





→ 8th ADVANCED TRAINING COURSE ON LAND REMOTE SENSING

10–14 September 2018 University of Leicester | United Kingdom

Forestry applications with Polarimetry and Interferometry (Lecture)

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SAR & Hyperspectral multi-modal Imaging and sigNal processing, **Electromagnetic modeling**



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Rennes - Britanny







Rennes - Britanny







Rennes - Britanny









To provide the minimum, but necessary, amount of knowledge required to understand and to practice :

SAR Polarimetry + Interferometry (Pol-InSAR)

for forestry applications



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🖉 Height Estimation from Inversion Procedures		
Input Master - Slave Directory		
C:/My_Data_Directory/Pol-InSAR_PolSARproSIM	_forest/master_slave_FER	
Output Master - Slave Directory		
C/My_Data_Directory/PoHnSAR_PoISARproSIM_forest/master_slave_FER		
Init Row 1 End Row	301 Init Col 1 End Col 301	
	Update List	
☐ Polarimetric Phase Centre Height Estimation	Polanmetric Channel HH 👤	
DEM Differencing Algorithm		
Coherence Amplitude Inversion Procedure		
Ground Phase Estimation and RVDG Inversion Procedure Median Window Size 21 Weighting Coherence Fraction Factor 0.4		
Top Phase Centre HV 👤 2D Kz File	Ground Phase Centre HH - W	
C:/My_Data_Directory/PoHnSAR_PoISARproSIM_forest/slave/kz,bin		
Run	Ем	

PolSARpro - practicals

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INVERSION PROCEDURES

- DEM Differencing Algorithm
- Coherence Amplitude Inversion Procedure
- Ground Phase Estimation &
- **RVOG Inversion Procedure**



🖉 Height Estimation from Inversion Proc	edures
Input Master - Slave Directory	
C:/My_Data_Directory/PoHnSAR_PoISARpr	SIM_forest/master_slave_FER
Output Master - Slave Directory	
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Init Row 1 End Row	301 Init Col 1 End Col 301
	Update List
F Polarimetric Phase Centre Height Estimatic	n Polarimetric Channel HH _
DEM Differencing Algorithm	
Coherence Amplitude Inversion Procedure	
Ground Phase Estimation and RVOG Inve	rsion Procedure
Median Window Size 21 💌	Weighting Coherence Fraction Factor 0.4
Top Phase Centre HV	Ground Phase Centre HH - W
C:/My_Data_Directory/Pol-InSAR_PolSARpr	pSIM_forest/slave/kz,bin
Run	(D Exit



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🦨 Height Estimation from Inversion Procedu		
Input Master - Slave Directory		
C./My_Data_Directory/PoHnSAR_PoISARproSIM_forest/master_slave_FER		
Output Master - Slave Directory		
C:/My_Data_Directory/PoHnSAR_PoISARproSIM_forest/master_slave_FER		
Init Row 1 End Row	301 Init Col 1 End Col 301	
	Update List	
F Polarimetric Phase Centre Height Estimation	Polarimetric-Channel HH	
DEM Differencing Algorithm		
Coherence Amplitude Inversion Procedure		
Ground Phase Estimation and RVDG Inversion Procedure		
Median Window Size 21 💌 🛋	Weighting Coherence Fraction Factor 0.4	
Top Phase Centre HV 💌	Ground Phase Centre HH - VV	
2D Kz File		
C:/My_Data_Directory/PoHnSAR_PolSARproSIM_forest/slave/kz,bin		
Run Hist	Exit	

Pol-InSAR

Interferometry

SAR



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2-D SAR imaging





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2-D SAR imaging





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2-D SAR imaging





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3-D In-SAR imaging







3-D In-SAR imaging





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3-D In-SAR imaging



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esa





Interferometric coherence γ



 $\gamma = \frac{E(s_1 s_2^*)}{\sqrt{E(s_1 s_1^*)E(s_2 s_2^*)}} = \frac{E(s_1 s_2^*)}{\sqrt{\overline{I}_1 \overline{I}_2}} = |\gamma| e^{j\phi}$



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S₂

Interferometric coherence γ



$$=\frac{E\left(s_{1}s_{2}^{*}\right)}{\sqrt{E\left(s_{1}s_{1}^{*}\right)E\left(s_{2}s_{2}^{*}\right)}}=\frac{E\left(s_{1}s_{2}^{*}\right)}{\sqrt{I_{1}I_{2}}}=|\gamma|e^{j\phi}$$

