

Project 3-4 (CAFFEWS)

Uncertainty Estimation through Bayesian Forecasting System (BFS)

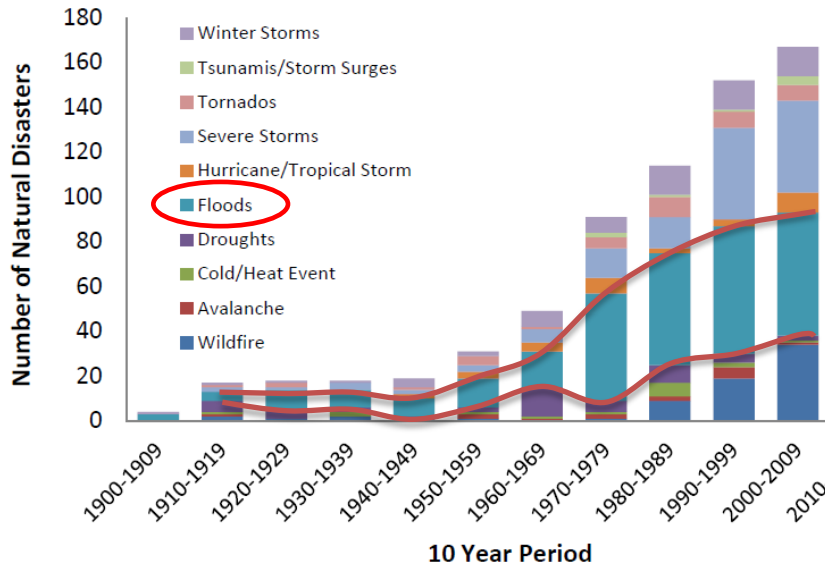
An Application to Humber River Watershed in Southern Ontario

Shasha Han and Paulin Coulibaly

Outline

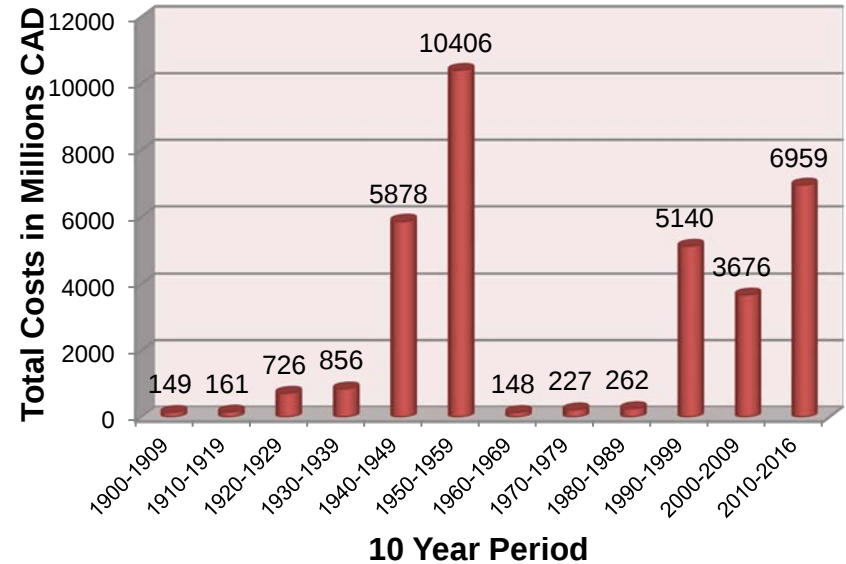
1. Introduction
2. What is BFS?
3. BFS method
4. Preliminary Application Results
5. Future Plans

Introduction



Frequency of Natural Disasters in Canada (1900-2011)

Source: Climate Change Adaptation Plan 2012



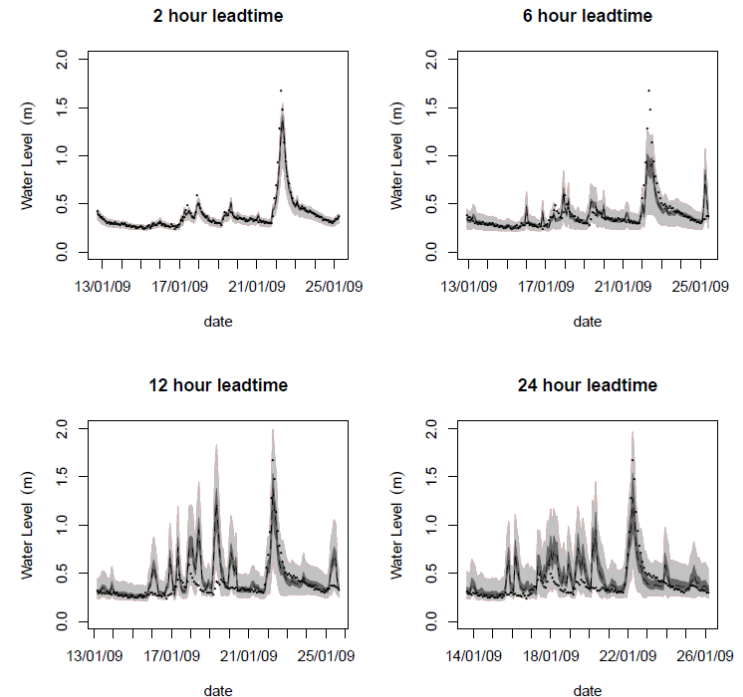
Estimated Total Costs of Large Floods in Canada

Data Source: Public Safety Canada

Introduction...

