

wk7-hw-bayes

Part 1

1. Use at least 2 different methods proving that the groups in section 3.3 of part 1 of the workshop are different.

```
myDataFrame = read.csv( file=paste(dataPath, "wk7/TwoGroupIQ.csv", sep="/") )
y = as.numeric(myDataFrame[, "Score"])
x = as.numeric(as.factor(myDataFrame[, "Group"]))
(xLevels = levels(as.factor(myDataFrame[, "Group"])))
```

```
## [1] "Placebo"    "Smart Drug"
```

```
(Ntotal = length(y))
```

```
## [1] 120
```

```
# Specify the data in a list, for later shipment to JAGS:
dataList = list(
  y = y ,
  x = x ,
  Ntotal = Ntotal ,
  meanY = mean(y) ,
  sdY = sd(y)
)
dataList
```

```
## $y
## [1] 102 107 92 101 110 68 119 106 99 103 90 93 79 89 137 119 126
## [18] 110 71 114 100 95 91 99 97 106 106 129 115 124 137 73 69 95
```



```

    real expLambda;
    unifLo = sdY/1000;
    unifHi = sdY*1000;
    normalSigma = sdY*100;
    expLambda = 1/29.0;
}
parameters {
    real<lower=0> nuMinusOne; //New: definition of additional parameter nu
    real mu[2];
    real<lower=0> sigma[2];
}
transformed parameters {
    real<lower=0> nu;           //New: new parameter nu
    nu=nuMinusOne+1;           //New: shifting nu to avoid zero
}
model {
    sigma ~ uniform(unifLo, unifHi);
    mu ~ normal(meanY, normalSigma);
    nuMinusOne~exponential(expLambda); //New: exponential prior for nu
    for (i in 1:Ntotal) {
        y[i] ~ student_t(nu, mu[x[i]], sigma[x[i]]);
    }
}

```

save

```
stanDsoRobust <- stan_model( model_code=modelString )
```

```

## In file included from D:/R-3.4.4/library/BH/include/boost/config.hpp:39:0,
##                  from D:/R-3.4.4/library/BH/include/boost/math/tools/config.hpp:13,
##                  from D:/R-3.4.4/library/StanHeaders/include/stan/math/rev/core/var.hpp:7,
##                  from D:/R-3.4.4/library/StanHeaders/include/stan/math/rev/core/gevv_vvv_vari.hpp:5,
##                  from D:/R-3.4.4/library/StanHeaders/include/stan/math/rev/core.hpp:12,
##                  from D:/R-3.4.4/library/StanHeaders/include/stan/math/rev/mat.hpp:4,

```

```
##          from D:/R-3.4.4/library/StanHeaders/include/stan/math.hpp:4,
##          from D:/R-3.4.4/library/StanHeaders/include/src/stan/model/model_header.hpp:4,
##          from file3a4c33b37610.cpp:8:
## D:/R-3.4.4/library/BH/include/boost/config/compiler/gcc.hpp:186:0: warning: "BOOST_NO_CXX11_RVALUE_REFERENCES"
## redefined
## # define BOOST_NO_CXX11_RVALUE_REFERENCES
## ^
## <command-line>:0:0: note: this is the location of the previous definition
## cclplus.exe: warning: unrecognized command line option "-Wno-ignored-attributes"
```

```
save(stanDsoRobust, file=paste(dataPath, "DSORobust1.Rds", sep="/"))
```

run

```
parameters = c( "mu" , "sigma" , "nu" )      # The parameters to be monitored
adaptSteps = 500                             # Number of steps to "tune" the samplers
burnInSteps = 1000
nChains = 4
thinSteps = 1
numSavedSteps<-5000
# Get MC sample of posterior:

stanFitRobust <- sampling( object=stanDsoRobust ,
                          data = dataList ,
                          pars=c('nu', 'mu', 'sigma'),
                          chains = 3,
                          cores= 3,
                          iter = 50000,
                          warmup = 300,
                          thin = 1 )
```

save

```
save(stanFitRobust, file=paste(dataPath, "StanRobustFit2Groups.Rdata", sep="/"))
# load(paste(dataPath, "StanRobustFit2Groups.Rdata", sep="/"))
```

