SAR Theory and Applications to Forest Cover and Disturbance Mapping and Forest Biomass Assessment

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Outline

- 1. Introduction: Why Forest Observation?
- 2. SAR Techniques of interest for forestry applications
- 3. SAR for Forestry Applications Some Basics
- 4. Forest Cover and Biomass Mapping Excurses
 - 1. BIOMASAR Hypertemporal C-band Data Assimilation
 - 2. Forest Cover Mapping Using Backscatter and Coherence
 - 3. Forest Biomass Mapping Using Backscatter and Coherence
 - 4. Polarimetry for Forest Cover Mapping
 - 5. INSAR Phase and Tree Height
 - 6. Seasonality of C-band Backscatter in Siberia
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 - 9. Mapping of woody cover in KNP using L-band backscatter



Introduction - Why Forest Observation?



Observe clear-felling (ALOS PALSAR, Multitemporal Composite, Siberia)

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Introduction - Why Forest Observation?



Observe damage by forest fires (mid-August 2010, fires close to Moscow)











The wood stack on the photo contains approx. 1.000.000 m3. It is 60 m wide, 16 m high, and more than 2 km long. The storm "Gudrun", which hit southern Sweden in January 2005 fell approx. 75.000.000 m3, which is almost the annual cut in Sweden. Photo: Ola Nilsson





- Stem Biomass is strongly related to the commercially interesting biomass.
- The major part of forest biomass is concentrated in the major trees. The contribution of minor trees (and hidden biomass) to total biomass is rather low
- Characterizing biomass using tree height will cover 75-95% of the vegetated earth and could directly characterize 80-90% of the aboveground biomass stock

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Introduction - Main components of biomass distribution



Field-measured biomass distributions for different sections (stem, bark, living and dead branches) of the tree for Scots pine and Norway spruce Kankare et al. (2013): Individual tree biomass estimation using terrestrial laser scanning, ISPRS Journal of Photogrammetry and Remote Sensing 75, pp. 64–75.

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DID YOU KNOW?

- Forests cover approximately 33% of the Earth's land surface (JENSEN, 2000)
- Forests play an important role in the global carbon cycle, since each year forests absorb approximately 1/12 of the Earth's atmospheric CO₂ stock (MALHI et al., 2002)
- Forested ecosystems account for app. 72% of the Earth's terrestrial carbon storage (MALHI et al., 2002)
- Therefore, Vegetation biomass is a *larger global store of carbon than the atmosphere* (FAO, 2009)
- Between 1850 and 2011, humans have released app. 480 Gt (480 BILLION TONS!!!) of CO₂ into the atmosphere through fossil fuel burning and land use changes (e.g. deforestation and fires) (GHASEMI et al., 2011)

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Forest Biomass

In Forestry, the biomass calculation is based on measurements of trunk diameter and height of sample populations of trees:

 $Biomass_{forest} = N \times \pi \times \left(\frac{1}{2} \text{ dbhmid}\right)^2 \times hmid \times \rho \times fz$

Biomass _{forest} [t/ha]	is defined as aboveground woody of trunk and branches where exceeding 7 cm diameter
dbh _{mid} [cm]	is the (dbh² weighted) mean diameter at breast height 1.3 m
<i>h_{mid}</i> [m]	is the height of the tree
ρ [g/cm³]	is the species-specific wood density
f_{z} []	is a form factor (= 0.4-0.5, constant in a first order approximation)
Ν	is the tree density (tree number per area unit)

The product of $N \times \pi \times \left(\frac{1}{2} \operatorname{dbh}_{mid}\right)^2$ is also called **basal area g**.



WHY DO WE NEED TO OBSERVE (GLOBAL) FOREST BIOMASS?

- For a better understanding and quantification of:
 - the global carbon cycle
 - global warming
 - · terrestrial carbon stocks and fluxes in forests
 - terrestrial carbon sources and sinks
- Information of forest biomass is needed to *support sustainable forest resource management*



- 3. Forest Biomass Mapping Using Backscatter and Coherence
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Correlation between SAR data and stem volume

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SAR Techniques: Interferometry

Coherence and INSAR phase contain information on forest

Interferometric Coherence – correlation of two complex SAR images



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SAR Techniques: Interferometric Coherence



Fig. 33: Temporal Change of the surface (PALLAN o. J.:o. S.).

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SAR Techniques: Interferometry

Coherence and INSAR phase contain information on forest

• Interferometric Phase $\Phi_{\text{flat earth}}$ $\Phi_{\text{topography}}$ Φ_{motion} $\Phi = \Phi_{\text{flatearth}} + \Phi_{\text{topography}} + \Phi_{\text{motion}} + \Phi_{\text{atmosphere}} + \Phi_{\text{noise}}$





A complex SAR image can be decomposed into ...



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Phase Difference of Two SAR Images



Phase in one SAR image looks random (→speckle effect!). Only after accurate co-registraton the phase difference reveals the interferogram.

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Phase Difference of Two SAR Images



Phase in one SAR image looks random (→speckle effect!). Only after accurate co-registraton the phase difference reveals the interferogram.





