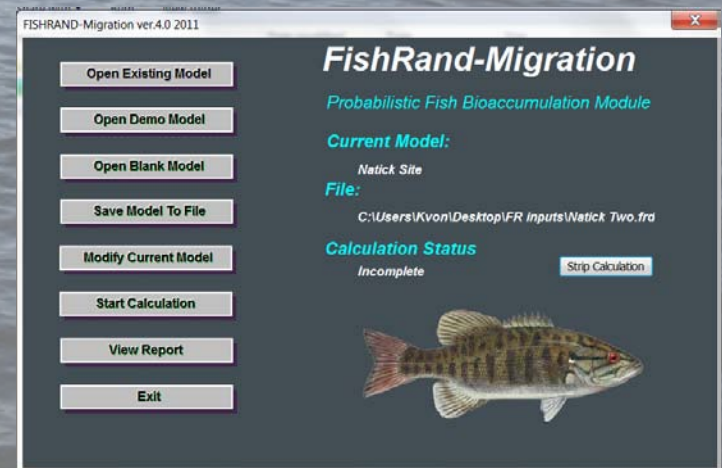


Spatially-explicit bioaccumulation modeling to support human health and ecological risk assessments in a decision analytic context

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Bioaccumulation Models

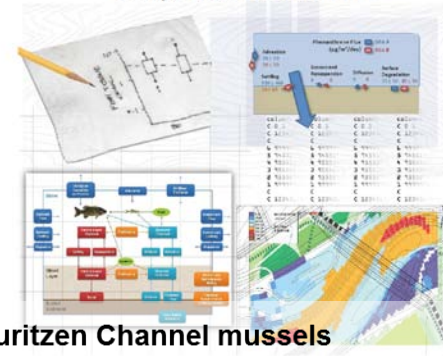
- Used to predict concentrations in aquatic organisms exposed to sediment-associated contaminants
- Used to evaluate changes in predicted organism concentrations (and resulting human health and ecological risks) as a result of different management alternatives
- Most, if not all, applications based on static exposures (e.g., SWAC, average, etc.)



Office of Superfund Remediation and
Technology Innovation

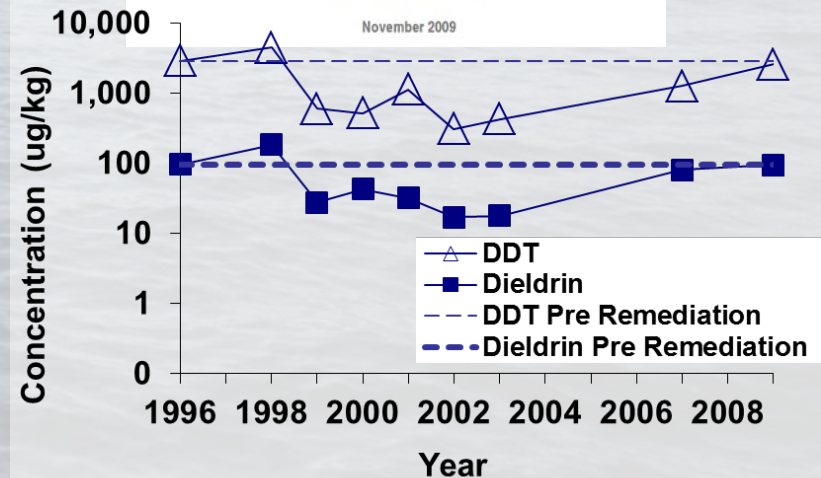
Sediment Assessment and Monitoring Sheet (SAMS) #2

Understanding the Use of Models in Predicting
the Effectiveness of Proposed Remedial Actions
at Superfund Sediment Sites



Lauritzen Channel mussels

OSWER Directive 9200.1-96FS



Characterizing Exposures

0 mg/kg outside
modeling domain

66.7
mg/kg in
sediment
within
modeling
domain



- External processing of static exposures represented by one value (average, SWAC, etc.)
- Bioaccumulation models do not represent fish behavior, foraging strategy, life history, habitat preferences
- Either exposed or not – doesn't capture dynamics of fish behavior
- Don't typically capture changes over time
- Doesn't capture uncertainty and variability