Print

```
print(stanFitRobust)
```

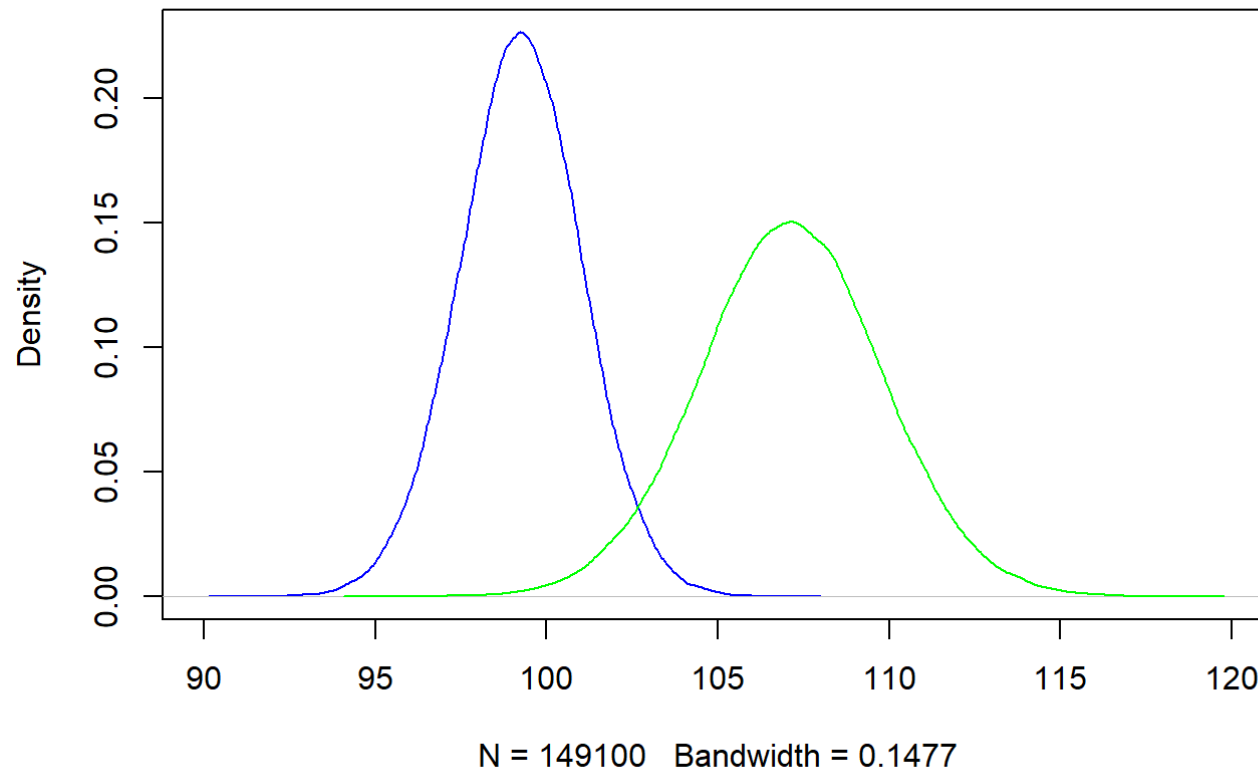
```
## Inference for Stan model: 13367de882eb2ef7e53d40ca8e14e25e.
## 3 chains, each with iter=50000; warmup=300; thin=1;
## post-warmup draws per chain=49700, total post-warmup draws=149100.
##
##               mean se_mean   sd    2.5%    25%    50%    75%    97.5%
## nu           3.89    0.01 1.70     1.94     2.82     3.51     4.49     8.03
## mu[1]        99.25    0.00 1.79    95.70    98.07    99.26   100.45   102.77
## mu[2]       107.15    0.01 2.69   101.86   105.36   107.14   108.93   112.45
## sigma[1]     11.34    0.01 1.74     8.30    10.13    11.21    12.41    15.13
## sigma[2]     17.96    0.01 2.73    13.05    16.05    17.82    19.71    23.73
## lp__        -451.36    0.01 1.64   -455.40  -452.19  -451.02  -450.15  -449.20
##               n_eff Rhat
## nu          102308    1
## mu[1]       149100    1
## mu[2]       149100    1
## sigma[1]    117279    1
## sigma[2]    134181    1
## lp__         66907    1
##
## Samples were drawn using NUTS(diag_e) at Tue May 15 18:29:58 2018.
## 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).
```

Matches with workshop.

```
dis1<-cbind(Mu=rstan::extract(stanFitRobust,pars="mu[1]")$'mu[1]',
            Sigma=rstan::extract(stanFitRobust,pars="sigma[1]")$'sigma[1]')
dis2<-cbind(Mu=rstan::extract(stanFitRobust,pars="mu[2]")$'mu[2]',
            Sigma=rstan::extract(stanFitRobust,pars="sigma[2]")$'sigma[2]')
denDis1<-density(dis1[, "Mu"])
denDis2<-density(dis2[, "Mu"])
```

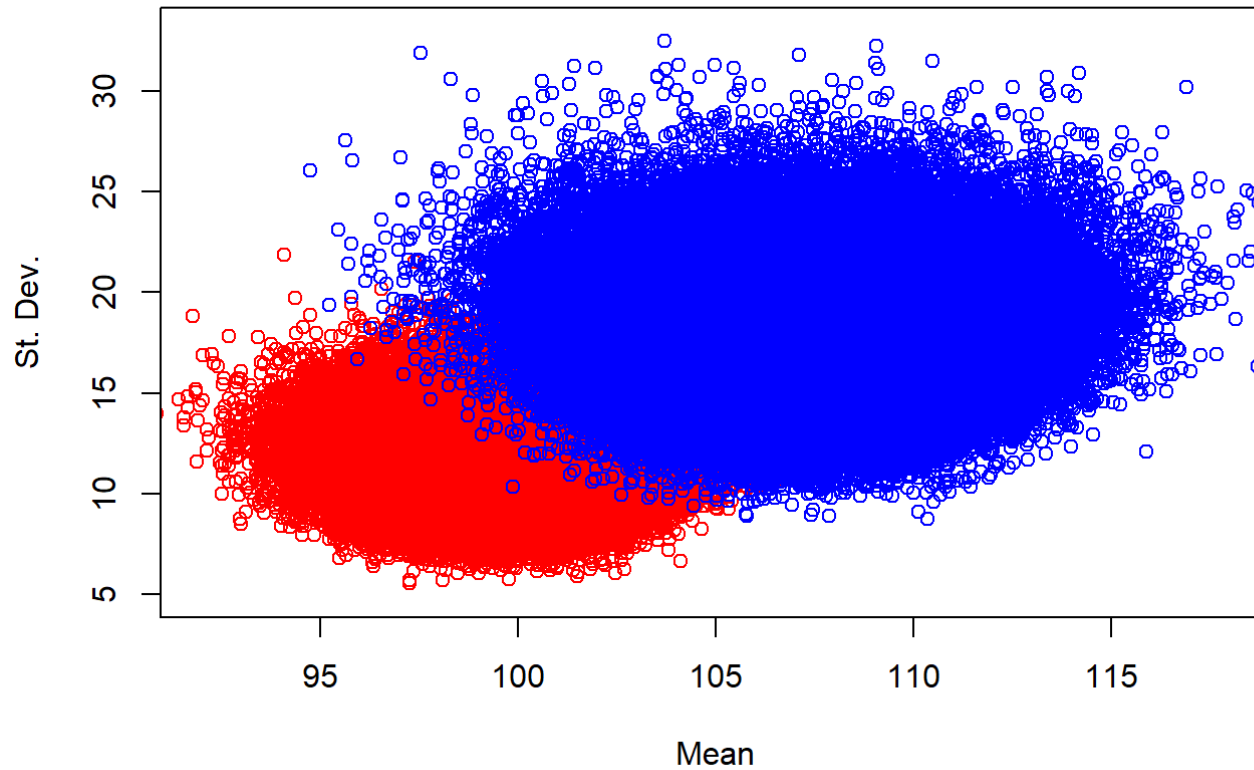
```
plot(denDis1,col="blue",xlim=c(90,120))  
lines(denDis2,col="green")
```

density.default(x = dis1[, "Mu"])



Plots

```
plot(dis1,xlim=c(92,118),ylim=c(5,33),col="red",xlab="Mean",ylab="St. Dev.")  
points(dis2,col="blue")
```



##Two different tests.

