

Cost Effective Flash Flood and Flood Warning using Existing Floodplain Models?

Seth Lawler, Dewberry

Dinakar Nimmala, Dewberry

Celso Ferreira, George Mason University



Introduction: Thinking Out loud

It's the **next thing**.

We are **not satisfied** leaving models on the shelf.

Dewberry Engineers **want** to do it.

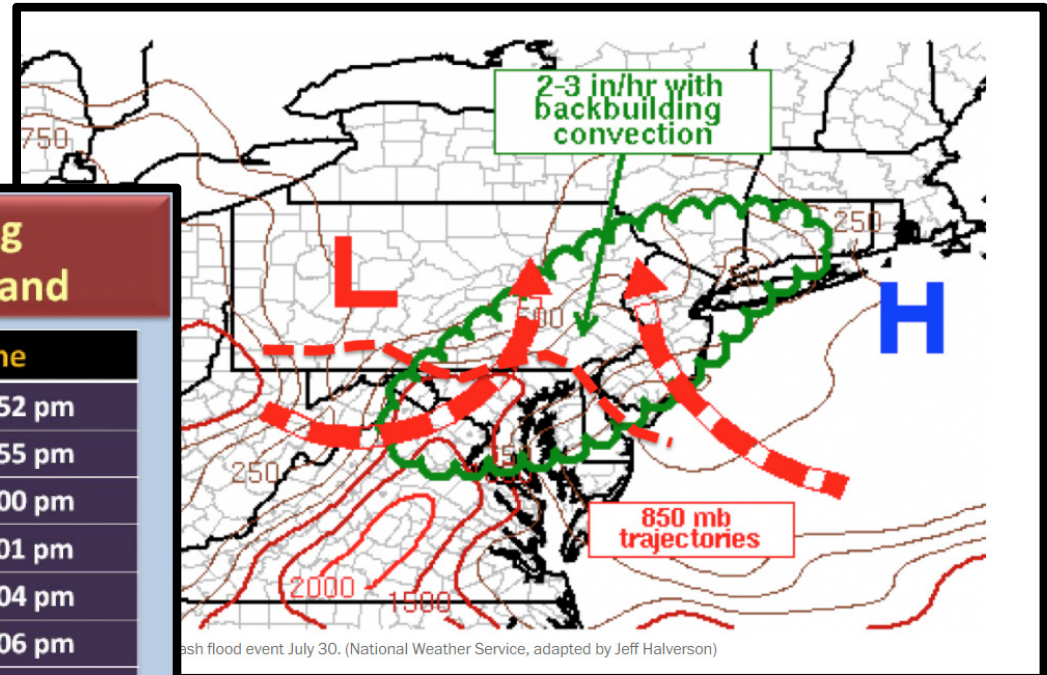
University Partners **are asking**.

Emergency Managers **are waiting**.

Opportunity to develop **open source solutions**.

Challenges for Forecasting: When?

Analysis at 5 p.m. that shows how air streams (heavy red, dashed arrows) at the 5,000-foot level were converging over Maryland.

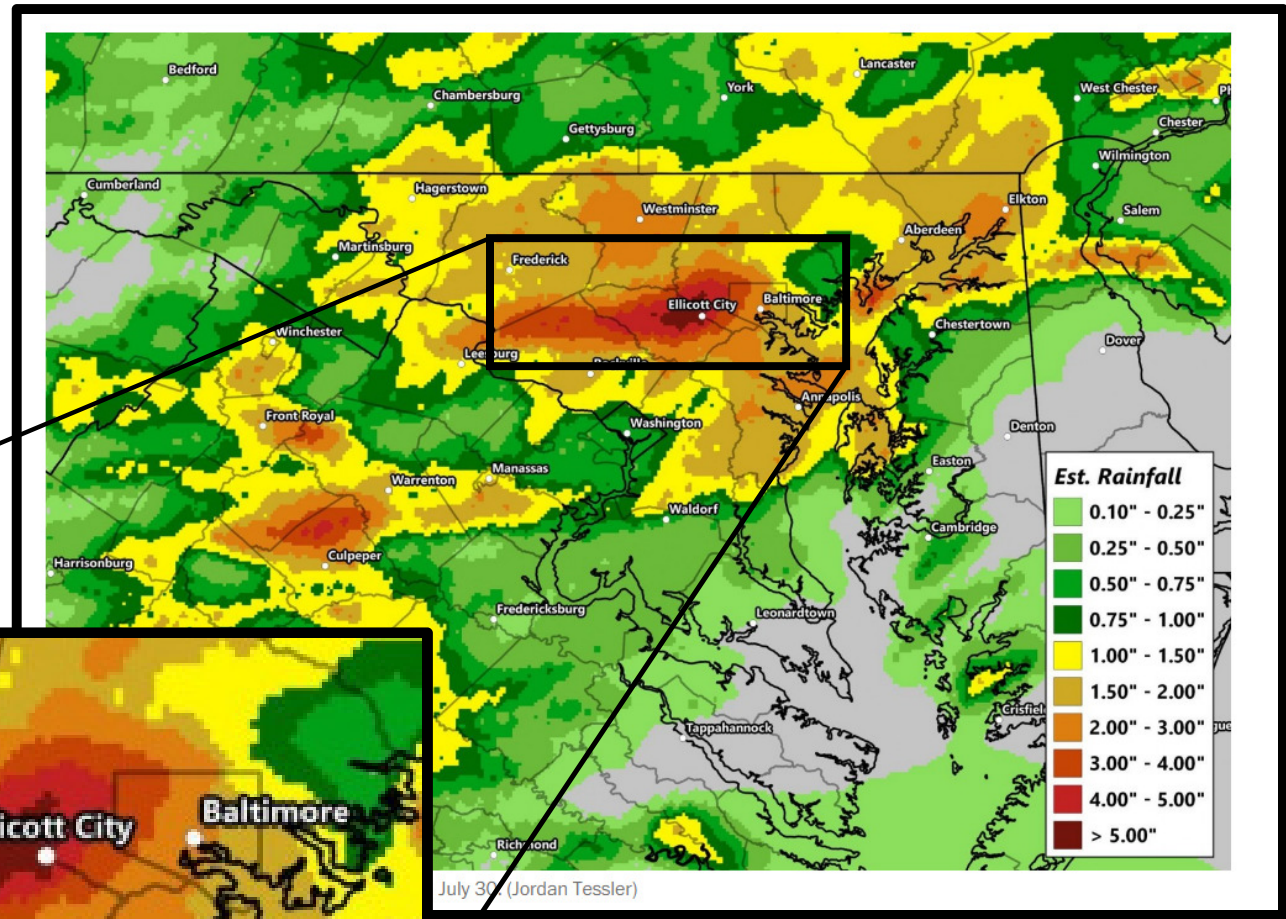
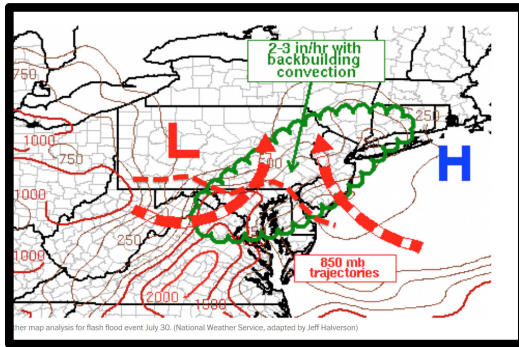


