

# Ch. 1 - Model comparisons

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## Contents

<b>Model Comparisons</b>	<b>1</b>
Model 1 - CFAR ~ GSH . . . . .	1
Model 2 - CFAR ~ GSH * Max Size . . . . .	4
Model 3 - CFAR ~ GSH + Max Size (No interaction) . . . . .	6
Model 4 - CFAR ~ GSH + (1 Max Size) (No interaction) . . . . .	9
Model 5 - CFAR ~ GSH + Phylogeny . . . . .	11
Model 5 - CFAR ~ GSH * Max Size + Phylogeny . . . . .	13
Model 6 - CFAR ~ GSH + Max Size + Phylogeny . . . . .	17

## Model Comparisons

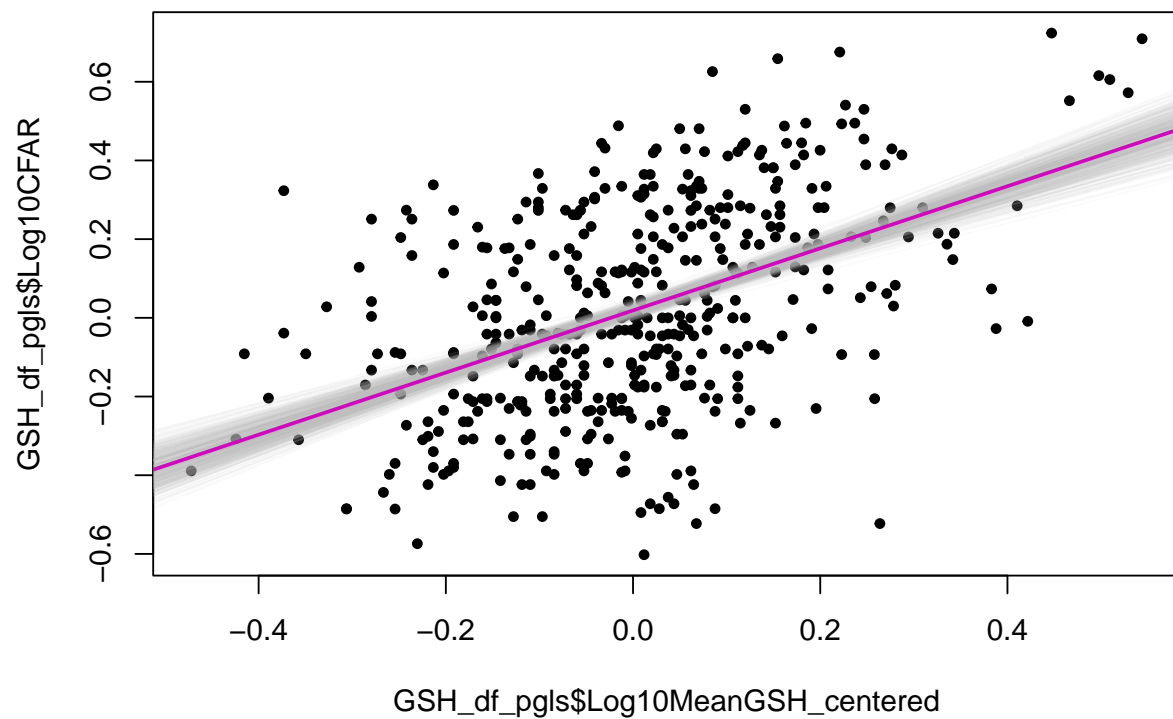
### Model 1 - CFAR ~ GSH

brms model

```
Model_plain <- brm(Log10CFAR ~ Log10MeanGSH_centered,  
  data = GSH_df_pgls,  
  family = gaussian(),  
  prior = Plain_prior  
)
```

```
fixef(Model_plain, summary = T)
```

```
##              Estimate Est.Error      Q2.5      Q97.5  
## Intercept          0.01878724 0.01121401 -0.003048758 0.04068312  
## Log10MeanGSH_centered 0.78943177 0.06909079  0.653733169 0.92180824
```



### STAN model

Data considered

```
data {
  int <lower=1> N;
  vector[N] x;
  vector[N] y;
}
```

The parameters accepted by the model

```
parameters {
  real alpha;
  real beta;
  real<lower=0> sigma;
}
```

Model form:

```
model {
  sigma ~ student_t(3, 0, 10);
```

```

y ~ normal(alpha + x * beta , sigma);
}

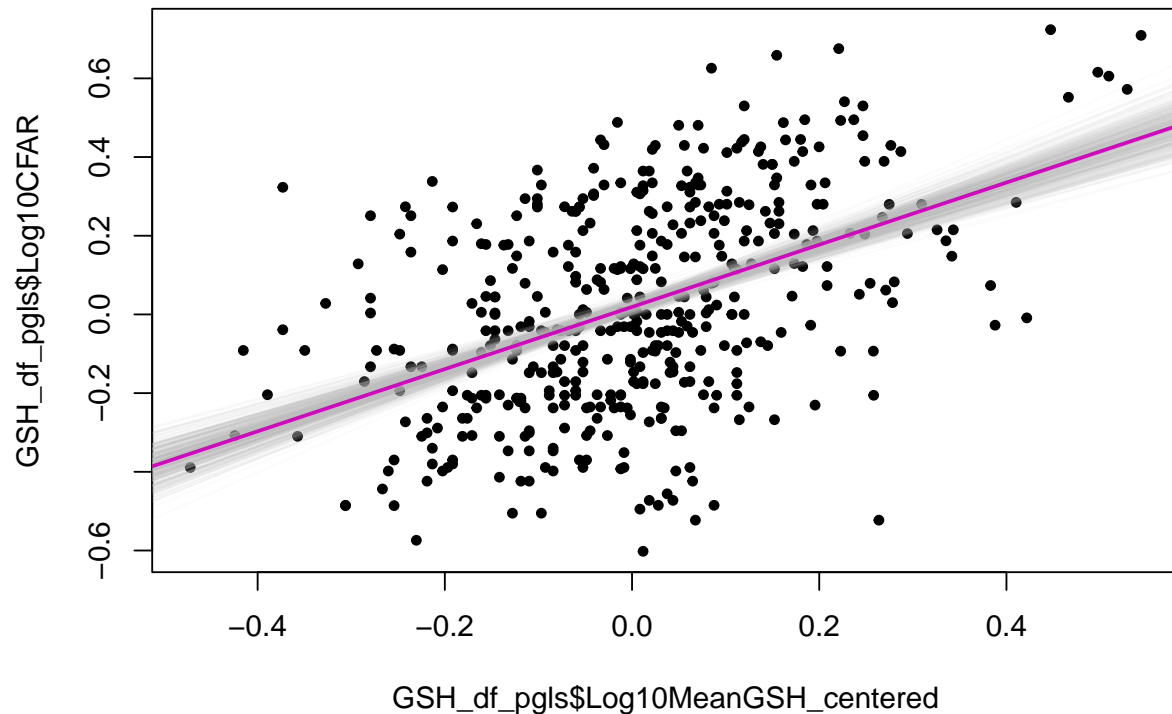
```

```
fit1
```

```

## Inference for Stan model: StanModel1.
## 4 chains, each with iter=1000; warmup=500; thin=1;
## post-warmup draws per chain=500, total post-warmup draws=2000.
##
##          mean se_mean   sd  2.5%  25%  50%  75%  97.5% n_eff Rhat
## alpha    0.02    0.00 0.01   0.00  0.01  0.02  0.03   0.04 2098   1
## beta     0.79    0.00 0.07   0.66  0.75  0.79  0.84   0.93 1707   1
## sigma    0.24    0.00 0.01   0.22  0.23  0.24  0.24   0.25 1951   1
## lp__    429.89    0.03 1.19 426.78 429.34 430.19 430.76 431.27 1176   1
##
## Samples were drawn using NUTS(diag_e) at Wed Mar 02 11:19:28 2022.
## For each parameter, n_eff is a crude measure of effective sample size,
## and Rhat is the potential scale reduction factor on split chains (at
## convergence, Rhat=1).

