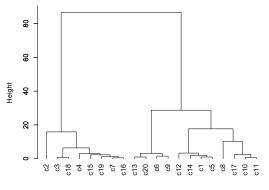
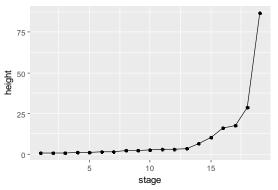
Ch2

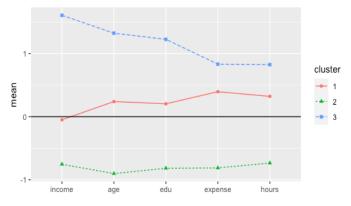
Hierarchical clustering

scale(x) - standardization (col by col)
d<-dist(x, method = "euclidean")^2 - distance
fit.hc<-hclust(d, method =
"complete"/"ward.D2"/"single"/"ward.D"/"average"/"centroid")
dendrogram or distance plot</pre>





cluster<-cutree(fit.hc, k)
centers<-aggregate(x = data, by = list(cluster = cluster), FUN = mean)
profile plot</pre>



K-means

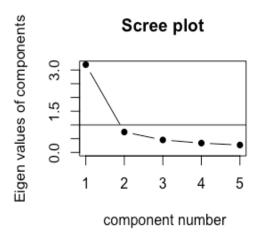
fit.km<-kmeans(x = data, centers = num/centers, algorithm = "Hartigan-Wong"/"Lloyd"/"Forgy"/"MacQueen") (set.seed(123) if num)

Ch3

Principal component method

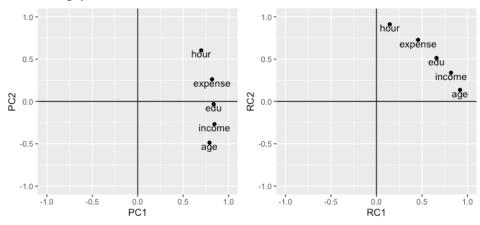
(fit.pc.full<-principal(data, nfactors = num, rotate = "none"))

scree(data, factor = F) (correlation matrix)

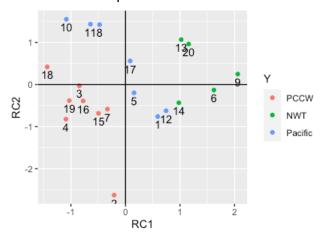


fit.pc.rot<-principal(data, nfactors = 2, rotate =
"varimax"/"quartimax"/"promax") (correlation matrix or covar = T)
(regression by default)</pre>

loading plots before and after factor rotations



factor score plot



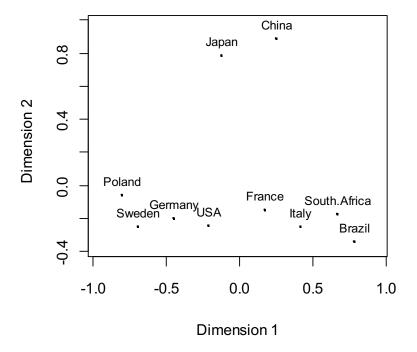
Maximum likelihood method

fit.pc.ml<-fa(data, nfactors = 2, rotate = "none"/"varimax"/"quartimax", fm = "ml") (correlation matrix or covar = T) (regression by default)

Ch4

Dissimilarity (distance) matrix

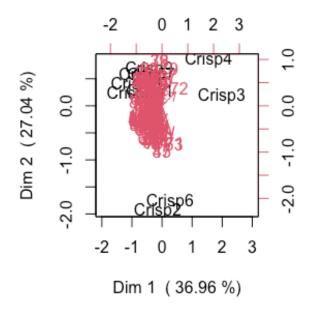
fit.dc<-mds(delta, ndim=2, type="interval"/"ratio"/"ordinal") (ordinal=nonmetric MDS, compare the stress) # perceptual map



Preference matrix

fit.pr<-cmdpref(pref, ndim = 2, monotone = T/F) (Rows as objects and columns as subjects. Small value for less preferable and large value for more preferable.)

joint plot for object coordinates and subject ideal vectors



**Disimilarity matrices provided by several individuals (INDSCAL model) **

fit.in<-indscal(delta, ndim=2, type="interval"/"ratio"/"ordinal") (ordinal=nonmetric MDS, compare the stress) # perceptual map for objects and weighting plot for subjects

