trump explore

Explore Points

Clean and save cleaned data

```
# Configure for environment
CLEAN_DATA = TRUE
if(CLEAN_DATA) {
  # Data clean functions
  agree_to_num=function(agree) {
    switch(as.character(agree),
           "Strongly Disagree" = as.numeric(0),
           "Disagree" = as.numeric(1/3),
           "Agree" = as.numeric(2/3),
           "Strongly Agree" = as.numeric(1))
  }
  sex_to_num=function(sex) {
    switch(as.character(sex),
           "Male" = as.numeric(0),
           "Female" = as.numeric(1))
  # Simplification - assumption of average age
  age_to_num=function(age) {
    switch(as.character(age),
           "17 or younger" = as.numeric(15),
           "18-20" = as.numeric(19),
           "21-29" = as.numeric(25),
           "30-39" = as.numeric(35),
           "40-49" = as.numeric(45),
           "50-59" = as.numeric(55),
           "60 or older" = as.numeric(65))
  }
  # resp 37 Cultural Marxism does not exist
  # resp 86 no data
  # Load data
  base_df=read.csv("C:\\dev\\bayes-present\\explore\\poll_data.csv")
  # Clean data
  base_df$Sharia=sapply(base_df$Sharia, agree_to_num, simplify = TRUE)
  base_df$Marx=sapply(base_df$Marx, agree_to_num)
  base_df$SS=sapply(base_df$SS, agree_to_num)
  base_df$Medi=sapply(base_df$Medi, agree_to_num)
  base_df$ChinaWar=sapply(base_df$ChinaWar, agree_to_num)
  base df$ChinaMoney=sapply(base df$ChinaMoney, agree to num)
  base_df$Trad=sapply(base_df$Trad, agree_to_num)
  base_df$Support=sapply(base_df$Support, agree_to_num)
  base_df$Sex=sapply(base_df$Sex, sex_to_num)
  base_df$Age=sapply(base_df$Age, age_to_num)
```

```
# Save data
write.csv(base_df, file="clean_data.csv", row.names=FALSE)
}
```

Load data

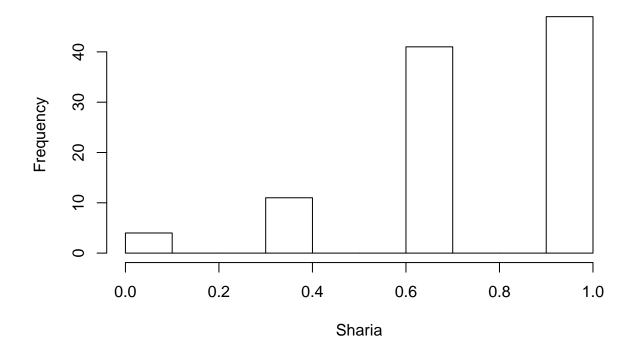
```
# Attach data

df=read.csv("C:\\dev\\bayes-present\\explore\\clean_data.csv")
attach(df)
```

Frequency of responses

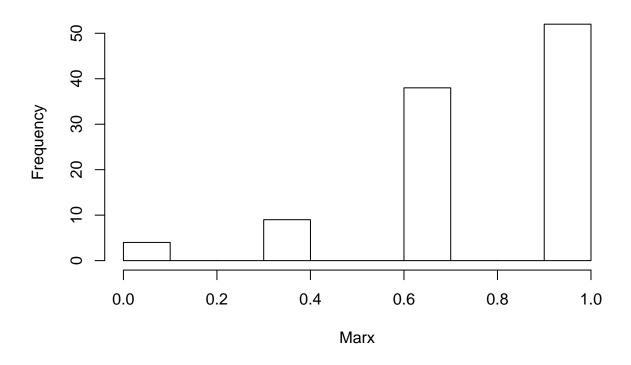
hist(Sharia)