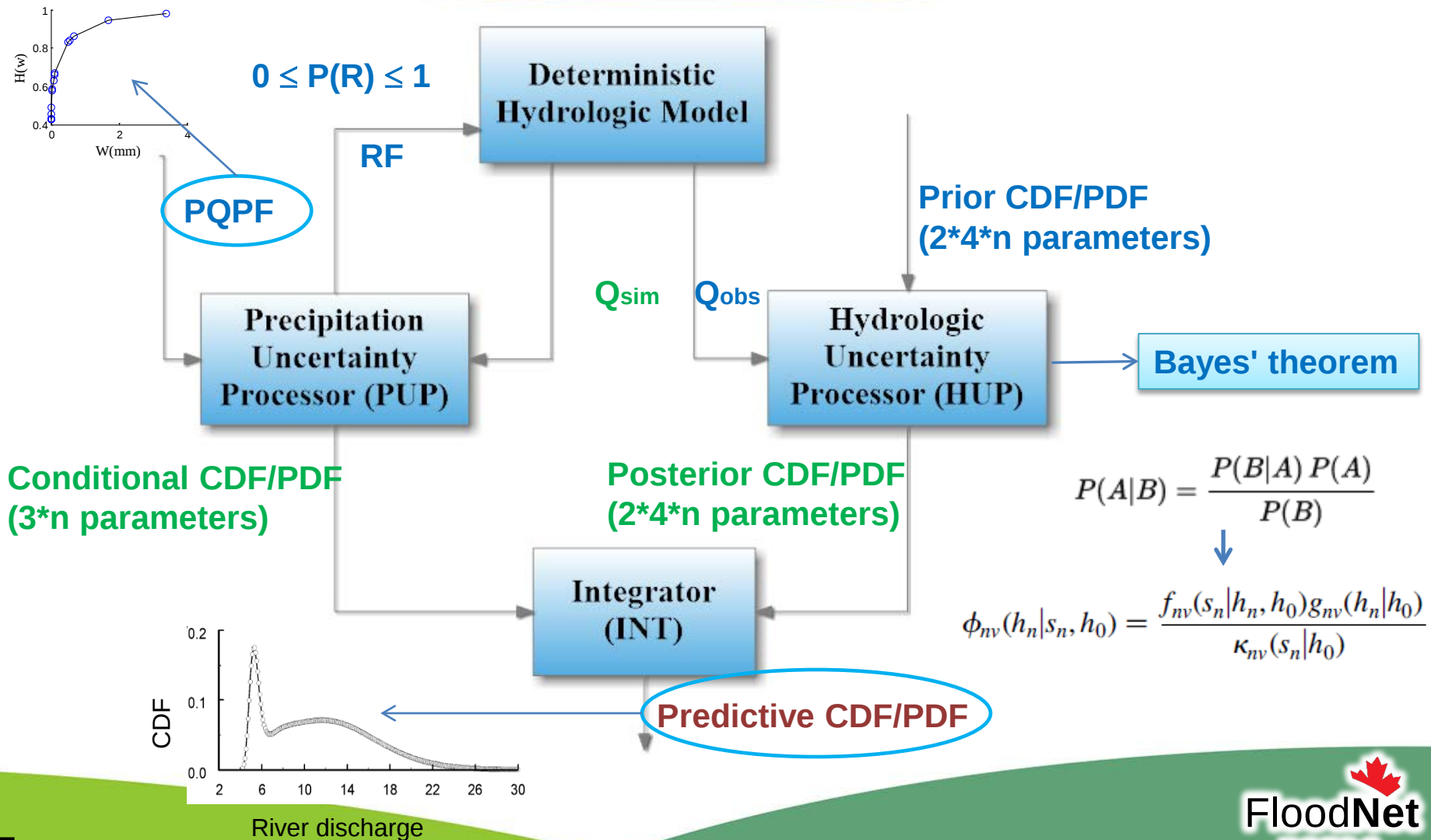
Deterministic forecast VS Probabilistic forecast ...

- Deterministic forecast: point forecast
- Probabilistic forecast: degree of confidence associated with point forecast
- This work presents a preliminary application of BFS to Humber river watershed to assess and reduce the flood forecast uncertainty.

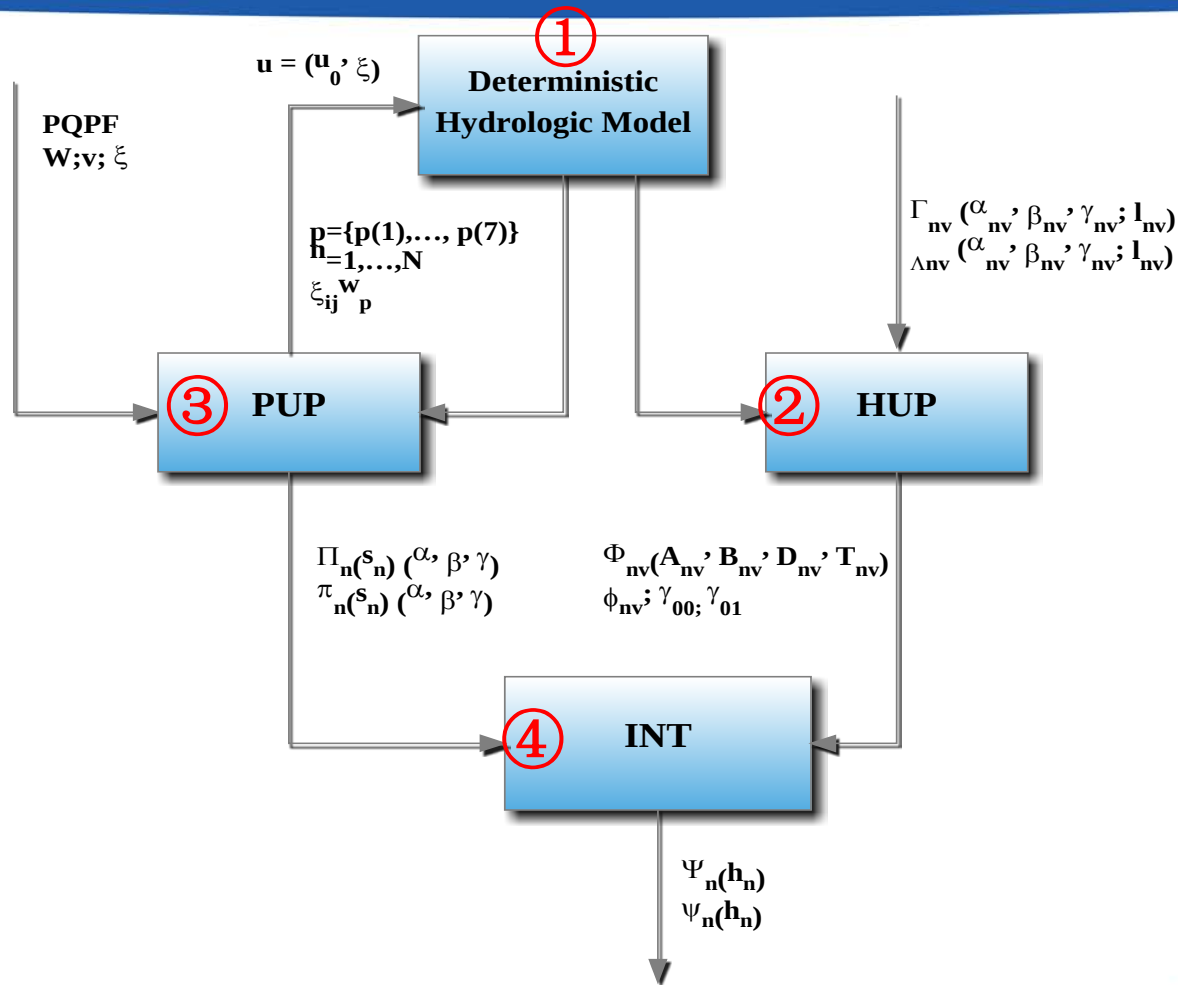


Comparison of observed & predicted water level (Weerts, A. H., et al, 2011)

What is BFS?



