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# A new Multi-Objective strategy to support model selection for environmental modelling

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# 1. Outline

## PURPOSES

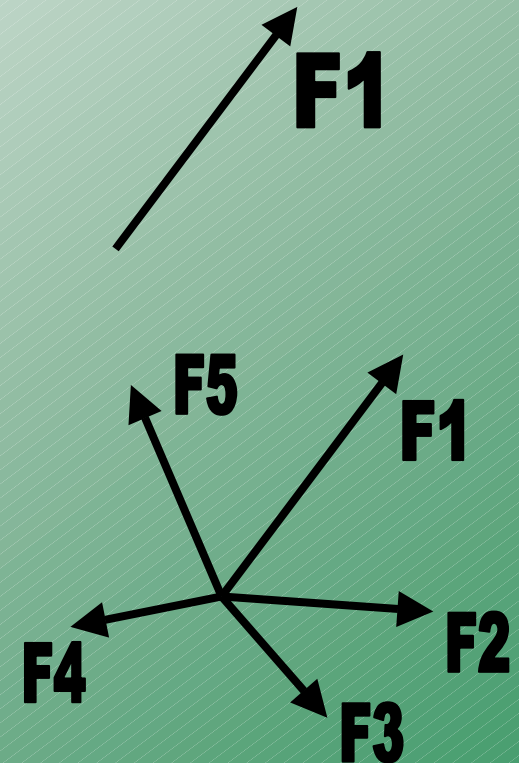
- Data augmentation
- On-line forecasting
- Off-line forecasting
- Scientific knowledge discovery

***EPR***

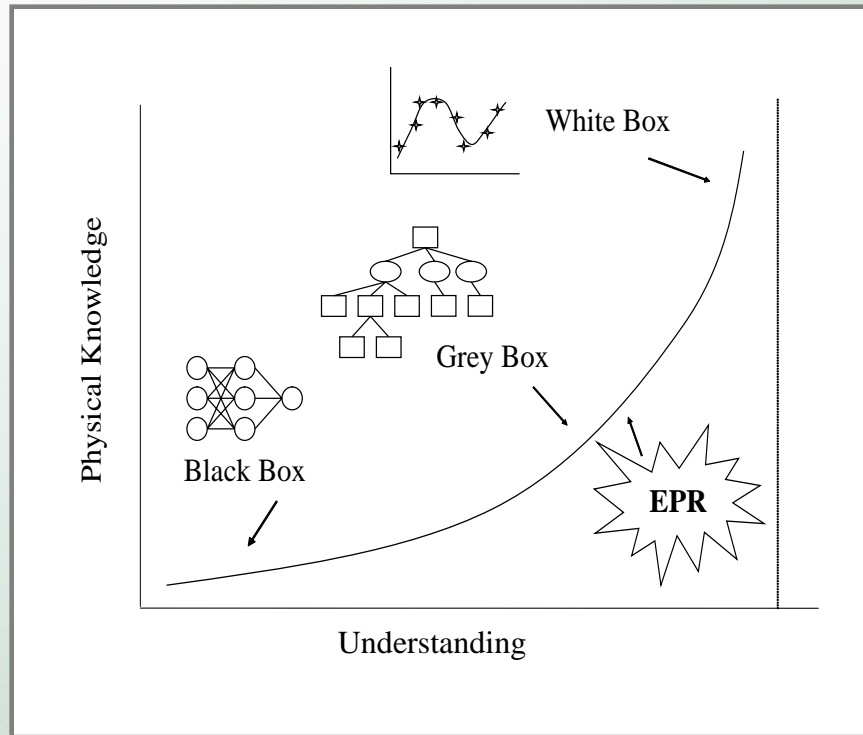
***Evolutionary Computing***

## 2. Multidimensional Modelling

- Old strategies:
  - Single-Objective approach.
    - One objective function is optimized.
    - One-dimensional scenario.
- Novel strategies:
  - Multi-Objective approach.
    - Multipurpose modelling.
    - Multidimensional scenario: fitness vs. complexity.

