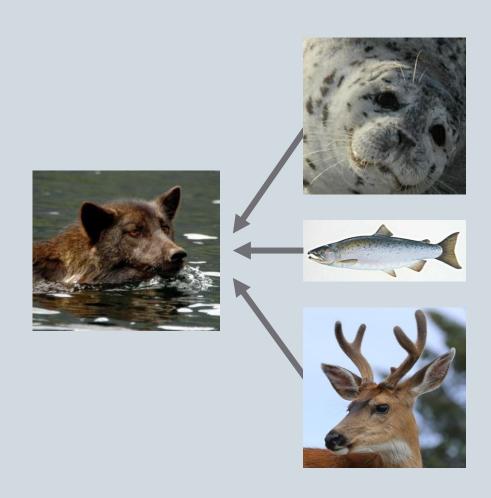
Bayesian stable isotope mixing models and the MixSIAR R package

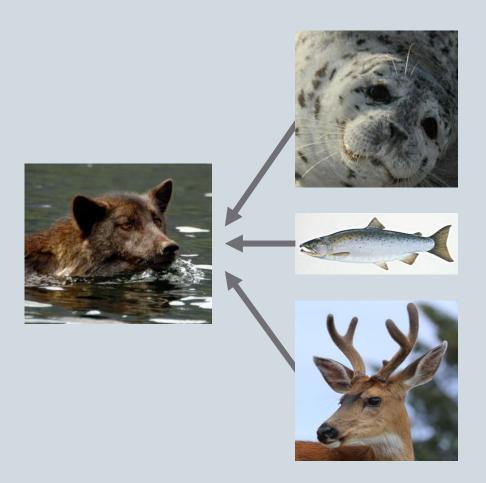
BRIAN STOCK

APRIL 3, 2023

Calculate source % to a mixture

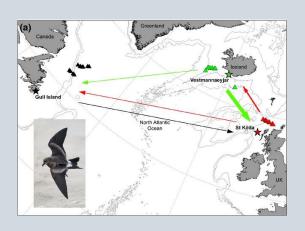


Calculate prey % to a diet



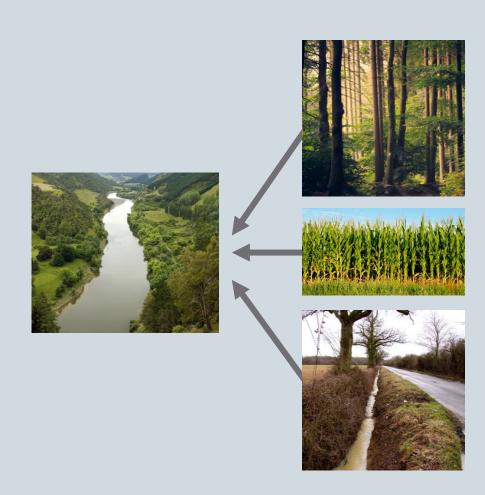
MixSIAR wolves example
http://brianstock.github.io/MixSIAR/articles/wolves ex.html

Calculate colony % to a bird

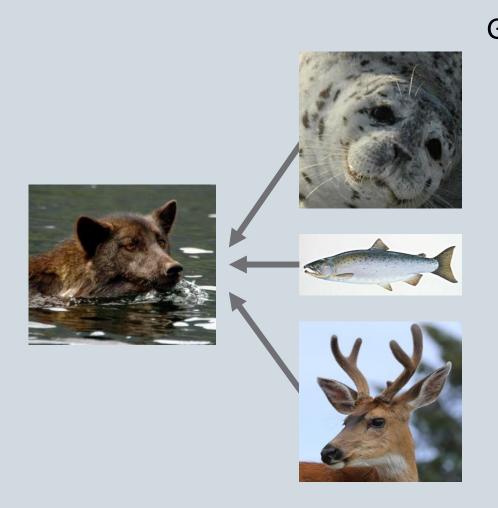




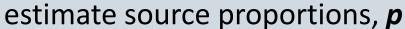
Calculate **soil** % to a **sediment**

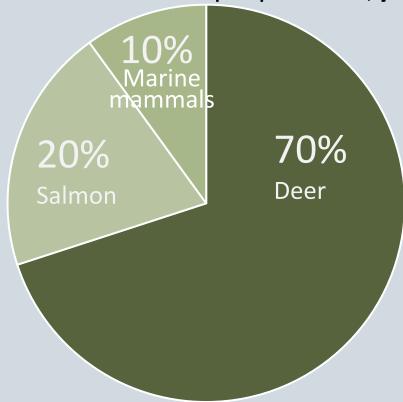


Calculate **prey** % to a **diet**

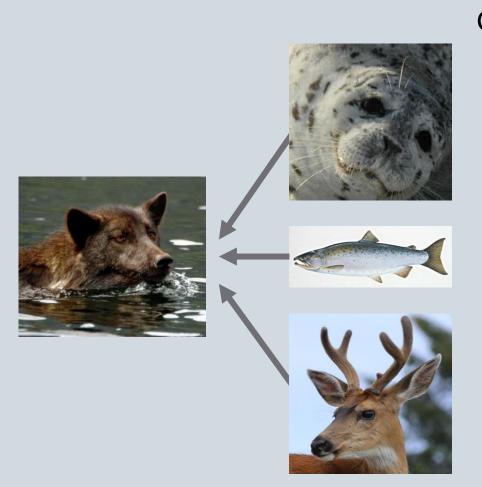


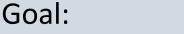
Goal:

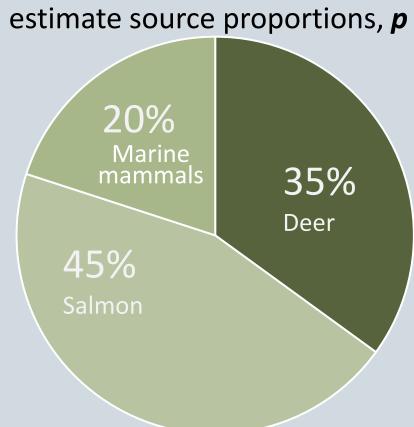




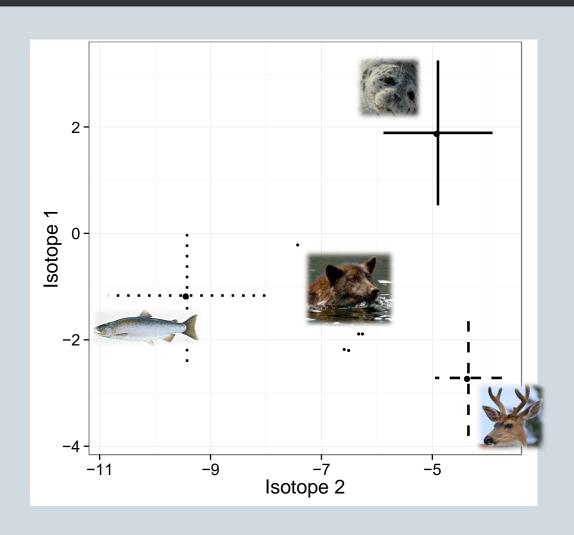
Calculate prey % to a diet

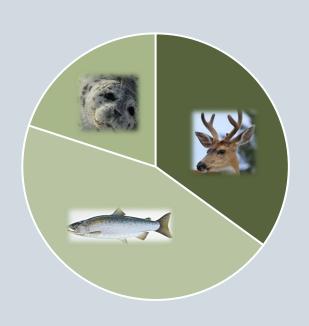




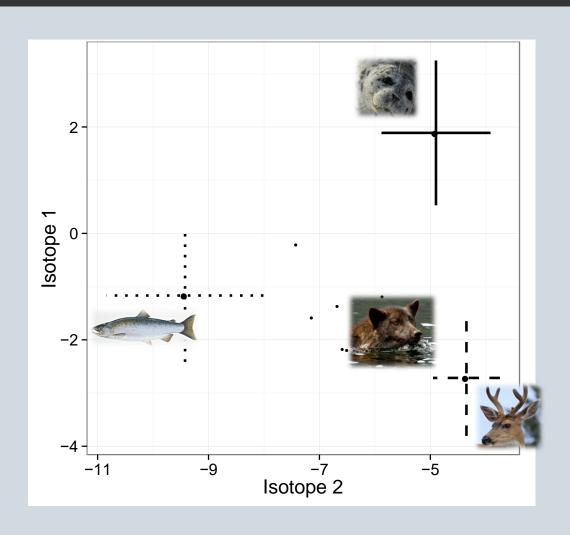


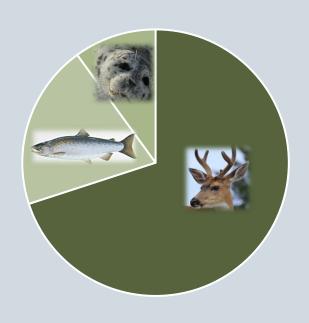
Using stable isotope data "biotracer"





Using stable isotope data "biotracer"

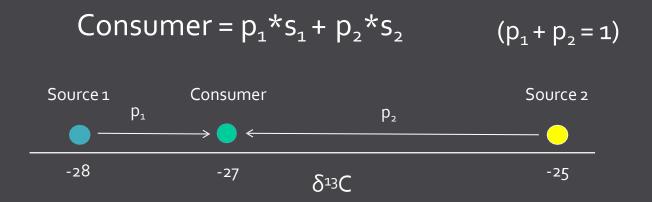




Linear mixing model:



Linear mixing model:

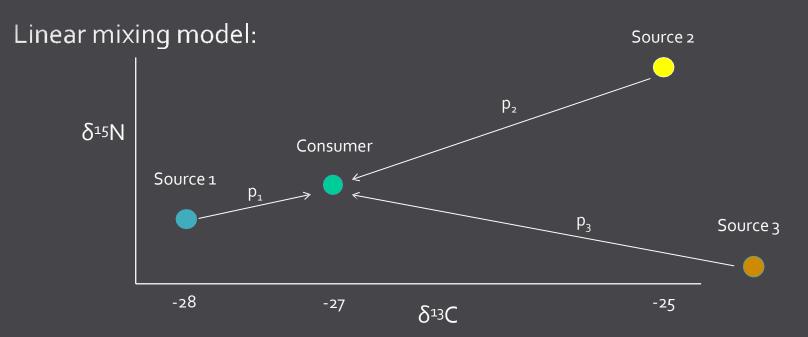


Linear mixing model:

$$-27 = 2/3(-28) + 1/3(-25)$$



Twice as close to source 1 than source 2



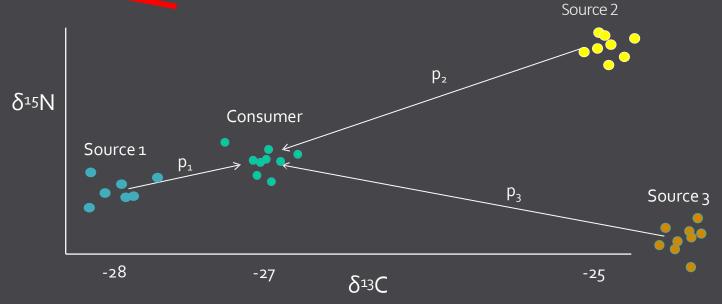
Consumer_C =
$$p_1s_{1C} + p_2s_{2C} + p_3s_{3C}$$