```



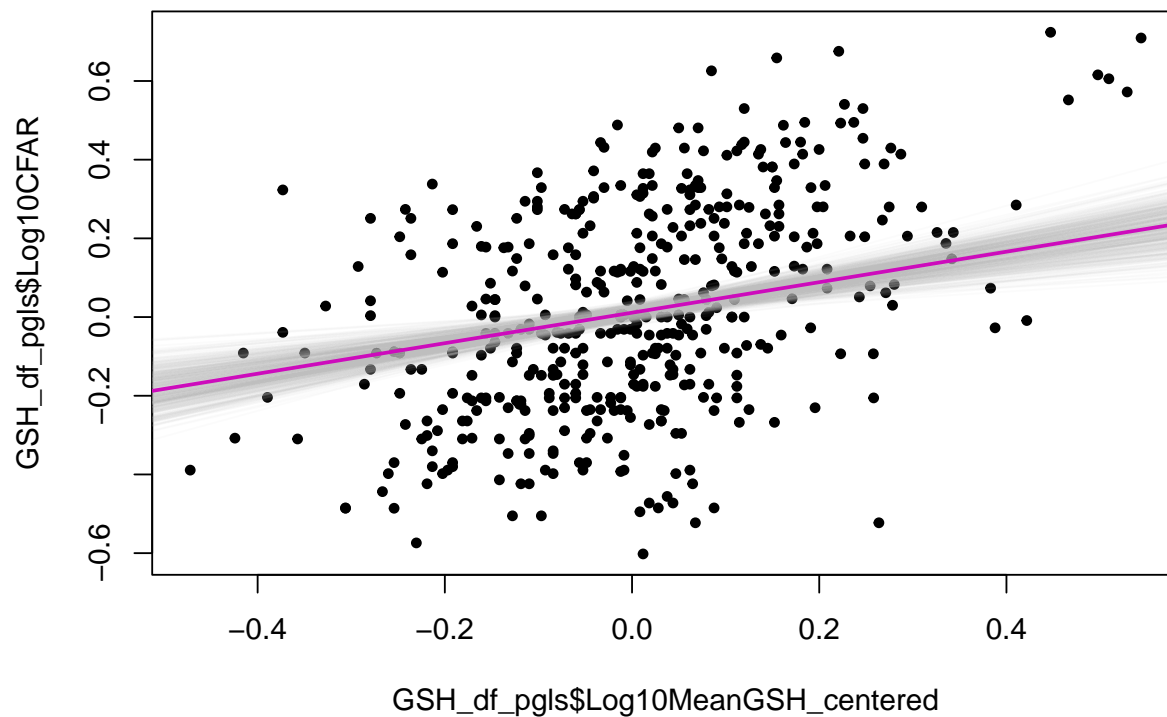
## Model 2 - CFAR ~ GSH \* Max Size

brms model

```
Model_size <- brm(Log10CFAR ~ Log10MeanGSH_centered * Log10MaxSize_centered,  
  data = GSH_df_pgls,  
  family = gaussian(),  
  prior = Size_prior  
)
```

```
fixef(Model_size, summary = T)
```

##	Estimate	Est.Error	Q2.5
## Intercept	0.01107801	0.01158262	-0.01069365
## Log10MeanGSH_centered	0.38761087	0.08262037	0.22932179
## Log10MaxSize_centered	0.30460627	0.04281846	0.22222482
## Log10MeanGSH_centered:Log10MaxSize_centered	0.26628115	0.16406086	-0.05284798
##	Q97.5		
## Intercept	0.03436049		
## Log10MeanGSH_centered	0.55212619		
## Log10MaxSize_centered	0.38716498		
## Log10MeanGSH_centered:Log10MaxSize_centered	0.58368860		



## STAN model

Data considered

```
data {  
  int <lower=0> N; // number of data points  
  int <lower=0> K; // number of predictors  
  matrix[N,K] x; // predictor matrix  
  vector[N] y; // CFAR  
}
```

The parameters accepted by the model.

