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Discontinuous Galerkin Finite Element Method for Wave Equations with Time-dependent Coefficients

Master Thesis

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Abstract

The symmetric interior penalty discontinuous Galerkin finite element method is presented for a second order elliptical problem in 1d which yields a symmetric, positive definite bilinear form for a sufficiently positive penalty parameter. This guarantees existence of and uniqueness of the discrete problem. Implementation and theoretical details are discussed. Numerical results confirm the optimal convergence rate of $\mathcal{O}(h^{r+1})$ in the L^2 -norm for \mathcal{P}^r -elements.

0.1 Introduction

The classical continuous Finite Element Method (FEM) is a widely used and very powerful

0.2 Overview

Chapter 1

DG for an Elliptic Problem

First we consider a time-independent elliptic problem. Not only is it useful for initiation to the subject to first consider a simpler elliptic problem, but it is also an essential preparational step in deriving the SIPG bilinear form for the elliptic part of the hyperbolic problem as well.

The goal of this chapter is to build all the necessary theoretical and practical tools to solve a given elliptic problem numerically and then experimentally test the method for different parameters. We will define the necessary notation and derive the SIPG variational formulation as well as in detail describe further implementation steps as for example what basis of the finite element space we chose and how to derive local matrices. Finally this chapter will also include some well established theoretical results in the context of discontinuous Galerkin methods. The derivation of the bilinear form is inspired by Chapter 1 in [7] as well as [2] and [4] for cross reference.

1.1 Problem

We consider the following elliptic model problem:

$$-(c(x)u'(x))' = f(x) \quad \forall x \in \Omega \quad (1.1)$$

$$u(0) = g_0, u(1) = g_1 \quad (1.2)$$

where $\Omega = (0, 1)$ is the domain, $g_0, g_1 \in \mathbb{R}$ are Dirichlet boundary conditions, $f \in L^2(\Omega)$ and $c \in C^1(\Omega)$ satisfies:

$$c_{\min} \leq c(x) \leq c_{\max} \quad \forall x \in \Omega$$

for $0 < c_{\min} \leq c_{\max}$. Multiplying the solution by a test function and integrating by parts over Ω we get the standard weak formulation:

Find $u \in \{v \in H^1(\Omega) \mid v(0) = g_0, v(1) = g_1\}$ such that:

$$a(u, v) = (f, v)_{L^2(\Omega)} \quad \forall v \in C_c^\infty(\Omega) \quad (1.3)$$

Where

$$a : H^1(\Omega) \times H^1(\Omega) \rightarrow \mathbb{R}, \quad (u, v) \mapsto \int_\Omega c(x)u'(x)v'(x)dx$$

defines the standard elliptic bilinear form on $H^1(\Omega)$ and

$$(u, v)_{L^2(\Omega)} = \int_\Omega uv dx$$

denotes the L^2 -inner product.

If we assume $f \in L^2(\Omega)$ and $c \in C^1(\Omega)$ we know by *Lax-Milgram* that there exists a unique (weak) solution of (1.3) and from the *elliptic regularity theory* we can conclude higher regularity of the weak solution, namely $u \in H^2(\Omega)$. For reference see chapter 6.2, 6.3 in [1].

1.2 Discretization

We will now discretize the domain Ω , note that the notation will persist throughout this thesis. Let $0 = x_0 < \dots < x_{N+1} = 1$ be the mesh faces, $I_n = (x_n, x_{n+1})$ for $n = 0, \dots, N$ be the elements and $\mathcal{T}_h = \{I_n\}_{n=0}^N$ a partition of Ω for some fixed $N \in \mathbb{N}$. Sometimes we will write N_h to emphasize the dependency on h . We denote the element length by $h_n = x_{n+1} - x_n$ for $n = 0, \dots, N$ and the global meshsize by $h = \max_n h_n$. Next we define the discontinuous finite element space

$$V_h^r(\mathcal{T}_h) = \{v \in L^2(\Omega) \mid v|_{I_n} \in \mathcal{P}^r(I_n)\}, \quad (1.4)$$

where $\mathcal{P}^r(I_n)$ denotes the space of polynomials $p : I_n \rightarrow \mathbb{R}$ of degree r for $r \in \mathbb{N}$. When the context allows it, we will denote the finite element space with just V_h for simplicity. V_h is our final approximation space in which the numerical solution lays. We observe that in contrast to a continuous finite element approximation space here the resulting solution is a priori discontinuous by construction. Furthermore we have here $V_h \not\subset H^1(\Omega)$. This is especially apparent in 1d due to the Sobolev embedding $H^1(\Omega) \subset C^0(\Omega)$. Any discontinuous element of V_h can therefore not be in $H^1(\Omega)$.

To proceed we will require the following trace operators:

Definition 1.1. Let $v : \Omega \rightarrow \mathbb{R}$ be piecewise continuous and let $n \in \{1, \dots, N\}$

(i) We denote

$$v(x_n^+) := \lim_{x \searrow x_n} v(x), \quad v(x_n^-) := \lim_{x \nearrow x_n} v(x)$$

the limit from above/below.

(ii) We define the **jump** at x_n as

$$[v(x_n)] := v(x_n^-) - v(x_n^+)$$

and the **average** at x_n as

$$\{\{v(x_n)\}\} := \frac{v(x_n^+) + v(x_n^-)}{2},$$

furthermore by convention we set:

$$[v(x_0)] := -v(x_0^+), \quad [v(x_{N+1})] := v(x_{N+1}^-), \quad \{\{v(x_0)\}\} := v(x_0^+), \quad \{\{v(x_{N+1})\}\} := v(x_{N+1}^-).$$

1.3 Variational Formulation

To derive the SIPG variational formulation, let $v \in V_h$ be a test function. As mentioned in section 1.1 we can assume that the coefficient $c \in C^1(\Omega)$ and the exact solution $u \in H^2(\Omega) \subset C^1(\Omega)$. Due to the discontinuity of the test function in contrast to continuous FEM we multiply u with v on each element I_n and integrate by parts locally

$$\int_{x_n}^{x_{n+1}} f v \, dx = - \int_{x_n}^{x_{n+1}} (cu')' v \, dx = \int_{x_n}^{x_{n+1}} cu' v' \, dx - cu' v \Big|_{x_n}^{x_{n+1}} \quad \forall n \in \{0, \dots, N\},$$

then sum over all elements

$$(f, v)_{L^2(\Omega)} = \sum_{n=0}^N \int_{I_n} cu' v' \, dx - \sum_{n=0}^{N+1} [c(x_n) u'(x_n) v(x_n)], \quad (1.5)$$

where we have used that $\sum_{n=0}^N w \Big|_{x_n}^{x_{n+1}} = w(x_{N+1}^-) - w(x_N^+) + w(x_N^-) - \dots - w(x_1^+) + w(x_1^-) - w(x_0^+) = \sum_{n=0}^{N+1} [w(x_n)]$ for any piece-wise continuous function w .

By our construction c, u' are continuous on Ω , this means

$$[c(x_n)u'(x_n)v(x_n)] = c(x_n)u'(x_n)[v(x_n)] = \{c(x_n)u'(x_n)\}[v(x_n)] \quad \forall n \in \{0, \dots, N+1\}, \quad (1.6)$$

and

$$[u(x_n)] = 0 \quad \forall n \in \{1, \dots, N\}. \quad (1.7)$$

To derive the final variational form we will now have to add two additional terms to (1.5):

Step 1. Firstly we need to symmetrize our currently non-symmetrical right hand side which will correspond to the SIPG bilinear form. To do so we add the symmetry term $-\sum_{n=0}^{N+1} \{c(x_n)v'(x_n)\}[u(x_n)]$ on both sides of (1.5) so we get

$$\begin{aligned} & (f, v)_{L^2(\Omega)} - g_1 c(x_{N+1}^-) v'(x_{N+1}^-) + g_0 c(x_0^+) v'(x_0^+) \\ &= \sum_{n=0}^N \int_{I_n} c u' v' dx - \sum_{n=0}^{N+1} \{c(x_n)u'(x_n)\}[v(x_n)] + \{c(x_n)v'(x_n)\}[u(x_n)], \end{aligned}$$

where on the left hand side of the equation we have applied (1.7) for the interior node contributions of the sum (which therefore vanish), and the boundary condition (1.2) ensuring the left hand side to be solely dependent on v .

Step 2. The bilinear form we seek to create will (for now) be defined on $V_h \times V_h$ meaning it will intake discontinuous functions. In particular the numerical solution will be a discontinuous function whereas the exact solution is continuous. To counterweigh this discrepancy we need to introduce a penalization mechanism, to minimize discontinuous behavior. Technically speaking this penalization term will guarantee coercivity of the bilinear form (see section 1.9).

Let $\sigma > 0$ constant, we define:

$$c_n := \begin{cases} \max(c(x_n^+), c(x_n^-)), & n = 1, \dots, N \\ c(x_n^+), & n = 0 \\ c(x_n^-), & n = N+1 \end{cases}, \quad h_n := \begin{cases} \min(h_n, h_{n-1}), & n = 1, \dots, N \\ h_n, & n \in \{0, N+1\} \end{cases}$$

with this we define our penalization parameter

$$a_n := \frac{\sigma c_n}{h_n} > 0 \quad \forall n \in \{0, \dots, N+1\}. \quad (1.8)$$

Similarly to Step 1 we can now add the penalty term $\sum_{n=0}^{N+1} a_n [u(x_n)][v(x_n)]$ on both sides of (1.5) and get the final *discrete* SIPG variational formulation.

Find $u_h \in V_h$ such that:

$$b_h(u_h, v) = \ell_h(v), \quad \forall v \in V_h, \quad (1.9)$$

where

$$\begin{aligned} b_h(u, v) &= \sum_{n=0}^N \int_{I_n} c u' v' dx - \sum_{n=0}^{N+1} \{c(x_n)u'(x_n)\}[v(x_n)] + \{c(x_n)v'(x_n)\}[u(x_n)] + \sum_{n=0}^{N+1} a_n [u(x_n)][v(x_n)], \\ \ell_h(v) &= (f, v)_{L^2(\Omega)} - g_1 c(x_{N+1}^-) v'(x_{N+1}^-) + g_0 c(x_0^+) v'(x_0^+) + a_{N+1} g_1 v(x_{N+1}^-) + a_0 g_0 v(x_0^+) \end{aligned}$$

for $u, v \in V_h$.

Remark 1.2. In higher dimensions ($d \in \{2, 3\}$) there are no point values appearing in the variational formulation, instead each point value evaluation becomes an integral over an element face. Since in practice higher dimensional problems are more interesting we quickly show how the variational formulation would

look like in arbitrary dimensions.

Let \mathcal{F}_h denote the set of all faces of the partition of a domain $\Omega \in \mathbb{R}^d$ and \mathcal{T}_h denote the set of all elements. The bilinear form and the linear form become

$$b_h(u, v) = \sum_{K \in \mathcal{T}_h} \int_K c \nabla u \cdot \nabla v \, dx - \sum_{F \in \mathcal{F}_h} \int_F \{c \nabla u\} \cdot [v] + \{c \nabla v\} \cdot [u] \, dS + \sum_{F \in \mathcal{F}_h} \int_F \mathbf{a}[u][v] \, dS,$$

$$\ell_h(v) = (f, v)_{L^2(\Omega)} - \sum_{F \in \mathcal{F}_h \cap \partial\Omega} \int_F [g] \cdot \{c \nabla v\} \, dS + \sum_{F \in \mathcal{F}_h \cap \partial\Omega} \int_F \mathbf{a}[g][v] \, dS.$$

1.4 Boundary Conditions

By adding the terms $-\sum_{n=0}^{N+1} \{c(x_n)v'(x_n)\}[u(x_n)]$, $\sum_{n=0}^{N+1} \mathbf{a}_n[u(x_n)][v(x_n)]$ on both sides of (1.5) we weakly imposed the Dirichlet boundary conditions into the variational form. This stands in contrast to how boundary conditions are usually imposed in continuous FEM. Indeed one could also impose them strongly, meaning we could define

$$V_h^r(\mathcal{T}_h) = \{v \in L^2(\Omega) | v|_{I_n} \in \mathcal{P}^r(I_n), v(x_0) = g_0, v(x_{N+1}) = g_1\},$$

but this solely as a side note, we will continue to work with purely weakly imposed boundary conditions.

One could alternatively desire to impose *Neumann* boundary conditions weakly, this slightly changes the variational formulation. We illustrate the idea on the following example boundary condition. A solution u should satisfy:

$$u(0) = g_0, \quad u'(1) \cdot n_1 = g_1,$$

where again $g_0, g_1 \in \mathbb{R}$ are the boundary values and n_1 denotes the outward normal of the domain at the upper boundary. In 1d we trivially have $n_1 = 1, n_0 = -1$, where n_0 denotes the outward normal at the lower boundary. Note that we need at least one Dirichlet boundary condition, else there exists no unique solution.

Now recall the initial incomplete formulation (1.5). First we take the Neumann boundary contribution $\{c(x_{N+1})u'(x_{N+1})\}[v(x_{N+1})]$ to the other side of the equation. We get

$$\sum_{n=0}^N \int_{I_n} cu'v' \, dx - \sum_{n=0}^N \{c(x_n)u'(x_n)\}[v(x_n)] = (f, v)_{L^2(\Omega)} + g_1 c(x_{N+1}^-)v'(x_{N+1}^-)$$

We have used $\{c(x_{N+1})u'(x_{N+1})\}[v(x_{N+1})] = c(x_{N+1}^-)u'(x_{N+1}^-)v(x_{N+1}^-) \cdot n_1 = g_1 c(x_{N+1}^-)v(x_{N+1}^-)$.

From here on we proceed similarly as in the Dirichlet case. The main difference is that we always omit the boundary face with the Neumann boundary condition.

We add the symmetry and penalty terms

$$-\sum_{n=0}^N \{c(x_n)v'(x_n)\}[u(x_n)] + \sum_{n=0}^N \mathbf{a}_n[u(x_n)][v(x_n)]$$

to both sides, using again that the real solution has zero jump on the interior faces and applying the boundary conditions we finally derive the variational form

$$\begin{aligned} & \sum_{n=0}^N \int_{I_n} cu'v' \, dx - \sum_{n=0}^N \{c(x_n)u'(x_n)\}[v(x_n)] + \{c(x_n)v'(x_n)\}[u(x_n)] + \sum_{n=0}^N \mathbf{a}_n[u(x_n)][v(x_n)] \\ &= (f, v)_{L^2(\Omega)} + g_0 c(x_0^+)v'(x_0^+) + \mathbf{a}_0 g_0 v(x_0^+) + g_1 c(x_{N+1}^-)v(x_{N+1}^-) \end{aligned}$$

1.5 Matrix-Vector System

We will now transform the variational form (1.9) into Matrix-Vector system form. To do so let $r \in \mathbb{N}$ denote the polynomial degree and consequently the element degree of freedom. Note that in this thesis we will only consider global polynomial degrees, meaning one set polynomial degree for all elements. Next let $\{\Phi_0, \dots, \Phi_M\}$ be a basis of V_h , where $M = \dim(V_h)$. We can represent the sought Galerkin approximation as $u_h = \sum_{j=0}^M \alpha_j \Phi_j \in V_h$ for coefficients $\alpha_j \in \mathbb{R}$. Then (1.9) is equivalent to:

$$\sum_{j=0}^M \alpha_j b_h(\Phi_j, \Phi_i) = \ell_h(\Phi_i) \quad \forall i = 0, \dots, M,$$

which corresponds to the system:

$$\mathbf{B}\mathbf{u} = \mathbf{l} \tag{1.10}$$

for $\mathbf{B} \in \mathbb{R}^{M \times M}$, $[\mathbf{B}]_{i,j} = b_h(\Phi_j, \Phi_i)$, $\mathbf{u} \in \mathbb{R}^M$, $[\mathbf{u}]_j = \alpha_j$, $\mathbf{l} \in \mathbb{R}^M$, $[\mathbf{l}]_j = \ell_h(\Phi_j)$.

1.6 Basis of Finite Element Space

There are many ways of choosing basis functions for finite element spaces. In this thesis we solely focus on elementwise nodal Lagrangian basis functions, although there are alternatives which might be just as valid depending on the goal one pursues, like for example a modal Legendre basis. Having decided to work with a Lagrange nodal basis means we need to decide on the placing of the nodes. In principle for a desired polynomial degree r we could decide any $r+1$ points per elements. Since we restrict ourselves to a single global polynomial degree (i.e. no p-adaptivity) it makes sense to choose the same type of placement for all elements since this will allow us to calculate the actual function values on a single reference element (see below).