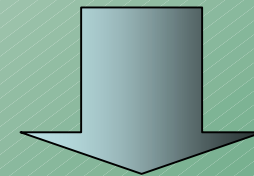


# 3.1 Evolutionary Polynomial Regression



$$y = \sum_{j=1}^m F(\mathbf{X}, f(\mathbf{X}), a_j) + a_0$$

**Hybrid Paradigm**

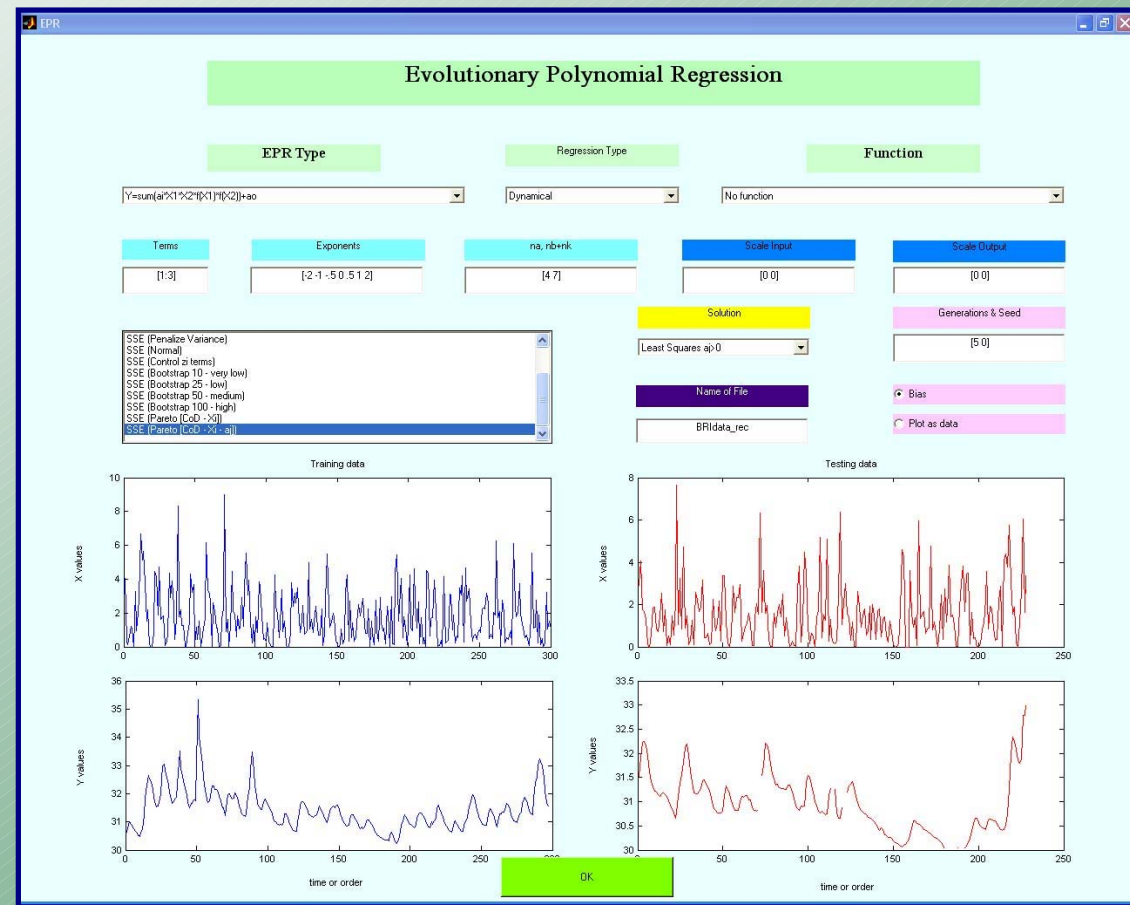


**GA + LS**

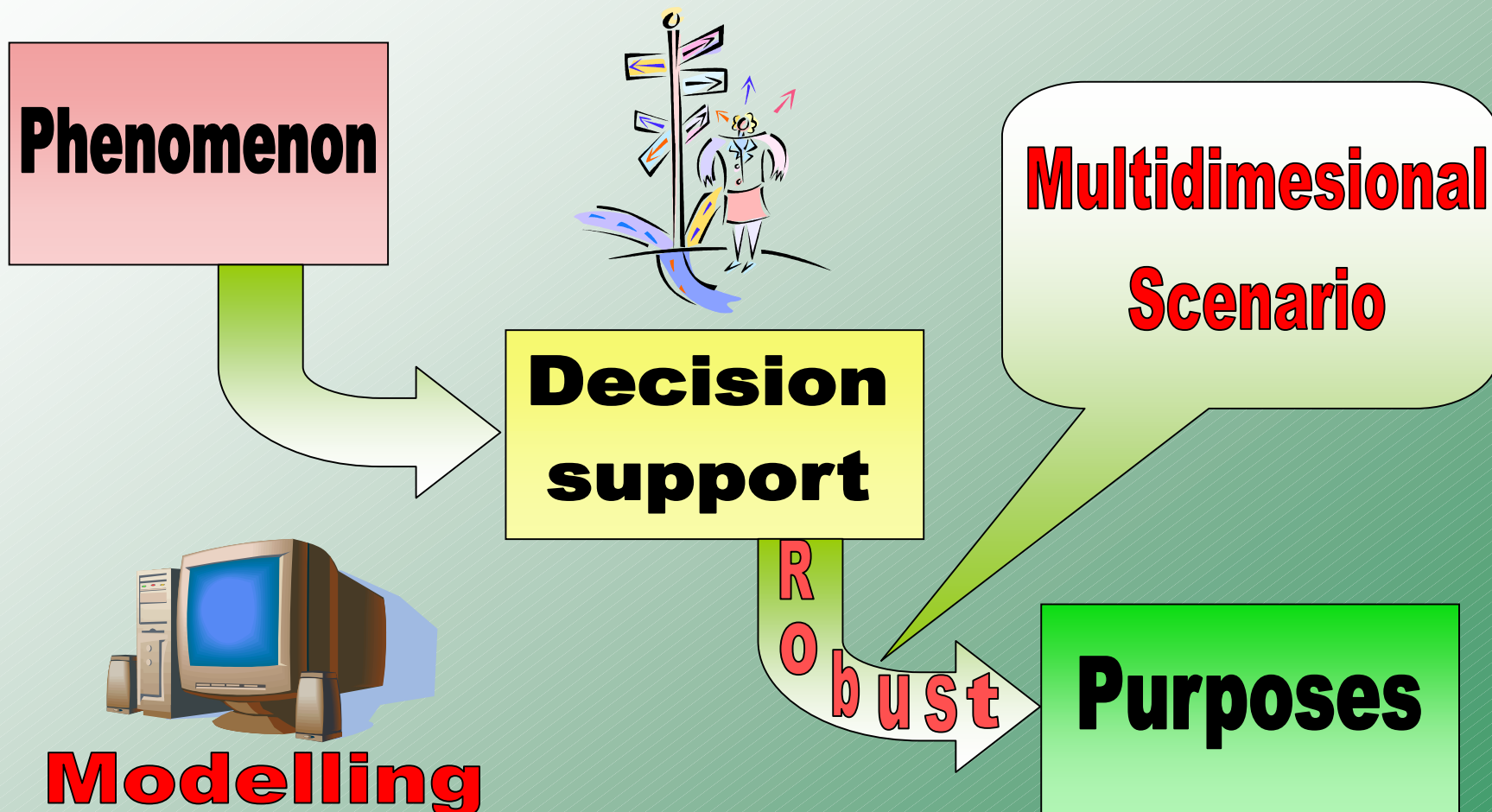
$$W_t = C_{\text{adim\_EPR}} = \sqrt{\frac{8}{f}} = 0.4746 \frac{ds}{hs} \cdot \ln \frac{2R}{ds} + \left( 29.6138 \ln \frac{hs}{ds} + 3.9319 \frac{R \cdot S}{ds \cdot \frac{\mathbf{R}_w}{10^5}} + 50.55 \right) 59914$$

## 3.2 MO-EPR: features

- Global method
- Parsimonious
- Non – overfitting
- Multiple models
- Interactive
- Multi-Objective
- LS  $a_j > 0$

