

Simultaneous Power-Based Localization of Transmitters for Crowdsourced Spectrum Monitoring

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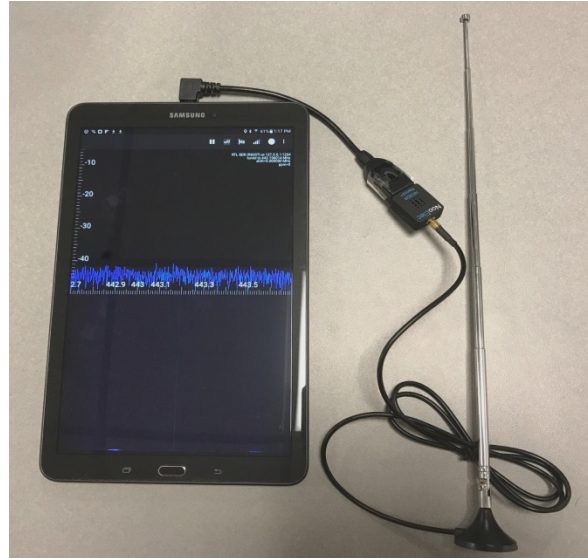
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Need for spectrum monitoring

As SDR becomes ubiquitous



Present: externally attached SDR

Future: mobile devices with integrated SDR

More threats will appear

- Virus affected multiple SDRs leading to jamming
- Selfish users altering SDR radio

How to prevent spectrum offence?

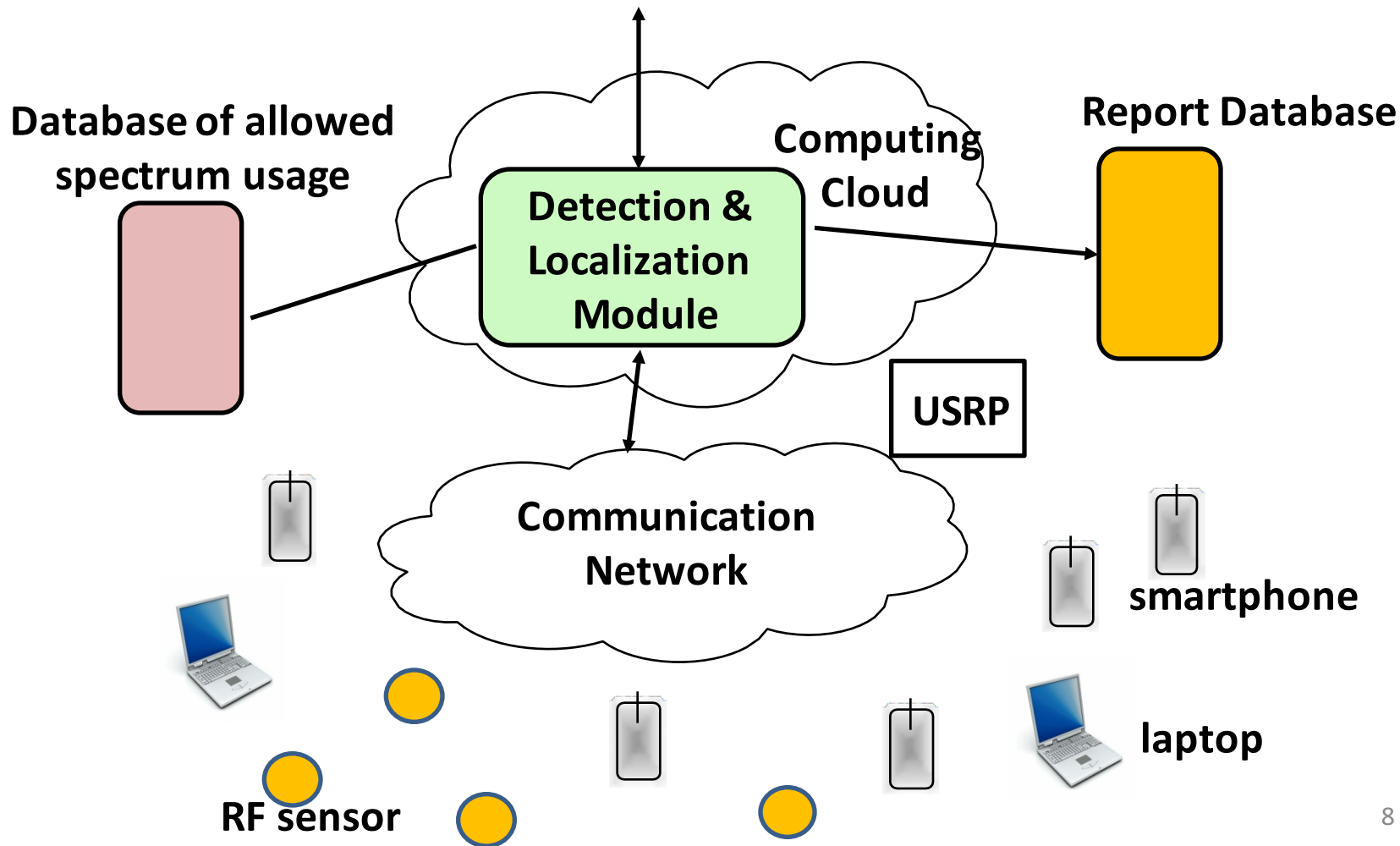
What method exists?

- FCC enforcement bureau
 - Relies on complaints and manual investigation
- Drawbacks
 - Time consuming
 - Human extensive
 - Expensive

