

## Adapting the AAPP microwave tests for EPS-SG: part 2 – surface tests

This documentation was developed within the context of the EUMETSAT Satellite Application Facility on Numerical Weather Prediction (NWP SAF), under the Cooperation Agreement dated 7 September 2021, between EUMETSAT and the Met Office, UK, by one or more partners within the NWP SAF. The partners in the NWP SAF are the Met Office, ECMWF, DWD and Météo France.

COPYRIGHT 2022, EUMETSAT, ALL RIGHTS RESERVED.

Change record			
Version	Date	Author / changed by	Remarks
1.0	15/08/2022	N C Atkinson	

## Table of Contents

<b>1.</b>	<b>INTRODUCTION .....</b>	<b>3</b>
<b>2.</b>	<b>SURFACE TEST IMPLEMENTED IN AAPP V8 .....</b>	<b>3</b>
<b>3.</b>	<b>THE TELSEM ATLAS.....</b>	<b>5</b>
<b>4.</b>	<b>MWS SIMULATION.....</b>	<b>6</b>
<b>5.</b>	<b>RESULTS FOR 8 CATEGORIES.....</b>	<b>8</b>
<b>6.</b>	<b>RESULTS FOR 6 CATEGORIES.....</b>	<b>11</b>
<b>7.</b>	<b>CONCLUSIONS .....</b>	<b>15</b>
<b>8.</b>	<b>REFERENCES .....</b>	<b>16</b>
<b>9.</b>	<b>APPENDIX – MEAN AND COVARIANCE TABLES FOR MWS.....</b>	<b>16</b>

## 1. INTRODUCTION

NWP SAF document NWPSAF-MO-TR-040 discusses the adaption of several AAPP microwave tests for use with the MicroWave Sounder (MWS) on EPS-SG.

In this document we consider in more detail the test that is intended to characterise the land or sea surface.

It should be noted that these tests make use only of the observations, and do not use a model background. They can therefore be useful for detecting inconsistencies between the data and the model.

## 2. SURFACE TEST IMPLEMENTED IN AAPP V8

This section is largely based on the earlier document NWPSAF-MO-TR-040, and is included for completeness.

The surface test implemented in AAPP version 8 (and earlier releases) makes use of three channels, at 23.8 GHz (channel 1), 31.4 GHz (channel 2) and 50.3 GHz (channel 3). These channels are common to AMSU, ATMS and MWS. The test works by looking for the surface type that minimises a cost function

$$J = (T_1 - \bar{T}_1, T_2 - \bar{T}_2, T_3 - \bar{T}_3) \mathbf{C}^{-1} \begin{pmatrix} T_1 - \bar{T}_1 \\ T_2 - \bar{T}_2 \\ T_3 - \bar{T}_3 \end{pmatrix} \quad (1)$$

where  $\mathbf{C}$  is a 3x3 covariance matrix,  $T_i$  is the brightness temperature for channel  $i$  and the overbar denotes a mean brightness temperature. Values of  $\mathbf{C}$ ,  $\bar{T}_1$ ,  $\bar{T}_2$  and  $\bar{T}_3$  are provided for 5 different fixed zenith angles ( $\sec(z) = 1.0, 1.25, 1.5, 1.75, 2.0$ ), and they are interpolated for each observation. Matrices are provided for 8 surface types, and a ninth (desert) is deduced afterwards from a test of channel 1 BT. The types are shown in Table 1. The principles for distinguishing the various types, based on emissivity characteristics (including variation with frequency) are given in Grody (1988).

**Table 1: Microwave surface types in AAPP**

Type number	Description
1	Bare young ice (i.e. new ice, no snow)
2	Dry land (i.e. dry with or without significant vegetation)
3	Dry snow (i.e. snow with water less than 2%, over land)
4	Multi-year ice (i.e. old ice with snow [assumed dry] cover)
5	Sea (i.e. open water, no islands, ice-free, WS=0 to 14m/s)
6	Wet forest (i.e. established forest with wet canopy)
7	Wet land (i.e. non-forested land with a wet surface)
8	Wet snow (i.e. snow with water content > 2%, over land or ice)

9	Desert
---	--------

The value of the cost function, after surface type identification, gives an indication of cloud liquid water. As stated in English (1997): *In simple terms, AMSU-A channel 1 is sensitive to total water vapour, channel 2 cloud liquid water and channel 3 absorber temperature. Because the covariance due to water vapour and temperature variations is large in channels 1 and 3, changes in the atmospheric profile have little impact on the cost, J. However, the introduction of liquid water, which increases channel 2 out of proportion, leads to a very rapid rise in cost. This test is sensitive enough to the presence of cloud liquid water to detect cloud amounts below 100 gm<sup>-2</sup>.*

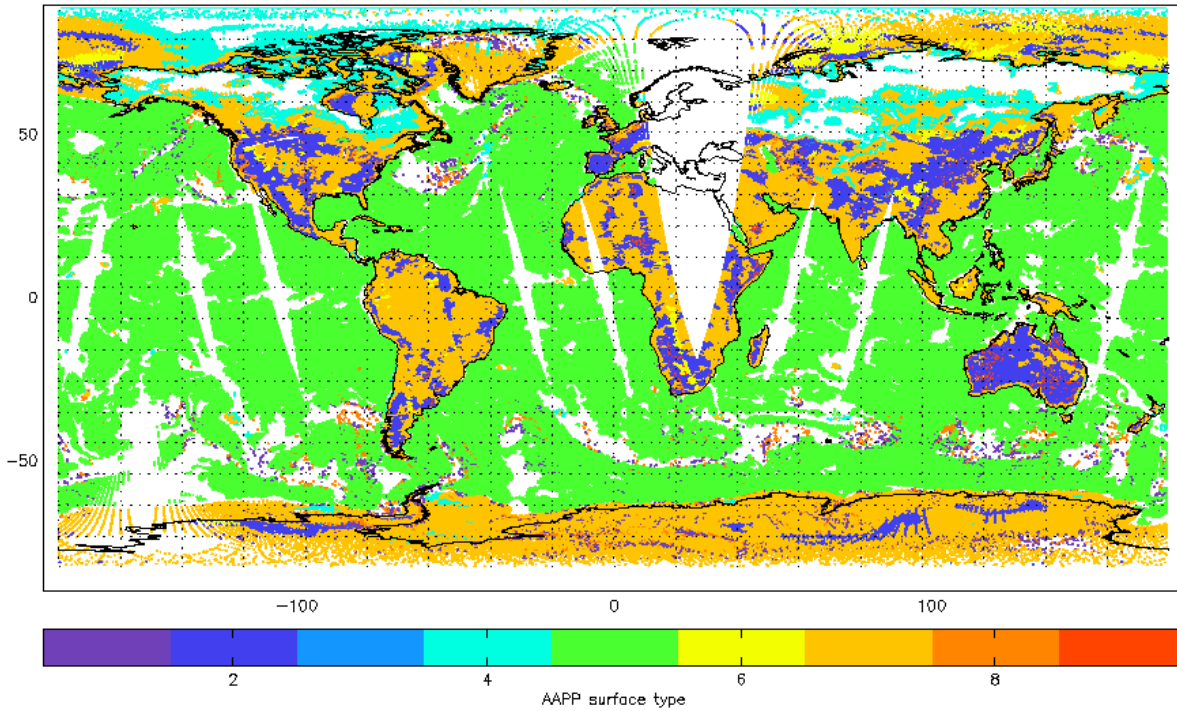
Note that matrix **C** is symmetric and its terms are all positive. **C**<sup>-1</sup> is also symmetric but can have negative off-diagonal terms.

