

Latest Advancements in Process Control in Refineries and Chemical Plants

*Author: Giovanni Franchi-Chemical Engineer – PhD Student
–University UCBM – Rome (Italy)*

1.Theme Description

The world production of chemicals in 2020 will increase of 144 Million of metric tons[\[1\]](#) with a market of 4.650 US\$ billion.[\[2\]](#) Automation and Process Control have a pivotal role in industrial plants, indeed, they allow to improve products quality, plants efficiency, the safety and reliability of the processes.

The automatic feedback controls were introduced in 1920-1930s mounted on the controlled equipment. Since then, process control has spread rapidly from the first digital devices at the end of 1950s up to Programmable Logic Controllers (PLCs) and Distributed Control System (DCS) in 1970s. Nowadays networks of computers manipulate thousands of variables, but 85-95% of feedback control loops are based on Proportional-Integrative-Derivative (PID) system developed in 1930s. Furthermore, the flow rates of liquid and gases are controlled by pneumatic valves.[\[3\]](#)

The use of advanced controls can increase plant's profit margin between 10-20% and reduce emissions of about 70%[\[4\]](#). Therefore, in the following sections PID tuning optimization, APC (Advanced Process Control) and MPC (Model Predictive Control) are described. Finally, an overview on the

latest software in process control and “smart control” are discussed.

2. Process Control

The Aspen Technology Inc. has defined five levels of maturity for a refinery and chemical plant depending on the control level, from level zero where no process simulation is used up to level four, where several models are reported in a single flowsheet and engineers can make decisions by monitoring key parameters.[\[5\]](#) As can be seen in Figure 1, a plant, usually runs in a safety zone called “comfort zone” away from constrain limits. With PID optimization and APC is possible to reduce from three to ten times the amplitude of oscillations working near constrain limits and increasing productivity and profit margins.[\[6\]](#)

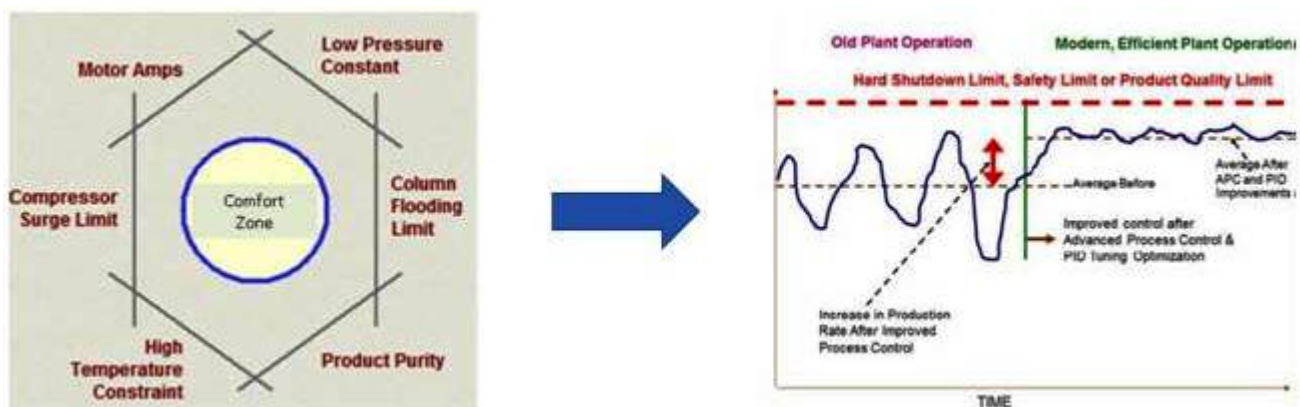


Figure 1 – “Comfort Zone” and Positive Effects by using PID tuning optimization and APC.6

Therefore, in this section PID tuning optimization and APC controls are described.

PID Controls

The PID controls are the most common control used into chemical and petrochemical plants due to easy implementation and robustness. (Figure 2)

The controller takes a corrective action depending on the magnitude of the error[7]:

- proportional action cuts off most of the errors;
- integral action reduces steady-state error or off-set;
- derivative action decreases maximum overshoot.

