



Master's Thesis Defense



Illumination Optimized Transmit Signals for Space-Time Multi-Aperture Radar

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January 23, 2006

Committee

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OUTLINE



❑ Introduction

- What are we doing?
- Why are we doing it ?
- What has been done before ?
- How is our work different ?

❑ Execution + Results

- How do we do it?
- Is it good?
- If yes, how much is it better than the work done previously?

❑ More Observations

- How close do we reach to the goal we started with initially?
- Is there anything more to it?

❑ Conclusions and Future work

- What did we learn?
- What more can be done?



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OBJECTIVE



- To come up with **illumination optimized** transmit signals for a space-time multi-aperture radar
- **Applications**
 - SAR
 - GMTI
 - AMTI
- **Why a distributed sensor ?**
 - More robust structure
 - Manageable power requirements
 - **Potential for space-time operation**
- **Project supported by AFRL**





WHY IS IT IMPORTANT?



- **Objective of any radar – SAR, GMTI or AMTI: Accurate detection and estimation of Targets**
 - Place energy on the regions it is interested in (targets)
 - Not waste any energy on regions it is not interested in (clutter)
 - Also distribute the energy equally on all targets
 - Make returns from all targets as dissimilar or uncorrelated as possible
- illumination optimization
(Maximizes SINR)
- ambiguity optimization
(Maximizes estimation)
- **Ways to control the radar performance:**
 - Add more transmitters/receivers
 - Modify the antenna array
 - **Change the radar transmit signal**



PRIOR WORK

OPTIMAL TRANSMIT SIGNAL CONSTRUCTION

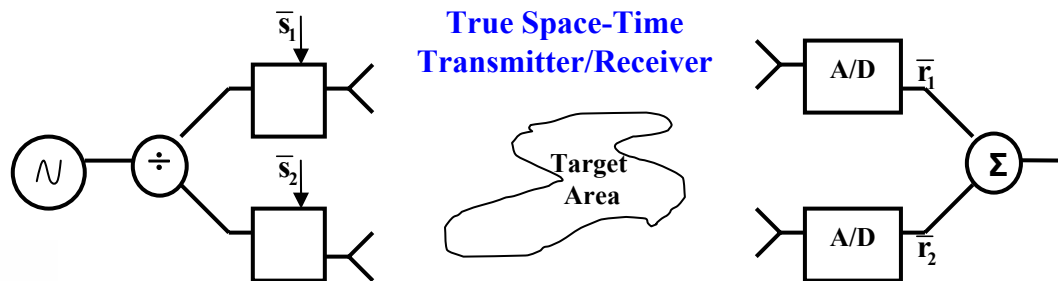
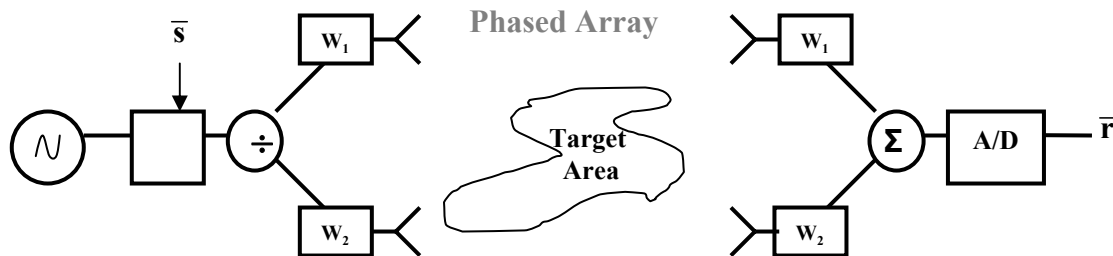
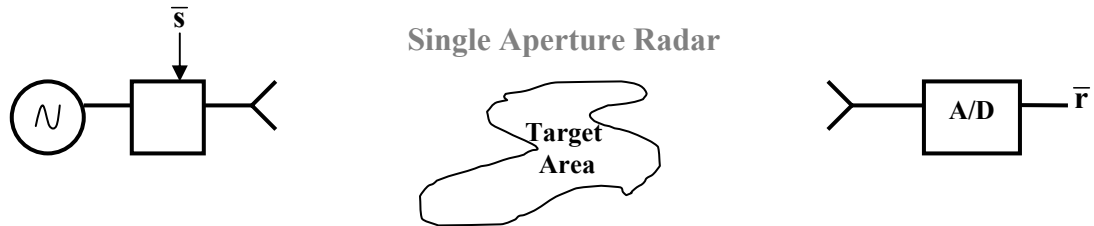


- **Optimal transmit signals/codes for illumination optimization**
 - Time-frequency codes
 - Optimized for a single target
 - Compromise between SINR and quality of radar waveform
 - Pseudo space-time codes → Phased arrays, essentially plain spatial codes
- **Nothing on the True Space-Time Transmitter**



OUR APPROACH

A TRUE SPACE-TIME TRANSMITTER

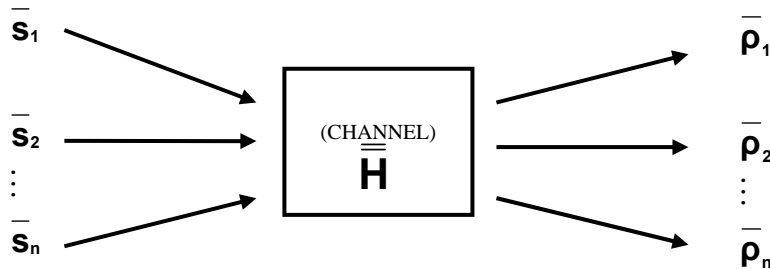




SPACE-TIME CODES IN COMMUNICATIONS

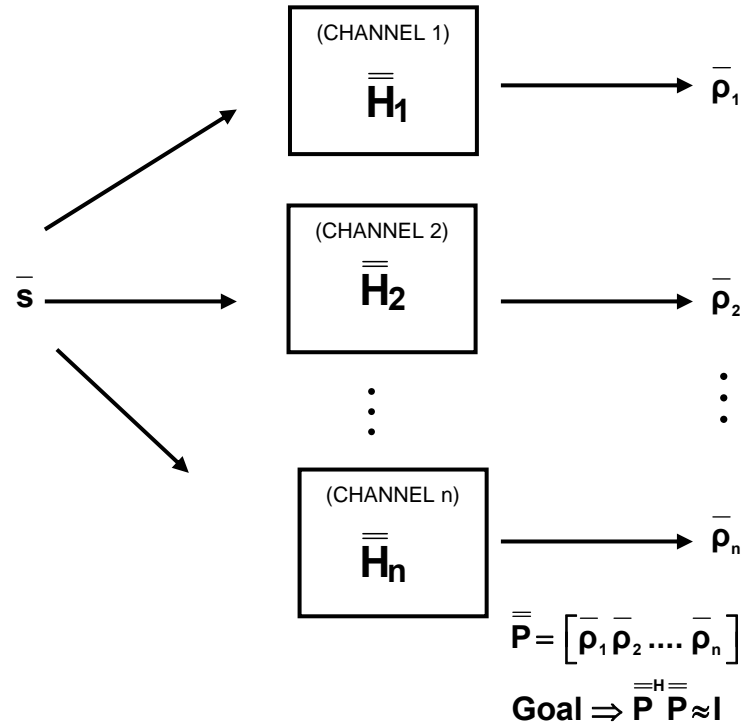


- **Space-time codes for communications**
 - A much easier problem



$$\bar{P} = [\bar{p}_1 \bar{p}_2 \dots \bar{p}_n] \quad \text{Goal} \Rightarrow \bar{P}^H \bar{P} \approx I$$

Space Time Codes for Communications



$$\bar{P} = [\bar{p}_1 \bar{p}_2 \dots \bar{p}_n] \quad \text{Goal} \Rightarrow \bar{P}^H \bar{P} \approx I$$

Space Time Codes for Radars



TRUE SPACE-TIME CODES - MOTIVATION



- **Better radar performance → Using same transmit power and radar resolution.**
- **Reduced receiver complexity**
- **Improved performance in multiple modes – e.g. SAR and GMTI**



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RADAR MODELS - THE BACKGROUND



- **We need radar models to:**
 - Simulate the space-time radar and the propagation from the transmitter to the targets
 - Model complex targets geometries consisting of any combination of point, distributed, stationary, moving, airborne and surface targets
 - Represent the radar transmit signal as a complex superposition of orthonormal basis functions
- **A transmission, target and propagation model and a transmit signal model were coded and implemented in MATLAB.**



TRANSMISSION, TARGET AND PROPAGATION MODEL



- **Space-Time Transmitter**

- The radar transmit signal is modeled as a superposition of N basis functions ϕ_n $\rightarrow s(\bar{z}) = \sum_{n=1}^N s_n \phi_n(\bar{z})$

- The basis functions ϕ are functions of 3-D space, time and frequency collectively spanning the entire timewidth, bandwidth and the spatial extent of the of the radar array $\rightarrow \bar{z} = [x \ y \ z \ t \ w]^T$

- The vector \mathbf{s} containing the complex weights s_n for each of the basis functions then represents the transmit signal completely. $\rightarrow \mathbf{s} = [s_1 \ s_2 \ s_3 \ \dots \ s_N]^T$



TRANSMISSION TARGET AND PROPAGATION MODEL



- **Targets**

- The joint target scattering response can also be modeled as a combination of N_t orthonormal basis functions ψ_t

$$\rightarrow \gamma(\bar{y}) = \sum_{t=1}^{N_t} \gamma_t \psi_t(\bar{y})$$

- The basis functions ψ are functions of 3-D space and radial velocity

$$\rightarrow \bar{y} = [x \ y \ z \ v_r]^T$$

- The vector γ of complex scattering coefficients γ_t then defines the set of illuminated targets completely

$$\rightarrow \gamma = [\gamma_1 \ \gamma_2 \ \gamma_3 \ \cdots \ \gamma_{N_t}]^T$$



