

Microsecond timescale (N)MPC for Power-Electronics applications

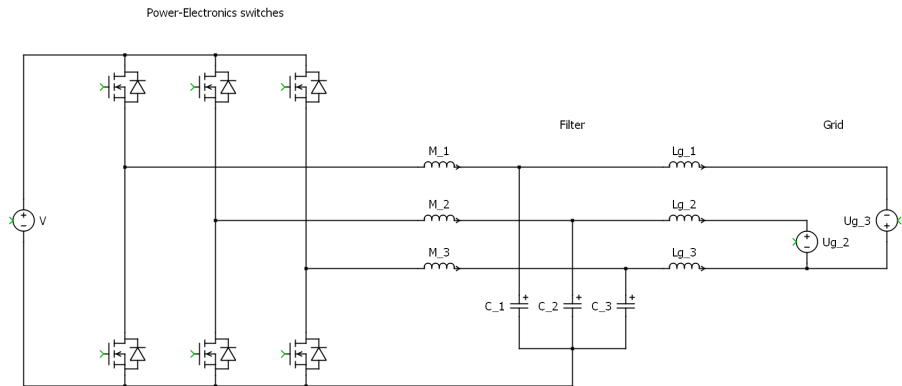
Benjamin Stickan

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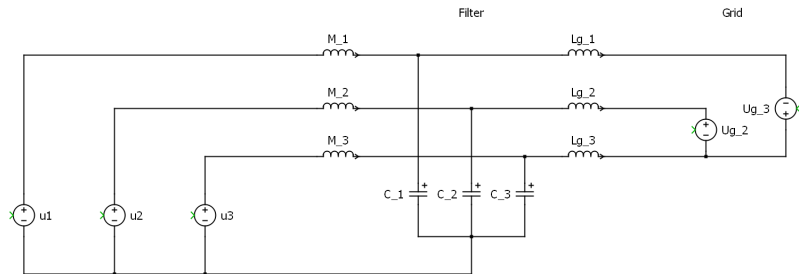
- 1 Recent work at Fraunhofer
- 2 C/GMRES

- Current project
 - EU project Netfficient
 - Local energy storage on the island of Borkum
 - Aim: support the grid
 - Reduce energy exchange between island and mainland
 - Storage System with 1 MW inverter (8x 125 kW)
- Recent work
 - Extension of inverter model to 3-phases
 - Development of EMPC-scheme for 125 kW inverter

Current System (1/3)



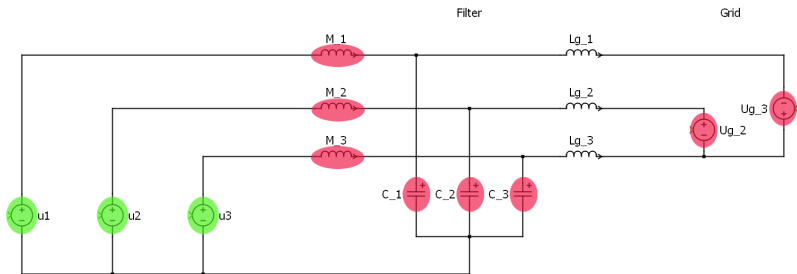
Current System (2/3)



Current System(3/3)

Measurements: Main inductor current, Capacitor Voltage, Grid

Controls: Average input voltages u_1 - u_3



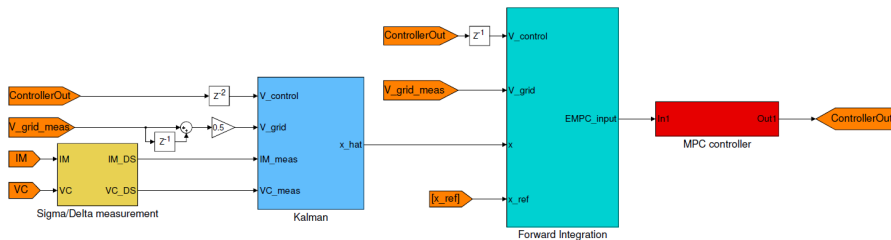
Control Scheme

Control loop frequency: 40 kHz (25 μ s)

