Statistical Shape Modelling & Inference (Assignment 2 & Assignment 3)

Importing requried libraries.

```
In [1]:
```

```
import cv2 as cv
import numpy as np
from matplotlib import pyplot as plt
```

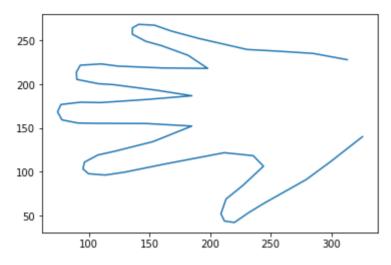
Asignment 2

- In statistical shape modelling, we first calculate the mean shape.
- Then find the covariance matrix of the points.
- Then we find the eigen value and eigen vector subspace.
- Finally we caclulate phi which represents set of eigen vectors.

In [3]:

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QUESTION 2
111
Read the data.
data = np.loadtxt('data/hands aligned train.txt',skiprows=1)
Caclulate mu
mu=np.mean(data,axis=1)
Calculate W = w-mu
W = data - mu.reshape(-1,1)
Caclulate covariance matrix W.Wt
cov = np.cov(W)
111
Find the eigen vectors and eigen value of covariance matrix
eigen value,eigen vectors=np.linalg.eig(cov)
Take the first 5 eigen vectors and find phi and sigma square.
111
K=5
sigma square = 1/(112-K)*(np.sum(eigen value[K+1:]))
L2 = np.diag(eigen value)
second = L2[:K,:K] - sigma square*np.eye(K,K)
phi = np.dot(eigen vectors[:,:K],np.sqrt(second))
Calculate wi with the weights given in the sheet.
wi=(mu + np.dot(phi,np.array([-0.4,-0.2,0,0.2,0.4])).real)
Plot the wi.
plt.suptitle("Statistical Shape Model , K = 5 and W = [-0.4, -0.2, 0, 0.2, 0.4]")
plt.plot(wi[:56],wi[56:])
plt.show()
plt.pause(2)
plt.close()
```

Statistical Shape Model , K = 5 and W = [-0.4, -0.2, 0, 0.2, 0.4]



As shown above, we get the transformed hand using the weights.

Assignment 3

In the inference step we combine the Statistical shape models with ICP. We use the following approach:

- 1. Calculate the w = mu + phi.h
- 2. Find the psuedo inverse components.
- 3. Caclulate psi based on test points.
- 4. Transform w using the new psi.
- 5. Caclulate the components of h.
- 6. Reiterate if shape not converged.

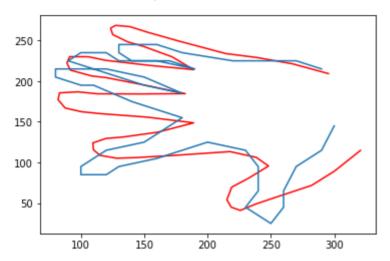
In [6]:

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OUESTION 3
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. . .
Read the test shape
test shape = np.loadtxt('data/hands aligned test.txt', skiprows=1)
h = np.array([0,0,0,0,0]) #Assume random h
h org = np.copy(h)
count iter=0
psi=None
A=None
while(True):
    w = mu + np.dot(phi,h) # w
    w=w.real
    Shape changes to calculate psuedo inverse equation components
    w xy = np.array([np.split(w,2)[0],np.split(w,2)[1]]).T
    A= []
    for point in w xy:
        A = A + [[point[0], point[1], 0, 0, 1, 0]]
        A = A + [[0, 0, point[0], point[1], 0, 1]]
    A = np.array(A)
    test split = np.split(test shape, 2)
    y = np.array([test_split[0], test_split[1]]).T.flatten()
    Calculate transformation vector psi
    psi = np.dot(np.dot(np.linalg.inv(np.dot(A.T, A)), A.T), y)
    psi 1 = psi[:4].reshape(2,2)
    psi 2 = psi[4:]
    Transform points according to psi
    w = np.dot(A,psi).real
    Caclulate parts of the h update equation
    second_term = y - (np.dot(A,psi))
    second term=np.array(np.split(second term, 56)).T.flatten()
    second sum = np.zeros((5,1))
    for n in range (56):
        phi_n=np.array([phi[n, :], phi[n+ 56, :]])
        first n = np.dot(psi 1,phi n).T.real
        second n = np.array([second term[n],second term[n+56]]).reshape(-1,1)
        second sum = second sum + np.dot(first n, second n)
    sum first = np.zeros((5,5))
    for n in range (56):
        phi_n = np.array([phi[n, :], phi[n + 56, :]])
        sum first += phi n.T.dot(psi 1.T).dot(psi 1).dot(phi n).real
    sum_first = sum_first+ sigma_square*np.eye(5,5)
    sum first inv = np.linalg.inv(sum first)
    111
    Evalulate h
```

```
h = np.dot(sum_first_inv,second_sum).real.flatten()
    print(h)
    Check if iterations are done.
    if count iter>15:
        plt.plot(np.array(np.split(w, 56))[:, 0], np.array(np.split(w, 56))[:, 1
], c='r')
        plt.show()
        break
    else:
        h org=h
    count_iter+=1
    Calculate RMS error
    RMS = np.sqrt(np.sum(np.square(w-y)))
    plt.suptitle("Inference Step - RMS" + str(RMS))
    plt.plot(np.array(np.split(w,56))[:,0],np.array(np.split(w,56))[:,1],c='r')
    plt.plot(test_shape[:56],test_shape[56:])
    plt.pause(0.2)
```

