

analysis

Setup

```
itr_total = 12000
quantiles=c(0.1, 0.5, 0.9)
dist_eps_list=c("stdN","student")
d=2
burn_in = 2000

simulation1 = data.frame("Distribution_Eps"=character(0), "Method"=character(0),
                        "p"=numeric(),
                        "beta1_MCSE"=numeric(), "beta1_IF"=numeric(), "beta1_mu"=numeric(),
                        "beta1_sd"=numeric(), "beta1_last"=numeric(),
                        "beta2_MCSE"=numeric(), "beta2_IF"=numeric(), "beta2_mu"=numeric(),
                        "beta2_sd"=numeric(),
                        "beta2_last"=numeric(), stringsAsFactors=F)
```

Append data

```
par(mfrow=c(2,3))
setwd('C:\\Users\\juyl\\Dropbox\\Study\\GitHub\\Bayesian-gibbs-sampler\\R')
for (p in quantiles){
  rqfit = rq(foodexp ~ income, data = engel, tau = p)
  for (dist_eps in dist_eps_list){
    load(paste0("Engel\\Gibbs with Scale\\", "GWS_", p, "_", dist_eps, "_betap.RData"))
    beta_p = list()
    for (i in 1:d){
      beta_p[[i]]=unlist(lapply(beta_p_record, function(x){return(x[i,])}))[burn_in+1:itr_total]
      plot(beta_p[[i]], type="l", main=paste("GWS_", p, "_", dist_eps, "_beta_", i), xlab="itr", ylab=paste("beta_", i))
      abline(h=summary(rqfit)$coefficients[i], col="red", lwd=3)
    }
    new_obs = list(dist_eps, "GWS", as.numeric(p),
      as.numeric(mcse(beta_p[[1]])$se), as.numeric(length(beta_p[[1]])/effectiveSize(beta_p[[1]])),
      as.numeric(mean(beta_p[[1]])), as.numeric(sd(beta_p[[1]])),
      as.numeric(beta_p[[1]][length(beta_p[[1]])]),
