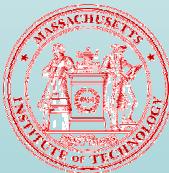


# Comparison of IDF Estimation Methods



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# IDF Estimation Procedure

$i(d, T)$  = average intensity in  $d$  with return period  $T$  (yr)

- **From Annual Maxima**  $i(d, T) = \text{upper } 1/T \text{ quantile of } I_{\max}(d, T)$
- **From POT values**  $T(i, d) = \frac{1}{\lambda(i^*) \cdot P[POT(d, i^*) > i - i^*]}, i^* = \text{threshold}$
- **From Marginal Distribution**  $\begin{cases} F_{I_{\max}(d)}(i) = [F_I(d)(i)]^{1/d} \\ i(d, T) = \text{see annual max} \end{cases}$
- **From  $I(t)$  process**
- **Hybrid** – *Combine marginal and annual -max information*

# Other IDF Estimation choices

## → Parameterization of $i(d, T)$

- Dependence on  $T$ ?  $\Rightarrow$  Quantiles of  $I_{max}$ , POT,  $I$
  - Dependence on  $d$ ? Smooth distribution parameters?  
Smooth quantiles?

## → Parameter estimation procedure

## MoM , PWM , ML , tail fitting .... ?

# Evaluation Criteria

- *Bias, variance, RMS error*
- *Sensitivity to outliers*
- *Does the optimal method depend on  $d$ ,  $T$  or record length  $D$  ?*

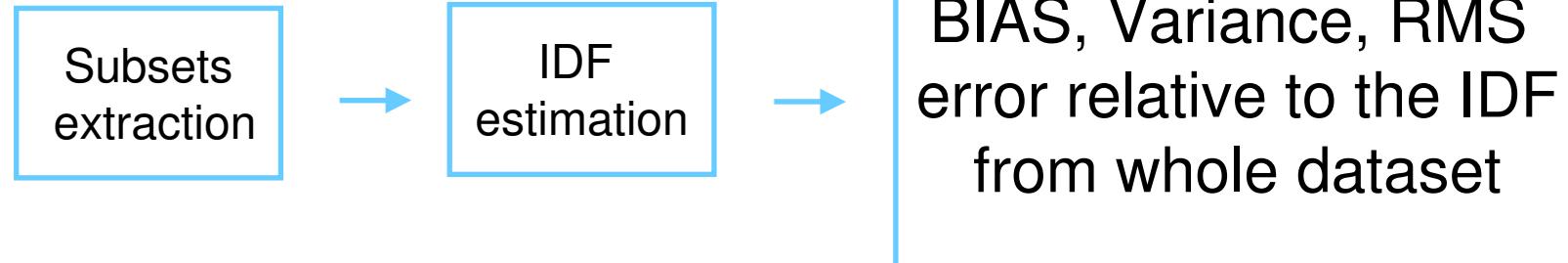
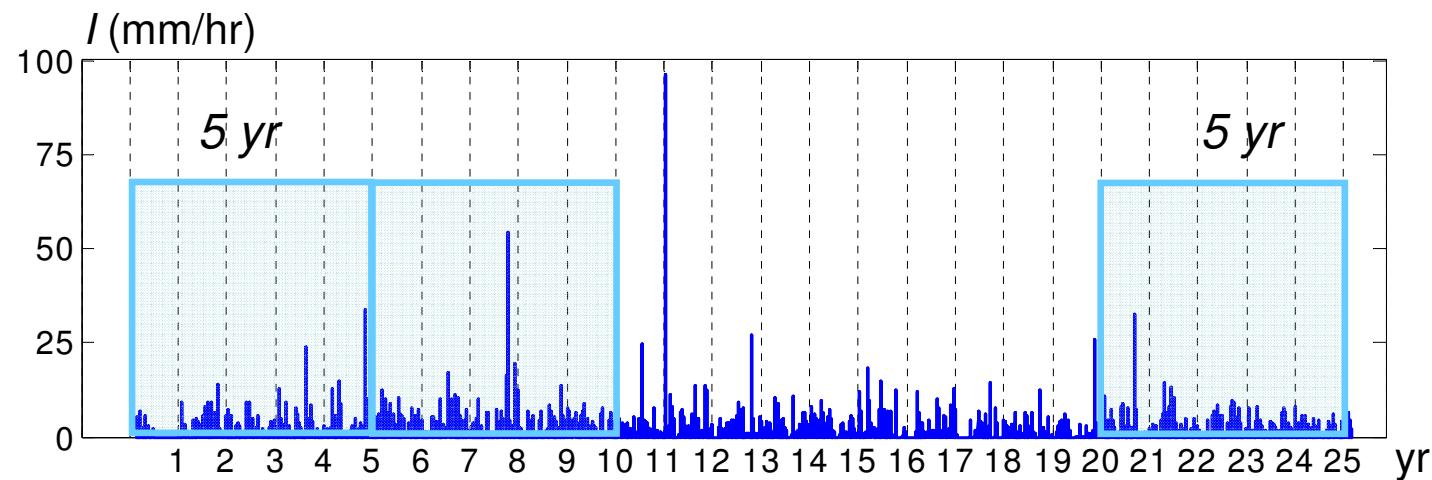
# Methods Compared

