

# Problem Statement

## Bayesian filtering problem

Consider CO<sub>2</sub> monitoring as a Bayesian filtering problem:

$$\mathbf{x}_k \sim p(\mathbf{x}_k \mid \mathbf{x}_{k-1}) \qquad \mathbf{y}_k \sim p(\mathbf{y}_k \mid \mathbf{x}_k), \quad k = 0, 1, 2, \dots$$

where

- ▶  $\mathbf{x}_k \in \mathbb{R}^n$  is the *state* (CO<sub>2</sub> saturation/pressure) vector at time  $t = k\Delta t$
- ▶  $\mathbf{y}_k \in \mathbb{R}^m$  is the *observation* vector
- ▶  $p(\mathbf{x}_k \mid \mathbf{x}_{k-1}) = p(\mathbf{x}_k \mid \mathbf{x}_{1:k-1}, \mathbf{y}_{1:k-1})$  is the Markovian *transition* probability
- ▶  $p(\mathbf{y}_k \mid \mathbf{x}_k) = p(\mathbf{y}_k \mid \mathbf{x}_{1:k}, \mathbf{y}_{1:k-1})$  is the *likelihood* of the *measurement* model

# Sequential Bayesian Inference

Calculate *posterior*  $p(\mathbf{x}_k | \mathbf{y}_{1:k})$  for the *state*,  $\mathbf{x}_k$ , recursively, via the *predictive* distribution:

$$\begin{aligned} p(\mathbf{x}_k | \mathbf{y}_{1:k-1}) &= \int p(\mathbf{x}_k | \mathbf{x}_{k-1}) p(\mathbf{x}_{k-1} | \mathbf{y}_{1:k-1}) d\mathbf{x}_{k-1} \\ &= \mathbb{E}_{\mathbf{x}_{k-1} \sim p(\mathbf{x}_{k-1} | \mathbf{y}_{1:k-1})} \left[ p(\mathbf{x}_k | \mathbf{x}_{k-1}) \right] \end{aligned}$$

followed by the *correction* step involving Bayes formula:

$$p(\mathbf{x}_k | \mathbf{y}_{1:k}) = \frac{\overbrace{p(\mathbf{y}_k | \mathbf{x}_k)}^{\text{likelihood}} \overbrace{p(\mathbf{x}_k | \mathbf{y}_{1:k-1})}^{\text{"prior"}}}{\int p(\mathbf{y}_k | \mathbf{x}_k) p(\mathbf{x}_k | \mathbf{y}_{1:k-1}) d\mathbf{x}_k}$$

- **Marginalization over state  $\mathbf{x}_{k-1}$  in Chapman-Kolmogorov integral, and**
- **Integral for evidence are both computationally unfeasible!**