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ALPHATECH, Inc.

KASSPER Workshop Presentation

Incorporating Past CPI Data into Knowledge-Aided STAP Processing

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Problem Summary



- **Standard STAP performance is degraded in nonhomogeneous terrain environments**
 - Single CPI training data is limited, due to range-varying clutter returns and limited range resolution
- **Past CPI data cubes contain information that, if exploited, could reduce training errors**
 - Strong clutter returns from urban areas, highly sloped terrain, etc. will be correlated over multiple CPIs
- **Challenges remain in utilizing past data:**
 - Location of clutter ridge in Doppler-angle space varies with CPI, due to varying platform geometry
 - Direct averaging of past-CPI samples into covariance estimate is not effective
- **Develop a technique that overcomes these challenges and improves clutter estimates (and consequently, STAP performance)**

Summary of Approach for Incorporating Past-CPI Data



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- **Build an earth-referenced “clutter map” based on multiple CPIs**
 - Requires registration of each CPI to a geodetic coordinate system
 - Estimate clutter reflectivity separately on each CPI
 - Average clutter reflectivity estimates over several CPIs to obtain smoothed estimates
 - May be generalized to parameters other than reflectivity
- **Form predicted clutter estimates on current CPI by indexing into clutter map**
 - Calculate range-dependent covariance matrices
- **Incorporate predicted covariance matrices into STAP weight vector calculation**
 - Knowledge-aided pre-whitening
 - Algorithm applied to the KASSPER Data Set 2

Steps in Formation of a Geodetic Clutter Map Based on Past-CPI Data



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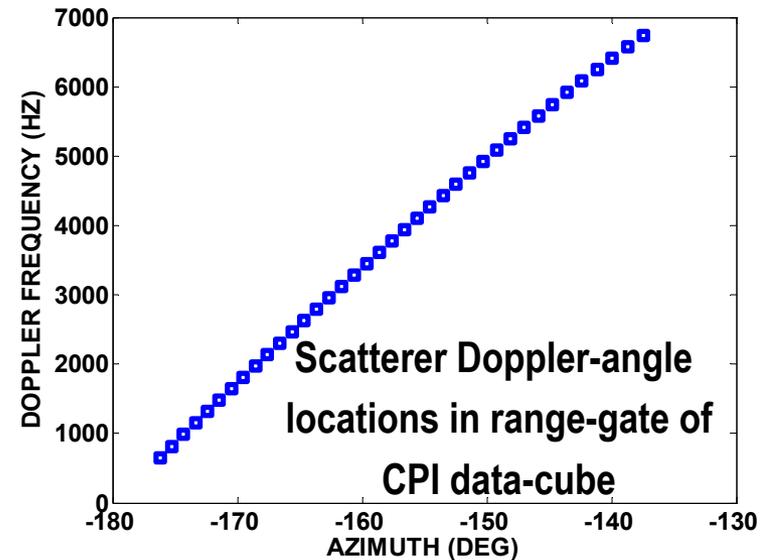
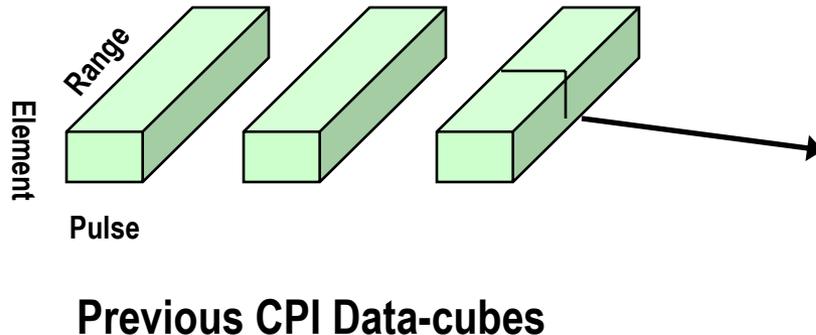


- 1. In each range-gate of past-CPI data-cube, define a set of point scatterers along the clutter ridge**
- 2. Register the scatterer locations to an earth-based coordinate system**
- 3. Estimate scatterer strengths using measured data**
- 4. Form an earth-based clutter map**

•Detailed description of these steps follows

1. Define Clutter Scatterer Locations

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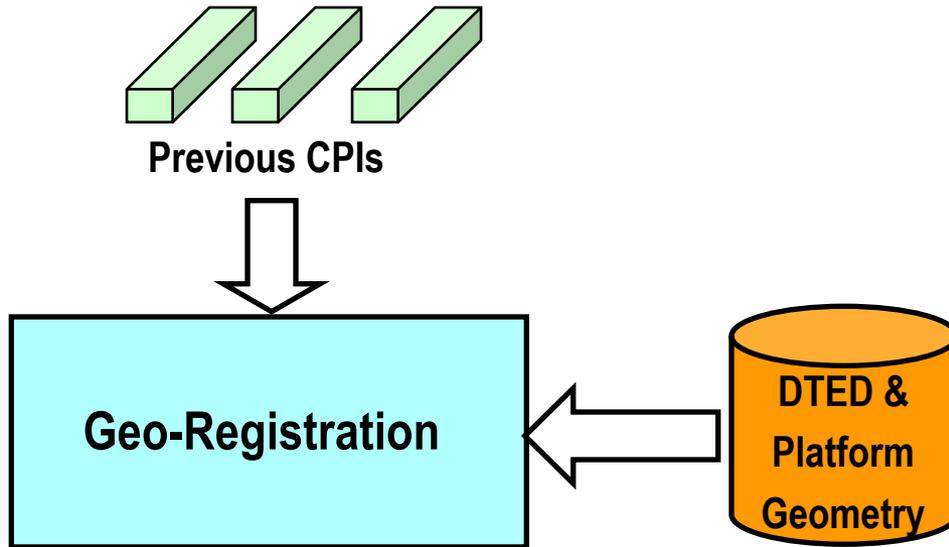