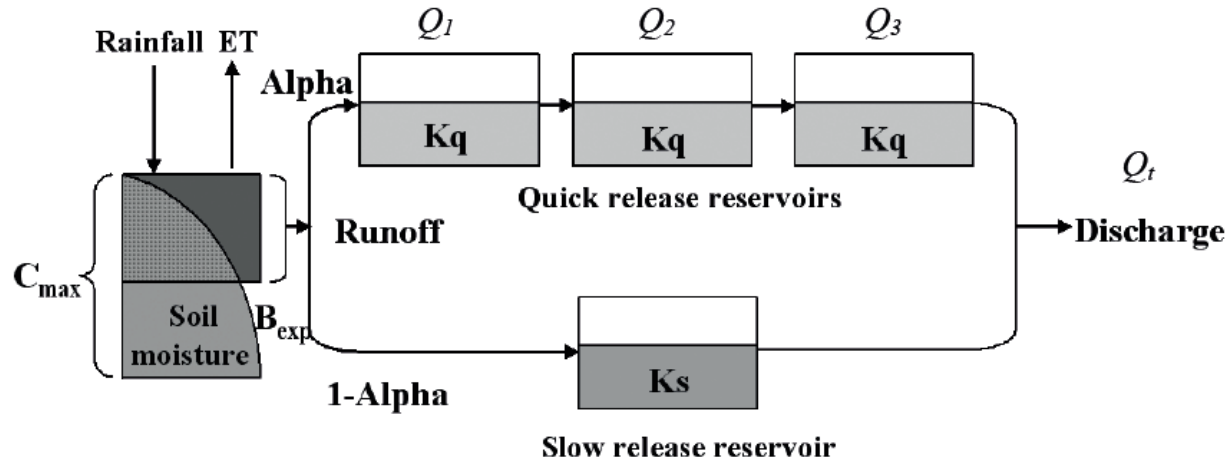
BFS Method Structure



Structure of Bayesian Forecasting System (BFS)

Deterministic Hydrologic Model

- ❖ Hydrologic model: HYMOD
- ❖ Parameter optimization : Monte Carlo method

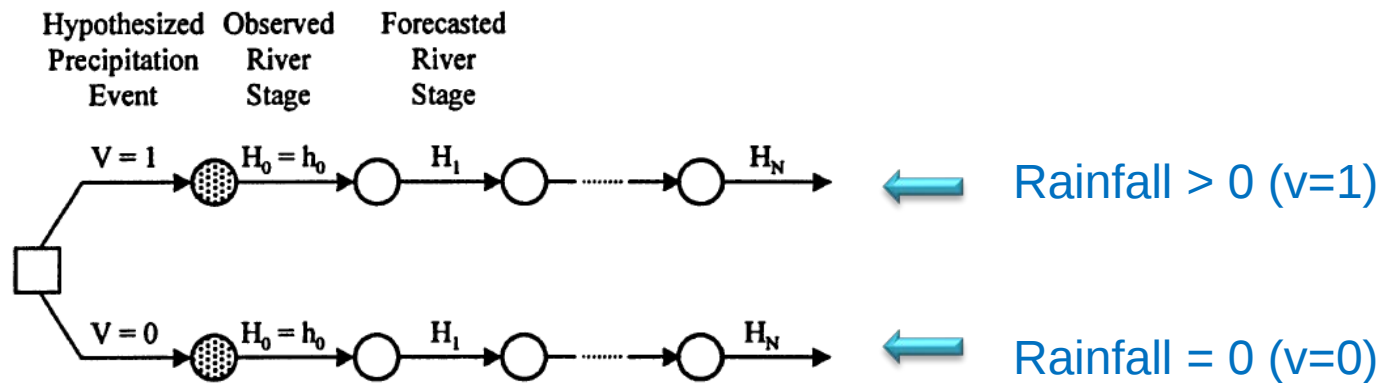
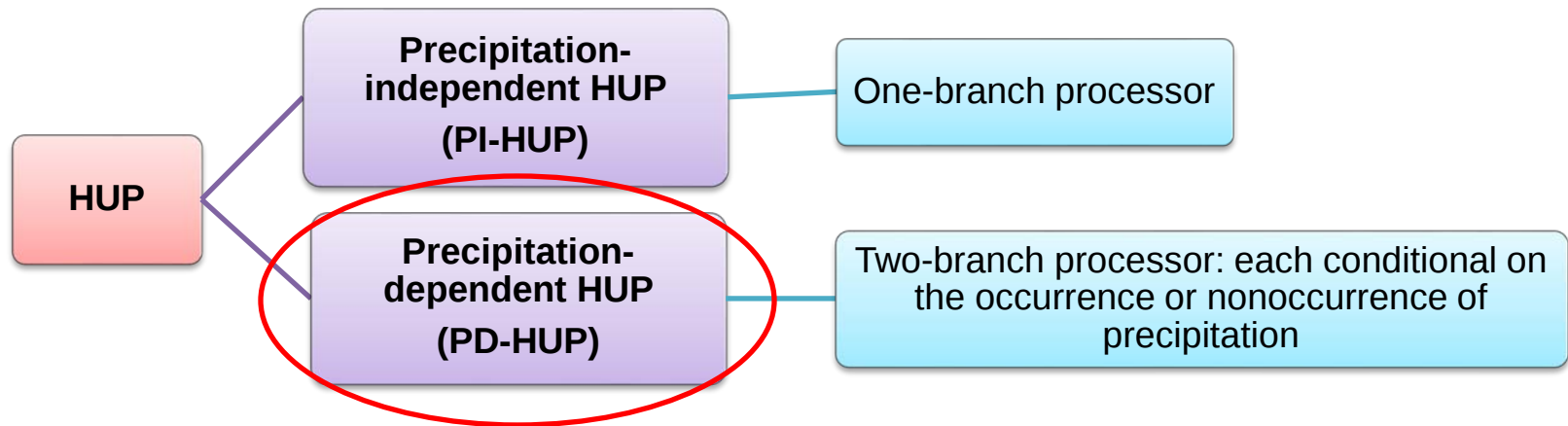


