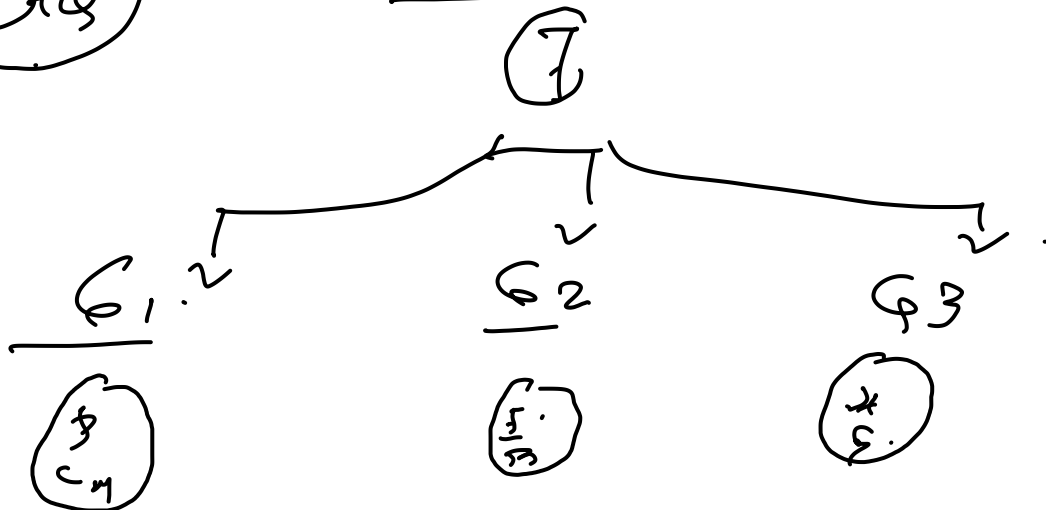
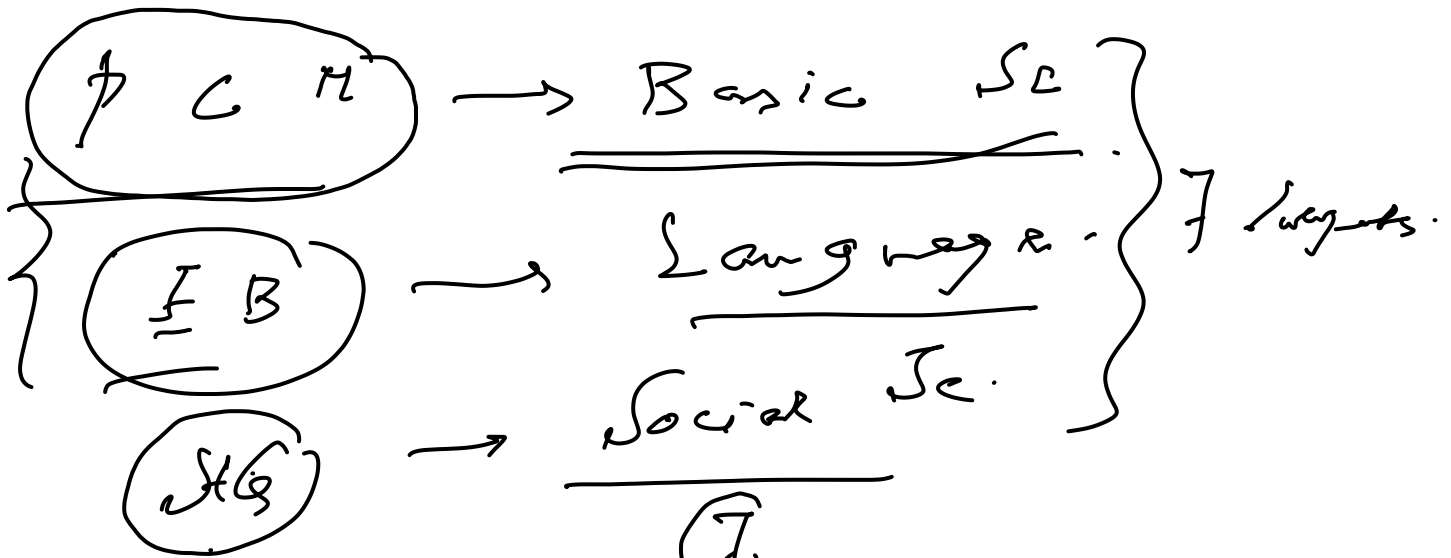


PLA

25.02.2024

Multivariate

- ① Dimension
- ② Sorting & grouping
- ③ Dependence among the variables $R \times C$
- ④ Predictive Time.
- ⑤ Prob. App. Construction & Testing



Trend:

(a) $\begin{matrix} \nearrow \\ \searrow \end{matrix}$

7 variables.

