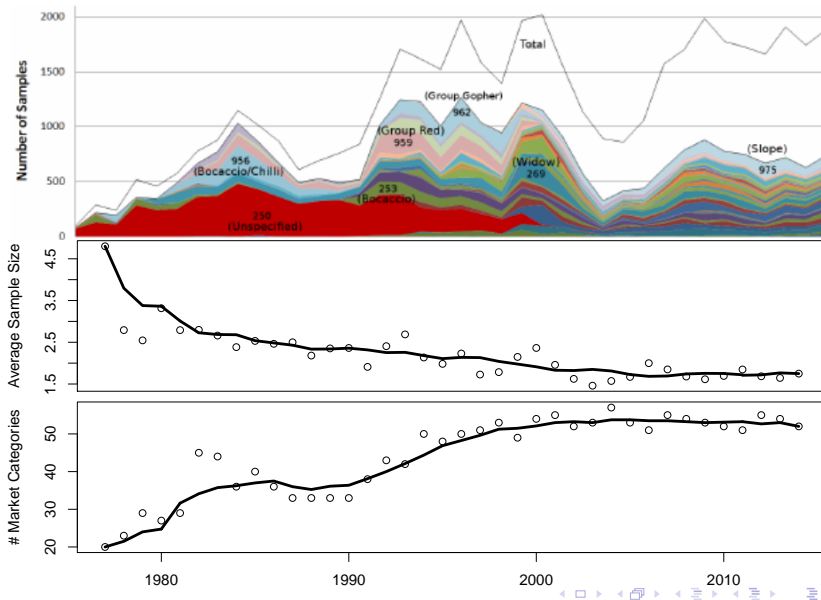


Improving Catch Estimation Methods in Sparsely Sampled, Mixed Stock Fisheries.

