



UNIVERSITAT POLITÈCNICA DE CATALUNYA
BARCELONATECH

Departament de Teoria del Senyal
i Comunicacions



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SAR Statistics and Polarimetric decompositions

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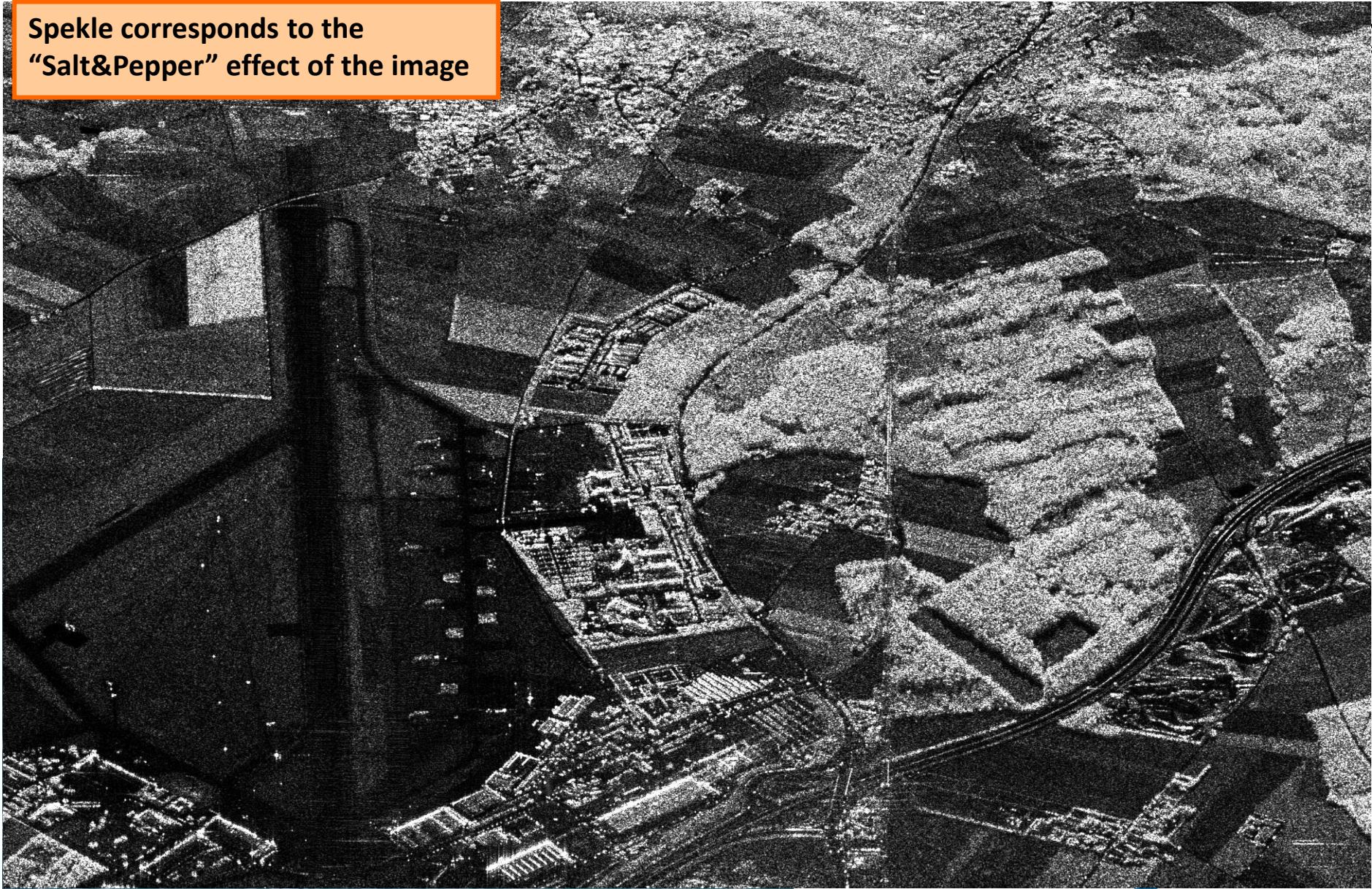


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- SAR Data Statistical Characterization
- PolSAR Data Statistical Characterization
- Information Estimation/Filtering

Speckle corresponds to the "Salt&Pepper" effect of the image

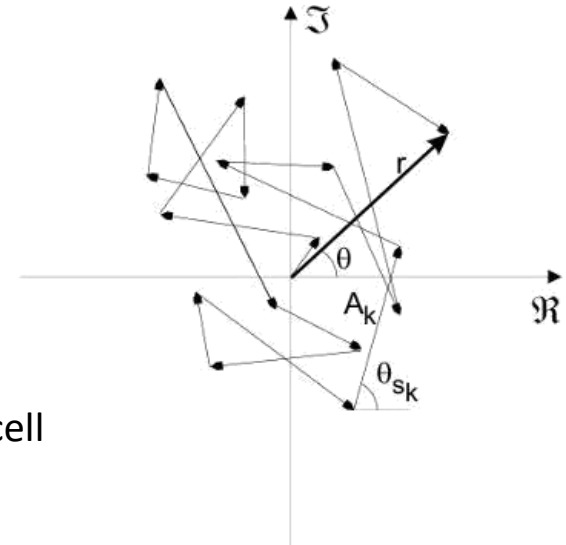


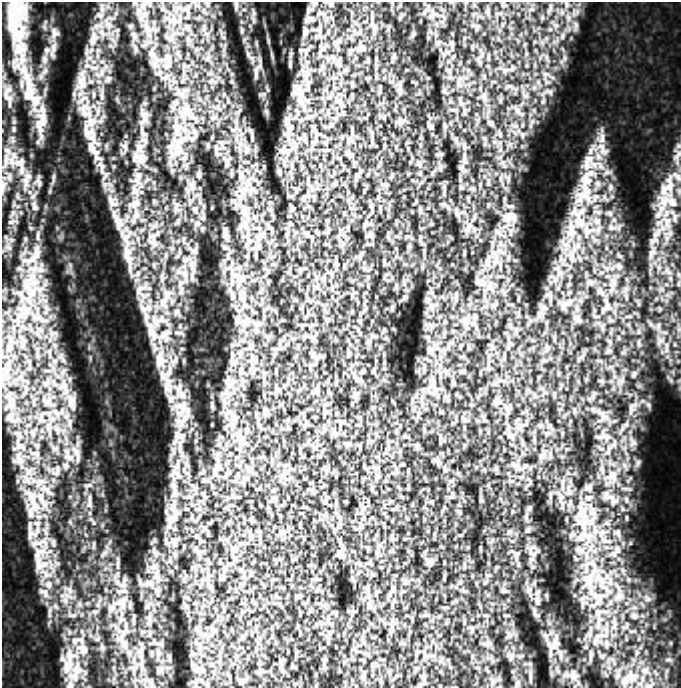
On the basis of the discrete scatter description

$$S(x, r) = \frac{1}{\sqrt{L}} \sum_{k=1}^L \sqrt{\sigma_k} e^{j\theta_k} h(x - x_k, r - r_k)$$

L: Number of point scatters embraced by the resolution cell

- L as a **deterministic** quantity
 - L = 1: or a dominating point scatter: **Deterministic scattering**
 - Rice/Rician model
 - L > 1: Partially developed speckle
 - Not solved model. Even numerical solution difficult
 - L >> 1: **Fully developed speckle**
 - Gaussian model
- L as a **stochastic** quantity
 - L characterized by a pdf: Image texture
 - K-distribution model

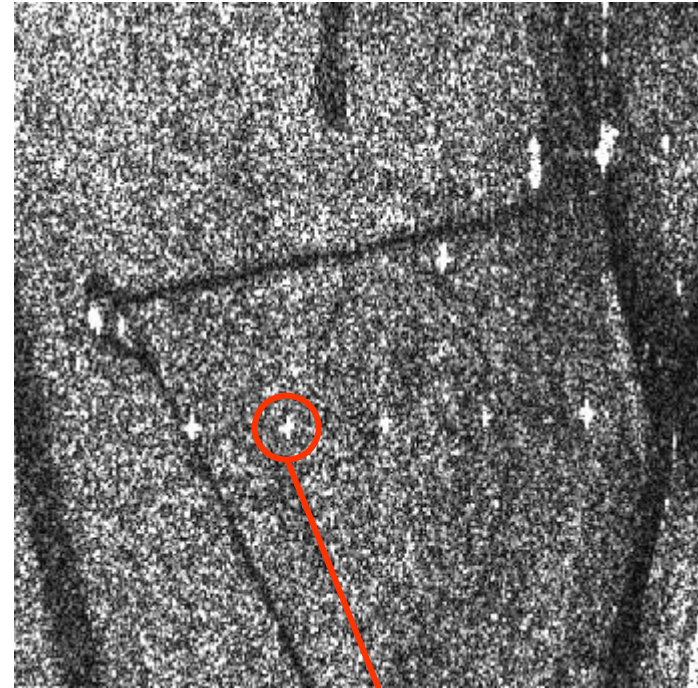




Fully Developed speckle

Bright points: Points where the interference is **constructive**

Dark points: Points where the interference is **destructive**



Corner reflector
Dominant scatter
No speckle

Speckle is an electromagnetic interference phenomenon



S_{hh} amplitude
E-SAR L-band system

- **Completely developed Speckle** (large L and no dominant scatterer)
 - When the number of scatterers is sufficiently large, and their spatial distribution is random (random phases) the central limit theorem applies

$$S = N_{C^2} \left(0, \sigma^2/2 \right)$$

- **Central Limit Theorem**

- **Real Part**

$$p_{\Re\{S\}}(\Re\{S\}) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2}\left(\frac{\Re\{S\}}{\sigma}\right)^2\right) \quad \Re\{S\} \in (-\infty, \infty) \quad \text{Gaussian pdf}$$

- **Imaginary Part**

$$p_{\Im\{S\}}(\Im\{S\}) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2}\left(\frac{\Im\{S\}}{\sigma}\right)^2\right) \quad \Im\{S\} \in (-\infty, \infty) \quad \text{Gaussian pdf}$$

- Real and imaginary parts are uncorrelated

$$E\{\Re\{S\}\Im\{S\}\} = 0$$

- **Amplitude: Rayleigh pdf**

$$p_r(r) = \frac{r}{\sigma^2} \exp\left(-\frac{1}{2}\left(\frac{r}{\sigma}\right)^2\right) \quad r \in [0, \infty)$$

$$E\{r\} = \sqrt{\frac{\pi}{2}}\sigma$$

$$E\{r^2\} = 2\sigma^2$$

$$\sigma_r^2 = E\{r^2\} - E^2\{r\} = \left(2 - \frac{\pi}{2}\right)\sigma^2$$

- **Intensity ($I=r^2$): Exponential pdf**

$$p_I(I) = \frac{1}{2\sigma^2} \exp\left(-\frac{I}{2\sigma^2}\right) \quad I \in [0, \infty)$$

$$E\{I\} = 2\sigma^2 \equiv \sigma$$

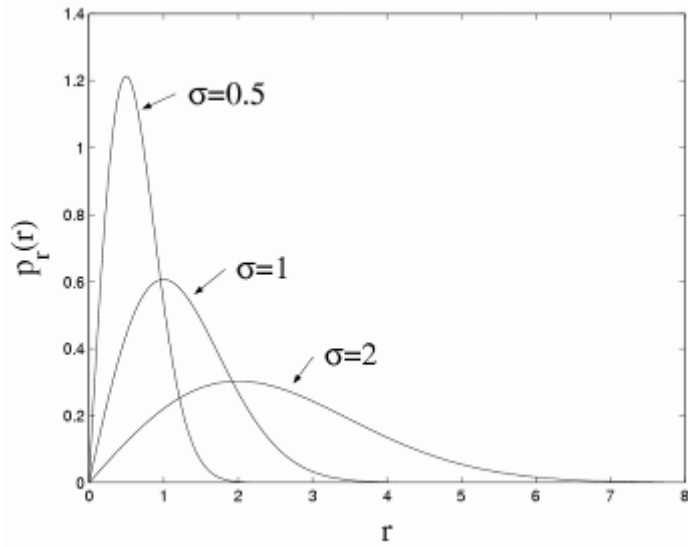
$$E\{I^2\} = 2(2\sigma^2)^2$$

$$\sigma_I^2 = E\{I^2\} - E^2\{I\} = (2\sigma^2)^2$$

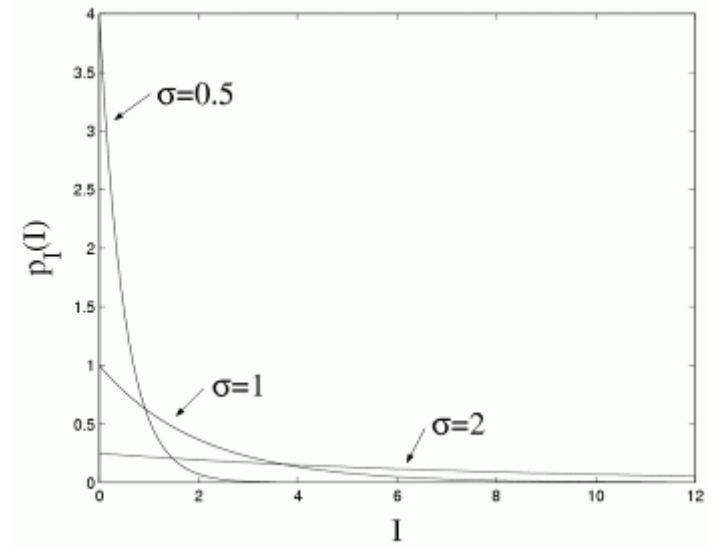
- **Phase: Uniform pdf.** Contains NO information

