Implementation and Performance Analyses of a Highly Efficient Algorithm for Pressure-Velocity Coupling

Implementierung und Untersuchung einer hoch effizienten Methode zur Druck-Geschwindigkeits-Kopplung

Master-Thesis von Fabian Gabel

Tag der Einreichung:

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Erklärung zur Master-Thesis

Hiermit versichere ich, die vorliegende Master-Thesis ohne Hilfe Dritter nur mit den angegebenen Quellen und Hilfsmitteln angefertigt zu haben. Alle Stellen, die aus Quellen entnommen wurden, sind als solche kenntlich gemacht. Diese Arbeit hat in gleicher oder ähnlicher Form noch keiner Prüfungsbehörde vorgelegen.

Darmstadt, den January 15, 2015

(F. Gabel)

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1 Introduction

This thesis is about.

- Cartesian Grid Components 3d
- Final Forms ideally integrals which are the starting point for finite volume methods

This section covers the set of fundamental equations for thermo-hydrodynamical problems which the numerical solution techniques of the following chapters are aiming to solve. Furthermore the notation regarding the physical quantities to be used throughout this thesis is introduced. The following paragraphs are based on (Kundu, Spurk, Ferziger, Anderson). For a thorough derivation of the matter to be presented the reader may consult the mentioned sources. Since the present thesis focusses on the application of finite-volume methods the focus lays on stating the integral forms of the relevant conservation laws. However in the process of deriving the final set of equations the use of differential formulations of the stated laws are required. Einstein's convention for taking sums over repeated indices is used to simplify certain expressions. For the remainder of this thesis non-moving inertial frames in a Cartesian coordinate system with the coordinates x_i are used. This approach is also known as *Eulerian approach*.

2.1 Conservation of Mass – Continuity Equation

The conservation law of mass embraces the physical concept that, neglecting relativistic and nuclear reactions, mass cannot be created or destroyed. Using the notion of a mathematical control volume, which is used to denote a constant domain of integration, one can state the integral mass balance of a control volume V with control surface S with surface normal unit vector $\mathbf{n} = (n_i)_{i=1,\dots,3}$ using Gauss' theorem as

$$\iiint_V \frac{\partial \rho}{\partial t} + \frac{\partial}{\partial x_i}(\rho u_i)\,\mathrm{d}V = \iiint_V \frac{\partial \rho}{\partial t}\mathrm{d}V + \iint_S \rho u_i n_i \mathrm{d}S = 0,$$

where ρ denotes the material density, t denotes the independent variable of time and $\mathbf{u} = (u_i)_{i=1,\dots,3}$ is the velocity vector field. Since this equation remains valid for arbitrary control volumes the equality has to hold for the integrands as well. In this sense the differential form of the conservation law of mass can be formulated as

$$\frac{\partial \rho}{\partial t} + \frac{\partial}{\partial x_i}(\rho u_i) = 0. \tag{1}$$

2.2 Conservation of Momentum – Cauchy-Equations

The conservation law of momentum, also known as Newton's Second Law, axiomatically demands the balance of the temporal change of momentum and the sum of all attacking forces on a body. Those forces can be divided into body forces and surface forces. Let $\mathbf{k} = (k_i)_{i=1,\dots,3}$ denote a mass specific force and $\mathbf{t} = (t_i)_{i=1,\dots,3}$ the stress vector. A first form of the integral momentum balance in the direction of x_i can be formulated as

$$\iiint_V \frac{\partial(\rho u_i)}{\partial t}\mathrm{d}V + \iint_S \rho u_i \left(u_j n_j\right)\mathrm{d}S = \iiint_V \rho k_i \mathrm{d}V + \iint_S t_i \mathrm{d}S.$$

In general the stress vector $\mathbf{t}$ is a function not only of the location $\mathbf{x} = (x_i)_{i=1,\dots,3}$ and of the time t but also of the surface normal unit vector $\mathbf{n}$. A central simplification can be introduced, namely Cauchy's stress theorem, which states that the stress vector is the image of the normal vector under a linear mapping $\mathbf{T}$. With respect to the Cartesian canonical basis $(\mathbf{e}_i)_{i=1,\dots,3}$ the mapping $\mathbf{T}$ is represented by the coefficient matrix $\left(\tau_{ji}\right)_{i,j=1,\dots,3}$ and Cauchy's stress theorem reads

$$\mathbf{t}(x,t,n) = \mathbf{T}(x,t,n) = \left(\tau_{ji} n_j\right)_{i=1,\dots,3}.$$

Assuming the validity of Cauchy's stress theorem one can derive Cauchy's first law of motion, which in differential form can be formulated as

$$\frac{\partial(\rho u_i)}{\partial t} + \frac{\partial}{\partial x_j}\left(\rho u_i u_j\right) = \rho k_i + \frac{\partial \tau_{ji}}{\partial x_j} \tag{2}$$

and represents the starting point for the modelling of fluid mechanical problems. One should note, that Cauchy's first law of motion does not take any assumptions regarding material properties, which is why the set of equations (1,2) is not closed in the sense that there exists a independent equation for each of the dependent variables.

