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# Nomenclature

## Acronyms

|       |   |
|-------|---|
| AANN  | Autoassociative Artificial Neural Network       |
| ABM   | Assumption Based Method                         |
| AEGIS | Abnormal Events Guidance and Information System |
| ANN   | Artificial Neural Network                       |
| ASM   | Abnormal Situation Management                   |
| BPN   | Backpropagation Artificial Neural Network       |
| DCS   | Distributed Control System                      |
| EKF   | Extended Kalman Filter                          |
| F&EI  | Fire and Explosion Index                        |
| FD    | Fault Diagnosis                                 |
| FDS   | Fault Diagnosis System                          |
| FL    | Fuzzy Logic                                     |
| FLS   | Fuzzy Logic System                              |
| FMEA  | Failures Modes and Effect Analysis              |
| GDR   | Generalized Delta Rule                          |
| GUI   | Graphical-User-Interface                        |
| HAZOP | Hazard and Operability study                    |
| IFAC  | International Federation of Automatic Control   |
| KB    | Knowledge Base                                  |
| KBES  | Knowledge Based Expert System                   |
| MF    | Membership function                             |
| MHI   | Material Hazard Index                           |
| MPCA  | Multiway Principal Component Analysis           |
| MSE   | Mean Square Error                               |
| NF    | Neuro-Fuzzy                                     |
| NLPCA | Nonlinear Principal Component Analysis          |
| OBM   | Observer Based Method                           |
| P&ID  | Piping and Instrument Diagram                   |
| PCA   | Principal Component Analysis                    |
| PCB   | Pseudo Continuous Block                         |
| PCEG  | Possible Cause-Effect Graph                     |
| PHA   | Preliminary Hazard Analysis                     |
| PLS   | Partial Least Squares                           |
| PNN   | Probabilistic Artificial Neural Network         |

|      |                                      |
|------|--------------------------------------|
| QTA  | Qualitative Trend Analysis           |
| RB   | Rule Based                           |
| RBFN | Radial Basis Function Neural Network |
| SDG  | Signed Directed Graphs               |
| SOM  | Self Organising Map                  |
| SPC  | Statistical Process Control          |
| SPE  | Squared Prediction Error             |

## Notation

|                 |  |
|-----------------|--|
| $a$             | ANN output vector  |
| $a_i$           | Output vector of the $i$ th layer in a BPN                             |
| $A_{pi}$        | Approximation on the $i$ th decomposition                              |
| $b$             | ANN bias vector  |
| $b_i$           | Bias vector of the $i$ th layer in a BPN                               |
| $db$            | Number of nodes in the bottleneck layer of an AANN                     |
| $D_i$           | Detail of the $i$ th wavelet decomposition                             |
| $dl$            | Dilation parameter of a wavelet  |
| $E$             | Residual matrix for a new batch  |
| $\underline{E}$ | Residual matrix for historical database                                |
| $f$             | Function   |
| $F$             | Fault vector   |
| $g$             | Target vector for ANN training   |
| $H_0$           | Low pass filter  |
| $H_1$           | High pass filter   |
| $i$             | Index for batches (or for samples)                                     |
| $j$             | Index for measurements (sensors)                                       |
| $J$             | Total number of measurements (sensors)                                 |
| $k$             | Index for time intervals   |
| $K$             | Total number of time intervals   |
| $l$             | Index for layers in an ANN   |
| $m$             | Index for nodes in an ANN  |
| $M_1$           | Vector of measurements from the plant that are the ANN's input         |
| $M_1K^{nf}$     | Matrix with the measurements profiles for the fault $nf$               |
| $M_1'K^{nf}$    | Matrix of extrema of the measurements for the fault $nf$               |
| $m_1$           | Length of vector $M_1$   |
| $M_2$           | Vector of measurements from the plant that are part of the FLS's input |
| $m_2$           | Total number of measurements that are part of the FLS's input          |
| $M_d$           | Number of nodes in the demapping layer of an AANN                      |
| $M_l$           | Number of nodes in the mapping layer of an AANN                        |
| $n_i$           | Activation status vector of the $i$ th layer in a BPN                  |
| $N_1$           | Output vector of the ANN in the proposed FDS                           |
| $n_1$           | Total number of "pre-faults" diagnosed by the ANN                      |
| $n$             | Activation status of a node in an ANN                                  |
| $N$             | Total number of nodes in a layer of an ANN                             |
| $N_f$           | Total number of defined faults   |
| $nf$            | Index for faults   |
| $\%P$           | Performance parameter for FDS  |
| $\%P^*$         | Modified $\%P$ which takes into account cases of false diagnosis       |
| $p$             | Principal components loading vector                                    |
| $P$             | Principal components loading matrix                                    |

|                 |   |
|-----------------|---|
| $P^T$           | Transpose of matrix $P$   |
| $q$             | ANN input vector  |
| $Q$             | Sum of squares of the residuals   |
| $R$             | Range of the $SPE_{js}$ % of the set of sensors                                 |
| $rl$            | Index for rules   |
| $RI$            | Total number of rules in a FLS  |
| $S_i$           | Total number of nodes of the $i$ th layer of an ANN                             |
| $S_n$           | Total number of nodes of the output layer of an ANN                             |
| $SPE_j$         | SPE calculated for the measurement $j$ and the corresponding output of the AANN |
| $SPE_{js}$      | 99% upper control limit of $SPE_j$ for a set of correct measurements            |
| $SPE_{js}\%$    | Percentage relation between $SPE_j$ of a new measurement and $SPE_{js}$         |
| $SPE_{sup}$     | 99% upper control limit of SPE for a set of correct measurements                |
| $t$             | Principal component scores vector   |
| $td$            | Time spent by a FDS for correct diagnosis                                       |
| $tl$            | Translation parameter of a wavelet  |
| $tss$           | Time with the plant in non-steady state after a fault occurs                    |
| $T$             | Principal component scores matrix   |
| $T^2$           | Overall measure of variability  |
| $v$             | Index for training patterns   |
| $W$             | ANN weight matrix   |
| $W_i$           | Weight matrix of the $i$ th layer in a BPN                                      |
| $X$             | Bidimensional matrix  |
| $\underline{X}$ | Tridimensional matrix   |
| $z'_{ct}$       | Output of the centroid defuzzification method                                   |
| $z'_{ht}$       | Output of the height defuzzification method                                     |
| $z'_{m}$        | Output of the mean of maximum defuzzification method                            |
| $z'_{nx}$       | Output of the maximum defuzzification method                                    |

### Greek symbols

|          |                          |
|----------|--------------------------|
| $a$      | Momentum term in the GDR |
| $d$      | Error signal in the GDR  |
| $e$      | Small number             |
| $h$      | Learning rate for GDR    |
| $m$      | Membership function      |
| $\sigma$ | Standard deviation       |
| $y$      | Mother wavelet function  |



