

# RF Environment behavior modeling based on 3-D Ray-Tracing and Neural Networks to mitigate multipath in indoor position estimation

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## **Part I: Problem Definition**

## **Part II: Proposed methodology for high accuracy position estimation**

2.1 Position Estimation Methods.

2.2 3-D Ray Tracing and Neural Networks Channel Model.

2.3 Integration of the position estimation solution for error minimization in TDoA.

## **Part III: Implementation and experimental results**

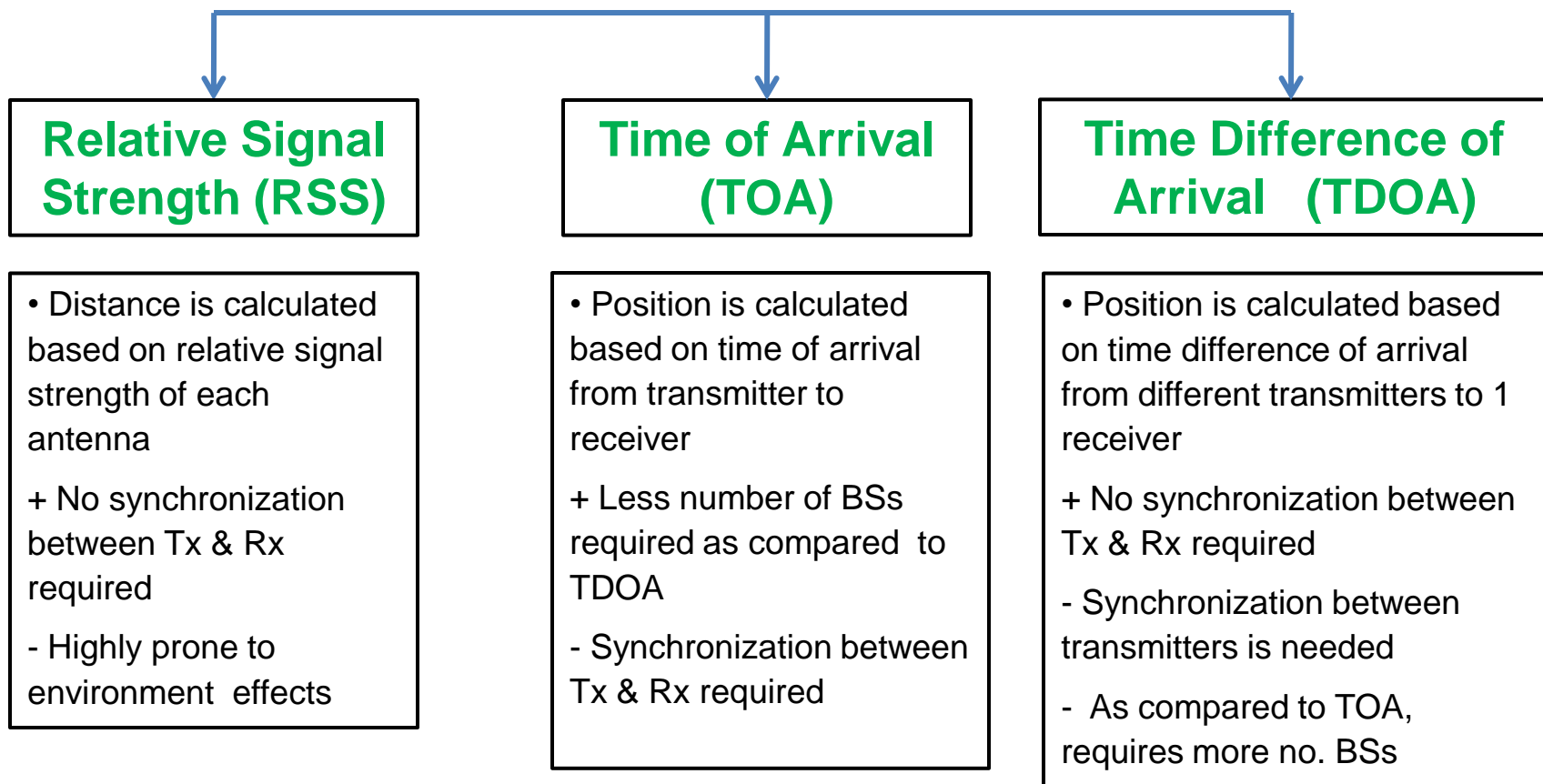
3.1 Measurement set-up.

