

Non-cooperative target localisation using rank based EDM approach

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Outline

- Introduction
- Problem Formulation
- Proposed Algorithm
- Results and Conclusion

Target Localisation

Locating RF emissions

To provide a greater level of situational awareness for units in an operational area

Locating their own units

SEARCH & RESCUE

Range based

TOA, TDOA, RSS

Range free

Knowledge of connectivity
possibilities

Localisation

Distributed

Sensors perform many
calculations and analysis to
locate the target

Centralised

Sensors pass information
to Fusion centre

Mobile Ad-hoc Sensor Network (MASNET)

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[2] Z. Ma and K. C. Ho, "ToA localisation in the presence of random sensor position errors," in IEEE Int. Conf. Acoust., Speech and Signal Process. (ICASSP), 2011, pp. 2468–2471.

[3] P. Singh and S. Agrawal, "TDoA based node localisation in WSN using neural networks," in Int. Conf. Commun. Syst. and Network Technologies, 2013, pp. 400–404.

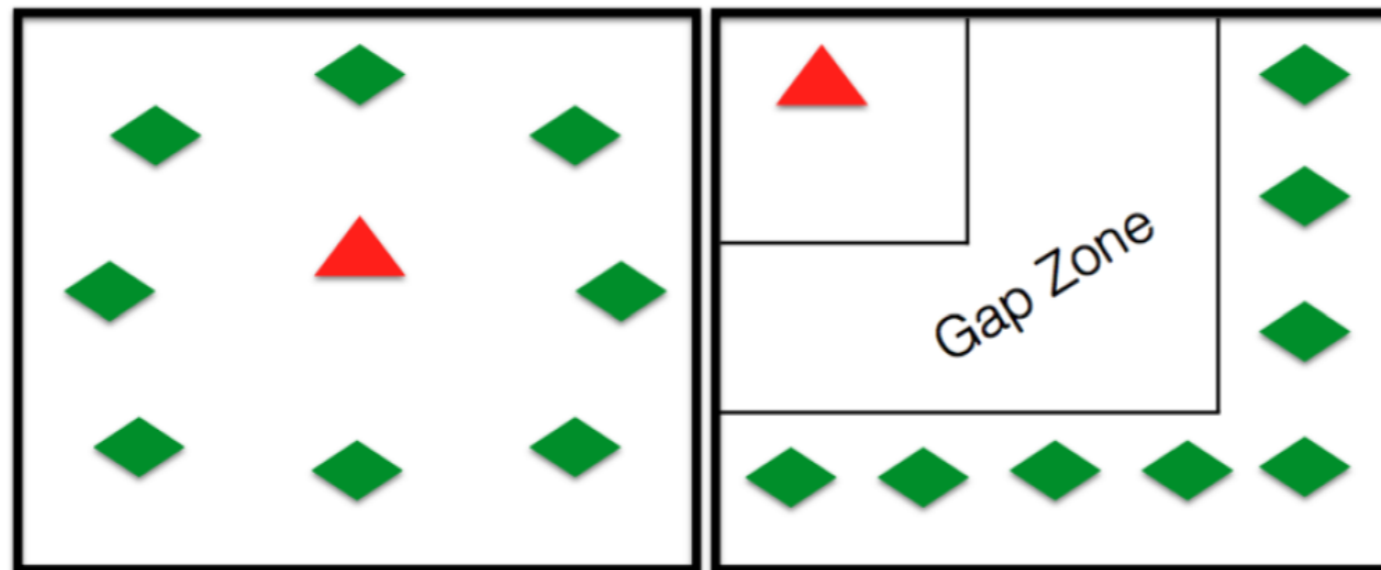
[4] A. Beck, P. Stoica, and J. Li, "Exact and approximate solutions of source localisation problems," IEEE Transactions on Signal Processing, vol. 56, no. 5, pp. 1770–1778, May 2008.

[5] L. Gui, A. Wei, and T. Val, "A range-free localization protocol for wireless sensor networks," in Int. Symp. Wireless Commun. Syst. (ISWCS), Paris, 2012, pp. 496–500.

