

STATS 271 Final Project

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```
library(mlmRev)
library(tidyverse)
library(rstan)
library(rstanarm)
library(gridExtra)
library(loo)
library(brms)
library(ggplot2)
theme_set(theme_minimal())
data(Contraception)
cont <- Contraception
#save.image("~/Desktop/271_project.RData")
load("~/Desktop/271_project.RData")
```

1. EDA and summary statistics

```
use_cont <- filter(cont, use==1)
group_cont_use <- cont %>% group_by(district) %>% summarise(use=mean(use))
group_region_use <- cont %>% group_by(urban) %>% summarise(use=mean(use))
table(cont$livch)
group_livch_use <- cont %>% group_by(livch) %>% summarise(use=mean(use))
mean(cont$use)
cont2 <- cont
cont2$age <- cont2$age + 30
test <- filter(cont2, age > 50 & age <= 55)
mean(test$use)

district_mean_hist <- group_cont_use %>% ggplot(aes(x=use)) +
  geom_histogram(binwidth=0.05, color="black", fill="white") +
  theme_bw() + ggtitle("Distribution of percentage of contraception use by district") +
  xlab("Percentage of contraception use") + theme(panel.grid.major = element_blank(),
                                                  panel.grid.minor = element_blank(),
                                                  panel.background = element_blank(), axis.line = element_line(colour = "black"))

cont_2 <- use_cont %>% ggplot(aes(x = factor(urban), fill = factor(livch))) +
  geom_bar(position="dodge") + theme_bw() +
  ggtitle("Use of contraception by urban/rural residence and number of living children") +
  xlab("Urban Residence") + theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
                                  panel.background = element_blank(), axis.line = element_line(colour = "black")) +
  scale_fill_discrete(name="Number of living children")
```

```
grid.arrange(district_mean_hist, cont_2, ncol=2)
```

2. Modeling and posterior checking

Model 1. Varying intercept model with no predictor

```
varying_inter_nopred_mod <- stan_glmer(formula = use ~ 1 + (1|district),  
  data = cont,  
  seed = 271,  
  cores = getOption("mc.cores", 4L),  
  family = binomial(link = "logit"))
```

```
summary(varying_inter_nopred_mod, digits = 3)
```

```
##  
## Model Info:  
## function:      stan_glmer  
## family:        binomial [logit]  
## formula:       use ~ 1 + (1 | district)  
## algorithm:     sampling  
## sample:        4000 (posterior sample size)  
## priors:        see help('prior_summary')  
## observations:  1934  
## groups:        district (60)  
##  
## Estimates:  
##
```

	mean	sd	10%	50%	90%
## (Intercept)	-0.540	0.089	-0.653	-0.537	-0.427
## b[(Intercept) district:1]	-0.456	0.208	-0.725	-0.452	-0.194
## b[(Intercept) district:2]	-0.046	0.333	-0.480	-0.045	0.382
## b[(Intercept) district:3]	0.302	0.499	-0.319	0.288	0.951
## b[(Intercept) district:4]	0.347	0.308	-0.046	0.345	0.743
## b[(Intercept) district:5]	-0.034	0.284	-0.406	-0.030	0.328
## b[(Intercept) district:6]	-0.279	0.254	-0.609	-0.273	0.044
## b[(Intercept) district:7]	-0.218	0.366	-0.683	-0.214	0.236
## b[(Intercept) district:8]	0.021	0.293	-0.354	0.022	0.398
## b[(Intercept) district:9]	-0.168	0.351	-0.626	-0.157	0.272
## b[(Intercept) district:10]	-0.603	0.424	-1.159	-0.587	-0.076
## b[(Intercept) district:11]	-1.005	0.427	-1.566	-0.977	-0.482
## b[(Intercept) district:12]	-0.070	0.311	-0.473	-0.064	0.325
## b[(Intercept) district:13]	0.118	0.328	-0.301	0.114	0.530
## b[(Intercept) district:14]	0.933	0.201	0.677	0.935	1.194
## b[(Intercept) district:15]	-0.020	0.337	-0.449	-0.021	0.405
## b[(Intercept) district:16]	0.405	0.354	-0.043	0.401	0.847
## b[(Intercept) district:17]	-0.209	0.346	-0.655	-0.203	0.226
## b[(Intercept) district:18]	-0.091	0.275	-0.442	-0.089	0.254
## b[(Intercept) district:19]	0.038	0.327	-0.375	0.038	0.458
## b[(Intercept) district:20]	0.058	0.371	-0.421	0.062	0.534
## b[(Intercept) district:21]	0.045	0.354	-0.408	0.050	0.494
## b[(Intercept) district:22]	-0.428	0.360	-0.905	-0.419	0.026
## b[(Intercept) district:23]	-0.222	0.389	-0.722	-0.216	0.266

```

## b[(Intercept) district:24] -0.635 0.409 -1.163 -0.625 -0.128
## b[(Intercept) district:25] 0.269 0.237 -0.040 0.270 0.566
## b[(Intercept) district:26] 0.025 0.388 -0.466 0.029 0.519
## b[(Intercept) district:27] -0.643 0.312 -1.054 -0.626 -0.253
## b[(Intercept) district:28] -0.424 0.275 -0.774 -0.420 -0.077
## b[(Intercept) district:29] -0.264 0.314 -0.661 -0.256 0.134
## b[(Intercept) district:30] 0.401 0.245 0.090 0.397 0.716
## b[(Intercept) district:31] 0.241 0.301 -0.144 0.234 0.623
## b[(Intercept) district:32] -0.433 0.350 -0.886 -0.425 0.004
## b[(Intercept) district:33] 0.122 0.380 -0.367 0.119 0.606
## b[(Intercept) district:34] 0.813 0.306 0.429 0.803 1.212
## b[(Intercept) district:35] 0.410 0.257 0.079 0.404 0.746
## b[(Intercept) district:36] -0.043 0.369 -0.514 -0.042 0.414
## b[(Intercept) district:37] 0.316 0.382 -0.183 0.313 0.808
## b[(Intercept) district:38] -0.168 0.393 -0.681 -0.169 0.328
## b[(Intercept) district:39] 0.338 0.313 -0.062 0.336 0.729
## b[(Intercept) district:40] 0.279 0.274 -0.063 0.278 0.637
## b[(Intercept) district:41] 0.333 0.323 -0.083 0.332 0.747
## b[(Intercept) district:42] 0.306 0.399 -0.202 0.300 0.805
## b[(Intercept) district:43] 0.502 0.282 0.147 0.497 0.867
## b[(Intercept) district:44] -0.410 0.346 -0.852 -0.412 0.021
## b[(Intercept) district:45] -0.112 0.282 -0.475 -0.110 0.249
## b[(Intercept) district:46] 0.534 0.213 0.265 0.533 0.806
## b[(Intercept) district:47] 0.202 0.383 -0.280 0.199 0.698
## b[(Intercept) district:48] 0.465 0.266 0.137 0.461 0.811
## b[(Intercept) district:49] -0.333 0.478 -0.946 -0.319 0.256
## b[(Intercept) district:50] 0.236 0.360 -0.217 0.234 0.698
## b[(Intercept) district:51] 0.265 0.295 -0.111 0.266 0.646
## b[(Intercept) district:52] 0.242 0.240 -0.067 0.239 0.557
## b[(Intercept) district:53] 0.112 0.343 -0.337 0.116 0.551
## b[(Intercept) district:55] -0.237 0.450 -0.814 -0.231 0.323
## b[(Intercept) district:56] 0.637 0.278 0.276 0.635 0.991
## b[(Intercept) district:57] -0.524 0.353 -0.978 -0.512 -0.089
## b[(Intercept) district:58] 0.231 0.298 -0.153 0.236 0.604
## b[(Intercept) district:59] -0.467 0.423 -1.012 -0.460 0.060
## b[(Intercept) district:60] -0.459 0.338 -0.893 -0.446 -0.035
## b[(Intercept) district:61] -0.511 0.300 -0.900 -0.499 -0.140
## Sigma[district:(Intercept),(Intercept)] 0.274 0.089 0.172 0.262 0.393
##
## Fit Diagnostics:
##      mean    sd   10%   50%   90%
## mean_PPD 0.393 0.015 0.373 0.393 0.412
##
## The mean_ppd is the sample average posterior predictive distribution of the outcome variable (for de
##
## MCMC diagnostics
##
##      mcse  Rhat  n_eff
## (Intercept) 0.002 1.002 1616
## b[(Intercept) district:1] 0.003 1.000 4142
## b[(Intercept) district:2] 0.005 0.999 5197
## b[(Intercept) district:3] 0.007 0.999 5508
## b[(Intercept) district:4] 0.004 0.999 6021
## b[(Intercept) district:5] 0.004 1.000 5397
## b[(Intercept) district:6] 0.004 1.000 4613

```

## b[(Intercept) district:7]	0.005	1.000	5412
## b[(Intercept) district:8]	0.004	1.000	5477
## b[(Intercept) district:9]	0.004	0.999	6521
## b[(Intercept) district:10]	0.006	1.000	4502
## b[(Intercept) district:11]	0.007	1.000	4254
## b[(Intercept) district:12]	0.004	1.000	6266
## b[(Intercept) district:13]	0.004	1.001	5372
## b[(Intercept) district:14]	0.003	1.000	3834
## b[(Intercept) district:15]	0.004	1.000	6235
## b[(Intercept) district:16]	0.005	1.000	4926
## b[(Intercept) district:17]	0.004	0.999	6865
## b[(Intercept) district:18]	0.004	1.000	4971
## b[(Intercept) district:19]	0.004	1.000	5295
## b[(Intercept) district:20]	0.005	0.999	5163
## b[(Intercept) district:21]	0.005	1.000	6193
## b[(Intercept) district:22]	0.005	0.999	6035
## b[(Intercept) district:23]	0.005	1.000	7110
## b[(Intercept) district:24]	0.006	0.999	4429
## b[(Intercept) district:25]	0.003	1.000	5123
## b[(Intercept) district:26]	0.005	1.000	6568
## b[(Intercept) district:27]	0.005	1.000	4761
## b[(Intercept) district:28]	0.004	0.999	5379
## b[(Intercept) district:29]	0.004	1.000	5286
## b[(Intercept) district:30]	0.004	1.000	4600
## b[(Intercept) district:31]	0.004	1.000	4815
## b[(Intercept) district:32]	0.004	1.000	6394
## b[(Intercept) district:33]	0.005	1.000	6574
## b[(Intercept) district:34]	0.004	0.999	5017
## b[(Intercept) district:35]	0.004	1.000	4694
## b[(Intercept) district:36]	0.005	0.999	5971
## b[(Intercept) district:37]	0.005	1.000	6009
## b[(Intercept) district:38]	0.005	1.000	5663
## b[(Intercept) district:39]	0.004	1.001	4835
## b[(Intercept) district:40]	0.004	1.000	4482
## b[(Intercept) district:41]	0.005	1.000	5091
## b[(Intercept) district:42]	0.005	1.000	6060
## b[(Intercept) district:43]	0.004	1.000	4553
## b[(Intercept) district:44]	0.005	1.000	5537
## b[(Intercept) district:45]	0.004	0.999	5705
## b[(Intercept) district:46]	0.003	1.000	3921
## b[(Intercept) district:47]	0.005	1.000	5410
## b[(Intercept) district:48]	0.004	0.999	4548
## b[(Intercept) district:49]	0.006	1.000	5843
## b[(Intercept) district:50]	0.005	1.001	5061
## b[(Intercept) district:51]	0.004	1.000	4903
## b[(Intercept) district:52]	0.004	1.000	4649
## b[(Intercept) district:53]	0.004	0.999	6174
## b[(Intercept) district:55]	0.006	0.999	5630
## b[(Intercept) district:56]	0.004	1.000	4354
## b[(Intercept) district:57]	0.005	1.000	4694
## b[(Intercept) district:58]	0.004	1.000	4809
## b[(Intercept) district:59]	0.005	0.999	6144
## b[(Intercept) district:60]	0.005	0.999	5588
## b[(Intercept) district:61]	0.004	1.000	4664

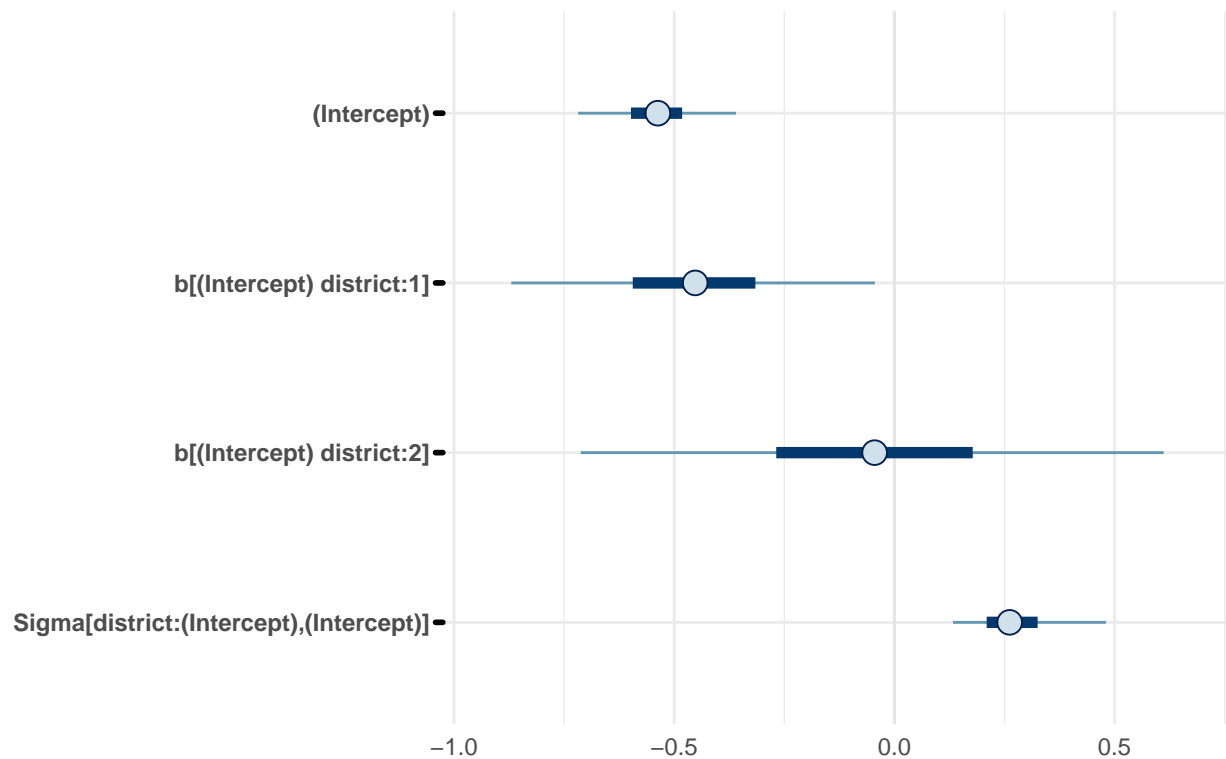
```
## Sigma[district:(Intercept),(Intercept)] 0.002 1.000 1396
## mean_PPD                                0.000 1.000 4276
## log-posterior                            0.270 1.001 818
##
## For each parameter, mcse is Monte Carlo standard error, n_eff is a crude measure of effective sample
prior_summary(varying_inter_nopred_mod)