Ttest to compare means with different group vars. Reject null here.

```
t.test(dis1[,1],dis2[,1], var.equal=F, paired=FALSE)
```

```
##  
##  Welch Two Sample t-test  
##
```

```
## data:  dis1[, 1] and dis2[, 1]
## t = -944.62, df = 259770, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  -7.913784 -7.881012
## sample estimates:
## mean of x mean of y
##  99.25166 107.14906
```

Test two: let's run a regression to see if the samples have significant impact on mu values.

```
test_dta = data.frame(mu = c(dis1[,1],dis2[,1]),
                      group = c(rep(1,length(dis1[,1])),rep(2,length(dis2[,1]))))
)

head(test_dta)
```

```
##      mu group
## 1 101.77434    1
## 2  96.56251    1
## 3 100.75549    1
## 4  97.60919    1
## 5  95.62384    1
## 6  98.43982    1
```

GLM

```
glm1 = glm(mu ~ as.factor(group), data = test_dta)
summary(glm1)
```

```
##
## Call:
## glm(formula = mu ~ as.factor(group), data = test_dta)
##
```



```
## Deviance Residuals:
##      Min        1Q      Median        3Q        Max
## -12.4110   -1.4432    0.0012    1.4422   11.9441
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    99.251659   0.005912 16789.1  <2e-16 ***
## as.factor(group)2  7.897398   0.008360   944.6  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 5.210728)
##
##      Null deviance: 6203430  on 298199  degrees of freedom
## Residual deviance: 1553829  on 298198  degrees of freedom
## AIC: 1338504
##
## Number of Fisher Scoring iterations: 2
```

Significant p values! Can see as well

Part 2

2. Analyze convergence of MCMC in section 5.1.4 of part 2 of the workshop, try to adjust parameters and rerun the process to obtain the a better quality of MCMC.

```
df = read.csv(file=paste(dataPath, "wk7/HierLinRegressData.csv", sep="/"))
head(df)
```

```
##   Subj    X    Y
## 1    1 60.2 145.6
## 2    1 61.5 157.3
## 3    1 61.7 165.6
## 4    1 62.3 158.8
```

```
## 5    1 67.6 196.1
## 6    1 69.2 183.9
```

```
dataList<-list(Ntotal=length(df$Y),
              y=df$Y,
              x=df$X,
              Nggroups = max(df$Subj),
              group = df$Subj
            )
```

Model desc

```
modelStringPanel = "
data {
  int<lower=1> Ntotal;
  vector[Ntotal] y;
  vector[Ntotal] x;
  int<lower=1> Nggroups;
  int<lower=1, upper=Nggroups> group[Ntotal];
}
transformed data {
  real meanY;
  real sdY;
  vector[Ntotal] zy; // normalized y
  real meanX;
  real sdX;
  vector[Ntotal] zx; // normalized x
  meanY = mean(y);
  sdY = sd(y);
  zy = (y - meanY) / sdY;
  meanX = mean(x);
  sdX = sd(x);
  zx = (x - meanX) / sdX;
}
parameters {
  real<lower=0> zsigma;
```

```

    real<lower=0> nu;
    real zbeta0mu;
    real zbetalmu;
    real<lower=0> zbeta0sigma;
    real<lower=0> zbetalsigma;
    vector[Ngroups] zbeta0;
    vector[Ngroups] zbeta1;
}
transformed parameters {
    real<lower=0> sigma;
    real beta0mu;
    real betalmu;
    vector[Ngroups] beta0;
    vector[Ngroups] beta1;
    // Transform to original scale:
    sigma = zsigma * sdY;
    beta0mu = meanY + zbeta0mu * sdY - zbetalmu * meanX * sdY / sdX;
    betalmu = zbetalmu * sdY / sdX;
    beta0 = meanY + zbeta0 * sdY - zbeta1 * meanX * sdY / sdX; // vectorized
    beta1 = zbeta1 * sdY / sdX; // vectorized
}
model {
    zsigma ~ uniform(0.001, 1000);
    nu ~ exponential(1/30.0);
    zbeta0mu ~ normal(0, 10.0^2);
    zbetalmu ~ normal(0, 10.0^2);
    zbeta0sigma ~ uniform(0.001, 1000);
    zbetalsigma ~ uniform(0.001, 1000);
    zbeta0 ~ normal(zbeta0mu, zbeta0sigma); // vectorized
    zbeta1 ~ normal(zbetalmu, zbetalsigma); // vectorized
    for (i in 1:Ntotal) {
        zy[i] ~ student_t(1+nu, zbeta0[group[i]] + zbeta1[group[i]] * x[i], zsigma);
    }
}
}"

```

Run

```
stanDsoRobustRegPanel <- stan_model( model_code=modelStringPanel )
```

```
## In file included from D:/R-3.4.4/library/BH/include/boost/config.hpp:39:0,  
##                  from D:/R-3.4.4/library/BH/include/boost/math/tools/config.hpp:13,  
##                  from D:/R-3.4.4/library/StanHeaders/include/stan/math/rev/core/var.hpp:7,  
##                  from D:/R-3.4.4/library/StanHeaders/include/stan/math/rev/core/gevv_vvv_vari.hpp:5,  
##                  from D:/R-3.4.4/library/StanHeaders/include/stan/math/rev/core.hpp:12,  
##                  from D:/R-3.4.4/library/StanHeaders/include/stan/math/rev/mat.hpp:4,  
##                  from D:/R-3.4.4/library/StanHeaders/include/stan/math.hpp:4,  
##                  from D:/R-3.4.4/library/StanHeaders/include/src/stan/model/model_header.hpp:4,  
##                  from file3a4c30c831ee.cpp:8:  
## D:/R-3.4.4/library/BH/include/boost/config/compiler/gcc.hpp:186:0: warning: "BOOST_NO_CXX11_RVALUE_REFERENCES"  
redefined  
## # define BOOST_NO_CXX11_RVALUE_REFERENCES  
## ^  
## <command-line>:0:0: note: this is the location of the previous definition  
## cclplus.exe: warning: unrecognized command line option "-Wno-ignored-attributes"
```

