

Differential SAR interferometry for estimating snow water equivalent in central Apennines complex orography from Sentinel-1 satellite within SMIVIA project

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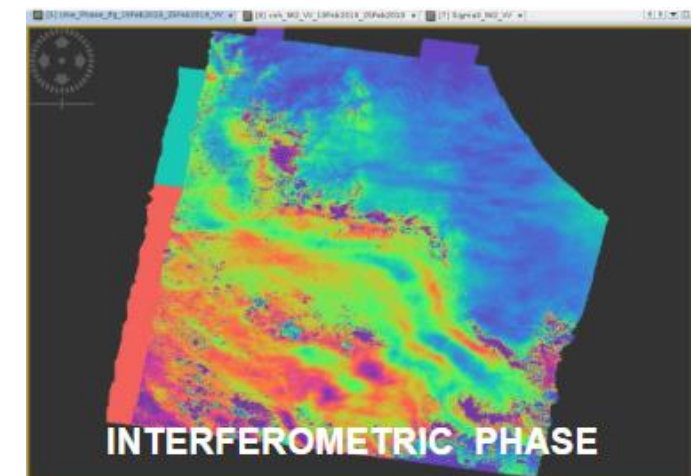
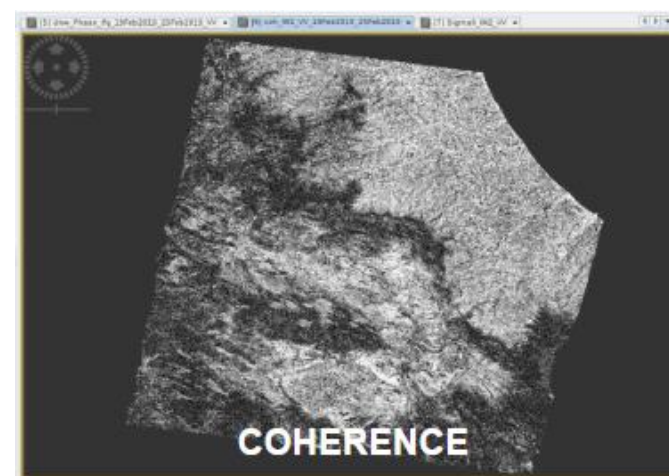
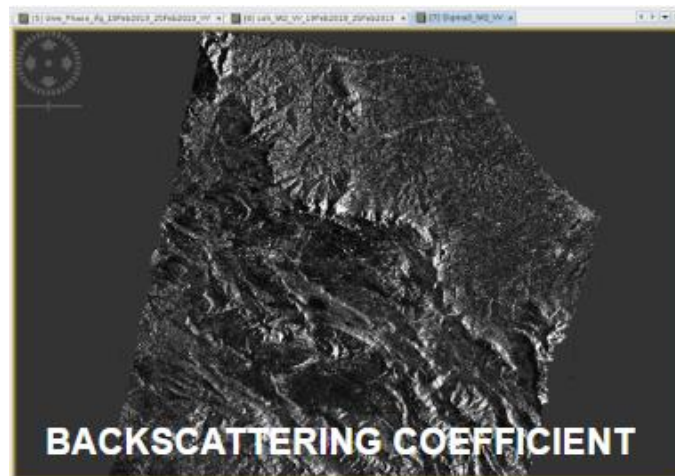
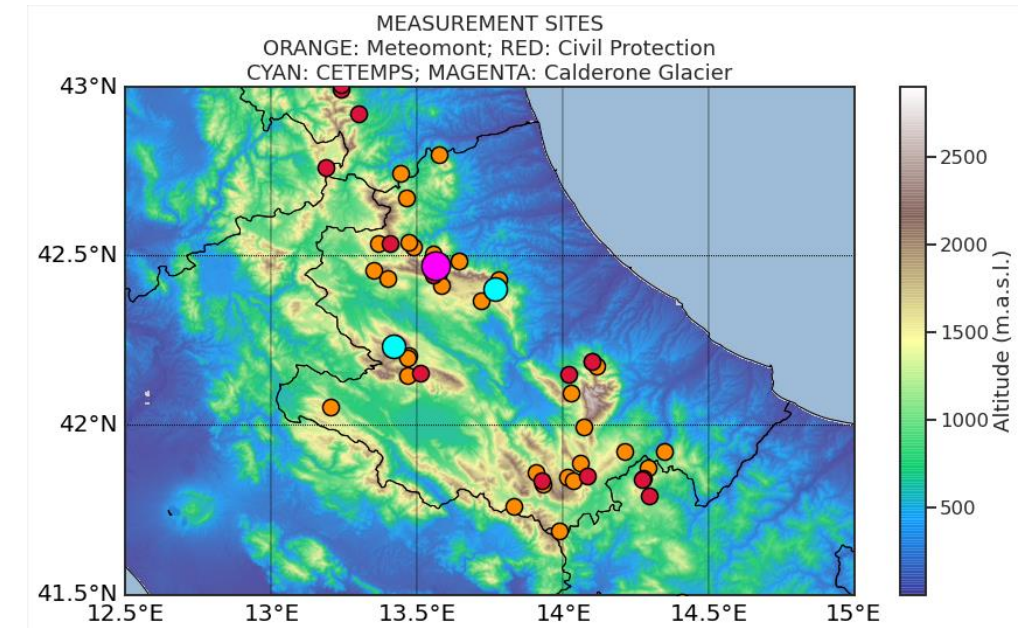


