

1 A Real-time Multi-source Flash Flood Verification Database in Support of WPC  
2 Research and Operations

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4 Brian A. Cosgrove (NOAA/NWS/OHD)

5 Faye Barthold, Thomas Workoff, Mark Klein, and David Novak (NOAA/NWS/WPC)

6 Wallace Hogsett (formerly NOAA/NWS/WPC)

7 Jonathan J. Gourley and Zac Flamig (NOAA/NSSL)

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9       Obtaining a complete and accurate assessment of flash flood occurrences is  
10 critical for verifying and improving the operational excessive rainfall and experimental  
11 flash flood forecasts produced by the National Weather Service’s Weather Prediction  
12 Center (NWS/WPC). Unfortunately, a single authoritative, comprehensive source of  
13 flash flood verification data does not currently exist. As such, WPC, with assistance  
14 from NOAA’s National Severe Storm Laboratory (NSSL), leveraged three real-time  
15 CONUS-wide hydrologic data sources to create a new experimental, merged, real-time  
16 verification dataset. These data sources include NWS flash flood Local Storm Reports  
17 (LSRs), NSSL Meteorological Phenomena Identification Near the Ground (mPING)  
18 reports, and United States Geological Survey (USGS) stream gage measurements. While  
19 each of these datasets is not without weaknesses, they also feature complementary  
20 strengths.

21       National Weather Service LSRs are an official NWS product, provide relatively  
22 dense coverage and, in many cases, include rich descriptive language of the event.  
23 However, they can be subjective in nature and are dependent on people actually

24 witnessing and reporting an event; darkness, low population density, and poor weather  
25 itself are factors that can limit the number of events observed. Event categorization,  
26 location and timestamp errors can also occur, as can long time lags in the submission of  
27 reports. Like LSRs, mPING reports are dependent on submission by end users, though in  
28 this case via a mobile app or a website. mPINGs suffer from similar categorization,  
29 coverage and quality control issues. Unlike LSRs, they do not differentiate between  
30 floods and flash floods, though NSSL examination of the mPING reports indicates that  
31 they are mainly flash floods. Even with these weaknesses, as with many crowd-sourced  
32 social media-type applications, there is strong potential for this data source to quickly  
33 grow over time as more people become mPING reporters.

34         The third and final component of the multi-source flash flood database centers on  
35 USGS stream gage reports. The only objective and automated source of the three, stream  
36 gage reports are underutilized for flash flood verification, and, to the best knowledge of  
37 the authors, this research effort represents the first CONUS-wide attempt at leveraging  
38 them for real-time verification of flash flooding. The dataset is comprised of stage and  
39 discharge data collected at approximately 10,000 automated USGS stream gages across  
40 the CONUS every 5-60 minutes. Gage coverage is dense in many areas of the country,  
41 and as an added strength, gages provide fully automated operation in all weather  
42 conditions. However, the reports are necessarily limited to stream locations and are  
43 sparse in some sections of the interior western US. Natural stream flow signals can also  
44 be contaminated by regulation (i.e., dams and diversions). Differing from the LSRs and  
45 mPINGs, these reports are not event-based and are not specifically aimed at isolating  
46 flash floods or floods. Rather, they are simply ongoing reports of the stage and discharge

47 of the river at a particular location, whether during drought-, average-, or flood-type  
48 conditions. To extract natural flash flood event signals from these observations, real-time  
49 data from each USGS basin smaller than 2000 km<sup>2</sup> are passed through a series of  
50 sequential filters, which include checks for exceedance of minor flood stage or two year  
51 return period flow, rate of rise, and total stage change.

52 Stream gage observations are downloaded and processed via automated scripts  
53 alongside mPING and LSR data. Upon retrieval, reports from each data source are  
54 inserted into a searchable Postgres database. Latitude, longitude and timestamp values  
55 are stored for all three data sources, with additional attributes (e.g., descriptive event  
56 remarks, stream rate-of-rise) archived as available. Since direct comparison of point-type  
57 verification observations to areal-type flash flood forecasts is challenging, the combined  
58 data are plotted in both point-type and areal-type fashions. Underpinning this is the  
59 Practically Perfect (PP) analysis technique, which converts point observations to  
60 contoured areas and is used by the NWS Storm Prediction Center (SPC) for verifying  
61 severe weather forecasts (Hitchens et al., 2013; personal communication Israel Jirak,  
62 SPC). The goal of this approach is to produce a flash flood forecast map that resembles  
63 that which would be produced by someone with perfect knowledge of future flash flood  
64 events. While the PP approach is a relatively simplistic spatial approach and ignores  
65 basin boundaries, it is suitable for broad CONUS-scale verification applications and has  
66 proved valuable in verifying both WPC Flash Flood and Intense Rainfall Experiment  
67 (FFAIR) predictions and, in an experimental fashion, WPC excessive rainfall forecasts.  
68 Many other government, academic, and private agencies focus on flash floods, and it is

69 expected that this database will prove useful to a wide variety of applications within those  
70 groups.

71 This presentation will explore the details of the database and will cover several  
72 verification case studies that leverage the flash flood information contained therein.