Here in contrast to continuous FEM, DG provides additional liberty, since there are no shared nodes over element faces we actually could decide to place the basis nodes purely inside of the element.

In this thesis we will not try to justify the choice of basis function too much and use Gauss-Lobatto quadrature nodes as basis nodes. This is a commonly used nodal basis in continuous FEM and we will also use it for our DG method. In Appendix A.2.3 of [5] the *Fekete points*, which in 1d coincide with the Gauss-Lobatto points, are recommended with the argument that this nodal basis yields a mass matrix with an optimal condition number. For more detailed information on choosing basis functions and alternative approaches see for example Appendix A.2 in [5].

Let $n \in \{1, \dots, N\}$ and $I_n \in \mathcal{T}_h$ be an arbitrary element. We denote $\hat{I} = (-1, 1)$ the *reference element* and $F_n : \hat{I} \rightarrow I_n$, $\xi \mapsto \frac{x_n + x_{n+1}}{2} + \frac{h_n}{2}\xi$ the *element map*. This now allows us to define a basis on the reference element and extend it to all elements using the element map.

For a fixed polynomial degree $r \geq 2$ let $\xi_0, \dots, \xi_r \in [-1, 1]$ be the Gauss-Lobatto nodes.

$$\begin{array}{c|c} r = 2 & \{-1, 1\} \\ r = 3 & \{-1, 0, 1\} \\ r = 4 & \{-1, -\frac{1}{\sqrt{5}}, \frac{1}{\sqrt{5}}, 1\} \\ \vdots & \vdots \end{array}$$

The inner nodes are given by the roots of L'_{r-1} , the derivative of the $r-1$ -th Legendre polynomial. We define the basis on the reference element as the Lagrangian nodal basis

$$\widehat{\phi}_i(\xi) := \prod_{\substack{j=0 \\ j \neq i}}^r \frac{\xi - \xi_j}{\xi_i - \xi_j}, \quad \text{for } i = 0, \dots, r \tag{1.11}$$

and define the basis functions on the element I_n as

$$\phi_i^n : I_n \rightarrow \mathbb{R}, \quad \phi_i^n(x) := \hat{\phi}_i(F_n^{-1}(x))$$

as a last step we extend the basis functions to the whole domain Ω by zero

$$\Phi_i^n : \Omega \rightarrow \mathbb{R}, \quad \Phi_i^n(x) := \begin{cases} \phi_i^n(x), & \text{for } x \in I_n \\ 0, & \text{else} \end{cases} \quad (1.12)$$

for $n = 0, \dots, N$ and $i = 0, \dots, r$. Clearly we have $\text{span}(\hat{\phi}_0, \dots, \hat{\phi}_r) = \mathcal{P}^r(\hat{I})$ and by extension $\text{span}(\Phi_0^0, \dots, \Phi_r^N) = V_h^r(\mathcal{T}_h)$. It is essential that our basis has only local support, meaning the basis functions are zero on most of the domain. This is the key property which allows the final matrices to be sparse. Choosing basis functions with global support, the computational cost would be unfeasible for small mesh sizes.

By having chosen a Lagrangian nodal basis the mesh nodes exactly coincide with the Gauss-Lobatto nodes on each element. To simplify the notation we introduce a *local-to-global* index map

$$T : \{0, \dots, N\} \times \{0, \dots, r\} \rightarrow \{1, \dots, M\} \quad (1.13)$$

where $M = (r+1)(N+1) = \dim(V_h)$. T takes an element index n and a local basis function index i as inputs and returns the globally assigned node index $T(n, i)$. T corresponds to the *connectivity matrix*. In the simplest case we have the global index ordered from left to right and get $T(n, i) = nr + i$.

1.7 Quadrature Rule

To discuss the assembly of the components of the system (1.10) in detail we will require a tool to approximate the integrals appearing in it, i.e. we need to choose a quadrature rule. The quadrature rule should primarily be as accurate as possible for the least amount of nodes necessary. So immediately one thinks of the Gauss-Legendre quadrature rule, which for $r+1$ quadrature nodes, is exact for polynomials of degree $2r+1$. In fact Gauss-Legendre is often the preferred choice due to its high accuracy. A downside of Gauss-Legendre is that it does not include element boundary nodes. We do require the values of our basis functions at the boundary nodes for the consistency, symmetry and penalty terms in (1.9), as well as the values of c at the element faces. All these values can be preloaded at the desired quadrature nodes, which makes it simpler to use a quadrature rule with nodes at the element boundary. This is why we choose the Gauss-Lobatto quadrature rule despite its lower accuracy of being exact for polynomials of degree $2r-1$ with $r+1$ quadrature nodes.

We could still use Gauss-Legendre and interpolate the element boundary values. Suppose we chose $r+1$ Gauss-Legendre quadrature nodes per element, then we require the element boundary values as well, so after all we need to preload the values of the basis functions and the coefficient c at $r+3$ nodes per element (the quadrature nodes plus the boundary nodes). If instead we choose $r+2$ Gauss-Lobatto quadrature nodes we achieve the same accuracy but use a node less.

An additional benefit of using Gauss-Lobatto quadrature nodes is that now the mesh nodes as well as the quadrature nodes are of the same type. This simplifies the program architecture and allows for the possibility of using *mass-lumping*, meaning we can let the basis nodes and the quadrature nodes coincide when assembling the mass matrix

$$[\mathbf{M}]_{i,j} = (\Phi_j, \Phi_i)_{L^2(\Omega)}.$$

When we use $r+1$ Gauss-Lobatto nodes to approximate the integral $\int_{\Omega} \Phi_j \Phi_i \, dx$ we introduce an error, since $\Phi_j \Phi_i \in \mathcal{P}^{2r}$, but the resulting mass-lumped mass matrix is diagonal, which is a crucial benefit in explicit time-marching schemes with continuous FEM spacial discretization. We compare the error created by using the inexact $r+1$ node Gauss-Lobatto quadrature rule for \mathcal{P}^r -elements with a higher order Gauss-Lobatto quadrature rule experimentally in 1.12.2, but we can already say that due to the decoupled element structure DG provides, the mass matrix is already block-diagonal and mass lumping provides no real benefit.

All of this being said the choice of the Gauss-Lobatto quadrature rule is in our case mainly preferential and also not the focus of this thesis.

1.8 Stiffness Matrix Assembly

With the basis functions in (1.12) defined we can now in detail investigate how to assemble the matrix \mathbf{B} in (1.10). To do so we firstly separate the bilinear form b_h into different components

$$\begin{aligned} a_h(u, v) &:= \sum_{n=0}^N \int_{I_n} c u' v' \, dx \\ b_h^{\text{cons}}(u, v) &:= \sum_{n=0}^{N+1} \{c(x_n) u'(x_n)\}[v(x_n)] + \{c(x_n) v'(x_n)\}[u(x_n)] \\ b_h^{\text{penal}}(u, v) &:= \sum_{n=0}^{N+1} \mathbf{a}_n[u(x_n)][v(x_n)] \end{aligned}$$

Let $u_h = \sum_{m=0}^N \sum_{j=0}^r \alpha_j^m \Phi_j^m \in V_h$ denote the Galerkin approximation, then as discussed in section 1.6 the discrete variational formulation (1.9) is equivalent to

$$\sum_{m=0}^N \sum_{j=0}^r \alpha_j^m \left(a_h(\Phi_j^m, \Phi_i^n) - b_h^{\text{cons}}(\Phi_j^m, \Phi_i^n) + b_h^{\text{penal}}(\Phi_j^m, \Phi_i^n) \right) = \ell_h(\Phi_i^n), \quad \forall n = 0, \dots, N, i = 0, \dots, r \quad (1.14)$$

which corresponds to the matrix vector system (1.10) where we can write

$$\mathbf{B} = \mathbf{A} - \mathbf{B}_{\text{cons}} + \mathbf{B}_{\text{penal}}$$

we will assemble the three (symmetric) matrices separately.

$$\begin{aligned} [\mathbf{B}]_{T(n,i), T(m,j)} &= b_h(\Phi_j^m, \Phi_i^n), & [\mathbf{A}]_{T(n,i), T(m,j)} &= a_h(\Phi_j^m, \Phi_i^n) \\ [\mathbf{B}_{\text{cons}}]_{T(n,i), T(m,j)} &= b_h^{\text{cons}}(\Phi_j^m, \Phi_i^n) & [\mathbf{B}_{\text{penal}}]_{T(n,i), T(m,j)} &= b_h^{\text{penal}}(\Phi_j^m, \Phi_i^n) \end{aligned}$$

where T is given by (1.13)

1.8.1 Assembly of \mathbf{A}

\mathbf{A} is assembled similarly to the standard stiffness matrix in continuous finite element. The main difference is that there is no overlap in the elementwise contributions. Each set of local (element) basis functions only contributes to the integrals over said element. We can rewrite $\mathbf{A} = \sum_{s=0}^N \mathbf{A}^{(s)}$, where $[\mathbf{A}^{(s)}]_{T(n,i), T(m,j)} = \int_{I_s} c(\Phi_j^m)'(\Phi_i^n)' \, dx$. Now since we have $\text{supp}(\Phi_i^n) \subset I_n$ the only non-zero entries of $\mathbf{A}^{(s)}$ are the ones where both $n = m = s$. Pulling back the integral to the reference element using the chain rule and the substitution $F_s^{-1}(x) = \xi$ we find

$$\int_{I_s} c(x) (\Phi_j^s)'(x) (\Phi_i^s)'(x) \, dx = \frac{2}{h_s} \int_{\hat{I}} c(F_s(\xi)) \hat{\phi}'_j(\xi) \hat{\phi}'_i(\xi) \, d\xi$$

This integral now only depends on the reference shape functions, the element length h_s and the values of the coefficient c . Using the Gauss-Lobatto quadrature rule we can approximate the integral only requiring the values of c at the nodes, whilst having the values of ϕ, ϕ' preloaded. The total assembly of \mathbf{A} can therefore be achieved by calculating a local contribution matrix $\hat{\mathbf{A}}^{(s)} \in \mathbb{R}^{(r+1) \times (r+1)}$ for each element I_s and adding it into \mathbf{A} .

Example 1.3. For $c \equiv 1$ with \mathcal{P}^1 -elements ($r = 1$) we have

$$\hat{\mathbf{A}}^{(s)} = \frac{1}{h_s} \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}$$

1.8.2 Assembly of B consistency part

As before we rewrite $\mathbf{B}_{\text{cons}} = \sum_{s=0}^{N+1} \mathbf{B}_{\text{cons}}^{(s)}$, where

$$[\mathbf{B}_{\text{cons}}^{(s)}]_{T(n,i), T(m,j)} = \{c(x_s) \Phi_j^{m'}(x_s)\} [\Phi_i^n(x_s)] + \{c(x_s) \Phi_i^{n'}(x_s)\} [\Phi_j^m(x_s)]$$

Interior Faces

First let $s \in \{1, \dots, N\}$ denote an interior face, we observe again, that the entries of $\mathbf{B}_{\text{cons}}^{(s)}$ can only be non-zero for an index tuple $(T(n, i), T(m, j))$ if $n, m \in \{s, s-1\}$ by the local support of the basis functions. Therefore assembling \mathbf{B}_{cons} again comes down to calculating a local contribution matrix

$$\widehat{\mathbf{B}}_{\text{cons}}^{(s)} = \begin{bmatrix} \mathbf{C}_{\text{cons}}^{(s-1,s-1)} & \mathbf{C}_{\text{cons}}^{(s-1,s)} \\ \mathbf{C}_{\text{cons}}^{(s,s-1)} & \mathbf{C}_{\text{cons}}^{(s,s)} \end{bmatrix} \in \mathbb{R}^{2(r+1) \times 2(r+1)}$$

for each interior face x_s consisting of four blocks we will now lay out in more detail. We discuss the boundary case separately. Using again the local support of the basis functions we find

$$\begin{aligned} [\mathbf{C}_{\text{cons}}^{(s-1,s-1)}]_{i,j} &= \frac{c(x_s^-)}{2} \Phi_j^{s-1'}(x_s^-) \Phi_i^{s-1}(x_s^-) + \frac{c(x_s^-)}{2} \Phi_i^{s-1'}(x_s^-) \Phi_j^{s-1}(x_s^-) \\ [\mathbf{C}_{\text{cons}}^{(s-1,s)}]_{i,j} &= \frac{c(x_s^+)}{2} \Phi_j^{s'}(x_s^+) \Phi_i^{s-1}(x_s^-) - \frac{c(x_s^-)}{2} \Phi_i^{s-1'}(x_s^-) \Phi_j^s(x_s^+) \\ [\mathbf{C}_{\text{cons}}^{(s,s)}]_{i,j} &= -\frac{c(x_s^+)}{2} \Phi_j^{s'}(x_s^+) \Phi_i^s(x_s^+) - \frac{c(x_s^+)}{2} \Phi_i^{s'}(x_s^+) \Phi_j^s(x_s^+) \end{aligned}$$

where we have used the definitions of jump and average in (1.1). Note that $\mathbf{C}_{\text{cons}}^{(s-1,s)} = (\mathbf{C}_{\text{cons}}^{(s,s-1)})^T$ by the symmetry of the bilinear form b_h^{cons} . Next we represent the values of the basis functions Φ at the element boundary by the values of the reference shape functions $\widehat{\phi}$

$$\begin{aligned} \Phi_i^s(x_s^+) &= \widehat{\phi}_i(-1), & \Phi_i^{s-1}(x_s^-) &= \widehat{\phi}_i(1) \\ \Phi_i^{s'}(x_s^+) &= \frac{2}{h_s} \widehat{\phi}'_i(-1), & \Phi_i^{s-1'}(x_s^-) &= \frac{2}{h_{s-1}} \widehat{\phi}'_i(1) \end{aligned} \quad (1.15)$$

which finally yields

$$\begin{aligned} [\mathbf{C}_{\text{cons}}^{(s-1,s-1)}]_{i,j} &= \frac{c(x_s^-)}{h_{s-1}} \widehat{\phi}'_j(1) \widehat{\phi}_i(1) + \frac{c(x_s^-)}{h_{s-1}} \widehat{\phi}'_i(1) \widehat{\phi}_j(1) \\ [\mathbf{C}_{\text{cons}}^{(s-1,s)}]_{i,j} &= \frac{c(x_s^+)}{h_s} \widehat{\phi}'_j(-1) \widehat{\phi}_i(1) - \frac{c(x_s^-)}{h_{s-1}} \widehat{\phi}'_i(1) \widehat{\phi}_j(-1) \\ [\mathbf{C}_{\text{cons}}^{(s,s)}]_{i,j} &= -\frac{c(x_s^+)}{h_s} \widehat{\phi}'_j(-1) \widehat{\phi}_i(-1) - \frac{c(x_s^+)}{h_s} \widehat{\phi}'_i(-1) \widehat{\phi}_j(-1) \end{aligned}$$

Example 1.4. Consider $c \equiv 1$ for \mathcal{P}^1 -elements ($r = 1$) with an equidistant mesh with meshsize h we have

$$\widehat{\mathbf{B}}_{\text{cons}}^{(s)} = \frac{1}{h} \begin{bmatrix} 0 & -1/2 & 1/2 & 0 \\ -1/2 & 1 & -1 & 1/2 \\ 1/2 & -1 & 1 & -1/2 \\ 0 & 1/2 & -1/2 & 0 \end{bmatrix}$$

Boundary Faces

For $s \in \{0, N+1\}$ and x_s a boundary face we now have a smaller local contribution matrix since the contribution can only come from the one element to which x_s belongs. Meaning we have $\widehat{\mathbf{B}}_{\text{cons}}^{(0)}, \widehat{\mathbf{B}}_{\text{cons}}^{(N+1)} \in$

$\mathbb{R}^{(r+1) \times (r+1)}$ with

$$\begin{aligned} [\hat{\mathbf{B}}_{\text{cons}}^{(0)}]_{i,j} &= -\frac{2c(x_0^+)}{h_0} \hat{\phi}'_j(-1) \hat{\phi}_i(-1) - \frac{2c(x_0^+)}{h_0} \hat{\phi}'_i(-1) \hat{\phi}_j(-1) \\ [\hat{\mathbf{B}}_{\text{cons}}^{(N+1)}]_{i,j} &= \frac{2c(x_{N+1}^-)}{h_{N+1}} \hat{\phi}'_j(1) \hat{\phi}_i(1) + \frac{2c(x_{N+1}^-)}{h_{N+1}} \hat{\phi}'_i(1) \hat{\phi}_j(1) \end{aligned}$$

Example 1.5. For $c \equiv 1$ with \mathcal{P}^1 -elements ($r = 1$) we have

$$\hat{\mathbf{B}}_{\text{cons}}^{(0)} = \frac{1}{h_0} \begin{bmatrix} 2 & -1 \\ -1 & 0 \end{bmatrix}, \quad \hat{\mathbf{B}}_{\text{cons}}^{(N+1)} = \frac{1}{h_{N+1}} \begin{bmatrix} 0 & -1 \\ -1 & 2 \end{bmatrix}$$

1.8.3 Assembly of B penalty part

Again we rewrite $\mathbf{B}_{\text{penal}} = \sum_{s=0}^{N+1} \mathbf{B}_{\text{penal}}^{(s)}$, where

$$[\mathbf{B}_{\text{penal}}^{(s)}]_{T(n,i), T(m,j)} = \mathbf{a}_s [\Phi_j^m(x_s)] [\Phi_i^n(x_s)]$$

We proceed analogously to 1.8.2.

