

Introduction

- We denote the full unapproximated solution of BSP problem as X-BSP

$$J(u) \doteq \mathbb{E}_z \left[\sum_{i=k+1}^{k+L} c_i (b[X_{i|k}], u_{i-1|k}) \right]$$

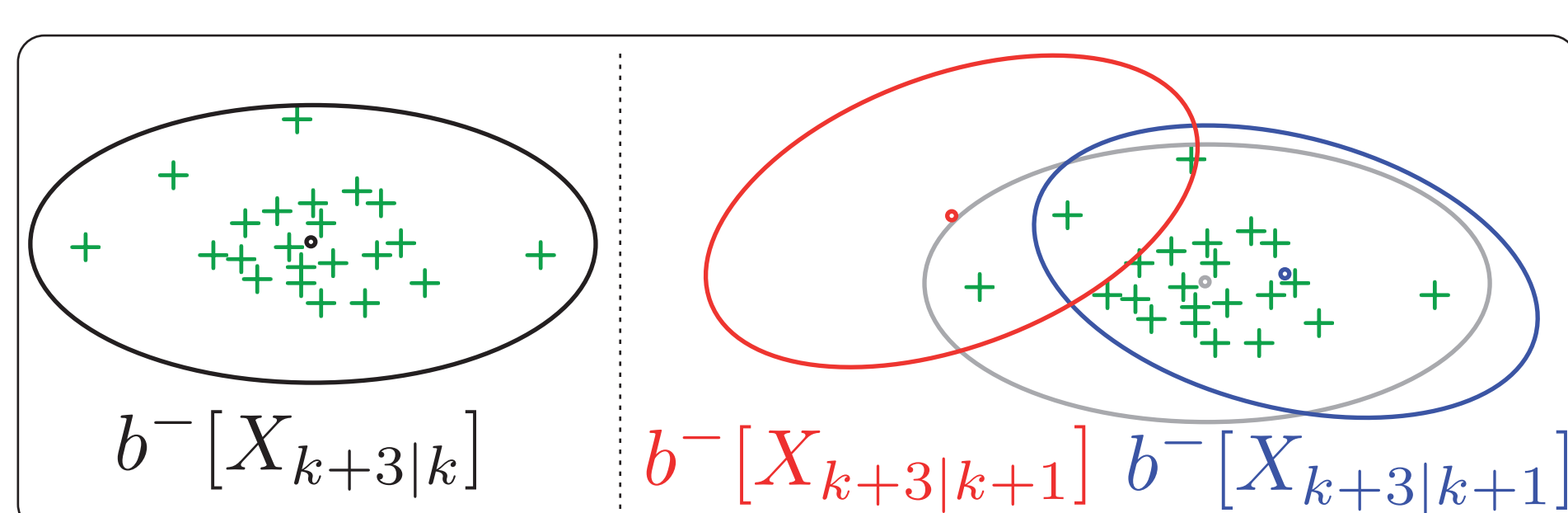
- The ML approximation for X-BSP is denoted by ML-BSP
- Calculation re-use in BSP has only been done over ML-BSP with restricting assumptions

Problem Statement

- Planning at time k - executed
- Optimal action chosen - $u_{k:k+L-1|k}^*$
- $u_{k:k+L-1|k}^* \in u_{k:k+L-1|k}^*$ executed
- Acquired measurements $z_{k+1:k+L|k+L}$
- Next planning is with respect to future measurements $z_{k+l+1:k+l+L|k+l}$
- Existing approaches perform these costly calculations from scratch
- Our goal** - develop approach for re-using calculations from precursory planning sessions
- Assumptions:
 - Precursory planning calculations are accessible
 - Horizon overlap, i.e. $l \in [1, L]$
 - $u_{k+l:k+L-1|k}^*$ partially resides in the set of candidate actions for planning time $k+l$

Key Observation

- Two successive X-BSP sessions from times k and $k+l$ are similar
- Calculations in planning time k can be re-used for planning time $k+l$
- By re-using samples we can avert from costly calculations at time $k+l$



(a) Belief distance affecting sample re-use

Approach

- Consider the objective calculations from planning time $k \forall u \in \mathcal{U}_k$

$$J(u) \approx \frac{1}{n} \sum_{\{z_{k+1|k}\}_1^n} [w_{k+1|k}^i \cdot c_{k+1|k} + \dots]$$

- and the desired objective for planning time $k+l \forall u' \in \mathcal{U}_{k+l}$

