

Software Lab Computational Engineering Science

Type-Generic Solution for Systems of Nonlinear Equations

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Informatik 12:
Software and Tools for Computational Engineering (STCE)

RWTH Aachen University

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Summary and Next Steps

Objective

- ▶ Introduction to design and implementation of a type-generic solution infrastructure for parameterized systems of nonlinear equations

Learning Outcomes

- ▶ You will understand
 - ▶ requirements, design, implementation of the sample code
- ▶ You will be able to
 - ▶ download, build and run the sample code.

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We consider parameterized systems of nonlinear equations

$$\mathbf{y} = \mathbf{F}(\mathbf{x}, \mathbf{p}) = 0$$

for $\mathbf{x} = \mathbf{x}(\mathbf{p})$ and with invertible Jacobians

$$\mathbf{F}' = \frac{d\mathbf{F}}{d\mathbf{x}}(\mathbf{x}, \mathbf{p}) \in \mathbf{R}^{n_s \times n_s}$$

of the residual $\mathbf{F} : \mathbf{R}^{n_s} \times \mathbf{R}^{n_p} \rightarrow \mathbf{R}^{n_s}$ at all pairs of states \mathbf{x} and parameters \mathbf{p} of interest.

The Newton method iteratively computes the fixed point

$$\mathbf{x} = \mathbf{x} - (\mathbf{F}'(\mathbf{x}, \mathbf{p}))^{-1} \cdot \mathbf{F}(\mathbf{x}, \mathbf{p}) .$$

See modules I and II on Newton method.

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Summary and Next Steps

I am looking for a software library for solving parameterized systems of nonlinear equations $F(\mathbf{x}(\mathbf{p}), \mathbf{p}) = 0$ allowing me to

- ▶ define the residual $F : \mathbb{R}^{n_s} \times \mathbb{R}^{n_p} \rightarrow \mathbb{R}^{n_s}$ as well as potentially required derivatives thereof
- ▶ define, store and extract the state $\mathbf{x} \in \mathbb{R}^{n_s}$ and the parameters $\mathbf{p} \in \mathbb{R}^{n_p}$
- ▶ extend the choice of nonlinear solvers (in addition to the Newton method)

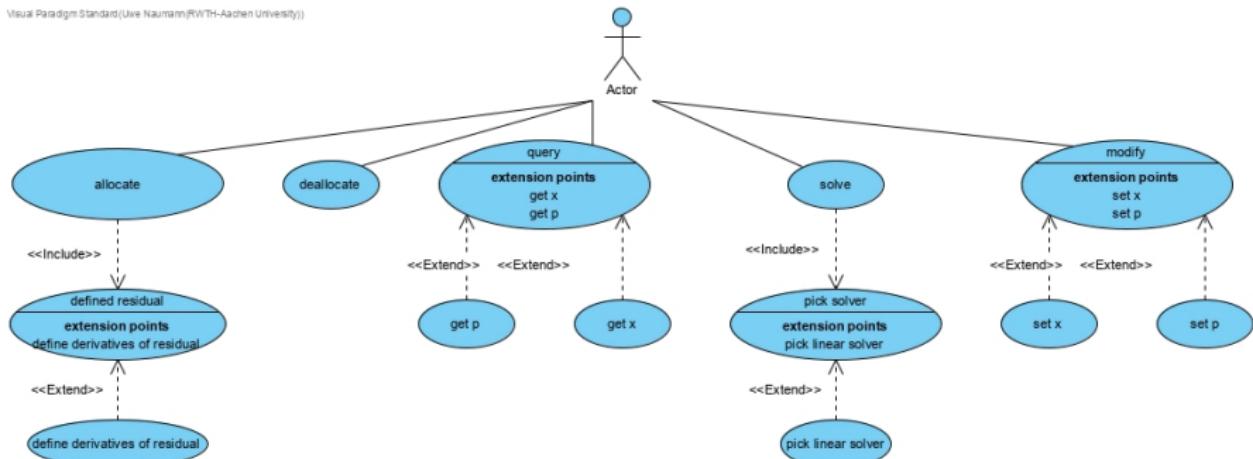
Algorithmic differentiation with dco/c++ shall be used for sensitivity analysis of functions of \mathbf{x} (e.g., $\|\mathbf{x}\|_2$) wrt. to perturbations in \mathbf{p} .

Most likely, it will have to be embedded into other software solutions, e.g., for the solution of systems of ordinary differential equations at some later stage.