$$p_\theta(\theta) = \frac{1}{2\pi} \quad \theta \in [-\pi, \pi)$$

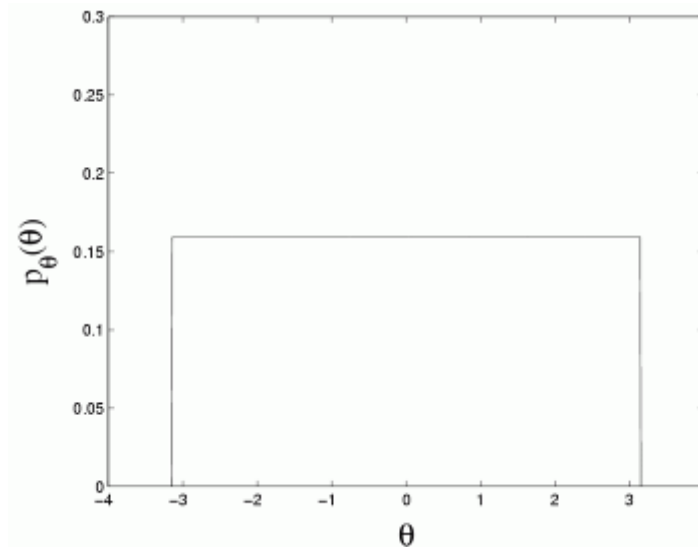
- Amplitude and phase are uncorrelated



Amplitude: Rayleigh pdf



Intensity: Exponential pdf

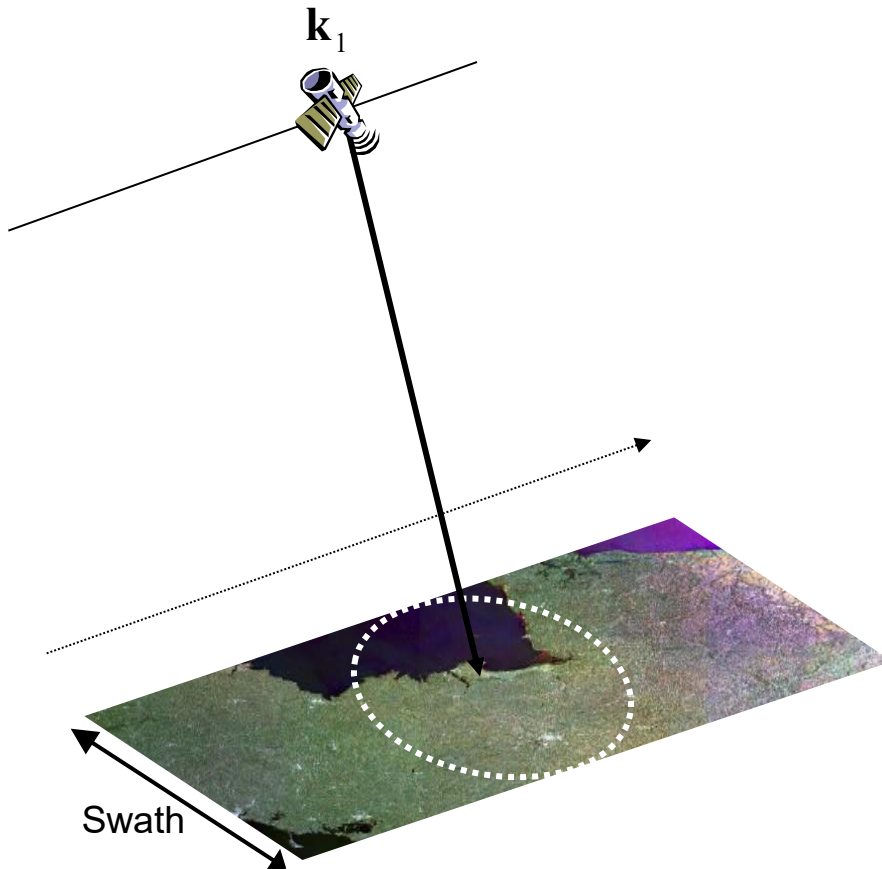


Phase: Uniform pdf

The Polarimetric SAR system acquires 3 complex SAR images

$$\text{Scattering matrix } \mathbf{S} = \begin{bmatrix} S_{hh} & S_{hv} \\ S_{hv} & S_{vv} \end{bmatrix}$$

$$\text{Scattering vector } \mathbf{k} = [S_{hh}, 2S_{hv}, S_{vv}]^T$$



The properties of the scattering vector follow from the properties of a single SAR image

- \mathbf{k} is **deterministic** for **point scatters**.
- \mathbf{k} is a **multidimensional random variable** for **distributed scatters** due to **speckle**.

SAR images characterized through second order moments

- **Second order moments** in multidimensional SAR data are **matrix quantities**

Characterization of random variables

- Probability Density Function (pdf)
- Statistical moments (mean, power, kurtosis, skewness...)

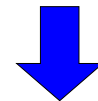
Zero-mean multidimensional complex Gaussian pdf

$$p_{\mathbf{k}}(\mathbf{k}) = \frac{1}{\pi^3 |\mathbf{C}|} \exp(-\mathbf{k}^H \mathbf{C}^{-1} \mathbf{k})$$

- **First** order moment $E\{\mathbf{k}\} = \mathbf{0}$
- **Second** order moment: Covariance matrix

$$\mathbf{C} = E\{\mathbf{k}\mathbf{k}^H\} = \begin{bmatrix} E\{|S_{hh}|^2\} & E\{S_{hh}S_{hv}^*\} & E\{S_{hh}S_{vv}^*\} \\ E\{S_{hv}S_{hh}^*\} & E\{|S_{hv}|^2\} & E\{S_{hv}S_{vv}^*\} \\ E\{S_{vv}S_{hh}^*\} & E\{S_{vv}S_{hv}^*\} & E\{|S_{vv}|^2\} \end{bmatrix}$$

$$E\{S_k S_l^*\} \neq 0 \quad k, l \in \{1, \dots, m\}, k \neq l$$



Correlated SAR images

The covariance matrix contains the **correlation structure** of the set of m SAR images

$$\mathbf{C} = E \{ \mathbf{k} \mathbf{k}^H \} = \begin{bmatrix} E \{ |S_{hh}|^2 \} & E \{ S_{hh} S_{hv}^* \} & E \{ S_{hh} S_{vv}^* \} \\ E \{ S_{hv} S_{hh}^* \} & E \{ |S_{hv}|^2 \} & E \{ S_{hv} S_{vv}^* \} \\ E \{ S_{vv} S_{hh}^* \} & E \{ S_{vv} S_{hv}^* \} & E \{ |S_{vv}|^2 \} \end{bmatrix}$$

Information

- Diagonal elements: **Power information**

$$E \{ S_k S_k^H \} = E \{ |S_k|^2 \} \quad k \in \{1, 2, \dots, m\}$$

- Off-diagonal elements: **Correlation information**

$$E \{ S_k S_l^H \} \quad k, l \in \{1, 2, \dots, m\}, k \neq l \quad \rho_{k,l} = \frac{E \{ S_k S_l^* \}}{\sqrt{E \{ |S_k|^2 \} \cdot E \{ |S_l|^2 \}}} = |\rho_{k,l}| e^{j\theta_{k,l}}$$

First order moment

**Multidimensional SAR data
descriptors
for distributed scatterers**

Second order moment

$$\mathbf{S} = \begin{bmatrix} S_{hh} & S_{hv} \\ S_{hv}^H & S_{vv} \end{bmatrix}$$

Scattering matrix

$$\mathbf{k} = [S_1, S_2, \dots, S_m]^T$$

Scattering vector

Description for
PolSAR data

**Point scatterers
characterization**

Description for
generalized multidimensional
SAR data

$$\mathbf{k}\mathbf{k}^H = \begin{bmatrix} S_{hh}S_{hh}^H & \sqrt{2}S_{hh}S_{hv}^H & S_{hh}S_{vv}^H \\ \sqrt{2}S_{hv}S_{hh}^H & 2S_{hv}S_{hv}^H & \sqrt{2}S_{hv}S_{vv}^H \\ S_{vv}S_{hh}^H & \sqrt{2}S_{vv}S_{hv}^H & S_{vv}S_{vv}^H \end{bmatrix}$$

Covariance matrix

$$\mathbf{k}\mathbf{k}^H = \begin{bmatrix} S_1S_1^H & S_1S_2^H & \dots & S_1S_m^H \\ S_2S_1^H & S_2S_2^H & \dots & S_2S_m^H \\ \vdots & \vdots & \ddots & \vdots \\ S_mS_1^H & S_mS_2^H & \dots & S_mS_m^H \end{bmatrix}$$

$$E\{\mathbf{k}\} = \mathbf{0}$$

Characterizes completely
the data distribution

**Distributed scatterers
characterization**

$$E\{\mathbf{k}\mathbf{k}^H\} = \mathbf{C}$$

Covariance matrix

$$p_{\mathbf{k}}(\mathbf{k}) = \frac{1}{\pi^m |\mathbf{C}|} \exp(-\mathbf{k}^H \mathbf{C}^{-1} \mathbf{k})$$

Covariance matrix estimation by means of a **MultiLook** (BoxCar)

- **Maximum likelihood** estimator: Sample covariance matrix

$$\mathbf{Z}_n = \frac{1}{n} \sum_{k=1}^n \mathbf{k}\mathbf{k}^H \quad \rightarrow \quad \text{Averaging usually done in the spatial domain (assuming homogeneous samples)}$$

- n represents the total number of samples employed to estimate the covariance matrix, taken a region (square, rectangular, adapted...)
- \mathbf{Z}_n as estimator of V
 - Does not consider signal morphology/heterogeneity
 - **Loss of spatial resolution**

The sample covariance matrix \mathbf{Z}_n is itself a multidimensional random variable

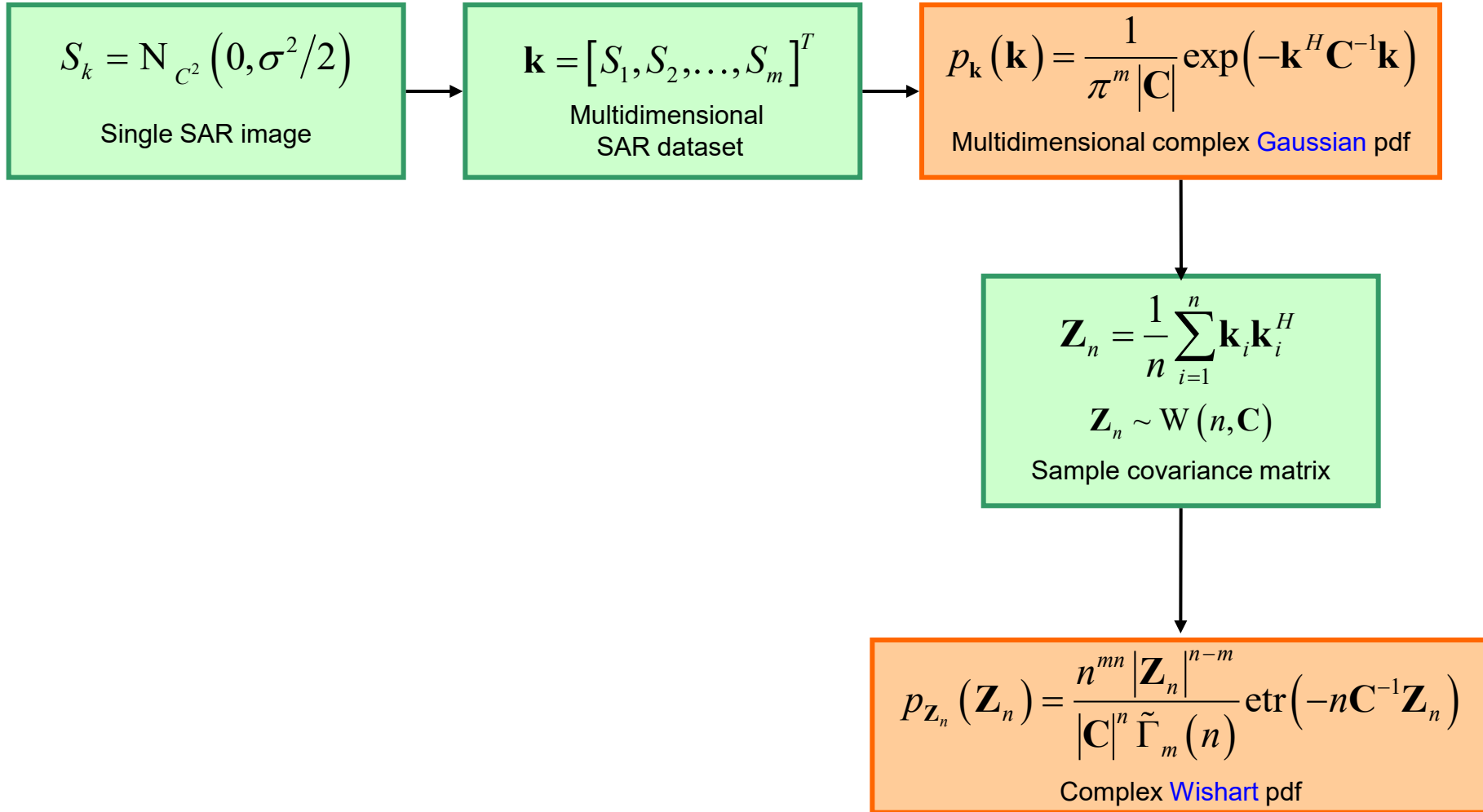
The sample covariance matrix \mathbf{Z}_n is characterized by the **complex Wishart distribution**

$$\mathbf{Z}_n \sim \mathbf{W}(n, \mathbf{C})$$

$$p_{\mathbf{Z}_n}(\mathbf{Z}_n) = \frac{n^{mn} |\mathbf{Z}_n|^{n-m}}{|\mathbf{C}|^n \tilde{\Gamma}_m(n)} \text{etr}(-n\mathbf{C}^{-1}\mathbf{Z}_n)$$

$$\tilde{\Gamma}_m(n) = \pi^{m(m-1)/2} \prod_{i=1}^m \Gamma(n-i+1)$$

- Multidimensional data distribution
- Valid for $n \geq m$, otherwise $|\mathbf{Z}_n|^{n-m}$ is equal to zero and the Wishart pdf is undetermined





