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Reducing Global Surface Emissivity Similarity Classification by Ecosystem Matching Karen I. Mohr¹ and Grant Petty²

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1. Introduction

Accurate land surface emissivity (LSE) estimates are critical to the successful inversion of passive microwave radiometric signatures into instantaneous rainfall rates. Improving the dynamic estimation of LSE for physically-based retrievals is a key objective of NASA's Precipitation Measuring Mission whose core satellite launch is in 2013. Because land surface emissivity is a function of land surface state variables such as surface roughness and wetness, the dynamic estimation of LSE using land surface models (LSM) is linked to the specification of both fixed and dynamic parameters describing land surface characteristics.



season and b) wet season in northern Mali and c) spring and d) fall by Mt Washington, NH. Note the surface roughness changes due to seasonal aridity in Mali and, in c vs. d, due to deciduous trees and snow in NH.

The long history of SSM/I observations has enabled Prigent, Aires, and colleagues to estimate surface emissivities from all available SSMI observations from 1993 to 2008, under clear sky conditions. Their emissivity database was subjected to a global cluster analysis of 6-channel monthly mean emissivities with covariances. The analysis resulted in 300 distinct surface classes over land and over high latitude, coastal, and continental shelf ocean regions in which seasonal variability is prominent.



Figure 2: Cluster analysis results: class 1 a) Jan and b) Jul; class 2 c) Jan and d) Jul; and class 8 e) Jan and f) Jul. Class 1 pertains only to high latitude oceans with artificial boundaries at the 45° parallels. Although pelagic tropical/sub-tropical oceans observed by TRMM are excluded from the analysis, coastal and continental shelf regions of the oceans at all latitudes are analyzed.

2. Approach

Each class map (12 maps/year) was converted into an Arc/Info format grid at 0.188° resolution from 80°N to 80°S and then all surface emissivity cluster classes for each month were combined into one grid. Some cells were not classifiable (< 5%). These cells and the tropical/subtropical oceans between 45°N-45°S were assigned "undefined" codes. In Figure 3 are the maps for January and July.







Figure 4: In a) are the number of times each gridcells is assigned a emissivity class. A grid cell of 7 (yellow) is assigned 7 different classes Jan to Dec.

In b) are emissivity classes 1-50 and c) classes 50-100, representing 80% and 10%, respectively, of the total number of gridcells analyzed. Key to b): DO deep ocean, MCO mid-continental ocean, ShO shallow (coastal) ocean, and DIW, SIW, EW = deep, shallow, ephemeral inland water. In a) note the variability in the middle to high northern latitudes, Antarctica, and the semi-arid regions of the tropics and sub-tropics. Class changes mainly take place interseasonally, but intraseasonal changes are common in regions with highly variable snow cover and sea ice [orange and red areas in a)].

3. Preliminary Results

In the GPROF ATBD Ver 1.0 (12/2010), it is proposed to reproduce a global selfsimilarity classification for each location/month with a goal of 15-25 classes. In this analysis, we looked to see if standard climate classifications had any value in grouping emissivity classes. Two standard classifications were tested vs. the monthly emissivity maps. These were Koeppen-Geiger (Rubel and Kottek 2010) and Bailey's Ecoregions (Bailey 1995, 1998), tested for spatial overlap and monthto-month stability. The overlap was determined using the Arc/Info COMBINE function. This utility assesses how much a series (2-50) of grids overlaps spatially by finding the common sets of gridcell values. It is essentially a spatiallydistributed form of clustering based on shared values.



Figure 5: Bailey's Ecoregions of the Continents. Unlike Koeppen-Geiger's climate regimes, this classification includes, is based on vegetation cover in addition to climate drivers affecting seasonal temperature and humidity changes

Bailey	s Ecoregions	s of the F	irst 10 Em	issivity	Classes
2	3	4	5	6	7

Dec	Öcean	Savanna Div	Rainforest Div	T/St Steppe Div	Ocean	Öcean	Ocean	T/St Desert Div	Subtropical Reg Mts	Ocean
an	Ocean	Savanna Div	Rainforest Div	T/St Steppe Div	Ocean	Ocean	Ocean	T/St Desert Div	Subtropical Reg Mts	Ocean
eb	Ocean	Savanna Div	Rainforest Div	T/St Steppe Div	Ocean	Ocean	Ocean	T/St Desert Div	Subtropical Reg Mts	Ocean
Mar	Ocean	Savanna Div	Rainforest Div	Savanna Div	Ocean	Ocean	Ocean	T/St Desert Div	Subtropical Reg Mts	Ocean
Apr	Ocean	Savanna Div	Rainforest Div	Savanna Div	Ocean	Ocean	Ocean	T/St Desert DIv	Subarctic Div	Ocean
May	Ocean	Savanna Div	Rainforest Div	Savanna Div	Ocean	Ocean	Ocean	T/St Desert Div	Subarctic Div	Ocean
un	Ocean	Savanna Div	Rainforest Div	Savanna Div	Ocean	Ocean	Ocean	T/St Desert Div	Subarctic Div	Ocean
ul	Ocean	Savanna Div	Rainforest Div	Savanna Div	Ocean	Ocean	Ocean	T/St Desert Div	Subarctic Reg Mts	Ocean
Aug	Ocean	Savanna Div	Rainforest Div	Savanna Div	Ocean	Ocean	Ocean	T/St Desert Div	Subarctic Reg Mts	Ocean
Sep	Ocean	Savanna Div	Rainforest Div	Savanna Div	Ocean	Ocean	Ocean	T/St Desert Div	Subarctic Div	Ocean
Det	Ocean	Savanna Div	Rainforest Div	Savanna Div	Ocean	Ocean	Ocean	T/St Desert Div	Subarctic Div	Ocean
Vov	Ocean	Savanna Div	Rainforest Div	Savanna Div	Ocean	Ocean	Ocean	T/St Desert Div	Subtropical Reg Mts	Ocean

the Bailey's Ecoregions.

Tables (top, middle, bottom): These tables illustrate how closely the emissivity classes, particularly the large classes (< 20) correspond to

The TOP table shows the principal classification

for the first 10 classes. Over 80% of the cells in

these classes fell into the regime listed for that

class each month. Note seasonal changes.

All 300	with more than 1 BER Div.	Top 20	wit than 1
Dec	29%	Dec	
Jan	31%	Jan	
Feb	31%	Feb	
Mar	28%	Mar	-
Apr	25%	Apr	
May	27%	May	
Jun	24%	Jun	
Jul	23%	Jul	
Aug	25%	Aug	
Sep	27%	Sep	
Oct	28%	Oct	
Nov	32%	Nov	

<mark>/ vs. SE</mark> lap lec	Percent of Spatial Overlap Globally 79%	The middle table shows that fewer than 4-5% of the top 20 classes had more than one regime listed for the class. There is more variability for the smaller, higher numbered classes, but less than one-third fell into multiple regimes.
an eb	77%	less than one third ten into matche regimesi
Nar opr tay un	78% 80% 81% 81%	The bottom table shows the overlap between the Bailey's map and the monthly emissivity maps. Cell-to-cell, the SE classes tend to fall into specific Bailey's Ecoregions.
ug	82%	Acknowledgments: Catherine Prigent and Felipe Aires for their SSM/I
ep Oct Iov	80% 80% 79%	Limatology. Discussions with Radiometer Algorithm working group members (Joe Turk, Christa Peters-Lidard, Yudong Tian, Ken Harrison, Tom Wilheit, Gail Jackson). Funded by the NASA Precipitation Measuring Mission.