



Po Basin And Public Administrations

Italian Department of Civil Protection

Po river basin Authority

Interregional Agency for the Po river basin

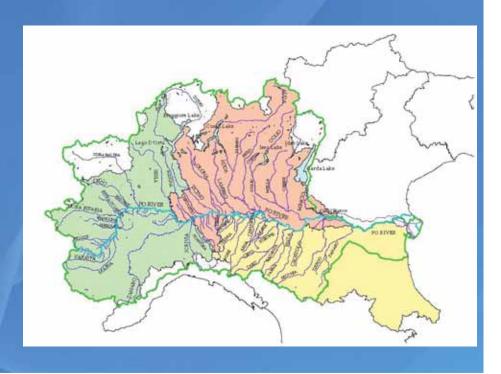
Emilia-Romagna Region

Lombardia Region

Piemonte Region

Valle d'Aosta Region

Veneto Region



Project objectives



To develop an adequate modelling system for planning and management



To develop a suitable forecasting system in real time applications

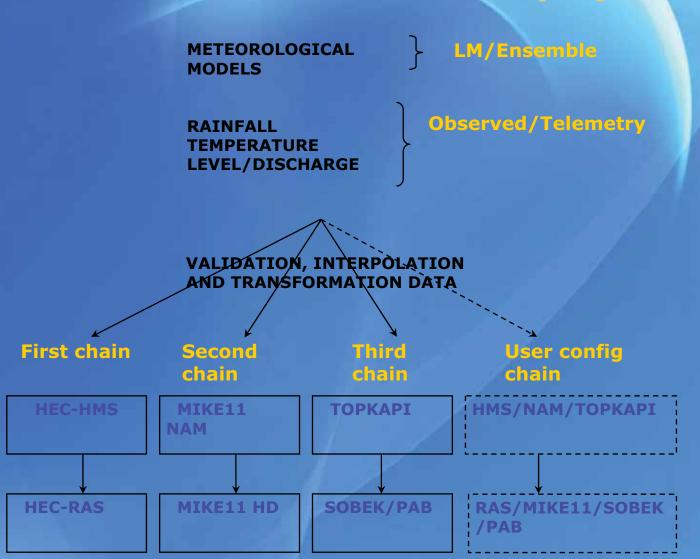


To help the organization of flood control services and soil defense services in advance, including civil protection measures to manage emergencies

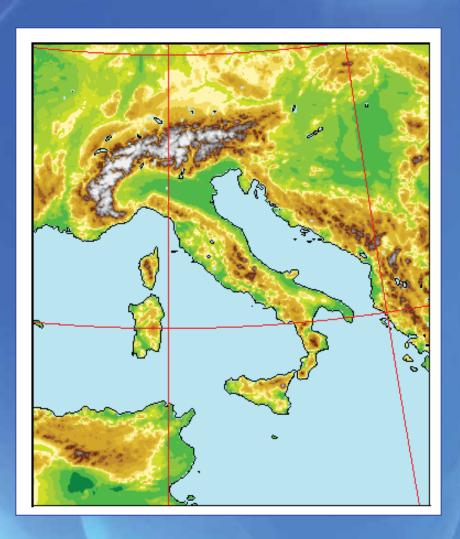
Po flood modelling system

- > Identification of observed hydrologic values from the monitoring network;
- ➤ Identification of the basin closures where each Centro Funzionale will provide results of their modelling system;
- > Identification of geometric schemes for flood propagation;
- > Analysis of frequency of data acquisition;
- > Realization of a number of forecasting modelling chains;
- Calibration and validation of modelling systems;
- Realtime data assimilation;
- > Realizations of user interfaces;
- > Realizations of tools for simulated scenarios of possible actions;
- > Realization of output system: fax, sms, web, ...

Schematic overview of the Po project

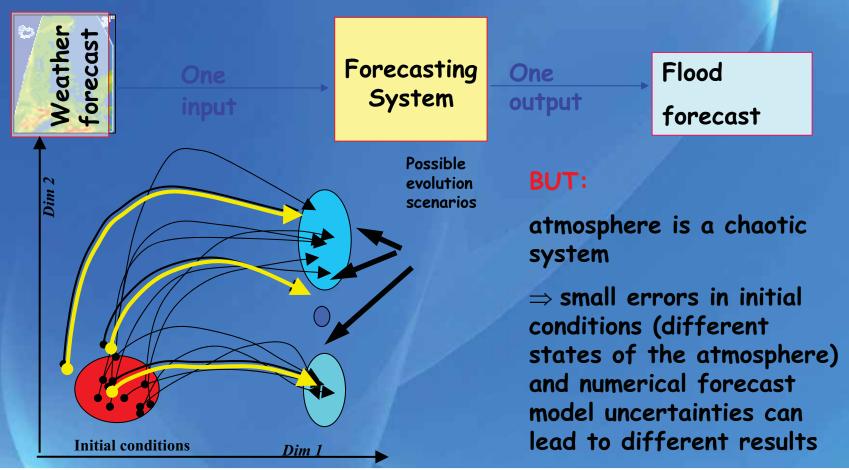


The LAMI Model



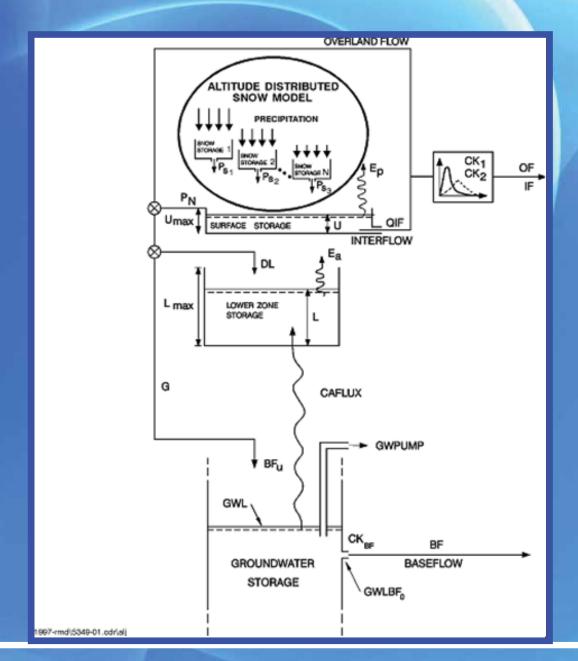
- •The Hydro-Meteorological Service of Emilia-Romagna, ARPA-SIM, has been using LM as the operational forecast model since 2001;
- · LM is run twice a day (at 00UTC and 12UTC) for 72 hours with a spatial horizontal resolution of 2.8 km and 40 layers in the vertical.
- The boundary conditions for LM are supplied by the global model of the DWD (one-way nesting) every hour. The initial condition is provided by a mesoscale data assimilation based on a nudging technique

From deterministic to probabilistic forecasting: Ensemble Prediction System (EPS-LEPS)



