

ror bound between $\tilde{V}_t^G(b)$ and $\tilde{V}_t^*(b)$, as our current results (Theorem 5) show only that it is bounded. We also intend to consider cases where its possible to sequentially process information from sensors and thus integrate our approach with adaptive submodularity.

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