Original data

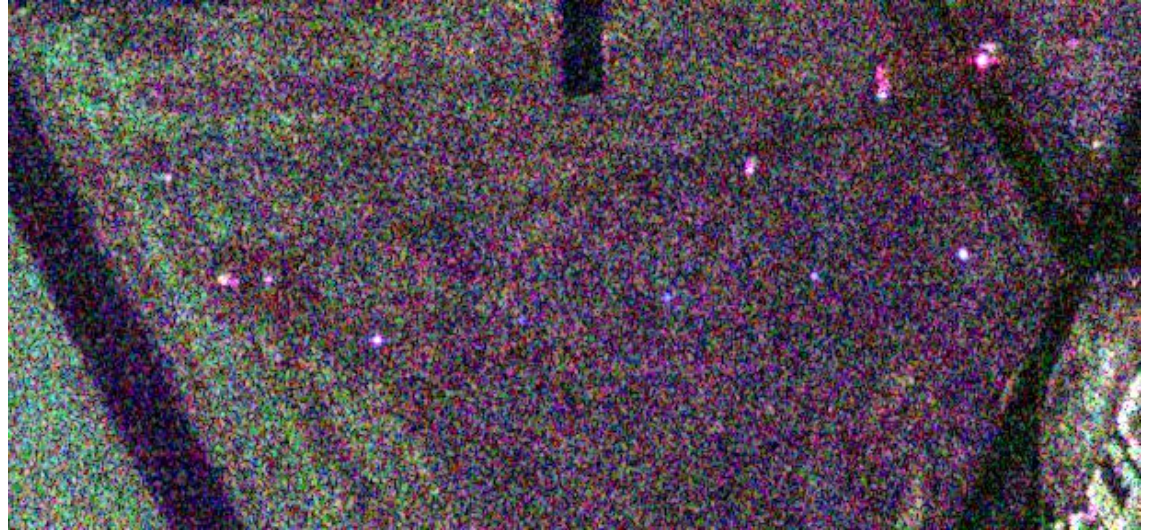
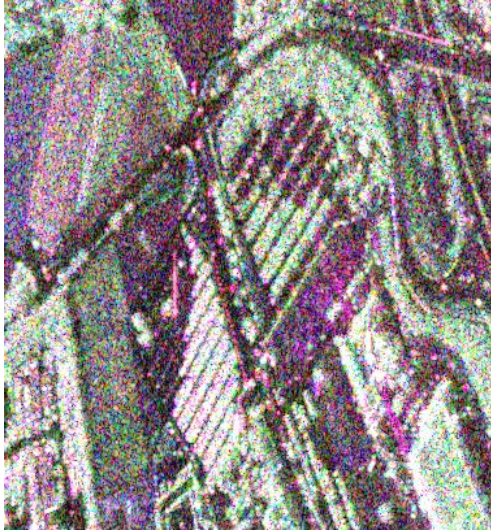


7x7 MLT data

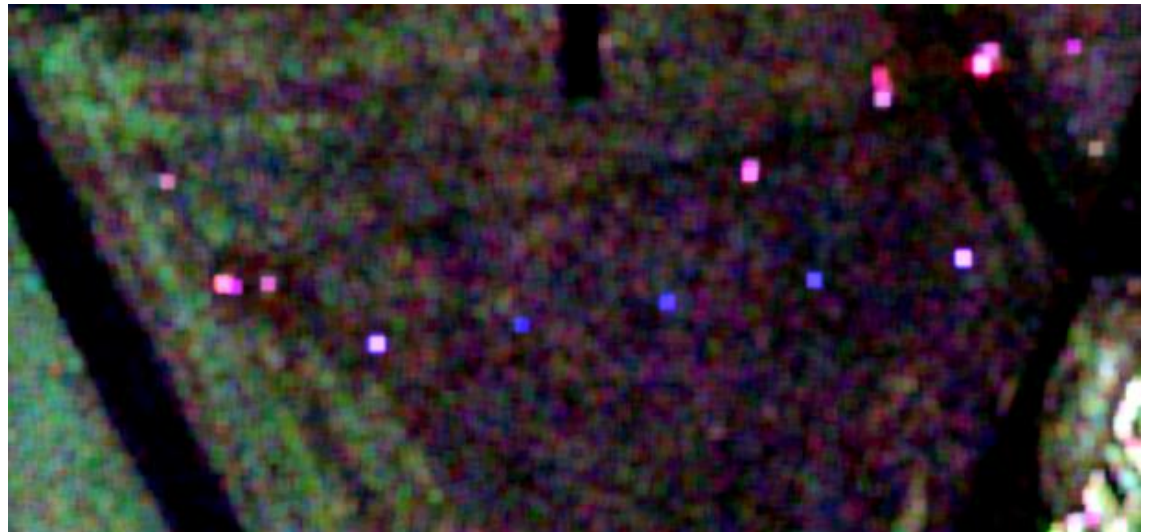
$|Shh-Svv|$ $2|Shv|$ $|Shh+Svv|$

L-band (1.3 GHz) fully PolSAR data
E-SAR system. Oberpfaffenhofen test area (D)

Original data



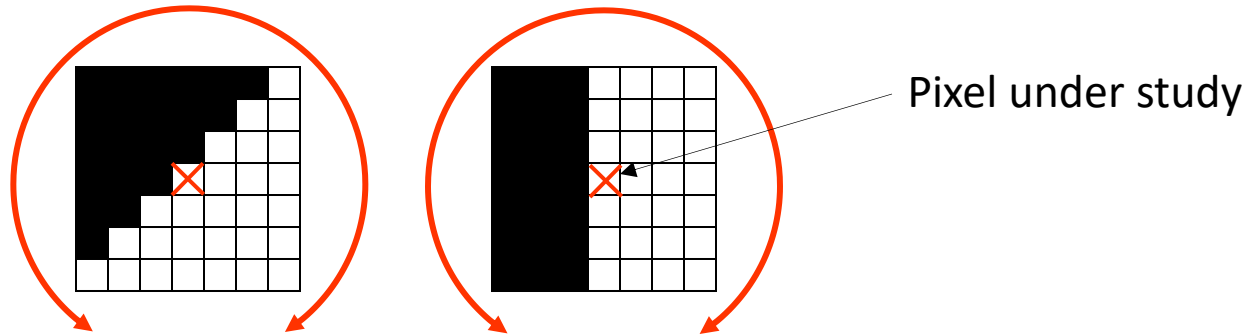
7x7 MLT data



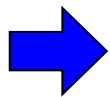
$$|Shh-Svv| \quad 2|Shv| \quad |Shh+Svv|$$

Refined Lee filter

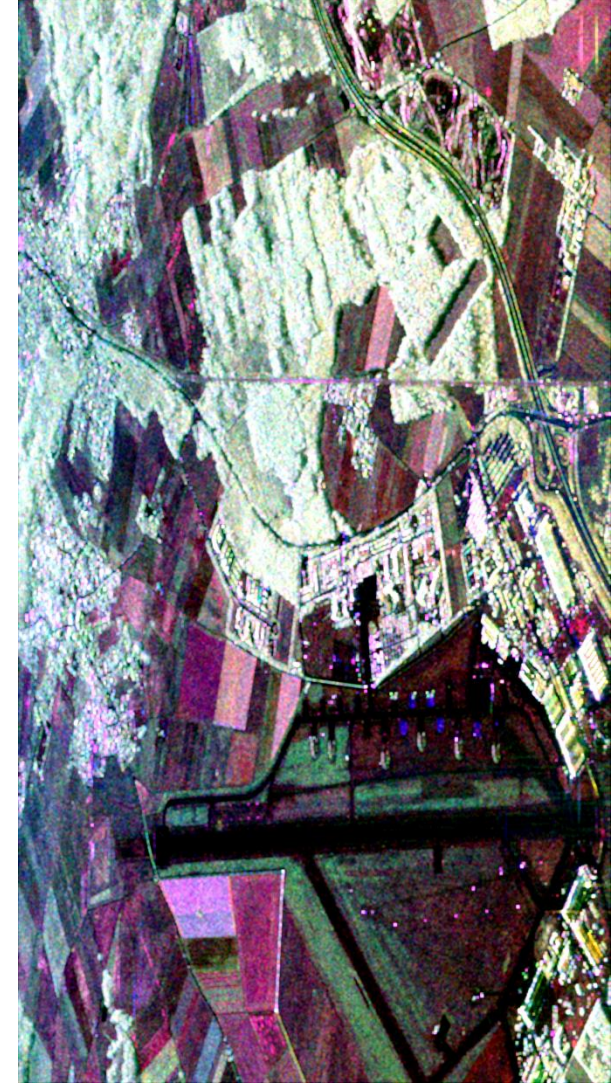
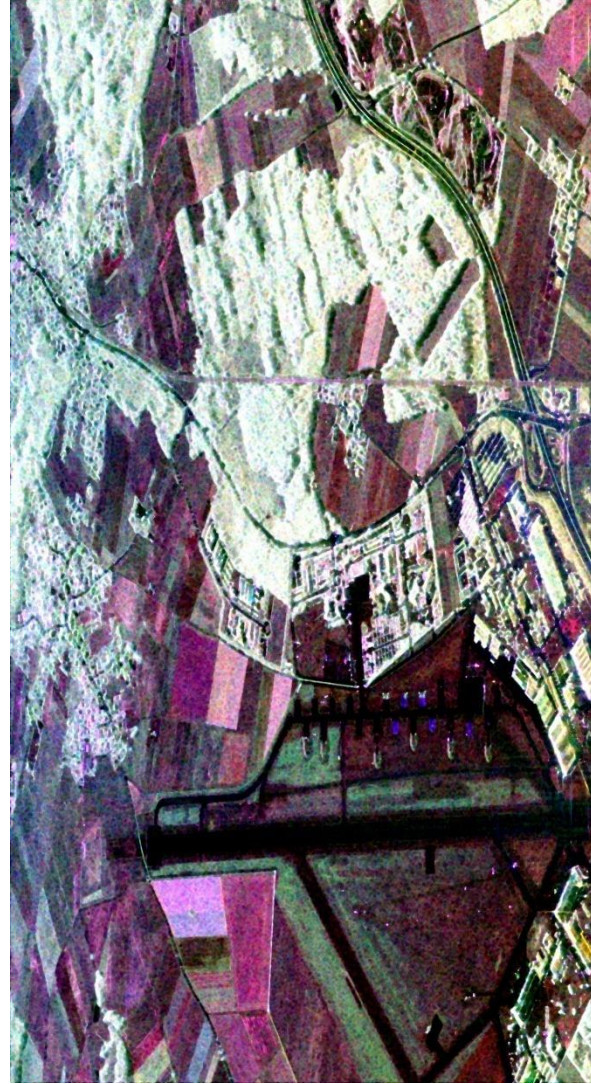
Statistics estimation in windows selected according to the signal morphology in order to retain edges, spatial feature and point targets



The directional windows try to adapt to image structure

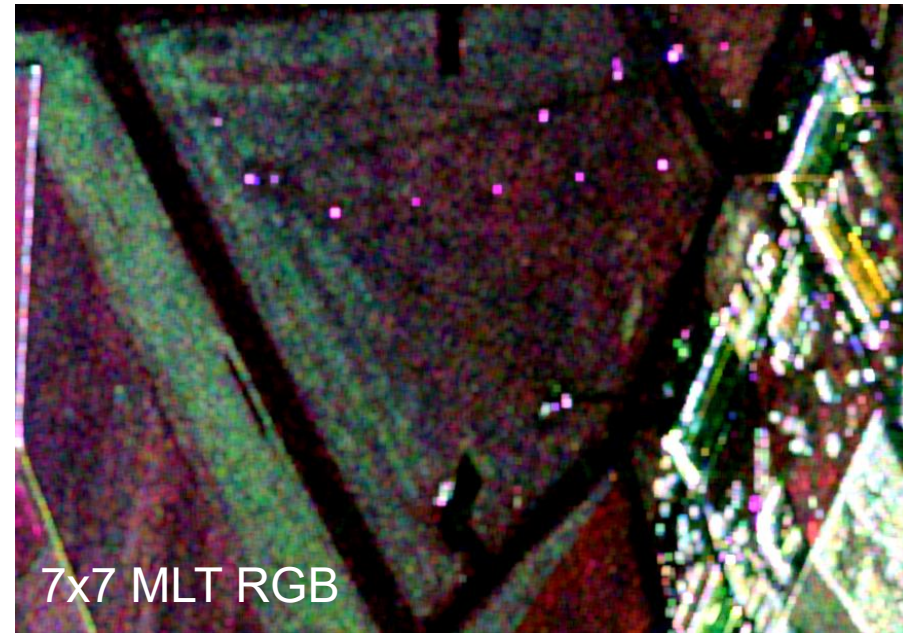
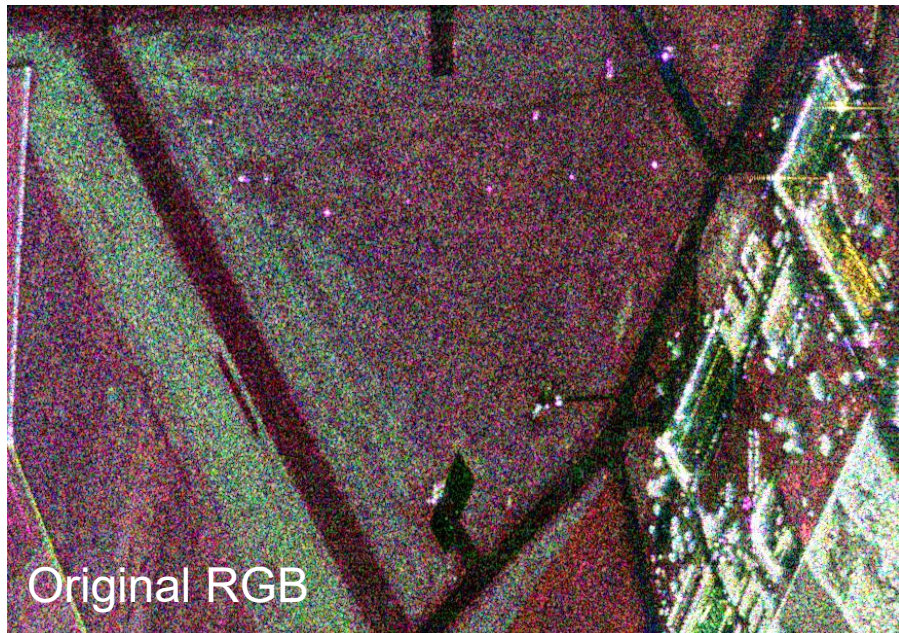