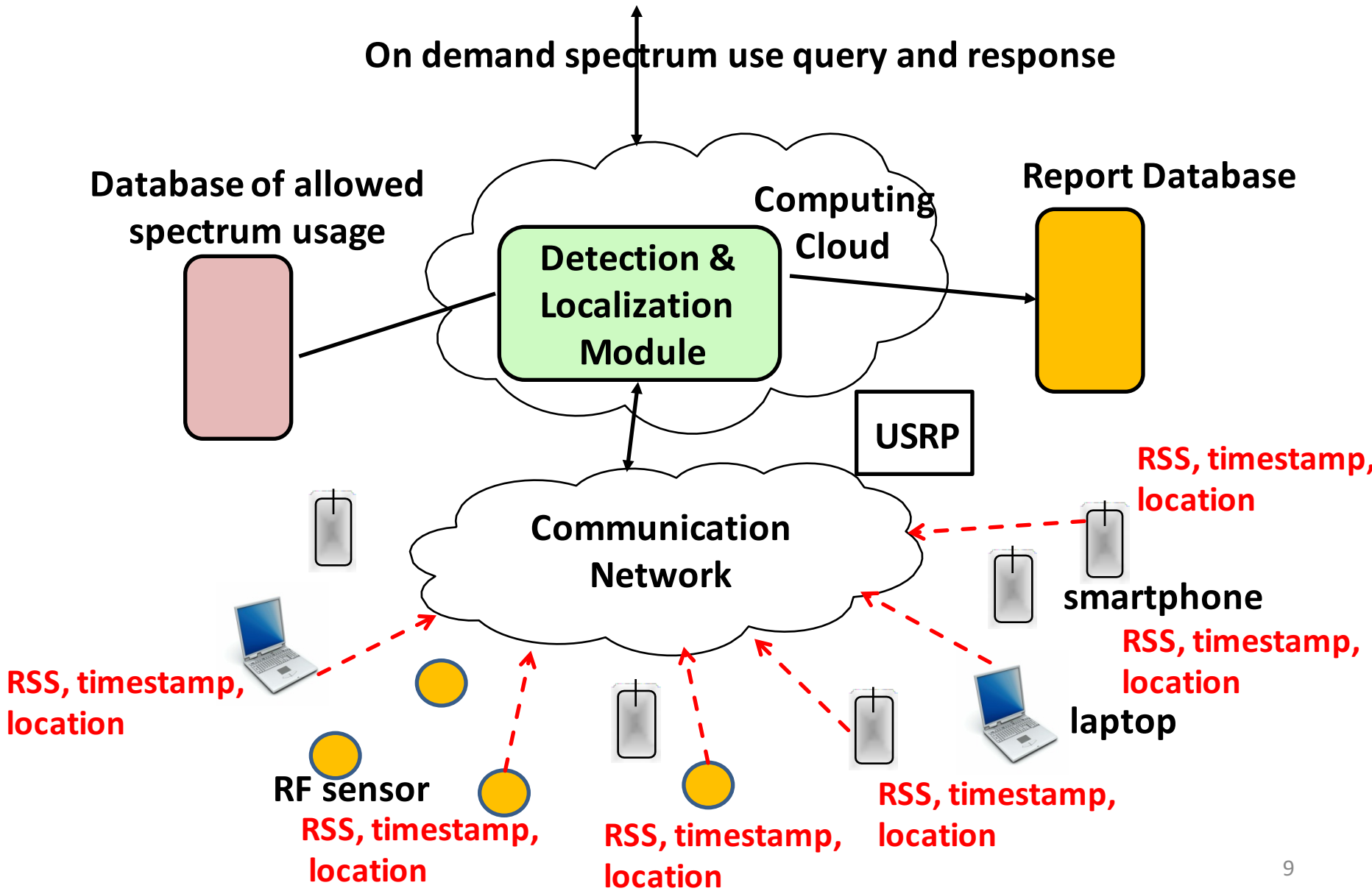
Need **efficient** spectrum monitoring

Detecting, localizing spectrum offenders using Crowdsourcing

On demand spectrum use query and response

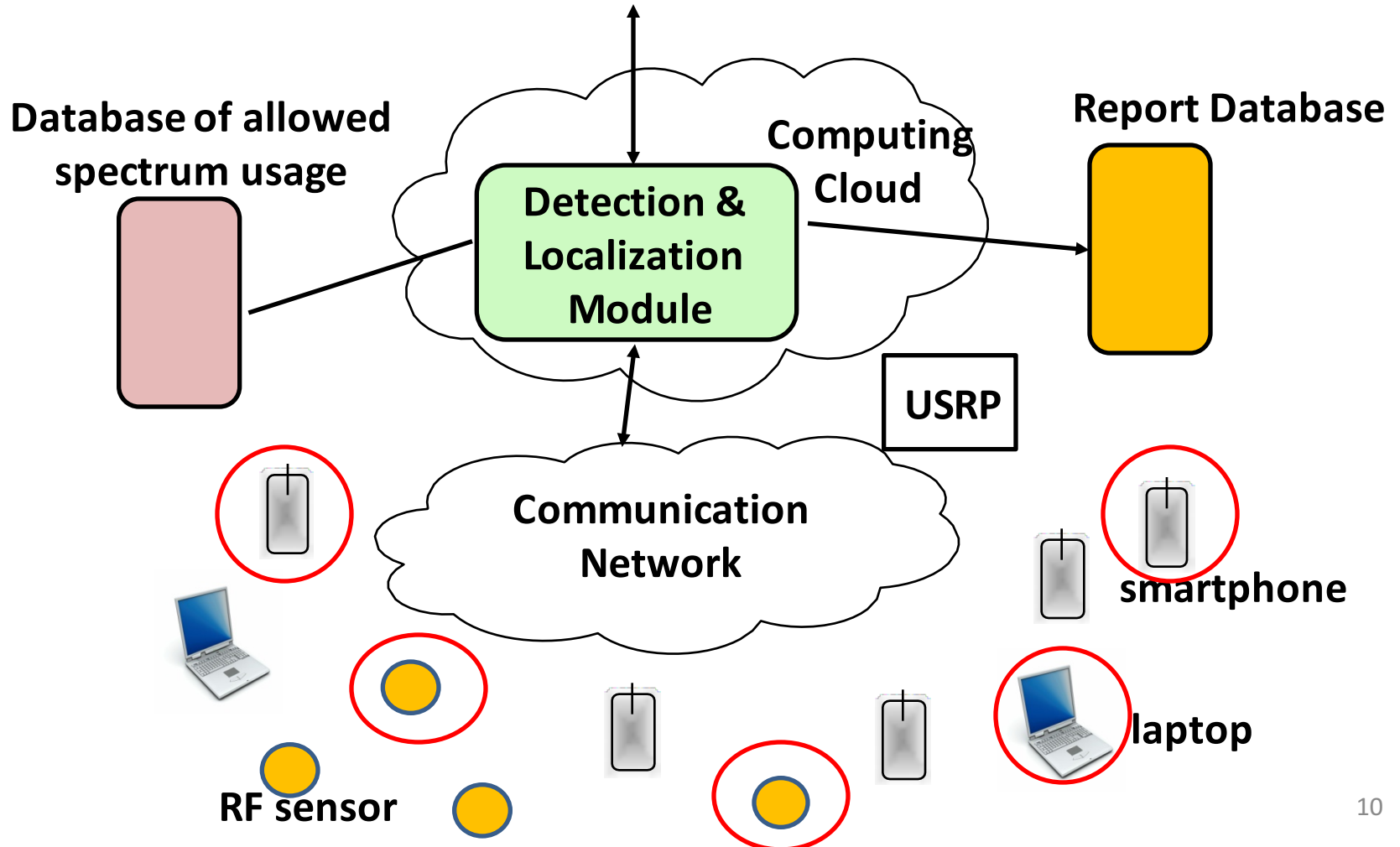


Report RSS, timestamp, location



Multi transmitter localization and selection of receivers

On demand spectrum use query and response



SPLIT

**Simultaneous Power-Based Localization of Transmitters for
Crowdsourced Spectrum Monitoring**

Outline

- SPLIT
 - Multi-source localization
- Selecting receivers and incentivizing them
 - For coverage of monitored area
 - Motivation to participate
- Experiments
 - Indoor, outdoor and test bed
- Summary

Heart of SPLOT

Localization of **simultaneously
active unknown** number of
transmitters

Challenges in multi source localization

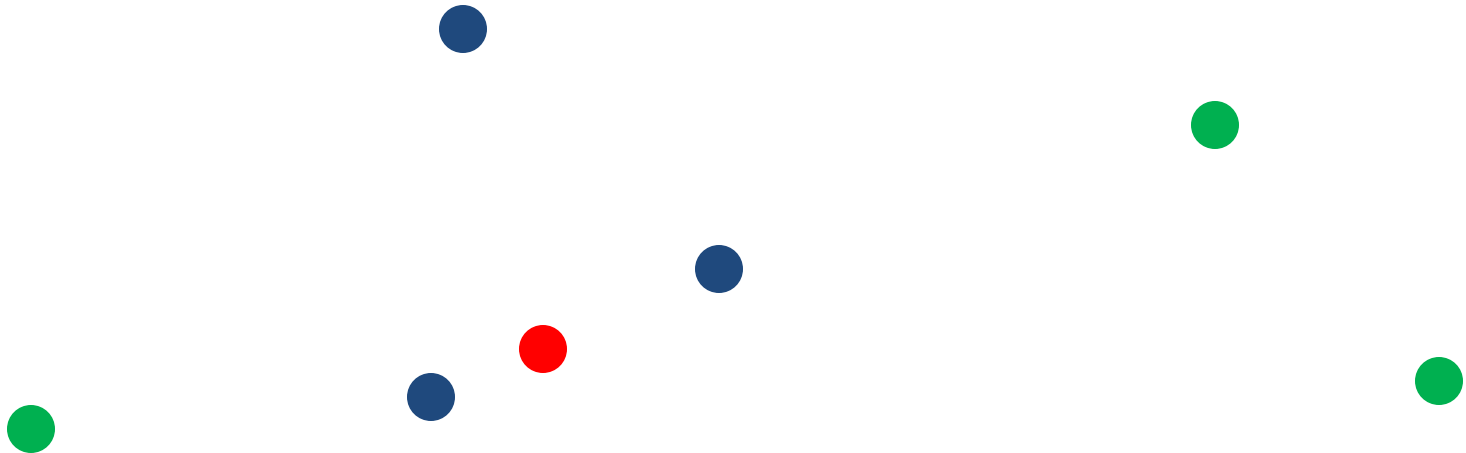
- Unknown number of transmitters
- Received power not separable at receiver
- Mobility of transmitters and receivers
- Unpredictability of receivers participation

Related work

- Quasi Expectation Maximization [Nelson'2009]
 - Number of transmitters known
- Locating multiple transmitters using mobile robots [Kim'2014]
 - Receiver's motion controlled, predictable

Multi source localization in SPLOT

- Observation 1 - receivers near transmitter observe higher power



- Transmitter
- Receiver with high received power
- Receiver with low received power

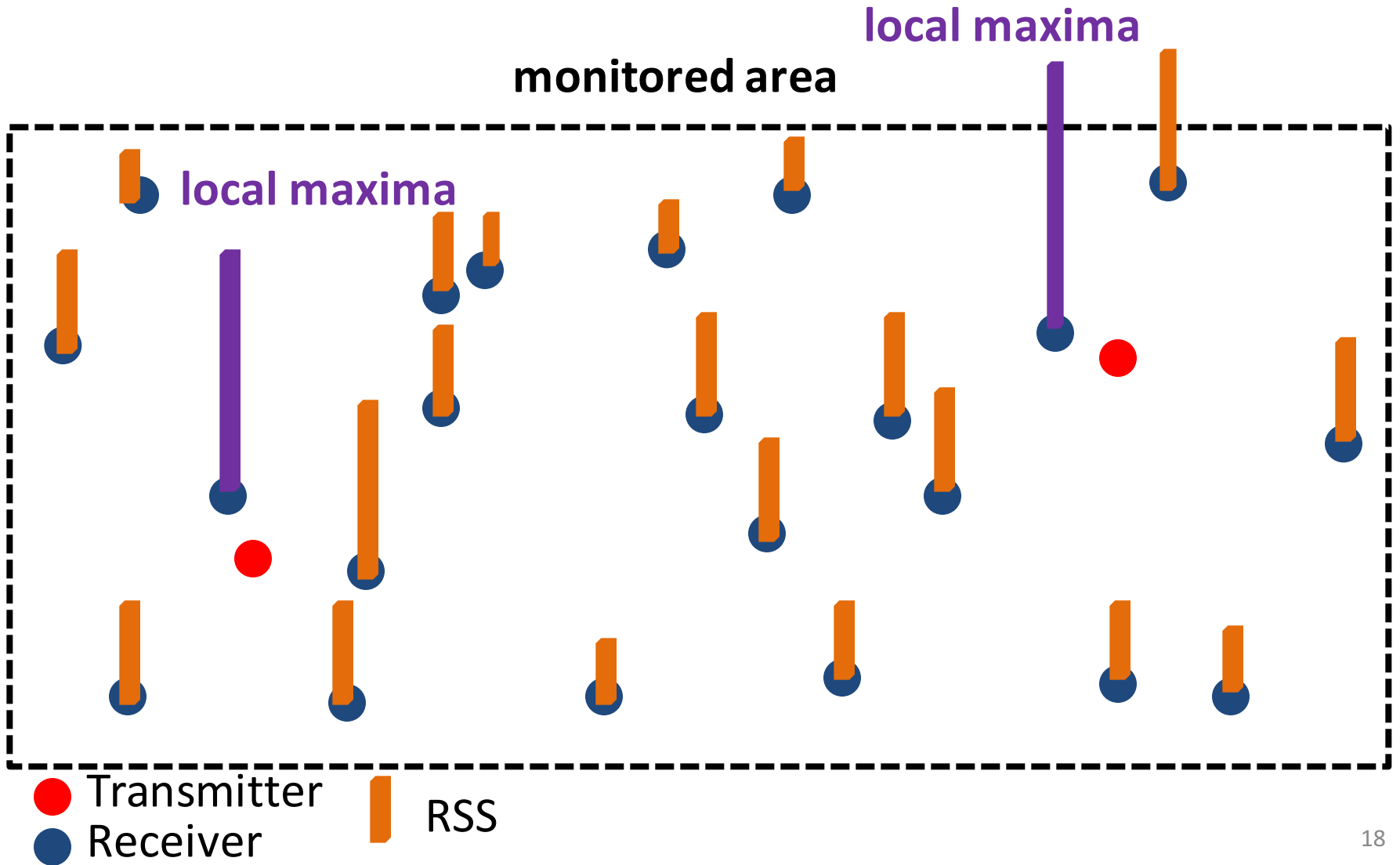
Multi source localization in SPLOT

- Observation 2 - RSS at receiver primarily affected by nearest transmitter



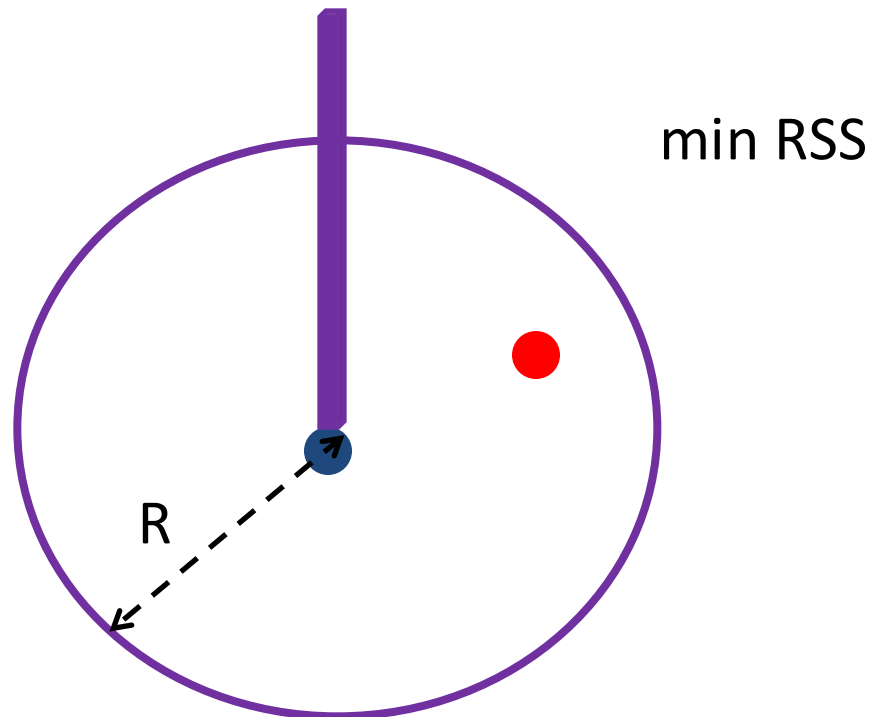
- Transmitter with high impact on receiver's RSS
- Transmitter with low impact on receiver's RSS
- Receiver

