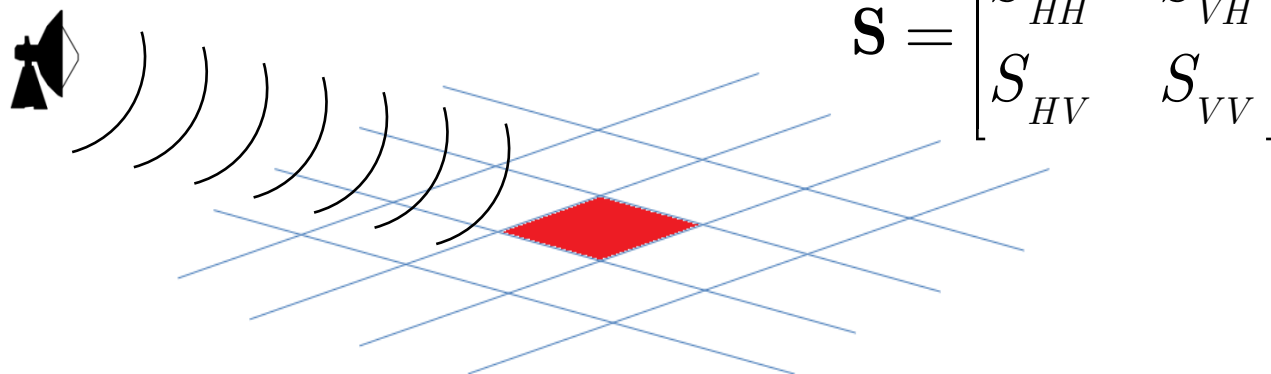


A New Model-Based Scattering Power Decomposition for Polarimetric SAR and Its Application in Analyzing Post-Tsunami Effects

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Background (1 / 5)

► POLSAR data



Coherency/Covariance Matrix

$$\mathbf{k}_p = \frac{1}{\sqrt{2}} \begin{bmatrix} S_{HH} + S_{VV} \\ S_{HH} - S_{VV} \\ 2S_{HV} \end{bmatrix}, \mathbf{k}_l = \begin{bmatrix} S_{HH} \\ \sqrt{2}S_{HV} \\ S_{VV} \end{bmatrix}$$

$$\mathbf{T} = \langle \mathbf{k}_p \mathbf{k}_p^H \rangle$$

$$\mathbf{C} = \langle \mathbf{k}_l \mathbf{k}_l^H \rangle$$

Second-Order Statistics

Background (2/5)

► Facts of coherency matrix

$$\mathbf{T} = \begin{bmatrix} T_{11} & T_{12} & T_{13} \\ T_{12}^* & T_{22} & T_{23} \\ T_{13}^* & T_{23}^* & T_{33} \end{bmatrix}$$

Diagram illustrating the structure of the coherency matrix \mathbf{T} . The matrix is Hermitian, meaning $T_{ij} = T_{ji}^*$. The diagonal elements T_{11} , T_{22} , and T_{33} are circled in red. The off-diagonal elements T_{12} , T_{13} , and T_{23} are circled in blue. Dashed arrows point from the blue circles to the text $\text{Re}(T_{12}), \text{Im}(T_{12})$, indicating that the real and imaginary parts of these elements are independent parameters.

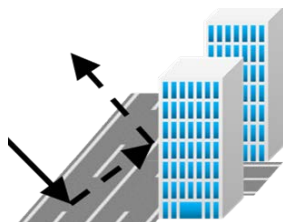
- Hermitian & Positive-semidefinite
- 9 independent (real) parameters

Background (3/5)

► Model-Based Decomposition of POLSAR Data

► Freeman-Durden Decomposition^[1]

$$\mathbf{T} = P_S \mathbf{T}_S + P_D \mathbf{T}_D + P_V \mathbf{T}_V$$



$$\mathbf{T}_S = \begin{bmatrix} 1 & \beta^* & 0 \\ \beta & |\beta|^2 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \mathbf{T}_D = \begin{bmatrix} |\alpha|^2 & \alpha^* & 0 \\ \alpha & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \mathbf{T}_V = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

► *Incomplete Utilization of Polarimetric Information: 5 out of 9*

► *Emergence of Negative Powers: $P_S, P_D < 0$ for certain pixels*

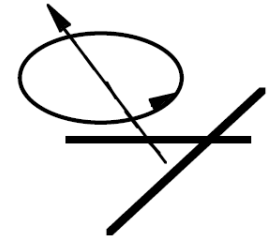
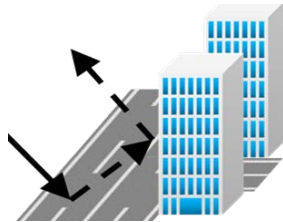
► [1] A. Freeman and S. L. Durden, "A three-component scattering model for polarimetric SAR data," *IEEE Trans. Geosci. Remote Sens.*, vol.36, pp.963-973, 1998.

