

Draft on BV Project

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1 The continuous and discrete problem

The continuous problem involves a low-order term with positive parameter $\alpha > 0$ and a given right-hand side $f \in L^2(\Omega)$ in a bounded polyhedral Lipschitz domain $\Omega \subset \mathbb{R}^n$ and minimizes

$$E(v) := \frac{\alpha}{2} \|v\|_{L^2(\Omega)}^2 + |v|_{\text{BV}(\Omega)} + \|v\|_{L^1(\partial\Omega)} - \int_{\Omega} f v \, dx \quad (1.1)$$

amongst all functions $v \in V := \text{BV}(\Omega) \cap L^2(\Omega)$. The BV seminorm $|v|_{\text{BV}}$ is the total variation $\int_{\Omega} |\nabla v| \, dx$ and well defined for $v \in \text{BV}(\Omega)$ and is equal to the $W^{1,1}$ seminorm for $v \in W^{1,1}(\Omega)$. The trace theorem for BV functions shows that they are Lebesgue functions along the boundary and so the term $\|v\|_{L^1(\partial\Omega)}$ is well defined and models homogeneous boundary conditions.

It is well established that the energy has a unique minimizer u in V and it is a corollary of a lemma of Brezis that, for a convex domain and $f \in H^1(\Omega)$, u belongs to $H_0^1(\Omega)$.

The nonconforming discretization of this section is nonconforming in two aspects. First it utilizes nonconforming P_1 finite element functions named after Crouzeix-Raviart [4],

$$\text{CR}_0^1(\mathcal{T}) := \{v_{\text{CR}} \in P_1(\mathcal{T}) \mid v_{\text{CR}} \text{ is continuous at midpoints of interior sides and vanishes at midpoints of boundary sides}\}.$$

Here and throughout this paper, $P_k(\mathcal{T})$ abbreviates the piecewise polynomials at total degree at most k with respect to a shape-regular triangulation \mathcal{T} of Ω into simplices with the set of nodes \mathcal{N} and set of sides \mathcal{F} (resp. interior sides $\mathcal{F}(\Omega)$); the L^2 projection onto $P_0(\mathcal{T})$ is the ~~associated~~ averaging operator Π_0 .

The subsequent discrete problem is nonconforming in a second way ^u it replaces the distributional gradient ∇ by ~~the~~ ^{the} piecewise action of it, written ∇_{NC} and, for any $v \in H^1(\mathcal{T})$, defined as $\nabla_{\text{NC}} v \in L^2(\Omega, \mathbb{R}^n)$ on $T \in \mathcal{T}$

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by $(\nabla_{\text{NC}} v)|_T := \nabla(v|_{\text{int}(T)})$; abbreviate the discrete substitute $|\cdot|_{1,1,\text{NC}} := \|\nabla_{\text{NC}} \cdot\|_{L^1(\Omega)}$ ~~of the BV(Ω) + $L^2(\partial\Omega)$ seminorm~~. It is well understood that

$$|v_{\text{CR}}|_{\text{BV}} + \|v_{\text{CR}}\|_{L^1(\partial\Omega)} = |v_{\text{CR}}|_{1,1,\text{NC}} + \sum_{F \in \mathcal{F}} \int_F |[v_{\text{CR}}]_F| ds$$

with the jump $[v_{\text{CR}}]_F$ of v_{CR} across the interior side F is easily computable for all $v_{\text{CR}} \in \text{CR}(\mathcal{T})$.

The ~~discrete~~ problem minimizes the nonconforming energy

$$E_{\text{NC}}(v_{\text{CR}}) := \frac{\alpha}{2} \|v_{\text{CR}}\|_{L^2(\Omega)}^2 + |v_{\text{CR}}|_{1,1,\text{NC}} - \int_{\Omega} f v_{\text{CR}} dx \quad (1.2)$$

amongst all ~~Crouzeix-Raviart functions~~ $v_{\text{CR}} \in \text{CR}_0^1(\mathcal{T})$.

The convex modulus function $|\cdot|$ has the multivalued subgradient sign, defined by $\text{sign } F := \{F/|F|\}$ for $F \in \mathbb{R}^n \setminus \{0\}$ and $\text{sign } 0 := \overline{B(0,1)}$ the closed unit ball in \mathbb{R}^n endowed with the Euclidean scalar product " \cdot ".

Theorem 1.1 (characterization of discrete solutions). (a) There exist a unique discrete minimizer $u_{\text{CR}} = \text{argmin } E_{\text{NC}}(\text{CR}_0^1(\mathcal{T}))$.

(b) The minimizer u_{CR} is equivalently characterized as the solution $u_{\text{CR}} \in \text{CR}_0^1(\mathcal{T})$ to the variational inequality

$$(f - \alpha u_{\text{CR}}, v_{\text{CR}} - u_{\text{CR}})_{L^2(\Omega)} \leq |v_{\text{CR}}|_{1,1,\text{NC}} - |u_{\text{CR}}|_{1,1,\text{NC}} \quad (\text{VI})$$

for all $v_{\text{CR}} \in \text{CR}_0^1(\mathcal{T})$.

(c) The minimizer u_{CR} is equivalently characterized by $u_{\text{CR}} \in \text{CR}_0^1(\mathcal{T})$ and the existence of some $\Lambda_0 \in P_0(\mathcal{T}; \overline{B(0,1)})$ with $\Lambda_0 \cdot \nabla_{\text{NC}} u_{\text{CR}} = |\nabla_{\text{NC}} u_{\text{CR}}|$ a.e. in Ω and

$$(f - \alpha u_{\text{CR}}, v_{\text{CR}})_{L^2(\Omega)} = (\Lambda_0, \nabla_{\text{NC}} v_{\text{CR}})_{L^2(\Omega)} \text{ for all } v_{\text{CR}} \in \text{CR}_0^1(\mathcal{T}). \quad (1.2a)$$

Proof. Standard arguments on the quadratic growth and the continuity of the discrete energy $E_{\text{NC}}(v_{\text{CR}})$ with respect to v_{CR} in the fixed finite-dimensional space $\text{CR}_0^1(\mathcal{T})$ provide the existence of a minimizer u_{CR} of E_{NC} in $\text{CR}_0^1(\mathcal{T})$. Moreover, this minimizer is equivalently characterized as the solution of the variational inequality and more details for (a) and (b) are omitted. One constructive way for the existence proof of either the discrete or the continuous minimization problem utilizes a regularization of the L^1 norm. For instance, the modulus $|\cdot|$ may be replaced by a differentiable upper bound $|\cdot|_{\varepsilon}$, defined for any $\varepsilon > 0$ by

$$|F|_{\varepsilon} := \sqrt{\varepsilon^2 + F \cdot F} \quad \text{for all } F \in \mathbb{R}^n.$$