Histogram of Sharia



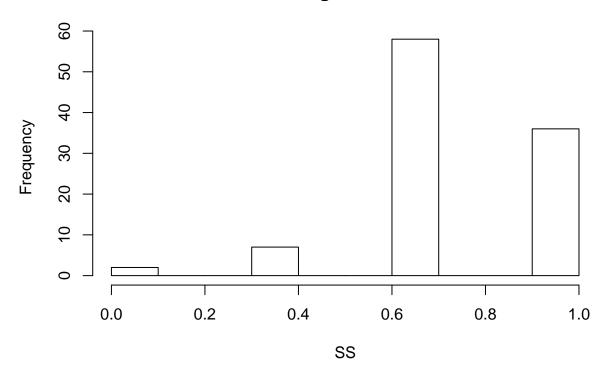
hist(Marx)

Histogram of Marx



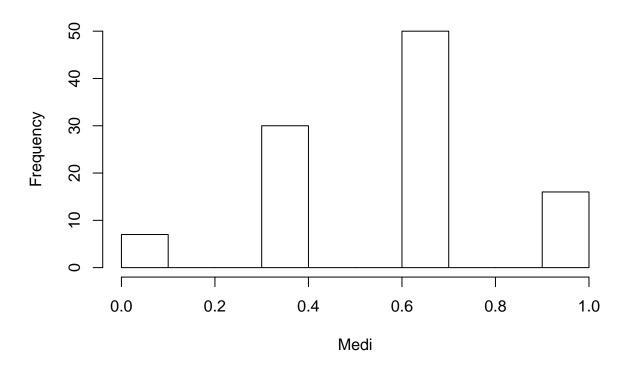
hist(SS)

Histogram of SS



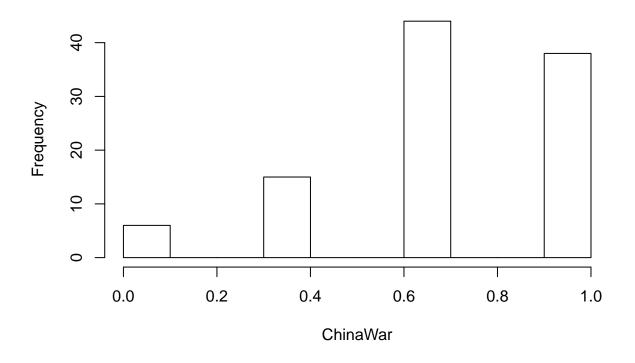
hist(Medi)

Histogram of Medi



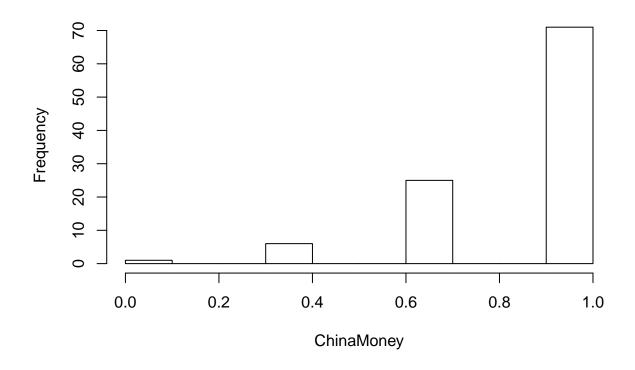
hist(ChinaWar)

Histogram of ChinaWar



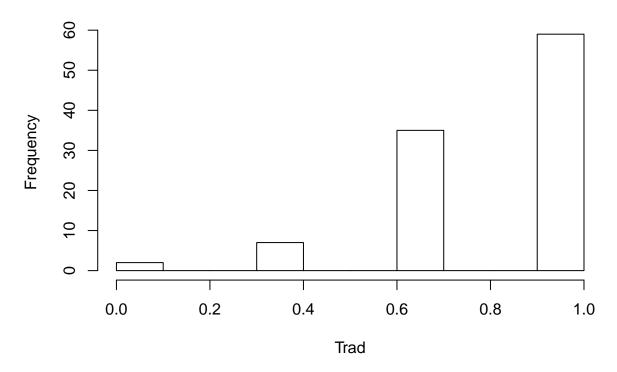
hist(ChinaMoney)

