

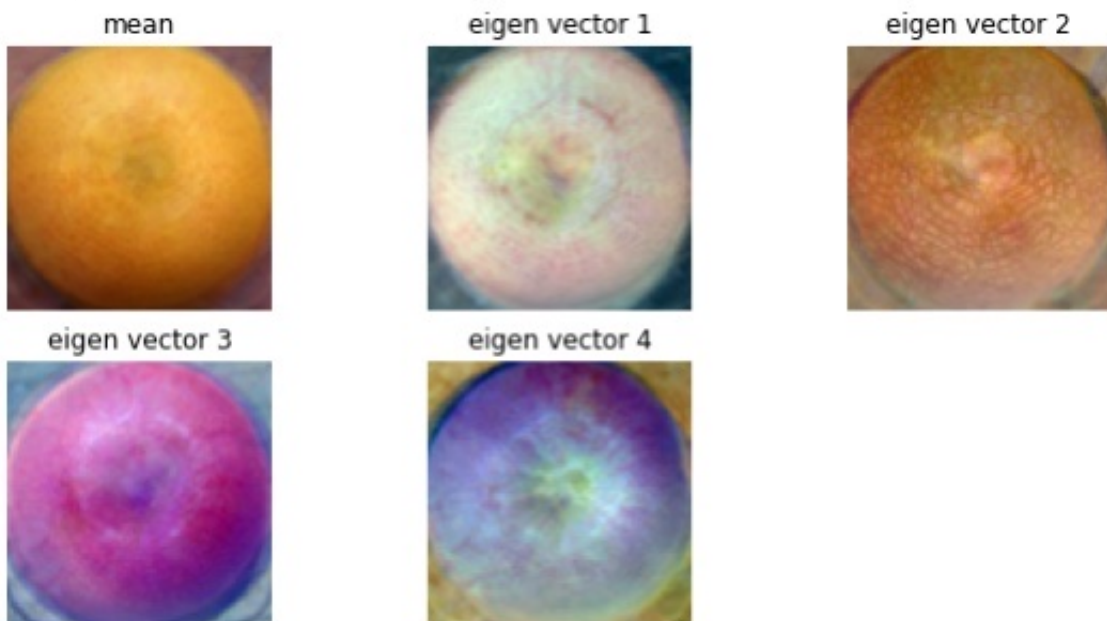
6. `codes → qc.py, qc.ipynb.`

a. Similar to the analysis done in the prev. question

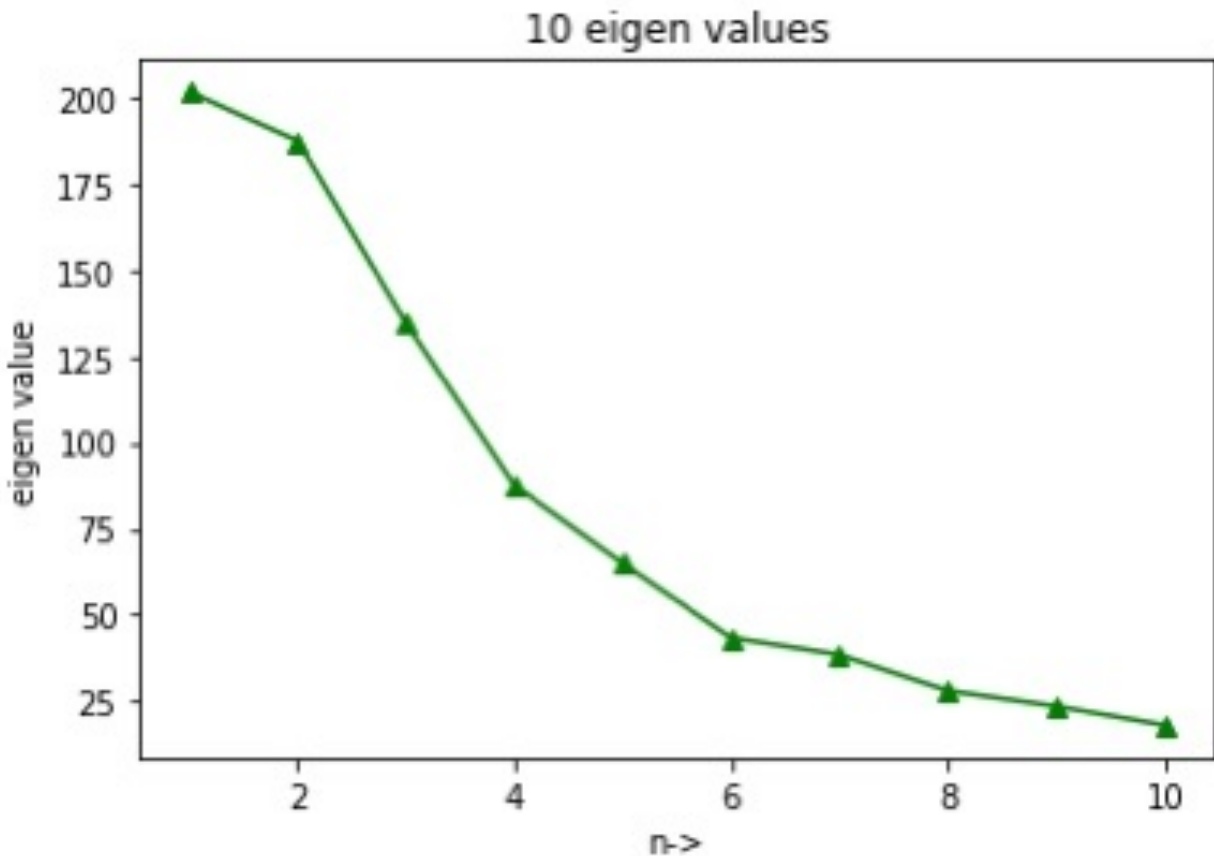
For this part we calculated the mean and covariance.