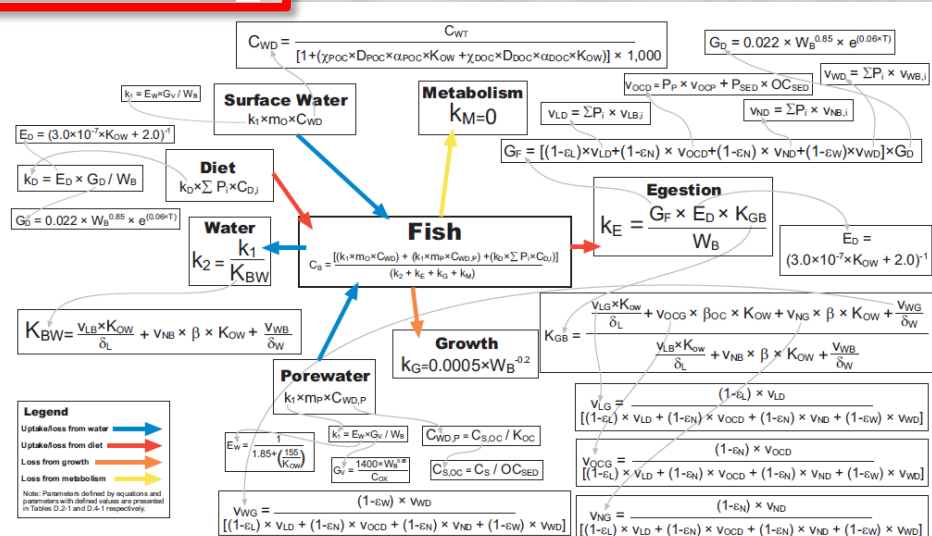
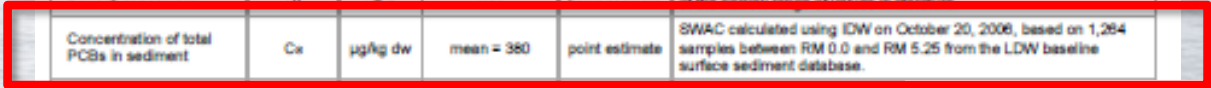
Bioaccumulation Models: State of the Application at Large Superfund Sites by Karl Gustavson, Katherine von Stackelberg, Igor Linkov, and Todd Bridges



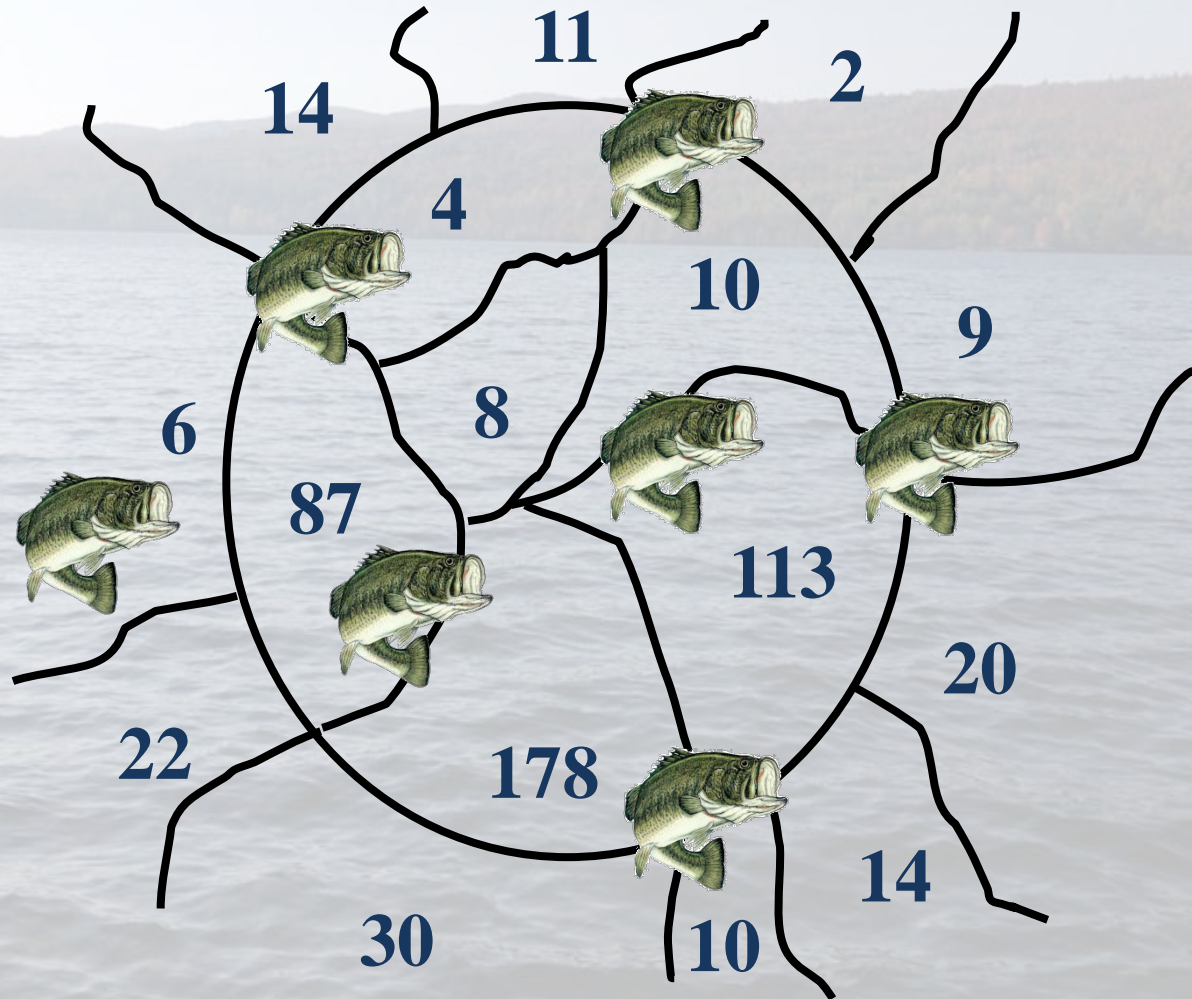
Table D.4-1, cont. Input parameter probability distribution statistics and point estimate values

