	-	-	-	-
	(2, 4)	304	1211	1207	2	-	-	-	-	-	-	-	-
	(2, 5)	304	1210	1203	2	-	-	-	-	-	-	-	-
	(3, 5)	304	1211	1205	2	32	153	177	4	32	153	177	4
AB2	-	304	1200	1189	2	32	145	167	4	32	145	167	4
AB3	-	304	1086	961	2	32	143	163	4	32	143	163	4

where we used two different weighting parameters, α , for the larger step-length. Only analyses that i) found all 9 solutions and ii) converged to $t = t_{max} = 1.2$ were deemed successful. The same correction scheme was used here as in Example 1.

The cases WLSE(2, 4) and WLST(2, 4) with $a = 0.1$ and WLST(2, 4) $a = 0.5$, passed through all solutions but also backtracked before eventually converging. The case of WLSE(2, 5), $a = 0.5$, while it converges, it passes through only 5 solutions and, therefore, it is not considered a successful analysis. We note that, due to the excessively large step-length of $\Delta s = 1$ for this

Table 5.3 Stationary points of 5.35

Root	Coordinates			Coordinates		
	x_1	x_2	Root	x_1	x_2	
1	3.5856	-1.8546	6	-0.1282	-1.9697	
2	3.3844	0.0826	7	-3.7792	-3.2829	
3	3.0059	1.9860	8	-3.0733	-0.0819	
4	0.0867	2.8841	9	-2.8050	3.1311	
5	-0.2681	-0.9329				

problem, only the cubic WLS predictors manage to perform on par with the Adams-Bashforth ones.

5.5.3 Numerical Example 3

We consider the following fixed point problem[111, 112, 100]:

$$F_i(x, t) = x_i - \lambda \left[\cos\left(i \sum_{j=1}^{10} x_j\right) \right] = 0 \quad (5.38)$$

or $F(x, \lambda) = 0$, $F : \mathbb{R}^{11} \rightarrow \mathbb{R}^{10}$. This is solved using the Probability-one homotopy[113, 55] for system 5.38:

$$\vec{H}(\vec{x}, t) = t \vec{G}(\vec{x}) + (1-t)(\vec{x} - \vec{x}_0) \quad (5.39)$$

The resulting solution path exhibits more than 40 turning points with respect to t and is considered very difficult to solve. Starting from $(\vec{x}_0, t_0) = (\vec{0}, 0) \in \mathbb{R}^{11}$, we follow the path until the first intersection with hyperplane $t = 1$. The so-called Normal Flow corrector is used in this test case, where the Moore-Penrose inverse of the $n \times (n+1)$ Jacobian of 5.39 is used during the Newton iterations. The results are shown in 5.4, where now for both step-lengths the optimal performance of the WLS predictors with respect to variations in the weighting parameter α are listed. This is because even for the smaller Δs , convergence was difficult to achieve. For reference, in an investigation[112] of the same problem with the same arc-length parameterization but with step-length control, the smallest reported step-length was 0.000195 and on average, it was between 0.1 and 0.2. The solution path of 5.39, projected on the $t - x_1$ plane, is shown in 5.5.

Table 5.4 Results for Example 3, $\epsilon_{tol} = 10^{-6}$, $t_0 = 0$, $t_{max} = 1$, Normal Flow.

Predictor	(m, k)	$\Delta s = 0.15$					$\Delta s = 0.3$				
		α	Ns	Nf	Nl	Ni	α	Ns	Nf	Nl	Ni
Euler	-	-	-	-	-	-	-	449	2091	1196	26
WLSE	(2, 3)	-	1165	4920	2592	30	-	-	-	-	-
	(2, 4)	*	-	-	-	-	0.40	291	1337	756	24
	(2, 5)	1.00	554	2378	1270	12	0.30	189	942	564	28
	(3, 5)	0.20	472	2042	1098	20	1.00	289	1363	785	26
WLST	(2, 3)	-	-	-	-	-	-	576	2761	1611	26
	(2, 4)	0.50	598	2521	1326	14	0.65	326	1585	934	29
	(2, 5)	0.80	608	2591	1375	13	0.95	297	1460	866	26
	(3, 5)	0.40	625	2739	1489	21	0.90	305	1458	848	30
WLSIT	(2, 3)	-	-	-	-	-	-	-	-	-	-
	(2, 4)	0.05	592	2507	1324	10	0.35	294	1402	815	24
	(2, 5)	1.00	601	2564	1362	9	0.80	302	1447	843	22
	(3, 5)	*	-	-	-	-	*	-	-	-	-
AB2	-	-	758	3156	1643	11	-	337	1546	875	12
AB3	-	-	651	2788	1489	18	-	-	-	-	-

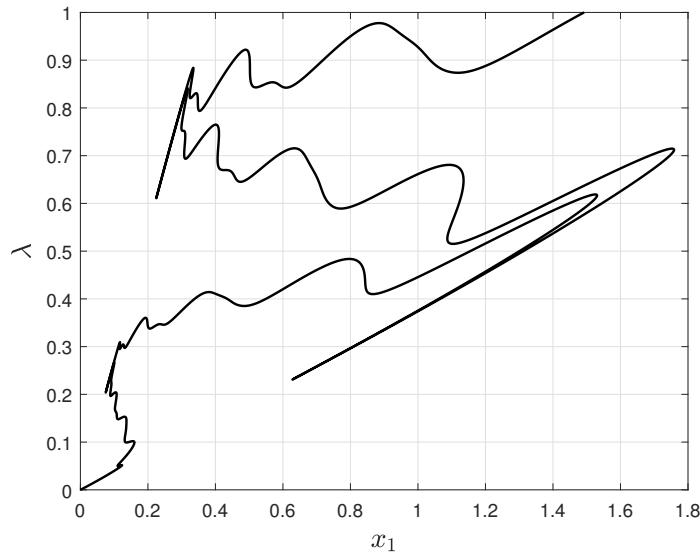


Fig. 5.5. Solution path of 5.39.

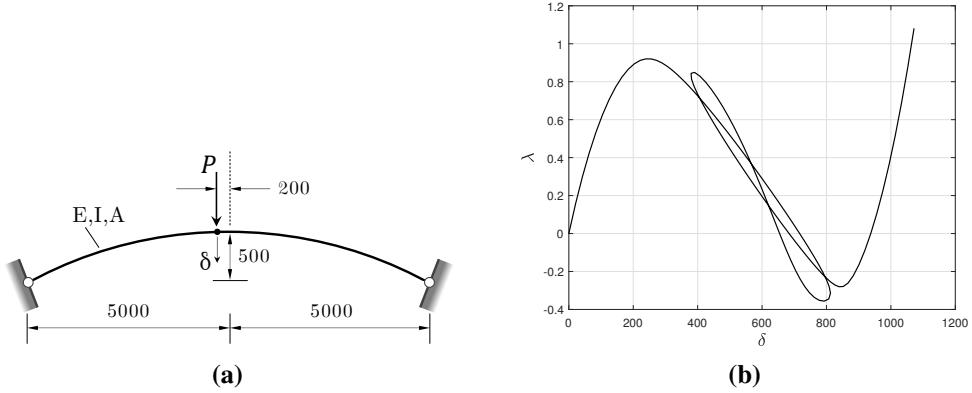


Fig. 5.6. (a) Shallow arch structure and loading, (b) equilibrium path.

