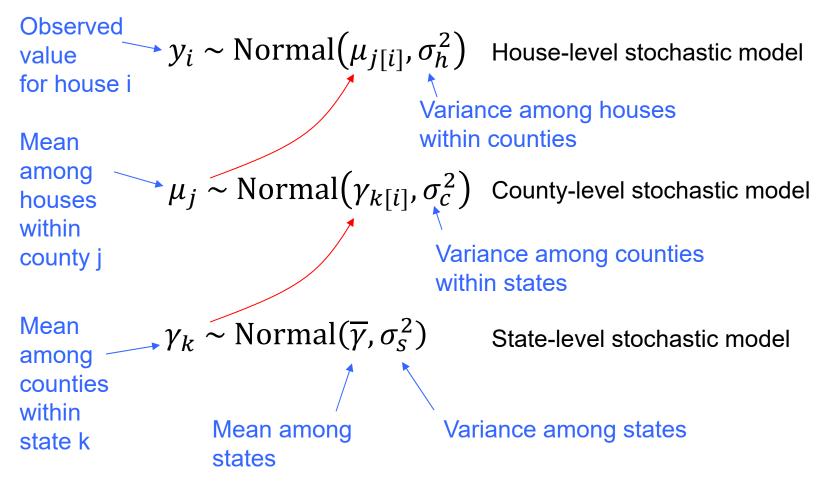
Multiple nested scales

Multilevel model, with 3 nested levels



Multiple nested scales

Alternative parameterization

$$y_i = \beta_0 + s_{k[i]} + c_{j[i]} + e_i \qquad \text{Residual: deviation of house i from county j mean}$$
 Overall Deviation of state k mean county j mean from overall mean from states mean state k mean
$$s_k \sim \text{Normal}(0, \sigma_s^2) \qquad \text{Variance among counties}$$

$$c_j \sim \text{Normal}(0, \sigma_c^2) \qquad \text{Variance among counties}$$
 within states

 $e_i \sim \text{Normal}(0, \sigma_e^2)$ Variance among houses

within counties

Multiple nested scales

Linear model syntax for Imer

If states and counties uniquely identified in data:

$$y \sim 1 + (1|state) + (1|county)$$

Otherwise you can form unique identifiers with the nested operator:

```
y \sim 1 + (1|state/county)
```

which expands to (you can also use this):

```
y \sim 1 + (1|state) + (1|state:county)
```

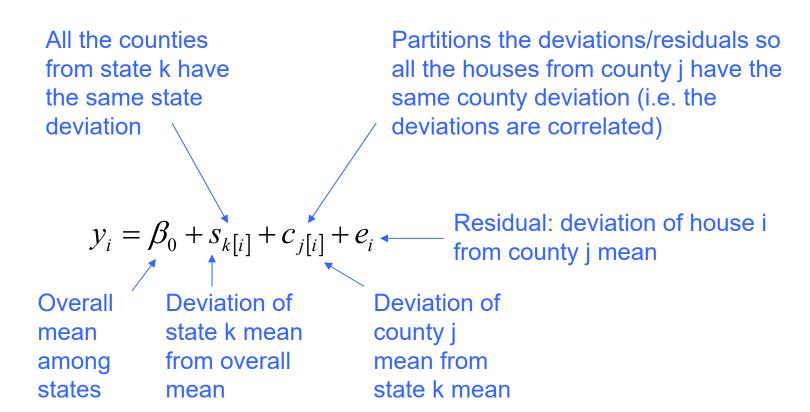
thus, the interaction state: county uniquely identifies all counties in the dataset

Autocorrelation

- When present: overoptimistic about the amount of information in the data
- The grouping terms account for autocorrelation (units within groups are more correlated than between groups)
- Specifically, their deviations are correlated.
 The estimated group deviations (i.e.
 "random effects") properly partition the
 correlation into scales of variation

Autocorrelation

How the multilevel model accounts for autocorrelation



LMMs

- Linear mixed models
 - aka linear mixed effects models
- Class of multilevel models
- Mixed = fixed + random effects
 - aka deterministic + stochastic effects
 - aka no pooling + partial pooling
- Fixed effects are linear
- Random effects are all Normal

GLMMs

- Generalized linear mixed models
- Fixed effects have linear predictor but link function can be nonlinear (e.g. log, logistic)
- Data level: exponential family distribution
 - e.g. Poisson, Binomial, Exponential, Normal
- Upper levels: Normal distribution