Consumer_N =
$$p_1s_{1N} + p_2s_{2N} + p_3s_{3N}$$

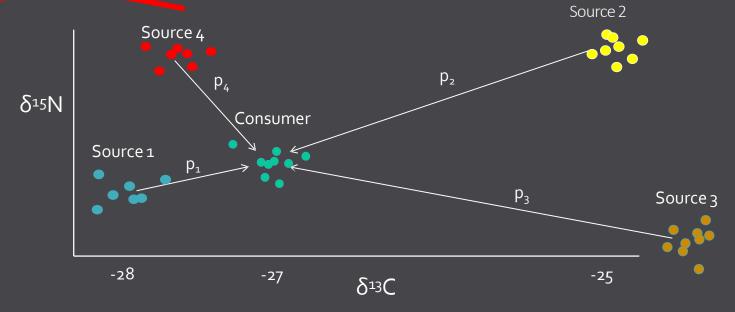
$$p_1 + p_2 + p_3 = 1$$

Introduction

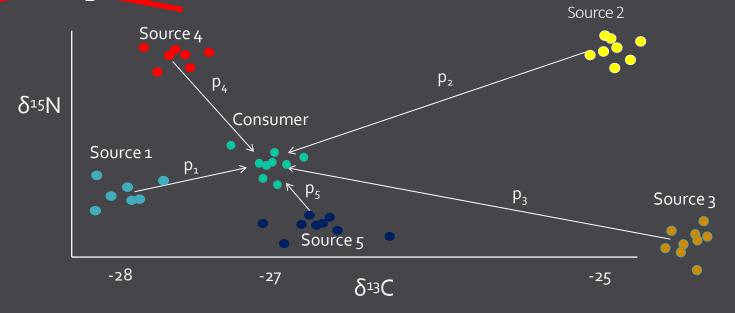




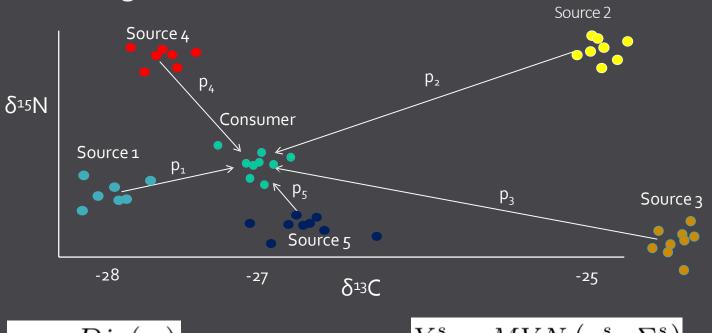
Linear mixing model:



Linear mixing model:

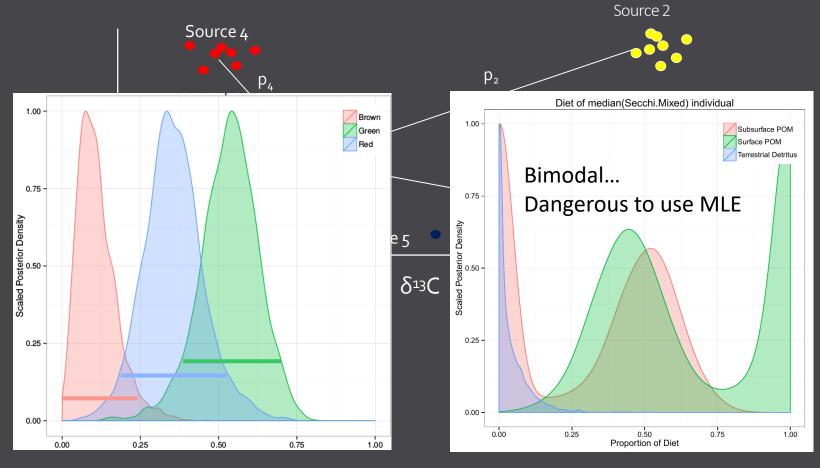


Bayesian mixing model:

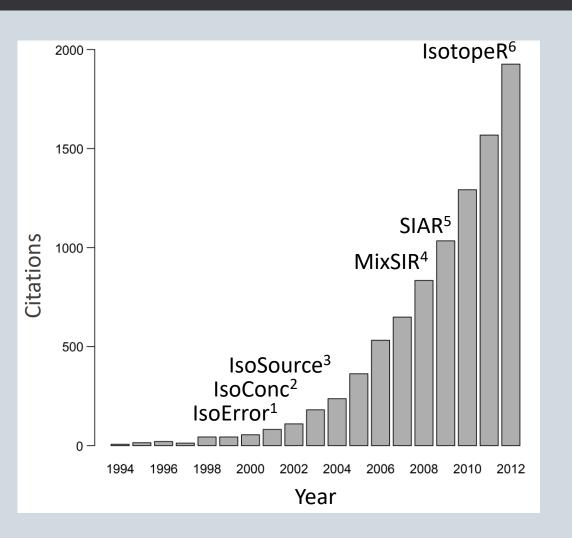


$$\begin{array}{l}
\mathbf{p} \sim Dir(\boldsymbol{\alpha}) & Y_{jk}^s \sim MVN\left(\mu_{jk}^s, \Sigma_k^s\right) \\
Y_{ij} \sim MVN\left(\sum_k p_k \mu_{jk}^s, \Sigma\right) & \mu_{jk}^s \sim \mathcal{N}(0, .001) \\
\hline
\tau_{jk} \sim gamma(.001, .001)
\end{array}$$

Bayesian mixing model:



MixSIAR incorporates years of methods development



¹Phillips and Gregg 2001a, 2001b

Uncertainty (ad hoc)

²Phillips and Koch 2002

Concentration dependence

³Phillips and Gregg 2003

Underdetermined system

⁴Moore and Semmens 2008

• Bayesian, MATLAB GUI

⁵Parnell et al. 2010

• Bayesian, residual error

Ward et al. 2010

Fit source means

⁶Hopkins and Ferguson 2012

Covariance

Parnell et al. 2013

MixSIAR: the general idea

1. Load data in R

2. Choose model options

```
if(resid_err){
cat("
    # Additive residual error (SIAR)
    for(iso in 1:n.iso){
    resid.prcsn[iso] ~ dgamma(.001,.001);
    resid.var[iso] <- 1/resid.prcsn[iso];
}
", file=filename, append=T)</pre>
```