Historic Rainfall and Flash Flooding Saturday Night in Ellicott City, Maryland

Duration	Rainfall Total	Time
1 minute	0.20"	7:51-7:52 pm
5 minutes	0.80"	7:50-7:55 pm
10 minutes	1.44"	7:50-8:00 pm
15 minutes	2.04"	7:46-8:01 pm
20 minutes	2.48"	7:44-8:04 pm
30 minutes	3.16"	7:36-8:06 pm
60 minutes	4.56"	7:30-8:30 pm
90 minutes	5.52"	7:00-8:30 pm
2 hours	5.92"	6:45-8:45pm

The storm total rainfall at Ellicott City was 6.50 inches. Based on the preliminary precipitation frequency estimates in NOAA Atlas 14 from the nearest location, the rainfall amounts with duration 10 minutes to 2 hours statistically have a less 0.1% chance of occurring in any given year, or a 1 in 1000 year event.

Challenges for Forecasting: Where?



<https://www.washingtonpost.com/news/capital-weather-gang/wp/2016/08/01/this-is-how-an-off-the-charts-flood-ravaged-ellcitt-city/>



Forecasting Framework

1. Transform Hydrologic Models
2. Develop Data Acquisition & Processing Schema
3. Automate Forcing & Input/Output
4. Bias Correction & Modeling Tasks
5. Evaluate/Configure Hydraulic connection
6. Post Processing/Forecast Product
7. Latency & Utility

Background: GMU Flood Hazard Lab

Hydro-Meteorological Monitoring Network

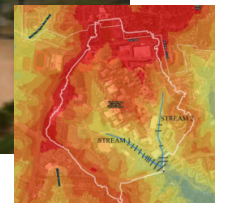
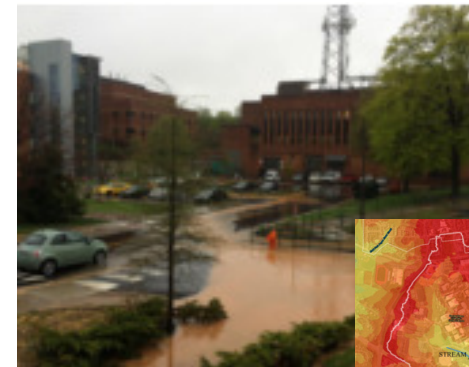
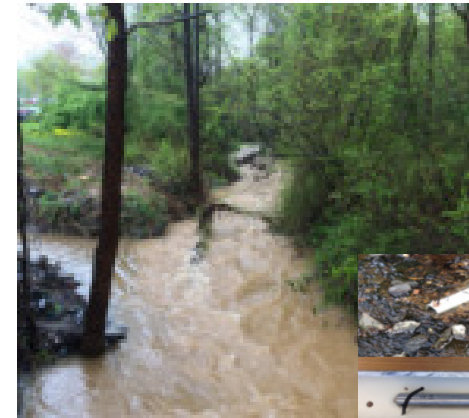
- ✓ Precipitation, wind, temperature...
- ✓ Water Levels, velocity, flow

Campus

- ✓ History of Flooding
- ✓ HEC-HMS
- ✓ HEC-RAS (1D2D), surveyed XS's

Poseidon Server for Real-time Flood Modeling

- ✓ Coastal Model : Running
- Riverine Model: In Progress



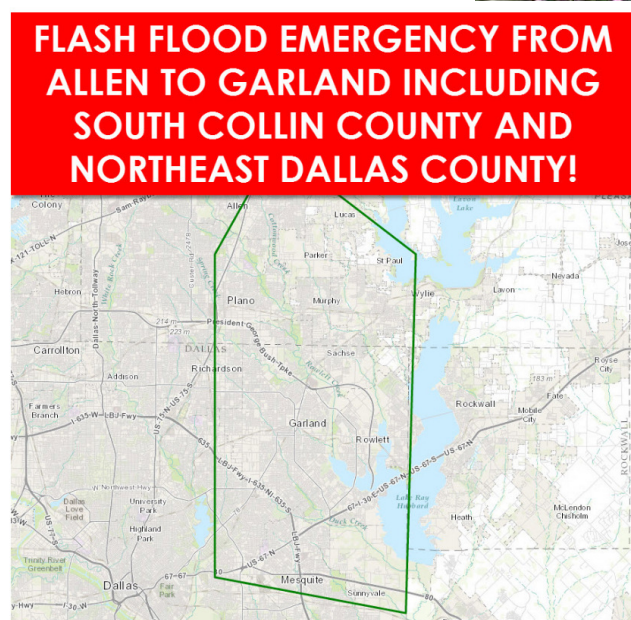
Background: Cameron Run Watershed

- High Resolution, Calibrated Models
- Nuisance/Flash Flooding
- Well Gaged
- Reservoir



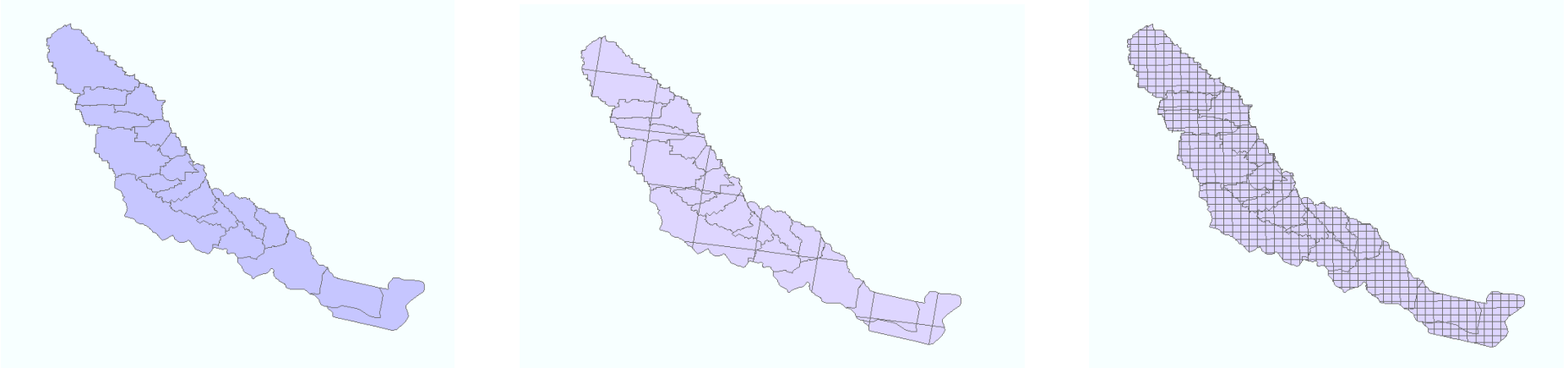
Background: Duck Creek Watershed

- Flash Flooding
- Structural Data
- Existing Models
- Extensive Ground Radar Network



