Code Appendix

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# Libraries
library(MVN); library(gridExtra); library(ztable); library(ICSNP); library(data.table)
library(devtools); library(ggbiplot); library(ggfortify); library(rrcov); library(MASS)
# Load data
# https://data.world/craiqkelly/usda-national-nutrient-db/workspace/file?filename=nndb_flat.csv
nutrition <- read.csv('nndb_flat.csv')</pre>
# Filter out some food groups, columns
nutrition <- subset(nutrition, nutrition[,2] %in% c("Baked Products",
  "Cereal Grains and Pasta", "Finfish and Shellfish Products",
  "Lamb, Veal, and Game Products", "Pork Products", "Vegetables and Vegetable Products",
  "Beef Products", "Dairy and Egg Products", "Fats and Oils", "Fruits and Fruit Juices",
  "Legumes and Legume Products", "Nut and Seed Products", "Poultry Products"))
nutrition$FoodGroup <- droplevels(nutrition$FoodGroup)</pre>
# Multivariate Normality Check
# https://cran.r-project.org/web/packages/MVN/vignettes/MVN.pdf
mardia <- mvn(data = nutrition[nutrition$FoodGroup %in%
  c('Beef Products', 'Vegetables and Vegetable Products'), c(8:30)], mvnTest = "mardia")
hz <- mvn(data = nutrition[nutrition$FoodGroup %in%
  c('Beef Products', 'Vegetables and Vegetable Products'), c(8:30)], mvnTest = "hz")
royston <- mvn(data = nutrition[nutrition$FoodGroup %in%
  c('Beef Products', 'Vegetables and Vegetable Products'), c(8:30)], mvnTest = "royston")
# Choose 2 populations: Beef, Vegetables
beef_names <- nutrition[nutrition$FoodGroup %in%</pre>
  c('Beef Products', 'Vegetables and Vegetable Products'),]
nut clean <- nutrition[nutrition$FoodGroup %in%</pre>
  c('Beef Products', 'Vegetables and Vegetable Products'),]
nut_clean$FoodGroup <- droplevels(nut_clean$FoodGroup)</pre>
beef_names$FoodGroup <- droplevels(beef_names$FoodGroup)</pre>
nut_clean <- nut_clean[order(nut_clean$FoodGroup),]</pre>
beef_names <- beef_names[order(beef_names$FoodGroup),]</pre>
nut_clean \leftarrow nut_clean[,c(2,8:30)]
beef_names \leftarrow beef_names[,c(2,4,8:30)]
# no NA's
any(is.na(nut_clean))
nrow(nut_clean)
sort(sapply(nut_clean[,-1], function(x) sum(x == 0)))
# Sample estimates
x <- nut_clean[,-1]</pre>
y <- factor(nut_clean[,1])</pre>
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comb_s_mean <- colMeans(x)</pre>
pop_2_s_mean <- sapply(x, function(w) tapply(w, y, mean))</pre>
sample_var_cov <- lapply(split(x, y), var)</pre>
# https://stackoverflow.com/questions/42860716/export-dataframe-to-pdf-png-in-r
comb s mean <- data.frame(comb s mean)</pre>
colnames(comb_s_mean) <- c("Sample Mean")</pre>
comb s mean$Nutrients <- rownames(comb s mean)</pre>
rownames(comb s mean) <- NULL</pre>
comb_s_mean$Nutrients <- gsub("_", " ", comb_s_mean$Nutrients)</pre>
comb_s_mean$`Sample Mean` <- round(comb_s_mean$`Sample Mean`, 4)</pre>
grid.table(comb_s_mean)
colnames(pop_2_s_mean) <- gsub("_", " ", colnames(pop_2_s_mean))</pre>
pop_2_s_mean <- t(pop_2_s_mean)</pre>
colnames(pop_2_s_mean) <- c("Beef", "Vegetables")</pre>