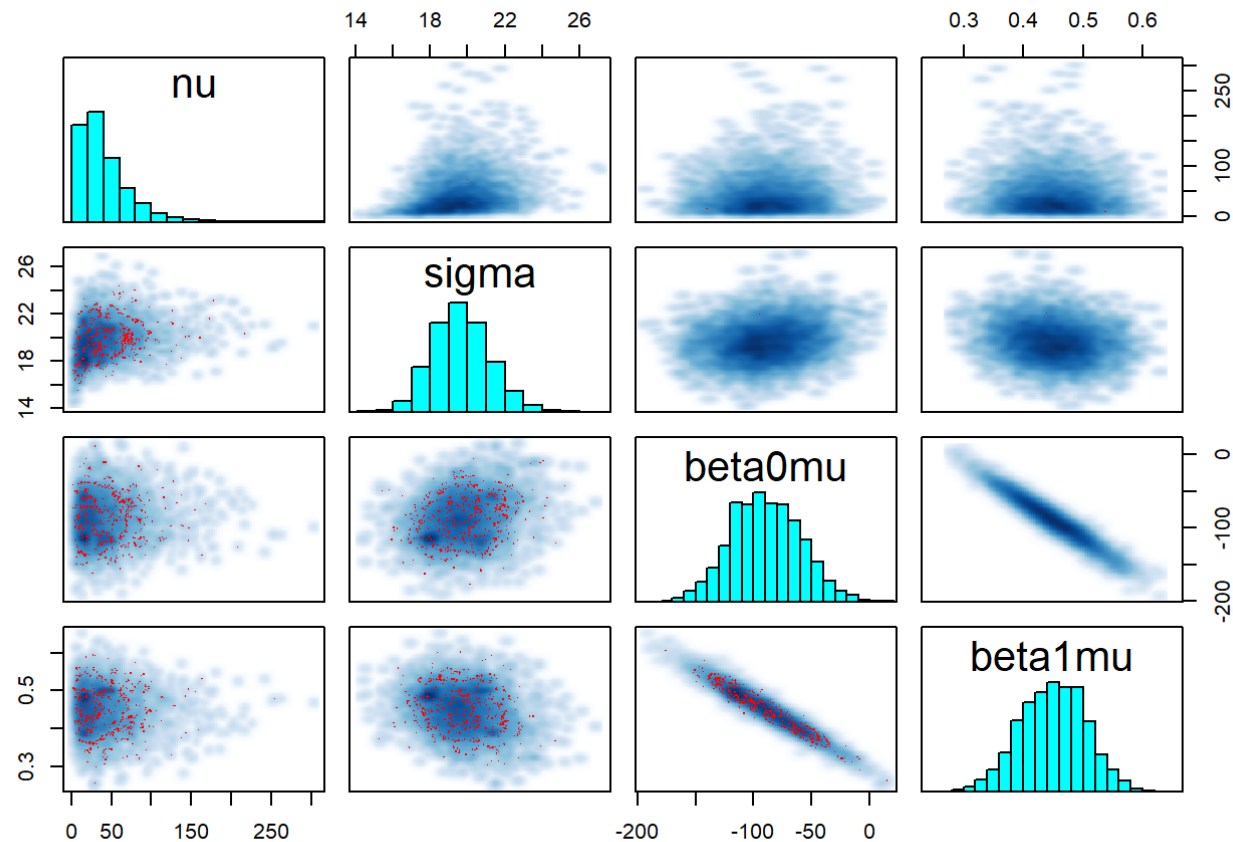
```
fit <- sampling (stanDsoRobustRegPanel,  
  data=dataList,  
  pars=c('nu', 'sigma', 'beta0mu', 'beta1mu', 'beta0', 'beta1',  
        'zbeta0sigma', 'zbeta1sigma'),  
  iter=5000,  
  chains = 4, cores = 4  
)
```

```
## Warning: There were 625 divergent transitions after warmup. Increasing adapt_delta above 0.8 may help. See  
## http://mc-stan.org/misc/warnings.html#divergent-transitions-after-warmup
```

```
## Warning: There were 3 chains where the estimated Bayesian Fraction of Missing Information was low. See  
## http://mc-stan.org/misc/warnings.html#bfmi-low
```

```
## Warning: Examine the pairs() plot to diagnose sampling problems
```

```
# pairs
pairs(fit, pars=c('nu', 'sigma', 'beta0mu', 'beta1mu'))
```



Pairs plot shows high cor between betas.

improving convergence

Increase adaptive delta.

