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Read the data from the hands_aligned_train.txt.new file
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with open('./data/hands_aligned_train.txt.new') as newData:
    lines = [line.rstrip('\n') for line in newData]
     data = [np.fromstring(line, dtype=int, sep=' ') for line in lines[1:]]
 # data = np.loadtxt('data/hands_aligned_train.txt.new', dtype=int, comments='#', delimiter=None)
Added the data in x
 x = np.array(np.array(data))
taking x transpose
 x = x.transpose()
 #print(x)
Calculating the mean
 x_mean = np.sum(x,axis=0)/x.shape[0]
 #print(x_mean)
 X = x[:] - x_mean
 #print(X)
Calculating the WWT from slide 49
 XXT = X @ np.transpose(X)
 #print(XXT)
WW^T = U\Sigma V^T V\Sigma U^T = UL^2 U^T
 u, e, v = np.linalg.svd(XXT)
 print(e)
 #e.sort()
 eCollection = 0
Since we need to take the N which is the minimum number of principal components preserving 90% of the
energy.
 N = 1
 for i in range (len(e)):
    eCollection += e[i]
    minimum_energy = eCollection / e.sum()
    if (minimum_energy > 0.9):
      N = i
      break
```

print("N-------",N) #print(u.shape)
$$\begin{split} \mathbf{W}\mathbf{W}^T &= \mathbf{U}\mathbf{L}^2\mathbf{U}^T \\ \hat{\sigma}^2 &= \frac{1}{D-K}\sum_{j=K+1}^D L_{jj}^2 \\ \hat{\Phi} &= \mathbf{U}_K(\mathbf{L}_K^2 - \hat{\sigma}^2\mathbf{I})^{1/2} \\ \end{aligned}$$
 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2 | 1/2

Wi calculation

#print(x_mean.shape)

#print(phi.shape)

$$\mathbf{w}_i \approx \boldsymbol{\mu} + \sum_{k=1}^K \phi_k h_{ik}$$

wi = x_mean + (phi * hi.reshape(1,N)).sum(axis=1)

PLOT

```
fig = plt.figure()
ax = fig.add_subplot()
wi = np.transpose(wi.reshape(wi.shape[0] // 2, 2, order='F'))
ax.plot(wi[0,:],wi[1,:])
plt.show()
```

