



# Cognitive/Adaptive Spatial Sampling

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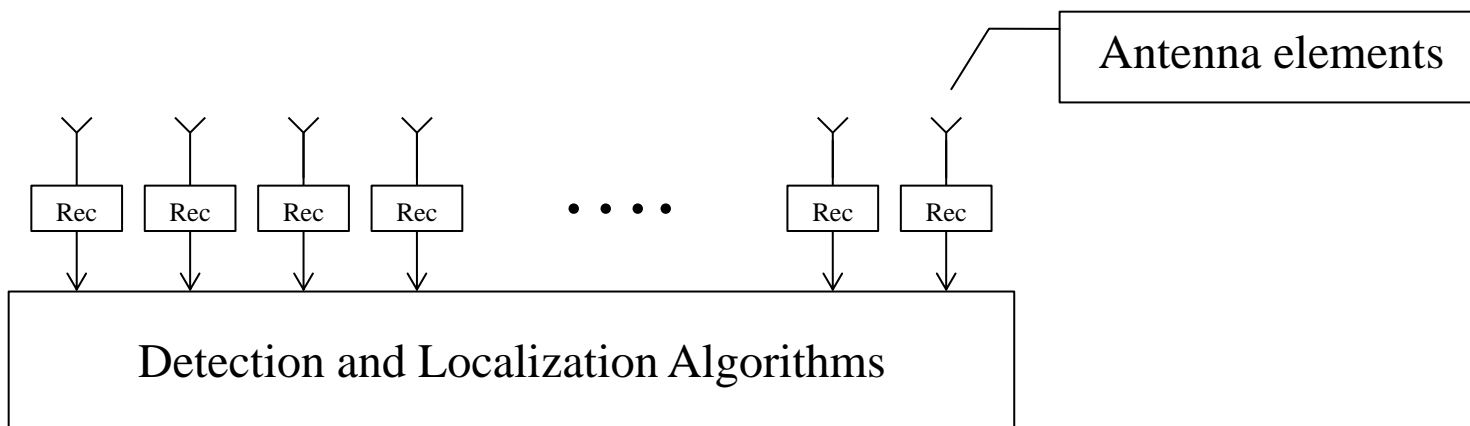
*Ben-Gurion University of the Negev*

**BGU Radar Symposium 2017**



# Introduction

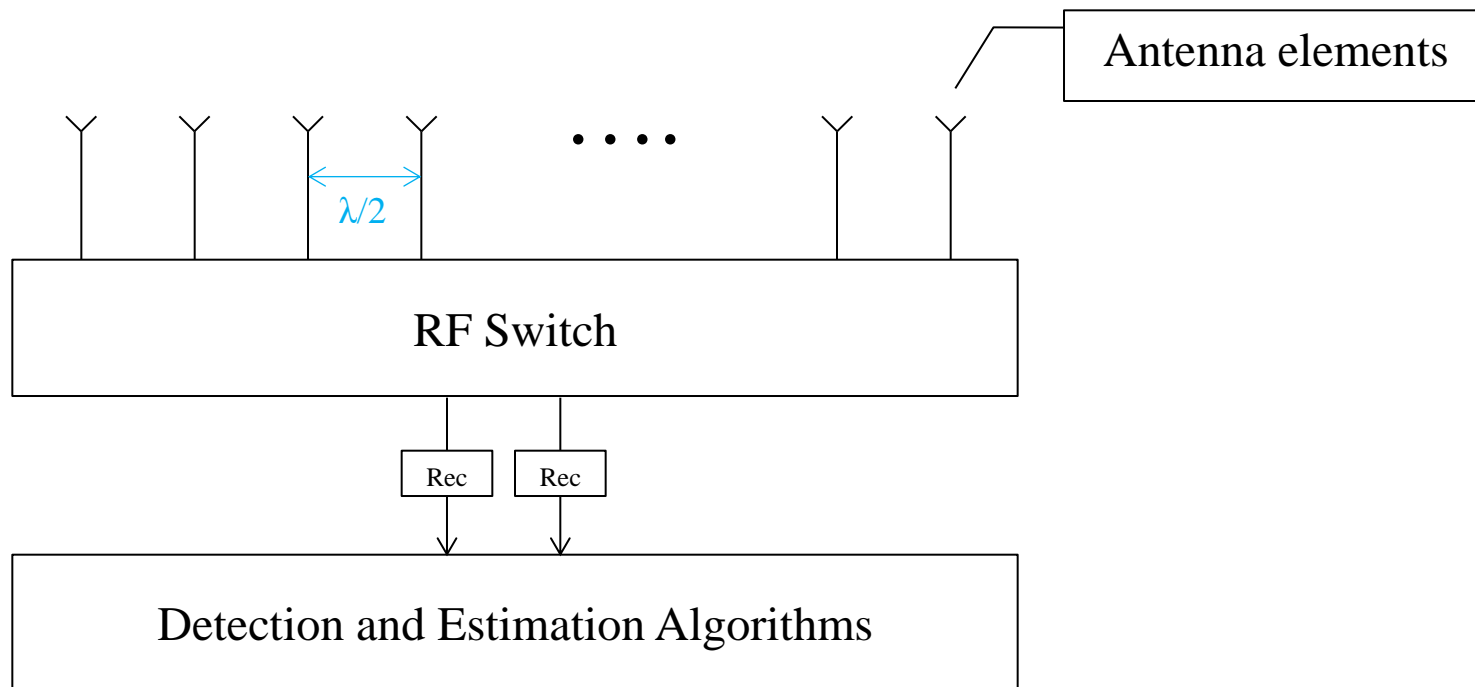
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- **Expensive receiving channels**
- Allows full flexibility in data processing
- Good accuracy without ambiguity in direction-of-arrival (DOA)



