

## HW5. for Multivariate Statistics II

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### Chapter 8. Correspondence Analysis(CRA) : Multiple CRA

The following data are the three-way table for the car accident record on the highway depending on whether the safety belt is worn.

[Table 1]

안전벨트	이탈	사망	
		예	아니오
착용	예	14	1
	아니오	483	411
미착용	예	497	462
	아니오	100	157

(1) Explain category variables and classification variables in [Table 1].

categorical variables : 안전벨트, 이탈, 사망

classification variables : 착용/미착용, 예/아니오, 예/아니오

(2) Make a data form of classification variables from [Table 1].

> accident

이탈 사망 안전벨트 Freq

```
1 yes yes 착용 14
2 no yes 착용 483
3 yes no 착용 1
4 no no 착용 411
5 yes yes 미착용 497
6 no yes 미착용 100
7 yes no 미착용 462
8 no no 미착용 157
```

### (3) Apply multiple CRA to the data in (2).

```
> mjca(table)
```

Eigenvalues:

	1	2
Value	0.147308	2e-06
Percentage	74.75%	0%

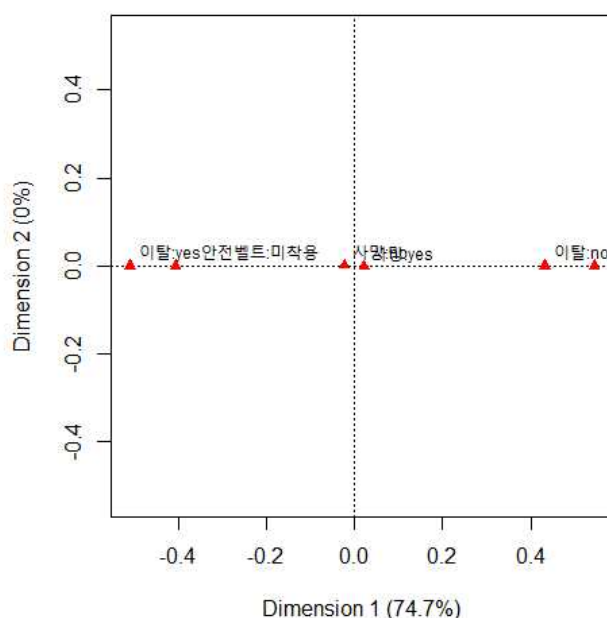
Columns:

	이탈:no	이탈:yes	사망:no	사망:yes	안전벨트:미착용	안전벨트:착용
Mass	0.180549	0.152784	0.161725	0.171608	0.190745	0.142588
ChiDist	0.669315	0.790946	0.595732	0.561426	0.629616	0.842259
Inertia	0.080883	0.095581	0.057396	0.054091	0.075615	0.101152
Dim. 1	1.124677	-1.329058	-0.061194	0.057670	-1.059518	1.417352
Dim. 2	0.114342	-0.135121	1.779101	-1.676648	0.034830	-0.046593

Dim1's GOF is 74.75%. Dim2's GOF is 0%. It can be shown the associations between classification variables by Dim1.

### (4) Describe some associations between categorical variables in the multiple CRA map.

**MCRA : 3-ways contingency table**



Relative positions of the points : similarities and differences among rows and columns categories.

"이탈:yes" and "안전벨트:미착용", "사망:no" are related to each other because they have same direction.

"이탈:no" and "안전벨트:착용", "사망:yes" are related to each other because they have same direction.

We can see that wearing the seat belt is related to lane departure and death.

## 2. [Exercise 8.10] (p.501)

### (1) Apply multiple CRA

Burt matrix

> B

	교육1	교육2	교육3	교육4	적성1	적성2	적성3	적성4	적성5	직업1	직업2	직업3	직업4
[1,]	1040	0	0	0	215	281	372	128	44	239	6	1	794
[2,]	0	1108	0	0	208	285	386	176	53	309	11	7	781
[3,]	0	0	1237	0	138	284	446	238	131	233	70	12	922
[4,]	0	0	0	968	83	197	385	186	117	53	199	215	501
[5,]	215	208	138	83	644	0	0	0	0	122	30	20	472
[6,]	281	285	284	197	0	1047	0	0	0	226	51	66	704
[7,]	372	386	446	385	0	0	1589	0	0	306	115	96	1072
[8,]	128	176	238	186	0	0	0	728	0	130	59	38	501
[9,]	44	53	131	117	0	0	0	0	345	50	31	15	249
[10,]	239	309	233	53	122	226	306	130	50	834	0	0	0
[11,]	6	11	70	199	30	51	115	59	31	0	286	0	0
[12,]	1	7	12	215	20	66	96	38	15	0	0	235	0
[13,]	794	781	922	501	472	704	1072	501	249	0	0	0	2998