3. write_JAGS_model.r

4. Run model in JAGS
5. Diagnostics
+ output in R

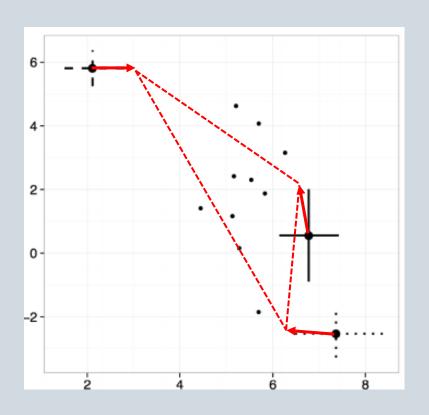
The plan

- 1. Introduction
- 2. Source fitting
- 3. Covariate effects

Don't think other software currently includes these (simmr)

- 4. Effects of priors
- 5. GitHub site + resources

Source fitting



Account for sampling error of the sources (allow source means to deviate from the sample means)

- 1. Raw data
 - Parnell et al. 2013
 - includes covariance
- 2. Mean / SD / n
 - Ward et al. 2010
- 3. Fixed
 - set n = 10000

*Source and TDF variance indistinguishable

$$\sqrt{\sigma_{source}^2 + \sigma_{discr}^2}$$

Source fitting

http://dx.doi.org/10.7717/peerj.5096



Analyzing mixing systems using a new generation of Bayesian tracer mixing models

Brian C. Stock¹, Andrew L. Jackson², Eric J. Ward³, Andrew C. Parnell⁴, Donald L. Phillips⁵ and Brice X. Semmens¹

Supplement 1

MixSIAR Model Descr

Contents

Account for sampling error of the sources (allow source means to deviate from the sample means)

- 1. Raw data
 - Parnell et al. 2013
 - includes covariance
- 2. Mean / SD / n
 - Ward et al. 2010
- 3. Fixed
 - set n = 10000

*Source and TDF variance indistinguishable

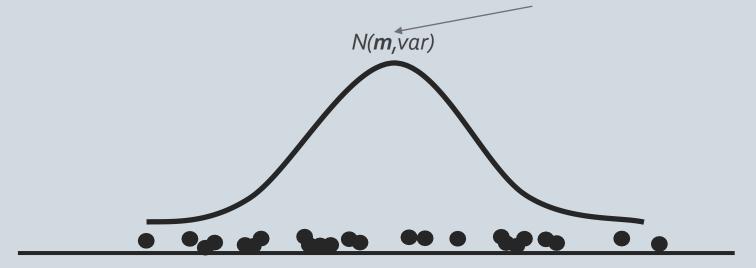
$$\sqrt{\sigma_{source}^2 + \sigma_{discr}^2}$$

Covariate effects in MixSIAR

No covariate effects...

$$p = [20\%, 50\%, 20\% 10\%]$$

Assumes that all consumers have the Same diet



Consumer δ¹³C

Covariate effects in MixSIAR

Transform p's

Linear regression in ILR-space

$$\mathbf{p} \sim Dir(\boldsymbol{\alpha})$$

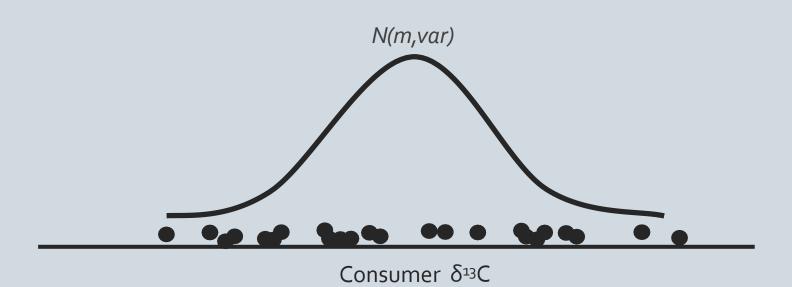
 $p_{ik} = inverseILR(ilr.global_k + ilr.fac1_{mk} + ilr.cont1_kCont1_i)$

Covariate effects in MixSIAR

Transform p's

Linear regression in ILR-space

Fixed effects

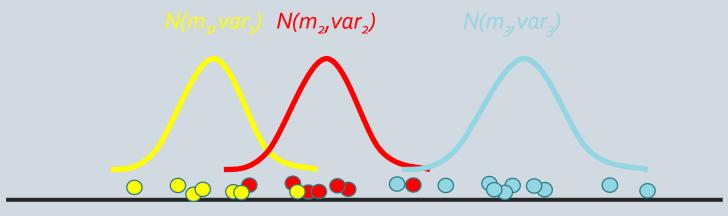


Fixed effects

Simplest, estimate mean **p** for different groups independently

Similar to running separate mixing models (e.g. SIAR, simmr), but

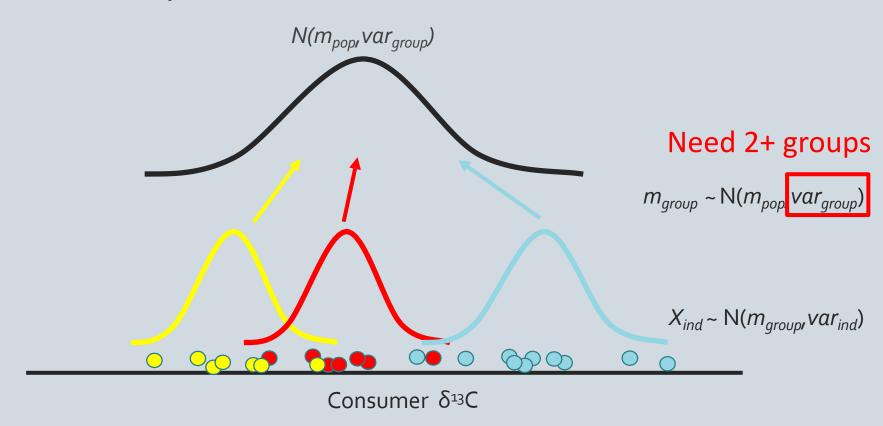
Source means, error terms fit 1x and shared across groups