Decision needs to take into account statistical distribution to avoid the introduction of bias or distortion



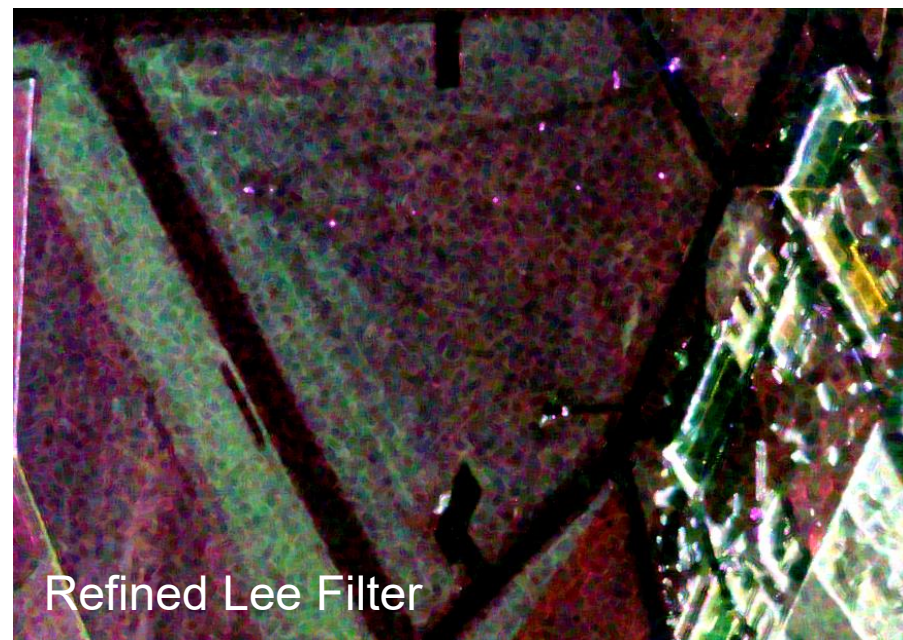
$|Shh|$ $|Shv|$ $|Svv|$

L-band (1.3 GHz) fully PolSAR data
E-SAR system. Oberpfaffenhofen test area (D)



$|Shh|$ $|Shv|$ $|Svv|$

L-band (1.3 GHz) fully **PoISAR** data
E-SAR system. Oberpfaffenhofen test area (D)



- Polarimetric Target Decomposition Theorems
- Coherent Decompositions
- Incoherent Decompositions
- Examples

Polarimetric Decompositions allow the interpretation of measured PolSAR data by decomposing them into more explainable terms

■ Coherent Decompositions

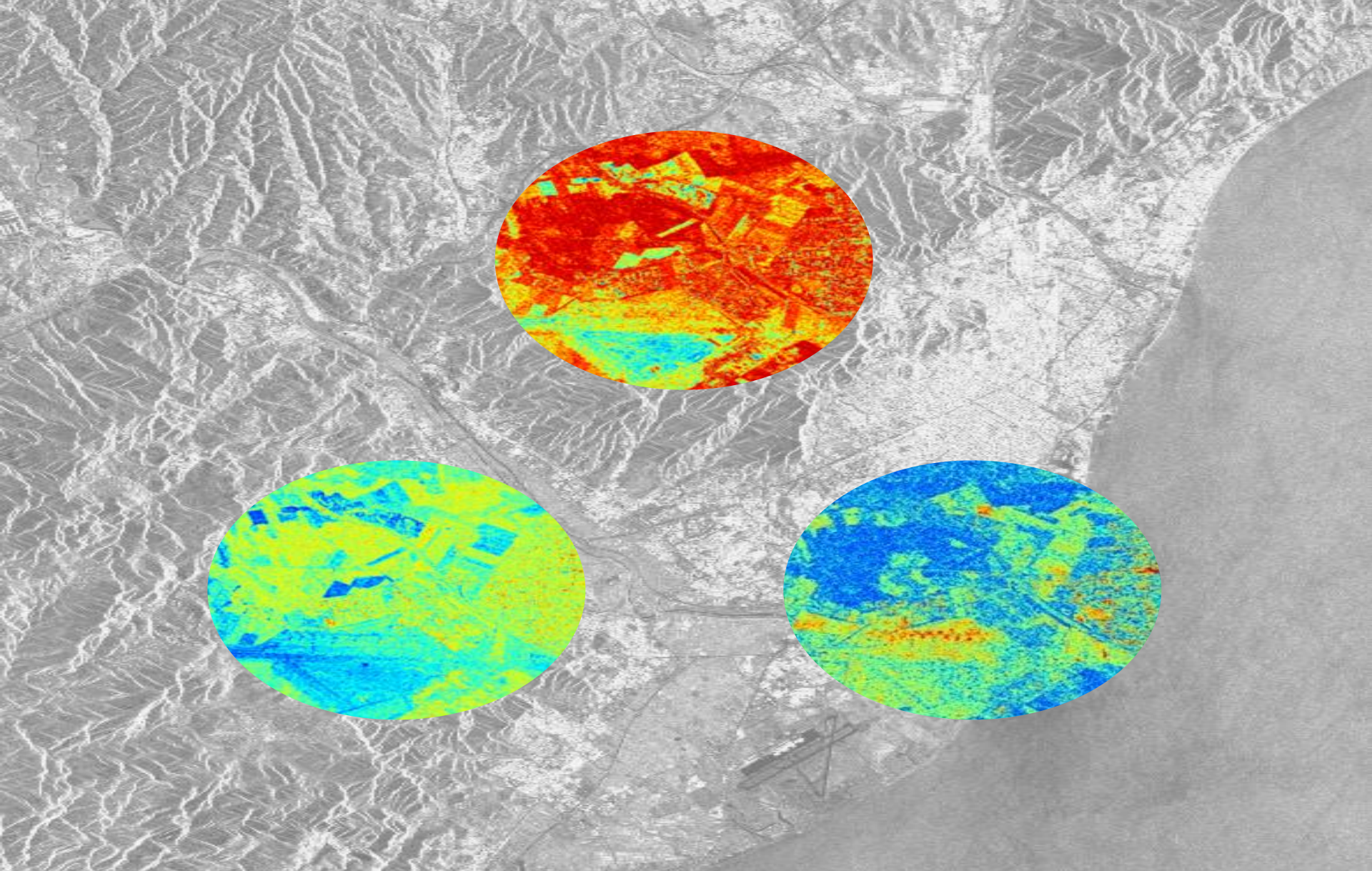
- Applied to the first order descriptors, i.e., the Scattering Matrix
- Valid for the interpretation of Pure or Deterministic scatters
- Decomposition of the data into canonical scattering mechanisms

$$\mathbf{S} = \sum_{i=1}^k c_i \mathbf{S}_i$$

■ Incoherent Decompositions

- Applied to second order descriptors, i.e., the Covariance and Coherency matrices
- Valid for the interpretation of Deterministic and Distributed scatters
- Decomposition of the data into simple scattering mechanisms that may, or not, admit an equivalent scattering matrix

$$\langle \mathbf{C} \rangle = \sum_{i=1}^k c_i \mathbf{C}_i \qquad \langle \mathbf{T} \rangle = \sum_{i=1}^k c_i \mathbf{T}_i$$



Coherent Decompositions

Based on the decomposition of the measured scattering matrix into the orthogonal Pauli decompositions basis

■ Basis components

$$\mathbf{S}_a = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad \mathbf{S}_b = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \quad \mathbf{S}_c = \frac{1}{\sqrt{2}} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad \mathbf{S}_d = \frac{1}{\sqrt{2}} \begin{bmatrix} 0 & -j \\ j & 0 \end{bmatrix}$$

Single bounce

Double bounce

Volume

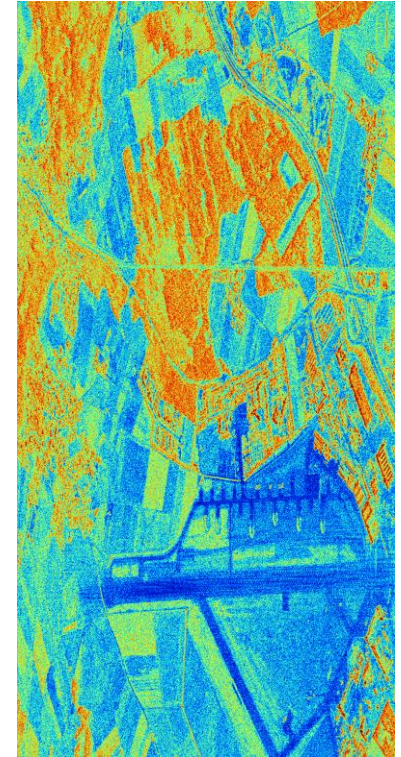
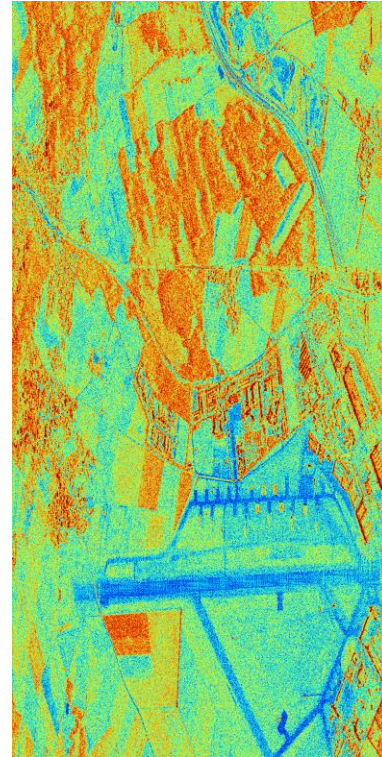
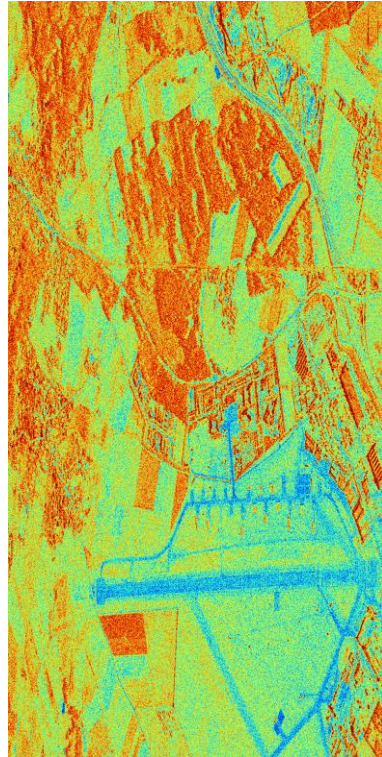
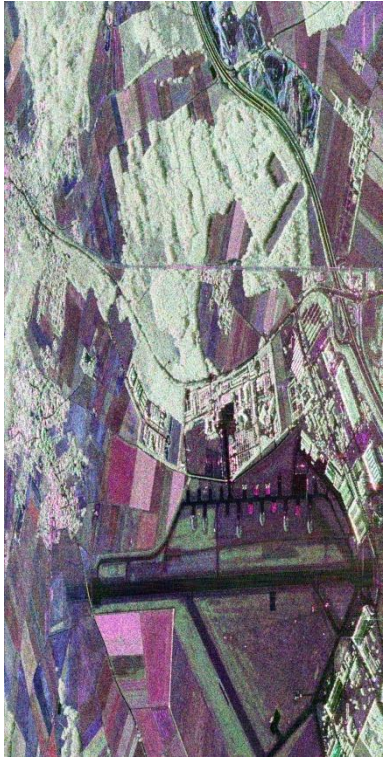
- Admit a physical interpretation into simple scattering mechanisms
- Orthogonal scattering mechanisms

■ Decomposition into three orthogonal scattering mechanisms (monostatic case)

$$\mathbf{S} = \begin{bmatrix} S_{hh} & S_{hv} \\ S_{hv} & S_{vv} \end{bmatrix} = \alpha \mathbf{S}_a + \beta \mathbf{S}_b + \gamma \mathbf{S}_c$$

■ Decomposition coefficients

$$\alpha = \frac{S_{hh} + S_{vv}}{\sqrt{2}} \quad \beta = \frac{S_{hh} - S_{vv}}{\sqrt{2}} \quad \gamma = \sqrt{2} S_{hv}$$



$|S_{hh}-S_{vv}|$ $2|S_{hv}|$ $|S_{hh}+S_{vv}|$

$$\alpha = \frac{S_{hh} + S_{vv}}{\sqrt{2}}$$

$$\beta = \frac{S_{hh} - S_{vv}}{\sqrt{2}}$$

$$\gamma = \sqrt{2}S_{hv}$$



Two type of decompositions

- **Model-based decompositions:** based on components that model physical scattering mechanisms
 - Simple physical interpretation
 - Assumptions concerning the type of scattering are necessary

- **Mathematical-based decompositions:** based on mathematical or spectral decompositions, as the eigenvalues/eigenvectors
 - Simple computation of the decomposition
 - No assumptions about the type of scattering are necessary
 - Physical interpretation of the different decomposition components are necessary

S

COHERENT DECOMPOSITION

E. KROGAGER (1990)

W.L. CAMERON (1990)

[K]

TARGET DICHOTOMY

J.R. HUYNEN (1970)

R.M. BARNES (1988)

TC

AZIMUTHAL SYMMETRY

EIGENVECTORS BASED DECOMPOSITION

S.R. CLOUDE (1985)

W.A. HOLM (1988)

MODEL BASED DECOMPOSITION

A.J. FREEMAN (1992)

EIGENVECTORS / EIGENVALUES ANALYSIS & MODEL BASED DECOMPOSITION

J.J. VAN ZYL (1992)

EIGENVECTORS / EIGENVALUES ANALYSIS ENTROPY / ANISOTROPY

S.R. CLOUDE - E. POTTIER (1996-1997)

Incoherent decomposition proposed by A. Freeman (1992)

- Decomposition

$\langle \mathbf{T} \rangle$ $\langle \mathbf{C} \rangle$ Three components scattering mechanism model