3.2 Experimental Results.

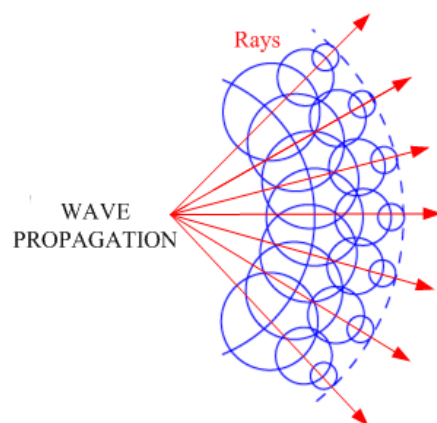
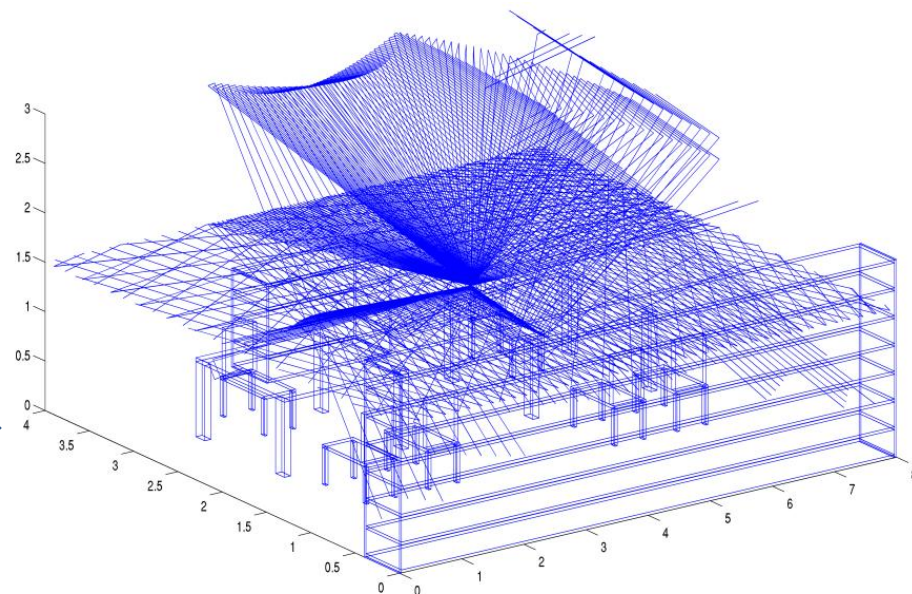
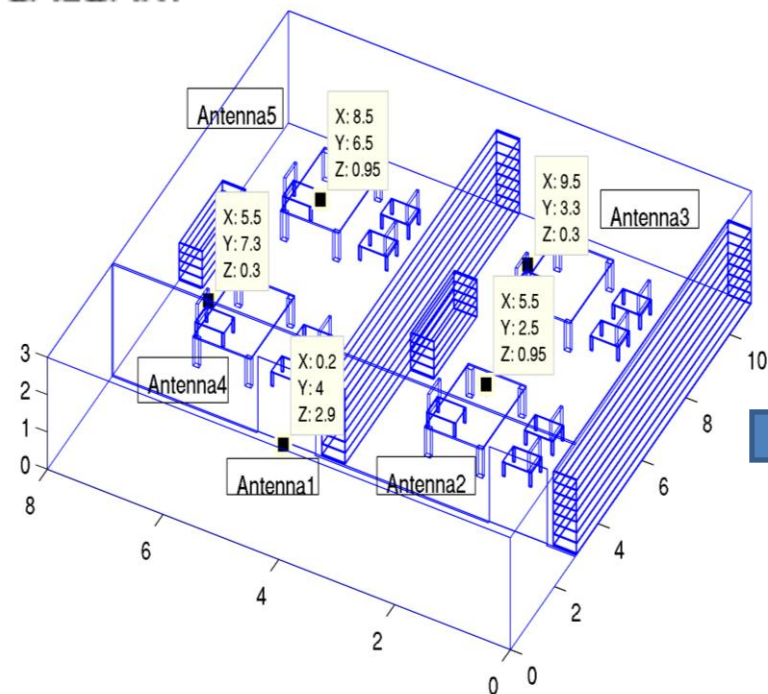
## **Part IV: 4. CONCLUSION**

- Highly accurate position estimation is very desirable in many applications.
- RF environment for communication are often characterized with strong multipath behavior.
- There is a need for characterizing the multipath behavior of the channel to mitigate this multipath effect.
- The empirical techniques offer low computational cost, but also low accuracy.
- Deterministic techniques, which are based on the calculation of electromagnetic field, offer high accuracy at the expense of very high computational complexity.