System Model

Centralised Range based Localisation

Mixed



Separated

▲ Target Node ◆ Sensor Node

- Each sensor acquires the time information of the received signal and transmit it to the fusion centre
- The fusion centre converts this information to the corresponding distances between target and sensors nodes

Problem Formulation

- Euclidean Distance Matrix (EDM) have useful properties and are used in applications such as crystallography, acoustics,...
- EDM is a symmetric hollow matrix of squared distance between points
- Rank of an EDM is at most “d+2”, where d is the dimensional space

Two sets of distance problems

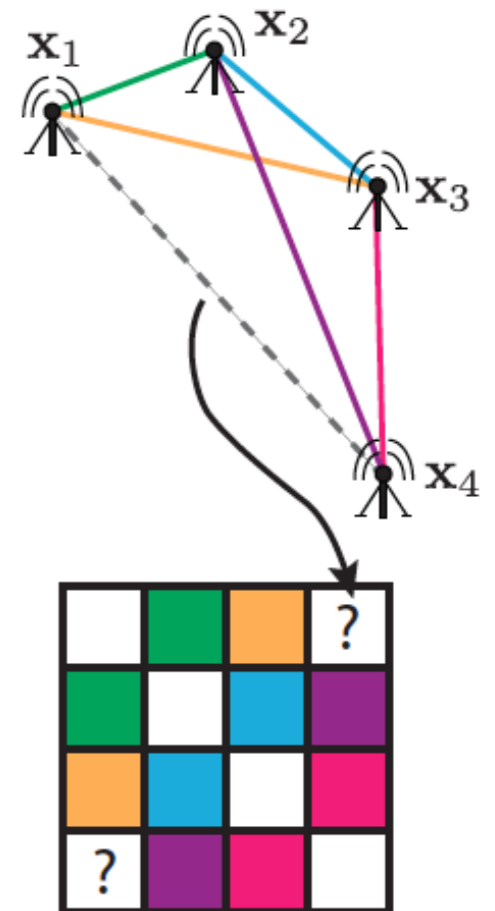
- Distances are noisy \equiv All sensors detect
- Some distances are missing \equiv Not all sensors detect

$$D = \begin{bmatrix} D_T & D_{TS} \\ D_{ST} & D_S \end{bmatrix}$$

D_T : distances between target nodes

D_{TS} : distances between target and sensor nodes

D_S : distances between sensor nodes (known)



Alternating Rank based EDM

Noisy
distances (D)

embedding
dimension (d)

Alternating rank based algorithm

- After algorithm's convergence, the relative positions for target and sensors are computed
- These positions may be rotated and/or translated
- Exploiting Procrustes Analysis, the absolute position of target is computed

