# Testing different imputation methods on PUMS (MCAR) - MICE

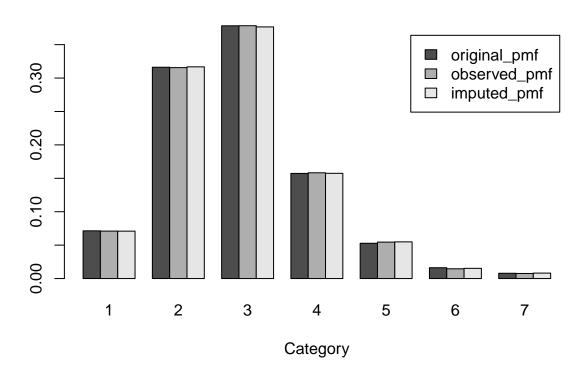
```
# load dataset: df
load('../Datasets/ordinalPUMS.Rdata')
# take 10,000 samples: df
set.seed(0)
n = 10000
sample <- sample(nrow(df), size = 10000)</pre>
df <- df[sample,]</pre>
# create MCAR scneario with 30% chance of missing: df_observed
missing_prob = 0.3
df_observed <- df
missing_col = colnames(df)[c(1,3,5,7,9,11)]
for (col in missing_col) {
  missing_ind <- rbernoulli(n,p = missing_prob)</pre>
  df_observed[missing_ind, col] <- NA</pre>
}
MICE
Create 5 imputed dataset
library(mice)
##
## Attaching package: 'mice'
## The following objects are masked from 'package:base':
##
##
       cbind, rbind
imputed_df <- mice(df_observed,m=5,print=F)</pre>
## Warning: Number of logged events: 150
Extract the 5 imputed dataset
d1 <- complete(imputed_df, 1)</pre>
d2 <- complete(imputed_df, 2)</pre>
d3 <- complete(imputed_df, 3)</pre>
d4 <- complete(imputed_df, 4)</pre>
d5 <- complete(imputed_df, 5)</pre>
imputed_sets = rbind(d1, d2, d3, d4, d5)
Diagnostics
Assess bivariate joint distribution
```

Assess trivariate joint distribution

numeric\_df = sapply(df, as.numeric)

# calculate rmse

### MICE: VEH



```
normalized_df = t(t(numeric_df-1)/(apply(numeric_df, MARGIN = 2, FUN = max)-1))
numeric_impute = sapply(d1, as.numeric)
normalized_impute = t(t(numeric_impute-1)/(apply(numeric_df, MARGIN = 2, FUN = max)-1))
missing_matrix = is.na(df_observed)

rmse = sqrt(sum((normalized_df[missing_matrix] - normalized_impute[missing_matrix])^2)/sum(missing_matr rmse

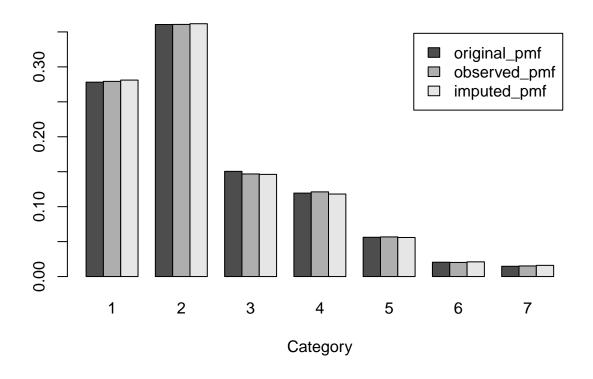
## [1] 0.2634679

# accuracy
acc = sum(numeric_df[missing_matrix] == numeric_impute[missing_matrix])/sum(missing_matrix)
acc

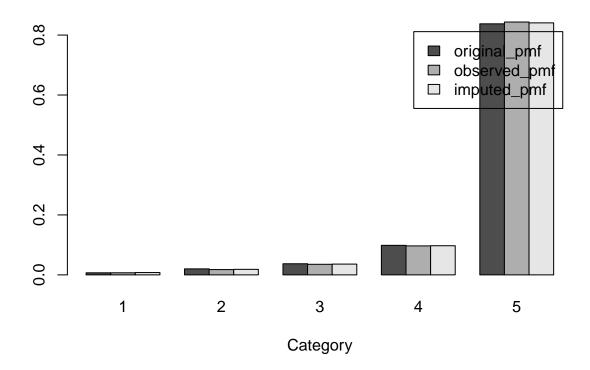
## [1] 0.3411563

# Actual vs Imputed values
col = 1
missing_indicator = missing_matrix[,col]
plot(as.integer(df[missing_indicator, col]), as.integer(d1[missing_indicator, col]), xlab = 'actual values')
```