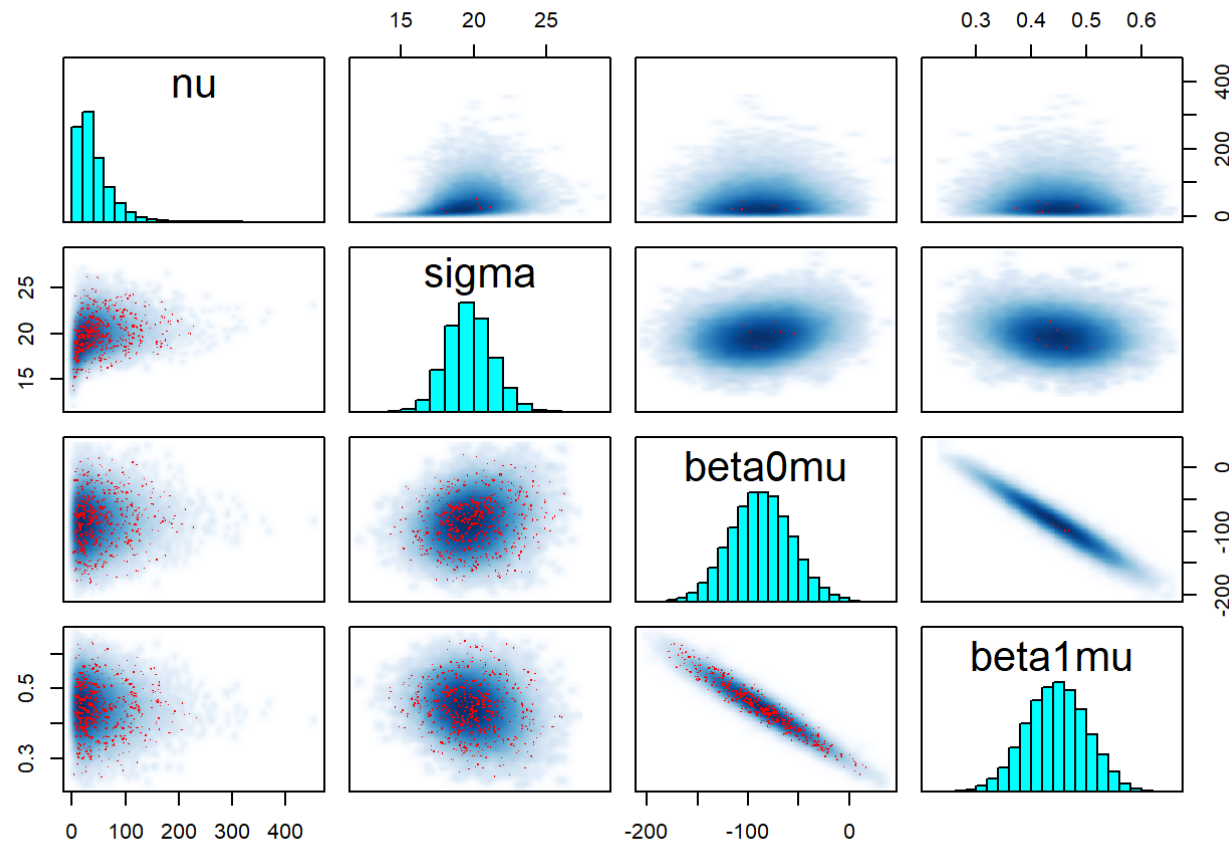
```
fit <- sampling (stanDsoRobustRegPanel,  
               data=dataList,  
               pars=c('nu', 'sigma', 'beta0mu', 'beta1mu', 'beta0', 'beta1',  
                     'zbeta0sigma', 'zbeta1sigma'),  
               iter=50000,  
               chains = 4, cores = 4,  
               control = list(adapt_delta = 0.999, stepsize = 0.01, max_treedepth = 15)  
               )
```

```
## Warning: There were 1524 divergent transitions after warmup. Increasing adapt_delta above 0.999 may help. See  
## http://mc-stan.org/misc/warnings.html#divergent-transitions-after-warmup
```

```
## Warning: There were 4 chains where the estimated Bayesian Fraction of Missing Information was low. See  
## http://mc-stan.org/misc/warnings.html#bfmi-low
```

```
## Warning: Examine the pairs() plot to diagnose sampling problems
```

```
pairs(fit, pars=c('nu', 'sigma', 'beta0mu', 'beta1mu'))
```



Still a lot of divergent points with a small step size and lhigh adapt_delta.

```
save(fit, file=paste(dataPath, "fitPanelModifControl11152016.Rdata", sep="/"))
```

Part 3

3. Consider data state.x77 from datasets used in multiple regression example in Statistical Analysis (MScA 31007).

Using LifeExp as response fit Gaussian and robust non-hierarchical regression models using Bayesian approach.

```
dta<-as.data.frame(state.x77)
colnames(dta)[4]<- "LifeExp"
colnames(dta)[6]<- "HSGrad"

head(dta)
```

```
##           Population Income Illiteracy LifeExp Murder HSGrad Frost   Area
## Alabama           3615   3624         2.1   69.05   15.1   41.3    20  50708
## Alaska             365   6315         1.5   69.31   11.3   66.7   152 566432
## Arizona           2212   4530         1.8   70.55    7.8   58.1    15 113417
## Arkansas          2110   3378         1.9   70.66   10.1   39.9    65  51945
## California        21198   5114         1.1   71.71   10.3   62.6    20 156361
## Colorado          2541   4884         0.7   72.06    6.8   63.9   166 103766
```

```
dim(dta)
```

```
## [1] 50  8
```

Let's do a simple regression.

```
## Create Stan data
dat <- list(N      = nrow(dta),
            p      = 8,
            Population = dta$Population,
            Income    = dta$Income,
            Illiteracy = dta$Illiteracy,
            Murder    = dta$Murder,
            HSGrad    = dta$HSGrad,
            Frost     = dta$Frost,
            Area      = dta$Area,
            LifeExp   = dta$LifeExp

            )
```


Model text.

```
modelStringPanel = "  
  
data {  
  // Define variables in data  
  // Number of observations (an integer)  
  int<lower=0> N;  
  // Number of parameters  
  int<lower=0> p;  
  // Variables  
  real LifeExp[N];  
  real Population[N];  
  real Income[N];  
  real Illiteracy[N];  
  real Murder[N];  
  real HSGrad[N];  
  real Frost[N];  
  real Area[N];  
  
}  
  
parameters {  
  // Define parameters to estimate  
  real beta[p];  
  
  // standard deviation (a positive real number)  
  real<lower=0> sigma;  
}  
  
transformed parameters {  
  // Mean  
  real mu[N];  
  for (i in 1:N) {  
    mu[i] <- beta[1] + beta[2]*Population[i] + beta[3]*Income[i] + beta[4]*Illiteracy[i] + beta[5]*Murder[i] + b  
eta[6]*HSGrad[i] + beta[7]*Frost[i] + beta[8]*Area[i];  
  }  
}
```

```

}

model {
  // Prior part of Bayesian inference (flat if unspecified)

  // Likelihood part of Bayesian inference
  LifeExp ~ normal(mu, sigma);
}

"