PARAMETER	SYMBOL	UNIT	VALUES*	DISTRIBUTION
Mean temperature of water column water	T	°C	mean = 11.2 SE = 0.307	normal
Dissolved oxygen concentration in water column water	C _{ox}	mg/L	mean = 7.93 SE = 0.203	normal
TSS concentration in water column water	C _{ss}	kg/L	mean = 5.8 × 10 ⁻⁶ SE = 8.8 × 10 ⁻⁷	normal
Density of seawater	δ _w	kg/L	1.03	point estimate
Concentration of total PCBs in sediment	C _s	µg/kg dw	mean = 380	point estimate
Sediment organic carbon	OC _{sed}	%	mean = 1.91 SE = 0.025	normal
Chemical Parameters				
Log octanol-water partition coefficient for total PCBs	log K _{ow}	L/kg	mean = 6.8 SE = 0.05	normal
Proportionality constant expressing the sorption capacity of NLOM for an organic chemical relative to that of octanol	β	unitless	mean = 0.035 SE = 0.005*	normal
Proportionality constant expressing the sorption capacity of NLOC for an organic chemical relative to that of octanol	β _{oc}	L/kg	0.35	point estimate
Rate constant for metabolic transformation of total PCBs	k _m	day ⁻¹	0	point estimate

Application of the Arnot and Gobas (2004a) FWM to the LDW required the selection of values for 114 input parameters (including dietary fractions). Because the Arnot and



