

**ANALYSIS OF LEAST SQUARES
FINITE ELEMENT METHODS FOR
THE NAVIER-STOKES EQUATIONS**

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SUMMARY

Introduction

- Notation
- What is a least squares (LS) method
- Least squares strategy: how and why

LS for the Navier-Stokes Equations

- VVP equations in 2D and in 3D
- ADN infomercial
- Boundary conditions
- Modified equations
- The LS methods
- Implementation issues

Error Analysis

- Abstract approximation theory
- Application to LS methods

Sobolev Spaces

$$W^{m,p}(\Omega) = \{u \in L^p(\Omega) \mid D^\alpha u \in L^p(\Omega) \forall |\alpha| \leq m\}$$

For example

$$L^2(\Omega) = \{u \mid \int_{\Omega} u^2 d\Omega < \infty\}$$

$$H^1(\Omega) = \{u \mid D^\alpha u \in L^2(\Omega), |\alpha| \leq 1\}$$

Norm for a function g belonging to $H^m(\Omega)$:

$$\|g\|_m^2 = \sum_{|\alpha| \leq m} \int_{\Omega} (D^\alpha g)^2 d\Omega$$

Sobolev spaces related to the Navier-Stokes equations:

$$L_0^2(\Omega) = \{u \in L^2(\Omega) \mid \int_{\Omega} u d\Omega = 0\}$$

$$H_0^1(\Omega) = \{u \in H^1(\Omega) \mid u = 0 \text{ on } \Gamma\}$$

$$\tilde{H}^1(\Omega) = \{u \in H^1(\Omega) \mid \int_{\Omega} u d\Omega = 0\}$$

Curls and Vector Products.

Curl operator in 3D:

$$\mathbf{curl} \mathbf{u} = \nabla \times \mathbf{u}$$

Curl operator(s) in 2D:

$$\mathbf{curl} \omega = \begin{pmatrix} \omega_y \\ -\omega_x \end{pmatrix} \quad \text{and} \quad \mathbf{curl} \mathbf{u} = u_{2x} - u_{1y}.$$

“Vector product” in 2D:

$$\phi \longrightarrow (0, 0, \phi)$$

$$\mathbf{u} \longrightarrow (u_1, u_2, 0)$$

$$\mathbf{v} \longrightarrow (v_1, v_2, 0)$$

then we define:

$$\phi \times \mathbf{u} \longrightarrow \phi \begin{pmatrix} -u_2 \\ u_1 \end{pmatrix}$$

$$\mathbf{u} \times \mathbf{v} \longrightarrow u_1 v_2 - u_2 v_1$$

Vorticity in 3D/2D:

$$\omega = \mathbf{curl} \mathbf{u} \quad (= \nabla \times \mathbf{u})$$

$$\omega = \mathbf{curl} \mathbf{u} \quad (= \nabla \times \mathbf{u})$$

What is a Least Squares Method

Boundary value problem

$$\begin{cases} \mathcal{L}U = F & \text{in } \Omega \\ \mathcal{R}U = 0 & \text{on } \Gamma \end{cases}$$

where $\mathcal{L} : \mathbf{X} \mapsto \mathbf{Y}$; \mathbf{X} and \mathbf{Y} are Hilbert spaces.

1. Least squares functional

$$J(U) = (\mathcal{L}U - F, \mathcal{L}U - F)_Y = \|\mathcal{L}U - F\|_Y^2.$$

2. Least squares principle

Seek $U \in \mathbf{X}$ such that $J(U) \leq J(V) \forall V \in \mathbf{X}$.

3. Euler-Lagrange equation

$$\delta J(U) = \lim_{\varepsilon \rightarrow 0} \frac{d}{d\varepsilon} J(U + \varepsilon V) = 0 \quad \forall V \in \mathbf{X}.$$

Equivalent variational problem

Seek $U \in \mathbf{X}$ such that

$$Q(U; V) = \mathcal{F}(V) \text{ for all } V \in \mathbf{X}.$$

Existence of minimizers.

