

SAR TOMOGRAPHY

Spectral Analysis (Specan) techniques

Laurent Ferro-Famil

ISAE-SUPAERO, Centre d'Etudes Spatiales de la Biosphère



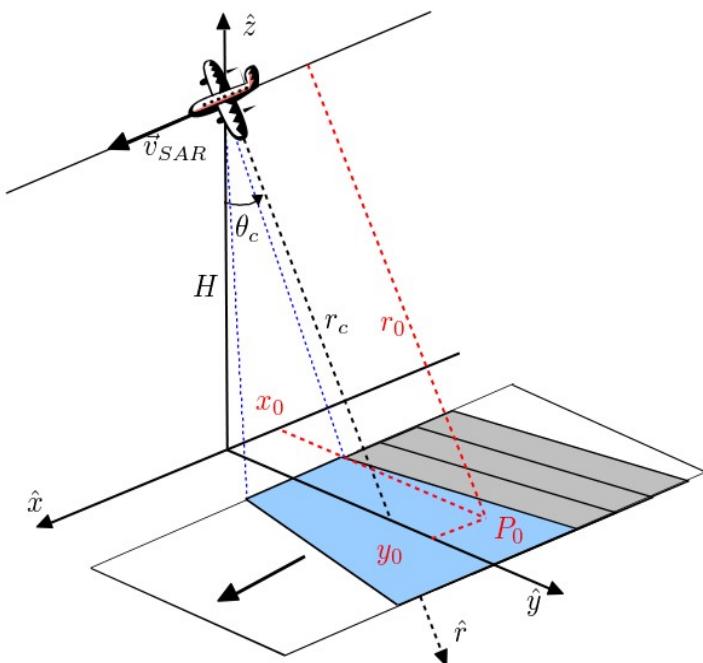
- *From 3-D SAR Imaging to the Beamformer*
- *PolTomSAR imaging using 1D Specan techniques*
- *Advanced PolTomSAR imaging using Specan*
- *Polarimetric TomoSAR tomography*
Full-Rank specan & and SKP decomposition
- *Spaceborne 3-D imaging using correlation SAR tomography*



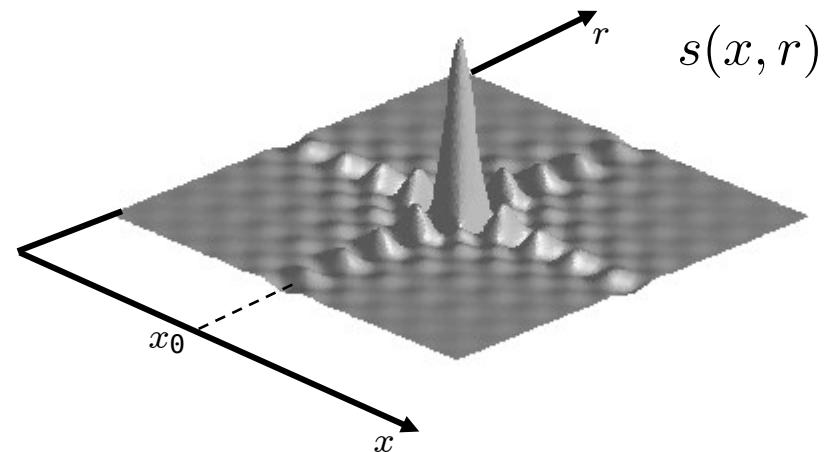
From 3-D Synthetic Aperture Imaging To the Beamformer



2-D SAR impulse response



2-D focused signal (x-r domain)



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

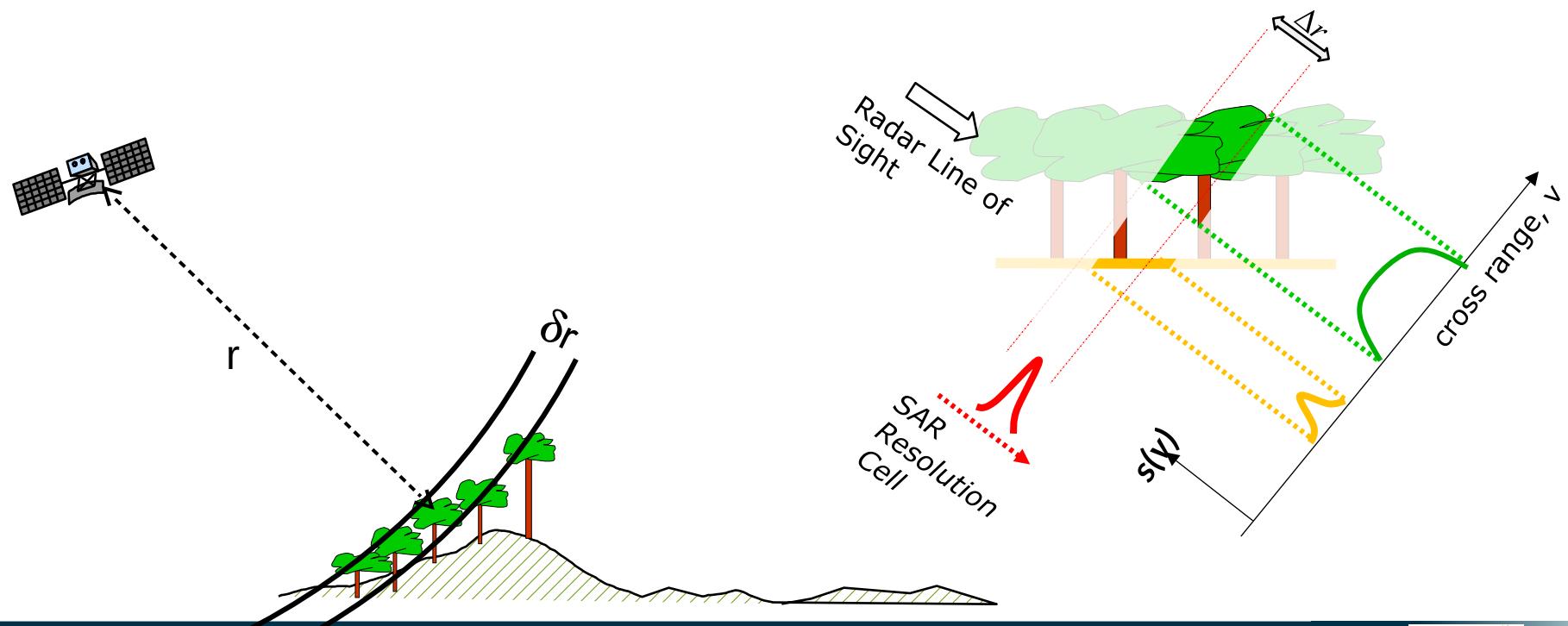
- \rightarrow **complex reflection coefficient**
- \rightarrow **delayed range impulse response**
- \rightarrow **delayed azimuth impulse response**
- \rightarrow **two-way propagation phase**

2-D SAR imaging

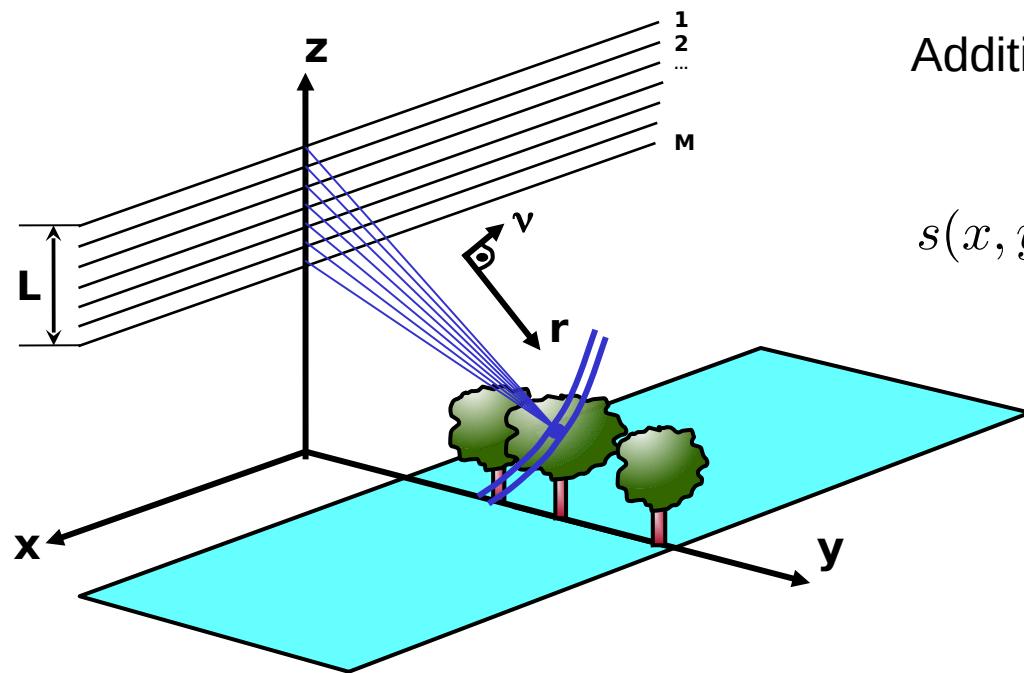
SAR imaging: coherent integration of a reflectivity density

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

$$s(x, r) \approx \int_{\mathcal{C}} a_c(x, r, \nu) e^{-j k_c r(\nu)} d\nu$$

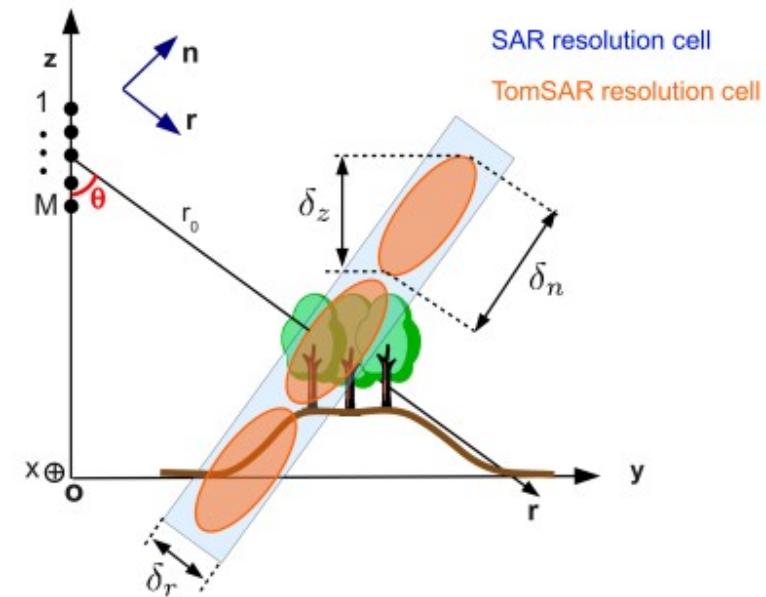


3-D SAR imaging



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

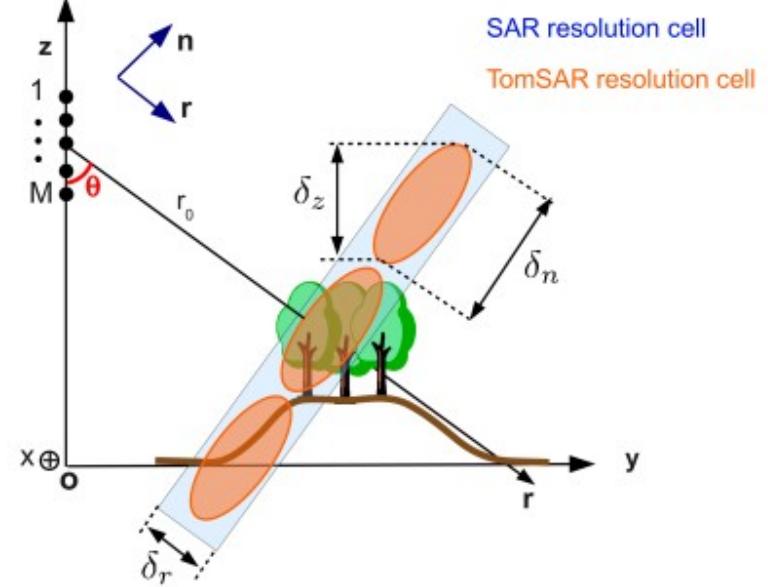
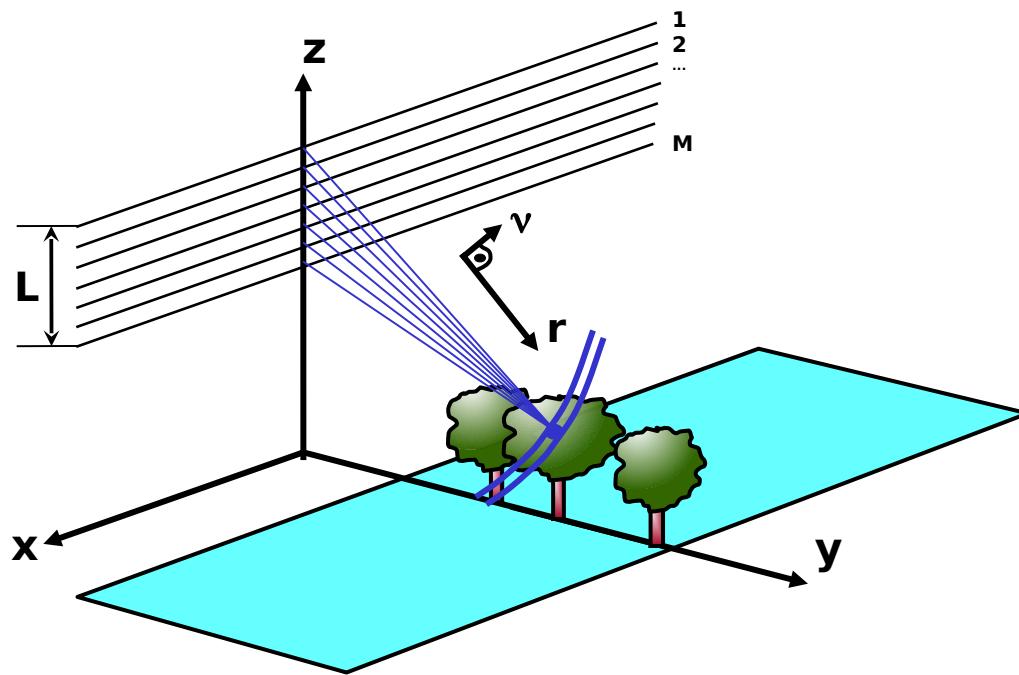
$$s(x, y, z) = \sum_{i=1}^M s_i(x, r(y, z)) e^{j k_c r(y, z)}$$



Vertical aperture : L_{tomo}

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

3-D SAR imaging



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

Interpolation

HF term

3-D SAR imaging

Co-registration on a reference plane

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

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

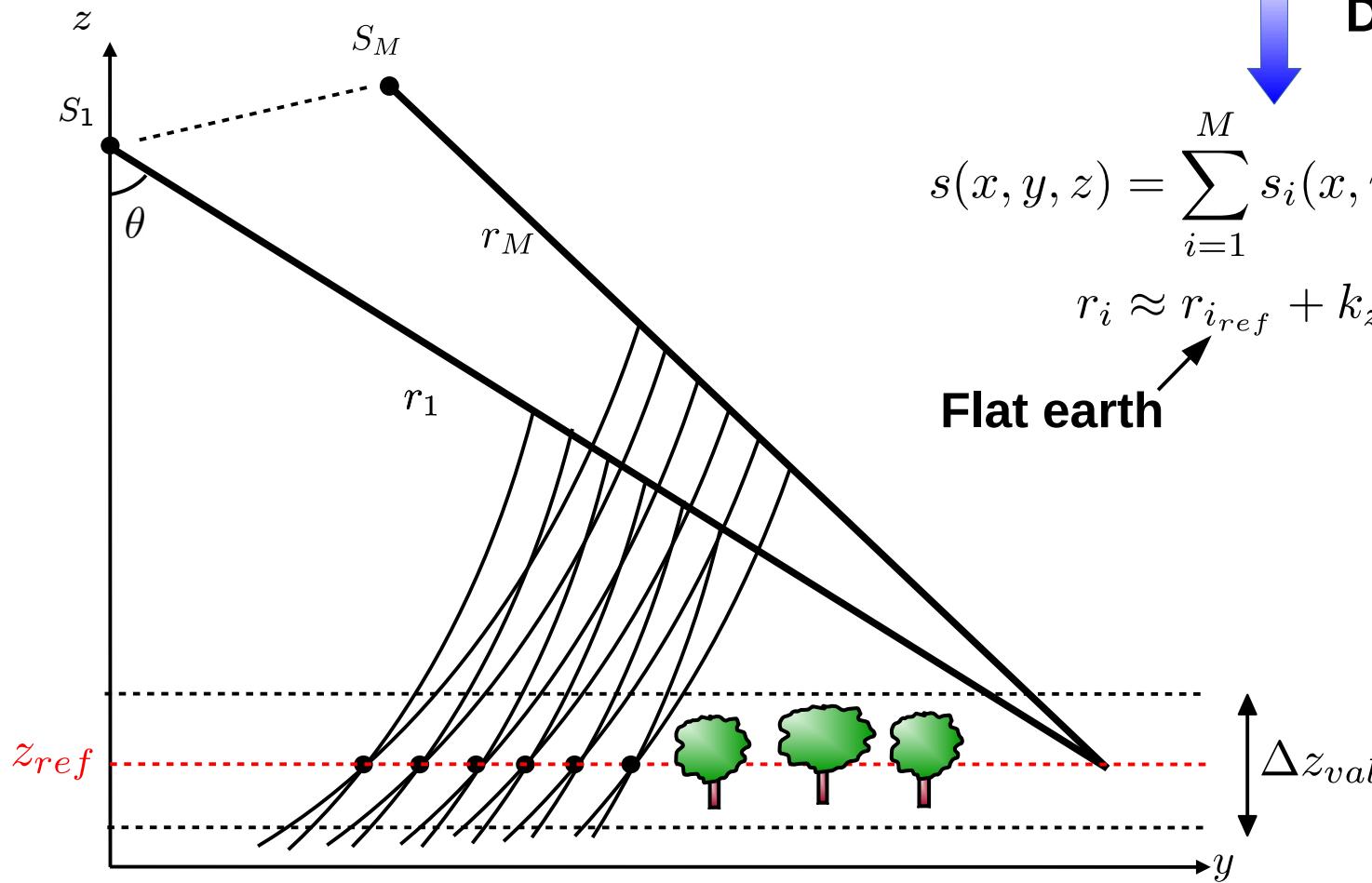
**Discretization
NN-interp.**

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

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

Flat earth

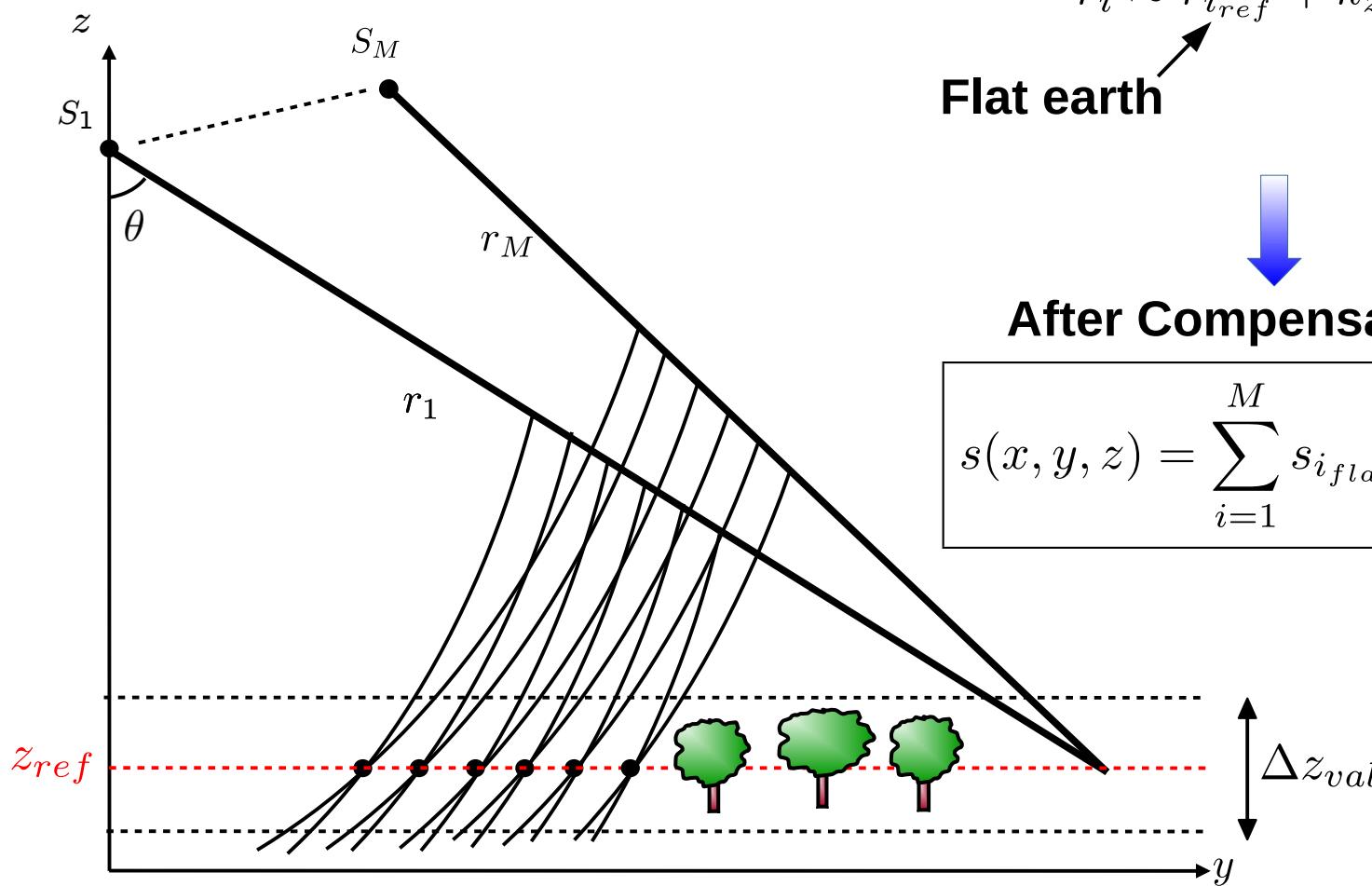
Elevation



3-D SAR imaging

Co-registration on a reference plane

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



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

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

Flat earth **Elevation**

After Compensation

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

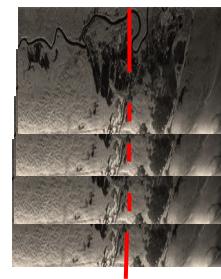
3-D SAR imaging: 2D + 1D processing

3-D Synthetic Aperture imaging

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

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

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



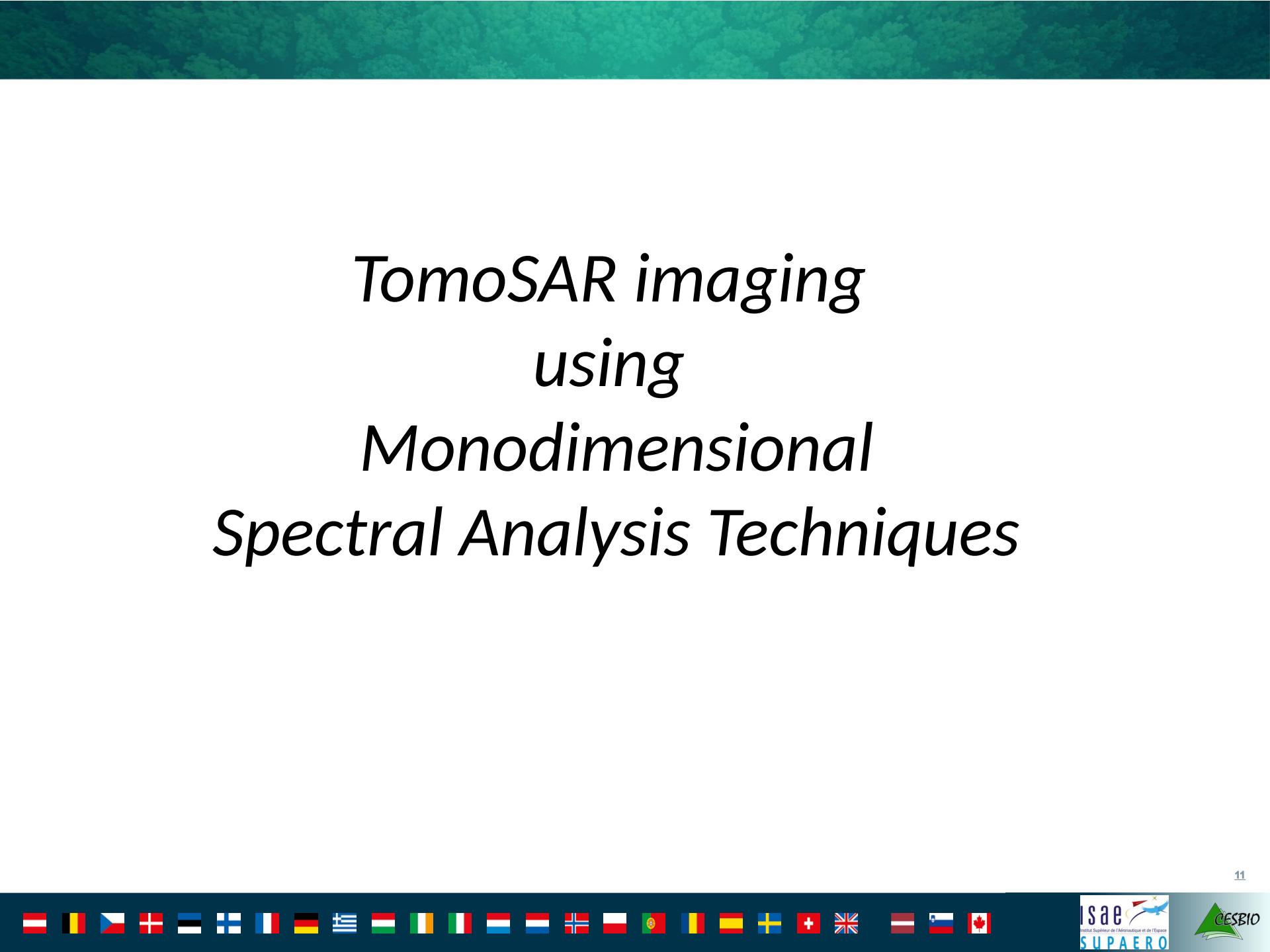
$$\Rightarrow \mathbf{y} = \begin{bmatrix} y_1 \\ \vdots \\ y_M \end{bmatrix} = \begin{bmatrix} s_1(x, r_{1_{ref}}) \\ \vdots \\ s_M(x, r_{M_{ref}}) \end{bmatrix}$$

1D Linear filter

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

Steering vector

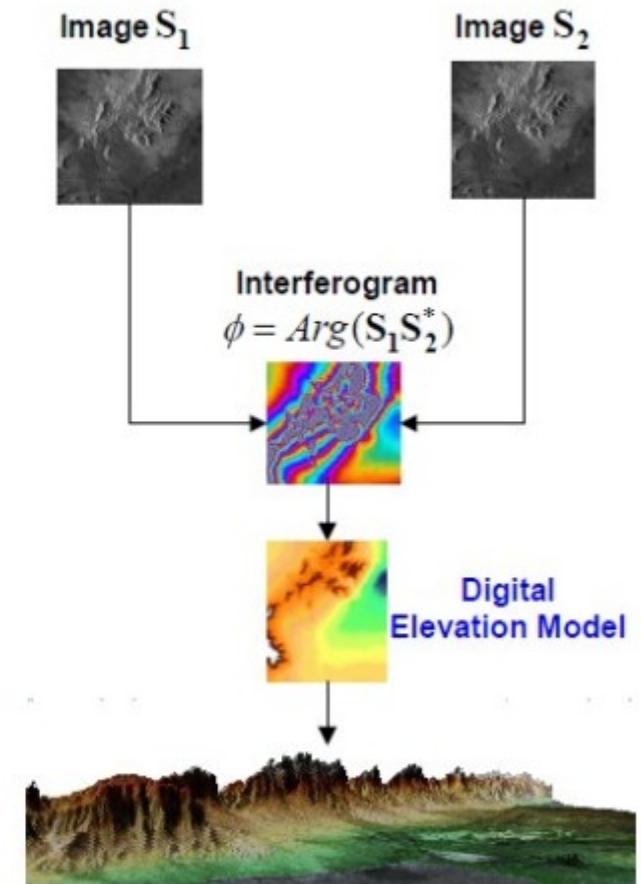
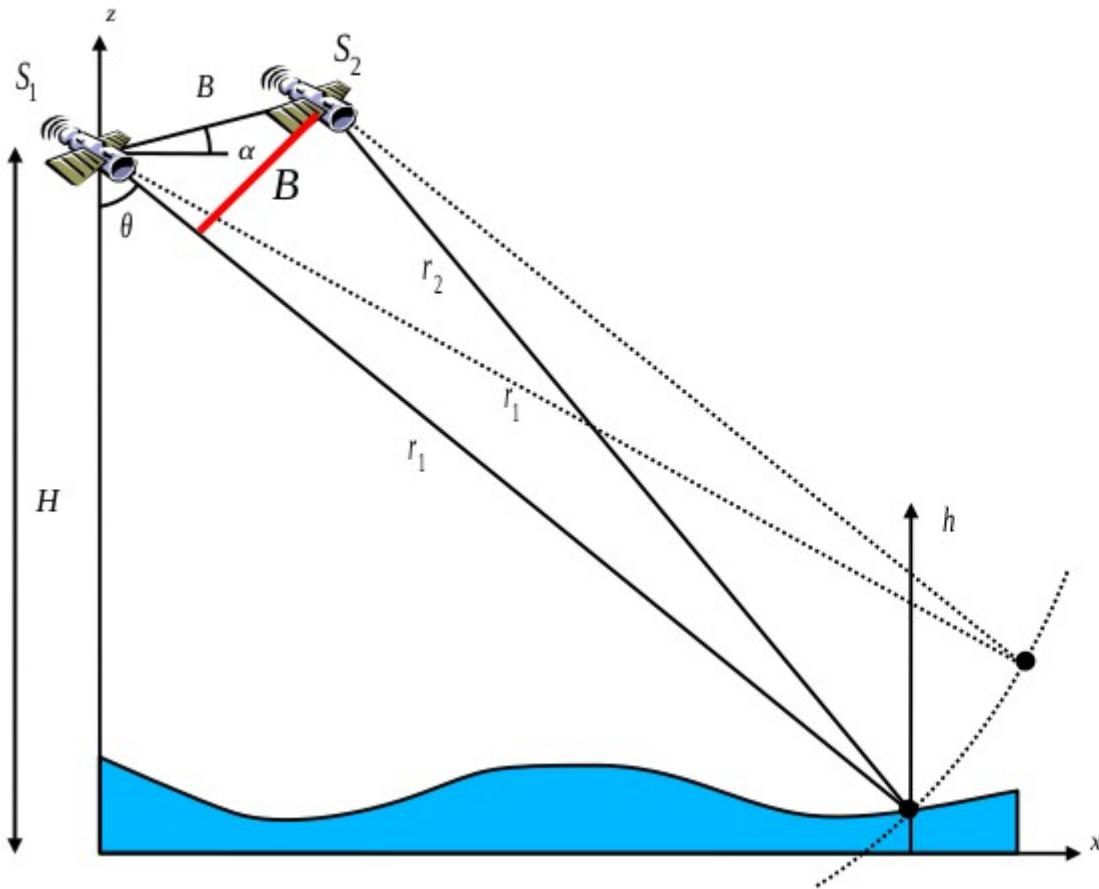
$$\mathbf{a} = [1, e^{j k_{z_2} z}, \dots, e^{j k_{z_M} z}]^T$$