Background (4/5)

► Model-Based Decomposition of POLSAR Data

► Yamaguchi Decomposition^[1]

$$\mathbf{T} = P_S \mathbf{T}_S + P_D \mathbf{T}_D + P_V \mathbf{T}_V + P_C \mathbf{T}_C$$



$$\mathbf{T}_S = \begin{bmatrix} 1 & \beta^* & 0 \\ \beta & |\beta|^2 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \mathbf{T}_D = \begin{bmatrix} |\alpha|^2 & \alpha^* & 0 \\ \alpha & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \mathbf{T}_V = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \mathbf{T}_C = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & \pm j \\ 0 & \mp j & 1 \end{bmatrix}$$

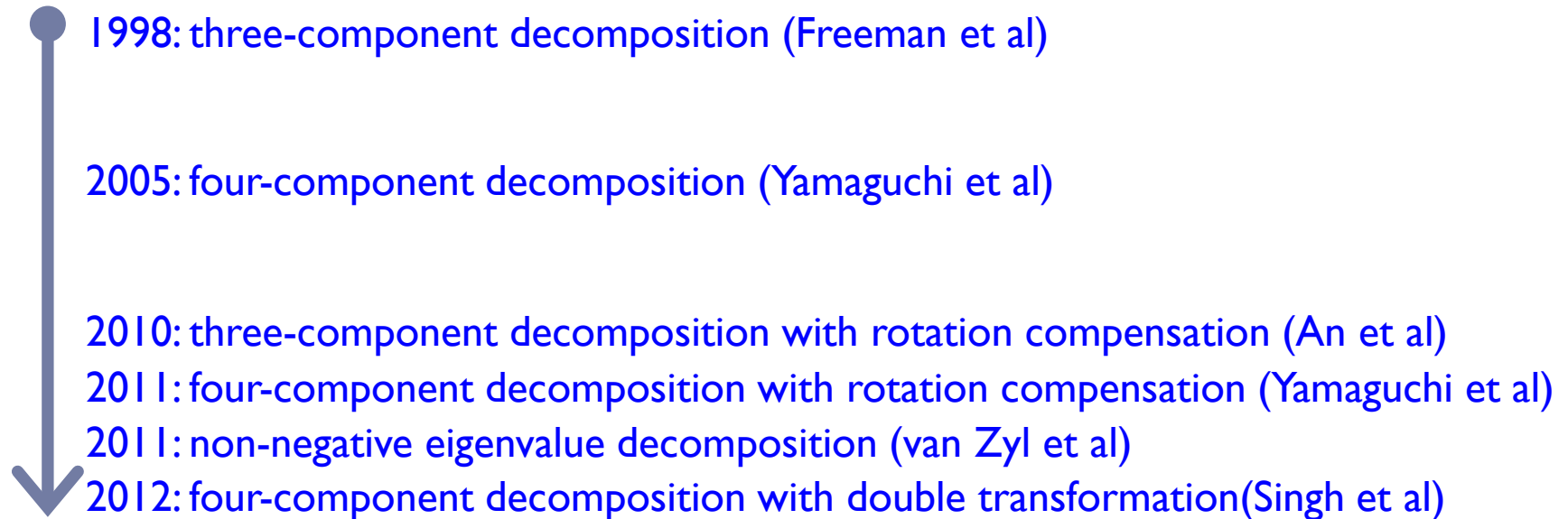
► *Incomplete Utilization of Polarimetric Information: 6 out of 9*

► *Emergence of Negative Powers: $P_S, P_D < 0$ for certain pixels*

► [1] Y. Yamaguchi, T. Moriyama, M. Ishido, and H. Yamada, "Four-component scattering model for polarimetric SAR image decomposition," *IEEE Trans. Geosci. Remote Sens.*, vol. 43, no. 8, pp. 1699-1706, Aug. 2005.

Background (5/5)

► Timeline of model-based decomposition



► Pending issues

- Prevention of negative powers
- Complete utilization of full-polarimetric information




Proposed decomposition (1 / 10)

► Problem formulation

$$\mathbf{T} = P_S \mathbf{T}_S + P_D \mathbf{T}_D + P_V \mathbf{T}_V$$



Single Scatterers



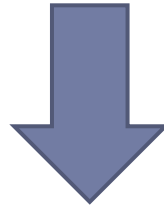
Random Volume Scattering

- **Condition #1:** \mathbf{T}_V is a known positive-definite Hermitian matrix
- **Condition #2:** \mathbf{T}_S and \mathbf{T}_D are unknown **single-rank** matrices
- **Condition #3:** $P_S, P_D, P_V \geq 0$

Proposed decomposition (2 / 10)

- ▶ Solving for volume scattering power

$$\mathbf{T} = P_S \mathbf{T}_S + P_D \mathbf{T}_D + P_V \mathbf{T}_V \Rightarrow \mathbf{T} - P_V \mathbf{T}_V = P_S \mathbf{T}_S + P_D \mathbf{T}_D$$



$$|\mathbf{T} - P_V \mathbf{T}_V| = 0$$

- ▶ All roots of $|\mathbf{T} - P_V \mathbf{T}_V| = 0$ are **non-negative**.
- ▶ **Minimum root** is volume scattering power.