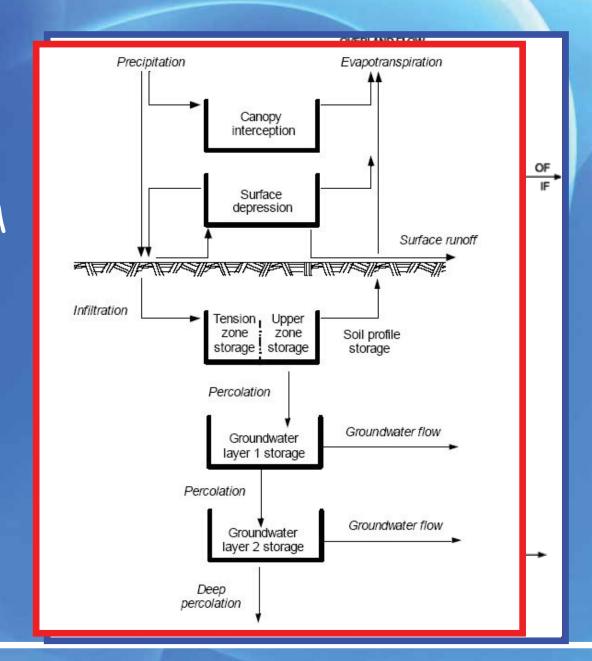
Hydrological models

· MIKE11-NAM



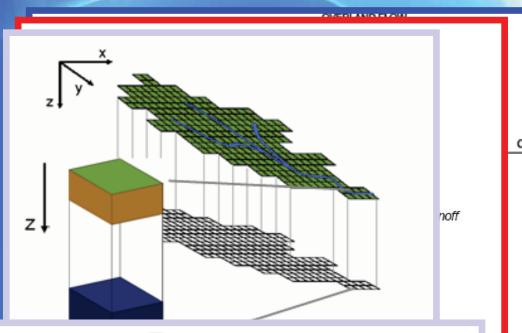
Hydrological models

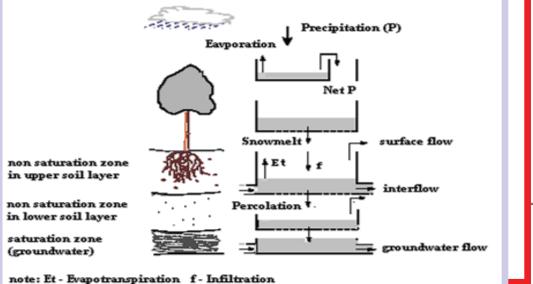
- · MIKE11-NAM
- · HEC-HMS



Hydrological models

- · MIKE11-NAM
- · HEC-HMS
- TOPKAPI

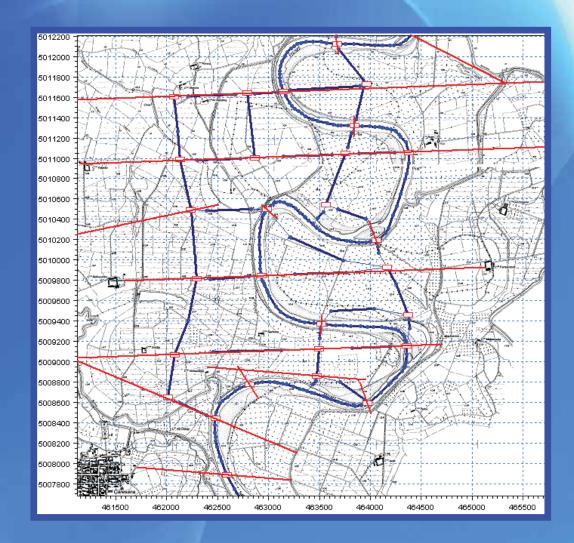






Hydrodinamic models

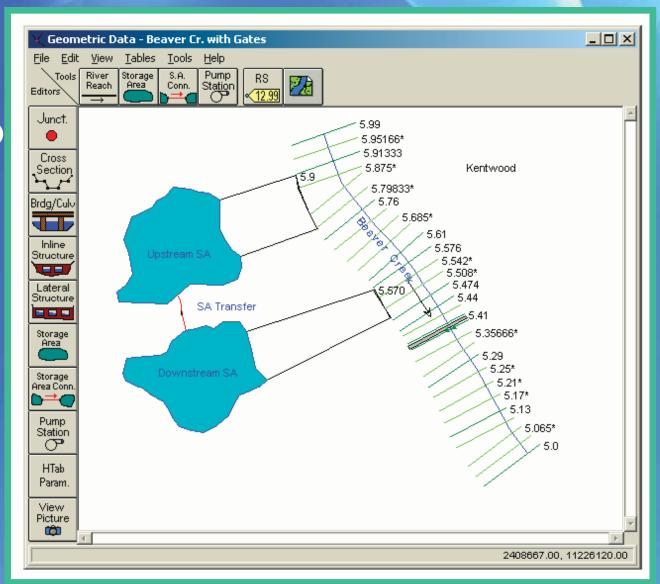
·MIKE11-HD



Hydrodinamic models

·MIKE11-HD

·HEC-RAS



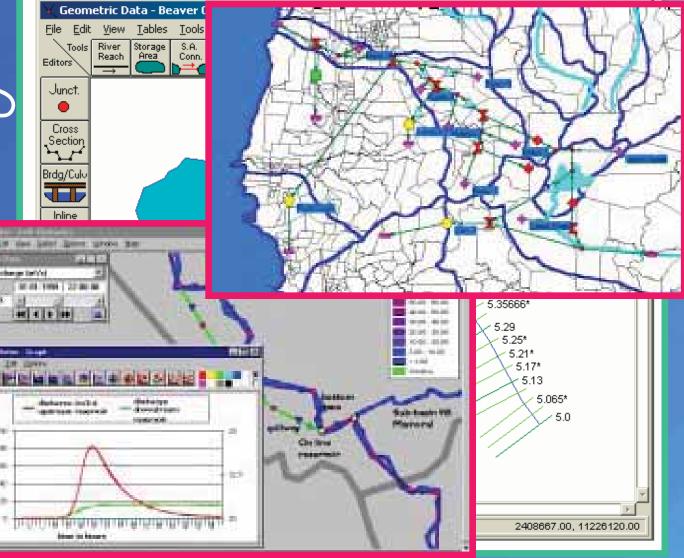


Hydrodinamic models

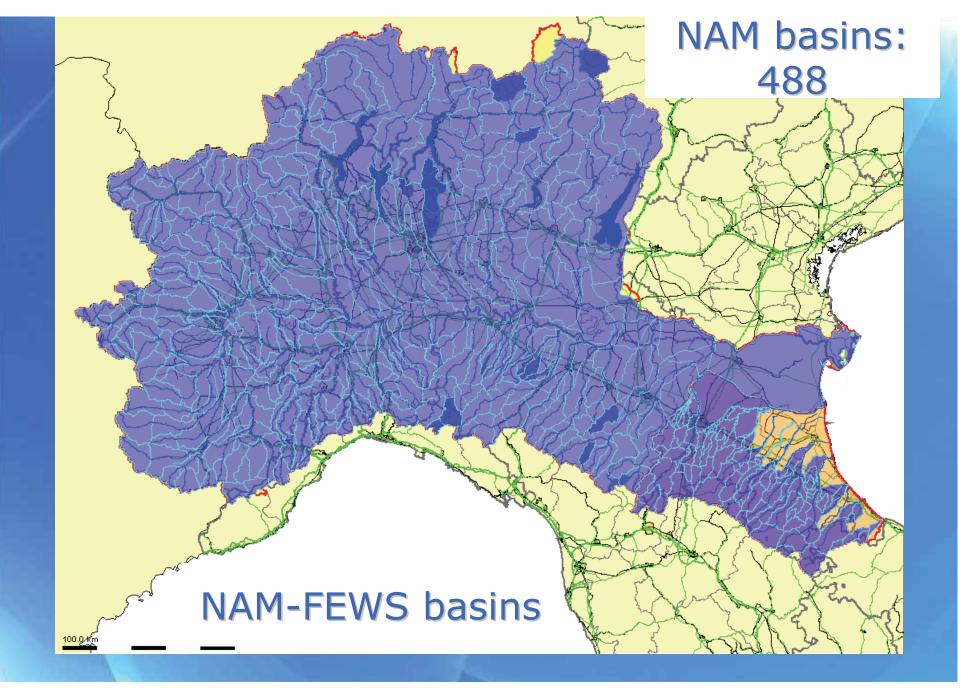
·MIKE11-HD

·HEC-RAS

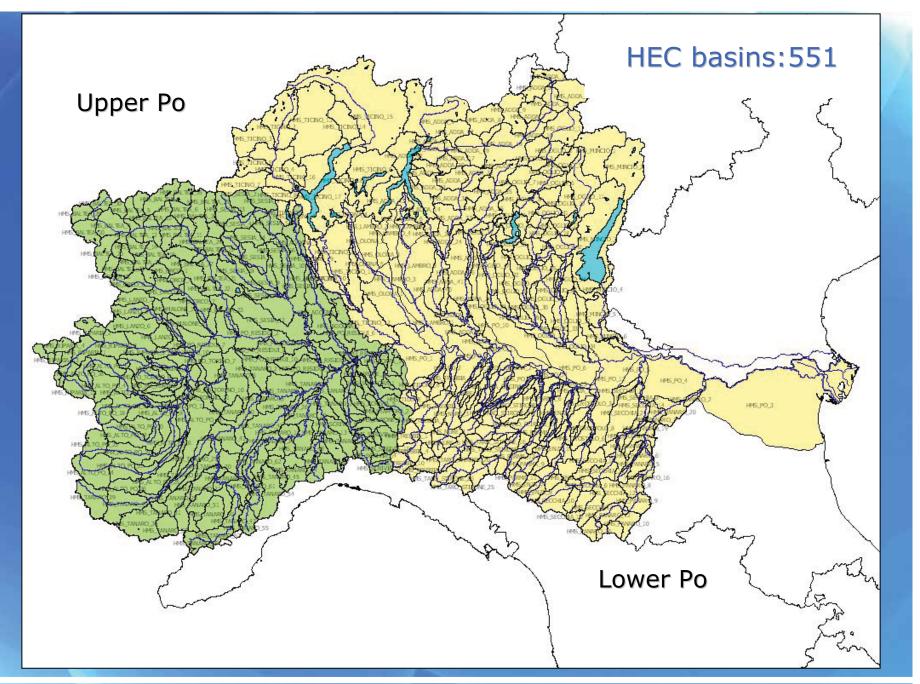
·SOBEK

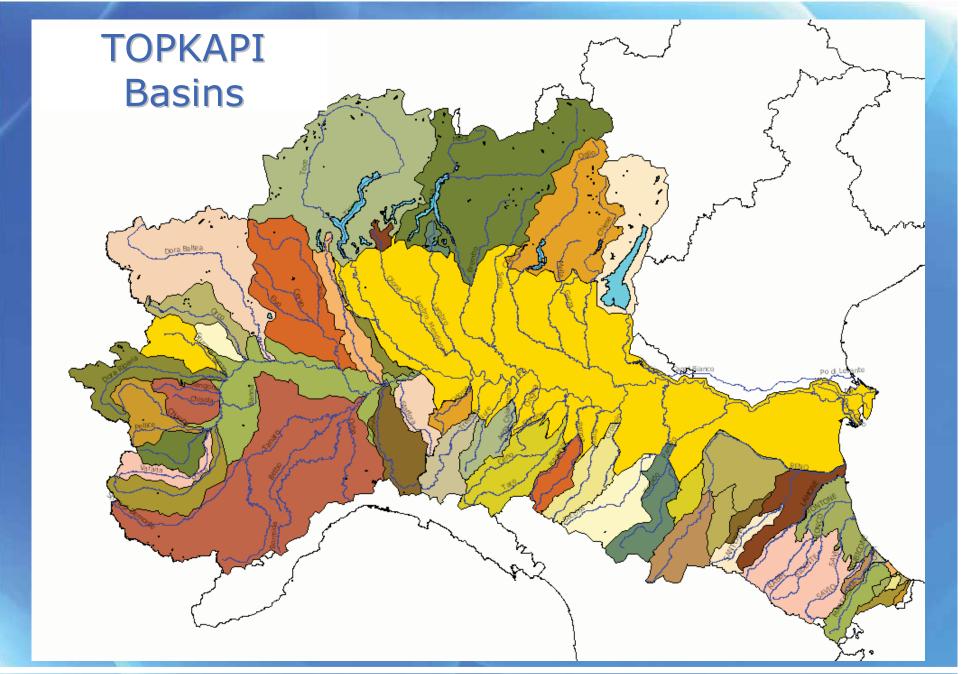








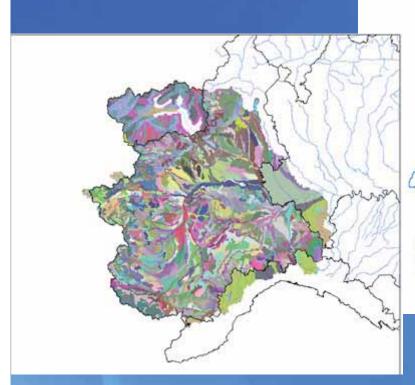


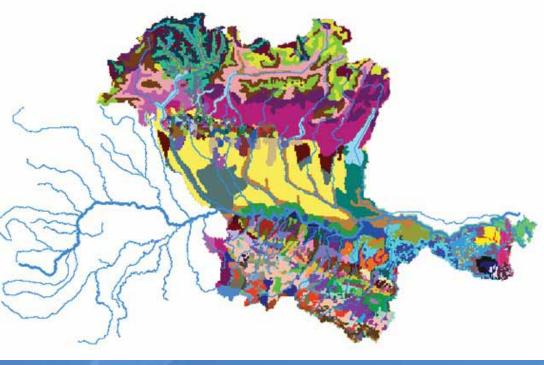




Soil types

- Emilia Romagna Region pedologic map (1:250000);
- Lombardia Region pedologic map (1:250000);
- Piemonte Region pedologic map (1:250000);
- European soil database (for missing data).



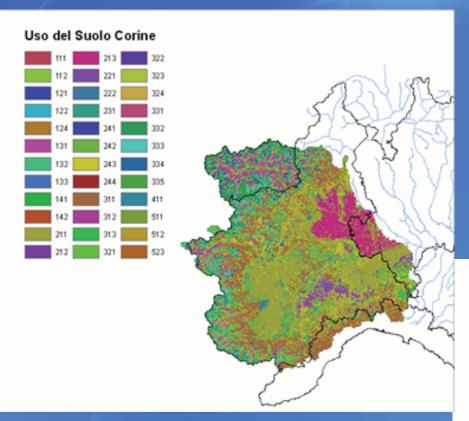


272 soil type classes:

- 76 Emilia-Romagna
- 61 Lombardia
- 82 Piemonte
- 53 European soil database

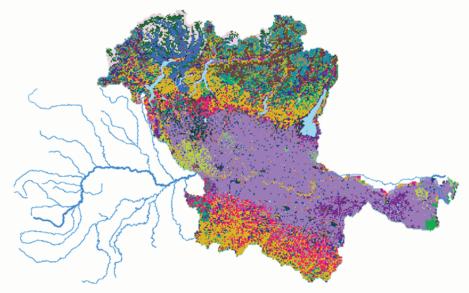


Land use



River Po basin land use characteristics have been individuated with "CORINE Program – LAND COVER Project".

