



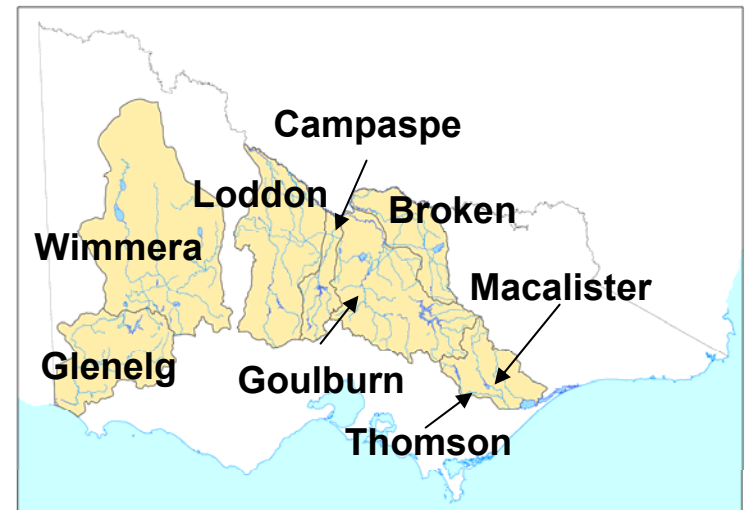
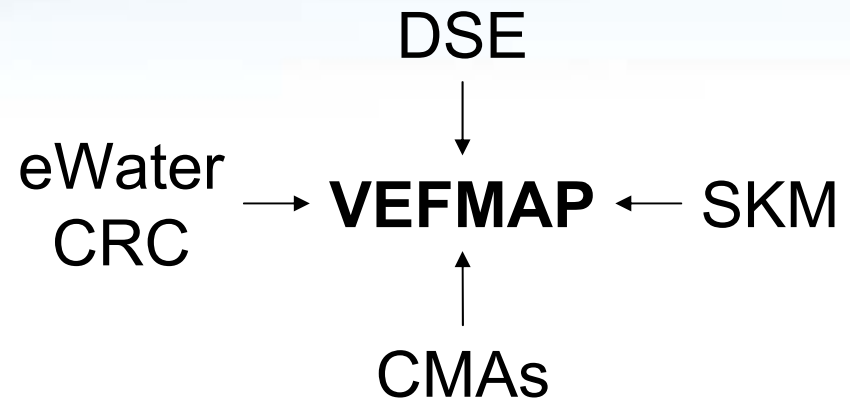
# Monitoring and Assessing the effects of Environmental Flows: the VEFMAP approach

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enterprise environment education

# Victorian Environmental Flows Monitoring and Assessment Program (VEFMAP)

- Partnership between managers and scientists
- Major Aims
  - Maximizing ability to detect ecosystem-level responses to environmental flows
    - Get creative
  - Doing so with the highest level of scientific integrity
    - Peer review
    - Publication