Proposed decomposition (3 / 10)

- ▶ Decomposing the Remaining Matrix $\mathbf{T}' = \mathbf{T} - P_V \mathbf{T}_V$

$$\mathbf{T}' = P_S \mathbf{k}_S \mathbf{k}_S^H + P_D \mathbf{k}_D \mathbf{k}_D^H$$

$$\mathbf{k}_S = \begin{bmatrix} \cos \delta_S \\ \sin \delta_S \cos \omega_S e^{j\phi_S} \\ \sin \delta_S \sin \omega_S e^{j\phi_S} \end{bmatrix} \quad \mathbf{k}_D = \begin{bmatrix} \cos \delta_D \\ \sin \delta_D \cos \omega_D e^{j\phi_D} \\ \sin \delta_D \sin \omega_D e^{j\phi_D} \end{bmatrix}$$

- ▶ Unknown parameters: **10**
 - ▶ Available equations from \mathbf{T}' : **8**
- Underdetermined problem**

Proposed decomposition (4 / 10)

- Decomposing the Remaining Matrix $\mathbf{T}' = \mathbf{T} - P_V \mathbf{T}_V$

- **Method 1: Eigendecomposition**^[1]

$$\mathbf{T}' = \lambda_1 \mathbf{k}_1 \mathbf{k}_1^H + \lambda_2 \mathbf{k}_2 \mathbf{k}_2^H, \lambda_1 \geq \lambda_2 > 0$$



Dominant Single Scatterer:
Optimal Single-Rank Approximation of \mathbf{T}'

$$\mathbf{k}_1 = \begin{bmatrix} k_1 \\ k_2 \\ k_3 \end{bmatrix} \begin{cases} |k_1| > |k_2| \rightarrow \text{Odd-scattering} \\ |k_1| < |k_2| \rightarrow \text{Even-scattering} \end{cases}$$

[1] J.J. van Zyl, M. Arii, and Y. Kim, "Model-based decomposition of polarimetric SAR covariance matrices constrained for nonnegative eigenvalues," *IEEE Trans. Geosci. Remote Sens.*, vol. 49, no. 9, pp. 1104-1113, Sept. 2011.

Proposed decomposition (5/10)

INPUT: \mathbf{T}, \mathbf{T}_V

1: Solve the cubic equation: $\det(\mathbf{T} - x\mathbf{T}_V) = 0 \Rightarrow x_1, x_2, x_3$

2: Determine the volume scattering power: $P_V = \min\{x_1, x_2, x_3\}, \mathbf{T}' = \mathbf{T} - P_V\mathbf{T}_V$

3: Solve the eigendecomposition problem: $\mathbf{T}' = \lambda_1 \mathbf{k}_1 \mathbf{k}_1^H + \lambda_2 \mathbf{k}_2 \mathbf{k}_2^H$

4: SWITCH

5: CASE: $|\mathbf{k}_1(1)| > |\mathbf{k}_1(2)|$ AND $|\mathbf{k}_2(1)| < |\mathbf{k}_2(2)|$

6: $P_S = \lambda_1, P_D = \lambda_2$

7: CASE: $|\mathbf{k}_1(1)| < |\mathbf{k}_1(2)|$ AND $|\mathbf{k}_2(1)| > |\mathbf{k}_2(2)|$

8: $P_S = \lambda_2, P_D = \lambda_1$

9: CASE: $|\mathbf{k}_1(1)| > |\mathbf{k}_1(2)|$ AND $|\mathbf{k}_2(1)| > |\mathbf{k}_2(2)|$

10: $P_S = \lambda_1 + \lambda_2, P_D = 0$

11: CASE: $|\mathbf{k}_1(1)| < |\mathbf{k}_1(2)|$ AND $|\mathbf{k}_2(1)| < |\mathbf{k}_2(2)|$

12: $P_S = 0, P_D = \lambda_1 + \lambda_2$

13: END

OUTPUT: P_S, P_D, P_V

Proposed decomposition (6/10)

- ▶ Decomposing the Remaining Matrix $\mathbf{T}' = \mathbf{T} - P_V \mathbf{T}_V$
 - ▶ **Method 2: Optimal Model-Fitting**

$$\mathbf{T}' = P_S \mathbf{T}(\beta, \theta) + P_D \mathbf{k}_D \mathbf{k}_D^H$$



Rotated Bragg Scatterer

$$\mathbf{T}(\beta, \theta) = \frac{1}{1 + |\beta|^2} \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos 2\theta & \sin 2\theta \\ 0 & -\sin 2\theta & \cos 2\theta \end{bmatrix} \begin{bmatrix} 1 & \beta^* & 0 \\ \beta & |\beta|^2 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos 2\theta & -\sin 2\theta \\ 0 & \sin 2\theta & \cos 2\theta \end{bmatrix}$$

Proposed decomposition (7 / 10)

