

Comparison of low dimensional profile models for the characterization of tropical forests using SAR tomography

Pierre-Antoine Bou ^(1,2), Laurent Ferro-Famil^(2,3), Frederic Brigui⁽¹⁾, Yue Huang⁽²⁾, Ludovic Villard⁽²⁾

pierre-antoine.bou@onera.fr

⁽¹⁾ ONERA, DEMR/TSRE, Palaiseau, France

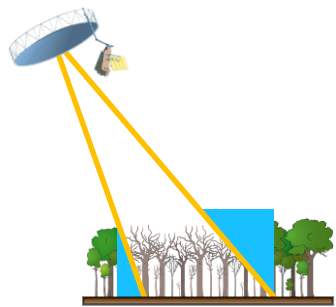
⁽²⁾ CESBIO, University of Toulouse, Toulouse, France

⁽³⁾ ISAE-SUPAERO, University of Toulouse, DEOS Dept., Toulouse, France

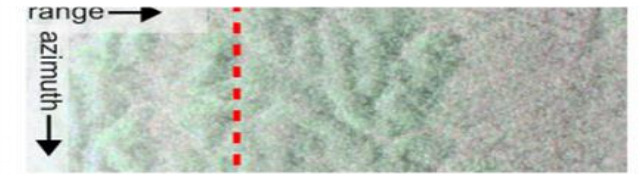
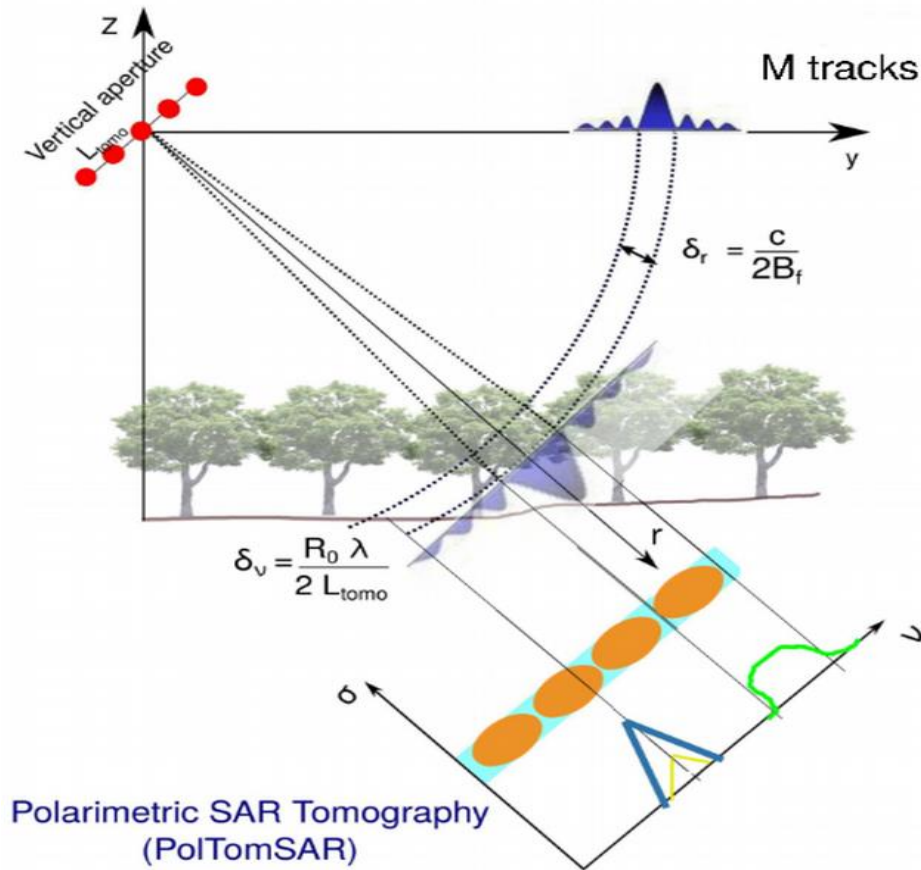
Context : Need for Parametric Tomography

BIOMASS mission

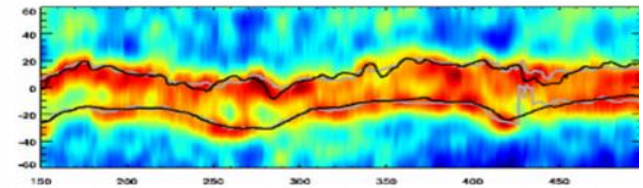
ONERA TropiSAR campaign



P-band radar



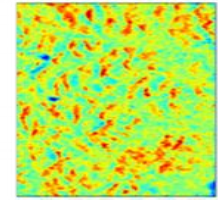
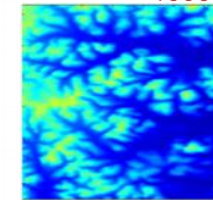
HH



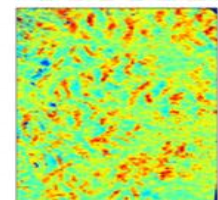
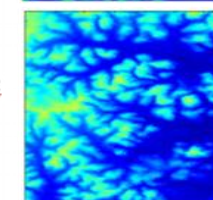
Sub-canopy DTM

Forest height

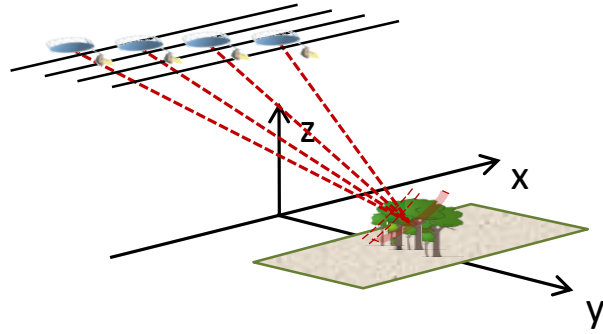
Lidar



TomoSAR

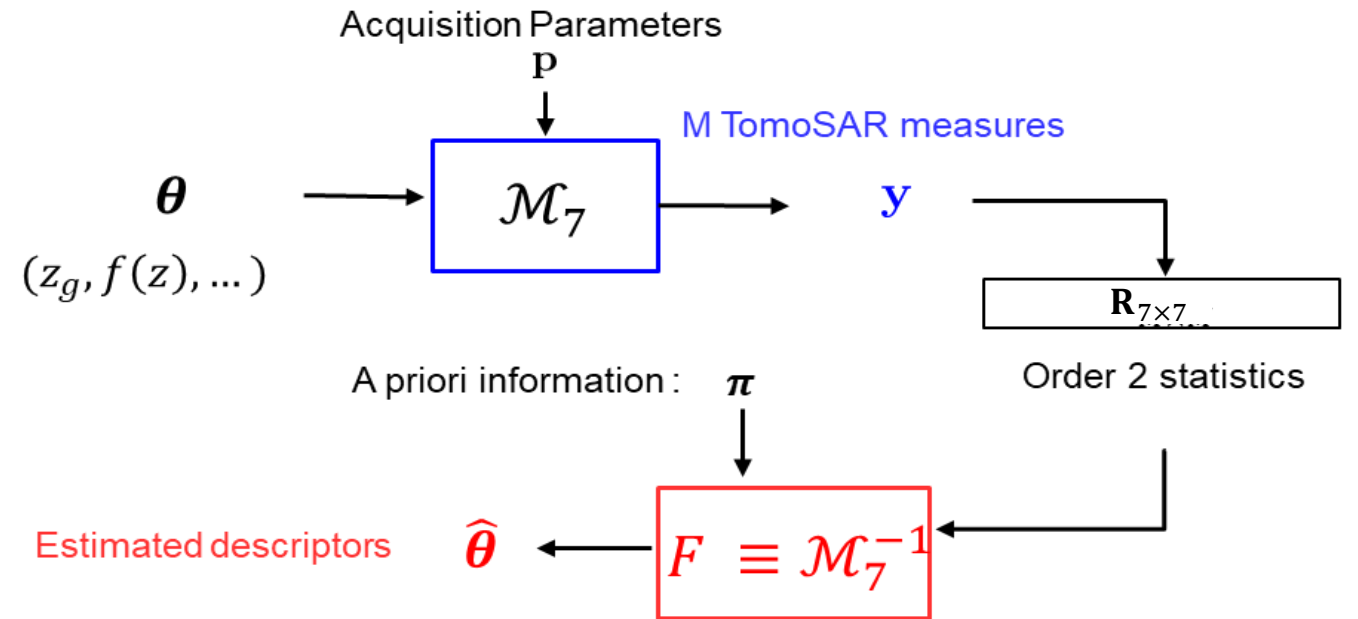


Parameter extraction in Tomography

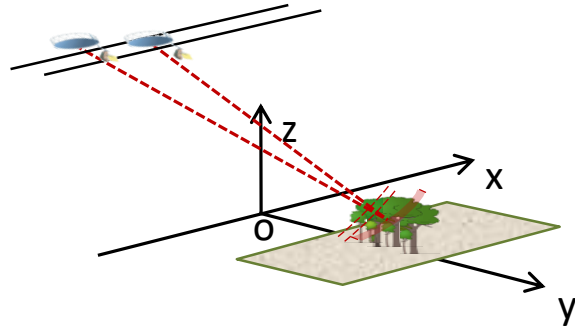


Tomographic phase
 Global coverage : 16 months
 M = 7 images (3 days REV)

