

# EVALUATION AND IMPLEMENTATION OF ADVANCED PROCESS CONTROL WITH COMPACTCRIO MATERIAL

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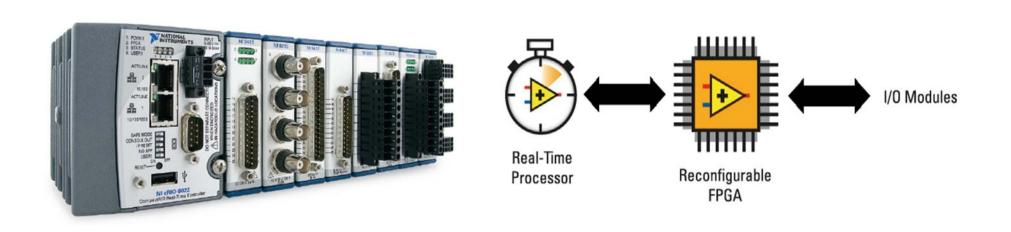
#### **MOTIVATION & STUDY**

Programmable Logic Controller (PLC) is commonly used in industries and research applications for process control. However a very complex process control may require algorithms and performances beyond the capability of PLC's, very high-speed or precision controls may also require other solutions.

cRIO has been used at CERN in a recent research conducted to implement advanced process controls (decoupling of Muti Input Multi Output process control, steady state feedback, Kalman filter) with the compactRIO (cRIO) material from National Instruments.

### **Compact RIO**

cRIO is a reconfigurable embedded control and acquisition system developed by National Instruments, integrating a real-time controller, a reconfigurable FPGA and I/O modules. The real-time controller consists of a powerful processor for performing autonomous and deterministic applications. The FPGA is a programmable logic circuit, comparable to the hardwired logic.

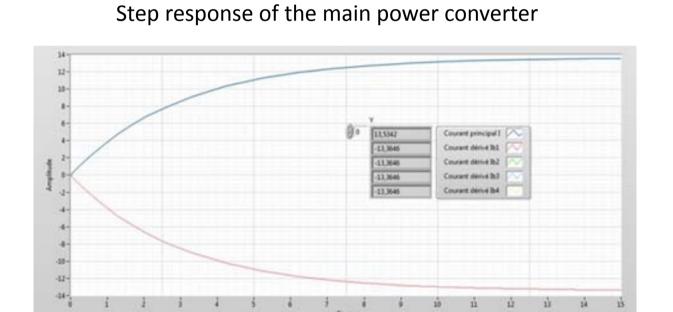


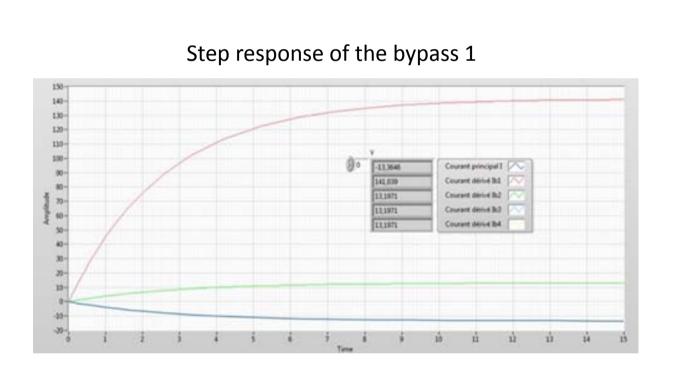
## MULTIPLE INPUT MULTIPLE OUTPUT

Performance assessment of the cRIO for advanced control systems: the CERN SPS beam transfer lines (several magnets in series supplied by one power converter and correction by bypass).

#### **PROCESS PROCESS MODEL** $U = L\frac{dI}{dt} + RI + L_1\frac{dI_1}{dt} + R_1I_1 + L_2\frac{dI_2}{dt} + R_2I_2 + L_3\frac{dI_3}{dt} + R_3I_3 + L_4\frac{dI_4}{dt} + R_4I_4$ $U_1 = L_1 \frac{dI_1}{dt} + R_1 I_1 + R_{C1} (I_1 - I)$ 22 airmants L=220mH R=73mΩ $U_2 = L_2 \frac{dI_2}{dt} + R_2 I_2 + R_{C2} (I_2 - I)$ $U_3 = L_3 \frac{dI_3}{dt} + R_3 I_3 + R_{C3} (I_3 - I)$ L1=20mH $U_4 = L_4 \frac{dI_4}{dt} + R_4 I_4 + R_{C4} (I_4 - I)$ $R1=7.7m\Omega$ **STATE-SPACE MODEL** Main Power Bypass 2 1.2=20mH R2=7,7mΩ Converter $\dot{x}(t) = Ax(t) + Bu(t)$ y(t) = Cx(t) + Du(t)Bypass 3 L3=20mH $R3=7.7m\Omega$ Bypass 4 L4=20mH

