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# Extreme Flood Frequency Analysis: Concepts, Philosophy and Strategies

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# Challenge

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Estimate frequency of rare floods  
(exceedance probabilities of 1 in 100,

or

1 in 10,000) using limited data provided  
by streamflow gauge at the site ...  
plus other sources of information.

# What Data Do We Use?

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Record at streamflow gauge at the site.

Historical data at the site

Paleoflood data at the site

Regional analysis of gaged data,  
historical and paleoflood information

# Bulletin 17B

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- Uniform flood frequency techniques used by US Federal agencies
- Bulletin not updated in 20+ years
  - despite significant amount of research
  - additional 30 years of data for skew map
  - better statistical procedures for censored data

# Bulletin 17C

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- Revision on the way
- Provides a tune up addressing
  - Use of historical information
  - Potentially Influential Low Floods (PILFs)
  - Censored & interval data
  - Confidence intervals & uncertainty analyses
- Also developing better regional skew

# Regional Skew Estimation

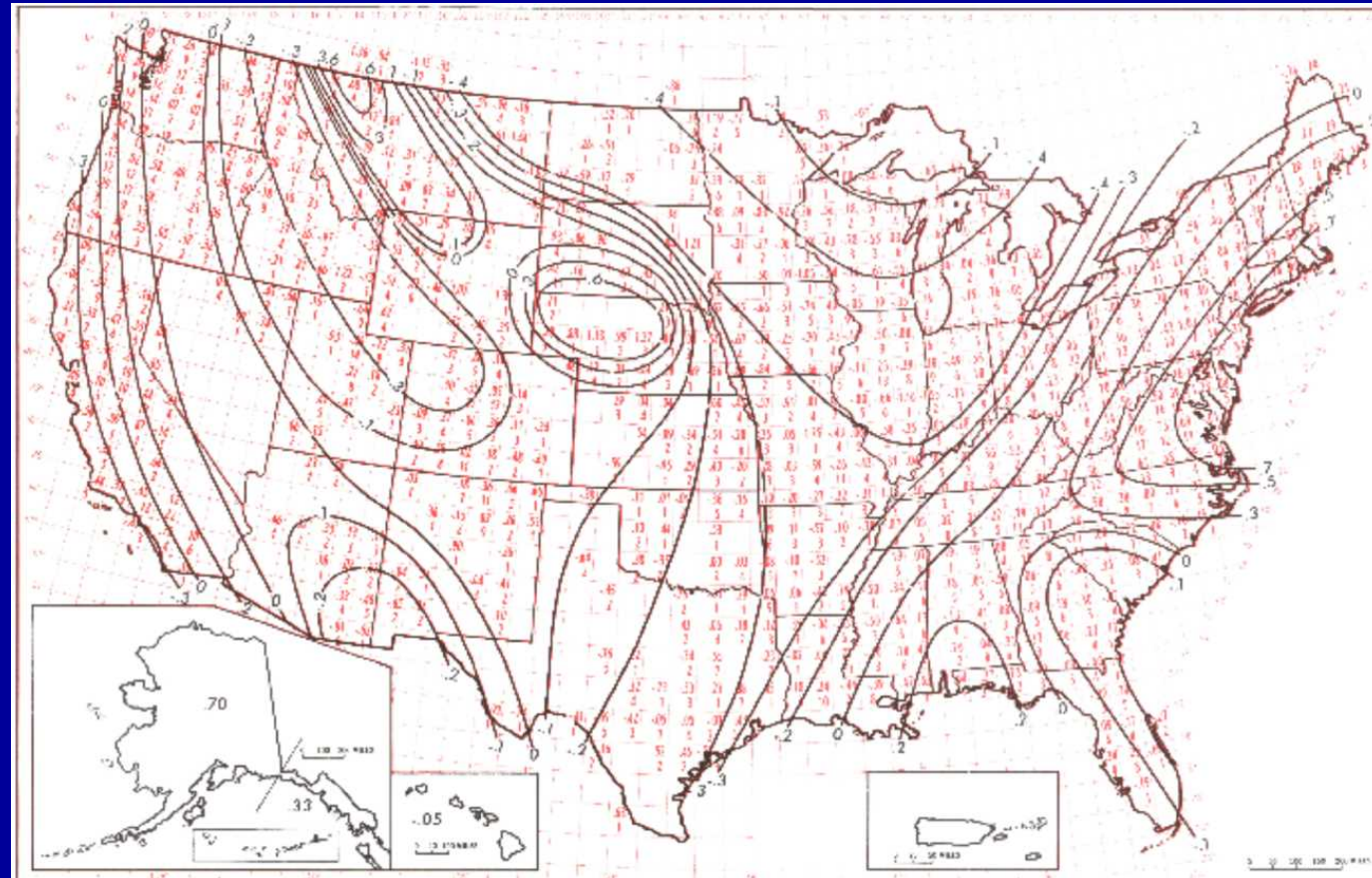
Regional skew  $G_g$   
from B17  
skew map

$$-0.5 < G_g < +0.7$$

$$\text{Map SE} = 0.55$$

$$\rightarrow \text{MSE}[G_g] = 0.302$$

Effective record  
length = 17 yrs

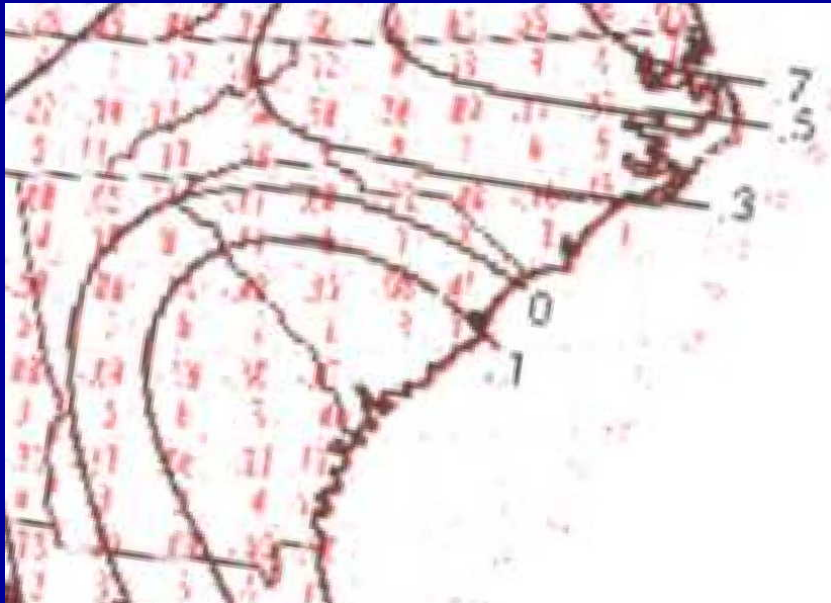


**GENERALIZED SKEW COEFFICIENTS OF LOGARITHMS OF ANNUAL MAXIMUM STREAMFLOW**  
**AVERAGE SKEW COEFFICIENT BY ONE DEGREE QUADRANGLES**

Lower number in each quadrangle is number of stream gaging stations for which the average shown above it was computed

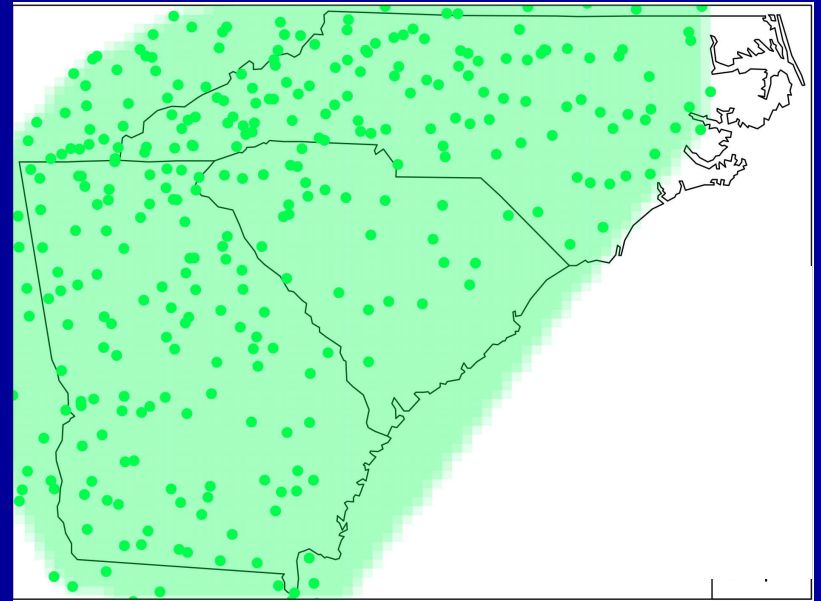
# Regional Skew Southeast

1976 (MSE = 0.302)  
2,972 sites nationally



Maps claims to be as  
good as 17 years

2009 (MSE = 0.14) revised 1/09  
342 *independent* watersheds



$\gamma = 0$  as good as 40  
years of record

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# Historical & Paleoflood data at a site:

Ways to extend the record.



