Freeze/thaw cycles and rain-on-snow

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Structure of this lecture

- 1. What is freeze/thaw and why are we interested in it?
- 2. Sensors, data and algorithms
- Applications with focus on events (esp. rain-on-snow)



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What is freeze/thaw and why are we interested in it?

- Temperature change which results in change of physical properties
- Assumption presence of water
- High dynamics diurnal variations common, minimum daily sampling required
- Monitoring of landsurface
 - Changes type throughout the season
 - We need to consider snow and/or soils



What is freeze/thaw and why are we interested in it?

- Temperature change which results in change of physical properties
- Assumption presence of water
- High dynamics diurnal variations common, minimum daily sampling required
- Relevant for e.g.
 - Soil processes, e.g. microbial activity \rightarrow carbon cycle
 - Snow properties of relevance for subsoil temperatures \rightarrow permafrost
 - Changes in snow structure \rightarrow wild life



Where does freeze/thaw occur?

 Often freeze/thaw is associated with permafrost – actually only characterizes the state of the top of the active layer, by definition not permafrost



Number of unfrozen days, Kim et al. 2011



Which sensors are used?

- Thermal: Landsurface temperature records require gap filling
- Microwave range primarily used
 - No problem with clouds & illumination
 - Sensitivity to freeze/thaw
 - High sampling rate with coarse spatial resolution
- Scatterometer (C-band, 5.3; Ku-band 13.4 GHz), Cband SAR experimentally
- Radiometer (1.4, 19, 37 GHz)



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Example Metop ASCAT 5.3 GHz, since 2007 (Metop A, B ...)







Daily coverage



Example Seawinds on QuikScat

13.4 GHz, 1999 - 2009



Perry 2000

Distribution of footprints and their time stamp 67.8 67.7 67.6 Latitude 67.5 67.4 2001-05-22 PM 10 km 67.3 86.0 86.4 86.8 87.2 Longitude

Bartsch, A., Kidd, R., Wagner, W. and Z. Bartalis (2007): Temporal and spatial variability of the beginning and end of daily spring freeze/thaw cycles derived from scatterometer data. Remote Sensing of Environment, 106(3), 360-374

Daily coverage





Example SSM/I 19, 37 GHz, 1987 -

Schlüssel (1996)



Figure 2.2 Daily coverage of the Earth by the Special Sensor Microwave/Imager.

extension to 1979 possible with SSMR

Algorithms transferable to AMSRE and AMSR2

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Example SMOS and SMAP with	1.4GHz
SMOS	SMAP
Soil Moisture and Ocean Salinity	Soil Moisture Active and Passive
MIRAS (Microwave Imaging Radiometer with Aperture Synthesis)	
ESA	NASA
Launched on November 2nd 2009	Launched on January 31, 2015

V and H polarization resolution ~40 km Global coverage in 3 days



Physical basis



also need to consider snow

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Source: Ulaby & Long (2015)



Snow interactions - radar

- Shorter wavelengths such as Ku-band interact with the snow particels
- Longer wavelengths penetrate dry snow and interact with the soil
- Interaction with ice layer at C-band

Dry snow, e.g.



Snow structure and wavelength crucial

(arrows do not represent magnitude, just symbolize interaction)



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Example record ASCAT versus QuikScat



Typical backscatter time series (in dB) for C-Band (blue crosses, source: ERS) and Ku-band (grey points, source: QuikScat) at Ust Usa (56.92 E, 65.97 N) for August 2003 to July 2004. Daily air temperature range in C extracted from the WMO512 dataset is shown as yellow bars. Diamonds represent Ku-band diurnal backscatter difference in dB (source: Bartsch 2010)



Methods and parameters of snow products (excluding glacier and sea ice) from spaceborne scatterometers, status 2010

Reference	Sensor	Method	Parameters
Boehnke & Wismann 1996	ERS (C-band)	Location specific summer (July) and winter (February) backscatter	Thaw timing
Frolking et al. 1999	NSCAT	Five day average backscatter and location specific difference from overall mean	Thaw timing
Kimball et al. 2001	NSCAT	Application of similar method as in Boehnke et al. 1996	Thaw timing
Kidd et al. 2003, Bartsch et al. 2007	SeaWinds QuikSCAT	Diurnal differences with respect to noise and multiple thaw periods	Start and end of major thaw period
Kimball et al. 2004a	NSCAT	Extension of Frolking et al. 1999	Start, end, primary thaw date
Kimball et al. 2004b	SeaWinds QuikSCAT	As in Kimball et al. 2004a	Start, end, primary thaw date plus autumn refreeze
Brown et al. 2007	SeaWinds QuikSCAT	Fixed threshold for deviation from winter (February) backscatter level	Thaw timing
Wang et al. 2008	SeaWinds QuikSCAT	Application of method from Frolking et al. 1999 to average evening backscatter with respect to summer mean values (August)	Snow-off date
Bartsch et al. 2010	SeaWinds QuikSCAT	Moving window with fixed threshold	Refreeze events
Naeimi et al. 2012	Metop ASCAT	Edge detection and location specific threshold from inflection point of temperature function	Surface status + - 1 month of average spring/autumn transition