This leads to a regularized nonconforming energy (1.2) with the substitution of $\int_{\Omega} |\nabla_{\text{NC}} v_{\text{CR}}|_{\varepsilon} dx$ for $|v_{\text{CR}}|_{1,1,\text{NC}}$. The same substitution applies to the discrete variational inequality in (b), which becomes an equality for $|\bullet|_{\varepsilon}$ is differentiable for any positive ε . For any $\varepsilon > 0$ there exists a unique minimizer to the regularized nonconforming energy and the necessary stationary condition applies for the smooth functional and results in that $u_{\text{CR},\varepsilon} \in \text{CR}_0^1(\mathcal{T})$ satisfies

$$(f - \alpha u_{\text{CR},\varepsilon}, v_{\text{CR}})_{L^2(\Omega)} = \int_{\Omega} \Lambda_{\varepsilon} \cdot \nabla_{\text{NC}} v_{\text{CR}} dx \text{ for all } v_{\text{CR}} \in \text{CR}_0^1(\mathcal{T})$$

with the abbreviation

$$\Lambda_{\varepsilon} := \frac{\nabla_{\text{NC}} u_{\text{CR},\varepsilon}}{\sqrt{\varepsilon^2 + |\nabla_{\text{NC}} u_{\text{CR},\varepsilon}|^2}} \in P_0(\mathcal{T}; \overline{B(0,1)}). \quad (1.3)$$

The test with $v_{\text{CR}} = u_{\text{CR},\varepsilon}$ and standard arguments reveal that $u_{\text{CR},\varepsilon} \in \text{CR}_0^1(\mathcal{T})$ is bounded (in any norm for the fixed finite-dimensional vector space $\text{CR}_0^1(\mathcal{T})$) as $\varepsilon \rightarrow 0^+$. Any accumulation point $(u_{\text{CR}}, \Lambda_0) \in \text{CR}_0^1(\mathcal{T}) \times P_0(\mathcal{T}; \mathbb{R}^n)$ of bounded subsequences $(u_{\text{CR},\varepsilon}, \Lambda_{\varepsilon})$ as $\varepsilon \rightarrow 0^+$ satisfies

$$(f - \alpha u_{\text{CR}}, v_{\text{CR}})_{L^2(\Omega)} = \int_{\Omega} \Lambda_0 \cdot \nabla_{\text{NC}} v_{\text{CR}} dx \text{ for all } v_{\text{CR}} \in \text{CR}_0^1(\mathcal{T}); \quad (1.4)$$

$$|\Lambda_0| \leq 1 \quad \text{and} \quad \Lambda_0 \cdot \nabla_{\text{NC}} u_{\text{CR}} = |\nabla_{\text{NC}} u_{\text{CR}}| \quad \text{a.e. in } \Omega. \quad (1.5)$$

Substitute the test function $v_{\text{CR}} \in \text{CR}_0^1(\mathcal{T})$ in (1.4) by $v_{\text{CR}} - u_{\text{CR}}$ (for some fixed limit u_{CR} and) any $v_{\text{CR}} \in \text{CR}_0^1(\mathcal{T})$ to obtain with (1.5) the identity

$$(f - \alpha u_{\text{CR}}, v_{\text{CR}} - u_{\text{CR}})_{L^2(\Omega)} = \int_{\Omega} \Lambda_0 \cdot \nabla_{\text{NC}} v_{\text{CR}} dx - |u_{\text{CR}}|_{1,1,\text{NC}}.$$

Since $|\Lambda_0| \leq 1$ a.e. in Ω implies $\int_{\Omega} \Lambda_0 \cdot \nabla_{\text{NC}} v_{\text{CR}} dx \leq |v_{\text{CR}}|_{1,1,\text{NC}}$, this becomes the discrete variational inequality (b). In other words, any selected accumulation point u_{CR} of the discrete solutions $u_{\text{CR},\varepsilon} \in \text{CR}_0^1(\mathcal{T})$ as $\varepsilon \rightarrow 0^+$ is equal to the unique solution u_{CR} in (b). This implies convergence $u_{\text{CR},\varepsilon} \rightarrow u_{\text{CR}}$ as $\varepsilon \rightarrow 0^+$ but in general not for Λ_{ε} from (1.3). However, any accumulation point Λ_0 (of the possibly many choices) leads to (1.4)-(1.5). \square

Remark 1.2 (dual variable). The condition (1.5) equivalently reads $\Lambda_0 \in \text{sign } \nabla_{\text{NC}} u_{\text{CR}}$ for the discrete minimizer u_{CR} . If $\nabla_{\text{NC}} u_{\text{CR}} \neq 0$ on $T \in \mathcal{T}$, then the dual variable $\Lambda_0 = \nabla_{\text{NC}} u_{\text{CR}} / |\nabla_{\text{NC}} u_{\text{CR}}|$ is unique on T .

Example 1 (nonuniqueness of dual variable). Given $f \equiv 0$, the unique minimizer vanishes $u_{\text{CR}} \equiv 0$ a.e. in Ω , while any $v_{\text{C}} \in S^1(\mathcal{T})$ with $|\text{Curl } v_{\text{C}}| \leq 1$ a.e. in Ω leads to $\Lambda_0 := \text{Curl } v_{\text{C}} \in P_0(\mathcal{T}; \overline{B(0,1)})$ with $\langle \Lambda_0, \nabla_{\text{NC}} v_{\text{C}} \rangle = 0$.

Theorem 1.3 (convergence). For any sequence $(\mathcal{T}_\ell)_{\ell \in \mathbb{N}}$ of meshes in \mathbb{T} with respective discrete solutions $(u_{\text{CR}}^{(\ell)})_{\ell \in \mathbb{N}}$ to (1.2) and maximal mesh-sizes $h_\ell \rightarrow 0^+$ as $\ell \rightarrow \infty$ the unique solution u to (1.1) satisfies

$$\lim_{\ell \rightarrow \infty} \|u - u_{\text{CR}}^{(\ell)}\|_{L^2(\Omega)} = 0.$$

2 Numerical realization

2.1 Iterative solve

Theorem 1.1.c characterizes the discrete solution u_{CR} of (1.2) by $u_{\text{CR}} \in \text{CR}_0^1(\mathcal{T})$ and the existence of some $\Lambda_0 \in \text{sign } \nabla_{\text{NC}} u_{\text{CR}}$ with

$$a(u_{\text{CR}}, v_{\text{CR}})_{L^2(\Omega)} + (\Lambda_0, \nabla_{\text{NC}} v_{\text{CR}})_{L^2(\Omega)} = (f, v_{\text{CR}})_{L^2(\Omega)} \quad (2.1)$$

for all $v_{\text{CR}} \in \text{CR}_0^1(\mathcal{T})$. The pair $(u_{\text{CR}}, \Lambda_0)$ is unique in the first component u_{CR} and computed with the following algorithm with (u_0, Λ_0) as the input and $v_0 := 0$ in $\text{CR}_0^1(\mathcal{T})$.

Algorithm 2.1 (primal-dual iteration).

