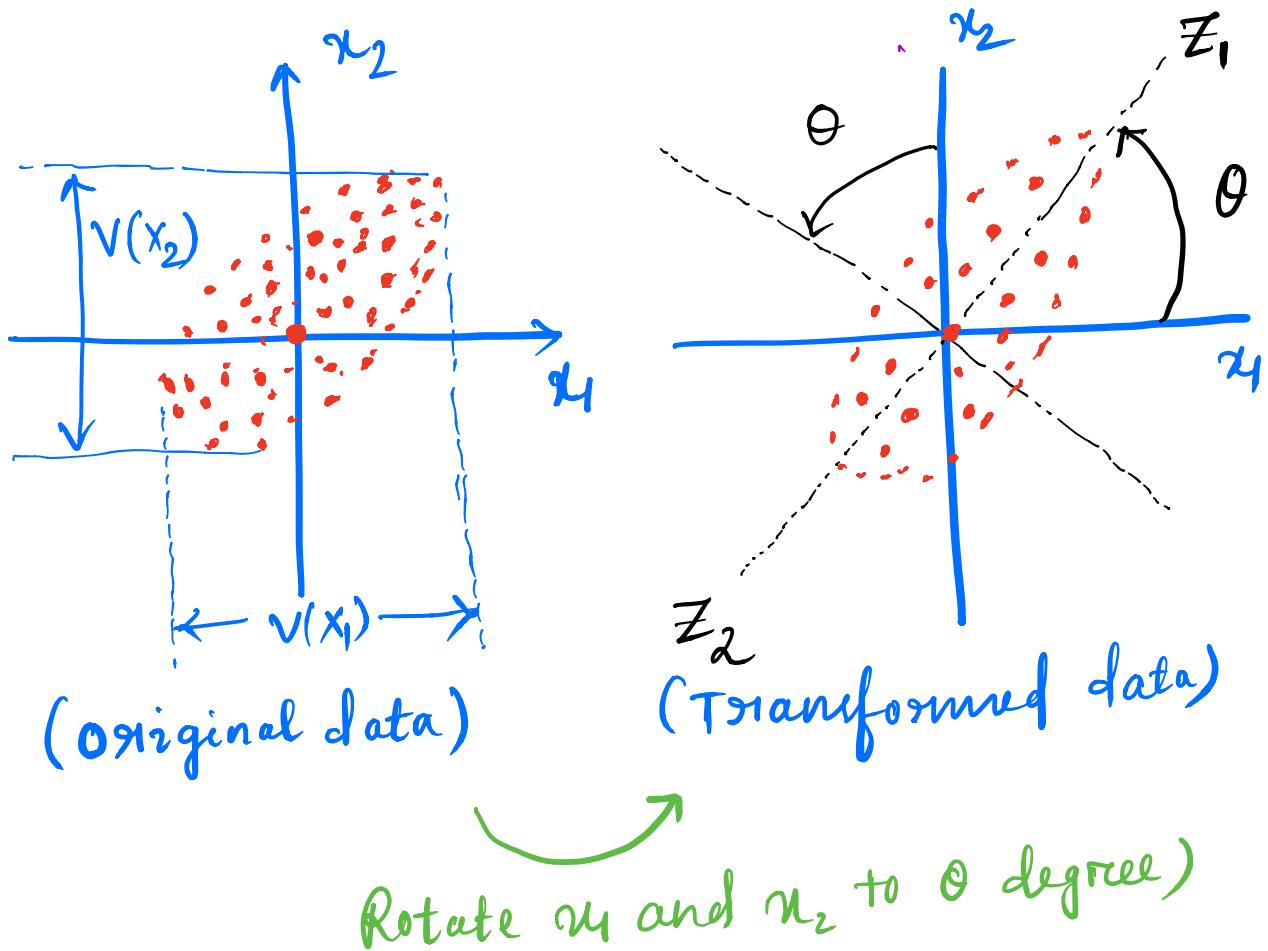
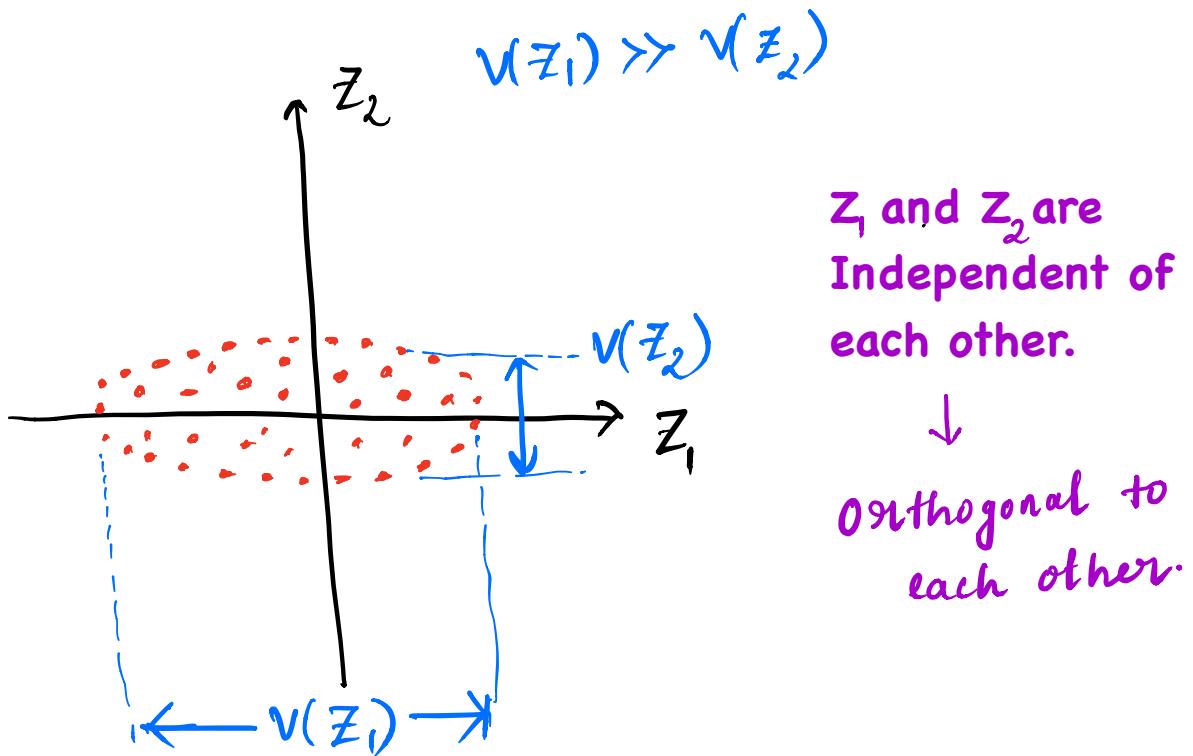