Cotopaxi volcano Ecuador (SRTM/X-SAR)











Coherence and InSAR phase



[MFFU Sommerschule 2000]









- Both techniques require at least two complex SAR images
- A) Same polarisation different position



Interferometry

B) Same position – different polarisation

















SAR Techniques: Polarimetry

- Investigation backscatter at different polarisations
- Computation of polarimetric parameters









Konstantinos Papathanassiou et al. (DLR) ESA PECS SAR Remote Sensing Course, July 2017, Vilnius





SAR Polarimetric Tomography





SAR Polarimetric Tomography



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SAR Polarimetric Tomography



Figure 6.14: Backscattering from reference surface (bare soil with very low vegetation).

Andreas Reigber - Dissertation

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SAR Polarimetric Tomography



Figure 6.15: Backscattering from forest stand 1 (spruce \sim 15-20m).

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Possible Scattering in Forest





C-band radar backscatter is more sensitive to structural properties of the forest if 1) the radar wave penetrates deeper into the canopy (e.g. frozen or dry conditions) and 2) if the backscatter from the ground is not strong (frozen or dry conditions, smooth soil)



Impact of different frequencies



LE TOAN et al. 2001: 4

Frequen-	Х	C	L	Р	VHF
cy band					
Main scatterers	Leaves, Twigs	Leaves Small branches	Branches	Branches & Trunk	Trunk









SATURATION PROBLEM

The saturation level of different wavelengths and polarizations depends on:

- wavelength (i.e. different bands, such as C, L, P)
- *polarization* (HV, HH and VV)
- **object characteristics** (vegetation stand structure and ground conditions)





Forest at different frequencies



- Variable response to water
- Variable response to open areas
- Can be used as indicator of environmental effects effecting the coherence



- Medium dynamic range
- Stable response to water
- Possible to identify
- agricultural fields
- Higher frame to frame variations



Forest at different frequencies



- Small dynamic range
- Variable response to water
- Variable response to open areas
- Can be used as indicator of environmental effects effecting the coherence



- Medium dynamic range
- Stable response to water • •
- Possible to identify agricultural fields
- Higher frame to frame variations



Higher contrast between forest/non forest

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- Higher sensitivity to forest volume
- Confusion between water and dense forest
- Frame to frame variations

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Linking SAR measures and Forest Parameters



Forest Parameters



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ENVISAT ASAR Wide Swath dataset

During 2003 and 2004 ENVISAT ASAR data in Wide Swath mode has been acquired over the study area of the SIBERIA-II Project; Several hundred ASAR scenes have been acquired, with a high degree of overlap between neighboring tracks







Modeling Example: A Water Cloud-like model

A water cloud with gaps is close to reality and easy to handle



Water cloud with gaps


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 σ_{gr}

 σ_{veg}

ß

 $\sigma_{for}^{o} = \sigma_{veg}^{o} \left(1 - e^{-\beta V} \right) + \sigma_{gr}^{o} e^{-\beta V}$ Unknown

ground backscatter canopy backscatter forest transmissivity coefficient

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Multi-temporal combination of single biomass estimates



- A multi-temporal combination of single estimates with weights determined by the backscatter contrast $\sigma^0_{\ veg}$ - $\sigma^0_{\ gr}$ allows obtaining the final estimate

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Multi-temporal combination of single biomass estimates



- · From a single image it is possible to identify sparse/dense forest patterns at most
- · From multi-temporal combination it is possible to identify biomass levels

(ESA BIOMASAR Project, Maurizio Santoro, 2007)



Retrieved GSV Map vs. in-situ data



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Impact of uncertainty of in situ data

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- The quality of the reference data can affect the retrieval statistics
- Cross-comparison with other EO data helped in bailing out extreme cases
- Retrieval statistics at full resolution embed a certain amount of error due to imprecision in the ref. data
- More correct results are obtained when aggregating



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Forest GSV + GLC2000 - Central Siberia - Pixel size: 1 km BIOMASAR GSV map of **Central Siberia** 1 km resolution 2,400,000 km² 500 **m³/ha** ENVISAT ASAR - Global • 400 Monitoring mode (Jan. 2005 -Feb. 2006) 300 GLC 2000 land cover used as 200 background 100

> 900 1000 1100 Easting (km)

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0 m³/ha









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Analysis of PALSAR data - FBS



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Forest Clear-cut separability

date	mode	position	
19MAY06	FBS	54°12'N 99°94'E	
19MAY06	FBS	55°59'N 99°58'E	
19MAY06	FBS	56°08'N 99°46'E	
14AUG06	FBS	54°12'N 101°56'E	
14AUG06	FBS	54°61'N 101°44'E	
27DEC06	FBS	56°84'N 104°16'E	
27DEC06	FBS	57°33'N 103°99'E	
13JAN07	FBS	56°83'N 103°62'E	
13JAN07	FBS	56°83'N 103°62'E	
11FEB07	FBS	56°84'N 104°18'E	
11FEB07	FBS	57°33'N 104°02'E	
28FEB07	FBS	56°84'N 103°64'E	