## Priors for model 'varying_inter_nopred_mod'
## -----
## Intercept (after predictors centered)
## ~ normal(location = 0, scale = 10)
##
## Covariance
## ~ decov(reg. = 1, conc. = 1, shape = 1, scale = 1)
## -----
## See help('prior_summary.stanreg') for more details

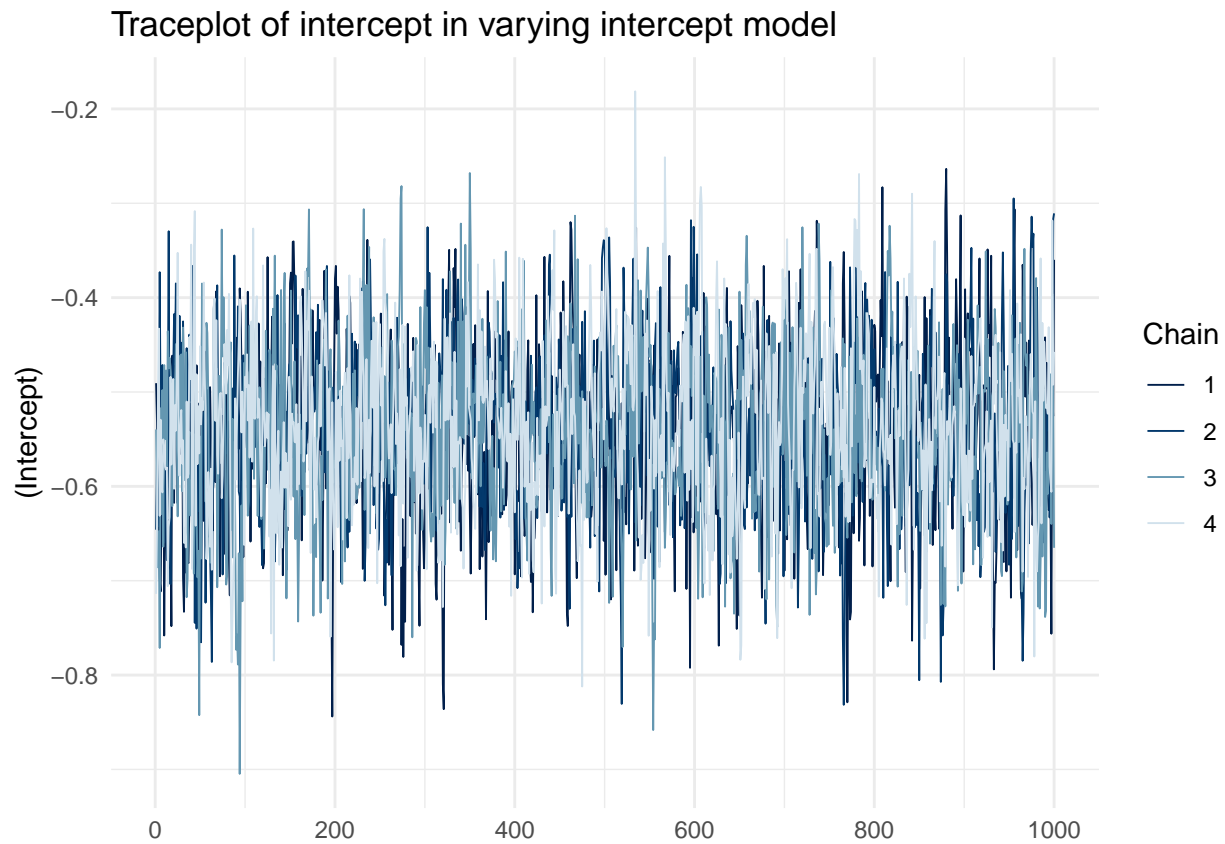
summary(varying_inter_nopred_mod, digits = 3,
  pars=c("(Intercept)", "b[(Intercept) district:1]",
        "b[(Intercept) district:2]", "Sigma[district:(Intercept),(Intercept)]"),
  probs = c(0.025, 0.5, 0.975))

plot(varying_inter_nopred_mod, pars=c("(Intercept)",
  "b[(Intercept) district:1]", "b[(Intercept) district:2]",
  "Sigma[district:(Intercept),(Intercept)]"),
  prob = 0.5, prob_outer = 0.95) + ggplot2::ggtitle("Posterior medians \n with 50% and 95% intervals")
```

Posterior medians
with 50% and 95% intervals in varying intercept moc



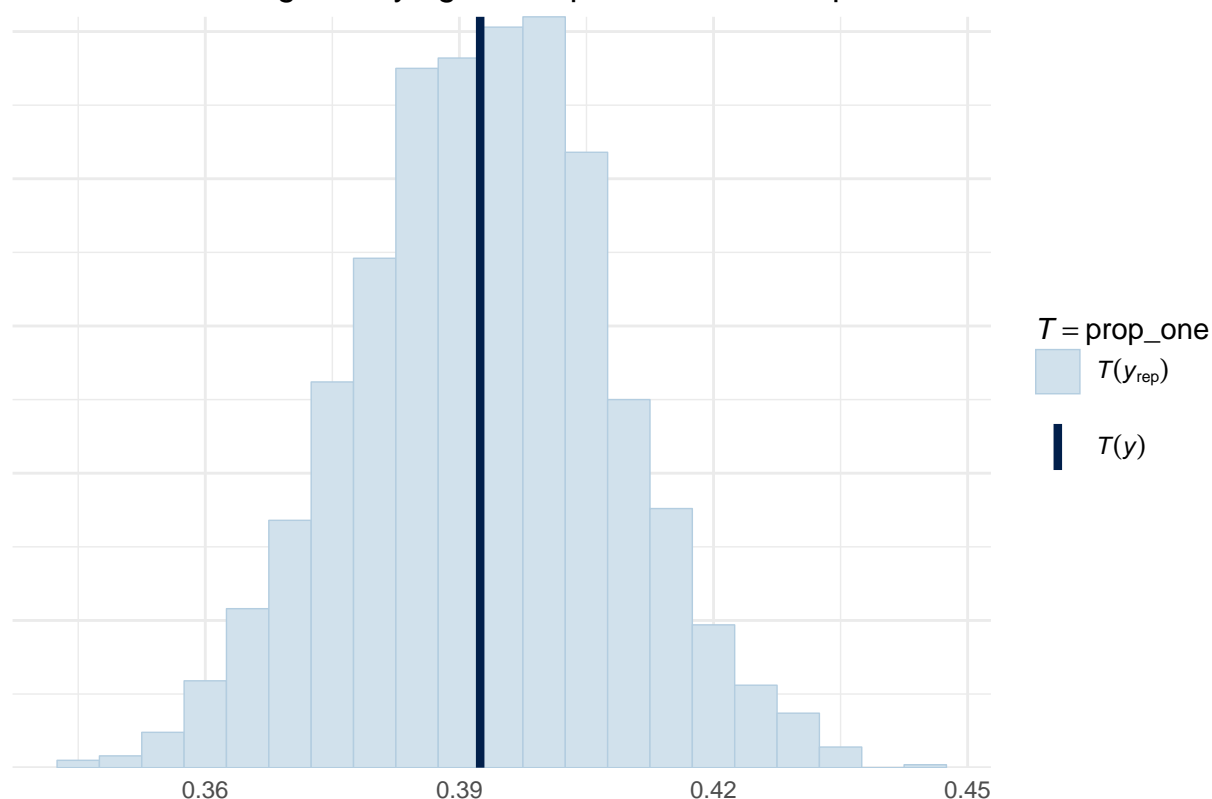
```
plot(varying_inter_nopred_mod, "trace", pars = "(Intercept)") +  
  ggplot2::ggtitle("Traceplot of intercept in varying intercept model")
```



Posterior checking

