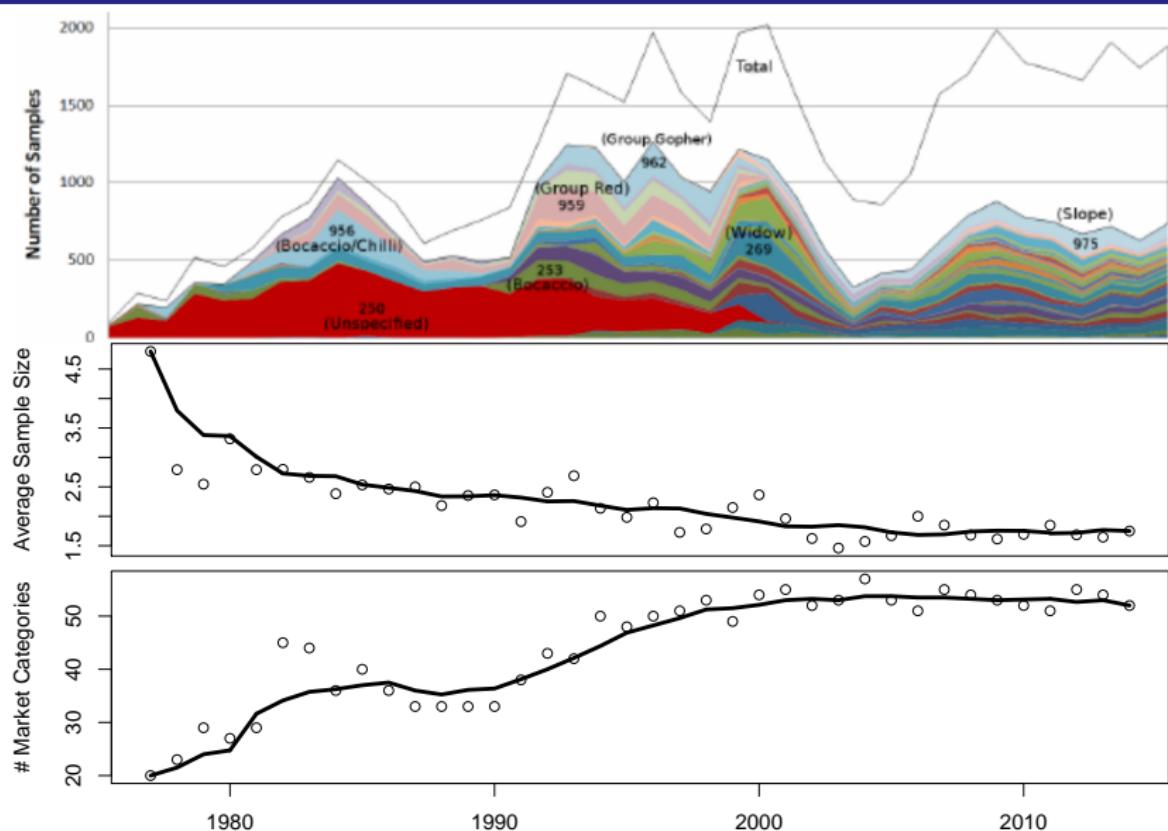


Improving Catch Estimation Methods in Sparsely Sampled, Mixed Stock Fisheries.