Input: $u_0 \in \text{CR}_0^1(\mathcal{T})$, $v_0 := 0$ in $\text{CR}_0^1(\mathcal{T})$, $\Lambda_0 \in P_0(\mathcal{T}; \overline{B(0, 1)})$, $\tau > 0$ for $j = 1, 2, \dots$

$$\tilde{u}_j := u_{j-1} + \tau v_{j-1} /$$

$$\Lambda_j := (\Lambda_{j-1} + \tau \nabla_{\text{NC}} \tilde{u}_j) / (\max\{1, |\Lambda_{j-1} + \tau \nabla_{\text{NC}} \tilde{u}_j|\}),$$

Solve linear system of equations

$$\frac{1}{\tau} a_{\text{NC}}(u_j, \bullet) + \alpha(u_j, \bullet)_{L^2(\Omega)} = \frac{1}{\tau} a_{\text{NC}}(u_{j-1}, \bullet) + (f, \bullet)_{L^2(\Omega)} - (\Lambda_j, \nabla_{\text{NC}} \bullet)_{L^2(\Omega)}$$

in $\text{CR}_0^1(\mathcal{T})$ for solution $u_j \in \text{CR}_0^1(\mathcal{T})$

$$v_j := \frac{1}{\tau} (u_j - u_{j-1}) / \tau$$

Output: sequence $(u_j, \Lambda_j)_{j \in \mathbb{N}}$ in $\text{CR}_0^1(\mathcal{T}) \times P_0(\mathcal{T}; \overline{B(0, 1)})$

2.2 Data structures

To realize Algorithm 2.1 numerically the data structures `c4n`, `n4e`, `s4e`, `n4s`, `n4sDb`, `n4sNb`, `s4n`, and `area4e` are recapitulated, see also [2] and section 7.8 of [3].

Let \mathcal{T} be a regular triangulation of the polygonal bounded Lipschitz domain

§ 2.2 Convergenz Analysis!
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I Give edge $E_j = \text{conv}\{P_\alpha, P_\beta\}$
in $T_1 \dots$

Ω into triangles with set of nodes \mathcal{N} and set of edges \mathcal{E} , and the set of inner nodes $\mathcal{N}(\Omega)$ and set of inner edges $\mathcal{E}(\Omega)$. The data structure of \mathcal{T} and \mathcal{E} is described as follows.

The j -th row $c4n(j,:) = [x \ y]$ of $c4n$ contains the coordinates of the j -th node $z_j = (x, y)$ of \mathcal{T} . $n4e(j,:) = [\alpha \ \beta \ \gamma]$ contains the three node numbers of the j -th triangle $T_j = \text{conv}\{P_\alpha, P_\beta, P_\gamma\}$ in counterclockwise order with the refinement edge $E_{\text{Ref}} = \text{conv}\{P_\alpha, P_\beta\}$ of T_j . $\mathcal{E}(T_j) = \{E_\alpha, E_\beta, E_\gamma\}$ (the j -th row $s4e(j,:) = [\alpha \ \beta \ \gamma]$ of $s4e$ contains the three side numbers of the j -th triangle T_j in counterclockwise order, where E_α is the refinement edge of T_j). $n4s(j,:) = [\alpha \ \beta]$ contains the two node numbers of the j -th edge $E_j = \text{conv}\{P_\alpha, P_\beta\}$. For a boundary edge $E_j = \text{conv}\{P_\alpha, P_\beta\} \in \mathcal{E} \setminus \mathcal{E}(\Omega)$ choose α and β such that the domain Ω is on the left-hand side of the vector from P_α to P_β . The following three lines of Matlab compute $n4s$.

```
allSides=[n4e(:, [1 2]); n4e(:, [2 3]); n4e(:, [3 1])];
[b, ind]=unique(sort(allSides, 2), 'rows', 'first');
n4s=allSides(sort(ind), :);
```

$n4sDb$ contains the numbers of the edges with Dirichlet boundary data and $n4sNb$ contains those of the edges with Neumann boundary data. With $n4s$ the computation of the data structure $s4n$ is possible which satisfies

$$s4n(\alpha, \beta) = s4n(\beta, \alpha) = \begin{cases} \ell & \text{if there exist } E_\ell = \text{conv}\{P_\alpha, P_\beta\}, \\ 0 & \text{else} \end{cases}$$

for any given nodes $\alpha, \beta \in \mathcal{N}$. $s4n = \text{sparse}(n4s(:, 1), n4s(:, 2), 1:S, N, N)$ creates a sparse matrix with

$$s4n(n4s(j, 1), n4s(j, 2)) = j \quad \text{for } j = 1, \dots, M := |\mathcal{E}|$$

and the last two parameters ensure that the resulting Matrix is $N \times N$ with $N = |\mathcal{N}|$.

```
M=size(n4s, 1); N=size(c4n, 1);
s4n=sparse(n4s(:, 1), n4s(:, 2), 1:M, N, N);
s4n=s4n + s4n';
```

The AFEM function `computeArea4e.m` computes the array `area4e` that contains the area of the j -th triangle of \mathcal{T} in its j -th row.

2.3 Basis of Crouzeix-Raviart functions [2]

Given an edge $E \in \mathcal{E}$ the associated edge-oriented basis function $\psi_E \in CR^1(\mathcal{T})$ is characterized by $\psi_E \in \mathcal{P}_1(T)$ and

$$\psi_E(\text{mid}(F)) = \delta_{EF} \text{ for all } F \in \mathcal{E}.$$

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Recall the (E_1, \dots, E_n) is an orthonormal basis of $E(u)$ described by the file \dots ? and abbreviated April 4, 2018

Then $(\psi_j)_{j=1, \dots, |\mathcal{E}|}$ is a basis of $CR^1(\Omega)$ and $(\psi_j|_{E_j} \in \mathcal{E}(\Omega))$ is a basis of $CR_0^1(\Omega)$. It is established that on a triangle $T \in \mathcal{T}$ with $E \in \mathcal{E}(T)$ the function ψ_E has the representation

$$\psi_E = 1 - 2\varphi_P \text{ in } T$$

using the barycentric coordinate φ_P with respect to the ~~corner~~ ^{vertex} $P \in \mathcal{N}(T)$ of T opposite to E .

2.4 Discrete gradients

To compute the non-conforming gradient of a Crouzeix-Raviart function, the gradients of the conforming basis (ϕ_1, ϕ_2, ϕ_3) functions on a triangle $T \in \mathcal{T}$ are ~~[(2), (3)]~~ **[1, 3]**

$$\begin{pmatrix} 1 & 1 & 1 \\ P_1 & P_2 & P_3 \end{pmatrix} \begin{pmatrix} \nabla \phi_1 \\ \nabla \phi_2 \\ \nabla \phi_3 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{pmatrix}.$$

Consequently

$$\begin{pmatrix} 1 & 1 & 1 \\ P_1 & P_2 & P_3 \end{pmatrix} \begin{pmatrix} \nabla \psi_1|_T \\ \nabla \psi_2|_T \\ \nabla \psi_3|_T \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 \\ P_1 & P_2 & P_3 \end{pmatrix} \begin{pmatrix} -2\nabla \phi_1 \\ -2\nabla \phi_2 \\ -2\nabla \phi_3 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ -2 & 0 \\ 0 & -2 \end{pmatrix}$$

computes the gradients of the CR basis functions. $c4n(n4e(j, :), :)$ $\in \mathbb{R}^{3 \times 2}$ are the coordinates of the nodes of a triangle $T = \text{conv}\{P_\alpha, P_\beta, P_\gamma\}$ with $n4e(\text{elem}, :) = [\alpha \ \beta \ \gamma]$, i.e.

$$c4n(n4e(j, :), :) = [P_\alpha^T; P_\beta^T; P_\gamma^T].$$

Hence for the triangle $T \in \mathcal{T}$ of number j ,

$$[\text{ones}(1, 3); c4n(n4e(j, :), :)] \setminus [\text{zeros}(1, 2); -2 * \text{eye}(2)]$$

are the gradients of the CR basis functions on T .