- In each range-gate of past-CPI data-cube, define a set of point scatterers along the clutter ridge
 - Defined by range/Doppler locations
 - Choose Doppler extent to span area of significant clutter (may be several Doppler ambiguities)

2. Geo-Register Scatterer Locations



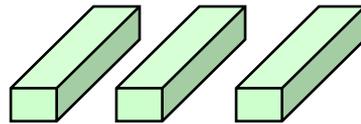
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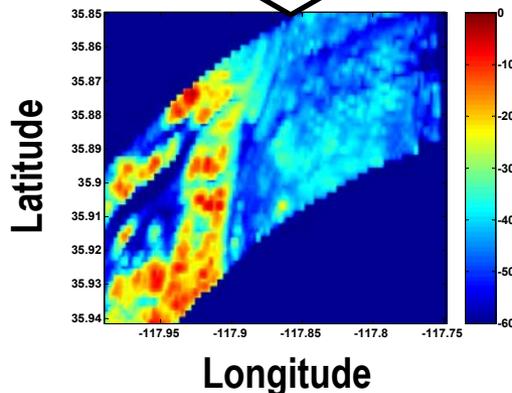
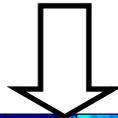
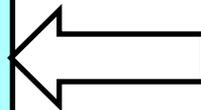
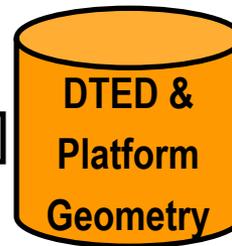
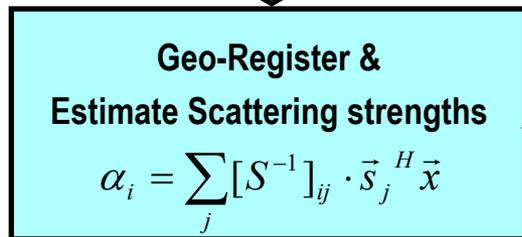
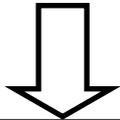
- **Register the scatterer locations to an earth-based coordinate system**
 - Requires computing intersection of range sphere, Doppler cone, and the earth's surface
 - Range sphere and Doppler cone defined by platform location and velocity
 - Earth's Surface is defined by digital terrain database (DTED)
 - Iterative approach required, as radius of earth surface varies with ground location

3 & 4. Estimate Clutter Reflectivities and Form Clutter Map

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Previous CPIs



- Estimate scatterer strengths using measured data
 - Space-time responses defined by azimuth and elevation angles, & Doppler
 - Strengths estimated using a minimum squared error criterion

- Form an earth-based clutter map
 - Normalize scatterer strengths by their area on the ground to obtain reflectivity estimates
 - Each cell of the map is an averaged reflectivity over multiple CPIs

Predicting Clutter Statistics of Current CPI Using Geodetic Clutter Map



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- **Define current-CPI scatterers along the clutter ridge**
 - Geodetic locations calculated as described on previous slide
- **Compute reflectivity of scatterers by indexing into geodetic clutter map**
- **Form covariance matrices by adding contributions from each scatterer:**

$$\tilde{R} = \sum_i \rho_i \cdot A_i \cdot \vec{s}_i \vec{s}_i^H$$

$\rho_i \equiv$ reflectivity of scatterer i

$A_i \equiv$ area of scatterer i

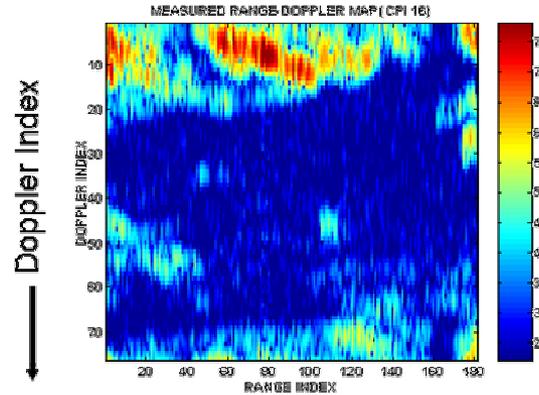
$\vec{s}_i \equiv$ space - time steering vector of scatterer i

Measured Range-Doppler Spectra of Four Different CPIs of KASSPER Data Set 2

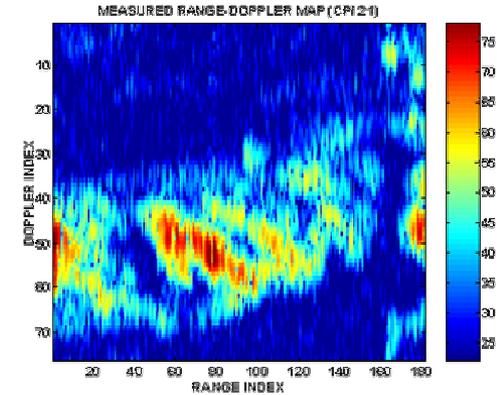


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Measured (CPI 16)



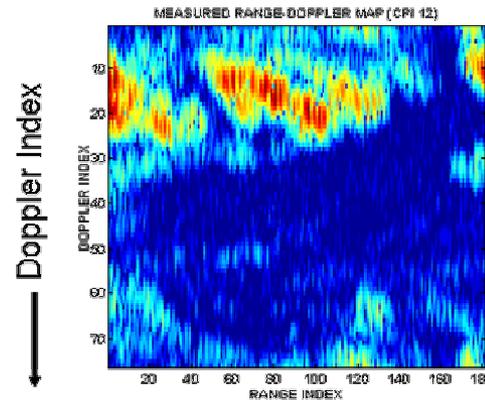
Measured (CPI 21)