Phase fringes Contour lines

3-D World



Ø

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 S_1

 $B_{\perp} = B\cos(\theta_1 - \alpha)$

 $\phi = \varDelta \phi_{1 \cdot 2} \approx \varDelta \phi_{topo} + \varDelta \phi_{fe}$

$$\Delta \phi_{topo} \propto rac{k_c B_{\perp}}{R_1 \sin(\theta_1)} \Delta h$$

$$k_c = \frac{4\pi f_c}{c} = \frac{4\pi}{\lambda_c}$$

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Complete InSAR phase-to-height processing chain













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Interferometric coherence γ





Arg(%)

DLR E-SAR L Band Pol-In SAR (1.5m x 3m) – Baseline 5m

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Interferometric coherence γ : decorrelation sources

 γ fixed by a set of external sources :

System :

- Thermal or system noise : SAR amplifiers, ADC, antennas ...
- Quantization noise
- Geometric decorrelation : Baseline, squint ...
- Azimuth : Doppler decorrelation ...
- Ambiguities ...
- Processing errors : coregistration, interpolation ...

Environment:

- Random media : Surface & Volumetric media e.g. forest ...
- Temporal variations : wind, flowing or plowing, building ...

 $\gamma = \gamma_{SNR} \cdot \gamma_{quant} \cdot \gamma_{amb} \cdot \gamma_{geo} \cdot \gamma_{az} \cdot \gamma_{proc} \cdot \gamma_{media} \cdot \gamma_{temp}$

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Interferometric coherence γ : decorrelation sources

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Environment : • Random media : Surface & Volumetric media e.g. forest • Temporal variations : wind, flowing or plowing, building ... $\gamma = \gamma_{SNR} \cdot \gamma_{quant} \cdot \gamma_{amb} \cdot \gamma_{geo} \cdot \gamma_{az} \cdot \gamma_{proc} \cdot \gamma_{media} \cdot \gamma_{temp}$ • 8th ADVANCED TRAINING COURSE ON LAND REMOTE SENSING





Interferometric coherence γ

$$\gamma = \frac{E\left(s_{1}s_{2}^{*}\right)}{\sqrt{E\left(s_{1}s_{1}^{*}\right)E\left(s_{2}s_{2}^{*}\right)}} = \frac{E\left(s_{1}s_{2}^{*}\right)}{\sqrt{\overline{I}_{1}\overline{I}_{2}}} \approx \frac{E\left(s_{1}s_{2}^{*}\right)}{\overline{I}} = |\gamma|e^{j\phi}$$

 $\overline{I}_1 \approx \overline{I}_2 \approx \overline{I}$

In-SAR signal formulation

$$s_{1}(x,r) = e^{-jkr_{\theta_{1}}} \int_{V} a_{c_{1}}(\vec{r}') e^{-j\vec{k}_{1}\cdot(\vec{r}'-\vec{r}_{\theta})} h(x-x',r-r') dv'$$

$$s_{2}(x,r) = e^{-jkr_{\theta_{2}}} \int_{V} a_{c_{2}}(\vec{r}') e^{-j\vec{k}_{2}\cdot(\vec{r}'-\vec{r}_{\theta})} h(x-x',r-r') dv'$$

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Interferometric coherence γ

$$\gamma = \frac{E\left(s_{1}s_{2}^{*}\right)}{\sqrt{E\left(s_{1}s_{1}^{*}\right)E\left(s_{2}s_{2}^{*}\right)}} = \frac{E\left(s_{1}s_{2}^{*}\right)}{\sqrt{\overline{I_{1}\overline{I_{2}}}}} \approx \frac{E\left(s_{1}s_{2}^{*}\right)}{\overline{I}} = |\gamma|e^{j\phi} \qquad \overline{I_{1}} \approx \overline{I_{2}} \approx \overline{I}$$

Volume complex reflectivity
$$a_{c_i}(\vec{r})$$

 $\overline{I_i}(x,r) = E(s_i(x,r)s_i^*(x,r)) = \int_V \sigma_{v_i}(\vec{r}') |h(x-x',r-r')|^2 dv'$
 $E(s_1(x,r)s_2^*(x,r)) = \int_V \sigma_{v_e}(\vec{r}')e^{-j(\vec{k}_1-\vec{k}_2)(\vec{r}'-\vec{r}_0)} |h(x-x',r-r')|^2 dv'$

With :

$$E\left(a_{c_1}(\vec{r})a_{c_2}^*(\vec{r'})\right) = \sigma_{v_a}(\vec{r})\,\delta(\vec{r}-\vec{r'})$$

 $E\left(a_{c_i}\left(\vec{r}\right)a_{c_i}^*\left(\vec{r}'\right)\right) = \sigma_{v_i}\left(\vec{r}\right)\delta\left(\vec{r}-\vec{r}'\right)$