grid.table(round(pop_2_s_mean, 4))
png('var_cov_mat.png', height = 1200, width = 4000)
grid.table(sample_var_cov$`Beef Products`)
dev.off()
# Barplot
# https://stackoverflow.com/questions/29639680/r-table-function-how-to-remove-0-counts
# https://www.r-graph-gallery.com/213-rotating-x-axis-labels-on-barplot/
par(mar = c(10,3,3,3))
nut_table <- sort(table(nutrition[,2]), decreasing = TRUE)</pre>
food_group_barplot <- barplot(nut_table, names.arg="",</pre>
                               las = 1, main = 'Barplot of Food Groups Frequency')
xlab_names <- names(nut_table)</pre>
xlab_names <- gsub(" Products", "", xlab_names)</pre>
xlab_names[2] <- "Vegetables"</pre>
text(food_group_barplot[,1], -3.7, srt = 60, adj= 1, xpd = TRUE, labels = xlab_names, cex=0.7)
# Distribution table
# https://www.quru99.com/r-sort-data-frame.html
# https://stackoverflow.com/questions/24428051/removing-display-of-row-names-from-data-frame
food_group_dist <- cbind(as.data.frame(prop.table(table(droplevels(nutrition[,2])))*100),</pre>
                          as.data.frame(table(droplevels(nutrition[,2])))[,2])
colnames(food_group_dist) <- c('Food Groups', 'Percentage (%)', 'Count')</pre>
food_group_dist <- food_group_dist[order(food_group_dist$Count, decreasing = TRUE),]</pre>
rownames(food_group_dist) <- NULL</pre>
ztable(food_group_dist, caption = 'Percentage and Count of Food Groups')
# Numerical summary
\# https://stackoverflow.com/questions/11346880/r-plot-multiple-box-plots-using-columns-from-data-frame
# https://stackoverflow.com/questions/23050928/error-in-plot-new-figure-margins-too-large-scatter-plot
# https://www.r-bloggers.com/setting-graph-margins-in-r-using-the-par-function-and-lots-of-cow-milk/
X_list <- split(nut_clean[,-1], nut_clean$FoodGroup)</pre>
colnames(X_list[[1]]) <- gsub("_", " ", colnames(X_list[[1]]))</pre>
colnames(X_list[[2]]) <- gsub("_", " ", colnames(X_list[[2]]))</pre>
X_list_summary <- lapply(X_list, function(x) sapply(x, summary))</pre>
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grid.table(t(round(X_list_summary$`Beef Products`, 4)))
grid.table(t(round(X_list_summary$`Vegetables and Vegetable Products`, 4)))
# Boxplots
par(mfrow=c(5,5))
par(mar=c(2,2,5,2))
nut_clean_bar <- subset(nut_clean, nut_clean$Fiber_g < 15 & nut_clean$VitA_mcg < 2000 &</pre>
       nut clean$VitB12 mcg < 40 & nut clean$VitC mg < 500 & nut clean$VitE mg < 3 &
       nut_clean$Folate_mcg < 200 & nut_clean$Niacin_mg < 20 & nut_clean$Riboflavin_mg < 1.5 &</pre>
       nut_clean$Thiamin_mg < 5 & nut_clean$Calcium_mg < 400 & nut_clean$Copper_mcg < 5 &
       nut_clean$Iron_mg < 20 & nut_clean$Magnesium_mg < 200 & nut_clean$Manganese_mg < 3 &
       nut_clean$Selenium_mcg < 100)</pre>
beef_names <- subset(beef_names, beef_names$Fiber_g < 15 & beef_names$VitA_mcg < 2000 &
       beef_names$VitB12_mcg < 40 & beef_names$VitC_mg < 500 & beef_names$VitE_mg < 3 &