Spatial RSS map



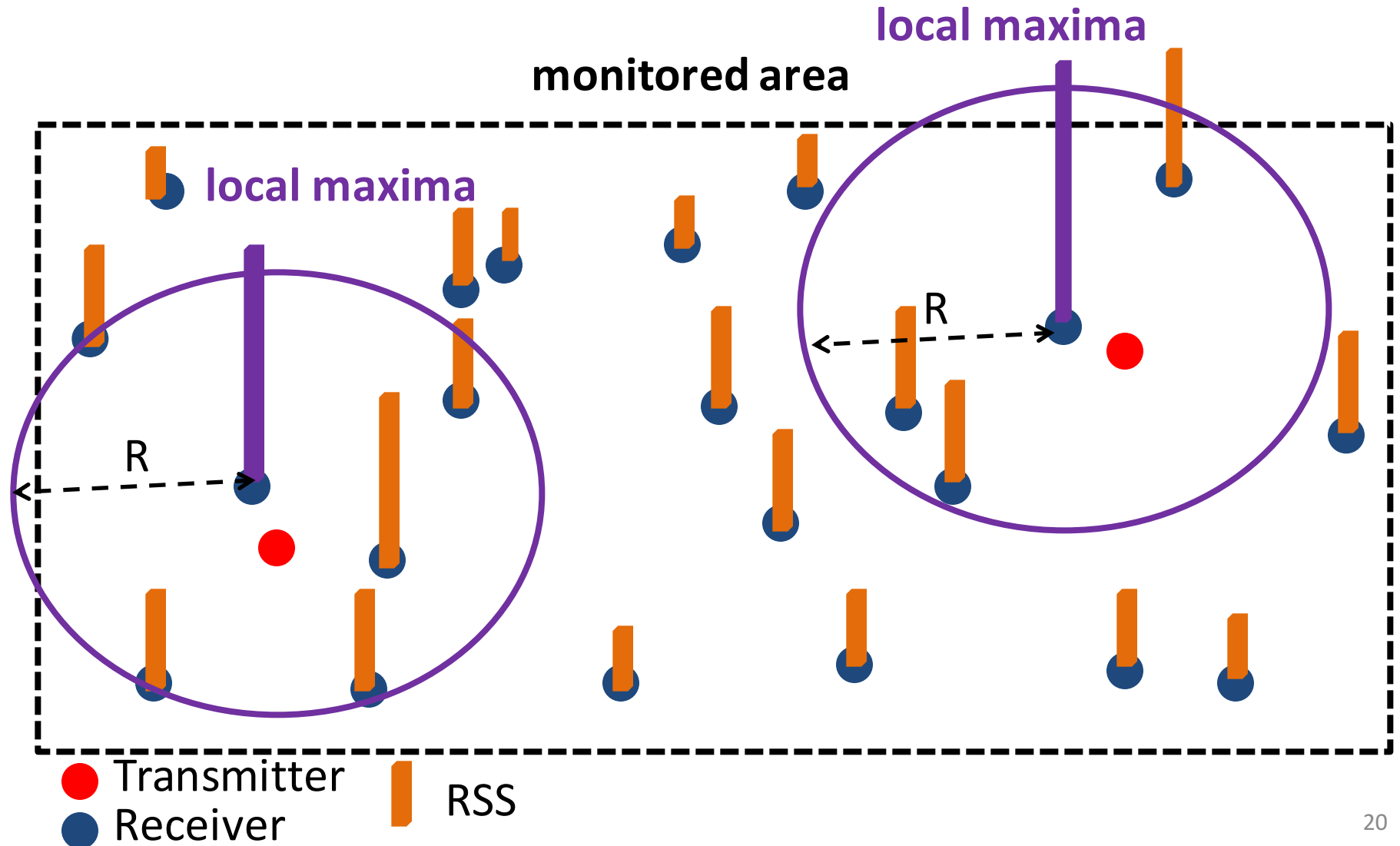
What is local maxima?

- If above minimum RSS that receiver would observe when there is a transmitter nearby and maximum locally

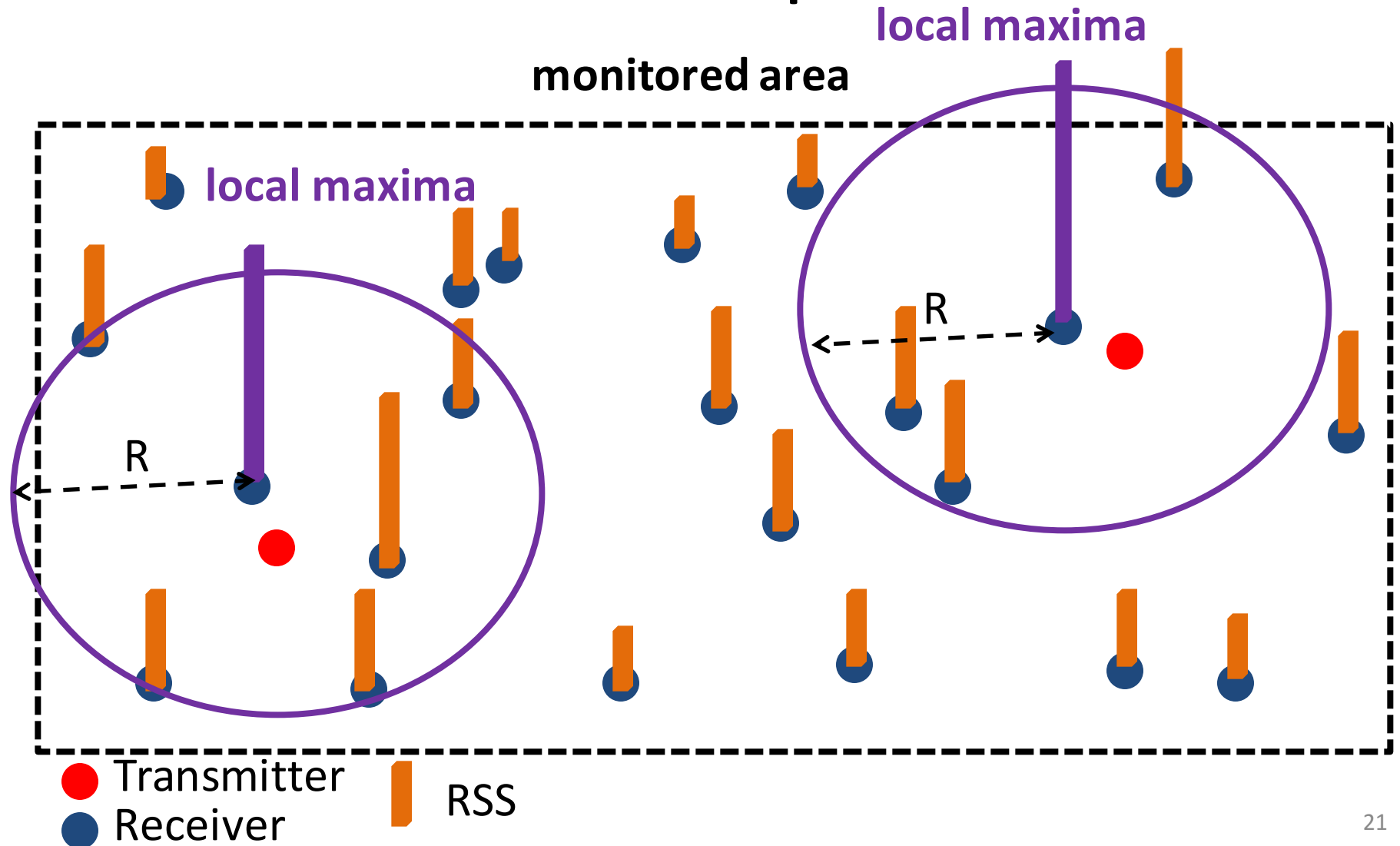


- Transmitter
- Receiver

Approximate region of presence for each transmitter

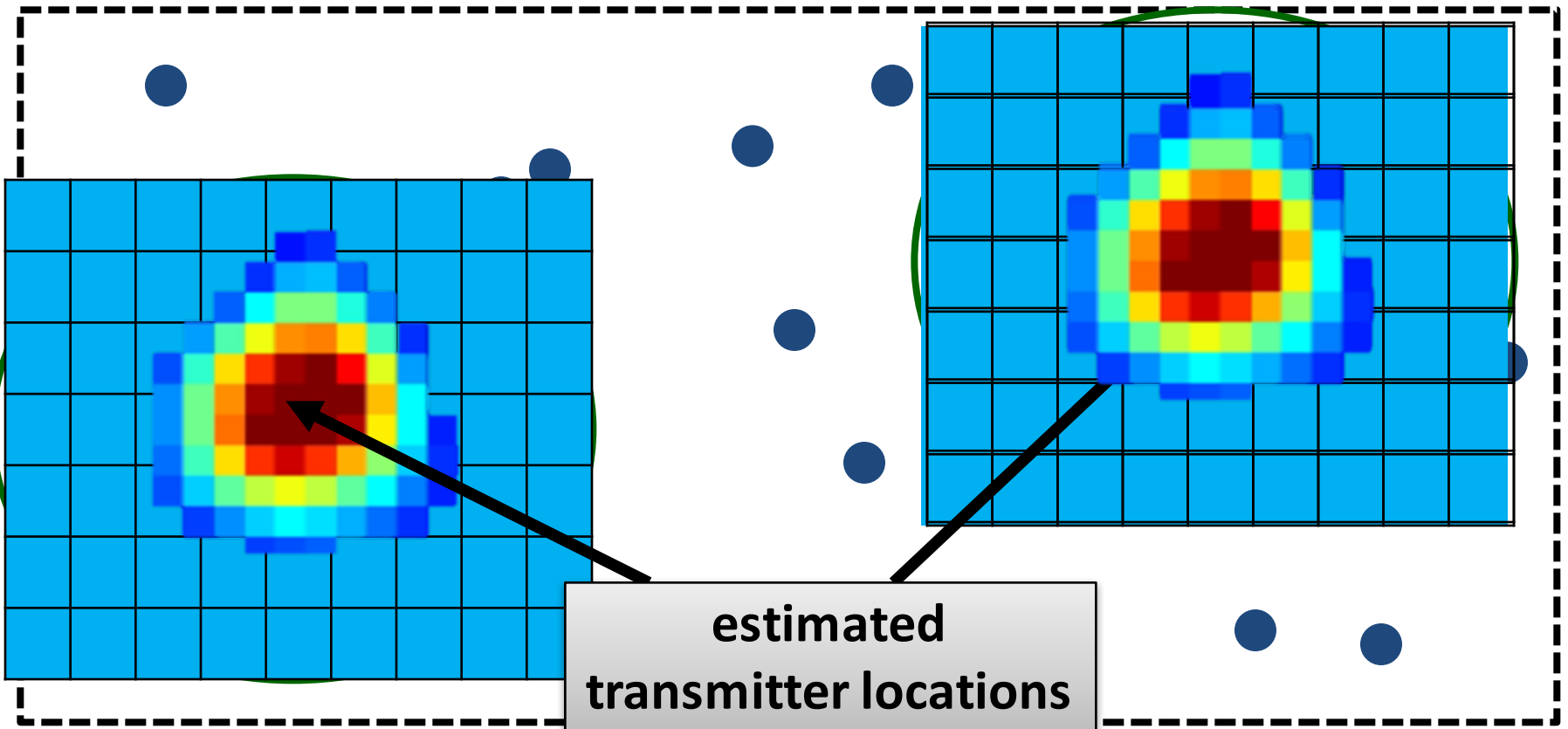


Set of single transmitter localization problems



Solve using matrix inversion

monitored area



- Transmitter
- Receiver

Matrix inversion approach

weight matrix

RSS
observed by
L receivers

$$\begin{bmatrix} y_1 \\ y_2 \\ \dots \\ y_L \end{bmatrix} = \begin{bmatrix} w_{11} & w_{12} & \dots & w_{1Q} \\ w_{21} & w_{22} & \dots & w_{2Q} \\ \dots & \dots & \dots & \dots \\ w_{L1} & w_{L1} & \dots & w_{LQ} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \dots \\ x_Q \end{bmatrix} + \begin{bmatrix} n_1 \\ n_2 \\ \dots \\ n_L \end{bmatrix}$$

noise and
fading

transmit
power
field

where

$$W_{ij} = \begin{cases} d_{ij}^{-\eta_p} & \text{for } d > \min PL \\ \min PL^{-\eta_p} & \text{otherwise} \end{cases}$$