Reflectivity density

Effective reflectivity density

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Interferometric coherence γ

$$\gamma = \frac{E\left(s_{1}s_{2}^{*}\right)}{\sqrt{E\left(s_{1}s_{1}^{*}\right)E\left(s_{2}s_{2}^{*}\right)}} = \frac{E\left(s_{1}s_{2}^{*}\right)}{\sqrt{\overline{I_{1}\overline{I_{2}}}}} \approx \frac{E\left(s_{1}s_{2}^{*}\right)}{\overline{I}} = |\gamma|e^{j\varphi}$$

 $\overline{I}_1 \approx \overline{I}_2 \approx \overline{I}$

$$\gamma = \frac{\int_{V} \sigma_{v_{e}}(\vec{r}') e^{-j(\vec{k}_{1}-\vec{k}_{2})(\vec{r}'-\vec{r}_{0})} |h(x-x',r-r')|^{2} dv'}{\int_{V} \sigma_{v}(\vec{r}') |h(x-x',r-r')|^{2} dv'}$$

With :

$$E\left(a_{c_i}(\vec{r})a_{c_i}^*(\vec{r}')\right) = \sigma_{v_i}(\vec{r})\,\delta(\vec{r}-\vec{r}')$$

Reflectivity density

$$E\left(a_{c_1}(\vec{r})a_{c_2}^*(\vec{r}')\right) = \sigma_{v_e}(\vec{r})\,\delta(\vec{r}-\vec{r}')$$

Effective reflectivity density

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In-SAR coherence decomposition sa

Interferometric coherence γ

$$v = \frac{\int_{V} \sigma_{v_{e}}(\vec{r}') e^{-j(\vec{k}_{1}-\vec{k}_{2})(\vec{r}'-\vec{r}_{0})} |h(x-x',r-r')|^{2} dv'}{\int_{V} \sigma_{v}(\vec{r}') |h(x-x',r-r')|^{2} dv'}$$

 $\gamma = \gamma_{temp} \cdot \gamma_{media} = \gamma_{temp} \cdot \left(\gamma_{x,y} \cdot \gamma_z \right)$

Surface / Volume

$$\gamma_{vol} = \gamma_z = \frac{\int_Z \sigma_{v_e}(\vec{r}') e^{jk_z(z-z_0)} dz'}{\int_Z \sigma_{v_e}(\vec{r}') dz'}$$

Volumetric / random media decorrelation

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Volume decorrelation

$$\gamma_z = \frac{\int \sigma_{v_e}(z) e^{jk_z(z-z_0)} dz'}{\int Z \sigma_{v_e}(z) dz'}$$

Decorrelation due to the vertical structure $\sigma_{v_e}(z) = A_{v_e}f(z)$



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- 2 significant and uncorrelated mechanisms : volume + underlying ground
- Iow density medium = No refraction

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Ground only

Ground layer $\sigma_{v_e}(z) = \sigma_g \delta(z - z_g)$

No volume

$$\gamma_z = \frac{\int \sigma_{v_e}(z) e^{jk_z(z-z_0)} dz'}{\int \int \sigma_{v_e}(z) dz'} = e^{jk_z z_g}$$

 $z_0 + z_g$ z_0

Z.

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 $\sigma_{v_o}(z)$

 σ_{g}
In-SAR coherence decomposition sa

7 + h

Random volume (RV) with null extinction





No underlying ground

Homogeneous medium

$$\gamma_{z} = \frac{\int_{z_{0}+z_{v_{0}}}^{z_{0}+u_{v}} \sigma_{v_{e}}(z) e^{jk_{z}(z-z_{0})} dz'}{\int_{z_{0}+z_{v_{0}}}^{z_{0}+h_{v}} \sigma_{v_{e}}(z) dz'} = e^{jk_{z}} \frac{(h_{v}+z_{v_{0}})}{2} sinc\left(\frac{k_{z}d}{2}\right)}$$

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Null extinction : $\sigma_{v_e}(z) = A_v$

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Random volume (RV) with non-null extinction



No underlying ground

Non-null extinction : κ_e

Homogeneous medium with elementary reflectivity density σ_v (

ctivity density
$$\sigma_{v_e}(z) = A_v e^{\frac{2\kappa_e}{\cos(\theta)}(z - (z_\theta + h_v))}$$