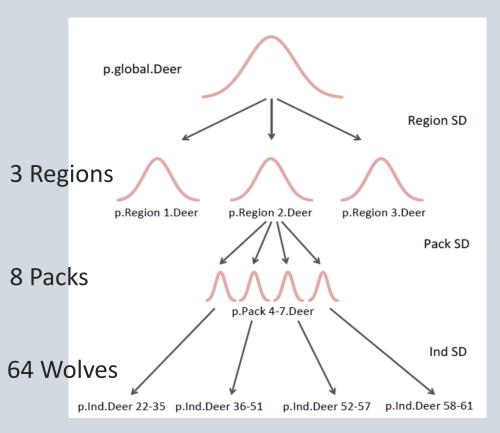
Consumer δ¹³C

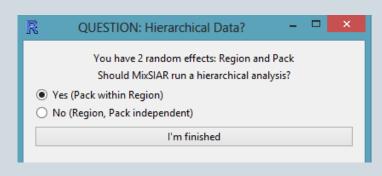
Random effects

More complex



Nested/hierarchical random effects

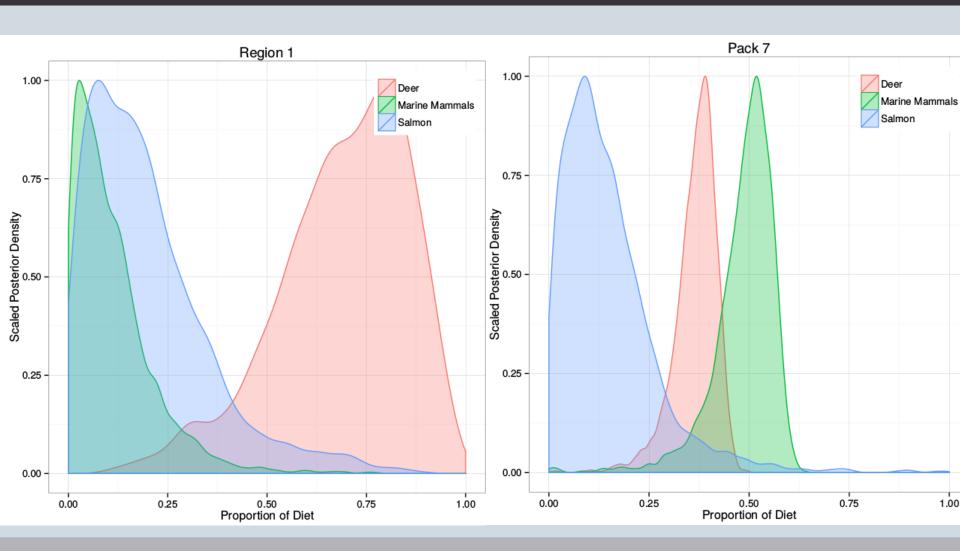




```
Hierarchical ilr.fac2.tot[f2,src] <- ilr.global[src] + ilr.fac1[factor1_lookup[f2],src] + ilr.fac2[f2,src];

Independent ilr.fac2.tot[f2,src] <- ilr.global[src] + ilr.fac2[f2,src];
```

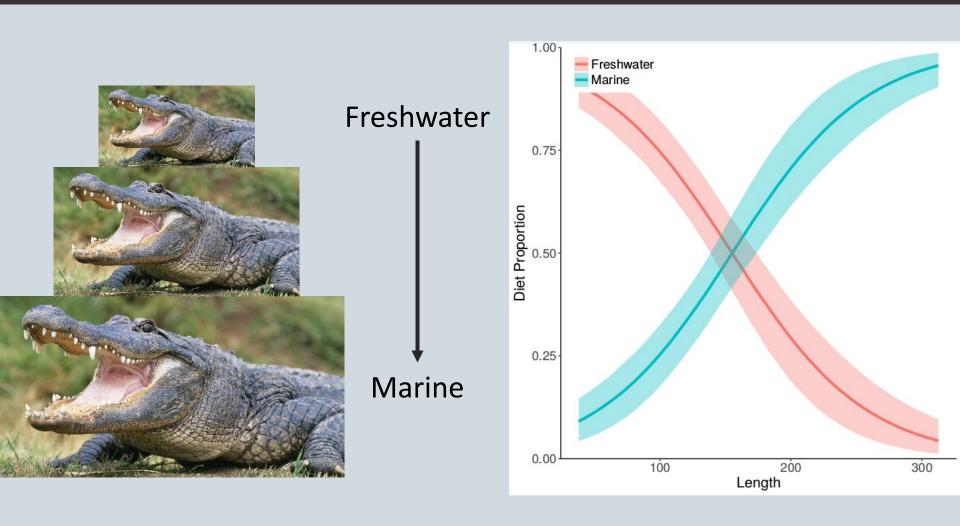
(Nested) Random effects: Wolves Ex



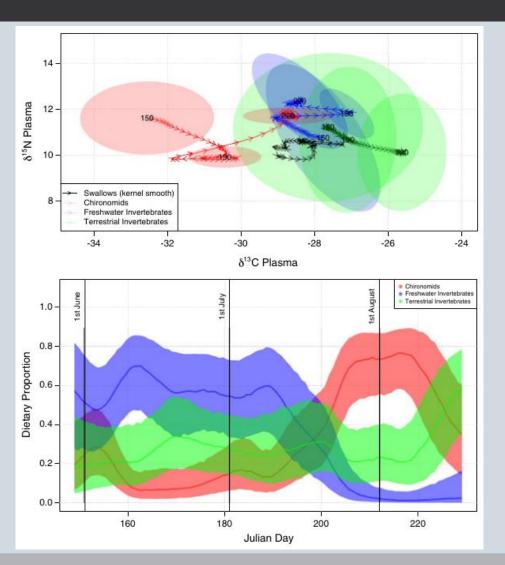
3. Covariate effects

Semmens et al. (2009)

