

# HTCondor goes fishing: A USGS National Fish Mercury Advisory Model



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# Validation of the National Descriptive Model of Mercury in Fish

Mercury, fish, and human health

A descriptive statistical model

Validation efficiency

The role of HTCondor



Monument to Minamata  
Mercury Poisoning Victims

# What's the deal with Mercury?

Powerful neurotoxin in humans

Bioaccumulates in fish

Generally, mercury concentration increases with background concentration and fish size

Important social justice implications

USGS as a national science agency is responsible for looking at these issues on a national scale

# What's the deal with Mercury?

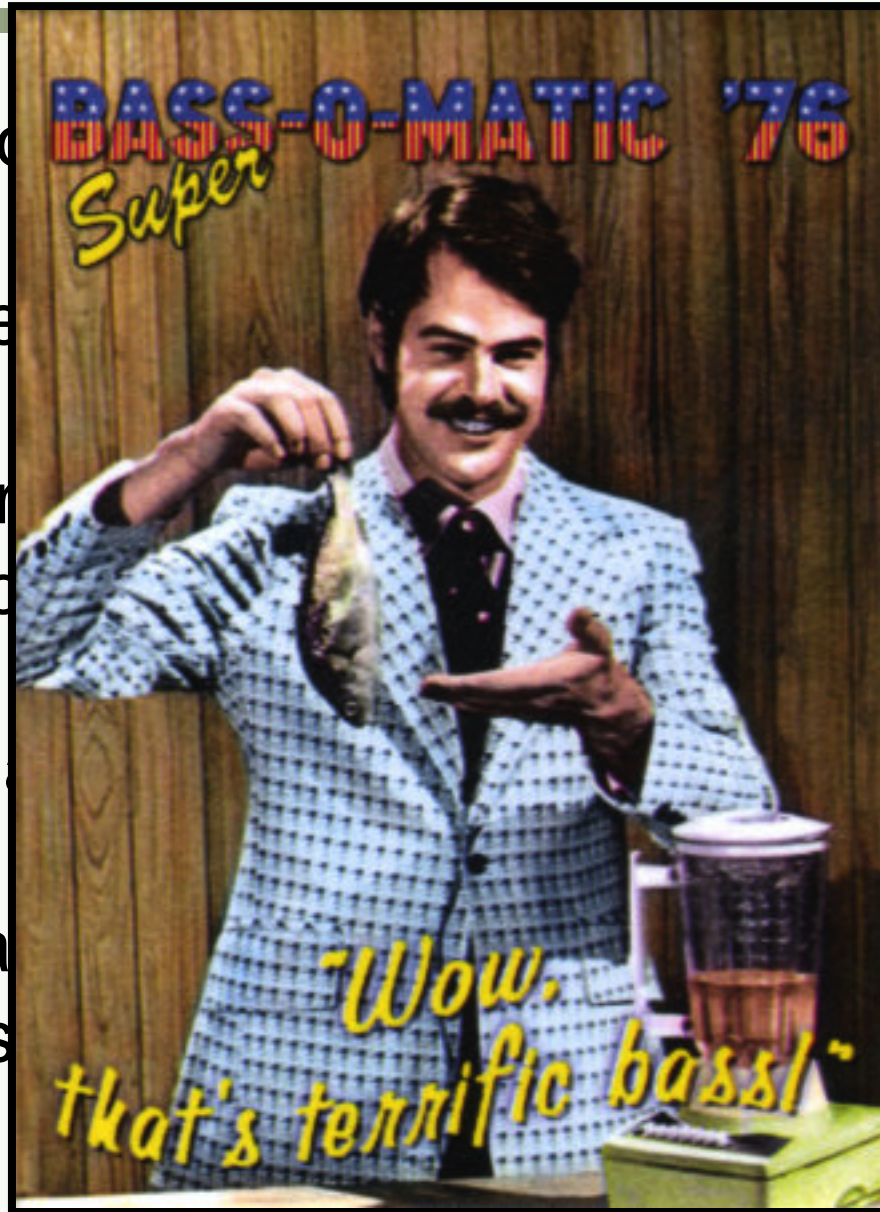
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# The National Descriptive Model of Mercury in Fish

Analysis of Covariance (general linear model) applied to a large, national fish-mercury data set.