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Random volume (RV) with non-null extinction



$$\gamma_{z} = \frac{\int_{z_{0}+z_{v_{0}}}^{z_{0}+n_{v}} \sigma_{v_{e}}(z) e^{jk_{z}(z-z_{0})} dz'}{\int_{z_{0}+z_{v_{0}}}^{z_{0}+h_{v}} \sigma_{v_{e}}(z) dz'} = e^{jk_{z}z_{v_{0}}} \frac{p}{p_{1}} \left(\frac{e^{p_{1}d}-1}{e^{pd}-1}\right) \qquad p = \frac{2\kappa_{e}}{\cos(\theta)} \\ p_{1} = \frac{2\kappa_{e}}{\cos(\theta)} + jk_{z}$$

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In-SAR coherence decomposition Sa





Non-null extinction : κ_e Underlying ground : $I_g = \sigma_g e^{-\frac{2\kappa_e d}{\cos(\theta)}} \delta(z - (z_\theta + z_g))$

Homogeneous medium with elementary reflectivity density $\sigma_{vol}(z) = A_v e^{\frac{2\kappa_e}{\cos(\theta)}(z-(z_\theta+h_v))}$

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Zn

Random volume over ground (RVoG)

$$\sigma_{v_e}(z) = I_g + \sigma_{vol}(z)|_{z \in [z_0 + z_{v_0} \dots z_0 + h_v]} \longrightarrow \gamma_z = \frac{\int \sigma_{v_e}(z) e^{jk_z(z - z_0)} dz'}{\int \sigma_{v_e}(z) dz'}$$

$$\gamma_{z} = \frac{\int_{z_{\theta}+z_{v_{\theta}}}^{z_{\theta}+h_{v}} \sigma_{vol}(z) e^{jk_{z}(z-z_{\theta})} dz' + I_{g} e^{jk_{z}z_{g}}}{\int_{z_{\theta}+z_{v_{\theta}}}^{z_{\theta}+h_{v}} \sigma_{vol}(z) dz' + I_{g}}$$

 $\gamma_z = \frac{\gamma_{vol} + \frac{I_g}{I_v} e^{jk_z z_g}}{1 + \frac{I_g}{I_v}} = e^{jk_z z_g} \frac{\widetilde{\gamma}_{vol} + m}{1 + m}$

$$\int_{z_{\theta}+z_{v_{\theta}}}^{z_{\theta}+h_{v}} \sigma_{vol}(z) e^{jk_{z}(z-z_{\theta})} dz' + I_{g} e^{jk_{z}z_{g}}$$

Z

 $I_v + I_g$

With :

$$\gamma_{vol} = \frac{1}{I_{v}} \int_{z_{0}+z_{v_{0}}}^{z_{0}+h_{v}} \sigma_{vol}(z) e^{jk_{z}(z-z_{0})} dz'$$

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In-SAR coherence decomposition sa

Random volume over ground (RVoG)

$$\widetilde{\gamma}_{vol} = e^{-jk_z z_g} \gamma_{vol}$$

$$k_{z} = \frac{k_{c}B_{\perp}}{R_{1}\sin(\theta_{1})}$$
$$k_{c} = \frac{4\pi f_{c}}{c}$$

Ground to volume intensity ratio : $m = \frac{I_g}{I_m}$

Observables (2): γ_z **Unknowns (4)**: z_g , m, $\gamma_{vol}(2)$

1 In-SAR acquisition vs. Complex RVOG structure : under-determined problem

 $\gamma_z = e^{jk_z z_g} \frac{\widetilde{\gamma}_{vol} + m}{1 + m}$

 \rightarrow another source of diversity is needed : polarization ?

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PART - 2





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Radar Polarimetry





Radar Polarimetry (Polar : polarisation Metry: measure) is the science of acquiring, processing and analysing the polarization state of an electromagnetic field

Radar Polarimetry deals with the full vector nature of polarized electromagnetic waves

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Radar Polarimetry



The POLARISATION information Contained in the waves backscattered from a given medium is highly related to:

its geometrical structure reflectivity, shape and orientation

its geophysical properties such as humidity, roughness, ...

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Radar Polarimetry





Forest Vegetation



Agriculture



Snow and Ice

- Forest Height
- Forest Biomass
- Forest Structure
- Canopy Extinction
- Underlying Topography
- Soil Moisture Content
- Soil roughness
- Height of Vegetation Layer
- Extinction of Vegetation Layer
- Moisture of Vegetation Layer

- Forest Ecology
- Forest Management
- Ecosystem Change
- Carbon Cycle
- Farming Management
- Water Cycle
- Desretification

- Topography
- Penetration Depth / Density
- Snow Ice Layer
- Snow Ice Extinction
- Water Equivalent