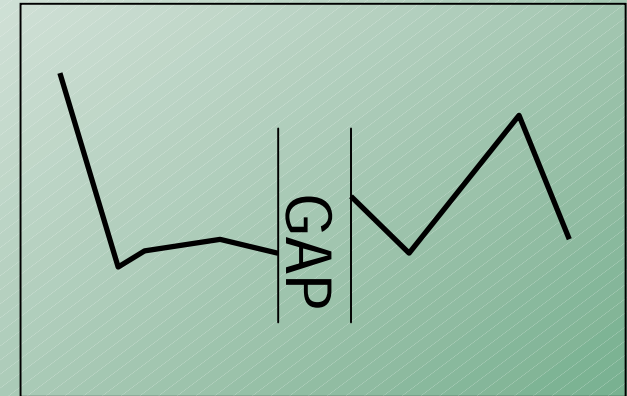


# 4. Environmental phenomena

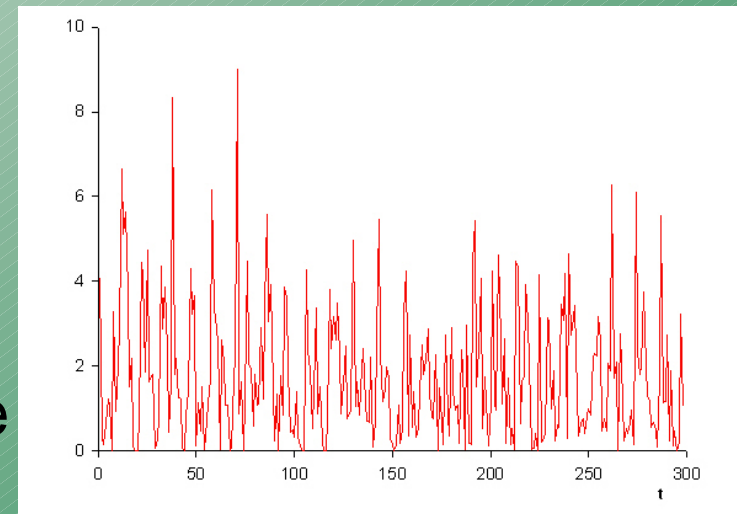


## 4.2 Environmental phenomena

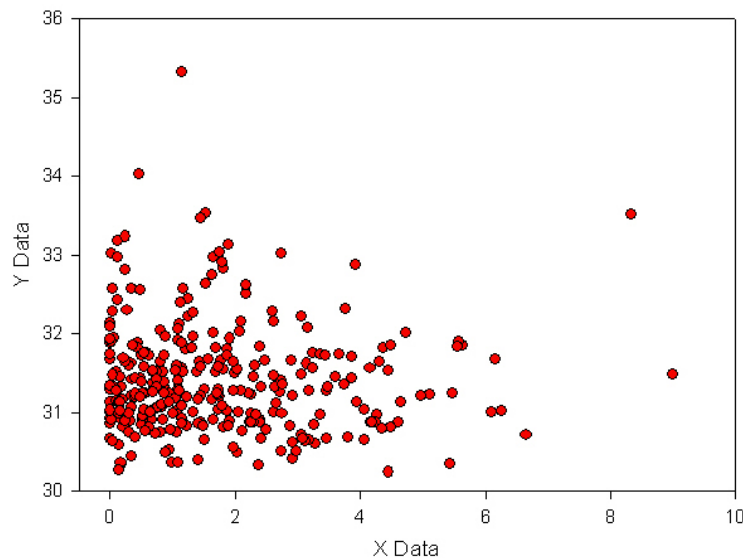