System states: 8

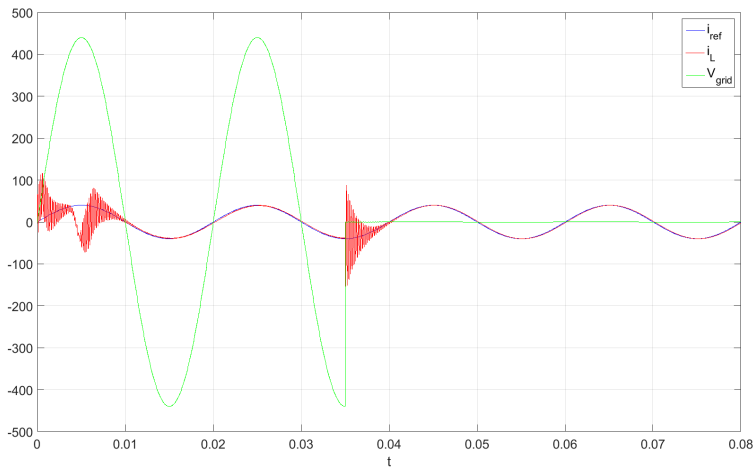
Estimator states: 20

Control variables: 3



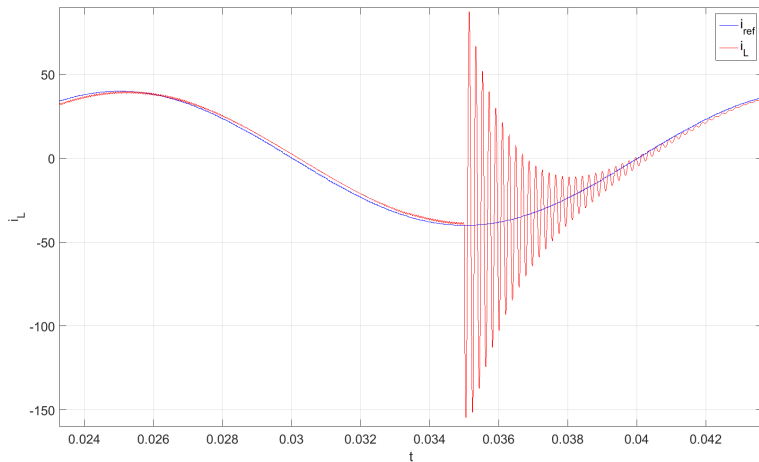
Simulation results (1/3)

- Controller: EMPC, $H=3$, move blocking over whole horizon
- Decision variables: 3 (only one phase shown here)



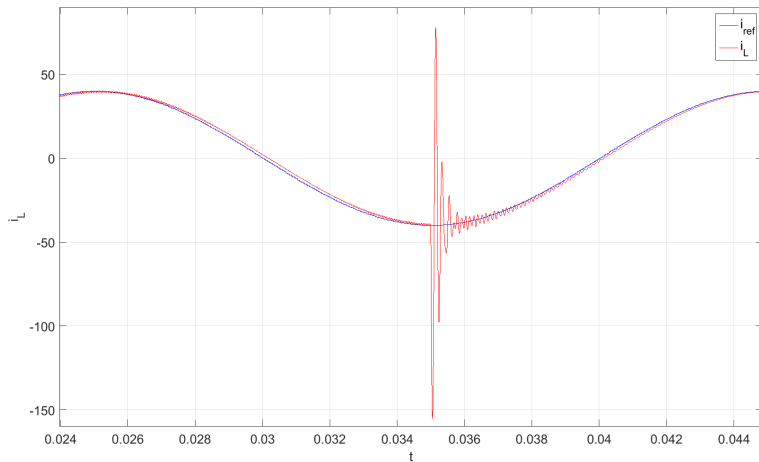
Simulation results (2/3)

- Controller: EMPC, $H=3$, move blocking over whole horizon
- Decision variables: 3



Simulation results (3/3)

- Controller: qpOASES, $H=5$
- Decision variables: 15



- How to solve problems fast and **reliable**?
- Are FPGAs an option?
 - Which methods would be suited?
 - How large is the effort?

- Developed by Toshiyuki Ohtsuka
- 'Indirect' approach
- Based on Pontryagin's maximum (minimum) principle

OCP formulation

Consider general nonlinear system equations:

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

$x(t) \in \mathbb{R}^n$: state vector

$u(t) \in \mathbb{R}^{m_u}$: input vector

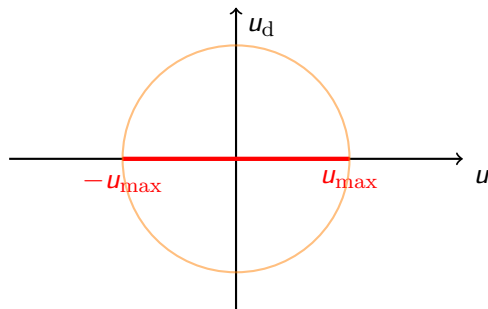
$p(t) \in \mathbb{R}^{m_p}$: vector of given time dependent variables.

Inequality Constraints

How deal with inequality constraints $|u| \leq u_{\max}$?