► Solution of Optimal Model-Fitting

► Step 1: rotation compensation

$$\mathbf{T}'(\theta) = \mathbf{R}^T(\theta) \mathbf{T}' \mathbf{R}(\theta) = \mathbf{R}^T(\theta) \left[\underline{P_S \mathbf{T}(\beta, \theta)} + \underline{P_D \mathbf{k}_D \mathbf{k}_D^H} \right] \mathbf{R}(\theta)$$

$$\mathbf{T}'(\theta) = \frac{P_S}{1 + |\beta|^2} \begin{bmatrix} 1 & \beta^* & 0 \\ \beta & |\beta|^2 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\mathbf{R}^T(\theta) \mathbf{k}_D = \begin{bmatrix} \cos \delta \\ \sin \delta \cos \omega e^{j\phi} \\ \sin \delta \sin \omega e^{j\phi} \end{bmatrix}$$

$$+ P_D \begin{bmatrix} \cos^2 \delta & \cos \delta \sin \delta \cos \omega e^{-j\phi} & \cos \delta \sin \delta \sin \omega e^{-j\phi} \\ \cos \delta \sin \delta \cos \omega e^{j\phi} & \sin^2 \delta \cos^2 \omega & \sin^2 \delta \cos \omega \sin \omega e^{j(\phi-\phi)} \\ \cos \delta \sin \delta \sin \omega e^{j\phi} & \sin^2 \delta \cos \omega \sin \omega e^{j(\phi-\phi)} & \sin^2 \delta \sin^2 \omega \end{bmatrix}$$

Proposed decomposition (8/10)

► Solution of Optimal Model-Fitting

► Step 2: solving unknowns in terms of $\mathbf{T}'(\theta)$

$$\mathbf{T}'(\theta) = \frac{P_S}{1 + |\beta|^2} \begin{bmatrix} 1 & \beta^* & 0 \\ \beta & |\beta|^2 & 0 \\ 0 & 0 & 0 \end{bmatrix} + P_D \begin{bmatrix} \cos^2 \delta & \cos \delta \sin \delta \cos \omega e^{-j\phi} & \cos \delta \sin \delta \sin \omega e^{-j\phi} \\ \cos \delta \sin \delta \cos \omega e^{j\phi} & \sin^2 \delta \cos^2 \omega & \sin^2 \delta \cos \omega \sin \omega e^{j(\phi-\varphi)} \\ \cos \delta \sin \delta \sin \omega e^{j\phi} & \sin^2 \delta \cos \omega \sin \omega e^{j(\varphi-\phi)} & \sin^2 \delta \sin^2 \omega \end{bmatrix}$$



$$\begin{aligned} P_S &= T'_{11}(\theta) + T'_{22}(\theta) - \frac{|T'_{13}(\theta)|^2 + |T'_{23}(\theta)|^2}{T'_{33}(\theta)} \\ \beta &= \frac{T'_{33}(\theta)[T'_{12}(\theta)]^* - T'_{23}(\theta)[T'_{13}(\theta)]^*}{T'_{11}(\theta)T'_{33}(\theta) - |T'_{13}(\theta)|^2} \\ P_D &= T'_{33}(\theta) + \frac{|T'_{13}(\theta)|^2 + |T'_{23}(\theta)|^2}{T'_{33}(\theta)} \\ \delta &= \sin^{-1} \left[\sqrt{\frac{|T'_{33}(\theta)|^2 + |T'_{23}(\theta)|^2}{|T'_{33}(\theta)|^2 + |T'_{23}(\theta)|^2 + |T'_{13}(\theta)|^2}} \right] \\ \omega &= \sin^{-1} \left[\sqrt{\frac{|T'_{33}(\theta)|^2}{|T'_{33}(\theta)|^2 + |T'_{23}(\theta)|^2}} \right] \\ \phi &= \arg\{T'_{23}(\theta)[T'_{13}(\theta)]^*\} \\ \varphi &= -\arg[T'_{13}(\theta)] \end{aligned}$$

For any θ , $P_S, P_D \geq 0$



Proposed decomposition (9 / 10)

► Solution of Optimal Model-Fitting

► Step 3: maximizing P_s with respect to θ

$$\max_{\theta} \left\{ T'_{11}(\theta) + T'_{22}(\theta) - \frac{|T'_{13}(\theta)|^2 + |T'_{23}(\theta)|^2}{T'_{33}(\theta)} \right\} \xrightarrow{\text{green arrow}} \theta_{\max} = \frac{1}{4} \left[\tan^{-1} \frac{(B + \sqrt{B^2 - C})d - a}{(B + \sqrt{B^2 - C})e - b} + k\pi \right]$$

$$B = \frac{ad + eb - cf}{d^2 + e^2 - f^2}$$

$$C = \frac{a^2 + b^2 - c^2}{d^2 + e^2 - f^2}$$

$$a = \frac{1}{2} [T'_{11}T'_{33} - |T'_{13}|^2 - T'_{11}T'_{22} + |T'_{12}|^2]$$

$$b = T'_{11} \operatorname{Re}(T'_{23}) - \operatorname{Re}[T'_{12}(T'_{13})^*]$$

$$c = T'_{22}T'_{33} - |T'_{23}|^2 + \frac{1}{2} (T'_{11}T'_{33} - |T'_{13}|^2 + T'_{11}T'_{22} - |T'_{12}|^2)$$

$$d = \frac{1}{2} (T'_{33} - T'_{22})$$

$$e = \operatorname{Re}(T'_{23})$$

$$f = \frac{1}{2} (T'_{22} + T'_{33})$$



Proposed decomposition (10/10)

