



A flexible and efficient spatiotemporal modeling strategy for trend analysis of North American Christmas Bird Counts

L. Miller

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1900 Christmas Bird Count

- 27 volunteers
- 25 count circles
- 18,500 birds counted
- 89 species recorded
- 2 Countries



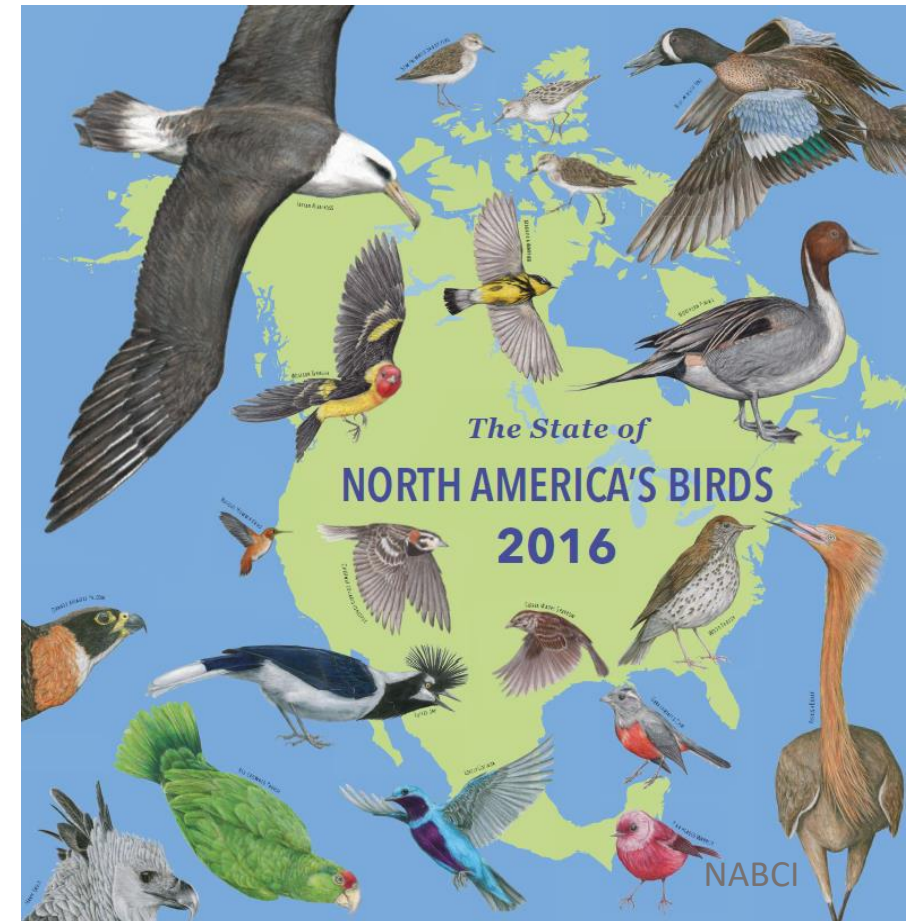
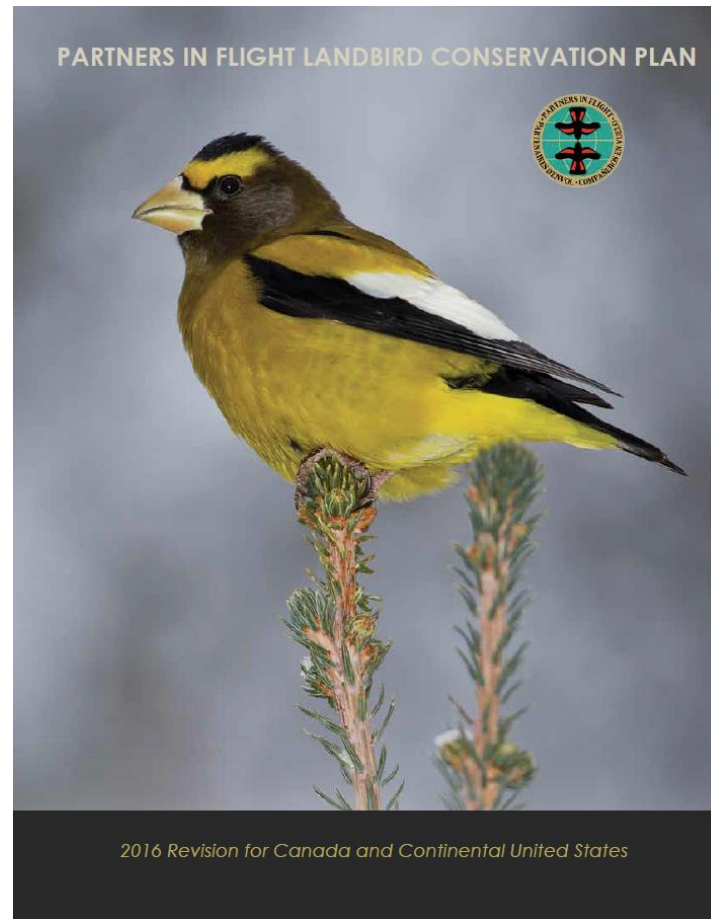
2017 Christmas Bird Count

- 73,153 volunteers
- 2,536 count circles
- 56,139,812 birds counted
- 2,636 species recorded
- Across North and South America, Caribbean, Pacific Islands



CBC Trends and Conservation

- Count trends
- 1966-2017
- 550 species
- Partners in Flight
- North American Bird Conservation Initiative



Standard Approach: Independent Strata

Analytical Strata and CBC Circles

States and Provinces



Bird Conservation Regions



Standard Approach: Model Counts

$$\log(\lambda_{i,j,t}) = S_i + \omega_j + \gamma_{i,t} + \beta_i(t - t^*) + \frac{B_i (\xi_{i,j,t}^{p_i} - 1)}{p_i} + \varepsilon_{i,j,t}$$

