

Dimensionality Reduction

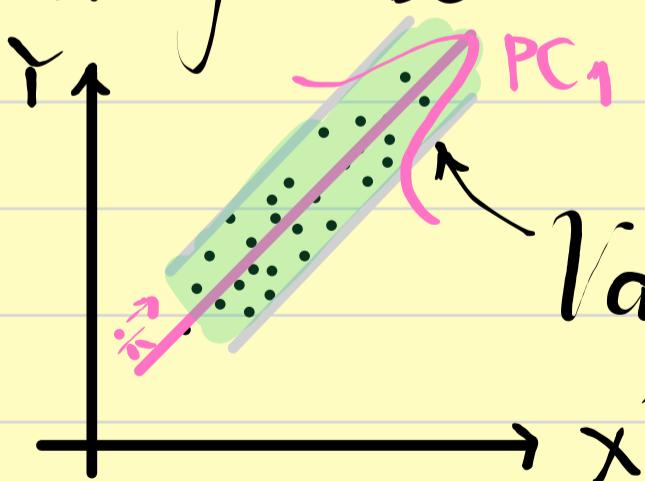
Examples Multidimensional Systems

- Management System
- Images/Videos

Principal component Analysis (PCA)

Definition: a PC is the Eigenvector of the Covariance Matrix.

Intuition: the PC₁ is the direction in which variability can be best explained.



Variability is smallest
in this direction

Covariance Matrix 3x3

KPI₁ KPI₂ KPI₃
 Cov₁ ≡ ≡ ≡
 Cov₂ ≡ ≡ ≡
 Cov₃ ≡ ≡ ≡
 ... ≡ ≡ ≡

$$\lambda = \begin{bmatrix} \text{VAR}(x) & \text{cov}(x,y) & \text{cov}(x,z) \\ \text{cov}(x,y) & \text{VAR}(y) & \text{cov}(y,z) \\ \text{cov}(x,z) & \text{cov}(y,z) & \text{VAR}(z) \end{bmatrix}$$

$$\text{VAR}(x) = \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n-1}$$

$$\text{cov}(x,y) = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{n-1}$$

$$\det(\lambda I - \lambda \lambda) = 0 \rightarrow \lambda \rightarrow \vec{v}$$

Example for 2 KPIs :

$$\cdot \text{KPI}_1 \stackrel{\text{€}}{\equiv} \text{Unit} \stackrel{\text{€}}{=} [17, 19, 23, 22] = x$$

$$\cdot \text{KPI}_2 \stackrel{\text{Revenue}}{\equiv} [34, 41, 46, 45] = y$$

$$\bar{x} = \frac{17+19+23+22}{4} = 20'25$$

$$\bar{y} = \frac{34+41+46+45}{4} = 41'5$$

$$\text{VAR}(x) = \frac{(17-20'25)^2 + (19-20'25)^2 + (23-20'25)^2 + (22-20'25)^2}{4} = 7'67$$

$$\text{VAR}(y) = \frac{(34-41'5)^2 + (41-41'5)^2 + (46-41'5)^2 + (45-41'5)^2}{4} = 29'67$$

$$\text{cov}(x,y) = \frac{(17-20^{1/2}5)(34-41^{1/2}5) + (19-20^{1/2}5)(41-41^{1/2}5) + (23-20^{1/2}5)(46-41^{1/2}5) + (27-20^{1/2}5)(45-41^{1/2}5)}{3}$$

$$= 14^{1/2}5$$

$$\mathcal{A} = \begin{bmatrix} 7^{1/2}67 & 14^{1/2}5 \\ 14^{1/2}5 & 29^{1/2}67 \end{bmatrix}$$

$$\det(\mathcal{A} - \lambda I) = 0 \rightarrow \det \left[\begin{bmatrix} 7^{1/2}67 - \lambda & 14^{1/2}5 \\ 14^{1/2}5 & 29^{1/2}67 - \lambda \end{bmatrix} \right] = 0 \rightarrow$$

$$\rightarrow (7^{1/2}67 - \lambda)(29^{1/2}67 - \lambda) - 14^{1/2}5^2 = 0 \rightarrow$$

$$\rightarrow 7^{1/2}67 \cdot 29^{1/2}67 - (7^{1/2}67 + 29^{1/2}67)\lambda + \lambda^2 - 14^{1/2}5^2 = 0$$