\mathbf{T}_S Single bounce
 \mathbf{C}_S scattering

\mathbf{T}_D Double bounce
 \mathbf{C}_D scattering

\mathbf{T}_V Volume
 \mathbf{C}_V scattering

$$\langle \mathbf{T} \rangle = f_S \mathbf{T}_S + f_D \mathbf{T}_D + f_V \mathbf{T}_V$$

$$\langle \mathbf{C} \rangle = f_S \mathbf{C}_S + f_D \mathbf{C}_D + f_V \mathbf{C}_V$$

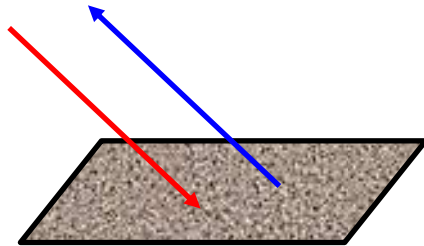
Decomposition coefficients

Incoherent decomposition proposed by A. Freeman (1992)

- Decomposition

$\langle \mathbf{T} \rangle$ $\langle \mathbf{C} \rangle$ \rightarrow Three components scattering mechanism model

\mathbf{T}_S Single bounce scattering
 \mathbf{C}_S



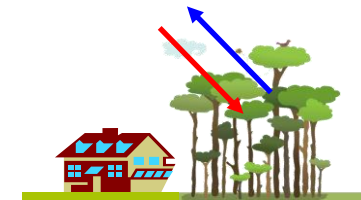
$$\mathbf{C}_S = f_S \begin{bmatrix} \beta^2 & 0 & \beta \\ 0 & 0 & 0 \\ \beta & 0 & 1 \end{bmatrix}$$

\mathbf{T}_D Double bounce scattering
 \mathbf{C}_D



$$\mathbf{C}_D = f_D \begin{bmatrix} \alpha^2 & 0 & -\alpha \\ 0 & 0 & 0 \\ -\alpha & 0 & 1 \end{bmatrix}$$

\mathbf{T}_V Volume scattering
 \mathbf{C}_V



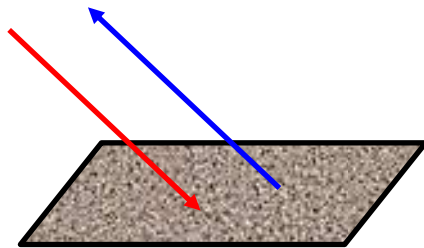
$$\mathbf{C}_V = f_V \begin{bmatrix} 1 & 0 & \frac{1}{3} \\ 0 & \frac{2}{3} & 0 \\ \frac{1}{3} & 0 & 1 \end{bmatrix}$$

Incoherent decomposition proposed by A. Freeman (1992)

- Decomposition

$\langle \mathbf{T} \rangle$ $\langle \mathbf{C} \rangle$ \rightarrow Three components scattering mechanism model

\mathbf{T}_S Single bounce scattering
 \mathbf{C}_S



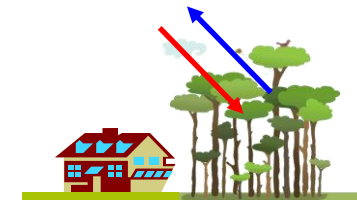
$$\mathbf{C}_S = f_S \begin{bmatrix} \beta^2 & 0 & \beta \\ 0 & 0 & 0 \\ \beta & 0 & 1 \end{bmatrix}$$

\mathbf{T}_D Double bounce scattering
 \mathbf{C}_D



$$\mathbf{C}_D = f_D \begin{bmatrix} \alpha^2 & 0 & -\alpha \\ 0 & 0 & 0 \\ -\alpha & 0 & 1 \end{bmatrix}$$

\mathbf{T}_V Volume scattering
 \mathbf{C}_V



$$\mathbf{C}_V = f_V \begin{bmatrix} 1 & 0 & \frac{1}{3} \\ 0 & \frac{2}{3} & 0 \\ \frac{1}{3} & 0 & 1 \end{bmatrix}$$

3 Components Scattering Mechanism Model

$$\langle \mathbf{C} \rangle = \begin{bmatrix} f_S \beta^2 + f_D \alpha^2 + f_V & 0 & f_S \beta - f_D \alpha + \frac{f_V}{3} \\ 0 & \frac{2f_V}{3} & 0 \\ f_S \beta - f_D \alpha + \frac{f_V}{3} & 0 & f_S + f_D + f_V \end{bmatrix}$$



5 Unknown real coefficients



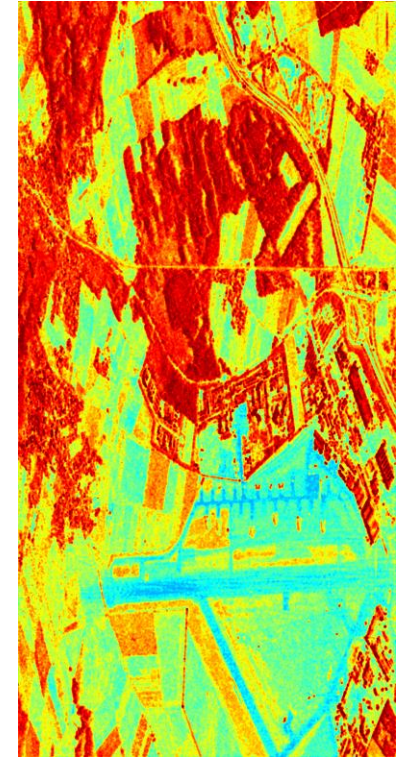
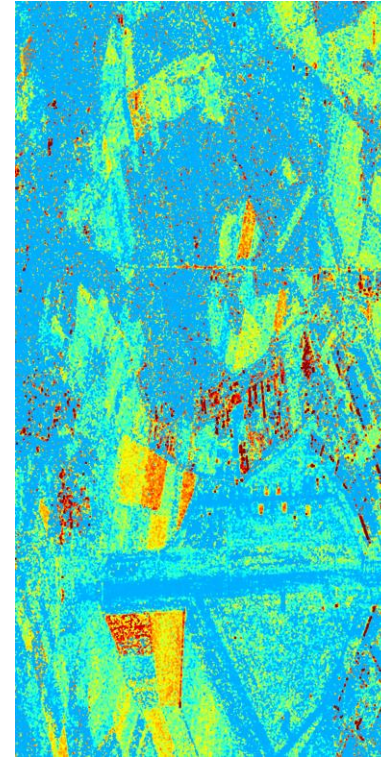
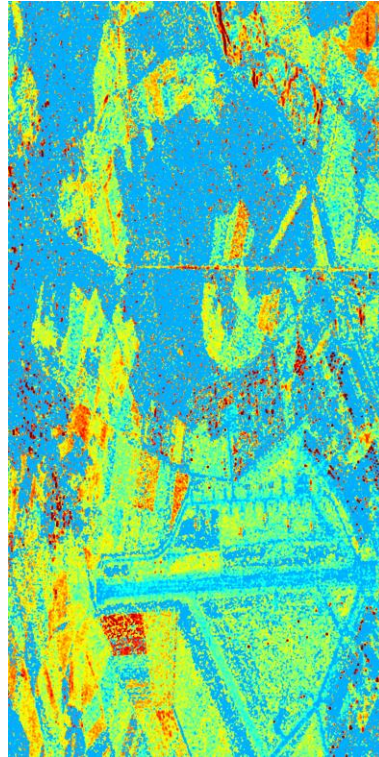
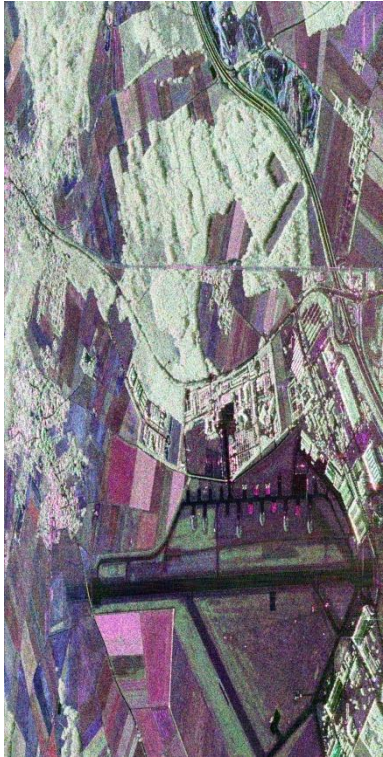
4 Observed equations



Negative power is observed

Assumptions must be considered !!!

7x7 Multilook

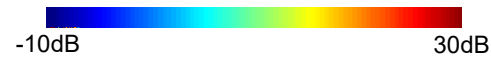


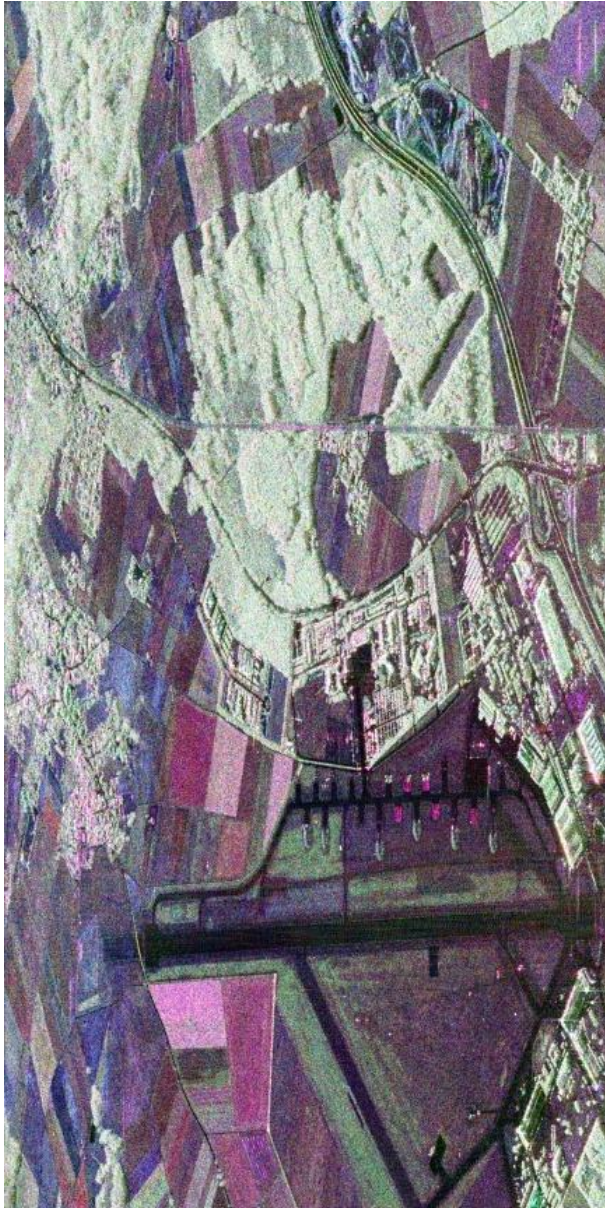
$|Shh-Svv|$ $2|Shv|$ $|Shh+Svv|$

f_S

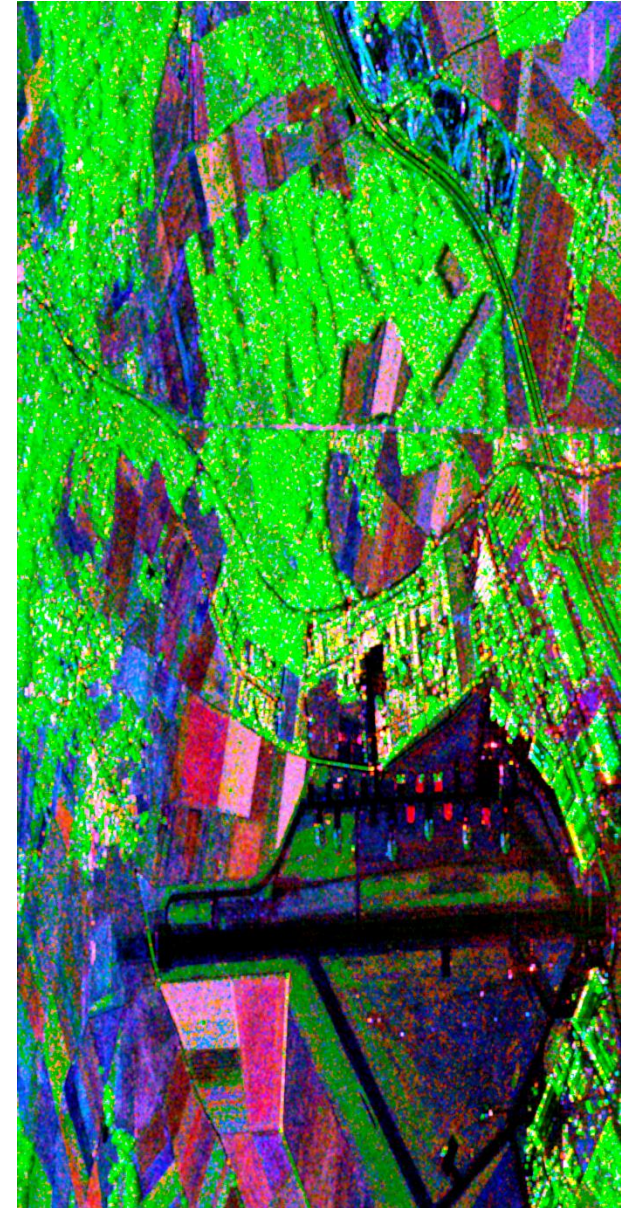
f_D

f_V





$|Shh-Svv|$ $2|Shv|$ $|Shh+Svv|$



$|fb|$ $|fv|$ $|fs|$

Decomposition proposed by Shane Cloude, based on the mathematical decomposition of the coherency matrix on its **eigenvalue and eigenvectors**

$$\mathbf{k}_p = \frac{1}{\sqrt{2}} \begin{bmatrix} S_{hh} + S_{vv} \\ S_{hh} - S_{vv} \\ 2S_{hv} \end{bmatrix}$$

Sample

$$\langle \mathbf{T} \rangle = \frac{1}{N} \sum_{i=1}^N \mathbf{k}_i \mathbf{k}_i^H = \frac{1}{N} \sum_{i=1}^N \mathbf{T}_i$$