INPUT:

\mathbf{T}, \mathbf{T}_V

- 1: Solve the cubic equation: $\det(\mathbf{T} - x\mathbf{T}_V) = 0 \Rightarrow x_1, x_2, x_3$
- 2: Determine the volume scattering power: $P_V = \min\{x_1, x_2, x_3\}, \mathbf{T}' = \mathbf{T} - P_V\mathbf{T}_V$
- 3: Determine the constants a, b, c, d, e, f, B, C
- 4: Obtain the rotation angle: $\theta = \frac{1}{4} \left[\tan^{-1} \frac{(B + \sqrt{B^2 - C})d - a}{(B + \sqrt{B^2 - C})e - b} + k\pi \right]$
- 5: Rotate the remaining matrix: $\mathbf{T}'(\theta) = \mathbf{R}^T(\theta)\mathbf{T}'\mathbf{R}(\theta)$
- 6: $|\beta| = \left| \frac{T'_{23}(\theta)[T'_{12}(\theta)]^* - T'_{13}(\theta)[T'_{23}(\theta)]^*}{T'_{11}(\theta)T'_{23}(\theta) - |T'_{13}(\theta)|^2} \right|, |\tan\delta\cos\omega| = \left| \frac{T'_{23}(\theta)}{T'_{13}(\theta)} \right|$

7:

SWITCH

8:

CASE: $|\beta| < 1$ **AND** $|\tan\delta\cos\omega| > 1$

9:

$$P_S = T'_{11}(\theta) + T'_{22}(\theta) - \frac{|T'_{13}(\theta)|^2 + |T'_{23}(\theta)|^2}{T'_{23}(\theta)}, P_D = T'_{33}(\theta) + \frac{|T'_{13}(\theta)|^2 + |T'_{23}(\theta)|^2}{T'_{23}(\theta)}$$

10:

CASE: $|\beta| > 1$ **AND** $|\tan\delta\cos\omega| < 1$

11:

$$P_S = T'_{33}(\theta) + \frac{|T'_{13}(\theta)|^2 + |T'_{23}(\theta)|^2}{T'_{23}(\theta)}, P_D = T'_{11}(\theta) + T'_{22}(\theta) - \frac{|T'_{13}(\theta)|^2 + |T'_{23}(\theta)|^2}{T'_{23}(\theta)}$$

12:

CASE: $|\beta| < 1$ **AND** $|\tan\delta\cos\omega| < 1$

13:

$$P_S = T'_{11}(\theta) + T'_{22}(\theta) + T'_{33}(\theta), P_D = 0$$

14:

CASE: $|\beta| > 1$ **AND** $|\tan\delta\cos\omega| > 1$

15:

$$P_S = 0, P_D = T'_{11}(\theta) + T'_{22}(\theta) + T'_{33}(\theta)$$

16:

END

OUTPUT:

P_S, P_D, P_V

Application to post-tsunami area (1 / 4)

▶ Experimental data

- ▶ **Sensor:**ALOS-PALSAR

- ▶ **Mode:** fully-polarimetric

- ▶ **Location:** Ishinomaki area

- ▶ **Data:**

 - ▶ ALPSRP257090760-PI.1__A (Nov. 21, 2010, before the earthquake)

 - ▶ ALPSRP277220760-PI.1__A (Apr. 8, 2011 after the earthquake,).