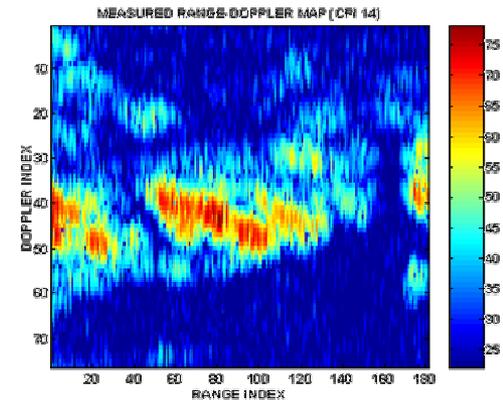
Mis-registration of clutter ridge is evident in these plots

Range-gate →

Measured (CPI 12)



Measured (CPI 14)



Range-gate →

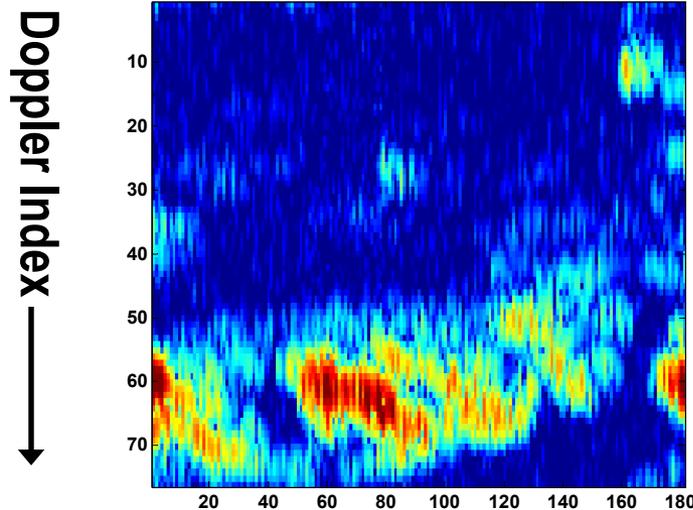
Comparison of Measured Spectrum With Predicted Reflectivity Map



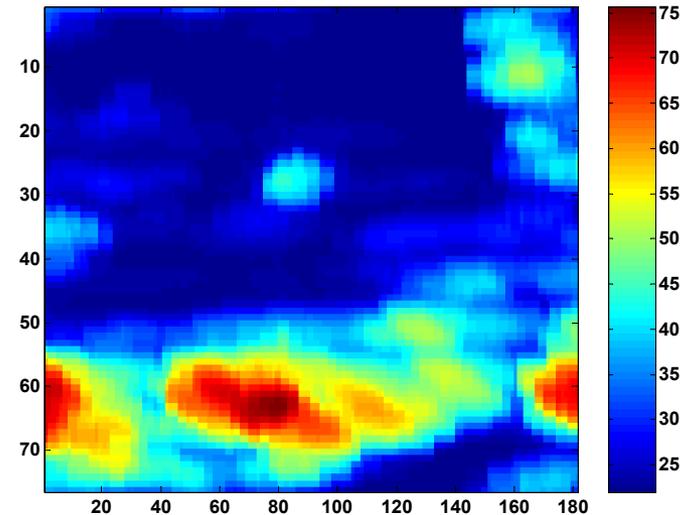
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Measured (CPI 22)



Calculated (CPI 22) based on 200m resolution clutter map



Range-gate →

- Procedure has registered the different CPIs and produced a smoothed estimate of the clutter spectrum
 - Results above reflect a 200 meter resolution map (approximately matching the width of a Doppler filter on the ground)
 - Allows improved prediction of range varying clutter statistics over standard range-averaging procedure, which smoothes features over much wider range



Incorporation of Calculated Covariance Matrices Into STAP Processing



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- **Using covariance matrices directly in STAP vector ($w=R^{-1}s$) does not produce good performance**
 - Due to steering vector errors, varying internal clutter motion (ICM), etc.
- **Calculated matrices were used to pre-whiten the data using the method of [1], with the following additional techniques employed:**
 - A covariance taper [2] was applied prior to pre-whitening to account for ICM
 - A reduced degree of freedom (DOF) post-Doppler, extended-factored (or “multi-bin”) algorithm was employed
 - A separate pre-whitening transformation was performed in each Doppler filter

[1] J. Bergin, J. Guerci, P. Techau, C. Teixeira, “Space-Time Beamforming with Knowledge-Aided Constraints”, Adaptive Sensor Array Processing Workshop 2003, 11-13 March 2003

[2] J. R. Guerci, “Theory and Application of Covariance Matrix Tapers for Robust Adaptive Beamforming”, IEEE. Trans. Signal Processing, Vol. 4, No. 4, April 1999



Evaluation of Algorithm Performance



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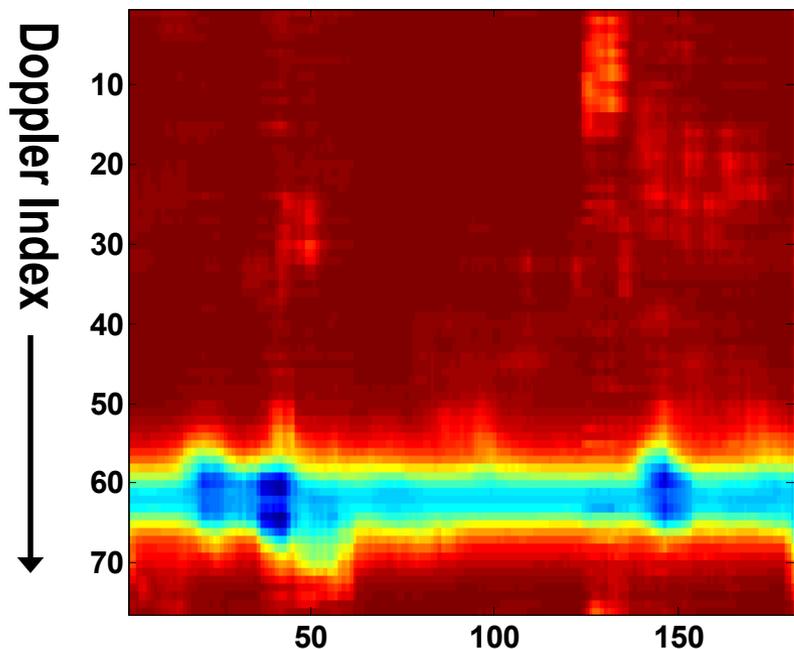
- **Performance of standard STAP vs. KASTAP was evaluated on KASSPER Data Set 2**
 - 5 adjacent filters employed in the post-Doppler algorithm, yielding 60 DOFs
 - Training window extent was 3 times number of DOFs (test cell excluded from covariance estimate)
 - Diagonal loading at noise floor employed
- **SINR Loss versus range and Doppler**
- **Detections and false alarms**
 - Adaptive matched filter (AMF) statistic used to determine threshold crossings
 - Receiver operating characteristic (ROC) curves obtained over 23 CPIs
- **Additionally, preliminary tracking performance was obtained using ALPHATECH's MHT tracker testbed**
 - Two-target scenario over 23 CPIs