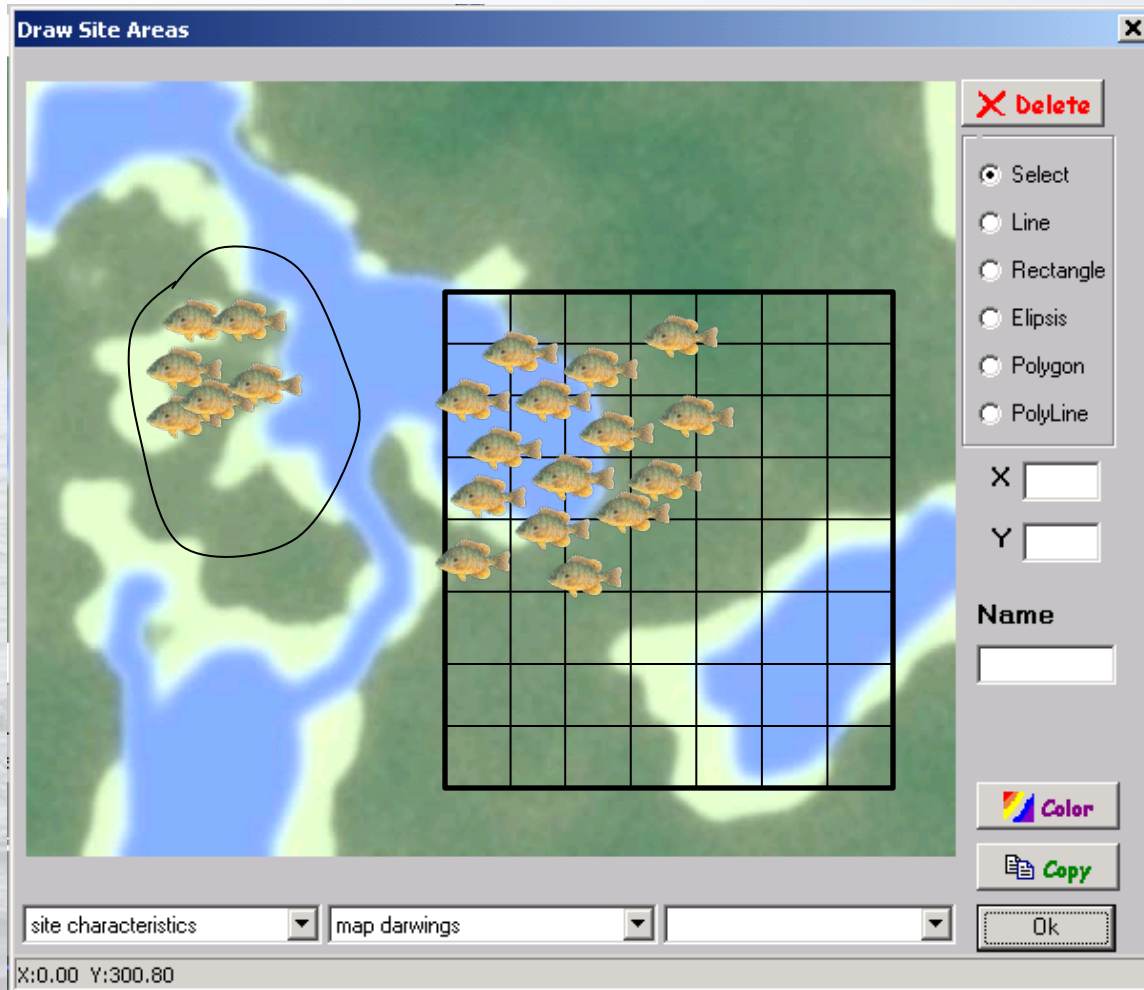
Spatial Heterogeneity in Exposure



Motivation for Spatially-Explicit Approach

- Spatial and temporal scales associated with ecological receptor exposure
- Fish have localized movements
 - Daily foraging strategies
 - Preferential habitat and foraging areas
- Seasonal habits
 - Migratory status
 - May leave modeling grid for parts of the year
 - Offshore movements

FishRand Approach



- Sampling from a population of fish
- Movements and foraging strategies contribute to the distribution of predicted tissue concentrations
- Takes advantage of GIS-based sediment concentrations
- Probabilistic linkages
 - Decision analytic approaches
 - Integration with economic and other data

Importance of Uncertainty and Variability to Predicted Risks from Trophic Transfer of PCBs in Dredged Sediments

Katherine E. von Stackelberg,^{1,*} Dmitriy Burmistrov,¹ Donna J. Vorhees,¹ Todd S. Bridges,² and Igor Linkov³

Biomagnification of organochlorine and other persistent organic contaminants by high trophic level organisms represents one of the most significant sources of uncertainty and variability in evaluating potential risks associated with disposal of dredged materials. While important to distinguish between population variability (e.g., true population heterogeneity in fish weight, and lipid content) and uncertainty (e.g., measurement error), they can be operationally difficult to define separately in probabilistic estimates of human health and ecological risk. We propose a disaggregation of uncertain and variable parameters based on (1) availability of supporting data; (2) the specific management and regulatory context (in this case, of the U.S. Army Corps of Engineers/U.S. Environmental Protection Agency tiered approach to dredged material management); and (3) professional judgment and experience conducting probabilistic risk assessments. We describe and quantitatively evaluate several sources of uncertainty and variability in estimating risk to human health from trophic transfer of polychlorinated biphenyls (PCBs) using a case study of sediments obtained from the New York-New Jersey Harbor and being evaluated for disposal at an open water off-shore disposal site within the northeast region. The estimates of PCB concentrations in fish and dietary dose of PCBs to humans ingesting fish are expressed as distributions of values, of which the arithmetic mean or mode represents a particular fractile. The distribution of risk values obtained using a food chain biomagnification model developed by Gobas^(1,2) by specifying distributions for input parameters disaggregated to represent either uncertainty or variability. Only those sources of uncertainty that could be quantified were included in the analysis. Results for several different two-dimensional Latin Hypercube analyses are provided to evaluate the influence of the uncertain versus variable disaggregation of model parameters. The analysis suggests that variability in human exposure parameters is greater than uncertainty bounds on any particular fractile, given the described assumptions.

KEY WORDS: Biomagnification; probabilistic risk assessment (PRA); polychlorinated biphenyls; dredged material; trophic transfer; uncertainty and variability



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The use of spatial modeling in an aquatic food web to estimate exposure and risk

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Todd S. Bridges^c

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Received 8 June 2001; accepted 15 October 2001

Abstract

This paper quantitatively evaluates interactions among foraging behavior, habitat preferences, site characteristics and the spatial distribution of contaminants in estimating PCB exposure concentrations for winter flounder at a hypothetical open water dredged material disposal site in the coastal waters of New York and New Jersey (NY–NJ). The models implemented in this study include a spatial submodel to account for spatial and temporal characteristics of fish exposure and a probabilistic adaptation of the Gobas bioaccumulation model to account for temporal variation in concentrations of polychlorinated biphenyls (PCBs) in sediment and water. We estimated the geographic distribution of an offshore winter flounder subpopulation based on species biology, including such variables as foraging area, habitat size, disposal site size and migration characteristics. We incorporated these variables together with an estimate of differential attraction to a management site within a spatially explicit model to assess the range of expected PCB exposures to a winter flounder population. The output of this modeling effort, flounder PCB tissue concentrations, provides exposure point concentrations for estimates of human health risk through ingestion of locally caught flounder. The risks obtained for the spatially non-explicit case are as much as one order of magnitude higher than those obtained after incorporating spatial and temporal characteristics of winter flounder foraging and seasonal migration. Incorporating spatial and temporal variables in food chain models can help support sediment management decisions by providing a quantitative expression of the confidence in risk estimates. © 2002 Elsevier Science B.V. All rights reserved.