HMS Model Set-Up: HRAP or SHG Grid

1. Transform Hydrologic Models



Result: A coarse meteorological grid forcing a highly resolved hydrologic model

Map HRAP Grid Cell to Intersecting HMS Subbasin:

HMS .mod file

```
HRAP.mod
1 Parameter Order: Xcoord Ycoord TravlLength Area
2 End:
3
4 Sub
5
6 End
7
8 Sub
9
10 End
11 End

SHG.mod
1 Parameter Order: Xcoord Ycoord TravlLength Area
2 End:
3
4 Subbasin: PB2.10
5 GridCell: 961 543 0.10000000000000000 0.61935
6 End:
7
8 Subbasin: CRtrib.11_1.13
9 GridCell: 961 544 0.10000000000000000 0.065952
10 GridCell: 962 544 0.10000000000000000 0.485423
11 End:
```

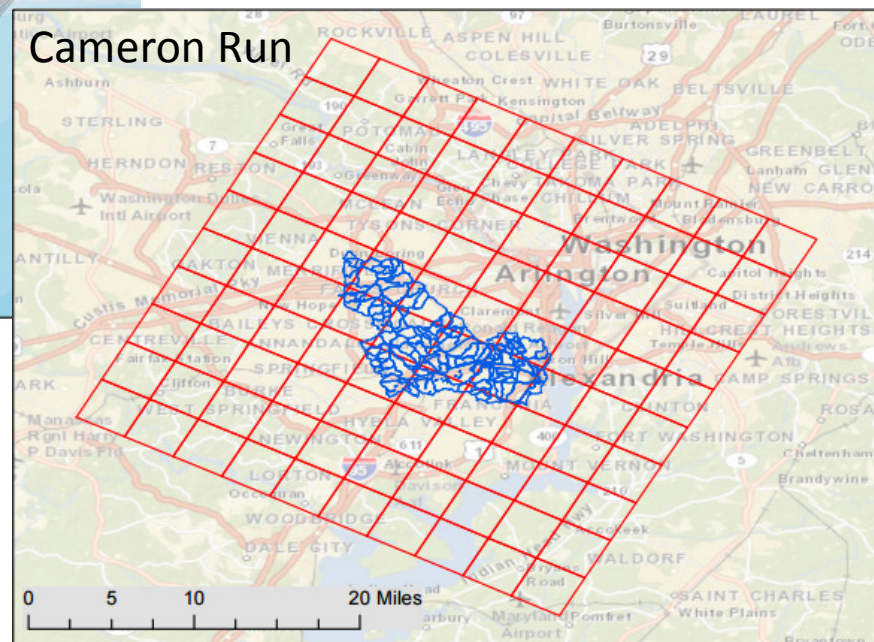
HMS Model Set-Up: QPF & HRAP

1. Transform Hydrologic Models



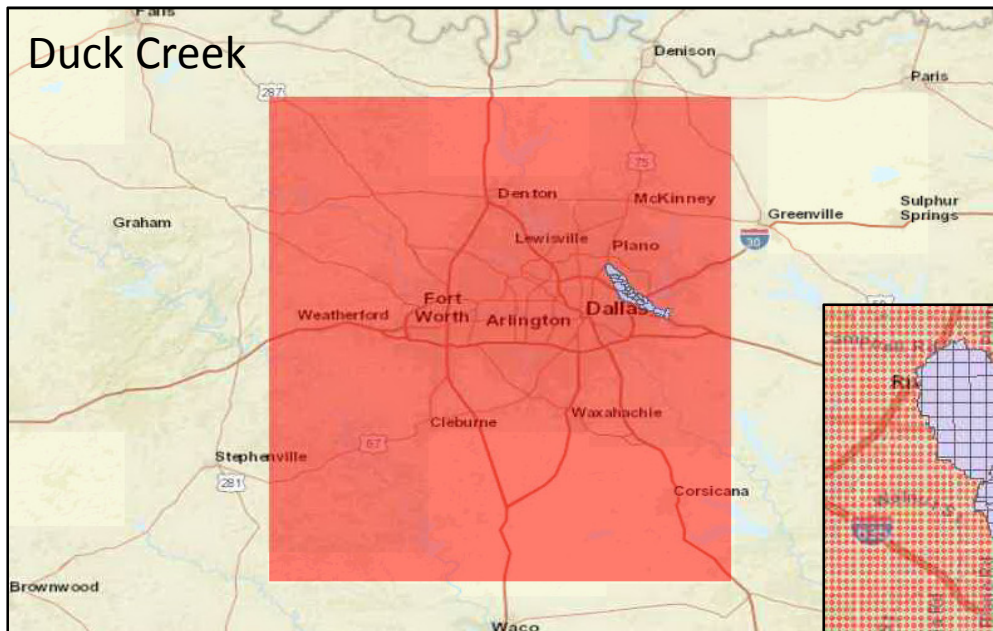
Mid-Atlantic River Forecast Center:
Quantitative Precipitation Forecast
(QPF) Coverage Area on the HRAP Grid

