User Manual of *ParNMPC* by Examples

A Code Generation Toolbox for Parallel Nonlinear Model Predictive Control with OpenMP

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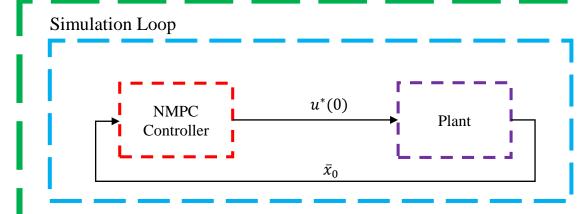
- Highly Parallelizable
- Fast Rate of Convergence
- *Parallel C/C++ Code Generation*
- Closed-loop Simulation
- Easy to Use

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Introduction

ParNMPC



ParNMPC is a Matlab software toolbox developed to carry out closed-loop NMPC simulation and parallel C/C++ code generation with OpenMP. ParNMPC is based on a highly parallelizable Newton-type algorithm, which can utilize N cores at most to calculate the optimal control input. To use ParNMPC, you have to define the NMPC Controller, the Plant and the Simulation Loop. Parameters and functions need to be defined are listed below.

NMPC Controller

$$\min_{x(\cdot), u(\cdot)} \int_0^T L(u(t), x(t), p(t)) dt$$
s.t. $\dot{x}(t) = f(u(t), x(t), p(t)),$

$$x(0) = \bar{x}_0,$$

$$C(u(t), x(t), p(t)) = 0, t \in (0, T].$$

Functions:

- f: dynamic model for controller design
 - L: cost function
- C: constraint

Parameters:

- x: state vector
- *u*: input vector
- *p*: parameter vector
 - \bar{x}_0 : initial state vector
- T: prediction horizon
- *N*: number of discretization grids

Plant

$$\dot{x}_{sim} = f_{sim}(u_{sim}, x_{sim}, p_{sim})$$

- f_{sim} : dynamic model of the plant
- p_{sim} : parameter in the plant model

Simulation Loop

Simulation parameters:

- Sampling interval T_s
- Simulation length

Introduction

• In practice, ParNMPC solves the discretized optimal control problem (OCP) for a given \bar{x}_0 , and apply u_1^* .

$$\min_{x(\cdot), u(\cdot)} \int_0^T L(u(t), x(t), p(t)) dt$$
s.t. $\dot{x}(t) = f(u(t), x(t), p(t))$,
$$x(0) = \bar{x}_0,$$

$$C(u(t), x(t), p(t)) = 0, t \in (0, T].$$
Discretization with the implicit Euler method
$$C(u(t), x(t), p(t)) = 0, t \in (0, T].$$
Discretization with the implicit Euler method
$$C(u_i, x_i, p_i) = 0, i = 1, \dots, N,$$

$$U = (u_1, u_2, \dots, u_N),$$

$$U = (u_1, u_2, \dots, u_N),$$

$$\Delta \tau = T/N.$$

• More specifically, *ParNMPC* solves the KKT conditions of the discretized OCP, i.e., the following nonlinear algebraic equations with respect to the costates $\{\lambda_i\}_{i=1}^N$, the multipliers $\{\mu_i\}_{i=1}^N$, the inputs $\{u_i\}_{i=1}^N$ and the states $\{x_i\}_{i=1}^N$:

$$\begin{bmatrix} x_{i-1} - x_i + f(u_i, x_i, p_i) \Delta \tau \\ C(u_i, x_i, p_i) \Delta \tau \\ H_u^T(\lambda_i, \mu_i, u_i, x_i, p_i) \Delta \tau \\ \lambda_{i+1} - \lambda_i + H_x^T(\lambda_i, \mu_i, u_i, x_i, p_i) \Delta \tau \end{bmatrix}_{i=1}^{N} = 0 \in \mathbb{R}^{N(2n_x + n_\mu + n_u)}$$

Here, $H(\lambda, \mu, u, x, p) \coloneqq L(u, x, p) + \lambda^T f(u, x, p) + \mu^T C(u, x, p)$ denotes the Hamiltonian, $\lambda \in \mathbb{R}^{n_x}, \mu \in \mathbb{R}^{n_\mu}, u \in \mathbb{R}^{n_u}, x \in \mathbb{R}^{n_x}, p \in \mathbb{R}^{n_p}, x_0 = \bar{x}_0$ and $\lambda_{N+1} = 0$.