Algorithm 1 Alternating Rank Based EDM Algorithm

Inputs: D, W, d, max.tolerance

Output: E

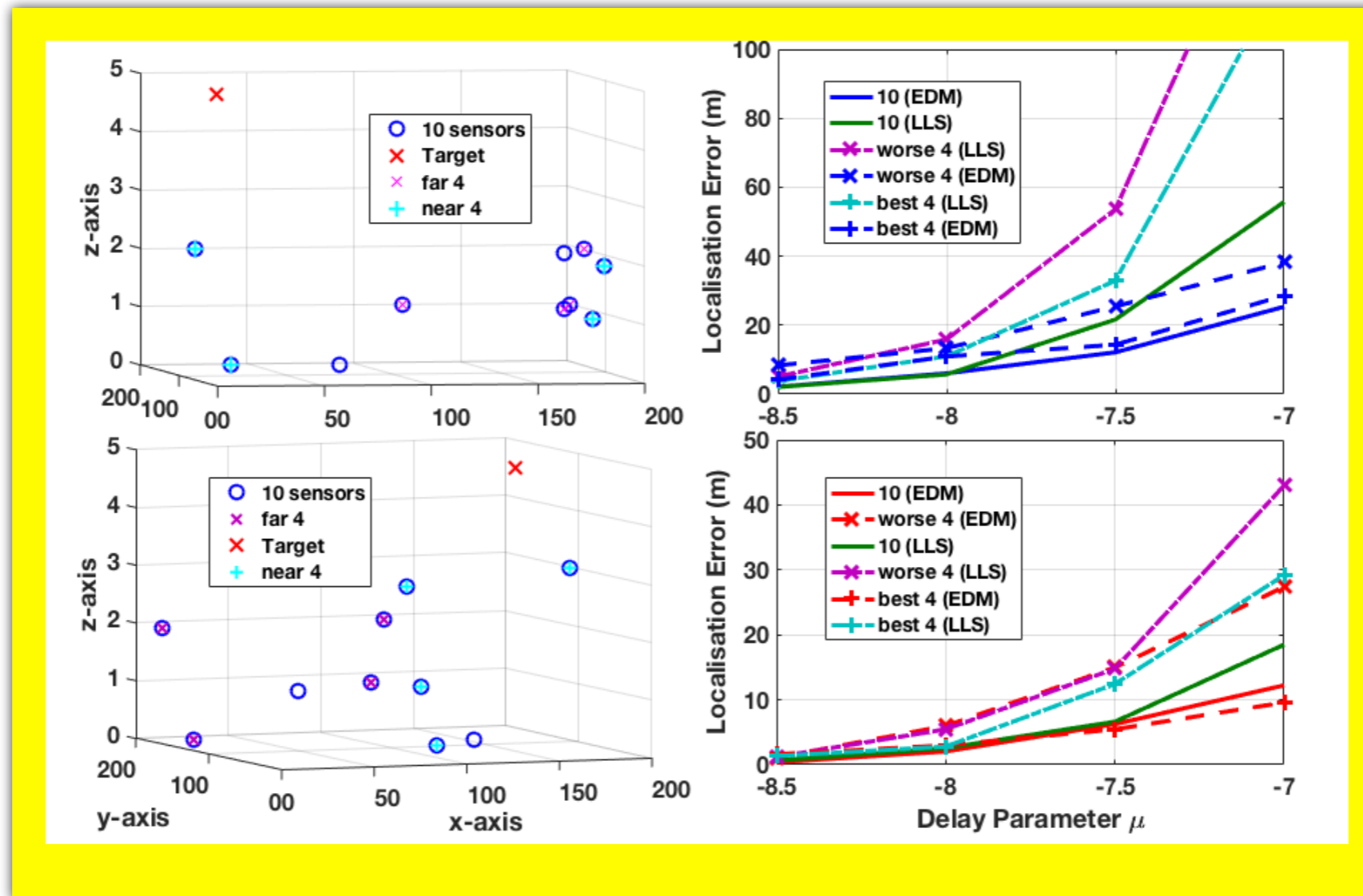
Initialization and Definitions:

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1:  $\mathbf{D}_{11^T - \mathbf{W}} \leftarrow \phi$       Initialize unknown entries
2:  $\mathbf{J} \leftarrow \mathbf{I} - \frac{1}{k} \mathbf{1}\mathbf{1}^T$     Define Geometric centering matrix
3: repeat
4:    $\mathbf{G} \leftarrow \frac{-1}{2} \mathbf{J}\mathbf{D}\mathbf{J}$       Compute Gram Matrix
5:    $\mathbf{U}, [\lambda]_{i=1}^k \leftarrow \text{EVD}(\mathbf{G})$   Eigenvalue Decomposition
6:    $\Sigma \leftarrow \text{diag}(\lambda_1, \dots, \lambda_d, 0, \dots, 0)$ 
7:   Compute  $\mathbf{G} \leftarrow \mathbf{U}\Sigma\mathbf{U}^T$ 
8:   Compute  $\mathbf{E} = \text{diag}(\mathbf{G})\mathbf{1}^T - 2\mathbf{G} + \mathbf{1}\text{diag}(\mathbf{G})$ 
9:   Compute  $e_1 = \|\mathbf{E} - \mathbf{D}\|_F$ 
10:   $\mathbf{E}_W \leftarrow \mathbf{D}_W$       Enforce known entries
11:   $\mathbf{E}_I \leftarrow 0$       Set Diagonal to zero
12:   $(\mathbf{E})_- \leftarrow 0$       Assign zeros to the negative entries
13:  Compute  $e_2 = \|\mathbf{E} - \mathbf{D}\|_F$ 
14:  if  $(e_1 < \text{max. tolerance}) \vee (e_2 < \text{max.tolerance})$  then
15:    return E
16:  else
17:     $\mathbf{D} \leftarrow \mathbf{E}$ 
18:  end if
19: until Convergence or MaxIter
20:  $\hat{\mathbf{X}} = \Sigma^{1/2} \mathbf{U}^T$ 

