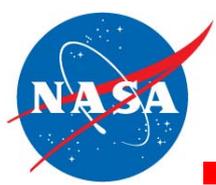


Soil moisture and biomass retrieval over forest areas using low frequency polarimetric SAR data

My-Linh Truong-Loi

Jet Propulsion Laboratory / California Institute of Technology

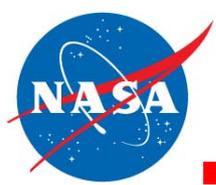
Postdoc seminar – December 11th 2012



Outline

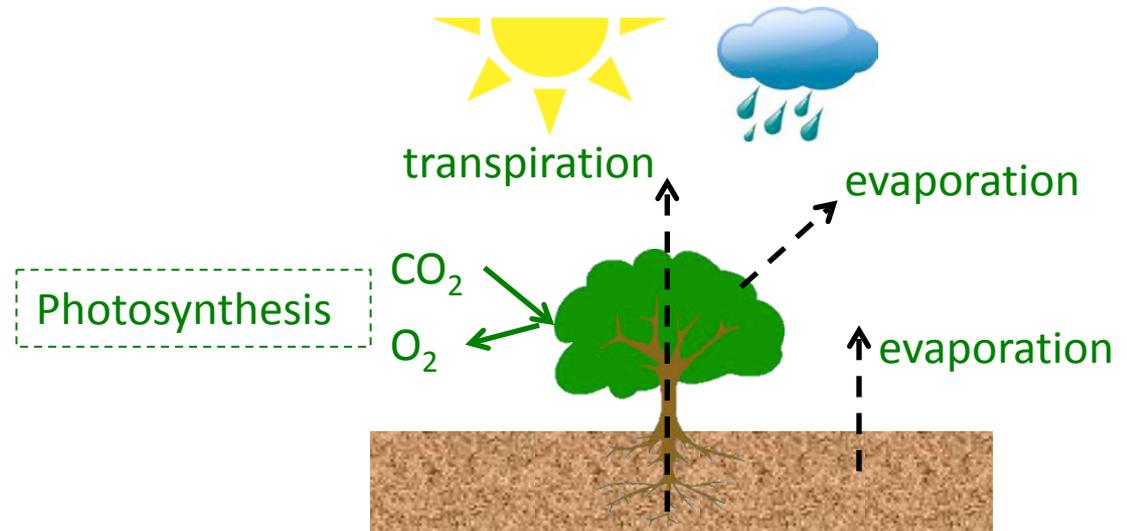
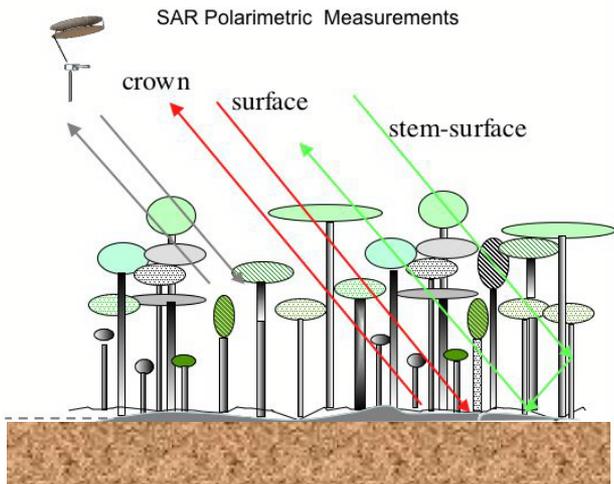
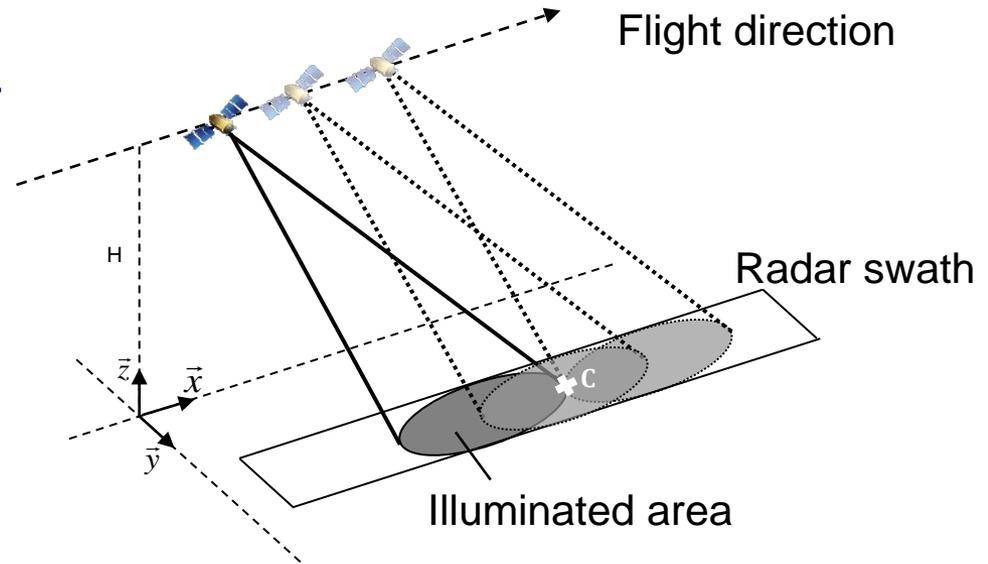
Objective: is it possible to estimate soil moisture under forest?

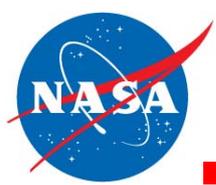
- Introduction
- Presentation of the model
- Inversion process
- Sensitivity analysis of the inversion process with simulated data
- Application with real data
- Explanation of field measurements



Context

- Synthetic aperture radar
- Low frequency
- Soil moisture





Context

Temperate evergreen forest



Open shrubland



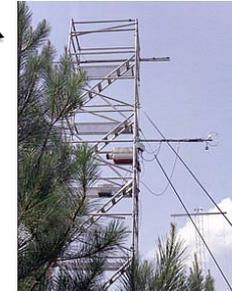
Boreal forest



Boreal transitional



Temperate mixed forest



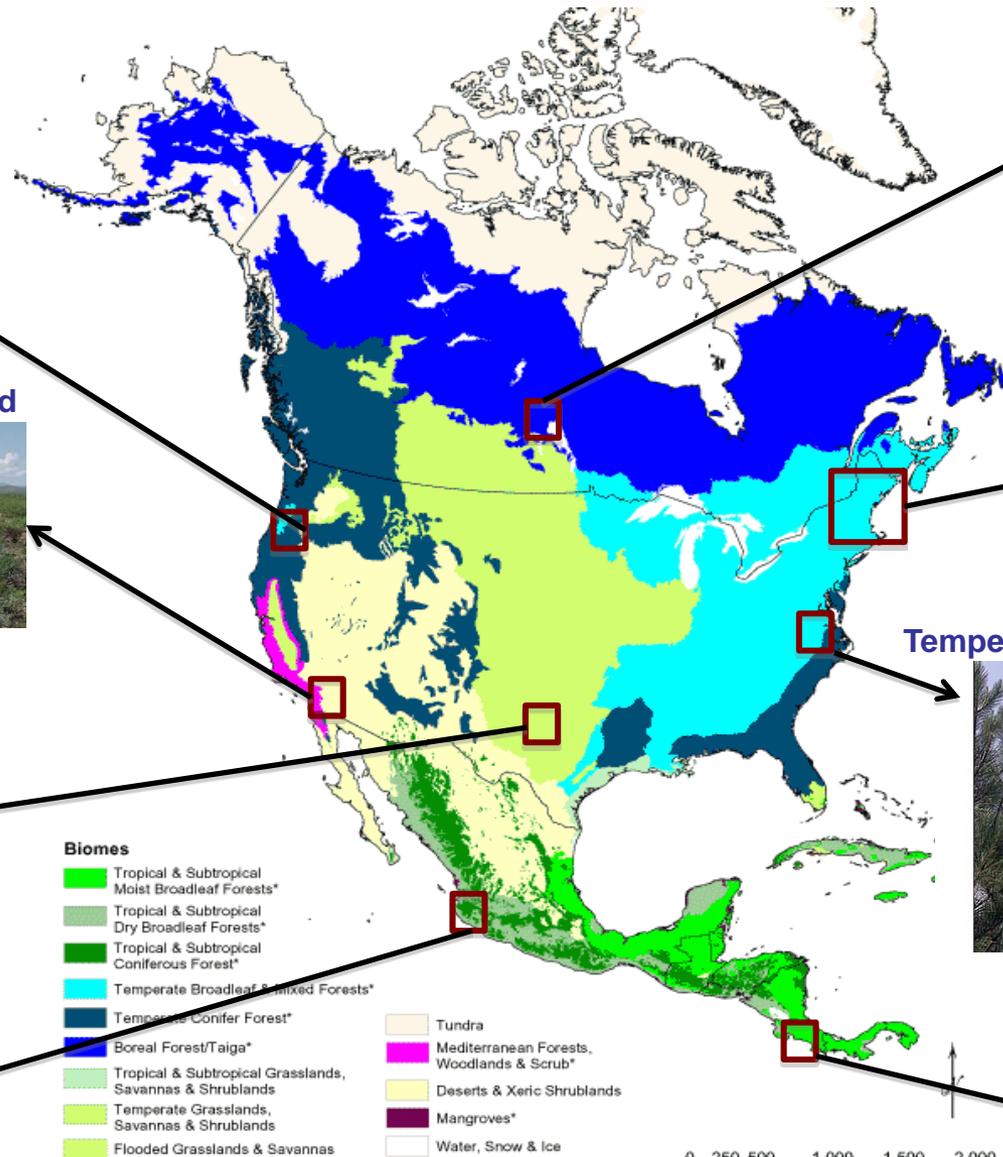
Tropical moist forest



Temperate grasslands



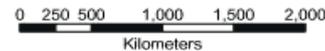
Subtropical dry forest

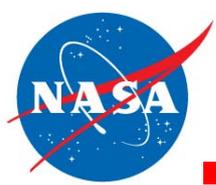


Biomes

- Tropical & Subtropical Moist Broadleaf Forests*
- Tropical & Subtropical Dry Broadleaf Forests*
- Tropical & Subtropical Coniferous Forest*
- Temperate Broadleaf & Mixed Forests*
- Temperate Conifer Forest*
- Boreal Forest/Taiga*
- Tropical & Subtropical Grasslands, Savannas & Shrublands
- Temperate Grasslands, Savannas & Shrublands
- Flooded Grasslands & Savannas
- Montane Grasslands & Shrublands
- Tundra
- Mediterranean Forests, Woodlands & Scrub*
- Deserts & Xeric Shrublands
- Mangroves*
- Water, Snow & Ice

* Included in fragmentation analyses





Forward model expression

→ Simplification of the distorted Born approximation

-complex

-requires detailed information about vegetation structure

$$\sigma_{HH}^0 = A_{HH} \cos \theta_0 W^{\alpha_{HH}} (1 - \exp(-B_{HH} W^{\beta_{HH}} \sec \theta_0)) + C_{HH} \Gamma_{HH} W^{\delta_{HH}} \sin(\theta_0) \exp(-B_{HH} W^{\beta_{HH}} \sec \theta_0) + S_{HH} \exp(-B_{HH} W^{\beta_{HH}} \sec \theta_0)$$

$$\sigma_{VV}^0 = A_{VV} \cos \theta_0 W^{\alpha_{VV}} (1 - \exp(-B_{VV} W^{\beta_{VV}} \sec \theta_0)) + C_{VV} \Gamma_{VV} W^{\delta_{VV}} \sin(\theta_0) \exp(-B_{VV} W^{\beta_{VV}} \sec \theta_0) + S_{VV} \exp(-B_{VV} W^{\beta_{VV}} \sec \theta_0)$$

$$\sigma_{HV}^0 = A_{HV} \cos \theta_0 W^{\alpha_{HV}} (1 - \exp(-B_{HV} W^{\beta_{HV}} \sec \theta_0)) + C_{HV} \Gamma_{HV} W^{\delta_{HV}} \sin(\theta_0) \exp(-B_{HV} W^{\beta_{HV}} \sec \theta_0) + S_{HV} \exp(-B_{HV} W^{\beta_{HV}} \sec \theta_0)$$