Schematic description of HYMOD (Quan et al., 2015)

Further improvement:

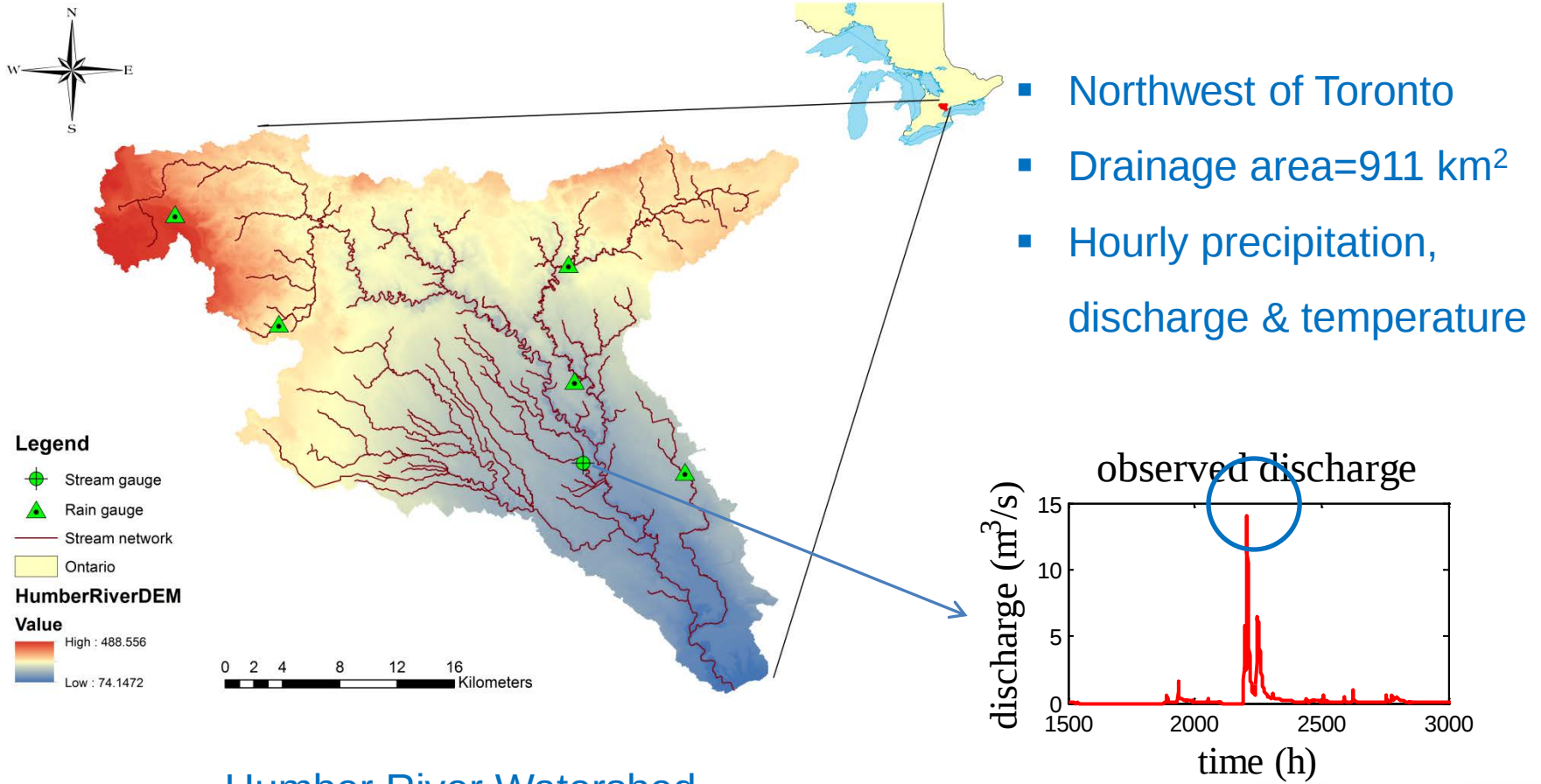
- ET: another ET calculation method instead of Hargreaves-Samani 1985
- Data: use most recent data

HUP (Hydrologic Uncertainty Processor)



PD-HUP (Krzysztofowicz & Herr, 2001)

Study Area and Data



Humber River Watershed

Data Pre-processing

Step 1: Define lead time n

$n = \text{lead time of precipitation} + \text{concentration time of basin}$

Step 2: Obtain Q_{obs} ($h_0, h_1, \dots, h_n$) & Q_{sim} ($s_0, s_1, \dots, s_n$) at time $t_0, t_1, \dots, t_n$ (for both $v=0$ & $v=1$)

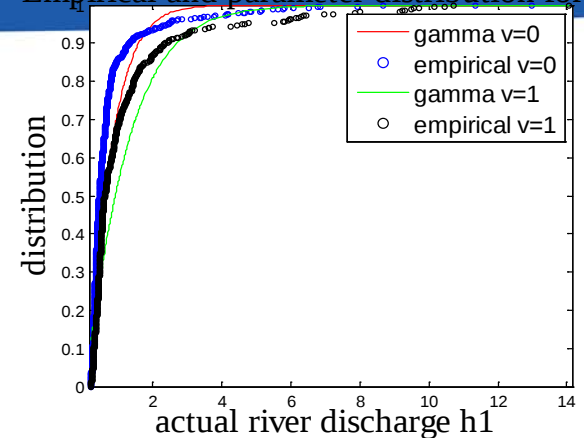
Step 3: Estimate the marginal distribution: Γ_{nv} for h_n , Λ_{nv} for s_n

