



# Fully Probabilistic Design of Multi-Agent System for Cloud Prediction

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## 1 Our context: Hierarchical FPD in Multiagent System

**General Task:** The optimal probabilistic design of multiagent system (MAS) for prediction of the unknown quantity of interest ( $x$ ) in large-scale system.

**Agents** = a set of sensors, a set of computer programmers, ...

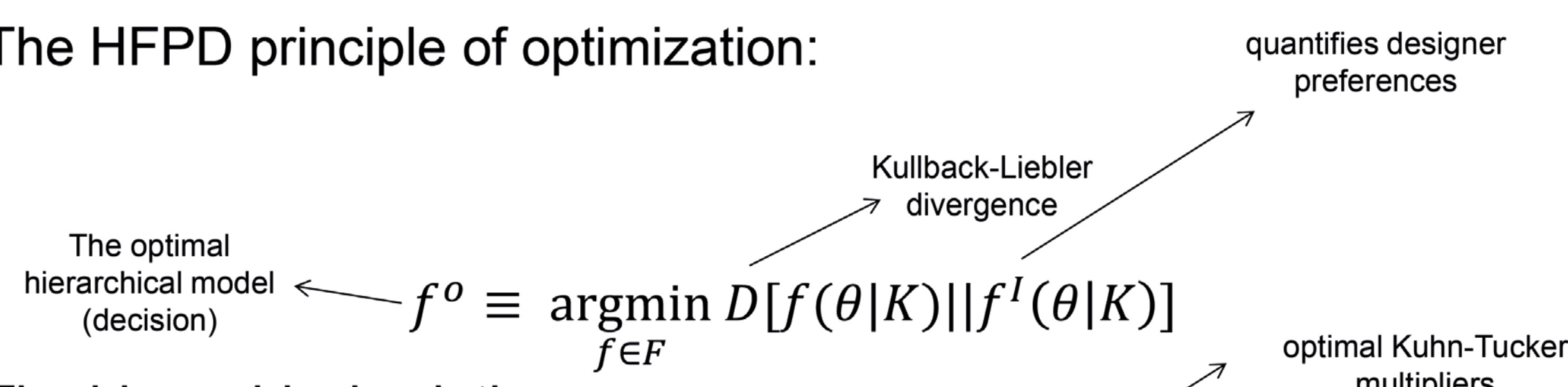
**Agent's task:** Each agent predicts  $x$  and reports known distribution

**Bayesian Moral Deliberator ( $I_0$ 's task):** Chooses the optimal distribution and proposes the uncertainty around their selection.

**Deliberator's knowledge :**

$$F = \{f(\theta|K); E_{f(\theta|K)} D(g_l||h(x|\theta)) \leq \beta_l, l = 1, \dots, s\}$$

The HFPD principle of optimization:



The hierarchical solution:

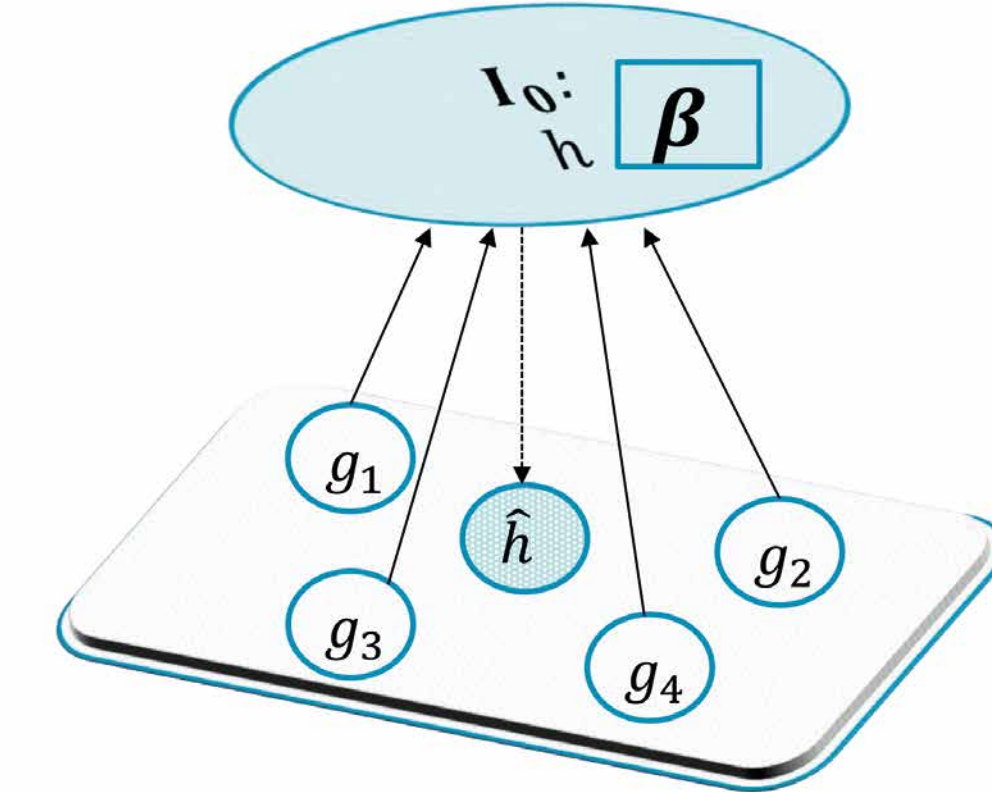
$$f^o(\theta|K) \propto f^l(\theta|K) \exp(-\sum_{l=1}^s \lambda_l D(g_l||h(x|\theta)))$$