2.3 Closing the System of Equations – Newtonian Fluids

As result of Cauchy's theorem the stress vector $\mathbf{t}$ can be specified once the nine components τ_{ji} of the coefficient matrix are known. As is shown in (Spurk usw.) by formulating the conservation law of angular momentum the coefficient matrix is symmetric,

$$\tau_{ji} = \tau_{ij}, \tag{3}$$

hence the number of unknown coefficients may be reduced to six unknown components. In a first step it is assumed that the coefficient matrix can be decomposed into fluid-static and fluid-dynamic contributions,

$$\tau_{ij} = -p\delta_{ij} + \sigma_{ij},$$

where p is the thermodynamic pressure, δ_{ij} is the *Kronecker*-Delta and σ_{ij} is the so called *deviatoric stress tensor*. For the fluids the studies that the present thesis performs it is sufficient to consider viscous fluids for which there exists a linear relation between the components of the deviatoric stress tensor and the symmetric part of the transpose of the jacobian of the velocity field $\left(S_{ij}\right)_{i,j=1,\dots,3}$,

$$S_{ij} = \frac{1}{2}\left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i}\right).$$

If one now imposes material-isotropy and the mentioned stress-symmetry (3) restriction it can be shown (Aries) that the constitutive equation for the deviatoric stress tensor reads

$$\sigma_{ij} = 2\mu S_{ij} + \lambda S_{mm}\delta_{ij},$$

where λ and μ denominate scalars which depend on the local thermodynamical state. Taking everything into account (2) can be formulated as the differential conservation law of momentum for newtonian fluids, better known as the *Navier-Stokes equations* in differential form:

$$\frac{\partial\left(\rho u_i\right)}{\partial t} + \frac{\partial}{\partial x_j}\left(\rho u_i u_j\right) = \rho k_i - \frac{\partial p}{\partial x_i} + \frac{\partial}{\partial x_j}\left(\mu\left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i}\right)\right) + \frac{\partial}{\partial x_i}\left(\lambda\frac{\partial u_m}{\partial x_m}\right) \tag{4}$$

2.4 Conservation Law for Scalar Quantities

The modelling of the transport of scalar quantities, convection, by a flow field $\mathbf{u}$ is necessary if the fluid mechanical problem to be analyzed includes for example heat transfer. Other scenarios that involve the necessity to model scalar transport surge, when turbulent flows are to be modeled by two-equation models like the k-ε-model (Pope).

Since this thesis focusses on the transport of the scalar temperature T this section introduces the conservation law for energy in differential form,

$$\frac{\partial\left(\rho T\right)}{\partial t} + \frac{\partial}{\partial x_j}\left(\rho u_j - \kappa\frac{\partial T}{\partial x_j}\right) = q_T, \tag{5}$$

where κ denotes the thermal conductivity of the modelled material and q_T is a scalar field representing sources and sinks of heat throughout the domain of the problem.

2.5 Necessary Simplification of Equations

Negligible viscous dissipation and and pressure work source terms in the enery equation (vakilipour)

The purpose of this section is to motivate and introduce further common simplifications of the previously presented set of constitutive equations.

2.5.1 Incompressible Flows and Hydrostatic Pressure

A common simplification when modelling low Mach number flows ($Ma < 0.3$), is the assumption of *incompressibility*, or the assumption of an *isochoric* flow. If one furthermore assumes homogeneous density ρ in space and time, a restrictive assumption that will be partially alleviated in the following section the continuity equation in differential form (1) can be simplified to

$$\frac{\partial u_i}{\partial x_i} = 0. \tag{6}$$

In other words: In order for a velocity vector field $\mathbf{u}$ to be valid for an incompressible flow it has to be free of divergence, or *solenoidal* (Aries).

If furthermore, one assumes also constant dynamic viscosity μ, which can be suitable in the case of isothermal flow or if the temperature differences within the flow are small, the Navier-Stokes equations in differential form can be reduced to

$$\frac{\partial (\rho u_i)}{\partial t} + \rho \frac{\partial}{\partial x_j}\left(u_i u_j\right) = \rho k_i - \frac{\partial p}{\partial x_i} + \frac{\partial}{\partial x_j}\left(\mu\left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i}\right)\right) \tag{7}$$

$$= \rho k_i - \frac{\partial p}{\partial x_i} + \mu \frac{\partial}{\partial x_j}\left(\frac{\partial u_i}{\partial x_j}\right) \tag{8}$$

by using *Schwartz*'s lemma to interchange the order of differentiation. A common simplification to further simplify the set of equations is the assumption of a volume specific force $\rho\mathbf{k}$ that can be modelled by a potential, such that is can be represented as the gradient of a scalar field $\Phi_\mathbf{k}$ as

$$-\rho k_i = \frac{\partial \Phi_\mathbf{k}}{\partial x_i}.$$

In the case of this thesis this assumption is valid since the mass specific force is the mass specific gravitational force $\mathbf{g} = (g_i)_{i=1,\dots,3}$ and the density is assumed to be constant, so the potential can be modelled as

$$\Phi_g = -\rho g_j x_j.$$

This term can be interpreted as the hydrostatic pressure p_{hyd} and can be added to the thermodynamical pressure p to simplify calculations.