Missing data



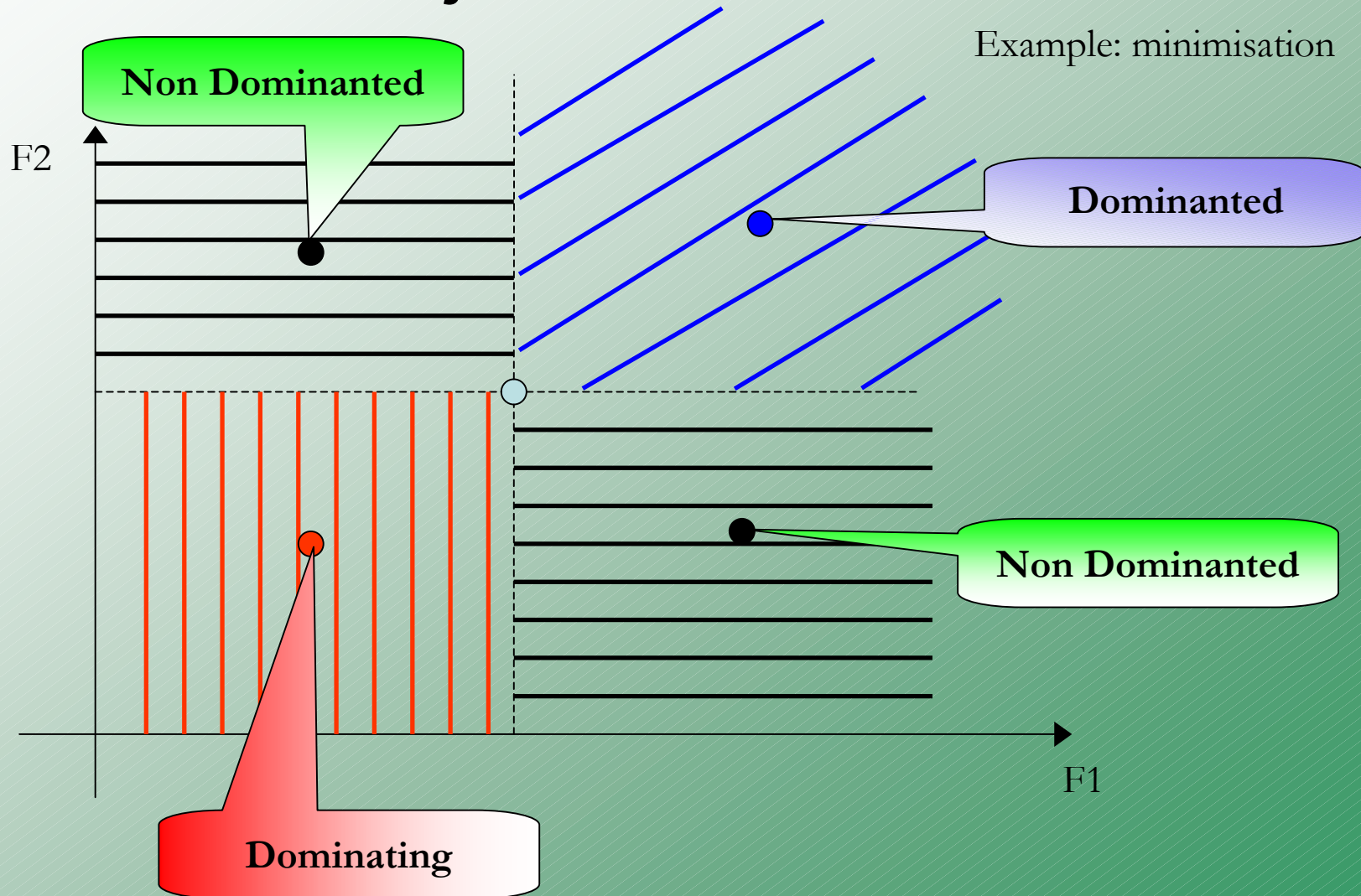
Non-linear dynamics



Non Gaussian noise



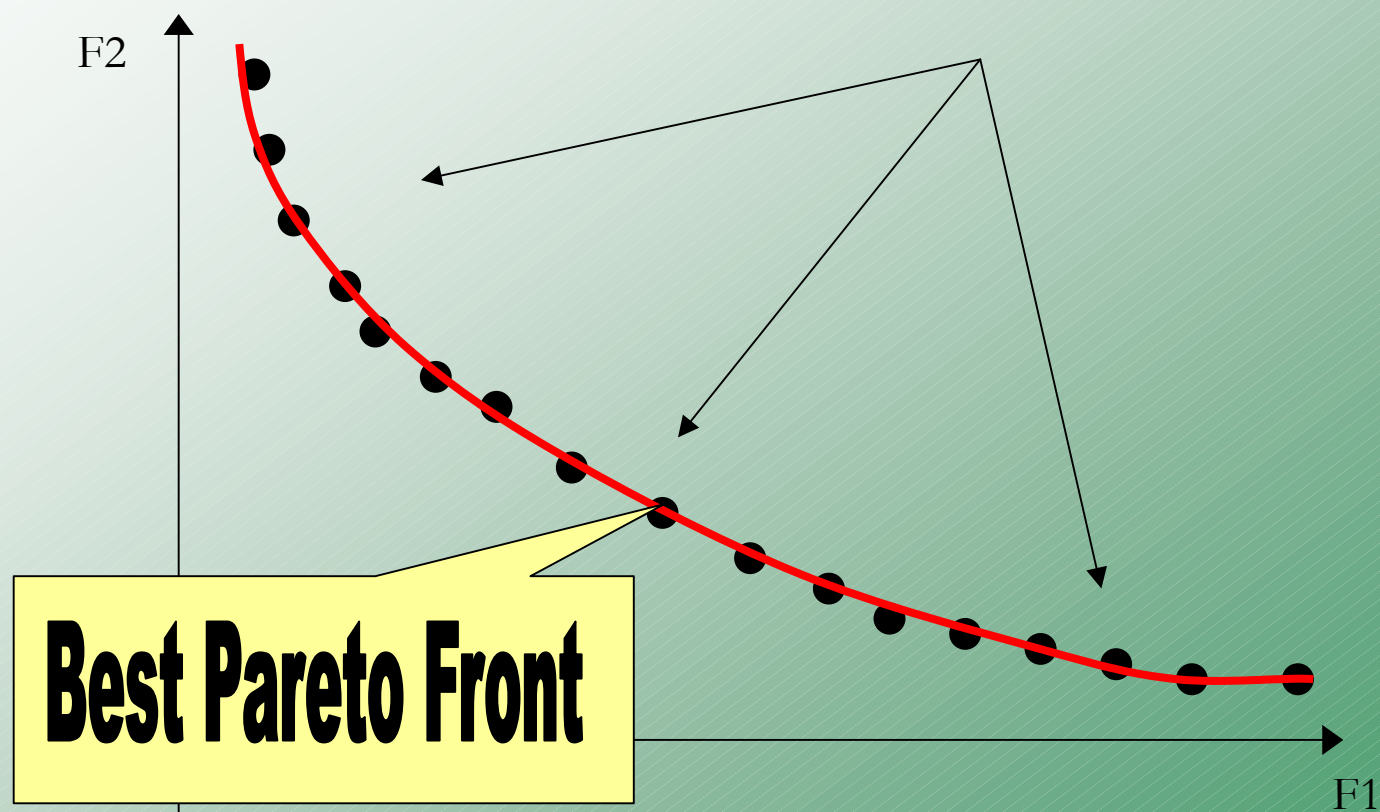
# 5.1 Multi-Objective search