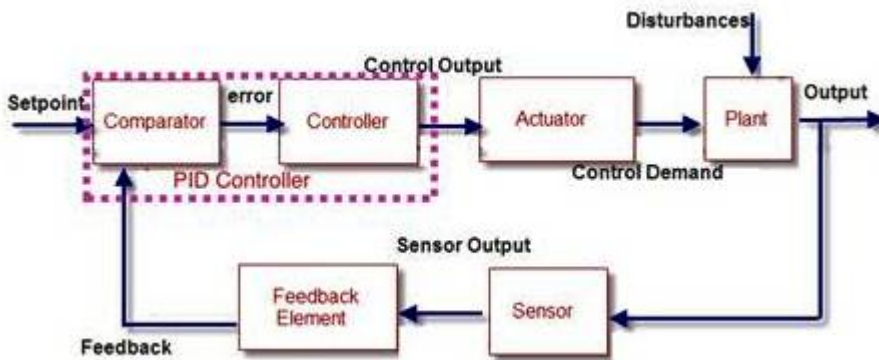


Figure 2 – PID control loop.[8]

In order to have desirable outputs it can be used a separate controller for each variable (decentralized strategy) or a single controller that manipulate all the variables (centralized strategy)3.It is usually used in parallel form expressed as follow:

Where, b = bias (steady-state) value; K_c = controller gain; $e(t)$ = error signal set equal to (set point – present value); τ_I = integral time or set time; τ_D = derivative time.

Tuning Optimization

There are several tuning techniques that allow to find proper PID parameters.

These methods were developed since 1940s and nowadays can be divided into two categories: Classical and Artificial Intelligent methods.

Classical methods include: **Ziegler and Nichols** and **Cohen and Coon Methods**. Ziegler and Nichols proposed two methods, the first called “step response” that can be applied only to open-loop stable plants. It considers, indeed, the response of industrial process such as a S-shape without overshoot. As can be seen in Figure 3, the delay time (L) represents the intersection of tangent line at inflection point of the curve and x-axes. While time constant (T) represents the intersection with steady state line. From these values is possible to find PID parameters.

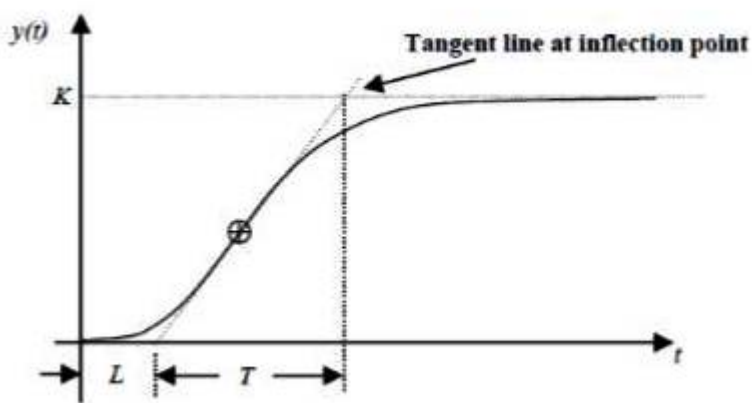


Figure 3 – Parameters of the First Ziegler Nichols Method. [\[9\]](#)

The second method is called “continuous cycling method”. It allows to find the critical frequency of the system by increasing the proportional gain until stability limit. The two parameters that describe the response of the system are K_{cu} (ultimate gain) and P_u (ultimate period). In Table 1 are shown the relationship between PID parameters and the two Ziegler and Nichols methods.

