

# SAR TOMOGRAPHY

## Spectral Analysis (Specan) techniques

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- *From 3-D SAR Imaging to the Beamformer*
- *PolTomSAR imaging using 1D Specan techniques*
- *Advanced PolTomSAR imaging using Specan*
- *Polarimetric TomoSAR tomography*

*Full-Rank specan & and SKP decomposition*

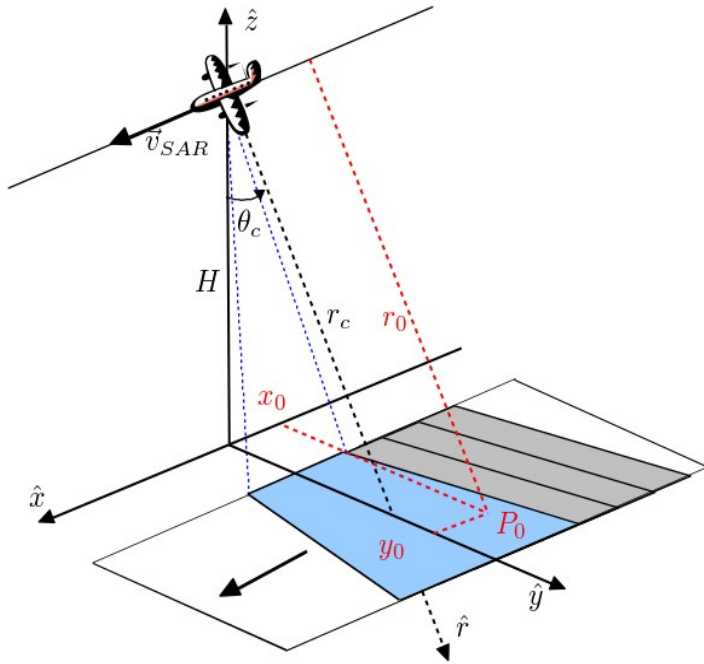
- *Spaceborne 3-D imaging using correlation SAR tomography*



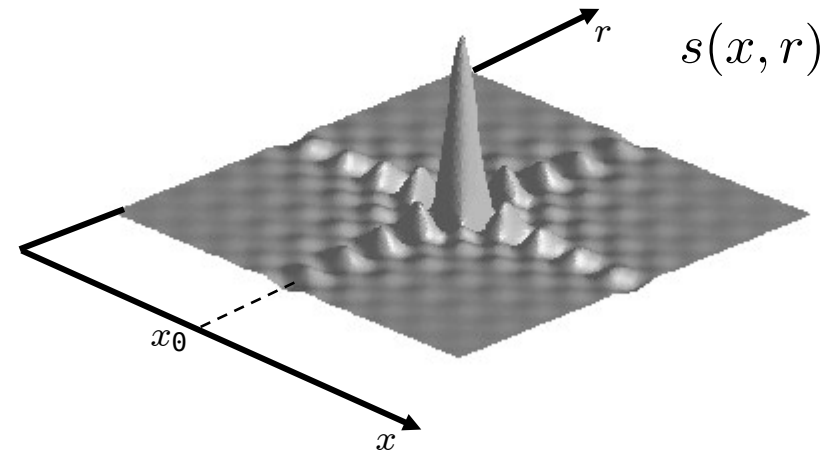
# *From 3-D Synthetic Aperture Imaging To the Beamformer*



# 2-D SAR impulse response



2-D focused signal (x-r domain)



$$s(x, r) = a_c h_r(d - r_0) h_a(x - x_0) \exp(-j \frac{4\pi}{\lambda_c} r_0)$$

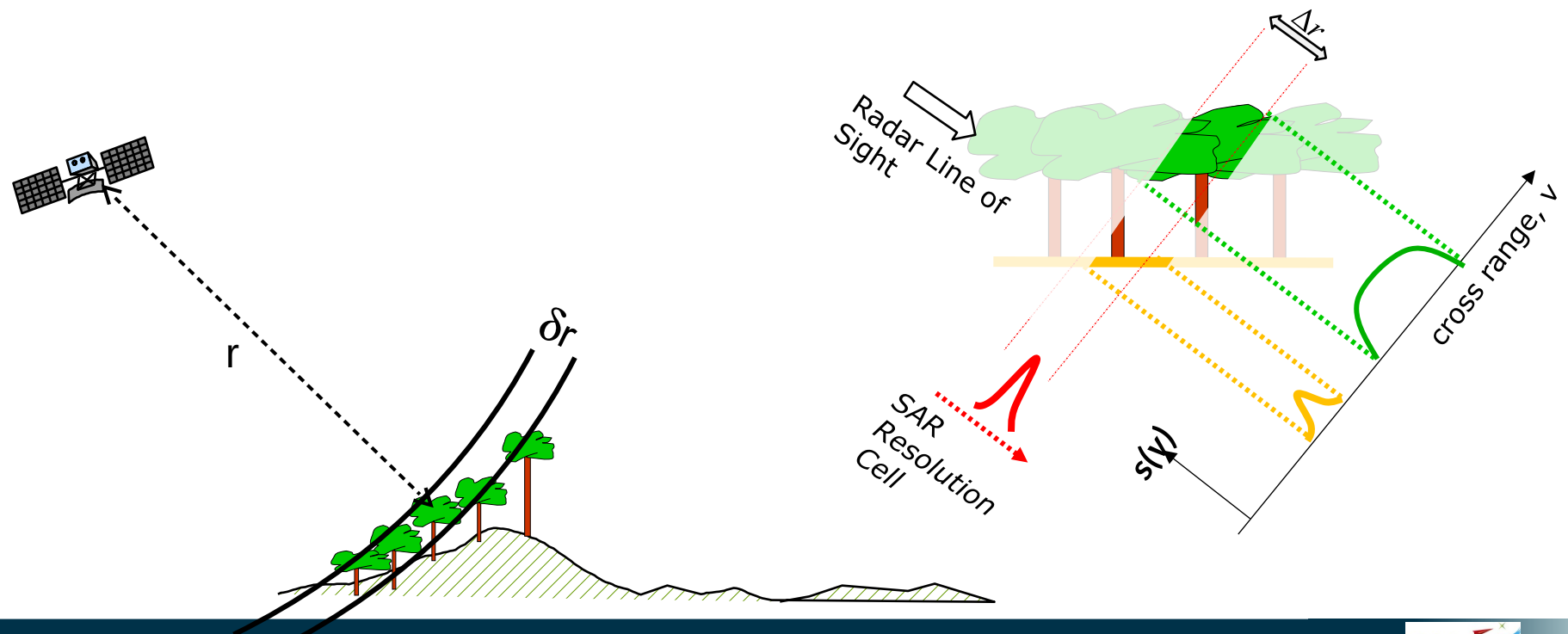
- **complex reflection coefficient**
- **delayed range impulse response**
- **delayed azimuth impulse response**
- **two-way propagation phase**

# 2-D SAR imaging

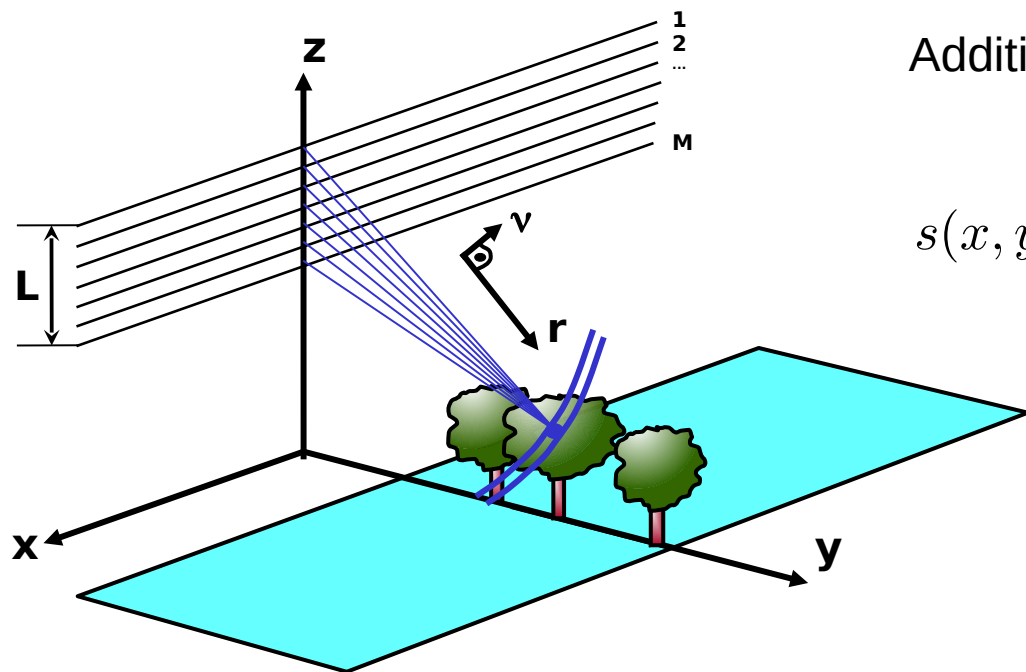
SAR imaging: coherent integration of a reflectivity density

$$s(x, r) = \int a_c(x', r', \nu') h(x' - x, r' - r) e^{-j \frac{4\pi}{\lambda_c} d(x' - x, r' - r)} dx' dr' d\nu$$

$$s(x, r) \approx \int_C a_c(x, r, \nu) e^{-jk_c r(\nu)} d\nu$$



# 3-D SAR imaging

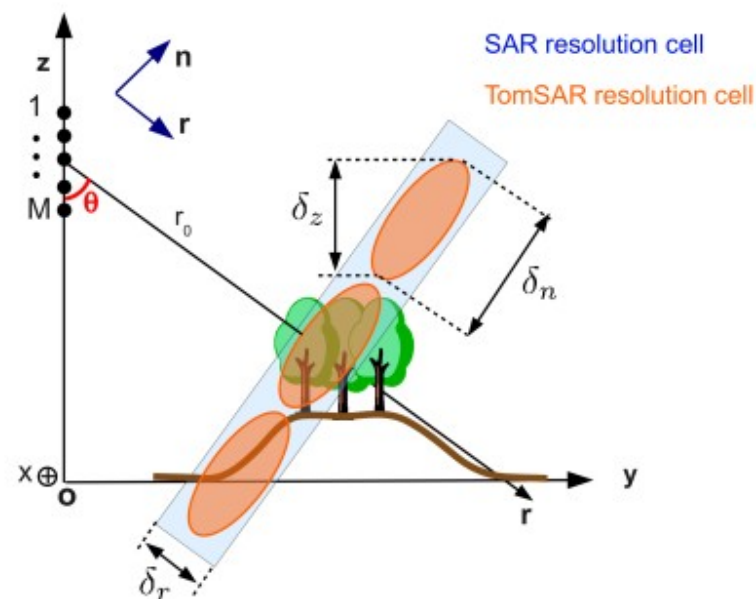


Additional aperture in elevation:  
2-D → 3-D focusing

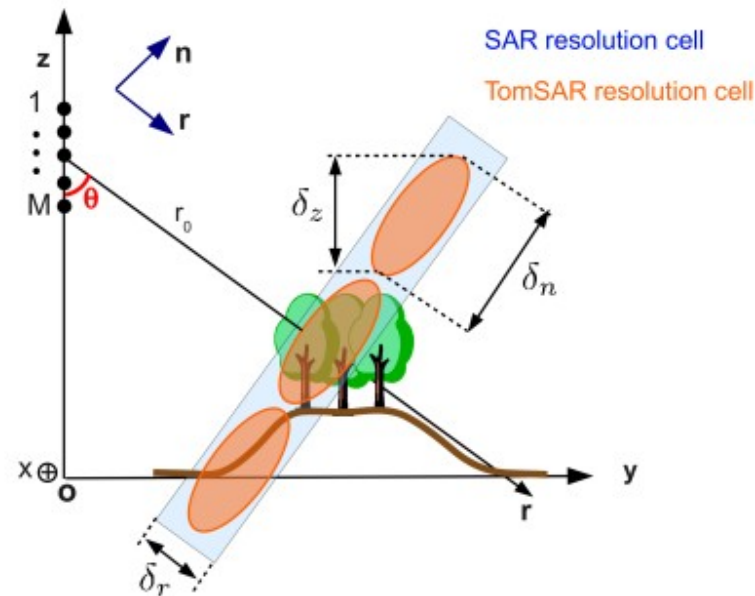
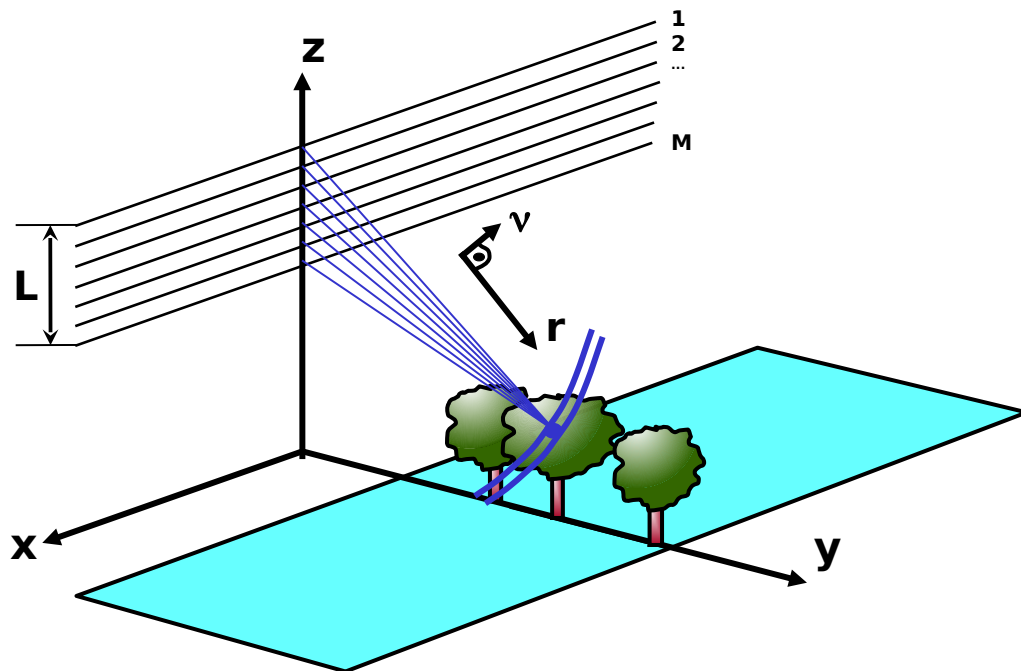
$$s(x, y, z) = \sum_{i=1}^M s_i(x, r(y, z)) e^{jk_c r(y, z)}$$

Vertical aperture :  $L_{tomo}$

Resolution :  $\delta_z = \delta_n \sin \theta$  with  $\delta_n = \frac{\lambda R_0}{2L_{tomo}}$



# 3-D SAR imaging



$$s(x, y, z) = \sum_{i=1}^M s_i(x, r(y, z)) e^{jk_c r(y, z)}$$

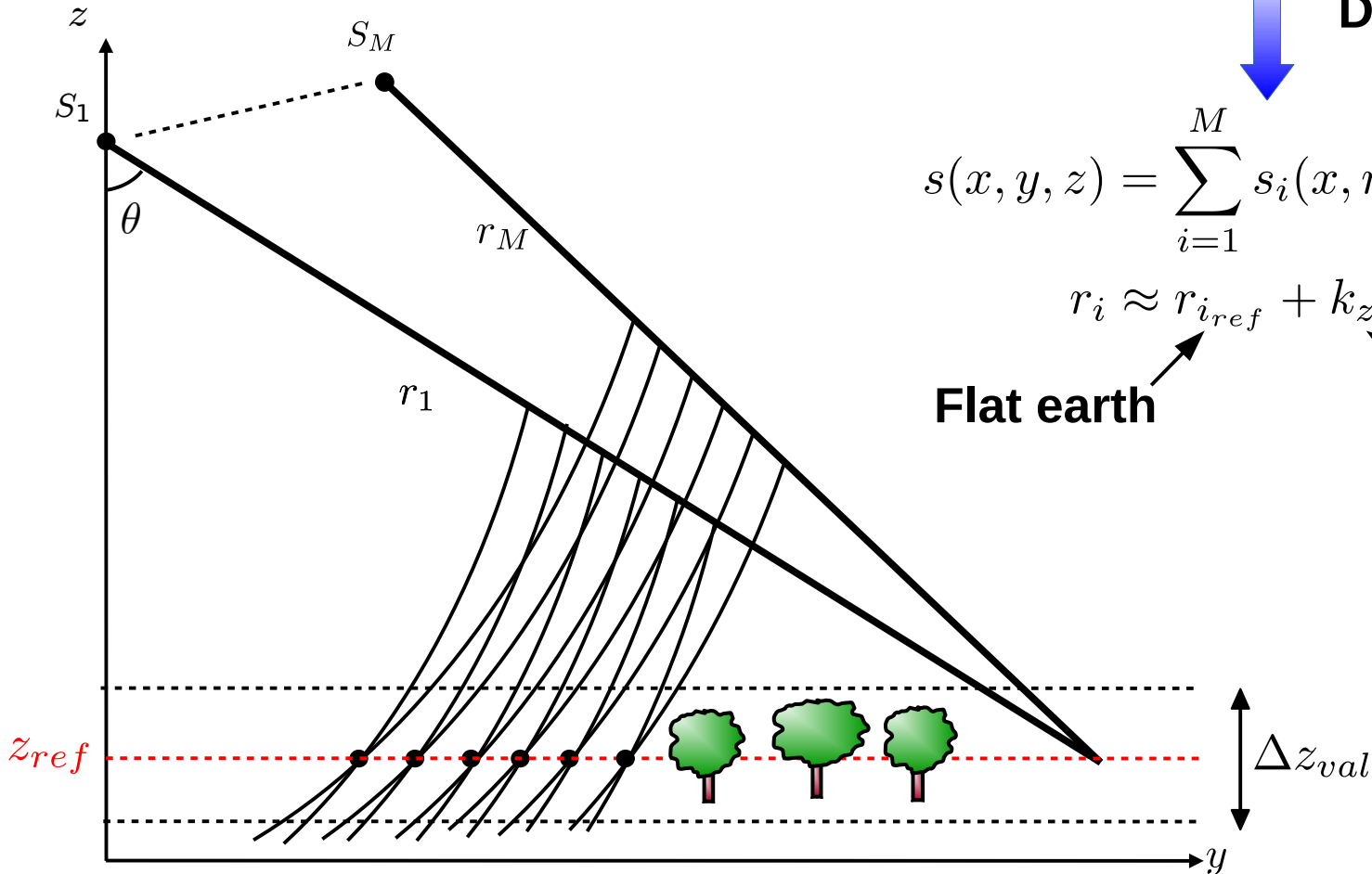
Interpolation

HF term

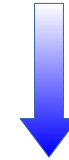
# 3-D SAR imaging

Co-registration on a reference plane

Valid for  $z \in z_{ref} \pm \Delta z_{val}/2$



$$s(x, y, z) = \sum_{i=1}^M s_i(x, r(y, z)) e^{jk_c r(y, z)}$$



**Discretization  
NN-interp.**

$$s(x, y, z) = \sum_{i=1}^M s_i(x, r_{i_{ref}}) e^{jk_c r_i}$$

$$r_i \approx r_{i_{ref}} + k_{z_i} z$$

**Flat earth**

**Elevation**

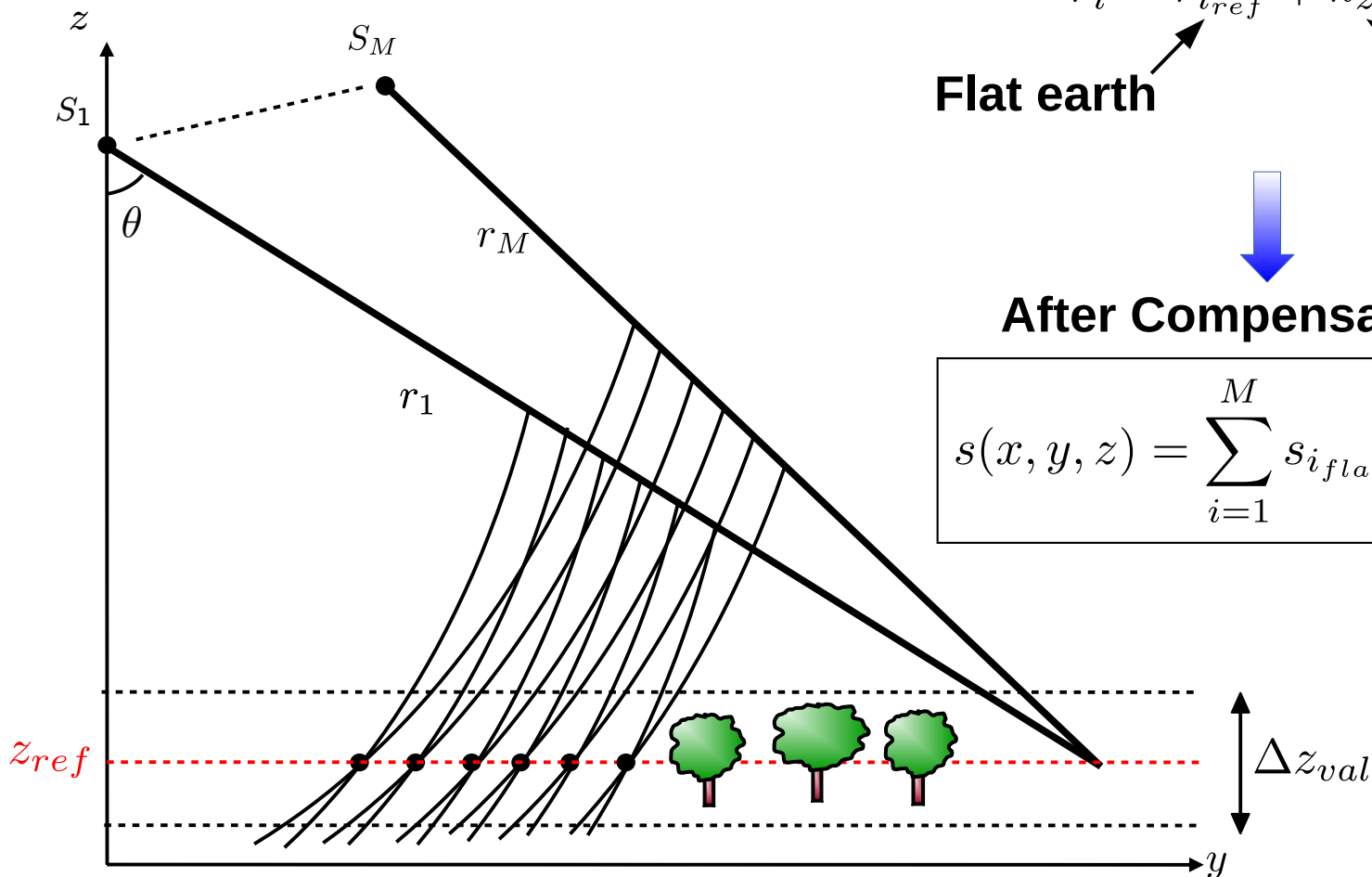




# 3-D SAR imaging

Co-registration on a reference plane

Valid for  $z \in z_{ref} \pm \Delta z_{val}/2$

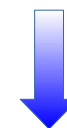


$$s(x, y, z) = \sum_{i=1}^M s_i(x, r_{i_{ref}}) e^{jk_c r_i}$$

$$r_i \approx r_{i_{ref}} + k_{z_i} z$$

Flat earth

Elevation



After Compensation

$$s(x, y, z) = \sum_{i=1}^M s_{i_{flat}}(x, r_{i_{ref}}) e^{jk_{z_i} z}$$

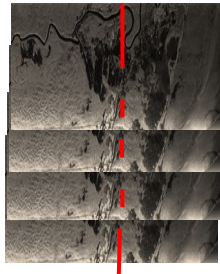
# 3-D SAR imaging: 2D + 1D processing

## 3-D Synthetic Aperture imaging

$$s(x, y, z) = \sum_{i=1}^M s_i(x, r_{i_{ref}}) e^{jk_{z_i} z}$$

Filter-like formulation for a given 2-D resolution cell

*Coregistered  
Resampled  
Flattened  
Single Look Complex  
(SLC) data*



$$\Rightarrow \mathbf{y} = \begin{bmatrix} y_1 \\ \vdots \\ y_M \end{bmatrix} = \begin{bmatrix} s_1(x, r_{1_{ref}}) \\ \vdots \\ s_M(x, r_{M_{ref}}) \end{bmatrix}$$

## 1D Linear filter

$$s(z) = \sum_{i=1}^M y_i e^{-jk_{z_i} z} = \mathbf{a}^H(z) \mathbf{y}$$

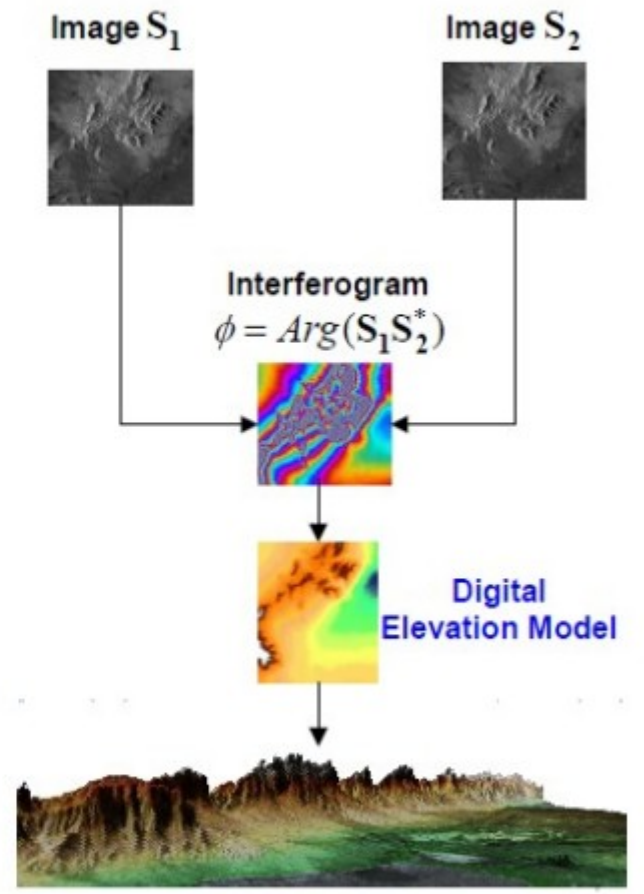
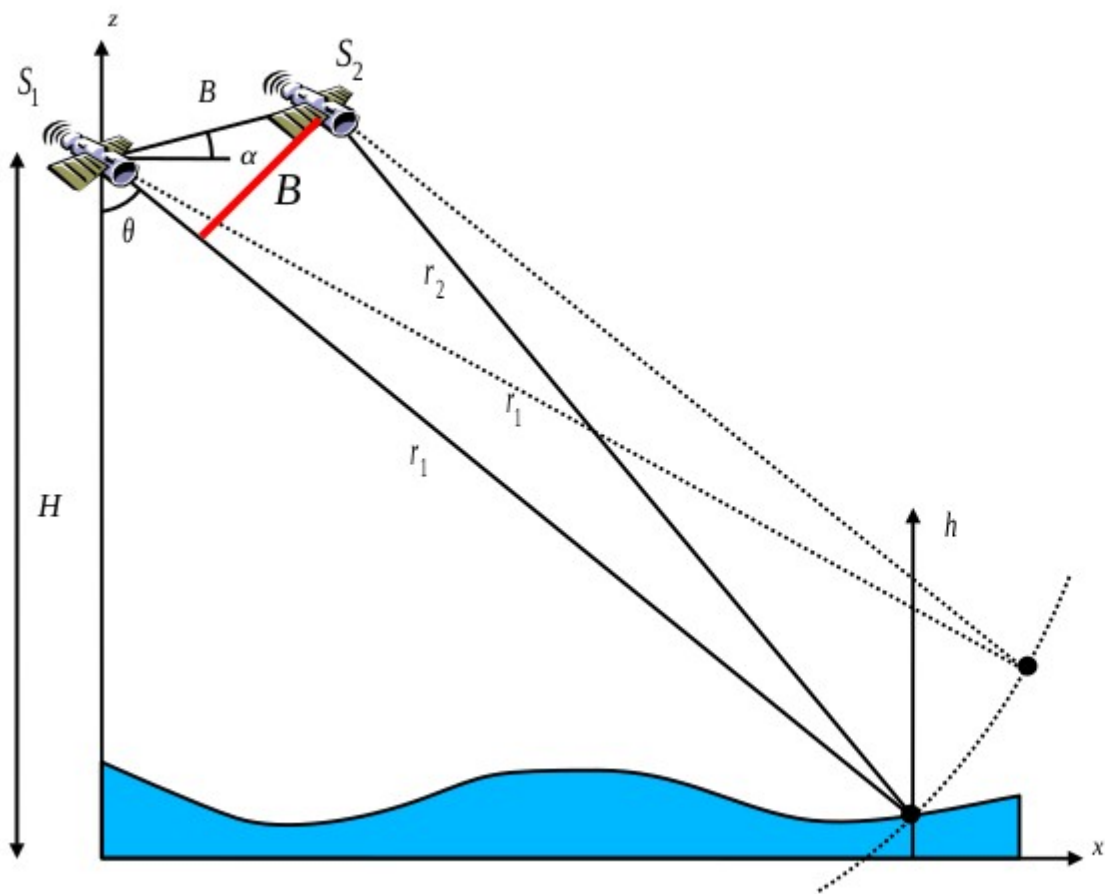
Steering vector

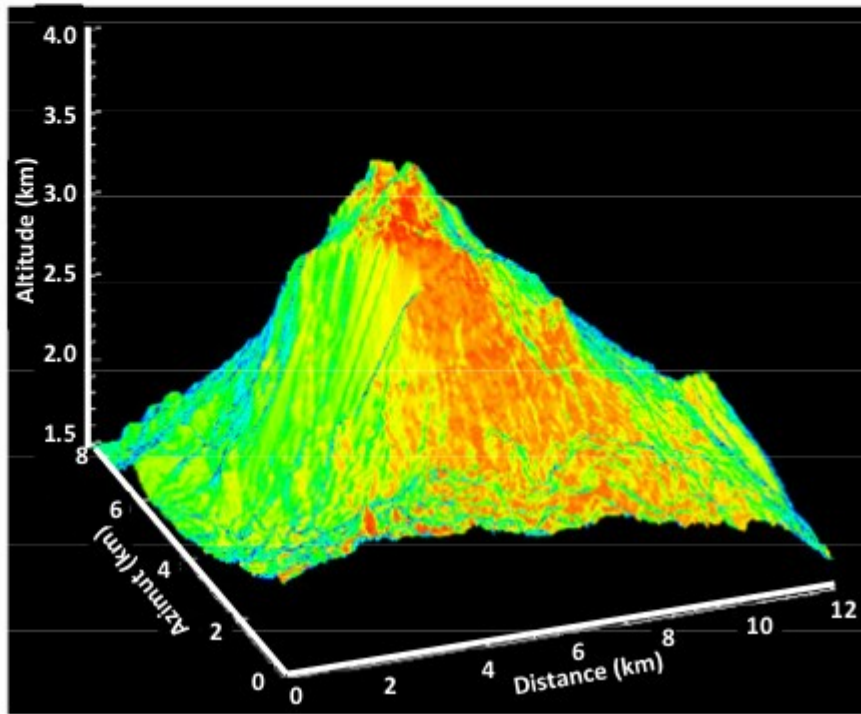
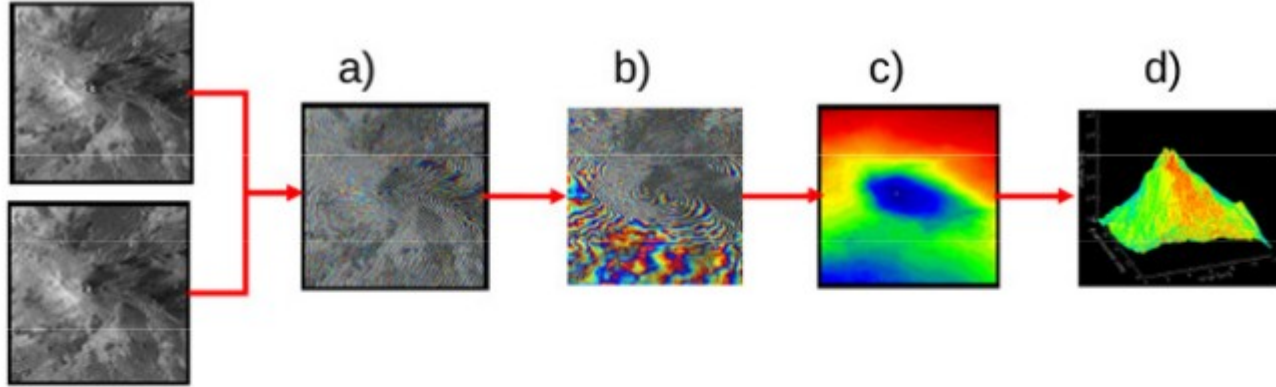
$$\mathbf{a} = [1, e^{jk_{z_2} z}, \dots, e^{jk_{z_M} z}]^T$$