It is not known exactly what database was used to establish this test. The AAPP Scientific Description states: “Mean brightness temperatures and BT covariance matrices have been calculated (with a radiative transfer model) for the 20 AMSU channels for different surface types, with no cloud liquid water.” Moreover, the document also says, “the coefficients are not well developed and its results are currently not very meaningful.”

It is likely that the FASTEM land/sea-ice component was used. This is configurable via 5 input parameters to represent different surface types for MW emissivities (the “Grody parameters”). This code is still available in the current FASTEM (part of RTTOV), but it is considered to be deprecated (J. Hocking, pers. comm.).

Different centres would be expected to have different uses for the AAPP surface test output. At the Met Office, it is used (i) to improve the profile surface type and to set “mismatch” flags if the AAPP surface type differs from the model type, and (ii) to set an appropriate emissivity over sea ice (B. Candy, pers. comm). The Met Office requirement is for the AAPP test to distinguish between (i) sea, (ii) multi-year ice and (iii) other surfaces.

Figure 1 shows the AMSU surface type for NOAA-18 and NOAA-19 observations for the 00Z cycle on 16<sup>th</sup> March 2021. We can see that sea areas are correctly identified as type 5. Land areas are mostly classified as dry land (type 2) or wet land (type 7); it is not obvious whether there is any skill in distinguishing them. The main areas of wet forest (type 6) are over polar sea-ice – which is not realistic! There are some areas classified as multi-year ice (type 4) in the Arctic, but also some areas in central Asia which are more likely to be desert. Almost no areas are actually classified as desert, though note that this was a night-time scene over the African deserts, where the land surface is cold.



**Figure 1: AMSU surface type from AAPP for samples with surface cost < 10K**

The test would need improvement, and coefficients would need to be re-generated, if it is to be useful for MWS on EPS-SG. In the next section we discuss use of the TELSEM atlas in performing this task.

### 3. THE TELSEM ATLAS

RTTOV has an option to use emissivity values from the TELSEM version 2 atlas (Prigent and Aires, 2015). This atlas comprises monthly files from which it is possible to derive surface emissivity at different microwave frequencies. It is also possible to extract the surface category – there are 18 surface categories provided, covering land and sea-ice, as shown in Figure 2.

Surface type	Class nb for each type	Class1 in TELSEM <sup>2</sup>	Class2 in TELSEM <sup>2</sup>	Broad characteristics
Sea-ice	1	10	11	Sea-ice marging
	2	10	12	New sea-ice
	3	10	13	New sea-ice
	4	9	14	Multi-year sea-ice
	5	8	15	Multi-year sea-ice
	6	7	16	Multi-year sea-ice
Snow and continental ice	1	TELSEM class	17	Snow (small grain)
	2	TELSEM class	18	Snow (medium grain)
	3	TELSEM class	19	Snow (large grain)
	4	TELSEM class	20	Continental ice
	5	TELSEM class	21	Continental ice
	6	TELSEM class	22	Continental Ice
Continents (snow and ice-free)	1	1	1	Tropical forest
	2	2	2	Dense vegetation
	3	3	3	Bushes
	4	4	4	Grassland
	5	5	5	Deserts
	6	10	10	Water surfaces

Figure 2: TELSEM surface types. The table is from Prigent and Aires, 2016

The resolution of the atlas is 0.25° in latitude and longitude. Interpolation software is provided, to allow emissivity to be estimated at any point.

For AAPP use, it is clearly necessary to reduce the number of categories, which will be discussed in subsequent sections.

#### 4. MWS SIMULATION

The NWP SAF Radiance Simulator (RadSim) was used to simulate MWS brightness temperatures for channels 1, 2 and 3 (23.8, 31.4 and 50.3 GHz respectively).

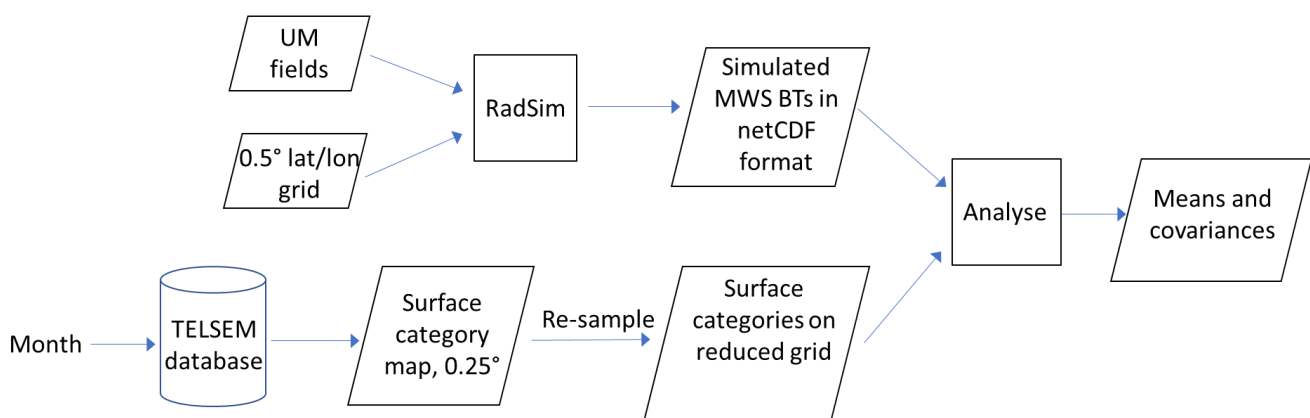
To generate the statistics from which the means and covariances can be derived, a rectangular (720 x 360) half-degree lat/lon grid was used – repeated 5 times for the 5 zenith angles. RTTOV was configured to use the TELSEM atlas.

Atmospheric state was taken from Met Office UM model fields. Two different sets of fields were used covering different seasons and times of day

- 20210316T0000Z
- 20220807T1200Z

This allows simulated brightness temperatures to be generated. The analysis could be extended later, if needed, to cover other seasons and times.

The TELSEM categories were extracted from the database, for the appropriate month. These should be consistent with the simulated BTs. The outputs were then combined and analysed according to the scheme illustrated in Figure 3.