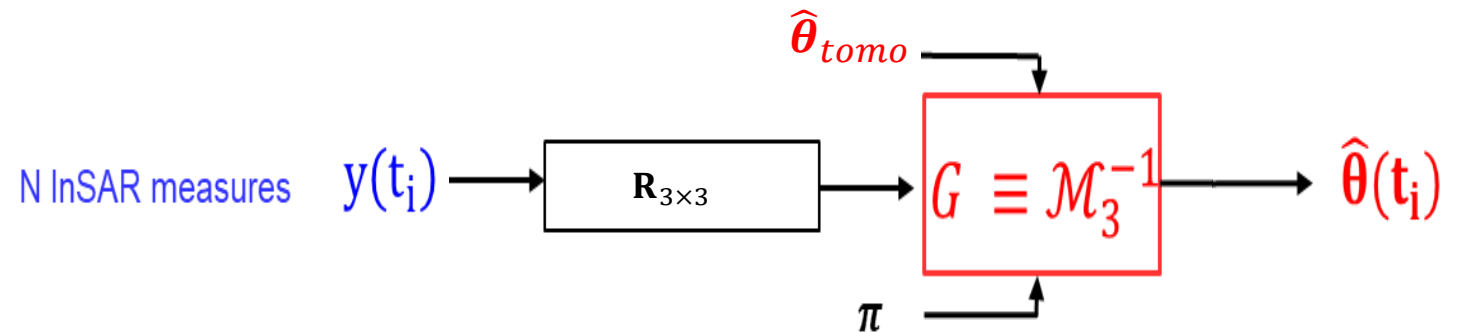
High Resolution



Parameter extraction by models with few parameters

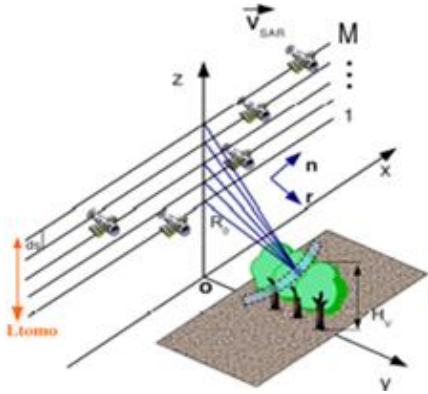


Interferometric cycles
Every 8 months
M = 3 images (3 days REV)
Low Resolution

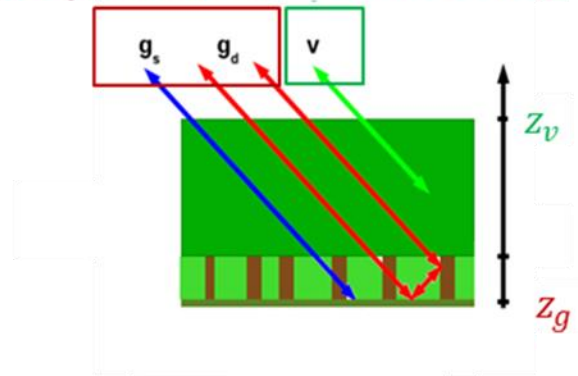


Objective : Choice of a low dimensionality model

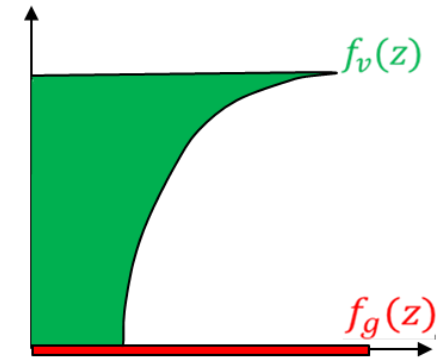
TomoSAR Forest Response Model



Scattering mechanisms :
Ground response **Volume response**



Vertical reflectivity Profile



$$\mathbf{R}_{M \times M} = I \begin{bmatrix} 1 & \gamma_{12} & \cdots & \gamma_{1M} \\ \vdots & \vdots & \ddots & \vdots \\ \gamma_{M1} & \cdots & \cdots & 1 \end{bmatrix} = \mathbf{R}_V + \mathbf{R}_g$$

$$\gamma(k_{z_{ij}}) = \frac{\int f_x(z) e^{jk_{z_{ij}}z} dz}{I}$$

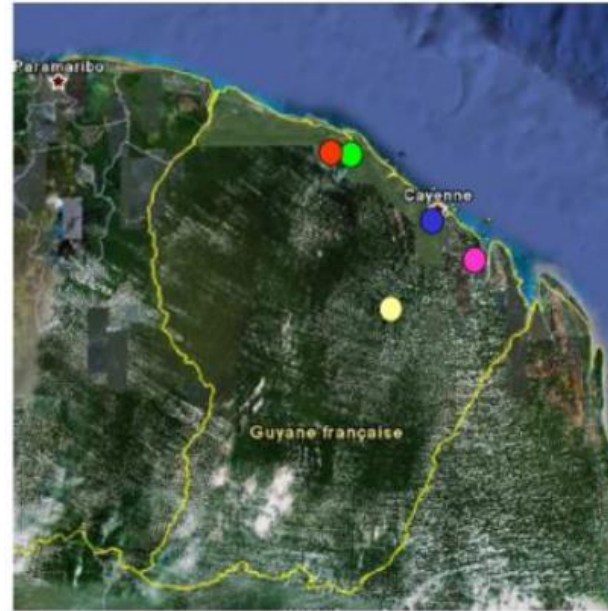
$$f_x(z) = \hat{f}_v(z) + \hat{f}_g(z)$$

Inverse Problem

Estimate $\hat{f}_v(z)$, $\hat{f}_g(z)$ using a low dimensional parametrization

Comparison of reflectivity profile basis

- TropiSAR Campaign, 2009
- ONERA SETHI
- P-band
- 6 pass
- $\delta_{az} = 1.245m$
- $\delta_{rg} = 1m$
- $\delta_z = 12.5m$



The ECOFOG Sites

- Nouragues
- Paracou
- Arboce

The Calibration site

- Rochambeau

Other site

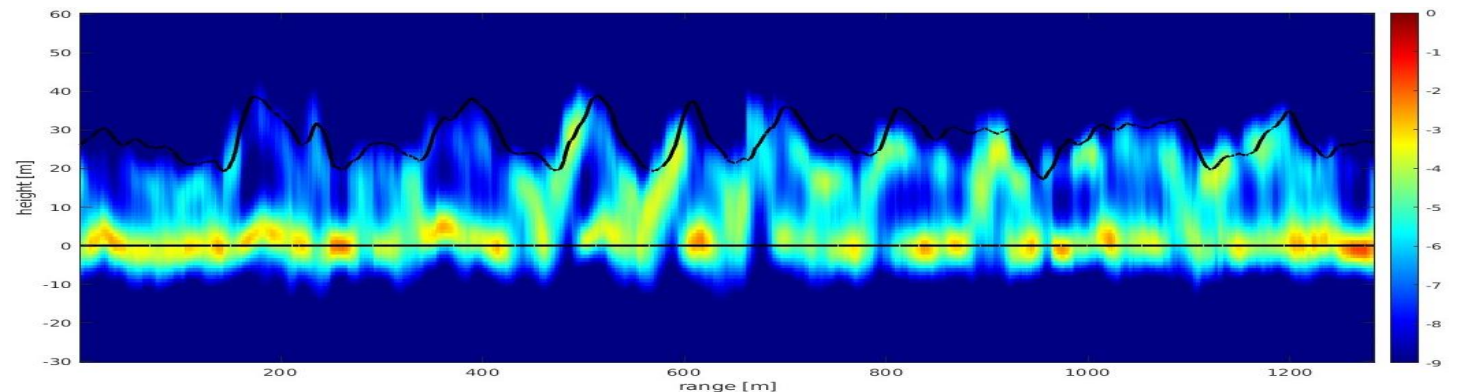
- Marais de Kaw

French Guyana test site



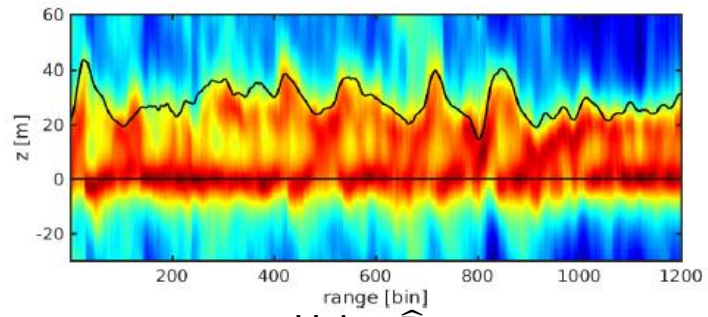
Courtesy ONERA

Capon Tomogram

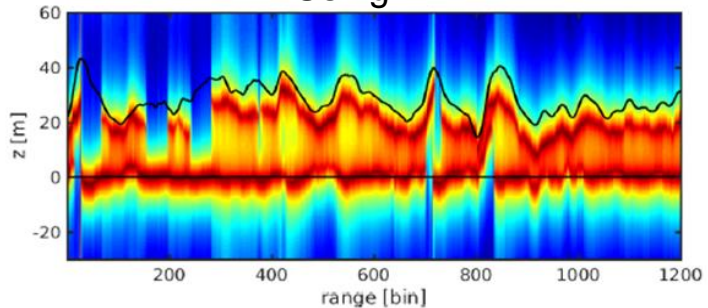


Estimation of the model order for single PolTomography

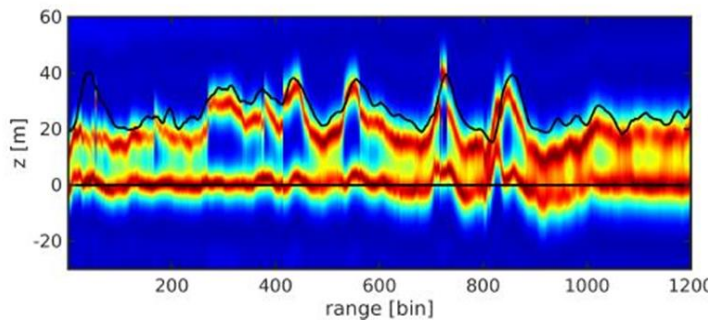
Capon Tomogram :



Using \hat{R}



Using an adaptive basis (wavelets)



Using an exponential Profile

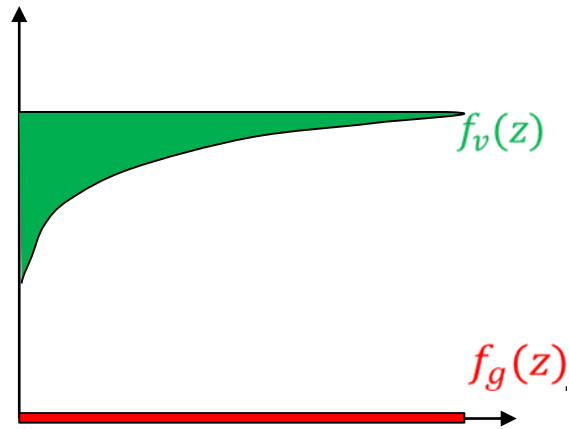
- Modelled with an adaptive basis & sparse signal estimation
↳ 2 components
"Forest structure characterization using SAR tomography and an adaptative estimation technique", EuSAR 2022