These can be used to calculate the gradient of $u_{CR} \in CR^1(T)$ on T since u_{CR} on T can be represented as

$$u_{CR}|_T = \sum_{E \in \mathcal{E}(T)} u_E \psi_E|_T.$$

Therefore

$$\nabla u_{CR}|_T = \sum_{E \in \mathcal{E}(T)} u_E \nabla \psi_E|_T.$$

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$u(s4e(j,:))$ determines the three coefficients $(u_E|E \in \mathcal{E})$ of u_{CR} on T . $s4e(j,:)$ contains the global edge numbers of T in counterclockwise order and the first global edge number of T in $s4e(j,:)$ is the number of the refinement edge E_{Ref} . This implies

$$u(s4e(j,1)) = u_{E_{Ref}}.$$

On the other hand

```
[ones(1,3); c4n(n4e(j,:),:)] \ [zeros(1,2); -2*eye(2)]
```

computes the three gradients of the basis functions on T in counterclockwise order and the first gradient with respect to the edge opposit to the first global node in $n4e(j,:)$, i.e. the first of the gradients is not with respect to the refinement edge of T .

$c4n(n4e(j,[3 \ 1 \ 2]),:)$ permutes the coordinates of the nodes such that the order of the gradients computed by

```
[ones(1,3); c4n(n4e(j,[3 1 2]),:)] \ [zeros(1,2); -2*eye(2)]
```

matches the order of the coefficients. Hence

```
du(j,:) = u(s4e(j,:))' * ...  
[ones(1,3); c4n(n4e(j,[3 1 2]),:)] \ [zeros(1,2); -2*eye(2)]
```

computes the gradient of u_{CR} on the j -th element T . The function `GradientNC.m` uses this to compute the gradients of u_{CR} on every triangle $T \in \mathcal{T}$. The inputs of the function are `c4n`, `n4e`, `s4e`, the number of elements `nrElems` in \mathcal{T} , and the edge-oriented components of u_{CR} in `u`. The output of the function is the array $du \in \mathbb{R}^{|\mathcal{T}| \times 2}$. It contains the gradient of u_{CR} on the j -th element in the j -th component.

```
function du = GradientNC(c4n,n4e,s4e,nrElems,u)  
    du = zeros(nrElems,2);  
    for elem = 1:nrElems  
        du(elem,:) = u(s4e(elem,:))' * ...  
            [ones(1,3);c4n(n4e(elem,[3 1 2]),:)] \ [zeros(1,2);-2*eye(2)];  
    end  
end
```

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2.5 Finite elemente matrices ~~([2], [3])~~

The entries of the non-conforming stiffness matrix are

$$A_{j,k} = a_{NC}(\psi_j, \psi_k) := \int_{\Omega} \nabla_{NC} \psi_j \cdot \nabla_{NC} \psi_k \, dx$$

and the entries of the ~~non-conforming~~ mass matrix are

$$M_{j,k} = (\psi_j, \psi_k)_{L^2(\Omega)} = \int_{\Omega} \psi_j \psi_k \, dx = |\omega_{E_j}| \delta_{j,k}$$

$$\omega_{E_j} := T_+ \cup T_- \dots$$

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The local non-conforming mass matrix on a triangle T reads

$$M(T) := \left(\int_T \psi_j \psi_k \, dx \right)_{j,k=1,2,3} = \frac{|T|}{3} I_{3 \times 3}$$

Furthermore the ~~non-conforming~~ ^{piecewise const} gradient ∇_{NC} of a CR basis function is constant on any triangle $T \in \mathcal{T}$, hence

$$\begin{aligned} A(T) &:= \left(\int_T \nabla \psi_j \cdot \nabla \psi_k \, dx \right)_{j,k=1,2,3} = \int_T 1 \, dx (\nabla \psi_j \cdot \nabla \psi_k)_{j,k=1,2,3} \\ &= |T| \begin{pmatrix} \nabla \psi_1 \\ \nabla \psi_2 \\ \nabla \psi_3 \end{pmatrix} \begin{pmatrix} \nabla \psi_1^T & \nabla \psi_2^T & \nabla \psi_3^T \end{pmatrix}. \end{aligned}$$

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Using this observations the function `FeMatrices.m` assembles the ~~non-conforming~~ mass matrix and non-conforming stiffness matrix. The inputs of the function are `c4n`, `n4e`, `s4e`, `area4e`, and the number of elements `nrElems` in \mathcal{T} . The outputs of the function are the non-conforming stiffness matrix A and the non-conforming mass matrix M as sparse matrices. First

```
Alocal = zeros(3,3,nrElems);
Mlocal = zeros(3,3,nrElems);
```

preallocates memory for the local non-conforming matrices. `Mlocal` and `Alocal` are both $3 \times 3 \times |\mathcal{T}|$ arrays. The following loop over all triangles uses this and the computation of the discrete gradients in section 2.4 to assemble the local non-conforming stiffness and mass matrices.

```
for elem = 1 : nrElems
    gradsNC = [ones(1,3); c4n(n4e(elem,:),[3 1 2]))'\ ...
               [zeros(1,2); -2*eye(2)];
    Alocal(:, :, elem) = area(elem) * (gradsNC * gradsNC');
    Mlocal(:, :, elem) = area(elem) * Mlocal(:, :, elem);
end
```

Then the global non-conforming stiffness and mass matrix can be assembled. `I` and `J` are used to identify local side numbers of the elements with the global side numbers of the triangulation by specifying the rows and columns the entries must have in the global matrices. Note that while assembling the sparse matrices, in case there already is an entry at a position in the matrix, the `sparse` function sums up the new number and the already existing one.

```
s4eT = s4e';
I = [s4eT; s4eT; s4eT];
J = [s4eT(:), s4eT(:), s4eT(:)]';
A = sparse(I(:), J(:), Alocal(:));
M = sparse(I(:), J(:), Mlocal(:));
```

The complete function reads

Knepper: klar erklären was A und M sind und wie dabei nummeriert wird!