**Figure 3: Processing scheme used in this study**

The ‘analyse’ step in Figure 3 is as follows:

1. For each observation, identify the surface category (discussed in subsequent sections) and the zenith angle category.
2. For each surface category  $i$  and zenith angle category  $j$ , compute the sum of the BTs and also the sum of  $\mathbf{B}_{ij} \mathbf{B}_{ij}^T$  (where  $\mathbf{B}_{ij}$  is a 3-element column vector of brightness temperatures for the 3 channels). The sums are stored in a file which can be added to in subsequent runs.
3. For each category, compute the mean BT:  $\mathbf{m}_{ij} = \Sigma(\mathbf{B}_{ij})/n_{ij}$  and covariance  $\mathbf{C}_{ij} = \Sigma(\mathbf{B}_{ij} \mathbf{B}_{ij}^T)/n_{ij} - \mathbf{m}_{ij} \mathbf{m}_{ij}^T$ , where  $n_{ij}$  is the number of observations in categories  $i$  and  $j$ .

The March and August runs were combined together before computing means and covariances.

Having computed the means and covariances, it is then possible to access the degree to which the different surface categories can be identified in the simulated BTs, by computing the cost function (eq. 1) for each category and identifying the category that minimises the cost function.

For this assessment, the sample locations and viewing geometry were taken from approximately 12 hours of ATMS data. Both NOAA-20 and Suomi NPP were used, in order to give a good distribution of viewing angles. AAPP “level 1d” files were used, retaining all 96 spots across scan, but thinning to 1 scan in 3. Note that the date and time of the ATMS observations is irrelevant as they are merely to provide a good geographical spread of sample locations and angles. The

reference matrices for the 5 zenith angles were interpolated for each observation – as in the operational version of AAPP.

## 5. RESULTS FOR 8 CATEGORIES

As noted previously, a scheme such as the one described above cannot expect to identify all 18 TELSEM categories. As a first attempt at reducing the number of categories, they were merged so as to give 8 categories similar to those implemented in AAPP v8.

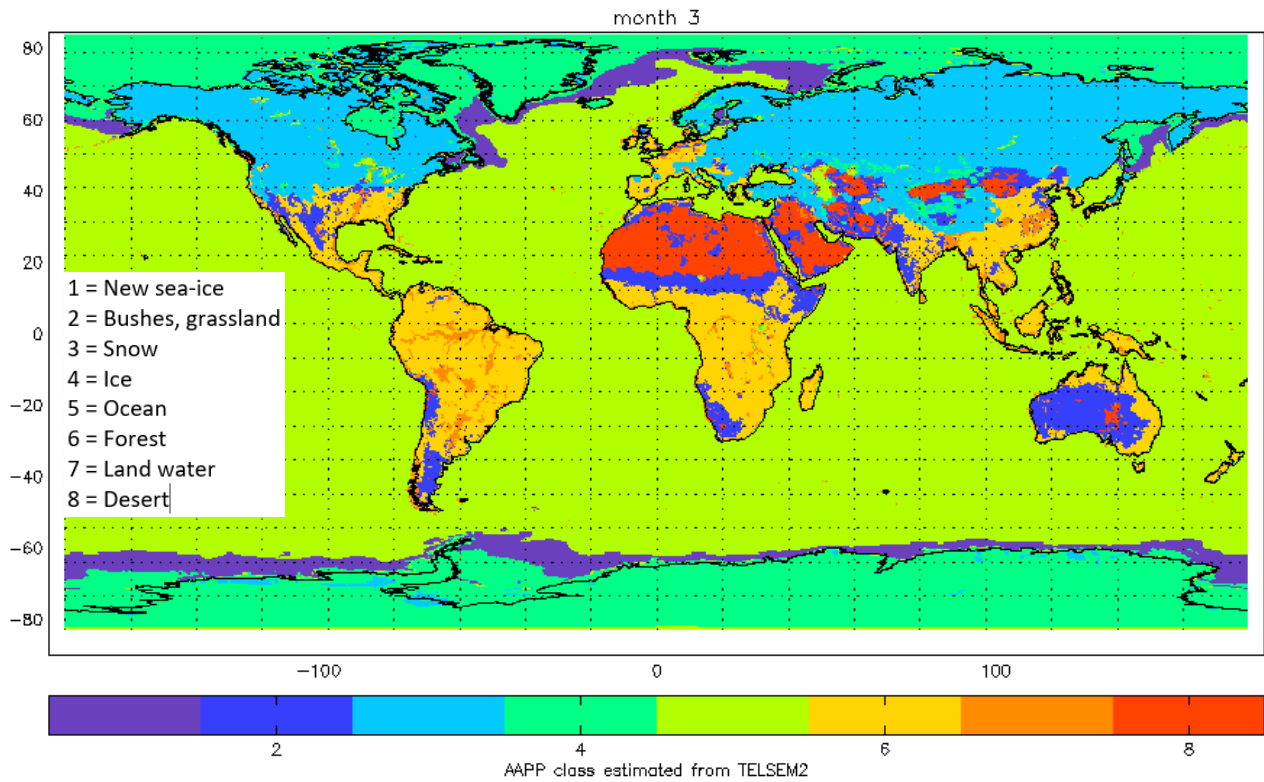
**Table 2: Translation of TELSEM categories to eight categories based on those in AAPP v8**

AAPP category number	TELSEM Class2	Characteristic
1	11, 12, 13	New sea-ice
2	3, 4	Bushes; grassland
3	17, 18, 19	Snow
4	20, 21, 22, 14, 15, 16	Continental ice; multi-year sea-ice
5	-	Ocean
6	1, 2	Tropical forest; dense vegetation
7	10	Water surfaces over land
8	5	Desert