$$\begin{aligned} \rho g_i - \frac{\partial p}{\partial x_i} &= \frac{\partial}{\partial x_i}\left(\rho g_j x_j\right) - \frac{\partial p}{\partial x_i} \\ &= \frac{\partial}{\partial x_i}\left(\rho g_j x_j\right) - \frac{\partial}{\partial x_i}\left(\hat{p} + p_{hyd}\right) \\ &= -\frac{\partial \hat{p}}{\partial x_i} \end{aligned} \tag{9}$$

Since in incompressible fluids only pressure differences matter, this has no effect on the solution. After finishing the calculations p_{hyd} can be calculated and added to the resulting pressure $\hat{p}$.

2.5.2 Variation of Fluid Properties – The Boussinesq Approximation

If modelling of an incompressible flow involves heat transfer fluid properties like the density change with varying temperature. If the variation of temperature is small one can still assume a constant density to maintain the structure of the advection and diffusion terms in (4) and only consider the changes of the density in the gravitational term. If linear variation of density with respect to temperature is assumed this approximation is called *Boussinesq*-approximation. In this case the Navier-Stokes equations are formulated using a reference pressure ρ_0 at the reference temperature T_0 and the now temperature dependent density ρ, with

$$\rho(T) = \rho_0 (1 - \beta (T - T_0)). \tag{10}$$

Here β denotes the coefficient of thermal expansion. Under the use of the Boussinesq-approximation the incompressible Navier-Stokes equations in differential form can be formulated as

$$\begin{aligned}\rho_0 \frac{\partial (u_i)}{\partial t} + \rho_0 \frac{\partial}{\partial x_j}\left(u_i u_j\right) &= \rho_0 g_i + (\rho - \rho_0) g_i - \frac{\partial p}{\partial x_i} + \mu \frac{\partial}{\partial x_j}\left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i}\right) \\ &= \frac{\partial}{x_i}\left(\rho_0 g_j x_j\right) + (\rho - \rho_0) g_i - \frac{\partial p}{\partial x_i} + \mu \frac{\partial}{\partial x_j}\left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i}\right) \\ &= -\frac{\partial \hat{p}}{\partial x_i} + (\rho - \rho_0) g_i + \mu \frac{\partial}{\partial x_j}\left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i}\right) \\ &= -\frac{\partial \hat{p}}{\partial x_i} - \rho_0 \beta (T - T_0) + \mu \frac{\partial}{\partial x_j}\left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i}\right)\end{aligned} \tag{11}$$

using $\rho \mathbf{g}$ as the mass specific force.

Talk about natural and forced convection. Differences for the solver algorithm. (s.a.) Peric P447 Talk about flows with variation in fluid properties -> mms has to map this behaviour (Buoyancy force driven, i.e. naturally convected fluid), mixed Convection Also talk about non-dimensional values like Prandtl number, Rayleigh and Reynolds Talk about the validity of this approxiation

2.6 Final Form of the Set of Equations

In the previous subsections different simplifications have been introduced which will be used throughout the thesis. The final form of the set of equations to be used is thereby presented. As further simplification the modified pressure $\hat{p}$ will be treated as p and since the use of the Boussinesq-approximation substitutes the variable ρ by a linear function of the temperature T the reference pressure ρ_0 for the remainder of this thesis will be referred to as ρ. Note that incompressibility has been taken into account:

$$\frac{\partial u_i}{\partial x_i} = 0. \tag{12}$$

$$\rho \frac{\partial (u_i)}{\partial t} + \rho \frac{\partial}{\partial x_j}\left(u_i u_j\right) = -\frac{\partial p}{\partial x_i} - \rho \beta (T - T_0) + \mu \frac{\partial}{\partial x_j}\left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i}\right) \tag{13}$$

$$\frac{\partial (\rho T)}{\partial t} + \frac{\partial}{\partial x_j}\left(\rho u_j - \kappa \frac{\partial T}{\partial x_j}\right) = q_T. \tag{14}$$

3 Finite Volume Method for Incompressible Flows – Theoretical Basics

This section deals with the fundamentals of the numerical solution via a finite volume method of the formerly presented set of partial differential equations. The focus of this section is, to provide an overview over the methods to be used in the present thesis. The information contained in this section is based on (Peric,Schäfer,Muzaferja,Jsak). The overview starts by mentioning the different grid types to be used and the discretization techniques to be applied. On the basis of integral formulations of the equations to be solved, the therein contained integrals and differential operators have to be discretized. Since the accuracy of the default concepts for discretizing differential operators degrades with decreasing grid quality, this chapter furthermore presents different approaches to handle corrections for cases in which the cause of degrading grid quality is increased non-orthogonality.