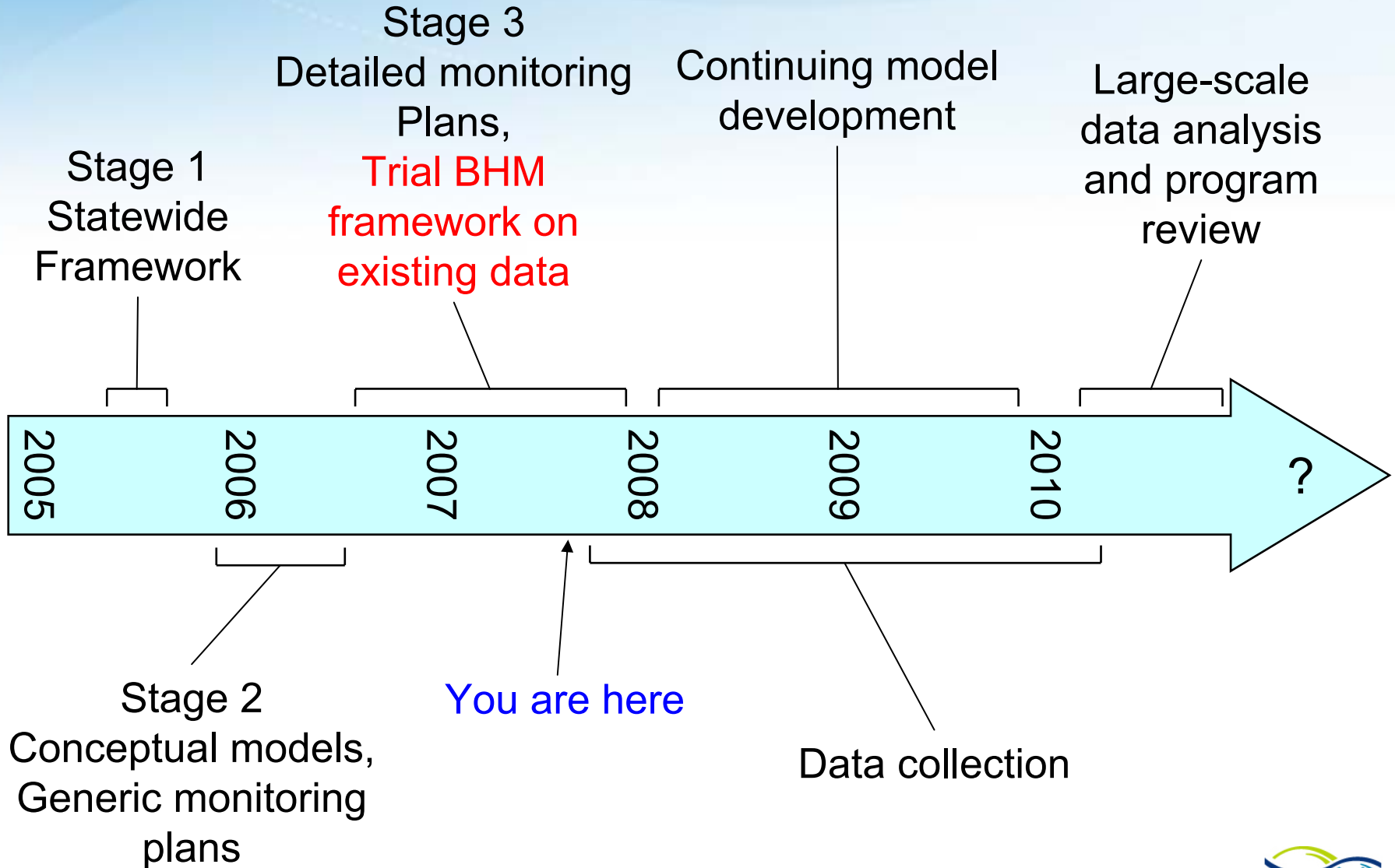
# The Approach

- Development of compatible monitoring programs across the state
  - Hypothesis based (conceptual models)
  - Prioritise monitoring effort based on
    - Conceptual understanding
    - Expected ability to detect a response
- Statewide analyses using Bayesian hierarchical models, and where possible combining data from different rivers

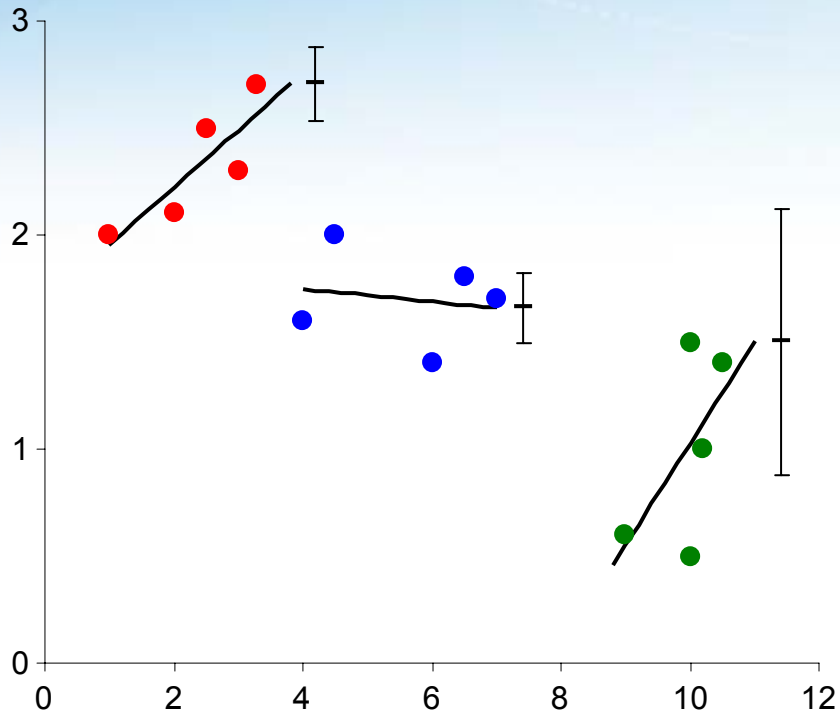
# Why Bayesian Hierarchical Modelling?

