```
// generated with brms 2.20.1
functions {
data {
  int<lower=1> N: // total number of observations
  vector[N] Y; // response variable
  int<lower=-1,upper=2> cens[N]; // indicates censoring
  int<lower=1> K_alpha; // number of population-level effects
  matrix[N, K_alpha] X_alpha; // population-level design matrix
  int<lower=1> K_beta1; // number of population-level effects
  matrix[N, K beta1] X beta1; // population-level design matrix
  int<lower=1> K_beta2; // number of population—level effects
  matrix[N, K_beta2] X_beta2; // population-level design matrix
  int<lower=1> K_gamma; // number of population—level effects
  matrix[N, K_gamma] X_gamma; // population—level design matrix
  int<lower=1> K_kappa; // number of population-level effects
  matrix[N, K kappa] X kappa; // population-level design matrix
  // covariates for non-linear functions
  vector[N] C_1;
  // data for group-level effects of ID 1
  int<lower=1> N_1; // number of grouping levels
  int<lower=1> M_1; // number of coefficients per level
  int<lower=1> J_1[N]; // grouping indicator per observation
  // group-level predictor values
  vector[N] Z_1_alpha_1;
  // data for group-level effects of ID 2
  int<lower=1> N_2; // number of grouping levels
  int<lower=1> M_2; // number of coefficients per level
  int<lower=1> J_2[N]; // grouping indicator per observation
  // group-level predictor values
  vector[N] Z_2_alpha_1;
  // data for group-level effects of ID 3
  int<lower=1> N_3; // number of grouping levels
  int<lower=1> M 3; // number of coefficients per level
  int<lower=1> J_3[N]; // grouping indicator per observation
  // group-level predictor values
  vector[N] Z 3 beta1 1;
  // data for group-level effects of ID 4
  int<lower=1> N 4; // number of grouping levels
  int<lower=1> M 4; // number of coefficients per level
  int<lower=1> J_4[N]; // grouping indicator per observation
  // group-level predictor values
  vector[N] Z 4 beta1 1;
  // data for group-level effects of ID 5
  int<lower=1> N_5; // number of grouping levels
  int<lower=1> M_5; // number of coefficients per level
  int<lower=1> J_5[N]; // grouping indicator per observation
  // group-level predictor values
  vector[N] Z 5 beta2 1;
  // data for group-level effects of ID 6
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int<lower=1> N 6; // number of grouping levels
  int<lower=1> M 6; // number of coefficients per level
  int<lower=1> J_6[N]; // grouping indicator per observation
  // group-level predictor values
  vector[N] Z 6 beta2 1:
  // data for group-level effects of ID 7
  int<lower=1> N 7; // number of grouping levels
  int<lower=1> M 7; // number of coefficients per level
  int<lower=1> J_7[N]; // grouping indicator per observation
  // group-level predictor values
  vector[N] Z 7 gamma 1;
  // data for group-level effects of ID 8
  int<lower=1> N_8; // number of grouping levels
  int<lower=1> M_8; // number of coefficients per level
  int<lower=1> J_8[N]; // grouping indicator per observation
  // group-level predictor values
  vector[N] Z 8 kappa 1;
  int prior_only; // should the likelihood be ignored?
}
transformed data {
parameters {
  vector[K_alpha] b_alpha; // regression coefficients
  vector[K_beta1] b_beta1; // regression coefficients
  vector[K_beta2] b_beta2; // regression coefficients
  vector[K_gamma] b_gamma; // regression coefficients
  vector[K_kappa] b_kappa; // regression coefficients
  real<lower=0> sigma; // dispersion parameter
  vector<lower=0>[M_1] sd_1; // group-level standard deviations
  vector[N_1] z_1[M_1]; // standardized group-level effects
  vector<lower=0>[M_2] sd_2; // group-level standard deviations
  vector[N_2] z_2[M_2]; // standardized group-level effects
  vector<lower=0>[M_3] sd_3; // group-level standard deviations
  vector[N 3] z 3[M 3]; // standardized group-level effects
  vector<lower=0>[M 4] sd 4; // group-level standard deviations
  vector[N 4] z 4[M 4]; // standardized group-level effects
  vector<lower=0>[M_5] sd_5; // group-level standard deviations
  vector[N_5] z_5[M_5]; // standardized group-level effects
  vector<lower=0>[M 6] sd 6; // group-level standard deviations
  vector[N 6] z 6[M 6]; // standardized group-level effects
  vector<lower=0>[M_7] sd_7; // group-level standard deviations
  vector[N 7] z 7[M 7]; // standardized group-level effects
  vector<lower=0>[M 8] sd 8; // group-level standard deviations
  vector[N_8] z_8[M_8]; // standardized group-level effects
}
transformed parameters {
  vector[N_1] r_1_alpha_1; // actual group-level effects