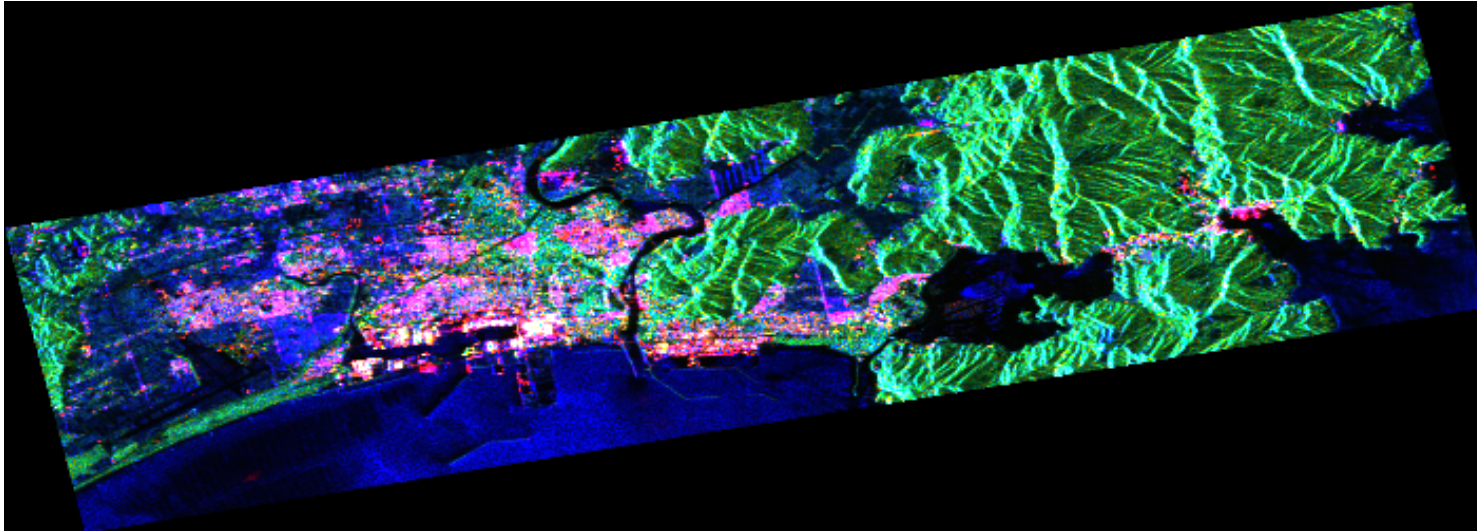
▶ Preprocessing

- ▶ Reprojection to the World Geodetic System (WGS 84).

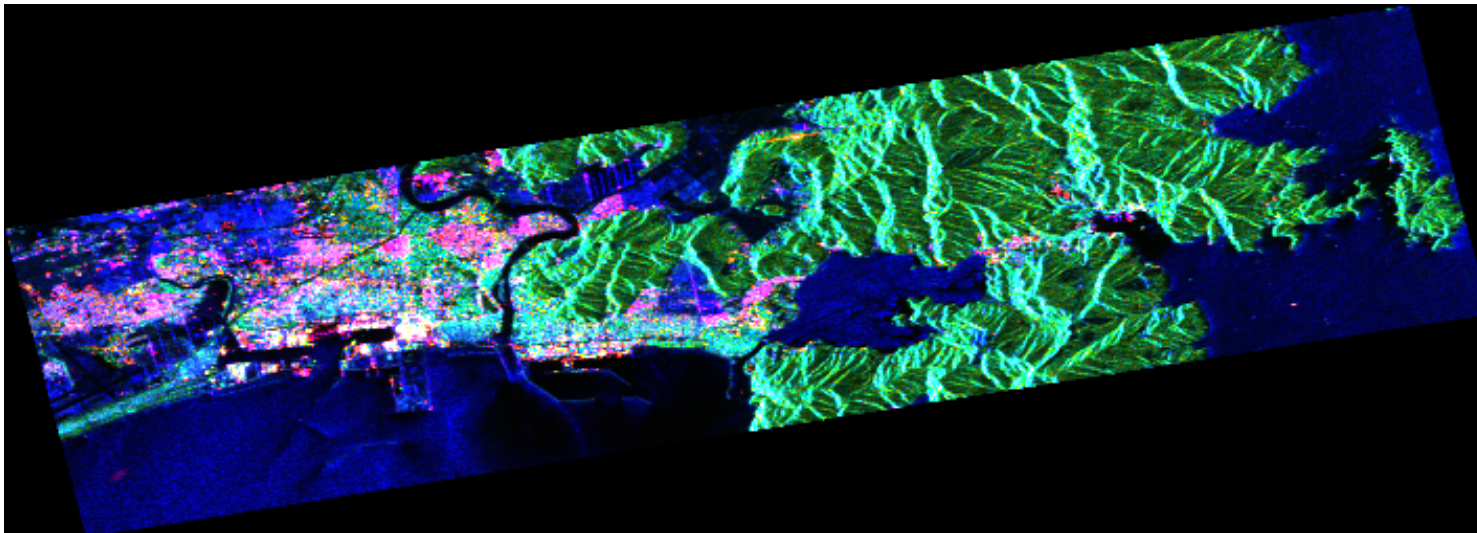
- ▶ Spatial multi-look processing with 9×9 pixels^[1].

[1] J.-S. Lee, T.L. Ainsworth, J.P. Kelly, and C. Lopez-Martinez, "Evaluation and bias removal of multilook effect on entropy/alpha/anisotropy in polarimetric SAR decomposition," *IEEE Trans. Geosci. Remote Sens.*, vol. 46, no. 10, pp. 3039-3052, Oct. 2008.