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```
function [A,M] = FeMatrices(c4n,n4e,s4e,area4e,nrElems)
    Alocal = zeros(3,3,nrElems);
    Mlocal = zeros(3,3,nrElems);

    for elem = 1 : nrElems
        gradsNC = [ones(1,3); c4n(n4e(elem,[3 1 2]),:)]\ ...
            [zeros(1,2); -2*eye(2)];
        Alocal(:,:,elem) = area(elem) * (gradsNC * gradsNC');
        Mlocal(:,:,elem) = area(elem) * eye(3)/3;
    end

    s4eT = s4e';
    I = [s4eT;s4eT;s4eT];
    J = [s4eT(:),s4eT(:),s4eT(:)]';
    A = sparse(I(:),J(:),Alocal(:));
    M = sparse(I(:),J(:),Mlocal(:));
end
```

2.6 ~~Reformulation of Algorithm 2.1 for implementation~~

In ~~an~~ iteration step of Algorithm 2.1 with given Λ_{j-1}, u_{j-1} and $v_{j-1} := \frac{u_{j-1} - u_{j-2}}{\tau}$, the equation that is to be solved in $CR_0^1(\mathcal{T})$ is ~~has~~

$$\frac{1}{\tau} a_{NC}(u_j, \bullet) + \alpha(u_j, \bullet)_{L^2(\Omega)} = \frac{1}{\tau} a_{NC}(u_{j-1}, \bullet) + (f, \bullet)_{L^2(\Omega)} - (\Lambda_j, \nabla_{NC} \bullet)_{L^2(\Omega)} \dots$$

This leads to the ~~problem of finding a solution to~~

$$\left(\frac{1}{\tau} A + \alpha M \right) x = b \quad (2.2)$$

with $(b)_k$

where $b_k := \left(\frac{1}{\tau} \nabla_{NC} u_{j-1} - \Lambda_j, \nabla_{NC} \psi_k \right)_{L^2(\Omega)} + (f, \psi_k)_{L^2(\Omega)}$ ~~for~~ $k=1, \dots$

The left-hand side matrix can easily be calculated after the non-conforming stiffness and mass matrix are assembled with

$$A/\tau + \alpha M;$$

2.7 ~~Right-hand side b~~

To obtain b the function `Integrals.m` evaluates the integrals

$$\int_T f \psi_k \, dx \text{ for all } T \in \mathcal{T} \text{ and for all } k = 1, \dots, |\mathcal{E}|.$$

The inputs are a function handle of the function f in `f`, `c4n`, `n4e`, the algebraic degree of exactness the numerical quadrature will have, and `area4e`. The output is, for $j = 1, 2, 3$, an array `tempj` that contains the value of the approximation of $\int_{T_k} f \psi_j \, dx$ in the k -th component for $k = 1, \dots, |\mathcal{T}|$.

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To use the AFEM packages function `integrate.m` the integrands have to be defined and therefore function handles of the local CR basis functions are needed.

```
phi1 = @(x)( x(:,1) );
phi2 = @(x)( x(:,2) );
phi3 = @(x)( 1 - x(:,1) - x(:,2) );

psi1 = @(x)( .1-2*phi3(x) );
psi2 = @(x)( 1-2*phi1(x) );
psi3 = @(x)( 1-2*phi2(x) );
```

phi_j for $j = 1, 2, 3$ are function handles of the barycentric coordinates on the reference triangle $T_{\text{Ref}} = \text{conv}((0, 0), (1, 0), (0, 1))$. They are used to define the CR basis functions psi_j for $j = 1, 2, 3$ on T_{Ref} . The edge numeration convention from `s4e` is used and therefore the first edge is the refinement edge.

Then the integrands are defined where the CR basis functions will be evaluated on the reference triangle $\text{psi}_j(\text{Gpts4ref})$, $j = 1, 2, 3$, and transformed to a element $T \in \mathcal{T}$.

```
INT1 = @(n4p, Gpts4p, Gpts4ref) psi1(Gpts4ref).*f(Gpts4p);
INT2 = @(n4p, Gpts4p, Gpts4ref) psi2(Gpts4ref).*f(Gpts4p);
INT3 = @(n4p, Gpts4p, Gpts4ref) psi3(Gpts4ref).*f(Gpts4p);
```

The integrals are evaluated via `integrate.m` and `area4e` will be used for the transformations from the reference triangle to the elements of \mathcal{T} .

```
temp1 = integrate(INT1, c4n, n4e, degree, area4e);
temp2 = integrate(INT2, c4n, n4e, degree, area4e);
temp3 = integrate(INT3, c4n, n4e, degree, area4e);
```

The complete function reads

```
function [temp1,temp2,temp3] = computeIntegrals(f,c4n,n4e,degree,area4e)
phi1 = @(x)( x(:,1) );
phi2 = @(x)( x(:,2) );
phi3 = @(x)( 1 - x(:,1) - x(:,2) );

psi1 = @(x)( 1-2*phi3(x) );
psi2 = @(x)( 1-2*phi1(x) );
psi3 = @(x)( 1-2*phi2(x) );

INT1 = @(n4p, Gpts4p, Gpts4ref) psi1(Gpts4ref).*f(Gpts4p);
INT2 = @(n4p, Gpts4p, Gpts4ref) psi2(Gpts4ref).*f(Gpts4p);
INT3 = @(n4p, Gpts4p, Gpts4ref) psi3(Gpts4ref).*f(Gpts4p);

temp1 = integrate(INT1, c4n, n4e, degree, area4e);
temp2 = integrate(INT2, c4n, n4e, degree, area4e);
temp3 = integrate(INT3, c4n, n4e, degree, area4e);
end
```

The function `RightHandSide.m` uses temp_j , $j = 1, 2, 3$, to compute the right-hand side b with

$$b_k := \left(\frac{1}{\tau} \nabla_{\text{NC}} u_{j-1} - \Lambda_j, \nabla_{\text{NC}} \psi_k \right)_{L^2(\Omega)} + (f, \psi_k)_{L^2(\Omega)} \text{ for all } k = 1, \dots, |\mathcal{E}|.$$

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The inputs of the function are $c4n$, $n4e$, $s4e$, the number of sides $nrSides$ in \mathcal{T} , $area4e$, the discrete gradient du of u_{j-1} computed by the function `GradientNC` in section 2.4, the parameter $\tau = \tau$, $\Lambda_j = \Lambda_j$, the number of elements $nrElems$ in \mathcal{T} , and $tempj$, $j = 1, 2, 3$, the arrays with the value of the integral $\int_{T_k} f \psi_j dx$ in the k -th component for $k = 1, \dots, |\mathcal{T}|$. The outputs are the right-hand side $b = b$ and the vector $temp = (\int_{\Omega} f \psi_k dx)_{k=1, \dots, |\mathcal{E}|}$ that will be used to compute the discrete energy.

First memory for b and $temp$ is preallocated

```
b = zeros(nrSides,1);
temp = zeros(nrSides,1);
```

Das ist uns alle klar, also ist der Hinkgrad damit!

Given $elem$ the number of an arbitrary element $T \in \mathcal{T}$

```
gradsNC = [ones(1,3); c4n(n4e(elem,[3 1 2]),:)] \ ...
    [zeros(1,2); -2*eye(2)];
```

computes the gradients of the CR basis functions on T ~~as seen~~ in section 2.4. Furthermore for all $k = 1, \dots, |\mathcal{E}|$

$$\begin{aligned} \tilde{b}_k &:= \left(\frac{1}{\tau} \nabla_{NC} u_{j-1} - \Lambda_j, \nabla_{NC} \psi_k \right)_{L^2(\Omega)} \\ &= \sum_{T \in \mathcal{T}} \left(\frac{1}{\tau} \nabla u_{j-1}|_T - \Lambda_j, \nabla \psi_k|_T \right)_{L^2(T)} \\ &= \sum_{T \in \mathcal{T}} |T| \left(\frac{1}{\tau} \nabla u_{j-1}|_T - \Lambda_j \right) \cdot \nabla \psi_k|_T \\ &= \sum_{elem=1}^{|\mathcal{T}|} area4e(elem) (du(elem,:)/tau - Lambda(elem,:)) gradsNC'_k \end{aligned}$$

Was ist das denn?

and

$$temp = (f, \psi_k)_{L^2(\Omega)} = \sum_{T \in \mathcal{T}} (f, \psi_k)_{L^2(T)} \quad ? \quad ne?$$

hier die like la hier ein Vektor!

Note that $b = \tilde{b} + temp$. Therefore