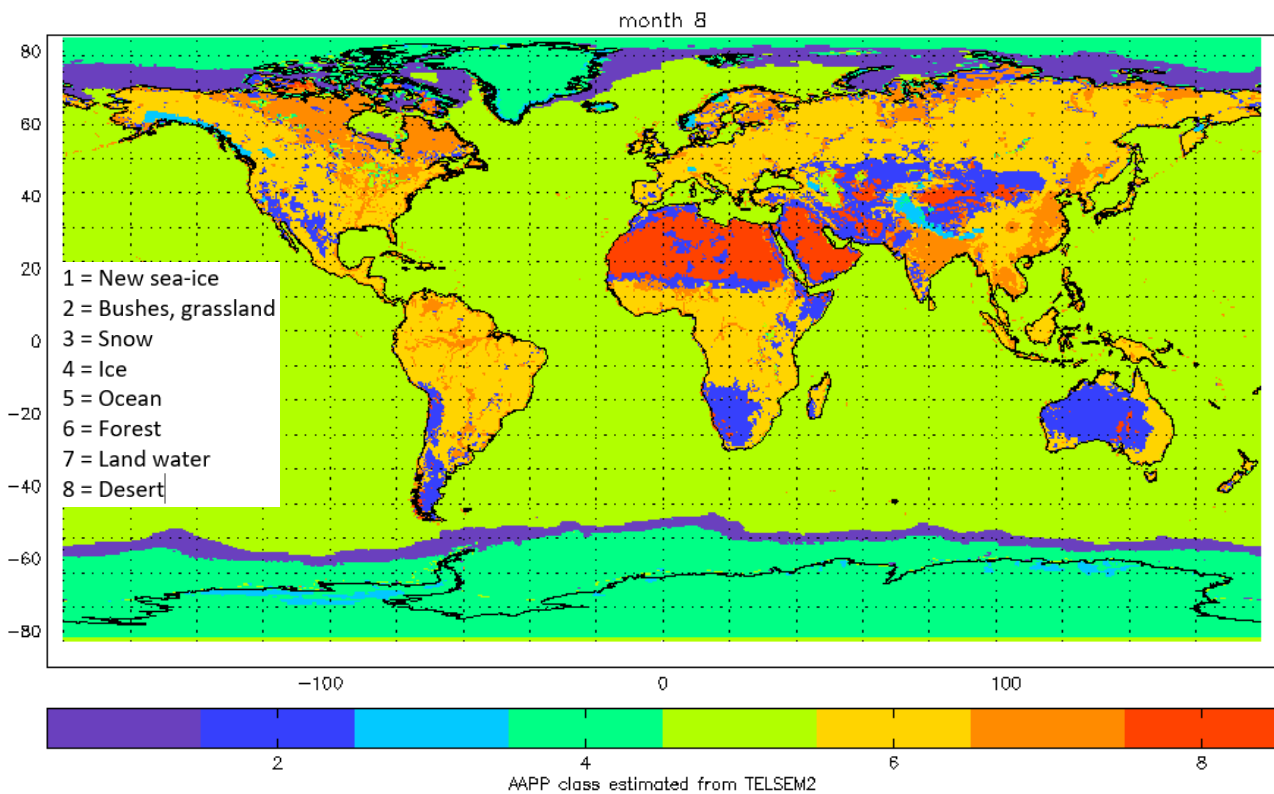
Note that there is no TELSEM category for “wet snow” – hence there are only 8 categories, not the 9 shown in Table 1.

Maps of the 8 classes corresponding to the TELSEM atlas are shown in Figure 4 for March and Figure 5 for August. Note the rather large extent of northern hemisphere snow in March (category 3), and the Antarctic sea-ice in August (category 4).



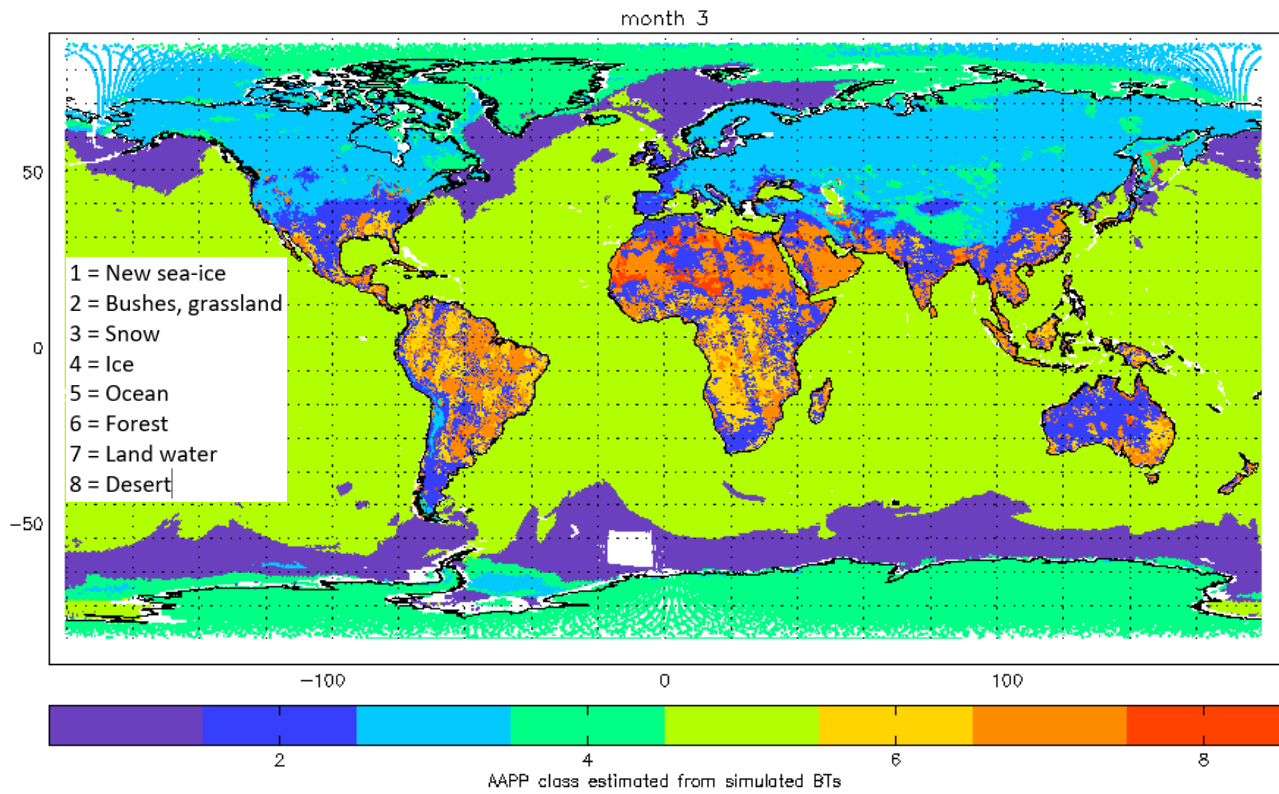


**Figure 4: TELSEM categories for March converted to 8 AAPP classes. Note that data poleward of 87.5°N have been excluded as the TELSEM atlas appears to be missing these points**