Keywords: Bioaccumulation; PCBs; Exposure assessment; Spatial; Probabilistic

SERDP Demonstration Project

DEFINE MODELING AREA:

- Base map with GIS-based spatially-defined exposures

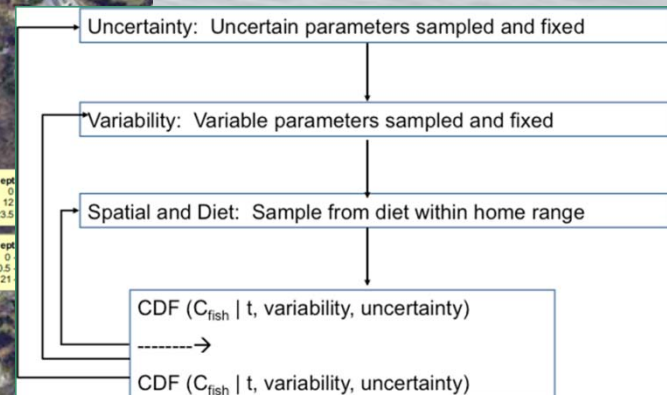
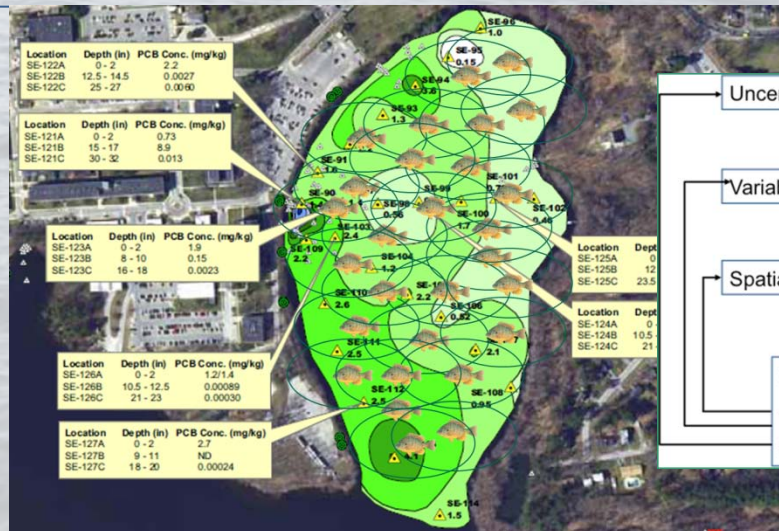
SPECIFY INPUTS:

- Simple food web (water column and benthic inverts, pumpkinseed, bluegill, yellow perch, largemouth bass)

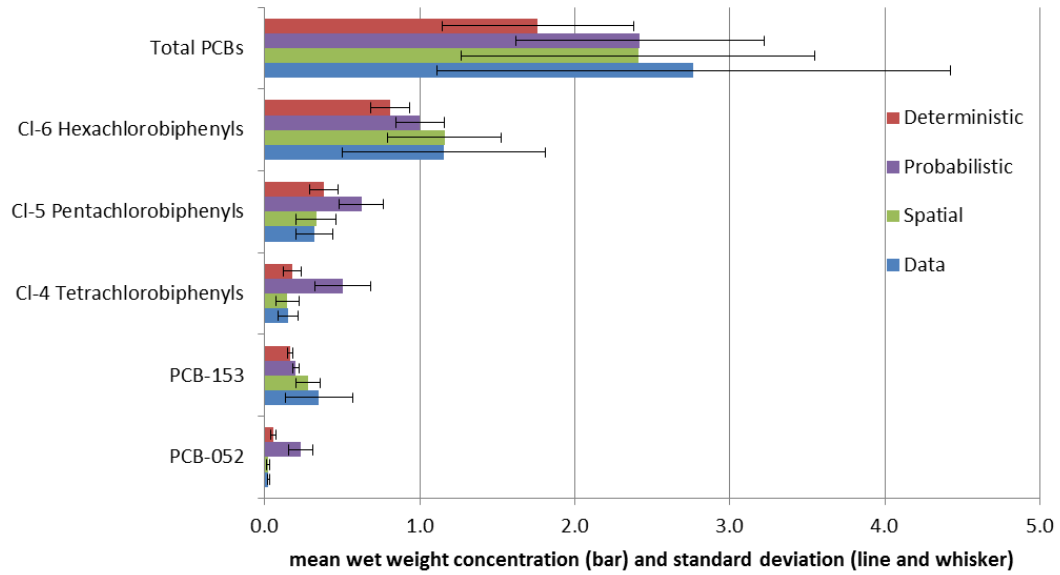
THREE MODEL RUNS:

- Inputs to all three runs identical except for sediment concentrations
 - Deterministic (SWAC)
 - Probabilistic (distribution but still averaged)
 - Spatial (in this case, deterministic but spatial; could be distributions)