Continuous effect: Alligator length



More complexity is possible ...way beyond MixSIAR



Multivariate spline model

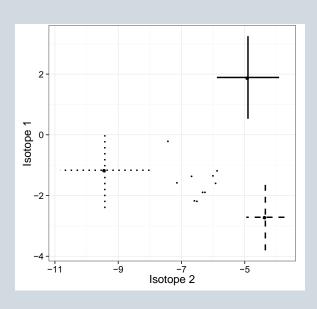
1. Time-varying source means and covariances

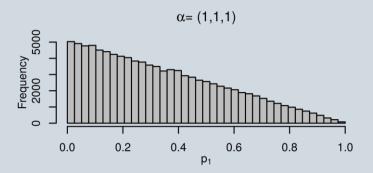
2. Time-varying diet proportions

4. Effect of priors/
"Bayesian mixing models are biased"

a. There is no "uninformative" prior

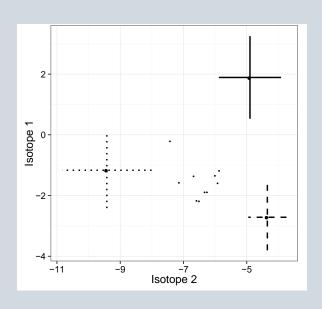
Problem: proportions are not independent!

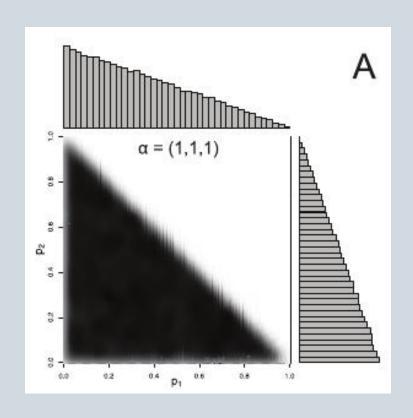




a. There is no "uninformative" prior

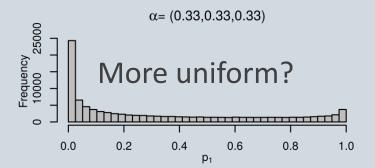
Problem: proportions are not independent!

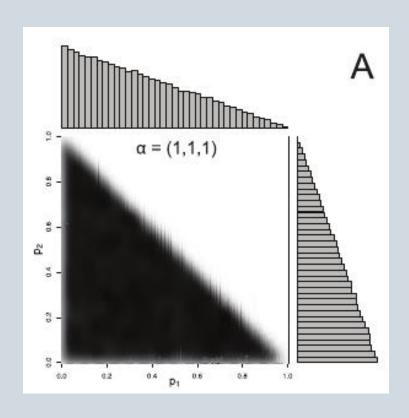




All **p** equally likely "uninformative" / generalist

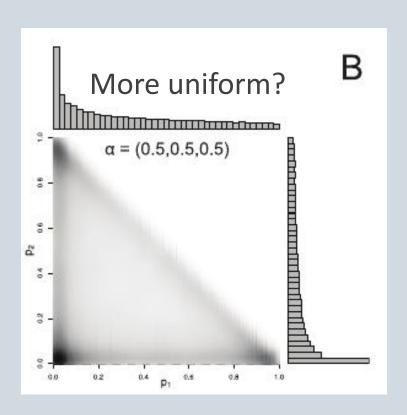
a. There is no "uninformative" prior



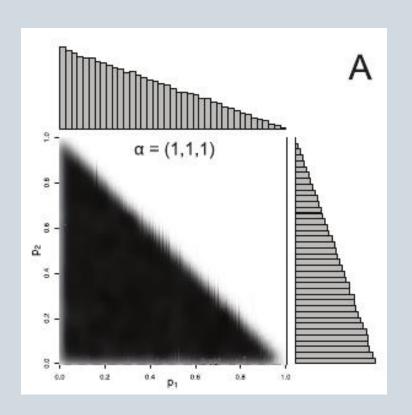


All **p** equally likely "uninformative" / generalist

a. There is no "uninformative" prior



Extreme p heavily weighted

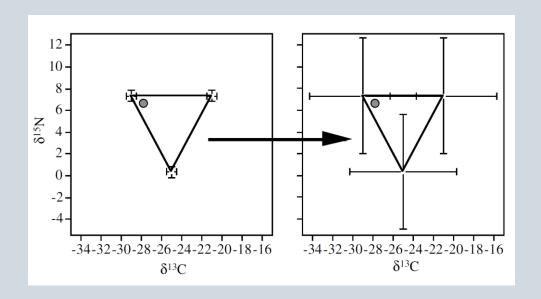


All **p** equally likely "uninformative" / generalist

4. Priors

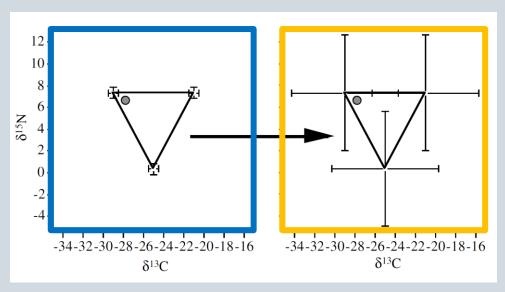
b. Effect of the "uninformative" prior

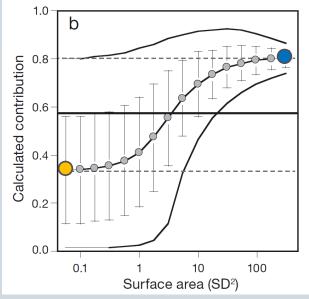
1. How good is your data?



b. Effect of the "uninformative" prior

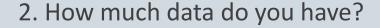
1. How good is your data?