W is the biomass (Mg/ha)

$$\Gamma_{pq} = R_p R_q^* \exp(-4k^2 s^2 \cos^2(\theta_0))$$

s is the rms height

K is the wavenumber

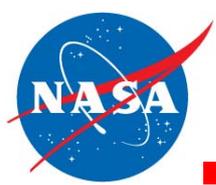
R_p and R_q being the Fresnel reflection coefficients of the ground for polarization p and q

S_{HH}, S_{VV} and S_{HV} being the scattering term from a rough surface model

θ₀ is the local incidence angle

α_{pq}, β_{pq}, δ_{pq} are structural parameters

A_{pq}, B_{pq} and C_{pq} are calibration factors



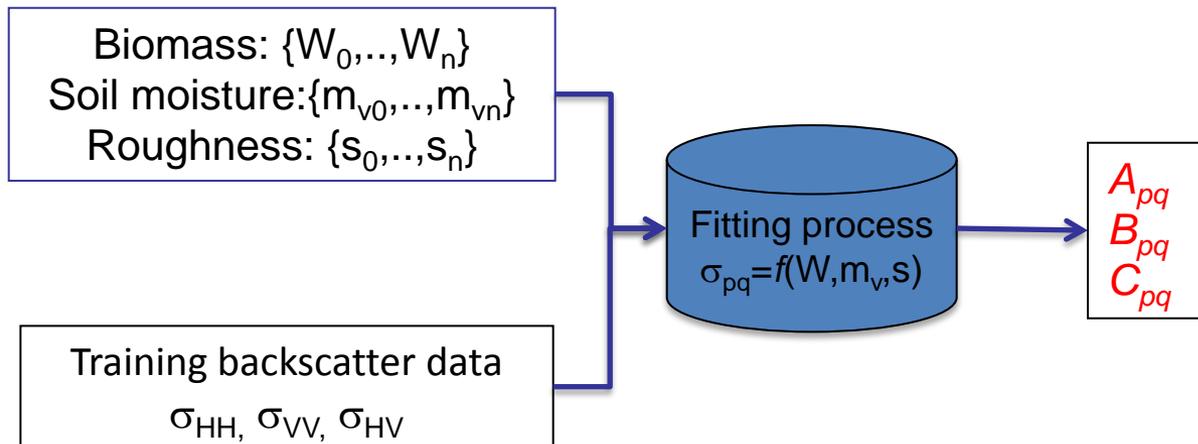
Training the model to site data

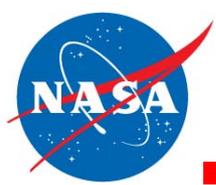
$$\sigma_{HH}^0 = A_{HH} \cos \theta_0 W^{\alpha_{HH}} (1 - \exp(-B_{HH} W^{\beta_{HH}} \sec \theta_0)) + C_{HH} \Gamma_{HH} W^{\delta_{HH}} \sin(\theta_0) \exp(-B_{HH} W^{\beta_{HH}} \sec \theta_0) + S_{HH} \exp(-B_{HH} W^{\beta_{HH}} \sec \theta_0)$$

$$\sigma_{VV}^0 = A_{VV} \cos \theta_0 W^{\alpha_{VV}} (1 - \exp(-B_{VV} W^{\beta_{VV}} \sec \theta_0)) + C_{VV} \Gamma_{VV} W^{\delta_{VV}} \sin(\theta_0) \exp(-B_{VV} W^{\beta_{VV}} \sec \theta_0) + S_{VV} \exp(-B_{VV} W^{\beta_{VV}} \sec \theta_0)$$

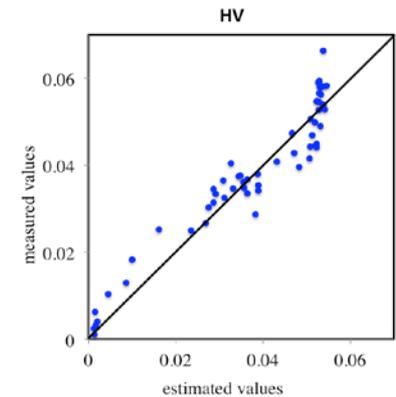
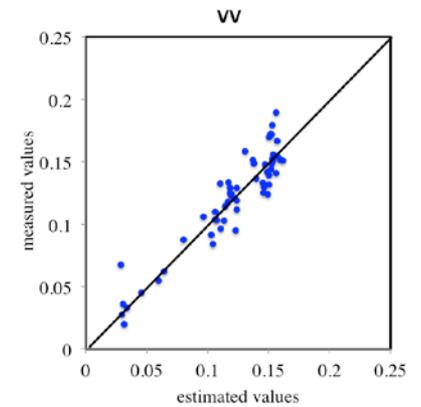
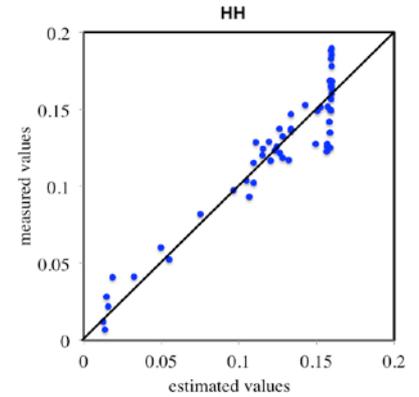
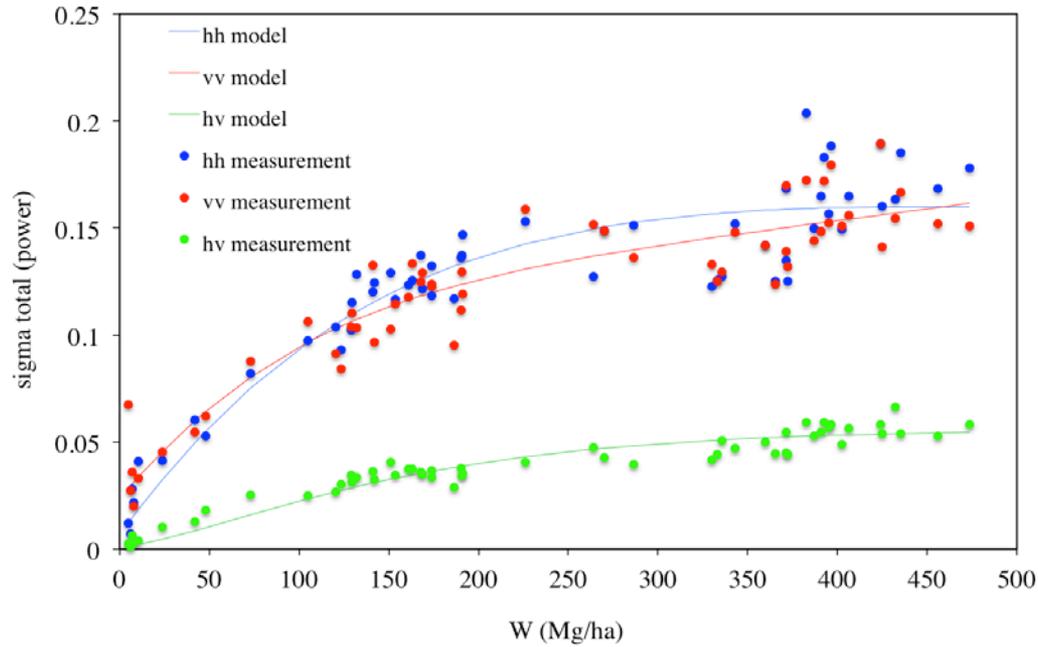
$$\sigma_{HV}^0 = A_{HV} \cos \theta_0 W^{\alpha_{HV}} (1 - \exp(-B_{HV} W^{\beta_{HV}} \sec \theta_0)) + C_{HV} \Gamma_{HV} W^{\delta_{HV}} \sin(\theta_0) \exp(-B_{HV} W^{\beta_{HV}} \sec \theta_0) + S_{HV} \exp(-B_{HV} W^{\beta_{HV}} \sec \theta_0)$$

- Calibrated SAR data
- Use average soil moisture and roughness for the site
- Use plot level biomass values
- Create a series of points to estimate coefficients A_{pq} , B_{pq} , C_{pq}

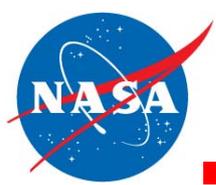




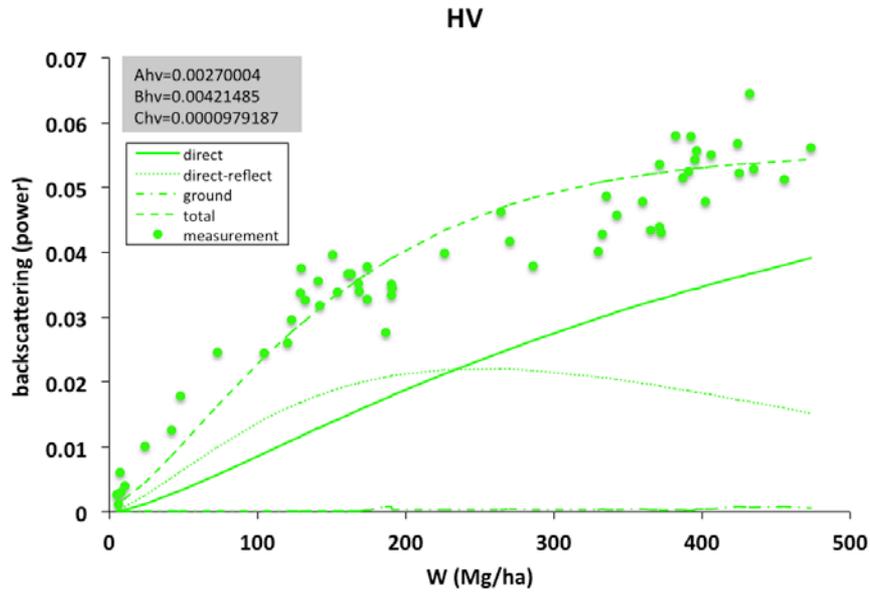
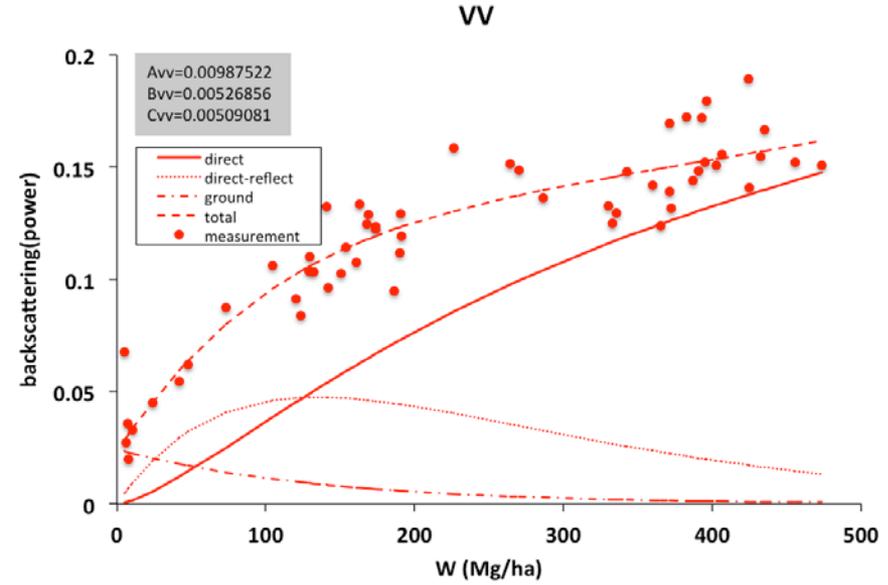
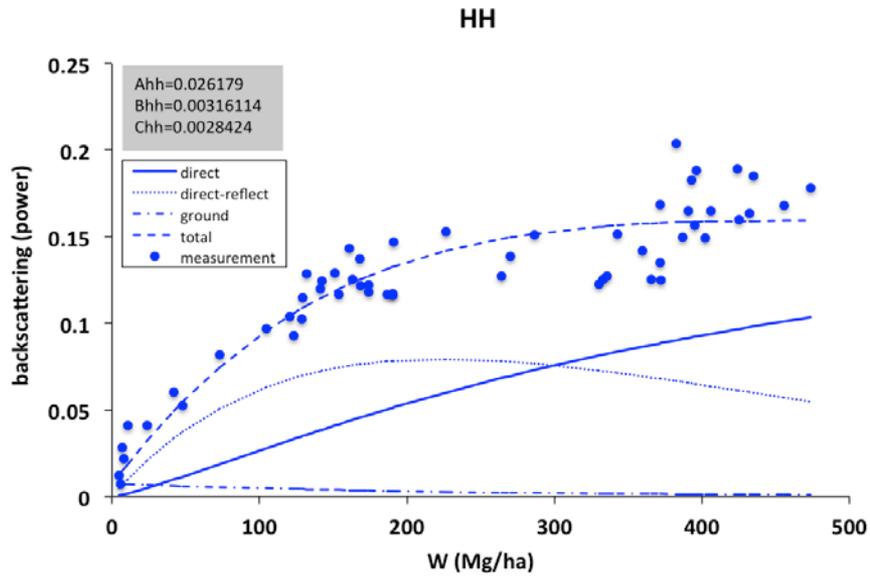
Model shape – tropical data