Comparison of SINR Loss (clutter-only training, CPI 22)

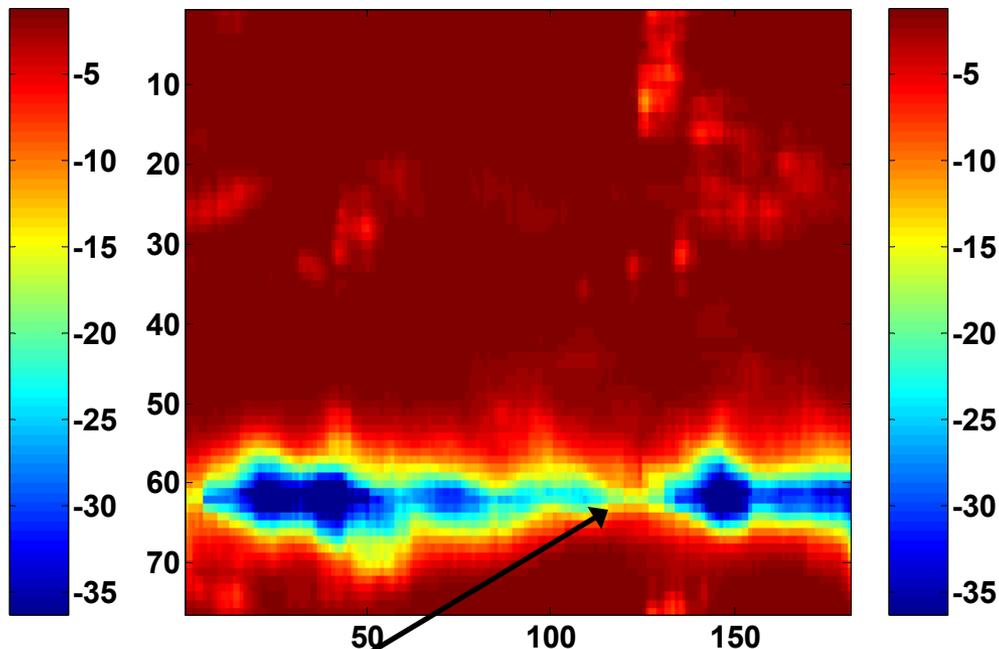


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Standard STAP



Knowledge-aided STAP utilizing past-CPI data



•SINR Loss improved significantly, resulting in reduced MDV

Contamination of Training Data by Targets and Discretets



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- **The large number of targets (~1200) in KASSPER Data Set 2 challenge the STAP training procedure**
- **Masking of detections was investigated to reduce contamination of training data by targets and discretets**
 - **With such a large number of targets, range-only masking causes excessive reduction of training data**
 - **In a post-Doppler framework, separate masking of the detections can be performed in each Doppler filter**
 - **The reduction in training data is much less severe**
 - **This Technique can be incorporated into a tracker feedback loop (“Tracker-in-the-Loop”)**

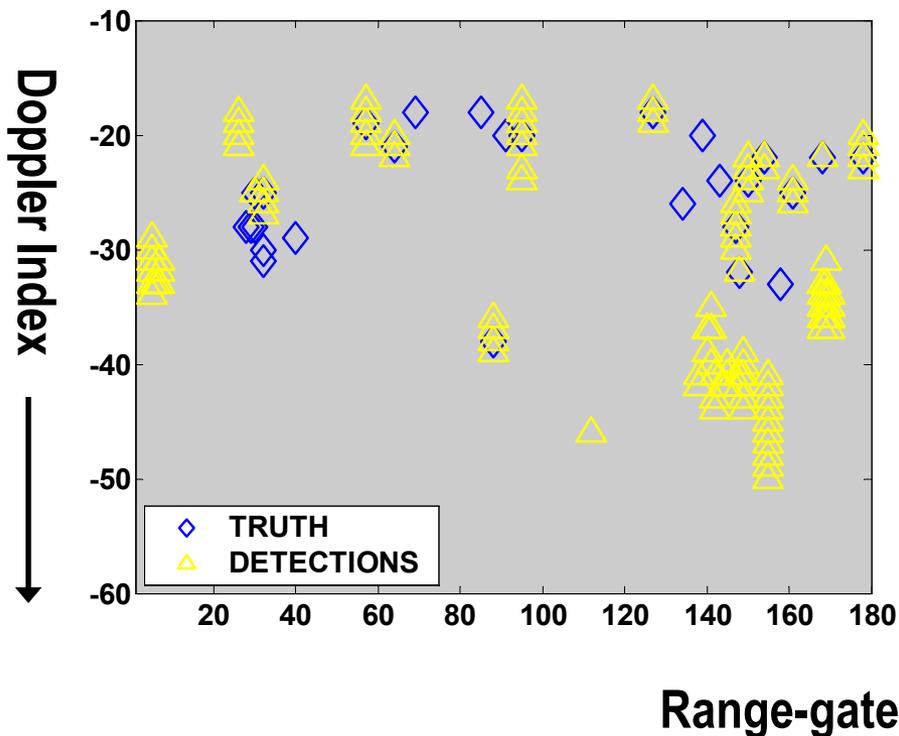
Detection vs. False Alarms (high target density region of CPI 67)



