

EUMETSAT Satellite Application Facility on Support to Operational Hydrology and Water Management



Product Validation Report (PVR) for product P-AC-SM2RAIN (H64)

Precipitation/Soil Moisture integrated product

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1 Introduction to the Product Validation Report

The Product Validation Report (PVR) contains all the useful information for users to acquire more about the limits and potentials of the precipitation product. The document collects all the information and all the results obtained from the Quality and Monitoring Assessment cluster.

The precipitation product under review, during the analyzes carried out by the cluster, is on "indevelopment" phase. The objective of this report is to determine and quantify both the limits and the potential of the product in order to assess the overall quality and make it available to end users.

The report is structured as follows:

In **Chapter 2 "<u>P-AC-SM2RAIN product (H64)</u>**" there is a brief description of the product and the precipitation algorithm retrieval. More details about this can be found on the PUM (Product User Manual) and on the ATBD (Algorithm Theoretical Baseline Document) of the products themselves.

The main part of this document is contained in **Chapters 3** "<u>Validation results: case study analysis</u>" and 4 "<u>Validation results: long-term analysis</u>": these are dedicated to the description of the validation results obtained for both case studies and analyses over a long period, respectively.

For more information about the analysis, validation, and quality assessment methodology, compare **Appendix 1** "<u>Validation strategy, methods and tools</u>"

All detailed information on the data used to compare the products is contained in Appendix 2 "Ground data used for validation activities".

Any further results obtained and not included in chapters 3 and 4, are included from **Appendix 3** onwards.

Finally, the last **Appendix 6** "<u>Acronyms</u>" is dedicated to the list of numerous acronyms used in this document.

For any errors, oversights, or requests for updates or changes, please contact us via H SAF project website: <u>http://h-saf.eumetsat.int</u>



2 P-AC-SM2RAIN product (H64)

Product H64 (P-AC-SM2R-PMW over the H SAF Extended Area) is based on the integration between precipitation measurements retrieved by LEO PMW instruments and soil moisture-derived rainfall obtained via SM2RAIN algorithm (Brocca et al., 2014). It refers to accumulated rainfall.

2.1 Algorithm principle

A short description of the algorithm used to estimate rainfall from soil moisture (SM) data and to merge it with Passive Microwave (PMW) rainfall estimates is given. SM2RAIN (Brocca et al., 2013) is an algorithm that inverts the soil water balance equation to estimate rainfall. The algorithm is applied to ASCAT SM observations (H SAF products H16 and H101) to obtain H101- and H16-based rainfall estimates that are then merged with the PMW-only H23 product (Marra et al., 2015, Panegrossi et al 2014, 2016, Ciabatta et al., 2017). The algorithm is implemented in the MATLAB programming language. SM2RAIN estimates rainfall starting from SM variations observed by a satellite sensor within a soil layer. The method has been applied to different satellite SM products (Brocca et al., 2014; Koster et al., 2016; Ciabatta et al., 2018; Massari et al., 2017). Moreover, the integration between SM2RAIN-derived and classical rainfall products was found to provide more reliable data (Ciabatta et al., 2015, 2017; Brocca et al., 2016; Tarpanelli et al., 2017) that can be used also for flood modelling (Massari et al., 2014; Ciabatta et al., 2016; Camici et al., 2018) and landslides forecasting (Brunetti et al., 2018). H64 product exploits two different satellite retrieval approaches. The classical top-down approach in which satellites sense clouds from above and the bottom-up approach which is based on SM2RAIN. The two are merged to produce the final rainfall product. The architecture of the H64 product generation chain is shown in Figure 1 and consists of two modules:

- 1. Rainfall estimation via SM2RAIN
- 2. Integration module



Figure 1: Architecture of H64 product generation chain



For more details on H64 algorithm and data processing please refer to the ATBD (Algorithm Theoretical Baseline Document) document.

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3 Validation results: case study analysis

3.1 Introduction

As reported in the Chapter Appendix 1 the common validation methodology is composed of large statistic (multi-categorical and continuous), and case study analysis. Both components are considered complementary in assessing the accuracy of the implemented algorithms. Large statistics helps in identifying existence of pathological behavior, selected case studies are useful in identifying the roots of such behavior, when present.

This Chapter collects the case study analysis performed by PPVG. The Chapter is structured by Country / Team, one section each. The analysis has been conducted to provide information to the User of the product on the variability of the performances with climatological and morphological conditions, as well as with seasonal effects.

Each section presents the case studies analysed giving the following information:

- description of the meteorological event;
- comparison of ground data and satellite products;
- visualization of ancillary data deduced by nowcasting products or lightning network;
- discussion of the satellite product performances;
- indication on the ground data (if requested) availability into the H SAF project.

3.2 Product information

Some main product information are summarized in table.

PRODUCT NAME: P-A	C-SM2RAIN (H64)	
PRODUCT DEVELOPER	Developers:	Contact point:
INSTITUTE:	Ciabatta L. (Leader),	<u>l.ciabatta@irpi.cnr.it</u>
CNR- IRPI	Brocca L.,	luca.brocca@irpi.cnr.it
	Massari C.	<u>c.massari@irpi.cnr.it</u>
OPERATIONAL CHAIN	Responsables:	Contact points:
INSTITUTE:	Zauli F.,	francesco.zauli@aeronautica.difesa.it
COMET	Melfi D.	davide.melfi@aeronautica.difesa.it



3.3 Case study analysis in Belgium (RMI)

3.3.1 Case study: 2 October 2019

PRODUCT NAME	H64
CASE STUDY PERIOD	2 October 2019
CASE STUDY AREA	Belgium
METEOROLOGICAL EVENT	Night showers and thunderstorms
VALIDATION INSTITUTE	RMI
PRODUCT DEVELOPER INSTITUTE	CNR-IRPI
OPERATIONAL CHAIN INSTITUTE	COMET

METEOROLOGICAL EVENT DESCRIPTION

Different low pressure systems from England to Scandinavia are moving eastwards influencing the weather over Belgium. As a results, unstable maritime air masses circulate in western Europe. After the passage of an occlusion during the night of 1 to 2 October, a flow of polar air is established between the high pressures in a vast area from the Azores to Greenland and the low pressures in the eastern parts of Europe.

This situation is summarized in the following maps, including fronts and barometric systems.



Figure 2: Analysis charts valid for 1 (left) and 2 (right) October 2019 at 12:00 and 00:00 UTC respectively

DATA/PRODUCTS USED

Reference data: Belgian Radar located at Wideumont, Wallonia (RMI) Precipitation product H64 (H-SAF/EUMETSAT) Weather charts (MetOffice analyses from Wetterzentrale archive)

RESULTS OF COMPARISON

In Figure 3 we can see the result of upscaling of radar data to the H64 grid on 2 October 2019, together with the corresponding H64 data. The upscaled radar data (i.e., radar data gridded over the satellite grid) serve as basis for comparison between radar and H64. The displayed precipitation values result from cumulation over 24 hours from the previous day at 06:00 UTC up to the current day at 06:00 UTC.



Figure 3: Upscaled radar (left), H64 (right) valid for 2 October 2019.

There is obviously a good agreement between radar and H64 regarding the spatial aspects of the precipitation distribution. However, the product H64 is apparently positively biased by a large margin. Although the number of grid cells exceeding a given amount of precipitation is not much different between the two images, the actual product values are considerably higher than the radar values with an absolute maximum of 36.7 mm for H64 compared to 15.6 mm for the radar.

For the case depicted in Figure 3 we have computed statistical scores shown in the following table.

H64 vs. RADAR	0.25 mm < rr < 1 mm	1 mm < rr < 10 mm	rr > 0.25 mm
ME [mm]	13.72	11.90	12.33
STD [mm]	0.00	7.43	6.99
MB [-]	16.71	3.14	2.86
RMSE [mm]	13.72	14.03	14.17
FSE [%]	15.71	2.52	2.14

H64 vs. RADAR	rr > 0.25 mm	rr > 1 mm
POD [-]	1.00	0.98
FAR [-]	0.00	0.00
CSI [-]	1.00	0.98

 Table 1: Statistical scores obtained from the comparison between H64 and radar acquisition, valid for 2 October 2019.

FINAL COMMENTS

1

In the present case of showers and thunderstorms during the night of 1 to 2 October, there is a good agreement in the spatial features of precipitation between radar and the product H64. However, the product appears consistently positively biased across all precipitation classes.



3.4 Case study analysis in Poland (IMWM-NRI)

3.4.1 Case study: 2 May 2019

PRODUCT NAME	H64
CASE STUDY PERIOD	20 th of May 2019
CASE STUDY AREA	Poland
METEOROLOGICAL EVENT	Severe precipitation events
VALIDATION INSTITUTE	IMWM-NRI
PRODUCT DEVELOPER INSTITUTE	CNR-IRPI
OPERATIONAL CHAIN INSTITUTE	COMET

METEOROLOGICAL EVENT DESCRIPTION

On the 20th of May 2019 one could observe the upper field slow transformation, a slow drift of the heat wedge of 500 hPa over the central and northwest part of European Russia and over the Baltic Sea and a large part of Scandinavia – this drift was not caused by any major pressure systems but was due to the natural heating of the continent. In the middle of that week, the warmer air from the south with Arctic cold were clashing over northern Scandinavia and the White Sea, and the contrast on the front was strongly strengthened, which was being stationary for a few days extending from Labrador through the southern end of Greenland, Iceland, Tromso region, White Sea, Karelia and Northern Ural. In that time, a major part of Europe, including Poland, was in the area of relatively warm, humid air flowing from the eastern and south-eastern countries, it was quite natural to develop convection phenomena in those conditions. The warmer volumes of air and water vapor rised together in a convective thermal chimney reaching a condensation level usually located 800 to 1200 meters above the ground. Convection clouds were developing in the afternoon and evening on that day, being laid out in a clear line of rain squalls, starting from Roztocze region, on the line of Lublin and Siedlce, Ostrołęka, in the night reaching to Podlasie and Mazury regions. The storms appeared in Zamość at 1500, in Lublin at 1600, and in Siedlce at 1700 local time. The width of the zone of the most intense rainfall and storms was ~ 70 km, covering a large part of the Lublin and eastern Mazovia, then Podlasie and Mazury regions. In addition to the intense storms and simulatanous precipitation in those regions (12-16 mm per hour), a local convective precipitation exceeded 30 mm/h. There was also a rapid drop in temperature recorded (from 25 to 15 degrees C) and a lot of squall impacts occured (up to 25 m/s) or more depend on the development of individual Cumulonibus clouds in situ. Thus, the inhabitants of the eastern part of the country did not have a quiet evening and night, the roads were blocked by broken branches and tree trunks, power lines, roofs, advertising panels etc. Sudden and lasting several hours precipitation events resulted in numerous floodings of meadows and fields, especially in Roztocze region and Lubelszczyzna region. The lower Silesia region was covered by a continuous precipitation which moved slowly in the NW direction, during the day this rainfall zone turned over the Lubusz and Western Pomerania regions into a storm area with strong convection, but







3-28-26-24-22-20-18-16-14-12-10-8 -6 -4 -2 0 2 4 6 8 10 12 14 16 18 20 22 24 26 28 30 32

Figure 5: 850 hPa Geopotential chart valid for the 20th of May 2019 at 00 UTC.



Figure 6: The lightning location map presents massive structures present on 20th of May 2019 and reasembling the overall storm layout. The system recorded 53056 CG-; 7039 CG+ and 145165 IC events on that day.

H64 product on 20.05.2019



Figure 7: Accumulated precipitation estimated by H64 product on 20th May, 2019 over Poland.



The lighting localization map shows the horizontal extent of the intensely developing electrical structures in the atmosphere over Poland on the 20th of May 2019 (Figure 6). The H64 cumulative precipitation product (Figure 7) reflects the lightning pattern and the precipi area well.

DATA/PRODUCTS USED

Reference data: Polish lightning detection and localization system PERUN Weather charts (www.wetterzentrale.de)

RESULTS OF COMPARISON

The visual comparison between the H64 product and the radar data field as well as ATS network data field shows a clear match between recorded area of rainfall and its localization.





Figure 9: 24h accumulated precipitation estimated by ATS network data on 20th May, 2019, over Poland.

The maximum intensity of precipitation observed by radar was 13 mm/h, at the same time rain gauge network maxed at 30 mm/h, while H64 recorded a maximum value of 55 mm/h.

FINAL COMMENTS

The H64 product has reflected the rainfall area as well as the overall layout of the precipitable clouds. The map presented above looks very alike the real precipitation over Poland on that day (especially E and N of the country). However, H64 overestimated the rainfall amounts almost twice in reference to the ATS, and over four times in reference to radar. Please note that the radar, ATS and H64 maps above weren't presented to scale [mm/h] which can make the comparison confusing.



3.5 Case study analysis in Slovakia (SHMI)

3.5.1 Case study: 2 September 2019

PRODUCT NAME	H64
CASE STUDY PERIOD	2 Sep 2019 00 UTC – 4 Sep 2019 00 UTC
CASE STUDY AREA	Slovakia
METEOROLOGICAL EVENT	Intense waving cold front over Slovakia
VALIDATION INSTITUTE	Slovak Hydrometeorological Institute (SHMI)
PRODUCT DEVELOPER INSTITUTE	CNR-IRPI
OPERATIONAL CHAIN INSTITUTE	COMET

METEOROLOGICAL EVENT DESCRIPTION

On 1st of September 2019, a ridge of high pressure spread over the Carpathian region from the northeast and on its backside, very warm, originally tropical air from the south to the southeast flowed into our area. At the same time, a cold front, connected by a pressure low with a center above the Norwegian Sea, advanced further east through Germany and the Alps. Behind this cold front a cold air mass began to penetrate over our territory on the next day. However, the weak airflow aloft in combination with the mountains slowed down the progress of the front in our area, and the front began to undulate, which contributed to large temperature differences within the territory as well as heavy rainfall on 2nd and 3rd of September 2019.



Figure 10:Synoptic analyses for 3rd September 2019, at 00:00 UTC over Europe.







Figure 11: MSG Airmass RGB imagery documenting passage of waiving cold front over Slovakia during 2-3 Sep 2019



Figure 12: MSG HRV imagery showing formation of prefrontal thunderstorms over North Western Slovakia on 2 Sep 2019 15:00 UTC (left) and waving front cloudiness bringing stratiform precipitation over Slovakia on 3 Sep 2019 06:00 UTC (right)

DATA/PRODUCTS USED

P-AC-SM2RAIN cumulated precipitation over 24h Meteosat imagery to document synoptic situation Cumulated precipitation from SHMU radars over 24h derived from CAPPI 2km product using quality indices, thresholded to overall radar QI > 0.6 Cumulated precipitation fields from SHMU radars (as above) upscaled into satellite projection

RESULTS OF COMPARISON

Based on visual comparison, the cumulated precipitation fields observed by H64 and radars match relatively well. Especially, in case of 3 Sep 2019 00:00 UTC (Figure 11, upper row), the precipitation structure is captured very well by the H64, except for a false detection of light precipitation in the South East part of Slovakia. Also, the precipitation totals are very similar in this case with an exception of overestimated maxima by the H64 near northern Slovakia-Moravia border. On 4 Sep 2019 00:00 UTC (Figure 11, lower row), the precipitation pattern detected by radars is less clearly recognized in the H64 field and the satellite product slightly overestimated light to medium precipitation.