The goal of the finite volume method is to provide algebraic equations which can be used to determine an approximate solution of a partial differential equation. This system of linear algebraic equations can be solved by means of algorithms to be presented in the end of this section. However since the Navier-Stokes equations are in general non-linear an intermediate step has to be taken, by linearizing the discrete equations. This leads to the need of an iteration process, the *Picard iteration*, which will be explained briefly.

3.1 Numerical Grid

In this subsection a brief overview of the general grid structure to be used in the present thesis is given. The main idea behind finite volume methods is to solve partial differential equations by integrating them over the specified continuous problem domain and dividing this domain into a finite number of subdomains, the so called control volumes. The result of the this finite partition of a continuous problem domain is called the numerical grid. The grid consists of a finite number of grid cells which represent the boundaries of a discrete domain of integration.

Regarding the treatment of domain boundaries and the ordering of the cells within the problem domain different types of numerical grids can be distinguished. The present thesis makes use of so called block structured grids with hexahedron cells. A structured grid is characterized by a constant amount of of grid cells in each coordinate direction. The high regularity of structured grids benefits the computational efficiency of algorithms to be used on this type of grid. A block structured grid consists of different grid blocks. of which each considered individually is structured but if the topology of the grid is considered it is unstructured. An example of a block structured grid with distinguishable grid blocks is given in figure 3.3.3. The use of block structured grids is motivated by the need to increase the adaptivity of structured grids by maintaining high computational efficiency.

Inside a structured grid block, cells with the shape of hexahedrons are used. In addition to the geometric boundaries of each control volume a numerical grid also provides a mapping that assigns to each control volume with index P a set of indexes of neighbouring control volumes $NB(P) := \{W, S, B, T, N, E\}$, which are named after the geographic directions. Figure 3.3.3 shows a single grid cell with its direct neighbours. The faces $\{S_w, S_s, S, b, S_t, S_n, S_e\}$ of each hexahedral control volume represent the mentioned geometric boundaries.

- talk about grid quality
- talk about local refinement
- talk about variable arrangement

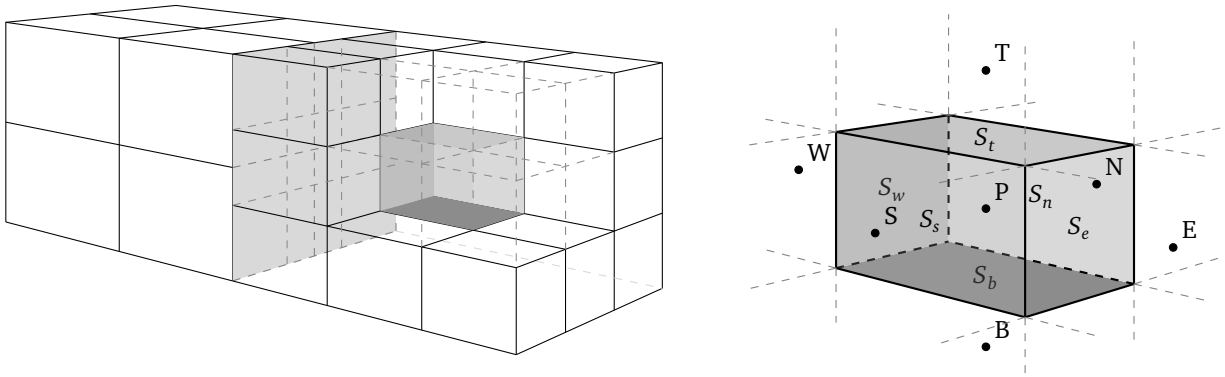

Figure 1: Blockstructured grid consisting of two blocks

3.2 Approximation of Integrals and Derivatives

In the course of transforming a partial differential equation into a system of linear algebraic equations, integrals and derivatives have to be approximated. The simplest method for approximating an integral is by using the *midipoint rule*. This rule is similar to the mean value theorem of integration, which states that there exists a point $\xi \in V$ for a riemann integrable function ϕ such that $\phi(\xi)\int_V \mathrm{d}V = \int_V \phi(x)\mathrm{d}V$. For the midpoint rule ξ is taken to be the center of mass of V. If the integration domain V is indeed a Volume, fortunately the calculation of $\phi(\xi)$ with $\xi := \left(\int_V x_i \mathrm{d}V / \int_V \mathrm{d}V\right)_{i=1,\dots,3}$ presents no difficulties since due to the collocated variable arrangement the value of ϕ is stored in the cell center, which corresponds to the location ξ. However if the domain of integration is a surface, a preceding interpolation step is necessary.

