



Washington University in St. Louis

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# **Parallel Matlab Computation for STAP Clutter Scattering Function Estimation and Moving Target Estimation**

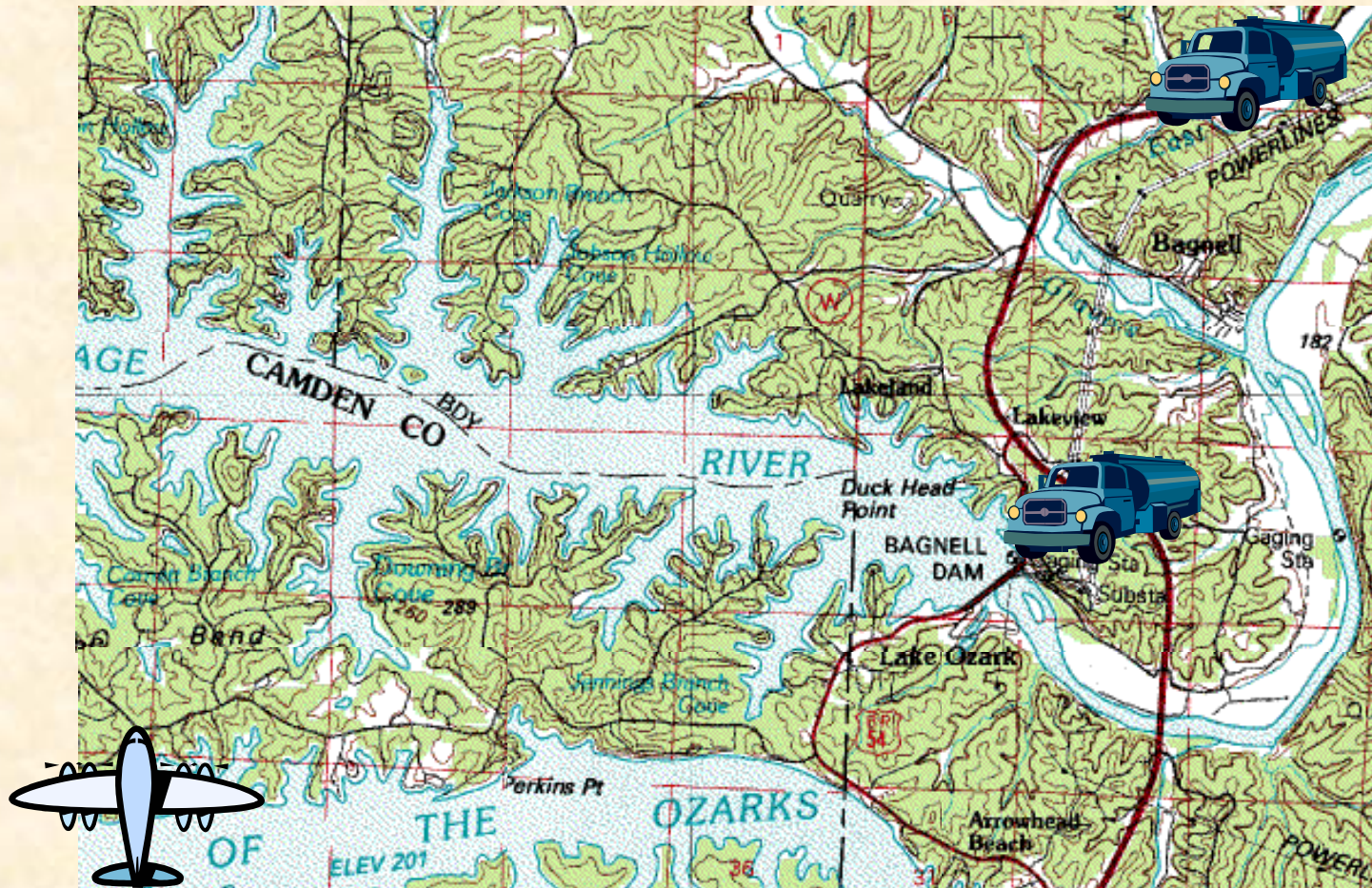
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# Context