Also we calculated the top 4 principle eigen vectors and showed them as images in the same plot below

In the following plot the first image is for mean and the rest are for eigen vectors.



Also we have found the top 10 eigen values and plotted them in the following graph.



b. Let A be a vector image and

$\alpha_1 v_1 + \alpha_2 v_2 + \alpha_3 v_3 + \alpha_4 v_4$ be the closest representation and

$\|A - \alpha_1 v_1 - \alpha_2 v_2 - \alpha_3 v_3 - \alpha_4 v_4\|_2$ be the norm of closeness

First we obtain that $\langle v_i, v_j \rangle = 0$ if $i \neq j$;

We claim that

$$\eta_1 = \langle AV_1 \rangle$$

$$\eta_2 = \langle AV_2 \rangle$$

$$\eta_3 = \langle AV_3 \rangle$$

$$\eta_4 = \langle AV_4 \rangle$$

Best guess of a vector if we know the eigen components and the PCA reduction at that vector.

Definition:

$\langle AV \rangle = A^T V$ is like a dot product at A, V
column vectors

Frobenius Norm:

$\|A\|_2$ for one-column is same as Frobenius but

$$\|A\|_{Fro} = \sqrt{\sum_{i=1}^n \sum_{j=1}^n |a_{ij}|^2}$$

Let $\mu + \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n$ be the best approximation we can make by minimizing Frobenius norm where

$$\alpha_1, \alpha_2, \dots, \alpha_n \in \mathbb{R}$$

v_1, v_2, \dots, v_n are eigen vecs / principal directions.

$\mu \rightarrow$ mean.

\rightarrow Let A be the original vector

$\|A - (\mu + \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n)\|_F$ should be minimum.

We see that ,

$$\|A - (\mu + \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n)\|_F^2 = \|A - \mu - (\alpha_1 v_1 + \dots + \alpha_n v_n)\|_F^2$$

$$= \text{trace} \left(\langle (A - \mu) - (\alpha_1 v_1 + \dots + \alpha_n v_n), (A - \mu) - (\alpha_1 v_1 + \dots + \alpha_n v_n) \rangle \right)$$

Observe that $\langle v_i, v_j \rangle = 0$ $\left[\begin{array}{l} \text{Symmetric matrix} \Rightarrow \\ \text{eigen vecs are orthogonal} \end{array} \right]$

$$= \text{trace} \left(\langle A-u \ A-u \rangle + \sum_{i=1}^n \eta_i^2 \langle V_i V_i \rangle - \eta_i \langle V_i A-u \rangle - \eta_i \langle A-u V_i \rangle \right)$$

$$= \text{trace} \left(\langle A-u \ A-u \rangle + \sum_{i=1}^n \eta_i^2 \langle V_i V_i \rangle - 2\eta_i \langle V_i A-u \rangle \right)$$

$$= \text{trace} \left(\langle A-u \ A-u \rangle + \sum_{i=1}^n \eta_i^2 - 2\eta_i \langle A-u V_i \rangle \right)$$

should be minimum.