- BACI methods generally not applicable
  - Shortage of Control and Reference Sites
  - eFlows do not have a Before/After boundary
- Data are messy
- Often sparse
- In general, don't conform to requirements of familiar frequentist analyses
- Bayesian Hierarchical Modelling (BHM)
  - More flexibility with models
  - Better ability to combine data in analyses, strengthening conclusions
- This has caused some unease
  - Trialling framework in 2007

# Implementation



# Why BHM?



Site 1:  $\beta = 0.27 \pm 0.17$

Site 2:  $\beta = -0.03 \pm 0.17$

Site 3:  $\beta = 0.48 \pm 0.62$

- Interested in regression slope ( $\beta$ ) at the site level, which should be similar
  - But few data and much unexplained variability at each site
  - Site-level estimates vary widely and can be very uncertain

# Bayesian Statistics – Bayes' Formula

Posterior Probability:  
Probability of model,  
given the data

Likelihood function:  
driven by data

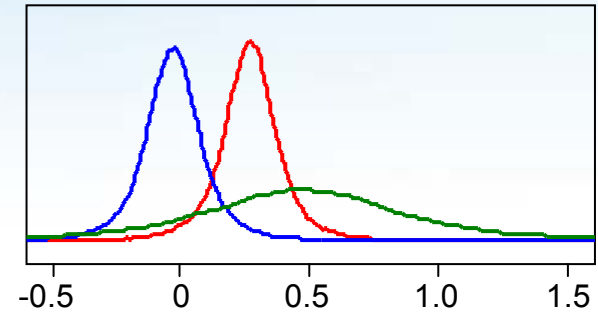
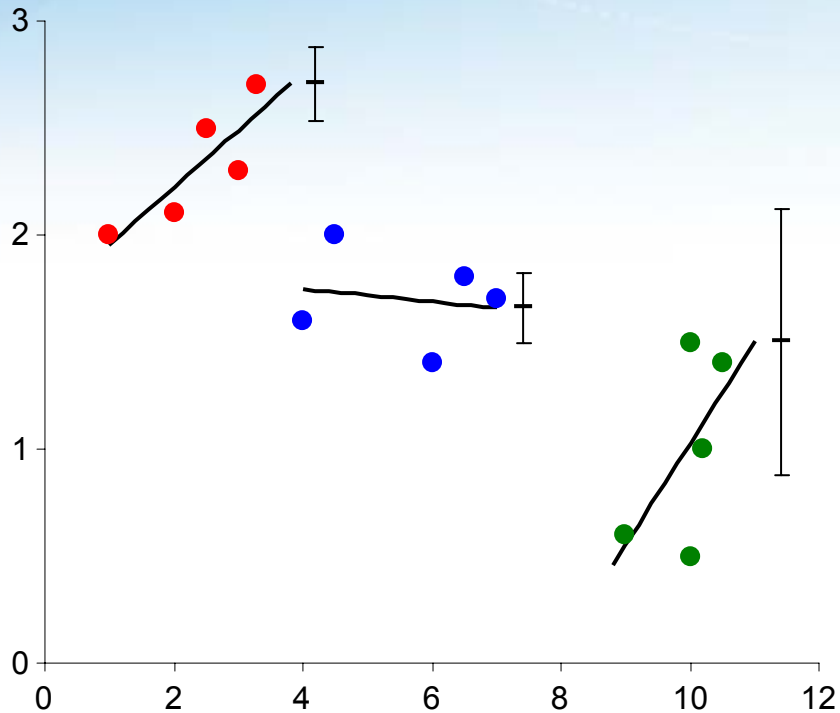
Prior Probability:  
Level of belief in  
the model before  
data collection