Extracted Area of Interest:
HMS Sub-basins for the
Cameron Run Watershed

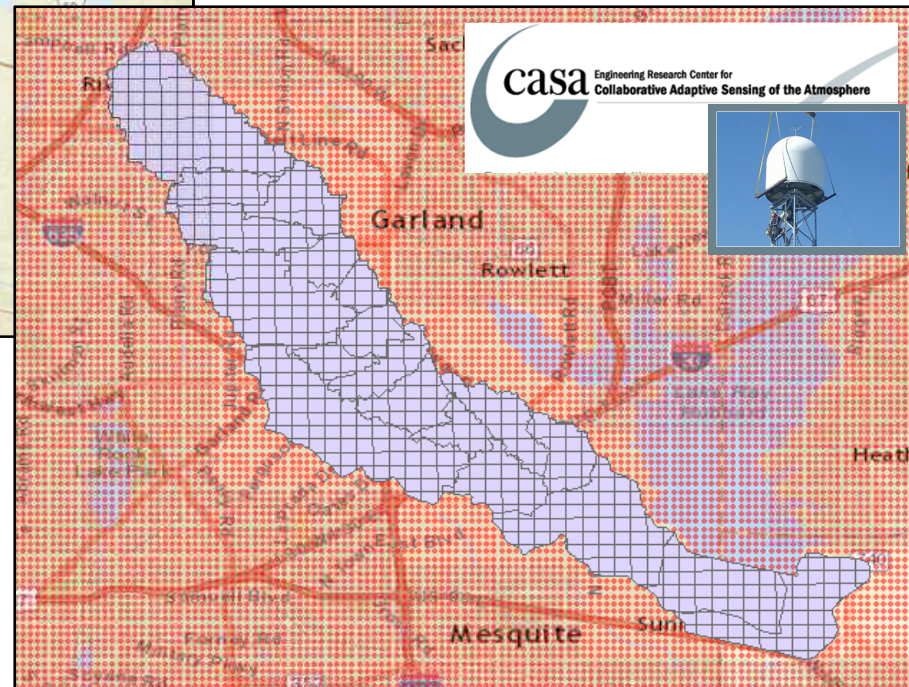


HMS Model Set-Up : Now-cast & SHG

1. Transform Hydrologic Models



Ground Radar Observations
at 10 min Past the hour: 250
meters



Now-casts available for 30
min-3 hour leading to and
during events

HMS Model: Meteorological Forcing

2. Develop Data Acquisition & Processing Schema

Python Library Development

- Read Met File Data:
 - Grib/Grib2
 - XMRG
 - NETCDF
- Extract Area of Interest
- Resample Data
- Write DSS files
- Run HMS

```
class Grid(object):  
  
    def __init__(self, grid, coords, cellsize):  
        self.grid = grid  
        self.coords = coords  
        self.cellsize = cellsize  
  
    def lats(self):  
        f = Dataset(self.grid, 'r')  
        lats = f.variables['Latitude'][:]  
        return lats  
  
    def lons(self):  
        f = Dataset(self.grid, 'r')  
        lons = f.variables['Longitude'][:]  
        return lons  
  
    def precip(self):  
        f = Dataset(self.g  
        precip = f.variab  
        return precip
```

```
class dssWriter:  
  
    def __init__(self, name):  
        self.name = name  
  
    def date(self):  
        f = self.name.split('_')  
        date = datetime.strptime(f[0]+f[1], '%d%b%Y%H%M%S')  
        return date  
  
    def dss_start(self):  
        dss_start = self.date().strftime('%d%b%Y:%H%M')  
        return dss_start  
  
    def delta_t(self):  
        f = self.name.split('_')  
        interval = findall('\d+|\D+', str(f[2][:-3]))  
        scalar, unit = interval[0], interval[1]  
  
        if unit == 'min.':  
            unit = 'minutes'  
        elif unit == 'hour':  
            unit = 'hours'
```



Getting a Hydrograph: Automation

3. Automate Forcing & Input/Output

© Randy Glasbergen
glasbergen.com

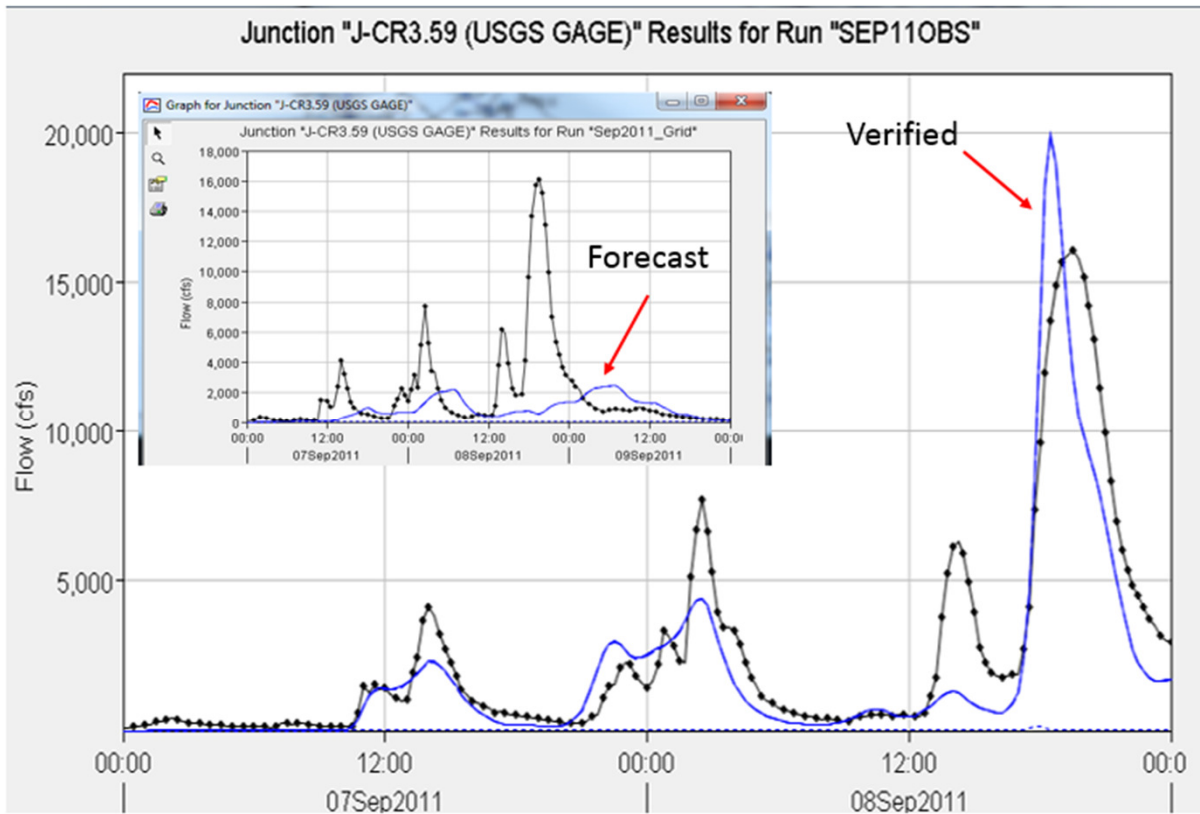


**“Tech support says the problem is located
somewhere between the keyboard and my chair.”**

How Dispensable is the Modeler?

Living with the hydrograph(s)

4. Bias Correction & Modeling Tasks



Where is the correction made....precipitation forecast, hydrograph?

When is bias correction Applied?

How?

