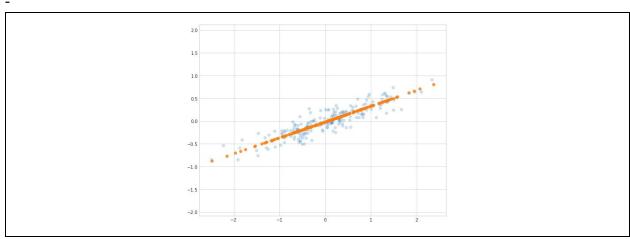
## 1) Principal Component Analysis

-Please see the notebook at the end of this document.

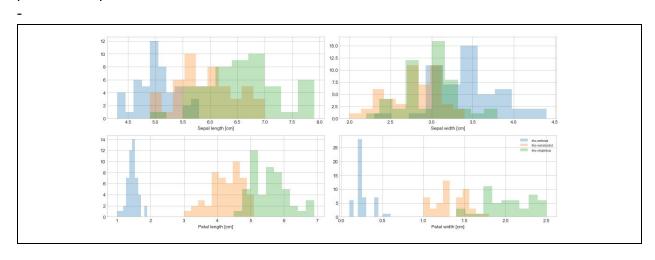
input principal components

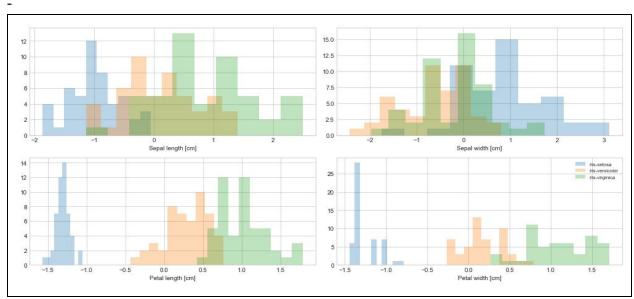
15
10
05
> 00
-0.5
-1.0
-1.5
-2
-1 0 1 2 -3
-4 -2 0 0 2 4
x component 1

They are vectors with respect to datapoints.



They are in the same direction as the main data points. It seems that the variance of the point in regard to the principal components is preserved, but in regard to the other one, the variance of points is not preserved.





The shape of histograms remains the same; however, the intervals in the horizontal axis are different.

```
Covariance matrix:
```

Eigenvectors

[[ 0.52237162 -0.37231836 -0.72101681 0.26199559]

[-0.26335492 -0.92555649 0.24203288 -0.12413481]

[ 0.58125401 -0.02109478 0.14089226 -0.80115427]

[ 0.56561105 -0.06541577 0.6338014 0.52354627]]

## Eigenvalues

[2.91081808 0.92122093 0.14735328 0.02060771]

Eigenvalues in descending order:

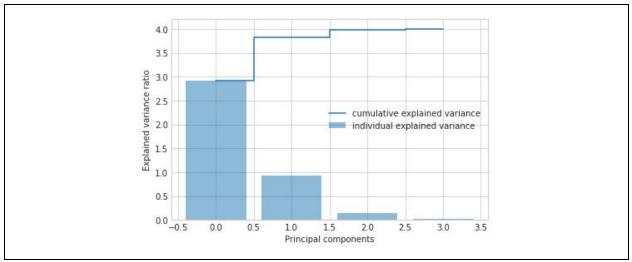
2.9108180837520528

0.9212209307072242

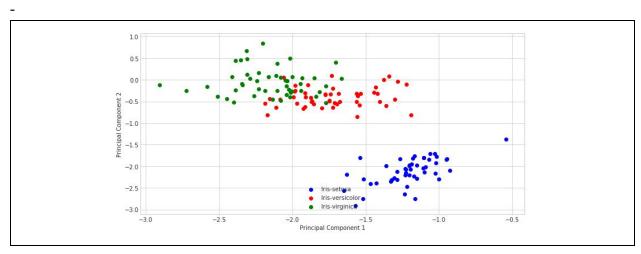
0.14735327830509573

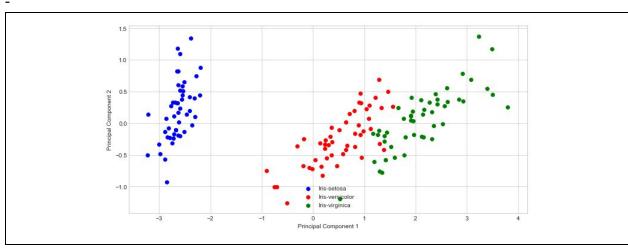
0.020607707235625487

\_

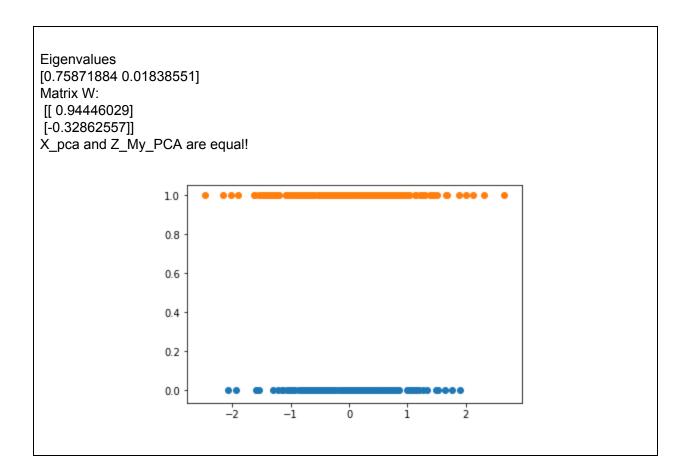


The variance for the two first principal components are large; the other two are not considerable.