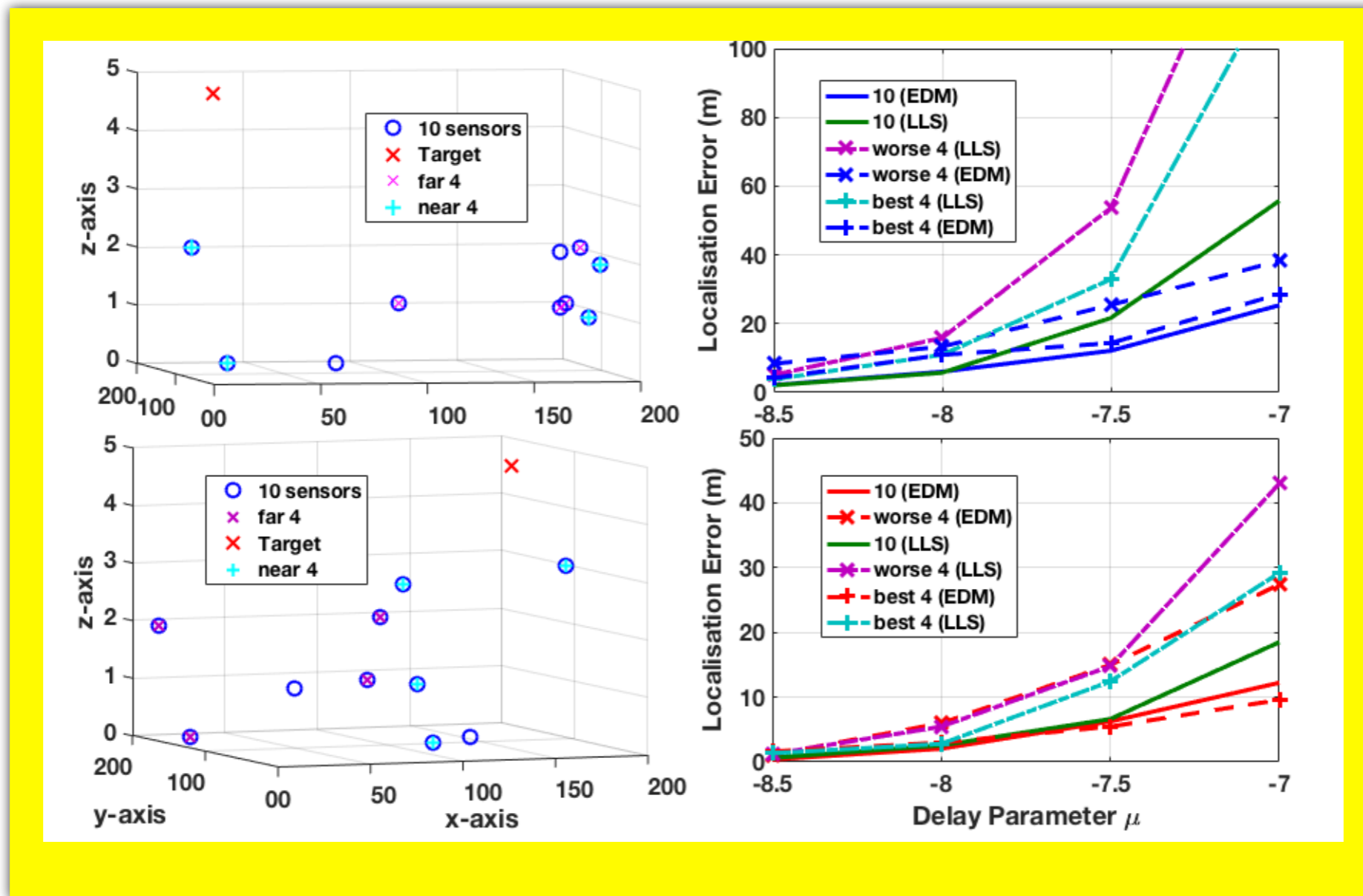
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Results