Forest Clear-cut separability



Class signatures basing on image objects including standard deviation and min/max: brown = clear cut (HH), green = forest (HH), X-axis labels test cases

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Forest Clear-cut separability

date	mode	position		separability: pixel / object
19MAY06	FBS	54°12'N 99°94'E	0.97	1.00
19MAY06	FBS	55°59'N 99°58'E	0.99	1.00
19MAY06	FBS	56°08'N 99°46'E	0.99	1.00
14AUG06	FBS	54°12'N 101°56'E	0.99	1.00
14AUG06	FBS	54°61'N 101°44'E	0.93	1.00
27DEC06	FBS	56°84'N 104°16'E	0.94	1.00
27DEC06	FBS	57°33'N 103°99'E	0.93	1.00
13JAN07	FBS	56°83'N 103°62'E	0.97	1.00
13JAN07	FBS	56°83'N 103°62'E	0.94	1.00
11FEB07	FBS	56°84'N 104°18'E	0.95	1.00
11FEB07	FBS	57°33'N 104°02'E	0.93	1.00
28FEB07	FBS	56°84'N 103°64'E	0.96	1.00



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Analysis of PALSAR data - PLR



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Analysis of PALSAR data - PLR

date	mode	position	
28AUG06	PLR	56°93'N 99°96'E	
28AUG06	PLR	57°42'N 99°78'E	
14SEP06	PLR	56°44'N 99°63'E	
14SEP06	PLR	54°12'N 101°56'E	
13OCT06	PLR	57°41'N 99°75'E	
17MAR07	PLR	56°45'N 99°67'E	
17MAR07	PLR	57°42'N 99°25'E	



Analysis of PALSAR data - PLR



Class signatures basing on image objects including standard deviation and min/max: brown = clear cut (HV), green = forest (HV), X-axis labels test cases

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Analysis of PALSAR data - PLR

date	mode	position	separability: pixel/object
28AUG06	PLR	56°93'N 99°96'E	0.50 (HH) 1.00 (HH) 0.88 (HV) 1.00 (HV) 0.53 (VV) 1.00 (VV)
28AUG06	PLR	57°42'N 99°78'E	0.51 (HH) 1.00 (HH) 0.93 (HV) 1.00 (HV) 0.43 (VV) 1.00 (VV)
14SEP06	PLR	56°44'N 99°63'E	0.64 (HH) 0.86 (HH) 0.85 (HV) 1.00 (HV) 0.59 (VV) 0.82 (VV)
14SEP06	PLR	54°12'N 101°56'E	0.75 (HH) 1.00 (HH) 0.94 (HV) 1.00 (HV) 0.75 (VV) 1.00 (VV)
13OCT06	PLR	57°41'N 99°75'E	0.65 (HH) 1.00 (HH) 0.99 (HV) 1.00 (HV) 0.39 (VV) 1.00 (VV)
17MAR07	PLR	56°45'N 99°67'E	0.31 (HH) 0.92 (HH) 0.74 (HV) 1.00 (HV) 0.32 (VV) 0.92 (VV)
17MAR07	PLR	57°42'N 99°25'E	0.27 (HH) 0.83 (HH) 0.71 (HV) 1.00 (HV) 0.24 (VV) 0.81 (VV)

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Analysis of PALSAR data – FBS Coherence (Winter)



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Analysis of PALSAR data – FBS Coherence (Winter)

date	mode	position	
27DEC06 11FEB07	FBS Coh.	56°84'N 104°16'E	
27DEC06 11FEB07	FBS Coh.	57°33'N 103°99'E	
13JAN07 28FEB07	FBS Coh.	56°84'N 103°62'E	
13JAN07 28FEB07	FBS Coh.	57°33'N 103°45'E	
01JAN07 16FEB07	FBS Coh.	56°35'N 102°69'E	
01JAN07 16FEB07	FBS Coh.	56°84'N 102°54'E	







Class signatures basing on image objects including standard deviation and min/max: brown = clear cut (coherence), green = forest (coherence), X-axis labels test cases



Analysis of PALSAR data - FBS Coherence (Winter)

date	mode	position		separability: pixel / object
27DEC06 11FEB07	FBS Coh.	56°84'N 104°16'E	0.99	1.00
27DEC06 11FEB07	FBS Coh.	57°33'N 103°99'E	0.99	1.00
13JAN07 28FEB07	FBS Coh.	56°84'N 103°62'E	0.98	1.00
13JAN07 28FEB07	FBS Coh.	57°33'N 103°45'E	0.98	1.00
01JAN07 16FEB07	FBS Coh.	56°35'N 102°69'E	0.98	1.00
01JAN07 16FEB07	FBS Coh.	56°84'N 102°54'E	0.99	1.00

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Object based signatures: forest, burnt/clear-cut

- Summer intensity seems slightly better suited than winter intensity
- (Relatively poor separability basing on PLR intensity is owing to the higher noise and speckle effect and to the reduced resolution

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01



Value of Coherence



HV / HH / Coherence

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Value of Coherence



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Forest Cover Mapping Using Intensity and Coherence