Employ dummy input u_d :

$$C := (u^2 + u_d^2 - u_{\max}^2) = 0$$



First Order Optimality Conditions

$$x_\tau^* = f(x^*, u^*, p),$$

$$x^*(0, t) = x(t),$$

$$\lambda_\tau^* = -H_x^T(x^*, \lambda^*, u^*, \mu^*, p),$$

$$\lambda^*(T, t) = \varphi_x^T(x^*(T, t), p(t+T)),$$

$$H_u(x^*, \lambda^*, u^*, \mu^*, p) = 0,$$

$$C(x^*, u^*, p) = 0,$$

with Hamiltonian

$$H(x^*, \lambda^*, u^*, \mu^*, p) := L(x^*, u^*, p) + \lambda^T f(x^*, u^*, p) + \mu^T C(x^*, u^*, p),$$

$\lambda^* \in \mathbb{R}^n$: costate vector

$\mu^* \in \mathbb{R}^{m_c}$: Lagrange multiplier associated with the equality constraints

Discretized Formulation (1/3)

$$x_{i+1}^*(t) = x_i^*(t) + f(x_i^*(t), u_i^*(t), p_i^*(t))\Delta\tau(t),$$

$$x_0^*(t) = x(t),$$

$$\lambda_i^*(t) = \lambda_{i+1}^*(t) + H_x^T(x_i^*(t), \lambda_{i+1}^*(t), u_i^*(t), \mu_i^*(t), p_i^*(t))\Delta\tau(t),$$

$$\lambda_N^*(t) = \varphi_x^T(x_N^*(t), p_i^*(t)),$$

$$H_u(x_i^*(t), \lambda_{i+1}^*(t), u_i^*(t), \mu_i^*(t), p_i^*(t)) = 0,$$

$$C(x_i^*(t), u_i^*(t), p_i^*(t)) = 0,$$

$$\Delta\tau(t) := T(t)/N$$

$$x_i(t)^* := x^*(i\Delta\tau, t)$$

$$p_i^*(t) := p(t + i\Delta\tau)$$

Discretized Formulation (2/3)

$$x_{i+1}^*(t) = x_i^*(t) + f(x_i^*(t), u_i^*(t), p_i^*(t))\Delta\tau(t),$$

$$x_0^*(t) = x(t),$$

$$\lambda_i^*(t) = \lambda_{i+1}^*(t) + H_x^T(x_i^*(t), \lambda_{i+1}^*(t), u_i^*(t), \mu_i^*(t), p_i^*(t))\Delta\tau(t),$$

$$\lambda_N^*(t) = \varphi_x^T(x_N^*(t), p_i^*(t)),$$

$$H_u(x_i^*(t), \lambda_{i+1}^*(t), u_i^*(t), \mu_i^*(t), p_i^*(t)) = 0,$$

$$C(x_i^*(t), u_i^*(t), p_i^*(t)) = 0,$$

Discretized Formulation (3/3)

Stack optimization variables in single vector:

$$U(t) := [u_0^{*\top}(t), \mu_0^{*\top}(t), \dots, u_{N-1}^{*\top}(t), \mu_{N-1}^{*\top}(t)]^\top \in \mathbb{R}^m,$$
$$m := m_u + m_c,$$

Solve:

$$F(U(t), x(t), t) := \begin{bmatrix} H_u^\top(x_0^*, \lambda_1^*, u_0^*, \mu_0^*, p(t)) \\ C(x_0^*, u_0^*, p(t)) \\ \vdots \\ H_u^\top(x_{N-1}^*, \lambda_N^*, u_{N-1}^*, \mu_{N-1}^*, p(t + (N-1)\Delta\tau)) \\ C(x_{N-1}^*, u_{N-1}^*, p(t + (N-1)\Delta\tau)) \end{bmatrix} = 0.$$

$$F(U(t), x(t), t) = 0$$

must be fulfilled at any time t .

$$\Rightarrow \dot{F}(U(t), x(t), t) = 0$$

$$\Leftrightarrow \dot{U} = F_U^{-1}(-F_x \dot{x} - F_t)$$

Find initial solution for $U(0)$ and then integrate over $\dot{U}(t)$.

Drawback: inverse and jacobians needed.

Forward Difference (1/3)

Approximate approach: employ forward difference approximation for products of Jacobians and vectors.

$$D_h F(U, x, t : W, w, \omega) := \frac{F(U + hW, x + hw, t + h\omega) - F(U, x, t)}{h}$$

Approximate

$$\dot{F} = F_U \dot{U} - F_x \dot{x} - F_t$$

by

$$D_h F(U, x, t : \dot{U}, \dot{x}, 1) = A_s F(U, x, t)$$

(A_s is a stable Matrix that stabilizes $F = 0$ to suppress numerical errors that may accumulate through numerical integration)

Forward Difference (3/3)

Result: linear equation of the form $Ax = b$:

$$D_h F(U, x + h\dot{x}, t + h : \dot{U}, 0, 0) = b(U, x, \dot{x}, t),$$

with $b(U, x, \dot{x}, t) := A_s F(U, x, t) - D_h F(U, x, t : 0, \dot{x}, 1)$.

The solution \dot{U} can then be integrated to obtain the next optimal control vector U

- How to solve linear equations fast?
 - → Ohtsuka uses Krylov subspace based iterative GMRES method
 - Theoretically direct solver (for infinite precision)
 - Matrices and vectors grow with every iteration
 - Can be restarted

- Similarities between Ohtsuka's discretization method and direct methods?
- How 'exact' is the continuation method? Applicable to Power-Electronics?
- Can the GMRES algorithm generally be implemented on an FPGA?
 - Alternative: other iterative solvers?

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