- Using a fixed basis : Exponential Volume + Ground
↳ Similar Results

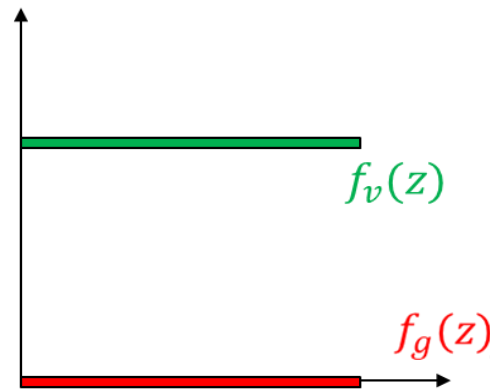
- Notice : the Volume component is always located **below** the LiDAR upper limit estimate

→ Fixed 2 components basis :
Parametric estimation of Volume & Ground

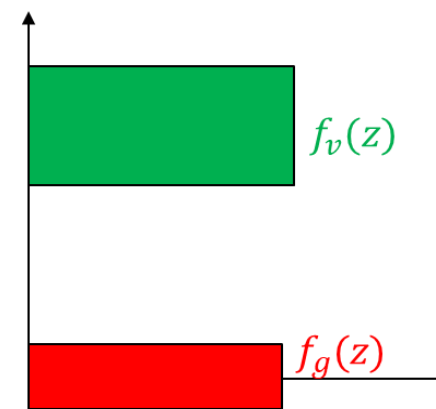
Proposed 2 components reflectivity models



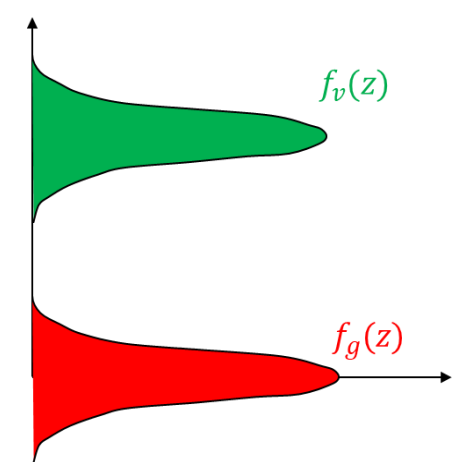
Exponential & Dirac



2 Diracs



2 Boxes

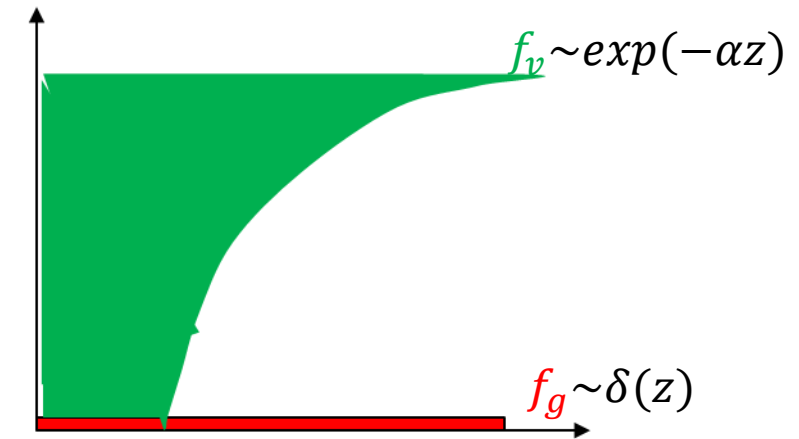
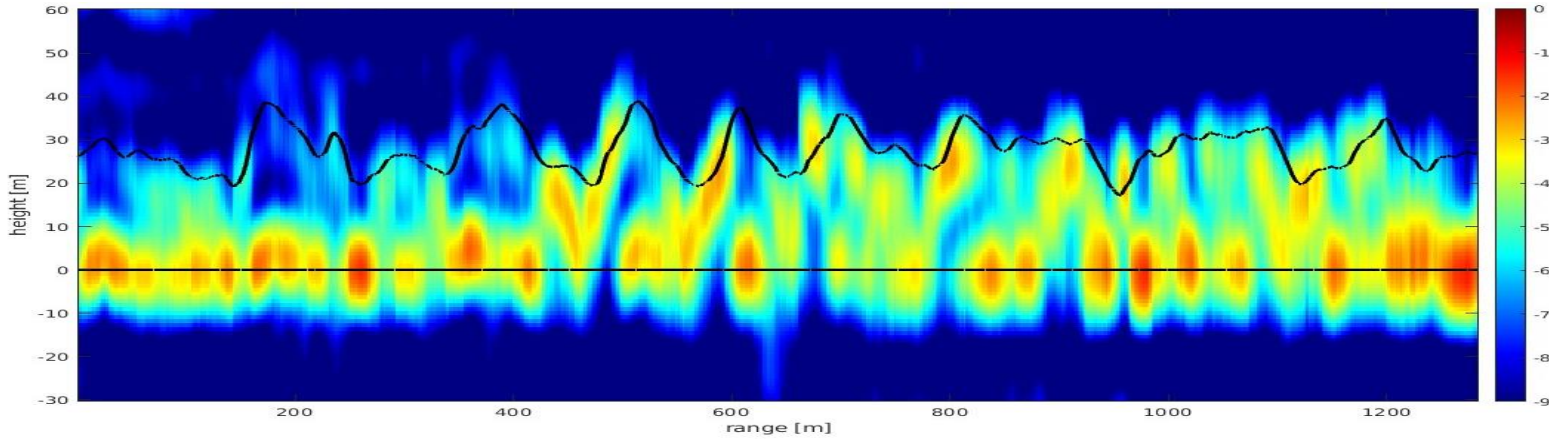


2 Gaussian components

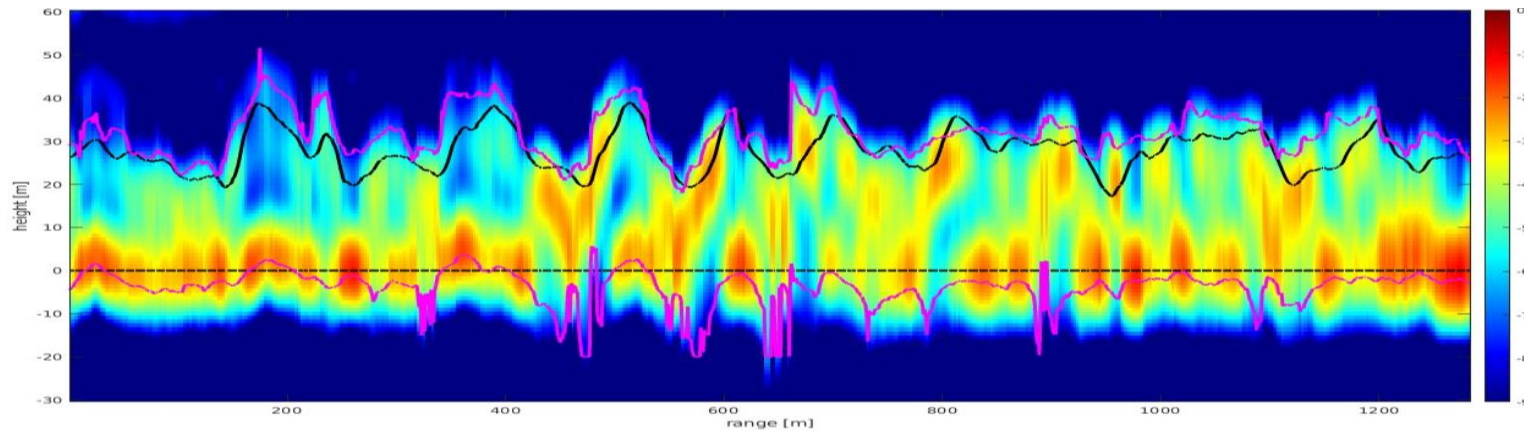
Retrieved z_g , z_v & reflectivity profile for a single polarization tomogram

Exponential model

Original Tomogram



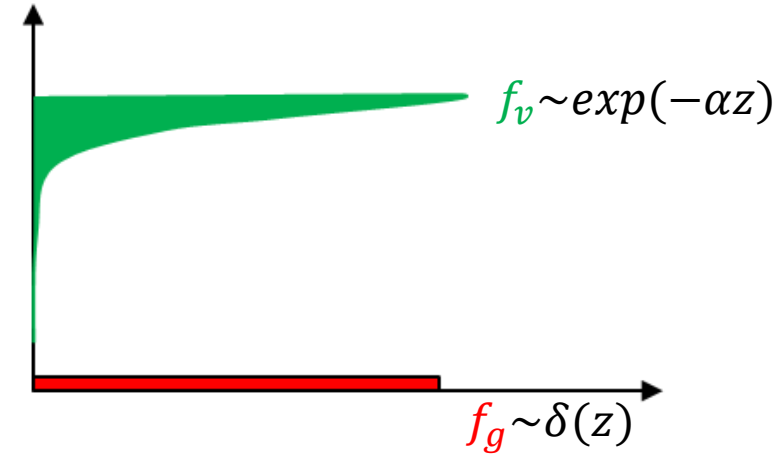
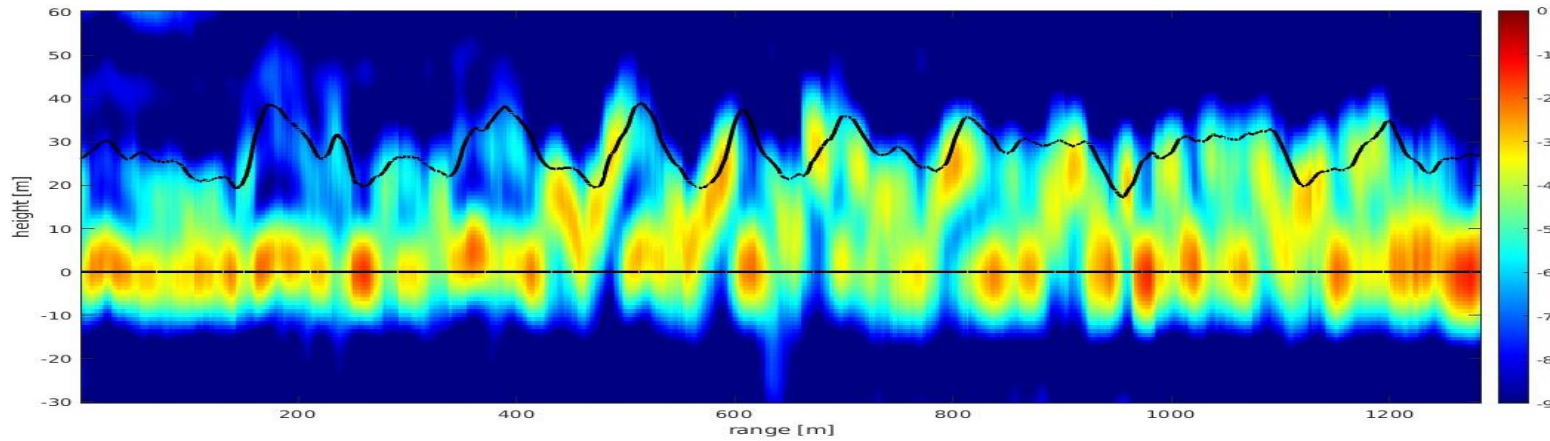
Reconstructed tomogram