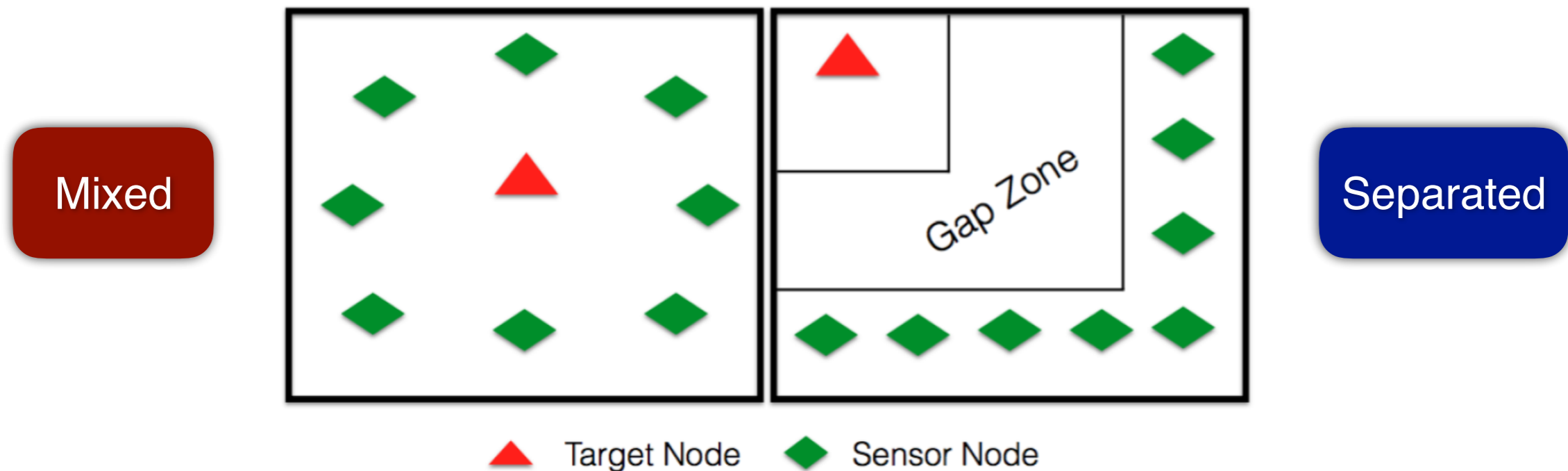


- Localisation error is the distance between the estimated and actual target positions.
- The delay parameter is the mean of exponential random variable
- Low delay values corresponds to less noise and vice versa