Installation

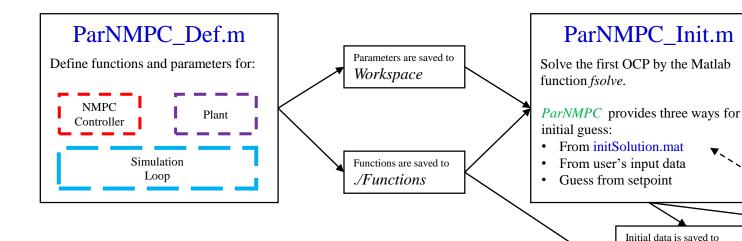
- 1. Download ParNMPC-master.zip
- 2. Extract files (5 folders):
 - CSTR
 - DoubleInvertedPendulum
 - Quadrotor
 - SemiActiveDamper
 - UserManual
- 3. Go into one of the application folders, e.g., *Quadrotor*
- 4. Files and folders in *Quadrotor*:
 - Folders
 - codegen (Pre-generated C code)
 - Functions (Functions and scripts)
 - Files:
 - ParNMPC_Def.m (Define the simulation problem)
 - ParNMPC_Init.m (Solve the first OCP)
 - ParNMPC.m (For simulation and code generation)
 - initialSolution.mat (Solution for the first OCP)
- ParNMPC is application-oriented and each application folder is self-contained.

System Requirements

- MATLAB 2016a or later
 - MATLAB Coder
 - Optimization Toolbox
 - Parallel Computing Toolbox
 - Symbolic Math Toolbox
- Compiler
 - C/C++ Compilers supporting OpenMP for code generation
 - See <u>Supported and Compatible Compilers</u>

- Tested Environments:
 - Windows 10 + MATLAB 2016a + Microsoft Visual C++ 2015 Professional
 - Windows 10 + MATLAB 2017b + Microsoft Visual C++ 2017 Community

Flow of *ParNMPC*



Steps of using *ParNMPC*:

- The user has to define the closed-loop simulation problem in ParNMPC Def.m and run this file.
- Run ParNMPC_Init.m to solve the very first OCP. The user has to select one method and provide the corresponding parameters to guess the solution (initial guess).
- 3. Run ParNMPC.m to check the simulation results. Generate code if all go well.
- Deploy the generated C/C++ code.

ParNMPC.m

GEN initData.mat

For simulation and code generation

- Matlab mode:
 - Simulation in Matlab in serial
- Code generation mode:
 - Simulation by C/C++ in parallel

Simulation results are saved to GEN log rec

Initial solution is saved to

initSolution.mat

Getting Started Semi-Active Damper

Simulation Loop:

- Sampling interval: 0.01 s
- Simulation length: 20 s

Plant:

Same as in NMPC:

$$f_{sim}(u,x,p_{sim}) = \begin{bmatrix} x_2 \\ ax_1 + bx_2u_1 \end{bmatrix}$$

NMPC Controller:

• Dynamics:

$$f(u, x, p) = \begin{bmatrix} x_2 \\ ax_1 + bx_2u_1 \end{bmatrix}, a = b = -1$$

• Constraints (Converted into equality constraint by introducing dummy input u_2 (see *ref*)):

$$-1 \le u_1 \le 1 \longrightarrow C(u, x, p) = u_1^2 + u_2^2 - 1$$

• Stage cost function:

$$L(u, x, p) = \frac{1}{2} (\|x - x_{ref}\|_{Q}^{2} + \|u - u_{ref}\|_{R}^{2} - r \cdot u)$$

$$Q = \text{diag}(10,10), R = \text{diag}(0.01,0.01), r = [0,0.1]$$

$$x_{ref} = [0,0]^{T}, u_{ref} = [0,1]^{T}$$

- Initial state: $\bar{x}_0 = [1,0]^T$
- Prediction horizon: 2 s
- Number of discretization grids: N = 36
- Max number of iterations: 5
- Tolerance: 1e-5

Semi-Active Damper – Problem Definition

Step 1:

Define functions and parameters as follows in ParNMPC_Def.m and run ParNMPC_Def.m.