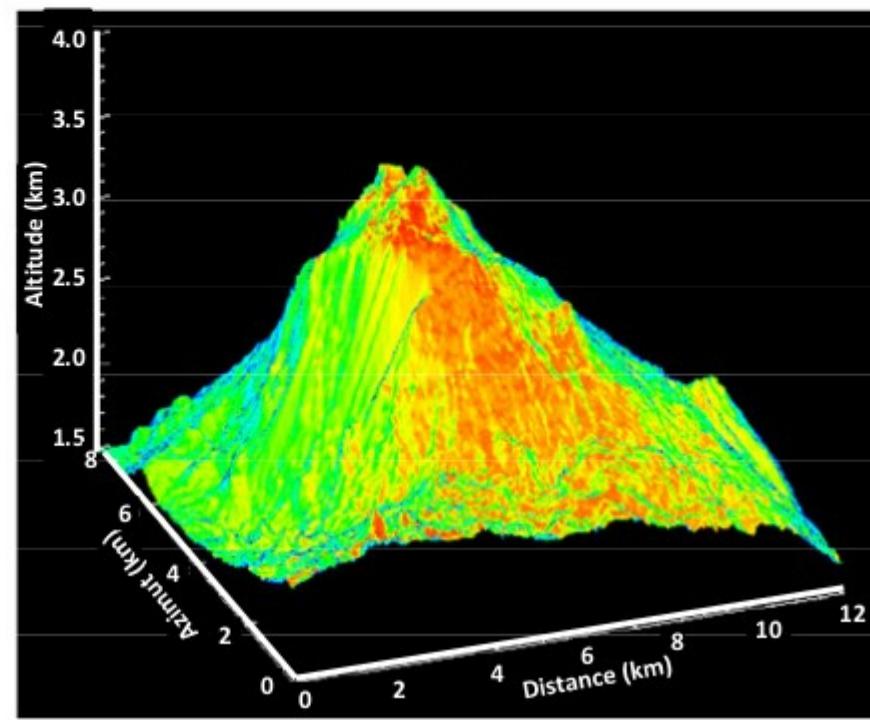
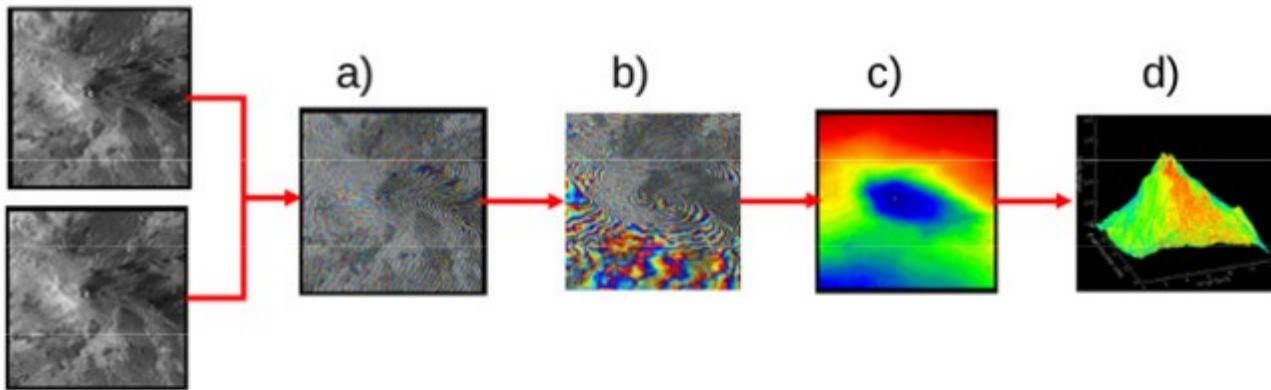


TomoSAR imaging using Monodimensional Spectral Analysis Techniques



Interferometric phase variations with height





Estimation of a single scatterer, M=2 images

InSAR way

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

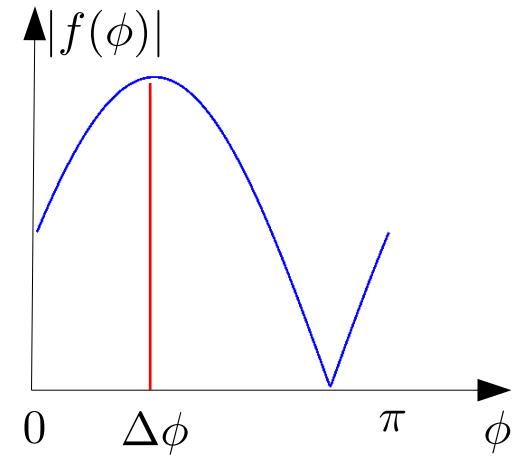
Linear filtering way

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

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

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

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



Estimation of several scatterers, M>2 images

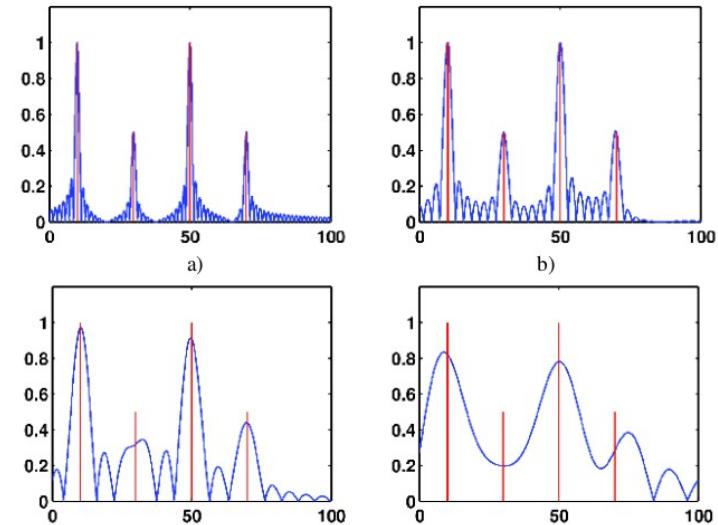
Estimation of several scatterers: MB InSAR way

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

Estimation of several scatterers: linear filtering way

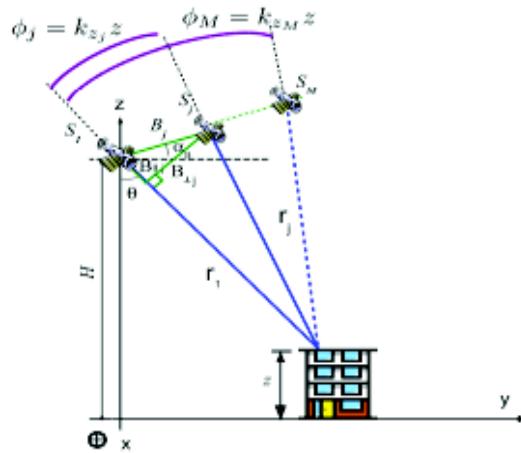
$$\mathbf{y} = \begin{bmatrix} s_1 \\ \vdots \\ s_M \end{bmatrix}, \mathbf{a}(z) = \begin{bmatrix} 1 \\ \vdots \\ e^{jk_{z_M} z} \end{bmatrix}$$
$$f(z) = \frac{\mathbf{a}^H(z)\mathbf{y}}{M} = \frac{\sum_m s_m e^{-jk_{z_m} z}}{M}$$

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



- 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

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

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

Spectral sampling: $\Delta k_z = \frac{k_c B_{\perp}}{r \sin \theta}$

Spectral bandwidth: $\Delta k_z = M \Delta k_z$

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

Fast
M times Slower

Periodic oscillating filter output

Tomographic imaging using specan

Uniform baseline sampling

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

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

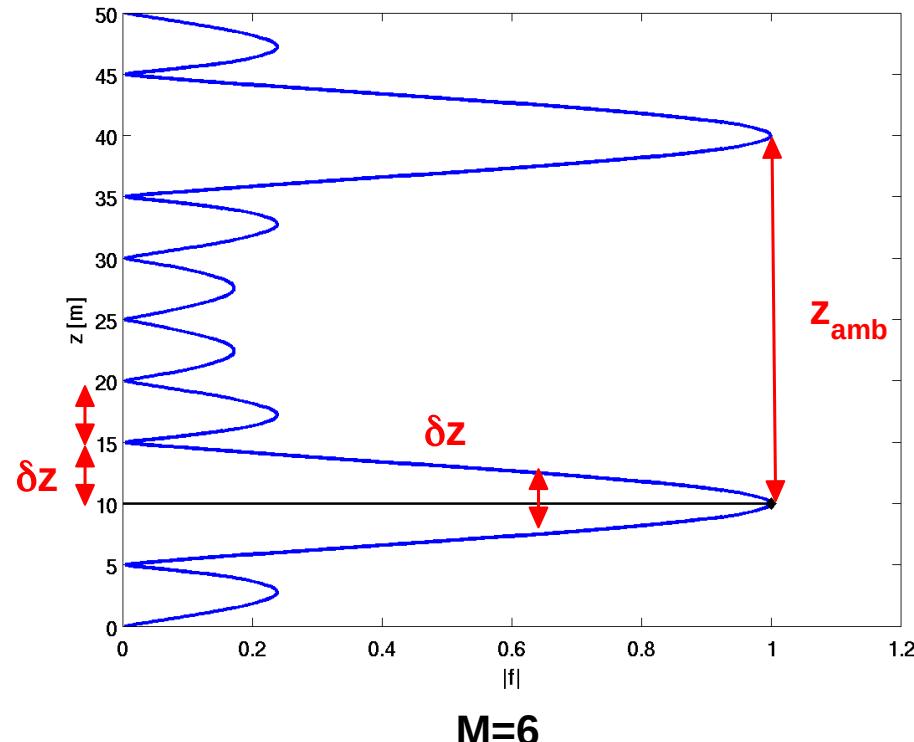
Fast → resolution

Slow → ambiguity

Spatial features of a tomogram

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

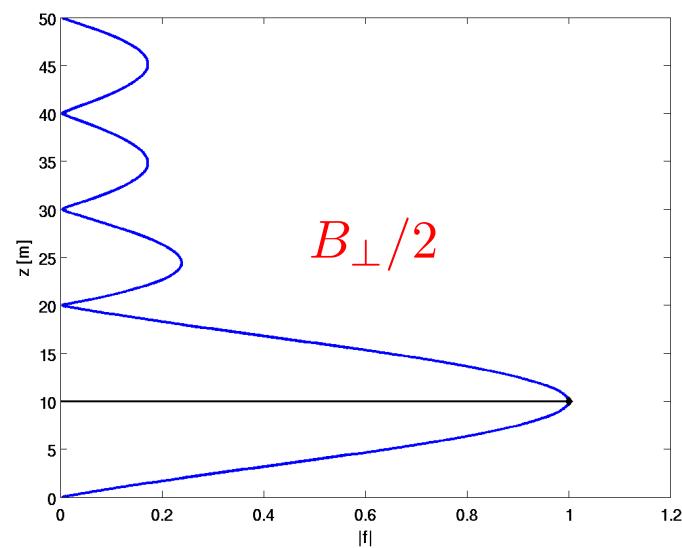
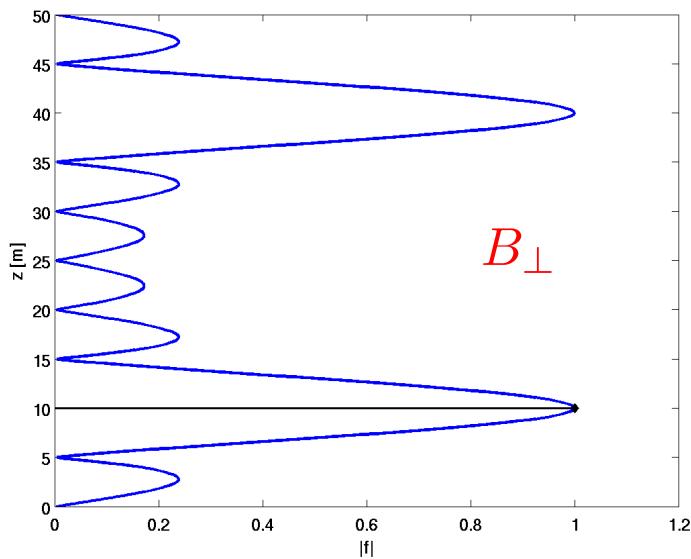
$$\delta z = \frac{2\pi}{\Delta k}, z_{amb} = \frac{2\pi}{dk}, \delta z = \frac{z_{amb}}{M}$$



Tomographic imaging using specan

M=6

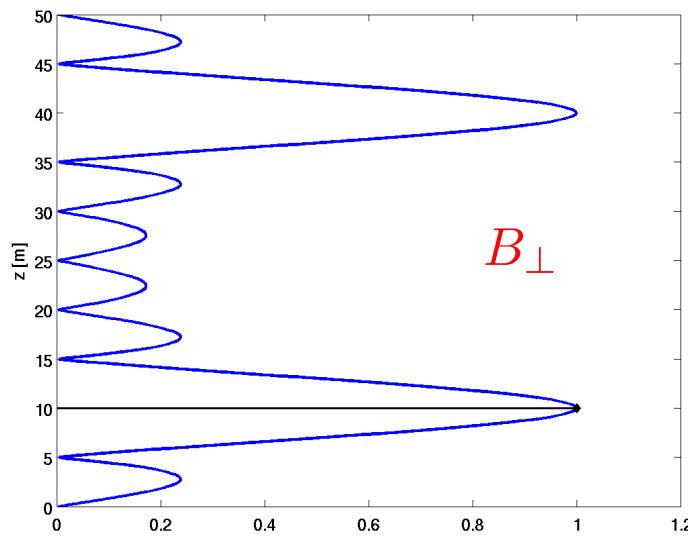
$$\Delta k_z \propto MB_{\perp}$$
$$dk_z = \frac{\Delta k_z}{M}$$



- Reduced resolution
- Improved ambiguity

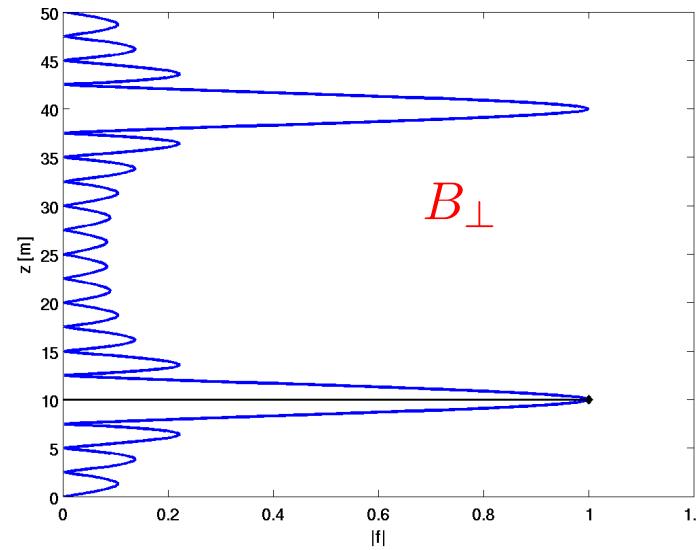
Tomographic imaging using specan

M=6



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

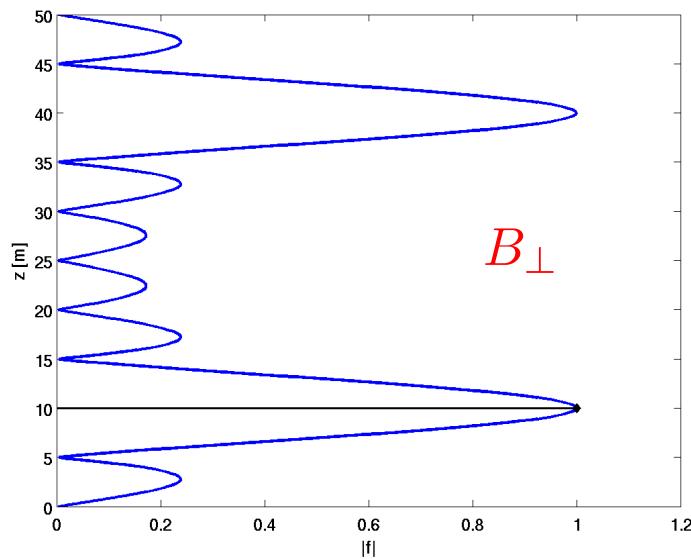
M=12



- Improved resolution
- Unchanged ambiguity

Tomographic imaging using specan

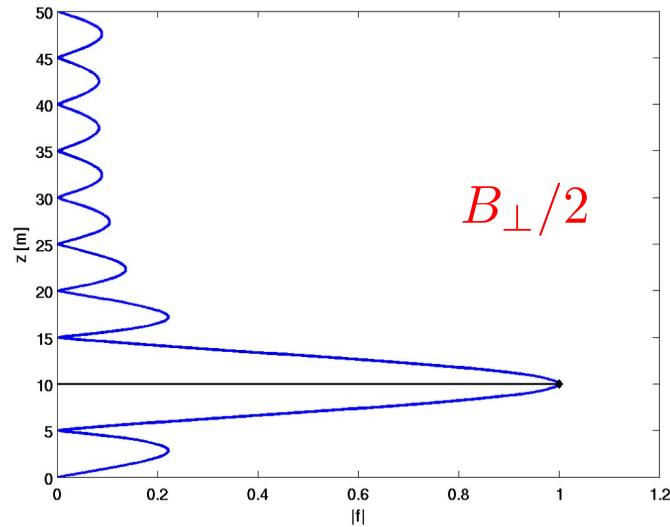
M=6



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

M=12

- **Unchanged resolution**
- **Improved ambiguity**

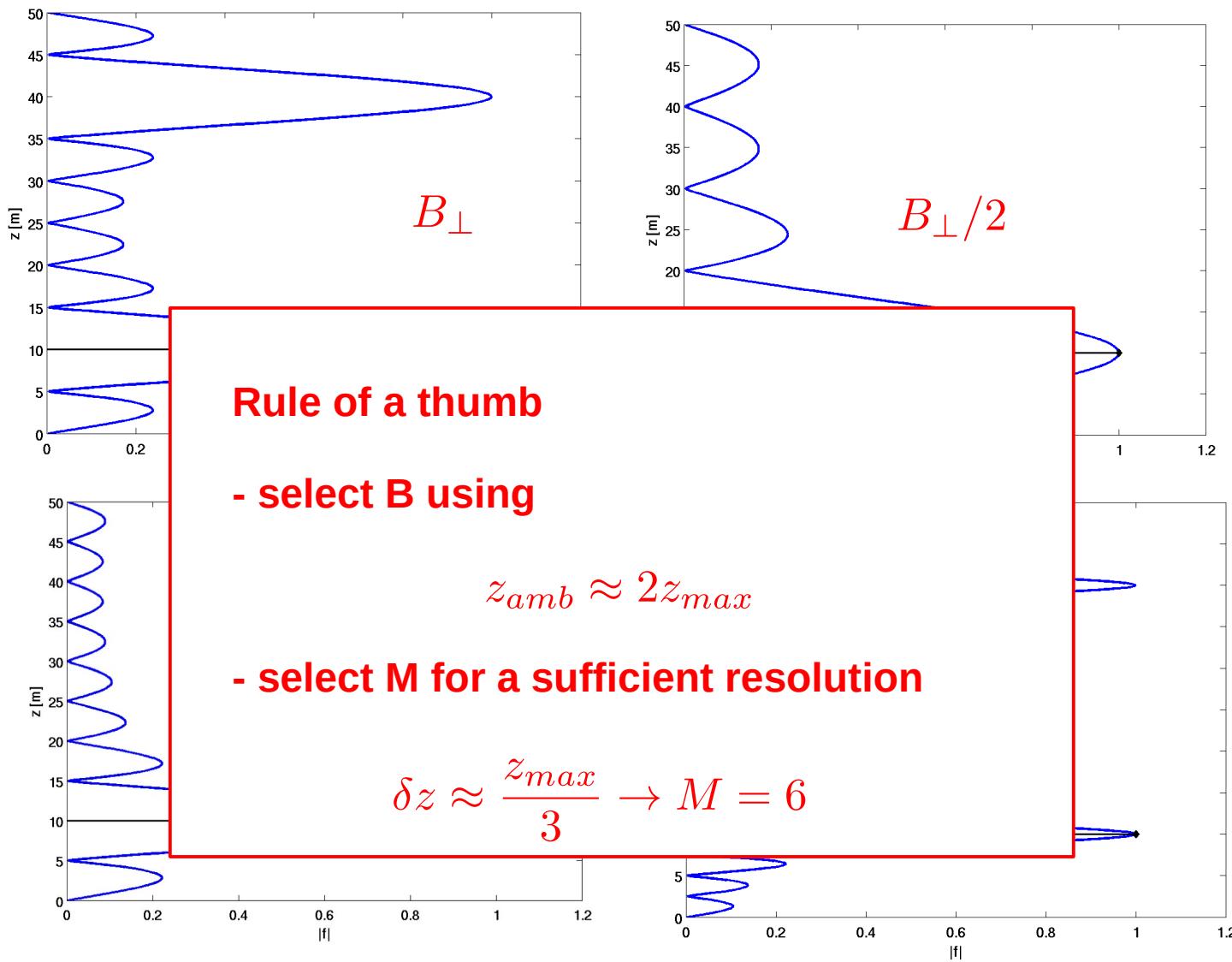


Tomographic imaging using specan

M=6

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

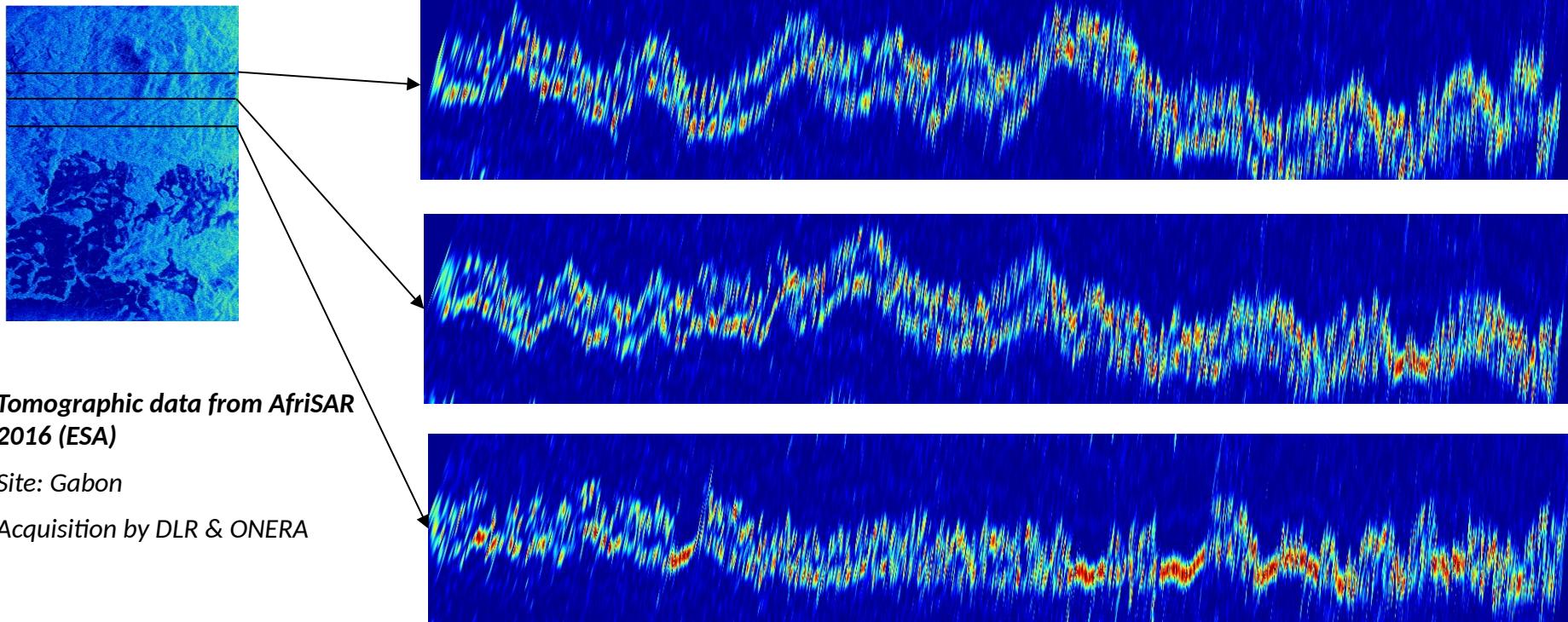
M=12



Tomographic imaging using specan

Single-look tomograms

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



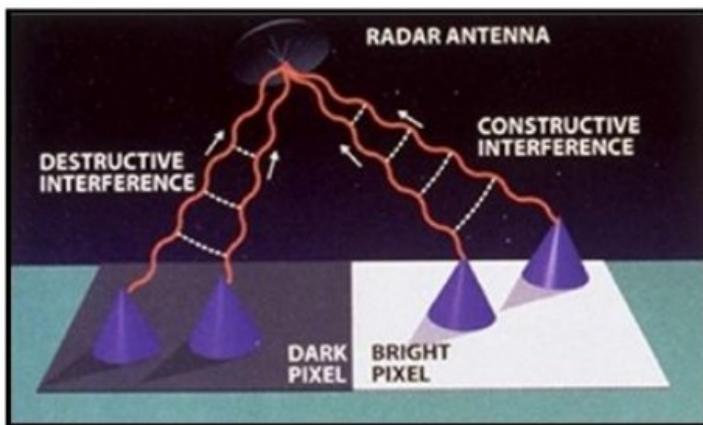
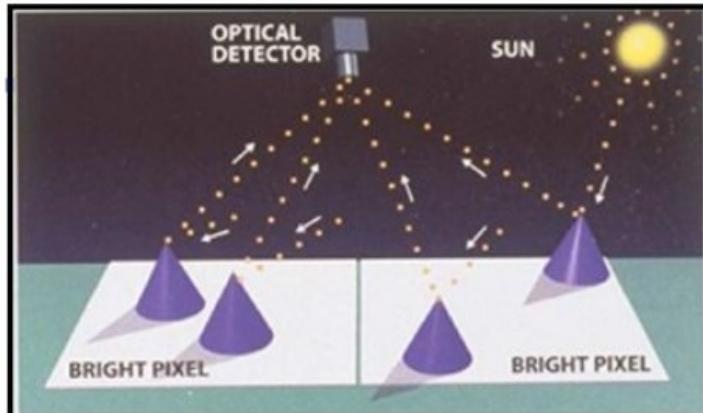
Noisy aspect due to speckle

TomoSAR imaging

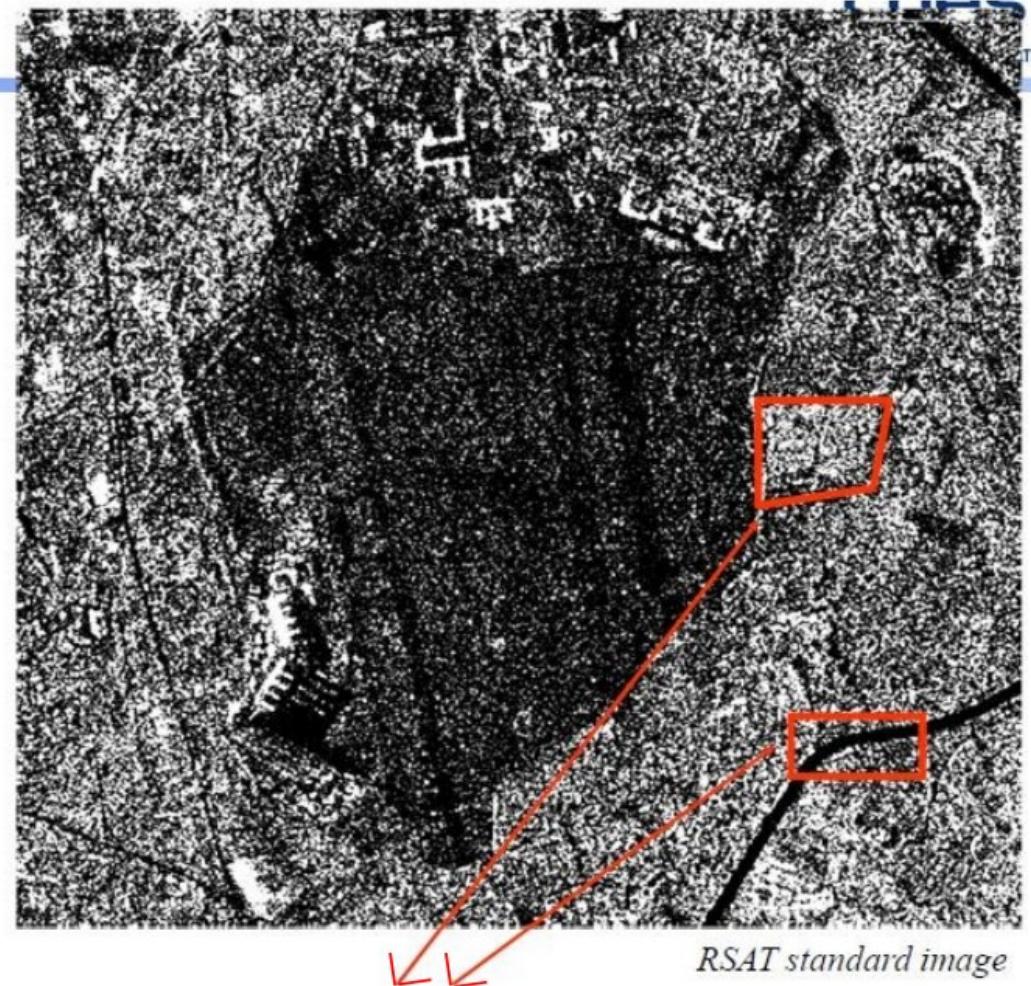
Using multilook

Specan methods

Speckle effect



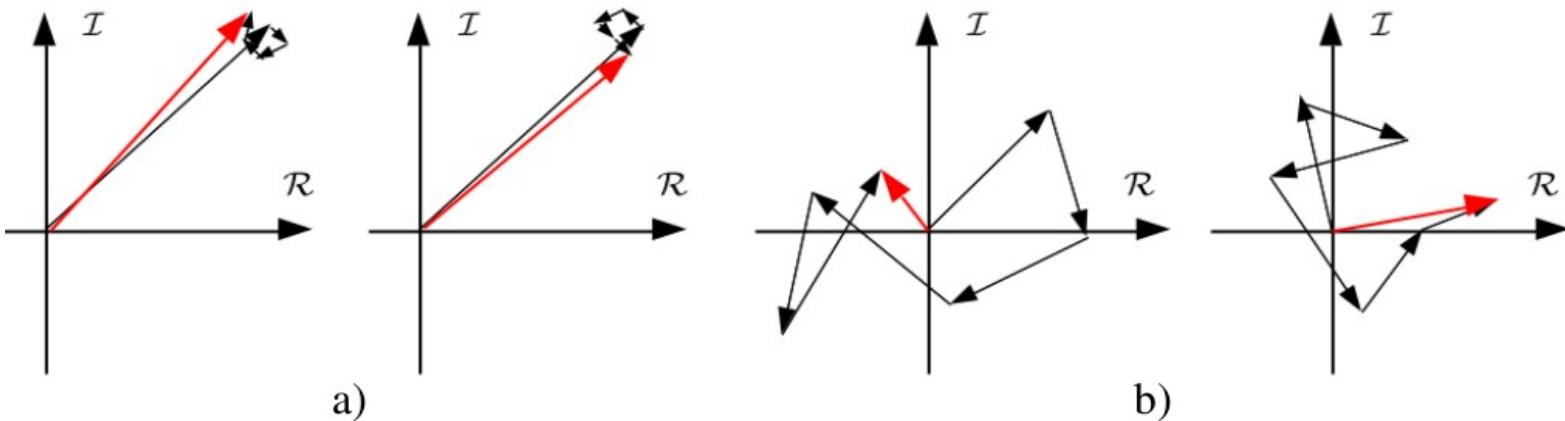
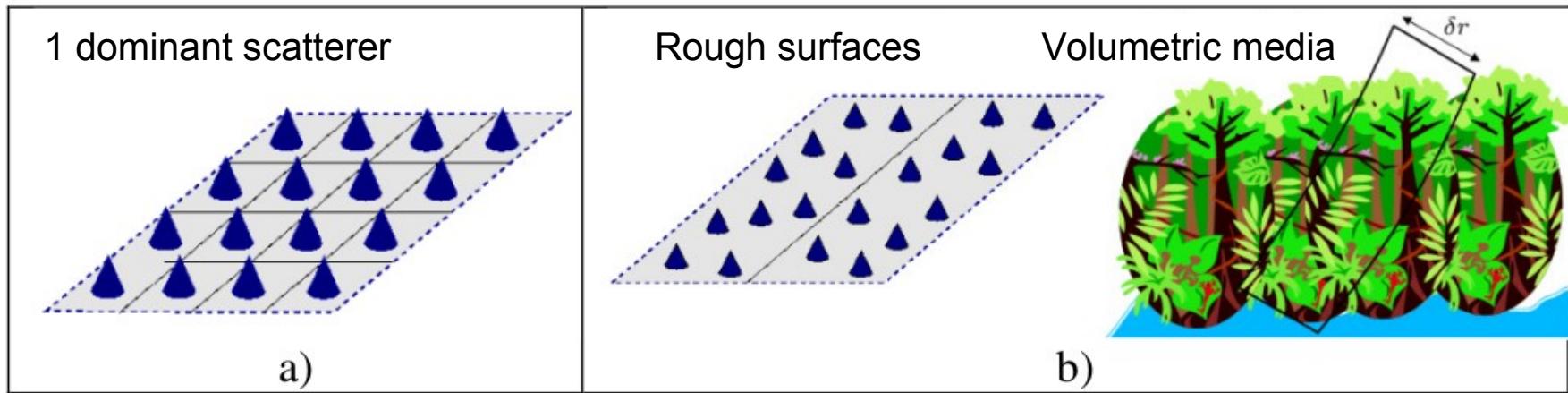
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Speckle: coherent effect that appears as a Multiplicative noise