Histogram of ChinaMoney



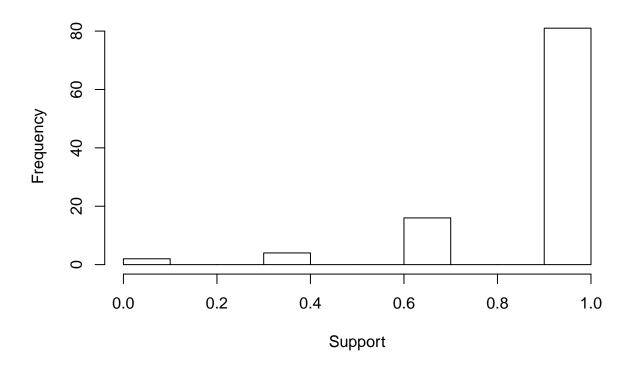
hist(Trad)

Histogram of Trad



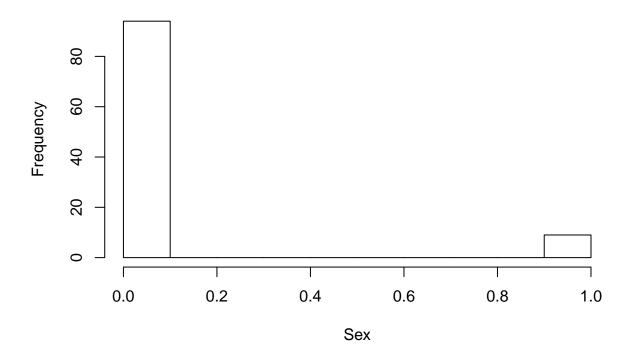
hist(Support)

Histogram of Support



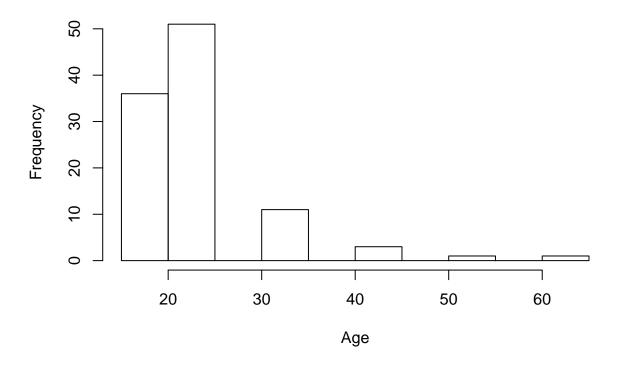
hist(Sex)

Histogram of Sex



hist(Age)

Histogram of Age



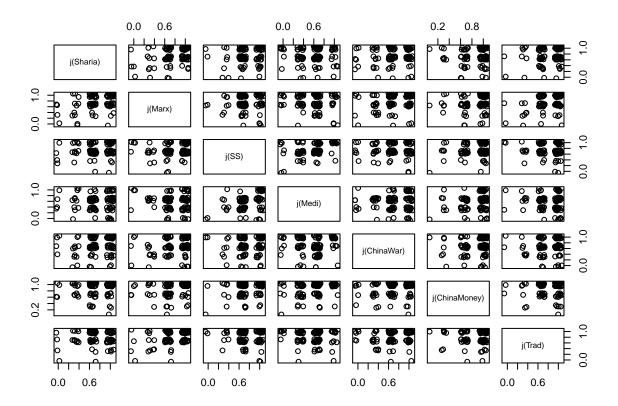
Scatterplot comparison of reponses

```
library(car)

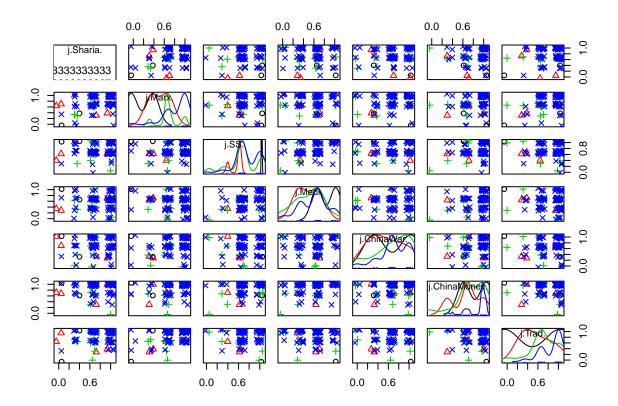
## Warning: package 'car' was built under R version 3.2.5

j=function(s) {
   return(jitter(s, factor = 1))
}

pairs(~j(Sharia)+j(Marx)+j(SS)+j(Medi)+j(ChinaWar)+j(ChinaMoney)+j(Trad), data=df)
```