#### **PROCESS SIMULATION**



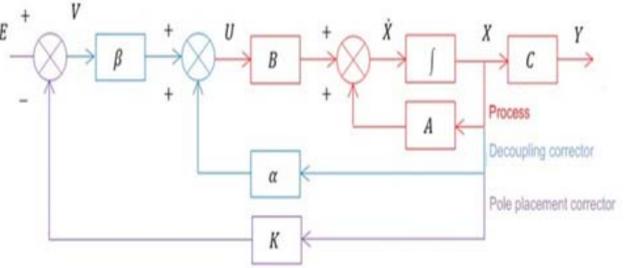


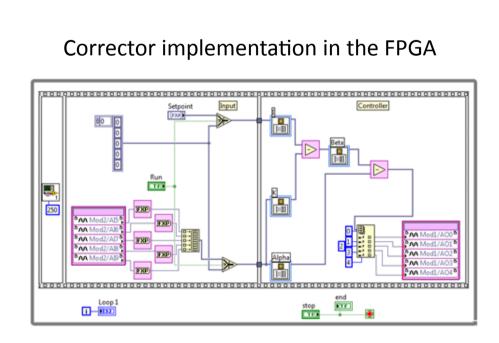
The MIMO process is coupled, the step response to an input affects all outputs.

### MIMO DECOUPLING & POLE PLACEMENT CORRECTOR

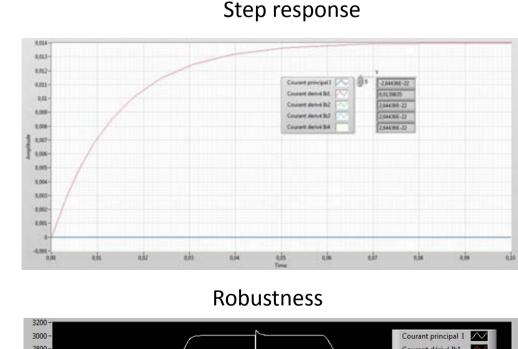
The goal of decoupling corrector is to limit the response of an input to an output with the state feedback  $U = \beta V + \alpha X$ In order to adjust the dynamic in close loop by pole placement, we add an additional corrector with the state feedback V = E - KX

Process, decoupling corrector and pole placement feedback





RESULTS



With the advanced process control (MIMO decoupling), the inputs are decoupled of the outputs and the process reacts as several independent subsystems. The dynamic of each subsystem can be adjusted (here with pole at -70, time constant is 14ms).

The robustness of the system is also insured, a perturbation on an output does not affects other subsystems.

The acquisition time of the I/O modules and the treatment for the MIMO advanced control is 10µs.

#### **KALMAN FILTER**

To study and evaluate the performance of the cRIO FPGA for signal processing, it is necessary to implement more complex algorithms.

#### **KALMAN FILTER THEORY**

The Kalman filter is a powerful and useful mathematical tool in the embedded system. It allows to estimate the state of a system, depending on its previous state, controls applied and noisy measurement.

Equations of a discrete Kalman filter are divided into two stages (prediction and correction):

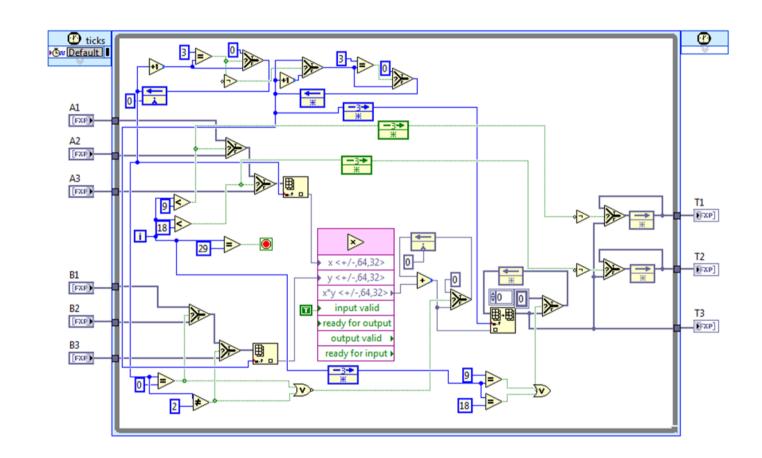
Step of prediction Step of correction 
$$\hat{x}_{k+1/k} = A_k \hat{x}_{k/k} + B_k u_k$$
 
$$\hat{x}_{k+1/k} = A_k P_{k/k} A_k^T + G_k Q_k G_k^T$$
 
$$\hat{x}_{k/k} = \hat{x}_{k/k-1} + K_k (y_k - C_k \hat{x}_{k/k-1})$$
 
$$P_{k/k} = (I - K_k C_k) P_{k/k-1}$$

And  $K_k$  is the optimal gain of the filter  $K_k = P_{k/k-1} C_k^T (C_k P_{k/k-1} C_k^T + R_k)^{-1}$ 