TRANSMISSION, TARGET AND PROPAGATION MODEL



- **Space-Time Receiver**

- The received signal can be also be represented as a weighted superposition of M orthonormal space-time basis functions φ_m $\rightarrow r(\bar{x}) = \sum_{m=1}^M r_m \varphi_m(\bar{x})$

- The basis functions φ are again functions of 3-D space, slow time and fast frequency. $\rightarrow \bar{x} = [x \ y \ z \ t \ w]^T$

- The vector \mathbf{r} of complex weights r_m completely defines the received space-time signal. $\rightarrow \mathbf{r} = [r_1 \ r_2 \ r_3 \ \dots \ r_M]^T$



TRANSMISSION TARGET AND PROPAGATION MODEL



- **Propagation**

- The transmitted, target and receive functions are related by the following convolution integral through the dyadic Green's propagation functions

$$r(\bar{x}) = \int \vec{H}(\bar{x}; \bar{y}) \cdot \gamma(\bar{y}) \cdot \int \vec{G}(\bar{y}; \bar{z}) \cdot s(\bar{z}) d\bar{z} d\bar{y}$$

- Also since,

$$r_m = \int r(\bar{x}) \varphi_m(\bar{x}) d\bar{x}$$

- We can simplify as follows

$$r_m = \int \varphi_m(\bar{x}) \int \vec{H}(\bar{x}; \bar{y}) \cdot \gamma(\bar{y}) \cdot \int \vec{G}(\bar{y}; \bar{z}) \cdot s(\bar{z}) d\bar{z} d\bar{y} d\bar{x}$$

$$r_m = \sum_{t=1}^{N_t} \gamma_t \sum_{n=1}^N s_n \int \varphi_m(\bar{x}) \int \vec{H}(\bar{x}; \bar{y}) \cdot \psi_t(\bar{y}) \cdot \int \vec{G}(\bar{y}; \bar{z}) \cdot \phi_n(\bar{z}) d\bar{z} d\bar{y} d\bar{x}$$

$$r_m = \sum_{t=1}^{N_t} \gamma_t \sum_{n=1}^N s_n H_{mn}^t$$

where, $H_{mn}^t = \int \varphi_m(\bar{x}) \int \vec{H}(\bar{x}; \bar{y}) \cdot \psi_t(\bar{y}) \cdot \int \vec{G}(\bar{y}; \bar{z}) \cdot \phi_n(\bar{z}) d\bar{z} d\bar{y} d\bar{x}$



TRANSMISSION TARGET AND PROPAGATION MODEL



$$\mathbf{r} = \sum_{t=1}^{N_t} \gamma_t \mathbf{H}_t \mathbf{s} + \mathbf{n}$$

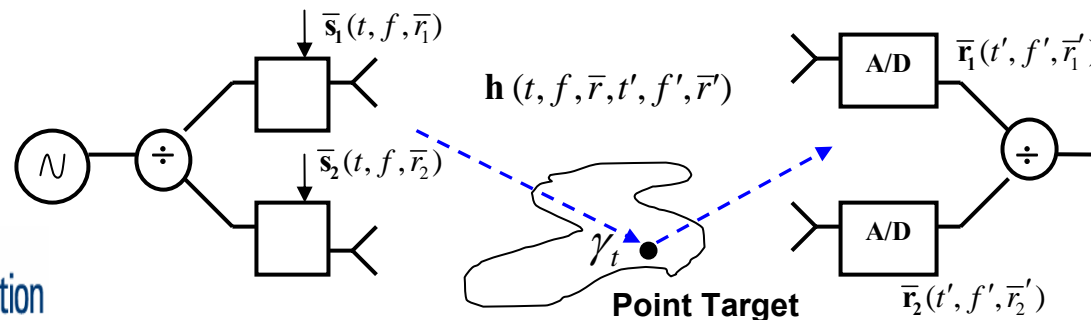
\mathbf{s} → is the transmit signal vector completely representing the transmitted signal

\mathbf{r} → is the received signal vector completely representing the received signal

γ_t → is the scattering coefficient for each target

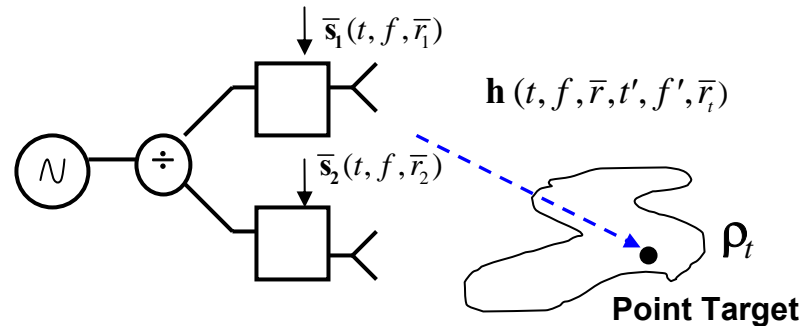
\mathbf{n} → measurement noise vector

\mathbf{H}_t → is a 2-D matrix relating the N transmitted samples to the M received samples for the t^{th} target – analogous to the convolution function of a two port network





ADJUSTMENT FOR THE ILLUMINATION OPTIMIZATION PROBLEM



- For the illumination optimization problem we just need to model the propagation from the transmitter to the targets
- The propagation matrix for each target \mathbf{H}_t is modified accordingly
- The normalized response at the target due to a transmit signal \mathbf{s} is given by

$$\rho_t = \mathbf{H}_t \mathbf{s}$$

- The set of N_t propagation matrices \mathbf{H}_t , and normalized responses ρ_t are critical parameters for all our algorithms and optimization procedures.



TRANSMIT SIGNAL MODEL



- Required for expanding the transmit signal as a weighted superposition of space-time orthonormal basis functions
- A time-frequency basis function consist of a train of U wideband pulses – the same pulse train is present at each antenna resulting in a space-time basis function
- The pulse trains at the same antenna have different delays and/or different phase weightings - \therefore the basis functions form an orthonormal set
- The different delays and phase weightings characterize the different fast and slow time basis functions available
- Each basis function has wide timewidth and bandwidth, and would make an adequate radar signal in itself
- Choice and number of time-frequency basis functions are important



TRANSMIT SIGNAL MODEL



- **Inputs to the model**

- f_c = carrier frequency (Hz)
- B = transmit signal bandwidth (Hz)
- f_o = pulse repetition frequency – PRF (Hz)
- U = integer number of pulses transmitted as part of the transmit signal
- Q = odd number of ‘fast-time’ basis functions
- P = odd number of ‘slow-time’ basis functions
- $g_s(t)$ = a ‘mother function’ used to generate new slow-time basis functions
- $G_f(w)$ = a ‘mother function’ used to generate new fast-time basis functions
- $\{\tau_q\}$ = Q time delay values used to generate all the fast-time basis functions
- $\{w_p\}$ = P frequency shift values used to generate all the slow-time basis functions

- $1/f_o = T_o$ = pulse repetition interval – PRI (sec)
- $UT_o = T$ = transmit signal timewidth (sec)
- $w_o = 2\pi f_o$ = angular pulse repetition frequency (radians/sec)
- $w_c = 2\pi f_c$ = angular carrier frequency (radians/sec)



TRANSMIT SIGNAL MODEL



- Any real valued temporal signal can be expressed as $v_s(t) = \text{Re}\{S(t)e^{-j\omega_c t}\}$
- $S(t)$ can be written as a weighted superposition of PQ complex basis functions

$$S(t) = \sum_p \sum_q S_{pq} \psi_{pq}(t) \quad \psi_{pq}(t) = s_p(t) \sum_u f_q(t - uT_o) e^{j\omega_c uT_o}$$

$$f_q(t) = g_f(t - \tau_q) e^{j\omega_c \tau_q}, \text{ where } \tau_q \ll T_o$$

$$S_p(\omega) = G_s(\omega - \omega_p), \text{ where } \omega_p \ll \omega_o$$

$$S_p(\omega) = \int_{-\infty}^{+\infty} s_p(t) e^{-j\omega t} dt \quad \text{and} \quad G_s(\omega) = S_o(\omega) \quad Q\tau_q \leq T_o \quad P\omega_p \leq \omega_o$$

- **Sampled Windowed Fourier Transform of $S(t)$** $S(t) = \sum_p \sum_q S_{pq} e^{j\omega_p t} g_s(t) \sum_u f_q(t - uT_o) e^{j\omega_c uT_o}$

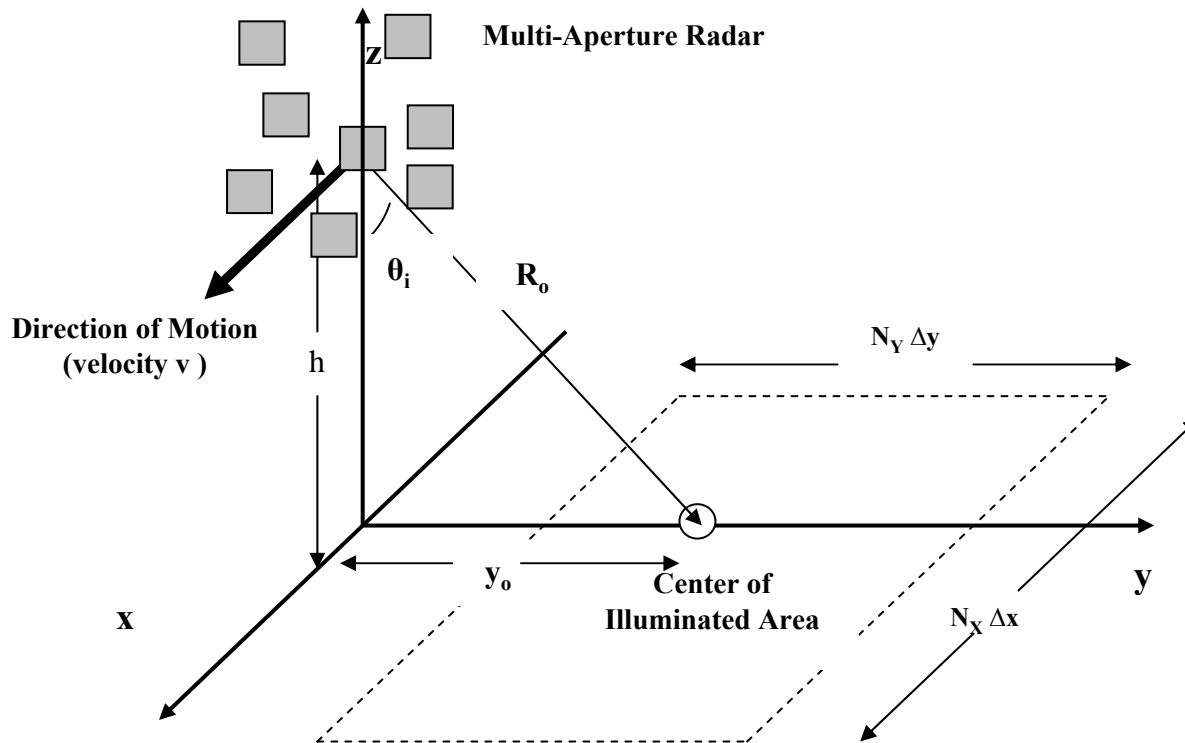
$$s(uT_o, \omega) = \sum_p g_s(uT_o) e^{ju\omega_p T_o} \sum_q S_{pq} G_f(\omega) e^{-j\omega uT_o} e^{-j(\omega - \omega_c)\tau_q}$$