      as.numeric(mcse(beta_p[[2]])$se), as.numeric(length(beta_p[[2]])/effectiveSize(beta_p[[2]])),
      as.numeric(mean(beta_p[[2]])), as.numeric(sd(beta_p[[2]])),
      as.numeric(beta_p[[2]][length(beta_p[[2]])]))
    simulation1[nrow(simulation1)+1,] = new_obs

#####
    load(paste0("Engel\\Gibbs without Scale\\", "GWOS_", p, "_", dist_eps, "_betap.RData"))
    beta_p = list()
    for (i in 1:d){
      beta_p[[i]]=unlist(lapply(beta_p_record, function(x){return(x[i,])}))[burn_in+1:itr_total]
      plot(beta_p[[i]], type="l", main=paste("GWOS_", p, "_", dist_eps, "_beta_", i), xlab="itr", ylab=paste("beta_", i))
      abline(h=summary(rqfit)$coefficients[i], col="red", lwd=3)
    }
    new_obs = list(dist_eps, "GWOS", as.numeric(p),
      as.numeric(mcse(beta_p[[1]])$se), as.numeric(length(beta_p[[1]])/effectiveSize(beta_p[[1]])),
      as.numeric(mean(beta_p[[1]])), as.numeric(sd(beta_p[[1]])),
      as.numeric(beta_p[[1]][length(beta_p[[1]])]),
      as.numeric(mcse(beta_p[[2]])$se), as.numeric(length(beta_p[[2]])/effectiveSize(beta_p[[2]])),
      as.numeric(mean(beta_p[[2]])), as.numeric(sd(beta_p[[2]])),
      as.numeric(beta_p[[2]][length(beta_p[[2]])]))
    simulation1[nrow(simulation1)+1,] = new_obs

#####
    load(paste0("Engel\\Double Exponential Prior\\", "DEP_", p, "_", dist_eps, "_betap.RData"))
    beta_p = list()
    for (i in 1:d){
      beta_p[[i]]=unlist(lapply(beta_p_record, function(x){return(x[i,])}))[burn_in+1:itr_total]
      plot(beta_p[[i]], type="l", main=paste("DEP_", p, "_", dist_eps, "_beta_", i), xlab="itr", ylab=paste("beta_", i))
