



Centre for
Process
Systems
Engineering

Imperial College London

Multi-Parametric Programming & Explicit Model Predictive Control

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Centre for Process Systems Engineering

Imperial College London

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Multi-Parametric Programming & Model Predictive Control – *are they meant for each other?*

Multi-Parametric Programming & Model Predictive Control – *are they meant for each other?*

YES!
**in many cases (but not
for all)**

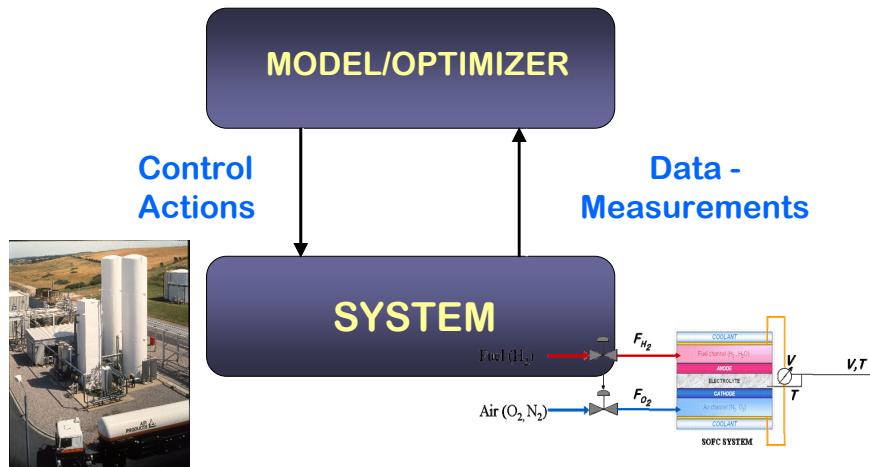
Outline

- Key concepts & historical overview
- Multi-parametric Programming Theory
 - Recent Developments & future directions
- Explicit Model Predictive Control Theory
 - Recent Developments & future directions
- A framework for multi-parametric programming & MPC
- MPC-on-a-Chip Applications
 - Recent Developments & future directions

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What is On-line Optimization?



What is Multi-parametric Programming?

- Given:

- a performance criterion to minimize/maximize
- a vector of constraints
- a vector of parameters

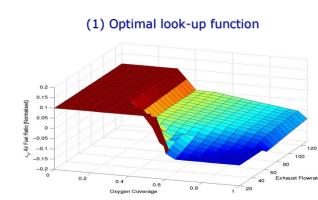
$$\begin{aligned}
 z(x) &= \min_u f(u, x) \\
 \text{s.t. } g(u, x) &\leq 0 \\
 x &\in \mathbb{R}^n \\
 u &\in \mathbb{R}^s
 \end{aligned}$$

What is Multi-parametric Programming?

- Given:
 - a performance criterion to minimize/maximize
 - a vector of constraints
 - a vector of parameters
 - Obtain:
 - the performance criterion and the optimisation variables as a function of the parameters
 - the regions in the space of parameters where these functions remain valid

Multi-parametric programming

$$\begin{aligned} z(x) &= \min_u f(u, x) \\ \text{s.t. } g(u, x) &\leq 0 \\ x &\in \mathbb{R}^n \\ u &\in \mathbb{R}^s \end{aligned}$$



Obtain optimal solution $u(x)$ as a function of the parameters x

Multi-parametric programming

Problem Formulation

$$\min_{\mathbf{u}_1, \mathbf{u}_2} (-3\mathbf{u}_1 - 8\mathbf{u}_2)$$

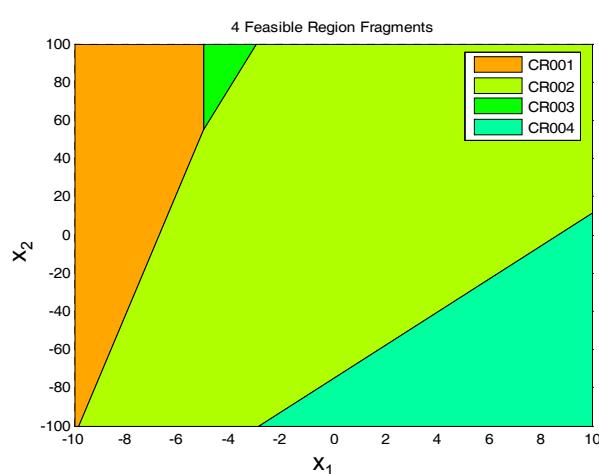
st.

$$\begin{bmatrix} 1 & 1 \\ 5 & -4 \\ -8 & 22 \\ -4 & -1 \end{bmatrix} \cdot \begin{bmatrix} \mathbf{u}_1 \\ \mathbf{u}_2 \end{bmatrix} + \begin{bmatrix} -1 & 0 \\ 0 & 0 \\ 0 & -1 \\ 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \end{bmatrix} + \begin{bmatrix} -13 \\ -20 \\ -121 \\ 8 \end{bmatrix} \leq \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$-10 \leq \mathbf{x}_1 \leq 10 \quad -100 \leq \mathbf{x}_2 \leq 100$$