On the other hand to transorm a partial differential equation into a linear algebraic equation it is necessary to discretize the differential operators of the equations. For numerical reasons two different discretization techniques are used in this thesis. A common task is to discretize expressions of the form

$$(\nabla\phi)_e \cdot \mathbf{n}_e,$$

where $(\nabla\phi)_e$ is the Gradient of ϕ on a boundary face S_e. One method is to directly interpret this expression as a directional derivative and approximate it with a central difference

$$(\nabla\phi)_e \cdot \mathbf{n}_e \approx \frac{\phi_P - \phi_E}{||\mathbf{x}_P - \mathbf{x}_E||_2}. \tag{15}$$

Another method would be to first calculate the cell center gradients $(\nabla\phi)_P$ and $(\nabla\phi)_E$ and interpolate them linearily before calculating the projection onto $\mathbf{n}_e$

$$(\nabla\phi)_e \cdot \mathbf{n}_e \approx \left[\gamma_e(\nabla\phi)_P + (1-\gamma_e)(\nabla\phi)_E\right] \cdot \mathbf{n}_e,$$

where $\gamma_e := ||\mathbf{x}_P - \mathbf{x}_e||_2 / ||\mathbf{x}_P - \mathbf{x}_E||_2$ is a geometric interpolation factor. For calculating the cell center gradients a method based on Gauss' integration theorem and the midpoint rule for volume integration is employed.

$$(\nabla\phi)_{i,P} = \left(\frac{\partial\phi}{\partial x_i}\right)_P \approx \frac{\int_V \left(\frac{\partial\phi}{\partial x_i}\right)_P \mathrm{d}V}{|V|} \tag{16}$$

Briefly explain the idea behing the quadrature via the midpoint rule. Talk about central differences and the approximation via the gauss theorem. Maybe talk about the resulting order of the truncation error.

3.3 Treatment of Non-Orthogonality of Grid Cells – Deferred Correction Approach

Unfortunately real applications involve complex geometries which in turn affects the orthogonality of the grid. On non-orthogonal meshes the directional derivative in direction of the face normal unit vector $\mathbf{n}_e$ can no longer be approximated as in (15). On the other side the exclusive usage of (16) is not desirable due to the bigger truncation error that comes with this approximation (PROOF?). Hence a compromise is made and the surface vector $\mathbf{S}_e := S_e\mathbf{n}_e$ is decomposed as

$$\mathbf{S}_e = \Delta + \mathbf{k}, \tag{17}$$

where Δ is parallel to the the vector $\mathbf{d}_e := (\mathbf{x}_E - \mathbf{x}_P)$ that directly connects the center of the control volume P with the center of its neighbour E. This vector controls the *orthogonal* contribution to the directional derivative. The vector $\mathbf{k}$ controls the influence of the *non-orthogonal* contribution. In the next paragraphs the three main decompositions of the surface vector $\mathbf{S}_e$ will be presented. However in each method the non-orthogonal contribution vanishes as expected, when an orthogonal grid is used. For simplicity the presentation of the decompositions is chosen to be two dimensional. An geometrical interpretation of the tree approaches is given in FIGURE.

3.3.1 Minimum Correction Approach

This is the approach as used in proposed in Muzaferja The reader should note, that even though Ferziger/Peric reference the work of Muzaferja they use a different approach to be presented in the next paragraph. This method is designed to keep the non-orthogonal contribution minimal by chosing $\mathbf{k}$ to be orthogonal to Δ, which leads to

$$\Delta = (\mathbf{d}\cdot\mathbf{S}_e)\frac{\mathbf{d}}{||\mathbf{d}||_2}.$$

The Influence of the non-orthogonal contribution increases with increasing non-orthogonality of the grid.

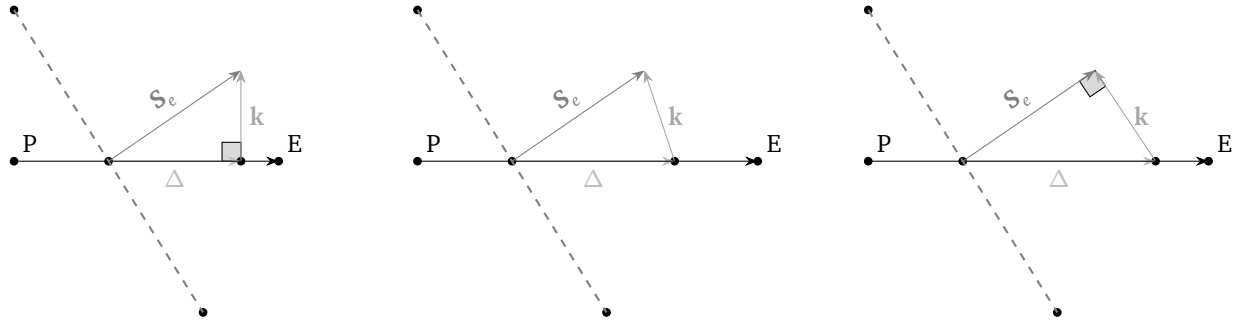

Figure 2: Blockstructured grid consisting of two blocks

3.3.2 Orthogonal Correction Approach

The following method for decomposing the surface normal vector is presented in Ferziger/Peric and the approach implemented in the developed solvers. In this approach a simple projection is used which is independent of the non-orthogonality of the grid. The orthogonal contribution thus is modelled as