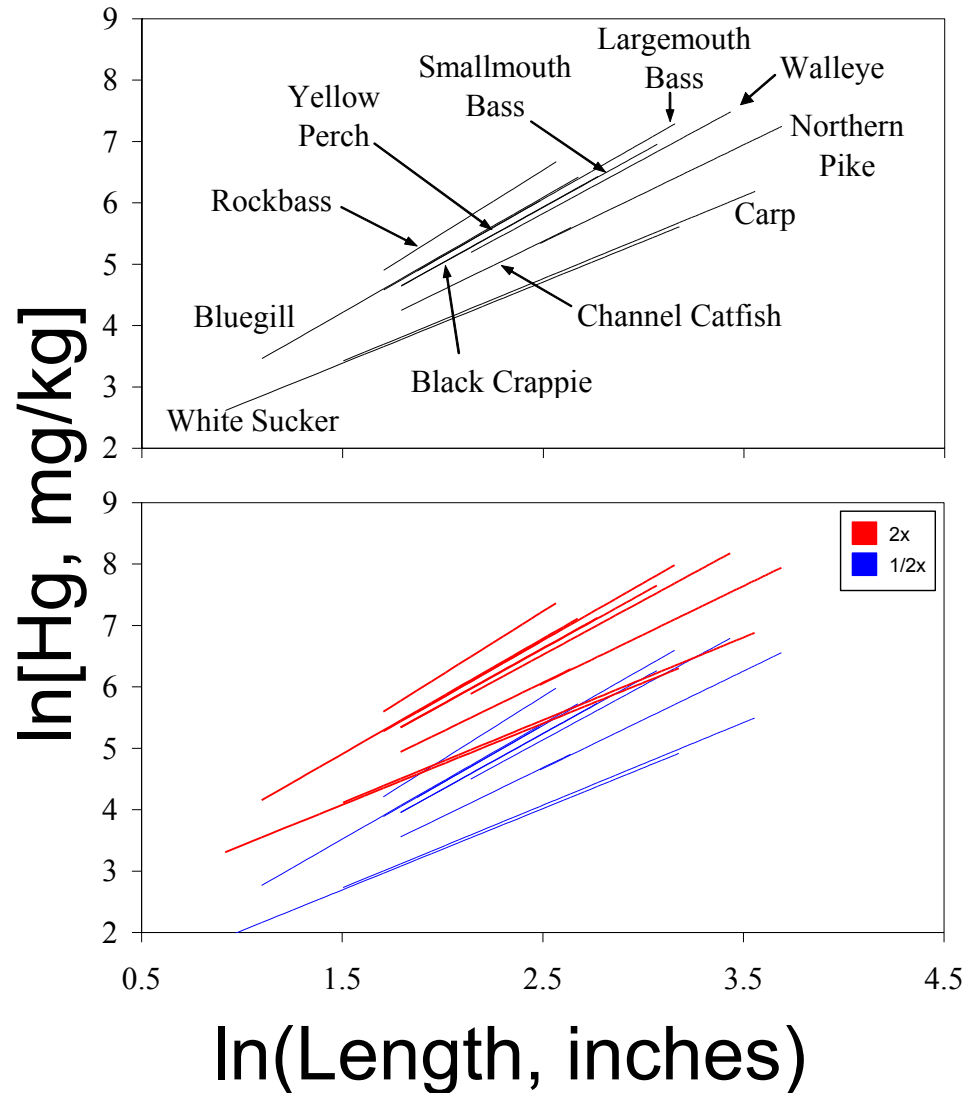
Relates fish [Hg] to fish length for many **events** and **species & cuts**, simultaneously.