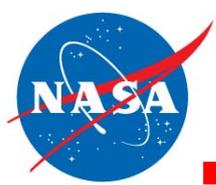
	HH	VV	HV
RMSE	0.016819935	0.0142891	0.00504441
R ²	0.930469068	0.933139812	0.955186855



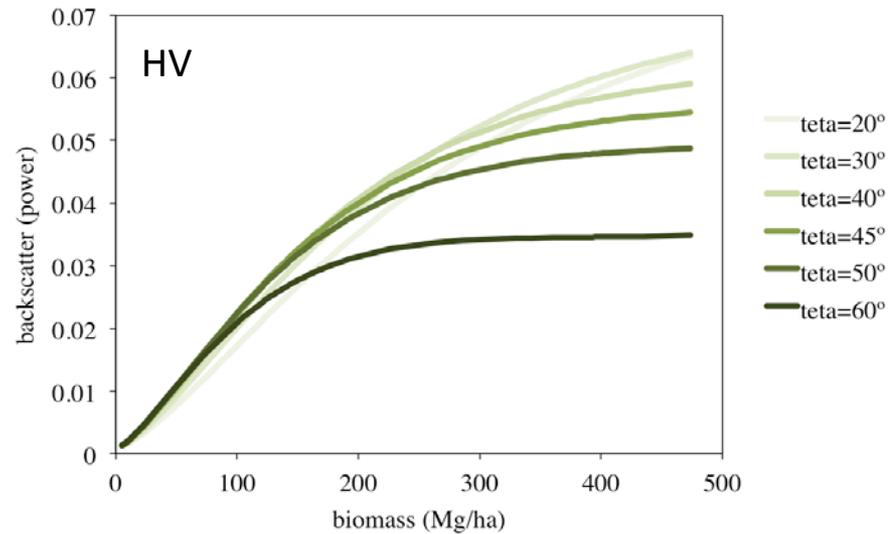
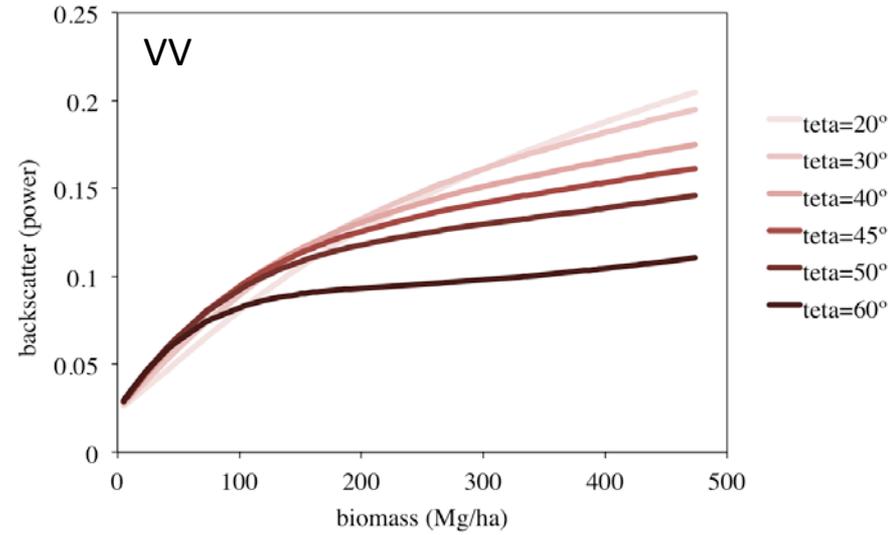
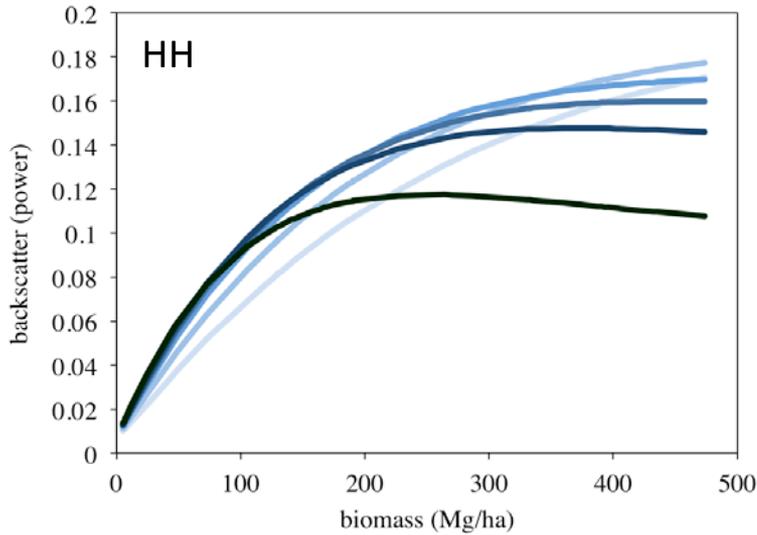
Shape of each scattering

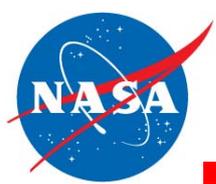


- - - - total
- volume
- double-bounce
- . . - surface
- • • • measurements

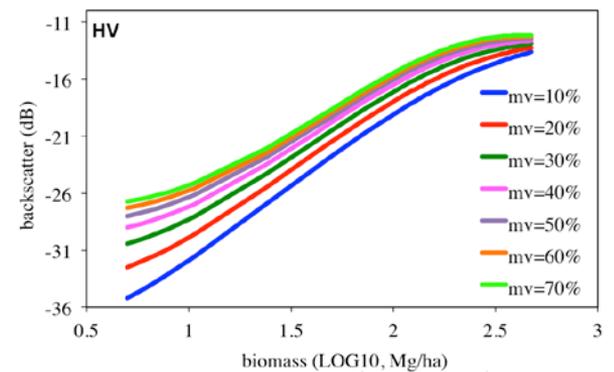
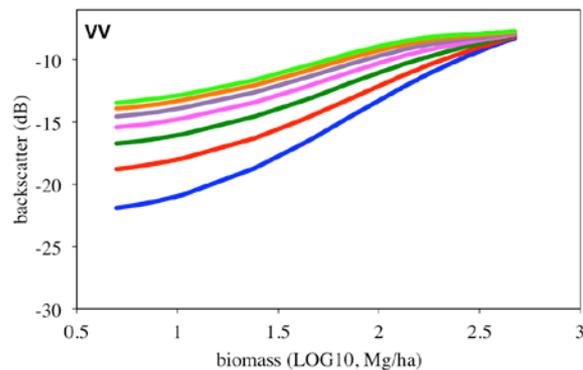
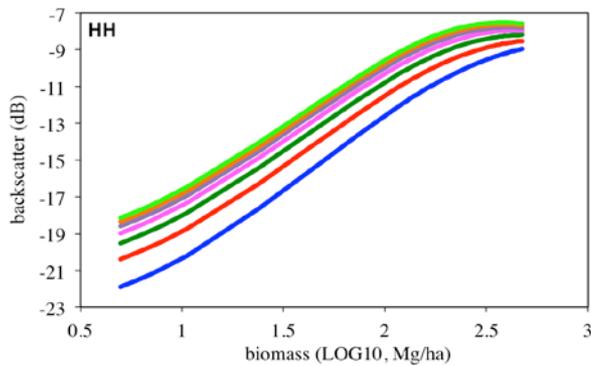
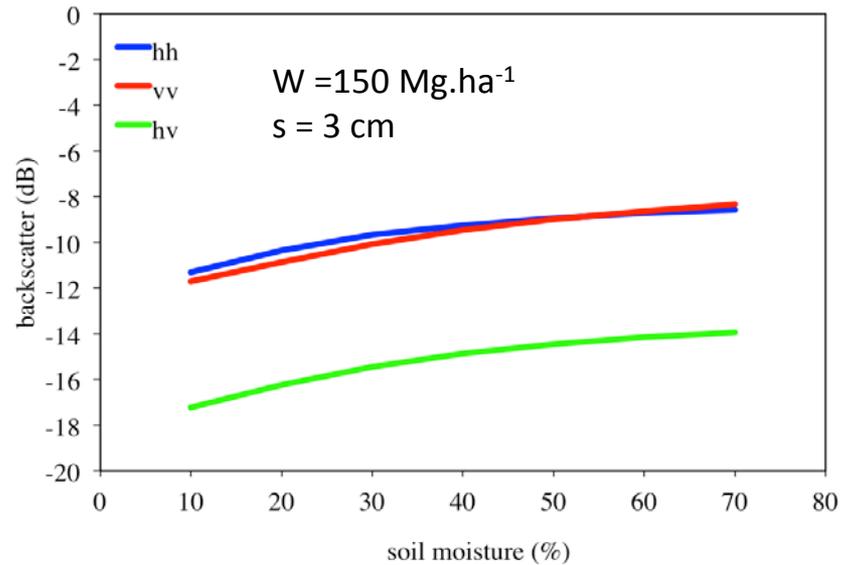