1. Intercept per stratum

2. Effort effect per stratum

3. Log-linear year effect per stratum

4. Stratum by year residual

5. Circle random effect

6. Overdispersion

Standard Approach: Abundance Index

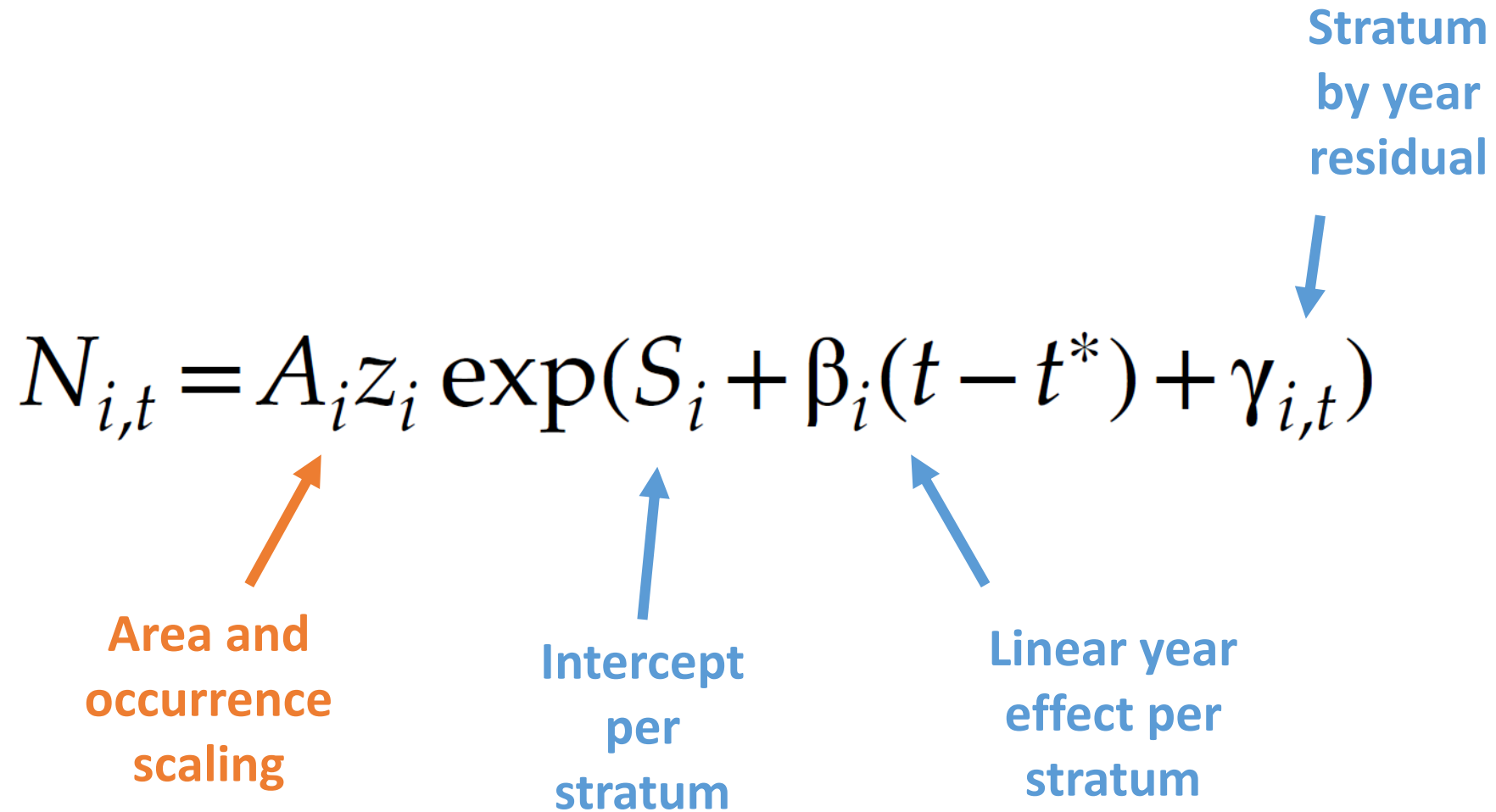
$$N_{i,t} = A_i z_i \exp(S_i + \beta_i(t - t^*) + \gamma_{i,t})$$

Stratum by year residual

Area and occurrence scaling

Intercept per stratum

Linear year effect per stratum



Standard Approach: Abundance Index Trend

$$T_i = \left\{ \frac{N_{i,t_b}}{N_{i,t_a}} \right\}^{1/(t_b - t_a)}$$

Different Approach: Emphasize Fine-Scaled Trends




Different Approach: Emphasize Fine-Scaled Trends

$$\log(\mu_{i,t}) = \alpha_i + \epsilon_i \log(E_{i,k,t}) + \tau_i T_{i,k,t} + \kappa_k$$