# Sources of Historical Information

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## Long-term Residences' Memory

Written accounts, markers, pictures

## Geomorphologic evidence

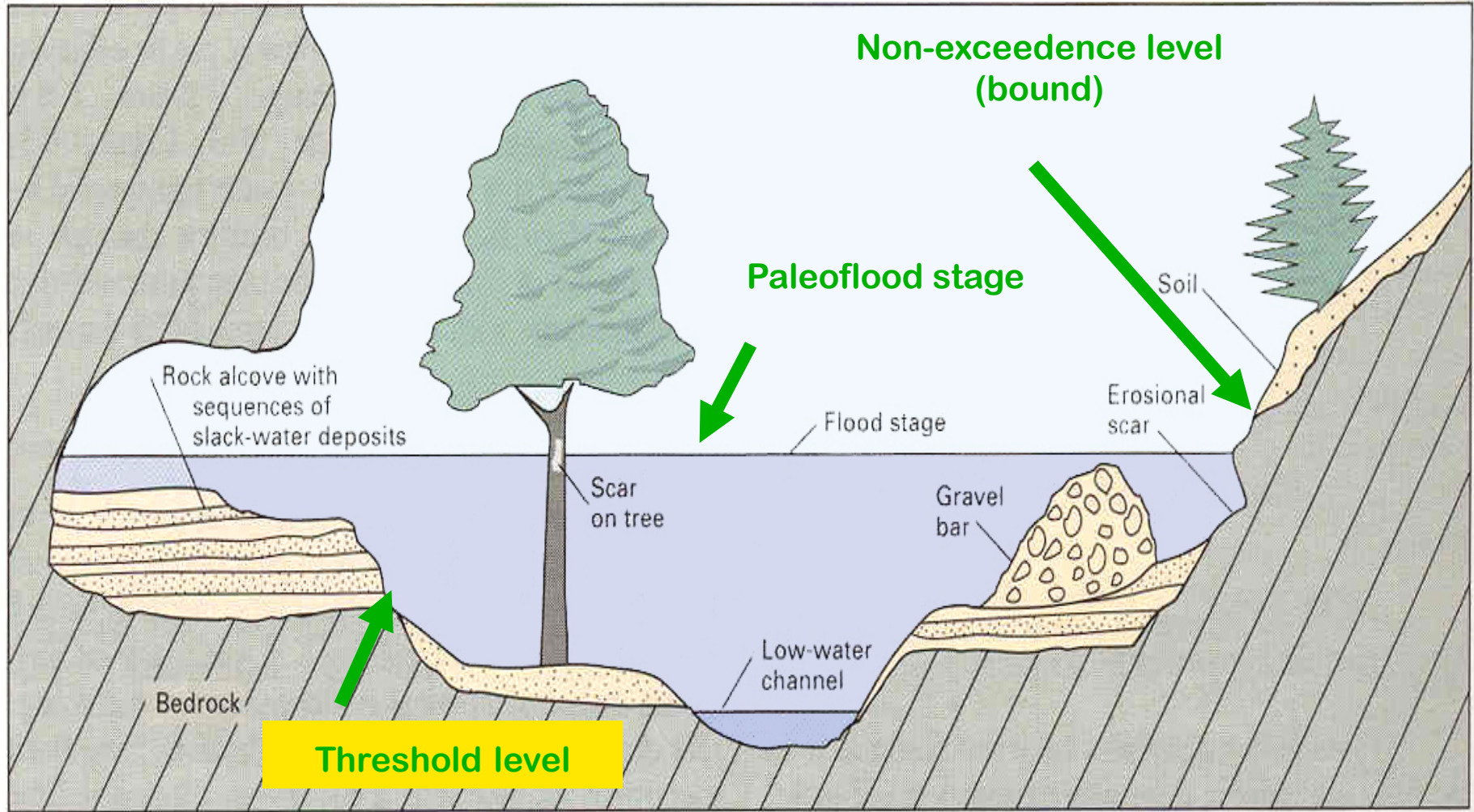
slack-water deposits

scour lines

high-water marks

undisturbed areas

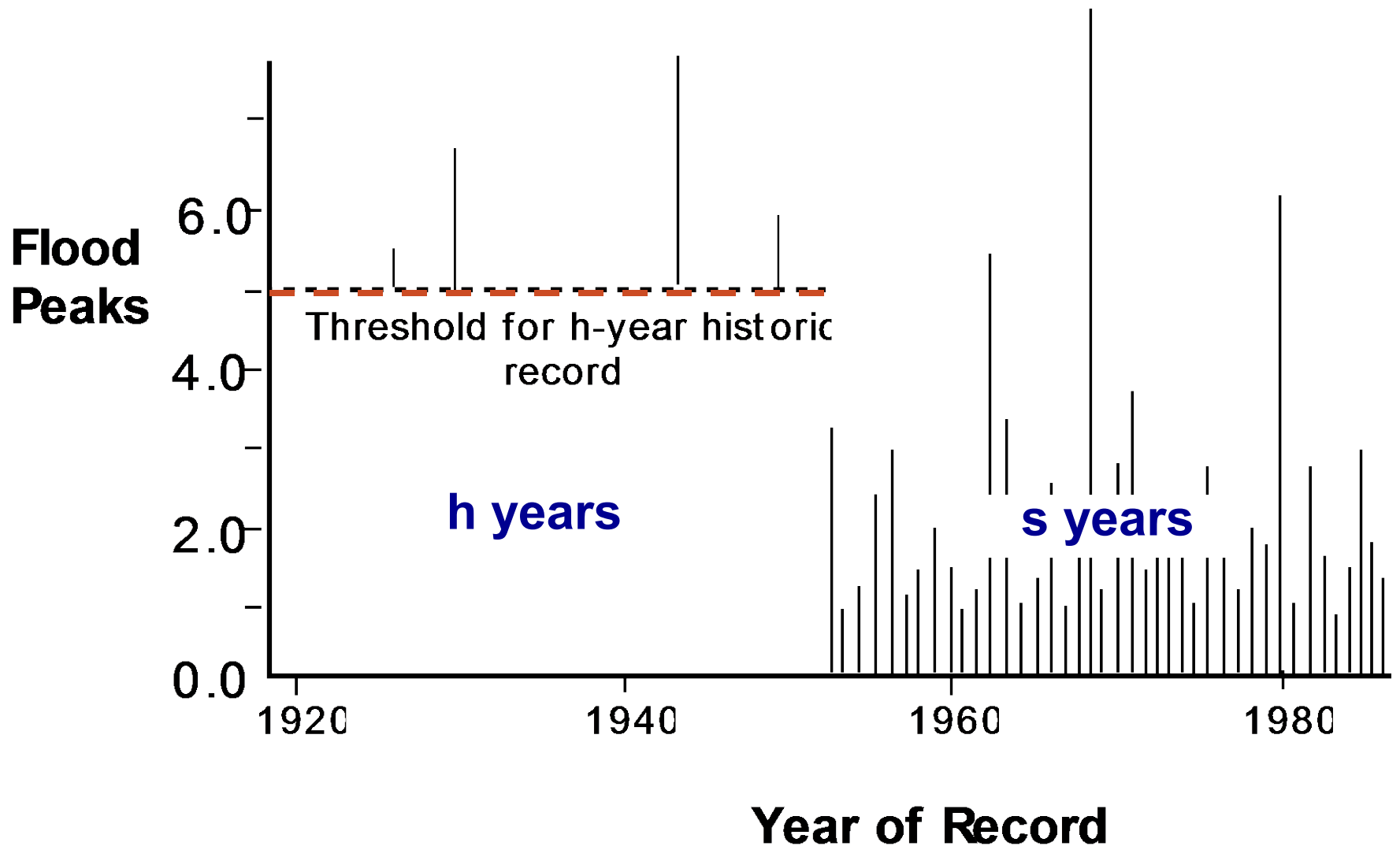
# Paleoflood Data



(Source: Jarrett 1991, modified from Baker 1987)

