Today

- Recap & questions from homework
- Pair programming

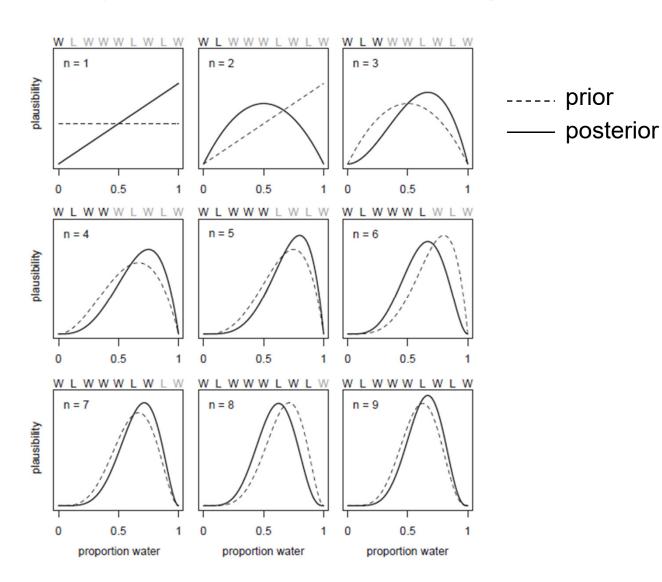
GitGUI & Gitk

- MacOS
 - install via homebrew
 - https://git-scm.com/downloads/mac
 - -\$ brew install git-gui

Main concepts McElreath 2

- Golem = algorithm
- Small world (model) vs large world (reality)
- Likelihood: counting all the ways data could have happened
- Bayesian updating: prior x likelihood
 - using counts (marbles)
 - using probabilities (marbles; divide by sum of numerator)
 - using distributions (p_water via Earth toss)

Bayesian updating



Pair programming

- Chapter 2
 - Coding Bayesian updating
 - Questions: 2M1, 2M2

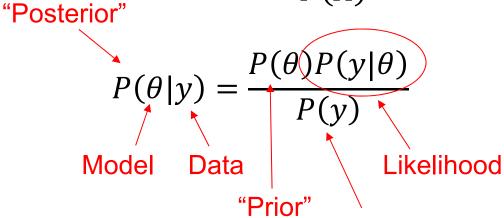
Components of model

- 1) Likelihood
 - "data story" = data generating process
 - from first principles, or "off the shelf"
- 2) Parameters
 - quantities that don't change
 - to be estimated
- 3) Prior distribution
- 4) Posterior distribution (inference)
 - histogram is the posterior

Bayesian inference

$$P(B|A) = \frac{P(B)P(A|B)}{P(A)}$$

Bayes' rule for two events A, B



Apply Bayes' rule to convert the likelihood into what we really want to know: the probability of the model given the data

Total probability of the data

P(y): probability added up or integrated over all of the models

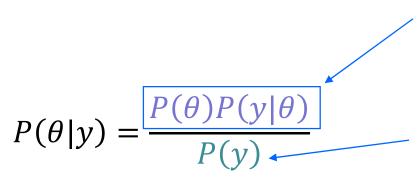
$$P(y) = \sum_{\theta} P(\theta) P(y|\theta)$$

Discrete parameter

$$P(y) = \int P(\theta)P(y|\theta) d\theta$$

Continuous parameter

Bayesian inference



Unstandardized posterior

Total probability standardizes the posterior to a probability (discrete parameter) or probability density (continuous parameter)

$$P(y) = \sum_{\theta} P(\theta)P(y|\theta)$$

Discrete parameter

$$P(y) = \int P(\theta)P(y|\theta) d\theta$$

Continuous parameter

Posterior distribution algorithm

posterior distribution

Discrete parameter

Algorithm

load data

for each parameter value

unstandardized posterior = prior * likelihood

total probability = sum(unstandardized posteriors)

for each parameter value

posterior probability =

unstandardized posterior / total probability

plot posterior probability vs parameter values

posterior distribution

$$P(y) = \sum_{\theta} P(\theta) P(y|\theta)$$

probability mass function (pmf)

 $P(\theta|y) = \frac{P(\theta)P(y|\theta)}{P(y)}$

Discrete parameter

- Uncommon
- Usually a latent (hidden) state
- Examples
 - male/female in population model
 - occupied/unoccupied in species occupancy model

Continuous parameter

```
P(\theta|y) = \frac{P(\theta)P(y|\theta)}{P(y)}
Exact algorithm
load data
define grid of parameter values
for each parameter value
  unstandardized posterior = prior * likelihood
total probability = integral(unstandardized posterior function)
for each parameter value
                                       area under curve
  posterior Pr density =
     unstandardized posterior / total probability
plot posterior Pr density vs parameter values ___ posterior
                                                      distribution
```

$$P(y) = \int P(\theta)P(y|\theta) d\theta$$
 probability density function (pdf)

Continuous parameter

 $P(\theta|y) = \frac{P(\theta)P(y|\theta)}{P(y)}$ Grid approximation algorithm load data define grid of parameter values with resolution r for each parameter value unstandardized posterior = prior * likelihood total probability = sum(unstandardized posteriors) * r for each parameter value approximate posterior Pr density = area under curve unstandardized posterior / total probability plot posterior Pr density vs parameter values ___ posterior distribution numerical integration $P(y) = \int P(\theta)P(y|\theta) d\theta$ probability density function (pdf)