```
for elem = 1 : nrElems
    gradsNC = [ones(1,3); c4n(n4e(elem,[3 1 2]),:)] \ ...
        [zeros(1,2); -2*eye(2)];
    bLocal = ( du(elem,:)/tau - Lambda(elem,:) ) * gradsNC';
    temp(s4e(elem,:)) = temp(s4e(elem,:)) + ...
        [temp1(elem),temp2(elem),temp3(elem)]';
    b(s4e(elem,:)) = b(s4e(elem,:)) + area4e(elem)*bLocal';
end
b = b + temp;
```

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computes \mathbf{b} in a loop over all triangles in \mathcal{T} .

The complete function reads

```
function [b,temp] = RightHandSide(c4n,n4e,s4e,nrSides,area4e,du,tau, ...
    Lambda,nrElems,temp1,temp2,temp3)
    b = zeros(nrSides,1);
    temp = zeros(nrSides,1);
    for elem = 1 : nrElems
        gradsNC = [ones(1,3); c4n(n4e(elem,[3 1 2]),:)] \ ...
            [zeros(1,2); -2*eye(2)];
        bLocal = ( du(elem,:)/tau - Lambda(elem,:) ) * gradsNC';
        temp(s4e(elem,:)) = temp(s4e(elem,:)) + ...
            [temp1(elem),temp2(elem),temp3(elem)]';
        b(s4e(elem,:)) = b(s4e(elem,:)) + area4e(elem)*bLocal';
    end
    b = b + temp;
end
```

2.8 Degrees of freedom

Since the solution u_j of the linear system of equations in algorithm 2.1 is a function in $\text{CR}_0^1(\mathcal{T})$ only the edge-oriented coefficients for non-Dirichlet edges, the degrees of freedom for this system, have to be computed. The function `DegreesOfFreedom.m` has the inputs $\mathbf{n4e}$, the number of sides $\mathbf{nrSides}$ of the triangulation \mathcal{T} , and $\mathbf{n4sDb}$ and computes the degrees of freedom \mathbf{dof} as output as a vector that contains the global numbers of the sides in \mathcal{E} that are not Dirichlet sides. First $\mathbf{s4n}$ is assembled to compute the Dirichlet boundary sides in $\mathbf{DbSides}$. For the j -th pair of nodes in $\mathbf{n4sDb}$

```
DbSides(j) = s4n(n4sDb(j,1),n4sDb(j,2));
```

computes the global number of the side that has this pair of nodes as end points. This number is saved in $\mathbf{DbSides}(j)$ as the number of the j -Dirichlet boundary side.

```
s4n = computeS4n(n4e);

DbSides = zeros(1,size(n4sDb,1));
for j = 1:size(n4sDb,1)
    DbSides(j) = s4n(n4sDb(j,1),n4sDb(j,2));
end
```

Now the line

```
dof = setdiff(1:nrSides,DbSides);
```

removes all numbers of Dirichlet boundary sides from a vector $1:\mathbf{nrSides} = [1 \ 2 \ 3 \ \dots \ \mathbf{nrSides}]$ and yields an array \mathbf{dof} that only contains the global numbers of the degrees of freedom. The complete code reads.

```
function dof = DegreesOfFreedom(n4e,nrSides,n4sDb)
    s4n = computeS4n(n4e);

    DbSides = zeros(1,size(n4sDb,1));
    for i = 1:size(n4sDb,1)
```

} bitte ins Hauptprogramm!

das muß doch noch vorne, oder?

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```

        DbSides(i) = s4n(n4sDb(i,1),n4sDb(i,2));
    end

    dof = setdiff(1:nrSides,DbSides);
end

```

~~3 Solving of 2.2~~ *← des ist eher trivial oder?*

With the left-hand side matrix $\frac{1}{\tau}A + \alpha M$, the right-hand side b and the degrees of freedom of the linear system 2.2 computed it is now solved for all non-Dirichlet boundary sides. The solution `uNew` contains the edge-oriented components of $u_j \in CR_0^1(\mathcal{T})$ and is initialized as

```
uNew = zeros(nrSides,1);
```

The system is only solved for the degrees of freedom `dof` and the zeros remain for the components with respect to Dirichlet boundary sides.

```
uNew(dof) = A(dof,dof)\b(dof);
```

4 Discrete energy

The function `Energy.m` computes the discrete energy

$$E_{\text{NC}}(\mathbf{V}) = \frac{\alpha}{2} \|\mathbf{V}\|_{L^2(\Omega)}^2 + \|\mathbf{V}\|_{1,1,\text{NC}} - \int_{\Omega} f \mathbf{V} dx. \quad \text{R } v = ?$$

The inputs are `area4e`, the edge-oriented components u of a Crouzeix-Raviart function $u_{\text{CR}} \in CR_0^1(\mathcal{T})$, the gradient du of u_{CR} , an array `tmp` ~~that contains~~ *:=* *←* the value of the integral $\int_{\Omega} f \psi_j dx$ in the j -th component for $j = 1, \dots, |\mathcal{E}|$, $\alpha = \alpha$, and the non-conforming mass matrix `MAMANC`. It holds for $u \in$

*bitt den Leser mit
demit langweilen!*

*Was denn nun?
tmp vektor?
 $\text{tmp}(j) = \int_{\Omega} f \psi_j dx$?*

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$CR^1(\mathcal{T})$

$$\begin{aligned}
 E_{NC}(u) &= \frac{\alpha}{2} \|u\|_{L^2(\Omega)}^2 + \|u\|_{1,1,NC} - \int_{\Omega} f u \, dx \\
 &= \frac{\alpha}{2} (u, u)_{L^2(\Omega)} + \int_{\Omega} |\nabla_{NC} u| \, dx - \sum_{j=1}^{|\mathcal{E}|} u(j) \int_{\Omega} f \psi_j \, dx \\
 &= \frac{\alpha}{2} \sum_{j,k=1}^{|\mathcal{E}|} u(j) (\psi_j, \psi_k)_{L^2(\Omega)} u(k) + \sum_{T \in \mathcal{T}} \int_T |\nabla u|_T \, dx - \sum_{j=1}^{|\mathcal{E}|} u(j) \text{temp}(j) \\
 &= \frac{\alpha}{2} u' * \text{MAMANC} * u + \sum_{j=1}^{|\mathcal{T}|} |T_j| |du(j, :)| - u' * \text{temp} \\
 &= \frac{\alpha}{2} u' * \text{MAMANC} * u + \text{area4e}' * \sqrt{\text{sum}(du.^2, 2)} - u' * \text{temp}
 \end{aligned}$$

Therefore the complete function reads:

```

function ENew = Energy(area4e,u,du,temp,alpha,MAMANC)
ENew = alpha/2 * u'*MAMANC*u + area4e'*sqrt(sum(du.^2,2)) - u'*temp;
end

```

5 Main function

Bessere Überschrift!

