Challenges in the Modeling and Parameterization of the Surface Contribution in the Microwaves

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+ Carlos Jimenez, Lise Kilic, Iris de Gelis, Filipe Aires
+ the ISSI group for the Ocean Emissivity model
+ many other collaborators



The next key challenge in Numerical Weather Prediction (NWP): the assimilation of all-surface radiances



That clearly requires an accurate estimation of the surface contribution

- for all surface-sensitive observations
- for all surfaces

Outline of this talk

1) General considerations

2) Ocean

Physically-based models, and their fast version

3) Other surfaces

Physically-based model??

Or emissivity parameterization based on the available surface information and satellite-derived emissivity?

The bright future of passive microwave observations in Europe



= Ice Cloud Imager (MetOp-SG B)

- Extension of the frequency range up to 664 GHz, with ICI
- Simultaneous observations between 1.4 and 36 GHz, with CIMR

For atmospheric characterization, the surface contribution is a source of noise.

Atmospheric 'windows' for surface characterization. The surface contribution is the information.

=> In both cases, the surface contribution has to be quantified!

An accurate estimate of the surface contribution is needed in the microwaves, for all surface types, at global scale

- across frequencies: from low microwaves to millimeter waves. Possibly including the infrared.
- **across observing conditions**: incidence angle, polarization. For both passive instruments (emissivity) and active instruments (backscattering).
- across applications: for NWP, for atmospheric retrieval as well as for the retrieval of surface properties.

Consistency required to optimize the exploitation of multi-frequency, multi-instrument capability, for both atmospheric and surface characterizations

Toward coupled land / ocean / atmosphere assimilation systems

OCEAN

How to accurately estimate the surface contribution in the microwaves at global scale?

Open ocean: a rather homogeneous surface (at least compared to the other surfaces)

=> Robust physically-based radiative transfer models exist.



Microwave sea surface emissivity models

Physically-based models

two-scale models valid from long microwaves to IR

(examples: Yueh, 1997; Dinnat et al., 2003; Yin et al., 2012; Dinnat et al., 2023)

Fast models parameterized from physically-based models (examples: FASTEM, TESSEM², SURFEM-Ocean) distributed with RTTOV or CRTM

Models fitted to satellite observations

(example: Remote Sensing System model, Meissner et al., 2012, 2014)

They all include:

- a sea water dielectric model
- a wind-driven roughness model
- a foam model (extent and emissivity)

An international team was formed, to work on the development of a

Reference Quality Model for Ocean Surface Emissivity and Backscatter

- Physically-based
- From the microwaves to the infrared
- **.** For both active and passive modes

A Reference Quality Model For Ocean Surface Emissivity And Backscatter From The Microwave To The Infrared

INTERNATIONAL SPACE SCIENCE INSTITUTE



https://www.issibern.ch/teams/oceansurfemiss/

English et al., BAMS, 2020; Dinnat et al., BAMS, 2023

- A physically-based reference ocean model was selected: PARMIO (Passive and Active Reference Microwave to Infrared Ocean) (Dinnat et al., 2003; Yin et al., 2012, 2016; Dinnat et al., 2023)
- Extensively evaluated with multiple observations at global scale (SMAP, AMSR, GMI, ATMS) (Kilic et al., JGR, 2019; Kilic et al., ESS, 2023).
- Adjustments made to the initial model to better fit the observations under cold temperatures and for high wind speeds.

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- Adjustments made to the initial model to better fit the observations under cold temperatures and for high wind speeds.
- A fast code derived from this model, with similar inputs as FASTEM, along with Jacobians and error estimates, SURFEM-Ocean, included in RTTOV 13 (Kilic et al., ESS, 2023), successfully tested at ECMWF (Geer et al., 2024) and now activated in the operational cycle.

Selection of the physically-based reference ocean model, its evaluation and its adjustments



A fast model derived from PARMIO, based on NN parameterization.