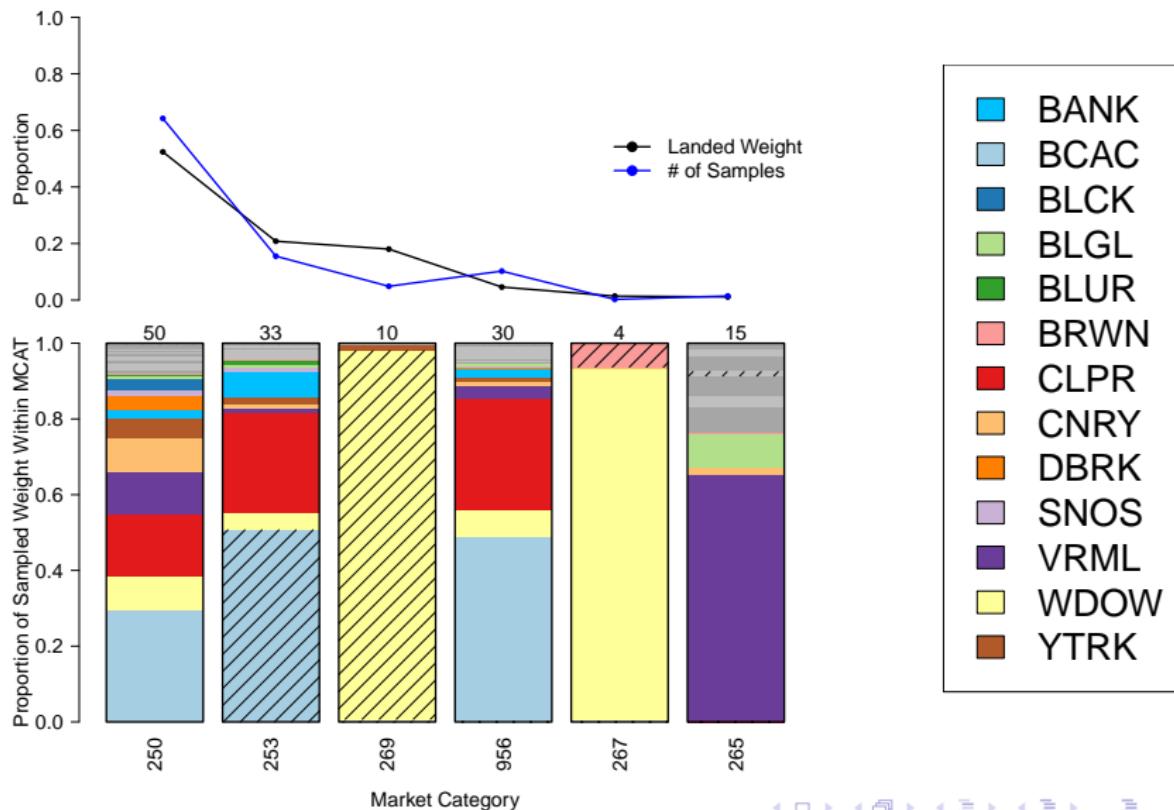
Nick Grunloh

UCSC :: CSTAR :: SWFSC :: NMFS

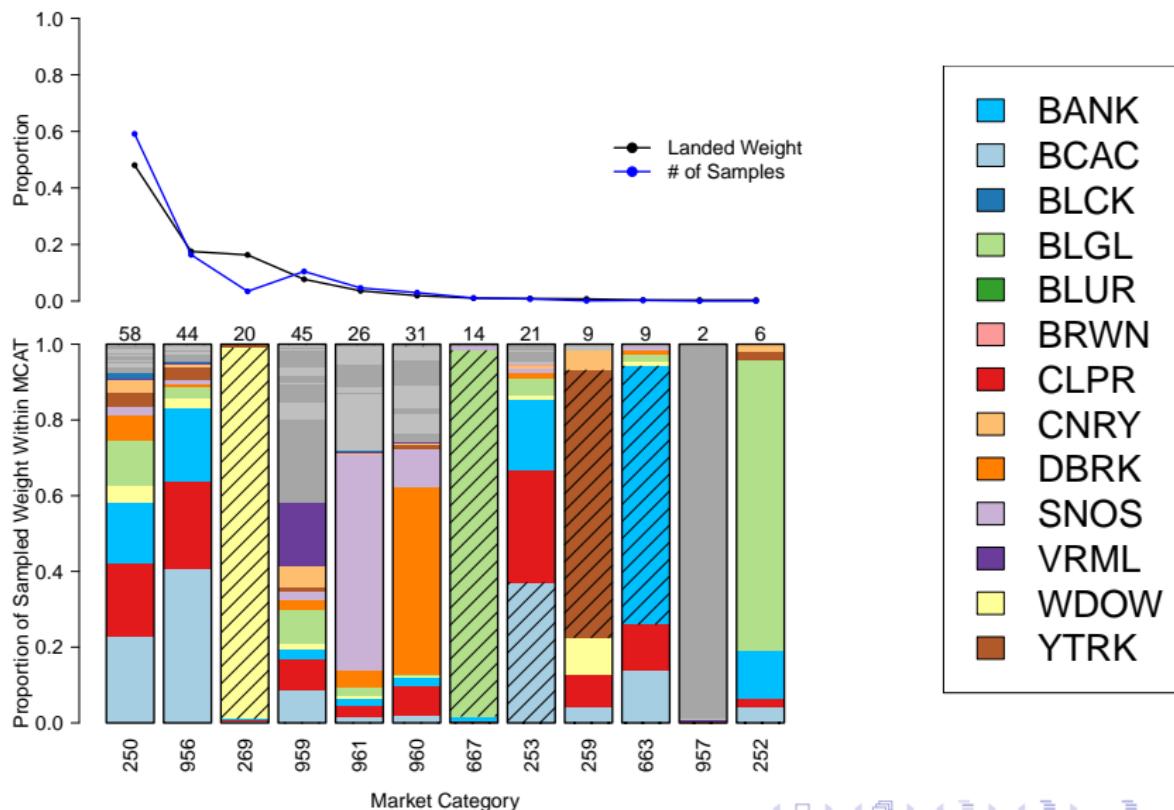
28 March 2018



1978–1982



1983–1990



sparse data - \downarrow Pooling and heirarchical 