The function `tvRegPrimalDual.m` realizes Algorithm 2.1. Its inputs are `c4n`, `n4e`, `n4sDb`, the initial data $u = u_0$ and $\text{Lambda} = \Lambda_0$, $f = f$, $\alpha = \alpha$, and `epsStop` for the termination criterion of the iteration. The outputs are the solution of the iteration u , the value `corr` of the last iteration step that satisfies `corr < epsStop`, and arrays `corrVec` and `energyVec` that contain the values of `corrj` and the discrete energy $E_{NC}(u_j)$ in the j -th component.

First τ is chosen and the function computes the number `nrElems` of triangles in \mathcal{T} . Then the AFEM functions `computeArea4e.m` and `computeS4e.m` compute the data structures `area4e` and `s4e`. The function uses `s4e` to compute the number `nrSides` of edges of \mathcal{T} in \mathcal{E} .

```

tau = 1/2;

nrElems = size(n4e,1);
area4e = computeArea4e(c4n,n4e);
s4e = computeS4e(n4e);
nrSides = max(max(s4e));

```

The function `DegreesOfFreedom.m` computes `dof` that contains the global side numbers of non-Dirichlet sides, the function `FeMatrices.m` computes the non-conforming stiffness and mass matrices, and the function `computeIntegrals.m`

Abbruchkriterien bitte zu in gepaart
 bitte in ein Unterlegende!
 → bitte die Dok erklären
 das ist richtig so also
 divide Regel!

also z.B. $\propto \frac{1}{2} \| \nabla u \|_{L^2(\Omega)}^2$ → MAMANC
 bitte MAMANC ist diagonal!
 Ziemlich also nicht anders als Pythagoras?

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computes for $j = 1, 2, 3$ the array `tempj` that contains the value of the approximation of $\int_{T_k} f\psi_j dx$ in the k -th component for $k = 1, \dots, |\mathcal{T}|$.

```
dof = DegreesOfFreedom(n4e,nrSides,n4sDb);

[STIMANC,MAMANC] = FeMatrices(c4n,n4e,s4e,area4e,nrElems);
A = STIMANC/tau+alpha*MAMANC;

[temp1,temp2,temp3] = computeIntegrals(f,c4n,n4e,200,area4e);
```

Then the function `GradientNC.m` computes the gradients of u_0 on every triangle in \mathcal{T} , $v_0 = 0$ is initialized, the value `corr` is initialized, `corrVec` and `energyVec` to save the corresponding information during the iteration are initialized, and the energy E is initialized.

```
du = GradientNC(c4n,n4e,u);

v = zeros(nrSides,1);

corr = epsStop+1;
corrVec = [];
energyVec = [];

E = 1;
```

After the initializations are completed a while loop realizes the iteration described in Algorithm 2.1. The termination criterion is reached when

$$\text{epsStop} \geq |E_{\text{NC}}(u_j) - E_{\text{NC}}(u_{j-1})|$$

← birk update of and/or Kikim!

In the beginning of the loop the gradient of v_{j-1} and $M = \Lambda + \tau \nabla_{\text{NC}} + \tau^2 \nabla_{\text{NC}}$ are computed.

```
dv = GradientNC(c4n,n4e,v);
M = Lambda + tau*(du + tau*dv);
```

The line

```
Lambda = bsxfun(@rdivide,M,max(1,sqrt(sum(M.^2,2))));
```

of MATLAB code computes Λ_j by row-wise division of M by $\max\{1, |M_j|\}$. The function `RightHandSide.m` computes the right-hand side b and an array `temp` that contains the value of the integral $\int_{\Omega} f\psi_j dx$ in the j -th component for $j = 1, \dots, |\mathcal{E}|$.

```
[b,temp] = RightHandSide(c4n,n4e,s4e,nrSides,area4e,du, ...
    tau,Lambda,nrElems,temp1,temp2,temp3);
```

~~Now~~

```
uNew = zeros(nrSides,1);
uNew(dof) = A(dof,dof)\b(dof);
```

computes $u_j = u_{\text{New}}$,

```
v=(uNew-u)/tau;
```

computes $v_j = v$, and

in last - birk knapper!

Needs!

A

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```
u = uNew;  
  
du = GradientNC(c4n,n4e,u);  
ENew = Energy(area4e,u,du,temp,alpha,MAMANC);
```

computes the gradient du of u_j for the next iteration step and to compute the energy with the function `Energy.m`. With the now computed energy E

```
corr = abs(ENew-E);  
E = ENew;  
energyVec(end+1) = E;  
corrVec(end+1) = corr;
```

computes `corr` to check whether to terminate the iteration or not. Also the values E and `corr` are saved in `energyVec` and `corrVec` for later use.

The complete function reads

```
function [u,corr,corrVec,energyVec] = tvRegPrimalDual(c4n,n4e,n4sDb, ...  
    u,Lambda,f,alpha,epsStop)  
  
    tau = 1/2;  
  
    nrElems = size(n4e,1);  
    area4e = computeArea4e(c4n,n4e);  
    s4e = computeS4e(n4e);  
    nrSides = max(max(s4e));  
  
    dof = DegreesOfFreedom(n4e,nrSides,n4sDb);  
  
    [STIMANC,MAMANC] = FeMatrices(c4n,n4e,s4e,area4e,nrElems);  
    A = STIMANC/tau+alpha*MAMANC;  
    [temp1,temp2,temp3] = computeIntegrals(f,c4n,n4e,200,area4e);  
    du = GradientNC(c4n,n4e,u);  
    v = zeros(nrSides,1);  
  
    corr = epsStop+1;  
    corrVec = [];  
    energyVec = [];  
  
    E = 1;  
    while corr > epsStop  
        dv = GradientNC(c4n,n4e,v);  
        M = Lambda + tau*(du + tau*dv);  
        Lambda = bsxfun(@rdivide,M,max(1,sqrt(sum(M.^2,2))));  
  