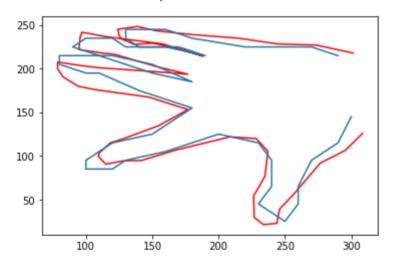
[-0.69036349 -2.20438239 -2.24082408 -1.38536341 1.82349612]

Inference Step - RMS163.03564385904167



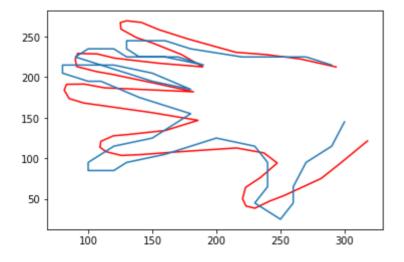
[-0.32635865 -0.49826725 -0.05970978 0.95481052 -0.07318516]

Inference Step - RMS77.45199642212476



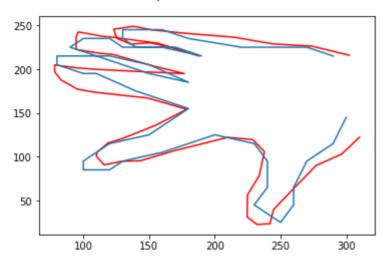
[-0.53549396 -1.89859183 -2.1998935 -1.84020999 1.81081344]

Inference Step - RMS143.82093358893405



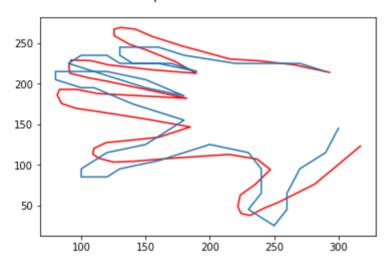
 $[-0.38715896 \ -0.67522255 \ -0.09353588 \ 1.12160094 \ -0.03141717]$

Inference Step - RMS87.19347944429104



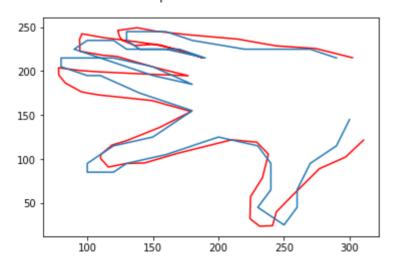
[-0.5044567 -1.78867928 -2.1580744 -1.87882086 1.73072291]

Inference Step - RMS138.27273173590066



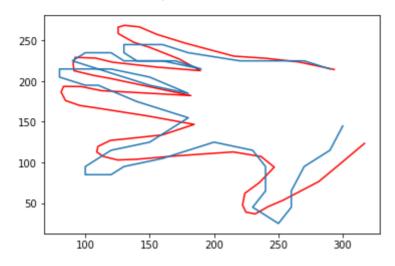
[-0.39868969 -0.75128628 -0.133197 1.09476994 0.05093057]

Inference Step - RMS89.83763108295099



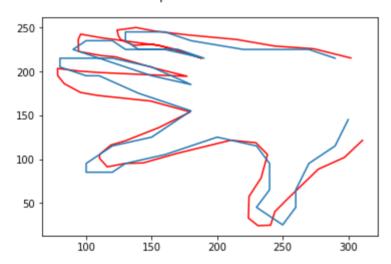
 $[-0.49562411 \ -1.73548558 \ -2.11169983 \ -1.82097766 \ 1.6333582]$

