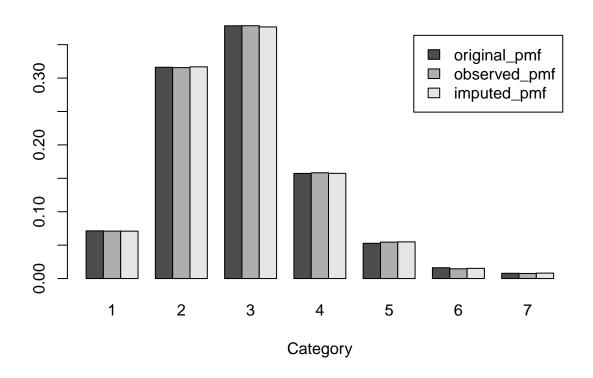
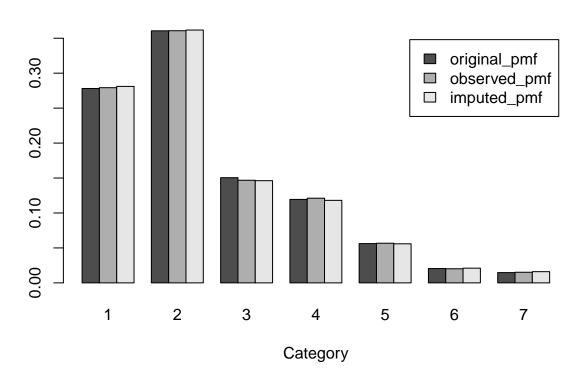
# Testing different imputation methods on PUMS (MCAR) - MICE

```
# 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
```

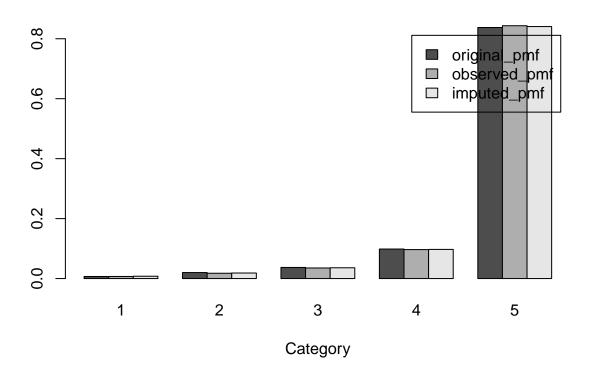
MICE: VEH



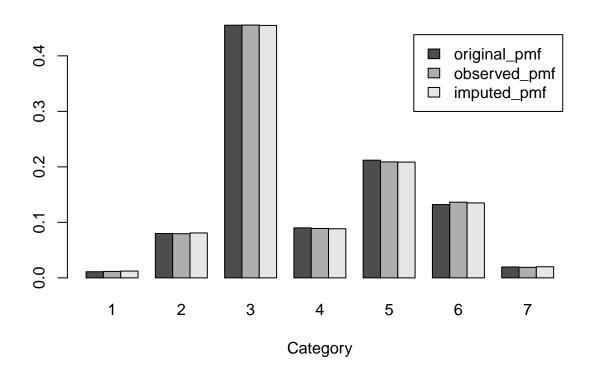
## MICE: NP



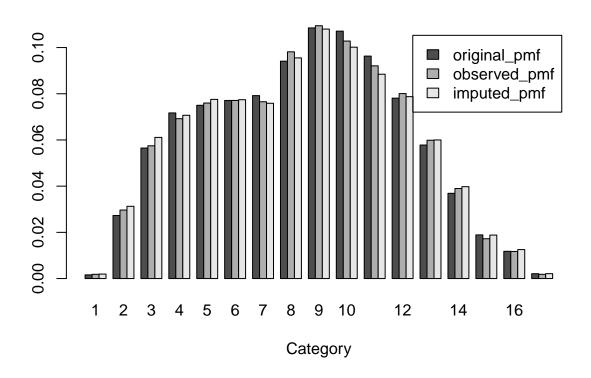
**MICE: ENG** 



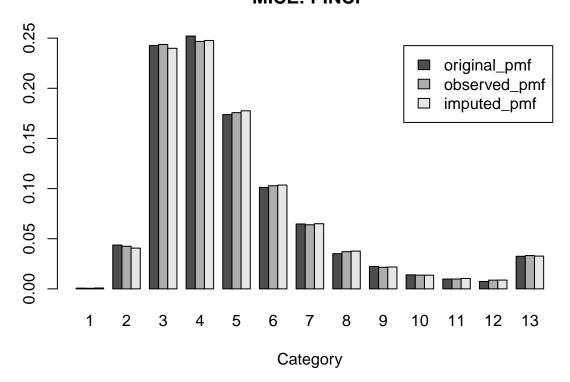
#### **MICE: SCHL**



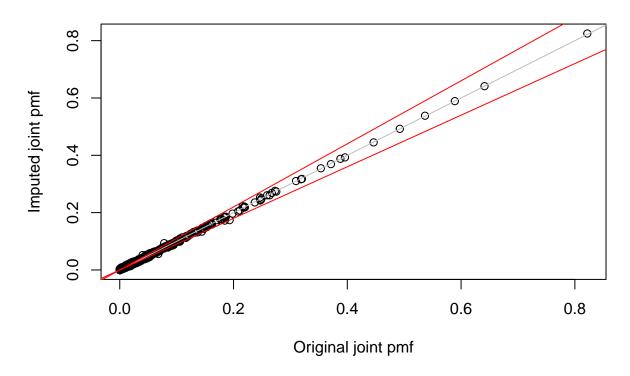
**MICE: AGEP** 







## Bivariate pmf, r square: 1



## Trivariate pmf, r square: 0.999