Example 4

Here, we investigate the performance of the proposed predictors in a practical application pertaining to elastostatics. More specifically, we are concerned with the geometrically nonlinear elastic response of the double pinned shallow arch shown in Fig. 5.6a. The structure is discretized into 42 hybrid NLP finite elements and solved for the external load applied with a geometric imperfection, as shown in 5.6b. The geometry and material properties considered here are the same as in [114]. The parameterized equilibrium system has the following form:

$$\vec{R}(\vec{u}, t) = \vec{F}_{int}(\vec{u}) - t\vec{P} \quad (5.40)$$

where \vec{F}_{int} is the vector of internal forces at the nodes, $P = 1300$ is the prescribed external load and \vec{u} is the nodal displacement vector. There are a total of 43 nodes, with 41 nodes in the interior. The total number of primary degrees of freedom (problem dimension) is $n = 41 \cdot 3 + 2 = 125$. Thus, $R : \mathbb{R}^{126} \rightarrow \mathbb{R}^{125}$. The equilibrium path, shown in 5.6b, is tracked from the undeformed configuration ($0 \in \mathbb{R}^{126}$) up to the first full application of external load $P = 1300$, at which point, $t = 1$. The results are listed in 5.5, where $a = 0.2$ for all WLS prediction variants.

Table 5.5 Results for Example 4, $\epsilon_{tol} = 10^{-6}$, $t_0 = 0$, $t_{max} = 1$, $\alpha = 0.2$.

Predictor	(m, k)	$\Delta s = 100$				$\Delta s = 280$			
		Ns	Nf	NI	Ni	Ns	Nf	NI	Ni
Euler	-	99	311	523	4	69	258	448	5
WLSE	(2, 3)	99	314	432	4	35	161	256	21
	(2, 4)	99	317	439	4	35	150	235	9
	(2, 5)	99	322	450	4	35	150	235	9
	(3, 5)	99	316	438	4	-	-	-	-
WLST	(2, 3)	99	312	418	4	35	151	234	13
	(2, 4)	99	314	433	4	35	154	241	11
	(2, 5)	99	319	444	4	35	156	246	11
	(3, 5)	99	317	440	4	35	159	252	15
WLSIT	(2, 3)	99	319	442	4	35	170	273	25
	(2, 4)	99	320	445	4	35	157	248	13
	(2, 5)	99	324	454	4	35	154	243	9
	(3, 5)	-	-	-	-	-	-	-	-
AB2	-	99	309	520	4	46	175	305	11
AB3	-	99	306	514	4	49	188	327	12

5.6 Closing remarks

As can be seen in the results of the previous examples, the WLS variants offer comparable and can even outperform the Euler and the AB2 & AB3 higher order predictors, which employ the exact tangent. For larger steps in particular, we see that achieving reliable estimates of the next solution point can greatly enhance the performance of the continuation process. The proposed prediction schemes do so without utilizing expensive second order information but by reserving an additional portion of available memory, not large, for storing the a number of previously converged solution points. In practice, WLS prediction is unlikely to use a polynomial order higher than $m = 4$ and, thus, at each step, we can simply ensure that the last $k = 5$ solution points are available in memory.

Appendices

APPENDIX A

Explicit expressions of the formulation are provided in this Appendix. We note that operations involving the symbol ∇ map scalars to vectors and vectors to matrices. For example, if $f : \mathbb{R}^p \rightarrow \mathbb{R}$ and $\vec{v} : \mathbb{R}^p \rightarrow \mathbb{R}^q$, then $\nabla f : \mathbb{R}^p \rightarrow \mathbb{R}^p$ and $\nabla \vec{v} : \mathbb{R}^p \rightarrow \mathbb{R}^{p \times q}$.

The Lagrangian function is used here, as presented in Eq. (2.30):

$$\mathcal{L}(\vec{y}, \vec{d}_g, \vec{\lambda}) = U - W + \vec{\lambda}^T \vec{h}_g \quad (\text{A.1})$$

where it is also reminded that $\vec{y} = [\vec{y}_1^T \quad \vec{y}_2^T \quad \dots \quad \vec{y}_m^T]^T$, $\vec{y}_i = [\vec{q}_i \quad \phi_i]^T$ and $\vec{q}_i = [\epsilon_i \quad \gamma_i \quad \kappa_i]^T$. The number m designates the total number of integration points in the structure.

A.0.1 Gradient of strain energy

The gradient of U is expressed as:

$$\nabla U = \begin{bmatrix} \nabla_y U \\ \nabla_{d_g} U \\ \nabla_\lambda U \end{bmatrix}$$

The gradient with respect to the strain vector \vec{y} gives the stress resultants at all cross-sections and has the following form for a particular cross-section i :

$$\nabla_{y_i} U = \begin{bmatrix} w_i N_i \\ w_i V_i \\ w_i M_i \\ 0 \end{bmatrix} = w_i \begin{bmatrix} \vec{F}_{sec}^{(i)} \\ 0 \end{bmatrix} \quad (\text{A.2})$$

it is usually computed by numerical integration of Eqs. (2.55) over the cross-section height. In the case of elastic analysis, the above vector can be computed explicitly for each cross-section, since then $\vec{\mathbf{F}}_{sec}^{(i)} = \begin{bmatrix} EA\epsilon_i & k_s GA\gamma_i & EI\kappa_i \end{bmatrix}^T$.

The gradients of U with respect to both $\vec{\mathbf{d}}_g$ and $\vec{\boldsymbol{\lambda}}$ provide zero vectors:

$$\nabla_{\mathbf{d}_g} U = \vec{\mathbf{0}} \in \mathbb{R}^{3N} \quad \text{and} \quad \nabla_{\boldsymbol{\lambda}} U = \vec{\mathbf{0}} \in \mathbb{R}^s$$

where $s = m + 3n_{el}$, n_{el} being the number of elements. Finally, we then get:

$$\nabla U = \begin{bmatrix} \nabla_{\mathbf{y}} U \\ \vec{\mathbf{0}} \\ \vec{\mathbf{0}} \end{bmatrix} \quad (\text{A.3})$$

A.0.2 Gradient of constraints

We restate the constraints equations from Eqs. (2.32a)-(2.32b):

$$\vec{\mathbf{h}}_g = \begin{bmatrix} \vec{\mathbf{h}}_g^A \\ \vec{\mathbf{h}}_g^B \end{bmatrix} = \begin{bmatrix} \mathbf{V}_1 \vec{\mathbf{d}}_g - \boldsymbol{\Lambda}^T \left[\sum_{i=1}^n w_i \mathbf{R}_i(\vec{\mathbf{q}}_i + \vec{\mathbf{E}}_1) - \ell \vec{\mathbf{E}}_1 \right] \\ \vec{\boldsymbol{\phi}} - \mathbf{V}_2 \vec{\mathbf{d}}_g - \mathbf{T} \vec{\boldsymbol{\kappa}} \end{bmatrix}$$