Estimation of the covariance matrix

■ Decomposition (i)

$$\langle \mathbf{T} \rangle = \mathbf{U} \mathbf{\Sigma} \mathbf{U}^{-1} = \begin{bmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \mathbf{u}_3 \end{bmatrix} \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix} \begin{bmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \mathbf{u}_3 \end{bmatrix}^H$$

- The eigenvectors \mathbf{u}_1 , \mathbf{u}_2 and \mathbf{u}_3 are orthonormal
- The eigenvalues are real $\lambda_1 > \lambda_2 > \lambda_3 \geq 0$

- Decomposition components

- The **eigenvectors** represent rank 1 scattering mechanisms that are related with a scattering matrix
- Parametrization of the eigenvectors matrix

$$\mathbf{U} = \begin{bmatrix} \cos(\alpha_1) & \cos(\alpha_2) & \cos(\alpha_3) \\ \sin(\alpha_1)\cos(\beta_1)e^{j\delta_1} & \sin(\alpha_2)\cos(\beta_2)e^{j\delta_2} & \sin(\alpha_3)\cos(\beta_3)e^{j\delta_3} \\ \sin(\alpha_1)\sin(\beta_1)e^{j\gamma_1} & \sin(\alpha_2)\sin(\beta_2)e^{j\gamma_2} & \sin(\alpha_3)\sin(\beta_3)e^{j\gamma_3} \end{bmatrix}$$

Target 1
Target 2
Target 3

Decomposition basis

- The **eigenvalues** represent the power associated to every target

 λ_1
 λ_2
 λ_3

Decomposition coefficients

- Decomposition (ii)

$$\langle \mathbf{T} \rangle = \mathbf{U} \mathbf{\Sigma} \mathbf{U}^{-1} = \lambda_1 \mathbf{u}_1 \mathbf{u}_1^H + \lambda_2 \mathbf{u}_2 \mathbf{u}_2^H + \lambda_3 \mathbf{u}_3 \mathbf{u}_3^H = \lambda_1 \mathbf{T}_1 + \lambda_2 \mathbf{T}_2 + \lambda_3 \mathbf{T}_3$$

- The coherency matrices \mathbf{T}_k are rank 1 matrices associated with a single scattering matrix
- It is possible to associate a probability to every scattering mechanism

$$P_i = \frac{\lambda_i}{\sum_{k=1}^3 \lambda_k} \qquad \text{Span} = \sum_{k=1}^3 \lambda_k$$

- Definition of the **mean dominant scattering mechanism**
 - Mean parameters

$$\bar{\alpha} = P_1 \alpha_1 + P_2 \alpha_2 + P_3 \alpha_3$$

- Analysis of the **eigenvalues**. These parameters classify the spectrum of scattering mechanisms

- **Entropy**: Degree of randomness or statistical disorder

$$H = -\sum_{i=1}^3 P_i \log_3(P_i)$$

- **Pure target** ($H=0$). One scattering mechanism present

$$\lambda_1 = \text{Span} \quad \lambda_2 = 0 \quad \lambda_3 = 0$$

- **Random target** ($H=1$). All scattering mechanisms equally probable

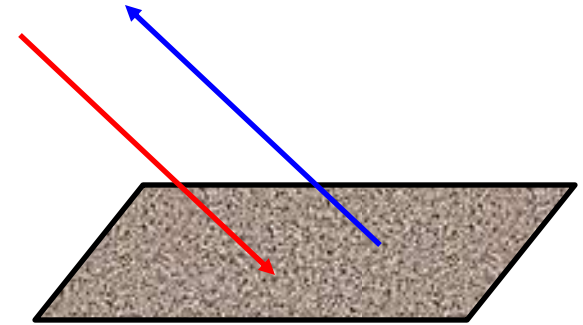
$$\lambda_1 = \lambda_2 = \lambda_3 = \text{Span} / 3$$

- **Anisotropy**: Relative importance of the 2nd and 3rd scattering mechanisms

$$A = \frac{\lambda_2 - \lambda_3}{\lambda_2 + \lambda_3}$$

- Physical interpretation of the average alpha angle parameter

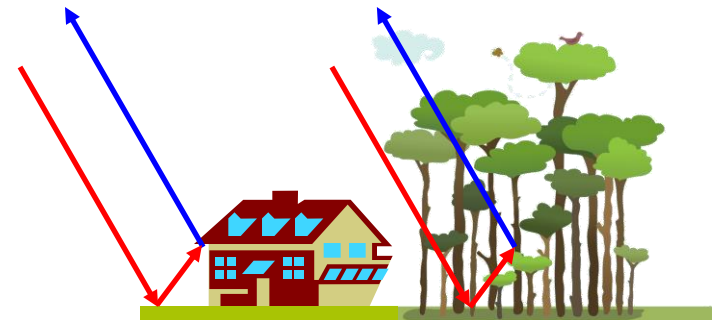
- Single bounce scattering, for instance, a surface



$$\alpha \approx 0$$

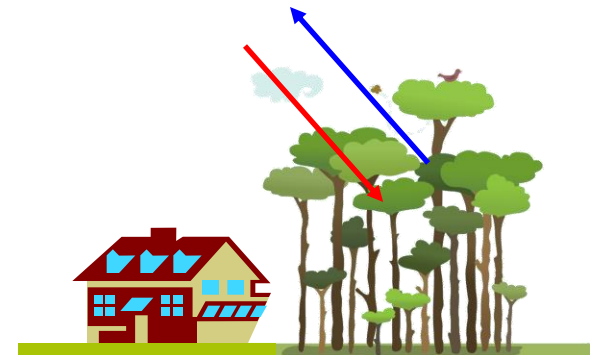
- Double bounce scattering

$$\alpha \approx \frac{\pi}{2}$$

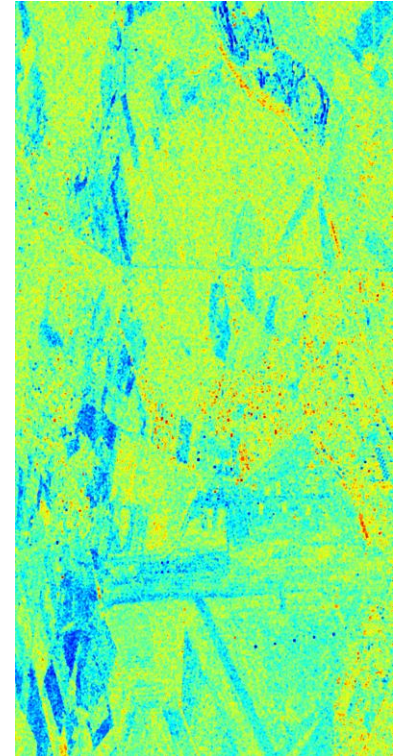
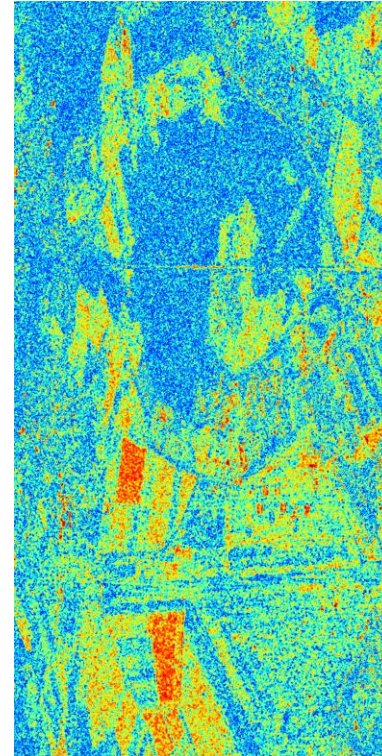
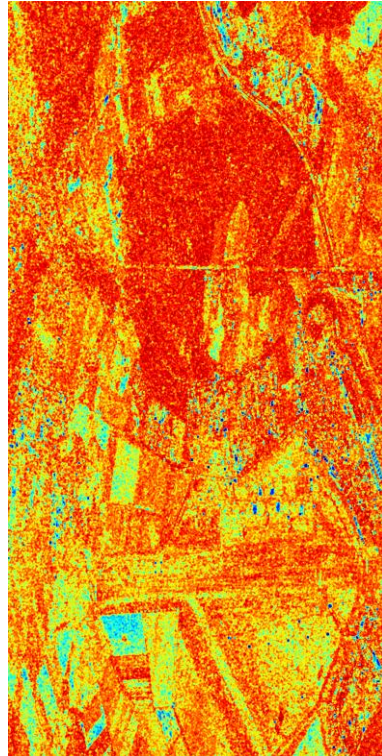
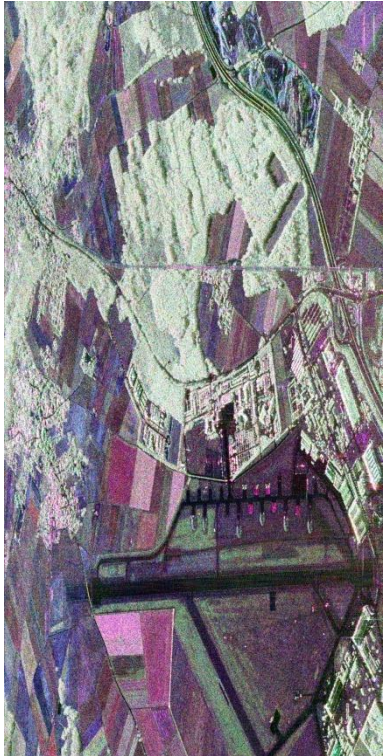


- Volume scattering

$$\alpha \approx \frac{\pi}{4}$$



7x7 Multilook

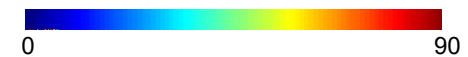
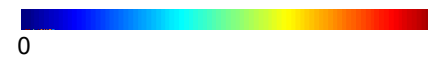
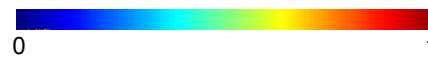


$|Shh-Svv|$ $2|Shv|$ $|Shh+Svv|$

H

A

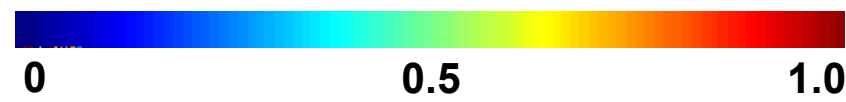
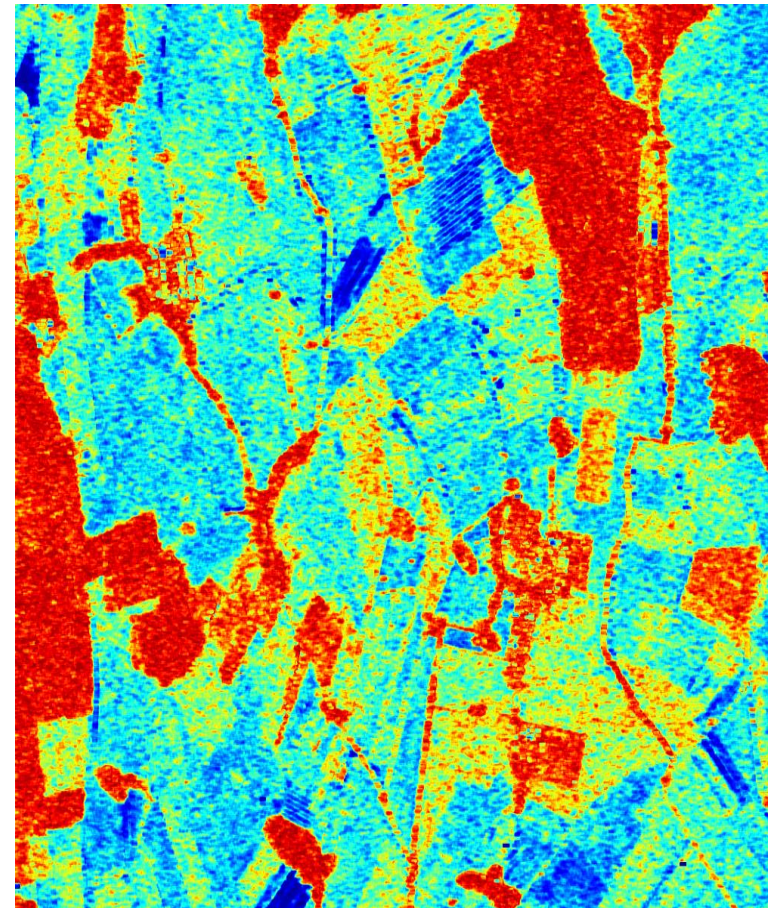
α



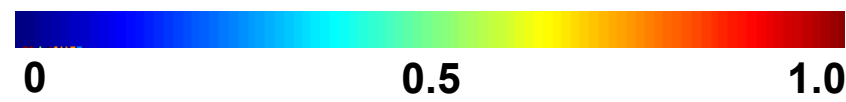
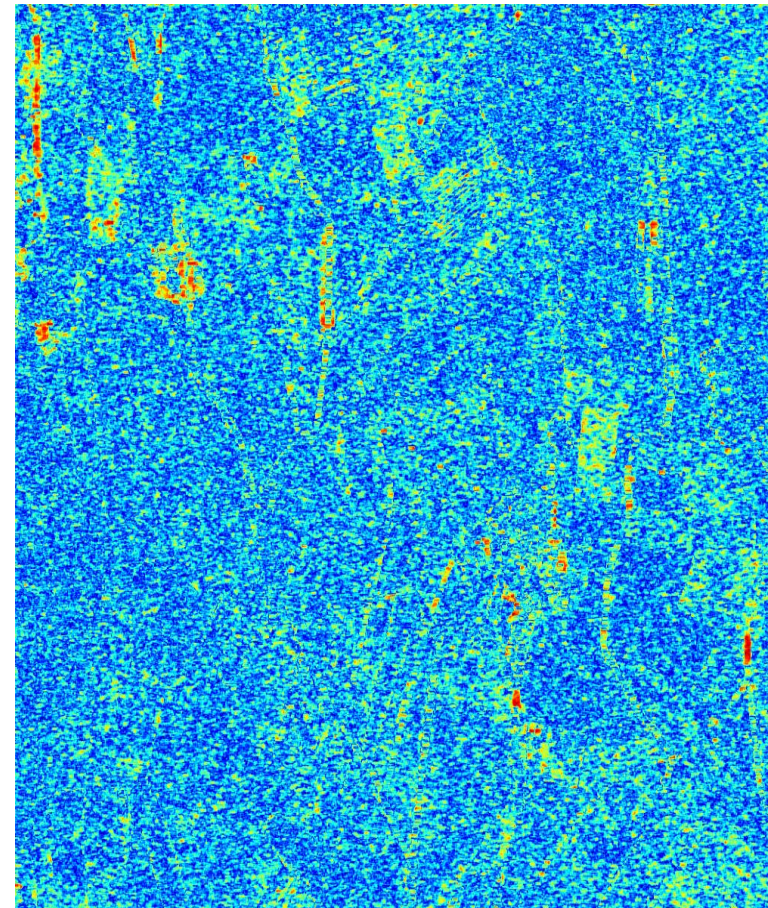