# Introduction

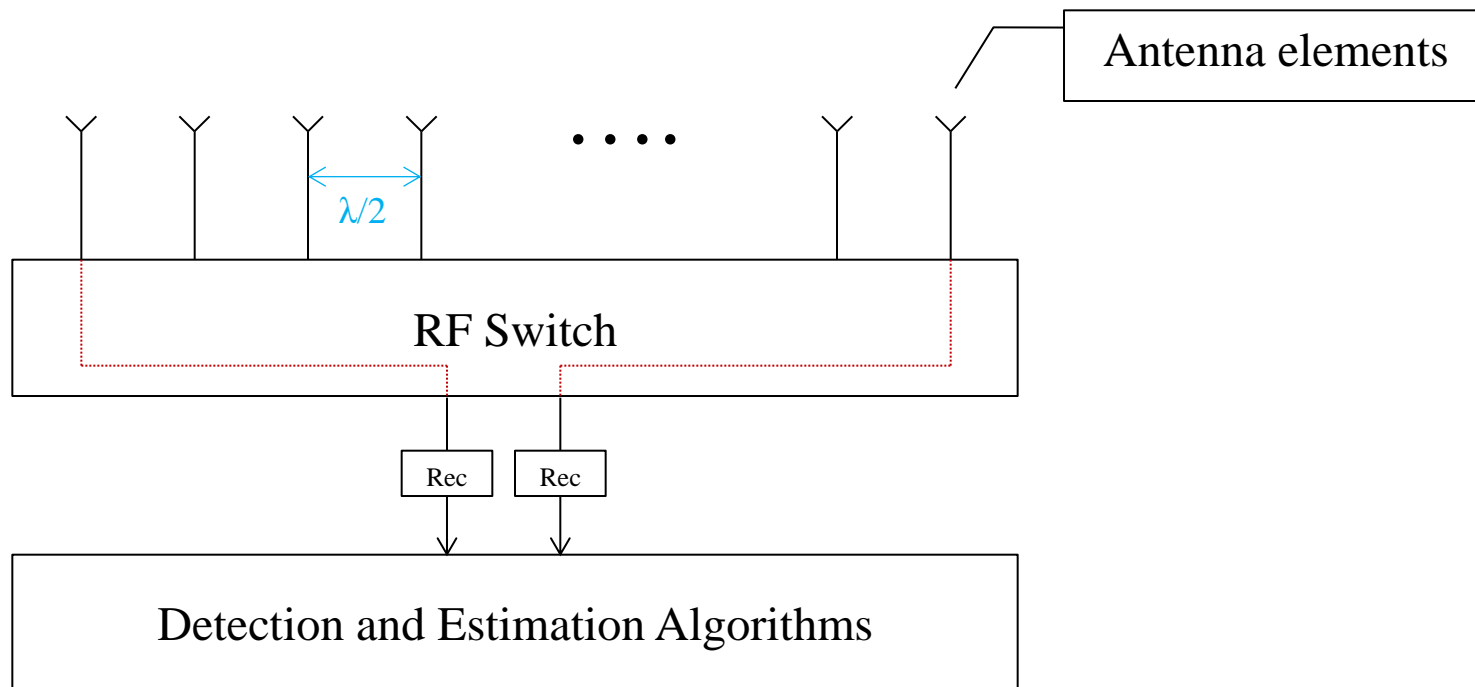


- Low cost

**What is the optimal switch configuration?**



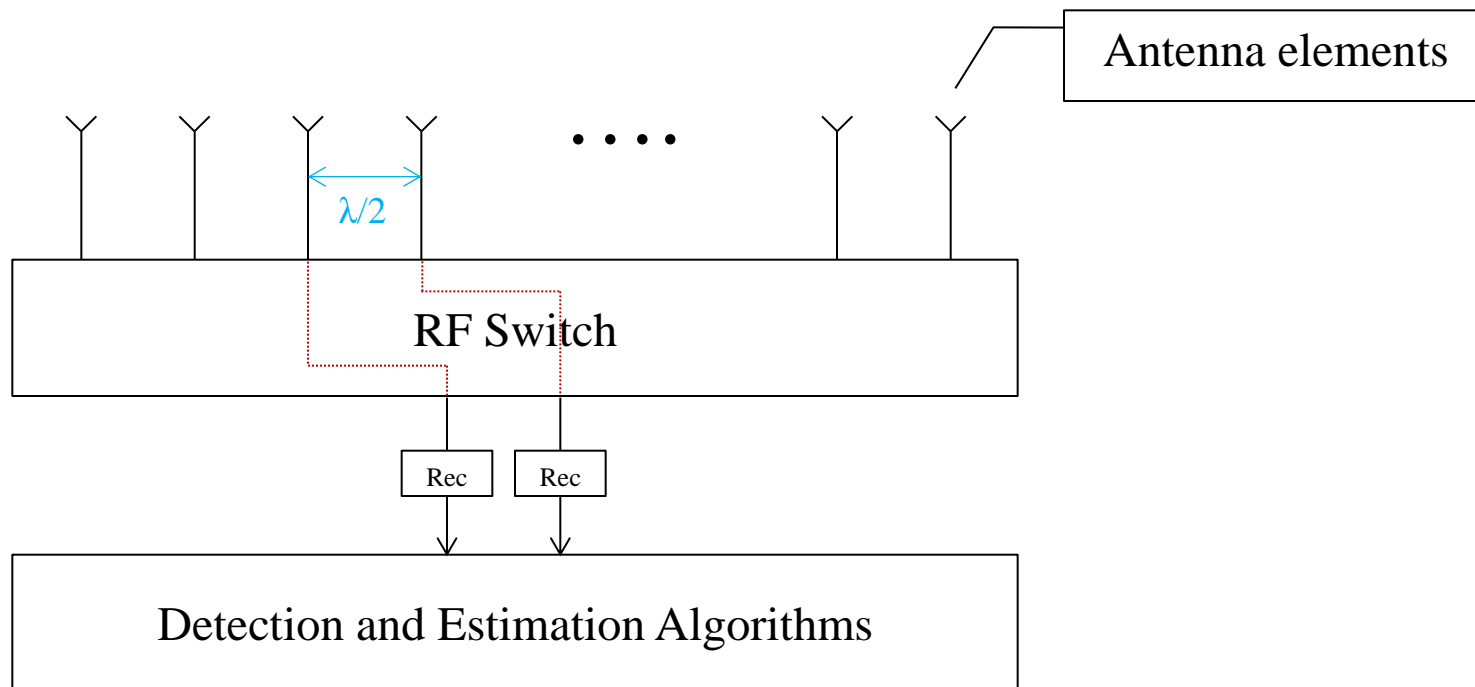
# Introduction



- Low cost
- Large aperture - high angular resolution/accuracy
- **Ambiguity in DOA**

