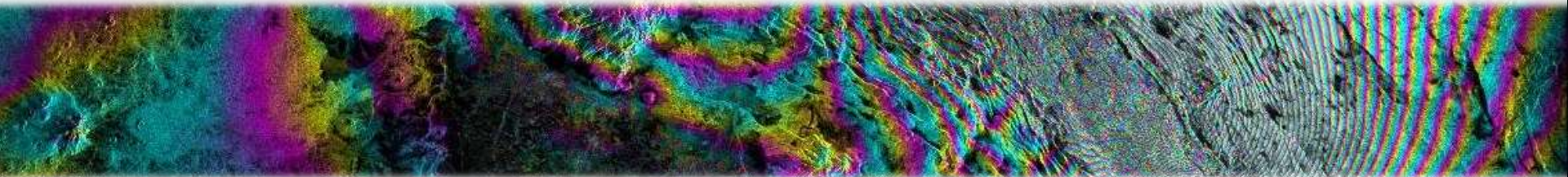


SYNTHETIC APERTURE RADAR FOR MAPPING OF FOREST DEGRADATION AND DEFORESTATION

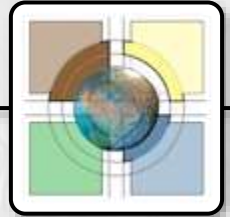
A ONE-WEEK TRAINING ON SAR

Lecturer: F.J. Meyer, Geophysical Institute, University of Alaska Fairbanks; fjmeyer@alaska.edu

Lecture 3: Concepts and Benefits of Polarimetric SAR



Outline:



POLARIMETRIC SIGNALS

POLARIMETRIC SAR ACQUISITION

INTERPRETATION OF POLARIMETRIC DATA

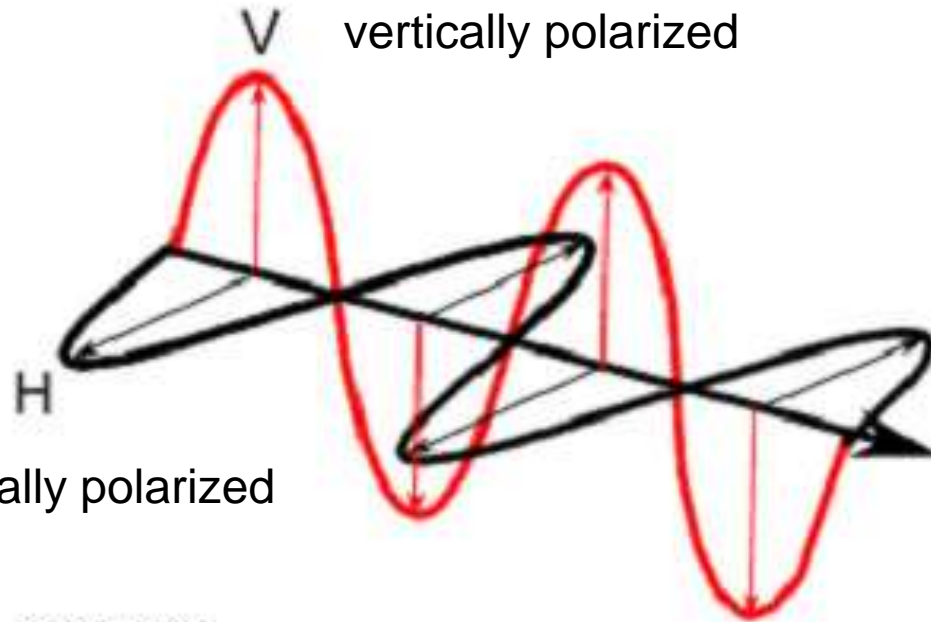
HIGHER-LEVEL POLARIMETRIC TECHNIQUES

EXAMPLES OF SAR IMAGE CLASSIFICATION

Polarization States of a Coherent Plane Wave

electric field vector

V vertically polarized



horizontally polarized

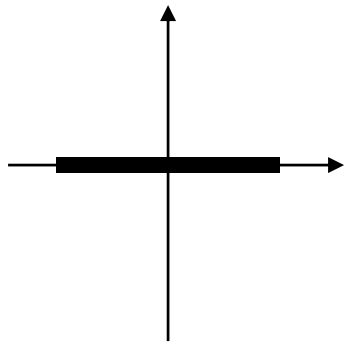
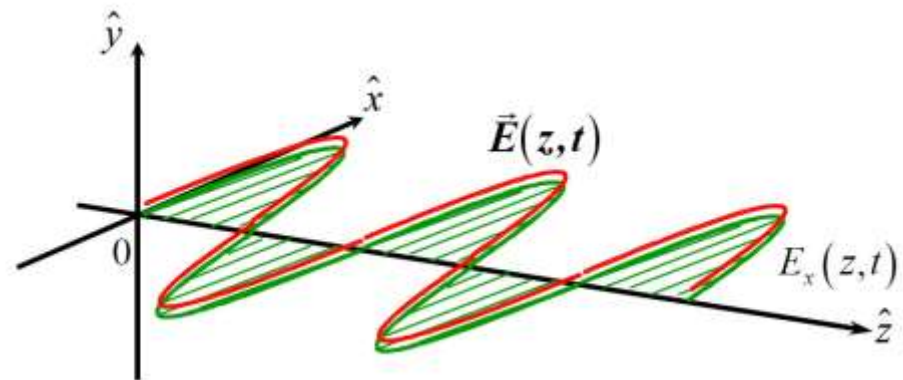
© CCRS / CCT



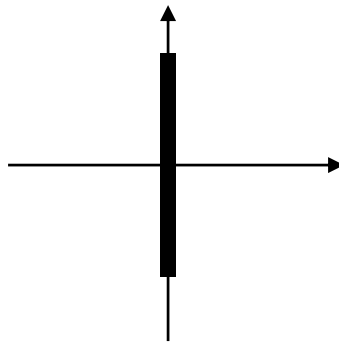
Linearly Polarized Signals

- Several stages of linear polarization possible

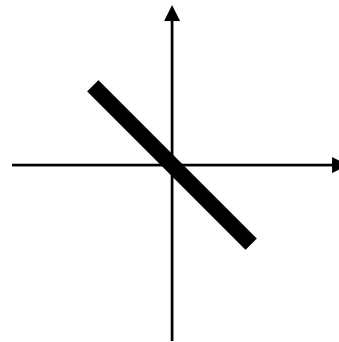
- Horizontal polarization (a)
- Vertical polarization (b)
- Linear -45° polarization (c)
- Linear $+45^\circ$ polarization (d)



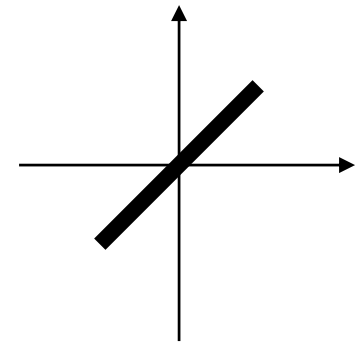
a)



b)



c)



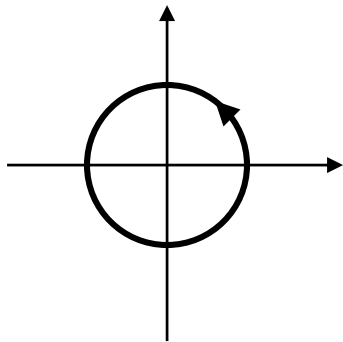
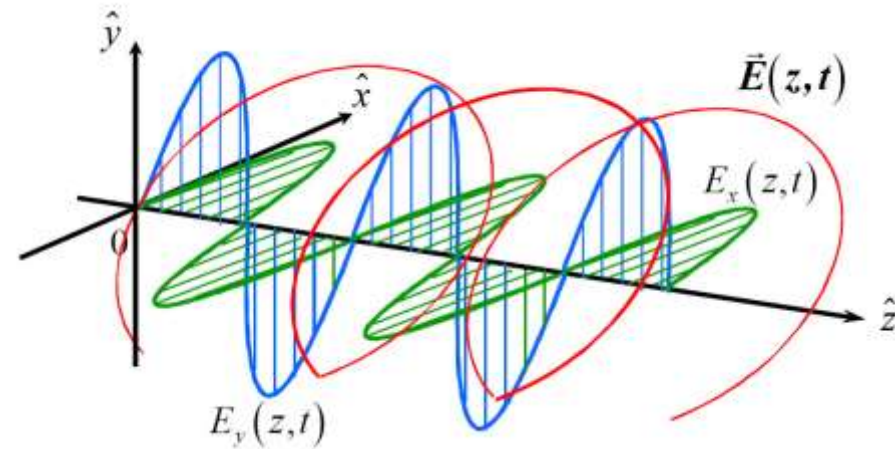
d)



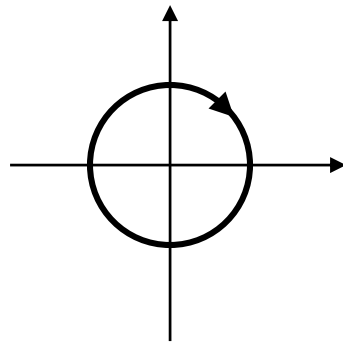
Circular and Elliptical Polarization

- **Combination of vertically and horizontally polarized signals**

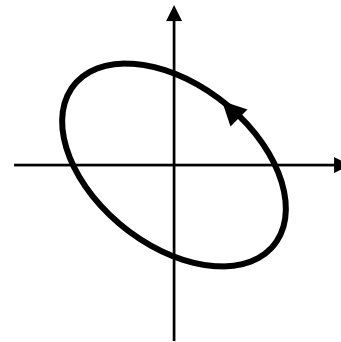
- Right circular polarization (a)
- Left circular polarization (b)
- Right elliptical polarization (c)
- Left elliptical polarization (d)



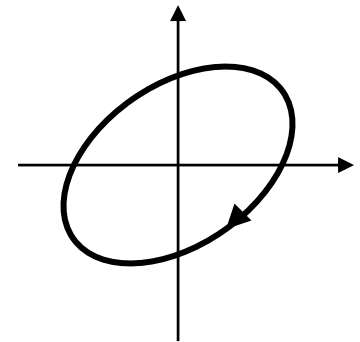
a)



b)



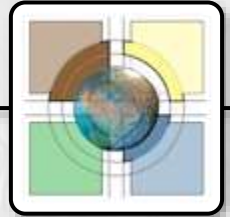
c)



d)

http://webphysics.davidson.edu/physlet_resources/dav_optics/examples/polarization.html

Outline:



POLARIMETRIC SIGNALS

POLARIMETRIC SAR ACQUISITION

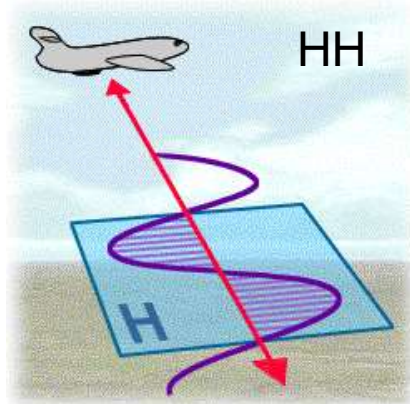
INTERPRETATION OF POLARIMETRIC DATA

HIGHER-LEVEL POLARIMETRIC TECHNIQUES

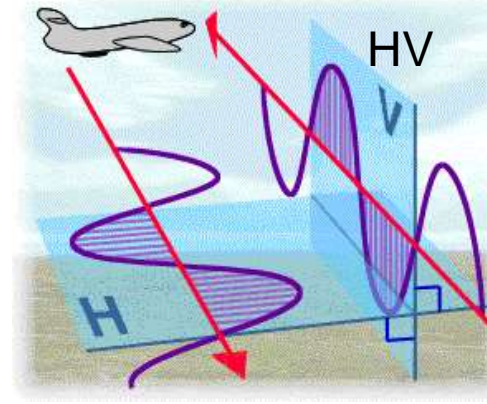
EXAMPLES OF SAR IMAGE CLASSIFICATION

Polarization of SARs

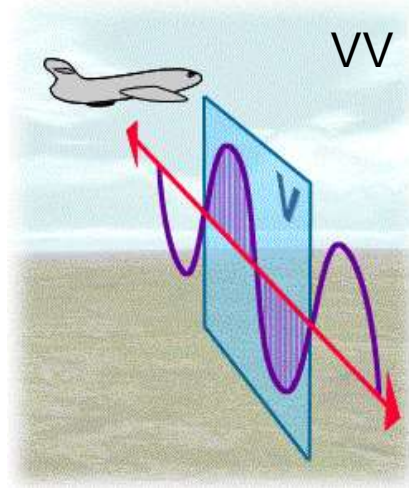
Horizontal Transmit
Horizontal Receive



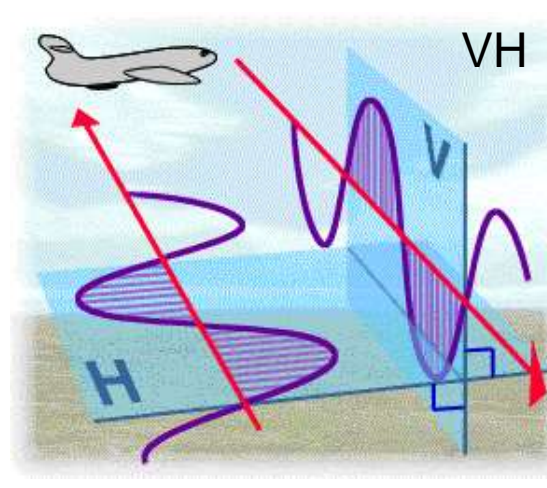
Horizontal Transmit
Vertical Receive



Vertical Transmit
Vertical Receive



Vertical Transmit
Horizontal Receive



Polarimetric SAR System Configurations

- **single pol:**
 - VV or HH (or possibly HV or VH)
- **dual pol:**
 - HH and HV, VV and VH, or HH and VV
- **quad pol (fully polarimetric):**
 - HH, VV, HV, and VH

relative phase between channels is important information



In a quad-pol SAR, every pixel is represented by a matrix of four complex numbers, representing ratios of received and transmitted electric-field amplitudes:

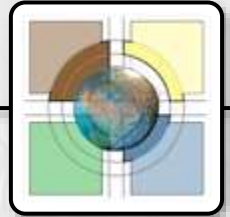
$$[S] = \begin{pmatrix} S_{HH} & S_{HV} \\ S_{VH} & S_{VV} \end{pmatrix} \quad (\text{scattering or Jones matrix})$$

For (monostatic) SARs: $S_{VH} = S_{HV}$ (reciprocity)

$$[S] = \begin{pmatrix} S_{HH} & S_{HV} \\ S_{HV} & S_{VV} \end{pmatrix} \quad (\text{Sinclair matrix})$$

⇒ 3 independent amplitudes + 2 independent phases per pixel

Outline:



POLARIMETRIC SIGNALS

POLARIMETRIC SAR ACQUISITION

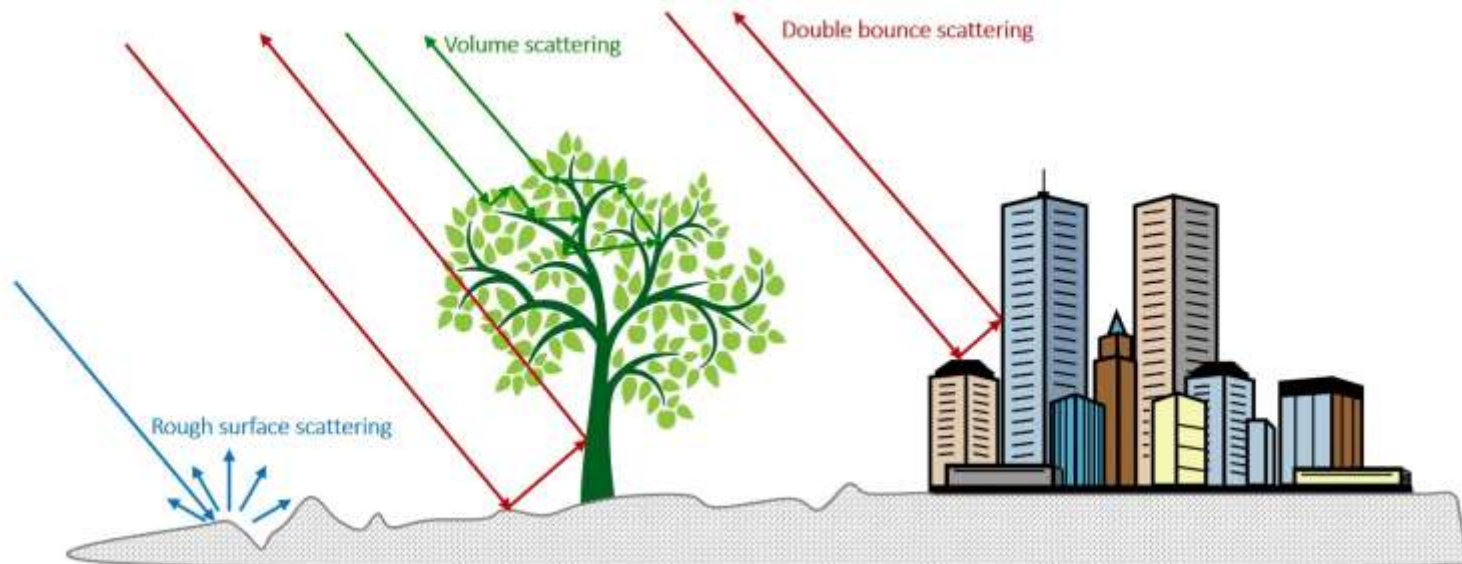
INTERPRETATION OF POLARIMETRIC DATA

HIGHER-LEVEL POLARIMETRIC TECHNIQUES

EXAMPLES OF SAR IMAGE CLASSIFICATION

What Does SAR See?