- Ecosystem Change
- Water Cycle
- Water Management

- Geometric Properties
- Dielectric Properties

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Urban Monitoring

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Scattering Polarimetry





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Scattering Polarimetry





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Scattering Polarimetry



Sinclair Color Coding



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Sinclair Matrix



POLARIMETRIC DESCRIPTORS

 $\begin{bmatrix} S \end{bmatrix} = \begin{bmatrix} S_{HH} & S_{HV} \\ S_{VH} & S_{VV} \end{bmatrix}$

TRANSMITTER:H & VRECEIVERS:H & V

<u>k</u> Target Vector

[*T*] 3x3 COHERENCY Matrix

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Coherency Matrix



MONOSTATIC CASE

PAULI SCATTERING VECTOR \underline{k}

$$\underline{k} = \frac{1}{\sqrt{2}} \begin{bmatrix} S_{HH} + S_{VV} & S_{HH} - S_{VV} & 2S_{HV} \end{bmatrix}^{T}$$

COHERENCY MATRIX [7]

$$\begin{bmatrix} T \end{bmatrix} = \underline{k} \cdot \underline{k}^{*T} = \begin{bmatrix} 2A_0 & C - jD & H + jG \\ C + jD & B_0 + B & E + jF \\ H - jG & E - jF & B_0 - B \end{bmatrix}$$

HERMITIAN MATRIX - RANK 1

A0, B0+B, B0-B : HUYNEN TARGET GENERATORS

[T] is closer related to Physical and Geometrical Properties of the Scattering Process, and thus allows a better and direct physical interpretation

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Target Generators



PHYSICAL INTERPRETATION



$$T_{11} = 2A_0 = |S_{HH} + S_{VV}|^2$$

$$T_{33} = B_0 - B = 2|S_{HV}|^2$$

$$T_{22} = B_0 + B = |S_{HH} - S_{VV}|^2$$

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Target Generators





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Target Generators





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SAR

Polarimetry





(Pol-InSAR)

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$$\langle [T_6] \rangle = \left\langle \underline{k} \cdot \underline{k}^{T*} \right\rangle = \begin{bmatrix} \left\langle \underline{k}_1 \cdot \underline{k}_1^{T*} \right\rangle & \left\langle \underline{k}_1 \cdot \underline{k}_2^{T*} \right\rangle \\ \left\langle \underline{k}_2 \cdot \underline{k}_1^{T*} \right\rangle & \left\langle \underline{k}_2 \cdot \underline{k}_2^{T*} \right\rangle \end{bmatrix} = \begin{bmatrix} \left\langle [T_1] \right\rangle & \left\langle [\Omega_{12}] \right\rangle \\ \left\langle [\Omega_{12}]^{T*} \right\rangle & \left\langle [T_2] \right\rangle \end{bmatrix}$$

POLARIMETRIC INTERFEROMETRIC COHERENCY MATRIX (6x6)

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POLSAR IMAGES $I_1 = \underline{w}_1^{T^*} \cdot \underline{k}_1$ and $I_2 = \underline{w}_2^{T^*} \cdot \underline{k}_2$

With: $(\underline{w}_1, \underline{w}_2)$ complex Unitary Vectors

$$\gamma(\underline{w}_1, \underline{w}_2) = \frac{\langle I_1 I_2^* \rangle}{\sqrt{\langle I_1 I_1^* \rangle \langle I_2 I_2^* \rangle}} = \frac{\langle \underline{w}_1 [\Omega_{12}] \underline{w}_2^{T*} \rangle}{\sqrt{\langle \underline{w}_1 [T_1] \underline{w}_1^{T*} \rangle \langle \underline{w}_2 [T_2] \underline{w}_2^{T*} \rangle}}$$

COMPLEX POLARIMETRIC INTERFEROMETRIC COHERENCE











$$\gamma(\underline{w}_1, \underline{w}_2) = \frac{\langle I_1 I_2^* \rangle}{\sqrt{\langle I_1 I_1^* \rangle \langle I_2 I_2^* \rangle}} = \frac{\langle \underline{w}_1 [\Omega_{12}] \underline{w}_2^{T*} \rangle}{\sqrt{\langle \underline{w}_1 [T_1] \underline{w}_1^{T*} \rangle \langle \underline{w}_2 [T_2] \underline{w}_2^{T*} \rangle}}$$

COMPLEX POLARIMETRIC INTERFEROMETRIC COHERENCE

QUESTION: WHICH POLARISATION COMBINATION LEADS TO THE MAXIMUM POSSIBLE INTERFEROMETRIC COHERENCE ?