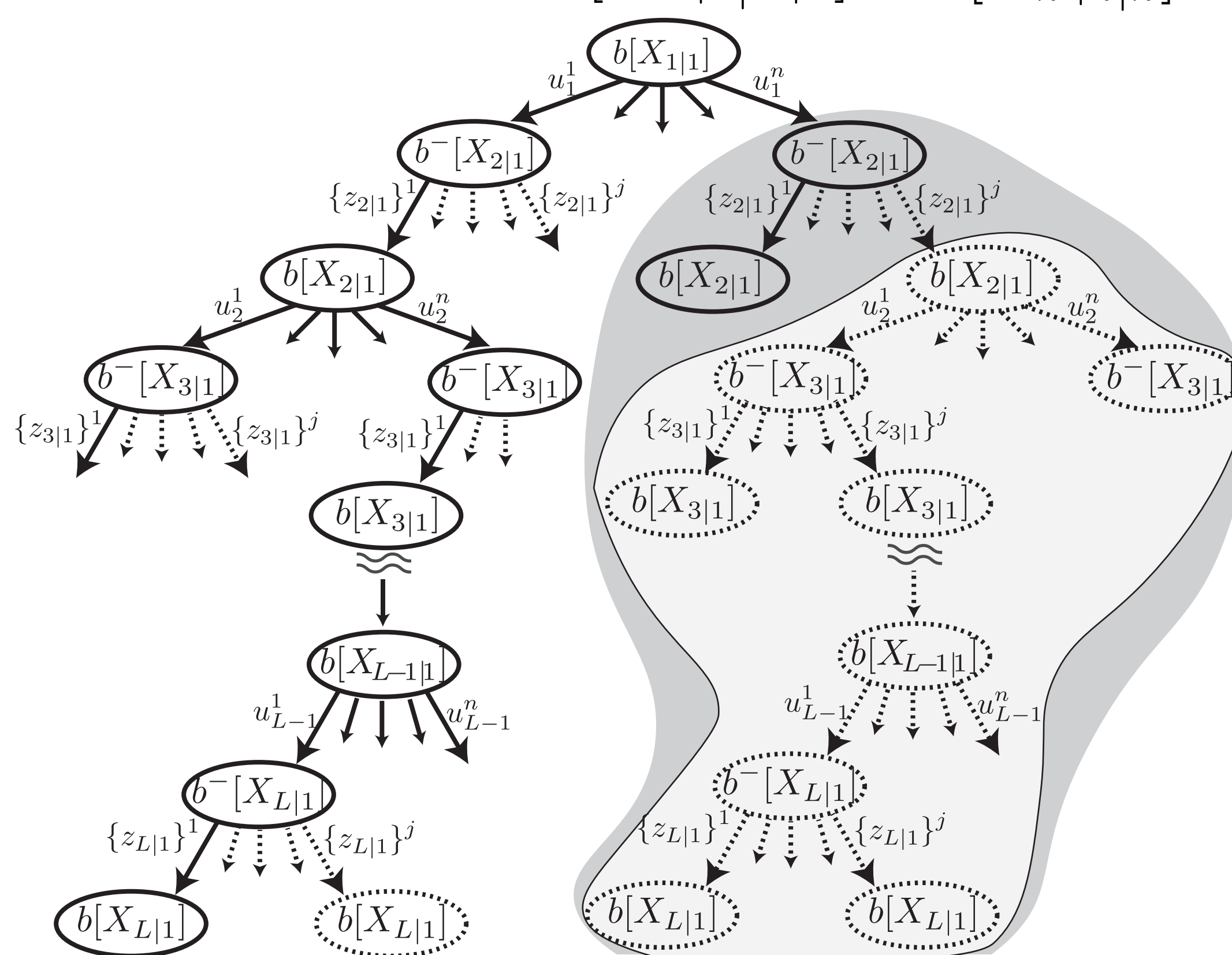
$$J(u') \approx \frac{1}{n} \sum_{\{z_{k+l+1|k+l}\}_1^n} [w_{k+l+1|k+l}^j \cdot c_{k+l+1|k+l} + \dots]$$

- Measurements from horizons $k+l+1$ to $k+L$ are sampled from similar distributions (3) $\mathbb{P}(z_{k+l+1|k} | H_{k+l+1|k}^-)$

$$(4) \mathbb{P}(z_{k+l+1|k+l} | H_{k+l+1|k+l}^-)$$

- Assuming (1) has been sampled from original distribution, e.g. (3), we get $w_{i|k}^j \doteq 1 \forall i, j$

- Denote belief $b[X_{k+l|k}]$ closest to the posterior belief $b[X_{k+l|k+l}]$ as $\tilde{b}[X_{k+l|k}]$



