CS 4442b Assignment 1

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Question 1

A)
$$P(\text{Water} = \text{warm} \mid \text{Play} = \text{yes}) = 2/3$$

$$P(\text{Water} = \text{warm} \mid \text{Play} = \text{no}) = 1$$
B)
$$P(\text{Play} = \text{yes} \mid \text{Water} = \text{warm}) = 2/3$$

$$P(\text{Play} = \text{no} \mid \text{Water} = \text{warm}) = 1/3$$
C)
$$P(\text{Play} = \text{yes} \mid \text{Forcast} = \text{same}) = 1$$

$$P(\text{Play} = \text{yes} \mid \text{Forcast} = \text{change}) = 1/2$$
D) With Laplace smoothing
$$P(\text{Water} = \text{warm} \mid \text{Play} = \text{yes}) = 3/5$$

$$P(\text{Water} = \text{warm} \mid \text{Play} = \text{no}) = 2/3$$

Question 2

Part A

$$k(x,z) = \beta k_1(x,z) - \alpha k_2(x,z)$$

This kernel is symmetric since both k_1 and k_2 are valid kernels and symmetric themselves. Then

$$k(z, x) = \beta k_1(x, z) - \alpha k_2(x, z)$$

= $\beta k_1(z, x) - \alpha k_2(z, x)$
= $\beta k_1(x, z) - \alpha k_2(x, z) = k(x, z)$

However this violates Mercer's theorem for the matrix K being positive and semidefinite whenever $\alpha \geq \beta$ and $k_2(x, z) < k_1(x, z)$

Part B

$$k(x,z) = k_1(x,z)k_2(x,z)$$

This kernel is symmetric.

$$k(z,x) = k_1(z,x)k_2(z,x)$$

= $k_1(x,z)k_2(x,z) = k(x,z)$

Since both the other kernels are symmetric.

Finally this kernel is positive semidefinite, since both k_1 and k_2 are positive semidefinite, then k will also be positive semidefinite for all values.

Part C

$$k(x, z) = f_1(x)f_1(z) + f_2(x)f_2(z)$$

Let $\phi(x) = [f_1(x), f_2(x)]^T$ then

$$k(x, z) = \phi^T(x)\phi(z)$$

$$= [f_1(x), f_2(x)] \begin{bmatrix} f_1(z) \\ f_2(z) \end{bmatrix}$$

$$= f_1(x)f_1(z) + f_2(x)f_2(z)$$

By defining the kernel as above, it is easy to see that it is symmetric

$$k(z, x) = \phi^{T}(z)\phi(x)$$

$$= [f_{1}(z), f_{2}(z)] \begin{bmatrix} f_{1}(x) \\ f_{2}(x) \end{bmatrix}$$

$$= f_{1}(x)f_{1}(z) + f_{2}(x)f_{2}(z) = k(x, z)$$

For any vector $a = [a_1, \dots, a_n]^T \epsilon R^n$ and $\phi(x)$ We have

$$aKa = \sum_{i=1}^{m} \sum_{j=1}^{m} a_i K_i j a_j$$

=
$$\sum_{i=1}^{m} \sum_{j=1}^{m} a_i (\phi(x_i) \phi(x_j)) a_j$$

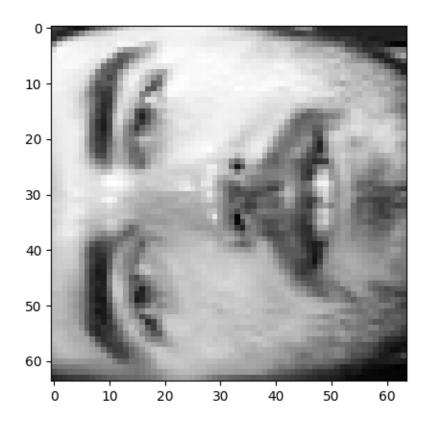
=
$$(\sum_{i=1}^{m} a_i \phi(x_i))^2 \ge 0$$

Therefore since the kernel is both symmetric and positive semidefinite, this is a valid kernel.

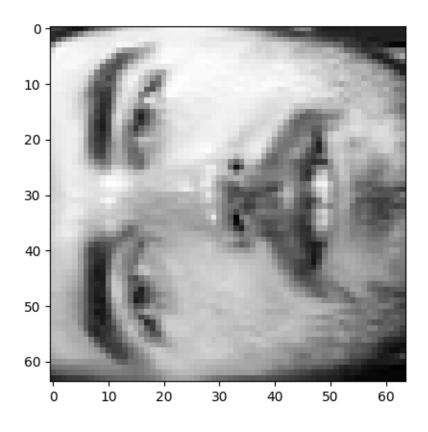
Question 3

Part A

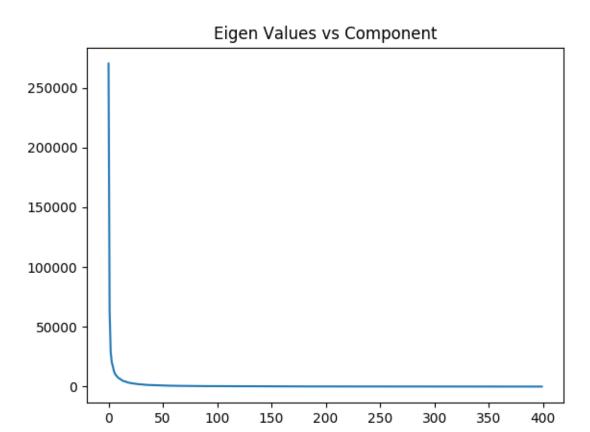
The 100-th image $\,$



Part B $\label{eq:bart} \mbox{After removing the mean } (m \approx 132.384...)$



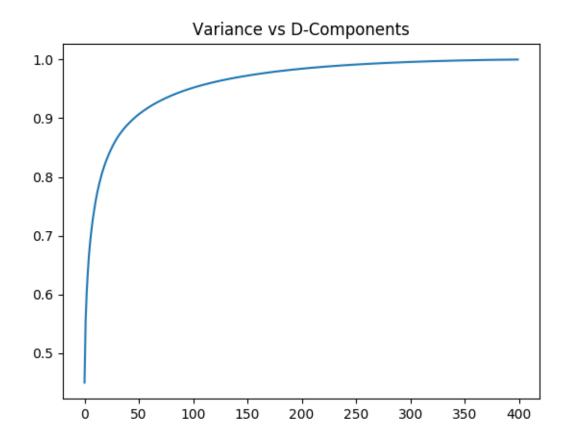
Part C



Part D

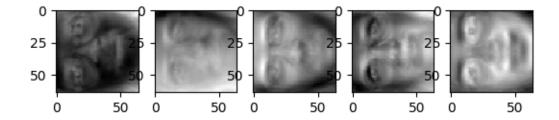
The eigen value computed for the 400-th component is $1.8307756442497e-23\approx 0$. Note the reason it is not zero exactly is due to either numpy using n-1 in its eigen value decomposition or due to floating point arithmetic percision. However, ignoring that fact, the reason that the n-th eigen value is 0, is due to the nature of the covariance matrix used for the components. When the k-th component is equal to the size of the set, the only possible eigen value is then 0, since there are no more dimensions that could be seperated from the n dimension data set. This is the point at which all previous variance has been accounted for.

Part E



To select the dimensionality of the data you would select a principle component which selects for most of the variance. From above any x value greater than 200 would suffice.

Part F



Part G