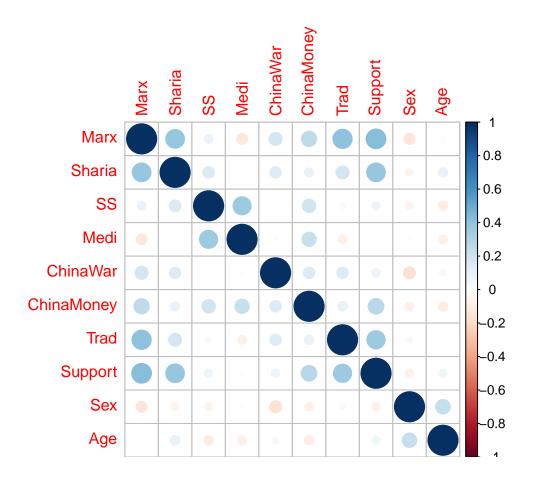
In terms of Support



Correlation Visualization

```
library(corrplot)
## Warning: package 'corrplot' was built under R version 3.2.5
```

```
M <- cor(df)
corrplot(M, method="circle")</pre>
```

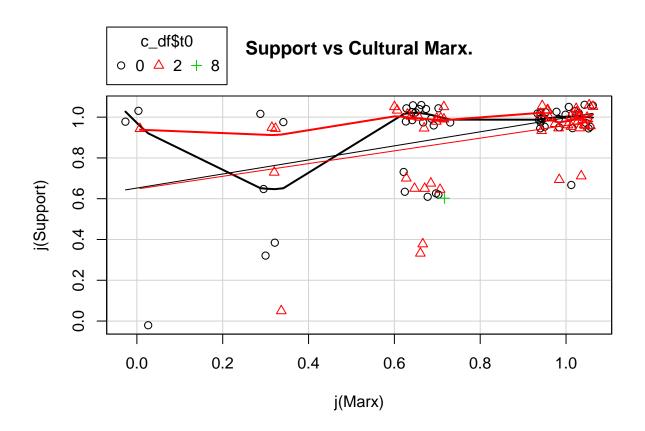


Load clustering results from classify_people.py

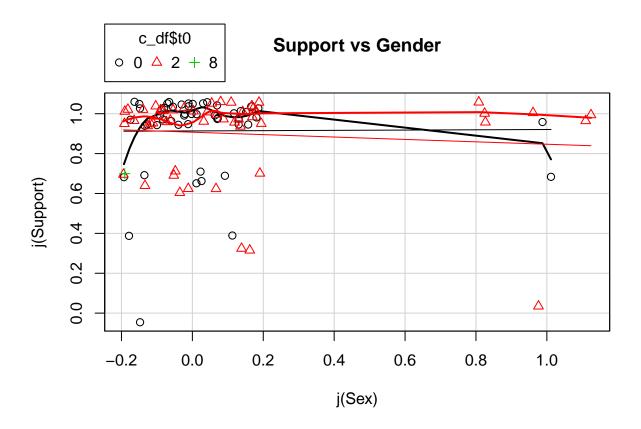
```
c_df=read.csv("C:\\dev\\bayes-present\\explore\\cluster_data_25.csv")
```

```
scatterplot(j(Support) ~ j(Marx)| c_df$t0, data=df,
main="Support vs Cultural Marx.")
```