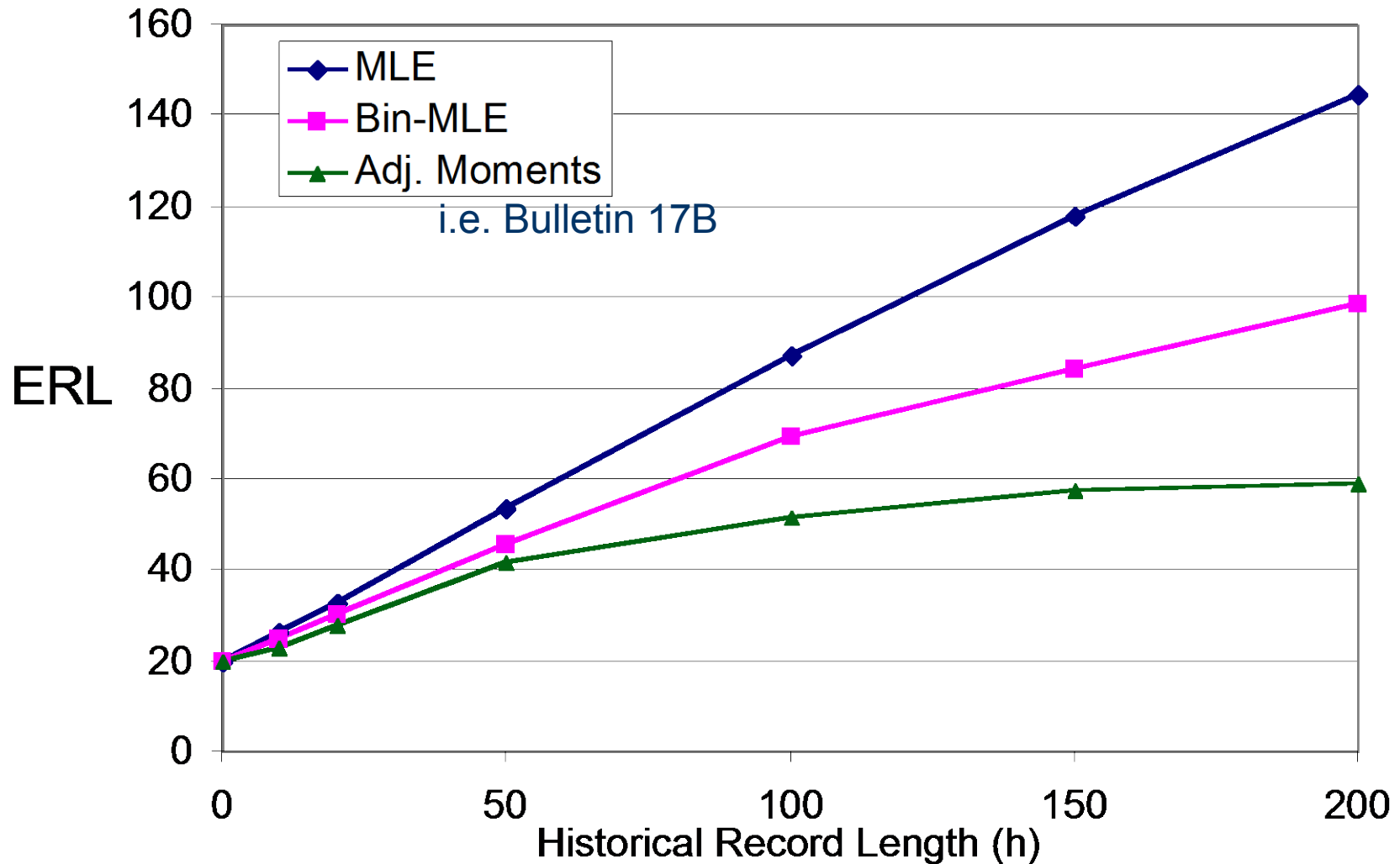
# Historical and Gauged Record

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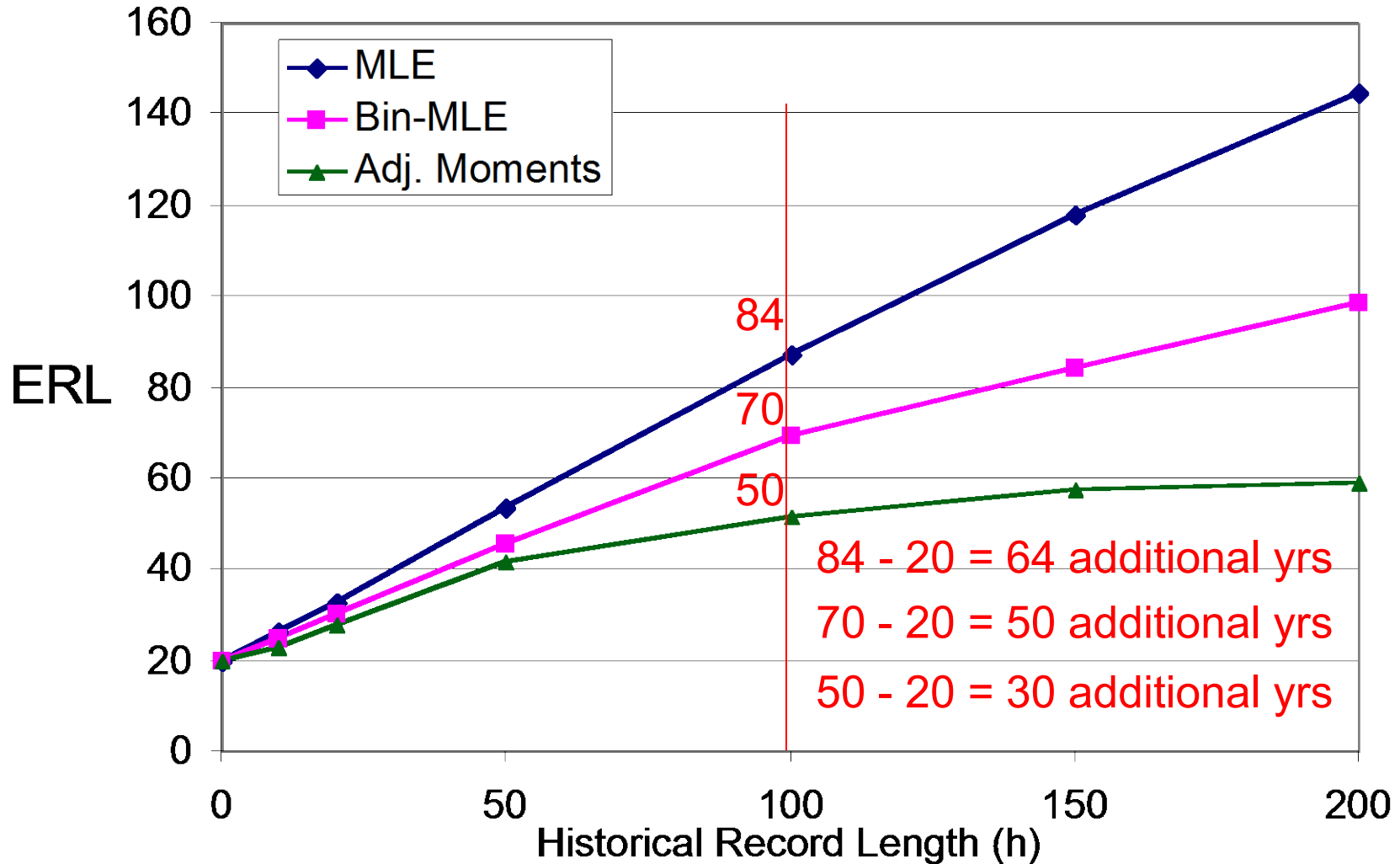
# Value of Historical Information