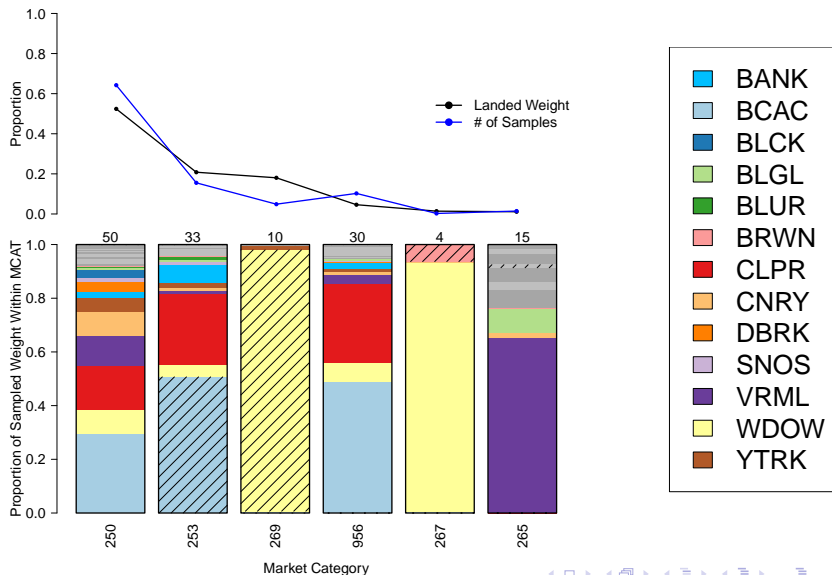
Nick Grunloh

UCSC :: CSTAR :: SWFSC :: NMFS

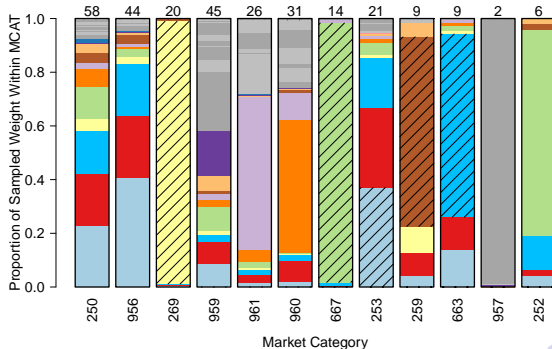
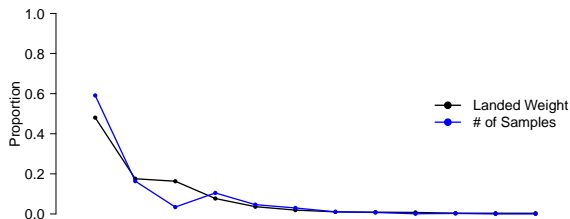
28 March 2018



1978–1982



1983–1990



- BANK
- BCAC
- BLCK
- BLGL
- BLUR
- BRWN
- CLPR
- CNRY
- DBRK
- SNOS
- VRML
- WDW
- YTRK

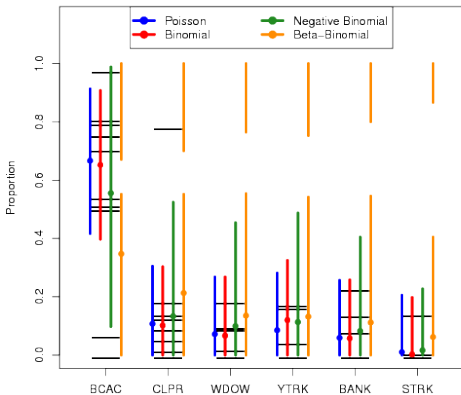
sparse data - $\hat{\gamma}$ Pooling and hierarchical models
integer overdispersion (Motivate next slide)

Likelihood

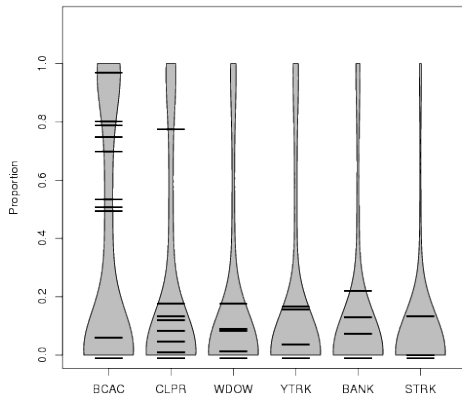
y_{ij} : i^{th} sample of the j^{th} species' integer weight from market category 250, in the Monterey port complex trawl fishery for the second quarter of 1982.

$$y_{ij} \sim \text{Pois}(\theta_j) \quad y_{ij} \sim \text{Bin}(\theta_j) \quad y_{ij} \sim \text{NB}(\theta_j, \phi) \quad y_{ij} \sim \text{BB}(\theta_j, \phi)$$

95% Predictive HDI Model Comparison



Beta-Binomial Posterior Predictive Species Compositions



	Poisson	Binomial	NB	BB
MSE	0.06412	0.06264	0.05171	0.04479
Δ DIC	1001.41	1230.60	5.03	0
Δ WAIC	1079.95	1323.75	3.43	0
$pr(M y)$	≈ 0	≈ 0	$\approx 10^{-7}$	$\approx 1 - 10^{-7}$

Beta-Binomial Model

$$y_{ijklm\eta} \sim \text{Beta-Binomial}(\mu_{ijklm\eta}, \sigma_{ijklm\eta}^2)$$

$$\mu_{ijklm\eta} = n \text{ logit}^{-1}(\theta_{ijklm\eta})$$

$$\sigma_{ijklm\eta}^2 = \mu_{ijklm\eta} \left(1 - \frac{\mu_{ijklm\eta}}{n}\right) \left(1 + (n-1) \rho\right)$$

$$\theta_{ijklm\eta} = \beta_0 + \beta_j^{(s)} + \beta_k^{(p)} + \beta_l^{(g)} + \beta_{m\eta}^{(t)}$$

$y_{ijklm\eta}$: i^{th} sample of the j^{th} species' integer weight, in the k^{th} port, caught with the l^{th} gear, in the η^{th} quarter, of year m , for a particular market category.