- This initial investigation was carried out in the framework of GSE Forest Monitoring
- Summer intensity and winter coherence images are used
- Intensities (FBD HH/HV) have been acquired during summer 2007 (K&C intensity stripes)
- For coherence estimation standard level 1.1 FBS scenes were applied
- 43 pairs have been acquired during winters 2006/2007 (cycles 8 & 9) and 2007/2008 (cycles 16 & 17)
- Each pair stems from consecutive cycles (46 days temporal baseline)
- During both winters suited weather conditions have been reported



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Forest Cover Mapping Using Intensity and Coherence





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Test area (light green patch, right image) in the centre of the prototype area





- · Classification is based on image segments (multiresolution segmentation algorithm)
- Nearest Neighbor algorithm was used
- · Defined target classes: forest, very low biomass forest and non-forest
- · For each class 20 samples have been selected



Example of segmented dataset

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Forest Cover Mapping Using Intensity and Coherence



SAR data (HV/HH/Coherence)

Map (forest: green, very low biomass forest: brownish green, non-forest: light brown)

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Forest Cover Mapping Using Intensity and Coherence

- The accuracy assessment for the whole monitoring area is basing on 1,000 point samples
- · The random sampling was stratified by class proportion
- Overall accuracy: 90.87%.



SAR data (HV/HH/Coherence)

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Map (forest: green, very low biomass forest: brownish green, non-forest: light brown)

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Forest Cover Mapping using Coherence and Backscatter







Is X-band backscatter useful for forest applications?

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100°39'0'E



Validation with TerraSAR-X

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lometers

100°36'0"E

Is X-band backscatter useful for forest applications?





Is X-band backscatter useful for forest applications?

- Method: Stratified Random Sampling Points
- Reference 25 High Resolution Spotlight TerraSAR-X Data randomly spread over the study area
- · Minimum of 5 sampling points per class

	Producers Accuracy	Users Accuracy
	[%]	[%]
Non Forest	92.6	90.9
Forest	95.1	92.3
Sparse Forest	92.6	96.6

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Demonstrating the Potential of ALOS PALSAR Backscatter and INSAR Coherence for Growing Stock Volume Estimation in Central Siberia





Christian Thiel Christiane Schmullius Friedrich-Schiller-University Jena, Germany

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Motivation





Radar backscatter and coherence as function of GSV for the inventory site Hrebtovsky S. The backscatter image (HV) polarisation was acquired at unfrozen conditions, while the data for the coherence image was acquired at frozen conditions.

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Forest Cover Mapping Using Intensity and Coherence

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- The random sampling was stratified by class proportion
- Overall accuracy: 90.87%.



SAR data (HV/HH/Coherence)

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Map (forest, very low biomass forest, non-forest)



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Outline

- 1. Area and test sites
- 2. PALSAR data
- 3. Summary of observations
- 4. Map generation approach
- 5. Results
- 6. Conclusions



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Site Characteristics

- Middle Siberian Plateau: southern part is dominated by hills up to 1700 m, northern part is plain with heights up to 500 m
- Continental climate, prec. 400-450 mm/y, most of the precipitation occurs in summer
- Territory is characterised by large area changes of forests such as forest fire, insect outbreaks, and intensive human activities
- Characteristic taiga forests (birch, pine, fir, aspen, larch, spruce, cedar) cover about 82% of the region





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Location	Chunsky N	Chunsky E	Primorsky	Bolshe	Shesta.	Nizhne	Irbeisky	Hrebt.
2006		30 Dec		28 Dec	—	—		
2007	20 Jun	14 Feb	18 Jan	12 Feb	13 Jan	11 Jan	10 Aug	6 Jan
	5 Aug	2 Jul	5 Mar	15 Aug	28 Feb	26 Feb	10 Nov	21 Feb
	20 Sep	17 Aug	21 Jul	30 Sep	16 Jul	14 Jul	26 Dec	9 Jul
	5 Nov	2 Oct	5 Sep	31 Dec	31 Aug	14 Oct		24 Aug
	21 Dec	17 Nov	21 Oct		16 Oct			9 Oct
2008	5 Feb	2 Jan	21 Jan	15 Feb	16 Jan	29 Feb	10 Feb	9 Jan
	22 Mar	17 Feb		2 Jul	2 Mar	16 Jul	27 Jun	24 Feb
	7 May	4 Jul		17 Aug	17 Apr	31 Aug	12 Aug	11 Jul
	22 Jun	19 Aug			18 Jul		28 Dec	26 Aug
	7 Aug	-			2 Sep			
2009	_	4 Jan		2 Jan	18 Jan	16 Jan	12 Feb	11 Jan
		19 Feb		17 Feb	5 Mar	3 Mar	30 Jun	26 Feb
	(mar.)				21 Jul	()()	15 Aug	14 Jul
					5 Sep		30 Sep	29 Aug
					21 Oct			14 Oct

2000

1600

Forest Forest

400

0 L 0

100

Stands

Number of 008

- PALSAR L-band (1,27 GHz) data
- 87 acquisitions, mode: FBS FBD
- Approx. 300 interferograms
- FBS: HH (28 MHz), FBD; HH/HV (14 MHz)
- Repetition rate: 46 days



XA

Pine 23%

...