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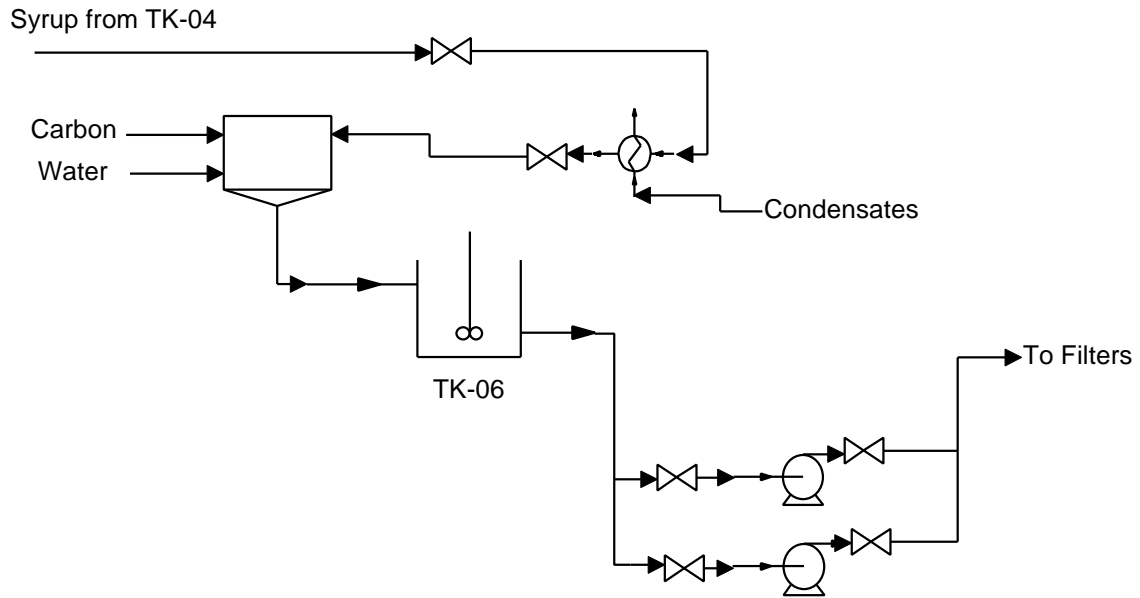
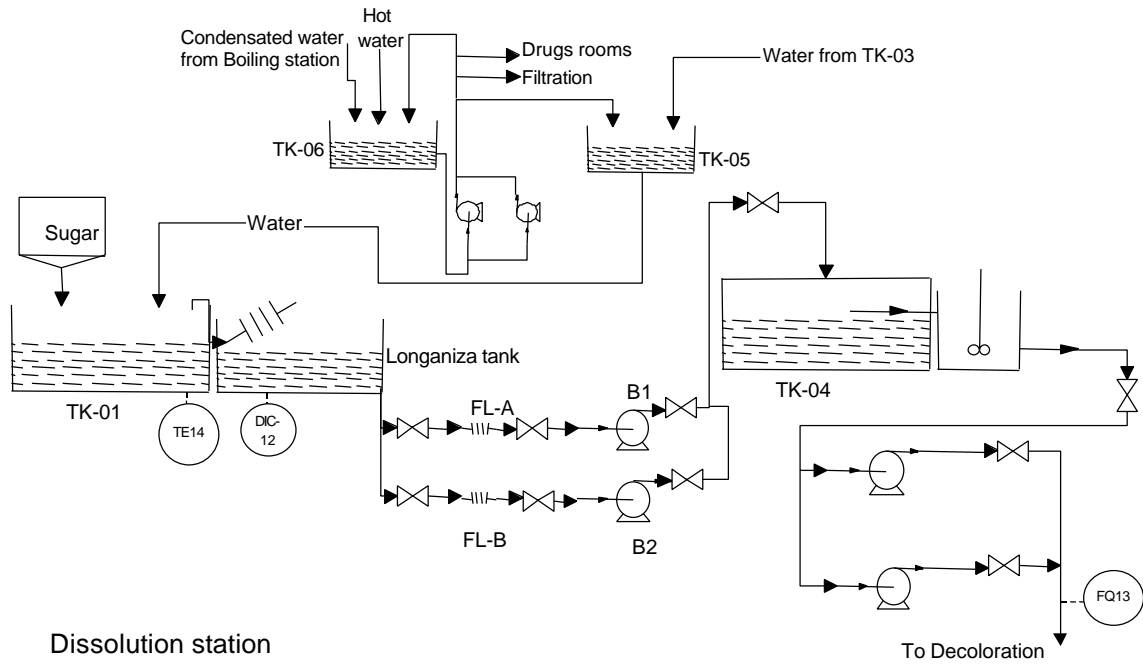


## Annex A

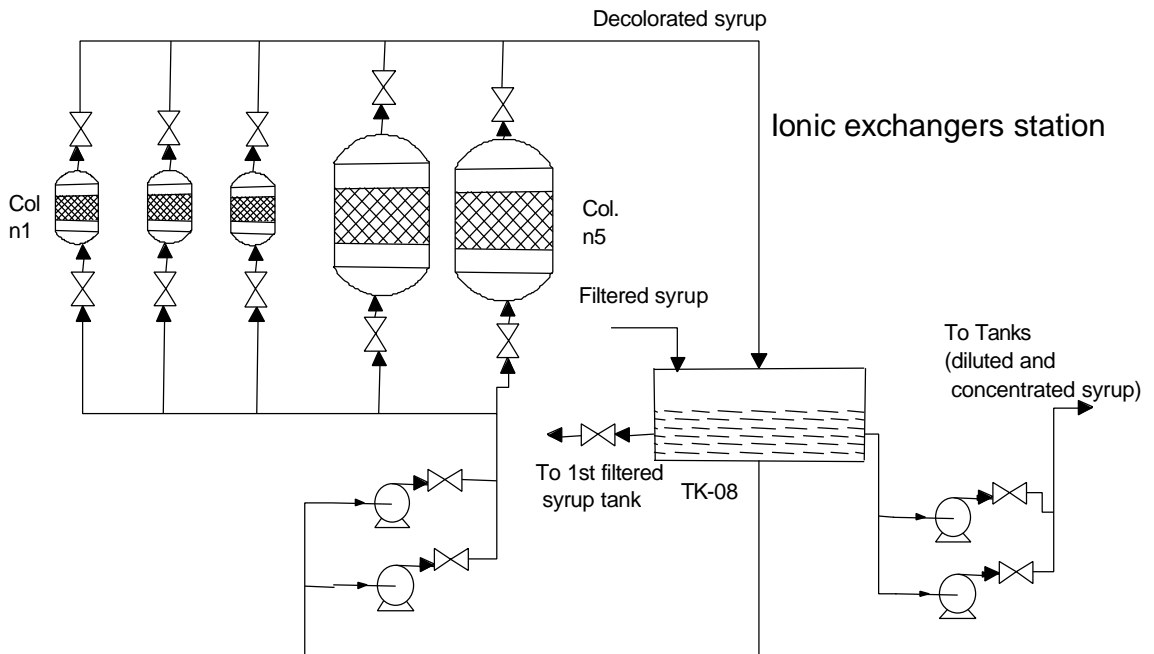
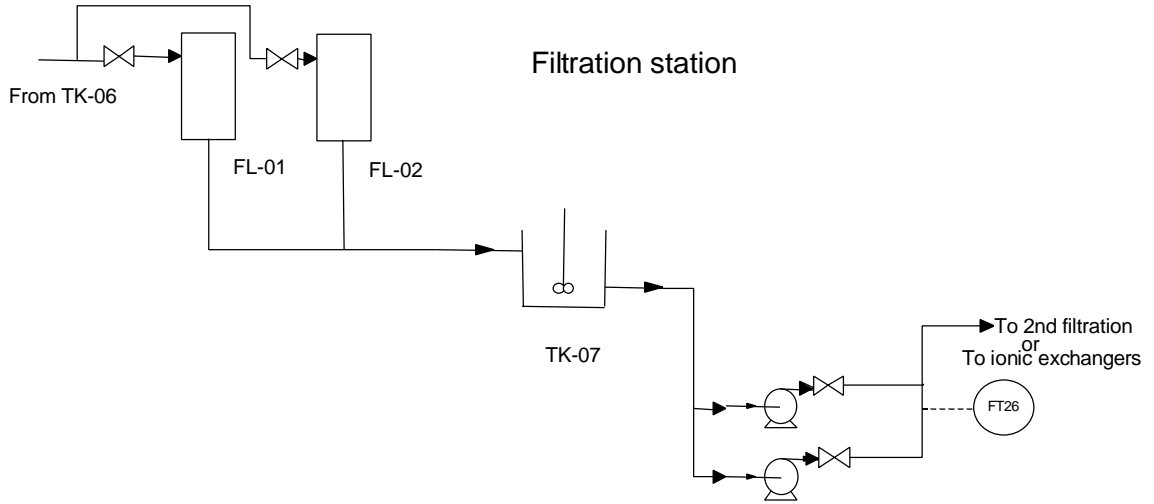
In this appendix, the simplified HAZOP analysis used for the FDS implementation, of the two sugar cane refinery case studies are reported:

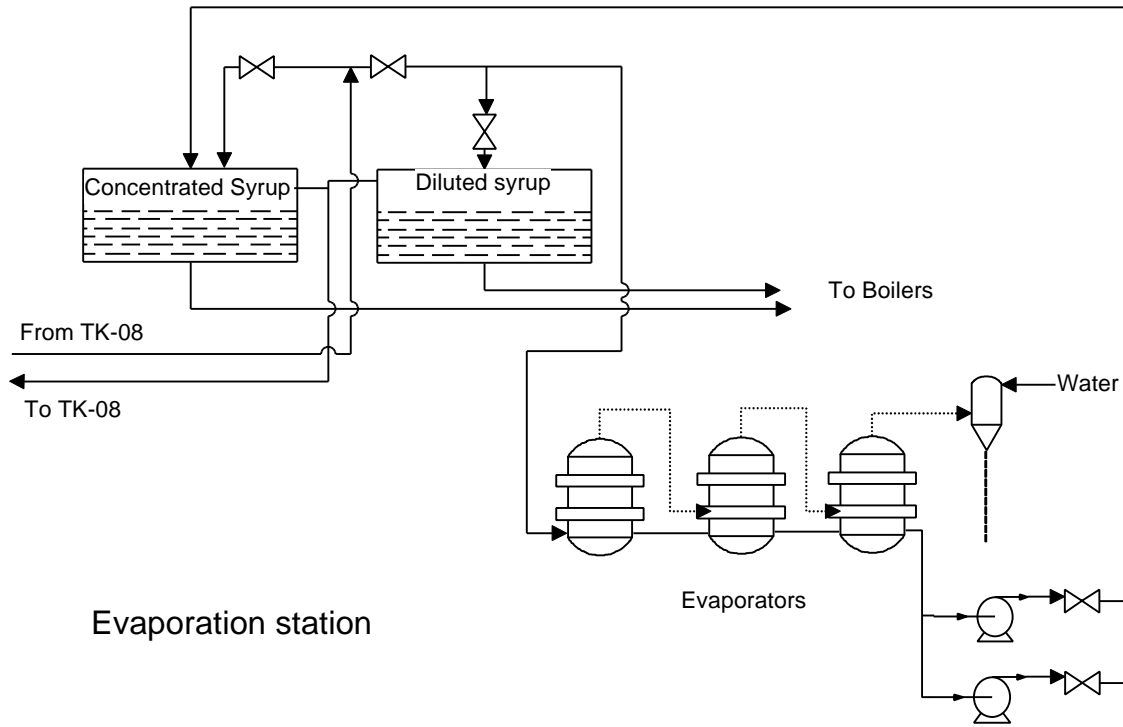
- Complejo Azucarera Concepción S.A. (CACSA), with the collaboration of its Technical Office.
- Complejo Agroindustrial Azucarero Camilo Cienfuegos (CAICC), with the collaboration of its plant engineers.

### CACSA - HAZOP analysis

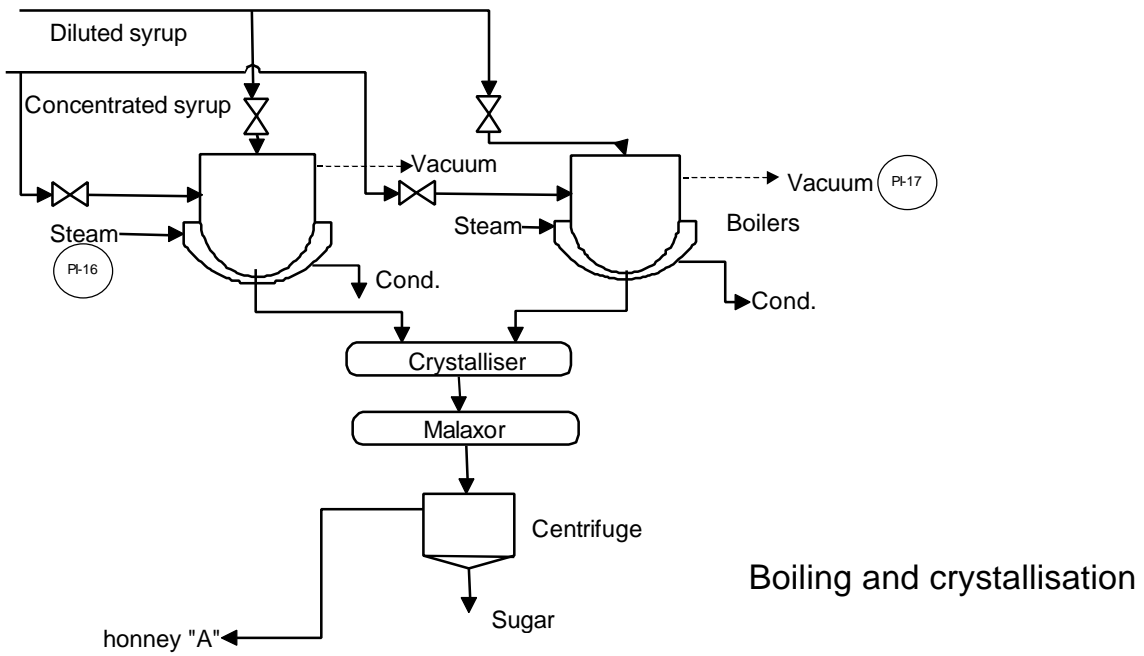








Evaporation station



Boiling and crystallisation

|  |                  |  |   |  |   |
|--|------------------|--|---|--|---|
| Dissolution station                            |                  |  |   |  |   |
| <b>Node: Sugar discharge to TK01</b>           |                  |  |   |  |   |
| <i>Variable</i>                                | <i>Guideword</i> | <i>Causes</i>  | <i>Consequences</i>                                   | <i>Safeguards</i>                            | <i>Recommendations</i>  |
| Flow   | No               | Raw sugar unavailable  | Refinery shutdown                                     | Cut in the Water supply                      | To operate having an adequate excess of sugar in stock                            |
|  |                  | Obstruction in the discharge   |   | Shutdown of pumps B1 & B2                    | Alternative discharges available  |
| <b>Node: Water discharge to TK01</b>           |                  |  |   |  |   |
| Flow   | No               | Piping broken<br>Fresh water tank empty  | High density in dissolution<br>idem                   | Cut in the Water supply<br>idem              | Piping preventive maintenance<br>Low level alarm in tank TK-05 (Fresh water tank) |
| Temperature                                    | Less             | Water temperature decrease in tank TK05 due to low water supply from tank TK03 | Difficulties in sugar dissolution                     | Temperature controller in tank TK01 (TIC-14) |   |
|  | More             | Water temperature increase in tank TK05 due to low water supply from tank TK06 | Possible sugar caramelisation                         | idem   | To install safety cold water stream   |
| <b>Node: Piping from Tank TK01 to pumps</b>    |                  |  |   |  |   |
| Flow   | Less             | Pipe leakage or obstruction  | Lack of syrup in tank TK04                            |  | Piping preventive maintenance   |
|  |                  | Pump malfunctioning  |   | Alternative pump                             | Pressure sensor in filters  |
| Temperature                                    | More             | High temperature in dissolver TK05   | Possible pump cavitation                              | TIC-14                                       |   |
|  |                  | Pumps (B1&B2) output valves closed   |   | idem   | Proximity sensor in valves  |
| <b>Node: Discharge of the "Longaniza" tank</b> |                  |  |   |  |   |
| Density  | Less             | Agitator broken (inefficient dissolution)                                      | Major energy consumption in the concentration process | DIC12 controller                             | Alternative agitators   |
|  | More             | Fault in density controller  | Decoloration difficulty                               | DIC12 controller                             | Redundancy control  |
| Possible crystals precipitation                |                  |  | idem  | idem   |   |

| Decoloration station                                  |                  |  |   |  |  |
|---|------------------|--|---|--|--|
| <b>Node: Syrup stream to the heat exchanger</b>       |                  |  |   |  |  |
| <i>Variable</i>                                       | <i>Guideword</i> | <i>Causes</i>                                      | <i>Consequences</i>                           | <i>Safeguards</i>                                | <i>Recommendations</i>   |
| Flow  | Less             | Pipe leakage or obstruction<br>Pump malfunctioning | Excessive syrup heating                       | Operating procedure<br>Shutdown of pumps B1 & B2 | Temperature controller in the heat exchanger<br>To operate having an adequate excess of sugar in stock |
|   | More             | Syrup excess in tank TK04                          | Temperature decrease                          |  | Temperature control in the heat exchanger  |
| <b>Node: Syrup to the activated carbon mixer tank</b> |                  |  |   |  |  |
| Flow  | Less             | Pipe leakage or obstruction                        | Possible excess of carbon                     | Operating procedure                              | Control in carbon adding   |
|   | More             | Syrup excess in tank TK04                          | Insufficient amount of carbon                 | idem   | idem   |
| Temperature   | Less             | Pipe leakage or obstruction<br>Pump malfunctioning | Inadequate decoloration<br>idem               |  | Temperature controller in the heat exchanger<br>idem   |
| <b>Node: Discharge to TK06</b>                        |                  |  |   |  |  |
| Flow  | No               | Mixer tank empty                                   | Lack of syrup with carbon to feed the filters |  | Low level alarm in TK06  |
| Solids concentration                                  | More             | Agitator broken                                    | Pipe obstruction                              |  | Shutdown alarm in the agitator   |
|   |                  | Carbon excess in the mixer                         | Solids deposit in pump                        |  | Control in carbon adding   |