Figure 13: 24-hour accumulated precipitation fields observed by H64 product (left column), SHMU radars upscaled to satellite grid (middle) and SHMU radars in original resolution (right column). Upper row: on 3 Sep 2019 00:00 UTC, lower row: on 4 Sep 2019 00:00 UTC.

Precipitation class (mm)	1 - 8	8 - 32	32 - 64	64 - 128	≥128	≥1
Number of satellite obs.	412	193	11	0	0	616
Number of radar obs.	296	121	2	0	0	419
Mean error (mm/h)	2.768	0.151	-0.785	-	-	1.995
Multiplicative bias	1.674	1.011	0.977	-	-	1.283
Correlation coefficient	0.341	0.617	-1.0	-	-	0.685
URD-RMSE (%)	176.5	58.7	38.1	-	-	151.7
Fractional Standard Error	113.4	53.3	38.2	-	-	79.9
(%)						
Nash-Sutcliffe	-4.131	-0.556	-45.982	-	-	0.117
Coefficient						

Results of statistical comparison of the satellite product with radars are shown in the following tables.

 Table 2: Selected scores of continuous statistics

Precipitation threshold (mm)	≥1	≥ 8
POD	0.967	0.724
FAR	0.342	0.564
CSI	0.643	0.374

 Table 3: Selected scores of dichotomous statistics

Both Mean Error and Multiplicative Bias confirm observed slight overestimation of light precipitation by the H64, especially for class 1-8 mm (see Table 1). Medium precipitation (class 32-64 mm) by the H64, on the other hand, exhibit slight underestimation. Results for this precipitation class, however, might be influenced by very low number of observations.

The POD in this case reached high values, especially for overall precipitation $AP \ge 1$ mm with a value very close to 1 (Table 2). The FAR reached a little bit less excellent results, especially for precipitation



threshold of 8 mm. The CSI score reached relatively high value for the 1 mm threshold but only less than 0.4 in case of the 8 mm threshold.

COMMENTS

The upscaled radar precipitation fields shown in this case study are limited to area with radar quality index higher than 0.6. Only the data lying inside this area are used for calculation of the statistical results.



4 Validation results: long-term analysis

4.1 Overview

Product ID (Acronym)	H64 (P-AC-SM2RAIN)			
Product name	Precipitation/Soil Moi	Precipitation/Soil Moisture integrated produ		
Algorithm version	Latest version:		1.0	
number	Version considered for	r Q.A.:	1.0	
Covered period	01/01/2019 - 31/12/2	2019		
Q.A. methods applied	Continuous statistics	ME, SD, MAE, MB, RMSE, CC, FSE		
	Multi-categorical	POD, FAR, CSI		
	statistics			
	Contributing	BE, BU, DE, HU, IT,	PL, SK, TU	
	countries			

The validation has been performed over the full 1-year period between January and December 2019 by European countries belonging to PPVG and supervised by DPC. The product release currently in force at the time of writing has been evaluated. The results are showed both for European (EU) area in comparison with ground (radar and rain gauge) data (section 4.2), and over the MSG Full Disk (FD) area using the Triple Collocation (TC) methodology (section 4.3). The surface type classification is taken in account (land, sea and coast areas) as well as the three precipitation classes (as defined in <u>Table 17</u>), even if only land areas are considered to assess the quality of this product. The validation procedure evaluates <u>only high-quality precipitation values (for both reference and satellite data)</u>. Satellite Field Of Views not fully covered by reference data (or with percentage of coverage less than 50%) are discarded by the Q.A. procedure in order to increase the significance of the statistical sample.

4.2 Validation results over EU area

Validation has been performed using rain gauge and radar data as reference in Belgium, Bulgaria, Germany, Italy, Hungary, Poland, Slovakia and Turkey, for the one-year period as above descripted. Each institute has used the Unique Common Code (UCC Version 2.0) developed by PPVG to evaluate every H SAF precipitation product in terms of different statistical scores. More than 100 million of ground and satellite pairs of data were analyzed. For H64 product, the UR depends by FSE score as indicated in <u>Table 4</u>.

Precipitation class	Threshold	Target	Optimal
≥1 mm/daily	200	150	100

Table 4: Accuracy requirements for product P-AC-SM2RAIN in term of FSE(%).

<u>The FSE score</u> (mathematical formula shown in <u>Table 18</u>) <u>defines the accuracy</u> of the H SAF product under analysis.

4.2.1 Monthly accuracy

The monthly accuracy of all daily accumulated precipitation estimates computed for H64 in comparison with radar, rain gauge and overall ground observations is shown in Figure 14 in panel a,



b and c respectively. In every panel, the background color indicates the region inside each requirement accuracy threshold as defined in <u>Table 4</u> and also used in <u>Table 8</u>. The black dotted line represents the mean FSE value over the full period (12 months). Panel d shows the percentage contribution for different instruments and observation's surfaces respect to the full dataset.

P-AC-SM2RAIN shows unstable monthly results. Scores are between 130% and 290% with better values respect to radar data (below the threshold). The overall results indicate an accuracy slightly above 200%, marginally above the threshold. The percentage contribution highlights how gauge represent the most part of the full dataset (79%).





4.2.2 Monthly continuous statistical scores

In Figure 15 the monthly trend for H64 product respect to radar over land (green line) and rain gauge (blue line) for all precipitation amounts above 1 mm/day is shown for six continuous statistical scores (ME, SD, MAE, MB, RMSE and FSE) indicated in section 0. Results for both radar and gauge have a similar trend: an overestimation above 5 mm daily is evident in almost of months. The MB also shows on average a factor two of overestimation. Finally, the FSE shows as radar and gauge results are similar with values (170% and 225%, respectively) around the threshold fixed to 200%. More continuous statistical scores are shown in Appendix.





Figure 15: Monthly continuous statistical scores resulting for H64 relatively to the accumulated precipitation above 1 mm/daily.

4.2.3 Multi-categorical statistics

The multi-categorical statistics are below shown. In each table, the first column indicates the precipitation classes of the satellite product, while along the columns are reported the ground precipitation classes. The first class detects the very low accumulated rain class with PC<1 mm/24h, the second class identifies the accumulated rain between $1 \le PC \le 10 \text{ mm}/24h$, while the last one classifies the higher accumulated rains (PC $\ge 10 \text{ mm}/24h$).



The H64 product doesn't fine detect the no-rain area: in less than 25% of cases over land the no rain area is well detected. The success of detection increases for higher accumulation: light rain is detected in 52% of cases with respect gauge measurements and 49% with radar ones.

Finally, higher accumulation amounts areas are well identified in 70% of cases in comparison with both gauge and radar measurements.

More multi-categorical statistics are shown in Appendix.

Radar Land				
Multi-Categorical Statistics				
	< 1 mm/day	[1 – 10 [mm/day	≥10 mm/day	
< 1 mm/day	21%	11%	4%	
[1 - 10[mm/day	66%	49%	26%	
≥10 mm/day	13%	40%	70%	

 Table 5: Multi-categorical table for product H64 – radar validation over land.

The precipitation classes along the columns (rows) are relative to ground (satellite) precipitation.

Gauge Land				
Multi-Categorical Statistics				
	< 1 mm/day	[1 - 10[mm/day	≥10 mm/day	
< 1 mm/day	23%	10%	4%	
[1 - 10[mm/day	66%	52%	27%	
≥10 mm/day	12%	38%	69%	

Table 6: Multi-categorical table for product H64 – rain gauge validation over land. The precipitation classes along the columns (rows) are relative to ground (satellite) precipitation.

Overall				
Multi-Categorical Statistics				
	< 1 mm/day	[1 - 10[mm/day	≥10 mm/day	
< 1 mm/day	22%	10%	4%	
[1 - 10[mm/day	66%	52%	27%	
≥10 mm/day	12%	38%	69%	

Table 7: Multi-categorical table for product H64 – Overall validation.

The precipitation classes along the columns (rows) are relative to ground (satellite) precipitation.

4.2.4 Product requirement compliance

I

The accuracy of H64 for the 1-year validation period (January – December 2019) respect to ground measurements and overall results are reported in <u>Table 8</u>.

The accuracy with radar estimates is below the threshold, while overall results exceed – slightly - the threshold (206%). The resulting overall result is slightly above the threshold required. Furthermore, the FSE is particularly sensitive to the low precipitations which constitute the weak point of H64. In fact, already above the ground threshold of 5 mm per day, the FSE is reduced to only 124% (as observable in Appendix).

H S	SAT	Pro	Product Validation Report - PVR-64 (Product H64 – P-AC-SM2RAIN)			Doc. No: SA Date: 27/02 Page: 28/90	AF/HSAF/PVR-64 2/2022 D	
Between ta optimal	rget and	Betwee and targ	tween threshold Threshold exceeded d target by < 50 %		Threshold exceeded by ≥ 50 %			
H64 vs Ground			Annual average of FSE (%)					
Precipitation	Requi	rement	(FSE %)	%) Reder (Land) Caus				
Class	thresh	target	optimal	r	Radar (Land) G		ge (Land)	OVERALL
> 1 mm/dav	200	150	100		170		224	206

Table 8: Product requirement and compliance analysis for product H64

Moreover, the correlation coefficient between H64 estimates and ground data (radar, gauge and overall) as shown below in Figure 16 highlights good performances with median values above 0.5 with gauge and above 0.6 in comparison with radar data. Percentages of data with CC above the three thresholds are also indicated in figure.



GAUGE RADAR GAUGE+RADAR Figure 16: Boxplot for correlation coefficient computed for H64 product in comparison with EU ground data (gauge, radar and overall). Percentage of grid points with CC above 0.5, 0.65 and 0.8 are also indicated.



4.3 Validation results over FD area

The validation procedure over FD area has been developed by the IRPI-CNR and conducted by Italian DPC. TC methodology is applied between P-AC-SM2RAIN, GPCC and GLDAS datasets over the overlapped domains. The accuracy of the satellite product is evaluated by the CC score: results for different areas and on monthly basis are here shown. More maps over the FD area are shown in Appendix.

4.3.1 Monthly accuracy

The UR score used in the TC methodology is the CC. The monthly trend obtained for the triplet of datasets used (daily accumulated precipitation data) is shown in the <u>Figure 17</u>. The blue bars indicate the results for the H SAF satellite product (P-AC-SM2RAIN); in green those obtained from GPCC data, while in orange those from GLDAS dataset. The colored background highlights the areas where thresholds are set for the chosen UR.

Focusing on the results from SEVIRI, the values range from 0.54 for the worst month (January 2019) to 0.61 for the best months (May 2019). The average value calculated on an annual basis is 0.61, which is quite close to the target.

GPCC obtains comparable results to H SAF product, with a very similar monthly trend.

GLDAS, on the other hand, has results in terms of CC that are lower than the previous ones.



igure 17: Monthly validation results (January - December 2019) in terms of CC (UR) score obtained by the P-AC-SM2RAIN (daily) product (blue bars) using the TC methodology. Background colours highlight the UR thresholds.



4.3.2 Multi-categorical statistics

In the following tables, the percentages of corrected estimates in three different precipitation classes are indicated. The percentages are normalized along each column that represent the reference for the precipitation (GLDAS and GPCC for the first two tables respectively). The first class detects the very low accumulated rain class with PC<1 mm/24h, the second class identifies the accumulated rain between $1 \le PC \le 10 \text{ mm}/24h$, while the last one classifies the higher accumulated rains (PC $\ge 10 \text{ mm}/24h$).

Comparing the first two tables in which P-AC-SM2RAIN estimates are classified respect to GLDAS and GPCC dataset respectively, it is possible to note better agreement with the first respect to second one, particularly for medium and higher precipitation class: 50% vs 39% and 43% vs 38%. Generally, the precipitation is well detected in all precipitation classes, more confident for very low accumulates (82%).

Overall					
Multi-Categorical Statistics					
P-AC-SM2RAIN vs GLDAS	< 1 mm/24h	[1 - 10[mm/24h	≥10 mm/24h		
< 1 mm/24h	82%	32%	15%		
[1 - 10[mm/24h	12%	50%	42%		
≥10 mm/24h	6%	18%	43%		

 Table 9: Multi-categorical table for product P-AC-SM2RAIN versus GLDAS over 24h – Overall FD validation. The precipitation classes along the columns (rows) are relative to GLDAS (H SAF) precipitation.

Overall					
Multi-Categorical Statistics					
P-AC- SM2RAIN vs GPCC	< 1 mm/24h	[1 - 10[mm/24h	≥10 mm/24h		
< 1 mm/24h	82%	45%	25%		
[1 - 10[mm/24h	13%	39%	37%		
≥10 mm/24h	5%	16%	38%		

 Table 10: Multi-categorical table for product P-AC- SM2RAIN versus GPCC over 24h – Overall FD validation. The precipitation classes along the columns (rows) are relative to GPCC (H SAF) precipitation.

In the table below, the percentages of estimates in agreement between the two datasets GLDAS and GPCC for three different precipitation classes are shown. The percentages are normalised along each column represented by GPCC for the first columns and GLDAS for the second columns in each class, respectively.

The agreement is very good for lower precipitation class (97-94%), but it decreases by increasing the precipitation accumulates. In fact, the agreement for the intermediate accumulated class ranges between 34% and 47%, while for higher accumulates ranges between 31 and 35% with respect GPCC and GLDAS, respectively.

Overall

Multi-Categorical Statistics

H SAF	Produ (Pr	ict Validatior oduct H64 – F	n Report - PV P-AC-SM2RAIN	R-64 Do Da I) Pa	c. No: SAF/HSA te: 27/02/2022 ge: 31/90	F/PVR-64
GLDAS vs GPCC and vice versa	< 1 m	< 1 mm/24h		mm/24h	≥10 m	um/24h
< 1 mm/24h	97%	94%	56%	40%	31%	18%
[1 - 10[mm/24h	3%	6%	34%	47%	38%	47%
≥10 mm/24h	0%	0%	10%	13%	31%	35%

 Table 11: Multi-categorical table for GLDAS versus GPCC (and vice versa) over 24h Overall FD validation. The precipitation classes in the first columns are relative to GLDAS dataset; in the second columns are relative to GPCC dataset.

