

Dimensionality Reduction

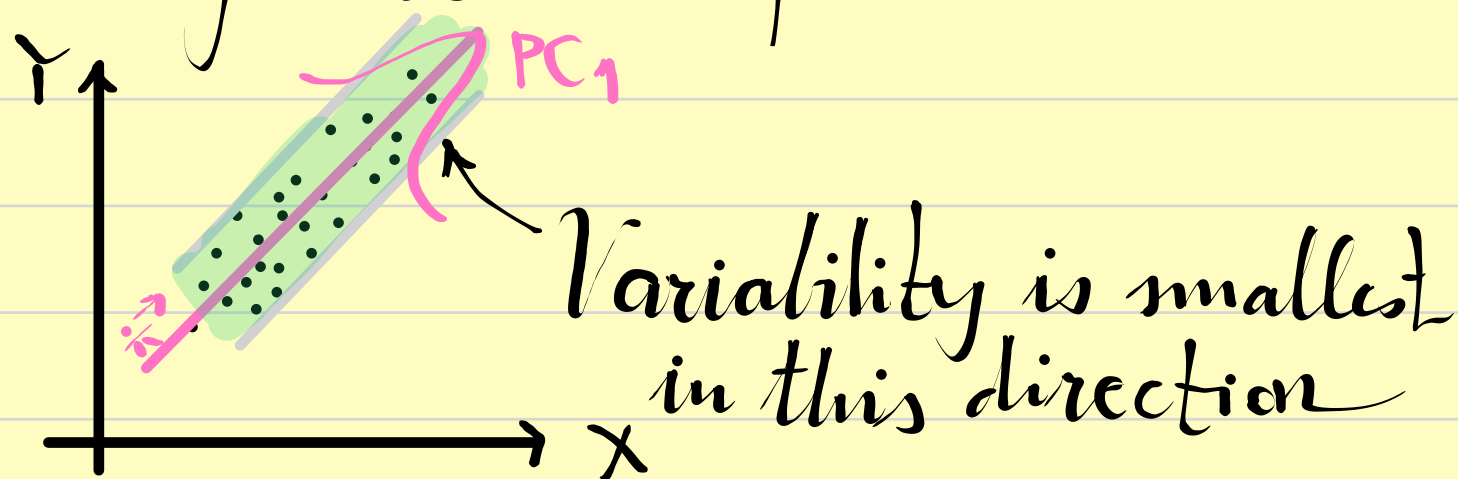
Examples Multidimensional Systems

- Management System
- Images/Videos

Principal component Analysis (PCA)

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

Intuition: the PC_1 is the direction in which variability can be best explained.



Covariance Matrix 3x3

	KPI ₁	KPI ₂	KPI ₃	
av ₁	≡	≡	≡	$A = \begin{bmatrix} \text{VAR}(x) & \text{cov}(x,y) & \text{cov}(x,z) \\ \text{cov}(x,y) & \text{VAR}(y) & \text{cov}(y,z) \\ \text{cov}(x,z) & \text{cov}(y,z) & \text{VAR}(z) \end{bmatrix}$
av ₂	≡	≡	≡	
av ₃	≡	≡	≡	
...	≡	≡	≡	

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

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

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

Example for 2 KPIs:

• KPI₁ ≡ ~~€~~ unit ≡ [17, 19, 23, 22] ≡ x

• KPI₂ ≡ Revenue ≡ [34, 41, 46, 45] ≡ y

$$\bar{x} = \frac{17+19+23+22}{4} = 20.25 \quad \bar{y} = \frac{34+41+46+45}{4} = 41.5$$

$$\text{VAR}(x) = \frac{(17-20.25)^2 + (19-20.25)^2 + (23-20.25)^2 + (22-20.25)^2}{4} = 1.61$$

$$\text{VAR}(y) = \frac{(34-41.5)^2 + (41-41.5)^2 + (46-41.5)^2 + (45-41.5)^2}{4} = 29.67$$