For notational simplicity, we assume here only one element in the derivations that follow. When more elements are used, their corresponding constraint vectors are stacked in vector form as $\vec{\mathbf{h}} = \begin{bmatrix} \vec{\mathbf{h}}_g^{1T} & \vec{\mathbf{h}}_g^{2T} & \dots & \vec{\mathbf{h}}_g^{n_{el}T} \end{bmatrix}^T$. The gradient with respect to the strain vector $\vec{\mathbf{y}}_i$ at a particular cross-section is given by:

$$\nabla_{\mathbf{y}_i} \vec{\mathbf{h}}_g^A = \begin{bmatrix} \nabla_{\mathbf{q}_i} \vec{\mathbf{h}}_g^A & \frac{d\vec{\mathbf{h}}_g^A}{d\phi_i} \end{bmatrix} = -w_i \boldsymbol{\Lambda}^T \begin{bmatrix} \mathbf{R}_i & \mathbf{R}_i \mathbf{X}[\vec{\mathbf{q}}_i + \vec{\mathbf{E}}_1] \end{bmatrix}$$

where:

$$\frac{d\mathbf{R}_i}{d\phi_i} = \mathbf{R}_i \mathbf{X} \quad \text{and} \quad \mathbf{X} = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

The gradient of $\vec{\mathbf{h}}_g^B$ with respect to strain vector $\vec{\mathbf{y}}_i$, is:

$$\nabla_{\mathbf{y}_i} \vec{\mathbf{h}}_g^B = \left[\nabla_{\mathbf{q}_i} \vec{\mathbf{h}}_g^B \quad \frac{d\vec{\mathbf{h}}_g^B}{d\phi_i} \right] = \left[\vec{\mathbf{0}} \quad \vec{\mathbf{0}} \quad -\mathbf{T}\hat{\mathbf{E}}_i \quad \hat{\mathbf{E}}_i \right]$$

where $\hat{\mathbf{E}}_i \in \mathbb{R}^n$ is a unit vector with the i -th component equal to one. Collectively, we have:

$$\nabla_{\mathbf{y}_i} \vec{\mathbf{h}}_g = \begin{bmatrix} -w_i \Lambda^T \mathbf{R}_i & -w_i \Lambda^T \mathbf{R}_i \mathbf{X} [\vec{\mathbf{q}}_i + \vec{\mathbf{E}}_1] \\ [\vec{\mathbf{0}} \quad \vec{\mathbf{0}} \quad -\mathbf{T}\hat{\mathbf{E}}_i] & \hat{\mathbf{E}}_i \end{bmatrix} \quad (\text{A.4})$$

The gradient with respect to the displacement vector is derived in a straightforward manner by simple derivations:

$$\nabla_{\mathbf{d}_g} \vec{\mathbf{h}}_g = \begin{bmatrix} \mathbf{V}_1 \\ -\mathbf{V}_2 \end{bmatrix} \quad (\text{A.5})$$

$$\mathbf{V}_1 = \begin{bmatrix} -\mathbf{I} & \mathbf{I} \end{bmatrix} \quad , \quad \mathbf{V}_2 = \begin{bmatrix} \vec{\mathbf{0}} & \vec{\mathbf{0}} & \vec{\mathbf{1}} & \vec{\mathbf{0}} & \vec{\mathbf{0}} & \vec{\mathbf{0}} \end{bmatrix}$$

$$\mathbf{I} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad , \quad \vec{\mathbf{0}}, \vec{\mathbf{1}} \in \mathbb{R}^n, \quad \vec{\mathbf{0}} = \begin{bmatrix} 0 & \dots & 0 \end{bmatrix}^T, \quad \vec{\mathbf{1}} = \begin{bmatrix} 1 & \dots & 1 \end{bmatrix}^T$$

Finally, the gradient with respect to the vector $\vec{\lambda}$ gives a zero matrix:

$$\nabla_{\lambda} \vec{h}_g = \mathbf{0} \in \mathbb{R}^{(n+3) \times (n+3)} \quad (\text{A.6})$$

The total constraints gradient is then expressed in block matrix form as:

$$\nabla \vec{h}_g = \begin{bmatrix} \nabla_{y_1} \vec{h}_g & \nabla_{y_2} \vec{h}_g & \cdots & \nabla_{y_n} \vec{h}_g & \nabla_{d_g} \vec{h}_g & \nabla_{\lambda} \vec{h}_g \end{bmatrix}$$

Note that in the case of small displacements, only the expression $\nabla_{y_i} \vec{h}_g^A$ has to be modified, so that the strain-displacement relations reduce to the classical linear Timoshenko theory:

$$\nabla_{y_i} \vec{h}_g^A = -w_i \Lambda^T \begin{bmatrix} 1 & -\phi_i & 0 & -\gamma_i \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

with the vector of constraints \vec{h}_g^A being:

$$\vec{h}_g^A = \mathbf{V}_1 \vec{d}_g - \Lambda^T \left[\sum_{i=1}^n w_i [\hat{\mathbf{R}}_i (\vec{q}_i + \vec{E}_1) - \vec{g}_i] - \ell \vec{E}_1 \right] \quad (\text{A.7})$$

where:

$$\hat{\mathbf{R}}_i = \begin{bmatrix} 1 & -\phi_i & 0 \\ \phi_i & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad \vec{g}_i = \begin{bmatrix} 0 \\ \epsilon_i \phi_i \\ 0 \end{bmatrix}$$

The vector \vec{g}_i is added in order to eliminate the second order term $\epsilon_i \phi_i$ arising from the imposition of the small displacement assumption $\sin \phi_i \approx \phi_i, \cos \phi_i \approx 1$ in Eq. (2.32a).

A.0.3 Gradient of the potential energy

The potential energy due to external loads $\vec{\mathbf{P}}$ is:

$$W = \sum_{j=1}^N \vec{\mathbf{P}}_j^T \vec{\mathbf{d}}_{g,j} = \vec{\mathbf{d}}_g^T \vec{\mathbf{P}}$$

The gradient of the potential energy is, again, straightforward in its derivation as it only depends on the displacement vector:

$$\nabla W = \begin{bmatrix} \nabla_{\mathbf{y}} W \\ \nabla_{\mathbf{d}_g} W \\ \nabla_{\lambda} W \end{bmatrix}$$

with $\nabla_{\mathbf{y}} W = \vec{\mathbf{0}} \in \mathbb{R}^m$, $\nabla_{\mathbf{d}_g} W = \vec{\mathbf{P}}$ and $\nabla_{\lambda} W = \vec{\mathbf{0}} \in \mathbb{R}^{n+3}$ for one element. Thus, we have:

$$\nabla W = \begin{bmatrix} \vec{\mathbf{0}} \\ \vec{\mathbf{P}} \\ \vec{\mathbf{0}} \end{bmatrix} \quad (\text{A.8})$$