```
prop_one <- function(y) mean(y == 1)  
prop_zero_test1 <- pp_check(varying_inter_nopred_mod, plotfun = "stat",  
                             stat = "prop_one", binwidth=0.005)  
prop_zero_test1 + ggtitle("Posterior checking of varying intercept model with no predictors")
```

Posterior checking of varying intercept model with no predictors



Model 2. Varying intercept model with predictors

all the predictors have the same slope, but allows intercept to vary among districts.

```
varying_intercept_mod <- stan_glmer(formula = use ~ 1 + urban + age + livch + (1|district),
  data = cont,
  seed = 271,
  cores = getOption("mc.cores", 4L),
  family = binomial(link = "logit"))
```

```
prior_summary(varying_intercept_mod)
```

```
## Priors for model 'varying_intercept_mod'
## -----
## Intercept (after predictors centered)
## ~ normal(location = 0, scale = 10)
##
## Coefficients
## Specified prior:
## ~ normal(location = [0,0,0,...], scale = [2.5,2.5,2.5,...])
## Adjusted prior:
## ~ normal(location = [0,0,0,...], scale = [2.50,0.28,2.50,...])
##
## Covariance
## ~ decov(reg. = 1, conc. = 1, shape = 1, scale = 1)
## -----
## See help('prior_summary.stanreg') for more details
```

```
summary(varying_intercept_mod, digits = 3)
```

```
##
## Model Info:
## function:      stan_glmer
## family:        binomial [logit]
## formula:       use ~ 1 + urban + age + livch + (1 | district)
## algorithm:     sampling
## sample:        4000 (posterior sample size)
## priors:        see help('prior_summary')
## observations:  1934
## groups:        district (60)
##
## Estimates:
##              mean    sd    10%    50%    90%
## (Intercept) -1.686  0.145 -1.875 -1.683 -1.504
## urbanY       0.729  0.122  0.571  0.730  0.884
## age        -0.026  0.008 -0.036 -0.027 -0.016
## livch1       1.103  0.153  0.906  1.098  1.303
## livch2       1.372  0.172  1.154  1.371  1.591
## livch3+      1.338  0.176  1.118  1.338  1.563
## b[(Intercept) district:1] -0.729  0.214 -1.007 -0.730 -0.462
## b[(Intercept) district:2] -0.041  0.347 -0.475 -0.042  0.393
## b[(Intercept) district:3]  0.226  0.455 -0.350  0.217  0.808
## b[(Intercept) district:4]  0.194  0.309 -0.201  0.195  0.594
## b[(Intercept) district:5]  0.049  0.290 -0.330  0.051  0.418
## b[(Intercept) district:6] -0.230  0.252 -0.563 -0.224  0.091
## b[(Intercept) district:7] -0.148  0.363 -0.611 -0.142  0.309
## b[(Intercept) district:8]  0.104  0.285 -0.270  0.109  0.464
## b[(Intercept) district:9] -0.235  0.324 -0.650 -0.232  0.179
## b[(Intercept) district:10] -0.403  0.398 -0.916 -0.386  0.093
## b[(Intercept) district:11] -0.763  0.417 -1.299 -0.739 -0.256
## b[(Intercept) district:12] -0.089  0.302 -0.468 -0.088  0.294
## b[(Intercept) district:13]  0.125  0.330 -0.288  0.127  0.539
## b[(Intercept) district:14]  0.612  0.218  0.330  0.615  0.888
## b[(Intercept) district:15] -0.057  0.338 -0.488 -0.047  0.370
## b[(Intercept) district:16]  0.563  0.346  0.128  0.556  1.002
## b[(Intercept) district:17] -0.143  0.332 -0.567 -0.138  0.279
## b[(Intercept) district:18] -0.081  0.270 -0.430 -0.081  0.257
## b[(Intercept) district:19]  0.113  0.317 -0.293  0.112  0.528
## b[(Intercept) district:20]  0.200  0.361 -0.261  0.199  0.658
## b[(Intercept) district:21]  0.004  0.349 -0.437  0.007  0.437
## b[(Intercept) district:22] -0.389  0.349 -0.841 -0.383  0.051
## b[(Intercept) district:23] -0.176  0.362 -0.649 -0.169  0.282
## b[(Intercept) district:24] -0.492  0.407 -1.022 -0.472  0.004
## b[(Intercept) district:25]  0.200  0.235 -0.100  0.201  0.506
## b[(Intercept) district:26]  0.020  0.384 -0.468  0.015  0.502
## b[(Intercept) district:27] -0.487  0.300 -0.876 -0.479 -0.107
## b[(Intercept) district:28] -0.360  0.276 -0.706 -0.358 -0.016
## b[(Intercept) district:29] -0.126  0.317 -0.535 -0.120  0.277
## b[(Intercept) district:30]  0.436  0.247  0.122  0.438  0.751
## b[(Intercept) district:31]  0.287  0.295 -0.089  0.284  0.663
## b[(Intercept) district:32] -0.411  0.334 -0.834 -0.398  0.009
## b[(Intercept) district:33] -0.083  0.362 -0.539 -0.081  0.385
```



```

## b[(Intercept) district:34]      0.740  0.307  0.349  0.725  1.155
## b[(Intercept) district:35]      0.232  0.267 -0.103  0.231  0.581
## b[(Intercept) district:36]     -0.097  0.363 -0.555 -0.097  0.357
## b[(Intercept) district:37]      0.337  0.371 -0.147  0.334  0.817
## b[(Intercept) district:38]     -0.203  0.382 -0.703 -0.191  0.278
## b[(Intercept) district:39]      0.469  0.326  0.062  0.467  0.893
## b[(Intercept) district:40]      0.138  0.276 -0.216  0.132  0.491
## b[(Intercept) district:41]      0.319  0.325 -0.088  0.311  0.743
## b[(Intercept) district:42]      0.222  0.385 -0.265  0.216  0.723
## b[(Intercept) district:43]      0.495  0.276  0.139  0.491  0.852
## b[(Intercept) district:44]     -0.292  0.338 -0.732 -0.288  0.131
## b[(Intercept) district:45]     -0.165  0.288 -0.539 -0.153  0.200
## b[(Intercept) district:46]      0.573  0.216  0.293  0.575  0.848
## b[(Intercept) district:47]      0.207  0.368 -0.255  0.206  0.669
## b[(Intercept) district:48]      0.440  0.283  0.075  0.438  0.795
## b[(Intercept) district:49]     -0.211  0.474 -0.819 -0.198  0.378
## b[(Intercept) district:50]      0.303  0.351 -0.141  0.300  0.758
## b[(Intercept) district:51]      0.053  0.280 -0.311  0.048  0.418
## b[(Intercept) district:52]      0.198  0.246 -0.118  0.196  0.516
## b[(Intercept) district:53]     -0.121  0.362 -0.584 -0.122  0.334
## b[(Intercept) district:55]     -0.341  0.419 -0.882 -0.334  0.196
## b[(Intercept) district:56]      0.595  0.283  0.236  0.594  0.962
## b[(Intercept) district:57]     -0.449  0.345 -0.888 -0.440 -0.010
## b[(Intercept) district:58]      0.231  0.297 -0.148  0.232  0.610
## b[(Intercept) district:59]     -0.439  0.414 -0.955 -0.430  0.075
## b[(Intercept) district:60]     -0.413  0.314 -0.824 -0.402 -0.032
## b[(Intercept) district:61]     -0.555  0.295 -0.938 -0.547 -0.181
## Sigma[district:(Intercept),(Intercept)] 0.242  0.083  0.149  0.232  0.346
##
## Fit Diagnostics:
##           mean    sd   10%   50%   90%
## mean_PPD 0.393  0.015 0.374 0.392 0.412
##
## The mean_ppd is the sample average posterior predictive distribution of the outcome variable (for de
##
## MCMC diagnostics
##
##           mcse  Rhat  n_eff
## (Intercept) 0.003 0.999 2143
## urbanY      0.002 1.000 4842
## age         0.000 1.000 3272
## livch1      0.003 1.000 3584
## livch2      0.003 1.000 2965
## livch3+     0.004 0.999 2305
## b[(Intercept) district:1] 0.003 1.000 4266
## b[(Intercept) district:2] 0.004 1.000 5978
## b[(Intercept) district:3] 0.006 1.000 6378
## b[(Intercept) district:4] 0.004 1.000 6126
## b[(Intercept) district:5] 0.004 1.000 5210
## b[(Intercept) district:6] 0.003 1.000 5437
## b[(Intercept) district:7] 0.005 1.000 6412
## b[(Intercept) district:8] 0.004 0.999 4445
## b[(Intercept) district:9] 0.004 1.000 6237
## b[(Intercept) district:10] 0.005 1.000 5493
## b[(Intercept) district:11] 0.006 1.000 4584

```