# *TomoSAR imaging using Monodimensional Spectral Analysis Techniques*



# Interferometric phase variations with height





# Estimation of a single scatterer, $M=2$ images

## InSAR way

$$\begin{cases} s_1 = a_c e^{j\xi} \\ s_2 = a_c e^{j\xi + \Delta\phi} \end{cases} \Rightarrow \begin{cases} \Delta\hat{\phi} = \arg(s_2 s_1^*) \\ \hat{I} = \frac{|s_1|^2 + |s_2|^2}{2} \end{cases}$$

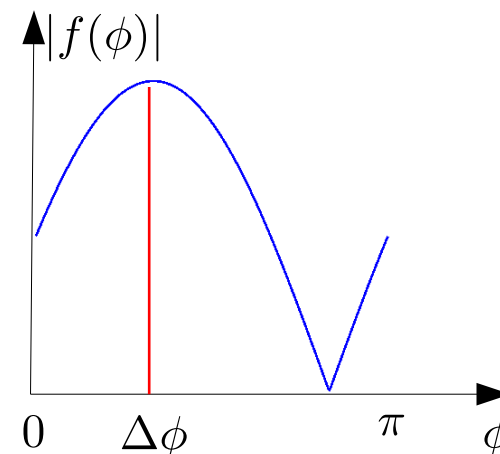
## Linear filtering way

$$\mathbf{y} = \begin{bmatrix} s_1 \\ s_2 \end{bmatrix} = a_c e^{j\xi} \begin{bmatrix} 1 \\ e^{j\Delta\phi} \end{bmatrix}, \mathbf{a}(\phi) = \begin{bmatrix} 1 \\ e^{j\phi} \end{bmatrix}$$

$$f(\phi) = \frac{\mathbf{a}^H(\phi)\mathbf{y}}{2} = \frac{s_1 + s_2 e^{-j\phi}}{2} = a_c e^{j\xi} \frac{1 + e^{-j(\Delta\phi - \phi)}}{2}$$

$$\Rightarrow \begin{cases} \Delta\hat{\phi} = \arg \max_{\phi} |f(\phi)|^2 \\ \hat{I} = |f(\Delta\hat{\phi})|^2 \end{cases}$$

- Phase estimation → linear filtering & search
- Filter output: reflectivity
- $\mathbf{a}(\phi)$  steering vector: **matched filter**



# Estimation of several scatterers, $M > 2$ images

## Estimation of several scatterers: MB InSAR way

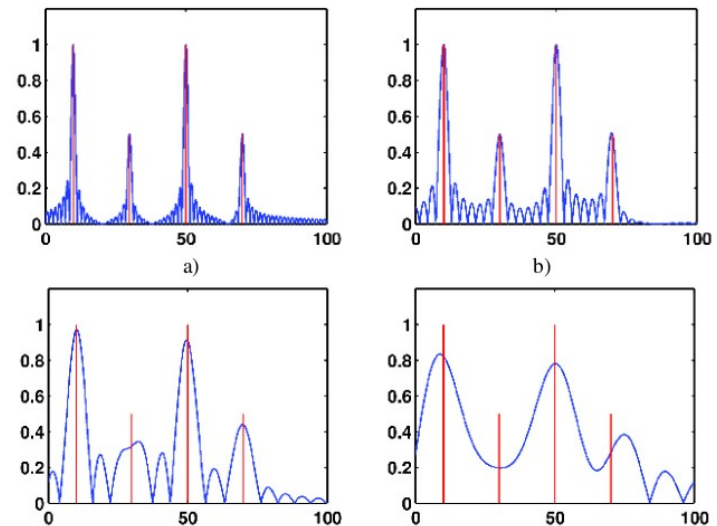
$$\{s_1, \dots, s_M\}, \quad s_m = \sum_{t=1}^{N_t} a_{ct} e^{j\xi_t} e^{jk_{z_m} z_t} \quad \Rightarrow \quad ???$$

## Estimation of several scatterers: linear filtering way

$$\mathbf{y} = \begin{bmatrix} s_1 \\ \vdots \\ s_M \end{bmatrix}, \quad \mathbf{a}(z) = \begin{bmatrix} 1 \\ \vdots \\ e^{jk_{z_M} z} \end{bmatrix}$$

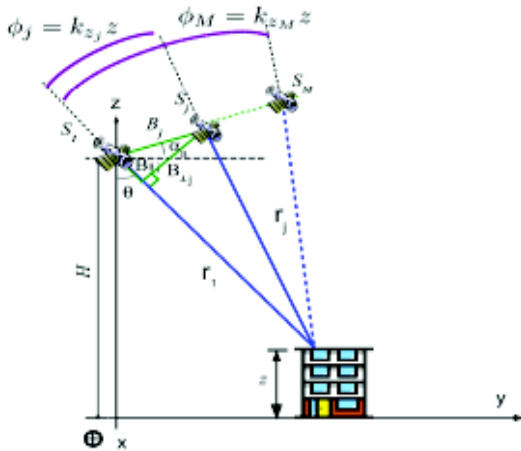
$$f(z) = \frac{\mathbf{a}^H(z)\mathbf{y}}{M} = \frac{\sum_m s_m e^{-jk_{z_m} z}}{M}$$

$$\Rightarrow \begin{cases} \hat{z}_t = \arg \max_{loc} |f(z)|^2 \\ \hat{I}_t = |f(\hat{z}_t)|^2 \end{cases}$$



- Matched filter: **Discrete Fourier Transform**
- **Tomographic focusing: spectral estimation problem**
- Estimation quality: depends on MB-inSAR configuration

# Tomographic imaging using specan



## Ideal acquired signal (single scatterer)

$$\mathbf{y} = a_c \mathbf{a}(z_0)$$

$$\text{with } \mathbf{a}(z_0) = [1, e^{jk_{z_2} z_0}, \dots, e^{jk_{z_M} z_0}]^T$$

## Uniform baseline distribution

$$B_{\perp i} = (i - 1)B_{\perp} \Rightarrow k_{z_i} = (i - 1)dk_z$$

$$\mathbf{a}(z) = [1, e^{jdk_z z}, \dots, e^{j(M-1)dk_z z}]^T$$

**Spectral sampling:**  $dk_z = \frac{k_c B_{\perp}}{r \sin \theta}$

**Spectral bandwidth:**  $\Delta k_z = Mdk_z$

$$|f(z)| = |a_c| \frac{|\mathbf{a}^H(z)\mathbf{a}(z_0)|}{M} = \frac{|a_c|}{M} \frac{|\sin(\pi \Delta k_z (z - z_0))|}{|\sin(\pi dk_z (z - z_0))|}$$

Fast

M times Slower

Periodic oscillating filter output



# Tomographic imaging using specan

Uniform baseline sampling

$$\mathbf{a}(z) = [1, e^{j\Delta k_z z}, \dots, e^{j(M-1)\Delta k_z z}]^T$$

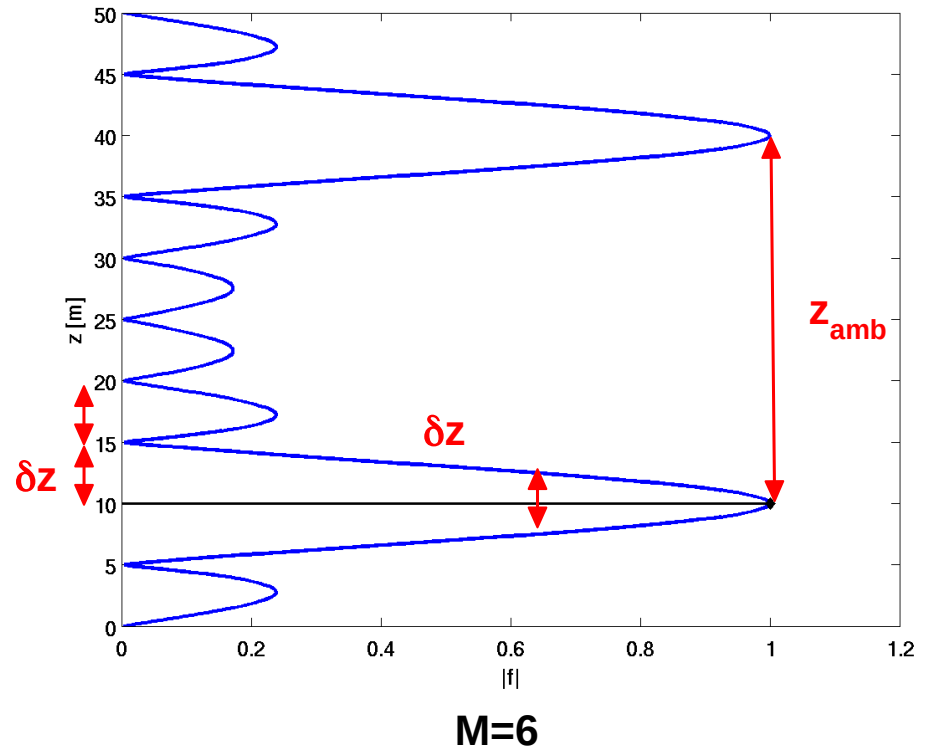
$$|f(z)| = |a_c| \frac{|\mathbf{a}^H(z)\mathbf{a}(z_0)|}{M} = \frac{|a_c|}{M} \frac{|\sin(\pi\Delta k_z(z - z_0))|}{|\sin(\pi\Delta k_z(z - z_0))|}$$

Fast → resolution  
 Slow → ambiguity

## Spatial features of a tomogram

- rapid oscillations: resolution
- band-limited: sidelobes
- sampled spectrum : spatial ambiguities

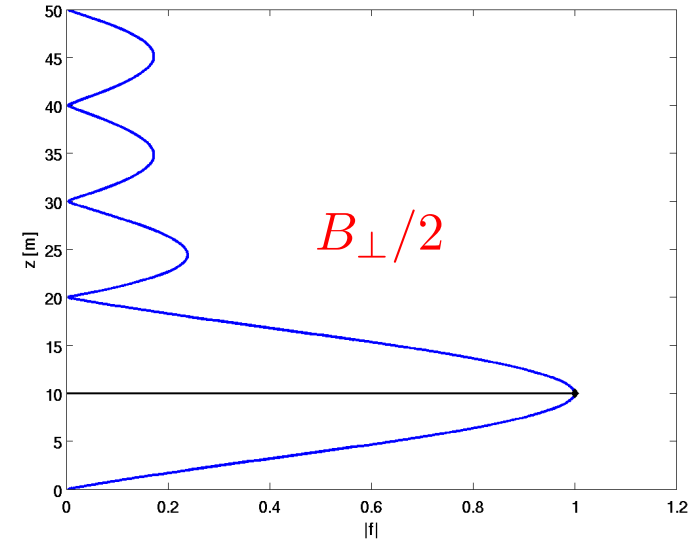
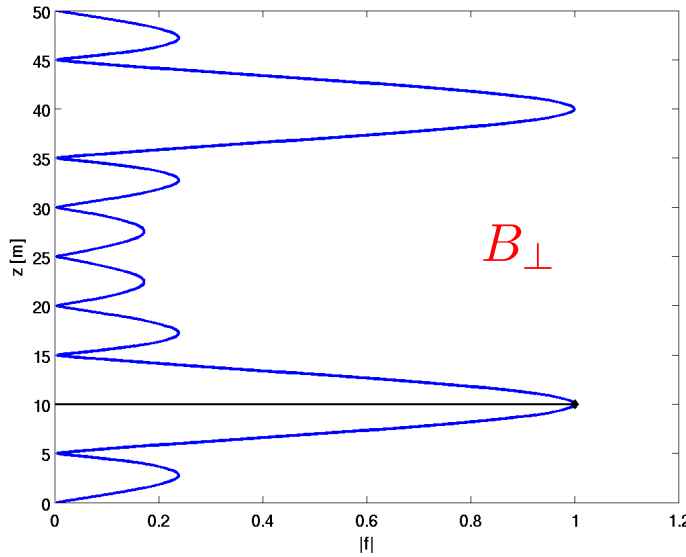
$$\delta z = \frac{2\pi}{\Delta k}, z_{amb} = \frac{2\pi}{dk}, \delta z = \frac{z_{amb}}{M}$$



# Tomographic imaging using specan

**M=6**

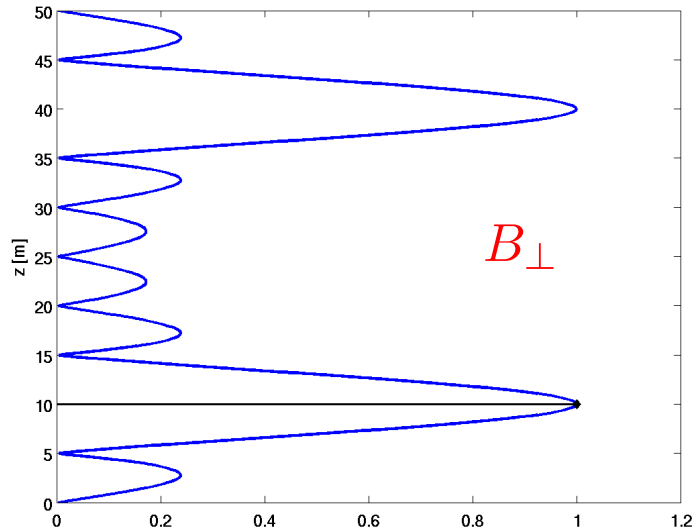
$$\Delta k_z \propto M B_{\perp}$$
$$dk_z = \frac{\Delta k_z}{M}$$



- **Reduced resolution**
- **Improved ambiguity**

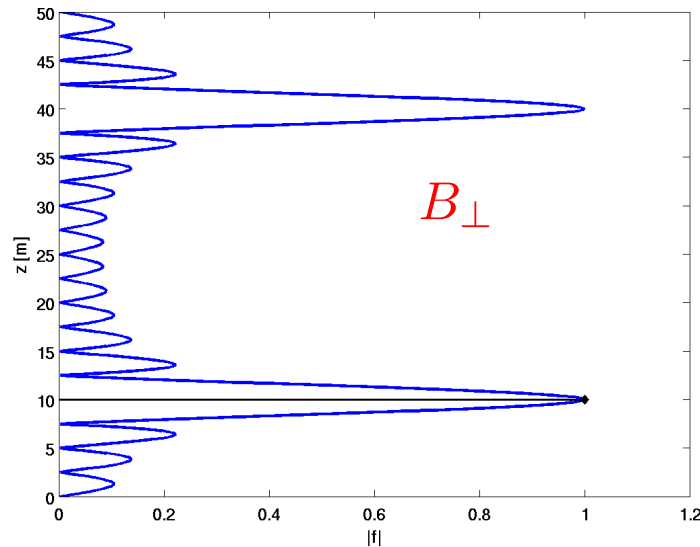
# Tomographic imaging using specan

**M=6**



$$\Delta k_z \propto M B_{\perp}$$
$$dk_z = \frac{\Delta k_z}{M}$$

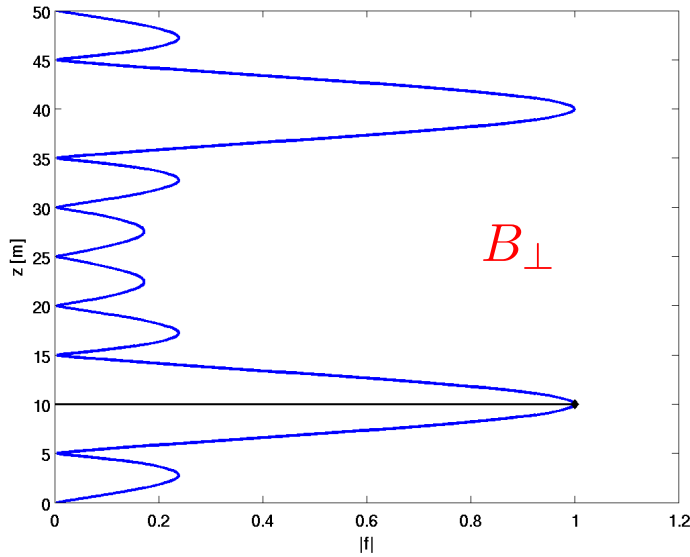
**M=12**



- Improved resolution
- Unchanged ambiguity

# Tomographic imaging using specan

**M=6**

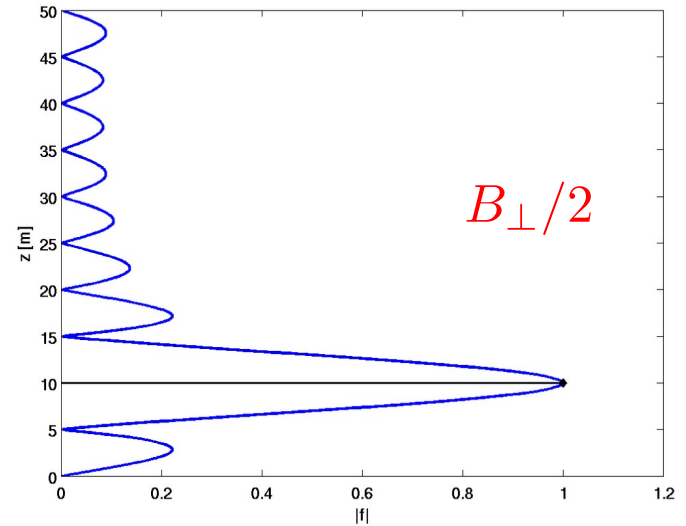


$$\Delta k_z \propto M B_{\perp}$$

$$dk_z = \frac{\Delta k_z}{M}$$

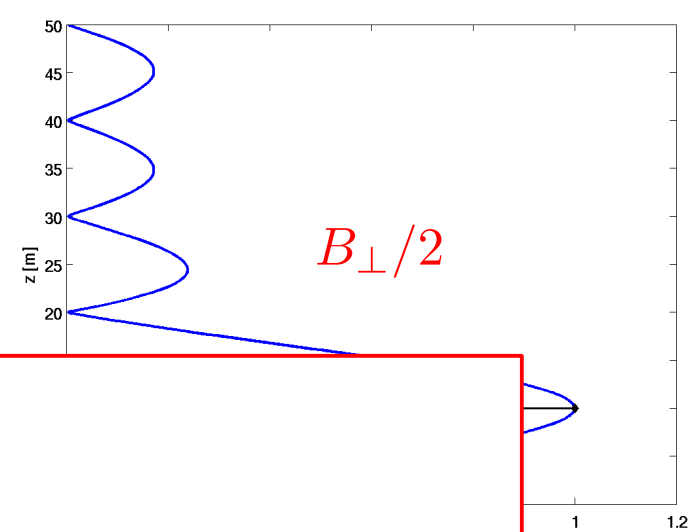
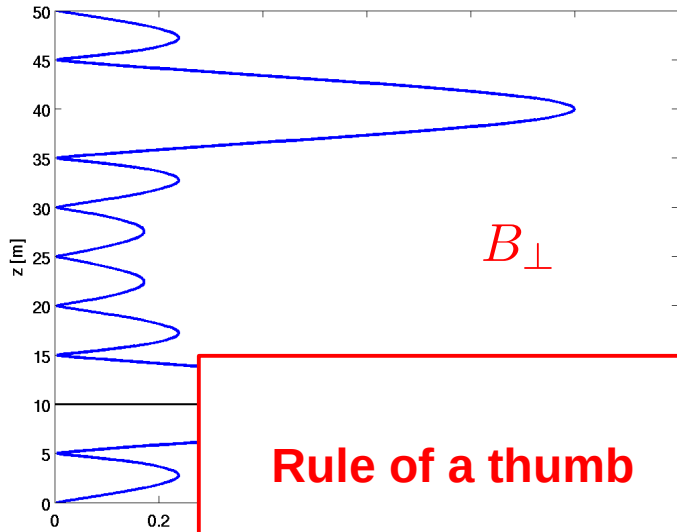
**M=12**

- Unchanged resolution
- Improved ambiguity



# Tomographic imaging using specan

**M=6**



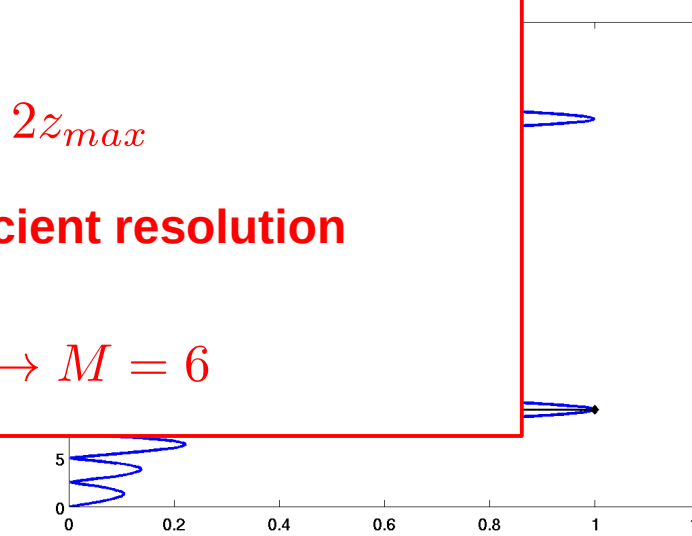
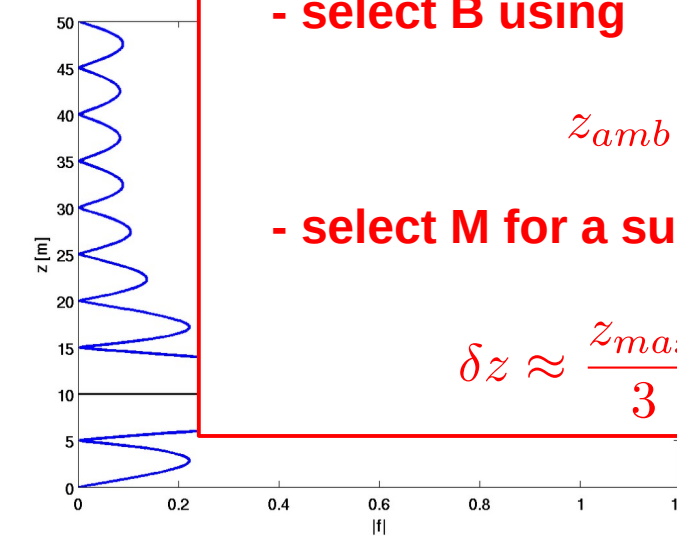
## Rule of a thumb

- select B using

$$z_{amb} \approx 2z_{max}$$

- select M for a sufficient resolution

$$\delta z \approx \frac{z_{max}}{3} \rightarrow M = 6$$



$$\Delta k_z \propto MB_{\perp}$$

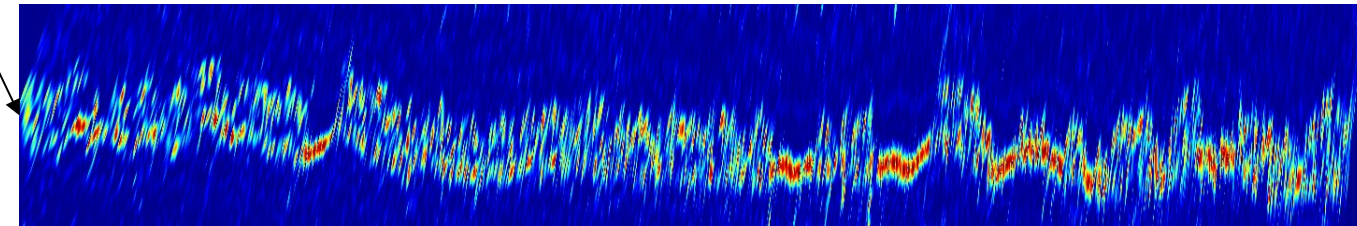
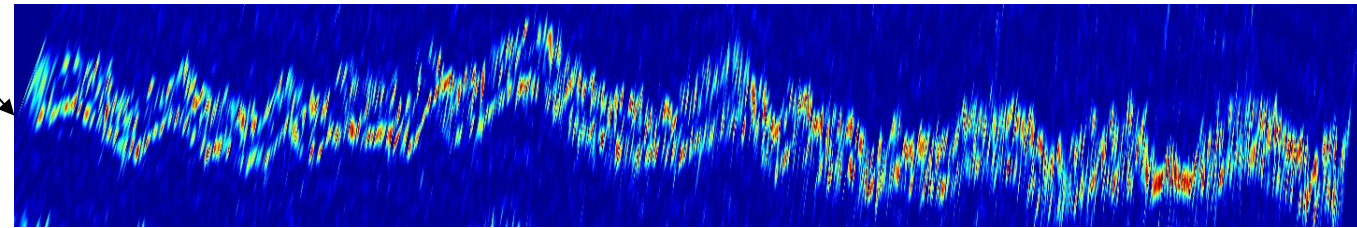
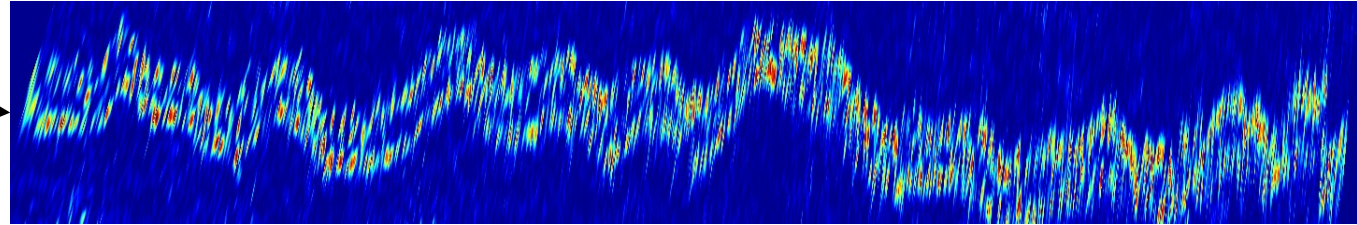
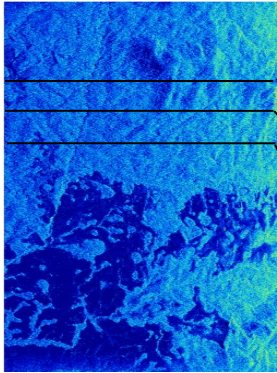
$$dk_z = \frac{\Delta k_z}{M}$$

**M=12**

# Tomographic imaging using specan

Single-look tomograms

$$\hat{I}(z) = \left| \frac{\mathbf{a}^H(z)\mathbf{y}}{M} \right|^2$$



Tomographic data from AfriSAR  
2016 (ESA)

Site: Gabon

Acquisition by DLR & ONERA

Noisy aspect due to speckle

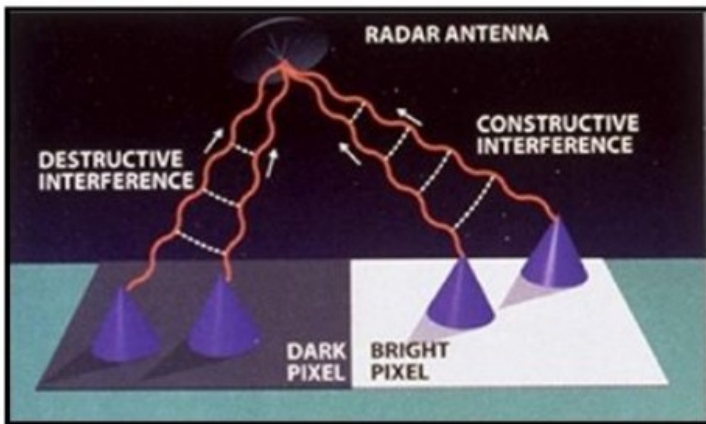
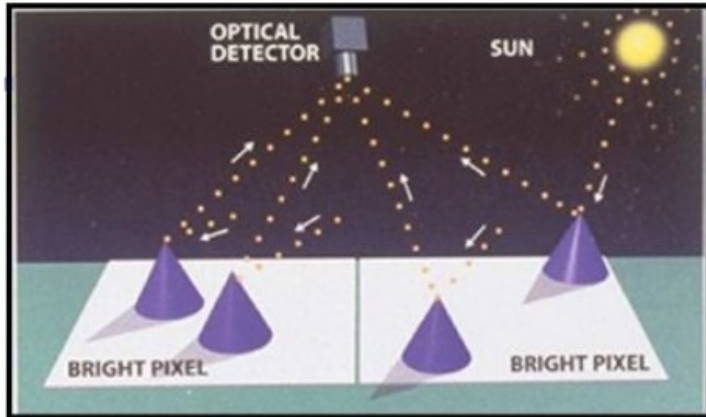
# *TomoSAR imaging*

## *Using multilook*

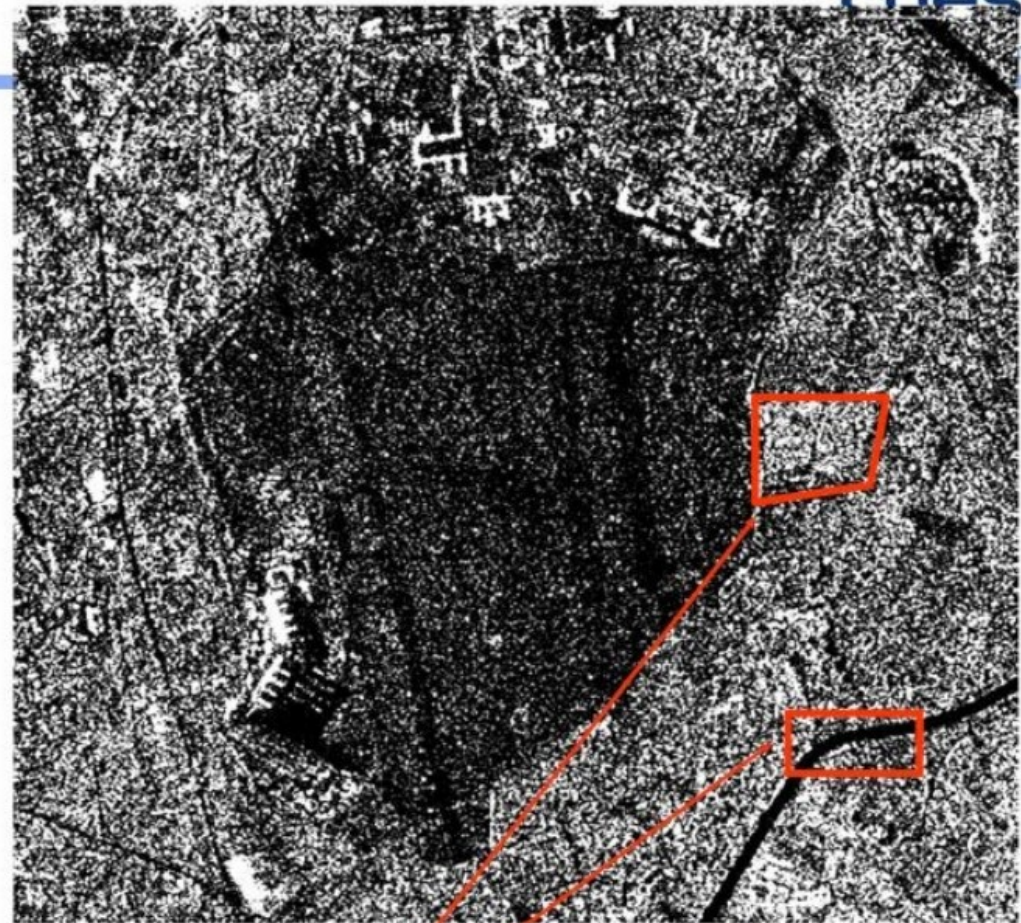
### *Specan methods*



# Speckle effect



© Scientific American



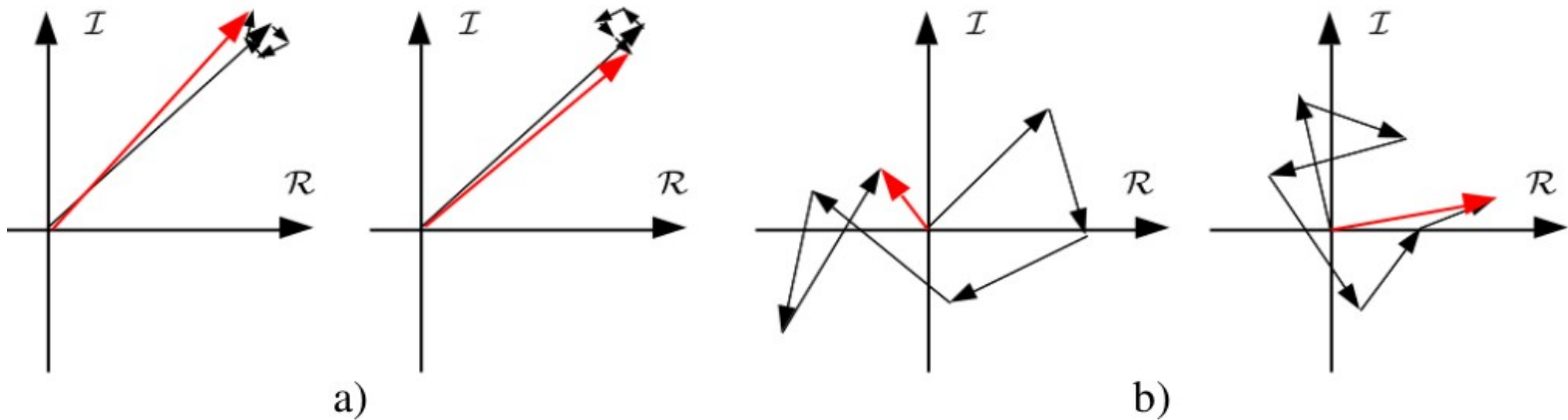
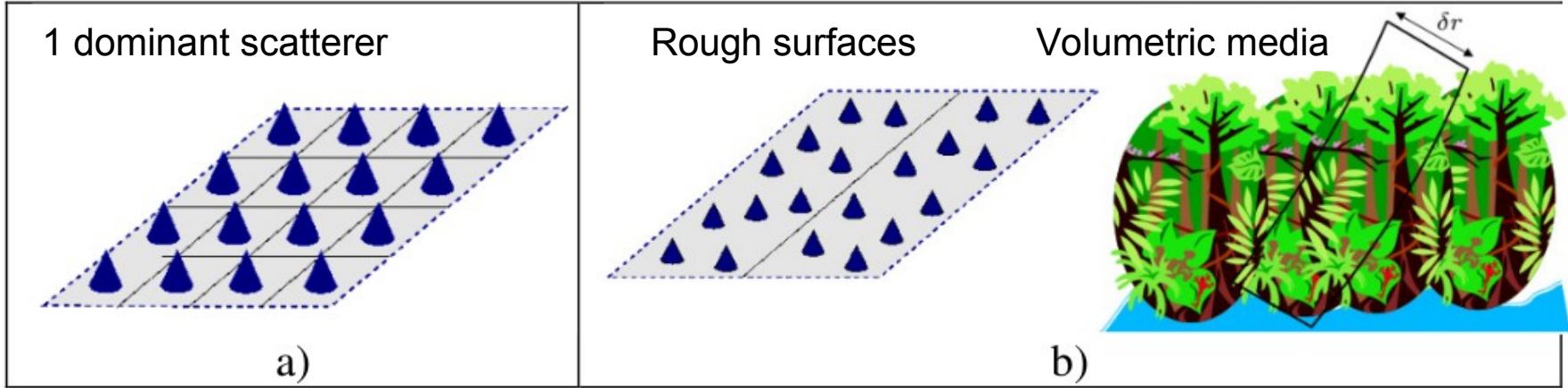
RSAT standard image