Speckle effect

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



Two realizations in both cases

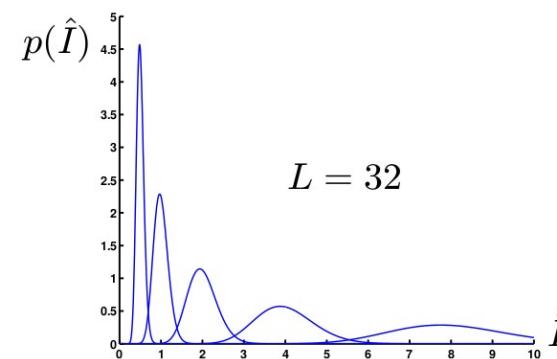
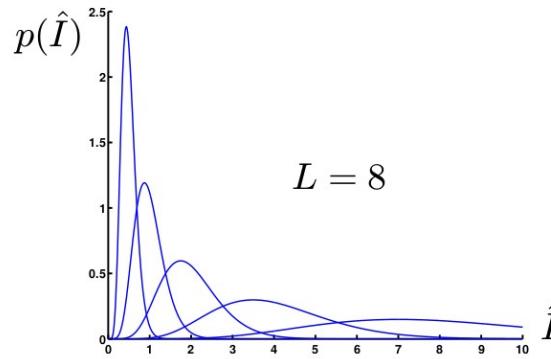
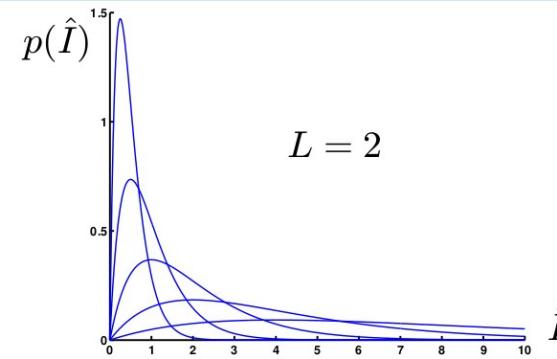
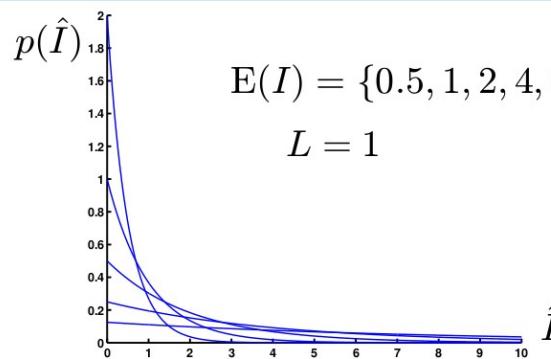
Speckle filtering

Unfiltered intensity image: exponential distribution

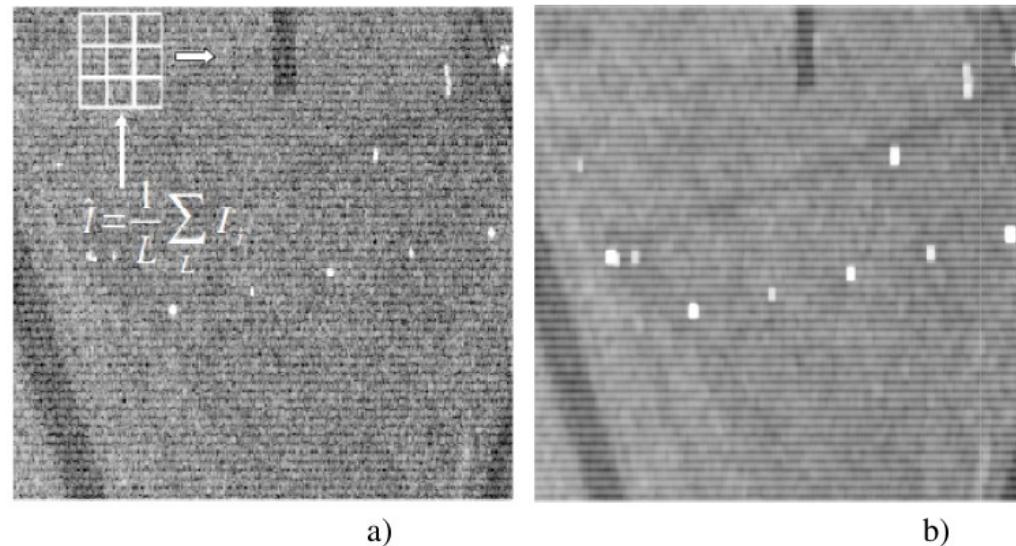
$$\hat{I} = |s(l)|^2, \quad \text{E}(\hat{I}) = I, \quad \text{var}(\hat{I}) = I^2$$

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

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

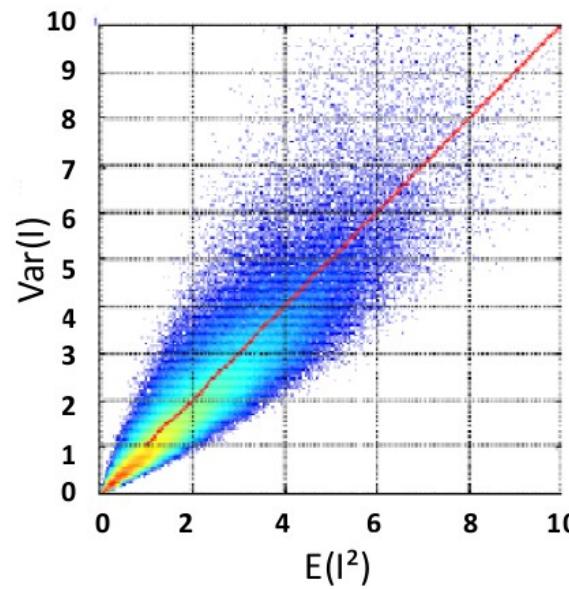


Speckle filtering



Equivalent Number of Looks

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

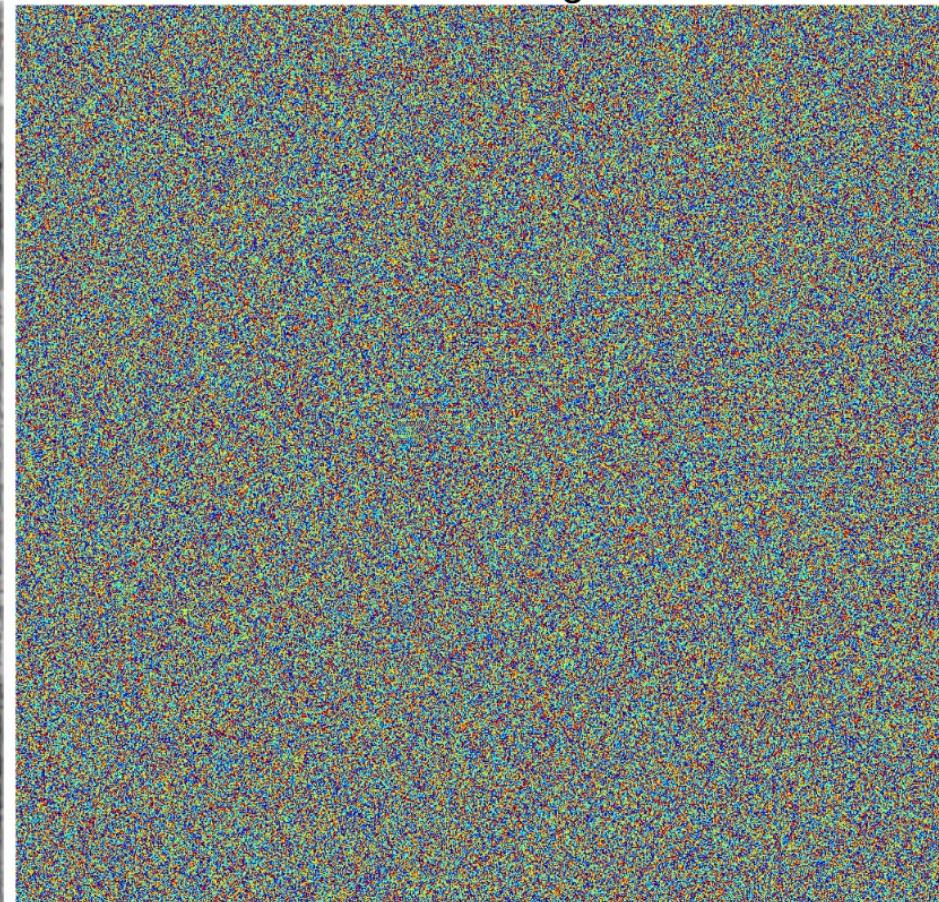


Speckle filtering

Intensity image



Phase image

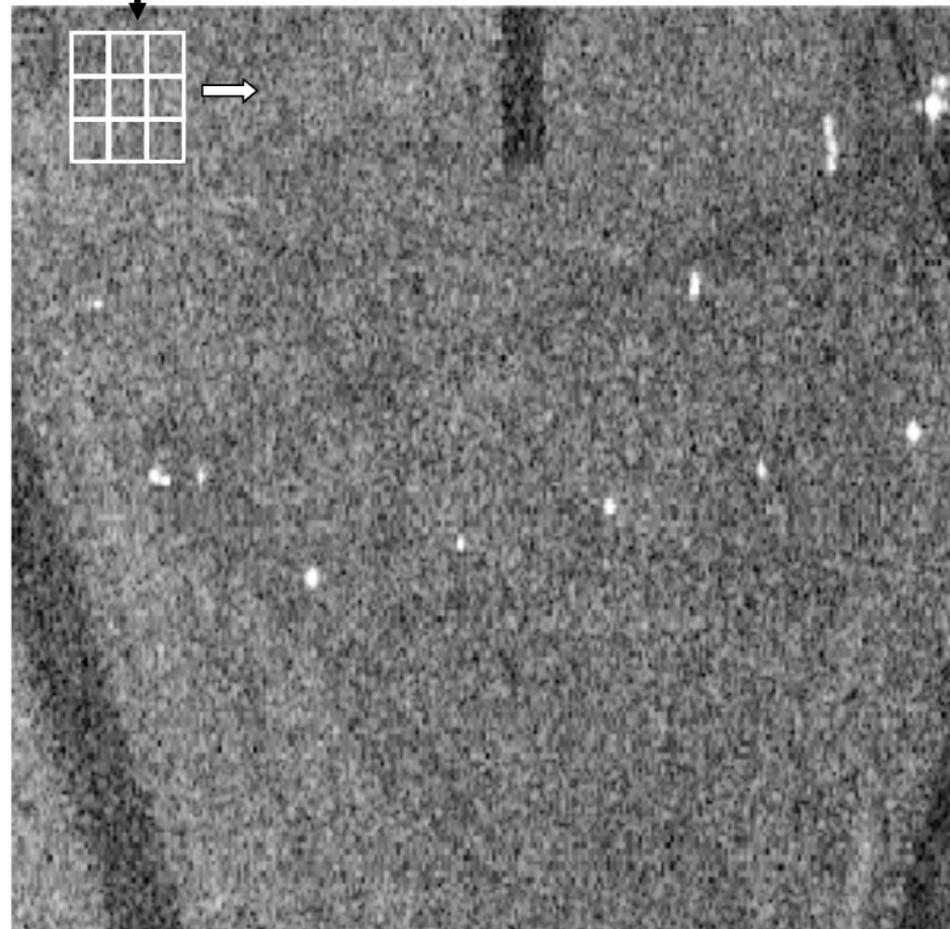
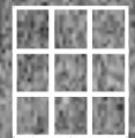


Speckle filtering

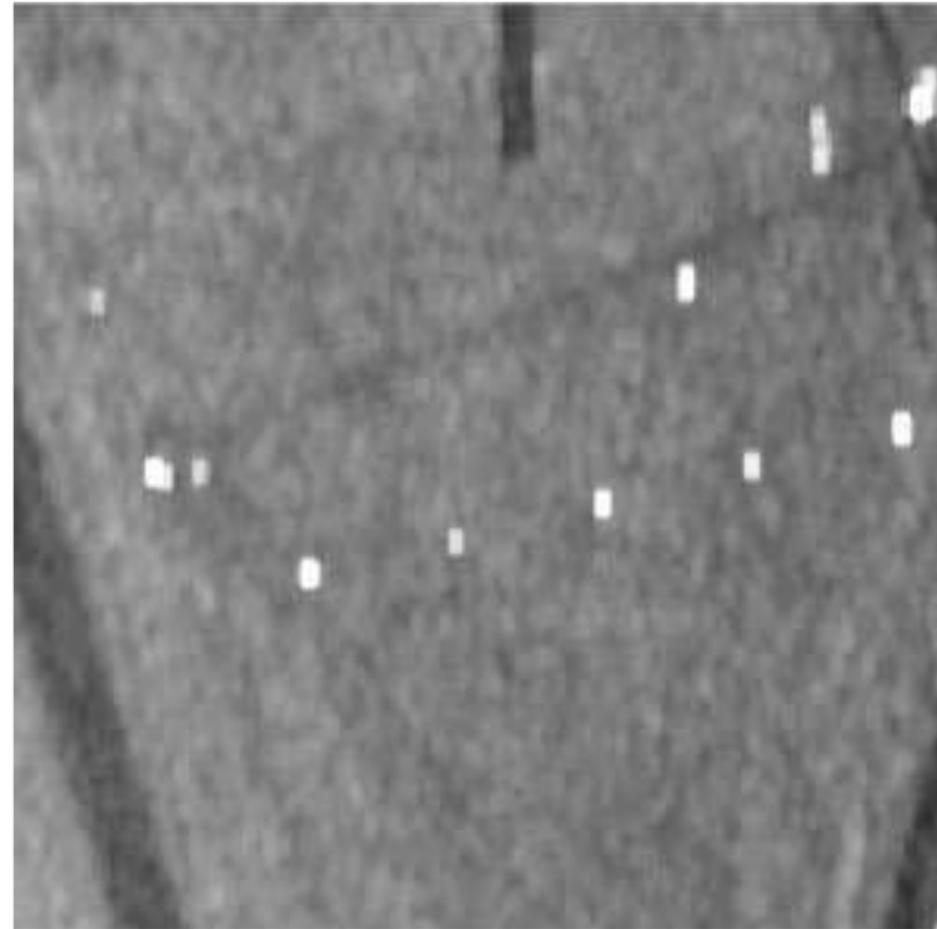
Intensity images

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

Single look image

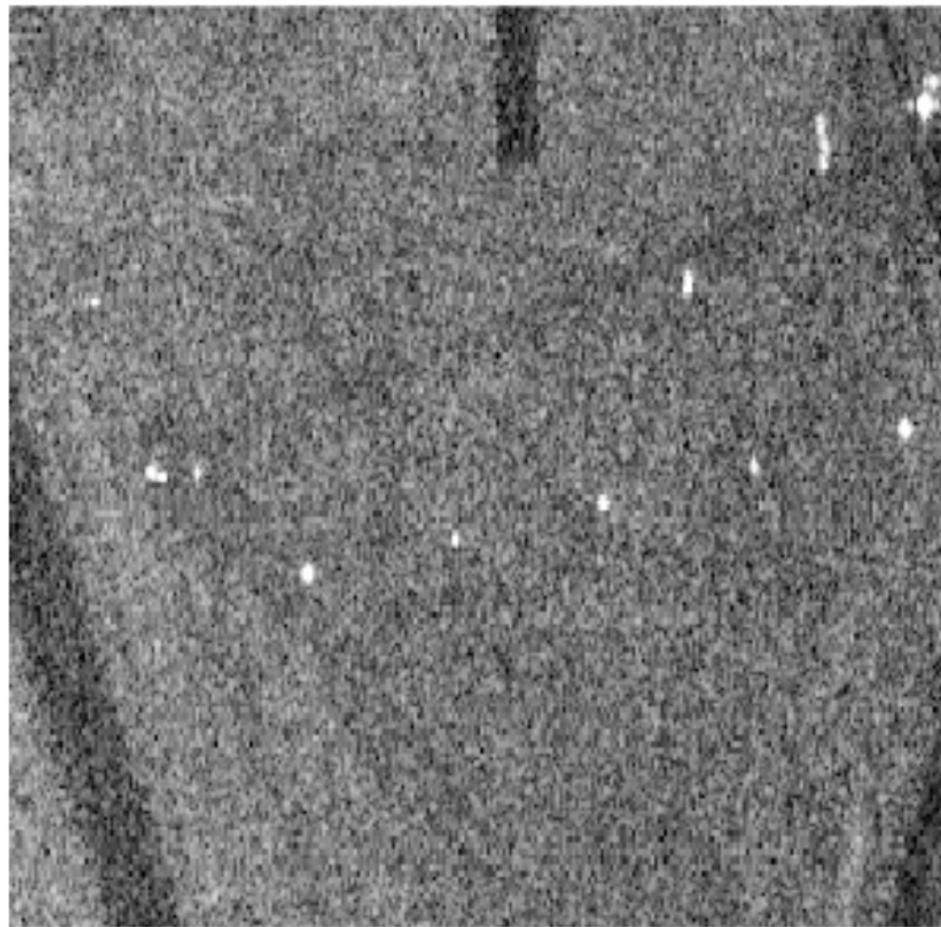


After spatial filtering (N*N boxcar)

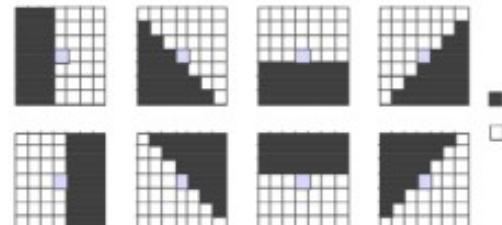
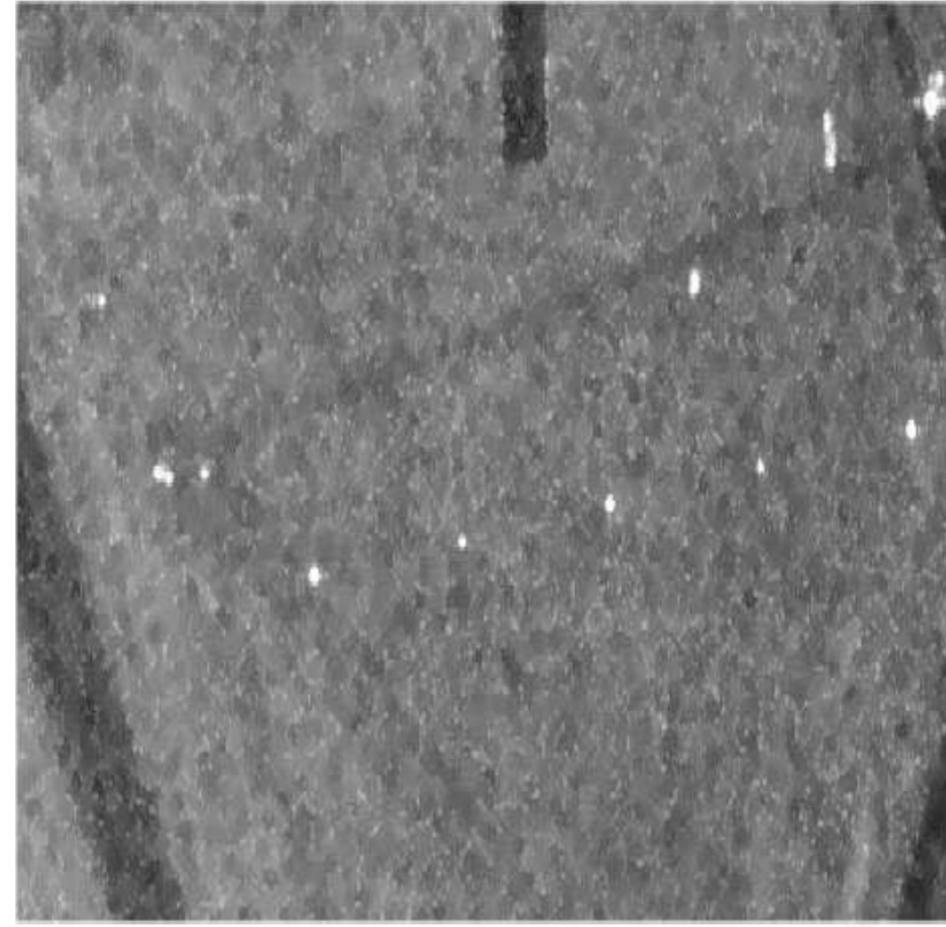


Speckle filtering

Single look image

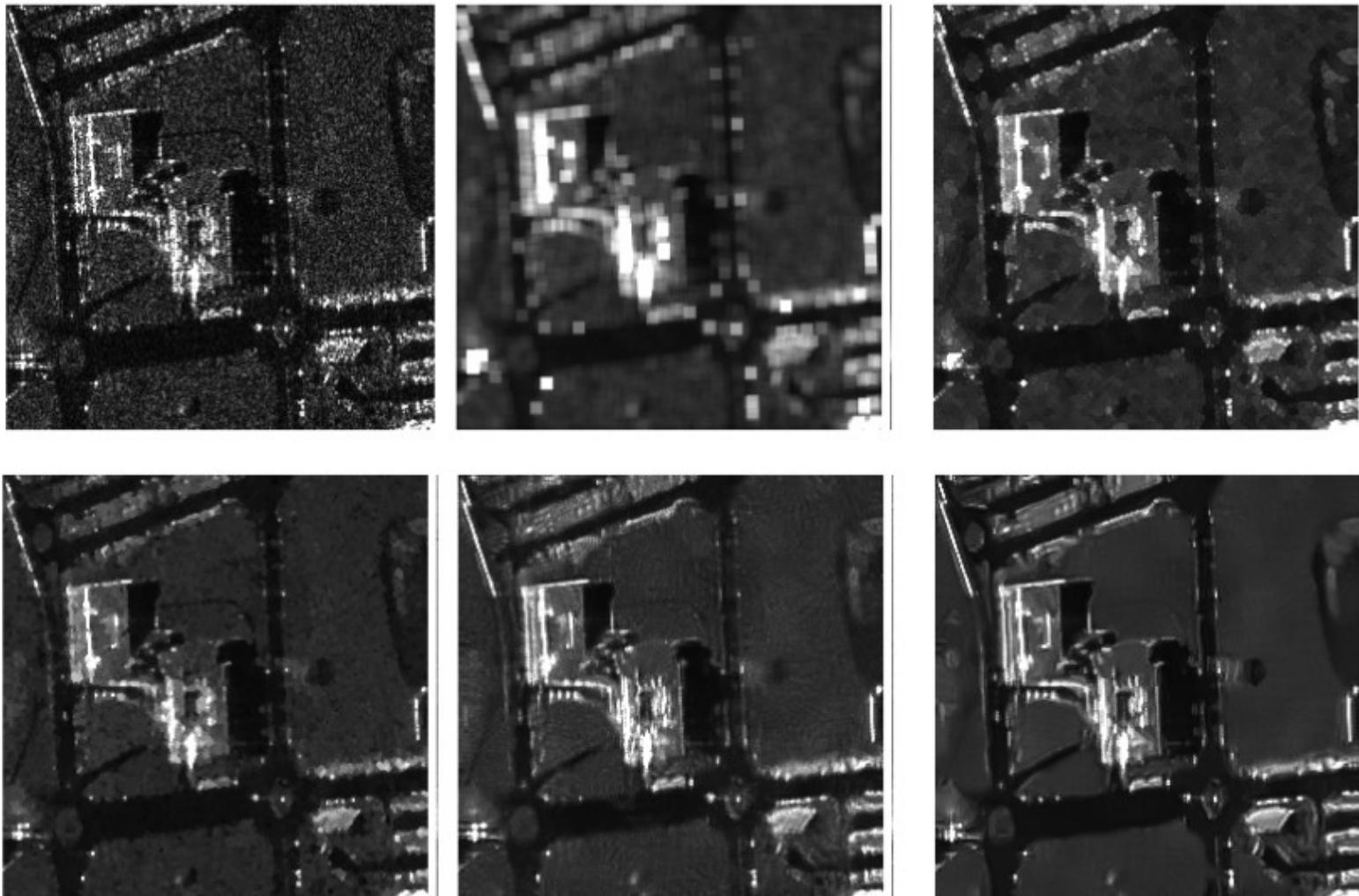


After spatial filtering (N*N Lee filter)



30

Speckle filtering



Non local speckle filtering

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Speckle filtering with tomographic data

Speckle filtering for monovariate SLC SAR images

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

Speckle filtering for multivariate SLC MB-InSAR images

$$\mathbf{y} = \begin{bmatrix} s_1 \\ \vdots \\ s_M \end{bmatrix}, \mathbf{a}(z) = \begin{bmatrix} 1 \\ \vdots \\ e^{jk_{z_M} z} \end{bmatrix} \quad f(z) = \frac{\mathbf{a}^H(z)\mathbf{y}}{M} = \frac{1}{M} \sum_m s_m e^{-jk_{z_m} z}$$
$$\hat{I}(z) = \frac{1}{L} \sum_{l=1}^L |f(z, l)|^2 = \frac{1}{M^2} \mathbf{a}^H(z) \widehat{\mathbf{R}} \mathbf{a}(z)$$

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

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

Speckle filtering with tomographic data

TomoSAR covariance matrix

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

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

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

Interferometric coherence estimate

$$\hat{\gamma}_{ij} = \frac{\hat{R}_{ij}}{\sqrt{\hat{I}_i \hat{I}_j}} \quad \hat{\phi}_{ij} = \arg(\hat{\gamma}_{ij}) \quad |\hat{\gamma}_{ij}| \leq 1$$

Tomographic imaging using specan

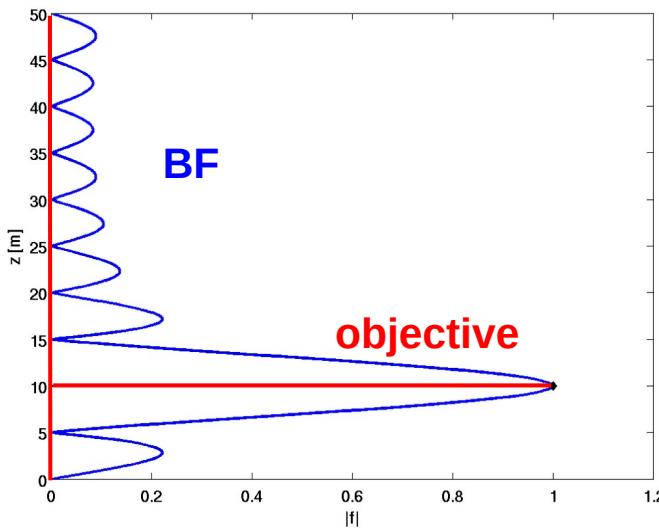
Beamformer is Fourier imaging

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

- 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

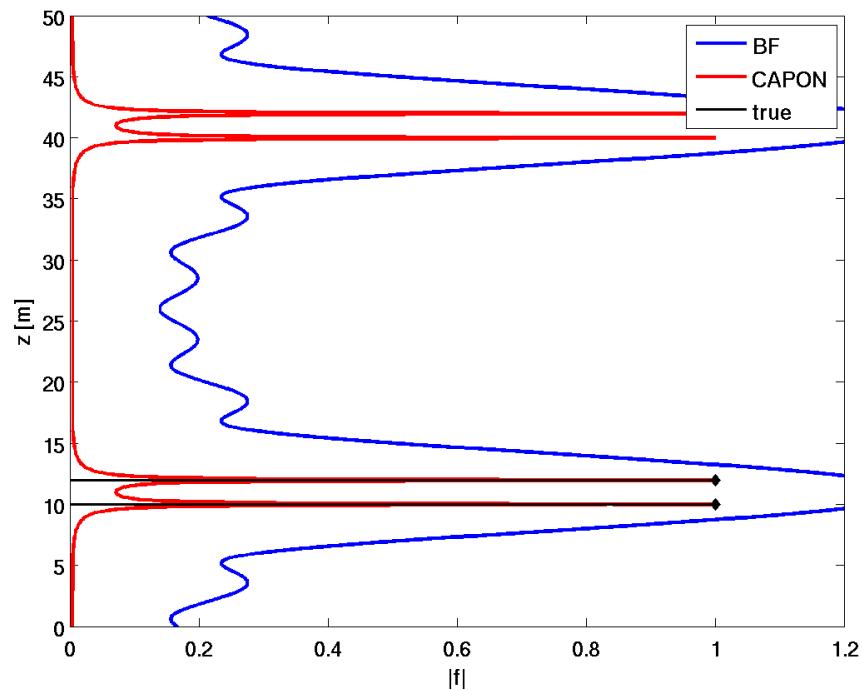
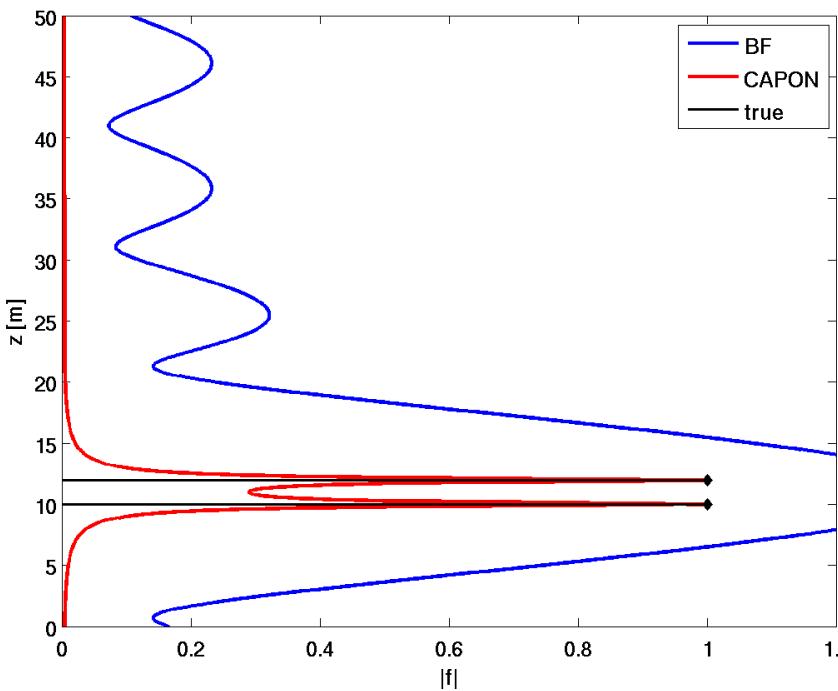


$$\mathbf{v}_{CP}(z) = \arg \min_{\mathbf{v}} \mathbb{E}(|\mathbf{v}^H \mathbf{y}|^2) \quad \text{s.t.} \quad \mathbf{v}^H \mathbf{a}(z) = 1$$