| Filtration station                           |                  |  |                               |                         |  |
|--|------------------|--|-------------------------------|-------------------------|--|
| <b>Node: Suspension input to the filters</b> |                  |  |                               |                         |  |
| <i>Variable</i>                              | <i>Guideword</i> | <i>Causes</i>                                      | <i>Consequences</i>           | <i>Safeguards</i>       | <i>Recommendations</i>                         |
| Flow   | Less             | Pipe leakage or obstruction<br>Pump malfunctioning | Inadequate filtration<br>idem |                         | Pressure sensor to stop the filtration<br>idem |
|  |                  |  |                               |                         |  |
| <b>Node: Input to the tank</b>               |                  |  |                               |                         |  |
| Flow   | Less             | Pipe leakage or obstruction<br>Pump malfunctioning | Empty tank<br>idem            | Low level alarm<br>idem |  |
| <b>Node: Output from the tank</b>            |                  |  |                               |                         |  |
| Flow   | Less             | Pipe leakage or obstruction<br>Pump malfunctioning | Empty tank<br>idem            | Low level alarm<br>idem |  |

|  |                                     |  |   |                     |                             |
|--|-------------------------------------|--|---|---------------------|-----------------------------|
| Ionic exchange station                           |                                     |  |   |                     |                             |
| <b>Node: Input to the ionic exchangers</b>       |                                     |  |   |                     |                             |
| <i>Variable</i>                                  | <i>Guideword</i>                    | <i>Causes</i>                                      | <i>Consequences</i>                             | <i>Safeguards</i>   | <i>Recommendations</i>      |
| Flow   | Less                                | Pipe leakage or obstruction                        | Inadequate operation of a ionic exchange column | Alternative columns | Install flowmeters          |
|  |                                     | Pump malfunctioning                                | idem  | idem                | idem                        |
|  |                                     | Obstruction in a ionic exchanger column            | idem  | idem                | idem                        |
|  | More                                | Pump malfunctioning                                | Deficient decoloration                          | idem                | idem                        |
| Overloading of TK08                              | Alternative tank for filtered syrup |  |   |                     |                             |
| Temperature                                      | More                                | Excessive return of concentrated syrup to the TK08 | Ionic exchange resins damage                    |                     | Temperature control in TK08 |
| <b>Node: Decolorated syrup discharge to TK08</b> |                                     |  |   |                     |                             |
| Flow   | Less                                | Pipe leakage or obstruction                        | Inadequate operation of a ionic exchange column | Alternative columns | Install flowmeters          |
|  |                                     | Pump malfunctioning                                | idem  | idem                | idem                        |
|  |                                     | Obstrucción de una columna de intercambio          | idem  | idem                | idem                        |
|  | More                                | Pump malfunctioning                                | Deficient decoloration                          | idem                | idem                        |
|  | Overloading of TK08                 | Alternative tank for filtered syrup                |   |                     |                             |
| Reverse  | Pumps switch off and valves open    |  |   | Retention valves    |                             |

| Evaporation station                             |                  |  |   |                     |  |
|---|------------------|--|---|---------------------|--|
| <b>Node: From TK08 to evaporators</b>           |                  |  |   |                     |  |
| <i>Variable</i>                                 | <i>Guideword</i> | <i>Causes</i>  | <i>Consequences</i>   | <i>Safeguards</i>   | <i>Recommendations</i>   |
| Flow  | No               | Pipe leakage or obstruction  | Shutdown of evaporation section   |                     | Install flowmeters   |
|   |                  | Pump malfunctioning  | idem  |                     | idem   |
|   |                  | TK08 empty   | idem  |                     | idem   |
|   | Less             | Pipe leakage or obstruction  | Do not operate at full capacity   |                     | Additional diluted syrup supply from an alternative stock tank |
|   |                  | Pump malfunctioning  | idem  |                     | idem   |
|   | Other than       | Valve openness in the line to concentrated syrup tank  | Dilution of concentrated syrup  | Operating procedure | To verify the necessity of this connexion                      |
|   |                  | Valve openness in the line to the diluted syrup tank   | Possible overloading of the tank  | High level return   | idem   |
| <b>Node: Syrup outputs from each evaporator</b> |                  |  |   |                     |  |
| Flow  | No               | Obstruction in the output  | Overloading in the previous evaporator<br>Low level in the following evaporator | Maintenance<br>idem | Flow control<br>idem   |
| Temperature                                     | Less             | Excessive deposit of incrustations in tubes<br>Fat layers in the exterior of the tubes<br>Escape in the base of the calandry<br>Incondensables blocking<br>Water entrance to the steam tube<br>Fault in level sensor |   |                     |  |
|   | More             | Evaporator low level<br>Steam at a higher pressure   | Sugar grains crystallisation<br>Syrup coloration                                |                     | Low level alarm<br>Control in the heating steam pressure       |

|  |                  |  |   |   |                          |
|--|------------------|--|---|---|--------------------------|
| <b>Boiling station</b>                   |                  |  |   |   |                          |
| <b>STAGE: Loading</b>                    |                  |  |   |   |                          |
| <b>Node: Input of concentrated syrup</b> |                  |  |   |   |                          |
| <i>Variable</i>                          | <i>Guideword</i> | <i>Causes</i>  | <i>Consequences</i>   | <i>Safeguards</i>   | <i>Recommendations</i>   |
| Flow                                     | Less             | Low level in the concentrated syrup deposit  |   |   |                          |
|  |                  | Insufficient addition of syrup by the operator                                       | Caramelisation of the splashing of small drops on the heating surface | Operating procedure that indicates that the heating surface has to be covered | Level control            |
| <b>STAGE: Grain growth</b>               |                  |  |   |   |                          |
| <b>Node: Input of heating steam</b>      |                  |  |   |   |                          |
| Pressure                                 | Less             | High steam demand from the plant which generates a sudden decrease in steam pressure | False grain formation   | Temperature increase by the reduction of vacuum                               | Adequate scheduling      |
| <b>Node: Vacuum generator</b>            |                  |  |   |   |                          |
| Vacuum level                             | More             | Vacuum level increase  | idem  | Addition of diluted syrup or water  | Individual vacuum system |
| Vacuum level                             | Less             | Vacuum decrease  | Possible sugar coloration due to temperature                          | Addition of diluted syrup or water  | Individual vacuum system |
| <b>STAGE: Crystallisation</b>            |                  |  |   |   |                          |
| <b>Node: Vacuum generator</b>            |                  |  |   |   |                          |
| Seed Humidity                            | More             | Defficient quality control   | Bubbles formation   | Adequate storage of the seeds   |                          |

## CAICC - HAZOP analysis

| HAZOP ANALYSIS. REFINERY PLANT.                |            |          |   |   |                   |   |
|--|------------|----------|---|---|-------------------|---|
|  | GUIDE WORD | VARIABLE | DEVIATION                               | POSSIBLE CAUSES   | CONSEQUENCES      | RECOMMENDATIONS   |
| NODE # 1:<br>DISCHARGE OF RAW SUGAR TO MINGLER | NO         | FLOW     | LACK OF SUGAR IN THE MINGLER            | LACK OF SUGAR IN STOCK  | REFINERY SHUTDOWN | TO HAVE SUFFICIENT STOCK OF SUGAR   |
|  | NO         | FLOW     | LACK OF SUGAR IN THE MINGLER            | THICK DISCHARGE OF THE HOPPER                                   | REFINERY SHUTDOWN | TO IMPROVE CURRENT DISCHARGE SYSTEM                                       |
|  | NO         | FLOW     | LACK OF SUGAR IN THE MINGLER            | WEIGHT OF THE BAND FOR THE SUGAR BROKEN                         | REFINERY SHUTDOWN | PREVENTIVE MAINTENANCE OF EQUIPMENT                                       |
|  | MORE       | FLOW     | INADEQUATED REGULATION IN THE DISCHARGE | SHUTDOWN OF THE DRIVERS AND ELEVATORS OF SUGAR DUE TO EXTRALOAD | REFINERY SHUTDOWN | TO MODIFY THE SYSTEM OF PROTECTION FOR EXTRALOAD AND THE SUGAR REGULATION |