POLARIMETRIC INTERFEROMETRIC COHERENCE OPTIMISATION PROCEDURE

S.R CLOUDE - K. PAPATHANASSIOU (1999)

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POLARIMETRIC INTERFEROMETRIC COHERENCE OPTIMISATION PROCEDURE

S.R CLOUDE – K. PAPATHANASSIOU (1999)

$$\gamma(\underline{w}_1, \underline{w}_2) = \frac{\langle I_1 I_2^* \rangle}{\sqrt{\langle I_1 I_1^* \rangle \langle I_2 I_2^* \rangle}} = \frac{\langle \underline{w}_1 [\Omega_{12}] \underline{w}_2^{T*} \rangle}{\sqrt{\langle \underline{w}_1 [T_1] \underline{w}_1^{T*} \rangle \langle \underline{w}_2 [T_2] \underline{w}_2^{T*} \rangle}}$$

Optimum Coherence set (3x3 eigenvector problem) :

$$\left(\underline{w}_{opt_1}, \underline{w}_{opt_2}\right) = \underset{(\underline{w}_1, \underline{w}_2)}{\operatorname{arg\,max}} \left(\gamma(\underline{w}_1, \underline{w}_2) |^2 \right)$$

$$\frac{\partial |\gamma(\underline{w}_1, \underline{w}_2)|^2}{\partial \underline{w}_1} = \frac{\partial |\gamma(\underline{w}_1, \underline{w}_2)|^2}{\partial \underline{w}_2} = 0$$

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POLARIMETRIC INTERFEROMETRIC COHERENCE OPTIMISATION PROCEDURE

S.R CLOUDE – K. PAPATHANASSIOU (1999)

 $\frac{\partial |\gamma(\underline{w}_{1}, \underline{w}_{2})|^{2}}{\partial \underline{w}_{1}} = \frac{\partial |\gamma(\underline{w}_{1}, \underline{w}_{2})|^{2}}{\partial \underline{w}_{2}} = 0$ $[T_{1}]^{-1}[\Omega_{12}][T_{2}]^{-1}[\Omega_{12}]^{T^{*}}\underline{w}_{opt_{1}} = |\gamma_{opt}|^{2}\underline{w}_{opt_{1}}$ $[T_{2}]^{-1}[\Omega_{12}]^{T^{*}}[T_{1}]^{-1}[\Omega_{12}]\underline{w}_{opt_{2}} = |\gamma_{opt}|^{2}\underline{w}_{opt_{2}}$

3 Real Eigenvalues (Optimum Coherence Values) : $\gamma_{opt_1} \ge \gamma_{opt_2} \ge \gamma_{opt_3} \ge 0$

3 Pairs of Eigenvectors (Optimum Scattering Mechanisms) : $\left\{ \underbrace{w_{opt_{1-1}}, \underbrace{w_{opt_{2-1}}}_{}, \left\{ \underbrace{w_{opt_{1-2}}, \underbrace{w_{opt_{2-2}}}_{}, \left\{ \underbrace{w_{opt_{1-3}}, \underbrace{w_{opt_{2-3}}}_{}, \underbrace{w_{opt_{2-3}}}_{} \right\} \right\}$

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PHYSICAL INTERPRETATION OF POLARIMETRIC INTERFEROMETRIC COHERENCES OPTIMISATION ALGORITHM





IN A PERFECT WORLD

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YX1-X2

PHYSICAL INTERPRETATION OF POLARIMETRIC INTERFEROMETRIC COHERENCES OPTIMISATION ALGORITHM







2HV

HH+VV

HH-VV

IN A REAL WORLD

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PHYSICAL INTERPRETATION OF POLARIMETRIC INTERFEROMETRIC COHERENCES OPTIMISATION ALGORITHM





IN AN OPTIMISED WORLD

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YX1-X2









Importance: Key parameter for forest management worldwide

Height as a product itself

- Phase of stand development
- Spatial height distribution (risk assessment, diversity)
- Management with esthetical protection goals
- Change of topography by forests (water runoff, skidding, roads)

Height as an input parameter

- Wood volume / forest biomass
- Site Index (with age and species information)

Height and density determine the microclimate and ecological processes within the forest

Height is dependent on and therefore reflects the site conditions







Which Height does POLinSAR measure ? H100 (forest standard)

- Most characteristic height in a forest
- Formed by the crown of the trees exposed to the sun light
- Typically concentrate most of the forest biomass
- Simple to measure: it is the intuitive forest height, measurement of few representative tree heights
- Sufficient for a good estimation





Forest Stand Type II

Very heterogeneously structured stands Large height variations on short distance Typical for natural uneven-aged forest

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MODELING AND PARAMETER ESTIMATION