Analysis

Use Cases

Visual Paradigm Standard (Uwe Naumann (RWTH-Aachen University))



- ▶ nonlinear system
 - ▶ allocation
 - ▶ deallocation
 - ▶ generic element types TS of elements of state and TP of parameters enabling **overloading** and corresponding matrix and vector types.
 - ▶ read/write access routines for **x** and **p**
 - ▶ implementation of residual and of required derivatives
- ▶ nonlinear solver
 - ▶ allocation
 - ▶ deallocation
 - ▶ solution of nonlinear system (e.g, Newton using libls)
 - ▶ abstraction for extensibility
 - ▶ **overloading**

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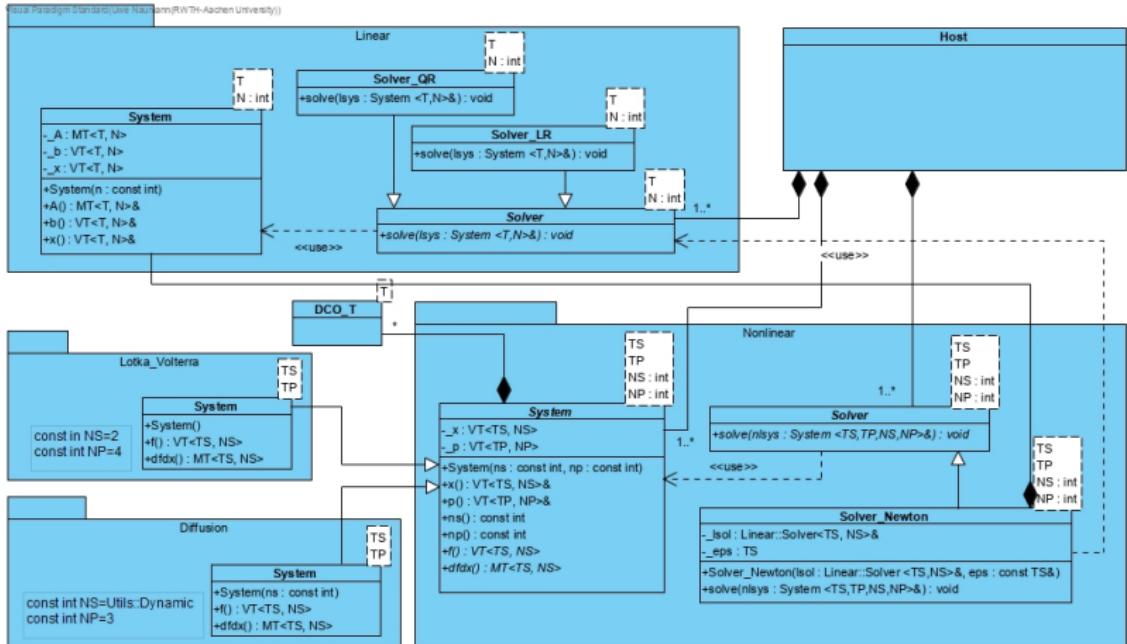
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Implementation

Library

```
1 doc
2 Doxyfile
3 Makefile
4
5 include
6 nonlinear_solver.hpp
7 nonlinear_solver_newton.hpp
8 nonlinear_system.hpp
9
10 Makefile
11
12 src
13     nonlinear_solver_newton.cpp
14     nonlinear_system.cpp
```

Implementation

nonlinear_system.hpp

```
1 #include <Eigen/Dense>
2
3 namespace Nonlinear {
4
5     template<typename TS, typename TP, int NS, int NP>
6     class System
7     {
8         public:
9             using VTS=Eigen::Matrix<TS,NS,1>;
10            using MTS=Eigen::Matrix<TS,NS,NS>;
11            using VTP=Eigen::Matrix<TP,NP,1>;
12
13         protected:
14             VTS _x; VTP _p;
15
16         public:
17             System(int,int);
18             int ns(); int np(); VTS& x(); VTP& p();
19             virtual VTS f()=0; virtual MTS dfdx()=0;
20
21     };
22 }
```