      abline(h=summary(rqfit)$coefficients[i], col="red", lwd=3)
    }
    new_obs = list(dist_eps, "DEP", as.numeric(p),
      mcse(beta_p[[1]])$se, length(beta_p[[1]])/effectiveSize(beta_p[[1]]),
      as.numeric(mean(beta_p[[1]])), as.numeric(sd(beta_p[[1]])),
      as.numeric(beta_p[[1]][length(beta_p[[1]])]),
      mcse(beta_p[[2]])$se, length(beta_p[[2]])/effectiveSize(beta_p[[2]]),
      as.numeric(mean(beta_p[[2]])), as.numeric(sd(beta_p[[2]])),
      as.numeric(beta_p[[2]][length(beta_p[[2]])]))
  }
```

```

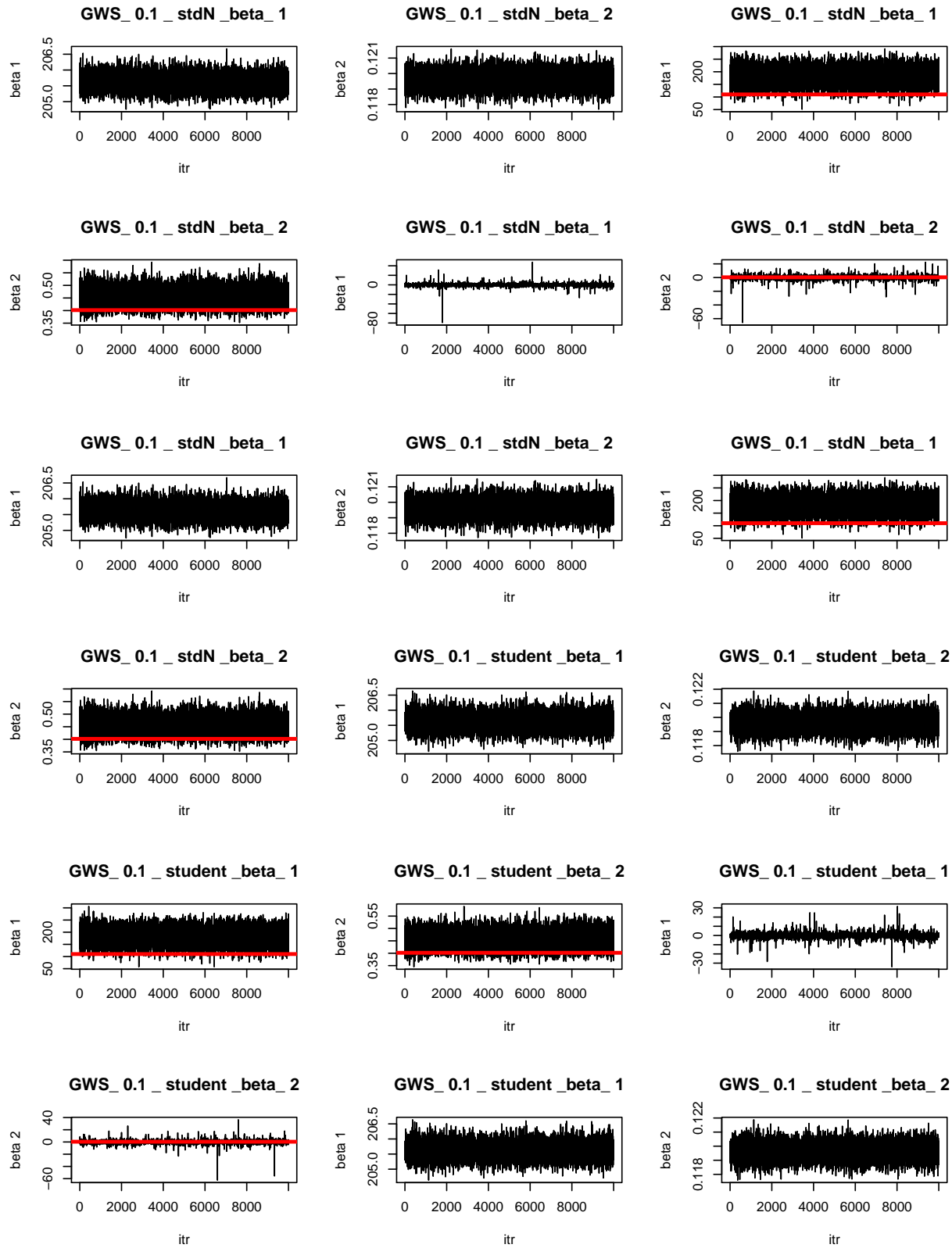
        as.numeric(beta_p[[2]][length(beta_p[[2]])]))
simulation1[nrow(simulation1)+1,] = new_obs

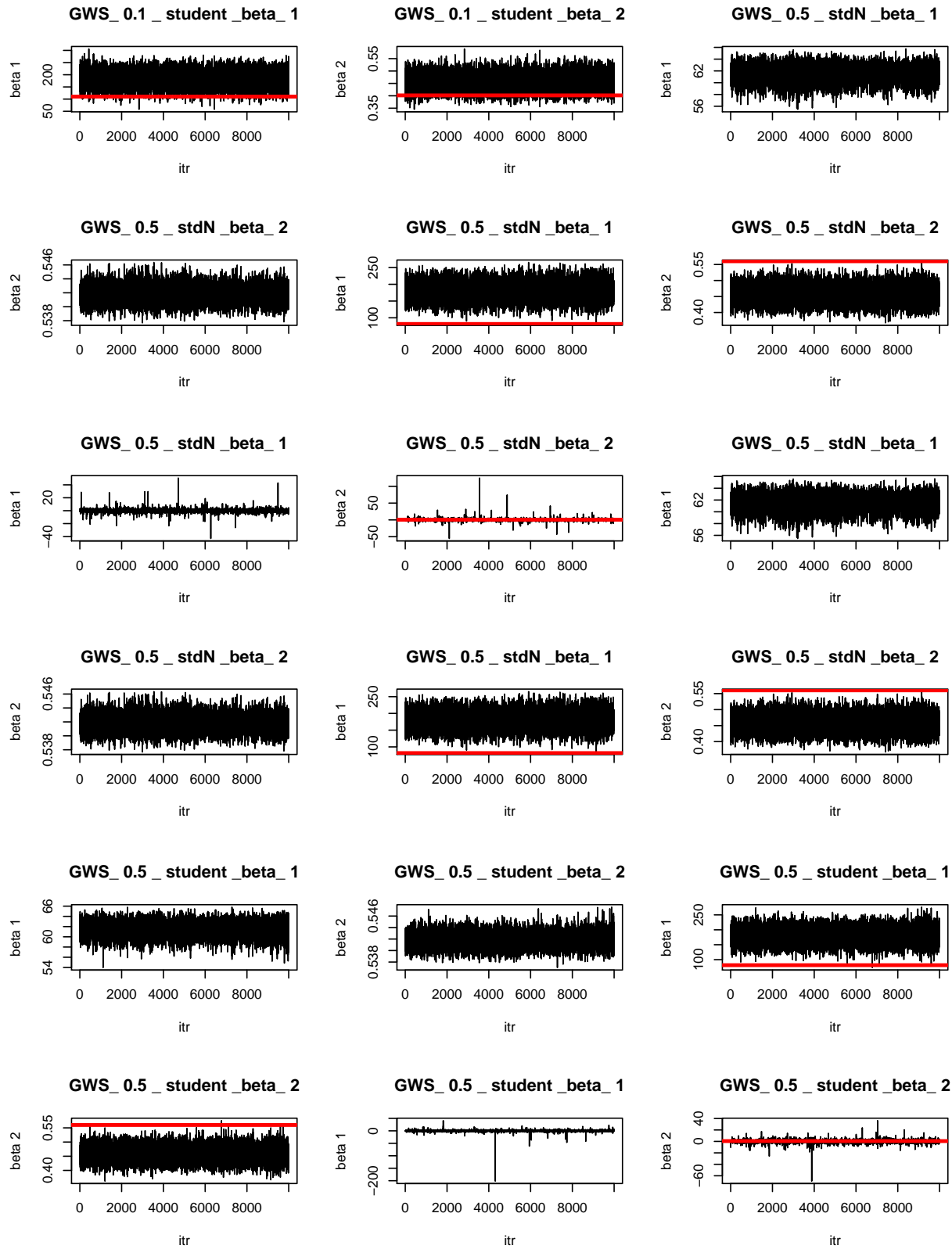
#####
load(paste0("Engel\\Tobit with Scale\\", "TWS_", p, "_", dist_eps, "_betap.RData"))
beta_p = list()
for (i in 1:d){
  beta_p[[i]]=unlist(lapply(beta_p_record, function(x){return(x[i,])}))[ (burn_in+1):itr_total]
  plot(beta_p[[i]], type="l", main=paste("GWS_", p, "_", dist_eps, "_beta_", i), xlab="itr", ylab=paste("beta_", i))
  abline(h=summary(rqfit)$coefficients[i], col="red", lwd=3)
}
new_obs = list(dist_eps, "TWS", as.numeric(p),
               as.numeric(mcse(beta_p[[1]])$se), as.numeric(length(beta_p[[1]])/effectiveSize(beta_p[[1]])),
               as.numeric(mean(beta_p[[1]])), as.numeric(sd(beta_p[[1]])),
               as.numeric(beta_p[[1]][length(beta_p[[1]])]),
               as.numeric(mcse(beta_p[[2]])$se), as.numeric(length(beta_p[[2]])/effectiveSize(beta_p[[2]])),