$$\rightarrow \lambda^2 - 37^{1/2}34\lambda + 17^{1/2}32 = 0$$

$$\rightarrow \lambda = \frac{37^{1/2}34 \pm \sqrt{37^{1/2}34^2 - 4 \cdot 17^{1/2}32}}{2} = \begin{cases} \lambda_1 = 36^{1/2}87 \\ \lambda_2 = 0^{1/2}47 \end{cases}$$

$$\begin{aligned} ax^2 + bx + c &= 0 \\ x &= \frac{-b \pm \sqrt{b^2 - 4ac}}{2a} \end{aligned}$$

$$\boxed{\lambda_1 = 36^1 87} \rightarrow A \cdot \vec{v}_1 = \lambda_1 \vec{v}_1 \rightarrow \begin{bmatrix} 7'67 & 14'5 \\ 14'5 & 29'67 \end{bmatrix} \begin{bmatrix} v_{11} \\ v_{12} \end{bmatrix} = 36^1 87 \begin{bmatrix} v_{11} \\ v_{12} \end{bmatrix}$$

$$\begin{aligned} \cdot 7'67 \cdot v_{11} + 14'5 \cdot v_{12} &= 36^1 87 v_{11} \\ \cdot 14'5 \cdot v_{11} + 29'67 \cdot v_{12} &= 36^1 87 \cdot v_{12} \end{aligned}$$

$$v_{11} = 1 \rightarrow v_{12} = \frac{1}{0'49} = 2'04$$

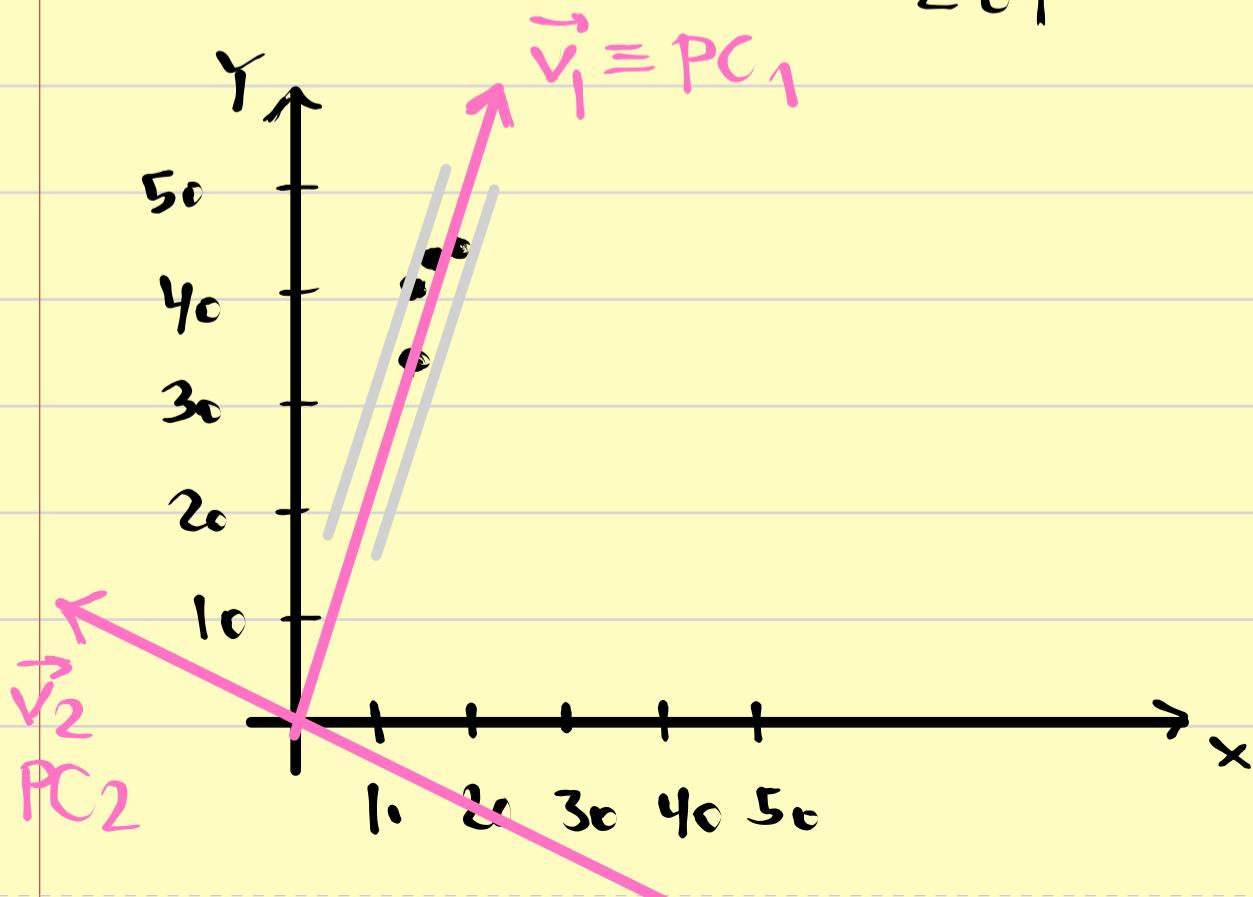
$$\vec{v}_1 = \begin{bmatrix} 1 \\ 2'04 \end{bmatrix}$$