## Location Estimation



# 3-D Ray Tracing (1/2)



## Information Extraction

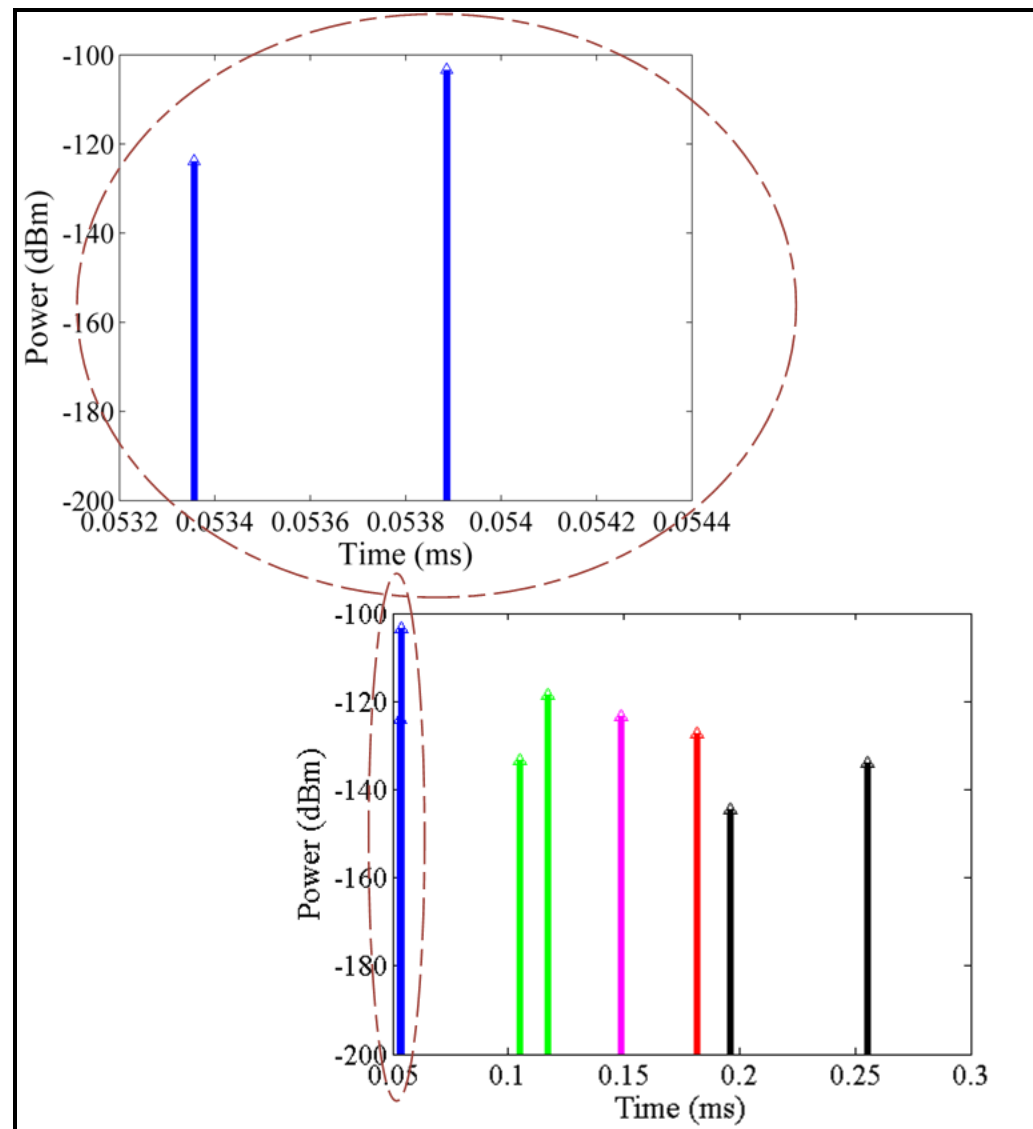
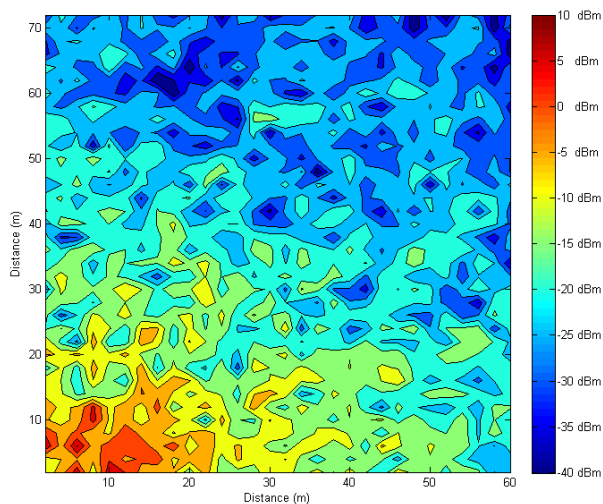
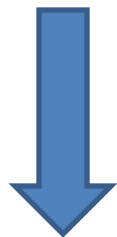
- Electric field created by transmitter
- Electric field at receiver antenna
- Received power
- Power Delay Profile
- Medium Delay
- Dispersion

# 3-D Ray Tracing (2/2)

**Power Delay Profiles**



**Power Maps**



# Neural networks

## *Why?*

- To lessen the burden of storing the data and computation for 3-D Ray-Tracing.
- To predict for the positions not visited by rays in Ray-Tracing.
- To adapt to the change in environment while learning from Ray-Tracing simulations.



# Neural Network Architecture

\*Two Hidden layer feedforward  
Neural network.