               as.numeric(mean(beta_p[[2]])), as.numeric(sd(beta_p[[2]])),
               as.numeric(beta_p[[2]][length(beta_p[[2]])]))
simulation1[nrow(simulation1)+1,] = new_obs

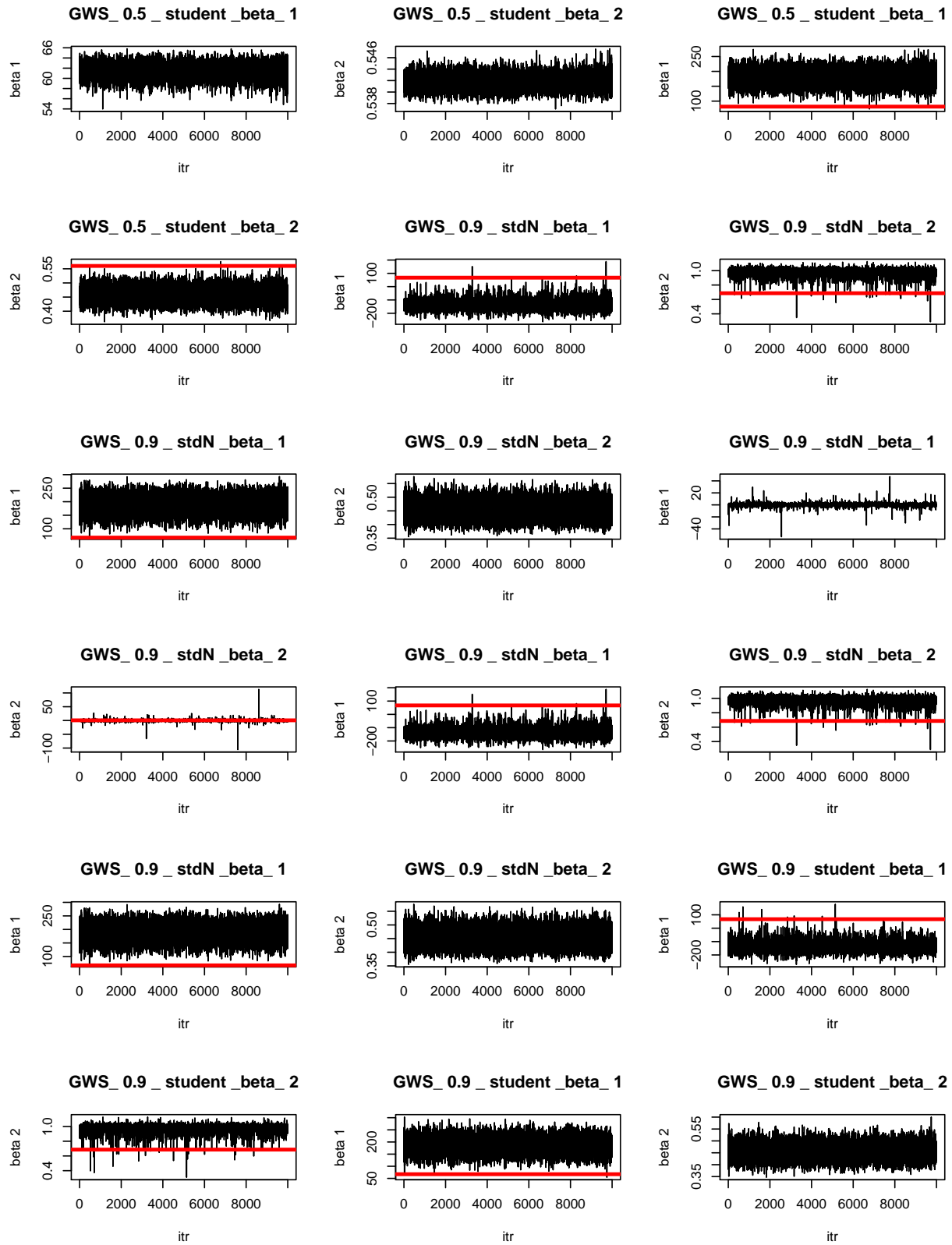
#####
load(paste0("Engel\\Tobit without Scale\\", "TWOS_", p, "_", dist_eps, "_betap.RData"))
beta_p = list()
for (i in 1:d){
  beta_p[[i]]=unlist(lapply(beta_p_record, function(x){return(x[i,])}))[ (burn_in+1):itr_total]
  plot(beta_p[[i]], type="l", main=paste("GWS_", p, "_", dist_eps, "_beta_", i), xlab="itr", ylab=paste("beta_", i))
  abline(h=summary(rqfit)$coefficients[i], col="red", lwd=3)
}
new_obs = list(dist_eps, "TWOS", as.numeric(p),
               as.numeric(mcse(beta_p[[1]])$se), as.numeric(length(beta_p[[1]])/effectiveSize(beta_p[[1]])),
               as.numeric(mean(beta_p[[1]])), as.numeric(sd(beta_p[[1]])),
               as.numeric(beta_p[[1]][length(beta_p[[1]])]),
               as.numeric(mcse(beta_p[[2]])$se), as.numeric(length(beta_p[[2]])/effectiveSize(beta_p[[2]])),
               as.numeric(mean(beta_p[[2]])), as.numeric(sd(beta_p[[2]])),
               as.numeric(beta_p[[2]][length(beta_p[[2]])]))
simulation1[nrow(simulation1)+1,] = new_obs

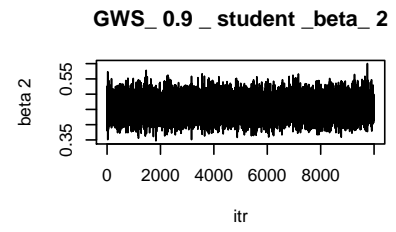
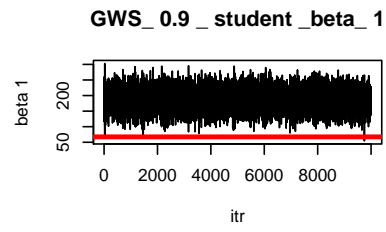
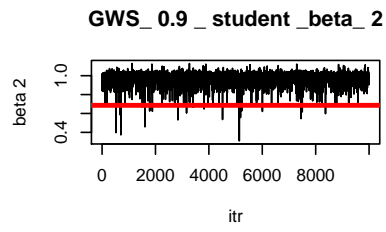
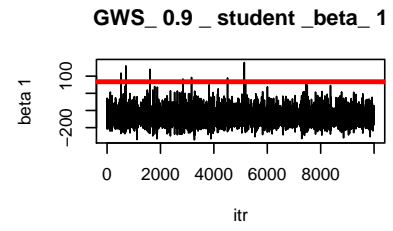
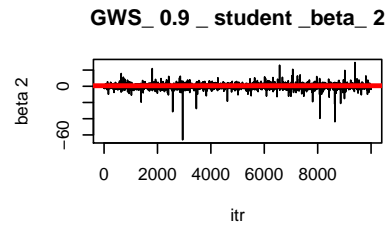
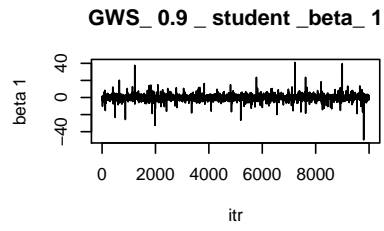
}
}

```









Clean

```
ord = c("stdN", "heteroN", "student")