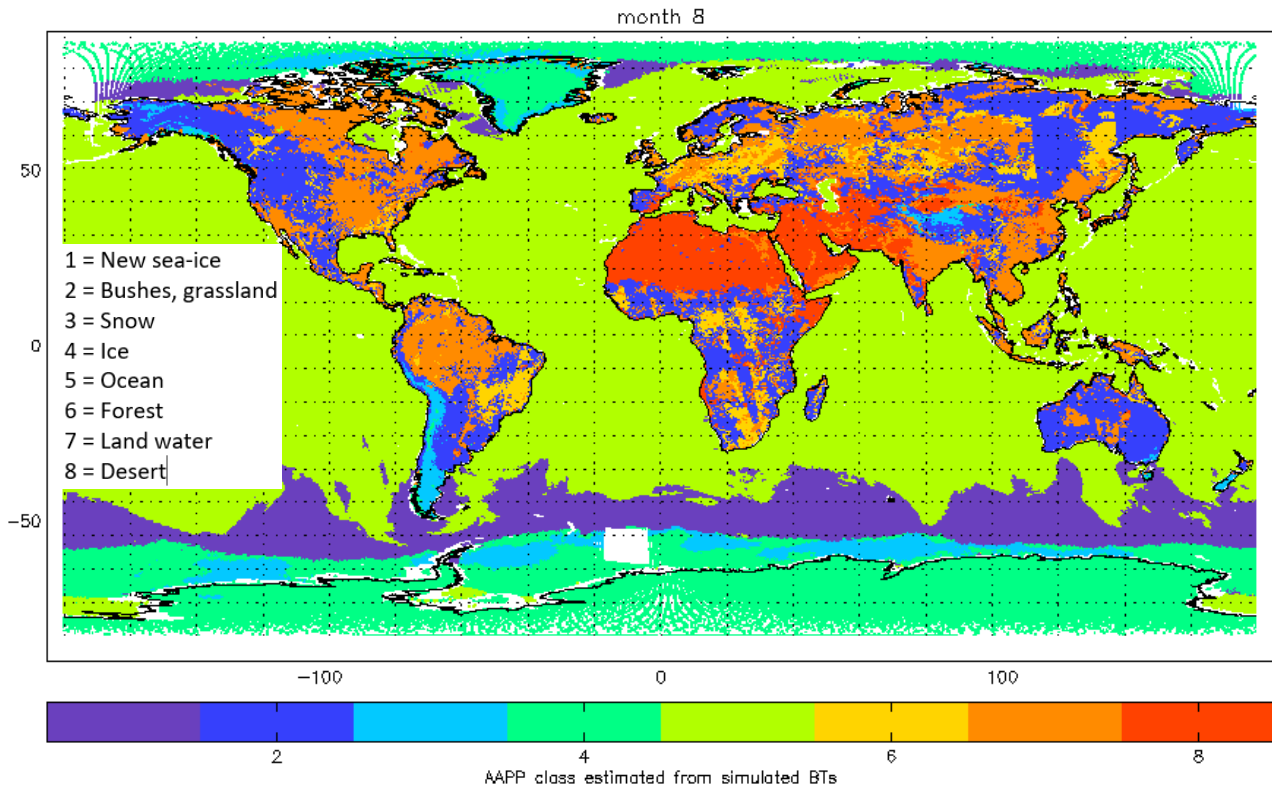


**Figure 5: TELSEM categories for August converted to 8 AAPP classes**

The retrieved surface types are shown in Figure 6 and Figure 7 for March and August respectively.



**Figure 6: Retrieved surface category for 00Z on 16 March 2021 using simulated MWS BTs from ATMS sample locations. Note that the gap in the south Atlantic is because the time period used was a little less than 12 hours.**



**Figure 7: Retrieved surface category for 12Z on 7 August 2022 using simulated MWS BTs from ATMS sample locations**

We can make the following observations:

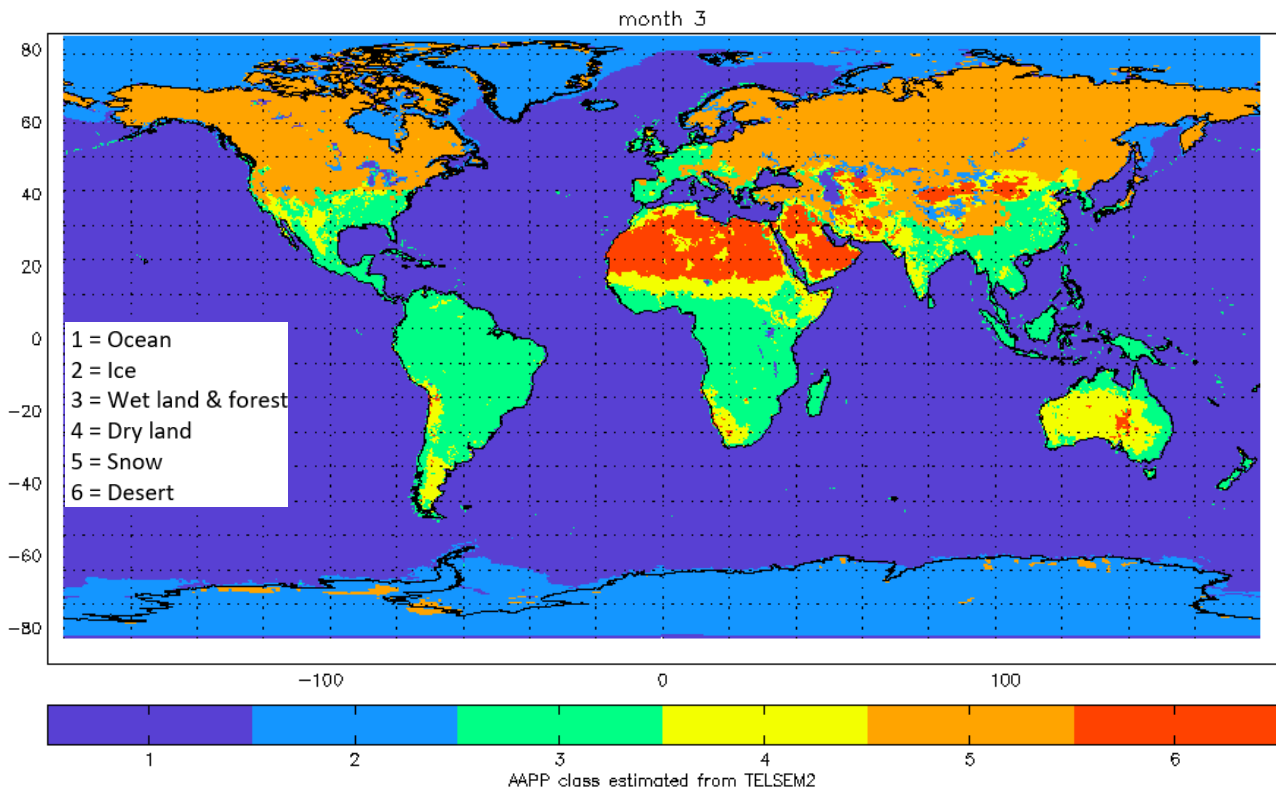
- The distinction between “new ice” and “ocean” appears to be weak. There is far too much “new ice” shown in the retrieved categories. If this was used to identify sea areas as part of a quality control system, a lot of data would be rejected.
- The land categories are rather mixed up.
- There is reasonable skill in the sea-ice.
- The “desert” category exists mainly over Africa, and is only detectable in the microwave radiances in the daytime, once the land surface has heated up