Bartsch, A. 2010: Ten years of SeaWinds on QuikSCAT for snow applications. Remote Sensing 04, 2(4)



C-band time series example



Advanced Scatterometer (ASCAT) backscatter (dB), air temperature (°C) and snow depth (cm) for the World Meteorological Organization (WMO) station Chokurdah Nr. 21946 (70.617, 147.883).

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Helena Bergstedt, Simon Zwieback, Annett Bartsch, Marina Leibman: Dependence of C-Band Backscatter on Ground Temperature, Air Temperature and Snow Depth in Arctic Permafrost Regions. Remote Sensing 01/2018; 10(1). austrian polar research institute

C-band – relationship with ground temperatures



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Helena Bergstedt, Simon Zwieback, Annett Bartsch, Marina Leibman: Dependence of C-Band Backscatter on Ground Temperature, Air Temperature and Snow Depth in Arctic Permafrost Regions. Remote Sensing 01/2018; 10(1).



Example ENVISAT ASAR Global monitoring mode

1km, background mode, C-band





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Sang-Eun Park, Annett Bartsch, Daniel Sabel, Wolfgang Wagner, Vahid Naeimi, Yoshio Yamaguchi: Monitoring freeze/thaw cycles using

ENVISAT ASAR Global Mode. Remote Sensing of Environment 12/2011; 115(12).



Example ENVISAT ASAR Global monitoring mode



Acquired when no other mode requested

Bartsch et al. 2009, Journal of Environmental Management

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A Bartsch, W Wagner, K Scipal, C Pathe, D Sabel, P Wolski: Global monitoring of wetlands — the value of ENVISAT ASAR global mode. Journal of Environmental Management 04/2008; 90(7):2226-33.

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Example ENVISAT ASAR Global monitoring mode

Edge detection – step function, does not allow to detect single freeze/thaw events





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101-

Sang-Eun Park, Annett Bartsch, Daniel Sabel, Wolfgang Wagner, Vahid Naeimi, Yoshio Yamaguchi: M<u>onitoring freeze/thaw cycles using</u> ENVISAT ASAR Global Mode. Remote Sensing of Environment 12/2011; 115(12).

Step function applied to Metop ASCAT – transition timing



Thaw

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Naeimi, V., Paulik, C., Bartsch, A., Wagner, W., Kidd, R. Park, S.-E., Elger, K., Boike, J. 2012: ASCAT Surface State Flag (SSF): Extracting Information on Surface Freeze/Thaw Conditions from Backscatter Data Using an Empirical Threshold-Analysis Algorithm. IEEE TGRS 50(7): 2566

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Freeze/thaw events from Metop ASCAT



Reanalyses temperature data for location specific determination of

logistic function





Wooded Tundra

Russia, Sektyakh

ASCAT 0⁰(40) (dB)

-40

-20 0 20 Temperature (°C)

Fig. 4. ASCAT normalized backscatter versus temperature in a grid point near to Sektyakh meteorological station in Russia. Red curve indicates the best logistic function fitted to the measurements. Blue line shows the inflection point of the logistic curve which is assumed as backscatter level at freeze/thaw point.

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Naeimi, V., Paulik, C., Bartsch, A., Wagner, W., Kidd, R. Park, S.-E., Elger, K., Boike, J. 2012: ASCAT Surface State Flag (SSF): Extracting Information on Surface Freeze/Thaw Conditions from Backscatter Data Using an Empirical Threshold-Analysis Algorithm. IEEE TGRS 50(7): 2566

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C-band location specific threshold





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Naeimi, V., Paulik, C., Bartsch, A., Wagner, W., Kidd, R. Park, S.-E., Elger, K., Boike, J. 2012: ASCAT Surface State Flag (SSF): Extracting Information on Surface Freeze/Thaw Conditions from Backscatter Data Using an Empirical Threshold-Analysis Algorithm. IEEE TGRS 50(7): 2566