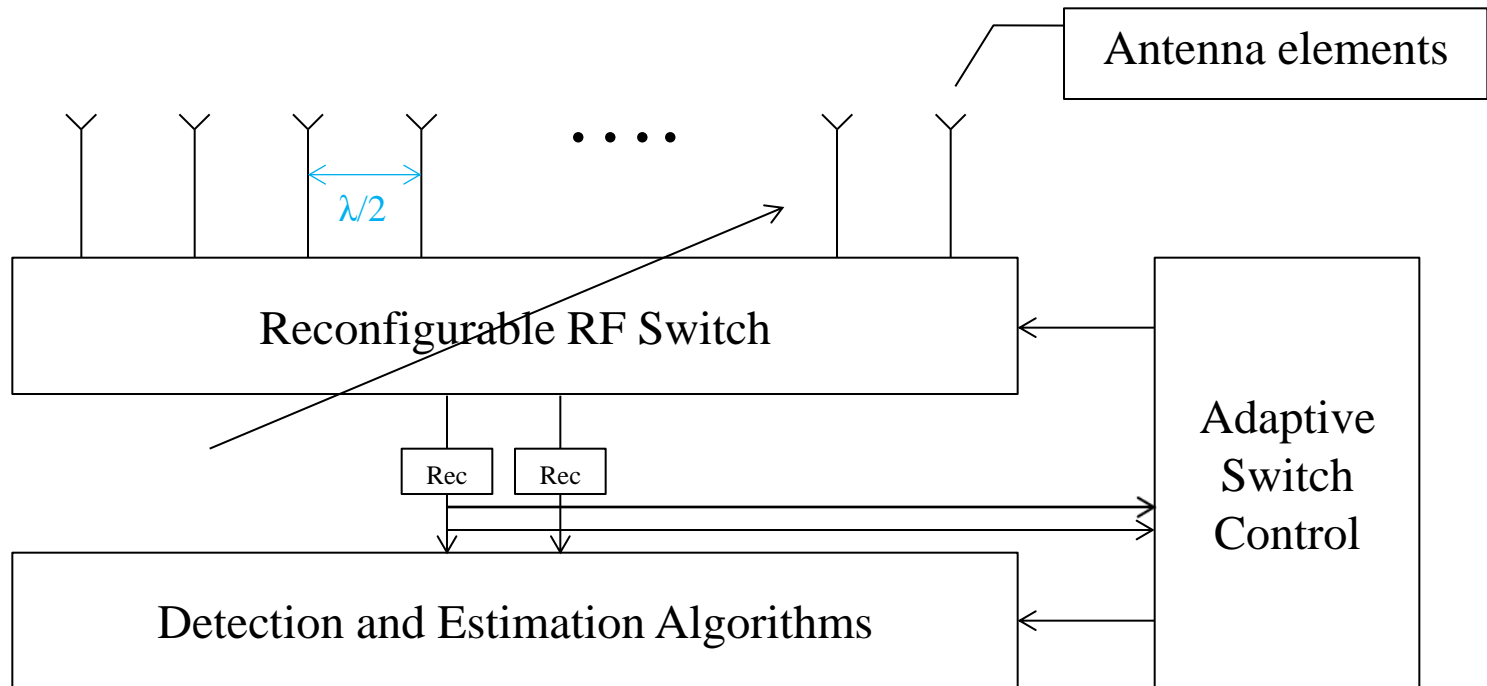


# Introduction



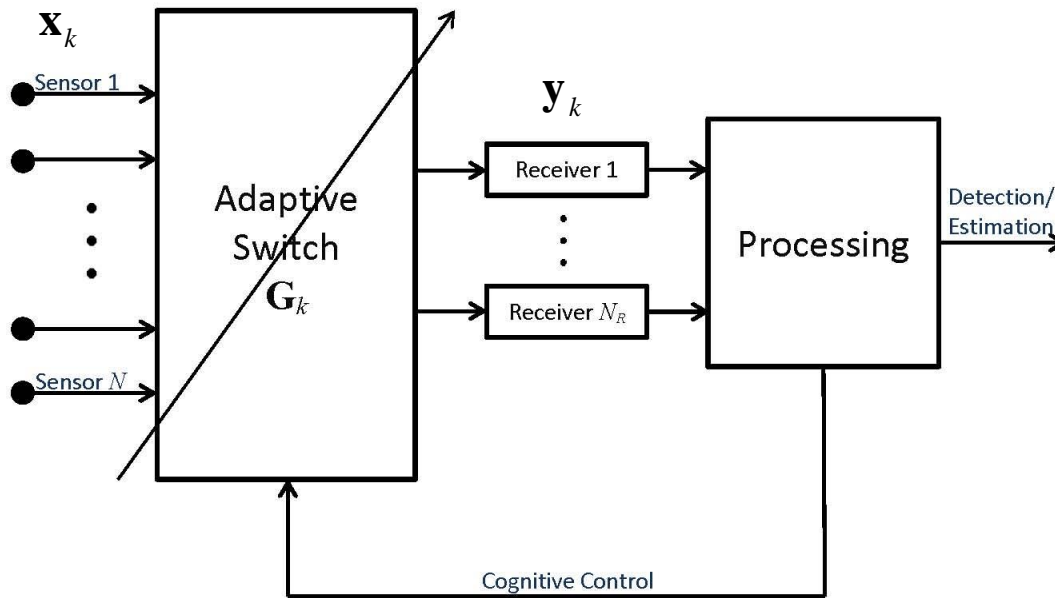
- Low cost
- **Small aperture - low angular resolution/accuracy**
- No DOA ambiguity

# Introduction



**Goal: Optimal adaptive switch control based on history observations.**

# Problem Statement



$$\mathbf{x}_k = \mathbf{a}(\theta)s_k + \mathbf{v}_k, \quad k = 1, 2, \dots$$

$$\mathbf{y}_k = \mathbf{G}_k (\mathbf{a}(\theta)s_k + \mathbf{v}_k),$$

$$[\mathbf{G}_k]_{ij} \in \{0, 1\} \text{ and } \mathbf{G}_k \mathbf{1}_N = \mathbf{1}_{N_R}$$

$$\mathbf{G}_k = \begin{bmatrix} 0 & 1 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & 0 & \dots & 1 \\ \vdots & & & & & \vdots \\ 0 & 0 & 0 & 1 & \dots & 0 \end{bmatrix}$$

Problem statement: optimize  $C(\mathbf{G}_k, \mathbf{X}^{(k-1)})$   
 $\mathbf{G}_k$

$\mathbf{X}^{(k-1)} = [\mathbf{x}_1, \dots, \mathbf{x}_{k-1}]$  - history measurements

$C(\cdot, \cdot)$  - Chosen criterion - e.g. conditional lower bound on the estimation accuracy.

# Problem Statement - Signal Models

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## Time - invariant signal :

$$\mathbf{y}_k = \mathbf{G}_k (\mathbf{a}(\theta)s + \mathbf{v}_k), \quad k = 1, 2, \dots$$

$$\mathbf{y}_k | s, \theta \sim CN(\mathbf{G}_k \mathbf{a}(\theta)s, \sigma^2 \mathbf{I}_{N_R})$$

Passive radar: (Observation time  $\times$  Bandwidth)  $\square$  1

Active radar: considering given Doppler cell with non-fluctuating target

## Time - varying signal :

$$\mathbf{y}_k = \mathbf{G}_k (\mathbf{a}(\theta)s_k + \mathbf{v}_k), \quad k = 1, 2, \dots$$

$$\mathbf{y}_k | \sigma_s^2, \theta \stackrel{i.i.d.}{\sim} CN(\mathbf{0}, \sigma_s^2 \mathbf{G}_k \mathbf{a}(\theta) \mathbf{a}^H(\theta) \mathbf{G}_k^H, \sigma^2 \mathbf{I}_{N_R})$$