HMS Model: Meteorological Forcing

4. Bias Correction & Modeling Tasks

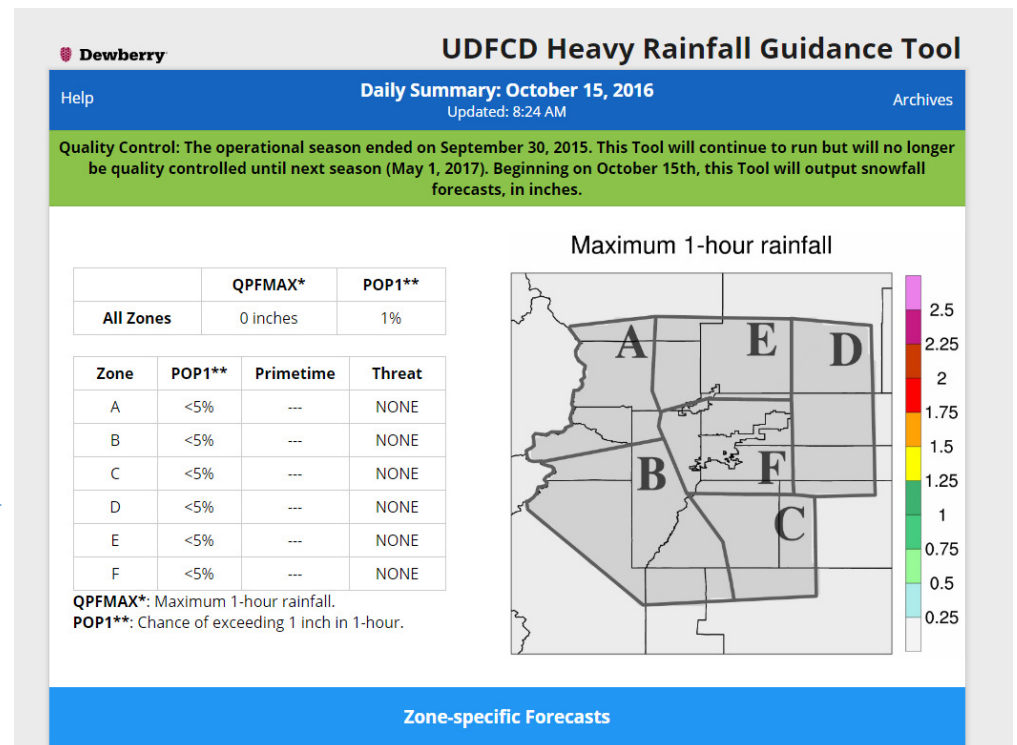
Table 4: Sources of QPF data.

Modeling center	Ensembles (Model/Family)	Total ensembles
National Severe Storms Laboratory (NSSL)*	1. arw / A 2. arw-ctl / A 3. arw-p1 / A 4. arw-n1 / A 5. arw-p2 / A 6. nmb-ctl / B 7. nmb-n1 / B 8. nmb-p1 / B 9. nmb-p2 / B	9
National Centers for Environmental Prediction (NCEP)*	10. hires-arw / C 11. hires-nmm / D 12. namnest-00Z / E 13. namnest-06Z / E	4
National Center for Atmospheric Research (NCAR)*	10 members (#14-23) Family: F	10*
	Total Ensembles	13 (ops) 23 (ops + testing)

*Research / operational center
*Operational center
*Not used in 2015 operations, only for testing

Creation of QPF's for Daily Colorado Flood Threat Guidance Tool

Developed by Dewberry Meteorologist & Hydro-meteorologists



Using all the tools

4. Bias Correction & Modeling Tasks

Statistical Methods

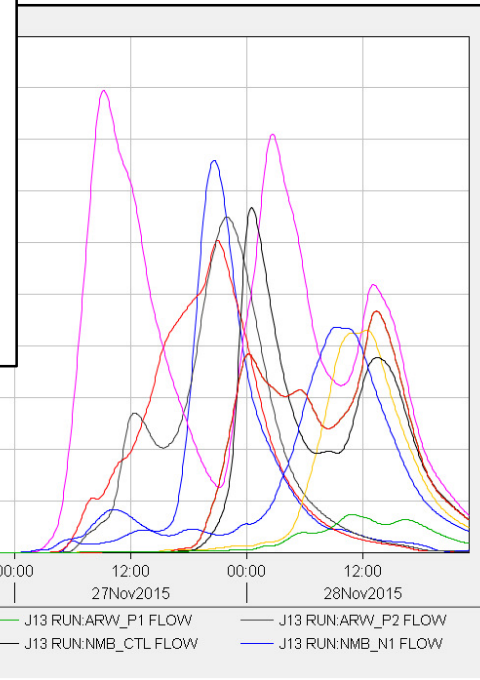
Ensemble Forecasts

Synthetic Storms

Table 4: Sources of QPF data.

Modeling center	Ensembles (Model/Family)	Total ensembles
National Severe Storms Laboratory (NSSL)*	1. arw / A 2. arw-ctl / A 3. arw-p1 / A 4. arw-n1 / A 5. arw-p2 / A 6. nmb-ctl / B 7. nmb-n1 / B 8. nmb-p1 / B 9. nmb-p2 / B	9
National Centers for Environmental Prediction (NCEP)*	10. hires-arw / C 11. hires-nmm / D 12. namnest-ooZ / E 13. namnest-o6Z / E	4
National Center for Atmospheric Research (NCAR)*	10 members (#14-23) Family: F	10*
Total Ensembles		13 (ops) 23 (ops + testing)

**Research / operational center*
**Operational center*
**Not used in 2015 operations, only for testing*

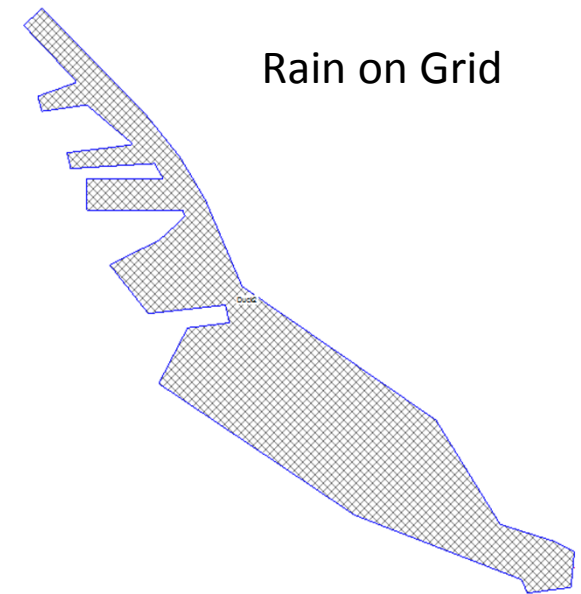
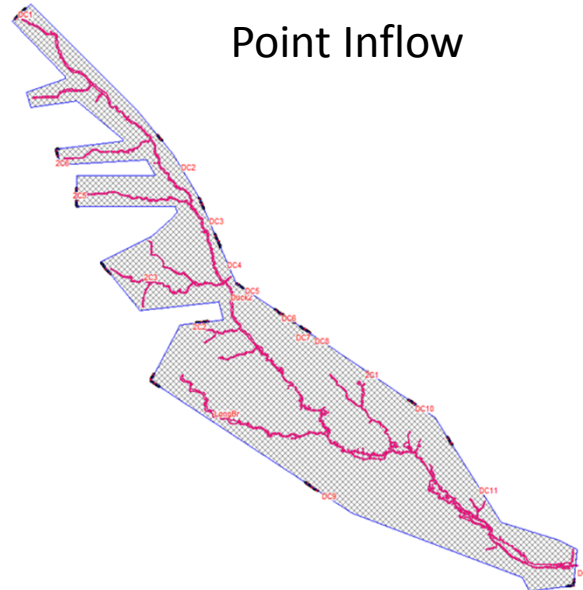
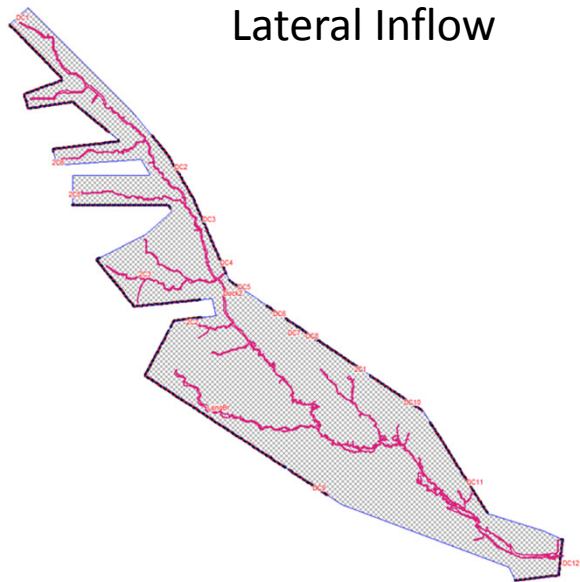