500

400

Asper 11%

Birch 28%

200 300 GSV [m³/ha]



- 1. Area and test sites
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1. Area and test sites

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Delineation of GSV Maps

- Random training data selection (20% of the forest inventory data)
- Training of empirical exponential model
- Pixel based model inversion
- Averaging intermediate GSV maps resulting in one backscatter based and in one coherence based GSV map
- Merging coherence and backscatter based GSV map
- Elimination of pixels with a GSV difference > 100 m³/ha (floodplains, change, water, urban etc.)
- Setting all negative GSV values to zero
- Assessing accuracy using the remaining 80% of the reference data



- 1. Area and test sites
- 2. PALSAR data
- 3. Summary of observations
- 4. Map generation approach
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- 6. Conclusions

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Example for delineation of GSV Map (Hrebtovsky site)

Data:

- 3 coherence images (frozen conditions)
- 6 HV backscatter images (unfrozen conditions)
- R² between coherence and GSV: 0.44 (average)
- R² between backscatter and GSV: 0.48 (average)
- Coherence saturation level: 250 m³/ha (average)
- Backscatter saturation level: 200 m³/ha (average)



Example for delineation of GSV Map (Hrebtovsky site)



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99°42'E

99°45'E

99°48'E

99°51'E

99*39'8

99°36'

99°33



Example for delineation of GSV Map (Hrebtovsky site)



Forest stand level based comparison of two SAR data based GSV maps for Hrebtovsky S

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Results for the other sites

	Chunsky E	Chunsky N	Shesta	Hrebt S	Nishni
R ² coh + int	0.79	0.79	0.54	0.57	0.83
R² coh	0.80	0.78	0.37	0.55	0.82
R² int	0.67	0.70	0.56	0.50	0.82
RMSE [m³/ha] coh + int	56.6	41.2	50.4	57.4	48.9
RMSE [m³/ha] coh	56.4	42.4	52.7	61.9	50.7
RMSE [m³/ha] int	71.1	50.3	56.2	59.1	56.1

Rel. RMSE approximately 25% for all sites



2. PALSAR data

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Conclusions

- Coherence at frozen conditions offers the largest potential for GSV estimation
 - Saturation at 230 m³/ha, R² between coherence and GSV is 0.58
 - Comparable results were found in other studies using ERS-1/2 Tandem data
- Backscatter less sensitive
 - Saturation at 75-100 m³/ha, R² between backscatter and GSV 0.42 (HH) 0.48 (HV)



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Conclusions

- Coherence at frozen conditions offers the largest potential for GSV estimation
 - Saturation at 230 m³/ha, R² between coherence and GSV is 0.58
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 - Saturation at 75-100 m³/ha, R² between backscatter and GSV 0.42 (HH) 0.48 (HV)
- Combination of backscatter and coherence led to improvement of GSV estimation, in particular exclusion of areas with contradictory GSV (coherence vs. backscatter) helpful

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Conclusions

- Coherence at frozen conditions offers the largest potential for GSV estimation
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- Combination of backscatter and coherence led to improvement of GSV estimation, in particular exclusion of areas with contradictory GSV (coherence vs. backscatter) helpful
- Demonstrated: Potential of ALOS PALSAR to map the *GSV* of the Siberian forest with a precision close to the accuracy of the conventional forest inventory data (relative RMSE approx. 25%)
- Data availability: At each region in Siberia in average 4 coherence images (temporal baseline 46 days) acquired at frozen conditions and 6 FBD backscatter images acquired at unfrozen conditions are available



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 SAR Techniques: Polarimetry

 1 Investigation backscatter at different polarisations

 2 Computation of polarimetric parameters





Pauli – Decomposition	S _{HH} – S _{VV}	Double Bounce
	23 _{HV}	Volume Scattering



Analysis of Polarimetric Parameters

- 1. Intensities
- 2. Polarimetric HHVV Coherence
- 3. Cloude decomposition parameters
- 4. Freeman decomposition parameters
- 5. Krogager decomposition parameters
- 6. Summary of separability measures

Class signature analysis



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Intensities - Overview



- LEFT: Summer conditions (28th August 2006)
- RIGHT: Autumn/Early winter conditions – beginning of freezing, leaves off (13th October 2006)

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winter summer -9 sigma nought HH [dB] Ī -10 -11 -12 ļ -13 -14 0 1 2 3 4 5 6 7 8 9 summer winter sigma nought HV [dB] -14 ∎ -16 -18 Ŧ Ŧ -20 Ŧ ē -22 -24 0 1 2 3 4 5 7 8 9 6 Signature plot of HV & HH intensity 1 & 6 = recent clear-cut 2 & 7 = former clear-cut 3 & 8 = fire scar 4 & 9 = forest Intensities

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- Displayed: Magnitude of HHVV Coherence
- Provides information on the scattering process
- Surface scattering creates high coherence, multiple scattering low values

 $\rho_{HHITT} = \left\langle \frac{S_{HH}S_{IT}^{*}}{\sqrt{\left\langle S_{HH}S_{HH}^{*}\right\rangle \left\langle S_{IT}S_{IT}^{*}\right\rangle }} \right\rangle$

HHVV Coh.