$$\ln[Hg+1] = \alpha_j + \beta_k \ln[length+1] + \varepsilon$$

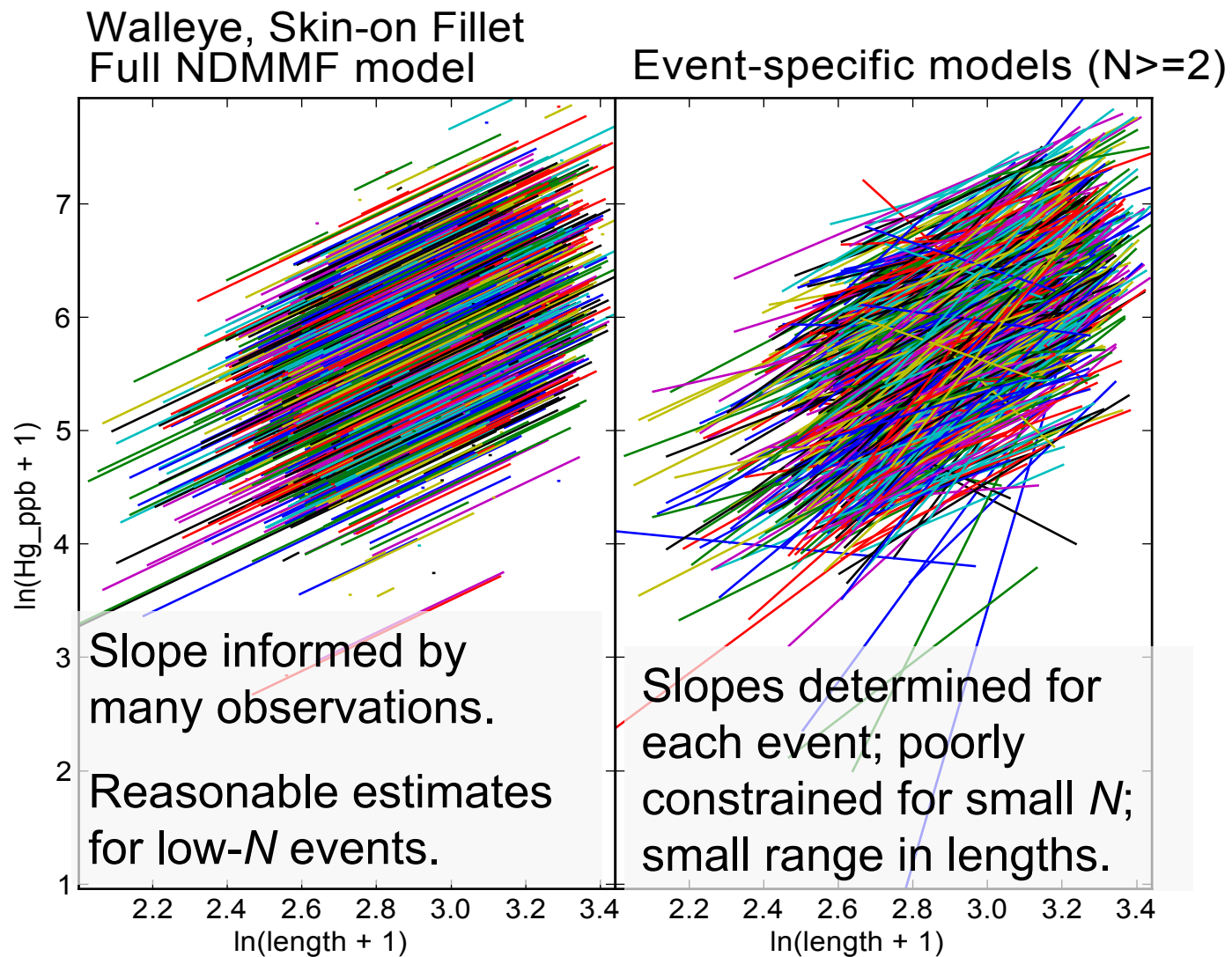
Intercept (event parameter) for  $j^{th}$  event. All fish samples collected at a site in a calendar year comprise an event.