simulation1$Distribution_Eps = factor(simulation1$Distribution_Eps, levels=ord)
simulation1 = simulation1[order(simulation1$Distribution_Eps),]
is.num <- sapply(simulation1, is.numeric)
simulation1[is.num] <- lapply(simulation1[is.num], round, 4)
simulation1
```

##	Distribution_Eps	Method	p	beta1_MCSE	beta1_IF	beta1_mu	beta1_sd
## 1	stdN	GWS	0.1	0.0024	1.0000	205.6222	0.2436
## 2	stdN	GWOS	0.1	0.1389	0.2230	183.1656	34.3504
## 3	stdN	DEP	0.1	0.0213	1.0000	-0.0078	2.1268
## 4	stdN	TWS	0.1	0.0024	1.0000	205.6222	0.2436
## 5	stdN	TWOS	0.1	0.1389	0.2230	183.1656	34.3504
## 11	stdN	GWS	0.5	0.0069	0.2313	61.5778	1.4369
## 12	stdN	GWOS	0.5	0.1430	0.2834	181.3803	27.0758
## 13	stdN	DEP	0.5	0.0205	1.0000	0.0075	2.0461
## 14	stdN	TWS	0.5	0.0069	0.2313	61.5778	1.4369
## 15	stdN	TWOS	0.5	0.1430	0.2834	181.3803	27.0758
## 21	stdN	GWS	0.9	0.3113	0.7344	-121.2961	35.7309
## 22	stdN	GWOS	0.9	0.1834	0.2138	187.8984	33.8837
## 23	stdN	DEP	0.9	0.0209	1.0000	0.0064	2.0857
## 24	stdN	TWS	0.9	0.3113	0.7344	-121.2961	35.7309
## 25	stdN	TWOS	0.9	0.1834	0.2138	187.8984	33.8837
## 6	student	GWS	0.1	0.0026	1.0727	205.6145	0.2586
## 7	student	GWOS	0.1	0.1554	0.2153	183.4250	33.9458
## 8	student	DEP	0.1	0.0196	1.0000	0.0091	1.9588
## 9	student	TWS	0.1	0.0026	1.0727	205.6145	0.2586
## 10	student	TWOS	0.1	0.1554	0.2153	183.4250	33.9458
## 16	student	GWS	0.5	0.0071	0.2223	61.5517	1.4918
## 17	student	GWOS	0.5	0.1349	0.2824	181.2703	26.9501
## 18	student	DEP	0.5	0.0296	1.0000	-0.0462	2.9588
## 19	student	TWS	0.5	0.0071	0.2223	