

# *Probability, Agnosticism, and Guarantees in Inductive Learning Processes*

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*(joint work with Simone Garatti and Algo carè)*



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Marco Fuhrman 60th birthday



“our” classroom at the Politecnico

Marco Fuhrman 60th birthday

# *Probability, Agnosticism, and Guarantees in Inductive Learning Processes*



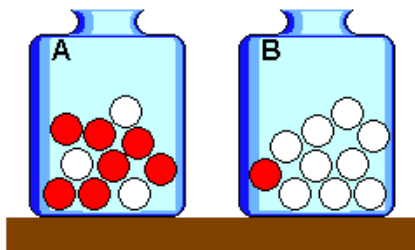
$R$  nominal value 2G Ohm  
standard deviation = 0.1G Ohm

$$Y = R + N$$

- *construct bi-variate distribution for  $(R, Y)$   $\longrightarrow$  use Bayes rule*
- *what does prior on  $R$  describes?  $\longrightarrow$  subjectivist interpretation of probability*

## a misconception

- *subjectivism goes hand in hand with fully probabilistic models (and Bayesian updating)*



$$p = \frac{m}{50}, \quad m \in \{0, \dots, 50\}$$

# agnosticism

- *Every probability distribution is possible*

→ *practical reasons*

→ *philosophical reasons*

# agnosticism

- *Every probability distribution is possible*

→ *practical reasons*



*goal: predict whether the defibrillator shock will be effective*



## agnosticism

- *Every probability distribution is possible*

→ *practical reasons*

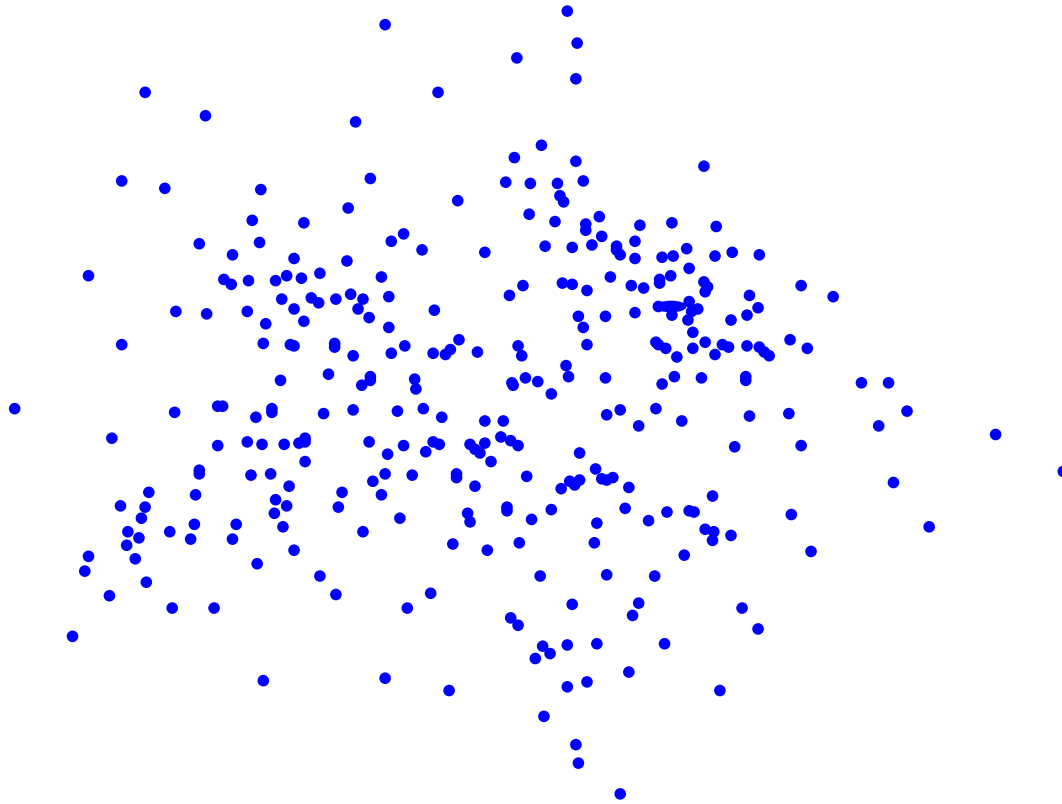
→ *philosophical reasons*

*can knowledge be created out of lack of knowledge in the light of observations?*

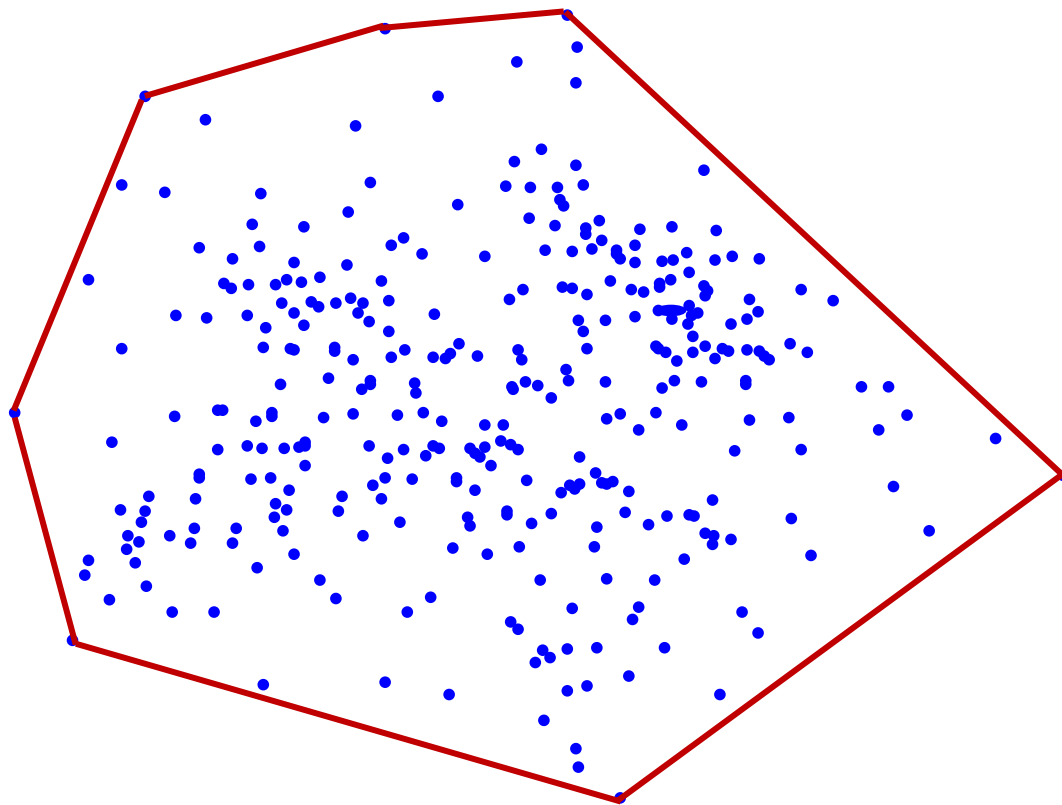
*is it possible to derive mathematically  
rigorous and practically useful data-driven  
techniques within an agnostic setup?*