| HAZOP ANALYSIS. REFINERY PLANT.   |            |          |                               |  |   |   |
|-----------------------------------|------------|----------|-------------------------------|--|---|---|
|                                   | GUIDE WORD | VARIABLE | DEVIATION                     | POSSIBLE CAUSES  | CONSEQUENCES                                      | RECOMMENDATIONS   |
| NODE # 2:<br>MINGLED OF RAW SUGAR | NO         | FLOW     | LACK OF SUGAR                 | LACK OF SUGAR SUPPLY   | REFINERY SHUTDOWN                                 | TO HAVE SUFFICIENT STOCK OF SUGAR   |
|                                   | LESS       | TEMP.    | LOW TEMPERATURE IN THE SYRUP  | INSUFFICIENT STEAM OF HEATING FOR THE REFINERY SYRUP                     | LOW QUALITY IN REFINED SUGAR                      | TO IMPROVE THE REGULATION OF TEMPERATURE IN THE SYRUP TANK<br>TO INCREASE TIME OF CENTRIFUGATION      |
|                                   | MORE       | TEMP.    | HIGH TEMPERATURE IN THE SYRUP | EXCESS OF STEAM TO REFINERY  | DISSOLVED SUGAR IN EXCESS IN THE REFINERY PROCESS | TO IMPROVE THE REGULATION OF TEMPERATURE AND TO ADD LESS SYRUP  |
|                                   | LESS       | DENSITY  | EXCESS OF SYRUP IN THE SYSTEM | DENSITY REGULATION OUT OF CONTROL  | DISSOLVED SUGAR IN EXCESS IN THE REFINERY PROCESS | TO IMPROVE THE REGULATION OF THE DENSITY CONTROL LOOP   |
|                                   | MORE       | DENSITY  | LACK OF SYRUP IN THE SYSTEM   | LACK OF SYRUP IN THE SUPPLY TANK AND FAULT IN THE DENSITY CONTROL SYSTEM | MINGLER BROKEN                                    | TO IMPROVE THE REGULATION OF THE DENSITY CONTROL LOOP<br>TO INSTALL LOW LEVEL ALARM IN THE SYRUP TANK |

| HAZOP ANALYSIS. REFINERY PLANT. |            |              |  |  |                                  |  |
|---------------------------------|------------|--------------|--|--|----------------------------------|--|
|                                 | GUIDE WORD | VARIABLE     | DEVIATION  | POSSIBLE CAUSES  | CONSEQUENCES                     | RECOMMENDATIONS  |
| NODE # 3:<br>CENTRIFUGATION     | NO         | FLOW         | LACK OF MAGMA SUPPLY                               | LOCKING OF THE VALVE IN LINE THAT FEEDS MIXER                                | REFINERY SHUTDOWN                | TO IMPROVE PREVENTIVE MAINTENANCE TO THE VALVE   |
|                                 |            |              | SHUTDOWN CENTRIFUGES                               | DISCHARGE BLOCKED IN THE CENTRIFUGES DUE TO HIGH LEVEL IN THE PRE DISSOLUTOR | REFINERY SHUTDOWN                | TO REVISE AND TO IMPROVE LEVEL CONTROL AT THE PRE DISSOLUTOR.<br>TO IMPROVE MAINTENANCE IN THE PUMP OF THE PRE DISSOLUTOR. |
|                                 | LESS       | POL IN SUGAR | DEFICIENT MAGMA QUALITY                            | MAGMA NOT FLUENT DUE TO THE LACK OF SYRUP                                    | LOW PRODUCTIVITY OF THE REFINERY | TO EXAMINE THE PREPARATION OF MAGMA  |
|                                 |            |              | INSUFFICIENT CLEANING WATER IN CENTRIFUGES         | CLEANING CYCLE IN CENTRIFUGES OUT OF CONTROL                                 | LOW PRODUCTIVITY OF THE REFINERY | TO REVISE THE AMOUNT OF CLEANING WATER   |
|                                 |            |              | INSUFFICIENT TEMP. IN THE CLEANING WATER OF CENTR. | WATER TEMPERATURE REGULATION OUT OF CONTROL                                  | LOW PRODUCTIVITY OF THE REFINERY | TO INSTALL AN ALARM FOR LOW TEMPERATURE IN WATER   |
|                                 |            |              | LOW VELOCITY OF CENTRIFUGATION                     | INSUFFICIENT TIME OF CENTRIFUGING IN THE CYCLE                               | LOW PRODUCTIVITY OF THE REFINERY | TO REVISE THE TIME OF CENTRIFUGING   |



| HAZOP ANALYSIS. REFINERY PLANT. |            |                     |                                    |  |                                     |                                    |
|---------------------------------|------------|---------------------|------------------------------------|--|-------------------------------------|------------------------------------|
|                                 | GUIDE WORD | VARIABLE            | DEVIATION                          | POSSIBLE CAUSES                          | CONSEQUENCES                        | RECOMMENDATIONS                    |
| NODE # 3:<br>CENTRIFUGATION     | MORE       | REFINED SUGAR COLOR | DEFICIENT CLEANING IN THE SUGAR    | INSUFFICIENT CLEANING IN THE CENTRIFUGES | EFFECTS ON QUALITY OF REFINED SUGAR | TO REVISE THE TIME OF CENTRIFUGING |
|                                 |            |                     | DEFICIENT PREPARATION OF THE MAGMA | EXCESS OF DENSITY IN THE MAGMA           | EFFECTS ON QUALITY OF REFINED SUGAR | TO REVISE THE QUALITY OF THE MAGMA |

| HAZOP ANALYSIS. REFINERY PLANT.       |            |          |   |  |                      |  |
|---------------------------------------|------------|----------|---|--|----------------------|--|
|                                       | GUIDE WORD | VARIABLE | DEVIATION   | POSSIBLE CAUSES  | CONSEQUENCES         | RECOMMENDATIONS  |
| NODE # 4:<br>DISSOLUTION OF RAW SUGAR | NO         | FLOW     | LACK OF PRE-DISSOLVED SUGAR                                 | LACK OF SUPPLY OF RAW SUGAR DUE TO INTERRUPTION IN CENTRIFUGES | SHUTDOWN REFINERY    | TO REVISE THE PROCESS OF CENTRIFUGATION                            |
|                                       |            |          | LACK OF PRE-DISSOLVED SUGAR                                 | THICK SUPPLY LINE  | SHUTDOWN DISSOLUTION | TO LOOK FOR AN ALTERNATIVE SOLUTION                                |
|                                       |            |          | LACK OF PRE-DISSOLVED SUGAR                                 | VALVE OF RECIRCULATION BLOCKED                                 | SHUTDOWN DISSOLUTION | TO IMPROVE THE SYSTEM OF FILTERING                                 |
|                                       |            |          | SHUTDOWN OF DISSOLUTION DUE TO HIGH LEVEL IN THE DISSOLUTOR | DEFICIENT PUMPING SYSTEM (23 & 24) OF THE CRUDE LIQUOR         | SHUTDOWN DISSOLUTION | TO IMPROVE THE MAINTENANCE SYSTEM TO IMPROVE THE SYSTEM OF FILTERS |
|                                       |            |          | THICK PUMPING SYSTEM  | AGITATOR BROKEN  | SHUTDOWN DISSOLUTION | TO REVISE AND IMPROVE THE MAINTENANCE OF AGITATORS                 |
|                                       |            |          | HIGH LEVEL IN THE TANK                                      | PUMPING BLOCKED DUE TO THE FULLNESS OF TANK 200                | SHUTDOWN DISSOLUTION |  |

| HAZOP ANALYSIS. REFINERY PLANT.       |            |          |                                      |  |  |   |
|---------------------------------------|------------|----------|--------------------------------------|--|--|---|
|                                       | GUIDE WORD | VARIABLE | DEVIATION                            | POSSIBLE CAUSES  | CONSEQUENCES   | RECOMMENDATIONS   |
| NODE # 4:<br>DISSOLUTION OF RAW SUGAR | LESS       | TEMP.    | INSUFFICIENT STEAM IN THE DISSOLUTOR | BAD FUNCTIONING OF THE REGULATION SYSTEM FOR THE TEMPERATURE | INSUFFICIENT DISSOLUTION PRESENCE OF CRYSTALS IN THE DISSOLUTION | TO REVISE AND IMPROVE TH MAINTENANCE OF THE REGULATION OF TEMP.                   |
|                                       |            |          | WATER WITH INSUFFICIENT TEMPERATURE  | INSUFFICIENT WATER HEATING IN THE SUPPLY TANK                | INSUFFICIENT DISSOLUTION PRESENCE OF CRYSTALS IN THE DISSOLUTION | TO IMPROVE THE HEATING OF WATER AND TO INSTALL AN ALARM FOR LOW WATER TEMPERATURE |
|                                       | MORE       | TEMP.    | STEAM EXCESS IN THE DISSOLUTOR       | TEMPERATURE REGULATOR OUT OF CONTROL                         | CARAMELISATION OF THE SUGAR                                      | TO IMPROVE TEMPERATURE CONTROL  |
|                                       | LESS       | DENSITY  | EXCESS OF WATER                      | DENSITY REGULATOR OUT OF CONTROL                             | CLARIFICATION DEFICIENT AND EXCESSIVE STEAM CONSUMPTION          | TO IMPROVE DENSITY CONTROL  |
|                                       |            |          | PRE-DISSOLUTION WITH LOW DENSITY     | DEFICIENT DISSOLUTION OF THE SUGAR IN THE GONDOLAS           | CLARIFICATION DEFICIENT AND EXCESSIVE STEAM CONSUMPTION          | TO IMPROVE THE MANIPULATION OF THE SYSTEM   |
|                                       | MORE       | DENSITY  | LACK OF WATER                        | DENSITY REGULATOR OUT OF CONTROL                             | THICK FLOW IN LINES AND PUMPS                                    | TO IMPROVE THE SYSTEM OF REGULATION   |



## Publications and Conferences

### **Control strategies applied to a chemical plant with interacting control loops**

Ruiz D., Nogués J.M. and Puigjaner L. (1999).