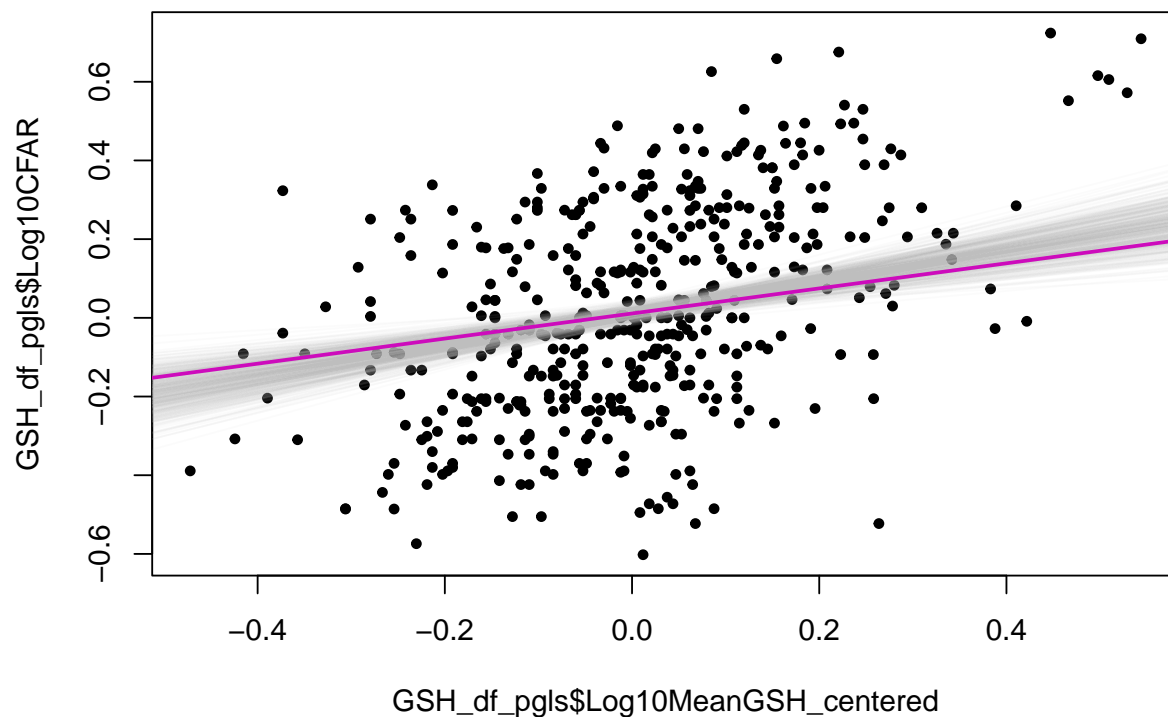
```
parameters {  
  real alpha;  
  vector[K] beta;  
  real<lower=0> sigma;  
}
```

The model to be estimated. We model the output ‘y’ to be normally distributed with mean ‘mu’ and standard deviation ‘sigma’.

```
model {  
  
  sigma ~ student_t(3, 0, 10);  
  
  y ~ normal(alpha + x * beta , sigma);  
}
```

fit2

```
## Inference for Stan model: StanModel2.  
## 4 chains, each with iter=1000; warmup=500; thin=1;  
## post-warmup draws per chain=500, total post-warmup draws=2000.  
##  
##           mean se_mean   sd  2.5%  25%   50%   75%  97.5% n_eff Rhat  
## alpha      0.01    0.00 0.01  -0.01   0.00   0.01   0.02   0.04  1841   1  
## beta[1]    0.39    0.00 0.08   0.23   0.33   0.39   0.44   0.55  1758   1  
## beta[2]    0.31    0.00 0.04   0.22   0.28   0.31   0.33   0.39  1742   1  
## beta[3]    0.26    0.00 0.16  -0.06   0.16   0.27   0.37   0.59  1975   1  
## sigma      0.22    0.00 0.01   0.21   0.22   0.22   0.23   0.24  2516   1  
## lp__      459.42    0.05 1.62 455.40 458.60 459.75 460.60 461.57   884   1  
##  
## Samples were drawn using NUTS(diag_e) at Wed Mar 02 11:20:54 2022.  
## For each parameter, n_eff is a crude measure of effective sample size,  
## and Rhat is the potential scale reduction factor on split chains (at  
## convergence, Rhat=1).
```



### Model 3 - CFAR $\sim$ GSH + Max Size (No interaction)

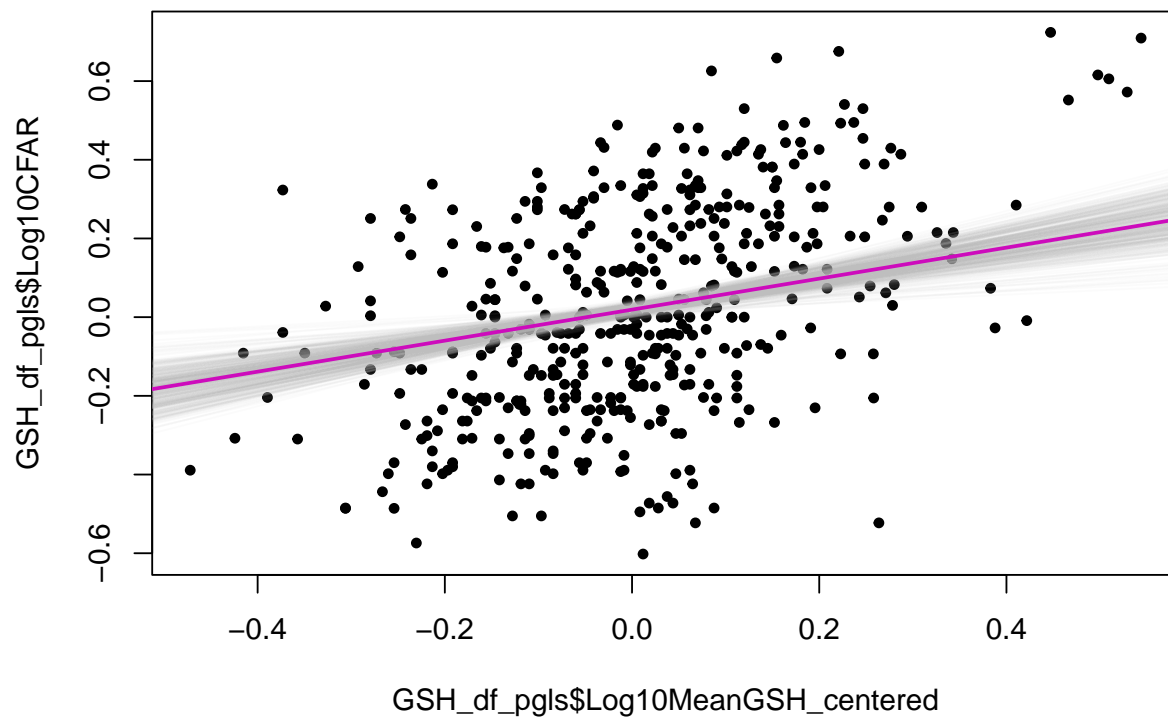
brms model

```
Model_size_2 <- brm(Log10CFAR ~ Log10MeanGSH_centered + Log10MaxSize_centered,
  data = GSH_df_pgls,
  family = gaussian(),
  prior = Size_prior_2)
```

```
summary(Model_size_2)
```

```
## Family: gaussian
## Links: mu = identity; sigma = identity
## Formula: Log10CFAR ~ Log10MeanGSH_centered + Log10MaxSize_centered
## Data: GSH_df_pgls (Number of observations: 456)
## Draws: 4 chains, each with iter = 2000; warmup = 1000; thin = 1;
## total post-warmup draws = 4000
##
## Population-Level Effects:
##               Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS
## Intercept           0.02     0.01  -0.00   0.04  1.00    3526
## Log10MeanGSH_centered 0.39     0.08   0.24   0.55  1.00    3567
## Log10MaxSize_centered 0.33     0.04   0.25   0.40  1.00    3528
```

```
##                               Tail_ESS
## Intercept                    2556
## Log10MeanGSH_centered       2768
## Log10MaxSize_centered       3043
##
## Family Specific Parameters:
##      Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sigma    0.22     0.01   0.21   0.24 1.00   3652   2981
##
## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS
## and Tail_ESS are effective sample size measures, and Rhat is the potential
## scale reduction factor on split chains (at convergence, Rhat = 1).
```