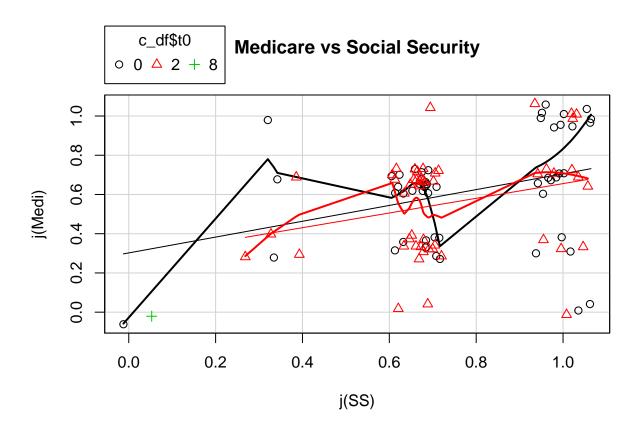
```
## Warning in smoother(.x[subs], .y[subs], col = col[i], log.x =
## logged("x"), : could not fit smooth
```



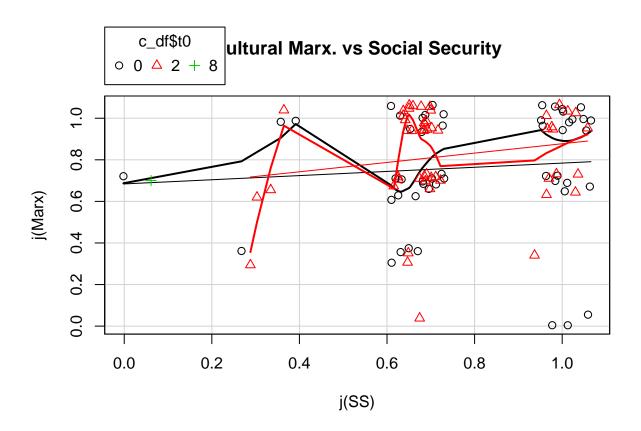
```
## Warning in smoother(.x[subs], .y[subs], col = col[i], log.x =
## logged("x"), : could not fit smooth
```



```
## Warning in smoother(.x[subs], .y[subs], col = col[i], log.x =
## logged("x"), : could not fit smooth
```



```
## Warning in smoother(.x[subs], .y[subs], col = col[i], log.x =
## logged("x"), : could not fit smooth
```



In terms of group

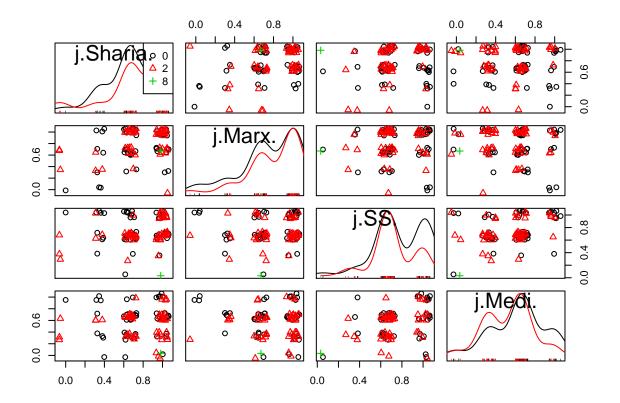
```
scatterplotMatrix(~j(Sharia)+j(Marx)+j(SS)+j(Medi)|c_df$t0, data=df,
reg.line=F, spread=FALSE, lwd=0, smoother=NULL)
```

```
## Warning in diag.panel(...): cannot estimate density for group 8
## Error in density.default(xx, adjust = adjust, na.rm = TRUE) :
## need at least 2 points to select a bandwidth automatically

## Warning in diag.panel(...): cannot estimate density for group 8
## Error in density.default(xx, adjust = adjust, na.rm = TRUE) :
## warning in diag.panel(...): cannot estimate density for group 8
## Error in density.default(xx, adjust = adjust, na.rm = TRUE) :
## need at least 2 points to select a bandwidth automatically

## Warning in diag.panel(...): cannot estimate density for group 8
## Error in density.default(xx, adjust = adjust, na.rm = TRUE) :
## need at least 2 points to select a bandwidth automatically

## need at least 2 points to select a bandwidth automatically
```

