National Weather Service (NWS)
Hydrologic Ensemble Forecast Service

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Outline

• Impetus for Hydrologic Ensemble Forecasts
• Concept and Design
• Implementation Status
• New Products and Services
• Future Efforts
Why use hydrologic ensemble forecasts?

National Research Council, 2006

“All prediction is inherently uncertain and effective communication of uncertainty information in weather, seasonal climate, and hydrological forecasts benefits users’ decisions. These uncertainties generally increase with forecast lead time and vary with weather situation and location. Uncertainty is thus a fundamental characteristic of weather, seasonal climate, and hydrological prediction, and no forecast is complete without a description of its uncertainty.” [emphasis added]
Why use hydrologic ensemble forecasts?

- Consistent feedback from customers and research community
  - 2006 National Research Counsel (NRC) report
  - 2008 Customer Feedback Insights (CFI) survey
  - River Basin Commission Stakeholder Engagements and Regional Water Conversations
- Aptima study (human centered engineering) validated need for water managers
- Multiple Internal NWS Service Assessments
  - Red River Floods in 1997 and 2009
  - Central U.S. Floods in 2008
  - Nashville Flooding in 2010
Why use hydrologic ensemble forecasts?

Goal: capture skill in weather ensembles

- Ensembles are standard in weather forecasting
- Include single models and “multi-model” ensembles
- Essential that water forecasts capture this information
- But, problematic to use them directly: wrong scale, biases
Why use hydrologic ensemble forecasts?

Goal: better-informed water decisions

At peak stage, HEFS says ~75% chance of Minor Flood or above, and ~25% chance of no flooding.

At peak stage, HEFS says ~50% chance of Minor Flood (25-75% of HEFS spread in Minor Flood band).

Risk of flooding

The current (issue) time is 12Z on 11 March.
**Why use hydrologic ensemble forecasts?**

**Goal: improve NWS hydrologic services**

<table>
<thead>
<tr>
<th>Feature</th>
<th>ESP (old service)</th>
<th>HEFS (new service)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Forecast time horizon</strong></td>
<td>Weeks to seasons</td>
<td>Hours to years, depending on the input forecasts</td>
</tr>
<tr>
<td><strong>Input forecasts (“forcing”)</strong></td>
<td>Historical climate data (i.e. weather observations) with some variations between RFCs</td>
<td>Short-, medium- and long-range weather forecasts</td>
</tr>
<tr>
<td><strong>Uncertainty modeling</strong></td>
<td>Climate-based. No accounting for hydrologic uncertainty or bias. Suitable for long-range forecasting only</td>
<td>Captures total uncertainty and corrects for biases in forcing and flow at all forecast lead times</td>
</tr>
<tr>
<td><strong>Products</strong></td>
<td>Limited number of graphical products (focused on long-range) and verification</td>
<td>A wide array of data and user-tailored products are planned, including standard verification</td>
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HEFS Service Level Objectives

Produce ensemble streamflow forecasts that:

• seamlessly span lead times from one hour to one year
• statistically calibrated (unbiased with reliable spread)
• consistent across time and space
• effectively capture information in NWS weather/climate models
• dependable (consistent with retrospective forecasts)
• adequately verified
• aid user’s decisions (compatible with Decision Support Systems)
**What is the HEFS?**

**Goal: quantify total uncertainty in flow**

- HEFS aims to “capture” observed flow consistently
- So, must account for total uncertainty & remove bias
- Total = forcing uncertainty + hydrologic uncertainty
What is the HEFS?

- **WPC/RFC forecasts** (1-5 days)
  - Meteorological Ensemble Forecast Processor (MEFP)
    - Correct forcing bias
    - Downscale (basin)
    - Merge in time
  - = forcing unc.
  - = hydro. unc.
  - = users

- **GEFS forecasts** (1-15 days)

- **CFSv2 forecasts** (16-270 days)

- **Climatology** (271+ days)

- **Ensemble Post-Processor (EnsPost)**
  - Correct flow bias
  - Add spread to account for hydro. model uncertainty

- **Hydrologic models (CHPS)**

- **Bias-corrected ensemble flow forecasts**

- **NWS and external user applications**

(MEFP forcing also available to users)
What is the HEFS?