Results



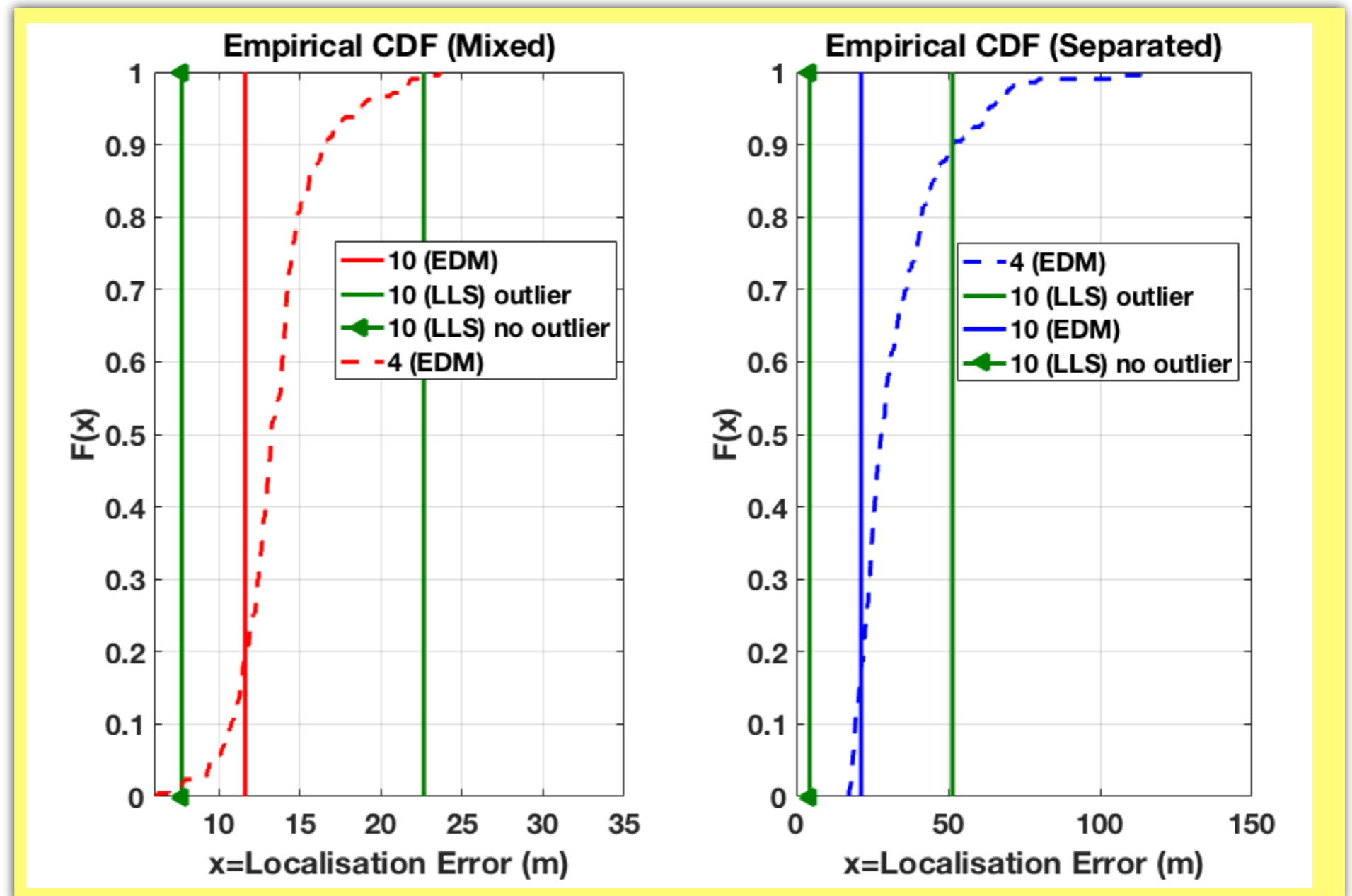
- Localisation error is larger for separated than mixed
- Localisation error using all ten and four sensors are close enough in low noise region
- LLS is worse than EDM due to outliers resulting from larger errors in the measured distances



- The sensor nodes in the separated scenario experience similar receiving time information, hence similar distance measurements.
- This leads to an EDM which is not well separated in terms of distances and hence not well conditioned.

Results

- For no outlier case, we assume that all sensors experience similar delay
- **Delay distribution for no outlier case is uniform with small standard deviation**



In some rare cases, using EDM can lead to better performance for chosen four sensors than using all ten sensors
Reason : Chosen four sensors are close to the target and experiencing overall small delay

Conclusion

- At high delay values the proposed algorithm outperforms the conventional LLS algorithm
- Localisation error depends on the amount of delay and the operating scenario for the sensors (separated or mixed)
- The sensor nodes in the separated scenario experience similar receiving time information which results in larger localisation error than in the mixed scenario
- The proposed algorithm is less sensitive to outliers

*Thank
you*

