Thursday

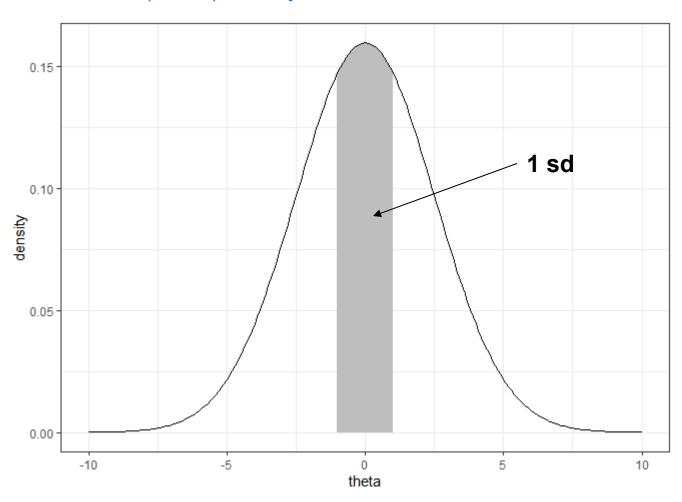
- Individual projects
 - see my initial comment in your repo
 - add more details if necessary (see HW)
 - individual meetings (5 mins)

Today

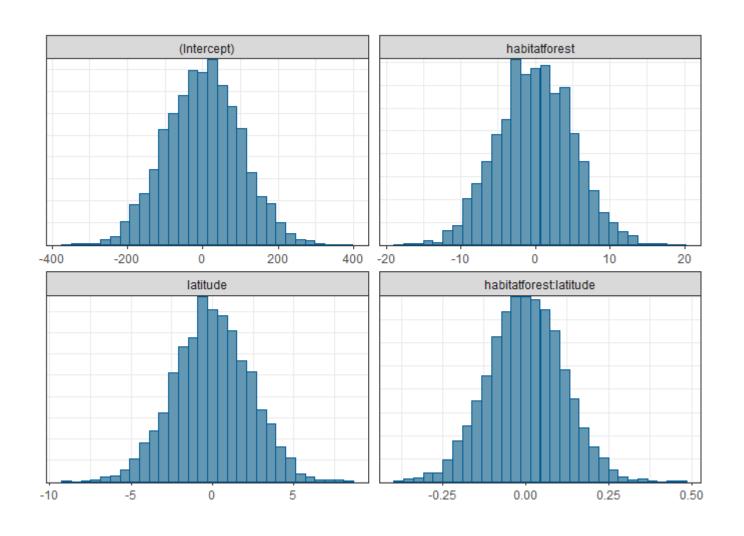
- Priors
- Bayesian model check
- Multilevel models (radon case study)
 - scales and groups
 - study design
 - models and code