$$s_{uv} = \sum_p g_s(uT_o) e^{ju\omega_p T_o} \sum_q S_{pq} G_f\left(\frac{v\omega_o}{2}\right) e^{-j\left(\frac{v\omega_o}{2} - \omega_c\right)\tau_q}$$

- **Defining** $\psi_{uv}^{pq} = g_s(uT_o) e^{ju\omega_p T_o} G_f\left(\frac{v\omega_o}{2}\right) e^{-j\left(\frac{v\omega_o}{2} - \omega_c\right)\tau_q}$ **we get** $s_{uv} = \sum_p \sum_q \psi_{uv}^{pq} S_{pq}$ OR, $\mathbf{s}^t = \boldsymbol{\Psi} \mathbf{S}^t$



RADAR GEOMETRY



Radar Parameters

Radar height (h)	183 km
Radar velocity (v)	780 m/sec
Look angle (θ_i)	45 degrees
Carrier frequency (f_c)	10 GHz
Horizontal distance to the center of target grid	183 km
Actual distance to the center of target grid (R_o)	258.8 km



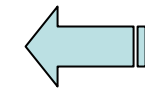
SPACE-TIME ILLUMINATION OPTIMIZATION – THE PROCESS



- If all scatterers are classified into two sets:
 - **Targets:** Scatterers we wish to illuminate or estimate
 - **Clutter:** Scatterers we do not wish to illuminate

- Then the **perfect transmit code** would:

- Illuminate all targets
- Not illuminate any clutter objects
- Distribute energy equally amongst all targets
- Make responses from all targets mutually orthogonal



**Illumination
Optimization**

- Unfortunately such a perfect transmit code does not exist!!

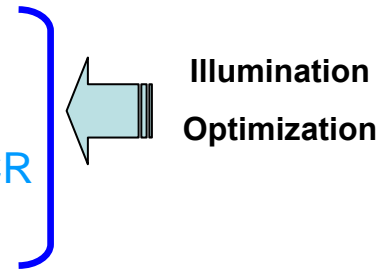


SPACE-TIME ILLUMINATION OPTIMIZATION

– OPTIMIZATION CRITERIA



- Instead we define a number of optimization criteria and try to satisfy them to the greatest possible extent
- If not perfect, then at least an optimal code
- Example of optimization criteria can be:
 - Maximize the total energy on all targets
 - Minimize the total energy on all clutter objects
 - Maximize the ratio of total signal (target) to clutter energy – SCR
 - Maximize the SCR for the target receiving the minimum SCR
 - Minimize the maximum correlation between any two targets





BASIC OPTIMIZATION CRITERIA

MAXIMUM TARGET ENERGY



- Total Energy on all target objects is given as

$$E_{targets} = \sum_{i \in targets} \rho_i' \rho_i = \mathbf{S}'\mathbf{A}\mathbf{S}$$

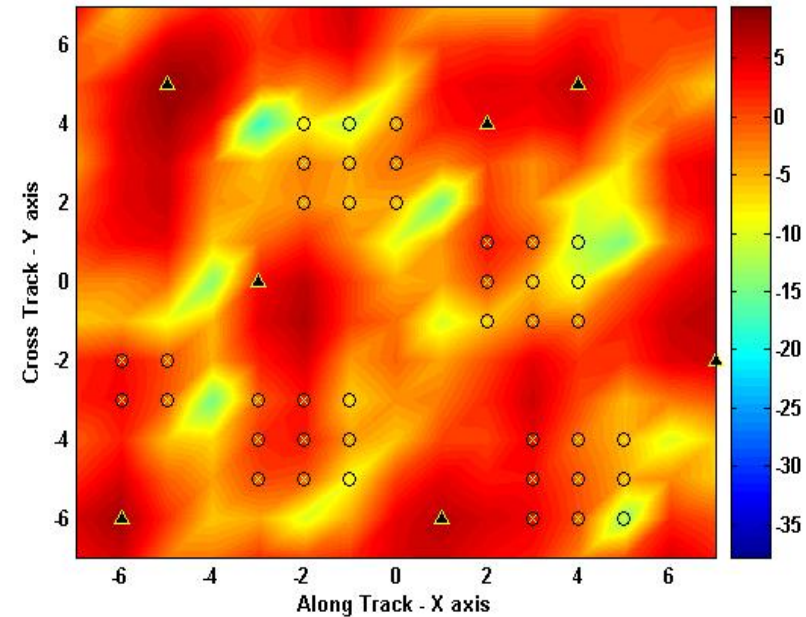
$$\text{where } \mathbf{A} = \sum_{i \in targets} \mathbf{H}_i' \mathbf{H}_i$$

- This energy is maximized when the eigen vector associated with the largest eigen value of the non-negative definite matrix \mathbf{A} is selected as the transmit code \mathbf{S}

$$\mathbf{A} = \sum_n \lambda_n^a \hat{e}_n^a \hat{e}_n^{a'}$$

$$\therefore \mathbf{S} = \hat{e}_n^a \text{ associated with } (\lambda_n^a)_{\max}$$

NOTE: For all cases 14 transmit antennas and 9 time-frequency basis functions, or a transmit signal dimension of 126



▲ Target Locations

○ Clutter Locations



BASIC OPTIMIZATION CRITERIA

MINIMUM CLUTTER ENERGY



- Total Energy on all clutter objects is given as

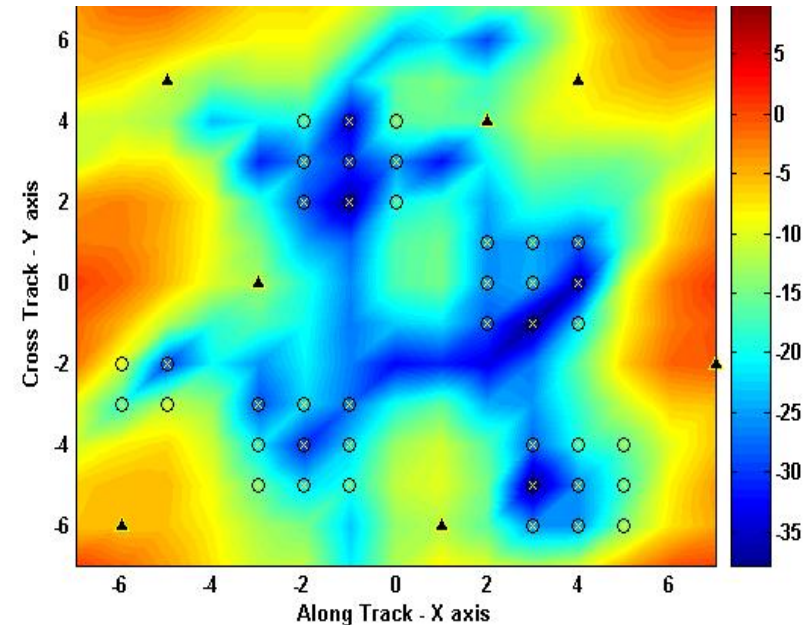
$$E_{clutter} = \sum_{j \in clutter} \rho_j' \rho_j = \mathbf{S}'\mathbf{B}\mathbf{S}$$

$$\text{where } \mathbf{B} = \sum_{j \in clutter} \mathbf{H}_j' \mathbf{H}_j$$

- This energy is minimized when the eigen vector associated with the smallest eigen value of the non-negative definite matrix \mathbf{B} is selected as the transmit code \mathbf{S}

$$\mathbf{B} = \sum_n \lambda_n^b \hat{e}_n^b \hat{e}_n^{b'}$$

$$\therefore \mathbf{S} = \hat{e}_n^b \text{ associated with } (\lambda_n^b)_{\min}$$



▲ Target Locations

○ Clutter Locations



BASIC OPTIMIZATION CRITERIA

MAXIMUM SCR



- The ratio of the signal to clutter energy is given as

$$SCR = \frac{E_{targets}}{E_{clutter}} = \frac{\mathbf{S}'\mathbf{A}\mathbf{S}}{\mathbf{S}'\mathbf{B}\mathbf{S}} = \frac{\tilde{\mathbf{S}}'\mathbf{C}\tilde{\mathbf{S}}}{\tilde{\mathbf{S}}'\tilde{\mathbf{S}}}$$