```

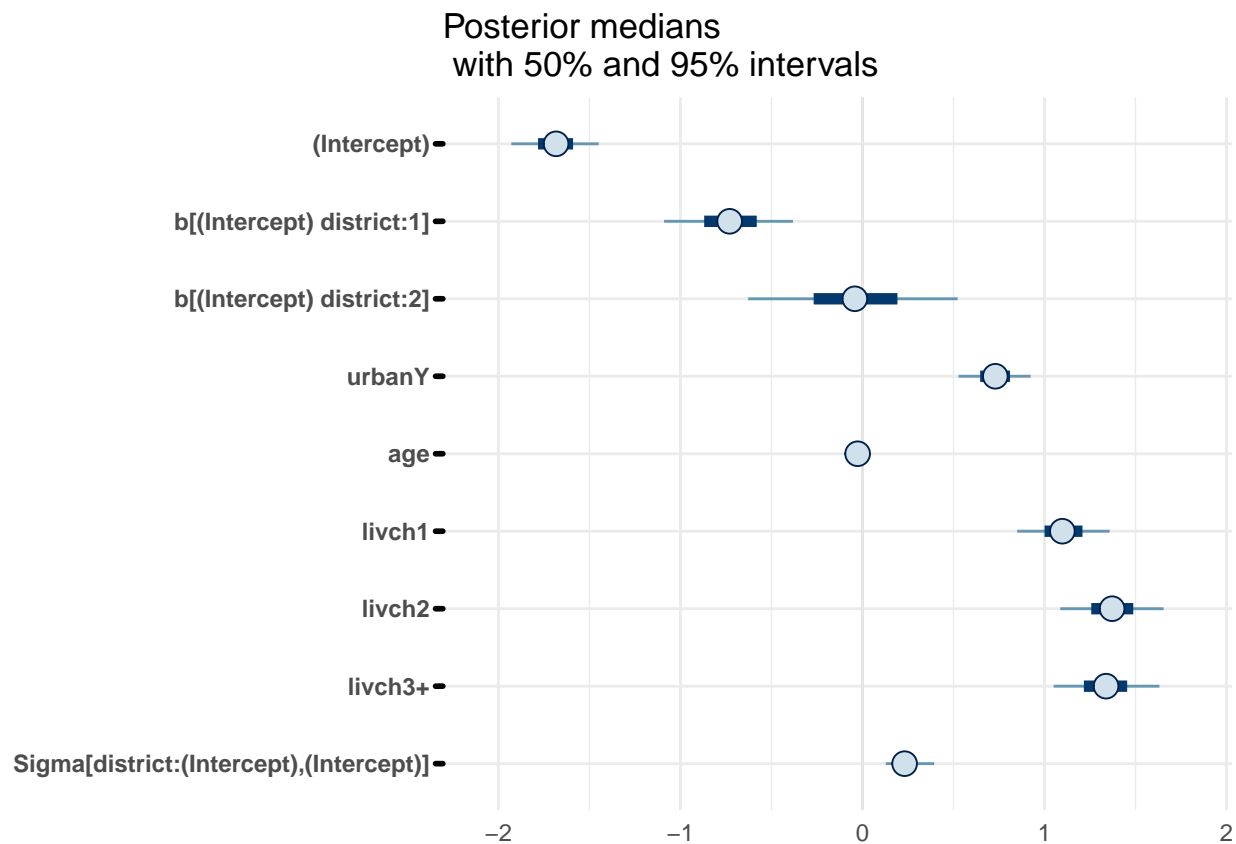
## b[(Intercept) district:12]          0.005 1.000 4501
## b[(Intercept) district:13]          0.004 0.999 6217
## b[(Intercept) district:14]          0.004 1.000 3665
## b[(Intercept) district:15]          0.004 0.999 6145
## b[(Intercept) district:16]          0.005 1.000 5244
## b[(Intercept) district:17]          0.004 1.000 5653
## b[(Intercept) district:18]          0.004 1.000 5680
## b[(Intercept) district:19]          0.004 1.000 6254
## b[(Intercept) district:20]          0.005 0.999 6041
## b[(Intercept) district:21]          0.004 0.999 6872
## b[(Intercept) district:22]          0.005 1.000 5425
## b[(Intercept) district:23]          0.005 1.000 5619
## b[(Intercept) district:24]          0.005 0.999 6021
## b[(Intercept) district:25]          0.004 1.000 4466
## b[(Intercept) district:26]          0.005 1.000 5702
## b[(Intercept) district:27]          0.004 0.999 4849
## b[(Intercept) district:28]          0.004 1.000 5076
## b[(Intercept) district:29]          0.004 0.999 7100
## b[(Intercept) district:30]          0.004 1.000 4775
## b[(Intercept) district:31]          0.004 1.000 5526
## b[(Intercept) district:32]          0.005 0.999 5248
## b[(Intercept) district:33]          0.005 1.000 6007
## b[(Intercept) district:34]          0.005 1.000 4176
## b[(Intercept) district:35]          0.003 1.000 6067
## b[(Intercept) district:36]          0.005 1.001 4680
## b[(Intercept) district:37]          0.005 1.000 5444
## b[(Intercept) district:38]          0.005 0.999 6624
## b[(Intercept) district:39]          0.005 0.999 4746
## b[(Intercept) district:40]          0.004 1.000 4888
## b[(Intercept) district:41]          0.004 0.999 5820
## b[(Intercept) district:42]          0.005 1.000 5873
## b[(Intercept) district:43]          0.004 1.000 4915
## b[(Intercept) district:44]          0.004 1.000 6647
## b[(Intercept) district:45]          0.004 1.000 4808
## b[(Intercept) district:46]          0.003 1.000 4686
## b[(Intercept) district:47]          0.005 1.000 4912
## b[(Intercept) district:48]          0.004 1.000 5647
## b[(Intercept) district:49]          0.006 0.999 6431
## b[(Intercept) district:50]          0.005 1.000 5399
## b[(Intercept) district:51]          0.004 1.000 5680
## b[(Intercept) district:52]          0.003 0.999 5515
## b[(Intercept) district:53]          0.005 1.000 5174
## b[(Intercept) district:55]          0.005 1.000 6301
## b[(Intercept) district:56]          0.004 0.999 5361
## b[(Intercept) district:57]          0.005 1.000 5653
## b[(Intercept) district:58]          0.004 0.999 5262
## b[(Intercept) district:59]          0.006 1.000 4746
## b[(Intercept) district:60]          0.004 0.999 5790
## b[(Intercept) district:61]          0.004 0.999 5122
## Sigma[district:(Intercept),(Intercept)] 0.002 1.001 1690
## mean_PPD                               0.000 1.000 4686
## log-posterior                          0.246 1.001 1027
##
## For each parameter, mcse is Monte Carlo standard error, n_eff is a crude measure of effective sample

```

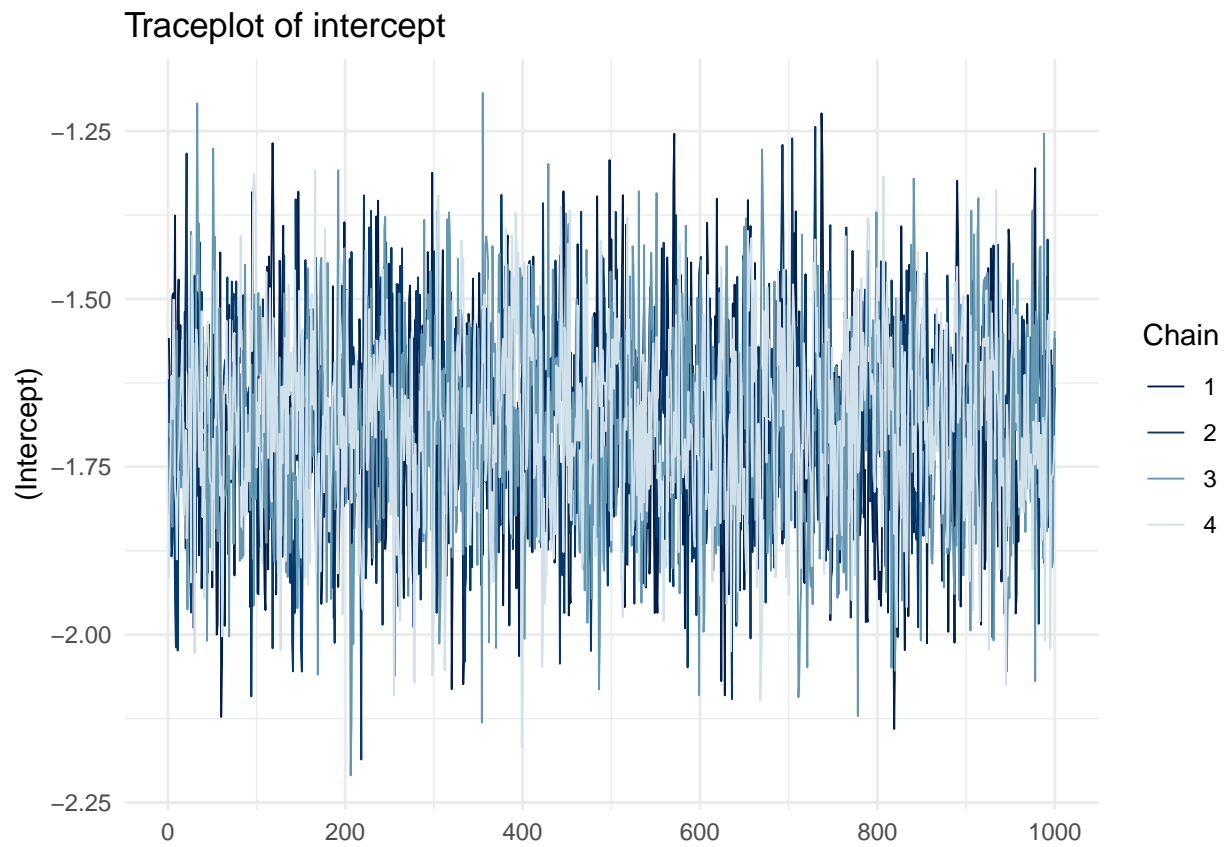
```
summary(varying_intercept_mod, digits = 3,
  pars=c("(Intercept)", "b[(Intercept) district:1]",
    "b[(Intercept) district:2]", "urbanY", "age", "livch1",
    "livch2", "livch3+",
    "Sigma[district:(Intercept),(Intercept)]"),
  probs = c(0.025, 0.5, 0.975))
```

Plotting