Interior Faces

Let $s \in \{1, \dots, N\}$ and x_s denote an interior face. As before we have that the entries of $\mathbf{B}_{\text{penal}}^{(s)}$ can only be nonzero for an index tuple $(T(n, i), T(m, j))$ if $n, m \in \{s, s-1\}$, similar to 1.8.2 we find that the assembly boils down to adding up local contributions represented in a local contribution matrix

$$\hat{\mathbf{C}}_{\text{penal}}^{(s)} = \begin{bmatrix} \mathbf{C}_{\text{penal}}^{(s-1,s-1)} & \mathbf{C}_{\text{penal}}^{(s-1,s)} \\ \mathbf{C}_{\text{penal}}^{(s,s-1)} & \mathbf{C}_{\text{penal}}^{(s,s)} \end{bmatrix} \in \mathbb{R}^{2(r+1) \times 2(r+1)}$$

and using (1.15), and the definition of the penalization parameter (1.8) we specifically find

$$\begin{aligned} [\mathbf{C}_{\text{penal}}^{(s-1,s-1)}]_{i,j} &= \mathbf{a}_s \hat{\phi}_j(1) \hat{\phi}_i(1) \\ [\mathbf{C}_{\text{penal}}^{(s-1,s)}]_{i,j} &= -\mathbf{a}_s \hat{\phi}_j(-1) \hat{\phi}_i(1) \\ [\mathbf{C}_{\text{penal}}^{(s,s)}]_{i,j} &= \mathbf{a}_s \hat{\phi}_j(-1) \hat{\phi}_i(-1) \end{aligned}$$

where again by symmetry of the penalty term we have $\mathbf{C}_{\text{penal}}^{(s-1,s)} = (\mathbf{C}_{\text{penal}}^{(s,s-1)})^T$ and \mathbf{a}_s only depends on the two adjacent elements I_{s-1}, I_s .

Example 1.6. Consider $c \equiv 1$ for \mathcal{P}^1 -elements ($r = 1$) with an equidistant mesh with meshsize h we have

$$\hat{\mathbf{B}}_{\text{penal}}^{(s)} = \frac{\sigma}{h} \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Boundary Faces

For $s \in \{0, N+1\}$ we again have only the respective boundary element contributing. So the local contribution matrices $\hat{\mathbf{B}}_{\text{penal}}^{(s)} \in \mathbb{R}^{(r+1) \times (r+1)}$ satisfy

$$[\hat{\mathbf{B}}_{\text{penal}}^{(0)}]_{i,j} = \mathbf{a}_0 \hat{\phi}_j(-1) \hat{\phi}_i(-1), \quad [\hat{\mathbf{B}}_{\text{penal}}^{(N+1)}]_{i,j} = \mathbf{a}_{N+1} \hat{\phi}_j(1) \hat{\phi}_i(1)$$

Example 1.7. For $c \equiv 1$ with \mathcal{P}^1 -elements ($r = 1$) we have

$$\hat{\mathbf{B}}_{\text{penal}}^{(0)} = \frac{\sigma}{h_0} \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \quad \hat{\mathbf{B}}_{\text{penal}}^{(N+1)} = \frac{\sigma}{h_{N+1}} \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

1.8.4 System Vector Assembly

We divide assembling the vector \mathbf{l} in (1.10) into two parts.

$$\mathbf{l} = \mathbf{l}_{\text{load}} + \mathbf{l}_{\text{bc}}$$

First we recall the assembly of the load vector \mathbf{l}_{load} , i.e. the vector containing the contributions of the forcing term f and secondly we will describe how to add the Dirichlet boundary condition contributions (\mathbf{l}_{bc}).

Load Vector

The assembly of the load vector is completely analogous to the continuous finite element case. Using the local support of Φ_i^n we can rewrite

$$\int_{\Omega} f \Phi_i^n dx = \sum_{s=0}^N \int_{I_s} f \Phi_i^n dx = \int_{I_n} f(x) \Phi_i^n(x) dx = \frac{h_n}{2} \int_{-1}^1 f(F_n(\xi)) \hat{\phi}_i(\xi) d\xi$$

meaning as before we can assemble $\mathbf{l}_{\text{load}} = \sum_{s=0}^N \mathbf{l}_{\text{load}}^{(s)}$ where

$$[\mathbf{l}_{\text{load}}^{(s)}]_{T(n,i)} = \int_{I_s} f \Phi_i^n dx = \delta_{n,s} \frac{h_n}{2} \int_{-1}^1 f(F_n(\xi)) \hat{\phi}_i(\xi) d\xi$$

which can be characterized by the local contribution vector $\hat{\mathbf{l}}_{\text{load}}^{(s)} \in \mathbb{R}^{r+1}$ defined as

$$[\hat{\mathbf{l}}_{\text{load}}^{(s)}]_i = \frac{h_s}{2} \int_{-1}^1 f(F_s(\xi)) \hat{\phi}_i(\xi) d\xi$$

In practice we approximate the integral using the Gauss-Lobatto quadrature rule.

Example 1.8. For f piecewise constant (i.e. $f|_{I_s} \equiv f_s \in \mathbb{R} \quad \forall s = 0, \dots, N$) with \mathcal{P}^1 -elements ($r = 1$) we have

$$\hat{\mathbf{l}}_{\text{load}}^{(s)} = \frac{f_s h_s}{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Dirichlet Boundary Condition Vector

We have

$$[\mathbf{l}_{\text{bc}}]_{T(n,i)} = -g_1 c(x_{N+1}^-) \Phi_i^n'(x_{N+1}^-) + g_0 c(x_0^+) \Phi_i^n'(x_0^+) + \mathbf{a}_{N+1} g_1 \Phi_i^n(x_{N+1}^-) + \mathbf{a}_0 g_0 \Phi_i^n(x_0^+)$$

where the entries can clearly only be non-zero at indices corresponding to boundary elements. We characterize the assembly using the local contribution vectors $\hat{\mathbf{l}}_{\text{bc}}^{(N+1)}, \hat{\mathbf{l}}_{\text{bc}}^{(0)} \in \mathbb{R}^{r+1}$ where

$$[\hat{\mathbf{l}}_{\text{bc}}^{(0)}]_i = \frac{g_0}{h_0} c(x_0^-) \hat{\phi}_i(-1) + \mathbf{a}_0 g_0 \hat{\phi}_i(-1), \quad [\hat{\mathbf{l}}_{\text{bc}}^{(N+1)}]_i = -\frac{g_1}{h_N} c(x_{N+1}^-) \hat{\phi}_i(1) + \mathbf{a}_{N+1} g_1 \hat{\phi}_i(1)$$

1.9 Existence of Discrete Solution

Firstly we will recall some basic definitions:

Definition 1.9. Let V be a normed vector space and $b : V \times V \rightarrow \mathbb{R}$ be a bilinear form.

(i) We say b is **continuous** if $\exists C_{cont} > 0$, such that

$$|b(u, v)| \leq C_{cont} \|u\| \|v\| \quad \forall u, v \in V$$

(ii) We say b is **symmetric** if

$$b(u, v) = b(v, u) \quad \forall u, v \in V$$

(iii) We say b is **coercive** if $\exists C_{coer} > 0$, such that

$$b(u, u) \geq C_{coer} \|u\|^2 \quad \forall u \in V$$

Since (1.9) corresponds to the finite dimensional system (1.10) uniqueness and existence of a solution are equivalent. The bilinear form b_h is *symmetric* by construction the goal of this section is to show that b_h is also *coercive* for $\sigma > 0$ big enough. From the coercivity of b_h it will follow that the matrix \mathbf{B} in (1.10) is positive definite and hence invertible, which means there exists a (unique) solution of (1.9).

Lemma 1.10. Let $V = \text{span}(\varphi_1, \dots, \varphi_M)$ be a finite dimensional normed vector space with $\dim(V) = M \in \mathbb{N}$ and let $b : V \times V \rightarrow \mathbb{R}$ be a symmetric, coercive bilinear form, then the matrix $[\mathbf{B}]_{i,j} = [b(\varphi_j, \varphi_i)]_{i,j} \in \mathbb{R}^{N \times N}$ is symmetric positive definite.

Proof. Clearly \mathbf{B} is symmetric.

Let $\mathbf{v} = (v_1, \dots, v_M) \in \mathbb{R}^M$ then $v = \sum_{i=1}^M v_i \varphi_i \in V$ and we have:

$$\mathbf{v}^T \mathbf{B} \mathbf{v} = \sum_{i,j=1}^M v_i v_j b(\varphi_j, \varphi_i) = b(v, v) \geq C_{coer} \|v\|^2$$

where we have used the bilinearity and the coercivity of b . □

1.9.1 Trace Inequalities

Here we recall what is in the theory called *trace inequalities*, which are inequalities bounding a boundary norm with a norm over the whole domain. In 1d this means estimating the absolute value of a function at a boundary node, by a norm over the whole interval, which simplifies the proofs a lot. Still let it be said here, that these kind of inequalities hold in higher dimensions as well. General results can for example be found in Ern [5].

This subsection will contain two results, one for general H^1 -functions and one for polynomials.

We start by considering the finite dimensional case, inspired by [8].

Lemma 1.11 (Discrete trace inequality). Let $r \geq 1$ be the polynomial degree, $a, b \in \mathbb{R}$ with $a < b$ and let $\mathcal{P}^r([a, b])$ denote the space of polynomials of degree r defined on $[a, b]$.

For any $v \in \mathcal{P}^r([a, b])$ we have:

$$(i) |v(a)|^2 \leq \frac{(r+1)^2}{|b-a|} \|v\|_{L^2([a,b])}^2$$

$$(ii) |v(b)|^2 \leq \frac{(r+1)^2}{|b-a|} \|v\|_{L^2([a,b])}^2$$

Proof. We will prove the statements first for the reference element $\hat{I} = [-1, 1]$ and then use a scaling argument

to show the general case by applying a simple substitution.

Step 1 (Setup).

We will make use of the Legendre orthonormal basis of $\mathcal{P}^r(\hat{I})$: Let P_0, \dots, P_r denote the Legendre polynomials on $\mathcal{P}^r(\hat{I})$. Recall the following well known facts (see for example [6]):

1. $\{P_0, \dots, P_r\}$ form an orthogonal basis of $\mathcal{P}^r(\hat{I})$ under the $L^2(\hat{I})$ inner product. Meaning:

$$\text{span}(P_0, \dots, P_r) = \mathcal{P}^r(\hat{I}), \quad \int_{-1}^1 P_i P_j d\xi = \begin{cases} \frac{2}{2i+1}, & \text{for } i=j \\ 0, & \text{for } i \neq j \end{cases}$$

2. $P_i(1) = 1, P_i(-1) = (-1)^i, \quad \forall i = 0, \dots, r$

Let $\psi_i = \sqrt{\frac{(2i+1)}{2}} P_i$ for $i = 0, \dots, r$ denote the normed basis function. Clearly we now have

$$\psi_i(-1) = (-1)^i \sqrt{\frac{2i+1}{2}}, \quad \psi_i(1) = \sqrt{\frac{2i+1}{2}}, \quad \int_{-1}^1 \psi_i \psi_j d\xi = \delta_{i,j}, \quad \forall i = 0, \dots, r$$

where $\delta_{i,j} = \begin{cases} 1, & \text{for } i=j \\ 0, & \text{for } i \neq j \end{cases}$, and hence $\{\psi_0, \dots, \psi_r\}$ form an orthonomal basis.

Step 2 (Proof on reference element).

For any $v \in \mathcal{P}^r(\hat{I})$ there exist coefficients $v_0, \dots, v_r \in \mathbb{R}$, such that $v = \sum_{i=0}^r v_i \psi_i$. By applying Cauchy-Schwarz we find

$$|v(-1)|^2 = \left| \sum_{i=0}^r v_i \psi_i(-1) \right|^2 \leq \left(\sum_{i=0}^r v_i^2 \right) \left(\sum_{i=0}^r \psi_i(-1)^2 \right) = \left(\sum_{i=0}^r v_i^2 \right) \left(\sum_{i=0}^r \frac{2i+1}{2} \right) = \left(\sum_{i=0}^r v_i^2 \right) \frac{(r+1)^2}{2}$$

and finally the orthonormality of the ψ_i yields

$$\frac{(r+1)^2}{2} \sum_{i=0}^r v_i^2 = \frac{(r+1)^2}{2} \sum_{i,j=0}^r v_i v_j \delta_{i,j} = \frac{(r+1)^2}{2} \|v\|_{L^2(\hat{I})}^2$$

This yields the first inequality for the reference element. The second inequality can be proven analogously.

Step 3 (Scaling argument).

Now we assume that $v \in \mathcal{P}^r([a, b])$. Using the affine (element) map

$$F : [-1, 1] \rightarrow [a, b], \xi \mapsto \frac{a+b}{2} + \frac{|b-a|}{2} \xi$$

we can pull v back to the reference element by defining $\hat{v}(\xi) := v(F(\xi))$ for all $\xi \in \hat{I}$. Clearly $\hat{v} \in \mathcal{P}^r(\hat{I})$ hence, by Step 2 we obtain

$$|v(a)|^2 = |\hat{v}(F^{-1}(a))|^2 = |\hat{v}(-1)|^2 \leq \frac{(r+1)^2}{2} \int_{-1}^1 \hat{v}(\xi)^2 d\xi = \frac{(r+1)^2}{2} \frac{2}{|b-a|} \|v\|_{L^2([a,b])}^2$$

where in the last equality we have applied a change of variable $x = F(\xi)$ to the integral. Applying the same line of reasoning to $|v(b)|^2$ proves both inequalities and so we are done.

□

Next we consider the continuous case.

Lemma 1.12 (Continuous trace inequality). *Let $I = (a, b) \subset \mathbb{R}$ be an element of length $h := b - a$, then for any $v \in H^1(\Omega)$*

$$(i) \quad |v(a)| \leq h^{-1/2} \|v\|_{L^2(I)} + h^{1/2} \|v'\|_{L^2(I)};$$

$$(ii) \quad |v(b)| \leq h^{-1/2} \|v\|_{L^2(I)} + h^{1/2} \|v'\|_{L^2(I)}.$$

Proof. We show (i), the proof of (ii) is analogous.

Define $\varphi(x) := 1 - \frac{(x-a)}{h}$, where $\varphi(a) = 1, \varphi(b) = 0$ and let $v \in H^1(I)$. By applying the fundamental theorem of calculus we can estimate

$$\begin{aligned} |v(a)| &= |v(a)\varphi(a) - v(b)\varphi(b)| = \left| \int_I (v\varphi)' dx \right| = \left| \int_I v'(x)\varphi(x) dx + \int_I v(x)\varphi'(x) dx \right| \\ &\leq \int_I |v'(x)| \underbrace{|\varphi(x)|}_{\leq 1} dx + \frac{1}{h} \int_I |v(x)| dx, \end{aligned}$$

where in the last step we have used that $|\varphi'(x)| = h^{-1}$. Using Cauchy-Schwarz we finally estimate

$$\begin{aligned} |v(a)| &\leq \left(\int_I 1 dx \right)^{1/2} \left(\int_I |v'(x)|^2 dx \right)^{1/2} + \frac{1}{h} \left(\int_I 1 dx \right)^{1/2} \left(\int_I |v(x)|^2 dx \right)^{1/2} \\ &\leq h^{1/2} \|v'\|_{L^2(I)} + h^{-1/2} \|v\|_{L^2(I)}, \end{aligned}$$

which proves (i). □

Recall the in previous sections established notations, let $r \in \mathbb{N}$ denote the polynomial degree and $V_h^r(\mathcal{T}_h)$ be the discrete subspace.

