

► Overview of Optimica by Johan Åkesson

Remark: Optimica will be used in a computer exercise, but you need not to learn any details about Modelica, AMPL, IPOPT etc.

► Extra example

Minimum-time problem for double integrator

Bring this hand-out to computer exercise CE5

Dynamic Optimization with Optimica, AMPL and IPOPT

Johan Åkesson

Dept. of Automatic Control
Lund Institute of Technology

Outline

- Motivation of Optimica
- System overview
- The Optimica language extension
- AMPL
- CSTR example in Optimica
- Working with Dymola and Optimica

Optimica – Motivation

- It is often cumbersome to use numerical optimization algorithms for large-scale systems
 - Code the dynamics in C or FORTRAN: inconvenient modeling formalism
 - Compute and code derivatives
 - Difficult to reuse code
- High-level modeling language Modelica increasingly used
 - Object-oriented component-based modeling
 - Libraries and reuse of code
- Modelica does not have language support for optimization
- The JModelica project and Optimica
 - Develop a prototype Modelica compiler in JastAdd/Java
 - Extend Modelica to support dynamic optimization – Optimica
 - Modelica used to express dynamics
 - Optimica used to express cost functions, constraints etc.

The Optimica Extension – Scope

$$\min_{u(t), p} \int_0^{t_f} L(x(t), u(t), p) dt + \phi(x(t_f))$$

subject to

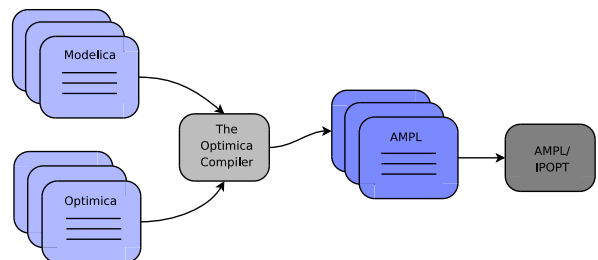
$$f(\dot{x}(t), x(t), u(t), p) = 0$$

and

$$\begin{array}{ll} c_e(x(t), u(t), p) = 0 & c_i(x(t), u(t), p) \leq 0 \\ c_{fe}(x(t_f), u(t_f), p) = 0 & c_{fi}(x(t_f), u(t_f), p) \leq 0 \\ c_{0e}(x(0), u(0), p) = 0 & c_{0i}(x(0), u(0), p) \leq 0 \end{array}$$

- Modelica to express dynamic constraint
- Optimica to express everything else

System Overview



- The Optimica compiler reads Modelica and Optimica models and outputs a set of AMPL files
- The AMPL files can be executed by the AMPL tools
- The problem is solved numerically by IPOPT

The Optimica Extension – Scope cont'd

$$\min_{u, p} M(x)$$

subject to

$$\begin{array}{l} f(\dot{x}, x, u, p)|_{\dot{x}=0} = 0 \\ c_i(x, u, p) \leq 0 \\ c_e(x, u, p) = 0 \end{array}$$

- Static optimization based on a static or dynamic Modelica model

The Optimica Compiler

- ▶ Extension of a Modelica compiler: The JModelica compiler for pure Modelica
- ▶ The following steps are performed in the compilation
 - ▶ **Flattening:** The original Modelica model is transformed into a flat differential equation. This involves instantiation of components and resolution of inheritance.
 - ▶ **Transcription:** The derivative operators in the flat model representation is transcribed using the collocation method based on Lagrange polynomials
 - ▶ **Code generation:** The flat transcribed representation is written, in the form of an NLP problem, to AMPL format
- ▶ The cumbersome and error-prone transcription procedure is performed automatically
- ▶ The user can focus on *formulation* of the optimization problem instead of *encoding* it

Optimica – Superimpose information

- ▶ Variable bounds (path constraints for variables)
`varName(lowerBound=-1,upperBound=1);`
- ▶ Let a Modelica parameter be free in the optimization
`oq varName(lowerBound=-1,upperBound=1);`
- ▶ Initial guess
`varName(lowerBound=-1,
 upperBound=1,
 initialGuess=0);`
- ▶ Specify free initial conditions for state variables
`varName(freeInitial(
 lowerBound=[-0.01;-0.001;-0.01;-0.001],
 upperBound=[0.01;0.001;0.01;0.001],
 initialGuess=[0.001;0;0;0])=true);`