convex hull

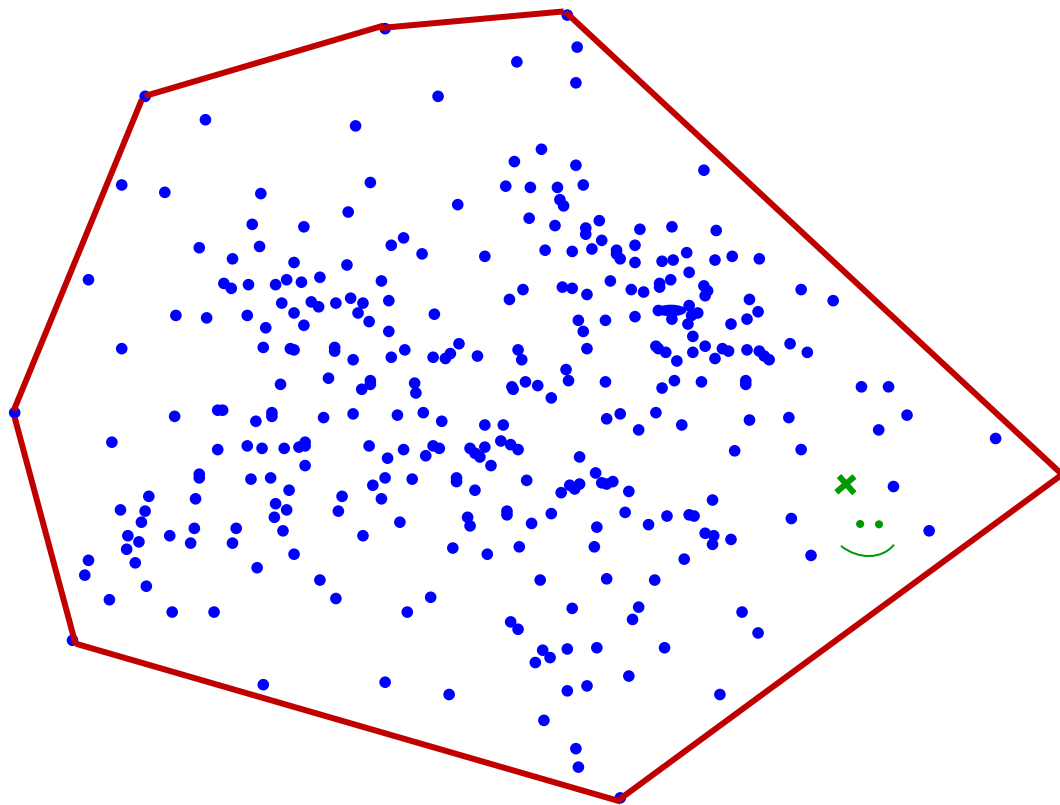
1000 points



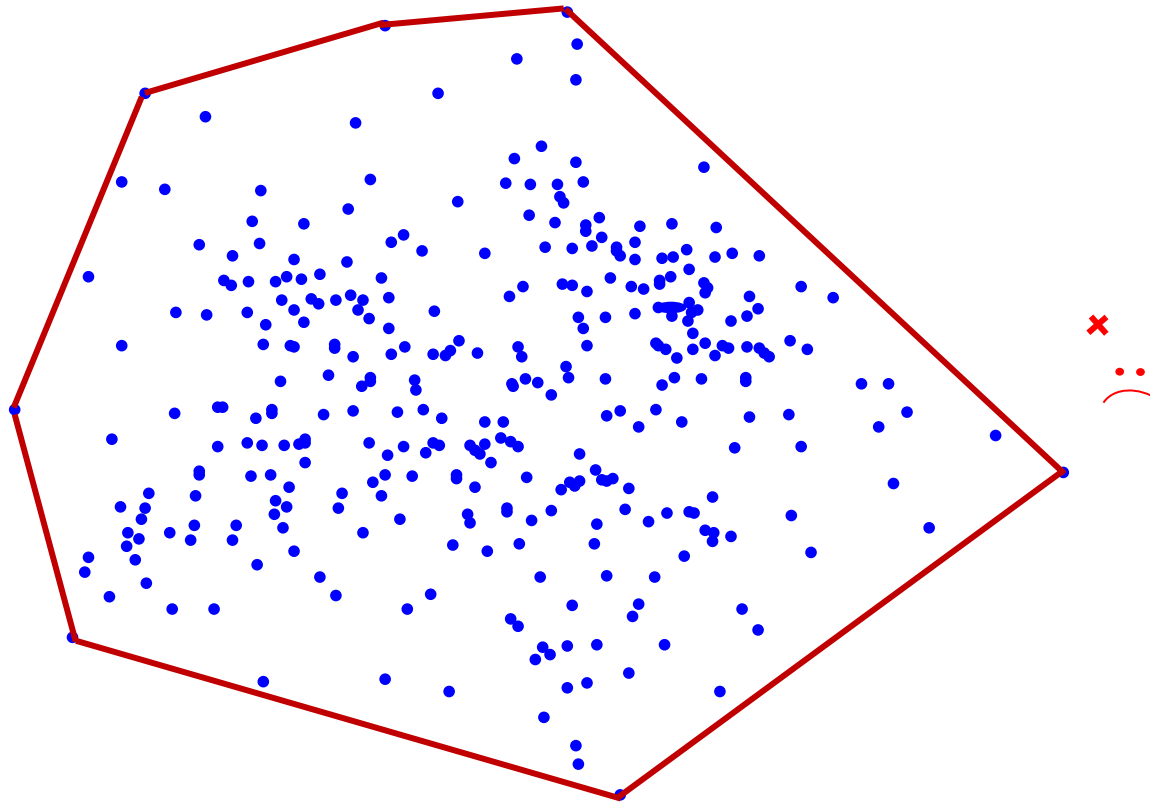
1000 points



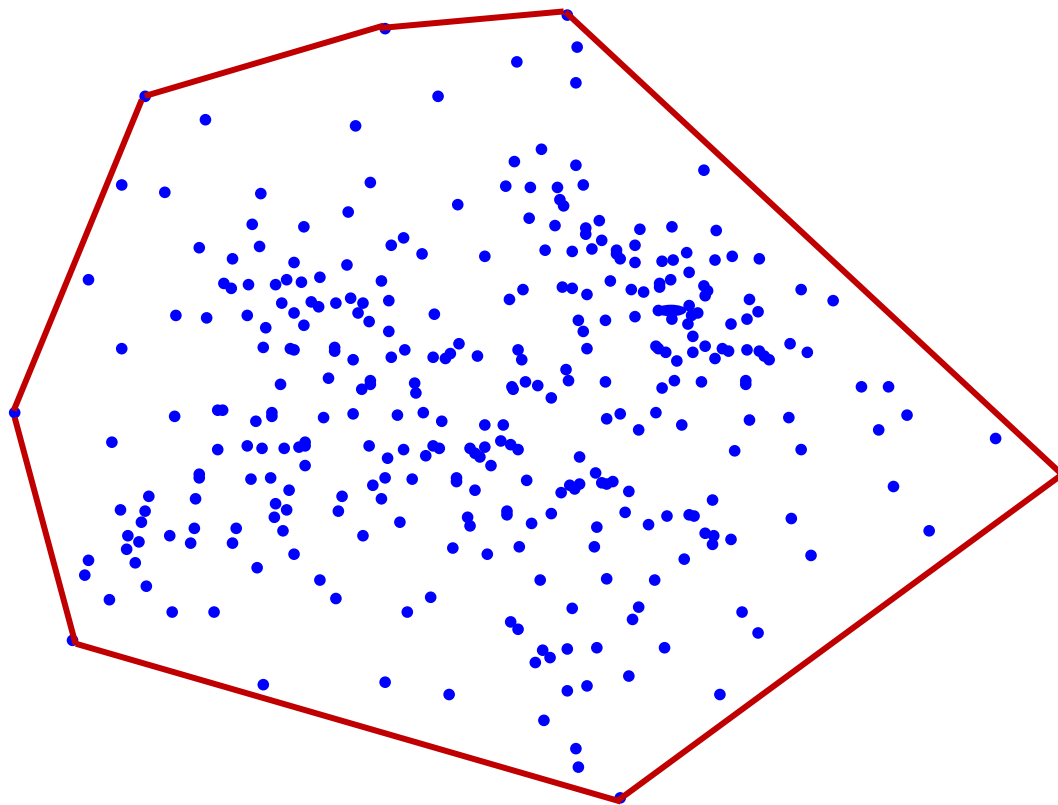
1000 points



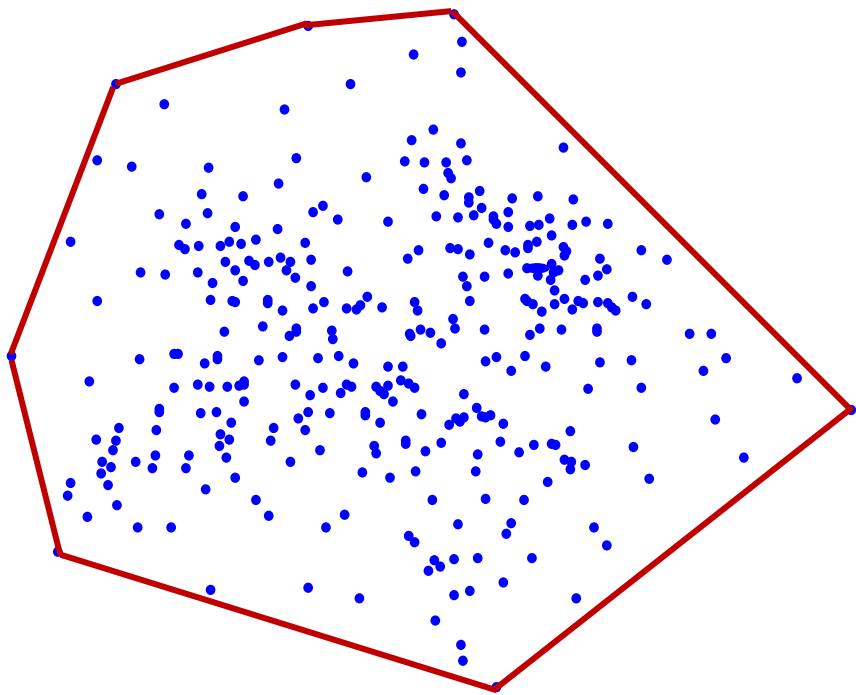
1000 points



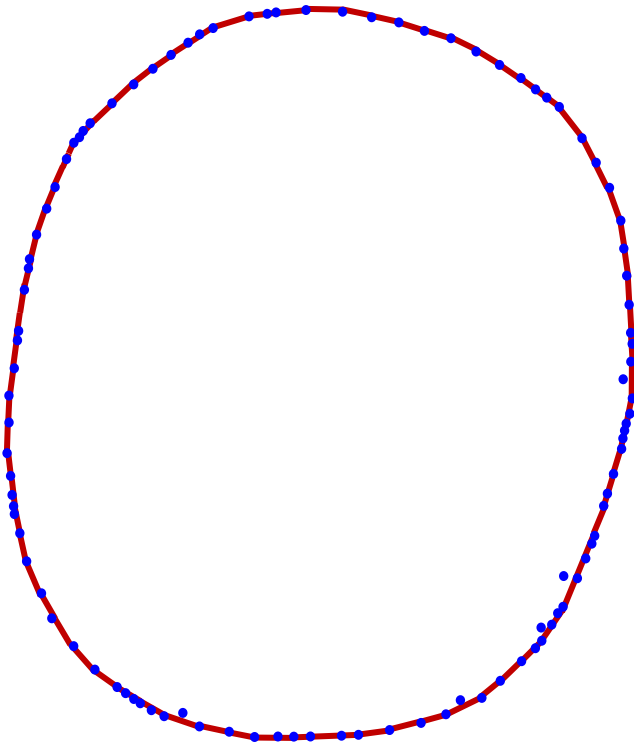
1000 points



7 points at the boundary



7 points at the boundary



988 points at the boundary

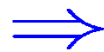


complexity

~~$p_1$~~ ,  $p_2$ ,  ~~$p_3$~~ ,  ~~$p_4$~~ ,  $p_5, \dots, \del{p_{1000}}$



“7 points at the boundary”