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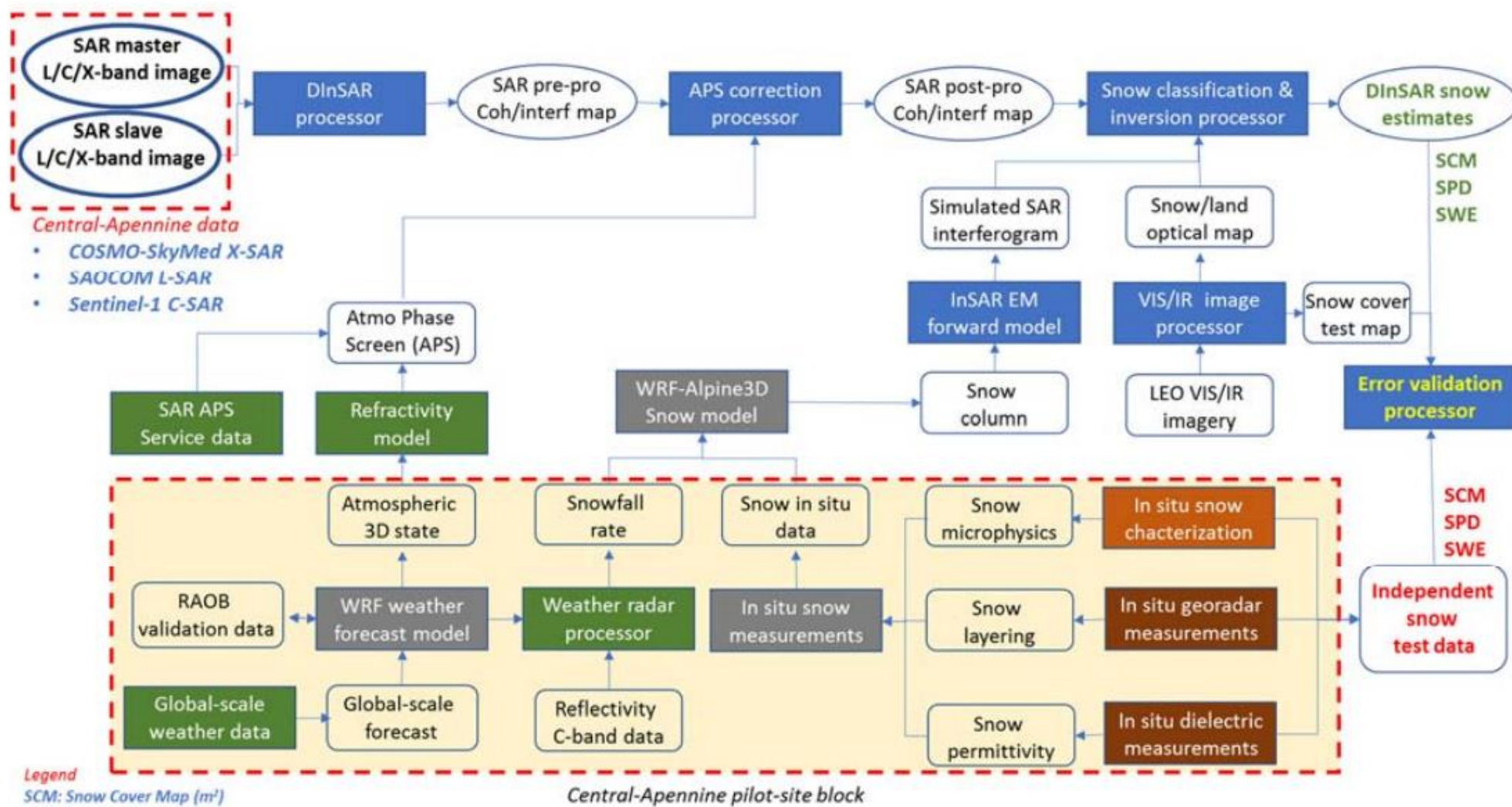