\*First Layer :

Activation function:  $\frac{1}{1-e^{-x}}$   
No. Of neurons: 4

\*Second Layer:

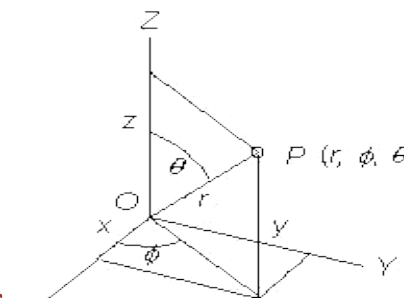
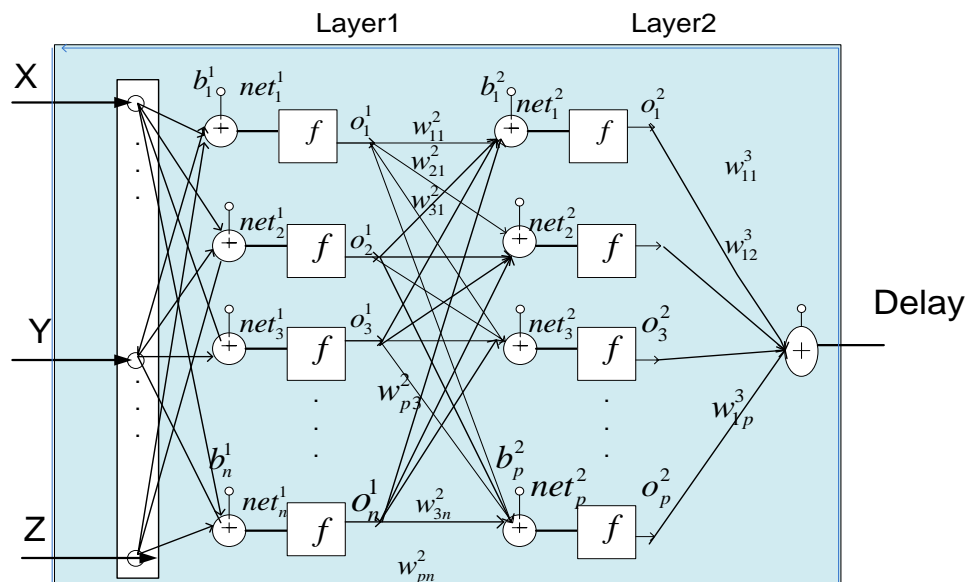
Activation function:  $\frac{e^x - e^{-x}}{e^x + e^{-x}}$   
No. Of neurons: 17

\*Inputs: Co-ordinates of point  
in the space.

\*Output: Delay error due to multipath.

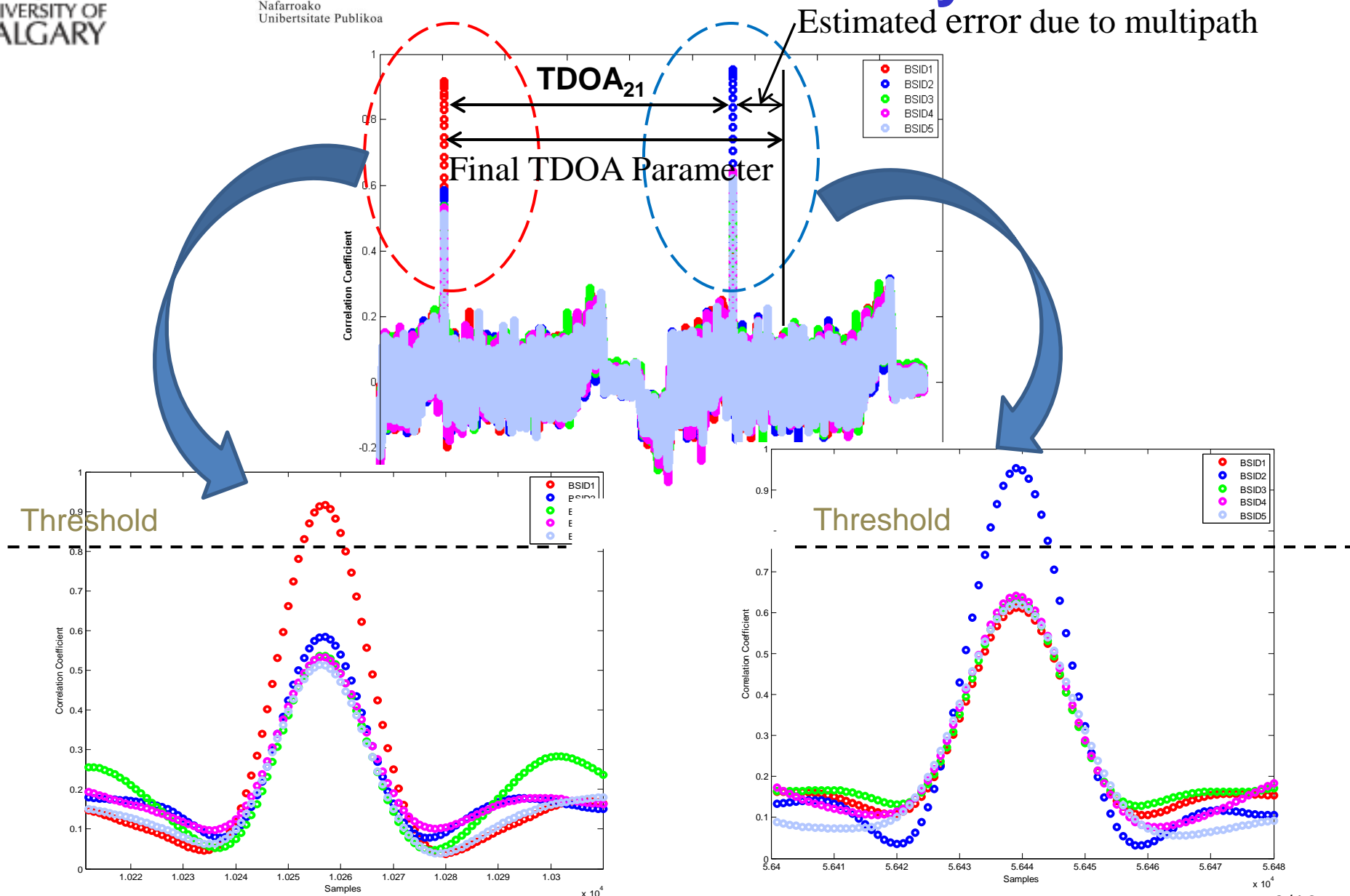
\*NN training uses Levenberg Marquardt Back  
Propagation algorithm for mean square error  
minimization.