Passive radar: non-negligible (Observation time  $\times$  Bandwidth)

Active radar: moving/fluctuating targets



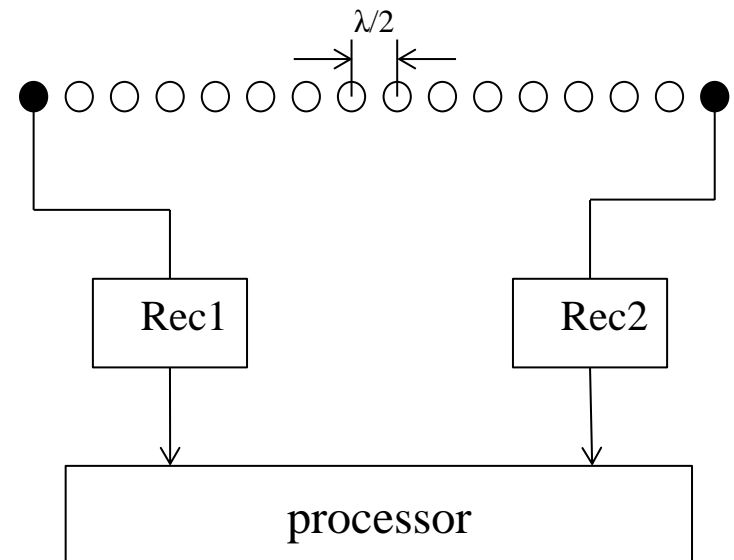
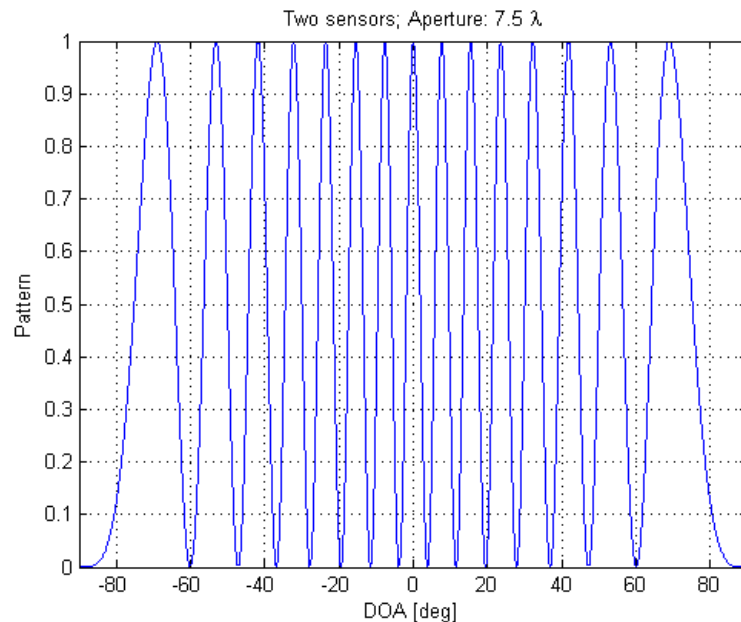
# Criterion for Antenna Selection

Bayesian Cramér-Rao bound (BCRB) is a popular tool for performance analysis.

Advantage: Simplicity

Disadvantage: Ignores “large errors” due to ambiguity.

Example: single source  $\mathbf{x}_k = \mathbf{a}(\theta)s + \mathbf{v}_k$ ,  $\mathbf{v}_k \sim CN(\mathbf{0}, \sigma^2 \mathbf{I})$



# Cognitive Antenna Selection (CASE)

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Proposed approach: Using the Bobrovsky-Zakai bound (BZB) as a criterion.

Advantage: Takes into account “large errors” due to grating lobes and ambiguity.

Disadvantage: Complexity

**Optimization criterion:**

$$\mathbf{G}_k^{opt}(\mathbf{X}^{(k-1)}) = \arg \min_{\mathbf{G}_k} BZB_{\varphi}(\mathbf{G}_k, \mathbf{X}^{(k-1)})$$

The optimization is performed using an iterative greedy approach:

At each iteration, the optimization is performed w.r.t. one switch (one row of the matrix  $\mathbf{G}_k$ ) while the other rows are fixed. This process is repeated for all the rows of  $\mathbf{G}_k$ .

$$\mathbf{G}_k = \begin{bmatrix} 0 & 1 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & 0 & \dots & 1 \\ \vdots & & & & & \vdots \\ 0 & 0 & 0 & 1 & \dots & 0 \end{bmatrix}$$

# Cognitive Antenna Selection (CASE)

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**Sequential BZB:**

$$BZB_{\theta}(\mathbf{G}_k, \mathbf{X}^{(k-1)}) = \sup_{\mathbf{h}} \frac{h_{\theta}^2}{B(\mathbf{G}_k, \mathbf{X}^{(k-1)}, \mathbf{h}) - 1} \quad \boldsymbol{\psi} = [\theta, s]^T, \quad \mathbf{h} = [h_{\theta}, h_s]^T$$

$$B(\mathbf{G}_k, \mathbf{X}^{(k-1)}, \mathbf{h}) = \mathbb{E} \left[ Z_k^{(k-1)} \cdot W^{(k-1)} \mid \mathbf{X}^{(k-1)} \right]$$

$$Z_k^{(k-1)} = \mathbb{E} \left( \underbrace{\frac{f^2(\mathbf{x}_k \mid \boldsymbol{\psi} + \mathbf{h}, \mathbf{X}^{(k-1)})}{f^2(\mathbf{x}_k \mid \boldsymbol{\psi}, \mathbf{X}^{(k-1)})}}_{\text{Current Info}} \mid \mathbf{X}^{(k-1)} \right) \quad W^{(k-1)} = \frac{f^2(\boldsymbol{\psi} + \mathbf{h} \mid \mathbf{X}^{(k-1)})}{\underbrace{f^2(\boldsymbol{\psi} \mid \mathbf{X}^{(k-1)})}_{\text{History Info}}}$$