Solution:

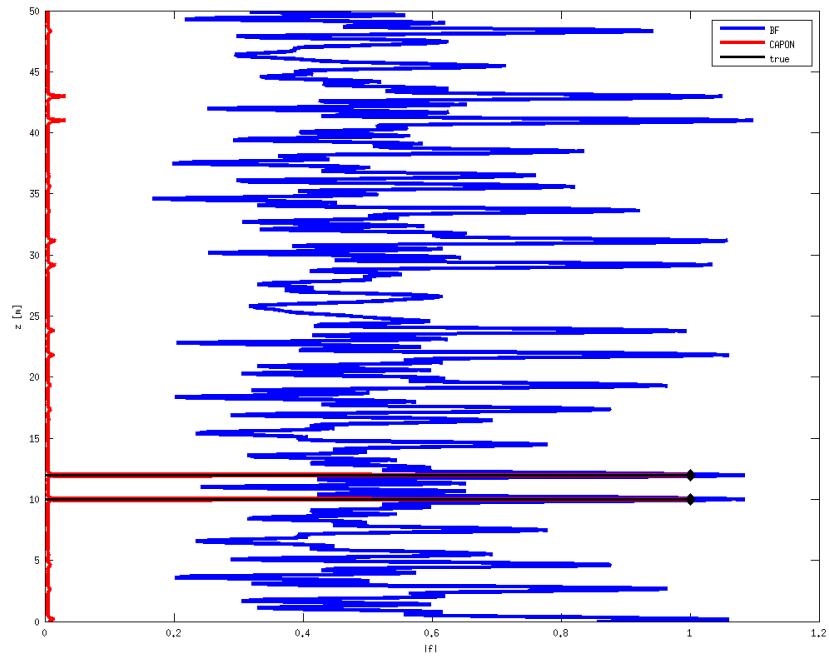
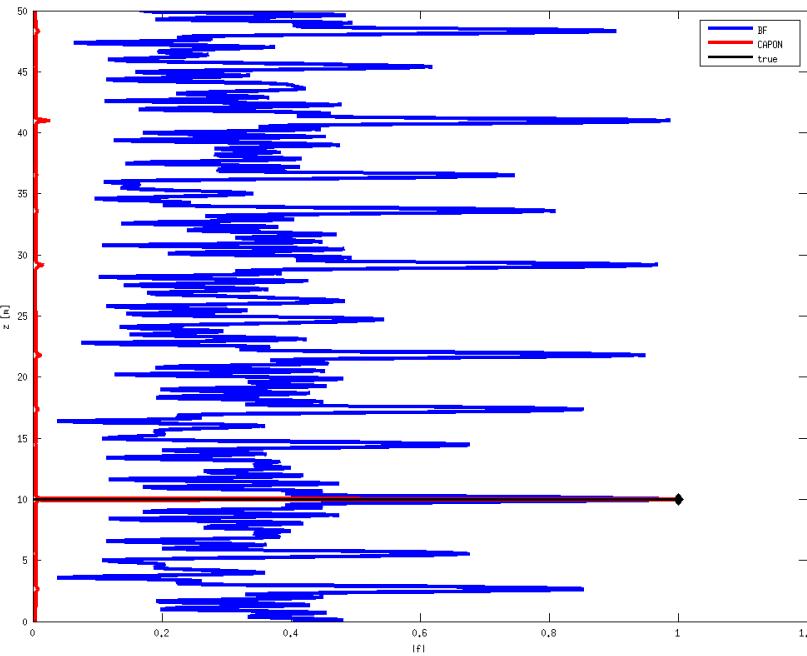
$$\hat{I}_{CP} = \frac{1}{\mathbf{a}^H(z) \hat{\mathbf{R}}^{-1} \mathbf{a}(z)}$$

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

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

- BF: quite stable w.r.t L

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

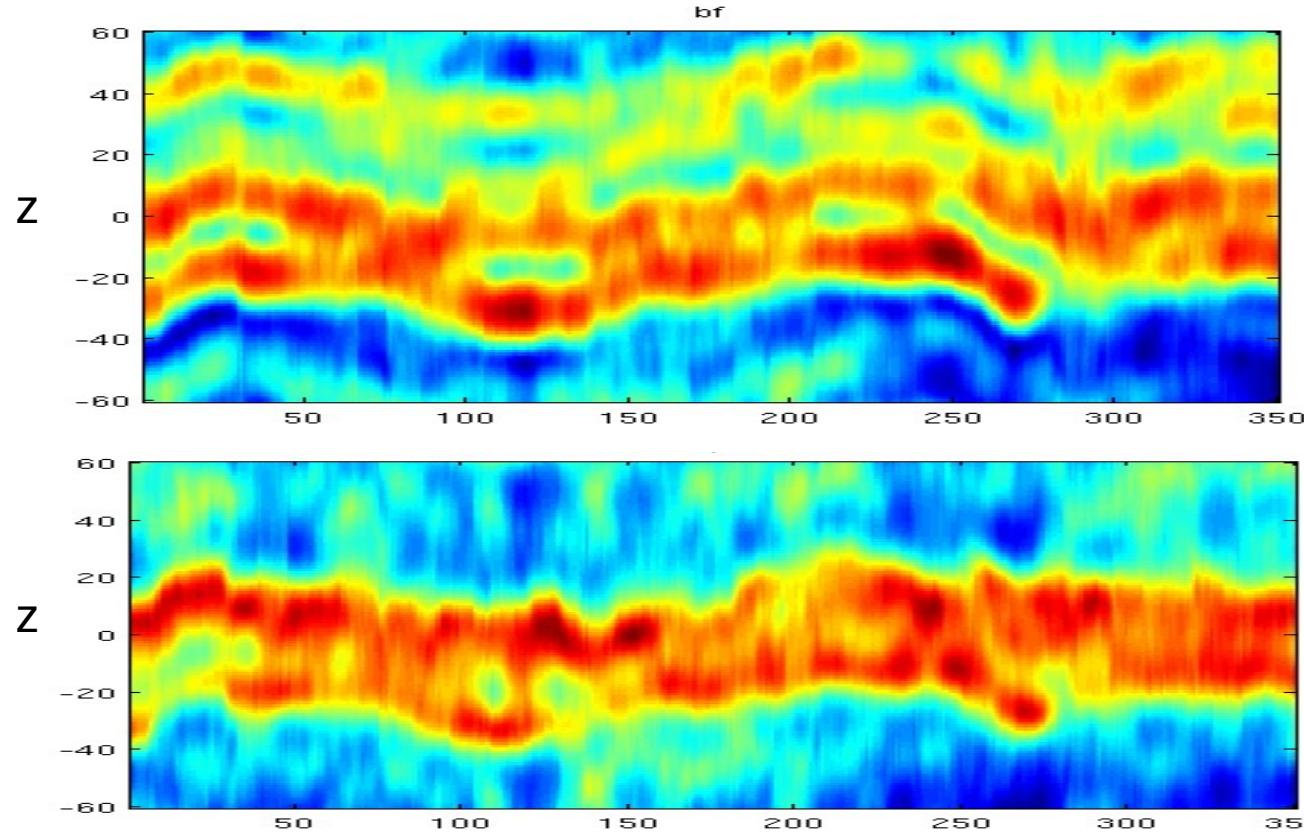
⇒ Diagonal Loading

$$\widetilde{\mathbf{R}} = \widehat{\mathbf{R}} + \alpha \mathbf{I}_M, \quad \alpha \geq 0$$

For large α (low SNR): CP → BF

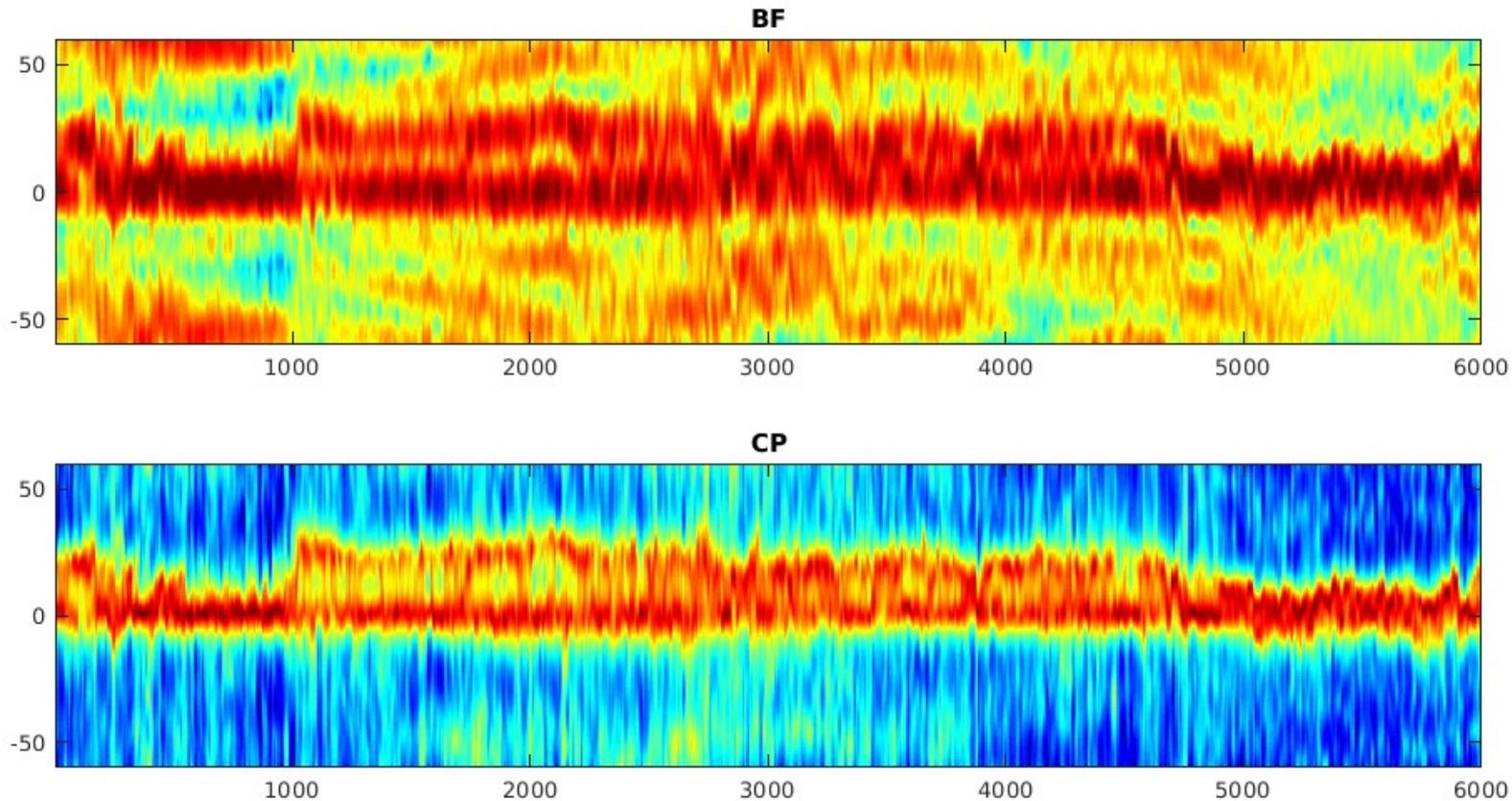
Tomographic imaging using specan

Tropical forest profile at P band with residual phase errors

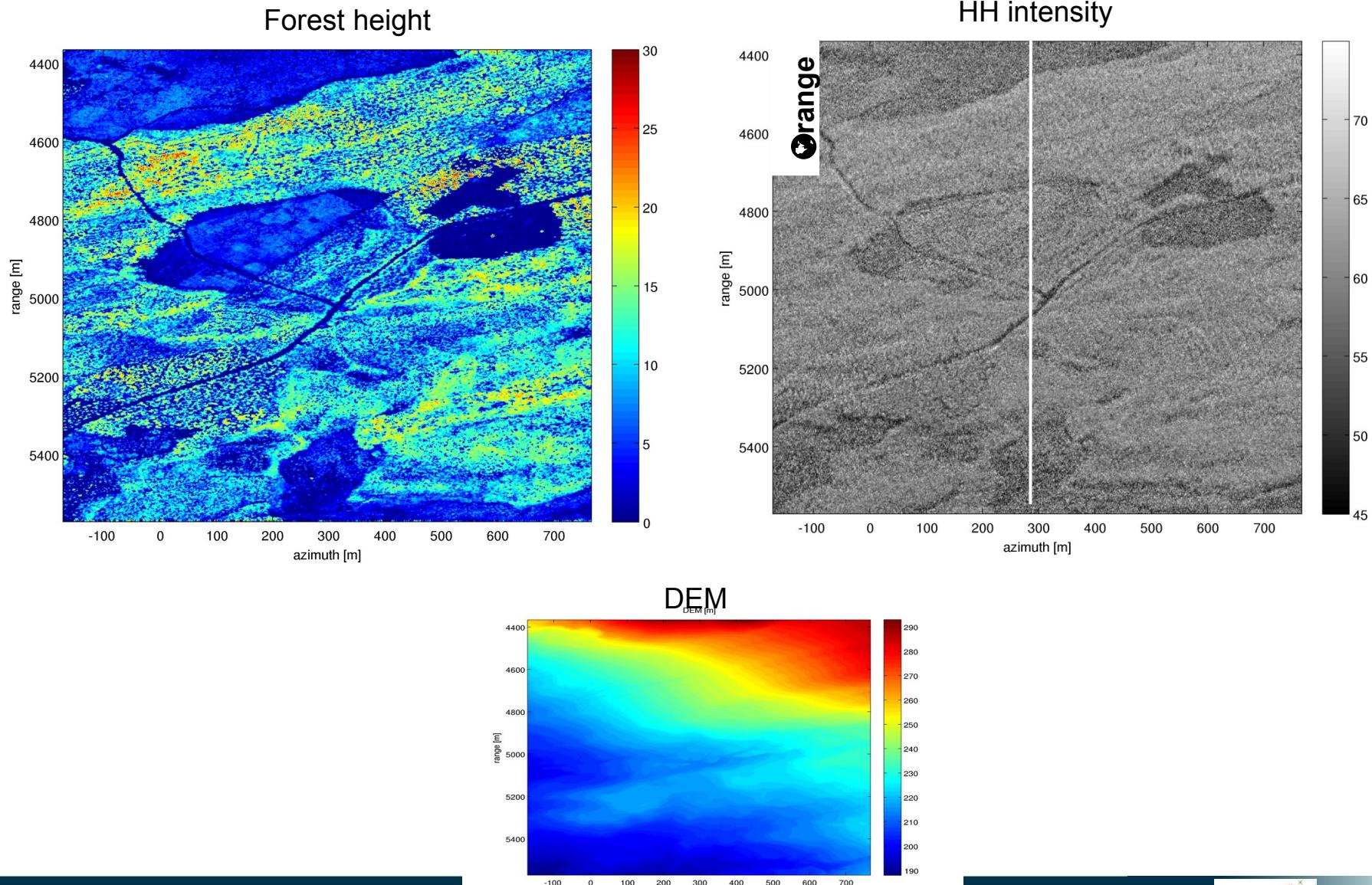


Tomographic imaging using specan

P band tomogram (Tomosense campaign) with residual phase errors

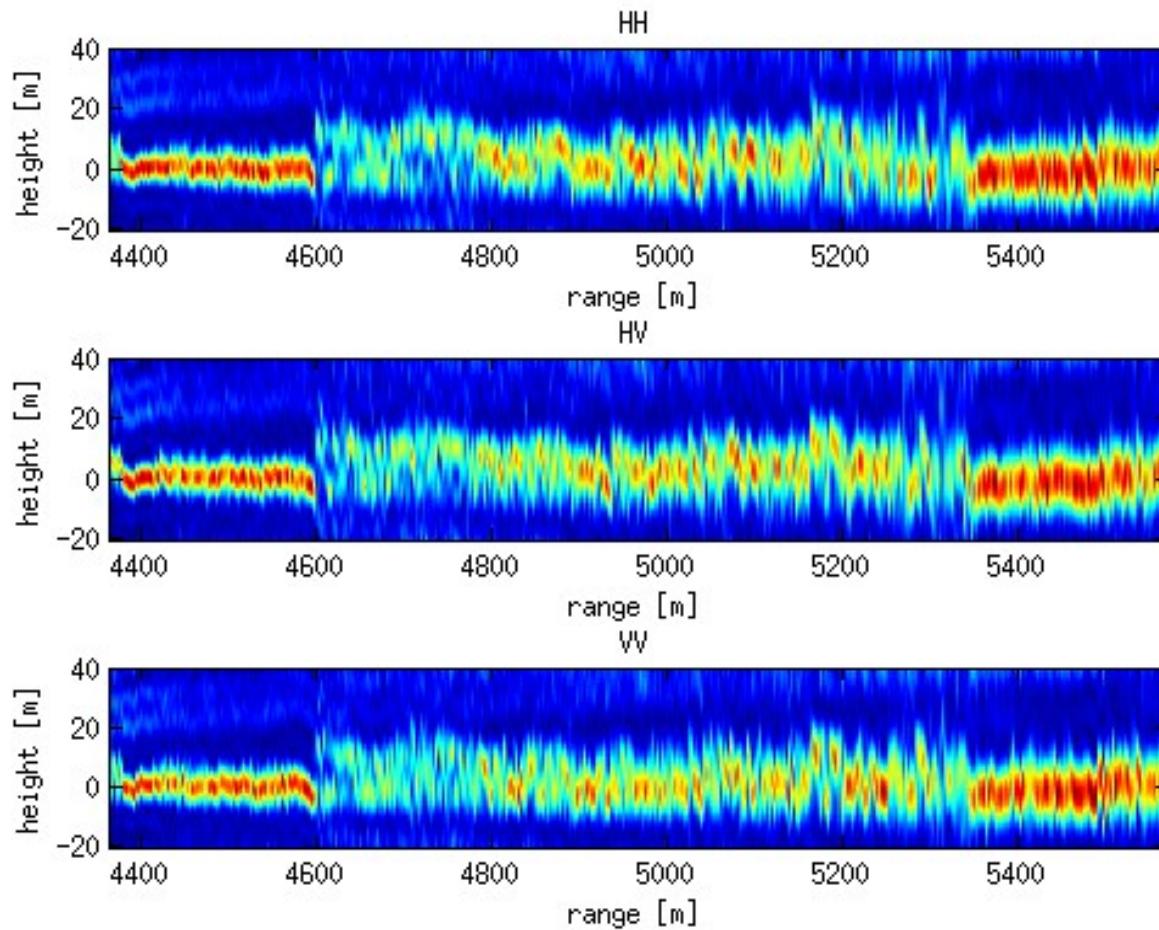


Case study: BIOSAR 2 data



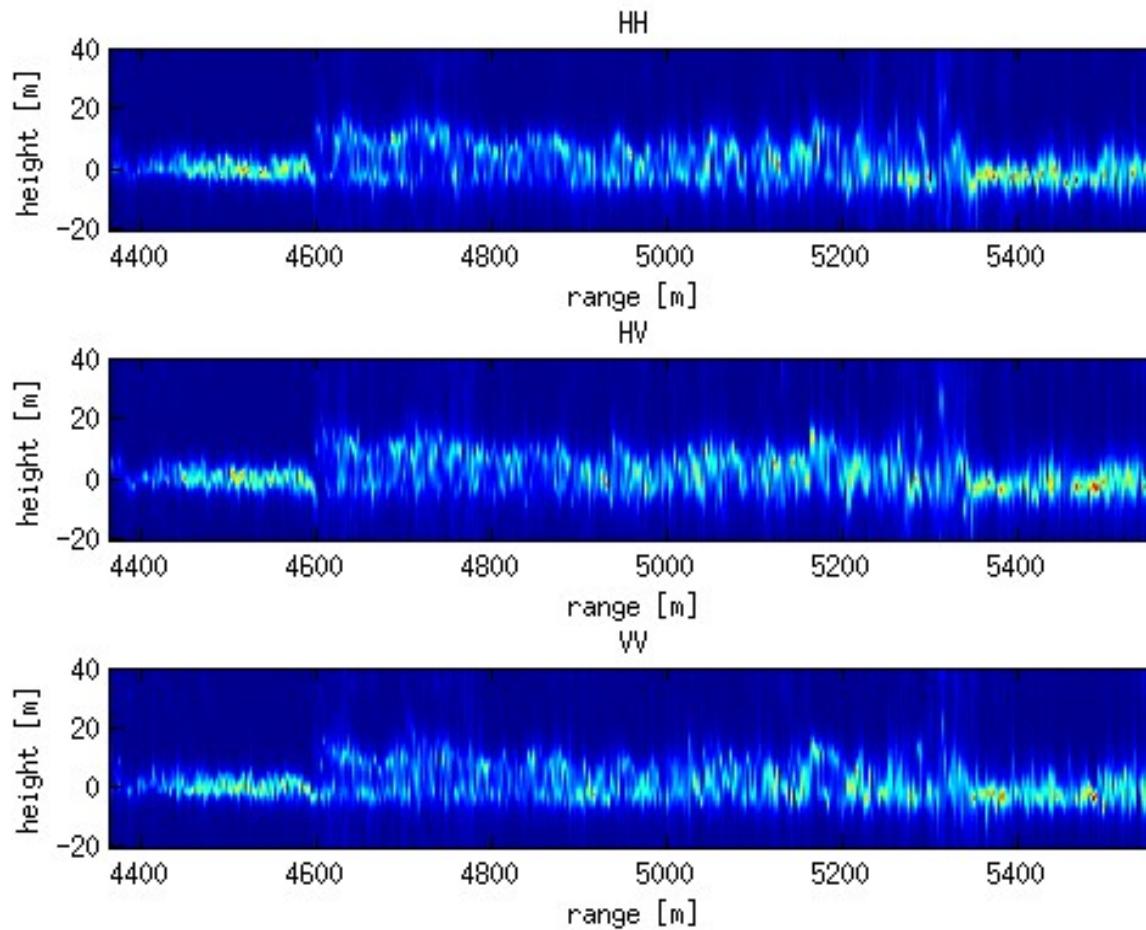
Case study: BIOSAR 2 data

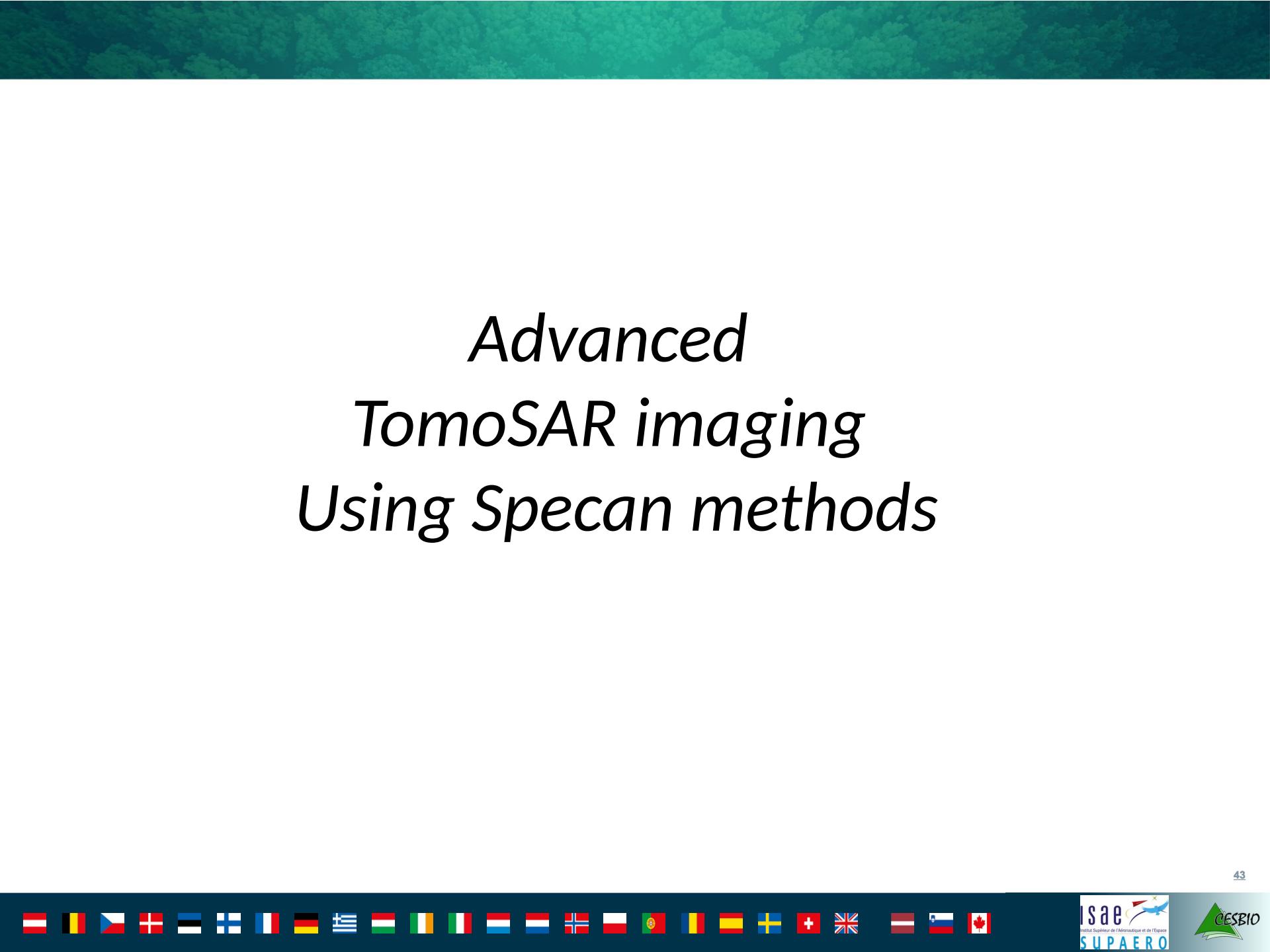
BF



Case study: BIOSAR 2 data

CAPON: processing OK ?





Advanced TomoSAR imaging Using Specan methods

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

Urban area test site

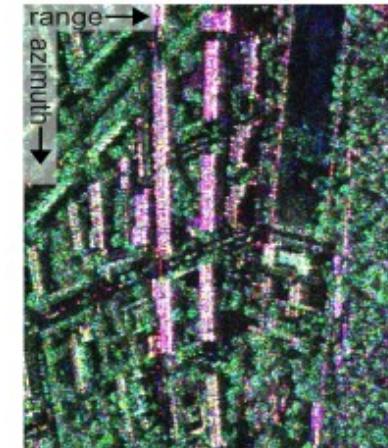
- Images over Dresden, 2000
- DLR's E-SAR at L-Band
- Resolution : $0.5 \text{ m} \times 2.5 \text{ m}$
- Fully polarimetric
- Dual-baseline InSAR

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

3 PolSAR images



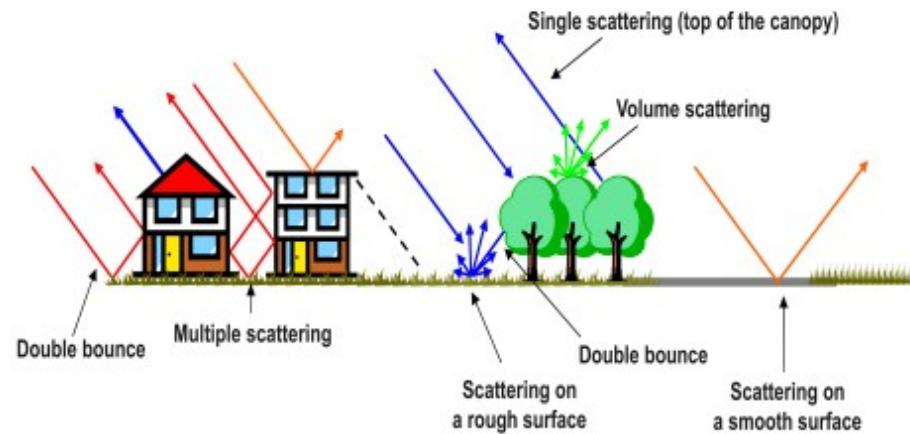
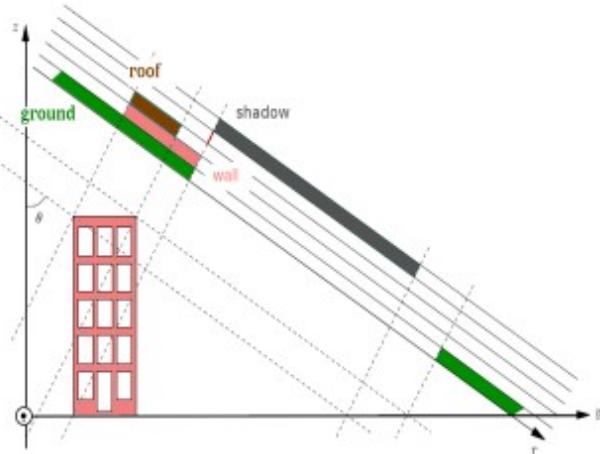
Pauli-coded SAR image



Optical image



SAR tomography over urban areas



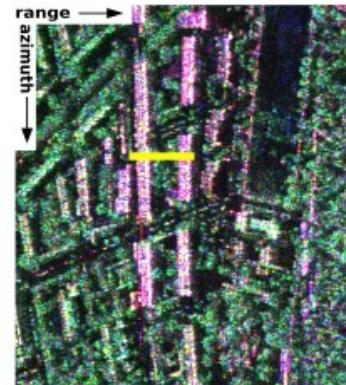
- L-band intermediate-resolution data sets
⇒ High-Resolution (HR) tomographic estimators
- 3 images
⇒ $N_s = 2$

Tomographic imaging using specan

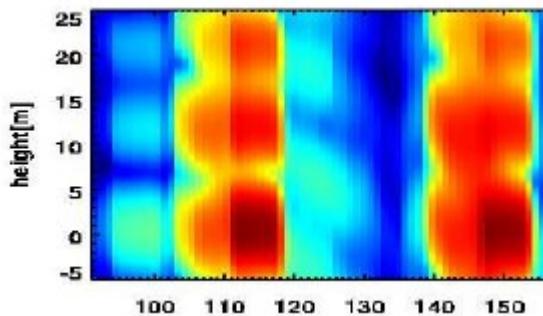
Critical configuration (**3 images**) in an urban environment at L band