**MEFP (“forcing processor”)**

- Does three things to raw forcing
  1. Adds sufficient spread to account for forecast errors
  2. Corrects systematic biases
  3. Downscales to basin
- The MEFP uses separate statistical models for temperature and precipitation
- The MEFP parameters are estimated using historical data (forecast archive or hindcasts)
- The outputs from the MEFP are FMAP and FMAT for a basin
What is the HEFS?

EnsPost ("flow processor")

- Does two things to flow forecast
  1. Adds spread to account for hydrologic model errors
  2. Corrects systematic biases
- Uses linear regression between observed flow and historical simulated flow (observed forcing)
- Scatter around line of best fit represents the hydrologic error (i.e. no forcing uncertainty)
- Prior observation ("persistence") also included in regression (not shown here)

\[
\hat{Z}_{\text{obs}}(t+1) = bZ_{\text{mod}}(t+1) + E(t+1)
\]
What is the HEFS?

- Forecast tool (real-time/hindcast)
- Supporting tool
- Future capability

- Raw weather and climate forecasts (GEFS, CFSv2,..)
- MEFP: pre-processor
- MEFP PE: parameters
- Unbiased forcing (basin scale)
- EVS: verification
- "Raw flow"
- "Corrected flow"
- Data assimilator
- Hydrologic Ensemble Processor
- Hydrologic data
- EnsPost: post-processor
- EnsPost PE: parameters
- GraphGen: products
- Verification results
HEFS Development Timeline

- 2011: Definition of initial version (1.0)

- 2012 - 2013: Develop prototype versions and training to five (Phase 1) RFCs

- 2014: Release version 1.0 to all RFCs and provide training to remaining RFCs
  - Includes tools and training for retrospective forecasts, i.e. validation

- 2015-16: RFCs begin implementation and validating at initial locations

- 2017-2018: RFCs expand implementation; OWP/RFCs address limitations of HEFS version 1
HEFS configured: 1,727 river locations
HEFS hindcast output validated: 164 river locations
Example of early application of HEFS

Managing NYC water supply

- Croton; Catskill; and Delaware
- Includes 19 reservoirs, 3 lakes; 2000 square miles
- Serves 9 million people (50% of NY State population)
- Delivers 1.1 billion gallons/day
- Operational Support Tool (OST) to optimize infrastructure, and avoid unnecessary ($10B+) water filtration costs
- HEFS forecasts are central to OST. The OST program has cost NYC under $10M
Example of early application of HEFS

“Mission critical decision to manage shutdown of RBWT Tunnel based on HEFS forecasts”

HEFS streamflow forecasts are used to optimize and validate the NYC OST for million/billion dollar applications

“HEFS forecasts critical to protecting NYC drinking water quality during high turbidity events”

Risk to water availability from Delaware Basin reservoirs

“HEFS forecasts help optimize rule curves for seasonal storage objectives in NYC reservoirs”

“HEFS forecasts used to determine risks to conservation releases”

(Cannonsville Reservoir Spillway)
Example of national HEFS product

AHPS short-range probabilistic product

See: http://water.weather.gov/ahps/
Example: short-range AHPS product

- Initial Experimental HEFS product depicts the uncertainty in short-range river forecasts
- Probability bands
  - Median (50%)
  - 25-75%
  - 10-90%
  - 5-95%
- 130 locations have experimental product on AHPS
- New river service locations will expand throughout 2016-17
- Feedback via survey
"HEFS showed ~85% chance of flooding for the 90 day period while historically it was about 40%. About a week after this forecast, the point almost hit moderate flood."

Eric Jones, ABRFC
Example: 10-Day Accumulated Reservoir Inflow

Accumulated Reservoir Inflow over next 10 days
- 1 day
- 2 day, etc.

Includes single value forecast
HEFS Challenges/Future Development

- Address performance in extreme events
- Effectively include the effects of reservoir regulation and other water management activities
- Improved hydrologic uncertainty estimation
- Automated Data Assimilation techniques
- Expand probabilistic product suite and leverage emerging NWS and NWC Data Services
Ensemble Forecast Challenge of a different kind

• Effectively communicate uncertainty information in a form and context that is useful to our customers
  • Education and training
  • Context, validation and verification
  • Compatibility with decision support tools

• Realize the full utility of this probabilistic information for optimized decisions
  • Internal NWS (WFO warning/hazard operations)
  • External partners and customers (Forecast Informed Reservoir Operations, EM response, etc.)
Summary