```
plot(varying_intercept_mod, pars=c("(Intercept)", "b[(Intercept) district:1]",
  "b[(Intercept) district:2]", "urbanY",
  "age", "livch1", "livch2", "livch3+",
  "Sigma[district:(Intercept),(Intercept)]")) +
  ggplot2::ggtitle("Posterior medians \n with 50% and 95% intervals")
```

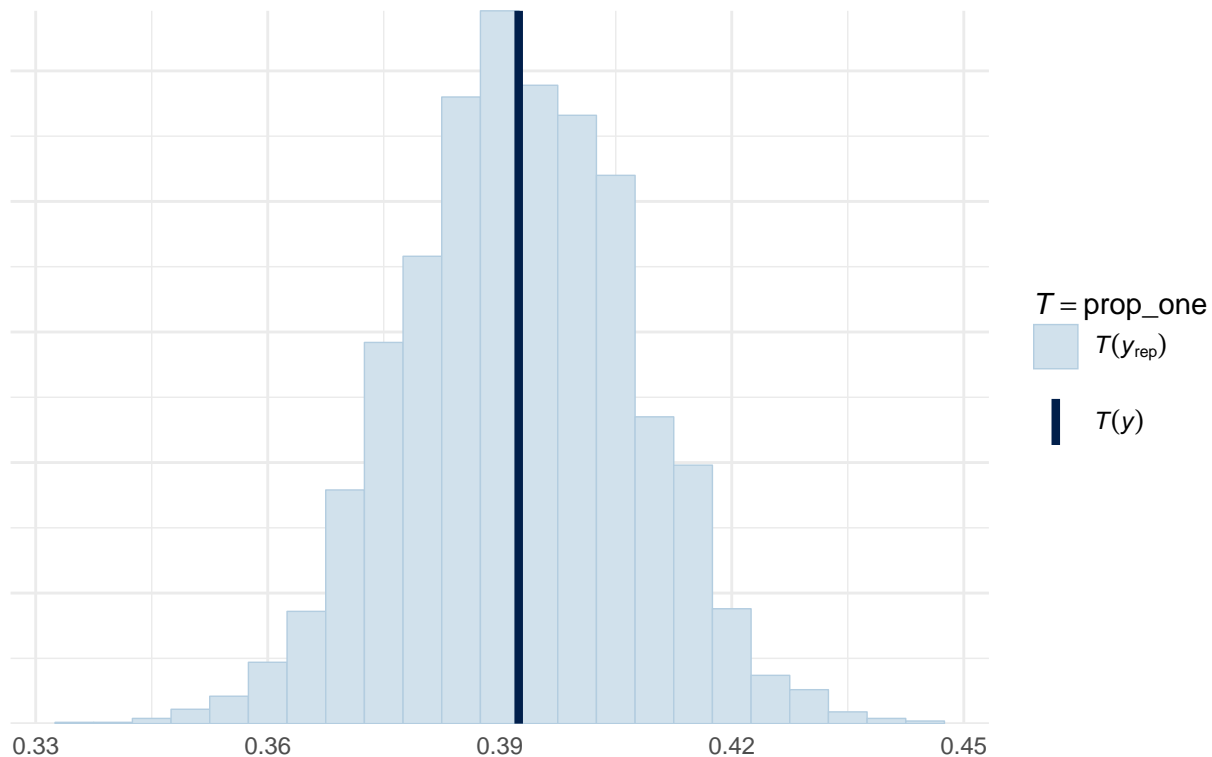


```
plot(varying_intercept_mod, "trace", pars = "(Intercept)") +
  ggplot2::ggtitle("Traceplot of intercept")
```



```
prop_zero_test2 <- pp_check(varying_intercept_mod, plotfun = "stat",  
                             stat = "prop_one", binwidth=0.005)  
prop_zero_test2 + ggtitle("Posterior checking of varying intercept  
model with predictors")
```

Posterior checking of varying intercept model with predictors



Model 3. Varying intercept varying slope model

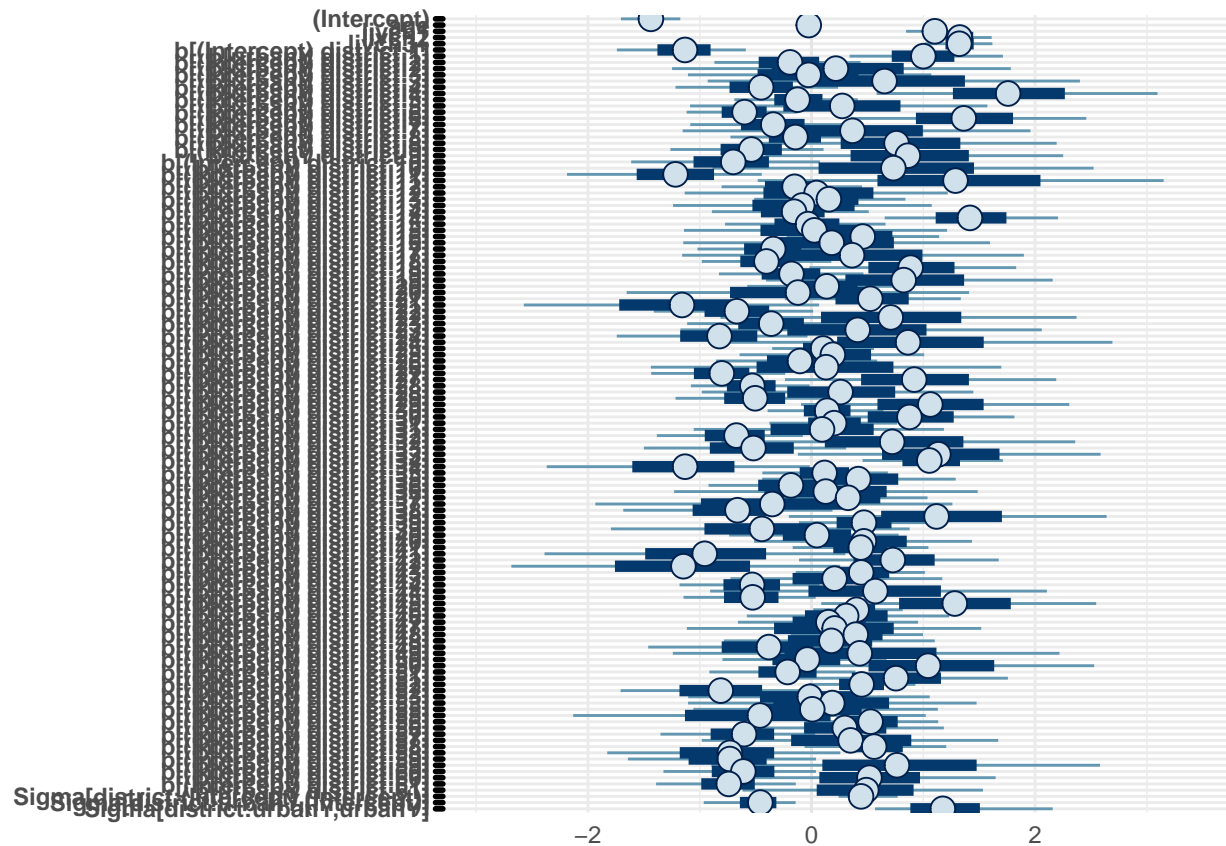
age and livch have the same slope, but allows the slope of urban and intercept to vary among the districts.

```
varying_inter_slope_mod <- stan_glmmer(formula = use ~ 1 + age + livch + (urban|district),
  data = cont,
  seed = 271,
  cores = getOption("mc.cores", 4L),
  family = binomial(link = "logit"))
```

```
prior_summary(varying_inter_slope_mod)
```

```
## Priors for model 'varying_inter_slope_mod'
## -----
## Intercept (after predictors centered)
## ~ normal(location = 0, scale = 10)
##
## Coefficients
##   Specified prior:
##     ~ normal(location = [0,0,0,...], scale = [2.5,2.5,2.5,...])
##   Adjusted prior:
##     ~ normal(location = [0,0,0,...], scale = [0.28,2.50,2.50,...])
##
## Covariance
## ~ decov(reg. = 1, conc. = 1, shape = 1, scale = 1)
## -----
## See help('prior_summary.stanreg') for more details
```