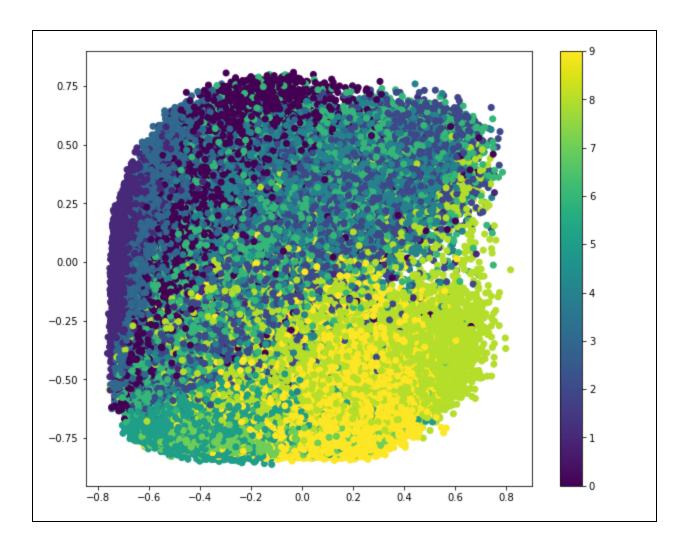


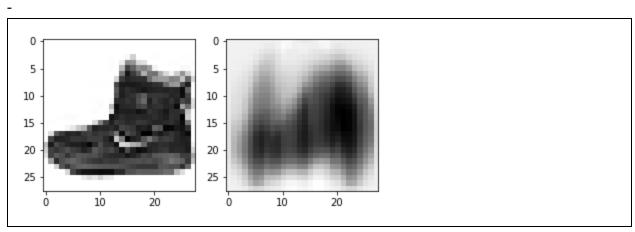
Eigenvectors [[ 0.94446029 -0.32862557] [ 0.32862557 0.94446029]]



## 2) Autoencoders for Dimensionality Reduction

\_





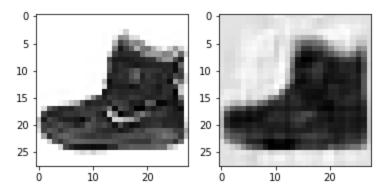
Epoch 1/5 60000/60000 [============] - 8s 138us/step - loss: 0.0186 Epoch 2/5

60000/60000 [========] - 8s 126us/step - loss: 0.0111 Epoch 3/5 60000/60000 [===========] - 5s 77us/step - loss: 0.0104 Epoch 4/5 60000/60000 [==============] - 6s 94us/step - loss: 0.0101 Epoch 5/5 60000/60000 [===============] - 6s 107us/step - loss: 0.0099

Out[28]:

<matplotlib.image.AxesImage at 0x1db41e04160>

<Figure size 720x720 with 0 Axes>



It is much better.

- KNN on raw data:

Time Elapsed for Learning: 0.5202689170837402 Time Elapsed for Prediction: 23.02728009223938

Accuracy: 0.8264

-

```
training for 2 dimensions
Epoch 1/5
Epoch 2/5
Epoch 3/5
7500/7500 [=============================] - 0s 43us/step - loss: 0.0596
Epoch 4/5
7500/7500 [====
             Epoch 5/5
7500/7500 [==============================] - 0s 39us/step - loss: 0.0508
for 2 dims scores are: [0.48369927 0.46333333 0.50500334 0.48098732 0.47765177]
_____
training for 5 dimensions
Epoch 1/5
Epoch 2/5
```

```
Epoch 3/5
7500/7500 [==============================] - 0s 36us/step - loss: 0.0449
Epoch 4/5
Epoch 5/5
for 5 dims scores are: [0.66666667 0.66666667 0.69513009 0.68112075 0.67311541]
_____
training for 7 dimensions
Epoch 1/5
Epoch 2/5
7500/7500 [=============] - 0s 41us/step - loss: 0.0481
Epoch 3/5
7500/7500 [=============================] - 0s 47us/step - loss: 0.0408
Epoch 4/5
Epoch 5/5
_____
training for 10 dimensions
Epoch 1/5
Epoch 2/5
Epoch 3/5
Epoch 4/5
Epoch 5/5
7500/7500 [==================================] - 0s 53us/step - loss: 0.0301
for 10 dims scores are: [0.75582169 0.71866667 0.74583055 0.77051368 0.74916611]
_____
training for 15 dimensions
Epoch 1/5
7500/7500 [=============================] - 0s 62us/step - loss: 0.0710
Epoch 2/5
Epoch 3/5
Epoch 4/5
7500/7500 [=============================] - 0s 58us/step - loss: 0.0284
Epoch 5/5
7500/7500 [=============================] - 0s 54us/step - loss: 0.0264
_____
training for 20 dimensions
```

Epoch 1/5
7500/7500 [===================================
Epoch 2/5
7500/7500 [===================================
Epoch 3/5
7500/7500 [===================================
Epoch 4/5
7500/7500 [===================================
Epoch 5/5
7500/7500 [===================================
for 20 dims scores are : [0.79041916 0.78533333 0.78519013 0.79052702 0.79853235]
best dim: 20
1 111-1

\_