Assume the *a priori* estimate

$$\|U\|_X \leq C\|\mathcal{L}U\|_Y.$$

Then $Q(\cdot; \cdot)$ is **coercive** on $\mathbf{X} \times \mathbf{X}$:

$$Q(U; U) = (\mathcal{L}U, \mathcal{L}U)_Y = \|\mathcal{L}U\|_Y^2 \geq C\|U\|_X^2.$$

\implies Existence and uniqueness of the minimizer will follow from the Lax-Milgram Lemma, if one can establish an *a priori* estimate for the PDE.

4. Discretization

Choose $\mathbf{X}^h \subset \mathbf{X}$ and then solve the problem

Seek $U^h \in \mathbf{X}^h$ such that

$$Q(U^h; V^h) = \mathcal{F}(V^h) \text{ for all } V^h \in \mathbf{X}^h.$$

- Typically, \mathbf{X} is a Sobolev space constrained by the boundary conditions.
- Discrete problem is equivalent to a linear system having *symmetric, positive definite* matrix.
- Approximations are optimally accurate.

Least Squares Strategy: How and Why

1. Transformation of the original PDE or system of PDEs to a *first order system*:

⇒ Discretization by C^0 finite elements may be possible.

⇒ Direct and optimal approximations of physically important fields, e.g., vorticity, stresses.

2. Identification of spaces \mathbf{X} and \mathbf{Y} such that an *a priori* estimate holds and, formulation of the LS functional for the (first order) system:

⇒ Existence and uniqueness of the minimizers

⇒ Stability of the discretizations is guaranteed by the inclusion $\mathbf{X}^h \subset \mathbf{X}$ and **inf-sup** (LBB) type conditions are not required for \mathbf{X}

⇒ Discretization results in linear systems with *symmetric, positive definite* matrices

⇒ Discrete equations can be solved by robust iterative methods (e.g., CG methods)

⇒ Assembly free methods are feasible.

Applications of least squares methods

- Least squares finite element methods are based on *minimization principle*:

⇒ very competitive when Galerkin formulation corresponds to a *saddle point* optimization, because inf-sup (LBB) condition is avoided.

- Boundary conditions can be imposed in a weak sense by including into the functional the term

$$\|\mathcal{R}U - G\|_{\Gamma}^2$$

⇒ Approximating functions need not satisfy the essential boundary conditions.

Examples of LS applications

- Stationary, incompressible flow
- Time dependent incompressible flow
- Convection-diffusion problems
- Purely hyperbolic problems

LS References

General

- 1970,73** - Bramble, Schatz (LS for $2m^{th}$ order BVP)
- 1973** - Baker (Simplified proofs for Bramble, et.al.)
- 1977** - Jespersen (LS for systems associated with elliptic PDE)
- 1979** - Glowinski et. al. (LS in H^{-1} norm)
- 1979** - Fix, Gunzburger, Nicolaides (LS for div - grad systems)
- 1979** - Wendland (LS for Petrovski elliptic systems)
- 1981** - Fix, Stephan (LS for domains with corners)
- 1985** - Aziz, Kellog, Stephens (LS for ADN systems)
- 1987** - Chang, Gunzburger (LSFEM for first order systems in 3-D)
- 1993** - Pehlivanov, Carey, Lazarov (LS for ODE's)
- 1994** - Lazarov, Manteuffel, McCormick (FOSLS: H_{div} norms)
- 1994** - Bramble, Lazarov, Pasciak (LS in H^{-1} norm)

Convection-diffusion and hyperbolic problems

1992 - Chen (semidiscrete LS)

1994 - Jiang (iteratively reweighted LS)

1994 - Tsang (LS for advective transport)

LS for incompressible viscous flows

1979 - Glowinski et. al. (LS for primitive variable Navier-Stokes in H^{-1} norms)

1990 - Chang, Jiang (LS for VVP Stokes)

1990 - Chang, Povinelli (LS in 3D)

1992 - Lefebvre, Peraire and Morgan (LS for VVP Navier-Stokes, 3D)

1993 - Bochev, Gunzburger (LS for VVP Stokes)

1993 - Bochev, Gunzburger (LS for velocity-pressure-stress Stokes)

1994 - Bochev (LS for VVP Navier-Stokes)

1994 - Bochev, Gunzburger (WLS for VVP Navier-Stokes)