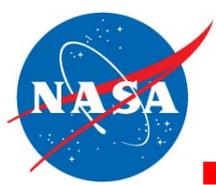
Incidence angle effect



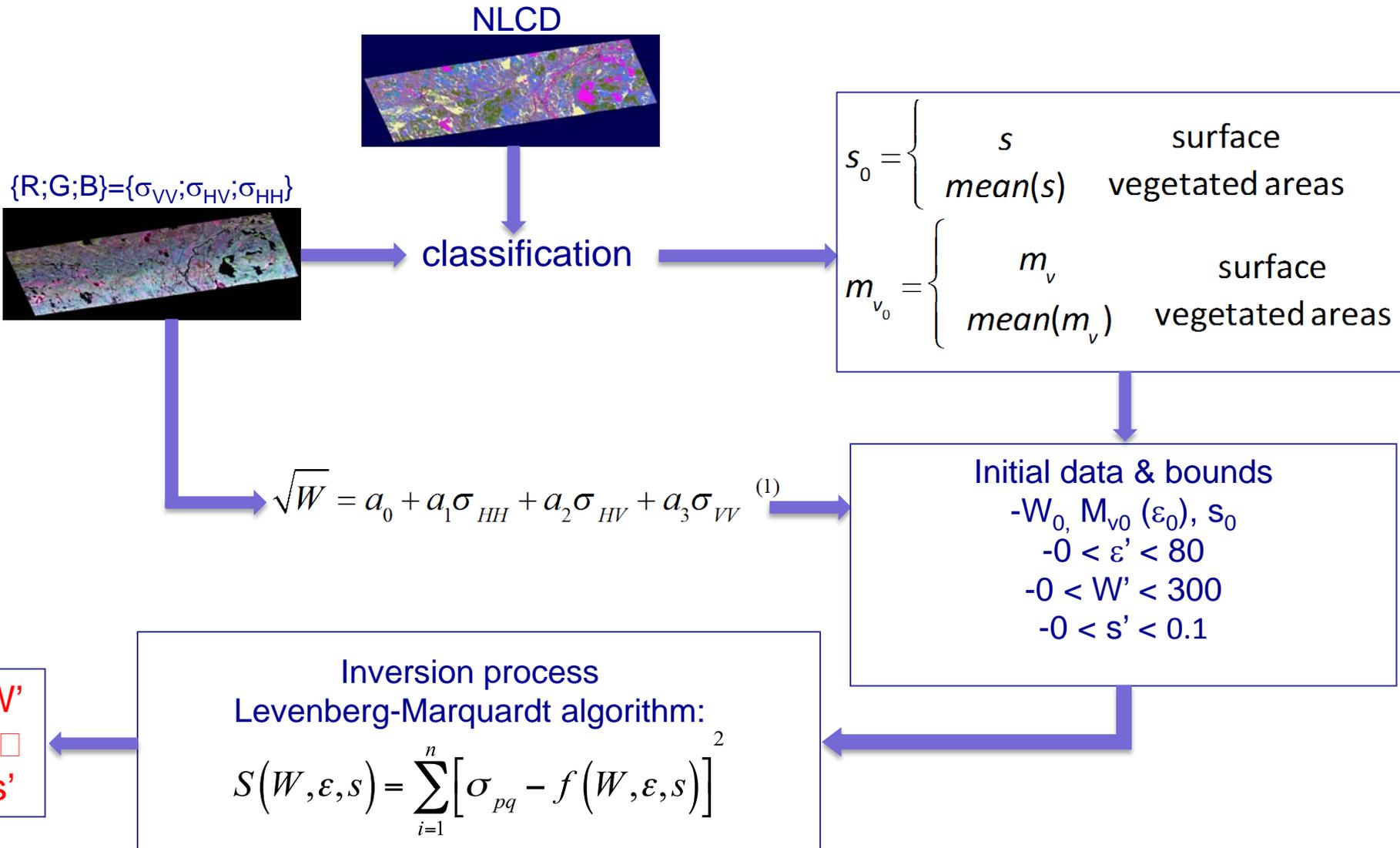


Soil moisture sensitivity





Inversion process



$$\sqrt{W} = a_0 + a_1 \sigma_{HH} + a_2 \sigma_{HV} + a_3 \sigma_{VV} \quad (1)$$

Initial data & bounds

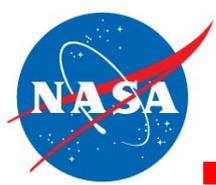
- $-W_0, M_{v_0} (\epsilon_0), s_0$
- $-0 < \epsilon' < 80$
- $-0 < W' < 300$
- $-0 < s' < 0.1$

Inversion process
Levenberg-Marquardt algorithm:

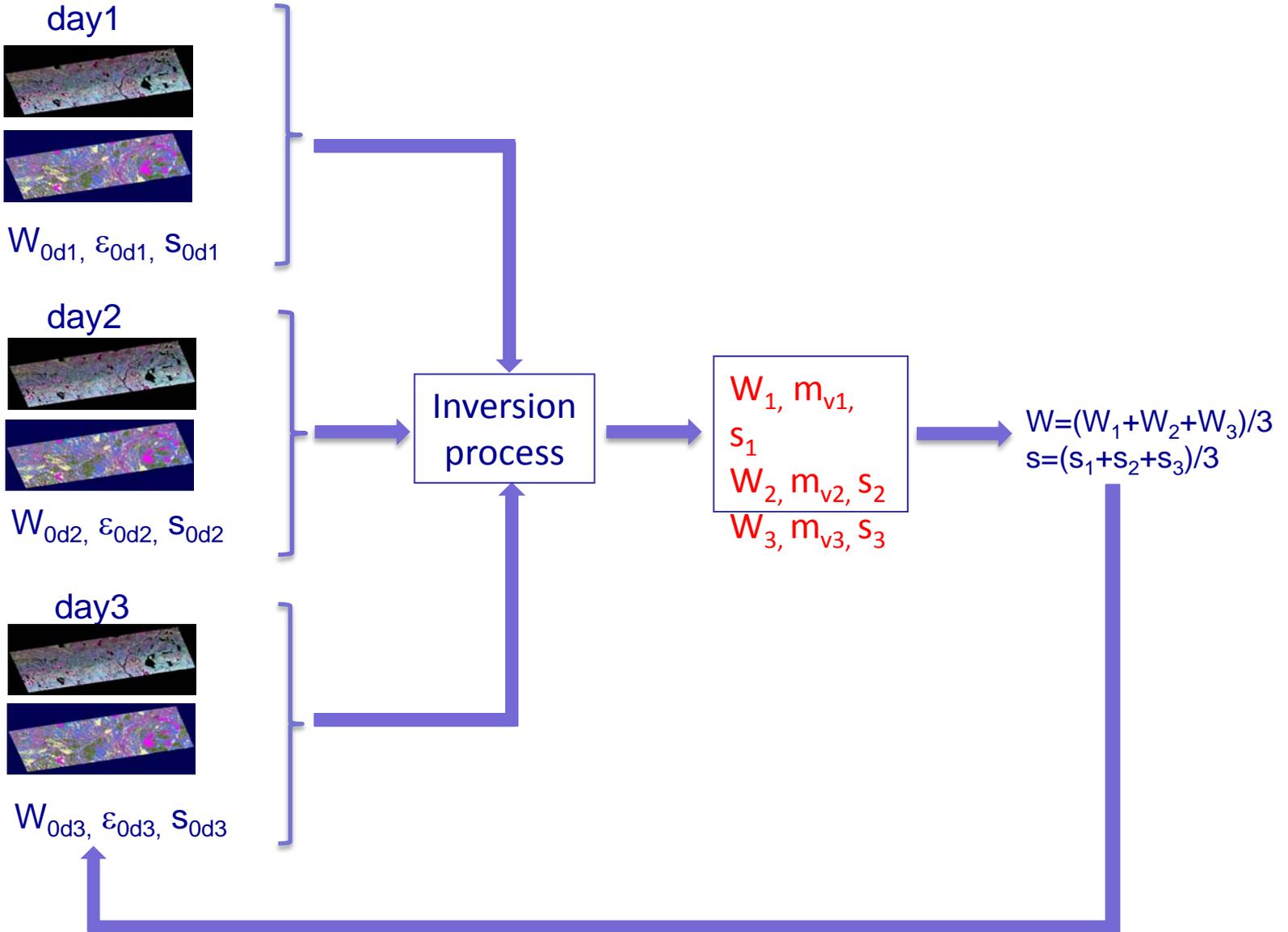
$$S(W, \epsilon, s) = \sum_{i=1}^n [\sigma_{pq} - f(W, \epsilon, s)]^2$$

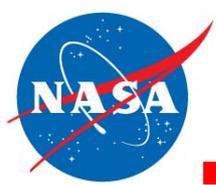
W'
 ϵ'
 s'

(1) "Impact of spatial variability of tropical forest structure on radar estimation of aboveground biomass", S. Saatchi, M. Marlier, R. L. Chazdon, D. B. Clark, A. E. Russell, Remote Sensing of Environment, vol. 115, no. 11, pp. 2836-2849, 2011.

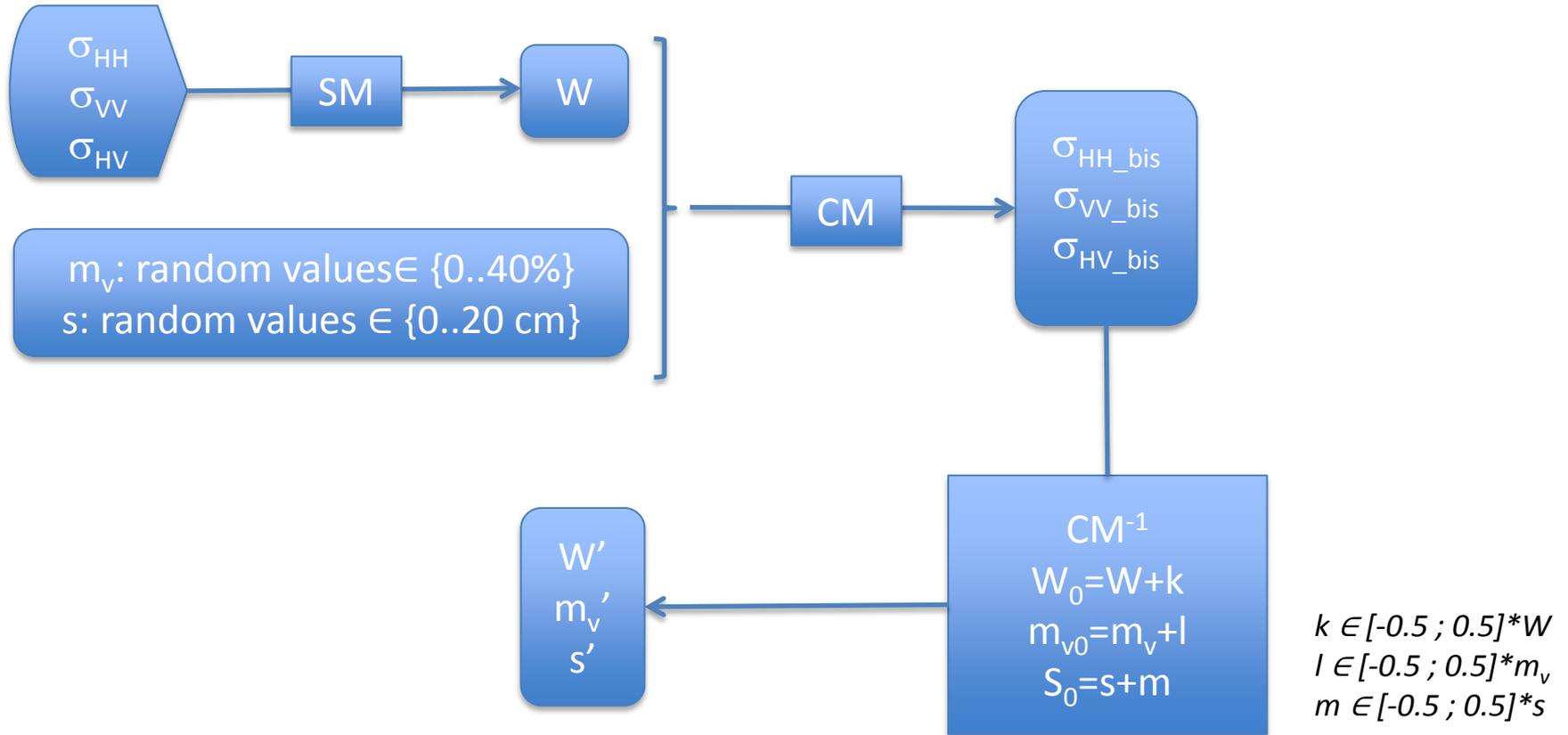


With time series measurements





Sensitivity analysis



$$SM: \sqrt{W} = a_0 + a_1 \sigma_{HH} + a_2 \sigma_{HV} + a_3 \sigma_{VV}$$

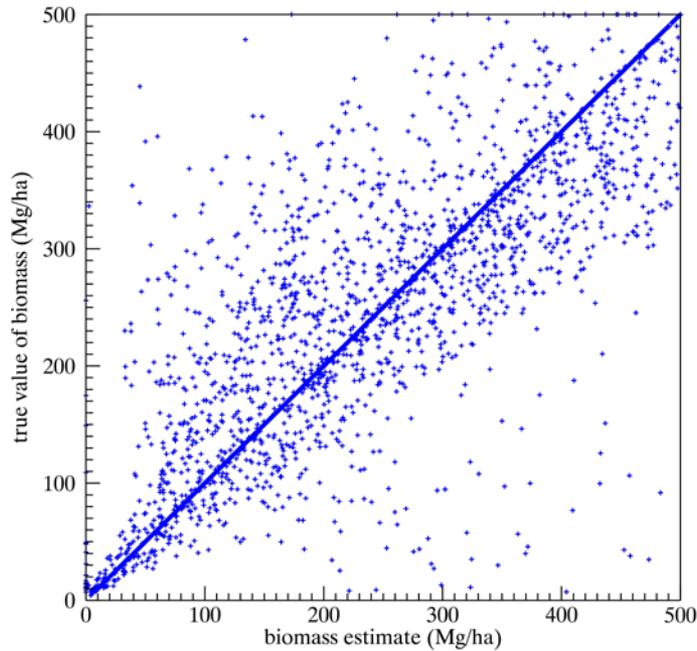
$$CM: \sigma_{HH}^0 = A_{HH} \cos \theta_0 W^{\alpha_{HH}} (1 - \exp(-B_{HH} W^{\beta_{HH}} \sec \theta_0)) + C_{HH} \Gamma_{HH} W^{\delta_{HH}} \sin(\theta_0) \exp(-B_{HH} W^{\beta_{HH}} \sec \theta_0) + S_{HH} \exp(-B_{HH} W^{\beta_{HH}} \sec \theta_0)$$