Method	Controller	K_c	τ_i	τ_D
Step Response	P	T/KL	-	-
	PI	0.9 T/KL	3.3 L	-
	PID	1.2 T/KL	2 L	0.5 L
Continuous Cycling	P	0.5 K_{CU}	-	-
	PI	0.45 K_{CU}	$P_U/1.2$	-
	PID	0.6 K_{CU}	$P_U/2$	$P_U/8$

Table 1–Parameters of the Step Response and Continuous Cycling Method. [10]’ [11]

Ziegler and Nichols methods are suitable for level control but not for flow, liquid pressure that require a rapid response. [12] In these cases Cohen and Coon methods are used. [13] This method finds three poles: two complex and one real that allow to minimize the integrated error and to have a decay ratio of about $\frac{1}{4}$. [14]

Artificial Intelligent Methods include dozens of methods. Some of them are described, such as **Genetic** and **Differential Evolution Algorithms**. For the other ones such as Simulated Annealing (SA), fuzzy system, Artificial Bee Colony (ABC), Particle Swarm Optimization (PSO) etc. it can be consult references [15], [16] .

Genetic Algorithm starts from a random population of binary strings called chromosomes each of them represent the solutions of the problem. The strings are encoded to real number that defines PID parameters. These values are elaborated by PID controller and the response is evaluated by means of objective function such as MSE (Mean Square Error), IAE (Integral Absolute Error), ISE (Integral Squared Error) etc. The fitness values are subjected to a process of selection, crossover and mutation until best fitness is obtained. [17]

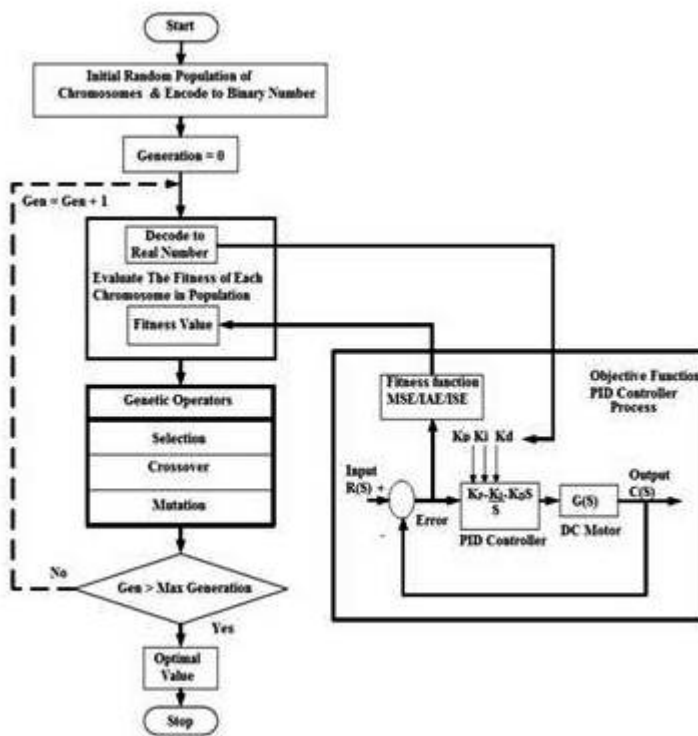


Figure 4 – Flow chart of Genetic Algorithm. [\[18\]](#)

Differential Evolution Algorithm, instead, starts from the initialization of real encoding matrix where rows represent PID parameter while columns represent i-th population vector. Each population is evaluated by PID controller and the result represent the fitness value. Then crossover step that involves target vector (first vector of population) and mutual vector (three random vectors from population are selected) generates a trial vector whom fitness values are elaborated by PID. Finally, fitness values of target vector and trial vector are compared to select the minimum value. In this way the individual of new population is generated. The algorithm stops when new population is completed.