Velocity-Vorticity-Pressure Navier-Stokes Equations

Differential equations in 2D

$$\nu \mathbf{curl} \omega + \mathbf{grad} r + \omega \times \mathbf{u} = \mathbf{f} \quad \text{in } \Omega \rightarrow 2$$

$$\mathbf{curl} \mathbf{u} - \omega = 0 \quad \text{in } \Omega \rightarrow 1$$

$$\mathbf{div} \mathbf{u} = 0 \quad \text{in } \Omega \rightarrow 1$$

Differential equations in 3D

$$\nu \mathbf{curl} \omega + \mathbf{grad} r + \omega \times \mathbf{u} = \mathbf{f} \quad \text{in } \Omega \rightarrow 3$$

$$\mathbf{curl} \mathbf{u} - \omega = 0 \quad \text{in } \Omega \rightarrow 3$$

$$\mathbf{div} \mathbf{u} = 0 \quad \text{in } \Omega \rightarrow 1.$$

Boundary conditions

$$\mathcal{R}U = 0 \quad \text{on } \Gamma$$

- \mathbf{u} - velocity
- 2D: $\omega = \mathbf{curl} \mathbf{u}$; 3D: $\omega = \mathbf{curl} \mathbf{u}$ - vorticity
- $r = p + 1/2|\mathbf{u}|^2$ - total head, ($p = \text{pressure}$)

$$\int_{\Omega} r \, d\Omega = 0$$

- $\nu = 1/Re$ - kinematic viscosity
- \mathbf{f} - body force

The ADN Theory

- Success of the LS method depends on the proper choice of the spaces \mathbf{X} and \mathbf{Y} (**coercivity!**)
- The spaces \mathbf{X} and \mathbf{Y} should be such that an *a priori* estimate holds for the linearized system
- The ADN theory identifies the spaces in which *a priori* estimates of the form $\|U\|_X \leq C\|\mathcal{L}U\|_Y$ hold for solutions of elliptic BVP
- The norms appearing in these estimates are chosen so that the operator \mathcal{L} and boundary operator \mathcal{R} satisfy a certain precise condition known as the *complementing condition*.
- The *complementing condition* guarantees that the boundary operator \mathcal{R} is compatible with the operator \mathcal{L} i.e., that the BVP is well-posed in the spaces \mathbf{X} and \mathbf{Y} .
- Different boundary conditions may result in different a priori estimates, i.e., the choice of \mathbf{X} and \mathbf{Y} may depend on \mathcal{R} .

In two-dimensions:

- Four equations and unknowns
- Elliptic system of total order 4
- Needs 2 conditions on Γ ; same as for the (\mathbf{u}, p) Navier-Stokes in 2D: *can use the same \mathcal{R} .*

In three-dimensions

- Seven equations and unknowns;
- The VVP system cannot be elliptic in the sense of Agmon, Douglas and Nirenberg because $\det \mathcal{L}^P(\xi + \tau \xi') = 0$ will have a real root.

To derive a well-posed system:

- We add a seemingly redundant equation and a new “slack” variable ϕ :

$$\operatorname{div} \omega = 0 \quad \text{in } \Omega;$$

$$\mathbf{curl} \mathbf{u} - \omega + \mathbf{grad} \phi = 0 \quad \text{in } \Omega.$$

- Elliptic system of total order 8
- Needs 4 conditions on Γ ; one more than the (\mathbf{u}, p) Navier-Stokes in 3D: *must supplement \mathcal{R} by an additional condition on Γ !*

Principal parts and a priori estimates

s_1	ω_y	r_x	0	0
s_2	$-\omega_x$	r_y	0	0
s_3	$-\omega$	0	$-u_{1y}$	u_{2x}
s_4	0	0	u_{1x}	u_{2y}
s/t	t_1	t_2	t_3	t_4

- $s_i \leq 0$ determine norms for the data
- $t_j \geq 0$ determine norms for the solution
- $\deg \mathcal{L}_{ij} \leq s_i + t_j$
- $\mathcal{L}^P = \{\mathcal{L}_{ij} \mid \deg \mathcal{L}_{ij} = s_i + t_j\}$