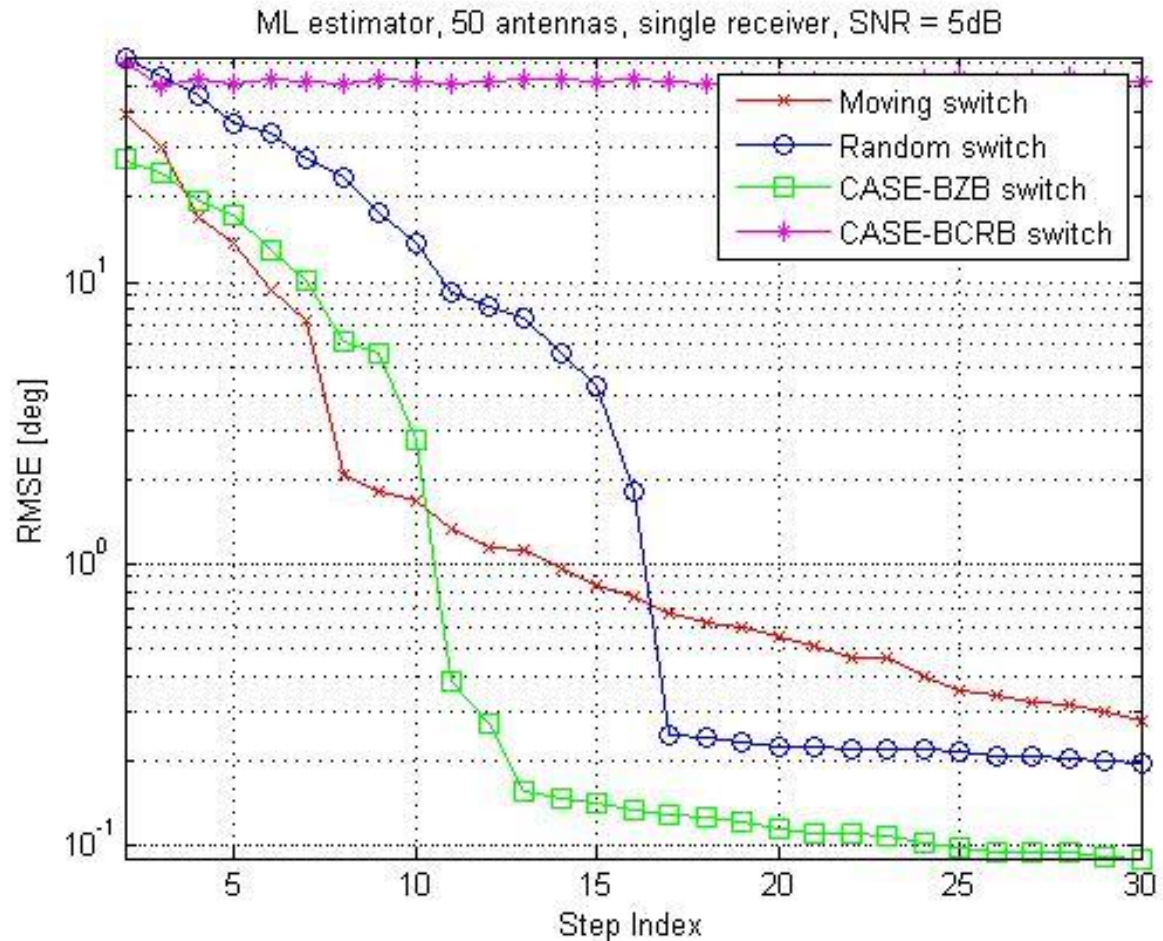
In our problem (time-invariant signals):

$$Z_k^{(k-1)} = \exp \left( \frac{2}{\sigma^2} \left\| \alpha \mathbf{G}_k \mathbf{a}(\theta) - (s + h_s) \mathbf{G}_k \mathbf{a}(\theta + h_{\theta}) \right\|^2 \right)$$

$$W^{(k-1)} = \exp \left( \frac{2}{\sigma^2} \sum_{m=1}^{k-1} \left[ -\left\| \mathbf{x}_m - (s + h_s) \mathbf{G}_k \mathbf{a}(\theta + h_{\theta}) \right\|^2 + \left\| \mathbf{x}_m - s \mathbf{G}_k \mathbf{a}(\theta) \right\|^2 \right] \right)$$

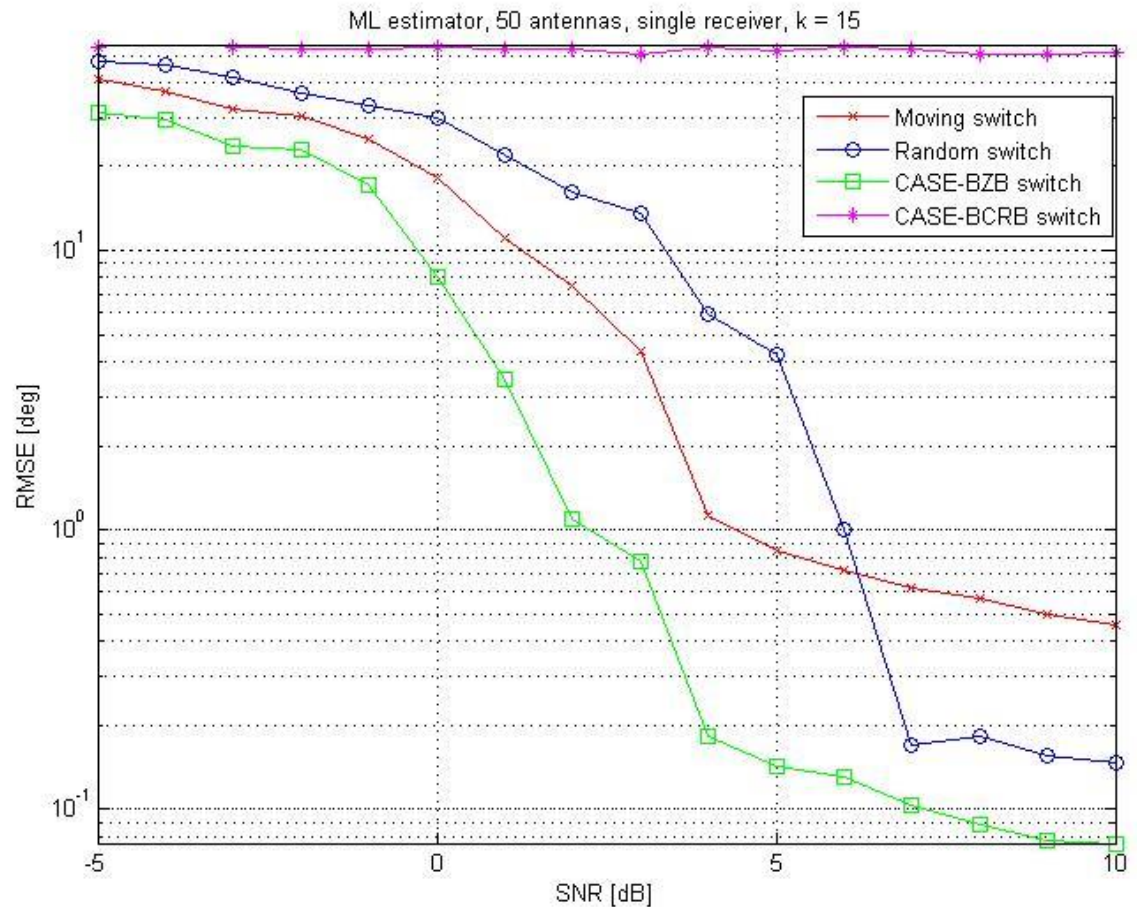
# Results - Single Target

Time-invariant signal  
1 Tx, 1 Rx  
50 antennas  
Azimuth  $\sim U(-80^\circ, 80^\circ)$   
SNR=5dB