Effective Record Length,  $X_0 = 90\%$ -tile



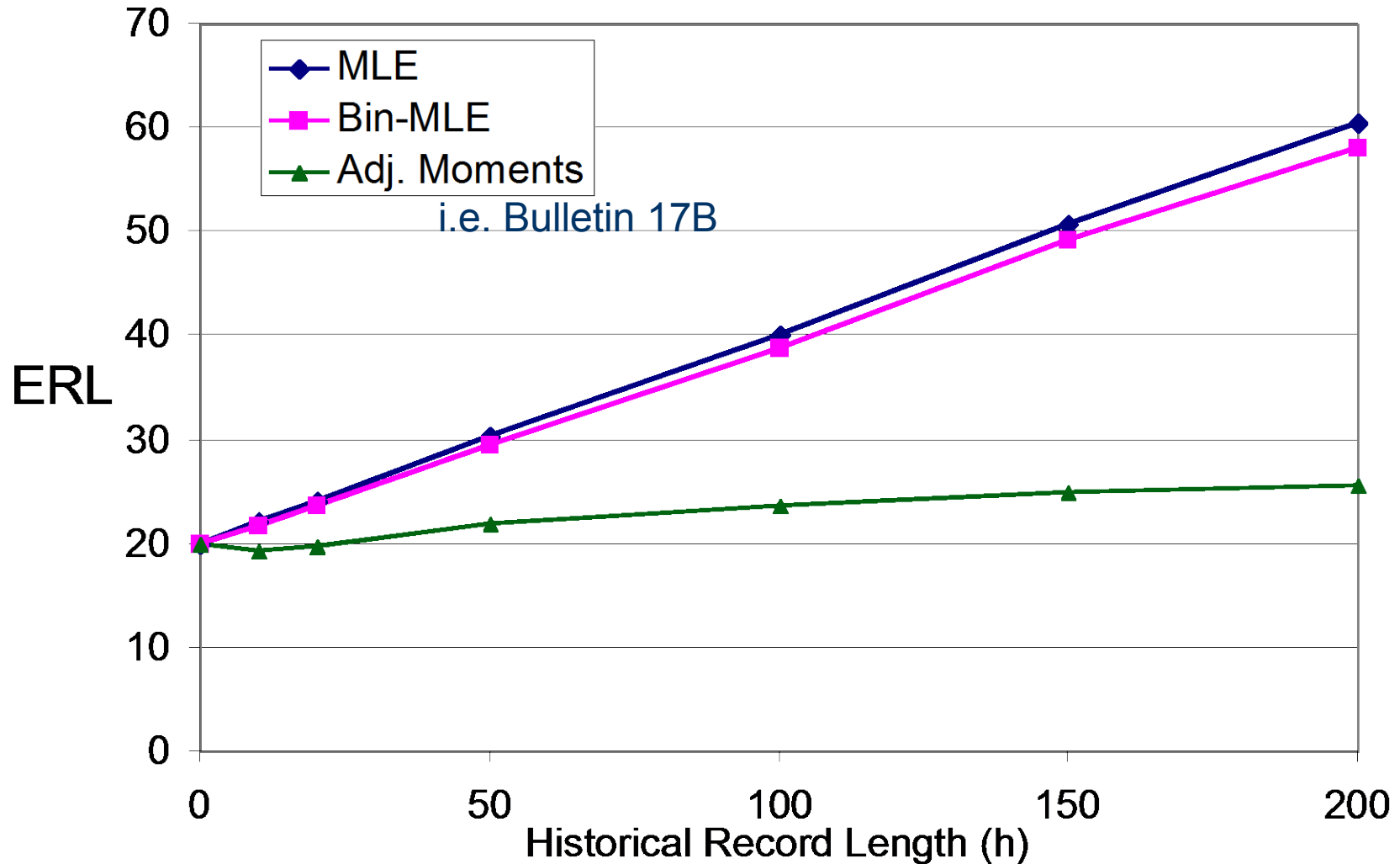
# Value of Historical Information

Effective Record Length,  $X_0 = 90\%$ -tile



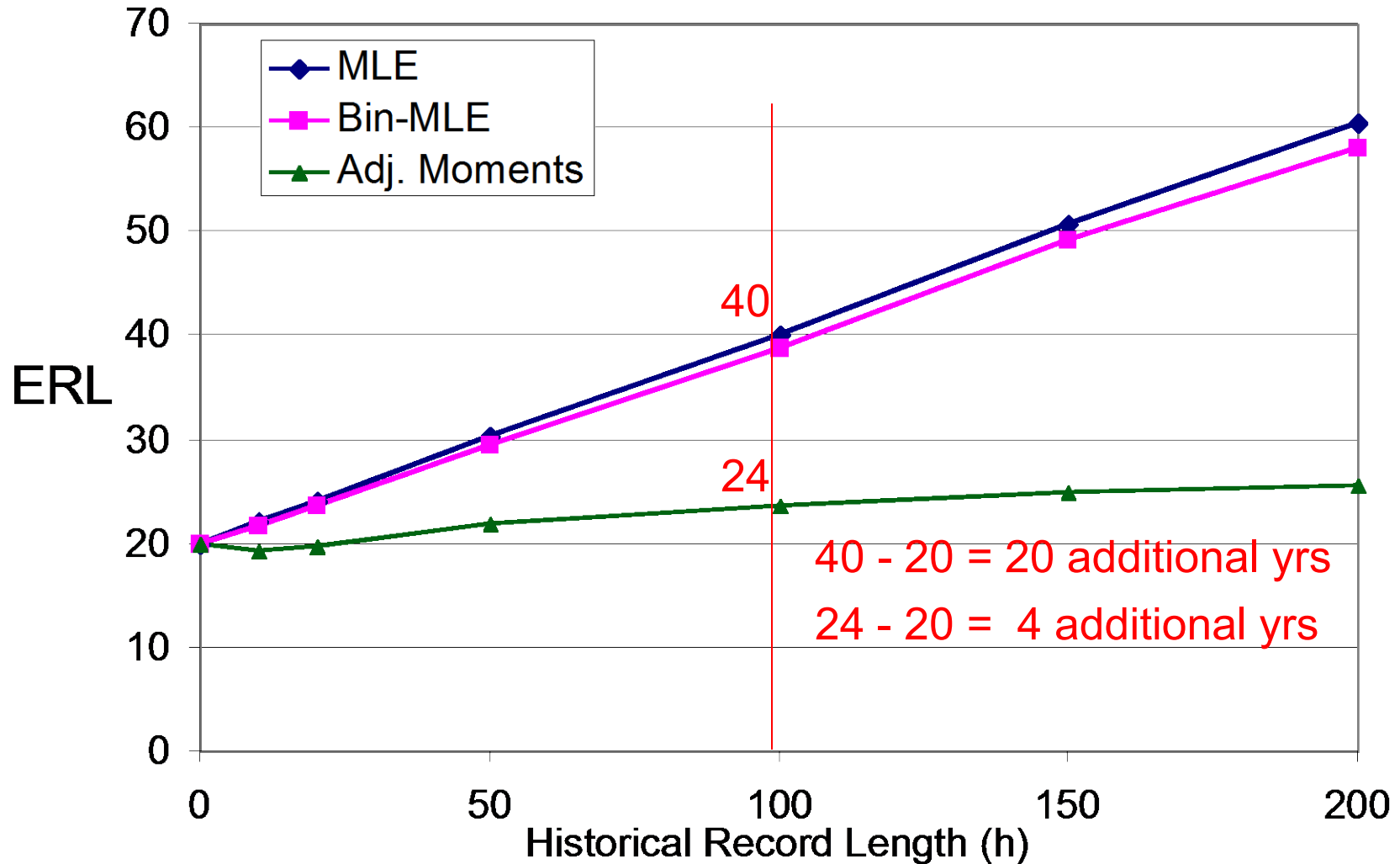
# Value of Historical Information

Effective Record Length,  $X_0 = 99\%$ -tile

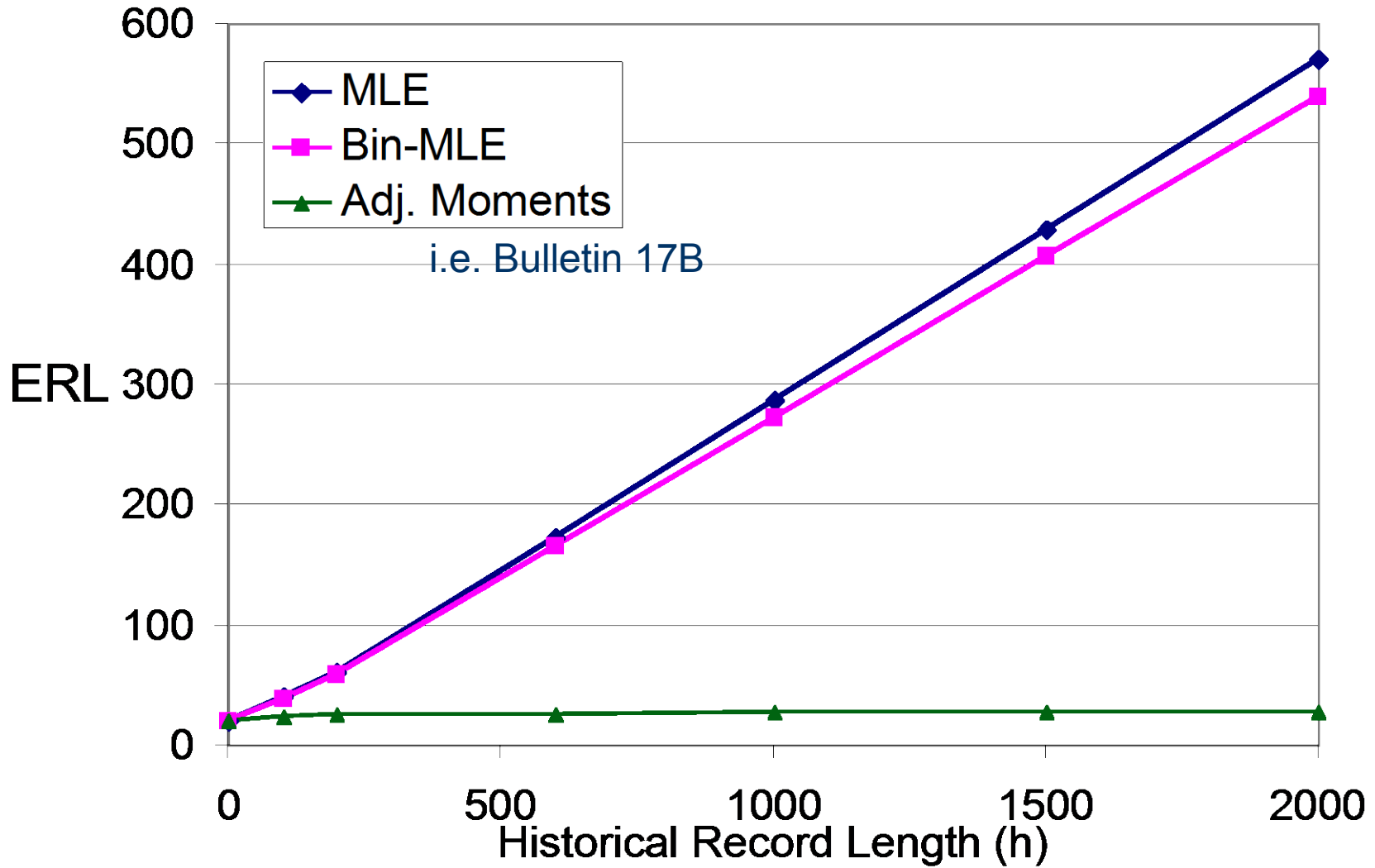


# Value of Historical Information

Effective Record Length,  $X_0 = 99\%$ -tile

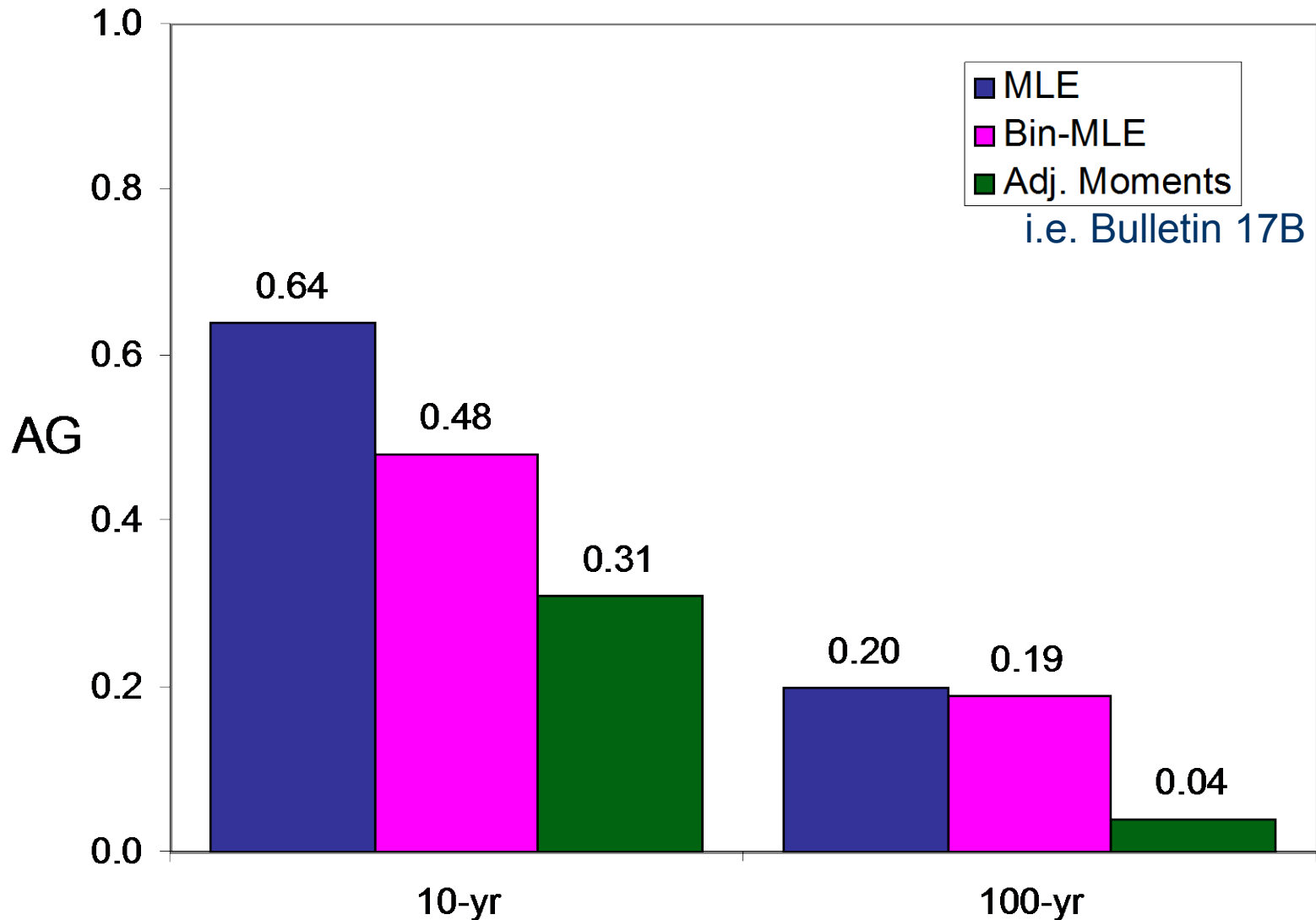