Multi-parametric programming

Critical Regions



Multi-parametric programming

Multi-parametric Solution

$$U = \begin{cases} \left[\begin{array}{cc} -0.33 & 0 \\ 1.33 & 0 \end{array} \right] \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} -1.67 \\ 14.67 \end{bmatrix} & \text{if } \begin{bmatrix} 1 & -0.031 \\ 1 & 0 \\ -1 & 0 \\ 0 & -1 \\ 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} -6.71 \\ -5 \\ 10 \\ 100 \\ 100 \end{bmatrix} \\ \dots \\ \left[\begin{array}{cc} 0.73 & -0.03 \\ 0.26 & 0.03 \end{array} \right] \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 5.5 \\ 7.5 \end{bmatrix} & \text{if } \begin{bmatrix} 1 & -0.115 \\ -1 & 0.031 \\ -1 & 0.045 \\ 1 & 0 \\ 0 & -1 \\ 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} 8.65 \\ 6.71 \\ 7.5 \\ 10 \\ 100 \\ 100 \end{bmatrix} \\ \dots \\ \left[\begin{array}{cc} 0 & 0 \\ 1 & 0 \end{array} \right] \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 13 \end{bmatrix} & \text{if } \begin{bmatrix} 1 & -0.045 \\ -1 & 0 \\ 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} -7.5 \\ 5 \\ 100 \end{bmatrix} \\ \dots \\ \left[\begin{array}{cc} 0 & 0.05 \\ 0 & 0.06 \end{array} \right] \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 11.8 \\ 9.8 \end{bmatrix} & \text{if } \begin{bmatrix} -1 & 0.11 \\ 1 & 0 \\ 0 & -1 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} -8.65 \\ 10 \\ 100 \end{bmatrix} \end{cases}$$

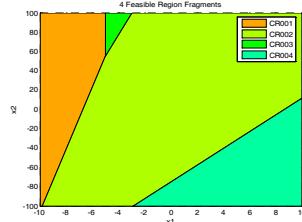
Multi-parametric programming

$$\min_u (-3u_1 - 8u_2)$$

s.t.

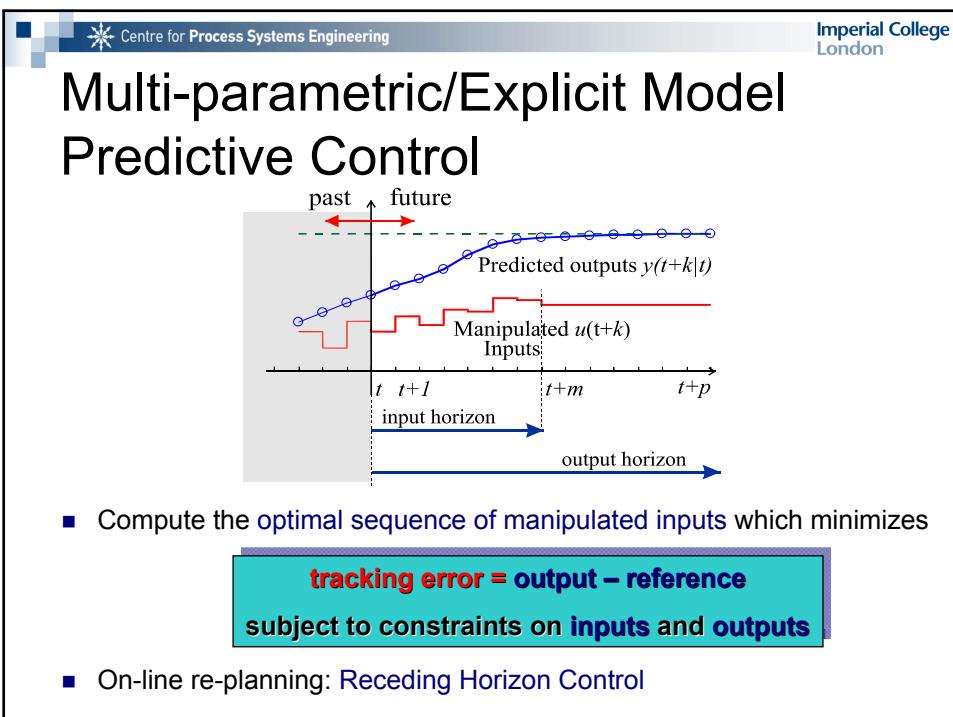
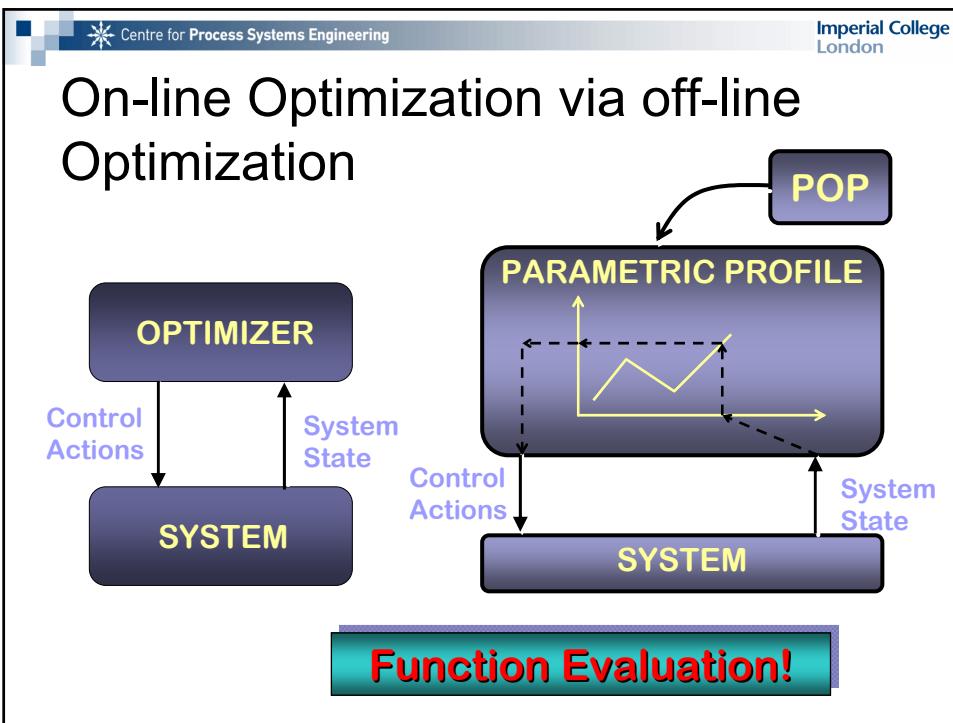
$$\begin{bmatrix} 1 & 1 \\ 5 & -4 \\ -8 & 22 \\ -4 & -1 \end{bmatrix} \cdot \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} + \begin{bmatrix} -1 & 0 \\ 0 & 0 \\ 0 & -1 \\ 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} -13 \\ -20 \\ -121 \\ 8 \end{bmatrix} \leq \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$-10 \leq x_1 \leq 10, -100 \leq x_2 \leq 100$$