PARMIO

A Reference Quality Model for Ocean Surface Emissivity and Backscatter from the Microwave to the Infrared

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https://doi.org/10.1175/BAMS-D-23-0023.1

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Earth and Space Science

RESEARCH ARTICLE 10.1029/2022EA002785

· An international team of experts has converged on the selection of a Passive and Active Reference Microwave to Infrared Ocean model · A fast version of the ocean microwave emissivity model has been developed. named SURface Fast Emissivity Model for Ocean (SURFEM-Ocean)

Key Points:

 SURFEM-Ocean is evaluated by comparisons with a large database of passive microwave satellite observations

Development of the SURface Fast Emissivity Model for Ocean (SURFEM-Ocean) Based on the PARMIO Radiative Transfer Model

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Abstract A Passive and Active Reference Microwave to Infrared Ocean (PARMIO) physical radiative

Technical Memo

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SURFEM-ocean microwave surface emissivity evaluated

SURFEM-Ocean reproduces well the PARMIO model



SURFEM-Ocean reproduces well the PARMIO model



Test of SURFEM-Ocean with multiple satellite observations: Biases with observations



(Kilic et al., ESS, 2023)

Test of SURFEM-Ocean with multiple satellite observations: Dependence with ocean parameters



limited error at low SST

(Kilic et al., ESS, 2023)

Test of SURFEM-Ocean with multiple satellite observations: Dependence with incidence angle



ATMS

Conclusion over ocean



- > Rather robust radiative transfer models exist.
- A reference physical model (PARMIO) with maintenance insured over time, available at <u>https://gitlab.com/nwpsaf/parmio.</u>
- > Preliminary testing for infrared emissivity and for active microwaves.
- A corresponding fast emissivity model developed for passive microwaves, based on NN (SURFEM-Ocean).
- > SURFEM-Ocean incorporated into RTTOV 13 and operational at ECMWF now.
- More info at <u>https://www.issibern.ch/teams/oceansurfemiss/</u>

LAND, SNOW, ICE SEA ICE

How to accurately estimate the surface contribution in the microwaves at global scale?

Open ocean: a rather homogeneous surface

=> Robust radiative transfer models exist

Land, snow, ice, sea ice: high heterogeneity and complex interaction with the radiation

=> Radiative transfer modeling very challenging surface reflection and scattering + volume scattering



The challenge of calculating emissivities with radiative transfer models at global scale

CRTM



Satellite-derived

(Prigent et al., JGR, 2015)

Specific microwave emissivity physical models

For snow, ice, sea ice



SMRT (Snow Microwave Radiative Transfer) (Picard et al., GMD, 2018, https://www.smrt-model.science/documentation.htm)

- Can handle several physical scattering assumptions, as well as multiple geophysical conditions.
- Passive and active microwaves
- A user friendly inter-active version available

Physically-based microwave emissivity models are very challenging for global applications, over land, snow, ice, and sea ice

- > Difficulty to capture the high spatial and temporal heterogeneity
- Complex interaction between the signal and the surfaces surface reflection and scattering + volume scattering... at the same time
- > Difficulty to access the necessary input parameters for the model
- Which are the key drivers of the signal variability?
 - \circ Are they included in the model inputs?
 - Are they available at large scale?

Satellite-derived microwave emissivity



Applied to window channels for SSM/I, AMSU, AMSR, GMI.... under clear sky only or imbedded into a full retrieval of the atmosphere and surface

E.g., Prigent et al., 1997, 2006; Aires et al., 2001; Karbou et al., 2005, Boukabara et al., 2018; Munchack et al., 2020...

In operational mode:

- Emissivities are calculated on line in window channels and propagated to other channels
- Or emissivity atlases are used

Satellite-derived microwave emissivity

$$\epsilon_p = \frac{Tb_p - T_{atm}^{\uparrow} - T_{atm}^{\downarrow} \times e^{-\tau(0,H)/\mu}}{e^{-\tau(0,H)/\mu} \times (T_{surf} - T_{atm}^{\downarrow})}$$

Sources of errors:

The surface temperature T_{surf}

- Tsurf=Tskin? Tskin from NWP model? From IR estimates (under clear sky conditions) ?
- Sub-surface contribution? Tsurf=Teff. Depends upon the frequency... (Effective emissivities)
- Clearly, the **dominant error**

The atmospheric contribution

- especially at high frequency
- adjusted when calculation within a full surface / atmosphere inversion model (as in NWP centers)

Specular approximation

• always valid? Is there a need to add a Lambertian contribution close to nadir and at high frequency, especially over snow and ice? (Matzler, GRSL, 2005; Karbou et al., GRSL, 2005; Harlow, IEEE, 2009)

Satellite-derived microwave emissivity

An analysis of emissivities has been derived from multiple satellites, to parameterize the frequency, angle, and polarization dependence of the emissivity for NWP applications (Prigent et al., IEEE, 2008).