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winter summer 0,8 pol. coherence [mag] 0,7 Ī Ī 0,6 0,5 0,4 0,3 0 1 2 3 5 6 7 8 9 4 1 & 6 = recent clear-cut 2 & 7 = former clear-cut 3 & 8 = fire scar 4 & 9 = forest HHVV Coh.


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- Roll invariant Eigenvector-Eigenvalue based decomposition of the coherency matrix
- Physical interpretability of concluding parameters
- Alpha indicates type of mean scattering mechanism
- Entropy and Anisotropy specify distribution of the scattering mechanisms
- · Displayed: Entropy

Cloude

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summer winter 1,0 0,9 0,8 entropy 0,7 ē 0,6 0,5 0 2 3 4 5 6 8 9 1 7 summer winter 50 45 Ŧ 40 alpha 35 Ŧ 30 25 20 0 2 7 8 9 1 5 6 3 4 1 & 6 = recent clear-cut 2 & 7 = former clear-cut 3 & 8 = fire scar 4 & 9 = forest Cloude

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- Separates backscattered power with a modelled covariance matrix into three fractions:
 Volume scattering (Pv), double bounce (Pd) and surface scattering (Ps)
- Not roll invariant and topography can affect the fractioning

Freeman

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summer winter 1000 400 Ŧ freeman [vol] 800 300 ē Ŧ 200 600 • • 100 400 200 0 0 1 2 7 8 9 3 6 4 5 Signature plot of Pv (volume scattering) 1 & 6 = recent clear-cut 2 & 7 = former clear-cut3 & 8 = fire scar4 & 9 = forest Freeman ESA PECS SAR Remote Sensing Course, July 2017, Vilnius

[•] Displayed: Pd/ Pv/ Ps



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- Coherent decomposition
- Factorises the scattering matrix as combination of three responses: sphere, helix and diplane
- Power scattered by each of these responses is given by |ks|², |kh|² and |kd|²
- Displayed: |kd|²

Krogager

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summer winter 24 11 23 22 Ŧ 10 krogager [kd] 9 21 Ŧ 8 ₫ 20 7 19 ¢ 6 18 5 9 0 1 2 3 4 5 6 7 8 Signature plot of |kd|² (diplane response) 1 & 6 = recent clear-cut 2 & 7 = former clear-cut 3 & 8 = fire scar 4 & 9 = forest Krogager

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Summary of separability measures

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Summary of separability measures

	1 - 2	1-3	1 -4	2 - 3	2 - 4	3-4
σ^0 HH	0,34	0,20	0,40	0,23	0,08	0,29
$\sigma^0 HV$	0,49	0,45	0,91	0,07	0,69	0,74
$\sigma^0 VV$	0,32	0,13	0,41	0,32	0,11	0,42
$ ho_{HHVV}$	0,20	0,44	0,78	0,28	0,72	0,54
Alpha	0,27	0,57	0,91	0,38	0,88	0,72
Entropy	0,32	0,58	0,89	0,35	0,88	0,80
Pv	0,71	0,65	0,99	0,15	0,91	0,95
kd ²	0,72	0,70	0,99	0,13	0,90	0,95

1 = recent clear-cut, 2 = former clear-cut

3 = fire scar, 4 = forest

Normalised Jefferies-Matusita distance

(1.0 = signatures separable; 0.0 = signatures inseparable)





Summary of separability measures

	1 - 2	1-3	1 -4	2 - 3	2 - 4	3-4
$\sigma^0 \mathrm{HH}$	0,34	0,20	0,40	0,23	0,08	0,29
$\sigma^0 \mathrm{HV}$	0,49	0,45	0,91	0,07	0,69	0,74
$\sigma^0 VV$	0,32	0,13	0,41	0,32	0,11	0,42
$ ho_{HHVV}$	0,20	0,44	0,78	0,28	0,72	0,54
Alpha	0,27	0,57	0,91	0,38	0,88	0,72
Entropy	0,32	0,58	0,89	0,35	0,88	0,80
Pv	0,71	0,65	0,99	0,15	0,91	0,95
$ \mathbf{kd} ^2$	0,72	0,70	0,99	0,13	0,90	0,95

1 = recent clear-cut, 2 = former clear-cut 3 = fire scar, 4 = forest Normalised Jefferies-Matusita distance (1.0 = signatures separable; 0.0 = signatures inseparable)

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Vegetation height estimation from SRTM



J. Kellndorfer et al. / Remote Sensing of Environment 93 (2004) 339-358

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Vegetation (and building) height estimation from E-SAR Data



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Trees acquired at superhigh resolution (X-band)