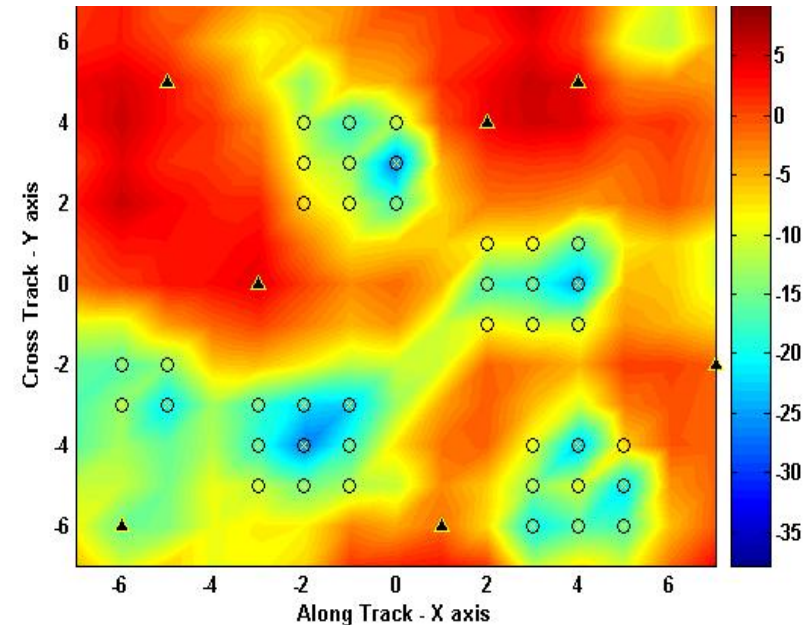
where $\mathbf{C} = (\mathbf{B}^{-1/2})' \mathbf{A} (\mathbf{B}^{-1/2})$ and $\tilde{\mathbf{S}} = \mathbf{B}^{1/2} \mathbf{S}$

- The SCR is maximized when the eigen vector associated with the largest eigen value of the non-negative definite matrix \mathbf{C} is selected, and the transmit code \mathbf{S} is determined from it

$$\mathbf{C} = \sum_n \lambda_n^c \hat{e}_n^c \hat{e}_n^{c'}$$

$\therefore \tilde{\mathbf{S}} \hat{e}_n^c$ associated with $(\lambda_n^c)_{\max}$

$$\text{and } \mathbf{S} = \mathbf{B}^{-1/2} \tilde{\mathbf{S}}$$

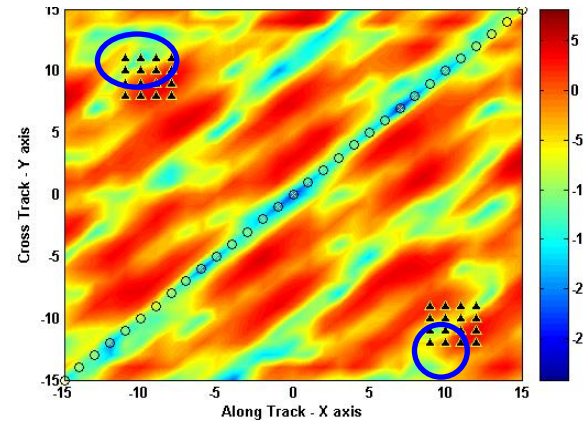
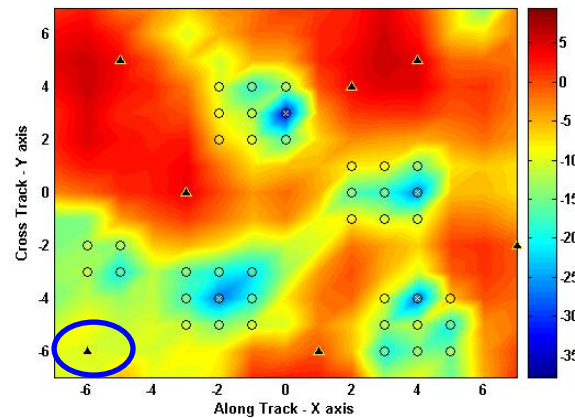


▲ Target Locations

○ Clutter Locations

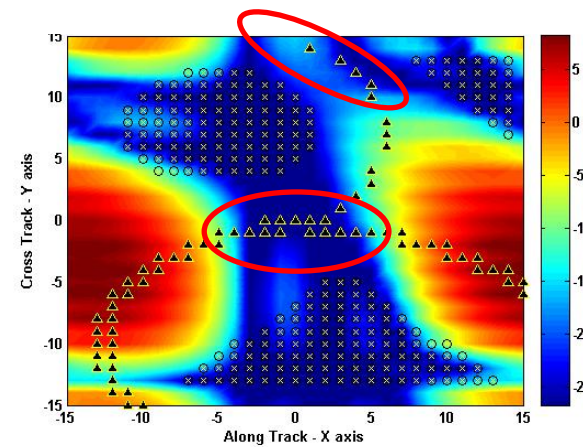
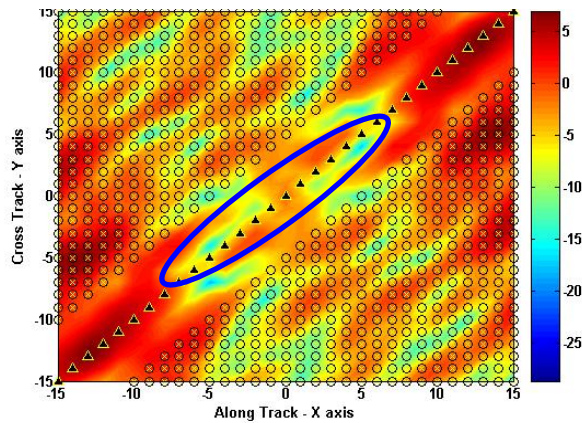


THE ORPHAN PROBLEM



▲ Target Locations

○ Clutter Locations





NEED FOR AN ADVANCED CRITERIA MAXI-MIN OR MINI-MAX



- We can define the alternate **maxi-min** or the **mini-max** criteria as:

The code which **maximizes** the **smallest** energy received by any target

The code which **minimizes** the **largest** energy received by any clutter object

The code which **maximizes** the SCR for the target with the **worst SCR**



THE MAXI-MIN PROCEDURE

- Finding the best maxi-min solution is difficult, finding the worst maxi-min is easy
- Project out enough of these worst dimensions from the finite dimensional transmit signal space → Converge to a good solution
- The SCR received by each target is defined by it's individual \mathbf{C}_i matrix:

$$SCR_i = \frac{E_a^i}{E_b} = \frac{\mathbf{S}'\mathbf{A}_i\mathbf{S}}{\mathbf{S}'\mathbf{B}\mathbf{S}} = \frac{\tilde{\mathbf{S}}'\mathbf{C}_i\tilde{\mathbf{S}}}{\tilde{\mathbf{S}}'\tilde{\mathbf{S}}}$$

- The smallest eigen value of a \mathbf{C}_i matrix - λ_i^{min} provides the worst SCR that the i^{th} target can receive, and the corresponding eigen vector $\tilde{\mathbf{e}}^{min}$ is the worst SCR solution for that particular target.
- Overall worst solution is then simply the $\tilde{\mathbf{e}}^{min}$ associated with the smallest of all individual minimum eigen values: $\lambda_{smallest}^{min} \rightarrow$ lower bound on SCR_{min}
- The lower bound on SCR_{min} is raised by restricting our solutions to an orthogonal subspace:

$$\mathbf{P}_{\perp}(l) = \mathbf{I} - \tilde{\mathbf{e}}^{min} \tilde{\mathbf{e}}^{min'}$$

$$\mathbf{C}_i(l+1) = \mathbf{P}_{\perp}'(l) \mathbf{C}_i(l) \mathbf{P}_{\perp}(l)$$



THE MAXI-MIN PROCEDURE



- Again look for the worst solution in the new subspace and project orthogonal to it
- All projections are orthogonal to each other \rightarrow the lower bound on SCR_{\min} monotonically increases
- We continue with this process till we are left with a single dimension – a vector
- This vector forms our optimal maxi-min transmit solution **S**
- The process is mathematically defensible – hence called the **True Maxi-min**



HEURISTIC MAXI-MIN



- Upper bound on SCR_{\min} is given by the smallest of all maximum eigen values for individual C_i matrices:

$$\lambda_{\text{smallest}}^{\max} = \min \{ \lambda_1^{\max}, \lambda_2^{\max}, \lambda_3^{\max}, \dots, \lambda_{N_t}^{\max} \} \geq SCR_{\min}$$

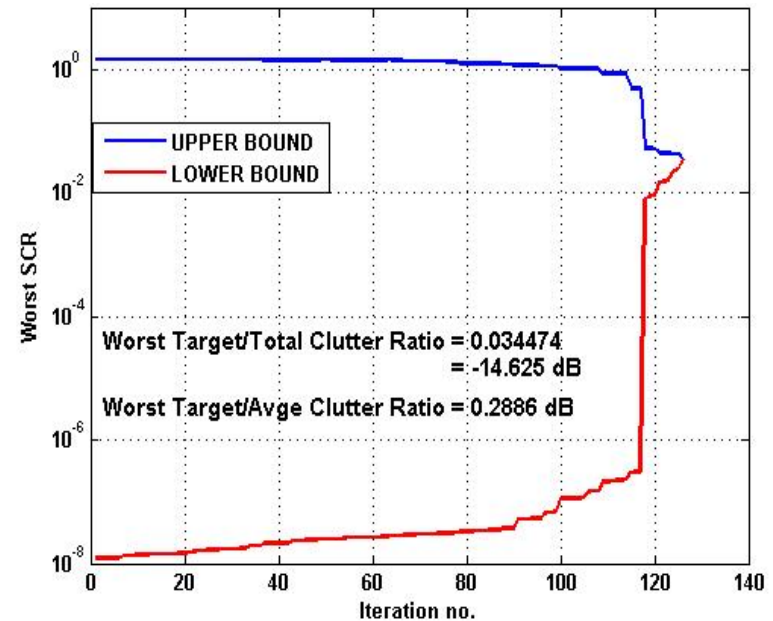
- The upper bound also comes down with every projection
- An alternative approach can be to try and keep the upper bound on SCR_{\min} as high as possible:
 - Find the weakest target ' t ', or the one with the smallest maximum eigen value $\lambda_{\text{smallest}}^{\max}$
 - Find the worst solution for this target \tilde{e}_t^{\min} i.e. the vector corresponding to the smallest eigen value of it's C matrix
 - Use this vector to form the projection matrix and repeat all steps as before
- This approach is called the **HEURISTIC MAXI-MIN** as it is not guaranteed to improve or preserve any bound, but is often seen to perform well - in fact most often even better than the earlier mathematically defensible **TRUE MAXI-MIN**