## STAN model

Data considered

```
data {
  int <lower=0> N; // number of data points
  int <lower=0> K; // number of predictors
  matrix[N,K] x; // predictor matrix
  vector[N] y; // CFAR
}
```

The parameters accepted by the model.

```
parameters {  
  real alpha;  
  vector[K] beta;  
  real<lower=0> sigma;  
}
```

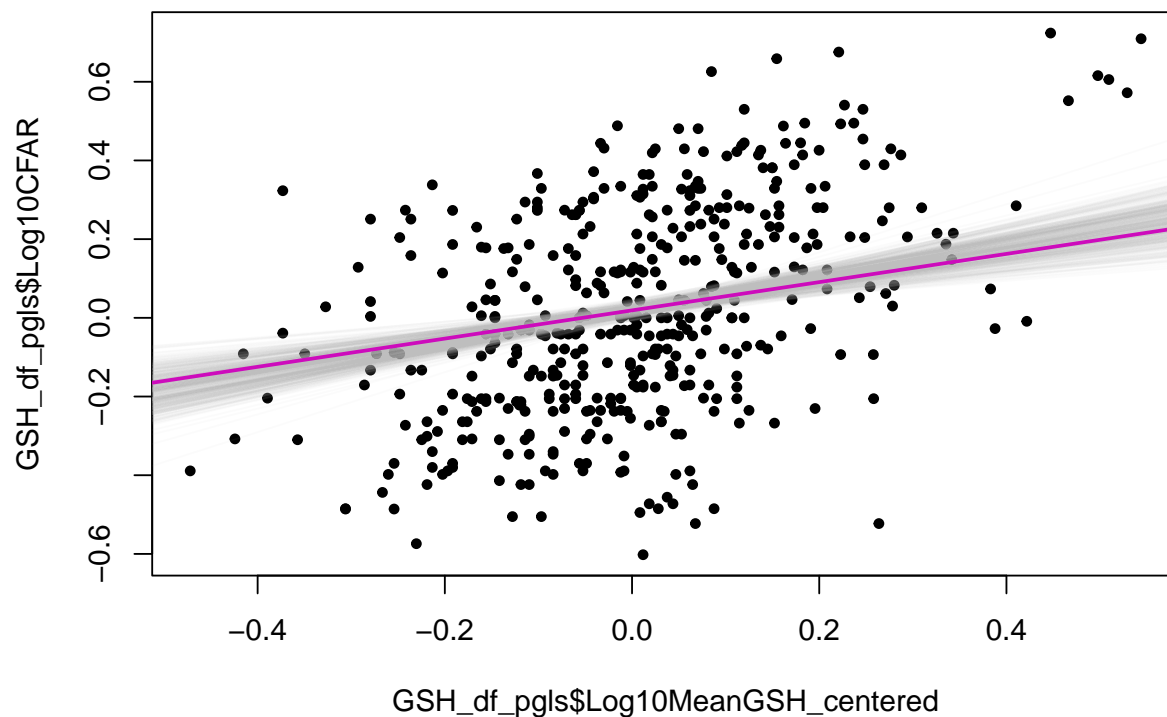
The model to be estimated. We model the output ‘y’ to be normally distributed with mean ‘mu’ and standard deviation ‘sigma’.

```
model {  
  
  sigma ~ student_t(3, 0, 10);  
  
  y ~ normal(alpha + x * beta , sigma);  
}
```

```
fit3
```

```
## Inference for Stan model: StanModel2.  
## 4 chains, each with iter=1000; warmup=500; thin=1;  
## post-warmup draws per chain=500, total post-warmup draws=2000.  
##  
##          mean se_mean   sd  2.5%   25%   50%   75%  97.5% n_eff Rhat  
## alpha      0.02    0.00 0.01   0.00   0.01   0.02   0.03   0.04  2075    1  
## beta[1]     0.39    0.00 0.08   0.23   0.34   0.39   0.45   0.55  1620    1  
## beta[2]     0.32    0.00 0.04   0.24   0.30   0.32   0.35   0.40  1550    1  
## sigma       0.22    0.00 0.01   0.21   0.22   0.22   0.23   0.24  1790    1  
## lp__        458.67    0.05 1.41 455.06 458.03 458.99 459.70 460.42   948    1  
##  
## Samples were drawn using NUTS(diag_e) at Wed Mar 02 11:22:15 2022.  
## For each parameter, n_eff is a crude measure of effective sample size,  
## and Rhat is the potential scale reduction factor on split chains (at  
## convergence, Rhat=1).
```