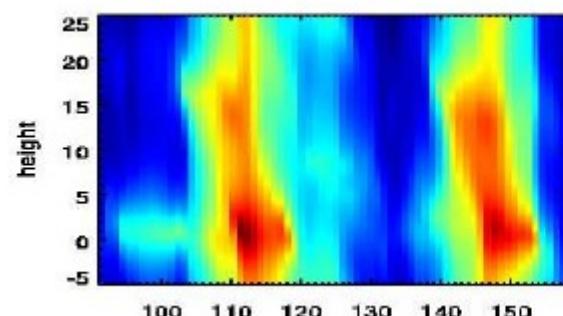
(a) Optical image



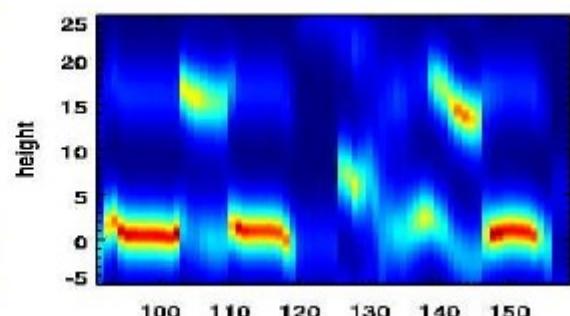
(b) SAR



BF



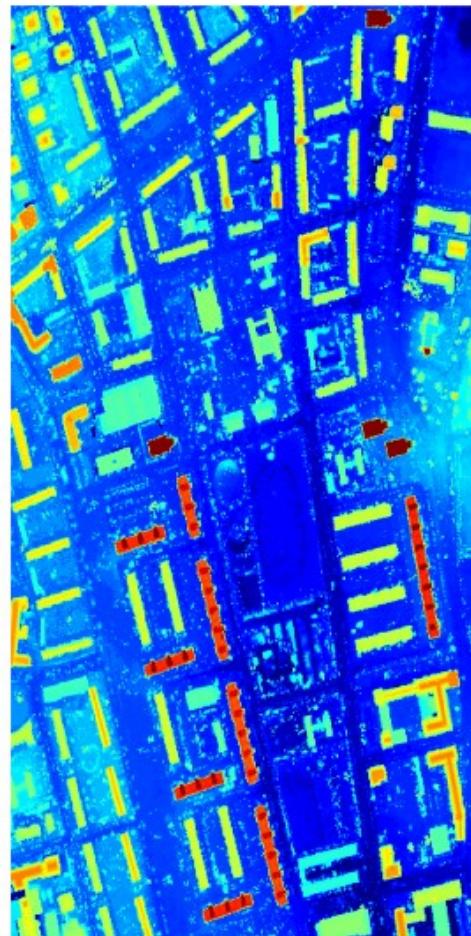
CAPON



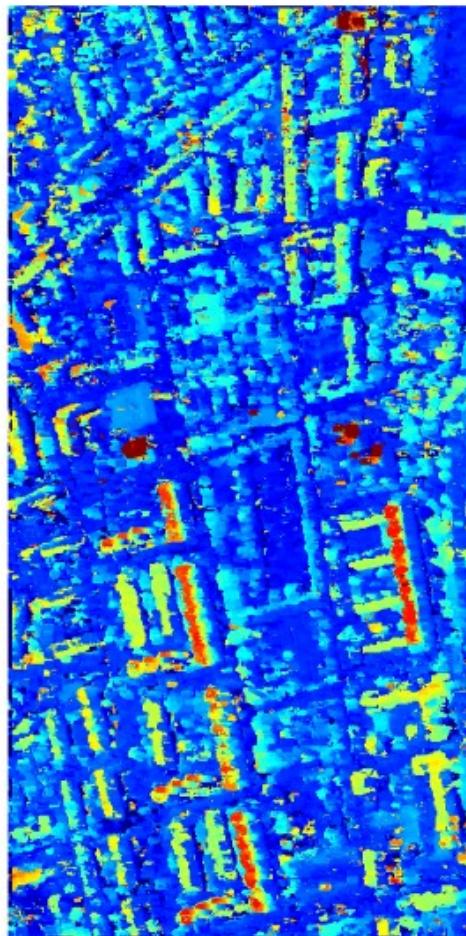
MUSIC

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

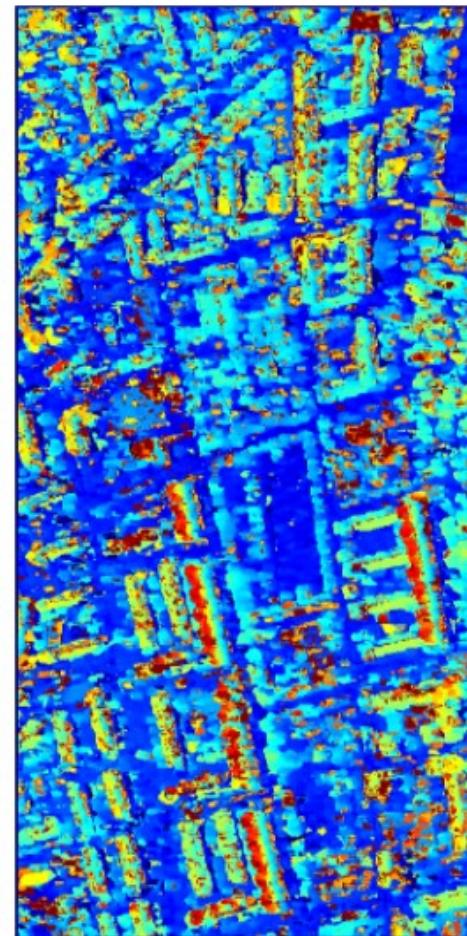
Polarimetric SAR tomography over urban areas



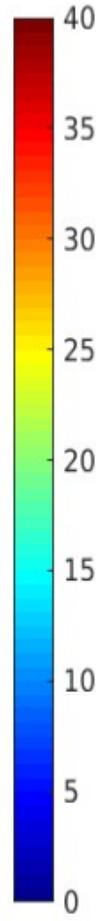
LiDAR



P-SSF

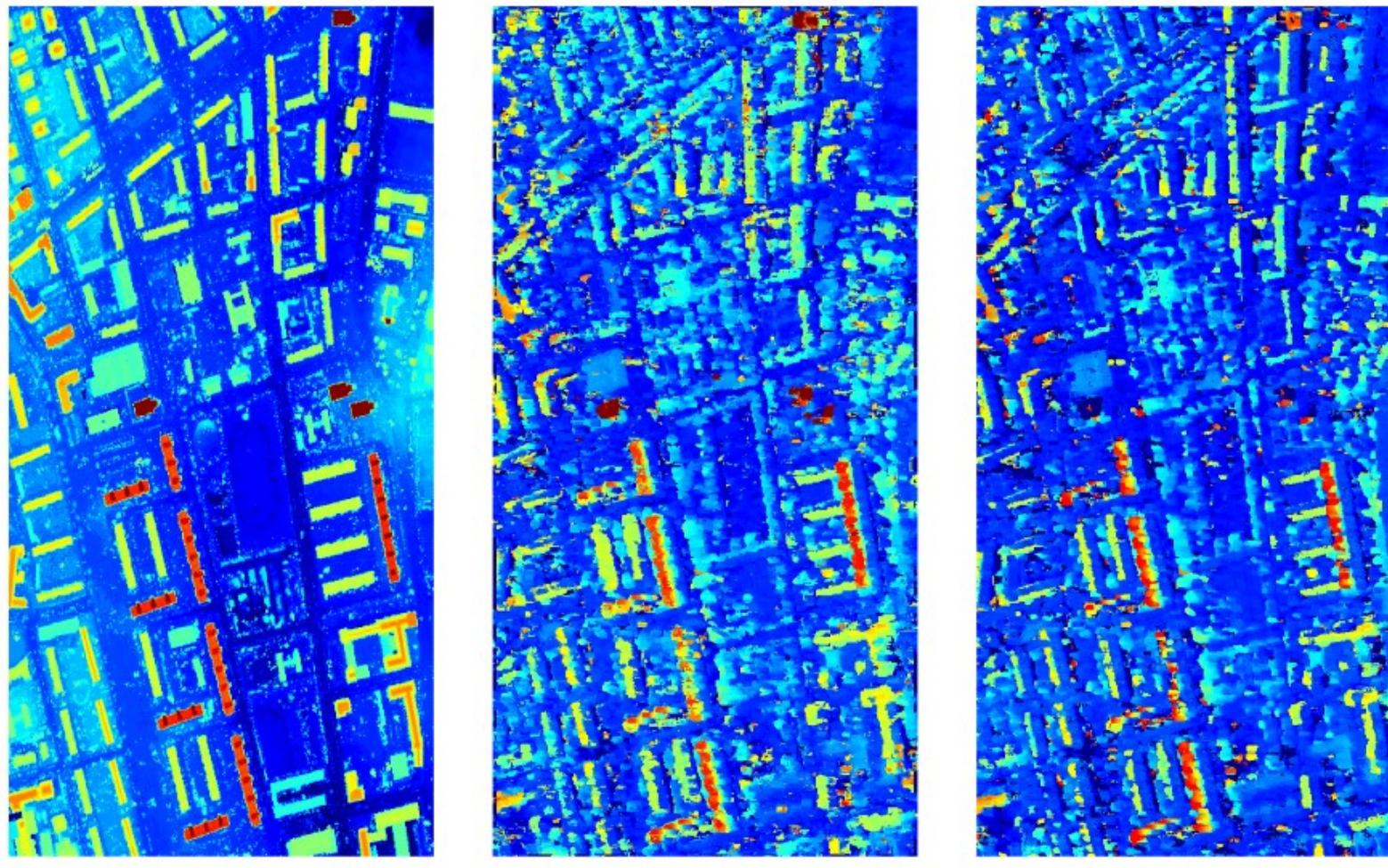


VV SSF



z

Polarimetric SAR tomography over urban areas



LiDAR

P-SSF

P-NSF

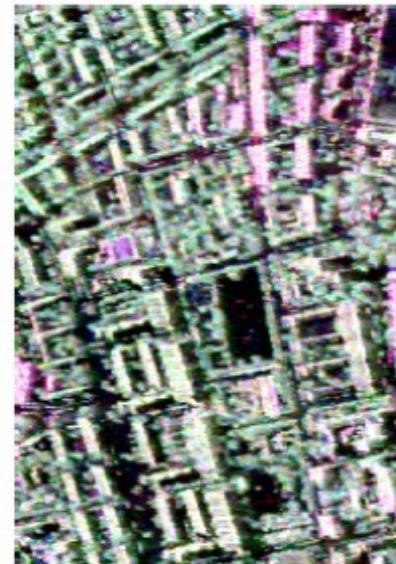
z

Polarimetric SAR tomography over urban areas

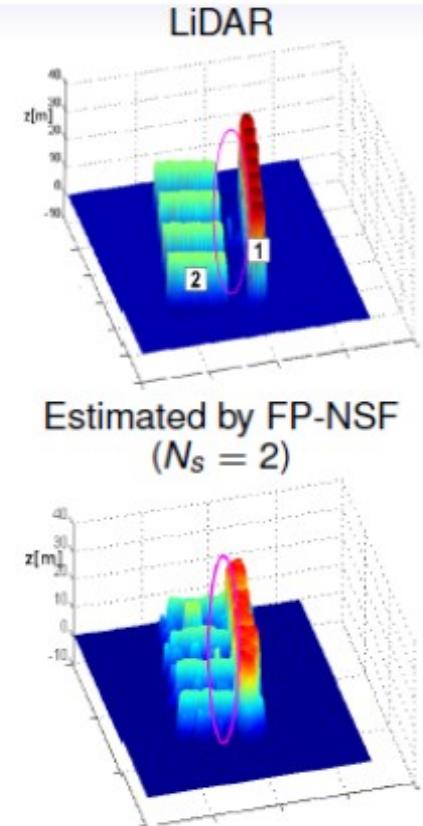
Building reconstruction



Google map



Pauli-coded



Difference between LiDAR and estimated surface

- projection of SAR imaging
- vegetation between B1 and B2

Polarimetric SAR tomography over urban areas

Building reconstruction

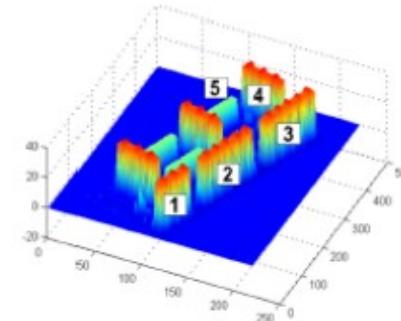


Google map

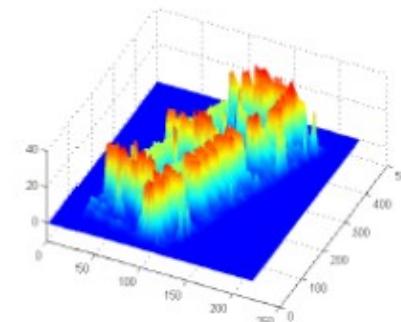


Bing map

LiDAR



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

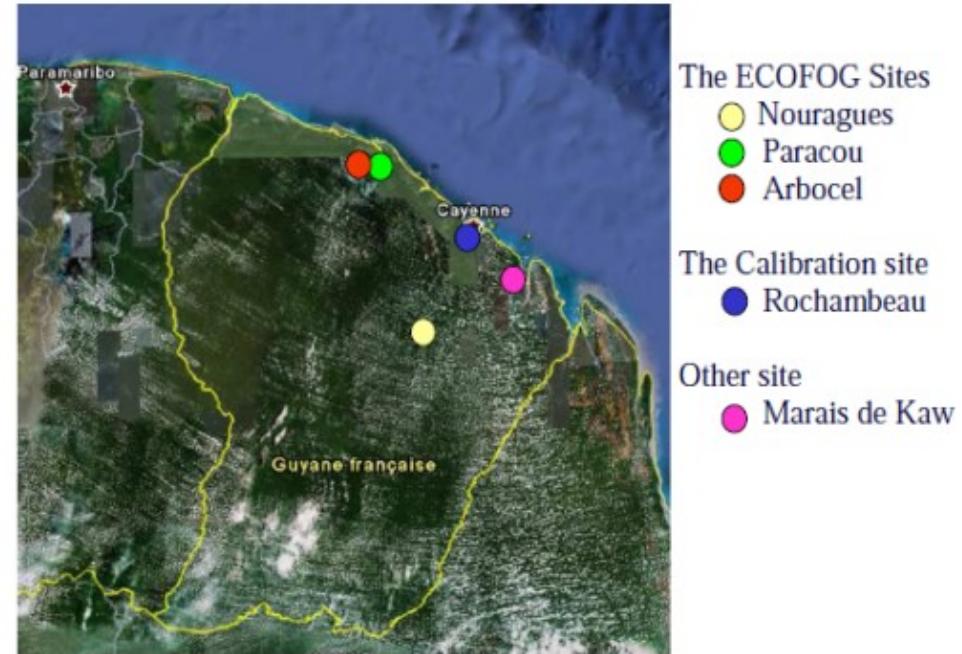


Averaged z[m]	B1	B2	B3	B4	B5
LIDAR	30.0	30.2	30.1	30.8	16.3
Estimated	27.5	27.8	27.5	27.3	16.1

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

Objectives

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

Tropical forest characterization

Tropical forest test site : Paracou

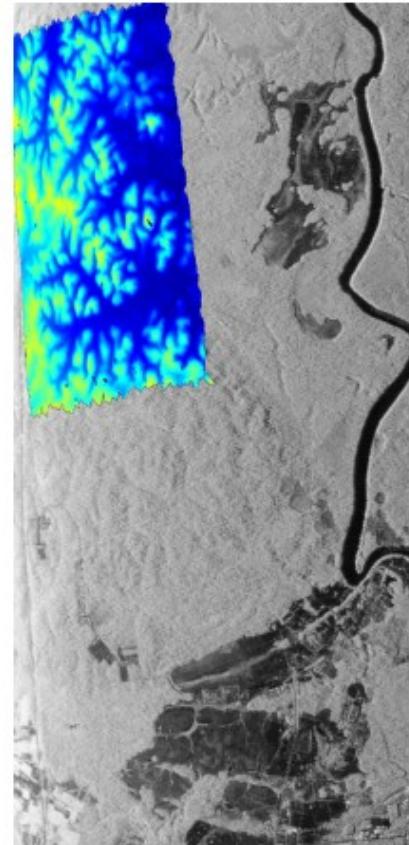
Optical image



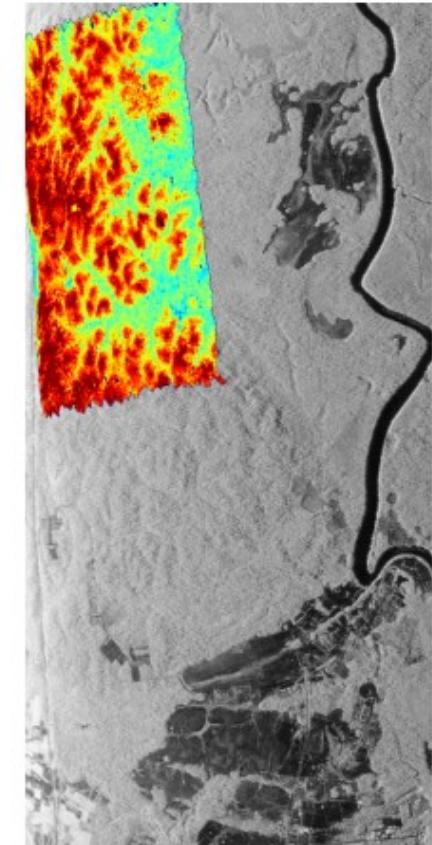
SAR image



LiDAR z_g



LiDAR z_{top}



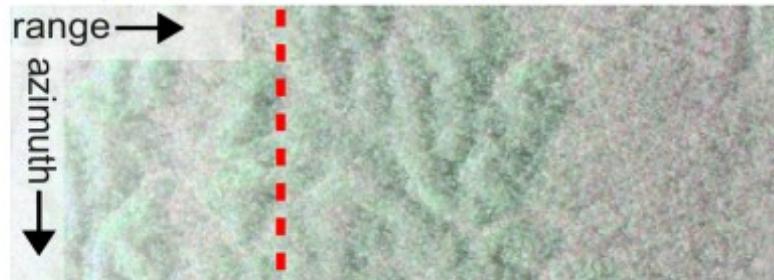
- Tropical forest environments
(savannah, undisturbed forests, logged plots...)

- Highly varying ground topography

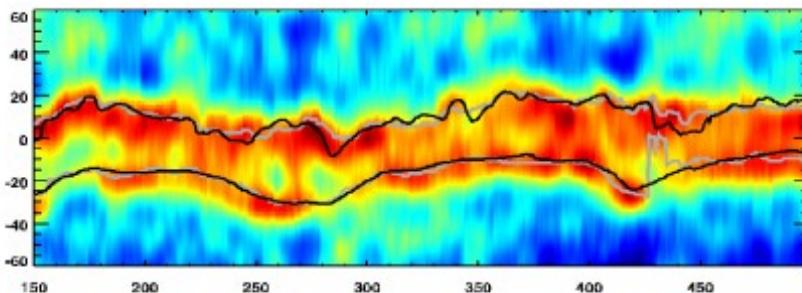
Tropical forest characterization

Tree height and ground topography estimation

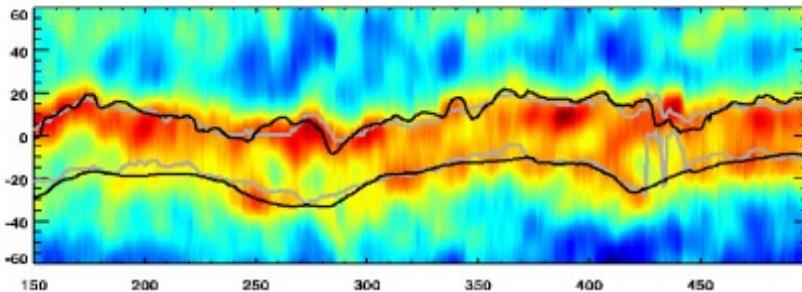
Hybrid spectral approach



HH



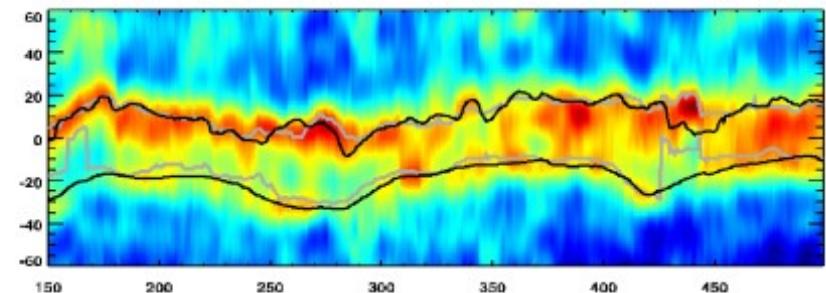
HV



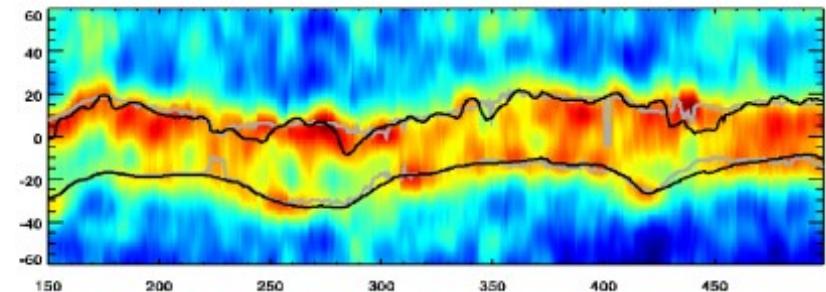
- Estimated profiles match LiDAR
- HH profiles : similar to FP case

LiDAR — TomSAR —

VV



FP

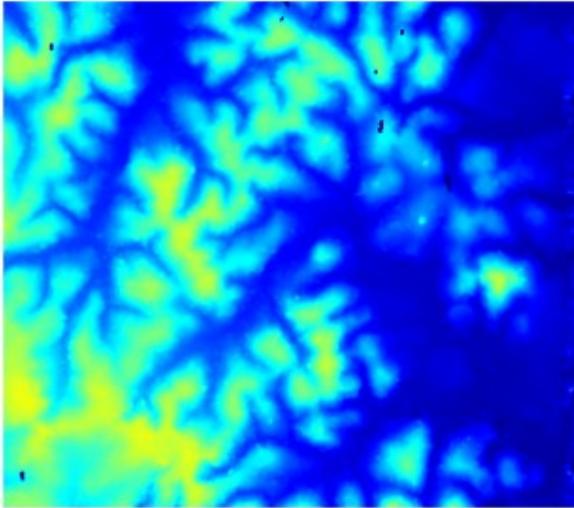


Tropical forest characterization

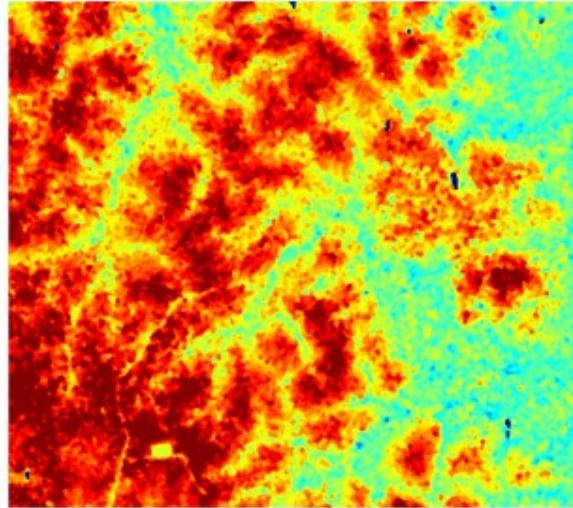
LiDAR data (ground range)

z_g, z_{top} 0 [m] 65[m]

z_g

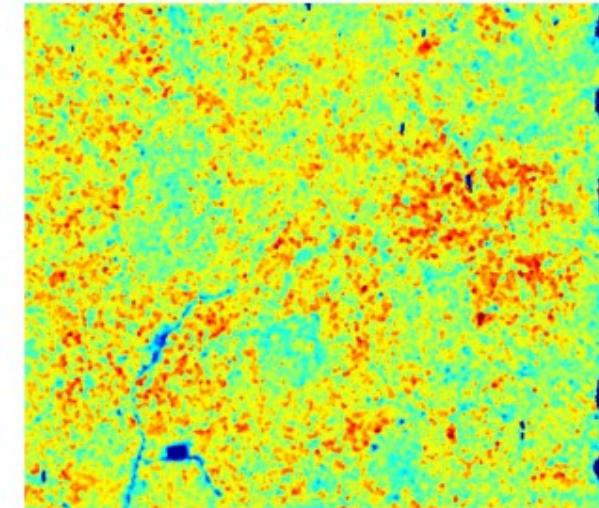


z_{top}



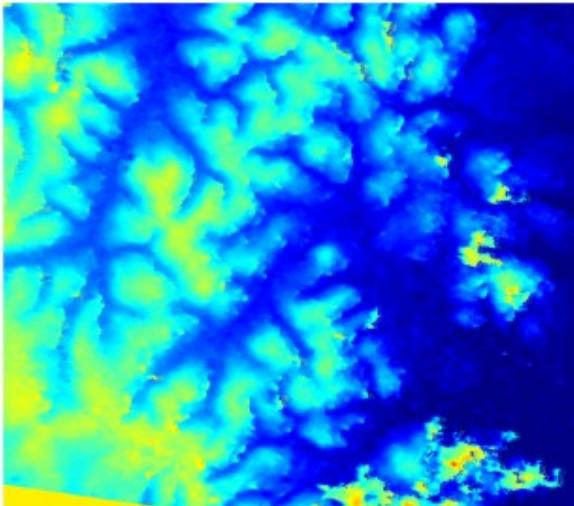
H_v 0 [m] 50[m]

H_v

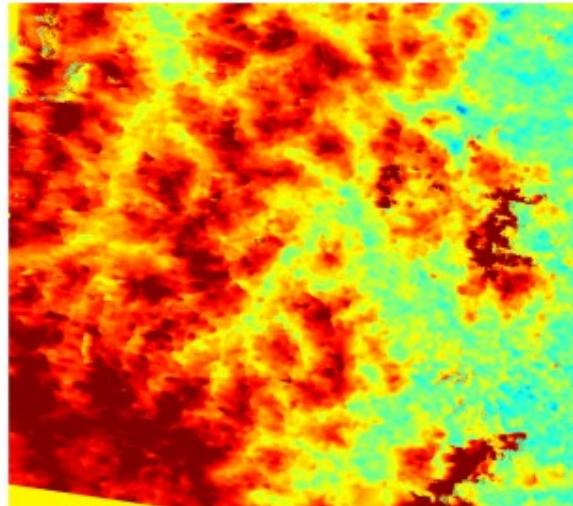


Estimated results (ground range)

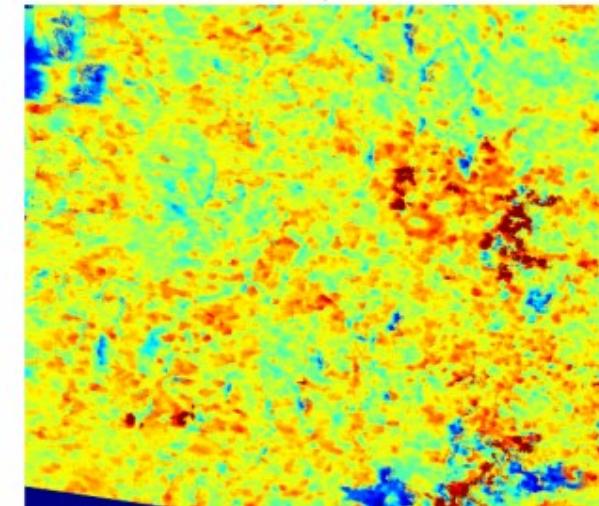
\hat{z}_g



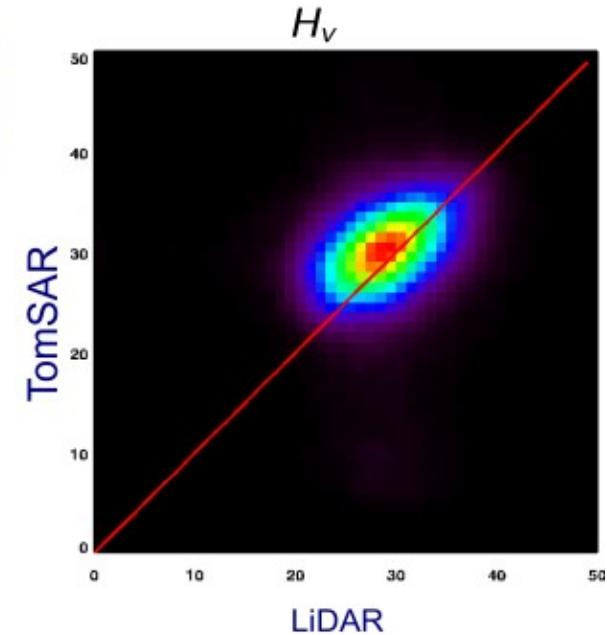
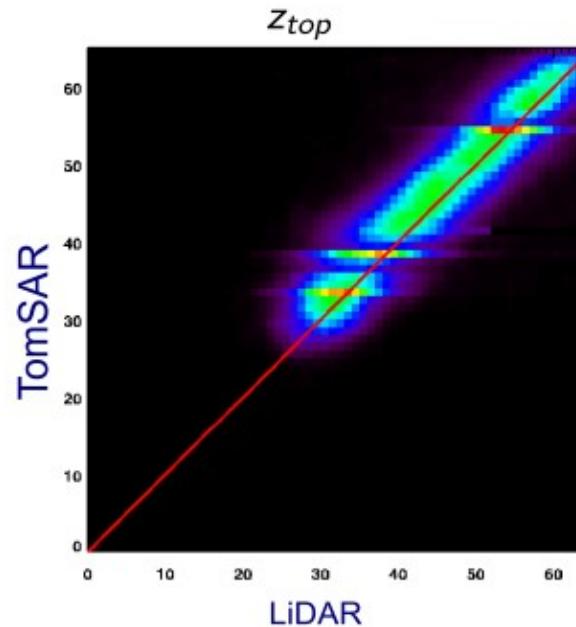
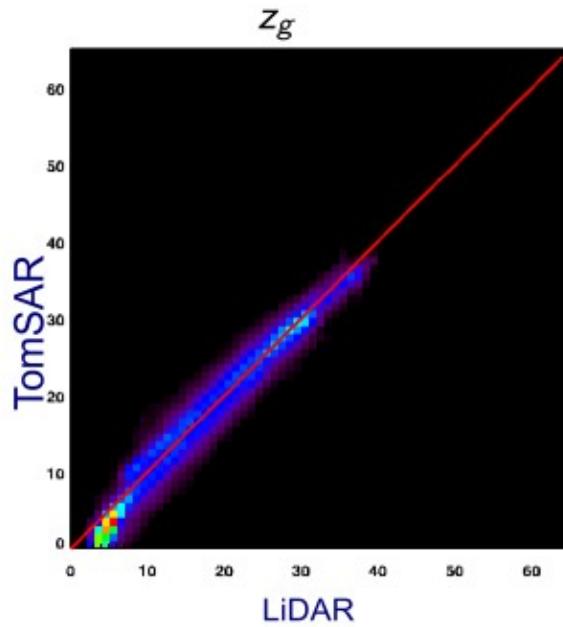
\hat{z}_{top}



\hat{H}_v



Tropical forest characterization

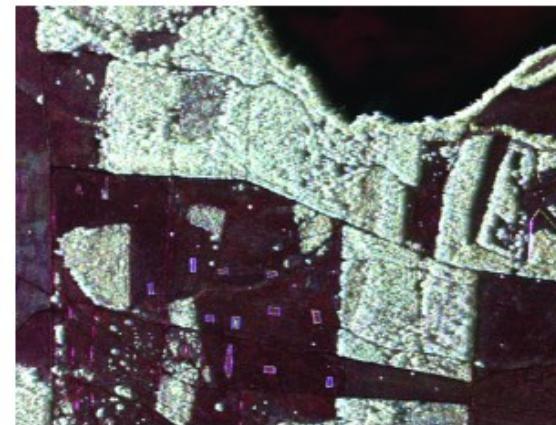
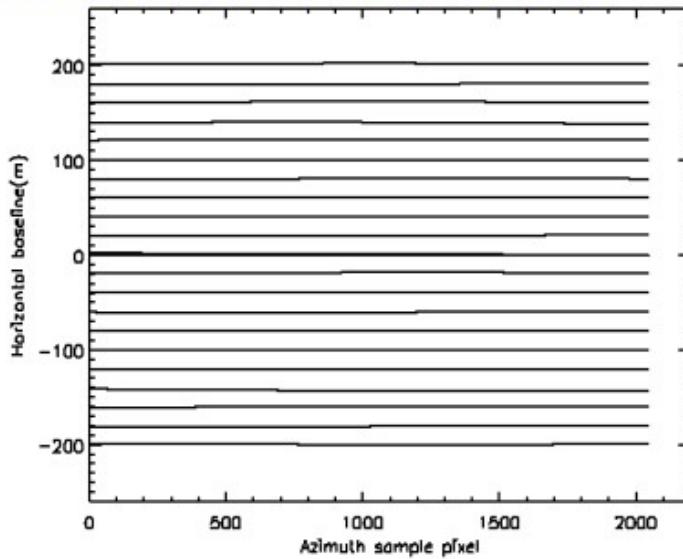


TomoSAR imaging of concealed objects

Above ground and under foliage objects observed at L band

- DLR E-SAR image over Dornstetten, Germany
- L-Band
- 21 tracks : average baseline 20m
- $\delta_z = 2m$

Horizontal baseline distribution



TomoSAR imaging of concealed objects

VV reflectivity tomograms



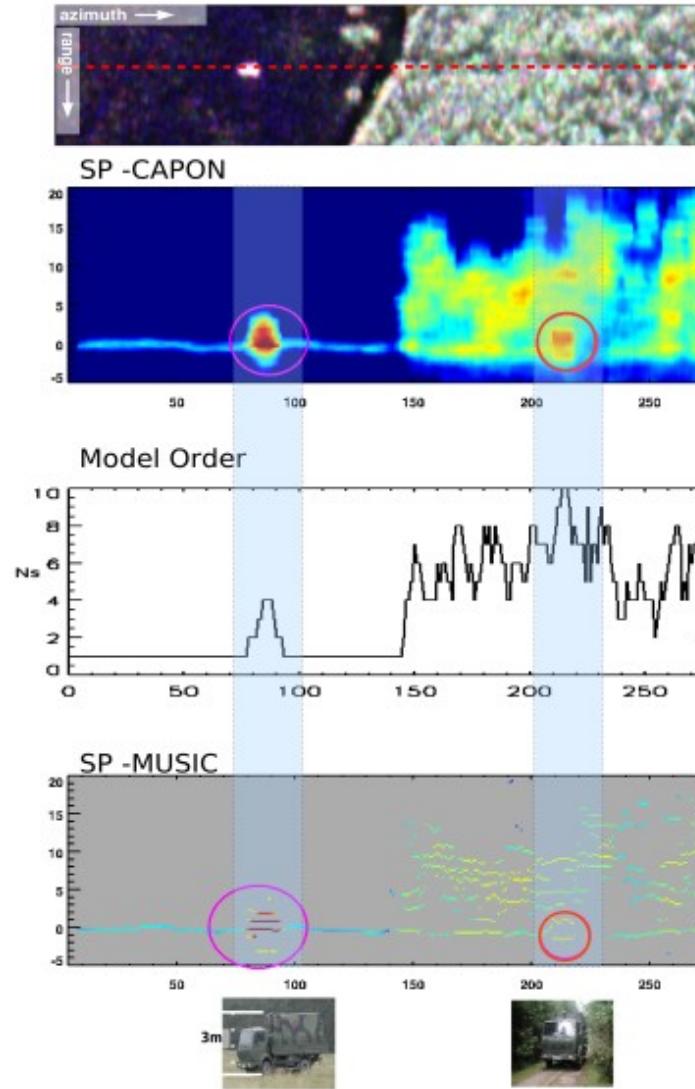
Capon :

limited resolution

⇒ overestimated H_{truck}

MUSIC :

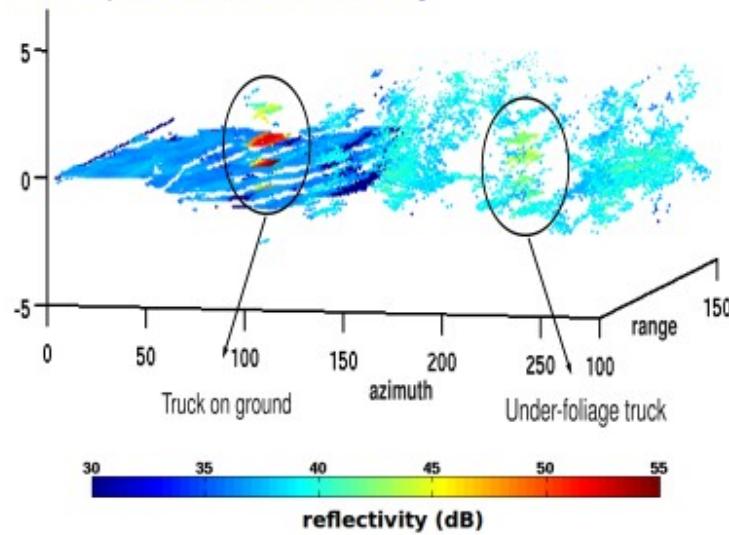
- ☺ Sub-canopy truck
⇒ hybrid scatterer
- ☹ Uncovered
⇒ coherent scatterer
- ☹ Spurious sidelobes.