Priors

Normal(0, 2.5) weakly informative



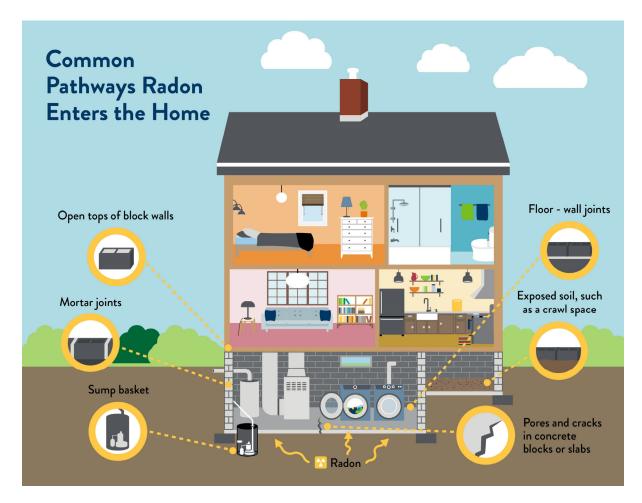
Prior predictive distribution



Bayesian model check

See script

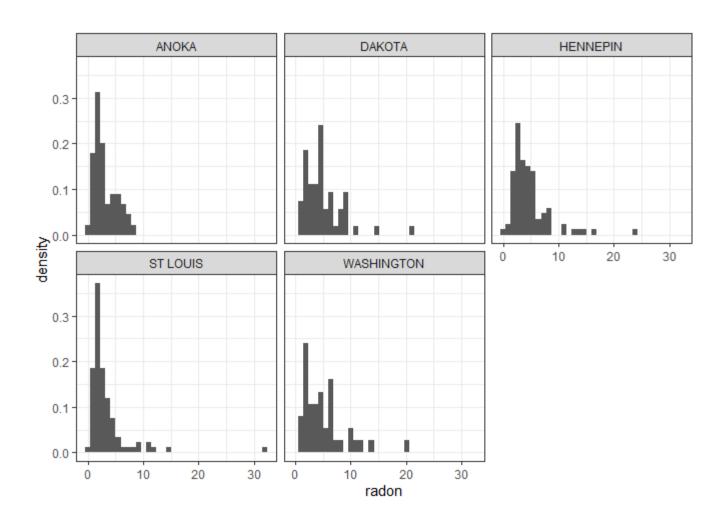
Radon case study



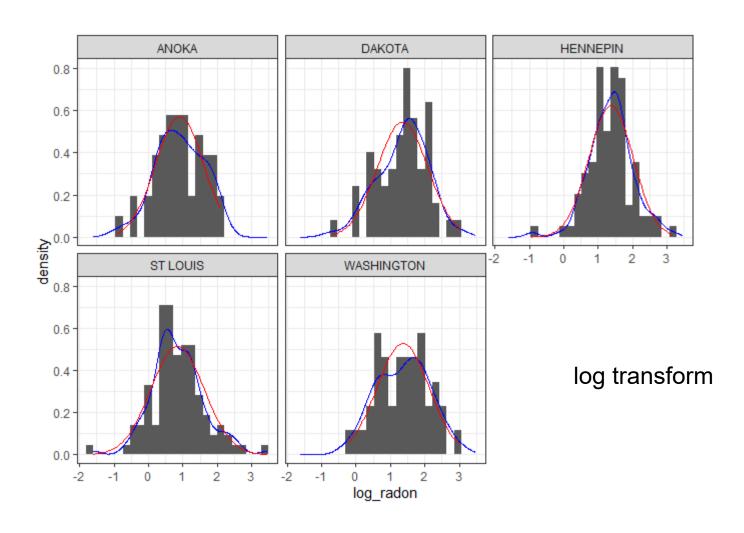
Expect: basement more susceptible

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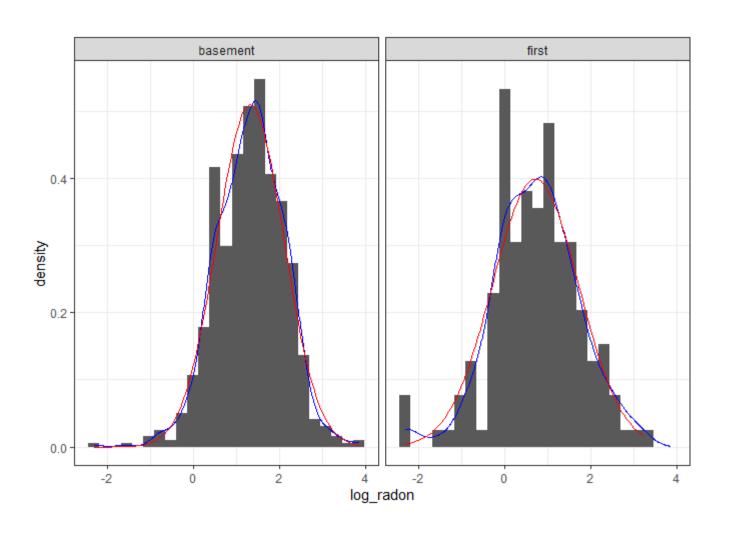
EDA – informal modeling



EDA – informal modeling



EDA – informal modeling



Art of multilevel models

- Sketch data design
- Sketch the data generating process
- Math equations
- Linear model syntax

Scales in radon data

- You identified these scales:
- State, county, house, floor, sensor
- Hierarchical structure: nested
- Data unit: measurement per sensor
- Potential groups: state, county, house, floor, sensor ... but 1 sample per house so house is smallest (unit scale)

