















$NP_{PF} = 3000$  particles. Figure 12 and Figure 13 report the KLD mean value over Scenario I and Scenario II, respectively. The measurement gaps (fixed for scenario I and variable for Scenario II) are considered for all filters, so that the approximation loss for CV-CT-PF and CV-CT-EKF can be observed. Apart from the initial settling phase of the filter, it is evident that the knowledge-based PF technique largely outperforms the EKF counterpart, yielding a much better capability in approximating the asymptotically true posterior PDF.

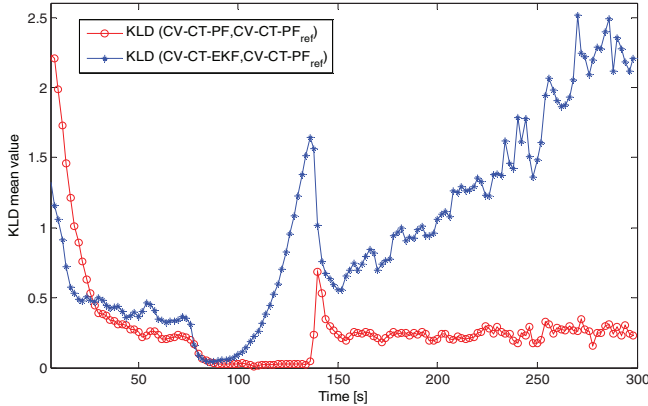


Figure 12. KLD mean value for Scenario I:  $KLD(CV-CT-PF, CV-CT-PF_{ref})$  and  $KLD(CV-CT-EKF, CV-CT-PF_{ref})$ ;  $\Delta_{GAP} = 60s$   $T = 2s$  and  $T_{OBS} = 300s$ .

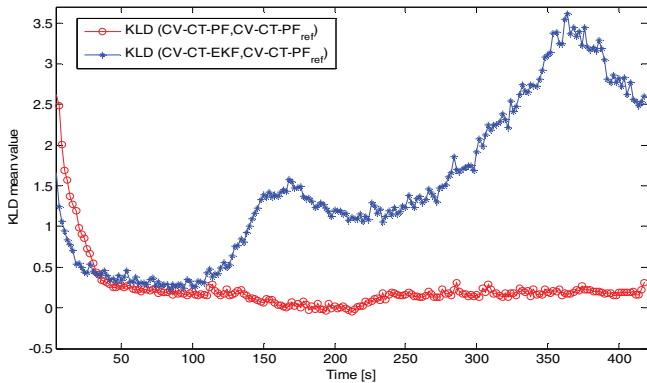


Figure 13. KLD mean value for Scenario II:  $KLD(CV-CT-PF, CV-CT-PF_{ref})$  and  $KLD(CV-CT-EKF, CV-CT-PF_{ref})$ ;  $\Delta_{GAP} = 72s$ ,  $T = 2s$  and  $T_{OBS} = 420s$ .

## VII. CONCLUSIONS

Two innovative filtering schemes have been proposed in this paper. They implement non-linear Bayesian techniques for target state estimation, and they both exploit a priori vessel route information. The former is derived from the Extended Kalman Filter, whereas the latter is a Particle Filter scheme. The non-sensor information drives the selection of the target dynamical model used in both filters. The performance analysis over simulated scenarios yields a significant improvement coming from the exploitation of the a priori information. In addition, the PF scheme demonstrates increased robustness with respect to the EKF as highly non-linear tracking conditions are experienced.

## ACKNOWLEDGMENT

The research leading to these results received funding from the EU's 7th Framework Programme (grant agreement no 238710). The research has been carried out in the MC IMPULSE project: <https://mcimpulse.isy.liu.se>.

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