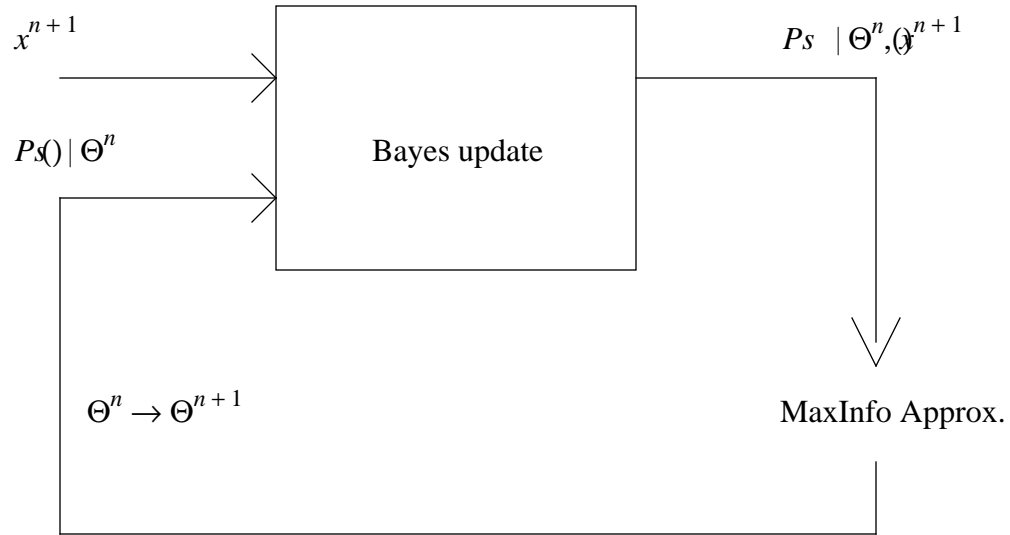


GKF Update Loop Equation



The elements going into $P_s(\mathcal{Y}) | \Theta^n$ are the prior, restricted to some knowledge H about the field, $P_s(\mathcal{Y} | H)$. (In the main text example, H is the set of known surface height field values.) and the Knowledge Representation (KR) distribution is $P(\theta) | \Theta^n$, which is the learned knowledge about the specifics of the surface at the n 'th iteration of the GKF.

These form the approximate posterior $P_s(\mathcal{Y}) | \Theta^n$ given by the integral over H of the product of the KR distribution and the prior distribution given H known, that is

$$P_s(\mathcal{Y}) | \Theta^n = \int P_s(\mathcal{Y} | H) P(\theta) | \Theta^n dH \quad (1)$$

At update $n + 1$, the new data and the approximate posterior from iteration n are incorporated using the likelihood $P_s(\mathcal{Y}^{n+1} | s)$ and Bayes' theorem to produce the data-dependent posterior written $P_s | \Theta^n, \mathcal{Y}^{n+1}$. Then, the new KR that captures an approximation to this exact posterior using (1) above with $n \rightarrow n + 1$ via Maximally informative statistical inference completes the GKF loop.

Figure 1 - Generalized Kalman Filter Update Loop