Matrix inversion approach

weight matrix

$$\begin{array}{l} \text{RSS} \\ \text{observed by} \\ \text{L receivers} \end{array} \begin{bmatrix} y_1 \\ y_2 \\ \dots \\ y_L \end{bmatrix} = \begin{bmatrix} w_{11} & w_{12} & \dots & w_{1Q} \\ w_{21} & w_{22} & \dots & w_{2Q} \\ \dots & \dots & \dots & \dots \\ w_{L1} & w_{L1} & \dots & w_{LQ} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \dots \\ x_Q \end{bmatrix} + \begin{bmatrix} n_1 \\ n_2 \\ \dots \\ n_L \end{bmatrix} \begin{array}{l} \text{noise and} \\ \text{fading} \end{array}$$

transmit power field

- Estimate \hat{x} from y
- Estimated transmitter location = $\arg \max X$
- Ill-posed inverse – use least square regularization

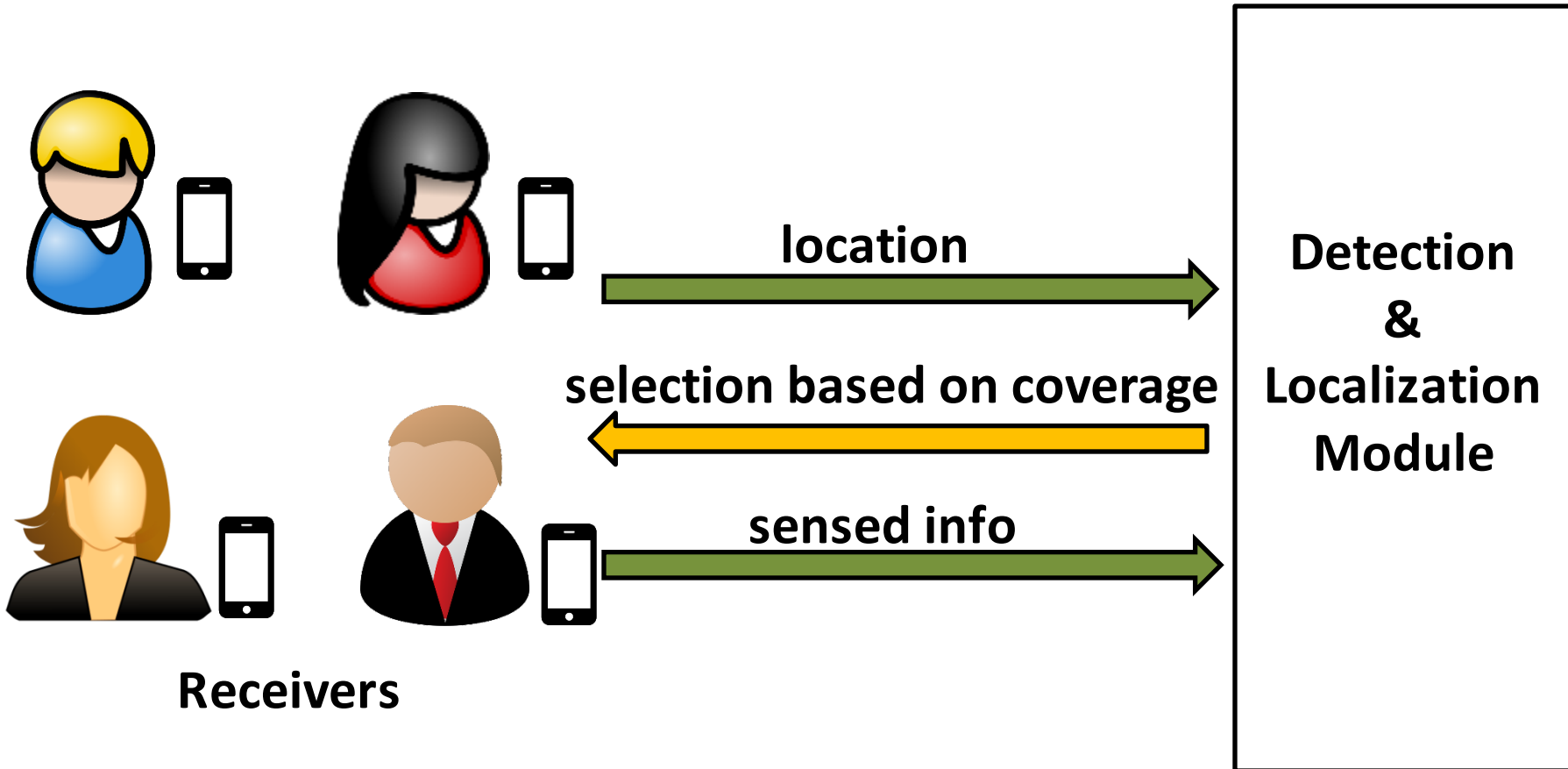
Selection of receivers

Which devices to select for Crowdsourcing?