50 DIFFERENT LAND USES



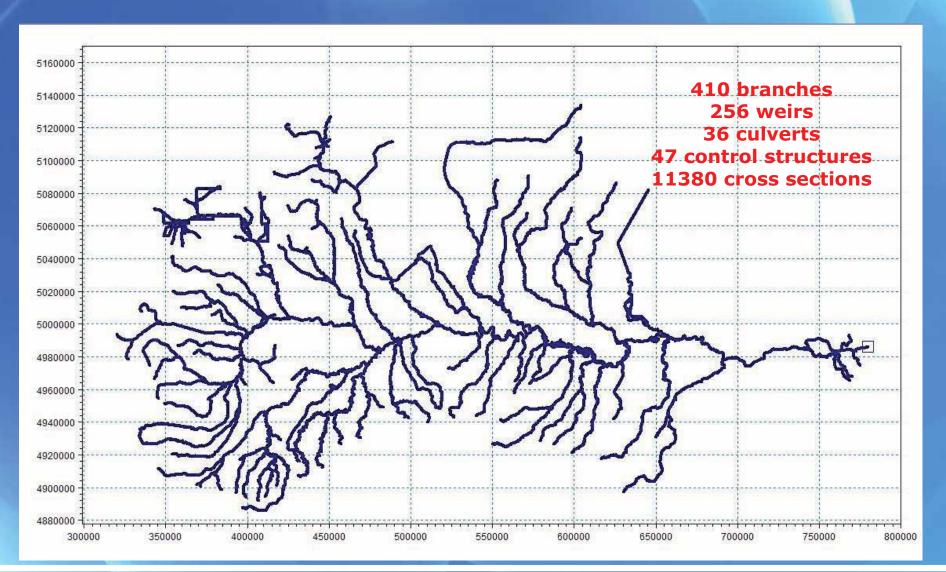
Cell dimensions for each sub-basin: 200 m, 250 m, 500 m, 1000 m

Bacino	Adda	Bacino	Trebbia	Bacino	Scrivia	Bacino	Orco
Rows	222	Rows	109	Rows	300	Rows	200
Cols	259	Cols	75	Cols	180	Cols	264
xCellSize	500	xCellSize	500	xCellSize	250	xCellSize	250
yCellSize	500	yCellSize	500	yCellSize	250	yCellSize	250
Bacino	Oglio	Bacino	Ticino	Bacino	Sesia	Bacino	Secchia
Rows	174	Rows	210	Rows	190	Rows	129
Cols	144	Cols	228	Cols	138	Cols	129
xCellSize	500	xCellSize	500	xCellSize	500	xCellSize	500
yCellSize	500	yCellSize	500	yCellSize	500	yCellSize	500
Bacino	Mincio	Bacino	Pianura	Bacino	Tanaro	Bacino	Tidone_Staffora
Rows	195	Rows	150	Rows	244	Rows	170
Cols	88	Cols	327	Cols	330	Cols	174
xCellSize	500	xCellSize	1000	xCellSize	500	xCellSize	250
yCellSize	500	yCellSize	1000	yCellSize	500	yCellSize	250
Bacino	Lambro_Olona	Bacino	Reno	Bacino	Varaita	Bacino	Lanzo
Rows	113	Rows	184	Rows	224	Rows	134
Cols	241	Cols	238	Cols	296	Cols	192
xCellSize	200	xCellSize	500	xCellSize	250	xCellSize	500
yCellSize	200	yCellSize	500	yCellSize	250	yCellSize	500
	Chiavenna-						
Bacino	Arda-Nure	Bacino	Romagna	Bacino	Pellice	Bacino	Parma
Rows	198	Rows	280	Rows	244	Rows	247
Cols	141	Cols	296	Cols	236	Cols	165
xCellSize	250	xCellSize	250	xCellSize	250	xCellSize	200
yCellSize	250	yCellSize	250	yCellSize	250	yCellSize	200
Bacino	Enza_Crostolo	Bacino	Agogna	Bacino	PianuraPie	Bacino	Taro
Rows	302	Rows	372	Rows	160	Rows	252
Cols	231	Cols	256	Cols	302	Cols	259
xCellSize	200	xCellSize	250	xCellSize	500	xCellSize	250
yCellSize	200	yCellSize	250	yCellSize	500	yCellSize	250
Bacino	Panaro	Bacino	DoraBaltea				
Rows	154	Rows	190				
Cols	99	Cols	212				
xCellSize	500	xCellSize	500				
yCellSize	500	yCellSize	500				

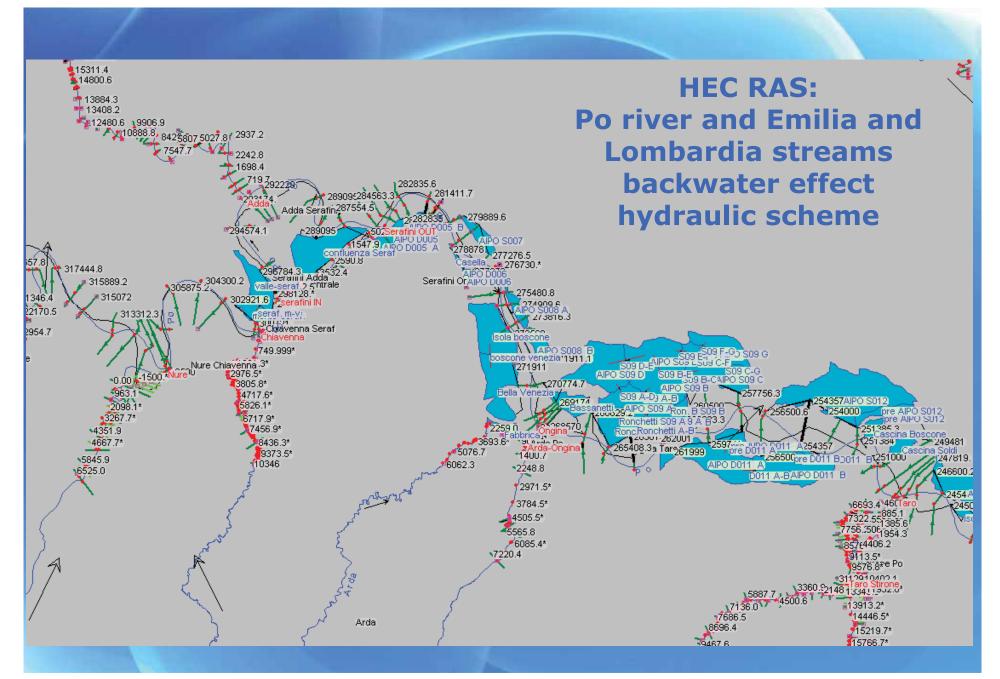
Total number of cells, covering the whole Po basin and Romagna Region: 285.000



