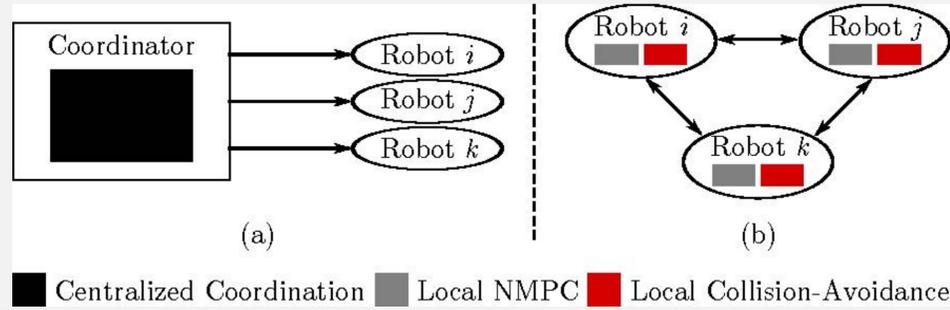


## 1. Introduction

Goal: safe motion planning in tight environments in a distributed way in real time.



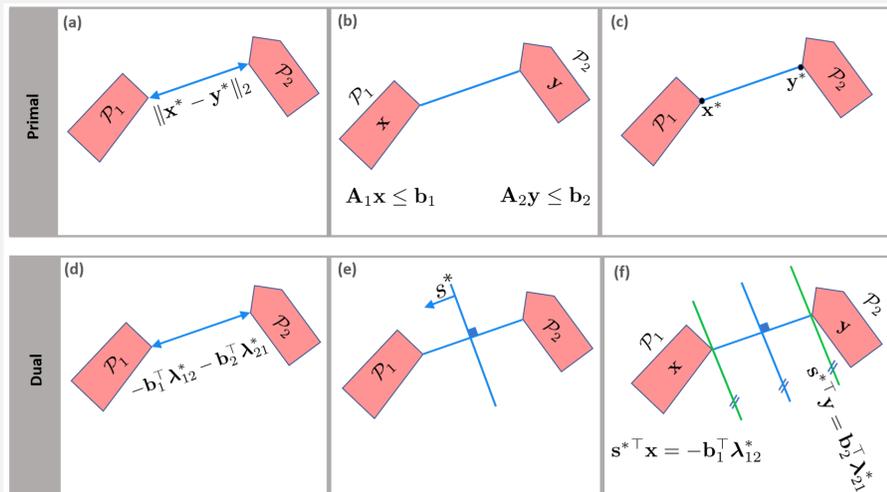
## 2. Collision Avoidance: Distance between Two Polytopes

### Primal Formulation

$$\text{dist}(\mathcal{P}_1, \mathcal{P}_2) = \min_{\mathbf{x}, \mathbf{y}} \{ \|\mathbf{x} - \mathbf{y}\|_2 \mid \mathbf{A}_1 \mathbf{x} \leq \mathbf{b}_1, \mathbf{A}_2 \mathbf{y} \leq \mathbf{b}_2 \}$$

### Dual Formulation

$$\begin{aligned} \text{dist}(\mathcal{P}_1, \mathcal{P}_2) &= \max_{\lambda_{12}, \lambda_{21}, \mathbf{s}} -\mathbf{b}_1^\top \lambda_{12} - \mathbf{b}_2^\top \lambda_{21} \\ \text{s.t. } &\mathbf{A}_1^\top \lambda_{12} + \mathbf{s} = 0, \quad \mathbf{A}_2^\top \lambda_{21} - \mathbf{s} = 0, \\ &\|\mathbf{s}\|_2 \leq 1, \quad -\lambda_{12} \leq 0, \quad -\lambda_{21} \leq 0 \end{aligned}$$



## 3. Centralized Formulation: Nonlinear MPC

$$\begin{aligned} \min_{\mathbf{u}^i(\cdot|t)} \quad & \text{Cost} \sum_{i=1}^M J^i(\mathbf{z}^i, \mathbf{u}^i) \\ \text{subject to } & \text{Dynamics } \mathbf{z}^i(k+1|t) = f(\mathbf{z}^i(k|t), \mathbf{u}^i(k|t)), \\ & \text{Initial Condition } \mathbf{z}^i(0|t) = \mathbf{z}^i(t), \\ & \text{State and Input Limits } \mathbf{z}^i(k|t) \in \mathcal{Z}, \quad \mathbf{u}^i(k|t) \in \mathcal{U}, \\ & \text{Collision Avoidance } \mathcal{P}(\mathbf{z}^i(k|t)) \cap \mathcal{P}(\mathbf{z}^j(k|t)) = \emptyset, \quad i \neq j \\ & \quad \forall i \in \mathcal{V}, \quad j \in \mathcal{N}_i, \text{ and } k \in \{1, 2, \dots, N\}. \end{aligned}$$

Robot Index

## 4. Distributed Formulation: Parallelization

Splitting the centralized optimization into smaller subproblems  
NMPC+CA to be solved in parallel.

### NMPC Optimization

$$\begin{aligned} \min_{\mathbf{u}^i(\cdot|t)} \quad & J^i(\mathbf{z}^i, \mathbf{u}^i) \\ \text{subject to } & \mathbf{z}^i(k+1|t) = f(\mathbf{z}^i(k|t), \mathbf{u}^i(k|t)), \\ & \mathbf{z}^i(0|t) = \mathbf{z}^i(t), \\ & \mathbf{z}^i(k|t) \in \mathcal{Z}, \quad \mathbf{u}^i(k|t) \in \mathcal{U}, \\ & (-\mathbf{b}^i(\mathbf{z}^i(k|t))^\top \bar{\lambda}_{ij}(k|t) \\ & \quad - \bar{\mathbf{b}}^j(\bar{\mathbf{z}}^j(k|t))^\top \bar{\lambda}_{ij}(k|t)) \geq d_{\min}, \\ & \mathbf{A}^i(\mathbf{z}^i(k|t))^\top \bar{\lambda}_{ij}(k|t) + \bar{\mathbf{s}}_{ij}(k|t) = 0, \\ & \text{for all } k \in \{1, 2, \dots, N\}, \end{aligned}$$

The bar symbol denotes fixed values.