	D	C	M	A	E	S	G
1	✓	✓	✓	✓	✓	✓	✓
2	✓	✓	✓	✓	✓	✓	✓
⋮	✓	✓	✓	✓	✓	✓	✓
100	✓	✓	✓	✓	✓	✓	✓

100 Rows

7 Cols

100 x 7

Step 1: Find mean & S.D. for each variable.

Step 2 (Cur Imp): Find the Covariance matrix.

Step 3: Find the Correlation matrix
because it is very simple to understand.

Covariance matrix

$$\begin{matrix} & \begin{matrix} x_1 & x_2 & x_3 \end{matrix} \\ \begin{matrix} x_1 \\ x_2 \\ x_3 \end{matrix} & \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \end{matrix} \quad \begin{matrix} \text{Diagonal matrix.} \\ \left. \begin{matrix} \sqrt{r_1} \\ \sqrt{r_2} \\ \sqrt{r_3} \end{matrix} \right\} \begin{matrix} C_{12} = C_{21} \\ C_{13} = C_{31} \\ C_{23} = C_{32} \end{matrix} \end{matrix}$$

when there are 3 variables.

You have to compute

$$(3C_1) = 3^2 \rightarrow \text{variables}$$

Q. To compute Covariance = $(3C_2) = 3$ ✓

$$\begin{matrix} x_1 \\ x_2 \\ x_3 \end{matrix} \begin{pmatrix} \cancel{v(x_1)} & \boxed{c_{12}} & \boxed{c_{13}} \\ c_{21} & \cancel{v(x_2)} & \boxed{c_{23}} \\ c_{31} & c_{32} & \cancel{v(x_3)} \end{pmatrix} \begin{matrix} 3 \times 3 \end{matrix}$$

$$\left. \begin{matrix} v(x_1) = 1 \\ v(x_2) = 5 \\ v(x_3) = 2 \end{matrix} \right\} \begin{matrix} c_{12} = c_{21} = -2 \\ c_{13} = c_{31} = 0 \\ c_{23} = c_{32} = 0 \end{matrix} \right\}$$

$$\begin{pmatrix} 1 & -2 & 0 \\ -2 & 5 & 0 \\ 0 & 0 & 2 \end{pmatrix} \begin{matrix} \\ \\ 3 \times 3 \end{matrix}$$

$$\begin{matrix} x_1 & x_2 & x_3 & x_4 \end{matrix}$$

$$\begin{matrix} \text{no of variables} = (4C_1) = 4 \\ \text{no of } \underline{\text{covariances}} = (4C_2) = 6 \end{matrix}$$

$$\begin{pmatrix} \cancel{v(x_1)} & c_{12} & c_{13} & c_{14} \\ c_{21} & \cancel{v(x_2)} & c_{23} & c_{24} \\ c_{31} & c_{32} & \cancel{v(x_3)} & c_{34} \\ c_{41} & c_{42} & c_{43} & \cancel{v(x_4)} \end{pmatrix} \begin{matrix} \\ \\ \\ 4 \times 4 \end{matrix}$$

$$r(x_1) = 1 \quad r(x_2) = 5 \quad r(x_3) = 3 \quad \checkmark$$

$$r(x_4) = 6$$

$$c_{12} = 2$$

$$c_{13} = 0$$

$$c_{14} = -2$$

$$c_{23} = 5$$

$$c_{24} = -1$$

$$c_{34} = 3$$

$$\checkmark 4 + 6 = \underline{(10)}$$

$$\begin{pmatrix} 1 & 2 & 0 & -2 \\ 2 & 5 & 5 & -1 \\ 0 & 5 & 3 & 3 \\ -2 & -1 & 3 & 6 \end{pmatrix} \quad \underline{4 \times 4}$$

7x7 Correlation

$$\begin{matrix} 1 & r_{12} & r_{13} & r_{14} & r_{15} & r_{16} & r_{17} \\ & 1 & & & & & \\ & & 1 & & & & \\ & & & 1 & & & \\ & & & & 1 & & \\ & & & & & 1 & \\ & & & & & & 1 \end{matrix}$$

variable.

$$c_{7(2)} = \frac{7 \times 6}{2} = \underline{(21)} \checkmark$$

P
 C
 M.

no. of correlation

$$= (3C_2) = 3$$

P
 C

G₂

$$= (2C_2) = 1$$

P
 C
 M

G₃

$$= (2C_2) = 1$$

7



3 Dimension Reduction

Banking

Element

Red.

