

→ EARTH OBSERVATION SUMMER SCHOOL

Earth System Monitoring & Modelling

30 July–10 August 2018 | ESA–ESRIN | Frascati (Rome) Italy

±三

Sea ice concentration – retrieval and assimilation

Leif Toudal Pedersen, Technical University of Denmark

Technical University of Denmark

*



AMSR2 on JAXA's GCOM-W





AMSR2 Channel Set									
Center Freq.	Band width	Pol.	Beam width	Ground res.	Sampling interval				
GHz	MHz		degree	km	km				
6.925/7.3	350	V/H	1.8	35 x 62	10				
10.65	100		1.2	24 x 42					
18.7	200		0.65	14 x 22					
23.8	400		0.75	15 x 26					
36.5	1000		0.35	7 x 12					
89.0	3000		0.15	3 x 5	5				



Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 2



Signature - Summary



- FY-ice emits almost like a black body
- MY-ice volume scattering reduce emissivity for shorter wavelengths
- Large polaristion difference for water surface

Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 3

European Space Agency 🛀 📰 🔤 🔤 📰 📕 📲 🔤 👘

Example of algorithm principle - Tie-points



$$C_T = \frac{T_B - T_B^W}{T_B^I + T_B^W}$$

where X is the new observation, W is the water signature and I is a point on the FY to MY line (the 100% line)

W, FY and MY are called tie-points



Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 4

Algorithm selection

45





Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 5

OSISAF/SICCI algorithm





European Space Agency 🛀 📰 🔤 🔤 📰 📲 🔤 🚛 🚱 📲 🧮 📲 🖬

Available PMR data and necessary processing steps





CDR	Algorithm / Channels	Instruments	Period	Grid resolution	Project
OSI-450	(19v,37v,37h)	SMMR SSM/I SSMIS	1979-2015	25x25 km	Martin Ma
SICCI2 25.0km	(18v,36v,36h)	AMSR-E AMSR2	2002-2011 2012- 2017	25x25 km	sea ice
SICCI2 50.0km	(06v,36v,36h)	AMSR-E AMSR2	2002-2011 2012- 2017	50x50 km	sea ice cci

Status October 2017:

The OSISAF CDR was released in May 2017: http://osisaf.met.no;

The ESA CCI CDRs are **released:** (http://cci.esa.int/data); March 2017, v2.0, 2002-2015; October 2017, v2.1, extended to 15th May 2017.

Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 8

Polar orbiting satellite - SSMIS & SSM/I daily coverage





Product comparison - 20130315





Apparent concentration of thin ice





Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 12

European Space Agency 🛀 📰 🖛 🛀 📰 📰 📰 European Space Agency

Sea ice concentration and melt ponds



Ice surface fraction is the fraction of the surface which is ice (NOT melt-pond and NOT lead/open water)







Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 14

Melt ponds & thin ice - summary



At the PMW channels we use for SIC retrievals, there is no difference between the emission of sea water (leads) and melt water (ponds). PMR SIC = ISF = 1 - (LeadFraction + MeltPondFraction) and all SIC algorithms underestimate the concentration of thin ice

- Melt ponds will be seen as open water, and cannot be distinguished from leads and other openings
- 100%, thin (<20cm) sea ice will be systematically retrieved as lower concentration of sea ice.
- 100%, thin (>20cm, <50cm) can also be biased low because it is typically more saline, smooth, snowfree.
- EO community cannot "fix" this consistently unless we bring external SIT information, e.g. from models.
- EO community provides operators f(SIC,SIT) to modelers for translating model (SIC,SIT,MPF) to PMR SIC



Sea ice concentration, area and extent



- What the satellites measure is more or less the fraction of the resolution cell covered by ice at the surface.
- Often data are provided on a finer grid than the resolution!
- It is the misinterpretation of melt ponds that has led to the invention of the concept of sea ice extent (all grid cells with a concentration above 15%)
- The 15% threshold was introduced since most sea ice concentration algorithms deploy a socalled weather filter to remove spurious weather induced ice but which on average also removes ice up to 15% (sometimes more).
- Note that sea ice extent calculation will depend on resolution finer resolution will lead to smaller extent – be sure to use the same resolution when comparing extents (Notz, 2014)