Inference Step - RMS135.4354133465122



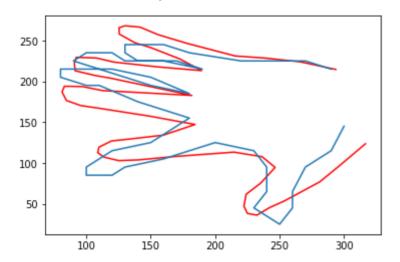
[-0.40131854 - 0.79638264 - 0.1751365 1.02303609 0.137447

Inference Step - RMS90.83496501750953



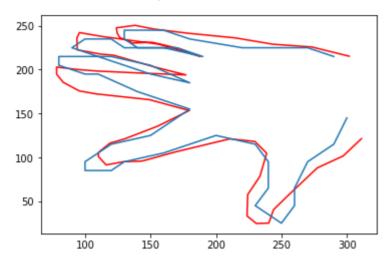
[-0.4909807 -1.69970951 -2.06546437 -1.74269675 1.5407586]

Inference Step - RMS133.31483637900817



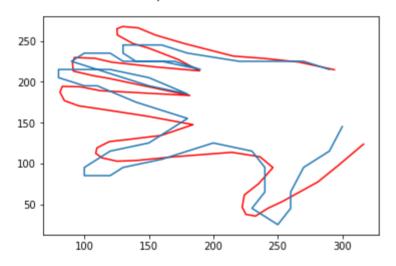
[-0.40248889 -0.83097024 -0.21622592 0.94335921 0.21639123]

Inference Step - RMS91.46567087787632



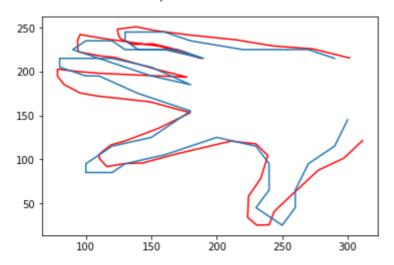
[-0.48721498 -1.67027891 -2.02131182 -1.66268668 1.45860339]

Inference Step - RMS131.48372412960975



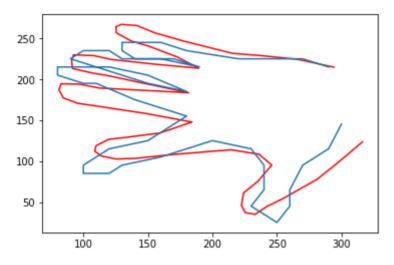
[-0.4035654 -0.86095721 -0.25542417 0.8649807 0.28550992]

Inference Step - RMS92.02161941177964



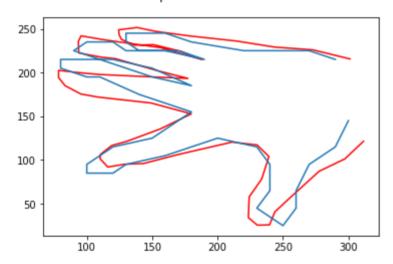
[-0.48370522 -1.64401086 -1.97967466 -1.5855443 1.38713612]

Inference Step - RMS129.83118374042158



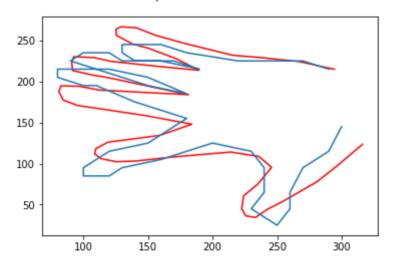
[-0.40476664 - 0.88813685 - 0.29256087 0.79013201 0.34537995]

Inference Step - RMS92.56591698868009



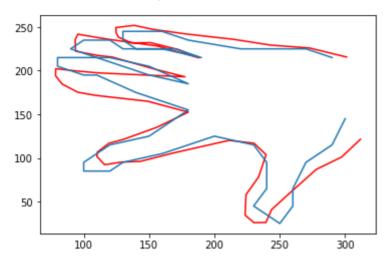
[-0.48034025 -1.61990717 -1.94048373 -1.51229789 1.32519032]

Inference Step - RMS128.31837687440856

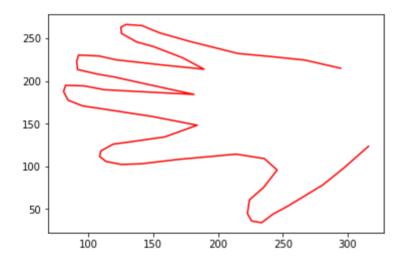


[-0.40609403 -0.91314856 -0.32772906 0.71924442 0.39720472]

Inference Step - RMS93.10874737777215



[-0.47710756 -1.59757918 -1.90354624 -1.44306137 1.27140758]



As see above, the shape gradually converges toward the given test shape.