(b) Selecting beliefs for re-use

- All beliefs from planning time k rooted in $\tilde{b}[X_{k+l|k}]$ are considered for re-use in planning time $k+l$
- We assume all samples can be re-used, will be relaxed in future work
- Incrementally update all candidate beliefs with actual information received up-to time $k+l$ (Farhi17icra)
- Since samples are re-used rather than freshly sampled, the j^{th} weight at the i^{th} horizon step is given by

$$w_{i|k}^j = \frac{\mathbb{P}(z_{i|k}^j | H_{i|k+l}^-)}{\mathbb{P}(z_{i|k}^j | H_{i|k}^-)} \quad \forall i, j$$

- Since (3) nor (4) can be directly calculated we use

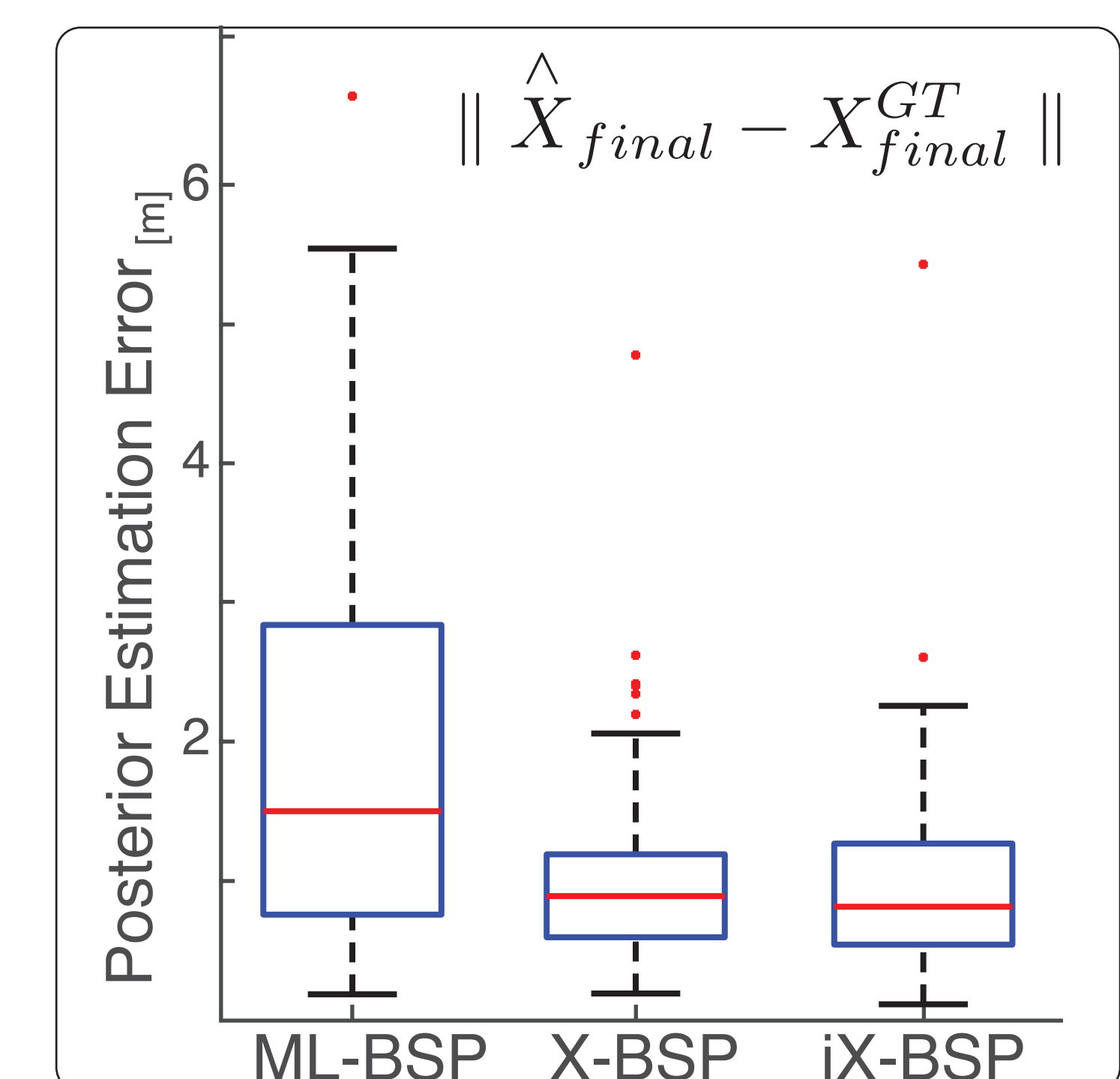
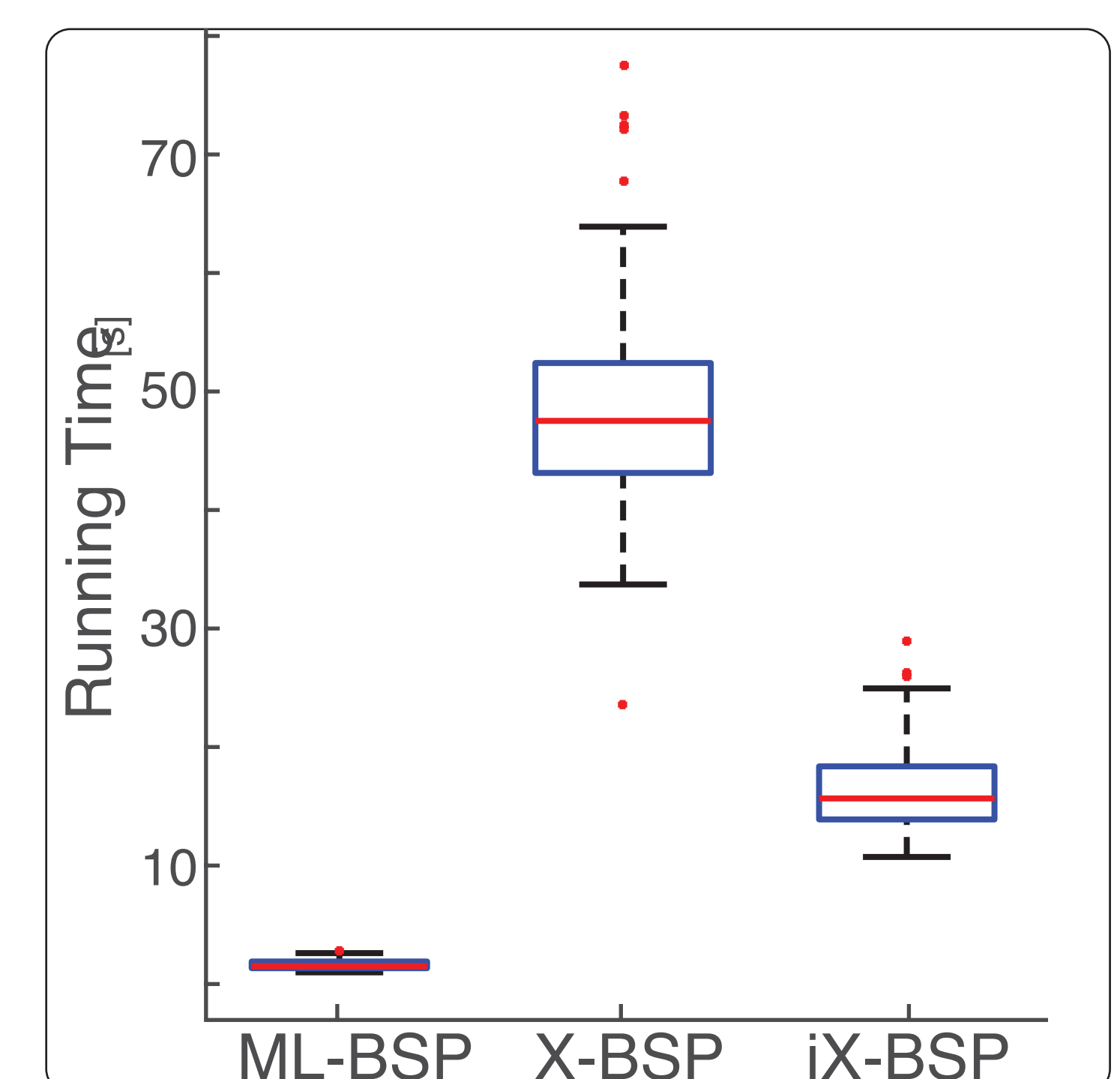
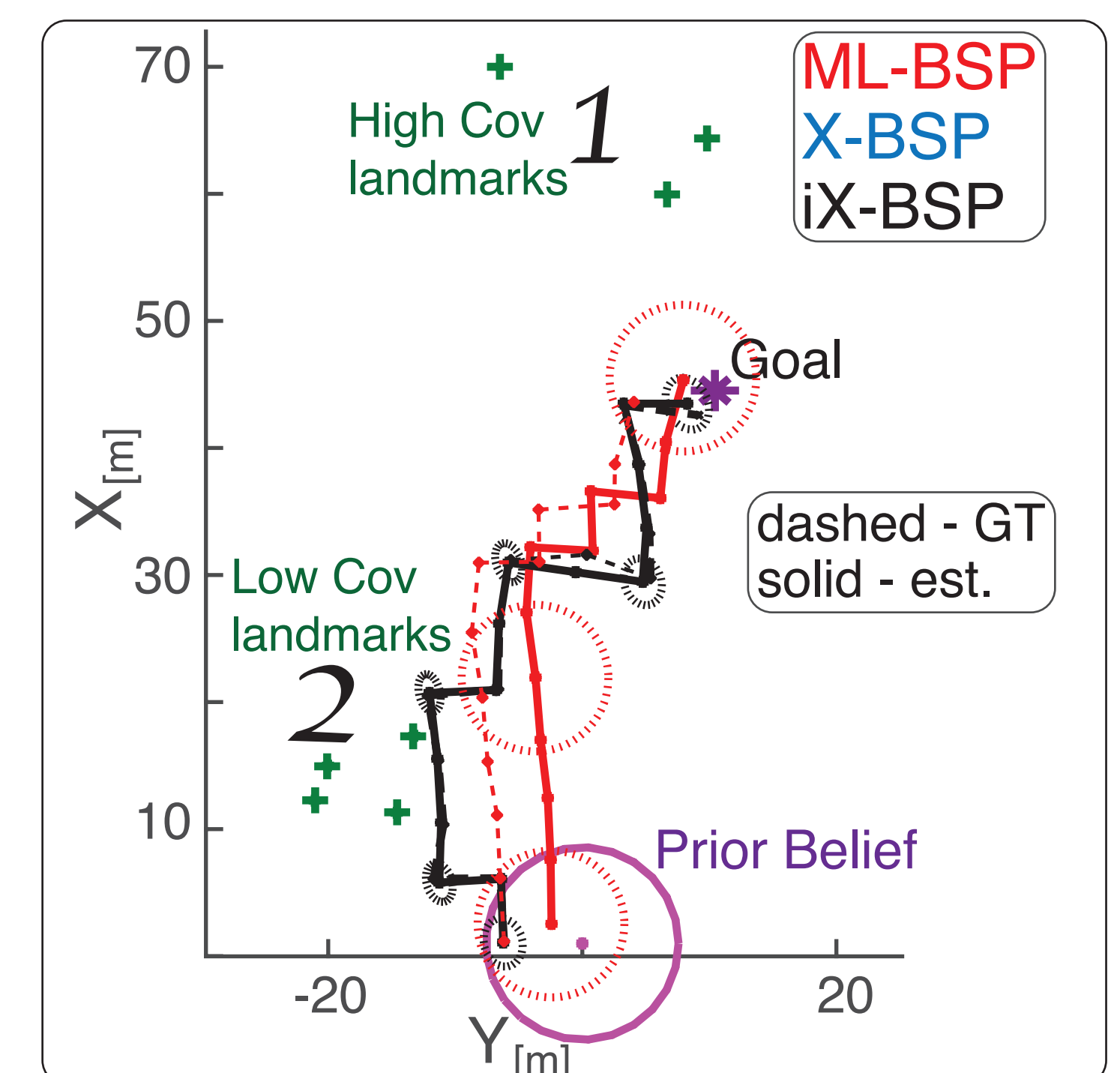
$$\mathbb{P}(z_{i|k}^j | H_{i|k}^-) \approx \frac{1}{n} \sum_{X_i} \mathbb{P}(z_{i|k}^j | X_i) \mathbb{P}(X_i | H_{i|k}^-)$$

- We complete the rest of the horizon ($k+L+1:k+L+l$) with X-BSP

- With solved beliefs and updated weights we can now solve (4) $\forall u' \in \mathcal{U}_{k+l}$

Results

- We performed 100 rollouts, each with a different sampled ground-truth for prior state. Map illustrates a single rollout.



Conclusions

- BSP using expectation can be efficiently updated using a precursory planning session, presenting iX-BSP
- iX-BSP provides the same statistical accuracy as X-BSP for a reduced computational effort
- Since iX-BSP alters the solution approach of the original, un-approximated, problem (X-BSP), we believe it can also be utilized for existing approximations of X-BSP

References

- E. I. Farhi and V. Indelman, "Towards efficient inference update through planning via jip - joint inference and belief space planning," in IEEE Intl. Conf. on Robotics and Automation (ICRA), 2017