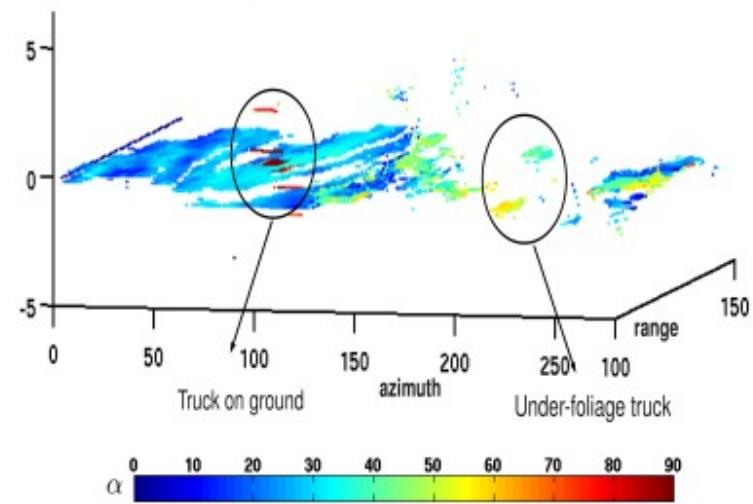
TomoSAR imaging of concealed objects

High Resolution tomograms of underfoliage objects

SSF : shape and reflectivity

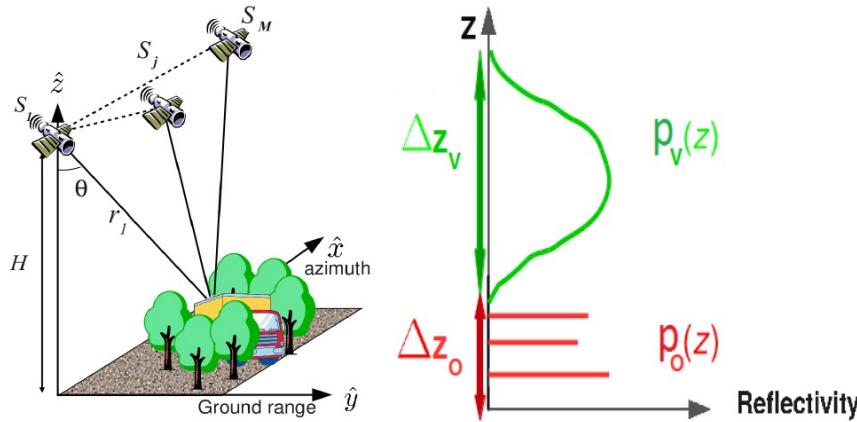


FP-NSF : scattering mechanisms



TomoSAR imaging of concealed objects

Sparse (compressive) sensing solution

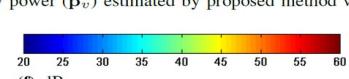
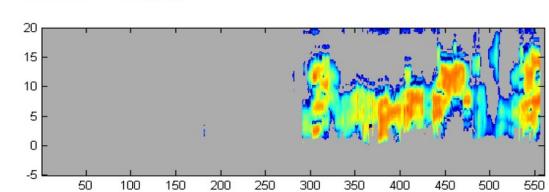
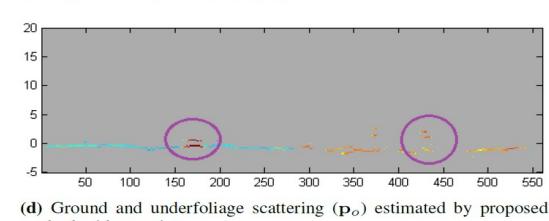
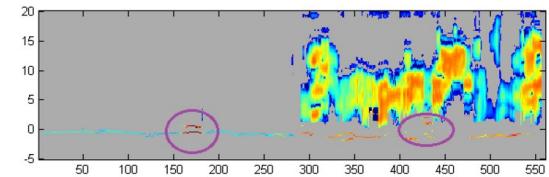
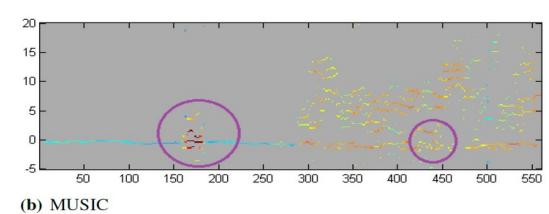
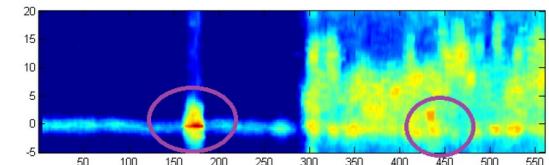
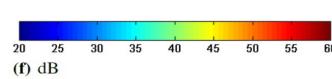
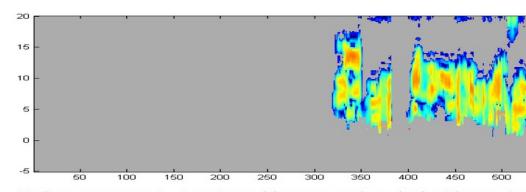
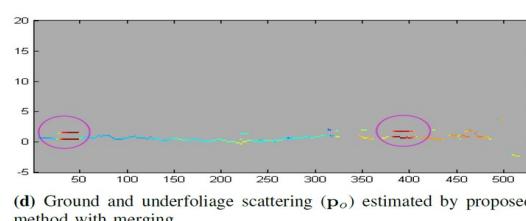
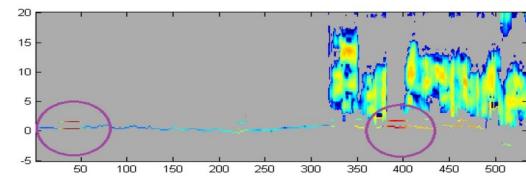
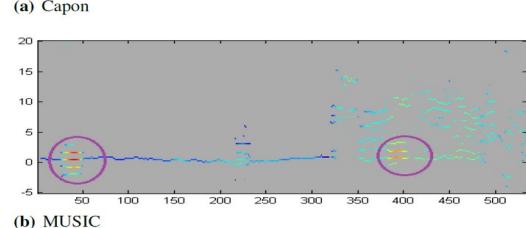
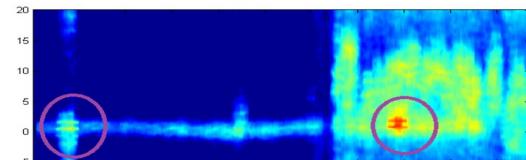
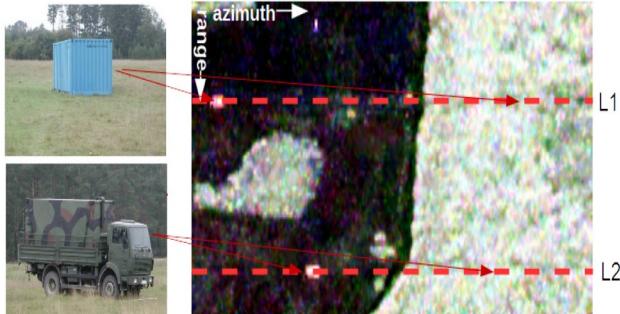


- a few wavelet components
- a few discrete contributions

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

$$\mathbf{B} = \begin{bmatrix} \mathbf{I}_{(N_o \times N_o)} & \mathbf{0} \\ \mathbf{0} & \Psi_{(N_v \times N_v)} \end{bmatrix} \in \mathbb{R}^{(N_s \times N_s)} \quad \mathbf{p} = [\mathbf{p}_o^T \quad \mathbf{p}_v^T]^T \in \mathbb{R}^{+N_s \times 1}$$

TomoSAR imaging of concealed objects

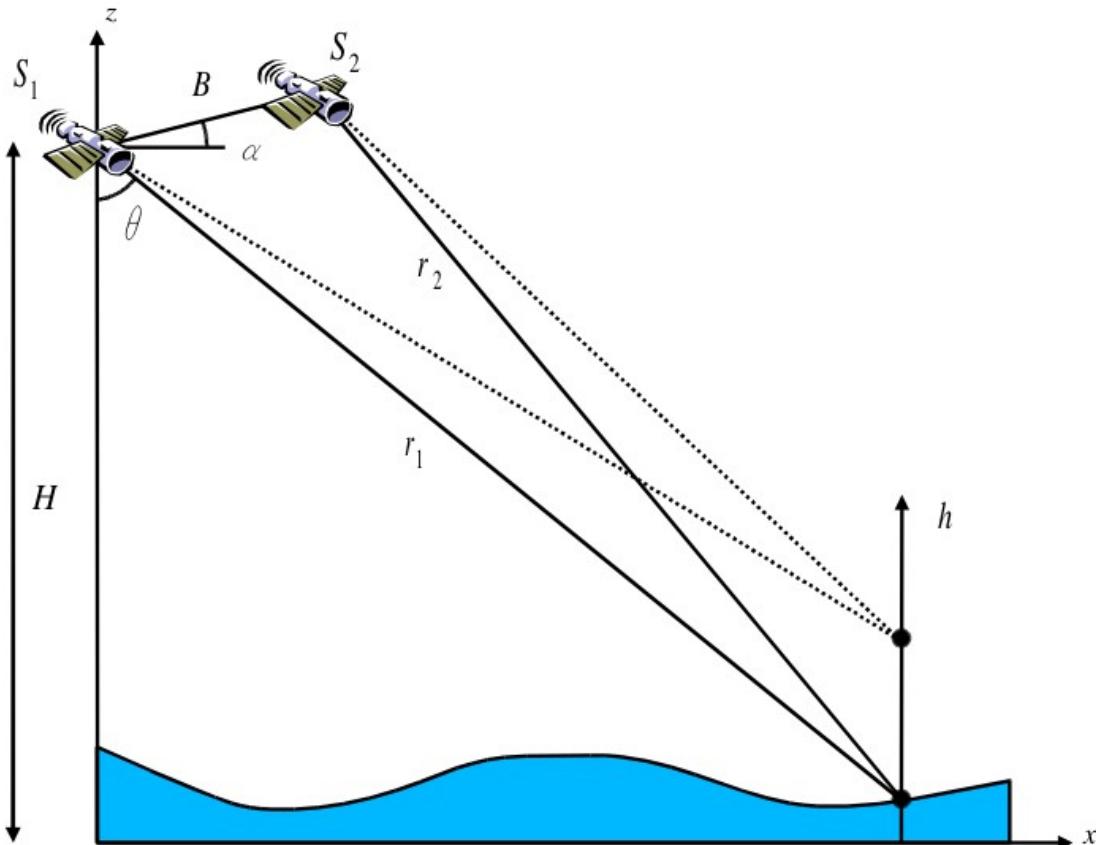




InSAR coherence analysis and TomoSAR modeling



InSAR coherence analysis



S_1 and S_2 are random variables

⇒ assumptions may not be fully verified

⇒ the interferometric phase difference is a random variable

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

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

Interferometric coherence

Joint interferometric representation

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

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

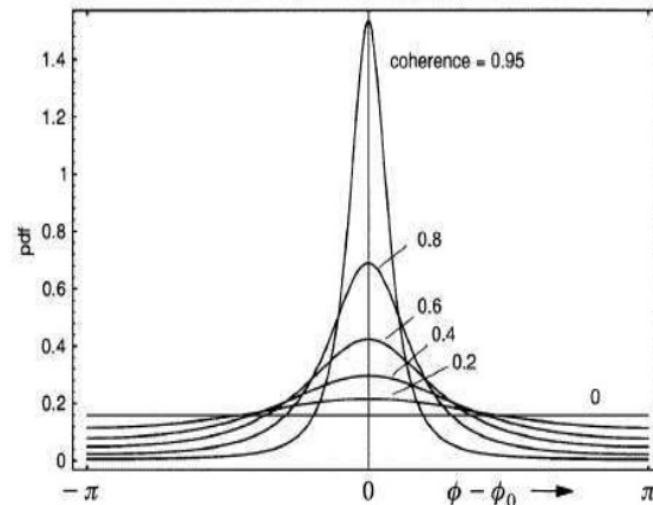
$|\gamma| = 1 \Rightarrow \phi = ?$ interferometric images are totally uncorrelated

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

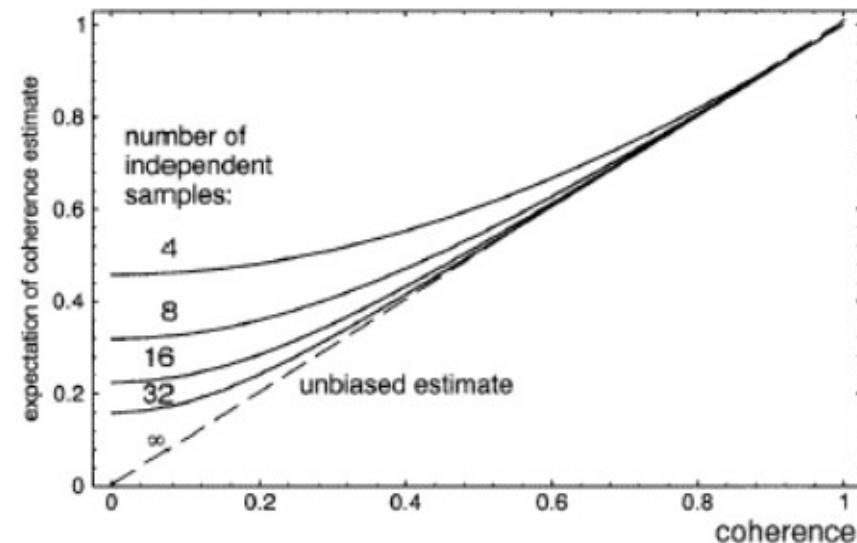
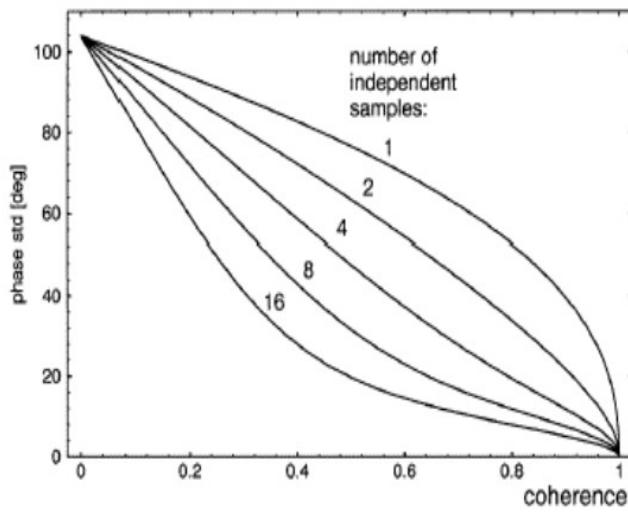


InSAR coherence analysis

Phase
pdf



$|\gamma|$: indicator of phase quality



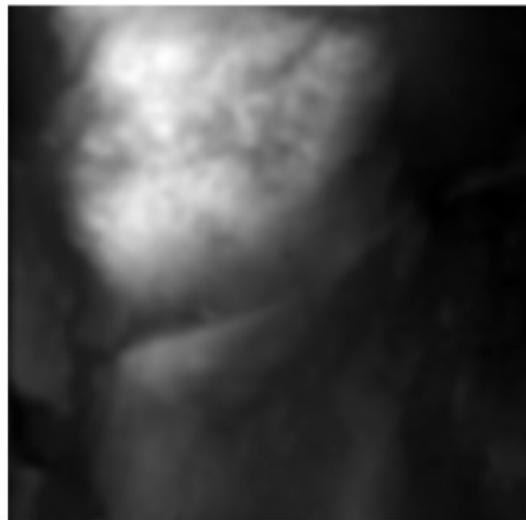
Large number of looks required to reduce:

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

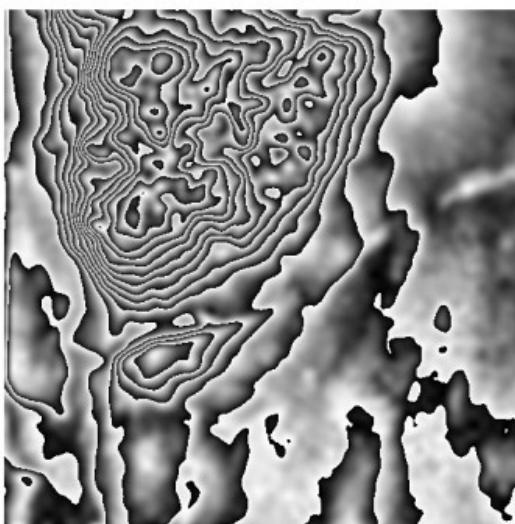
$$\text{bias: } |\gamma| - E(|\hat{\gamma}|) \geq 0$$

InSAR coherence analysis

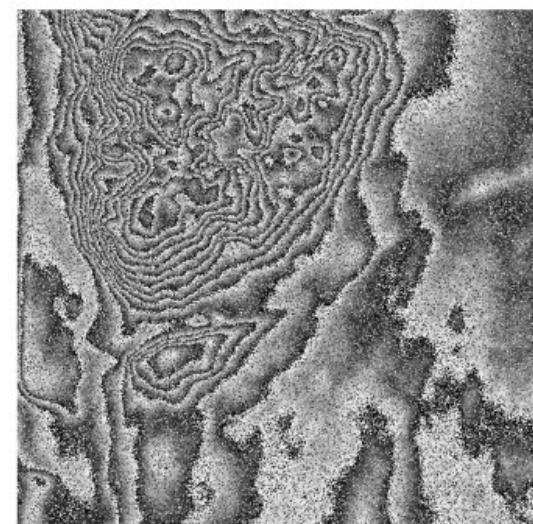
Single-look inSAR phase



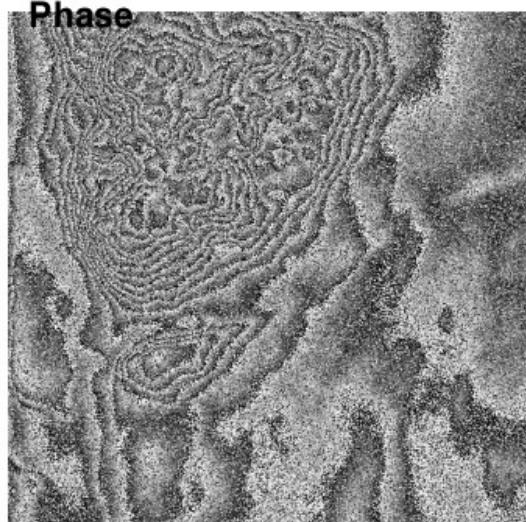
Absolute «True»
Phase



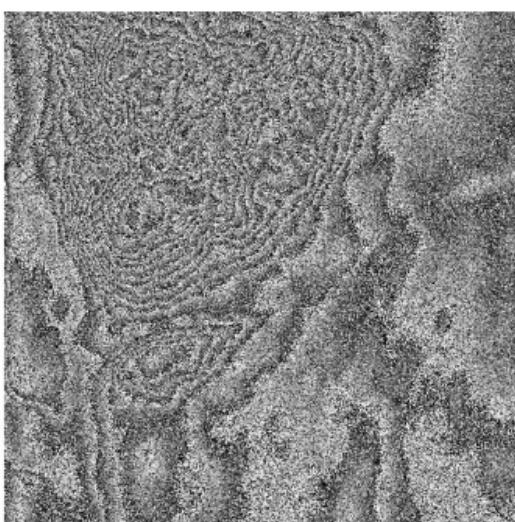
Coherence=1.0 L=1



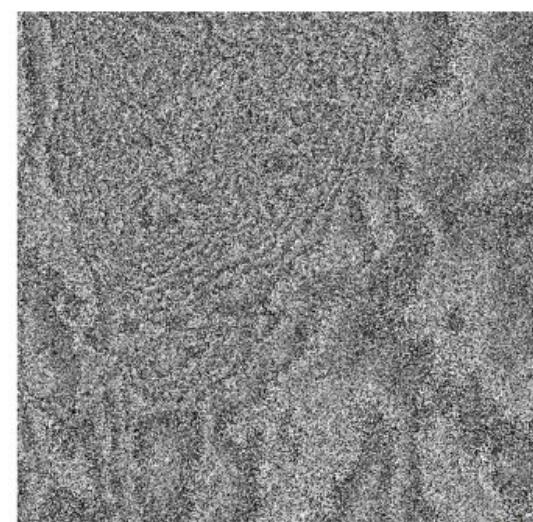
Coherence=0.8 L=1



Coherence=0.6 L=1



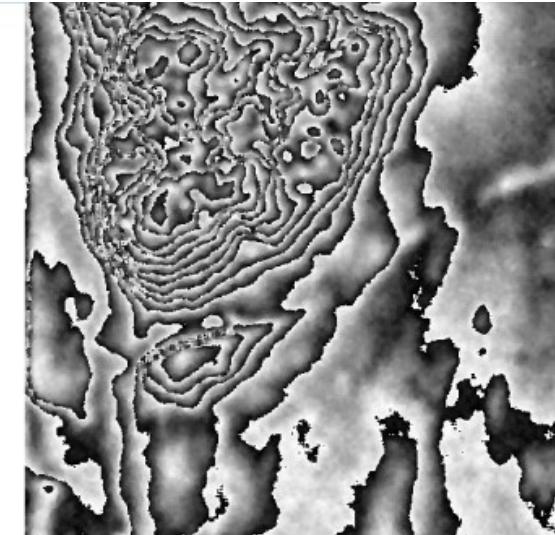
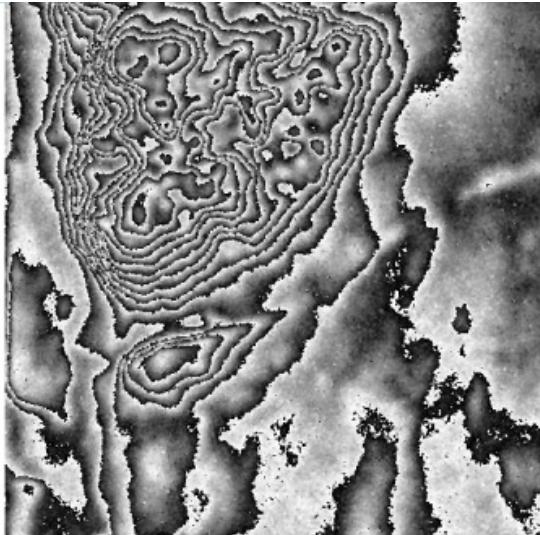
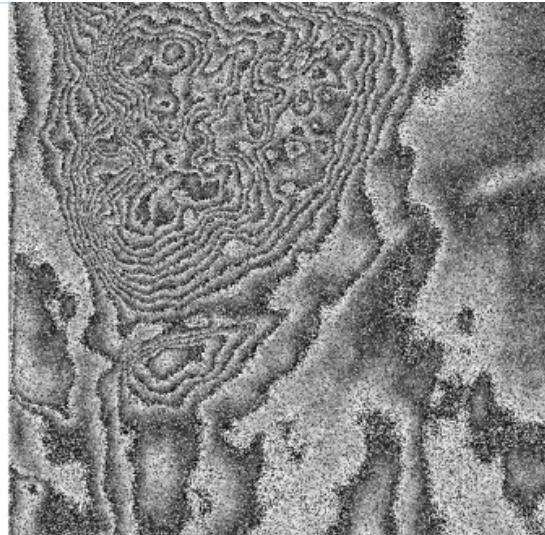
Coherence=0.4 L=1



Coherence=0.2 L=1

InSAR coherence analysis

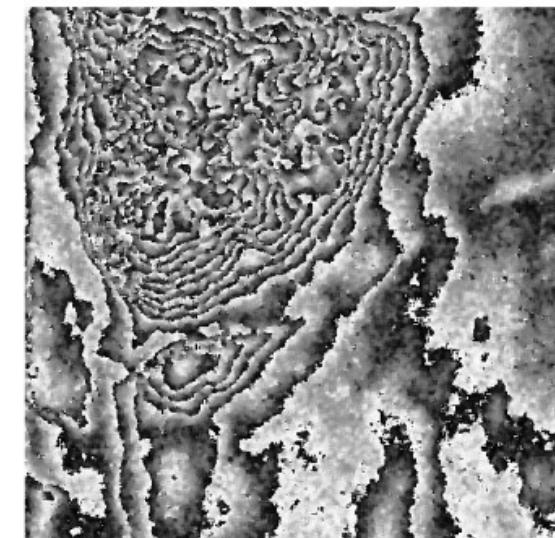
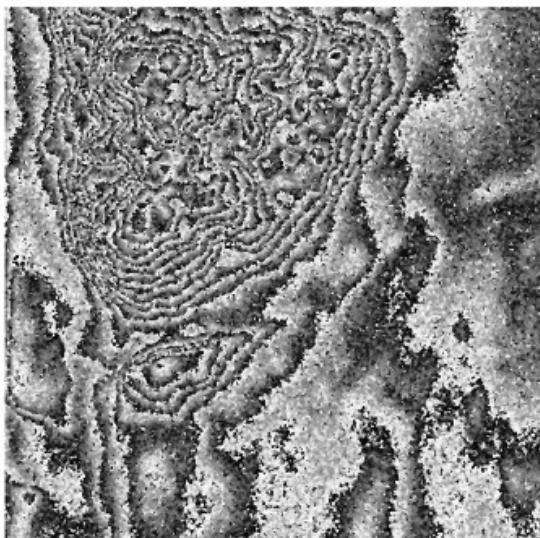
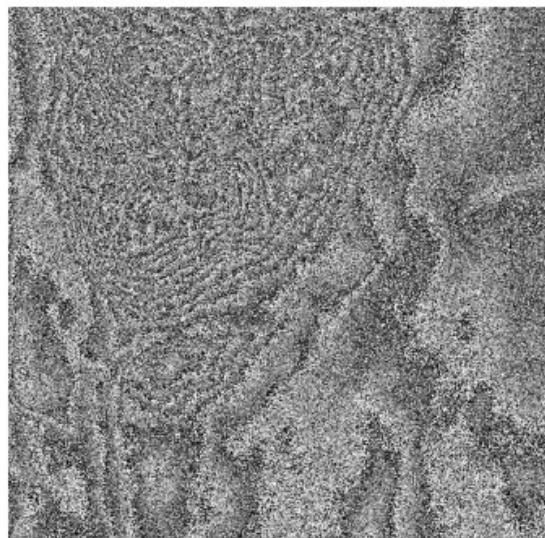
Multi-look inSAR phase



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, γ , 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 ...