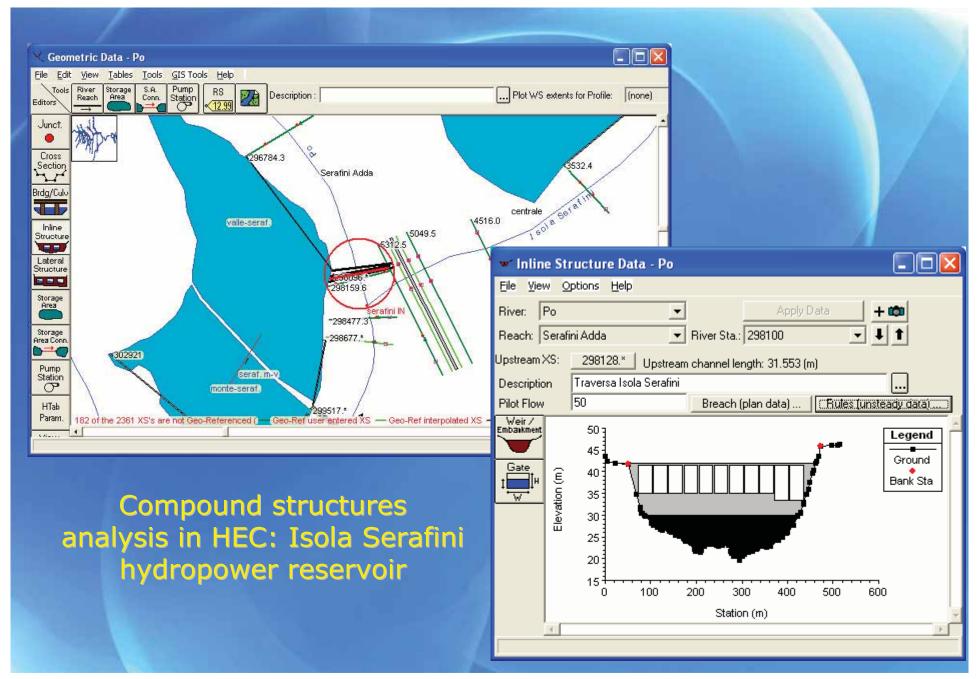
MIKE MODEL SCHEMATIZATION APPLIED TO THE PO BASIN



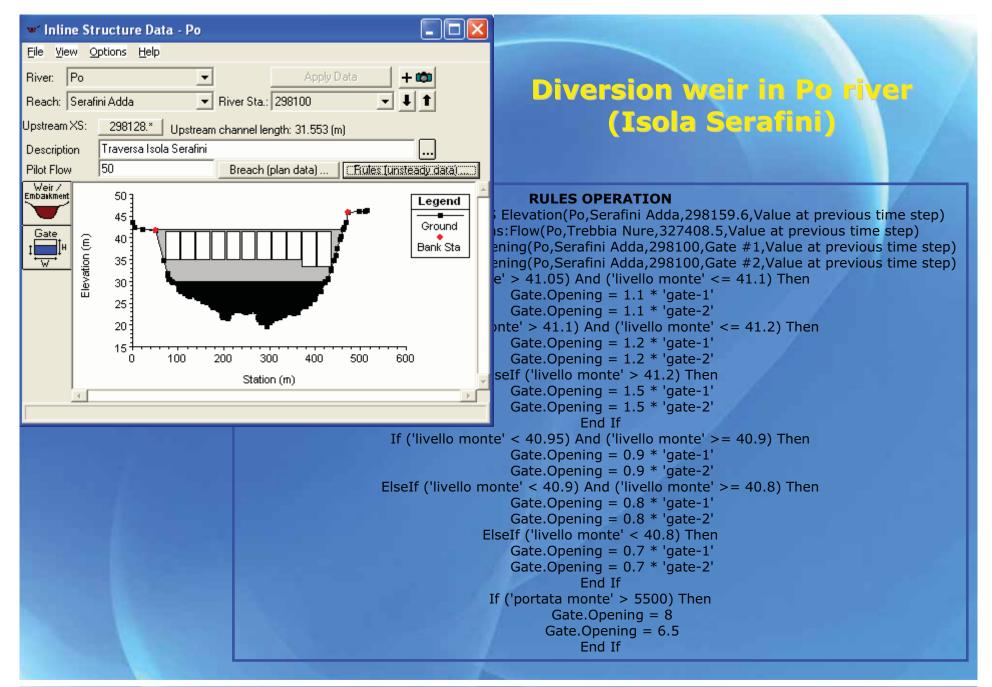




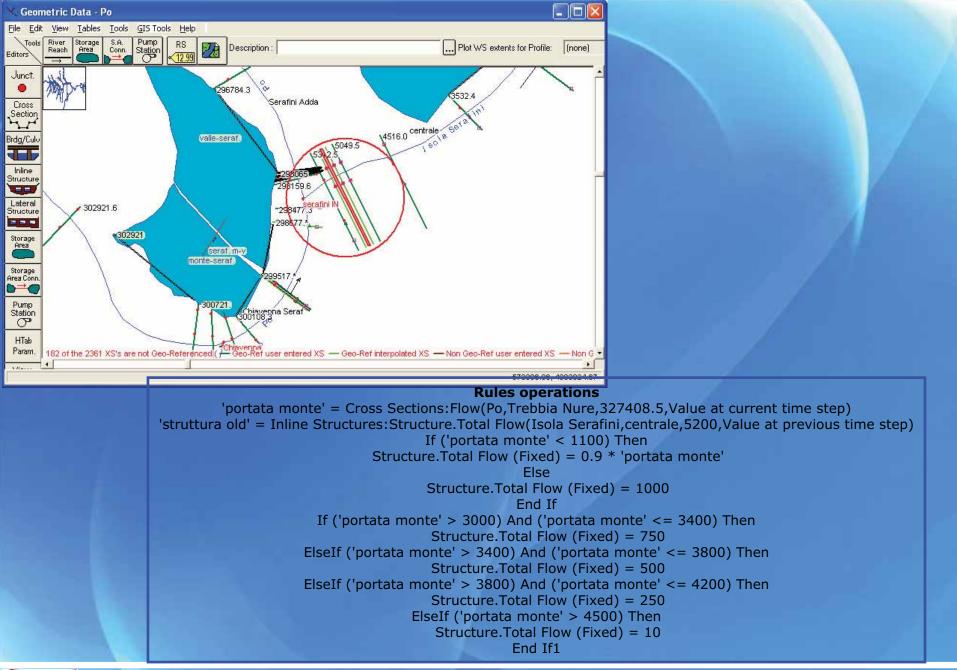






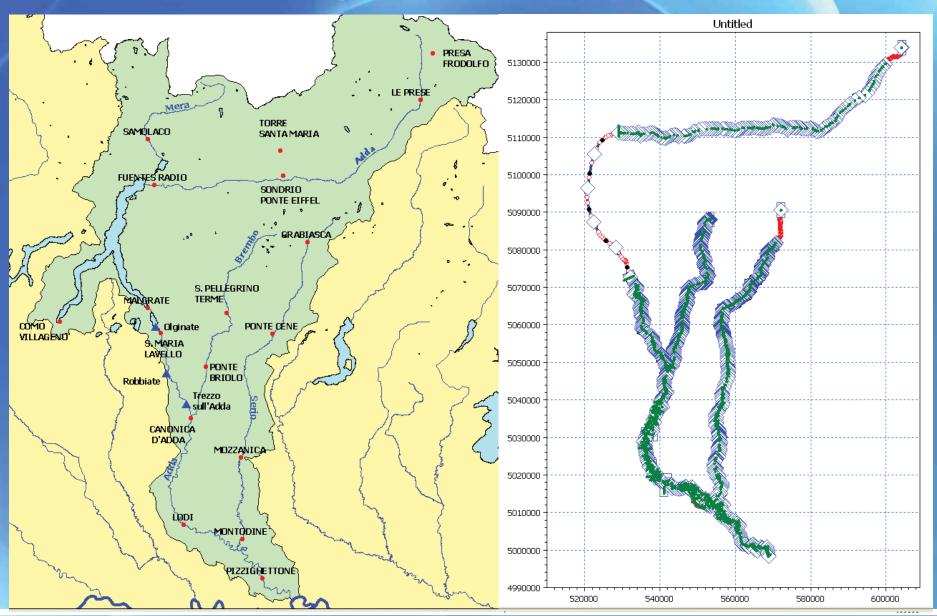




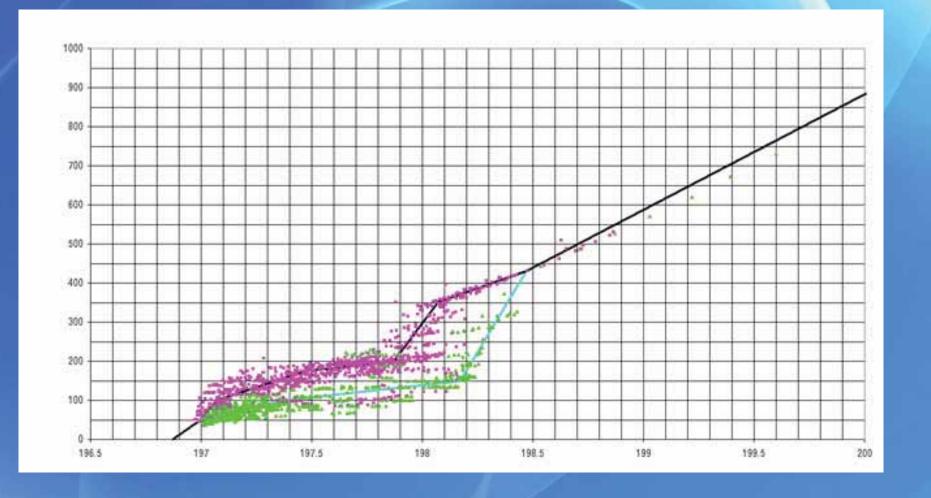




Mike model implemented on Adda and Como Lake

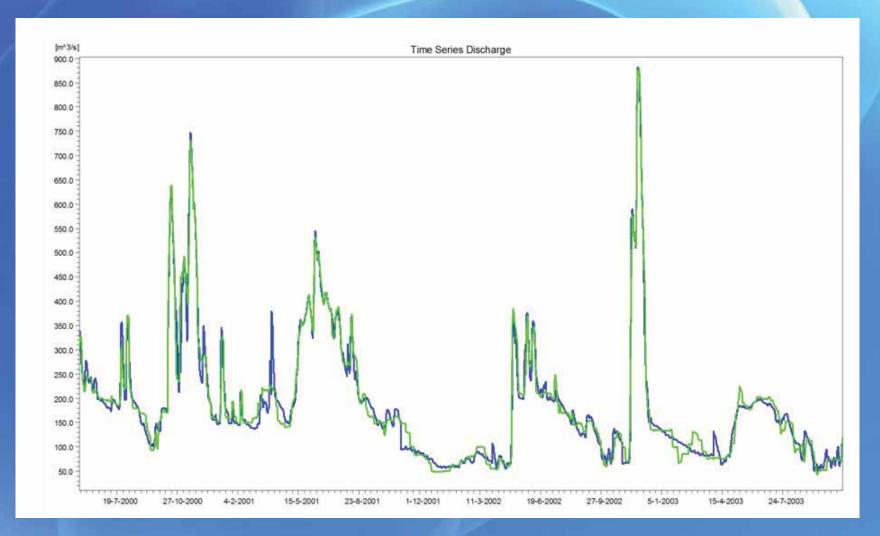






Water level and discharge at the downstream section of Como lake and relations used in the numerical model.

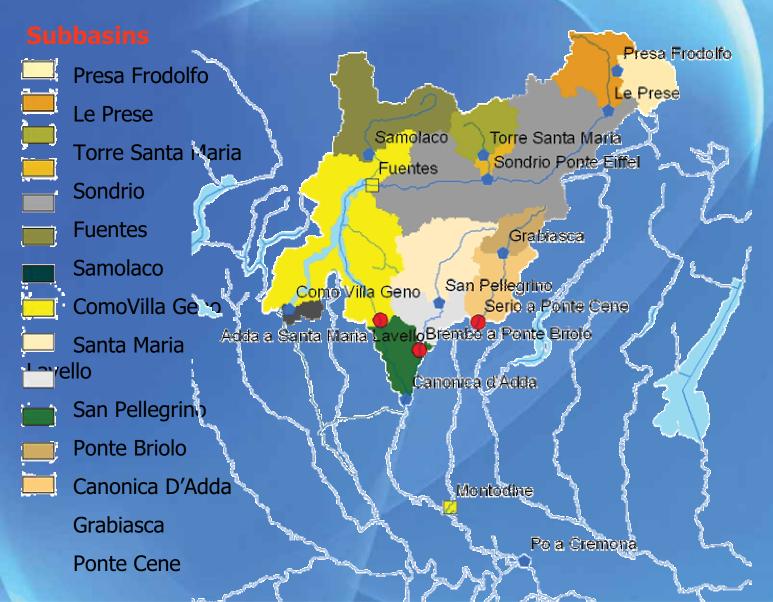




Data comparison between the observed outflow of the Lake – daily mean (green line) and the modelled outflow (blue line)

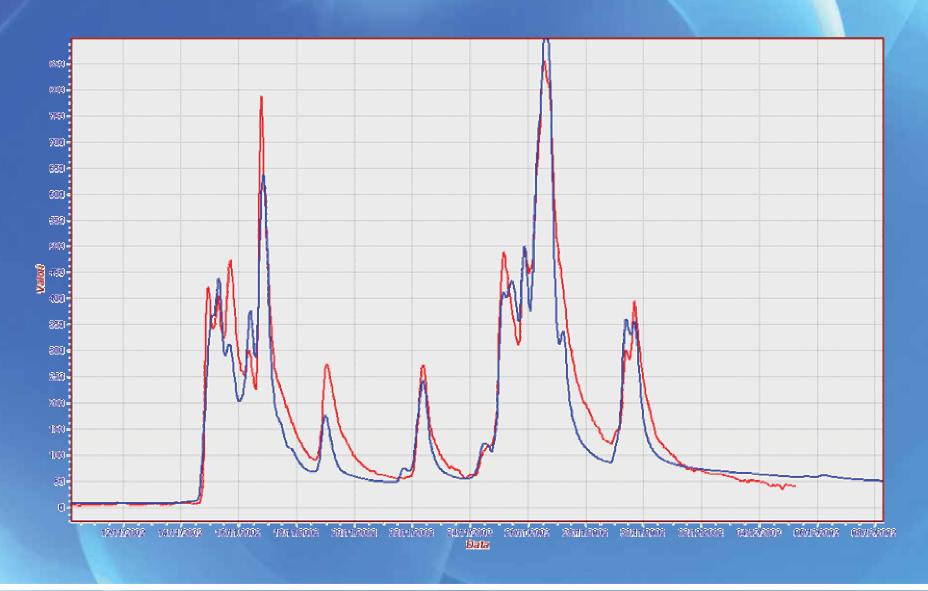


ADDA - Sub basins

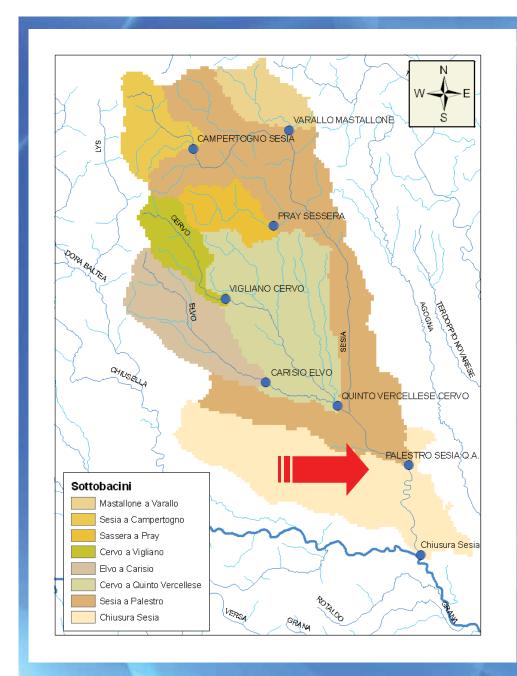




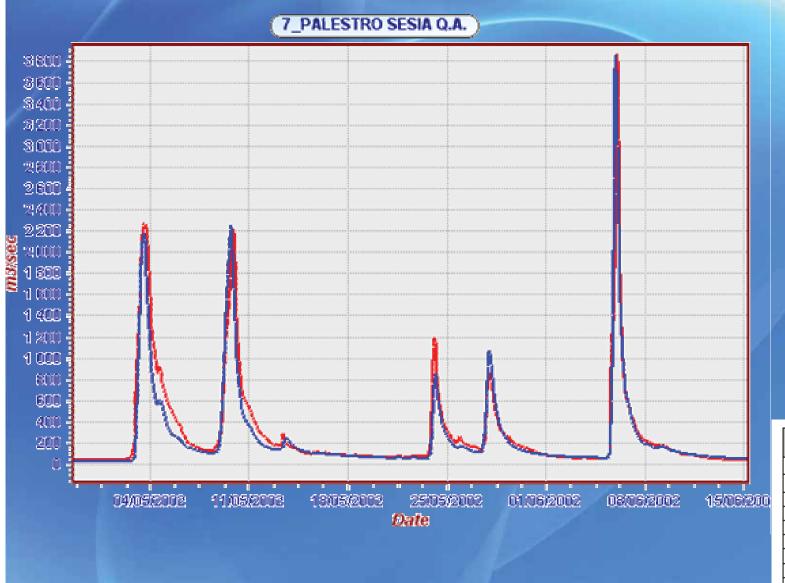
Example of discharge reconstruction at Ponte Briolo