$j \in \{1, \dots, J\}$ Species

$k \in \{1, \dots, K\}$ Ports

$l \in \{1, \dots, L\}$ Gears

$m \in \{1, \dots, M\}$ Years

$\eta \in \{1, \dots, H\}$ Quarters

Time Model

(M1)

$$\beta_{m\eta}^{(t)} = \beta_m^{(y)} + \beta_{\eta}^{(q)}$$

$$\beta_m^{(y)} \sim N(0, 32^2)$$

$$\beta_{\eta}^{(q)} \sim N(0, 32^2)$$

(M2)

$$\beta_{m\eta}^{(t)} = \beta_m^{(y)} + \beta_{\eta}^{(q)}$$

$$\beta_m^{(y)} \sim N(0, v^{(y)})$$

$$\beta_{\eta}^{(q)} \sim N(0, v^{(q)})$$

(M3)

$$\beta_{m\eta}^{(t)} = \beta_m^{(y)} + \beta_{\eta}^{(q)} + \beta_{m\eta}^{(y:q)}$$

$$\beta_m^{(y)} \sim N(0, v^{(y)})$$

$$\beta_{\eta}^{(q)} \sim N(0, v^{(q)})$$

$$\beta_{m\eta}^{(y:q)} \sim N(0, v)$$

(M4)

$$\beta_{m\eta}^{(t)} = \beta_{m\eta}^{(y:q)}$$

$$\beta_{m\eta}^{(y:q)} \sim N(0, v)$$

(M5)

$$\beta_{m\eta}^{(t)} = \beta_{m\eta}^{(y:q)}$$

$$\beta_{m\eta}^{(y:q)} \sim N(0, v_{\eta})$$

(M6)

$$\beta_{m\eta}^{(t)} = \beta_{m\eta}^{(y:q)}$$

$$\beta_{m\eta}^{(y:q)} \sim N(0, v_m)$$

Priors

$$\beta_0 \propto 1$$

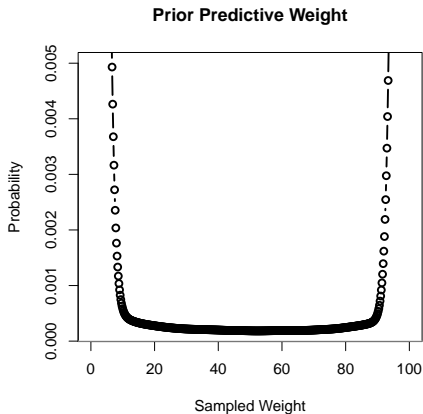
$$\beta_j^{(s)} \sim N(0, 32^2)$$

$$\beta_k^{(p)} \sim N(0, 32^2)$$

$$\beta_l^{(g)} \sim N(0, 32^2)$$

$$\text{logit}(\rho) \sim N(0, 2^2)$$

$$\nu \sim IG(1, 2 \times 10^3) \quad \forall \quad \nu$$



1978-1982

	M1	M2	M3	M4	M5	M6
MSE	0.12725	0.12704	0.12680	0.12237	0.12724	0.12657
Δ DIC	2558.56	2259.94	2013.21	0	2175.32	2174.71
Δ WAIC	2562.65	2263.58	2009.32	0	2171.18	2170.56
$pr(M y)$	≈ 0	≈ 0	≈ 0	≈ 1	≈ 0	≈ 0