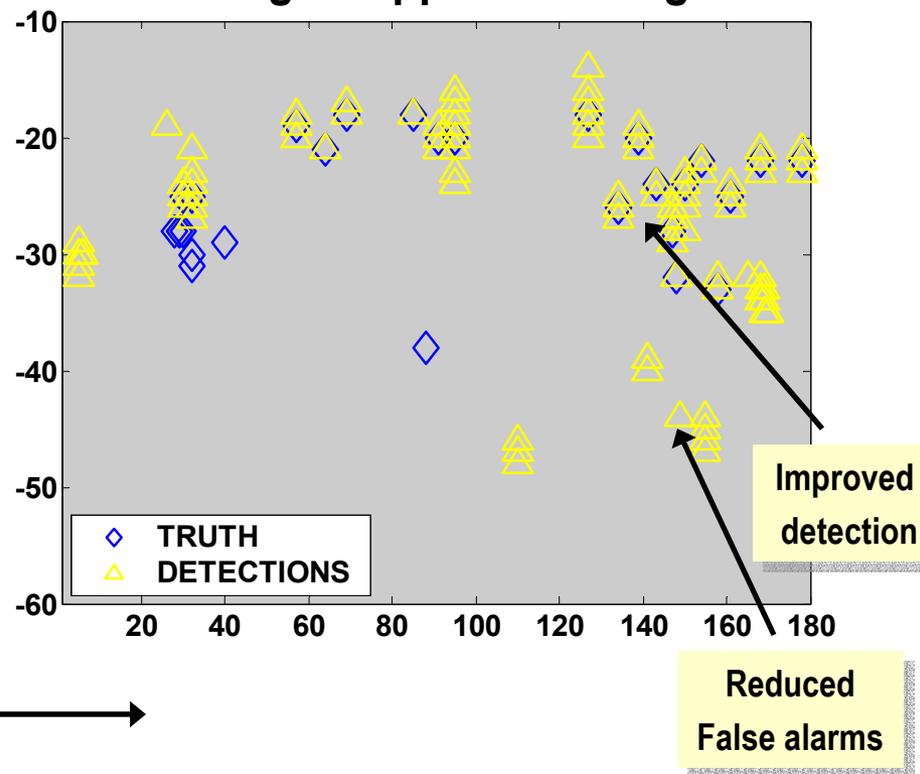
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Standard STAP



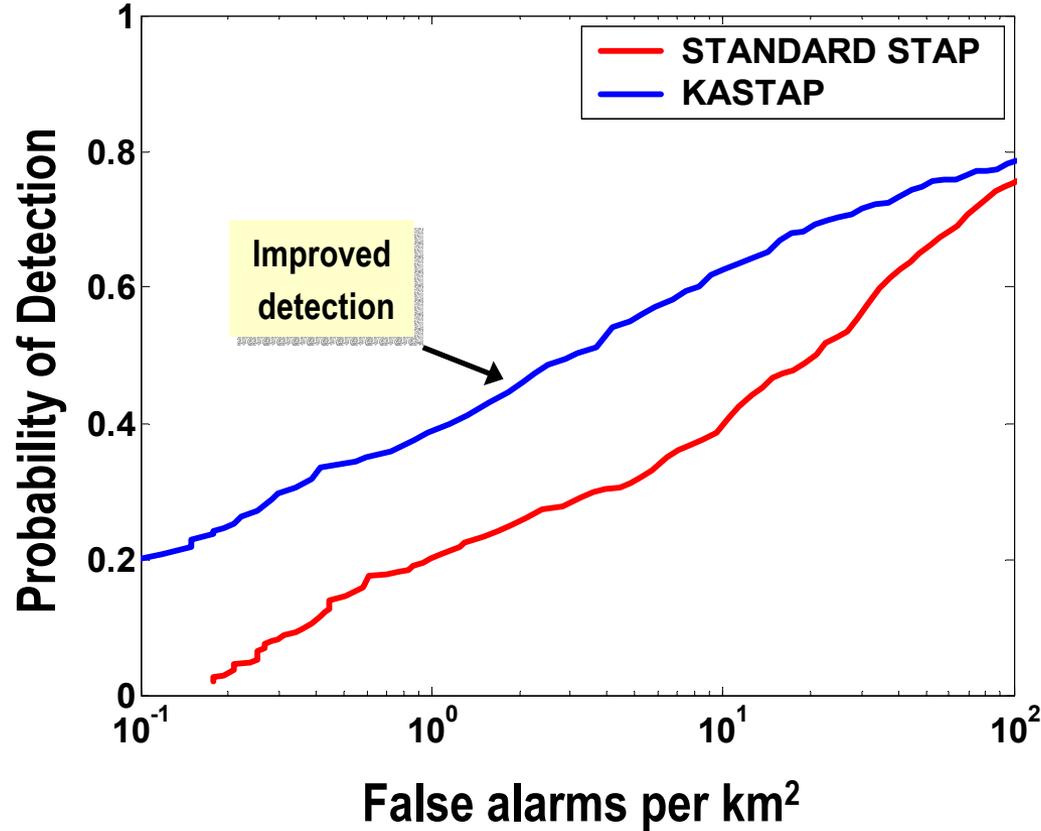
KASTAP with pre-whitening and range-Doppler masking



•A significant benefit is seen from employing KASTAP in this region



ROC Curves



Performance computed over 23 CPIs (3 km²/CPI)
AMF test statistic used

Summary of Tracker Performance STAP vs. KASTAP



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STAP:

Detection Performance:

Target 1: **10 detections 13 misses**

Target 2: 15 detections 8 misses

False alarms: **11**

Tracker Performance:

Target 1: **Track starts at CPI 11**

Target 2: Track starts at CPI 6

Tracks maintained

KASTAP:

Detection Performance:

Target 1: **15 detections 8 misses**

Target 2: 15 detections 8 misses

False alarms: **5**

Tracker Performance:

Target 1: **Track starts at CPI 6**

Target 2: Track starts at CPI 6

Tracks maintained

Movie of Tracker Performance with Standard STAP Processing

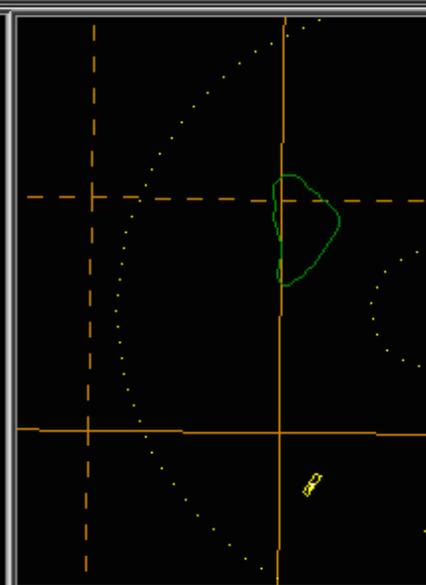
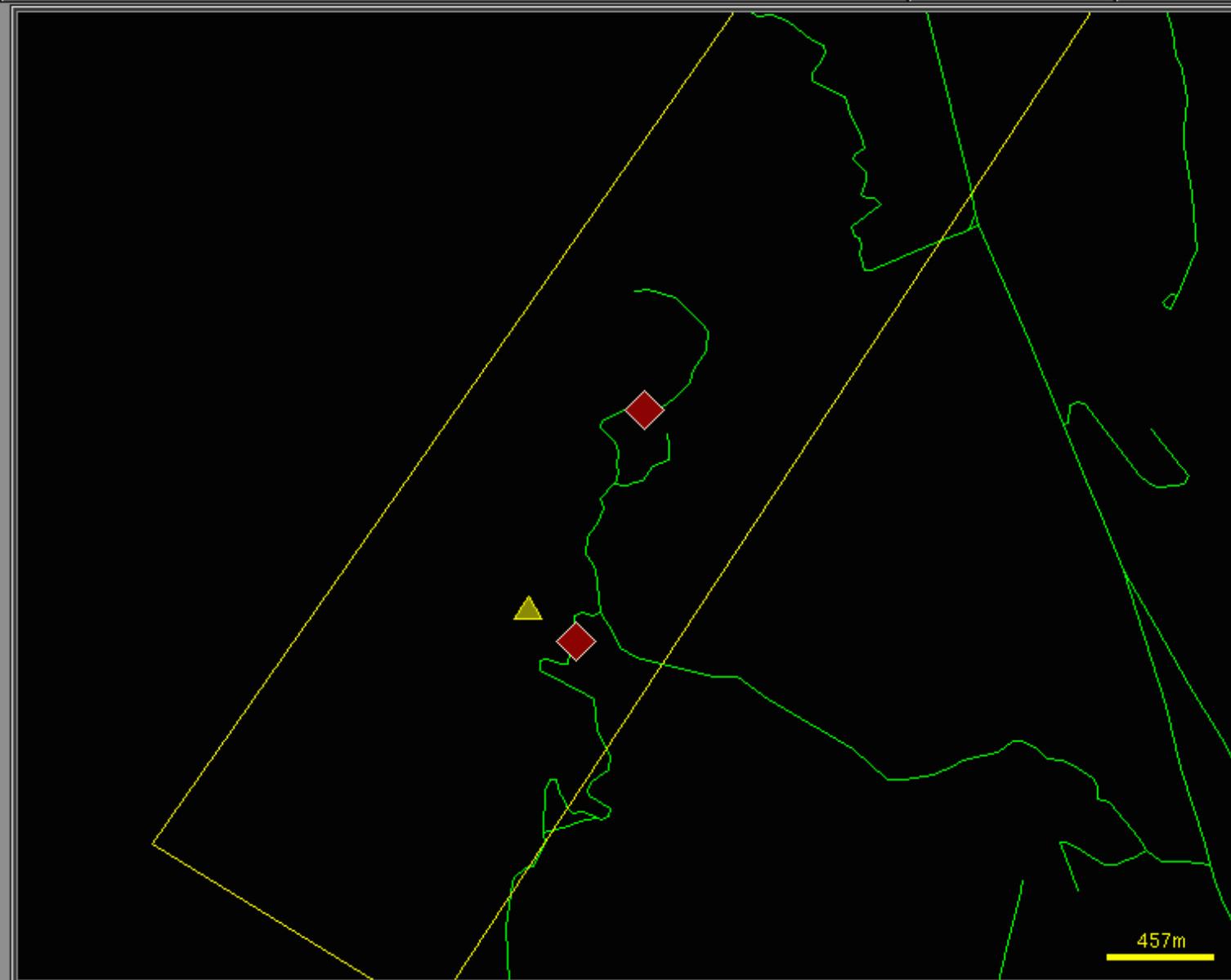


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Tracker Legend

	Detection Report
	Truth
	Track Report (Off-Road)
	Track Report (On-Road)



Track Info	
Track ID:	0
Track Count:	2
Report Index:	-1
Last Update [sec]:	001-00:00:00
Status:	Dropped
Detections:	--
Missed Detections:	--
JTIDS TQ:	15
Location:	0.00000,0.00000
Speed [knots] [m/s]:	0.0 0.0
Heading/Aspect [deg]:	0.0 5.0
Length/Width [m]:	-- --
Radial Velocity[m/s]:	-0.0
Type:	Unknown
Node/Link/Dist/Rate:	--
Group count:	--
Group len/wid [m]:	--