A.0.4 Gradient of the Lagrangian

We substitute in Eqs. (2.35) the explicit expressions derived here in the Appendix:

$$\nabla_{\mathbf{y}_i} \mathcal{L} = w_i \begin{bmatrix} \vec{\mathbf{F}}_{sec}^{(i)} \\ 0 \end{bmatrix} + \begin{bmatrix} -w_i \Lambda^T \mathbf{R}_i & -w_i \Lambda^T \mathbf{R}_i \mathbf{X} [\vec{\mathbf{q}}_i + \vec{\mathbf{E}}_1] \\ [\vec{\mathbf{0}} \quad \vec{\mathbf{0}} \quad -\mathbf{T} \hat{\mathbf{E}}_i] & \hat{\mathbf{E}}_i \end{bmatrix}^T \vec{\boldsymbol{\lambda}} \quad (\text{A.9})$$

$$\nabla_{\mathbf{d}_g} \mathcal{L} = - \begin{bmatrix} \vec{\mathbf{0}} \\ \vec{\mathbf{P}} \\ \vec{\mathbf{0}} \end{bmatrix} + \begin{bmatrix} \mathbf{V}_1 \\ -\mathbf{V}_2 \end{bmatrix}^T \vec{\boldsymbol{\lambda}} \quad (\text{A.10})$$

$$\nabla_{\lambda}\mathcal{L}=\vec{h}_g \tag{A.11}$$

APPENDIX B

The Hessian form in Eq. (2.36), is restated here:

$$\mathbf{H} = \begin{bmatrix} \nabla_{\mathbf{y}\mathbf{y}}^2 \mathcal{L} & \mathbf{0} & \nabla_{\mathbf{y}} \vec{\mathbf{h}}_g^T \\ \mathbf{0} & \mathbf{0} & \nabla_{\mathbf{d}_g} \vec{\mathbf{h}}_g^T \\ \nabla_{\mathbf{y}} \vec{\mathbf{h}}_g & \nabla_{\mathbf{d}_g} \vec{\mathbf{h}}_g & \mathbf{0} \end{bmatrix}$$

The gradients of $\vec{\mathbf{h}}_g$ have been derived in the previous section. Thus, for the complete specification of the second order information we need only to determine the second derivative matrix with respect to the strain vector $\vec{\mathbf{y}}$. Since different integration points are independent of one another, then it is clear that $\nabla_{\mathbf{y}_i \mathbf{y}_j}^2 \mathcal{L} = \mathbf{0}$ for $i \neq j$. Therefore, the matrix $\nabla_{\mathbf{y}\mathbf{y}}^2 \mathcal{L}$ has a block diagonal form:

$$\nabla_{\mathbf{y}\mathbf{y}}^2 \mathcal{L} = \begin{bmatrix} \nabla_{\mathbf{y}_i \mathbf{y}_i}^2 \mathcal{L} & \mathbf{0} & \cdots & \mathbf{0} \\ \mathbf{0} & \nabla_{\mathbf{y}_i \mathbf{y}_i}^2 \mathcal{L} & \cdots & \mathbf{0} \\ \vdots & \ddots & & \vdots \\ \mathbf{0} & \cdots & \mathbf{0} & \nabla_{\mathbf{y}_m \mathbf{y}_m}^2 \mathcal{L} \end{bmatrix}$$

and making use of Eq. (2.60) gives:

$$\nabla_{\mathbf{y}_i \mathbf{y}_i}^2 \mathcal{L} = \nabla_{\mathbf{y}_i \mathbf{y}_i}^2 U + \nabla_{\mathbf{y}_i} ([\nabla_{\mathbf{y}_i} \vec{\mathbf{h}}_g]^T \vec{\boldsymbol{\lambda}}) \quad (\text{B.1})$$

As already seen, the second derivative matrix of the strain energy yields the generalized section

stiffness of cross-section i :

$$\nabla_{\mathbf{y}_i}^2 U = w_i \begin{bmatrix} \mathbf{k}_{sec}^{(i)} & \mathbf{0} \\ \mathbf{0} & 0 \end{bmatrix}$$

where $\mathbf{k}_{sec}^{(i)}$ is given by Eq. (2.58) after numerical integration of the cross-section. We only now need to derive the explicit form for the second term of the right hand side in Eq. (B.1). To this end, we separate the Lagrange multiplier vector into two parts, which correspond to vectors $\vec{\mathbf{h}}_h^A$ and $\vec{\mathbf{h}}_h^B$ respectively:

$$\vec{\boldsymbol{\lambda}} = \begin{bmatrix} \vec{\boldsymbol{\lambda}}^A \\ \vec{\boldsymbol{\lambda}}^B \end{bmatrix}$$

Eq. (A.9) can be now restated as:

$$\nabla_{\mathbf{y}_i} \mathcal{L} = w_i \begin{bmatrix} \vec{\mathbf{F}}_{sec}^{(i)} \\ 0 \end{bmatrix} - w_i \begin{bmatrix} \mathbf{R}_i^T \boldsymbol{\Lambda} \\ [\vec{\mathbf{q}}_i + \vec{\mathbf{E}}_1]^T \mathbf{X}^T \mathbf{R}_i \boldsymbol{\Lambda} \end{bmatrix} \vec{\boldsymbol{\lambda}}^A + \begin{bmatrix} \vec{\mathbf{0}}^T \\ \vec{\mathbf{0}}^T \\ -\hat{\mathbf{E}}_i^T \mathbf{T}^T \\ \hat{\mathbf{E}}_i^T \end{bmatrix} \vec{\boldsymbol{\lambda}}^B \quad (\text{B.2})$$

Noticing that the gradient of $\vec{\mathbf{h}}_g^B$ yields components independent of strains, we conclude that the second derivatives of \mathcal{L} will include only the part $\vec{\boldsymbol{\lambda}}^A$, as:

$$\nabla_{\mathbf{y}_i}^2 \mathcal{L} = \nabla_{\mathbf{y}_i}^2 U + \mathbf{Z}_i \quad (\text{B.3})$$

where:

$$\mathbf{Z}_i = \begin{bmatrix} \mathbf{0}_{(3 \times 3)} & \vec{\mathbf{t}}_i \\ \vec{\mathbf{t}}_i^T & \vec{\mathbf{t}}_i^T \mathbf{X} [\vec{\mathbf{q}}_i + \vec{\mathbf{E}}_1] \end{bmatrix} \quad \text{and} \quad \vec{\mathbf{t}}_i = -w_i \mathbf{X}^T \mathbf{R}_i^T \boldsymbol{\Lambda} \vec{\boldsymbol{\lambda}}^A$$

and the second derivative matrix of the Lagrangian is now fully determined.

APPENDIX C

First, we express the effective stress vector $\vec{\zeta}_{n+1}$ purely in terms of λ , using eq. (3.23a) and the fact that $\vec{n} = \sqrt{\frac{3}{2}} \frac{\mathbf{V}\vec{\zeta}}{\|\vec{\zeta}\|_{\mathbf{V}}}$:

$$\left[\mathbf{I} + \frac{\lambda \sqrt{\frac{3}{2}}}{\|\vec{\zeta}_{n+1}\|_{\mathbf{V}}} \left[\mathbf{C}^{el} \mathbf{V} + H_{kin} \mathbf{I} \right] \right] \vec{\zeta}_{n+1} = \vec{\zeta}_{n+1}^{TR} \quad (\text{C.1})$$

In the first phase of a plastic step we assume that no plastic flow occurs (Elastic Prediction), therefore, for the purposes of the iterative strategy we can consider the following identity:

$$f_{n+1} = 0 \rightarrow \|\vec{\zeta}_{n+1}\|_{\mathbf{V}} = q_n + \lambda \sqrt{\frac{2}{3}} \frac{\partial q_n}{\partial e_n^{pl}} \quad (\text{C.2})$$

