

A Medium Term Multireservoir Hydrothermal Coordination Model by Stochastic Programming

A. Ramos, S. Cerisola, J.M. Latorre
Universidad Pontificia Comillas
R. Bellido, A. Perea
Iberdrola

Contents

- > Introduction
- > System modeling
- Model description
- Case study
- Conclusions



Renaissance of hydro scheduling models

- Nowadays, under a deregulated framework electric companies manage their own generation resources and need detailed operation planning tools
- In the next future, high penetration of intermittent generation is going to force the electric system operation
- Hydro and storage hydro plants are going to play a much more important role due to their flexibility and complementary use with intermittent generation



Medium term model (i)

- Hydroelectric model deals only with hydro plants
- Hydrothermal model manages simultaneously both hydro and thermal plants
- Thermal units considered individually. So rich marginal cost information used for guiding hydro scheduling
- Hydro plants considered also individually. No aggregation or disaggregation process for hydro input and output is needed
- Obtain a feasible solution for each hydro plant is very difficult because the problem requires a huge amount of data and by the complexity of hydro subsystems



Medium term model (ii)

- Determines:
 - the optimal yearly operation of all the thermal and hydro power plants
 - taking into account multiple basins and multiple cascaded reservoirs connected among them
- Cost minimization model because the main goal is medium term hydro operation
- Upper level: stochastic market equilibrium model
- Lower level: daily simulation model



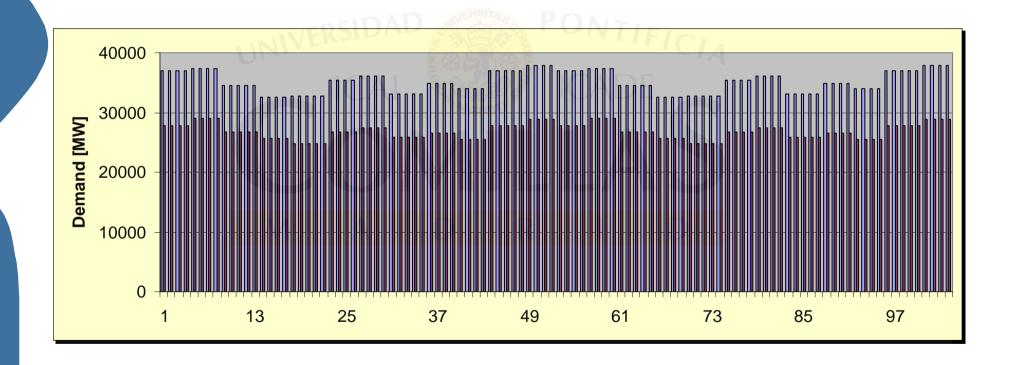
Contents

- > Introduction
- > System modeling
- Model description
- Case study
- Conclusions



Demand

 Weekly demand with two load levels (peak and off-peak each week)





Hydro subsystem

- Different modeling approach for hydro reservoirs depending on:
 - Owner company
 - Relevance of the reservoir
- Reservoirs belonging to other companies modeled in energy units [GWh]
- Own reservoirs modeled in water units [hm³, m³/s]
- Important reservoirs modeled with water head effects
- Very diverse hydro subsystem:
 - Hydro reservoir volumes from 0.15 to 2433 hm³
 - Hydro plant capacities from 1.5 to 934 MW



