

CWVS: Critical Window Variable Selection

CWVS_Example

[1] Simulate data for analysis:

- Setting the reproducibility seed and initializing packages for data simulation:

```
set.seed(4679)
```

```
library(CWVS)
```

```
library(boot) #Inverse logit transformation
```

- Setting the global data values:

```
n<-5000 #Sample size
```

```
m<-27 #Number of exposure time periods
```

```
x<-matrix(1,
```

```
  nrow=n,
```

```
  ncol=1) #Covariate design matrix
```

```
z<-matrix(rnorm(n=(n*m)),
```

```
  nrow=n,
```

```
  ncol=m) #Exposure design matrix
```

```
for(j in 1:m){
```

```
  z[,j]<-(z[,j] - median(z[,j]))/IQR(z[,j]) #Data standardization (interquartile range)
```

```
}
```

- Setting the values for the statistical model parameters:

```
beta_true<- -0.30
```

```
theta_true<-rep(0.60, times=m)
```

```
gamma_true<-c(rep(0, times=12),
```

```
  rep(1, times=4),
```

```
  rep(0, times=11))
```

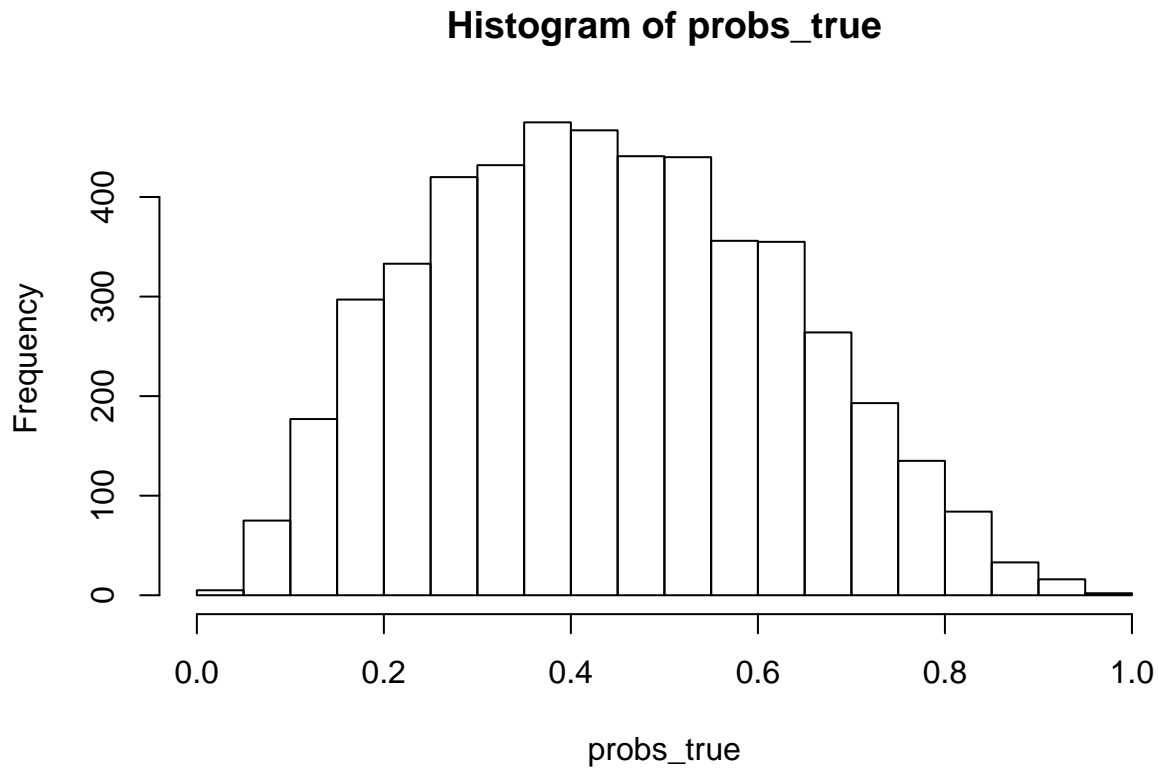
```
alpha_true<-gamma_true*theta_true
```

```
logit_p_true<-x%%beta_true +
```

```
  z%%alpha_true
```

```
probs_true<-inv.logit(logit_p_true)
```

```
hist(probs_true)
```



- Simulating the analysis dataset:

```
y<-rbinom(n=n,
          size=1,
          prob=probs_true)
```

[2] Fit GPCW to estimate critical windows of susceptibility:

```
results<-CWVS(mcmc_samples = 10000,
              y = y, x = x, z = z,
              metrop_var_phi1_trans = 1.00,
              metrop_var_phi2_trans = 1.00,
              metrop_var_A11_trans = 0.05,
              metrop_var_A22_trans = 0.50,
              metrop_var_A21 = 1.00)
```

```
## Progress: 10%
## phi1 Acceptance: 37%
## phi2 Acceptance: 37%
## A11 Acceptance: 17%
## A22 Acceptance: 24%
## A21 Acceptance: 22%
## *****
## Progress: 20%
## phi1 Acceptance: 36%
## phi2 Acceptance: 37%
## A11 Acceptance: 18%
## A22 Acceptance: 25%
```

```

## A21 Acceptance: 20%
## *****
## Progress: 30%
## phi1 Acceptance: 35%
## phi2 Acceptance: 37%
## A11 Acceptance: 18%
## A22 Acceptance: 23%
## A21 Acceptance: 20%
## *****
## Progress: 40%
## phi1 Acceptance: 34%
## phi2 Acceptance: 37%
## A11 Acceptance: 18%
## A22 Acceptance: 25%
## A21 Acceptance: 20%
## *****
## Progress: 50%
## phi1 Acceptance: 35%
## phi2 Acceptance: 37%
## A11 Acceptance: 18%
## A22 Acceptance: 24%
## A21 Acceptance: 21%
## *****
## Progress: 60%
## phi1 Acceptance: 35%
## phi2 Acceptance: 36%
## A11 Acceptance: 19%
## A22 Acceptance: 24%
## A21 Acceptance: 22%
## *****
## Progress: 70%
## phi1 Acceptance: 34%
## phi2 Acceptance: 36%
## A11 Acceptance: 19%
## A22 Acceptance: 25%
## A21 Acceptance: 22%
## *****
## Progress: 80%
## phi1 Acceptance: 34%
## phi2 Acceptance: 36%
## A11 Acceptance: 19%
## A22 Acceptance: 24%
## A21 Acceptance: 21%
## *****
## Progress: 90%
## phi1 Acceptance: 34%
## phi2 Acceptance: 37%
## A11 Acceptance: 19%
## A22 Acceptance: 23%
## A21 Acceptance: 21%
## *****
## Progress: 100%
## phi1 Acceptance: 34%
## phi2 Acceptance: 36%

```

```
## A11 Acceptance: 19%
## A22 Acceptance: 21%
## A21 Acceptance: 21%
## *****
```

[3] Analyzing Output:

```
par(mfrow=c(2,2))
plot(results$beta[1, 1001:10000],
     type="l",
     ylab="beta",
     xlab="Sample")
abline(h=beta_true,
      col="red",
      lwd=2) #True value
plot(rowMeans(results$alpha[,1001:10000]),
     pch=16,
     ylab="alpha",
     xlab="Time")
points(alpha_true,
      pch=16,
      col="red") #True values
plot(rowMeans(results$gamma[,1001:10000]),
     pch=16,
     ylab="gamma",
     xlab="Time")
points(gamma_true,
      pch=16,
      col="red") #True values
```