UPPER/LOWER BOUND CONVERGENCE



- The **TRUE** algorithms just aim to increase the lower bound on SCR_{\min} after each step
- However for most case, the upper bound turns out to be the more critical of the two bounds
- Thus the **HEURISTIC** algorithms are usually seen to be more effective than the **TRUE** algorithms

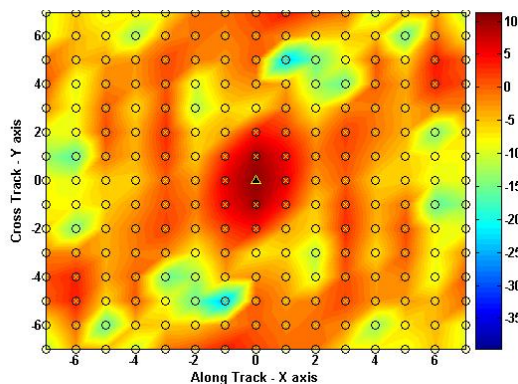




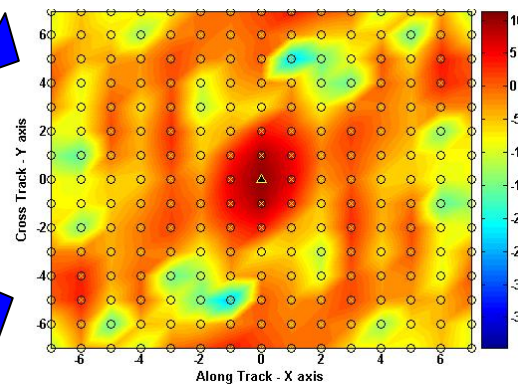
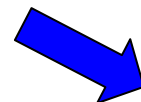
HEURISTIC AND TRUE SCR CONVERGENCE



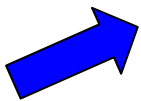
True Maxi-min



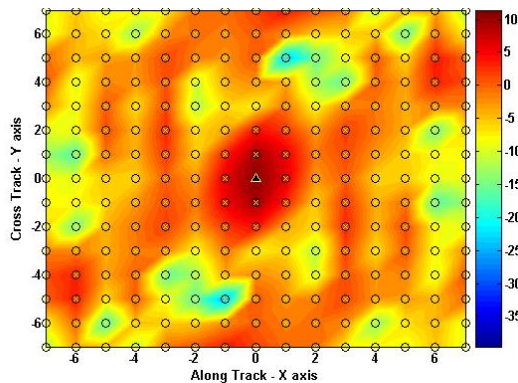
True SCR Convergence



Maximum SCR



Heuristic Maxi-min



Heuristic SCR Convergence



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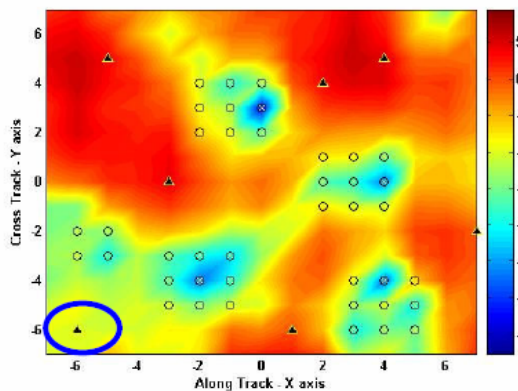
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ORPHAN PROBLEM SOLUTION

