# Localization of RFid tags by means of an antenna array: influence of the array shape

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#### **Goal of the project: Localization of RFid tags inside a warehouse**

- System principle: Estimation of the direction of arrival and intensity of the tag response with an antenna array.
- Main difficulties to overcome:
  - Multiple propagation path
  - Low signal to noise ratio of the received signal ( $_{SNR \simeq 0 \text{ dB}}$ )

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– Limited size of the array for cost reason





Figure 5 : Evolution of the theoretical and experimental mean square error and Cramer Rao Bound with the signal to noise ratio for the direction of arrival of one signal and 16 snapshots

• For a given array configuration, the estimation error varies with the direction of arrival of the incident signal. For example, a square array with 9 antennas regularly spaced exhibits



Figure 1 : RFTag – Localisation of RFid tags in warehouse

# Direction of arrival estimation by a maximum likelihood based algorithm

#### 1. Propagation model.

The presence of multipath components is simulated by delayed version of the signal, coming from different directions. Supposing that the tag response is a narrowband signal, the propagation model is given by:

$$x_{k}(t) = \frac{1}{d_{0}}g_{k}(\theta_{0},\varphi_{0}) \cdot e^{j\frac{\omega_{c}}{c}(x_{k}\cos\theta_{0}\cos\varphi_{0}+x_{k}\sin\theta_{0}\cos\varphi_{0})}s_{0}(t) + \sum_{i=1}^{N_{R}}\frac{1}{d_{i}}\alpha_{i}g_{k}(\theta_{i},\varphi_{i}) \cdot e^{j\frac{\omega_{c}}{c}(x_{k}\cos\theta_{i}\cos\varphi_{i}+x_{k}\sin\theta_{i}\cos\varphi_{i})}s_{i}(t) + n_{k}(t)$$
  
This model can be rewritten as :  $\mathbf{x}(t) = \mathbf{A}(\theta,\varphi)\mathbf{s}(t) + \mathbf{n}(t)$ 

## 2. Principle of the maximum likelihood estimation scheme

→ To seek the parameters  $(\theta, \varphi)$  maximizing the likelihood of the measured data according to the propagation model used:

$$p\left(\boldsymbol{x}(t) \middle/ \boldsymbol{\theta}, \boldsymbol{\varphi}, \boldsymbol{\sigma}^{2}, \boldsymbol{s}(t)\right) = \frac{1}{\pi^{L}} \cdot \frac{1}{(\boldsymbol{\sigma}^{2})^{L}} \cdot \exp\left(\left\|\boldsymbol{x}(t) - \boldsymbol{A}(\boldsymbol{\theta}, \boldsymbol{\varphi})\boldsymbol{s}(t)\right\|^{2} \middle/ \boldsymbol{\sigma}^{2}\right)$$

Under the hypothesis of independent snapshots, the function the estimation of the azimuth  $\theta$  and elevation  $\varphi$  angle are obtained by minimizing the function:

$$\min_{\boldsymbol{\theta},\boldsymbol{\varphi},\boldsymbol{\sigma}^{2},\boldsymbol{s}(t)} l_{DML}\left(\boldsymbol{\theta},\boldsymbol{\varphi},\boldsymbol{\sigma}^{2},\boldsymbol{s}(t)\right) = \min_{\boldsymbol{\theta},\boldsymbol{\varphi},\boldsymbol{\sigma}^{2},\boldsymbol{s}(t)} \left\{ L\log \boldsymbol{\sigma}^{2} + \frac{1}{\boldsymbol{\sigma}^{2} \cdot N} \sum_{i=1}^{N} \left\| \boldsymbol{x}(t) - \boldsymbol{A}(\boldsymbol{\theta},\boldsymbol{\varphi})\boldsymbol{s}(t) \right\|^{2} \right\}$$

## **Optimization of the antenna array shape**

1. Antenna array = spatial sampling

performance degradation for very oblique signals. Less regular structures (like minimum redundancy array) or "Y" shape arrays can prevent such problem



Figure 6 : Evolution of the MSE error on  $\theta$  for 9 antennas square array

Figure 7 : Evolution of the MSE error  $\theta$ on for 7 antennas "Y" array

3. Main characteristics used to compare array's

Two characteristics were used to compare array properties:

- Slope of the Cramer Rao Bound (representative of the estimation error at high SNR) (Figure 8)
- SNR threshold when using the DML algorithm (Figure 9)

As these characteristics vary with the direction of arrival, mean value on  $\theta$  and  $\phi$  were calculated and compared for different shape of array.