## 6. RESULTS FOR 6 CATEGORIES

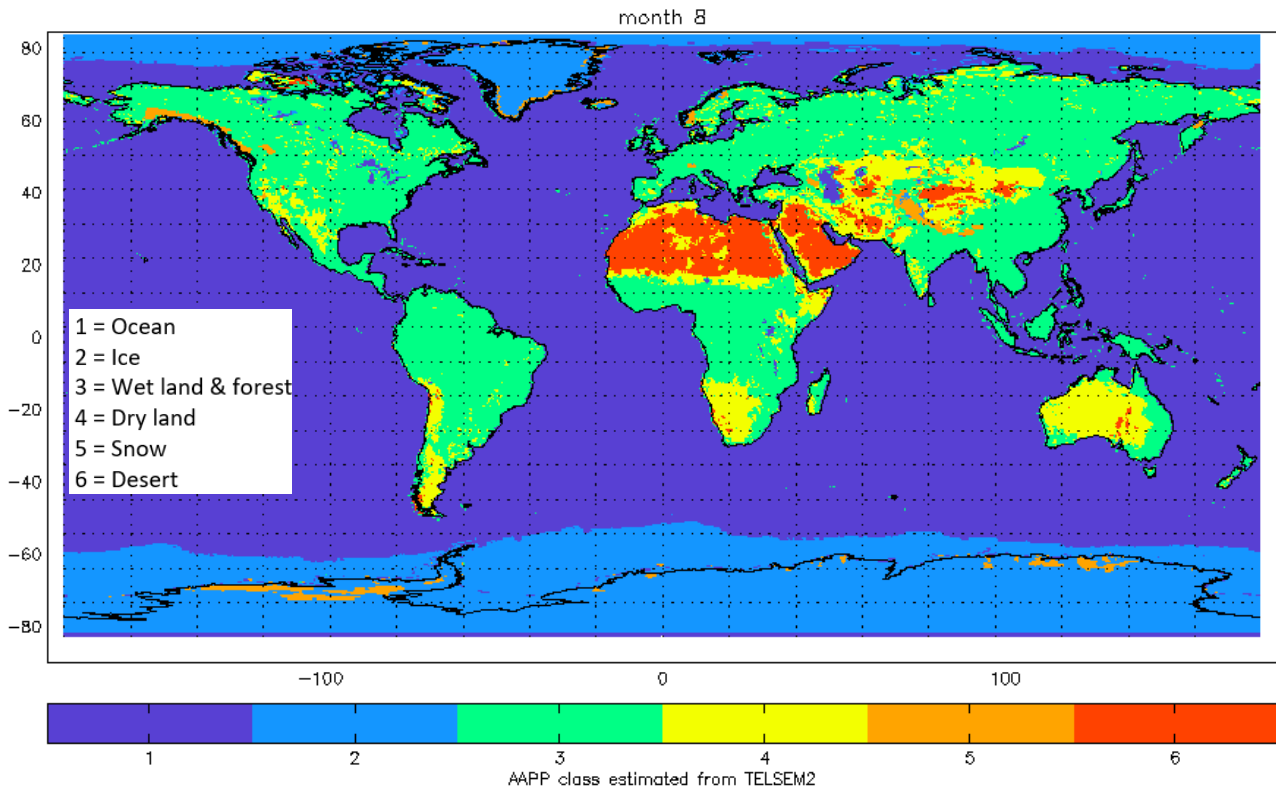
To attempt to improve these aspects, the number of categories was reduced to 6 by merging new sea-ice with ocean, and merging water surfaces over land with dense vegetation. See Table 3, Figure 8 and Figure 9.

**Table 3: Translation of TELSEM categories to six categories**

category number	TELSEM Class2	Characteristic
1	None, 11, 12, 13	Ocean and new sea-ice
2	20, 21, 22, 14, 15, 16	Multi-year ice
3	1, 2, 10	Wet land and forest
4	3, 4	Dry land
5	17, 18, 19	Snow
6	5	Desert