**Problem:** Detect ground moving targets  
in the presence of ground clutter



# Context

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- Wide-area surveillance airborne radar
- Arbitrary flight path
- Multiple sensors and Doppler pulses
- Space-time adaptive processing (STAP)
  - Better knowledge of the clutter covariance matrix gives better detection performance

**Objective:** Estimate the clutter covariance matrix and detect moving targets

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# Approach

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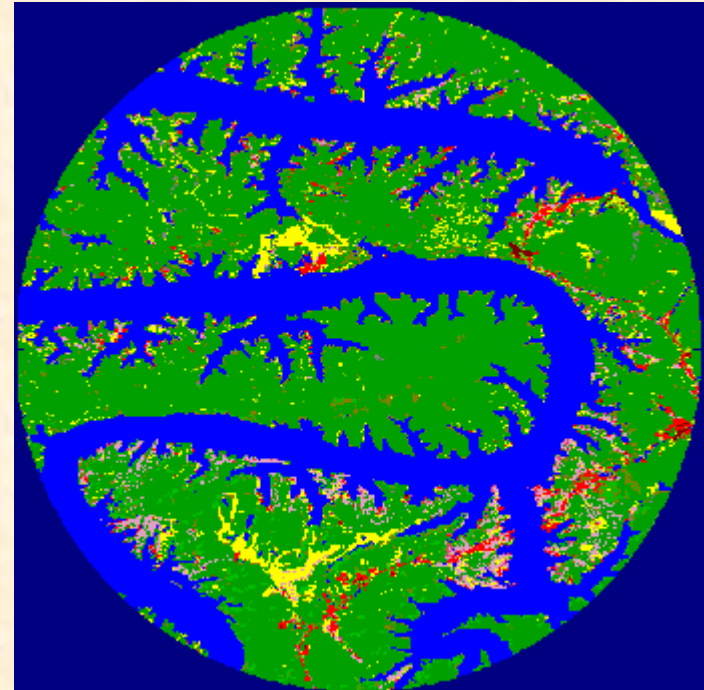
- Ground subdivided into pixels or ground patches
  - Known range and angle of each patch with respect to airborne platform
  - Known illumination pattern
  - Received data: sum of returns from targets and all patches on the ground
  - Prior knowledge is available:
    - Digital Terrain Elevation Maps
    - Land use information
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# Terrain Simulation

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- **Region of Interest**

- Lake of the Ozarks
- 15 km diameter
- 197,316 pixels
- 30m resolution



# Datasets

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- Obtained from USGS Seamless Data Server
    - 30m resolution
  - Digital Elevation Model
    - Used for modeling geometry
  - Land Use
    - Scattering function based on 21 classes of land cover
      - 9 primary classes
        - Water, Developed, Barren, Forested Upland, Shrubland, Non-Natural Woody, Herbaceous Upland Natural/Semi-natural Vegetation, Herbaceous Planted/Cultivated, Wetlands
      - Each class contains one or more categories, e.g.
        - Open Water, High-Intensity Residential, Deciduous Forest, Row Crops
    - Scattering function chosen arbitrarily for simulation
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# Coordinate Systems

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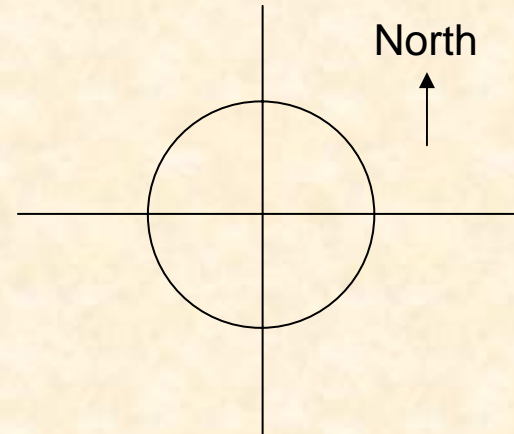
- Datasets referenced in spherical coordinates
    - Latitude, Longitude, Elevation
  - Convert to Cartesian Coordinates
    - Simpler to use over small region
    - Computations can be made independent of Earth model
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# Coordinate Conversion