Speckle: **coherent effect** that appears as a **Multiplicative noise**



# Speckle effect

$$s(x, r) \approx \int_C a_c(x, r, \nu) e^{-jk_c r(\nu)} d\nu$$



Two realizations in both cases

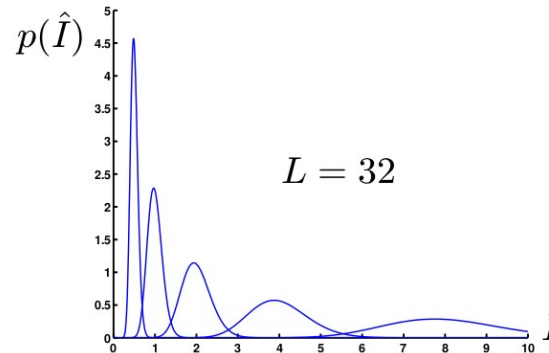
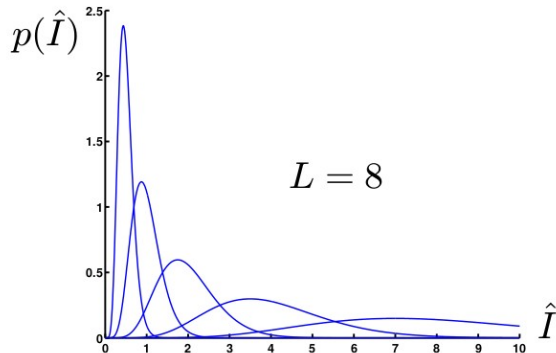
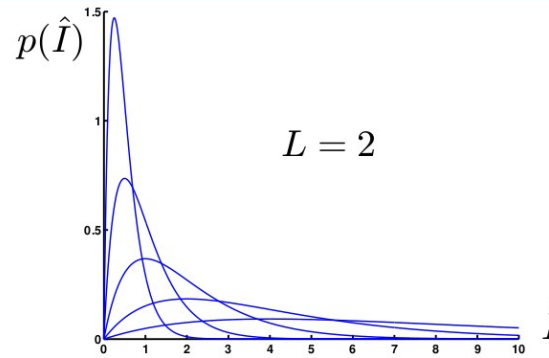
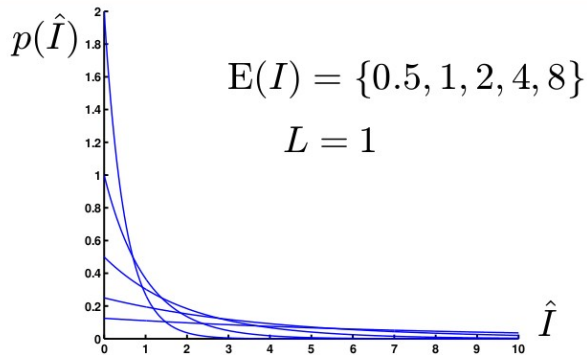
# Speckle filtering

Unfiltered intensity image: exponential distribution

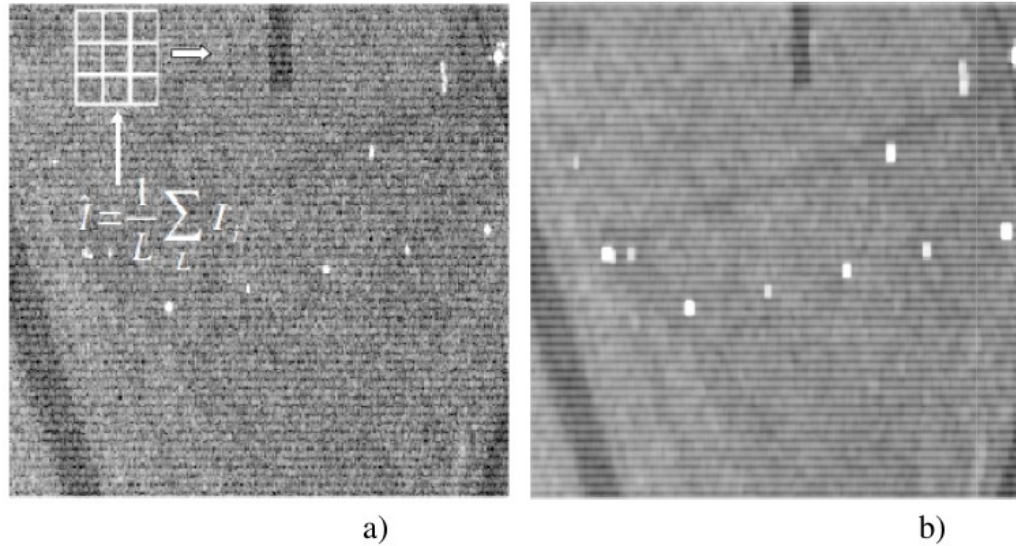
$$\hat{I} = |s(l)|^2, \quad E(\hat{I}) = I, \quad \text{var}(\hat{I}) = I^2$$

L independent samples (looks): ML estimate has chi2 distribution

$$\hat{I} = \frac{1}{L} \sum_{l=1}^L |s(l)|^2, \quad E(\hat{I}) = I, \quad \text{var}(\hat{I}) = \frac{I^2}{L}$$

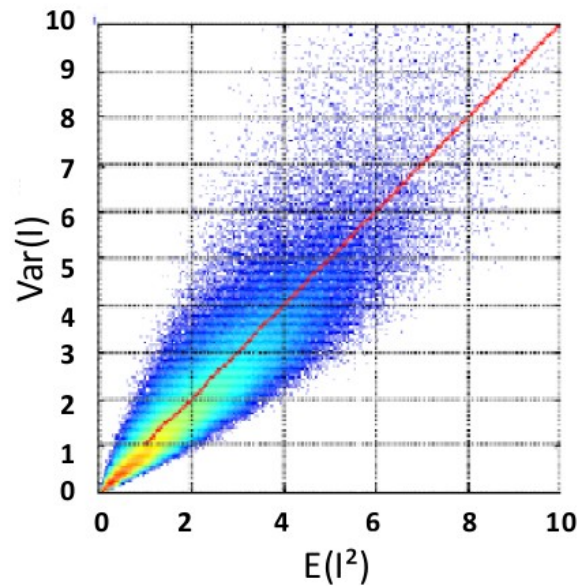


# Speckle filtering



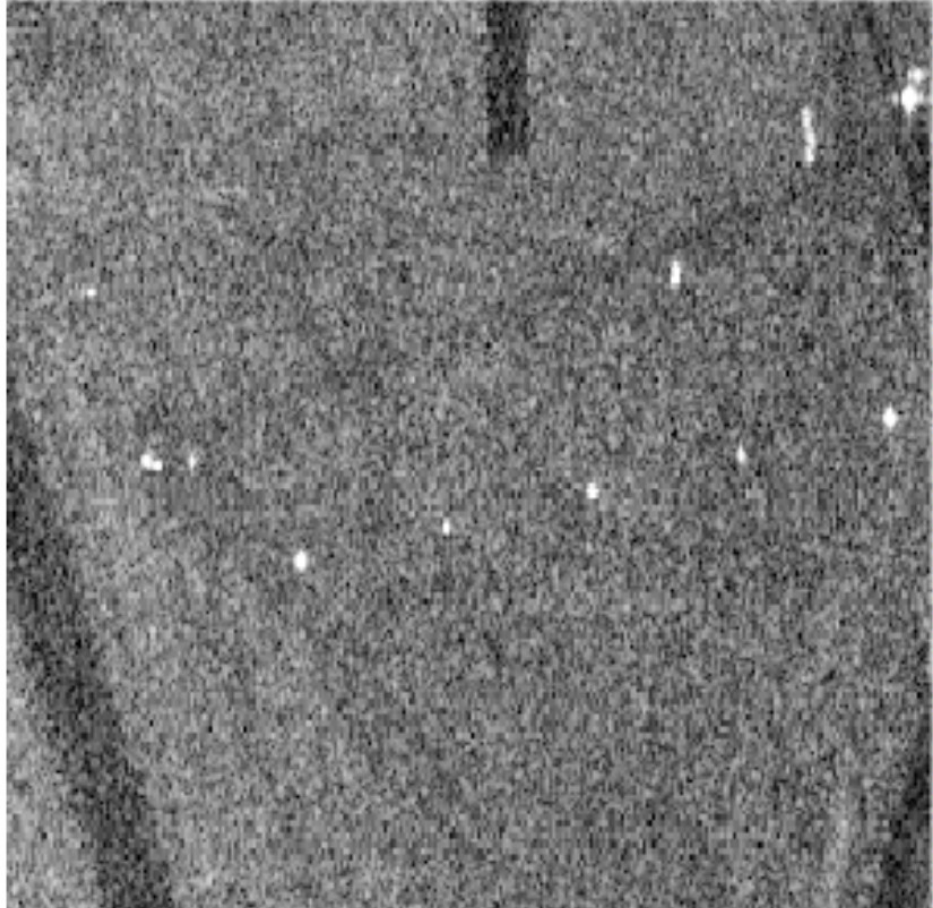
Equivalent Number of Looks

$$ENL = \frac{E(\hat{I})^2}{\text{var } \hat{I}}$$

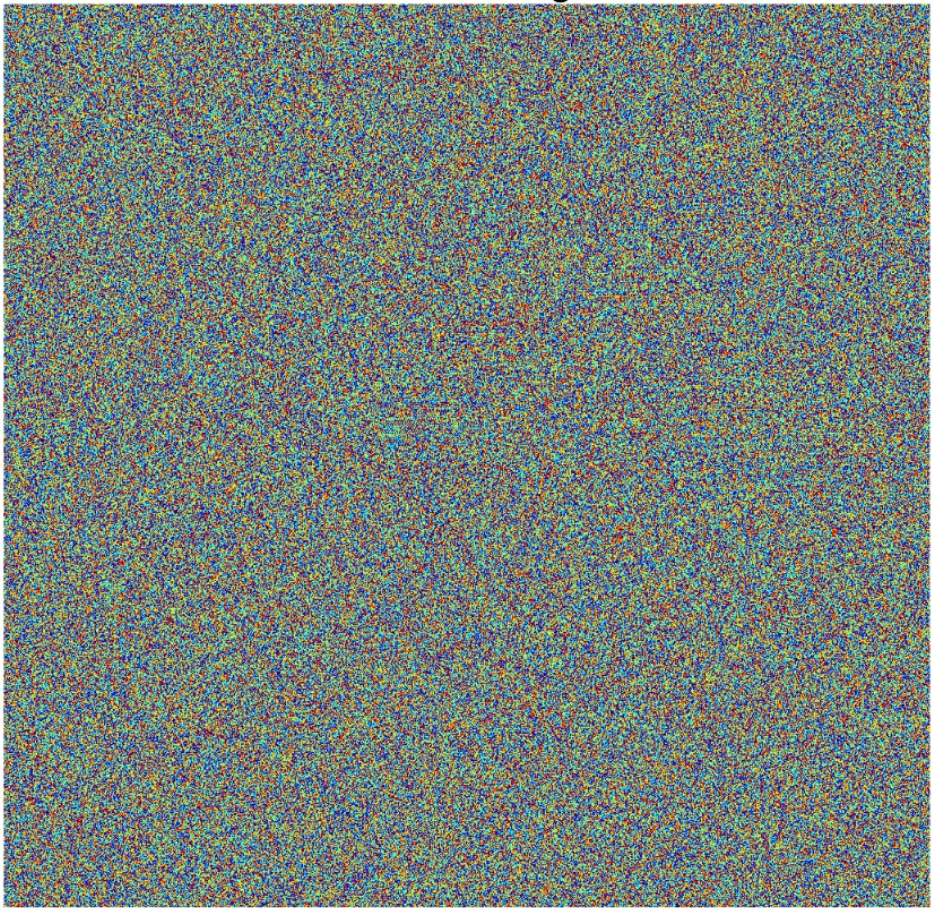


# Speckle filtering

Intensity image



Phase image

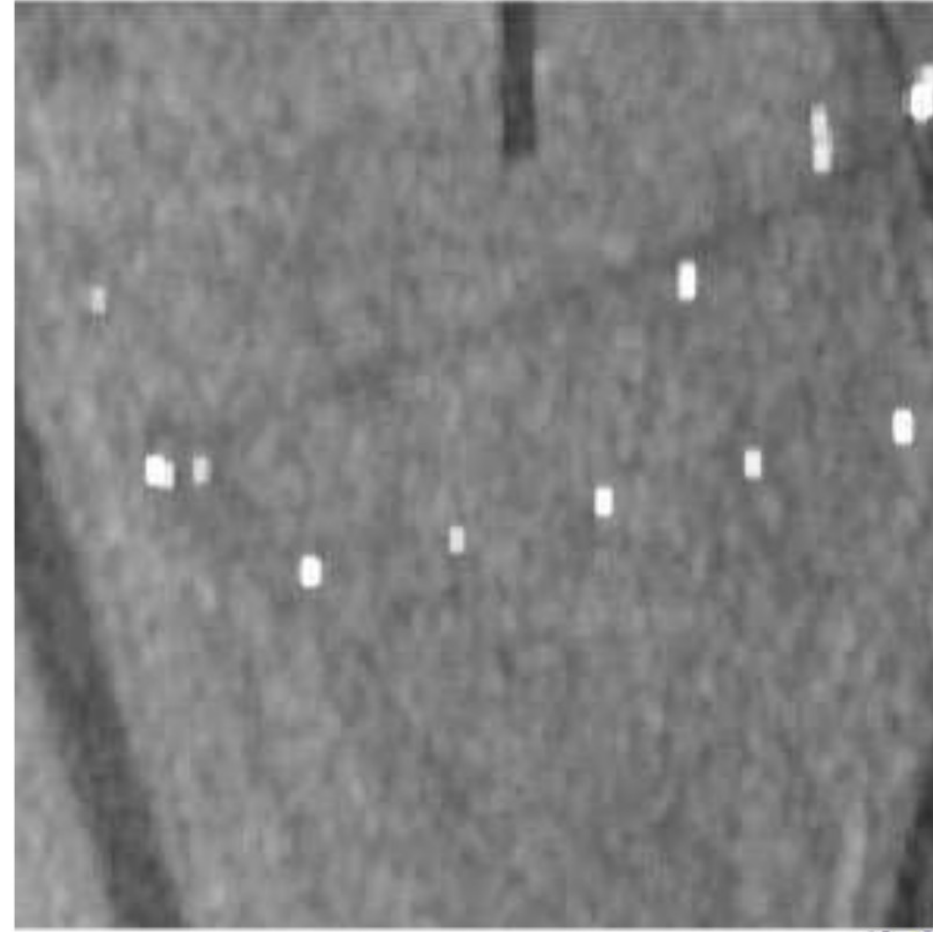
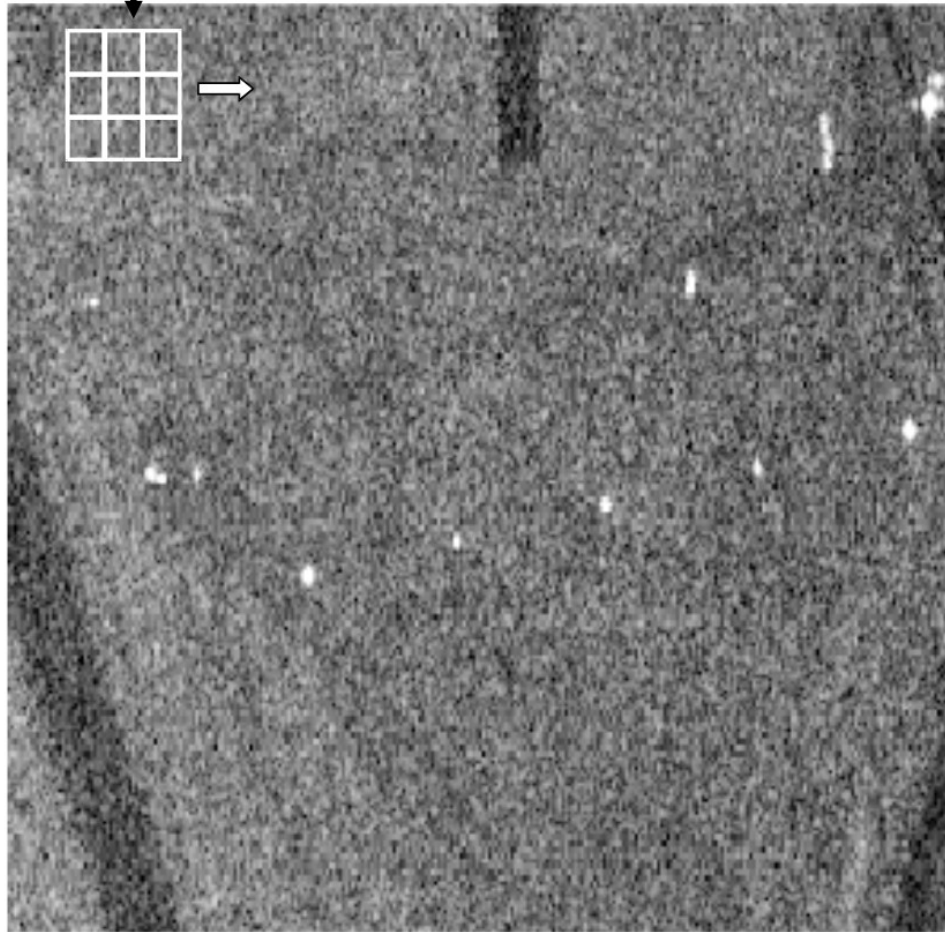


## Intensity images

$$\hat{I} = \frac{1}{L} \sum_L I_i$$

Single look image

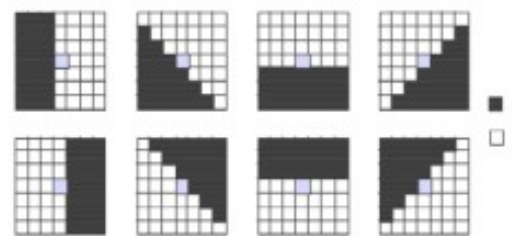
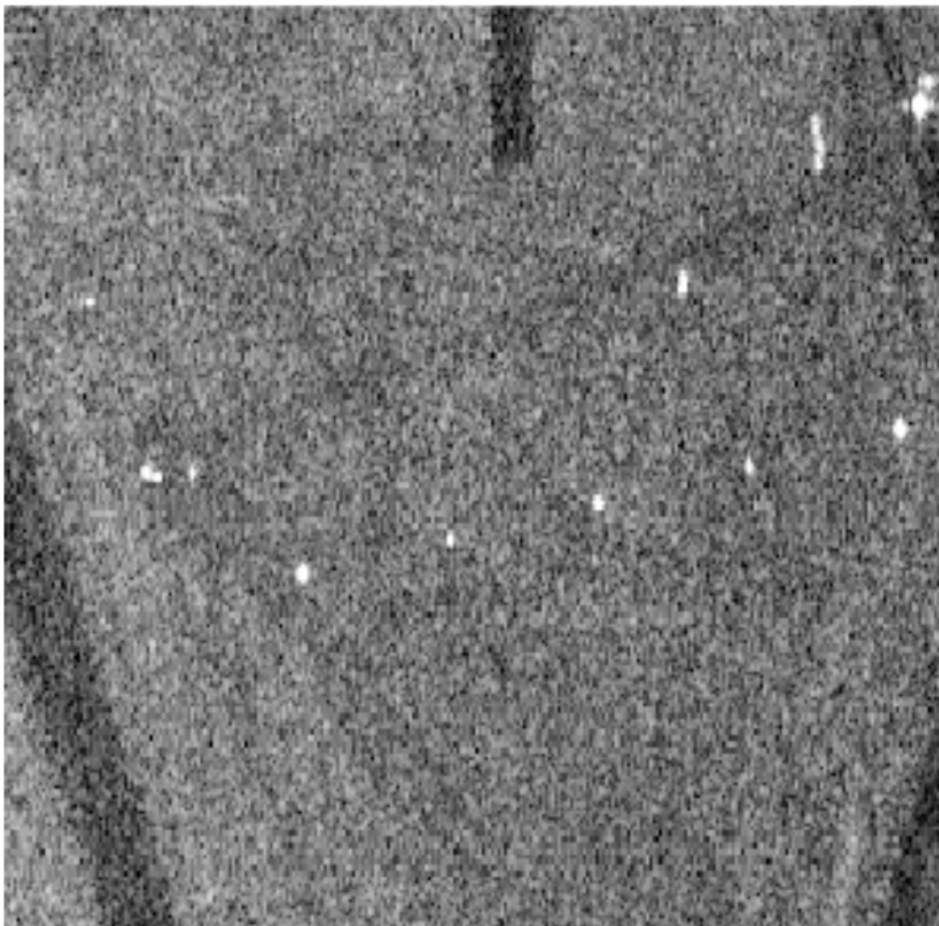
After spatial filtering (N\*N boxcar)



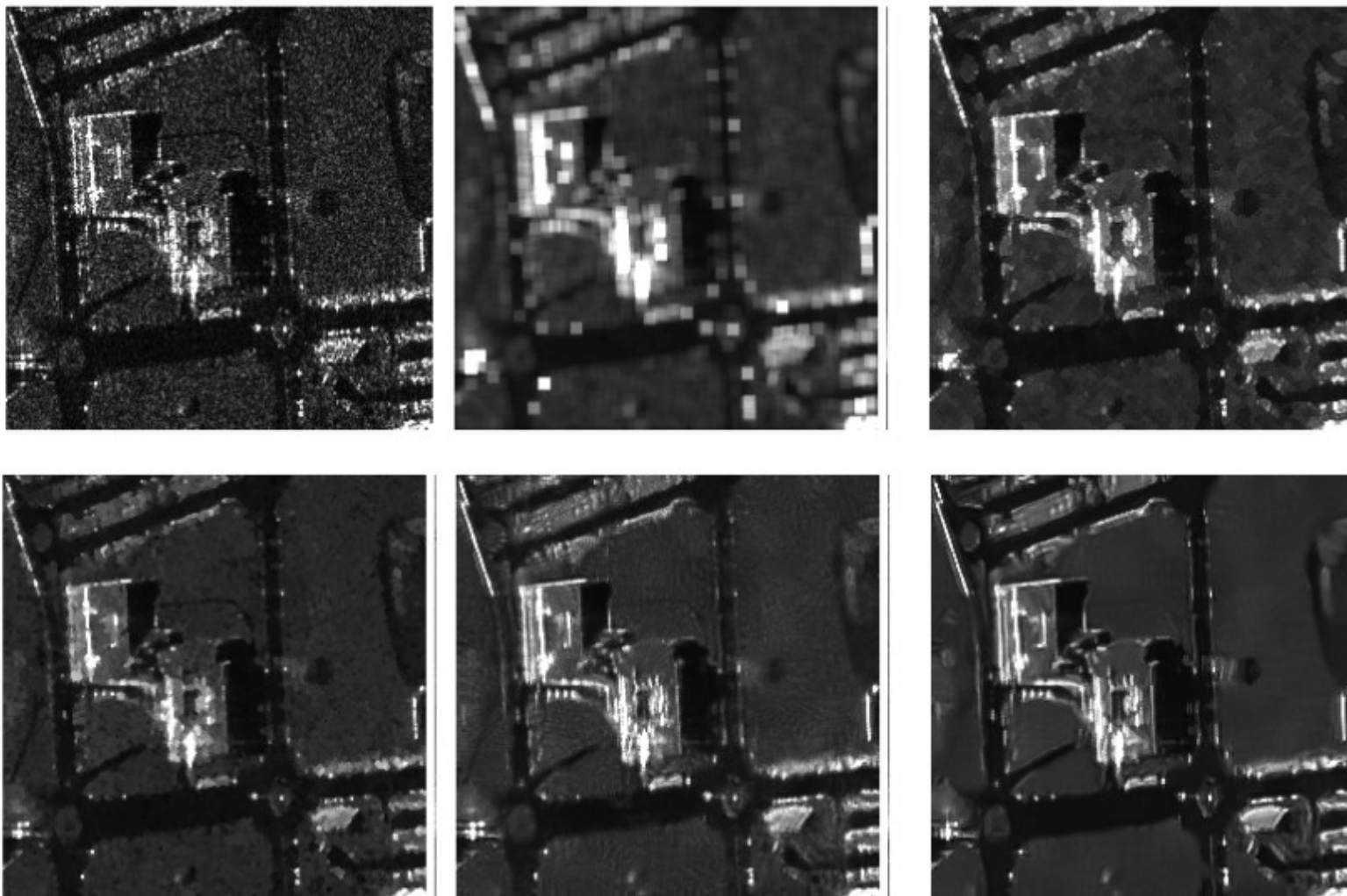
# Speckle filtering

Single look image

After spatial filtering (N\*N Lee filter)



# Speckle filtering



Non local speckle filtering

# Speckle filtering with tomographic data

## Speckle filtering for monivariate SLC SAR images

$$\hat{I} = \frac{1}{L} \sum_{l=1}^L |s(l)|^2, \quad \mathbb{E}(\hat{I}) = I, \quad \text{var}(\hat{I}) = \frac{I^2}{L}$$

## Speckle filtering for multivariate SLC MB-InSAR images

$$\mathbf{y} = \begin{bmatrix} s_1 \\ \vdots \\ s_M \end{bmatrix}, \quad \mathbf{a}(z) = \begin{bmatrix} 1 \\ \vdots \\ e^{jk_{z_M} z} \end{bmatrix} \quad f(z) = \frac{\mathbf{a}^H(z)\mathbf{y}}{M} = \frac{1}{M} \sum_m s_m e^{-jk_{z_m} z}$$

$$\hat{I}(z) = \frac{1}{L} \sum_{l=1}^L |f(z, l)|^2 = \frac{1}{M^2} \mathbf{a}^H(z) \hat{\mathbf{R}} \mathbf{a}(z)$$

## L-look (ML) estimate of the TomoSAR covariance matrix

$$\hat{\mathbf{R}} = \frac{1}{L} \sum_{l=1}^L \mathbf{y}(l)\mathbf{y}^H(l) \quad \mathbb{E}(\hat{\mathbf{R}}) = \mathbf{R}$$



# Speckle filtering with tomographic data

## TomoSAR covariance matrix

$$\hat{\mathbf{R}} = \frac{1}{L} \sum_{l=1}^L \mathbf{y}(l) \mathbf{y}^H(l) \quad \mathbf{E}(\hat{\mathbf{R}}) = \mathbf{R}$$

$$\hat{\mathbf{R}} = \frac{1}{L} \sum_{l=1}^L \mathbf{y}(l) \mathbf{y}^H(l) = \begin{bmatrix} R_{11} & R_{12} & \dots & R_{1M} \\ R_{12}^* & R_{22} & \dots & R_{2M} \\ & & \ddots & \\ R_{1M}^* & R_{2M}^* & \dots & R_{MM} \end{bmatrix}$$

$$\hat{R}_{ii} = \frac{1}{L} \sum_{l=1}^L y_i(l) y_i^*(l) = \hat{I}_i$$

$$\hat{R}_{ij} = \frac{1}{L} \sum_{l=1}^L y_i(l) y_j^*(l) = \sqrt{\hat{I}_i \hat{I}_j} \hat{\gamma}_{ij}$$

## Interferometric coherence estimate

$$\hat{\gamma}_{ij} = \frac{\hat{R}_{ij}}{\sqrt{\hat{I}_i \hat{I}_j}}$$

$$\hat{\phi}_{ij} = \arg(\hat{\gamma}_{ij})$$

$$|\hat{\gamma}_{ij}| \leq 1$$

# Tomographic imaging using specan

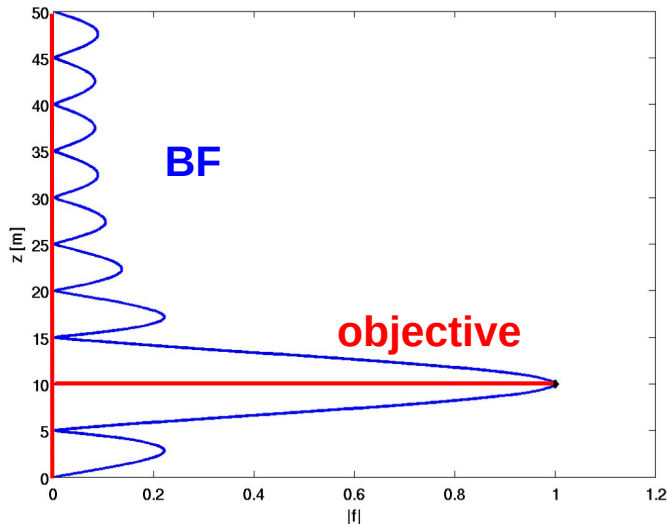
## Beamformer is Fourier imaging

$$\hat{I}_{BF}(z) = \frac{1}{M^2} \mathbf{a}^H(z) \hat{\mathbf{R}} \mathbf{a}(z)$$

- Excellent (optimal) statistical accuracy
- Fourier resolution:  $\delta z = \frac{2\pi}{\Delta k}$
- Cannot handle closely spaced scatterers
- High sidelobes