Stochasticity sources

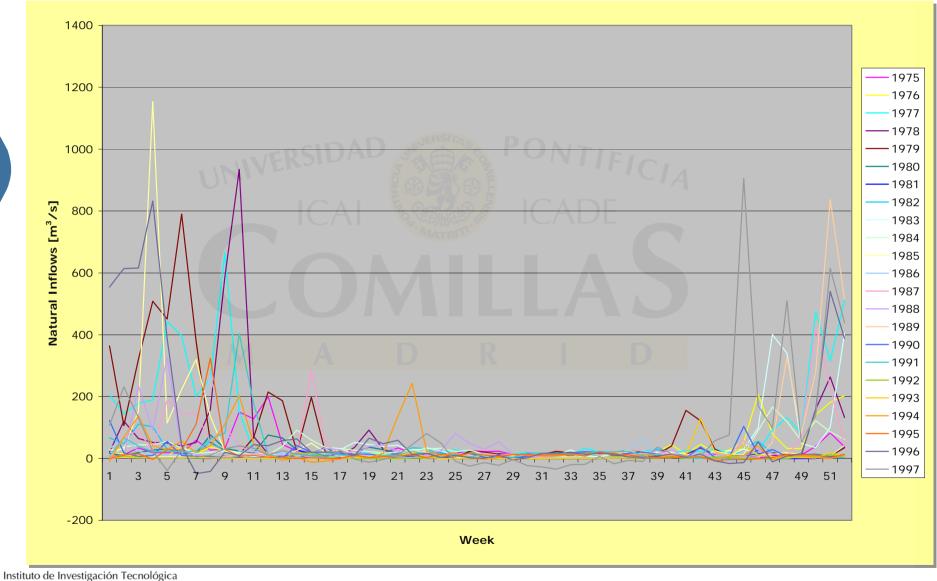
 Natural hydro inflows (clearly the most important factor in Spanish electric system)

Year	Hydro energy	Index	Probability of being
	Available [TWh]	ERSITAS	exceeded [%]
2001	32.9	1.13	32
2002	20.9	0.72	87
2003	33.2	1.15	30
2004	22.7	0.79	80
2005	12.9	0.45	100
2006	24.0	0.83	70

- Changes in reservoir volumes are significant because of:
 - Stochasticity in hydro inflows
 - Seasonal pattern of inflows and
 - Capacity of the reservoir with respect to the inflows



Historical natural inflows



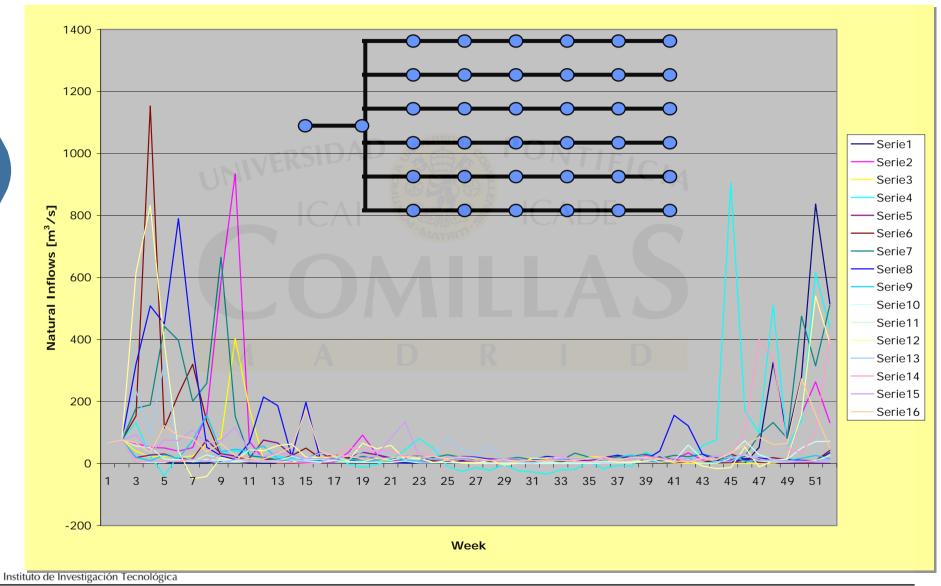


Scenario tree generation

- A multivariate scenario tree is obtained by neural gas clustering technique that simultaneously takes into account the main stochastic series and their spatial and temporal dependencies.
- Very extreme scenarios can be artificially introduced with a very low probability
- Number of scenarios generated enough for medium term operation planning



Natural inflows: scenario tree





Contents

- > Introduction
- System modeling
- Model description
- Case study
- Conclusions



Constraints: Generation and load balance

Generation of thermal units

- + Generation of storage hydro units
- Consumption of pumped hydro units
- = Demand