Model 4 - CFAR  $\sim$  GSH + (1|Max Size) (No interaction)

brms model

```
Model_size_3 <- brm(Log10CFAR ~ Log10MeanGSH_centered + (1|Log10MaxSize_centered),
  data = GSH_df_pgls,
  family = gaussian(),
  prior = Size_prior_3,
  control = list(adapt_delta = 0.9)) #needed to make model converge ...
'''
```

```
'''r
```

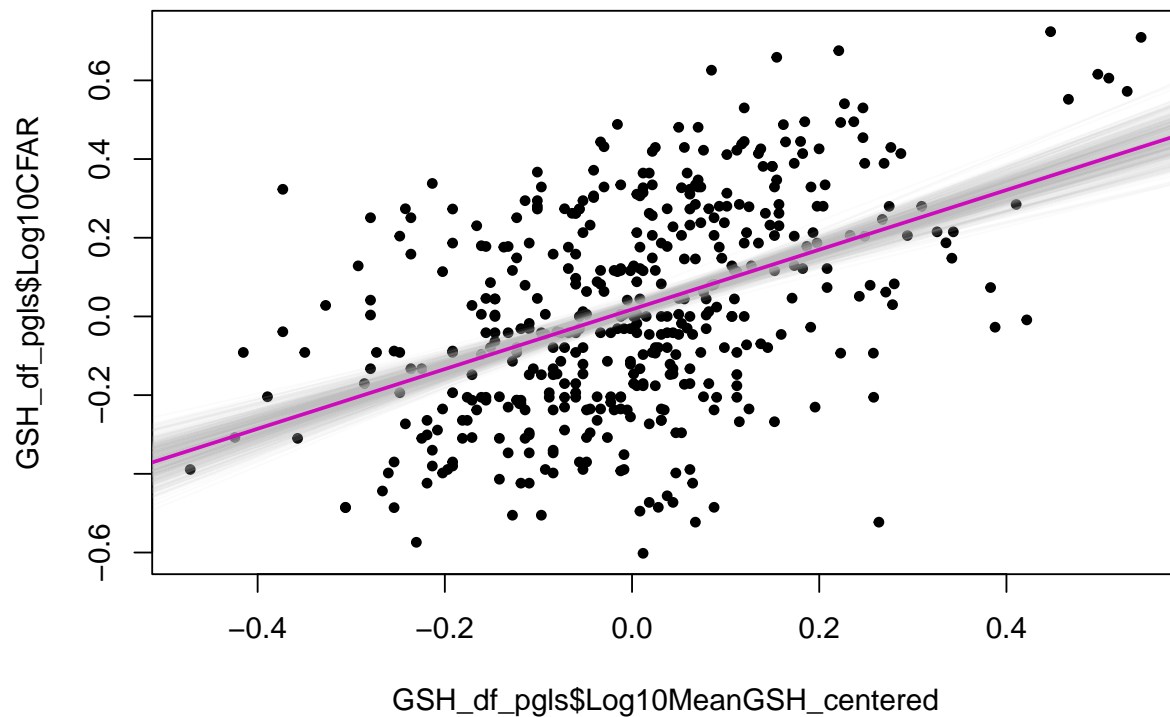
```
summary(Model_size_3)
```

```
## Family: gaussian
## Links: mu = identity; sigma = identity
## Formula: Log10CFAR ~ Log10MeanGSH_centered + (1 | Log10MaxSize_centered)
## Data: GSH_df_pgls (Number of observations: 456)
## Draws: 4 chains, each with iter = 2000; warmup = 1000; thin = 1;
## total post-warmup draws = 4000
##
## Group-Level Effects:
## ~Log10MaxSize_centered (Number of levels: 222)
```

```

##               Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sd(Intercept)    0.06     0.03    0.01    0.11 1.01     435     525
##
## Population-Level Effects:
##               Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS
## Intercept         0.02     0.01   -0.01    0.04 1.00     5528
## Log10MeanGSH_centered 0.76     0.07    0.62    0.90 1.00     4392
##               Tail_ESS
## Intercept         2718
## Log10MeanGSH_centered 2985
##
## Family Specific Parameters:
##               Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sigma           0.23     0.01    0.21    0.25 1.01     917     1516
##
## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS
## and Tail_ESS are effective sample size measures, and Rhat is the potential
## scale reduction factor on split chains (at convergence, Rhat = 1).