# Effective Record Length, $X_0 = 99\%$ -tile

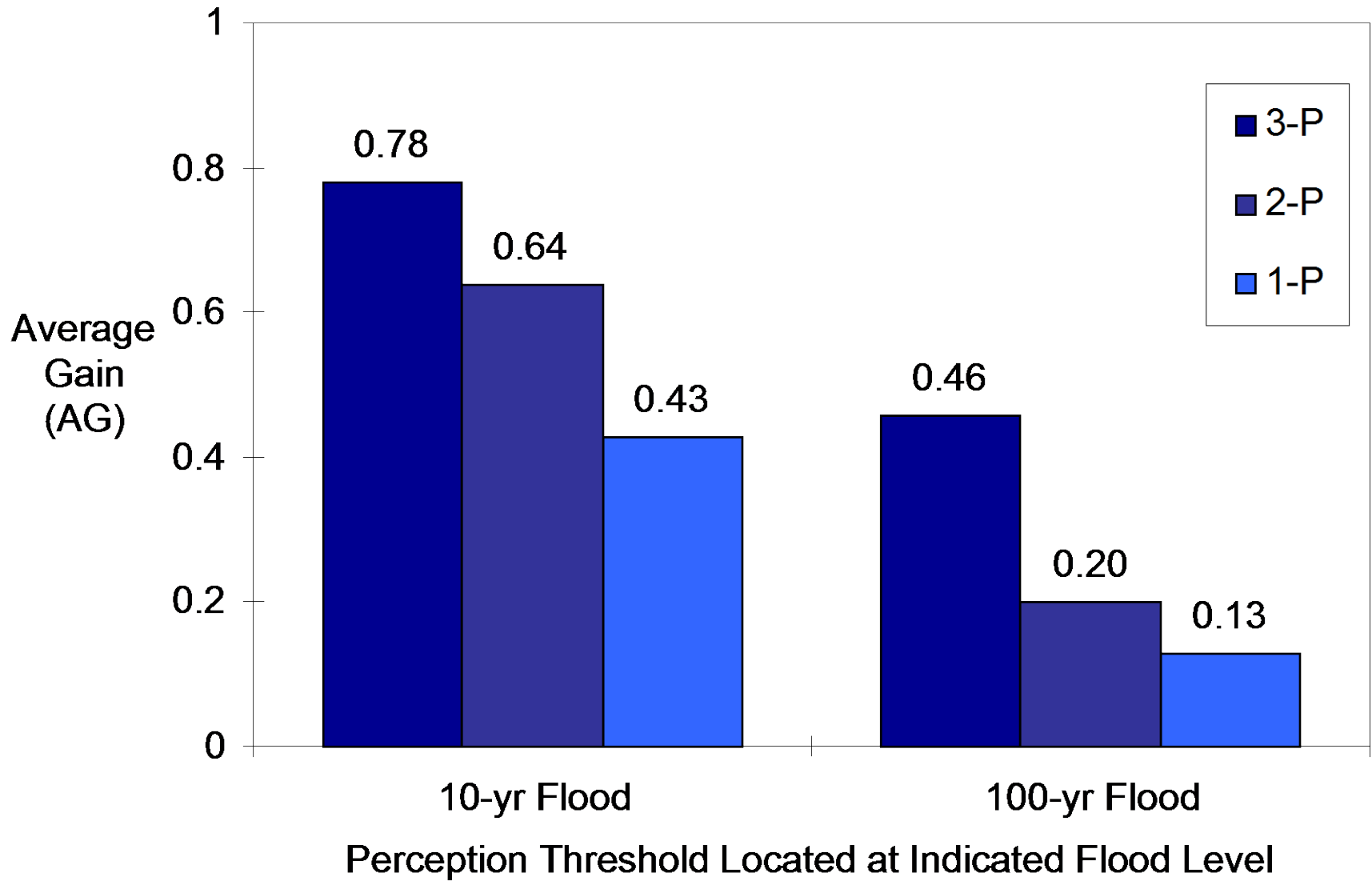




# Average Gains when Estimating 100-year Flood by fitting 2-Parameter Lognormal Distribution ( $s = 20$ years; $h = 100$ years)



# Average Gains when using MLE's to Estimate 1, 2, and 3 Lognormal Parameters



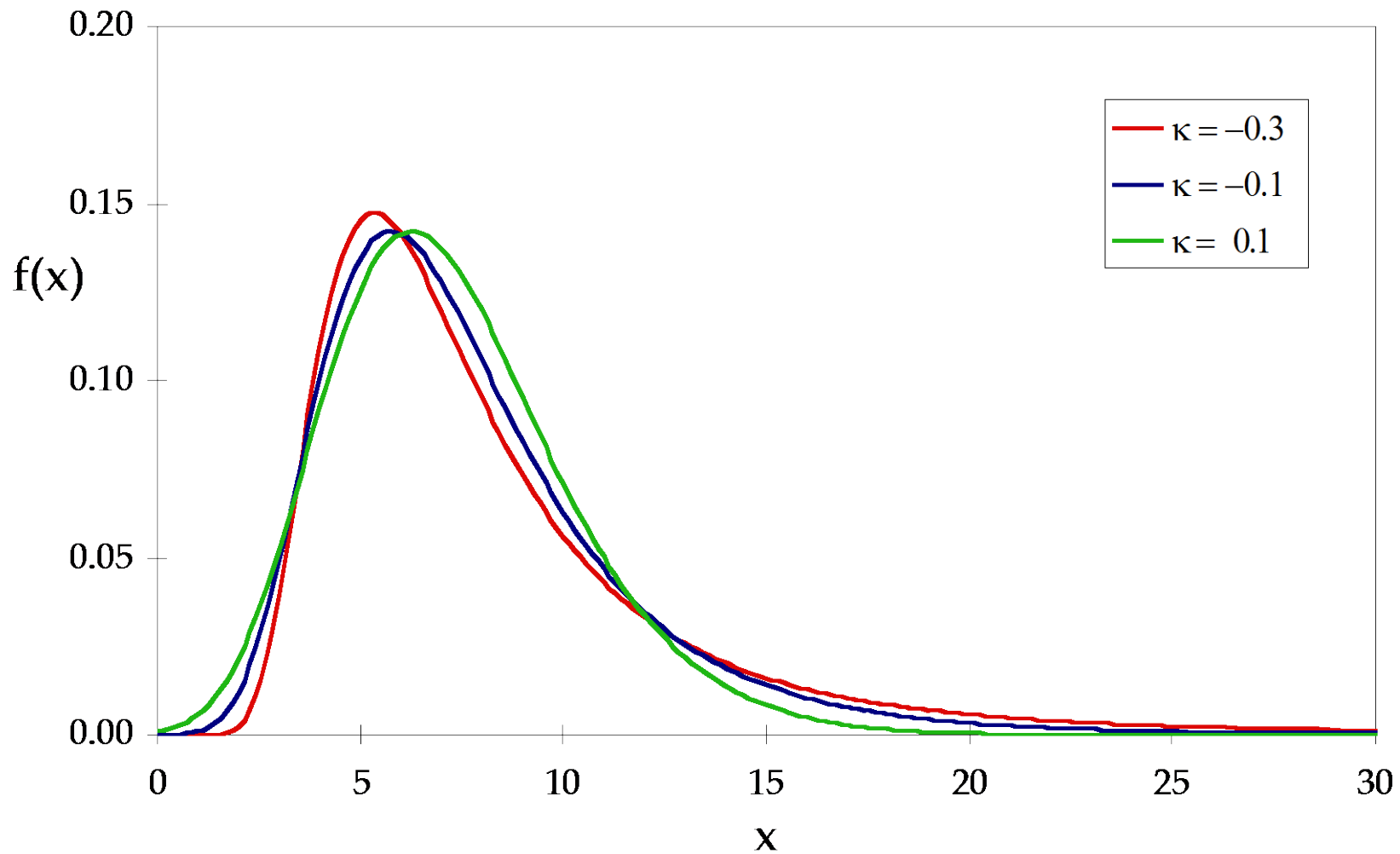
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# Observation:

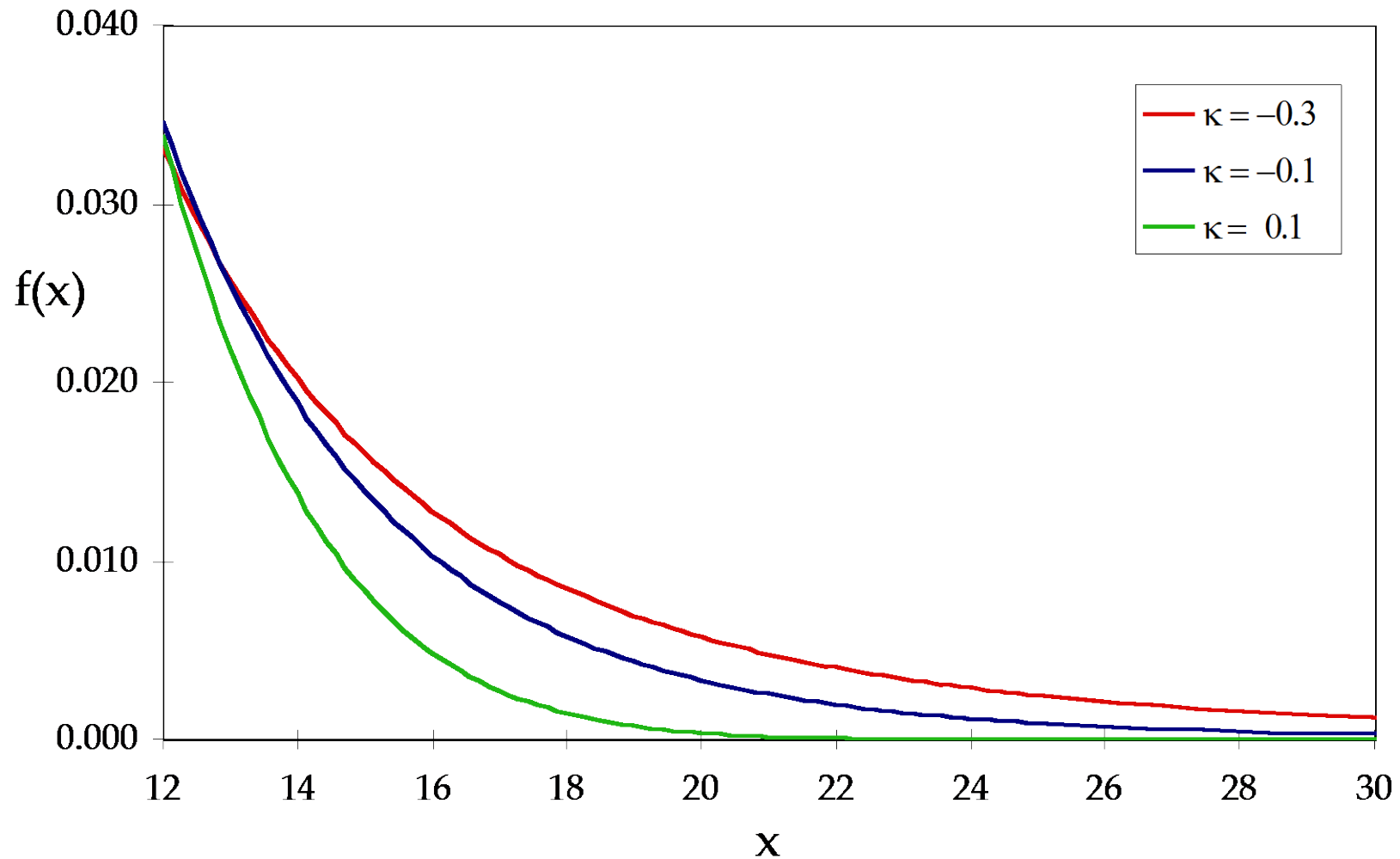
Much larger *Average Gains* when  
estimating 3 parameters.