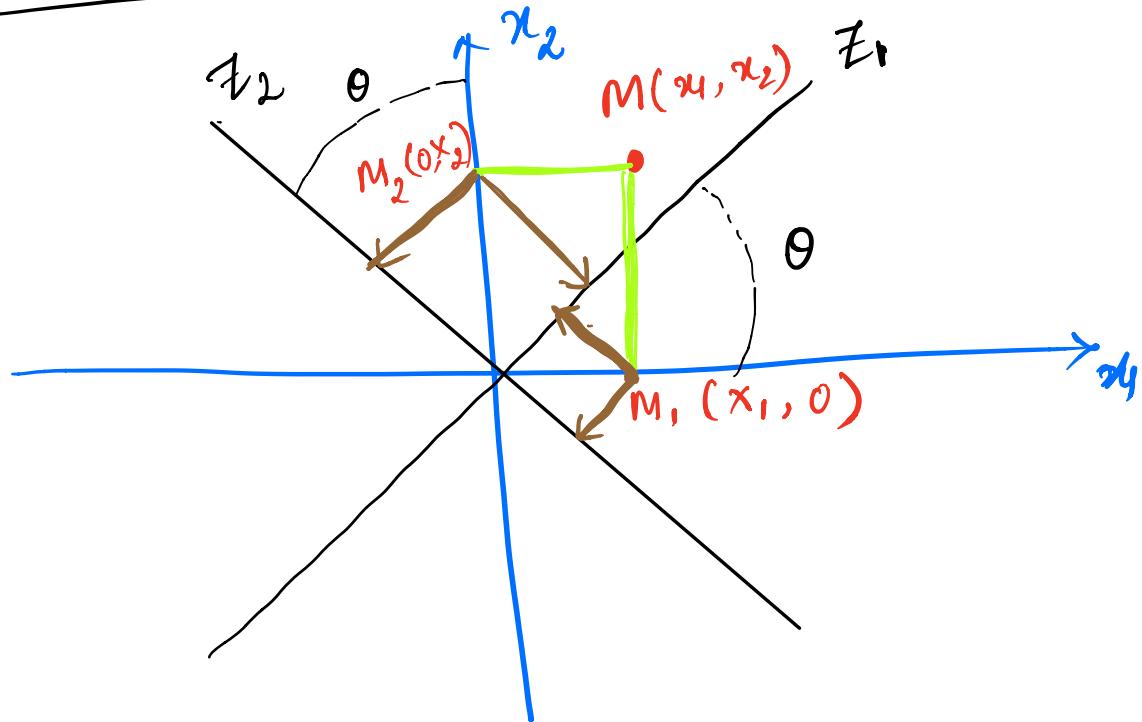
PCA

Principal Component Analysis





How to Project a Point from x_1, x_2 to z_1, z_2



In Real time some columns are important and some columns are not.