1. Global and
CAR-random
abundance index for
reference year



2. Global and
CAR-random
effort slope
(SVC)



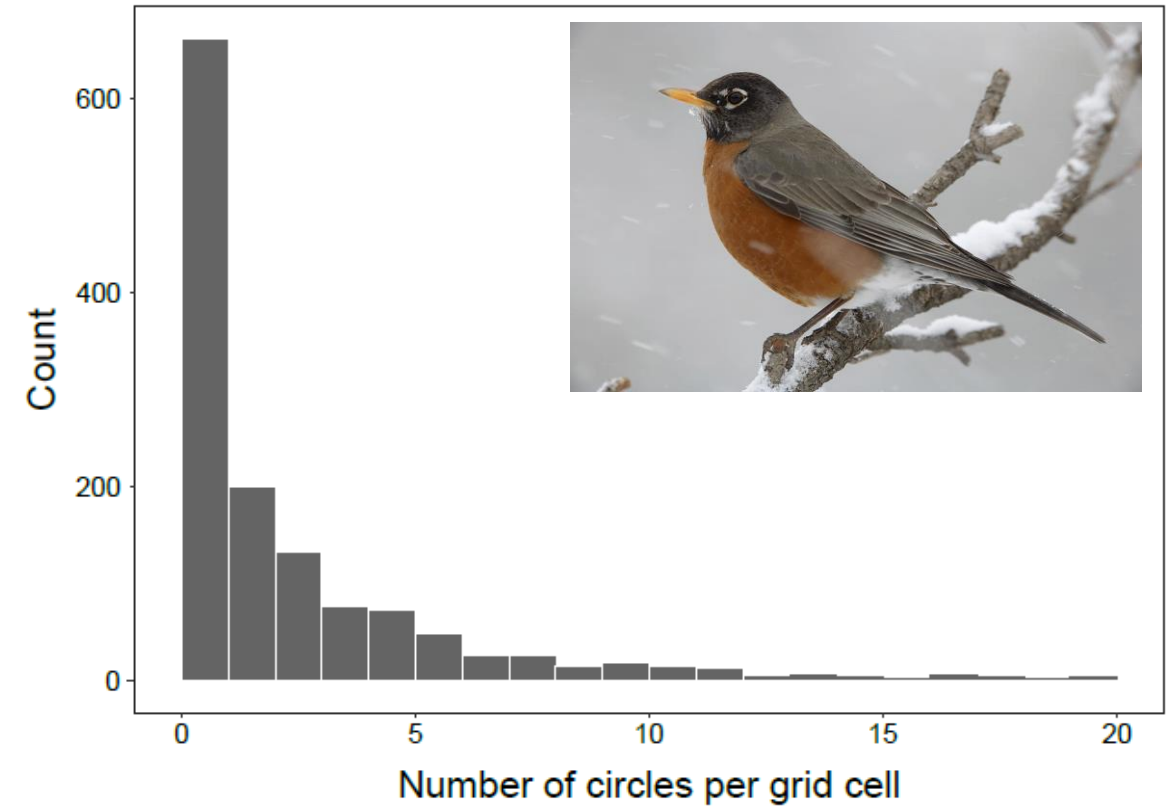
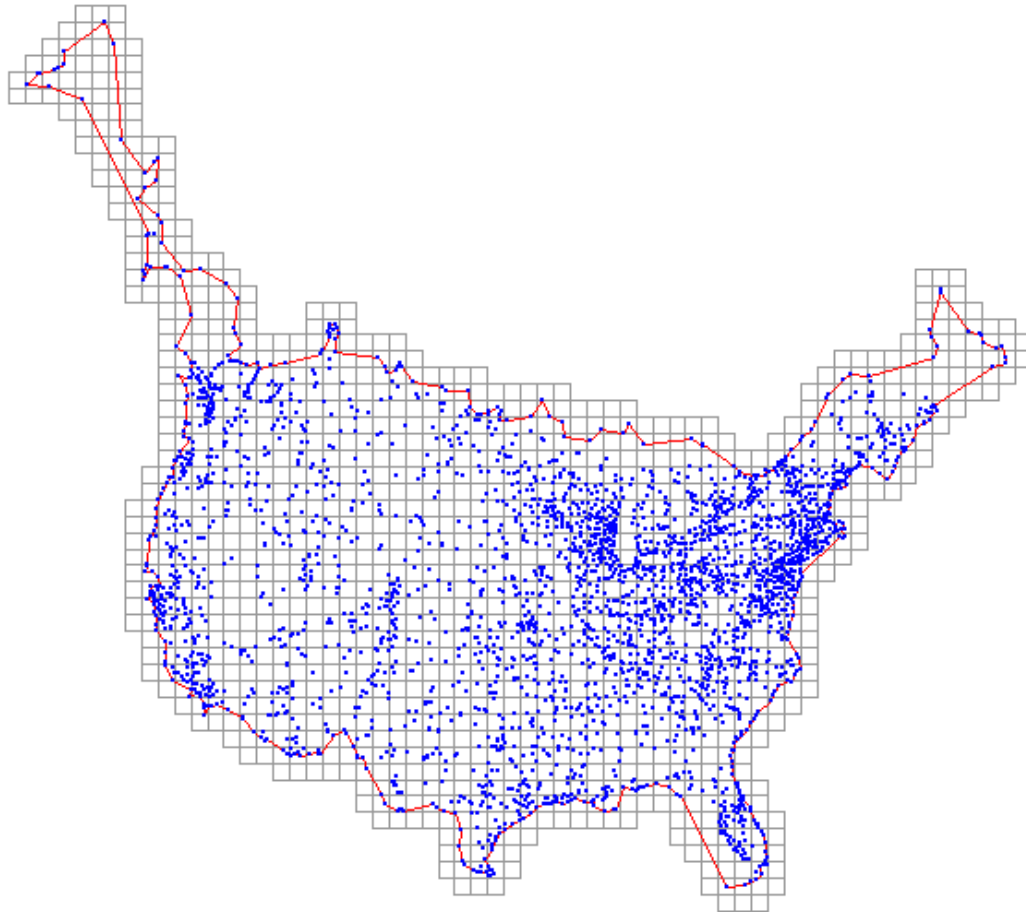
3. Global and
CAR-random year
slope (SVC)



4. Exchangeable
random circle
effect



Example: American Robin



Example: Computing with R-INLA

Model statement

```
formula <- count ~ 1 + f(grid_id1, model="besag", graph=g) + log_hrs +  
  f(grid_id2, log_hrs, model="besag", graph=g) + std_yr + f(grid_id3, std_yr,  
  model="besag", graph=g) + f(circle, model="iid")
```

Function call

```
result <- inla(formula, family="nbinomial", data=modeling_data,  
  control.compute=list(cpo=T, waic=T, config=T),  
  control.inla=list(int.strategy='eb'))
```