# Results - Single Target

1 Tx, 1 Rx,  
50 antennas  
Azimuth  $\sim U(-80^\circ, 80^\circ)$   
 $k=15$  pulses



# Results - Single Target

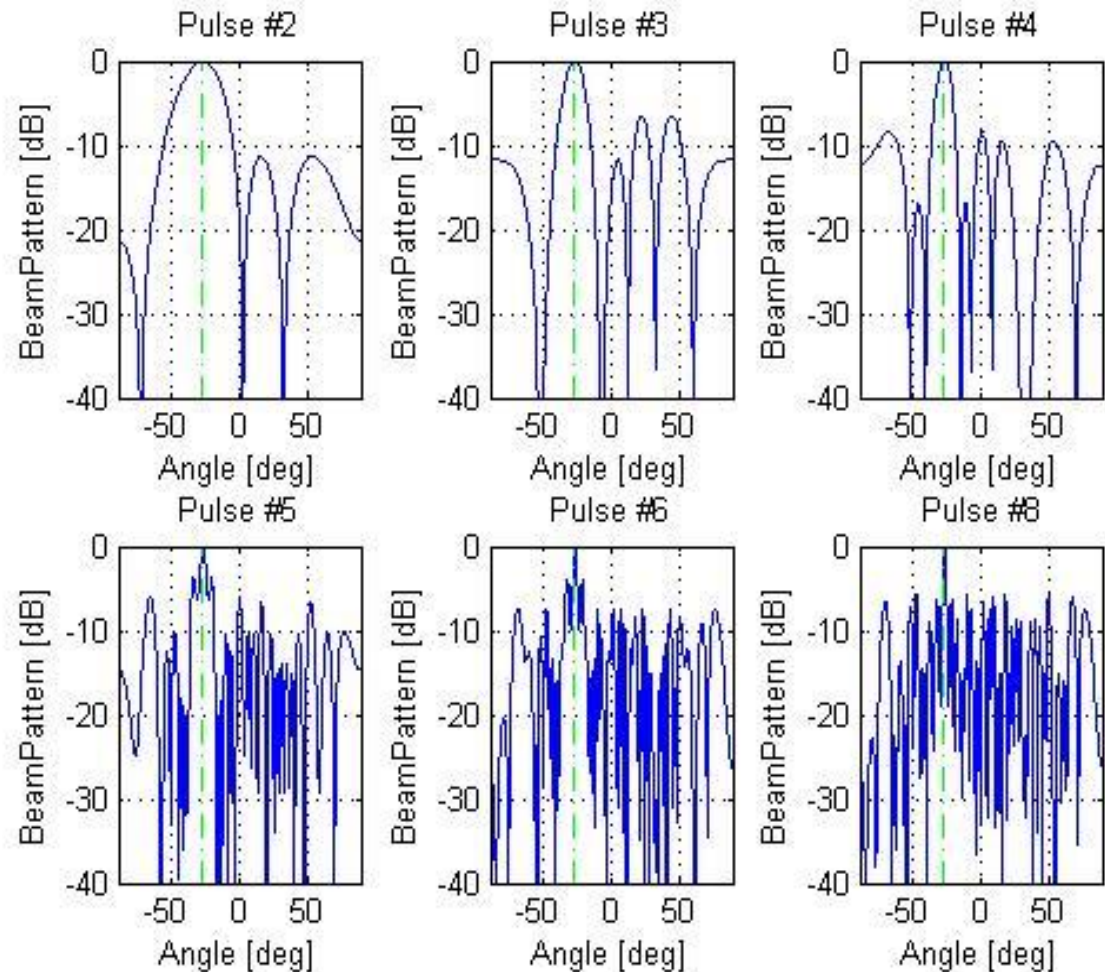
## Beampattern

1 Tx, 2 Rx

50 antennas

Azimuth  $\sim U(-80^\circ, 80^\circ)$

SNR=0dB



# Results - Single Target

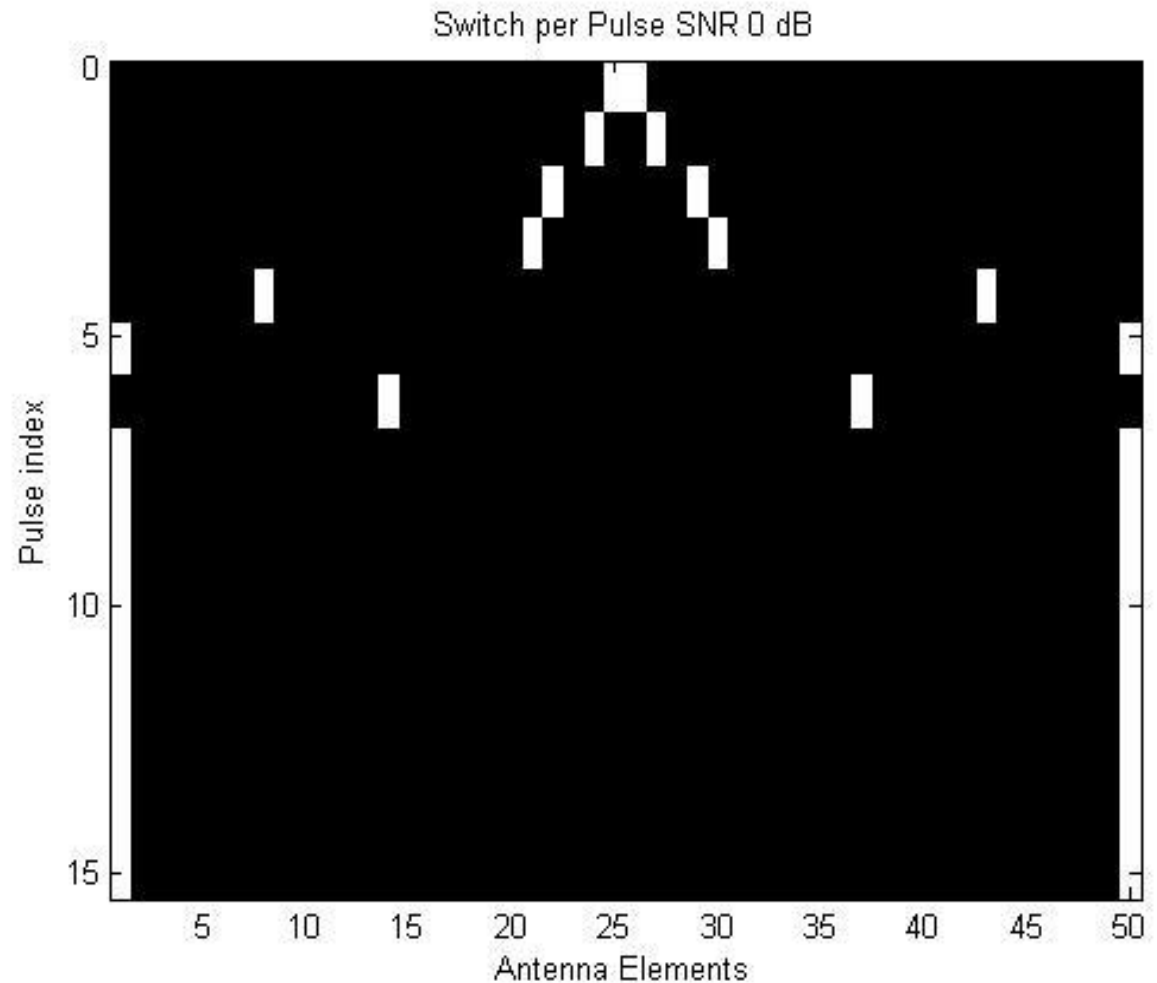
## Switching

1 Tx, 2 Rx

50 antennas

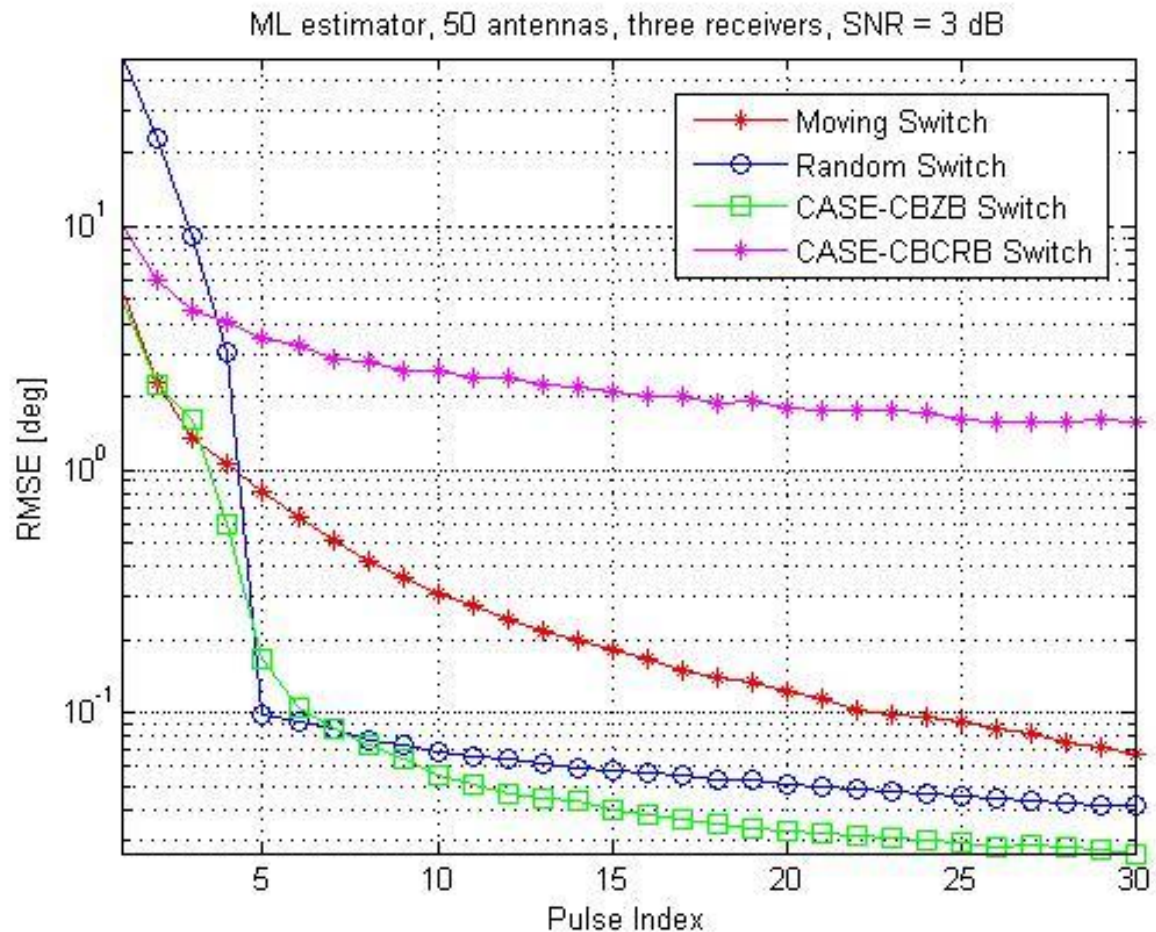
Azimuth  $\sim U(-80^\circ, 80^\circ)$

SNR=0dB



# Results – Two Targets

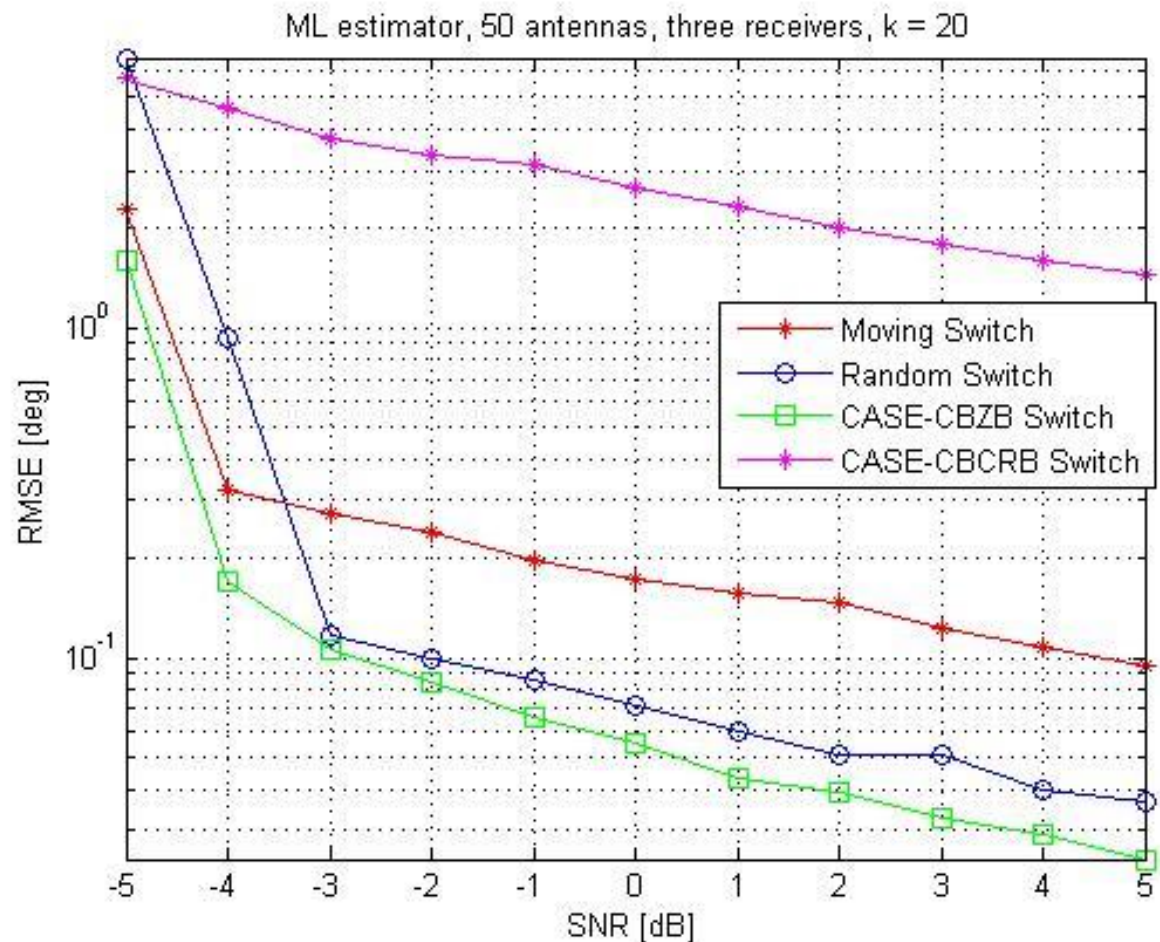
Time-varying signal  
1 Tx, 3 Rx  
50 antennas  
Two targets:  
Azimuth  $\sim U(-80^\circ, 80^\circ)$   
SNR=3dB  
20 snapshots per pulse