0

3

?

P
 C
 M

G₁. How many members in the group?

In G₁. no of members = 3.

In G_2

no of members = 2

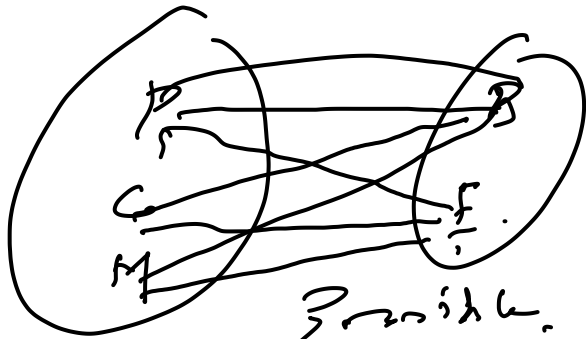
17 " " = 2.

In G_3

Association among
 G_1 & G_2 .

occur between

G_1 & G_3 G_2 & G_3 .



possible
no of association

= 6

G_1 & G_3

6

(7₂)

G_2 & G_3
4

6 + 6 + 4
= 16

$$21 = 5 + 16$$

Total

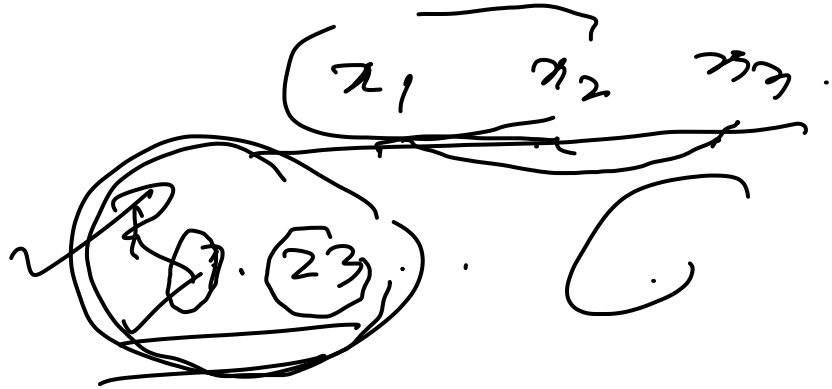
Correlation among

x_1, y_1 & z_1

(x, y)

x_2

y_2



$$x_1 = DV$$

$$x_2 \text{ \& } x_3 = IV$$

Three types of model

- (1) Independent
 - (2) Hypothetical
 - (3) Saturated
- (1) \rightarrow (2)

$x_1 \quad x_2 \quad x_3$
They are completely ind.

$$r_{12} = 0 \quad r_{13} = 0 \quad r_{23} = 0$$

$$I = \begin{matrix} & x_1 & x_2 & x_3 \\ \begin{matrix} x_1 \\ x_2 \\ x_3 \end{matrix} & \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \end{matrix}$$

Identity or unit matrix

$$\begin{vmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{vmatrix} = 1 \quad |I| = 1$$

$$\begin{matrix} x_1 \\ x_2 \\ x_3 \end{matrix} \begin{pmatrix} 1 & 0.99 & 0.98 \\ 0.99 & 1 & 0.90 \\ 0.97 & 0.90 & 1 \end{pmatrix} \quad \underline{\underline{3 \times 3}}$$

$$r_{12} = 0.99$$

$$r_{13} = 0.97$$

$$r_{23} = 0.90$$

$$\boxed{|A| \rightarrow 0} \quad \swarrow$$

If p -value > 0.05
 \Rightarrow No linear association between two variables.

$$\text{p-value} \leq 0.05$$

$$\left\{ \begin{array}{l} 9 \text{ variables} = \text{Iv.} \\ 1 \text{ Dr.} \end{array} \right\}$$

To get a model
 To improve R^2 .