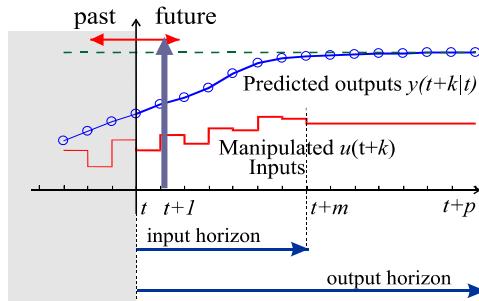


$$U = \begin{cases} \left[\begin{array}{cc} -0.333 & 0 \\ 1.333 & 0 \end{array} \right] \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} -1.6667 \\ 14.6667 \end{bmatrix} & \text{if } \begin{bmatrix} 1 & -0.03125 \\ 1 & 0 \\ -1 & 0 \\ 0 & -1 \\ 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} -6.71875 \\ -5 \\ 10 \\ 100 \\ 100 \end{bmatrix} \\ \dots \\ \left[\begin{array}{cc} 0.7333 & -0.0333 \\ 0.26667 & 0.0333 \end{array} \right] \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 5.5 \\ 7.5 \end{bmatrix} & \text{if } \begin{bmatrix} 1 & -0.115385 \\ -1 & 0.03125 \\ -1 & 0.0454545 \\ 1 & 0 \\ 0 & -1 \\ 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} 8.65938 \\ 6.71875 \\ 7.5 \\ 10 \\ 100 \\ 100 \end{bmatrix} \\ \dots \\ \left[\begin{array}{cc} 0 & 0 \\ 1 & 0 \end{array} \right] \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 13 \end{bmatrix} & \text{if } \begin{bmatrix} 1 & -0.0454545 \\ -1 & 0 \\ 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} 5 \\ 100 \end{bmatrix} \\ \dots \\ \left[\begin{array}{cc} 0 & 0.05128 \\ 0 & 0.0641 \end{array} \right] \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 11.8462 \\ 9.80769 \end{bmatrix} & \text{if } \begin{bmatrix} 1 & -0.115385 \\ 1 & 0 \\ 0 & -1 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} -8.65385 \\ 10 \\ 100 \end{bmatrix} \end{cases}$$

Only 4 optimization problems solved!



Multi-parametric/Explicit Model Predictive Control



- Compute the optimal sequence of manipulated inputs which minimizes

Solve a QP at each time interval
- On-line re-planning: Receding Horizon Control

Multi-parametric Programming Approach

- State variables → Parameters
- Control variables → Optimization variables
- MPC → Multi-Parametric Programming problem
- Control variables → F(State variables)

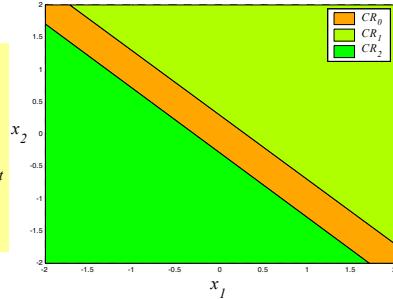
Multi-parametric Quadratic Program

Explicit Control Law

$$J(x(t)) = \min_{u_{t|t}, u_{t+1|t}} \sum_{j=0}^1 \left\{ \mathbf{x}_{t+j|t}^T \mathbf{x}_{t+j|t} + 0.01 \mathbf{u}_{t+j|t}^2 \right\} + \mathbf{x}_{t+2|t}^T P \mathbf{x}_{t+2|t}$$