Andreas R. Brenner and Ludwig Roessing, Radar Imaging of Urban Areas by Means of Very High-Resolution SAR and Interferometric SAR, IEEE TRANSACTIONS ON GEOSCIENCE AND REMOTE SENSING, VOL. 46, NO. 10, OCTOBER 2008

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Trees acquired at high resolution (L-band)



Forest Edge in L-Band

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Seasonal behaviour of C-Band Backscatter in Siberian Forests



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Seasonal behaviour of C-Band Backscatter in Siberian Forests



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Eastern & Western Scene – combined signature plot:

Mean backscatter for forest and non-forest (signatures merged from previous 3 forest and 2 non-forest classes)

Bars denote min and max respectively

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Separability analysis

	burnt/clear-cut vs. forest
14.02.2006	0,38
05.03.2006	0,49
21.03.2006	0,34
25.04.2006	0,78
14.05.2006	0,23
18.06.2006	0,11
04.07.2006	0,36
23.07.2006	0,24
27.08.2006	0,11
12.09.2006	0,46
05.11.2006	0,38
21.11.2006	0,27

Eastern & Western Scene: Normalised Jefferies-Matusita distances

Separability analysis performed on pixel level

1.0 = signatures separable 0.0 = signatures inseparable

Mean separability for forest and non-forest (signatures merged from previous 3 forest and 2 non-forest classes)

Best overall separability: 25th April



Separability analysis

- High separability of forest/non-forest at late April / early May is also evident for other scenes (were no complete time series was available) – next slide
- Where available, late April / early May scenes were utilised for map production, if not available less suited acquisition dates had to be applied









C-band scattering processes in late April



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Background

The boreal zone (in particular Siberia) is characterised by unique environmental conditions

Winter:

- Trees are frozen, almost transparent, backscatter significantly reduced, environmental conditions are very stable
- · Snow hardly impacts the scattering
- · Soil is also frozen, changes in soil moisture do not appear
- · Very low temporal decorrelation, great potential for forest biomass estimation

Thawing "season":

- Wet snow cover
- High level of heterogeneity in space and time (snow cover, moisture, state of forest)
- · Most unsuitable time

Summer:

- · Temporal decorrelation (rainfall, changing soil moisture and interception water, wind)
- Repeat pass coherence for forest is assumed being in general much smaller compared to mid-winter
- · However, not much is known about L-band mid-summer coherence (some work by Eriksson)

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104°E



Site Characteristics

- Central Siberia in Russia (Irkutsk Oblast, Krasnoyarsk Kray)
- Middle Siberian Plateau: southern part is dominated by hills up to 1700 m, northern part is plain with heights up to 500 m
- Characteristic taiga forests (spruce, birch, larch, pine, aspen etc.) cover about 82% of the region
- Territory is characterised by large area changes of forests such as forest fire, and intensive human activities
 Set 99'E 99'E 99'E 100'E 100'E





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Methodology of Investigation - Overview

- 1. Generation of subsets from original frames covering forest inventory data
- 2. Computation of mean coherence per forest stand new entity: forest stand
- 3. Computation of various statistical parameters
- 4. Fit of empirical exponential model (compare Askne & Santoro, 2005)
- 5. Creation of plots: stem volume vs. coherence
- 6. Check of perpendicular baseline \rightarrow rejection of coherence data with baseline > $\frac{1}{2}$ of critical baseline
- 7. Check of weather conditions

$$\gamma_{vol} = ae^{\frac{-vol}{c}} + b\left(1 - e^{\frac{-vol}{c}}\right)$$



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PALSAR Data

Location	Chunsky N	Chunsky E	Primorsky	Bolshe Murtinsky	Shesta- kovsky	Nizhne Udinsky	Irbeisky	Hrebtovsky
Track	T475	T473	T466	T481	T0463	T0471	T0478	T0468
Frame	F1150	F1150	F1110	F1140	F1130	F1100	F1100	F1190
Date		30 Dec 06	18 Jan 07	28 Dec 06	13 Jan 07	11 Jan 07		6 Jan 07
		14 Feb 07	5 Mar 07	12 Feb 07	28 Feb 07	26 Feb 07		21 Feb 07
	20 Jun 07	2 Jul 07	21 Jul 07	15 Aug 07	16 Jul 07	14 Jul 07		9 Jul 07
	5 Aug 07	17 Aug 07	5 Sep 07	30 Sep 07	31 Aug 07		10 Aug 07	24 Aug 07
	20 Sep 07	2 Oct 07	21 Oct 07		16 Oct 07	14 Oct 07		9 Oct 07
		17 Nov 07			16 Jan 08		10 Nov 07	9 Jan 08
	5 Nov 07				2 Mar 08	29 Feb 08	26 Dec 07	24 Feb 08
	21 Dec 07			31 Dec 07	17 Apr 08		10 Feb 08	11 Jul 08
	5 Feb 08	2 Jan 08	21 Jan 08	15 Feb 08	18 Jul 08	16 Jul 08	27 Jun 08	26 Aug 08
	22 Mar 08	17 Feb 08			2 Sep 08	31 Aug 08	12 Aug 08	
	7 May 08				18 Jan 09	16 Jan 09	28 Dec 08	11 Jan 09
	22 Jun 08	4 Jul 08		2 Jul 08	5 Mar 09	3 Mar 09	12 Feb 09	26 Feb 09
	7 Aug 08	19 Aug 08		17 Aug 08	21 Jul 09		30 Jun 09	14 Jul 09
		4 Jan 09		2 Jan 09	5 Sep 09		15 Aug 09	29 Aug 09
		19 Feb 09		17 Feb 09	21 Oct 09		30 Sep 09	14 Oct 09