Application to post-tsunami area (2/4)



Nov. 21, 2010

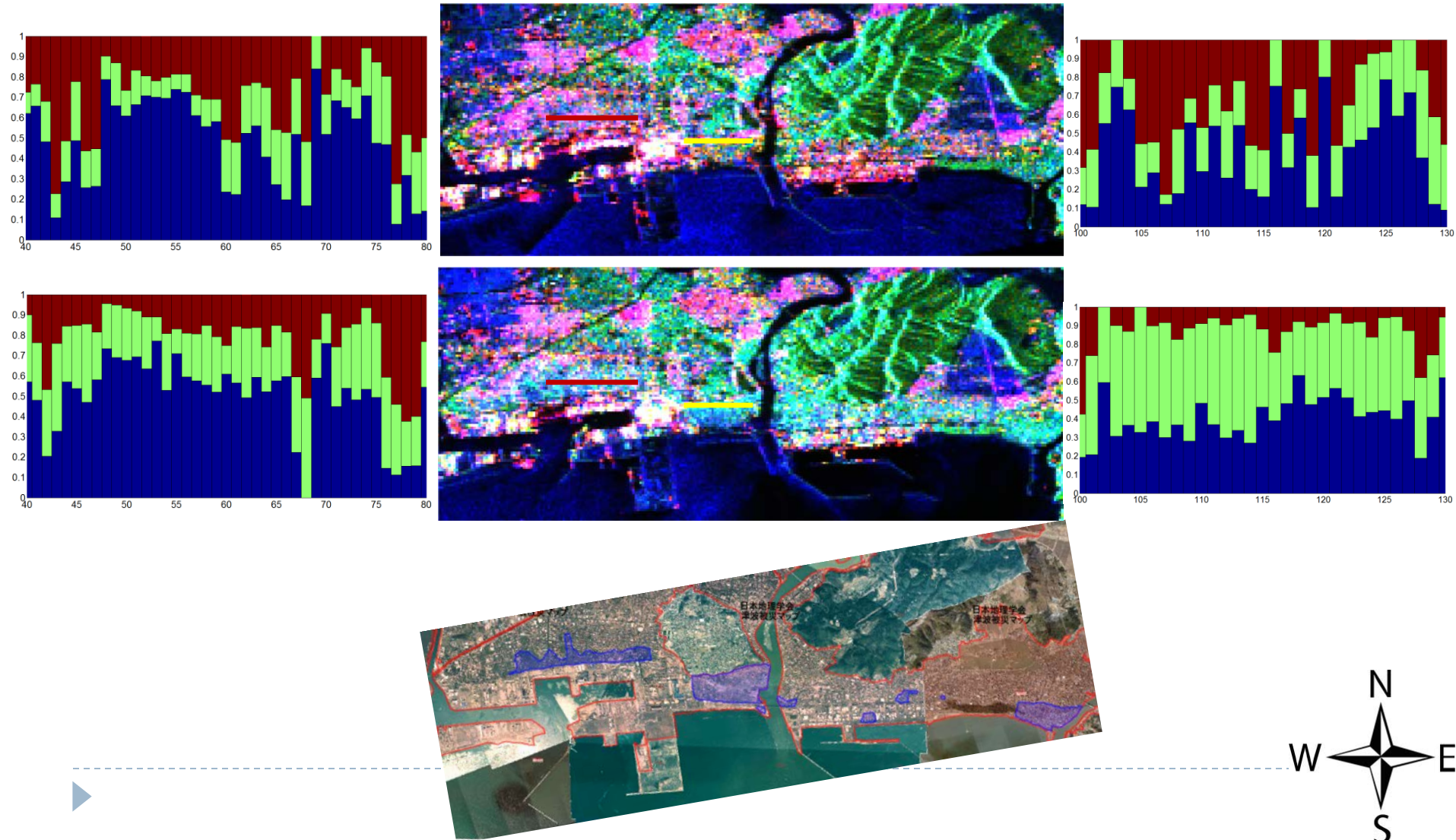


Apr. 08, 2011

double-boucne
volume
surface

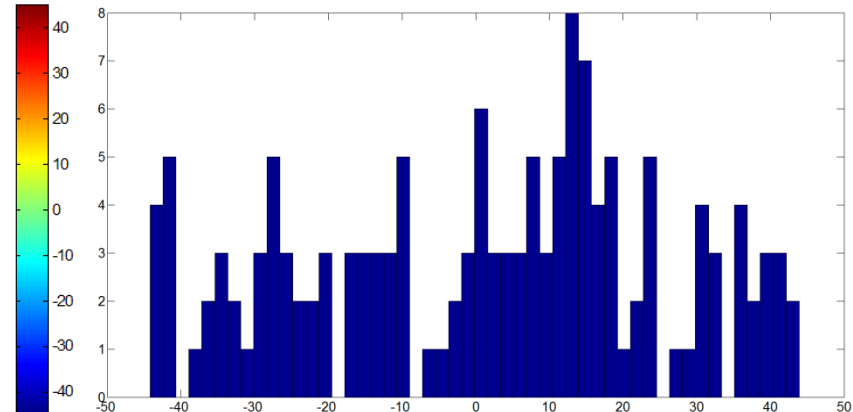
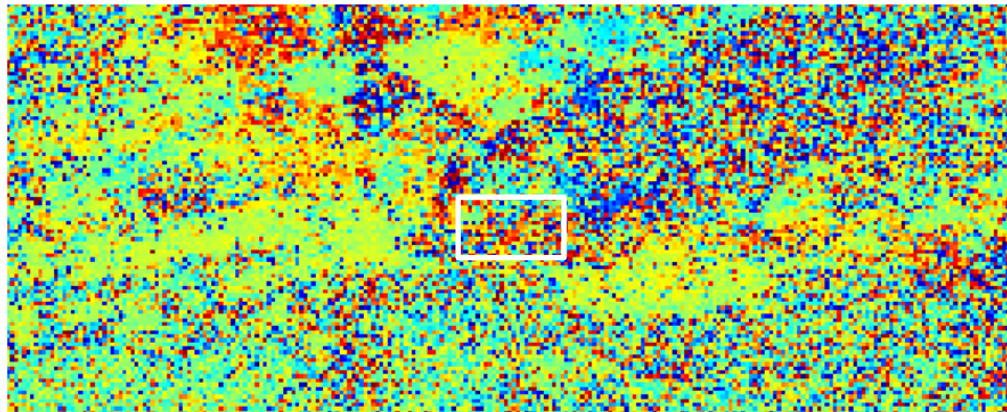
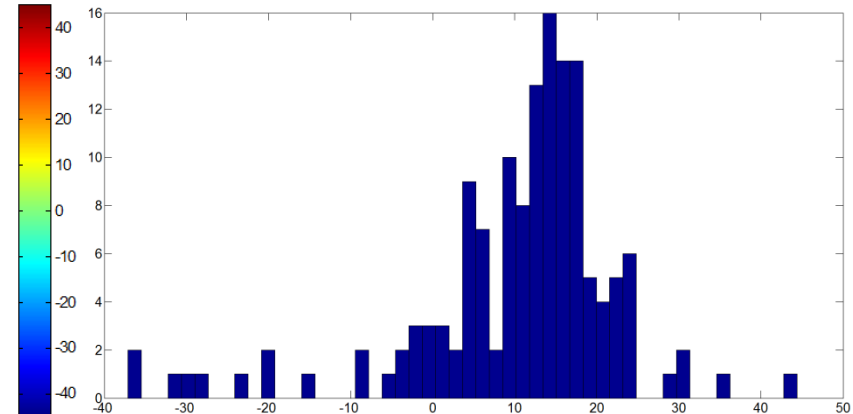
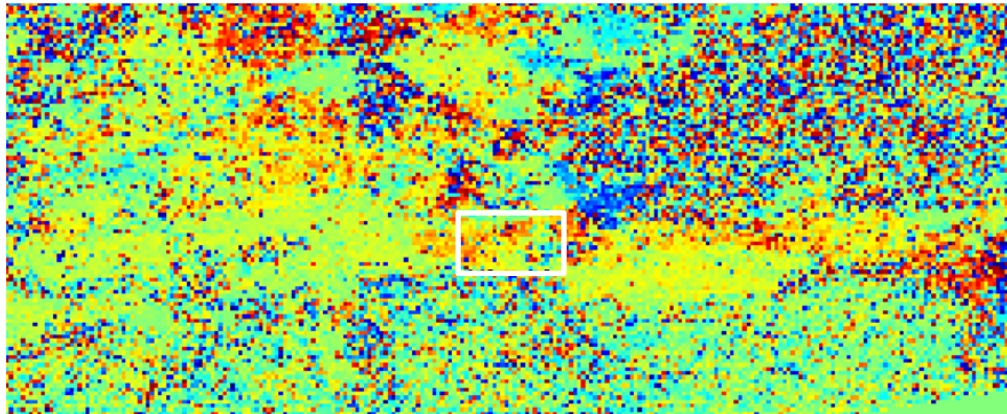
Application to post-tsunami area (3/4)

► Ishinomaki City



Application to post-tsunami area (4/4)

► Orientation angle



Conclusion

▶ New scattering power decomposition

- ▶ Measured coherency matrix is assumed to be the combined contribution of volume scattering and two single scatterers.
- ▶ Volume scattering power is uniquely equal to the minimum root of $|\mathbf{T} - P_V \mathbf{T}_V| = 0$.
- ▶ Methods to discriminate two single scatterers.
 - ▶ *Eigendecomposition method*
 - ▶ *Optimal model fitting method*

▶ Preliminary results in post-tsunami analysis

- ▶ Volume scattering power tends to increase in tsunami affected area.
- ▶ In urban areas, double-bounce scattering power tends to decrease due to the destruction of man-made structures.
- ▶ Surface scattering power tends to increase in totally washed-out area^[1].
- ▶ Orientation angle tends to randomize in tsunami affected area^[2].

[1] Y. Yamaguchi, "Disaster monitoring by fully polarimetric SAR data acquired with ALOS-PALSAR," *Proceedings of IEEE*, vol. 100, no. 10, Oct. 2012.

▶ [2] S.-W. Chen and M. Sato, "Tsunami damage investigation of built-up areas using multitemporal spaceborne fully polarimetric SAR images," *IEEE Trans. Geosci. Remote Sens.*, in press.

THANK YOU!