61.5517	1.4918
## 20	student	TWOS	0.5	0.1349	0.2824	181.2703	26.9501
## 26	student	GWS	0.9	0.3201	0.8315	-119.6984	36.4621
## 27	student	GWOS	0.9	0.1278	0.2273	188.0345	33.5837
## 28	student	DEP	0.9	0.0213	1.0000	-0.0043	2.1340
## 29	student	TWS	0.9	0.3201	0.8315	-119.6984	36.4621
## 30	student	TWOS	0.9	0.1278	0.2273	188.0345	33.5837
##	beta1_last	beta2_MCSE	beta2_IF	beta2_mu	beta2_sd	beta2_last	
## 1	205.4694	0.0000	0.9454	0.1196	0.0005	0.1199	
## 2	184.2547	0.0001	0.2288	0.4547	0.0349	0.4493	
## 3	0.0579	0.0215	1.0765	-0.0046	2.0398	-0.6097	
## 4	205.4694	0.0000	0.9454	0.1196	0.0005	0.1199	
## 5	184.2547	0.0001	0.2288	0.4547	0.0349	0.4493	
## 11	59.6592	0.0000	0.5973	0.5417	0.0012	0.5427	
## 12	149.7276	0.0001	0.2880	0.4567	0.0279	0.4900	
## 13	-2.0986	0.0259	1.0000	0.0380	2.5898	-0.5499	
## 14	59.6592	0.0000	0.5973	0.5417	0.0012	0.5427	
## 15	149.7276	0.0001	0.2880	0.4567	0.0279	0.4900	
## 21	-105.4173	0.0004	1.0073	0.9711	0.0441	0.9383	
## 22	207.5443	0.0002	0.2229	0.4535	0.0344	0.4319	
## 23	-1.9056	0.0268	1.0000	-0.0065	2.6815	4.1610	


```
## 24 -105.4173    0.0004    1.0073    0.9711    0.0441    0.9383
## 25  207.5443    0.0002    0.2229    0.4535    0.0344    0.4319
## 6   205.7859    0.0000    0.9956    0.1196    0.0006    0.1194
## 7   143.9234    0.0002    0.2239    0.4543    0.0343    0.4943
## 8    -0.0860    0.0210    1.0000   -0.0152    2.1021    0.8898
## 9   205.7859    0.0000    0.9956    0.1196    0.0006    0.1194
## 10  143.9234    0.0002    0.2239    0.4543    0.0343    0.4943
## 