Constraints: Minimum and maximum operating hours of thermal units

- Introduced to model:
 - Unavailability of thermal units
 - Domestic coal subsidies
 - CO2 Emission allowances
 - Capacity payments
- They are not separable by period

 $minimum \leq Yearly operation hours of each thermal unit for each scenario \leq maximum$

 $minimum \le Average \ yearly \ operation \ hours \ of \ each$ $thermal \ unit \le maximum$



Constraints: Water balance for large reservoirs

Reservoir volume at the beginning of the period

- + Natural inflows
- Spills from this reservoir
- + Spills from upstream reservoirs
- + Turbined water from upstream storage hydro plants
- + Pumped water from downstream pumped hydro plants
- Turbined and pumped water from this reservoir
- = Reservoir volume at the end of the period



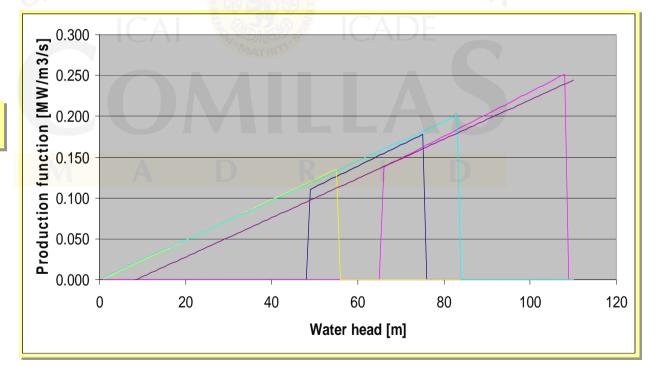
Constraint: Water head effects

• Power generation is the product (nonlinear function) of the flow and the production function $P = O \times PF$

Production function PF depends linearly on plant water

head

$$PF = \alpha H_p$$

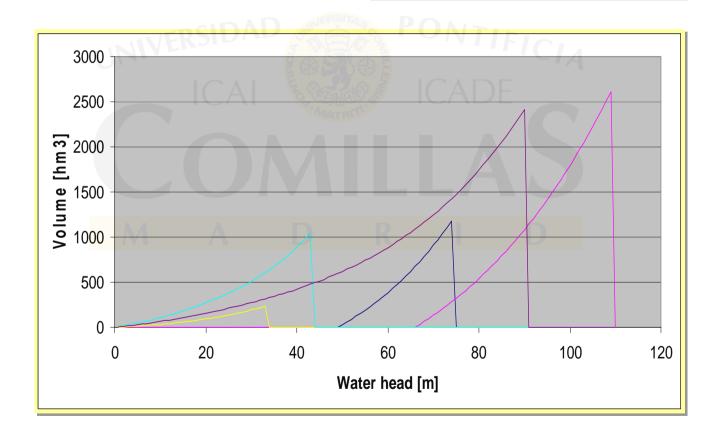




Constraint: Volume as a function of the head

 Reservoir volume depends quadratically (nonlinearly) on reservoir water head

 $V = \beta + \beta' H_r + \beta'' H_r^2$





Constraint: Water heads

Water head of the reservoir = forebay level - reference level

Water head of the plant = forebay level of the reservoir – tailrace level of the plant

Tailrace level of the plant = max [forebay level of downstream reservoir, reference tailrace level of the plant]



Constraint: operation limits

Reservoir volumes between limits for each hydro reservoir

Power operation between limits for each unit



Multiobjective function

- Thermal plant variable costs
- Penalties introduced in the objective function for softening several additional constraints:
 - Final reservoir volumes
 - Exceeding operating rule curves (minimum and maximum)
 - Minimum and maximum yearly operation hours of thermal units



Model results

- Results for each period and load block and for each scenario
 - Storage hydro, pumped hydro and thermal plant operation
 - Reservoir management
 - Basin and river production
 - Marginal costs
- Byproduct
 - Optimal water release tables for different stochastic natural inflows and reservoir levels (obtained by stochastic nested Benders' decomposition) used by a lower level daily simulation model