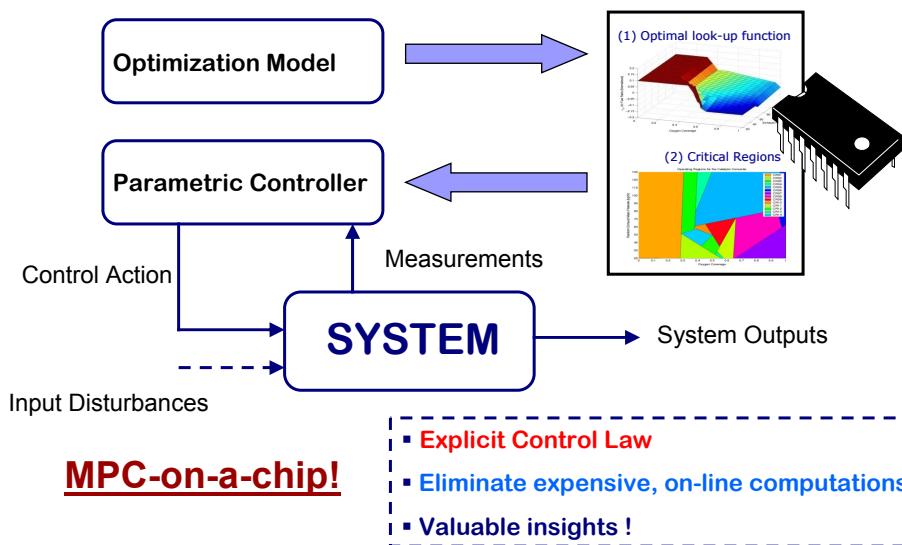
s.t. $\mathbf{x}_{t+j+1|t} = \begin{bmatrix} 0.7326 & -0.0861 \\ 0.1722 & 0.9909 \end{bmatrix} \mathbf{x}_{t+j|t} + \begin{bmatrix} 0.0609 \\ 0.0064 \end{bmatrix} \mathbf{u}_{t+j|t}$

$-2 \leq \mathbf{u}_{t+j|t} \leq 2 \quad j=1,2 \quad \mathbf{x}_{t|t} = \mathbf{x}(t)$



$$\mathbf{u}(t) = \begin{cases} [-6.8355 \ -6.8585] \mathbf{x}(t) & \text{if } \begin{bmatrix} 0.7059 & 0.7083 \\ -0.7059 & -0.7083 \end{bmatrix} \mathbf{x}(t) \leq \begin{bmatrix} 0.2065 \\ 0.2065 \end{bmatrix} \\ -2 & \text{if } [-0.7059 \ -0.7083] \mathbf{x}(t) \leq -0.2065 \\ 2 & \text{if } [0.7059 \ 0.7083] \mathbf{x}(t) \leq -0.2065 \end{cases}$$

Multi-parametric Controllers



Key milestones-Historical Overview

AIChE J., Perspective (2009)

- ❑ Number of publications

	Multi-Parametric Programming	Multi-Parametric MPC & applications
Pre-1999	>100	0
Post-1999	~50	~120

- ❑ 2002 Automatica paper ~ 420 citations
- ❑ Multi-parametric programming – until 1992 mostly analysis & linear models
- ❑ Multi-parametric/explicit MPC – post-2000 much wider attention

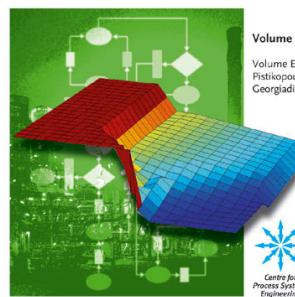
Process Systems Engineering

Efstratios N. Pistikopoulos,
Michael C. Georgiadis, Vivek Dua (Eds.)

WILEY-VCH

Multi-Parametric Programming

Theory, Algorithms and Applications



Volume 1

Volume Editors:
Pistikopoulos,
Georgiadis, Dua

Process Systems Engineering

Efstratios N. Pistikopoulos,
Michael C. Georgiadis, Vivek Dua (Eds.)

WILEY-VCH

Multi-Parametric Model-Based Control

Theory and Applications



Volume 2

Volume Editors:
Pistikopoulos,
Georgiadis, Dua

Patented Technology

- Improved Process Control, European Patent No EP1399784, 2004
- Process Control Using Co-ordinate Space, United States Patent No US7433743, 2008



**Multi-Parametric
Programming & Model
Predictive Control – *are
they meant for each other?***

YES!