Generally, the agreement is better in comparison with GLDAS: this is true also for validation with P-AC-SM2RAIN product. It is important to note that the best agreements are reached by the H SAF product estimates in comparison with the GLDAS dataset for medium (50%) and higher (43%) precipitation accumulates.

4.3.3 Product requirement compliance

User requirements and compliance analysis for H SAF product are reported below. The UR result for P-AC-SM2RAIN product is satisfactory with value equal to 0.61 that agree to index between threshold and target.

Between target and	Between threshold	Threshold exceeded	Threshold exceeded by \geq
optimal	and target	by <50%	50%

P-AC-SM2RAIN by TC	Annual average of CC			
Precipitation Class	Requirement (CC)		OVERALL	
	thresh	target	optimal	
≥0 mm daily	0.50	0.65	0.80	0.61

 Table 12: UR and compliance analysis for P-AC-SM2RAIN (24h) product over FD area by TC methodology.



5 Conclusions

The H64 (P-AC-SM2RAIN) product has been validated by the PPVG over one year of data (January - December 2019). Each Country/Team has provided long statistics analysis and case studies using ground data (radar and rain gauge) as reference following the common validation methodology reported in section Appendix 1. A huge effort has been made in the development of the Unique Common Code (UCC) used by all members of the validation cluster to improve it. The use of a UCC guarantees that the results obtained by every partner are obtained in the same way.

In addition, a validation outside the European area was performed. The Quality Assessment of H64 product was also evaluated over the full MSG disk coverage area using the TC methodology. This new validation procedure over the FD area was developed by the CNR-IRPI and results were managed by Italian DPC and reported in this document.

The Q. A. procedure takes into account the status of the background surface (including snow or frozen soil), the precipitation regime (deep convective, convective, stratiform and light stratiform) and the quality values by ground/reference datasets and by satellite estimates. Precipitation with poor and missing quality flags are rejected by the procedure that correspond to data seriously corrupted not suitable for validation.

Each Country/Team has independently evaluated the quality of its radar data to optimize the satellite data validation. Much effort has been made to define a standard radar quality data. Rain gauge quality data has been standardized to all members by use of the same interpolation method (GRISO).

Different case study analysis of H64 have been reported in section 3. Stratiform and convective precipitations during summer and winter periods have been analysed in different geographical regions. Rain gauges with high-frequency refresh time, radar data and nowcasting tools have been used to highlight different characteristics of the satellite product. The case studies proposed have pointed out as H64 product often appears positively biased mainly for light precipitation.

In section 4 statistical analysis obtained for the entire 1-year period have been presented. To assess the degree of compliance of the product with user requirements, each Country/Team has provided the monthly statistical scores. The results have been showed using both European ground (radar and rain gauge) data over land areas and by TC methodology using measurements provided by GLDAS and GPCC projects.

The results of the Precipitation Validation Programme are reported in this Product Validation Report (PVR). H64 meets the target accuracy requirement using the TC methodology over FD area, while exceeds slightly the threshold in comparison with ground data over EU area. This analysis highlights as H64 product tends to overestimate precipitation, mainly low accumulated rates. However, the correlation coefficient indicates a good agreement with respect European ground data but also respect with global datasets. Finally, contingency tables show as the precipitation areas are good observed in all events, even if the detection of lower accumulated estimates is not always accurate and represents the main weakness of this product, as also highlighted in most of case studies provided.



END MAIN DOCUMENT



Appendix 1 Validation strategy, methods and tools

The quality assessment procedure, methodologies and instruments used to assess the performances of precipitation products are described in this chapter.

A1.1 Validation team and work plan

To evaluate the satellite precipitation product accuracy, a Validation Group has been established by the beginning of the Validation Phase in the H SAF project. The Precipitation Product Validation team is composed of experts from the National Meteorological and Hydrological Institutes of Belgium, Bulgaria, Germany, Hungary, Italy, Poland, Slovakia, and Turkey (<u>Table 13</u>). Hydrologists, meteorologists, and precipitation ground data experts, coming from these countries are involved in the product validation activities (<u>Table 14</u>).



Table 13: Structure of the Precipitation products validation team.
--

Validation team for precipitation products							
Name	Institute	Country	e-mail				
Silvia Puca (Leader)	Civil Protection Department (DPC)	Italy	silvia.puca@protezionecivile.it				
Marco Petracca	Civil Protection Department (DPC) National Research Council (CNR - ISAC)	Italy	M.Petracca@isac.cnr.it				
Alexander Toniazzo	Civil Protection Department (DPC)	Italy	alexander.toniazzo@protezionecivile.it				
Alessandra Mascitelli	Civil Protection Department (DPC)	Italy	alessandra.mascitelli@protezionecivile.it				
Gianfranco Vulpiani	Civil Protection Department (DPC)	Italy	gianfranco.vulpiani@protezionecivile.it				
Emanuela Campione	Civil Protection Department (DPC)	Italy	emanuela.campione@protezionecivile.it				
Pierre Baguis	Royal Meteorological Institute of Belgium (RMI)	Belgium	Pierre.Baguis@meteo.be				
Emmanuel Roulin	Royal Meteorological Institute of Belgium (RMI)	Belgium	Emmanuel.Roulin@meteo.be				
Eram Artinyan	National Institute of Meteorology and Hydrology (NIMH)	Bulgaria	eram.artinian@meteo.bg				
Petko Tsarev	National Institute of Meteorology and Hydrology (NIMH)	Bulgaria	petko.tsarev@meteo.bg				
Georgy Koshinchanov	National Institute of Meteorology and Hydrology (NIMH)	Bulgaria	georgy.koshinchanov@meteo.bg				
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Peter Krahe	Bundesanstalt für Gewässerkunde (BfG)	Germany	krahe@bafg.de				
Márta Diószeghy	Hungarian Meteorological Service (OMSZ)	Hungary	dioszeghy.m@met.hu				
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Federico Porcu'	Department of Physics and Astronomy, University of Bologna (UniBo)		Italy	federico.porcu@unibo.it			
Bozena Lapeta	Institute of Meteorology and Water Management (IMWM)		Poland	Bozena.Lapeta@imgw.pl			
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Ladislav Méri	Slovensk (SHMÚ)	ý Hydrometeorologický Ústav	Slovakia	ladisl	ladislav.meri@shmu.sk		
Abmot Öztopal	Ictanbul	Turkov	ortor	al@itu.adu.tr			

Table 14: List of the people involved in the validation of H SAF precipitation products (PPVG)

The Precipitation products validation programme started with a first workshop in Rome, 20-21 June 2006, soon after the H SAF Requirements Review (26-27 April 2006). The first activity was to lay down the Validation plan, that was finalised as first draft early as 30 September 2006. After the first Workshop, other ones followed, at least one per year to exchange experiences, problem solutions and to discuss possible improvement of the validation methodologies. Often the Precipitation Product Validation workshop are joined with the Hydrological validation group.

The results of the Product Validation Programme are reported in this Product Validation Report (PVR) and are published in the validation section of the H SAF web page. A new structure and visualization of the validation section of H SAF web page is in progress to consider the user needs. This validation web section is continuously updated with the last validation results and studies coming from the Precipitation Product Validation Group (PPVG).

A1.2 Validation objects and issues

The products validation activity has to serve multiple purposes:

- to provide input to the product developers for improving calibration for better quality of baseline products, and for guidance in the development of more advanced products;
- to characterise the product error structure in order to enable the Hydrological validation programme to appropriately use the data;
- to provide information on product error to accompany the product distribution in an open environment, after the initial phase of distribution limited to the so-called "beta users".

Validation is a challenging task in the case of precipitation, both because the sensing principle from space is very much indirect, and because of the natural space-time variability of the precipitation field (sharing certain aspects with fractal fields), that poses severe sampling problems.

It is known that an absolute 'ground reference' does not exist. In the H SAF project the validation is based on comparisons of satellite products with **European ground data**: radar, rain gauge and radar integrated with rain gauge. During the Development phase some main problems have been pointed out. First of all, the importance to characterize the error associated to the ground data used by PPVG.



Secondly to develop software for all steps of the Validation Procedure, a software available to all the members of the PPVG. The radar and rain gauge Working Group (WG) have been composed to solve these problems.

In CDOP-3, with the release of more than 30 products **over the MSG full disk area**, the Validation Cluster had to develop new methodologies to compare precipitation estimates on almost global area coverage. The Associated Scientist analysis (H_AS16_03 DPC/CNR-ISAC 2016) has been identified the DPR (Dual-frequency Precipitation Radar) onboard of GPM-CO (Global Precipitation Measurement – Core Observatory) satellite as worthy instrument reference for the estimation of instantaneous precipitation on a global scale. In particular, the 2A-DPR NS V05 (**DPR**) was considered as most suitable product for potential use within the H SAF Precipitation Product Validation activity **for instantaneous precipitation estimates.** For more details, refer to Sebastianelli, 2017.

For accumulated precipitation products, instead, the **TC** (Brocca et al., 2014) methodology was used to perform the validation activity. TC requires the simultaneous availability of three products with mutually uncorrelated errors with similar spatial coverage, resolution and accumulation time.

In the following three sections, the validation methodologies and data used as reference to perform the comparisons are described.

A1.3 Validation methodology respect to GROUND reference data

From the beginning of the project it was clear the importance to define a common validation procedure in order to make the results obtained by several institutes comparable and to better understand their meanings. The main steps of this methodology have been identified during the development phase inside the validation group, in collaboration with the product developers, and with the support of ground data experts. This common procedure has given rise to a single common code for all members of the PPVG, named Unique Common Code (UCC). This common validation methodology is based on ground data (radar and rain gauge) comparisons to produce **large statistic** (multi-categorical and continuous), and **case study analysis.** Both components (large statistic and case study analysis) are considered complementary in assessing the accuracy of the implemented algorithms. Large statistics helps in identifying existence of pathological behaviour, selected case studies are useful in identifying the roots of such behaviour, when present.

The main steps of the validation procedure are:

- ground data error analysis: radar and rain gauge;
- point measurements (rain gauge) spatial interpolation;
- up-scaling of radar data versus satellite grid (radar data gridded on satellite grid);
- temporal comparison of precipitation products (satellite and ground);
- statistical scores (continuous and multi-categorical) evaluation;
- case study analysis.

Ground data and tools used for validation

Both rain gauge and radar data have been used for H SAF product validation. Working groups have been set up to solve specific items in the validation procedure and to develop a common software. A complete knowledge of the ground-based data characteristics, used within the PPVG, was the first


step necessary to define the procedure to select the most reliable data (ground reference) and to understand the validation results.



Figure 18: The network of 8,404 rain gauges used for H SAF precipitation products validation

The rain gauge networks of PPVG is composed of approximately 8400 stations across 8 Countries (Figure 18). A key characteristic of such networks is the distance between each raingauge and the closest one, averaged over all the instruments considered in the network and it is a measure of the raingauge density. Instruments number and density are summarized in the following Table 3.

Country	Total number of gauges *	Average minimum distance (km)
Belgium	92	15.2
Bulgaria	123	25.2
Germany	2,299	12.9
Hungary	270	17.0
Italy	2,934	11.3
Poland	540	24.0
Slovakia	911	13.6
Turkey	1,235	26.5

* the number of raingauges could vary from day to day due to operational efficiency within a maximum range of 10-15%.

Table 15: Number and density of raingauges within H SAF validation Group

Most of the gauges used in the National networks by the PPVG Partners are of the tipping bucket type, and hourly cumulated.



74 C-band radars (Figure 19) are used by the H SAF PPVG for assessing the satellite product accuracy. An inventory on radar data networks and products used in PPVG has pointed out that all the institutes involved in the PPVG declared the system are kept in a relatively good status and all of them apply some correction factors in their processing chain of radar data. Only the radar data, which passes the quality control of the owner Institute, are used by the PPVG for validation activities. Please note that the Validation procedure is the same for all countries of PPVG.



Figure 19: The networks of 74 C-band radars used by the H SAF PPVG. Note1: Turkish radars are not used in validation activities. Note2: Only one out of four belgian radars is shown.

Instruments number and average minimum distance in each country are summarized in Table 16.

Country	Total number of radar	Average minimum distance (km)
Belgium	4	120
Bulgaria	-	-
Germany	16	163
Hungary	4	190
Italy	22	141
Poland	8	186
Slovakia	4	137
Turkey	16*	253

 Table 16: Number and density of radars used by the H SAF PPVG.

 * Not used in validation activities.

Common procedure for the validation



The UCC developed by PPVG during CDOP-2 and improved in CDOP-3 has been used to validate satellite data respect to radar and rain gauge data considered as ground reference.

Common procedure for the validation with RADAR data

Selection of satellite pixels falling into the region of interest

In order to avoid time-consuming useless calculation, every country restricts the validation to a specific Area of Interest (normally the area covered by the RADAR data of the country), which is detected implicitly by the common validation algorithm.

Taking into account quality index information

The UCC considers the quality index for each radar pixel. It depends mainly by distance from radar site and by interferences or beam-blocking. Each country independently calculates the quality of its data. This quality information is used for validation purposes since CDOP2.

Selection of the RADAR data time-synchronous with the satellite ones

The RADAR data whose temporal characteristics are congruent with the data of the satellite product to be compared (instantaneous or mean value or daily cumulated) are selected. For instantaneous acquisitions, the maximum time difference allowed for satellite comparison is 10 minutes.

Up-scaling of RADAR data at the product satellite resolution

Radar data, whose spatial resolution $(0.25 \div 1 \text{ km})$ is typically greater than satellite products $(5 \div 25 \text{ km})$, is rescaled to the satellite product grid (we refer it as "radar upscaled"). The information on the radar data density for each grid point is kept in order to eliminate those grid points that are not representative of the radar data (whose spatial coverage is limited or less than 50%).

Calculation of corresponding satellite and RADAR rainfall values

For each single satellite file, a separate up-scaling procedure reads the look up table and assigns to each satellite pixel the RADAR rainfall average calculated from the values of the radar pixels belonging to the satellite pixel in the look-up table.

Averaging is simply arithmetical as investigations so far have shown that the averaging method does not have an impact on the statistical scores.

The flag indicating if the satellite pixel is coast, land or sea is matched to each satellite-radar data pair calculated in this step.

Common procedure for the validation with RAIN GAUGE data

Selection of satellite pixels falling into the region of interest:

In order to avoid time-consuming useless calculation, every country restricts the validation to a specific Area of Interest (normally the area covered by the rain gauge data of the country), which is detected implicitly by the common validation algorithm.

Selection of rain gauge data synchronous with the satellite ones

Gauges with different cumulation intervals are considered, and if the interval is longer than the time resolution of the product (15 or 30 minutes), more satellite images are averaged. For H68 product two consecutive images (2 x 30' every image) are summed for comparing with one-hourly accumulated gauges.