$$\text{Cov}(X, Y) = \frac{(17-20'25)(34-41'5) + (19-20'25)(41-41'5) + (23-20'25)(46-41'5) + (22-20'25)(45-41'5)}{3}$$

$$= 14'5$$

$$A = \begin{bmatrix} 7'67 & 14'5 \\ 14'5 & 29'67 \end{bmatrix}$$

$$\det(A - \lambda I) = 0 \rightarrow \det \begin{bmatrix} 7'67 - \lambda & 14'5 \\ 14'5 & 29'67 - \lambda \end{bmatrix} = 0 \rightarrow$$

$$\rightarrow (7'67 - \lambda)(29'67 - \lambda) - 14'5^2 = 0 \rightarrow$$

$$\rightarrow 7'67 \cdot 29'67 - (7'67 + 29'67)\lambda + \lambda^2 - 14'5^2 = 0$$

$$\rightarrow \lambda^2 - 37'34\lambda + 17'32 = 0$$

$$\rightarrow \lambda = \frac{37'34 \pm \sqrt{37'34^2 - 4 \cdot 17'32}}{2} = \begin{matrix} \nearrow \lambda_1 = 36'87 \\ \searrow \lambda_2 = 0'47 \end{matrix}$$

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

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

$$\begin{aligned} \cdot 7'67 \cdot v_{11} + 14'5 \cdot v_{12} &= 36'87 v_{11} \\ \cdot 14'5 \cdot v_{11} + 29'67 \cdot v_{12} &= 36'87 v_{12} \end{aligned} \rightarrow \begin{aligned} v_{11} &= \frac{14'5}{29'67 - 36'87} v_{12} = 0'49 v_{12} \end{aligned}$$

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

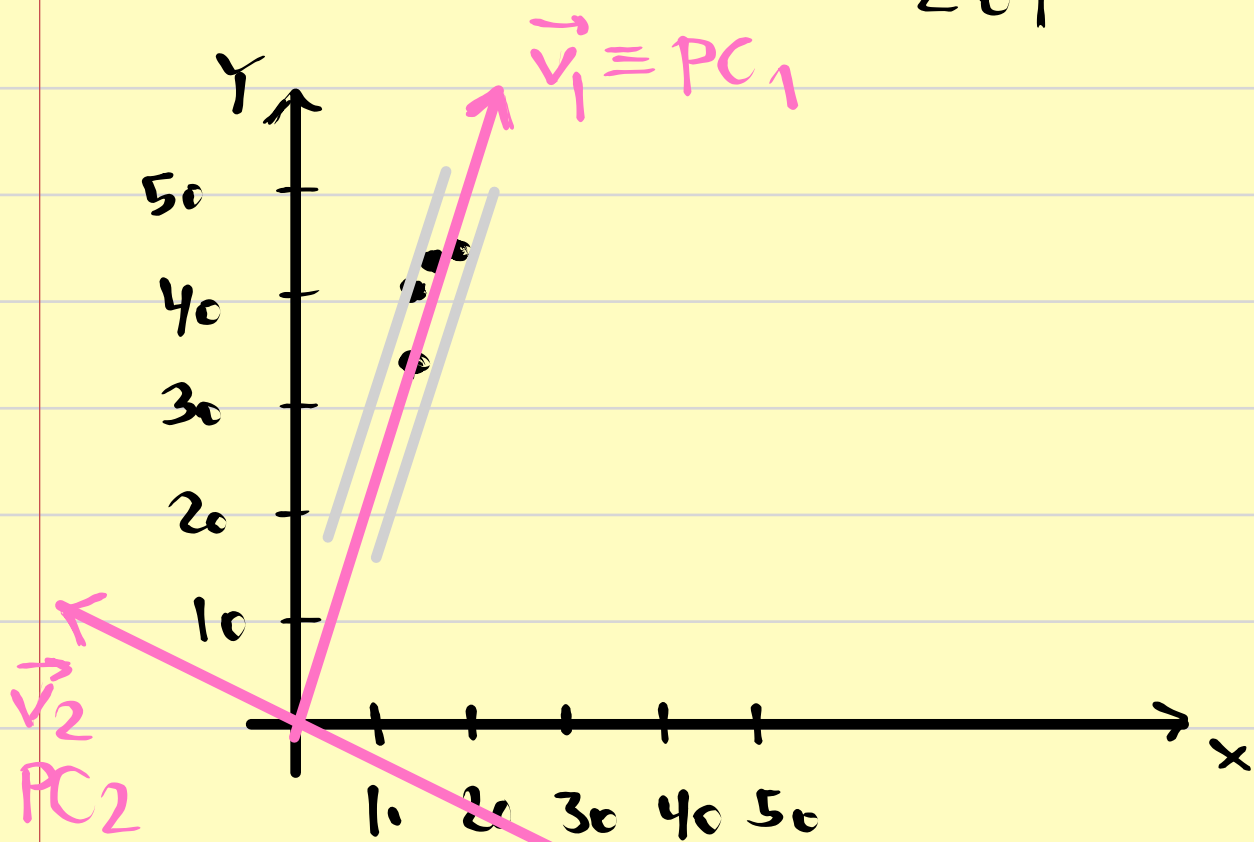
$$\boxed{\vec{v}_1 = \begin{bmatrix} 1 \\ 2'04 \end{bmatrix}}$$