       beef_names$Folate_mcg < 200 & beef_names$Niacin_mg < 20 & beef_names$Riboflavin_mg < 1.5 &
       beef_names$Thiamin_mg < 5 & beef_names$Calcium_mg < 400 & beef_names$Copper_mcg < 5 &
       beef_names$Iron_mg < 20 & beef_names$Magnesium_mg < 200 & beef_names$Manganese_mg < 3 &
       beef_names$Selenium_mcg < 100)</pre>
levels(nut_clean_bar$FoodGroup) = c("Beef", "Vegs")
levels(beef_names$FoodGroup) = c("Beef", "Vegs")
colnames(nut_clean_bar) <- gsub("_", " ", colnames(nut_clean_bar))</pre>
colnames(beef_names) <- gsub("_", " ", colnames(beef_names))</pre>
# create boxplot
for (i in 2:24) {
 boxplot(nut_clean_bar[,i] ~ FoodGroup,
          data = nut clean bar, main = names(nut clean bar)[i])
}
mtext("Boxplot of Nutrients for Beef and Vegetable Products",
      side = 3, line = -1, outer = TRUE, cex = 0.8)
dev.off()
### Confidence Interval
# One-sample inference about a mean vector
# Compute sample mean vector and
# sample covariance matrix
# https://stackoverflow.com/questions/3369959/moving-columns-within-a-data-frame-without-retyping
# filter data
beef_filter <- nut_clean_bar[nut_clean_bar$FoodGroup == 'Beef', c(-1,-6)]</pre>
beef_names <- beef_names[beef_names$FoodGroup == 'Beef', 2]</pre>
vege_filter <- nut_clean_bar[nut_clean_bar$FoodGroup == 'Vegs', c(-1,-6)]</pre>
xbar <- colMeans(beef filter)</pre>
xvar <- var(beef_filter)</pre>
# One-at-a-time confidence interval for mu_1, \ldots, mu_p
univariate_test <- sapply(beef_filter, function(x) t.test(x, mu = round(mean(x))))
uni_test_pval <- as.data.frame(unlist(univariate_test[3,]))</pre>
colnames(uni_test_pval) <- c('p-value')</pre>
uni_test_pval$Nutrient <- rownames(uni_test_pval)</pre>
uni_test_pval <- subset(uni_test_pval, select = c('Nutrient', 'p-value'))
rownames(uni_test_pval) <- NULL</pre>
uni_test_pval[,2] <- round(uni_test_pval[,2], 4)
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uni_test_pval[uni_test_pval[,2] > 0.05,]
grid.table(uni_test_pval)
uni_ci <- round(t(as.data.frame(univariate_test[4,])), 4)</pre>
colnames(uni_ci) <- c("Lower Bound", "Upper Bound")</pre>
grid.table(uni_ci)
# Bonferroni confidence interval for mu 1,...,mu p
bonf test <- sapply(beef filter, function(x) t.test(x,</pre>
  conf.level = (1-(0.05)/ncol(beef filter)), mu = round(mean(x))))
bonf_ci <- round(t(as.data.frame(bonf_test[4,])), 4)</pre>
colnames(bonf_ci) <- c("Lower Bound", "Upper Bound")</pre>
grid.table(bonf ci)
bonf_test_pval <- min(as.data.frame(unlist(bonf_test[3,])))</pre>
# Hotelling confidence interval for mu_1,...,mu_p
p_hotelling <- ncol(beef_filter)</pre>
n_hotelling <- nrow(beef_filter)</pre>
hotelling_crit_value <- ((p_hotelling*(n_hotelling - 1))/(n_hotelling - p_hotelling)) *
       qf(p = .95, df1 = p_hotelling, df2 = (n_hotelling - p_hotelling))
hotelling_lb <- xbar - sqrt(hotelling_crit_value * diag(xvar))</pre>
hotelling_ub <- xbar + sqrt(hotelling_crit_value * diag(xvar))</pre>
hotelling_ci <- round(as.data.frame(cbind(hotelling_lb, hotelling_ub)), 4)