Step 4: Normal quantile transform

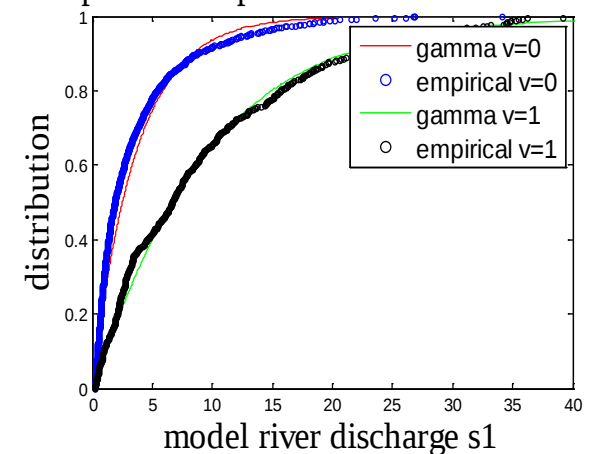
$$W_n = Q^{-1}(\Gamma_{nv}(H_n)), \quad n = 0, 1, \dots, N,$$

$$X_n = Q^{-1}(\bar{\Lambda}_{nv}(S_n)), \quad n = 1, \dots, N.$$

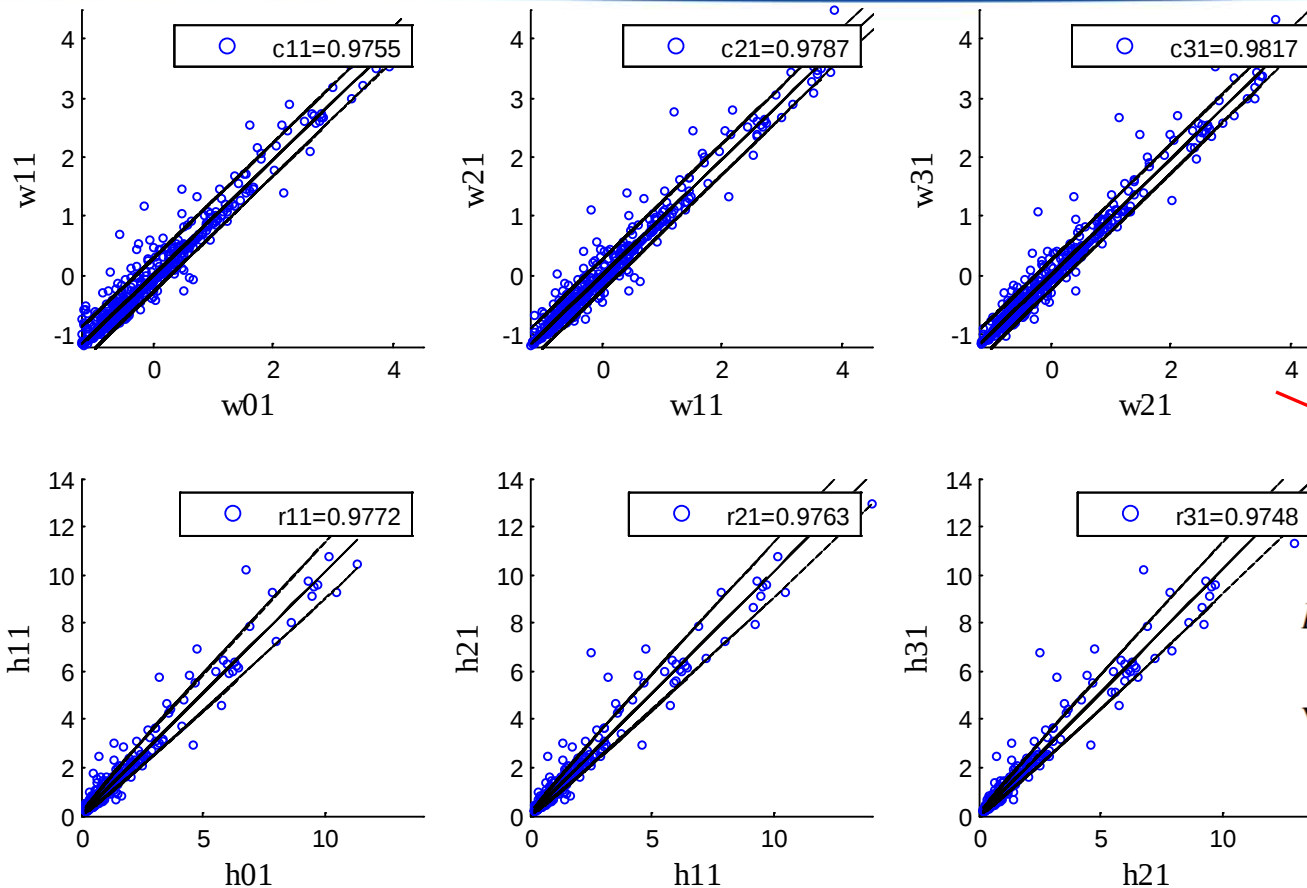
Empirical and parameter distribution for h_1



Empirical and parameter distribution for s_1



Preliminary Regression Results



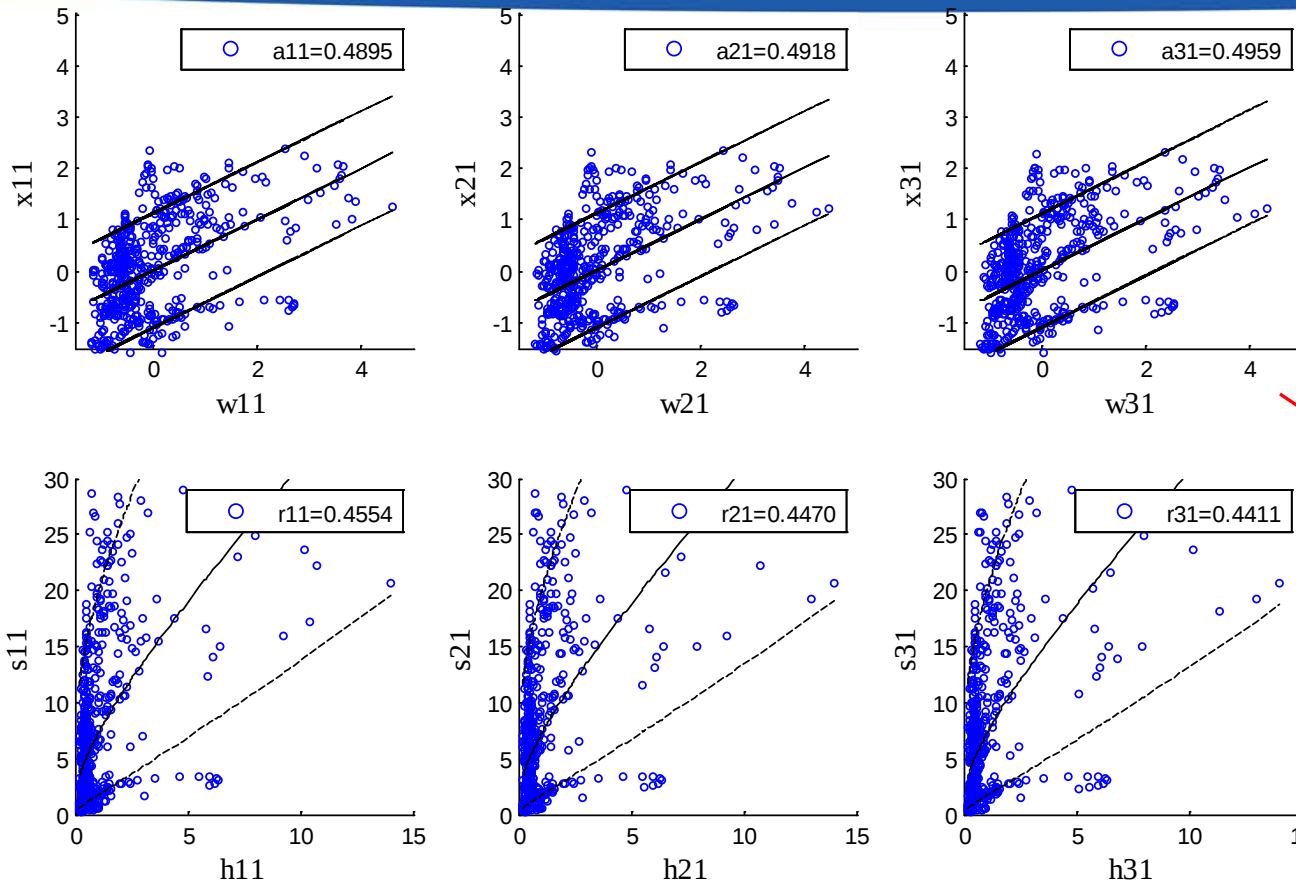
Dependence parameters of the transition densities:

$$E(W_n | W_{n-1} = w_{n-1}, V = v) = c_{nv} w_{n-1},$$

$$\text{Var}(W_n | W_{n-1} = w_{n-1}, V = v) = 1 - c_{nv}^2.$$

Dependence structure of transition densities (linear regression and 80% central credible interval. Upper: in transformed space; Lower: in original space)

Preliminary Regression Results



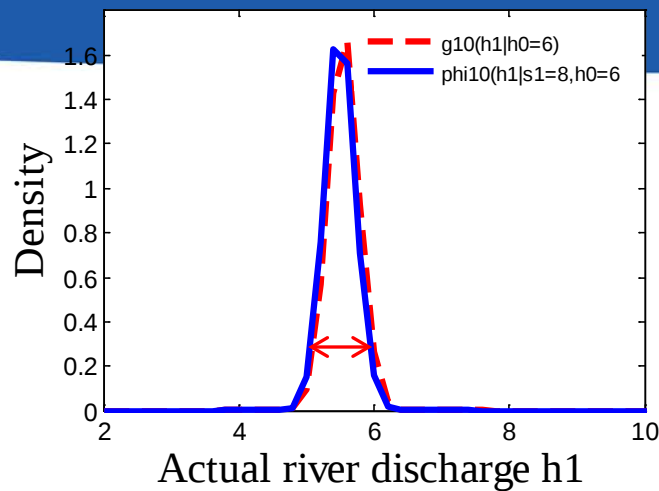
Dependence parameters of the likelihood function:

$$E(X_n | W_n = w_n, W_0 = w_0, V = v) = a_{nv}w_n + d_{nv}w_0 + b_{nv},$$

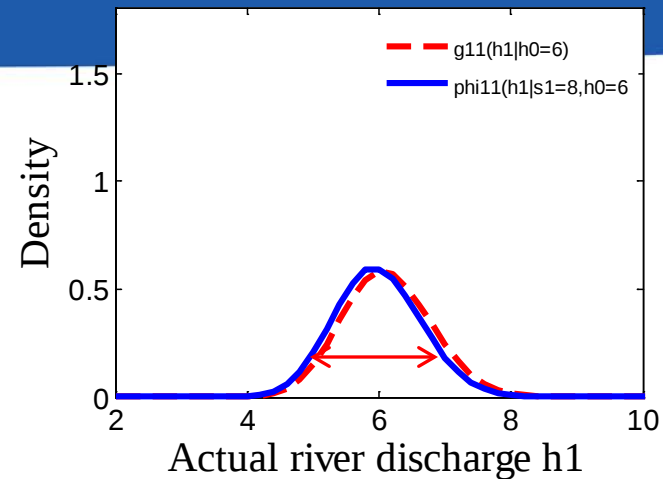
$$\text{Var}(X_n | W_n = w_n, W_0 = w_0, V = v) = \sigma_{nv}^2$$

Dependence structure of likelihood function (linear regression and 80% central credible interval. Upper: in transformed space; Lower: in original space)

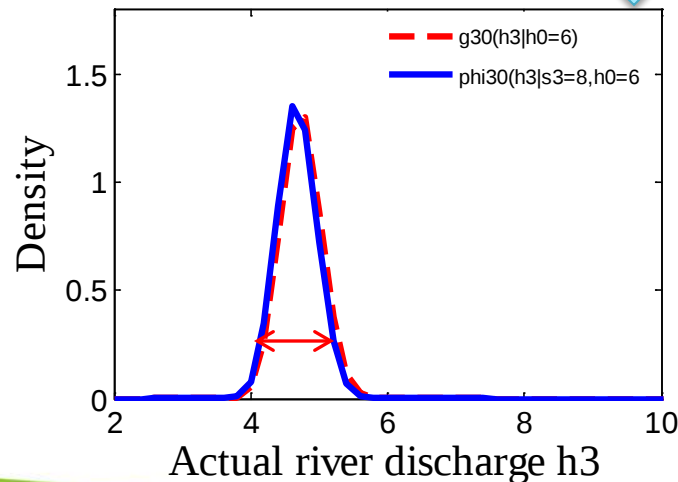
HUP Result: Prior density vs Posterior Density



Lead time $n=1$ ($v=0$)



Lead time $n=1$ ($v=1$)



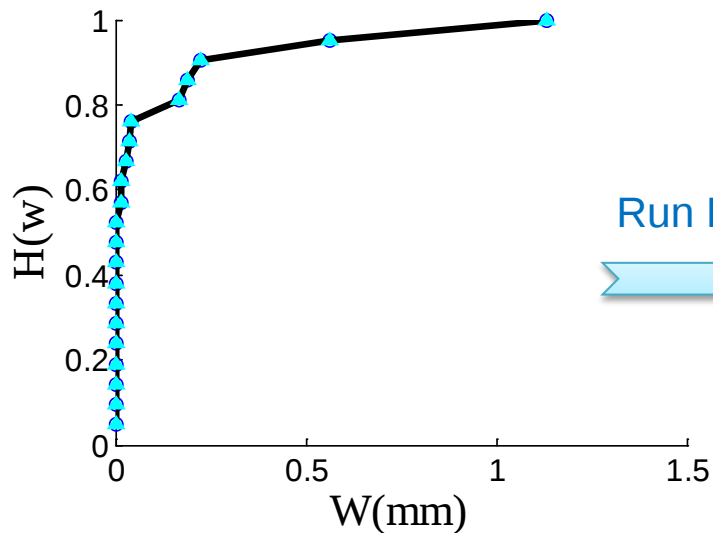
Lead time $n=3$ ($v=0$)