Interpolation of the rain gauge data:



All partners of the Validation Group have been used the same interpolation technique, named GRISO (a like-Kriging interpolation technique for rain gauge data, Pignone et al. 2010; Feidas et al. 2018), to get spatially continuous rainfall maps (over 5x5 km grid) from individual gauge measurements.

The GRISO technique is the interpolation method chosen for the common validation.

Taking into account quality index information

GRISO technique produces a quality index map for each instantaneous acquisition as function of distance from each rain gauge station. This quality information was used for validation purposes in CDOP2.

Matching between satellite and rain gauge data:

The satellite data is matched with the rain gauge interpolated grid using the nearest-neighbor method.

A1.4 Validation methodology of INstantaneous precipitation products over Full Disk coverage

As the validation with respect to ground data, even this methodology was developed in communion with European experts belonging to the VC. But, differently from the first one, this is not performed by all countries but only by Italian DPC because all DPR products over global area are freely available from GPM website. All instantaneous satellite precipitation products, with extension area over FD area, are evaluated following these main steps:

- regridding of DPR and H SAF data versus a regular 0.5° equi-distance grid;
- temporal and spatial matching between precipitation products;
- statistical scores (continuous and multi-categorical) evaluation;

The methodology, as the previous one, produces <u>large statistic</u> (multi-categorical and continuous) scores.

DPR products used for validation

The spatial coverage of both rain gauge and ground radar networks is not suitable to detect precipitation on a global scale. At the contrary, satellite observations provide estimates on a synoptic scale, although there are some issues related to their accuracy. It was discussed in the Visiting Associated analysis in comparison with ground radar network (Sebastianelli, 2017). The DPR is a Dual frequency Precipitation Radar located on board of the GPM Core Observatory (Figure 20). It uses the Ka (~35 GHz) and Ku bands (~13 GHz) to construct three-dimensional precipitation and drop size distribution maps. The GPM Core Observatory (Figure 20) flies in a non-sun-synchronous orbit at 65° inclination to cover a larger latitudinal extension with respect to the TRMM orbit, which extended from 35°S to 35°N. Both Ku- and Ka-band radars perform cross-track type scans (perpendicular to the direction of the satellite motion) estimating the precipitation during the day and the night over land and ocean. The Ku-band radar performs a normal scan (NS) acquisition mode that is composed by 49 footprints (IFOV) of 5 km in diameter. In fact, away from the scanning center, footprints tend to widen and overlap (edge effects) because of a geometric distortion. The term swath indicates the width of each scan of 245 km. The range resolution is 250 m. The Ka-band radar can perform a matched scan (MS) or a high sensitivity scan (HS) acquisition mode. The MS footprints match the central 25 footprints of the Ku-band and the range resolution is 250 m. Therefore, MS scan is composed of 25 footprints of 5 km in diameter and the swath is 125 km. When Ka-band radar



operates in HS mode footprints are interlaced with the matched beams, the range resolution is 500 m and there are 24 footprints along a swath. Figure shows the different DPR scanning modes with respect the flight direction.

It must to be noted that the range resolution is different from the spatial resolution. In fact, the sampling is carried out for 19 km above the sea level and then along the vertical there are many footprints of 250 m height (range resolution). In addition, footprint size decreases as the sampling height increases due to the antenna aperture. The sampling distance between the centers of two adjacent footprints is 5.2 km, and it is constant throughout the scan to the edges. Apart the other problems which affects the DPR estimates, the main issues deal with the attenuation and the ground clutter. The K-band radar estimates are affected by attenuation when they sample through very intense precipitations (convective cells). Ground clutter is a non-meteorological echo which causes an overestimate of precipitations.

DPR products (level 2A) referred to single frequency radar are 2A-Ku, 2A-Ka-MS and 2A-Ka-HS, as showed in Figure 21. Three different DPR products combining Ka and Ku bands precipitation rate estimates (prEs) also exist depending on the IFOV to which data are referred. The IFOV can be related to the NS Ku-band, or to the MS or HS Ka-band, and the corresponding DPR products for prEs are 2A-DPR-NS, 2A-DPR-MS and 2A-DPR-HS, respectively. Results of Visiting Associated activity (Sebastianelli, 2017) highlight as 2A-DPR-NS product performs better with respect to ground-based radar estimates. For this reason, the prEs by 2A-DPR-NS product (hereafter also referred as DPR-NS) was used as precipitation reference to validate the H SAF satellite precipitation products.



Figure 20: The GPM Core Observatory and the GMI and DPR ground tracks.



whereas matched and high sensitivity scans are performed by Ka-band radar.

Common procedure for the validation with DPR-NS

The comparison between DPR-NS and H SAF precipitation products is performed following the procedure below described.

Selection of synchronous data (first check)

Both DPR and H SAF filenames contain the start and finish time. Only files with consistent time intervals are considered and evaluated to avoid time-consuming useless computation.

Data re-gridding over regular grid

Both, DPR and H SAF data are re-gridded over the same equi-distance 0.5° grid. All satellite parameter (continuous) values are averaged over the new grid cell. For discrete parameters (such as sea/coast/land flag or precipitation phase flag) the most frequent value is considered.

Temporal and spatial matching between the two regridded data

DPR and H SAF gridded data are temporal- and spatial- matched. Only overlapping grid cells with maximum time difference within ± 15 minutes (from the average time of validity of the H SAF product) are stored and evaluated for statistical score analysis.

A1.5 Validation methodology of ACcumulated precipitation products over Full Disk coverage



In order to assess the quality of accumulated precipitation products over FD area the new TC methodology analysis was successfully tested by Brocca et al. (2014). This technique requires three wholly independent rainfall datasets on global scale without the availability of ground-based rainfall accumulation data.

Given three estimates of the same variable, the main assumptions of the TC method are the (i) stationarity of the statistics, (ii) linearity between the three estimates (vs. the same target) across all timescales and (iii) existence of uncorrelated error between the three estimates.

TC provides error and correlations of three products if each of the ones is afflicted by mutually independent errors.

The main steps of the TC validation procedure are:

- temporal (daily) and spatial $(0.5^{\circ} \times 0.5^{\circ} \text{ grid})$ matching between three precipitation datasets;
- TC main procedure;
- statistical scores (continuous and multi-categorical) evaluation;

The methodology, as the previous one, produces <u>large statistic</u> (multi-categorical and continuous) scores.

For more details on the TC main procedure, see the H SAF Visiting Scientist Final Report titled "Leveraging coincident soil moisture and precipitation products for improved global validation of satellite-based rainfall products" by Chen F. and Crow W. T. : <u>https://hsaf.meteoam.it/VisitingScientist/GetDocument?fileName=Final_Report_H_AS18_04.pdf</u>

Data and products used

In this analysis, TC is applied to the rainfall accumulation estimates derived from H SAF satellite product and by GPCC (Global Precipitation Climatology Centre, <u>https://climatedataguide.ucar.edu/climate-data/gpcc-global-precipitation-climatology-centre</u>) and GLDAS (Global Land Data Assimilation System, <u>https://ldas.gsfc.nasa.gov/gldas</u>) projects.

GPCC

The GPCC provides gridded gauge-analysis products derived from quality-controlled station data. In this analysis is used the GPCC First Guess Daily Product (hereafter named as GPCC): daily global land-surface precipitation based on the station database (SYNOP) available via the Global Telecommunication System (GTS) of the World Meteorological Organization (WMO) at the time of analysis (3 - 5 days after end of the analysis month). This product contains the daily totals for a month on a regular latitude/longitude grid with a spatial resolution of 1.0° x 1.0° latitude by longitude. Interpolation is made for the daily relative quota of the monthly total, i.e., the daily total divided by the monthly total, the latter has the DOI:10.5676/DWD GPCC/FG M 100. The temporal coverage of the dataset ranges from January 2009 to the most recent month for which GTS based SYNOP data is available, i.e. the previous month, 3-5 days after its completion. In Figure 22 is shown an example of spatial extension and resolution of the GPCC product.



(Product H64 – P-AC-SM2RAIN)

GPCC Monitoring Product Gauge-Based Analysis 1.0 degree precipitation anomaly for May 2012 in mm/month (deviation from normals 1951/2000) (grid based)



Figure 22: Example of data available in GPCC. The daily precipitation data used in TC methodology have the spatial extent as shown in the figure and the spatial resolution of 1.0 degree

GLDAS

NASA Global Land Data Assimilation System Version 2 (GLDAS-2) has three components: GLDAS-2.0. GLDAS-2.1, and GLDAS-2.2. GLDAS-2.0 is forced entirely with the Princeton meteorological forcing input data and provides a temporally consistent series from 1948 through 2014. GLDAS-2.1 is forced with a combination of model and observation data from 2000 to present. GLDAS-2.2 product suites use data assimilation (DA), whereas the GLDAS-2.0 and GLDAS-2.1 products are "open-loop" (i.e., no data assimilation). The choice of forcing data, as well as DA observation source, variable, and scheme, vary for different GLDAS-2.2 products.

The 3-hourly data product was simulated with the Noah Model 3.6 in Land Information System (LIS) Version 7 (https://lis.gsfc.nasa.gov/sites/default/files/LIS/public_7_3_releases/LIS_usersguide.06-dec-2021.pdf). The data product contains 36 land surface fields from January 2000 to present. In this analysis GLDAS-2.1 data are used. These data are archived and distributed in NetCDF format (DOI:10.5067/E7TYRXPJKWOQ). In Figure 23 is shown an example of spatial extension and resolution of the GLDAS product used.





Figure 23: Example of data available in GLDAS. The 3-hourly precipitation data used in TC methodology have the spatial extent as shown in the figure and the spatial resolution of 0.25 degree

GLDAS 2.1 uses GPCC observations via the ingestion of the disaggregated Global Precipitation Climatology Project (GPCP) V1.3 Daily Analysis precipitation fields. In order to assess whether this has an impact on the TC analysis, pixel-based cross-correlated errors between GPCC and GLDAS by quadruple collocation (QC) analysis have been calculated (Perdicca et al., 2015, Gruber et al., 2016) using GPCC first guess data, GLDAS 2.1, ERA5 and SM2RAIN-ASCAT derived daily rainfall at the global scale. QC is a viable way to estimate the cross-correlated errors between two of the datasets included in the quadruplet (see Chen et al. 2020).

The analysis has been carried out at 0.25° of spatial resolution for the year 2017. Despite the use of a single year, the analysis shows that cross correlated errors between GPCC and GLDAS are very low (R=0.05 in median). If we zoom over the H SAF Extended Area (HEA) the median value decreases to 0.044 (see Figure 24 and Figure 25). In the inset it is reported the histogram of the R-values, showing how most of the pixels provided very low values of error correlation. This point should assure the independence of the datasets used for performing the analysis and thus, should allow to use TC analysis for validating H SAF products not only over the HEA, but also at the global scale.





Figure 24: Cross correlated errors between GPCC and GLDAS over global scale for the year 2017.



Figure 25: Cross correlated errors between GPCC and GLDAS over H SAF extened area for the year 2017.



A1.6 Large statistic

The large statistical analysis allows to point out the existence of pathological behavior in the satellite product performance. The application of the same validation technique step by step is guaranteed in all institutes take part of the PPVG and in both validation methodologies above described.

The large statistical analysis in PPVG is based on the evaluation of monthly and seasonal *Continuous verification* and *Multi-Categorical* statistical scores on one full year of data. It was decided to evaluate both continuous and multi-categorical statistics to give a complete view of the error structure associated to the H SAF product. Since the accuracy of precipitation measurements depends on the type of precipitation or, to simplify matters, on the intensity or accumulated precipitation, the verification is carried out on three precipitation classes (for accumulated precipitation products) as descripted in <u>Table 17</u>.

Accumulated	1	2	3	
Precipitation Classes (CR)	≥ 1 mm/daily	≥5 mm/ daily	≥ 10 mm/ daily	
Table 17: Classes for evaluating cumulated precipitation products				

The impact of different background is also considered in the product performances. Statistical scores are separately computed for land, sea and coast areas. The Precipitation Product Validation Leader collects all validation results as computed by European institutes, verifies the consistency of these results and evaluates the monthly and seasonal common statistical results as reported in Chapter 4.

Continuous statistics

Continuous statistics are provided for each month and season of assessment. The main statistical scores are here listed:

Score	Acronym	Range	Perfect score	Calculation
Number of Satellite samples	NS	N.A.	N.A.	N.A.
Number of Reference (radar/rain gauge) samples	N	N.A.	N.A.	N.A.
Mean Error or Bias	ME	- ∞ to ∞	0	$ME = \frac{1}{N} \sum_{k=1}^{N} (sat_k - obs_k)$
Mean Absolute Error	MAE	0 to ∞	0	$MAE = \frac{1}{N} \sum_{k=1}^{N} sat_k - obs_k $
Standard Deviation	SD	0 to ∞	0	$SD = \sqrt{\frac{1}{N}\sum_{k=1}^{N} (sat_k - obs_k - ME)^2}$
Multiplicative Bias	МВ	-∞to∞	1	$MB = \frac{\frac{1}{N} \sum_{1}^{N} sat_{K}}{\frac{1}{N} \sum_{1}^{N} obs_{K}}$
Correlation Coefficient	СС	- 1 to 1	1	$CC = \frac{\sum_{k=1}^{N} (sat_k - \overline{sat})(obs_k - \overline{obs})}{\sqrt{\sum_{k=1}^{N} (sat_k - \overline{sat})^2 \sum_{k=1}^{N} (obs_k - \overline{obs})^2}}$

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Score	Acronym	Range	Perfect score	Calculation
Root Mean Square Error	RMSE	0 to ∞	0	$RMSF = \sqrt{\frac{1}{N} \sum_{n=1}^{N} (sat_{n-n} obs_{n-n})^{2}}$
(or Root Mean Square Difference)				$\sqrt{N} \sum_{k=1}^{k} \sqrt{N} \sum_{k=1}^{k} \sqrt{N}$
Fractional Standard Error (%)	FSE	0 to ∞	0	FSE = [RMSE / obs] *100%

Table 18: Continuous statistical scores

In the <u>Table 18</u>

- N represents the total number of observation samples and equivales to all satellite/observation pairs for computing all the statistical scores;

- NS indicates the number of product satellite estimates with given characteristics (e.g.; with estimated rain rate > 1 mm/h);

- the index "k" represents the spatial and temporal grid point at the scale of the common reference grid;

- *obs* and *sat* stand for rainfall value acquired by reference observations and satellite estimations, respectively.