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- Key concepts & historical overview
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 - Recent Developments & future directions

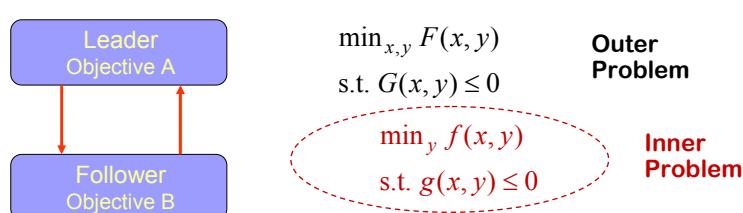
Multi-parametric Programming Theory

mp-LP	Gass & Saaty [1954], Gal & Nedoma [1972], Propoi [1975], Adler and Monterio [1992], Gal [1995], Acevedo and Pistikopoulos[1997], Dua et al [2002], Pistikopoulos et al [2007]
mp-QP	Townsley [1972], Propoi [1978], Best [1995], Dua et al [2002], Pistikopoulos et al [2002,2007]
mp-NLP	Fiacco [1976], Kojima [1979], Bank et al [1983], Fiacco [1983], Fiacco & Kyoarisis [1986], Acevedo & Pistikopoulos [1996], Dua and Pistikopoulos [1998], Pistikopoulos et al [2007]
mp-DO	Sakizlis et al.[2002], Bansal [2003], Sakizlis et al [2005], Pistikopoulos et al [2007]
mp-GO	Fiacco [1990], Dua et al [1999,2004], Pistikopoulos et al [2007]
mp-MILP	Marsten & Morin [1975], Geoffrion & Nauss [1977], Joseph [1995], Acevedo & Pistikopoulos [1997,1999], Dua & Pistikopoulos[2000]
mp-MINLP	McBride & Yorkmark [1980], Chern [1991], Dua & Pistikopoulos [1999], Hene et al [2002], Dua et al [2002]

Multi-parametric Programming Theory – Recent Developments

- Multi-parametric NLP (mp-NLP)
- Dynamic Programming
- Global Optimization
 - Bilevel/Multilevel Programming
 - Multi-parametric MILP

Bilevel/Multilevel Programming



- Even linear case is non-convex
- Multi-parametric Programming approach
 - Follower's problem - Multi-parametric Programming Problem
 - Leader's problem - single optimization problems
- Methodology applies to Multi-level problems
- Applications in Hierarchical Decision Making/Control

Multi-parametric Programming Theory – Future Directions

- Dynamic Systems
 - Multi-parametric Dynamic & Mixed Integer Optimization (mp-DO, mp-MIDO)
 - Valid bounds on number of control laws (Sakizlis et al, 2006) – approximations
- Global Optimization
 - Exploit model structure
 - General classes of nonlinear, mixed integer & dynamic systems
- Revisit fundamentals/basics
 - In search of a new ‘breakthrough’
 - Avoid active set strategy

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- A framework for multi-parametric programming & MPC

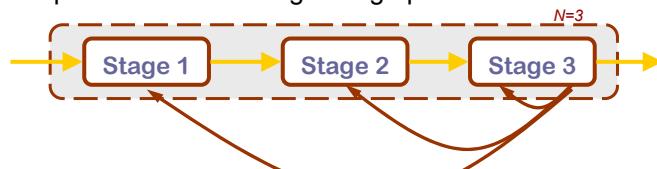
Multi-parametric/Explicit Model Predictive Control Theory

mp-MPC	Pistikopoulos [1997, 2000], Bemporad, Morari, Dua & Pistikopoulos [2000], Sakizlis & Pistikopoulos [2001], Tondel et al [2001], Pistikopoulos et al [2002], Bemporad et al [2002], Johansen and Grancharova [2003], Sakizlis et al [2003], Pistikopoulos et al [2007]
mp-Continuous MPC	Sakizlis et al.[2002], Kojima & Morari[2004], Sakizlis et al [2005], Pistikopoulos et al [2007]
Hybrid mp-MPC	Bemporad et al [2000], Sakizlis & Pistikopoulos [2001], Pistikopoulos et al [2007]
Robust mp-MPC	Kakalis & Pistikopoulos [2001], Bemporad et al [2001], Sakizlis et al [2002], Sakizlis & Pistikopoulos [2002], Sakizlis et al [2004], Olaru et al [2005], Faisca et al [2008]
mp-DP	Nunoz de la Pena et al [2004],Pistikopoulos et al [2007],Faisca et al [2008]
mp-NMPC	Johansen [2002], Bemporad [2003], Sakizlis et al [2007], Dobre et al [2007], Narciso & Pistikopoulos [2009]

Multi-parametric/Explicit Model Predictive Control Theory – Recent Developments

■ Robust Explicit MPC (Faísca et al, 2007, Pistikopoulos et al, 2009)

- MPC problem posed in a dynamic programming framework - decoupled into smaller single-stage problem



- At each stage reduced (Robust) mp-QP is solved
 - Consider control, state and constraints only at current stage
 - Immunize constraints against uncertainty (Robustification)