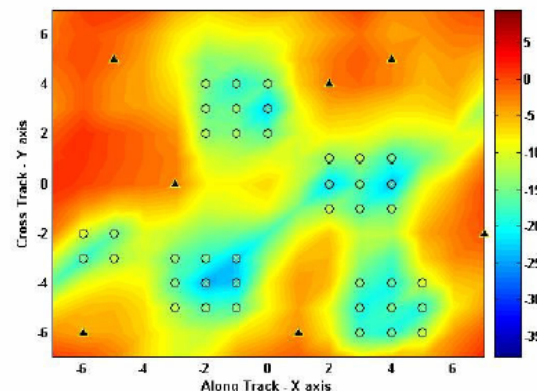


Maximum SCR



$SCR_{\min} = -2.73$ dB

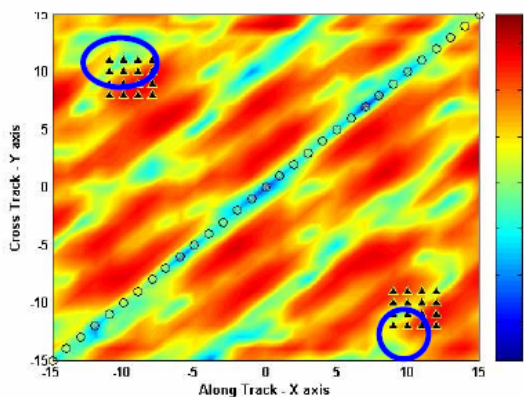
Heuristic SCR Convergence



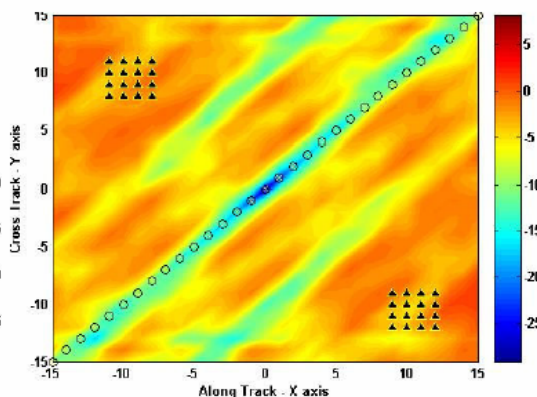
$SCR_{\min} = 11.38$ dB

▲ Target Locations

○ Clutter Locations



$SCR_{\min} = -2.21$ dB



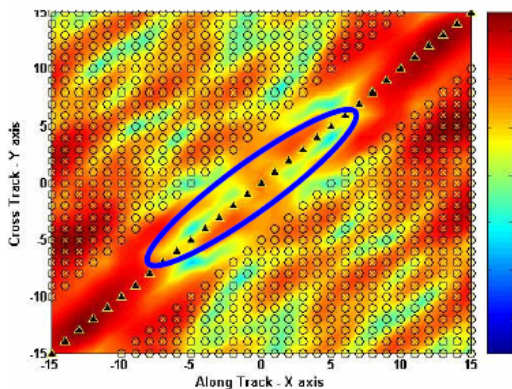
$SCR_{\min} = 8.84$ dB



ORPHAN PROBLEM SOLUTION

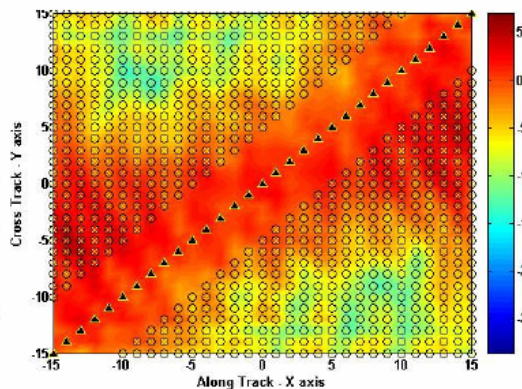


Maximum SCR

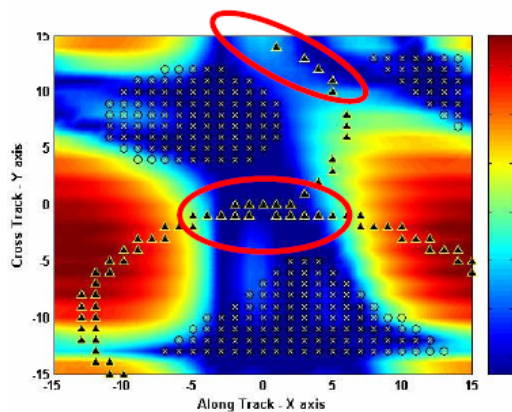


$SCR_{\min} = -4.54$ dB

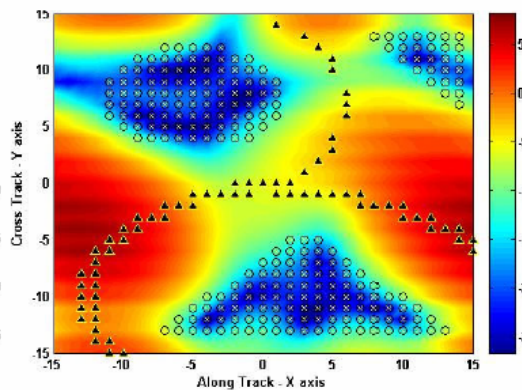
Heuristic SCR Convergence



$SCR_{\min} = 1.35$ dB



$SCR_{\min} = -13.82$ dB



$SCR_{\min} = 6.96$ dB

▲ Target Locations

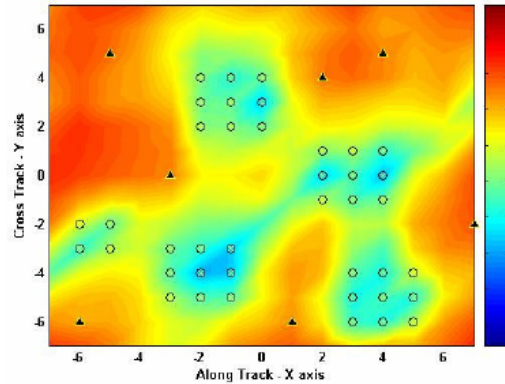
○ Clutter Locations



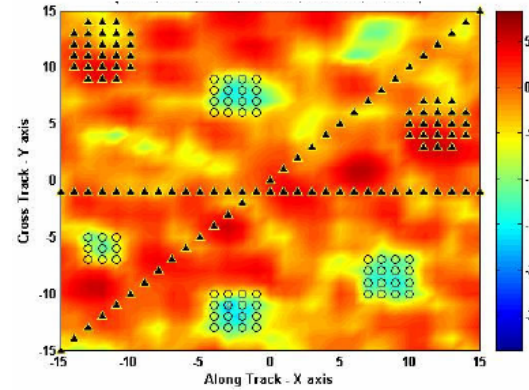
COMPARISON OF BOTH MAXI-MINS



Heuristic SCR Convergence



Heuristic SCR Convergence



Maximum SCR
 $SCR_{\min} = -2.73$ dB

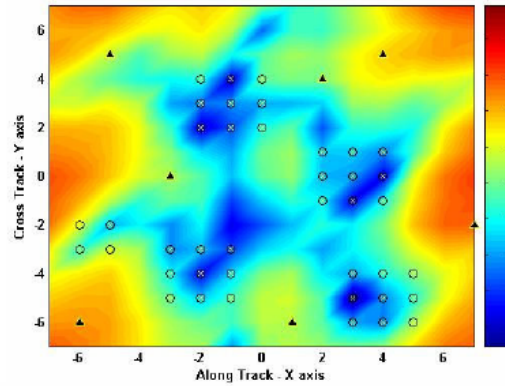
$SCR_{\min} = 11.38$ dB

BETTER

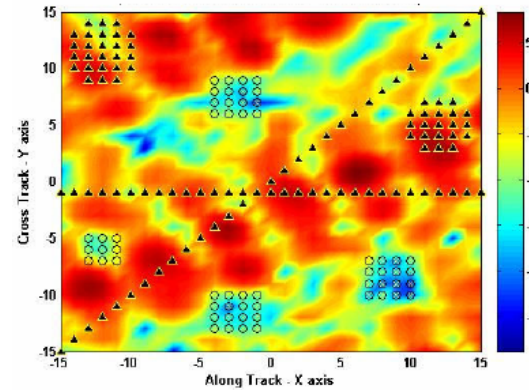
$SCR_{\min} = 2.51$ dB

Maximum SCR
 $SCR_{\min} = -4$ dB

True SCR Convergence



True SCR Convergence



$SCR_{\min} = 4.29$ dB

$SCR_{\min} = -1.47$ dB



PERFORMANCE ANALYSIS



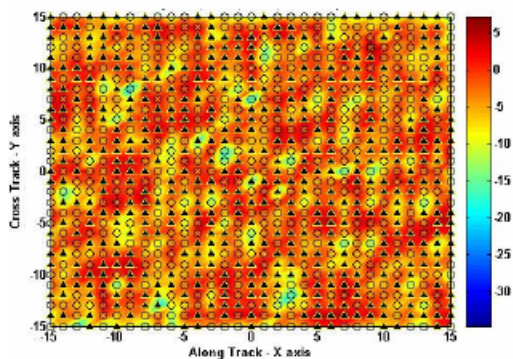
- **Champion Algorithm** → **Heuristic SCR Convergence**



DEPENDENCE ON TARGET SCENARIO

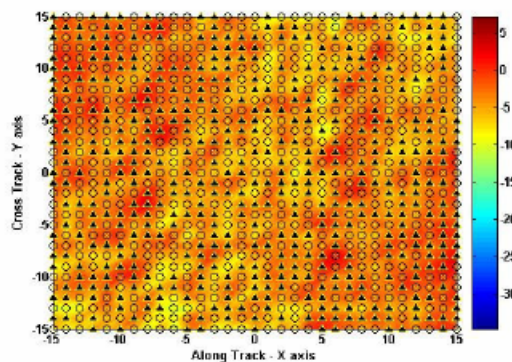


Maximum SCR



$SCR_{\min} = -20$ dB

Heuristic SCR Convergence

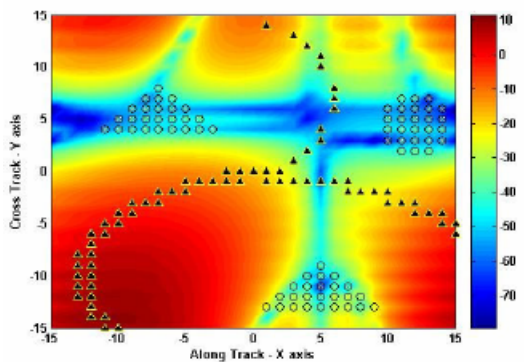


$SCR_{\min} = -8.4$ dB

▲ Target Locations

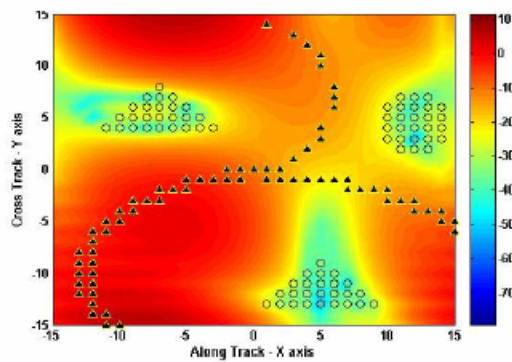
○ Clutter Locations

Maximum SCR



$SCR_{\min} = -23.64$ dB

Heuristic SCR Convergence



$SCR_{\min} = 7.19$ dB

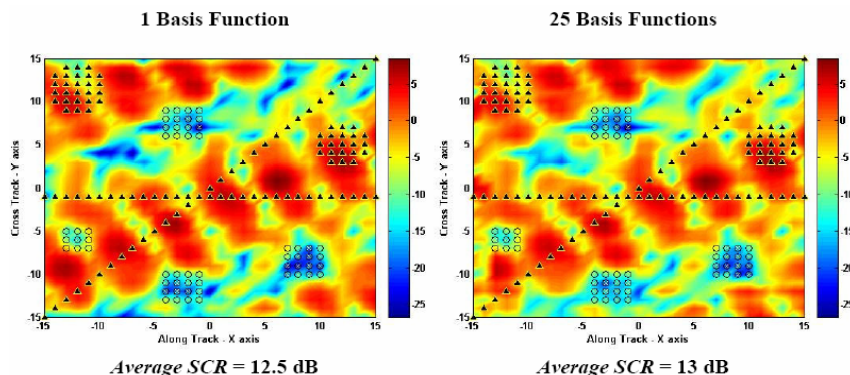


DEPENDENCE ON NUMBER OF BASIS FUNCTIONS (SIGNAL DIMENSION)

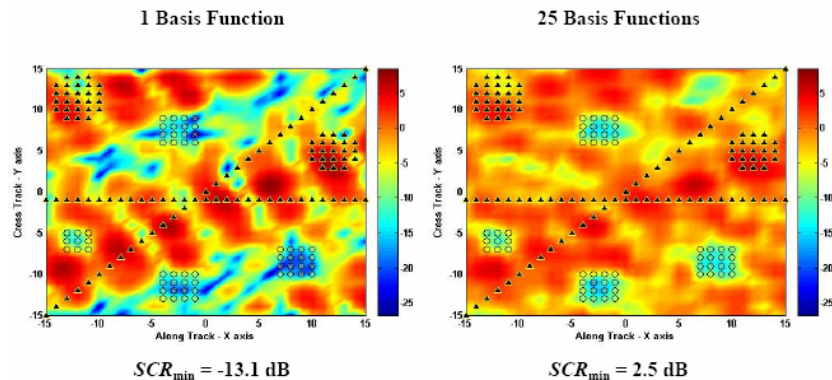


- Standard code performance is fairly insensitive to the increase in the number of time-frequency basis functions → spatial beamforming

- Maxi-min performance depends greatly on the number of basis functions → true space-time solutions



Maximum SCR



Heuristic SCR Convergence

Note: 1 basis function case essentially implies a spatial code

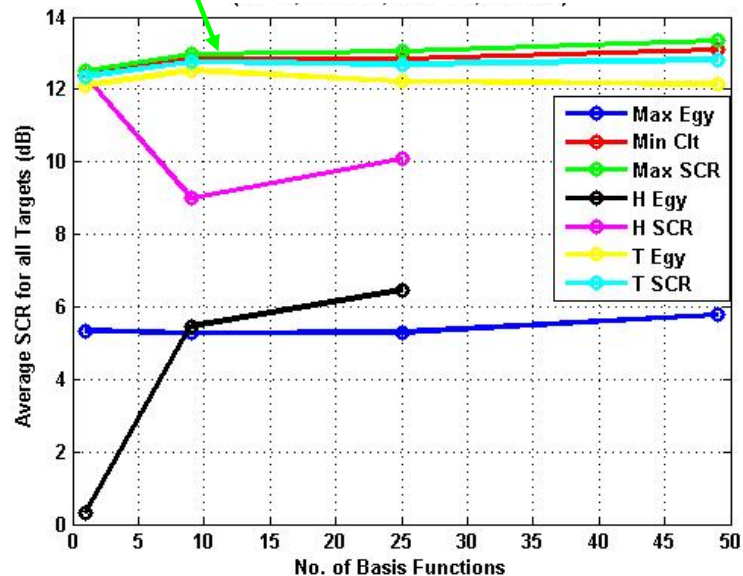


EFFECT OF INCREASING THE NUMBER OF BASIS FUNCTIONS ON DIFFERENT CODES



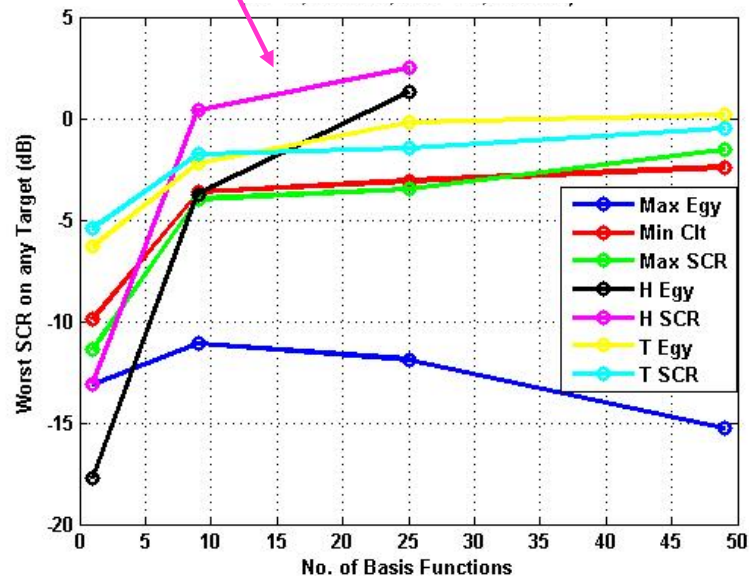
Not affected by increase in basis fn's – Spatial Beamforming equally effective

Average SCR



Greatly affected by increase in basis fn's – Spatial Beamforming not as good

Worst SCR





OUTLINE



❑ Introduction

- What are we doing?
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- How is our work different ?