Optimica – Cost Function and Constraints

- ▶ Cost function specified in section optimization
- ▶ Lagrange cost function
`minimize(lagrangeIntegrand=li_exp);`
- ▶ Terminal cost function
`minimize(terminalCost=tc_exp);`
- ▶ Combining Lagrange and terminal cost
`minimize(lagrangeIntegrand=li_exp,
 terminalCost=tc_exp);`
- ▶ Constraints declared in section subject to
`y<=x^2; // Path constraint
initial cos(x)>=0.4 // Initial constraint
terminal y=4; // Terminal constraint`

Example – CSTR: Optimica

```
class CSTR
optimization
  grid(finalTime=fixedFinalTime(finalTime=60),
        nbrElements=300);
  minimize(lagrangeIntegrand = q1*(cref-c)^2+
        q2*(Tref-T)^2+
        r*(uref-u)^2);
end CSTR;
```

- ▶ Mesh with 300 elements and fixed final time 60 s
- ▶ Lagrange cost function

The Optimica Extension

- ▶ An Optimica model is complementing a Modelica model
 - ▶ All declarations in the Modelica model are (implicitly) visible in the corresponding Optimica description
- ▶ Language elements
 - ▶ Superimpose bounds on variables
 - ▶ Mark Modelica parameters as free in optimization
 - ▶ Specify initial guesses
 - ▶ Mesh
 - ▶ Cost function
 - ▶ Constraints
- ▶ Initial guesses can be read from file

Optimica – Mesh

- ▶ A collocation method use a *mesh* consisting of a number of finite elements
- ▶ Declared in section optimization
- ▶ Mesh with fixed final time
`grid(finalTime=fixedFinalTime(finalTime=tf),
 nbrElements=n_el);`
- ▶ Mesh with free final time
`grid(finalTime=openFinalTime(initialGuess=tf_ig,
 lowerBound=tf_lb,
 upperBound=tf_ub),
 nbrElements=n_el);`
- ▶ Static optimization
`grid(static=true);`

Example – CSTR: Modelica

```
model CSTR
  parameter Real beta=1/20, J=100, cf=7.6;
  parameter Real alpha=1.95e-4, Tf=300, k=300;
  parameter Real Tc=290, N=5;
  parameter Real yc=Tc/J/cf, yf=Tf/J/cf;
  parameter Real c0=0.1367, T0=0.7293;
  parameter Real cref=0.0944, Tref=0.7766, uref=340;
  parameter Real q1=1, q2=1, r=0.000001;
  Real c(start=c0), T(start=T0),
  input Real u;
equation
  der(c)=beta*(1-c)-k*exp(-N/T)*c;
  der(T)=beta*(yf-T)+k*exp(-N/T)*c-alpha*(T-yc)*u;
end CSTR;
```

Running the Optimica Compiler

- ▶ `optimicac` arguments: argument
 - ▶ Name of Optimica file
 - ▶ Name of Modelica file
 - ▶ Name of Modelica class to optimize
 - ▶ Name of initial guess file (optional)
- ▶ TOC output

```
> optimicac CSTR.op CSTR.mo CSTR
Parsing CSTR.mo...
Checking for errors in Modelica file...
0 errors detected...
Instantiation starts...
Instantiation finished!
Parsing CSTR.op...
Checking for errors in Optimica file...
0 errors detected...
Generating AMPL code...
Code generation finished!
```

- ▶ Model definition
CSTR.mod
- ▶ Data definition
CSTR.dat
- ▶ Initial guess file
CSTR.InitialGuess.mod
- ▶ Constraints
CSTR.Constraint.mod
- ▶ Cost
CSTR.Cost.mod
- ▶ Script for generating result file
CSTR.GenLogFile.run
- ▶ Script for loading problem definitions and solving problem
CSTR.run

AMPL – Example

```

ampl: var x >=-10 <=10;
ampl: var y=(x-3)^2+3;
ampl: minimize COST: y;
ampl: solve;
... Ipopt 3.2.0: Optimal Solution Found
ampl: display x;
x = 3
ampl: display y;
y = 3
ampl: subject to CONSTR: x>=5;
ampl: solve;
... Ipopt 3.2.0: Optimal Solution Found
ampl: display x;
x = 5
ampl: display y;
y = 7

```

Static Optimization Example

- ▶ Modelica class


```

class StaticOpt
  Real x[4];
end StaticOpt;

```
- ▶ Optimica class


```

class StaticOpt
  x(lowerBound={1,1,1,1}, upperBound={5,5,5,5},
    initialGuess={1,5,5,1});
  optimization
    grid(static=true);
    minimize(terminalCost=x[1]*x[4]*(x[1]+x[2]+x[3])+x[3]);
  subject to
    x[1]*x[2]*x[3]*x[4]>=25;
    x[1]^2+x[2]^2+x[3]^2+x[4]^2=40;
  end StaticOpt;