⑨

⑦✓

Step wise Regression

✓

②

$$R^2 = 0.743 \checkmark$$

✓ ⑨

$$R^2 = 0.791 \rightarrow \text{Maximen} =$$

①

$$R^2 = 0.49 \rightarrow \text{Minimen}$$

②

$$R^2 = 0.633 \checkmark$$

Min
0.49
(1)

0.633
(2)

0.743
(2)

Max
0.791
(3)

$\phi(R)$ R^2

Statistically significant

02.03.2024

Data matrix

1 DV
2 IV

Regression

R^2 , AD , R^2

VIF

VIF for
ind variable
are very high
amongst ind.

Multi Collinearity

High correlation
between
variables.

More variables
occur this high

VIF. value
Regression failed.

Input 3 / IV: → Apply
PCA

Factor Analysis.

- Step ① For the theoretical concept we calculate correlation matrix.
- ② For applied purpose, we will correlation matrix.

	X_1	X_2	X_3	✓
1	0	0	0	
2	0	0	0	
100	0	0	0	

\leftarrow mean of X_1
 \leftarrow " X_2
 \leftarrow " X_3

\leftarrow $\begin{pmatrix} p_1 \\ p_2 \\ p_3 \end{pmatrix}$

$\frac{100 \times 3}{3 \times 10}$
 $\sqrt{a_1}$ C_{12}
 $\sqrt{a_2}$ C_{13}
 $\sqrt{a_3}$ C_{23}

	σ_{12}	σ_{13}	σ_{23}
	-2	0	
		5	0
			2

$\frac{2}{3 \times 3}$

where $\text{Row} = (01 \Rightarrow) \text{Diagonal}$
 $\text{Row} \neq (01 \Rightarrow) \text{off diagonal}$

$$\rho = \begin{pmatrix} 1 & r_{12} = -0.893 & 0 \\ -0.893 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}_{3 \times 3}$$

$$r_{12} = \frac{C_{12}}{\rho_1 \rho_2} = \frac{-2}{\sqrt{1} \sqrt{5}}$$

$$= \frac{-2}{2.24}$$

$$= \underline{-0.893}$$

Step (3) . We can apply PCA.
[Principle Component Analysis]
 or Covariance matrix
 or we can apply or
 Correlation matrix

For understanding PCA. you
have to know polynomial
& roots.

✓ $2x + 3 = 0 \Rightarrow$ 1st degree polynomial
 $x = -3/2 \Rightarrow (-3/2)$ root

LHS = $2x + 3$

RHS = 0.

= $2(-3/2) + 3 = 0 = \underline{\text{RHS}}$

$x^2 - 5x + 6 = 0$

(Two) ✓

$(x-2)(x-3) = 0$

$\therefore x = 2 \text{ or } 3$. Roots

Degree of Polynomial = $2/3/4$
 no of roots = $2/3/4$

P C M. ~~13~~ E H G. (7).



No of variables = 7.
 We can construct a vector
 degree polynomial

We get vector (roots).

Eigen values.

$$\tilde{Z} = \begin{pmatrix} 1 & -2 & 0 \\ -2 & 5 & 0 \\ 0 & 0 & 2 \end{pmatrix}_{3 \times 3}.$$

✓
 ✓
 ✓
 ✓
 P (A.

Diagonal -

of How much variance
 can be explained?

$$|\Sigma - \tau I| = 0 \Rightarrow \text{Polynomial}$$

$$\begin{vmatrix} 1 & -2 & 0 \\ -2 & 5 & 0 \\ 0 & 0 & 2 \end{vmatrix} - \tau \begin{vmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{vmatrix} = 0$$

$$\begin{vmatrix} \overset{a_{11}}{1-\tau} & \overset{a_{12}}{-2} & \overset{a_{13}}{0} \\ -2 & 5-\tau & 0 \\ 0 & 0 & 2-\tau \end{vmatrix} = 0$$

$$(1-\tau) \left[(5-\tau)(2-\tau) \right] - (-2) \left\{ -2(2-\tau) \right\} = 0$$

$$(1-\tau)(5-\tau)(2-\tau) - 4(2-\tau) = 0$$

3rd degree Polynomial

$$\left. \begin{aligned} \tau_1 &= 5.83 \\ \tau_2 &= 2.00 \\ \tau_3 &= 0.17 \end{aligned} \right\} \begin{aligned} \tau_1 + \tau_2 + \tau_3 &= 8 \\ &= \tau(\tau_1) + \tau(\tau_2) + \tau(\tau_3) \end{aligned}$$

$$\begin{vmatrix} \overset{a_{11}}{1-\tau} & \overset{a_{12}}{-2} & \overset{a_{13}}{0} \\ \overset{a_{21}}{-2} & \overset{a_{22}}{5-\tau} & \overset{a_{23}}{0} \\ \overset{a_{31}}{0} & \overset{a_{32}}{0} & \overset{a_{33}}{2-\tau} \end{vmatrix} = 0$$

$$\Delta = a_{11} (a_{22} a_{33} - a_{23} a_{32}) - a_{12} (a_{21} a_{33} - a_{23} a_{31}) + a_{13} (a_{21} a_{32} - a_{22} a_{31})$$