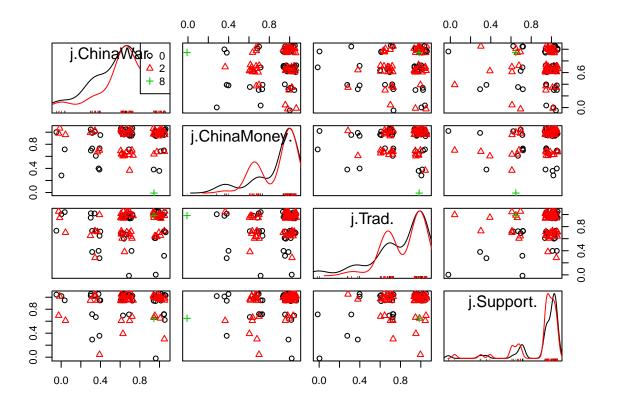


```
## Warning in diag.panel(...): cannot estimate density for group 8
## Error in density.default(xx, adjust = adjust, na.rm = TRUE) :
## need at least 2 points to select a bandwidth automatically

## Warning in diag.panel(...): cannot estimate density for group 8
## Error in density.default(xx, adjust = adjust, na.rm = TRUE) :
## warning in diag.panel(...): cannot estimate density for group 8
## Error in density.default(xx, adjust = adjust, na.rm = TRUE) :
## need at least 2 points to select a bandwidth automatically

## Warning in diag.panel(...): cannot estimate density for group 8
## Error in density.default(xx, adjust = adjust, na.rm = TRUE) :
## need at least 2 points to select a bandwidth automatically

## need at least 2 points to select a bandwidth automatically
```



principle component

```
pcs = prcomp(df, scale=TRUE)
```

bayesmix

```
library(bayesmix)
```

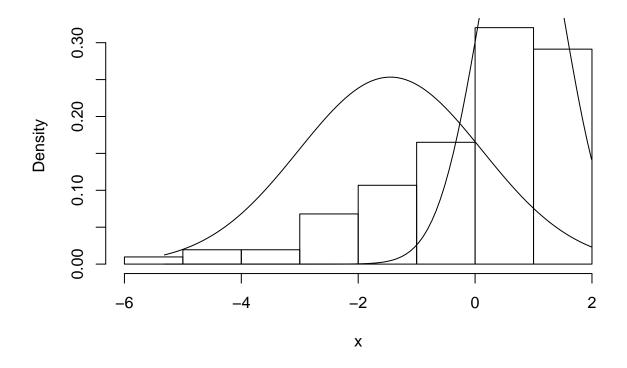
Warning: package 'bayesmix' was built under R version 3.2.5

```
get_mm=function(vec, p_string) {
    model = BMMmodel(
       vec, k = 2,
       priors = list(kind = "independence", parameter=p_string),
       no.empty.classes=TRUE
)
    control = JAGScontrol(variables = c("mu", "tau", "eta", "S"), burn.in = 1000, n.iter = 5000, seed = 1
    z = JAGSrun(vec, model = model, control = control, initialValues=list(S0=4))
    return(z)
}
mm1 = get_mm(pcs$x[,1],"priorsUncertain")
```