The ADN a priori estimates

- \mathcal{L}^P is uniformly elliptic
- \mathcal{L}^P satisfies the Supplementary Condition (2D)
- BCs satisfy the Complementing Condition

$$\|U\|_X = \sum_{i=1}^n \|v_j\|_{t_j} \leq C \sum_{i=1}^n \|f_i\|_{-s_i} = C \|\mathcal{L}U\|_Y$$

Classification of the Boundary Conditions

The principal part \mathcal{L}^P of the VVP linearized equations is not unique! There are two sets of indices such that \mathcal{L}_1^P and \mathcal{L}_2^P are *uniformly elliptic*.

Type 1 BC: CC holds with

$$\mathcal{L}_1^P = \begin{pmatrix} \mathbf{curl} \omega + \mathbf{grad} r \\ \mathbf{curl} \mathbf{u} \\ \mathbf{div} \mathbf{u} \end{pmatrix}$$

$$\underbrace{s_1 = s_2 = 0}_{\text{momentum}}, \quad s_3 = s_4 = 0 \quad \text{in 2D},$$

$$\underbrace{s_1 = \dots = s_4 = 0}_{\text{momentum and redundant}}; \quad s_5 = \dots = s_8 = 0 \quad \text{in 3D}.$$

Example: $\mathcal{R}U = (\mathbf{u} \cdot \mathbf{n}, r)$

Type 2 BC: CC holds with

$$\mathcal{L}_2^P = \begin{pmatrix} \mathbf{curl} \omega + \mathbf{grad} r \\ \boxed{-\omega} + \mathbf{curl} \mathbf{u} \\ \mathbf{div} \mathbf{u} \end{pmatrix}$$

$$\underbrace{s_1 = s_2 = 0}_{\text{momentum}}, \quad s_3 = s_4 = -1 \quad \text{in 2D},$$

$$\underbrace{s_1 = \dots = s_4 = 0}_{\text{momentum and redundant}}; \quad s_5 = \dots = s_8 = -1 \quad \text{in 3D}.$$

Example: $\mathcal{R}U = \mathbf{u}$

The Modified Navier-Stokes Equations:

In two-dimensions

Let ω_0 , r_0 and \mathbf{u}_0 solve

$$\begin{aligned} \mathbf{curl} \omega + \frac{1}{\nu} \mathbf{grad} r &= \frac{1}{\nu} \mathbf{f} && \text{in } \Omega \\ \mathbf{curl} \mathbf{u} - \omega &= 0 && \text{in } \Omega \\ \mathbf{div} \mathbf{u} &= 0 && \text{in } \Omega \\ \mathcal{R}U &= 0 && \text{on } \Gamma \end{aligned}$$

Least squares methods will be formulated for

$$\begin{aligned} \mathbf{curl} \omega + \mathbf{grad} r + \nu^{-1}(\omega_0 \times \mathbf{u}_0) + \\ \nu^{-1}(\omega_0 \times \mathbf{u} + \omega \times \mathbf{u}_0 + \omega \times \mathbf{u}) &= 0 && \text{in } \Omega \\ \mathbf{curl} \mathbf{u} - \omega &= 0 && \text{in } \Omega \\ \mathbf{div} \mathbf{u} &= 0 && \text{in } \Omega \\ \mathcal{R}U &= 0 && \text{on } \Gamma \end{aligned}$$

In three-dimensions

The “redundant” equation

$$\mathbf{div} \omega = 0$$

must be added for the *stability* of the method. The slack variable is identically zero and can be ignored

LS for the Navier-Stokes Equations

Least squares functional:

$$\begin{aligned}
 J(\omega, \mathbf{u}, r) &= \\
 &= \frac{1}{2} \left(\|\mathbf{curl} \omega + a(\omega, \mathbf{u}) + b(\omega, \mathbf{u}, \omega_0, \mathbf{u}_0) + \mathbf{grad} r\|_0^2 \right. \\
 &\quad \left. + \|\mathbf{curl} \mathbf{u} - \omega\|_s^2 + \|\mathbf{div} \mathbf{u}\|_s^2 \right),
 \end{aligned}$$

where $s = 0, 1$ for Type 1,2 BCs and

$$\begin{aligned}
 b(\omega, \mathbf{u}, \xi, \mathbf{v}) &= \frac{1}{\nu} (\omega \times \mathbf{u} + \xi \times \mathbf{v}) \\
 a(\omega, \mathbf{u}) &= b(\omega_0, \mathbf{u}, \omega, \mathbf{u}_0).
 \end{aligned}$$