$$\sigma_{VV}^0 = A_{VV} \cos \theta_0 W^{\alpha_{VV}} (1 - \exp(-B_{VV} W^{\beta_{VV}} \sec \theta_0)) + C_{VV} \Gamma_{VV} W^{\delta_{VV}} \sin(\theta_0) \exp(-B_{VV} W^{\beta_{VV}} \sec \theta_0) + S_{VV} \exp(-B_{VV} W^{\beta_{VV}} \sec \theta_0)$$

$$\sigma_{HV}^0 = A_{HV} \cos \theta_0 W^{\alpha_{HV}} (1 - \exp(-B_{HV} W^{\beta_{HV}} \sec \theta_0)) + C_{HV} \Gamma_{HV} W^{\delta_{HV}} \sin(\theta_0) \exp(-B_{HV} W^{\beta_{HV}} \sec \theta_0) + S_{HV} \exp(-B_{HV} W^{\beta_{HV}} \sec \theta_0)$$

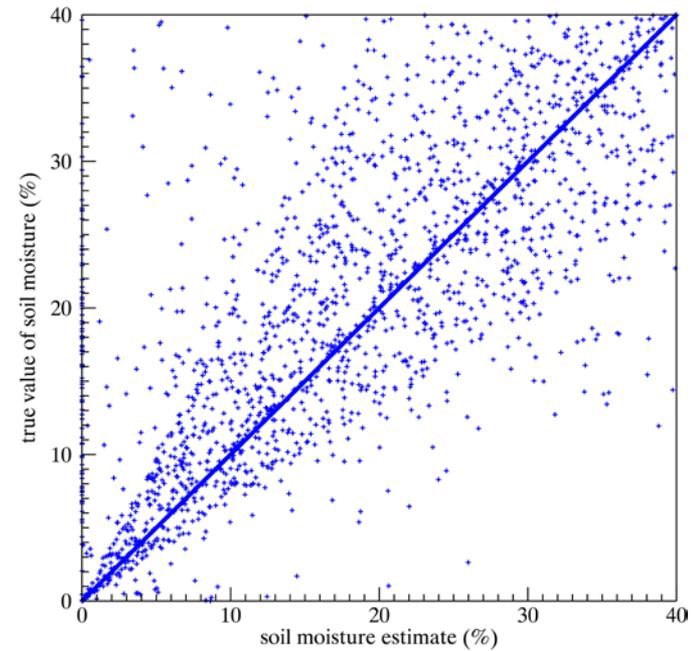
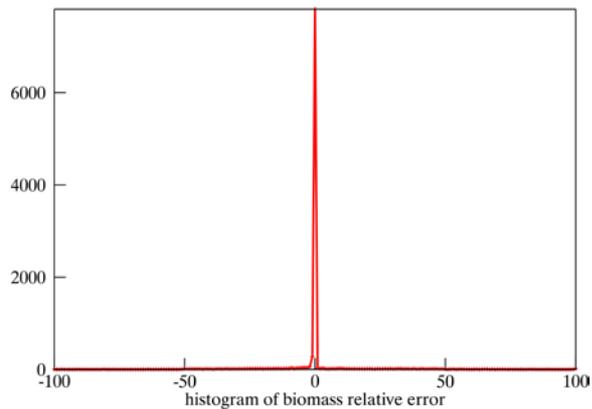


Sensitivity analysis - results



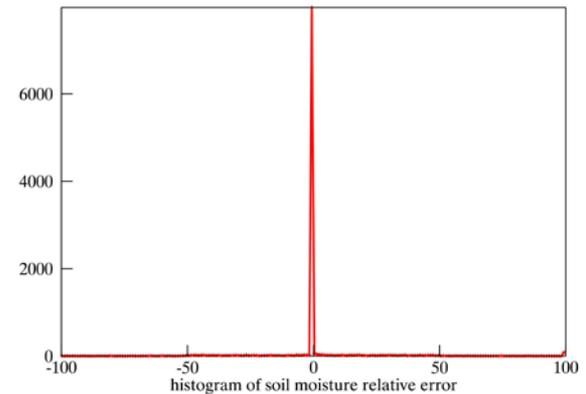
RMSE = 50 Mg/ha

88% of pixels have a relative error ~ 0



RMSE = 4.35 %

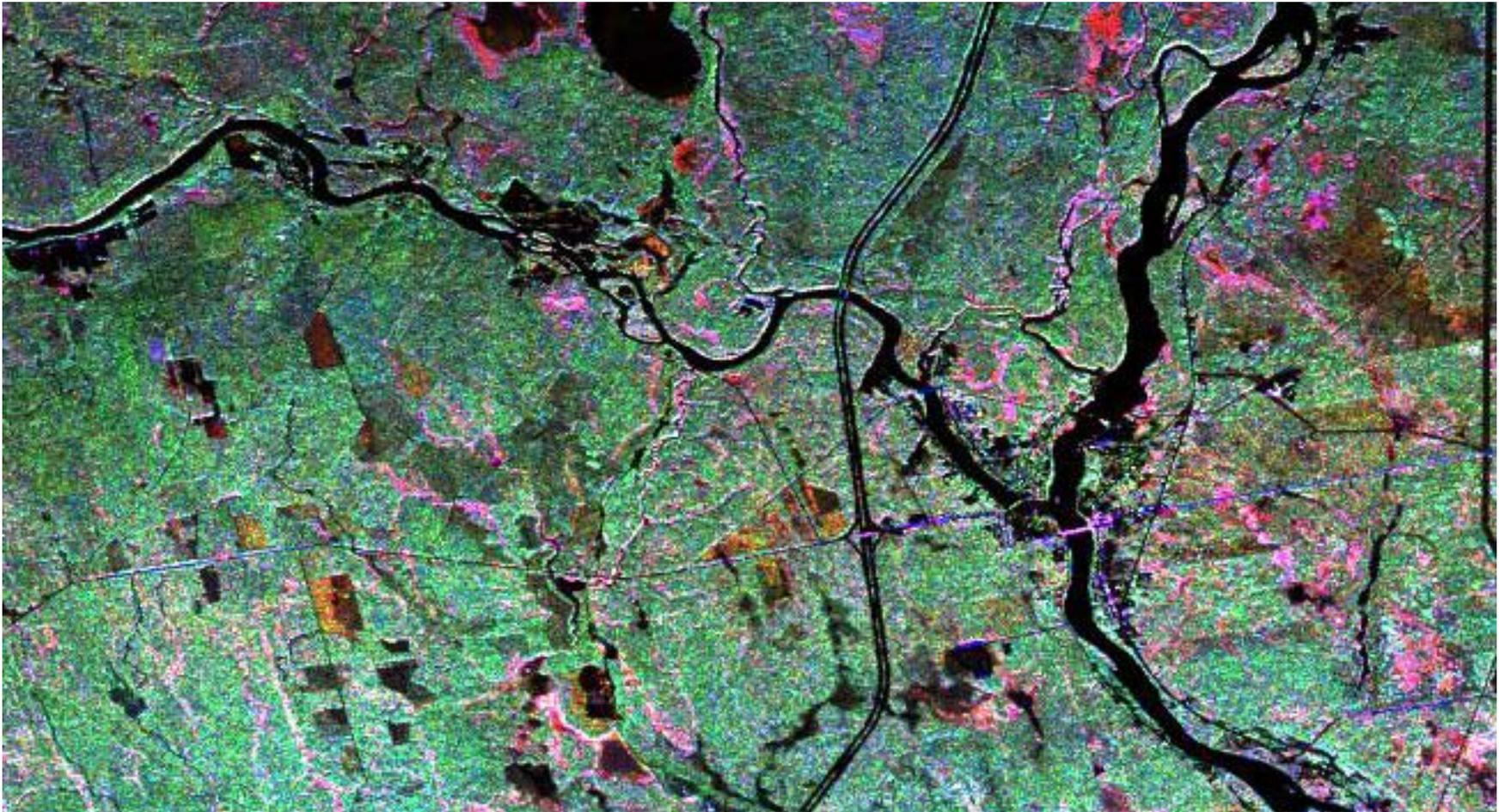
90% of pixels have a relative error ~ 0





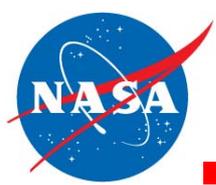
Application

AirSAR data - Howland forest – Maine - October 1994

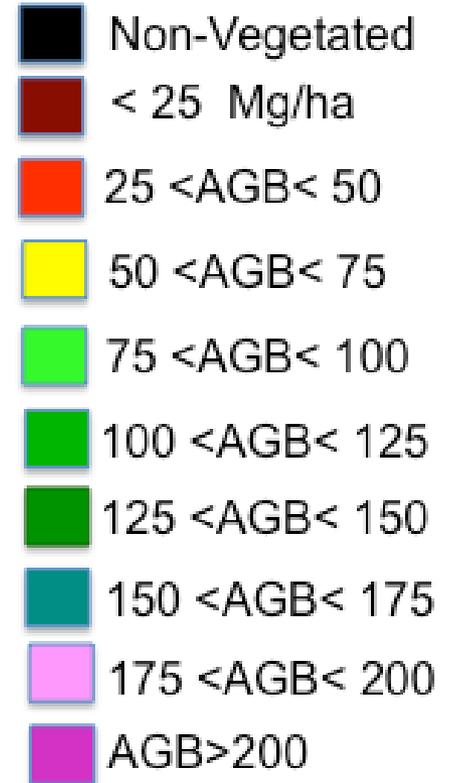
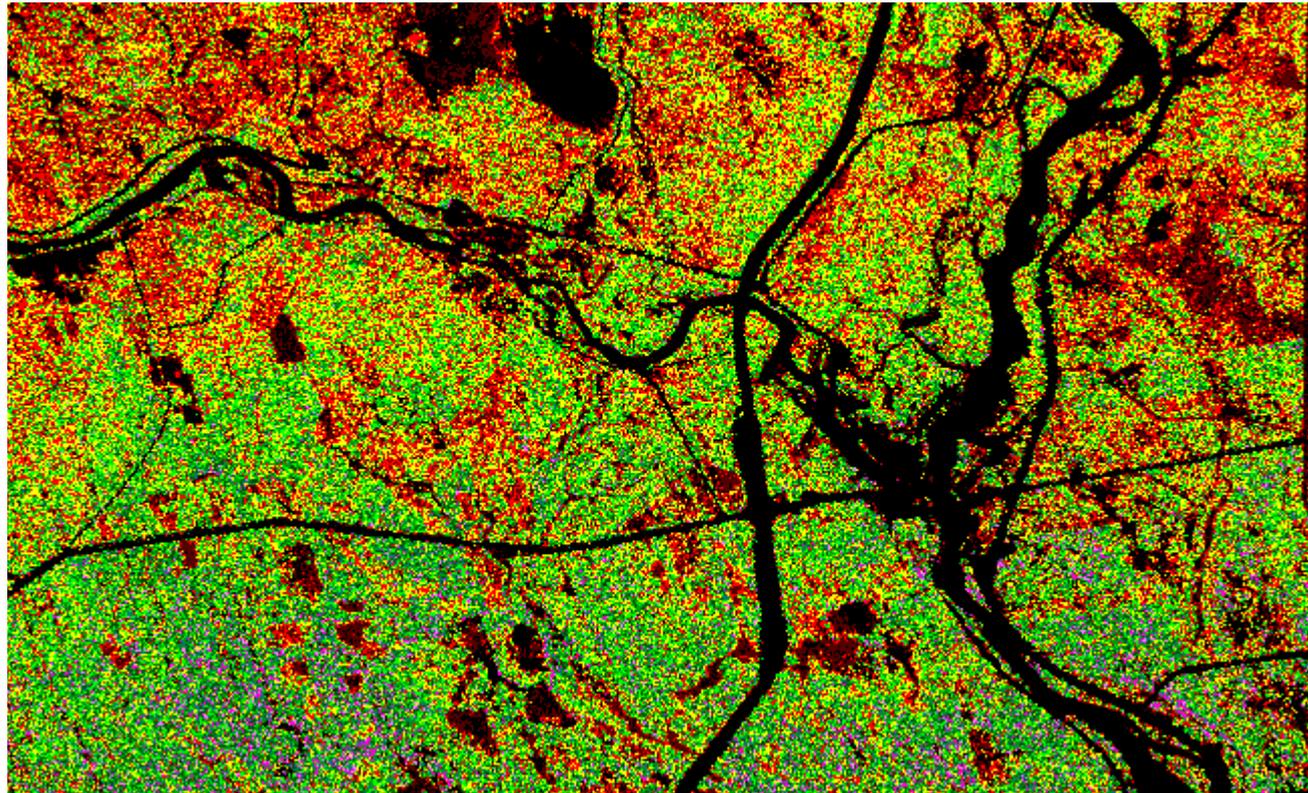


