

1. standardize the Cov matrix
(normalize the data)
2. Computing the Cov matrix
relation b/w 2 features (totally diff)
on the standardized fit transform fn.
data.

for how many features we want to consider

3. E-values & vectors
calculate using n how much var is explained along a vector.
(n x n) matrix

How well are the datasets are related along the line.

find PC's (like 3) multiply with Y ; to effectively reduce dim

how to take a diff transform such that it leads to different subspace.

we need to find a transformation

regression: squares are min

classification:

Direction along highest & lowest variance.

1st E-vector \rightarrow Dirⁿ of highest variance.

compute cov matrix for all features

cumulative var: ex: 30 vectors / components

If we take all 30 PC's \Rightarrow original -

max var: 100%

If we truncate some components

\Rightarrow ex 15 vectors that can explain the original at 99%

implying we don't need to consider other 15 vectors

5. Reducing the dimensions

 $Y \rightarrow$ Eigen vectors

Take all the feature labels

In SD :

all the vectors get multiplied get decomposed along a line.

dot product along x direction

Ex: Take 100% var

we have a vector along $[1, 2]$ to be decomposed into 2 vectors.

$$\begin{bmatrix} 1 \\ 2 \end{bmatrix} = x [1, 0] + x [0, 1]$$

Basis vectors for ^{original subspace.} transformationis same as PC components multiplied with ~~matrix~~ ^{matrix}.

defined subspace with basis vectors