Euler-Lagrange equation

Seek $(\omega, \mathbf{u}, r) \in \mathbf{X}_s$ such that

$$\begin{aligned}
 Q((\omega, \mathbf{u}, r); (\xi, \mathbf{v}, q)) &= \\
 &= (\mathbf{curl} \omega + \mathbf{grad} r + a(\omega, \mathbf{u}) + b(\omega_0, \mathbf{u}_0, \omega, \mathbf{u}), \\
 &\quad \mathbf{curl} \xi + \mathbf{grad} q + a(\xi, \mathbf{v}) + b(\xi, \mathbf{u}, \omega, \mathbf{v}))_0 \\
 &\quad + (\mathbf{curl} \mathbf{u} - \omega, \mathbf{curl} \mathbf{v} - \xi)_s \\
 &\quad + (\mathbf{div} \mathbf{u}, \mathbf{div} \mathbf{v})_s = 0 \quad \text{for all } (\xi, \mathbf{v}, q) \in \mathbf{X}_s.
 \end{aligned}$$

where

$$\mathbf{X}_s = [H^1(\Omega) \times \tilde{H}^1(\Omega) \times H^{s+1}(\Omega)^2] \cap [\mathcal{R}U = 0].$$

Least Squares Finite Element Method

1. Choose the discrete space $\mathbf{X}_s^h \subset \mathbf{X}_s$:
2. Solve the Euler-Lagrange equation

Seek $U^h = (\omega^h, \mathbf{u}^h, r^h) \in \mathbf{X}_s^h$ such that

$$Q(U^h; V^h) = 0 \quad \forall V^h = (\xi^h, \mathbf{v}^h, q^h) \in \mathbf{X}_s^h.$$

Implementation Issues

The Euler-Lagrange equation is a nonlinear system that must be solved in an iterative manner.

Newton's method:

- Locally has quadratic convergence
- In a neighborhood of a minimizer the Hessian is symmetric and positive definite.

\implies Continuation methods are required to get an initial approximation inside the attraction ball:

- **Continuation along the constant:** simple, but cannot handle turning points
- **Continuation along the tangent:** can be made to handle turning points.

Navier-Stokes Equations: Advantages of the Least Squares Methods

- Approximating spaces are not subject to the **LBB** (**inf-sup**) condition
- All unknowns can be approximated by the *same finite element space*
- Newton linearization results in *symmetric, positive definite* linear systems, at least in the neighborhood of a solution:
 - ⇒ Using a properly implemented continuation (with respect to the Reynolds number) techniques, a solution method can be devised that will only encounter symmetric, positive definite linear systems in the solution process
 - ⇒ Robust iterative methods can be used
 - ⇒ A solution method that is *assembly free* even at an element level can be devised
- No *artificial boundary conditions* for ω need be introduced at boundaries at which \mathbf{u} is specified
- Accurate vorticity approximations are obtained.

Error Analysis of the Least Squares Method: Abstract Approximation Theory

Abstract problem (Girault, Raviart, 1984)

$$F(\lambda, U) \equiv U + T \cdot G(\lambda, U) = 0,$$

where $\Lambda \subset \mathbb{R}$ is compact interval; \mathbf{X} and \mathbf{Y} are Banach spaces and $T \in L(\mathbf{Y}, \mathbf{X})$.

Regular branch of solutions

Assume that $\{(\lambda, U(\lambda)) \mid \lambda \in \Lambda\}$ is such that

$$F(\lambda, U(\lambda)) = 0 \quad \text{for } \lambda \in \Lambda.$$

1. The set $\{(\lambda, U(\lambda)) \mid \lambda \in \Lambda\}$ is called *branch of solutions* if the map $\lambda \rightarrow U(\lambda)$ is a continuous function from Λ into \mathbf{X}
2. The set $\{(\lambda, U(\lambda)) \mid \lambda \in \Lambda\}$ is called *regular branch* if $D_U F(\lambda, U(\lambda))$ is an isomorphism from \mathbf{X} into \mathbf{X} for all $\lambda \in \Lambda$.