$$P(H | X) = \frac{P(X | H)P(H)}{P(X)}$$

Total probability of  
the data occurring



# Non-informative prior distribution



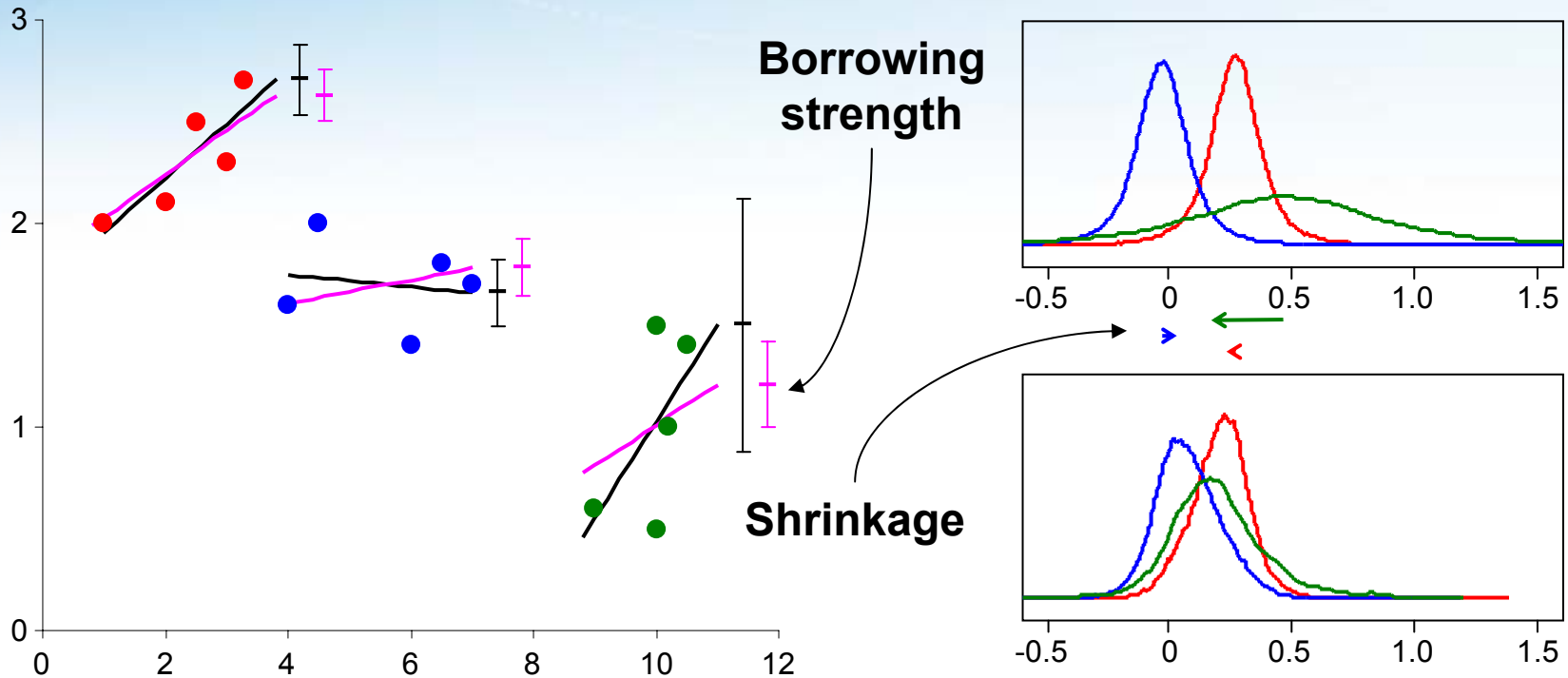
Site 1:  $\beta = 0.27 \pm 0.17$

Site 2:  $\beta = -0.03 \pm 0.17$

Site 3:  $\beta = 0.48 \pm 0.62$



# Non-informative prior distribution vs. hierarchical priors



Site 1:  $\beta = 0.27 \pm 0.17 \rightarrow 0.21 \pm 0.12$

Site 2:  $\beta = -0.03 \pm 0.17 \rightarrow 0.08 \pm 0.14$

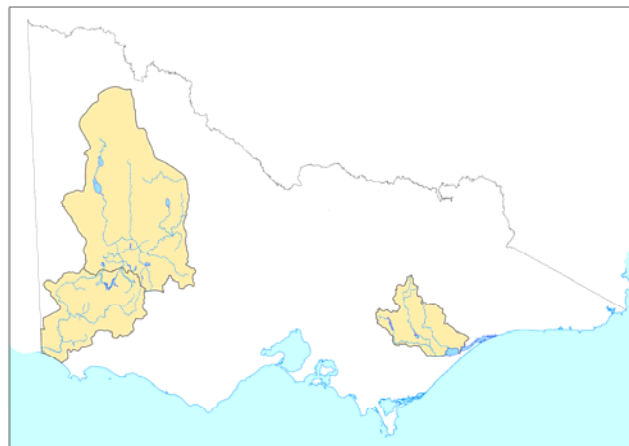
Site 3:  $\beta = 0.48 \pm 0.62 \rightarrow 0.20 \pm 0.21$