In case of correlation matrix of three variables -

$$\begin{pmatrix} 1 & -0.89 & 0 \\ -0.89 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \quad 3 \times 3$$

$\lambda_1, \lambda_2, \lambda_3 = 3$

Objective of PCA.

Dimension Reduction.

Why? No. of Eigen values > 1 ~~homogeneous~~
 = no. of ~~groups~~

§

$$\begin{array}{l}
 \lambda_1 > 1 \\
 \lambda_2 > 1 \\
 \lambda_3 < 1 \\
 \lambda_4 < 1 \\
 \lambda_5 < 1 \\
 \lambda_6 < 1
 \end{array}
 \left. \vphantom{\begin{array}{l} \lambda_1 \\ \lambda_2 \\ \lambda_3 \\ \lambda_4 \\ \lambda_5 \\ \lambda_6 \end{array}} \right\}
 \begin{array}{l}
 \text{Two Eigen} \\
 \text{values} > 1 \\
 \text{Four Eigen} \\
 \text{values} < 1 \\
 \lambda_1 + \lambda_2 + \lambda_3 + \lambda_4 + \lambda_5 \\
 + \lambda_6 = 6
 \end{array}$$

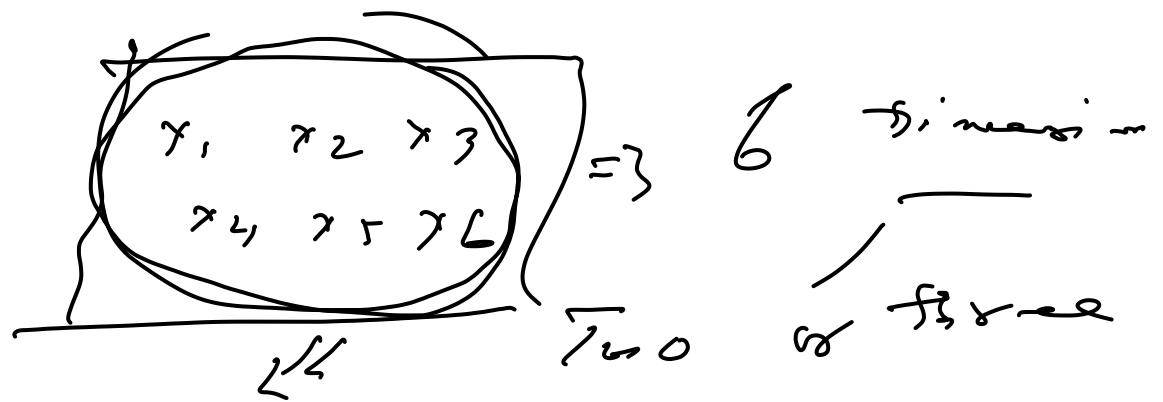
Contribution of 1st Eigen value is 32.53%
 Contribution of 2nd Eigen value is 27.1%

Contribution of other four eigen values (Balancing figure).

$$\begin{array}{l}
 \lambda_1 = 3.253 \quad \left\{ \begin{array}{l} 3.253/6 = 54.2\% \\ 1.654/6 = 27.6\% \end{array} \right. \\
 \lambda_2 = 1.654
 \end{array}$$

Total Contribution by Two eigen values = 81.1% Explained
 by the Two eigen values.

Remaining (100 - 81.1) = 18.9% is
 explained by the four eigen values < 1.