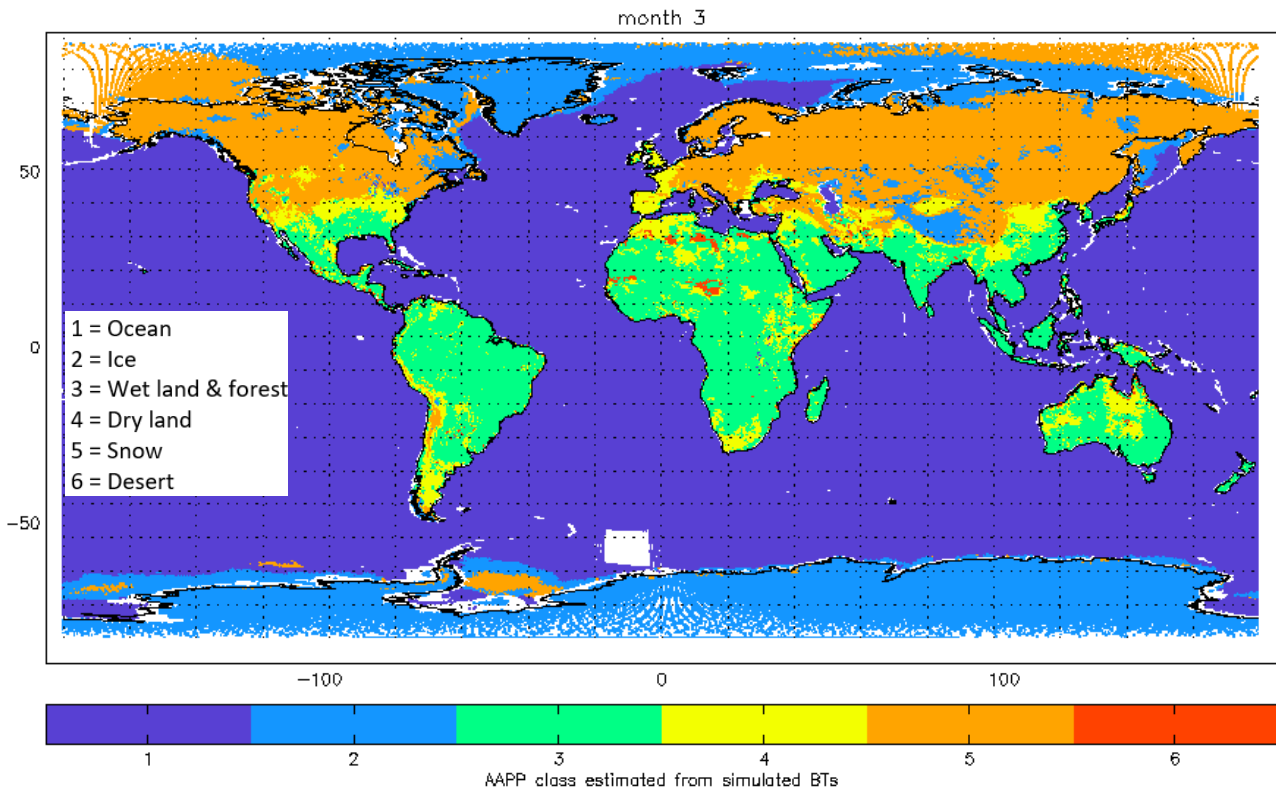


**Figure 8: TELSEM categories for March converted to 6 AAPP classes**

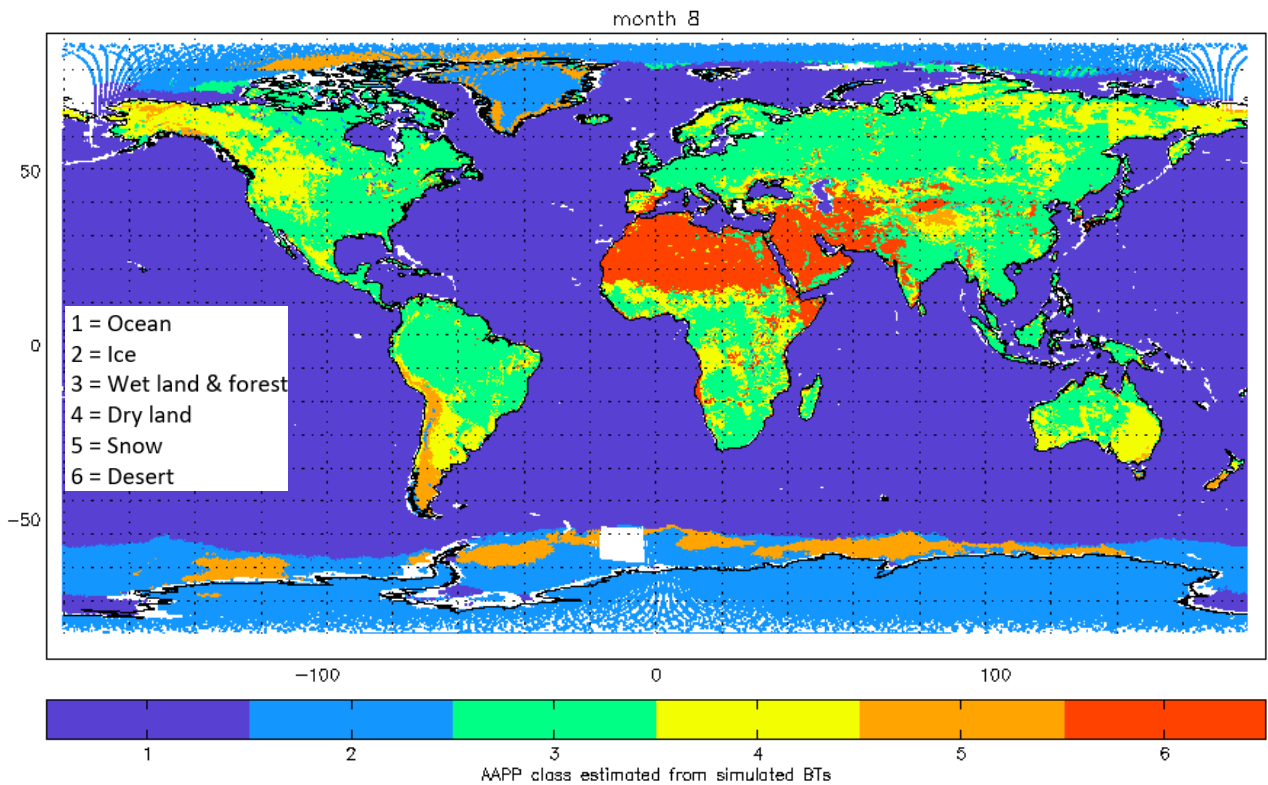


**Figure 9: TELSEM categories for August converted to 6 AAPP classes**

The corresponding retrievals are shown in Figure 10 and Figure 11.



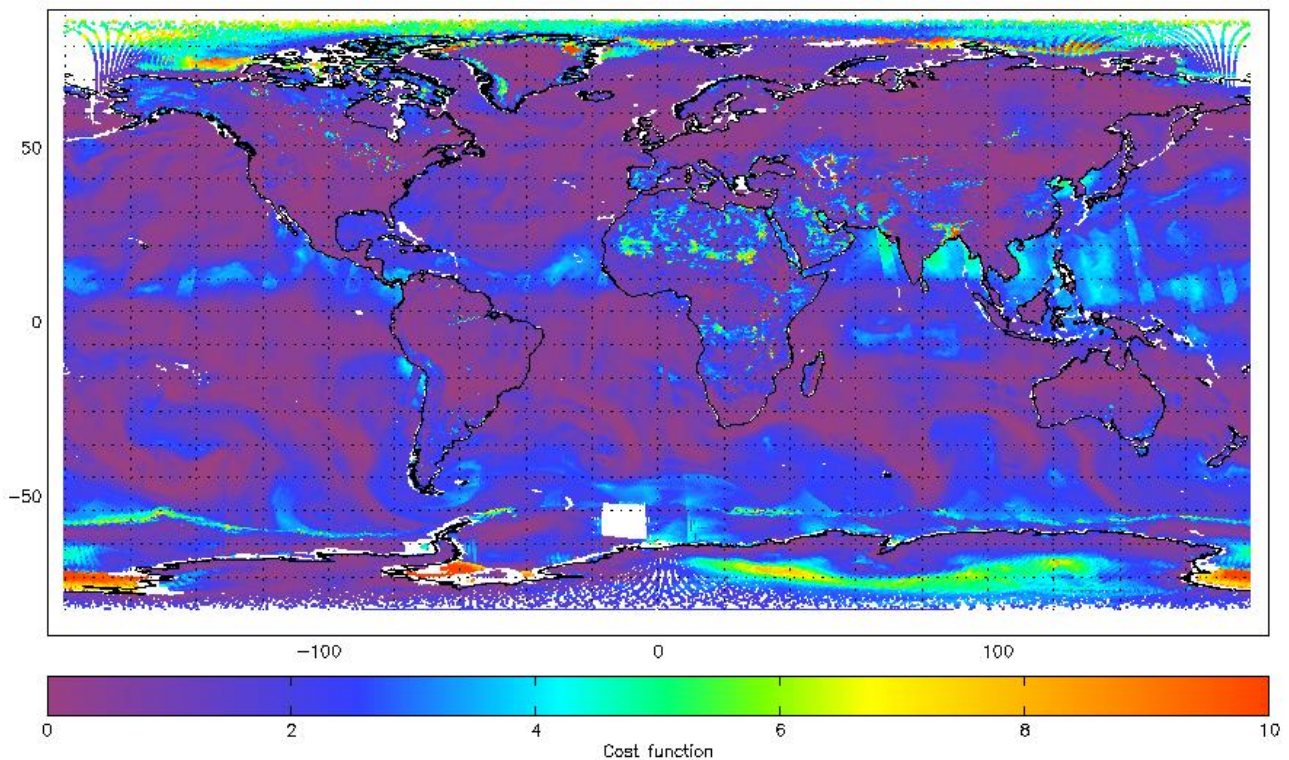
**Figure 10: Retrieved surface category (1 to 6) for 00Z on 16 March 2021 using simulated MWS BTs from ATMS sample locations**



**Figure 11: Retrieved surface category (1 to 6) for 12Z on 7 August 2022 using simulated MWS BTs from ATMS sample locations**

With six categories, the retrieval is somewhat improved. The ocean, sea-ice and snow categories are broadly correct. The desert is reasonable during the day. There is rather limited skill in distinguishing between dry land and vegetated land, but that is of secondary importance as far as AAPP users are concerned (infrared techniques work better in that regard).