Method	Parameters	Estimation Procedure
$I_{max}$ (AM)	<ul style="list-style-type: none"> <li>For each <math>d</math>: GEV parameters</li> <li>For each <math>T</math>: smooth IDF values in <math>d</math> (pars. <math>\theta, \eta</math>)</li> </ul>	PWM for each $d$ LS for each $T$
	<ul style="list-style-type: none"> <li>Dependence on <math>d</math>: pars. <math>\theta, \eta</math></li> <li>Dependence on <math>T</math>: pars. GEV</li> </ul>	Koutzoyiannis et al. 1998
POT	<ul style="list-style-type: none"> <li>For each <math>d</math>: threshold <math>i^*</math>, 3 GP pars.</li> </ul>	PWM and LS
Marginal (MD)	<ul style="list-style-type: none"> <li>For each <math>d</math>: 3 pars. for LN tail</li> </ul>	MoM ( $1^\circ, 2^\circ, 3^\circ$ )
I(t) process MF	<ul style="list-style-type: none"> <li><math>K(q)</math>, 2 <math>\beta</math>-LN parameters, <math>D_{max}</math></li> </ul>	MoM ( $1^\circ, 2^\circ, 3^\circ$ )
Hybrid (marginal+ $I_{max}$ )	<ul style="list-style-type: none"> <li>Same as marginal + calibration to mean annual max</li> </ul>	