Correlation matrix $\rightarrow 6 \times 6$

PCA. λ ^{help of} determined

$\lambda_1, \lambda_2, \dots, \lambda_6$

$\lambda_1 > \lambda_2 > \dots$

How much explained by 1st Eigen value

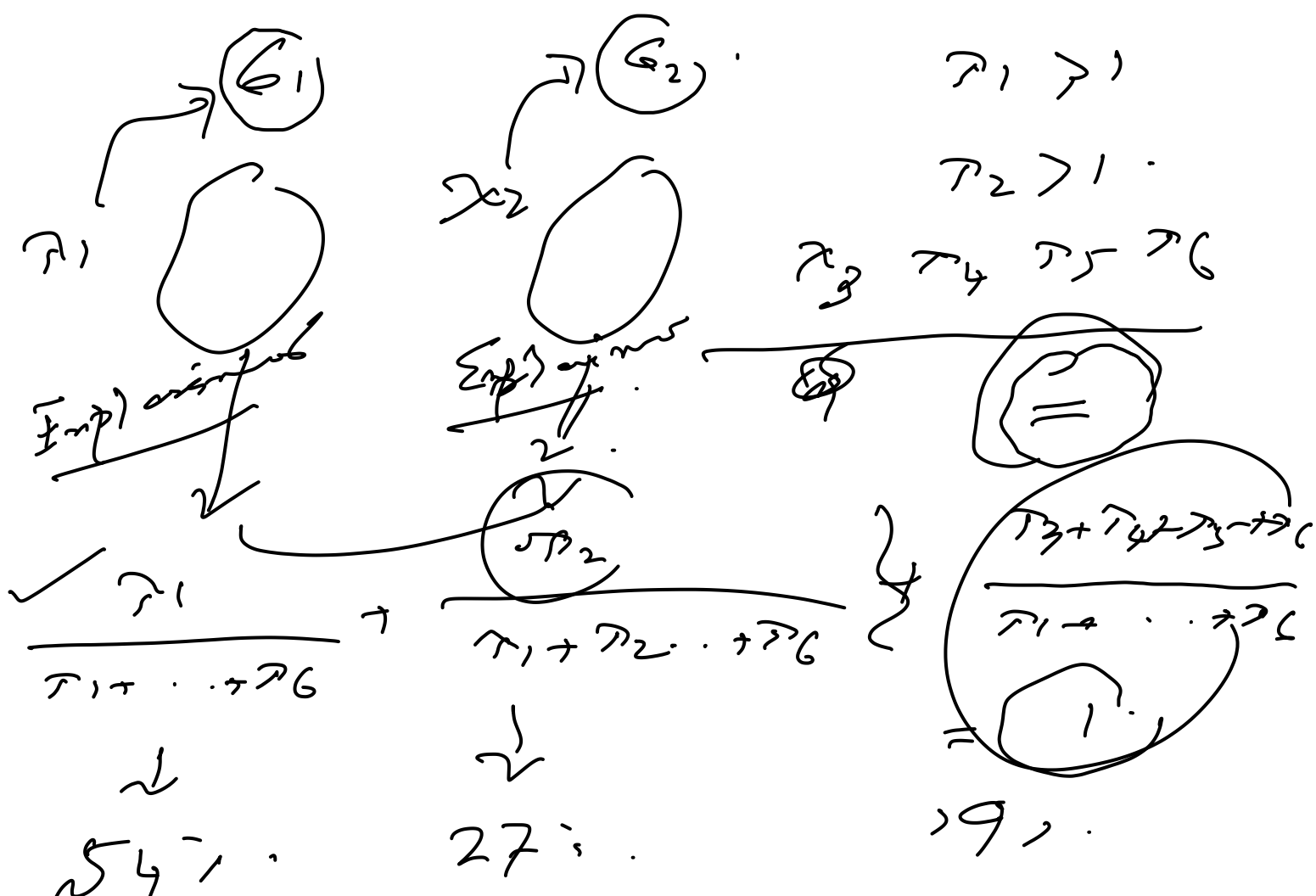
(Good)

> 50%

4

(20%)

==



81 19 $= 100$
 Explained Unexplained
 Dimension is reduced



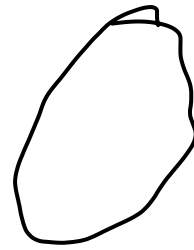
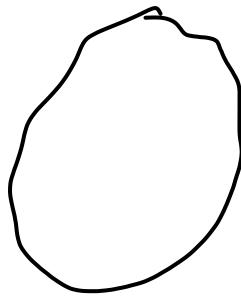
Explained $> 70\%$
 11/2 Explained $> 90\%$

PCA - helped us to
 determine the number of
 possible homogeneous groups
 with the help of eigen
 value;

BIG

=

numbers



Since. Eigen value based on
 = Correlation matrix & correlates
 occur pairwise so any group
 containing minimum two variables

$x_1 \quad x_2 \quad x_3 \quad x_4 \quad x_5 \quad x_6$