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- First Stage
  - Origin at Earth's center
  - Use Geodetic Reference System 1980 (GRS80)
- Second Stage
  - Move origin to center of region of interest
  - Elevation along Z-axis
  - North along positive Y-axis

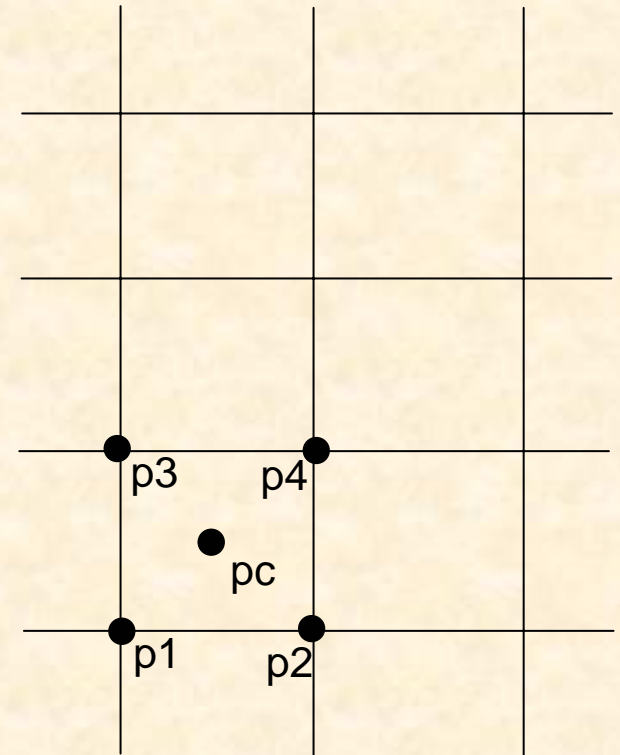




# Coordinate Systems

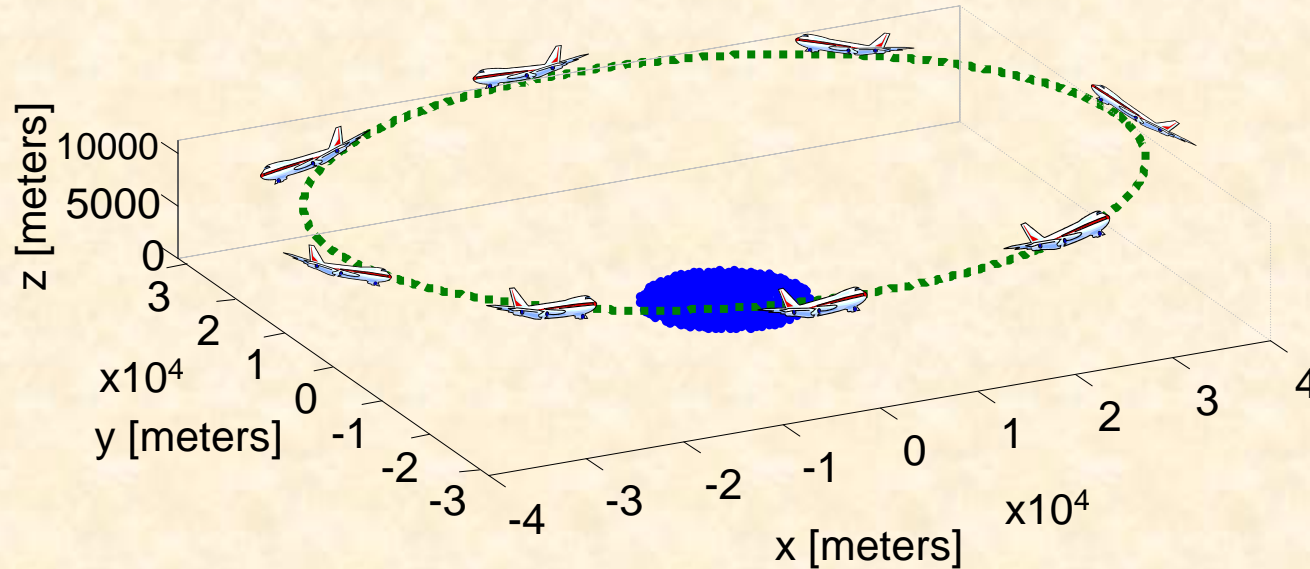
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- Adjacent data samples grouped into patches
  - Each patch, or pixel, contains:
    - Location for each corner
    - Location of center
    - Scattering function
    - Normal vectors
- 197,316 pixels in all



