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Sea ice data assimilation for NWP and operational ice services at Environment and Climate Change Canada

ECMWF Workshop on observations and analysis of sea-surface temperature and sea ice for NWP and Climate Applications Mark Buehner, Alain Caya, Alex Komarov Meteorological Research Division Lynn Pogson, Michael Ross Canadian Ice Service 22-25 January 2018



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Contents

- Motivation for sea ice data assimilation at ECCC
- Data assimilation methodology
- Recent addition of new observations
- Current research projects



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The need for coupled Atmosphere-**Ice**-Ocean prediction



ECCC requires ice-ocean forecasts and information services for:

- Sea ice prediction
 - Improved automated analyses and forecasts for the Canadian Ice Service - to complement manual ice chart analyses
 - Identify/predict high pressure areas dangerous for ships
- Improved weather and wave prediction
 - Timescales from days to seasons
 - Sea ice, tropical cyclones, surface interactions
- Emergency response
 - Comprehensive trajectory modelling capacity
 - E.g. dispersion of pollutants





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Environnement et Changement climatique Canada Davidson et al., SCOR, 2013

Ice-ocean modelling with

Applications and domains

CICE

1/4° Global

Surface currents

Operational Experimental In development

- Global 1° resolution (CanSIPS-GI
 - Seasonal forecasting
- Global 1/4° resolution (GIOPS)
 - Medium-monthly forecasting
 - Fully-coupled for NWP
- N. Atlantic & Arctic 1/12° (RIOPS)
 - Short-to-medium range forecasting
 - Coupled HRDPS-Polar for YOPP
- Great Lakes 2km and Gulf of St. • Lawrence 5km
 - Short-term forecasting





120⁰E

180⁰W

120°W

60°W

30⁰N

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1/12° N. Atlantic and Arctic

Coupled Global Forecast

Recently became operational

G Smith, J-M Belanger, F Roy, ...

- Coupled model:
 - Atm: GEM 25km
 - Ocean: NEMO-ORCA025 (1/4°)
 - Ice: CICE4
 - Uncoupled DA
- Evaluation of summer trials
 - 10 day forecasts 15 Jun–31 Aug, 2014
 - Significant forecast improvements over most areas
 - Shown: 850hPa geopotential height versus ERA-Interim





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Regional / global ice concentration analyses (Buehner et al. 2016)

- Regional: ~ 5.0 km resolution; Global: ~10 km resolution
- 4 analyses per day
- background = analysis 6 hours earlier
- total ice concentration (3DVar) and analysis-error stddev estimate (KF)
- observations assimilated:
 - CIS image analyses and daily and regional ice charts, lake bulletins
 - SSM/I, SSM/IS, AMSR2
 - ASCAT
 - AVHRR (ice/water)
- ice is removed where SST > 4°C
- ice field is "corrected" where analysiserror stddev is high



1768 × 1618 grid points



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Sea ice data assimilation cycle



Passive microwave data SSMI, SSMIS, AMSR2

- Assimilation:
 - Total ice concentration estimated from NASA Team 2 (NT2) retrieval algorithm
 - Use "footprint" observation operator that aggregates gridded ice lacksquareconcentration over footprint of instrument
- Quality control reject data when:
 - Surface Air Temperature (SAT) > 0°C (melt ponds)
 - Retrieved ice concentration is not zero AND
 - Sea Surface Temperature (SST) is above 4°C <u>OR</u>
 - Historical Frequency of Occurrence of ice is 0 OR
 - Wind speed > 25 knots (Wind filter)
- Background check:
 - Reject entire observation swath with bad/corrupted data (based) on average RMS difference with background state)

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Observation footprints



- Footprint observation operator important for combining information from sensors with such different resolution
- Observation rejected if footprint touches land, removing most low resolution obs near coast and in narrow channels



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Impact of observation quality control Example: July 8, 2007

Without QC



With QC



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Effect of wind filter

Without wind filter

With wind filter



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Analysis error stddev

- Due to QC procedures, may not assimilate data in a particular region for an extended period
- Therefore, also estimate analysis error stddev with simplified KF and simple error growth model
- Where ice is removed due to SST > 4°C, stddev set to zero
- Used when initializing forecasts: only replace model value at grid points with low uncertainty
- Used for forecast verification: only consider grid points with low uncertainty





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Ice / water classification from AVHRR(visible and infrared)We developed a sim



SOURCE: CIS DAILY AND REGIONAL ICE CHARTS / SOURCE: CARTES QUOTIDIENNES ET RÉGIONALES DES GLACES DU SC



Environment and Climate Change Canada Environnement et Changement climatique We developed a simple classification procedure (ice / water / ambiguous / cloud)

Distinguishing cloud from ice is difficult, but critically important

Example: 2011-08-05 16Z Western Arctic

(after thick cloud removed)