\*Normalized mean square error (NMSE) of 0.018nS is  
achieved while trained network is used with another set of data  
from the same scenario, where maximum error is of 0.02nS.

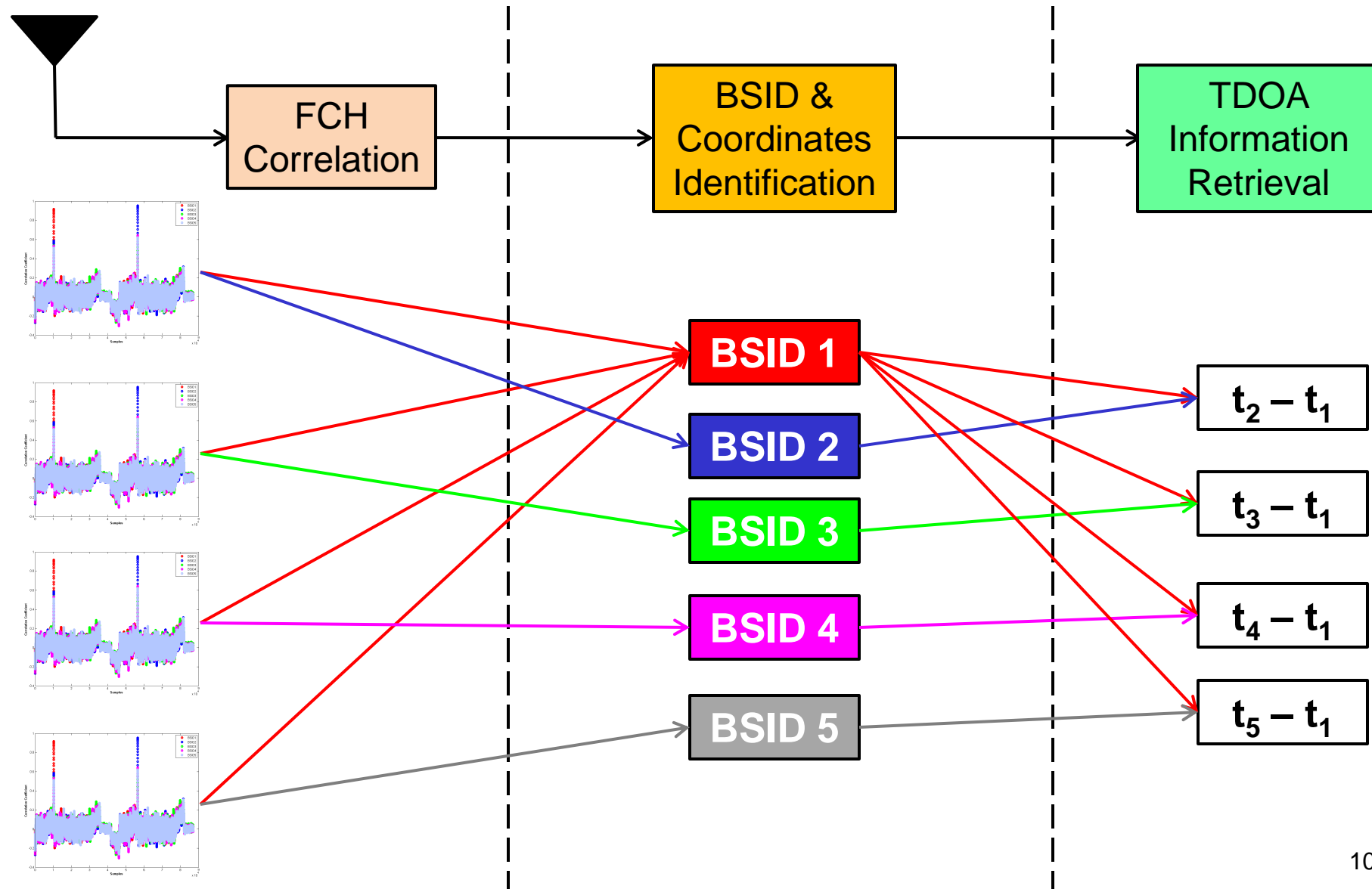




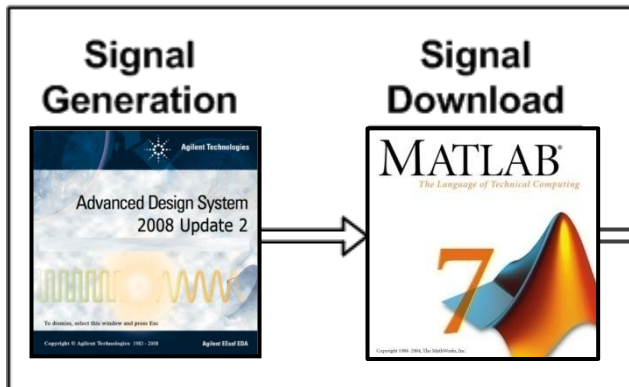
# Waveform Analysis



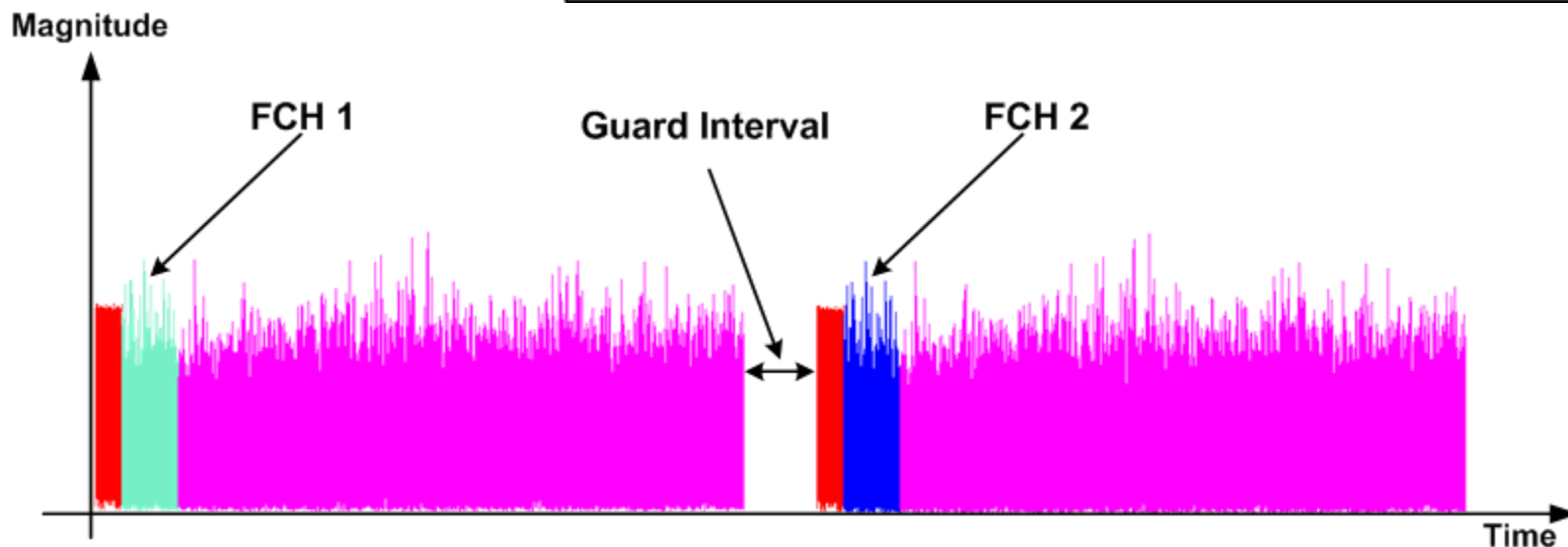
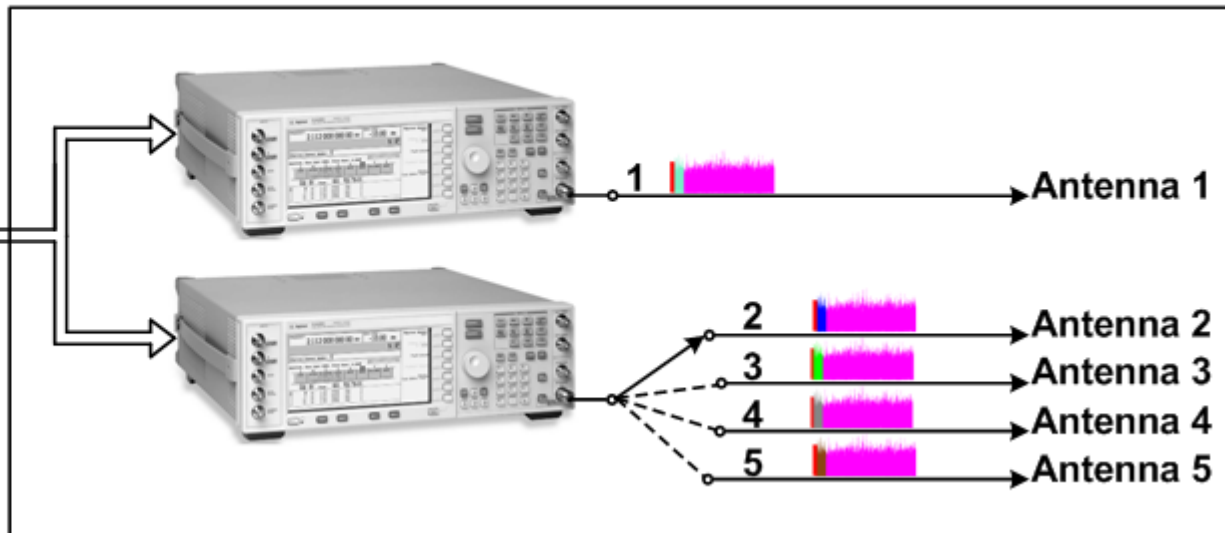
# Waveform Analysis