Scope of inference

- Houses in Minnesota ... potentially restricted by whatever criteria were used to include houses in the study
- If random sample of houses then scope would be houses generally in Minnesota
- It clearly wasn't a random sample, so warning: results might be biased
- Limited to basement and 1st floor

Sketch

State = extent scope of inference

County (- grouping variable

House

Floor breaked

Not a grouping

Not a grouping

OF houses in each

variable since

no replicate hors

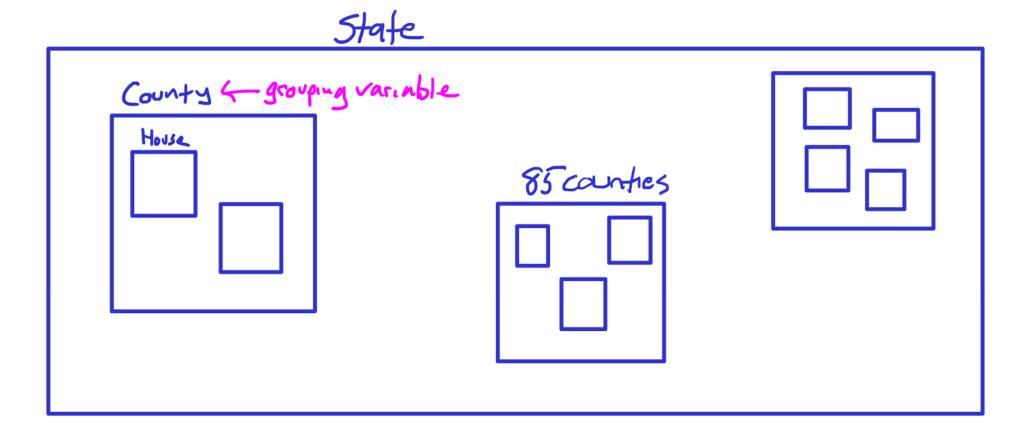
ensors.

Fixed vs random

- Floor: fixed effect
 - deterministic model for parameters
 - shared characteristic
- County: random effect
 - stochastic model for parameters
 - model by group (parameters vary by group)
 - model accounts for correlation among houses within counties

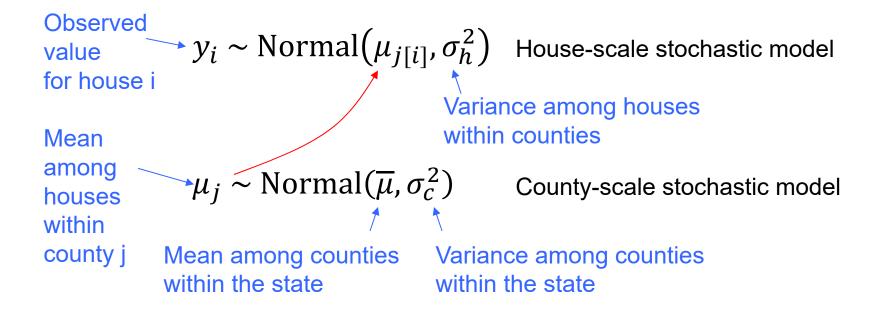
We could alternatively model county as a fixed effect: model parameters would be deterministic

Model 1: random effects



Writing model 1

Basic multilevel model, no predictors



j[i] is the county (j) of house i

Simulate model 1

algorithm

```
set mu_bar
set sigma_c
set sigma_h
set n_c #number of counties
set n_h #number of houses

for j in 1:n_c
    mu_j = normal(mu_bar, sigma_c)

for i in 1:n_h
    y_i = normal(mu_j[i], sigma_h)
```

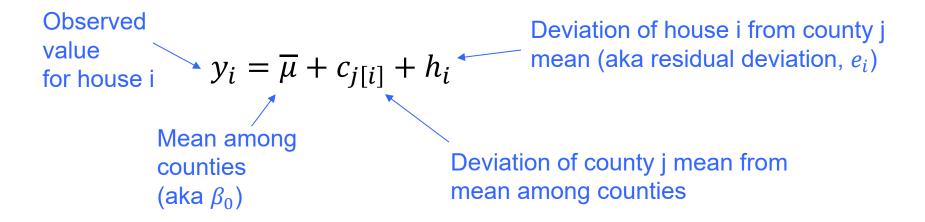
Data story:

First, county determines background radiation level

Then, house determines inside radiation level

Alt parameterization

Additive decomposition



$$c_j \sim {
m Normal}(0, \sigma_c^2)$$
 Variance among counties stochastic model $h_i \sim {
m Normal}(0, \sigma_h^2)$ Variance among houses within counties stochastic model

Sim alt parameterization

algorithm

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for j in 1:n_c
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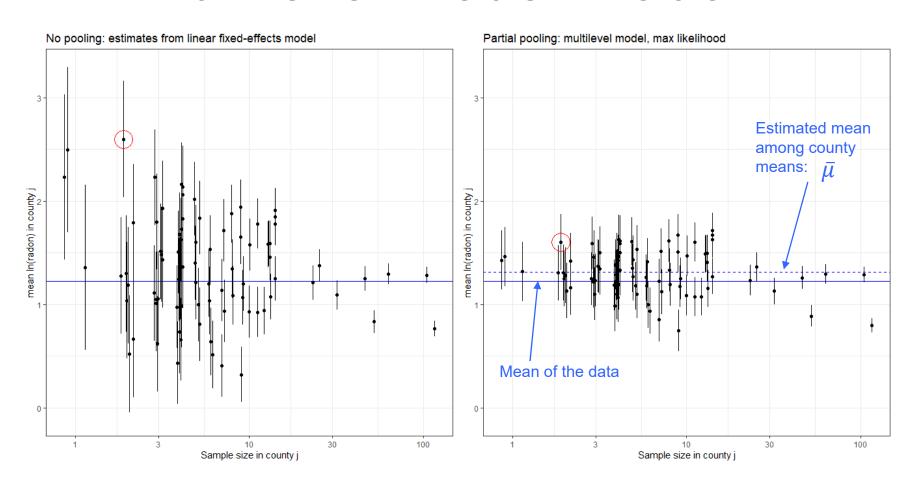
for i in 1:n_h
    h_i = normal(0, sigma_h)
    y_i = mu_bar + c_j[i] + h_i
```

Linear model syntax

```
log_radon \sim 1 + (1|county)
```

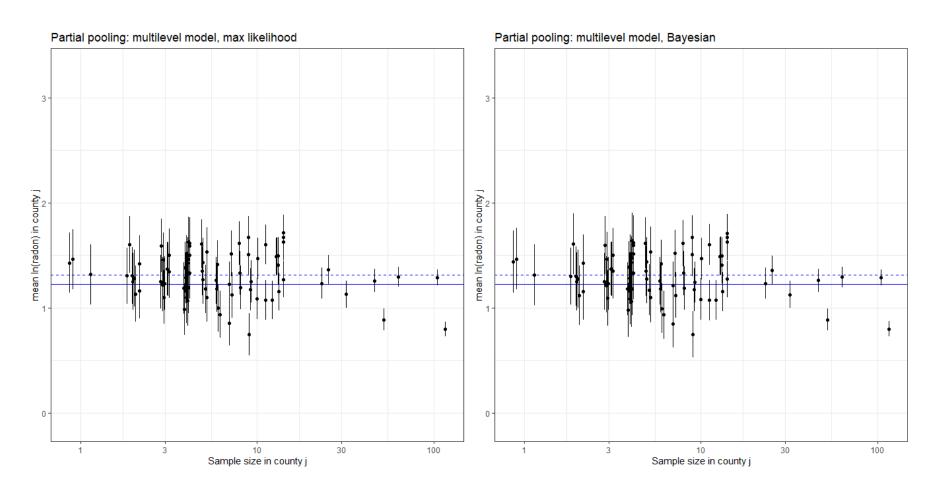
Code for the radon example radon_multilevel_EDA.md radon_multilevel_M1.md

Multilevel model - radon



Shrinkage

Multilevel model - radon



Maximum likelihood

Bayesian