Objectives of the study:

- Retrieve snow properties in the Italian Central Apennine with Differential Interferometric spaceborn SAR data, using some auxiliary data.
- Estimate snow cover extension and classify snow (dry snow, wet snow).
- Estimate snow depth.
- Validate estimates with in-situ data obtained with measurement campaigns.
- Provide useful information in terms of avalanche warning, monitoring of climate change evolution, flood forecasting and water volumes expected for the hydric supply.



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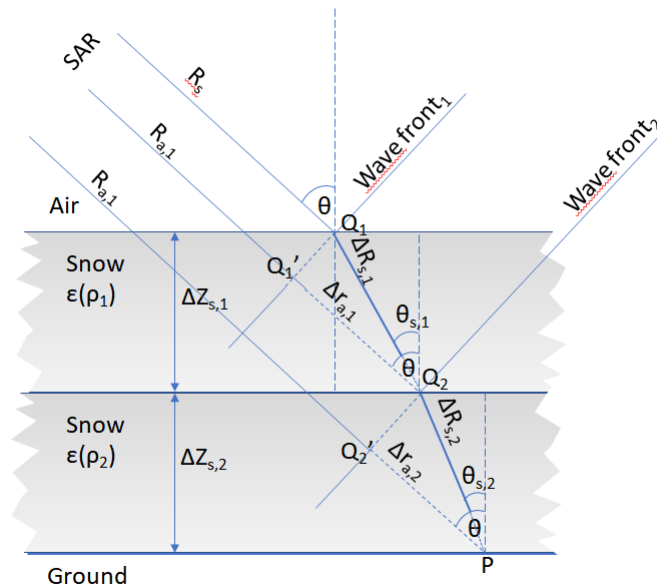


THE SMIVIA
INTEGRATED
APPROACH

DInSAR snow retrieval method

The signal delay caused by snow refraction can be measured **if snow permittivity is known**.

Therefore snow depth variation between two dates can be estimated by calculating the SAR signal phase difference between two satellite passes.



Snow Permittivity

- Snow permittivity depends on snow water fraction and snow density values.
- Different models are available to estimate snow permittivity.