# Simulation Setup

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- Platform moves around region of interest
    - Actual flight path is arbitrary
  - Eight looks
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# SIMULATION PARAMETERS

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- Platform
    - Flies in circular path around region
    - Radius 25 km
    - Altitude 7 km
    - 8 different viewpoints
  - Radar
    - $f_c$ : 10 GHz
    - BW: 10 MHz
    - PRF: 2 KHz
    - Pulses per CPI: 38
    - ULA elements: 12
    - Range gates: 990
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# Geometry Parameters

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- Geometry dependent parameters required for simulation
    - Range to each pixel
    - Projected area of each pixel
      - Incident energy incorporates range and projected area of patch
    - Occluded pixels
      - Patches hidden from radar are removed using Z-buffer algorithm
        - Patches sorted by distance from radar
        - Any patch facing backwards or directly behind another is removed
    - Angle between platform's velocity vector and line of sight to each pixel
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# Datacube Generation

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- Received data from a single patch

Return from  $n^{\text{th}}$  path is a random variable

$$u_k(n) \sim \mathcal{CN}(0, \lambda_k(n)\sigma_n)$$
$$\mathbf{z}_k = \mathbf{s}_k + \mathbf{n}_k + \sum_{n=1}^N u_k(n) \mathbf{a}_k(n)$$

return from targets

receiver noise

$$\mathbf{n}_k(n) \sim \mathcal{CN}(0, \epsilon \mathbf{I}_M)$$

Radar response vector

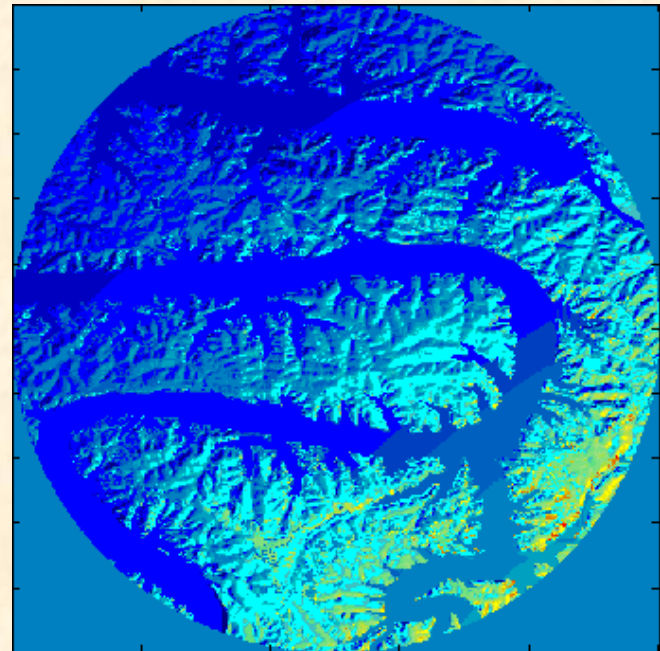
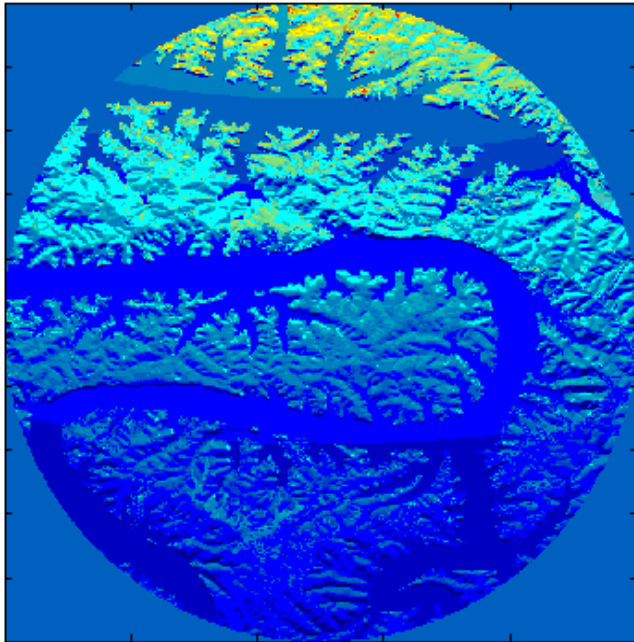
The diagram illustrates the equation for the received data vector  $\mathbf{z}_k$ . It shows the sum of three components: the return from targets  $\mathbf{s}_k$ , receiver noise  $\mathbf{n}_k$ , and a sum of  $N$  paths. Each path  $n$  consists of a random variable  $u_k(n)$  multiplied by the radar response vector  $\mathbf{a}_k(n)$ . Annotations with arrows point from the text labels to the corresponding terms in the equation: 'return from targets' points to  $\mathbf{s}_k$ , 'receiver noise' points to  $\mathbf{n}_k$ , 'Radar response vector' points to  $\mathbf{a}_k(n)$ , and the text 'Return from  $n^{\text{th}}$  path is a random variable' points to  $u_k(n)$ . The distribution for  $u_k(n)$  is given as  $\mathcal{CN}(0, \lambda_k(n)\sigma_n)$  and for  $\mathbf{n}_k(n)$  as  $\mathcal{CN}(0, \epsilon \mathbf{I}_M)$ .