```



STAN model

TBD

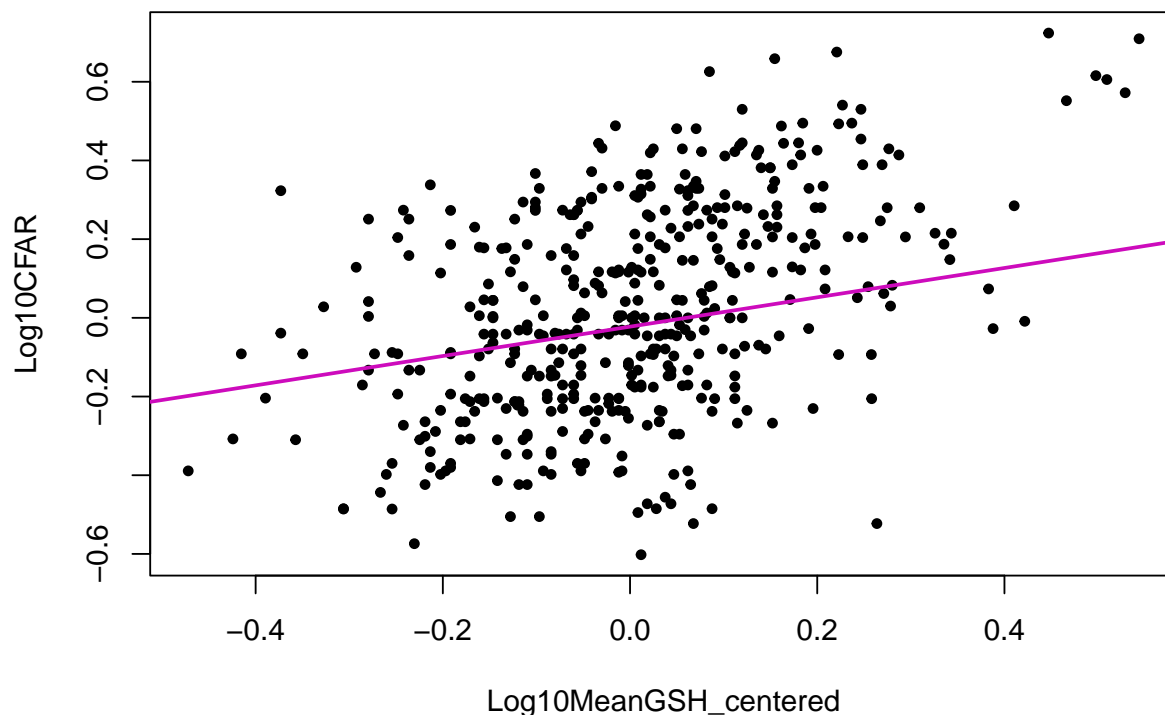
## Model 5 - CFAR ~ GSH + Phylogeny

### PGLS model

```
pglsMod1 <- gls(Log10CFAR ~ Log10MeanGSH_centered, correlation = corPagel(phy = SampleTree, value = 0.5
```

```
summary(pglsMod1)
```

```
## Generalized least squares fit by maximum likelihood
##   Model: Log10CFAR ~ Log10MeanGSH_centered
##   Data: GSH_df_pgls
##       AIC      BIC    logLik
##   -427.4206 -410.9306 217.7103
##
## Correlation Structure: corPagel
##   Formula: ~Binomial
##   Parameter estimate(s):
##     lambda
## 0.7030228
##
## Coefficients:
##              Value Std.Error   t-value p-value
## (Intercept)   -0.0224754 0.06282858 -0.357726  0.7207
## Log10MeanGSH_centered  0.3728837 0.05950418  6.266513  0.0000
##
## Correlation:
##              (Intr)
## Log10MeanGSH_centered -0.065
##
## Standardized residuals:
##      Min      Q1      Med      Q3      Max
## -2.7884571 -0.6933649  0.1151372  1.0885641  2.9034051
##
## Residual standard error: 0.2147249
## Degrees of freedom: 456 total; 454 residual
```



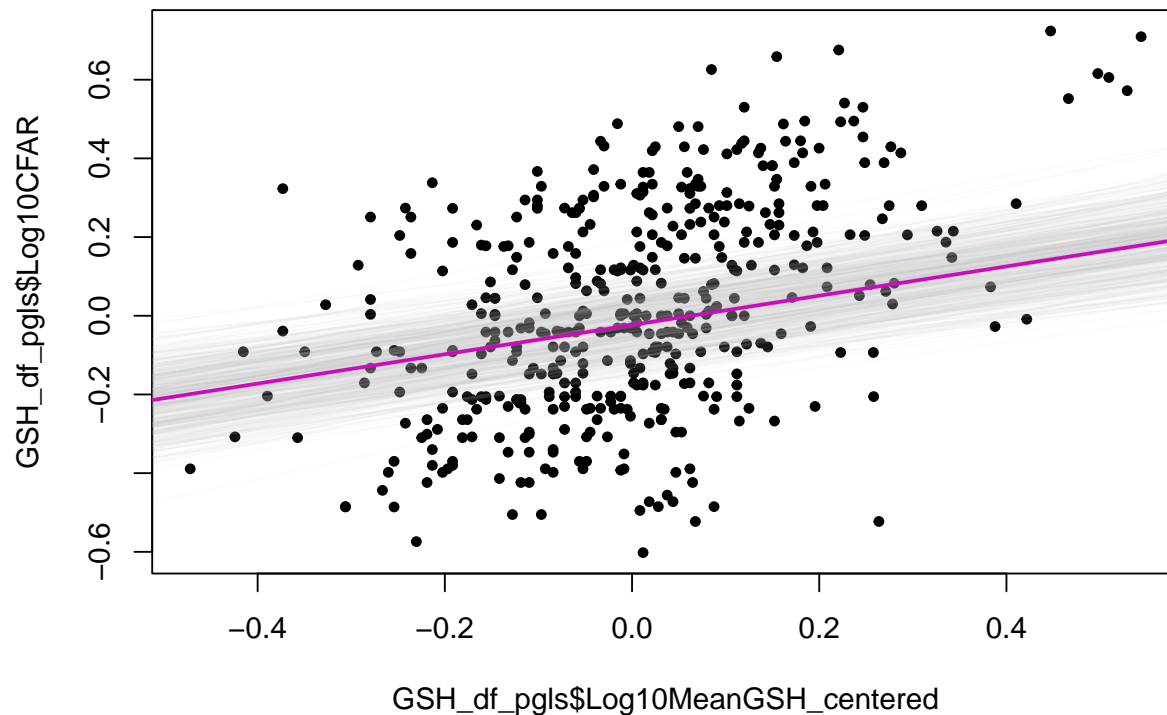
brms model

```
Model_simple <- brm(Log10CFAR ~ Log10MeanGSH_centered + (1|gr(Binomial, cov = A)),
  data = GSH_df_pgls,
  family = gaussian(),
  data2 = list(A = A),
  prior = Simple_prior,
  cores = 2)
```

```
summary(Model_simple)
```

```
## Family: gaussian
## Links: mu = identity; sigma = identity
## Formula: Log10CFAR ~ Log10MeanGSH_centered + (1 | gr(Binomial, cov = A))
## Data: GSH_df_pgls (Number of observations: 456)
## Draws: 4 chains, each with iter = 2000; warmup = 1000; thin = 1;
## total post-warmup draws = 4000
##
## Group-Level Effects:
## ~Binomial (Number of levels: 456)
## Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sd(Intercept) 0.01 0.00 0.01 0.01 1.00 1200 1938
##
## Population-Level Effects:
```

```
##               Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS
## Intercept          -0.02    0.06   -0.15    0.10 1.00    1257
## Log10MeanGSH_centered 0.37    0.06    0.25    0.49 1.00    4238
##               Tail_ESS
## Intercept           2076
## Log10MeanGSH_centered 3483
##
## Family Specific Parameters:
##      Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sigma    0.12    0.01    0.11    0.13 1.00    1717    2848
##
## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS
## and Tail_ESS are effective sample size measures, and Rhat is the potential
## scale reduction factor on split chains (at convergence, Rhat = 1).
```