1983-1990*

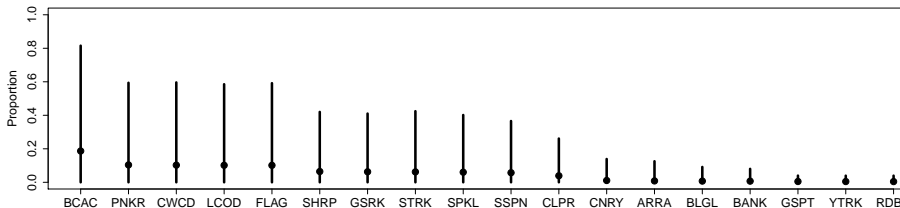
	M1	M2	M3	M4	M5	M6
MSE	0.12724	0.12704	0.12680	0.12237	0.12723	0.12657
Δ DIC	2558.56	2259.94	2013.21	0	2175.32	2174.71
Δ WAIC	2562.65	2263.58	2009.32	0	2171.18	2170.56
$pr(M y)$	≈ 0	≈ 0	≈ 0	≈ 1	≈ 0	≈ 0

Posterior Predictive Weight

$$p(y_{jklm\eta}^* | \mathbf{y}) = \iint \text{BB}(y_{jklm\eta}^* | \mu_{jklm\eta}, \sigma_{jklm\eta}^2) P(\mu_{jklm\eta}, \sigma_{jklm\eta}^2 | \mathbf{y}) d\mu_{jklm\eta} d\sigma_{jklm\eta}^2$$

$$\pi_{jklm\eta}^* = \frac{y_{jklm\eta}^*}{\sum_j y_{jklm\eta}^*} \quad \mathbf{y}_{klm\eta}^* \neq \mathbf{0}$$

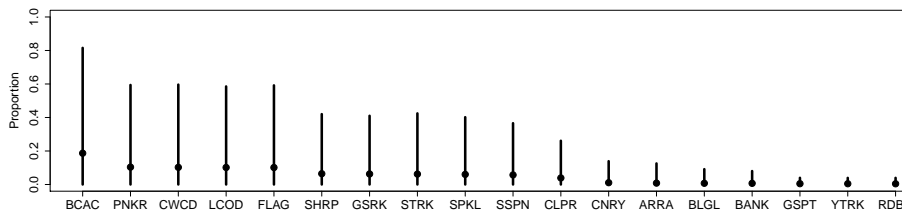
motivate prediction for filling holes/hindcasting
show a 100 pound BCAC distribution



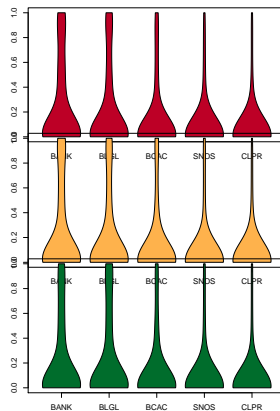
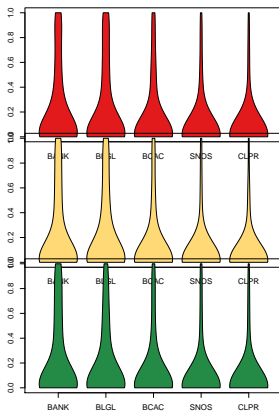
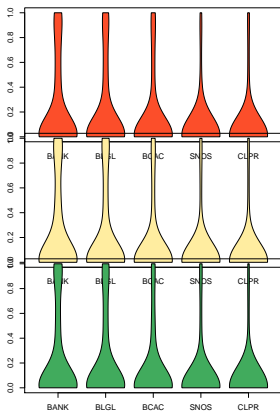
Species Composition

$$\pi_{jklm\eta}^* = \frac{y_{jklm\eta}^*}{\sum_j y_{jklm\eta}^*} \quad \mathbf{y}_{jklm\eta}^* \neq \mathbf{0}$$