Slope (length parameter) for  $k^{th}$  species-cut combination.

# Conceptualized model for 11 species, 3 events

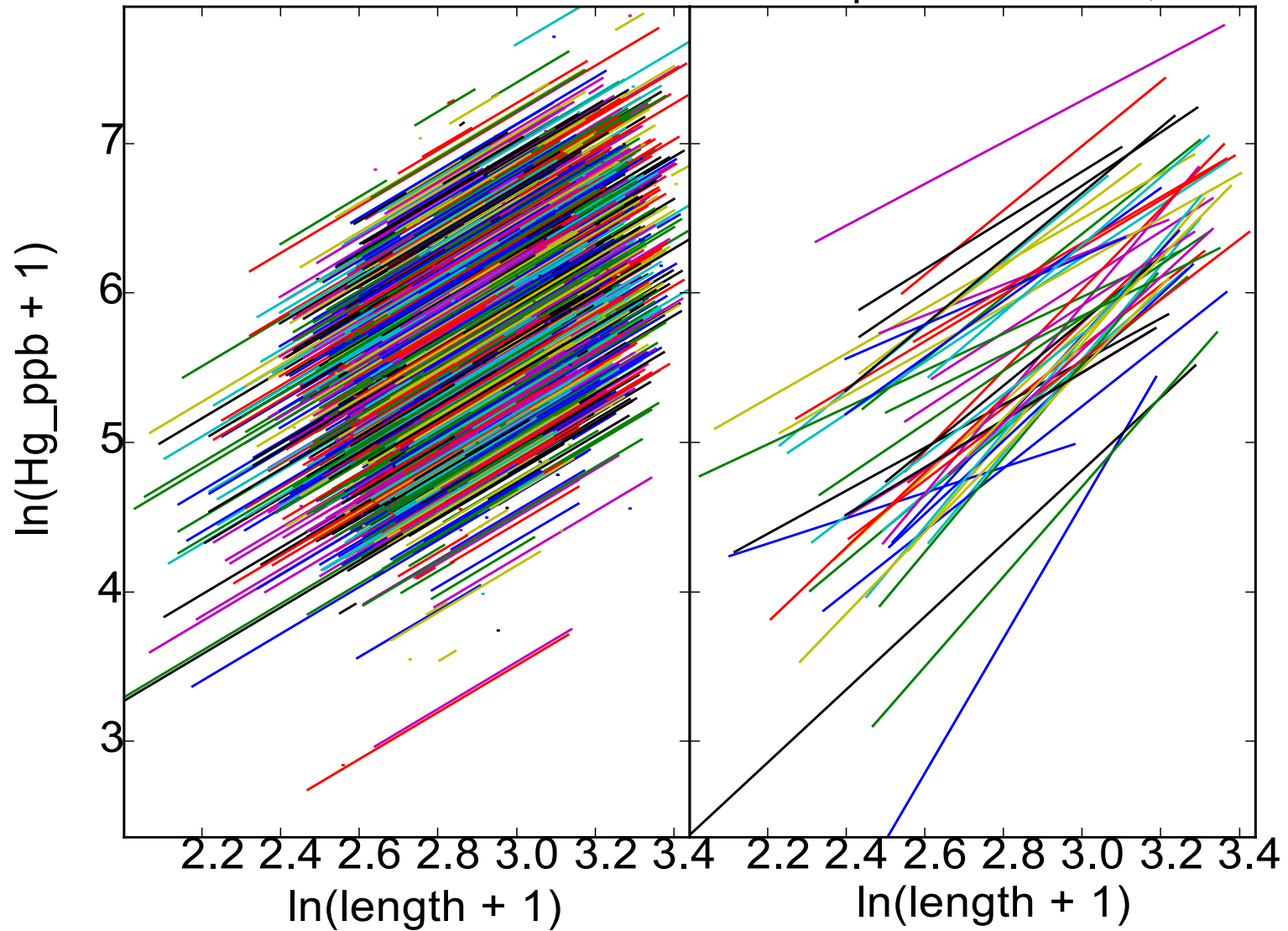


# Model assumes constant slope for species-cut



Walleye, Skin-on Fillet  
Full NDMMF model

Event-specific models,  $n \geq 10$





# Strength in numbers! Large *inference space*



Sampled species at an event

Can estimate (predict) [Hg] for same event, for *all* species in data set.