- Hydrologic uncertainty grows when precipitation occurs
- Hydrologic uncertainty increase with increasing lead time

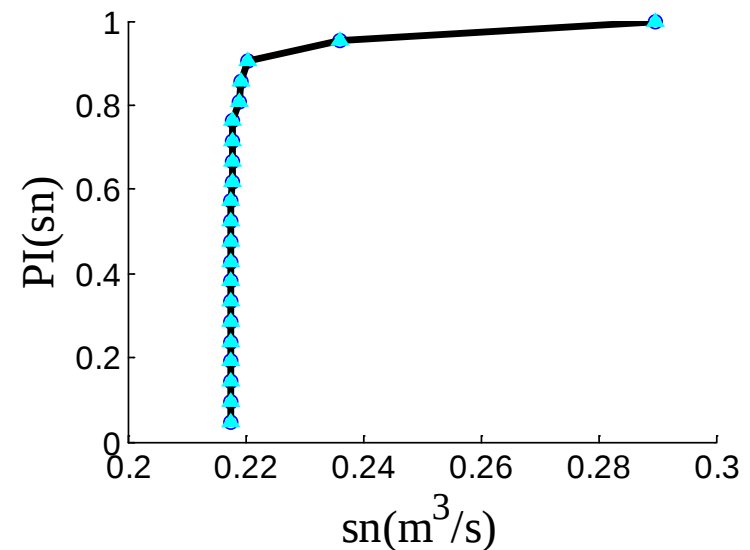
PUP (Precipitation Uncertainty Processor)

Input data: EC GEPS

- o $H(w)$: probability distribution of precipitation amount
- o v : probability of precipitation occurrence



Run HYMOD



Forecast distribution $H(w)$ of the basin average precipitation amount W at 0600 on June 4, 2015

Distribution $PI(sn)$ of model river discharge sn for W on June 4, 2015

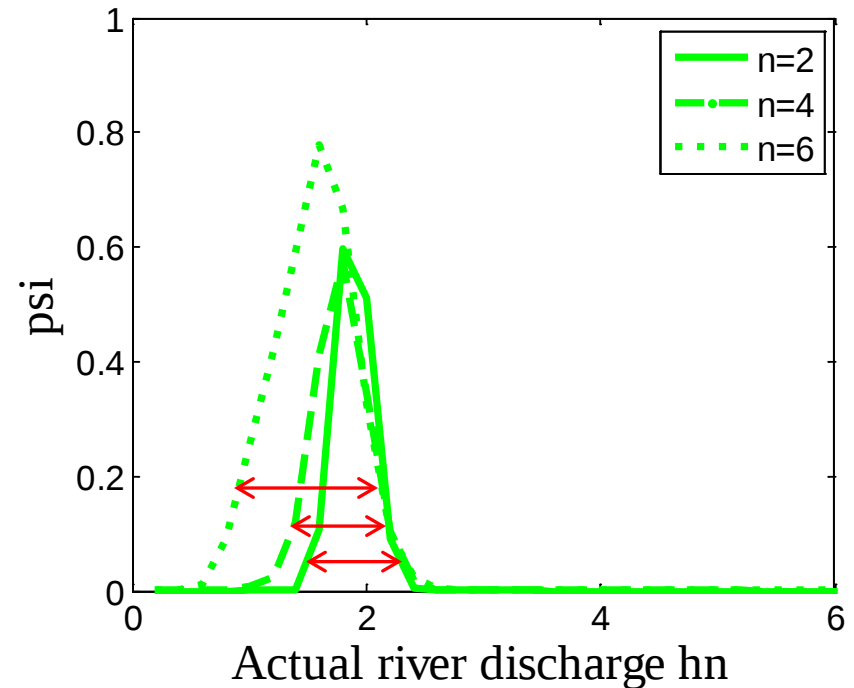
INT (Integrator)

Predictive distribution

$$\Psi_n(h_n) = \frac{\gamma_{00}(h_0)(1-\nu)}{\gamma_0(h_0)} \Phi_{n0}(h_n|s_{n0}, h_0) + \frac{\gamma_{01}(h_0)\nu}{\gamma_0(h_0)} \int_{s_{n0}}^{\infty} \Phi_{n1}(h_n|s_n, h_0) \pi_{n1}(s_n) ds_n$$

