

B. The Python Language

The Raspberry PI III has installed Python version 2 and 3. Python is a language versatile, used in Web development and many other types of applications. The name "Python" is a tribute to the English comedy group Monty Python. It runs on servers of mega-sites like Google and YouTube, in the printers computing clusters Industrial Light and Magic, in NASA laboratories and AstraZeneca Pharmaceuticals, and in games such as Civilization IV and EVE-Online. Python offers the following features also available in more complicated languages such as Java and C++.

In Process control, Python has Control-specific libraries (Control) and manipulation of arrays (Numpy), with functions similar to Matlab Software. It has diversified functions for plotting data in multiple libraries (matplotlib). These libraries contain a set of classes and functions that implement common operations for analysis and design of control systems. Python offers Compatibility Control Pack with Matlab (control.matlab) is available and offers a lot of functions corresponding to common commands in Matlab Control Systems Toolbox. Python-control must be imported into the python import command. Python-Control uses the Numpy, which is the basic package for scientific computing Python. It is a Python library which supports arrays and multidimensional arrays, owning a large collection of mathematical functions for quick operations with arrays, including mathematical, logical operations, handling forms, selections, i/o, Fourier transform, basic linear algebra, basic statistics operations, simulations, etc.

III. PREDICTIVE DMC CONTROL SYSTEM

A. Predictive DMC Controller

The model-based Predictive Control, MPC (Model Predictive Control) is a type of control algorithm, which uses the process model, to predict the behavior of the system controlled. Every sampling period, the optimization problem is solved and is calculated future actions to control. However, just the first control action is sent to the plant. In the next time, the parameters of the model are updated with the data of the plant and the optimization problem is recalculated. The use in the industry model-based Predictive control is growing, mainly because your feature to deal with the constraints of the process, as well as your anticipatory action to correct the error. The technique of type DMC is showing widely accepted due to the simplicity of the algorithms and the use of the step response model, which requires little information for the dynamic system characterization, used to predict the effect of future control actions on exit, determined by minimizing errors in the future. The approach of MPC controllers is presented in Figure 2.

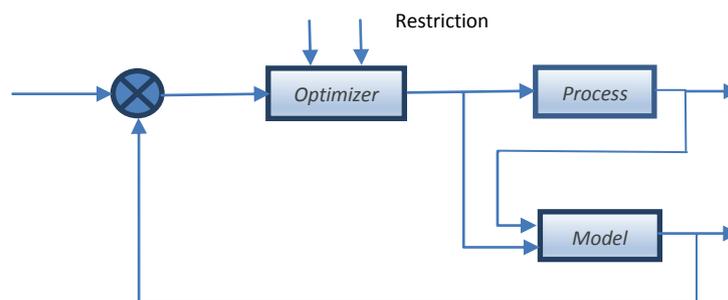


Figure. 2 Predictive MPC Controller

The future outputs for a given N , called horizon of prediction, are predicted for each time t , using the process template. The outputs predicted $\hat{y}(t+k|t)$, for $k = 1 \dots N$ depend on the knowledge of the values until the moment " t " (past inputs and outputs) and signs of future control $u(t+k|t)$, $k = 0 \dots N-1$, which are provided by the system and calculated.

The set of futures control signals are calculated by a certain optimization role to keep the process as close as possible to the path of reference $w(t+k)$, that can be the set point, or an approximation thereof. This role usually takes the form of quadratic function the error, between the sign of predicted output and the reference predicted trajectory. The control effort is included in objective function in most cases.

Control signal $u(t)$ is forwarded to the process, while others are rejected because in the next instant of time $y(t + 1)$ be known, the steps be replayed.

The goal of the DMC controller is to calculate the actions of control, represented by the manipulated variables, increments through the minimization of cost equation (1):

$$J(t) = \sum_{k=N_1}^{N_2} \lambda_y [\hat{y}(t+k) - w(t+k)]^2 + \sum_{k=1}^{N_u} \lambda \Delta u^2(t+k-1)$$

Equation 1

Where N_2 is the output forecast horizon, the horizon ($N_u \leq N_2$). λ_y and λ_u are the weights in the control and output signals, respectively. $\hat{y}(t+k)$ is the output forecast at the moment $(t+k)$, using the information available at time t . $w(t)$ is the reference at time t and $\Delta u(t)$ is the increase in the control signal at time t .

The prediction along the horizon is given by (2):

$$\hat{y}(t+k|t) = \sum_{i=1}^k g_i \Delta u(t+k-i) + \sum_{i=k+1}^N g_i \Delta u(t+k-i)$$

Equation 2

Where $\sum_{i=k+1}^N g_i \Delta u(t+k-i)$ is the free response of the system. This and the part of the answer that does not depend on future control actions, and is given by (3):

$$f(t+k) = y_m(t) + \sum_{i=1}^{\infty} (g_{k+1} - g_i) \Delta u(t-i)$$

Equation 3

Where g_i is obtained from the system step response.

IV. DESIGN OF DMC CONTROL SYSTEM

A. Process Description

The area under test comprised a supply sector of a sanitation company, responsible for the supply of 1500 People. A booster shown in figure 3, consisting of a motor-pump set of 7.5 CVs is responsible for the supply of treated water. This booster has a frequency Inverter triggering a motor pump set. At the critical point of the supply sector was installed a pressure transmitter.