## Capon's solution: constrained beamformer

Objective: minimize output power, with unitary gain at the height of interest

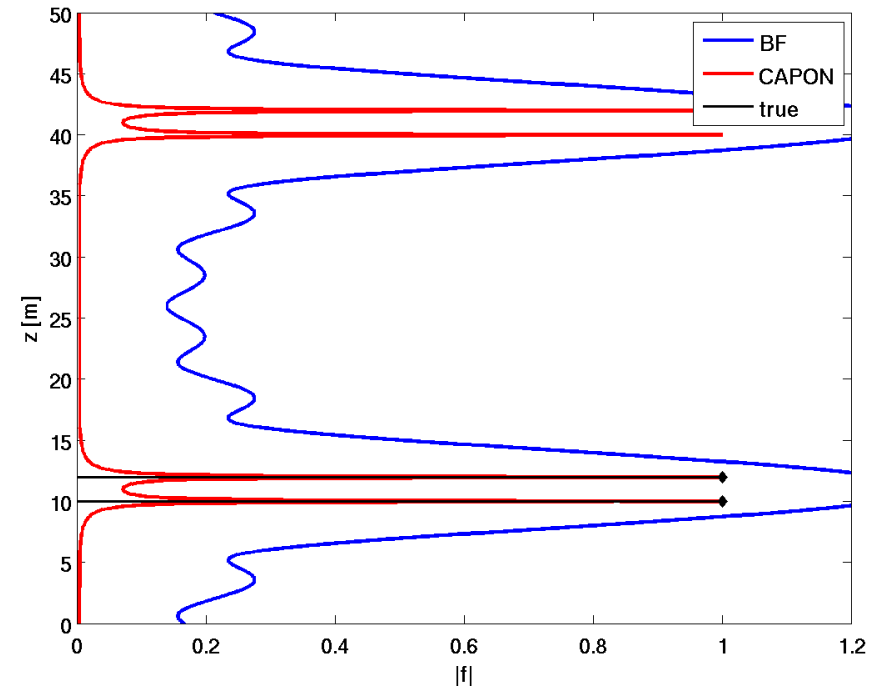
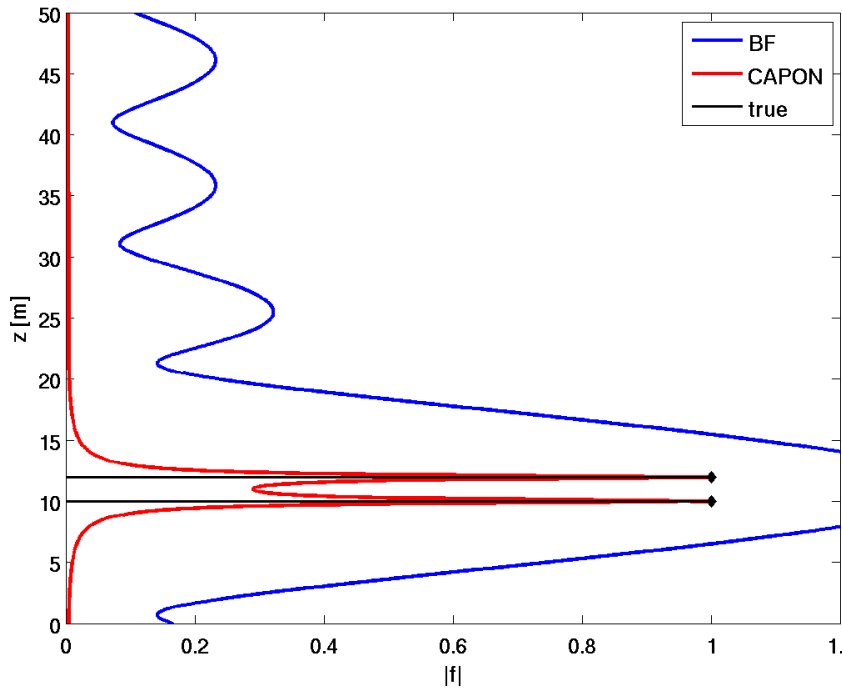


$$\mathbf{v}_{CP}(z) = \arg \min_{\mathbf{v}} E(|\mathbf{v}^H \mathbf{y}|^2) \quad \text{s.t.} \quad \mathbf{v}^H \mathbf{a}(z) = 1$$

Solution:

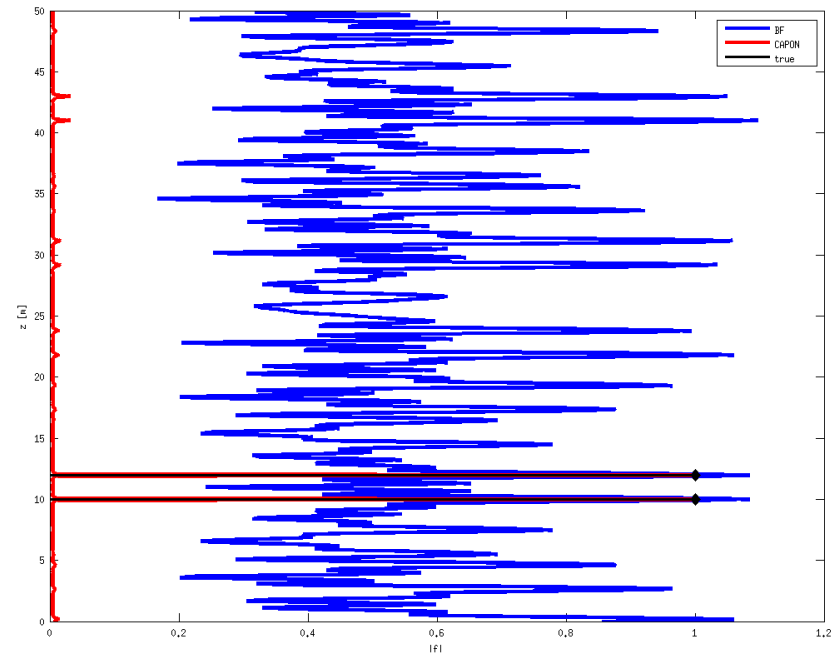
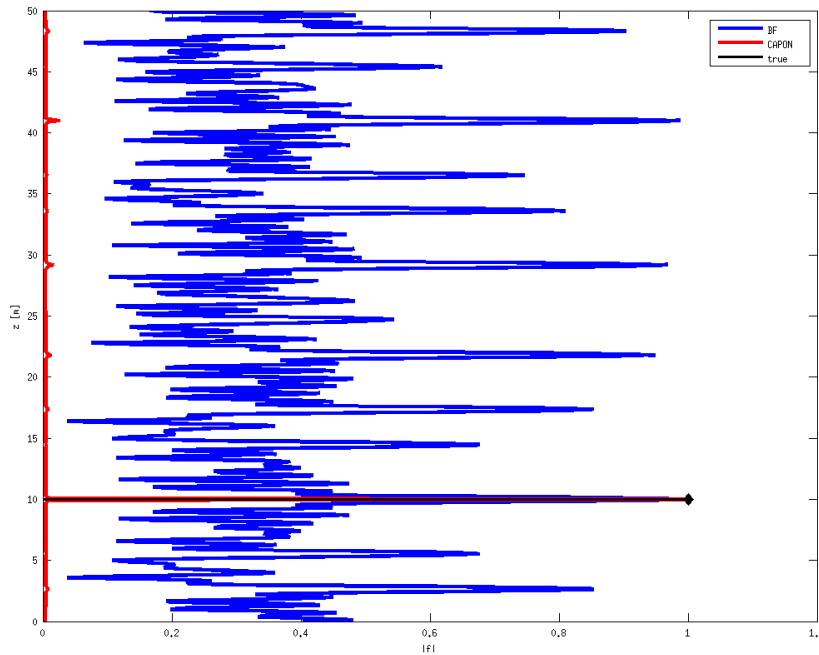
$$\hat{I}_{CP} = \frac{1}{\mathbf{a}^H(z) \hat{\mathbf{R}}^{-1} \mathbf{a}(z)}$$

# Tomographic imaging using specan



- Capon: significantly improved resolution
- Resolution improvement is a function of the Signal to Noise Ratio (SNR)
- For regular baselines, BF & Capon are equally affected by ambiguities

## Irregular baseline sampling: logscale distribution



- BF: strongly affected by ambiguities
- CAPON: asynchronous ambiguities are considered as perturbations and filtered (may be dangerous!). Good resolution performance preserved

# Tomographic imaging using specan

## Practical implementation

- Asymptotic ( $L \rightarrow +\infty$ ) estimators

$$I_{BF}(z) = \frac{\mathbf{a}^H(z) \mathbf{R} \mathbf{a}(z)}{M^2}$$

$$I_{CP}(z) = \frac{1}{\mathbf{a}^H(z) \mathbf{R}^{-1} \mathbf{a}(z)}$$

- In practice, spatial averaging

$$\mathbf{R} \rightarrow \hat{\mathbf{R}} = \frac{1}{L} \sum_{l=1}^L \mathbf{y}(l) \mathbf{y}^H(l)$$

- BF: quite stable w.r.t L

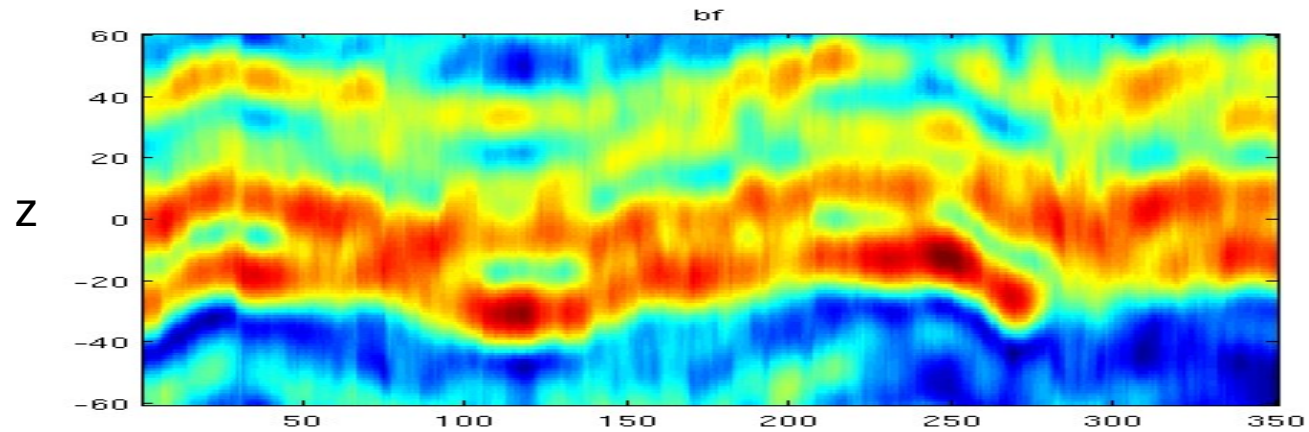
- Capon may suffer from a poor covariance matrix conditioning: sufficient ENL needed

$$\Rightarrow \text{Diagonal Loading} \quad \tilde{\mathbf{R}} = \hat{\mathbf{R}} + \alpha \mathbf{I}_M, \quad \alpha \geq 0$$

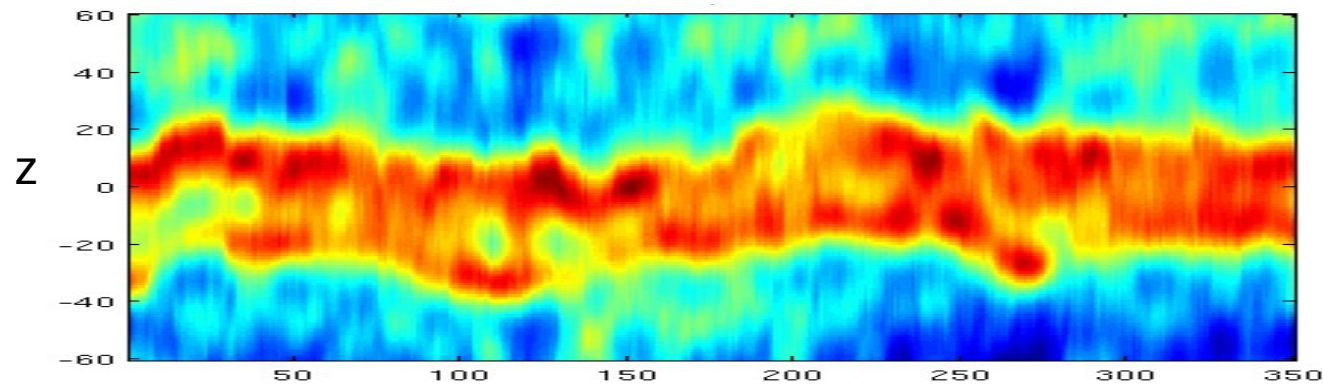
For large  $\alpha$  (low SNR): CP  $\rightarrow$  BF

# Tomographic imaging using specan

Tropical forest profile at P band with residual phase errors



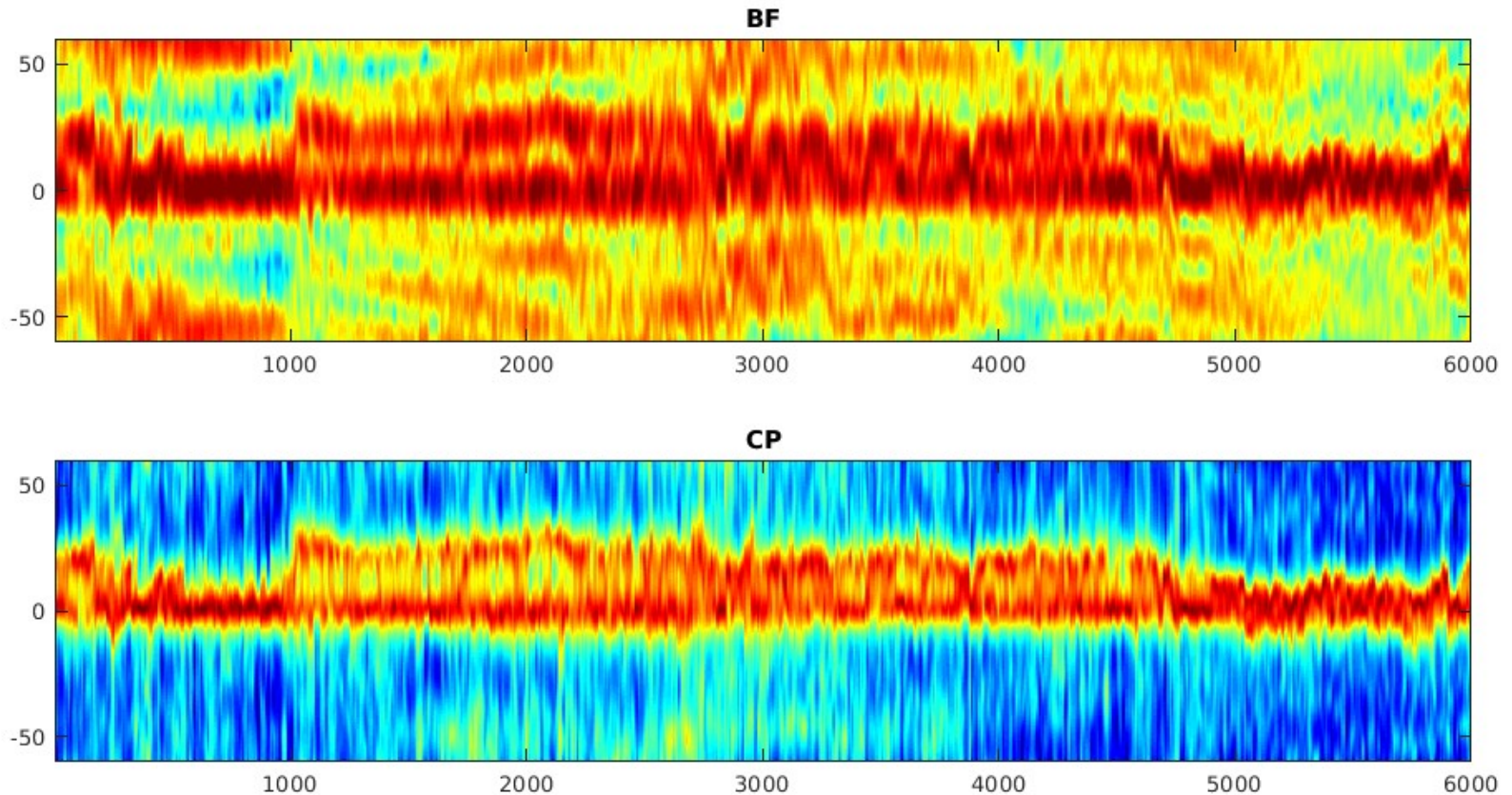
BF



Capon

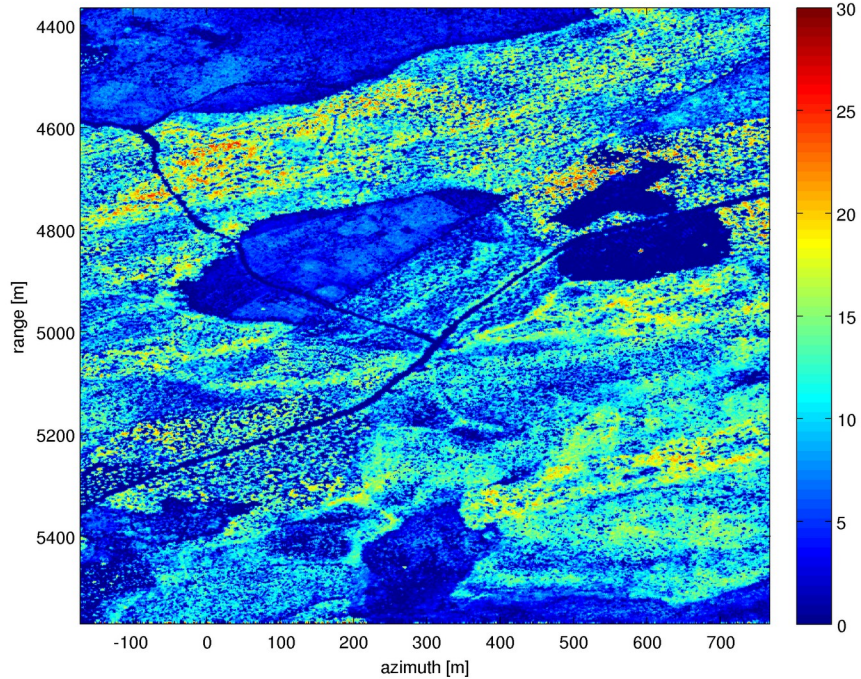
# Tomographic imaging using specan

P band tomogram (Tomosense campaign) with residual phase errors

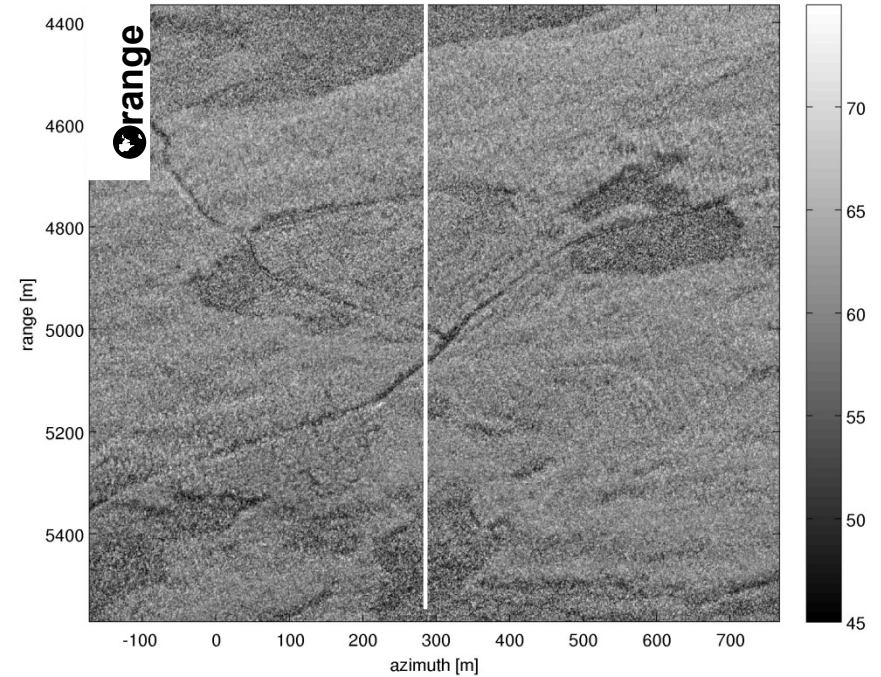


# Case study: BIOSAR 2 data

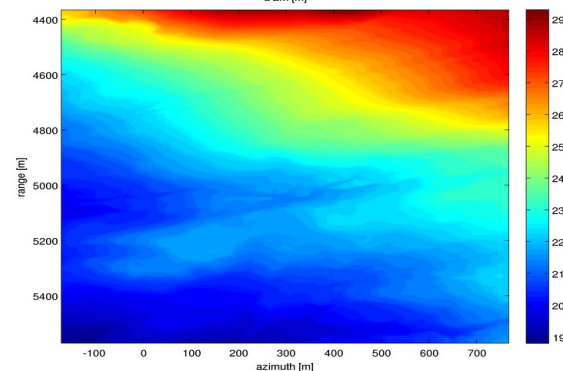
## Forest height



## HH intensity



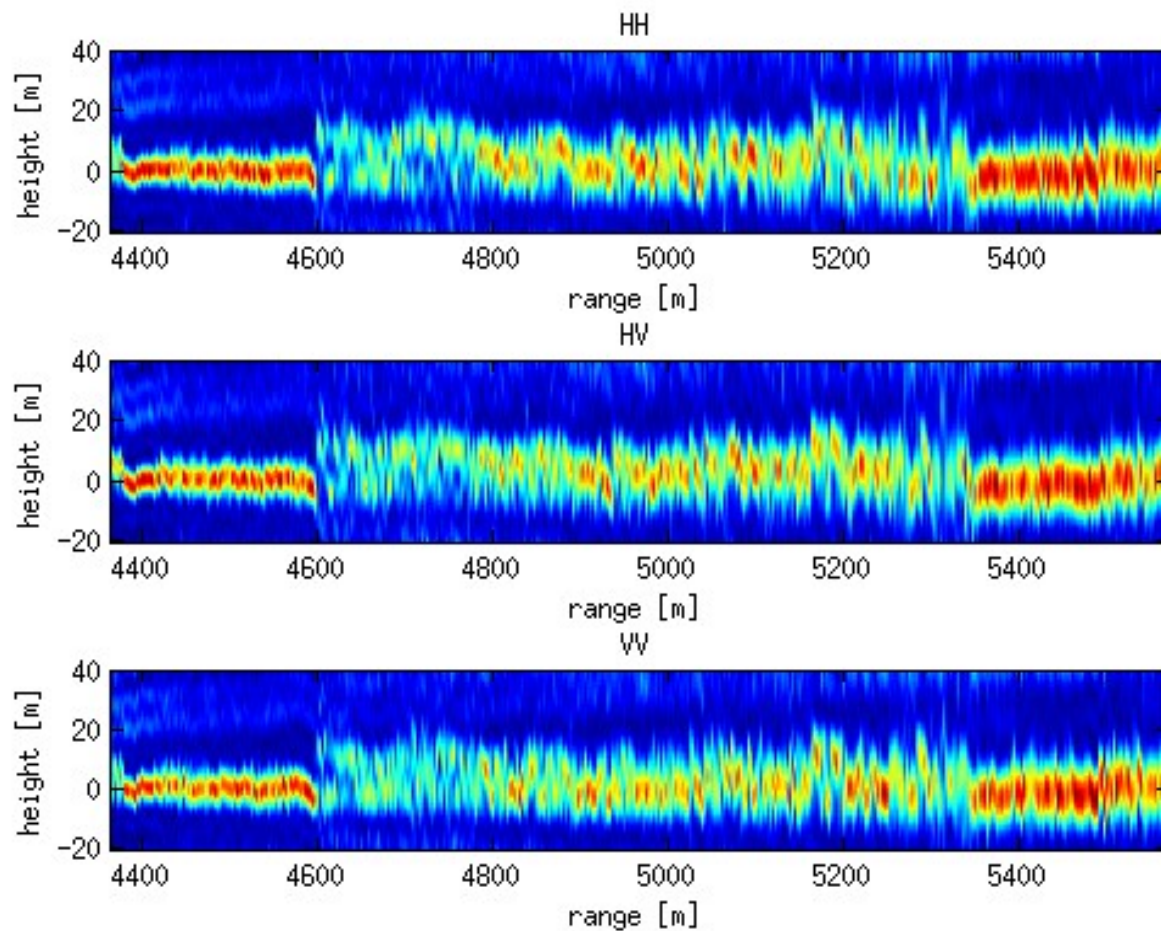
## DEM



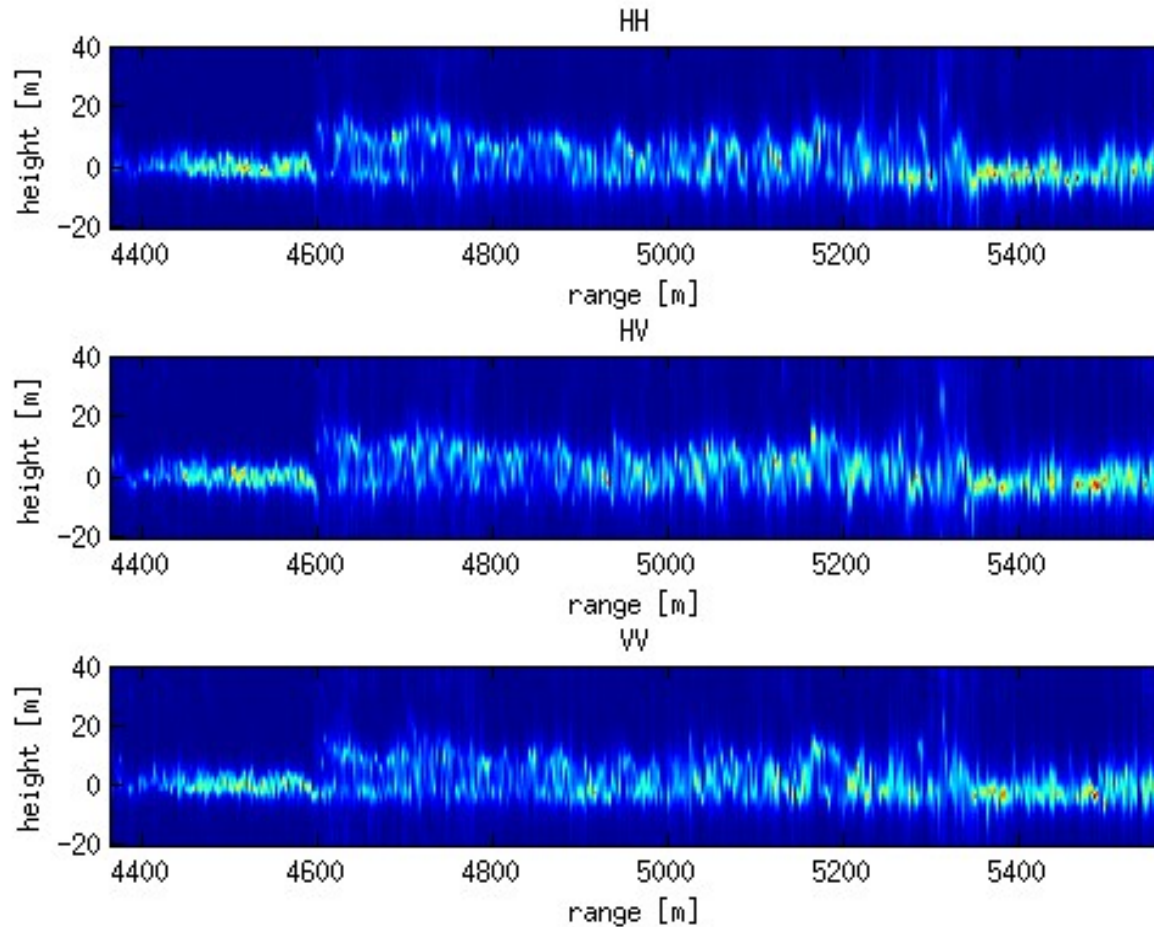


# Case study: BIOSAR 2 data

## BF



## CAPON: processing OK ?



# *Advanced TomoSAR imaging Using Specan methods*



# 3-D imaging of an urban area using a minimal configuration

## Urban area test site

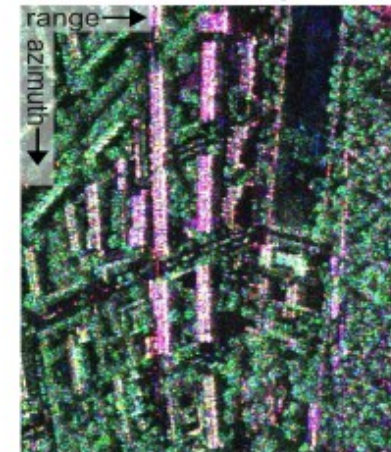
- Images over Dresden, 2000
- DLR's E-SAR at L-Band
- Resolution : 0.5 m × 2.5 m
- Fully polarimetric
- Dual-baseline InSAR

Baselines	$H_{am}$
10 m	55-73 m
40 m	14-18 m

**3 PolSAR images**



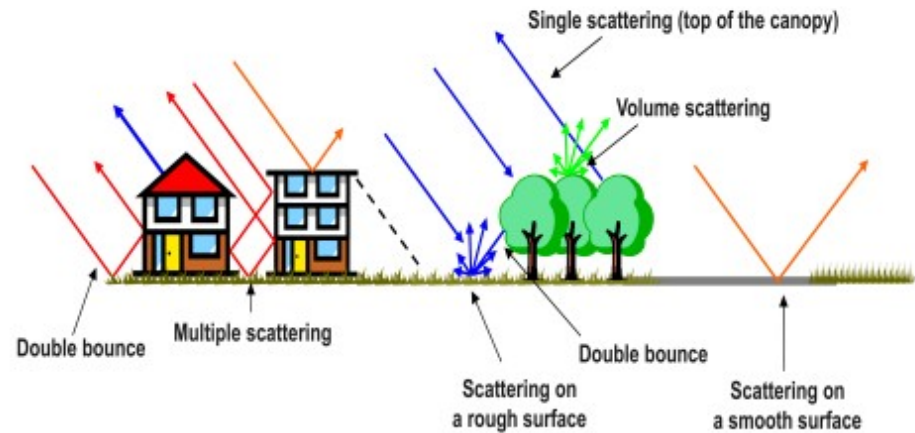
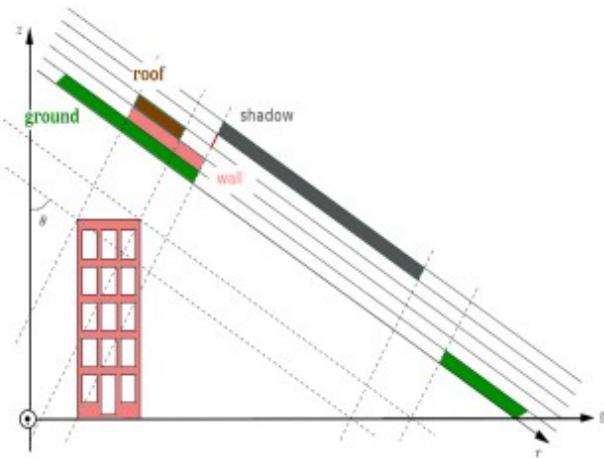
Pauli-coded SAR image



Optical image



# SAR tomography over urban areas



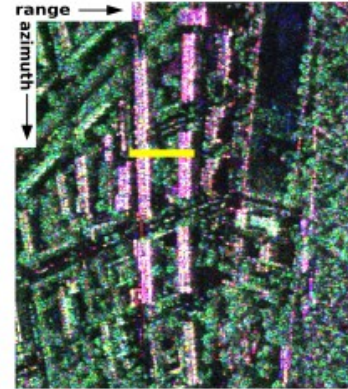
- L-band intermediate-resolution data sets  
⇒ High-Resolution (HR) tomographic estimators
- 3 images  
⇒  $N_s = 2$

# Tomographic imaging using specan

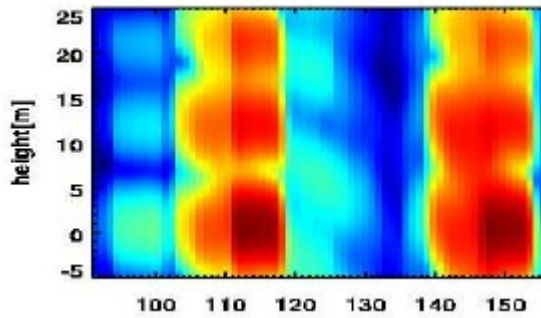
Critical configuration (**3 images**) in an urban environment at L band



