## **Correspondence Analysis (CA)**

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Correspondence analysis provides a graphical method of exploring the relationship between categorical variables in a contingency table.

## In summary:

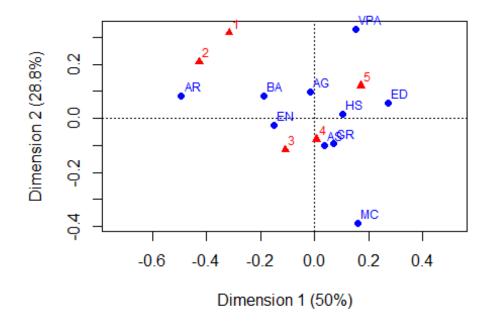
- We are looking for similarity (or short distance) between row variables and column variables.
- In the CA, the distance between column variables of a contingency table is defined as the weighted sum of squared difference between proportions (probabilities) in the columns. Same process for finding the distance between row variables.

(This process is explained in Video Lecture 15)

- In the CA, column points that are close together represent column categories with similar profiles (conditional distributions) across columns. Same interpretation for row points.
- If the angle between the row points A and column points B is less than 90-degree angle P(A|B)>P(A), which means the likelihood of A's occurrence given B is larger than the likelihood of A's occurrence in the entire sample space.
- If the angle between the row points A and column points, B is more than 90-degree angle P(A|B)<P(A), which means the likelihood of A's occurrence given B is less than the likelihood of A's occurrence in the entire sample space.

## **Example:** TTU College Graduation Survey (2002)

```
grad <- read.csv("http://westfall.ba.ttu.edu/isqs6348/Rdata/pgs.csv", header</pre>
= T)
nrow(grad)
## [1] 2002
ncol(grad)
## [1] 124
attach(grad)
#make a contingency table for college vs general rating
tbl = table(COL, FacTeaching)
tbl
        FacTeaching
##
## COL
                           5
           1
              2
                  3
                      4
           4 15 26 78 56
##
    AG
##
    AR
           3
              4
                   6
                      16
                           4
##
    AS
          12 24 124 290 171
           9 28 44 116
##
    BA
                         66
##
    DUAL
           0
              0
                  2
                       0
                          0
              6 26 113 93
##
    ED
           3
##
    EN
           5
              36 65 168 86
##
           0
              3
                  8 27 15
    GR
##
    HS
           1
              5 17 41 33
                      25
##
    MC
           0
              0
                  3
                          6
    VPA
           4
              7 10
##
                     37 44
library(ca)
grad.ca <- ca(tbl <- tbl[-5, ]) # removing dual</pre>
plot(grad.ca)
```



## How to interpret?

 The distribution of AS and GR among ratings is more similar than the distribution of VPA among ratings.