- At Radar wavelength, scattering is very physical and can be described as a series of bounces on scattering interfaces
- Three main scattering mechanisms dominate:
 - **Scattering on (rough) surfaces:** Water, bare soils, roads – Scattering strongly dependent on surface roughness and sensor wavelength (see Slide #36)
 - **Double-bounce scattering:** Buildings, tree trunks, light poles – little wavelength dependence
 - **Volume Scattering:** Vegetation; dry soils with high penetration – strongly dependent on sensor wavelength and dielectric properties of medium



Polarimetric Dependence of Scattering Principles

Relative scattering strength by polarization:

- **Pure Surface Scattering:** $|S_{VV}| > |S_{HH}| > |S_{HV}| \text{ or } |S_{VH}|$
- **Double Bounce Scattering:** $|S_{HH}| > |S_{VV}| > |S_{HV}| \text{ or } |S_{VH}|$
- **Volume Scattering:** main source of $|S_{HV}| \text{ and } |S_{VH}|$



Visual Interpretation

- Simplest method for classifying polarimetric imagery
- Present data in a color image to support interpretation
- Suggested “realistic looking” color assignment:
 - HH = red HV = green VV = blue
 - Water reflections have high VV
 - Vegetation has higher than average HV



Visual Interpretation – Color Composites



RGB Coding:

HH-VV

2HV

HH+VV

So-called
“Pauli”
decomposition

- **If all four polarimetric channels (HH, HV, VH, VV) available:**
 - RGB Composites can be created
 - Automatic image classification in classes “Rough Surface”, “Vegetation”, and “Urban Structures” can be created
- **Polarimetric Decompositions Mathematically Challenging:**
 - While we will skip most of polarimetric theory, I am providing more detailed information in the slides for your review at a later time
- **I will provide you with a very brief overview and a pointer to software that can be used for polarimetric processing**



Sinclair Decomposition

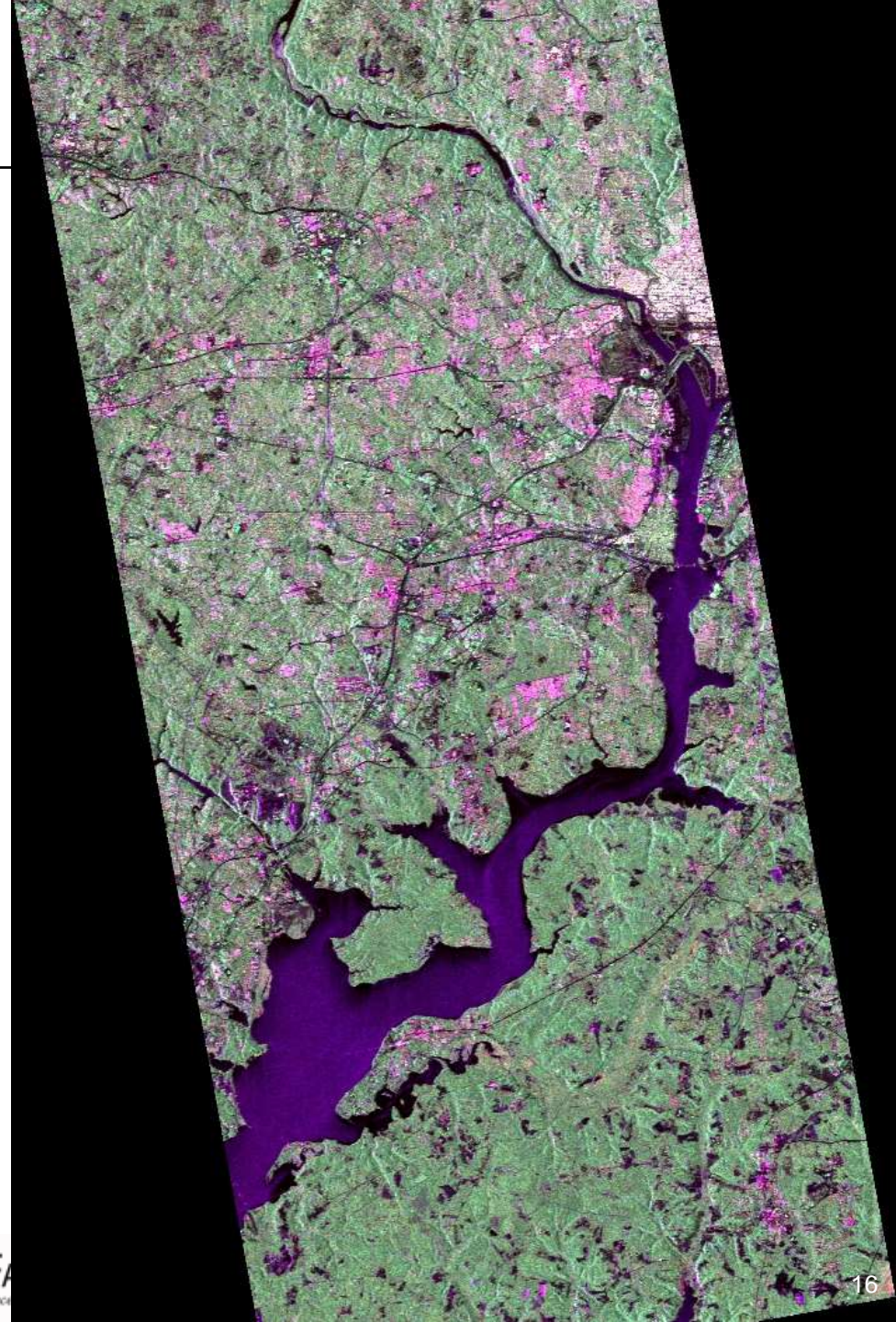
A Simple Method for RGB Creation

- Definition:

$$- \vec{s} = \begin{bmatrix} S_{HH} \\ S_{HV} + S_{VH} \\ S_{VV} \end{bmatrix} \begin{array}{l} \text{Double bounce dominated} \\ \text{Volume dominated} \\ \text{Surface dominated} \end{array}$$

- Example:

- ALOS PALSAR (L-band) data over Washington, D.C.



The Pauli Scattering Vector

The Simplest of all Polarimetric Decompositions

- In the monostatic case (where you can assume that $HV = VH$) the Pauli Vector reduces to three elements:

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

Surface scattering
Double bounce scattering
Volume scattering

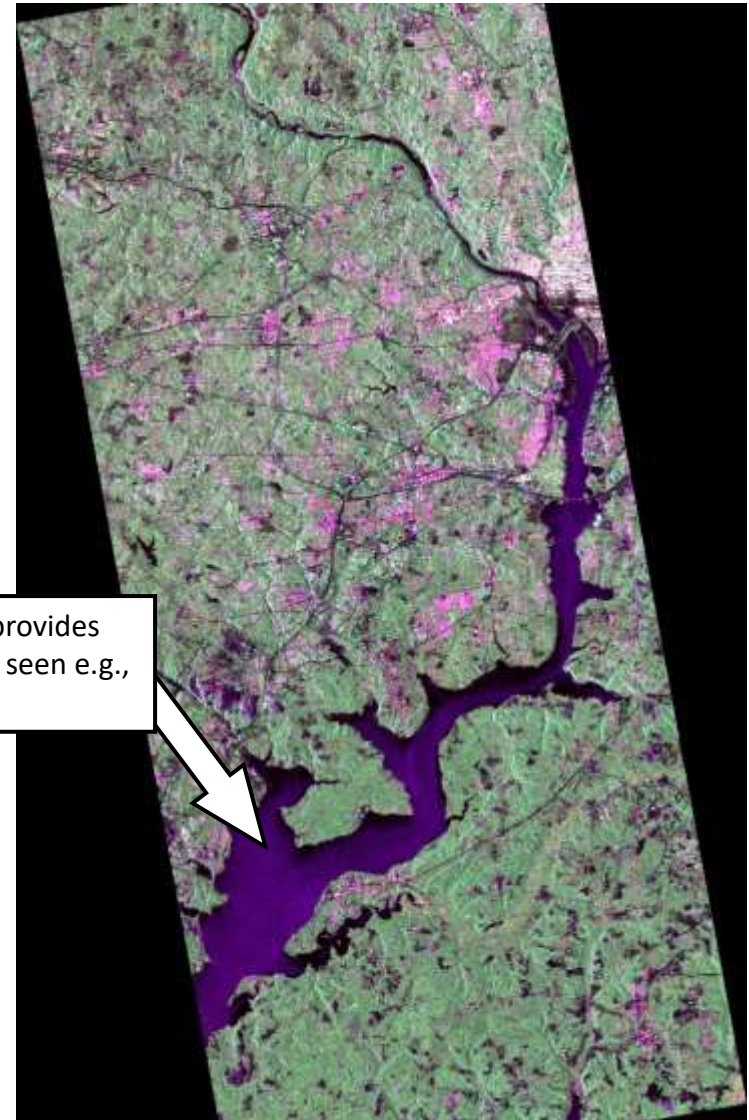
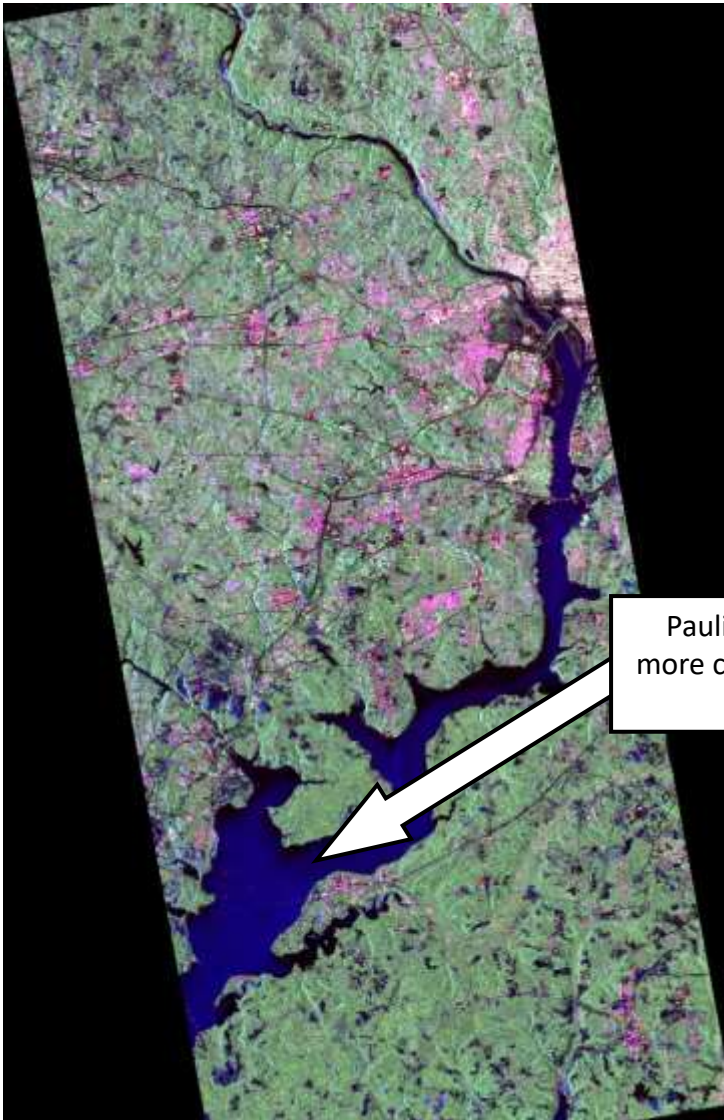
- It makes sense to represent these parameters in an RGB image with **red = double bounce**, **green = volume scattering**, and **blue = single bounce**

Pauli & Sinclair Decomposition

Washington, D.C., Example

Pauli decomposition ($HH-VV$, $HV+VH$, $HH+VV$)
(even bounce, volume, odd bounce)

Sinclair decomposition (HH , $HV+VH$, VV)
(preferred double bounce, volume, preferred surface)



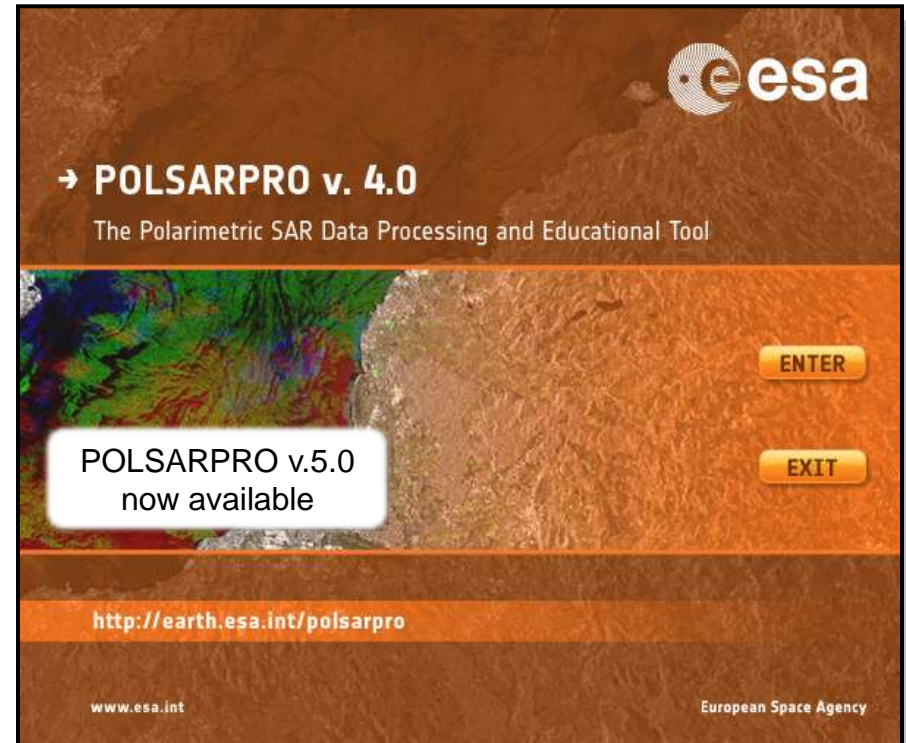
Pauli Decomposition provides
more correct "colors" as seen e.g.,
in water

The Pauli Vector – A Polarimetric Decomposition

- **Technically speaking, the Pauli Vector is a “Polarimetric Decomposition”:**
 - Take a complex matrix of correlated parameters
 - Decompose into sum of (orthogonal) component matrices, each of which has some simpler physical interpretation
- **Other Typical Decomposition Techniques include:**
 - Cholesky decomposition
 - Eigendecomposition
 - Singular Value Decomposition
 - Pauli Decomposition
 - ...
- **Goals:** Decompose complex correlated measurements into uncorrelated components that are easier to interpret

Available Software Tools

- PolSARPro is a freely available software tool provided by ESA
- Can be downloaded at:
<https://earth.esa.int/web/polsarpro/home>
- Includes all available polarimetric decomposition methods for dual-pol and quad-pol SAR data
- Output files of PolSARPro can be imported into MapReady for geocoding.
- In addition to the software also tutorials to Polarimetry can be downloaded



What To Do When Only Dual-Pol Data Are Available?

- **Full-polarimetric SAR data are rare and most modern SAR sensors only provide images in two polarizations**
 - E.g.: Sentinel-1 provides only the channels VV and VH
- The previously-shown RGB images cannot be created (at least one channel missing)
- **There are still some benefits of dual-pol SAR data:**
 - The VH channel contains information on volume scattering → good indicator for vegetation
 - Dual-pol data can still be used to create RGB images (see next slide)

What To Do When Only Dual-Pol Data Are Available?