- Response at a single range gate
    - Sum over all patches in range gate
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# ILLUMINATION

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Illumination from different looks



# Scattering Function Estimation

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- Prototype designed and tested first
    - Implements EM algorithm
    - Uses a Small-Scale Dataset with 554 pixels
  - EM algorithm requires response vector for each pixel, in each look
    - For Small-Scale Simulation
      - 2,020,992 complex doubles
      - 30.8 MB of data
  - Large-Scale Simulation contains 197,316 pixels
    - 719,808,768 complex doubles
    - 10.73 GB
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# Memory Reducing Techniques

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- Maintain only the Doppler and spatial vectors
    - Compute Kronecker product as needed
    - Reduces requirements to 1.17 GB
  - Lookup Table
    - Finely sampled table containing Doppler and spatial vectors
    - Indexed by a single value
    - Further reduces memory requirements
      - 10.64 MB when using a 10,000 entry table
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# The Need For Parallelism

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- The EM Algorithm can be parallelized in multiple ways
    - Across looks
    - Across range gates
  - Parallelism improves the algorithm
    - Significant speedup in processing time
    - Additional physical memory available
      - Only 150 MB needed per look for the response vectors (250 MB when all other necessary data are included)
    - More effective cache
      - Possible gain when using a Lookup Table
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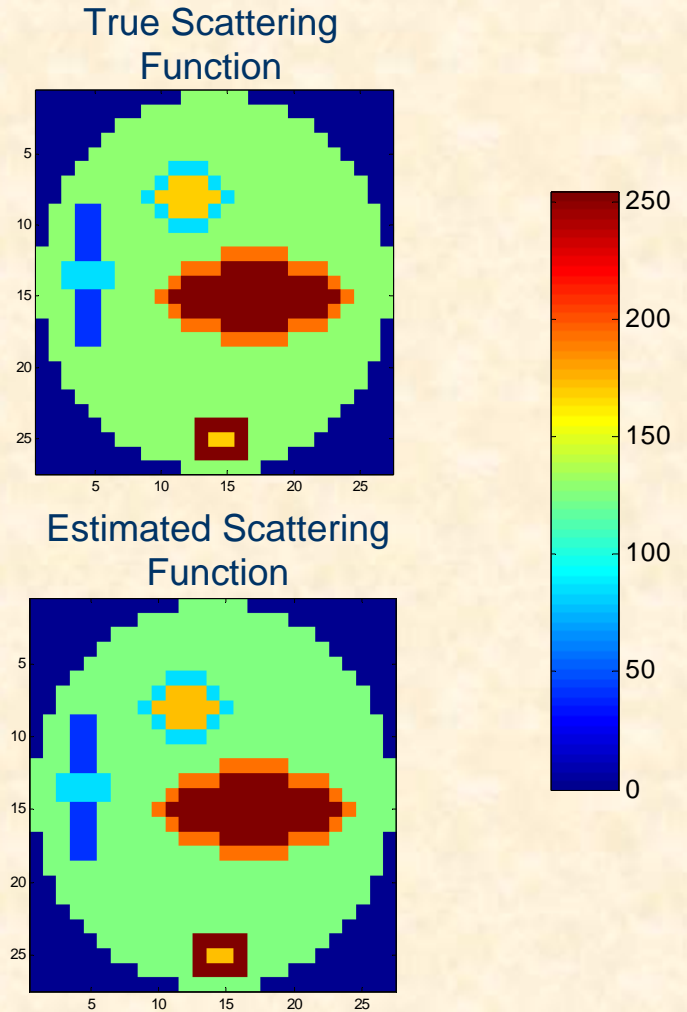
# Parallelism Using MatlabMPI

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- MatlabMPI provides parallel interface
    - Allows passing of messages between multiple systems that share a file system
  - Use 9 parallel threads (1 master, 8 slaves)
    - Slaves perform iterations of the EM algorithm on a single look
    - Master provides slaves with data and collects results from each iteration
  - Messages only sent at beginning and end of each iteration
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# Results

- Small-Scale Simulation
  - Provides results identical to prototype version
  - Runs 4% slower than non-parallel version
    - Computation for a single look is too fast to gain from parallelism
    - Message passing overhead too large
    - Not a problem for full-scale simulation