## Calculation of the Cramer Rao Bound

The Cramer Rao Bound on angle accuracy can be calculated from [2]:

$$E\left[\left(\hat{\theta}-\theta\right)\left(\hat{\theta}-\theta\right)^{T}\right] = \mathbf{R}_{\varepsilon} \ge \left[\frac{2N \cdot L \cdot SNR^{2}}{\left(1+L \cdot SNR\right)}\right]^{-1} \cdot \left[\left[\frac{\partial \mathbf{a}^{H}}{\partial \theta}\frac{\partial \mathbf{a}}{\partial \theta} - \frac{1}{L}\frac{\partial \mathbf{a}^{H}}{\partial \theta}\mathbf{a}\mathbf{a}^{H}\frac{\partial \mathbf{a}}{\partial \theta}\right]\right]$$

As the second factor of the right expression regroups all dependencies from antennas position, it is used as performance indicator for a particular array.

 $\rightarrow$  The shape of the array has a dramatic influence on the estimation performance.

 $\rightarrow$  The larger array, the higher accuracy

BUT

 $\rightarrow$  The cost of the array and the demodulation system is proportional to the number of antennas

→ If <u>all</u> antenna are spaced with <u>more than a half-wavelength</u>, grating lobes appears and ambiguities occurs in the estimation process (Nyquist criteria for spatial sampling):



Figure 2 : Array pattern of an antenna array where all antennas are spaced by a half wavelength Figure 3 : Array pattern of an antenna array where all antennas are spaced by a wavelength

Under these constraints, numerous array shapes possible: antennas can be regularly spaced by  $\frac{\lambda}{2}$  or irregularly disposed:



#### Estimation of the SNR threshold

A theoretical expression of the mean square error for likelihood estimation [1] is used to estimate the SNR threshold. In this expression, contributions from outliers and from local error are separately evaluated:

$$E\left[\left(\hat{\theta}-\theta\right)^{2}\right] = P\left[no \ outlier\right]E\left[\left(\hat{\theta}-\theta\right)^{2} \mid no \ outlier\right] + P\left[outlier\right]E\left[\left(\hat{\theta}-\theta\right)^{2} \mid outlier\right]$$

The SNR threshold is then defined as the signal to noise value at which the mean square error exceeds the Cramer Lower Bound by 10%.



Figure 8 : Evolution of the slope of the Cramer Rao Bound on  $\theta$  for different shape of array



**Figure 4 : Some examples of antenna array shape** 

 $\rightarrow$  Irregular arrays have a greater aperture but the array pattern presents higher sidelobes which can decrease the estimation accuracy at low SNR.

So, which one provides the best results for our problem?

- 2. General remarks on antenna disposition:
- Evolution of the estimation error with signal to noise ratio:
  - The DML algorithm for DOA estimation is an efficient estimation algorithm. For a number of snapshots sufficiently high, the estimation error attains the Cramer Lower Bound which is a theoretical lower bound on the variance estimation of any estimator.
  - However, below a certain signal to noise ratio, the error on the direction of arrival rises sharply: secondary lobes of the array pattern generates incorrect global maximum in the likelihood function. As a result, the antenna array system cannot be used below this SNR threshold.

Figure 9 : Evolution the array SNR threshold on  $\theta$  with the number of antenna for different shape of the array

#### **Conclusion**:

- For a given number of antennas, irregular arrays provide better accuracy at high signal to noise ratio.
- For a given number of antennas, the SNR threshold is generally lower for regular arrays than for irregular arrays.
- Some regular arrays like square and "L" shape arrays exhibit performance deterioration for very oblique signals. Simulations have shown that the use of "Y" arrays limits this problem.
  As a result, a four antennas "Y" shape array has been selected for a first prototype.

#### Bibliography

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