TELSEM²

Tool to Estimate Land Surface Emissivities at Microwaves and Millimeter waves (**distributed with RTTOV and CRTM**) (Aires et al., JQSRT, 2011; Wang et al., JAOT, 2017)

- Global atlases of emissivity for all continental and sea-ice surfaces, from 18 to 700 GHz, monthly mean, at 25 km resolution.
- Inputs: lat, lon, month, frequency, and incidence angle.
- Outputs: emissivities in V and H polarizations, along with error covariances
- Realistic FIRST GUESS estimates

Space and time variability of the microwave land surface emissivity



Space and time variability of the microwave land surface emissivity



Satellite-derived surface emissivity provides reasonable spatial and temporal variabilities, as well as frequency co-variabilities, over land, snow, sea ice.

But they do not tell about the key geophysical parameters that drive their variability...

For consistent surface and atmospheric inversion, how to relate the satellite-derived surface contribution to the geophysical parameters?

Given the limitations of the physically-based forward operator over land,

Possibility to derive statistical forward operators anchored to the satellite observations, and consistent for multi-frequency, multi-instrument operation?

Toward coupled land atmosphere assimilation system...

1) Revise and extend the satellite-derived emissivity database

- Satellite-derived emissivity calculated for the full time series of SSM/I, SSMIS, AMSR, SMAP, and SMOS, over the continents and sea ice
 a large emissivity database
- With ERA5 inputs for the atmosphere and the surface temperature (Tskin)

- 1) Revise and extent the satellite-derived emissivity database
- 2) Select potential predictors for the parameterization of the emissivities
 - Avalaible easily over long time series
 - Preferably from reanalysis (ERA5), from well-recognized community models when reanalysis information not enough
 - Not always the expected physical parameters that play the key role... Be ready for surprises...

- 1) Revise and extent the satellite-derived emissivity database
- 2) Select potential predictors for the parameterization of the emissivities
- 3) Statistically relate the satellite-derived emissivities to the relevant geophysical parameters available at global scale
 - Use of machine learning method, to account for complex relationships between geophysical parameters and emissivity
 - Neural Network parameterization suggested

- 1) Revise and extent the satellite-derived emissivity database
- 2) Select potential predictors for the parameterization of the emissivities
- 3) Statistically relate the satellite-derived to the relevant geophysical parameters available at global scale
- 4) Derive a physics-aware statistical parameterization of the surface emissivity
 - as a function of :
 - instrument characteristics (frequency, incidence angle, polarization)
 - geophysical variables from reanalysis, Land Surface Model / Sea Ice Model

- 1) Revise and extent the satellite-derived emissivity database
- 2) Select potential predictors for the parameterization of the emissivities
- 3) Statistically relate the satellite-derived to the relevant geophysical parameters available at global scale
- 4) Derive a physics-aware statistical parameterization of the surface emissivity

 \Rightarrow An example for snow and sea ice emissivity parameterization. See presentation from Iris de Gelis later today.

Conclusion over land, snow, ice, sea ice



- Physically-based radiative transfer land surface models are still very challenging for large scale applications, under multiple instrument conditions and diverse environments.
- Physics-aware ML parameterization of the surface emissivity as a function of :
 - instrument characteristics (frequency, incidence angle, polarization)
 - geophysical variables from reanalysis and from external Land Surface Models or Ocean-Sea ice models, when not enough information available in the reanalysis

Results to be shown for snow and sea ice (Iris de Gélis, later in the day)

Thank you!

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Generic land surface microwave emission models

Community Microwave Emission Model (CMEM) at ECMWF

(Drusch et al., JHM, 2009, de Rosnay et al., RSE, 2020)

Specific work at 1.4 GHz, for SMOS



Modular configuration of CMEM. For each module components, a choice of parameterizations is available. Parameterizations in bold are those used in this paper. Different combinations of CMEM using three different dielectric models, four roughness models and three vegetation optical depths models are compared, leading to 36 configurations evaluated against SMOS observations.