Multi-parametric/Explicit Model Predictive Control Theory – Recent Developments

■ Robust Explicit MPC Example

$$x_{t+j|t} = Ax_{t+j|t} + Bu_{t+j|t}$$

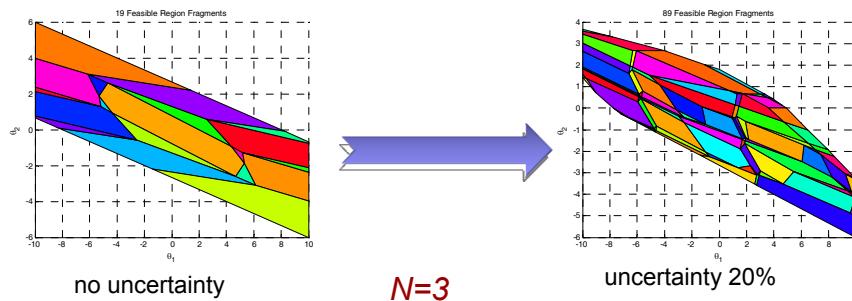
$$u_{\min} \leq u_{t+j|t} \leq u_{\max}$$

$$x_{\min} \leq x_{t+j|t} \leq x_{\max}$$

$$A = [a_y] \in \mathbb{R}^{n \times n}, B = [b_y] \in \mathbb{R}^{n \times m}$$

$$a_y \in \{a_y : |a_y - \bar{a}_y| \leq \varepsilon \bar{a}_y\}$$

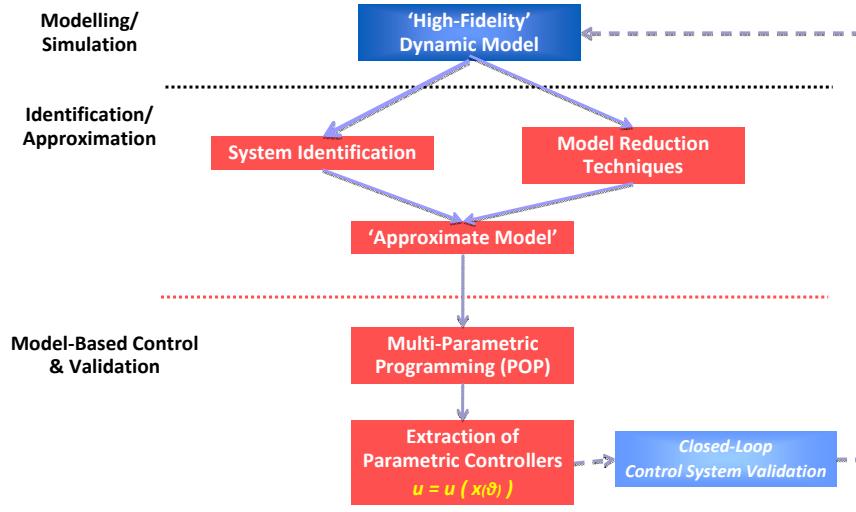
$$b_y \in \{b_y : |b_y - \bar{b}_y| \leq \varepsilon \bar{b}_y\}$$



Multi-parametric/Explicit Model Predictive Control Theory – Future Directions

- Robust & Hybrid Explicit Control
 - Hybrid Systems - with Nonlinear, Continuous-time Dynamics (mp-DO, mp-MIDO)
 - Robust Stability of mp-(N)MPC
- Model Reduction/Approximations
 - Nonlinear model order reduction methods with mp-NMPC
 - Identification methods (& uncertainty modelling)
- Estimation
- Explicit Nonlinear MPC (mp-NMPC)
 - Linked to Multi-parametric Nonlinear Programming (mp-NLP)
 - Constrained Dynamic Programming - nonlinear systems
 - Continuous-time Dynamic models (mp-DO, mp-MIDO)

A framework for multi-parametric programming & MPC (*Pistikopoulos 2008, 2009*)



**Multi-Parametric
Programming & Model
Predictive Control – *are
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**YES!
But a lot of work
remains to be done!**

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MPC-on-a-chip Applications – Recent Developments

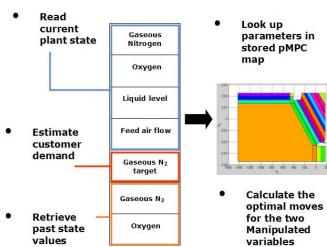
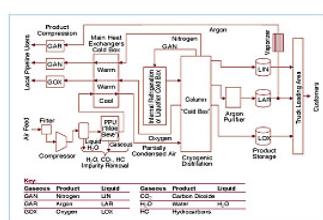
- Process Control
 - Air Separation (Air Products)
 - Hybrid PSA/Membrane Hydrogen Separation (EU/HY2SEPS, KAUST)
- Automotive
 - Active Valve Train Control (Lotus Engineering)
- Energy Systems
 - Hydrogen Storage (EU/DIAMANTE)
 - Fuel Cell

MPC-on-a-chip Applications – Recent Developments

- Biomedical Systems (MOBILE - ERC Advanced Grant Award)
 - Drug/Insulin, Anaesthesia and Chemotherapeutic Agents Delivery Systems
- Imperial Racing Green
 - Fuel cell powered Student Formula Car
- Aeronautics (EPSRC)
 - (Multiple) Unmanned Air Vehicles – with Cranfield University

Small Air Separation Units

(Air Products, Mandler et al, 2006)



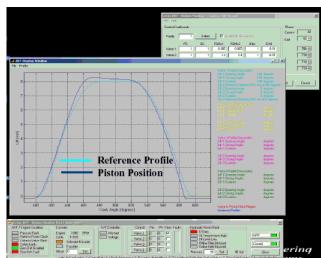
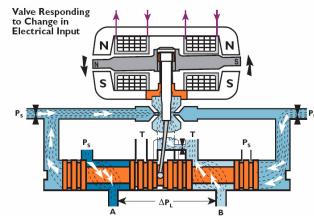
Implementation of look-up operation of pMPC for a small Nitrogen generator.