The optimal predictor:

$$\hat{h}(x) = E_{f^o}[h(x|\theta)|K]$$

## 2 Some Competing Approaches

**Schematic paradigm:**



» **Logarithmic pool**

$$h_{lo}(x) \equiv \operatorname{argmin}_h \sum_{l=1}^s \alpha_l D[h(x)||g_l(x)] \propto \prod_{l=1}^s g_l^{\alpha_l}(x)$$

» **Linear pool**

$$h_{li}(x) = \operatorname{argmin}_h \sum_{l=1}^s \eta_l D[g_l(x)||h(x)] = \sum_{l=1}^s \eta_l g_l(x)$$

» **Dynamic Diffusion estimate**

$$h_i(x) \propto \prod_{j \in N_i} g_j^{v_{ij}}(x)$$

**Challenging points:**

In pervious approaches: How to choose  $\alpha_l$ ,  $\eta_l$  and  $v_{ij}$ ?

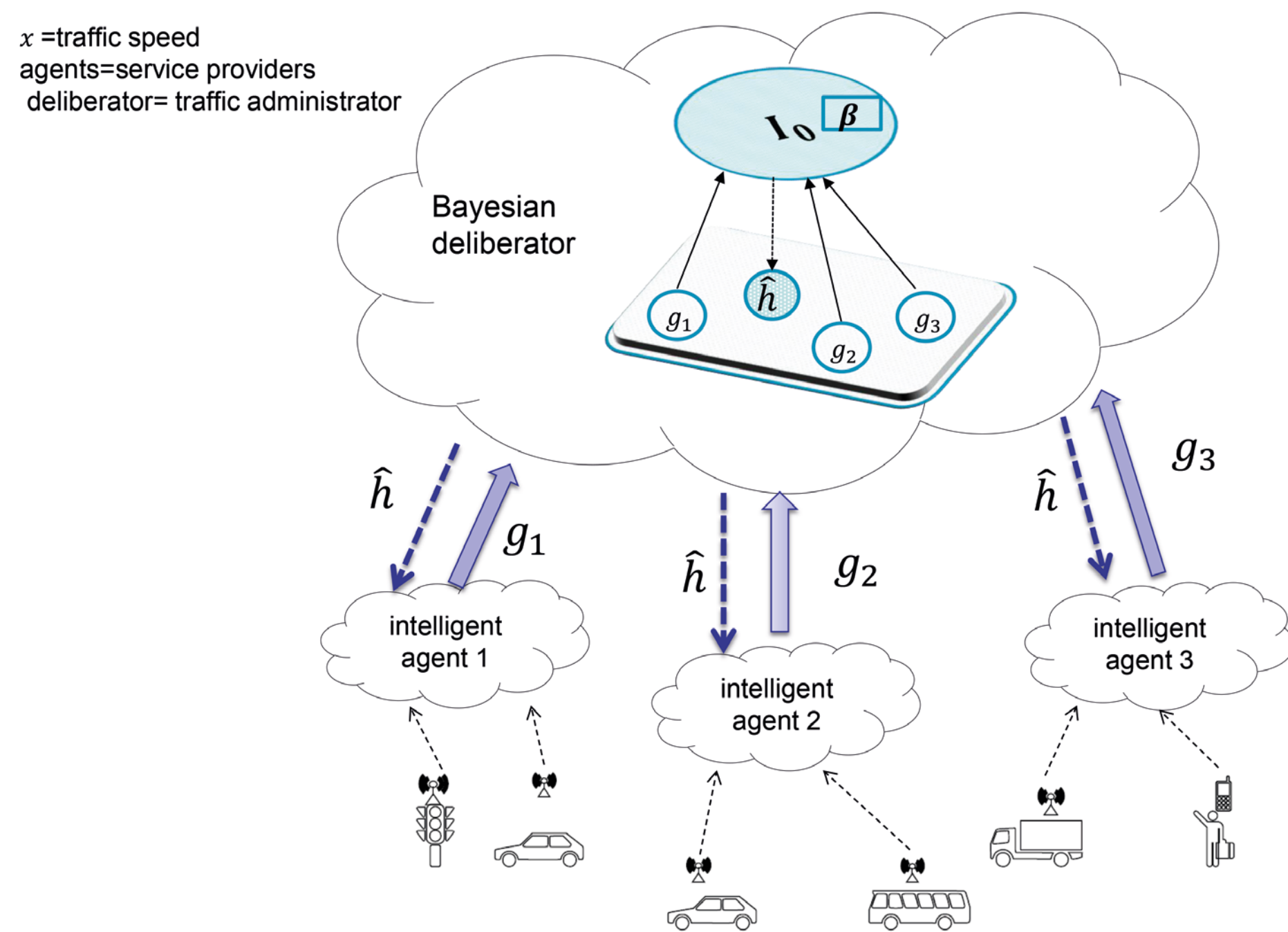
In our approach: How to design  $\lambda_l$ ,  $\beta_l$ ?