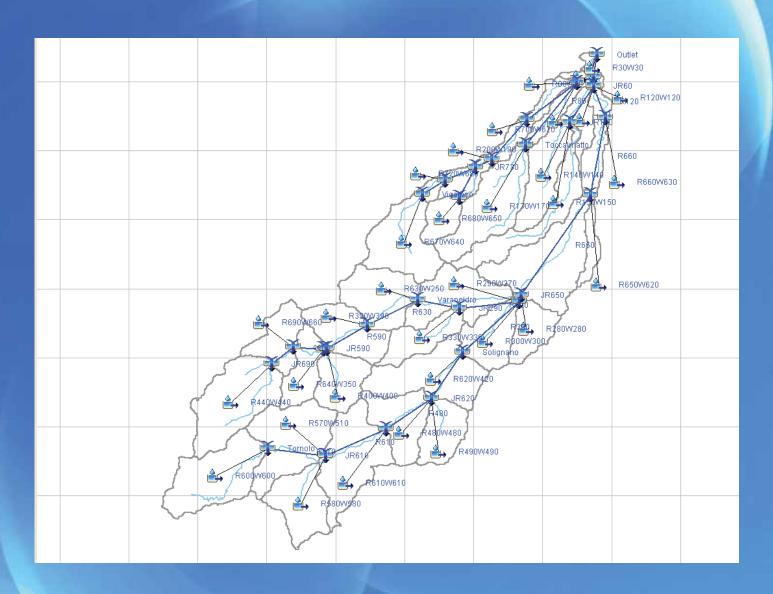
Sesia calibration at Palestro: year 2002



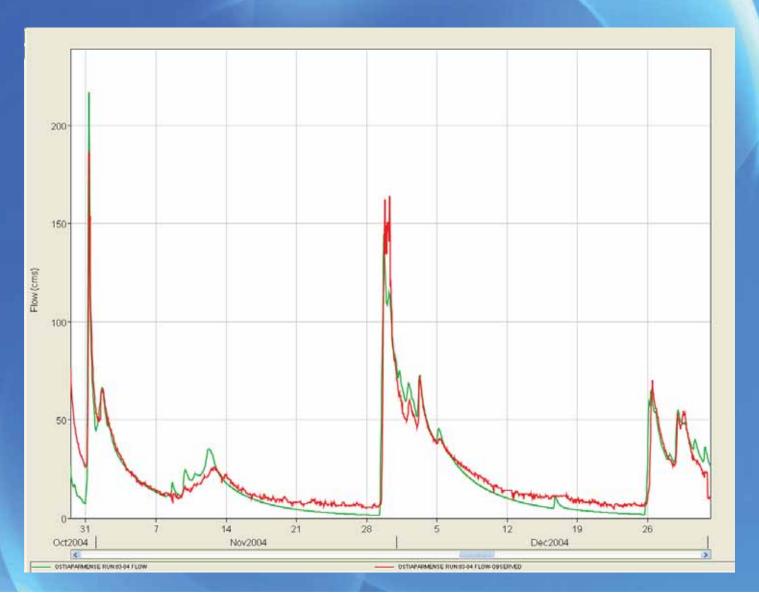
Palestro - E2 01/05/2002 - 17/06/2002							
Q _{max} Osservata [m³/s]	3865.22						
Q _{max} Calcolata [m³/s]	3854.91						
Shift [h]	2						
MAE	65.26						
RMSE	155.78						
R2	0.95						
EV	0.89						
E	0.89						
VC	0.75						
CMM	0.94						
d	0.97						
N.Dati	1151						

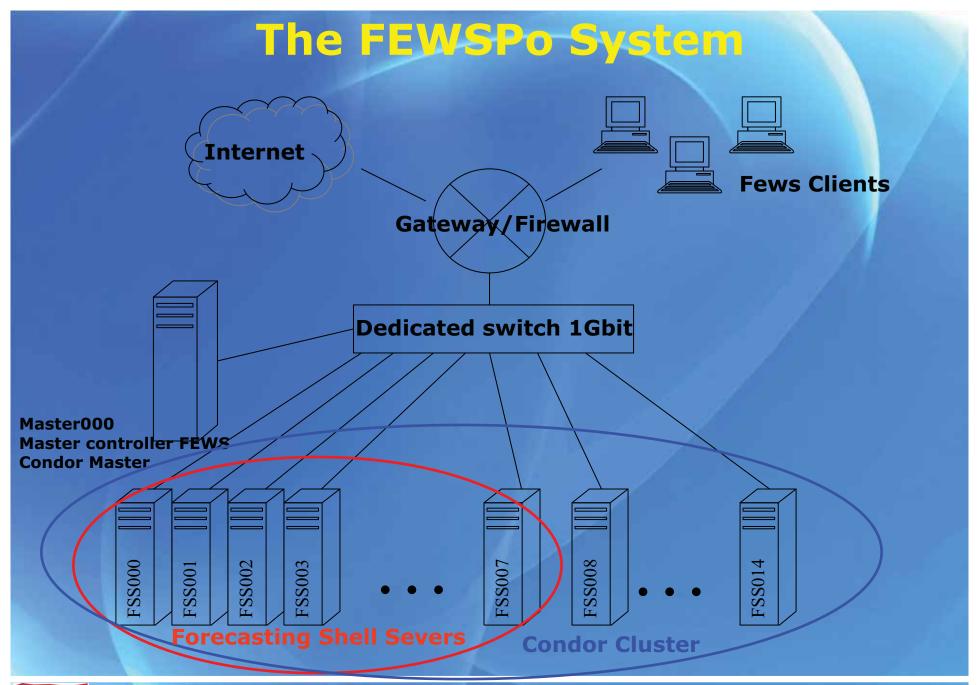


HEC model - Taro basin



HEC Calibration - Taro at Ostia Parmense







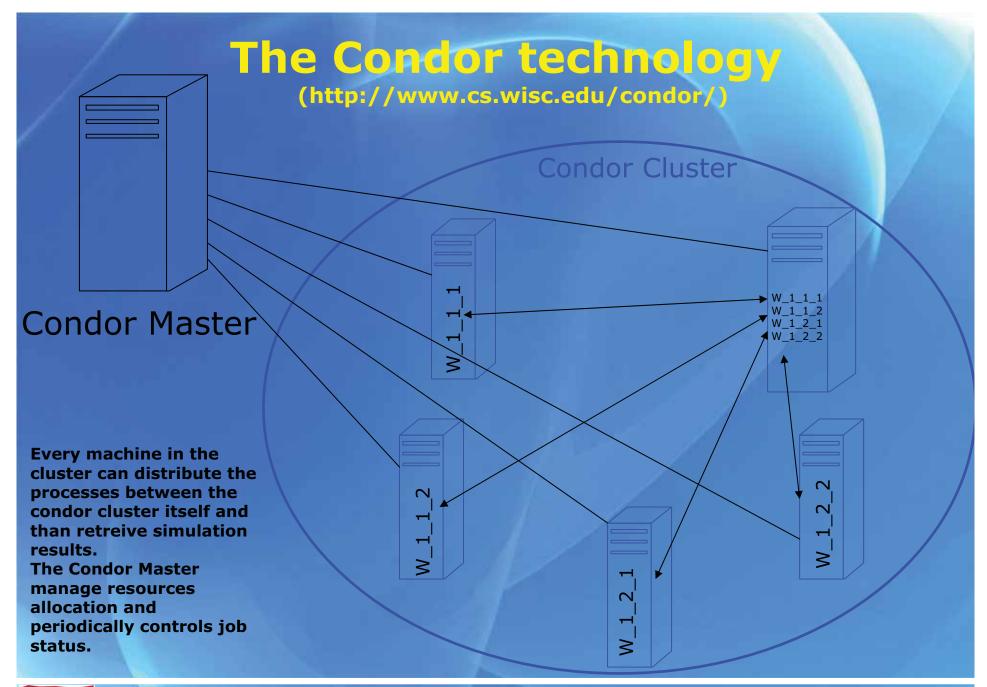
The FEWSPo System: HARDWARE



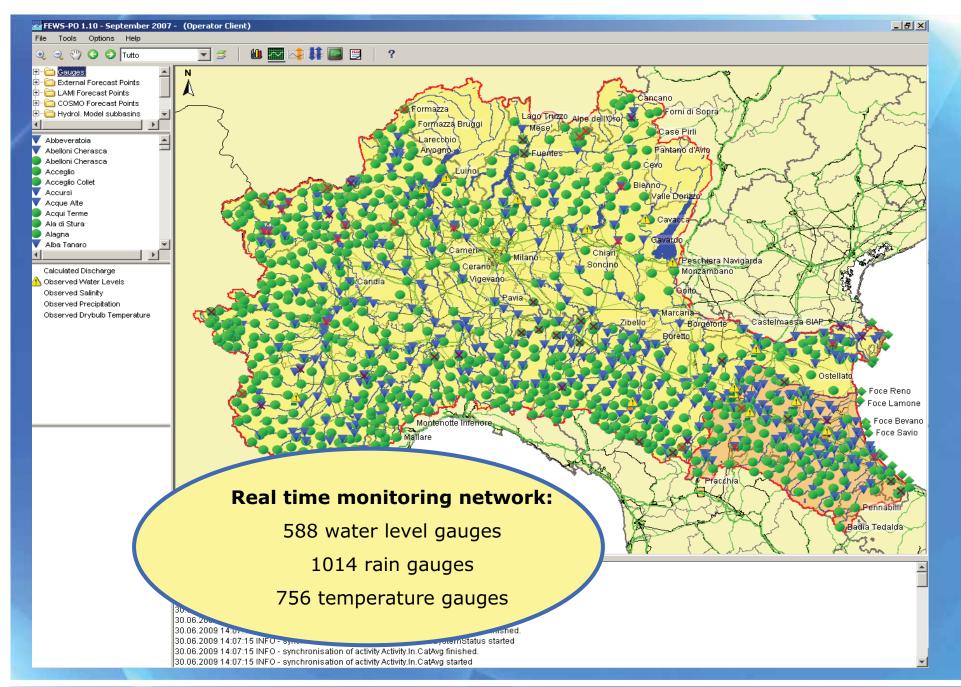
16 Servers (multi processor – 4 GB ram each)

A total of more than 140 CPU cores 3 TB Storage (> 10 yrs)