```
plot(varying_inter_slope_mod)
```



```
summary(varying_inter_slope_mod, digits = 3)
```

```
##
## Model Info:
## function:      stan_glm
## family:        binomial [logit]
## formula:       use ~ 1 + age + livch + (urban | district)
## algorithm:     sampling
## sample:        4000 (posterior sample size)
## priors:        see help('prior_summary')
## observations:  1934
## groups:        district (60)
##
## Estimates:
```

	mean	sd	10%	50%	90%
## (Intercept)	-1.438	0.159	-1.644	-1.437	-1.234
## age	-0.025	0.008	-0.035	-0.025	-0.015
## livch1	1.106	0.158	0.904	1.103	1.309
## livch2	1.326	0.176	1.100	1.325	1.549
## livch3+	1.322	0.182	1.086	1.322	1.559
## b[(Intercept) district:1]	-1.144	0.352	-1.598	-1.132	-0.697
## b[urbanY district:1]	1.007	0.418	0.496	1.000	1.555
## b[(Intercept) district:2]	-0.202	0.396	-0.710	-0.193	0.294
## b[urbanY district:2]	0.238	0.934	-0.928	0.218	1.405
## b[(Intercept) district:3]	-0.031	0.679	-0.886	-0.026	0.817

## b[urbanY district:3]	0.705	1.024	-0.558	0.653	2.041
## b[(Intercept) district:4]	-0.458	0.436	-1.017	-0.451	0.091
## b[urbanY district:4]	1.786	0.756	0.849	1.758	2.754
## b[(Intercept) district:5]	-0.125	0.335	-0.562	-0.125	0.294
## b[urbanY district:5]	0.270	0.816	-0.752	0.275	1.283
## b[(Intercept) district:6]	-0.608	0.300	-1.001	-0.599	-0.227
## b[urbanY district:6]	1.377	0.645	0.557	1.362	2.204
## b[(Intercept) district:7]	-0.357	0.434	-0.927	-0.342	0.188
## b[urbanY district:7]	0.383	0.954	-0.752	0.366	1.584
## b[(Intercept) district:8]	-0.151	0.348	-0.597	-0.144	0.301
## b[urbanY district:8]	0.812	0.810	-0.200	0.760	1.878
## b[(Intercept) district:9]	-0.546	0.416	-1.080	-0.539	-0.019
## b[urbanY district:9]	0.885	0.810	-0.142	0.861	1.920
## b[(Intercept) district:10]	-0.727	0.517	-1.394	-0.700	-0.078
## b[urbanY district:10]	0.762	1.044	-0.520	0.732	2.114
## b[(Intercept) district:11]	-1.244	0.533	-1.936	-1.217	-0.595
## b[urbanY district:11]	1.318	1.122	-0.095	1.288	2.708
## b[(Intercept) district:12]	-0.163	0.386	-0.659	-0.152	0.328
## b[urbanY district:12]	0.057	0.730	-0.886	0.045	0.987
## b[(Intercept) district:13]	0.153	0.414	-0.370	0.155	0.681
## b[urbanY district:13]	-0.079	0.699	-0.964	-0.083	0.807
## b[(Intercept) district:14]	-0.171	0.426	-0.730	-0.156	0.366
## b[urbanY district:14]	1.425	0.469	0.839	1.417	2.023
## b[(Intercept) district:15]	-0.041	0.439	-0.616	-0.031	0.515
## b[urbanY district:15]	0.028	0.719	-0.869	0.026	0.954
## b[(Intercept) district:16]	0.459	0.411	-0.060	0.458	0.966
## b[urbanY district:16]	0.194	0.840	-0.856	0.181	1.240
## b[(Intercept) district:17]	-0.351	0.390	-0.860	-0.342	0.140
## b[urbanY district:17]	0.371	0.940	-0.794	0.362	1.555
## b[(Intercept) district:18]	-0.402	0.357	-0.850	-0.404	0.057
## b[urbanY district:18]	0.897	0.568	0.188	0.886	1.623
## b[(Intercept) district:19]	-0.183	0.393	-0.694	-0.179	0.328
## b[urbanY district:19]	0.843	0.790	-0.158	0.824	1.832
## b[(Intercept) district:20]	0.129	0.423	-0.414	0.136	0.656
## b[urbanY district:20]	-0.121	0.941	-1.297	-0.121	1.078
## b[(Intercept) district:21]	0.540	0.486	-0.078	0.526	1.165
## b[urbanY district:21]	-1.198	0.808	-2.246	-1.159	-0.186
## b[(Intercept) district:22]	-0.679	0.436	-1.254	-0.668	-0.121
## b[urbanY district:22]	0.730	0.966	-0.481	0.708	1.998
## b[(Intercept) district:23]	-0.370	0.444	-0.944	-0.361	0.191
## b[urbanY district:23]	0.413	0.983	-0.825	0.413	1.691
## b[(Intercept) district:24]	-0.849	0.520	-1.505	-0.824	-0.209
## b[urbanY district:24]	0.906	1.033	-0.361	0.863	2.273
## b[(Intercept) district:25]	0.108	0.279	-0.253	0.103	0.464
## b[urbanY district:25]	0.190	0.501	-0.453	0.191	0.827
## b[(Intercept) district:26]	-0.112	0.439	-0.667	-0.104	0.444
## b[urbanY district:26]	0.126	0.948	-1.050	0.130	1.333
## b[(Intercept) district:27]	-0.813	0.368	-1.289	-0.802	-0.352
## b[urbanY district:27]	0.937	0.742	0.014	0.919	1.895
## b[(Intercept) district:28]	-0.536	0.324	-0.953	-0.530	-0.119
## b[urbanY district:28]	0.252	0.745	-0.671	0.260	1.194
## b[(Intercept) district:29]	-0.515	0.412	-1.048	-0.504	0.005
## b[urbanY district:29]	1.072	0.714	0.162	1.063	1.996
## b[(Intercept) district:30]	0.135	0.310	-0.268	0.139	0.530

## b[urbanY district:30]	0.897	0.552	0.197	0.876	1.623
## b[(Intercept) district:31]	0.202	0.349	-0.242	0.202	0.654
## b[urbanY district:31]	0.086	0.685	-0.781	0.094	0.950
## b[(Intercept) district:32]	-0.692	0.400	-1.226	-0.672	-0.197
## b[urbanY district:32]	0.735	0.957	-0.458	0.721	1.979
## b[(Intercept) district:33]	-0.548	0.556	-1.271	-0.519	0.140
## b[urbanY district:33]	1.171	0.813	0.153	1.135	2.225
## b[(Intercept) district:34]	1.072	0.383	0.584	1.054	1.568
## b[urbanY district:34]	-1.155	0.708	-2.064	-1.131	-0.282
## b[(Intercept) district:35]	0.118	0.341	-0.301	0.117	0.554
## b[urbanY district:35]	0.415	0.532	-0.267	0.418	1.102
## b[(Intercept) district:36]	-0.190	0.436	-0.771	-0.184	0.361
## b[urbanY district:36]	0.122	0.825	-0.922	0.127	1.168
## b[(Intercept) district:37]	0.322	0.438	-0.240	0.327	0.887
## b[urbanY district:37]	-0.342	0.976	-1.562	-0.351	0.906
## b[(Intercept) district:38]	-0.697	0.577	-1.436	-0.663	0.009
## b[urbanY district:38]	1.168	0.840	0.126	1.118	2.283
## b[(Intercept) district:39]	0.472	0.368	0.010	0.469	0.947
## b[urbanY district:39]	-0.449	0.802	-1.458	-0.444	0.567
## b[(Intercept) district:40]	0.041	0.459	-0.554	0.048	0.615
## b[urbanY district:40]	0.465	0.583	-0.271	0.468	1.208
## b[(Intercept) district:41]	0.442	0.366	-0.027	0.441	0.904
## b[urbanY district:41]	-0.976	0.819	-2.050	-0.955	0.039
## b[(Intercept) district:42]	0.743	0.546	0.071	0.727	1.438
## b[urbanY district:42]	-1.175	0.894	-2.331	-1.146	-0.076
## b[(Intercept) district:43]	0.445	0.356	-0.009	0.444	0.903
## b[urbanY district:43]	0.213	0.574	-0.509	0.206	0.941
## b[(Intercept) district:44]	-0.539	0.381	-1.028	-0.531	-0.054
## b[urbanY district:44]	0.575	0.930	-0.572	0.571	1.745
## b[(Intercept) district:45]	-0.541	0.362	-1.012	-0.527	-0.084
## b[urbanY district:45]	1.296	0.739	0.375	1.281	2.241
## b[(Intercept) district:46]	0.406	0.244	0.103	0.403	0.721
## b[urbanY district:46]	0.314	0.553	-0.387	0.310	1.014
## b[(Intercept) district:47]	0.155	0.486	-0.470	0.150	0.772
## b[urbanY district:47]	0.201	0.811	-0.809	0.207	1.261
## b[(Intercept) district:48]	0.401	0.359	-0.056	0.395	0.867
## b[urbanY district:48]	0.167	0.570	-0.579	0.178	0.884
## b[(Intercept) district:49]	-0.407	0.609	-1.193	-0.381	0.353
## b[urbanY district:49]	0.453	1.056	-0.853	0.431	1.835
## b[(Intercept) district:50]	-0.050	0.455	-0.634	-0.037	0.521
## b[urbanY district:50]	1.084	0.848	0.023	1.043	2.190
## b[(Intercept) district:51]	-0.226	0.395	-0.737	-0.214	0.271
## b[urbanY district:51]	0.759	0.601	-0.011	0.755	1.541
## b[(Intercept) district:52]	0.448	0.294	0.071	0.448	0.824
## b[urbanY district:52]	-0.817	0.538	-1.516	-0.812	-0.132
## b[(Intercept) district:53]	-0.022	0.662	-0.858	-0.015	0.825
## b[urbanY district:53]	0.182	0.781	-0.822	0.186	1.174
## b[(Intercept) district:55]	0.020	0.676	-0.795	0.007	0.871
## b[urbanY district:55]	-0.499	0.959	-1.747	-0.462	0.704
## b[(Intercept) district:56]	0.531	0.366	0.062	0.528	1.001
## b[urbanY district:56]	0.299	0.535	-0.378	0.297	0.960
## b[(Intercept) district:57]	-0.625	0.424	-1.181	-0.603	-0.098
## b[urbanY district:57]	0.348	0.807	-0.680	0.349	1.366
## b[(Intercept) district:58]	0.562	0.384	0.070	0.560	1.055