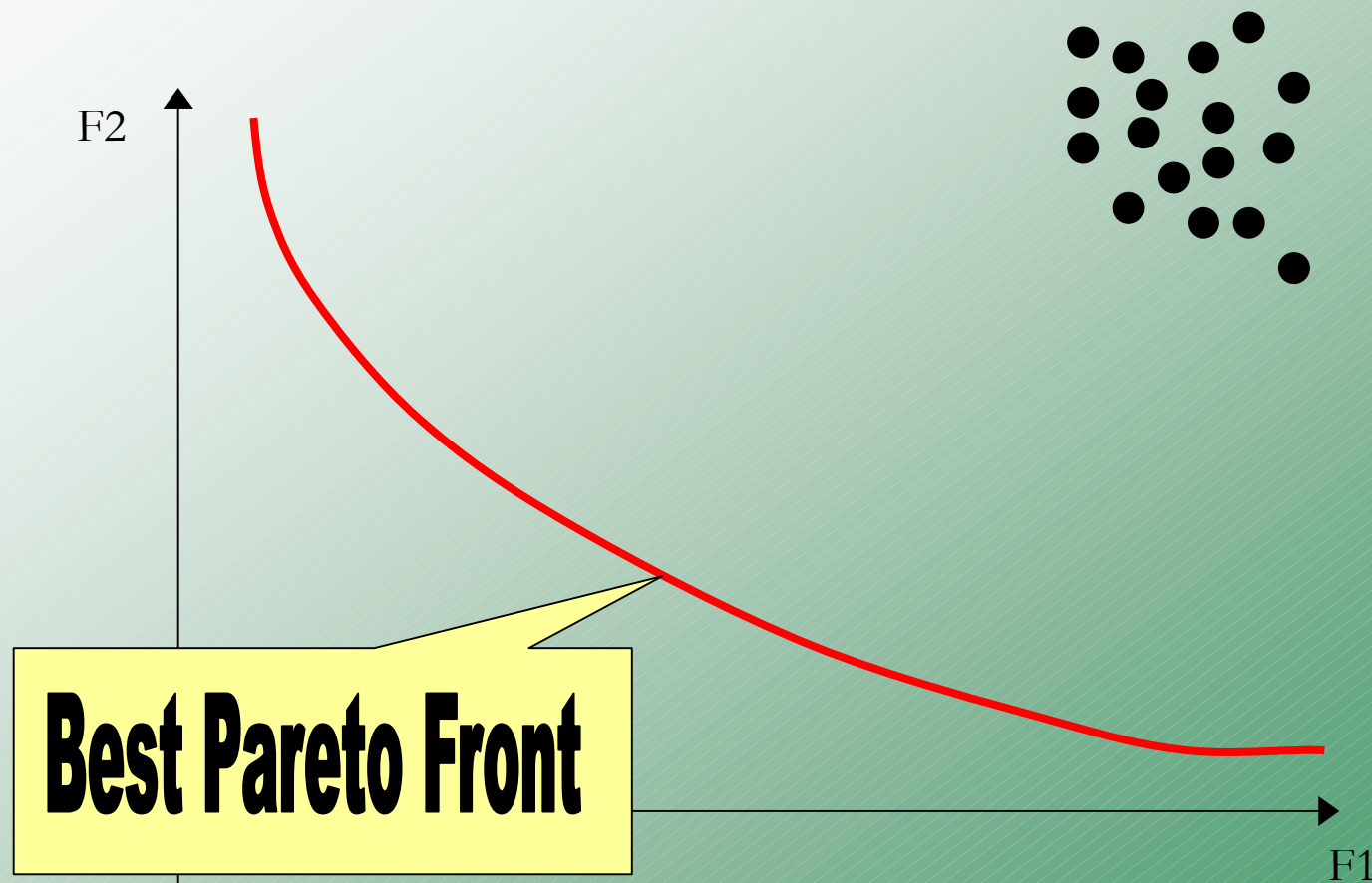
# 5.2 Multi-Objective search



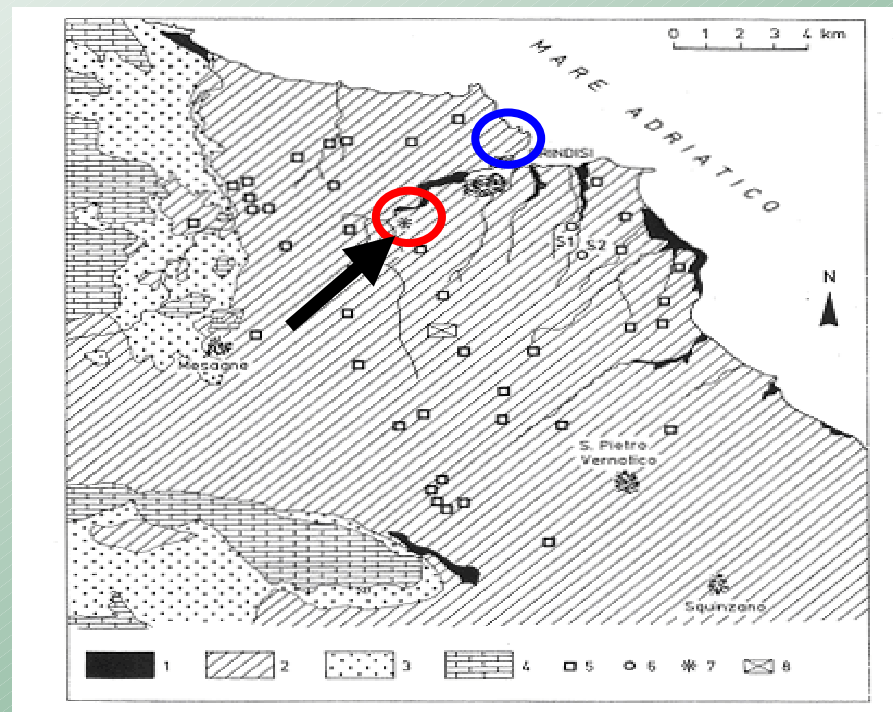
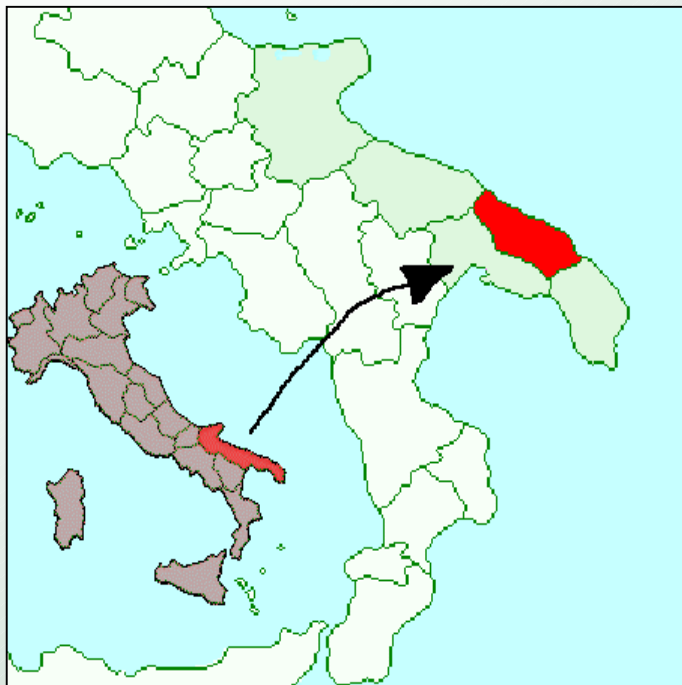
## 5.2 Multi-Objective search