PFCT

001-00:00:10.000 001-00:03:50.000

GOTO Location, Frame, or Time [0] OC

Cursor Position

Geo [d]:	35.86028,-
From- Sensor	M
Az [d]:	211.0 27
GR [m]:	52859 11
SR [m]:	53347

TRUTH	REPRT	TRACK
INFO	INFO	INFO
More	More	More

Movie of Tracker Performance with Knowledge-Aided STAP Processing

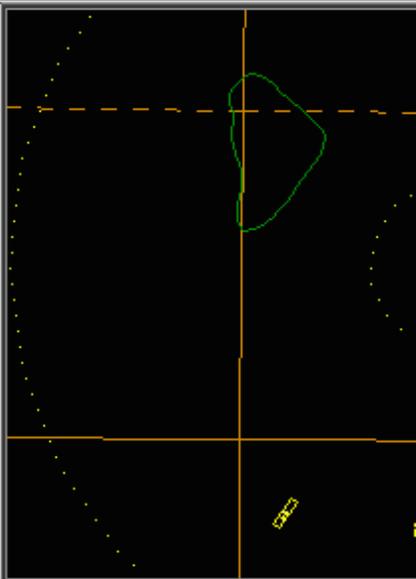
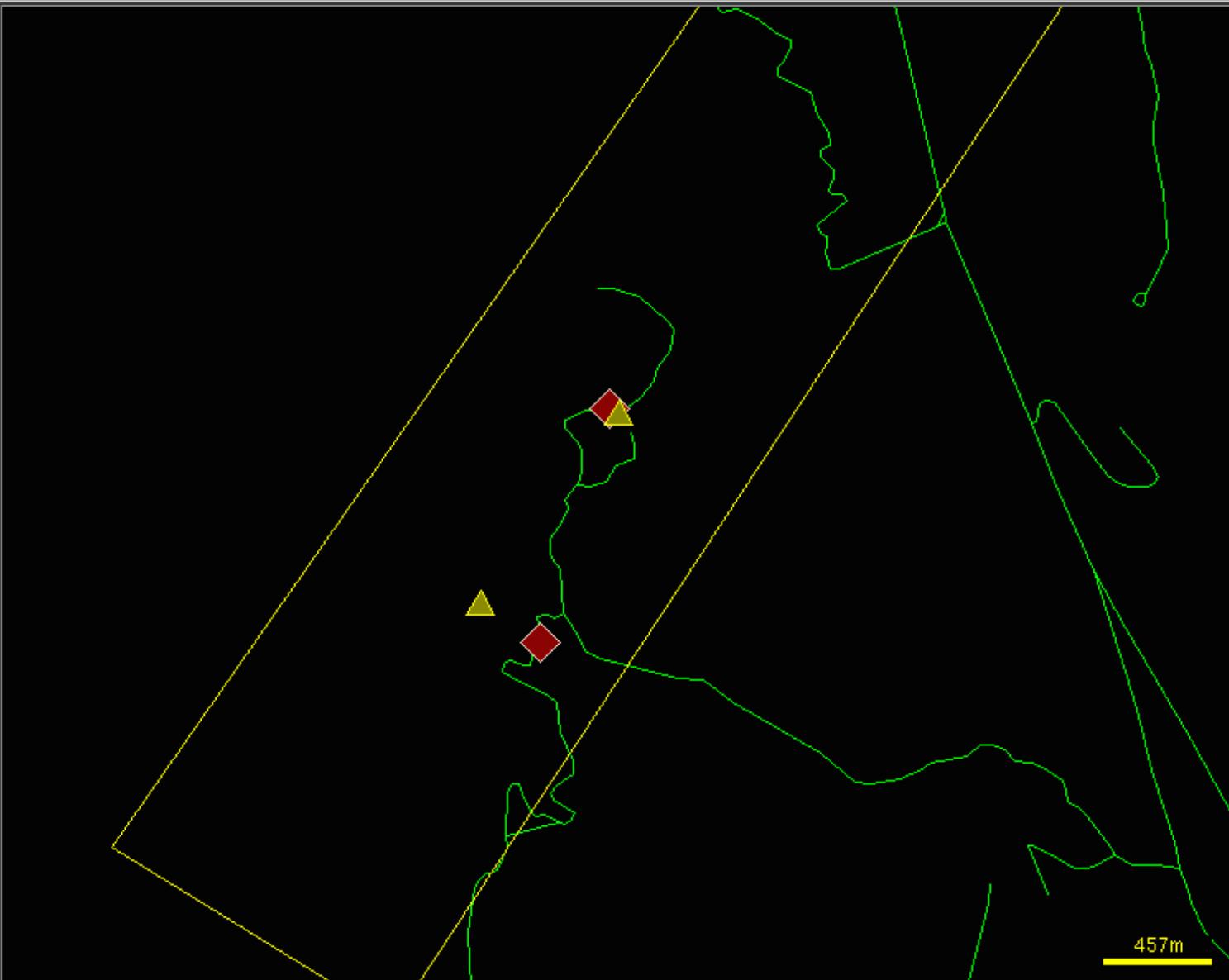


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Tracker Legend

	Detection Report
	Truth
	Track Report (Off-Road)
	Track Report (On-Road)



Track Info	
Track ID:	0
Track Count:	2
Report Index:	-1
Last Update [sec]:	001-00:00:00
Status:	Dropped
Detections:	--
Missed Detections:	--
JTIDS TQ:	15
Location:	0.00000,0.00000
Speed [knots] [m/s]:	0.0 0.0
Heading/Aspect [deg]:	0.0 9.0
Length/Width [m]:	-- --
Radial Velocity [m/s]:	-0.0
Type:	Unknown
Node/Link/Dist/Rate:	-- --
Group count:	--
Group len/wid [m]:	--