- **Dual-pol color image has the following channels:**

- simple bounce (one dominant scattering signature; e.g., “pixel has only surface scattering) with no volume scattering
- Simple bounce (polarized) with some volume scattering
- volume (depolarized) scattering

Blue

Red

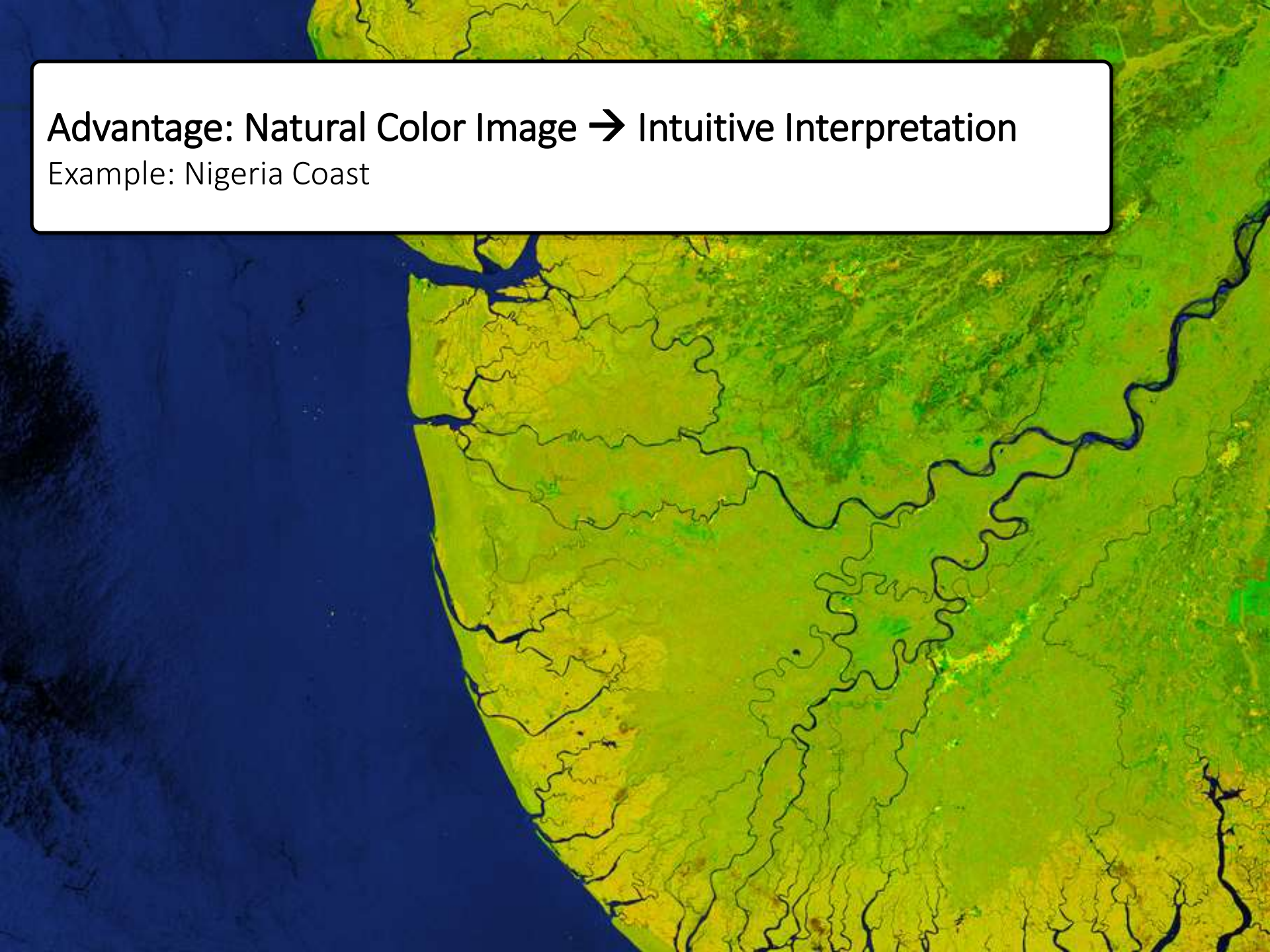
Green

Dual-pol RGB Image:
Coast of Nigeria (near
Warri)



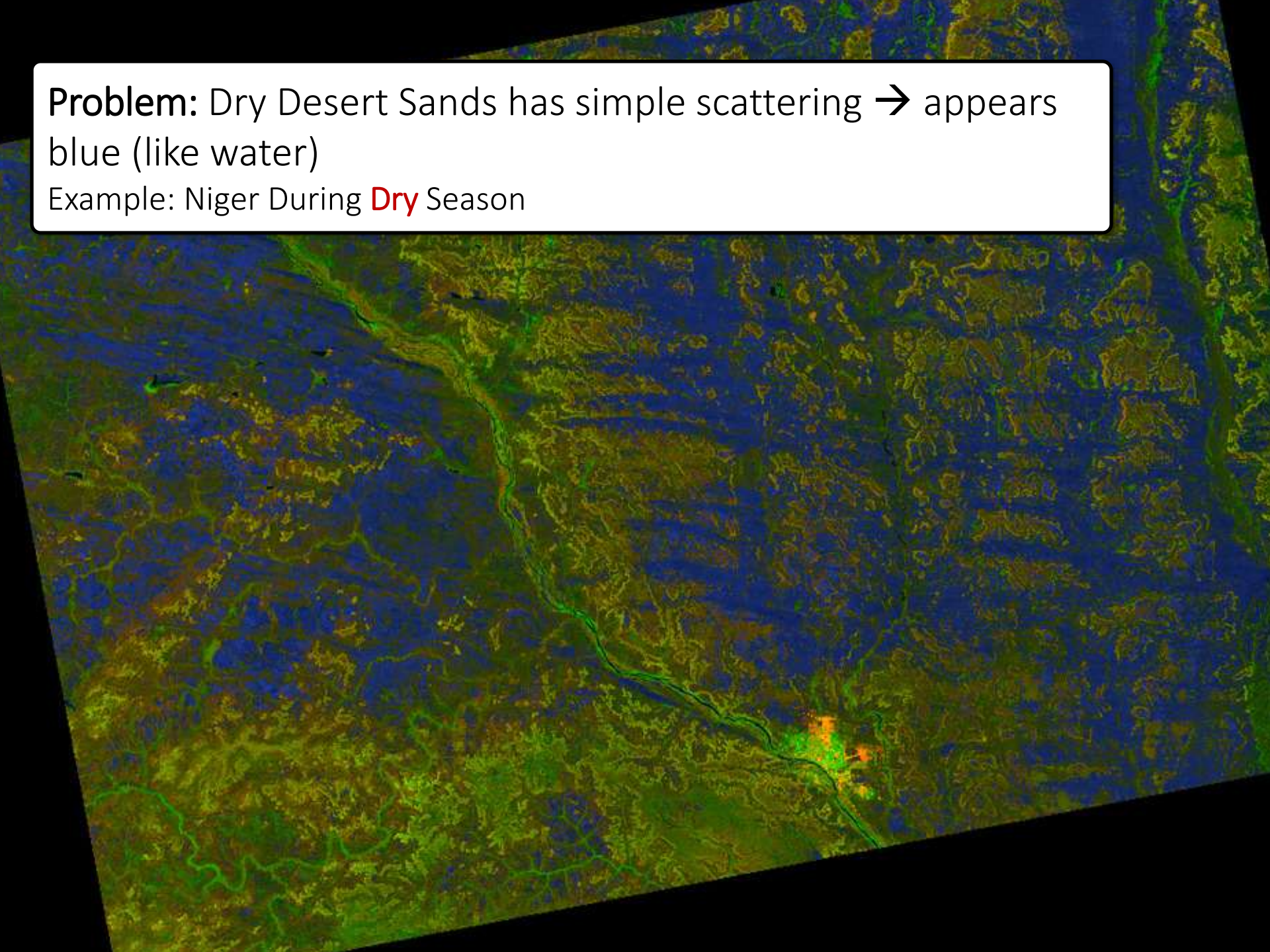
Advantage: Natural Color Image → Intuitive Interpretation

Example: Nigeria Coast



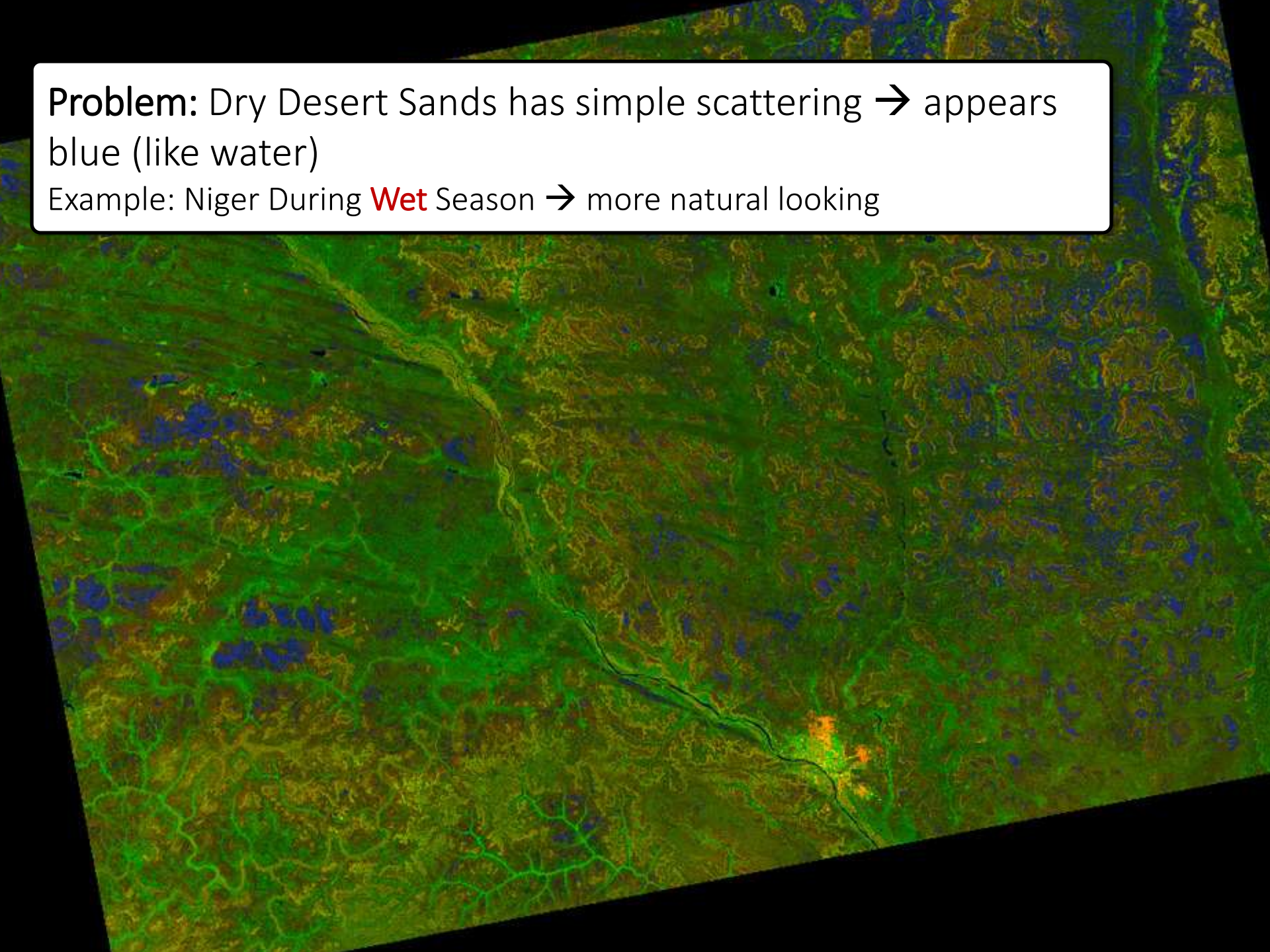
Problem: Dry Desert Sands has simple scattering → appears blue (like water)

Example: Niger During **Dry** Season

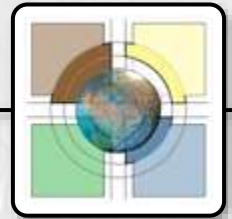


Problem: Dry Desert Sands has simple scattering → appears blue (like water)

Example: Niger During **Wet** Season → more natural looking



Outline:



POLARIMETRIC SIGNALS

POLARIMETRIC SAR ACQUISITION

INTERPRETATION OF POLARIMETRIC DATA

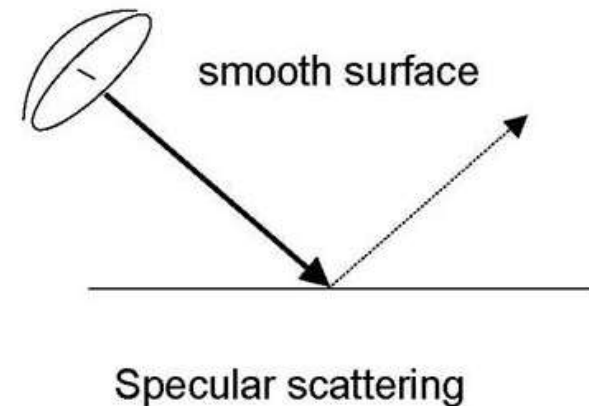
HIGHER-LEVEL POLARIMETRIC TECHNIQUES

EXAMPLES OF SAR IMAGE CLASSIFICATION

Scattering from Smooth Surfaces

Meaning of “Smooth”:

irregularities of surface \ll wavelength



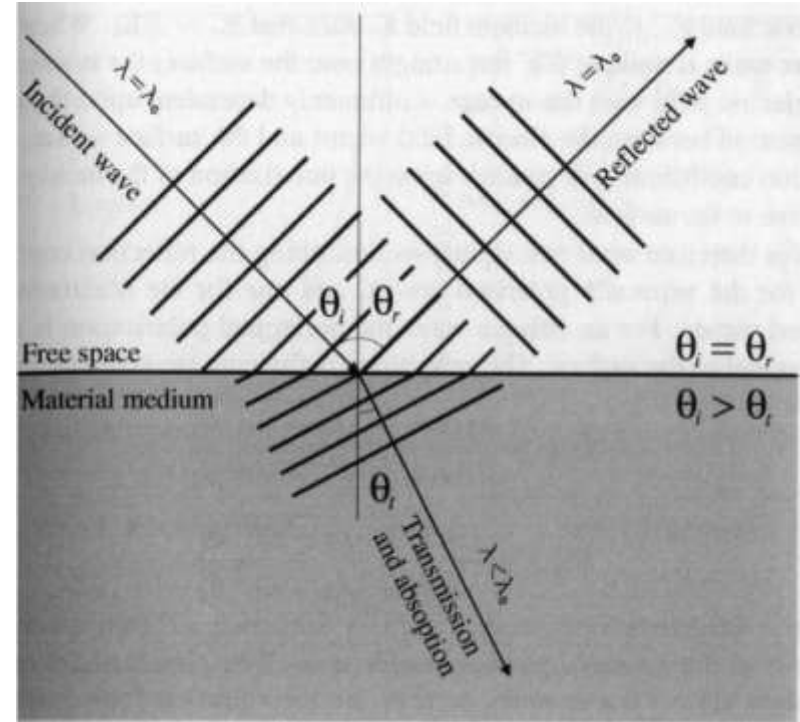
Scattering from Smooth Surfaces

- EM wave hitting interface at angle θ_i is partly **scattered** and partly **transmitted**
- For smooth surfaces, **scattering** forms coherent peak in specular direction
$$\theta_r = \theta_i$$
- **Part of signal enters medium** and is slowed down causing redirection (refraction) according to Snell's Law

$$\frac{\sin \theta_i}{\sin \theta_t} = \frac{n_2}{n_1}$$

- Percentage of reflected energy is given by **reflection coefficient** R relating incident electric field E_i to reflected electric field E_r

$$E_r = R \cdot E_i$$



From: Woodhouse (2006)

Quantifying the Surface Reflection Coefficient

- R depends on relative orientation between electric field vector and surface
- This means: R depends on the polarization of incident wave
- For linear polarization two equations are needed to describe R_{VV} (vertical to surface in transmit and receive) and R_{HH} (horizontal to surface)