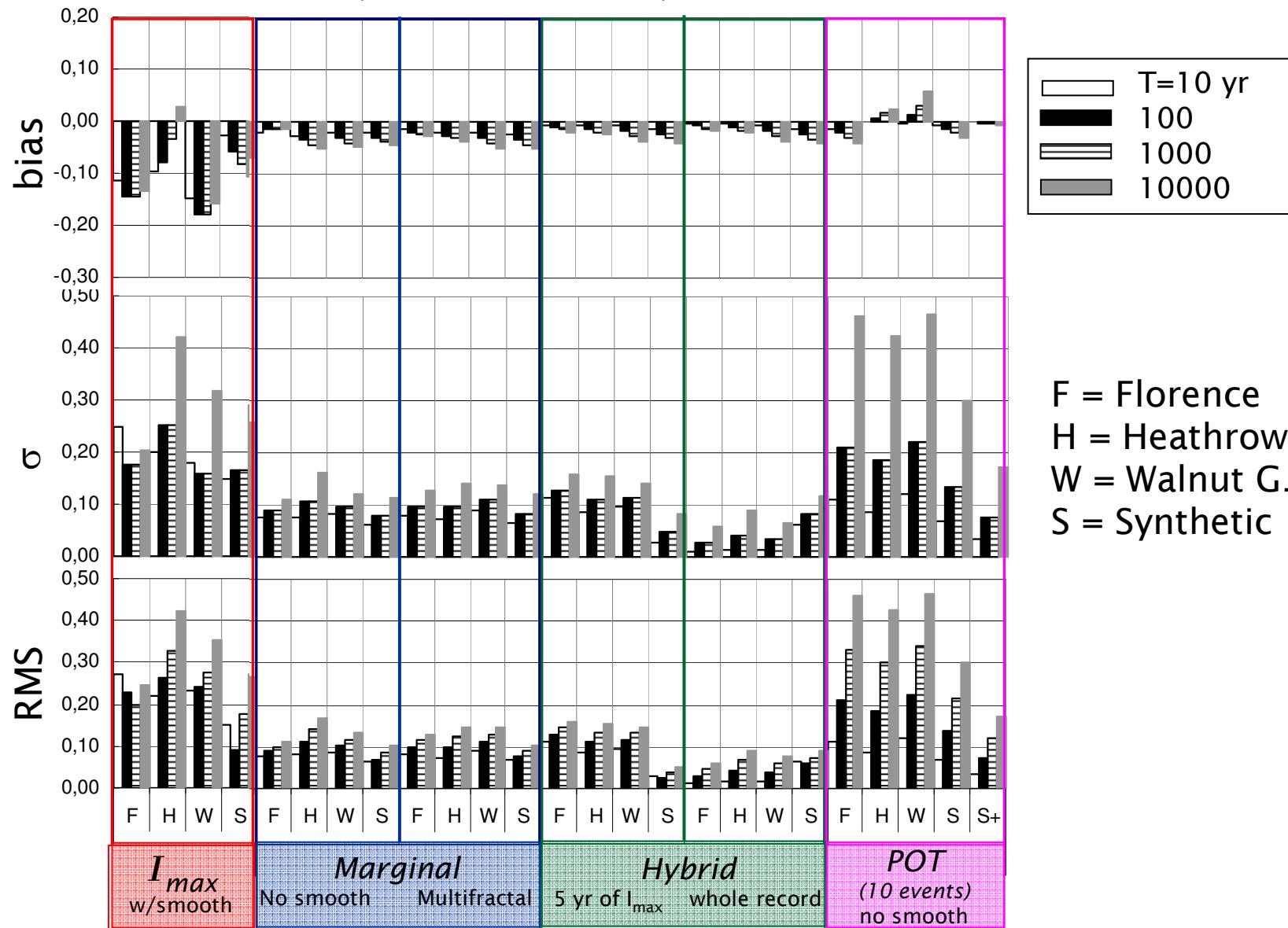
# Data sets used for evaluation

	n° years	mean Intensity mm/hr	rainy fraction %	mean $ I  > 0$ mm/hr	Comments
<b>Florence</b>	24	0.087	7.4	1.18	outliers: 1966, (long $d$ )
<b>Heathrow</b>	51	0.068	8.7	0.78	outliers: 1959, 1970, (short $d$ )
<b>Walnut Gulch</b>	49	0.035	2.3	1.53	-
<b>Synthetic</b>	1000	0.102	9.6	1.07	$\beta$ -LN cascade $C_\beta = 0.46$ , $C_{LN} = 0.06$ , $D_{max} = 9.56 \text{ days}$