...unless unconnected to at least one event



# This Model was Published but not Validated

2004 NDMMF solved by Maximum Likelihood regression, using PROC LIFEREG in SAS 9.3 on 64-bit Unix (35 hour forward run time)

Data set updated in 2006:

102,000 observations; 10,000 sites across U.S.

Used ~99,000 observation subset, where  $N > 1$ , and all samples were “connected” to data base

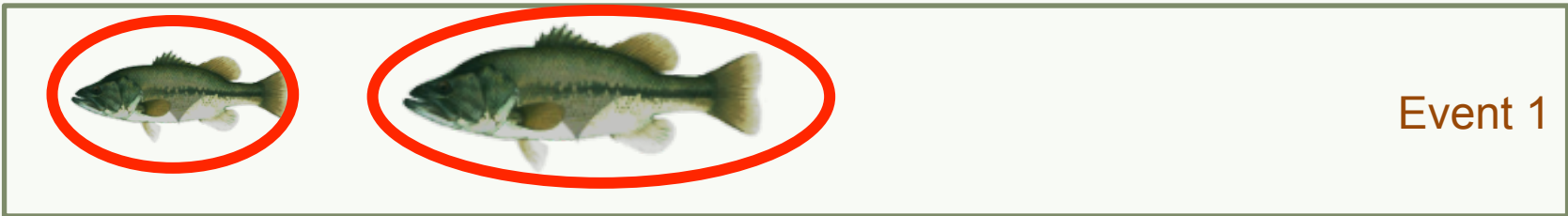
Validation study in 2010-2012:

Python code replaced manual spreadsheet fiddling to establish connectivity of events

C-code facsimile model that converges in seconds

Run in parallel on 76 CPUs with HTCondor

# Leave-One-Out Cross-Validation



...through all observations, and all events where  $N \geq 2$ .



# The Role of HTCondor

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High Throughput computing can increase efficiency...

...but more importantly, it can enable science that could not otherwise be done.