STAN model

TBD

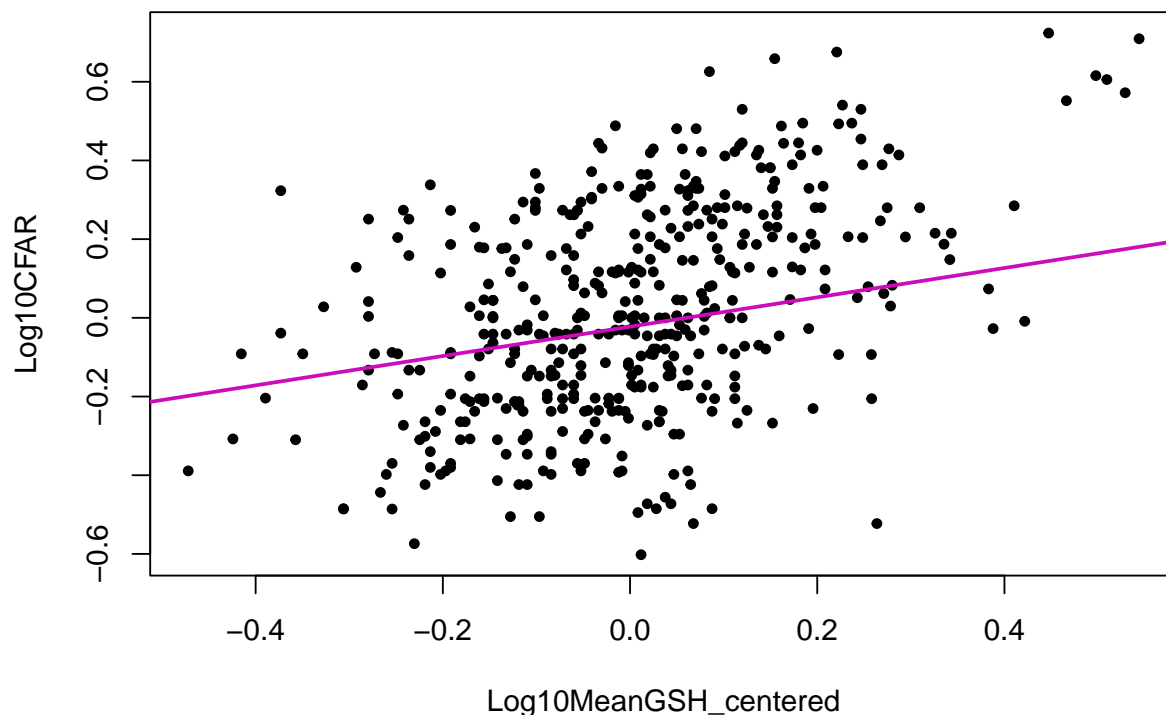
Model 5 - CFAR ~ GSH \* Max Size + Phylogeny

PGLS model

```
pglsMod2 <- gls(Log10CFAR ~ Log10MeanGSH_centered * Log10MaxSize_centered, correlation = corPagel(phy =
```

```
summary(pglMod2)
```

```
## Generalized least squares fit by maximum likelihood
## Model: y ~ x
## Data: GSH_df_pgl
##      AIC      BIC    logLik
## -427.4206 -410.9306 217.7103
##
## Correlation Structure: corPagel
## Formula: ~Binomial
## Parameter estimate(s):
##      lambda
## 0.7030228
##
## Coefficients:
##              Value Std.Error   t-value p-value
## (Intercept) -0.0224754 0.06282858 -0.357726  0.7207
## x           0.3728837 0.05950418  6.266513  0.0000
##
## Correlation:
## (Intr)
## x -0.065
##
## Standardized residuals:
##      Min      Q1      Med      Q3      Max
## -2.7884571 -0.6933649  0.1151372  1.0885641  2.9034051
##
## Residual standard error: 0.2147249
## Degrees of freedom: 456 total; 454 residual
```



brms model

```
Model_BS <- brm(Log10CFAR ~ Log10MeanGSH_centered * Log10MaxSize_centered + (1|gr(Binomial, cov = A)),
  data = GSH_df_pgls,
  family = gaussian(),
  data2 = list(A = A),
  prior = BS_prior,
  sample_prior = TRUE, chains = 4, cores = 2)
```

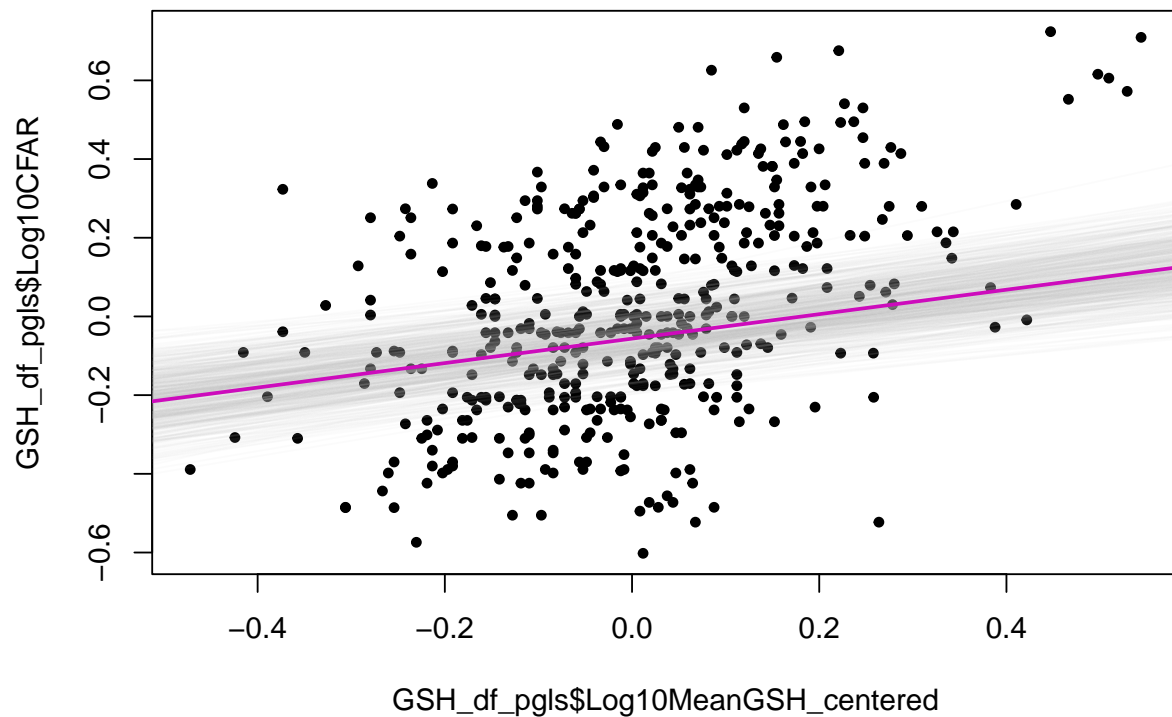
```
summary(Model_BS)
```

```
## Family: gaussian
## Links: mu = identity; sigma = identity
## Formula: Log10CFAR ~ Log10MeanGSH_centered * Log10MaxSize_centered + (1 | gr(Binomial, cov = A))
## Data: GSH_df_pgls (Number of observations: 456)
## Draws: 4 chains, each with iter = 2000; warmup = 1000; thin = 1;
## total post-warmup draws = 4000
##
## Group-Level Effects:
## ~Binomial (Number of levels: 456)
## Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sd(Intercept) 0.01 0.00 0.01 0.01 1.00 1303 2376
##
## Population-Level Effects:
```

```

##                                     Estimate Est.Error l-95% CI
## Intercept                         -0.06      0.06    -0.17
## Log10MeanGSH_centered              0.31      0.06     0.19
## Log10MaxSize_centered              0.13      0.04     0.06
## Log10MeanGSH_centered:Log10MaxSize_centered 0.43      0.12     0.20
##                                     u-95% CI Rhat Bulk_ESS Tail_ESS
## Intercept                         0.05 1.00    1824    2402
## Log10MeanGSH_centered              0.43 1.00    6595    3478
## Log10MaxSize_centered              0.20 1.00    4866    3342
## Log10MeanGSH_centered:Log10MaxSize_centered 0.67 1.00    5734    3590
##
## Family Specific Parameters:
##      Estimate Est.Error l-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sigma    0.12      0.01     0.11     0.13 1.00    2164    3283
##
## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS
## and Tail_ESS are effective sample size measures, and Rhat is the potential
## scale reduction factor on split chains (at convergence, Rhat = 1).