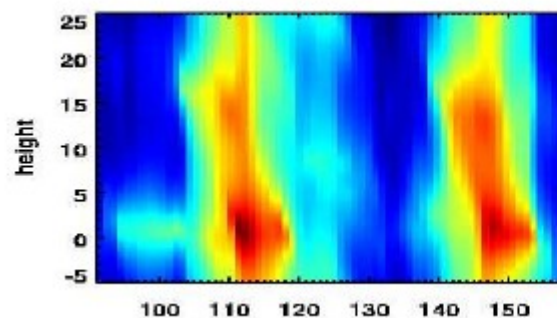
(a) Optical image



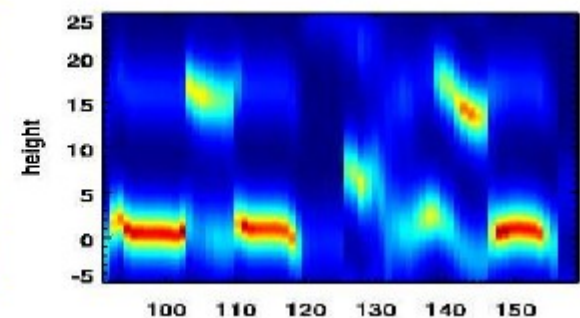
(b) SAR



BF



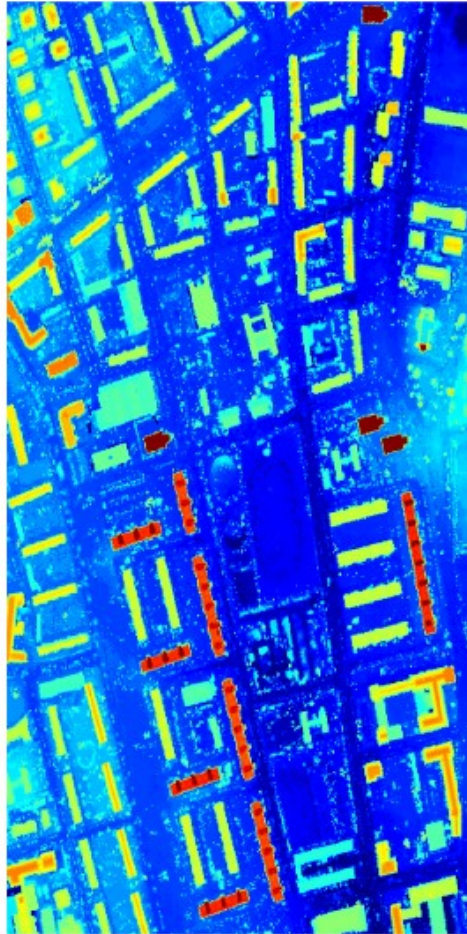
CAPON



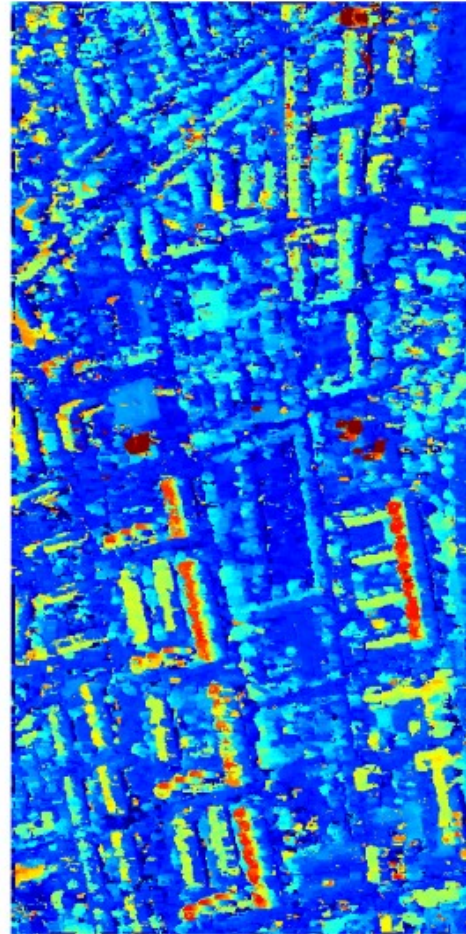
MUSIC

- Strictly speaking, Capon's technique is not HR, but is very convenient
- MUSIC (and some other techniques) is HR

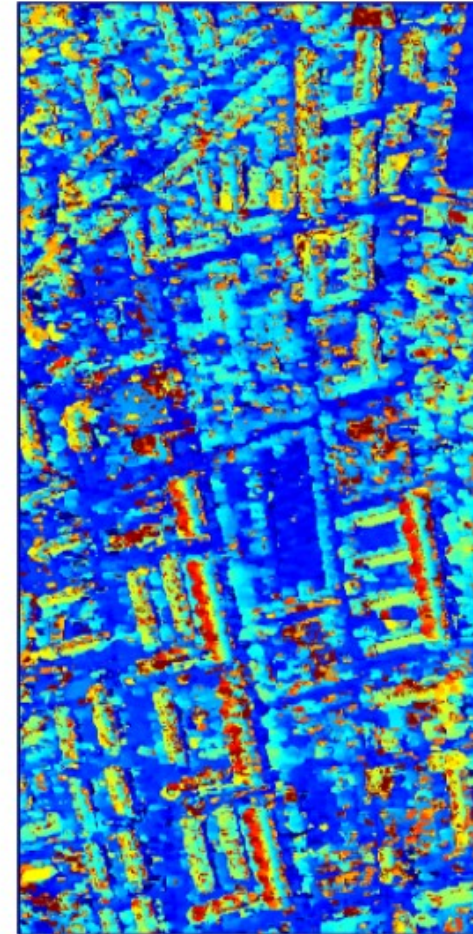
# Polarimetric SAR tomography over urban areas



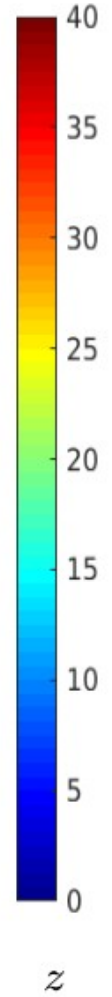
LiDAR



P-SSF

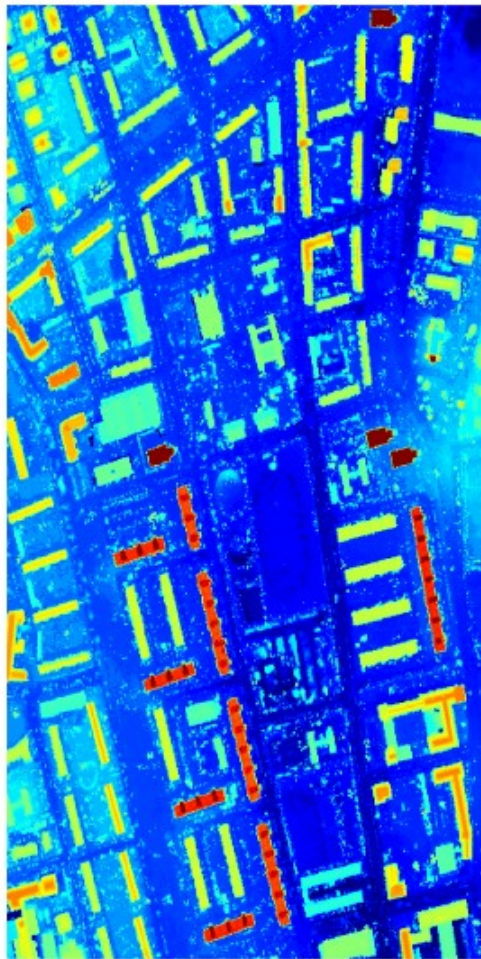


VV SSF

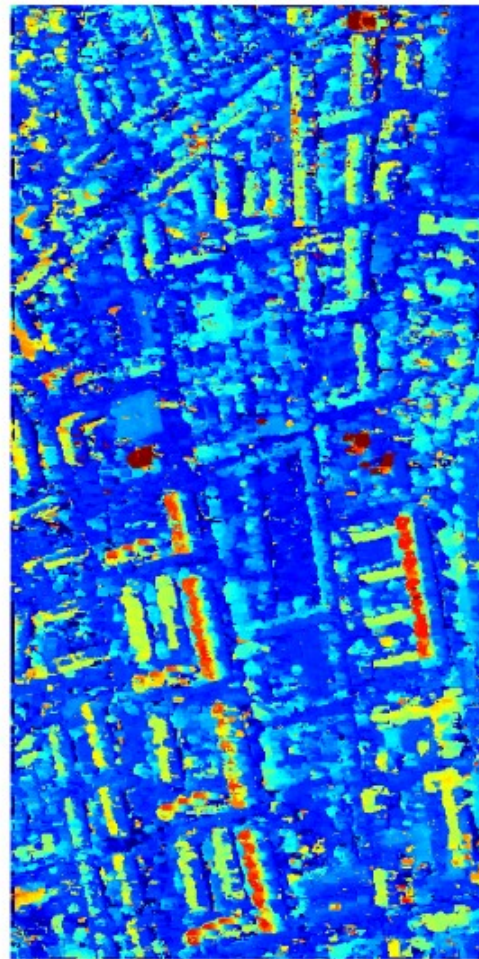


z

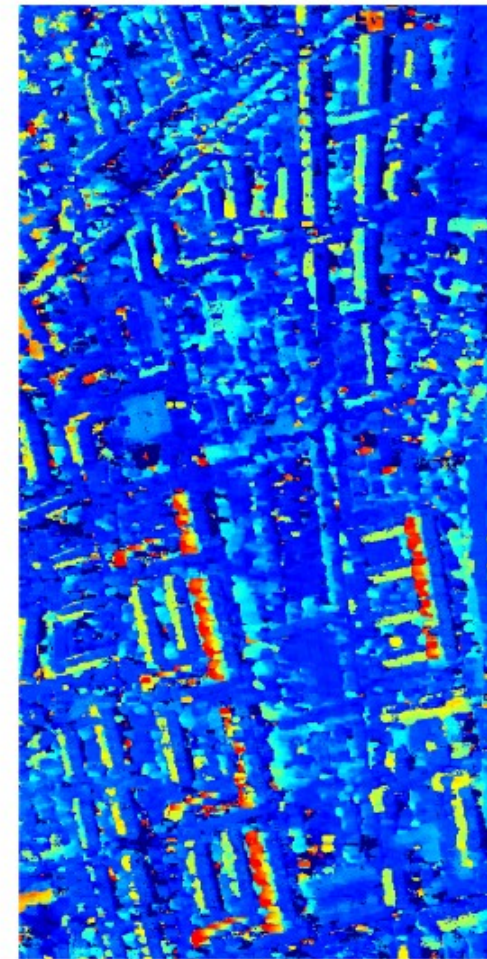
# Polarimetric SAR tomography over urban areas



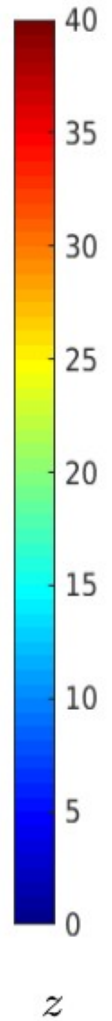
LiDAR



P-SSF



P-NSF



z



# Polarimetric SAR tomography over urban areas

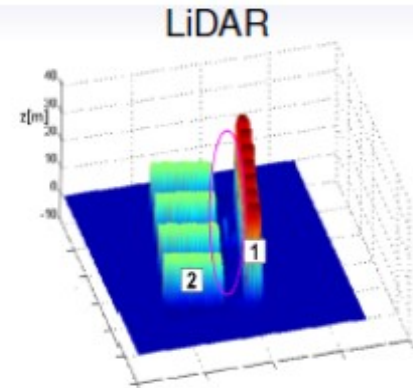
## Building reconstruction



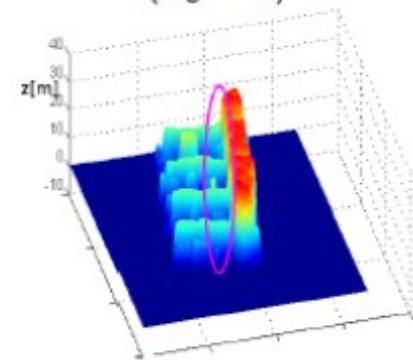
Google map



Pauli-coded



Estimated by FP-NSF  
( $N_s = 2$ )



### Difference between LiDAR and estimated surface

- projection of SAR imaging
- vegetation between B1 and B2

# Polarimetric SAR tomography over urban areas

## Building reconstruction

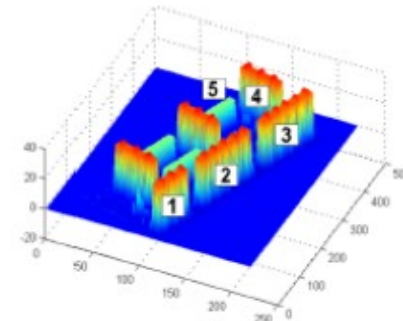


Google map

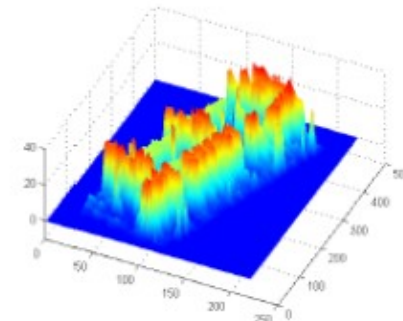


Bing map

LIDAR



Estimated by FP-NSF  
( $N_s = 2$ )



Averaged z[m]	B1	B2	B3	B4	B5
LIDAR	30.0	30.2	30.1	30.8	16.3
Estimated	27.5	27.8	27.5	27.3	16.1



# Tropical forest characterization

## Tropical forest test site and objectives

- TropiSAR Campaign, 2009
- ONERA SETHI
- P-Band
- 6 tracks
- $\delta_{az} = 1.245m$   
 $\delta_{rg} = 1m$
- $\delta_z = 12.5m$
- Ground truth
  - LiDAR data
  - Biomass measurements for 16 ROIs



Courtesy ONERA

### The ECOFOG Sites

- Nouragues
- Paracou
- ArboceI

### The Calibration site

- Rochambeau

### Other site

- Marais de Kaw

## Objectives

- Tree height, underlying ground topography estimation
- Forest vertical structure characterization
- Biomass monitoring

# Tropical forest characterization

## Tropical forest test site : Paracou

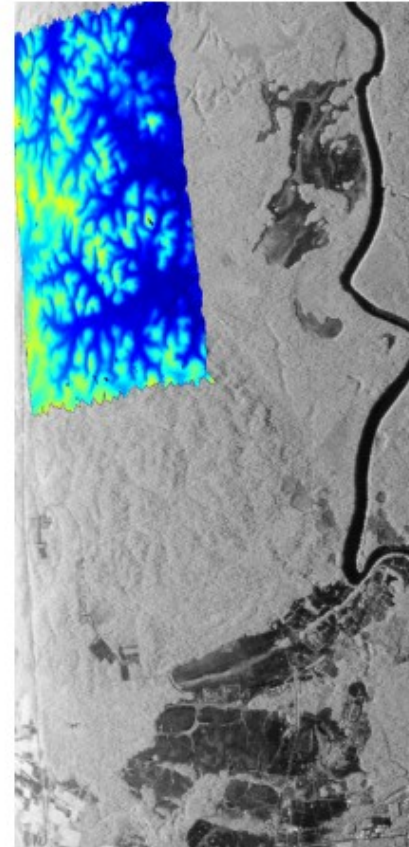
Optical image



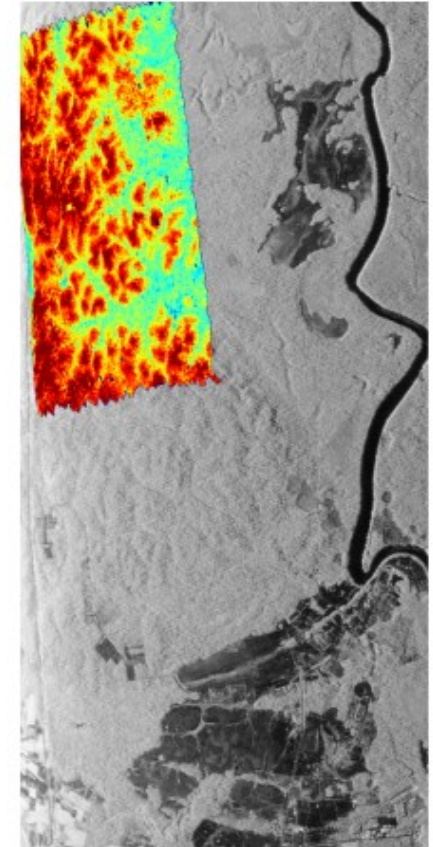
SAR image



LiDAR  $z_g$



LiDAR  $z_{top}$



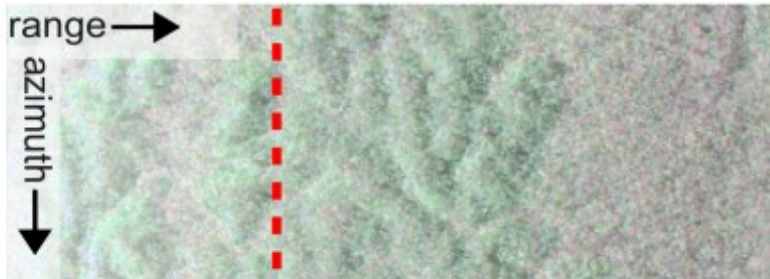
- Tropical forest environments (savannah, undisturbed forests, logged plots...)

- Highly varying ground topography

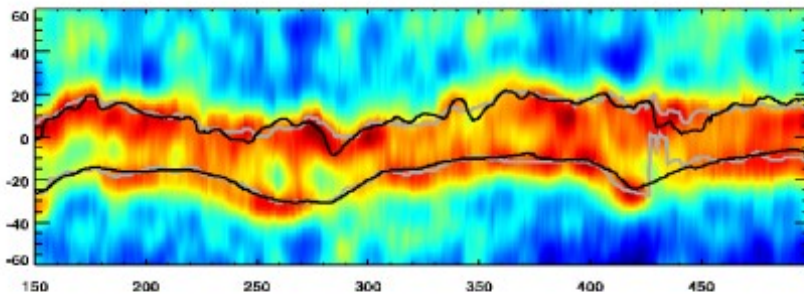
# Tropical forest characterization

## Tree height and ground topography estimation

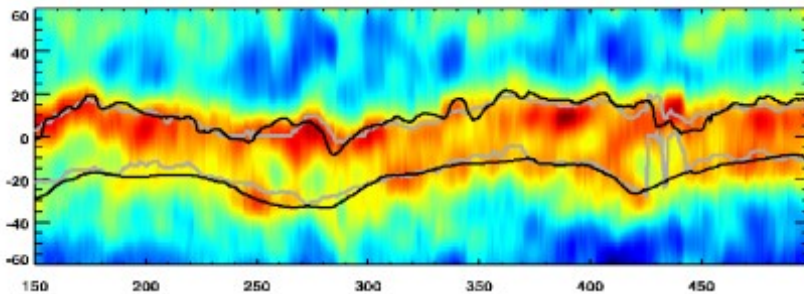
### Hybrid spectral approach



HH



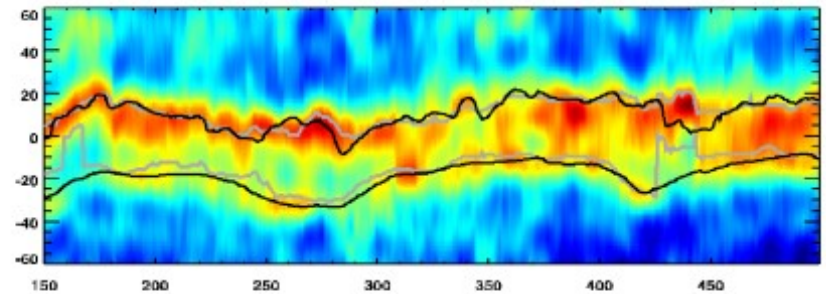
HV



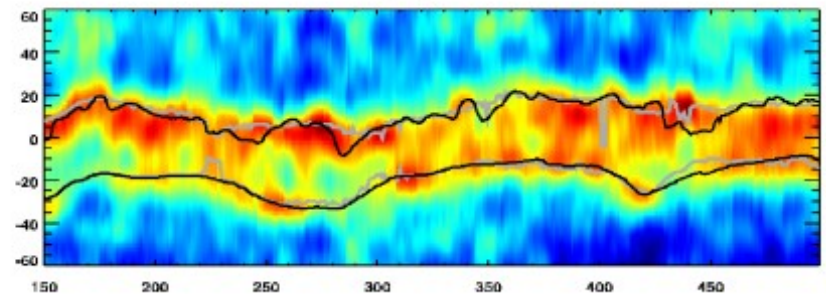
- Estimated profiles match LiDAR
- HH profiles : similar to FP case

LiDAR — TomSAR —

VV



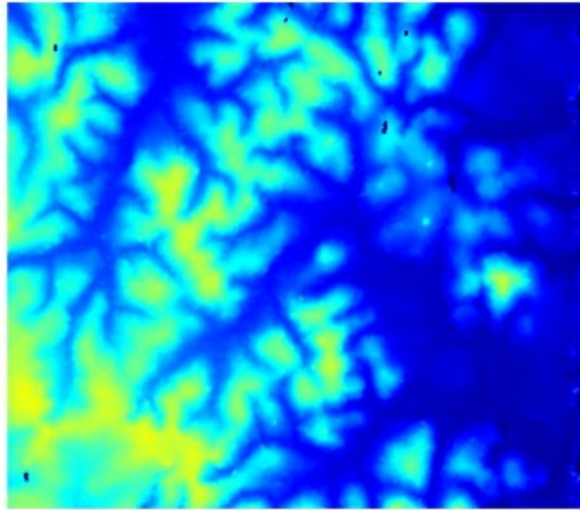
FP



# Tropical forest characterization

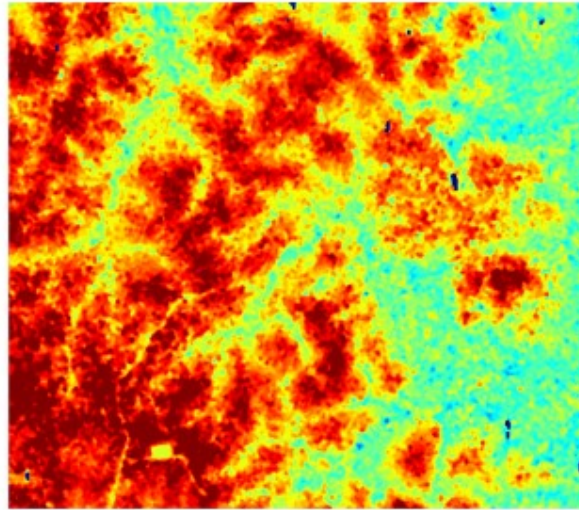
LiDAR data (ground range)

$Z_g$



$Z_g, Z_{top}$

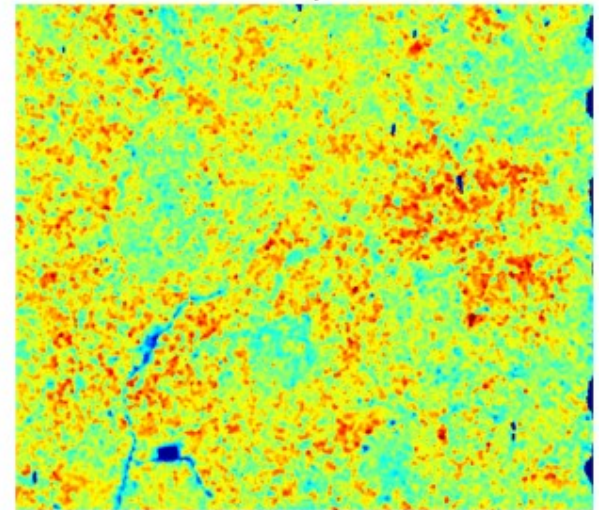
$Z_{top}$



$H_v$

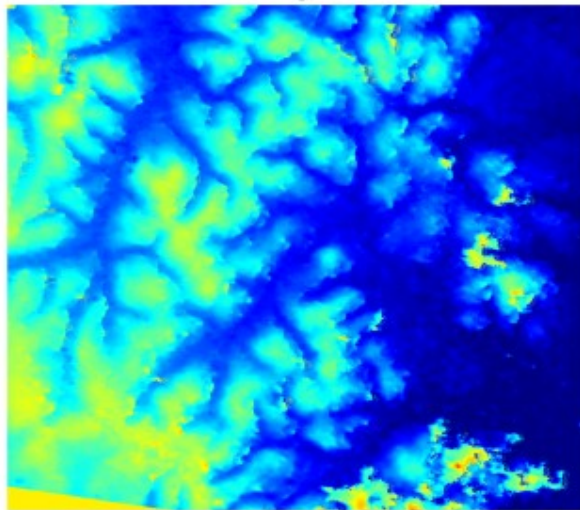
$H_v$  0 50[m]

$H_v$

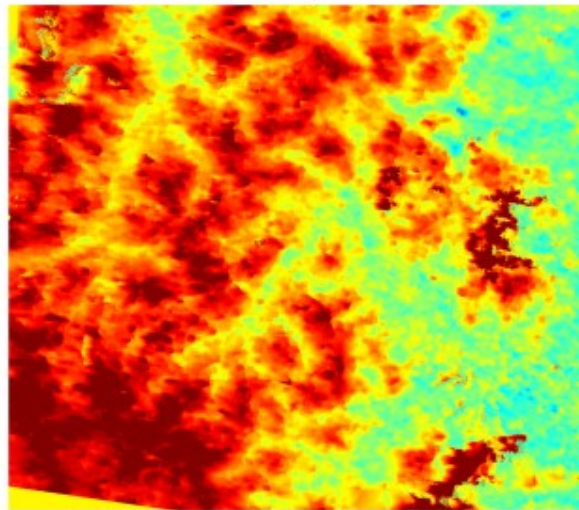


Estimated results (ground range)

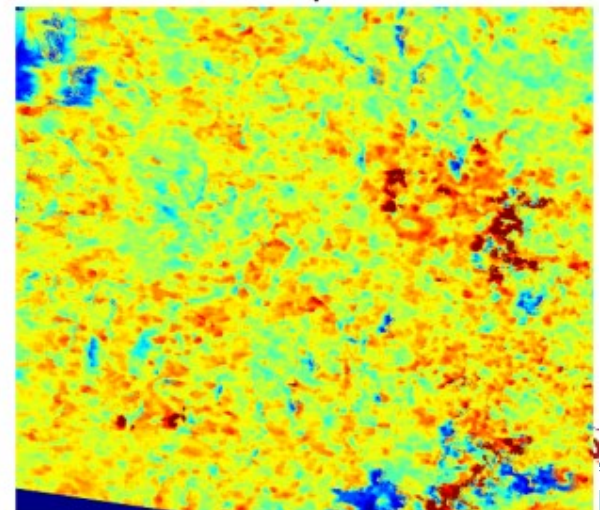
$\hat{Z}_g$



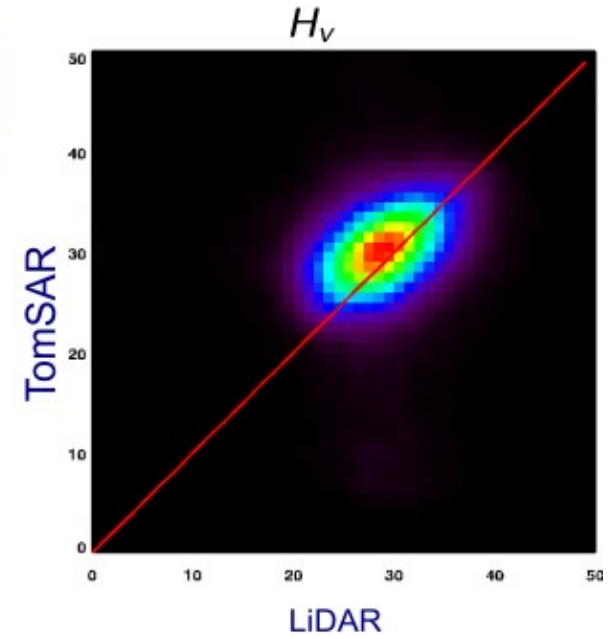
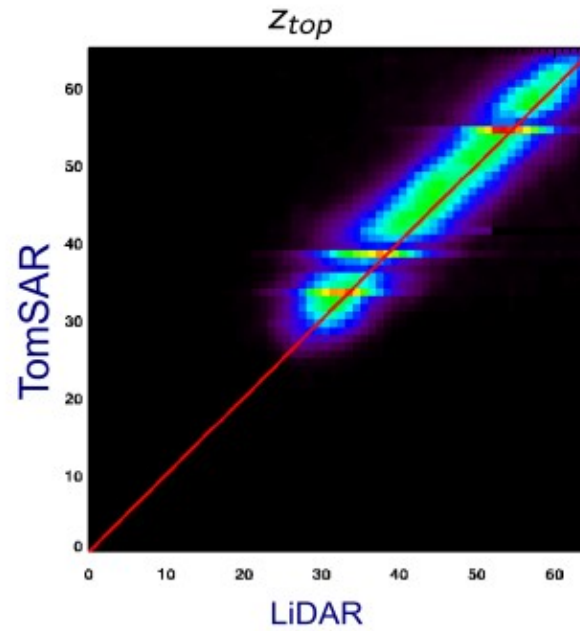
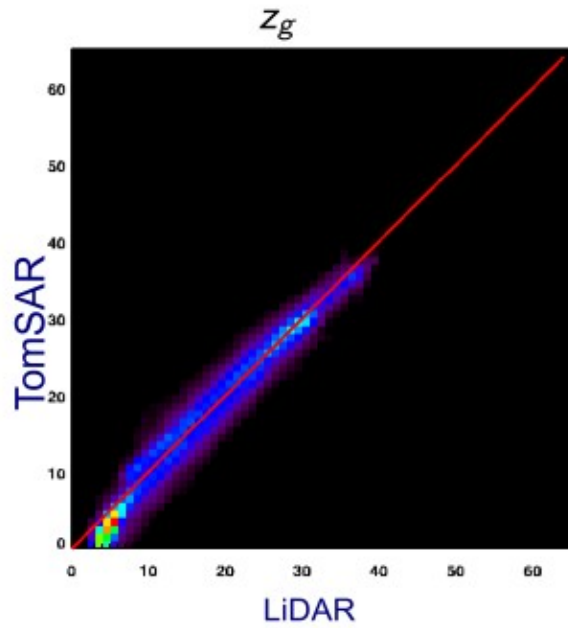
$\hat{Z}_{top}$



$\hat{H}_v$



# Tropical forest characterization

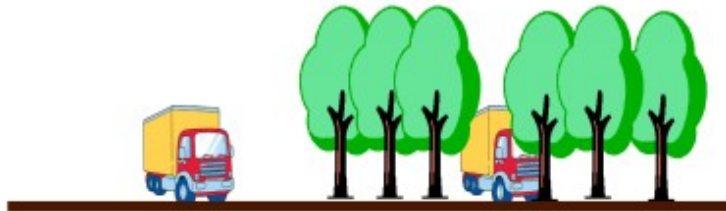






# TomoSAR imaging of concealed objects

## VV reflectivity tomograms



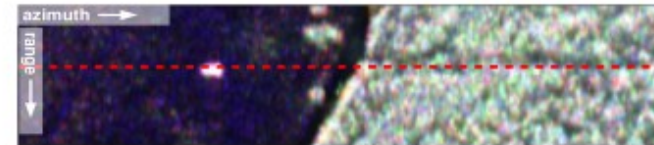
Capon :

limited resolution

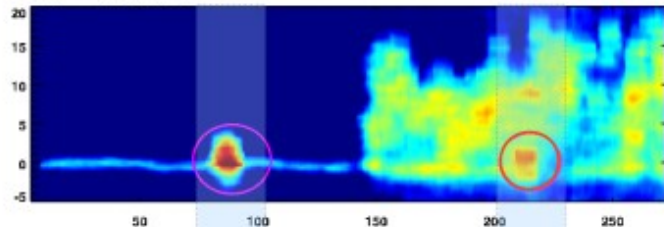
⇒ overestimated  $H_{truck}$

MUSIC :

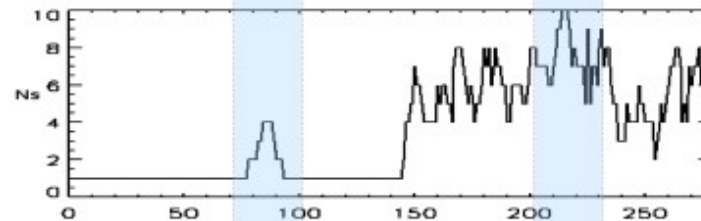
- ☺ Sub-canopy truck  
⇒ hybrid scatterer
- ☹ Uncovered  
⇒ coherent scatterer
- ☹ Spurious sidelobes.