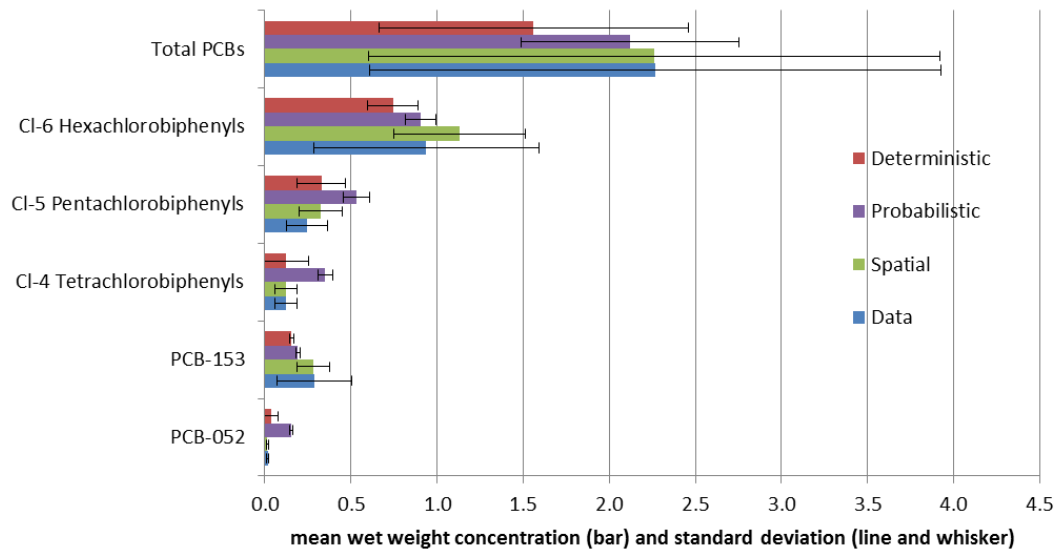
Input	Units	Distribution Type	Parameters	Source
Water column invertebrates lipid content	%	triangular (min, mode, max)	0.3, 1.0, 3	Assumed
Sediment invertebrates lipid	%	triangular (min, mode, max)	3.8, 1.1, 4.7	KAMM v1.0 documentation
Pumpkinseed				
Energy	g	triangular (min, mode, max)	34, 42, 47	FS Report, Table 3-1 2007 data
Lipid	%	triangular (min, mode, max)	1, 2, 2.9	FS Report, Table 3-1 2007 data
Bluegill				
Energy	g	triangular (min, mode, max)	70, 98, 135	ICF International 2008
Lipid	%	triangular (min, mode, max)	0.68, 0.61, 2.8	ICF International 2008
Yellow Perch				
Energy	g	triangular (min, mode, max)	50, 134, 175	ICF International 2008
Lipid	%	triangular (min, mode, max)	0.1, 0.43, 1.1	ICF International 2008
Largemouth Bass				
Weight	g	normal (μ, σ, min, max)	526 (186) 273-965	ICF International 2008
Lipid	%	normal (μ, σ, min, max)	0.83 (0.94) 0.05-4.43	ICF International 2008
Lag K _{oc}				
PCBs	kg/kg	normal (μ, σ, min, max)	6.6 (0.7) 4.1-9.1	(Elder and Bellie 1996; Hansen et al. 1999; Hawker & Connell 1988)
PCB52	kg/kg	point estimate	5.99	(Elder and Bellie 1996; Hansen et al. 1999; Hawker & Connell 1988)
PCB153	kg/kg	point estimate	7.66	(Elder and Bellie 1996; Hansen et al. 1999; Hawker & Connell 1988)
Homolog 4	kg/kg	triangular (min, mode, max)	5.45, 5.96, 6.43	(Elder and Bellie 1996; Hansen et al. 1999; Hawker & Connell 1988)
Homolog 5	kg/kg	triangular (min, mode, max)	5.72, 6.39, 7.32	(Elder and Bellie 1996; Hansen et al. 1999; Hawker & Connell 1988)
Homolog 6	kg/kg	triangular (min, mode, max)	6.24, 6.8, 7.31	(Elder and Bellie 1996; Hansen et al. 1999; Hawker & Connell 1988)
Total Organic Carbon	%	normal (μ, σ, min, max)	1.7 (0.07) 0.71-3.8	ICF International 2008
Temperature	deg C	triangular (min, mode, max)	8, 13, 18	Assumed (Average based on FS Table 3-1)
Water Concentration (ng/L)				
PCBs	ng/L	point estimate	1.0	nominal value (FS Table 3-1 states "1")
PCB52	ng/L	point estimate	0.03	nominal value (FS Table 3-1 states "0.03")
PCB153	ng/L	point estimate	0.03	nominal value (FS Table 3-1 states "0.03")
Homolog 4	ng/L	point estimate	0.03	nominal value (FS Table 3-1 states "0.03")
Homolog 5	ng/L	point estimate	0.03	nominal value (FS Table 3-1 states "0.03")
Homolog 6	ng/L	point estimate	0.03	nominal value (FS Table 3-1 states "0.03")



Largemouth Bass (mg/kg ww)



Yellow Perch (mg/kg ww)



