

# CS 199 BD Homework 3

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## 1 Linear Regression

Upon building a linear regression, we noticed that there are a lot of data points which were missing values. In order to build our linear regression, we removed these values.

## 2 R Code

```
crime<-read.csv('communities.data', header=FALSE)
crime<-crime[c(-1,-2,-3,-4,-5)]
crime<-crime[sample(nrow(crime)),]
crime[crime == '?'] <- NA
#replace all '?' with NA
drop_cols <- crime[complete.cases(crime), ] # only take
the variables w/ all the values
library(DAAG)
fit<-lm(V128 ~ V6+ V7+ V8+ V9+ V10+ V11+ V12+ V13+ V14+
V15+ V16+ V17+ V18+ V19+ V20+ V21+ V22+ V23+ V24+ V25+
V26+ V27+ V28+ V29+ V30+ V32+ V33+ V34+ V35+ V36+ V37
+ V38+ V39+ V40+ V41+ V42+ V43+ V44+ V45+ V46+ V47+
V48+ V49+ V50+ V51+ V52+ V53+ V54+ V55+ V56+ V57+ V58+
V59+ V60+ V61+ V62+ V63+ V64+ V65+ V66+ V67+ V68+ V69
+ V70+ V71+ V72+ V73+ V74+ V75+ V76+ V77+ V78+ V79+
V80+ V81+ V82+ V83+ V84+ V85+ V86+ V87+ V88+ V89+ V90+
V91+ V92+ V93+ V94+ V95+ V96+ V97+ V98+ V99+ V100+
V101+ V119+ V120+ V121+ V126, data=drop_cols)
cv.lm(df=drop_cols, fit, m=4)
library('MASS')
boxcox(fit, lambda = seq(0, 1, 1/10), plotit=TRUE)
crime_lambda1 = drop_cols
for (i in 1:nrow(crime_lambda1)){
  crime_lambda1[i, ncol(crime_lambda1)] = (crime_lambda1[
    i, ncol(crime_lambda1)]^0.3-1)/0.3
```

```

}

fit2<-lm(V128 ~ V6+ V7+ V8+ V9+ V10+ V11+ V12+ V13+ V14+
  V15+ V16+ V17+ V18+ V19+ V20+ V21+ V22+ V23+ V24+ V25+
  V26+ V27+ V28+ V29+ V30+ V32+ V33+ V34+ V35+ V36+ V37
  + V38+ V39+ V40+ V41+ V42+ V43+ V44+ V45+ V46+ V47+
  V48+ V49+ V50+ V51+ V52+ V53+ V54+ V55+ V56+ V57+ V58+
  V59+ V60+ V61+ V62+ V63+ V64+ V65+ V66+ V67+ V68+ V69
  + V70+ V71+ V72+ V73+ V74+ V75+ V76+ V77+ V78+ V79+
  V80+ V81+ V82+ V83+ V84+ V85+ V86+ V87+ V88+ V89+ V90+
  V91+ V92+ V93+ V94+ V95+ V96+ V97+ V98+ V99+ V100+
  V101+ V119+ V120+ V121+ V126, data=crime_lambda1)
cv.lm(df=crime_lambda1, fit2, m=4)

##### KNN
#####

comm<-read.csv('communities.data', header=FALSE);
comm<-comm[c(-1,-2,-3,-4,-5)]
#delete the first 5 vars which are not predictive

library('FNN')
#comm = comm[sample(nrow(comm)),]
comm[comm == '?'] <- NA
#replace all '?' with NA
full <- comm[complete.cases(comm),]
#then only take the ones w/ all the values
lapply(full, as.numeric)

full = subset(full, select=c(V31, V102, V103, V104, V105
, V106, V107, V108, V109, V111, V110, V112, V113, V114
, V115, V116, V117, V118, V122, V123, V124, V125, V127
))
comm_full = subset(comm, select=c(V31, V102, V103, V104,
V105, V106, V107, V108, V109, V111, V110, V112, V113,
V114, V115, V116, V117, V118, V122, V123, V124, V125,
V127))
#do some nearest neighbor stuff
wtrain <- full[1:100,1:(ncol(full)-1)]
wtrl <- full[1:100,(ncol(full))]
wtest <- full[101:200, 1:(ncol(full)-1)]
wtel <- full[101:200, ncol(full)]
#results = knn(wtrain, wtest, wtrl, k = 10, algorithm="
cover_tree")
results = knn(wtrain, wtest, wtrl, k = 21, algorithm="
cover_tree")

```

```

plot(as.numeric(results), as.numeric(wtel))
library(impute)
imputed=impute.knn(as.matrix(comm_full), k=10)
fit<-lm(V128 ~ V6+ V7+ V8+ V9+ V10+ V11+ V12+ V13+ V14+
        V15+ V16+ V17+ V18+ V19+ V20+ V21+ V22+ V23+ V24+ V25+
        V26+ V27+ V28+ V29+ V30+ V32+ V33+ V34+ V35+ V36+ V37
        + V38+ V39+ V40+ V41+ V42+ V43+ V44+ V45+ V46+ V47+
        V48+ V49+ V50+ V51+ V52+ V53+ V54+ V55+ V56+ V57+ V58+
        V59+ V60+ V61+ V62+ V63+ V64+ V65+ V66+ V67+ V68+ V69
        + V70+ V71+ V72+ V73+ V74+ V75+ V76+ V77+ V78+ V79+
        V80+ V81+ V82+ V83+ V84+ V85+ V86+ V87+ V88+ V89+ V90+
        V91+ V92+ V93+ V94+ V95+ V96+ V97+ V98+ V99+ V100+
        V101+ V119+ V120+ V121+ V126, data=as.data.frame(
        imputed$data))
cv.lm(df=as.data.frame(imputed$data), fit, m=4)

```