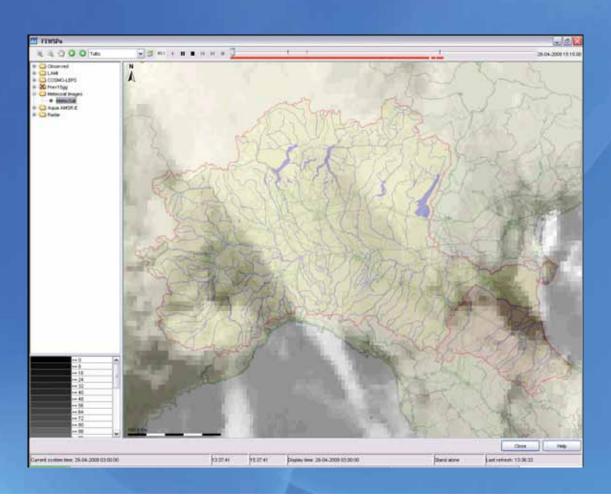




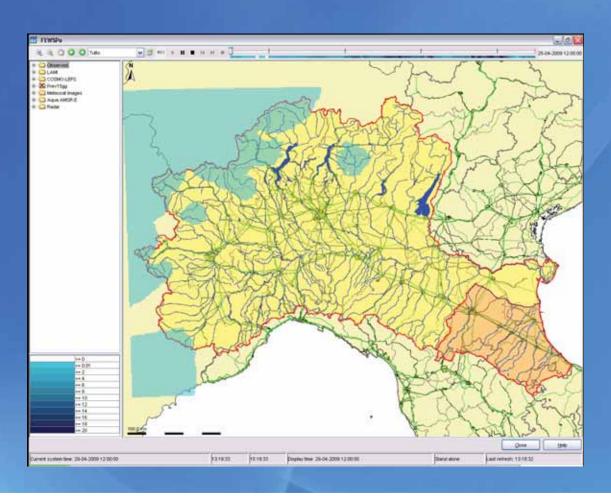




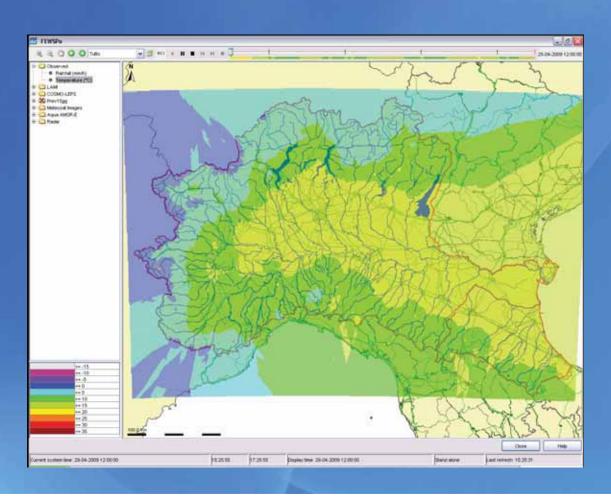
PO RIVER FLOOD EVENT IN APRIL 2009



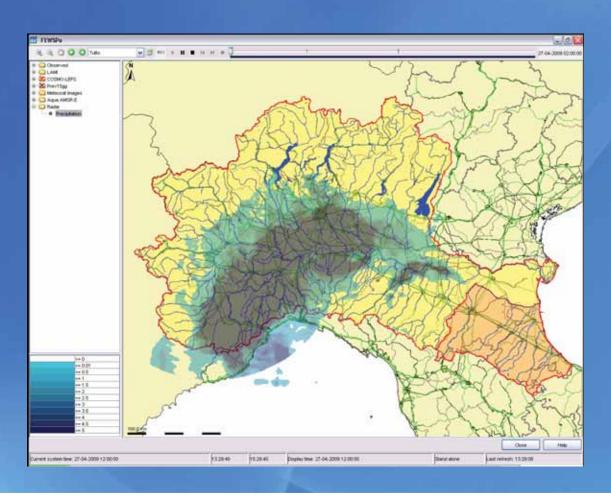
Observed precipitation



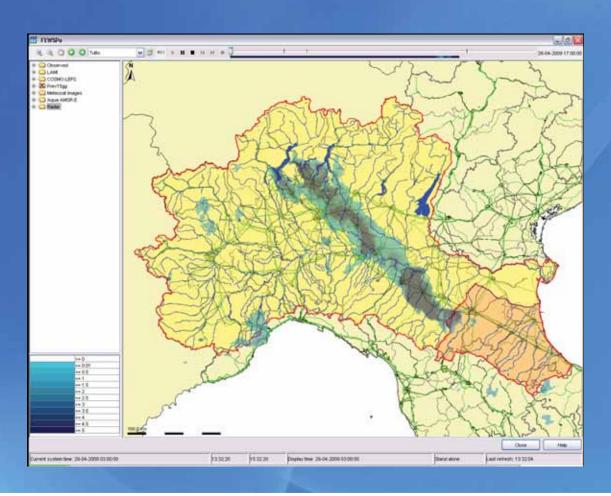
Observed temperature



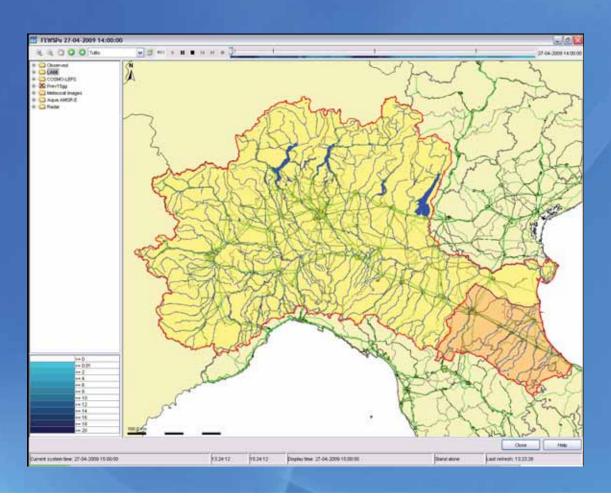
Radar precipitation 1/2



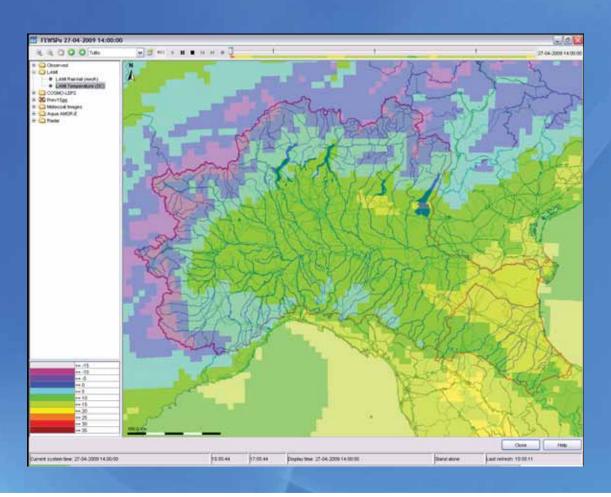
Radar precipitation 2/2



LAMI rainfall forecast



LAMI temperature forecast



LAMI FORECAST 27 apr h20 RADAR COMPOSITE MAP 27 apr h20

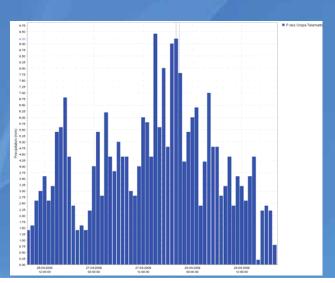
RADAR COMPOSITE MAP



SPATIAL INTERPOLATION BASED ON OBSERVATIONS 27 apr h20

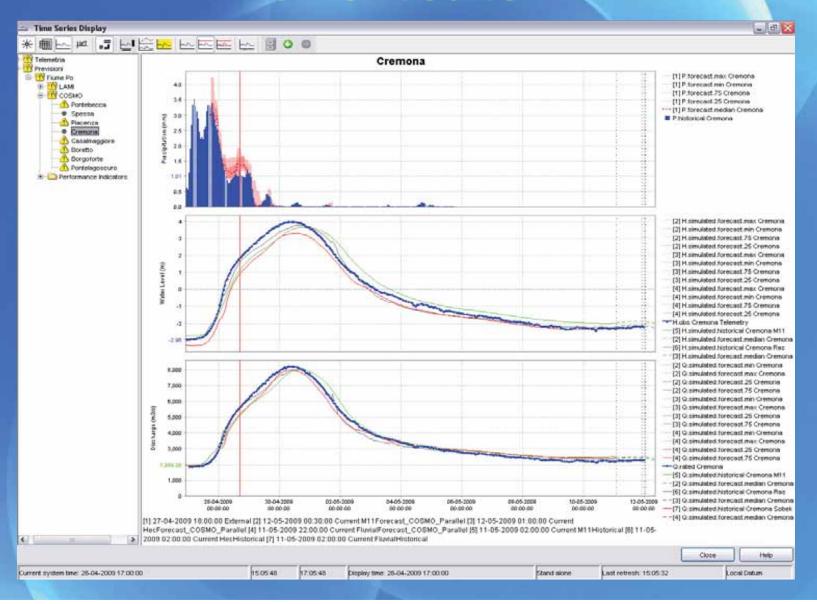


RAINFALL OROPA (PIEMONTE) 27 apr h20



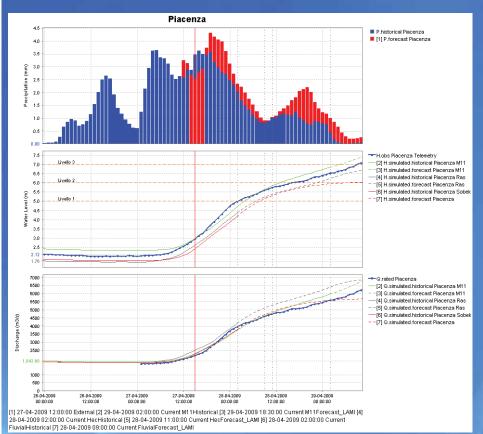


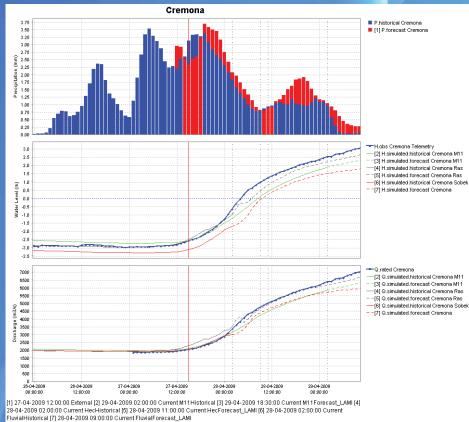
Shortcuts





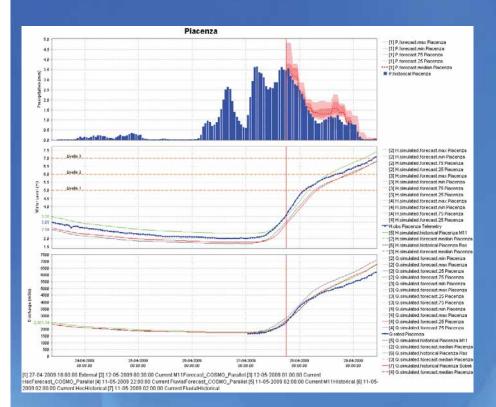
LAMI mean areal rainfall

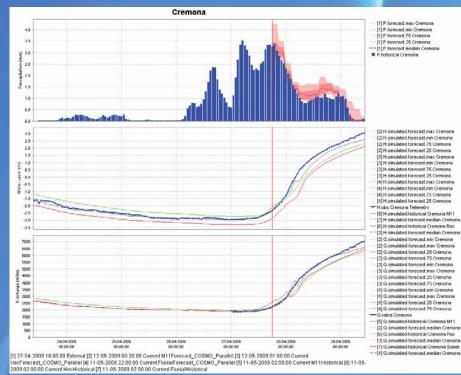






COSMO mean areal rainfall





POST-ELABORATION:

PREDICTIVE UNCERTAINTY

How to use in the best way measures and models to improve reliability and efficiency of operational decisions

DECISION MAKING UNDER UNCERTAINTY

In many operational problems, such as

- Flood warning;
- Flood emergency management;
- Reservoir management;
- Etc.

decision makers must take important decisions under the uncertainty of future events.



THE DEFINITION OF PREDICTIVE UNCERTAINTY

Predictive Uncertainty can be defined as the probability of occurrence of a future value of a predictand (such as water level, discharge or water volume) conditional on:

- 1) prior observations and knowledge
- 2) all the information that can be obtained on the future value, which is typically embodied in one or more meteorological, hydrological and hydraulic model forecasts.