```

```

## Run Stan
resStan = stan(model_code = modelStringPanel, data = dat,
               chains = 3, iter = 3000, warmup = 500, thin = 10)

```

```

## In file included from D:/R-3.4.4/library/BH/include/boost/config.hpp:39:0,
##             from D:/R-3.4.4/library/BH/include/boost/math/tools/config.hpp:13,
##             from D:/R-3.4.4/library/StanHeaders/include/stan/math/rev/core/var.hpp:7,
##             from D:/R-3.4.4/library/StanHeaders/include/stan/math/rev/core/gev_vvv_vari.hpp:5,
##             from D:/R-3.4.4/library/StanHeaders/include/stan/math/rev/core.hpp:12,
##             from D:/R-3.4.4/library/StanHeaders/include/stan/math/rev/mat.hpp:4,
##             from D:/R-3.4.4/library/StanHeaders/include/stan/math.hpp:4,
##             from D:/R-3.4.4/library/StanHeaders/include/src/stan/model/model_header.hpp:4,
##             from file3a4c51cb34cc.cpp:8:
## D:/R-3.4.4/library/BH/include/boost/config/compiler/gcc.hpp:186:0: warning: "BOOST_NO_CXX11_RVALUE_REFERENCES"
redefined
## # define BOOST_NO_CXX11_RVALUE_REFERENCES
## ^
## <command-line>:0:0: note: this is the location of the previous definition
## cclplus.exe: warning: unrecognized command line option "-Wno-ignored-attributes"
##
## SAMPLING FOR MODEL '1c25c7f835d87f1e691e83fba3abf98c' NOW (CHAIN 1).
##
## Gradient evaluation took 0 seconds
## 1000 transitions using 10 leapfrog steps per transition would take 0 seconds.
## Adjust your expectations accordingly!

```

```
##
##
## Iteration:    1 / 3000 [  0%] (Warmup)
## Iteration:   300 / 3000 [ 10%] (Warmup)
## Iteration:   501 / 3000 [ 16%] (Sampling)
## Iteration:   800 / 3000 [ 26%] (Sampling)
## Iteration:  1100 / 3000 [ 36%] (Sampling)
## Iteration:  1400 / 3000 [ 46%] (Sampling)
## Iteration:  1700 / 3000 [ 56%] (Sampling)
## Iteration:  2000 / 3000 [ 66%] (Sampling)
## Iteration:  2300 / 3000 [ 76%] (Sampling)
## Iteration:  2600 / 3000 [ 86%] (Sampling)
## Iteration:  2900 / 3000 [ 96%] (Sampling)
## Iteration:  3000 / 3000 [100%] (Sampling)
##
## Elapsed Time: 3.185 seconds (Warm-up)
##                7.48 seconds (Sampling)
##                10.665 seconds (Total)
##
##
## SAMPLING FOR MODEL '1c25c7f835d87f1e691e83fba3abf98c' NOW (CHAIN 2).
##
## Gradient evaluation took 0 seconds
## 1000 transitions using 10 leapfrog steps per transition would take 0 seconds.
## Adjust your expectations accordingly!
##
##
## Iteration:    1 / 3000 [  0%] (Warmup)
## Iteration:   300 / 3000 [ 10%] (Warmup)
## Iteration:   501 / 3000 [ 16%] (Sampling)
## Iteration:   800 / 3000 [ 26%] (Sampling)
## Iteration:  1100 / 3000 [ 36%] (Sampling)
## Iteration:  1400 / 3000 [ 46%] (Sampling)
## Iteration:  1700 / 3000 [ 56%] (Sampling)
## Iteration:  2000 / 3000 [ 66%] (Sampling)
## Iteration:  2300 / 3000 [ 76%] (Sampling)
## Iteration:  2600 / 3000 [ 86%] (Sampling)
```

```
## Iteration: 2900 / 3000 [ 96%] (Sampling)
## Iteration: 3000 / 3000 [100%] (Sampling)
##
## Elapsed Time: 3.46 seconds (Warm-up)
##               15.217 seconds (Sampling)
##               18.677 seconds (Total)
##
##
## SAMPLING FOR MODEL '1c25c7f835d87f1e691e83fba3abf98c' NOW (CHAIN 3).
##
## Gradient evaluation took 0 seconds
## 1000 transitions using 10 leapfrog steps per transition would take 0 seconds.
## Adjust your expectations accordingly!
##
##
## Iteration:    1 / 3000 [ 0%] (Warmup)
## Iteration:   300 / 3000 [ 10%] (Warmup)
## Iteration:   501 / 3000 [ 16%] (Sampling)
## Iteration:   800 / 3000 [ 26%] (Sampling)
## Iteration:  1100 / 3000 [ 36%] (Sampling)
## Iteration:  1400 / 3000 [ 46%] (Sampling)
## Iteration:  1700 / 3000 [ 56%] (Sampling)
## Iteration:  2000 / 3000 [ 66%] (Sampling)
## Iteration:  2300 / 3000 [ 76%] (Sampling)
## Iteration:  2600 / 3000 [ 86%] (Sampling)
## Iteration:  2900 / 3000 [ 96%] (Sampling)
## Iteration:  3000 / 3000 [100%] (Sampling)
##
## Elapsed Time: 2.341 seconds (Warm-up)
##               25.76 seconds (Sampling)
##               28.101 seconds (Total)
```