Type of optimization problem

- Deterministic approaches:
 - Network Flows
 - LP
 - NLP
 - MILP
 - commitment of thermal or hydro units
 - piecewise linear approximation of water head effects
- Stochastic approaches:
 - Stochastic Dynamic Programming (SDP)
 - Stochastic Linear Programming. Decomposition approaches (Benders, Lagrangian Relaxation, Stochastic Dual Dynamic Programming)
 - Stochastic Nonlinear Programming



Solution algorithm

- Algorithm:
 - Successive LP
 - Direct solution by a NLP solver
- Very careful implementation
 - Natural scaling of variables
 - Use of simpler expressions
 - Initial values and bounds for all the nonlinear variables computed from the solution provided by linear solver (CPLEX 10.2 IPM)
 - Nonlinear solver (CONOPT 3.14)



Model implementation

- General hydro topology
- Spreadsheet-based graphical interface
- GAMS-based optimization model



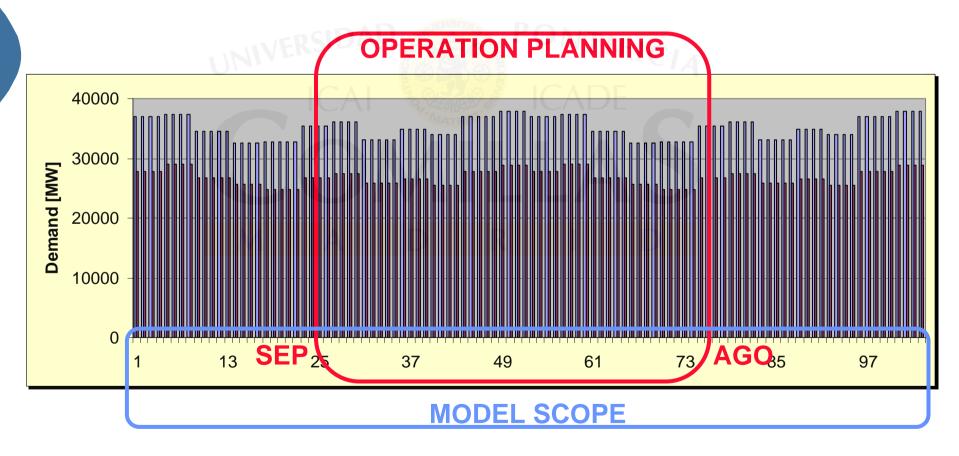
Contents

- > Introduction
- > System modeling
- ➤ Model description
- Case study
- Conclusions



Model use for operation planning

- Two-year long scope for one year operation planning
- Avoid initial and terminal effects





Two-year long case study

Spanish electric system

- 130 thermal units
- 3 main basins with 50 hydro reservoirs/plants and 2 pumped storage hydro plants
- 12 scenarios

Problem size:

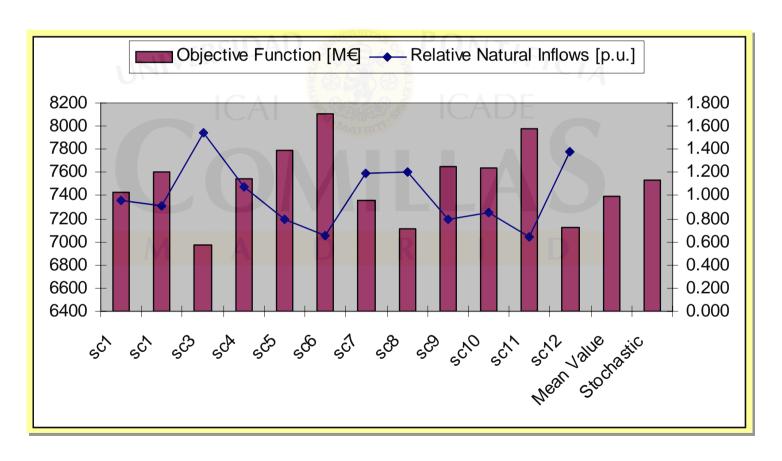
- 297913 constraints
- 530379 variables
- 1430949 non zero elements
- 17904 nonlinear variables
- 17904 nonlinear constraints