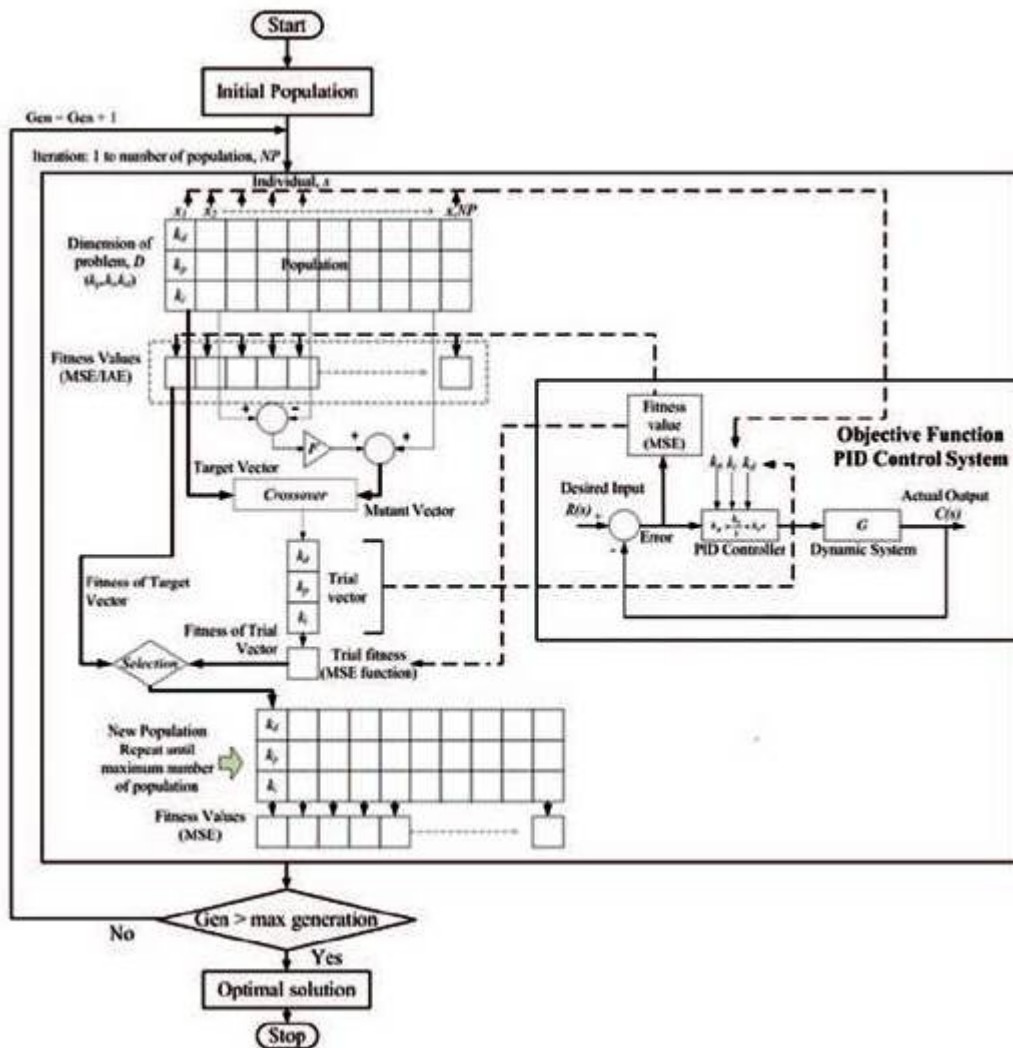


Figure 5 – Flow Chart of Differential Evolution. [\[19\]](#)

The main limit of feedback control is that corrective action takes place only when output is perturbed from its set point. Therefore, more advanced controls were developed such as PID plus feed forward that allows to intervene before disturbance takes place or cascade composed by two controllers, two sensors and one actuator acting on two processes in series. [\[20\]](#) Despite PID controllers ensure good stability and suppression of the disturbances, process performance optimization fails due to the multivariable nature of it and the complex interactions between controlled variables. Therefore, Advanced Process Control are necessary. [\[21\]](#)

3. APC

The APC includes all software that allows to control critical variables and predicts quality in real-time such as:

- **Statistical Process Control (SPC)** that uses a random sampling and statistical analysis to identify causes outside the process that changed the quality of a product. This method is used especially for manufacturing lines, but it fits well also process where output can be measured. [\[22\]](#)
- **Run2Run (R2R)** is widespread in the semiconductor industry (for more information see Moyene et al. [\[23\]](#)), but it can be applied also to batch processes such as chemical vapor deposition or batch chemical reactors¹⁰. The quality of the product is evaluated at the end of the run and the set-points are changed between two successive runs. Therefore, this control is used when there aren't enough on-line measurements of the interest products.
- **Model Predictive Control (MPC)** described more in detail later.