16   60.4416    0.0000    0.5559    0.5418    0.0013    0.5431
## 17  187.3763    0.0001    0.2894    0.4568    0.0277    0.4454
## 18   -0.5662    0.0197    1.0000   -0.0339    1.9749   -0.0959
## 19   60.4416    0.0000    0.5559    0.5418    0.0013    0.5431
## 20  187.3763    0.0001    0.2894    0.4568    0.0277    0.4454
## 26  -91.4538    0.0005    1.0557    0.9687    0.0456    0.9629
## 27  159.1309    0.0001    0.2349    0.4533    0.0342    0.4819
## 28    0.3276    0.0207    1.0000   -0.0122    2.0749   -1.6529
## 29  -91.4538    0.0005    1.0557    0.9687    0.0456    0.9629
## 30  159.1309    0.0001    0.2349    0.4533    0.0342    0.4819
```

quantreg

```
rqfit <- rq(foodexp ~ income, data = engel, tau = quantiles)
summary(rqfit)
```

```
##
## Call: rq(formula = foodexp ~ income, tau = quantiles, data = engel)
##
## tau: [1] 0.1
##
## Coefficients:
##              coefficients lower bd  upper bd
## (Intercept) 110.14157      79.88753 146.18875
## income       0.40177       0.34210   0.45079
##
## Call: rq(formula = foodexp ~ income, tau = quantiles, data = engel)
##
## tau: [1] 0.5
##
## Coefficients:
##              coefficients lower bd  upper bd
## (Intercept)  81.48225     53.25915 114.01156
## income       0.56018     0.48702   0.60199
##
## Call: rq(formula = foodexp ~ income, tau = quantiles, data = engel)
##
## tau: [1] 0.9
##
## Coefficients:
##              coefficients lower bd  upper bd
## (Intercept)  67.35087     37.11802 103.17399
## income       0.68630     0.64937   0.74223
```