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C. Paulik, A. Bartsch, D. Sabel, W. Wagner, C. Duguay, A. Soliman: Intercomparison of active microwave derived surface status and MODIS land surface temperature at high latitudes. Proceedings of 2012 IEEE International Geoscience & Remote Sensing Symposium; 07/2012





• Daily maps from ASAR GM for 25x25km area



• Dependance on topography and landcover

Images courtesy of H. Bergstedt



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• Tundra



Figure 10. Percentage of thawed SAR pixel within one ASCAT grid cell during transitional periods (indicated by the DOY), including DOY of the last surface state change in the ASCAT product (red symbols) for the selected grid cells with the assigned landscape type Tundra (Laptev Sea 1, Laptev Sea 2, Laptev Sea 3, Laptev Sea 4, Laptev Sea 5), for the ASCAT [42] and SAR [42] datasets.



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Helena Bergstedt, Annett Bartsch: Surface State across Scales; Temporal and Spatial Patterns in Land Surface Freeze/Thaw Dynamics. Geosciences (Switzerland) 08/2017; 7(3):65.



Lake dominated tundra



Figure 9. Percentage of thawed SAR pixel within one ASCAT grid cell during transitional periods (indicated by the DOY), including DOY of the last surface state change in the ASCAT product (red symbols) for the selected grid cells with the assigned landscape type Lake dominated tundra (Mackenzie 1, Mackenzie 2, Mackenzie 3, Mackenzie 4), for the ASCAT [42] and SAR [42] datasets.



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101-

Helena Bergstedt, Annett Bartsch: Surface State across Scales; Temporal and Spatial Patterns in Land Surface Freeze/Thaw Dynamics. Geosciences (Switzerland) 08/2017; 7(3):65.

• Forest



Figure 11. Percentage of thawed SAR pixel within one ASCAT grid cell during transitional periods (indicated by the DOY), including DOY of the last surface state change in the ASCAT product (red symbols) for the selected grid cells with the assigned landscape type Forest (Laptev Sea 6, Laptev Sea 7, Laptev Sea 8, Laptev Sea 9), for the ASCAT [42] and SAR [42] datasets.



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Helena Bergstedt, Annett Bartsch: Surface State across Scales; Temporal and Spatial Patterns in Land Surface Freeze/Thaw Dynamics. Geosciences (Switzerland) 08/2017; 7(3):65.

Summary radar

- Allows investigations with higher spatial resolution using SAR, but data availability is the major constrain
- but good SAR records available for C-band
- Continuity currently only ensured for C-band
- High potential for Ku-band, specifically for changes in the snow pack
- Simple threshold methods are usually not sufficient
- Threshold needs to be defined using temperature data from in situ or reanalyses



SSM/I and SSMR

Zhang & Armstrong (2001)

- SSM/I 19-GHz and 37-GHz vertically-polarized TBs discriminate frozen ground from unfrozen ground over prairie soils, using the following equation: (TB(37V) - TB(19V))/18 < 0 and TB(37V) < 258.2 K
- A similar approach is used with the SMMR data using the 18-Ghz and 37-Ghz channels.

Kim et al. (2011): seasonal threshold approach using $T_B(37V)$ at pm, also applied to AMSR-E and AMS2

$$\Delta T_{bp}(x,t) = \frac{(T_{bp}(x,t) - FrozRef(x))}{(ThawRef(x) - FrozRef(x))} \qquad \Delta T_{bp}(x,t) > T(x,t)$$
$$\Delta T_{bp}(x,t) < T(x,t)$$



slide courtesy of K. Rautiainen (FMI)



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K.Rautiainen, T. Parkkinen, J.Lemmetyinen, M.Schwank, A.Wiesmann, J. Ikonen, C.Derksen, S.Davydov, A. Davydova, J. Boike, M.Langer, M. Drusch, J.Pulliainen (2016): SMOS prototype algorithm for detecting autumn soil freezing. Remote Sensing of Environment 180 (2016) 346–360



slide courtesy of K. Rautiainen (FMI)



Soil freezing has following effects to L-band passive signal: Increased and saturated observed brightness temperature T_B Decrease in polarization difference: $T_B^V - T_B^H$

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K.Rautiainen, T. Parkkinen, J.Lemmetyinen, M.Schwank, A.Wiesmann, J. Ikonen, C.Derksen, S.Davydov, A. Davydova, J. Boike, M.Langer, M. Drusch, J.Pulliainen (2016): SMOS prototype algorithm for detecting autumn soil freezing. Remote Sensing of Environment 180 (2016) 346–360

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slide courtesy of K. Rautiainen (FMI)

• Observed T_B values converted to $FF_{NPR} = \frac{T_B^V - T_B^H}{T_B^V + T_B^H}$

- NPR Normalized polarization ratio
- Pixel-wise freeze and thaw references determined using all available data (one freeze and one thaw reference per pixel)



slide courtesy of K. Rautiainen (FMI)

Frost depth observation network operated by Finnish

Environmental Institute (SYKE)

Pixel-wise FF values compared against observed frost depths.