#include "../src/nonlinear_system.cpp"

```
1  namespace Nonlinear {  
2  
3      template<typename TS, typename TP, int NS, int NP>  
4      System<TS,TP,NS,NP>::System(int ns, int np)  
5          : _x(VTS::Zero(ns))  
6          , _p(VTP::Zero(np)) {}  
7  
8      template<typename TS, typename TP, int NS, int NP>  
9      int System<TS,TP,NS,NP>::ns() { return _x.size(); }  
10  
11     template<typename TS, typename TP, int NS, int NP>  
12     int System<TS,TP,NS,NP>::np() { return _p.size(); }  
13  
14     template<typename TS, typename TP, int NS, int NP>  
15     typename System<TS,TP,NS,NP>::VTS&  
16     System<TS,TP,NS,NP>::_x() { return _x; }  
17  
18     template<typename TS, typename TP, int NS, int NP>  
19     typename System<TS,TP,NS,NP>::VTP&  
20     System<TS,TP,NS,NP>::_p() { return _p; }  
21  
22 }
```

```
1 #include "nonlinear_system.hpp"
2
3 namespace Nonlinear {
4
5     template<typename TS, typename TP, int NS, int NP>
6     class Solver {
7         public:
8             virtual void solve(System<TS,TP,NS,NP>&)=0;
9         };
10    }
11 }
```

Implementation

nonlinear_solver_newton.hpp

```
1 #include "linear_solver.hpp"
2 #include "nonlinear_system.hpp"
3 #include "nonlinear_solver.hpp"
4
5 namespace Nonlinear {
6
7     template<typename TS, typename TP, int NS, int NP>
8     class Solver_Newton : public Solver<TS,TP,NS,NP> {
9         Linear::Solver<TS,NS>& _lsol;
10        TS _eps;
11    public:
12        Solver_Newton(Linear::Solver<TS,NS>&, const TS&);
13        void solve(System<TS,TP,NS,NP>&);
14    };
15
16 }
17
18 #include "../src/nonlinear_solver_newton.cpp"
```

Implementation

nonlinear_solver_newton.cpp

```
1  namespace Nonlinear {  
2  
3      template<typename TS, typename TP, int NS, int NP>  
4      Solver_Newton<TS,TP,NS,NP>::Solver_Newton(Linear::Solver<TS,NS>& lsol, const TS&  
5          eps) : _lsol(lsol), _eps(eps) {}  
6  
6      template<typename TS, typename TP, int NS, int NP>  
7      void  
8      Solver_Newton<TS,TP,NS,NP>::solve(System<TS,TP,NS,NP>& nlsys) {  
9          Linear::System<TS,NS> lsys(nlsys.ns());  
10         while (nlsys.f().norm()>_eps) {  
11             lsys.A()=nlsys.dfdx(); lsys.b()=-nlsys.f();  
12             _lsol.solve(lsys);  
13             nlsys.x()+=lsys.x();  
14         }  
15     }  
16 }  
17 }
```

Implementation Building

There is nothing to do for a C++ template (“header-only”) library.

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Overview

```
1 doc
2 Doxyfile
3 Makefile
4
5 include
6 nonlinear_system_diffusion.hpp
7 nonlinear_system_lotka_volterra.hpp
8 nonlinear_system_toy.hpp
9
10 Makefile
11
12 src
13 nonlinear_system_diffusion.cpp
14 nonlinear_system_lotka_volterra.cpp
15 nonlinear_system_toy.cpp
16
17 toy.cpp
18 diffusion.cpp
19 lotka_volterra.cpp
```

Application

Toy Problem: $x^2 - p = 0$ (nonlinear_system.toy.hpp)

```
1 #include "nonlinear_system.hpp"
2
3 namespace Toy {
4
5     const int NS=1;
6     const int NP=1;
7
8     template<typename TS, typename TP>
9     class System : public Nonlinear::System<TS,TP,1,1>
10    {
11         using Nonlinear::System<TS,TP,NS,NP>::x;
12         using Nonlinear::System<TS,TP,NS,NP>::p;
13     public:
14         System();
15         typename Nonlinear::System<TS,TP,NS,NP>::VTS f();
16         typename Nonlinear::System<TS,TP,NS,NP>::MTS dfdx();
17    };
18
19 }
20
21 #include "../src/nonlinear_system_toy.cpp"
```