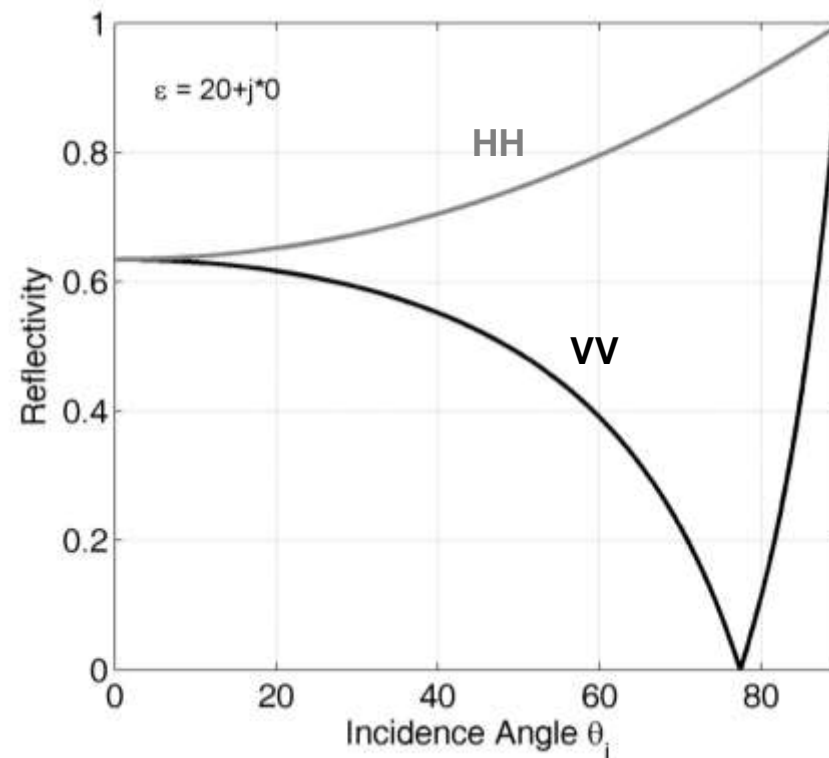
$$\begin{aligned} R_{VV} &= \frac{\varepsilon \cdot \cos \theta_i - \sqrt{\varepsilon - \sin^2 \theta_i}}{\varepsilon \cdot \cos \theta_i + \sqrt{\varepsilon - \sin^2 \theta_i}} \\ R_{HH} &= \frac{\cos \theta_i - \sqrt{\varepsilon - \sin^2 \theta_i}}{\cos \theta_i + \sqrt{\varepsilon - \sin^2 \theta_i}} \\ R_{HV} = R_{VH} &= 0 \end{aligned}$$

with ε being complex dielectric constant

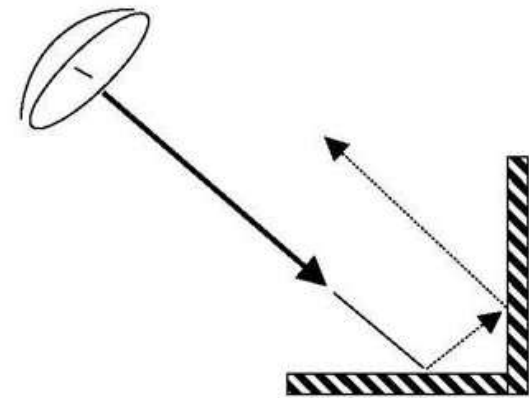
- **Fresnel reflection coefficients** R_{VV} and R_{HH} , defined between 0 and 1

Magnitude of Fresnel Reflection Coefficients

- One can see from both the Fresnel equations as well as the plot that $R_{VV} = R_{HH}$ if $\theta_i = 0^\circ$ and if $\theta_i = 90^\circ$
- R_{HH} is higher than $R_{VV} \rightarrow$ **surface backscatter stronger in horizontal polarization**



Scattering from Edges and Corners

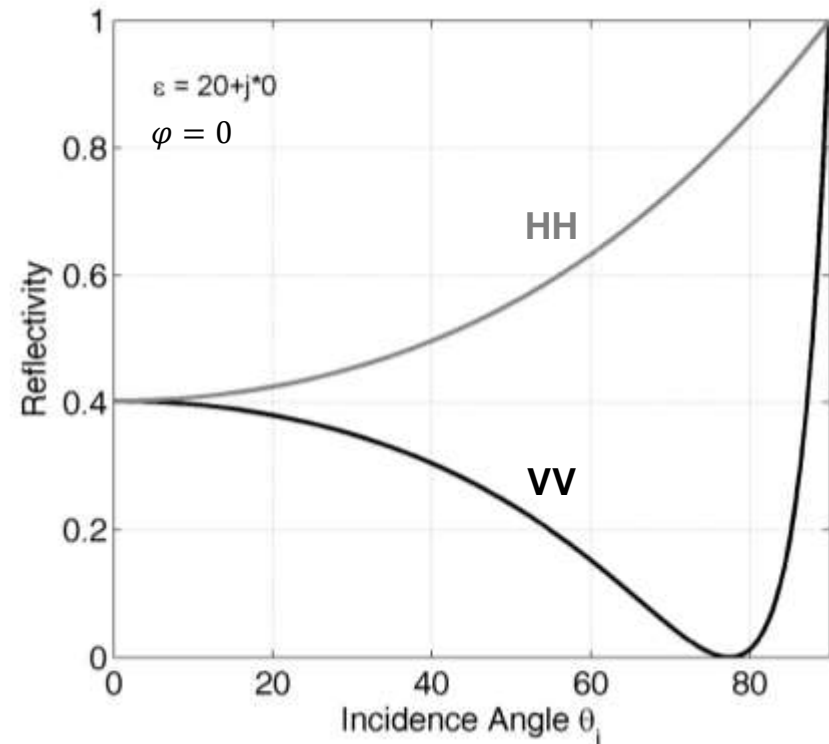


Edge and Corner Reflectors



Scattering on Edges and Corners

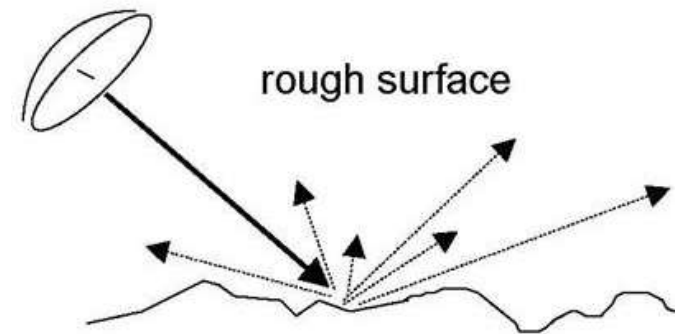
- If ε of both scattering surfaces is identical:
 - Scattering coefficient in HH: $R_{HH} \cdot R_{HH}$
 - Scattering coefficient in VV: $R_{VV} \cdot R_{VV} \exp(j\varphi)$ with R_{VV} from Slide 32 and φ being a potential material dependent phase shift relative to the HH signal
- Double-bounce scattering higher in horizontal (R_{HH}) than in vertical polarization (R_{VV})
- $R_{VH} = R_{HV} = 0$



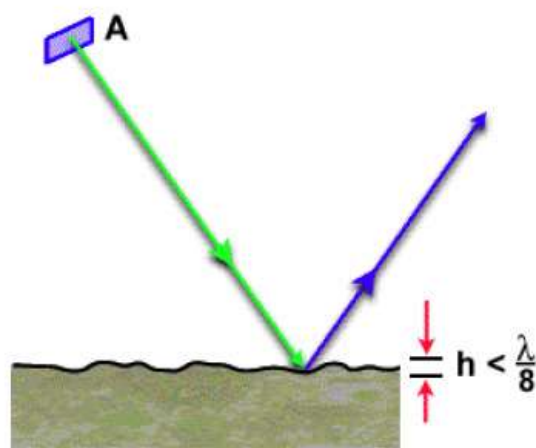
Scattering from Rough Surfaces

Meaning of “Rough”:

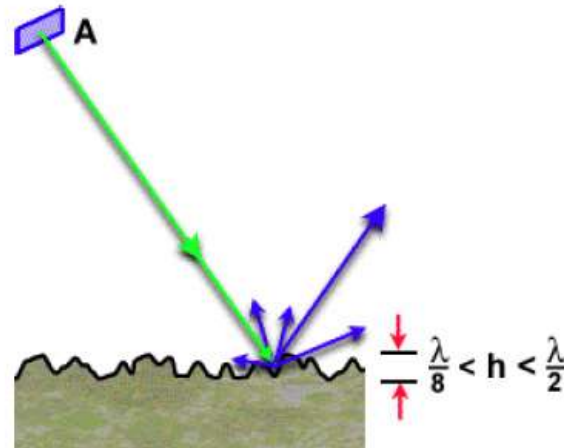
irregularities of surface in the order of larger than wavelength



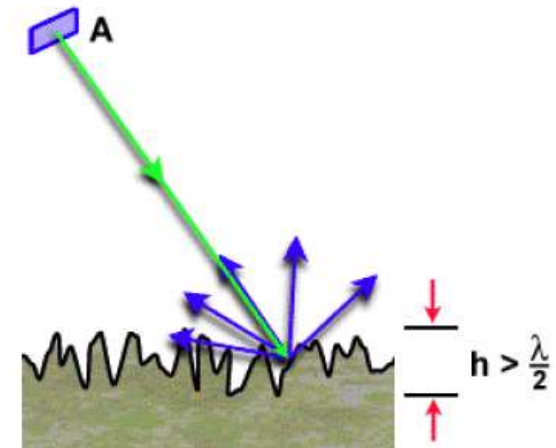
When is a Surface Rough? And how do Surfaces of Varying Roughness Scatter?



Smooth surface
Specular reflection
No return



Intermediate roughness
Moderate return



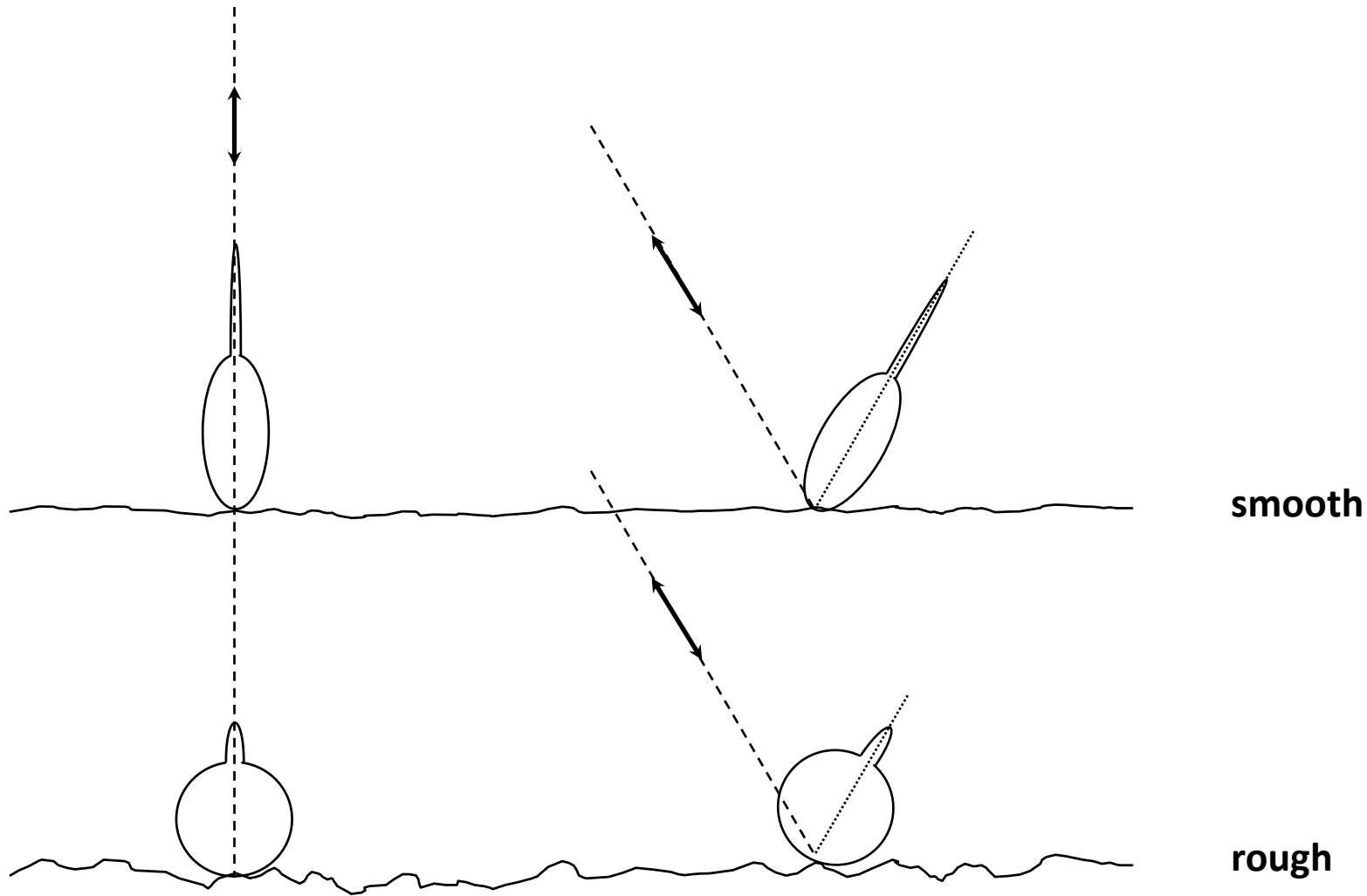
Rough surface
Diffuse scattering
Strong return

- Several criterion for roughness were developed that differ in strictness including the Fraunhofer criterion (a strict criterion) states:

$$h_{rough} > \frac{\lambda}{32 \cos \theta_i}$$



Schematic Analysis of Incidence Angle Dependence of Scattering



Wavelength and Surface Roughness

$$\lambda = 2\text{ cm} \dots 25\text{ cm}$$



rough



smooth

Scattering Processes: Bragg Scattering

- **Polarimetric Dependence of Bragg scattering:**

- **Horizontal polarization:**

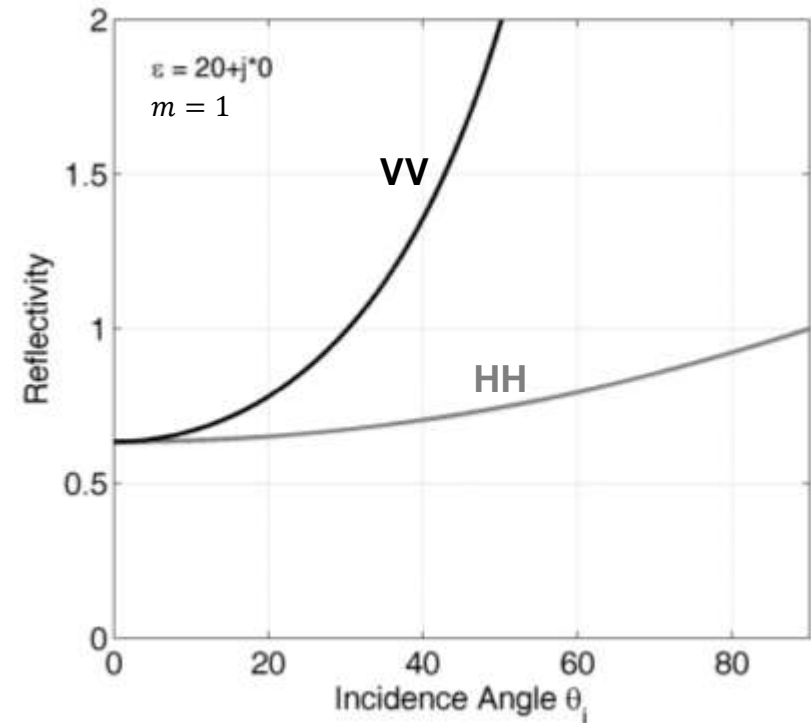
$$R_{HH} = m \cdot \frac{\cos\theta_i - \sqrt{\epsilon_r - \sin^2\theta_i}}{\cos\theta_i + \sqrt{\epsilon_r - \sin^2\theta_i}}$$

- **Vertical polarization:**

$$R_{VV} = m \cdot \frac{(\epsilon_r - 1)[\sin^2\theta_i - \epsilon_r(1 + \sin^2\theta_i)]}{(\epsilon_r \cdot \cos\theta_i + \sqrt{\epsilon_r - \sin^2\theta_i})^2}$$

- **Cross polarizations:**

$$R_{HV} = R_{VH} = 0$$



... where ϵ is the dielectric constant of the surface
and m depends on surface roughness

Polarimetric Dependence of Scattering Principles

Relative scattering strength by polarization:

- **Pure Surface Scattering:** $|S_{VV}| > |S_{HH}| > |S_{HV}| \text{ or } |S_{VH}|$
- **Double Bounce Scattering:** $|S_{HH}| > |S_{VV}| > |S_{HV}| \text{ or } |S_{VH}|$
- **Volume Scattering:** main source of $|S_{HV}|$ and $|S_{VH}|$



Sinclair Decomposition

- Definition:

$$- \vec{s} = \begin{bmatrix} S_{HH} \\ S_{HV} + S_{VH} \\ S_{VV} \end{bmatrix} \begin{array}{l} \text{Double bounce dominated} \\ \text{Volume dominated} \\ \text{Surface dominated} \end{array}$$

- Example:

- ALOS PALSAR (L-band) data over Washington, D.C.