colnames(hotelling_ci) <- c("Lower Bound", "Upper Bound")</pre>
grid.table(hotelling ci)
# Compute Hotelling statistic
p <- length(colnames(beef_filter))</pre>
n <- nrow(beef filter)</pre>
nullmean <- round(xbar)</pre>
d <- xbar-nullmean
t2 <- n*t(d)%*%solve(xvar)%*%d
cval <- (n-1)*p/(n-p)*qf(0.95,p,n-p)
t2mod <- (n-p)*t2/(p*(n-1))
pval \leftarrow 1- pf(t2mod,p,n-p)
cat("Hotelling T-squared statistic", fill=T)
cat("p-value", fill=T)
pval
# alternative using the function in the ISCP package
HotellingsT2(X = beef_filter, mu = nullmean)
# Confidence Region
# https://stackoverflow.com/questions/35805555/return-max-correlation-and-row-name-from-corr-matrix
beef_cor <- cor(beef_filter[,-1])</pre>
beef_cor_triangle <- setDT(melt(beef_cor))[Var1 != Var2, .SD[which.max(value)], keyby=Var1]</pre>
max(beef_cor_triangle$value)
# Protein and Fat
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datatemp<-beef_filter[,c(2, 3)]</pre>
colnames(datatemp)<-c("Protein", "Fat")</pre>
conf.reg<-function(xdata,alpha){</pre>
  if(ncol(xdata)!=2) stop("Only for bivariate normal")
  n<-nrow(xdata)</pre>
  xbar<-colMeans(xdata)
  S<-cov(xdata)
  es<-eigen(S)
  e1<-es$vec ** diag(sqrt(es$val))
  r1 < -sqrt(qf(alpha,2,n-2)) * sqrt(2*(n-1)/(n*(n-2)))
  theta<-seq(0,2*pi,len=250)
  v1<-cbind(r1*cos(theta), r1*sin(theta))
  pts < -t(xbar - (e1%*%t(v1)))
  plot(pts,type="l",main="Confidence Region for Bivariate Normal",xlab=colnames(xdata)[1],ylab=colnames
  segments(0,xbar[2],xbar[1],xbar[2],lty=2) # highlight the center
  segments(xbar[1],0,xbar[1],xbar[2],lty=2)
  th2<-c(0,pi/2,pi,3*pi/2,2*pi)
                                   #adding the axis
  v2<-cbind(r1*cos(th2), r1*sin(th2))
  pts2<-t(xbar-(e1%*%t(v2)))
  segments(pts2[3,1],pts2[3,2],pts2[1,1],pts2[1,2],lty=3)
  segments(pts2[2,1],pts2[2,2],pts2[4,1],pts2[4,2],lty=3)
}
conf.reg(datatemp,alpha=0.95)
# Compute 95% simultaneous confidence intervals for the two mean values
p<-ncol(datatemp)</pre>
n<-nrow(datatemp)</pre>
S<-cov(datatemp)</pre>
xbar<-colMeans(datatemp)
mu1.L=xbar[1]-sqrt(((n-1)*p/(n-p))*qf(0.95,p,n-p))*sqrt(S[1,1]/n)
mu1.U=xbar[1]+sqrt(((n-1)*p/(n-p))*qf(0.95,p,n-p))*sqrt(S[1,1]/n)
mu2.L=xbar[2]-sqrt(((n-1)*p/(n-p))*qf(0.95,p,n-p))*sqrt(S[2,2]/n)
mu2.U=xbar[2]+sqrt(((n-1)*p/(n-p))*qf(0.95,p,n-p))*sqrt(S[2,2]/n)
c(mu1.L,mu1.U)
c(mu2.L,mu2.U)
lines(c(mu1.L,mu1.L),c(0.53,mu2.U),lty=2,col=2,lwd=2)
lines(c(mu1.U, mu1.U), c(0.53, mu2.U), lty=2, col=2, lwd=2)
lines(c(0.49,mu1.U),c(mu2.L,mu2.L),lty=2,col=2,lwd=2)
lines(c(0.49,mu1.U),c(mu2.U,mu2.U),lty=2,col=2,lwd=2)
# Compute 95% Bonferroni confidence intervals for the two mean values
mu1.LB=xbar[1]-qt(0.05/(2*p),n-1,lower.tail=F)*sqrt(S[1,1]/n)
mu1.UB=xbar[1]+qt(0.05/(2*p),n-1,lower.tail=F)*sqrt(S[1,1]/n)
mu2.LB=xbar[2]-qt(0.05/(2*p),n-1,lower.tail=F)*sqrt(S[2,2]/n)
mu2.UB=xbar[2]+qt(0.05/(2*p),n-1,lower.tail=F)*sqrt(S[2,2]/n)
c(mu1.LB,mu1.UB)
c(mu2.LB,mu2.UB)
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```