The FSE score represents the accuracy for H SAF satellite precipitation products. The User requirements thresholds are below indicated:

Accuracy for $CR \ge 1 \text{ mm/daily}$				
Threshold	Target	Optimal		
FSE% = 200%	FSE% = 150%	FSE% = 100%		
T 11 1				

 Table 19: Precipitation products user requirements (UR)

The FSE score is not appropriate to define the accuracy in the TC methodology. In this case the Correlation Coefficient (CC) is used to define the product accuracy with thresholds as below indicated:

Accuracy for all precipitation rates (CR ≥ 0 mm/daily)			
Threshold	Target	Optimal	
CC = 0.50	CC = 0.65	CC = 0.80	

 Table 20: UR for accumulated precipitation products using TC methodology

Multi Categorical statistics

Multi categorical statistics are derived by the following contingency table:

		Observation		
		yes	no	total
	yes	hits	false alarms	forecast yes
Satellite	no	misses	correct negatives	forecast no

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	total	observed yes	observed no	total]

Table 21: Multi-categorical statistics contingency table

where:

hit:	$Sat_k \ge R_{th} and Obs_k \ge R_{th}$
miss:	$Sat_k < R_{th} and Obs_k \ge R_{th}$
false alarm:	$Sat_k \ge R_{th} and Obs_k \le R_{th}$
correct negative:	$Sat_k < R_{th} and Obs_k < R_{th}$

 \mathbf{R}_{th} is the <u>threshold between the "rain" and "no rain"</u> conditions. The scores evaluated from the contingency table are:

Score	Acronym	Range	Perfect score	Calculation
Probability Of Detection	POD	0 to 1	1	$POD = \frac{hits}{hits + misses} = \frac{hits}{observed yes}$
False Alarm Rate	FAR	0 to 1	0	$FAR = \frac{falsealarms}{hits + falsealarms} = \frac{falsealarms}{forecastyes}$
Critical Success Index	CSI	0 to 1	1	$CSI = \frac{hits}{hits + misses + false alarm}$

 Table 22: Multi-categorical statistics scores

A1.7 Case study analysis

Each institute, in addition to the large statistics verification, produces a case study analysis based on *the knowledge and experience of the institute itself*, following a standard format as below reported. The institute decides whether to use ancillary data such as lightning data, SEVIRI images, the output of numerical weather prediction and nowcasting products.

The main sections of the standard format are:

- description of the meteorological event;
- comparison of ground data and satellite products;
- visualization of ancillary data;
- discussion of the satellite product performances;
- indication on the ground data (if requested) availability into the H SAF project.

Case study analysis are reported in Chapter 3.

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Appendix 2 Ground data used for validation activities

In the following sections the precipitation ground data networks used in the PPVG are described: radar and rain gauge data the following countries: Belgium, Bulgaria, Germany, Hungary, Italy, Poland, Slovakia, and Turkey. It is well known that radar and rain gauge rainfall estimation is influenced by several error sources that should be carefully handled and characterized before using these data as reference for ground validation of any satellite-based precipitation products.

The rain gauge in PPVG is composed by more than 8000 instruments across the partner Countries. These data are, as usual, irregularly distributed over ground and are generally deduced by tipping bucket type instruments. Moreover, most of the measurements are hourly cumulated. So probably the raingauge networks used in this validation activities are surely appropriated for the validation of cumulated products (1 hour and higher), while for the validation of instantaneous estimates the use of hourly cumulated ground measurements could introduce a large error. Moreover, the revisiting time (3,4 hours) of the product makes impossible or not reasonable to validate the product for 1-24 hours cumulated interval. The first object of PPVG (*Rain Gauge- WG*) was to quantitatively estimate the errors introduced in the validation procedure comparing the instantaneous satellite precipitation estimation with the rain gauge precipitation cumulated on different intervals.

The radar data in the PPVG is composed by 74 C-band radars across the 7 countries: Belgium, Germany, Hungary, Italy, Slovakia, Poland, Turkey. The rain gauge network responsible declared that the systems are kept in a relatively good status. The rain gauge inventory pointed out that different correction factors are applied. This means that the corresponding rainfall estimates are diverse, and the estimation of their errors cannot be homogenized. The first step in PPVG (*Radar –WG*) was to define a quality index on the base of the study performed by the Slovakian team and the scheme published by J. Szturc et all. 2008. The evaluation of this quality index has allow to increase the confidence of radar estimates with the selection of more reliable radar data in the PPVG.

In this chapter a description of the ground data available in the PPVG is reported country by country. chapter has the object to provide ground data information and to highlight their error sources.



A1.1 Ground data in Belgium (IRM)

Radar Data

The network

There are four radars in Belgium (Figure 26 and Table 23): two operated by the RMI (in Wideumont - Ardennes range - and Jabbeke - near the coast), one by Belgocontrol (Zaventem airport near Brussels) and the radar coverage is also provided by a fourth radar in France, operated by Meteo France. Of particular interest is the Wideumont radar, located in one of the highest locations of the country with clear horizon in every direction. This is very important in order to have the best possible precipitation picture in this hilly area giving rise to many tributaries of the Meuse river.



Figure 26: Meteorological radars in Belgium (elevation data from Danielson, J.J., and Gesch, D.B., 2011, Global multiresolution terrain elevation data 2010 (GMTED2010): U.S. Geological Survey Open-File Report 2011–1073, 26 p.)

Radar location	Frequency Band	Polarization	Reflectivity Z = aR ^b
Wideumont	С	Single	a = 200, b = 1.6
Jabbeke	С	Dual	a = 200, b = 1.6
Zaventem	С	Single	a = 200, b = 1.6
Avesnes	С	Dual	a = 200, b = 1.6

Table 23: Meteorological radars in 1	Belgium, mai	n features
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Data processing

Raw reflectivity data are contaminated by off-shore wind farms and marine traffic clutter over the North Sea. The sea clutter is detected and removed using the module available in Selex Rainbow 5



software for this purpose. Rainfall rate estimation is based on a PCAPPI at a given height (depending on the radar) combined with the Marshall-Palmer relationship between reflectivity and rainrate. For grid points where several estimates are available from different radars, the maximum value is taken. A 5-min accumulation is generated based on the rainrates at t and t-5. If the rainrate at t-5 is missing, the rainrate at t-10 is used to generate the 5-min accumulations at both t and t-5. This means than 1 missing file is tolerated. Accumulations of higher durations are made by summing accumulations of lower duration.

Acknowledgement (elevation data for the radar map)

Danielson, J.J., and Gesch, D.B., 2011, Global multi-resolution terrain elevation data 2010 (GMTED2010): U.S. Geological Survey Open-File Report 2011–1073, 26 p.



A1.2 Ground data in Bulgaria (NIMH)

Rain gauge

The network

The maximum number of available manually measured daily accumulated rain gauges is up-to 300, irregularly distributed over the country. These stations are measured every day at 6:30 UTC by emptying the collected in the past 24hours rain.

The hourly measuring automatic rain gauges are varying on daily basis and range from 70 units to 130 units. Number of stations is varying mainly because in winter months not heating gauges data is discarded from the operational database when air temperature drops below 0° C. Other specific measurement errors that are detected by the operators as funnel clogging, sensor failure etc.

The average minimum distance between closest stations is about 20 km. Most dense network of automatic gauges is built in South-Central Bulgaria where a number of European funded projects permitted to purchase and install more than 50 telemetric gauges. Spatial distribution of automatic gauges is described in (Naldzhiyan et al., 2017)¹. One of main objectives was to consider mountain structures because of the need to measure snowfall accumulation in winter months.

This points out that the distribution of gauges could be able to describe the spatial structures of precipitation fields in case of wintertime rainfall. This objective is reached in central and South Bulgaria but much less in Western and Eastern parts of the country.



Figure 27: Spatial distribution of automatic telemetric gauges in Bulgaria (NIMH)

In following figure the distribution of working stations over Bulgaria is shown.

¹ Naldzhyian A, Georguiev O., Artinyan E., 2017: "From the sensors to the models, integrated hydro-meteorological systems in NIMH – BAS, Bulgaria". International Conference on Automatic weather stations ICAWS-2017.



Figure 28: Distribution of the automatic stations of the Bulgaria network collected by NIMH.

<u>The instruments – hourly measuring rain gauges:</u>

- About 50 raingauges are of weighing type so they can measure snowfall without heating; however some of them have orifice heating (where 220V supply is available). Examples of such sensors are Vaisala VRG101, SEBA TRW 200cm² and SUTRON TPG
- About 80 raingauges are of tipping bucket type so they need 220 V supply to switch on the heater in winter conditions, however half of them are installed beside rivers so 220V supply is not available; The sensors types are mostly SEBA RG50, DELTA-OHM 400cm² and MTX 400 cm²
- Most of the raingauges have a minimum detected quantity of 0.1 mm, others have 0.2 mm.
- The maximum rain rate (with acceptable quality) that can be measured by the gauges ranges between 33 and 120 mm⁻¹ over one minute, depending on the manufacturer.

The rainrate is measured over 1 minute and 1-hour accumulation intervals depending on the hardware specifications.

At the moment, the NIMH officially provides only daily data from manually measured rain gauges. Shorter accumulation times could be available for scientific studies but not publicly distributed.

The data processing

Quality control is performed on the data, after daily visual comparison check, but only on YES/NO basis. When rain sensor fails it may be seen in few days or not be seen by the operators specially when it doesn't rain.

For NIMH internal usage hourly accumulated rain data is converted into 3 h sums and then interpolated using a kriging technique to 8 km regular grid. The method also incorporates the 24h

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accumulated data from manually measured tin cans, thus enhancing the spatial and vertical quality of the field (Artinian et al., 2007)²

For H SAF validations a subset of country's automatic gauges is used (between 70 and 90) because of the much denser network in South-Central Bulgaria – an area of about 34000 km². Other parts of the country have much sparse gauge networks, data from which is not suitable to be interpolated using the GRISO technique.



Figure 29: The area in Bulgaria used for H SAF validation with hourly accumulated rain data

² Artinyan, E. et al, 2007: "Modelling the water budget and the riverflows of the Maritsa basin in Bulgaria", Modelling the water budget and the riverflows of the Maritsa basin in Bulgaria. Hydrology and Earth System Sciences. 12. 10.5194/hessd-4-475-2007.



A1.3 Ground data in Germany (BfG)

The H SAF products are validated for the territory of Germany by use of two observational ground data sets: SYNOP - precipitation data based on the network of synoptical stations, provided by the German Weather Service (DWD) and RADOLAN-RW - calibrated precipitation data based on the radar network of DWD and calibrated by DWD by use of measurements at precipitation stations.

Data	Number/Resolution	Time interval	Delay	Annotation
Synoptical stations	~ 200	6h / 12h	Near-real- time	
Precipitation stations	~ 1100	hourly	Near-real- time	Automatic precipitation stations
RADOLAN RW	16 German radar sites, ~1 km x ~1 km	1 hour,	Near-real- time	Quantitative radar composite product RADOLAN RW (Radar data after adjustment with the weighted mean of two standard procedures)

 Table 24: Precipitation data used at BfG for validation of H SAF products

Rain gauge

The network

The data used are compiled from \sim 1300 rain gauges. About 1000 are operated by DWD while about 300 are operated by other German authorities. The average minimum distance between stations is 17 km.

The instruments

The measurement instruments are precipitation sensors OTT PLUVIO of Company Ott^{3 4}. They continually and precisely measure quantity and intensity of precipitation in any weather, based on balance principle with temperature compensation (heated funnel) and by an electronic weighing cell. The absolute measuring error is less than 0.04 mm for a 10 mm precipitation amount and the long-term (12months) stability is better than 0.06 mm. The operating temperature ranges from -30° C to $+45^{\circ}$ C. The minimum detected quantity (sensitivity) is 0,05 mmh⁻¹. The maximum possible measured rain rate is 3000 mmh⁻¹. The operational accumulation interval theoretically is one minute.

The data processing

Continuous, automatic measurement of liquid and solid precipitation data are collected, accumulated (intervals: from 1hour until 1day) and provided as SYNOP tables by DWD. These data are error corrected and quality controlled in four steps with checks of completeness, climatologic temporal/spatial consistency and marginal checks.

³ <u>http://www.ott.com/web/ott_de.nsf/id/pa_ottpluvio2_vorteile.html?OpenDocument&Click=</u>

⁴ Precipitation amount and intensity measurements with the Ott Pluvio, Wiel Wauben, Instrumental Department, INSA-IO, KNMI, August 26, 2004



Figure 30: (left): Network of rain gauges in Germany - Figure 31: (right): Pluvio with Remote Monitoring Module

Radar data

Radar-based real-time analyses of hourly precipitation amounts for Germany (RADOLAN) is a quantitative radar composite product provided in near-real time by DWD. Spatial and temporal high-resolution, quantitative precipitation data are derived from online adjusted radar measurements in real-time production for Germany. Radar data are calibrated with hourly precipitation data from automatic surface precipitation stations.⁵

The combination of hourly point measurements at the precipitation stations with the five-minuteinterval radar signals of the 16 weather radars (C-Band Doppler) provides gauge-adjusted hourly precipitation sums for a \sim 1km x \sim 1km raster for Germany in a polar stereographic projection.

Radar site	Latitu	ıde	Long	itude	WMO	Radar site	Latitu	de	Longi	tude	WMO
	(N)		(E)		No.		(N)		(E)		No.
München	48°	20'	11°	36'	10871	Rostock	54°	10'	12°	03'	10169
	14"		46"				35"		33"		
Frankfurt	50°	01'	08°	33'	10630	Ummendorf	52°	09'	11°	10'	10356
	25"		34"				39"		38"		
Hamburg	53°	37'	09°	59'	10147	Feldberg	47°	52'	08°	00'	10908
	19"		52"				28"		18"		
Berlin-	52°	28'	13°	23	10384	Eisberg	49°	32'	12°	24'	10780
Tempelhof	43"		17"				29"		15"		
Essen	51°	24'	06°	58'	10410	Flechtdorf	51°	18'	08°	48'	10440
	22"		05"				43"		12"		
Hannover	52°	27'	09°	41'	10338	Neuheilenbach	50°	06'	06°	32'	10605
	47''		54"				38"		59"		

5

http://www.dwd.de/bvbw/appmanager/bvbw/dwdwwwDesktop? nfpb=true& windowLabel=dwdwww main book&T1460994925114492118088 1gsbDocumentPath=Navigation%2FWasserwirtschaft%2FUnsere Leistungen%2FRadarniederschlagsprodukte%2FRADOLAN%2Fradolan node.ht ml%3F nnn%3Dtrue&switchLang=en& pageLabel= dwdwww spezielle nutzer forschung fkradar

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H SAF				(Product H64 – P-AC-SM2RAIN)				Date: 27 Page: 59	7/02/20 9/90)22		
												-
Emden	53°	20'	07°	01'	10204	Türkheim	48°	35'	09°	47'	10832	
	22"		30"				10'	,	02"			
Neuhaus	50°	30'	11°	08'	10557	Dresden	51°	07'	13°	46'	10488	
	03"		10"				31'	,	11"			

Table 25: Location of the 16 meteorological radar sites of the DWD



Figure 32: (left) radar compound in Germany (March 2011) ; Figure 33: (right) location of ombrometers for online calibration in RADOLAN; squares: hourly data provision (about 500), circles: event-based hourly data provision (about 800 stations)6.

