D2A-BSP: Distilled Data Association Belief Space Planning with Performance Guarantees Under Budget Constraints

Supplementary Material

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This document provides supplementary material to the paper [1]. Therefore, it should not be considered a self-contained document, but instead regarded as an appendix of [1]. Throughout this report, all notations and definitions are with compliance to the ones presented in [1].

1 Incrementally adapting $\mathcal{LB}\left[\eta\right], \mathcal{UB}\left[\eta\right]$

We denote the bounds presented in Theorem 4 in [1] as $\mathcal{LB}[\eta|b_k^s]$, $\mathcal{UB}[\eta|b_k^s]$, i.e. with respect to a simplified belief b_k^s with M_k^s components. Given a belief component $r_k \notin M_k^s$ with associated weight w_k^r , we denote $M_k^{s+1} \triangleq M_k^s \cup r_k$. By definition (see eq. (11) in [1]) the simplified belief at time k for M_k^{s+1} components is given by

$$b_k^{s+1} \triangleq \sum_{j=1}^{M_k^{s+1}} w_k^{s+1,j} b_k^j \quad , \quad w_k^{s+1,j} \triangleq \frac{w_k^j}{w_k^{m,s+1}}, \tag{1}$$

where w_k^j corresponds to the original belief component weight (see eq. (3) in [1]) and $w_k^{m,s+1} = w_k^{m,s} + w_k^r$. As such, $\mathcal{LB}\left[\eta|b_k^{s+1}\right]$, $\mathcal{UB}\left[\eta|b_k^{s+1}\right]$ represent the bounds for the measurement likelihood η given a simplified belief b_k^{s+1} with M_k^{s+1} components. Using eq. (28) in [1] and (1) we define

$$\eta^{s+1} \triangleq \sum_{i}^{|L|} \sum_{j}^{M_k^{s+1}} \tilde{\zeta}_{k+1}^{i,j} w_k^{s+1,j}. \tag{2}$$

We now present how to incrementally adapt the lower and upper bounds. We begin by writing the lower bound with respect to the simplified belief b_k^{s+1} using (2) and get the recursive update rule

$$\mathcal{LB}\left[\eta|b_{k}^{s+1}\right] = \eta^{s+1}w_{k}^{m,s+1} = \sum_{i}^{|L|}\sum_{j}^{M_{k}^{s+1}}\tilde{\zeta}_{k+1}^{i,j}w_{k}^{j} = \sum_{i}^{|L|}\sum_{j}^{M_{k}^{s}}\tilde{\zeta}_{k+1}^{i,j}w_{k}^{j} + \sum_{i}^{|L|}\tilde{\zeta}_{k+1}^{i,r}w_{k}^{r} = \eta^{s}w_{k}^{m,s} + \sum_{i}^{|L|}\tilde{\zeta}_{k+1}^{i,r}w_{k}^{r} = \mathcal{LB}\left[\eta|b_{k}^{s}\right] + \sum_{i}^{|L|}\tilde{\zeta}_{k+1}^{i,r}w_{k}^{r}.$$
(3)

Using similar derivations the recursive update rule for the upper bound is given by

$$\mathcal{UB}\left[\eta|b_{k}^{s+1}\right] = \eta^{s+1}w_{k}^{m,s+1} + (1 - w_{k}^{m,s+1})\sigma\sum_{i}^{|L|}\alpha^{i} = \eta^{s}w_{k}^{m,s} + \sum_{i}^{|L|}\tilde{\zeta}_{k+1}^{i,r}w_{k}^{r} + (1 - w_{k}^{m,s} - w_{k}^{r})\sigma\sum_{i}^{|L|}\alpha^{i} = \eta^{s}w_{k}^{m,s} + (1 - w_{k}^{m,s})\sigma\sum_{i}^{|L|}\alpha^{i} + \sum_{i}^{|L|}\tilde{\zeta}_{k+1}^{i,r}w_{k}^{r} - w_{k}^{r}\sigma\sum_{i}^{|L|}\alpha^{i} = \mathcal{UB}\left[\eta|b_{k}^{s}\right] + w_{k}^{r}\sum_{i}^{|L|}\left[\tilde{\zeta}_{k+1}^{i,r} - \sigma\alpha^{i}\right].$$

$$(4)$$

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2 Incrementally adapting $\mathcal{LB}[\mathcal{H}], \mathcal{UB}[\mathcal{H}]$

We follow similar derivations as in Section 1 and denote the bounds presented in Theorem 2 in [1] as $\mathcal{LB}\left[\mathcal{H}|b_k^s\right]$, $\mathcal{UB}\left[\mathcal{H}|b_k^s\right]$, i.e. with respect to a simplified belief b_k^s with M_k^s components. Given a belief component $r_k \notin M_k^s$ with associated weight w_k^r , we denote $M_k^{s+1} \triangleq M_k^s \cup r_k$. Using (1) we also denote the bounds over the cost term, given a simplified belief b_k^{s+1} with M_k^{s+1} components, as $\mathcal{LB}\left[\mathcal{H}|b_k^{s+1}\right]$, $\mathcal{UB}\left[\mathcal{H}|b_k^{s+1}\right]$. Deriving a direct recursive update rule for these bounds is not trivial. Instead, we show how each term in $\mathcal{LB}\left[\mathcal{H}|b_k^{s+1}\right]$, $\mathcal{UB}\left[\mathcal{H}|b_k^{s+1}\right]$ can be incrementally updated individually. Using (2) we begin with a recursive update rule for η^{s+1} given by

$$\eta^{s+1} = \sum_{i}^{|L|} \sum_{j}^{M_k^{s+1}} \tilde{\zeta}_{k+1}^{i,j} w_k^{s+1,j} = \frac{1}{w_k^{m,s+1}} \left[\eta^s w_k^{m,s} + \sum_{i}^{|L|} \tilde{\zeta}_{k+1}^{i,r} w_k^r \right]. \tag{5}$$