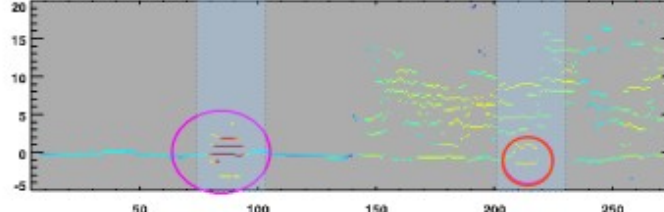
SP -CAPON



Model Order



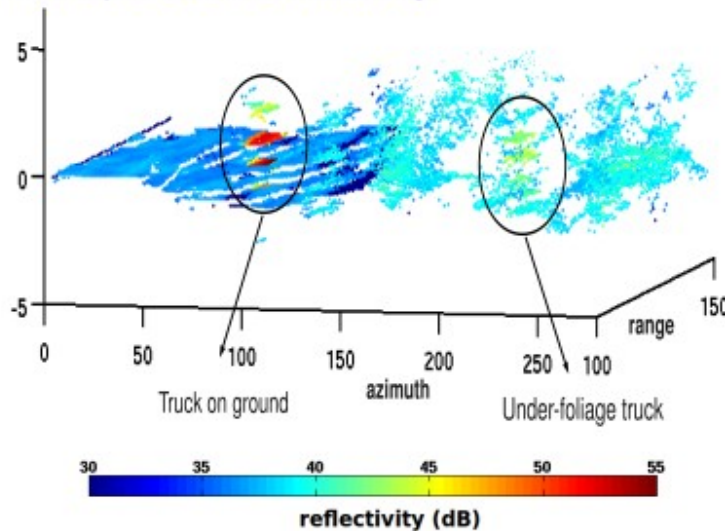
SP -MUSIC



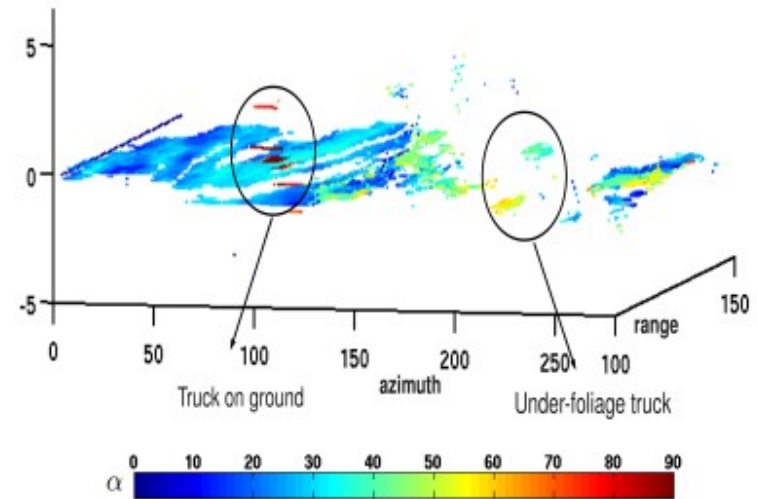
# TomoSAR imaging of concealed objects

## High Resolution tomograms of underfoliage objects

SSF : shape and reflectivity

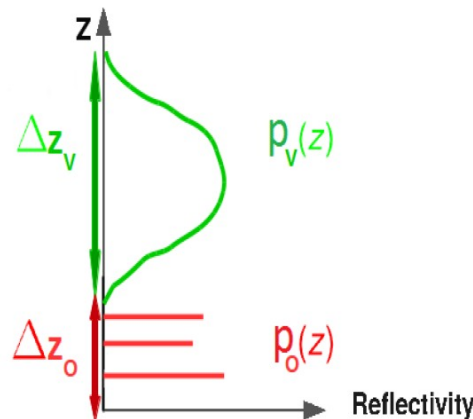
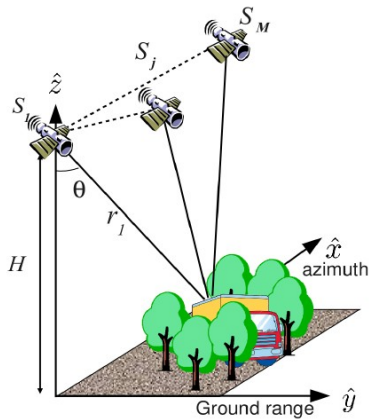


FP-NSF : scattering mechanisms



# TomoSAR imaging of concealed objects

## Sparse (compressive) sensing solution



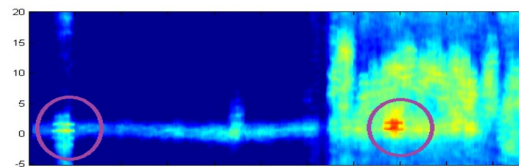
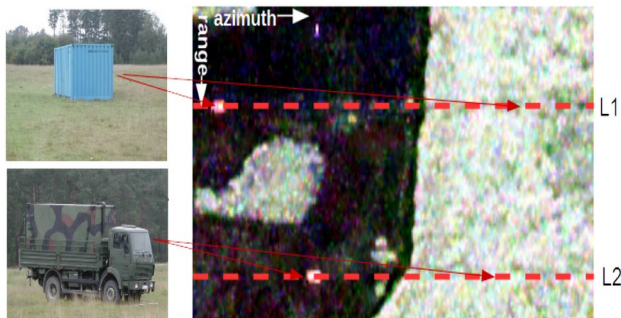
- a few wavelet components
- a few discrete contributions

$$\min_{\mathbf{p}} \|\mathbf{B}\mathbf{p}\|_1 \text{ subject to } \|\mathbf{R} - \hat{\mathbf{R}}\|_F \leq \epsilon \quad \hat{\mathbf{R}} = \mathbf{A}(\mathbf{z}) \text{diag}(\mathbf{p})\mathbf{A}^H(\mathbf{z})$$

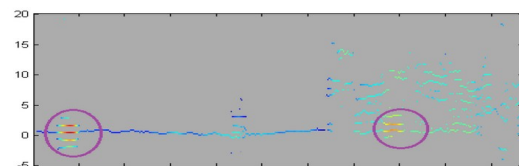
$$\mathbf{B} = \begin{bmatrix} \mathbf{I}_{(N_o \times N_o)} & \mathbf{0} \\ \mathbf{0} & \Psi_{(N_v \times N_v)} \end{bmatrix} \in \mathbb{R}^{(N_s \times N_s)}$$

$$\mathbf{p} = [\mathbf{p}_o^T \quad \mathbf{p}_v^T]^T \in \mathbb{R}^{+N_s \times 1}$$

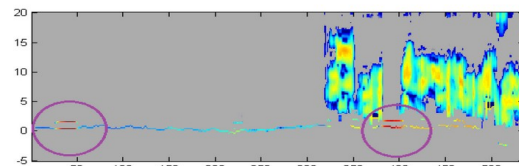
# TomoSAR imaging of concealed objects



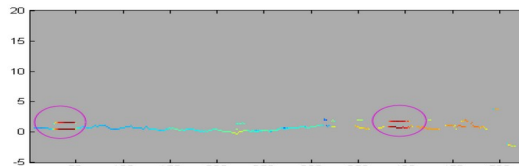
(a) Capon



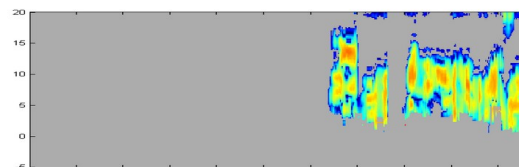
(b) MUSIC



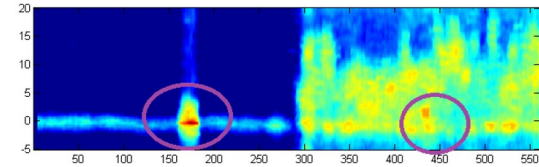
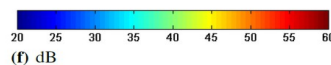
(c) Proposed method with merging



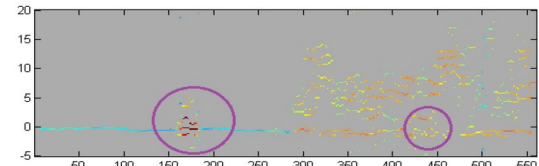
(d) Ground and underfoliage scattering ( $p_o$ ) estimated by proposed method with merging



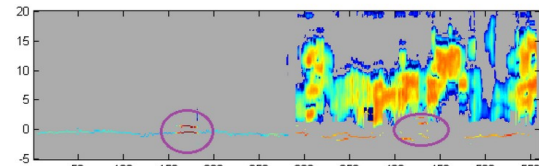
(e) Canopy power ( $p_v$ ) estimated by proposed method with merging



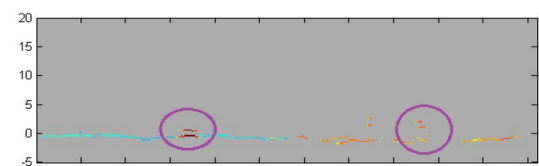
(a) Capon



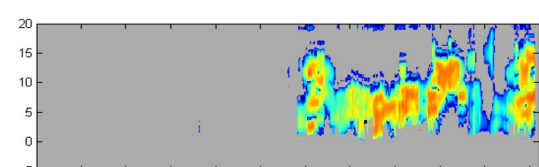
(b) MUSIC



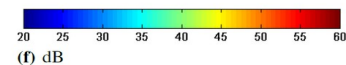
(c) Proposed method with merging



(d) Ground and underfoliage scattering ( $p_o$ ) estimated by proposed method with merging



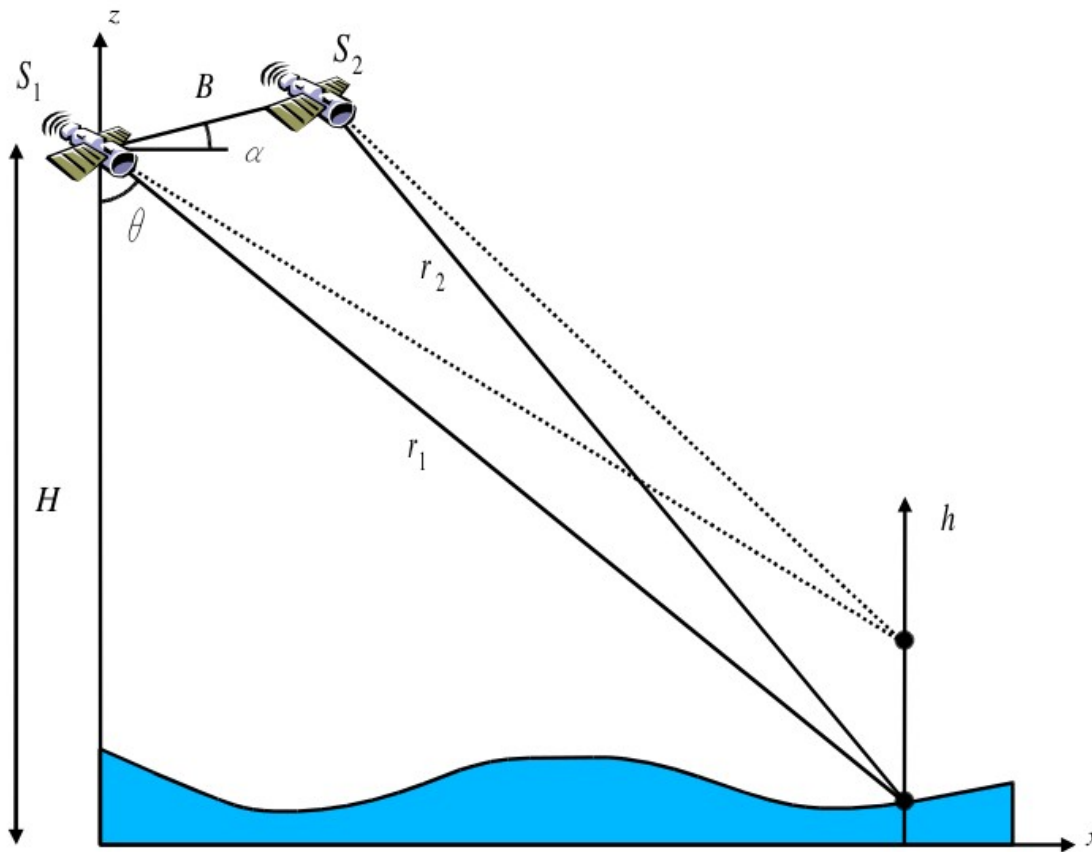
(e) Canopy power ( $p_v$ ) estimated by proposed method with merging



# *InSAR coherence analysis and TomoSAR modeling*



# InSAR coherence analysis



Coherent SAR image pair

$$s_1 = \sqrt{I_1} e^{j\phi_1} = \sqrt{I_1} e^{j(-kr_1 + \phi_{obj1})}$$

$$s_2 = \sqrt{I_2} e^{j\phi_2} = \sqrt{I_2} e^{j(-kr_2 + \phi_{obj2})}$$

Assumptions

$$I_1 \approx I_2 \quad \text{and} \quad \phi_{obj1} \approx \phi_{obj2}$$

Interferometric phase difference

$$\Delta\phi_{12} = \arg(s_1 s_2^*)$$

$S_1$  and  $S_2$  are random variables

⇒ assumptions may not be fully verified

⇒ the interferometric phase difference is a random variable



## Interferometric coherence

Joint interferometric representation

$$\mathbf{k} = \begin{bmatrix} s_1 \\ s_2 \end{bmatrix} \sim \mathcal{N}_c(\mathbf{0}, \mathbf{C})$$

$$\text{with } \mathbf{C} = \mathbf{E}(\mathbf{k}\mathbf{k}^\dagger) = \begin{bmatrix} \mathbf{E}(s_1 s_1^*) & \mathbf{E}(s_1 s_2^*) \\ \mathbf{E}(s_1^* s_2) & \mathbf{E}(s_2 s_2^*) \end{bmatrix} = \begin{bmatrix} \overline{I_1} & \gamma \sqrt{\overline{I_1} \overline{I_2}} \\ \gamma^* \sqrt{\overline{I_1} \overline{I_2}} & \overline{I_2} \end{bmatrix}$$

Interferometric coherence : normalized correlation coefficient

$$\gamma = \frac{\mathbf{E}(s_1 s_2^*)}{\sqrt{\overline{I_1} \overline{I_2}}} = |\gamma| e^{j\phi} \quad |\gamma| \leq 1 \quad \text{Cauchy-Schwarz inequality}$$

$|\gamma| = 1 \Rightarrow \phi = \Delta\phi_{12}$  interferometric assumptions are fulfilled

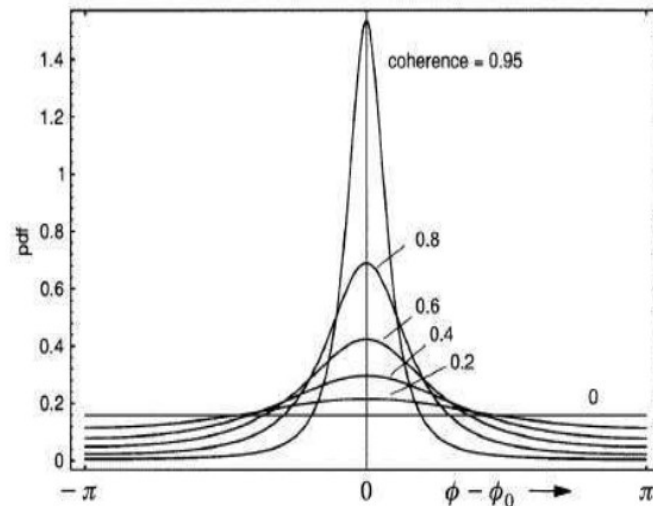
$|\gamma| = 0 \Rightarrow \phi = ?$  interferometric images are totally uncorrelated

$|\gamma|$  is an indicator of the interferometric information (and phase) quality

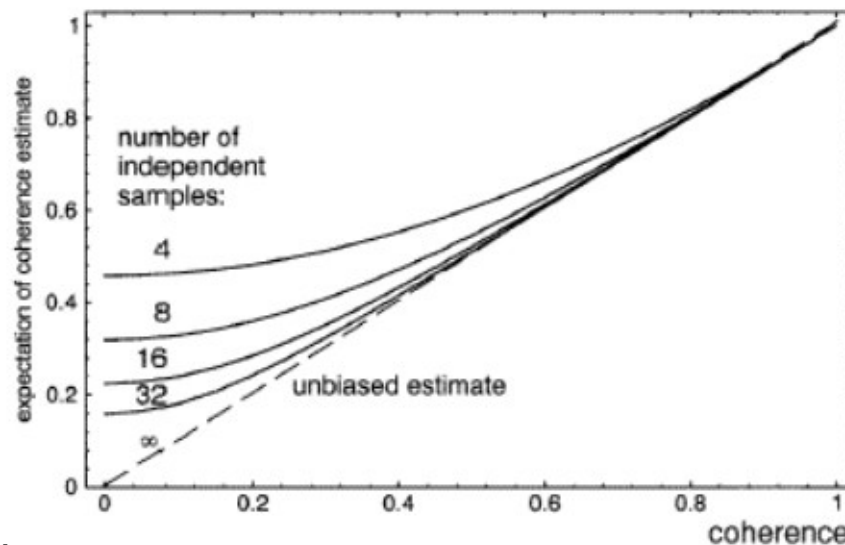
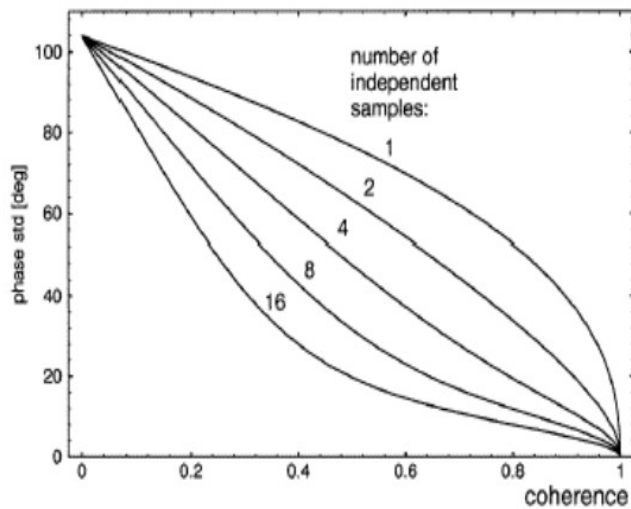


# InSAR coherence analysis

Phase pdf



$|\gamma|$  : indicator of phase quality



Large number of looks required to reduce:

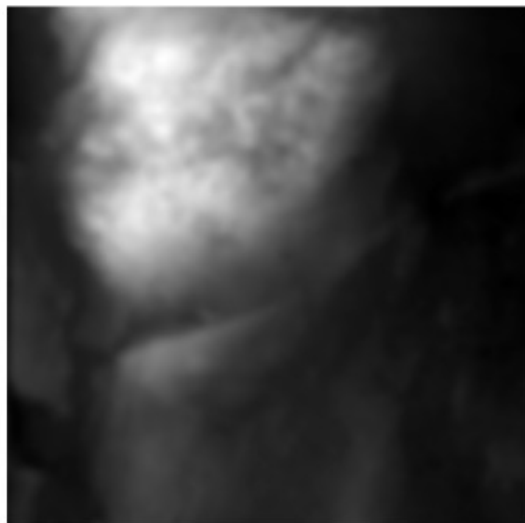
$$\sigma_{\hat{\phi}} \text{ as } |\gamma| \rightarrow 0$$

$$\text{bias: } |\gamma| - E(|\hat{\gamma}|) \geq 0$$

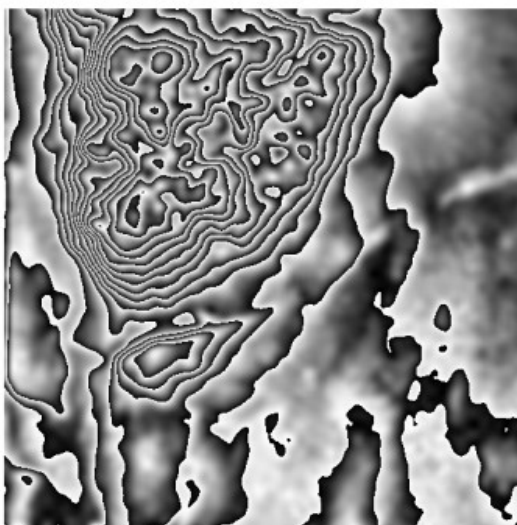


# InSAR coherence analysis

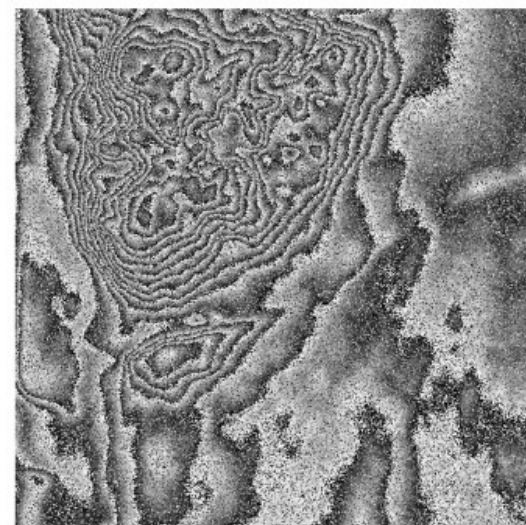
## Single-look inSAR phase



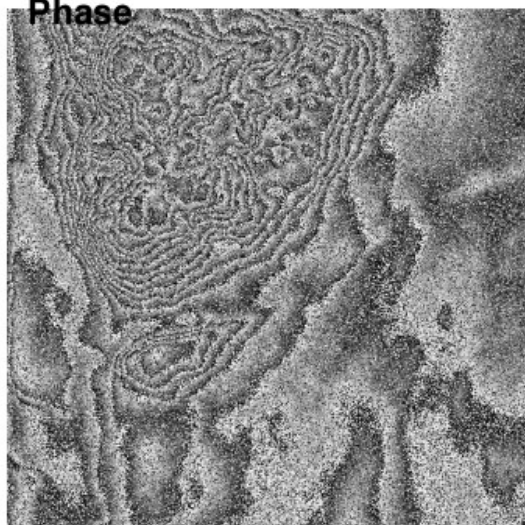
Absolute «True»  
Phase



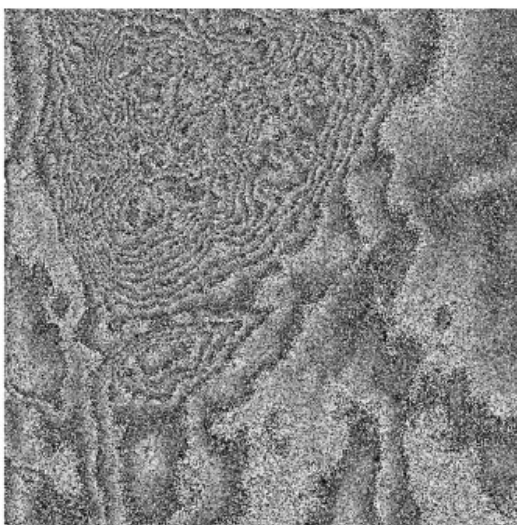
Coherence=1.0 L=1



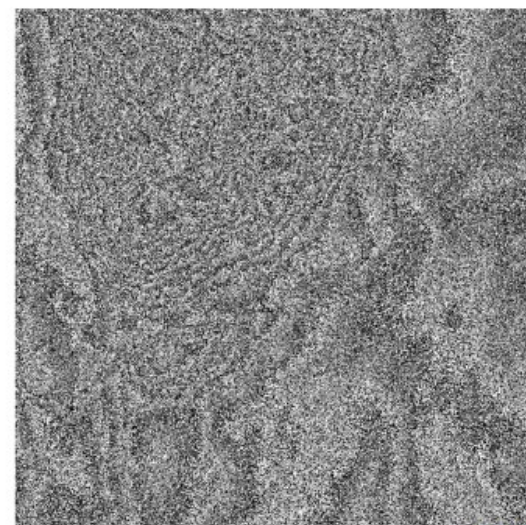
Coherence=0.8 L=1



Coherence=0.6 L=1

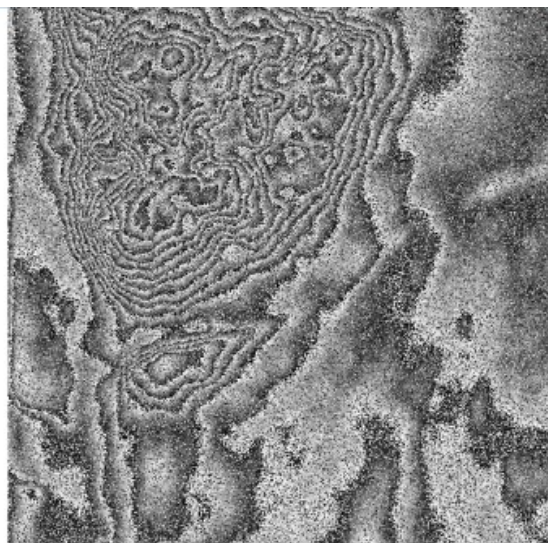


Coherence=0.4 L=1

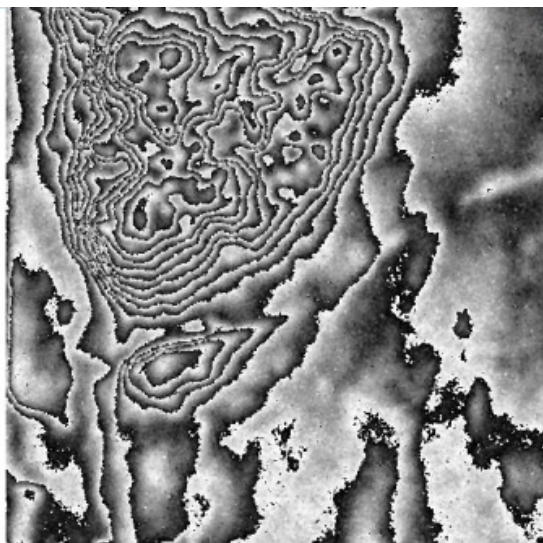


Coherence=0.2 L=1

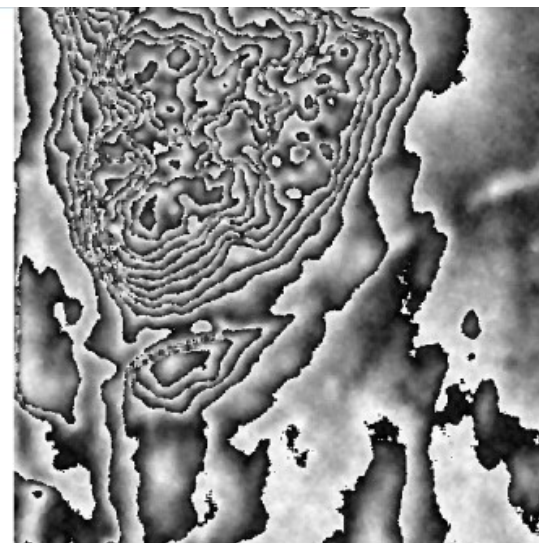
## Multi-look inSAR phase



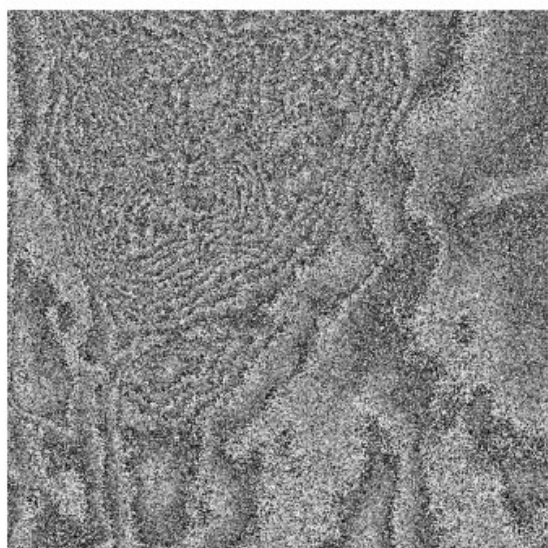
Coherence = 0.7      L=1



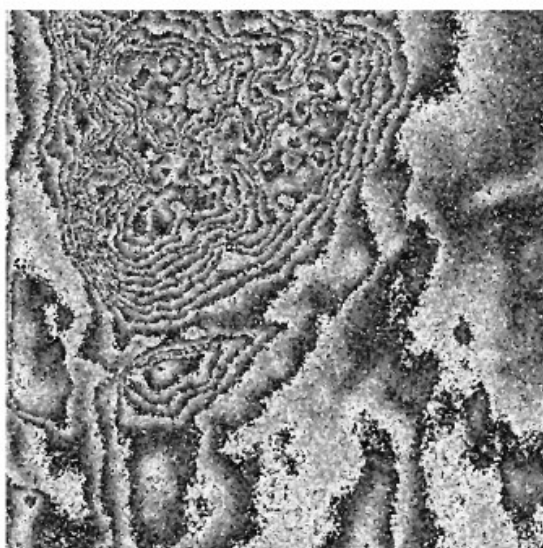
Coherence = 0.7      L=8



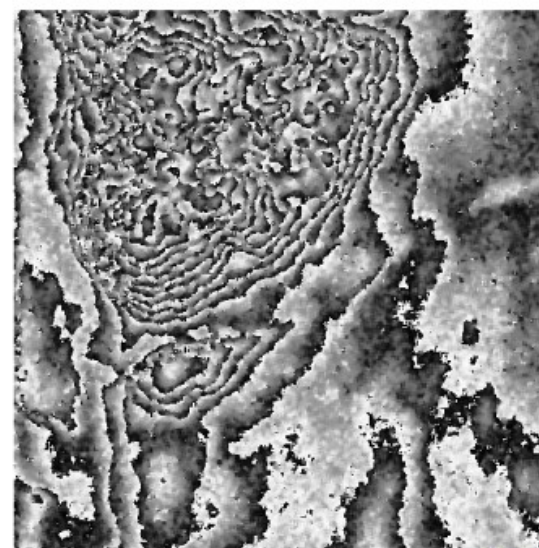
Coherence = 0.7      L=16



Coherence = 0.3      L=1



Coherence = 0.3      L=8



Coherence = 0.3      L=16

# InSAR coherence decomposition

The true value of the coherence,  $\gamma$ , is fixed by a set of external sources :

Thermal or system noise : SAR amplifiers, ADC, antennas ...

Geometric decorrelation : Baseline, squint ...

Volume decorrelation : Volumetric media e.g. forest ...

Temporal variations : wind, flowing or plowing, building ...

Processing errors : coregistration, interpolation ...