The flowchart of online calibration method applied in RADOLAN is depicted in Figure 34

⁶ Bartels, H.: Projekt RADOLAN. Routineverfahren zur Online-Aneichung der Radarniederschlagsdaten mit Hilfe von automatischen Bodenniederschlagsstationen (Ombrometer), Abschlussbericht 2004





A1.4 Ground data in Hungary (OMSZ)

The radar network

The main data used for validation in Hungary would be the data of meteorological radars. There are four C-band dual polarized Doppler weather radars operated routinely by the OMSZ-Hungarian Meteorological Service. The location of the four Hungarian radars and main measurement characteristics are listed in <u>Table 26</u>. All four radars are calibrated periodically, with an external (calibrated) TSG, the periodicity is kept every 1 year.

Year of installation	Location	Radar type	Parameters measured
1999	Budapest	Dual-polarimetric Doppler radar	Z,ZDR,KDP,ΦDP
2003	Napkor	Dual-polarimetric Doppler radar	Z,ZDR,KDP,ΦDP
2004	Poganyvar	Dual-polarimetric Doppler radar	Z,ZDR,KDP,ΦDP
2014	Szentes	Dual-polarimetric Doppler radar	Z,ZDR,KDP,ΦDP

 Table 26: Main characteristics of the Hungarian radar network

Instrument characteristics

The Hungarian radar network is composed by four Doppler radars, which are measuring in the Cband, mainly at same frequencies. The scan strategy is the same for all the radars. The parameters of the instruments and the measurement campaigns are listed in <u>Table 27</u>.

	Budapest	Napkor	Poganyvar	Szentes	
Frequency band	requencyC-Band,and5625MHz		C-Band, 5610MHz	C-Band, 5640MHz	
Polarization (Single/Double)	rization gle/Double) double d		double	double	
Doppler capability (Yes/No)	ppler pability Yes Yes es/No)		Yes	Yes	
Scan strategy: elevations,	scan freq: 5 min				
maximum nominal range	0 0.5 1.1 1.9 3.0 4.7 7.0 10.0 14.2	0 0.5 1.1 1.9 3.0 4.7 7.0 10.0 14.2	0 0.5 1.1 1.9 3.0 4.7 7.0 10.0 14.2	0 0.5 1.1 1.9 3.0 4.7 7.0 10.0 14.2	

		Product Validation Report - PVR-64				Doc. No: SAF/HSAF/PVR-64		
		(Product H64 – P-AC-SM2RAIN)			Date: 27/02/2022			
					Page: 62/90			
distance, range Range 2 resolution Resolution		240 Km	Range 240 Km	Range 240	Km	Range 240 Km		
		tion:125	Resolution:125	Resolution:125		Resolution:125		
	т		m	m		m		

Table 27: Characteristics of the four radar instruments in Hungary

Data processing and radar products

Radar field corrections

Radar measurements are influenced by many error sources that should be minimized as much as possible. As such, in case of the Hungarian radar data many correction methods are applied to filter out false radar reflectivity measurements. Clutter removal, WLAN filter and clear-air echo filter is implemented in the processing chain of all four-radar data. The beam blockage correction is also implemented in the processing chain in order to correct serious underestimation of precipitation amounts behind mountains. Attenuation correction (the attenuation of electromagnetic waves in water environment, water drops) was implemented in 2019. Hungary does not apply VPR (Vertical Profile Reflectivity) correction.

Precipitation intensity is derived from radar reflectivity with the help of an empirical formula, the Marshall-Palmer equation ($R=a*Z^b$, where a=200, b=1.6). From the four radar images a composite image over the territory of Hungary is derived every 5 minutes applying the maximum reflectivity in one column method, in order to make adjustments in overlapping regions.

Rain gauge correction of the radar precipitation fields

The non-corrected precipitation field can be corrected by rain gauge measurements. In Hungary, we do not make corrections to instantaneous 5 minutes radar data. In our institute, we only use a correction for the total precipitation for 1, 3-, 6-, 12- and 24-hour periods.

For the accumulated products, we use a special method to accumulate rainfalls: we interpolate the 5minutes measurements for 1-minute grid by the help of displacement vectors also measured by the radar, and then sum up the images which we got after the interpolation. It is more precise especially when we have storm cells on the radar picture, because a storm cell moves for 5 minutes and thus we do not get continuous precipitation fields when we sum up only with 5 minute periods. This provides satisfying results. However, there is still a need for rain-gauge adjustment because there are obviously places (behind mountains) that the radar does not see.

The radars are corrected with rain gauge data every hour. The correction method using rain gauge data for 1-hour total precipitation consists of two kinds of corrections: the spatial correction which becomes dominant in the case of precipitation extended over a large area, whereas the other factor, the distance correction factor prevails in the case of sparse precipitation. These two factors are weighted according to the actual situation. The weighting factor depends on the actual effective local station density, and also on the variance of the differences of the bias between radar and rain gauge measurements. On the whole, we can say that our correction method is efficient within a radius of 100 km from the radar. In this region, it gives a final underestimation of about 10%, while at bigger distance; the underestimation of precipitation fields slightly increases.

Resolution, projection, threshold of detection

The resolution of the radar data used for validation is 1km by 1km. This is true for the accumulated and the instantaneous products as well. Hungarian radar data is available operationally in stereographic (S60) projection.

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Raingauge network

Distribution of the raingauge stations in Hungary

The automated precipitation measurement network of OMSZ Hungarian Meteorological Service (OMSZ) consist of the following types of instruments:

96 instruments are tipping bucket type (34 Lambrecht Lambrecht 15183H, 62 Lambrecht 15180H), and 37 instruments are weighing type (4 Lambrecht 15184H, 23 OTT Pluvio2, 8 Geonica Datarain4000, 2 EWS HWI).

In addition, 142 weighing type instruments (OTT Pluvio2) belonging to the General Directorate of Water Management of Hungary (OVF) are also maintained by OMSZ and integrated into its network. Altogether more than 270 automated raingauges are available, but the number is constantly increasing with new installations.

All data of the precipitation network are collected in every 10 minutes and they undergo on a quality check procedure.

Traditional precipitation measurements by human observers (either spacialists or amateurs) are not used for HSAF validation work in Hungary.

The map below shows the Hungarian automated precipitation measurement network, the stations of OMSZ are red and the stations of OVF are green.



Figure 35: Hungarian automated precipitation measurement network.



A1.5 Ground data in Italy (DPC, UniBo)

Rain gauge

l

The network

The maximum number of available raingauges is about 3000, irregularly distributed over the surface. On the average, however, a number of stations have low quality data, failure or data transmission problems and their data are missing (-9999 recorded). This number of no data stations is highly varying on hourly/daily basis and ranges from few units to a hundred. In case of data acquired but not transmitted/recorded, the first transmitted measure is the cumulated value over the time when the data were not transmitted.

The average minimum distance between closest stations is about 11 km, with a very high variance: in some regions (such as Tuscany in central Italy) it is below 5 km, while in Emilia Romagna (Po Valley) it is more than 20 km. A study of the decorrelation distance between stations as function of the mutual distance has been carried out for the 2009 dataset. The decorrelation distance is defined as the minimum distance between two observations that makes the Pearson correlation coefficient between the two measures decrease below e⁻¹. Results are shown in Figure 36, where the decorrelation distance is a large variability of this parameter from higher values (around 60 km for cold months when large precipitating systems dominate and reduces to roughly 10 km when small scale convection is more likely to occur (warm months).

This points out that the distribution of gauges could be able to describe the spatial structures of precipitation fields in case of wintertime rainfall, while may be inadequate for spring/summer convective events.



Figure 36: Correlation between rainrates detected by two close stations as function of the distance between the two stations. Colors refer to the month along 2009

In Figure 37 the distribution of working stations over Italy is shown.





Figure 37: Distribution of the raingauge stations of the Italian network collected by DPC.

The instruments

This section provides the following information:

- All the available raingauge are of tipping bucket type;
- Most of the raingauge have a minimum detected quantity of 0.2 mm, others have 0.1 mm.
- The maximum rainrate that can be measured by the gauges ranges between 300 and 500 mm⁻¹ over one minute, depending on the manufacturer.

The rainrate is measured over different cumulation intervals by the different local administrations managing the network, but the data disseminated are all integrated over 60 minutes.

At the moment, the National network made available by DPC provides only hourly data, shorter cumulation times could be available for case studies after specific agreements with local management authorities.

Only a small subset (about 300 stations) of gauges have heated funnel, especially in alpine regions (such as Valle d'Aosta and Piedmont), and this is a clear source of errors in both summer (due to hailfall) and in autumn/winter (due to snowfall).

The data processing

To homogenize the two ground datasets, rain gauge data, preprocessed according to range, persistence, step, and spatial consistency (Shafer et al. 2000) to screen out suspect values, have been interpolated over a regular grid (5 km x 5 km) through the Random Generator of Spatial Interpolation from uncertain Observations (GRISO). The GRISO (Pignone et al. 2010; Feidas et al. 2018) is an improved kriging-based technique implemented by the International Centre on Environ- mental Monitoring (CIMA Research Foundation). This technique preserves the values observed at the rain gauge location, allowing for a dynamical definition of the covariance structure associated with each



rain gauge by the interpolation procedure. Each correlation structure depends both on the rain gauge location and on the accumulation time considered. GRISO is adopted by all European participating countries in the H SAF validation procedure (Puca et al. 2014). The resulting grid is a 5x5 km regular grid with 240 columns and 288 lines. Moreover, a Digital elevation model is used to provide a mask of Italy in order to: 1) screen out sea-pixels too far from the coastlines and 2) process the pixels with the elevation above sea level.

Radar data

The network

The Italian Department of Civil Protection (DPC) is the authority leading the national radar coverage project in order to integrate the pre-existent regional systems. Currently, the radar network is composed by 22 systems (20 C-band and 2 X-band systems), most of them with dual-polarization. The network is composed by 8 C-band fixed regional installations (five of them are polarimetric), five systems owned by the Italian company for air navigation services (ENAV), 9 dual-polarization systems managed by DPC (7 using C-band and 2 X-band).

The Figure 38 shows the spatial radar coverage of the Italian territory.



Figure 38: Italian radar network coverage. The green and blue radar symbol stands for dual- and single-polarization system, respectively.

Radar Data processing

The operational radar processing chain is briefly described in this section.



It aims at compensating or at least identifying most of the uncertainty sources conditioning the radar rainfall estimation process (Friedrich et al., 2006). Among them, the following error sources are primarily considered: contamination by non-weather returns (clutter), Partial Beam Blocking (PBB), beam broadening at increasing distances, vertical variability of precipitation (Germann and Joss, 2002; Joss and Lee, 1995; Marzano et al., 2004) and rain path attenuation (Bringi and Chandrasekar, 2001; Carey et al., 2000; Testud et al., 2000; Vulpiani et al., 2008). Every error source is quantified through specific tests ending with the estimation of specific (partial) quality matrices and, when possibile, is compensated for. The overall data quality (Q) is then obtained as a combination of the partial quality matrices. The quality model described in Rinollo et al. (2013) is embedded within the overall processing chain schematically depicted in Figure 24.

In this schematic representation, the sequential flow among consecutive computational steps is specified by black arrows, while the blue ones identify the data input (or output) to (or from) a specific processing module.

The processing chain can be summarized through the following few steps as follows:

- i. As typical, the raw volumetric data must be first filtered from non-weather returns. This step is here achieved using the fuzzy-logic approach proposed in Vulpiani et al. (2012) for polarimetric radar systems.
- ii. The next step is the correction for Partial Beam Blocking (PBB) based on the retrieved 3-D occlusion map (Bech et al., 2003) that, assuming the e.m. waves propagate in a standard atmosphere, is evaluated only once for a given radar scanning strategy.
- iii. The rain path attenuation is just qualitatively evaluated in case the considered radar system has single-polarization capability (Rinollo et al., 2013), otherwise it is compensated for by means of the differential phase shift that needs to be preliminarly processed. In this framework, the iterative moving-window range derivative approach proposed in Vulpiani et al. (2012) is applied here.
- iv. The range-related deterioration of radar data quality is modeled through a non-linear function as in Rinollo et al. (2013).
- v. Once the attenuation is evaluated and, eventually, compensated for through the so-called ZPHI method (Testud et al., 2000), the overall data quality is computed as geometric mean of the partial quality matrices.

 $Q = q_{clutter} \cdot q_{vertical} \cdot q_{PBB} \cdot q_{distance} \cdot q_{attenuation}$

- vi. The retrieved mean Vertical Profile of Reflectivity (VPR) is applied to the entire volumetric scan with the aim to use all the observations along the vertical to retrieve the surface rainfall rate. All the clutter-filtered and attenuation-corrected (if applicable) PPIs are projected at ground by means of the average Vertical Profile of Reflectivity (VPR).
- vii. The Surface Rainfall Intensity (SRI) map is computed as a quality-weighted average of each rain rate map, obtained by each ground-projected reflectivity sweep (Vulpiani et al., 2014).
- viii. The SRI composite is built by combining the single-radar rainfall maps through a squaredquality-weighted approach. In case of dual-polarization systems, the composite rainfall retrieval algorithm proposed in Vulpiani and Baldini (2013).

As descripted in Petracca et al. (2018), only radar data with Q values greater than 0.60 are used for comparison with satellite data.



Figure 39: Schematic representation of the Italian radar data processing chain.



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A1.6 Ground data in Poland (IMWM)

Rain gauge

<u>The network</u>

The maximum number of rain gauges in the Polish ATS (Automatic Telemetric Station) national network is 950. Each ATS post is equipped with two independent rain gauges of the same sort. One of them is heated during the winter period and the other one is not. Therefore precipitation information is derived from 475 points at the time. Fact that rainfall is measured by two equally sensitive instruments two meters away from each other at the same post, enables to apply simple in situ data quality control during summertime. During winter non-heated rain gauge is covered with a cup to prevent it from being clogged by the ice and damaged. Because of that the precipitation information derived from ATS network in winter cannot be verified using this method. It can be stated that during the wintertime precipitation information might be a slightly bigger measuring error.

The number of rain gauges available for H SAF validation activities varies from day to day due to operational efficiency of ATS network in Poland and depends on large number of independent factors. It can be stated that the number varies between 330 and 475 rain gauges for each day of operational work.