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.

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

Introduction
oooooo

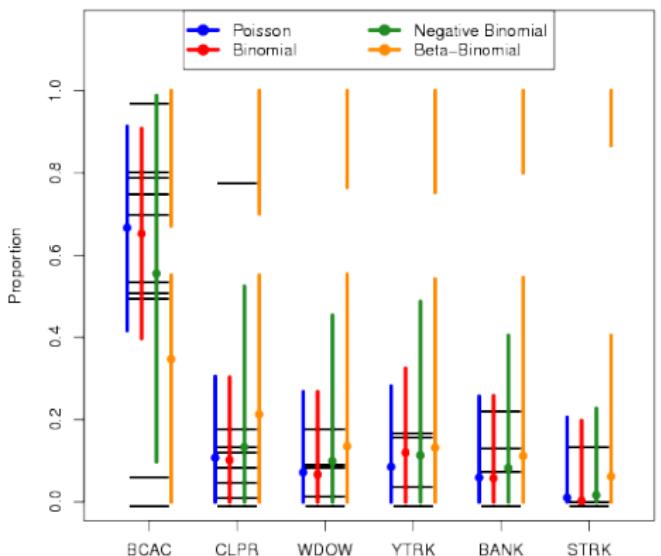
Modeling
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Prediction
ooo