Compiling model graph

```
Declaring variables
##
##
      Resolving undeclared variables
##
      Allocating nodes
## Graph information:
##
      Observed stochastic nodes: 105
##
      Unobserved stochastic nodes: 108
##
      Total graph size: 549
##
## Initializing model
plot_hist_with_norms=function(x, mm) {
  # Assume mm is a mixture model with two components
  hist(x, freq=FALSE)
  xfit=seq(min(x), max(x), length=100)
  y1=dnorm(xfit, mean=mean(mm$results[,"mu[1]"]), sd=sqrt(mean(mm$results[,"sigma2[1]"])))
  y2=dnorm(xfit, mean=mean(mm$results[,"mu[2]"]), sd=sqrt(mean(mm$results[,"sigma2[2]"])))
  lines(xfit, y1)
  lines(xfit, y2)
}
plot_hist_with_norms(pcs$x[,1], mm1)
```

Histogram of x



```
get_average_groups=function(mm, n) {
  av=c()
```

```
for(i in 1:n) {
    av[i] = mean(mm$results[,paste("S[", i, "]", sep="")])
}
    return(av)
}

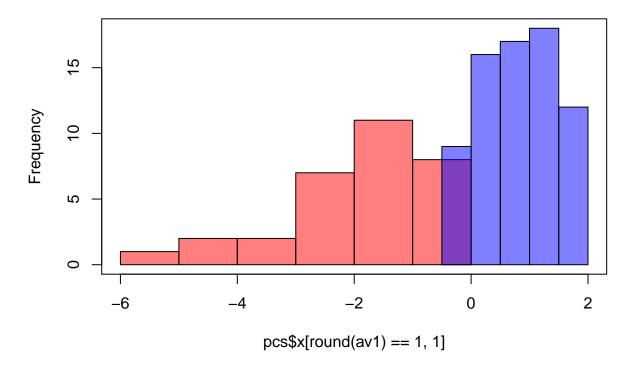
# Read in EM with two fixed clusters over 10000 trials
c_2_df=read.csv("C:\\dev\\bayes-present\\explore\\cluster_data_2_comps_10000_trials.csv")

c_2_avg=c()
for(i in 1:103) {
    c_2_avg[i] = mean(t(c_2_df[i,])) + 1
}

av1=get_average_groups(mm1, 103)

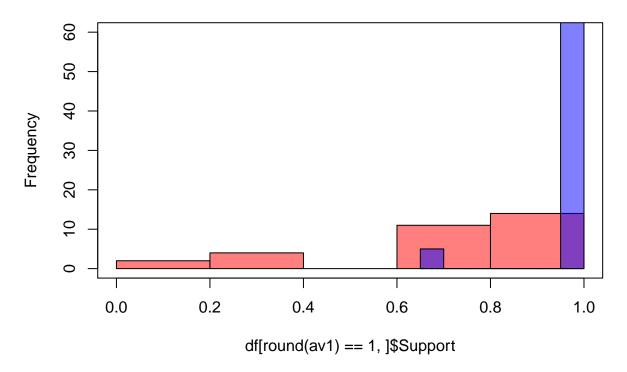
# Histogram Colored (blue and red)
hist(pcs$x[round(av1)==1,1], col=rgb(1,0,0,0.5), xlim=c(-6, 2), ylim=c(0, 18))
hist(pcs$x[round(av1)==2,1], col=rgb(0,0,1,0.5), add=T)
box()
```

Histogram of pcsx[round(av1) == 1, 1]