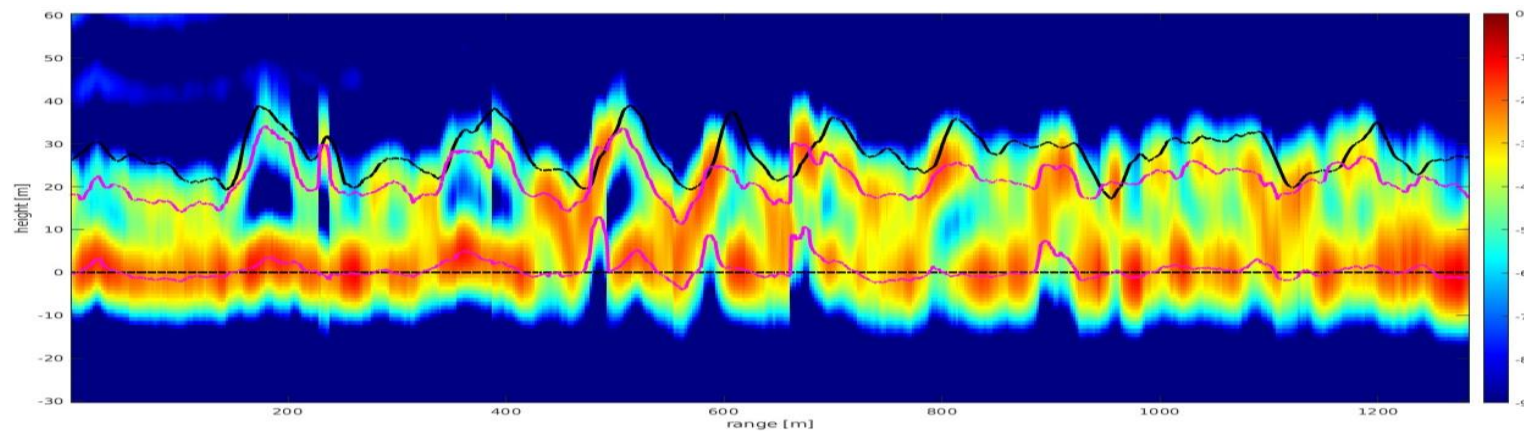
- Overestimated volume
- Underestimated ground

Exponential model

Original Tomogram



Constraint : large α values



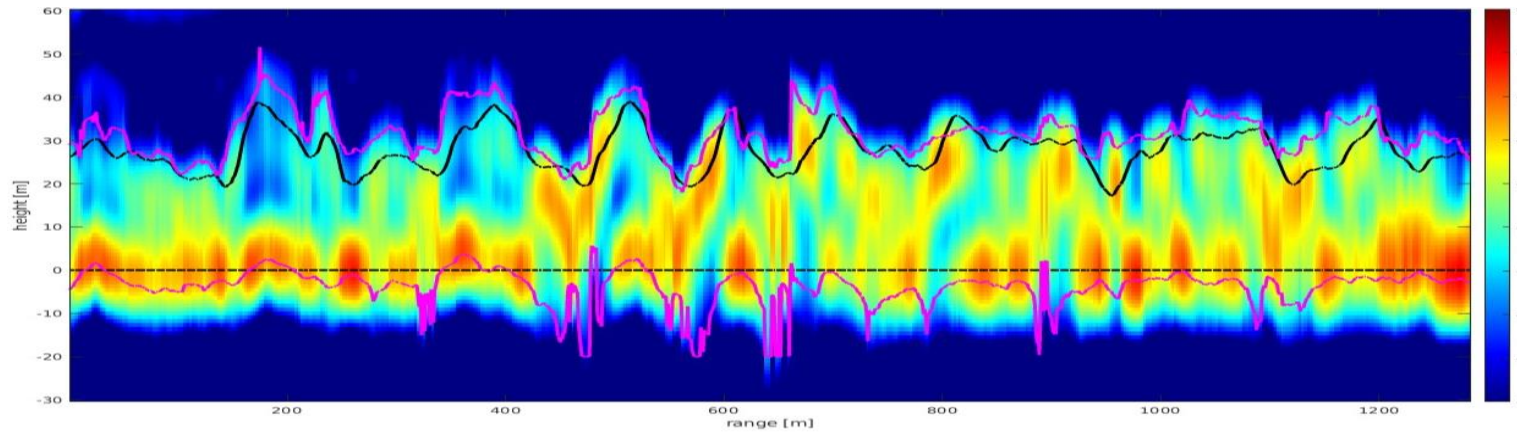
Constraint retrieval :

Correct z_g, z_v estimates

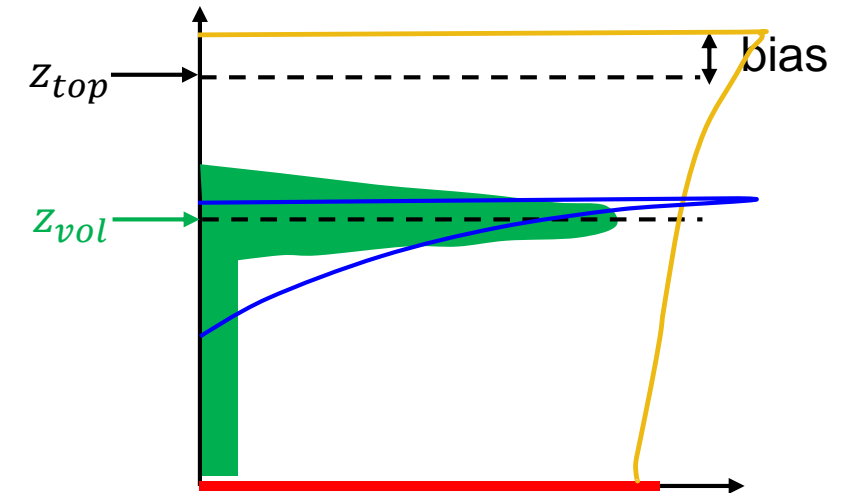
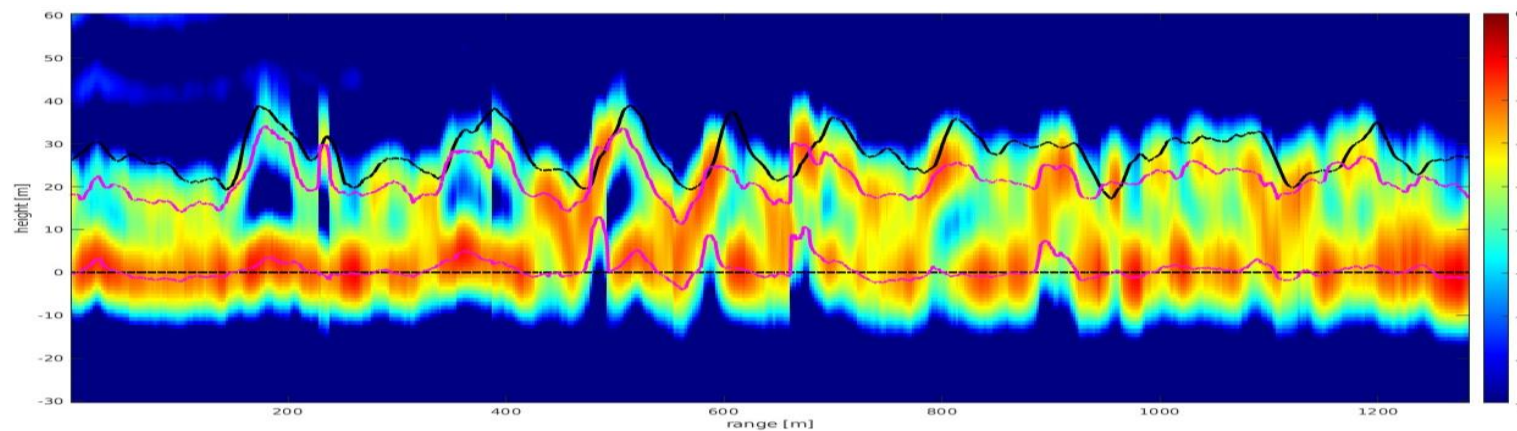
→ Restriction to narrow exponential profiles