Simulation Loop:

Plant:

```
% Definition of dimension
pSimDim = 0; % dimension of the parameter in the
simulation model

% Definition of system dynamics f_sim(u,x,p_sim)
fSim([u;x;pSim]) =
        [x(2); a * x(1) + b * x(2) * u(1)];

% init pSim
pSimVal = zeros(pSimDim,1);
```

NMPC Controller:

```
% Definition of dimensions
xDim = 2; % dimension of states
uDim = 2; % dimension of inputs including dummy inputs
muDim = 1; % dimension of constraints
pDim = 0; % dimension of parameters for the controller
% Definition of system dynamics f(u,x,p)
a = -1;
b = -1;
f([u;x;p]) = [x(2); a * x(1) + b * x(2) * u(1)];
% Definition of equality constraint C(u,x,p)
uMax = 1;
uMin = -1;
uBar = (uMax + uMin)/2;
C([u;x;p]) = (u(1) - uBar)^2 + u(2)^2 - (uMax - uBar)^2;
% Definition of cost function L(u,x,p)
Q = diag([10, 10]);
R = diag([0.01, 0.01]);
r = [0, 0.1];
xRef = [0;0];
uRef = [0; uMax];
L([u;x;p]) = 0.5*(x-xRef).'*Q*(x-xRef)...
             +0.5*(u-uRef).'*R*(u-uRef)...
             -0.5*r*u;
% Definition of NMPC settings
x0Value = [1;0]; % init state
T = 2; % prediction horizon
N = 24; % number of discretization grids
pVal = zeros(pDim, N); % parameters
MaxIterNum = 5; % max num of iterations per update
tolerance = 1e-5; % tolerance of the KKT conditions to terminate
```

Semi-Active Damper – Initialization

Step 2:

Define the initialization method by initMethod = INIT_BY_FILE in ParNMPC_Init.m and run ParNMPC_Init.m, solution will be initialized from file initialSolution.mat.

- If initialSolution.mat is not available, another method should be selected by either initMethod = INIT_BY_INPUT or INIT_BY_REF_GUESS, and the corresponding parameters should be provided.
- The optimal solution will be automatically saved to initialSolution.mat.

Semi-Active Damper – Simulation & Code Generation

Step 3:

- 1. Run ParNMPC.m to validate the closed-loop simulation. The results will be saved to GEN_log_rec.mat.
 - This step is not in parallel. Only the generated C/C++ code is in parallel.
 - If the simulation doesn't go well, it is most likely that the first OCP is not well solved, or the measured state and time-dependent parameter change dramatically.
- 2. Code generation:
 - 1. Set up your compiler by *mex -setup*.
 - 2. Open the MATLAB Coder app.
 - 3. Select ParNMPC.m.
 - 4. Click Generate.

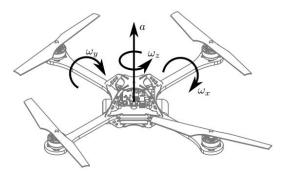


Semi-Active Damper – Deployment

Step 4:

- 1. Open Visual Studio.
- 2. Create an empty Win32 Console Application project.
- 3. Change to Release x64 mode.
- 4. Add *.h and *.c files in $\langle codegen \rangle lib \rangle ParNMPC$ to project.
- 5. Add *.h and *.c files in \codegen\lib\ParNMPC\examples to project.
- 6. Add the folder $\langle codegen \rangle lib \rangle ParNMPC$ to $Properties \rightarrow C/C++ \rightarrow General \rightarrow Additional Include Directories.$
- 7. Properties \rightarrow C/C++ \rightarrow Language \rightarrow Open MP Support: Yes (/openmp).
- 8. Compile and run.