CMEM modules	Choice of parameterizations	
	Short name	Reference
Soil module:		
Dielectric mixing model	Dobson	(Dobson et al., 1985),
	Mironov	Mironov et al. (2004)
	Wang	Wang and Schmugge
		(1980)
Effective temperature model	Surface temperature	
	forcing,	
	Choudhury	Choudhury et al. (1982)
	Wigneron	Wigneron et al. (2001)
	Holmes	Holmes et al. (2006)
Soil roughness model	Choudhury	(Choudhury et al., 1979)
	Wign07	Wigneron et al. (2007)
	Wign01	Wigneron et al. (2001)
	Texture dependent	citepatbd:10
	Wegmüller	Wegmüller and Mätzler
		(1999)
Vegetation module:		
Vegetation optical depth model	Wegmüller	(Wegmüller et al., 1995)
	Jackson	Jackson and O'Neill
		(1990)
	Kirdyashev	Kirdyashev et al. (1979)
	Wigneron	Wigneron et al. (2007)
Snow module:		
Snow emission model	HUT single layer	Pulliainen et al. (1999)
	model	

- Large errors before adjustements at 1.4 GHz (improvement after bias correction)
- Applicable to other frequencies, with consistent hypotheses and inputs?

Test of SURFEM-Ocean with multiple satellite observations: Dependence with ocean parameters



SURFEM-Ocean less precise than LOCEAN or RSS models (that are fitted for L-band).

SMAP

×10⁴ ×10⁴ 10 **Dependence with ocean parameters** 3 TBobs-TBsim (K) TBobs-TBsim (K) Number per bin Number per bin 8 2 2 8 6 6 4 0 36 V 2 2 - 1 -7 -7 ×10⁴ 10 ×10⁴ 3 3 GMI TBobs-TBsim (K) TBobs-TBsim (K) Number per bin Number per bin 8 2 2 8 6 6 4 0 0 **89V** 2 -1 -7 _7 ×10⁴ ×10⁴ 3 -FASTEM-6 TBobs-TBsim (K) TBobs-TBsim (K) Number per bin Number per bin 2 8 2 8 RSS LOCEAN 6 SURFEM 0 4 0 2 166V -1 -2 270 -2 0 0 0 10 15 20 280 290 300 5 OWS (m/s) SST (K)

Test of SURFEM-Ocean with multiple satellite observations:



WP4.1 Radiative transfer

• Evaluation of codes implemented for the scene simulation and inversions





WP4.1 Radiative transfer

• Evaluation of codes implemented for the scene simulation and inversions

- Limited dependence with the TCWV even at higher frequencies with WindSat.
- Large differences in the behaviours between WindSat and AMSR2.
- Large biases observed with AMSR2, as already evidenced in previous works (Kilic et al., 2020)



Configuration of PARMIO

- Dielectric constants: lacksquare
 - Meissner and Wentz, 2004 and 2012
- Wave spectrum: ۲
 - Durden and Vesecky, 1985 with the amplitude coefficient multiplied by 1.25.
- Foam coverage: ۲
 - New estimation

 $if U_{10} < 20$ $Fc = 6.25e^{-6} \times U_{10}^3$ else



- Foam emissivity: ۲
 - Anguelova and Gaiser, 2013 with a foam effective thickness of 2mm.

<u>Approach</u>

Within PARMIO the ocean emissivity is decomposed as follows:

$$e_{p} = e_{p}N + e_{p}0 + e_{p}1 \times \cos(\varphi) + e_{p}2 \times \cos(2\varphi)$$

For $p = V$ or H polarizations

 $e_q = e_q 1 \times sin(\varphi) + e_q 2 \times sin(2\varphi)$ For q = S3 or S4 (the 3rd or 4th stokes parameters)

With φ the relative wind direction

- $e_v N$ and $e_H N$ are the neutral emissivities (i.e. for a flat sea / no wind). They are estimated with the dielectric constant module from Meissner and Wentz [2004, 2012] and the Fresnel equations.
- $e_v 0$ and $e_H 0$ are the isotropic emissivities due to the wind. They are estimated with a neural network (NN).
- $e_v 1$, $e_v 2$, $e_H 1$, $e_H 2$, $e_{s_3} 1$, $e_{s_3} 2$, $e_{s_4} 1$ and $e_{s_4} 2$ are the anisotropic emissivities due to the wind. They are estimated with a second NN.

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