```



STAN model

TBD



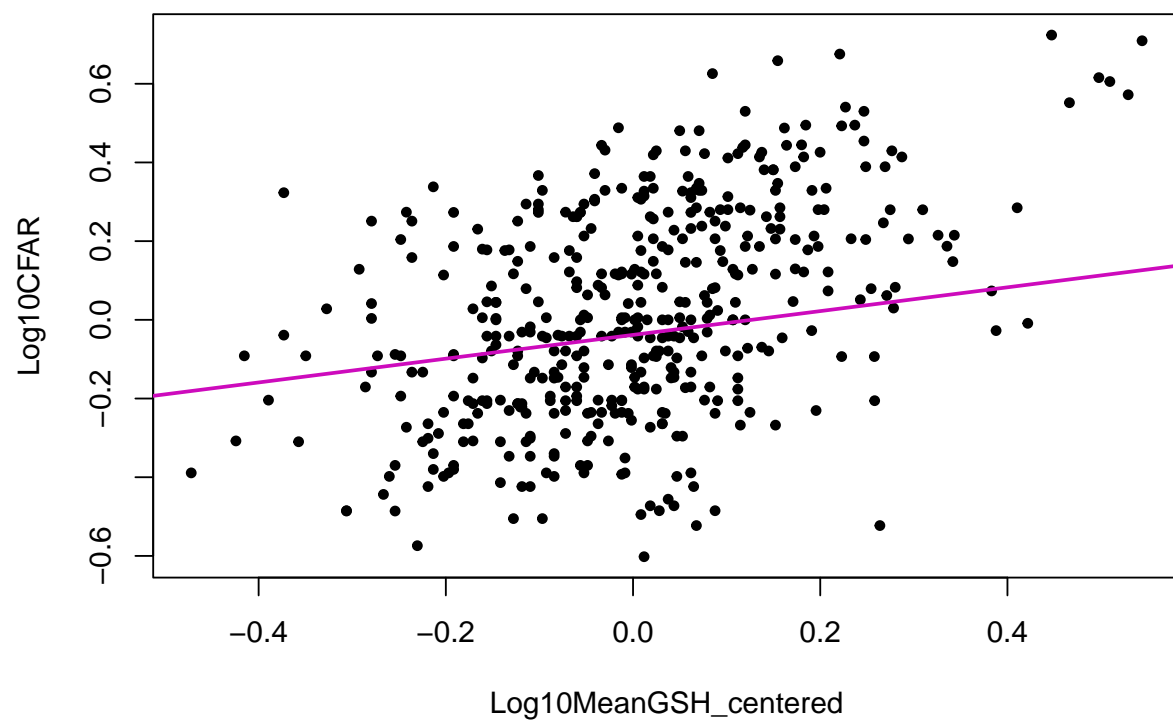
## Model 6 - CFAR ~ GSH + Max Size + Phylogeny

### PGLS model

```
pglsMod3 <- gls(Log10CFAR ~ Log10MeanGSH_centered + Log10MaxSize_centered, correlation = corPagel(phy =
```

```
summary(pglsMod3)
```

```
## Generalized least squares fit by maximum likelihood
## Model: Log10CFAR ~ Log10MeanGSH_centered + Log10MaxSize_centered
## Data: GSH_df_pgls
##      AIC      BIC    logLik
## -447.4392 -426.8268 228.7196
##
## Correlation Structure: corPagel
## Formula: ~Binomial
## Parameter estimate(s):
##      lambda
## 0.6786106
##
## Coefficients:
##              Value Std.Error   t-value p-value
## (Intercept)   -0.03827951 0.05915461 -0.647110  0.5179
## Log10MeanGSH_centered  0.30222706 0.06034147  5.008613  0.0000
## Log10MaxSize_centered  0.16260955 0.03412415  4.765233  0.0000
##
## Correlation:
##              (Intr) L10MGS
## Log10MeanGSH_centered -0.050
## Log10MaxSize_centered -0.056 -0.268
##
## Standardized residuals:
##      Min      Q1      Med      Q3      Max
## -2.8353741 -0.5472167  0.2440250  1.1919410  2.8367324
##
## Residual standard error: 0.2047976
## Degrees of freedom: 456 total; 453 residual
```



brms model

TBD

STAN model

TBD