X	Y	Z
1	1	1
0.5	0	0
0.25	1	1
0.35	1.5	1.5
0.45	1	1
0.57	2	2.1
0.62	1.1	1

Is Z- is adding
any new
Information
beyond what is
already contained
in y?
Y and Z are
highly
correlated.

Derivation

$$\begin{bmatrix} 2 & 3 \\ 2 & 1 \end{bmatrix}_{2 \times 2} \times \begin{bmatrix} 3 \\ 2 \end{bmatrix}_{2 \times 1} = \begin{bmatrix} 12 \\ 8 \end{bmatrix}_{2 \times 1} = 4 \times \begin{bmatrix} 3 \\ 2 \end{bmatrix}$$

↓ ↓ ↓ ↓
Input matrix(A) eigen vector(v) eigen value(λ) eigen vector(v)

$$AV = \lambda \cdot V$$

Calculating Eigen values:

$$A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix}$$

$$AV = \lambda V$$

$$\Rightarrow AV - \lambda V = 0$$

$$\Rightarrow (A - \lambda I) = 0$$

$\Rightarrow (A - \lambda I) = 0$, where I is the identity matrix.

$$\begin{aligned} \text{Then, } |A - \lambda I| &= \left| \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right| \\ &= \left| \begin{bmatrix} -\lambda & 1 \\ -2 & -3 - \lambda \end{bmatrix} \right| \\ &= (-\lambda \times (-3 - \lambda)) \\ &= \lambda^2 + 3\lambda + 2 \end{aligned}$$

$$\boxed{\lambda_1 = -1, \quad \lambda_2 = -2} \rightarrow \text{Eigen values}$$

Step-1

find co-variance matrix

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

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

e.g: covariance matrix of 3 rows & 3 columns:

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

Annotations:
 covariance of itself is nothing but variance
 cov(x, x) → cov(x, y) → cov(x, z)
 cov(y, x) → cov(y, y) → cov(y, z)
 cov(z, x) → cov(z, y) → cov(z, z)

Step-2

calculate the eigen vectors and eigen values
of the covariance matrix.

Step-3

Select m eigen vectors that corresponds to the
m largest eigen values to be the new axis.

PCA

Large Table

X1	X2	X3	X4	X5
*	*	*	*	*
*	*	*	*	*
*	*	*	*	*
*	*	*	*	*
*	*	*	*	*
*	*	*	*	*
*	*	*	*	*
*	*	*	*	*
*	*	*	*	*
*	*	*	*	*
*	*	*	*	*
*	*	*	*	*
*	*	*	*	*
*	*	*	*	*
*	*	*	*	*
*	*	*	*	*

Covariance matrix

$$\begin{pmatrix} * & * & * & * & * \\ * & * & * & * & * \\ * & * & * & * & * \\ * & * & * & * & * \\ * & * & * & * & * \end{pmatrix}$$

Eigenstuff

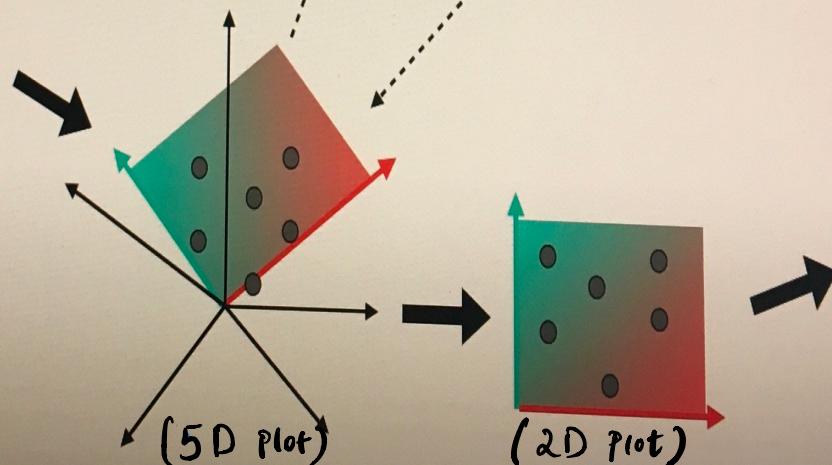
$$\begin{array}{ll} v_1 & \lambda_1 \\ v_2 & \lambda_2 \\ v_3 & \lambda_3 \\ v_4 & \lambda_4 \\ v_5 & \lambda_5 \end{array}$$

Big

Small

Small Table

W1	W2
*	*
*	*
*	*
*	*
*	*
*	*
*	*
*	*
*	*
*	*
*	*
*	*
*	*
*	*
*	*



PCA is an orthogonal linear transformation that transfers the data to a new coordinate system such that the greatest variance by any projection of the data comes to lie in the 1st coordinate, the second greatest variance lies in the 2nd coordinates and so on - - -