Journal: *Hungarian Journal of Industrial Chemistry* (HU ISSN: 0133-0276), Vol. **27**, pp 73-79.

Presented at the 4th International Workshop on Mathematical Modelling in Chemical Engineering, in Bad Honnef (Germany), August 10-14, 1998.

Keywords: multivariable control, internal model control structure, neural networks

#### Abstract

Interaction among control loops in multivariable systems has been the subject of much research in the last decades. Distillation columns that are widely used in the process industries are the typical example of these systems. In this work, Wood and Berry test-bed problem has been used due to the strong interaction among the control variables. Therefore, different control strategies have been applied to this binary distillation column with two interacting control loops. A diagonal controller structure with proportional-integral controllers using several tuning methods has been evaluated. Two Internal Model Control structures have been implemented utilising in one of them a neural network model of the plant. Model identification by neural networks has also been presented.

### **On-line process fault detection and diagnosis in plants with recycle**

Ruiz D., Nogués J.M. and Puigjaner L. (1999).

Journal: *Computers and Chemical Engineering* (ISSN: 0098-1354), Vol. **23S**, pp S219-S222.

Presented at the 9th European Symposium on Computer Aided Process Engineering (ESCAPE 9), in Budapest (Hungary), May 8-11, 1999.

Keywords: fault diagnosis, plants with recycle, artificial neural networks, fuzzy logic

#### Abstract

A process fault detection and diagnosis system is performed for the complex case of plant-wide control in processes with recycle in which the control system is the inventory control. It is considered an artificial neural network based supplement of a fuzzy system in a block-oriented configuration. A methodology for designing the system is described. As a case study, a

chemical plant with a recycle stream is considered. Faults in the supply of raw materials and in controllers are simulated. The performance of the system in handling simultaneous faults is also analysed. A comparison is made against a classification method (artificial neural networks) and an inference method (knowledge - based system).

### **Artificial Neural Networks applied to on-line fault diagnosis in chemical plants**

Ruiz D., Nougues J.M. and Puigjaner L. (1999).

Proceedings 7th IEEE International Conference on Emerging Technologies and Factory Automation, Editor: J. M. Fuertes, (ISBN 0-7803-5670-5), Vol. 2, pp 977-986.

Presented at the 7th IEEE International Conference on Emerging Technologies and Factory Automation, Barcelona (Spain), October 18-22, 1999.

Keywords: artificial neural networks, fault diagnosis, chemical plants

#### **Abstract**

Different kinds of artificial neural networks are compared regarding with their application to the fault diagnosis in steady state chemical processes. Their performance is studied taking into account the influence of some design parameters. Faults in sensors are considered separately by using auto-associative neural networks and a proposed algorithm. The developments have been applied to two case studies. The first one corresponds to a chemical plant with recycle. The second one has been carried out in a fluidised bed coal gasifier, at a pilot plant scale. In this last case, performance of the selected and optimised neural network approach is compared with a statistical technique: Principal Component Analysis. The methodology of implementation and optimisation of the artificial neural network approach for fault diagnosis shows promising results. This approach can be used to complement a knowledge-based approach for robust fault detection and diagnosis in chemical plants.

### **Integrated information system for monitoring, scheduling and control applied to batch chemical processes**

Canton J., Ruiz D., Benqlilou Ch., Nougues J.M. Puigjaner L. (1999)

Proceedings 7th IEEE International Conference on Emerging Technologies and Factory Automation, Editor: J. M. Fuertes, (ISBN 0-7803-5670-5), Vol. 1, pp 211-217.

Presented at the 7th IEEE International Conference on Emerging Technologies and Factory Automation, Barcelona (Spain), October 18-22, 1999.

Keywords: batch chemical plants, information system

#### **Abstract**

The effective operation of the batch processing based enterprise requires exploitation of computer integrated manufacturing (CIM) developments and the adaptation of these developments for the specific features of batch operations. The aim of this work is to present a general overview of a global methodology which allows to find a practical solution for the control of a multipurpose batch chemical plant. A case study carried out in a pilot plant for the simulation of batch chemical processes is also described.

### **On-line process fault diagnosis system for chemical process industries**

Ruiz D., Nogués J.M. and Puigjaner L. (1999)

Presented at the American Institute of Chemical Engineers Meeting '99 (Paper 208ae), in Dallas (U.S.A.), October 31 - November 5, 1999.

Keywords: fault diagnosis, artificial neural networks, industrial applications

#### **Abstract**

A process fault detection and diagnosis (PFD&D) system has been applied to different cases of chemical plants. It considers an artificial neural network (ANN) structure supplemented with a fuzzy system in a block-oriented configuration. The system combines the adaptive learning diagnostic procedure of the ANN and the transparent deep knowledge representation of a structured form of knowledge based expert system. This work is focused on the ANN block. Different ANN architectures have been compared. Back-propagation networks show to be less effective than radial basis function networks in diagnosing process and controllers' faults. Self-organising maps show low resolution but they can be useful to handle unsuspected faults. A performance index for comparison purposes is proposed. False diagnosis is also considered. The ANN is trained with steady state conditions of past faults. Faults in sensors are also contemplated. In this last case Nonlinear Principal Component Analysis is implemented. It uses an auto-associative neural network. Influence of the number of neurons in the bottleneck layer is studied. A comparison against conventional Principal Component Analysis is shown. The features and the performance of the PFD&D system are presented in several scenarios: academic, pilot plant scale and industrial cases. The academic case consists of a chemical plant with a recycle stream. The pilot plant scale case consists of a fluidised bed coal gasifier. One industrial case contemplates a real plant of linear alquil benzene. Another industrial case corresponds to a real sugar cane refinery factory. Future work is related to the integration of the PFD&D system to the scheduling and management levels.

### **Neural computation for abnormal situation management in a cane sugar refinery**

Ruiz D., Nogués J.M. and Puigjaner L. (1999)

Presented at the American Institute of Chemical Engineers Meeting '99 (Paper 51g), in Dallas (U.S.A.), October 31 - November 5, 1999.

Keywords: fault diagnosis, artificial neural networks, sugar industry

#### **Abstract**

In this paper, a Fault Diagnosis System (FDS) is applied to a real sugar cane refinery factory. The FDS is composed by an Artificial Neural Network (ANN) structure supplemented with a Knowledge Based Expert System (KBES) in a block-oriented configuration. The system combines the adaptive learning diagnostic procedure of the ANN and the transparent deep knowledge representation of the KBES. The information needed to design and implement the FDS includes a historical database, a Hazard and Operability study (HAZOP) and a model of the plant. The refinery sugar process encloses raw sugar dissolution, syrup treatment, boiling, crystallisation, centrifuging and sugar drying. Some examples of the system performance are shown. They are related to the Boiling step. In this step, the crystallisation of saccharose begins. Its adequate control has strong influence in quality product. Pressure steam to the boilers and the vacuum level must be constants. The ANN block of the FDS shows to be efficient by anticipating faults. It takes advantage of the historical database of past faults. On the other hand, the use of the KBES allows a transparent representation of the knowledge of the plant. In addition, the quick corrective actions implementation is straightforward. By this way the FDS can be robust enough to manage abnormal situations. The development is very important to increase the efficiency in the sugar industry.

### **Monitoring and fault diagnosis of batch processes using a neuro-fuzzy approach**

Ruiz D., Nougues J.M. and Puigjaner L. (1999)

Presented at the 8th Mediterranean Congress of Chemical Engineering (Session: Process systems engineering), in Barcelona (Spain), November 10-12, 1999.

