

Comments with "Tight Collision Probability for UAV Motion Planning in Uncertain Environment"

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Abstract—This is the document accompanying the paper "Tight Collision Probability for UAV Motion Planning in Uncertain Environment" to clarify two issues related to [1] and [2].

I. ISSUE I

Consider two ellipsoids $\mathcal{E}_1 = \mathcal{E}^n(B, \mathbf{b})$ and $\mathcal{E}_2 = \mathcal{E}^n(C, \mathbf{c})$, the collision status of them is identified by:

$$\mathbf{y}^T A \mathbf{y} \leq 1/\lambda_0^2(M'), \quad (1)$$

where $\mathbf{y} = \mathbf{b} - \mathbf{c}$, $\lambda_0(\cdot)$ is to calculate the minimal eigenvalue, and $A, M' \in \mathbb{R}^{2n \times 2n}$ are functions of $B, C, \mathbf{b}, \mathbf{c}$. The equation (1) can be written explicitly as

$$\mathbf{y}^T A(B, C, \mathbf{b}, \mathbf{c}) \mathbf{y} \leq 1/\lambda_0^2(M'(B, C, \mathbf{b}, \mathbf{c})), \quad (2)$$

Thomas et al. [1] claimed to find the exact solution of the collision probability of \mathcal{E}_1 and \mathcal{E}_2 by reformulate (1) as

$$\begin{aligned} P(\mathbf{y}^T A \mathbf{y} \leq 1/\lambda_0^2(M')) &= P(v \leq 1/\lambda_0^2(M')) \\ &= F_v(1/\lambda_0^2(M')) \end{aligned} \quad (3)$$

By regarding \mathbf{b} and \mathbf{c} as Gaussian distributed random variables, the collision probability is transformed into the CDF of the quadratic form in the random variable \mathbf{y} . However, this transformation is only valid if A and $1/\lambda_0^2(M')$ are deterministic variables. Unfortunately, [1] seems to overlook the fact that A and $1/\lambda_0^2(M')$ are functions of \mathbf{b} and \mathbf{c} , and therefore, are random variables. Instead, they use the mean values of \mathbf{b} and \mathbf{c} to calculate A , $1/\lambda_0^2(M')$, and (3).

II. ISSUE II

Our iterative trajectory optimization algorithm is benchmarked with [2]–[5]. For the Scene I in our paper, [2] gives the results presented in Figure 1, using their open source code (https://rebrand.ly/castillo_RAL2020benchmark). However, they carelessly took the variance as the standard deviation in the implementation of [2], [3], [5]. The rectified results is shown in Figure 2. The issue has been confirmed with the author of [2].

REFERENCES

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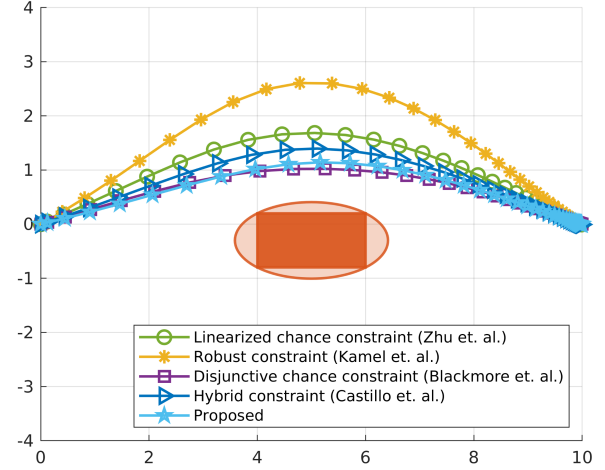


Fig. 1. Benchmark result of [2]

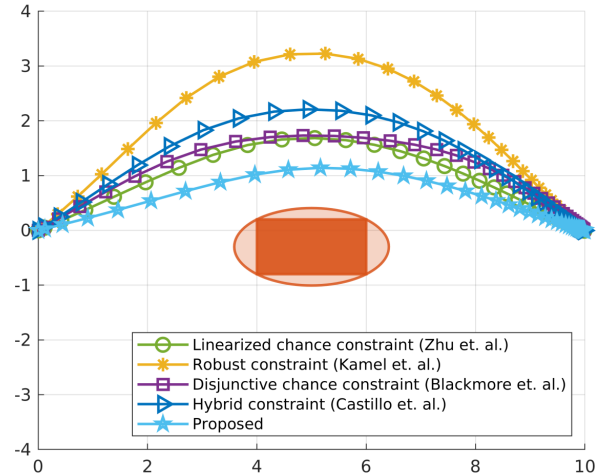


Fig. 2. Benchmark result of our implementation

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