Definition 1.13. *We define the **energy norm** on V_h by*

$$\|v\|_\epsilon^2 := \sum_{n=0}^N \int_{I_n} c(x)v'(x)^2 dx + \sum_{n=0}^{N+1} \mathbf{a}_n [v(x_n)]^2 \quad (1.16)$$

where \mathbf{a} denotes the penalization term in (1.8).

Lemma 1.14. $\|\cdot\|_\epsilon$ defines a norm on V_h .

Proof. Clearly we have $\|\lambda v\|_\epsilon = |\lambda| \|v\|_\epsilon$ for all $\lambda \in \mathbb{R}, v \in V_h$.

By definition we have $\mathbf{a}, c > 0$ and by extension $\|v\|_\epsilon \geq 0$ for all $v \in V_h$. Suppose now that $\|v\|_\epsilon = 0$ for some $v \in V_h$, then we must have $v|_{I_n} \equiv \text{const}$ and $[v(x_n)] = 0$ for all n . So v must be constant on all elements and have a jump of zero at the element boundaries. These two facts combined imply that v is constant on all of Ω . By the definition of the jump at the boundary nodes of Ω it immediately follows that $v = 0$. Clearly $\|0\|_\epsilon = 0$, therefore $\|\cdot\|_\epsilon$ is positive definite.

Using $[v(x_n) + w(x_n)] = [v(x_n)] + [w(x_n)] \quad \forall v, w \in V_h, n = 0, \dots, N+1$ we find

$$\begin{aligned} \|v + w\|_\epsilon &\leq \left(\sum_{n=0}^N (\|\sqrt{c}v'\|_{L^2(I_n)} + \|\sqrt{c}w'\|_{L^2(I_n)})^2 + \sum_{n=0}^{N+1} (\sqrt{\mathbf{a}_n} ([v(x_n)] + [w(x_n)]))^2 \right)^{1/2} \\ &\leq \|v\|_\epsilon + \|w\|_\epsilon \end{aligned}$$

where in the last inequality we have used the triangle inequality of the euclidian vector norm on \mathbb{R}^{2N+3} , with the vector given as

$$\mathbf{v} = [\|\sqrt{c}v'\|_{L^2(I_0)}, \dots, \|\sqrt{c}v'\|_{L^2(I_N)}, \sqrt{\mathbf{a}_0}[v(x_0)], \dots, \sqrt{\mathbf{a}_{N+1}}[v(x_{N+1})]]^T$$

this shows the triangle inequality for $\|\cdot\|_\epsilon$ and hence it is a norm. \square

Theorem 1.15. Let $r \in \mathbb{N}$, the bilinear form b_h in (1.9) is continuous on $V_h^r(\mathcal{T}_h)$ and if furthermore $\sigma \geq \frac{6(r+1)^2 c_{\max}}{c_{\min}}$, b_h is also coercive on $V_h^r(\mathcal{T}_h)$. The coercivity and continuity constants are given by

$$C_{coer} = \frac{1}{2}, \quad C_{cont} = (3 + \frac{5}{4}C_\sigma)$$

where $C_\sigma = \frac{(r+1)^2 c_{\max}}{\sigma c_{\min}}$

Proof. **Step 1 (Coercivity).**

Let $w \in V_h$. Note that

$$b_h(w, w) = \|w\|_\epsilon^2 - 2 \sum_{n=0}^{N+1} \{c(x_n)w'(x_n)\}[w(x_n)] \quad (1.17)$$

To derive the coercivity of b_h we will estimate the term $2 \sum_{n=0}^{N+1} \{c(x_n)w'(x_n)\}[w(x_n)]$ from above applying Lemma 1.11 and additional smaller tools:

Using Young's inequality: $2ab \leq a^2 + b^2, \forall a, b \in \mathbb{R}$, we estimate

$$\begin{aligned} 2 \sum_{n=0}^{N+1} \{c(x_n)w'(x_n)\}[w(x_n)] &= 2 \sum_{n=0}^{N+1} \{c(x_n)w'(x_n)\} \left(\frac{\mathbf{a}_n}{2}\right)^{-1/2} \left(\frac{\mathbf{a}_n}{2}\right)^{1/2} [w(x_n)] \\ &\leq 2 \sum_{n=0}^{N+1} \frac{\{c(x_n)w'(x_n)\}^2}{\mathbf{a}_n} + \frac{1}{2} \sum_{n=0}^{N+1} \mathbf{a}_n [w(x_n)]^2 \end{aligned} \quad (1.18)$$

Recalling $\mathbf{a}_n = \sigma \mathbf{c}_n \mathbf{h}_n^{-1}$ from (1.8) and noting the relations $\mathbf{h}_n \leq h_n, \mathbf{c}_n^{-1} \leq c(x_n^-)^{-1}, c(x_n^+)^{-1}$ we find

$$\begin{aligned} \mathbf{a}_n^{-1} c(x_n^+) &\leq \frac{h_n}{\sigma}, \quad \mathbf{a}_n^{-1} c(x_n^-) \leq \frac{h_{n-1}}{\sigma}, \quad \forall n = 1, \dots, N \\ \mathbf{a}_0^{-1} c(x_0^+) &= \frac{h_0}{\sigma}, \quad \mathbf{a}_{N+1}^{-1} c(x_{N+1}^-) = \frac{h_N}{\sigma} \end{aligned}$$

applying this and the usefull inequality $(a+b)^2 \leq 2a^2 + 2b^2$ yields

$$\begin{aligned} 2 \sum_{n=0}^{N+1} \frac{\{c(x_n)w'(x_n)\}^2}{\mathbf{a}_n} &= 2 \sum_{n=1}^N \frac{1}{4\mathbf{a}_n} \left(c(x_n^-)w'(x_n^-) + c(x_n^+)w'(x_n^+) \right)^2 + \frac{2}{\mathbf{a}_0} \left(c(x_0^+)w'(x_0^+) \right)^2 + \frac{2}{\mathbf{a}_{N+1}} \left(c(x_{N+1}^-)w'(x_{N+1}^-) \right)^2 \\ &\leq 2 \sum_{n=1}^N \frac{1}{2\sigma} \left(h_{n-1} c(x_n^-)w'(x_n^-)^2 + h_n c(x_n^+)w'(x_n^+)^2 \right) + \frac{2h_0}{\sigma} c(x_0^+)w'(x_0^+)^2 + \frac{2h_N}{\sigma} c(x_{N+1}^-)w'(x_{N+1}^-)^2 \\ &\leq \frac{c_{\max}}{\sigma} \sum_{n=1}^N \left(h_{n-1} w'(x_n^-)^2 + h_n w'(x_n^+)^2 \right) + \frac{2c_{\max}h_0}{\sigma} w'(x_0^+)^2 + \frac{2c_{\max}h_N}{\sigma} w'(x_{N+1}^-)^2 \end{aligned} \quad (1.19)$$

Since $w \in V_h$ is a (broken) polynomial, we can apply Lemma 1.11 elementwise and find

$$w'(x_n^+)^2, w'(x_{n+1}^-)^2 \leq \frac{(r+1)^2}{h_n} \|w'\|_{L^2(I_n)}^2 \quad \forall n = 0, \dots, N \quad (1.20)$$

By combining (1.19), (1.20) and inserting $1 = c_{\min}c_{\min}^{-1} \leq c(x)c_{\min}^{-1} \quad \forall x \in \Omega$ we find

$$2 \sum_{n=0}^{N+1} \frac{\{c(x_n)w'(x_n)\}^2}{\mathbf{a}_n} \leq 3C_\sigma \sum_{n=0}^N \|\sqrt{c}w'\|_{L^2(I_n)}^2 \quad (1.21)$$

$$\text{for } C_\sigma := \frac{(r+1)^2 c_{\max}}{\sigma c_{\min}} > 0.$$

Finally putting together (1.17), (1.18) and (1.21) yields

$$\begin{aligned} b_h(w, w) &\geq \|w\|_\epsilon^2 - 3C_\sigma \sum_{n=0}^N \|\sqrt{c}w'\|_{L^2(I_n)}^2 - \frac{1}{2} \sum_{n=0}^{N+1} \mathbf{a}_n [w(x_n)]^2 \\ &= (1 - 3C_\sigma) \sum_{n=0}^N \|\sqrt{c}w'\|_{L^2(I_n)}^2 + \frac{1}{2} \sum_{n=0}^{N+1} \mathbf{a}_n [w(x_n)]^2 \\ &\geq \frac{1}{2} \|w\|_\epsilon^2 \end{aligned}$$

for $\sigma \geq \frac{6(r+1)^2 c_{\max}}{c_{\min}}$, which proves the coercivity of b_h on V_h .

Step 2 (Continuity).

The proof of the continuity of b_h uses similar ideas as the coercivity proof. Let $u, v \in V_h$, by using Cauchy-Schwarz we immediately get

$$\begin{aligned} |b_h(u, v)| &\leq \sum_{n=0}^N \|\sqrt{c}u'\|_{L^2(I_n)} \|\sqrt{c}v'\|_{L^2(I_n)} + \sum_{n=0}^{N+1} |\{c(x_n)u'(x_n)\}[v(x_n)]| \\ &\quad + \sum_{n=0}^{N+1} |\{c(x_n)v'(x_n)\}[u(x_n)]| + \sum_{n=0}^{N+1} \mathbf{a}_n |[u(x_n)][v(x_n)]| \\ &=: T_{\text{ell}} + T_{\text{cons}}^{(u)} + T_{\text{cons}}^{(v)} + T_{\text{penal}} \end{aligned} \quad (1.22)$$

The goal is now to estimate the consistency terms T_{cons} from above by something of the form $\sum_{n=0}^{N+1} t_n(u)s_n(v) + \sum_{n=0}^{N+1} t_n(v)s_n(u)$, such that together with the terms $T_{\text{ell}}, T_{\text{penal}}$ we can use discrete Cauchy-Schwarz on the sums and hence separate them into a product of the two energy norms $C_{\text{cont}}\|u\|_h\|v\|_h$ scaled by a positive constant.

We will show the estimate of $T_{\text{cons}}^{(u)}$, the procedure to estimate $T_{\text{cons}}^{(v)}$ is analogous.

First rewrite

$$T_{\text{cons}}^{(u)} = \sum_{n=0}^{N+1} |\{c(x_n)u'(x_n)\}\mathbf{a}_n^{-1/2}\mathbf{a}_n^{1/2}[v(x_n)]| \quad (1.23)$$

Next again using the definition of \mathbf{a} and estimates as in Step 1 we find for interior faces $n = 1, \dots, N$

$$|\{c(x_n)u'(x_n)\}\mathbf{a}_n^{-1/2}| \leq \frac{1}{2} \sqrt{\frac{\mathbf{h}_n}{\sigma}} \sqrt{c_{\max}} (|u'(x_n^-)| + |u'(x_n^+)|)$$

and for the boundary faces

$$|\{c(x_0)u'(x_0)\}\mathbf{a}_0^{-1/2}| \leq \sqrt{\frac{\mathbf{h}_0}{\sigma}} \sqrt{c_{\max}} |u'(x_0^+)|, \quad |\{c(x_{N+1})u'(x_{N+1})\}\mathbf{a}_{N+1}^{-1/2}| \leq \sqrt{\frac{\mathbf{h}_N}{\sigma}} \sqrt{c_{\max}} |u'(x_{N+1}^-)|$$

Applying Lemma (1.11) yields for $\beta_n(u) := \sqrt{C_\sigma} \|\sqrt{c}u'\|_{L^2(I_n)}$, $n = 0, \dots, N$

$$\begin{aligned} |\{c(x_n)u'(x_n)\}| \mathbf{a}_n^{-1/2} &\leq \frac{\beta_{n-1}(u)}{2} + \frac{\beta_n(u)}{2} \quad \text{for } n = 1, \dots, N \\ |\{c(x_0)u'(x_0)\}| \mathbf{a}_0^{-1/2} &\leq \beta_0(u) \\ |\{c(x_{N+1})u'(x_{N+1})\}| \mathbf{a}_{N+1}^{-1/2} &\leq \beta_N(u) \end{aligned}$$

which we can now plug back into (1.23) to get

$$T_{\text{cons}}^{(u)} \leq \beta_0(u)\gamma_0(v) + \beta_N(u)\gamma_{N+1}(v) + \sum_{n=1}^N \frac{\beta_{n-1}(u)}{2}\gamma_n(v) + \sum_{n=1}^N \frac{\beta_n(u)}{2}\gamma_n(v) \quad (1.24)$$

for $\gamma_n(v) := \sqrt{\mathbf{a}_n} |[v(x_n)]| \forall n = 0, \dots, N+1$. By furthermore denoting $\alpha_n(u) := \|\sqrt{c}u'\|_{L^2(I_n)}$ we can represent

$$T_{\text{ell}} = \sum_{n=0}^N \alpha_n(u)\alpha_n(v), \quad T_{\text{penal}} = \sum_{n=0}^{N+1} \gamma_n(u)\gamma_n(v)$$

and in total for

$$\begin{aligned} \mathbf{u} &:= [\alpha_0(u), \dots, \alpha_N(u), \beta_0(u), \beta_N(u), \frac{\beta_0(u)}{2}, \dots, \frac{\beta_{N-1}(u)}{2}, \frac{\beta_1(u)}{2}, \dots, \frac{\beta_N(u)}{2}, \\ &\quad \gamma_0(u), \gamma_{N+1}(u), \gamma_1(u), \dots, \gamma_N(u), \gamma_1(u), \dots, \gamma_N(u), \gamma_0(u), \dots, \gamma_{N+1}(u)]^T \in \mathbb{R}^{6N+7} \\ \mathbf{v} &:= [\alpha_0(v), \dots, \alpha_N(v), \gamma_0(v), \gamma_{N+1}(v), \gamma_1(v), \dots, \gamma_N(v), \gamma_1(v), \dots, \gamma_N(v), \\ &\quad \beta_0(v), \beta_N(v), \frac{\beta_0(v)}{2}, \dots, \frac{\beta_{N-1}(v)}{2}, \frac{\beta_1(v)}{2}, \dots, \frac{\beta_N(v)}{2}, \gamma_0(v), \dots, \gamma_{N+1}(v)]^T \in \mathbb{R}^{6N+7} \end{aligned}$$

we get

$$\begin{aligned} T_{\text{ell}} + T_{\text{cons}}^{(u)} + T_{\text{cons}}^{(v)} + T_{\text{penal}} &\leq \mathbf{u}^T \mathbf{v} \leq |\mathbf{u}| |\mathbf{v}| \\ &\leq \left(\sum_{n=0}^N \left(1 + \frac{5}{4}C_\sigma\right) \|\sqrt{c}u'\|_{L^2(I_n)}^2 + 3 \sum_{n=0}^{N+1} \mathbf{a}_n [u(x_n)]^2 \right)^{1/2} \left(\sum_{n=0}^N \left(1 + \frac{5}{4}C_\sigma\right) \|\sqrt{c}v'\|_{L^2(I_n)}^2 + 3 \sum_{n=0}^{N+1} \mathbf{a}_n [v(x_n)]^2 \right)^{1/2} \\ &\leq C_{\text{cont}} \|u\|_\epsilon \|v\|_\epsilon \end{aligned}$$

where $C_{\text{cont}} := (3 + \frac{5}{4}C_\sigma)$. This last estimate together with 1.22 proves the continuity of b_h . \square

1.10 Extension of the Bilinear Form

To simplify the theory we will continue to assume the exact solution to be in $H^2(\Omega)$. When we intend to extend the bilinear form from V_h to a bigger space which contains the exact solution we run into the problem, that $V_h \not\subset H^1(\Omega)$, by extension we also have $V_h \not\subset H^2(\Omega)$. So in contrast to continuous FEM we have to use a bigger space allowing for discontinuities and including both V_h and $H^2(\Omega)$.