# Plot the confidence intervals together with the confidence ellipse:
lines(c(mu1.LB, mu1.LB), c(0.53, mu2.UB), lty=3, col=3, lwd=2)
lines(c(mu1.UB, mu1.UB), c(0.53, mu2.UB), lty=3, col=3, lwd=2)
lines(c(0.49,mu1.UB),c(mu2.LB,mu2.LB),lty=3,col=3,lwd=2)
lines(c(0.49,mu1.UB),c(mu2.UB,mu2.UB),lty=3,col=3,lwd=2)
legend("topright",
       legend = c("95% Confidence Interval T^2", "95% Confidence Interval Bonferroni"),
       col = c("red", "green"), lty = 2, cex = 0.7)
#### two-sample Hotelling's T2 test -----
# now we perform the two-sample Hotelling T^2-test
HotellingsT2(beef_filter, vege_filter)
n<-c(nrow(beef_filter),nrow(vege_filter))</pre>
p<-ncol(beef_filter)</pre>
xmean1<-colMeans(beef_filter)</pre>
xmean2<-colMeans(vege_filter)</pre>
d<-xmean1-xmean2
S1<-var(beef filter)
S2<-var(vege_filter)
Sp<-((n[1]-1)*S1+(n[2]-1)*S2)/(sum(n)-2)
t2 <- t(d)%*%solve(sum(1/n)*Sp)%*%d
alpha < -0.05
cval <- (sum(n)-2)*p/(sum(n)-p-1)*qf(1-alpha,p,sum(n)-p-1)
# since we reject the null, we use the simultaneous confidence intervals
# to check the significant components
# simultaneous confidence intervals
alpha < -0.05
 wd < -sqrt(((n[1]+n[2]-2)*p/(n[1]+n[2]-p-1))*qf(1-alpha,p,n[1]+n[2]-p-1))*sqrt(diag(Sp)*sum(1/n)) \\
Cis<-cbind(d-wd,d+wd)</pre>
cat("95% simultaneous confidence interval","\n")
Cis <- as.data.frame(Cis)</pre>
colnames(Cis) <- c("Lower Bound", "Upper Bound")</pre>
Cis <- round(Cis, 4)</pre>
grid.table(Cis)
Cis[Cis[,1] < 0 \& Cis[,2] > 0,]
#Bonferroni simultaneous confidence intervals
wd.b \leftarrow qt(1-alpha/(2*p),n[1]+n[2]-2) *sqrt(diag(Sp)*sum(1/n))
Cis.b<-cbind(d-wd.b,d+wd.b)
cat("95% Bonferroni simultaneous confidence interval","\n")
Cis.b <- as.data.frame(Cis.b)</pre>
colnames(Cis.b) <- c("Lower Bound", "Upper Bound")</pre>
Cis.b <- round(Cis.b, 4)</pre>
grid.table(Cis.b)
Cis.b[Cis.b[,1] < 0 & Cis.b[,2] > 0,]
```

```
# both component-wise simultaneous confidence intervals
# do not contain 0, so they have significant differences.
### PCA
# subset only beef, remove sugar and labels
beef.pc <- princomp(beef_filter, cor=T)</pre>
# Showing the coefficients of the components:
summary(beef.pc,loadings=T)
# Showing the eigenvalues of the correlation matrix:
beef_lambda <- (beef.pc$sdev)^2</pre>
lambda_df <- as.data.frame(beef_lambda)</pre>
colnames(lambda_df) <- c("Lambda")</pre>
lambda_df$PC <- rownames(lambda_df)</pre>
rownames(lambda_df) <- NULL</pre>
lambda_df <- subset(lambda_df, select = c('PC', 'Lambda'))</pre>
grid.table(lambda df)
# A scree plot:
par(mfrow=c(1,2))
plot(1:(length(beef.pc$sdev)), (beef.pc$sdev)^2, type='b',
     main="Scree Plot", xlab="Number of Components", ylab="Eigenvalue Size")
### plot proportion version
plot(cumsum(beef_lambda/sum(beef_lambda)),
     main = 'Proportion of Variance from Cumulative Eigenvalues',
     ylab = 'Proportion of variance',
     xlab = 'Eigenvalue Index')
abline(h = .8)
legend("bottomright", legend = "80% Cutoff", lty = 1)
beef_loadings <- round(beef.pc$loadings[,1:3], 4)</pre>
grid.table(beef_loadings)