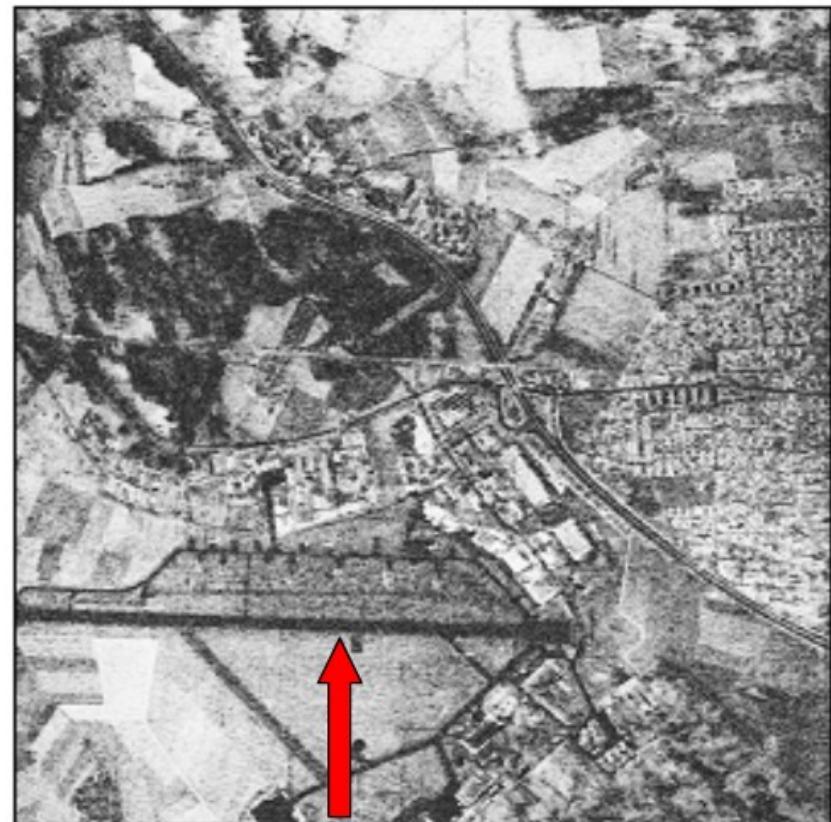
$$\gamma = \gamma_{th} \cdot \gamma_{geom} \cdot \gamma_{vol} \cdot \gamma_{temp} \cdot \gamma_{proc}$$

Thermal or system decorrelation

Intensity image



Coherence image

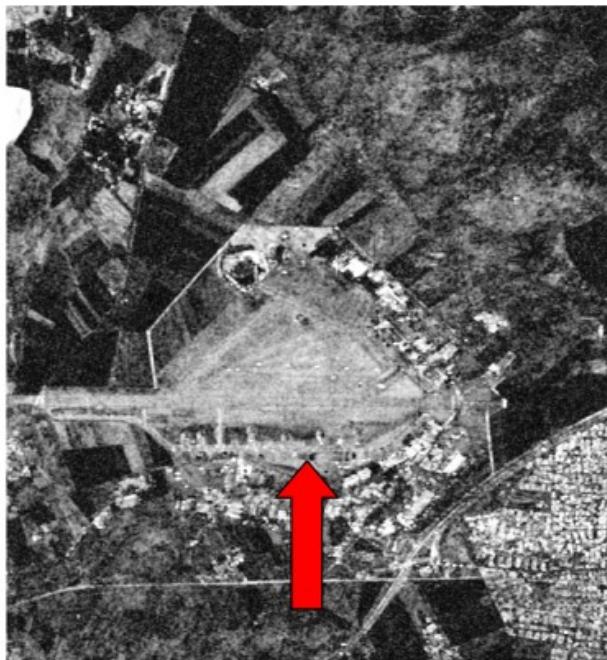


InSAR coherence decomposition

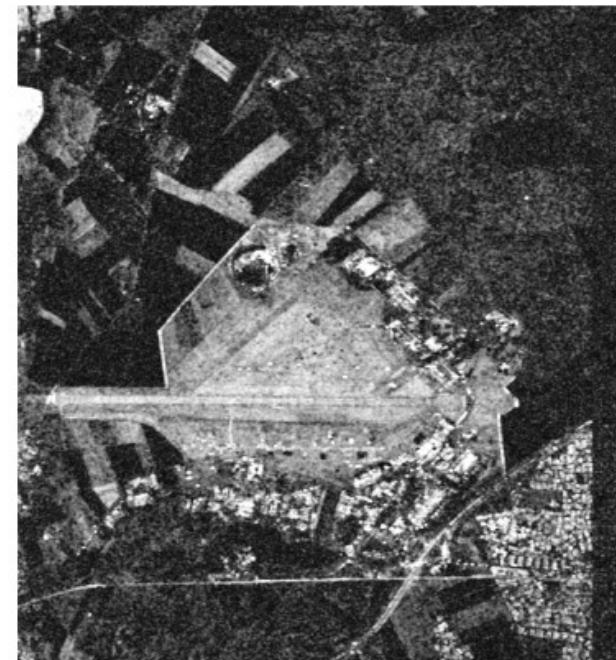
Temporal decorrelation



1 hour, 20m baseline



3 months, 0m baseline



1 year, 0m baseline

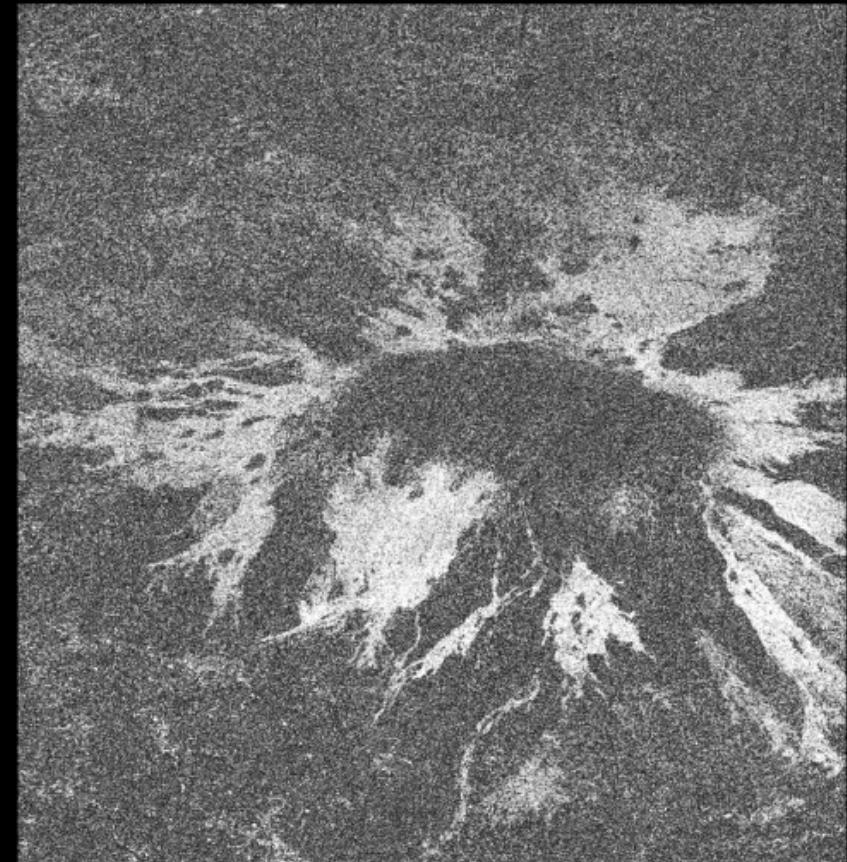
InSAR coherence decomposition

Temporal decorrelation

Coherence Maps



1 day ERS-1/ERS-2



70 days ERS-1/ERS-1

Test Site: Mt. Etna/Italy



Volume decorrelation

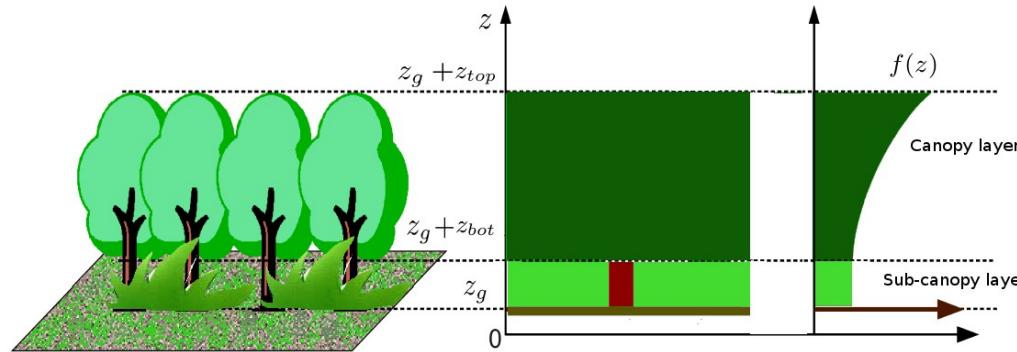
1 hour, 20 baseline



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 $\gamma = \gamma_{th} \quad \gamma_{proc} \quad \gamma_{temp} \quad \gamma_{surf} \quad \gamma_z$
- Decorrelation due to vertical structure :

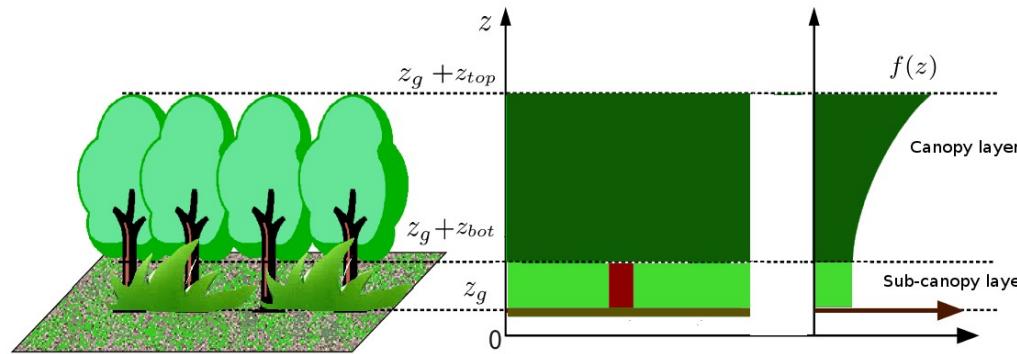
$$\gamma_z = \frac{\int \sigma_{v_e}(z) e^{j k_z z} dz}{\int \sigma_{v_e}(z) dz}$$

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

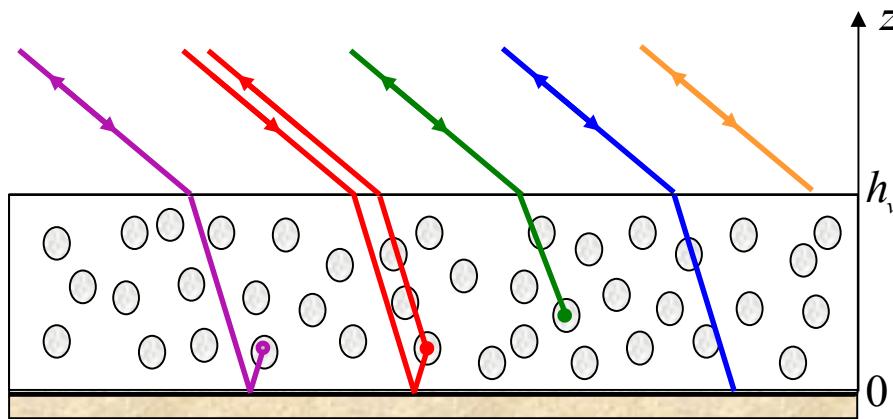
- Fourier transform-like **coherence-structure relationship**

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

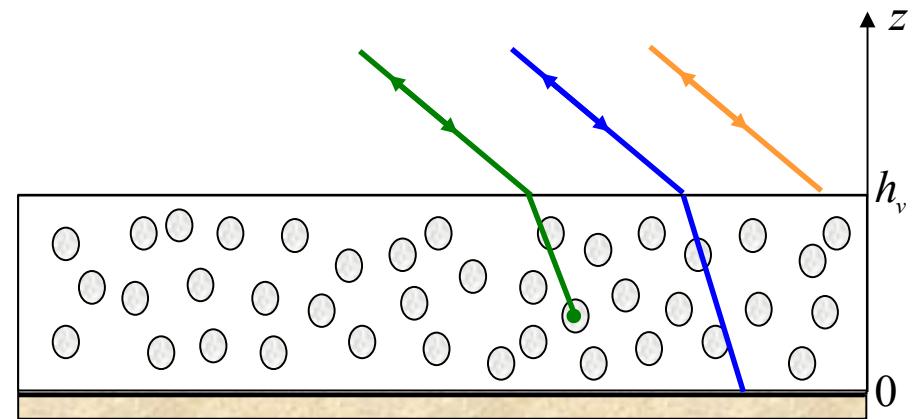
InSAR RVOG model



Modeling at order 1



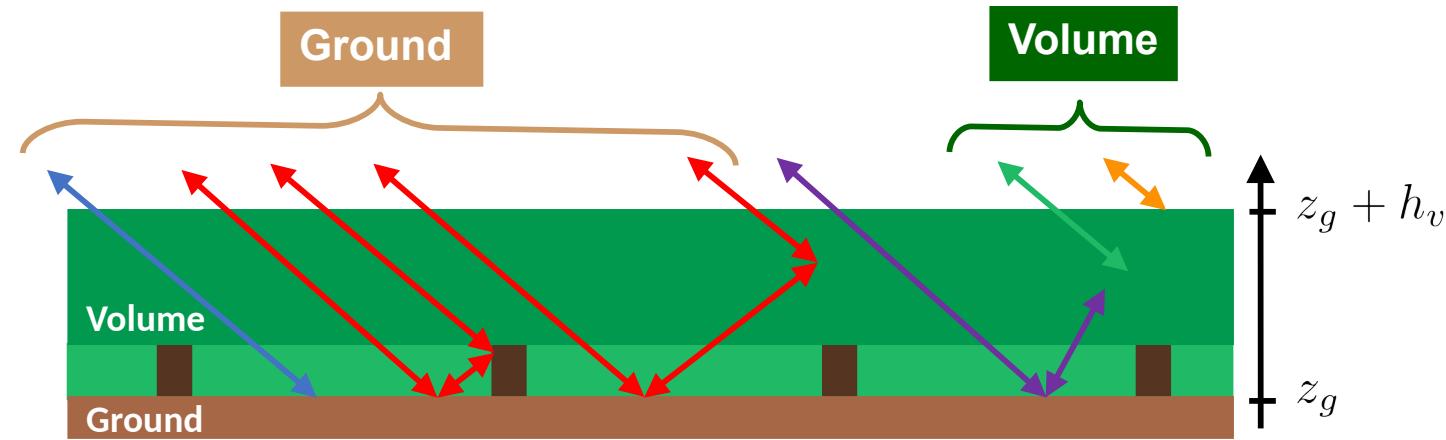
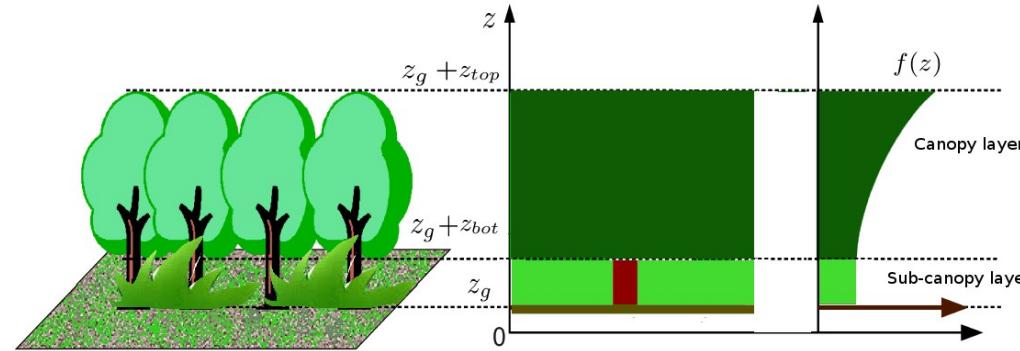
Parameter Estimation



Parameter estimation often requires to simplify models

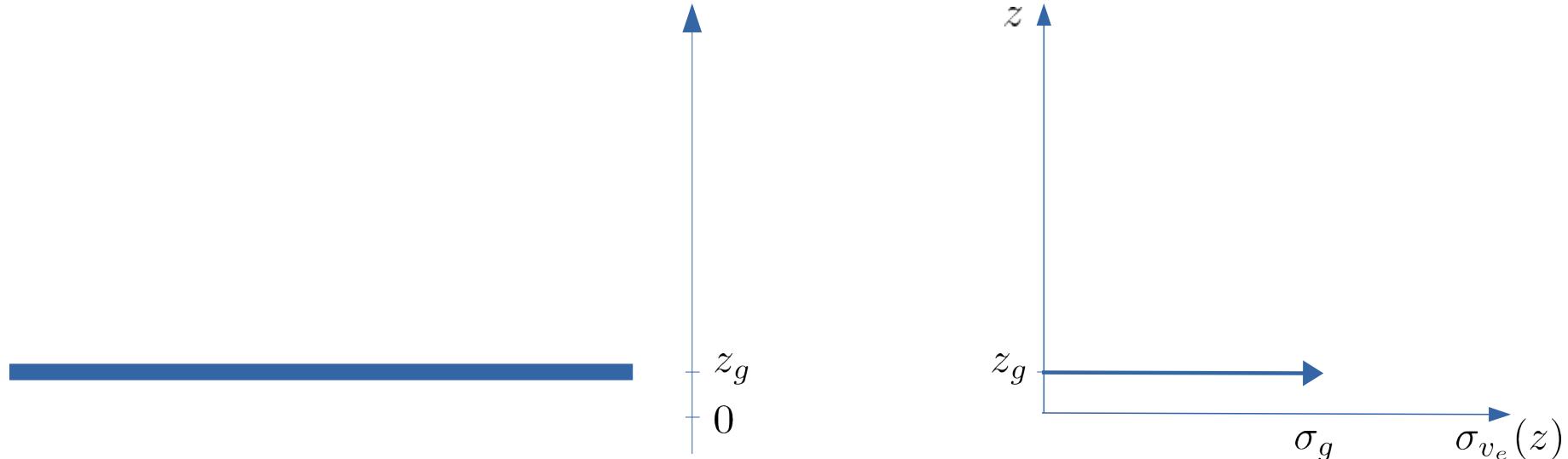
- omitting negligible terms
- merging contributions that cannot be discriminated (e.g. ground and double-bounce)

InSAR RVOG model



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

Ground only



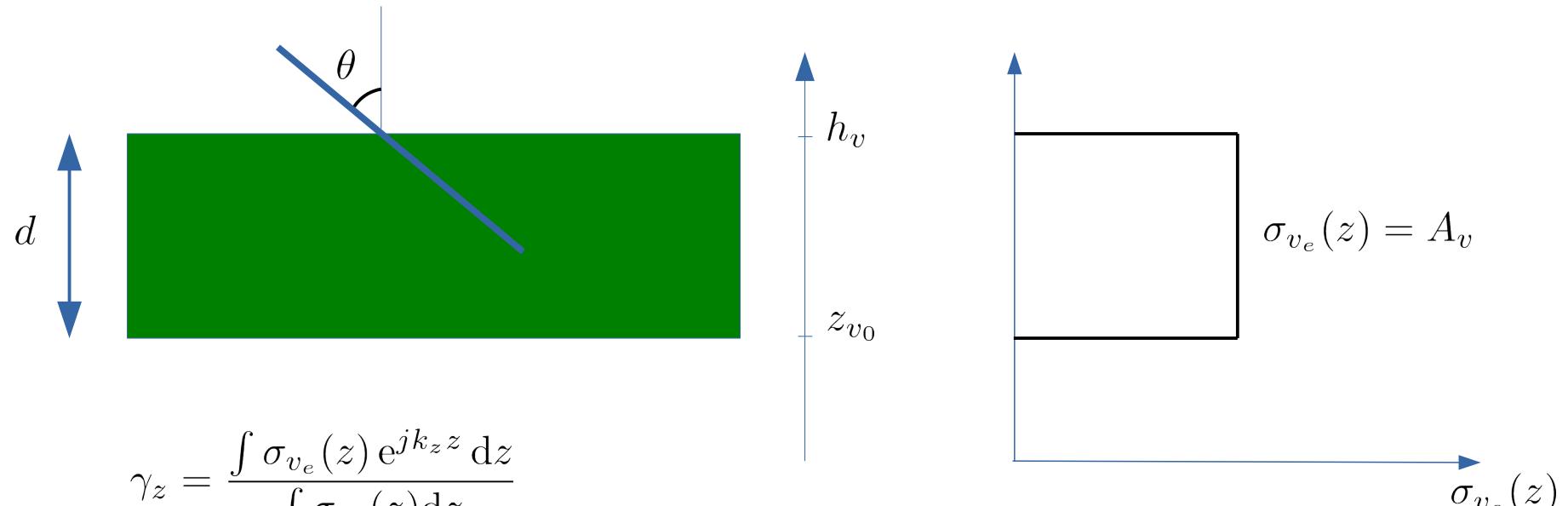
$$\gamma_z = \frac{\int \sigma_{v_e}(z) e^{j k_z z} dz}{\int \sigma_{v_e}(z) dz}$$

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

InSAR well adapted to topography estimation

InSAR RVOG analysis

Non attenuating random volume only



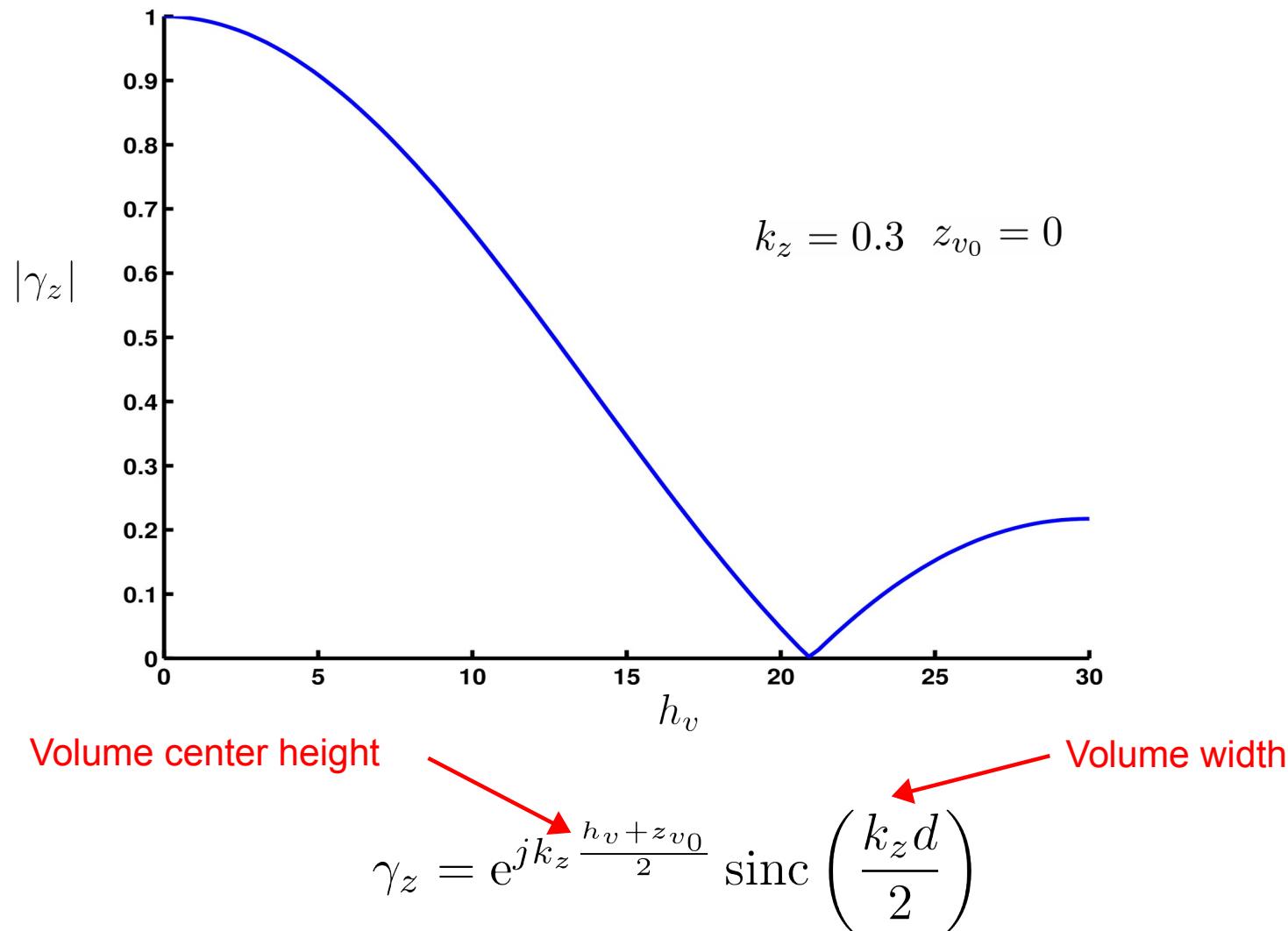
No underlying ground

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

$$\gamma_z = \frac{1}{d} \int_{z_{v_0}}^{h_v} e^{j k_z z} dz$$

$$\gamma_z = e^{j k_z \frac{h_v + z_{v_0}}{2}} \operatorname{sinc}\left(\frac{k_z d}{2}\right)$$

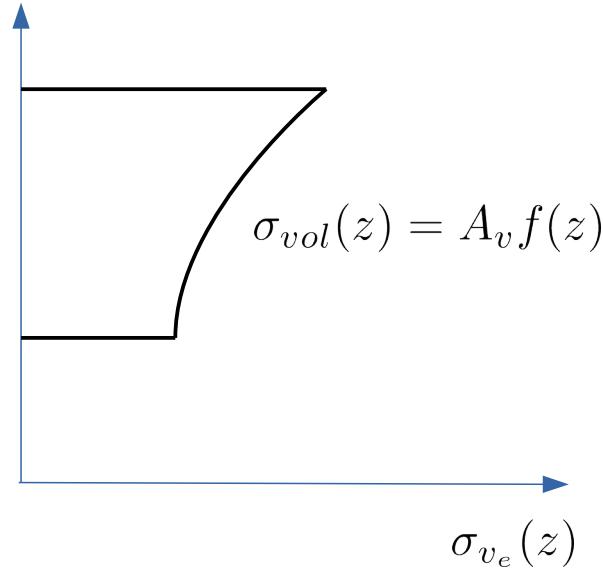
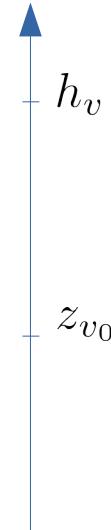
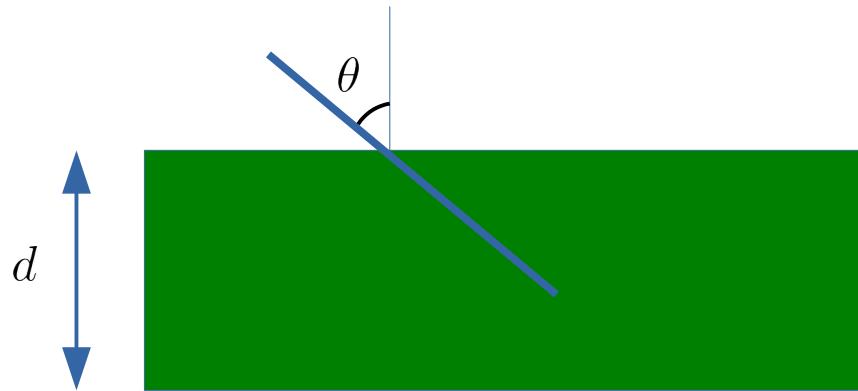
InSAR RVOG analysis



InSAR well adapted to volume analysis under specific conditions

InSAR RVOG analysis

Attenuating random volume only



Linear differential extinction

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

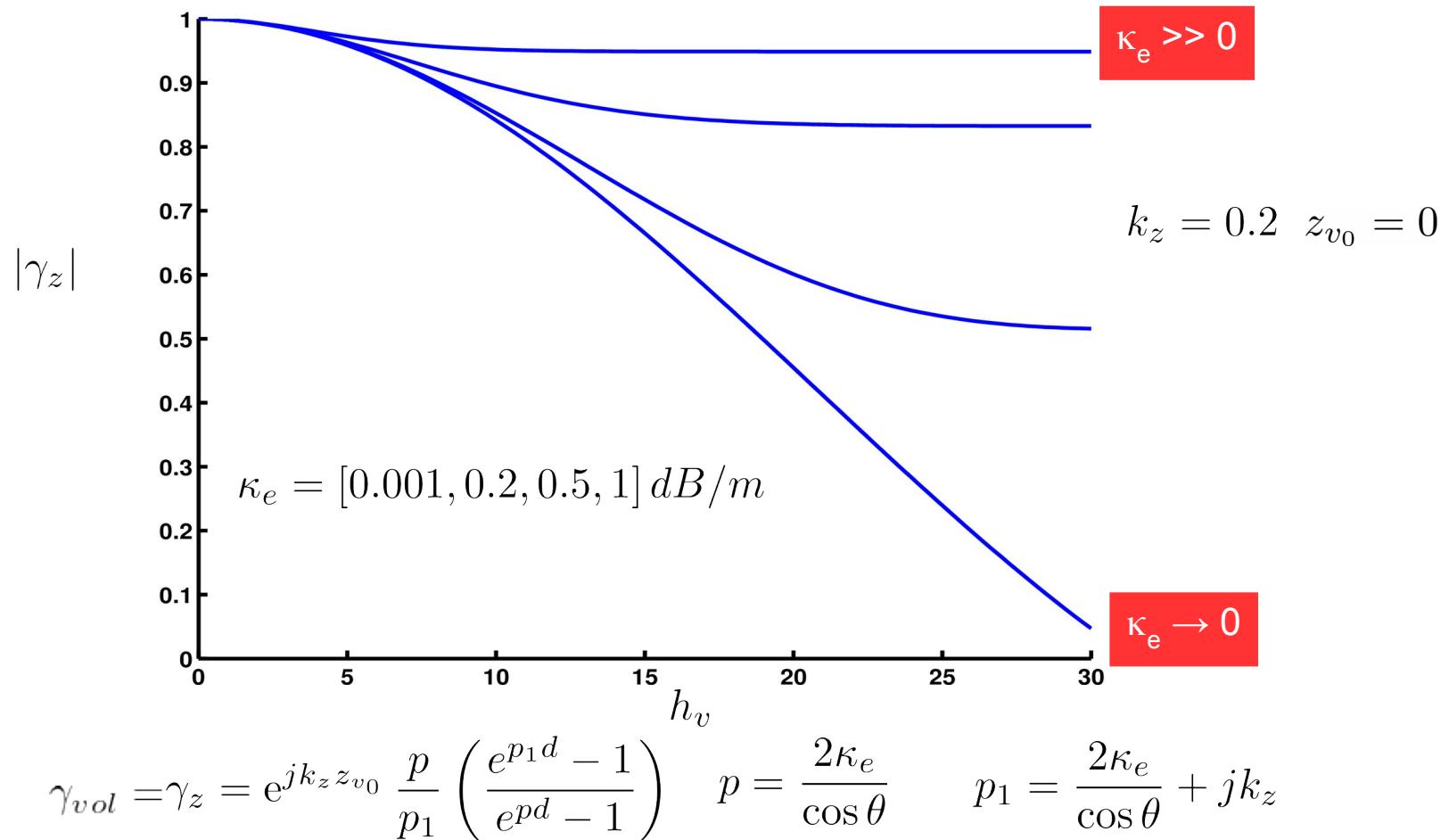
Effective reflectivity density
constant extinction

