

# Distributed TDOA- based indoor source localisation

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# Outline

- Background
- Problem Formulation
- Pruning out erroneous TOA measurements
- TDOA-based localization
- Experiments
- Conclusion

# Background

- Localisation techniques
  - In the hospital
  - In museums
- Outdoor environment: GPS
- Indoor environment:  
TOA, TDOA, RSS, AOA

# Problem Formulation

- Time difference of arrival techniques
- $M$  receivers with known location

$$\mathbf{r}_1, \mathbf{r}_2, \dots, \mathbf{r}_M$$

- $N$  transmitters whose locations are to be estimated

$$\mathbf{s}_1, \mathbf{s}_2, \dots, \mathbf{s}_N$$

- The time of arrival (TOA) information

$$t_{ij} = c^{-1} \|\mathbf{r}_i - \mathbf{s}_j\| + n_{ij}$$

# Pruning out erroneous TOA measurements

- Erroneous TOA measurements
- TOA measurements

$$t_{ij} = \|\mathbf{r}_i - \mathbf{s}_j\|$$



$$-(\mathbf{r}_i - \mathbf{r}_1)^T (\mathbf{s}_j - \mathbf{s}_1) = 0.5(t_{ij}^2 - t_{1j}^2 - t_{i1}^2 + t_{11}^2)$$



$$\mathbf{R}^T \mathbf{S} = \mathbf{T}$$

# Pruning out erroneous TOA measurements

- The set of all  $N - 1$  unique combinations of the set of TOA measurements

$$U_{N-1} = \binom{S_N}{N-1}$$

- For a specific combination, construct  $\mathbf{T}_u$ , and compute  $e_u$

$$e_u = \|\mathbf{T}_u\|_F^2 = \sum_{i=1}^{N_r} \sigma_i^2(\mathbf{T}_u)$$

# Pruning out erroneous TOA measurements

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**Algorithm 1** Pruning incorrect TOA measurements

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1. For  $n = 0, 1, N - N_{min}$
2. Generate the set of all possible combinations of the set  $S_{N-n}$

$$U_{N-n+1} = \binom{S_{N-n}}{N-n+1}$$

3. For each  $u \in U_{N-n+1}$ , construct  $\mathbf{T}_u$  and compute  $e_u$ .
4. Update the best TOA sets,

$$S_{N-n+1} = \begin{cases} \arg \min_s e_u & \text{if } \min e_u / \max e_u < \alpha \\ S_{N-n} & \text{otherwise} \end{cases}$$

5. End if  $S_{N-n+1} = S_{N-n}$
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# TDOA-based localization

## — Centralised localisation 1

- TDOA measurements

$$\Delta t_{1j} = t_1 - t_j = \frac{\|\mathbf{s} - \mathbf{r}_1\|}{c} - \frac{\|\mathbf{s} - \mathbf{r}_j\|}{c} + n_{1j}$$

- Define

$$d_j^2 = (x_j - x_0)^2 + (y_j - y_0)^2$$

$$K_j^2 = x_j^2 + y_j^2$$

- For receiver  $j$ :

$$-x_j x_0 - y_j y_0 = d_{j1} d_1 + \frac{1}{2}(d_{j1}^2 - K_j^2)$$



# TDOA-based localization

## — Centralised localisation 1

- Vector form:

$$\mathbf{A}_1 \mathbf{s} = d_1 \mathbf{b}_1 + \mathbf{c}_1$$

- Least-square problem:

$$\min_s \|\mathbf{A}_1 \mathbf{s} - (d_1 \mathbf{b}_1 + \mathbf{c}_1)\|_2^2$$

- Solution:

$$\hat{\mathbf{s}} = (\mathbf{A}_1^T \mathbf{A}_1)^{-1} \mathbf{A}_1^T (d_1 \mathbf{b}_1 + \mathbf{c}_1)$$

# TDOA-based localization

## — Centralised localisation 2

- Least-square problem:

$$\min_{\mathbf{s}_k} \|\mathbf{A}_k \mathbf{s}_k - (d_k \mathbf{b}_k + \mathbf{c}_k)\|_2^2$$

- Solution:

$$\hat{\mathbf{s}}_k = (\mathbf{A}_k^T \mathbf{A}_k)^{-1} \mathbf{A}_k^T (d_k \mathbf{b}_k + \mathbf{c}_k)$$