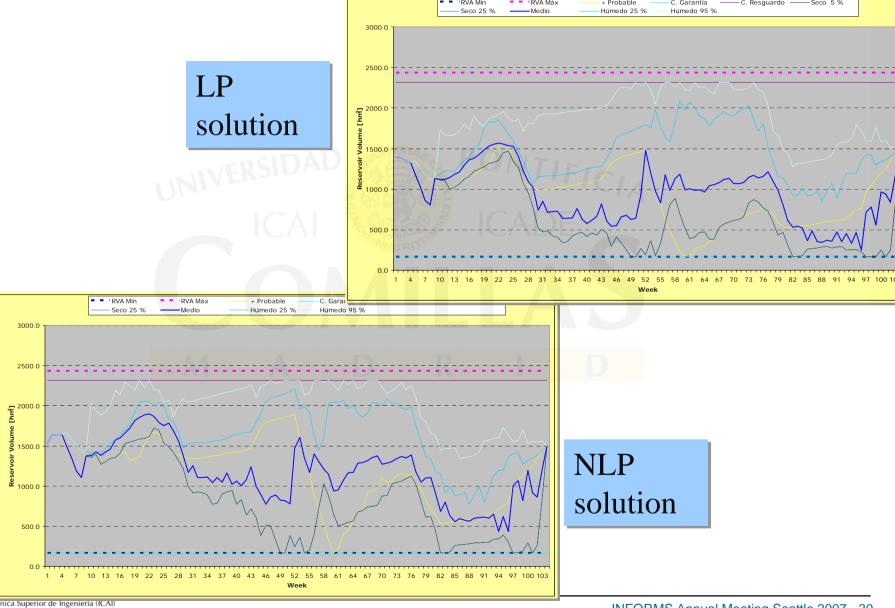
Scenario analysis and stochastic measures

- Value of the stochastic solution (VSS) = 133 M€
- Expected value with perfect information (EVWPI) = 7526 M€





Hydro reservoir operation (i)

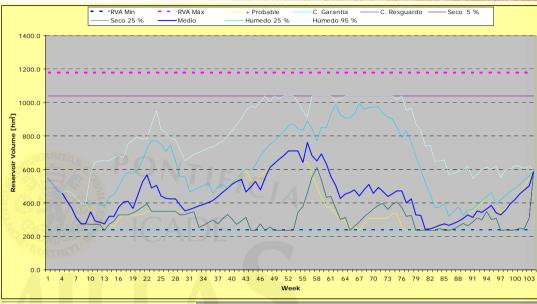


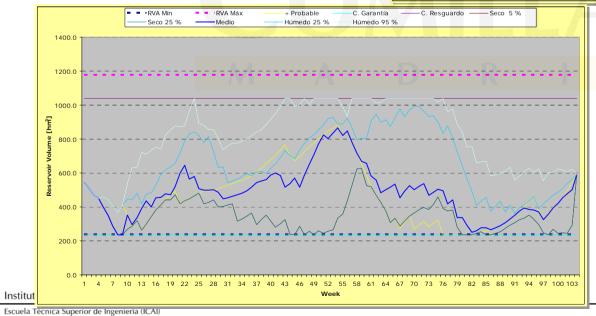


Escuela Técnica Superior de Ingeniería (ICAI) Universidad Pontificia Comillas

Hydro reservoir operation (ii)







NLP solution



Contents

- > Introduction
- System modeling
- ➤ Model description
- Case study
- **Conclusions**



Conclusions

- Medium term stochastic hydrothermal model for a complex multi-reservoir and multi-cascaded hydro subsystem
- Nonlinear water head effects modeled for large reservoirs
- Stochastic nonlinear optimization problem solved directed by a nonlinear solver given a close initial solution provided by a linear solver





A Medium Term Multireservoir Hydrothermal Coordination Model by Stochastic Programming

A. Ramos, S. Cerisola, J.M. Latorre
Universidad Pontificia Comillas
R. Bellido, A. Perea
Iberdrola