Species	Observed Mean (mg/kg ww)	Deterministic Case (mg/kg ww)	Probabilistic (No Spatial) (mg/kg ww)	Spatially-Explicit Model Results (mg/kg ww)	RPD Deterministic	RPD Probabilistic	RPD Spatially-Explicit
Yellow Perch							
PCB-052	0.016	0.037	0.155	0.015	78%	162%	-7%
PCB-153	0.289	0.156	0.191	0.281	-60%	-41%	-3%
Cl-4 Tetrachlorobiphenyls	0.124	0.122	0.351	0.122	-1%	96%	-1%
Cl-5 Pentachlorobiphenyls	0.247	0.328	0.53	0.323	28%	73%	27%
Cl-6 Hexachlorobiphenyls	0.938	0.745	0.903	1.130	-23%	-4%	19%
Total PCBs	2.266	1.56	2.12	2.260	-37%	-7%	0%
Bluegill							
PCB-052	0.008	0.016	0.079	0.006	69%	164%	-21%
PCB-153	0.072	0.049	0.043	0.086	-39%	-51%	17%
Cl-4 Tetrachlorobiphenyls	0.045	0.05	0.171	0.049	10%	116%	8%
Cl-5 Pentachlorobiphenyls	0.087	0.111	0.192	0.108	24%	75%	22%
Cl-6 Hexachlorobiphenyls	0.240	0.239	0.283	0.362	0%	16%	41%
Total PCBs	0.582	0.623	0.79	0.848	7%	30%	37%
Largemouth Bass							
PCB-052	0.023	0.054	0.231	0.021	81%	164%	-8%
PCB-153	0.348	0.161	0.198	0.276	-74%	-55%	-23%
Cl-4 Tetrachlorobiphenyls	0.149	0.175	0.502	0.146	16%	109%	-2%
Cl-5 Pentachlorobiphenyls	0.321	0.379	0.622	0.331	16%	64%	3%
Cl-6 Hexachlorobiphenyls	1.154	0.81	1.000	1.160	-35%	-14%	1%
Total PCBs	2.767	1.76	2.420	2.410	-44%	-13%	-14%
RPD = relative percent difference calculated as (predicted-observed)/average(predicted,observed)							
green values indicate lowest RPD; blue values indicate within 50% of observed							

Example Application and Linkage to Decision Analytic Framework

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Decision Analytic Strategies for Integrating Ecosystem Services and Risk Assessment

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(Submitted 22 June 2012; Returned for Revision 7 August 2012; Accepted 28 December 2012)

EDITOR'S NOTE

This paper is one of 8 articles generated from the SETAC Special Symposium: Ecosystem Services, from Policy to Practice (15-16 February 2012, Brussels, Belgium). The symposium aimed to give a broad overview of the application of the ecosystem services concept in environmental assessment and management, against the background of the implementation of the European environmental policies such as the biodiversity agenda, agricultural policy, and the water framework directive.

Special Series

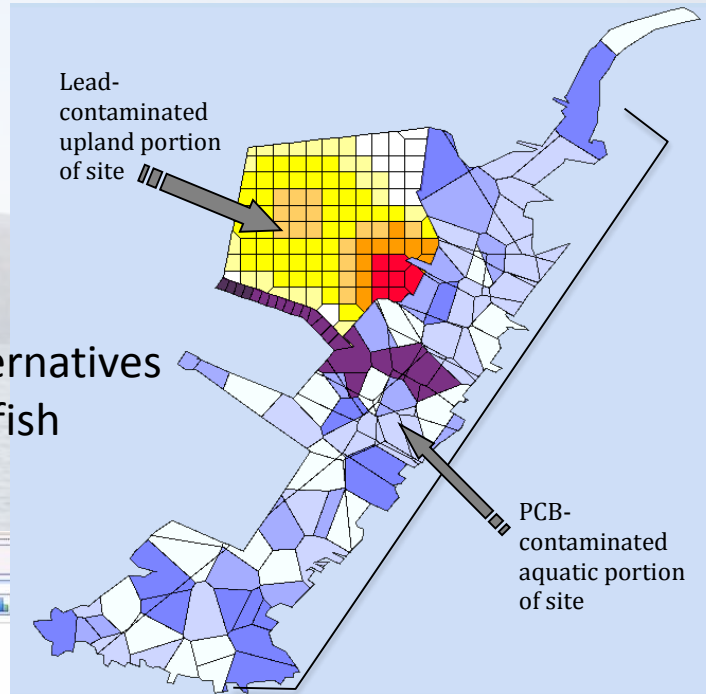
ABSTRACT

Ecosystem services as a concept and guiding principle are enjoying wide popularity and endorsement from high-level policy thinkers to industry as support for sustainability goals continue to grow. However, explicit incorporation of ecosystem services into decision making still lacks practical implementation at more local scales and faces significant regulatory and technical constraints. Risk assessment represents an example of a regulatory process for which guidance exists that makes it challenging to incorporate ecosystem service endpoints. Technical constraints exist in the quantification of the relationships between ecological functions and services and endpoints valued by humans, and the complexity of those interactions with respect to bundling and stacking. In addition, ecosystem services, by their very definition, represent an anthropogenic construct with no inherent ecological value, which, in practical terms, requires a far more inclusionary decision making process explicitly incorporating a greater diversity of stakeholder values. Despite these constraints, it is possible, given a commitment to sustainable decision making, to simplify the process based on strategic outcomes (e.g., identifying desired end-states in general terms). Decision analytic techniques provide a mechanism for evaluating tradeoffs across key ecosystem services valued by stakeholders and to develop criteria drawn from the entire spectrum of stakeholders in evaluating potential alternatives. This article highlights several examples of ways in which ecosystem service endpoints can be incorporated into the decision-making process. *Integr Environ Assess Manag* 2013;9:260–268. © 2013 SETAC