Different Ways Of Representing Polarimetric Data

- Commonly used polarimetric target descriptors include:

- **[S]** Sinclair matrix
- **k** Pauli Scattering Vector

- **[K]** Kennaugh Matrix
- **[T]** Coherency Matrix
- **[C]** Covariance Matrix

Will not be discussed here

- Pauli Vector, Kennaugh matrix, and Coherency matrix are closely related to physical properties of the scatter → allow insight into surface structure
- Sinclair Matrix and Covariance matrix are directly related to the system measurables

The Pauli Scattering Vector

- The Scattering matrix can be decomposed by a set of “Pauli Matrices” Ψ_p resulting in the Pauli vector \underline{k}

$$\underline{k} = \frac{1}{2} \text{trace}([S][\Psi_p])$$

where

$$[\Psi_p] = \left\{ \sqrt{2} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \sqrt{2} \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \sqrt{2} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \sqrt{2} \begin{bmatrix} 0 & -j \\ j & 0 \end{bmatrix} \right\}$$

- Resulting in:

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

Surface scattering

Double bounce scattering

Volume scattering



Related to physical properties of the scatterers

The Pauli Scattering Vector

- In the monostatic case (where you can assume that $HV = VH$) the Pauli Vector reduces to three elements:

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

	Surface scattering
	Double bounce scattering
	Volume scattering

- It makes sense to represent these parameters in an RGB image with **red = double bounce**, **green = volume scattering**, and **blue = single bounce**

The Pauli Vector – A Polarimetric Decomposition

- **Technically speaking, the Pauli Vector is a “Polarimetric Decomposition”:**
 - Take a complex matrix of correlated parameters
 - Decompose into sum of (orthogonal) component matrices, each of which has some simpler physical interpretation
- **Typical Decomposition Techniques include:**
 - Cholesky decomposition
 - Eigendecomposition
 - Singular Value Decomposition
 - Pauli Decomposition
 - ...
- **Goals:** Decompose complex correlated measurements into uncorrelated components that are easier to interpret

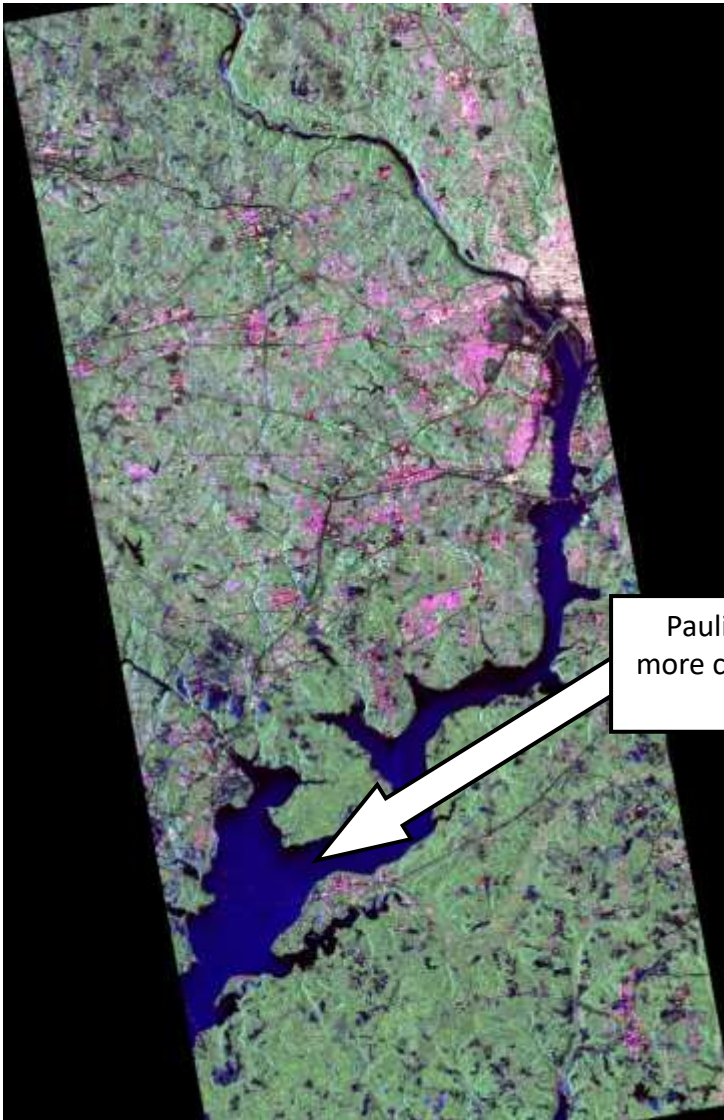


Pauli & Sinclair Decomposition

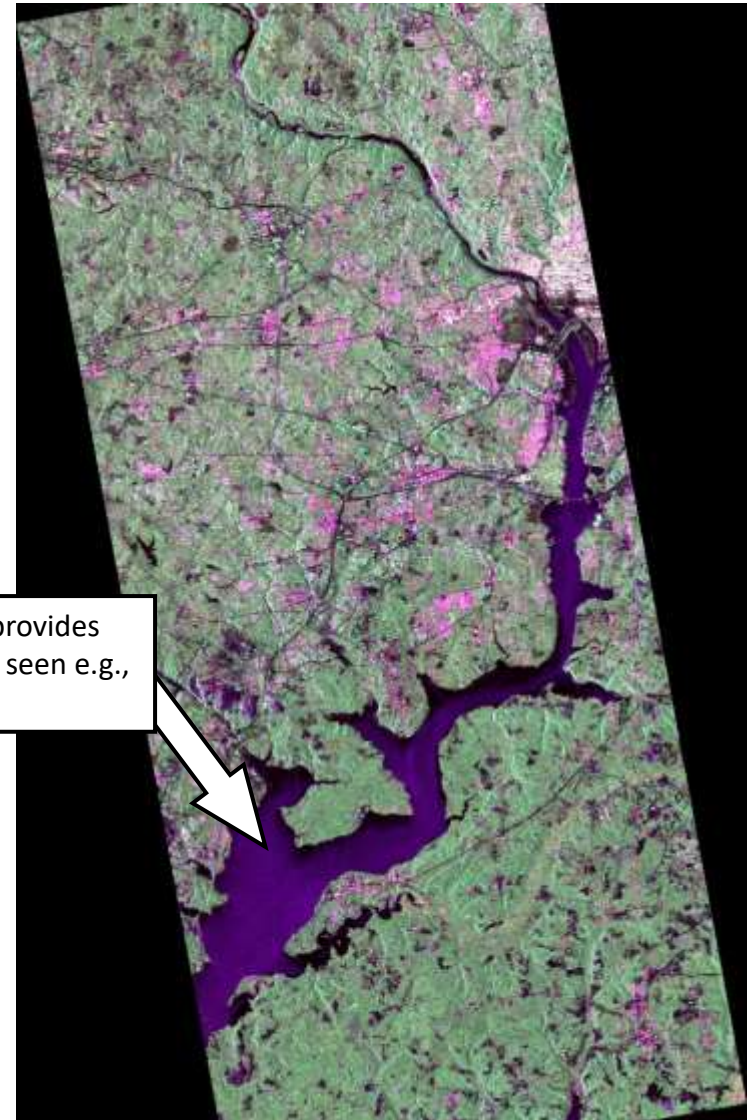
Washington, D.C., Example

Pauli decomposition ($HH-VV$, $HV+VH$, $HH+VV$)
(*even bounce*, *volume*, *odd bounce*)

Sinclair decomposition (HH , $HV+VH$, VV)
(*preferred double bounce*, *volume*, *preferred surface*)



Pauli Decomposition provides
more correct "colors" as seen e.g.,
in water



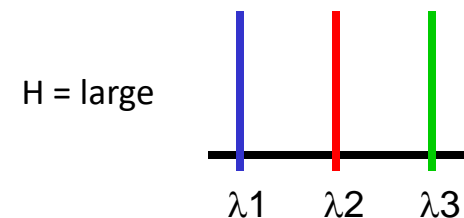
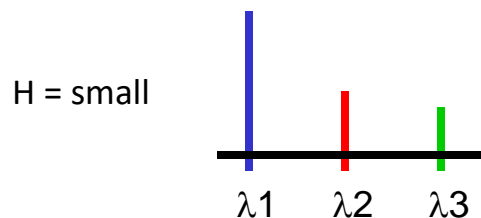
Scene Classification based on Polarimetric Decompositions

- **Unsupervised classifiers have been introduced in the recent years**
 - Based on physical properties of target and dominant scattering mechanism and NOT on dataset
 - No prior knowledge about scene content or terrain classes necessary
- **Main decomposition-based classifiers:**
 - Van Zyl: “odd bounce,” “even bounce,” “diffuse”
 - Freeman: Three component scattering model based classification
 - Cloude & Pottier: H / α Classifier
 - Cloude & Pottier: $H / A / \alpha$ Classifier
 - Yamaguchi: Four Component scattering model based classification (extension of Freeman)
- **Newest methods combine unsupervised methods (e.g., H / A) with a subsequent supervised classifier to improve results**



The Unsupervised H / α Classifier

- **Eigendecomposition** of the coherency matrix (i.e., reduction large number of parameters into an orthogonal set of parameters that contain meaningful information) → Physical interpretation of Eigenvalues
 - **Polarimetric Entropy (H)** [calculated from ratio of Eigenvalues]:
 - Represents *randomness* of scattering with $H = 0$ indicates single scattering mechanism and $H = 1$ represents random mix of mechanisms



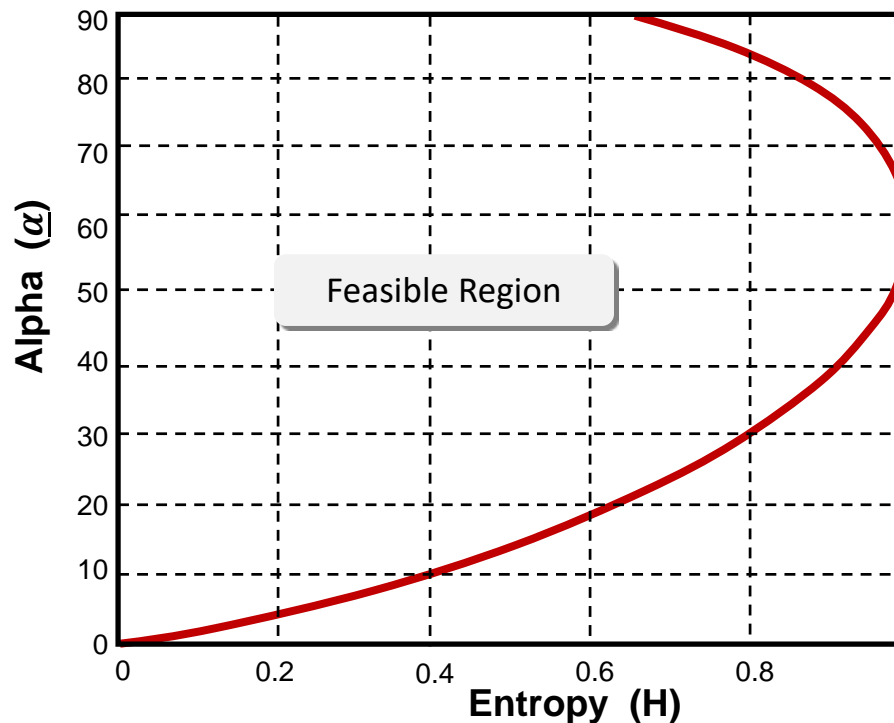
- **The (average) $\bar{\alpha}$ Angle** [calculated from Eigenvector]: Indicative of dominant scattering mechanism
 - If $\bar{\alpha} \approx 0^\circ$ → surface scattering
 - If $\bar{\alpha} \approx 45^\circ$ → volume scattering
 - If $\bar{\alpha} \approx 90^\circ$ → dihedral or multiple scattering

$$\bar{\alpha} = \sum_{i=1}^3 \alpha_i P_i$$

with $P_i = \frac{\lambda_i}{\lambda_1 + \lambda_2 + \lambda_3}$
and α_i = angle relative to vertical

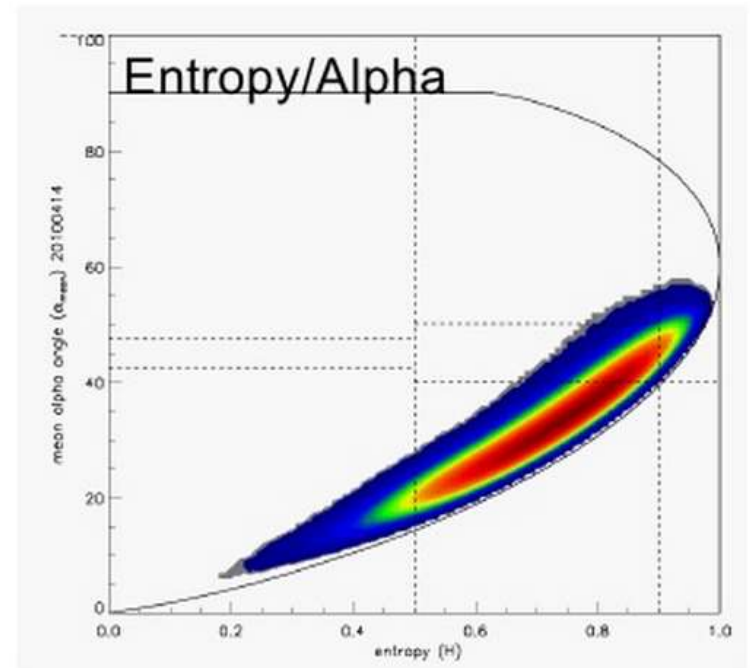
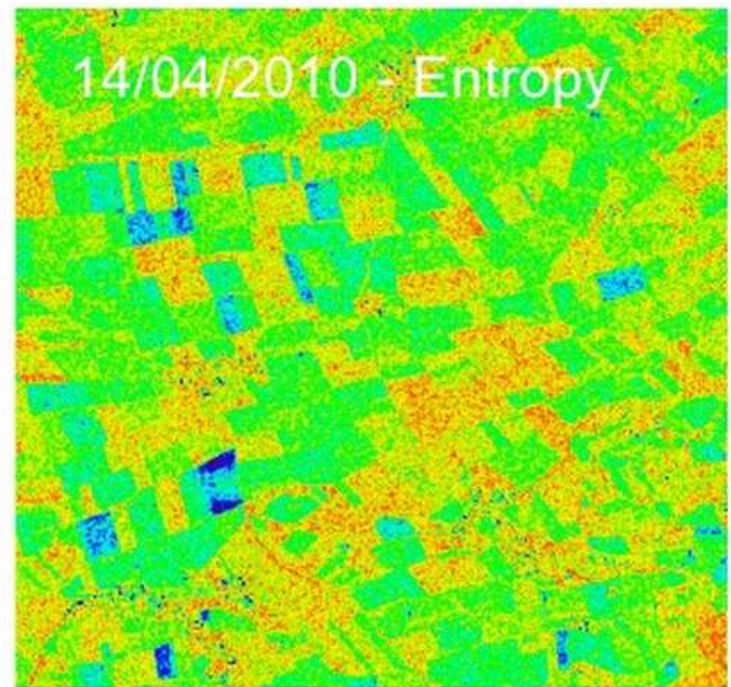
The Unsupervised H/α Classifier

- The H/α diagram is shown below
- By definition, all image pixels will have H/α values within the feasible region