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

$$\begin{aligned} \cdot 7'67 \cdot v_{21} + 14'5 \cdot v_{22} &= 0'47 v_{21} \\ \cdot 14'5 \cdot v_{21} + 29'67 \cdot v_{22} &= 0'47 v_{22} \end{aligned} \rightarrow \begin{aligned} v_{21} &= \frac{14'5}{-7'2} v_{22} = -2'01 v_{22} \end{aligned}$$

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

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

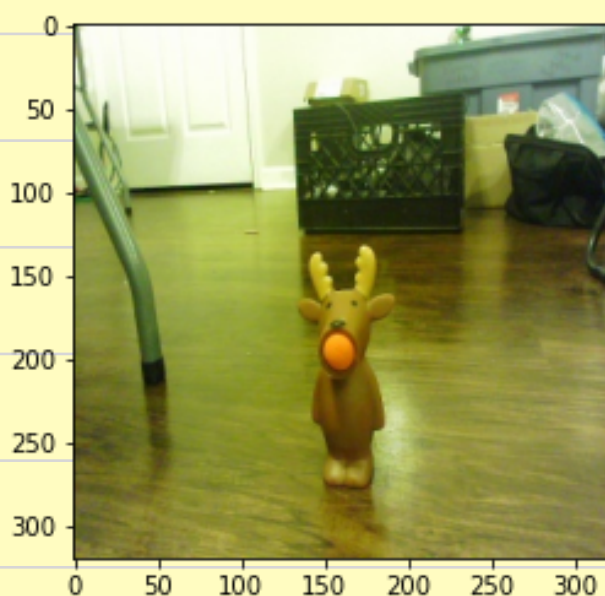