No prior data used, just the expectation that the sites are related

- Sites with less data / greater uncertainty more affected
- Results for data-rich sites will be practically unaffected

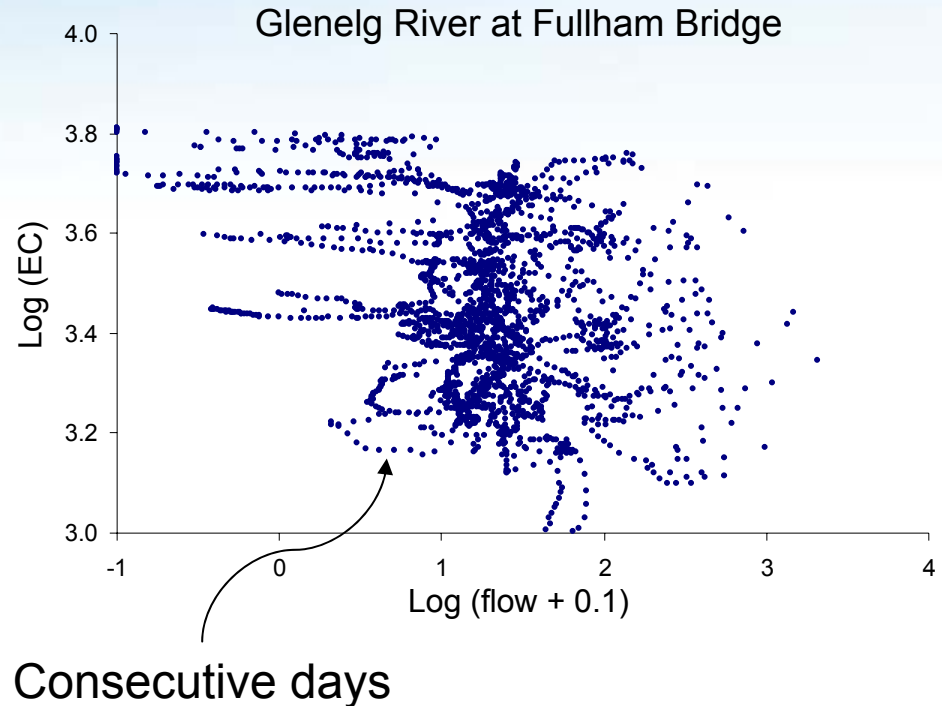
# Testing the analytical framework

- Need to demonstrate efficacy of BHM before large scale analysis of data in 2010
  - Possible analyses driven by data availability
  - Not answering eflows questions of primary interest
- Models applied to existing data
  - Salinity (EC) in Glenelg and Wimmera rivers
  - Fish (Australian Smelt) in the Thomson River