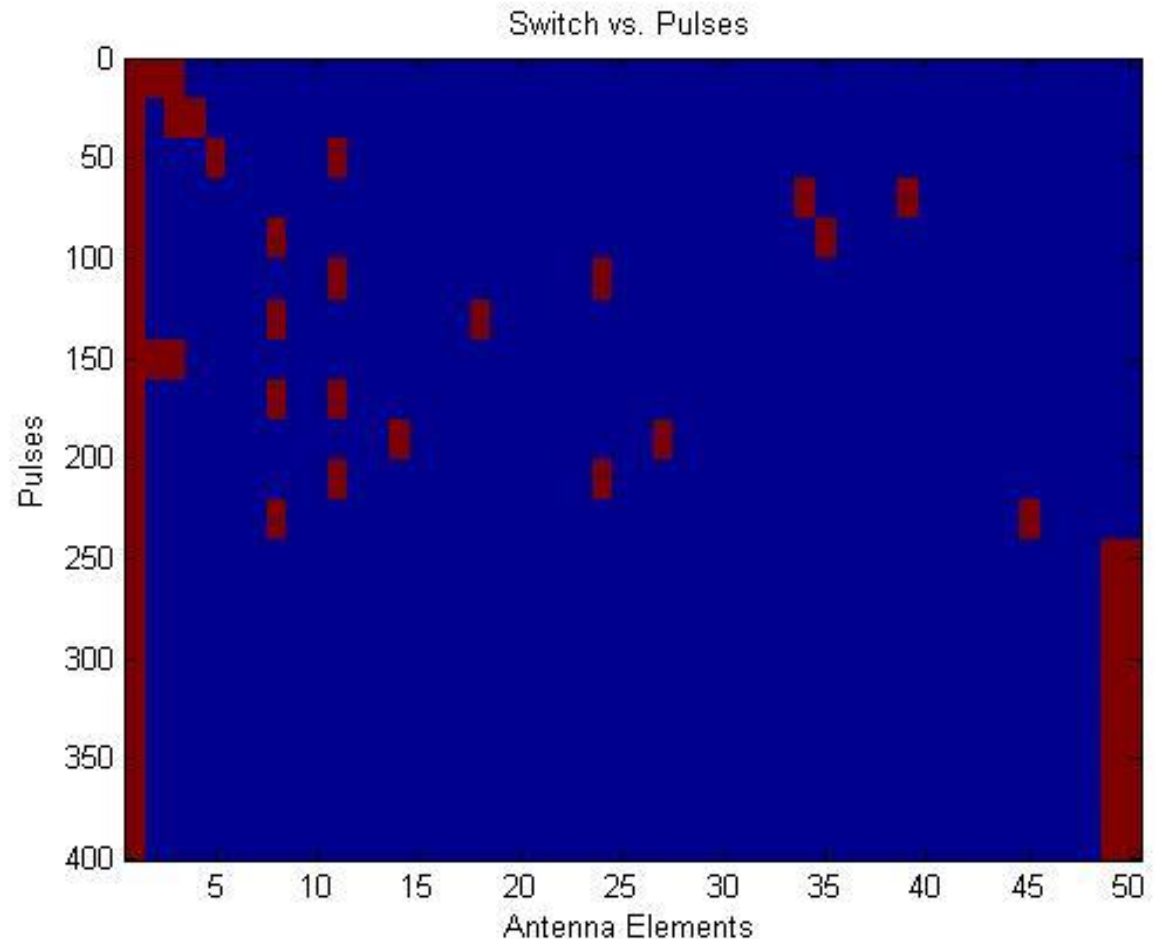
# Results – Two Targets

Time-varying signal  
1 Tx, 3 Rx  
50 antennas  
Two targets:  
Azimuth  $\sim U(-80^\circ, 80^\circ)$   
 $k=20$  pulses  
20 snapshots per pulse



# Results – Two Targets

Time-varying signal  
1 Tx, 3 Rx  
50 antennas  
Two targets:  
Azimuth  $\sim U(-80^\circ, 80^\circ)$   
 $k=20$  pulses  
20 snapshots per pulse



# Conclusion and Future Research

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- A new adaptive antenna selection approach was presented based on minimizing performance bounds for DOA estimation.
- The BCRB was found to be inappropriate as a criterion, since it takes into account only “local errors”. Therefore, the BZB was chosen as a criterion.
- Using small number of receiving channels, this approach allows to exploit the benefit of large array aperture, while avoiding ambiguity in DOA.
- The proposed technique was tested in the presence of one or two sources, and exhibited very good performance compared to other tested methods.
- Future/current research:
  - Considering hardware beamformer (phased array) instead of switch.
  - Taking into account Doppler information: needs considering joint DOA-Doppler ambiguity function (Doppler and DOA are not decoupled anymore).
  - Taking range information with multiple DOA/range distributed targets.



**Thank you!**

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