Coherence Images - Examples Chunsky N - Winter-Winter (Temporal Baseline 46 d)



05nov07_21dec07

21dec07_05feb08

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Coherence Images - Examples Chunsky N - Winter-Winter (Temporal Baseline 46 d)







05feb08_20jun07

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05nov07_20jun07



Coherence Images – Examples Chunsky N – Winter-Summer



20jun07_05aug07

05aug07_20sep07

0

B₁: 4,060 m (B_c: ~ 6,500 m)

22jun08_07aug08

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Coherence Images – Examples Chunsky N – Summer-Summer (Temp. Baseline 46 d)





100

2007/09/05

2007/10/21

100

200

300

 $B_{t} = 46 \text{ d}$

 $B_{\perp} = 435 \text{ m}$

200

(e) unfrozen, B_⊥ < 1 km

300

2

0.6

0.4

0

R²=0.15

a=0.22 b=0.342 c=77.6933

400

 $B_{\perp} = 52 \text{ m}$

2007/02/28

2009/03/05

*B*_t = 736 d

 $B_{\perp} = 1,291 \text{ m}$

100

100

200

(f) frozen

200

GSV [m³/ha]



200

 $B_1 = 343 \text{ m}$

2007/08/05

2007/12/21

*B*_t = 138 d

 $B_{\perp} = 815 \text{ m}$

1)

0.6

0.

100

100

200

(d) dissimilar conditions

300

300

2

0.6

0.4

R²=0.37

a=0.203 b=0.148 c=61.6791



300

R²=0.37

a=0.399 b=0.203 c=115.178

300





0,3 ESA PECS SAR Remote Sensing Course, July 2017, Vilnius

0,6 0,9 1,2 B₁[km] 1,8



1.0 2.0 3.0 4.0 B [km] 6.0

Summary of all PALSAR Observations







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X-band coherence over the Thuringian Forest



Nicolas Ackermann



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Introduction

- Context:
 - The monitoring of forested areas represents a great challenge in the context of the actual climate change and the development of the wood industry activities.
 - Cosmo-SkyMed, with a constellation of 4 satellites, constitutes a promising instruments for the retrieval of forest biophysical parameters.



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Introduction

- Context:
 - The monitoring of forested areas represents a great challenge in the context of the actual climate change and the development of the wood industry activities.
 - Cosmo-SkyMed, with a constellation of 4 satellites, constitutes a promising instruments for the retrieval of forest biophysical parameters.
- Objectives:
 - Can X-band data be useful for forest biomass assessment?
 - Investigate the X-band backscatter intensity and interferometric coherence.





Test site & data

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X-band InSAR coherence

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Interferometric coherence - Single pass -0 0 B_n=259 m 0.4 CSK image 30oct11 - 31oct11 0.2 - 1 day repeat pass -B_n=296 m 0.0 0 100 200 300 400 500 600 Growing Stock Volume [m3/ha]


Conclusions

- Investigations of the CSK, TSX and TDX backscatter intensity and interferometric coherence have been conducted.
- Conifers and Broadleaves amplitude signal can be separated with CSK HH.
- High temporal decorrelation in X-band repeat pass acquisitions (even with 1 day).
- X-band single pass coherence show potential for estimating biomass

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- 1. Simultaneous occurrence of patches of trees, shrubs and grasses
 - Pronounced seasonal variations (e.g. dry and rainy seasons)
 - ightarrow savannas are very heterogeneous, dynamic and sensitive ecosystems



2.

Fig. 1: Predicted tree-grass ratios across rainfall gradients (SANKARAN et al. 2004: 482)



- 3. Status of savannas and their temporal dynamics (e.g. vegetation height, woody cover, AGB)
- 4. Woody cover affects the carbon and water cycles, fire regimes, nutrient cycling and soil erosion

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(Weather stations acronyms: NHL: Nhlanguleni; SKZ: Skukuza; TAL: Talamati; TSH: Tshokwane)

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125 m

200 m

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25 m



Results/Discussion



Fig. 4: Mean R² between PALSAR HH backscatter intensity and LiDAR-based woody cover for three seasons at four different aggregation levels (DRY dry season; EWET end of rainy season; MWET middle of rainy season)

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Fig. 5: Comparison between the PALSAR-based woody cover (left) and LiDAR-based woody cover (right) for the test sites L1 and L2

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Results/Discussion



Fig. 6: SAR-based prediction of woody cover plotted again LiDAR-based observed woody cover. Red line is the regression line, and dotted line is the 1:1 line

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