  vector[N_2] r_2_alpha_1; // actual group-level effects
  vector[N_3] r_3_beta1_1; // actual group-level effects
  vector[N_4] r_4_beta1_1; // actual group-level effects
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vector[N_5] r_5_beta2_1; // actual group-level effects
  vector[N_6] r_6_beta2_1; // actual group-level effects
  vector[N_7] r_7_gamma_1; // actual group-level effects
  vector[N_8] r_8_kappa_1; // actual group-level effects
  real lprior = 0; // prior contributions to the log posterior
  r 1 alpha 1 = (sd 1[1] * (z 1[1]));
  r 2 alpha 1 = (sd 2[1] * (z 2[1]));
  r 3 beta1 1 = (sd 3[1] * (z 3[1]));
  r_4_{beta1_1} = (sd_4[1] * (z_4[1]));
  r 5 beta2 1 = (sd 5[1] * (z 5[1]));
  r 6 beta2 1 = (sd_6[1] * (z_6[1]));
  r_7_{gamma_1} = (sd_7[1] * (z_7[1]));
  r_8_{kappa_1} = (sd_8[1] * (z_8[1]));
  lprior += student_t_lpdf(sigma | 3, 0, 2.5)
    - 1 * student_t_lccdf(0 | 3, 0, 2.5);
  lprior += student_t_lpdf(sd_1 | 3, 0, 2.5)
    - 1 * student_t_lccdf(0 | 3, 0, 2.5);
  lprior += student_t_lpdf(sd_2 | 3, 0, 2.5)
    - 1 * student_t_lccdf(0 | 3, 0, 2.5);
  lprior += student_t_lpdf(sd_3 | 3, 0, 2.5)
    - 1 * student_t_lccdf(0 | 3, 0, 2.5);
  lprior += student_t_lpdf(sd_4 | 3, 0, 2.5)
    - 1 * student_t_lccdf(0 | 3, 0, 2.5);
  lprior += student_t_lpdf(sd_5 | 3, 0, 2.5)
    - 1 * student_t_lccdf(0 | 3, 0, 2.5);
  lprior += student_t_lpdf(sd_6 | 3, 0, 2.5)
    - 1 * student_t_lccdf(0 | 3, 0, 2.5);
  lprior += student_t_lpdf(sd_7 | 3, 0, 2.5)
    - 1 * student_t_lccdf(0 | 3, 0, 2.5);
  lprior += student t lpdf(sd 8 | 3, 0, 2.5)
    - 1 * student_t_lccdf(0 | 3, 0, 2.5);
}
model {
  // likelihood including constants
  if (!prior only) {
    // initialize linear predictor term
    vector[N] nlp_alpha = rep_vector(0.0, N);
    // initialize linear predictor term
    vector[N] nlp beta1 = rep vector(0.0, N);
    // initialize linear predictor term
    vector[N] nlp_beta2 = rep_vector(0.0, N);
    // initialize linear predictor term
    vector[N] nlp_gamma = rep_vector(0.0, N);
    // initialize linear predictor term
    vector[N] nlp_kappa = rep_vector(0.0, N);
    // initialize non-linear predictor term
    vector[N] mu;
    nlp_alpha += X_alpha * b_alpha;
    nlp beta1 += X beta1 * b beta1;
    nlp_beta2 += X_beta2 * b_beta2;
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nlp qamma += X qamma * b qamma;
         nlp_kappa += X_kappa * b_kappa;
         for (n in 1:N) {
              // add more terms to the linear predictor
              nlp_alpha[n] += r_1_alpha_1[J_1[n]] * Z_1_alpha_1[n] +
r_2_alpha_1[J_2[n]] * Z_2_alpha_1[n];
         for (n in 1:N) {
              // add more terms to the linear predictor
              nlp beta1[n] += r 3 beta1 1[J 3[n]] * Z 3 beta1 1[n] +
r_4_{beta1_1[J_4[n]]} * Z_4_{beta1_1[n]};
         for (n in 1:N) {
              // add more terms to the linear predictor
              nlp_beta2[n] += r_5_beta2_1[J_5[n]] * Z_5_beta2_1[n] +
r_6_{beta2_1[J_6[n]]} * Z_6_{beta2_1[n]};
         for (n in 1:N) {
             // add more terms to the linear predictor
              nlp_{gamma}[n] += r_7_{gamma}[J_7[n]] * Z_7_{gamma}[n];
         for (n in 1:N) {
              // add more terms to the linear predictor
              nlp_{kappa}[n] += r_8_{kappa_1}[J_8[n]] * Z_8_{kappa_1}[n];
         for (n in 1:N) {
              // compute non-linear predictor values
             mu[n] = (nlp_alpha[n] + nlp_beta1[n] * C_1[n] + nlp_beta2[n] *
nlp_{gamma}[n] * log((exp((C_1[n] - nlp_kappa[n]) / nlp_gamma[n]) +
exp(-(C_1[n] - nlp_kappa[n]) / nlp_gamma[n])) / (exp(nlp_kappa[n] / nlp_gamma[n])) / (exp(nlp_kappa[n]) / nlp_gamma[n]) / (exp(nlp_kappa[n]) / nlp_gamma[n]) / nlp_gamma[n])
nlp_gamma[n]) + exp( - nlp_kappa[n] / nlp_gamma[n]))));
         for (n in 1:N) {
         // special treatment of censored data
              if (cens[n] == 0) {
                  target += normal_lpdf(Y[n] | mu[n], sigma);
              } else if (cens[n] == 1) {
                  target += normal_lccdf(Y[n] | mu[n], sigma);
              } else if (cens[n] == -1) {
                  target += normal lcdf(Y[n] | mu[n], sigma);
         }
    }
    // priors including constants
    target += lprior;
    target += std_normal_lpdf(z_1[1]);
    target += std_normal_lpdf(z_2[1]);
    target += std_normal_lpdf(z_3[1]);
    target += std normal lpdf(z 4[1]);
    target += std_normal_lpdf(z_5[1]);
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target += std_normal_lpdf(z_6[1]);
target += std_normal_lpdf(z_7[1]);
target += std_normal_lpdf(z_8[1]);
}
generated quantities {
}
```