## Digital Signal Processing

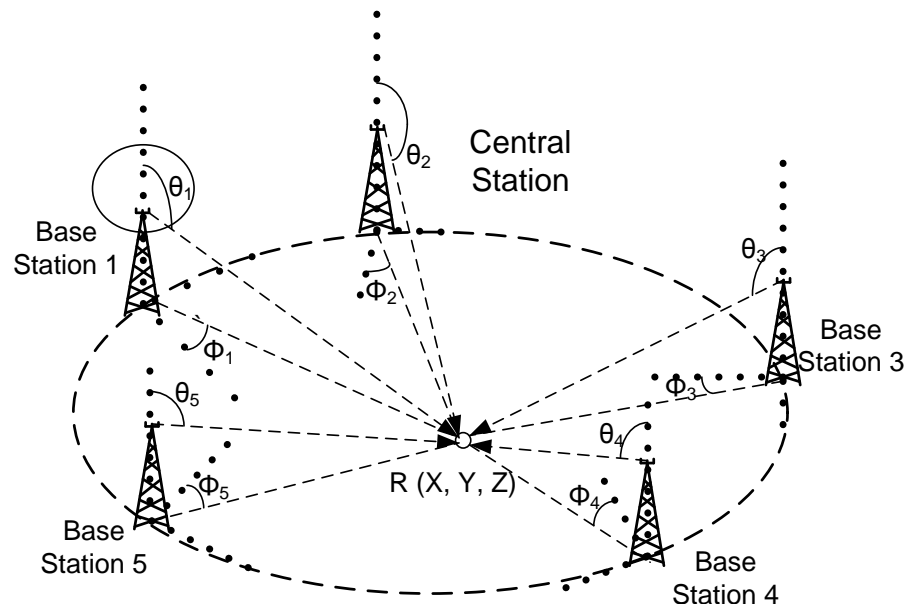


## Signal conditioning and Time Multiplexing



# Proposed Location Estimation Improvement

1. Initial TDOA Based Position Estimation
2. Multipath Characterization :
  - Ray Tracing
  - Neural Network
3. Broadcasting of error due to multipath to receiver.
4. Final TDOA Based Position Estimation with multipath correction at receiver.



# Experimental Setup

BS # 1

BS # 2

BS # 3

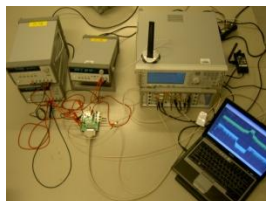
## MAX2830 RF Transceiver

- 2.4 – 2.5 GHz zero-IF
- OFDM compatible
- -75 dBm Rx Sensitivity
- Gain control interface