Getting Started Quadrotor



Simulation Loop:

- Sampling interval: 0.01 s
- Simulation length: 20 s

Plant:

• Same as in NMPC:

 $\ddot{X} = \alpha(\cos\gamma\sin\beta\cos\alpha + \sin\gamma\sin\alpha)$

 $\ddot{Y} = a(\cos \gamma \sin \beta \sin \alpha - \sin \gamma \cos \alpha)$

 $\ddot{Z} = a \cos \gamma \cos \beta - g$

 $\dot{\gamma} = (\omega_X \cos \gamma + \omega_Y \sin \gamma) / \cos \beta$

 $\dot{\beta} = -\omega_X \sin \gamma + \omega_Y \cos \gamma$

 $\dot{\alpha} = \omega_X \cos \gamma \tan \beta + \omega_Y \sin \gamma \tan \beta + \omega_Z$

NMPC Controller:

States:

$$x = [X \ \dot{X} \ Y \ \dot{Y} \ Z \ \dot{Z} \ \gamma \ \beta \ \alpha]^T \in \mathbb{R}^9$$

• Dynamics:

 $\ddot{X} = a(\cos\gamma\sin\beta\cos\alpha + \sin\gamma\sin\alpha)$

 $\ddot{Y} = a(\cos\gamma\sin\beta\sin\alpha - \sin\gamma\cos\alpha)$

 $\ddot{Z} = a\cos\gamma\cos\beta - g$

 $\dot{\gamma} = (\omega_X \cos \gamma + \omega_Y \sin \gamma) / \cos \beta$

 $\dot{\beta} = -\omega_X \sin \gamma + \omega_Y \cos \gamma$

 $\dot{\alpha} = \omega_X \cos \gamma \tan \beta + \omega_Y \sin \gamma \tan \beta + \omega_Z$

• Constraints (Converted to equality constraints

by introducing dummy inputs u_5 , u_6 , u_7 , u_8):

$$\begin{bmatrix} 0 \\ -1 \\ -1 \\ -1 \\ -1 \end{bmatrix} \le \begin{bmatrix} a \\ \omega_X \\ \omega_Y \\ \omega_Z \end{bmatrix} \le \begin{bmatrix} 11 \\ 1 \\ 1 \\ 1 \end{bmatrix} \longrightarrow C(u, x, p) = \begin{bmatrix} (a - 5.5)^2 + u_5^2 - 5.5^2 \\ \omega_X^2 + u_6^2 - 1 \\ \omega_Y^2 + u_7^2 - 1 \\ \omega_Z^2 + u_8^2 - 1 \end{bmatrix}$$

• Stage cost function:

$$L(u, x, p) = \frac{1}{2} (\|x - x_{ref}\|_{Q}^{2} + \|u - u_{ref}\|_{R}^{2} - r \cdot u)$$

 $Q = diag(10,1,2,1,10,1,1,1,1), R = blkdiag(I_4, 0.01 \cdot I_4)$

 $r = [zeros(1,4), 0.1 \cdot ones(1,4)]$

 $u_{ref} = [g, 0, 0, 0, 3.4167, 1, 1, 1]^T$

 $x_{ref}(t) = zeros(9,1), t \in [0,10]$

 $x_{ref}(t) = [1.5,0,1.5,0,1.5,0,0,0,0]^T, t \in (10,20]$

- Initial state: $\bar{x}_0 = [1,0,1,0,1,0,0,0,0]^T$
- Prediction horizon: 1 s
- Number of discretization grids: N = 24
- Max number of iterations: 5
- Tolerance: 5e-3

Getting Started Quadrotor – Problem Definition

Step 1:

Define functions and parameters as follows in ParNMPC_Def.m and run ParNMPC_Def.m.