maximum spatial coverage

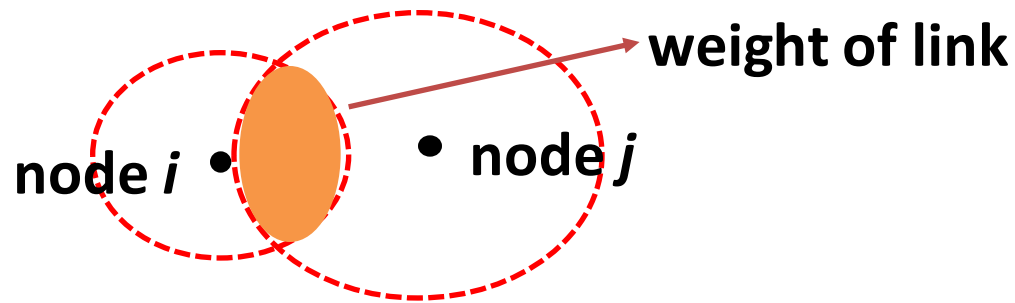
Sampling model



Sampling is repeated after time interval T

Receivers selection overview

- Construct weighted graph $G = (V, E, W)$



- Define new metric - degree expansion

$$DX(S) = \sum_{i \in N(S)} \min_{j \in S} w_{i,j}$$

- Minimum overlap a set with its neighbors

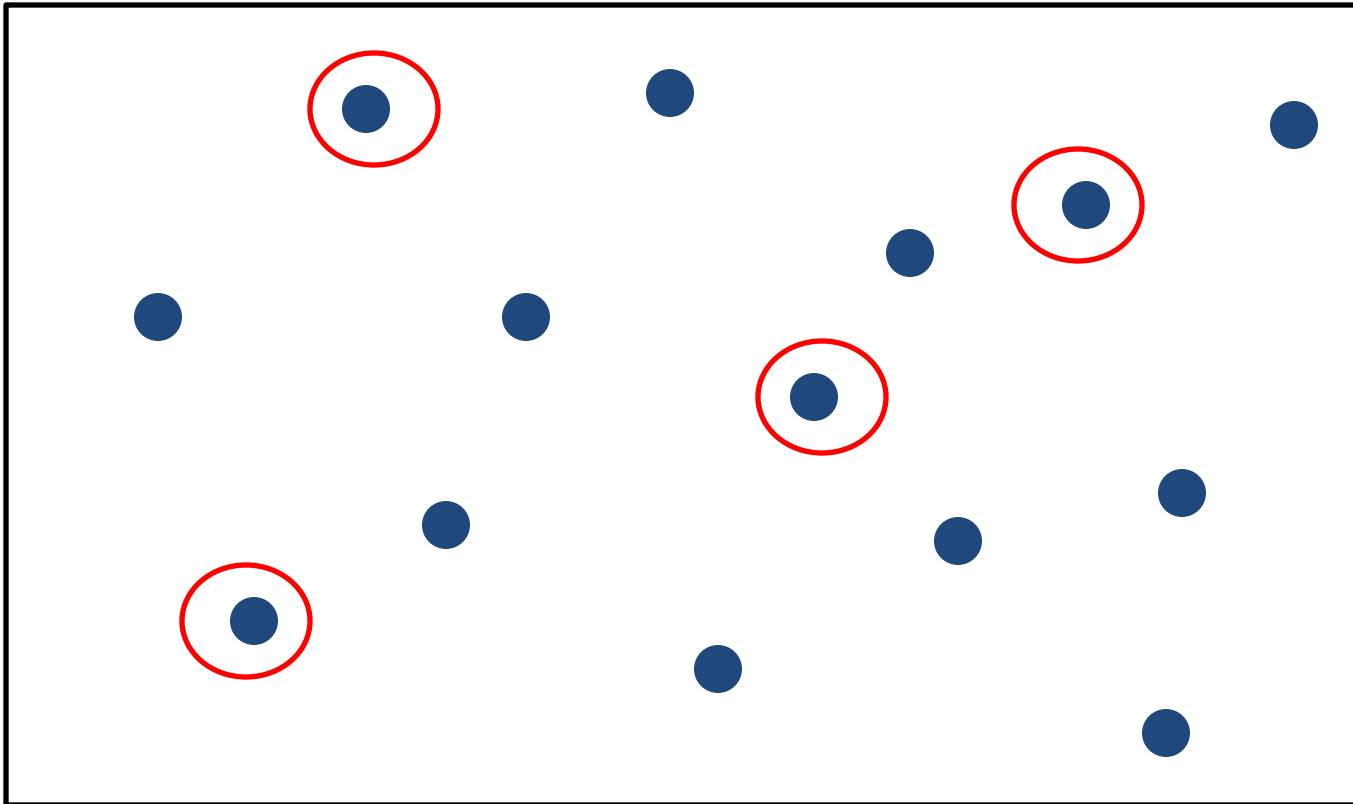
neighborhood of a set

Maximizing coverage

➤ For maximizing coverage solve: $\arg \max_{S \subseteq V} DX(S)$

NP-hard problem

Receiver selection by Greedy algorithm



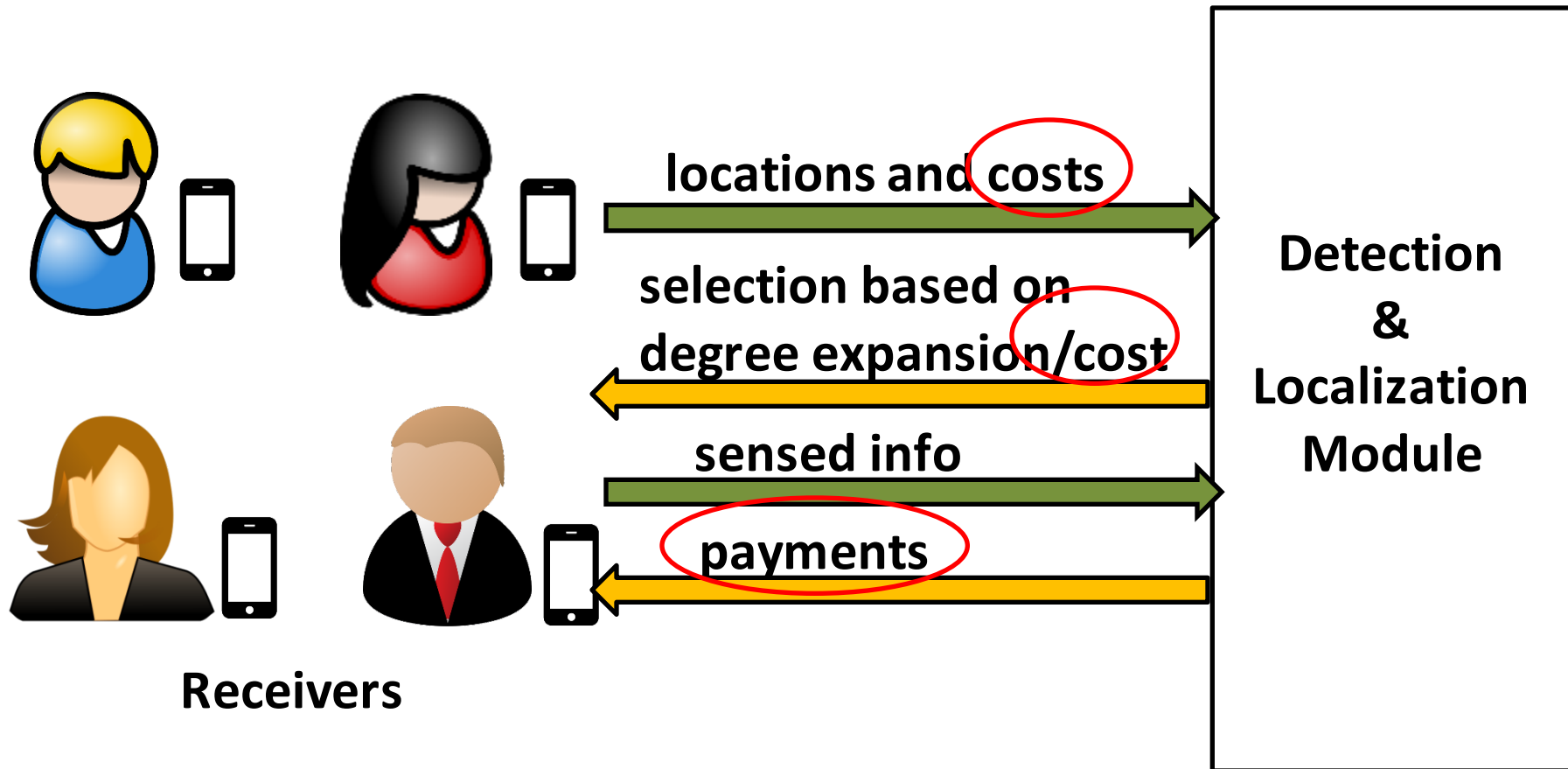
Greedy heuristic, selects RXs based on marginal contributions m_i to degree expansion

- Selected receivers
- Available receivers

How to incentivize participation?



Sampling and incentive model



Sampling is repeated after time interval T

Truthful receiver selection

- Need algorithm for selection that motivates truthful reporting of cost

Truthful greedy receiver selection

- Consider both cost, marginal degree expansion

$$\max_{S \subseteq V} \sum_{i \in S} \frac{m_i}{c_i}$$

- Incentive compatibility (truthfulness) – motivates truthful reporting of cost

utility when acting truthfully $p_i(c_i) - c_i \geq p_i(c'_i) - c_i \quad \forall i, c'_i$ declared sensing cost

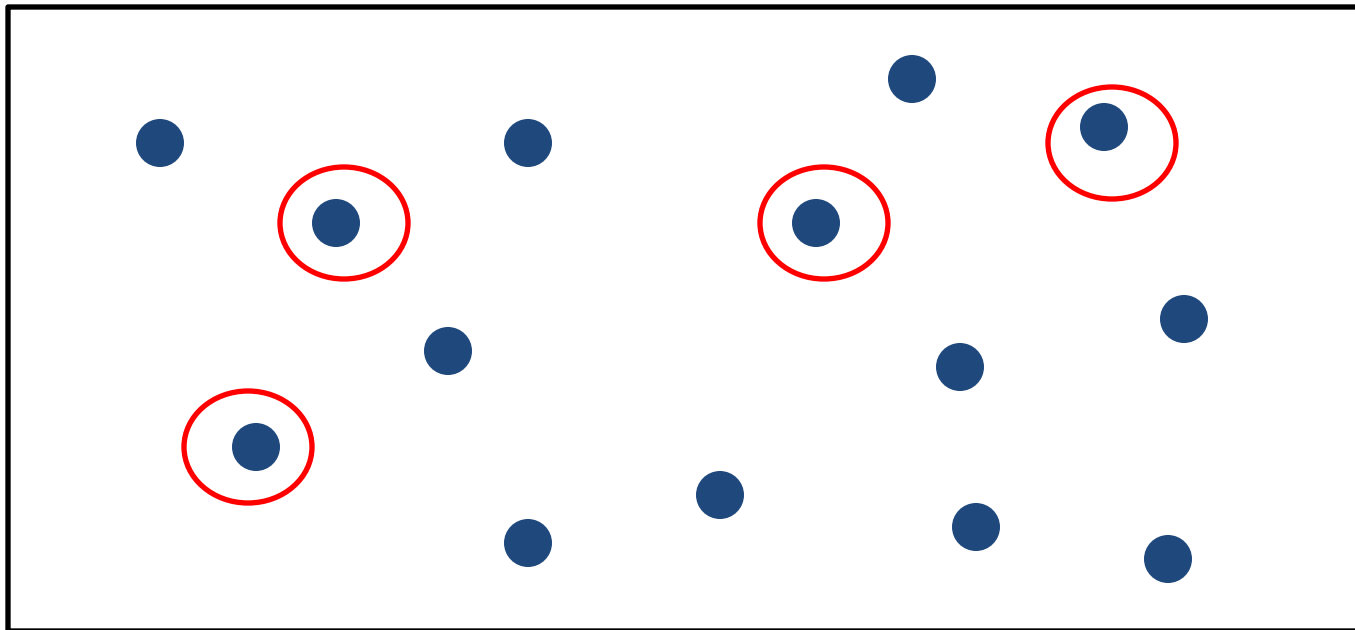
actual sensing cost

- Individual rationality (provides incentive to participate) - reward amount greater than cost

payment for i $p_i(c_i) - c_i \geq 0 \quad \forall i$

Truthful greedy receiver selection

- **Sampling** - Greedy algorithm that minimizes cost and maximizes marginal degree expansion



Selects RXs
based on
 $\frac{m_i}{c_i}$

- Selected receivers
- Available receivers

Truthful greedy receiver selection

- **Payment** – maximum value for cost that a node can declare and still win

Extensions to truthful greedy receiver selection

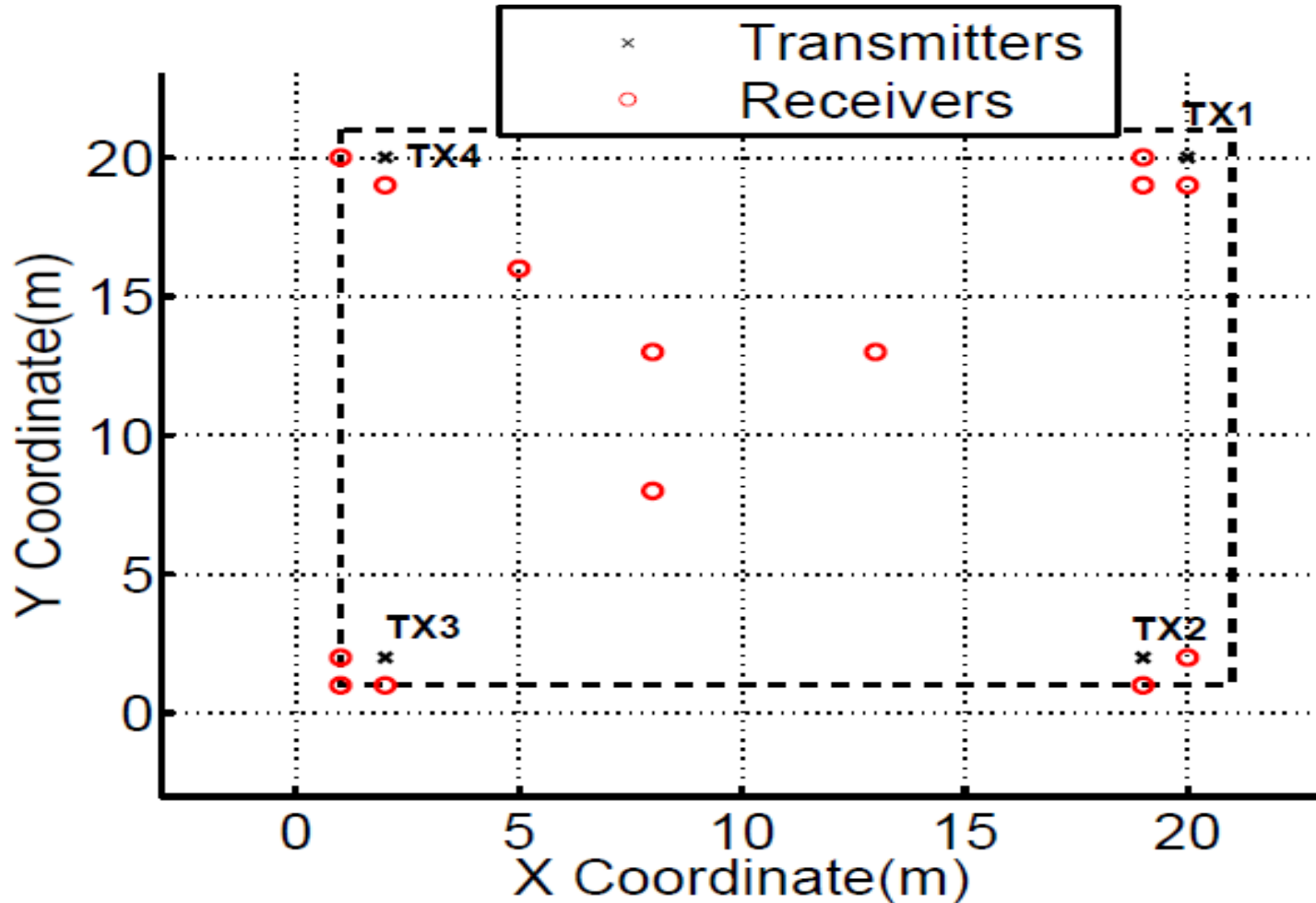
- Budget-feasible truthful greedy
 - Given budget B , maximize $DX(S)$
- Mobility aware budget-feasible truthful greedy
 - Maximize $DX(S)$ under budget B , also available during T
- Incentive compatibility (truthfulness)
- Individual rationality (reward greater than cost)

Evaluation of SPLOT

Evaluation metric

- **Localization error** - RMS error between estimated, actual transmitter location
- **Cardinality error** - fraction of time when number of estimated transmitter incorrect
- **Optimal Sub-Pattern Assignment (OSPA) metric**
 - penalizes error in number of estimated TX
 - localization error, cardinality error combined