“if only these 7 points are maintained,  
“Opt” gives the same solution”

complexity

~~$p_1$~~ ,  $p_2$ ,  ~~$p_3$~~ ,  ~~$p_4$~~ ,  $p_5, \dots, \del{p_{1000}}$



“7 points at the boundary”  $\Rightarrow$  “if only these 7 points are maintained,  
“Opt” gives the same solution”

*complexity of  
representation*

intuitively: high complexity  $\Rightarrow$  high risk

## wait-&-judge scenario optimization

Theorem (with S. Garatti)

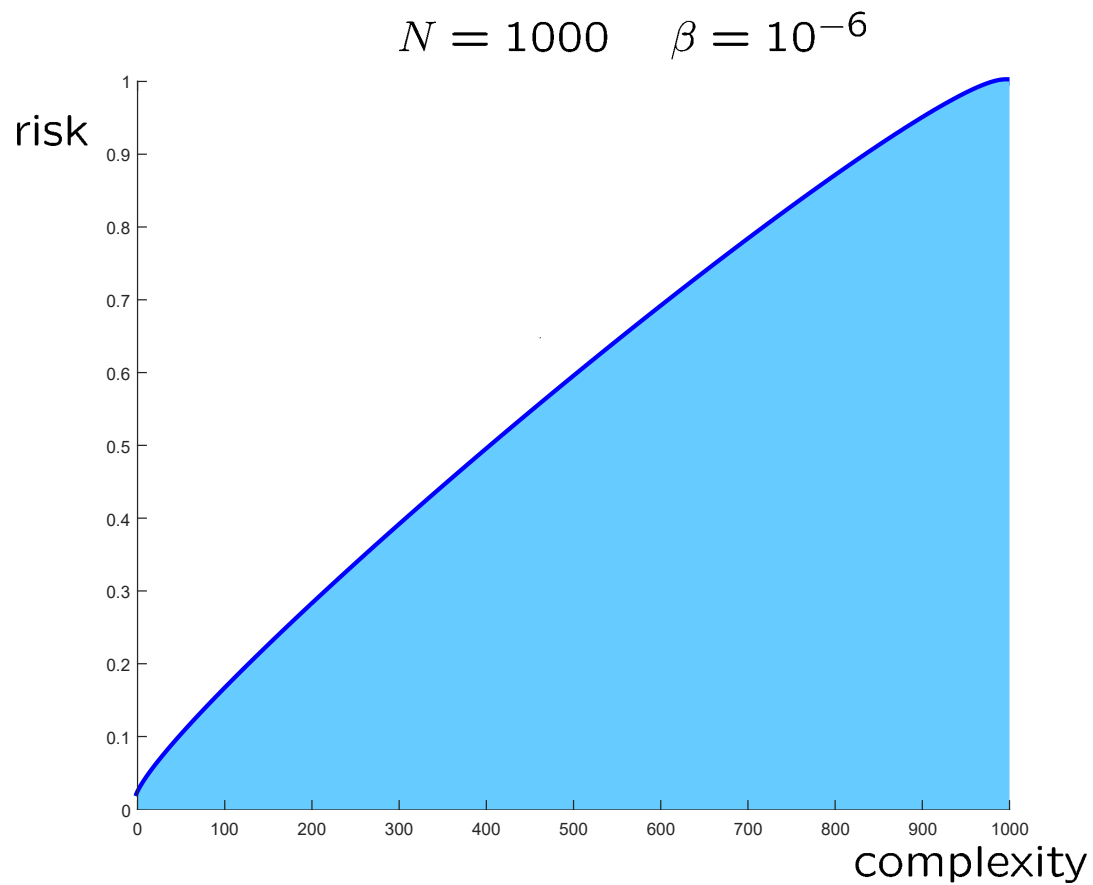
$$Pr\{\text{risk} \leq \epsilon(\text{complexity})\} \geq 1 - \beta$$

where

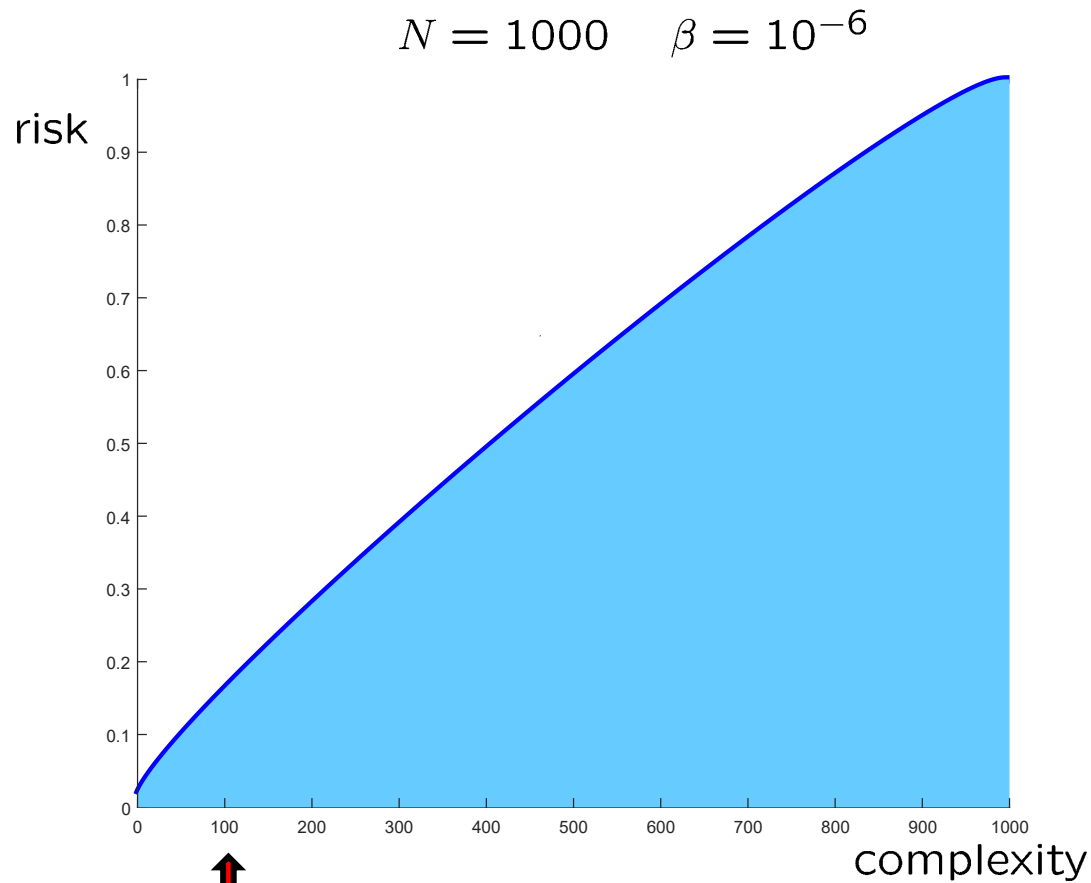
$$\beta = \inf_{\xi(\cdot) \in \mathcal{P}_N} \xi(1)$$