An example of APC implementation is described by Howes et al.⁶ for a lubrication of oil process. The system consists of 12 manipulated variables, 28 controlled variables and 11 feedforward controls. By means of Pitops software developed by Pi Control the plant has increased rate of production of 5% saving about 1.3 M€. The software, indeed, gives the parameters of the system in 10 minutes basing on historical data without step tests. Other examples are Canada's Yara Belle Plaine Inc. and South Korea's LG Petrochemical Corp. The first applied APC techniques to a nitric acid plant reducing methane emissions by 25% maintaining high temperature combustion. While the latter, applied to a naphtha cracking, allows to improve yield by 5%, reducing energy consumption in cold side by 8% and saving 100.000 \$/y. [\[24\]](#)

MPC

The precursor of MPC is represented by LQG (Linear Quadratic Gaussian) developed by Kalman in 1960s, but the first MPC generation appeared in 1970s with IDC0M (developed within ADERSA) and DMC (developed within Shell Oil). Nowadays we have reached the fifth generation where Honeywell, AspenTech and Shell dominate the markets[\[25\]](#). The MPC is suitable to describe the behaviour of MIMO (Multi-Input, Multi-Output) processes.

As can be seen in Figure 6 a classical plant control provides for different hierarchical levels[\[26\]](#): a plant wide-optimization, a local economic optimizer and a dynamic constraint control. Usually this is done by several PID controls lead-lag (L/L) blocks and high/low select logic. With MPC, shown in more detail on the right end, this can be achieved with better results by taking an action on the difference between actual and predictive value (residuals).

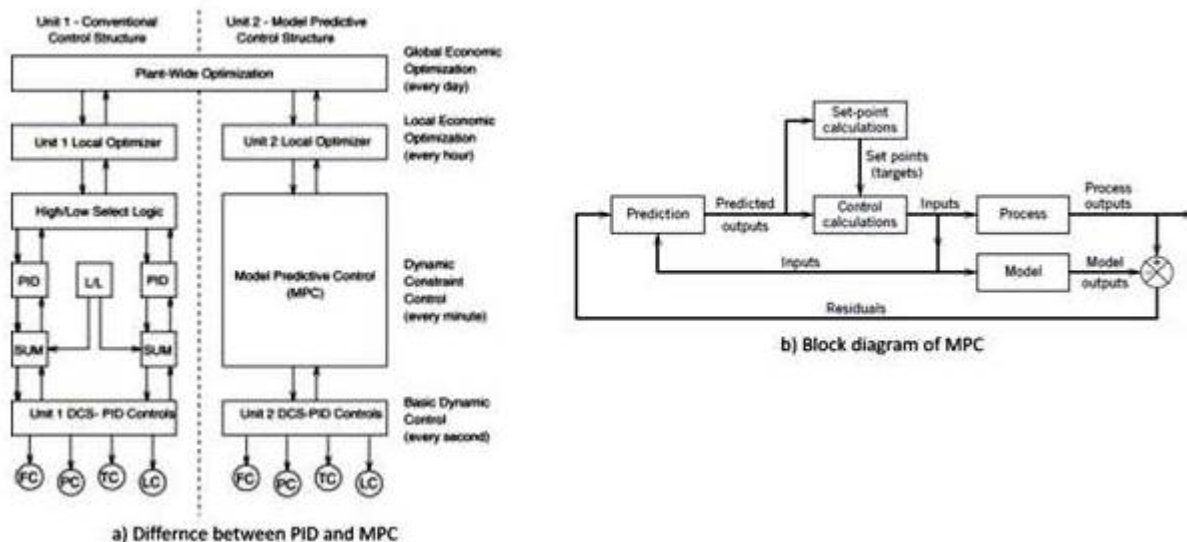


Figure 6—Comparison between a PID and MPC controllers.[\[27\]](#)[\[28\]](#)

Commercial MPC software

Nowadays the main MPC software includes: DMCplus developed by AspenTech, SMOc of Shell Global and RMPCT of Honeywell. The following is a brief description of these, for more details see Lahiri[\[29\]](#).