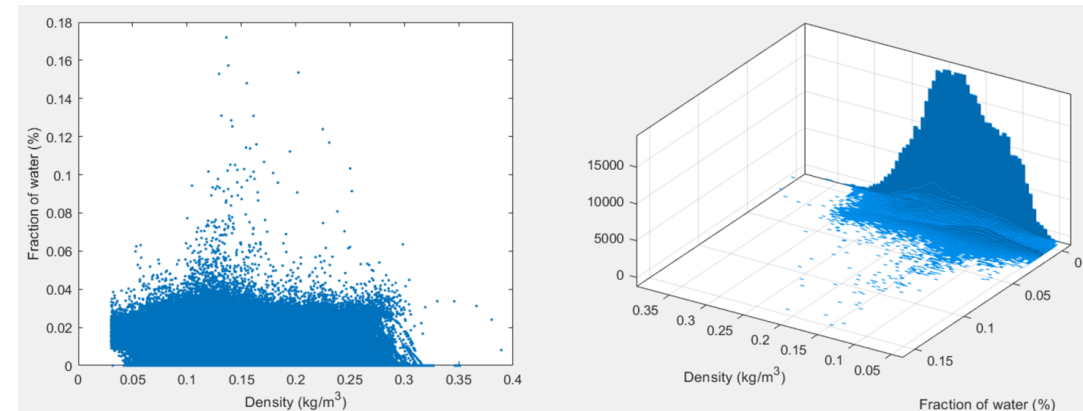
- A very simple model can be used as a rough approximation:

$$\epsilon_{\text{snow}} = \epsilon_{\text{dry-snow}} (1 - f_w) + \epsilon_{\text{water}} f_w$$

- More complex models are available in the literature

For water fraction and snow density values two options have been explored:

- a synthetic scenario with uniform distributions of values;
- a realistic scenario (shown below) made of estimates calculated by a dynamic snow-mantle evolution model (Alpine3D).