# Effect of Flow on Salinity

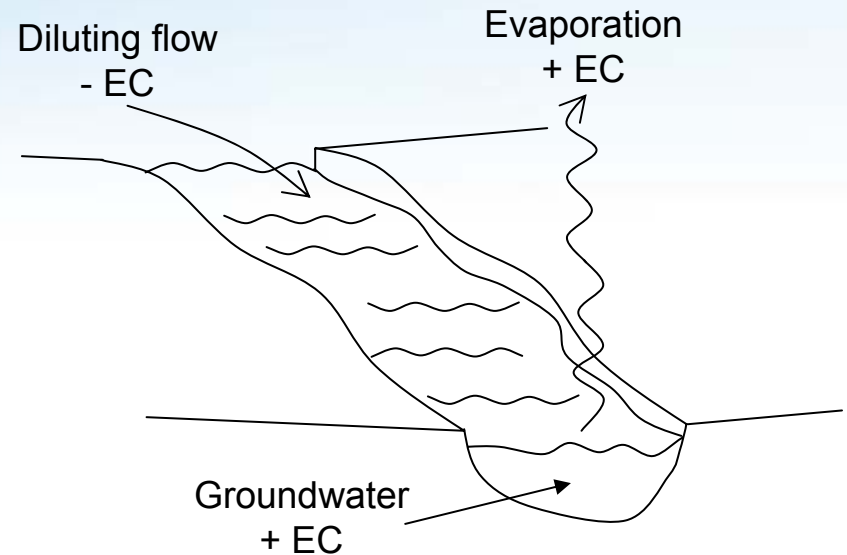
- What is the relationship between flow and EC?
  - Pretty poor!
- Lots of data, but highly autocorrelated
  - Model needs to take advantage of this



salinity = previous salinity + flow effects + non-flow effects

# Conceptual model → statistical model

- Hypothesise
  - ‘Background’ rate of salinity increase
  - Rate of salinity decrease proportional to flow
    - Scale according to summer low flow recommendation, which often aim to ‘maintain’ or ‘improve’ water quality




$$EC_i = EC_{i-1} - k_1 \frac{Q_i}{Q_R} + k_2$$

- Parameter of main interest
  - $p(k_1 > 0)$  “flow reduces EC”

# Implementation and Results

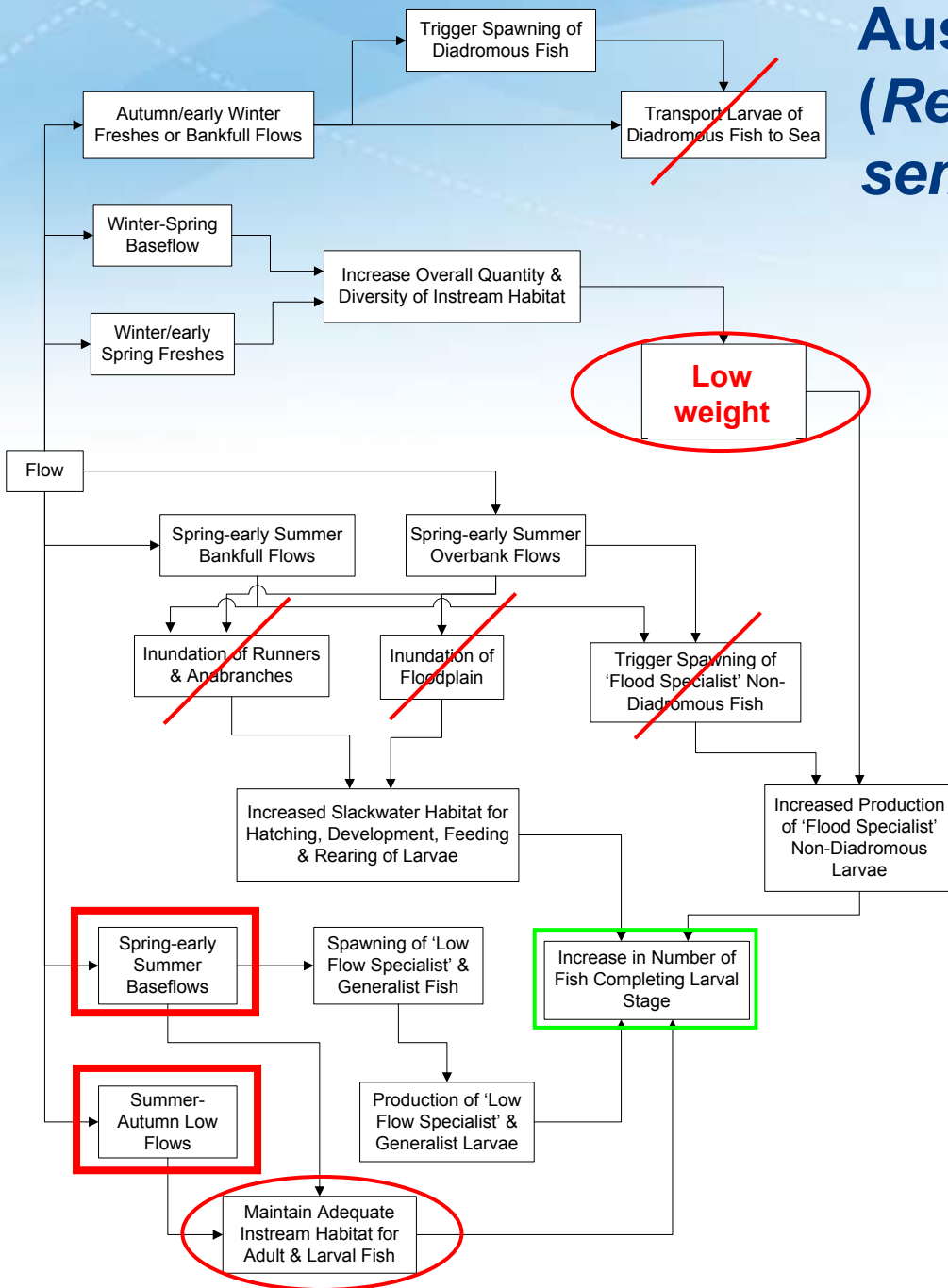
- 8938 ‘summer’ EC measures, 10 sites, 2 rivers
- Modelled at site, river and multi-river hierarchies