## 5.2 Multi-Objective search

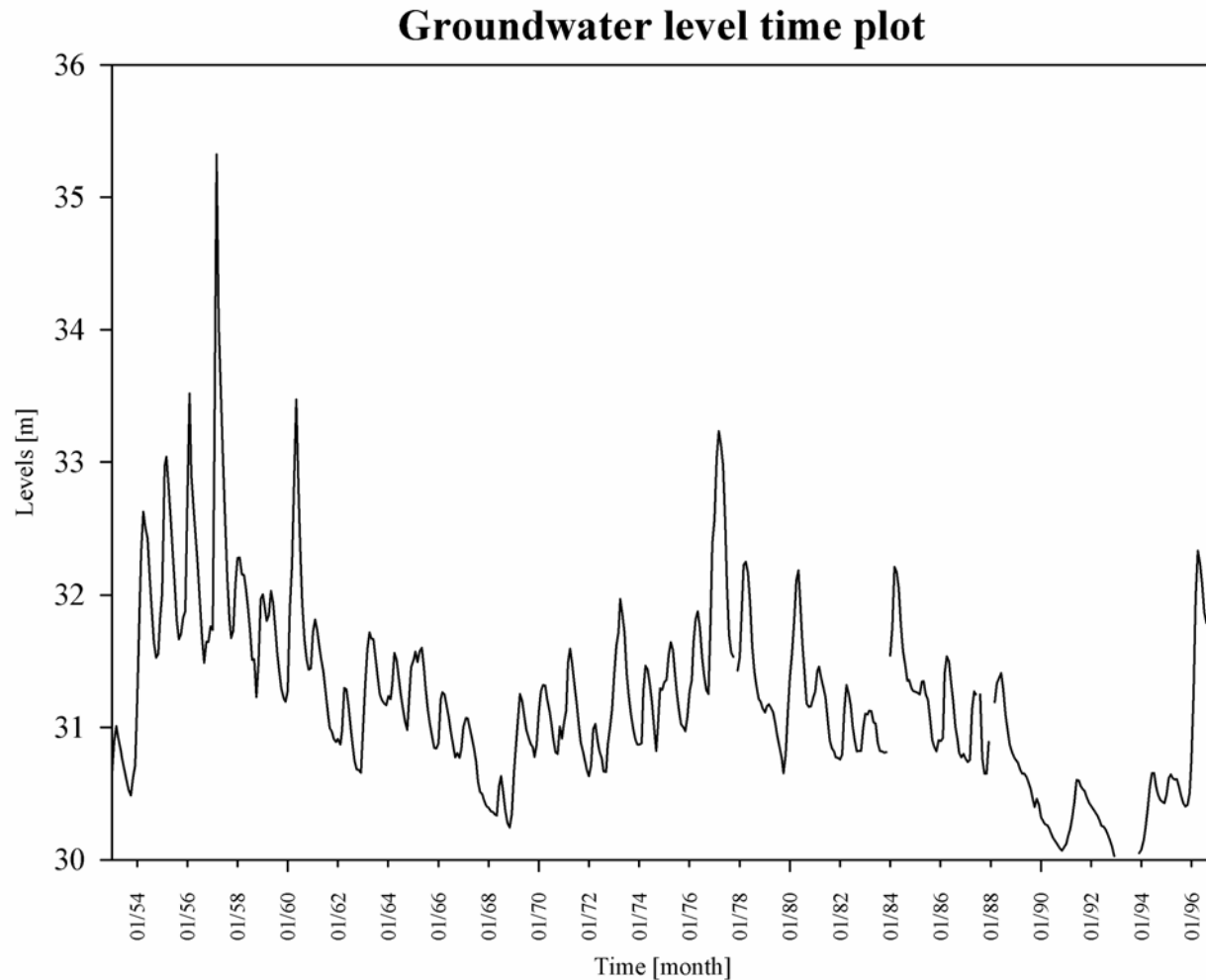


## 6.1 Case study



- Rainfall vs. Groundwater levels
- Brindisi - ITALY
- Monthly sampled data

# 6.2 GW time series

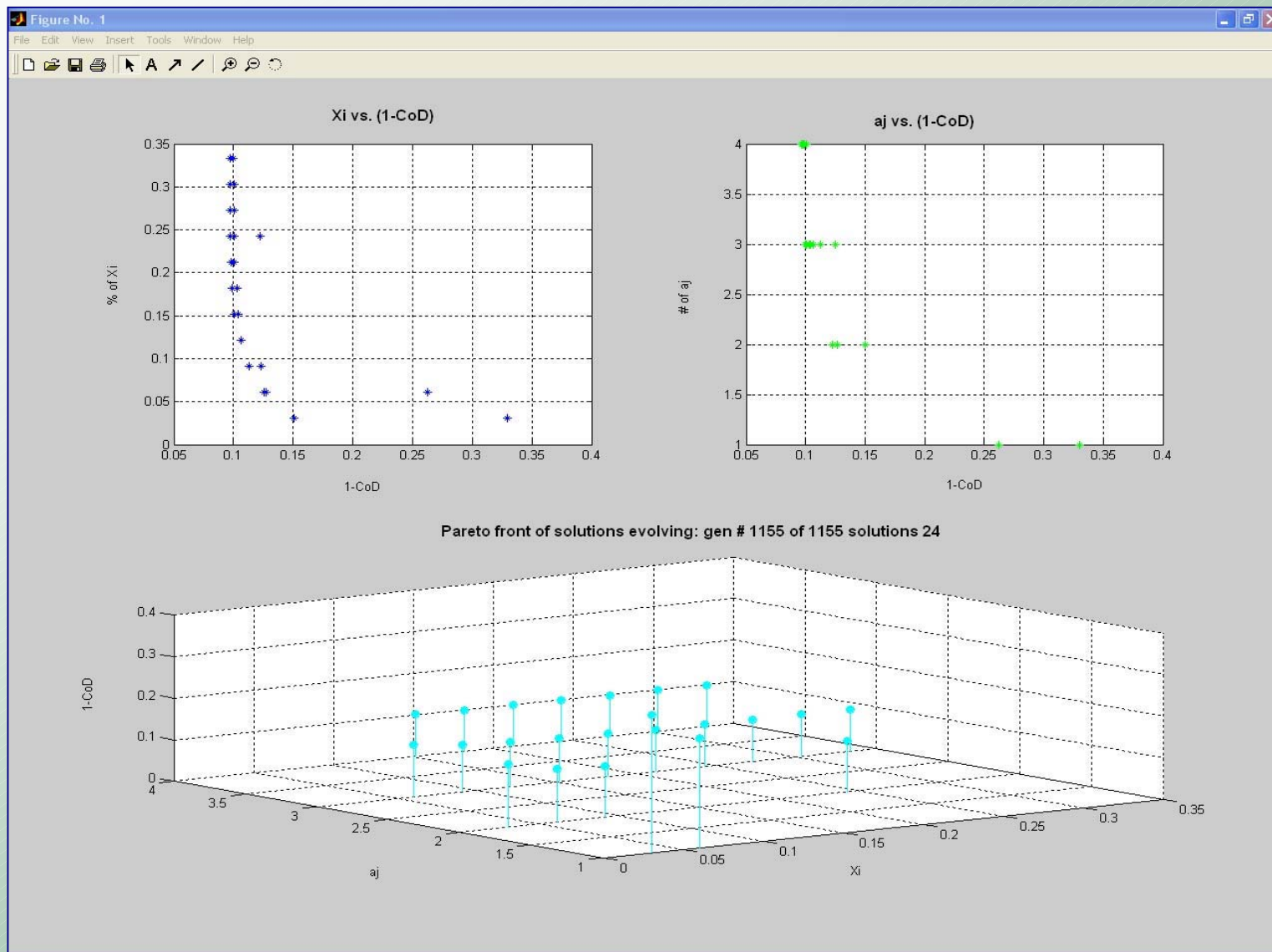