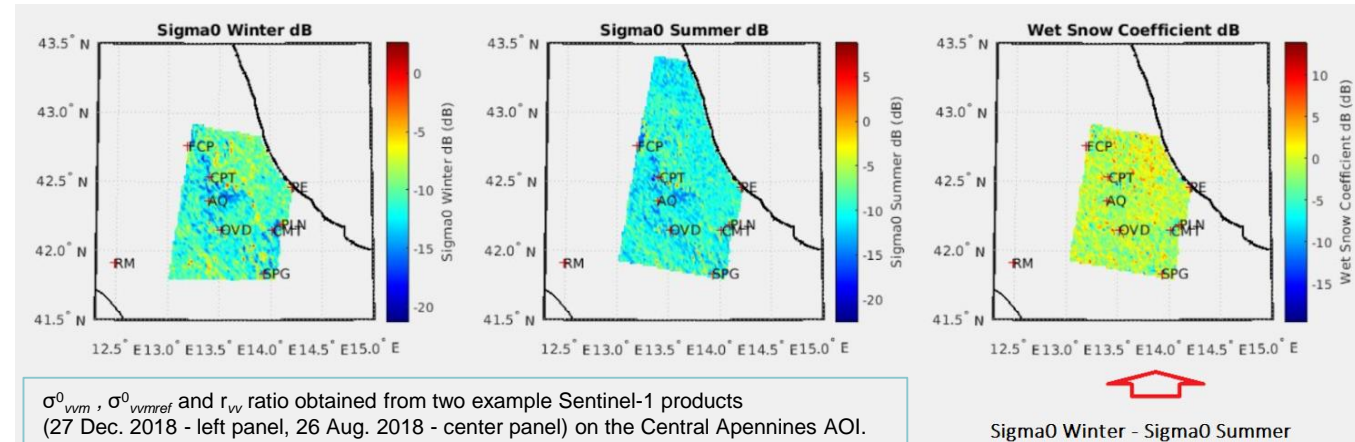
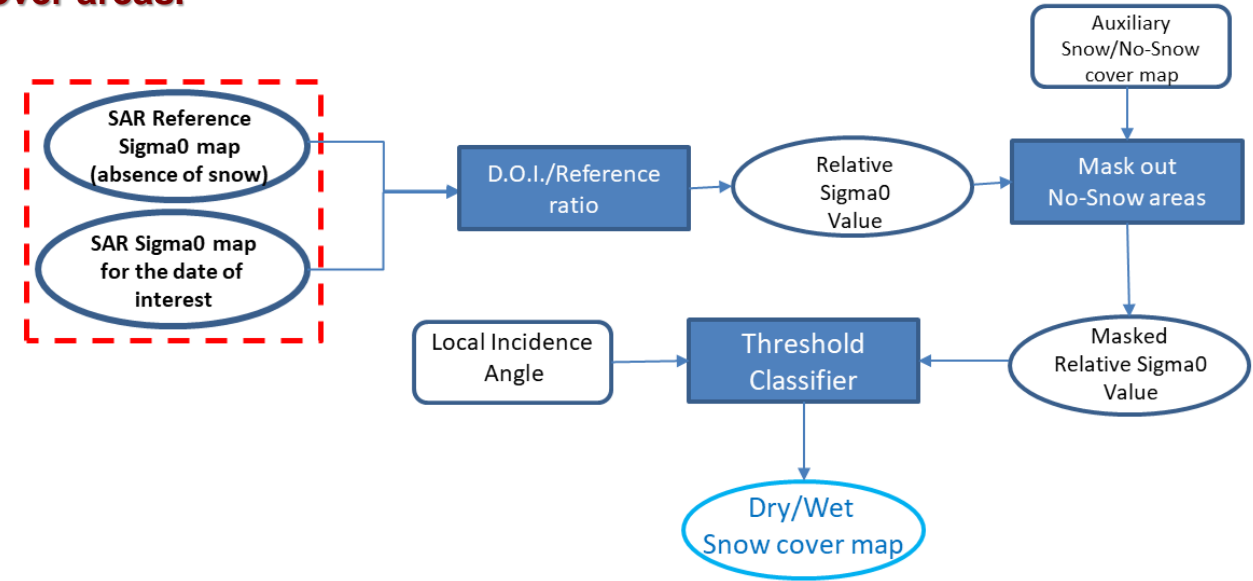
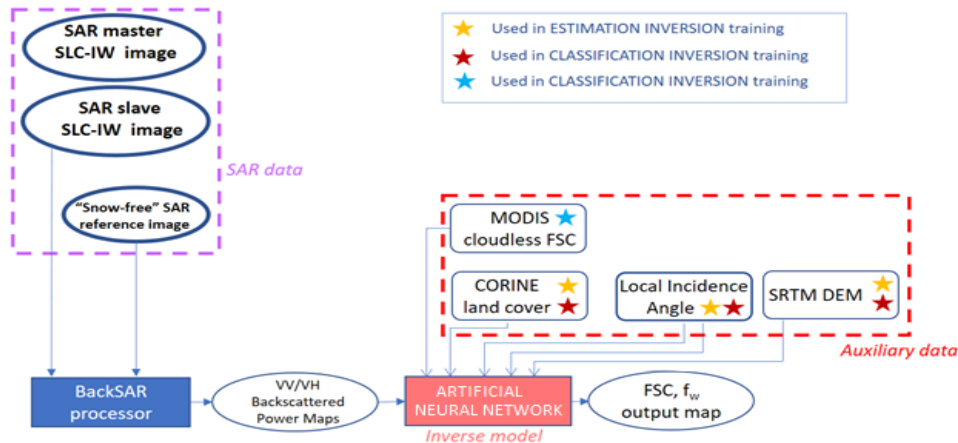


SNOW CLASSIFICATION: identification of dry snow and wet snow cover areas.

- A first approach has been adopted for the snow classification problem which employed the backscattering-coefficient threshold-based algorithm proposed by Nagler-Rott for the dry-wet snow classification task.

$$r_{vv}(x, y) = \frac{\sigma_{vv}^0(x, y)}{\sigma_{vv}^0(x, y)_{ref}} = \begin{cases} \leq r_{0wet} & \text{Wet snow} \\ > r_{0wet} & \text{Dry snow} \end{cases}$$

- This approach showed poor results on the considered AOI (Italian Central Apennines)
- Another approach based on an Artificial Neural Network (ANN) is currently being experimented.



SNOW ESTIMATION: retrieval of snow depth and snow density.

- The **PAI algorithm**, which is obtained by inverting the DInSAR equation with $\Delta h_s = \Delta Z_s$:

$$\Delta \hat{h}_s(x, y) = \frac{\Delta \Phi_{ppsm}(x, y)}{2 k_i (\sqrt{\epsilon_{Rrs} - \sin^2 \theta_l} - \cos \theta_l)}$$

The main limitation of the PAI inversion algorithm is that it requires ρ and f_w values to be known in order to perform the inversion; since ρ and f_w are, in general, not known, they need to be assigned estimated values, potentially resulting in a poor accuracy of the model.