Exponential model : Interpretation

No Constraint



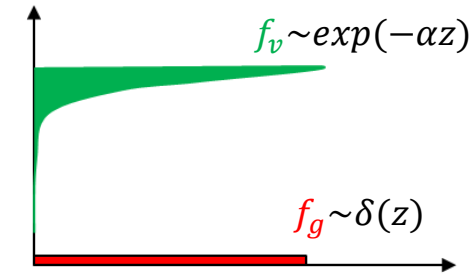
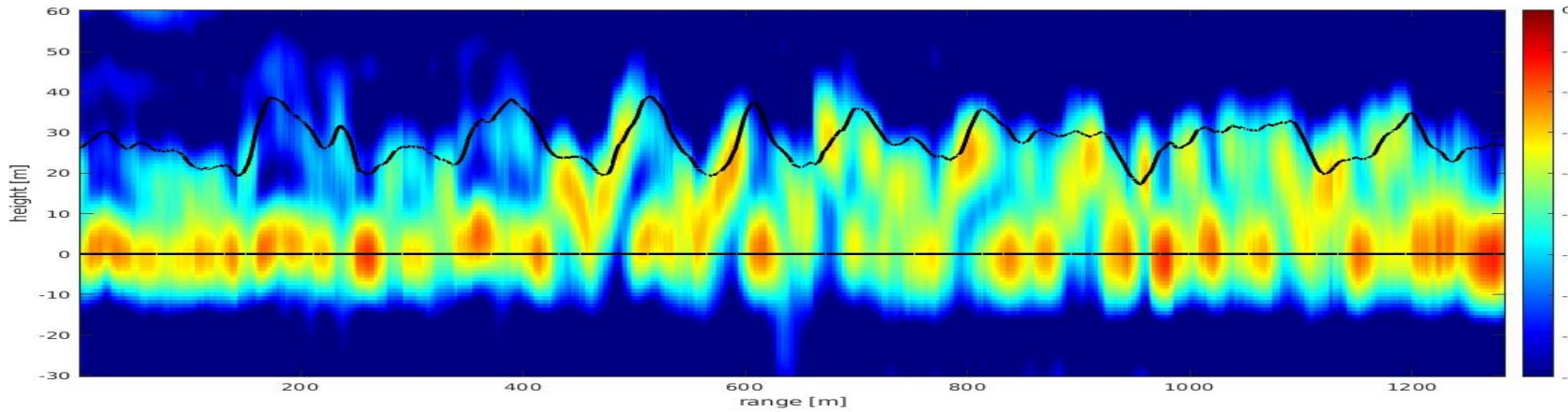
Constraint : large α



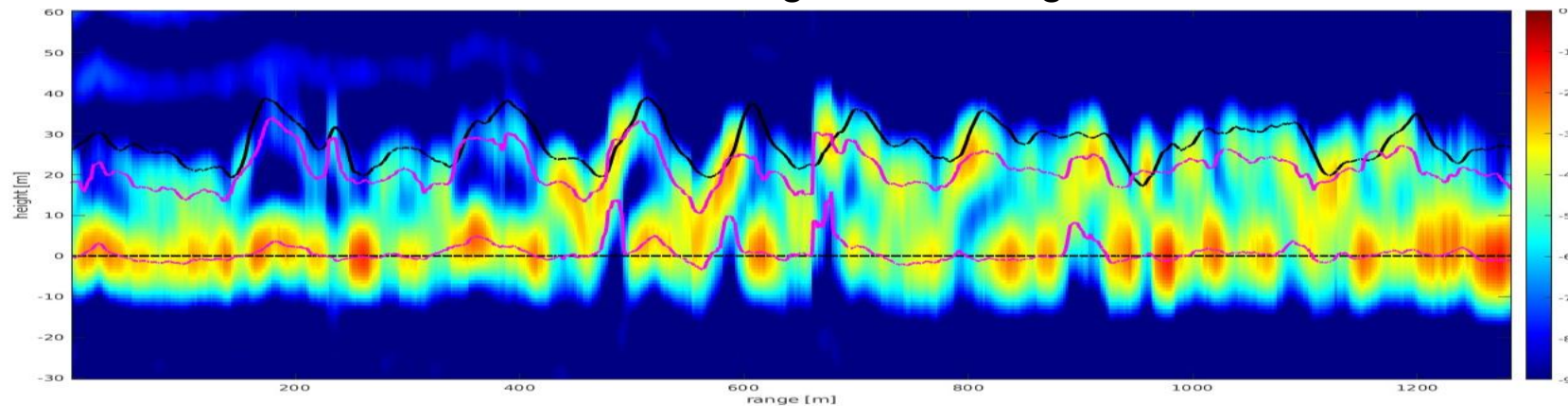
- Small α :
 - arbitrary bias
 - Ambiguous ground component estimate
- Large α : fit to z_{vol}
- Reflectivity fit could be improved

Exponential model with global decorrelation terms

Original Tomogram



Reconstructed tomogram with large α



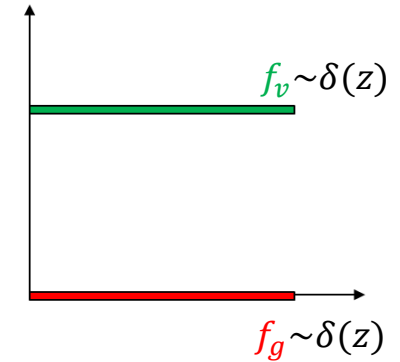
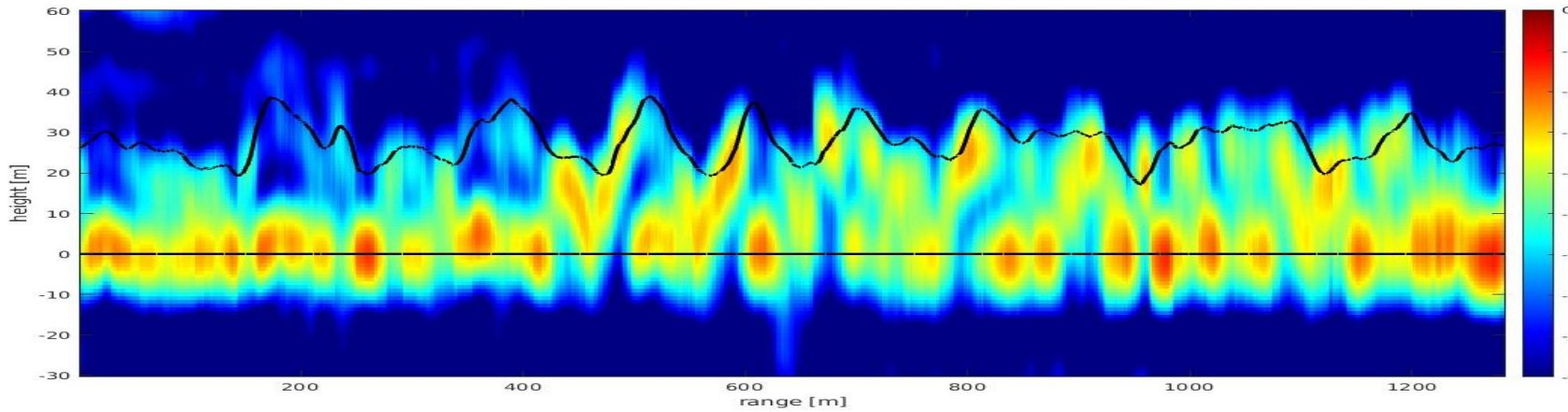
Inclusion of a global decorrelation term :

- SNR
- Spectral shift (range geometry)

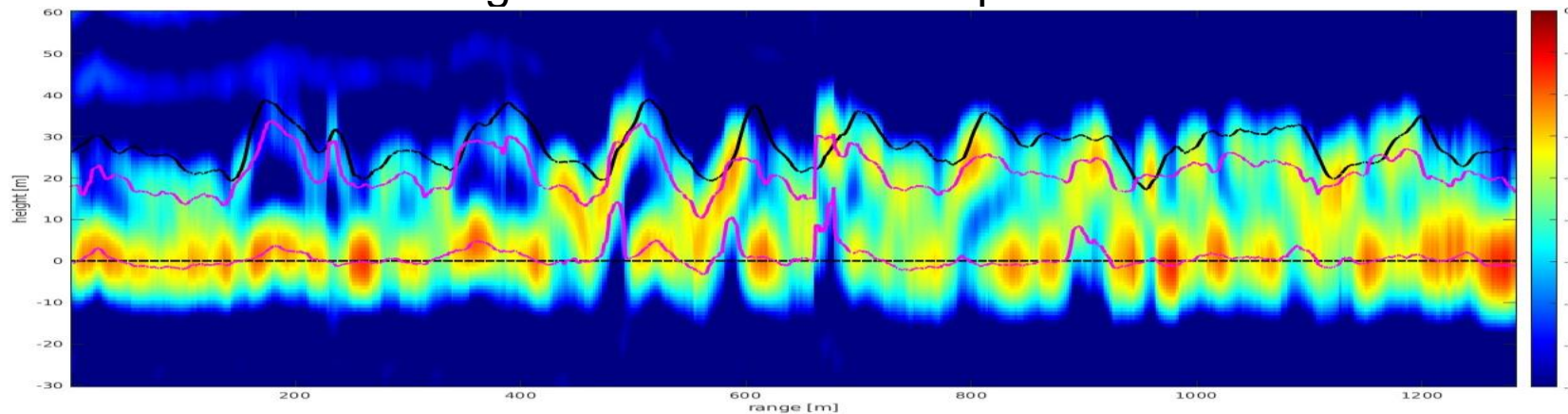
Very good Ground, Volume estimates & reflectivity fit