$$\text{subject to: } \begin{aligned} \frac{1}{k!} \frac{d^k}{dt^k} \xi(t) &\geq \binom{N}{k} t^{N-k} \cdot \mathbf{1}\{t \in [0, 1 - \epsilon(k)]\} & t \in [0, 1], \\ k &= 0, 1, \dots, N. \end{aligned}$$

# wait-&-judge scenario optimization

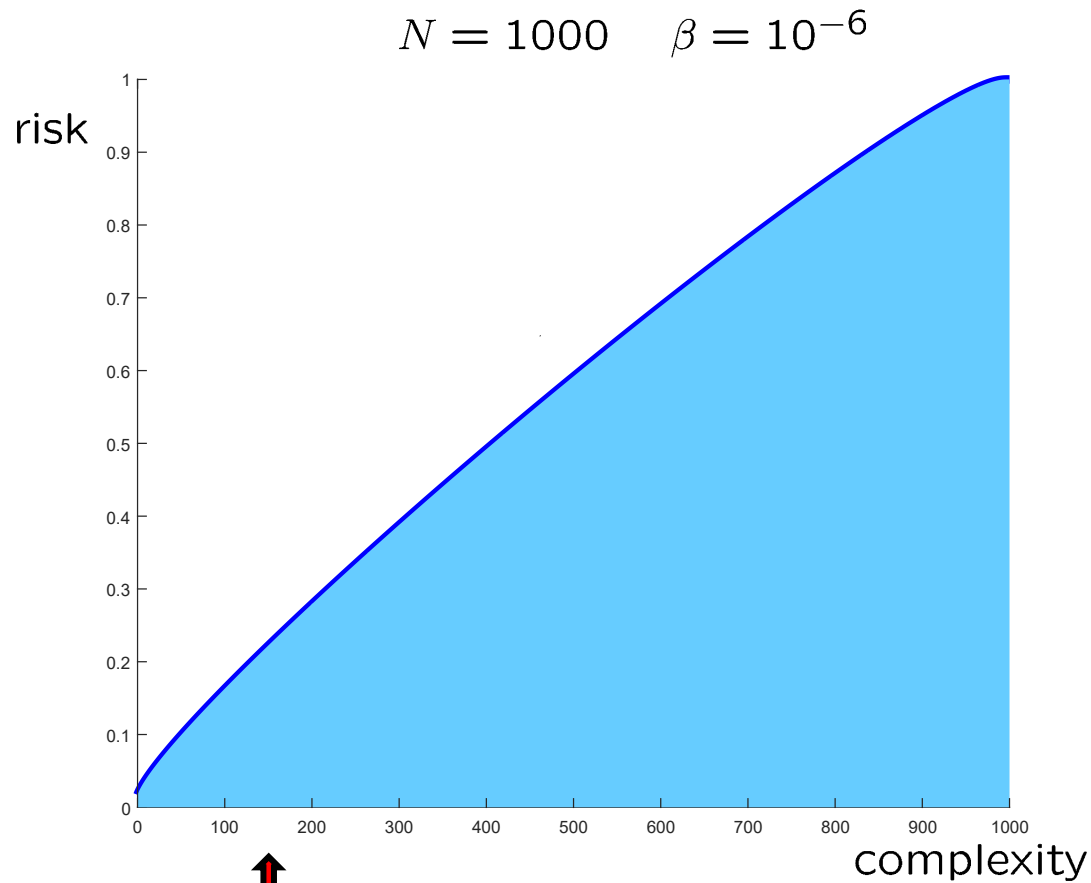


# wait-&-judge scenario optimization



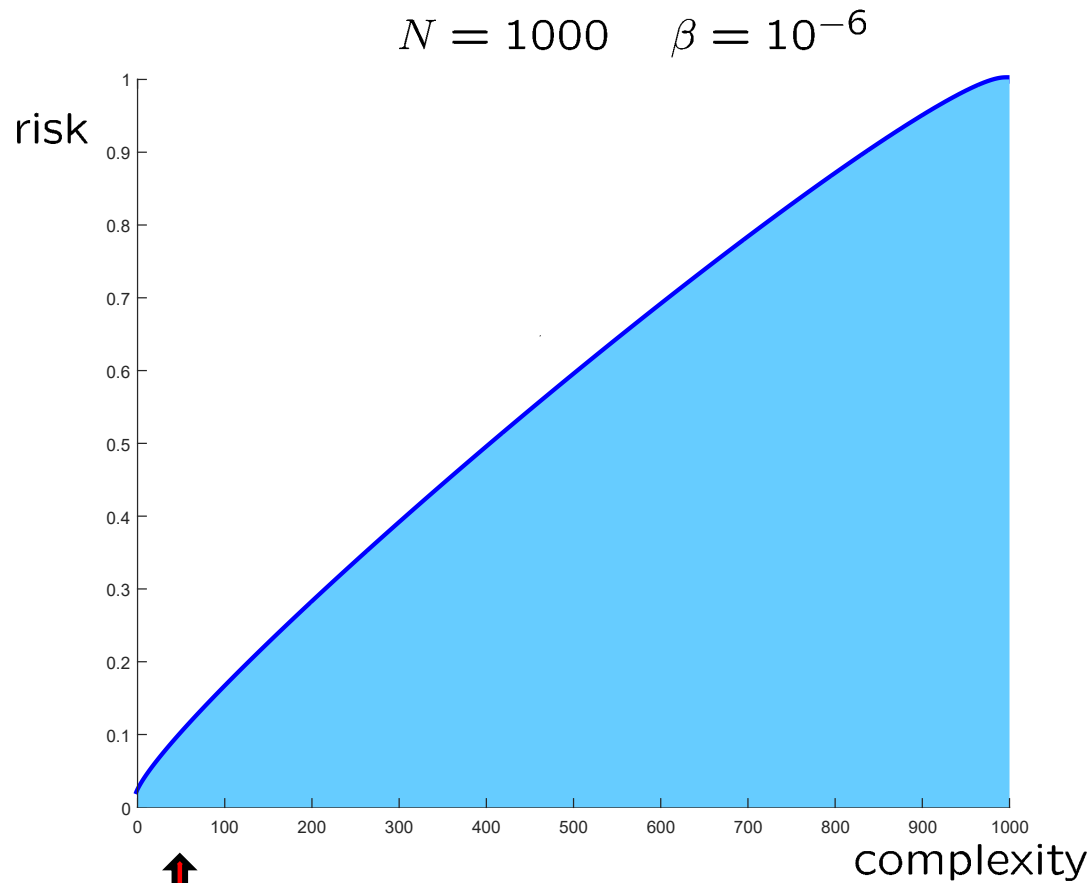
complexity = 100  $\rightarrow$  risk  $\leq 0.16$

# wait-&-judge scenario optimization



complexity = 150  $\rightarrow$  risk  $\leq 0.22$

# wait-&-judge scenario optimization



complexity = 50  $\rightarrow$  risk  $\leq 0.099$