# Leave-One-Out Cross-Validation

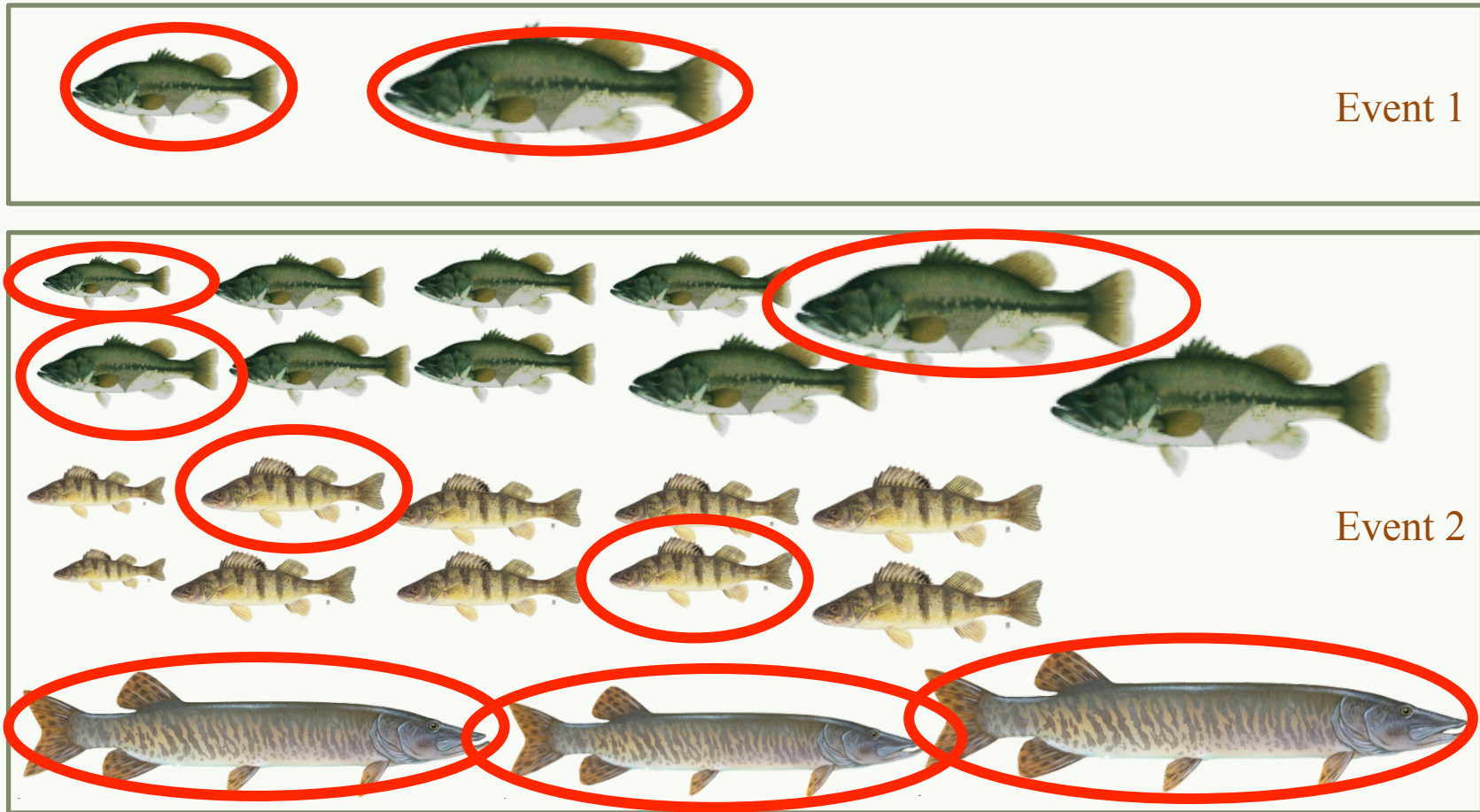
~100,000 runs x 35 hours = 400 CPU *years*

Reducing run times to ~3 minutes and distributing over 76 CPUs = 1-2 CPU *days*

*Beware the notification setting!*

```
notification = Never
universe = vanilla
log = log/worker_$(Cluster).log
output = log/worker_$(Cluster)_$(Process).out
error = log/worker_$(Cluster)_$(Process).err
requirements = ( (OpSys == "LINUX"))
executable = worker.sh ...
```

# Repeated Random Sub-Sampling



Randomly drop 10% of samples. Remove unconnected species and  $N=1$  events. Run model. Repeat 1,000 times (with replacement).

Describe model performance by specific cases:  
Species & Cut, State, Water body type, Site.