The cost function (defined in equ. (1)) corresponding to Figure 11, is shown in Figure 12. In AAPP, the cost function is normalised by dividing by the square of the number of channels, and a threshold for cloud is specified as 5.5 in the normalised cost function. Nearly all points in Figure 12 would therefore be classified as cloud-free – which is as expected since no clouds were included in the RTTOV simulation.



**Figure 12: Cost function (un-normalised) corresponding to Figure 11. AAPP subsequently normalises the cost by dividing by the square of the number of channels**

## 7. CONCLUSIONS

A six-category standalone surface test has been devised for use with MWS. It is proposed to implement this test in AAPP version 9, in the module that will accommodate EPS-SG. Mean BTs and covariances are shown in the Appendix.

There may be benefits to carrying out a similar analysis using post-launch observed BTs rather than simulated – provided that effective cloud-clearing is available.

## 8. REFERENCES

English S.J., R. J. Renshaw, P. C. Dibben, and J. R. Eyre, 1997: "The AAPP module for identifying precipitation, ice cloud, liquid water, and surface type on the AMSU-A grid". *Proceedings of the Ninth International TOVS Study Conference*, J. R. Eyre, Ed., ECMWF, 119–130

[http://library.ssec.wisc.edu/research\\_Resources/publications/pdfs/ITSC9/english01\\_ITSC9\\_1997.pdf](http://library.ssec.wisc.edu/research_Resources/publications/pdfs/ITSC9/english01_ITSC9_1997.pdf)

N. C. Grody, "Surface identification using satellite microwave radiometers" in *IEEE Transactions on Geoscience and Remote Sensing*, vol. 26, no. 6, pp. 850-859, Nov. 1988.

Catherine Prigent and Filipe Aires, 2016: "Tool to Estimate Land Surface Emissivity at Microwaves and Millimeter waves", EUMETSAT Project EUM/CO/14/4600001473/CJA, available from <https://nwp-saf.eumetsat.int/site/software/rttov/download/> (included in TELSEM download).

## 9. APPENDIX – MEAN AND COVARIANCE TABLES FOR MWS

```

AmeanBT =
! Surface 1
  172.2773    159.1212    222.3246
  176.1041    158.9730    228.6729
  187.5106    168.8058    237.4759
  199.1255    179.9392    244.6851
  209.0610    189.8634    249.8916
! Surface 2
  212.6208    212.0501    222.1425
  205.3085    204.4546    221.0775
  204.9024    202.7020    221.8196
  205.5846    202.0665    222.8645
  206.5107    201.8338    223.9334
! Surface 3
  271.9104    269.4530    273.8154
  272.9976    269.6643    273.6113
  273.7021    269.8256    273.1757
  274.4722    270.2219    272.7511
  275.1930    270.6958    272.2836
! Surface 4
  277.1412    275.7126    275.9280
  276.7378    274.7915    274.9898
  275.5508    272.9698    273.4575
  274.9624    271.7766    272.3121
  274.7267    271.0152    271.4140
! Surface 5
  240.4295    237.0165    239.6501
  238.0909    232.1580    238.3965
  236.8806    229.3316    238.1806
  236.3975    227.5217    238.4265
  236.2766    226.3215    238.8519
! Surface 6
  283.4961    282.0514    282.7207
  283.1535    281.0606    281.7412

```



281.6212	278.6832	279.8548
280.8539	277.1559	278.4293
280.6029	276.2811	277.3721

Acov =

! Surface 1

572.7359	331.3117	193.5178
331.3108	284.2204	114.6667
193.5161	114.6702	77.4203

723.8991	394.4729	228.0416
394.4720	309.8286	117.2430
228.0381	117.2395	85.3178

764.4414	401.4511	236.9489
401.4546	288.8710	109.8010
236.9559	109.7992	90.6207

760.3864	394.5778	240.1043
394.5778	263.9452	107.1194
240.1061	107.1177	97.1071

742.9573	388.8227	242.4440
388.8245	248.0406	109.7385
242.4440	109.7350	103.4850

! Surface 2

865.9026	852.0393	584.0352
852.0393	849.1563	600.5607
584.0352	600.5607	505.3091

957.8180	883.4231	571.1099
883.4231	827.6511	536.8618
571.1099	536.8618	435.6057

937.1466	850.7619	548.7581
850.7619	790.5874	505.8399
548.7581	505.8399	417.6859

905.3373	816.4185	531.2586
816.4185	758.0861	484.2149
531.2586	484.2149	405.3534

880.1515	792.0907	517.0356
792.0907	737.6231	468.9370
517.0356	468.9370	393.8088

! Surface 3

677.9994	727.1792	367.2439
727.1792	798.4813	395.8819
367.2439	395.8819	208.1688

633.4822	700.0149	318.2056
700.0149	796.7484	352.1977
318.2056	352.1977	169.6163

518.6329	578.9856	241.2693
578.9856	673.0345	268.2259
241.2693	268.2259	122.0426

# Adapting the AAPP microwave tests for EPS- SG: part 2 – surface tests

418.7396	469.3496	183.3384
469.3496	553.2678	202.5348
183.3384	202.5348	90.4195

343.1251	386.2328	144.3139
386.2328	459.7285	158.2054
144.3139	158.2054	71.6888

! Surface 4

216.7694	218.3131	176.7972
218.3131	221.5473	178.4317
176.7972	178.4317	148.8197

219.3762	221.3307	170.8387
221.3307	225.7368	172.5833
170.8387	172.5833	139.4114

216.8966	218.0800	162.8190
218.0800	222.0403	163.8111
162.8190	163.8111	128.4772

213.5354	213.9895	155.5059
213.9895	217.8300	155.4934
155.5059	155.4934	119.0975

210.3403	211.1320	148.9119
211.1320	215.0100	148.8125
148.9119	148.8125	111.1047

! Surface 5

357.5314	376.8039	265.7387
376.8039	409.8458	296.0162
265.7387	296.0162	230.6145

384.4513	404.5731	260.6769
404.5731	444.7213	294.9822
260.6769	294.9822	212.8816

385.4705	411.0099	248.2015
411.0099	462.4702	284.6516
248.2015	284.6516	191.5368

378.6977	410.3182	233.3786
410.3182	471.5127	270.6679
233.3786	270.6679	171.8042

369.5492	406.9485	218.8254
406.9485	476.6099	255.7111
218.8254	255.7111	153.7717

! Surface 6

350.0343	358.7999	285.8337
358.7999	369.3788	293.4213
285.8337	293.4213	235.7079

340.0107	351.3412	265.8422
351.3412	365.0915	274.9252
265.8422	274.9252	210.3610

# Adapting the AAPP microwave tests for EPS- SG: part 2 – surface tests

323.4083	334.3326	238.8365
334.3326	348.1589	247.0750
238.8365	247.0750	178.6523
313.0638	325.0145	219.8418
325.0145	340.1192	228.1414
219.8418	228.1414	156.5159
303.8927	317.4441	203.5820
317.4441	334.9548	212.3494
203.5820	212.3494	138.6153