## wait-&-judge scenario optimization

suppose: no concentrated mass

Theorem (with S. Garatti)

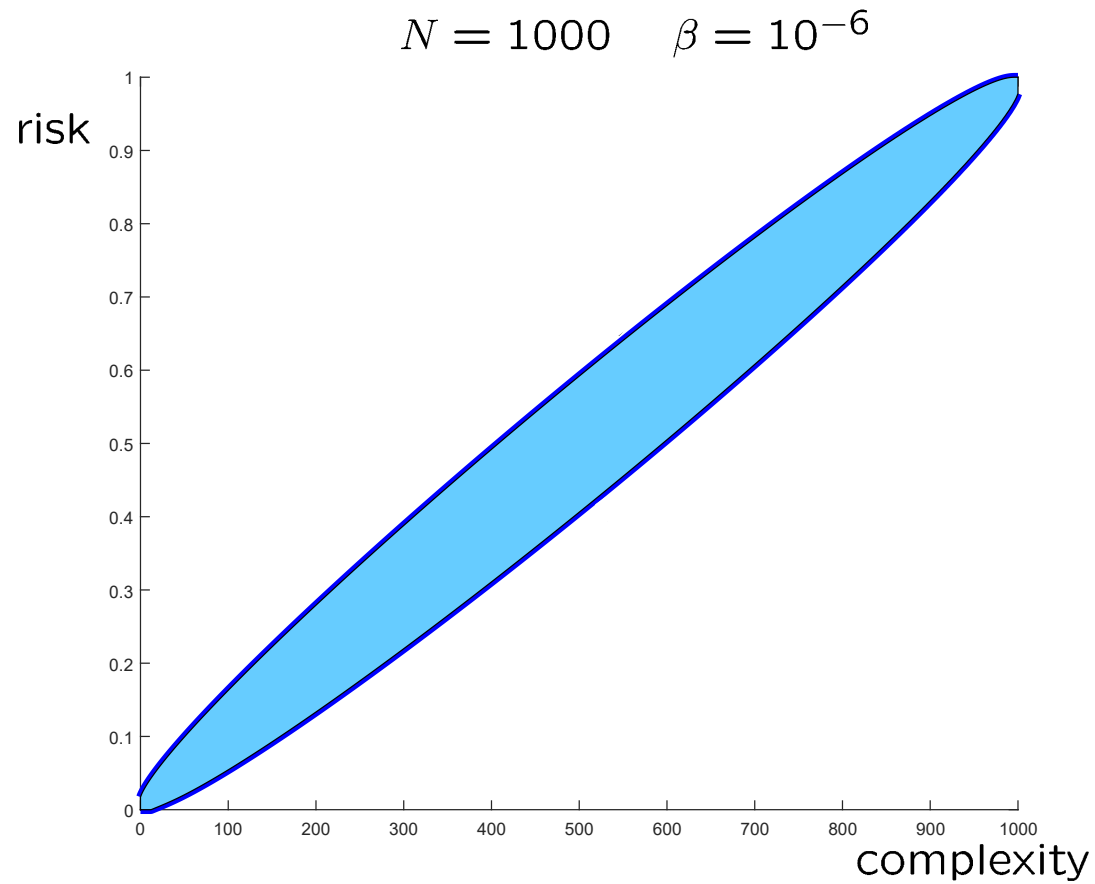
$$\Pr\{\underline{\epsilon}(\text{complexity}) \leq \text{risk} \leq \bar{\epsilon}(\text{complexity})\} \\ \geq 1 - \beta$$

where, with the position  $t = 1 - \epsilon$ ,  $\underline{\epsilon}$  and  $\bar{\epsilon}$  are given by:

$$\binom{N}{k} t^{N-k} - \frac{\beta}{2N} \sum_{i=k}^{N-1} \binom{i}{k} t^{i-k} - \frac{\beta}{6N} \sum_{i=N+1}^{4N} \binom{i}{k} t^{i-k} = 0.$$