- Average location estimation:

$$\hat{\mathbf{s}} = \frac{1}{M} \sum_{k=1}^M \hat{\mathbf{s}}_k$$

# TDOA-based localization

## — Distributed localisation

- Distributed localisation:

$$\min_{\mathbf{s}_k} \sum_{k=1}^{M_k} \|\mathbf{A}_k \mathbf{s}_k - (d_k \mathbf{b}_k + \mathbf{c}_k)\|_2^2,$$

subject to  $\mathbf{s}_k = \mathbf{s}_j, \forall (k, j) \in E$

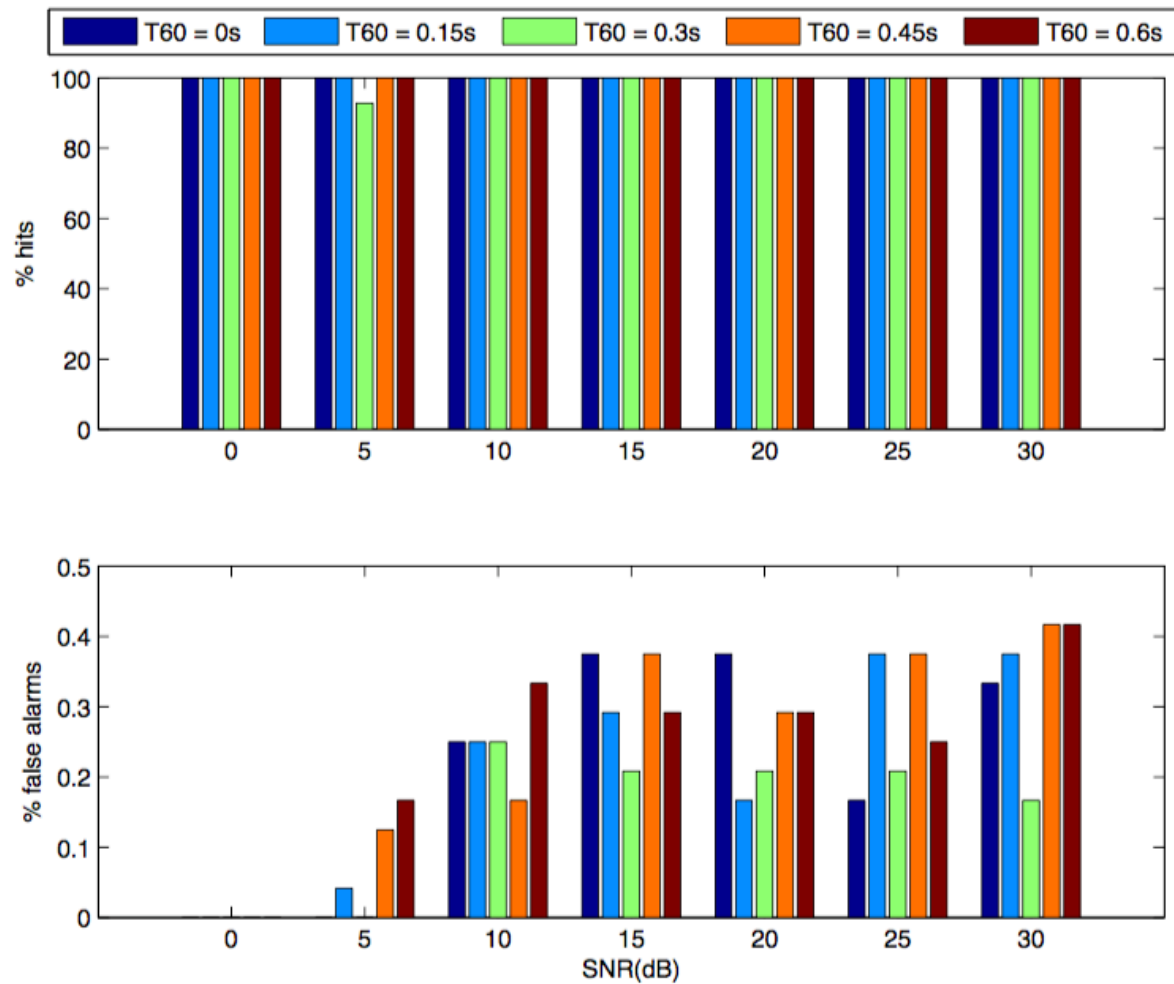
- Solution: standard solvers.