Keywords: Ecosystem services Decision analysis MCDA Influence diagrams Sustainability

Evaluation of Remedial Alternatives

Criteria against which alternatives are evaluated (predicted fish tissue concentrations → probabilistic risk results)



TT Model v2

MAVT

Evaluate Remedial Alternatives Dw

New criterion

Criterion name

Attributes

- prcnt_tot
- rpd
- PCB upland
- area
- prcnt_tot
- pCBS
- cancer
- child_hi
- adult_hi
- Lead upland
- area
- prcnt_tot

Function type for criterion value calculation

Average

OK Cancel

C1 (Blood lead) VF

C2 (Cancer) VF

C3 (Child HI) VF

C4 (EcoRisk) VF

C5 (Habitat) VF

C6 (Adult HI) VF

C7 (Cost) VF

C8 (Community acceptability) VF

A0 (No Action)

A1 (Remediate the upland s...)

A2 (Remediate sediments on...)

A3 (Remediate upland soils...)

A4 (Institutional control)

Remedial alternatives to be evaluated

Output Statistics from FR Link Directly to Risk Assessment Model to Criteria

Statistics included in the report:

Central Tendency	Variability
<input type="checkbox"/> Mean	<input type="checkbox"/> Stdev
<input type="checkbox"/> Median	<input type="checkbox"/> Variance
<input type="checkbox"/> Mode	<input type="checkbox"/> CV
<input type="checkbox"/> Geom. Mean	<input type="checkbox"/> Geom. Stdev

Group statistics by time intervals:

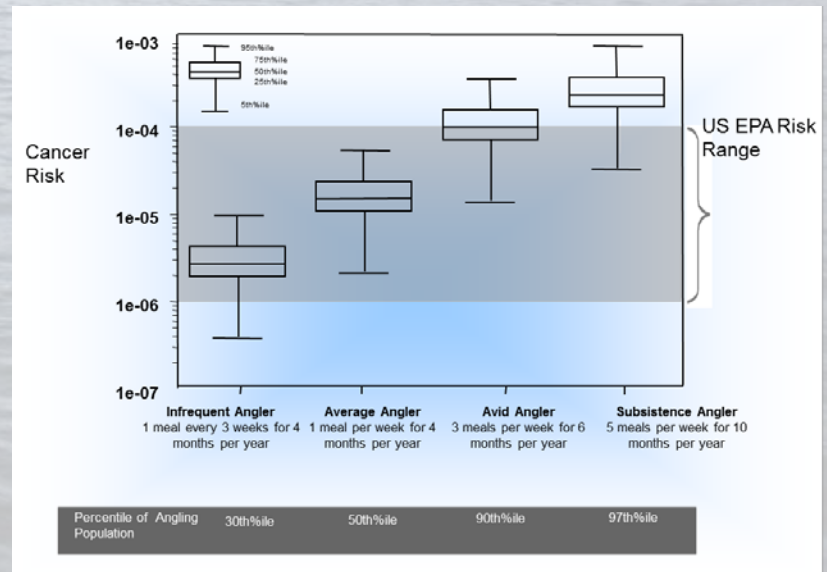
#	PERIOD BEGINNING	PERIOD ENDING
1		
2		
3		
4		
5		
6		
7		

Full tables:

- Include Percentiles of Wet Weight Concentration for Food Items
- Include Percentiles of Lipid Normalized Concentration for Food Items
- Include Percentiles of Wet Weight Concentration for Fish
- Include Percentiles of Lipid Normalized Concentration for Fish
- Report Inputs?

Model can generate different statistics of predicted tissue concentrations for use in risk assessment model

In this example, links directly to risk model within a GIS-based decision analytic framework



Conclusions

- Spatially-explicit bioaccumulation model provides greater realism in how sediment exposures influence predicted aquatic organism concentrations
- Probabilistic framework provides more information for decision makers and integration with complementary analyses
- Better evaluation of impact of management alternatives
 - Remove all “hot spots”
 - Remove all sediments above some threshold – truncate distributions
- Decision analytic approaches allow transparent evaluation of alternatives

Thank You!!

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1.508.596.4209

Acknowledgments

FishRand was developed with funding from the Army Corps of Engineers. Currently, the model is available by request from Dr. von Stackelberg; plans are underway to make it publicly available via web download.