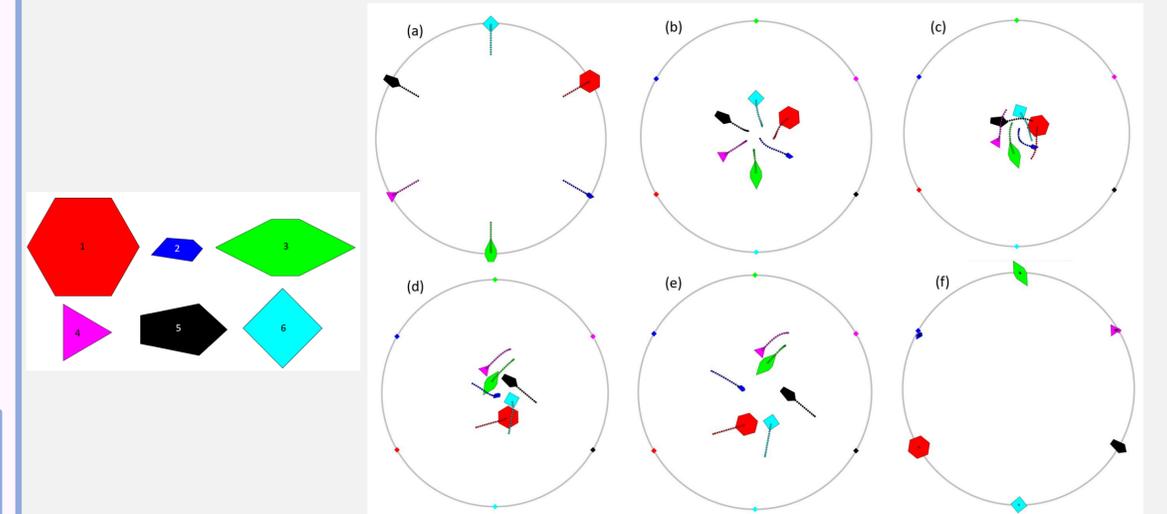
### Collision Avoidance (CA) Optimization

$$\begin{aligned} \max_{\lambda_{ij}(\cdot|t), \lambda_{ji}(\cdot|t), \mathbf{s}_{ij}(\cdot|t)} \quad & -\bar{\mathbf{b}}^i(\bar{\mathbf{z}}^i(k|t))^\top \lambda_{ij}(k|t) - \bar{\mathbf{b}}^j(\bar{\mathbf{z}}^j(k|t))^\top \lambda_{ji}(k|t) \\ \text{subject to } & \bar{\mathbf{A}}^i(\bar{\mathbf{z}}^i(k|t))^\top \lambda_{ij}(k|t) + \mathbf{s}_{ij}(k|t) = 0, \\ & \bar{\mathbf{A}}^j(\bar{\mathbf{z}}^j(k|t))^\top \lambda_{ji}(k|t) - \mathbf{s}_{ij}(k|t) = 0 \\ & (-\bar{\mathbf{b}}^i(\bar{\mathbf{z}}^i(k|t))^\top \lambda_{ij}(k|t) \\ & \quad - \bar{\mathbf{b}}^j(\bar{\mathbf{z}}^j(k|t))^\top \lambda_{ji}(k|t)) \geq d_{\min}, \\ & \|\mathbf{s}_{ij}(k|t)\|_2 \leq 1, \quad -\lambda_{ij}(k|t) \leq 0, \\ & -\lambda_{ji}(k|t) \leq 0, \text{ for all } i \in \mathcal{V}, \quad j \in \mathcal{N}_i, \\ & \text{for all } k \in \{1, 2, \dots, N\}. \end{aligned}$$

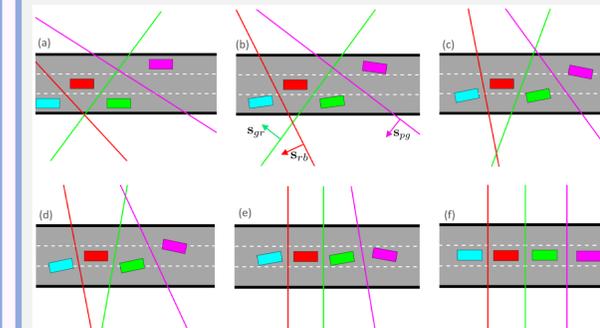
Alternating

## 5. Results: Formation and Reconfiguration

### Robots with Different Polytopic Shapes



### Connected and Automated Vehicles



#	Centralized		Distributed			
	Avg.	Max.	NMPC		CA	
2	1.3851	2.7482	0.1457	0.3621	0.0022	0.0022
			0.1102	0.3313	0.0021	0.0022
			Total Avg. = 0.1507			
3	2.7680	4.6817	0.1848	0.3933	0.0025	0.0028
			0.1206	0.3169	0.0024	0.0026
			Total Avg. = 0.1514			
4	12.8331	28.7763	0.1919	0.4211	0.0030	0.0024
			0.1125	0.2602	0.0027	0.0029
			Total Avg. = 0.1757			

Computation Time (Sec)

## 6. Conclusion

Using the proposed distributed algorithm the heterogenous team of robots can plan dynamically feasible trajectories in real time and navigate safely in tight environments to reach their goals.