Discretizations

- Choose a discrete subspace $\mathbf{X}^h \subset \mathbf{X}$;
- Choose $T_h \in L(\mathbf{Y}, \mathbf{X}^h)$ to be an approximating operator for the linear part T of F .
- Then, consider the approximate problem

$$F^h(\lambda, U^h) \equiv U^h + T_h \cdot G(\lambda, U^h) = 0.$$

- Approximation in F^h is introduced by approximating only the linear operator T :

$\implies F^h$ has the same differentiability properties as the nonlinear map F .

Abstract approximation result

We make the following assumptions:

A1. $\{(\lambda, \tilde{U}(\lambda)) | \lambda \in \Lambda\}$ is a branch of regular solutions;

A2. G is a C^2 mapping $G : \Lambda \times \mathbf{X} \mapsto \mathbf{Y}$;

A3. All second derivatives $D_U^2 G$ are bounded on all bounded subsets of $\Lambda \times \mathbf{X}$;

A4. There exists a space $\mathbf{Z} \subset \mathbf{Y}$, with a continuous imbedding, such that

$$D_U G(\lambda, U) \in L(\mathbf{X}, \mathbf{Z}) \quad \forall U \in \mathbf{X};$$

A5. The operator T_h satisfies conditions

$$\begin{aligned} \lim_{h \rightarrow 0} \|(T - T_h)g\|_X &= 0 \quad \forall g \in \mathbf{Y}; \\ \lim_{h \rightarrow 0} \|T - T_h\|_{L(Z, X)} &= 0. \end{aligned}$$

Then, for h sufficiently small there exists a unique C^2 function $\lambda \mapsto U^h \in \mathbf{X}^h$, s.t. $\{(\lambda, U^h(\lambda)) | \lambda \in \Lambda\}$ is a branch of regular solutions of $F^h(\lambda, U^h) = 0$ and

$$\|U^h - U\|_X \leq C \|(T - T_h) \cdot G(\lambda, U)\|_X \quad \forall \lambda \in \Lambda$$

Application to the least squares method

$$\begin{aligned}
 Q((\omega, \mathbf{u}, r); (\xi, \mathbf{v}, q))_s &= \\
 &= (\mathbf{curl} \omega + \mathbf{grad} r, \mathbf{curl} \xi + \mathbf{grad} q)_0 \\
 &\quad + (\mathbf{curl} \mathbf{u} - \omega, \mathbf{curl} \mathbf{v} - \xi)_s + (\mathbf{div} \mathbf{u}, \mathbf{div} \mathbf{v})_s .
 \end{aligned}$$

$$\mathbf{Y}_s = H^{-1}(\Omega) \times H^{-1}(\Omega) \times H^{-1-s}(\Omega)^2$$

The operator T/T_h :

$\forall g \in \mathbf{Y}_s ; Tg = U \in \mathbf{X}_s$ if and only if U solves

Seek $U \in \mathbf{X}_s$ such that

$$Q(U; V)_s = (g, V) \quad \forall V \in \mathbf{X}_s .$$

For T_h take $U^h, V^h \in \mathbf{X}_s^h$. T and T_h are the Stokes LS solution operator and its discretization.

The operator G :

$\forall U \in \mathbf{X}_s ; G(\lambda, U) = g \in \mathbf{Y}_s$ if and only if

$$\begin{aligned}
 &(\mathbf{curl} \omega + \mathbf{grad} r + a(\omega, \mathbf{u}) + b(\omega, \mathbf{u}, \omega_0, \mathbf{u}_0), \\
 &\quad b(\xi, \mathbf{u}_0, \xi, \mathbf{u}) + b(\omega_0, \mathbf{v}, \omega, \mathbf{v}))_0 \\
 &+ (a(\omega, \mathbf{u}) + b(\omega, \mathbf{u}, \omega_0, \mathbf{u}_0), \mathbf{curl} \xi + \mathbf{grad} q)_0 \\
 &= (g_1, \xi)_0 + (g_2, \mathbf{v})_0 + (g_3, q)_0 \quad \forall (\xi, \mathbf{v}, q) \in \mathbf{X}_s .
 \end{aligned}$$