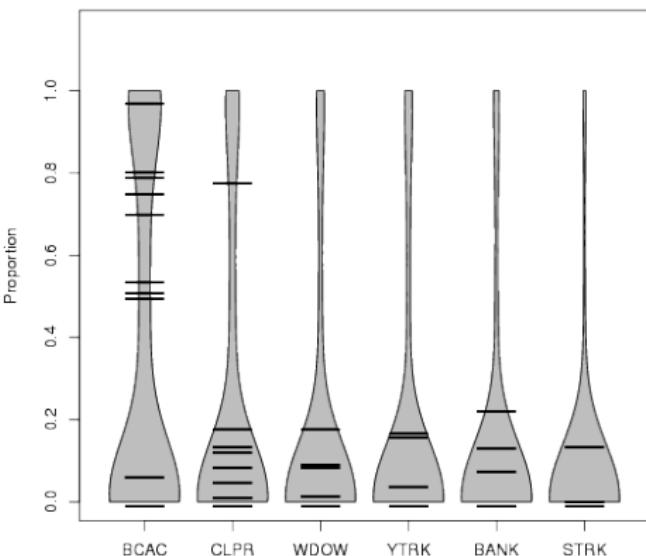
BMA
ooooooo

Extra
oo

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

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

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

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

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

$y_{ijklm\eta}$: 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

$$\beta_0 \propto 1$$

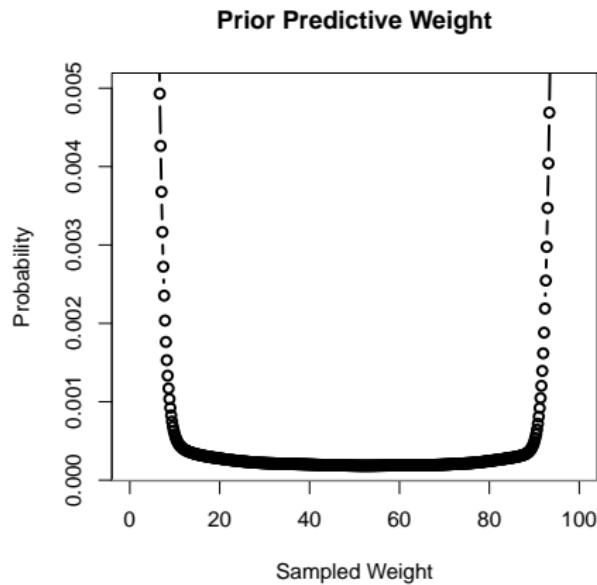
$$\beta_j^{(s)} \sim N(0, 32^2)$$

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

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

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

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



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 Distributions

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

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

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

show sppComp distribution for some strata

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BMA
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Extra
oo

Expansion

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Modeling
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BMA
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Extra
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instructive example of port pooling w/ Bell number and constraints

Introduction
ooooo

Modeling
oooooooo

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northern269.pdf

southern269.pdf

Nick Grunloh

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

Bayesian Model Averaging (BMA)

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

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

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

if f only depends on M through θ , then

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

* 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 | | | | | | | | | | |

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BMA
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select port pooling results

- Red stuff
- Species Composition Proof

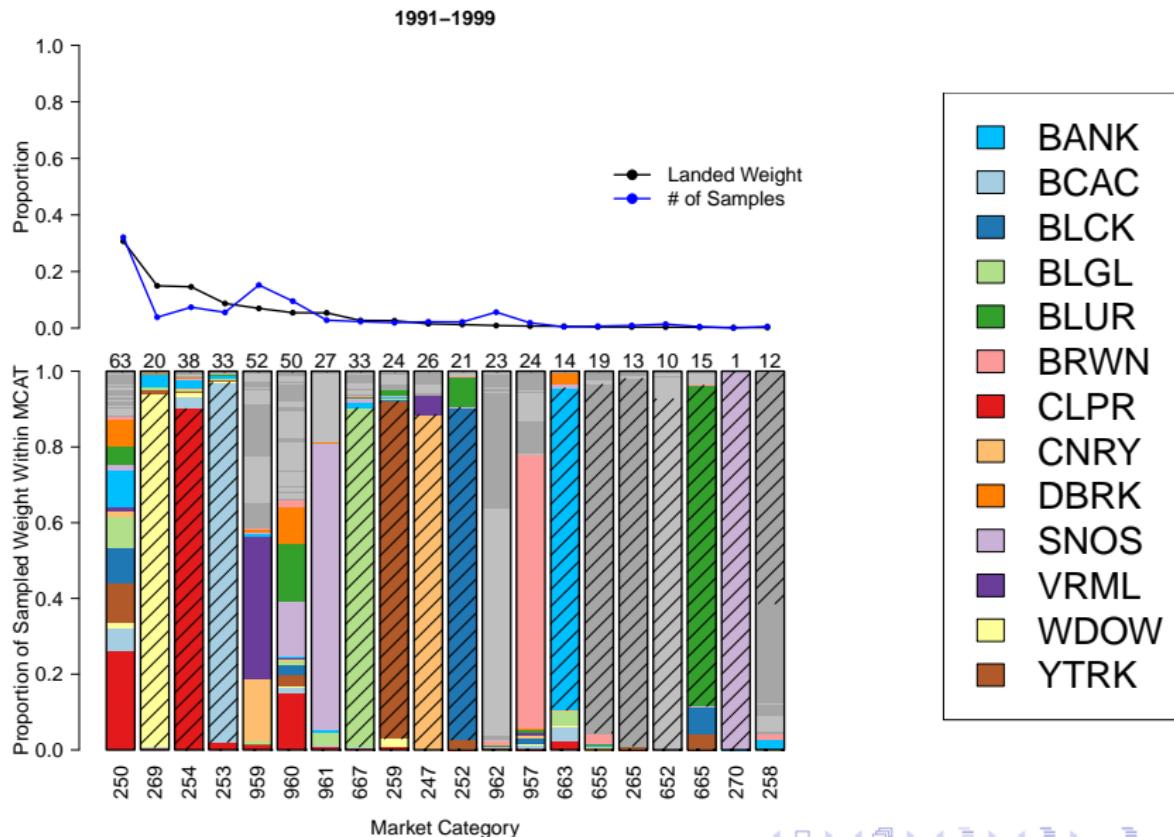
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2000–2015