**Advantages of our approach**

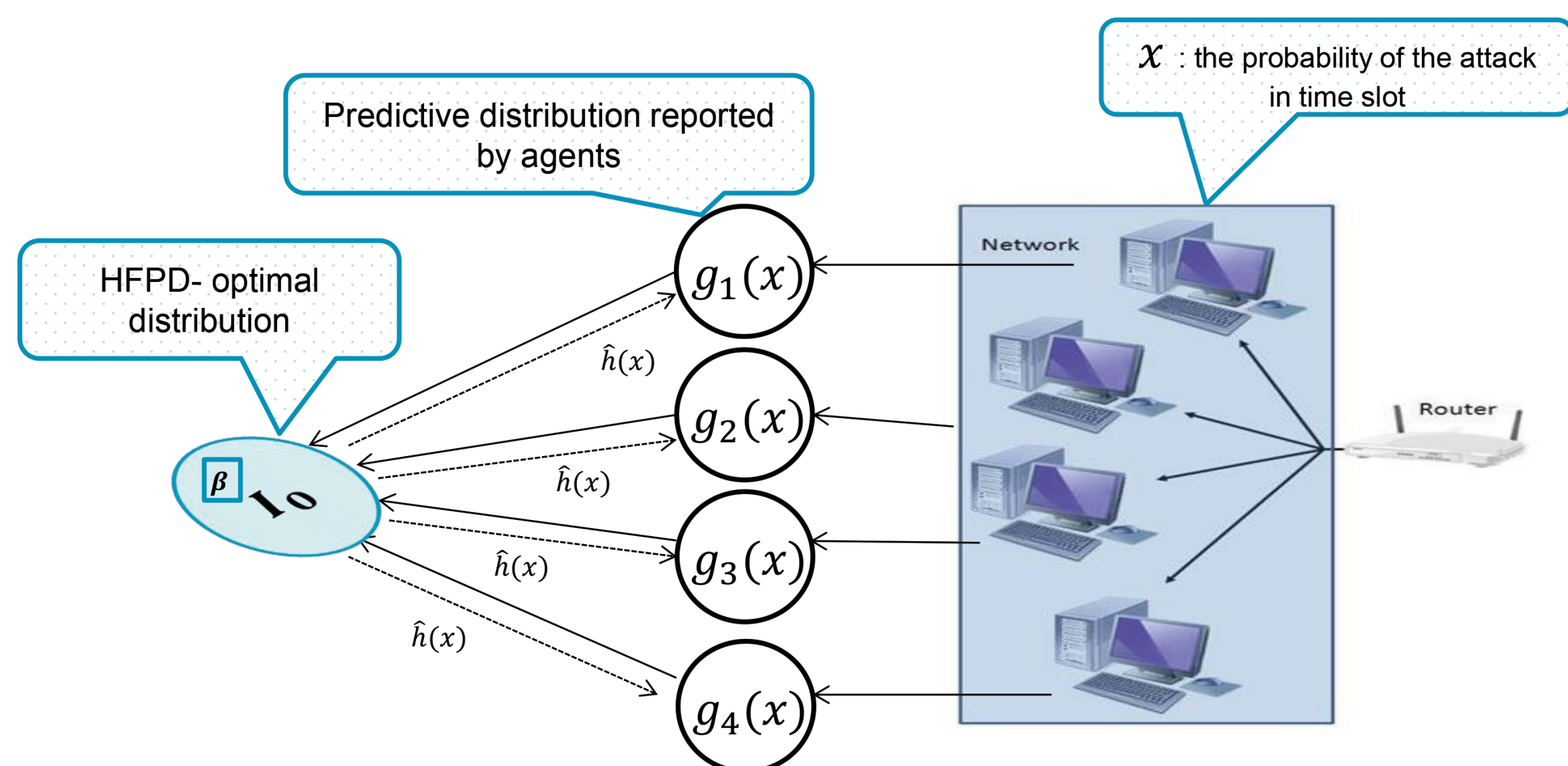
- » Many parametric forms of  $h$  can be explored under this approach
- » It is furnished with a measure of uncertainty around estimation of distribution,  $\hat{h}(x)$

