Algebra of Principal Component Analysis $\left(Y_c, Q = I_p, D = \frac{1}{n}I_n\right)$

Data:
$$\mathbf{Y} = \begin{bmatrix} 2 & 1 \\ 3 & 4 \\ 5 & 0 \\ 7 & 6 \\ 9 & 2 \end{bmatrix}$$
 Centre each column on its mean: $\mathbf{Y}_c = [y - \bar{y}] = \begin{bmatrix} -3.2 & -1.6 \\ -2.2 & 1.4 \\ -0.2 & -2.6 \\ 1.8 & 3.4 \\ 3.8 & -0.6 \end{bmatrix}$
where $\overline{\mathbf{y}} = \begin{bmatrix} 5.2 & 2.6 \end{bmatrix}^t$

$$\mathbf{S}_I = \mathbf{SQ} = \frac{1}{5} \mathbf{Y}_c^t \mathbf{Y}_c$$
 is the covariance matrix (2 variables): $\mathbf{S}_I = \begin{bmatrix} 6.56 & 1.28 \\ 1.28 & 4.64 \end{bmatrix}$

Equation for eigenvalues and eigenvectors of $S_I : (S_I - \lambda_k I) a_k = 0$

- Axis a_k
- Loadings (principal factors) $u_k = \mathbf{Q}a_k$ because $\mathbf{Q} = \mathbf{I}$

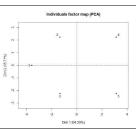
Eigenvalues:
$$\lambda_1 = 7.2$$
; $\lambda_2 = 4.0 \Rightarrow I_q = tr(\mathbf{S}_I) = \lambda_1 + \lambda_2 = 11.2$

Matrix of eigenvalues :
$$\Lambda = \begin{bmatrix} 7.2 & 0 \\ 0 & 4.0 \end{bmatrix}$$

Matrix of eigenvectors:
$$\mathbf{U} = \begin{bmatrix} \mathbf{u}_1 & \mathbf{u}_2 \end{bmatrix} \approx \begin{bmatrix} 0.89 & -0.45 \\ 0.45 & 0.89 \end{bmatrix}$$

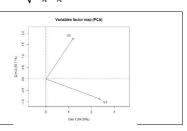
Coord of the 5 objects in ordination space: $\mathbf{F} = \mathbf{Y}_c \mathbf{U} \Rightarrow$ Principal components

$$\mathbf{F} = \begin{bmatrix} \mathbf{F}^1 & \mathbf{F}^2 \end{bmatrix} \approx \begin{bmatrix} -3.58 & 0 \\ -1.34 & 2.24 \\ -1.34 & -2.24 \\ 3.13 & 2.24 \\ 3.13 & -2.24 \end{bmatrix}$$
with F'F=1



Coord of the 2 variables in ordination space: $G^k = \sqrt{\lambda_k} a_k$

$$\mathbf{G} = \begin{bmatrix} \mathbf{G}^1 & \mathbf{G}^2 \end{bmatrix} \approx \begin{bmatrix} 2.4 & -0.89 \\ 1.2 & 1.79 \end{bmatrix}$$



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# exemple à la main => sous R : résultats au signe près
# ACP (Yc, Ip, (1/n) In)
y < -c(2,1,3,4,5,0,7,6,9,2)
Y<-matrix(v,ncol=2,byrow = T); Y
Yc<-scale(Y,scale=FALSE);Yc
dim(Yc) ; n=nrow(Yc)
SI < -(1/n) *t(Yc) % * % Yc ; SI
eig=eigen(SI) ; eig
#cP
F=Yc%*%eig$vectors; round(F,2)
#coord des variables
G < -matrix(0,2,2)
G[,1]=sqrt(eig$values[1])%*%eig$vectors[,1]
G[,2]=sqrt(eig$values[2])%*%eig$vectors[,2]
round(G,2)
library(FactoMineR)
PCA(as.data.frame(Yc), scale.unit=FALSE)
#resultat avec PCA au signe près!
```