Ice concentration from CIS image analysis

Assimilation strategy:

- Water observations assimilated as 0%, Ice assimilated as:
- 85% when background concentration < 85%
- otherwise, rejected
 (background already consistent with obs)



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Comparison of Ice / Water from AVHRR and IMS (NOAA manual ice/water product)



- Few observations in winter (insufficient solar illumination)
- Generally high accuracy (> 99%), but lower in summer

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Ice / water retrieval from RADARSAT-2 Komarov and Buehner, TGRS, 2017

- Assimilation of SAR data could improve ice analyses, particularly in narrow channels – very high resolution, not limited by cloud or solar illumination
- Challenge: Backscatter signal from ice and open water often overlap
- Most conventional automated methods for ice/water classification do not provide sufficiently high level of confidence required for data assimilation

Goal: Develop a technique for automated classification of RADARSAT-2 data (**ice / water / ambiguous**) based on estimation of probability of ice/water



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Key variables related to presence of ice

After some exploratory work, chose the following variables:

- 1. Difference between wind speed from SAR (HH-HV) [Komarov et al., TGRS, 2014] and from NWP
- 2. HH-HV spatial correlation
- 3. Spatial standard deviation of SAR-derived wind speed

Each is computed over a 2km x 2km window (41x41 50m pixels)

Benefits:

- Independent of instrument parameters (incidence angle, noise floor)
- Indirect use of HH and HV backscatter signals
- Good potential to transfer to other SAR instruments (e.g. RCM: RADARSAT Constellation Mission)

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Key variables related to presence of ice Probability distributions based on > 10,000 images



Presence of ice/water is obtained from CIS manual image analyses

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Key variables related to presence of ice Probability of ice as function of 3 variables



3D Ice Probability Model (IPM)

$$P(\vec{x}) = \frac{1}{1 + e^{f(\vec{x})}}$$

Probability of ice computed from 3 predictors using Bayes and logistic regression fit to 5 years of training data (*f* is 3rd order polynomial of 3 predictors)

Probability of ice computed from IPM at each location and classified as follows:

 $\begin{array}{ll} \mathsf{P}(\mathsf{x}) > 0.95 & \textbf{ice} \\ \mathsf{P}(\mathsf{x}) < 0.05 & \textbf{water} \\ \mathsf{Else} & \textbf{ambiguous} \end{array}$

Also used an adaptive probability threshold to slightly improve accuracy

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IPM Testing Results

Using CIS image analyses for verification

Independent testing subset (Year 2013) considering only 0 and 100% concentrations

P_{threshold} – adaptive thresholding

	# samples	Classified [%]	Misclassified [%]	Unknown [%]	Accuracy [%] Nc / (Nc + Nm)
Pure ice	990,638	83.78	0.24	15.98	99.71
Pure water	1,490,240	54.02	0.13	45.86	99.76
Pure ice & water	2,480,878	65.90	0.18	33.92	99.73



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Results with all ice concentration values Fraction of SAR ice/water retrievals as a function of CIS manual image analysis concentration



Fraction of ice/water samples

$$F_i = N_i / (N_i + N_w)$$

$$F_w = 1 - F_i$$

Different scales of IPM model and Image Analysis polygon:

Image Analysis polygon





Verification against IMS

Independent IMS product is available daily, consistent coverage over entire domain of RADARSAT-2 images.

IPM model was applied to all available 7411 RADARSAT-2 images in 2013.

	# samples	Classified [%]	Misclassified [%]	Unknown [%]	Accuracy [%] Nc / (Nc + Nm)
Ice	24,719,977	75.81	2.58	21.61	96.71
Water	12,813,635	55.53	2.90	41.57	95.04
Ice & Water	37,533,612	68.89	2.69	28.43	96.24



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Example comparison with IMS Ice edge dynamics – one of the worst cases



Ice concentration from ice/water retrievals Obtained by counting fraction of ice retrievals

Ice/water retrievals at 2.05 km spacing

Ice concentration based on 3x3 ice/water retrievals



10

20

30



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50

70

80

60

100

90

Sea ice data assimilation

Other ongoing and future work

- Apply approach used for AVHRR also to VIIRS data (similar channels, higher resolution, larger swath width)
- Based on high accuracy of ice/water RADARSAT-2 retrieval, evaluate impact of assimilating in 3DVar for estimating ice concentration
- Currently developing ice thickness analysis (3DVar) based on SMOS+CryoSat-2, together with estimated analysis error stddev (KF)
- Improve passive microwave retrievals by using RTM to account for atmospheric effects and using dynamic tie-points
- Started migrating sea ice and SST analyses into the modular software framework used for atmospheric 4D-EnVar → facilitate strongly coupled DA of atmosphere-ice-upper-ocean (through H and B)



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