Shape parameter is hard to pin down  
with short records, and their estimators  
are sensitive to extreme observations.

# Generalized Extreme Value (GEV) distribution



# GEV Prob. Density Function large x



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# Index Flood & Regionalization

Substituting Space for Time.

# Principles for better estimates of extreme flood probabilities

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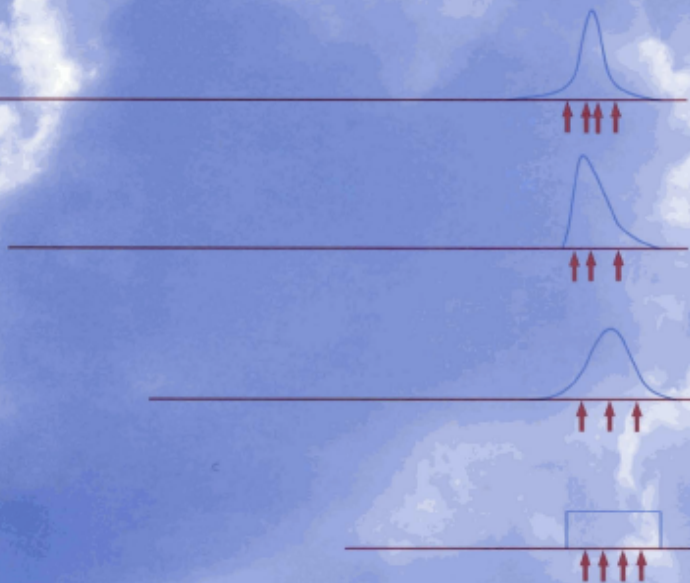
## *U.S. National Research Council report (1988)*

- Substitute space for time using "regionalization."
- Introduce additional structure into models based upon physical insight and hydrologic experience.
- Focus on extremes or "tails" of flood distribution, so less important and common floods do not distort fitted distribution for extreme events.

# Regional Frequency Analysis

## Analysis

An  
Approach  
Based on  
L-Moments



J. R. M. Hosking and J. R. Wallis

Hosking and Wallis  
(1997)

Development of  
L-moments for regional  
flood frequency  
analysis.



Systematic  
Historical  
Records

1

2

3

4

K

*Index Flood  
procedure  
combines many  
flood records*

Dimensionless  
Regional  
Model

$$y_t = \frac{x_t}{\bar{x}}$$

CV  $y_p$

Quantile  
Estimator

Systematic  
Record  
Mean

$$\hat{x}_p = \hat{\mu} y_p$$

Systematic  
Historical  
Records

1

2

3

4

K

Dimensionless  
Regional  
Model

$$y_t = \frac{x_t}{\bar{x}}$$

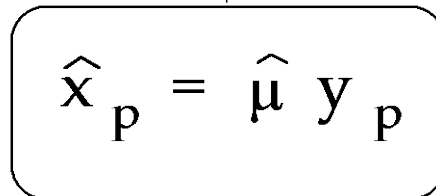
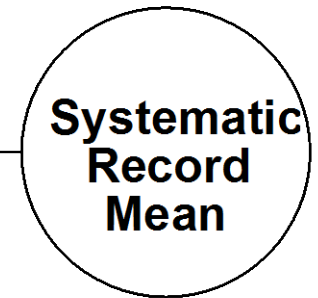
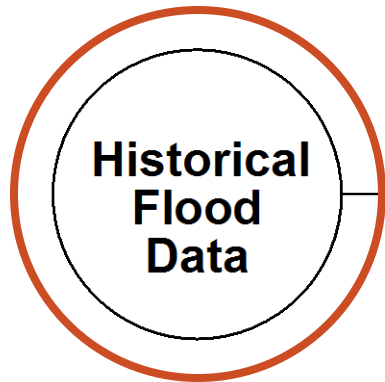
CV  $y_p$

Historical  
Flood  
Data

Generalized  
Maximum  
Likelihood

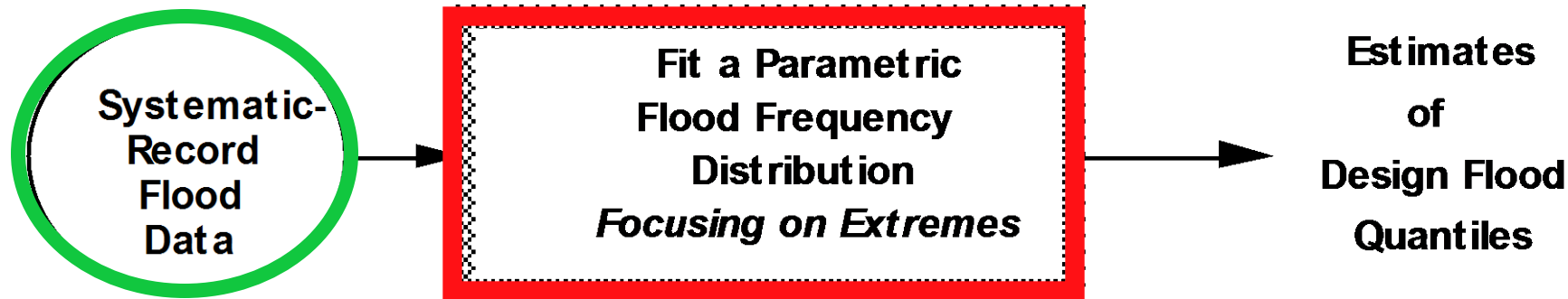
Systematic  
Record  
Mean

$$\hat{x}_p = \hat{\mu} y_p$$

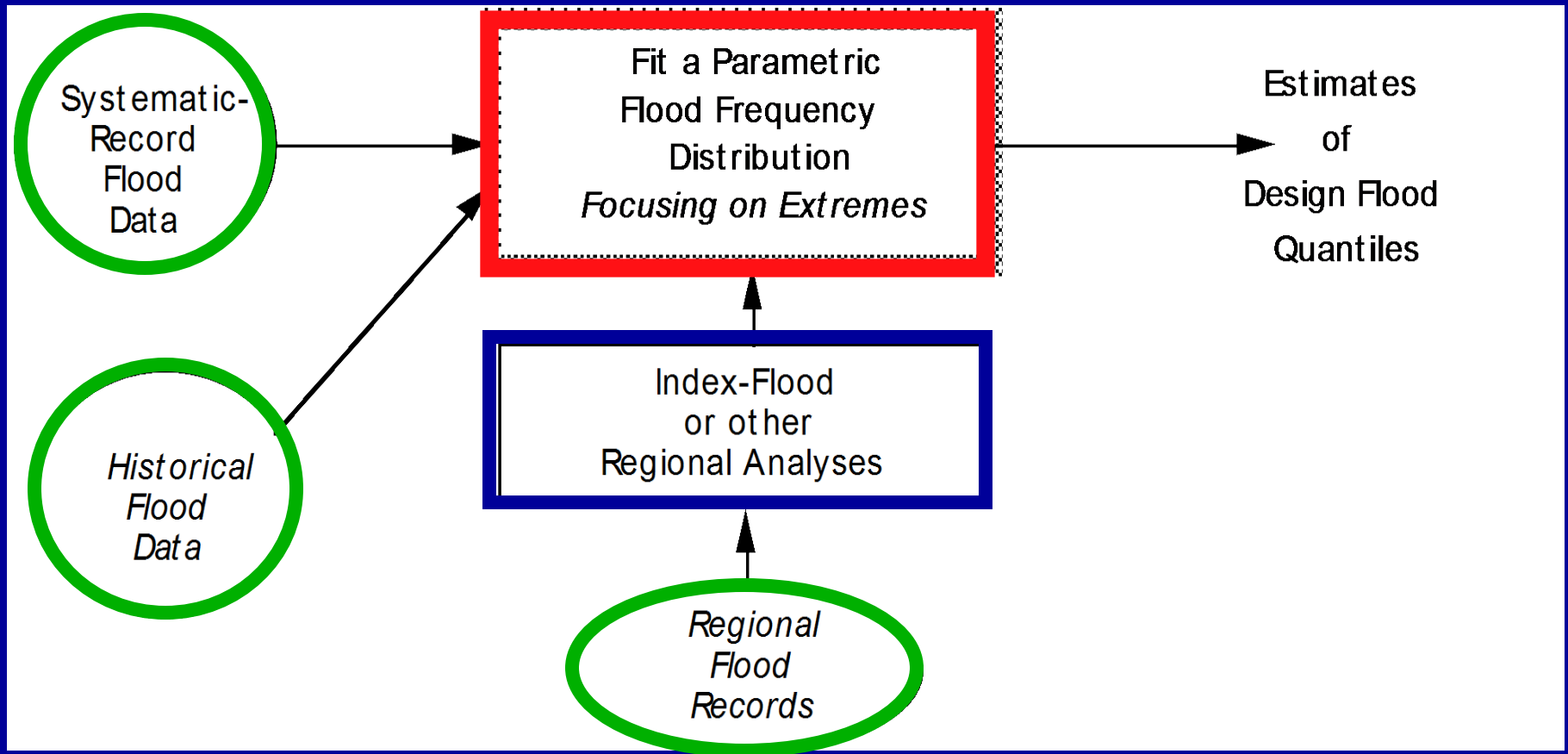


# Flood Frequency Analysis Paradigm

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# New Paradigm for Flood Frequency Analysis



# Credibility of Estimators

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- At site record: 50 - 100 years
- With historical info: 100-200 years
- Regional 20 records:  $20 \times 50 = 1,000$  years
  
- At-site & paleo. info: 1,000 years
- At-site & regional paleofloods 10,000 years

See USBR(1999, 2004) for similar analysis  
of limits of credibility of empirical estimators

# Empirical Analysis Assumes

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- Observed records contains representatives of critical events that represent LARGE & RARE floods at a site.
- Nature is “smooth” and “continuous” so that fitting a mathematically nice GEV or LP3 distribution is reasonable for extrapolation.