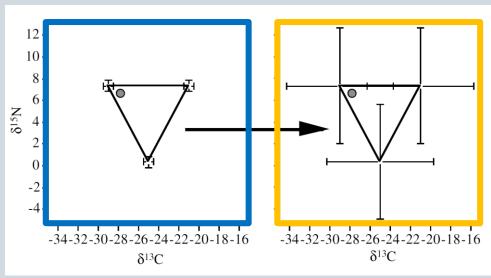


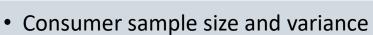


b. Effect of the "uninformative" prior

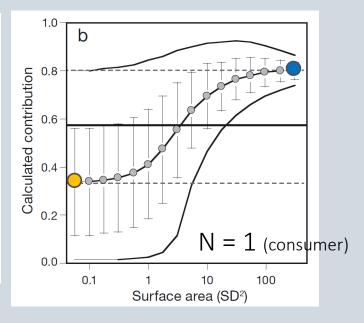
1. How good is your data?







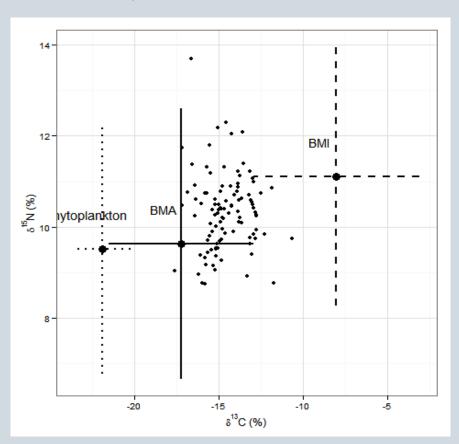
- Source sample size and variance
- Source geometry (are you unlucky?)



c. Source geometry

High source uncertainty*

- Collect more samples
- TDF experiment



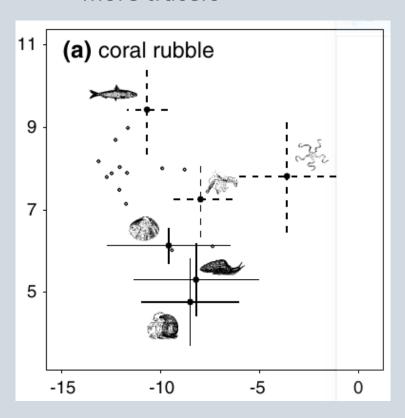
*Source and TDF variance indistinguishable

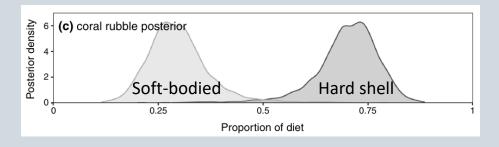
$$\sqrt{\sigma_{source}^2 + \sigma_{discr}^2}$$

c. Source geometry

Too many sources...

- combine (before/after)
- informative prior(s)
- more tracers



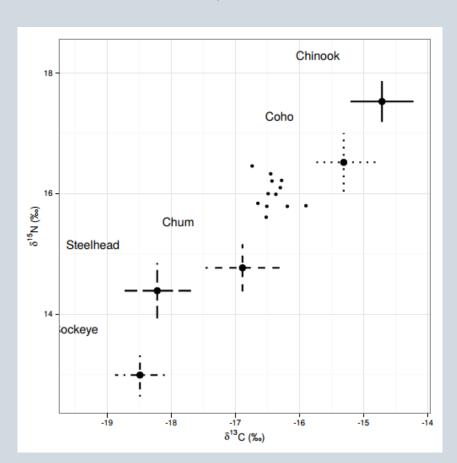


https://github.com/brianstock/MixSIAR/blob/master/inst/example_scripts/mixsiar_script_mantis.R

c. Source geometry

Sources confounded, multiple solutions...

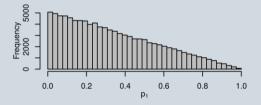
informative prior (stomach or fecal contents, visual obs, opinion)

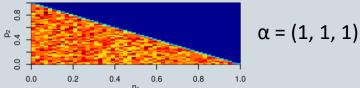


*Source and TDF variance indistinguishable

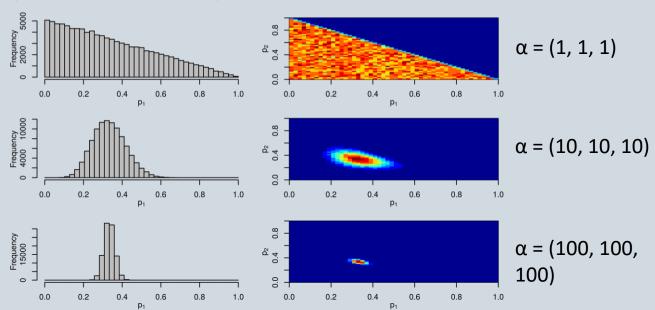
$$\sqrt{\sigma_{source}^2 + \sigma_{discr}^2}$$

You control the mean proportions AND the variance ("informativeness")





You control the mean proportions AND the variance ("informativeness")



You control the mean proportions AND the variance ("informativeness")

30



8



25



You control the mean proportions AND the variance ("informativeness")

30

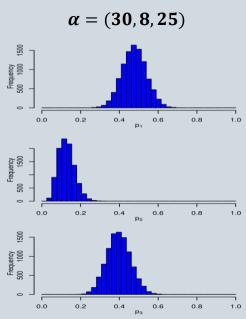


8

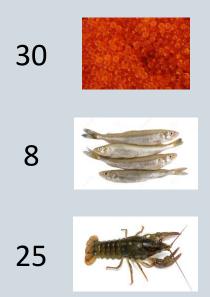


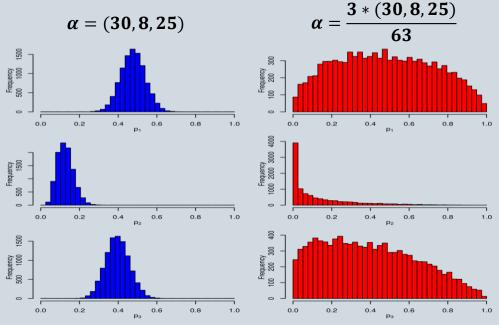
25





You control the mean proportions AND the variance ("informativeness")





Where do I get MixSIAR?