        [b,temp] = RightHandSide(c4n,n4e,s4e,nrSides,area4e,du, ...  
            tau,Lambda,nrElems,temp1,temp2,temp3);  
  
        uNew = zeros(nrSides,1);  
        uNew(dof) = A(dof,dof)\b(dof);  
        v=(uNew-u)/tau;  
        u = uNew;  
        du = GradientNC(c4n,n4e,u);  
        ENew = Energy(area4e,u,du,temp,alpha,MAMANC);  
  
        corr = abs(ENew-E);
```

eben zwei das noch Input!

was ist das? free nodes?
keine Unknoten
bitte!

Was soll das werden?

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6 Numerical Experiments

in the RHS $f(x) := g(|x|)$ is

Consider f defined on the unit circle as a function of the radius r by

$$g(r) := \begin{cases} \alpha - 12(2 - 9r) & \text{for } 0 \leq r \leq \frac{1}{6}, \\ \alpha(1 + (6r - 1)^\beta) - \frac{1}{r} & \text{for } \frac{1}{6} \leq r \leq \frac{1}{3}, \\ 2\alpha + 6\pi \sin(\pi(6r - 2)) - \frac{1}{r} \cos(\pi(6r - 2)) & \text{for } \frac{1}{3} \leq r \leq \frac{1}{2}, \\ 2\alpha(\frac{5}{2} - 3r)^\beta + \frac{1}{3} & \text{for } \frac{1}{2} \leq r \leq \frac{5}{6}, \\ -3\pi \sin(\pi(6r - 5)) + \frac{1}{r} \frac{(1 + \cos(\pi(6r - 5)))}{2} & \text{for } \frac{5}{6} \leq r \leq 1, \end{cases}$$

let $\alpha = \beta = 1$ and the exact solution to (1.1) with right hand side f reads

$$u(r) = \begin{cases} 1 & \text{for } 0 \leq r \leq \frac{1}{6}, \\ 1 + (6r - 1)^\beta & \text{for } \frac{1}{6} \leq r \leq \frac{1}{3}, \\ 2 & \text{for } \frac{1}{3} \leq r \leq \frac{1}{2}, \\ 2(5/2 - 3r)^\beta & \text{for } \frac{1}{2} \leq r \leq \frac{5}{6}, \\ 0 & \text{for } \frac{5}{6} \leq r \leq 1. \end{cases}$$

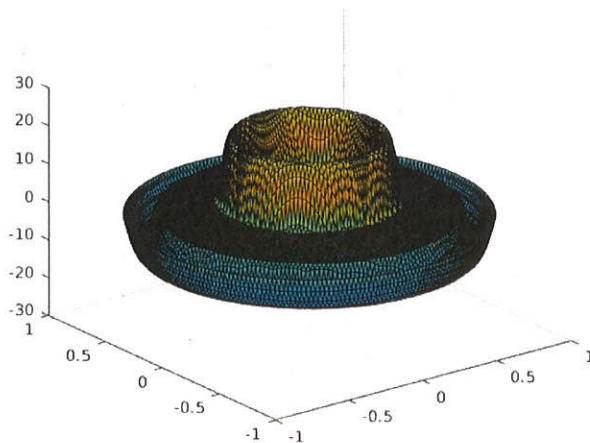


Figure 6.1: Plot of the right-hand side f with $\alpha = \beta = 1$.

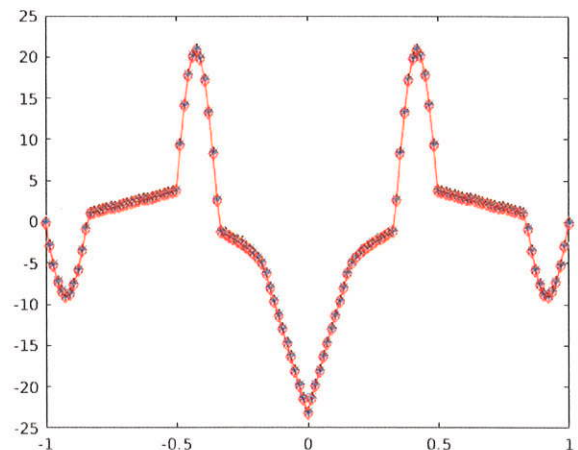


Figure 6.2: Plot of the right-hand side f along the axes with $\alpha = \beta = 1$.

Könnte z.B. denn nicht auch $\Omega = (-1, 1)^2$ nehmen?

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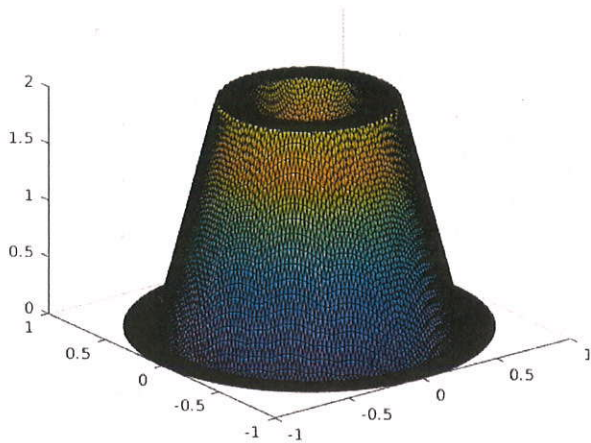


Figure 6.3: Plot of the exact solution u with $\alpha = \beta = 1$.

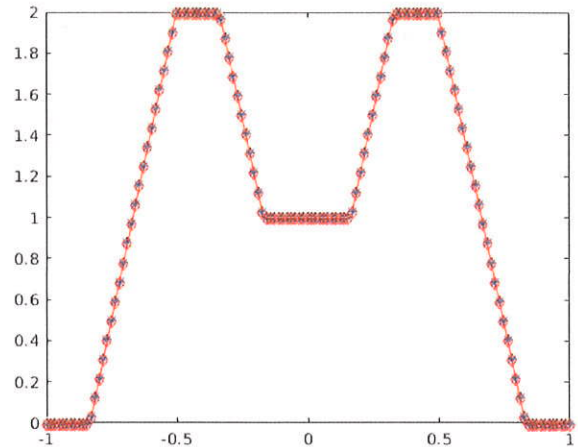


Figure 6.4: Plot of the exact solution u along the axes with $\alpha = \beta = 1$.

The discrete energy $E_{\text{NC}}(I_{\text{NC}}u)$ of the exact solution u on a triangulation that approximates the unit circle develops as in Table 1 with respect to the number of nodes.

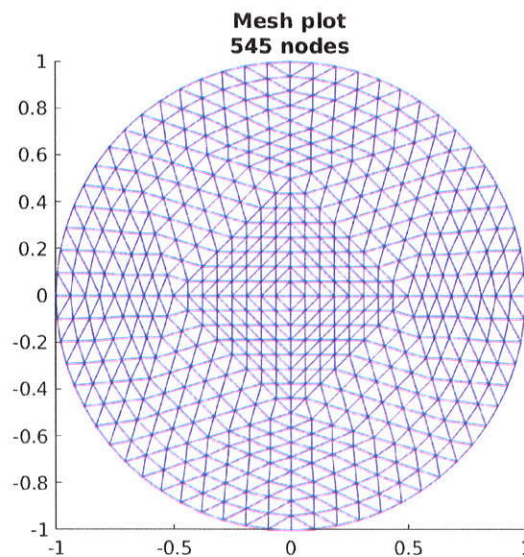


Figure 6.5: Example of a triangulation approximating the unit circle.

The values of $E_{\text{NC}}(I_{\text{NC}}u)$ are rounded. For the triangulation with 2099201 nodes the value of $E_{\text{NC}}(I_{\text{NC}}u)$ is approximately -2.05802391003896 . Therefore the discrete energy of the solutions of the discrete problem might converge from above to some real value close to $E_u := -2.05802391003896$.

gib das hier zur
Doch immer! ?

floating
point represents
of real
numbers!

sure as a reference
value.