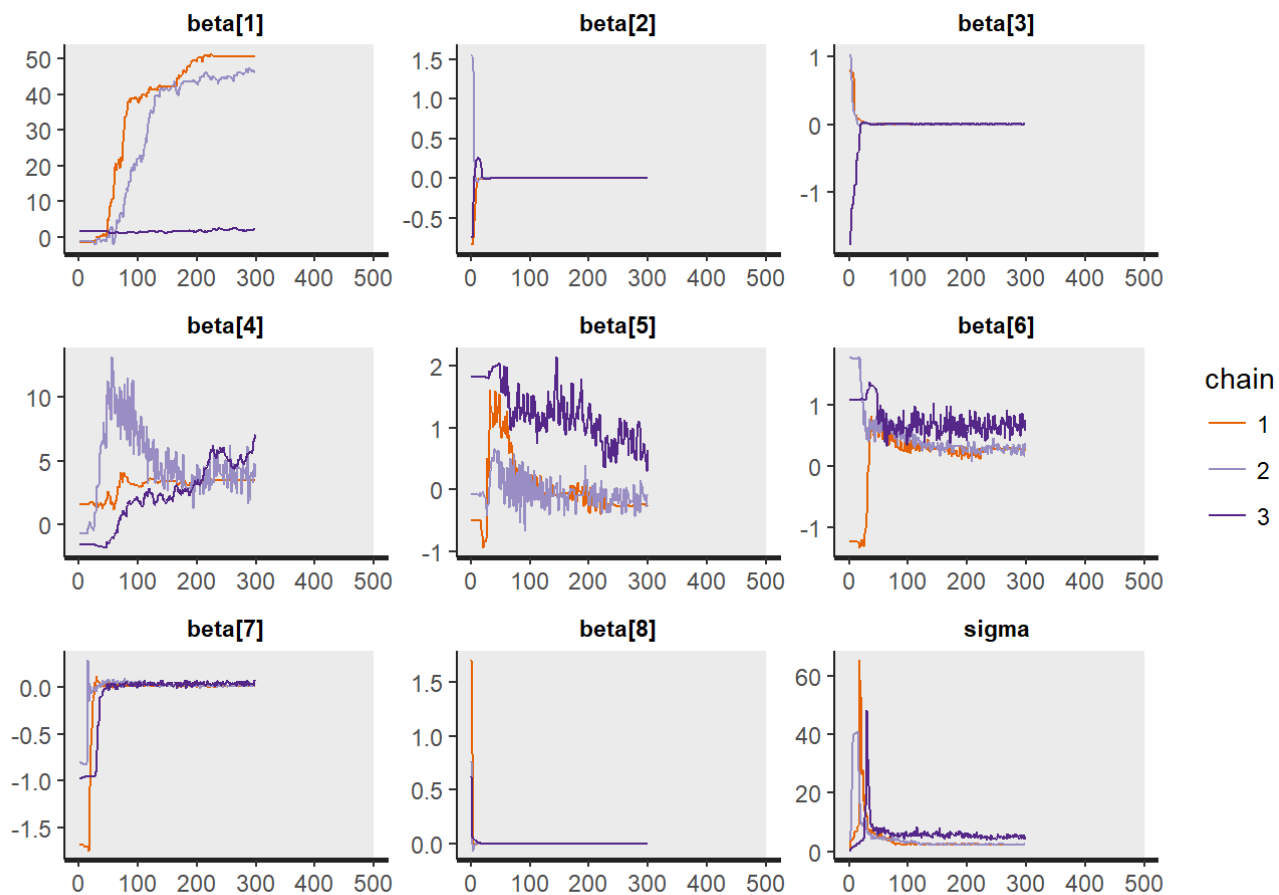
```
## Warning: There were 295 divergent transitions after warmup. Increasing adapt_delta above 0.8 may help. See
## http://mc-stan.org/misc/warnings.html#divergent-transitions-after-warmup
```

```
## Warning: There were 340 transitions after warmup that exceeded the maximum treedepth. Increase max_treedepth above 10. See  
## http://mc-stan.org/misc/warnings.html#maximum-treedepth-exceeded
```

```
## Warning: There were 2 chains where the estimated Bayesian Fraction of Missing Information was low. See  
## http://mc-stan.org/misc/warnings.html#bfmi-low
```

```
## Warning: Examine the pairs() plot to diagnose sampling problems
```

```
## Show traceplot  
traceplot(resStan, pars = c("beta", "sigma"), inc_warmup = TRUE)
```



```
#
print(resStan)
```

```
## Inference for Stan model: 1c25c7f835d87f1e691e83fba3abf98c.
## 3 chains, each with iter=3000; warmup=500; thin=10;
## post-warmup draws per chain=250, total post-warmup draws=750.
##
##               mean se_mean   sd    2.5%    25%    50%    75%   97.5% n_eff
## beta[1]   26.65   14.45 20.60    1.17    1.91   38.75   44.86   50.50     2
```