A very common approach (see for example [2]) is to choose a vector space sum

$$V := V_h + H^2(\Omega) = \{v \in L^2(\Omega) \mid v = v_h + \tilde{v}, \text{ for } v_h \in V_h, \tilde{v} \in H^2(\Omega)\} \quad (1.25)$$

Alternatively Rivière proposes the usage of *broken Sobolev spaces* in [7]. That is, for a given partition \mathcal{T}_h of Ω we can define

$$H^k(\mathcal{T}_h) := \{v \in L^2(\Omega) \mid v|_I \in H^k(I), \forall I \in \mathcal{T}_h\}$$

for some $k \in \mathbb{N}_0$, this allows for discontinuities and we have $V_h \in H^2(\mathcal{T}_h)$. We will use V as defined in (1.25). Now we extend the bilinear form b_h and the linear functional ℓ_h from V_h to V and formally get

$$b : V \times V \rightarrow \mathbb{R}, \quad \ell : V \rightarrow \mathbb{R} \quad (1.26)$$

where

$$\begin{aligned} b(u, v) &= \sum_{n=0}^N \int_{I_n} c u' v' \, dx - \sum_{n=0}^{N+1} \{c(x_n) u'(x_n)\}[v(x_n)] + \{c(x_n) v'(x_n)\}[u(x_n)] + \sum_{n=0}^{N+1} \mathbf{a}_n[u(x_n)][v(x_n)] \\ \ell(v) &= (f, v)_{L^2(\Omega)} - g_1 c(x_{N+1}^-) v'(x_{N+1}^-) + g_0 c(x_0^+) v'(x_0^+) + \mathbf{a}_{N+1} g_1 v(x_{N+1}^-) + \mathbf{a}_0 g_0 v(x_0^+) \end{aligned}$$

for $u, v \in V$.

This yields the SIPG variational formulation

Find $u \in V$ such that:

$$b(u, v) = \ell(v), \quad \forall v \in V \quad (1.27)$$

Furthermore we extend the energy norm $\|\cdot\|_\epsilon$ as presented in (1.16) to V . Note that this still defines a norm, since in lemma 1.14 we never explicitly need the finite dimensionality of V_h .

Lower Regularity Solutions

The H^2 -regularity assumption on the solution is a common one in the context of convergence theory, but still a rather strict one in general. If we assume only H^1 -regularity, which is a reasonable assumption to make, we encounter the problem of undefined trace values. Recall that from the trace theorem it is known that L^2 -regularity is not enough to yield uniquely defined trace values. In 1d this becomes apparent, since point values of L^2 -functions are not well defined. In particular for a solution $u \in H^1(\Omega)$, we have that $u'(x_n^+), u'(x_n^-)$ are not well defined. Extending the bilinear form b_h to V requires some additional work, which is often done in one of two ways:

1. Lifting operators (see [4])
2. L^2 -projection (see [2])

We will completely circumvent this issue by assuming H^2 -regularity.

Recall that in the proof of Theorem 1.15 for coercivity and continuity of the bilinear form we have used the finite dimensionality of the polynomial spaces over each element in \mathcal{T}_h . This means that the coercivity and continuity properties cannot be simply applied to our extended form b . For this we would have to introduce either lifting or the L^2 -projection as mentioned before.

1.10.1 Consistency of the SIPG Variational Formulation

The final matrix-vector system we solve (yielding our numerical solution) is equivalent to the fully discrete variational formulation (1.9), which in turn is a discretization of not the original weak problem (1.3), but rather the SIPG variational formulation (1.27). We now have to show, that any solution of the SIPG variational formulation is also a solution of the original weak problem and vice versa to ensure our Galerkin approximation approximates the wanted exact solution, i.e. that the SIPG variational formulation is consistent with the weak formulation of the problem.

Theorem 1.16. *The SIPG variational formulation is consistent. That is, for $u \in H^2(\Omega)$ we have: u is a solution of (1.3) if and only if u solves (1.27)*

Proof. Step 1 (" \Leftarrow ").

First suppose $u \in H^2(\Omega)$ is a solution of the SIPG variational formulation (1.27), then clearly

$$b(u, v) = \ell(v) \quad \forall v \in C_c^\infty(\Omega) \subset H^2(\Omega) \quad (1.28)$$

since u, v are continuous we have

$$[u(x_n)] = [v(x_n)] = 0 \quad \forall n \in \{1, \dots, N\}$$

and $v \in C_c^\infty(\Omega)$ implies

$$[v(x_0)] = [v(x_{N+1})] = \{c(x_0)v'(x_0)\} = \{c(x_{N+1})v'(x_{N+1})\} = 0$$

therefore (1.28) becomes

$$a(u, v) = (f, v)_{L^2(\Omega)} \quad \forall v \in C_c^\infty(\Omega) \quad (1.29)$$

where a is the elliptic bilinear form of the weak formulation in (1.3).

Now to recover the boundary conditions first note that since $u \in H^2(\Omega)$ and u satisfies (1.29) we have

$$-(cu')' = f \quad \text{a.e. in } \Omega \quad (1.30)$$

From here we follow exactly the derivation of the discrete SIPG variational formulation in section 1.3. We multiply (1.30) by a test function $v \in V$, integrate over the elements $I \in \mathcal{T}_h$ by parts, sum up over all elements and add the symmetry and penalty terms corresponding to the SIPG bilinear form, hence we find

$$b(u, v) = (f, v)_{L^2(\Omega)} - c(x_{N+1}^-)u(x_{N+1}^-)v'(x_{N+1}^-) + c(x_0^+)u(x_0^+)v'(x_0^+) + \alpha_{N+1}u(x_{N+1}^-)v(x_{N+1}^-) + \alpha_0u(x_0^+)v(x_0^+) \quad (1.31)$$

Subtracting (1.31) from (1.27) yields

$$\left(-c(x_{N+1}^-)v'(x_{N+1}^-) + \alpha_{N+1}v(x_{N+1}^-) \right) \left(u(x_{N+1}^-) - g_1 \right) + \left(c(x_0^+)v'(x_0^+) + \alpha_0v(x_0^+) \right) \left(u(x_0^+) - g_0 \right) = 0$$

Since the above equality holds for any $v \in V$ we can choose fitting test functions v to show that u satisfies the boundary conditions.

Step 2.

Now suppose $u \in H^2(\Omega)$ is a solution of the weak problem (1.3), as already argued in Step 1 we have

$$-(cu')' = f \quad \text{a.e. in } \Omega$$

this yields

$$b(u, v) = (f, v)_{L^2(\Omega)} - c(x_{N+1}^-)u(x_{N+1}^-)v'(x_{N+1}^-) + c(x_0^+)u(x_0^+)v'(x_0^+) + \alpha_{N+1}u(x_{N+1}^-)v(x_{N+1}^-) + \alpha_0u(x_0^+)v(x_0^+)$$

by the derivation steps recalled in Step 1. Applying the boundary conditions, which are fixed by the weak formulation (1.3), shows that u indeed satisfies the SIPG variational formulation (1.27) and we are done.

□

1.11 Error Analysis

In this section we analyze the error between the Galerkin approximation resulting from the discrete SIPG formulation and an exact solution. We will focus on the error in the energy norm $\|\cdot\|_h$ but also mention the resulting error measured in the L^2 -norm.

We start by recalling an important polynomial approximation result for finite elements.

Fix a polynomial degree $r \geq 1$ and let $I = (a, b) \subset \mathbb{R}$ be an element of length $h = b - a$. We denote the linear interpolation operator by

$$\mathcal{I}_h^r : C(\bar{I}) \rightarrow \mathcal{P}^r(I), \quad (1.32)$$

where the distinct interpolation nodes are $\xi_0, \dots, \xi_r \in I$. For our purpose the exact positioning of these nodes can be arbitrary. For readability if the context allows it we will denote the i -th derivative of a function v by $\frac{d^i}{dx^i}v = v^{(i)}$.

Lemma 1.17 (Interpolant estimate). *Let $v \in H^{r+1}(I)$, it holds that*

$$|v - \mathcal{I}_h^r v|_{H^m(I)} \leq h^{r+1-m} |v|_{H^{r+1}(I)} \quad \forall m \in \{0, \dots, r+1\}$$

where $|v|_{H^{r+1}(I)} = \|v^{(r+1)}\|_{L^2(I)}$ denotes the H^{r+1} -seminorm.

Proof. Fix $m \in \{0, \dots, r\}$, we start by first proving the H^m -seminorm

$$|v - \mathcal{I}_h^r v|_{H^m(I)} \leq h^{r+1-m} |v|_{H^{r+1}(I)} \quad (1.33)$$

Define $e := v - \mathcal{I}_h^r v \in H^{r+1}(I) \subset C^r(I)$, by construction of the interpolation operator we have $e(\xi_i) = 0$ for all $i \in \{0, \dots, r\}$, where the ξ_i are the interpolation nodes.

By applying Rolle's theorem in succession we can find r zeroes of e' , $r-1$ zeroes of e'', \dots , and one zero of $e^{(r)}$. Clearly in total there exist some points $y_0, \dots, y_r \in I$ such that

$$e^{(i)}(y_i) = 0 \quad \forall i \in \{0, \dots, r\}.$$

Next we use the fundamental theorem of calculus inductively, let $x_m \in I$ arbitrary and write

$$\begin{aligned} e^{(m)}(x_m) &= \underbrace{e^{(m)}(y_m)}_{=0} + \int_{y_m}^{x_m} e^{(m+1)}(x_{m+1}) \, dx_{m+1} \\ &= \underbrace{\int_{y_m}^{x_m} e^{(m+1)}(y_{m+1}) \, dx_{m+1}}_{=0} + \int_{y_m}^{x_m} \int_{y_{m+1}}^{x_{m+1}} e^{(m+2)}(x_{m+2}) \, dx_{m+1} \, dx_{m+2} \\ &= \dots = \underbrace{\int_{y_m}^{x_m} \dots \int_{y_{r-1}}^{x_{r-1}} e^{(r)}(y_r) \, dx_r \dots dx_{m+1}}_{=0} + \int_{y_m}^{x_m} \dots \int_{y_r}^{x_r} e^{(r+1)}(x_{r+1}) \, dx_{r+1} \dots dx_{m+1}. \end{aligned}$$

Squaring both sides of this equation and integrating over I yields

$$\begin{aligned} \|e^{(m)}\|_{L^2(I)}^2 &= \int_I \left(\int_{y_m}^{x_m} 1 \cdot \int_{y_{m+1}}^{x_{m+1}} \dots \int_{y_r}^{x_r} e^{(r+1)}(x_{r+1}) \, dx_{r+1} \dots dx_{m+2} \, dx_{m+1} \right)^2 \, dx_m \\ &\leq \int_I \underbrace{|x_m - y_m|}_{\leq h} \int_{y_m}^{x_m} \left(\int_{y_{m+1}}^{x_{m+1}} \dots \int_{y_r}^{x_r} e^{(r+1)}(x_{r+1}) \, dx_{r+1} \dots dx_{m+2} \right)^2 \, dx_{m+1} \, dx_m \\ &\leq h \int_I \int_I \left(\int_{y_{m+1}}^{x_{m+1}} \dots \int_{y_r}^{x_r} e^{(r+1)}(x_{r+1}) \, dx_{r+1} \dots dx_{m+2} \right)^2 \, dx_{m+1} \, dx_m \\ &\leq h^2 \int_I \left(\int_{y_{m+1}}^{x_{m+1}} \dots \int_{y_r}^{x_r} e^{(r+1)}(x_{r+1}) \, dx_{r+1} \dots dx_{m+2} \right)^2 \, dx_{m+1} \\ &\leq \dots \leq h^{2(r+1-m)} \int_I |e^{(r+1)}(x_{r+1})|^2 \, dx_{r+1} = h^{2(r+1-m)} |e|_{H^{r+1}(I)}^2, \end{aligned}$$

where we have applied Cauchy-Schwarz inductively. But now since $\mathcal{I}_h^r v \in \mathcal{P}^r(I)$ it holds that $(\mathcal{I}_h^r v)^{(r+1)} \equiv 0$ in I , which implies that $e^{(r+1)} = v^{(r+1)}$ and therefore $|e|_{H^{r+1}}^2 = |v|_{H^{r+1}}^2$. This proves (1.33).

Since we have proven (1.33) for an arbitrary $m \in \{0, \dots, r\}$ we can conclude that

$$|v - \mathcal{I}_h^r v|_{H^m(I)} \leq h^{r+1-m} |v|_{H^{r+1}(I)} \quad \forall m \in \{0, \dots, r\}.$$

If $m = r+1$ then we immediately have $|v - \mathcal{I}_h^r v|_{H^m(I)} = |v|_{H^{r+1}(I)}$.

□

Suppose $u \in V$ satisfies the SIPG variational formulation (1.27) and $u_h \in V_h$ satisfies the discrete SIPG variational formulation (1.9) (here the polynomial degree does not matter), then since $V_h \subset V$ it holds that

$$b(u - u_h, v) = 0 \quad \forall v \in V_h. \quad (1.34)$$

This is commonly known as *Galerkin orthogonality*.

We will now estimate the error of the Galerkin approximation in reference to the continuous solution. To do so recall the following setup.

Let $r \geq 1$ be a polynomial degree, $\mathcal{T}_h = \{I_n\}_{n=0}^{N_h}$, be a partition of Ω with meshsize $h = \max_{n \in \{0, \dots, N_h\}} h_n > 0$, where $h_n = |x_{n+1} - x_n|$ and $I_n = (x_n, x_{n+1})$ for $n = 0, \dots, N_h$.

Definition 1.18. We define the linear interpolation operator

$$\mathcal{I}_h^r : C^0(\Omega) \rightarrow V_h^r(\mathcal{T}_h) \cap C^0(\Omega),$$

such that the face nodes x_n are interpolated for all $n \in \{0, \dots, N_h\}$, i.e. for $v \in C^0(\Omega)$ we have

$$\begin{aligned}\mathcal{I}_h^r v(x_n^-) &= v(x_n) & \forall n \in \{1, \dots, N_h + 1\}, \\ \mathcal{I}_h^r v(x_n^+) &= v(x_n) & \forall n \in \{0, \dots, N_h\}.\end{aligned}$$

Note that for $r > 2$ we require $r - 1$ additional interpolation nodes in each element $I \in \mathcal{T}_h$ besides the two element boundary nodes, for our purpose these can be arbitrary as long as they are distinct.

Theorem 1.19 (Convergence in energy norm). Let $\sigma > 0$ big enough, such that the SIPG bilinear form is coercive on $V_h^r(\mathcal{T}_h)$. Let $u \in H^{r+1}(\Omega)$ be a solution of the SIPG variational formulation (1.27) and $u_h \in V_h^r(\mathcal{T}_h)$ be its Galerkin approximation, i.e. the solution of (1.9).

Then there exists a constant $C_\epsilon(C_{coer}, c_{\max}, \sigma) > 0$ independent of the meshsize h and the solution u , such that

$$\|u - u_h\|_h \leq C_\epsilon h^r |u|_{H^{r+1}(\Omega)}. \quad (1.35)$$

Remark 1.20. 1. The regularity assumption on the solution u is essential, if u is of lower regularity than the polynomial degree, then the optimal convergence rate is not achieved.

2. Recall c_{\max} is the maximum of the coefficient c in the pde (1.1) and C_{coer} is the coercivity constant from theorem 1.15.

Proof. **Step 1.**

We interpolate the solution u and shall for readability write $\mathcal{I}_h^r u = \mathcal{I}_h u \in V_h$. By the triangle inequality we can estimate

$$\|u - u_h\|_\epsilon \leq \|u - \mathcal{I}_h u\|_\epsilon + \|\mathcal{I}_h u - u_h\|_\epsilon. \quad (1.36)$$

By the construction of the interpolant we know that $[(u - \mathcal{I}_h u)(x_n)] = 0$ for all $n \in \{0, \dots, N_h + 1\}$, so the first term of (1.36) becomes

$$\begin{aligned}\|u - \mathcal{I}_h u\|_\epsilon &\leq \|\sqrt{c}(u - \mathcal{I}_h u)'\|_{L^2(\Omega)} \leq \sqrt{c_{\max}} \left(\sum_{n=0}^{N_h} |u - \mathcal{I}_h u|_{H^1(I_n)}^2 \right)^{1/2} \\ &\leq \sqrt{c_{\max}} h^r \left(\sum_{n=0}^{N_h} |u|_{H^{r+1}(I_n)}^2 \right)^{1/2} = \sqrt{c_{\max}} h^r |u|_{H^{r+1}(\Omega)},\end{aligned} \quad (1.37)$$

where in the last inequality we have used the interpolant estimate (lemma 1.17).