$|Shh|$ $|Shv|$ $|Svv|$

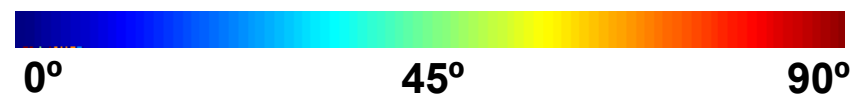
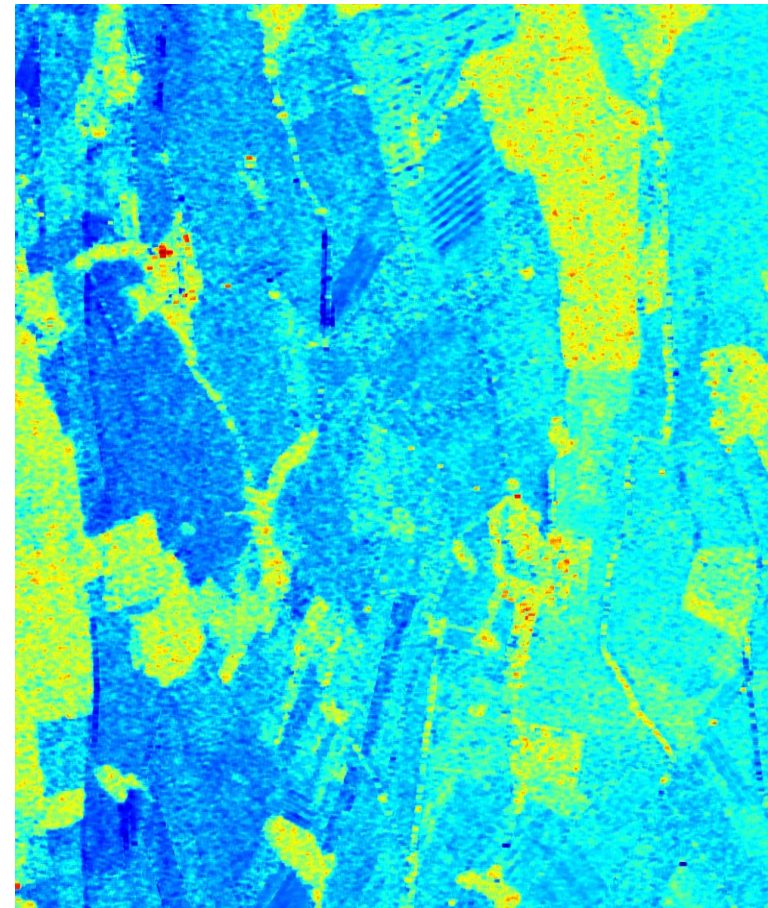
Entropy (H)



Anisotropy (A)



Alpha (α)



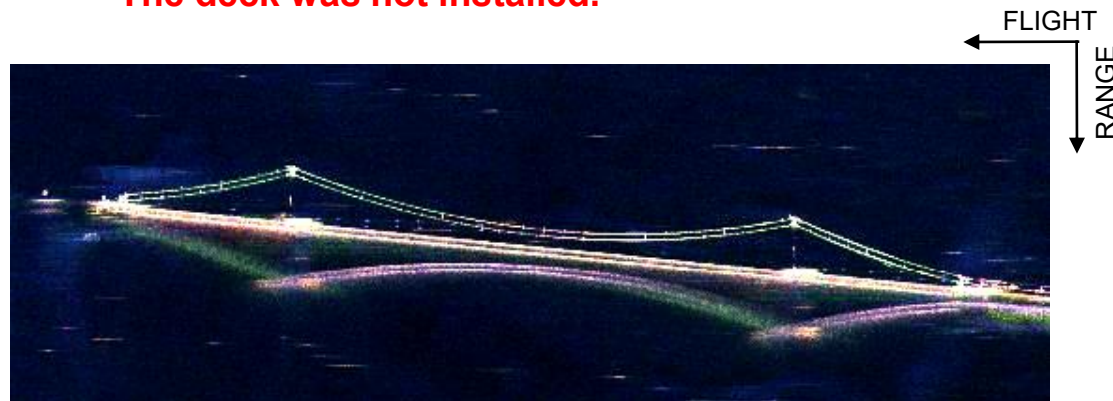


POLARIMETRIC TARGET SIGNATURES: An Example

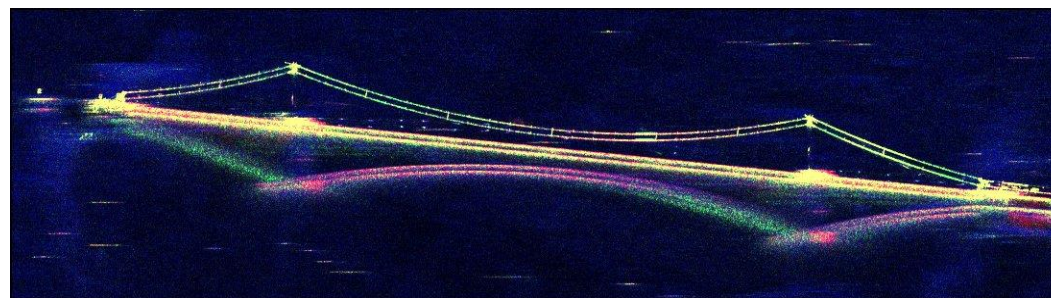
High-resolution POLSAR signature of a suspension bridge under construction
The deck was not installed.



Aerial Photo



Pauli Decomposition, $|HH-VV|$, $|HV|$, $|HH+VV|$



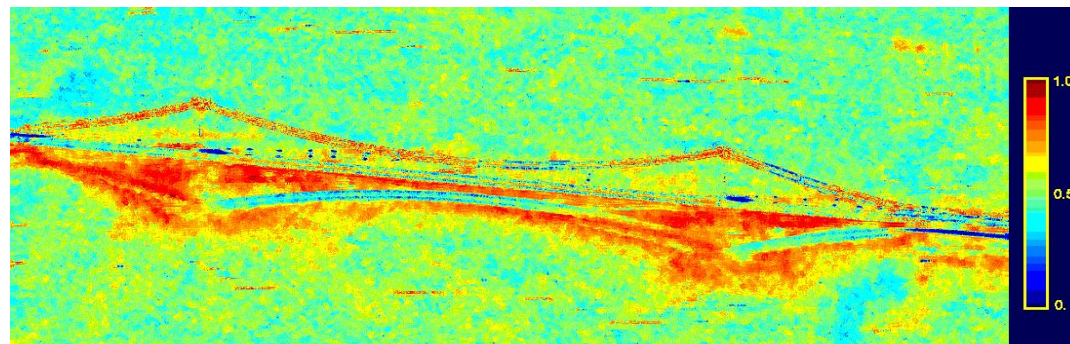
Krogager SDH Decomposition
Blue= Sphere Green = Helix Red = Diplane.

EMISAR C-Band Polarimetric SAR Image of StoreBelt Bridge

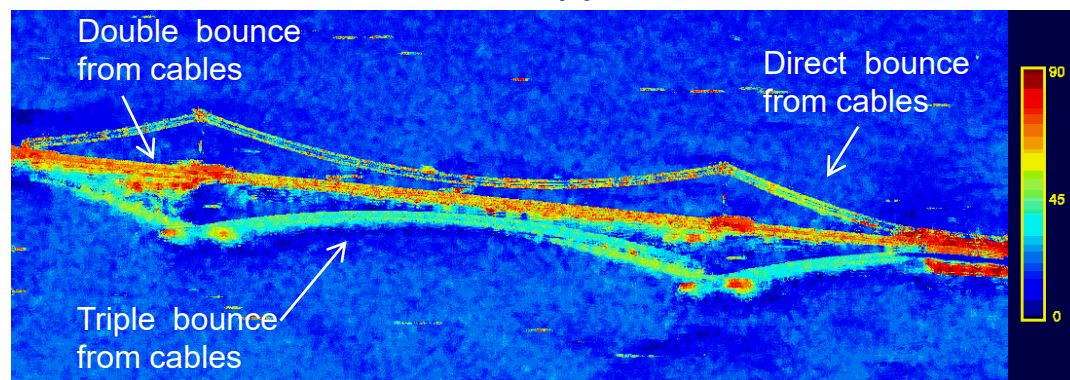
High-resolution POLSAR signature of a suspension bridge under construction
The deck was not installed.



Aerial Photo



H Entropy

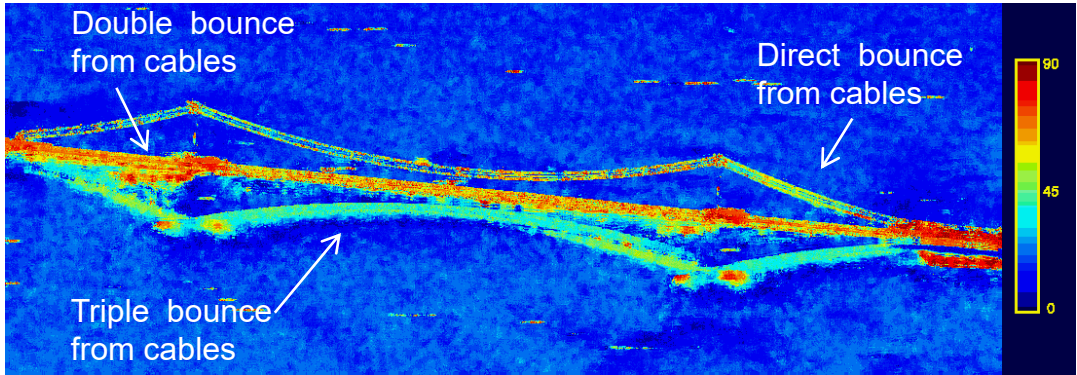
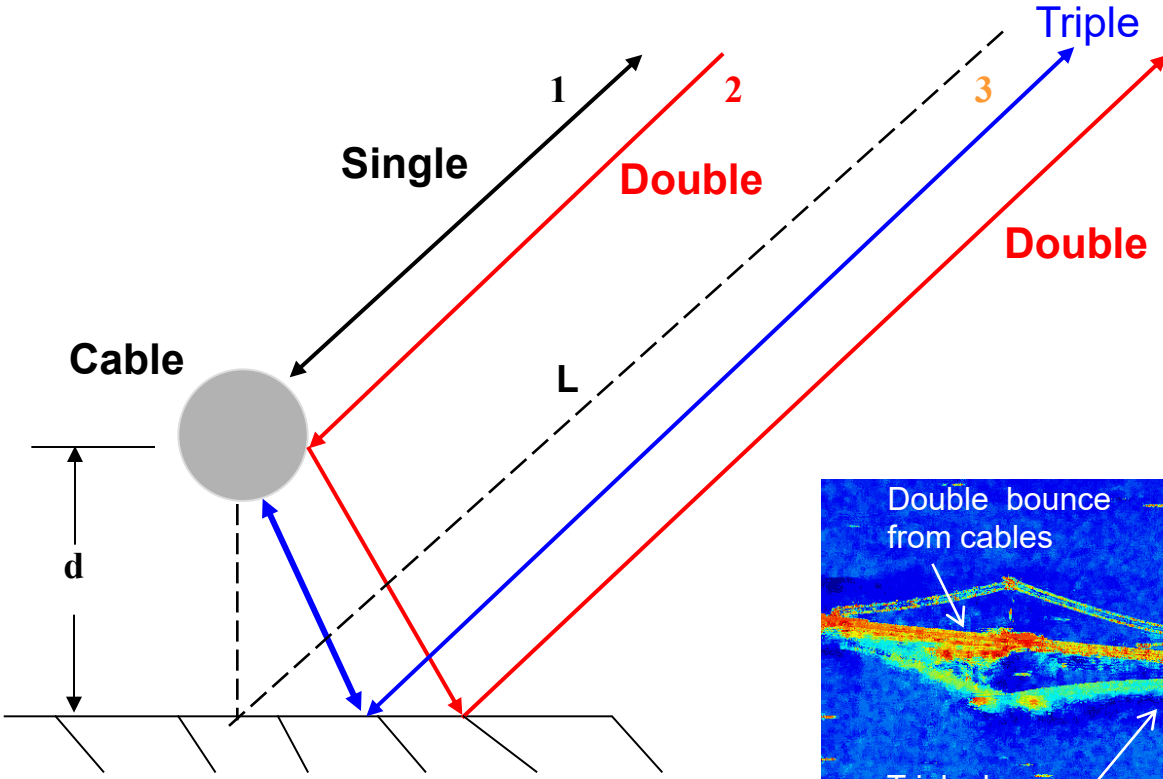


α Angle

Dominant scattering mechanisms are extracted by applying target decomposition:
Blue= Surface Green = Dipole Red = Double Bounce.