## Agilent 89600 VSA

- 2 MHz -78 MHz

## Mobile Node



**MAXIM**

Agilent Technologies

## Hawking Antennas

- 2.4 – 2.5 GHz
- 7 dBi gain

Receiver's Position

## Central Node



## Agilent E4440A VSA

- Up to 26.5 GHz
- 75 dB dynamic range

## Agilent VSA Software

- 1 ms capturing window

## Agilent E4438C ESG

- Up to 15 dBm 6 GHz

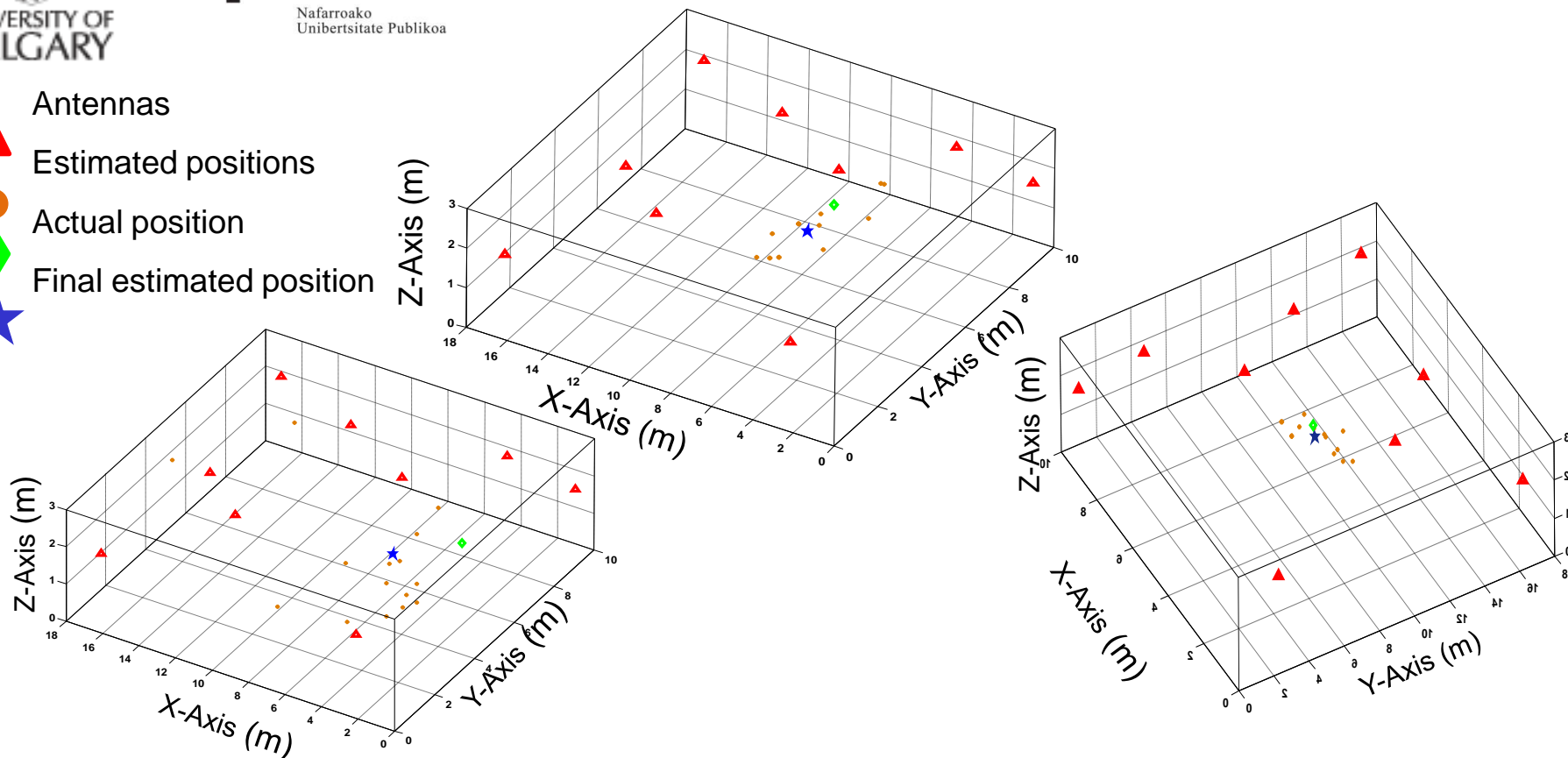
Agilent Technologies

BS # 5

BS # 4

# Measurements' Results

- Antennas
- Estimated positions
- Actual position
- Final estimated position



Receiver Position	Error before multipath correction (m)	Error after multipath correction (m)	Error reduction
Position 3	0.8201	0.6017	26.63%
Position 2	2.6416	2.6063	1.34%
Position 1	2.5846	1.444	44.13%

- A method to improve the accuracy and performance of a ray-tracing algorithm using feed forward neural networks has been developed.
- The proposed algorithm is used to improve the position estimation using TDOA.
- The proposed method only needs a software upgrade in the device.
- It has been shown that with this method an error reduction up to 44.14% is achieved.
- We propose this method as an inexpensive solution to improve the location estimation of TDOA based algorithms for environments characterized such as emergency scenes and indoor environment.



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**Thank you for your attention**