## 6.3 Case study: assumptions

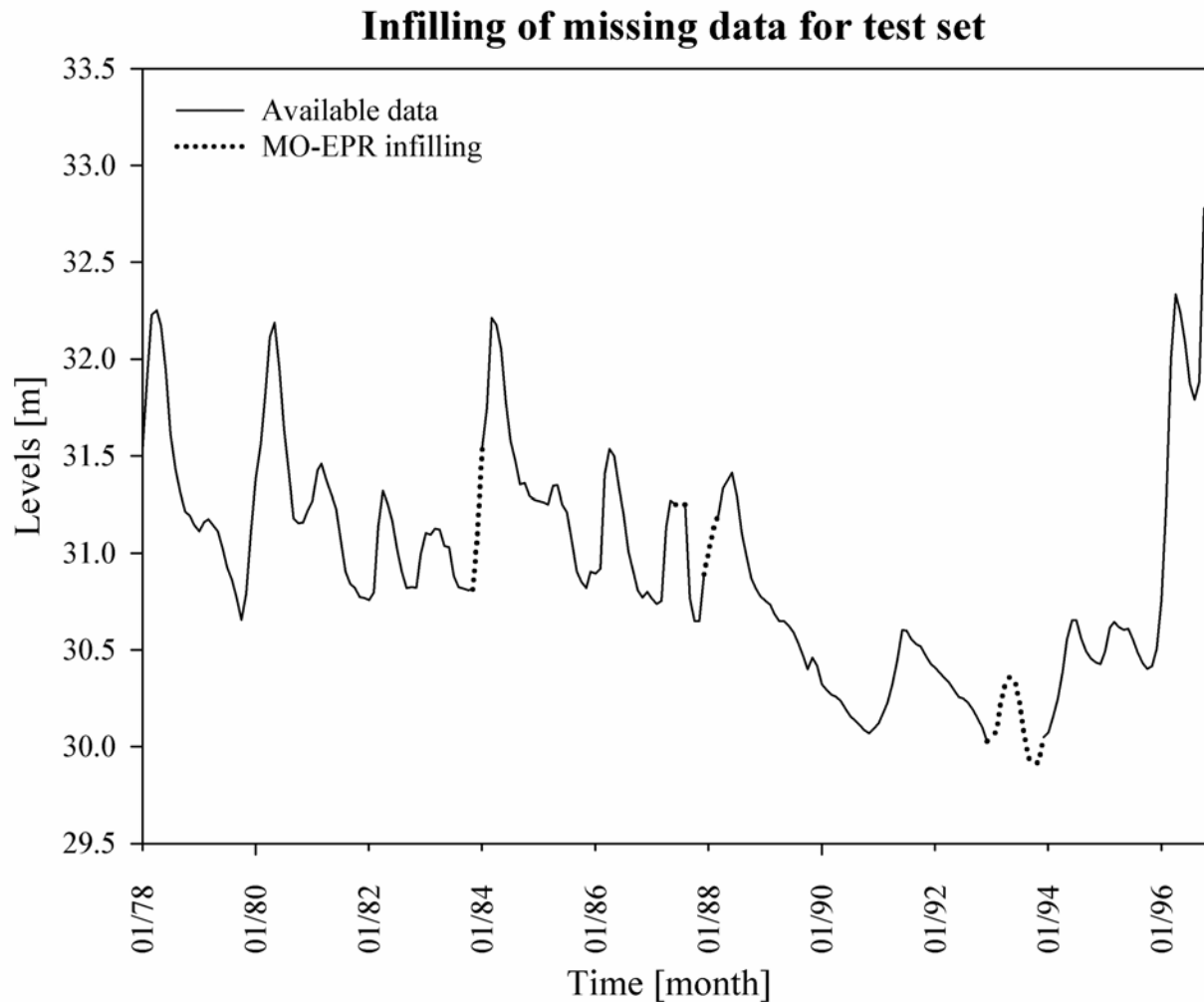
- Polynomially structured
- $m_{max} = 4$
- Training set
- 3 objective functions:
  - Fitness
  - # of monomial BB (complexity)
  - # of time each input is involved (complexity)
- Model evaluation on unseen data