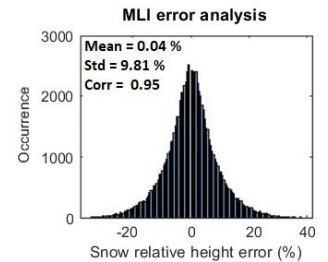
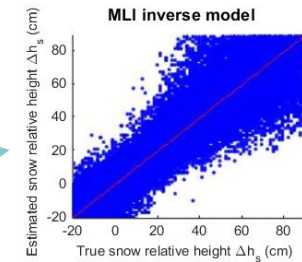
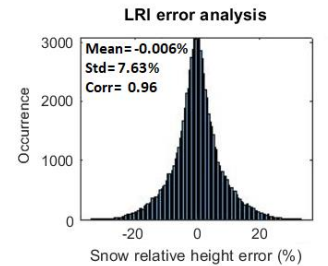
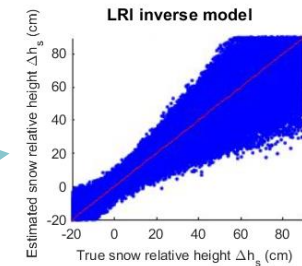
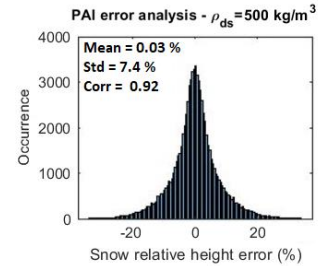
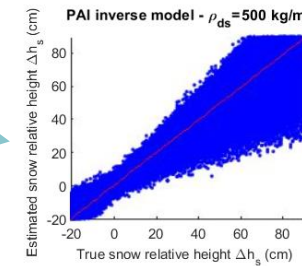
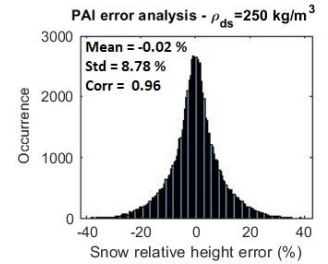
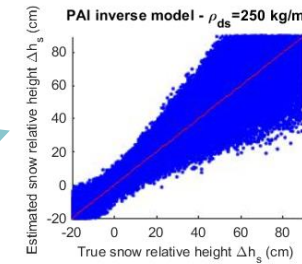
- The **LRI algorithm** which is based on the statistical linear-regression approach and is described by the following equation:

$$\Delta \hat{h}_s(x, y) = a_{h0} + a_{h1} \Delta \Phi_{ppsm}(x, y)$$

•

- The **MLI algorithm** which uses a statistical approach where the error probability density function is maximized, under a Gaussian hypothesis its negative argument is minimized.

$$\begin{bmatrix} \Delta \hat{h}_s(x, y) \\ \hat{\rho}_s(x, y) \end{bmatrix} = \underset{\rho_s}{\operatorname{argmin}} \left\{ [\Delta \Phi_{ppsm}(x, y) - \Delta \Phi_{pps}(x, y)]^2 \right\}$$



CONCLUSIONS

Current status of the study:

- We have implemented some dielectric models for snow permittivity estimation.
- We have used a realistic scenario based on a snow-mantle evolution model (Alpine-3D)
- We have implemented and tested a threshold-based algorithm which uses the backscattering-coefficient as input. This algorithm showed poor results on the Italian Central Apennine AOI.
- We have implemented and tested three different inverse models for snow depth estimation. They use differential SAR phase as input. MLI showed better performances.

Future developments:

- Multiband/multimission integrated analysis: combine C-Band (Sentinel-1), L-band (SAOCOM), and X-band (Cosmo SkyMed 1st and 2nd generation) data to take advantage of the different characteristics of each band.
- Implementing Artificial Neural Network based models to improve accuracy.
- Validation of the results with in-situ data.



Prof. Frank Silvio Marzano
1963 – 2022