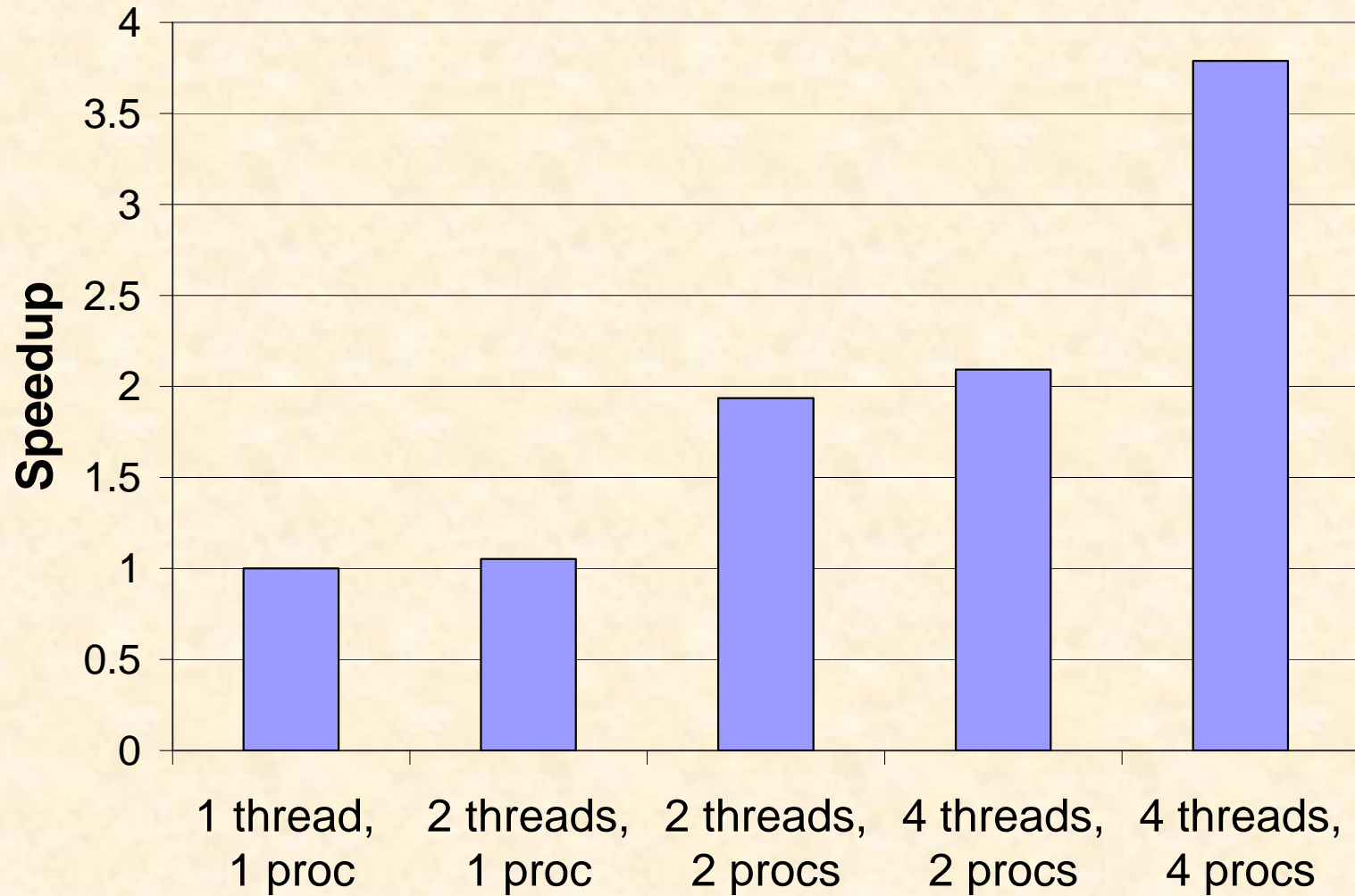
# Results

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- Full-Scale Simulation
    - Does not use Lookup Table
    - To avoid large messages, some inputs are read from disk