Setup 1 : Orbit test bed



- Open environment, no obstructions and USRP2 nodes used

Setup 1: Localization error for single transmitter

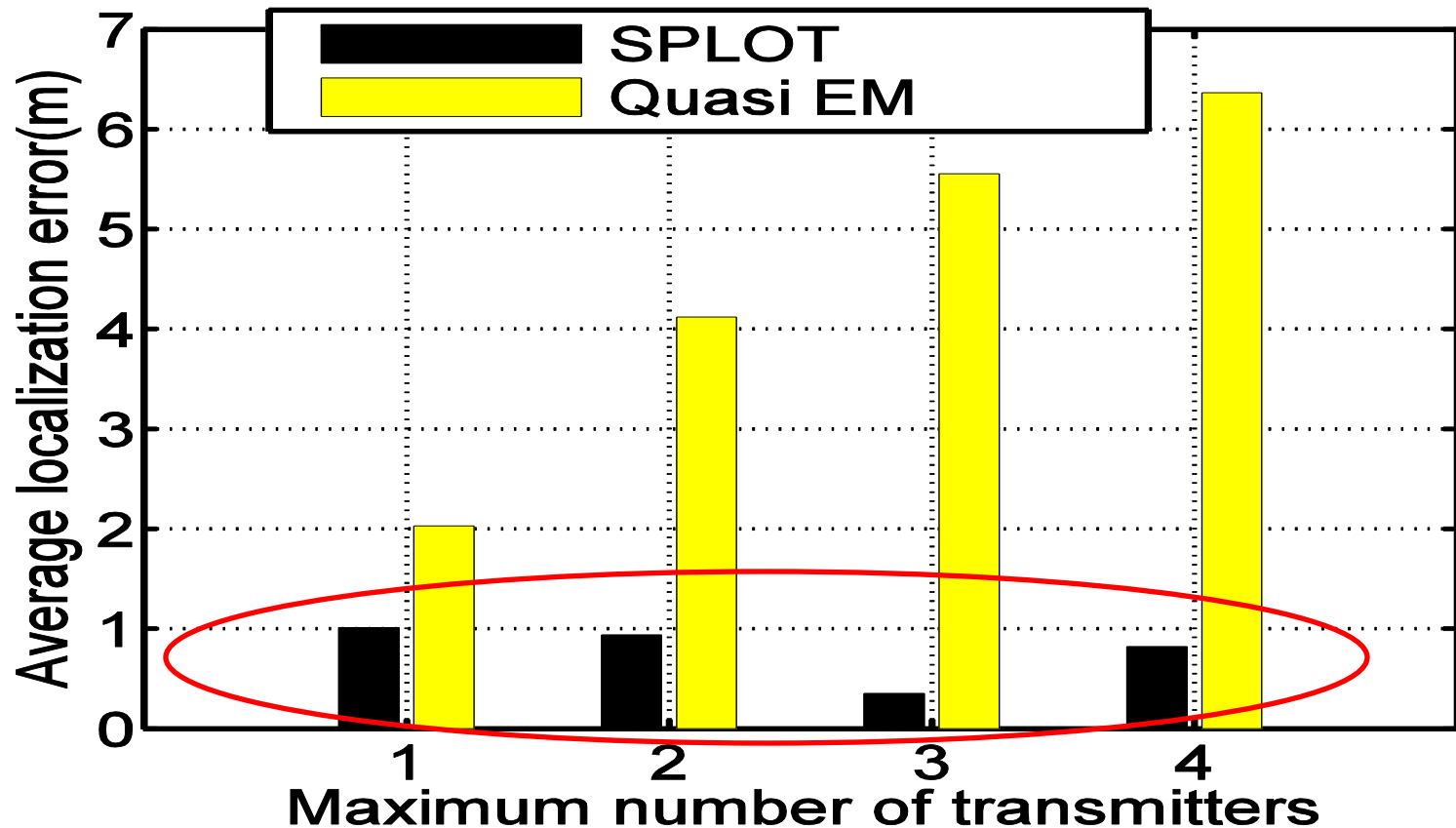
Transmitter	Localization error(m)	
	Matrix Inversion	MLE approach
TX1	1.41	1.01
TX2	0.5	1.03
TX3	1.11	0.97
TX4	1.41	5.08

average localization error **1.11m** and variance **0.18m**

average localization error **2.02m** and variance **4.1m**

matrix Inversion is better

Setup 1: Localization error for multiple transmitter



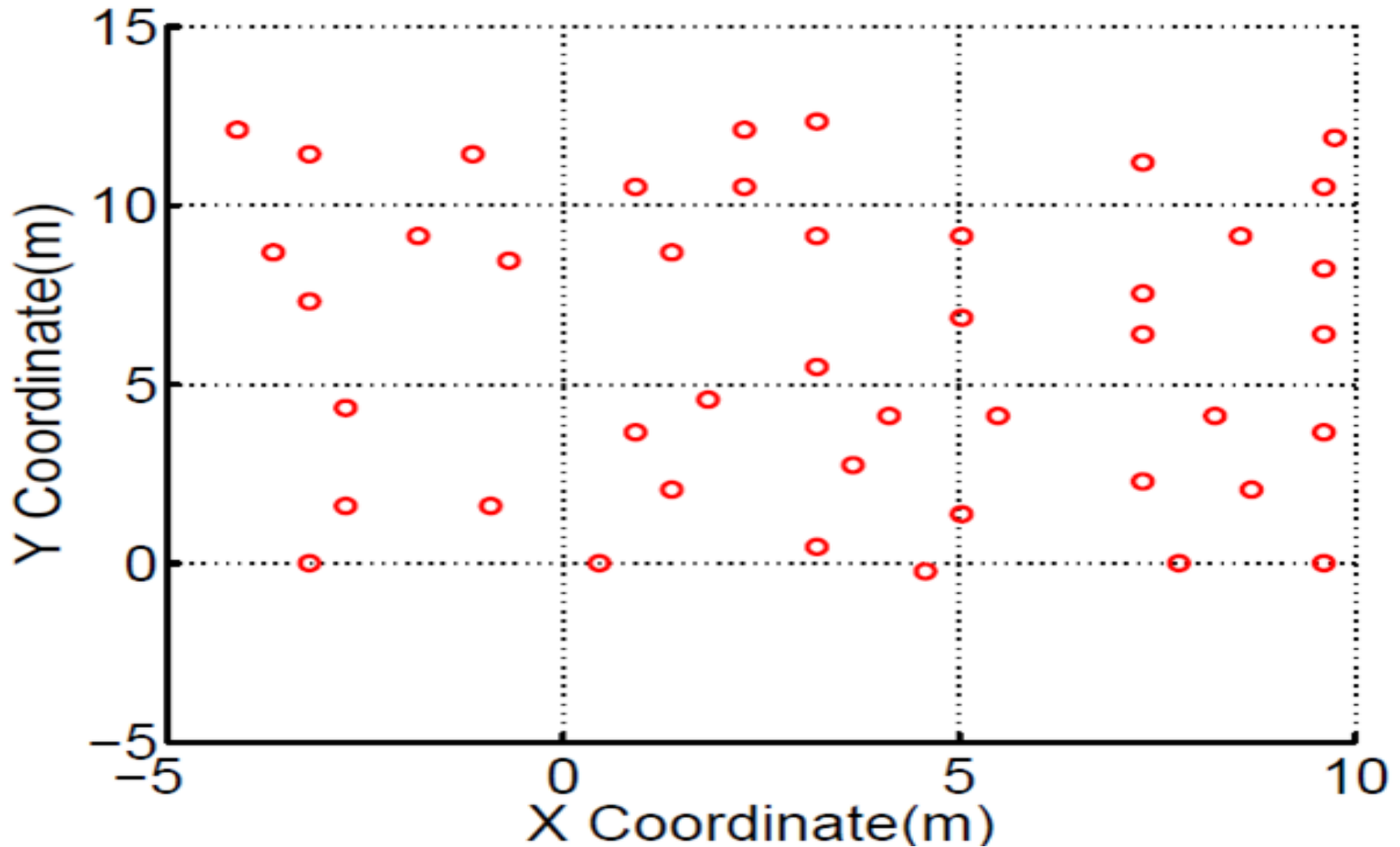
- SPLIT is independent of number of transmitter
- Unlike Quasi EM, SPLIT doesn't need number of transmitter

Setup 1: Cardinality error and OSPA metric

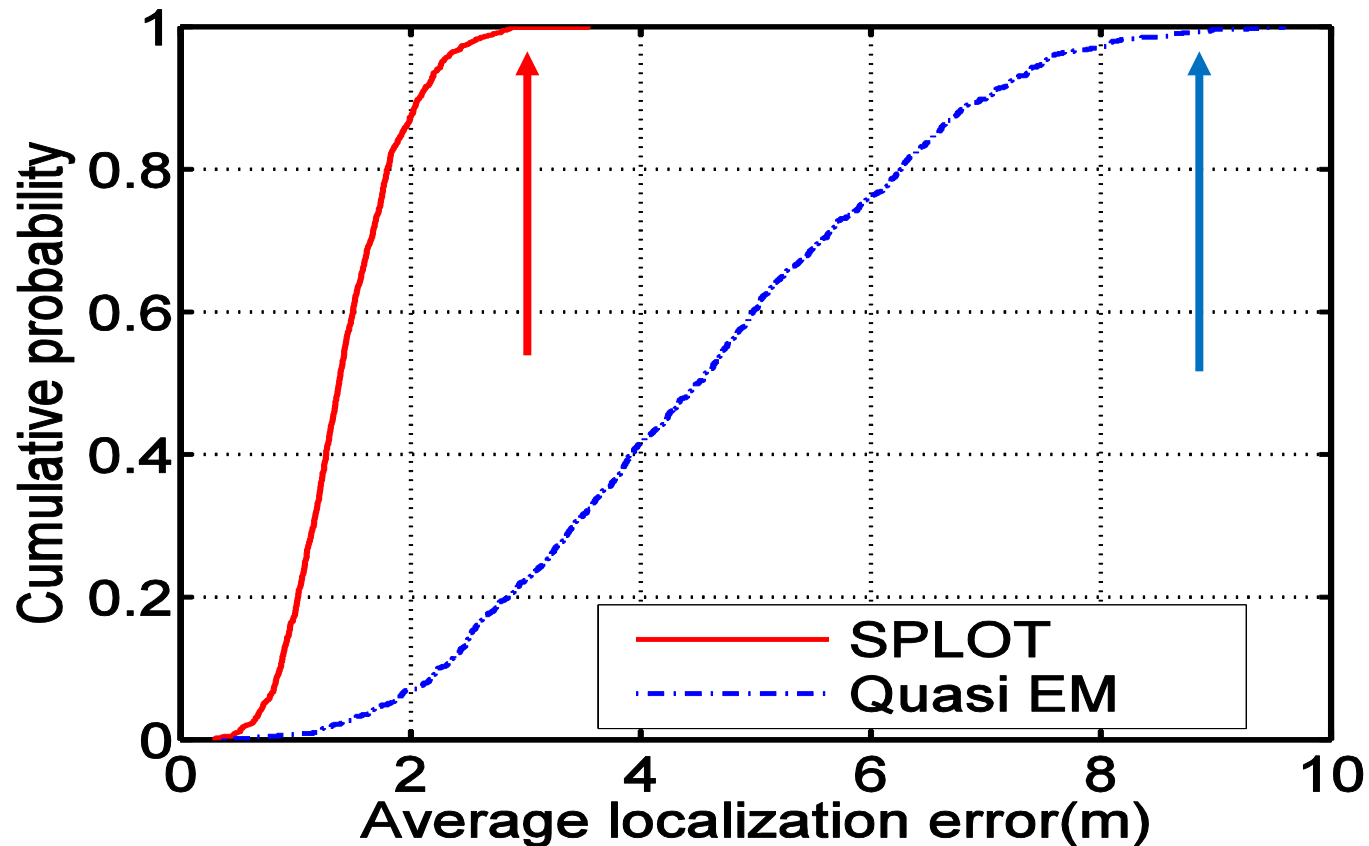
Maximum number of transmitters	OSPA metric			Cardinality error
	$g=1$ (m)	$g=2$ (m)	$g=5$ (m)	
1	0.79	1.01	1.01	0
2	0.91	1.17	1.17	0
3	1.04	1.36	1.45	0.05
4	1.14	1.51	1.79	0.14