$$\Delta = S_e \frac{\mathbf{d}}{||\mathbf{d}||_2}.$$

3.3.3 Over-Relaxed Approach

The last approach is used in Jsak and Darwish and is characterized by an increasing influence of the non-orthogonal contribution with decreasing grid orthogonality, as opposed to the minimum correction approach. The orthogonal contribution is calculated as

$$\Delta = S_e^2 \frac{\mathbf{d}}{\mathbf{d} \cdot \mathbf{S}_e}.$$

- note that each method should vanish on orthogonal grids
- Stick to Jsak and refer to Muzaferja's PHD thesis and Ferziger/Peric.
- Note the discrepancy in Perics book's referral to Muzaferja's thesis.
- Talk about the benefits of each approximation, maybe with example calculation
- Talk about the overall convergence rates
- note that this only accelerates the solver. The fully converged solution should be independent of the treatment of the non-orthogonalities. (Proof?)

3.4 Numerical Solution of Non-Linear Systems

Introduce the concept of the Picard-Iteration as a linearization technique. Introduce the notions of inner and outer iterations. Refer to later chapters when it comes to deferred correction.

3.5 Numerical Solution of Linear Systems

The discretization techniques to be presented in the following chapter in composition with a Picard iteration transform the approximate solution of a system of partial differential equations into the solution of algebraic equations via linear systems.

3.5.1 Stone's SIP Solver

Basic Idea as in Schäfer or Peric. Emphasize that this is a very problem specific approach, that cannot be generalized that easily in opposition to the general purpose linear solvers from PETSc. Present ether BiCGStab or GMRES, the one which performs better and is used throughout the thesis.

3.5.2 Krylov Subspace Methods

- General concept of cyclic vector spaces of $\mathbb{R}^n$,
- talk about bases of krylov subspaces and the arnoldi algorithm, talk about polynomials and linear combinations
- mention the two major branches (minimum residual approach, petrov and ritz-galerkin approach)
- name some representative ksp algorithms, importance of preconditioning, not as detailed as in bachelor thesis
- in cases there is a nonempty Nullspace what happens?

to be solved Since this thesis focusses on the analysis of a fully coupled numerical solution algorithm, special emphasis is put on the comparison between segregated and coupled solution methods for solving the set of partial differential equations. Concretely this is achieved by presenting the discretization technique used for the application of the SIMPLE algorithm and later on modifying those results to be applicable to use in a fully coupled solution algorithm. Since on top of the Navier-Stokes equations a scalar transport equation for the temperature is solved, different methods to realize velocity-temperature-coupling and vice versa are discussed.

4.1 Discretization of the Mass Balance

4.2 Discretization of the Momentum Balance

4.2.1 Semi Discretized Linearized Form of the Navier-Stokes Equations

4.2.2 Calculation of Mass Flux – Rhie-Chow Interpolation

4.2.3 Discretization of the Convective Term

4.2.4 Discretization of the Diffusive Term

4.2.5 Discretization of the Source Term

4.2.6 Assembly of Linear Systems – Final Form of Equations

Coefficients of matrices for momentum are identical except in case of different factors for under-relaxation (underrelaxation (Andersson))(when does this happen) for the main diagonal coefficient. Small example in code, then show image of assembled system.

4.3 Discretization of the Generic Transport Equation

4.4 The SIMPLE-Algorithm

4.4.1 Pressure Correction Equation

4.4.2 Characteristic Properties of Projection Methods

Under-relaxation, slow convergence, inner iterations outer iterations, relative tolerances, also talk about staggered and collocated variable positioning

4.4.3 Dependence on Under-Relaxation – The Pressure-Weighted Interpolation Method

Present an approach for Under-Relaxation independent converged solution. Conduct the proof to show it really works. Present the results for different under-relaxation factors

4.4.4 Coupling of Temperature Equation

Explicit coupling through source term in momentum balances (Boussinesq-Approximation)

4.5 Boundary Conditions on Domain and Block Boundaries

Introduce chapter by talking about the nature of partial differential equations (Hackbusch). Always start with a simple implementation for the generic transport equation, then specialize to Navier-Stokes equation.

4.5.1 Dirichlet Boundary Condition

Only talk about dirichlet for velocities not for pressure.