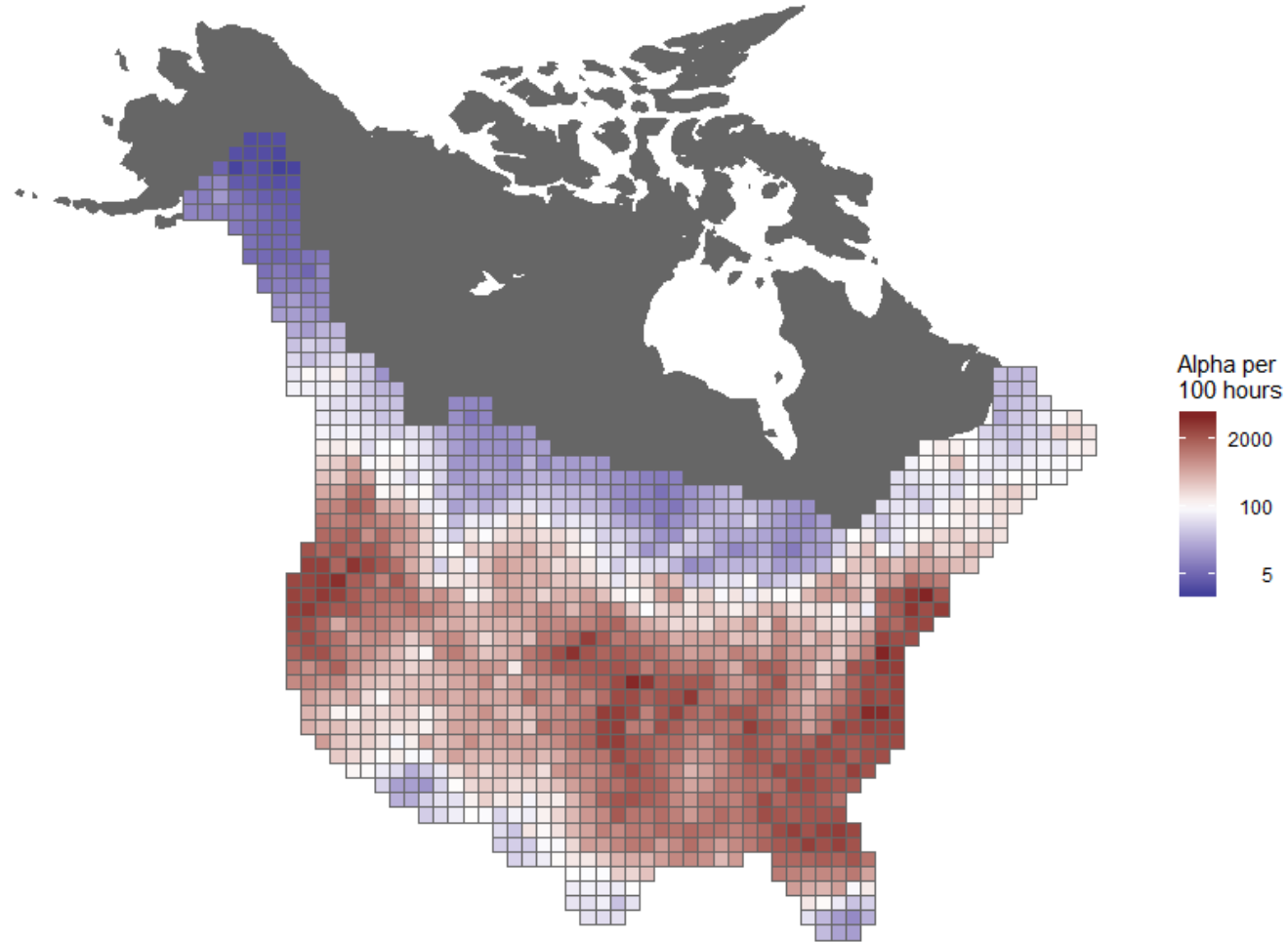
Example: Computing with R-INLA

Time used: 61.68 min

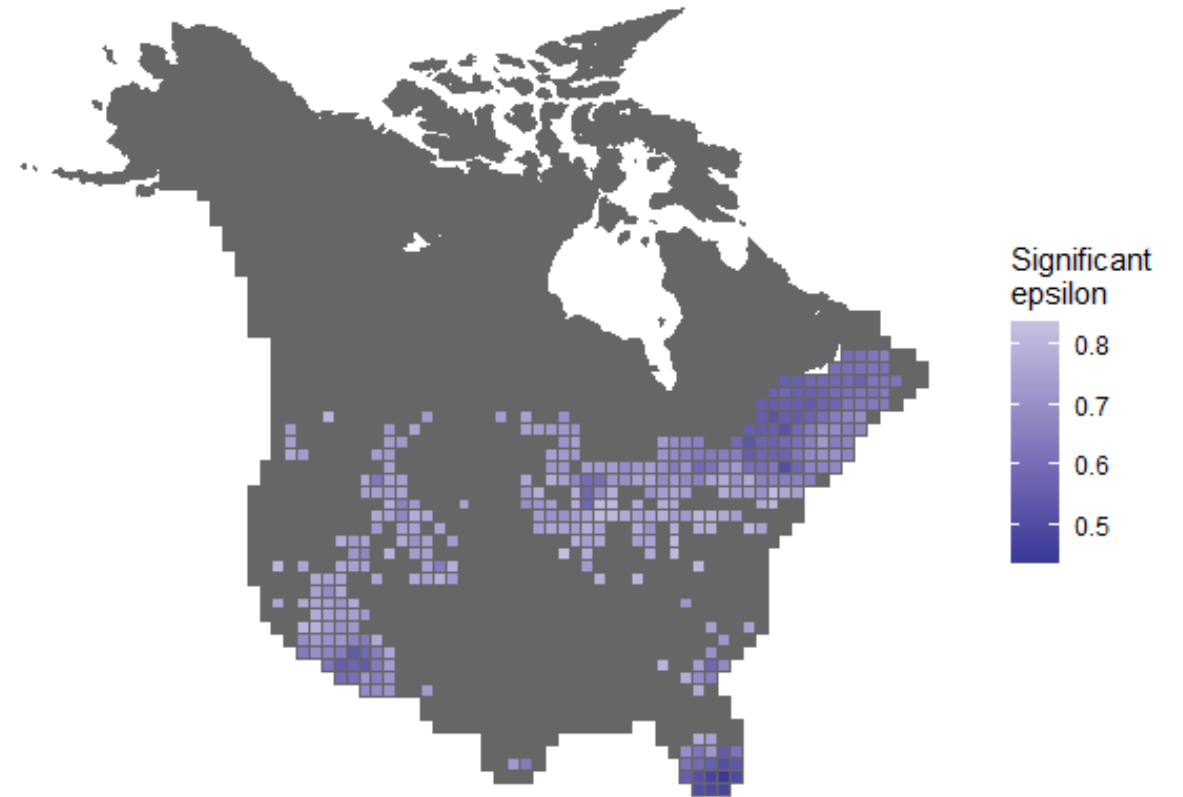
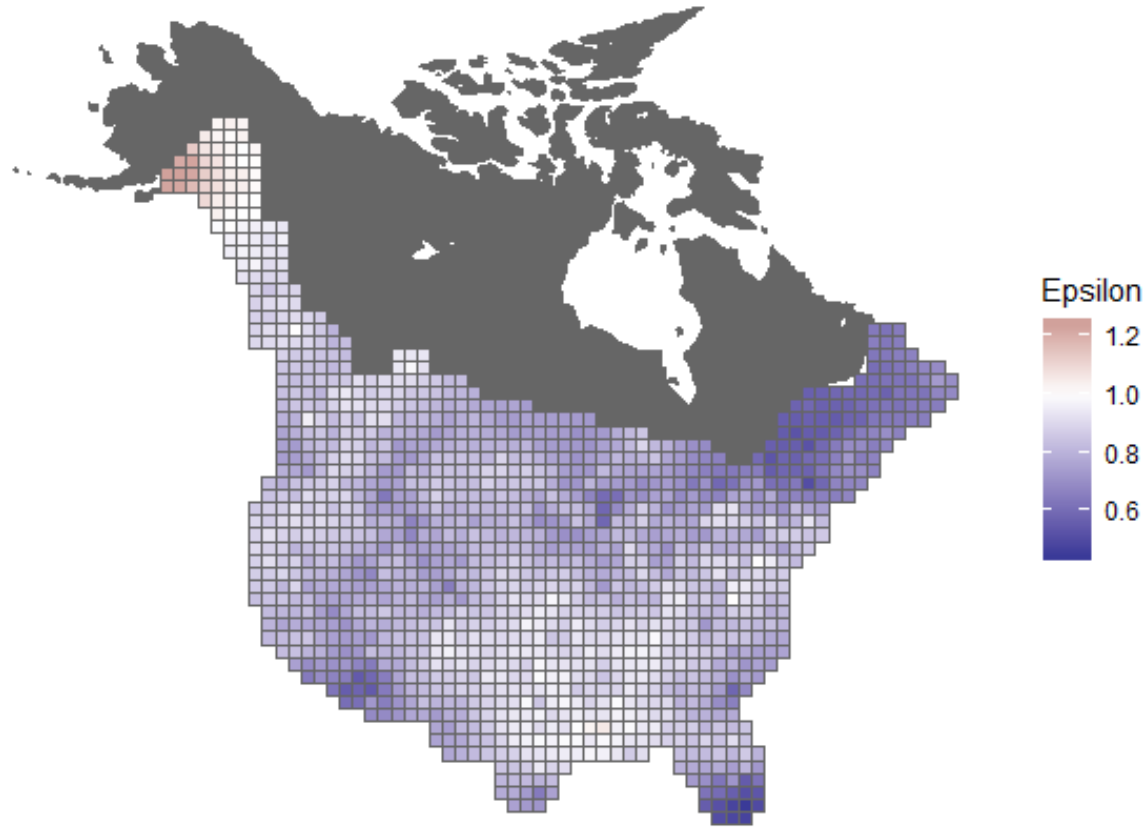
Fixed effect:	0.025quant	0.5quant	0.975quant
Global intercept	0.602	0.772	0.941
Global log_hrs effect	0.774	0.815	0.857
Global std_yr effect	0.022	0.025	0.027