- Cardinality error increase with number of TX
- OSPA metric increases slightly for high penalty

Setup 2 : Cluttered office



Setup 2: Localization error for 2 transmitters



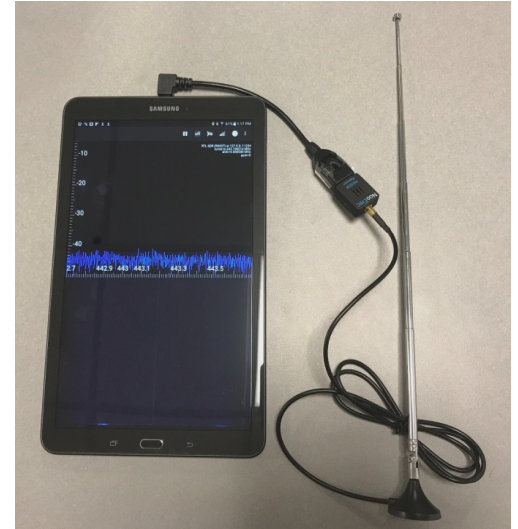
Average localization error of SPLIT significantly less than Quasi EM

Implementations with mobile transmitters and receivers

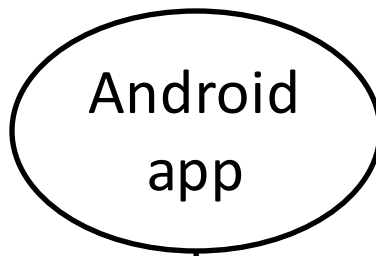
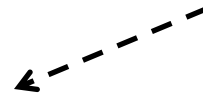
Implementation Equipments



BaoFeng BF-F8HP
Power 1W

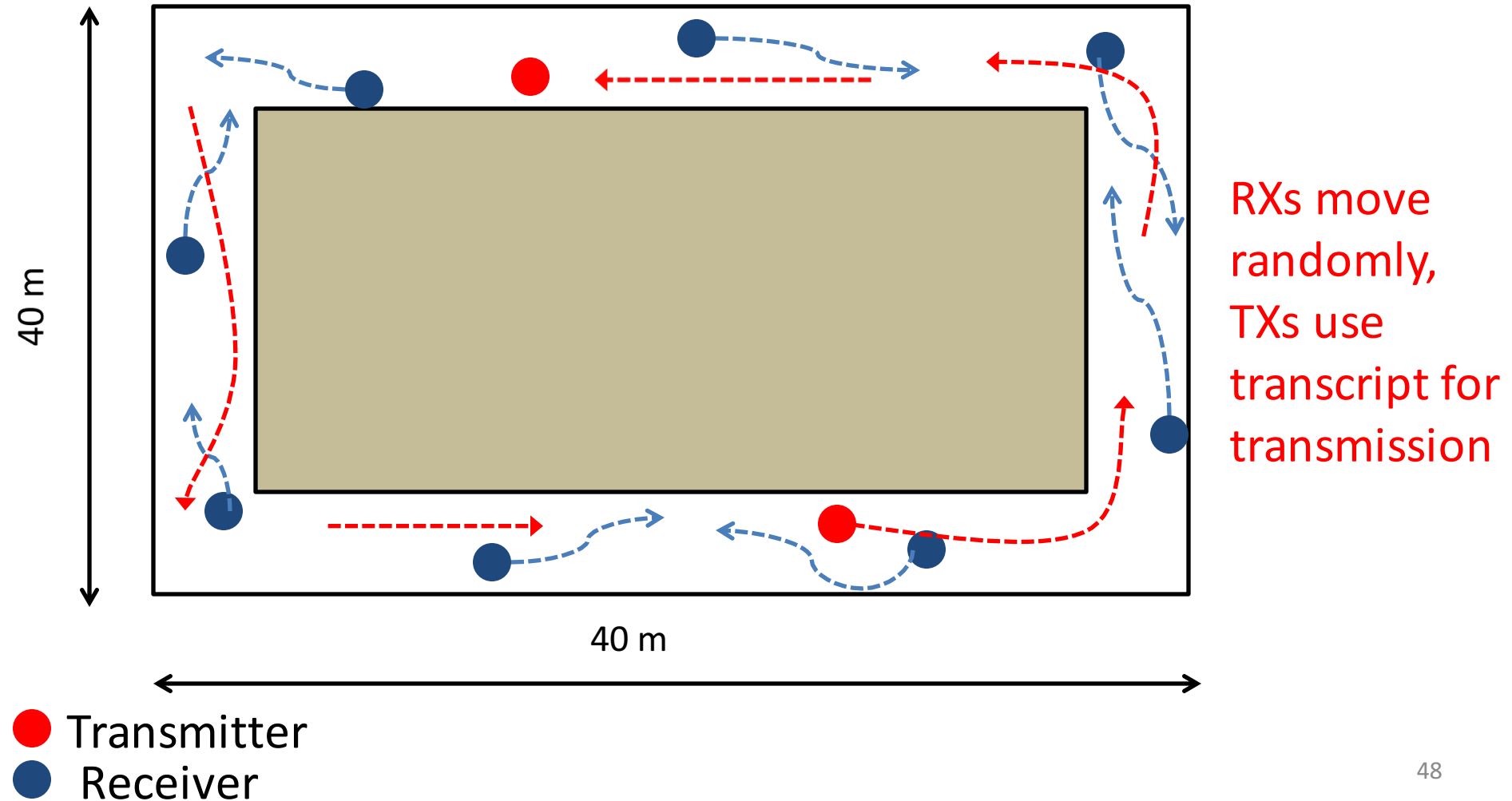


Realtek dongle
(RTL-SDR) with
USB cable

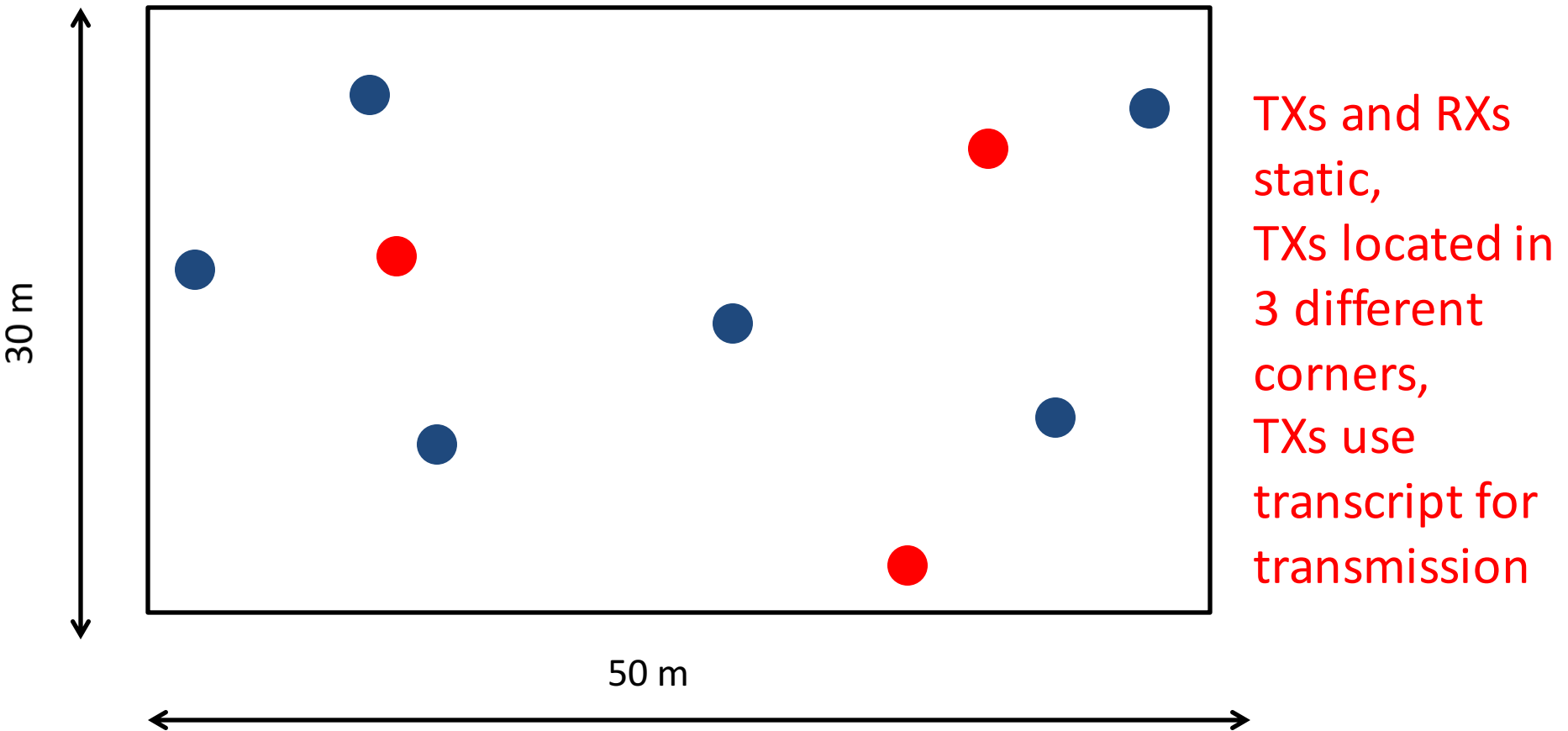


<time, location, RSS>

Experiment B (Environment 1) (Indoor, university building)