Data Assimilation

Rating Curves

RAS Model: Forcing

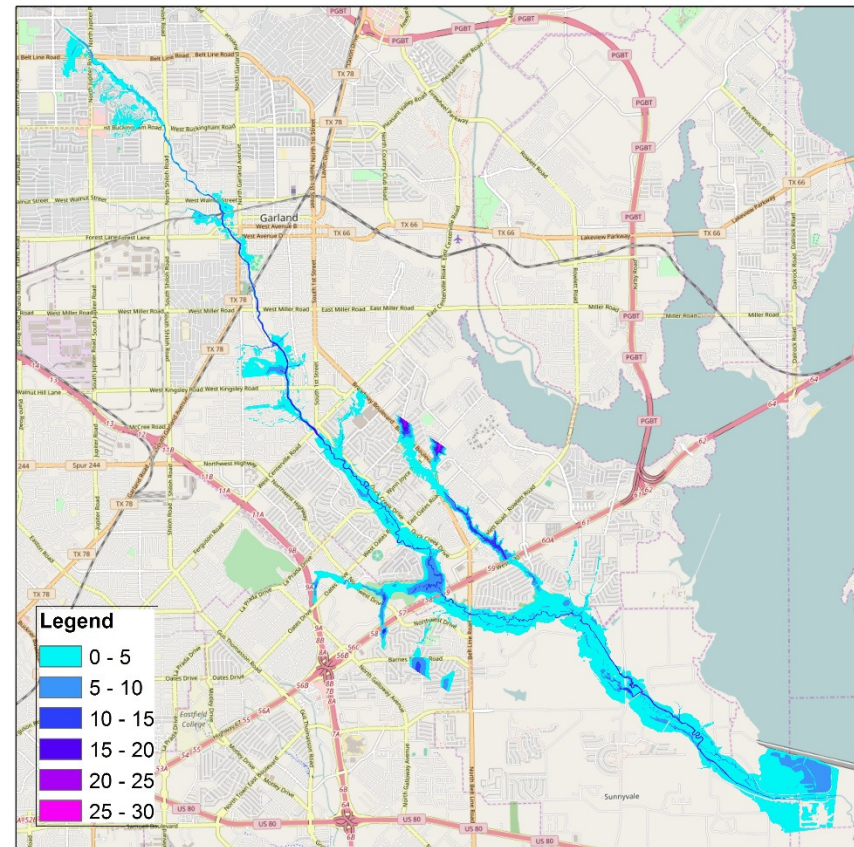
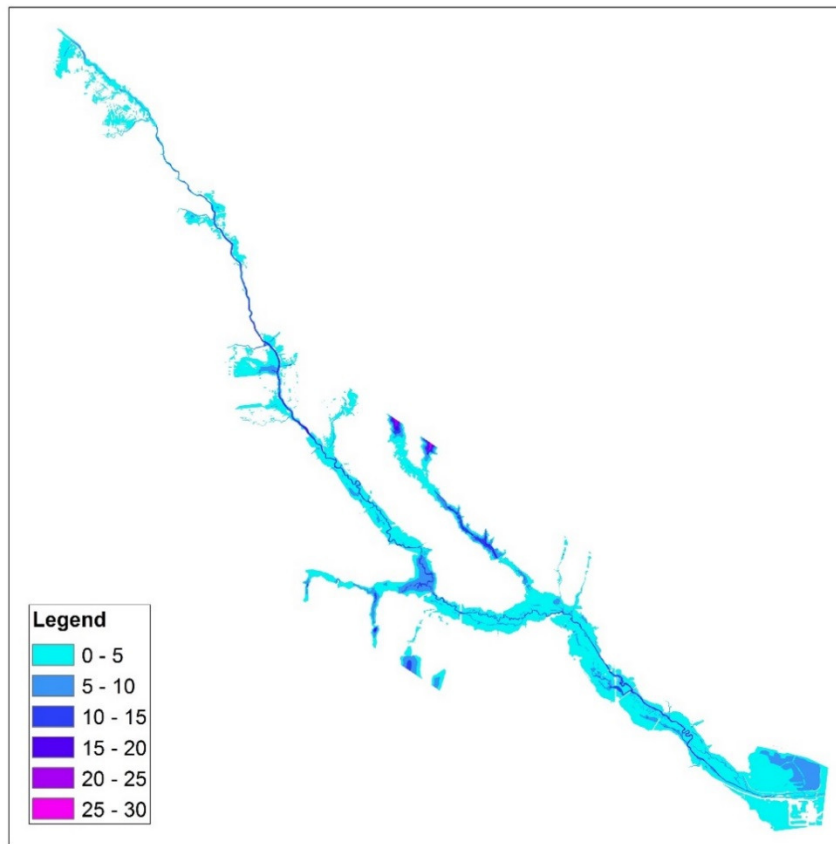
5. Configure Hydraulic connection: 2D



Cells	Cell Size (ft)	Computational Time-step (min)	Run Time (hh:mm:ss)
60,000	200	5	0:03:30
120,000	100		0:14:30
360,000	50		0:58:00

RAS Model: Post Processing

6. Post Processing/Forecast Product



HEC-RAS 5.0 Post Processing

Precip ==> Flow ==> Floodplain Extent

7. Latency & Utility

t0 = Precipitation Forecast Available

t1 = Precipitation Data processed

t2 = Hydrologic Simulation

t3 = Hydraulic Simulation

t4 = Floodplain/Flow Forecast Available

$$\text{Latency} = t_0 + t_1 + t_2 + t_3 + t_4$$

At hour 00, Latency (minutes) = 5 + 10 + 5 + 25: **45minutes**

Utility = Success – (False Alarms + Latency + Non-Detection)