# Estimating Flood Risk

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B17B

PFHA (?)



**Empirical**

**Imagine**

Use observed  
Flood Data

Construct using  
regional flood,  
rainfall, climate,  
storm-track, &  
orographic data

T = 100 to 1000 comfortably

T = 10,000 to 1 million

# Addressing Uncertainty

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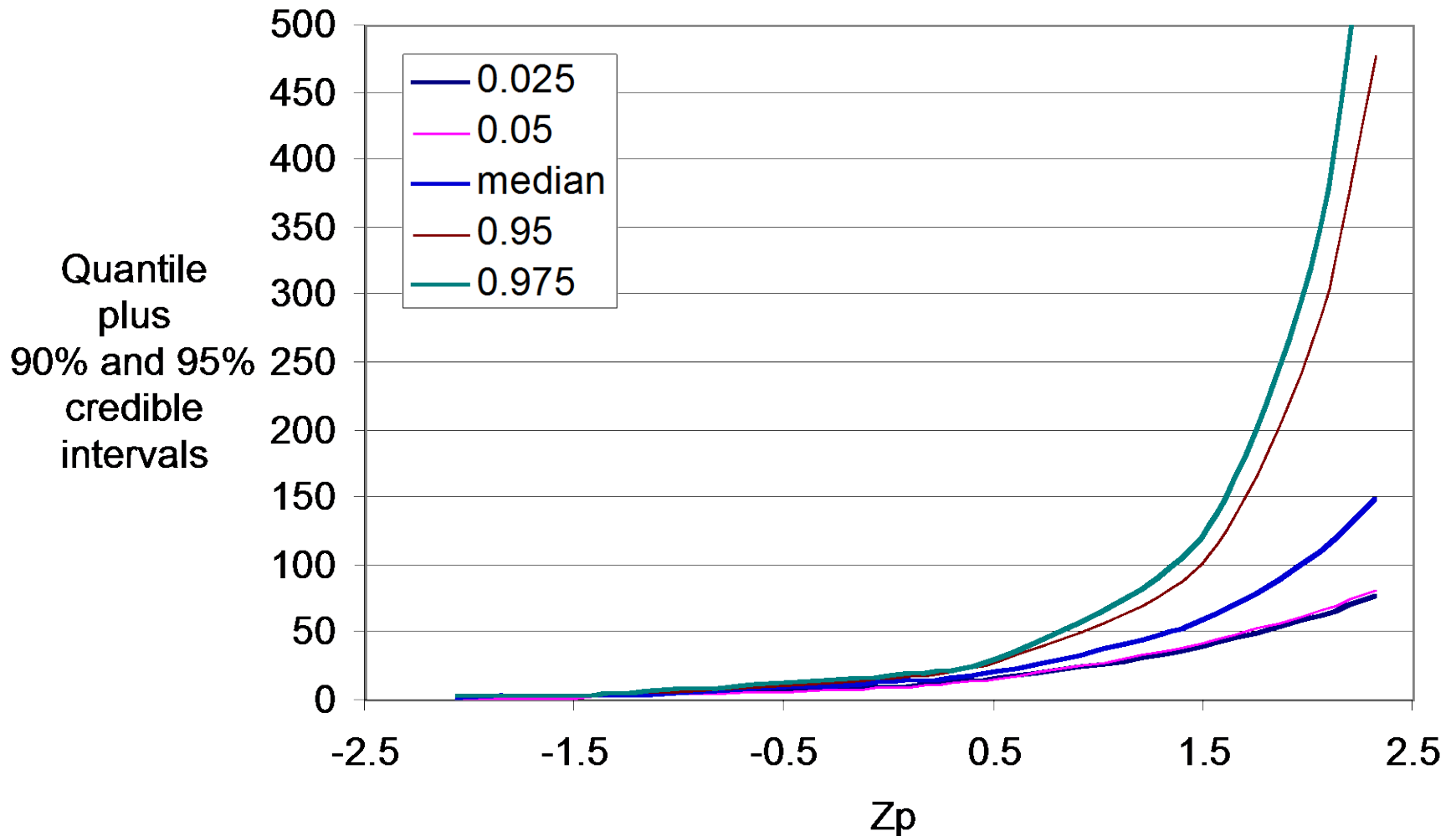
- Consider

**Natural Variability** (*Aleatory Uncertainty*) versus **Knowledge Uncertainty** (*Epistemic Uncertainty*-Limited data to estimate parameters)?

- Do they combined to get total risk?
- Do we give nominal risk with uncertainty bounds reflecting epistemic uncertainty?
- **Flood risk estimates in US (following Bulletin 17B) include only aleatory uncertainty.**



# MCMC for LP3 Quantiles with only systematic information, $s = 100$



# Illustration of Uncertainty

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Drew samples of size  $n = 100$

From 2-parameter lognormal distribution.

Fit 2-parameter lognormal distribution AND

Fit 3-parameter LP3 (with log moments)

Estimated events exceeded with probability of

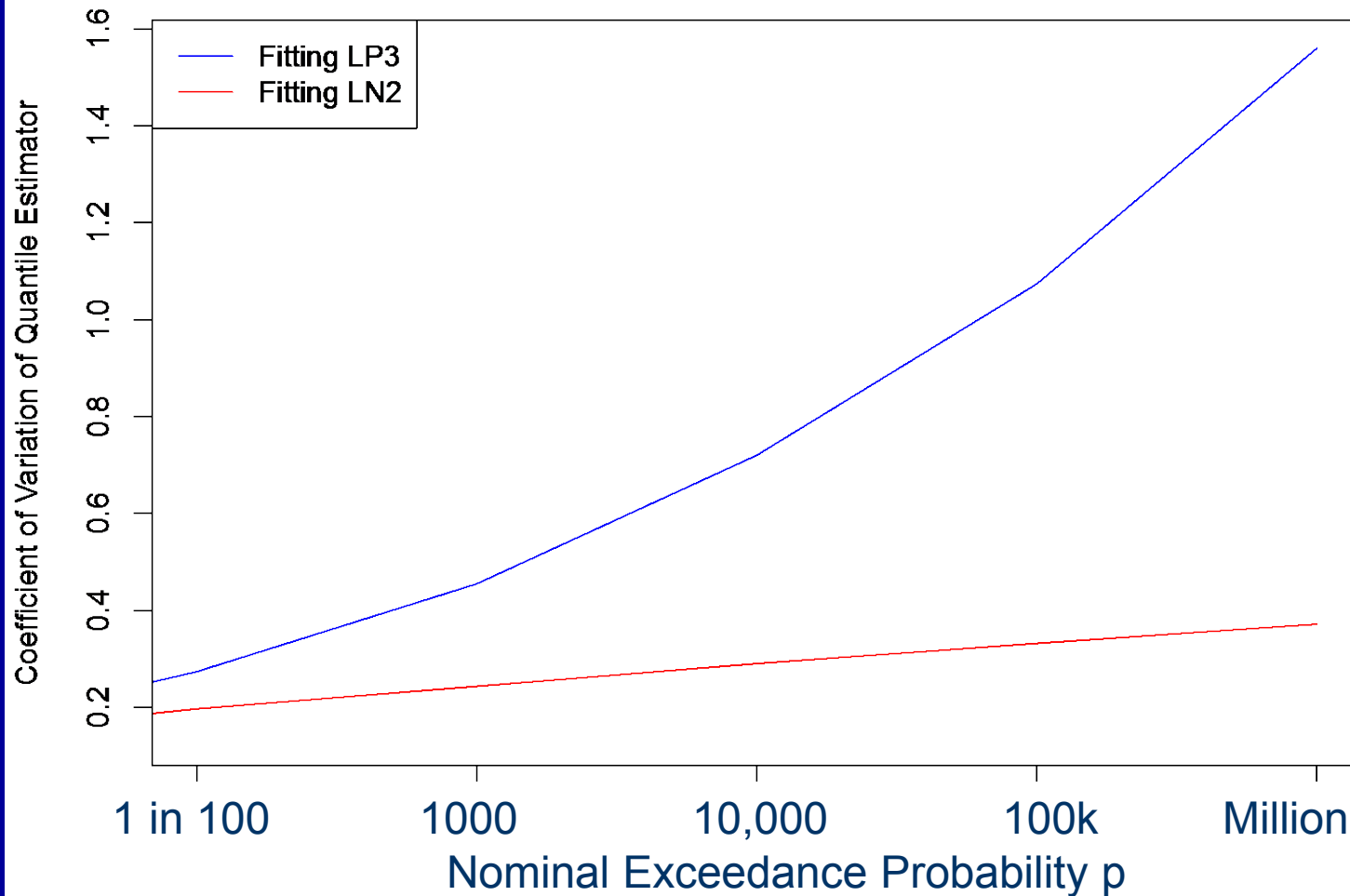
1 in 100 through 1 in a million.

Report ratio of exceedance probability of “design floods”  
to anticipated exceedance probability.

and Coefficient of Variation of “design flood” estimators.