Observation operator(s)

Translate model state to observation space using simple models such as

- Melt-pond correction
- Thin ice correction

Using model ice concentration, thickness distribution and melt pond fraction

Curtesy Thomas Lavergne, MET, Norway

https://figshare.com/articles/A_step_back_is_a_move _forward/5501536/1



A step back is a move forward ∂C Agreement (within uncertainty range) radiances Y: value, quantity, uncertainty Satellite Retrievals Obs Oper Obs Ope Obs Ope X=lce Surface Fraction X: what the variable is, its definition Model

European Space Agency *

∂t



Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 17

Observation operator(s)







esa

Simple ice/snow emissivity/TB model





TB as a function of

- f_{MY} Multi-year ice fraction
- T_{ice} (snow surface temperature)
- h_{snow}(snow thickness)
- h_{ice} (ice thickness)
- T_{DN} Downwelling TB (known from atmosphere RTM)

 $T_{ice}, h_{ice}, h_{snow} \rightarrow T(z)$ using fixed thermal conductivities for snow, FY and MY respectively Salinity + T(z) -> brine volume, $V_b(z)$ -> dielectric profile ($\epsilon(z)$) -> $\kappa_a(z)$ -> penetration depth (D_p). Dielectric profile ($\epsilon(z)$) -> $\kappa_a(z)$ -> $\kappa_a(z)$ and later a simple $\kappa_s(z)$ profile.

Reflections only at air/snow and snow/ice interface – Include scattering from surface roughness.

Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 20

Ice emissivity - Simple radiative transfer model



- Perform radiative transfer using standard formulation.
- Incidence angle variation in snow and ice (Snell's law). $n_1^* \sin(\theta_1) = n_2^* \sin(\theta_2)$, where n_1 and n_2 are the refractive indices in medium 1 and 2 respectively. This needs to be done for both air/snow and snow/ice.
- Assume no reflection at other layer boundaries.

$$T_{B} = \sec \theta \int_{-(h_{ice} + h_{snow})}^{0} T(z)\kappa_{a}(z)e^{-\tau(z,0)}dz \quad (+\text{scattering contribution})$$
$$\tau(z_{1}, z_{2}) = \sec \theta \int_{z_{1}}^{z_{2}} (\kappa_{a}(z) + \kappa_{s}(z))dz$$

Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 21

Simple optimal estimation – definitions



p is a vector of

- Total ice concentration,
- MY-fraction,
- Ice temperature,
- Sea Surface Temperature (SST)
- Wind speed,
- Column water vapour
- Cloud liquid water



Measurement (radiometer sensitivity) and model error characterized by covariance S_e

And we have additional information from Climatology (typical values for the state variables (mean p_o and covariance S_p)

Due to the non-linearities in the model, we need to iterate:

$$\mathbf{p}_{n+1} = \mathbf{p}_n + \left(\mathbf{S}_p^{-1} + \mathbf{M}_n^T \mathbf{S}_e^{-1} \mathbf{M}_n\right)^{-1} \left(\mathbf{M}_n^T \mathbf{S}_e^{-1} \left(\mathbf{T}_A - \mathbf{T}_{A,n}\right) + \mathbf{S}_p^{-1} \left(\mathbf{p}_0 - \mathbf{p}_n\right)\right)$$

Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 22



Estimation theory – Measurement noise

 $T_A = Mp + e$

$$\hat{\mathbf{p}} = (\mathbf{M}^{t}\mathbf{M})^{-1} \mathbf{M}^{t} \mathbf{T}_{ap}$$
 (7.6)

$$\mathbf{T}_{\mathbf{A}} = \mathbf{M}\mathbf{p} + \mathbf{e} \tag{7.7}$$

$$\hat{\mathbf{p}} = (\mathbf{M}^{t} \mathbf{S}_{e}^{-1} \mathbf{M})^{-1} \mathbf{M}^{t} \mathbf{S}_{e}^{-1} \mathbf{T}_{A}$$
 (7.8)

$$\hat{\mathbf{S}} = (\mathbf{M}^{t} \mathbf{S}_{e}^{-1} \mathbf{M})^{-1}$$
 (7.9)

$$\hat{\mathbf{p}} = \hat{\mathbf{S}} \left(\mathbf{S}_{\mathbf{p}}^{-1} \mathbf{p}_{0} + \mathbf{M}^{t} \mathbf{S}_{e}^{-1} \mathbf{T}_{A} \right)$$
 (7.13)

 $\hat{\mathbf{S}} = (\mathbf{S}_{p}^{-1} + \mathbf{M}^{t} \mathbf{S}_{e}^{-1} \mathbf{M})^{-1}$ (7.11)

Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 23

European Space Agency 🛀 📰 🖛 🖳 📲 🔚 📲 🚍 👫 📥 🚺 📲 🚍 🖬 👯 🚘 🚺



OE AMSR-E retrieval - February 4, 2006





Runs operationally with AMSR2 data -Results available through www.seaice.dk java sea ice browser



Water Vapour







Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 25

Diagonal elements of S matrix – estimated uncertainty







DTU | 31/07/2018 | Slide 26



Cost function $(T_B(p)-T_Bobs)$



Discrepancies along strong gradients due to mismatching footprints Discrepancies inside Arctic Ocean due to too simple ice forward model

Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 27

Conclusions



Satellite data products are NOT the truth

All SIC algorithms underestimate thin ice All SIC algorithms see melt-ponds as 0% ice Algorithms overestimate SIC in Summer due to difficulty in tie-point definition

Icecharts are NOT the truth either

They typically overestimate intermediate ice concentrations They are not necessarily consistent from day to day

4 reasons why icechart and PMR SIC products differ

- 1. Thin ice is underestimated in PMR SIC
- 2. Melt ponds are seen as open water in PMR SIC (PMR SIC = ISF)
- 3. Wet snow on ice may lead to regional overestimation tie-points are hemispheric
- 4. Ice charts often overestimate intermediate concentrations

However, when used with caution, satellite data (and icecharts) provide a wealth of useful information about sea ice and it's snow cover

Data providers should provide quantitative estimates of known issues – ask them for specifications for observation operators!!

Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 28

•



Ressources

OSISAF operational and arctived Sea Ice products (netCDF & quicklooks) http://osisaf.met.no/p/ice/index.html

ESA CCI Sea Ice CDR (netCDF) (AMSR-E/AMSR2) https://icdc.cen.uni-hamburg.de/1/projekte/esa-cci-sea-ice-ecv0.html

ESA CCI and H2020 SPICES projects Round Robin Data Package

https://figshare.com/articles/Reference_dataset_for_sea_ice_concentration/6626549

Lavergne, T., Sørensen, A. M., Kern, S., Tonboe, R., Notz, D., Aaboe, S., Bell, L., Dybkjær, G., Eastwood, S., Gabarro, C., Heygster, G., Killie, M. A., Kreiner, M. B., Lavelle, J., Saldo, R., Sandven, S., and Pedersen, L. T.: Version 2 of the EUMETSAT OSI SAF and ESA CCI Sea Ice Concentration Climate Data Records, The Cryosphere Discuss., https://doi.org/10.5194/tc-2018-127, in review, 2018.

Ivanova, N., Pedersen, L. T., Tonboe, R. T., Kern, S., Heygster, G., Lavergne, T., Sørensen, A., Saldo, R., Dybkjær, G., Brucker, L., and Shokr, M.: Inter-comparison and evaluation of sea ice algorithms: towards further identification of challenges and optimal approach using passive microwave observations, The Cryosphere, 9, 1797-1817, https://doi.org/10.5194/tc-9-1797-2015, 2015.

Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 29

Sea ice Observations, Modelling and Data Assimilation



Sea Ice Analysis and Forecasting

Towards an Increased Reliance on Automated Prediction Systems

Edited by Tom Carrieres, Mark Buehner, Jean-François Lemieux and Leif Toudal Pedersen



Sea Ice Analysis and Forecasting (Book)

Towards an Increased Reliance on Automated Prediction Systems

Edited by Tom Carrieres, Mark Buehner, Jean-François Lemieux, Leif Toudal Pedersen

This book provides an advanced introduction to the science behind automated prediction systems, focusing on sea ice analysis and forecasting. Starting from basic principles, fundamental concepts in sea ice physics, remote sensing, numerical methods, and statistics are explained at an accessible level. Existing operational automated prediction systems are described and their impacts on information providers and end clients are discussed. The book also provides insight into the likely future development of sea ice services and how they will evolve from mainly manual processes to increasing automation, with a consequent increase in the diversity and information content of new ice products. With contributions from world-leading experts in the fields of sea ice remote sensing, data assimilation, numerical modelling, and verification and operational prediction, this comprehensive reference is ideal for students, sea ice analysts, and researchers, as well as decision-makers and professionals working in the ice service industry.

Cambridge University Press, Fall 2017

Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 30

End of part 2



Thank you 🙂

European Space Agency

÷

Include volume scattering



$$T_{B} = \frac{1}{\cos(\theta)} \int_{-(hice+hsnow)}^{0} T(z)\kappa_{a}(z)e^{-\tau(z,hsnow)}dz + \frac{1}{\cos(\theta)} \int_{-(hice+hsnow)}^{0} T_{DN}(z)\kappa_{s}(z)dz$$

$$\tau(z_{1}, z_{2}) = \frac{1}{\cos(\theta)} \int_{z_{1}}^{z_{2}} (\kappa_{a}(z) + \kappa_{s}(z))dz$$

$$T_{DN}(z) = T_{DN}(0)e^{-\tau(0,z)} + \frac{1}{\cos(\theta)} \int_{z}^{0} T(z)\kappa_{a}(z)e^{-\tau(z,hsnow)}dz$$
Left Touchal Pedersen | DTU | 31/07/2018 | Slide 32

+



Icechart to OSISAF SIC intercomparison

East 100 95 Norwegia Greenland Area 75 Met.no ice concentration [%] 50 See also: Karvonen, Juha, Jouni Vainio, Marika Marnela, Patrick Eriksson, and Tuomas Niskanen, A Comparison Between High-Resolution EO-Based and Ice 20 Analyst-Assigned Sea Ice Concentrations, IEEE JOURNAL OF SELECTED TOPICS IN APPLIED 5 Danish EARTH OBSERVATIONS AND 0 REMOTE SENSING, VOL. 8, NO. 4, **APRIL 2015** 0 10 15 20 30 35 55 60 65 70 85 85 90 95 95 5 40 45

Danmarks Meteorologiske Institut

European Space Agency 🛀 📰 🔤 🔤 🔛 📲 🔤 📰 📰 🔤 👘

DMI regional NE ice concentration [%]

Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 34

Icechart vs OSISAF - SIC intercomparison



Danmarks

Meteorologiske Institut esa

Icechart vs OSI-SAF ice concentration

Sea Ice Concentration (OSI SAF vs NIC) Comparison 2017-07-20



Danmarks Meteorologiske Institut

CIS Icechart vs Sentinel-1 SAR





CIS daily ice chart on July



Sentinel-1 SAR upped July 1201/022011 Zilde 37

*

Icechart to OSISAF SIC intercomparison





DMI Greenland Overview Ice Chart





Met.no Svalbard Ice Chart



Leif Toudal Pedersen | DTU | 31/07/2018 | Slide 38



