

## 1<sup>st</sup> Step: data pre-processing

Raw dataset (input)

$I_i$ : individual,  $i = 1, \dots, n$

$V_j$  = categorical variable,  $j = 1, \dots, m$

	$V_1$	$V_2$	$\dots$	$V_j$	
$I_1$					
$I_2$					
$\vdots$					
$I_i$					

Multiple Correspondence Analysis

Multiple Correspondence Analysis result  
 $PC_j$  = Principal Component (continuous variable),  $j = 1, \dots, m$

	$PC_1$	$PC_2$	$\dots$	$PC_j$	
$I_1$					
$I_2$					
$\vdots$					
$I_i$					

Threshold  
Inertia percentage (at least 50%)

Dataset for Hierarchical clustering (output)

Selected  $PC, k < j$   
 $PC_k, k = 1, \dots, p$

	$PC_1$	$PC_2$	$\dots$	$PC_k$	
$I_1$					
$I_2$					
$\vdots$					
$I_i$					

## 2<sup>nd</sup> Step: Hierarchical Cluster Analysis

Dataset for Hierarchical clustering (input)

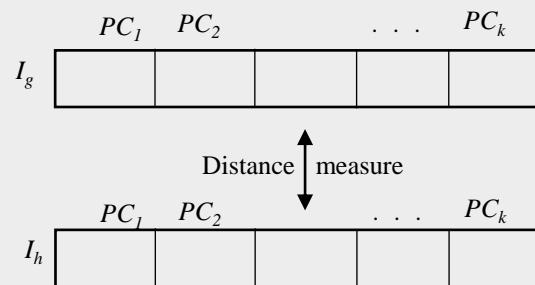
Selected  $PC, k < j$   
 $PC_k, k = 1, \dots, p$

	$PC_1$	$PC_2$	$\dots$	$PC_k$	
$I_1$					
$I_2$					
$\vdots$					
$I_i$					

Hierarchical Cluster Analysis

Starting agglomerative process:  
distance between  $p$ -dimensional observations

(Euclidian distance)



Ward's algorithm  
Agglomeration of clusters

Hierarchical Cluster Analysis result

Hierarchical tree (output)

