P(A extracts features and reduces the number of features

In some problem set, if there are 1000 of features I we try to train ML model it will overfit

When machine is trained to take a decision, it considers features. They require space and computation for training If we can reduce the number of features we can reduce complexity of Model

If we remove few features, we can make the model simpler

But removing features is loss of data

If we remove the features that are less important, we will get lower loss.

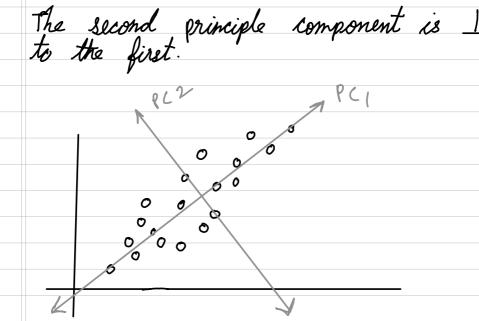
PCA automatically finds most important features to keep

Data is of above form (20 data) we want to convert it into 1D to with lowest loss, we can

1) data representation

In order to minimize the loss we want a line that covers the maximum variation

The line that covers the maximum data points is the principle component



P(A is also known as k-1 transform and notelling transform

It is used in data compression

World without P(A long court we drop x or y sxis? Principle components are I axes with least loss

Steps in PCA

O standardize dataset
$$\frac{x-x}{3}$$

$$3 = \sqrt{(\varkappa - \bar{\varkappa})^{2}}$$
2 Calculate Covariance matrix
$$(ov(\chi, y) = \underbrace{\Xi(\chi - \bar{\varkappa}) * (y - \bar{y})}_{N}$$

3 Calculate eigenvalues & eigenvectors
$$\begin{bmatrix} CoV \end{bmatrix} - \times I = O - solve$$

PCA

Principle Component Analysis

Used to identify most important points or features (01) Data

Reduce dimention from 2 to 1

step 1: No of features = 2 No of samples = 4

step 2: Calcutate Mean

 $\overline{\chi} = 4 + 8 + 13 + 7 = 4$

 $\hat{y} = \frac{11 + 4 + 5 + 14}{1} = \frac{1}{1}$

Note step 3: Covariance Matrix for data $S = \frac{1}{N-1} \underbrace{\frac{1}{2} \left(X; -\overline{X}\right)^{2}}_{i=1} \underbrace{\frac{1}{2} \left(X - \overline{X}\right)^{2}}_{2 \times 2X} \underbrace{\frac{1}{2} \left(X - \overline{X}\right)^{2}}_{2 \times 2X}$ 2 x2-2x x+ x

 $(6v(Y, X) = \frac{1}{1.1}((4-8) + \cdots) = 14$

 $Cov(x,y) = \frac{1}{N-1} \frac{S}{S}(x-\overline{x})(y-\overline{y})$ - 1 (4-8) x (11-85) + ...

Cov(y,y) = Coul(y,y) = -11 $Cov(y,y) = \frac{1}{N-1} (\frac{1}{2}, \dots) = \frac{2}{3}$

Matrix (x,x) (y,y) = | 14 -11 | (y,x) (y,y) = | -11 23 |

= - []

Step 4 Eigenvalue, eigenvector

 $(S-\lambda I)V=0$ $dd(S-\lambda I)=0 \qquad \text{or } V=0$

λ ⇒ eigenvalue √ ⇒ eigenvector

Sy = AV

 $\begin{bmatrix} 14 & -11 \\ -11 & 23 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = 0$

solving for h A, = 30.384 λ2 = 6.615

 $\lambda^2 - 31\lambda + 20 = 0$

(S-)1) V = 0 $\begin{vmatrix}
14 - \lambda & -11 \\
-11 & 23 - \lambda
\end{vmatrix}$ $\begin{vmatrix}
V_1 \\
V_2
\end{vmatrix} = \begin{vmatrix}
0 \\
0
\end{vmatrix}$

First principle component 1, = 30 381

eigenvector for the principle component

Consider highest value of a

 $(14-\lambda)V_1 + (-11)V_2 = 0 - (1)$ $(-11)V_1 + (23-2)V_2 = 0$

From () V1 = 11

-16.3848

X = 30.38 V2 14-30.38 V, - 11 V2= -16.384 8

-16.38 V 112+((6.28)2

Normalized eigen vectors

Step 5: Derive 2 to 1 using 1st)

 $\begin{bmatrix} V_1 & V_2 \end{bmatrix} \qquad \begin{bmatrix} \chi_1 - \overline{\chi} \\ y_1 - \overline{y} \end{bmatrix}$ [0.5514 - 0.8303] [4-8] [1-8.5]

 $rac{1}{1} = -4.3052$

 $\ell_{13} = 5.69$ $\ell_{14} = -5.1238$

 $\begin{bmatrix} V_1 V_2 \end{pmatrix} \begin{bmatrix} \chi_2 - \overline{\chi} \\ \chi_2 - \overline{\chi} \end{bmatrix}$ $\begin{bmatrix}
 0.5574 - 0.8303
 \end{bmatrix}
 \begin{bmatrix}
 0
 \\
 -3.5
 \end{bmatrix}
 = +3.7386$

Final Answer

J-4.3052 3.7386 5.6928 -5.1238