Hyperparameter:	0.025quant	0.5quant	0.975quant
1/Overdispersion	0.444	0.449	0.453
Precision alpha CAR	0.673	0.688	0.703
Precision epsilon CAR	33.055	33.691	34.340
Precision tau CAR	678.391	707.756	738.433
Precision kappa iid	0.927	0.969	1.012

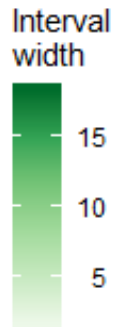
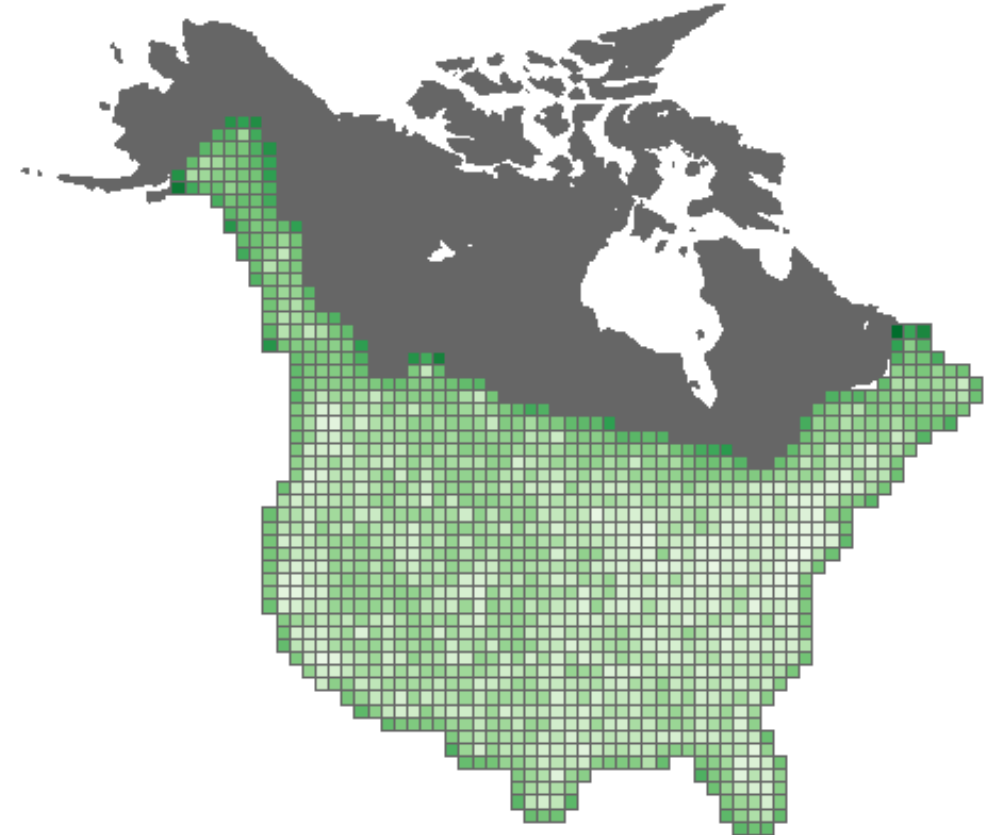
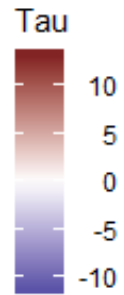
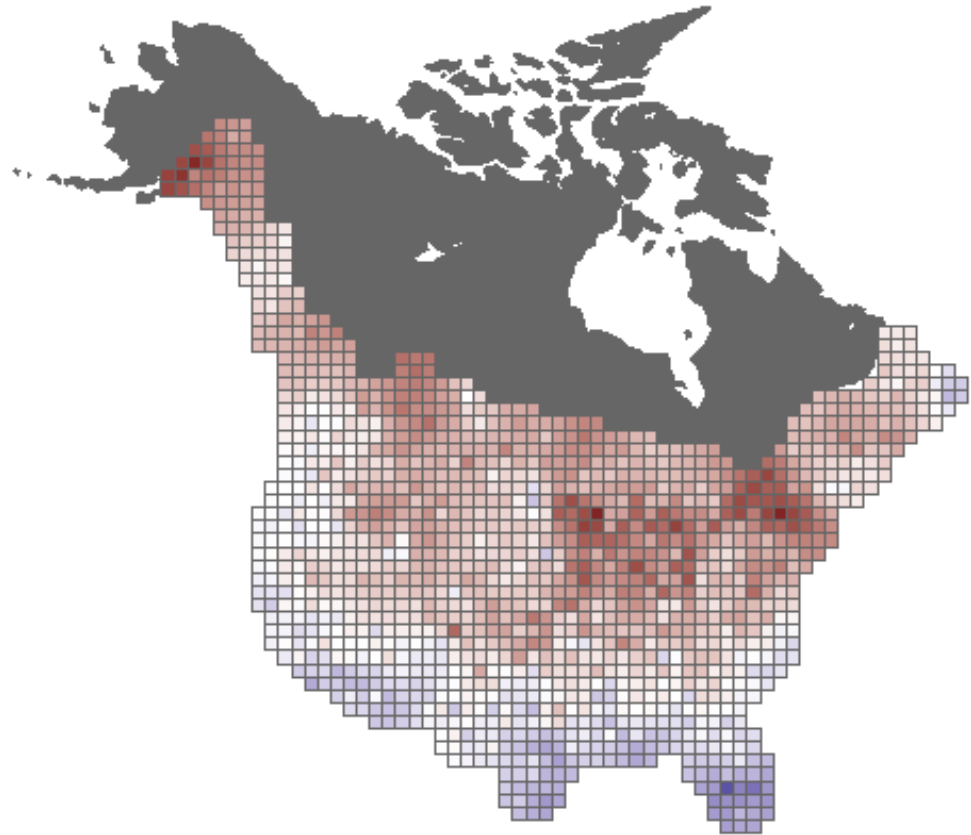
Alpha: 2017 Abundance Indices