GOTO Location, Frame, or Time [0] 001-00:00:10.000



Cursor Position	
Geo [d]:	35.86241, -
From-Sensor [M]:	27
Az [d]:	211.3 27
GR [m]:	52772 11
SR [m]:	53269

TRUTH	REPRT	TRACK
INFO	INFO	INFO
More	More	More



Summary

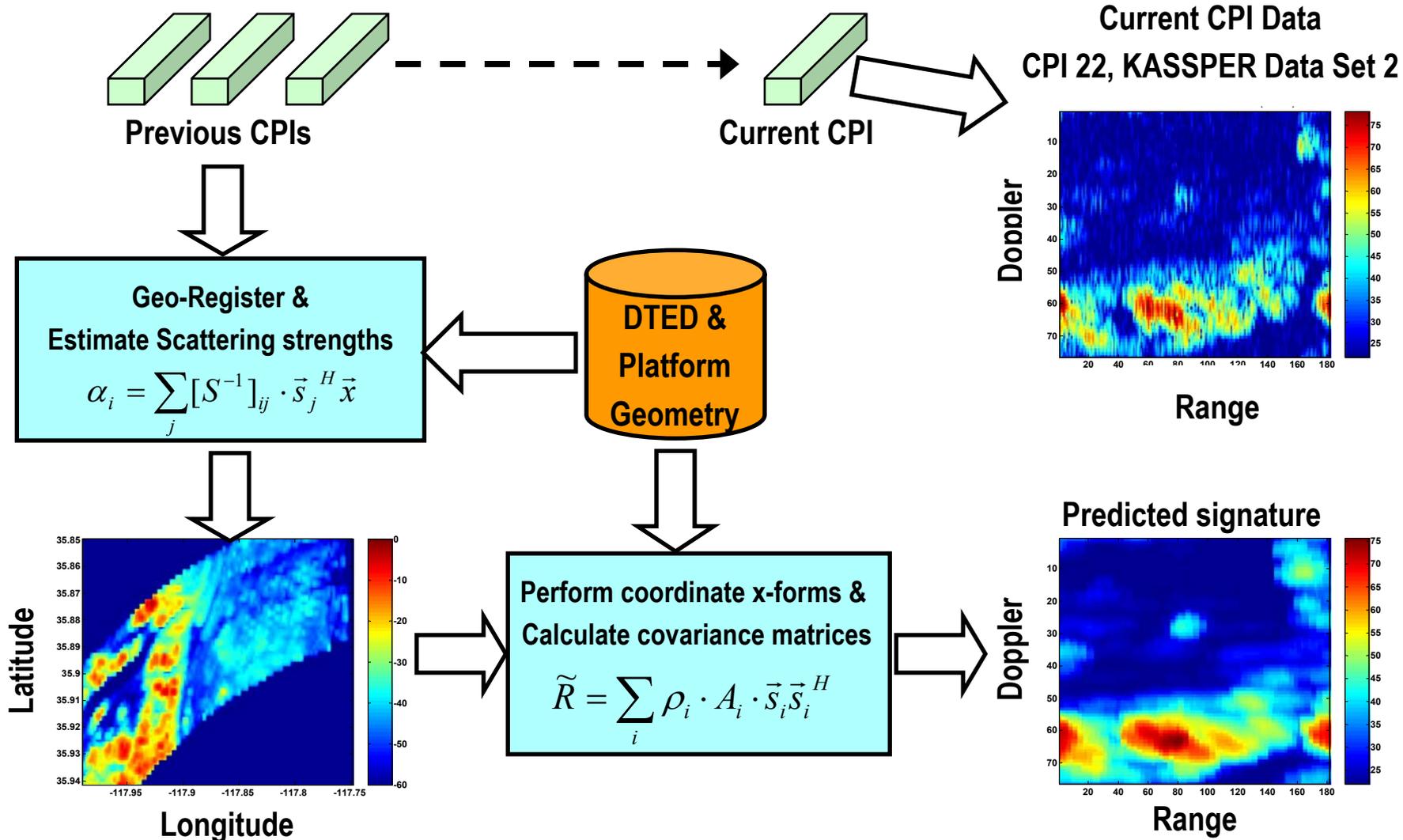


- **A technique to incorporate past-CPI data into knowledge-aided STAP processing has been described**
 - Algorithm employs a clutter map, based on multiple CPIs, in a knowledge-aided pre-whitening procedure
- **This technique has been applied to KASSPER Data Set 2 with encouraging results**
 - Improved performance in a post-Doppler framework has been demonstrated
- **Preliminary tracking performance has been obtained using ALPHATECH's MHT tracker testbed**
- **More extensive evaluations will be conducted**

Utilizing Past-CPI Data to Predict Clutter Statistics



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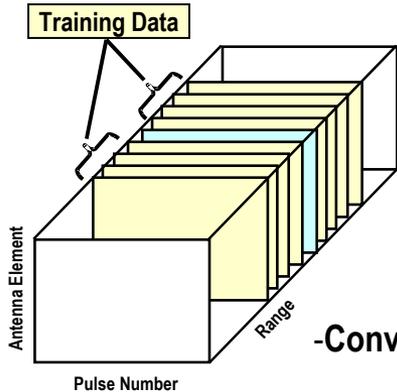


Combining Clutter Predictions to Improve STAP Performance



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Current CPI
Data-Cube



Incorporate
knowledge into STAP
weight vector calculation

Calculate STAP weights

-Conventional covariance estimates

$$\hat{R} = \frac{1}{K} \sum_i \bar{x}_i \bar{x}_i^H + \gamma \cdot I$$

-Knowledge-derived covariance matrices

$$\tilde{R} = \sum_i \rho_i \cdot A_i \cdot \vec{s}_i \vec{s}_i^H \circ T + \gamma \cdot I$$

-Tracker predictions of target range/Doppler
locations

SINR Loss Performance
(KASSPER Data Set 2, CPI 22, range 1035)