$$y = \sum_{j=1}^m F(\mathbf{X}, f(\mathbf{X}), a_j) + a_0$$

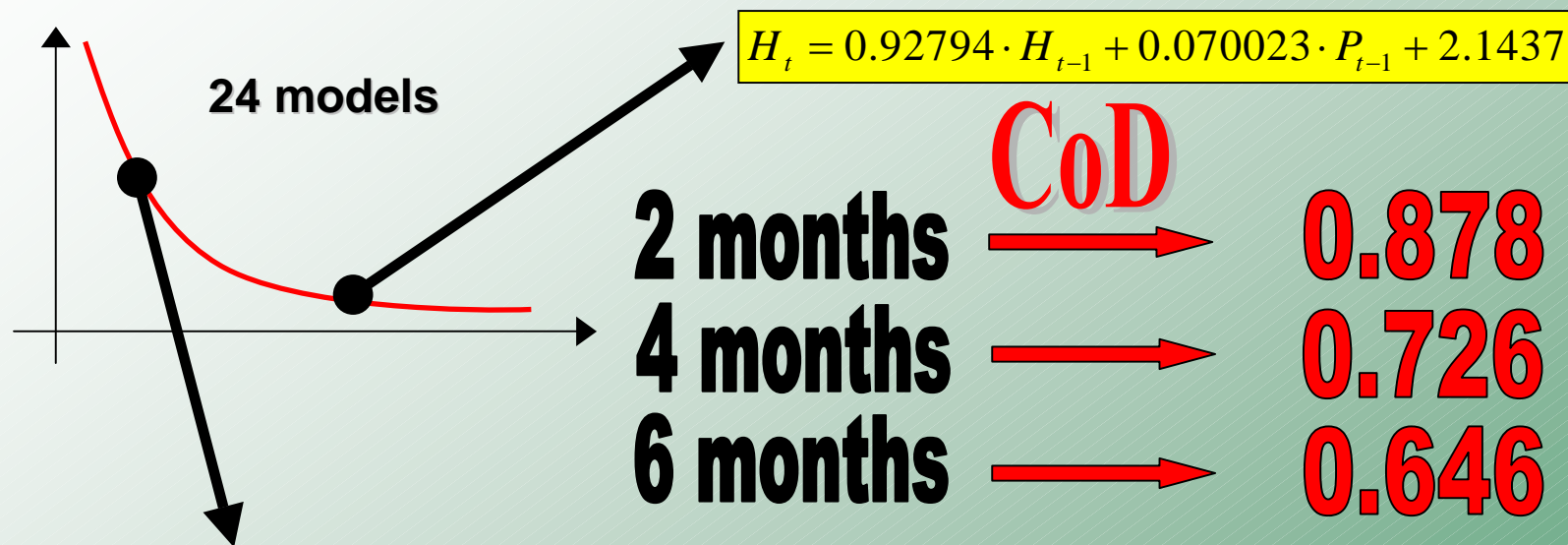
# 6.4 Pareto Front of solutions



# 6.5 Purpose: infilling of data



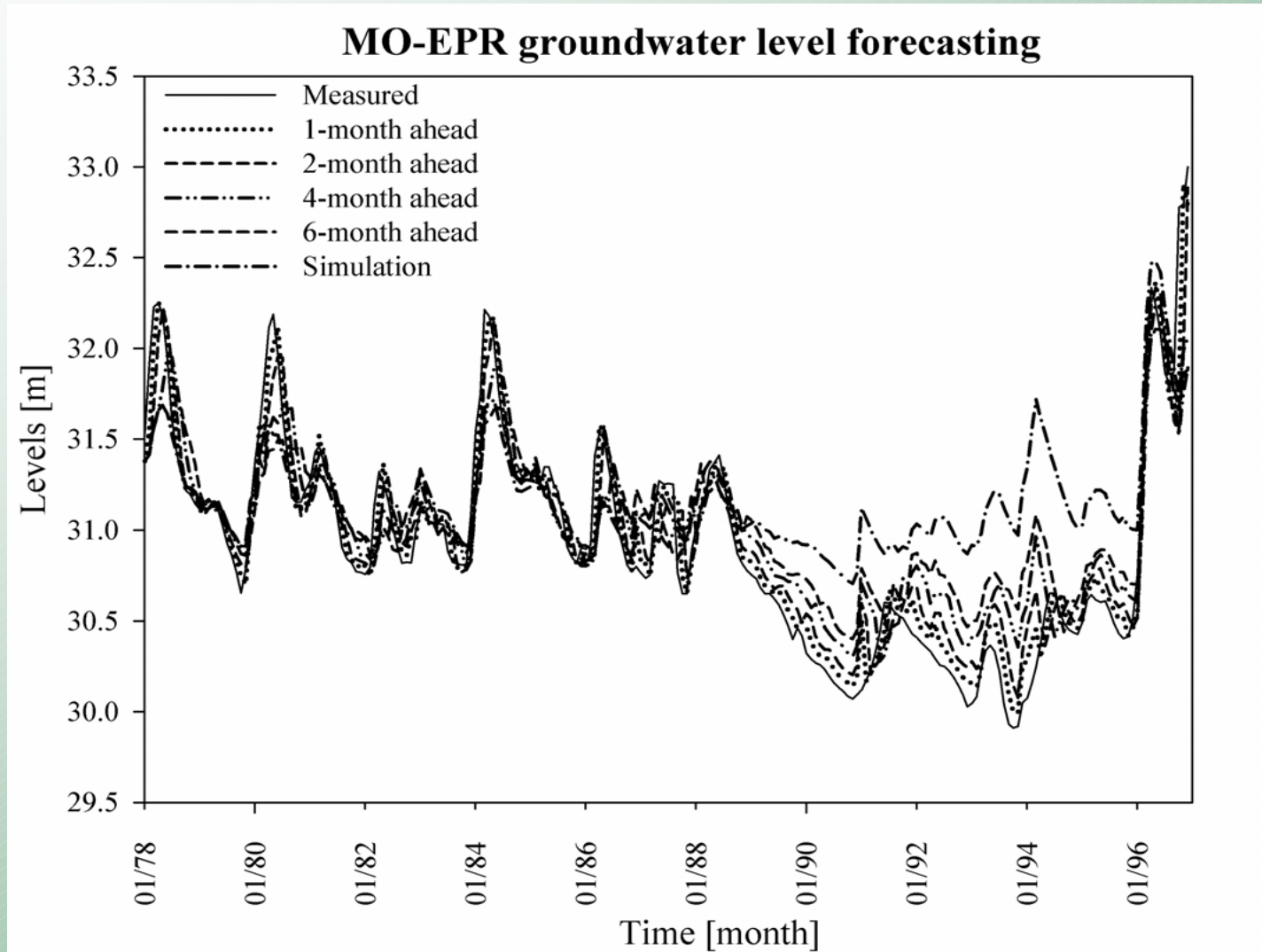
## 6.6 MO-EPR results – Test set



$$H_t = 0.10417 \cdot H_{t-1}^2 \cdot H_{t-2}^{-0.5} + 0.024556 \cdot P_{t-1} \cdot P_{t-2}^{0.5} \cdot P_{t-3}^{0.5} + 0.0054314 \cdot P_t^2 + 12.9706$$

1 months	CoD	0.956
Off-line prediction		0.242

# 6.7 MO-EPR results - Test set



# 7. Conclusions

- Multidimensional modelling:
  - Not only statistically based choice
  - “Pareto” surface based choice
- Wide range of solution/purposes:
  - Infilling
  - Prediction
  - Etc.
- Proposal: MO-EPR (Not unique!)

# 7. Conclusions

<http://www.poliba.it/taranto/software/hydroinformatics/index.htm>

**Thank you!**