Step 2.

Only the second term of (1.36) remains to be estimated. Define $e_h := u_h - \mathcal{I}_h u \in V_h$ and note that

$$b(e_h, e_h) = \underbrace{b(u_h - u, e_h)}_{=0} + b(u - \mathcal{I}_h u, e_h),$$

by Galerkin orthogonality (1.34). Using the coercivity of the bilinear form b on V_h we can estimate

$$\begin{aligned}
C_{\text{coer}} \|e_h\|_h^2 &\leq b(e_h, e_h) = b(u - \mathcal{I}_h u, e_h) \\
&= \underbrace{\left| \sum_{n=0}^{N_h} \int_{I_n} c(u - \mathcal{I}_h u)' e'_h \, dx \right|}_{=: T_1} - \underbrace{\sum_{n=0}^{N_h+1} \{c(x_n)(u - \mathcal{I}_h u)'(x_n)\}[e_h(x_n)]}_{=: T_2} \\
&\quad - \sum_{n=0}^{N_h+1} \{c(x_n)e'_h(x_n)\} \underbrace{[(u - \mathcal{I}_h u)(x_n)]}_{=0} + \sum_{n=0}^{N_h+1} \mathbf{a}_n \underbrace{[(u - \mathcal{I}_h u)(x_n)][e_h(x_n)]}_{=0} \\
&= |T_1 - T_2|.
\end{aligned}$$

We first estimate $|T_1|$, using Cauchy-Schwarz and Young's inequality: $ab \leq \frac{a^2}{2\varepsilon} + \frac{b^2\varepsilon}{2}$, for $a, b \in \mathbb{R}, \varepsilon > 0$, we find

$$\begin{aligned}
|T_1| &\leq \sqrt{c_{\max}} \sum_{n=0}^{N_h} |u - \mathcal{I}_h u|_{H^1(I_n)} \|\sqrt{c} e'_h\|_{L^2(I_n)} \leq \sqrt{c_{\max}} \left(\sum_{n=0}^{N_h} |u - \mathcal{I}_h u|_{H^1(I_n)}^2 \right)^{1/2} \underbrace{\left(\sum_{n=0}^{N_h} \|\sqrt{c} e'_h\|_{L^2(I_n)}^2 \right)^{1/2}}_{\leq \|e_h\|_\epsilon} \\
&\leq \frac{c_{\max}}{C_{\text{coer}}} \sum_{n=0}^{N_h} |u - \mathcal{I}_h u|_{H^1(I_n)}^2 + \frac{C_{\text{coer}}}{4} \|e_h\|_h^2,
\end{aligned}$$

having used $\varepsilon = \frac{C_{\text{coer}}}{2}$. Finally applying lemma 1.17 again yields

$$|T_1| \leq \frac{c_{\max}}{C_{\text{coer}}} h^{2r} |u|_{H^{r+1}(\Omega)}^2 + \frac{C_{\text{coer}}}{4} \|e_h\|_h^2. \quad (1.38)$$

To estimate T_2 we make again use of Young's inequality with $\varepsilon = \frac{\mathbf{a}_n}{2} C_{\text{coer}}$ and find

$$\begin{aligned}
|T_2| &= \left| \sum_{n=0}^{N_h+1} \{c(x_n)(u - \mathcal{I}_h u)'(x_n)\}[e_h(x_n)] \right| \leq \left| \sum_{n=0}^{N_h+1} \frac{\{c(x_n)(u - \mathcal{I}_h u)'(x_n)\}^2}{\mathbf{a}_n C_{\text{coer}}} \right| + \frac{C_{\text{coer}}}{4} \sum_{n=0}^{N_h+1} \mathbf{a}_n [e_h(x_n)]^2 \\
&\leq \left| \sum_{n=0}^{N_h+1} \frac{\{c(x_n)(u - \mathcal{I}_h u)'(x_n)\}^2}{\mathbf{a}_n C_{\text{coer}}} \right| + \frac{C_{\text{coer}}}{4} \|e_h\|_\epsilon^2.
\end{aligned}$$

Fix $n \in \{1, \dots, N_h\}$, then x_n is an interior face node and shared boundary of the elements I_{n-1}, I_n . Using the continuous trace inequality (lemma 1.12) and consequently the interpolant estimate (lemma 1.17) we estimate

$$\begin{aligned}
|(u - \mathcal{I}_h u)'(x_n^+)|^2 &\leq 2h_n^{-1} |u - \mathcal{I}_h u|_{H^1(I_n)}^2 + 2h_n |u - \mathcal{I}_h u|_{H^2(I_n)}^2 \\
&\leq 2h_n^{-1} h_n^{2r} |u|_{H^{r+1}(I_n)}^2 + 2h_n h_n^{2(r-1)} |u|_{H^{r+1}(I_n)}^2 = 4h_n^{2r-1} |u|_{H^{r+1}(I_n)}^2,
\end{aligned}$$

and analogously we find

$$|(u - \mathcal{I}_h u)'(x_n^-)|^2 \leq 4h_{n-1}^{2r-1} |u|_{H^{r+1}(I_{n-1})}^2$$

for the lower element. Recall that $\mathbf{a}_n = \frac{\sigma \max(c(x_n^+), c(x_n^-))}{\min(h_{n-1}, h_n)}$ (see 1.8), using this and the estimates above we find

$$\begin{aligned}
\frac{\{c(x_n)(u - \mathcal{I}_h u)'(x_n)\}^2}{\mathbf{a}_n C_{\text{coer}}} &= \frac{c(x_n)^2 \min(h_{n-1}, h_n)}{\sigma C_{\text{coer}} \max(c(x_n^+), c(x_n^-))} \frac{|(u - \mathcal{I}_h u)'(x_n^-) + (u - \mathcal{I}_h u)'(x_n^+)|^2}{4} \\
&\leq \frac{c(x_n)}{\sigma C_{\text{coer}}} \frac{\min(h_{n-1}, h_n)}{2} \left(|(u - \mathcal{I}_h u)'(x_n^-)|^2 + |(u - \mathcal{I}_h u)'(x_n^+)|^2 \right) \\
&\leq \frac{c(x_n)}{\sigma C_{\text{coer}}} \frac{\min(h_{n-1}, h_n)}{2} \left(4h_{n-1}^{2r-1} |u|_{H^{r+1}(I_{n-1})}^2 + 4h_n^{2r-1} |u|_{H^{r+1}(I_n)}^2 \right) \\
&\leq \frac{2c_{\max}}{\sigma C_{\text{coer}}} h^{2r} \left(|u|_{H^{r+1}(I_{n-1})}^2 + |u|_{H^{r+1}(I_n)}^2 \right).
\end{aligned}$$

Similarly for the boundary nodes, i.e. $n \in \{0, N_h + 1\}$ we find

$$\frac{\{c(x_n)(u - \mathcal{I}_h u)'(x_n)\}^2}{\mathbf{a}_n C_{\text{coer}}} \leq \frac{4c_{\max}}{\sigma C_{\text{coer}}} h^{2r} |u|_{H^{r+1}(I_n)}^2.$$

This gives us now the tools to complete the estimate of $|T_2|$, we find

$$|T_2| \leq \frac{6c_{\max}}{\sigma C_{\text{coer}}} h^{2r} |u|_{H^{r+1}(\Omega)}^2 + \frac{C_{\text{coer}}}{4} \|e_h\|_{\epsilon}^2. \quad (1.39)$$

Combining the estimates (1.38), (1.39) yields

$$C_{\text{coer}} \|e_h\|_{\epsilon}^2 \leq \frac{C_{\text{coer}}}{2} \|e_h\|_{\epsilon}^2 + \frac{c_{\max}}{C_{\text{coer}}} \left(1 + \frac{6}{\sigma}\right) h^{2r} |u|_{H^{r+1}(\Omega)}^2.$$

By subtracting $\frac{C_{\text{coer}}}{2} \|e_h\|_{\epsilon}^2$ from both sides and then multiplying by $\frac{2}{C_{\text{coer}}}$ we find

$$\|e_h\|_{\epsilon}^2 \leq \frac{2c_{\max}}{(C_{\text{coer}})^2} \frac{6 + \sigma}{\sigma} h^{2r} |u|_{H^{r+1}(\Omega)}^2.$$

This, together with the estimate (1.37), we can now finally plug back into (1.36) and get

$$\|u - u_h\|_{\epsilon} \leq \sqrt{c_{\max}} \left(1 + \frac{\sqrt{2(\sigma + 6)}}{C_{\text{coer}} \sqrt{\sigma}}\right) h^r |u|_{H^{r+1}(\Omega)} = C_{\epsilon} h^r |u|_{H^{r+1}(\Omega)},$$

for $C_{\epsilon} := \sqrt{c_{\max}} \left(1 + \frac{\sqrt{2(\sigma + 6)}}{C_{\text{coer}} \sqrt{\sigma}}\right)$, which proofs (1.35). \square

Next we show that measuring the error in the L^2 -norm gives us a power of h more than in the energy norm. This coincides with the convergence rates known from continuous FEM.

Theorem 1.21 (L^2 -convergence). *Under the same assumptions as in theorem 1.19, there exists a constant $C_L(C_{\epsilon}, \Omega, c, \sigma) > 0$ independent of the meshsize h and the solution u , such that*

$$\|u - u_h\|_{L^2(\Omega)} \leq C_L h^{r+1} |u|_{H^{r+1}(\Omega)}. \quad (1.40)$$

Proof. We will prove the theorem using a duality argument often referred to as *Aubin-Nitsche trick*.

Step 1.

Define the error $e := u - u_h \in V$ and consider the dual problem

$$-(c\varphi')' = e \quad \text{in } \Omega, \quad (1.41)$$

$$\varphi(0) = \varphi(1) = 0. \quad (1.42)$$

By Lax-Milgram there exists a unique weak solution $\varphi \in H_0^1(\Omega)$ of (1.41)-(1.42). Furthermore by elliptic regularity it holds that $\varphi \in H^2(\Omega)$ with

$$\|\varphi\|_{H^2(\Omega)} \leq \tilde{C} \|e\|_{L^2(\Omega)}, \quad (1.43)$$

where $\tilde{C} > 0$ is a constant only dependent on Ω and the coefficient $c \in C^1(\Omega)$ (see for example Evans chapter 6.3 [1]). By the consistency of the SIPG variational formulation (see theorem 1.16) φ satisfies

$$b(\varphi, v) = (e, v)_{L^2(\Omega)} \quad \forall v \in V.$$

Note that the right-hand side contains only the L^2 -term since the dual problem is posed with homogeneous Dirichlet boundary conditions (1.42). In particular since $e \in V$ we have $b(\varphi, e) = \|e\|_{L^2(\Omega)}^2$.

Next we interpolate φ using the \mathcal{P}^1 -interpolant \mathcal{I}_h^1 (not \mathcal{P}^r), this means $\mathcal{I}_h^1\varphi \in V_h^r(\mathcal{T}_h)$ since $r \geq 1$. Now using Galerkin orthogonality (1.34) and the symmetry of the bilinear form b we find

$$\|e\|_{L^2(\Omega)}^2 = b(\varphi, e) = b(e, \varphi - \mathcal{I}_h^1\varphi),$$

which in turn we can estimate using Cauchy-Schwarz as follows:

$$\begin{aligned} b(e, \varphi - \mathcal{I}_h^1\varphi) &= \sum_{n=0}^{N_h} \int_{I_n} c e' (\varphi - \mathcal{I}_h^1\varphi)' \, dx - \sum_{n=0}^{N_h+1} \{c(x_n)(\varphi - \mathcal{I}_h^1\varphi)'(x_n)\}[e(x_n)] \\ &\quad - \sum_{n=0}^{N_h+1} \{c(x_n)e'(x_n)\} \underbrace{[(\varphi - \mathcal{I}_h^1\varphi)(x_n)]}_{=0} + \sum_{n=0}^{N_h+1} \mathbf{a}_n[e(x_n)] \underbrace{[(\varphi - \mathcal{I}_h^1\varphi)(x_n)]}_{=0} \\ &\leq \sqrt{c_{\max}} \left(\sum_{n=0}^{N_h} \int_{I_n} |(\varphi - \mathcal{I}_h^1\varphi)'|^2 \, dx \right)^{1/2} \underbrace{\left(\sum_{n=0}^{N_h} \int_{I_n} c |e'|^2 \, dx \right)^{1/2}}_{\leq \|e\|_\epsilon} \\ &\quad + \left(\sum_{n=0}^{N_h+1} \frac{\{c(x_n)(\varphi - \mathcal{I}_h^1\varphi)'(x_n)\}^2}{\mathbf{a}_n} \right)^{1/2} \underbrace{\left(\sum_{n=0}^{N_h+1} \mathbf{a}_n[e(x_n)]^2 \right)^{1/2}}_{\leq \|e\|_\epsilon}. \end{aligned}$$

We know from theorem 1.19 that $\|e\|_\epsilon \leq C_\epsilon h^r |u|_{H^{r+1}(\Omega)}$, and from the proof that

1. $|\varphi - \mathcal{I}_h^1\varphi|_{H^1(\Omega)}^2 \leq h^2 |\varphi|_{H^2(\Omega)}^2 \quad \forall n \in \{0, \dots, N_h\},$
2. $\frac{\{c(x_n)(\varphi - \mathcal{I}_h^1\varphi)'(x_n)\}^2}{\mathbf{a}_n} \leq \gamma_n \frac{c_{\max}}{\sigma} h^2 \left(|\varphi|_{H^2(I_{n-1})}^2 + |\varphi|_{H^2(I_n)}^2 \right) \quad \forall n \in \{0, \dots, N_h+1\},$

where $\gamma_n = 2$ for $n \in \{1, \dots, N_h\}$ and $\gamma_n = 4$ for $n \in \{0, N_h+1\}$. Putting all of the above together we find

$$\|e\|_{L^2(\Omega)}^2 \leq \left(\sqrt{c_{\max}} + \sqrt{\frac{6c_{\max}}{\sigma}} \right) C_\epsilon h^{r+1} |\varphi|_{H^2(\Omega)} |u|_{H^{r+1}(\Omega)} \leq \left(1 + \sqrt{\frac{6}{\sigma}} \right) \sqrt{c_{\max}} C_\epsilon \tilde{C} \|e\|_{L^2(\Omega)} |u|_{H^{r+1}(\Omega)},$$

where we have used the estimate (1.43). Dividing by $\|e\|_{L^2(\Omega)}$ on both sides finally yields

$$\|e\|_{L^2(\Omega)} \leq C_L h^{r+1} |u|_{H^{r+1}(\Omega)},$$

where $C_L = \left(1 + \sqrt{\frac{6}{\sigma}} \right) \sqrt{c_{\max}} C_\epsilon \tilde{C}$.

□

1.12 Numerical Results

1.12.1 Rate of Convergence

To replicate the theoretical rate of convergence we first consider a sequence of uniform meshes. Let $h_l = 2^{-l}$ denote the global (uniform) meshsize and $\mathcal{T}_h^{(l)}$ denote the partitions of Ω for $l = 2, \dots, 9$. As an example we have $\mathcal{T}_h^{(2)} = \{(0, 0.25), (0.25, 0.5), (0.5, 0.75), (0.75, 1)\}$. To begin we worked with the constant coefficient $c \equiv 1$. In this case choosing $\sigma = 10(r+1)^2$ was more than enough to guarantee positive definiteness of the system matrix. We tested our methods programmed in MATLAB using some very simple exact solutions $u \in C^1(\Omega)$, approximated them numerically finding some $u_h \in V_h^r(\mathcal{T}_h^{(l)})$ (for an a priori fixed polynomial degree r) and calculated the L^2 -norm

$$\|u - u_h\|_{L^2(\Omega)} = \left(\int_{\Omega} |u(x) - u_h(x)|^2 \, dx \right)^{1/2} = \left(\sum_{n=0}^{N_l} \int_{I_n} |u(x) - u_h(x)|^2 \, dx \right)^{1/2}$$

and the broken H^1 -norm

$$\|u - u_h\|_{H^1(\mathcal{T}_h^{(l)})} = \left(\sum_{n=0}^{N_l} \int_{I_n} |u'(x) - u'_h(x)|^2 dx \right)^{1/2}.$$

Note that even though u_h is not in $H^1(\Omega)$ and we can therefore not calculate the normal H^1 -norm, we do have

$$u, u_h \in H^1(\mathcal{T}_h^{(l)}) = \{v \in L^2(\Omega) | v|_{I_n} \in H^1(I_n) \text{ for } n = 0, \dots, N_l\}$$

so taking the broken Sobolev norm makes sense also for V_h functions.