❑ Execution + Results

- How do we do it?
- Is it good?
- If yes, how much is it better than the work done previously?

❑ More Observations

- How close do we reach to the goal we started with initially?
- Is there anything more to it?

❑ Conclusions and Future work

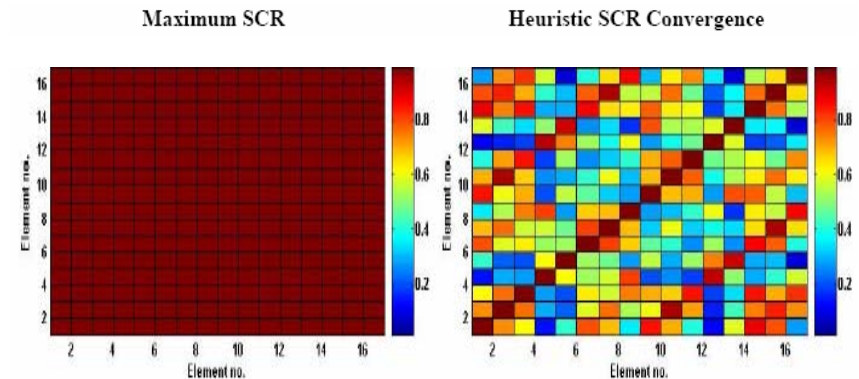
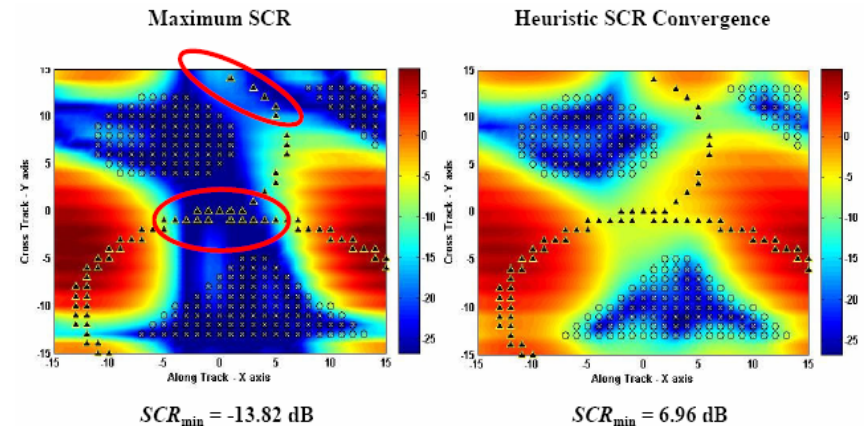
- What did we learn?
- What more can be done?



FORM OF THE TRANSMIT SIGNAL



- Recall our initial goal – To come up with optimal and true space-time codes, i.e. *different time-frequency signals propagate on different transmitters*
- How close do we reach to this goal ?
- Results show that for the basic codes the temporal signals on the different elements are perfectly correlated – i.e. **pure spatial beamforming**
- While for the maxi-mins the individual signals are typically only partially correlated – **true space-time operation**



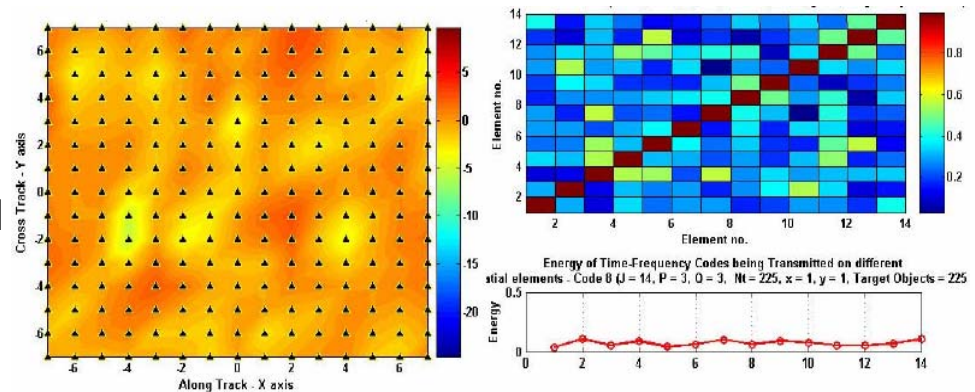


FORM OF THE TRANSMIT SIGNAL

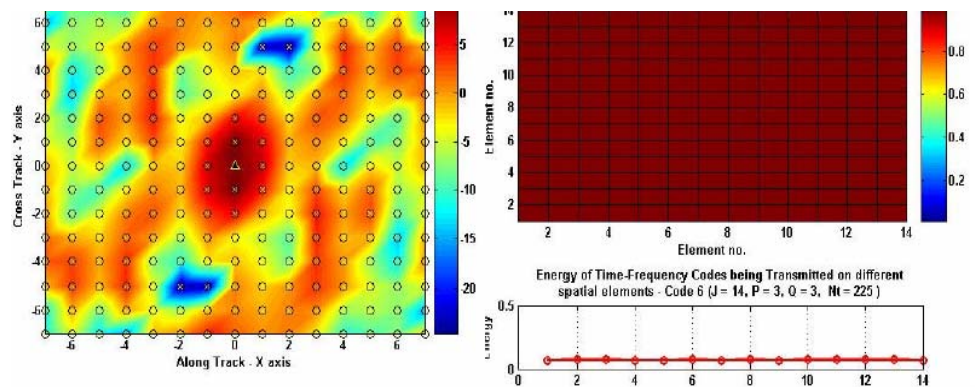


- Nothing in the algorithm tells it what solutions to converge to
- The structure that exists can be used for synthesizing both spatial and space-time solutions
- It just converges to the optimal solution for the particular case
- **Any other means to synthesize identical illumination patterns (except by transmitting dissimilar transmit signals on different antennas) is not possible**

Heuristic SCR Convergence



Heuristic SCR Convergence



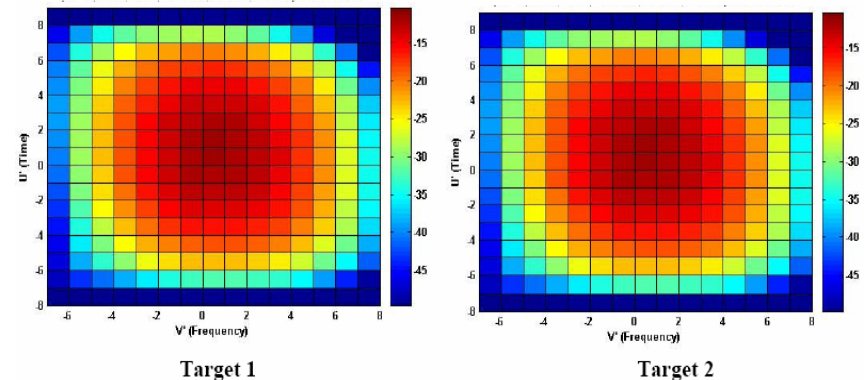


FORM OF THE INCIDENT SIGNAL ON TARGETS

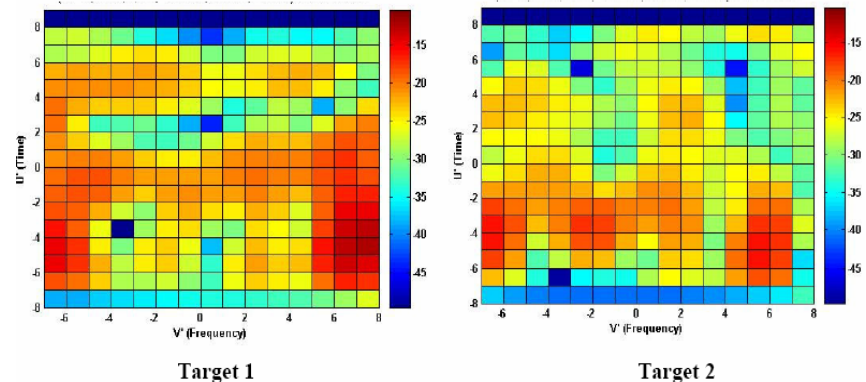


- More interesting than the form of transmit signal is the form of incident signals on the targets
- Resulting signal at any target is due to the coherent summation of all the individual temporal signals of different transmitters
- **Separable or spatial codes**
 - Individual temporal signals identical
 - *Resulting time-frequency spectra also identical at the different target locations*
- **Non-separable or space-time codes**
 - Coherent summation of *dissimilar* temporal signals of different antennas
 - *Time-frequency spectra completely different at different target locations*
- **Potential for target resolution**

Magnitude Response due to Maximum SCR (Spatial Code)



Magnitude Response due to Heuristic SCR Convergence (True Space-Time Code)