σ_{VV} ; σ_{HV} ; σ_{HH}

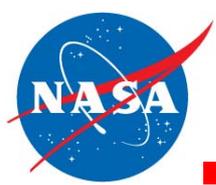
Pixel Size: 1 arcsec



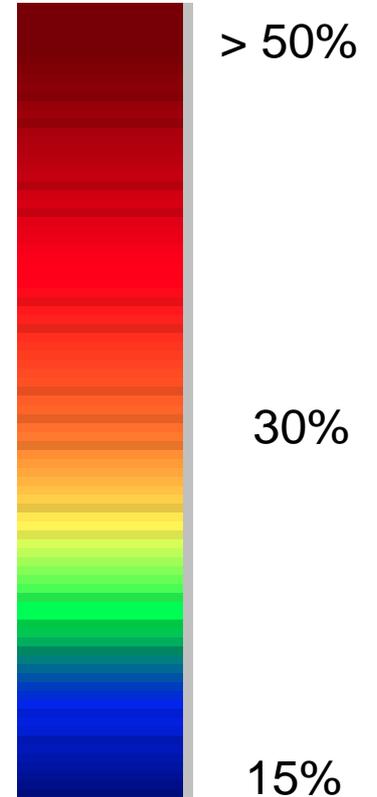
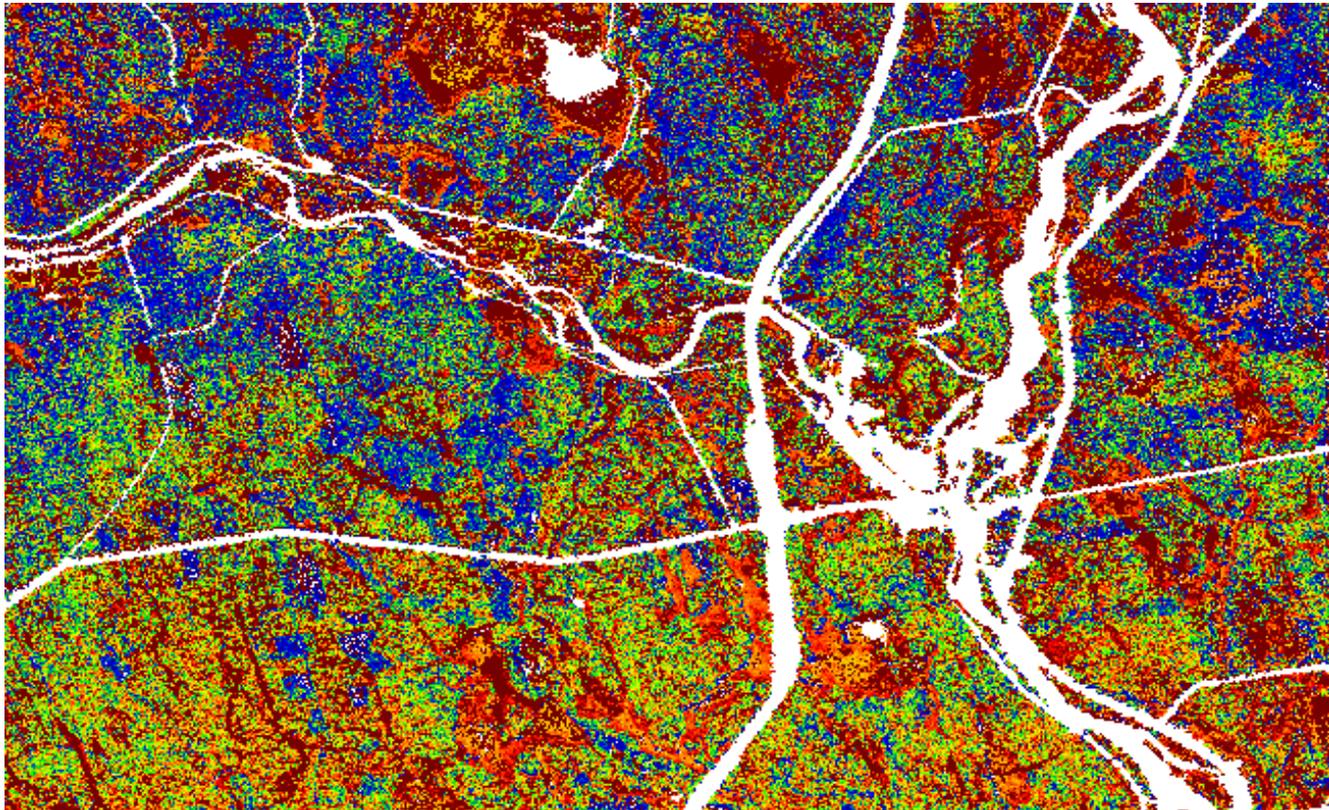
Aboveground Vegetation Biomass



$0 < W < 300 \text{ Mg/ha}$
 $\text{Mean}(W) = 64 \text{ Mg/ha}$



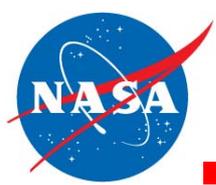
Soil moisture map



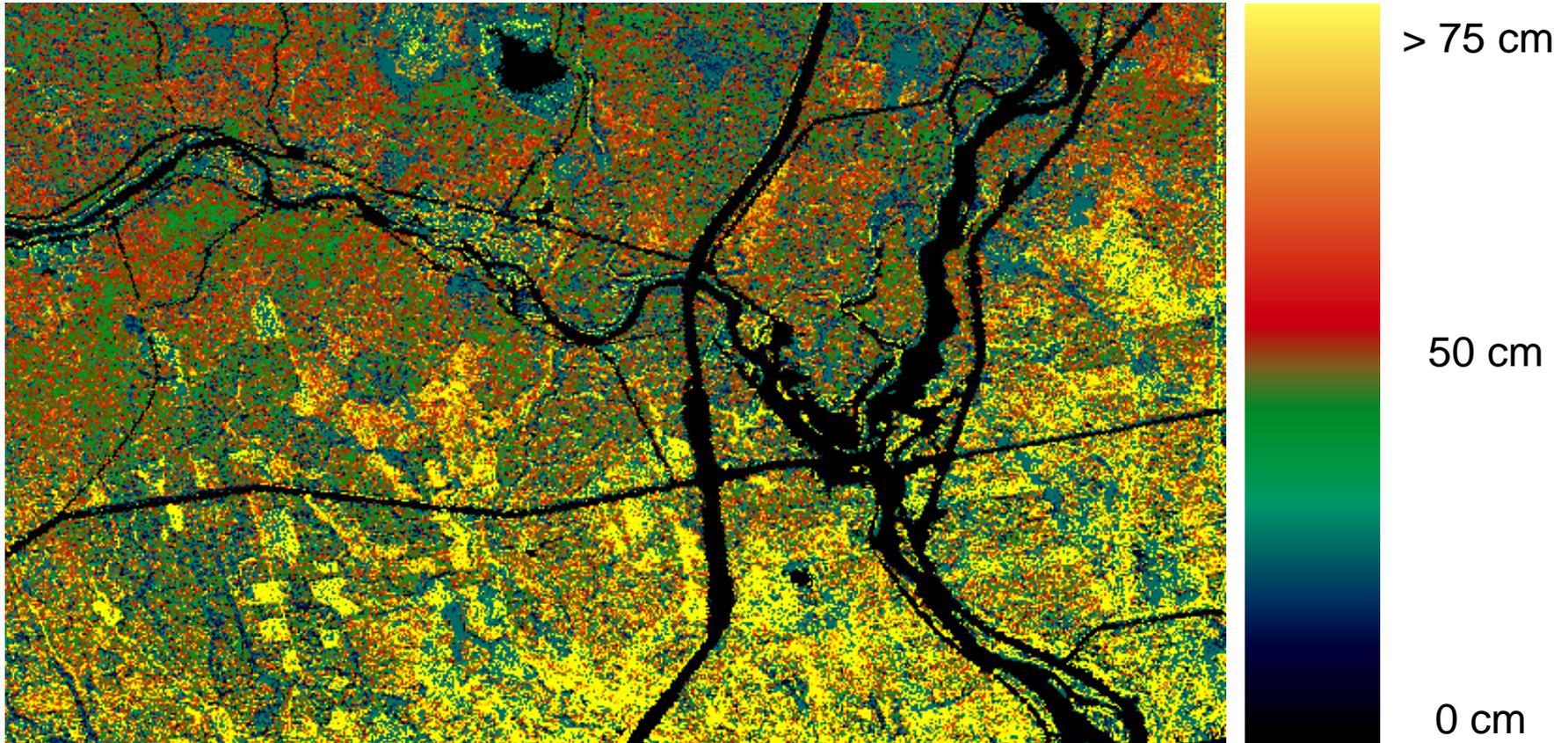
$0 < mv < 50\%$

Ground measurement = 18.4%

Estimated value on this particular point = 21.5%

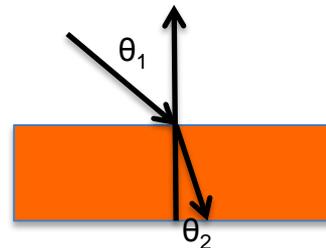


Soil penetration depth

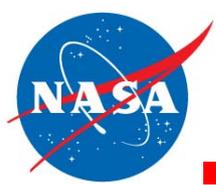


$$d = \frac{-\log_e(0.1)\lambda}{2\pi n_i}; \quad n_i = \text{Im}(\sqrt{\epsilon_r})$$

$$\text{Depth} = d \cos \theta_2, \quad \sin \theta_2 \sqrt{\epsilon_2} = \sin \theta_1 \sqrt{\epsilon_1}$$