# wait-&-judge scenario optimization

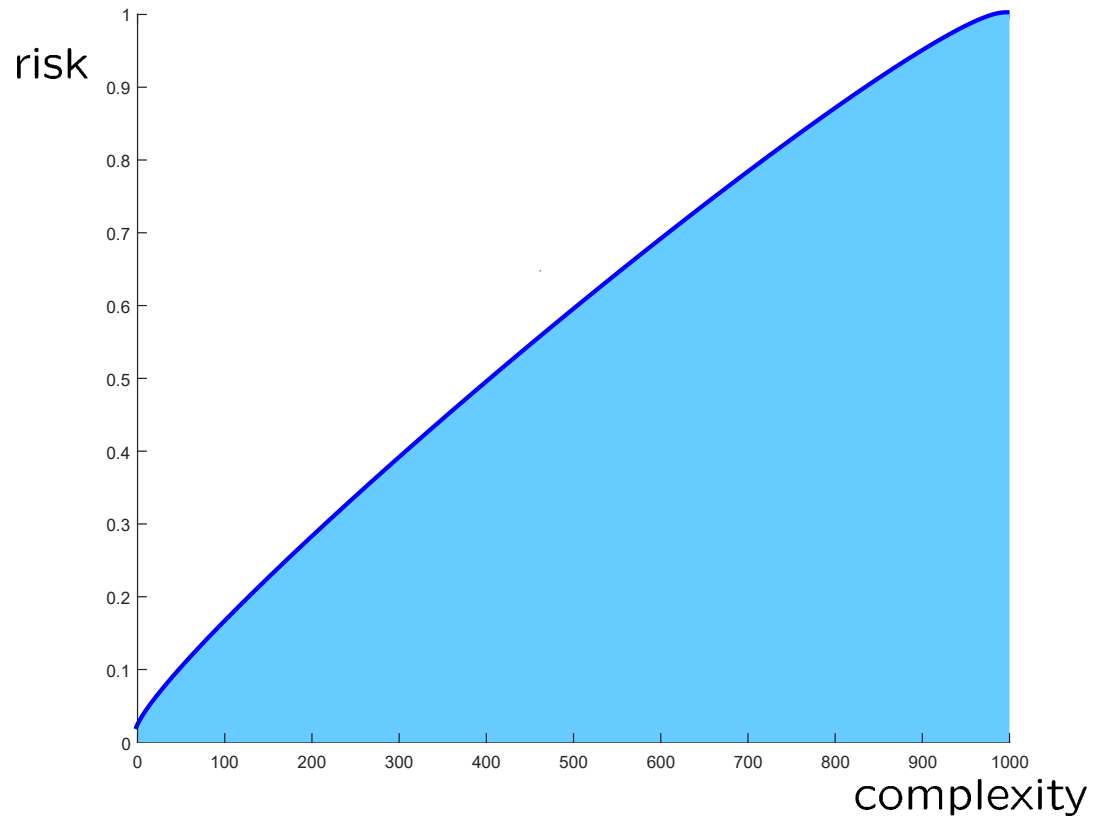


*how general is all this?*

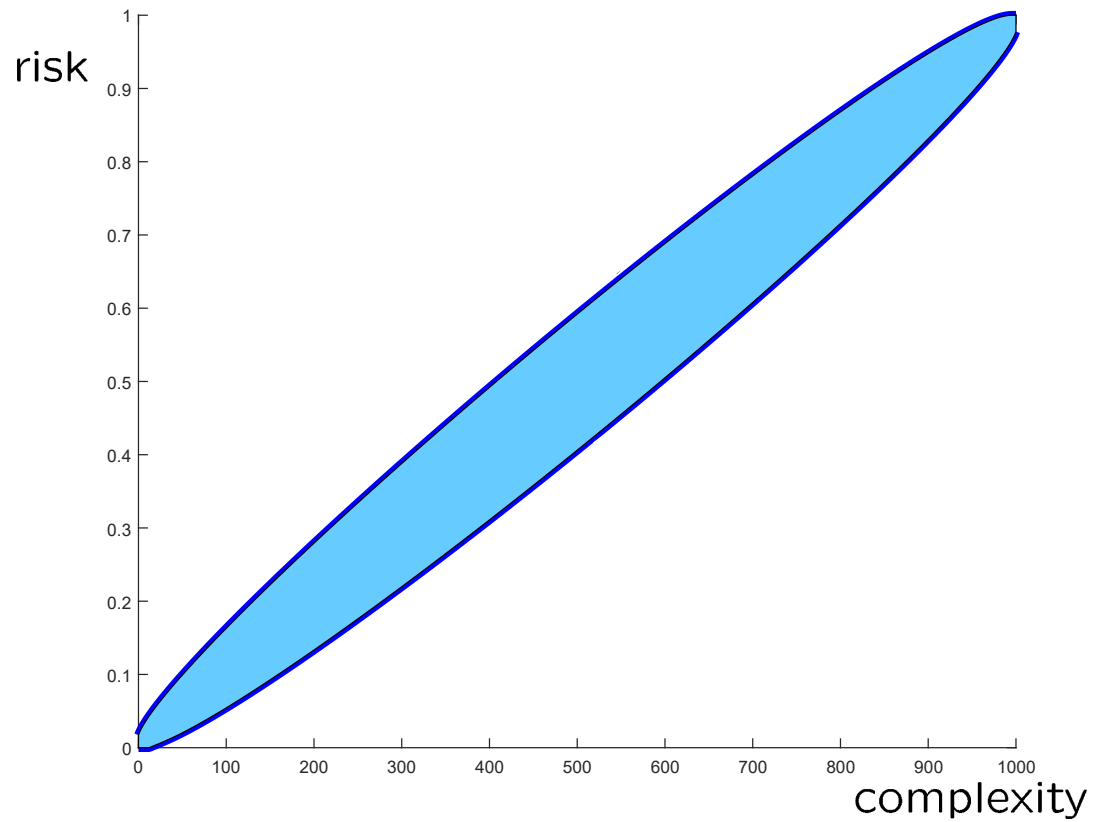
# consistent decision-making

- $\delta$  = observation
  - $\mathcal{D}$  = set of decisions
  - each  $\delta$  has associated a subset  $\mathcal{D}_\delta \subseteq \mathcal{D}$ , the set of decisions that are *appropriate* for  $\delta$
- (a) **permutation invariance:**  $P(\delta_1, \delta_2, \dots, \delta_n) = P(\delta_{i_1}, \delta_{i_2}, \dots, \delta_{i_n})$ ;
- (b) **stability in the case of confirmation:** if  $P(\delta_1, \delta_2, \dots, \delta_n)$  is appropriate for  $m$  new observations  $\delta_{n+1}, \dots, \delta_{n+m}$ , then  
 $P(\delta_1, \delta_2, \dots, \delta_n, \delta_{n+1}, \dots, \delta_{n+m}) = P(\delta_1, \delta_2, \dots, \delta_n)$ ;
- (c) **responsiveness to contradiction:** if there is at least one observation  $\delta_{n+i}$  for which  $P(\delta_1, \delta_2, \dots, \delta_n)$  is not appropriate, then  $P(\delta_1, \delta_2, \dots, \delta_n, \delta_{n+1}, \dots, \delta_{n+m}) \neq P(\delta_1, \delta_2, \dots, \delta_n)$ .