Figure. 3Supply structure

The acquisition of the data was carried out in the frequency range lift operation 45 to 56hz. 1500 samples were taken, figure 4.

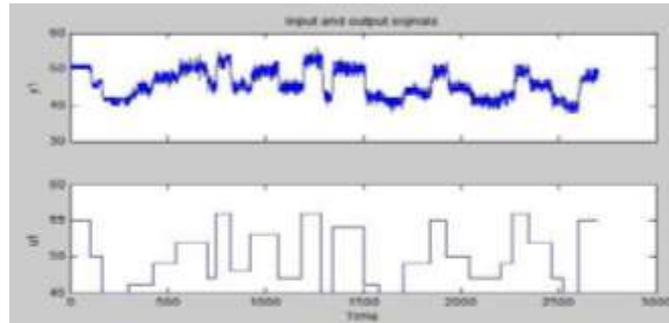


Figure. 4 Collected data

The identification and validation of transfer function relating the pressure at the critical point and the frequency of operation of the motor pump set, were made using the Ident tool of Matlab. (4)

$$G(s) = \frac{0.96e^{-2.59s}}{s(1 + 29.81s)}$$

Equation 4

B. Python Control DMC Implementation

For this project implementation in Python programming language was used mainly Control, matplotlib and numpy libraries. Due the Raspberry PI does not have analog inputs, we used the converter analog/digital PIC 16f877A as input interface, which makes reading the set-point and the process variable, converts to digital and passes to the Raspberry PI via I2C Protocol. How I2C communication each byte is transmitted at a time and the signal converted has two bytes, each variable is broadcast with two bytes. SPL/Sph are the bytes from the set-point and PvL/PvH are the bytes of the process variable, and are mounted in Raspberry after reception.

The prediction horizon, horizon, sampling interval control and weighting control signals and output have been set as follows:

HP = 7 (HP is the forecast horizon)

HC = 5 (HC is the control horizon)

TS = .04 (sampling interval) Limit = 1800/Ts

lick = .8 (Consideration of the control signal)

Alpha = 0.3 (factor reference forecast)

Delta = 0.6 (output signal Weighting)

The control signal output served directly on the PWM of the Raspberry PI, controlling the inverter. The range of PWM DutyCycle varies from 0 to 100%, where 0 represent 45Hz frequency and 100% represent 56Hz of operating frequency.

To test implemented controller, was generated a random reference signal (r - line in blue). Based on cost function, the control signal from the controller (u - line in purple) is generated controlling the output (y - line in green), figure 5.

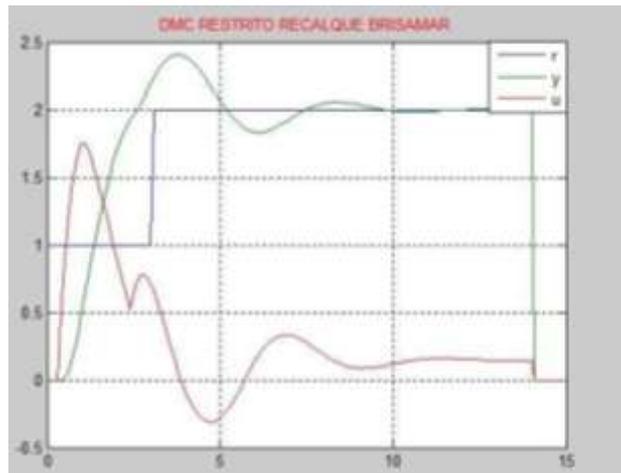


Fig. 5 Implemented DMCController

To evaluate the DMC controller, was implemented a PID controller and tuned by the Ziegler-Nichols method, where through comparison of performance index, settling time and overshoot.

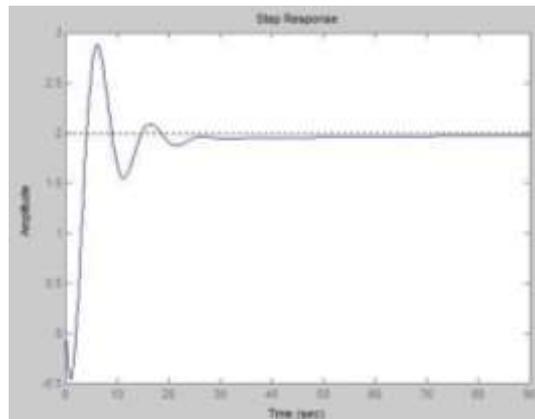


Fig. 6 Implemented PID signal.

V. CONCLUSION

The results obtained showed the feasibility of implementing the DMC in Python language. Although there are differences in implementation in relation to Matlab, the two languages control functions and a variety of compatible libraries, being the main advantage of implementing cost reduction, since Python is free.

Raspberry implementation IP enabled boarding the controller in a compact device with wide variety of features, which allowed the direct interface to the drive.

Both controllers in the test process stabilized, but note the anticipatory action in the DMC predictive control, providing a faster response of the front controller set-point variations.

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