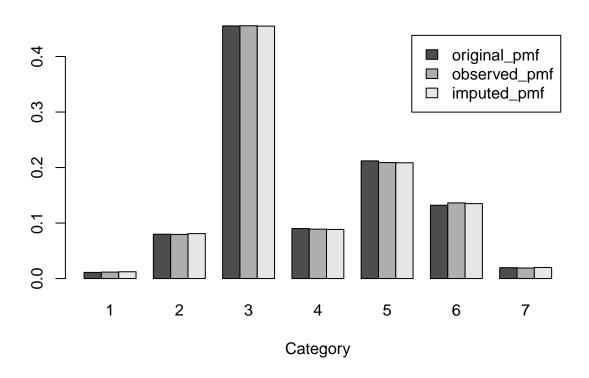
MICE: NP



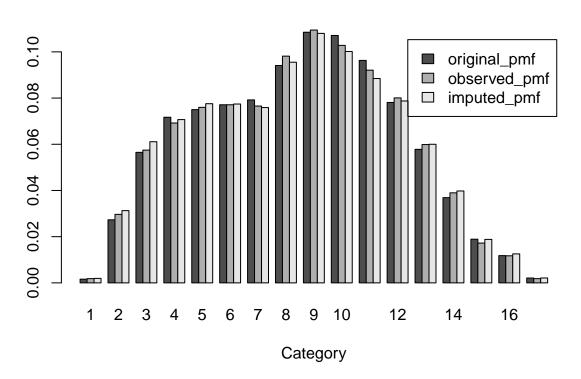
**MICE: ENG** 

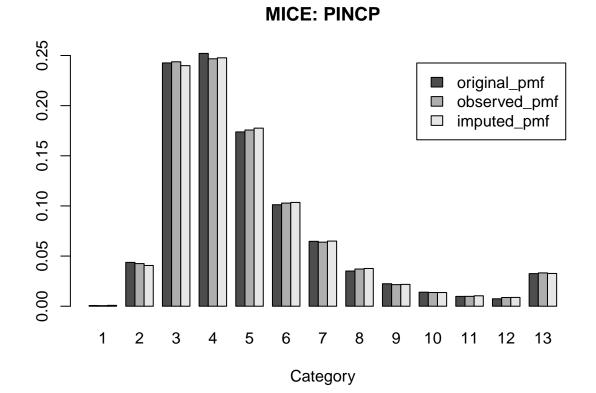


**MICE: SCHL** 

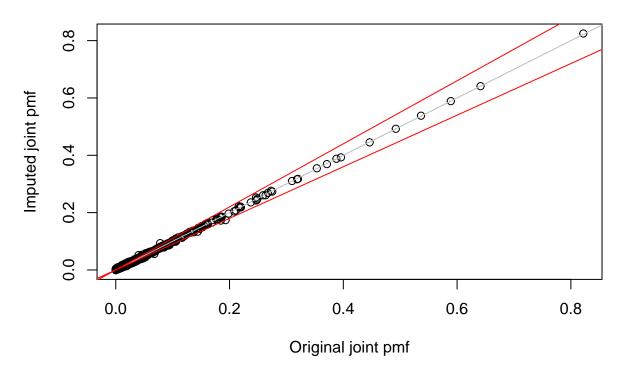


## **MICE: AGEP**

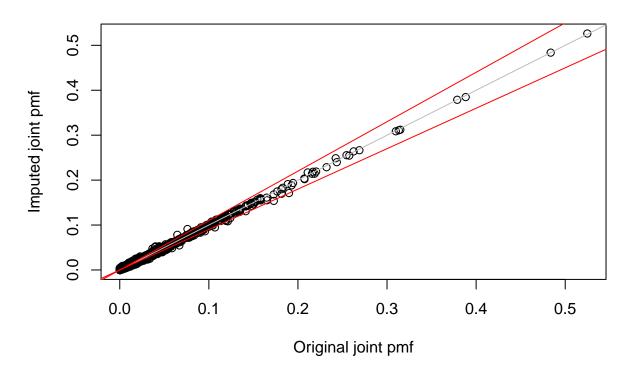




## Bivariate pmf, r square: 1



Trivariate pmf , r square: 0.999



## Actual vs Imputed values: VEH