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

Backscattered volume intensity

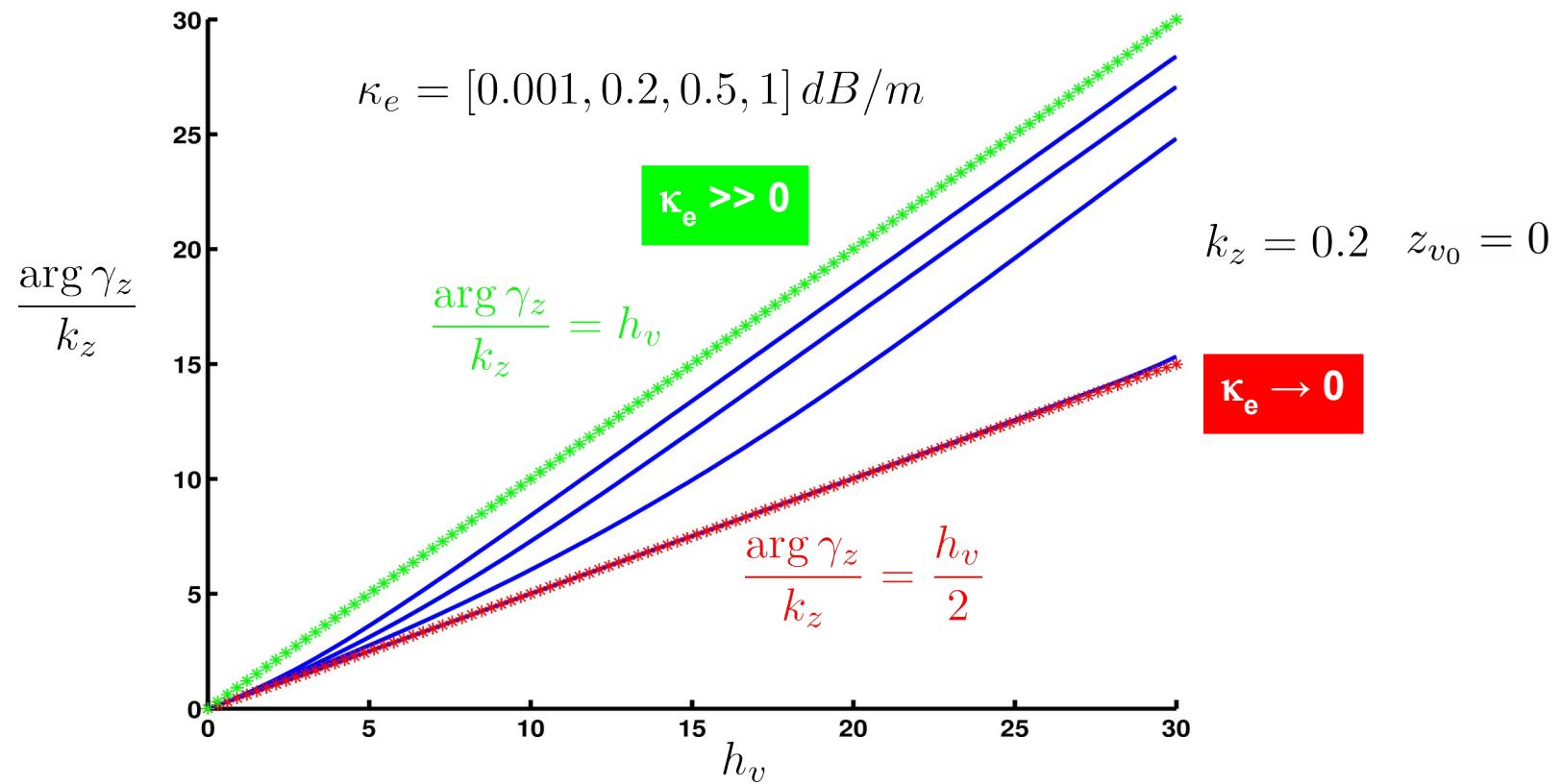
$$I_v = \int_{z_{v_0}}^{h_v} \sigma_{vol}(z) dz = \int_{z_{v_0}}^{h_v} A_v f(z) dz$$

InSAR RVOG analysis



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

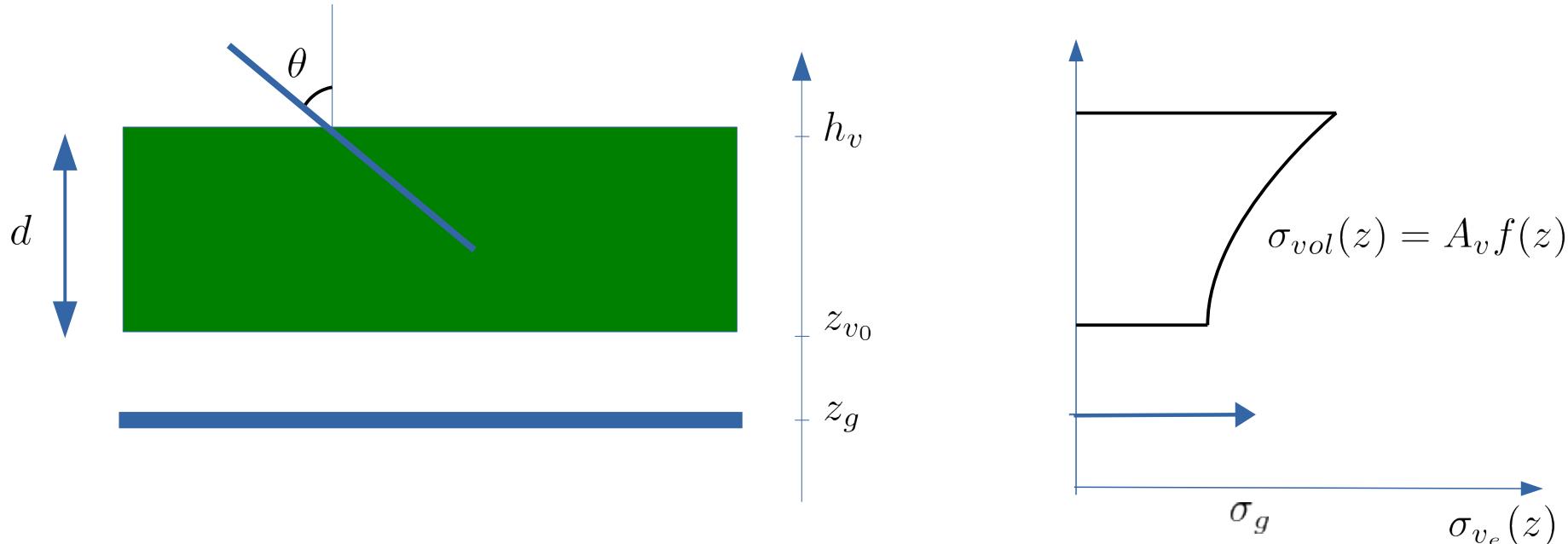
InSAR RVOG analysis



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

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

$$\boxed{\gamma_z = \frac{\gamma_{vol} + m e^{j k_z z_g}}{1 + m}}$$

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

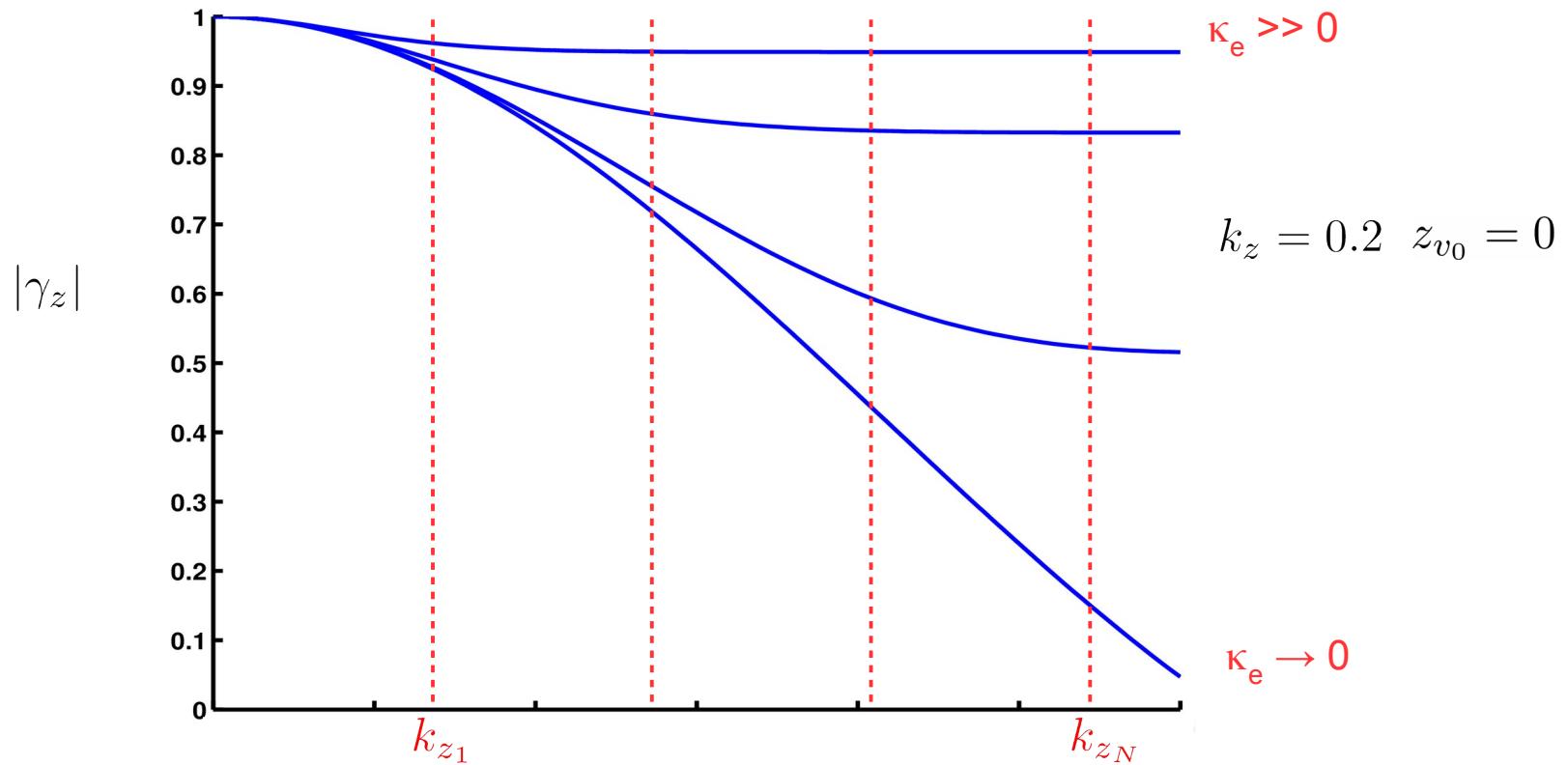
$$m \rightarrow 0 \Rightarrow \begin{cases} \arg \gamma_z \approx \arg \gamma_{vol} \\ |\gamma_z| \leq 1 \end{cases} \quad m \rightarrow +\infty \Rightarrow \begin{cases} \arg \gamma_z \approx \phi_g \\ |\gamma_z| = 1 \end{cases}$$

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

InSAR based RVOG analysis: under-determined problem

→ another source of diversity is needed : polarization ?

TomoSAR (MB-InSAR) RVOG analysis

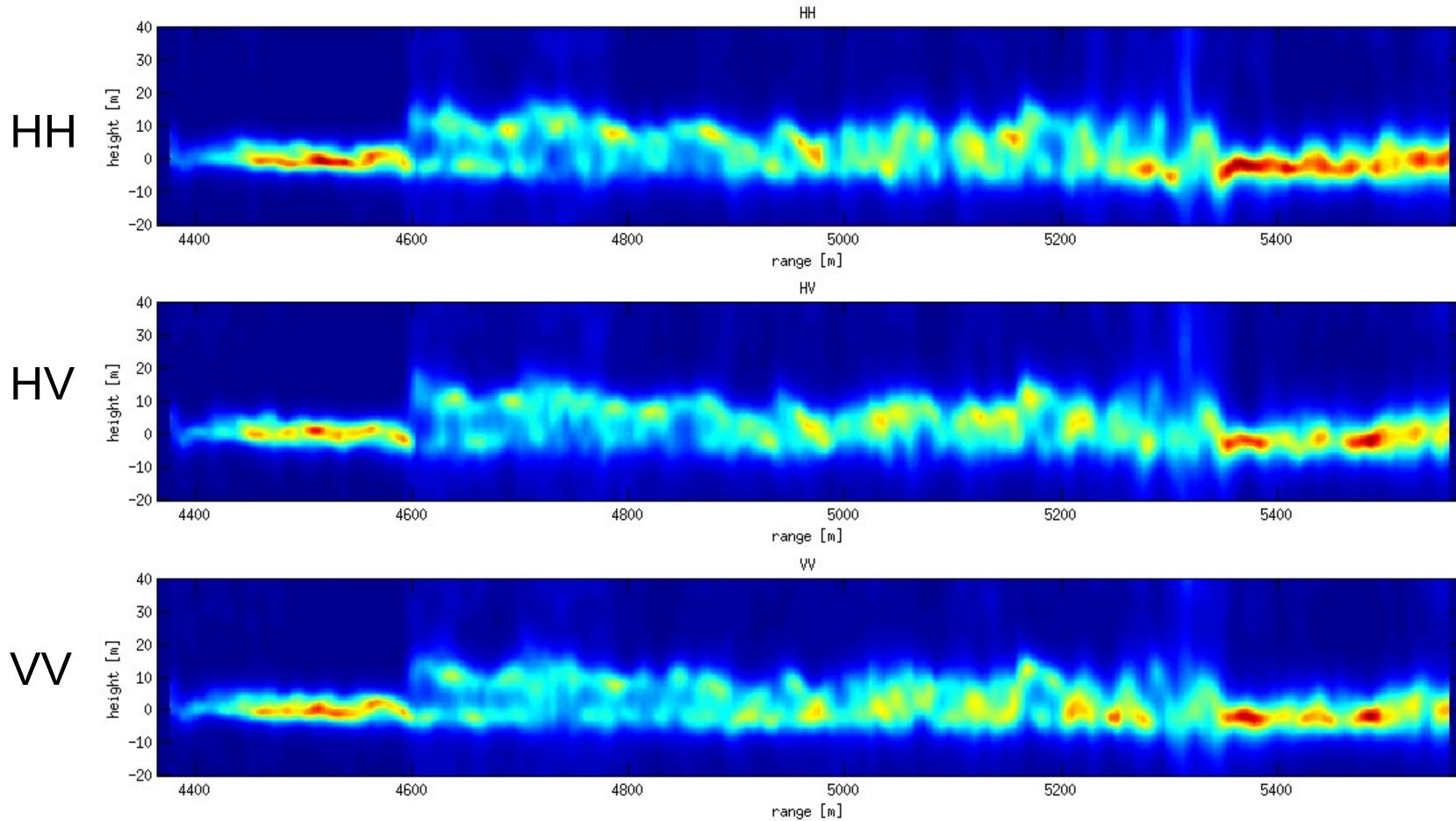


Additional spatial diversity

- $\{\gamma_z(k_{z_n})\}_{n=1}^N \longrightarrow$ {
- 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

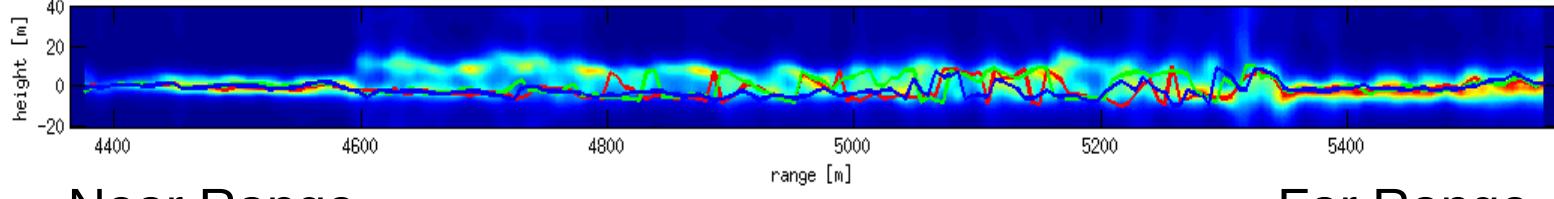
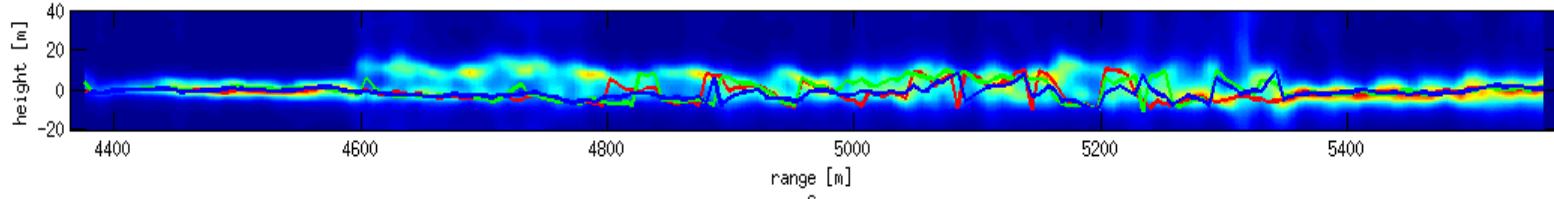
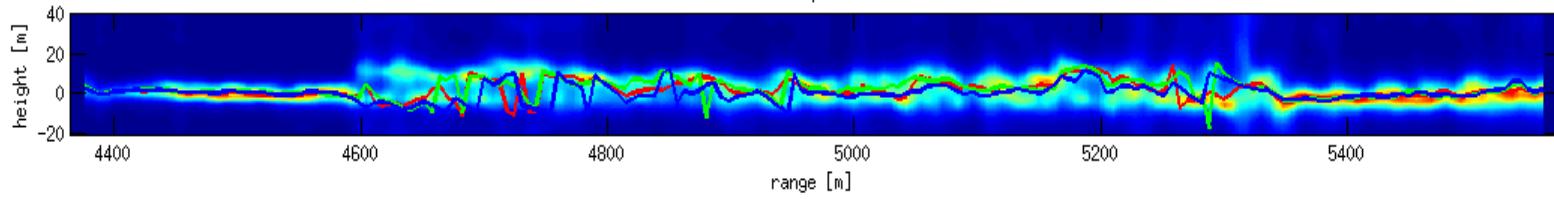
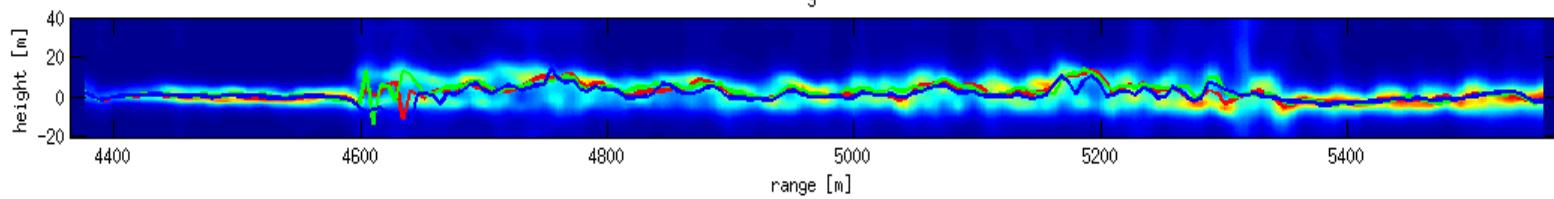
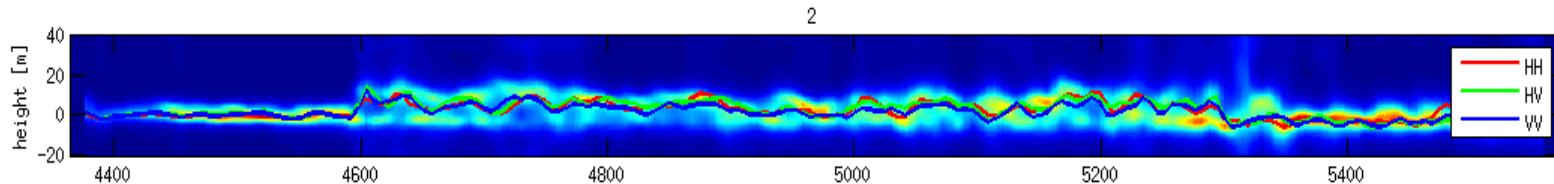
L-band BIOSAR2, Capon tomograms



InSAR phases, polarization & TomoSAR

InSAR phase center heights

$h(\phi_{12})$



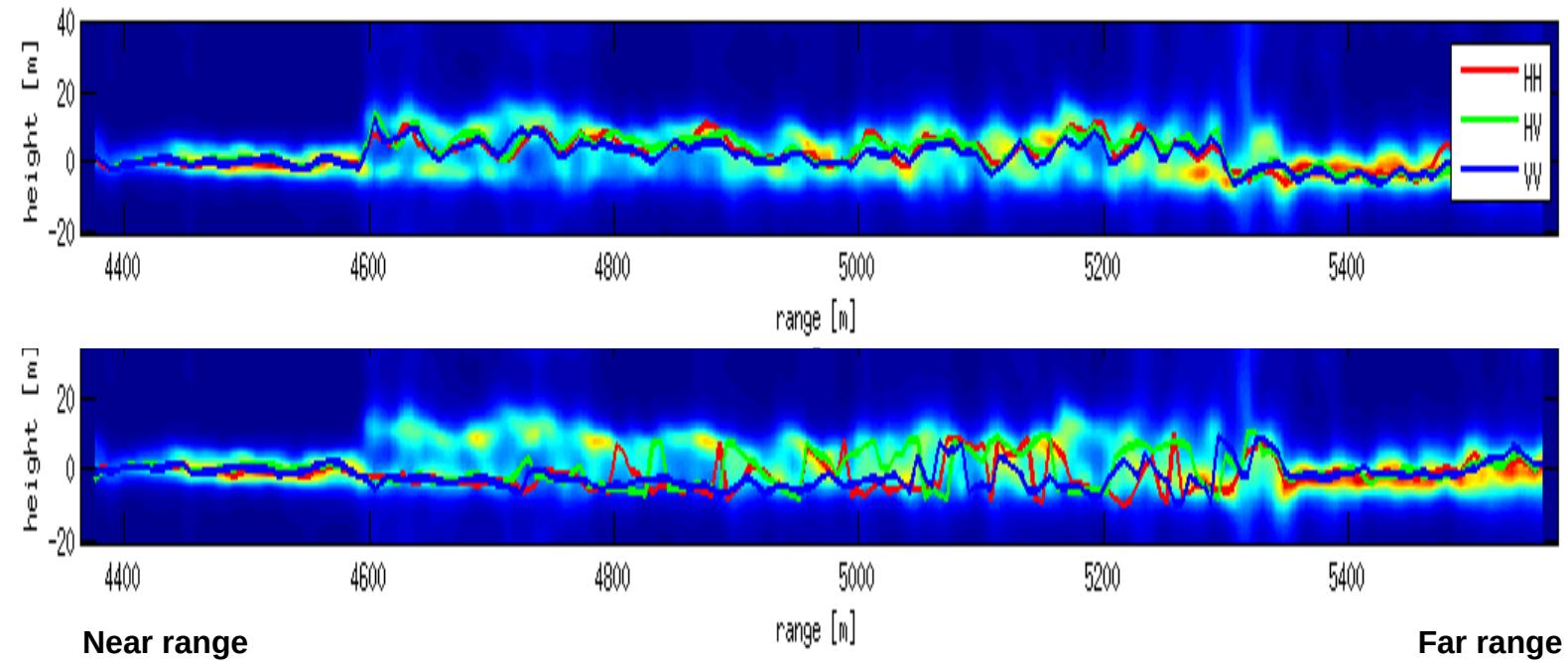
Near Range

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

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

Campaign	BioSAR 2008 - ESA
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 range)

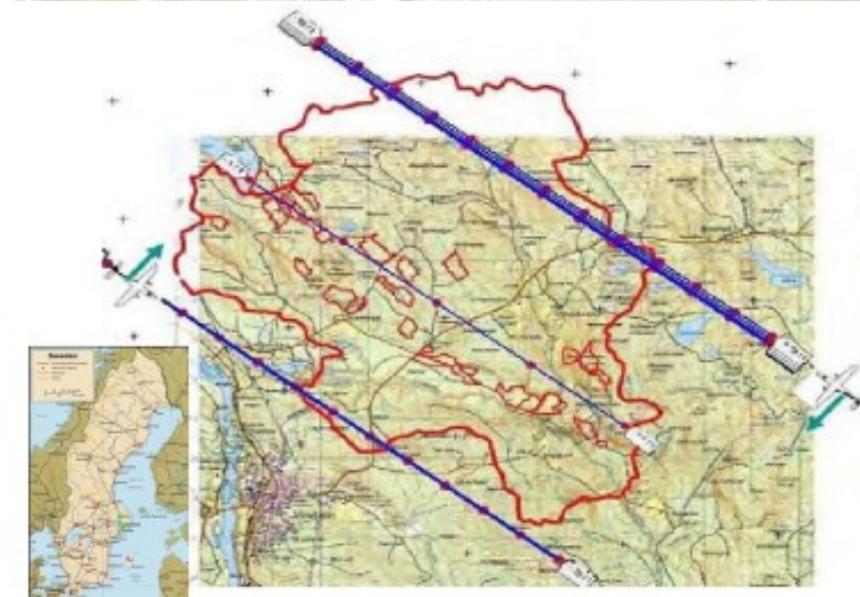
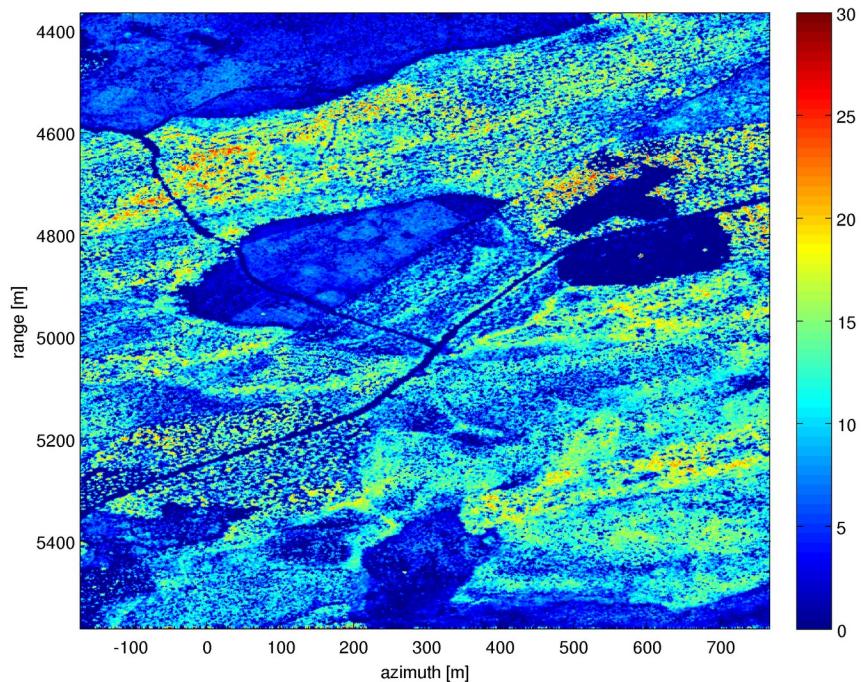
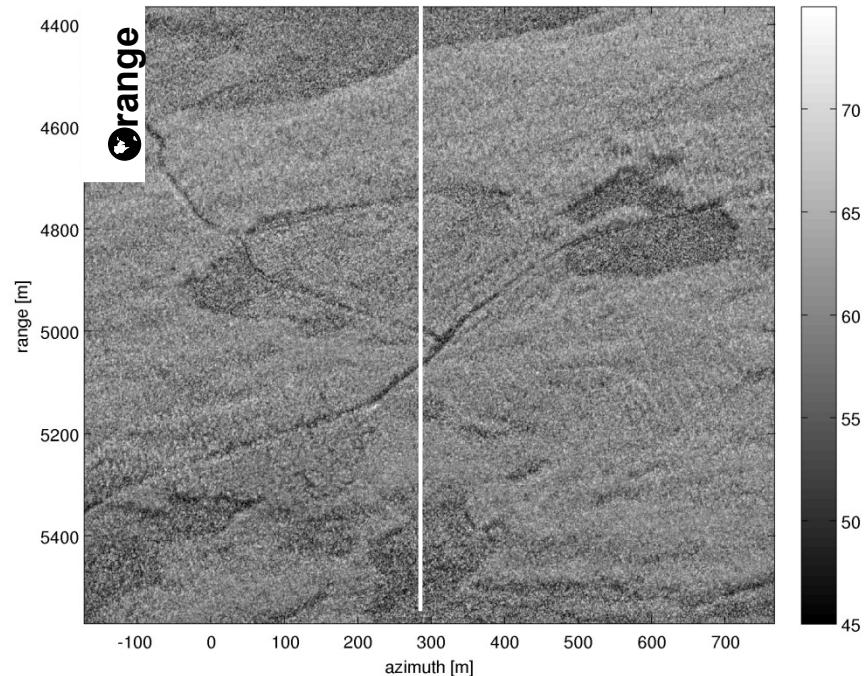


Illustration of coherence features

Forest height



HH intensity



DEM

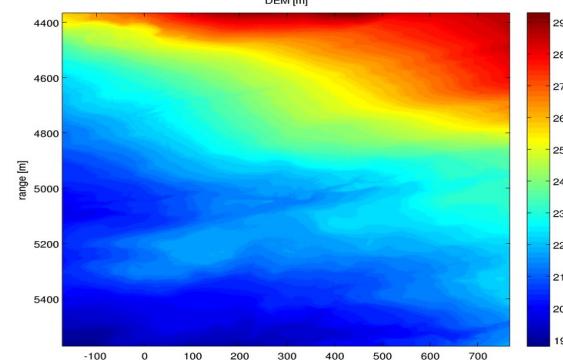
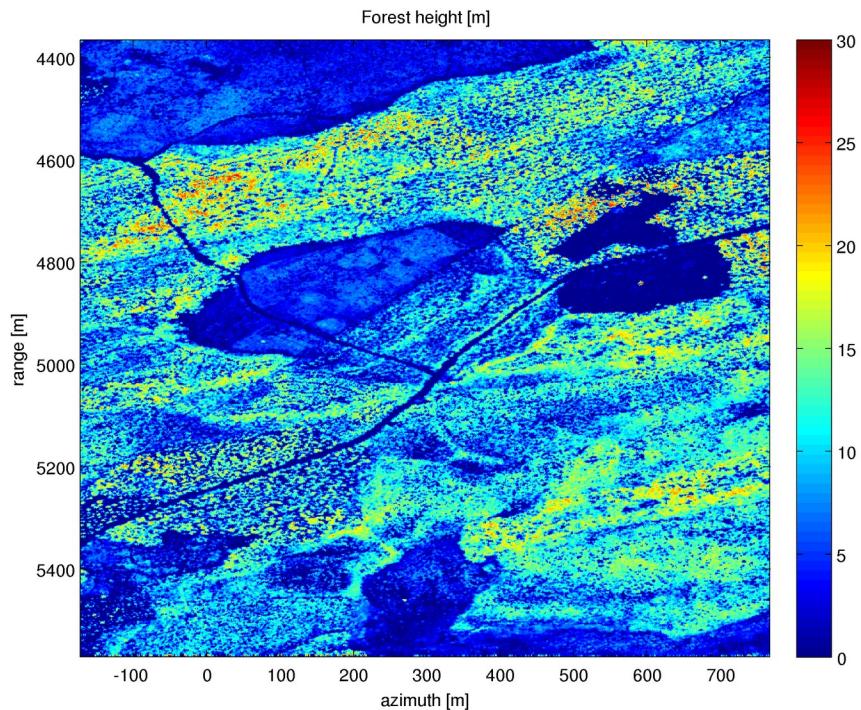
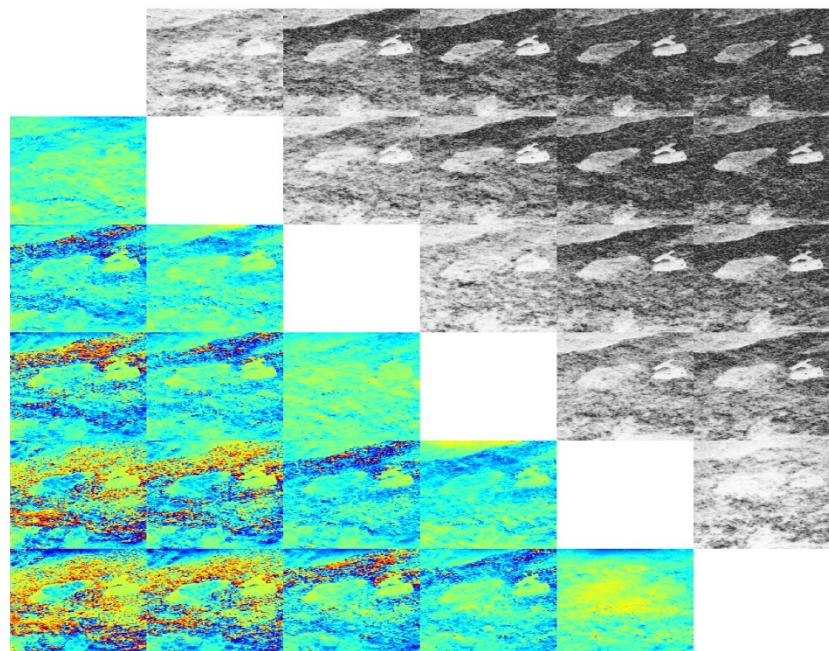


Illustration of coherence features

Forest height



Multi-baseline InSAR coherences and phases - VV



DEM