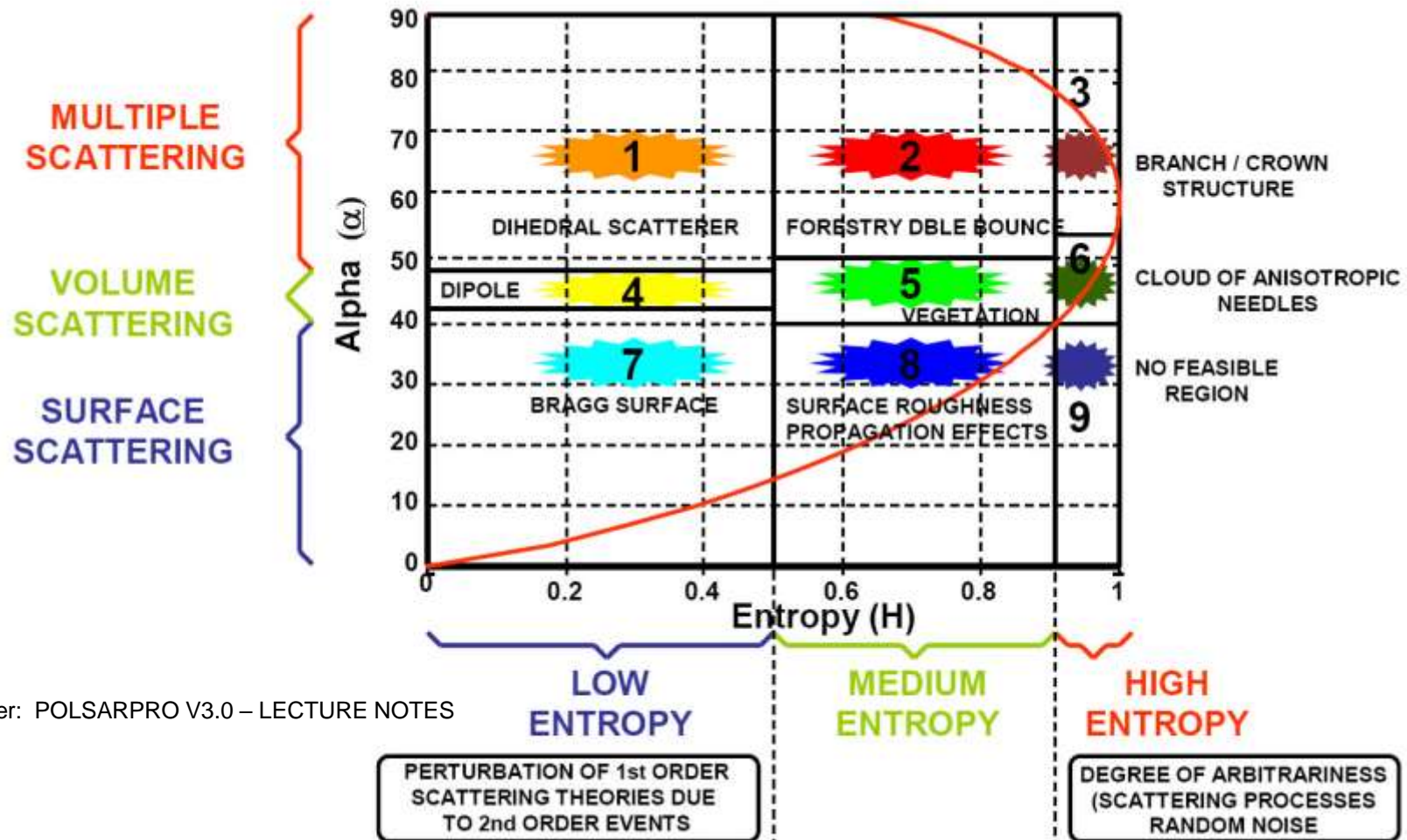
Example of SAR Data in H/α Diagram

- TerraSAR-X SAR Data
- Agricultural Fields are shown
- Data Characteristics:
 - Spread across a range of H values with average on $H \approx 0.8$
 - Low α indicates dominance of surface scattering
 - Indication of various density of crops on fields



The Unsupervised H / α Classifier

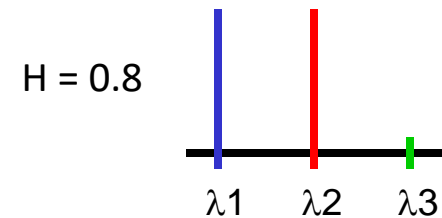
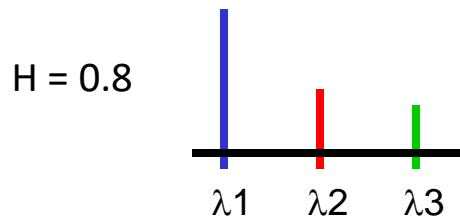
- The H/ α plane is partitioned into nine classes
- Classes are chosen on general properties of scattering



W. Boerner: POLSARPRO V3.0 – LECTURE NOTES

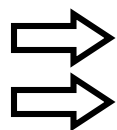
Refinement: The Unsupervised H / A / α (Entropy/**Anisotropy**/Alpha) Classifier

- In cases of high Entropy ($H > 0.7$), the H / α classification scheme is not unique.
- E.g. we can not discriminate between the following scattering cases:



- **Anisotropy** was introduced to solve these problems by distinguishing cases where $I_2 \approx I_3$ from cases where I_2 is large and I_3 is small

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

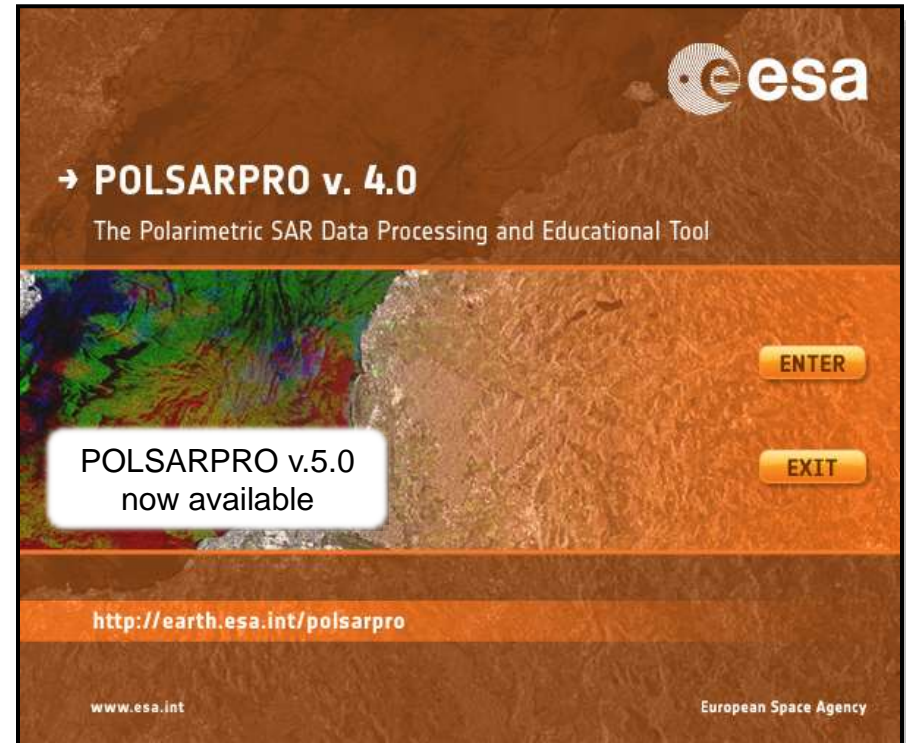


Complementary to Entropy H

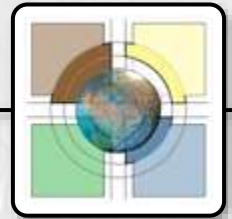
Additional discrimination when $H > 0.7$

Available Software Tools

- PolSARPro is a freely available software tool provided by ESA
- Can be downloaded at:
<https://earth.esa.int/web/polsarpro/home>
- Includes all available polarimetric decomposition methods for dual-pol and quad-pol SAR data
- Output files of PolSARPro can be imported into MapReady for geocoding.
- In addition to the software also tutorials to Polarimetry can be downloaded



Outline:



POLARIMETRIC SIGNALS

POLARIMETRIC SAR ACQUISITION

INTERPRETATION OF POLARIMETRIC DATA

EXAMPLES OF SAR IMAGE CLASSIFICATION

The H / α Classifier

Example Natural Environment

- Mixed natural environment in Germany



Optical image



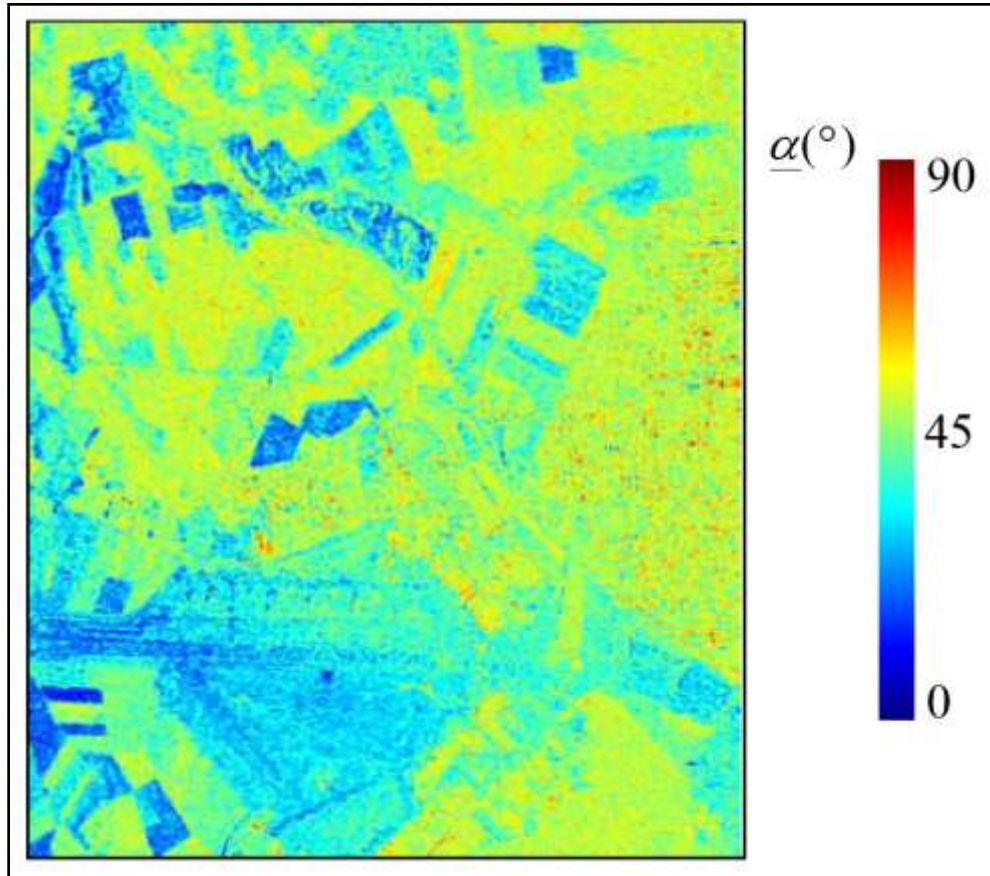
Polarimetric SAR color composite

The H / α Classifier

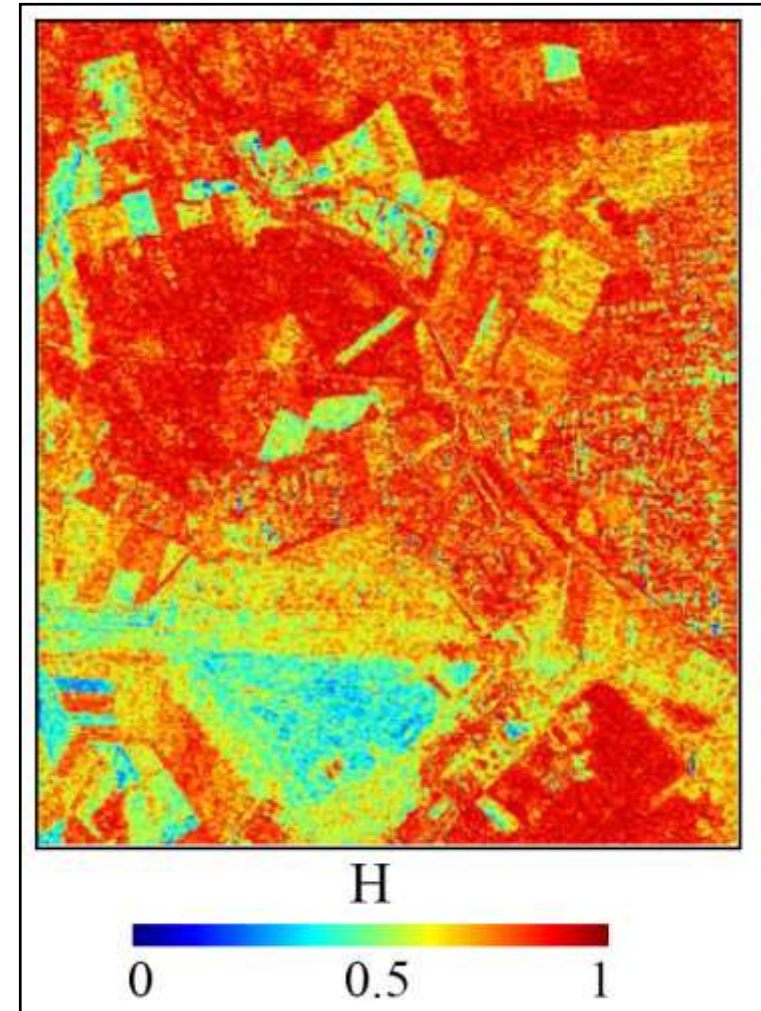
Example Natural Environment

- Mixed natural environment in Germany

α angle



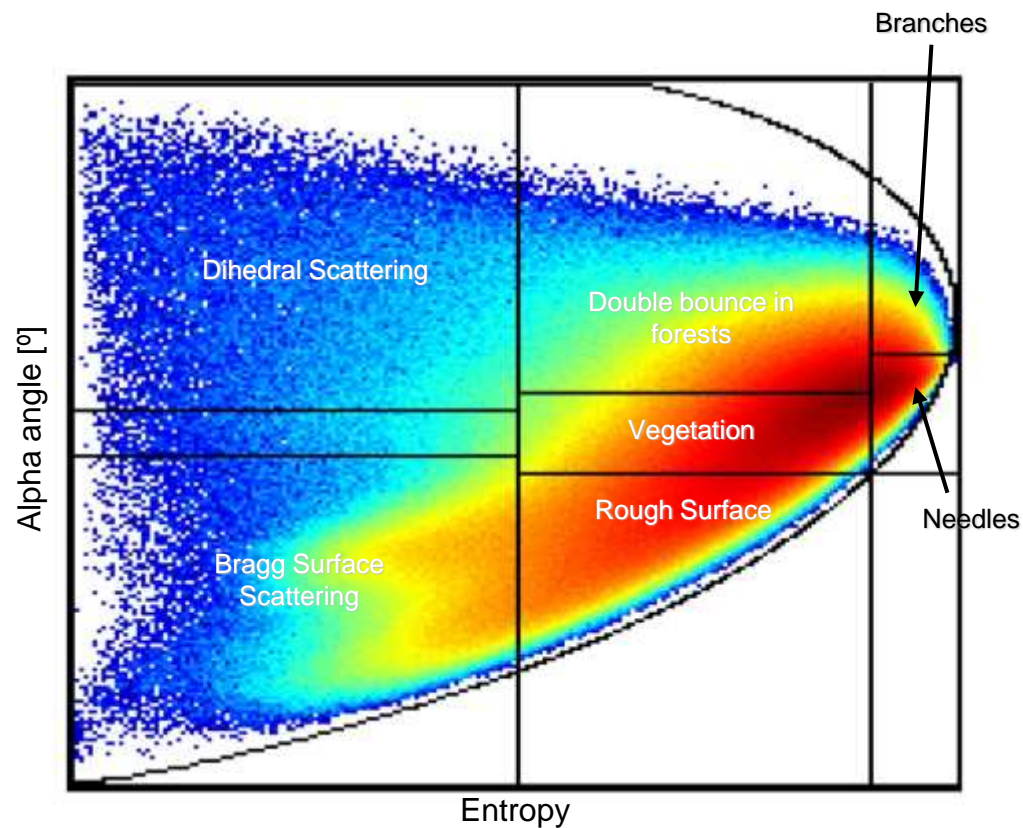
Entropy



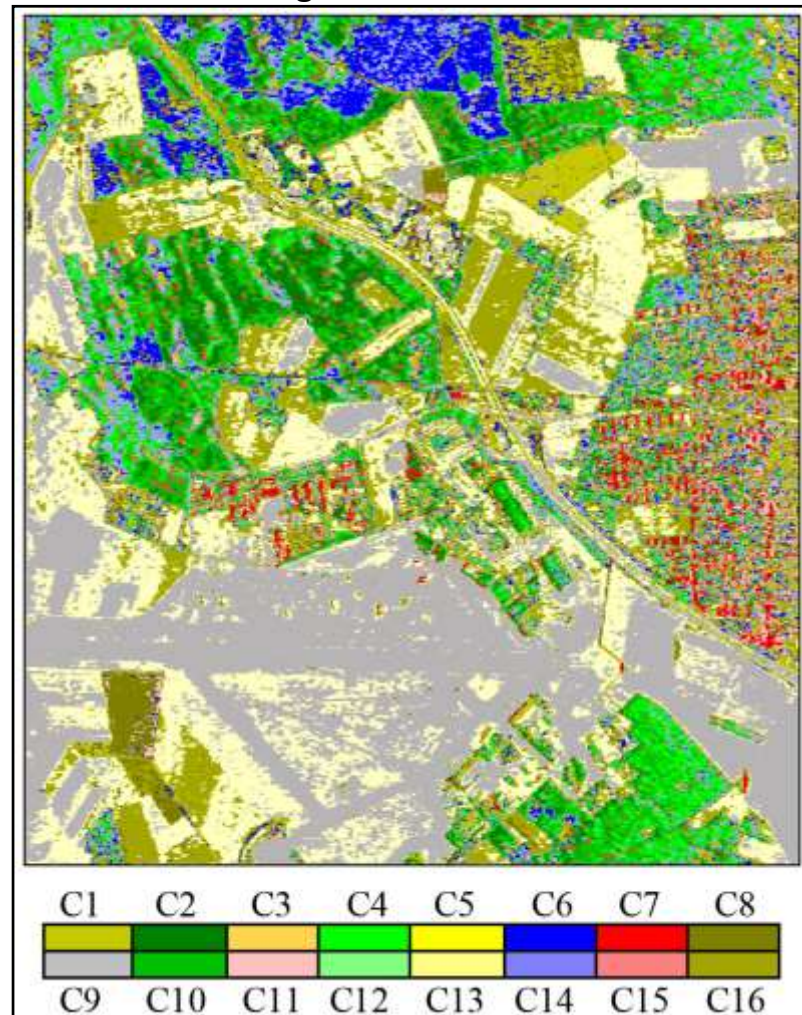
Courtesy: I. Hajnsek/DLR

The H / α Classifier

Example Natural Environment



Segmentation Result



Courtesy: I. Hajnsek/DLR

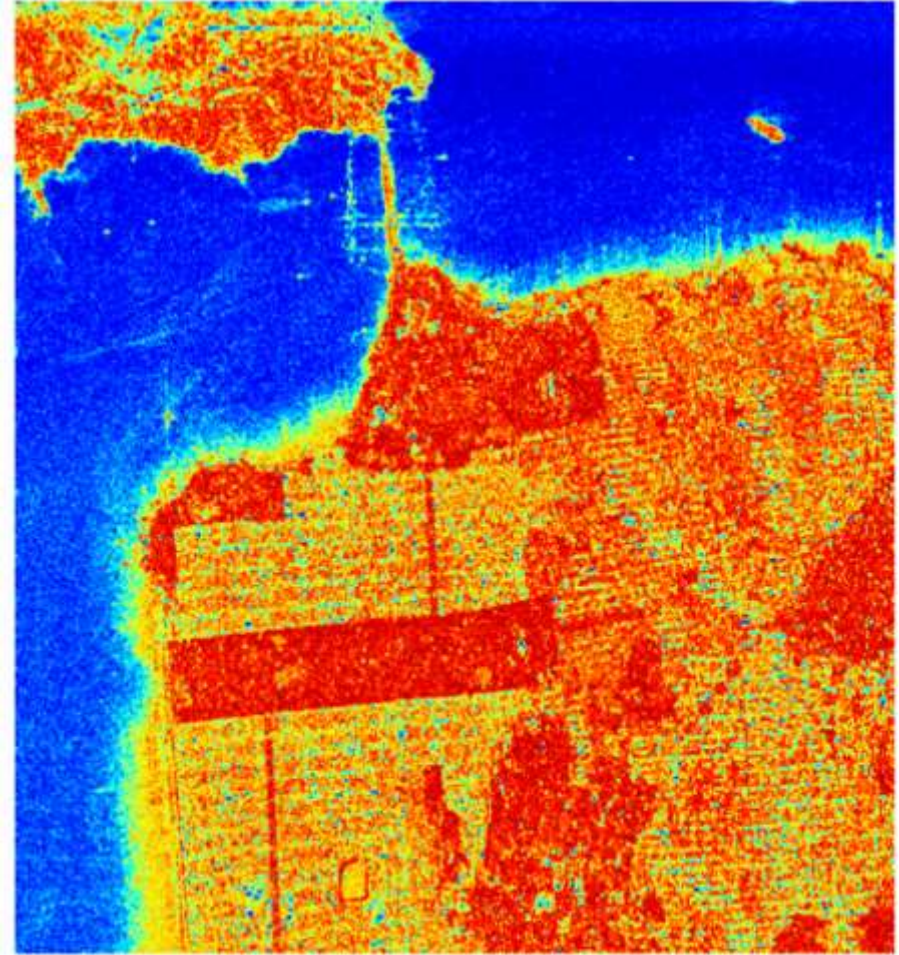
The H / α Classifier

Example: San Francisco Bay Area



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Pauli representation



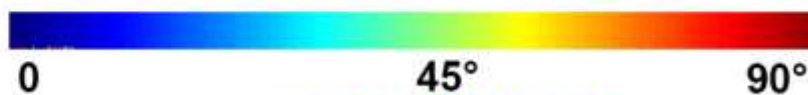
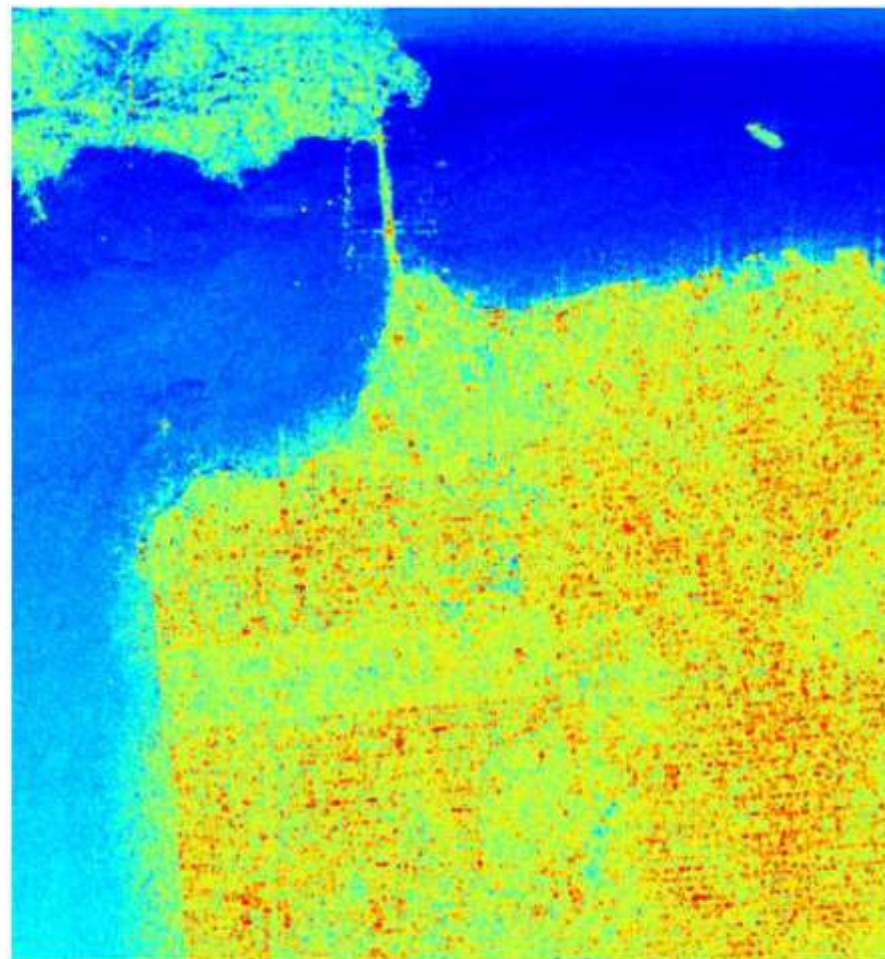
The H / α Classifier

Example: San Francisco Bay Area



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Pauli representation

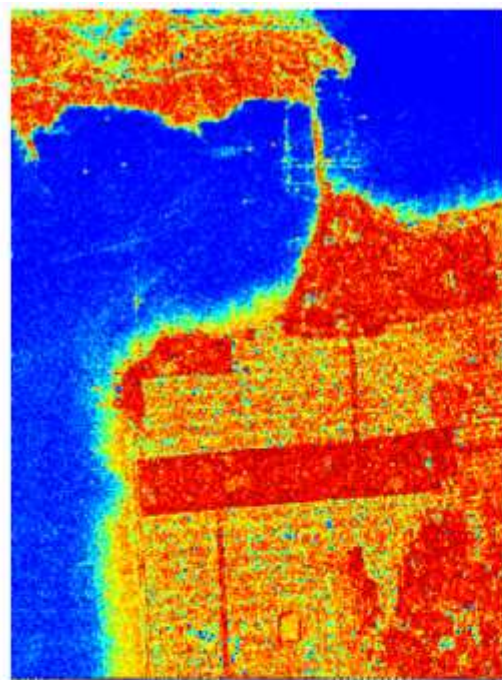


α PARAMETER



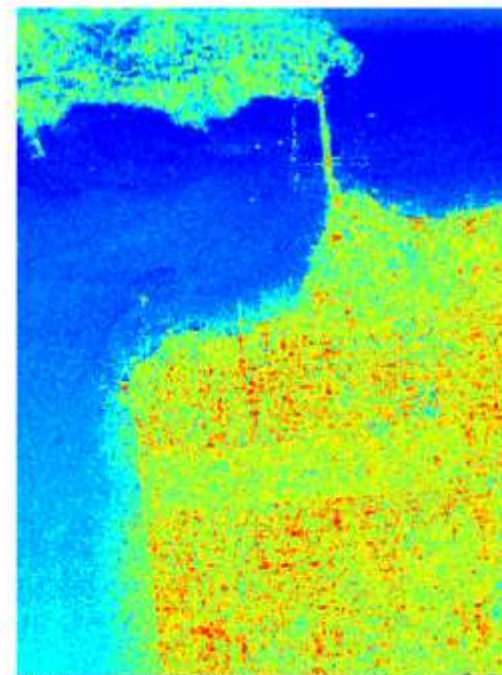
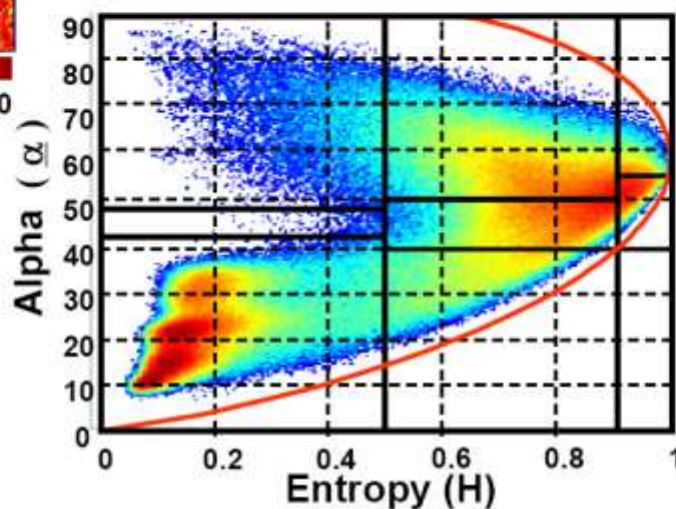
The H / α Classifier

Example: San Francisco Bay Area



0 0.5 1.0
H

POLSAR DATA
DISTRIBUTION
IN THE
H / α PLANE



0 45° 90°
 α

© E. Pottier – L. Ferro-Famil (05/2009)

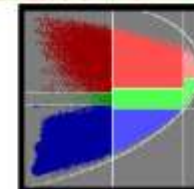
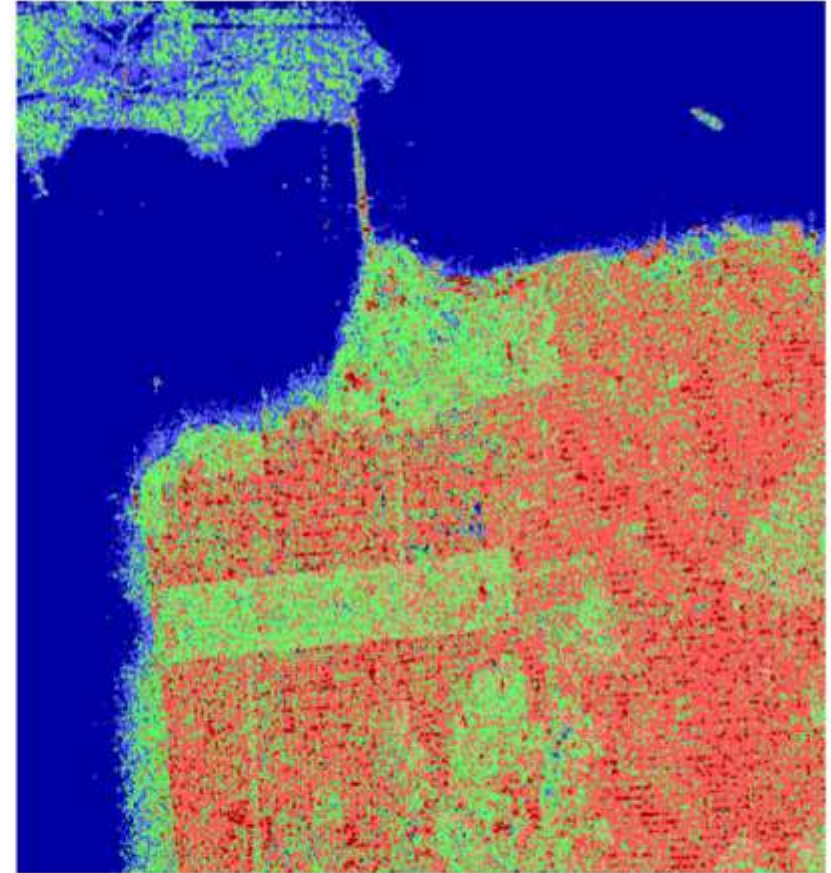
The H / α Classifier

Example: San Francisco Bay Area

H - α classification



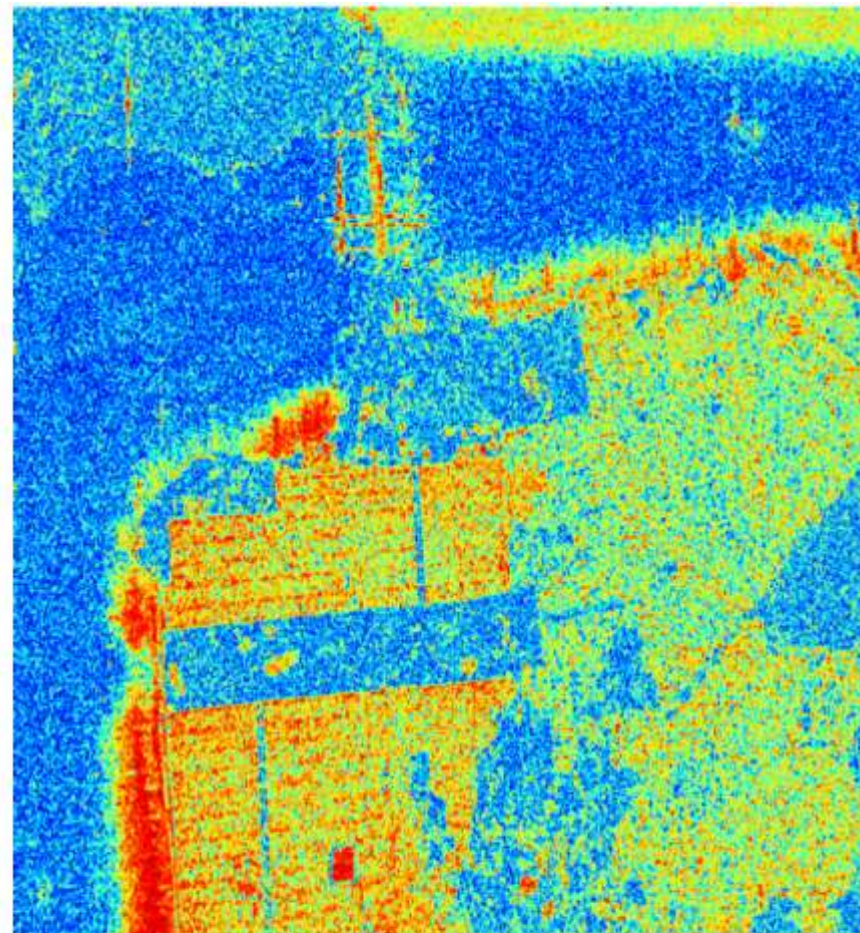
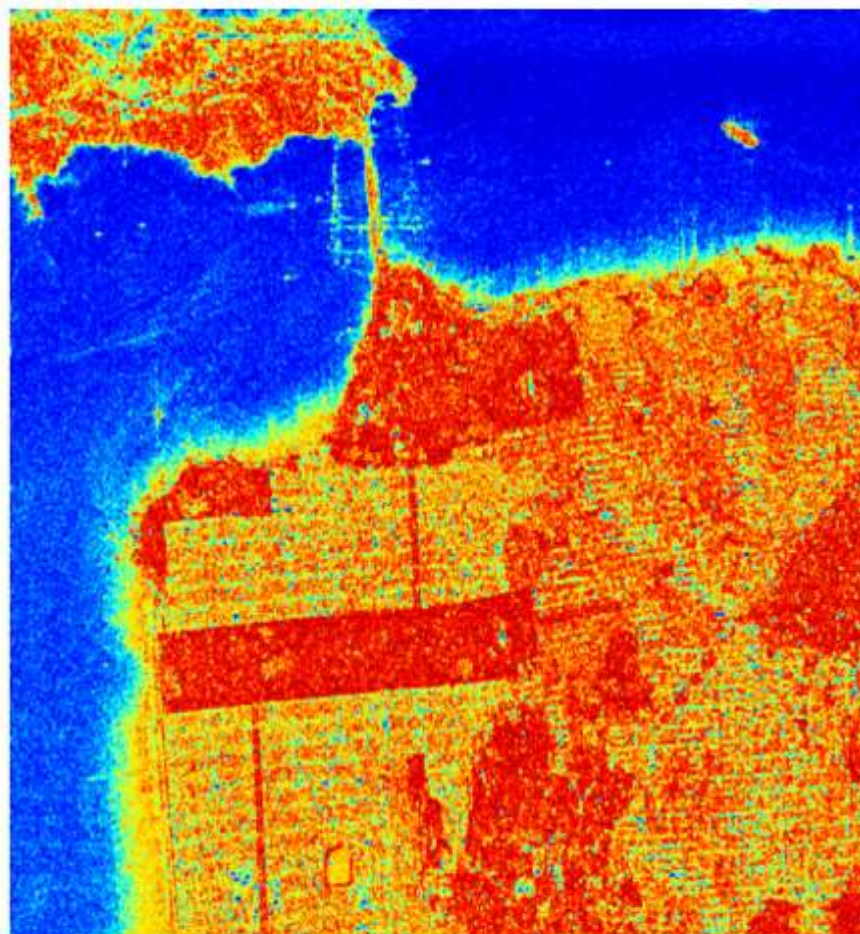
Pauli representation