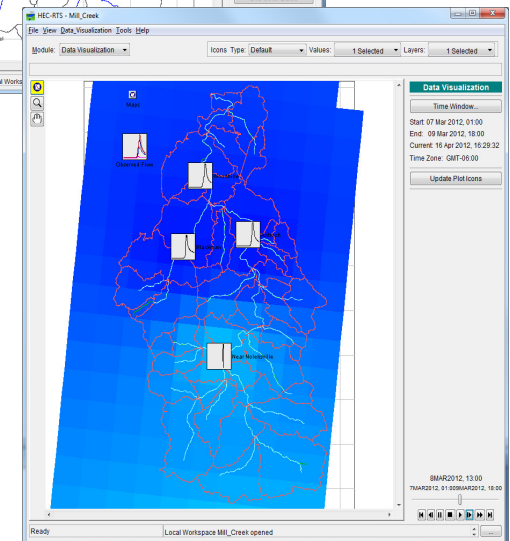
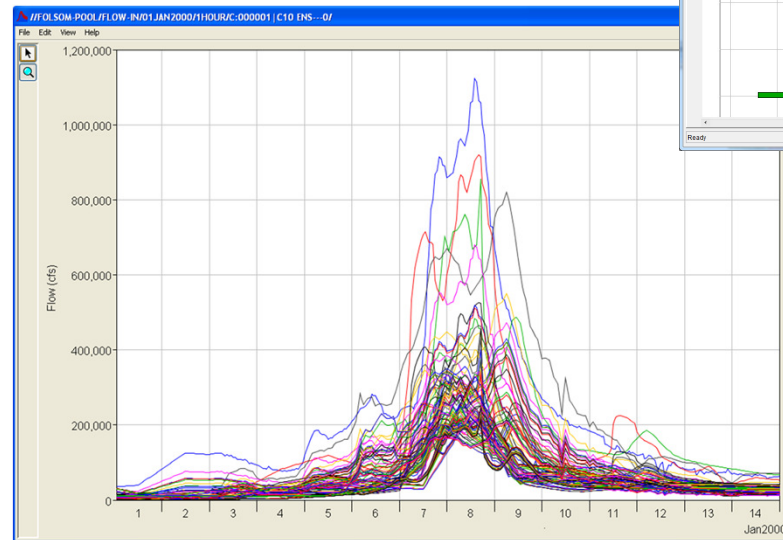
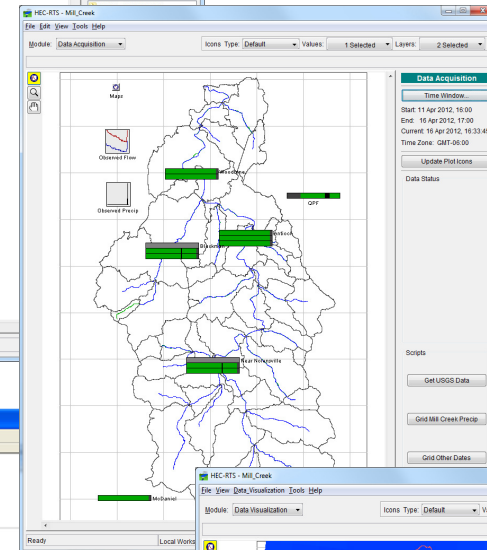
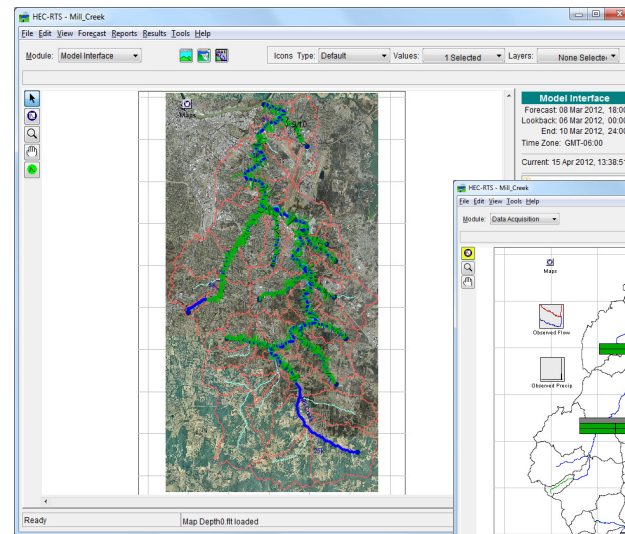
USACE

- Corps Water Management System (CWMS)

- HEC Products
- Oracle Database
- Unix Operating System
- Data Retrieval Client Based

- Real Time Simulation (RTS)

- HEC Products
- HEC-DSS
- Windows PC
- Scripting Capability

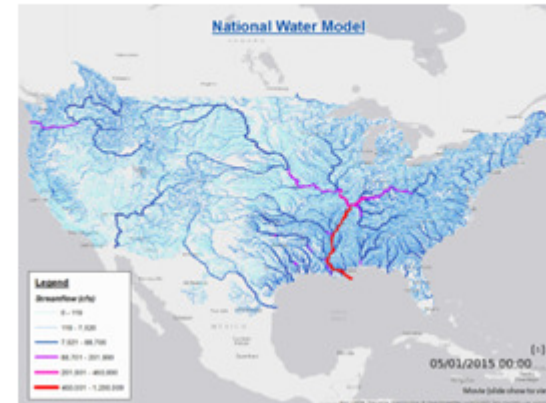


NWS: National Water Model

Current River Forecast Points (~3,600)