Modeling Establishment of scattering model [M] In Radar Scatterer

Radar Observables

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Scatterer Parameters

Parameter Estimation

Inversion of the scattering model [M]

 $\begin{vmatrix} Scatterer \\ Parameters \end{vmatrix} = [M]^{-1} \begin{vmatrix} 0 \\ 0 \end{vmatrix}$

Radar Observables

Requirements on [M]: 1. Correctness in Interpretation and prediction of the observables 2. Simplicity in terms of parameters in order to be determined

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=[M]







Simplifications : Only 2 significant mechanisms Low density medium ⇒ No refraction

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In-SAR data statistics



Interferometric coherence γ : decorrelation sources

 γ fixed by a set of external sources :

System :

- Thermal or system noise : SAR amplifiers, ADC, antennas ...
- Quantization noise
- Geometric decorrelation : Baseline, squint ...
- Azimuth : Doppler decorrelation ...
- Ambiguities ...
- Processing errors : coregistration, interpolation ...

Environment:

- Random media : Surface & Volumetric media e.g. forest ...
- Temporal variations : wind, flowing or plowing, building ...

 $\gamma = \gamma_{SNR} \cdot \gamma_{quant} \cdot \gamma_{amb} \cdot \gamma_{geo} \cdot \gamma_{az} \cdot \gamma_{proc} \cdot \gamma_{surf} \cdot \gamma_{vol} \cdot \gamma_{temp}$

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2 Layer Combined Surface and random Volume Scattering

$$\gamma_{z}(\underline{w}) = e^{j\phi_{0}} \frac{\widetilde{\gamma}_{vol} + m(\underline{w})}{1 + m(\underline{w})}$$

 $m(\underline{w}) = \frac{\text{Surface Scattering Contribution}}{\text{Volume Scattering Contribution}}$

G / V ratio

B. Treuhaft (2000), S.R. Cloude (2003)

POLARIZATION DEPENDENT

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\underline{w}_{ν} Polarisation Channel corresponding to Volume Scattering

$$\gamma_z(\underline{w}_v) \underset{m \mapsto 0}{\longmapsto} = e^{j\phi_0} \widetilde{\gamma}_{vol}$$

2HV



 \underline{w}_s Polarisation Channel corresponding to Surface Scattering

$$\gamma_{z}(\underline{w}_{s}) = e^{j\phi_{0}} \frac{\widetilde{\gamma}_{vol} + m(\underline{w})}{1 + m(\underline{w})} \underset{m \mapsto \infty}{\longmapsto} e^{j\phi_{0}}$$

HH-VV



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DEM Differencing Algorithm

$$\begin{array}{l} \gamma_{z}(\underline{w}_{v}) = e^{j\phi_{0}}\widetilde{\gamma}_{vol} \\ \gamma_{z}(\underline{w}_{s}) \mapsto e^{j\phi_{0}} \end{array} \right\} \quad \mapsto \quad \gamma_{z}(\underline{w}_{v}) = \gamma_{z}(\underline{w}_{s})\widetilde{\gamma}_{vol} \approx \gamma_{z}(\underline{w}_{s})\alpha \ e^{jk_{z}h_{v}} \end{array}$$

$h_{v} \approx \frac{\arg[\gamma_{z}(\underline{w}_{v})] - \arg[\gamma_{z}(\underline{w}_{s})]}{k_{z}}$

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Coherence Amplitude Inversion Procedure

Assumption : Only volume scattering is present

$$\gamma_z(\underline{w}_v) = e^{j\phi_0} \widetilde{\gamma}_{vol} \quad \mapsto \quad |\gamma_z(\underline{w}_v)| = |\widetilde{\gamma}_{vol}|$$

$$\underset{d}{\operatorname{arg\,min}} \left| \left| \gamma_{z}(\underline{w}_{v}) \right| - \left| \frac{p}{p_{1}} \left(\frac{e^{p_{1}d} - 1}{e^{pd} - 1} \right) \right| \right|$$

1-D search procedure with Look-Up-Table (LUT)

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Topographic Phase Estimation

$$\begin{array}{l} \gamma_{z}(\underline{w}_{v}) = e^{j\phi_{0}} \widetilde{\gamma}_{vol} \\ \gamma_{z}(\underline{w}_{s}) = e^{j\phi_{0}} \frac{\widetilde{\gamma}_{vol} + m(\underline{w})}{1 + m(\underline{w})} \end{array} \right\} \quad \mapsto \quad e^{j\phi_{0}} = \frac{\gamma_{z}(\underline{w}_{s}) - \gamma_{z}(\underline{w}_{v})(1 - L)}{L} \\ \\ m(w_{s}) \end{array}$$