```

## b[urbanY district:58] -0.756 0.628 -1.586 -0.730 0.033
## b[(Intercept) district:59] -0.763 0.522 -1.432 -0.731 -0.132
## b[urbanY district:59] 0.807 1.073 -0.501 0.764 2.163
## b[(Intercept) district:60] -0.621 0.417 -1.159 -0.610 -0.100
## b[urbanY district:60] 0.523 0.673 -0.345 0.518 1.399
## b[(Intercept) district:61] -0.751 0.377 -1.236 -0.741 -0.275
## b[urbanY district:61] 0.478 0.653 -0.368 0.490 1.303
## Sigma[district:(Intercept),(Intercept)] 0.466 0.166 0.282 0.443 0.680
## Sigma[district:urbanY,(Intercept)] -0.495 0.261 -0.823 -0.458 -0.203
## Sigma[district:urbanY,urbanY] 1.244 0.517 0.679 1.174 1.886
##
## Fit Diagnostics:
##      mean    sd   10%   50%   90%
## mean_PPD 0.392 0.014 0.374 0.392 0.411
##
## The mean_ppd is the sample average posterior predictive distribution of the outcome variable (for de
##
## MCMC diagnostics
##
##      mcse  Rhat  n_eff
## (Intercept) 0.004 1.002 1262
## age 0.000 1.001 3493
## livch1 0.003 1.000 3552
## livch2 0.003 1.001 2932
## livch3+ 0.004 1.001 2526
## b[(Intercept) district:1] 0.006 1.000 3212
## b[urbanY district:1] 0.006 1.000 5024
## b[(Intercept) district:2] 0.007 1.001 3376
## b[urbanY district:2] 0.012 1.000 5614
## b[(Intercept) district:3] 0.009 1.001 5313
## b[urbanY district:3] 0.014 0.999 5188
## b[(Intercept) district:4] 0.009 1.002 2154
## b[urbanY district:4] 0.012 0.999 3716
## b[(Intercept) district:5] 0.006 1.001 3541
## b[urbanY district:5] 0.011 1.000 5962
## b[(Intercept) district:6] 0.006 1.001 2166
## b[urbanY district:6] 0.009 0.999 5456
## b[(Intercept) district:7] 0.006 1.001 4606
## b[urbanY district:7] 0.015 1.000 4136
## b[(Intercept) district:8] 0.006 1.001 3431
## b[urbanY district:8] 0.012 1.000 4918
## b[(Intercept) district:9] 0.008 1.001 2944
## b[urbanY district:9] 0.012 1.000 4209
## b[(Intercept) district:10] 0.008 1.000 4071
## b[urbanY district:10] 0.018 1.000 3215
## b[(Intercept) district:11] 0.011 1.001 2487
## b[urbanY district:11] 0.026 1.002 1792
## b[(Intercept) district:12] 0.006 1.000 4537
## b[urbanY district:12] 0.010 1.000 5626
## b[(Intercept) district:13] 0.005 1.000 5744
## b[urbanY district:13] 0.009 0.999 6636
## b[(Intercept) district:14] 0.009 1.001 2377
## b[urbanY district:14] 0.007 1.000 3927
## b[(Intercept) district:15] 0.006 0.999 5128
## b[urbanY district:15] 0.009 0.999 6569

```

## b[(Intercept) district:16]	0.006	1.001	5138
## b[urbanY district:16]	0.012	0.999	5245
## b[(Intercept) district:17]	0.006	1.000	4158
## b[urbanY district:17]	0.014	1.000	4716
## b[(Intercept) district:18]	0.006	1.002	3484
## b[urbanY district:18]	0.007	1.000	5872
## b[(Intercept) district:19]	0.007	1.000	3583
## b[urbanY district:19]	0.011	1.000	4796
## b[(Intercept) district:20]	0.006	1.000	4461
## b[urbanY district:20]	0.011	0.999	6735
## b[(Intercept) district:21]	0.007	1.000	5269
## b[urbanY district:21]	0.013	1.001	3821
## b[(Intercept) district:22]	0.006	1.000	4724
## b[urbanY district:22]	0.017	1.001	3413
## b[(Intercept) district:23]	0.007	1.000	4472
## b[urbanY district:23]	0.015	1.000	4193
## b[(Intercept) district:24]	0.009	1.000	3515
## b[urbanY district:24]	0.019	1.001	3012
## b[(Intercept) district:25]	0.005	1.000	2989
## b[urbanY district:25]	0.006	1.000	6057
## b[(Intercept) district:26]	0.006	0.999	4712
## b[urbanY district:26]	0.013	1.000	5469
## b[(Intercept) district:27]	0.006	1.000	3294
## b[urbanY district:27]	0.011	1.000	4227
## b[(Intercept) district:28]	0.005	1.000	3982
## b[urbanY district:28]	0.012	1.000	4032
## b[(Intercept) district:29]	0.007	1.000	3171
## b[urbanY district:29]	0.010	0.999	5193
## b[(Intercept) district:30]	0.006	1.001	2526
## b[urbanY district:30]	0.007	1.000	5962
## b[(Intercept) district:31]	0.006	1.000	3713
## b[urbanY district:31]	0.010	1.000	5116
## b[(Intercept) district:32]	0.007	1.000	3735
## b[urbanY district:32]	0.016	1.000	3400
## b[(Intercept) district:33]	0.010	1.001	3149
## b[urbanY district:33]	0.011	1.000	5383
## b[(Intercept) district:34]	0.005	1.001	5341
## b[urbanY district:34]	0.011	0.999	4520
## b[(Intercept) district:35]	0.005	1.000	4866
## b[urbanY district:35]	0.006	0.999	7500
## b[(Intercept) district:36]	0.006	1.000	5881
## b[urbanY district:36]	0.011	1.000	5360
## b[(Intercept) district:37]	0.006	1.000	5603
## b[urbanY district:37]	0.013	1.000	5343
## b[(Intercept) district:38]	0.011	1.000	2574
## b[urbanY district:38]	0.013	1.000	4173
## b[(Intercept) district:39]	0.005	0.999	5420
## b[urbanY district:39]	0.010	0.999	5865
## b[(Intercept) district:40]	0.007	1.000	4491
## b[urbanY district:40]	0.008	0.999	5736
## b[(Intercept) district:41]	0.005	1.000	4858
## b[urbanY district:41]	0.012	1.000	4483
## b[(Intercept) district:42]	0.009	1.001	4047
## b[urbanY district:42]	0.014	1.000	3978

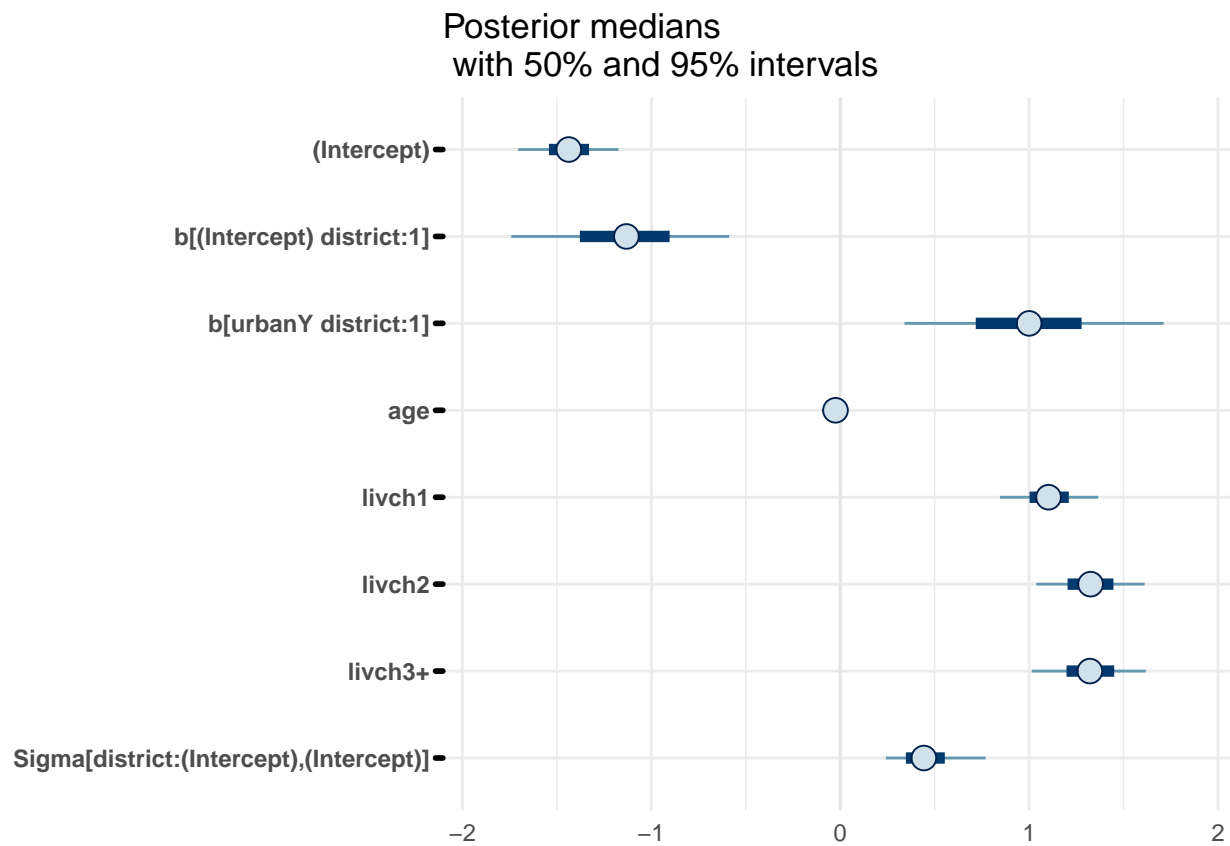
```
## b[(Intercept) district:43]          0.005 1.000 4351
## b[urbanY district:43]              0.007 0.999 6713
## b[(Intercept) district:44]          0.006 1.001 4175
## b[urbanY district:44]              0.015 1.000 4024
## b[(Intercept) district:45]          0.007 1.000 2824
## b[urbanY district:45]              0.012 1.000 4093
## b[(Intercept) district:46]          0.004 1.000 2984
## b[urbanY district:46]              0.007 1.000 6191
## b[(Intercept) district:47]          0.006 1.000 5792
## b[urbanY district:47]              0.010 1.000 6332
## b[(Intercept) district:48]          0.005 1.000 4300
## b[urbanY district:48]              0.008 1.000 5625
## b[(Intercept) district:49]          0.009 1.001 5037
## b[urbanY district:49]              0.017 1.000 3955
## b[(Intercept) district:50]          0.008 1.000 3308
## b[urbanY district:50]              0.013 1.000 4416
## b[(Intercept) district:51]          0.008 1.002 2659
## b[urbanY district:51]              0.008 1.000 5953
## b[(Intercept) district:52]          0.004 1.002 4701
## b[urbanY district:52]              0.006 1.000 7207
## b[(Intercept) district:53]          0.009 1.000 5784
## b[urbanY district:53]              0.010 1.000 5759
## b[(Intercept) district:55]          0.010 1.000 4716
## b[urbanY district:55]              0.014 1.000 4985
## b[(Intercept) district:56]          0.006 1.000 4234
## b[urbanY district:56]              0.007 0.999 6497
## b[(Intercept) district:57]          0.006 1.000 4413
## b[urbanY district:57]              0.012 1.001 4615
## b[(Intercept) district:58]          0.006 1.000 4726
## b[urbanY district:58]              0.008 1.000 5492
## b[(Intercept) district:59]          0.008 1.000 4295
## b[urbanY district:59]              0.019 1.001 3145
## b[(Intercept) district:60]          0.006 1.001 4198
## b[urbanY district:60]              0.010 1.000 5006
## b[(Intercept) district:61]          0.006 0.999 4558
## b[urbanY district:61]              0.009 0.999 5176
## Sigma[district:(Intercept),(Intercept)] 0.005 1.001 1232
## Sigma[district:urbanY,(Intercept)]      0.010 1.007 704
## Sigma[district:urbanY,urbanY]           0.016 1.003 1048
## mean_PPD                               0.000 1.000 4193
## log-posterior                          0.359 1.002 971
##
```

For each parameter, mcse is Monte Carlo standard error, n_eff is a crude measure of effective sample

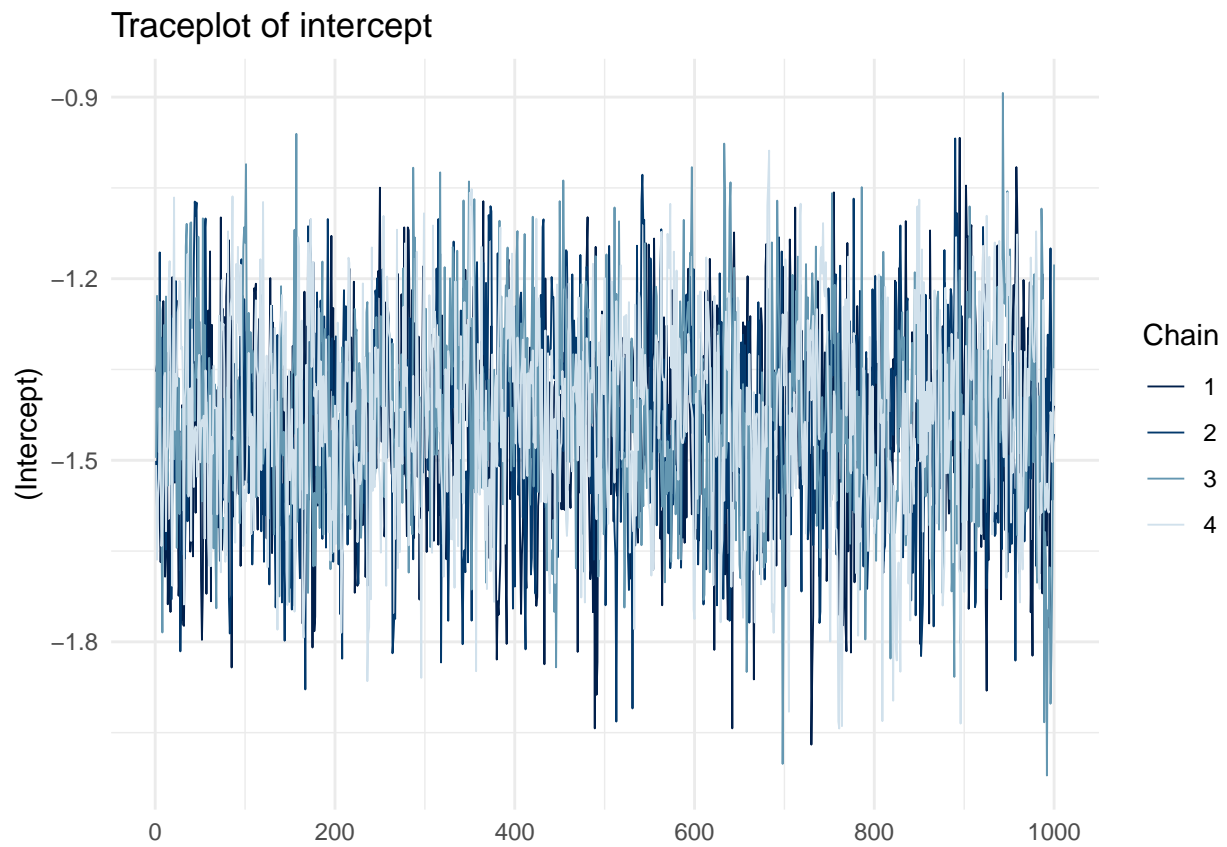
```
summary(varying_inter_slope_mod, digits = 3,
  pars=c("(Intercept)", "b[(Intercept) district:1]",
    "b[urbanY district:1]", "age", "livch1", "livch2",
    "livch3+", "Sigma[district:(Intercept),(Intercept)]"),
  probs = c(0.025, 0.5, 0.975))
```