Firstly to ensure the exactness of the method we chose the exact solution $u(x) = x^r$ and approximated it using \mathcal{P}^r elements. As expected we got a minimal floating-point error of around 10^{-15} in both L^2 - and $H^1(\mathcal{T}_h^{(l)})$ -norms, slightly increasing for increasing l due to the growing condition number of the system matrix. This indicates exactness of the method.

Next we considered the exact solution $u(x) = e^{-x} \sin(x)$, we observed the with the theory conforming convergence rates $\mathcal{O}(h^r), \mathcal{O}(h^{r+1})$, in the $H^1(\mathcal{T}_h)$ - and L^2 -norm respectively for \mathcal{P}^r -elements. As visible in figures 1.1, 1.2

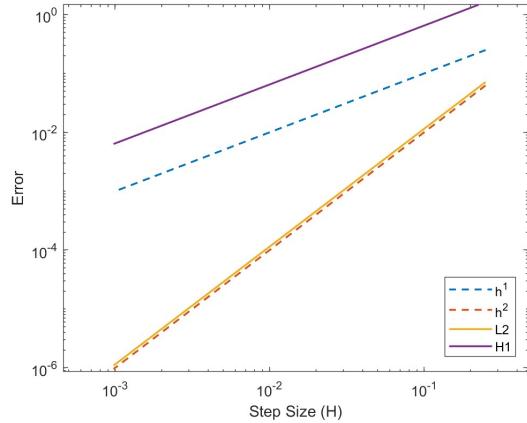


Figure 1.1: Errors of SIPG for P^1 -elements

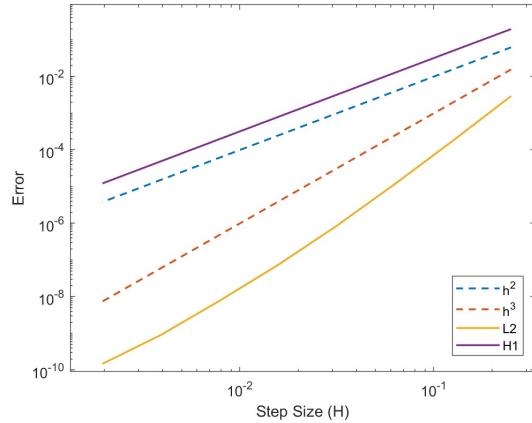


Figure 1.2: Errors of SIPG for P^2 -elements

We plotted the numerical solution for $\mathcal{P}^1, \mathcal{P}^2$ -elements in the figures 1.3, 1.4 respectively on the partition $\mathcal{T}_h^{(3)}$, i.e. for 8 elements with meshsize $h = \frac{1}{8}$. By eye the solution seems continuous, this is due to the high penalization parameter $\sigma = 10(r + 1)^2$. In fact the solution is not continuous and the boundary conditions are not exact either, the error is just not noticeable by eye.

If we reduce the penalization parameter enough such that the bilinear form is no longer coercive we can observe the jumps growing and the discontinuity becomes apparent as is visible in figures 1.5, 1.6 where we have set $\sigma = 1.1$. Maybe interesting to observe is the effect of enforcing boundary conditions weakly. Clearly here the Dirichlet boundary condition $u(0) = 0$ is not fulfilled by the numerical solutions, which would be the case for strongly enforced boundary conditions.

We repeated the tests above using a non constant yet still smooth coefficient $c(x) = \sin(10x) + 2$. We observed the same convergence rates but slightly higher errors overall. This was to be expected since for a non-constant c we introduce a quadrature error. We also observed the same convergence rates when we repeated all the experiments with a sequence of (nested) non-uniform meshes.

1.12.2 Influence of Quadrature Rule on the Convergence Rate

As noted in section 1.7 by using $r + 1$ Gauss-Lobatto nodes (in the context of \mathcal{P}^r -elements) as quadrature nodes and basis nodes at the same time introduces an error when assembling the mass matrix of the system.

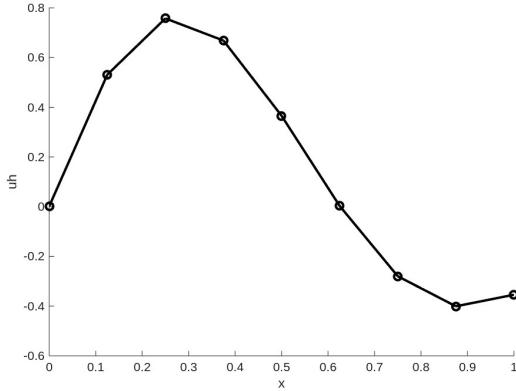


Figure 1.3: numerical SIPG-approximation for P^1 -elements

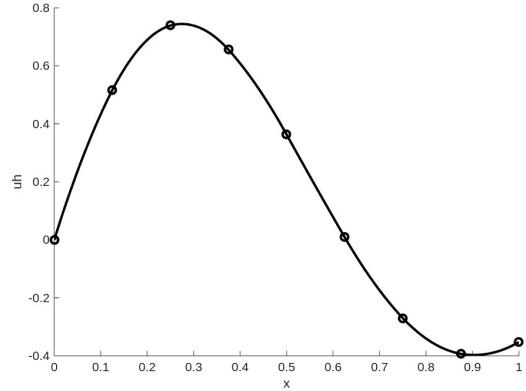


Figure 1.4: numerical SIPG-approximation for P^2 -elements

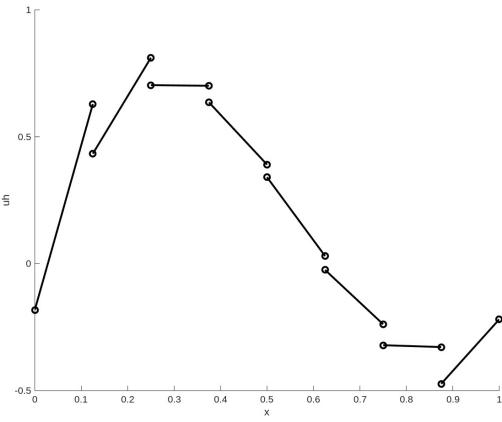


Figure 1.5: numerical SIPG-approximation for P^1 -elements with small penalty

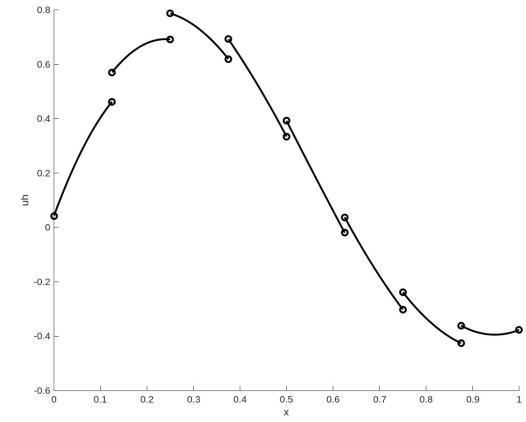


Figure 1.6: numerical SIPG-approximation for P^2 -elements with small penalty

In this subsection we experimentally compare the results of using $r + 1$ Gauss-Lobatto quadrature nodes to approximate the integrals versus using the exact integration values. To be precise we fix a polynomial degree r and our Lagrangian basis with $r + 1$ Gauss-Lobatto nodes as specified in section 1.6 and approximate all integrals first using the Gauss-Lobatto quadrature rule with $r + 1$ nodes, then with $r + 2$ nodes and compare the two.

First we have to consider a slightly different elliptical problem, which requires a mass matrix.

$$-(c(x)u'(x))' + u(x) = f(x) \quad \forall x \in \Omega$$

$$u(0) = g_0, u(1) = g_1$$

We can apply the exact same tools as in the derivation of the variational formulation in section 1.3 and get the discrete SIPG variational formulation.

Find $u_h \in V_h$ such that:

$$b_h(u_h, v) + (u_h, v)_{L^2(\Omega)} = \ell_h(v), \quad \forall v \in V_h \quad (1.44)$$

Now analogously to section 1.5 we write $u_h = \sum_{m=0}^N \sum_{j=0}^r \alpha_j^m \Phi_j^m \in V_h$ and find that (1.44) is equivalent to

$$\sum_{m=0}^N \sum_{j=0}^r \alpha_j^m \left(b_h(\Phi_j^m, \Phi_i^n) + (\Phi_j^m, \Phi_i^n)_{L^2(\Omega)} \right) = \ell_h(\Phi_i^n) \quad \forall i \in \{0, \dots, r\}, \{n = 0, \dots, N\}$$

which in turn is equivalent to the Matrix-Vector system

$$(\mathbf{B} + \mathbf{M})\mathbf{u} = \mathbf{l} \quad (1.45)$$

where as before $[\mathbf{B}]_{T(n,i), T(m,j)} = b_h(\Phi_j^m, \Phi_i^n)$, $[\mathbf{u}]_{T(m,j)} = \alpha_j^m$, $[\mathbf{l}]_{T(n,i)} = \ell_h(\Phi_i^n)$ and furthermore $[\mathbf{M}]_{T(n,i), T(m,j)} = (\Phi_j^m, \Phi_i^n)_{L^2(\Omega)}$.

Next we considered the same setting as described in subsection 1.12.1 and tested the convergence rates for $\mathcal{P}^1, \mathcal{P}^2$ -elements, where we chose the exact solution $u(x) = e^{-x} \sin(x)$ and $c(x) = \sin(10x) + 2$.

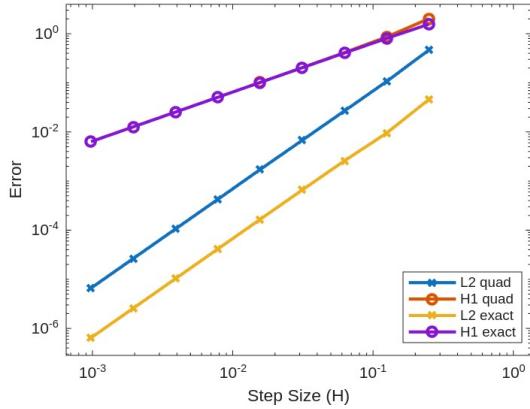


Figure 1.7: Comparison of convergence rates of higher order vs lower order quadrature for \mathcal{P}^1 -elements

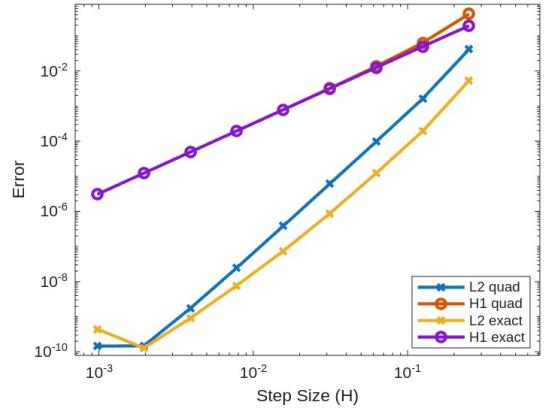


Figure 1.8: Comparison of convergence rates of higher order vs lower order quadrature for \mathcal{P}^2 -elements

Method A: "Error Quad"

This method corresponds to exactly doing what we have been doing before. We assembled the matrices of the system (1.45) as described in section 1.8 and approximated the integrals appearing in the entries of \mathbf{A} , \mathbf{M} , \mathbf{l} using the $r + 1$ node Gauss-Lobatto quadrature rule, where $r \in \{1, 2\}$, meaning the quadrature nodes coincide with the basis nodes. We calculated the $L^2, H^1(\mathcal{T}_h)$ -error between the numerical solution and the exact solution on all meshes and called it L2 quad, H1 quad respectively.

Method B: "Error Exact"

For this method we assembled the matrices \mathbf{A} , \mathbf{M} , \mathbf{l} for the system (1.45) using $r + 2$ Gauss-Lobatto quadrature nodes. This quadrature is exact for polynomials of order $2r + 1$, meaning the entries of the mass matrix \mathbf{M} are calculated exactly. Next we solved the system yielding the numerical solution and in turn calculated the L^2 , $H^1(\mathcal{T}_h)$ -errors on all meshes and called it L2 exact, H1 exact respectively.

In the figures 1.7, 1.8 we compared the error rates of Method A and B and observed no significant difference but a slight downward shift of L2 exact on a logarithmic scale in comparison to L2 quad.

Chapter 2

DG for the Wave Equation

In the last chapter we have derived the necessary theoretical tools and built code to solve the time-independent elliptic problem. This elliptic problem coincides with a time-independent version of the hyperbolic problem

$$u_{tt} - (cu_x)_x = f.$$

We will use a *method of lines* approach to first discretize in space with discontinuous Galerkin finite elements, using what we have derived in the last chapter, and then discretize in time using leapfrog time-integration to solve the problem numerically. This chapter is structured similarly to the last one. First we pose the problem and recall some analysis on regularity and existence, uniqueness of solutions. Next we discretize both in time and space deriving the fully discrete scheme. We will then again elaborate on the actual implementation of the program as well as recalling convergence theory. Finally we will present the numerical experiments reproducing the expected convergence rates where exact solutions are given and simulate the behavior of waves propagating through a waveguide with material properties varying in space and time.

2.1 Problem

Let $\Omega = (0, 1)$ be the domain (waveguide) and let $T > 0$ be the endtime. The 1d wave equation with a time-dependent coefficient is given by

$$u_{tt}(x, t) - (c(x, t)u_x(x, t))_x = f(x, t) \quad \forall (x, t) \in \Omega \times (0, T], \quad (2.1a)$$

$$u(0, t) = g_0(t), \quad u(1, t) = g_1(t) \quad \forall t \in [0, T], \quad (2.1b)$$

$$u(x, 0) = u_0(x), \quad u_t(x, 0) = v_0(x) \quad \forall x \in \Omega. \quad (2.1c)$$

Similarly to the elliptic case we require the coefficient $c \in C^1(\overline{\Omega} \times [0, T])$ to be bounded by

$$0 < c_{\min} \leq c(x, t) \leq c_{\max} < \infty \quad \forall (x, t) \in \overline{\Omega} \times [0, T]. \quad (2.2)$$

We assume the forcing term $f \in L^2(0, T; L^2(\Omega))$, the initial displacement $u_0 \in H^1(\Omega)$, and the initial velocity $v_0 \in L^2(\Omega)$. First we assume homogeneous Dirichlet boundary conditions, meaning $g_0 \equiv g_1 \equiv 0$. Multiplying the pde (2.1a) by a test function $v \in C_c^\infty(\Omega)$ and integrating by parts gives us the weak formulation:

Find $u \in L^2(0, T; H_0^1(\Omega))$, with $u_t \in L^2(0, T; L^2(\Omega))$, $u_{tt} \in L^2(0, T; H^{-1}(\Omega))$, such that

$$\langle u_{tt}(t), v \rangle + a(u(t), v; t) = (f(t), v)_{L^2(\Omega)} \quad \forall v \in C_c^\infty(\Omega) \quad \text{for a.e. } t \in (0, T), \quad (2.3a)$$

$$u(\cdot, 0) = u_0, \quad u_t(\cdot, 0) = v_0, \quad (2.3b)$$

where

$$a(u(t), v; t) = \int_{\Omega} c(x, t) u_x(x, t) v_x(x) dx$$

denotes the standard (time-dependent) elliptic bilinear form and $\langle \cdot, \cdot \rangle$ denotes the duality pairing between $H^{-1}(\Omega)$ and $H_0^1(\Omega)$. Note that both the time derivatives as well as the space derivatives have to be understood in a weak sense. It is well known that the problem (2.3) is well posed (see section 7.2 Theorem 3 in [1]). Furthermore we even have

$$u \in C^1([0, T]; L^2(\Omega)) \cap C([0, T]; H_0^1(\Omega)),$$

which ensures the validity of the initial conditions (2.3b).