- As HEFS is rolled out over the next few years, applications for decision support are expected to expand dramatically
  - Warning Operations
  - Emergency Services
  - Resource Management (water supply, fisheries, ecosystems, recreation, navigation)
  - Hydropower production
- The future of NWS water intelligence resides in our ability to support optimized risk-based decisions
  - Ensemble-based (probabilistic) forecasting is foundational
  - Utilizing NCEP model reforecasts in HEFS dramatically expands the utility of output for decisions, but creates a requirement for NOAA/NWS to continue to produce robust reforecasts
Questions

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Backup slides
Phased science validation

Three phases completed & documented

Phase I: medium-range (1-14 days), frozen GFS (discont.)
  • Selected basins in four RFCs (AB, CB, CN, MA)

Phase II: long-range (1-330 days), GEFS+CFSv2+CLIM
  • Selected basins in MA and NE (in support of NYCDEP)

Phase III: medium-range, latest GEFS
  • Same design as Phase I to establish gain from GEFS

See: http://www.nws.noaa.gov/oh/hrl/general/indexdoc.htm

Two papers forthcoming in JoH special issue (Brown et al.)
Evaluation - *Strengths*

- Supports implementation of HEFS by identifying problems, optimizing calibration, and providing info for risk-based decision support
- The HEFS broadly performs as anticipated
  - Captures skill in forcing inputs (weather and climate), including skill from multiple temporal scales. Corrects for biases.
  - Produces streamflow forecasts that are skillful and reliable (includes hydrologic uncertainty and corrects for hydrologic biases)
  - For short/medium-range forecasts, the **GEFS adds meaningful extra lead time** compared to the frozen GFS (1-2 days for precipitation, 2-4 days for temperature, 1-2+ days for streamflow)
  - At most locations, EnsPost is critically important for reliable streamflow forecasting. Improvements have been made to address “noisy” time-series, but HEFS improvements are an ongoing process
What is the HEFS?

Ensemble Verification Service

- Supports verification of HEFS including for precipitation, temperature and streamflow
- Verification of all forecasts or subsets based on prescribed conditions (e.g. seasons, thresholds, aggregations)
- Provides a wide range of verification metrics, including measures of bias and skill
- Requires a long archive of forecasts or hindcasts
- GUI or command-line operation
Ensemble Forecasting Challenge

- Mesh ensemble forcings from short, medium, and long range techniques.

- Mesh ensemble forcings:
  - mesoscale wx models
  - medium range wx models
  - long range global circulation models

- Downscaling:
  - variable
  - time

- Forecaster skill

- Climate forecasts and indexes
- Ensure forecast ensembles maintain spatial and temporal relationships across many scales.

- Similarly, ensure consistency between precipitation and temperature is preserved in the forecast ensembles.

Irrational outcomes
Meteorological Ensemble Forecast Processor

**Short-Range**
- HPC/RFC forecasts

**Medium-Range**
- GEFS forecasts

**Long-Range**
- CFSv2 forecasts
- Climatology

**Ensembles**
- (days 1-5)
- (Day 1-14)
- (out to 8/9 months)
- (out to one year)

**Merging**
- Calibrated short- to long-range forcing ensembles
MEFP Methodology

Goal: Produce reliable ensemble forcings that capture the skill and quantify the uncertainty in the source forecasts.

Key Idea: Condition the joint distribution of single-valued forecasts and the corresponding observations using the forecast.

- Use forecasts from multiple models to cover short- to long-range.
- Model the joint probability distribution between the single-valued forecast and the corresponding observation from historical records.
- Sample the conditional probability distribution of the joint distribution given the single-valued forecast.
- Rank ensembles based on the magnitude of the correlation coefficients between forecast and observation for the time scales and associated forecast sources.
- Generate blended ensembles (using Schaake Shuffle) iteratively for all time scales from low correlation to high correlation.
Ensemble Forecast Challenge

- Accurately incorporate the impacts of reservoirs and diversions
- Reservoir models only approximate the actual operator decisions
- Reliable information about diversions is rarely available
- Significant impact on “actual” flows
- Very important to many user groups
Ensemble Forecasting Challenge

- Maintain coherence between deterministic and ensemble forecasts
What is needed for partners?