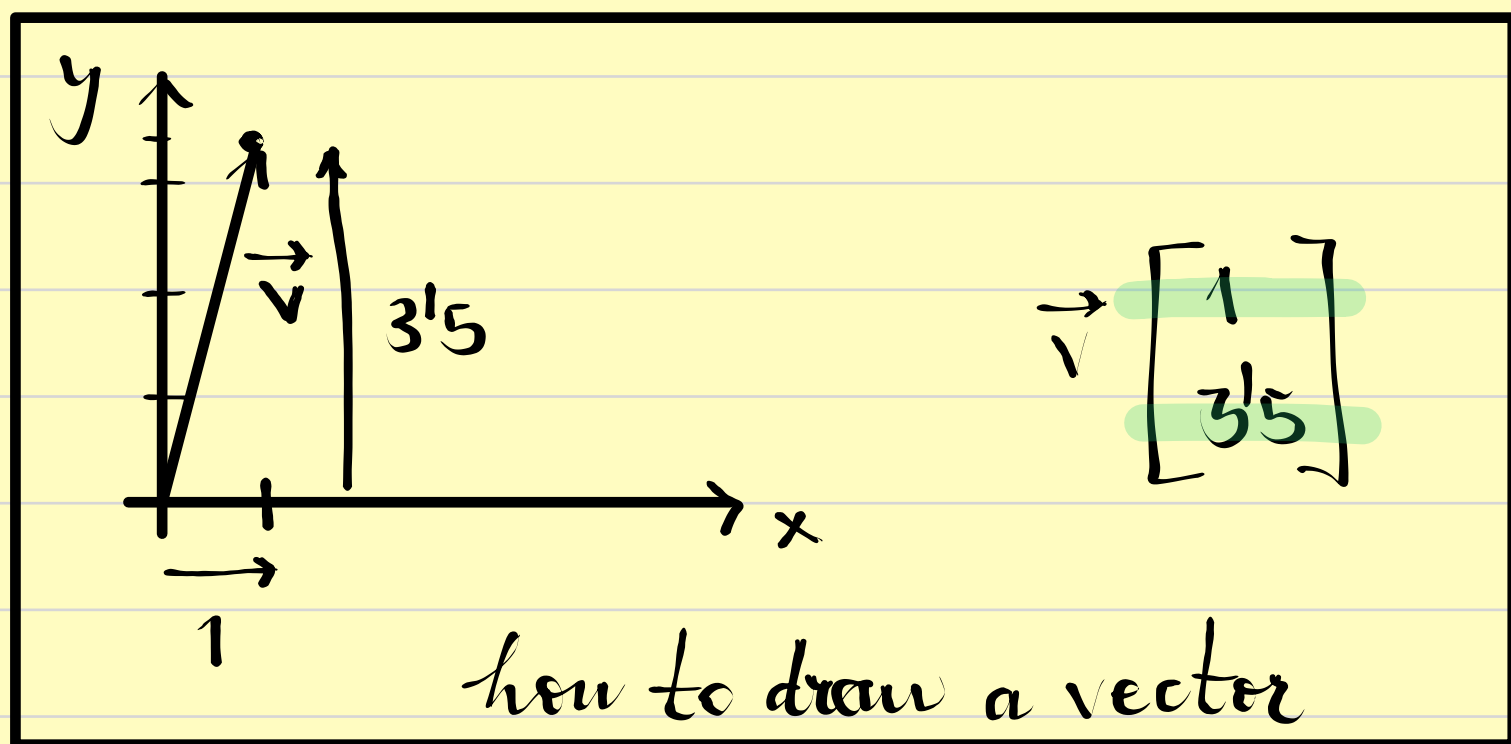


$$\begin{aligned} \lambda_1 &= 36'87 \\ \lambda_2 &= 0'47 \end{aligned}$$

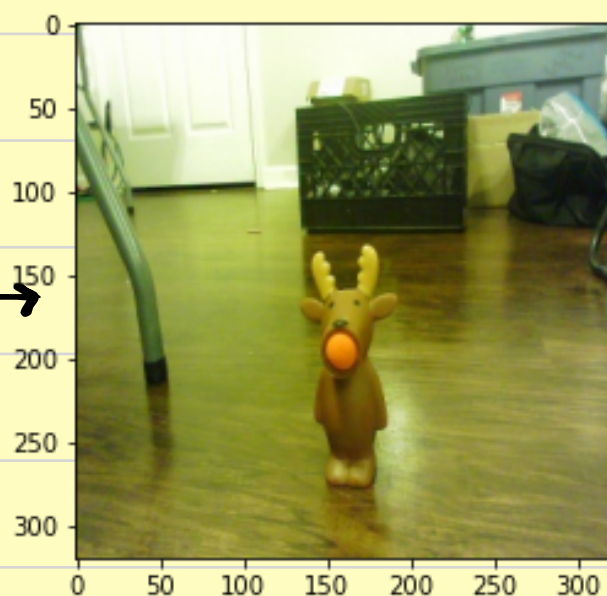
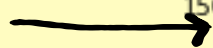
The amount of variability explained by each PE is given by the eigenvalues:

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

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



310 x 310



20 PCAs

.zip Datacompression