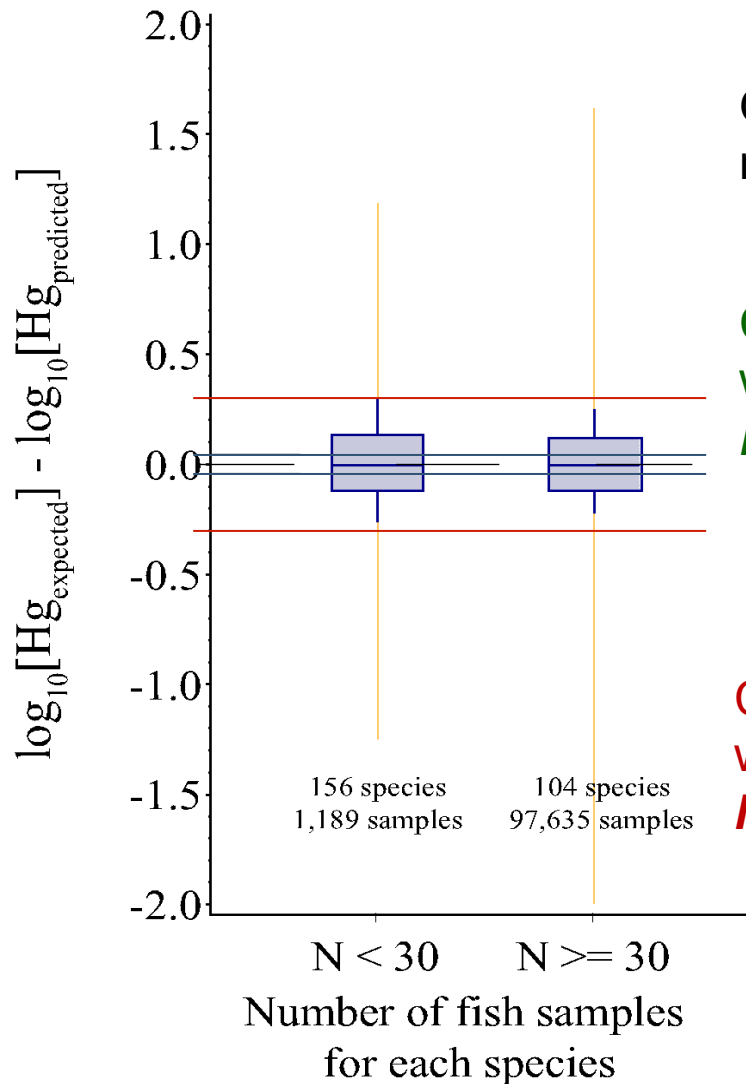
*Prediction Errors:*

$$PE = \log_{10}[Hg_{expected}] - \log_{10}[Hg_{predicted}]$$

Where:

$$[Hg_{expected}] = \begin{cases} [Hg_{observed}] & \text{if detected} \\ \text{inverse Mills ratio} & \text{if} \\ [Hg_{observed}] < \text{reporting limit.} \end{cases}$$

# Summary of Prediction errors: bias & variability



Overall, median PE  $\approx 0$ , indicating no *bias*.

Cases where median  $[\text{Hg}_{\text{predicted}}]$  is within 10% of  $[\text{Hg}_{\text{expected}}]$ :  
**low bias**

Cases where >80% of  $[\text{Hg}_{\text{predicted}}]$  are within 0.5x–2x of  $[\text{Hg}_{\text{expected}}]$ :  
**low variability**

# Conclusions

NDMMF incorporates greater inference space than traditional models

Predicts for species & cuts that were not sampled in an event; traditional models cannot.

Validation shows low bias for all states and water-body types; and for most species & cuts and sites

Some small- $N$  cases showed larger bias. As more data are added, parameter estimates are better constrained.

HTCondor is essential to enable large-scale analysis

It not only made our work more efficient, but enables science application that would otherwise not be possible



# NDMMF Future Applications

More complete and efficient fish consumption advisories using a consistent national model

1x10<sup>6</sup>+ samples from Environment Canada can be added to expand the range of applicability

# Acknowledgements

**Funding provided by  
The USGS National Water Quality Assessment Program**

**Thanks to  
All the State and Local agencies for providing data  
Steve Wentz (USEPA) for writing the first version of NDMMF**

**IT support  
CIDA: Ben Feinstein, Daniel Kester, and Chad Ingle  
Wisconsin Water Science Center: Ryan Heath and Dave Owens**

**USGS-CIDA Environmental Modeling Unit  
Randy Hunt, Harry House, Nate Booth, Scott Lewein**

**CHTC  
Miron Livny, Brooklin Gore, Todd Tannenbaum,  
Vladimir Brik, Cathrin Weiss**