• Demonstrating the skill/value in these forecasts
  • Verification Information
  • Event specific
• Communicating effectively
  • Understandable
  • Enhanced formats
  • Data Services
• Commitment to overcoming hurdles
  • Policy
  • Legislative mandates
  • Bureaucracy
  • Entrenched process
Summary

- NWS has an established practice of probabilistic forecasting at the long range, but there is much more potential in that information to be exploited.

- At the short range, NWS is just beginning to really determine how best to use streamflow ensemble output.

- The communication and process challenges may be as difficult as the technical challenges of producing reliable/skillful ensemble forecasts.
Evaluation - *Weaknesses*

- For long-range forecasting, HEFS value is more modest
  - The CFSv2 generally adds little or no skill when compared to climatology
  - But MEFP should be able to capture skill with improved climate information

- Current Ensemble Post-Processor is limited
  - Lumps all hydrologic uncertainties and biases, which reduces ability to model the total uncertainty effectively
  - Primarily benefits short-range ensembles (lead time 1-5 days)
  - Issues with temporal consistency (discontinuities): needs science solution
  - Automated data assimilation needed as a long-term investment (reduces lumping together of uncertainties, hence reduces pressure on EnsPost)
  - Not designed for correcting regulated flows. Should leverage HRC work on this

- Some issues with the MEFP forcings
  - "Canonical events", which try to capture skill at different temporal scales. Causing problems with lack of smoothness/discontinuities in P and T.
  - Biases in Probability of Precipitation (PoP). Currently under investigation.
Phased validation of the HEFS

- Temperature, precipitation and streamflow validated
- See: [www.nws.noaa.gov/oh/hrl/general/indexdoc.htm](http://www.nws.noaa.gov/oh/hrl/general/indexdoc.htm)

1. **First phase**: short- to medium-range (1-15 days)
   - GEFS forcing used in the MEFP
   - Selected basins in four RFCs (AB, CB, CN, MA)

2. **Second phase**: long-range (1-330 days)
   - GEFS (15 days) and CFSv2 (16-270 days)
   - Climatology (∼ESP) after 270 days
   - Selected basins in MARFC and NERFC
MEFP forcing

- Skill of the MEFP with GEFS forcing inputs
- Positive values mean fractional gain vs. climatology (e.g. 50% better on day 1 at FTSC1)
- MEFP temperature generally skillful, even after 14 days
- MEFP precipitation skillful during first week, but skill varies between basins
HEFS streamflow

- Skill of HEFS streamflow forecasts (including EnsPost)
- Positive values mean fractional gain vs. climatology (≈ESP)
- HEFS forecasts consistently beat climatology (by up to 50% for short-range)
- Both MEFP and EnsPost contribute to total skill (separate contribution not shown)
Long-range forecasts

- Example of MEFP precipitation forecasts from Walton, NY
- Beyond one week of GEFS, there is little skill vs. climatology
- In other words, the CFSv2 adds little skill for the long-range (but forcing skill may last >2 weeks in flow)
- If climate models improve in future, HEFS can be updated
Summary and conclusions

Ensemble forecasts are the future

- Forecasts incomplete unless uncertainty captured
- Ensemble forecasts are becoming standard practice
- HEFS implementation, products, and validation is ongoing and expanding
- Initial validation results are promising

HEFS will evolve and improve

- Science and software will improve through feedback
- Guidance will improve through experience
- We are looking forward to supporting end users!
Additional resources


What are ensemble forecasts?

A collection of forecasts to capture uncertainty

- Single-valued forecasts are known to be imperfect (data and models contain errors)
- For example, multiple weather models predict multiple hurricane tracks →
- Ensemble forecasts capture these uncertainties by producing an “ensemble” of weather (or water) forecasts
- Each ensemble member represents one possible outcome (e.g. one track)

Hurricane Irene track forecasts, 08/22/11
What are ensemble forecasts?

A collection of forecasts to capture uncertainty

- Single-valued forecasts are known to be imperfect
- An ensemble provides a collection of forecasts
- Each ensemble member is one possible outcome
Why use hydrologic ensemble forecasts?

Demand from the science community
• Single-valued forecasts are primitive and can mislead
• Ensemble techniques are rapidly becoming standard

Demand from operational forecasters
• For **simple** and **objective** ways to assess uncertainty
• For clear products to communicate uncertainty

Demand from users of water forecasts
• Increasingly, water decisions seek to evaluate **risks**
• Range of possible outcomes needed to assess risk