# Experiments

## — Experimental Setup

- Room: 6×5×4 m
- 8 receivers, 30 transmitters
- A hit: an erroneous TOA measurement being detected
- A false alarm: a correct TOA measurement being classified as erroneous

# Experiments

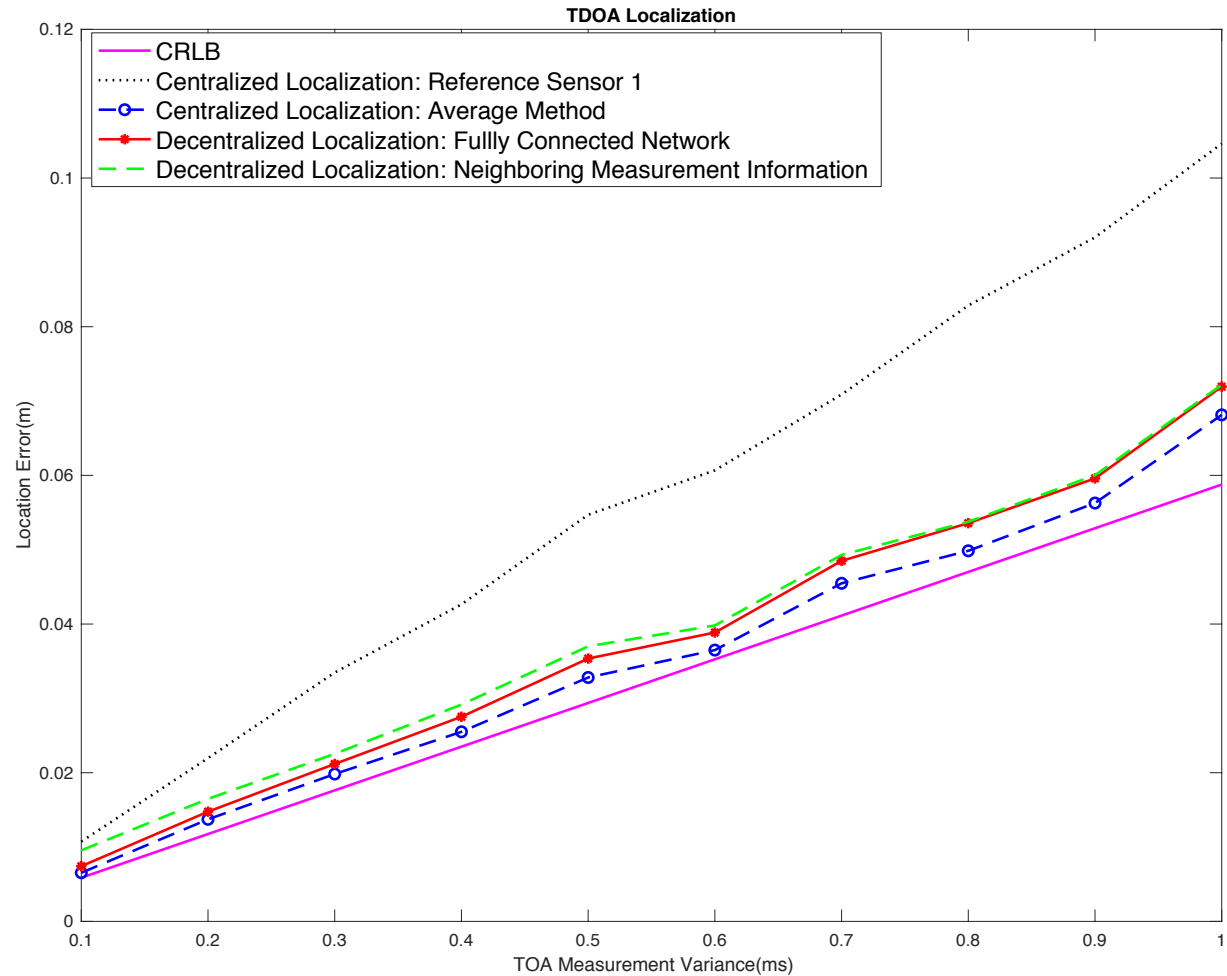


# Experiments

## — Experimental Setup

- Room:  $10 \times 10$  m
- 8 receivers uniformly placed on the boundary
- Two receivers can communicate when the distance is less than 10 m
- Propagation speed:  $c = 340$  m/s

# Experiments



# Conclusions

- An algorithm to prune out erroneous TOA measurements.
- TDOA-based localisation
  - Centralised localisation
  - Distributed localisation
- Accessing to neighbouring information doesn't decrease the accuracy.