Mean minimum distance between precipitation measuring ATS posts (between each pair of rain gauges) in Polish national network is 13,3 km.



Figure 40: ATS national network in Poland

The instruments

All rain gauges working within Polish ATS national network are MetOne tipping bucket type instruments. Minimum detected quantity that can be measured by those rain gauges is 0,1 mm/h which means that each tilt of rain gauge bucket adds 0,1mm to the total sum of the measured precipitation. During very heavy precipitation events MetOne rain gauges tend to underestimate real precipitation by factor of 10%. Maximum measured rainrate (mmh^{-1}) by MetOne instruments in Poland was recorded in 5.06.2007 at ATSO Koscielisko Kiry at the foot of Tatra Mountains. The recorded values



reached 65 mm/h. Operational cumulation interval *(min)* of ATS network rain gauges is set for 10 minutes and can be adjusted according to given needs. There is possibility to have very short cumulation intervals for case studies - theoretically 1 minute - but not on every given precipitation post. It depends on local DCS settings.

The data processing

As stated above the data quality control can be achieved by comparison on two rainfall datasets collected by two independent rain gauges at the same ATS post. It is done operationally during summertime. There is no such possibility during the winter because of lack of non-heated rain gauge dataset. In case that one pair of rain gauges at the same ATS post provide two different rainfall readings the higher one is taken into account.

No specialization technique is used for standard validation process. However, for some case studies, the Natural Neighbor technique is applied for satellite and ground precipitation data. To match the precipitation information with satellite data spatial and temporal matching are applied.

- <u>Spatial matching</u>: for each given satellite pixel, the posts situated within that pixel were found. The pixel size was taken into account, however, its shape was assumed to be rectangular. If more than one rain gauge were found within one satellite pixel, the ground rain rate value was calculated as a mean of all rain gauges measurements recorded within that pixel;
- <u>Temporal matching</u>: satellite derived product is combined with the next corresponding ground measurement. As the ground measurements are made with 10 minute time resolution, the maximum interval between satellite and ground precipitation is 5 minutes.

Radar data

The Polish meteorological radar network called POLRAD (initially Doppler Radar System METEOR) consists of eight devices and was produced by Gematronic Weather Radar Systems GmbH SELEX Sistemi Integrati GmbH. The system had been installed in Poland starting from year 2001 and has its 20th anniversary this year. The software running within this system is Rainbow 5 (and it's later updates). The radars are using band C (frequency 5,6 GHz and subsequently 5,4 cm wavelength). All of the radars are Doppler instruments operationally run in 10 min. scan frequency and two of them (Ramża and Pastewnik) in dual polarisation mode.

Station location	Latitude	Longitude	Elevation a.m.s.l.	Antenna height a.t.l.	Radar type
Legionowo	52°24'18,79''	20°57'39,28''	89,0 m	29,0 m	Doppler radar METEOR 1500 C
Rzeszów	50°06'49,79''	22°02'12,09''	206,5 m	30,0 m	Doppler radar METEOR 1500 C
Brzuchania	50°23'39,13''	20°05'00,35''	388,5 m	35,0 m	Doppler radar METEOR 500 C
Ramża	50°09'04,59''	18°43'29,86''	320,7 m	36,0 m	Dual polarisation Doppler radar METEOR 1600C
Pastewnik	50°53'32,81''	16°02'22,17''	666,5 m	23,5 m	Dual polarisation Doppler radar METEOR 1600C

Eight basic radars working within the POLRAD network.
P H	I SAF	Product Validation Report - PVR-64 (Product H64 – P-AC-SM2RAIN)			Doc. Date Page	Doc. No: SAF/HSAF/PVR-64 Date: 27/02/2022 Page: 73/90	
Poznań	52°24'47,73''	16°47'49,40''	95,1 m	35,0 1	n	Doppler radar METEOR 500 C	
Świdwin	53°47'40,25"	15°50'16,97"	121,1 m	30,0 1	n	Doppler radar METEOR 500 C	
Gdańsk	54°23'03,17''	18°27'23,00''	135,2 m	20,0 1	n	Doppler radar METEOR 1500 C	

Table 28: Radars in Poland

The original radar arrangement in Poland (when installed).



Figure 41: Radar dislocation in Poland

The radar ranges drawn above represent both 125 km and 250 km ranges respectively.

The spatial resolution of the radar scan is 1 km, the data resolution in azimuth is 1° on average, hence within a circumference of a given radius 200 km one elevation angle returns 72 000 virtual measurement points. The basic max reflectance radar information is gathered from 10 elevation angles so one measurement cycle (max reflectance only, no Doppler wind information received) returns 720 000 points. To make this information valid, one point requires 256 levels – 8 bits of information. Concluding, one reflectance only measurement cycle requires 720 kB.

Since the beginning of the millennium Poland has operational meteorological radar data exchange with neighbouring countries. The current composite radar map of Poland looks as follows (incorporating radar data received on the basis of transnational radar data exchange).



Figure 42: Radar composite map in Poland

The operational radar data exchange is maintained with Germany, Czechia, Denmark and Slovak Republic. The cooperation was initiated in 1998 within Baltrad and Opera projects and is still kept valid.

Foreign radars that provide data on operational basis.

Station location	Longitude	Latitude
Brdy-Praha	13.817800	49.658300
Skalky	16.788500	49.501100
Kojsovska	20.987277	48.782895
Maly Javornik	17.153100	48.256100
Kubinska Hola	19.249350	49.271670
Spani Laz	19.257430	48.240430
Protzel	13.858210	52.648660
Rostock	12.058070	54.175660
Dresden	13.768630	51.124630
Bornholm	14.887517	55.112750

Table 29: Operational foreign radars.

EUMETSAT	Product Validation Report - PVR-64	Doc. No: SAF/HSAF/PVR-64
H SAF	(Product H64 DAC SM2PAIN)	Date: 27/02/2022
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The non-operational radar data exchange is carried out with Belarus, Ukraine and Latvia. The data aren't operationally provided and sometimes are very burdened with artificial errors, hence they are not included in the operational composite radar map of Poland.

Station location	Longitude	Latitude		
Laukuva	22.239500	55.609040		
Traky	25.106780	54.626220		
Grodno	24.048889	53.651943		
Minsk	28.040277	53.865833		
Brzesc	23.898611	52.115276		
Lviv	23.900514	49.848378		
Table 30: Non-operational foreign reduce				

Foreign radars that provide test or non-operational data.

Table 30: Non-operational foreign radars.

It is worth mentioning that in the current year 2022 the POLRAD system is being upgraded to new instruments and software. The work will continue in the next year (2023). All new radars will be Doppler class magnetron instruments with dual polarisation. In addition, there will be two completely new radars located in Użranki and Góra Św. Anny.

The new radars that will be installed in 2023.

Station location	Longitude	Latitude
Użranki	21,4121	53,8557
Góra Św. Anny	18,1530	50,4640

Table 31: New radars with forthcoming installation



A1.7 Ground data in Slovakia (SHMÚ)

Rain gauge

<u>The network</u>

In Slovakia there are overall 98 automatic rain gauge stations potentially available for the H SAF project. The real number of usable gauges varies with time because on average about 20 of them are out of operation.

Mean minimum distance between rain-gauges in the complete network is 7,74 km. Map of the rain gauge network in Slovakia containing also climatological and selected hydrological stations is shown in next figure.



Figure 43: Map of SHMÚ rain gauge stations: green – automatic (98), blue – climatological (586), red - hydrological stations in H SAF selected test basins (37)

The instruments

Type of all the automatic rain gauges is tipping bucket (without heating of the funnel). The gauges are able to measure precipitation rates ranging from 0,1 to 200 mm/h at 10 min operational accumulation interval. Shorter accumulation interval of 1 min is also possible which makes the instruments suitable for case studies in the H SAF project.

The data processing

The rain gauge data are not used at SHMÚ directly for the H SAF precipitation validation but they are utilized as the input to the INCA precipitation analysis system which is supposed to become a new validation tool. Prior the INCA analysis the rain gauge data are interpolated onto the regular 1x1 km grid using the inverse-distance-squared (IDS) interpolation method. Only the 8 nearest rain gauge stations are taken into account in the interpolation in order to reduce occurrence of precipitation bull-eyes artifact.

SHMÚ performs the offline automatic and manual quality check of the rain gauge data. In frame of the INCA system a quality control technique called blacklisting has been developed which avoids the data from systematically erroneous rain gauges to enter the analysis. Currently the blacklisting is used in manual mode only.



Radar data The network

The Slovak meteorological radar network consists of 4 radars (see next figure). One is situated at the top of Maly Javornik hill near city Bratislava, second one is on the top of Kojsovska hola hill close to the city Kosice. The third and fourth radars are installed on Kubinska hola in Orava region and on Spani laz in the south of Central Slovakia. All of them are Doppler, dual-polarization C-band radars of the same type.



Figure 44: Map of SHMÚ radar network; the rings represent maximum operational range of the radars - 240 km

The instruments

The radars are operated and technically maintained by SHMÚ. Receivers of radars are calibrated regularly by means of internal test signal generator (TSG) every 6 months. The peak power of the transmitted pulses is calibrated with the same periodicity using calibrated power meters.

The basic parameters of SHMÚ radars are summarized in following table.

	Maly Javornik	Kojsovska hola	Kubinska hola	Spani laz
Frequency band	C-Band, YesC-Band, YesC-Band, Yes5605 MHz5645 MHz5630 MHz		C-Band, Yes 5615 MHz	
Polarization (Single/Double)	Double	Double	Double	Double
Doppler capability (Yes/No)	Yes	Yes	Yes	Yes
Scan strategy: scan frequency, elevations,	Scan frequency: 5 min	Scan frequency: 5 min	Scan frequency: 5 min	Scan frequency: 5 min

H SAF	Product Va (Product	Product Validation Report - PVR-64 (Product H64 – P-AC-SM2RAIN)			Doc. No: SAF/HSAF/PVR-64 Date: 27/02/2022 Page: 78/90		
maximum nominal rangeEl clistance, 0. 2. 7. 26range resolution7. 26Ra 25	evations (deg): 0 0.5 1.0 1.5 0 2.7 3.4 4.4 0 11.4 18.3 5.7 ange: 240 km esolution: 0m	Elevations (deg): 0.0 0.5 1.0 1.5 2.0 2.7 3.4 4.4 7.0 11.4 18.3 26.7 Range: 240 km Resolution: 250m	Elevati 0.0 0.2 2.0 2.7 7.0 1 26.7 Range: Resolut 250m	ons (deg): 5 1.0 1.5 7 3.4 4.4 1.4 18.3 240 km tion:	Elevations (deg): 0.0 0.5 1.0 1.5 2.0 2.7 3.4 4.4 7.0 11.4 18.3 26.7 Range: 240 km Resolution: 250m		

Table 32: Characteristics of the SHMÚ radars

The data processing

For ground clutter removal the GIP frequency domain filter is used. Isolated bins in the range and azimuth direction are removed by the speckle removal filters. The data with intensities around the noise level and below are eliminated using the LOG threshold.

The measured radar reflectivity is corrected for atmospheric (clear-air) attenuation of the radar beam. RLAN interference is removed using the Interference filter and SQI thresholding in the signal processor. The radar reflectivity is then corrected for attenuation of the radar beam in the precipitation. In the final step, dual-pol filtering of the non-meteorological echoes is applied on the reflectivity data which removes also most of the remaining interference.

Correction for vertical profile of reflectivity (VPR) is not applied at SHMÚ. However beam blocking correction is being used in the quality-checking step for the H SAF validation due to complicated orographical conditions in Slovakia. Software filter for the RLAN interference detected by radars is currently in development at SHMÚ.

Quality-based radar composite CAPPI 2 km products from all radars is used for the H SAF validation. The composition algorithm uses several quality-checking algorithms, where distance from radar, beam-blockage, spike-detection (similarity with neighbor values) and comparison with NWCSAF cloud-type and clout-top-height are considered and evaluated as quality indexes. The resulting value in the overlapping area of several radars is computed as weighted average of values from different radars, where weights are evaluated quality indexes.

Precipitation intensity is derived from radar reflectivity according to the Marshall-Palmer equation $(Z=a*R^b)$ with constant coefficients valid for stratiform rain (a=200, b=1.6).

No raingauge correction of the derived instantaneous precipitation is applied. Effect of elevating radar beam with increasing range and beam attenuation is reduced by limiting the validation area to rain effective range of 120 km for both radars in the composite.

The instantaneous precipitation products are provided in Mercator projection with approximately 1 km resolution. Threshold for precipitation detection is 0,02 mm/h. Time resolution of the current instantaneous products is 5 minutes, for the products prior to April 2010 it was 10 minutes and prior to August 2009 15 minutes.

Precipitation accumulation in case of 3-hourly interval is based on integration of 5 (10 or 15) minutes instantaneous measurements in time period of 3 hours. Accumulated precipitation for intervals of 6, 12 and 24 hours is calculated as a sum of the 3-hourly accumulated precipitation. At least 92% of instantaneous measurements must exist in relevant time period for the 3-hourly accumulated product to be produced.

No rain gauge correction of the accumulated precipitation is applied but the same limitation of validation area is used as for the instantaneous product. Threshold for precipitation detection of the 3-hourly accumulated product is 0,5 mm. Geographical projection and space resolution of the accumulated products are the same as those of instantaneous product (see above).



For validation of H SAF precipitation products it is necessary to know errors distribution of used ground truth data – in case of SHMÚ it is precipitation intensity and accumulated precipitation measured by Slovak radar network. For this purpose a study called "SHMU study on evaluation of radar measurements quality indicator with regards to terrain visibility" has been elaborated (/hsaf/WP6000/WP6100/precipitation/WG_groups/WG2-radar/WG-2-3_radar quality indication v1.doc). To find distribution of errors in radar range next steps had to be done:

- simulations of terrain visibility by radar network using 90m digital terrain model
- statistical comparison of radar data against independent rain gauge data measurements
- derivation of dependence (regression equation) describing the errors distribution in radar range with regard to terrain visibility, based on rain gauge and radar data statistical evaluation computation of error distribution maps using regression equation and terrain visibility

Main results of this study are shown in next figure. It is evident that the best visibility of SHMU radars corresponds to the lowest PR-RMSE-RMSE of 60% displayed by light violet colors. PR-RMSE-RMSE is of quite homogeneous distribution with average of 69% in prevalent lowlands of Slovakia displayed by bluish colors. But in central and north-west mountainous areas this error exceeds 100%.