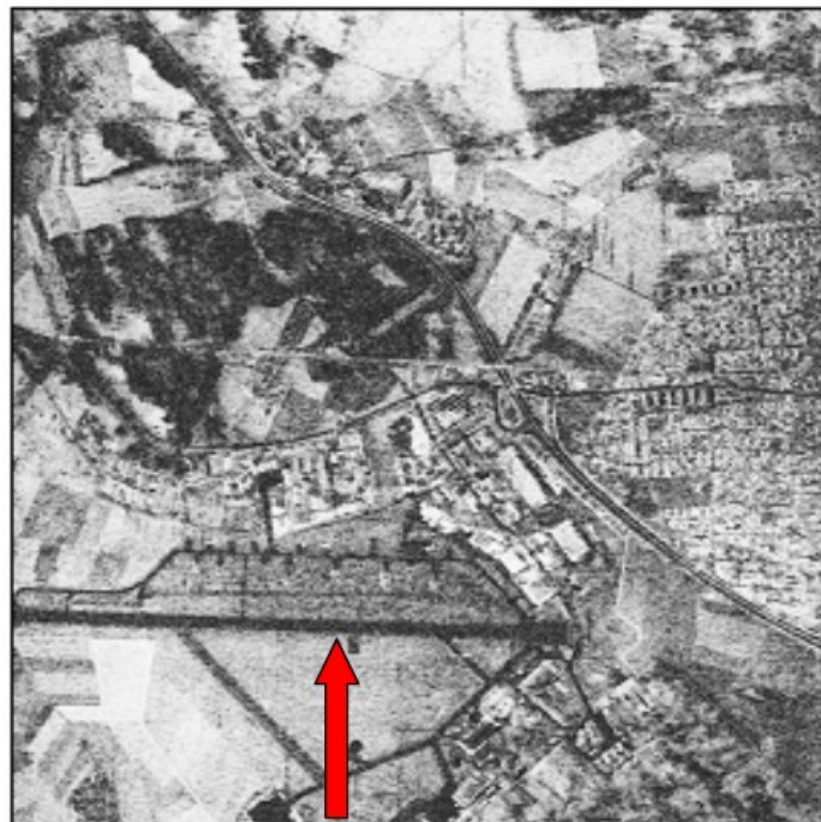
$$\gamma = \gamma_{th} \cdot \gamma_{geom} \cdot \gamma_{vol} \cdot \gamma_{temp} \cdot \gamma_{proc}$$

## Thermal or system decorrelation

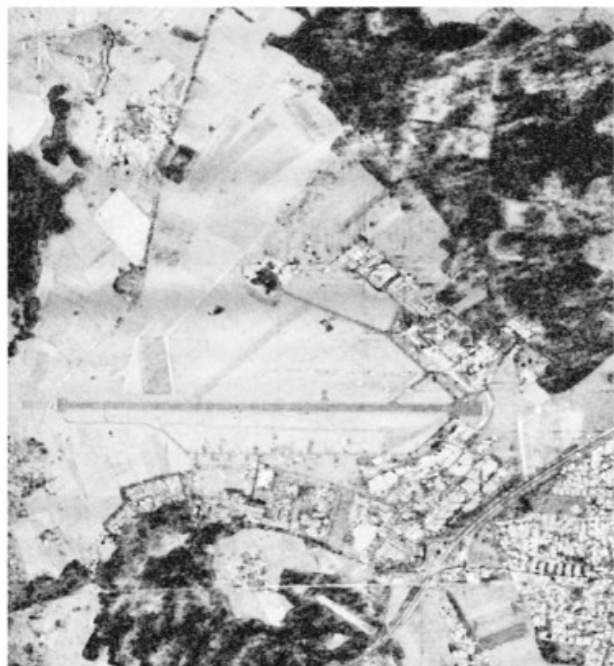
Intensity image



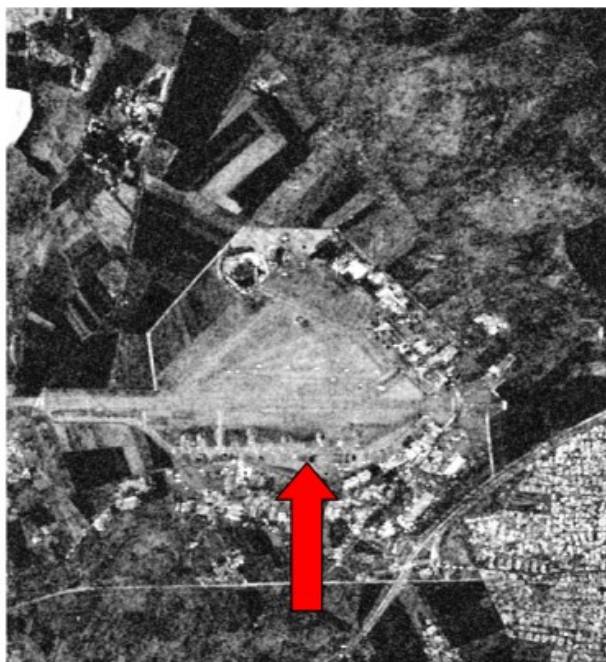
Coherence image



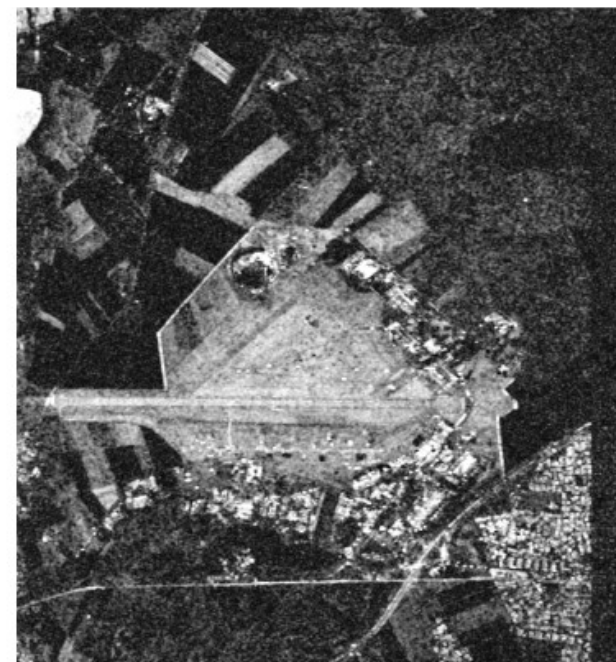
## Temporal decorrelation



1 hour, 20m baseline



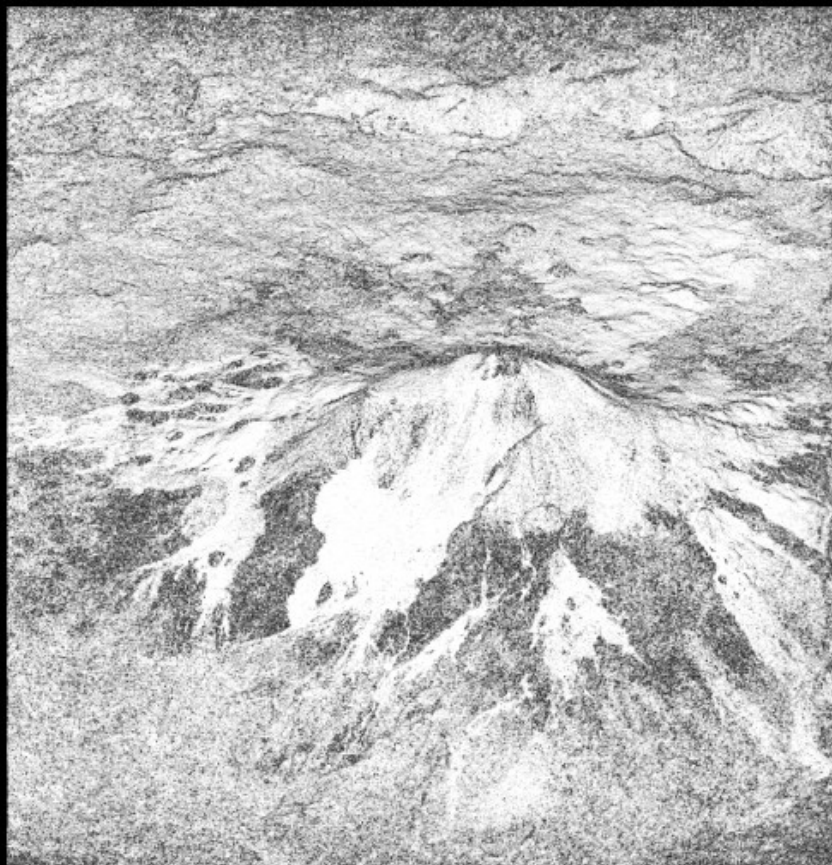
3 months, 0m baseline



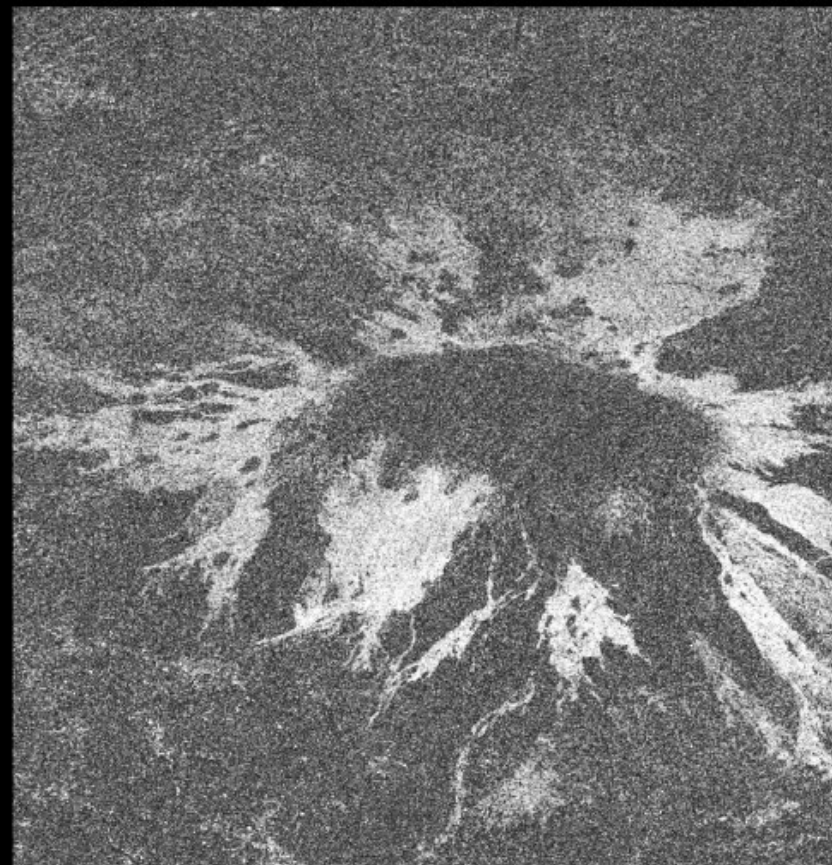
1 year, 0m baseline

## Temporal decorrelation

### Coherence Maps



1 day ERS-1/ERS-2



70 days ERS-1/ERS-1

Test Site: Mt. Etno/Italy



## Volume decorrelation

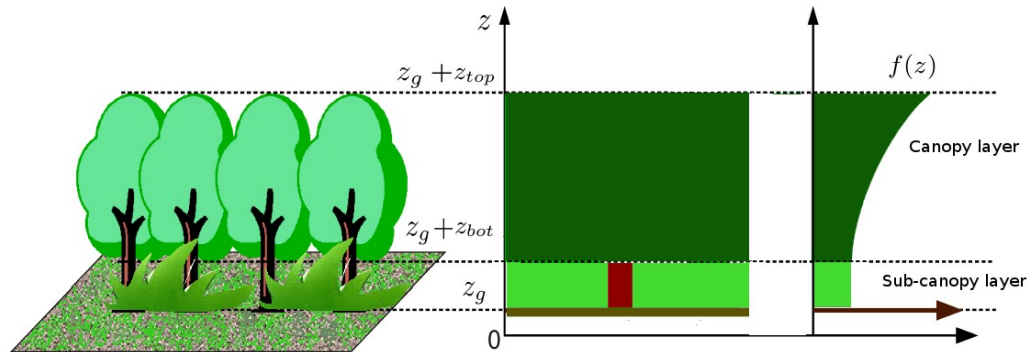
1 hour, 20 baseline



# InSAR vertical decorrelation over volumes

## Volumetric media inSAR response modeling

- Vertical reflectivity structure  $\sigma_{v_e}(\vec{r}) = \sigma_{v_e}(z) = A_{v_e} f(z)$



- InSAR coherence  $\gamma = \gamma_{th} \gamma_{proc} \gamma_{temp} \gamma_{surf} \gamma_z$

- Decorrelation due to vertical structure :

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

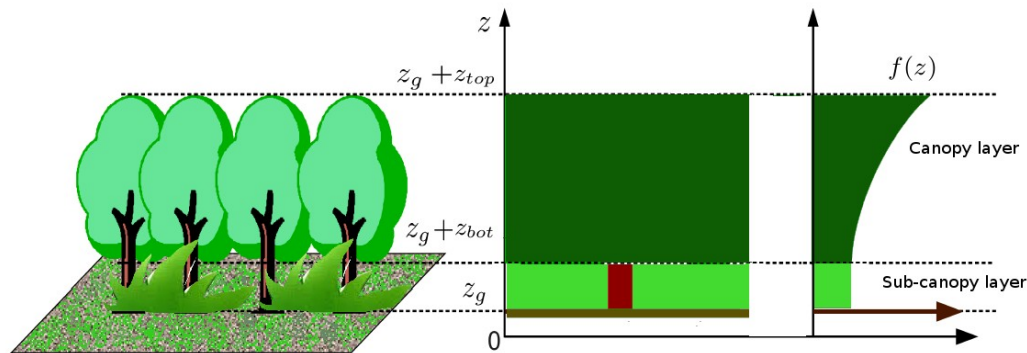
$$k_z = \frac{k_c B_{\perp}}{r \sin \theta}$$

- Fourier transform-like **coherence-structure relationship**

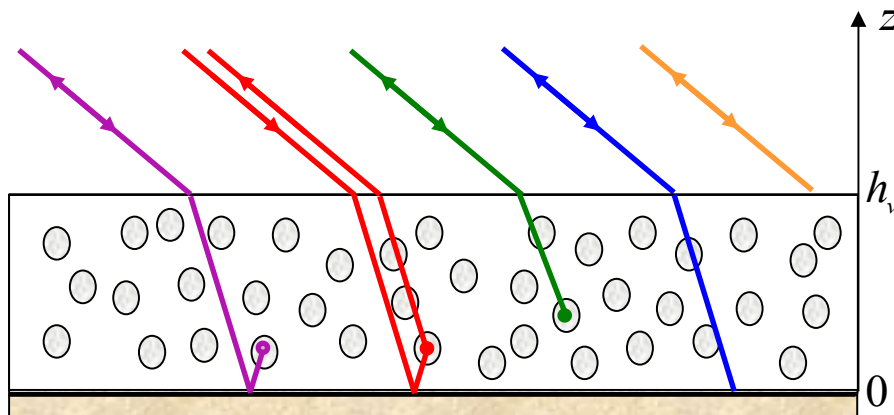
$$\gamma_z \xleftrightarrow{FT} \sigma_{v_e}(z)$$



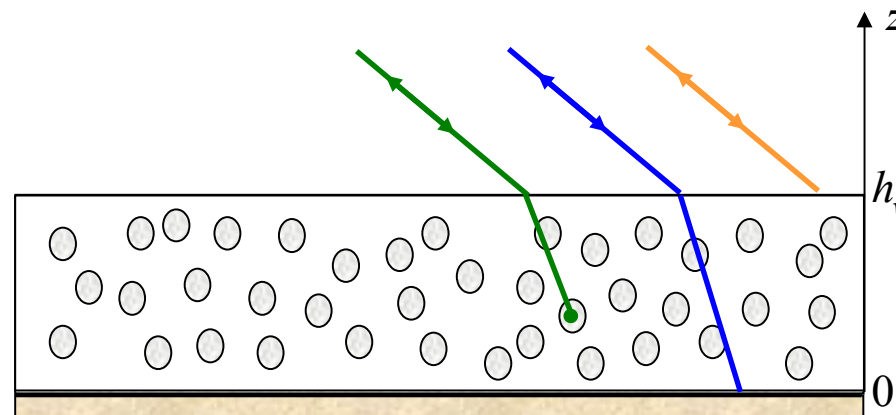
# InSAR RVOG model



## Modeling at order 1



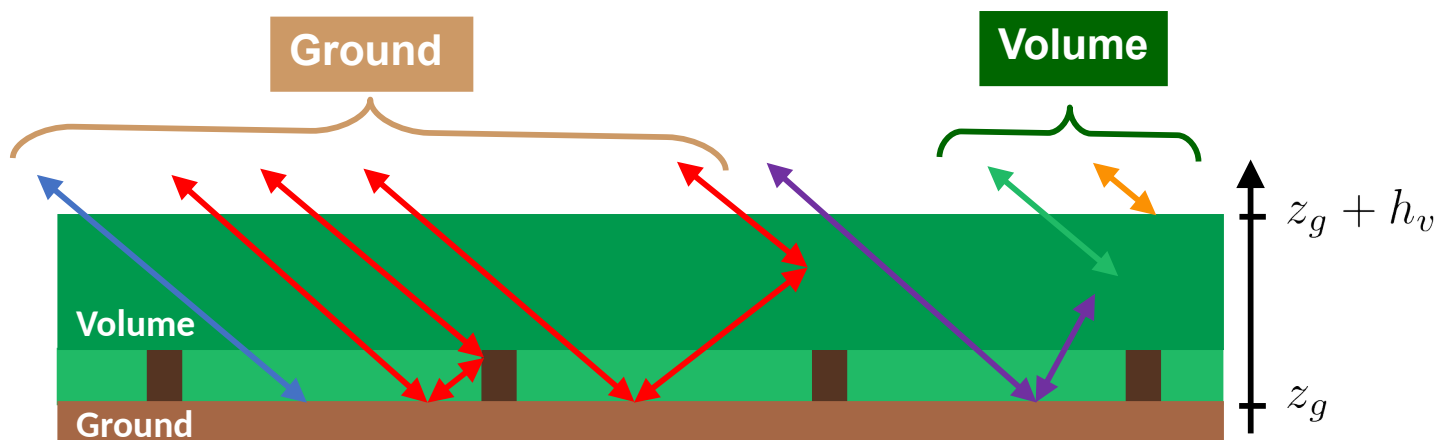
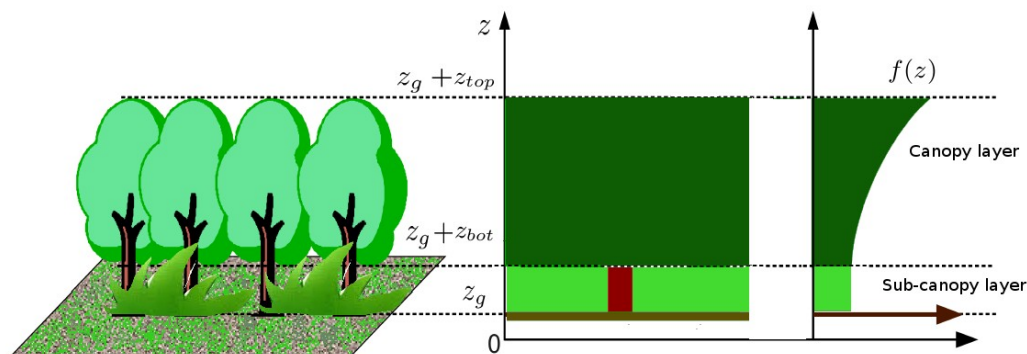
## Parameter Estimation



Parameter estimation often requires to simplify models

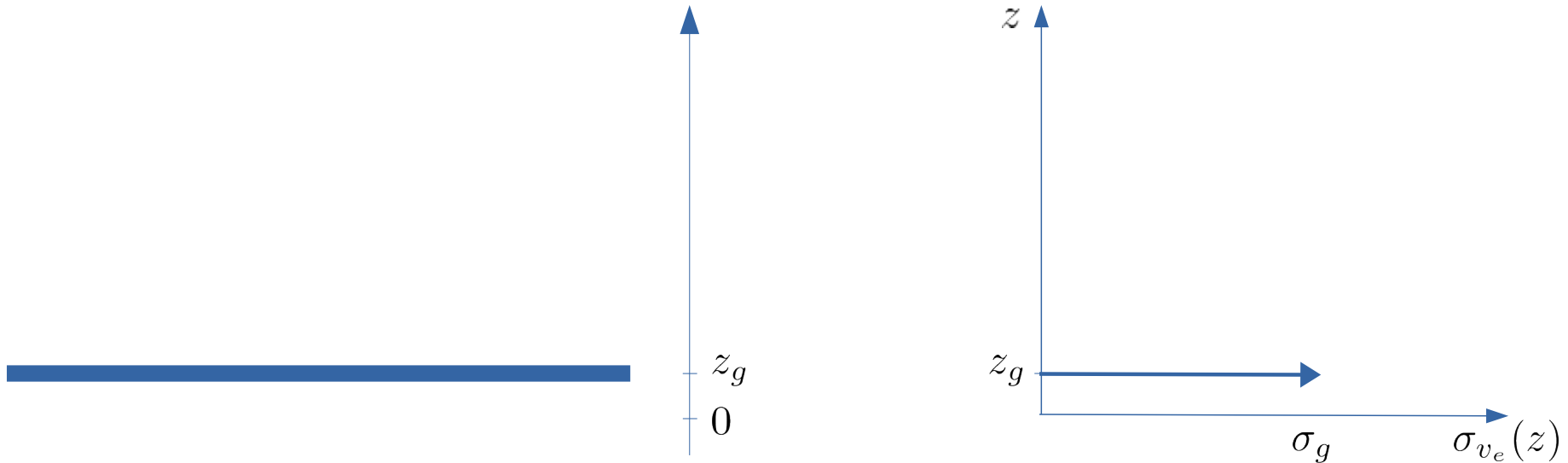
- omitting negligible terms
- merging contributions that cannot be discriminated (e.g. ground and double-bounce)

# InSAR RVOG model



- 2 significant and uncorrelated mechanisms :
  - ⇒ volume + underlying ground
- low density medium
  - ⇒ no refraction

## Ground only

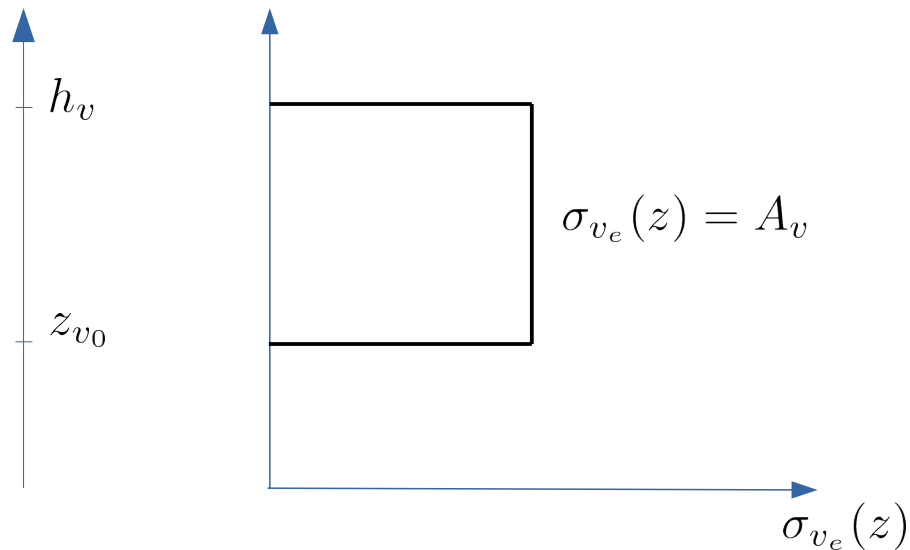
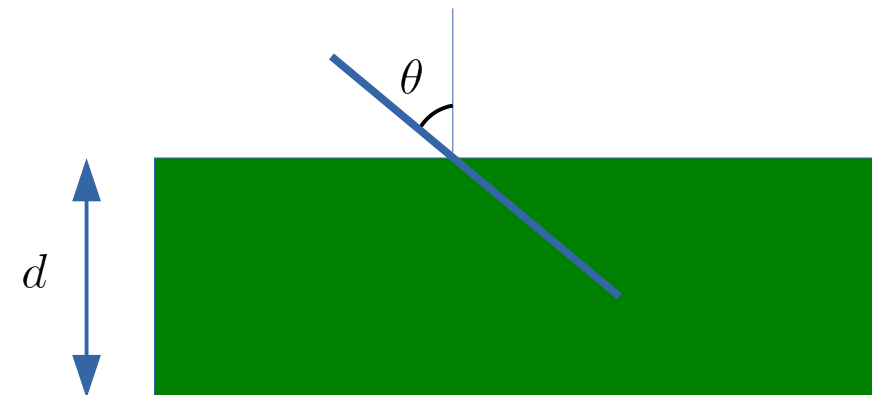


$$\gamma_z = \frac{\int \sigma_{v_e}(z) e^{jk_z z} dz}{\int \sigma_{v_e}(z) dz}$$

$$\sigma_{v_e}(z) = \sigma_g \delta(z - z_g) \Rightarrow \gamma_z = e^{jk_z z_g}$$

InSAR well adapted to topography estimation

## Non attenuating random volume only



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

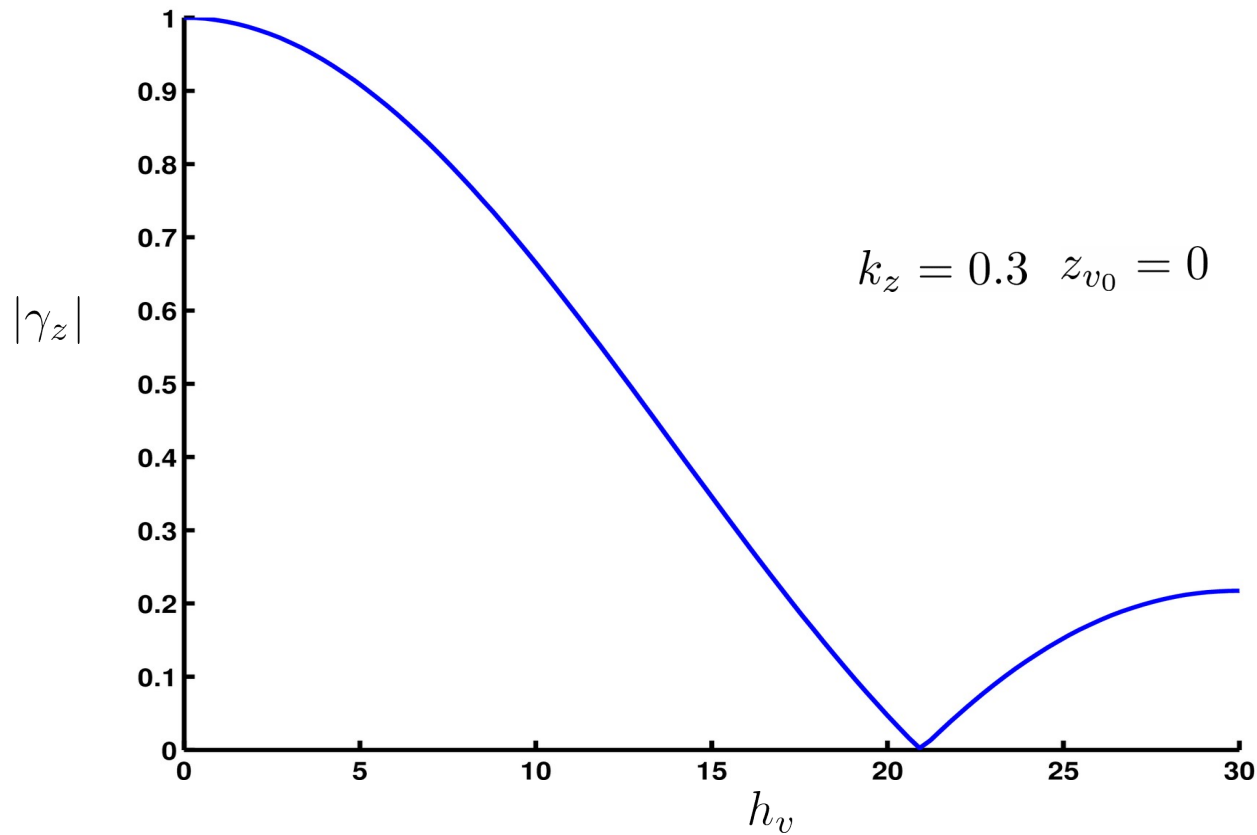
**No underlying ground**

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

$$\gamma_z = \frac{1}{d} \int_{z_{v0}}^{h_v} e^{jk_z z} dz$$

$$\gamma_z = e^{jk_z \frac{h_v + z_{v0}}{2}} \text{sinc} \left( \frac{k_z d}{2} \right)$$

# InSAR RVOG analysis



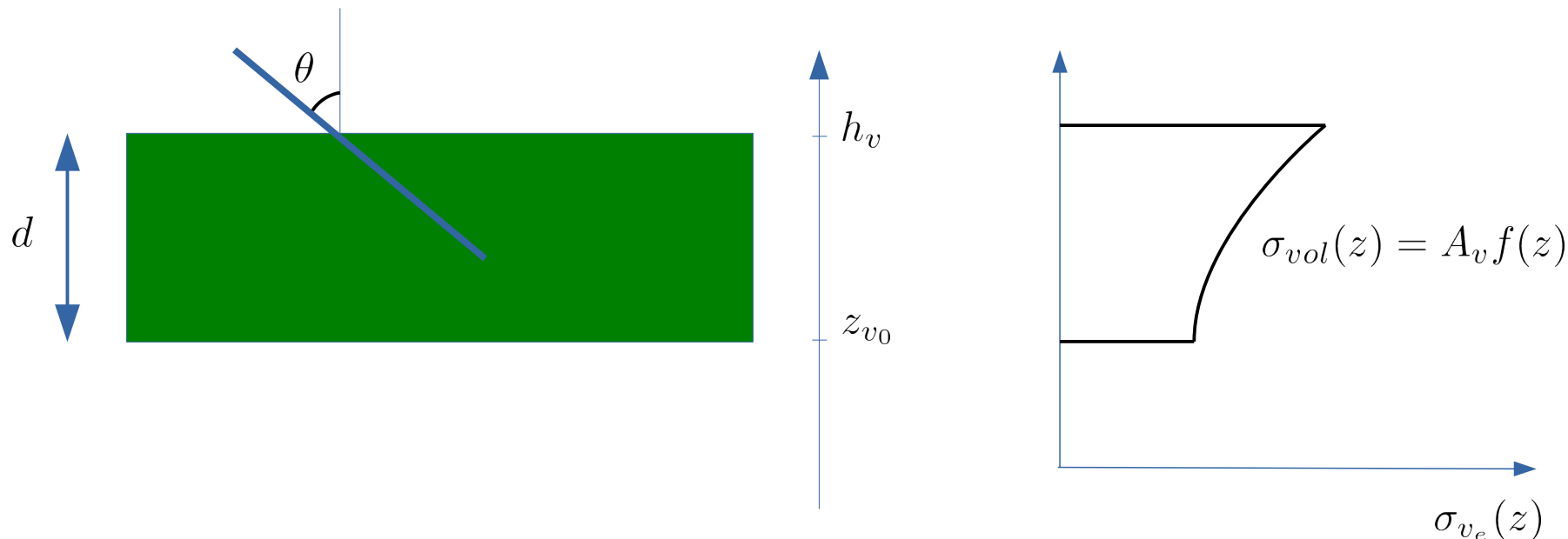
Volume center height

Volume width

$$\gamma_z = e^{jk_z \frac{h_v + z_{v0}}{2}} \operatorname{sinc} \left( \frac{k_z d}{2} \right)$$

InSAR well adapted to volume analysis under specific conditions

## Attenuating random volume only



Linear differential extinction

$$dI = -\kappa_e I ds = -\frac{\kappa_e}{\cos \theta} I dz$$

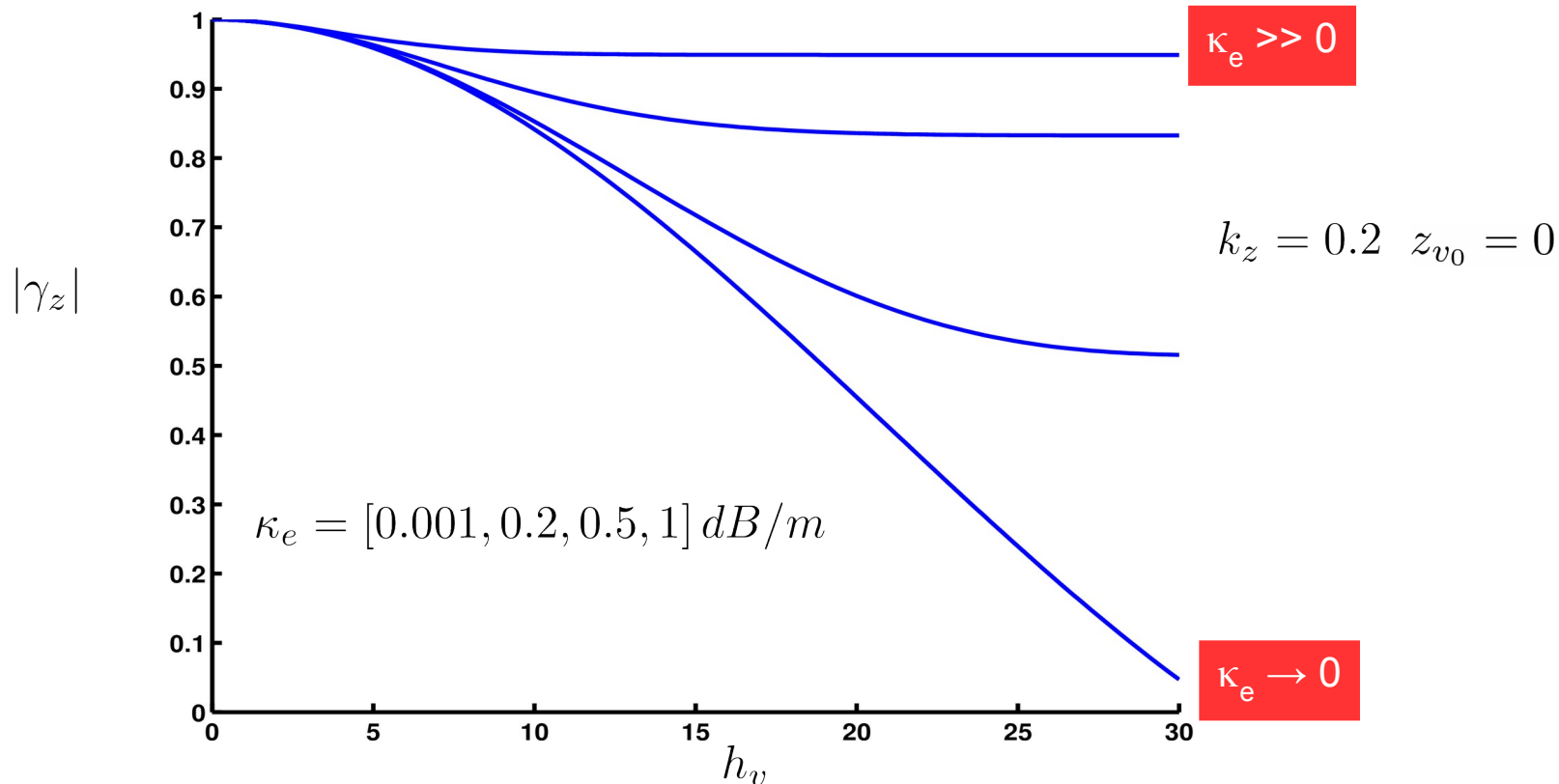
Effective reflectivity density  
**constant extinction**

$$\sigma_{vol}(z) = A_v e^{-2 \frac{\kappa_e}{\cos \theta} (h_v - z)} = A_v f(z)$$

Backscattered volume intensity

$$I_v = \int_{z_{v0}}^{h_v} \sigma_{vol}(z) dz = \int_{z_{v0}}^{h_v} A_v f(z) dz$$