CRAN (older)

- 1. Download and install/update R
- 2. Download and install JAGS
- 3. Open R and run:

install.packages("MixSIAR")

GitHub (latest)

- Download and install/update R
- Download and install JAGS
- 3. Open R and run:

```
library(devtools)
install_github("brianstock/MixSIAR")
```

MixSIAR: GUI RIP

Graphical User Interface (GUI) deprecated... RGtk2...



MixSIAR manual

https://github.com/brianstock/MixSIAR/blob/master/inst/mixsiar manual small.pdf

L	Introduction	
	1.1 What is the MixSIAR GUI?	
	1.2 What is the MixSIAR model framework? How does it relate to SIAR and Mix	S
	1.3 Model options	
	1.4 Great! How do I get started using MixSIAR?	
	Installation	
	2.1 Windows	
	2.2 Mac OS X	
	2.3 Linux	
	Running the working examples	
	3.1 Wolves Example	
	3.2 Geese Example	
	3.3 Lake Example	
	3.4 Palmyra Example	
	3.5 Killer Whale Example	
	3.6 Isopod Example	
	3.7 Cladocera Example	
	3.8 Storm-petrel Example	
	3.9 Snail Example	
	3.10 Mantis Example	
	3.11 Alligator Example	
	Using MixSIAR with your own data	
	4.1 Running your own data	
	4.2 Data file format and loading	
	4.3 Using the MixSIAR script version	
	4.4 GitHub Issues page	
	4.5 Citing MixSIAR	
	Introductions to Bayesian Analysis	
	5.1 Convergence Diagnostics	

3.1.4 Making an isospace plot

Once the mixture, source, and discrimination data are loaded, you can click "Make isospace plot". Your plot should match that of Figure 2. If you want to save the isospace plot as a .pdf or .png, make sure either/both of the appropriate boxes are checked. You can also change the name of the file here (default is to save the plot as "isospace_plot.pdf").

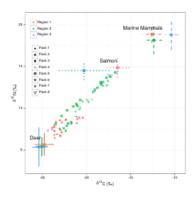


Figure 2: **Stable isotope input for the Wolves Example.** Mixture data (wolves) are by Region (color) and Pack (shape). Source data are by Region and have been adjusted by discrimination means **and SDs**. Error bars indicate ± 1 SD, the combined source+discrimination SD calculated as $\sqrt{\sigma_{source}^2 + \sigma_{discr'}^2}$, under the assumption of independence.

You should ALWAYS look at the isospace plot—this is a good check that the data is loaded correctly, and that the isospace geometry makes sense. If the mixture data are well outside the source polygon, you have a serious violation of mixing model assumptions, and it must be true that either 1) You're missing a source, or 2) You're using an incorrect discrimination factor. MixSIAR, like SIAR, fits a residual error term, and thus will always find a solution even if it is nonsensical.

MixSIAR model description

http://dx.doi.org/10.7717/peerj.5096



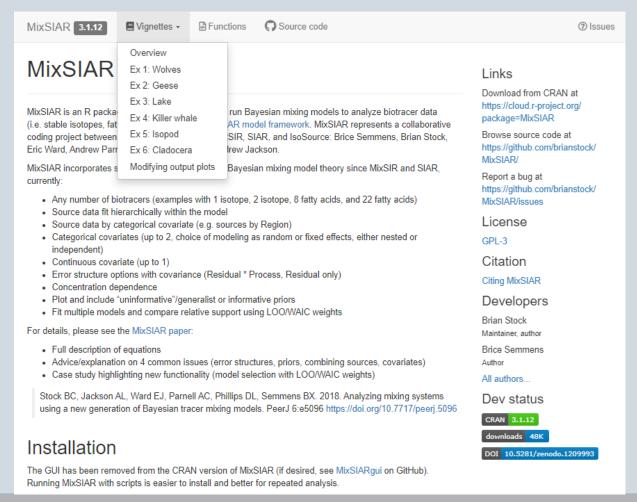
Analyzing mixing systems using a new generation of Bayesian tracer mixing models

Brian C. Stock¹, Andrew L. Jackson², Eric J. Ward³, Andrew C. Parnell⁴, Donald L. Phillips⁵ and Brice X. Semmens¹

Supplement 1: advice on model setup

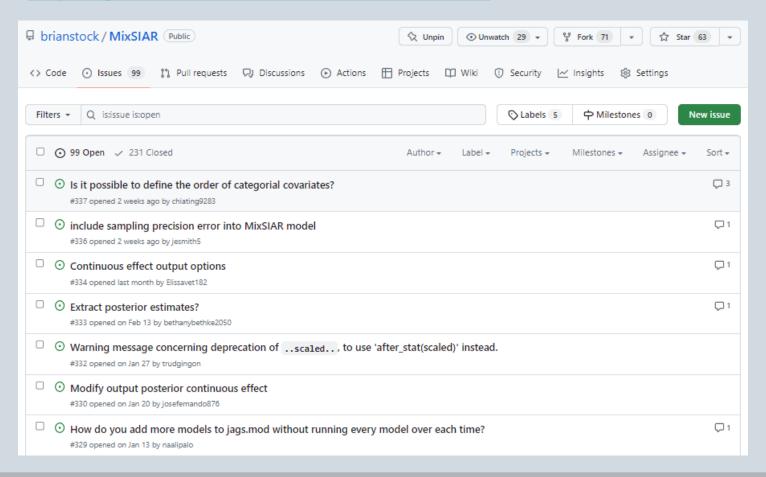
MixSIAR website + vignettes

http://brianstock.github.io/MixSIAR/



GitHub Issues page

https://github.com/brianstock/MixSIAR/issues/



GitHub Issues page