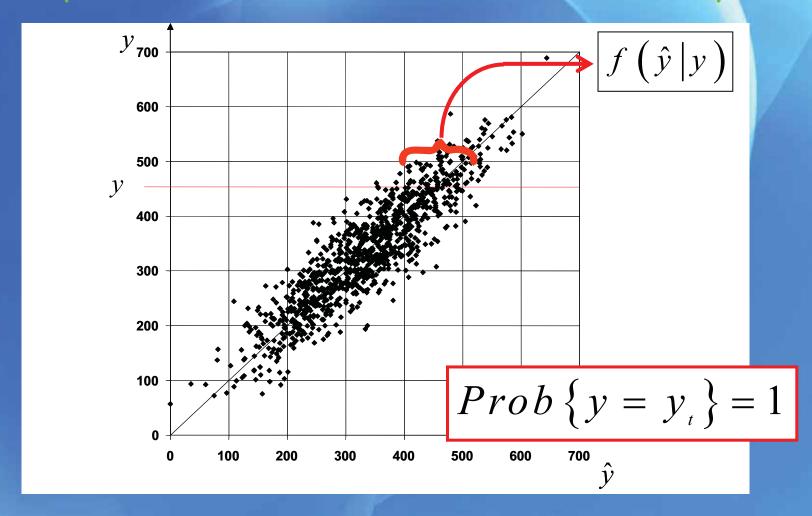


THE DEFINITION OF PREDICTIVE UNCERTAINTY

PU is obviously the uncertainty that we have on the occurrence of a real future value, as for instance the water level in 12 hours from now.

This must not be confused with "Emulation Uncertainty".

Representation of Emulation Uncertainty

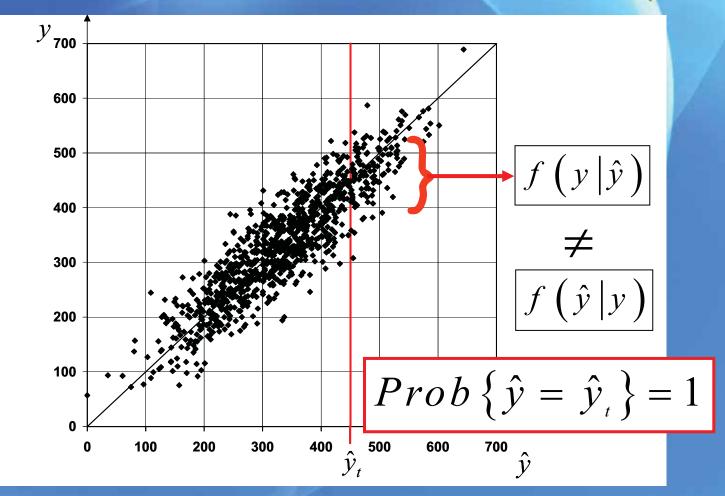


Predictand





Representation of Predictive Uncertainty



Predictand



Model



MODEL AND PARAMETER UNCERTAINTY

Anyway, presently available experiences show that marginalising parameters uncertainty, although statistically correct, does not produce substantial differences from using a best fit parameter set.

This is mostly due to the fact that the nearly best parameters produce predictions that are closely related among them, while the posterior probability, of the worst parameter sests is obviously very low.



MODEL AND PARAMETER UNCERTAINTY

This is why it is more interesting to approach the problem in terms of

few alternative models of widely different nature.

I.e. a physically based model, a conceptual model and a data driven model.

This has given rise to the development of several multi-model

Predictive Uncertainty Processors



Available Single or Multi-model Predictive Uncertainty Processors

Hydrological Uncertainty Processor

Krzysztofowicz, 1999; Krzysztofowicz and Kelly, 2000

Bayesian Model Averaging

Raftery et al., 2003;

Model Conditional Processor

Todini, 2008.

Krzysztofowicz, R., 1999. Bayesian theory of probabilistic forecasting via deterministic hydrologic model. Water Resour. Res., 35, 2739-2750.

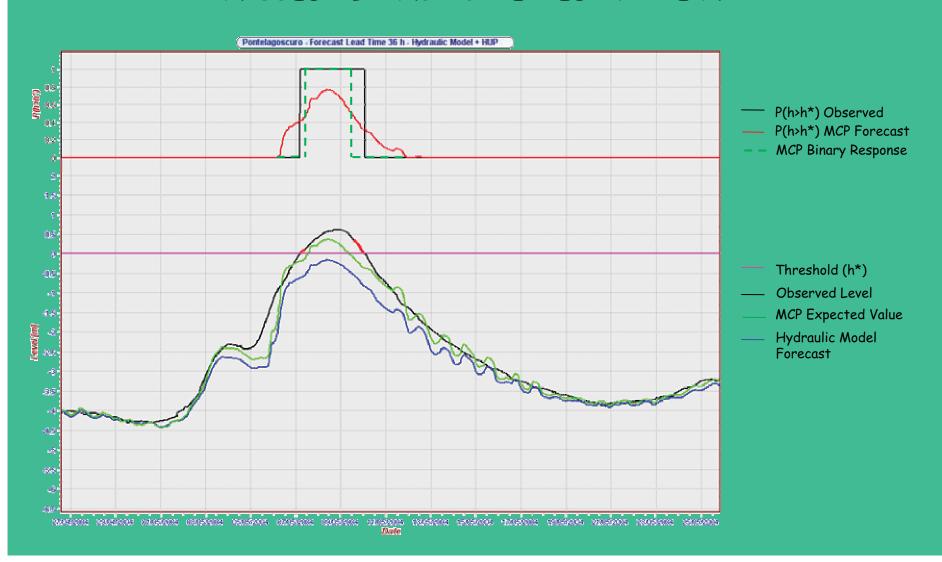
Raftery, A. E., F. Balabdaoui, T. Gneiting, and M. Polakowski, 2003. Using Bayesian model averaging to calibrate forecast ensembles, Tech. Rep. 440, Dep. of Stat., Univ. of Wash., Seattle.

Todini, E., 2008. A model conditional processor to assess predictive uncertainty in flood forecasting. Intl. J. River Basin Management. Vol. 6 (2), 123-137.