Keywords: batch process, process monitoring, fault diagnosis, neural networks

#### **Abstract**

A methodology to design a Fault diagnosis system (FDS) for batch chemical plants is presented. The FDS is composed by an Artificial Neural Network (ANN) structure supplemented with an inference system in a block oriented configuration. The system combines the adaptive learning diagnostic procedure of the ANN and the transparent deep knowledge representation of the expert system. The proposed FDS is also compared with other approach based on multiway Principal Component Analysis. The FDS's potential is illustrated using a pilot plant and an industrial case study.

### **An integrated architecture for information management in batch chemical processes**

Cantón J., Ruiz D., Benqlilou Ch., Nougues J.M. and Puigjaner L. (1999)

Presented at the 8th Mediterranean Congress of Chemical Engineering (Session: Process systems engineering), in Barcelona (Spain), November 10-12, 1999.

Keywords: batch process, process integration, modeling and simulation, process management and control

#### **Abstract**

The effective operation of the batch processing based enterprise requires exploitation of computer integrated manufacturing (CIM) developments and the adaptation of these developments for the specific features of batch operations. The aim of this work is to present a general overview of a global methodology which allows to find a practical solution for the control of a multipurpose batch chemical plant.

### **A Hybrid Neural Network-First Principles Approach for Process Modeling**

Benqlilou Ch., Ruiz D., Nougues J.M. and Puigjaner L. (1999)

Presented at the 8th Mediterranean Congress of Chemical Engineering (Session: Process systems engineering), in Barcelona (Spain), November 10-12, 1999.

Keywords: hybrid modelling, neural networks

#### **Abstract**

This work is based mainly on the use of Artificial Neural Network (ANN) in collaboration with the a priori knowledge of the process for modelling. The motivation to combine the parametric and non parametric model comes from different properties of two models. While ANN has good approximation and interpolation properties it show limited extrapolation capacity. On the other hand First Principles Model (FPM) is limited to deal with non linearities but often demonstrate more robust extrapolation behaviour. Whereas only modelling of the process is discussed in this work our goal is to proceed in the direction of process control and particularly to a model based predictive control scheme. The structure investigated for the purpose of process modelling is a hybrid parallel model referred to a non-linear error correction. The effective operation of the batch processing based enterprise requires exploitation of computer integrated manufacturing (CIM) developments and the adaptation of these developments for the specific features of batch operations. The aim of this work is to present a general overview of a global methodology which allows to find a practical solution for the control of a multipurpose batch chemical plant.

### **Teaching the plant-wide control complexity from the beginning**

Basualdo M., Ruiz D., Sequeira S. and Pedridio J. (1999)

Presented at the 8th Mediterranean Congress of Chemical Engineering (Session: Process systems engineering), in Barcelona (Spain), November 10-12, 1999.

Keywords plant-wide control, recycles

#### Abstract

This paper examines the practical use of simulation, within process plant-wide control application areas, and the benefits released by the undergraduate engineering students. In an environment of change, as the control of a chemical process, it offers an attractive tool to teach more efficiently the "real" control problem. The discussion draws on a study of the employment of simulation, which serves as an authentic pilot plant, to design conventional SISO and MIMO feedback controllers implemented over a typical plant which contains both separation and reaction stages. The overall steps from the identification of that system, including the design and tuning of the controllers conform a good practical work. The student can verify, by using dynamic simulation, which is the "cost" of obtaining bad models which drive to bad controller designs. Several proofs, for load and reference changes, are carried out by designing PID and PI for SISO bottom composition control over the separation stage. Then, this design is affected with other loops included in the stripper and finally the effect of including the reactor for conforming the plant is analysed. The need for retuning the controllers in order to improve the overall performance is shown. For SISO case several tuning methodologies are tested including model based statements. For MIMO problem the methodology proposed by Belanger and Luyben (1997) for controlling a plant with recycle is used.

### **Sistema de diagnosis de fallos en plantas químicas (*Fault diagnosis system in chemical plants*)**

Ruiz D., Nogués J.M. and Puigjaner L. (1999).

Proceedings 3th Users Meeting MATLAB 99, Editor: S. Dormido, (ISBN 84-699-1358-1), pp 297-302.

Presented at the 3th Users Meeting MATLAB 99, in Madrid (Spain), November 17-19, 1999.

Keywords: fault diagnosis, neural networks, fuzzy logic, communication via DDE, applications

#### Abstract

In this work the usefulness of MATLAB for the development of a fault diagnosis system applied to chemical plants is shown. The fault diagnosis system used consists in a combination of an artificial neural network and a fuzzy logic system. An example is shown where the program LABWINDOWS CVI is utilised for the simulation of a chemical plant and the communication via Dynamic Data Exchange (DDE) with the fault diagnosis system programmed in MATLAB is utilised.

### **Fault diagnosis system support for reactive scheduling in multipurpose batch chemical plants**

Ruiz, D., Nogués, J. M., Cantón J., Espuña A. and Puigjaner, L. (2000).

Book chapter: Computer Aided Chemical Engineering Series, Vol. 8 (Editor: S. Pierucci), Elsevier, Amsterdam, ISBN: 0-444-50520-2, pp 745-750.

Presented at the 10th European Symposium on Computer Aided Process Engineering (ESCAPE 10), in Florence (Italy), May 7-10, 2000.

Keywords: fault diagnosis, scheduling, batch plants

#### **Abstract**

In this work, a simple strategy for the development and implementation of a Fault Diagnosis System (FDS) that interacts with a schedule optimiser in batch chemical plants is presented. The proposed FDS consists in an Artificial Neural Network (ANN) structure supplemented with a Knowledge Based Expert System (KBES) in a block-oriented configuration. The information needed to implement the FDS includes a historical database of past batches, a Hazard and Operability (HAZOP) analysis and a model of the plant. A motivating case study is presented to show the results of the proposed methodology.

### **The use of process dynamic simulation for learning to design digital controllers**

Basualdo M., Salcedo J.B. and Ruiz, D. (2000)

Book chapter: Computer Aided Chemical Engineering Series, Vol. 8 (Editor: S. Pierucci), Elsevier, Amsterdam, ISBN: 0-444-50520-2, pp 259-264.

Presented at the 10th European Symposium on Computer Aided Process Engineering (ESCAPE 10), in Florence (Italy), May 7-10, 2000.

Keywords: dynamic simulation, digital controllers, education

#### **Abstract**

The discussion presented in this paper draws on a comparative study based on several design techniques of digital feedback controllers implemented as SISO structure of a binary distillation column. By using rigorous process models, which serves as an authentic pilot plant, offers an attractive tool to learn more efficiently the "real" control problem. It must be noted that many textbooks have presented several design techniques about this subject but they did not compare over chemical processes. Generally the overall conclusions are based on linear transfer functions only. Chemical processes represent a real challenger for developing efficient digital control design techniques. The overall steps from the identification of the nonlinear, with dead time and inverse response system, including the design and tune of the controllers are presented. Hence, is easier to verify which is the "cost" of obtaining bad models which drive to bad controller designs. Several proofs, for load and reference changes are carried out by designing discrete controllers such as PID, Ragazzini and W transform methodologies by using MATLAB-SIMULINK software. The rigorous model of the distillation column is developed through an S-function of SIMULINK.



### **Neural network based framework for fault diagnosis in batch chemical plants**

Ruiz D., Nogués J.M., Calderón Z., España A. and Puigjaner L. (2000).

Journal: *Computers and Chemical Engineering* (ISSN: 0098-1354), Vol. **24** (2-7), pp 777-784.

Presented at the 7th International Symposium on Process Systems Engineering 2000 (PSE 2000), in Keystone - Colorado (U.S.A.), July 16-21, 2000.

Keywords: fault diagnosis, batch plants, artificial neural networks

#### Abstract

In this work, an Artificial Neural Network (ANN) based framework for Fault Diagnosis in batch chemical plants is presented. The proposed FDS consists in an Artificial Neural Network (ANN) structure supplemented with a Knowledge Based Expert System (KBES) in a block-oriented configuration. The system combines the adaptive learning diagnostic procedure of the ANN and the transparent deep knowledge representation of the KBES. The information needed to implement the FDS includes a historical database of past batches, a Hazard and Operability (HAZOP) analysis and a model of the batch plant. The historical database that includes information related with normal and abnormal operating conditions is used to train the ANN structure. The deviations of the on-line measurements from a reference profile are processed by a multi-scale wavelet in order to determine the singularities of the transients and to reduce the dimensionality of the data. The processed signals are the inputs of an ANN. The ANN's outputs are the signals of the different suspected faults. The HAZOP analysis is useful to build the process deep knowledge base (KB) of the plant. This base relies on the knowledge of the operators and engineers about the process and allows formulating artificial intelligence algorithms. The case study corresponds to a batch reactor. The FDS performance is demonstrated through the simulation of different process faults. The FDS proposed is also compared with other approach based on multi-way Principal Component Analysis.