Toy Problem: $x^2 - p = 0$ (nonlinear_system.toy.cpp)

```
1  namespace Toy {  
2  
3      template<typename TS, typename TP>  
4          System<TS,TP>::System() : Nonlinear::System<TS,TP,NS,NP>(NS,NP) {}  
5  
6      template<typename TS, typename TP>  
7          typename Nonlinear::System<TS,TP,NS,NP>::VTS  
8          System<TS,TP>::f() {  
9              typename Nonlinear::System<TS,TP,NS,NP>::VTS r;  
10             r(0)=-x(0)*_x(0)-_p(0);  
11             return r;  
12         }  
13  
14     template<typename TS, typename TP>  
15         typename Nonlinear::System<TS,TP,NS,NP>::MTS  
16         System<TS,TP>::dfdx() {  
17             typename Nonlinear::System<TS,TP,NS,NP>::MTS drdx;  
18             drdx(0,0)=2*_x(0);  
19             return drdx;  
20         }  
21     }  
22 }
```

```
1 #include<iostream>
2
3 #include "nonlinear_system_toy.hpp"
4 #include "linear_solver_qr.hpp"
5 #include "nonlinear_solver_newton.hpp"
6
7 int main() {
8     using T=double;
9     Toy::System<T,T> nlsys;
10    Linear::Solver_QR<T,Toy::NS> lsol;
11    Nonlinear::Solver_Newton<T,T,Toy::NS,Toy::NP> nlsol(lsol,1e-7);
12    nlsys.x()(0)=1;
13    nlsys.p()(0)=2;
14    nlsol.solve(nlsys);
15    std::cout << "x=" << nlsys.x() << std::endl;
16    return 0;
17 }
```

Application Building

```
1 EXE=$(addsuffix .exe, $(basename $(wildcard *.cpp)))
2 CPPC=g++
3 CPPC_FLAGS=-Wall -Wextra -pedantic -Ofast -march=native
4 EIGEN_DIR=$(HOME)/Software/Eigen
5
6 BASE_DIR=$(HOME)/Documents/gitlab/e-learning/SP-CES/code
7 LIBLS_INC_DIR=$(BASE_DIR)/LINEAR_SYSTEM/libls/include
8 LIBNLS_INC_DIR=$(BASE_DIR)/NONLINEAR_SYSTEM/libnls/include
9 LIBNLS_APPS_INC_DIR=$(BASE_DIR)/NONLINEAR_SYSTEM/libnls_apps/include
10
11 all : $(EXE)
12
13 %.exe : %.cpp
14     $(CPPC) $(CPPC_FLAGS) -I$(EIGEN_DIR) -I$(LIBLS_INC_DIR) -I$(LIBNLS_INC_DIR) -I$(LIBNLS_APPS_INC_DIR) $< -o $@
15
16 clean :
17     rm -fr $(EXE)
18
19 .PHONY: all clean
```

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We consider the algorithmic differentiation of the solution of $x^2 - p = 0$ wrt. p in with dco/c++.

Setting $p^{(1)} = 1$ in

$$x^{(1)} = \frac{dx}{dp} \cdot p^{(1)}$$

yields the derivative as the tangent of the state x .

See [module I](#) on algorithmic differentiation and [source code](#).

Case Studies

Building

```
1 EXE=$(addsuffix .exe, $(basename $(wildcard *.cpp)))
2 CPPC=g++
3 CPPC_FLAGS=-Wall -Wextra -pedantic -Ofast -march=native
4 EIGEN_DIR=$(HOME)/Software/Eigen
5 DCO_DIR=$(HOME)/Software/dco
6 DCO_INC_DIR=$(DCO_DIR)/include
7 DCO_LIB_DIR=$(DCO_DIR)/lib
8 DCO_FLAGS=-DDCO_DISABLE_AUTO_WARNING
9 DCO_LIB=dcoc

10
11 BASE_DIR=$(HOME)/Documents/gitlab/e-learning/SP_CES/code
12 LIBLS_INC_DIR=$(BASE_DIR)/LINEAR_SYSTEM/libls/include
13 LIBNLS_INC_DIR=$(BASE_DIR)/NONLINEAR_SYSTEM/libnls/include
14 LIBNLS_APPS_INC_DIR=$(BASE_DIR)/NONLINEAR_SYSTEM/libnls_apps/include

15
16 all : $(EXE)

17
18 %.exe : %.cpp
19     $(CPPC) $(CPPC_FLAGS) $(DCO_FLAGS) -I$(EIGEN_DIR) -I$(DCO_INC_DIR)
20         -I$(LIBLS_INC_DIR) -I$(LIBNLS_INC_DIR) -I$(LIBNLS_APPS_INC_DIR) -L$(
21             DCO_LIB_DIR) $< -o $@ -I$(DCO_LIB)
22 ...
23 ...
```

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Summary and Next Steps

Summary

- ▶ Discussion of requirements, design and implementation of a type-generic solution infrastructure for parameterized systems of nonlinear equations

Next Steps

- ▶ Download, build and run the sample code.
- ▶ Inspect the sample code.
- ▶ Continue the course to find out more ...