DMCplus derives by the fusion of Dynamic Matrix Control (DCM) and Setpoint Multivariable Control Architecture (SCMA). The software is composed by several packages and allows to simulate Finite Impulse Response (FIR), linear Multi-Input, Multi-Output (MIMO) and nonlinear Multi-Input, Single Output (MISO) State Space. Recently AspenTech has introduced the Adaptive Process Control that reduces the possibility to have “flipping” behaviour of the plants due to the difference between the plant and the controller model. This package, indeed, forced the system to work in an optimum area instead of optimum setpoint. In this way, the performances of the plants aren't compromised. The model, also, is implemented by using historical data of the process and adjusting online the parameters with an adaptive model, saving time.

SMO has been used in more than 430 applications such as crude distillation, hydrocracking, styrene etc.

It includes several packages:

- **AIDA^{pro}** is an offline software that analyzes data from open and closed loop defining the mathematical model to be used by taking into account unmeasured disturbances.
- **MD^{pro}** is an offline software that allows to control base and multivariable control loops by means of statistical methods.
- **RQE^{pro}** uses a Kalman filter that allows to adapt the model on process conditions. It gives parameters for online process reducing maintenance and increasing stability of it.

- **SMOC^{pro}** is used for multivariable control system optimization. This package uses a grey-box approach where input and output are defined by means of intermediate variables. In this way the engineers can take a cascade correction optimizing the response of the process.

RMPCT manages process with huge errors and large interactions between controlled variables. The controller doesn't follow a specific trajectory but can move through any trajectory within the constraints defined by a "funnel". At the same time, the controlled variables aren't forced to keep the set points, but can change in a range. In this way disturbances are rejected and the control is optimized.

Smart Control

In the era of digital devices, "smart control" for chemical and petrochemical industry can have a key role in reducing costs, saving materials and increasing production rates. The idea is to create intelligent networks where flowsheet and variables are optimized in real time.[\[30\]](#) Emerson represents the leading company in the sector providing smart solutions both for old and new refinery such as electronic marshalling and HART (Highway Automated Remote Transducer) protocol. The former allows to eliminate cross-wires, to reduce the space occupied and the time for add new I/O interfaces[\[31\]](#); the HART protocol, instead, matches the characteristics of analog and control system removing repetitive problem and predicting unexpected failures.[\[32\]](#) Several companies have implemented smart controls such as Chevron/PDVSA in Petropiar refinery saving 70 M\$ in two years, reducing by 40% cost for pre-commissioning and commissioning, and 60% the losses due to instruments faults.[\[33\]](#) In China, Sinopec launched four pilot

plants (Jiujiang, Zhenhai, Maoming, and Yanshan) using advanced control and online optimization. This allowed to increase profits of about 10% (i.e. at Yanshan and at Maoming profits increased of 25.12 million of CNY and 41.94 million of CNY respectively).³⁰

4. Conclusion

Process Control is very common in refineries and chemical plants. It was used for the first time in 1920-1930s and today is essential to respect product quality, safety and reliability of the processes. Despite the progress of technologies, 85-95% of feedback control loops are based on PID controllers and the main system controls are dated in 1985. The value of technologies that has reached the end-life and with more than 20 years is about 65US\$ billion dollars and 53 US\$ billion dollars respectively.³³ Therefore, several tuning optimizations such as Artificial Intelligent Methods (Genetic and Differential Evolution Algorithms) together with Advanced Process Control (APC) have been described. Furthermore, some examples of the advantages offered by the implementation of APC have been shown and the latest software for Model Predictive Control (MPC) have been illustrated such as DMCplus, SMOC and RMPCT. In this scenario the “smart control” in chemical and petrochemical plants can have a pivotal role in reducing costs increase profits and create safer plants. The current estimate provides that plant’s profit margin can improve of about 10-20% while emissions can decrease of about 70%.

[1]

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[\[2\]](#)

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[\[3\]](#)

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