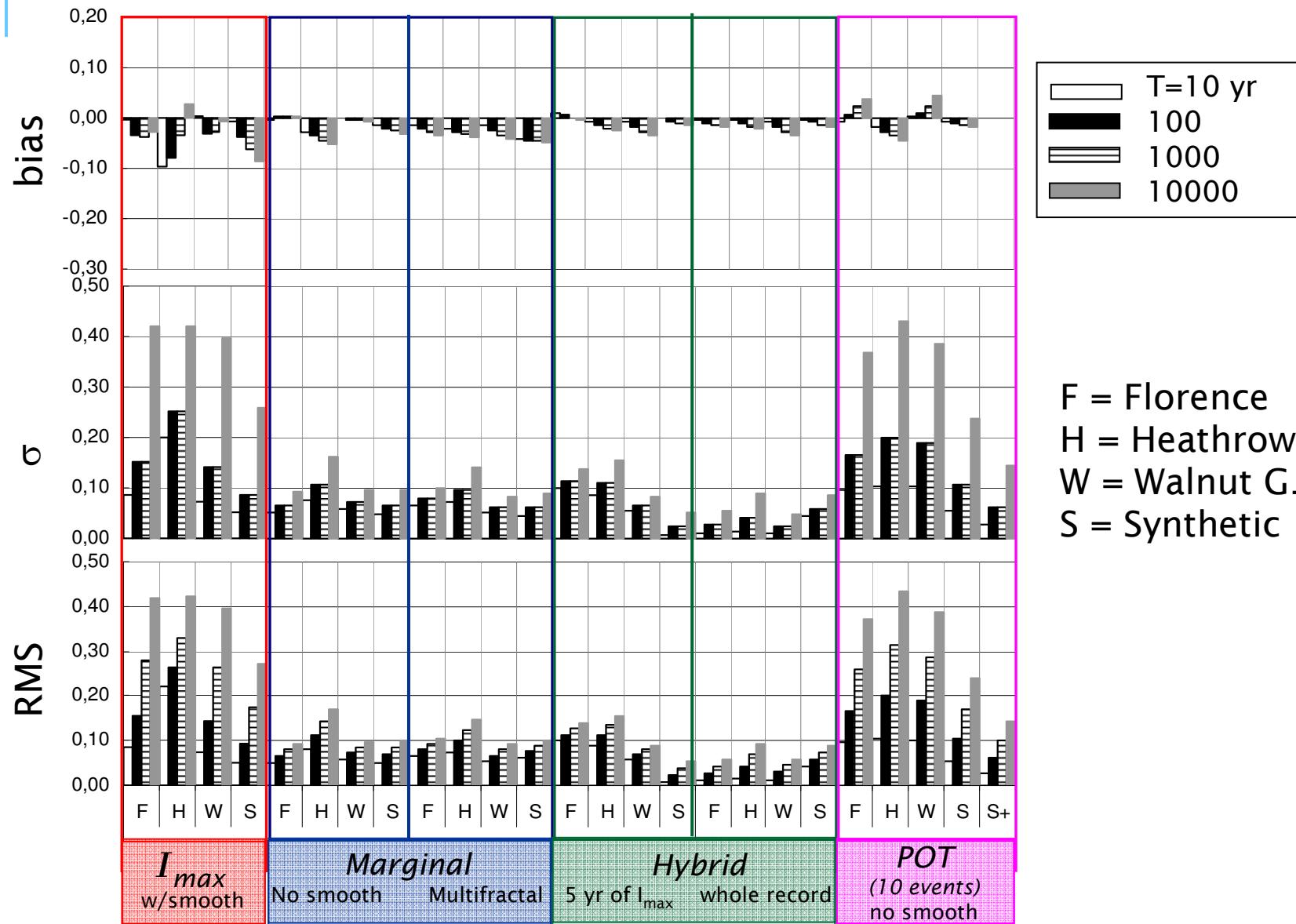
# Bias and Variance Comparison Using Subsets of the Record



# Bias, Variance, RMS error -1hr



# Bias, Variance, RMS error -24 hr



# Conclusions

- ***Bias, variance, RMS error, outliers sensitivity:***
  - Marginal or Hybrid < POT < Annual-max
- ***Does the optimal method depend on  $d$ ,  $T$  or record length  $D$ ?***
  - If  $I_{max}$  is available for many years → Hybrid is best;
  - Expected dependence on  $D$ :
    - Annual-max best for very long  $D$ ;
    - POT competitive for Intermediate  $D$  (a few decades);
    - Marginal best for short  $D$  (a few years).

These trends are qualitatively true, but Marginal is generally best for all  $D$

- Models that use shape pars. (GEV, GP) have large error variance (for small D, large T) benefit from smoothing, but smoothing affect bias;
- Bias/ $\sigma$  are similar for all methods when  $T \approx D$ ; increasing T enhances error variance, whereas bias is less variable.

# Thank you for your attention!

## *Acknowledgments*



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