```

Summary

- ▶ Modelica and Optimica for high-level dynamic optimization problems
- ▶ Transcription automated by the Optimica compiler
- ▶ AMPL – A language (and tool) for mathematical programming
- ▶ Numerical solver IPOPT used to solve the resulting NLP
- ▶ Results can be loaded into Matlab or Dymola

- ▶ A language for mathematical programming
 - ▶ Variable and set declarations
 - ▶ Cost function
 - ▶ Constraints
 - ▶ Model and data separation
 - ▶ But no derivative operator...
- ▶ The AMPL shell offers a scripting language which enables definition and execution of optimization problems
- ▶ Interfaces to a large number of numerical solvers (including IPOPT)
- ▶ Offers an API with sparsity patterns, first and second order derivatives to solvers
- ▶ Numerical solver invoked from AMPL shell

CSTR Example cont'd

- ▶ Execution in AMPL


```

ampl: include CSTR.run;

```
- ▶ Result is written to file CSTR_res.txt in Dymola (textual) format
- ▶ Load results into Matlab


```

>> d=loadsym('CSTR_res.txt');

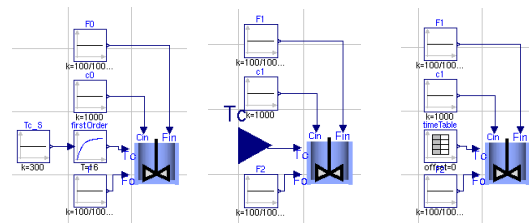
```
- ▶ Retrieve trajectories


```

>> [t,c]=getvar(d,'c');
>> [t,T]=getvar(d,'T');
>> [t,u]=getvar(d,'u');

```

Working with Dymola



Generation of initial guess

Optimization model

Verify optimal solution

Extra example—Minimum Time Control

Problem: Use bounded control, $u \in [-1, 1]$, and bring the states of the double integrator to the origin as fast as possible.

Free end-time t_f

$$\begin{aligned} \min_{u: [0, t_f] \rightarrow [-1, 1]} t_f &= \min_{u: [0, t_f] \rightarrow [-1, 1]} \int_0^{t_f} 1 \, dt \\ \dot{x}_1(t) &= x_2(t) \\ \dot{x}_2(t) &= u(t) \\ \psi(x(t_f)) &= (x_1(t_f), x_2(t_f))^T = (0, 0)^T \end{aligned}$$

Adjoint equations $\dot{\lambda}_1(t) = 0$, $\dot{\lambda}_2(t) = -\lambda_1(t)$ gives

$$\lambda_1(t) = c_1, \quad \lambda_2(t) = c_2 - c_1 t$$

With $u(t) = \zeta = \pm 1$, we have

$$\begin{aligned} x_1(t) &= x_1(0) + x_2(0)t + \zeta t^2/2 \\ x_2(t) &= x_2(0) + \zeta t \end{aligned}$$

Eliminating t gives curves (parabolas)

$$x_1(t) \pm x_2(t)^2/2 = \text{const}$$

These define the *switch curves*, where the optimal control switch.

Remark: See also solution to exam March 8, 2005 for more details

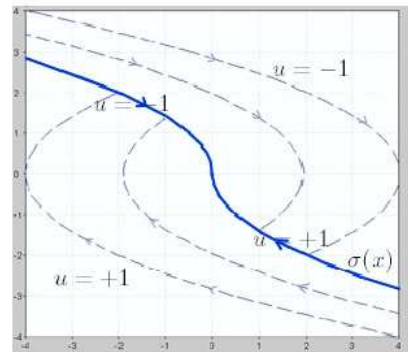
Hamiltonian

$$H = L + \lambda^T f = 1 + [\lambda_1 \lambda_2] \begin{bmatrix} x_2 \\ u \end{bmatrix}$$

The optimal control is the control which minimizes

$$\begin{aligned} u^*(t) &= \arg \min_{u \in [-1, 1]} H = \arg \min_{u \in [-1, 1]} 1 + \lambda_1(t)x_2^*(t) + \lambda_2(t)u \\ &= \begin{cases} 1, & \lambda_2(t) < 0 \\ -1, & \lambda_2(t) \geq 0 \end{cases} \end{aligned}$$

Note: This is "bang-bang control".



Combination of solution curves: Follow one parabola using maximum or minimum control signal until you hit the switch curve σ , switch to minimum/maximum control and then follow that parabola to the origin.