In the case of linear isotropic hardening, eq. (C.2) is exact since $\frac{\partial q}{\partial e^{pl}} = H_{iso}$. Otherwise the isotropic modulus used in the context of the fully implicit integration is replaced with the one at the previous step for the purposes of the iterative strategy and only for that. This technique is done so that $\vec{\zeta}_{n+1}$ is recast only in terms of λ . Substituting (C.2) into (C.1) and solving for $\vec{\zeta}_{n+1}$ we get:

$$\vec{\zeta}_{n+1} = \Omega \vec{\zeta}_{n+1}^{TR} \quad (\text{C.3})$$

Let $a_1 = E + \frac{3}{2}H_{kin} + \frac{\partial q_n}{\partial e_n^{pl}}$ and $a_2 = 3G + \frac{3}{2}H_{kin} + \frac{\partial q_n}{\partial e_n^{pl}}$. Then:

$$\Omega = \left[\mathbf{I} + \frac{\lambda \sqrt{\frac{3}{2}}}{q_n + \lambda \sqrt{\frac{2}{3}} \frac{\partial q_n}{\partial e_n^{pl}}} \left[\mathbf{C}^{el} \mathbf{V} + H_{kin} \mathbf{I} \right] \right]^{-1} = \text{diag} \left(\frac{q_n + \lambda \frac{\partial q_n}{\partial e_n^{pl}}}{q_n + \lambda a_1}, \frac{q_n + \lambda \frac{\partial q_n}{\partial e_n^{pl}}}{q_n + \lambda a_2} \right) \quad (\text{C.4})$$

Substitution of (C.3) into $f(\vec{\zeta}_{n+1}, q_{n+1}) \equiv f_{n+1} = 0$ yields the scalar equation to be solved

iteratively for λ :

$$f_{n+1} = \sqrt{\frac{3}{2}} \sqrt{(\vec{\zeta}_{n+1}^{TR})^T \Omega \mathbf{V} \Omega \vec{\zeta}_{n+1}^{TR}} - q_{n+1} = 0 \quad (\text{C.5})$$

The derivative of f_{n+1} with respect to λ involves finding the derivative of the diagonal matrix Ω and a chain rule with respect to q_{n+1} . It can be found that the derivative is:

$$\frac{df_{n+1}}{d\lambda} = -\frac{1}{q_n} \left[\frac{n_{11}\zeta_{11}^{TR}(E + \frac{3}{2}H_{kin})}{(1 + \frac{\lambda a_1}{q_n})^2} + \frac{n_{12}\zeta_{12}^{TR}(3G + \frac{3}{2}H_{kin})}{(1 + \frac{\lambda a_2}{q_n})^2} \right] - (1 + \lambda) \frac{\partial q_{n+1}}{\partial e_{n+1}^{pl}} \quad (\text{C.6})$$

APPENDIX D

For the inversion of the block matrix (3.25a) we perform block elimination considering $\mathbf{A} = \mathbf{C}^{el} + \lambda\boldsymbol{\Psi} - \lambda^2\boldsymbol{\Psi}_{n+1}\left[H_{kin}\mathbf{V} + \lambda\boldsymbol{\Psi}_{n+1}\right]^{-1}\boldsymbol{\Psi}_{n+1}$ as the Schur complement. Notice that while the matrix $\lambda\boldsymbol{\Psi}_{n+1}$ is not invertible, the diagonal ones are. Omitting the subscript referring to the step, inversion of system (3.25a) yields:

$$\begin{bmatrix} d\vec{\sigma} \\ d\vec{\alpha} \end{bmatrix} = \begin{bmatrix} \boldsymbol{\Xi}_{11} & \boldsymbol{\Xi}_{12} \\ \boldsymbol{\Xi}_{21} & \boldsymbol{\Xi}_{22} \end{bmatrix} \begin{bmatrix} d\vec{\epsilon} - d\lambda\vec{n} \\ d\lambda\vec{n} \end{bmatrix} \quad (\text{D.1})$$

and define $\boldsymbol{\Gamma} = \lambda\boldsymbol{\Psi}\left(H_{kin}^{-1}\mathbf{V} + \lambda\boldsymbol{\Psi}\right)^{-1}$. Then, for the block matrices $[\boldsymbol{\Xi}_{ij}]$ we have:

- $\boldsymbol{\Xi}_{11} = \mathbf{A}^{-1}$, the inverse of the Schur complement.
- $\boldsymbol{\Xi}_{12} = \boldsymbol{\Xi}_{11}\boldsymbol{\Gamma}$
- $\boldsymbol{\Xi}_{21} = \boldsymbol{\Xi}_{12}^T$
- $\boldsymbol{\Xi}_{22} = \left(H_{kin}^{-1}\mathbf{V} + \lambda\boldsymbol{\Psi}\right)^{-1} + \boldsymbol{\Gamma}\boldsymbol{\Xi}_{11}\boldsymbol{\Gamma}$

where the fact that matrices $\boldsymbol{\Psi}$ and $\boldsymbol{\Gamma}$ are symmetric was used. The vector \vec{N} in (3.28) is given by:

$$\vec{N} = \frac{[\boldsymbol{\Xi}_{11} + \boldsymbol{\Xi}_{12}]\vec{n}}{\|\vec{n}\|_M} \quad (\text{D.2})$$

with $\mathbf{M} = \boldsymbol{\Xi}_{11} + \boldsymbol{\Xi}_{22} + \boldsymbol{\Xi}_{12} + \boldsymbol{\Xi}_{21}$. Finally, parameter δ is defined as:

$$\delta = \frac{\frac{\partial q}{\partial e^{pl}}}{\|\vec{n}\|_M^2} \quad (\text{D.3})$$

The calculation can be simplified significantly if second and higher order terms in λ are ignored.

BIBLIOGRAPHY

- [1] E. Reissner, “On one-dimensional finite-strain beam theory: The plane problem,” *Zeitschrift für Angewandte Mathematik und Mechanik*, vol. 23, no. 5, pp. 795–804, 1972.
- [2] J. Simo, “A finite strain beam formulation. The three-dimensional dynamic problem. Part I,” *Computer Methods in Applied Mechanics and Engineering*, vol. 49, no. 1, pp. 55–70, 1985.
- [3] J. Simo and L. Vu-Quoc, “A three-dimensional finite-strain rod model. Part II: Computational aspects,” *Computer Methods in Applied Mechanics and Engineering*, vol. 58, no. 1, pp. 79–116, 1986.
- [4] J. Simo and L. Vu-Quoc, “On the dynamics of flexible beams under large overall motions—the plane case: Part I,” *Journal of Applied Mechanics*, vol. 4, no. 53, pp. 849–854, 1986.
- [5] A. Neuenhofer and F. C. Filippou, “Geometrically nonlinear flexibility-based frame finite element,” *Journal of Structural Engineering*, vol. 124, no. 6, pp. 704–711, 1998.
- [6] M. Schulz and F. C. Filippou, “Non-linear spatial timoshenko beam element with curvature interpolation,” *International Journal for Numerical Methods in Engineering*, vol. 50, no. 4, pp. 761–785, 2001.
- [7] A. Cardona and M. Geradin, *Flexible Multibody Dynamics: A Finite Element Approach*, English. UK: John Wiley and Sons, 2001.
- [8] D. Hodges, “Proper definition of curvature in nonlinear beam kinematics,” *American Institute of Aeronautics and Astronautics Journal*, vol. 22, no. 12, pp. 1825–1827, 1984.
- [9] D. Luenberger and Y. Yu, *Linear and Nonlinear Programming*, English. New York, NY: Springer, 2008.