Using equations (24) and (28) in [1] we write the recursive update rule for \mathcal{H}^{s+1} , i.e. the cost given a simplified belief b_k^{s+1}

$$\mathcal{H}^{s+1} = -\frac{1}{\eta^{s+1}} \sum_{i}^{|L|} \sum_{j}^{M_{k}^{s-1}} \left[\tilde{\zeta}_{k+1}^{i,j} w_{k}^{s+1,j} log \left(\tilde{\zeta}_{k+1}^{i,j} w_{k}^{s+1,j} \right) \right] + log \left(\eta^{s+1} \right) =$$

$$-\frac{1}{\eta^{s+1}} \left[\sum_{i}^{|L|} \sum_{j}^{M_{k}^{s}} \left[\frac{\tilde{\zeta}_{k+1}^{i,j} w_{k}^{j}}{w_{k}^{m,s+1}} log \left(\frac{\tilde{\zeta}_{k+1}^{i,j} w_{k}^{j}}{w_{k}^{j}} log \left(\frac{\tilde{\zeta}_{k+1}^{i,j} w_{k}^{j}}{w_{k}^{j}} log \left(\frac{\tilde{\zeta}_{k+1}^{i,$$

Using Theorem 2 in [1] we explicitly write the lower bound with respect to the simplified belief b_k^{s+1}

$$\mathcal{LB}\left[\mathcal{H}|b_{k}^{s+1}\right] = \frac{\eta^{s+1}w_{k}^{m,s+1}}{\mathcal{UB}\left[\eta|b_{k}^{s+1}\right]}\left[\mathcal{H}^{s+1} - log(\eta^{s+1})\right] - \frac{w_{k}^{m,s+1}}{\mathcal{UB}\left[\eta|b_{k}^{s+1}\right]} \sum_{i}^{|L|} \sum_{j}^{M_{k}^{s+1}} \tilde{\zeta}_{k+1}^{i,j}w_{k}^{s+1,j}log\left(\frac{w_{k}^{m,s+1}}{\mathcal{LB}\left[\eta|b_{k}^{s+1}\right]}\right) = \frac{\eta^{s+1}w_{k}^{m,s+1}}{\mathcal{UB}\left[\eta|b_{k}^{s+1}\right]} \left[\mathcal{H}^{s+1} - log(\eta^{s+1})\right] - \frac{w_{k}^{m,s+1}\eta^{s+1}}{\mathcal{UB}\left[\eta|b_{k}^{s+1}\right]}log\left(\frac{w_{k}^{m,s+1}}{\mathcal{LB}\left[\eta|b_{k}^{s+1}\right]}\right), \tag{7}$$

and observe that each term can be incrementally updated individually using (5), (6) and Section 1. Similarly, using Theorem 2 in [1], we explicitly write the upper bound with respect to the simplified belief b_k^{s+1}

$$\mathcal{UB}\left[\mathcal{H}|b_{k}^{s+1}\right] = \frac{\eta^{s+1}w_{k}^{m,s+1}}{\mathcal{LB}\left[\eta|b_{k}^{s+1}\right]}\left[\mathcal{H}^{s+1} - log(\eta^{s+1})\right] - \frac{w_{k}^{m,s+1}}{\mathcal{LB}\left[\eta|b_{k}^{s+1}\right]}\sum_{i}^{|L|}\sum_{j}^{M_{k}^{s+1}}\tilde{\zeta}_{k+1}^{i,j}w_{k}^{s+1,j}log\left(\frac{w_{k}^{m,s+1}}{\mathcal{UB}\left[\eta|b_{k}^{s+1}\right]}\right) - \gamma log\left(\frac{\gamma}{|L|\left|\neg M_{k}^{s+1}\right|}\right) = \frac{\eta^{s+1}w_{k}^{m,s+1}}{\mathcal{LB}\left[\eta|b_{k}^{s+1}\right]}\left[\mathcal{H}^{s+1} - log(\eta^{s+1})\right] - \frac{w_{k}^{m,s+1}\eta^{s+1}}{\mathcal{LB}\left[\eta|b_{k}^{s+1}\right]}log\left(\frac{w_{k}^{m,s+1}}{\mathcal{UB}\left[\eta|b_{k}^{s+1}\right]}\right) - \gamma log\left(\frac{\gamma}{|L|\left|\neg M_{k}^{s+1}\right|}\right), \tag{8}$$

where $\gamma \triangleq 1 - \frac{\eta^{s+1} w_k^{m,s}}{\mathcal{LB}[\eta|b_k^{s+1}]}$. We observe that each term can be incrementally updated individually using (5), (6) and Section 1.

References

[1] M. Shienman and V. Indelman. D2a-bsp: Distilled data association belief space planning with performance guarantees under budget constraints. In *IEEE Intl. Conf. on Robotics and Automation (ICRA)*, 2022. Submitted.