# Modeling changes in synchrony

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## Study questions:

- ► How has the phenology of different trophic levels changed over time and are there differences across suites of interactions?
- ➤ Are the observed trends in shifts in interactions similar using data from single-species studies, or should we be using biologically relevant paired species data?

# Building the model from simple to complex...

- How has the phenology of different trophic levels changed over time and what differences are there across suites of interactions?
  - To compare findings with those of Thackeray and Cohen, I am analyzing the data as single species using a simpler model

$$d\hat{oy} = \alpha_{sp_i} + \beta(year_i)$$
  
 $\alpha_s p \sim N(\mu, \sigma)$   
 $doy_i \sim N(d\hat{oy}, \sigma_i)$ 

# Building the model from simple to complex...

Are the observed trends in shifts in interactions similar using data from single-species studies, or should we be using biologically relevant paired species data?

$$do\hat{y_{r_i}} = \alpha_{r_i} + \beta_{r_i}(year_i)$$
  
 $\beta_{r_i} \sim N(\mu_r, \sigma_r)$   
 $doy_{r_i} \sim N(do\hat{y_{r_i}}, \sigma_{doy})$ 

$$egin{aligned} ext{doy}_{c_i} &= lpha_{c_i} + eta_{ ext{shift}}( ext{year}_i) \ eta_{ ext{shift}} &\sim (eta_{c_i} - eta_{r_i}) \ eta_{c_i} &\sim ext{N}(\mu_c, \sigma_c) \ ext{doy}_{c_i} &\sim ext{N}( ext{doy}_{c_i}, \sigma_{ ext{doy}}) \end{aligned}$$

# Generating data:

#### Test data variables:

- Study
- Species
- Event
- Year
- DOY

	Study	Species	Event	Year	DOY	S
1	1	51	4	1	206.9588	
2	1	51	4	2	208.6147	
3	1	51	4	3	212.5579	
4	1	51	4	4	222.5058	
5	1	51	4	5	223.9829	
6	1	51	4	6	232.6540	

#### Stan model:

```
/ The input data is a vector 'y' of length 'N'
 int<lower=0> N: //No. obs
 int<lower=0> Nspp; //No. spp
 int species[N]; // Grouping by species
 vector[N] year;
 real vpred[N]; //DOY of pheno event
// The parameters accepted by the model. Our model
// accepts two parameters 'mu' and 'sigma'.
parameters {
 //seting random values for intercepts
 real mu_a_sp;
  real<lower=0> siama a sp:
  real<lower=0,upper=365> a[Nspp]; // intercept for species
  real mu b sp:
  real<lower=0> siama_b_sp:
  real b[Nspp]: // slopes for species
  real<lower=0> siama v:
transformed parameters(
  real mu_y[N]; //individual mean
 for(i in1:N){
    mu_y[i]=a[species[i]]+b[species[i]]*year[i];
// The model to be estimated. We model the outputn'y' to be normally distribute
model {
 a ~ normal(mu_a_sp, sigma_a_sp); //using the previously defined values here
 b ~ normal(mu b sp. siama b sp):
 ypred ~ normal(mu_y, sigma_y);
 mu_a_sp ~normal(188, 20);
  siama_a_sp ~normal(0,20);
  mu_b_{sp} \sim normal(5,2);
  siama_b_sp ~normal(0,2);
  sigma_y ~normal(0,2);
```

## Stan model output:

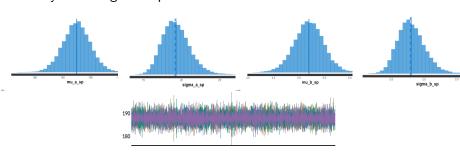
# Priors: //Priors mu\_a\_sp ~normal(188, 20); sigma\_a\_sp ~normal(0,20); mu\_b\_sp ~normal(5,2); sigma\_b\_sp ~normal(0,2); sigma\_y ~normal(0,2);

### Model output:

```
Inference for Stun model: singlesp_randslopes_goo.
4 Chains, each with iter=8000; warmp=8000; thin=1;
post-warmp draws per chain-4000, total post-warmp draws=16000.

mu_a_sp 187.50 0.01 2.43 182.70 185.87 187.51 189.12 192.31 37664 1
sigmo_a_sp 187.50 0.01 2.43 182.70 185.87 187.51 189.12 192.31 37664 1
sigmo_a_sp 192.80 0.01 1.75 16.71 18.62 19.15 20.93 21.31 48550 1
mu_b_sp 5.20 0.00 0.29 4.62 5.01 5.20 5.39 5.77 36173 1
sigmo_sp 2.31 0.00 0.21 1.95 2.16 2.29 2.43 2.72 28629 1
sigmo_sp 2.31 0.00 0.00 1.91 1.96 1.98 2.00 2.05 29088 1
0[1] 14.99 0.01 1.51 143.02 244.99 15.99 147.01 149.72 6578 1
0[2] 175.20 0.00 0.72 113.87 177.71 175.20 175.09 176.62 28136 1
0[3] 16.13 0.01 1.36 18.87 1.1604-916.31 816.23 104.06 82 23124 1
0[4] 143.18 0.00 0.56 142.09 142.80 143.19 143.56 144.28 26551 1
0[5] 193.00 0.01 0.01 197.53 195.76 194.02 00.03 15.51 51.53 10.23 10.24 25551 1
```

#### Shiny stan diagnostic plots:



Parameter: mu a sp

## **Next Steps:**

- More accurately represent slopes in the test data
- Manipulate posteriors to detect differences across groups
- ► Test on real data
- Repeat for a joint model