- [10] K. Washizu, Variational Methods in Elasticity and Plasticity, English. Pergamon Press, 1975.
- [11] J. Simo and T. Hughes, Computational Inelasticity, English. New York,NY: Springer, 1998.
- [12] C. P. Andriotis, K. G. Papakonstantinou, and V. K. Koumousis, “Nonlinear programming hybrid beam-column element formulation for large-displacement elastic and inelastic analysis,” Journal of Engineering Mechanics, vol. 144, no. 10, p. 04018096, 2018.
- [13] M. Crisfield, “A fast incremental/iterative solution procedure that handles ”snap-through”,” Computers & Structures, vol. 13, no. 1–3, pp. 55–62, 1981.
- [14] G. Cowper, “The shear coefficient in timoshenko’s beam theory,” Journal of Applied Mechanics, vol. 33, no. 2, pp. 335–340, 1966.
- [15] S. Mazzoni, F. McKenna, M. Scott, and G. Fenves, “Opensees command language manual,” Pacific Earthquake Engineering Research (PEER) Center, vol. 264, 2006.
- [16] A. Neuenhofer and F. C. Filippou, “Evaluation of nonlinear frame finite-element models,” Journal of Structural Engineering, vol. 123, no. 7, pp. 958–966, 1997.
- [17] ABAQUS/Standard User’s Manual, English. United States: Dassault Systèmes Simulia Corp, 2019.
- [18] K. Bishopp and D. Drucker, “Large deflection of cantilever beams,” Quarterly of Applied Mathematics, vol. 3, no. 3, pp. 272–275, 1945.
- [19] K. Mattiasson, “Numerical results from large deflection beam and frame problems analysed by means of elliptic integrals,” International Journal for Numerical Methods in Engineering, vol. 17, no. 1, pp. 145–153, 1981.

- [20] M. Batista, “A closed-form solution for Reissner planar finite-strain beam using Jacobi elliptic functions,” *International Journal of Solids and Structures*, vol. 87, pp. 153–166, 2016.
- [21] J. Reddy, “On locking-free shear deformable beam finite elements,” *Computer Methods in Applied Mechanics and Engineering*, vol. 149, no. 1-4, pp. 113–132, 1997.
- [22] G. Prathap and G. Bhashyam, “Reduced integration and the shear-flexible beam element,” *International Journal for Numerical Methods in Engineering*, vol. 18, no. 2, pp. 195–210, 1982.
- [23] O. Zienkiewicz, R. Taylor, and J. Too, “Reduced integration technique in general analysis of plates and shells,” *International Journal for Numerical Methods in Engineering*, vol. 3, no. 2, pp. 275–290, 1971.
- [24] S.-L. Lee, F. S. Manuel, and E. C. Rossow, “Large deflections and stability of elastic frames,” *Journal of Engineering Mechanics Division*, vol. 94, no. 2, pp. 521–548, 1968.
- [25] M. R. Horne, “Elastic-plastic failure loads of plane frames,” *Proceedings of the Royal Society of London A: Mathematical, Physical and Engineering Sciences*, vol. 274, no. 1358, pp. 343–364, 1963.
- [26] H. Santos, “Complementary-energy methods for geometrically non-linear structural models: An overview and recent developments in the analysis of frames,” *Archives of Computational Mechanics*, vol. 18, no. 4, p. 405, 2011.
- [27] J. Simo and R. Taylor, “A return mapping algorithm for plane stress elastoplasticity,” *International Journal for Numerical Methods in Engineering*, vol. 22, no. 3, pp. 649–670, 1986.
- [28] A. Papachristidis, M. Fragiadakis, and M. Papadrakakis, “A 3d fibre beam-column element with shear modelling for the inelastic analysis of steel structures,” *Computational Mechanics*, vol. 45, no. 6, pp. 553–572, 2010.

- [29] A. Saritas and F. C. Filippou, “Frame element for metallic shear-yielding members under cyclic loading,” *Journal of Structural Engineering*, vol. 135, no. 9, pp. 1115–1123, 2009.
- [30] P. Ceresa, L. Petrini, R. Pinho, and R. Sousa, “A fibre flexure–shear model for seismic analysis of rc-framed structures,” *Earthquake Engineering & Structural Dynamics*, vol. 38, no. 5, pp. 565–586, 2009.
- [31] J. N. Gregori, P. M. Sosa, M. F. Prada, and F. C. Filippou, “A 3d numerical model for reinforced and prestressed concrete elements subjected to combined axial, bending, shear and torsion loading,” *Engineering Structures*, vol. 29, no. 12, pp. 3404–3419, 2007.
- [32] A. Kagermanov and P. Ceresa, “Fiber-section model with an exact shear strain profile for two-dimensional rc frame structures,” *Journal of Structural Engineering*, vol. 143, no. 10, p. 04017132, 2017.
- [33] J. C. Simo and T. J. Hughes, *Computational inelasticity*. Springer Science & Business Media, 2006, vol. 7.
- [34] M. L. Wilkins, “Calculation of elastic-plastic flow,” California Univ Livermore Radiation Lab, Tech. Rep., 1963.
- [35] R. H. Dodds Jr, “Numerical techniques for plasticity computations in finite element analysis,” *Computers & Structures*, vol. 26, no. 5, pp. 767–779, 1987.
- [36] F. De Angelis and R. L. Taylor, “An efficient return mapping algorithm for elastoplasticity with exact closed form solution of the local constitutive problem,” *Engineering Computations*, vol. 32, no. 8, 2015.
- [37] J. Clausen, L. Damkilde, and L. Andersen, “Efficient return algorithms for associated plasticity with multiple yield planes,” *International Journal for Numerical Methods in Engineering*, vol. 66, no. 6, pp. 1036–1059, 2006.

- [38] O. Hopperstad and S Remseth, “A return mapping algorithm for a class of cyclic plasticity models,” *International Journal for Numerical Methods in Engineering*, vol. 38, no. 4, pp. 549–564, 1995.
- [39] W. M. Scherzinger, “A return mapping algorithm for isotropic and anisotropic plasticity models using a line search method,” *Computer Methods in Applied Mechanics and Engineering*, vol. 317, pp. 526–553, 2017.
- [40] A. R. Hartloper, A. de Castro e Sousa, and D. G. Lignos, “Constitutive modeling of structural steels: Nonlinear isotropic/kinematic hardening material model and its calibration,” *Journal of Structural Engineering*, vol. 147, no. 4, p. 04021031, 2021.
- [41] J. C. Simo and R. L. Taylor, “Consistent tangent operators for rate-independent elasto-plasticity,” *Computer methods in applied mechanics and engineering*, vol. 48, no. 1, pp. 101–118, 1985.
- [42] M Ortiz and E. P. Popov, “Accuracy and stability of integration algorithms for elasto-plastic constitutive relations,” *International journal for numerical methods in engineering*, vol. 21, no. 9, pp. 1561–1576, 1985.
- [43] E. A. de Souza Neto, D. Peric, and D. R. Owen, *Computational methods for plasticity: theory and application*, John Wiley & Sons, 2011.
- [44] K. D. Hjelmstad and E. P. Popov, “Cyclic behavior and design of link beams,” *Journal of Structural Engineering*, vol. 109, no. 10, pp. 2387–2403, 1983.
- [45] J. Argyris, B Boni, U Hindenlang, and M Kleiber, “Finite element analysis of two-and three-dimensional elasto-plastic frames—the natural approach,” *Computer Methods in Applied Mechanics and Engineering*, vol. 35, no. 2, pp. 221–248, 1982.
- [46] R. A. Ridha and L. H. Lee, “Inelastic finite deformation of planar frames,” *Journal of the Engineering Mechanics Division*, vol. 97, no. 3, pp. 773–789, 1971.