Model with two Diracs & decorrelation terms

Original Tomogram



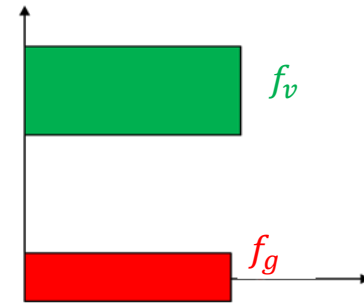
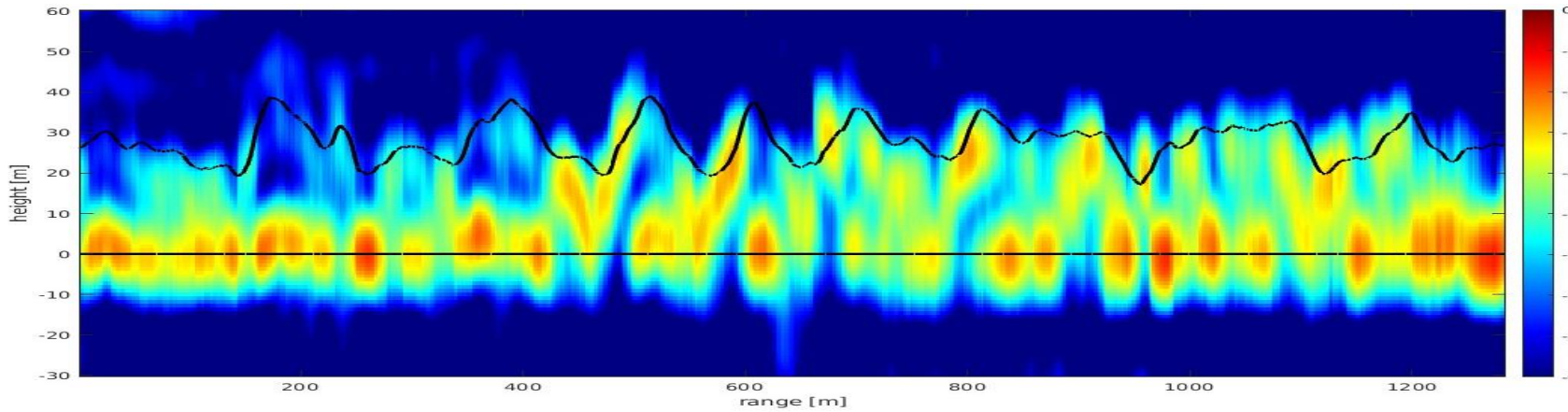
Tomogram with 2 Diracs components



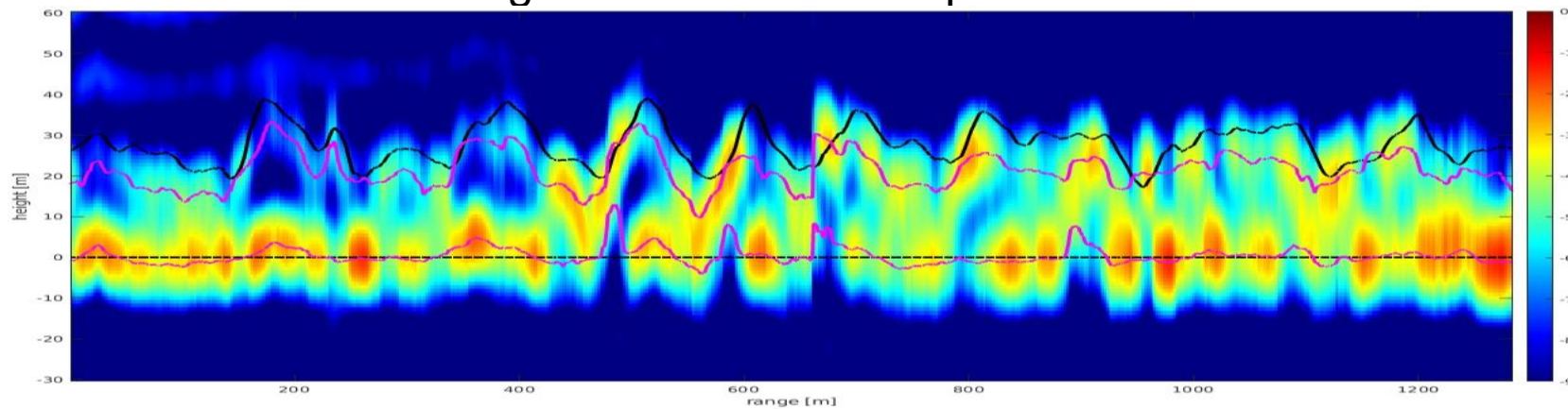
- Very good z_g , z_v fit & excellent reflectivity fit

Box model for the volume, the ground & decorrelation terms

Original Tomogram



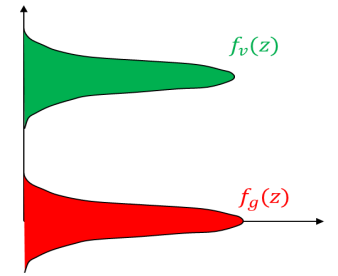
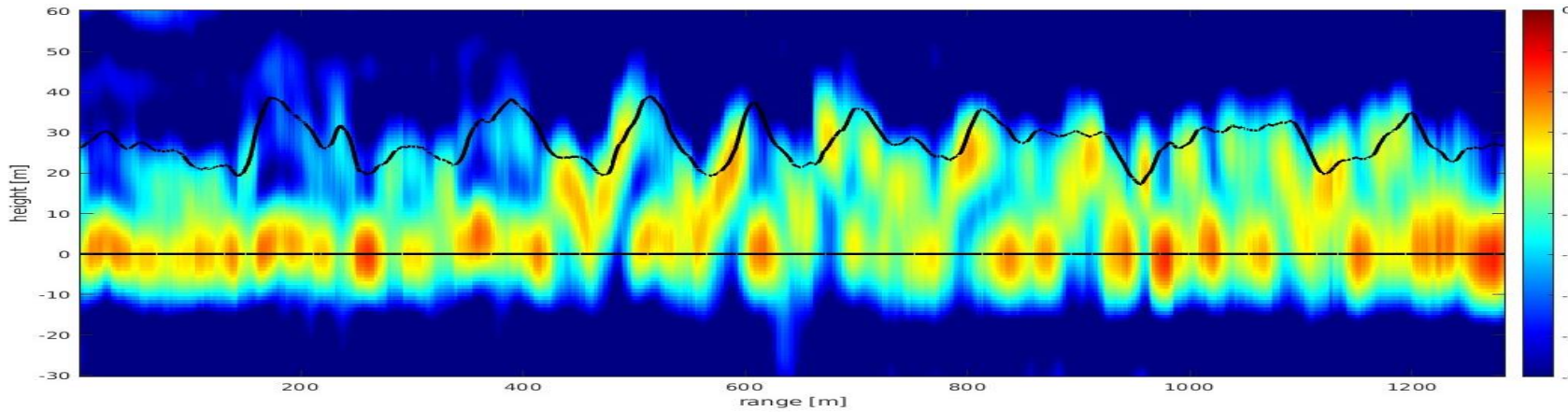
Tomogram with 2 Box components



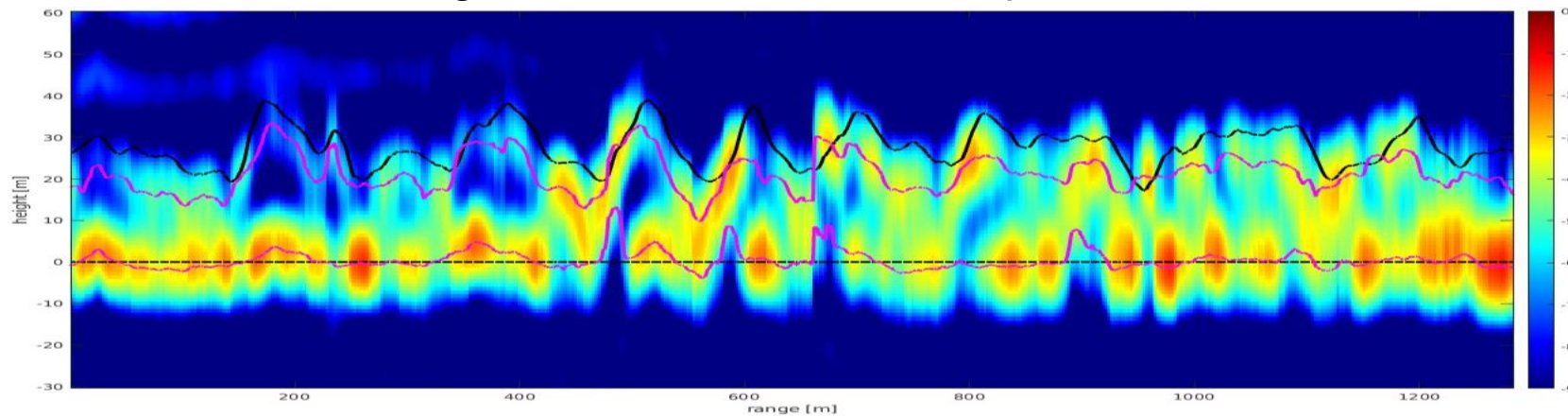
- Very good z_g , z_v fit & excellent reflectivity fit