More hierarchical 

River	Site Name	Sites $p(k_1 > 0)$	Sites(Rivers) $p(k_1 > 0)$	Sites(Rivers(State)) $p(k_1 > 0)$
Glenelg	Fulham Bridge	1.00	1.00	1.00
	Harrow	0.82	0.82	0.82
	Burkes Bridge	0.99	0.98	0.98
	Dergholm	0.67	0.64	0.64
	Sandford	0.58	0.57	0.57
	Dartmoor	0.00	0.00	0.00
Wimmera	Walmer	0.10	0.10	0.11
	U/S Dimboola	0.20	0.24	0.24
	Lochiel Railway Br.	1.00	1.00	1.00
	Tarranyurk	1.00	1.00	1.00

- Sensible results for Glenelg (except Dartmoor)
- Wimmera results harder to interpret
- Possibility of model inadequacies (e.g. saline fronts increasing EC)

# Australian Smelt (*Retropinna semoni*)



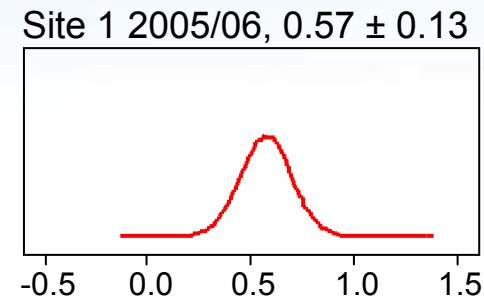
- Non-diadromous
- Non 'flood-specialist'
- Little floodplain, runner or anabranch habitat in the Thomson
- Pre-spawning condition of adults probably has little effect
- Concentrate on summer low flows and the slow-flow habitat for young fish