## beta[2]	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	5
## beta[3]	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.01	3
## beta[4]	3.87	0.77	2.03	0.48	2.95	3.48	4.68	9.67	7
## beta[5]	0.33	0.43	0.61	-0.36	-0.14	0.05	0.90	1.59	2
## beta[6]	0.46	0.13	0.20	0.19	0.29	0.37	0.63	0.88	2
## beta[7]	0.03	0.01	0.02	0.00	0.02	0.03	0.04	0.07	8
## beta[8]	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	2
## sigma	3.72	1.11	1.53	2.37	2.44	2.69	5.10	6.83	2
## mu[1]	67.42	0.19	1.63	64.14	66.55	67.52	68.27	70.65	75
## mu[2]	72.33	1.05	3.89	66.31	69.82	71.73	74.04	81.68	14
## mu[3]	72.02	0.25	1.56	68.28	71.17	72.06	73.03	75.04	40
## mu[4]	64.94	2.07	2.91	59.54	62.09	66.12	67.44	68.56	2
## mu[5]	72.91	0.48	2.57	68.35	71.37	72.65	74.22	78.95	28
## mu[6]	75.61	1.92	2.77	72.24	73.28	74.41	78.19	80.97	2
## mu[7]	75.44	0.61	1.89	72.54	74.08	75.09	76.56	79.79	10
## mu[8]	72.66	1.16	1.84	70.13	71.32	72.11	73.87	76.77	3
## mu[9]	70.84	1.07	2.05	67.73	69.59	70.38	71.58	76.06	4
## mu[10]	68.51	0.11	1.38	65.86	67.77	68.42	69.18	71.65	144
## mu[11]	77.19	1.15	2.82	73.24	75.16	76.42	78.78	83.78	6
## mu[12]	70.11	0.21	1.20	67.23	69.55	70.22	70.79	72.26	32
## mu[13]	73.25	1.86	2.77	69.75	71.11	72.33	75.25	79.23	2
## mu[14]	70.17	0.37	1.00	68.35	69.59	70.07	70.66	72.47	7
## mu[15]	70.98	0.64	1.20	67.92	70.36	71.25	71.81	72.69	4
## mu[16]	71.16	0.17	0.91	69.23	70.66	71.12	71.62	73.02	29
## mu[17]	65.39	1.32	2.08	61.23	63.61	66.09	67.02	68.25	2
## mu[18]	69.63	1.04	2.10	64.55	68.63	70.02	70.94	73.33	4
## mu[19]	68.73	1.72	2.55	63.19	66.81	69.69	70.66	71.90	2
## mu[20]	73.44	2.11	3.10	69.62	71.02	72.37	75.73	80.31	2
## mu[21]	74.00	0.37	1.40	71.30	73.27	73.96	74.66	77.24	14
## mu[22]	72.57	1.91	2.79	69.15	70.40	71.50	74.69	78.70	2
## mu[23]	70.80	0.99	1.64	66.71	69.84	71.34	71.94	72.96	3
## mu[24]	67.14	1.56	2.41	61.80	65.49	67.78	69.03	70.44	2
## mu[25]	67.58	0.41	1.33	64.19	67.07	67.81	68.33	69.81	11
## mu[26]	69.60	0.48	1.33	66.36	69.00	69.92	70.45	71.53	8
## mu[27]	70.79	0.65	1.20	67.77	70.14	71.08	71.65	72.47	3
## mu[28]	78.14	4.71	6.52	70.67	72.85	74.88	84.45	90.35	2
## mu[29]	72.76	0.14	1.20	70.33	72.06	72.70	73.31	75.36	73

```

## mu[30] 73.54 0.36 1.53 71.20 72.54 73.30 74.34 77.30 18
## mu[31] 73.23 0.72 2.25 68.60 72.19 73.15 74.17 78.37 10
## mu[32] 73.88 1.01 2.35 70.52 72.18 73.36 75.16 79.68 5
## mu[33] 66.11 1.32 2.04 61.62 64.50 66.75 67.66 68.83 2
## mu[34] 70.02 1.41 2.55 63.31 69.03 70.65 71.65 73.51 3
## mu[35] 71.04 0.06 1.20 68.93 70.29 70.93 71.66 73.51 343
## mu[36] 67.58 1.25 1.84 63.95 65.95 68.27 69.11 70.12 2
## mu[37] 68.68 0.66 1.93 64.32 67.66 69.05 70.08 71.47 9
## mu[38] 69.68 1.18 2.02 64.82 68.55 70.27 71.05 72.48 3
## mu[39] 69.26 1.95 2.97 62.35 67.23 70.38 71.42 73.10 2
## mu[40] 67.30 1.41 2.20 62.03 65.92 68.02 68.89 70.23 2
## mu[41] 67.32 2.01 2.88 61.37 64.76 68.62 69.71 70.78 2
## mu[42] 66.95 0.87 1.48 63.75 65.85 67.38 68.08 69.01 3
## mu[43] 65.91 2.46 3.67 57.85 63.10 67.14 68.73 70.83 2
## mu[44] 73.48 0.25 1.91 70.17 72.32 73.29 74.38 78.28 60
## mu[45] 71.71 0.48 1.47 69.26 70.80 71.47 72.39 75.15 9
## mu[46] 70.82 0.59 1.25 68.98 69.97 70.57 71.41 74.00 5
## mu[47] 71.32 0.40 1.86 67.54 70.21 71.33 72.26 75.52 22
## mu[48] 65.03 2.31 3.19 59.32 61.74 66.49 67.79 68.88 2
## mu[49] 69.83 1.22 1.83 65.63 68.37 70.55 71.19 72.06 2
## mu[50] 74.32 1.77 2.63 71.17 72.22 73.34 76.45 79.75 2
## lp__ -80.16 16.42 23.21 -116.17 -105.24 -67.16 -59.69 -52.83 2
## Rhat
## beta[1] 2.97
## beta[2] 1.22
## beta[3] 1.70
## beta[4] 1.62
## beta[5] 2.67
## beta[6] 1.81
## beta[7] 1.27
## beta[8] 1.96
## sigma 2.55
## mu[1] 1.04
## mu[2] 1.07
## mu[3] 1.06
## mu[4] 2.65
## mu[5] 1.06

```



```
## mu[6]    2.31
## mu[7]    1.26
## mu[8]    1.68
## mu[9]    1.43
## mu[10]   1.02
## mu[11]   1.28
## mu[12]   1.08
## mu[13]   2.02
## mu[14]   1.33
## mu[15]   1.31
## mu[16]   1.09
## mu[17]   1.78
## mu[18]   1.46
## mu[19]   1.96
## mu[20]   2.12
## mu[21]   1.19
## mu[22]   2.22
## mu[23]   1.54
## mu[24]   1.85
## mu[25]   1.34
## mu[26]   1.16
## mu[27]   1.33
## mu[28]   2.91
## mu[29]   1.05
## mu[30]   1.14
## mu[31]   1.31
## mu[32]   1.23
## mu[33]   1.77
## mu[34]   1.40
## mu[35]   1.03
## mu[36]   2.33
## mu[37]   1.26
## mu[38]   1.45
## mu[39]   1.95
## mu[40]   1.80
## mu[41]   2.23
## mu[42]   1.54
```

```
## mu[43] 1.99
## mu[44] 1.06
## mu[45] 1.10
## mu[46] 1.23
## mu[47] 1.11
## mu[48] 2.94
## mu[49] 1.86
## mu[50] 2.00
## lp__ 2.92
##
## Samples were drawn using NUTS(diag_e) at Wed May 16 00:09:45 2018.
## 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).
```

```
resStan@model_pars
```

```
## [1] "beta" "sigma" "mu" "lp__"
```

Checking plot

```
plot(resStan, pars=c('beta', 'sigma', 'mu'))
```

```
## ci_level: 0.8 (80% intervals)
```

```
## outer_level: 0.95 (95% intervals)
```