Exponential fit => thresholds defined



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K.Rautiainen, T. Parkkinen, J.Lemmetyinen, M.Schwank, A.Wiesmann, J. Ikonen, C.Derksen, S.Davydov, A. Davydova, J. Boike, M.Langer, M. Drusch, J.Pulliainen (2016): SMOS prototype algorithm for detecting autumn soil freezing. Remote Sensing of Environment 180 (2016) 346–36



L Band: SMAP

• The algorithm uses the normalized polarization ratio (NPR) of SMAP radiometer measurements defined by:

```
NPR = (TBV-TBH)/(TBV+TBH)
```

- A seasonal scale factor D(t) is defined for an observation acquired at time t as:
 Dt=(NPR(t)-NPR(fr))/(NPR(th)-NPR(fr))
- where NPR(t) is the normalized polarization ratio calculated at time t, for which a freeze/thaw classification is sought, and NPR(fr) and NPR(th) are normalized polarization ratios corresponding to the frozen and thawed reference states, respectively. Threshold value is 0.5

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C.Derksen, X. Xu, R. S. Dunbar, A. Colliander, Y.Kim, J.S. Kimball, T. A. Black, E.Euskirchen, A. Langlois, M. M. Loranty, P. Marsh, K. Rautiainen, A.Roy, A. Royer, J. Stephens, Retrieving landscape freeze/thaw state from Soil Moisture Active Passive (SMAP) radar and radiometer measurements, Remote Sensing of Environment, Volume 194, 2017, Pages 48-62.



L Band: SMAP

Freeze and thaw references



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C.Derksen, X. Xu, R. S. Dunbar, A. Colliander, Y.Kim, J.S. Kimball, T. A. Black, E.Euskirchen, A. Langlois, M. M. Loranty, P. Marsh, K. Rautiainen, A.Roy, A. Royer, J. Stephens, Retrieving landscape freeze/thaw state from Soil Moisture Active Passive (SMAP) radar and radiometer measurements, Remote Sensing of Environment, Volume 194, 2017, Pages 48-62.



Summary – passive microwave

- Varying approaches
 - Use of different frequencies
 - Use of different polarizations
 - Summer and winter references used
- Threshold needs to be defined using temperature data from in situ or reanalyses



Validation

- Air temperature
- Snow height
- Soil temperatures not that many sites, but e.g. permafrost boreholes
- Need to convert continuous values to yes/no information



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Validation

WMO and GTN-P borehole data (point versus 25km gridded ASCAT)

Agreement with SSF [%]	Temperature	Station
91.79	surface	Nadym R1
90.36	air	Nadym R1
82.75	0.02m below ground	R3 Marre Sale
80.13	0.5m below ground	R3 Marre Sale
80.36	0m surface	R33 Borehole 3
71.90	0.5m below ground	R33 Borehole 3

ASCAT (METOP-A) SSF
Unfrozen
Frozen
Vnfrozen
Koven Unknown
Frecipitaion and SWE (Snow Water Equivalent) data are extracted from GLDAS-NOAH dataset







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Naeimi, V., Paulik, C., Bartsch, A., Wagner, W., Kidd, R. Park, S.-E., Elger, K., Boike, J. 2012: ASCAT Surface State Flag (SSF): Extracting Information on Surface Freeze/Thaw Conditions from Backscatter Data Using an Empirical Threshold-Analysis Algorithm. IEEE TGRS 50(7): 2566

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Applications

- Entire Arctic
- Basins
- Locally





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Application – melting snow

Timing of snow melt from diurnal difference Ku-band



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Bartsch, A. 2010: Ten years of SeaWinds on QuikSCAT for snow applications. Remote Sensing 04, 2(4)