- Enable advanced MPC for small separation units
 - Optimize performance
 - Minimize operating costs
 - Satisfy product and equipment constraints
- Parametric MPC ideally suited
 - Supervises existing regulatory control
 - Off-line solution with minimum on-line load
 - Runs on existing PLC
 - Rapid installation compared to traditional MPC
- Advantages of Parametric MPC
 - 5% increased throughput
 - 5% less energy usage
 - 90% less waste
 - Installation on PLC in 1-day



Active Valve Train Control

(*Lotus Engineering, Kosmidis et al, 2006*)



(b) 8mm trapezoidal profile

- Active Valve Trains (AVT):

- Optimum combustion efficiency, Reduced Emissions, Elimination of butterfly valve, Cylinder deactivation, Controlled auto-ignition (CAI), Quieter operation

- Basic idea:

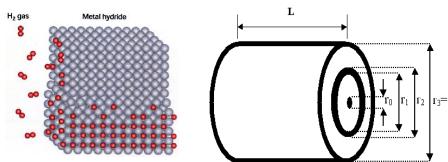
- Control System sends signal to valve
- This actuates piston attached to engine valve
- Enables optimal control of valve timing over entire engine rpm range

Challenges for the AVT control

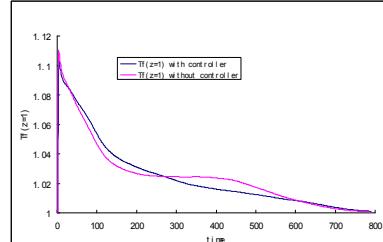
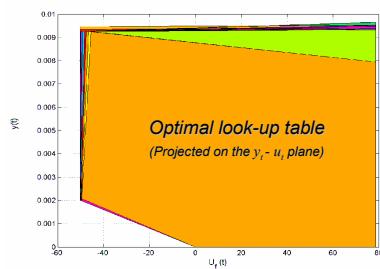
- **Nonlinear system dynamics:** Saturation, flow non-linearity, variation in fluid properties, non-linear opening of the orifices
- **Robustness** to various valve lift profiles
- **Fast dynamics** and sampling times (0.1ms)

Multi-parametric Control of H₂ Storage in Metal-Hydride Beds

(*EU-DIAMANTE, Georgiadis et al, 2008*)



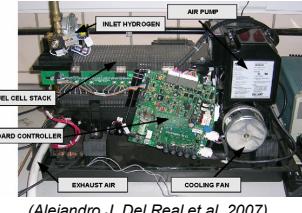
- Tracking the optimal temperature profile
- Ensure economic storage – expressed by the total required storage time
- Satisfy temperature and pressure constraints





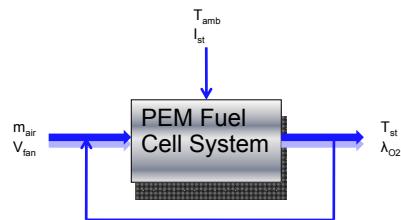
Fuel Cell Systems

- On-going implementation on 1.2kW Ballard PEM Fuel Cell
- Objective is to control oxygen excess ratio and stack temperature
- Controller is implemented on a DSP(digital Signal Processing) board
- Collaborative work with Fuel Cell Control Lab (**University of Seville**)

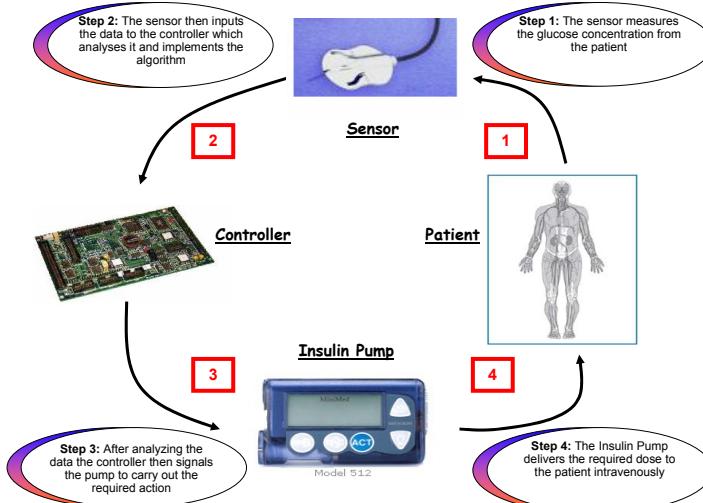


(Alejandro J. Del Real et al. 2007)

u: m_{air}, T_{st}
d: T_{amb}, I_{st}
y: T_{st}, V_{st}
θ: $x_t, T_{amb}, I_{st}, T_{st}, V_{st}, T_{st,sp}, V_{st,sp}$