  - Execution Environment
    - 2.4 GHz Pentium IV processors w/ hyperthreading
    - 1 GB RAM each
    - 4 nodes
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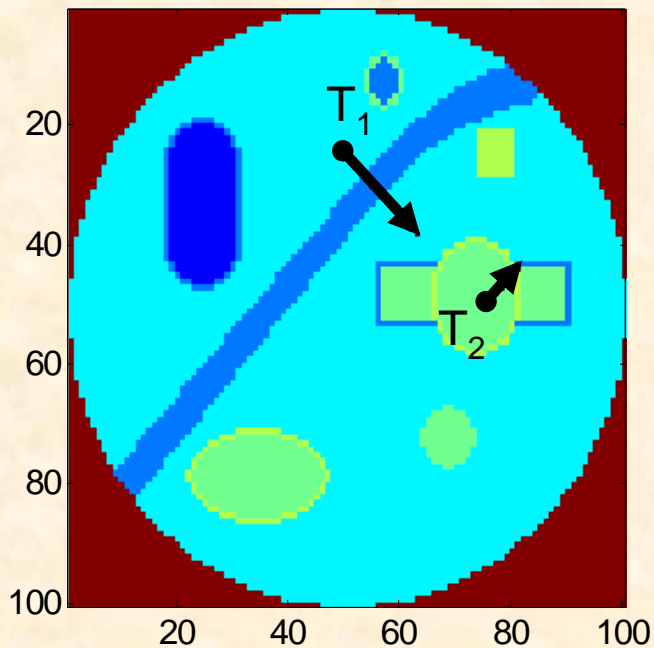