Application – melting snow

Melt area versus river discharge



QuikScat derived daily basin melt area in % of the Lena Kyusyur basin (black solid thick line), Uppler Lena Solyanka basin (solid thin grey line), Aldan basin (thick grey solid line) and river discharge in m3/s at corresponding stations (dashed lines)

Bartsch (2010b)

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A. Bartsch: Monitoring of Terrestrial Hydrology at High Latitudes with Scatterometer Data. Geoscience and Remote Sensing.New

Achievements, 02/2010



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Application – melting snow





Bartsch (2010b)

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A. Bartsch: Monitoring of Terrestrial Hydrology at High Latitudes with Scatterometer Data. Geoscience and Remote Sensing.New

Achievements, 02/2010





Ku-band diurnal cycling

Microbial activity starting when snow starts to melt



Fig. 9.8 Days with significant backscatter changes due to diurnal thaw and refreeze at Zotino (60.75° N, 89.38° E), smoothed (5 days) spring daily accumulated CO_2 fluxes (*solid line*) and mean daily air temperature (*dashed line*) in 2000 (source: TCOS Siberia). Duration of major and final melt period is indicated by *grey shading*. Adapted from Bartsch et al. (2007)



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A. Bartsch, W. Wagner, R. Kidd: Remote Sensing of Spring Snowmelt in Siberia. Environmental Change in Siberia, 01/2010: pages 135-155



Surface state from ASCAT

Globpermafrost WebGIS

www.alobpermafrost.info



Number of frozen days

http://maps.awi.de/map/map.html?cu=globpermafrost_arctic&sb=e&zm=1&ctr=[69.78107758124735,-97.23345073130245]&lyr=["globpermafrostarctic::GTNP WMO Thermal State of Permafrost TSP","globpermafrostarctic::GTNP WMO Circumpolar Active Layer Moinitoring CALM","globpermafrost-arctic::EU-INTERACT Research and Monitoring Stations","globpermafrost-arctic::International Arctic Research and Monitoring Networks","globpermafrost-arctic::Average Number of Frozen Days per Year from Metop ASCAT 2007-2012"]#mapcontent

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ASCAT Surface state and ground temperatures

Number of frozen days Mean Annual Ground temperature



Melting snow as frozen soil

$$R^2 = 0.64$$

Melting snow as thawing soil

$$R^2 = 0.66$$

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Kroisleitner, C., Bartsch, A., and Bergstedt, H.: Potential permafrost distribution and ground temperatures based on surface state obtained from microwave satellite data, The Cryosphere Discuss., https://doi.org/10.5194/tc-2017-162, in review, 2017.

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ASCAT Surface state and ground

Temperatures from lineat fit

Neglects snow and soil properties

Black lines are permafrost zones (source: Brown et al. 1997)





140°W 150°W 170°O 140°O 150 Metop ASCAT FT 2007 - 2012 MEAN N.07 MAGT at coldest sensor depth in °C -14.8 - -10 -9.9 - -5 -4.9 - -2 -1.9 - -1 -0.9 - 0 0.1 - 1 In situ MAGT in °C Permafrost types 1.1 - 2 continuous < -1 2.1 - 5 ---- discontinuous -1-0 5.1 - 10 0 - 1 sporadic 20°0 30°O 10.1 - 19 ······ isolated . >1

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Kroisleitner, C., Bartsch, A., and Bergstedt, H.: Potential permafrost distribution and ground temperatures based on surface state obtained from microwave satellite data, The Cryosphere Discuss., https://doi.org/10.5194/tc-2017-162, in review, 2017.

 Southern Yamal peninsula has been affected by a rain-on-snow (ROS) event in November 2006.

- The ROS event and subsequent refreezing with formation of ice crusts forced a major change in migration.
- Some of the brigades were additionally affected by an event to the west in
 January and as they migrated back northwards across the snowpack, which still consisted of the previous ice layers. The loss amounted to 25% of the animals including deaths and stillbirths resulting from exhaustion and poor nutrition of pregnant females.



Snow profile taken on the 19th of November 2006. (Photo: Florian Stammler)



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Annett Bartsch, Timo Kumpula, Bruce C Forbes, Florian Stammler: Detection of Snow Surface Thawing and Refreezing in the Eurasian

Arctic Using QuikSCAT: limplications for Reindeer Herding. Ecological Applications 12/2010; 20(8):2346-58.

A. Bartsch, et al. 2010

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Arctic Using QuikSCAT: limplications for Reindeer Herding. Ecological Applications 12/2010; 20(8):2346-58.



 Increase of snow depth in a very short time?