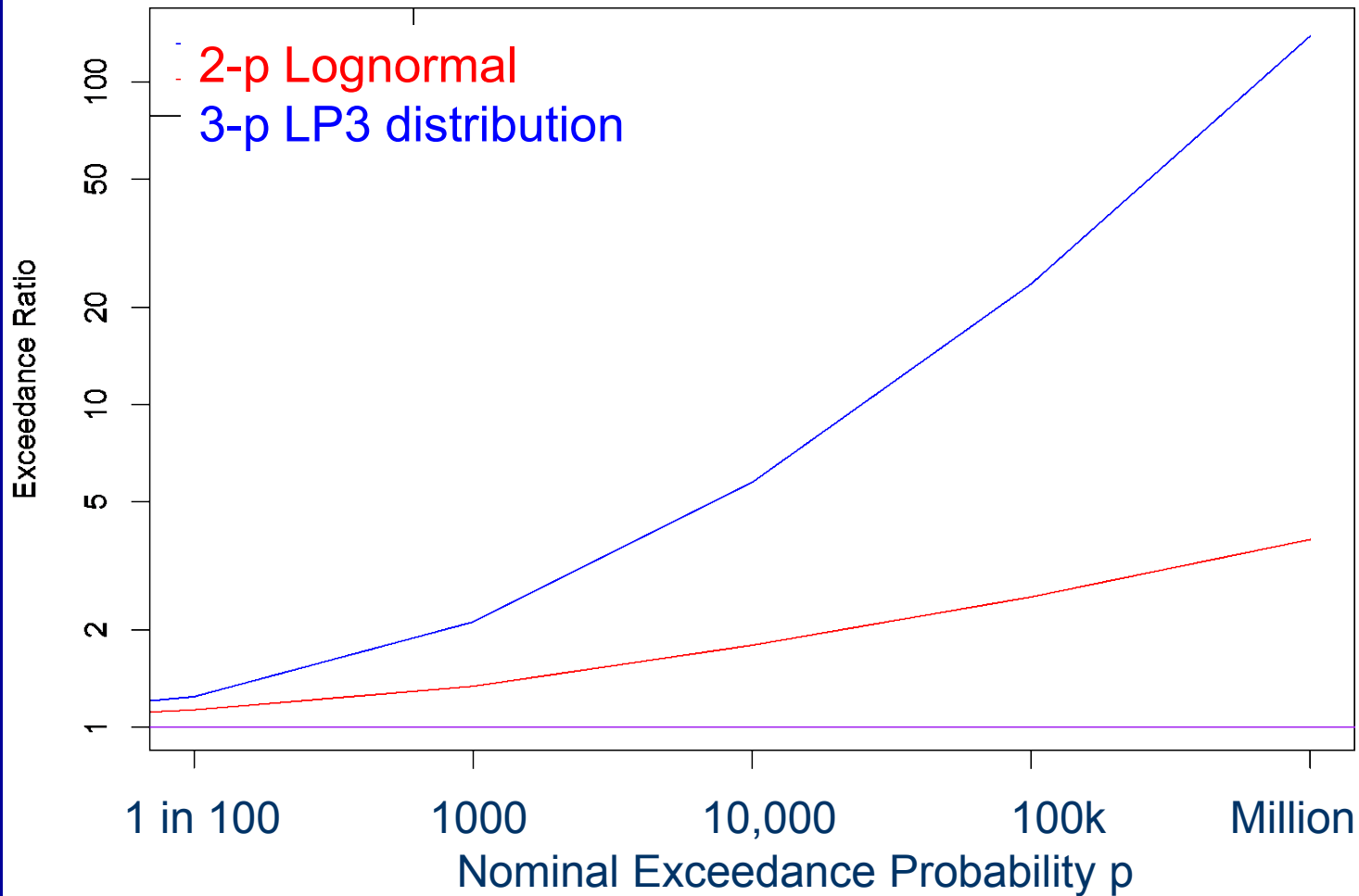
# Coefficient of Variation of Design Flood Estimators

Consider 2-p Lognormal, and 3-p LP3 distributions

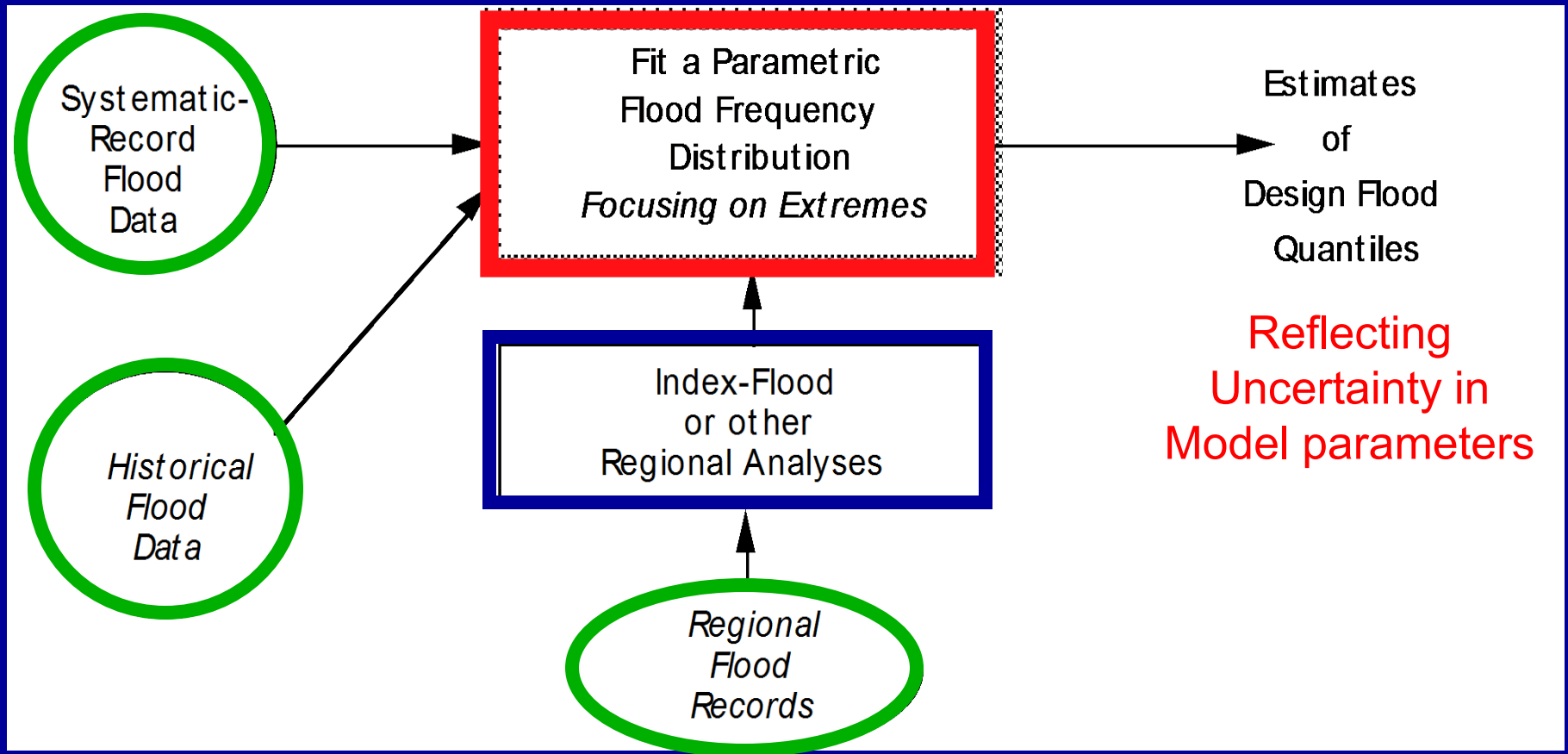


# Error in Exceedance Probability

Ratio exceedance probability of “design flood” to  $p$



# New Paradigm for Flood Frequency Analysis



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# Historical advice

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“We must accomplish a task  
more difficult to many minds  
than daring to know.  
We must dare to be ignorant.”

Concluding two sentences:

Karl Pearson

*The Grammar of Science*, 1892

# Stationarity Is Dead: Whither Water Management?

P. C. D. Milly,<sup>1\*</sup> Julio Betancourt,<sup>2</sup> Malin Falkenmark,<sup>3</sup> Robert M. Hirsch,<sup>4</sup> Zbigniew W. Kundzewicz,<sup>5</sup> Dennis P. Lettenmaier,<sup>6</sup> Ronald J. Stouffer<sup>7</sup>

“In view of the magnitude and ubiquity of the hydroclimatic change apparently now under way, however, we assert that stationarity is dead and **should no longer serve as a central, default assumption** in water-resource risk assessment and planning. Finding a suitable successor is crucial for human adaptation to changing climate.”

Science, Feb. 2008



# Climate Change – So What?

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- (1) Gaged records tell us what common floods look like and where we start.
- (2) Then we need to estimate the change.
- (3) But how to determine where we go from here?

# Climate Change & Risk Projection

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- Imagine 50-100 year record of floods
- Plus paleoflood record illustrating largest floods over 1000-2000 years.
- Fit a line to estimate 1-in-100,000 risk ?
- Paleoflood record likely includes periods of greater & lesser risk, so risk extrapolation OVER estimates actual risk in our time frame.

# Summary

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- Flood records for last century do not reveal reliably flood risk of 1 in 10,000 or 1 in a million.
- At-site Historical and Paleoflood information helps.
- Regionalization of recent & paleoflood records helps
- Extrapolation more uncertain with 3-parameter LP3.
- Climate change confuses things
  - Public loss of confidence in analysis
  - Extrapolating beyond paleoflood data likely conservative

# Issues for Discussion

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- Records of real floods may be insufficient for our purposes: analysis instead needs to *imagine*.
- Uncertainty can be large part of computed total risk.
  - Are we eady to combine variability & knowledge uncertainty into a total risk?
  - Can we do the computation appropriately using honest informative Bayesian priors ?

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Thank you for your attention.

Jery Stedinger

# Donald Rumsfeld, Secretary of Defense

under President George W. Bush from 2001 to 2006  
said at Department of Defense news briefing, February 2002:

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"Reports that say that something hasn't happened are always interesting to me, because as we know,

there are *known knowns*;

there are things we know we know. We also know there are *known unknowns*; that is to say

we know there are some things we do not know.

But there are also *unknown unknowns*

-- the ones we don't know we don't know.

And if one looks throughout the history of our country and other free countries, it is the latter category that tend to be the difficult ones."

# Generalized Extreme Value (GEV) distribution

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Gumbel's Type I, II & III Extreme Value distr.:

$$F(x) = \exp\{ - [ 1 - (\kappa/\alpha)(x-\xi) ]^{1/\kappa} \} \text{ for } \kappa \neq 0$$

$$\mu_X = \xi + (\alpha/\kappa) [1 - \Gamma(1+\kappa)]$$

$$\text{Var}[ X ] = (\alpha/\kappa)^2 [ \Gamma(1+2\kappa) - \Gamma^2(1+\kappa) ]$$

$$E[ (X-\mu_X)^3 ] = (\alpha/\kappa)^3 [ -\Gamma(1+3\kappa) + 3 \Gamma(1+\kappa) \Gamma(1+2\kappa) - 2\Gamma^3(1+\kappa) ]$$

$\kappa$  = shape;  $\alpha$  = scale,  $\xi$  = location.

Mostly

$$-0.25 < \kappa \leq 0$$