show a BCAC species comp distribution

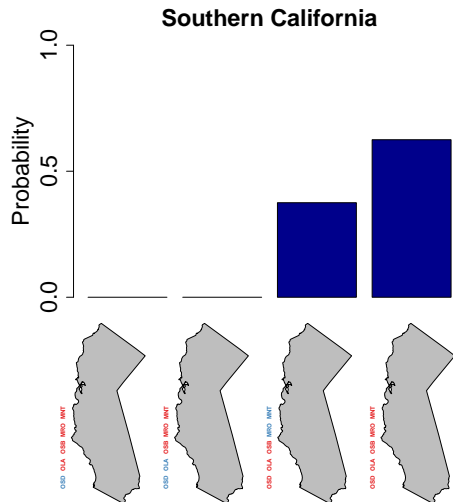
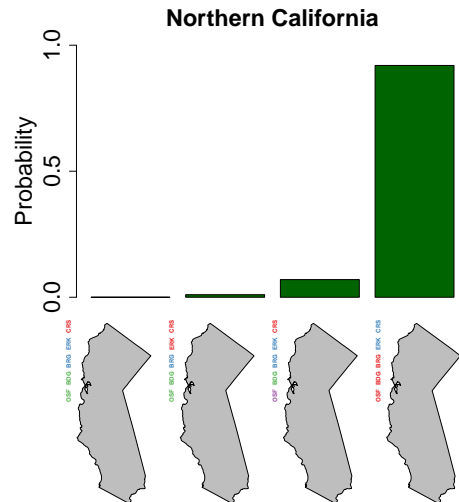


show sppComp distribution for some strata

HKL**NET****TWL**

Expansion

instructive example of port pooling w/ Bell number and constraints



Bayesian Model Averaging (BMA)

Consider a set of Models (M) indexed by ι :

$$\omega_{\iota} = Pr(M_{\iota}|y) = \frac{p(y|M_{\iota})p(M_{\iota})}{\sum_{\iota} p(y|M_{\iota})p(M_{\iota})}$$

$$\bar{p}(\theta|y) = \sum_{\iota} \omega_{\iota} p(\theta|y, M_{\iota})$$

if f only depends on M through θ , then

$$\bar{p}(y^*|y) = \int f(y^*|\theta) \bar{p}(\theta|y) d\theta$$

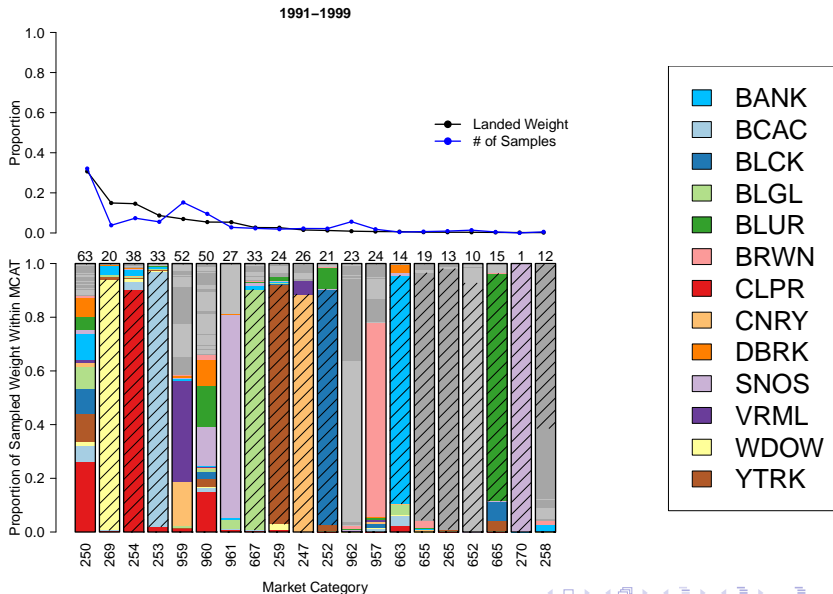
* Hoeting, J. A., Madigan, D., Raftery, A. E., and Volinsky, C. T. (1999). Bayesian model averaging: a tutorial.

Statistical science, 382-401.

MCAT 250										
ω	0.32	0.14	0.13	0.12	0.02	0.02	0.02	0.02	0.02	0.02
CRS										
ERK										
BRG										
BDG										
OSF										
MNT										
MRO										
OSB										
OLA										
OSD										

select port pooling results

- Red stuff
- Species Composition Proof



2000–2015