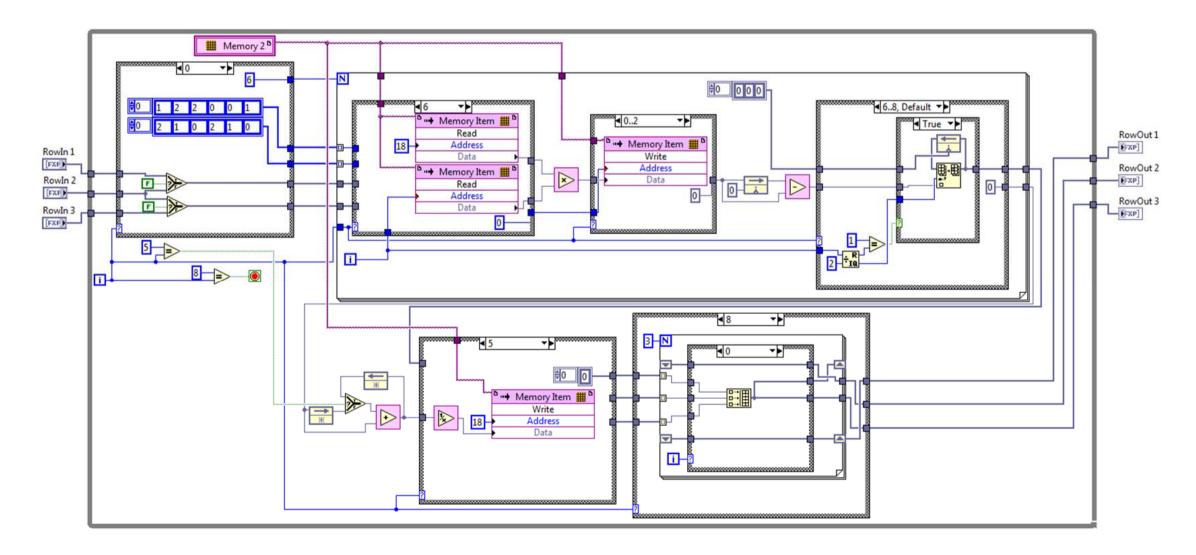
- $\widehat{\mathcal{X}}$  is the estimation of the state correction
- is the estimation error
- $G_k$  is the matrix specifying the noise vector involved in the model of evolution
- $H_{m k}$  is the matrix specifying the noise vector involved in the model of observation

#### KALMAN FILTER IMPLEMENTATION

The toolbox of FPGA LabVIEW provides a single tool to multiply a matrix with constant coefficients with a vector. For Kalman filter, it is necessary to perform matrix multiplication with variable coefficients with vector and also matrix multiplication by matrix (matrix being variable coefficients).

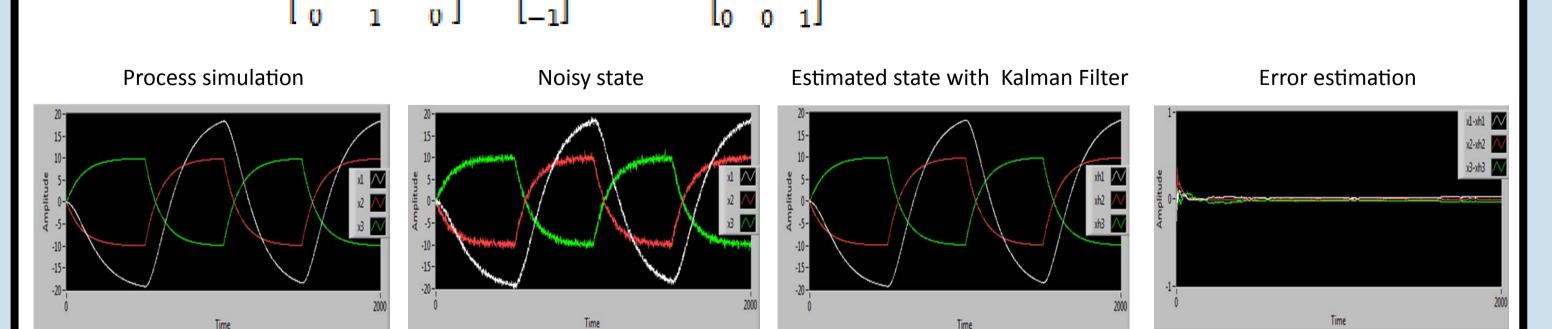


The calculation of the inverse matrix has been implemented in the FPGA with the Laplace formula  $A^{-1} = \frac{1}{\det A} t com A$ 



### **RESULTS**

In order to test Kalman filter implemented in the FPGA, the test was realized on a process with the following state-space representation



The algorithm performs in 2020 cycles, or 50µs with a clock rate of 40 MHz.

# **SUMMARY**

The speed of the FPGA in the cRIO and the toolbox from Labview allows the implementation of complex calculations in order to achieve the advanced control process command and the signal processing. The cRIO technology is a very good solution in comparison to the PLC for a fast and advanced control process, the computation time is about few 10µs against 20ms to 50ms for the PLC (depending on the complexity of the process and the number of I/O modules).