# What seems to be a reasonable number of PCs to use?
# Plotting the PC scores for the sample data in the space of the first two principal components:
set.seed(135)
beef_pc_index <- sample(1:nrow(beef.pc$scores))[1:50]</pre>
par(pty="s")
plot(beef.pc$scores[beef_pc_index,1],
     beef.pc$scores[beef_pc_index,2],
     xlab="PC 1", ylab="PC 2", type ='n', lwd=2)
text(beef.pc$scores[beef_pc_index,1],
     beef.pc$scores[beef_pc_index,2], cex=0.7, lwd=2)
biplot(beef.pc)
## explore the subgroups
ggbiplot(beef.pc, ellipse = TRUE, varname.adjust = 2) +
 ggtitle("Beef Biplot")
score1 = beef.pc$scores[,1]
score2 = beef.pc$scores[,2]
std.score1 = (score1 - mean(score1))/sd(score1)
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std.score2 = (score2 - mean(score2))/sd(score2)
grp1 = which((std.score1 > -3 & std.score1 < -1) & (std.score2 > 3 & std.score2 < 4.5))</pre>
beef_names[grp1]
grp2 = which((std.score1 > 4))
beef_names[grp2]
beef names[grepl("Beef, Australian, Wagyu", beef names)]
grp3 = which((std.score1 > 2.5 & std.score1 < 4) & (std.score2 > -1 & std.score2 < 1.5))
beef_names[grp3]
grp4 = which((std.score1 > -1 & std.score1 < 0) & (std.score2 > 2.5 & std.score2 < 3.1))
beef_names[grp4]
grp5 = which((std.score1 > -2 & std.score1 < 2) & (std.score2 > -3 & std.score2 < 2.5))
sort(beef_names[grp5])
grp6 = which((std.score2 > 4.5))
sort(beef_names[grp6])
a = apply(beef.pc$scores[grp6,], 2, median)
b = apply(beef.pc$scores[-grp6,], 2, median)
colnames(nutrition)
names(nut clean) [-c(1, 6)]
plot(a, type = 'b')
points(x = 1:22, y = b)
# https://stackoverflow.com/questions/2370515/how-to-get-row-index-number-in-r
beef_names[which(as.numeric(rownames(beef.pc$scores)) == '7516')]
beef_names[which(as.numeric(rownames(beef.pc$scores)) == '7519')]
### data visualization
### beef and vegetable biplot
X \leftarrow \text{nut\_clean}[,c(-1, -6)]
groupid <-nut_clean[,1]</pre>
X.pca <- princomp(~.,data=X)</pre>
summary(X.pca,loadings=TRUE)
plot(X.pca$scores[,1],X.pca$scores[,2],xlab="PC1",ylab="PC2",
     pch=rep(1:2,n),col=groupid,main="Iris data")
ggbiplot(X.pca, alpha = 0.1, ellipse = TRUE, groups = groupid) + xlim(-2, 0.5) + ylim(-2, 1)
legend("bottomleft",legend=levels(groupid),pch=1:3,col=1:3,cex=0.7)
pc1 = X.pca$scores[,1]
pc2 = X.pca$scores[,2]
spc1 = (pc1-mean(pc1))/sd(pc1)
spc2 = (pc2-mean(pc2))/sd(pc2)
outlier1 = which.min(spc2)
outlier2 = which(spc1 < -1)</pre>
# redo pc without outliers
new_groupid <-nut_clean[-c(outlier1, outlier2),1]</pre>
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levels(new_groupid) = c("Beef", "Vegetable")
new_X.pca <- princomp(~.,data=X[-c(outlier1, outlier2),])</pre>
ggbiplot(new_X.pca, ellipse = TRUE, groups = new_groupid) +
  ggtitle("Beef and Vegetable Biplot")
### LDA
par(mar=c(4,4,2,1))
# http://www.cookbook-r.com/Manipulating_data/Randomizing_order/
set.seed(135)
comb_filter <- rbind(beef_filter[sample(1:nrow(beef_filter))[1:50], c(1,18)],</pre>
                      vege_filter[sample(1:nrow(vege_filter))[1:50], c(1,18)])
comb_filter$Label <- c(rep('Beef', 50), rep('Vegetables', 50))</pre>
lda.obj<-lda(Label~.,data=comb filter,prior=c(1,1)/2)</pre>
plda<-predict(object=lda.obj,newdata=comb_filter)</pre>
#plot the decision line
gmean <- lda.obj$prior %*% lda.obj$means</pre>
const <- as.numeric(gmean %*%lda.obj$scaling)</pre>
slope <- - lda.obj$scaling[1] / lda.obj$scaling[2]</pre>
intercept <- const / lda.obj$scaling[2]</pre>
#Plot decision boundary
plot(comb_filter[,1:2], pch=rep(c(18,20), each=50), col=rep(c(2,4), each=50),
     main = 'LDA of Energy and Magnesium from Beef and Vegetables')
\# plot(salmon[,2:3],pch=rep(c(18,20),each=50),col=rep(c(2,4),each=50))
abline(intercept, slope)
# legend("topright", legend=c("Alaskan", "Canadian"), pch=c(18,20), col=c(2,4))
legend("topright",legend=c("Beef","Vegetables"),pch=c(18,20),col=c(2,4))
#A Stacked Histogram of the LDA Values
ldahist(data = plda$x[,1], g=comb_filter$Label, main = 'Stacked Histogram of LDA Values')
# Confusion matrix
table(comb_filter$Label,plda$class)
```