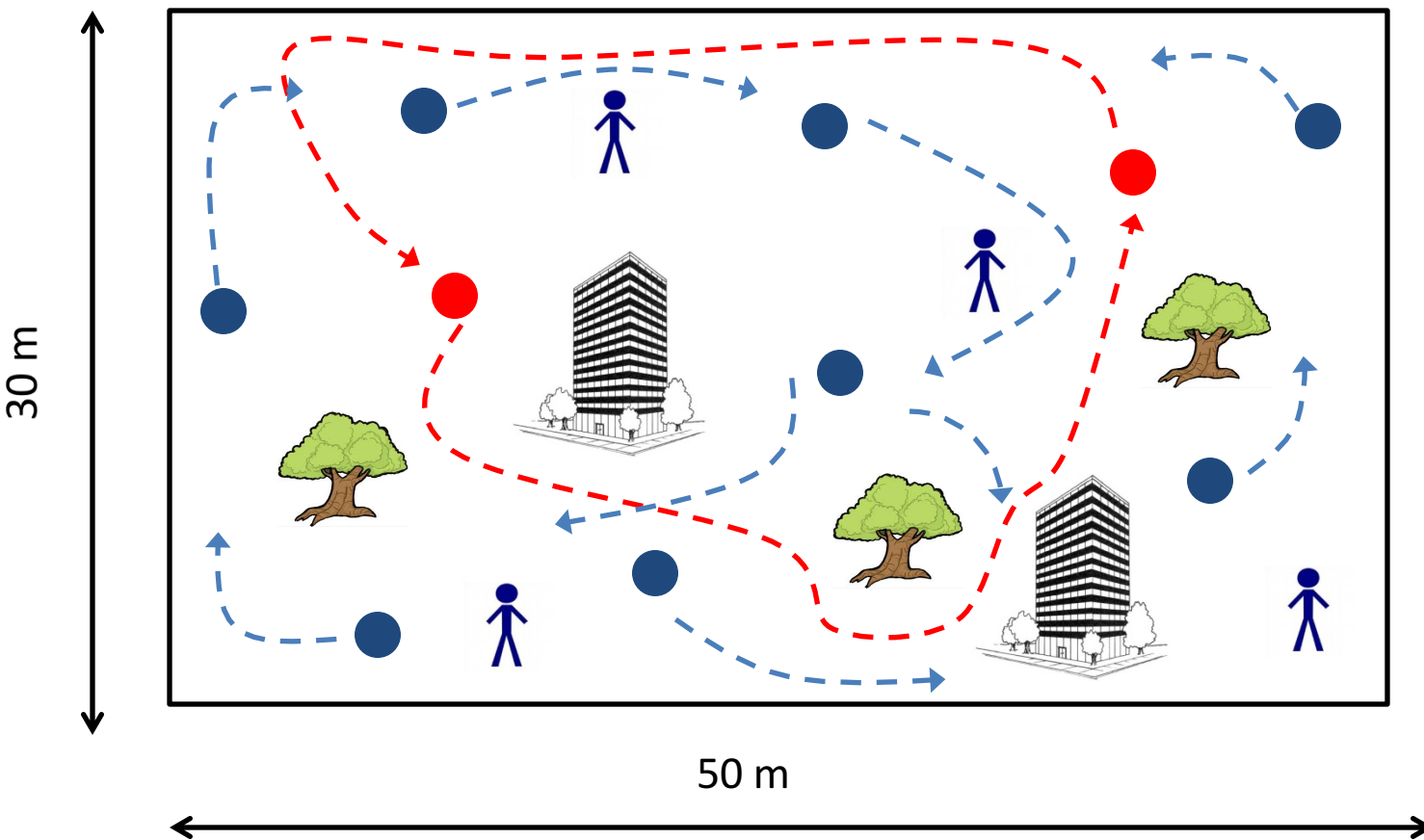


Experiment D (Environment 2) (Indoor, university food court)



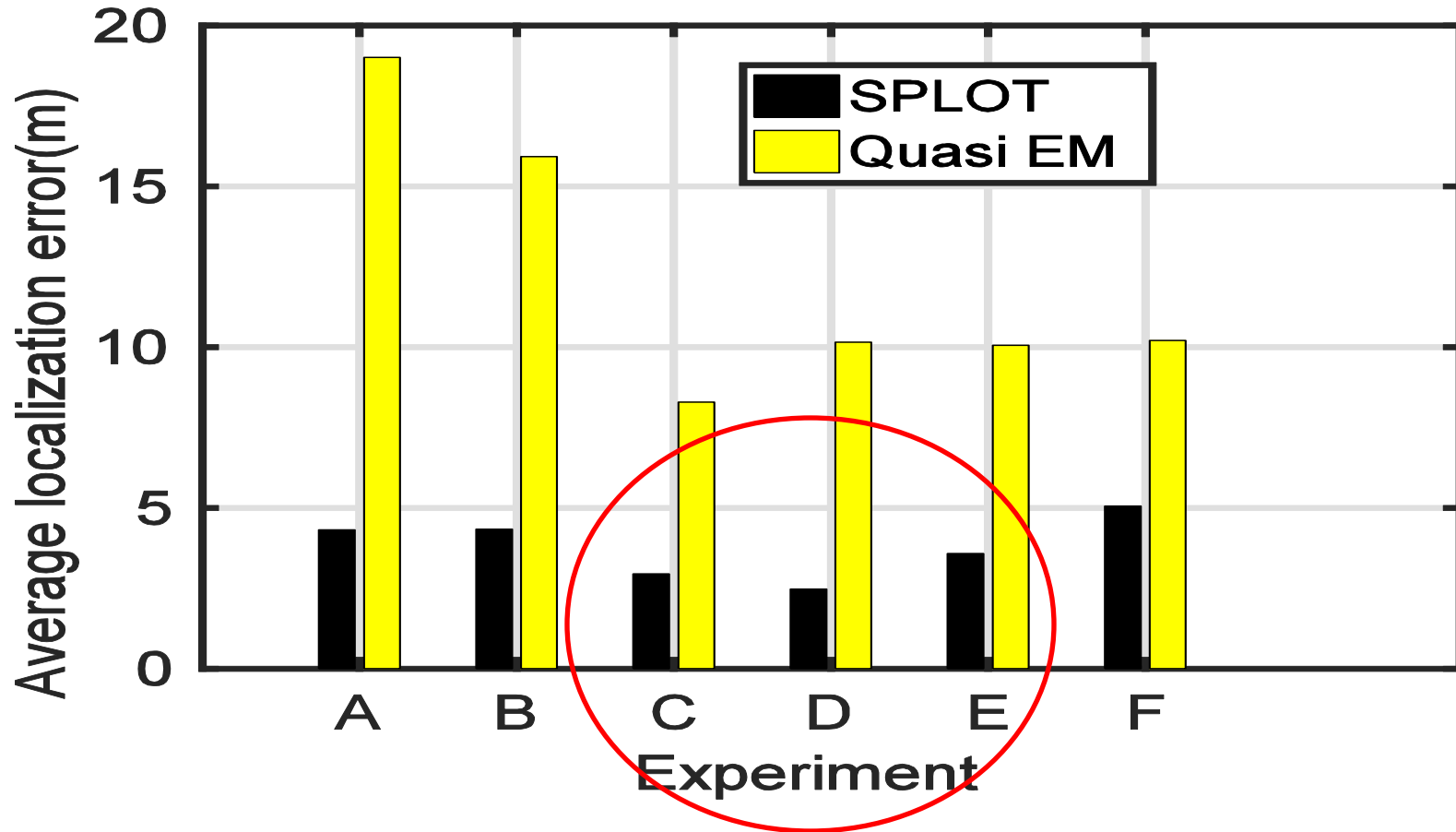
- Transmitter
- Receiver

Experiment F (Environment 3) (Outdoor, university campus)



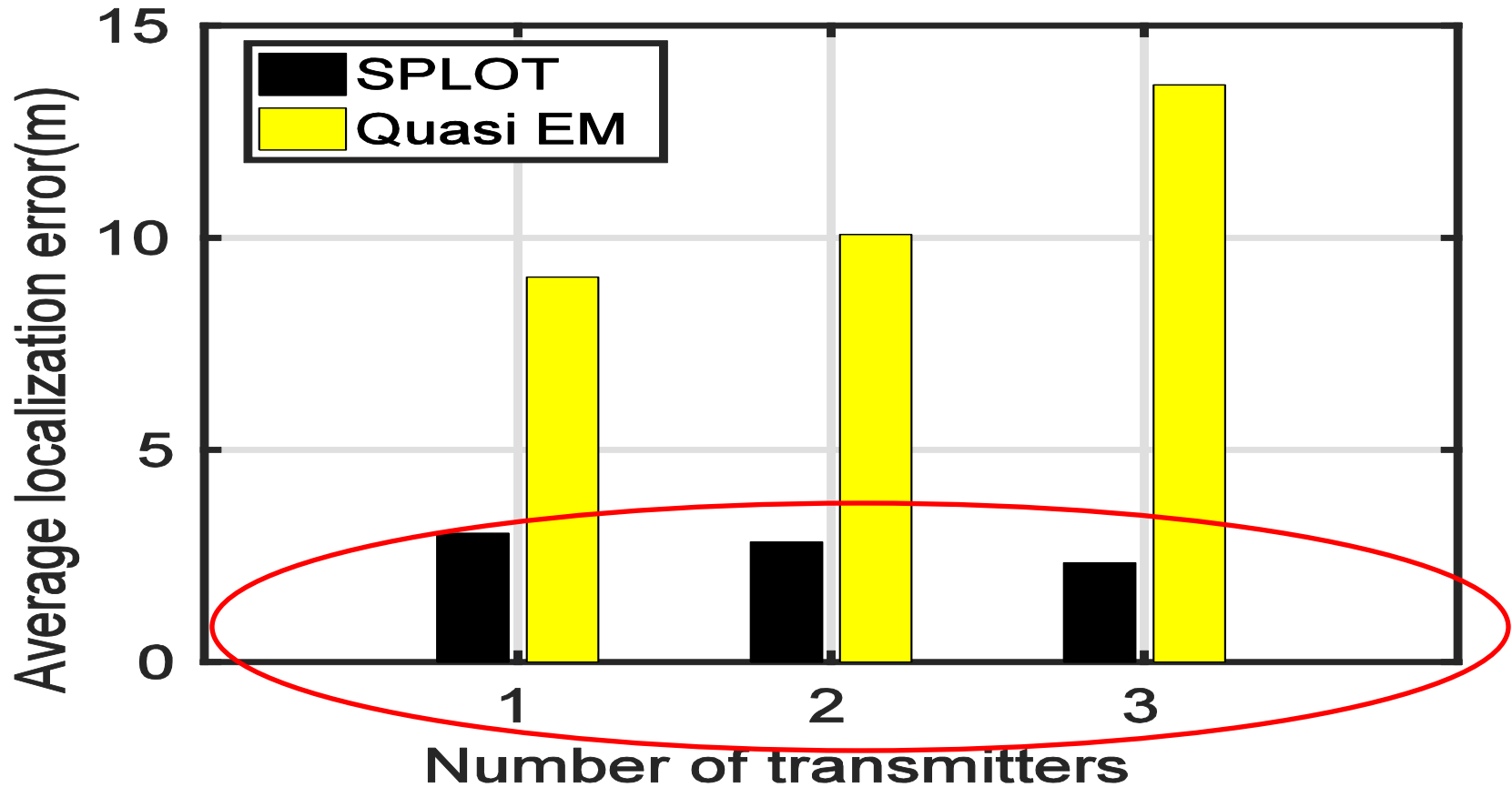
TXs, RXs move at normal walking speed for 7 minutes, TXs transmit continuously

Implementation results: Localization error



Decrease in localization error C, D, E (static)

Experiment D: Impact of number of transmitters



SPLIT is independent of number of transmitters

Conclusion

- **SPLOT** - simultaneous localization of multiple transmitters using crowdsourcing
- **Receiver selection** – selecting set of mobile sensing devices
- **Incentives** – for participating mobile sensing devices
- **Evaluation** – using multiple experiments

Advantages

- Using crowdsourcing to solve **multiple** transmitters localization problem
- Considering mobility, truthfulness and coverage problem all together
- Using simple yet efficient and accurate method to solve real-time problem

Disadvantage

- Hard to guarantee numbers of receivers
- Hard to solve when two transmitters are very close from each other
- The number of transmitters tested is not that convincing
- How to guarantee they report range truth?

Thank You
Questions?