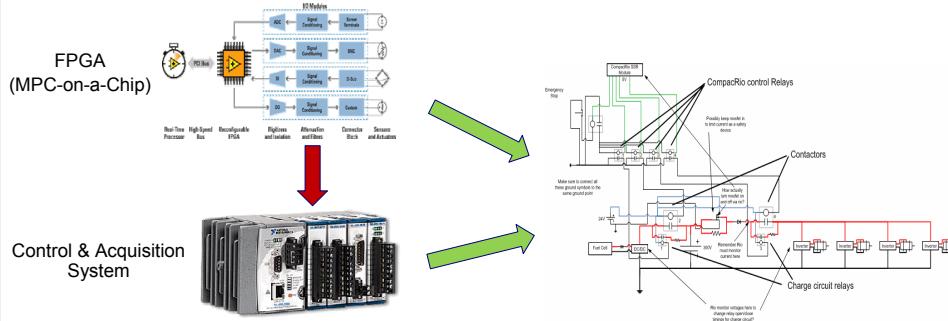
Biomedical Systems (MOBILE ERC Advanced Grant)



Imperial Racing Green Car



- Student Formula Project
 - Control of Start-up/Shut-down of the FC
 - Traction Motion Control



Multi-Parametric Programming & Model Predictive Control – *are they meant for each other?*

YES!

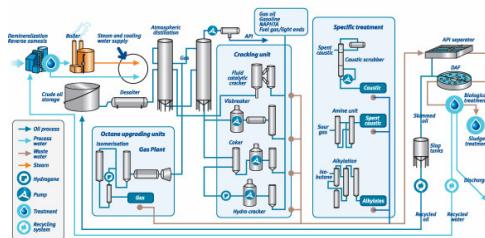
Success stories – generality?

MPC-on-a-chip – Future Directions

- Application types for Multi-parametric Programming & MPC
 - **Type 1** - Large scale and expensive industrial processes with slow/medium dynamics
 - **Type 2** - Medium scale and cost industrial processes with medium/fast dynamics
 - **Type 3** - Small scale and inexpensive processes/equipment with medium/fast dynamics

MPC-on-a-chip – Future Directions

- **Type 1** – Large scale and expensive industrial processes with slow/medium dynamics

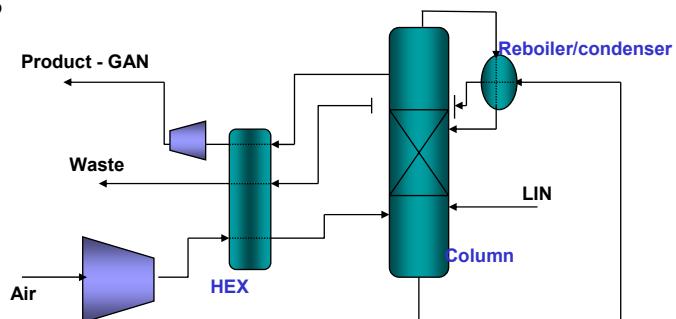


MPC-on-a-chip – Future Directions

- **Type 1** - Large scale and expensive industrial processes with slow/medium dynamics
 - Control hardware/software availability
 - MPC implementation mainly via online optimization
 - Explicit MPC can play a role for low level process control
 - Hybrid (on-line + off-line) approach possible

MPC-on-a-chip – Future Directions

- **Type 2** – medium scale and cost industrial processes with medium/fast dynamics

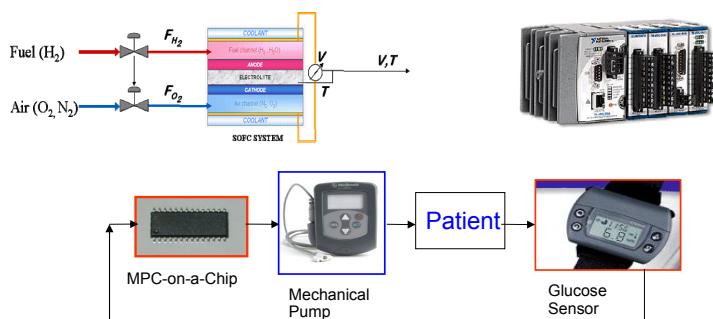


MPC-on-a-chip – Future Directions

- **Type 2** – medium scale and cost industrial processes with medium/fast dynamics
 - Limited Control hardware/software availability
 - Online optimization/MPC usually prohibitive
 - Multi-parametric MPC **ideal** – proved in previous applications (Air Separation, Automotive)

MPC-on-a-chip – Future Directions

- **Type 3** – small scale and inexpensive processes/equipment with medium/fast dynamics



MPC-on-a-chip – Future Directions

- **Type 3** – small scale and inexpensive processes/equipment with medium/fast dynamics
 - Available control hardware/software limited - not suitable for online MPC
 - Multi-parametric MPC technology **essential**
 - MPC-on-a-Chip part of embedded (all-in-one) system
 - Suitable for new technologies (FPGA, wireless)

**Multi-Parametric
Programming & Model
Predictive Control – *are
they meant for each other?***

YES
**in most cases (but not
for all)**

*"We owe to our parents our being
and to our teachers our well being"*

Alexander the Great

... to our teachers

- Professor Anastasios Karabelas
- Professor Stavros Nychas



THANK YOU!