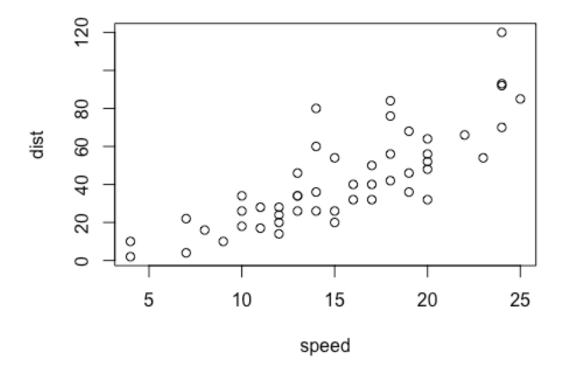
## **R Notebook**

This is an R Markdown Notebook. When you execute code within the notebook, the results appear beneath the code.

Try executing this chunk by clicking the *Run* button within the chunk or by placing your cursor inside it and pressing *Cmd+Shift+Enter*.

plot(cars)



Add a new chunk by clicking the *Insert Chunk* button on the toolbar or by pressing *Cmd+Option+I*.

When you save the notebook, an HTML file containing the code and output will be saved alongside it (click the *Preview* button or press *Cmd+Shift+K* to preview the HTML file).

The preview shows you a rendered HTML copy of the contents of the editor. Consequently, unlike *Knit*, *Preview* does not run any R code chunks. Instead, the output of the chunk when it was last run in the editor is displayed.

pdf("output.pdf")

```
install.packages('bayesm') library(bayesm) data(margarine) head(margarine)
library(mlogit) library(tinytex)
#Q1
data = margarine$choicePrice #Average
apply(as.matrix(margarine$choicePrice[,c(3:12)]),2,mean) #dispersion
apply(as.matrix(margarinechoicePrice[, c(3:12)]),2,sd)apply(as.matrix(margarinechoi
cePrice[,c(3:12)]),2,range) apply(as.matrix(margarine$choicePrice[,c(3:12)]),2,var)
#Market share choice frequency nrow(data)
tb share <-table(data$choice) names(tb share) <-
colnames(data)[3:12][as.numeric(names(tb share))] tb share
#market share by product characteristics(choice frequency by price bins:below average,
over average) df <-data.frame() for(i in c(1:10)){ df_new <-
data[which(data$choice==i),c('hhid','choice',colnames(data)[i+2])] colnames(df new)[3] <-
'price' df <-rbind(df,df_new) }
avg price <-tapply(dfprice, df choice, mean) avg price <-
data.frame(choice=names(avg_price),avg_prc=avg_price)
df <-merge(df,avg_price,by='choice')</pre>
over_average <-tapply(ifelse(dfprice >=
df avg_prc,1,0),dfchoice, sum)names(over_average) <
-colnames(data)[3:12][as.numeric(names(over_average))]below_average <
-tapply(ifelse(dfprice < dfavg_nrc, 1,0), df choice,sum) names(below_average) <-
colnames(data)[3:12][as.numeric(names(below average))]
fre_by_price_bins<-rbind(tb_share,over_average,below_average)</pre>
#Illustrate the mapping between observed attributes and choices The combination of
Parkay brand and stick type has the largest market share. The lowest market share is the
combination of House Brand and tub Type. From the aspect of product type, the market
share of stick is generally larger than the market share of tub. From the aspect of Brands,
Parkay is the most popular brand.
#Q2 First Model
#choice matrix #choice matrix #choice matrix
clogit=function(param){ data = margarine$choicePrice ni=nrow(data) nj=ncol(data[,3:12])
df < -data.frame() for( i in c(1:10)){ df_new < -
data[which(data$choice==i),c('hhid','choice',colnames(data)[i+2])] colnames(df_new)[3] <-
'price' df <-rbind(df,df_new) }
dfchoice < -as. factor(dfchoice) df <-cbind(df, as. data. frame(model. matrix(~choice-
1,df))) Y=df[,4:13]
```

```
intercept=matrix(rep(param[1:nj-1],each=ni),nrow=ni,ncol=nj-1)
XB=cbind(0,intercept)+data[,3:12] param[nj] eXB=exp(XB) prob=eXB/rowSums(eXB)
llike=sum(Ylog(prob)) return(-llike) } set.seed(1000) param=runif(10,-1,1)
model1=optim(param,clogit,method="BFGS") model1$par
#interpretation of coefficient increasing in price will decrease the probability of choosing
choice 1-9 but increase the probability of choosing choice 10.
#Q3 Second Model
data = margarinechoicePricedata2 = margarinedemos data3 =
merge(data,data2[,c(1:2)],by='hhid')
mlogit
mlogit = function(param) { data = data3 Income = data1ncomechoice = datachoice
ni = nrow(data) nj = length(unique(choice)) ut = mat.or.vec(ni,nj) # multinomial logit pn1 =
param[1:nj]
# multinomial logit for (j in 1:nj) {
ut[,j] = Income*pn1[j] \} #exp(XB) prob = exp(ut)
#sprob = rowsums(prob) # sum_j exp(XB) denominator prob =
sweep(prob,MARGIN=1,FUN="/",STATS=rowSums(prob)) # an example of how to
construct
avg_param <-c() for(j in 1:nj){ avg_param<-c(avg_param,mean(prob[,j]*param[j])) } #</pre>
match prob to actual choices probc = NULL me=mat.or.vec(ni,nj)
for (i in 1:ni) { probc[i] = prob[i, choice[i]] for (j in 1:nj) { me[i,j] = probc[i]*(param[j]-i)
avg_param[j]) }
} probc[probc>0.999999] = 0.999999 probc[probc<0.000001] = 0.000001 like =
sum(log(probc)) result = list(like=-like,me=me) return(result) }
set.seed(1000) param = rep(0,10)
like_fun = function(param){ result = mlogit(param) like=result$like return(like) } model <-
optim(param,like fun,method='BFGS') model
#Interprete coefficient on family Increasing in family income could reduce the probability
of choosing choice 1.
#Q4
#marginal effect of first model
#marginal effect of Second model me=mlogit(modelpar)me summary(me) Increasing one
unit of income could change the probability of choosing the choice.
```

#Q5 #combining dataset of price and income as data4 data = margarinechoicePricedata2 = margarinedemos data4 = merge(df,data2[,c(1:2)],by='hhid')