- [47] M Amir, K. Papakonstantinou, and G. Warn, “Scaled spherical simplex filter and state-space damage-plasticity finite-element model for computationally efficient system identification,” *Journal of Engineering Mechanics*, vol. 148, no. 2, p. 04021138, 2022.
- [48] E. L. Allgower and K. Georg, *Introduction to numerical continuation methods*. SIAM, 2003.
- [49] W. C. Rheinboldt, “Numerical continuation methods: A perspective,” *Journal of computational and applied mathematics*, vol. 124, no. 1-2, pp. 229–244, 2000.
- [50] H. B. Keller, “Global homotopies and newton methods,” in *Recent advances in numerical analysis*, Elsevier, 1978, pp. 73–94.
- [51] T.-Y. Li and J. A. Yorke, “A simple reliable numerical algorithm for following homotopy paths,” *Analysis and computation of fixed points*, pp. 73–91, 1980.
- [52] S. N. Chow, J. Mallet-Paret, and J. A. Yorke, “Finding zeroes of maps: Homotopy methods that are constructive with probability one,” *Mathematics of Computation*, vol. 32, no. 143, pp. 887–899, 1978.
- [53] S.-N. Chow, J. Mallet-Paret, and J. A. Yorke, “A homotopy method for locating all zeros of a system of polynomials,” in *Functional differential equations and approximation of fixed points*, Springer, 1979, pp. 77–88.
- [54] L. T. Watson, “Globally convergent homotopy methods: A tutorial,” *Applied Mathematics and Computation*, vol. 31, pp. 369–396, 1989.
- [55] L. T. Watson, “Globally convergent homotopy algorithms for nonlinear systems of equations,” *Nonlinear Dynamics*, vol. 1, no. 2, pp. 143–191, 1990.
- [56] E. L. Allgower, “A survey of homotopy methods for smooth mappings,” in *Numerical solution of nonlinear systems of equations*, Springer, 1981, pp. 1–29.
- [57] W. C. Rheinboldt, “Solution fields of nonlinear equations and continuation methods,” *SIAM Journal on Numerical Analysis*, vol. 17, no. 2, pp. 221–237, 1980.

- [58] W. C. Rheinboldt and J. V. Burkardt, “A locally parameterized continuation process,” *ACM Transactions on Mathematical Software (TOMS)*, vol. 9, no. 2, pp. 215–235, 1983.
- [59] T. Wayburn and J. Seader, “Homotopy continuation methods for computer-aided process design,” *Computers & Chemical Engineering*, vol. 11, no. 1, pp. 7–25, 1987.
- [60] M. Kojima and R. Hirabayashi, “Continuous deformation of nonlinear programs,” in *Sensitivity, Stability and Parametric Analysis*, Springer, 1984, pp. 150–198.
- [61] H Gfrerer, H. Wacker, W Zulehner, and J Guddat, “Path-following methods for kuhn-tucker curves by an active index set strategy,” in *Systems and Optimization*, Springer, 1985, pp. 111–131.
- [62] J. Guddat, F. G. Vazquez, and H. T. Jongen, *Parametric optimization: singularities, pathfollowing and j* Springer, 1990.
- [63] H. Gfrerer, J Guddat, *et al.*, “A globally convergent algorithm based on imbedding and parametric optimization,” *Computing*, vol. 30, no. 3, pp. 225–252, 1983.
- [64] W. I. Zangwill and C. Garcia, “Equilibrium programming: The path following approach and dynamics,” *Mathematical Programming*, vol. 21, no. 1, pp. 262–289, 1981.
- [65] A. B. Poore, “Bifurcations in parametric nonlinear programming,” *Annals of Operations Research*, vol. 27, no. 1, pp. 343–369, 1990.
- [66] H. T. Jongen and G.-W. Weber, “On parametric nonlinear programming,” *Annals of Operations Research*, vol. 27, no. 1, pp. 253–283, 1990.
- [67] C. Tiahrt and A. Poore, “Bifurcation and persistance of minima in nonlinear parametric programming,” in *Computation and Control*, Springer, 1989, pp. 343–353.
- [68] G. A. Wempner, “Discrete approximations related to nonlinear theories of solids,” *International Journal*, vol. 7, no. 11, pp. 1581–1599, 1971.

- [69] P. Bergan, G Horrigmoe, B Bråkeland, and T. Søreide, “Solution techniques for non-linear finite element problems,” *International Journal for Numerical Methods in Engineering*, vol. 12, no. 11, pp. 1677–1696, 1978.
- [70] P. Bergan and T. Soreide, “Solution of large displacement and instability problems using the current stiffness parameter,” *Finite elements in nonlinear mechanics*, vol. 2, pp. 647–669, 1978.
- [71] J.-L. Batoz and G. Dhatt, “Incremental displacement algorithms for nonlinear problems,” *International Journal for Numerical Methods in Engineering*, vol. 14, no. 8, pp. 1262–1267, 1979.
- [72] E Riks, “An incremental approach to the solution of snapping and buckling problems,” *International journal of solids and structures*, vol. 15, no. 7, pp. 529–551, 1979.
- [73] E. Ramm, “Strategies for tracing the nonlinear response near limit points,” in *Nonlinear finite element analysis*. Springer, 1981, pp. 63–89.
- [74] L. T. Watson, M. P. Kamat, and M. H. Reaser, “A robust hybrid algorithm for computing multiple equilibrium solutions,” *Engineering Computations*, 1985.
- [75] P. Sideris, M. Nikoukalam, and M Salehi, “A generalized normal flow method with online step controls for pushover analysis of nonlinear softening structures,” *Engineering Structures*, vol. 143, pp. 232–244, 2017.
- [76] G. D. Byrne and L. A. Baird, “Distillation calculations using a locally parameterized continuation method,” *Computers & chemical engineering*, vol. 9, no. 6, pp. 593–599, 1985.
- [77] A Ushida and L. Chua, “Tracing solution curves of non-linear equations with sharp turning points,” *International journal of circuit theory and applications*, vol. 12, no. 1, pp. 1–21, 1984.