Gaussian model for the volume, the ground & decorrelation terms

Original Tomogram

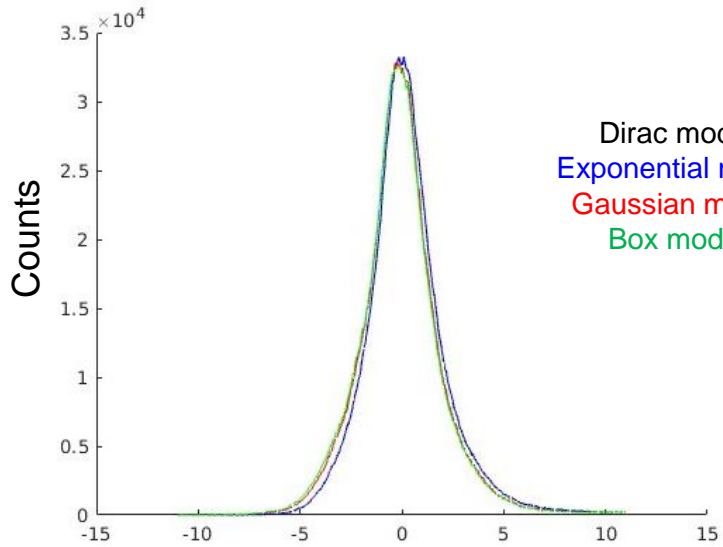


Tomogram with 2 Gaussian components



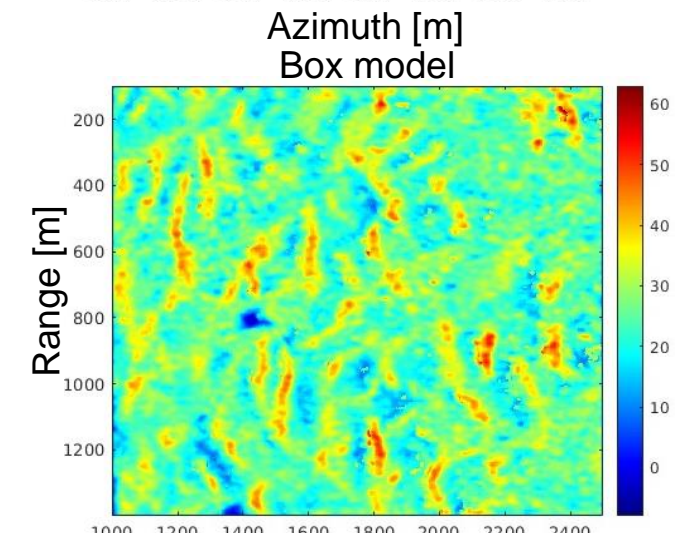
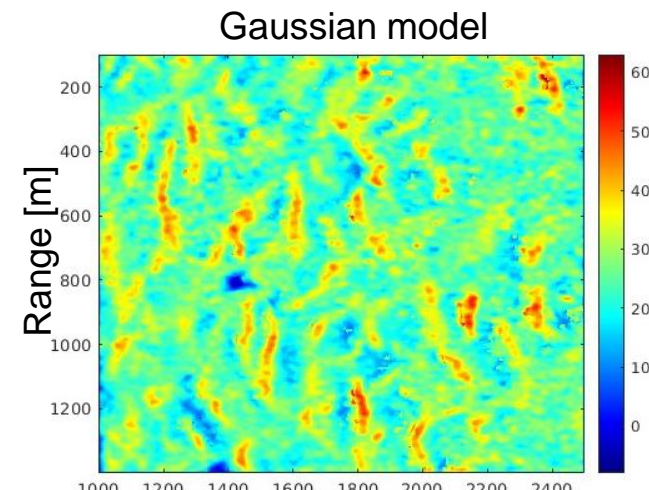
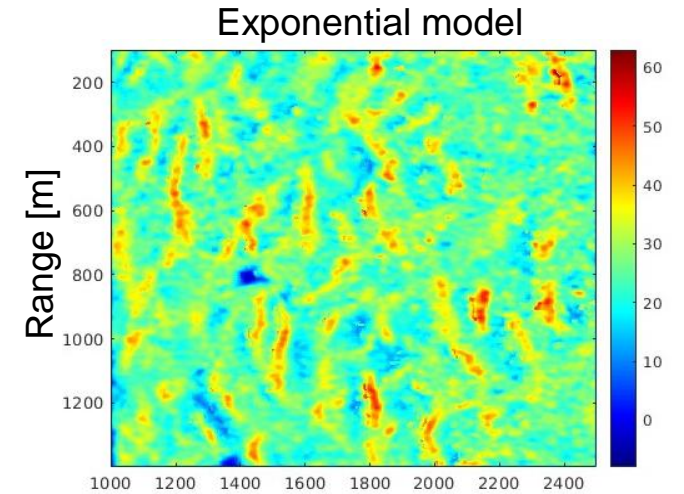
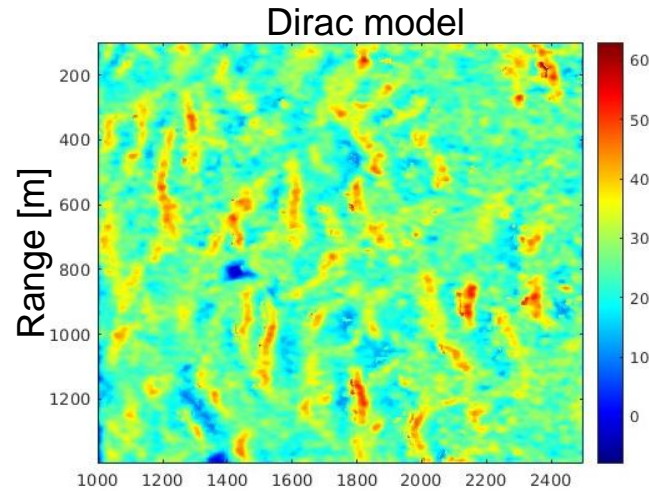
- Very good z_g , z_v fit & excellent reflectivity fit

Comparison of the ground estimates



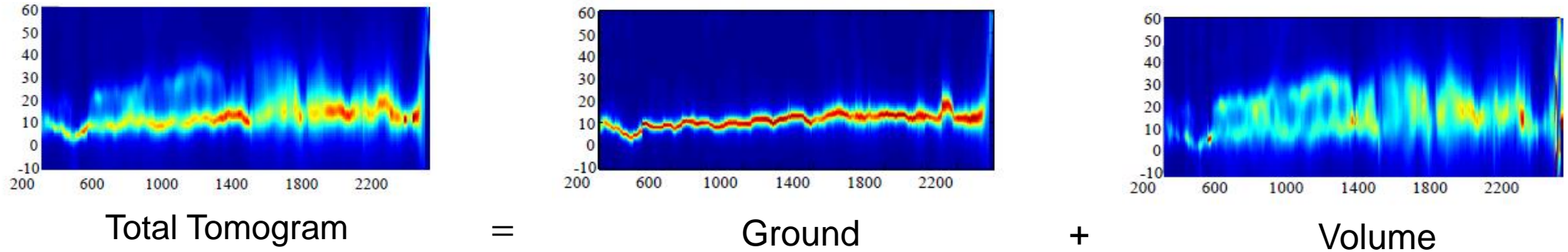
Ground vs LiDAR

Dirac model
Exponential model
Gaussian model
Box model

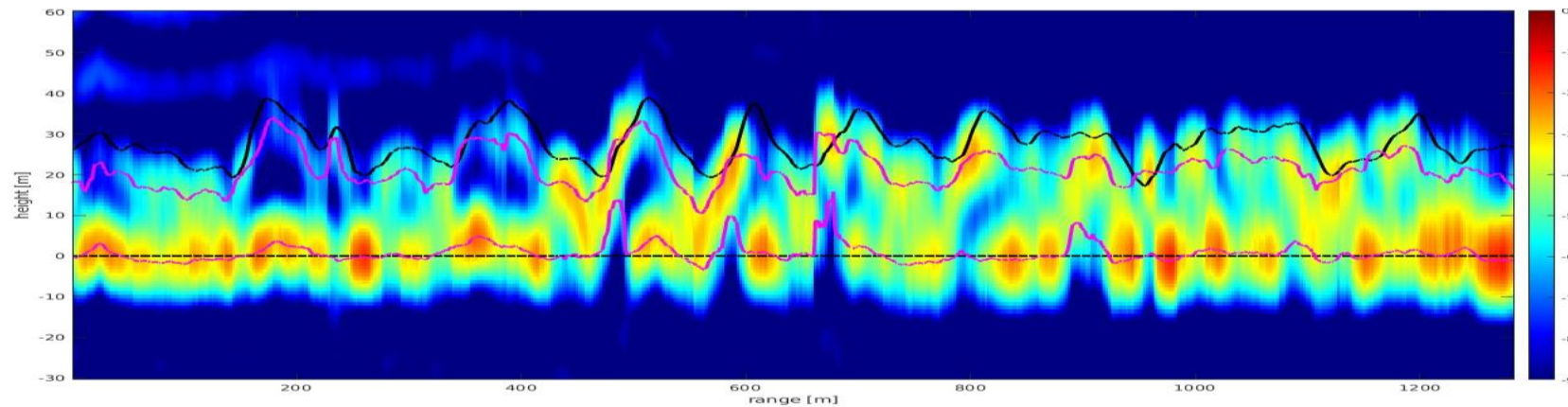


Validation of the parametric tomographic approach

- Polarimetric method : SKP decomposition

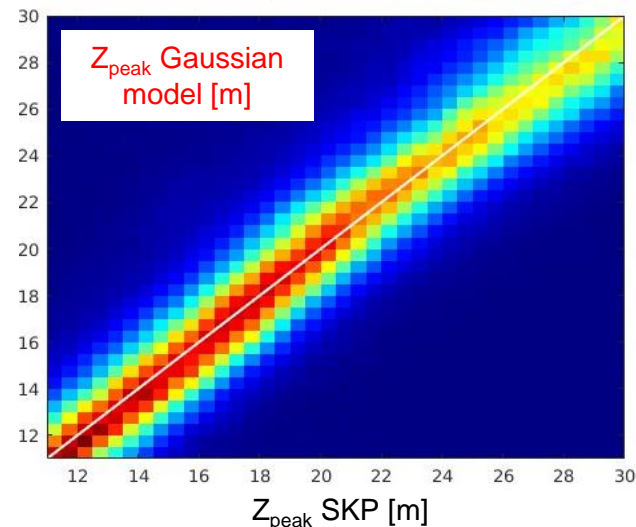
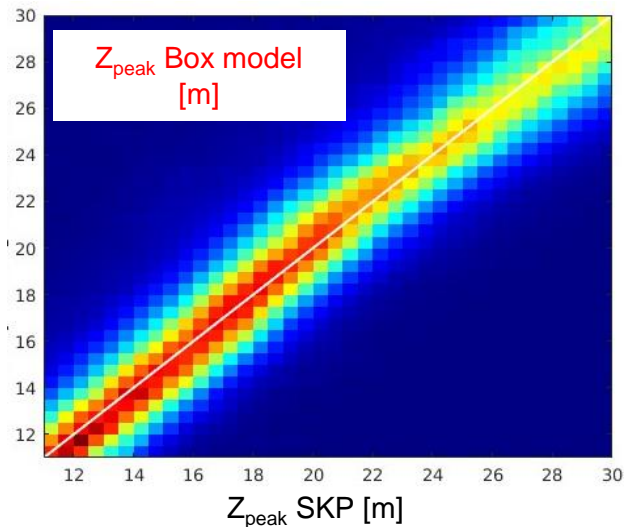
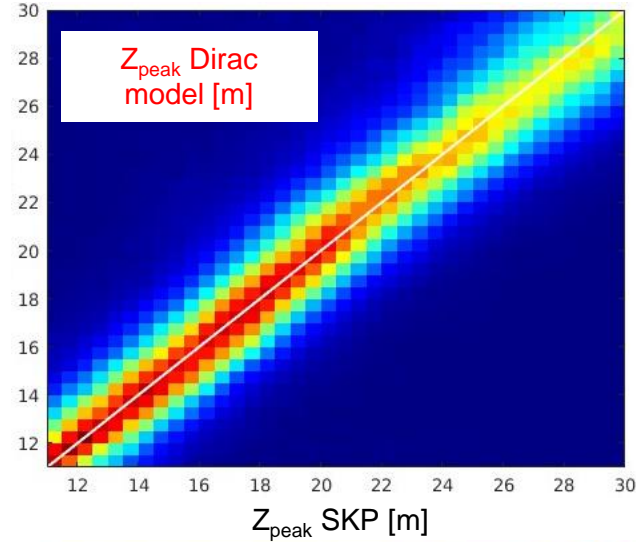
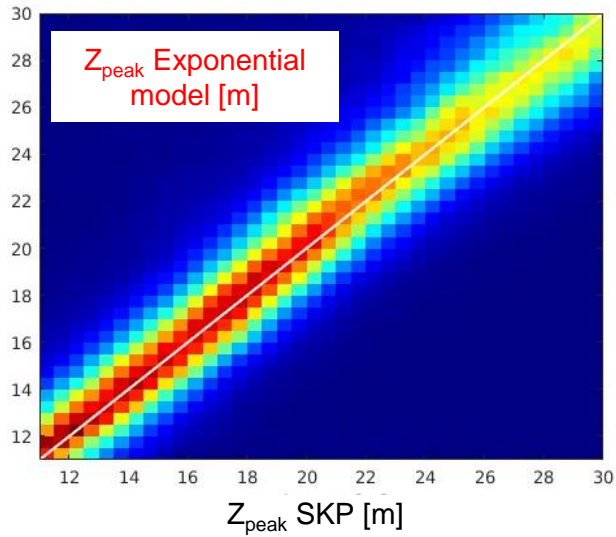