4.5.2 Neumann Boundary Condition

Problematics of outlet boundary conditions

4.5.3 Symmetry Boundary Condition

4.5.4 Wall Boundary Condition

Note that there are different approaches. Explain which approach is used and why (memory efficiency)

4.5.5 Block Boundary Condition

5 Finite Volume Method for Incompressible Flows – Coupled Approach

5.1 The Coupled Algorithm

5.1.1 Pressure Equation

5.1.2 Characteristic Properties of Coupled Solution Methods

No Underrelaxation needed, higher memory requirements

Bad condition, singularity, usually faster convergence if efficient linear solver is chosen, coupling in Buoyancy flows (s.a. Peric page 448, Galpin Raithby) Design of algorithm does not need to inforce continuity (is inherently fullfilled because of the coupling of the equations)

Explicitely mention the differences

- Implicit treatment of Pressure Gradient
- Implicit Treatment of Temperature possible
- Boussinesq approximation brings velocity-to-temperature-coupling (vakilipour), Newton-Raphson Linearization
- Temperature dependent densities also possible

5.2 Coupling the Temperature Equation

5.2.1 Decoupled Approach

5.2.2 Velocity-Temperature Coupling

5.2.3 Temperature-Velocity/Pressure Coupling – Newton-Raphson Linearization

5.3 Boundary Conditions on Domain and Block Boundaries

5.3.1 Dirichlet Boundary Condition for Velocity

5.3.2 Dirichlet Boundary Condition for Pressure

5.3.3 Symmetry Boundary Condition

5.3.4 Wall Boundary Condition

5.3.5 Block Boundary Condition

5.4 Assembly of Linear Systems – Final Form of Equations

6 CAFFA Framework

6.1 PETSc Framework

Keep in mind not to copy the manual

6.1.1 About PETSc

Bell Prize, MPI Programming

6.1.2 Basic Data Types

Vec,Mat (Different Matrix Types and Their effect on complex methods)

6.1.3 KSP and PC Objects and Their Usage

Singularities

6.1.4 Profiling

PETSc Log

6.1.5 Common Errors

Optimization, Interfaces, (ROWMAJOR,COLUMNMAJOR), Compiler Errors not helpful, Preallocation vs. Mallocs

6.2 Grid Generation and Conversion

Generation of block structured locally refined grids with non-matching block interfaces, neighbouring relations are represented by a special type of boundary conditions; Random number generator to move grid points within a epsilon neighbourhood while maintaining the grid intact. Show in a graph how preallocation impacts on runtime.

6.3 Preprocessing

Matching algorithm – the idea behind clipper and the used projection technique; alt.: Opencascade. Efficient calculation of values for discretization. Important for dynamic mesh refinement, arbitrary polygon matching, parallelizable due to easier interface

6.4 Implementation Details of CAFFA

6.4.1 MPI Programming Model

Basic idea of distributed memory programming model, emphasize the differences to shared memory model. Have a diagram at hand that shows how CAFFA sequentially works (schedule) and point out the locations where and of which type (global reduce, etc.) communication is, or when synchronization is necessary.

- after each solve
- pressure reference
- error calculation
- gradient calculation

Point out that one should try to minimize the number of this points such that parallel performance stays high. Better to calculate Velocity and Pressure Gradients at once not by seperately calling this routine.

6.4.2 Convergence Control

Explain how the criterion for convergence is met

6.4.3 Modi of Calculation

there are different modi of calculation, (NS segregated, then scalar; NS and Scalar Segregated; NS coupled and Scalar segregated; Fully coupled (wath out with fully coupled, this term seems to have already another meaning)). Note that for comparison of solvers it is crucial to develop programs on the same basis. This establishes comparability.

6.4.4 Indexing of Variables and Treatment of Boundary Values

Describe MatZeroValues and how it is used to simplify the code. Also loose a word on PCREDISTRIBUTE its advantages and downsides. Compliance of PETSc zero based indexing and CAFFA indexing which considers boundary values. Problems with boundary entries

6.4.5 Field Interlacing

Realization through special arrangement of variables and the use of index sets (subvector objects) and/or preprocessor directives. Advantages (there was a paper I cited in my thesis). Note that not all variables are interlaced (Velocities are, but their gradients are not). Great impact on Matrix structure.

6.4.6 Domain Decomposition, Exchange of Ghost Values and Parallel Matrix Assembly

- Ghost values are stored in local representations of the global vector (state the mapping for those entries).
- Matrix coefficients are calculated on one processor and sent to the neighbour.
- Preallocation as crucial aspect for program performance. For the coupled system the matrix is assembled in a 2-3 step process to save memory for coefficients.
- Present a simple method for balancing the matrix related load by letting PETSc take care of matrix distribution.
- Use Spy function of Matlab to visualize the sparse matrices. Point out advantages of calculating coefficients for the neighbouring cells locally (no need to update mass fluxes, geometric data doesn't need to be shared, small communication overhead since processors assemble matrix parts that don't belong to them (visualize)).
- Paradigm: Each time new information is available perform global updates. Advantages of using matrices: Show structure of matrix when using arbitrary matching vs. higher memory requirements vs. better convergence

6.5 Postprocessing

Visualization of Results with Paraview and Tecplot Export matrices as binaries and visualize them using matlab scripts.