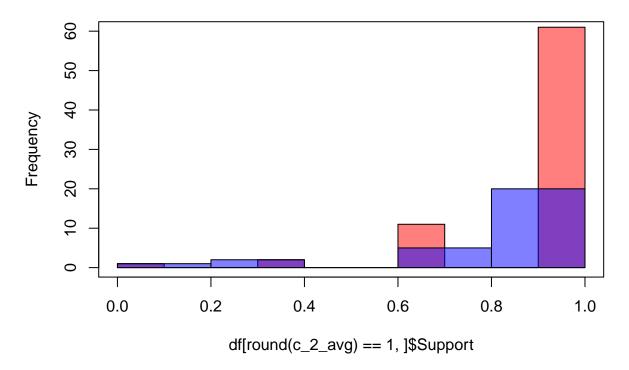
```
# Histogram Colored (blue and red)
hist(df[round(av1)==1,]$Support, col=rgb(1,0,0,0.5), ylim=c(0, 60))
hist(df[round(av1)==2,]$Support, col=rgb(0,0,1,0.5), add=T)
box()
```

Histogram of df[round(av1) == 1,]\$Support



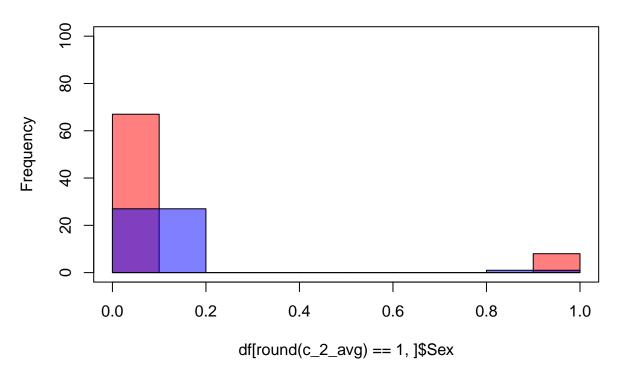
```
hist(df[round(c_2_avg)==1,]$Support, col=rgb(1,0,0,0.5), ylim=c(0, 60))
hist(df[round(c_2_avg)==2,]$Support, col=rgb(0,0,1,0.5), add=T)
box()
```

Histogram of df[round(c_2_avg) == 1,]\$Support



```
hist(df[round(c_2_avg)==1,]$Sex, col=rgb(1,0,0,0.5), ylim=c(0, 100))
hist(df[round(c_2_avg)==2,]$Sex, col=rgb(0,0,1,0.5), add=T)
box()
```

Histogram of df[round(c_2_avg) == 1,]\$Sex



```
percentage_female_c2_1 = mean(df[round(c_2_avg)==1,]$Sex)
percentage_female_c2_2 = mean(df[round(c_2_avg)==2,]$Sex)
```

```
# Gets differential entropy of observations which are
# assumed to follow multivariate normal
dif_entropy=function(obs) {
  k=ncol(obs)
  return((k/2*(1+log(2*pi))) + (1/2*log(det(cov(obs)))))
}
random_entropy_draws=function(obs) {
  ITERATIONS = 1000
  trial=c()
  sizes=c()
  for(it in 1:ITERATIONS) {
    # Make sure we ignore degenerate cases
    while(!is.numeric(ent) | !is.finite(ent)) {
      partition=sample(0:1,nrow(obs),replace=T)
      ent=dif_entropy(df[partition==1,])
      sizes[it] = sum(partition)
      trial[it] = ent
    }
  }
  print(paste(ITERATIONS, "trials"))
  print(paste("Average trial size:", mean(sizes)))
```

```
return(trial)
}
mc_draws=random_entropy_draws(df)
## [1] "1000 trials"
## [1] "Average trial size: 51.736"
print(mean(mc_draws))
## [1] 2.68561
print(quantile(mc_draws, c(.025, .975)))
##
       2.5%
               97.5%
## 1.643874 3.617058
print((av=dif_entropy(df)))
## [1] 3.130841
em_1=dif_entropy(df[round(c_2_avg)==1,])
em_2=dif_entropy(df[round(c_2_avg)==2,])
bay_1=dif_entropy(df[round(av1)==1,])
bay_2=dif_entropy(df[round(av1)==2,])
```

Clustering has

Correlation Visualization

```
library(corrplot)
M <- cor(df[round(c_2_avg)==1,])
corrplot(M, method="circle")</pre>
```