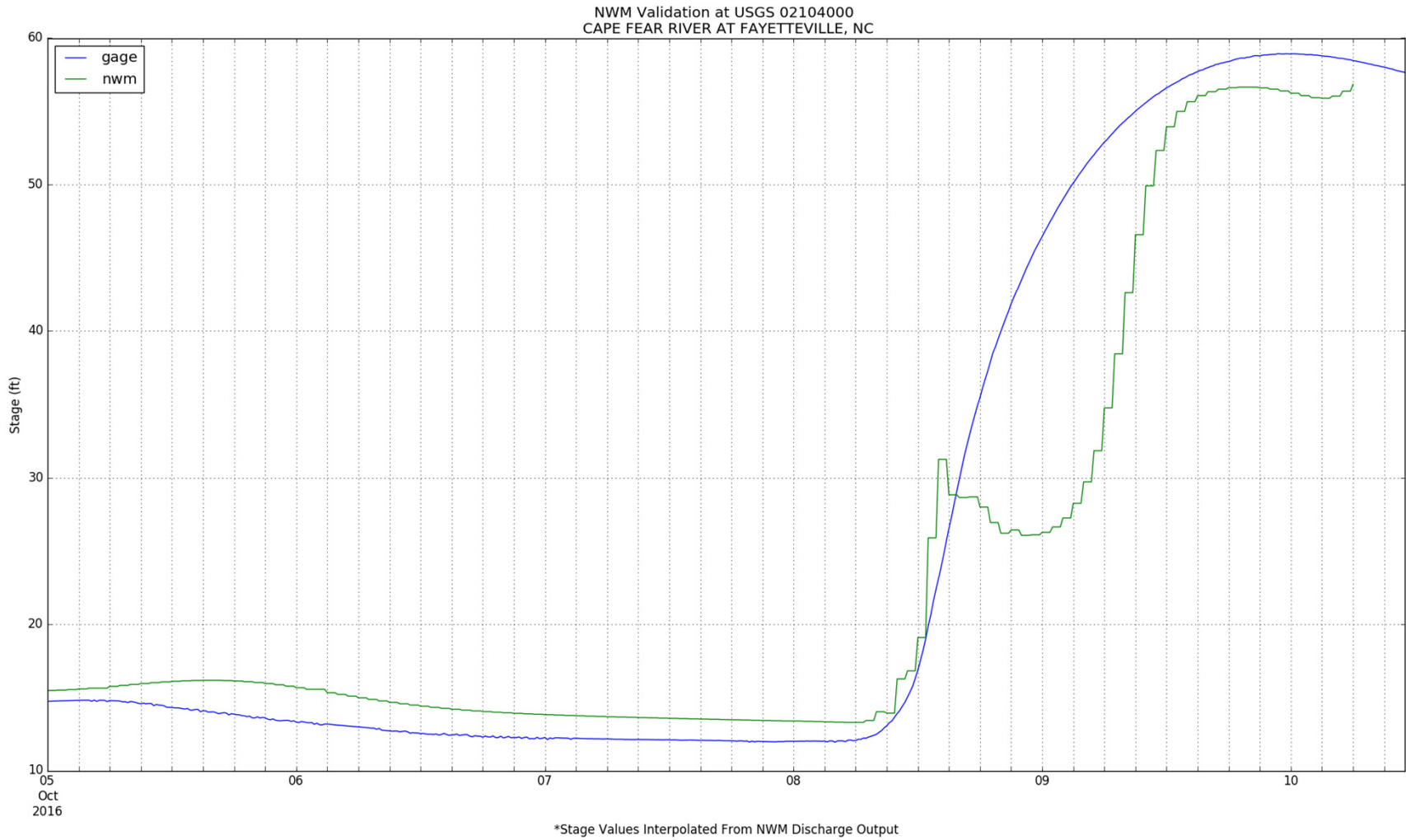
NWM Streamflow Output Points (~2.7 mil)



Model Details

Analysis & Assimilation	Short-Range	Medium-Range	Long-Range
Cycling Frequency			
Hourly	Hourly	Daily at 06Z	Daily Ens (16 mem)
Forecast Duration			
- 3 hrs	0-15 hours	0-10 days	0-30 days
Forecast Latency (latency of external forcing data accounts for most of delay)			
1 hour	1 hour 45 mins	6 hours	19 hours
Meteorological Forcing			
MRMS blend/ HRRR/ RAP bkgnd.	Downscaled HRRR/RAP blend	Downscaled GFS	Downscaled & bias-corrected CFS
Spatial Discretization & Routing			
1km/250m/NHDPlus Reach	1km/250m/NHDPlus Reach	1km/250m/NHDPlus Reach	1 km/NHDPlus Reach
Assimilation of USGS Obs			
Reservoirs (1260 water bodies parameterized with level pool scheme)			

NWS: National Water Model



Open Source Tools



GMU Ferreira Research Group

George Mason University, Department of Civil, Environmental and Infrastructure Engineering

Fairfax, VA <http://frg.vse.gmu.edu> cferrei3@gmu.edu



```
Code Issues 0 Pull requests 0 Projects 0 Wiki Pulse Graphs Settings
Branch: master FloodForecast / USGS_Data_Plotter.ipynb Find file Copy path
slawler Rename USGS_Data_Plotter (1).ipynb to USGS_Data_Plotter.ipynb 8370be1 6 days ago
1 contributor
255 lines (254 sloc) 6.01 KB Raw Blame History
Step-by-step guide to retrieving data from USGS API
Example:
Get streamflow data from selected gage for time period of interest & plot timeseries results
In [ ]: # Import Libraries
%matplotlib inline
import pandas as pd
import requests
import json
from datetime import datetime
from collections import OrderedDict

In [ ]: # Enter Desired Data
gage = "01646500" # USGS Gage
y0, m0, d0 = 2013, 4, 30 # Start date (year, month, day)
y1, m1, d1 = 2014, 5, 10 # End date
parameter = "00060" # Parameter
obser = "StreamFlow" # Observed data Requested
dformat = "json" # Data Format
url = "http://waterservices.usgs.gov/nwis/iv" # USGS API

In [ ]: # Create Datetime Objects
start = datetime(y0, m0, d0, 0)
stop = datetime(y1, m1, d1, 0)

# Format Datetime Objects for USGS API
first = datetime.date(start).strftime('%Y-%m-%d')
last = datetime.date(stop).strftime('%Y-%m-%d')
```



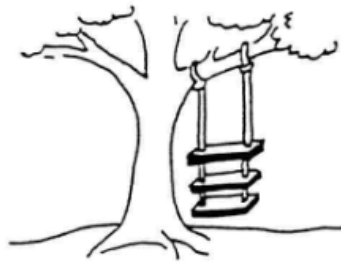
Open source, interactive data science and scientific computing across over 40 programming languages.



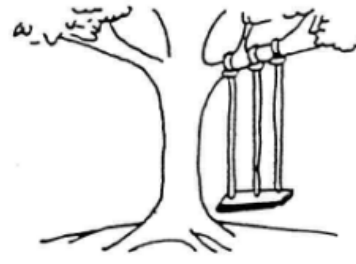
Summary & Conclusions

- In many cases, existing models can be refitted for forecasting
- Tools are readily available for transforming to an operational forecast capacity
- Forecast can be done at regional and local levels to augment national operations
- Water Resources Engineers need to become more familiar with the data types and data tools in use by meteorologist
- Flood Alerts: Operational forecast maps
- Competing Products are our friends

Questions?



As proposed by the project sponsors



As specified in the project request



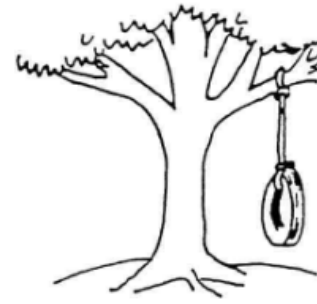
As designed by the senior analyst



As produced by the programmers



As installed at the user's site



What the user wanted

Tree Swing graphic by S High 1993 - from Businessballs.com/treeswing.htm 2013