7 Verification of CAFFA

Different parts, describe incremental approach, only present final results. Refer to next section for Validation of CAFFA

7.1 Theoretical Discretization Error

present the Taylor-Series Expansion

7.2 Method of Manufactured Solutions

basically sum up the important points of salari's technical report, symmetry of solution/domain/grid is bad point out that mms is not able to detect errors in the physical model Also loose a word or two about discontinuous manufactured solutions

7.3 Exact and Manufactured Solutions for the Navier-Stokes Equations and the Energy Equation

Not always there is an exact solution. Divergence free approach. Presentation of the used manufactured solution. What if solution is not divergence free? Derivation of equations and modifications to continuity equation. analyze the problem of too complicated manufactured solutions. also use temperature dependent density function. Explain why global mass conservation in a discrete sense is important and how it can be achived. Special domain, vanishing manufactured solution or symmetric manufactured solutions if a higher approximation of boundary fluxes is not feasible.

- `http://scicomp.stackexchange.com/questions/6943/manufactured-solutions-for-incompressible-navier-stokes`

- http://link.springer.com/article/10.1007/BF00948290
- http://physics.stackexchange.com/questions/60476/exact-solutions-to-the-navier-stokes-equations
- http://www.annualreviews.org/doi/pdf/10.1146/annurev.fl.23.010191.001111

7.4 Measurement of Error and Calculation of Order

Different error measures (L2-Norm,completeness of function space, consistency etc.)

7.4.1 Testcase on Single Processor on Orthogonal Locally Refined Grid

7.4.2 Testcase on Multiple Processors on Non-Orthogonal Locally Refined Grid

Give a measure of the grid quality.

8 Comparison of Solver Concepts

8.1 Convergence Behaviour on Locally Refined Block Structured Grids with Different Degrees of Coupling

Show how the implicit treatment of block boundaries maintains (high) convergence rates. Plot Residual over number of iterations.

8.2 Parallel Performance

8.2.1 Employed Hardware and Software – The Lichtenberg-High Performance Computer

- Networking
- Mem Section and processes in between islands (calculating across islands)
- Versioning information (PETSc,INTEL COMPILERS,CLIPPER,MPI IMPLEMENTATION,BLAS/LAPACK)
- Software not designed to perform well on desktop PCs.

8.2.2 Measures of Performance

- Maße definieren
- Nochmal Hager,Wellein studieren
- Guidelines for measuring performance (bias through system processes or user interaction), only measure calculation time do not consider I/O in the beginning and the end
- Cite Schäfer and Peric with their different indicators for parallel efficiency, load balancing and numerical efficiency

8.2.3 Preliminary Upper Bounds on Performance – The STREAM Benchmark

Pinning of processes (picture), preliminary constraints by hardware and operating systems, identification of bottlenecks and explain possible workarounds, history and results of STREAM. Bandwidth as Bottleneck, how to calculate a Speedup estimate based on the measured bandwidth. PETSc Implementation of STREAM

8.2.4 Discussion of Results for Parallel Efficiency

8.2.5 Speedup Measurement for Analytic Test Cases

8.3 Test Cases with Varying Degree of Non-Linearity

As Peric says I want to prove that the higher the non-linearity of NS, the better relative convergence rates can be achieved with a coupled solver. Fi

8.3.1 Transport of a Passive Scalar – Forced Convection

8.3.2 Buoyancy Driven Flow – Natural Convection

8.3.3 Flow with Temperature Dependent Density – A Highly Non-Linear Test Case

Maybe I could consider two test cases, one with oscillating density and one with a quadratic polynomial. Interesting would be also to consider the dependence of convergence on another scalar transport equation

8.4 Realistic Testing Scenarios – Benchmarking

Also consider simple load balancing by distributing matrix rows equally

8.4.1 Flow Around a Cylinder 3D – Stationary

Describe Testing Setup (Boundary conditions and grid). Present results and compare them with literature.

8.4.2 Flow Around a Cylinder 3D – Instationary

- `http://www.featflow.de/en/benchmarks/cfdbenchmarking/flow/dfg_flow3d/dfg_flow3d_configuration.html`

Describe Testing Setup (Boundary conditions and grid). Present results and compare them with literature.

8.4.3 Heat-Driven Cavity Flow

- `http://www.featflow.de/en/benchmarks/cfdbenchmarking/mit_benchmark.html`

Describe Testing Setup (Boundary conditions and grid). Present results and compare them with literature.

8.5 Realistic Testing Scenario – Complex Geometry

9 Conclusion and Outlook

Turbulence (turbulent viscosity has to be updated in each iteration), Multiphase (what about discontinuities), GPU-Accelerators, Load-Balancing, dynamic mesh refinement, Counjugate Heat Transfer with other requirements for the numerical grid, grid movement, list some papers here) Identify the optimal regimes / conditions for maximizing performance. Each solver concept has its strengths and weaknesses. Try other variants of Projection Methods like SIMPLEC, SIMPLER, PISO or PIMPLE (OpenFOAM)