EMISAR C-Band Polarimetric SAR Image of StoreBelt Bridge

Multi – Bounce Scattering mechanisms



Roundtrip distances:

Single bounce return: $2(L - d \cos\theta)$

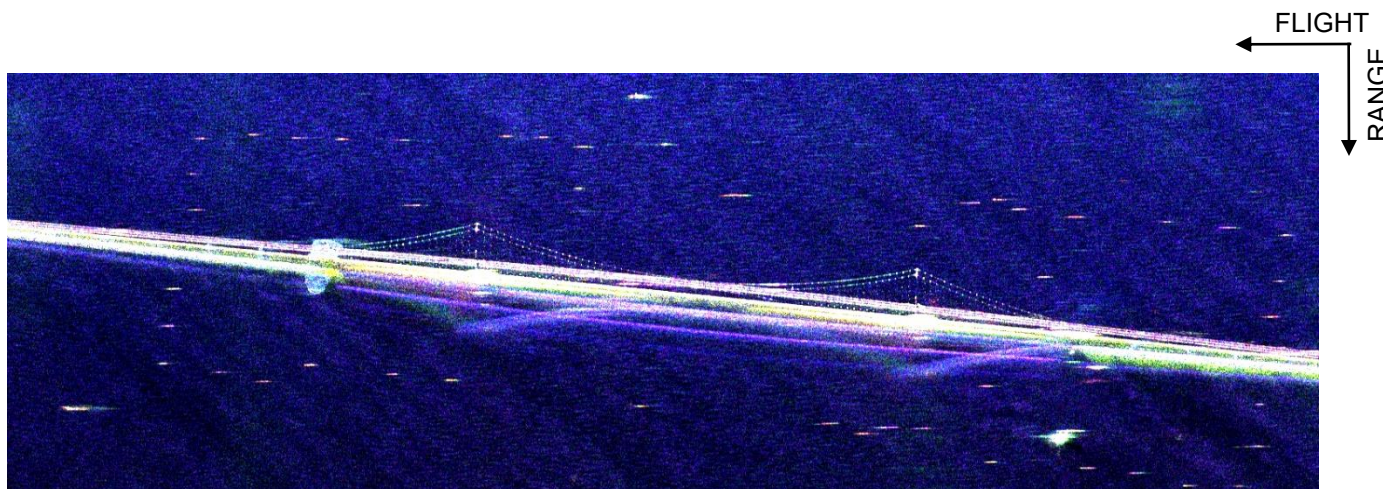
Double bounce return: $2L$ ← Independent of height

Triple bounce return: $2(L + d \cos\theta)$

θ is the local incidence angle.

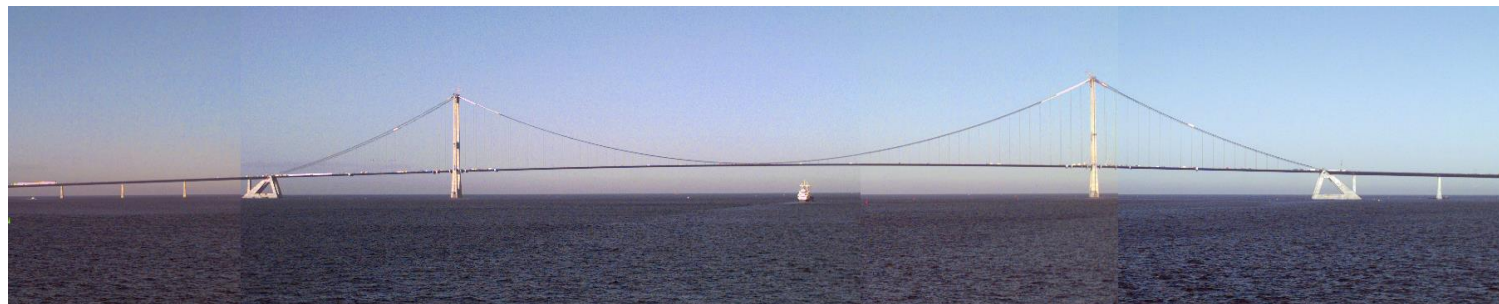
α Angle

High-resolution POLSAR signature of a suspension bridge after completion.
The deck is installed.

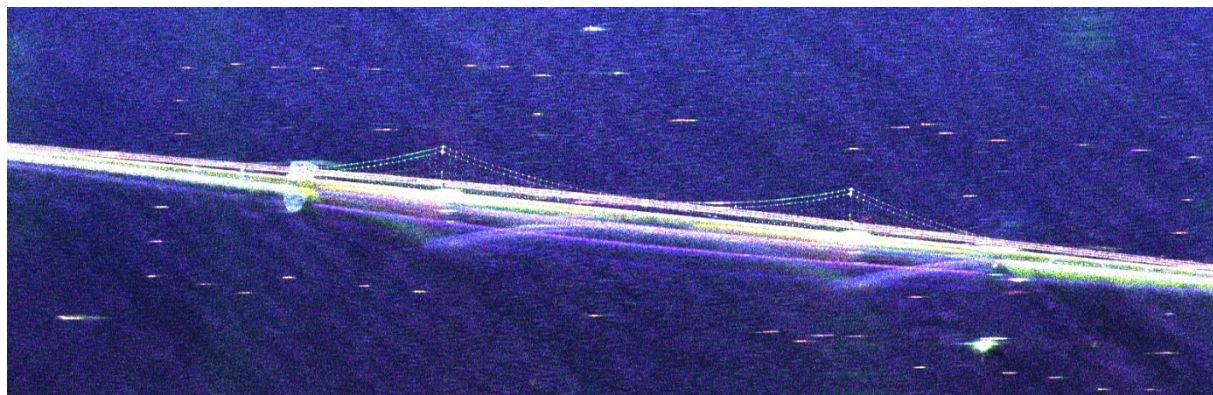


$|HH-VV|$, $|HV|$, $|HH+VV|$

EMISAR C-Band Polarimetric SAR Image of StoreBelt Bridge

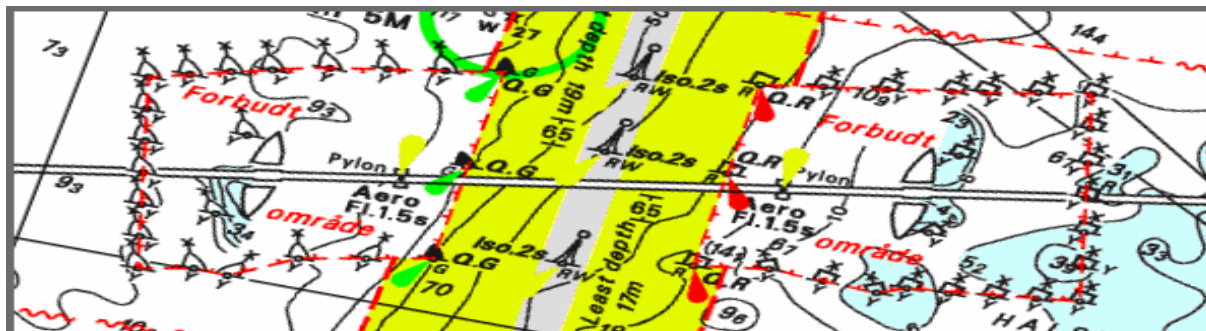


Buoy

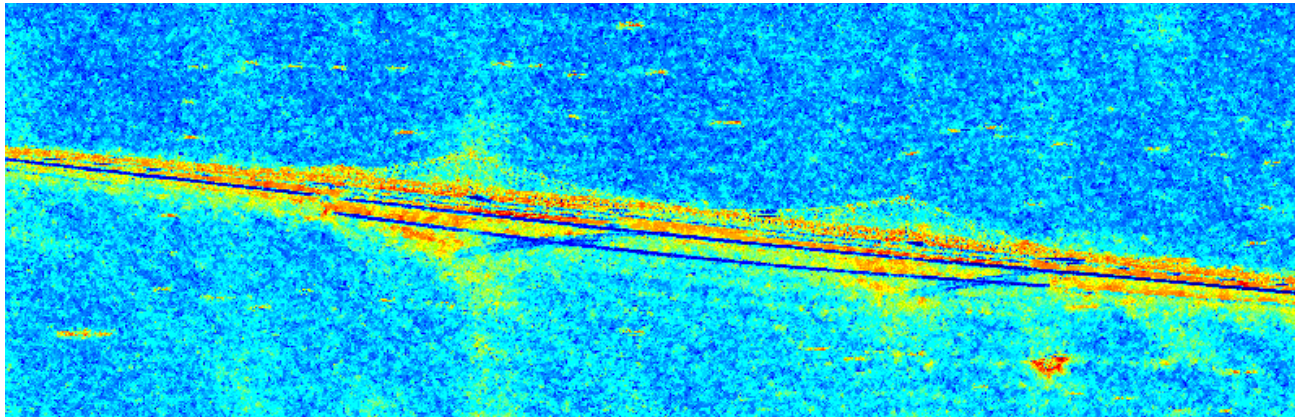


Buoy

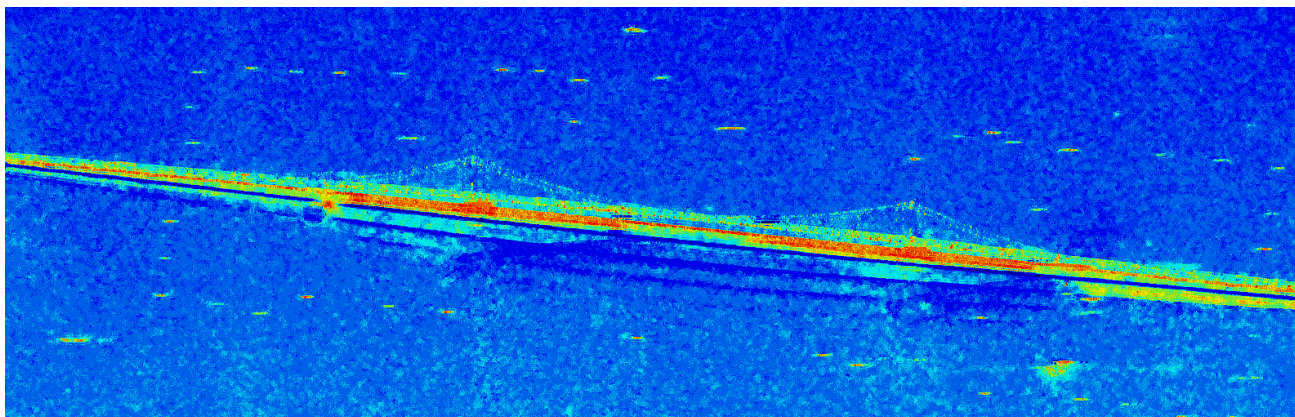
$$|HH-VV|, |HV|, |HH+VV|$$



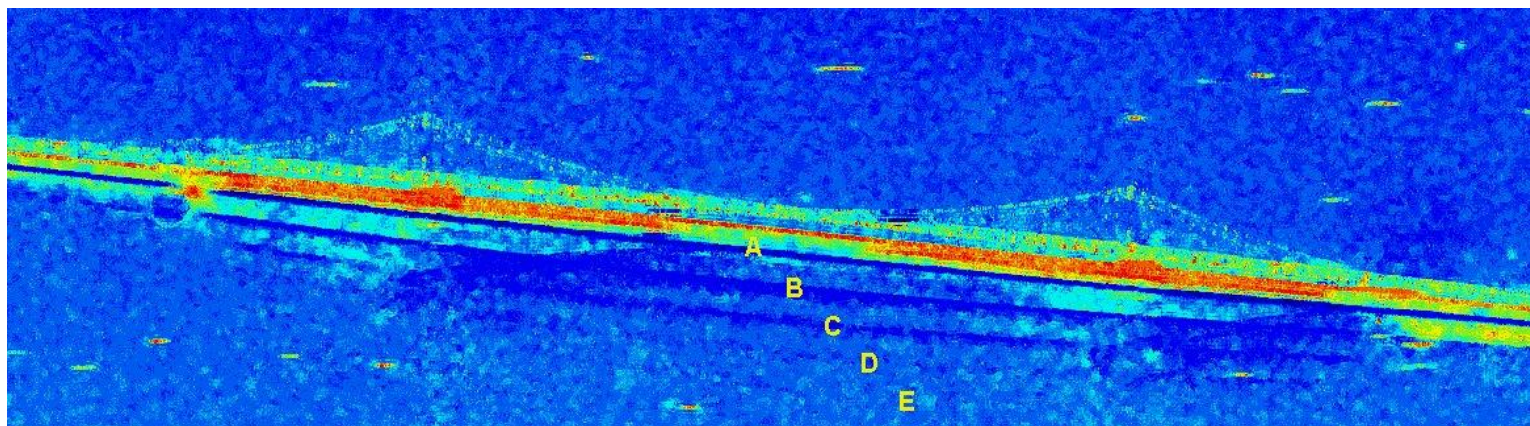
Navigation Map of Storebelt, Denmark



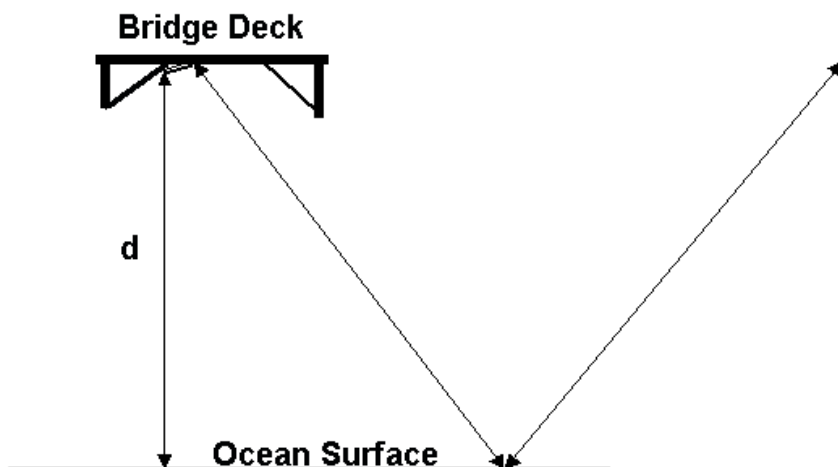
H Entropy



α Angle



α Angle



Roundtrip distances:

A: Triple bounces, $2(L+d\cos\theta)$

B: 5 (or 7) bounces, $2(L+d\cos\theta)+2d$

C: 7 (or 9) bounces, $2(L+d\cos\theta)+4d$

D: 9 (or 11) bounces, $2(L+d\cos\theta)+6d$

(...) indicates additional two bounces from the bottom of the deck.

RESONANT CAVITY

Dominant scattering mechanisms are extracted by applying target decomposition:

Blue = Surface Green = Dipole Red = Double Bounce