- [78] R. N. Borkovsky, U. Doraszelski, and Y. Kryukov, “A user’s guide to solving dynamic stochastic games using the homotopy method,” *Operations Research*, vol. 58, no. 4-part-2, pp. 1116–1132, 2010.
- [79] D. Besanko, U. Doraszelski, Y. Kryukov, and M. Satterthwaite, “Learning-by-doing, organizational forgetting, and industry dynamics,” *Econometrica*, vol. 78, no. 2, pp. 453–508, 2010.
- [80] P. J.-J. Herings and R. Peeters, “Homotopy methods to compute equilibria in game theory,” *Economic Theory*, vol. 42, no. 1, pp. 119–156, 2010.
- [81] M. Bartoň and V. M. Calo, “Optimal quadrature rules for odd-degree spline spaces and their application to tensor-product-based isogeometric analysis,” *Computer Methods in Applied Mechanics and Engineering*, vol. 305, pp. 217–240, 2016.
- [82] P. J. Barendrecht, M. Bartoň, and J. Kosinka, “Efficient quadrature rules for subdivision surfaces in isogeometric analysis,” *Computer Methods in Applied Mechanics and Engineering*, vol. 340, pp. 1–23, 2018.
- [83] L. T. Watson, “Probability-one homotopies in computational science,” *Journal of Computational and Applied Mathematics*, vol. 140, no. 1-2, pp. 785–807, 2002.
- [84] S. Smale, “A convergent process of price adjustment and global newton methods,” *Journal of Mathematical Economics*, vol. 3, no. 2, pp. 107–120, 1976.
- [85] C. Garcia and F. Gould, “Relations between several path following algorithms and local and global newton methods,” *Siam Review*, vol. 22, no. 3, pp. 263–274, 1980.
- [86] W. C. Rheinboldt, “Numerical analysis of continuation methods for nonlinear structural problems,” in *Computational Methods in Nonlinear Structural and Solid Mechanics*, Elsevier, 1981, pp. 103–113.
- [87] W.-J. Lin, J. Seader, and T. Wayburn, “Computing multiple solutions to systems of interlinked separation columns,” *AIChE Journal*, vol. 33, no. 6, pp. 886–897, 1987.

- [88] A. C. Sun and W. D. Seider, “Homotopy-continuation method for stability analysis in the global minimization of the gibbs free energy,” *Fluid Phase Equilibria*, vol. 103, no. 2, pp. 213–249, 1995.
- [89] R. Menzel and H. Schwetlick, “Parametrization via secant length and application to path following,” *Numerische Mathematik*, vol. 47, no. 3, pp. 401–412, 1985.
- [90] M. Kuno and J. Seader, “Computing all real solutions to systems of nonlinear equations with a global fixed-point homotopy,” *Industrial & engineering chemistry research*, vol. 27, no. 7, pp. 1320–1329, 1988.
- [91] G. Powell and J. Simons, “Improved iteration strategy for nonlinear structures,” *International Journal for Numerical Methods in Engineering*, vol. 17, no. 10, pp. 1455–1467, 1981.
- [92] K.-J. Bathe and E. N. Dvorkin, “On the automatic solution of nonlinear finite element equations,” *Computers & Structures*, vol. 17, no. 5-6, pp. 871–879, 1983.
- [93] A. Klarbring, “On discrete and discretized non-linear elastic structures in unilateral contact (stability, uniqueness and variational principles),” *International journal of solids and structures*, vol. 24, no. 5, pp. 459–479, 1988.
- [94] G. Björkman, “Path following and critical points for contact problems,” *Computational Mechanics*, vol. 10, no. 3, pp. 231–246, 1992.
- [95] W. C. Rheinboldt, *An adaptive continuation process for solving systems of nonlinear equations*. SIAM, 1975.
- [96] L. T. Watson, “Numerical linear algebra aspects of globally convergent homotopy methods,” *SIAM review*, vol. 28, no. 4, pp. 529–545, 1986.
- [97] P. Deuflhard, B. Fiedler, and P. Kunkel, “Efficient numerical pathfollowing beyond critical points,” *SIAM journal on numerical analysis*, vol. 24, no. 4, pp. 912–927, 1987.
- [98] R. Seydel and V. Hlavacek, “Role of continuation in engineering analysis,” *Chemical engineering science*, vol. 42, no. 6, pp. 1281–1295, 1987.

- [99] A Šalgovič, V Hlaváček, and J Ilavský, “Global simulation of countercurrent separation processes via one-parameter imbedding techniques,” *Chemical Engineering Science*, vol. 36, no. 10, pp. 1599–1604, 1981.
- [100] H. Schwetlick and J. Cleve, “Higher order predictors and adaptive steplength control in path following algorithms,” *SIAM journal on numerical analysis*, vol. 24, no. 6, pp. 1382–1393, 1987.
- [101] J.-J. Gervais and H. Sadiky, “A continuation method based on a high order predictor and an adaptive steplength control,” *ZAMM-Journal of Applied Mathematics and Mechanics/Zeitschrift für Angewandte Mathematik und Mechanik*, vol. 84, no. 8, pp. 551–563, 2004.
- [102] M. Syam, “Interpolation predictors over implicitly defined curves,” *Computers & Mathematics with Applications*, vol. 44, no. 8-9, pp. 1067–1076, 2002.
- [103] M. I. Syam, “Cubic spline interpolation predictors over implicitly defined curves,” *Journal of computational mathematics*, vol. 157, no. 2, pp. 283–295, 2003.
- [104] L. T. Watson, S. C. Billups, and A. P. Morgan, “Algorithm 652: Hompack: A suite of codes for globally convergent homotopy algorithms,” *ACM Transactions on Mathematical Software (TOMS)*, vol. 13, no. 3, pp. 281–310, 1987.
- [105] B. N. Lundberg and A. B. Poore, “Numerical continuation and singularity detection methods for parametric nonlinear programming,” *SIAM Journal on Optimization*, vol. 3, no. 1, pp. 134–154, 1993.
- [106] E. Allgower, C.-S. Chien, K Georg, and C.-F. Wang, “Conjugate gradient methods for continuation problems,” *Journal of Computational and Applied Mathematics*, vol. 38, no. 1-3, pp. 1–16, 1991.
- [107] H. B. Keller, “Lectures on numerical methods in bifurcation problems,” *Applied Mathematics*, vol. 217, p. 50, 1987.

- [108] B. N. Lundberg and A. B. Poore, “Variable order adams-bashforth predictors with an error-stepsize control for continuation methods,” *SIAM journal on scientific and statistical computing*, vol. 12, no. 3, pp. 695–723, 1991.
- [109] W. Mackens, “Numerical differentiation of implicitly defined space curves,” *Computing*, vol. 41, no. 3, pp. 237–260, 1989.
- [110] M. Kubíček, “Algorithm 502: Dependence of solution of nonlinear systems on a parameter [c5],” *ACM Transactions on Mathematical Software (TOMS)*, vol. 2, no. 1, pp. 98–107, 1976.
- [111] L. T. Watson, “A globally convergent algorithm for computing fixed points of c2 maps,” *Applied Mathematics and Computation*, vol. 5, no. 4, pp. 297–311, 1979.
- [112] K. Georg, “On tracing an implicitly defined curve by quasi-newton steps and calculating bifurcation by local perturbations,” *SIAM Journal on Scientific and Statistical Computing*, vol. 2, no. 1, pp. 35–50, 1981.
- [113] S. N. Chow, J. Mallet-Paret, and J. A. Yorke, “Finding zeroes of maps: Homotopy methods that are constructive with probability one,” *Mathematics of Computation*, vol. 32, no. 143, pp. 887–899, 1978.
- [114] M. J. Clarke and G. J. Hancock, “A study of incremental-iterative strategies for nonlinear analyses,” *International Journal for Numerical Methods in Engineering*, vol. 29, no. 7, pp. 1365–1391, 1990.

VITA

Vita may be provided by doctoral students only. The length of the vita is preferably one page. It may include the place of birth and should be written in third person. This vita is similar to the author biography found on book jackets.