```
plot(varying_inter_slope_mod, pars=c("(Intercept)",
  "b[(Intercept) district:1]",
  "b[urbanY district:1]", "age",
  "livch1", "livch2", "livch3+",
  "Sigma[district:(Intercept),(Intercept)]")) +
```

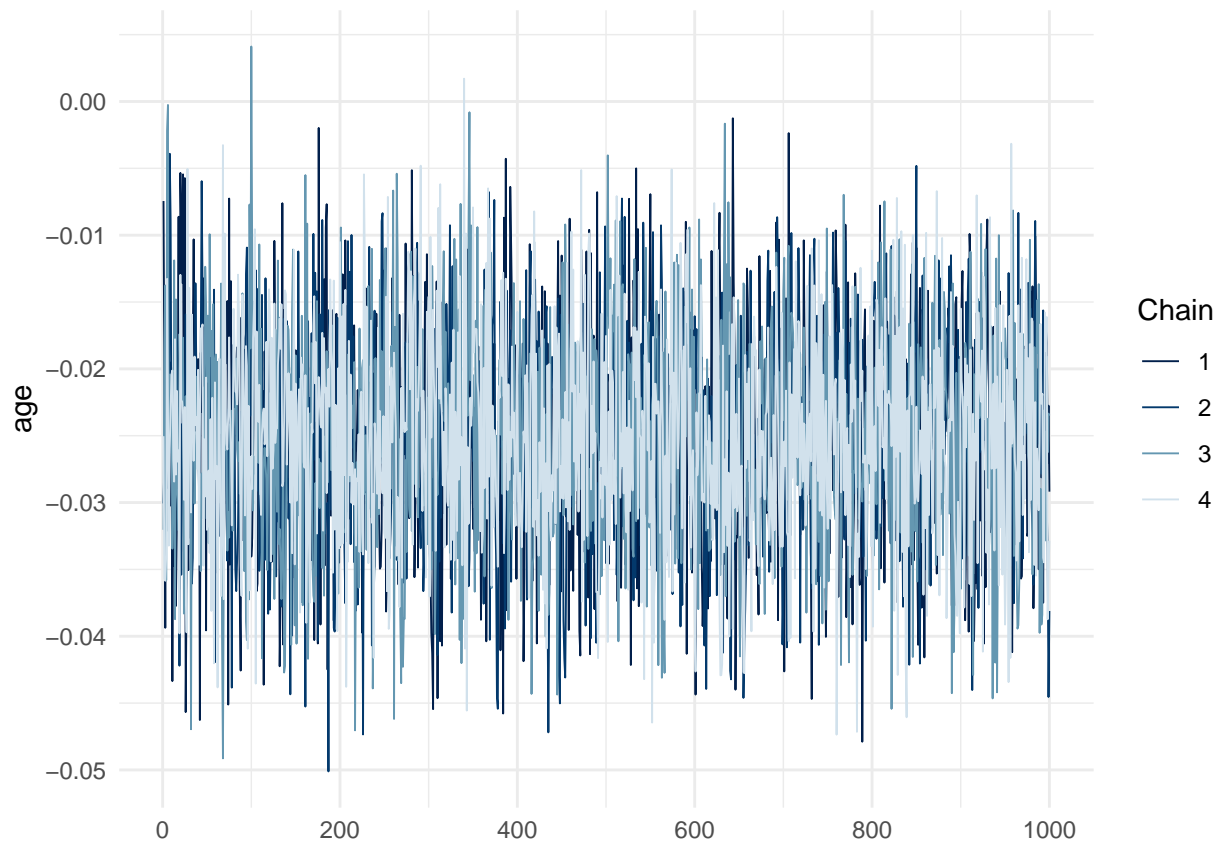
```
ggplot2::ggtitle("Posterior medians \n with 50% and 95% intervals")
```



```
plot(varying_inter_slope_mod, "trace", pars = "(Intercept)") + ggplot2::ggtitle("Traceplot of intercept")
```



```
plot(varying_inter_slope_mod, "trace", pars = "age")
```

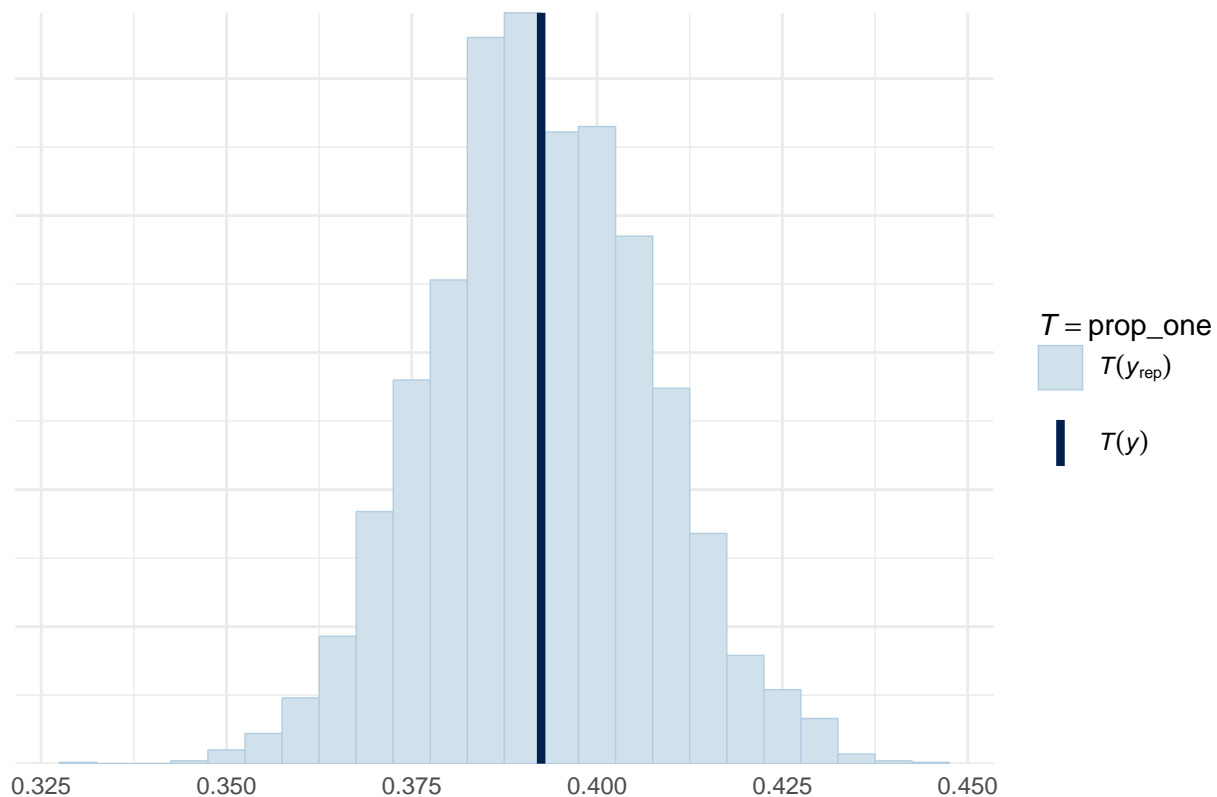


Posterior checking

```
yrep_mod3 <- posterior_predict(varying_inter_slope_mod, draws = 2000)
# mean(ytrue)
# hist(apply(yrep_mod3, 1, mean))

prop_one <- function(y) mean(y == 1)
prop_zero_test3 <- pp_check(varying_inter_slope_mod,
  plotfun = "stat", stat = "prop_one", binwidth=0.005)
prop_zero_test3 + ggtitle("Posterior checking of varying intercept varying slope model")
```

Posterior checking of varying intercept varying slope model



3. Model comparison

```
varying_inter_nopred_mod$waic <- rstanarm::waic(varying_inter_nopred_mod)
varying_intercept_mod$waic <- rstanarm::waic(varying_intercept_mod)
varying_inter_slope_mod$waic <- rstanarm::waic(varying_inter_slope_mod)

varying_inter_nopred_mod$loo <- rstanarm::loo(varying_inter_nopred_mod,
                                              cores = getOption("mc.cores", 4))
varying_intercept_mod$loo <- rstanarm::loo(varying_intercept_mod,
                                             cores = getOption("mc.cores", 4))
varying_inter_slope_mod$loo <- rstanarm::loo(varying_inter_slope_mod,
                                              cores = getOption("mc.cores", 4))

loo_compare(varying_inter_nopred_mod, varying_intercept_mod, varying_inter_slope_mod,
            criterion = "loo")

## Model comparison based on LOO-CV:
##               elpd_diff se_diff
## varying_inter_slope_mod    0.0    0.0
## varying_intercept_mod   -6.3    6.5
## varying_inter_nopred_mod -59.5   10.8

loo_compare(varying_inter_nopred_mod, varying_intercept_mod, varying_inter_slope_mod,
            criterion = "waic")

## Model comparison based on WAIC:
```

```

##                                elpd_diff se_diff
## varying_inter_slope_mod      0.0        0.0
## varying_intercept_mod       -6.5        6.5
## varying_inter_nopred_mod    -59.8       10.8

model_list <- stanreg_list(varying_inter_nopred_mod, varying_intercept_mod, varying_inter_slope_mod)
loo_model_weights(model_list)

## Method: stacking
## -----
##                                weight
## varying_inter_nopred_mod 0.000
## varying_intercept_mod   0.342
## varying_inter_slope_mod  0.658

Bayesian model averaging (not contained in the final report)

fit1 <- brm(formula = use ~ 1 + urban + age + livch + (1|district),
             data = cont,
             seed = 271,
             cores = getOption("mc.cores", 4L),
             family = bernoulli(link = "logit"))

summary(fit1)

fit2 <- brm(formula = use ~ 1 + age + livch + (urban|district),
             data = cont,
             seed = 271,
             cores = getOption("mc.cores", 4L),
             family = bernoulli(link = "logit"))

summary(fit2)
avg_pred <- pp_average(fit1, fit2)

```