With:
$$L = \frac{m(\underline{w}_s)}{1 + m(\underline{w}_s)}$$

$$\hat{\phi}_0 = \arg[\gamma_z(\underline{w}_s) - \gamma_z(\underline{w}_v)(1-L)]$$

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Estimation of *L* :

$$e^{j\phi_0} = \frac{\gamma_z(\underline{w}_s) - \gamma_z(\underline{w}_v)(1-L)}{L}$$

$$\left|\frac{\gamma_z(\underline{w}_s) - \gamma_z(\underline{w}_v)(1-L)}{L}\right|^2 = 1 \quad \Rightarrow \quad AL^2 + BL + C = \theta$$

With :
$$A = |\gamma_z(\underline{w}_v)|^2 - 1$$
 $B = 2\Re\left\{ (\gamma_z(\underline{w}_s) - \gamma_z(\underline{w}_v)) \gamma_z^*(\underline{w}_s) \right\}$
 $C = |\gamma_z(\underline{w}_s) - \gamma_z(\underline{w}_v)|^2$
 $L = \frac{-B - \sqrt{B^2 - 4AC}}{2A}$

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esa





Topographic Phase Estimation

$$A = |\gamma_{z}(\underline{w}_{v})|^{2} - 1$$

$$\{\gamma_{z}(\underline{w}_{s}), \gamma_{z}(\underline{w}_{v})\} \implies B = 2\Re\{(\gamma_{z}(\underline{w}_{s}) - \gamma_{z}(\underline{w}_{v}))\gamma_{z}^{*}(\underline{w}_{s})\}$$

$$C = |\gamma_{z}(\underline{w}_{s}) - \gamma_{z}(\underline{w}_{v})|^{2}$$

$$L = \frac{-B - \sqrt{B^2 - 4AC}}{2A}$$

$$\hat{\phi}_0 = \arg[\gamma_z(\underline{w}_s) - \gamma_z(\underline{w}_v)(1-L)]$$

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Random Vegetation Over Ground (RVoG) inversion procedure

$$\underset{d,\kappa_e}{\operatorname{arg\,min}} \left| \gamma_z(\underline{w}_v) - e^{j\hat{\phi}_0} \frac{p}{p_1} \left(\frac{e^{p_1 d} - 1}{e^{p d} - 1} \right) \right|$$

Expensive 2-D search procedure



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PolSARpro - practicals





	oherence Estima	tion	X
- Input Master D	irectory		
C:/My_Data_Di	rectory/Pol-InSAR	_PolSARproSIM_forest/master	
Input Slave Di	rectory		
C:/My_Data_Di	rectory/Pol-InSAR	_PolSARproSIM_forest/slave_FEF	
- Output Master	Slave Directory		
C;/My_Data_Di	rectory/Pol-InSAR	_PolSARproSIM_forest/master_sla	ve_FER / 🔤
Init Row	1 End	Row 301 Init Col	1 End Col 301
- Complex Cohe	rences		
- Linear-	- Circular -	Pauli	Optimal
Г⊽ НН	R IL	₩ HH + W F HV + VH	🔽 SVD. 🥅 L. MinMax
I HV	🔽 LR	₩ HH W T HH.W*	T PD T L Diff
ΓW	⊢ BB		IT NR
- Numerical R	adius	Loci MinMax	- Loci Diff
Theta1	Theta3	Num Points	Num Points
Box Car Wir	ndow	I ВМР	
Row 7	Col 7	Averaging Ro	aw Cal
Bun	1	Hist	Exit
- i dan			

Complex Coherence Estimation

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PolSARpro - practicals





🖉 Height Estimation from Inversion Procedures				
Input Master - Slave Directory				
C./My_Data_Directory/Pol-InSAR_PolSARproSIM_forest/master_stave_FER				
Dutput Master - Slave Directory	_			
C:/My_Data_Directory/PoHnSAR_PoISARproSIM_forest/master_slave_FER /				
Init Row 1 End Row 301 Init Col 1 End Col	301			
Update List				
Polarimetric Phase Centre Height Estimation Polarimetric Channel HH				
I DEM Differencing Algorithm				
Coherence Amplitude Inversion Procedure				
Ground Phase Estimation and RVOG Inversion Procedure Median Window Size 121 Weighting Coherence Fraction Factor	0.4			
Top Phase Centre HV Ground Phase Centre HH - VV	J			
2D Kz File				
C:/My_Data_Directory/PoHnSAR_PoISARproSIM_forest/slave/kz,bin				
Run Hist 2				

Height estimation Inversion procedures

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