We can furthermore deduce existence and uniqueness of the weak problem for inhomogeneous Dirichlet boundary conditions, if they are sufficiently regular. Suppose for example $g_0 = g(0, \cdot), g_1 = g(1, \cdot)$ for some $g \in C^2([0, T]; H^2(\Omega))$. We can write $u = w + g$, where $w \in C^1([0, T]; L^2(\Omega)) \cap C([0, T]; H_0^1(\Omega))$ is the unique (weak) solution of

$$\begin{aligned} w_{tt} - (cw_x)_x &= f - g_{tt} + (cg_x)_x && \text{in } \Omega \times (0, T], \\ w(0, \cdot) &= w(1, \cdot) = 0 && \text{in } [0, T], \\ w(\cdot, 0) &= u_0 - g(\cdot, 0), \quad w_t(\cdot, 0) = v_0 - g_t(\cdot, 0) && \text{in } \Omega. \end{aligned}$$

With this we can conclude the problem to be well-posed and continue with the discretization procedure.

2.2 Variational Formulation and Fully-Discrete-Scheme

2.2.1 Discretization in Space

We start by discretizing in space. To simplify and clarify calculations we assume the solution to be $u \in C^2([0, T]; H^2(\Omega))$. We fix some time $t \in [0, T]$, the discretization in space is analogous to the elliptical case for each time t (see section 1.3 in chapter 1), so we will only briefly recall the steps.

First we discretize the domain by choosing a partition \mathcal{T}_h of Ω using the notation introduced in the last chapter. We multiply (2.1) by a test function $v \in V_h$, where $V_h^r(\mathcal{T}_h)$ is the discontinuous finite element space defined (1.4), multiply over each element $I \in \mathcal{T}_h$ by parts and finally sum up over all elements. After adding the symmetry and penalty terms necessary for the SIPG variational form we find the semi-discrete SIPG variational formulation:

Find $u_h \in C^2([0, T]; V_h)$ with $u_h(\cdot, 0) = \mathcal{I}_h u_0, u_t(\cdot, 0) = \mathcal{I}_h v_0$ such that

$$(\partial_t^2 u_h(t), v)_{L^2(\Omega)} + b(u_h, v; t) = \ell(v; t) \quad \forall v \in V_h, t \in [0, T], \quad (2.4)$$

where

$$\begin{aligned} b(u, v; t) &= \sum_{n=0}^N \int_{I_n} c(x, t) u_x(x, t) v_x(x) dx - \sum_{n=0}^{N+1} \{c(x_n, t) u_x(x_n, t)\}[v(x_n)] + \{c(x_n, t) v_x(x_n)\}[u(x_n, t)] \\ &\quad + \sum_{n=0}^{N+1} \mathbf{a}_n(t)[u(x_n, t)][v(x_n)], \\ \ell(v; t) &= (f(t), v)_{L^2(\Omega)} - g_1(t)c(x_{N+1}^-, t)v_x(x_{N+1}^-) + g_0(t)c(x_0^+, t)v_x(x_0^+) \\ &\quad + \mathbf{a}_{N+1}(t)g_1(t)v(x_{N+1}^-) + \mathbf{a}_0(t)g_0(t)v(x_0^+). \end{aligned}$$

Recall $\mathcal{I}_h : C(\bar{\Omega}) \rightarrow V_h$ denotes the interpolation operator (1.32) (one could also use L^2 -projection). Clearly we can not expect the semi-discrete solution to satisfy continuous initial conditions, therefore the semi-discrete formulation must introduce projected initial condition.

Next we can rewrite (2.4) into matrix-vector form. To do so let Φ_1, \dots, Φ_M be a basis of V_h and therefore write

$$u_h(x, t) = \sum_{n=1}^M \alpha_n(t) \Phi_n(x).$$

We can plug this into the semi-discrete SIPG variational formulation (2.4), using linearity and the fact that testing against elements in a vector space is equivalent to testing against basis functions yields the matrix-vector system of ODEs

$$\mathbf{M}\ddot{\mathbf{u}}(t) + \mathbf{B}(t)\mathbf{u}(t) = \mathbf{l}(t) \quad \forall t \in [0, T]. \quad (2.5)$$

Here $[\mathbf{M}]_{i,j} = (\Phi_j, \Phi_i)_{L^2(\Omega)}$ is the *mass matrix*, which in our case is not time-dependent.

$[\mathbf{B}(t)]_{i,j} = b(\Phi_j, \Phi_i; t)$ is the *stiffness matrix*, similar to the elliptic case but now with a time-dependent coefficient c and $[\mathbf{l}(t)]_i = \ell(\Phi_i; t)$ the *load vector*, also time-dependent. Finally $[\mathbf{u}(t)]_i = \alpha_i(t)$ denotes the solution vector with the time-dependent coefficients as entries uniquely determining the semi-discrete solution.

2.2.2 Discretization in Time

With this we come to the time-discretization. (2.5) is a second order system of ODEs, meaning we can use a chosen time-integration scheme to discretize the solution in time. One benefit of using discontinuous Galerkin to discretize in space is that the mass matrix \mathbf{M} is block-diagonal, since by the construction of the basis functions, each element is fully decoupled from the rest, this simplifies solving any linear system with the mass matrix as the system matrix. We can capitalize on this property of DG-FEM by choosing an explicit time integration scheme.

First we choose a (stable) stepsize $\Delta t > 0$, we write $t_m = m \cdot \Delta t \in [0, T]$ and denote

$$\mathbf{u}_m := \mathbf{u}(t_m), \quad \mathbf{B}_m := \mathbf{B}(t_m), \quad \mathbf{l}_m := \mathbf{l}(t_m).$$

We introduce a second order finite difference quotient to approximate the second order time derivative

$$\ddot{\mathbf{u}}_m \approx \frac{\mathbf{u}_{m+1} - 2\mathbf{u}_m + \mathbf{u}_{m-1}}{\Delta t^2},$$

plugging this into (2.5) and reforming yields the fully discrete, explicit leapfrog scheme

$$\mathbf{M}\mathbf{u}_{m+1} = \Delta t^2 \mathbf{l}_m + (2\mathbf{M} - \Delta t^2 \mathbf{B}_m)\mathbf{u}_m - \mathbf{u}_{m-1}. \quad (2.6)$$

This being a multistep scheme we have to initialize $\mathbf{u}_0, \mathbf{u}_1$. To do so we use the interpolated values of the initial displacement to initialize \mathbf{u}_0 , meaning we can write $\mathcal{I}_h u_0 = \sum_{n=1}^M \alpha_n(0) \Phi_n$ for some coefficients $\alpha_n(0) \in \mathbb{R}$ and get

$$[\mathbf{u}_0]_i = \alpha_i(0).$$

In fact since we have chosen a Lagrangian basis for the space V_h , finding the coefficients $\alpha_i(0)$ corresponds to evaluating the to be interpolated function u_0 at the corresponding node x_i on which the basis function Φ_i is stationary, i.e. $\alpha_i(0) = u_0(x_i)$.

For \mathbf{u}_1 we Taylor-expand

$$\mathbf{u}_1 \approx \mathbf{u}(\Delta t) = \mathbf{u}_0 + \Delta t \mathbf{v}_0 + \frac{\Delta t^2}{2} \ddot{\mathbf{u}}(0) + \mathcal{O}(\Delta t^3),$$

using (2.5) we can rewrite $\ddot{\mathbf{u}}(0) = \mathbf{M}^{-1}(\mathbf{l}_0 - \mathbf{B}_0 \mathbf{u}_0)$ and define

$$\mathbf{u}_1 := \mathbf{u}_0 + \Delta t \mathbf{v}_0 + \frac{\Delta t^2}{2} \mathbf{M}^{-1}(\mathbf{l}_0 - \mathbf{B}_0 \mathbf{u}_0).$$

To guarantee stability of the leapfrog scheme Δt has to be chosen small enough to satisfy the CFL condition, here this means specifically that the matrix $\mathbf{M} - \frac{\Delta t^2}{4} \mathbf{B}(t)$ has to be symmetric positive definite (SPD) for all times. This is equivalent to requiring that

$$\Delta t < 2 \frac{\|\mathbf{M}\|_2}{\|\mathbf{B}(t)\|_2} \quad \forall t \in [0, T],$$

where $\|\mathbf{M}\|_2^2 = \sigma_{\max}(\mathbf{M})$ is the spectral matrix norm. Indeed the expression is well-defined, since $\mathbf{B}(t)$ itself is SPD for all $t \in [0, T]$. This follows from a slight adaptation of the coercivity proof of the SIPG bilinear form on V_h (Theorem 1.15). For further information on the stability of leapfrog see [3].

2.3 Absorbing Boundary Conditions

When simulating a wave a classical example of a domain on which to do so is a *wave guide*, a rectangular domain through which the wave is propagated. Since we restrict ourselves to the 1d case in this thesis, trivially every domain interval we can choose corresponds to a sort of waveguide. We are interested in simulating a wave which propagates through this waveguide only coming from one direction. Lets say we want to send a wave through the waveguide originating from the lower boundary of the domain. We can achieve that by imposing a corresponding Dirichlet or Neumann boundary condition at the point of entry. A question arises:

What boundary condition should be imposed at the upper boundary?

The answer may differ depending on the purpose of the simulation. The problem with using a simple Dirichlet or Neumann boundary condition at the upper boundary as well is that the incident wave is directly influenced and modified by the upper boundary condition. Lets say we impose a Dirichlet boundary condition at the lower boundary and a homogeneous Neumann boundary condition at the upper boundary

$$u(0) = \sin(x - t), \quad u_x(1) = 0,$$

then the upper boundary will reflect the full wave back. This clearly pollutes the incident wave and if we wanted to observe how the incident wave changes by passing through an inhomogeneous medium, the reflected wave would overshadow any slight modulation stemming from variations in the material. In reality in this context we are interested in simulating the problem on an unbounded domain. We would like the wave to just propagate through the upper boundary and not return. To simulate an artificial boundary of this kind we need to impose artificial boundary conditions, these are often called *absorbing*-, *transparent*-, or *non-reflecting* boundary conditions.

We impose the first order absorbing boundary condition

$$\frac{\partial u}{\partial t} + \sqrt{c} \frac{\partial u}{\partial n} = 0 \tag{2.7}$$

at the upper boundary (exit of the waveguide), which corresponds to

$$u_t(10, \cdot) + \sqrt{c(10, \cdot)} u_x(10, \cdot) = 0 \quad \text{in } [0, T].$$

To implement this boundary condition into our time-marching scheme we have to first treat the bilinear form b at the upper boundary as if we were to implement a Neumann boundary condition (see section 1.4), meaning after integrating by parts and summing over all elements we don't introduce the symmetry and penalty terms at the corresponding node x_{N+1} , instead we apply the absorbing boundary condition and get

$$-\{c(x_{N+1}, t) \partial_x u_h(x_{N+1}, t)\}[v(x_{N+1})] = \{\sqrt{c(x_{N+1}, t)} \partial_t u_h(x_{N+1}, t)\}[v(x_{N+1})].$$

the semi-discrete scheme (2.4) turns into

$$(\partial_t^2 u_h(t), v)_{L^2(\Omega)} + b(u_h, v; t) + \{\sqrt{c(x_{N+1}, t)} \partial_t u_h(x_{N+1}, t)\}[v(x_{N+1})] = \ell(v; t) \quad \forall v \in V_h, t \in [0, T],$$

where

$$\begin{aligned} b(u, v; t) &= \sum_{n=0}^N \int_{I_n} c(x, t) u_x(x, t) v_x(x) dx - \sum_{n=0}^N \{c(x_n, t) u_x(x_n, t)\}[v(x_n)] + \{c(x_n, t) v_x(x_n)\}[u(x_n, t)] \\ &\quad + \sum_{n=0}^N \mathbf{a}_n(t)[u(x_n, t)][v(x_n)], \end{aligned}$$

$$\ell(v; t) = (f(t), v)_{L^2(\Omega)} + g_0(t) c(x_0^+, t) v_x(x_0^+) + \mathbf{a}_0(t) g_0(t) v(x_0^+)$$

are slightly modified.

And the semi-discrete matrix-vector system becomes

$$\mathbf{M}\ddot{\mathbf{u}}(t) + \mathbf{B}(t)\mathbf{u}(t) + \mathbf{R}(t)\dot{\mathbf{u}}(t) = \mathbf{l}(t) \quad \forall t \in [0, T],$$

where $[\mathbf{R}(t)]_{i,j} = \{\sqrt{c(x_{N+1}, t)}\Phi_j(x_{N+1}, t)\}[\Phi_i(x_{N+1})]$ is a matrix where the only non-zero entry is the one at the bottom right corner where $i = j = \dim(V_h)$ and Φ_i the basis centered at x_{N+1} is.

From here on we can use a (second order) centered finite difference to approximate

$$\dot{\mathbf{u}}(t) \approx \frac{\mathbf{u}(t + \Delta t) - \mathbf{u}(t - \Delta t)}{2\Delta t}$$

and similarly to before we find the slightly modified, fully-discrete leapfrog scheme

$$\left(\mathbf{M} + \frac{\Delta t}{2}\mathbf{R}_m\right)\mathbf{u}_{m+1} = \Delta t^2\mathbf{l}_m + (2\mathbf{M} - \Delta t^2\mathbf{B}_m)\mathbf{u}_m + \left(\frac{\Delta t}{2}\mathbf{R}_m - \mathbf{I}\right)\mathbf{u}_{m-1}.$$

We have to also make a slight adjustment to the initialization of the solution at the first time-step \mathbf{u}_1 . By using a Taylor-expansion and the semi-discrete scheme we have just derived we can define

$$\mathbf{u}_1 := \mathbf{u}_0 + \Delta t\mathbf{v}_0 + \frac{\Delta t^2}{2}\mathbf{M}^{-1}(\mathbf{l}_0 - \mathbf{B}_0\mathbf{u}_0 - \mathbf{R}_0\mathbf{v}_0).$$

2.4 Reproducing Convergence Rates

To test our code we reproduced the expected convergence rates for smooth solutions and coefficients. We chose the waveguide $\Omega = (0, 10)$ as domain and fixed the endtime $T = 10$. Next we defined two exact solutions

$$u_1(x, t) = e^{-(x-t+2)^2}, \quad u_2(x, t) = \sin(x - t - \pi)$$

and the coefficients

$$c_1(x, t) = 1, \quad c_2(x, t) = (\sin(x) + 2)(\cos(t) + 2), \quad c_3(x, t) = \sin(x) + 2, \quad c_4(x, t) = \sin(t) + 2.$$

We imposed Dirichlet boundary conditions on both boundary points based on the exact solution and calculated the required forcing term $f_{i,j} = \partial_t^2 u_i - \partial_x(c_j \partial_x u_i)$ such that the chosen exact solution u_i solves the problem. In total for $i \in \{1, 2\}$ and $j \in \{1, 2, 3, 4\}$ we solved the pde:

Find $u \in C^2([0, T]; H^2(\Omega))$ such that

$$u_{tt} - (c_j u_x)_x = \partial_t^2 u_i - \partial_x(c_j \partial_x u_i) \quad \text{in } \Omega \times [0, T], \tag{2.8a}$$

$$u(0, \cdot) = u_i(0, \cdot), \quad u(10, \cdot) = u_i(10, \cdot) \quad \text{in } [0, T], \tag{2.8b}$$

$$u(\cdot, 0) = u_i(\cdot, 0), \quad u_t(\cdot, 0) = \partial_t u_i(\cdot, 0) \quad \text{in } \Omega. \tag{2.8c}$$

We considered a sequence of uniform meshes of meshsize $h \in \{1, \frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \frac{1}{16}, \frac{1}{32}\}$. For each mesh we fixed the stepsize $\Delta t = \frac{h}{100(r+1)}$, where $r \in \{1, 2\}$ is the polynomial degree of the finite element space. We calculated the numerical solution u_h for all timesteps using leapfrog in time and SIPG in space as presented in (2.6). We calculated the L^2 -error, the broken H^1 -error and the error in the energy norm between the exact solution u_i and the numerical solution u_h at the final time T , meaning we considered the quantities

$$\|u_i(\cdot, T) - u_h(\cdot, T)\|_{L^2(\Omega)}, \quad \|u_i(\cdot, T) - u_h(\cdot, T)\|_{H^1(\mathcal{T}_h)}, \quad \|u_i(\cdot, T) - u_h(\cdot, T)\|_h.$$

Appendix A

Prerequisites

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