Simulation Loop:

```
Ts = 0.01; %sampling interval simuLength = 10; %simulation length in seconds
```

Plant:

NMPC Controller:

```
% Definition of dimensions
xDim = 9; % dimension of states
uDim = 8; % dimension of inputs including dummy inputs
muDim = 4; % dimension of constraints
pDim = xDim+uDim; % dimension of parameters for the controller
% Definition of system dynamics f(u,x,p)
q = 9.81;
f([u;x;p]) = [x(2);...
              u(1)*(cos(x(7))*sin(x(8))*cos(x(9)) + sin(x(7))*sin(x(9)));...
              u(1)*(cos(x(7))*sin(x(8))*sin(x(9)) - sin(x(7))*cos(x(9)));...
              u(1)*cos(x(7))*cos(x(8)) - q;...
             (u(2)*\cos(x(7)) + u(3)*\sin(x(7)))/\cos(x(8));...
             -u(2)*sin(x(7)) + u(3)*cos(x(7));...
             u(2)*cos(x(7))*tan(x(8)) + u(3)*sin(x(7))*tan(x(8)) + u(4)];
% Definition of equality constraint C(u,x,p)
uMax = [11;1;1;1];
uMin = [0;-1;-1;-1];
uBar = (uMax + uMin)/2;
C([u;x;p]) = [(u(1) - uBar(1))^2 + u(5)^2 - (uMax(1)-uBar(1))^2;...
              (u(2) - uBar(2))^2 + u(6)^2 - (uMax(2) - uBar(2))^2;...
              (u(3) - uBar(3))^2 + u(7)^2 - (uMax(3)-uBar(3))^2;...
              (u(4) - uBar(4))^2 + u(8)^2 - (uMax(4) - uBar(4))^2;
% Definition of cost function L(u,x,p)
Q = diag([10, 1, 2, 1, 10, 1, 1, 1, 1]);
R = diag([1, 1, 1, 1, ...])
         0.01, 0.01, 0.01, 0.01 ]);
r = [0,0,0,0,...
    1,1,1,1]*0.1;
L([u;x;p]) = 0.5*(x-p(1:xDim)).'*Q*(x-p(1:xDim))...
             +0.5*(u-p(pDim-uDim+1:pDim)).'*R*(u-p(pDim-uDim+1:pDim))...
             -0.5*r*u;
% Definition of NMPC settings
x0Value = [1;0;1;0;1;0;0;0;0]; % initial state
T = 1.0; % prediction horizon
N = 24; % number of discretization grids
xRefConstant = [0;0;0;0;0;0;...
               0:0:01:
uRefConstant = [q;0;0;0;...
               sqrt(-(g - uBar(1))^2 + (uMax(1)-uBar(1))^2);...
                uMax(2);uMax(3);uMax(4)];
pVal = repmat([xRefConstant;uRefConstant],1,N); % init parameters
MaxIterNum = 5; % max num of iterations per update
tolerance = 5e-3; % tolerance of the KKT conditions to terminate
```

Getting Started Quadrotor – Initialization

• Step 2: refer to <u>Semi-Active Damper – Initialization</u>

Quadrotor-Simulation & Code Generation

• Step 3: refer to <u>Semi-Active Damper – Simulation & Code Generation</u>

Quadrotor—Deployment

• Step 4: refer to <u>Semi-Active Damper – Deployment</u>

Main Reasons Why *ParNMPC* Fails

- 1. The first OCP is not well solved. *ParNMPC* utilizes the warm-start strategy and highly depends on the solution of the first OCP. *ParNMPC* may fail even when the norm of the KKT conditions (first-order optimality) for the first OCP is small enough. In general, a good solution is smooth enough. In order to find a good solution to the first OCP, the method of continuation can be employed. For example, the solution of the first OCP can be initialized from the solution of an easy-to-solve problem, which has less active constraints or a shorter prediction horizon.
- 2. Dramatic changes of the measured state or time-dependent parameter. *ParNMPC* converges locally and requires that the current solution is close to the last solution. The current version of *ParNMPC* doesn't contain any globalization techniques such as line-search and trust-region methods. Make sure that the state and parameter don't vary too much with respect to time.
- 3. Ill-conditioned problem. It is usually cased by the settings of the weighting matrices.

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