Figure 45: Map of relative RMSE (left) and Mean Error (right) over the SHMÚ radar composite

Similar studies that have been carried out in the PPVG on comparison of radar data with rain gauge data have shown in general that RMSE error associated with radar fields depends considerably on radar minimum visible height above the rain gauge especially in mountainous countries. In lowlands this dependence is not so significant, but no negligible. The reason can be the location of radar sites at the top of hills and impossibility of the lowest elevation to reach the lowland's surface. In case of Slovakia the PR-RMSE error of radar accumulated fields is between 60-90%, with an average PR-RMSE value of 69,3%. Mean Error specified for 24-hours cumulated precipitation is -4,42mm or converted into instantaneous precipitation -0,184 mm/h. RMSE specified for 24-hours cumulated precipitation is 9,48mm or converted into instantaneous precipitation 0,395 mm/h.

Complete SHMU study is available on the H SAF ftp server:

/hsaf/WP6000/WP6100/precipitation/WG_groups/WG2-radar/WG-2-3_radar quality indication_v1.doc



A1.8 Ground Data in Turkey

Rain gauge

The network

356 Automated Weather Observation Station (AWOS) distributed over the country are used for the validation of the satellite precipitation products in the H SAF project. The average distance between the AWOS sites is 40.5 km.



Figure 46: Map Turkish rain gauge stations

The instruments

The gauge type of the network is tipping bucket where each has a heated funnel. The minimum detection capability of the gauge is 0.2 mm per tip. In the maximum capacity of the instrument is 720 mm/h at most. The operational accumulation interval is 1 minute, so that alternative cumulation intervals such as 5, 10, 20, 30 minutes are possible.

Data processing

Quality control

High quality of the ground data is critical for performing the validation of the precipitation products. The validation results or statistics can provide meaningful feedbacks for the product developers and additionally the products can be used reliably only if there is a confidence present about the ground data at a certain level. For this reason, some predefined quality assurance (QA) tests are considered for the precipitation data in order to define the confidence level. First of all, a flagging procedure is defined as described in next table

QA Flag Value	QA Status	Brief Description
0	Good	Datum has passed all QA Test
1	Suspect	There is concern about accuracy of datum
2	Failure	Datum is unstable

 Table 33: The precipitation data QA tests are summarized as follows.

Range Test



This test is used to see if any individual precipitation observation falls within the climatological lower and upper limits. The test procedures applied in the study are as follows.

IF $\text{Lim}_{\text{Lower}} \leq \text{Obser}_{j,t} \leq \text{Lim}_{\text{Upper}}$ **THEN** $\text{Obser}_{j,t}$ flag is 'Good' **IF** $\text{Obser}_i > \text{Lim}_{\text{Upper}}$ **OR** $\text{Obser}_{j,t} < \text{Lim}_{\text{Lower}}$ **THEN** $\text{Obser}_{j,t}$ flag is 'Failure'

 Lim_{Lower} and Lim_{Upper} thresholds are separately determined for each station on a monthly basis. At any specific site, all the observed monthly data is considered for determination of the upper and lower limits. By applying this test, each observation is flagged either by 'Good' or 'Failure' label depending on the comparison tests mentioned above.

Step Test

It is used to see if increment/decrement between sequential observations in time domain is in acceptable range or not. The applied test procedure is,

 $\begin{array}{ll} \textbf{IF} & |Obser_{j,t}\text{-}Obser_{j,t-1}| < Step_j \textbf{ THEN } Obser_{i,t} \text{ flag is 'Good'} \\ \textbf{IF} & |Obser_{j,t}\text{-}Obser_{j,t-1}| > Step_j \textbf{ THEN } Obser_{i,t} \text{ flag is 'Suspect'} \end{array}$

Step_j threshold is determined again for each site on a monthly basis. For each site, the dataset containing the absolute difference of the sequential observations is determined by considering the observations for the matching month. The 99.9 % cumulative histogram value of the dataset is set as the Step_j threshold for the related site and month.

Persistence Test

Persistence test is used to determine if any group of observations are due to instrument failures. The test procedure applied is defined as,

- **IF** $T < \Delta$ **THEN** Flag for all Obser in T : 'Good'
- **IF** $T > \Delta$ **THEN** Flag for **all** Obser in T : 'Suspect'

where T is the total number of the sequentially repeating observations forward in time and Δ is the possible maximum number of sequentially repeating observations. As in the other two tests, Δ threshold is determined for each site on a monthly basis. For any site, the data belonging to the same month is taken into account to determine the repeating number of the sequential observations. Then, 99.9 % cumulative histogram value of the repeating number dataset is assigned as the Δ amount for the corresponding site and month. Since there is a high possibility of no-precipitation data (zero), the sequential zero observations are excluded in this test during the determination of the Δ threshold amount and application of the test.

QA Test procedure

By applying the control procedures of the QA test mentioned above, each individual precipitation observation receives three flags referring to the corresponding test. For the corresponding observation if all the test flag is not 'Good' then the observation is excluded from the validation process.

Use of spatialization technique

Due to the time and space structure of precipitation and to the sampling characteristics of both the precipitation products and observations used for validation, care has to be taken to bring data into comparable and acceptable range. At a given place, precipitation occurs intermittently and at highly fluctuating rates. Various maps, time series analysis, statistical and probabilistic methodologies are employed in the validation procedure classically, but some additional new aspects such as the spatial coverage verification model of point cumulative semivariogram (PCSV) approach (Şen and Habib, 1998) are proposed for usage in this work.



Each precipitation product within the H SAF project represents a foot print geometry. Among these, H01 and H02 products represent an elliptical geometry while H03B and H05 have a rectangular geometry. On the other hand, the ground observation (rain gauge) network consists of point observations. The main problem in the precipitation product cal/val activities occurs in the dimension disagreement between the product space (area) and the ground observation space (point). To be able to compare both cases, either area to point (product to site) or point to area (site to product) procedure has to be defined. However, the first alternative seems easier. The basic assumption in such an approach is that the product value is homogenous within the product footprint. Next figure presents satellite footprint (FOV) centers of the H01 and H02 products, an elliptical footprint for the corresponding center (area within the yellow dots) and Awos ground observation sites. The comparison statistic can be performed by considering just the sites in the footprint area. Although this approach is reasonable on the average but it is less useful in spatial precipitation variability representation. The comparison is not possible when no site is available within the footprint area.



Figure 47: H01 and H02 products footprint centers with a sample footprint area as well as the Awos ground observation sites.

Alternatively, the point to area approach is more appealing for the realistic comparison of the precipitation product and the ground observation. This approach is simply based on the determination of the true precipitation field underneath the product footprint area. To do so, the footprint area is meshed and precipitation amounts are estimated at each grid point by using the precipitation observations at the neighboring Awos sites as shown in Figure 48. A 3x3 km grid spacing is considered for the products with elliptical geometry while 2x2 km spacing is considered for the products with rectangular geometry. For any grid point, Awos sites within the 45 km for the time period of April-September (convective type) and 125km for the rest (stratiform type) are taken into consideration. At each grid point, the precipitation amount is estimated by,

$$Z_{m} = \frac{\sum_{i=1}^{n} W(r_{i,m}) Z_{i}}{\sum_{i=1}^{n} W(r_{i,m})}$$
(4.13.1)

where Z_m is the estimated value and $W(r_{i,m})$ is the spatially varying weighting function between the i-th site and the grid point m.



Figure 48: Meshed structure of the sample H01 and H02 products footprint.

Determination of the $W(r_{i,m})$ weighting function in Equation 1 is crucial. In open literature, various approaches are proposed for determining this function. For instance, Thiebaux and Pedder (1987) suggested weightings in general as,

$$W(\mathbf{r}_{i,m}) = \begin{cases} \left(\frac{\mathbf{R}^2 - \mathbf{r}_{i,m}^2}{\mathbf{R}^2 + \mathbf{r}_{i,m}^2}\right)^{\alpha} & \text{for } \mathbf{r}_{i,m} \le \mathbf{R} \\ \mathbf{0} & \text{for } \mathbf{r}_{i,m} \ge \mathbf{R} \end{cases}$$
(4.13.2)

where R is the radius of influence, $r_{i,m}$ is the distance from point i to point m to the point and α is a power parameter that reflects the curvature of the weighting function. Another form of geometrical weighting function was proposed by Barnes (1964) as,

$$W(\mathbf{r}_{i,m}) = \exp\left[-4\left(\frac{\mathbf{r}_{i,m}}{\mathbf{R}}\right)^{\alpha}\right]$$
(4.13.3)

Unfortunately, none of these functions are observation dependent but suggested on the basis of the logical and geometrical conceptualizations only. They are based only on the configuration, i.e. geometry of the measurement stations and do not take into consideration the natural variability of the meteorological phenomenon concerned. In addition, the weighting functions are always the same from site to site and time to time. However, in reality, it is expected that the weights should reflect to a certain extent the regional and temporal dependence behavior of the phenomenon concerned.

For the validation activities, the point cumulative semi-variogram technique proposed by Şen and Habib (1998) is used to determine the spatially varying weighting functions. In this approach, the weightings not only vary from site to site, but also from time to time since the observed data is used. In this way, the spatial and temporal variability of the parameter is introduced more realistically to the validation activity.

Matching approach



The temporal and spatial matching approaches are applied separately in the validation of the satellite products. As for the temporal matching, the product time is taken into account and 5 minute window (t-2 to t+3) is considered for estimation of the average rainrate for each site.

For the spatial matching, the mesh grid size of 3kmX3km is constructed for each IFOV area. For each grid point, the rainrate is estimated by taking the 5 minute averaged rainrate amounts observed at the nearby AWOS sites within the radius distance of 45 km(for convective type) or 125 km(for stratiform type) considering the weighting of each site with respect to the grid point(Equation 1). The weighting amounts are derived from the spatially varying weighting functions obtained by using the semi-variogram approach (Şen and Habib,1998). Finally, the Gaussian filter is applied to the estimations at the mesh grid of the IFOV area to get the average rainrate. Then, this amount is compared with the satellite precipitation product amount for the validation purposes.



Appendix 3 Continuous statistical scores and maps

H64 vs Grd	RADA	R LAND	RADA	R SEA	RADAR	COAST	GAUG	E LAND	OVE	RALL
	≥5 mm/24h	≥10 mm/24h	≥5 mm/24h	≥10 mm/24h	≥5 mm24/h	≥10 mm/24h	≥5 mm/24h	≥10 mm/24h	≥5 mm/24h	≥10 mm/24h
Nsat	353,590	147,229	22,582	15,783	23,822	12,457	1,286,249	467,527	1,686,243	642,996
Nref	51,753	28,182	23,316	12,368	10,633	6,005	117,733	46,679	203,435	93,234
ME (mm)	0.62	-2.54	3.96	4.69	2.47	1.45	3.03	-0.49	2.50	-0.30
SD (mm)	14.02	16.07	25.64	30.93	22.18	26.25	13.73	17.45	16.17	20.17
MAE (mm)	9.79	11.70	15.17	19.66	13.77	16.96	9.23	11.39	10.29	12.94
МВ	1.04	0.88	1.27	1.22	1.16	1.07	1.26	0.97	1.19	0.99
сс	0.43	0.38	0.43	0.39	0.40	0.36	0.29	0.24	0.35	0.30
RMSE (mm)	14.03	16.27	25.94	31.28	22.32	26.29	14.06	17.45	16.36	20.17
FSE (%)	92	74	178	148	145	121	119	91	124	99

Table 34: Continuous statistical scores for H64 vs Ground over European area.



Figure 49: FSE maps for H64 in comparison with radar (on the left) and with gauge (on the right). Median value is also indicated.



Figure 50: Correlation maps for H64 in comparison with radar (on the left) and with gauge (on the right). Median value is also indicated.



Appendix 4 Multi-categorical statistics

H64 VS Ground		RADAR LAND	RADAR SEA	RADAR COAST	GAUGE LAND	OVERALL
	POD	0.93	0.49	0.64	0.93	0.89
≥0.25	FAR	0.67	0.15	0.50	0.74	0.72
mm/24h	MISS	0.07	0.51	0.36	0.07	0.11
	CSI	0.32	0.45	0.39	0.25	0.27
	POD	0.90	0.53	0.66	0.91	0.87
≥1	FAR	0.80	0.24	0.63	0.84	0.82
mm/24h	MISS	0.10	0.47	0.34	0.09	0.13
	CSI	0.20	0.46	0.31	0.16	0.17

 Table 35: Probability Of Detection (POD), False Alarm Ratio (FAR), Missing (MISS) and Critical Success Index (CSI) for H64 vs Radar over Land, Sea, Coast, Gauge over Land and Overall surfaces for different accumlated precipitation thresholds over the European area.



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Figure 51: RMSE map over FD area. Mean value equals 5.4 mm/24h.





Figure 52: CC map over FD area. Mean value equals 0.61 (between threshold and target).



Figure 53: CC maps over FD area computed for best and worst month: February 2019 (on the left) and May 2019 (on the right). Mean values are also indicated.



Appendix 6 Acronyms

AMSU	Advanced Microwave Sounding Unit
ATMS	Advanced Technology Microwave Sounder
ATS	Automatic Telemetric Station
AWOS	Automated Weather Observation Station
BE	Belgium
BfG	German Federal Institute of Hydrology
CAPPI	Constant altitude plan position indicator
CSI	Critical Success Index
DE	Germany
DPC	Italian Department of Civil Protection
ECMWF	European Centre for Medium-Range Weather Forecasts
EFOV	pixel Extension Field Of View
ENAV	Italian air navigation service provider
EU	European
FAR	False Alarm Ratio
FD	Full Disk
FMI	Finnish Meteorological Institute
GRD	Ground
GRISO	Rainfall Generator of Spatial Interpolation from Observation
HU	Hungary
IFOV	Instantaneous Field Of View
IMWM	Institute of Meteorology and Water Management - Poland
INCA	Integrated Nitrogen model for CAtchments
IRM	Institut Royal Meteorologique - Belgium
	Italy
HAF-UNMUA	Matialian Air Force – Operational Meteorological Center
MAE	Multiplicative Absolute Error
MB	Multiplicative Blas
METOD	Metaomological Operational Sociality
MUS	Microwaya Humidity Soundar
MW	MicroWave
NF	North-Fast
NW	North-West
OMSZ	Hungarian Meteorological Service
PL	Poland
POD	Probability Of Detection
PPVG	Precipitation Product Validation Group
PR	Precipitation Rate
PR-RMSE	Product Requirement Root Mean Square Error
PUM	Product User Manual
PVR	Product Validation Report
RD	Radar
RG	Raingauge
RMSE	Root Mean Square Error
SAF	Satellite Application Facility
SE	South-East
SHMU	Slovak hydrometeorological institute
SK	Slovakia
SRI	Surface Rainfall Intensity
STD	Standard Deviation
SW	South-West
TSMS	Turkish State Meteorological Service
TU	Turkey



(Product H64 – P-AC-SM2RAIN)

UCCUnique Common CodeWGWorking GroupZAMGZentralanstalt für Meteorologie und Geodynamik