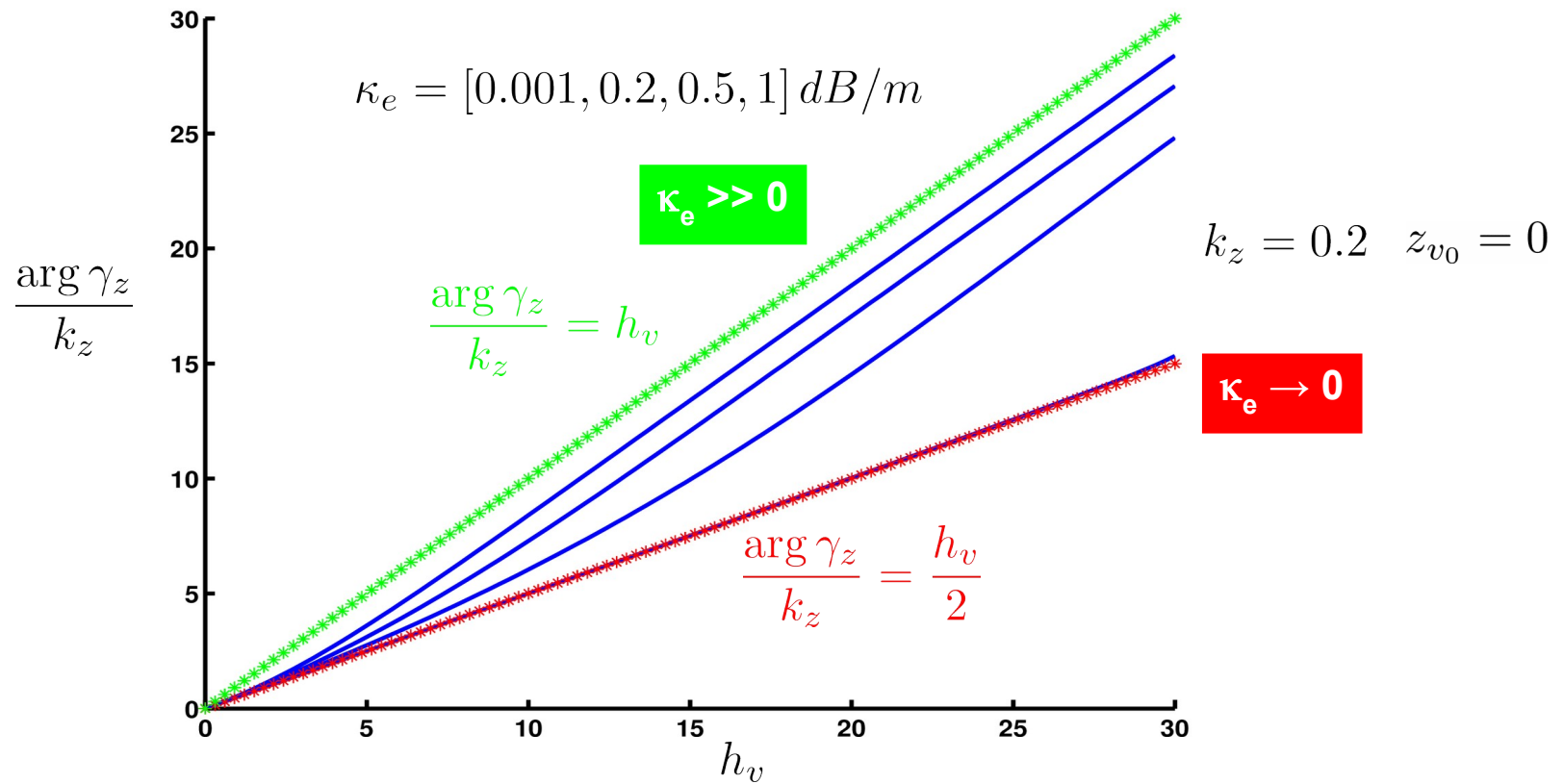
# InSAR RVOG analysis



$$\gamma_{vol} = \gamma_z = e^{jk_z z_{v0}} \frac{p}{p_1} \left( \frac{e^{p_1 d} - 1}{e^{p d} - 1} \right) \quad p = \frac{2\kappa_e}{\cos \theta} \quad p_1 = \frac{2\kappa_e}{\cos \theta} + jk_z$$

InSAR  $|\gamma_z| \rightarrow \hat{h}_v$  ambiguous estimation

# InSAR RVOG analysis

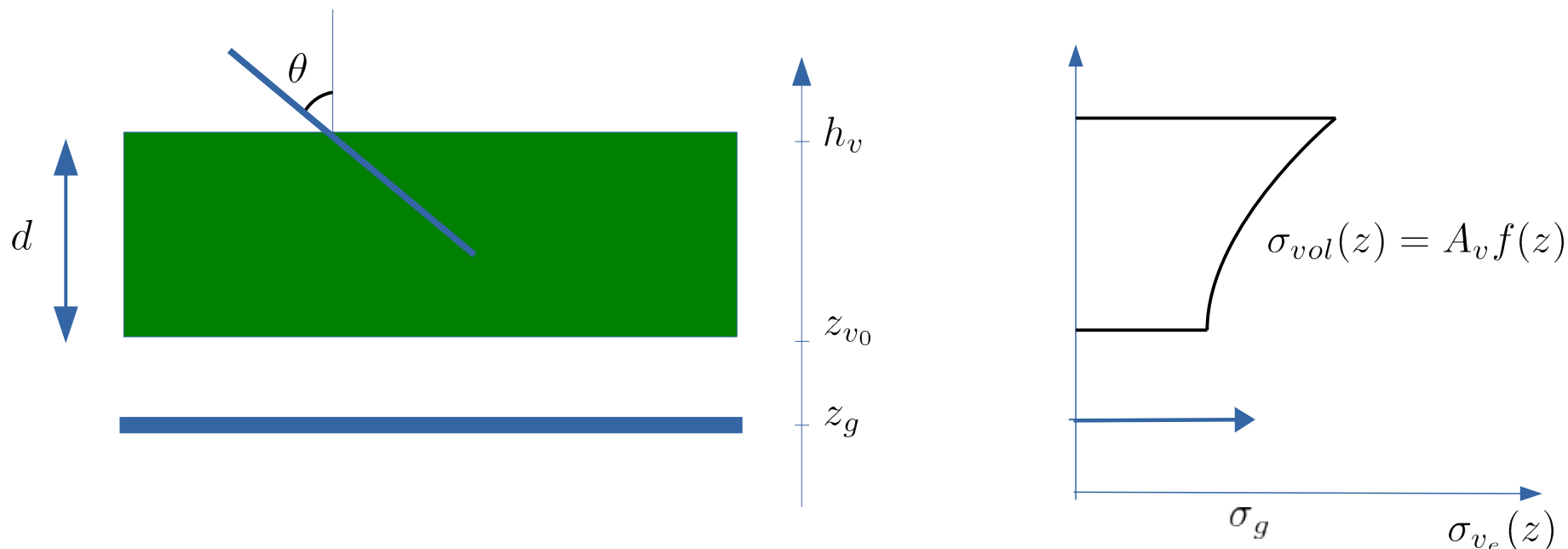


InSAR  $\arg(\gamma_z) \rightarrow \hat{h}_v$  ambiguous estimation

Unambiguous solution for known  $\sigma_{vol}(z)$  shape :  $|\gamma_z|, \arg(\gamma_z) \rightarrow \hat{h}_v$



## Attenuating random volume and ground



Backscattered volume intensity 
$$I_v = \int_{z_{v0}}^{h_v} \sigma_{vol}(z) dz = \int_{z_{v0}}^{h_v} A_v f(z) dz$$

Backscattered ground intensity 
$$I_g = f(z_{v0}) \sigma_g = e^{-2 \frac{\kappa_e}{\cos \theta} d} \sigma_g$$

# InSAR RVOG analysis

- Coherence formulation  $\sigma_{v_e}(z) = \sigma_{vol}(z) + \delta(z - z_g)I_g$

$$\gamma_z = \frac{\int \sigma_{v_e}(z) e^{jk_z z} dz}{\int \sigma_{v_e}(z) dz} = \frac{\int \sigma_{vol}(z) e^{jk_z z} dz + I_g e^{jk_z z_g}}{\int \sigma_{vol}(z) dz + I_g}$$

$$\gamma_z = \frac{\gamma_{vol} + m e^{jk_z z_g}}{1 + m}$$

- Ground to volume intensity ratio  $m = \frac{I_g}{I_v} \in \mathbb{R}^+$
- Coherence interpretation

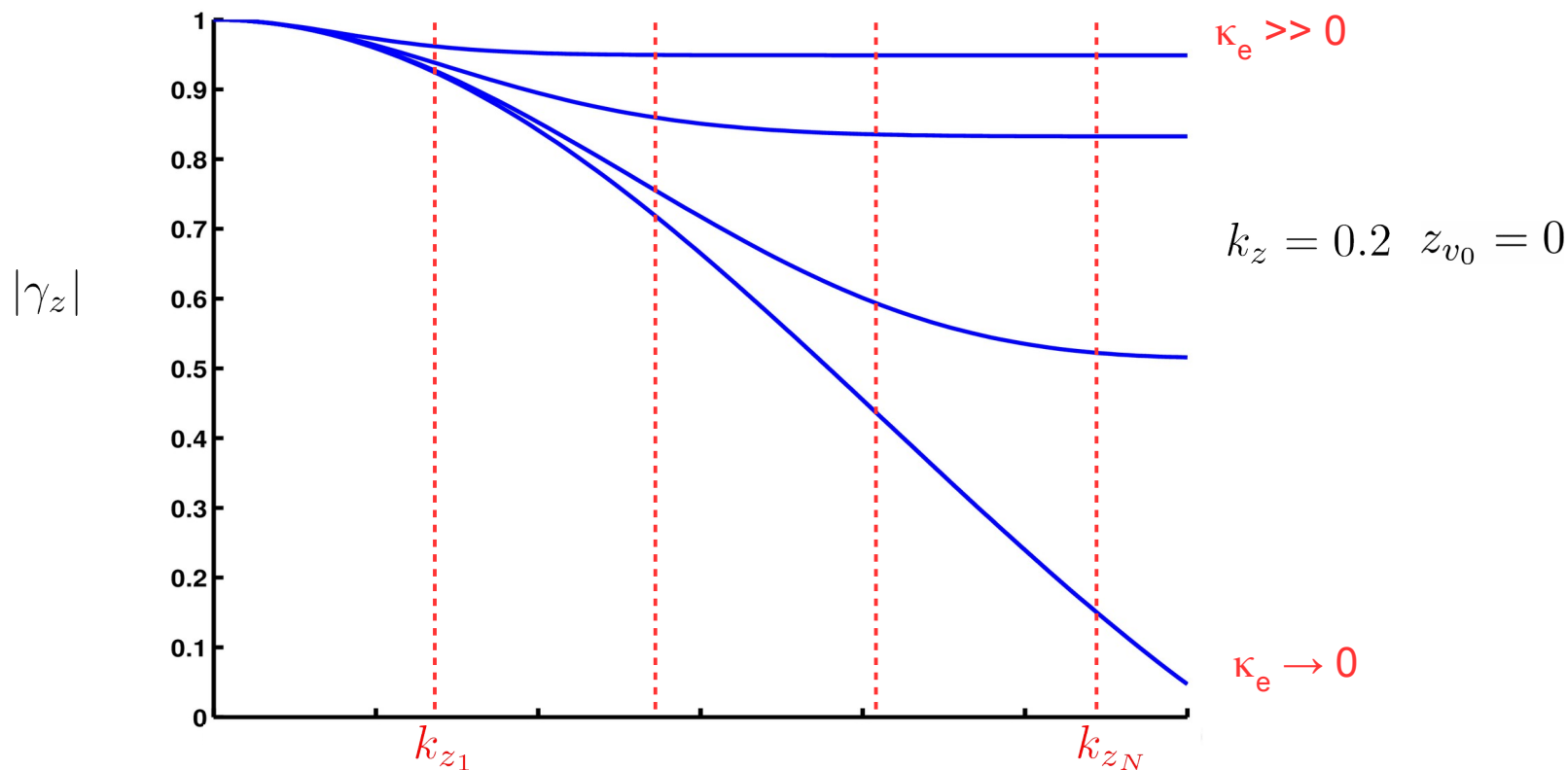
$$m \rightarrow 0 \Rightarrow \begin{cases} \arg \gamma_z \approx \arg \gamma_{vol} \\ |\gamma_z| \leq 1 \end{cases} \quad m \rightarrow +\infty \Rightarrow \begin{cases} \arg \gamma_z \approx \phi_g \\ |\gamma_z| = 1 \end{cases}$$

$$0 < m < +\infty \Rightarrow ?$$

**InSAR based RVOG analysis: under-determined problem**

**→ another source of diversity is needed : polarization ?**

# TomoSAR (MB-InSAR) RVOG analysis

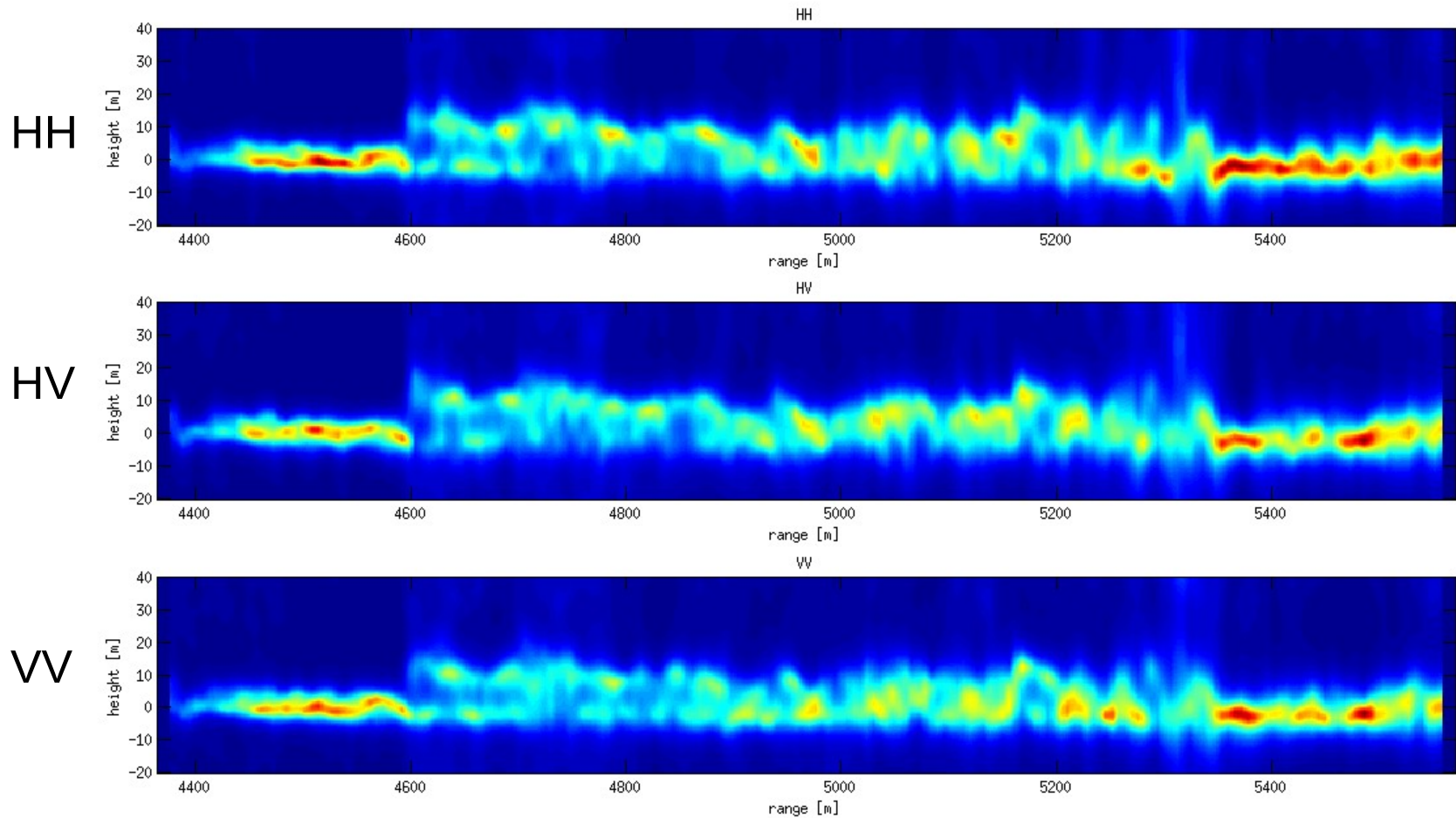


## Additional spatial diversity

- $\{\gamma_z(k_{z_n})\}_{n=1}^N \rightarrow$
- Unambiguous height estimation for known  $f(z)$  shape
  - Estimation of  $f(z)$  or **non-parametric analysis**
  - PolTomoSAR (MB-Pol-InSAR) :  $\{\gamma_z(k_{z_n}), \mathbf{w}\}_{n=1}^N$

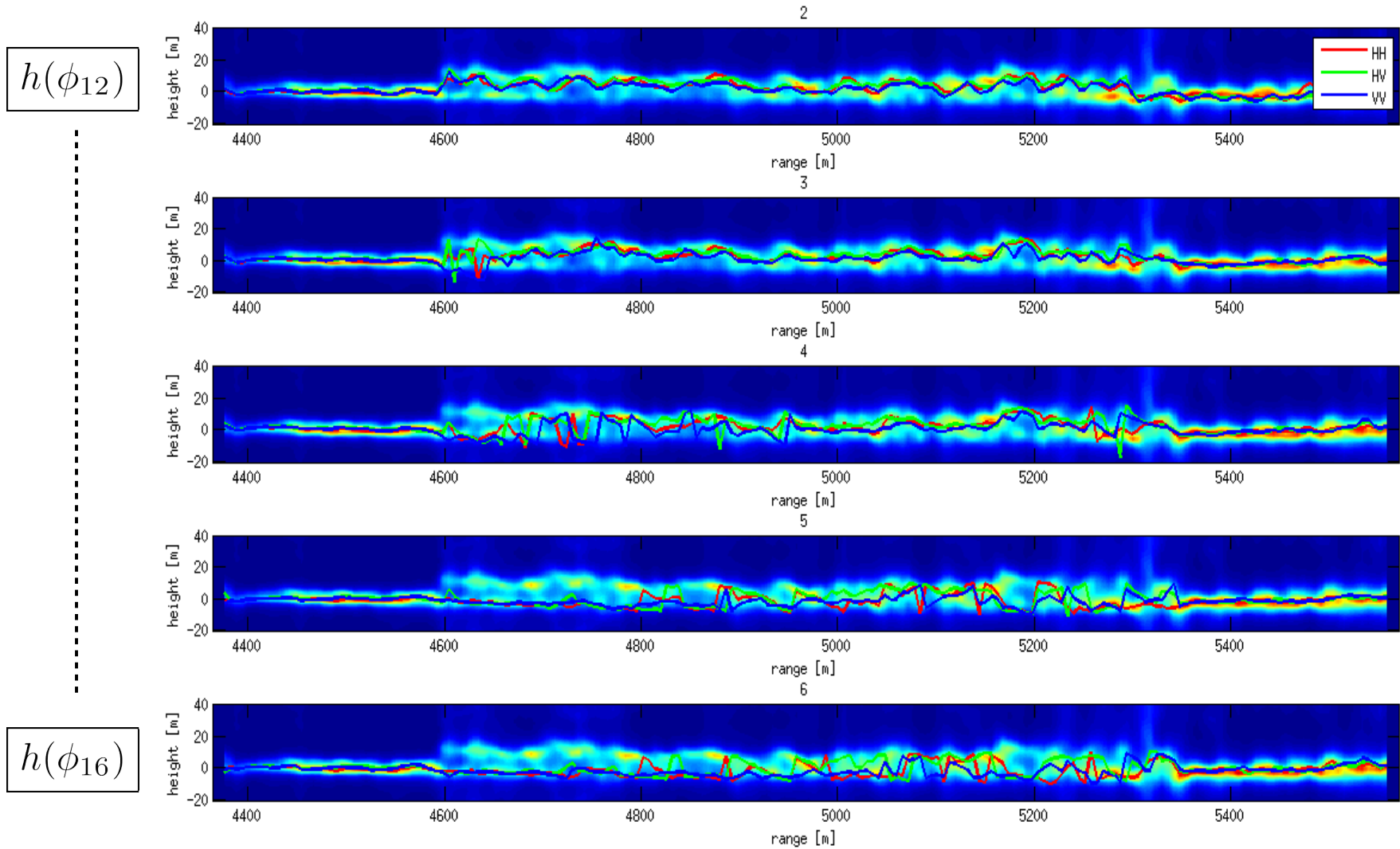
# InSAR phases, polarization & TomoSAR

## L-band BIOSAR2, Capon tomograms



# InSAR phases, polarization & TomoSAR

## InSAR phase center heights



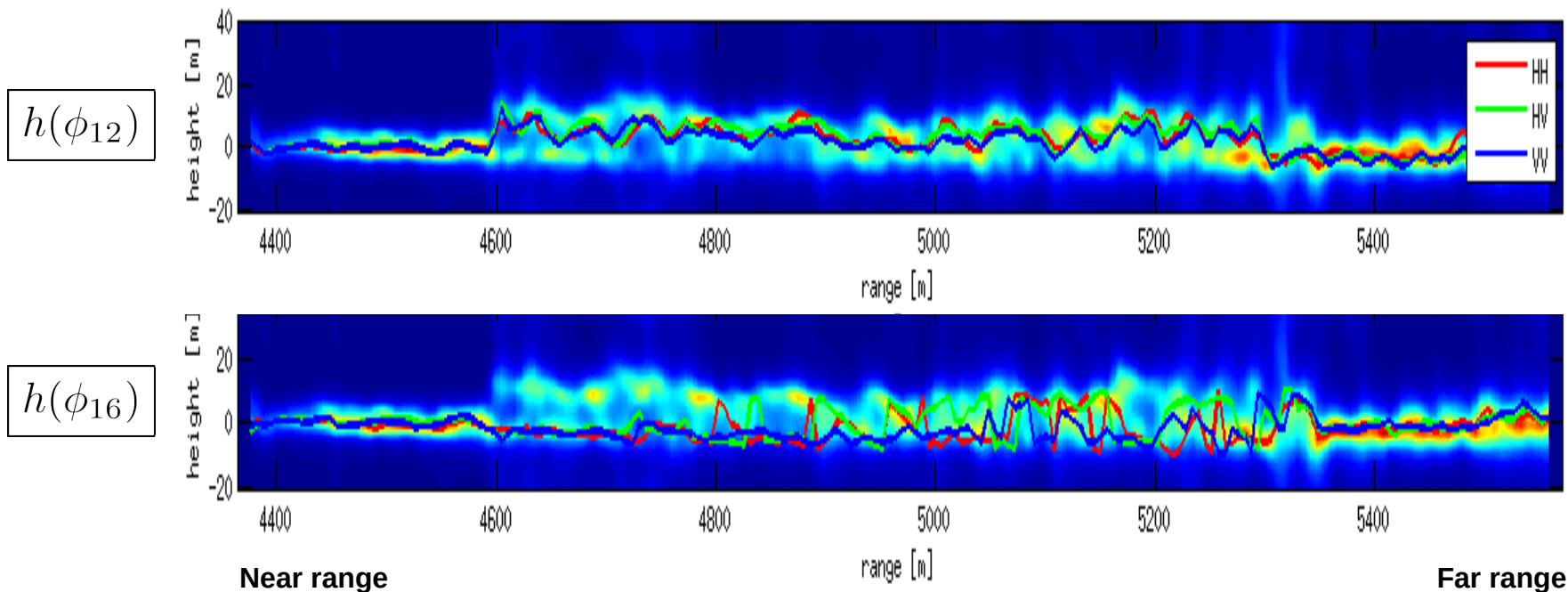
Near Range

$$B_{i+1} \geq B_i$$

Far Range

# InSAR phases, polarization & TomoSAR

## Polarimetric diversity POL-InSAR phase center heights



### Single-baseline PolInSAR:

- Phase Center height diversity not always guaranteed
- Requires **specific  $k_z$**  (baseline) values: **adequate volume decorrelation**

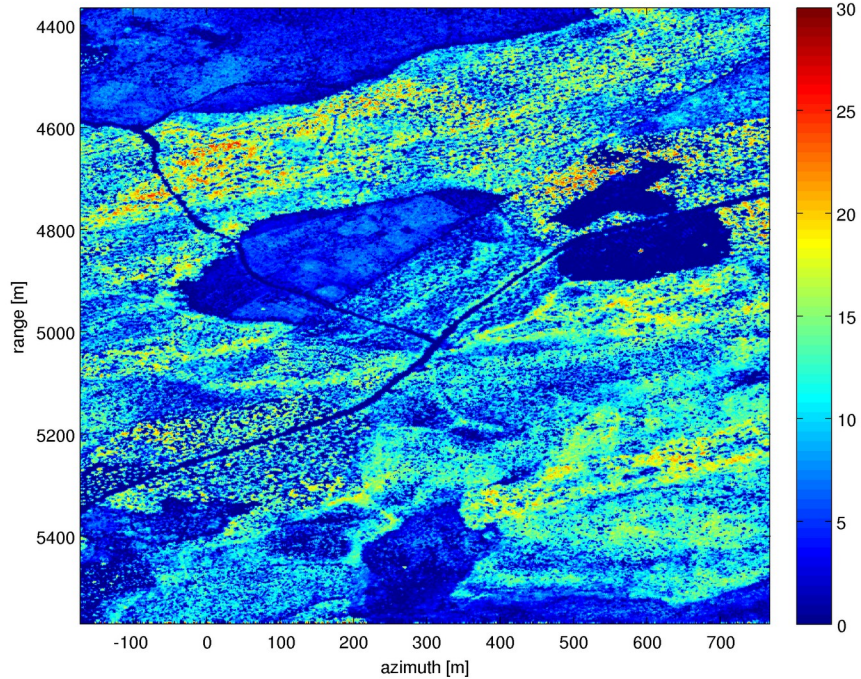
# Illustration of coherence features

<b>Campaign</b>	<b>BioSAR 2008 - ESA</b>
System	E-SAR - DLR
Site	Krycklan river catchment, Northern Sweden
Scene	Boreal forest Pine, Spruce, Birch, Mixed stand
Topography	Hilly
Tomographic Tracks	6 + 6 - Fully Polarimetric (South-West and North- East)
Carrier Frequency	P-Band and L-Band
Slant range resolution	1.5 m
Azimuth resolution	1.6 m
Vertical resolution (P- Band)	20 m (near range) to >80 m (far range)
Vertical	6 m (near range) to 25 m (far

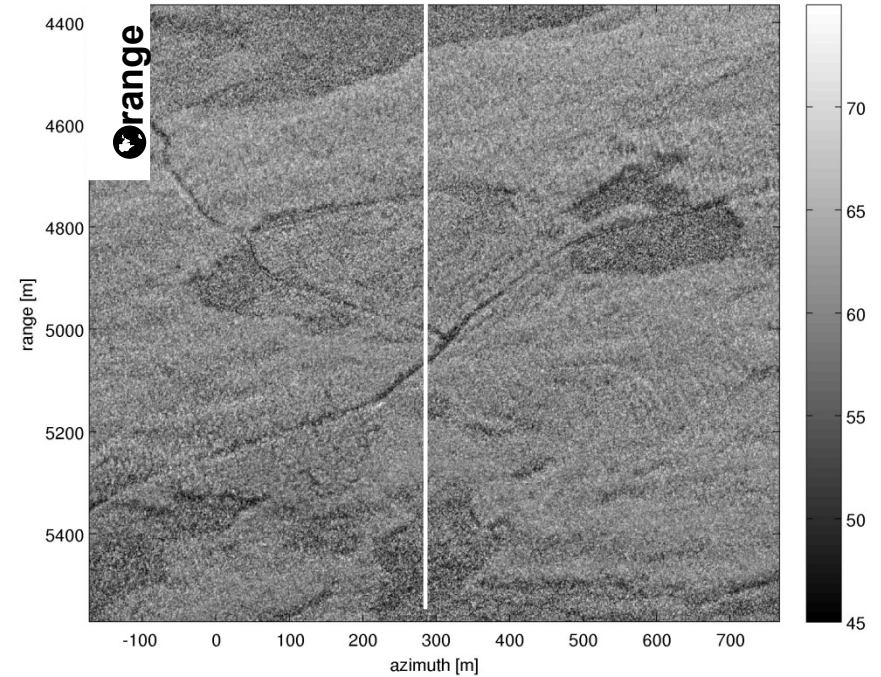


# Illustration of coherence features

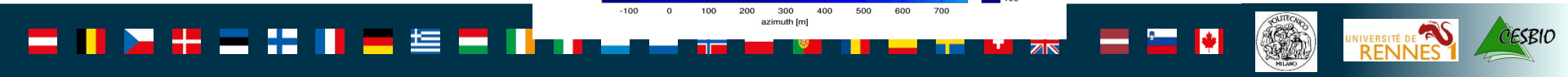
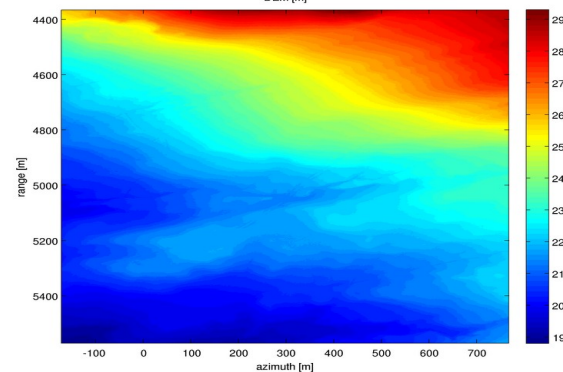
## Forest height



## HH intensity



## DEM

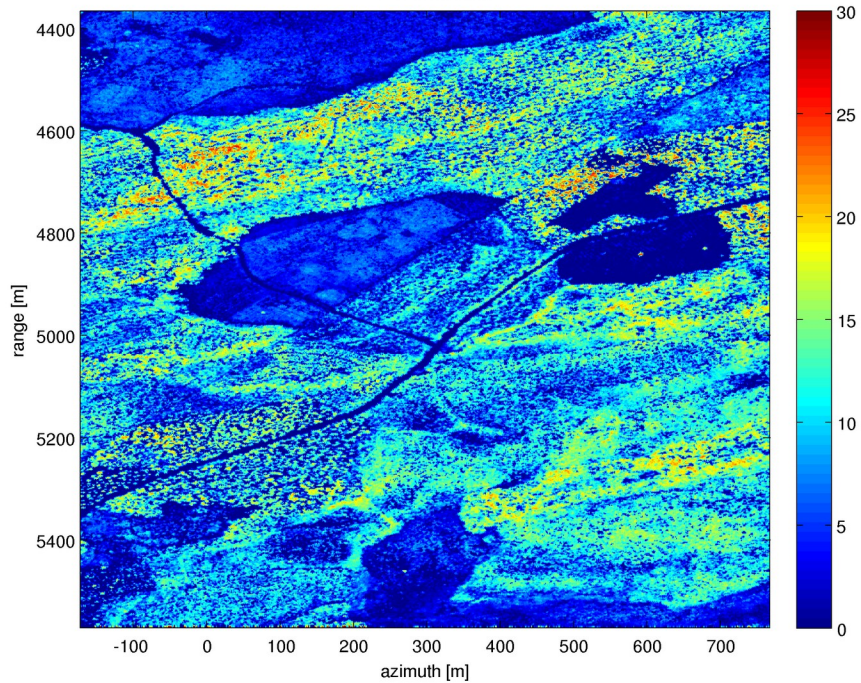




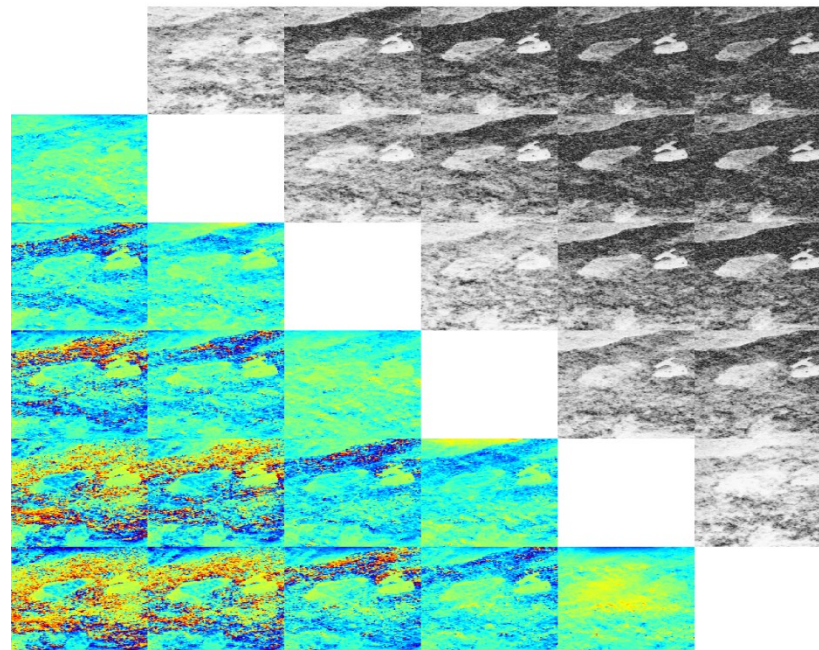
# Illustration of coherence features

## Forest height

Forest height [m]



## Multi-baseline InSAR coherences and phases - VV



## DEM

DEM [m]