- Low rank model fit



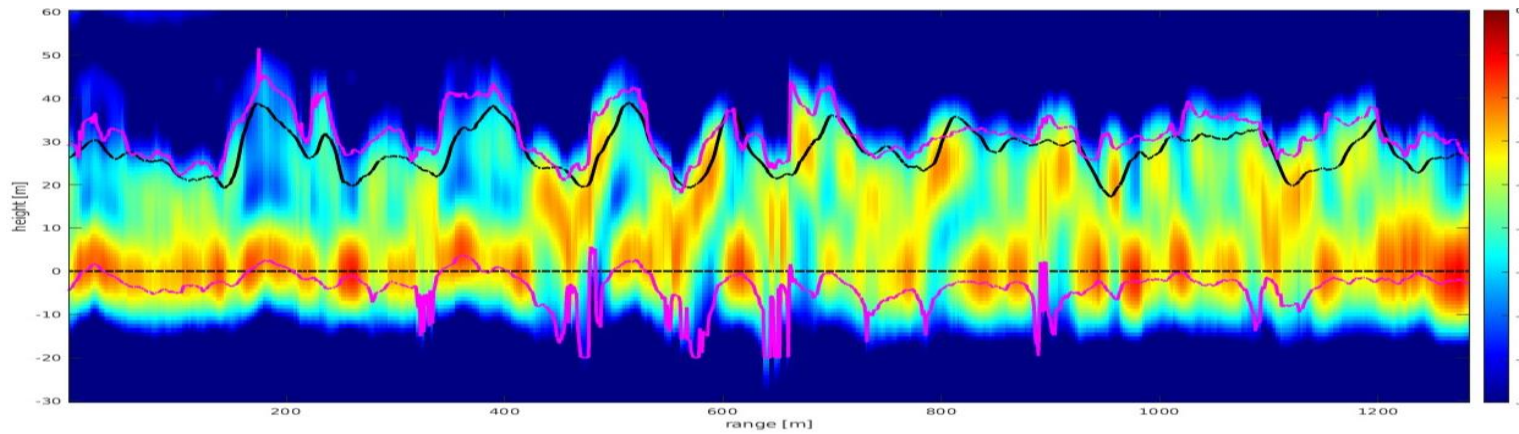
Relationship between z_{vol} fit & z_{vol} SKP

Validation of the parametric tomographic approach

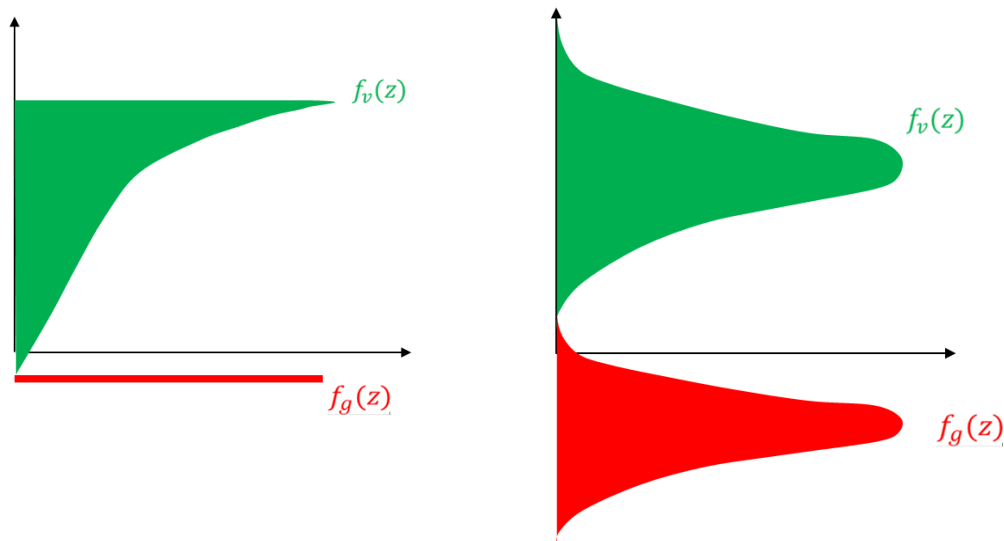


- $Z_v \equiv Z_{\text{peak}_{\text{volsKP}}}$
- Single Pol HH fit is equivalent to full Pol SKP approach

Conclusion :



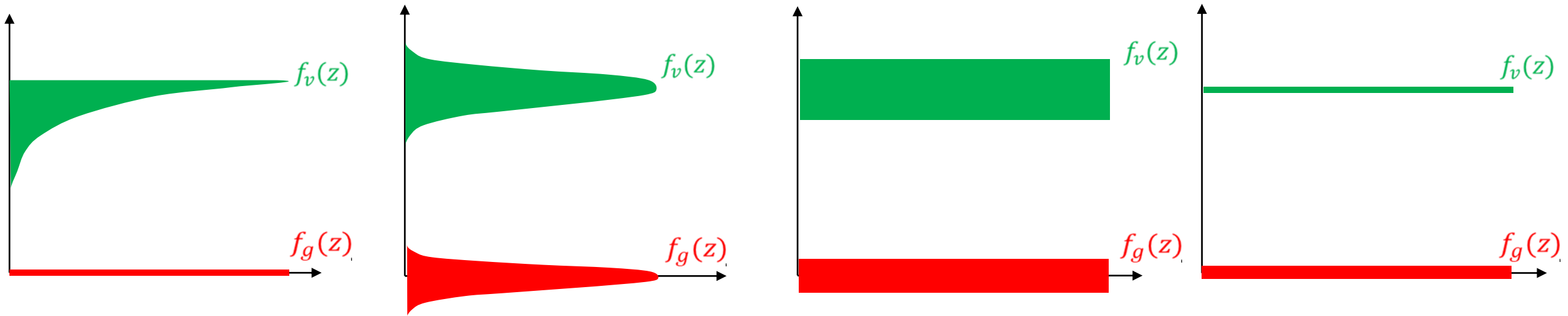
- Overestimated z_{top}
- Underestimated z_{ground}



Wide volume shape

- exaggerated spread to account for decorrelation
- Ambiguous estimation of ground component

Conclusion :



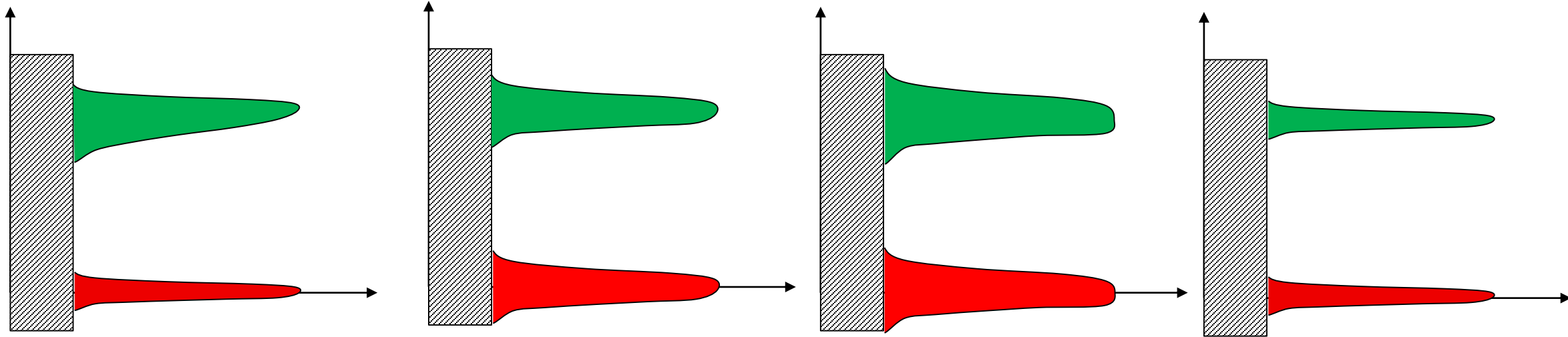
All **low rank** models lead to comparable estimates of ground & volume height

Accurate reflectivity modelling requires decorrelation terms

HH model fit estimates are similar to SKP ones

Conclusion :

Introduction of a decorrelation term to adjust the reconstruction & be similar to the original tomogram



After accounting for decorrelation terms all models converge to similar rounded narrow shape

Next step

- Application to BIOMASS like configuration with less resolution
- Synergy of BIOMASS acquisition modes

