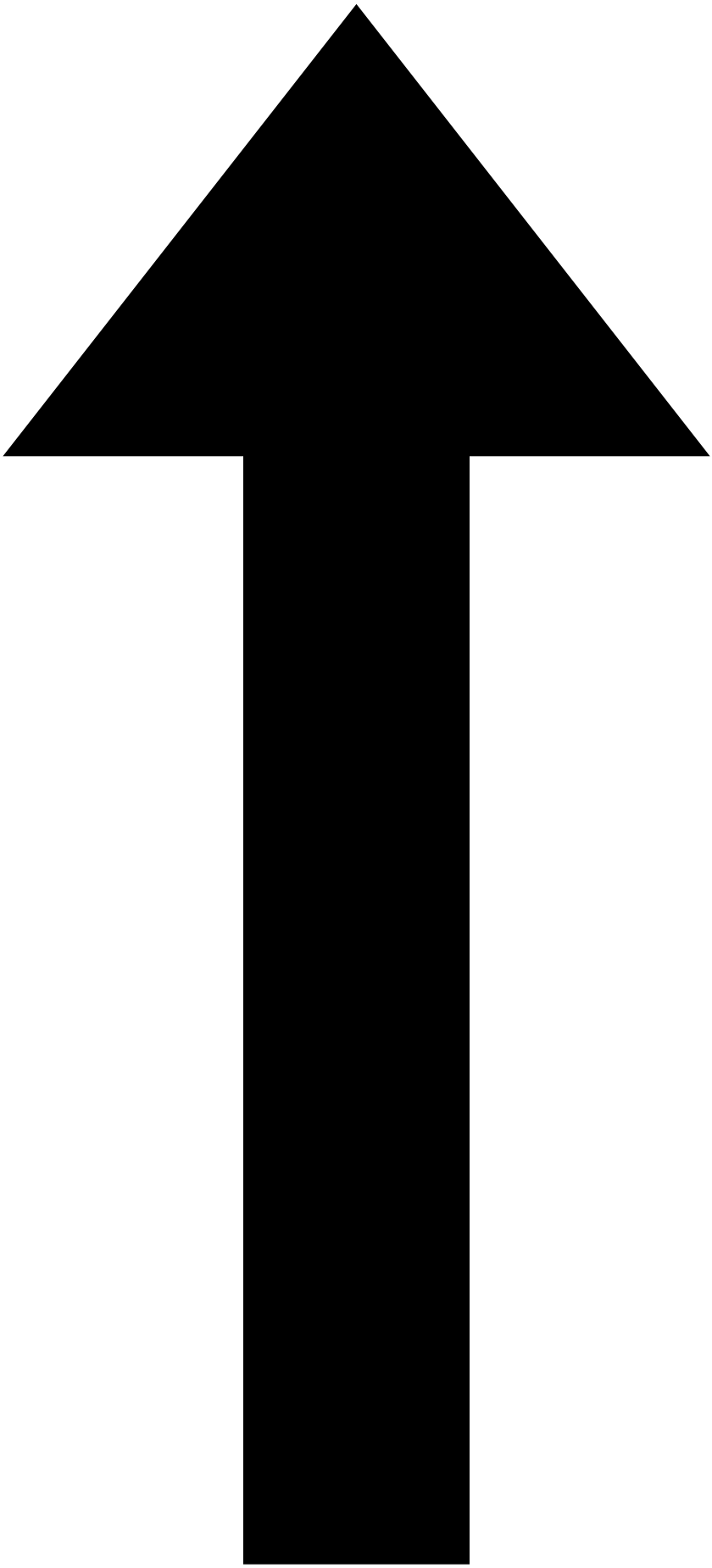
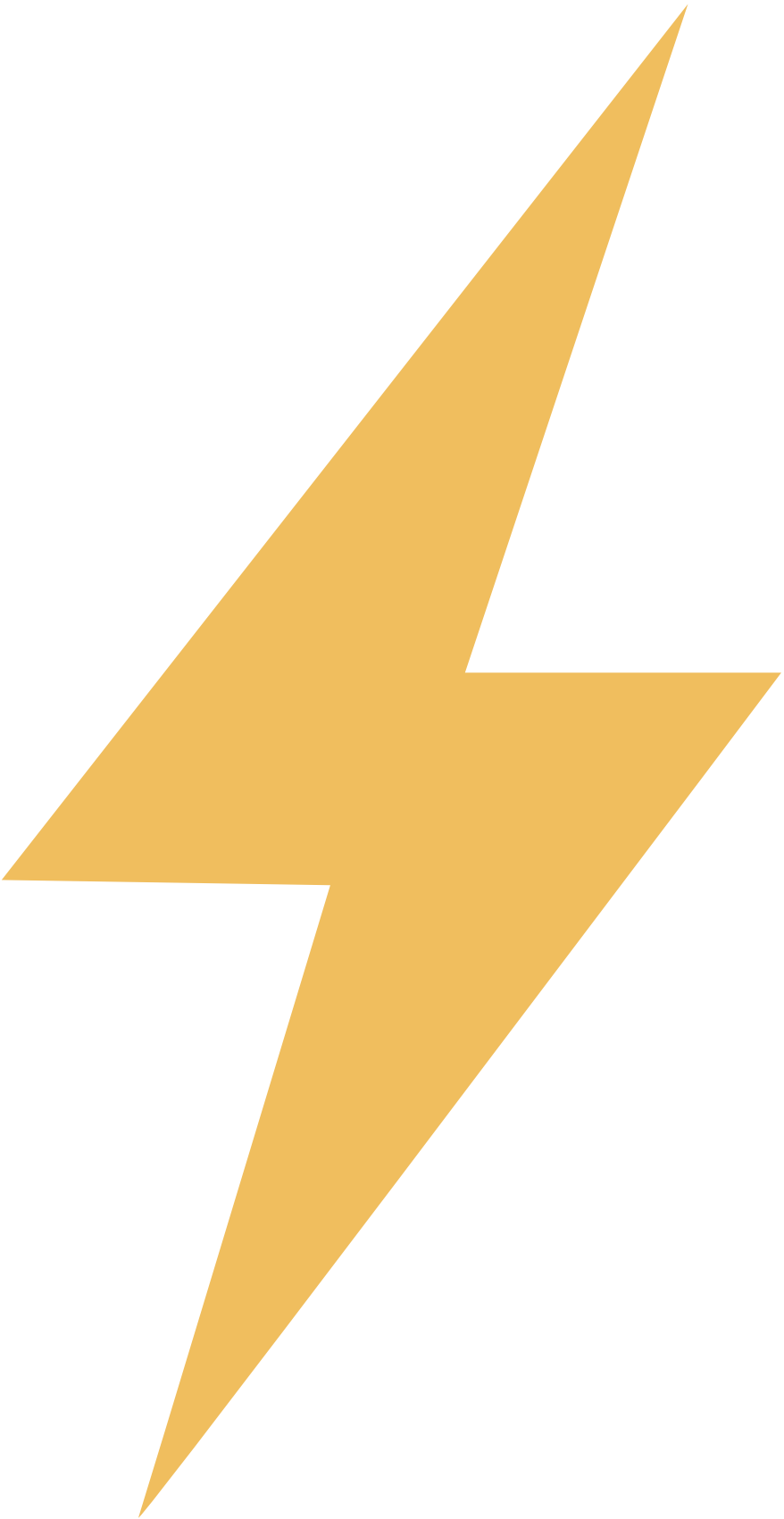
Posterior samples

```
> samples
# A tibble: 2,000 × 3
             a           b      sigma
   <dbl[1d]> <dbl[1d]> <dbl[1d]>
 1      115.     0.889       4.78
 2      109.     1.02        5.30
 3      112.     0.928       5.07
 4      111.     0.949       5.30
 5      111.     0.955       5.04
 6      115.     0.872       5.19
 7      109.     1.01        5.13
 8      117.     0.844       5.00
 9      115.     0.882       4.94
10      112.     0.939       4.95
# … with 1,990 more rows
```






FIT!

$$y_i \sim Normal(\mu_i, \sigma)$$

$$\mu_i = \alpha + \beta x_i$$

$$\alpha \sim Normal(0, 20)$$

$$\beta \sim lognormal(0, 1)$$

$$\sigma \sim Exponential(1)$$

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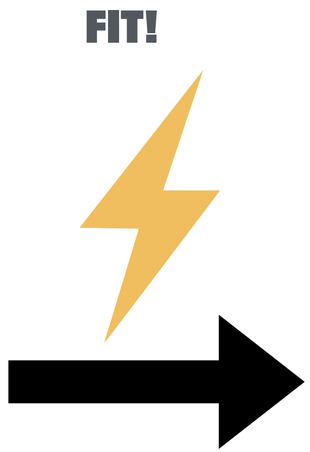


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```

Posterior mean estimates

$$E_y[\theta] =$$

```
> colMeans(samples)
          a           b       sigma
112.9296580   0.9253803   5.0453651
```