# Characterising flow for habitat

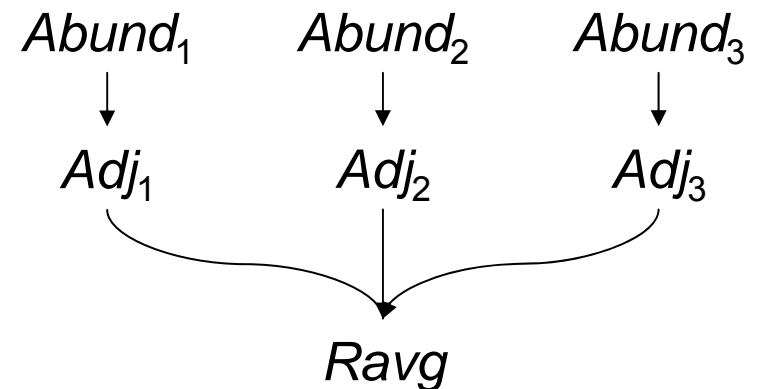
- VEFMAP monitoring will measure slow flow habitat
  - But we don't have that yet
- Summer flows in the Thomson generally exceed recommendations
  - Highest: Reach 4a (2005-2006)
    - 149 ML d<sup>-1</sup> ( $Q_R = 20 \text{ ML d}^{-1}$ )
- May expect this to negatively affect fish that need slow flow habitat for larvae and juveniles
- Characterise summer flow in terms of average proportion of recommendation

# Analysis

- Worries over fish data
  - Turbidity and flow on day of sampling reduce sampling efficiency
    - Include these effects in the model as covariates
    - Leads to uncertain 'data'



- Analyses conducted at reach scale
  - Flow data availability
  - Fish expected to respond at this scale
- Parameters of main interest
  - $\beta$  at reach level
  - $p(\beta < 0)$  – negative effect of high flow



$$Ravg = \alpha + \beta \cdot \frac{Q}{Q_R}$$

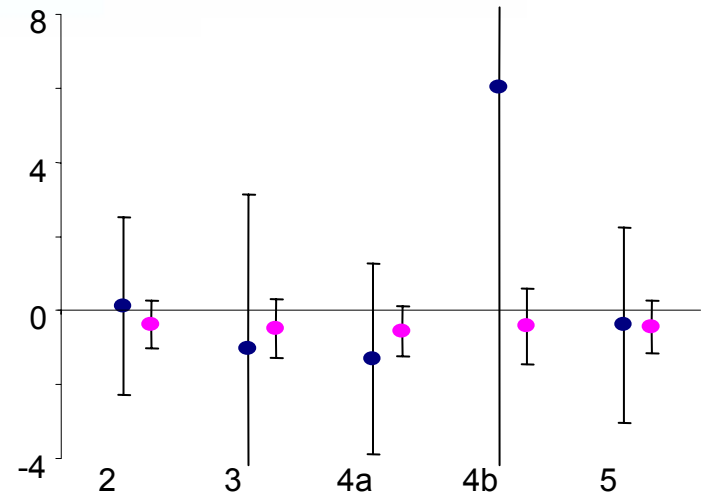


# Implementation and Results

- 44 site-level estimates (5 reaches, 3 years)
- Modelled at reach and river levels

More hierarchical

Reach Location		Reach $p(\beta < 0)$	River $p(\beta < 0)$
2	Thomson Dam - Aberfeldy R.	0.44	0.73
3	Aberfeldy R. - Cowwarr Weir	0.67	0.77
4a	Old Thomson River	<b>0.88</b>	<b>0.82</b>
4b	Rainbow Creek	0.21	0.74
5	Rainbow Creek - Macalister R.	0.60	0.76



- Largest effect in most flow-affected reach (4a)
- Weak positive effect in reach 4b (where  $Q < Q_R$ )
  - HM obscures this effect – model inadequacy (we should regress against habitat rather than flow)
- HM shows weak evidence of river-scale effects of high summer flows on abundance of Australian Smelt.

# Synthesis

- Bayesian framework allows analyses not possible with frequentist techniques
  - Autocorrelation model for EC
  - Within analysis adjustment for Tu and dFlow for Smelt
- Different effects of hierarchical modelling
  - Driven by data availability
  - Very helpful for data-poor analyses
- Results highlight possible inadequacies in model structure
  - Continue development
  - But a very promising start

# Where to now?

- Publish
- Complete implementation of monitoring programs
- Pray for rain
- Undertake further development work on model structures during 2008-09
- Major analysis of data 2010
- Review of program
  - Implementing changes based on lessons learned
    - Flow recommendations
    - Monitoring programs

# Conclusions

- VEFMAP's cooperative approach is helping to establish compatible monitoring programs
- Bayesian Hierarchical Modelling shows promise in identifying the effects of flow on ecosystem response

# Acknowledgements

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