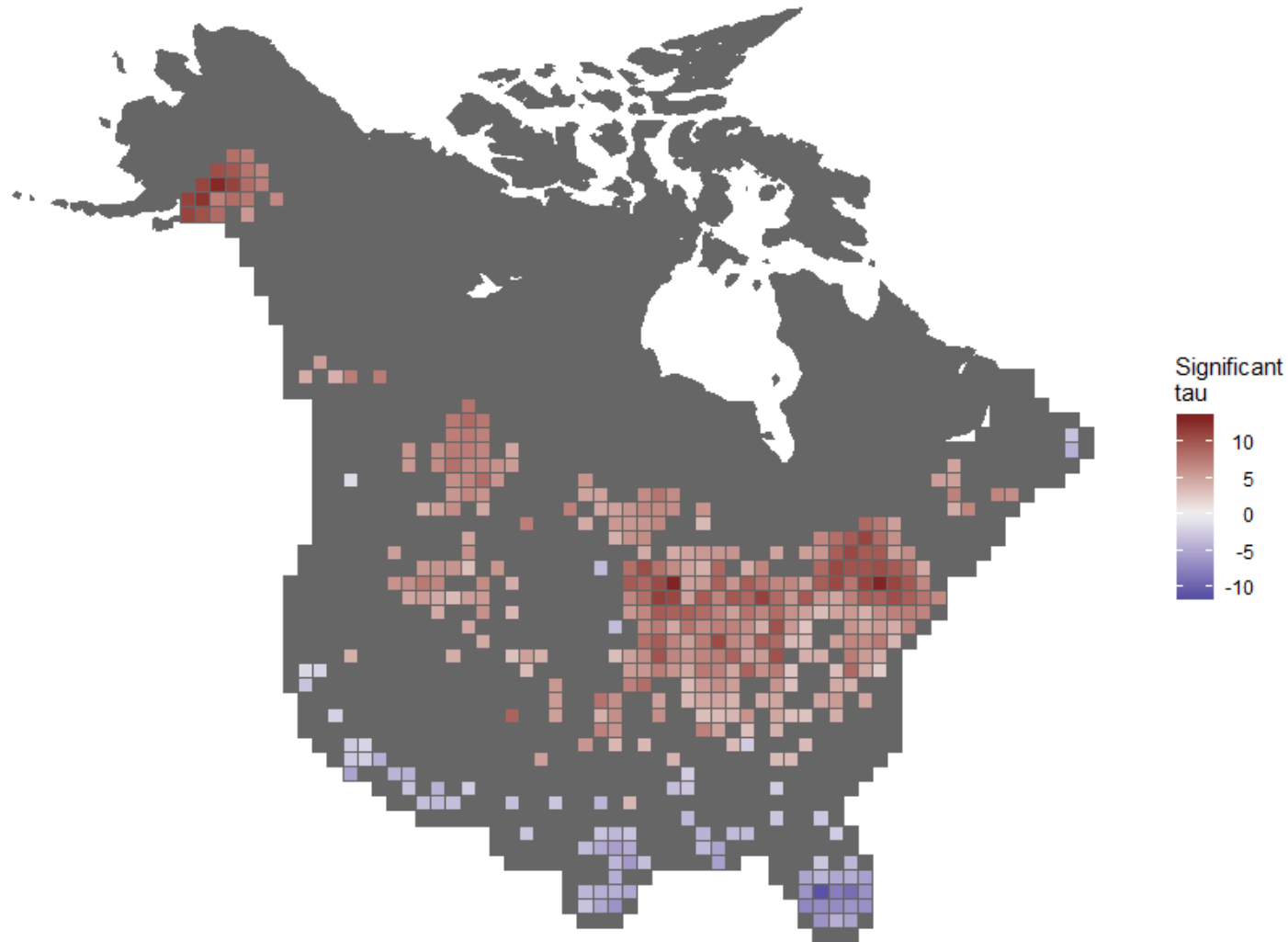
Epsilon: Effort Effects



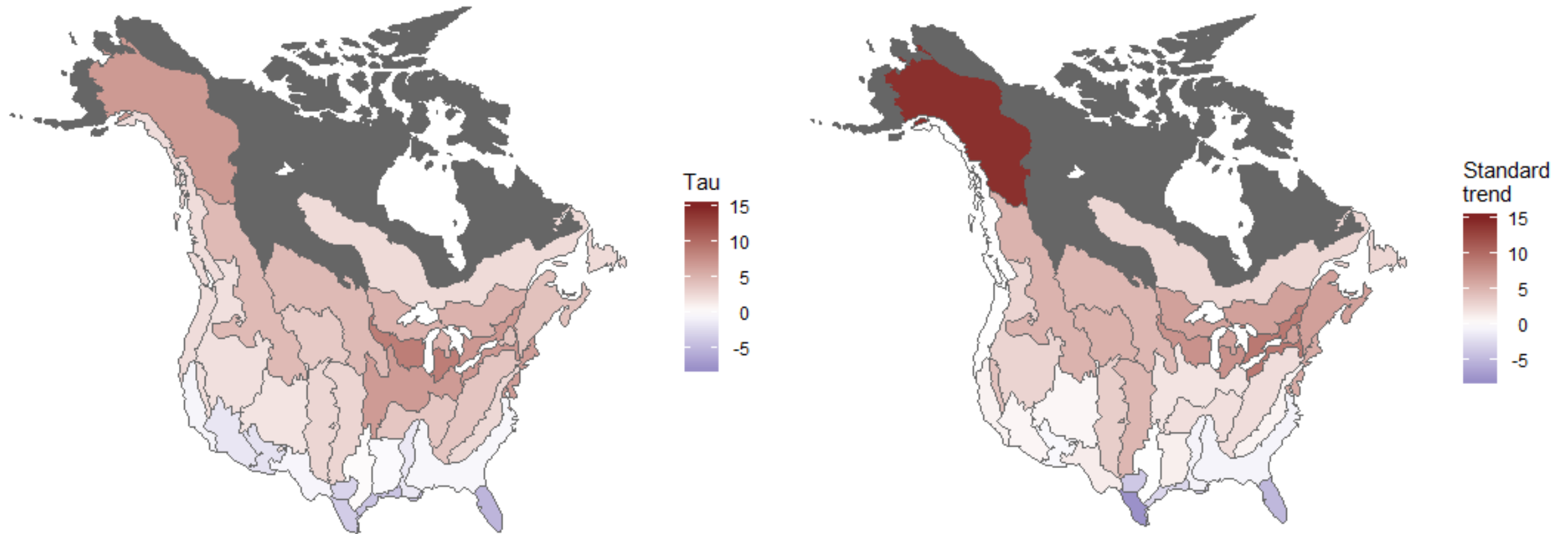
Tau: Year Effects



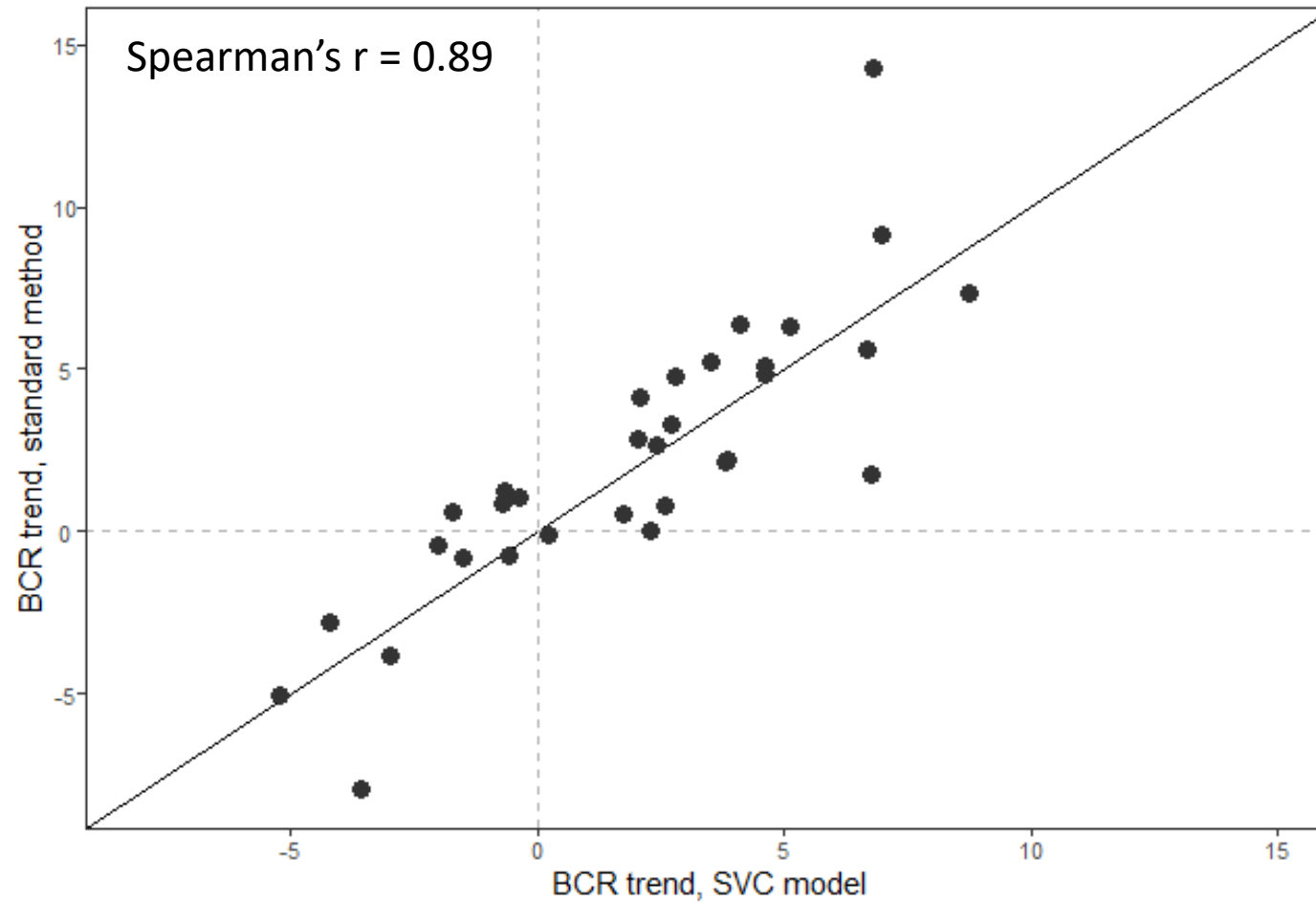