## 3 Possible application contexts

» **Traffic management:** traffic prediction

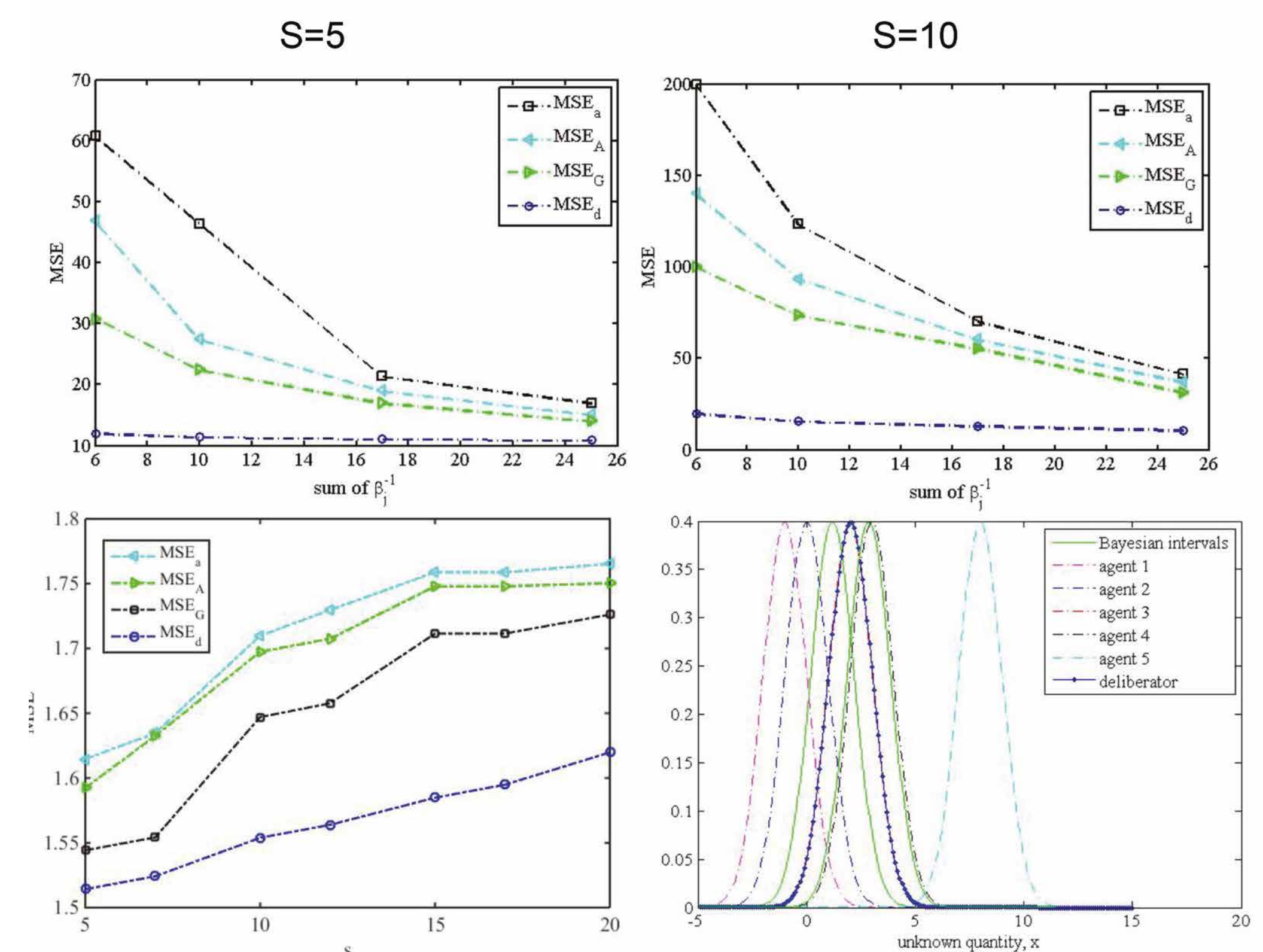


» **Monitoring network security:** prediction of potential security attack



## 4 Simulation Results

» MSE as a measure of performance:



## Conclusion and Future work

- » We propose a universal and consistent framework for how knowledge is expressed and shared in MAS regardless of the type of data.
- » We set up fully tractable context for limited definition of the nature of the deliberator, we aim to relax this definition to more general case.
- » How to express the deliberator's relationship with the agents.
- » Extension of Hierarchical FPD to more general cases where agents have partially common support
- » Adopting the proposed approach in the context of Bayesian filtering for tracking purposes