# wait-&-judge scenario optimization



# wait-&-judge scenario optimization

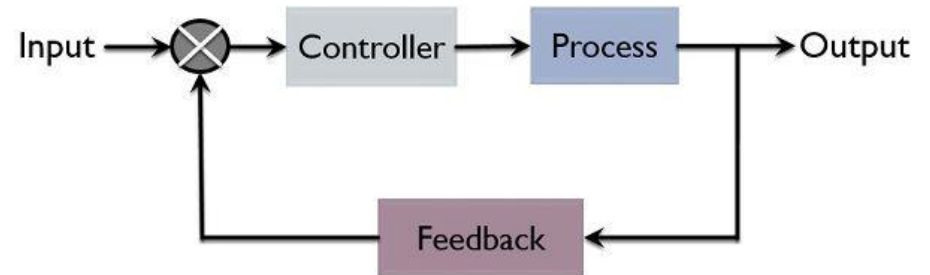


# robust optimization

$$\min_{x \in \mathcal{X}} c(x)$$

subject to:  $x \in \mathcal{X}_{\delta_i}, \quad i = 1, \dots, N$

*for example: Robust Control*



## relaxed schemes

$$\begin{array}{ll} \min_{x \in \mathcal{X}} & c(x) \\ \text{subject to:} & x \in \mathcal{X}_{\delta_i}, \quad i = 1, \dots, N \end{array}$$

## relaxed schemes

$$\min_{x \in \mathcal{X}} c(x)$$

subject to:  $f(x, \delta_i) \leq 0, \quad i = 1, \dots, N$



## relaxed schemes

$$\min_{x \in \mathcal{X}, \xi_i \geq 0} \quad c(x) + \rho \sum_{i=1}^N \xi_i$$

subject to:  $f(x, \delta_i) \leq \xi_i, \quad i = 1, \dots, N$

## relaxed schemes

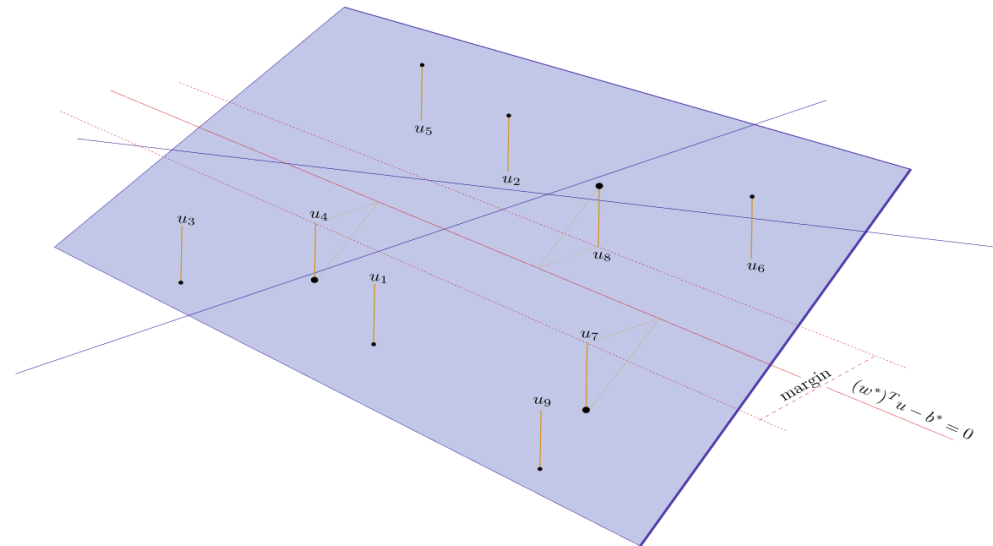
$$\min_{x \in \mathcal{X}, \xi_i \geq 0} c(x) + \rho \sum_{i=1}^N \xi_i$$

$$\text{subject to: } f(x, \delta_i) \leq \xi_i, \quad i = 1, \dots, N$$

*for example, SVM:*

$$\min_{w \in \mathbb{R}^d, b \in \mathbb{R}, \xi_i \geq 0} \|w\|^2 + \rho \sum_{i=1}^N \xi_i$$

$$\text{subject to: } 1 - y_i(\langle w, u_i \rangle - b) \leq \xi_i$$



more ...

- *constraint discarding*

- *CVaR = Conditional Value at Risk*

...

## beyond Popper

*- complexity is an impartial judge of reliability*

*and*

*- inductive procedures with continuous updates to observations are valid if complexity acts as a referee to assess model risk.<sup>(\*)</sup>*

*(\*) In “Conjectures and refutations: the growth of scientific knowledge”, Karl Popper explicitly condemns the practice of adapting theories to observations. Speaking of the Marxist theory of history, he writes that its followers “re-interpreted both the theory and the evidence in order to make them agree. [...] They thus gave a ‘conventionalist twist’ to the theory; and by this stratagem they destroyed its much advertised claim of scientific status”*



*a beacon called "Marco"*

***Happy birthday, Marco!***

Please, send me comments/remarks/observations: [marco.campi@unibs.it](mailto:marco.campi@unibs.it)

A **monograph** that links philosophical aspects to mathematical results can be downloaded from here:

<https://marco-campi.unibs.it/pdf-pszip/inductive%20methods.pdf>

The monograph is not yet published. **I would greatly appreciate receiving your comments, even on partial aspects of the work.** Thank you!