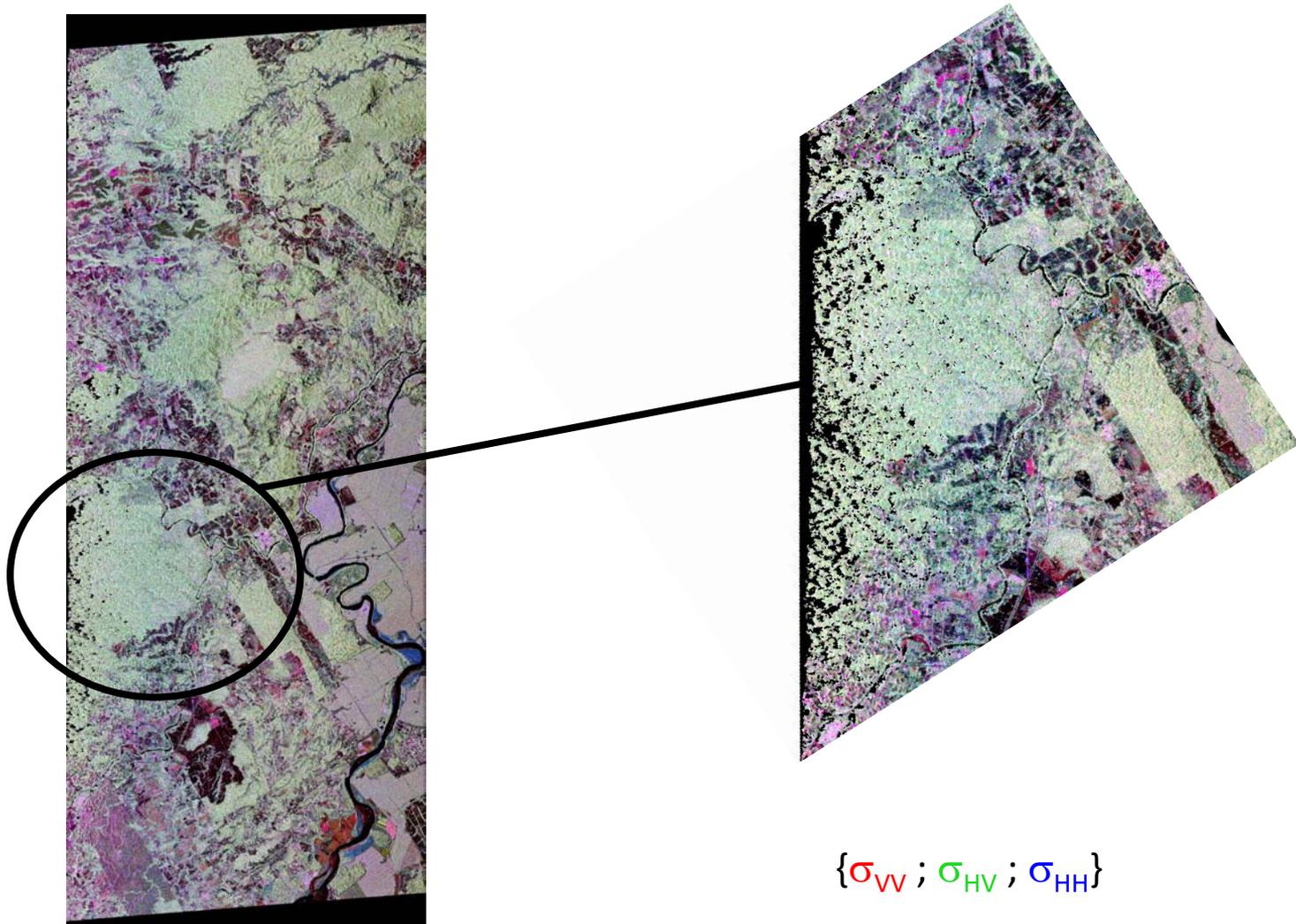


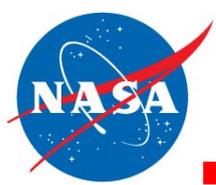
$0 < d < 75 \text{ m}$
Mean(d) = 50 cm



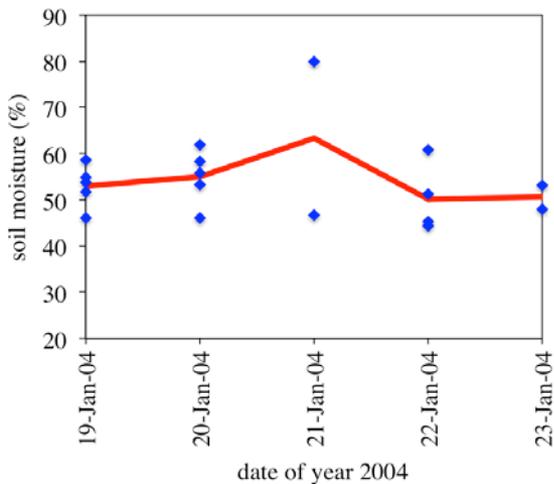
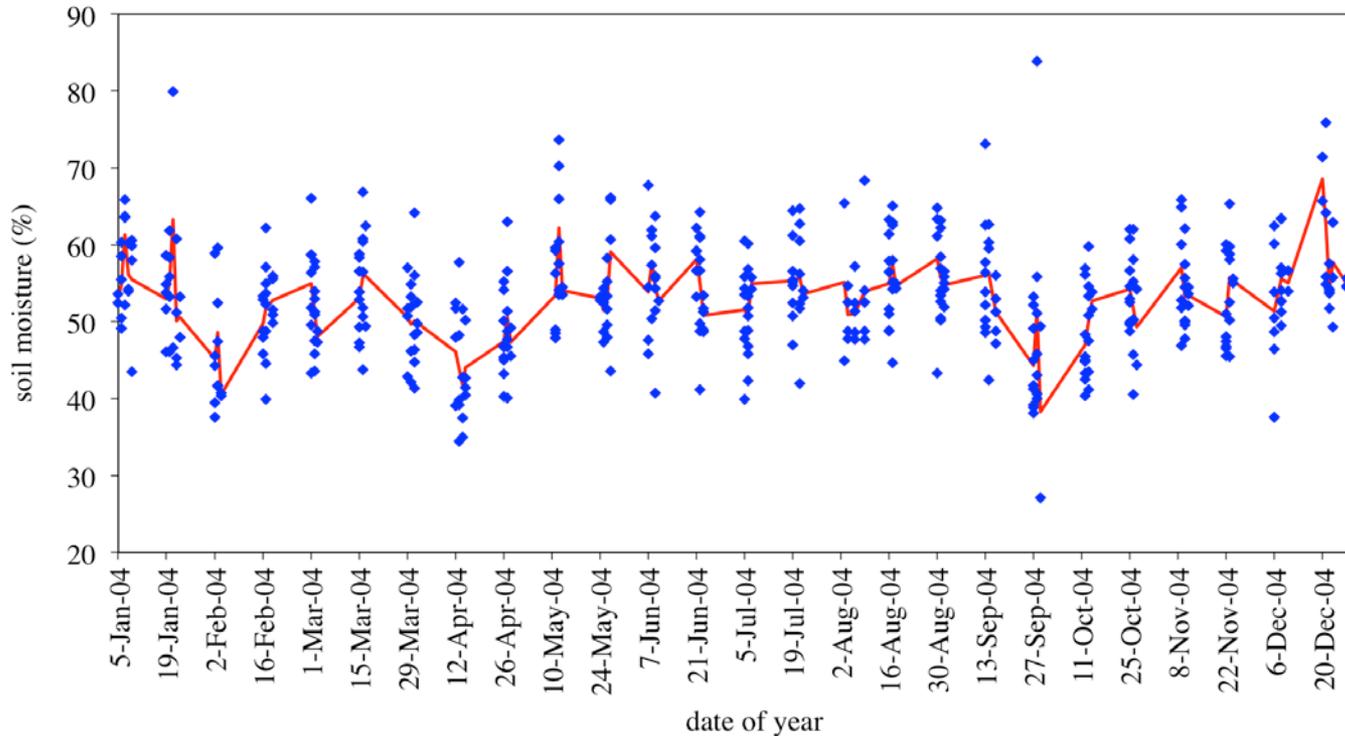
La Selva – Costa Rica

AirSAR data - La Selva - Costa Rica – March 6th, 2004

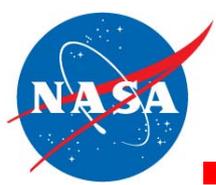




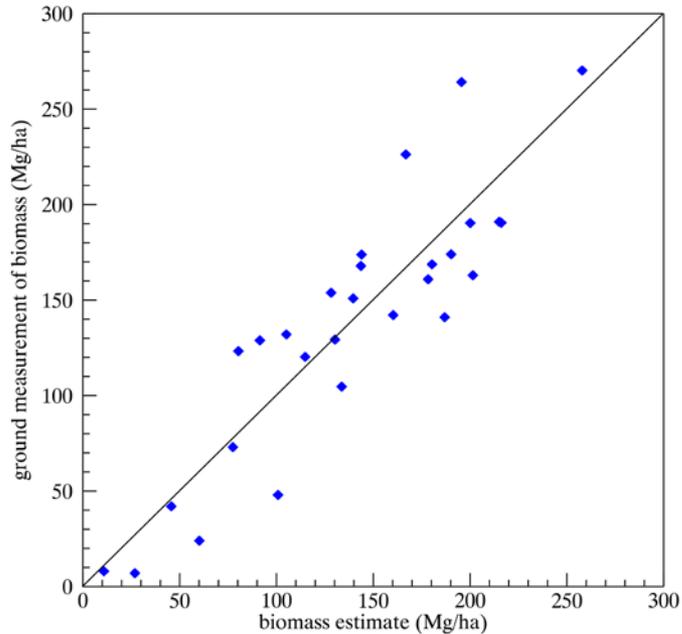
La Selva volumetric soil moisture in 2004



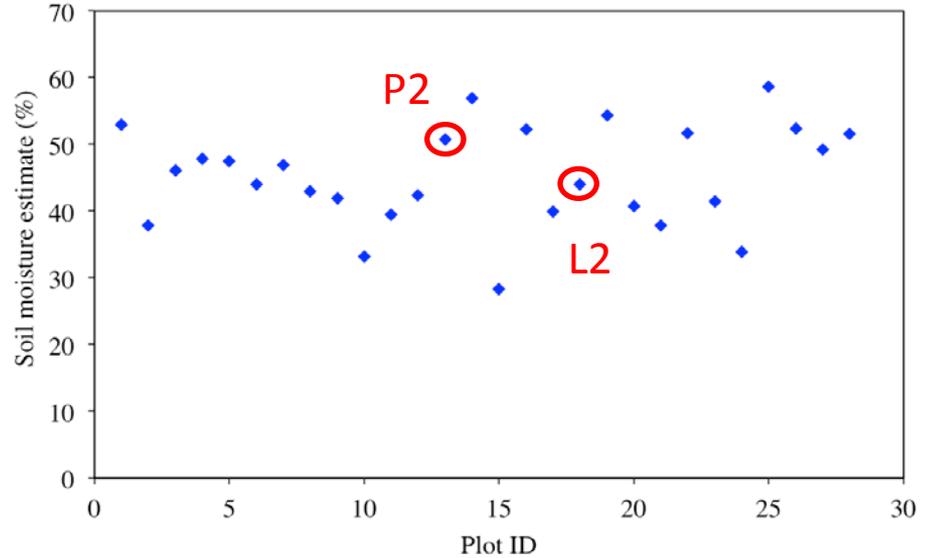
Courtesy of Dr. Diego Dierick for providing the data