$$\boxed{\lambda_2 = 0^1 47} \rightarrow A \cdot \vec{v}_2 = \lambda_2 \cdot \vec{v}_2 \rightarrow \begin{bmatrix} 7'67 & 14'5 \\ 14'5 & 29'67 \end{bmatrix} \begin{bmatrix} v_{21} \\ v_{22} \end{bmatrix} = 0^1 47 \begin{bmatrix} v_{21} \\ v_{22} \end{bmatrix}$$

$$\begin{aligned} \cdot 7'67 \cdot v_{21} + 14'5 \cdot v_{22} &= 0^1 47 v_{21} \\ 14'5 \cdot v_{21} + 29'67 \cdot v_{22} &= 0^1 47 \cdot v_{22} \end{aligned}$$

$$v_{21} = 1 \rightarrow v_{22} = \frac{-1}{2'01} = -0'497$$

$$\vec{v}_2 = \begin{bmatrix} 1 \\ -0'497 \end{bmatrix}$$



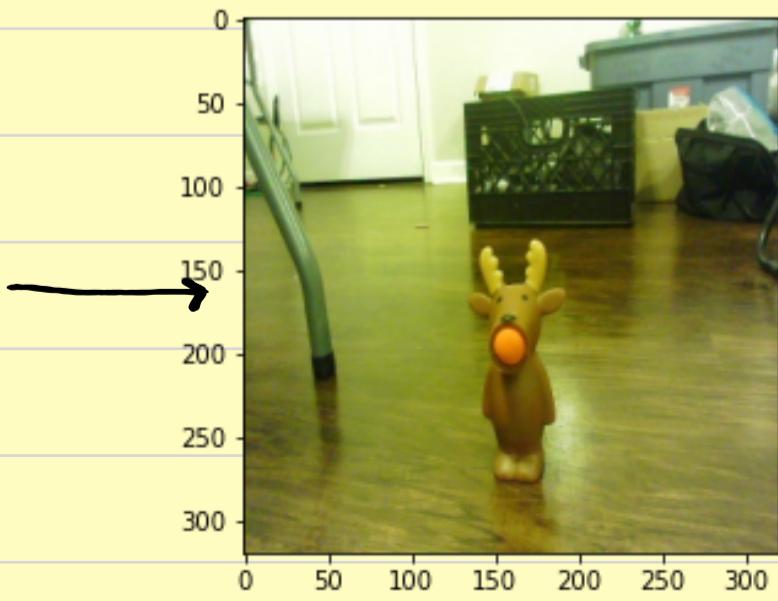
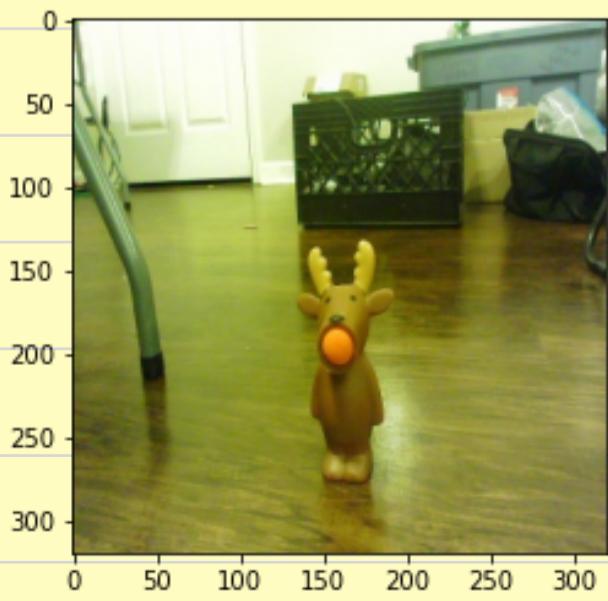
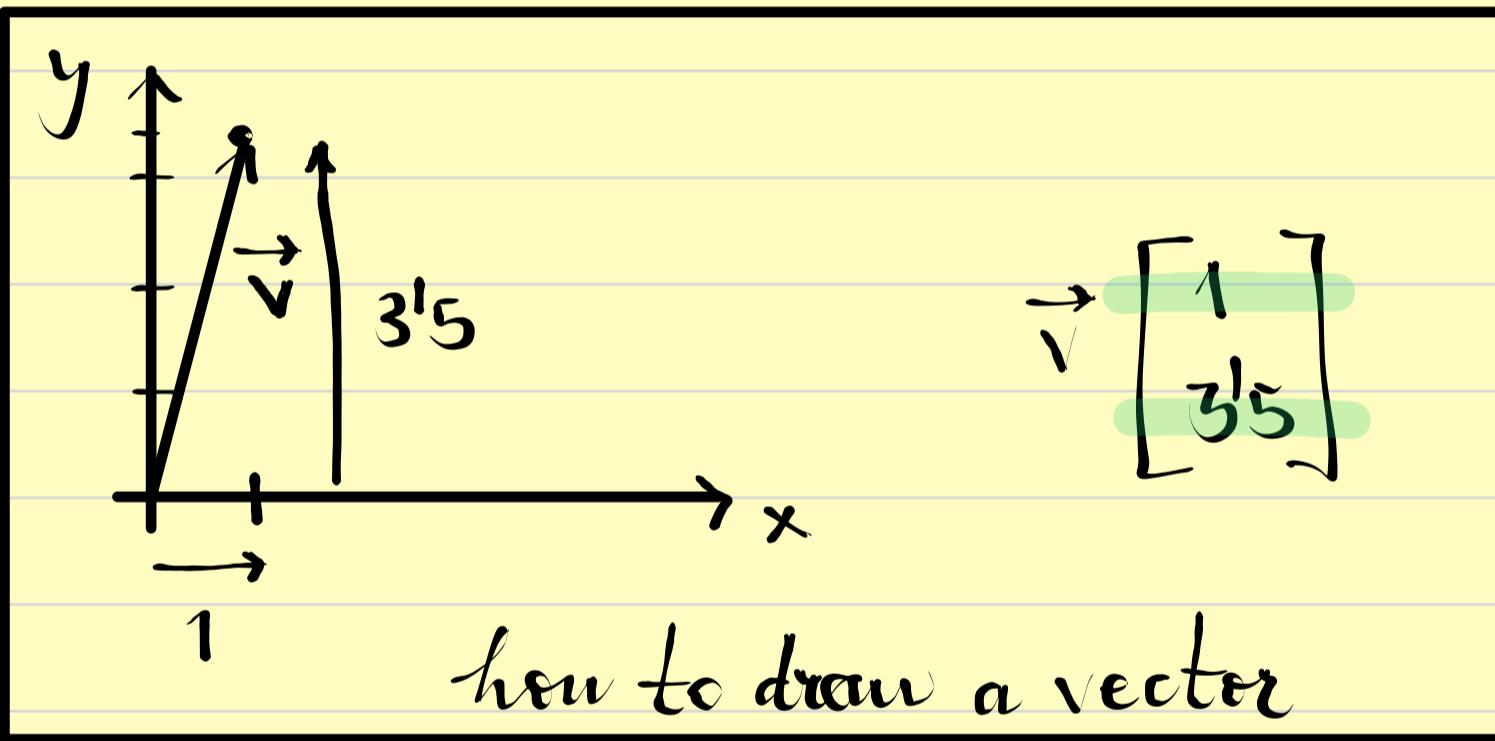
$$\lambda_1 = 36^1 87$$

$$\lambda_2 = 0^1 47$$

The amount of variability explained by each PC is given by the eigenvalues:

$$\% \text{ VAR}(\vec{v}_1) = \frac{\lambda_1}{\sum \lambda_i} \cdot 100 = \frac{36'87 \cdot 100}{36'87 + 0'47} = 98'7\%.$$

$$\% \text{ VAR}(\vec{v}_2) = \frac{\lambda_2}{\sum \lambda_i} \cdot 100 = \frac{0'47 \cdot 100}{36'87 + 0'47} = 1'3\%.$$



310×310

20 PCAs

.zip Datacompression