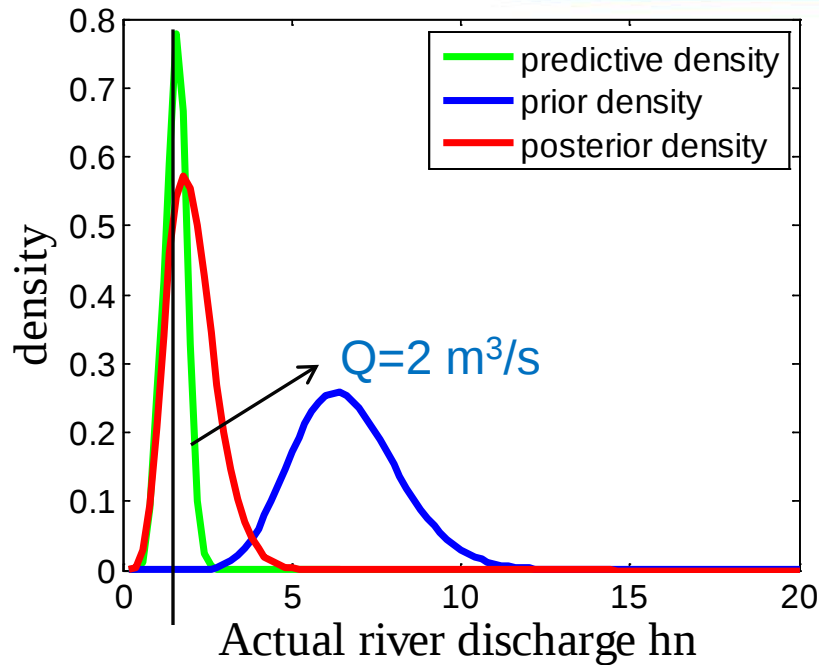
 From PUP From HUP

- Total uncertainty grows with increasing lead time

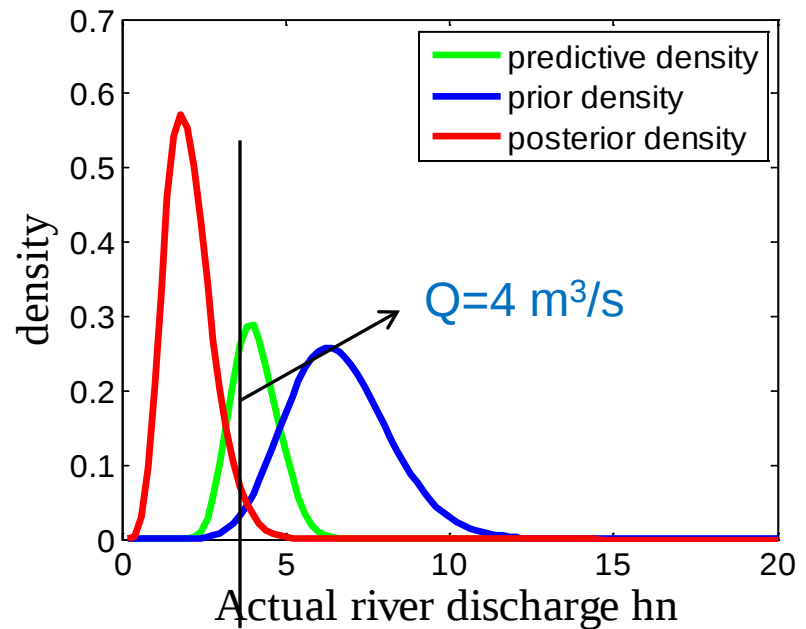


Comparison of predictive density at different lead time (2, 4, 6 hr)

INT Preliminary Results



Comparison of prior density, posterior density & predictive density conditional on $h_0=6$ & $sn=2$



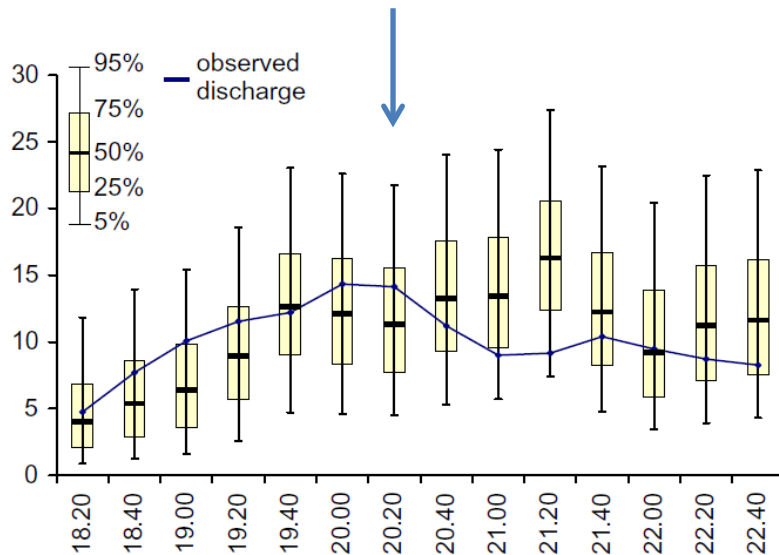
Comparison of prior density, posterior density & predictive density conditional on $h_0=6$ & $sn=4$

- The predictive density can reduce uncertainty and give a more accurate estimation compared with prior density

Future plans

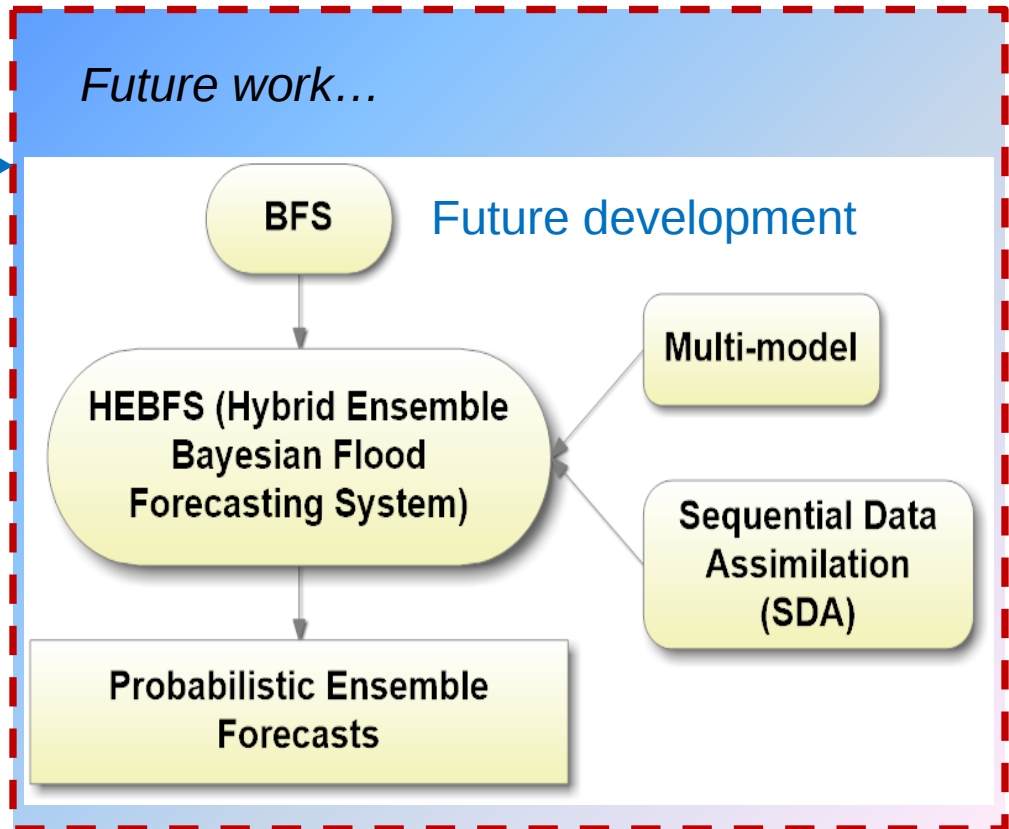
Present work...

- Optimize HYMOD and BFS method



BFS uncertainty bounds (Biondi, D., & De Luca, D. L., 2012)

Future work...



THANK YOU!