Po at Pontelagoscuro: 36 h forecast Mike11 and MCP processed PU

MISSED ALARMS CORRECTED BY THE MCP



Po at Pontelagoscuro: 36 h forecast Mike11 and MCP processed PU

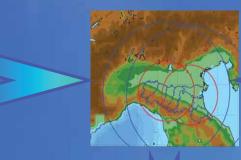
FALSE ALARM CORRECTED BY THE MCP

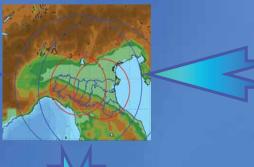


Composite radar



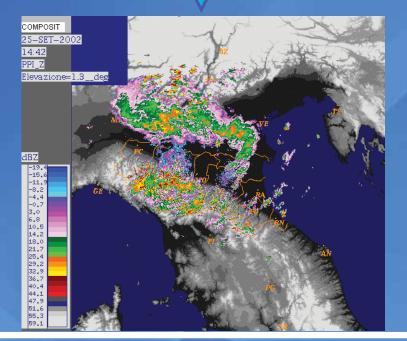
Gattatico



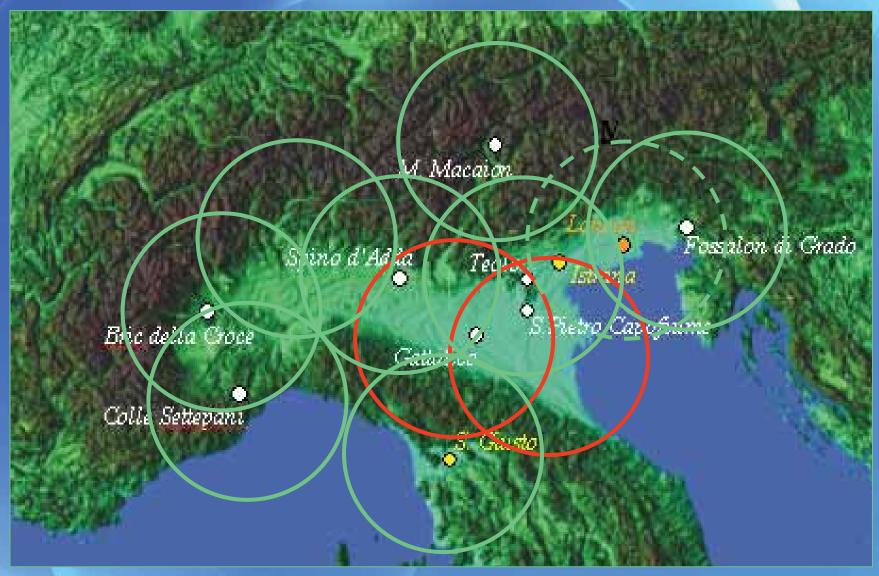




San Pietro Capofiume



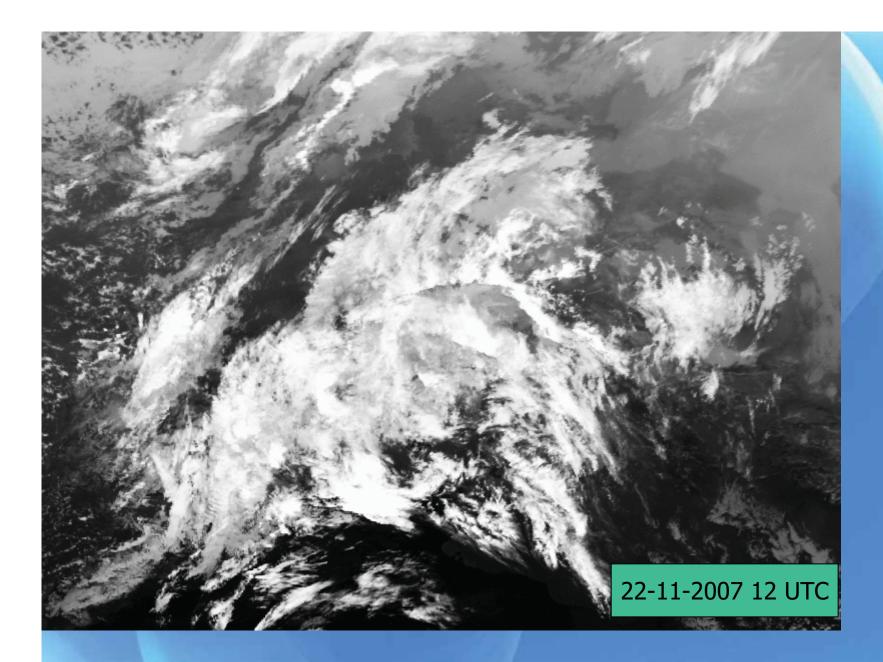
The radar network

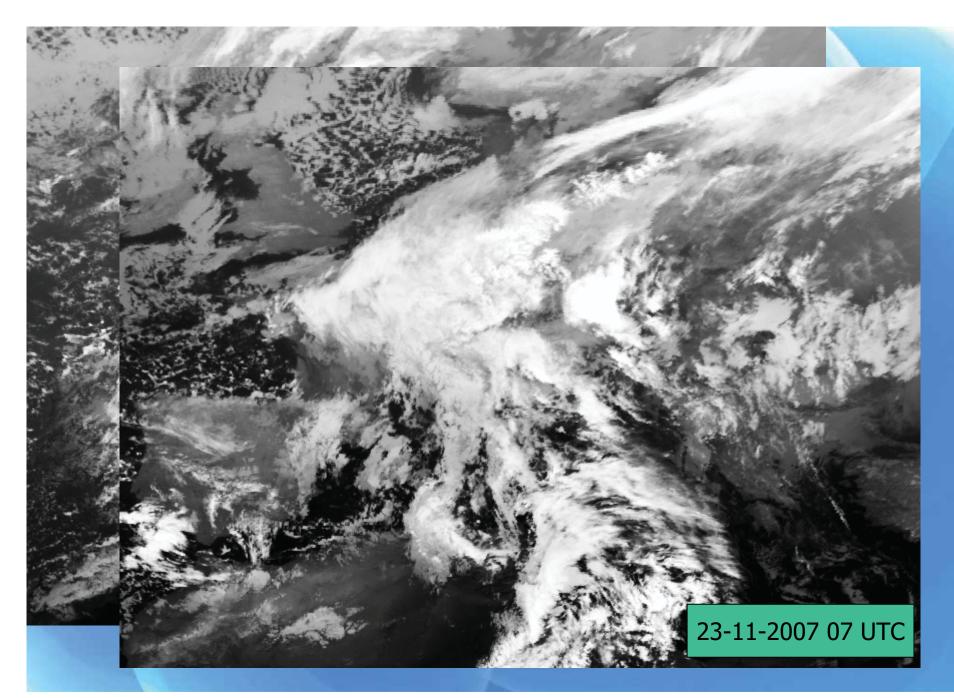




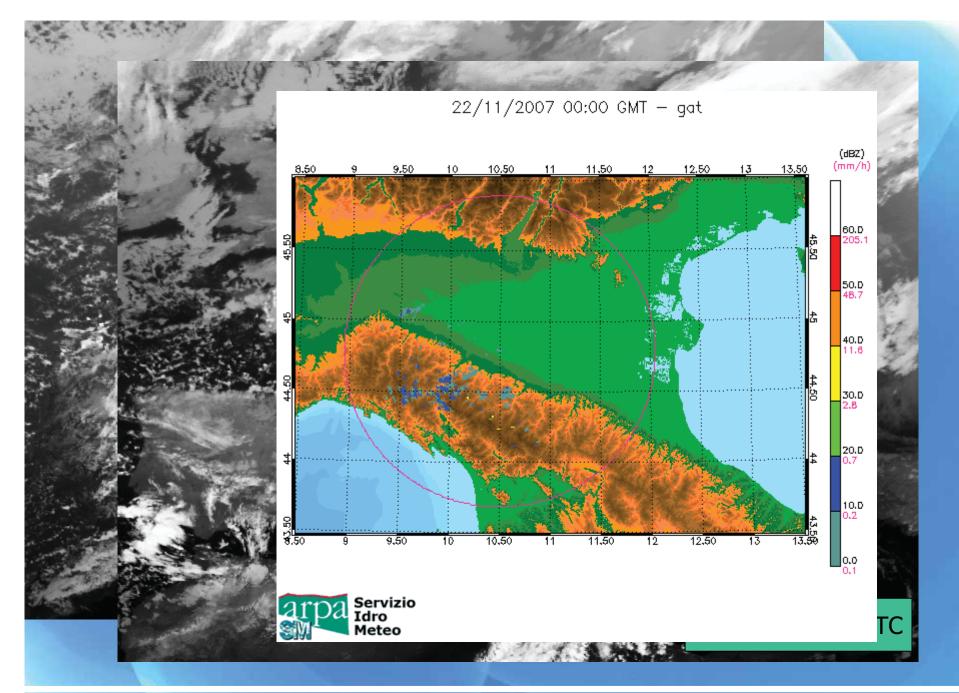
Event description

- A mesoscale system pass over Italy between 22-23 November 2007.
- Severe rain developed on the Apennine
- Maximum of precipitation in the study area was around 300 mm in 3 day











Polar Volume (Z) Raw data in polar coordinates

Data Processing

Z & Quality Flags

Quality descriptor

Clutter clearing Anaprop correction Beam blocking **VPR**

Reflectivitiy corrected data and associated quality flags corresponding to the different algorithms applied

Function that syntesize the quality flags defined

Z & quality

Surface estimated reflectivity and quality data

RADAR QRE Compositing

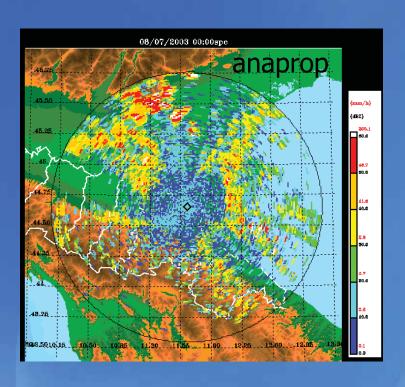
Precip. Field & quality characterisation

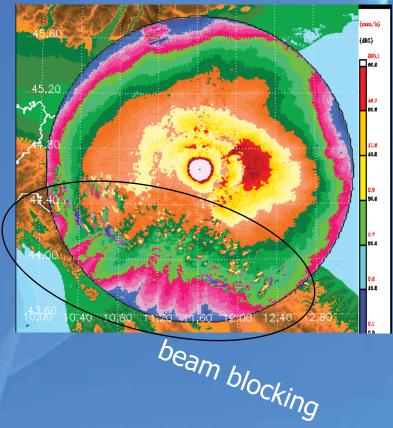
Error estimation

Raingauge comparison sh Floods and Fluvial Flooding - Cagliari, 26-28th May 2010

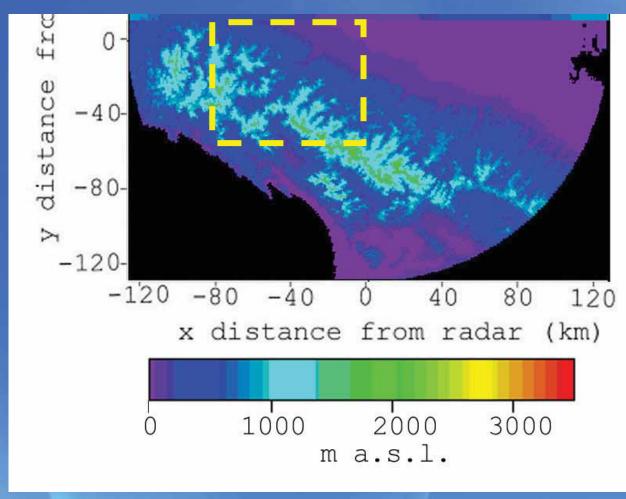


Just few samples



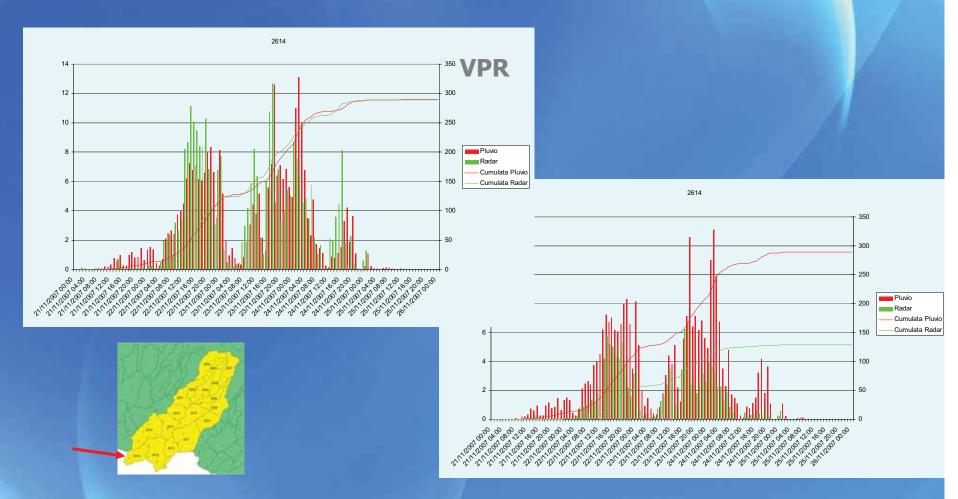


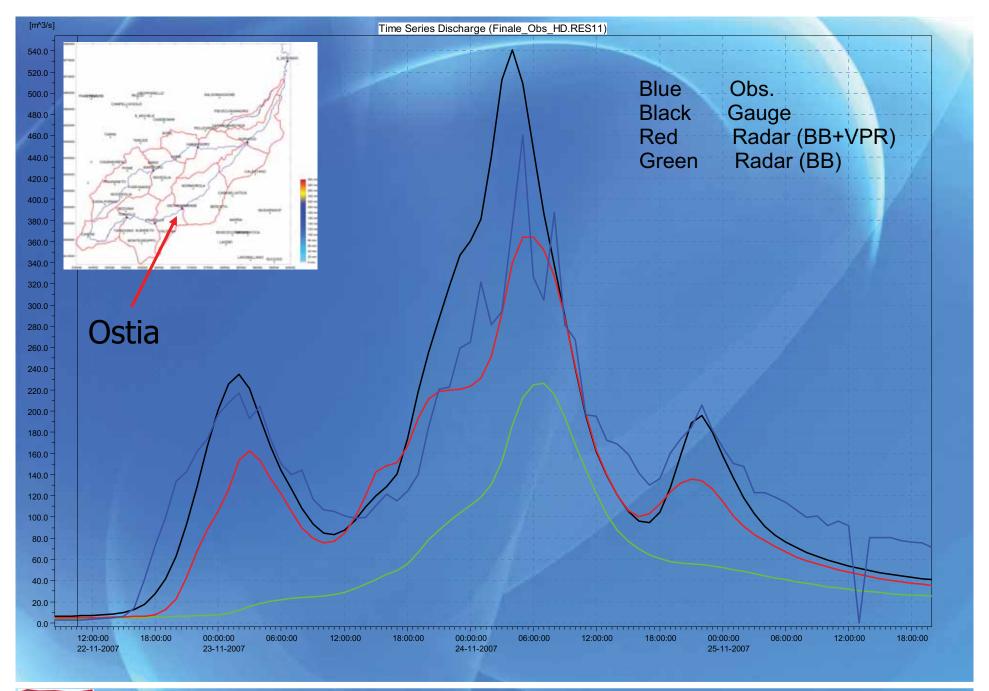
Radar visibility of the Taro Catchment



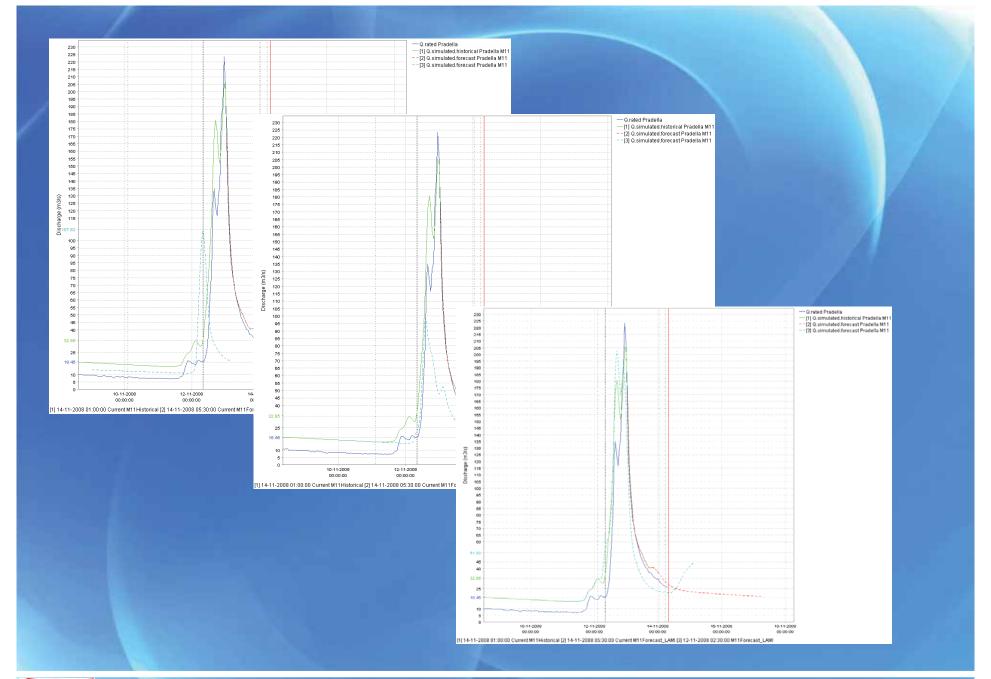
- DTM
- BB
- Elev. Selection
- Residual BB
- Height of beam center above dtm

Areal averaged precipitation timeseries









Thank you for your attention