Snow structure change

Meteorological records Russia (WMO512)





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Annett Bartsch, Timo Kumpula, Bruce C Forbes, Florian Stammler: Detection of Snow Surface Thawing and Refreezing in the Eurasian

Arctic Using QuikSCAT: limplications for Reindeer Herding. Ecological Applications 12/2010; 20(8):2346-58.

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A. Bartsch, et al. 2010

ASCAT Rain-on-Snow and Sea Ice

• Evidence for autumn atmospheric warming and precipitation increases over Arctic coastal lands in proximity to Barents and Kara sea ice loss.



Source: SSMI derived sea ice extent on NSIDC ESA Cryosphere remote sensing training course 2018



ASCAT backscatter change in dB

Bruce C. Forbes, Timo Kumpula, Nina Meschtyb, Roza Laptander, Marc Macias-Fauria, Pentti Zetterberg, Mariana

Verdonen, Anna Skarin, Kwang-Yul Kim, Linette N. Boisvert, Julienne C. Stroeve,

Annett Bartsch: Sea ice, rain-on-snow and tundra reindeer nomadism in Arctic Russia. Biology letters 11/2016; 12(11):20160466

Change detection, short term increase of backscatter





Bartsch (2010): Ten years of SeaWinds on QuikSCAT for snow applications Remote Sensing.

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QuikScat Snow structure change



Average number of events



Average size of events

Median 470 km²

Wilson, R.R., Bartsch A, Joly, K., Reynolds, J.H., Orlando, A., Loya, W.M. (2013): Frequency, timing, extent, and size of winter thaw-refreeze events in Alaska 2001-

2008 detected by remotely sensed microwave backscatter data. Polar Biology

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QuikScat Snow structure change



K. A. Semmens, J.

Ramage, A. Bartsch, and G. E. Liston (2013). Early

snowmelt events: detection, distribution, and

significance in a major subarctic watershed.

Environmental Research Letters

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Rain-on-Snow



Nov.-Feb. 2000-2009

Bartsch (2010): Ten years of SeaWinds on QuikSCAT for snow applications Remote Sensing.

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The extreme warm spell and ROS events in January–February 2012 had major implications for the society and wildlife in Svalbard.

- (a) Slush avalanches caused closed roads and schools and destroyed a bridge in the major settlement Longyearbyen (photo: Kjersti Strømmen).
- (b) A thick layer of ground-ice built up on roads and airport runways in Longyearbyen (photo: Øystein Varpe) and Ny-Ålesund.
- (c) A wild female reindeer struggles to find food on the ice-encapsulated tundra in Reindalen one week subsequent to the warm spell and ROS (photo: Brage B Hansen).

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Hansen et al. 2014: Warmer and wetter winters: characteristics and implications of an extreme weather event in the High Arctic. ERL

Rain-on-Snow



Nov.-Feb. 2000-2009

Bartsch (2010): Ten years of SeaWinds on QuikSCAT for snow applications Remote Sensing. Wild Svalbard reindeer (Rangifer tarandus platyrhynchus), the Svalbard rock ptarmigan (Lagopus muta hyperborea), and the sibling vole (Microtus levis), and one shared consumer, the arctic fox (Vulpes lagopus).

The community's population fluctuations are mainly driven by rain-on-snow events

Hansen, B.B; V. Grøtan, R. Aanes, B.-E. Sæther, A. Stien, E. Fuglei, R.A. Ims, N.G. Yoccoz, A.Ø. Pedersen. Climate Events Synchronize the Dynamics of a Resident Vertebrate Community in the High Arctic. Science, 18 Jan. 2013



Westermann, S., Boike, J., Langer, M., Schuler, T. V., and Etzelmüller, B.: Modeling the impact of wintertime rain events on the thermal regime of permafrost, The Cryosphere, 5, 945-959, https://doi.org/10.5194/tc-5-945-2011, 2011.

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Summary freeze/thaw

- Monitoring of landsurface
 - Changes type throughout the season: We need to consider snow and/or soils
- Relevant for e.g.
 - Soil processes, e.g. microbial activity \rightarrow carbon cycle
 - Snow properties of relevance for subsoil temperatures \rightarrow permafrost
 - Changes in snow structure \rightarrow wild life



List of references I

Bartsch, A., Kidd, R., Wagner, W. and Z. Bartalis (2007): Temporal and spatial variability of the beginning and end of daily spring freeze/thaw cycles derived from scatterometer data. Remote Sensing of Environment, 106(3), 360-374, doi:10.1016/j.rse.2006.09.004

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