### **Integration of Fault Diagnosis and Reactive Scheduling in Batch Chemical Plants**

Ruiz, D., Nogués, J. M., Cantón J., España A. and Puigjaner, L. (2000)

Presented at the 3rd Conference PRES 2000 - Process integration, modelling and optimisation for energy saving and pollution reduction (Ref. H6.4), in Praha (Czech Republic), August 27-31, 2000.

Keywords: fault diagnosis, scheduling, batch plants

#### Abstract

This paper is a considerable extension of the ideas presented previously (Ruiz et al., 2000a). A simple strategy for the development and implementation of a Fault Diagnosis System (FDS) that interacts with a schedule optimiser in batch chemical plants is shown, taking into account the energy savings. The proposed FDS consists in an Artificial Neural Network (ANN) structure supplemented with a Knowledge Based Expert System (KBES) in a block-oriented configuration. The system combines the adaptive learning diagnostic procedure of the ANN and the transparent deep knowledge representation of the KBES. The information needed to implement the FDS includes a historical database of past batches, a Hazard and Operability (HAZOP) analysis and a model of the plant. Two motivating case studies are presented to show the results of the proposed methodology. The first one corresponds to a fed-batch reactor. In this example, the FDS performance is demonstrated through the simulation of different process faults. The second case study corresponds to a multipurpose batch plant. In this case, the results of the reactive scheduling are shown by simulating different abnormal situations. A performance comparison is made against the traditional scheduling approach without the support of the proposed FDS. Large savings in energy and economic resources can be obtained.

### **Enhancement of Prediction and Extrapolation Properties By using Hybrid Neural Network Model**

Benqlilou Ch., Ruiz D., Espuña A. and Puigjaner L. (2000)

Presented at the 3rd Conference PRES 2000 - Process integration, modelling and optimisation for energy saving and pollution reduction (Ref. H3.7), in Praha (Czech Republic), August 27-31, 2000.

Keywords: dynamic modelling, hybrid neural networks, prediction, extrapolation

#### Abstract

In this work a comparison between parallel Hybrid Neural Network models (HNN), Artificial Neural Network (ANN) models and a First Principal Models (FPM) is performed in terms of their extrapolation and predictions properties. As motivation example a simple but highly non-linear dynamic process, the dynamic response of pH in a Stirred Continuous Tank Reactor has been chosen.

### **One experience of teaching the pass from analog to discrete PID designs**

Basualdo M., Boselli I. and Ruiz D. (2000)

Proceedings of the the 17th Argentinean Congress on Automatic Control, pp 117-122.

Presented at the 17th Argentinean Congress on Automatic Control, in Buenos Aires (Argentina), September 11-13, 2000.

Keywords: digital controller designs, nonlinear system, dynamic simulation

#### Abstract

The discussion presented in this paper draws on a study of how to teach more friendly and effectively the pass from analog to digital PID designs. By using rigorous process models, which serves as an authentic pilot plant, offers an attractive tool to learn more efficiently the "real" control problem. It must be noted that many textbooks have presented several design techniques about this subject but they did not compare over chemical processes. Generally, the overall conclusions are based on linear transfer functions only. Chemical processes represent a real challenge for developing efficient digital control design techniques. The overall steps from the identification of the nonlinear, with dead time system, including the design and tuning of the controllers are presented. Hence, it is easy to verify which is the "cost" of obtaining bad models which drive to bad controller designs. Several proofs, for load and reference changes are carried out by designing discrete PID controllers accounting different algorithms and sampling periods. This work is done as a part of series of practical works by using the same plant implemented in MATLAB-SIMULINK software.

### **A proposal to speed-up the implementation of an Abnormal Situation Management framework into real chemical plants**

Ruiz D., Nougués J.M. Benqlilou Ch., Ruiz C. and Puigjaner L. (2000)

Presented at the American Institute of Chemical Engineers Meeting 2000 (Paper 255b), in Los Angeles (U.S.A.), November 12-17, 2000.

Keywords: fault diagnosis, industrial applications, GCO

#### Abstract

The aim of this work is to present a proposal of implementation in a real industrial plant of a support framework for abnormal situation management. The basic strategy is: to simplify, to automate and to integrate. The main feature of the technology developed is that it takes advantage of existing software packages that are familiar to plant engineers (e.g. Plant Information System) and a commercial process simulator. Based on three sources of

information (a historical database, a HAZOP analysis and a regular plant model), the support framework is developed and easily implemented into the real plant. It consists of a pre-processing module which performs a variety of key tasks using plant data such as Trend generation, Principal Component Analysis, Data Reconciliation, filtering and denoising. Some of the outputs of this pre-processing module are the inputs of the Fault Diagnosis System (FDS). This FDS is a combination of a pattern recognition approach based on neural networks and a Fuzzy Logic System (FLS) in a block oriented configuration. The FLS has also the function of alarm handling. The outputs of the FDS can be used by an advanced control module in order to take control actions, or by the operators who are responsible for decision-making or by other levels in the information system as the scheduling system. The case study corresponds to a real petrochemical plant consisting in a train of distillation columns where a group of n-paraffines are separated from kerosene. Fluctuations in pressure of the hot-oil to the reboilers have been considered as suspected faults. The real plant has been simulated with HYSYS, using the DCS driver to allow communication with other applications. In this case, the other application is MATLAB, where the pre-processing system and the FDS are run. The pattern recognition approach can be developed using NerOnline studio software. This package was useful to perform data analysis, neural network training and validation. Integration of the developed platform with the existing hardware and software in our facilities is underway. Future work includes the application of the presented technology in a sugar cane refinery.

### **Fault diagnosis support system for complex chemical plants**

Ruiz D., Nogués J.M. and Puigjaner L. (2001).

Journal: *Computers and Chemical Engineering* (ISSN: 0098-1354), Vol. **25**, pp 151-160.

Keywords: fault diagnosis, plants with recycle, artificial neural networks, fuzzy logic

#### **Abstract**

A process fault detection and diagnosis system (PFD&D) is proposed for complex chemical plants. The system combines an artificial neural network based supplement of a fuzzy system in a block-oriented configuration. A methodology for designing the system is described. As a motivating example, a chemical plant with a recycle stream is considered. Faults in the supply of raw materials and in controllers are simulated. The performance of the system in handling simultaneous faults is also analysed. A comparison of the proposed approach is made with a classification method (artificial neural networks) and inference methods (knowledge - based system). Results of system implementation in a fluidised bed coal gasifier at pilot plant scale are also shown.

### **On-line fault diagnosis system support for reactive scheduling in multipurpose batch chemical plants**

Ruiz, D., Nogués, J. M., Cantón J., España A. and Puigjaner, L. (2001)

Journal: *Computers and Chemical Engineering* (in press)

Keywords: fault diagnosis, scheduling, batch plants

#### **Abstract**

In this work, a simple strategy for the development and implementation of a Fault Diagnosis System (FDS) that interacts with a schedule optimiser in batch chemical plants is presented. The proposed FDS consists in an Artificial Neural Network (ANN) structure supplemented with a Knowledge Based Expert System (KBES) in a block-oriented configuration. The system combines the adaptive learning diagnostic procedure of the ANN and the transparent deep knowledge representation of the KBES. The information needed to implement the FDS includes a historical database of past batches, a Hazard and Operability (HAZOP) analysis and a model of the plant. Two motivating case studies are presented to show the results of the proposed methodology. The first one corresponds to a fed-batch reactor. In this example, the FDS

performance is demonstrated through the simulation of different process faults. The second case study corresponds to a multipurpose batch plant. In this case, the results of reactive scheduling are shown by simulating different abnormal situations. A performance comparison is made against the traditional scheduling approach without the support of the proposed FDS.

### **Dynamic Cross-Functional Factory-to-Business Links in the Batch Industry**

Badell M., Ruiz D. and Puigjaner L. (2001)

To be presented at the 11th European Symposium on Computer Aided Process Engineering (ESCAPE-11), in Scanticon Comwell Kolding (Denmark), May 27-30, 2001.

Keywords: Enterprise Resource Planning system, short-term planning, financial and production cross-functional link, Web based systems, fault diagnosis system

#### **Abstract**

The lack of a cross-functional factory-to-business link between the shop floor and the necessary supply chain relation to e-business creates a gap. In order to bridge this gap a web-business-plant route is developed in a pilot plant using the TicTacToe sequencing algorithm. A web-based order management system is created to generate optimal plans taking into account the factory logistic status and detailed information of the real plant through a fault diagnosis system (FDS) with re-scheduling capabilities.