With the identification

$$U = (\omega, \mathbf{u}, r), \quad U^h = (\omega^h, \mathbf{u}^h, r^h) \quad \text{and} \quad \lambda = \frac{1}{\nu}$$

the Euler-Lagrange equation and its discretization can be cast into the canonical forms

$$U + T \cdot G(\lambda, U) = 0 \quad \text{and} \quad U^h + T_h \cdot G(\lambda, U^h) = 0$$

Concerning the error estimates one can show that the assumptions [A1.] - [A5.] hold. As a result one can prove the following:

Theorem 1 *Assume that $\{(\lambda, \tilde{U}(\lambda)) | \lambda \in \Lambda\}$ is a regular branch of (sufficiently smooth) solutions. Then, for h sufficiently small, the discrete Euler-Lagrange equation has a unique branch $\{(\lambda, U^h(\lambda)) | \lambda \in \Lambda\}$ of regular solutions such that*

$$\begin{aligned} \|\omega(\lambda) - \omega^h(\lambda)\|_1 &+ \|\mathbf{u}(\lambda) - \mathbf{u}^h(\lambda)\|_{1+s} \\ &+ \|r(\lambda) - r^h(\lambda)\|_1 \\ &\leq Ch (\|\omega(\lambda)\|_2 + \|\mathbf{u}(\lambda)\|_{2+s} + \|r(\lambda)\|_2) \end{aligned}$$

where $s = 0$ for Type 1 and $s = 1$ for Type 2 BCs.

Table 1: Boundary Conditions

	Boundary conditions	3D	2D	Type
BC1	Velocity	\mathbf{u}	\mathbf{u}	2
	Slack variable	ϕ	-	
BC1A	Velocity	\mathbf{u}	\mathbf{u}	2
	Normal vorticity	$\boldsymbol{\omega} \cdot \mathbf{n}$	-	
BC2	Pressure	p	p	1
	Normal velocity	$\mathbf{u} \cdot \mathbf{n}$	$\mathbf{u} \cdot \mathbf{n}$	
	Normal vorticity	$\boldsymbol{\omega} \cdot \mathbf{n}$	-	
	Slack variable	ϕ	-	
BC2A	Pressure	p	p	not well posed in 3D 1 in 2D
	Normal velocity	$\mathbf{u} \cdot \mathbf{n}$	$\mathbf{u} \cdot \mathbf{n}$	
	Tangential vorticity	$\mathbf{n} \times \boldsymbol{\omega} \times \mathbf{n}$	-	
BC3	Pressure	p	p	1 in 2D 2 in 3D
	Tangential velocity	$\mathbf{n} \times \mathbf{u} \times \mathbf{n}$	$\mathbf{u} \cdot \mathbf{t}$	
	Slack variable	ϕ	-	
BC3A	Pressure	p	p	1
	Tangential velocity	$\mathbf{n} \times \mathbf{u} \times \mathbf{n}$	$\mathbf{u} \cdot \mathbf{t}$	
	Normal vorticity	$\boldsymbol{\omega} \cdot \mathbf{n}$	-	
BC4	Normal velocity	$\mathbf{u} \cdot \mathbf{n}$	$\mathbf{u} \cdot \mathbf{n}$	1
	Tangential vorticity	$\mathbf{n} \times \boldsymbol{\omega} \times \mathbf{n}$	$\boldsymbol{\omega}$	
	Slack variable	ϕ	-	
BC5	Tangential velocity	$\mathbf{n} \times \mathbf{u} \times \mathbf{n}$	$\mathbf{u} \cdot \mathbf{t}$	1
	Tangential vorticity	$\mathbf{n} \times \boldsymbol{\omega} \times \mathbf{n}$	$\boldsymbol{\omega}$	
BC6	Vorticity	$\boldsymbol{\omega}$	$\boldsymbol{\omega}$	not well posed
	Pressure	p	p	