# Results



# Detection Example

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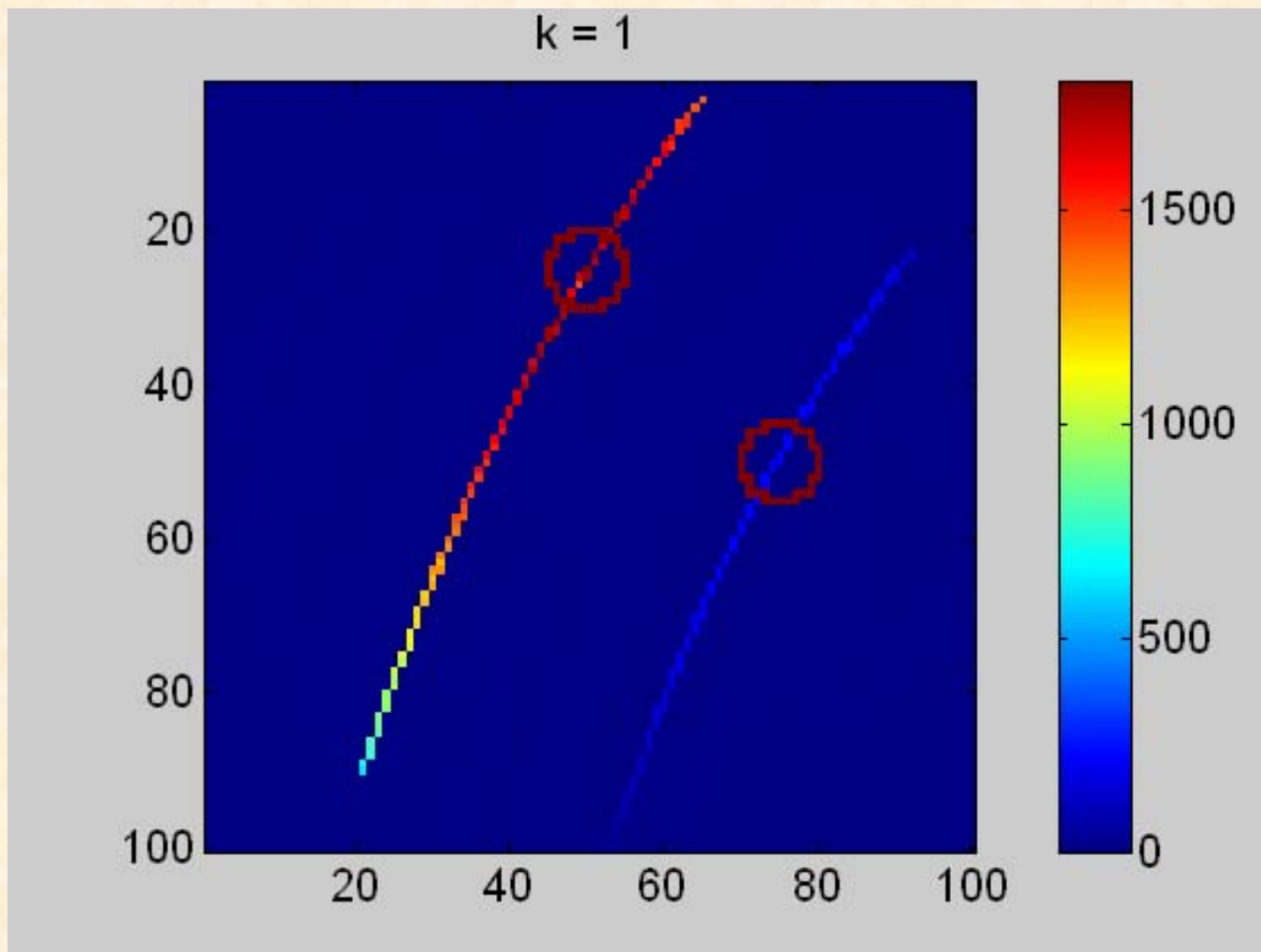


- Two artificial targets
- In small-scale environment
- Binary detection problem

$$\mathbf{z}_k = \begin{cases} \mathbf{s}_k + \mathbf{A}_k \mathbf{u}_k + \mathbf{n}_k & H_1 \\ \mathbf{A}_k \mathbf{u}_k + \mathbf{n}_k & H_0 \end{cases}$$

- Use adaptive matched filter
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# Adaptive Matched Filter Detector



# Current and Future Work

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- Completing the Full-Scale Simulation
    - Long runtimes are still a problem
  - Moving Target Estimation on Full-Scale System
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