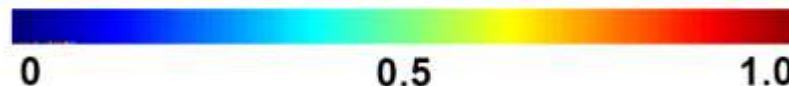
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The H / A / α Classifier

Example: San Francisco Bay Area



ENTROPY (H)



ANISOTROPY (A)

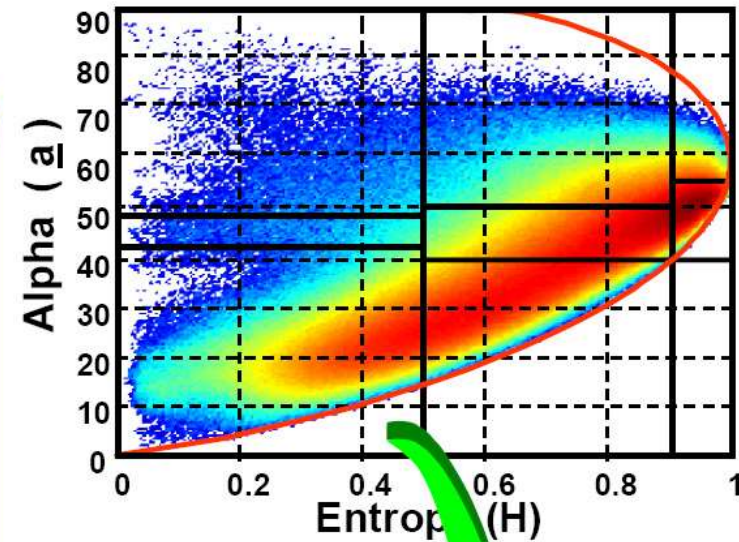
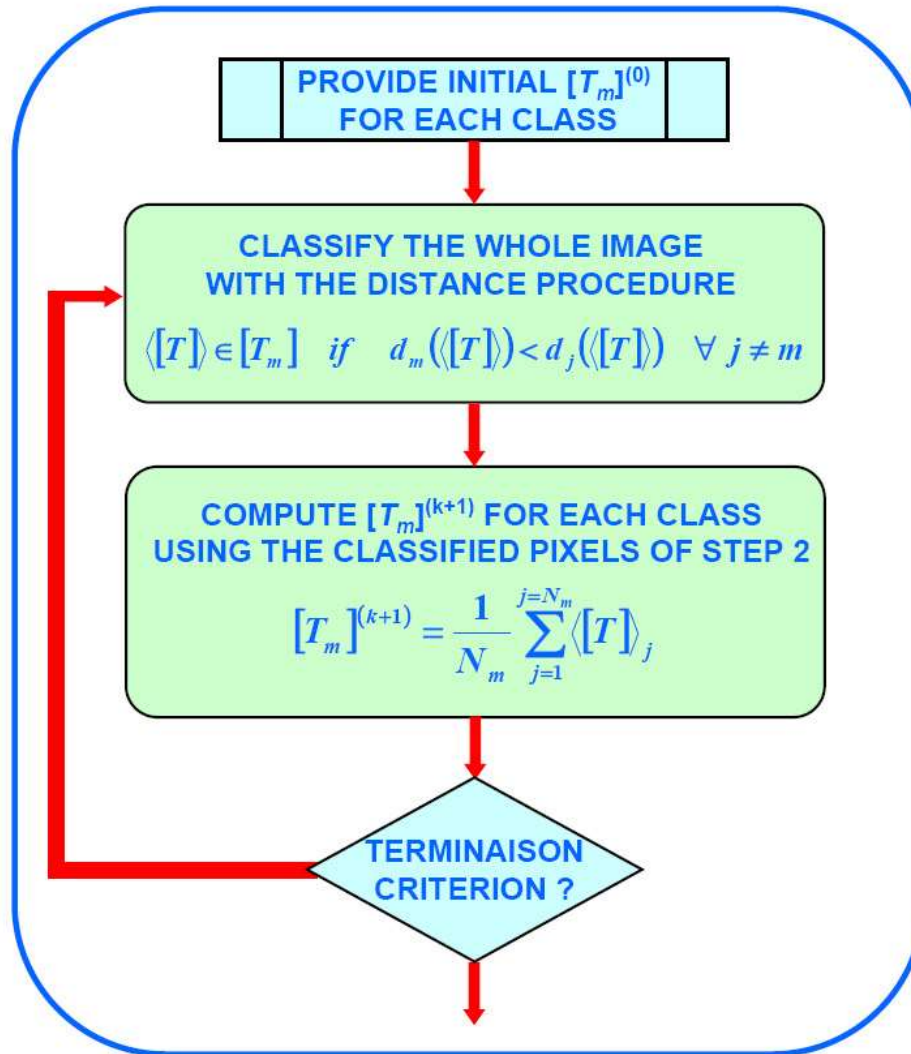
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Unsupervised Classification based on the H / A / α Decomposition

- **Disadvantage of H / A / α classification schemes:**
 - H / α classification space is sub-divided into 9 basic zones
 - Location of the boundaries is fairly arbitrary and generic
 - Therefore, H / A / α classification is not optimal and more advanced ways for determining class boundaries are sought.
- **Solution:**
 - In newer approaches H / A / α classification results are combined with a subsequent Wishart or maximum likelihood classification to improve performance.
 - H / A / α classification results provide initial class centers that are improved in subsequent classification

Unsupervised Classification based on the H / A / α Decomposition

- Example of k-mean classification procedure:



$$[T_m]^{(0)} = \frac{1}{N_m} \sum_{k=1}^{k=N_m} \langle [T] \rangle_k$$

Cluster Center of the class m
(Lee 1998)



Unsupervised Classification based on the H / A / α Decomposition

- Example of k-mean classification procedure:

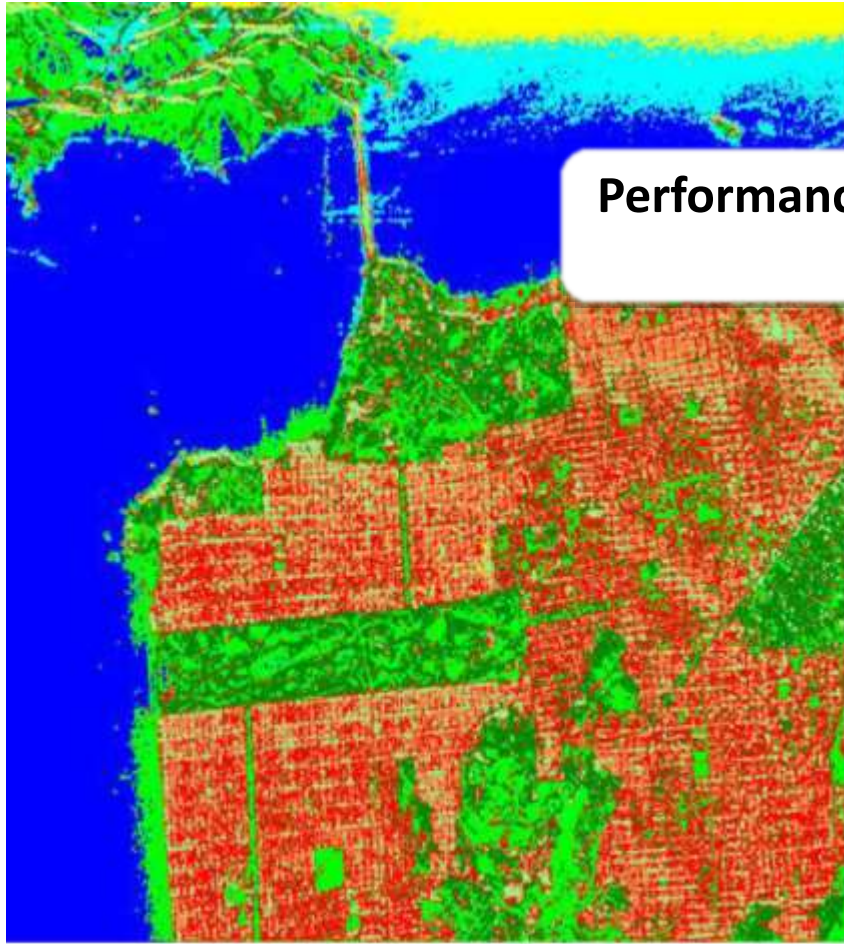


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Unsupervised Classification based on the $H / A / \alpha$ Decomposition

H/α + subsequent classifier



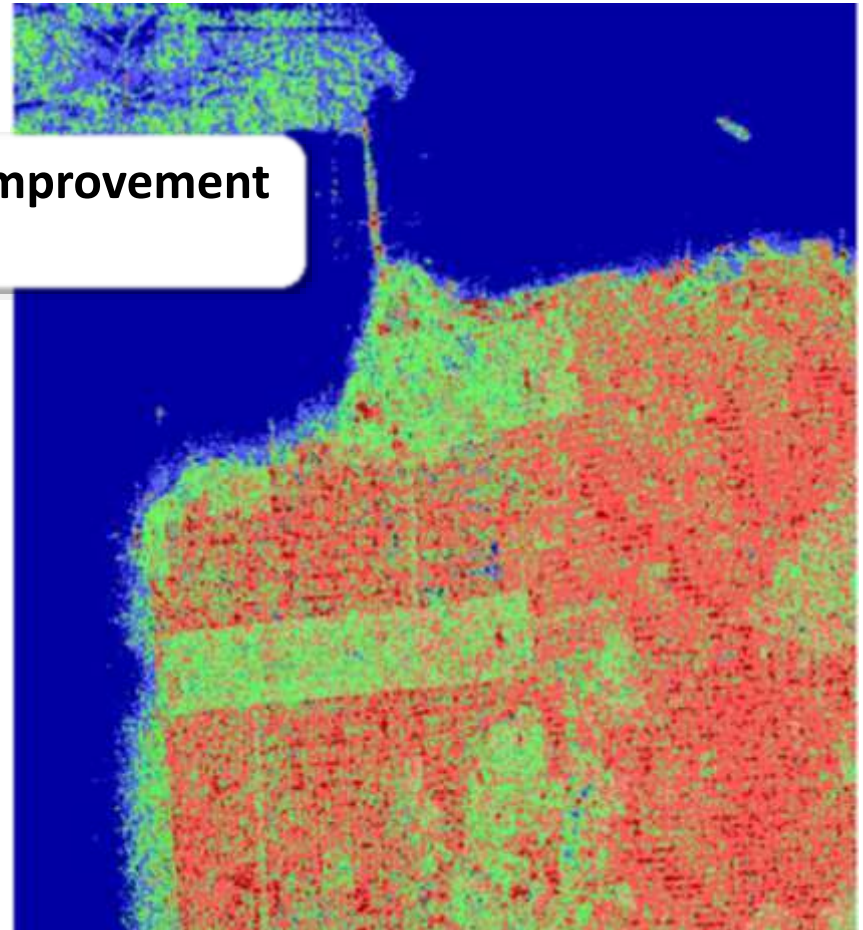
C1 C2 C3 C4 C5 C6 C7 C8



Performance improvement



H/α only



Polarimetric Classification

The Role of the Observation Wavelength

- Different penetration in different frequencies produce different polarimetric signatures

C-band data: Polarimetric data is volume dominated



Australian Pasture

Courtesy of J.S. Lee



4th Iteration (15 classes)



Polarimetric Classification

The Role of the Observation Wavelength

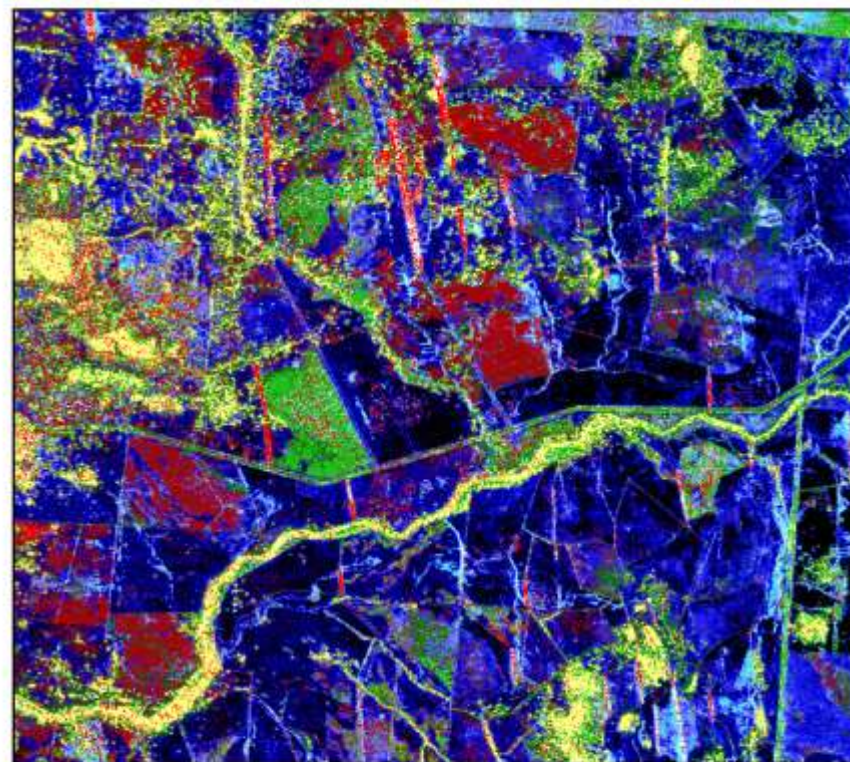
- Different penetration in different frequencies produce different polarimetric signatures

L-band data: Higher penetration causes more surface signal

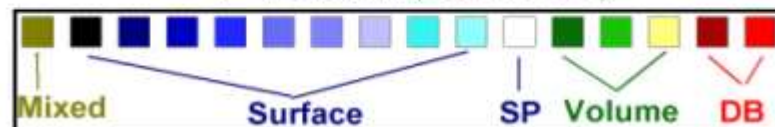


Australian Pasture

Courtesy of J.S. Lee



4th Iteration (15 classes)





QUESTIONS?

UP NEXT: WHERE TO GET SAR DATA