Tau: Significant Year Effects



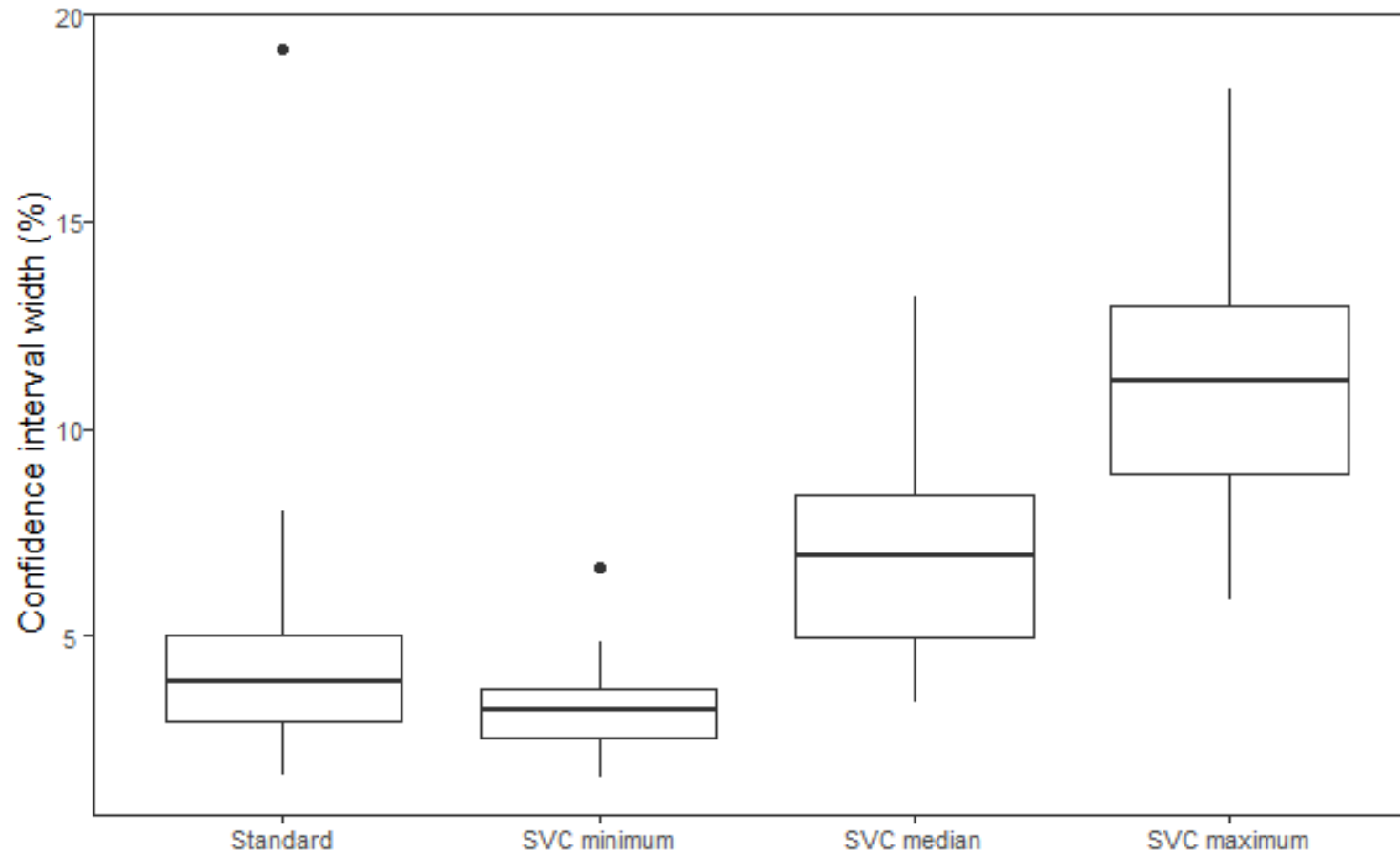
Aggregation: Trend Estimates Compared



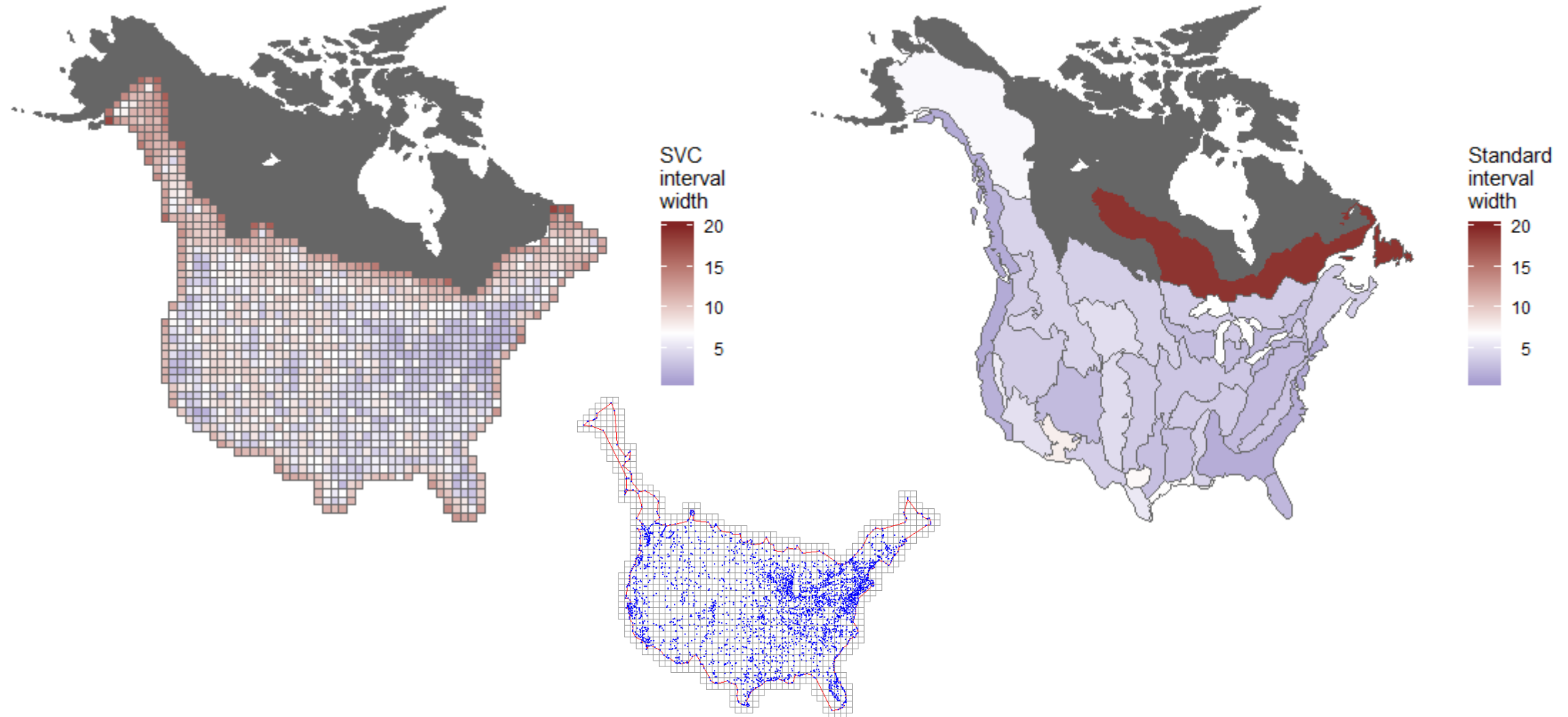
Aggregation: Trend Estimates Compared



Aggregation: Trend Precision Compared



Aggregation: Trend Precision Compared





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