Results on tropical forest - La Selva/Costa Rica

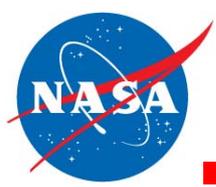


RMSE = 30.5 Mg/ha

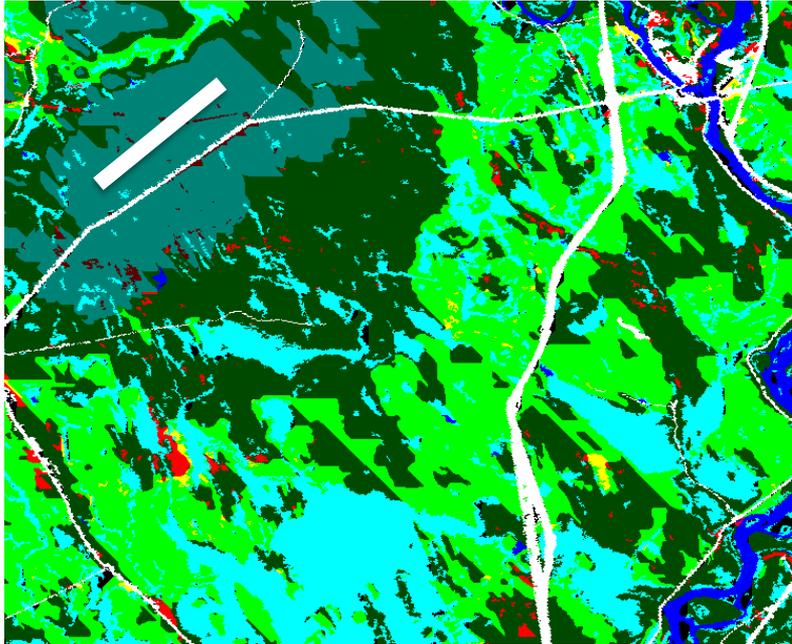


$28.3 \% < m_v < 58.7 \%$
Average(m_v)=45.2 %
P2: $m_v = 48.8\%$, $m_v' = 50.8\%$
L2: $m_v = 47.4\%$, $m_v' = 43.9\%$

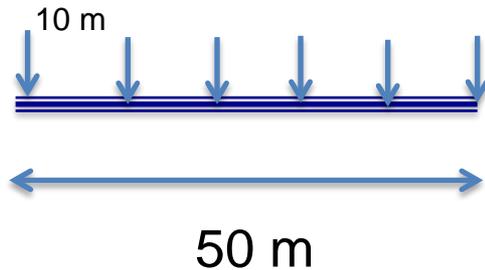
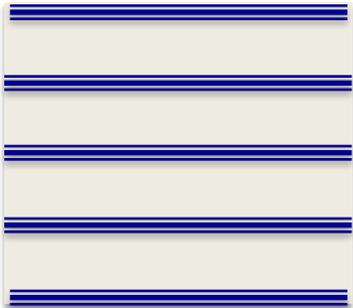
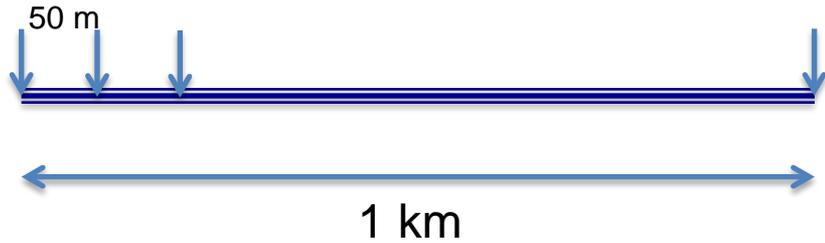
Field campaigns



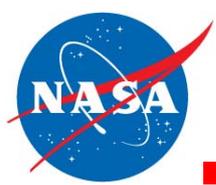
Sampling Strategy



1 km transects with sampling at 50m intervals with GPS at each location



Collect 5 parallel 50 m transects
With sampling at 10m intervals
with GPS at each location

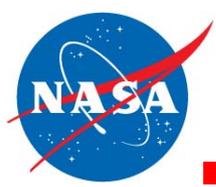


Soil Moisture TDR Sensors

2 units

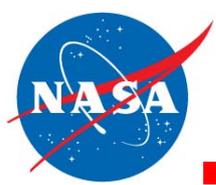
6050X1 TRASE SYSTEM I



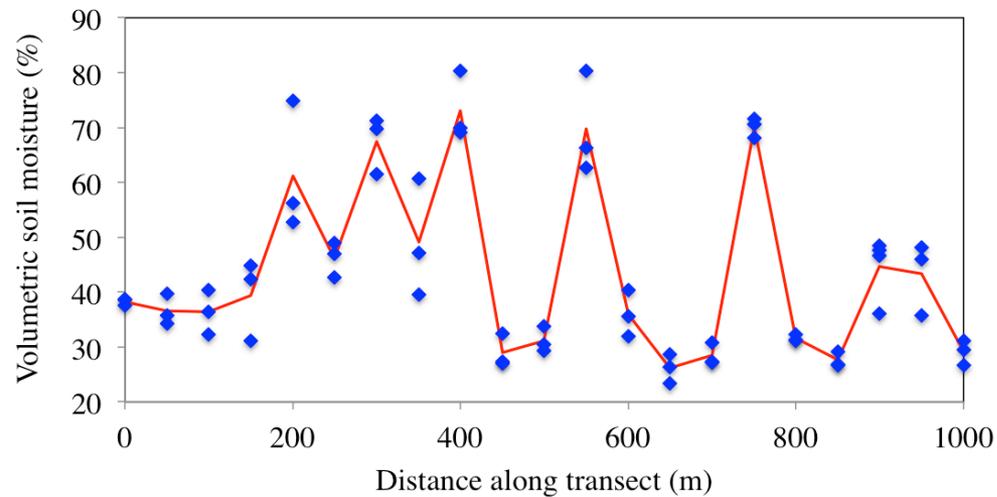
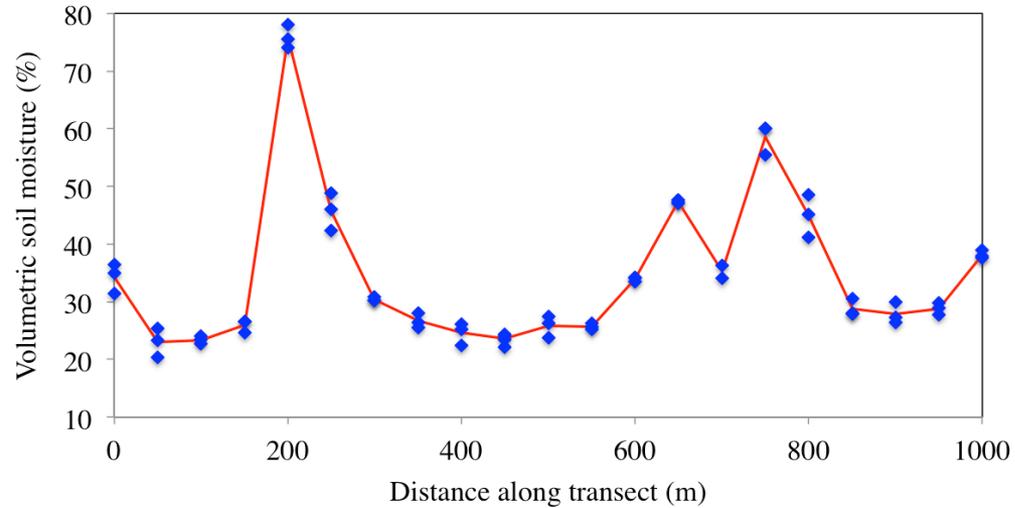


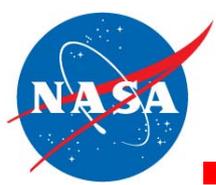
Soil Moisture TDR Sensors



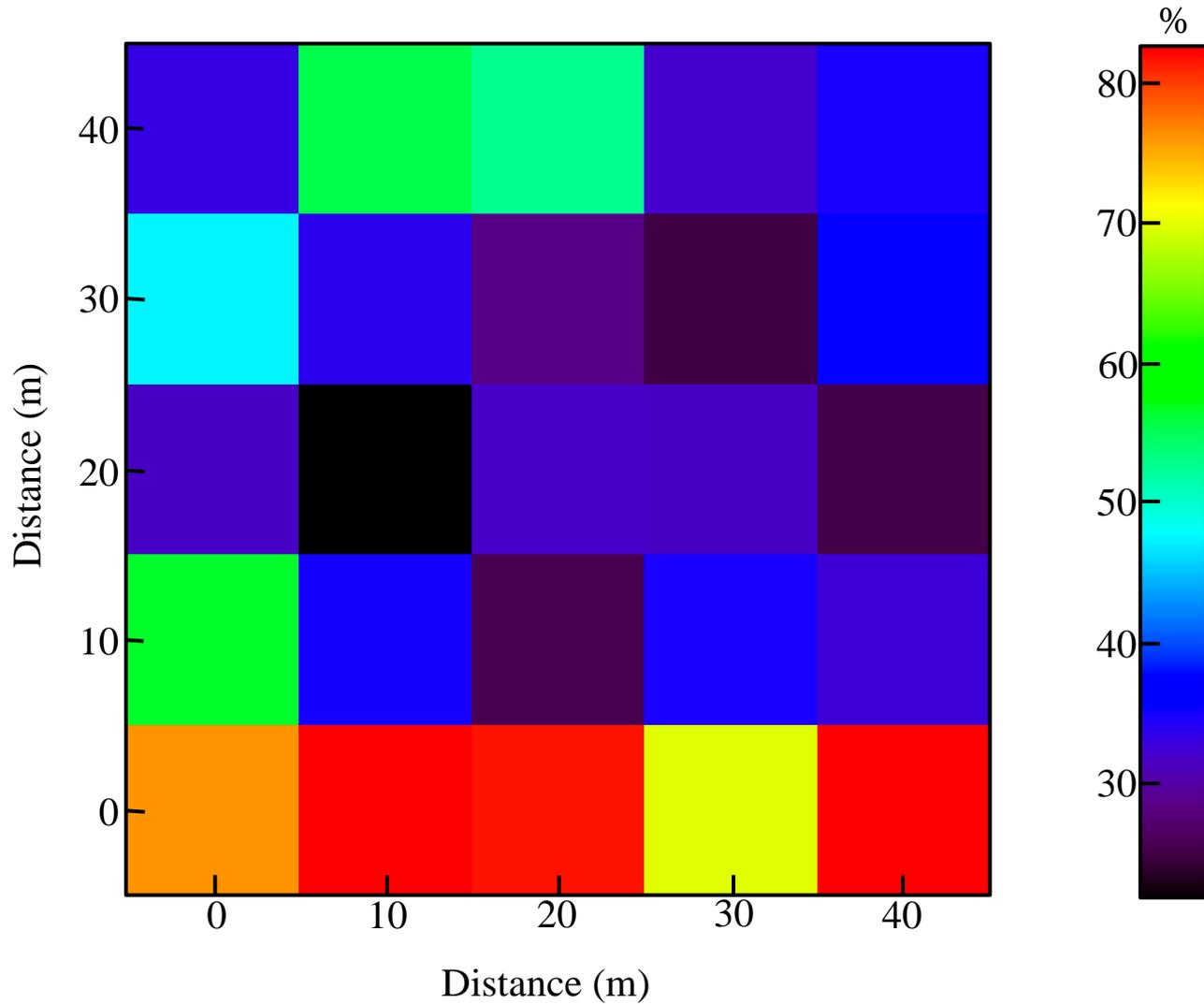


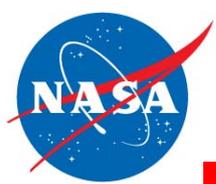
Field measurements examples





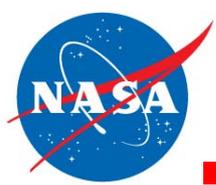
Spatial variation of soil moisture





Summary & Perspectives

- Semi-empirical model
 - Is a simplification of the distorted Born model
 - Requires ground measurements to set up calibration coefficients
 - Is sensitive to:
 - Biomass
 - Soil moisture
 - Incidence angle
- Inversion process
 - Uses LM method to invert the forward model
 - Needs initialization and constraints
 - Shows good results over simulated data
 - Biomass: RMSE = 50Mg/ha for a biomass level up to 500 Mg/ha
 - Soil moisture: RMSE = 4.35% for soil moisture up to 40%
 - Estimates soil moisture and biomass with AirSAR P-band data acquired over
 - Howland forest in October 1994 with an absolute error of 3%.
 - La Selva forest in March 2004 with a RMSE of 30.5 Mg/ha for a biomass level up to 300 Mg/ha and soil moisture in the expected range.
- Field measurements show the spatial variation of soil moisture.
- Future works: topographic effect, using AirMOSS SAR data



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