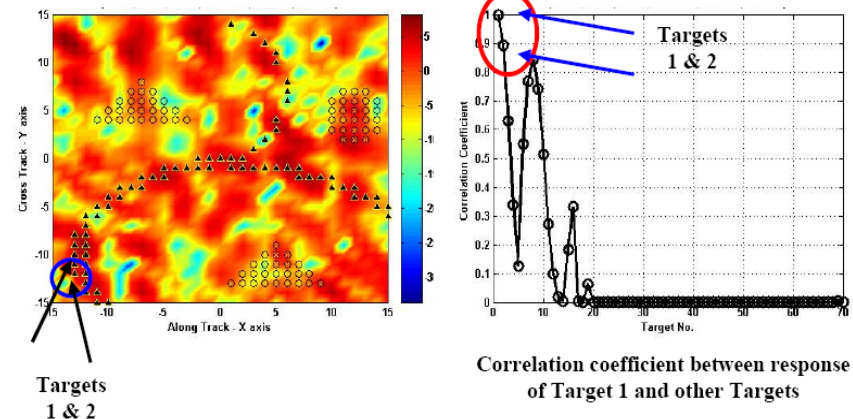
FORM OF THE SIGNAL INCIDENT ON TARGETS – ANOTHER PROSPECT



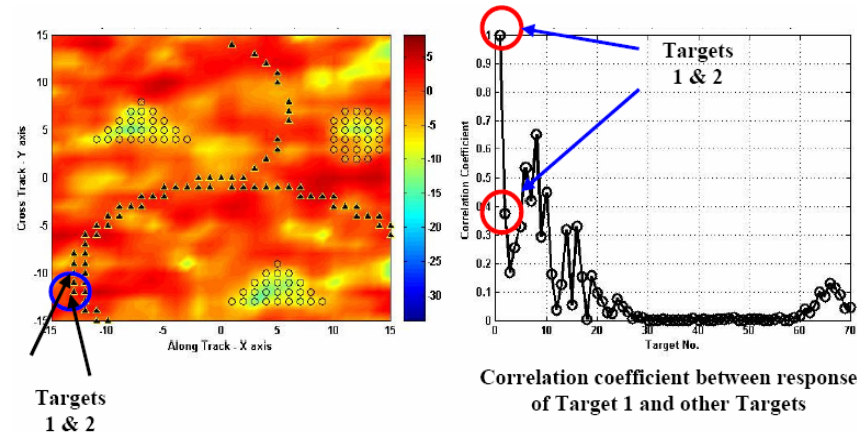
- Owing to the dissimilar magnitude responses, the cross-correlation between even those targets can be reduced that are non-resolvable in delay and doppler.

- Or the **main lobe of the time-frequency ambiguity function can be narrowed**

Maximum Energy (Spatial Code)



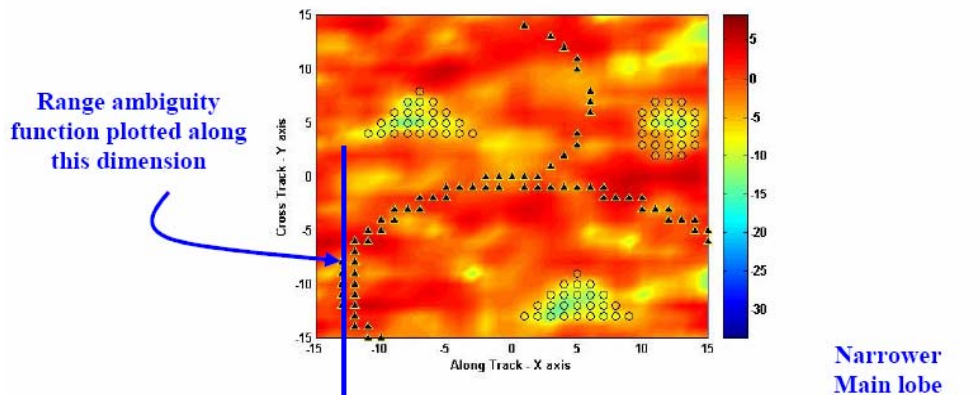
Heuristic SCR Convergence (Space-Time Code)



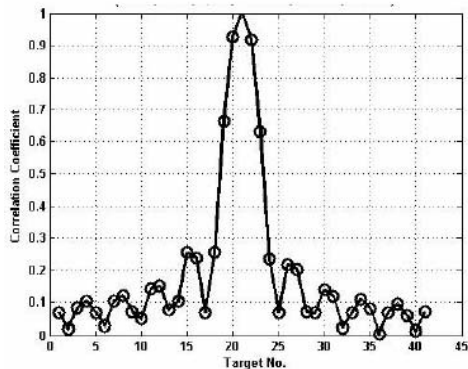


SPACE-TIME CODES

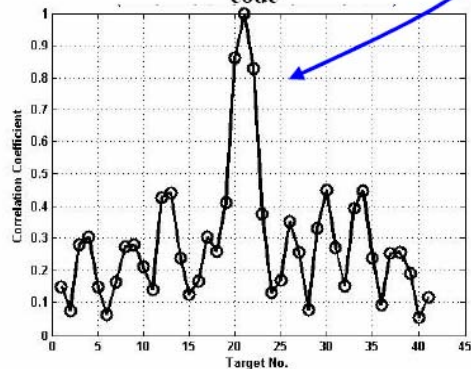
AMBIGUITY FUNCTION MAIN LOBE WIDTH REDUCTION



1 Basis Function – spatial code

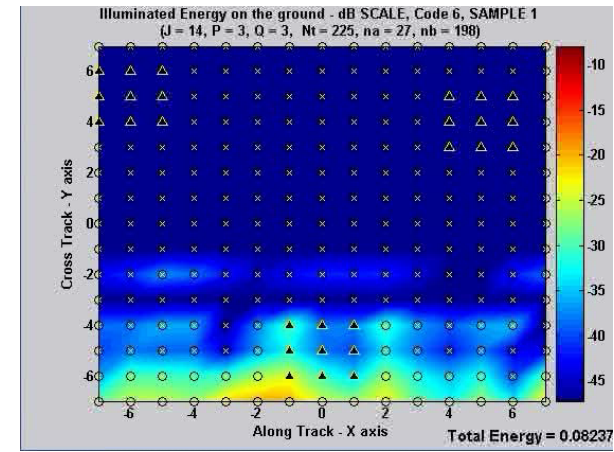
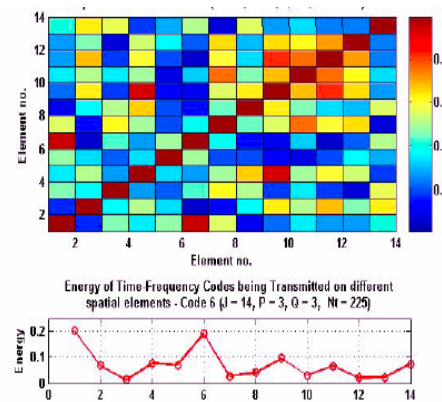
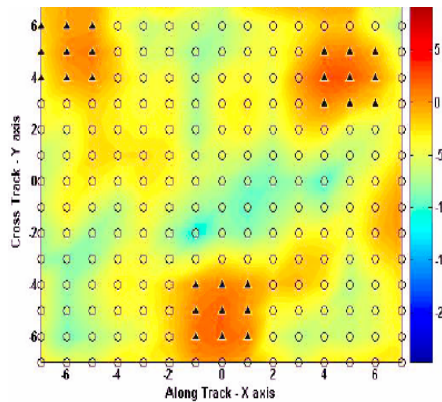


25 Basis Function – space-time code





SPACE-TIME CODES ONE FINAL PERSPECTIVE



- **True Space Time Codes** - the illumination pattern can change from pulse to pulse or frequency to frequency or even sample to sample, giving it additional versatility
- **Spatial Codes** - all antennas propagate the same temporal signal, and thus their coherent summation results in a constant illumination pattern with respect to time and frequency on the ground



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CONCLUSIONS



- **For exploiting the true potential of a multi-aperture radar, dissimilar temporal signals need to propagate on the different spatial elements.**
- **True space-time signals were successfully constructed for several illumination optimization criteria, using a union of vector-matrix radar models and linear algebraic techniques.**
- **Further, space-time codes that distribute energy appropriately across the available timewidth and bandwidth for different targets can potentially be constructed - and thus a range of radar requirements satisfied.**



FUTURE WORK



- **Improve the computational efficiency of our algorithms**
- **Adaptive Space-Time codes**
 - Information theoretic selection criteria
- **Hybrid or Multi-Mode operation**
 - Different radar modes across different spatial locations
- **GMTI and AMTI applications**
- **Unification with Space-Time Ambiguity Optimization**
- **Algorithm performance evaluation for other basis functions and spatial array arrangements.**



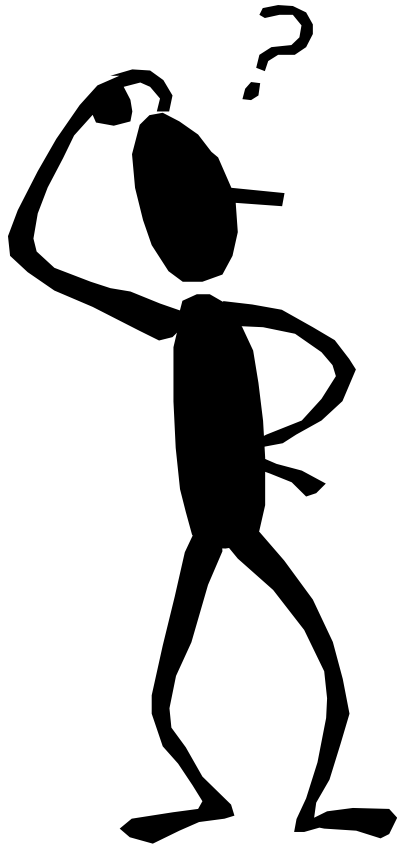
PUBLICATIONS



- Jim Stiles, Vishal Sinha, Atulya Deekonda. **“Optimal Space-Time Transmit Signals for Multi-Mode Radar”**, *2006 International Waveform Diversity & Design Conference, Lihue, Hawaii, USA, Jan 22-27*



THANK YOU!



QUESTIONS ??