Coordinate

> Cb2

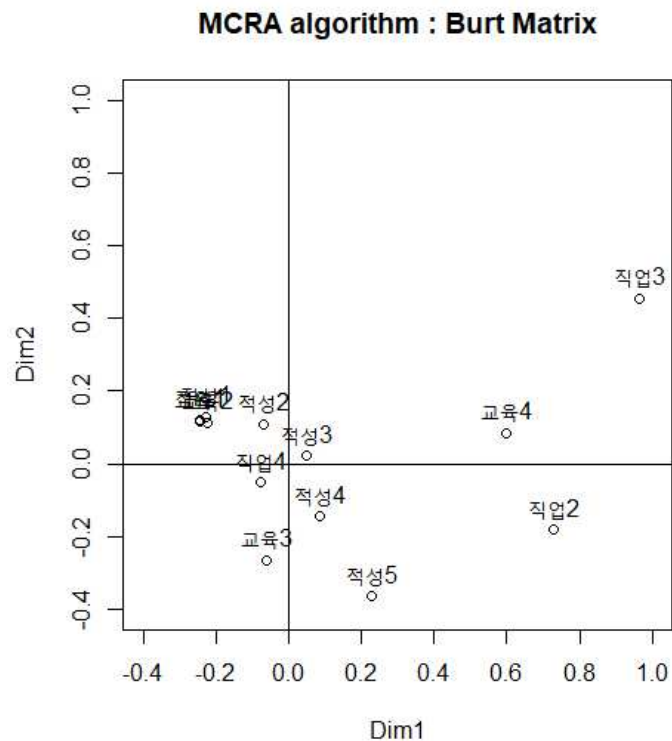
	[.1]	[.2]
교육1	-0.24275450	0.11799191
교육2	-0.22574360	0.11321747
교육3	-0.06306682	-0.26602887
교육4	0.59979570	0.08359625
적성1	-0.22720684	0.13062126
적성2	-0.06906785	0.10974704
적성3	0.04903160	0.02106607
적성4	0.08463271	-0.14601324
적성5	0.22930848	-0.36580175
직업1	-0.24599999	0.12024326
직업2	0.72887452	-0.18158516
직업3	0.96369167	0.45393605
직업4	-0.07663831	-0.05170930

GOF

> rbind(round(lam, 3), round(fit, 3))

	[.1]	[.2]	[.3]	[.4]	[.5]	[.6]	[.7]	[.8]	[.9]	[.10]	[.11]	[.12]	[.13]
[1,]	0.273	0.148	0.126	0.116	0.112	0.109	0.104	0.093	0.082	0.024	0	0	0
[2,]	23.039	12.500	10.620	9.779	9.425	9.160	8.799	7.810	6.885	1.983	0	0	0

(2) What is GOF of multiple CRA map. And interpret that.



Dim1's GOF is 23.039%. Dim2's GOF is 12.5%.

Relative positions of the points : similarities and differences among rows and columns categories.

"적성3~5" and "교육4", "직업2~3" are related to each other because they have same direction.

We can see that high aptitude levels is related to high educational levels. In this situation, these are related to professional and educational professions.

### 3. Consider the [Data 8.5.1] (view.txt) Economic Outlook and Economic Policy Preference Survey Data.

#### (1) Apply multiple CRA to the data

```
> head(view)
```

	sex	age	income	view	policy
1	1	2	2	2	2
2	1	1	1	1	2
3	2	1	1	1	1
4	1	2	2	2	2
5	2	1	1	2	2
6	1	2	1	2	1

```
> mjca(view)
```

Eigenvalues:

	1	2
Value	0.178447	0.001018
Percentage	85.18%	0.49%

Columns:

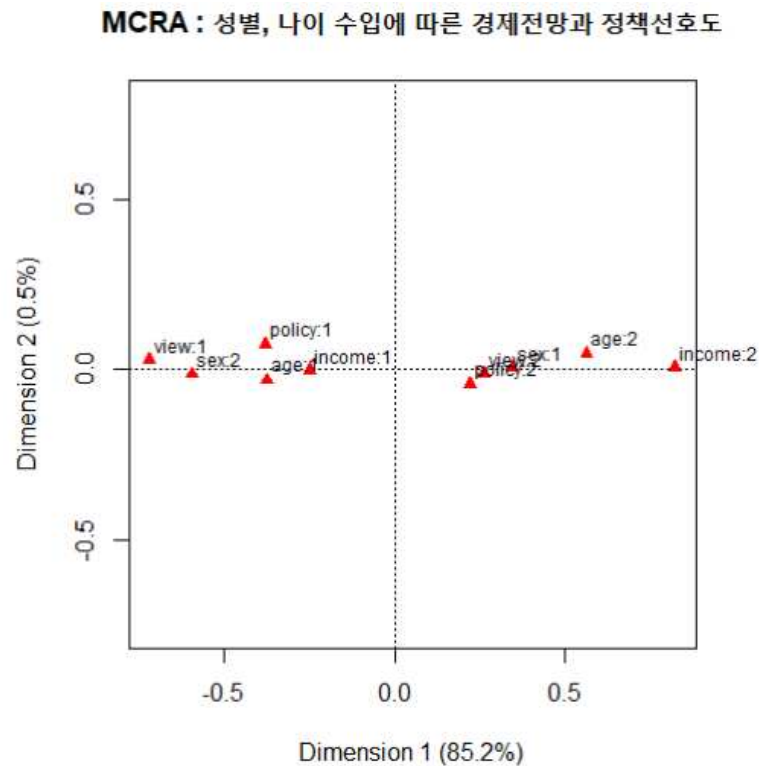
	sex:1	sex:2	age:1	age:2	income:1	income:2
Mass	0.126667	0.073333	0.120000	0.080000	0.153333	0.046667
ChiDist	0.465567	0.804162	0.528152	0.792229	0.341826	1.123143
Inertia	0.027455	0.047423	0.033473	0.050210	0.017916	0.058868
Dim. 1	0.815374	-1.408374	-0.888077	1.332115	-0.591738	1.944283
Dim. 2	0.230783	-0.398625	-0.984092	1.476138	-0.068827	0.226146

	view:1	view:2	policy:1	policy:2
Mass	0.053333	0.146667	0.073333	0.126667
ChiDist	0.998656	0.363147	0.718208	0.415804
Inertia	0.053190	0.019342	0.037827	0.021900
Dim. 1	-1.701435	0.618704	-0.896624	0.519098
Dim. 2	0.906130	-0.329502	2.329320	-1.348553

Dim1's GOF is 85.18%. Dim2's GOF is 0.49%. It can be shown the associations between classification variables by Dim1.

(2) Describe some associations between categorical variables in the multiple CRA map.



Relative positions of the points : similarities and differences among rows and columns categories.

"age:1" and "income:1", "sex:2", "policy:1", "income:1" are related to each other because they have same direction.

"age:2" and "income:2", "sex:1", "policy:2", "income:2" are related to each other because they have same direction.

We can see that women who have under the age of 30 and under the income of 10 million won are optimistic about the economic outlook and the government's economic policies. In contrast, men who have over the age of 30 and over the income of 10 million won are pessimism about the economic outlook and the government's economic policies.

```

setwd("E:/학교/2020 2학기 정호재/다변량통계학2/실습/20201117/Rdata")
Freq <-array(c(14,483,1,411,497,100,462,157))
이탈<-c("yes", "no")
사망<- c("yes", "no")
안전벨트<-c("착용","미착용")
data<-expand.grid(이탈=이탈, 사망=사망, 안전벨트=안전벨트)
accident<-cbind(data, Freq)
accident

table<-xtabs(Freq~이탈+사망+안전벨트, data=accident)
table

#install.packages("ca")
library(ca)
mjca(table)
plot(mjca(table),main="MCRA : 3-ways contingency table")
#####
apitude_burt<-read.table("apitude.txt",header=T)

B<-as.matrix(apitude_burt)
B
P <- B / sum(B)
round(P,3)
cm <- apply(P, 2, sum)
Dc<-diag(1/sqrt(cm))
eP <- cm %*% t(cm)
Y <- (P - eP) / sqrt(eP)
round(Y,3)

# Singular Value Decomposition
svd.Y<-svd(Y)
V<-svd.Y$v
Dl<-diag((svd.Y$d)^2)

lam<-(svd.Y$d)^2
fit<-lam/sum(lam)*100
rbind(round(lam, 3),round(fit, 3))

Cb<- Dc%*%V%*%Dl
rownames(Cb)<-colnames(apitude_burt)

Cb2<-Cb[, 1:2]

```

Cb2

```
lim<-range(pretty(Cb))
```

```
plot(Cb2, xlab="Dim1", ylab="Dim2", xlim=lim, ylim=lim, main="MCRA algorithm : Burt Matrix")
text(Cb2, colnames(apitude_burt), pos=3, col=1)
abline(v=0, h=0)
```

```
#####
```

```
view<-read.table("view.txt",header=T)
```

```
view<-view[,-1]
```

```
head(view)
```

```
library(ca)
```

```
mjca(view)
```

```
par(pty="s")
```

```
plot(mjca(view), main="MCRA : 성별, 나이 수입에 따른 경제전망과 정책선호도")
```