We can clearly see that for (1-D)
(1-column matrix) $\text{trace}(\langle AB \rangle) = \langle AB \rangle$ and for

$$\sum_{i=1}^n \eta_i^2 - 2\eta_i \langle V_i A-u \rangle \text{ to be minimum}$$

$$\eta_i = \langle V_i A-u \rangle = \langle A-u V_i \rangle$$

Hence the best approximation of A is

$$u_1 + \sum_{i=1}^n \langle A-u V_i \rangle V_i$$

where $\langle A - \mu | V_i \rangle = i^{\text{th}}$ value of PCA analysis of A

$$\text{PCA}(A) = (A - \mu)^T \cdot v = J$$

$$\langle A - \mu | V_i \rangle = J[0, i] \Rightarrow 1 \times n \text{ vector.}$$

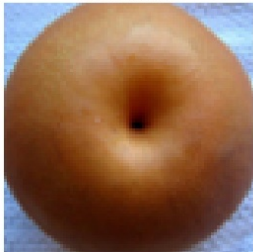
Best approximation gives PCA J is

$$\mu + \sum_{i=1}^n J[0, i] V_i$$

Hence we can see that PCA component which are determined by $\langle A | V_i \rangle$ are the best approximation to that coordinates

\Rightarrow The following are image retransformed compared to original ones.

original image1



approximate image1



original image2



approximate image2



original image3



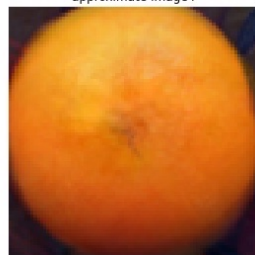
approximate image3



original image4



approximate image4



original image5



approximate image5



original image6



approximate image6



original image7



approximate image7



original image8



approximate image8



original image9



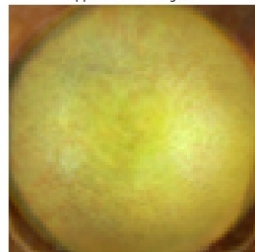
approximate image9



original image10



approximate image10



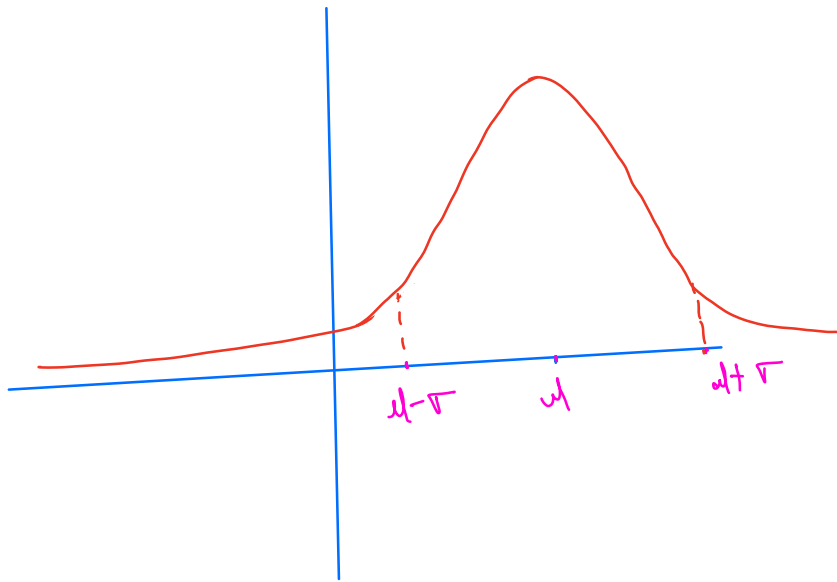


C. Generating random Images ,

$$e \leftarrow k_1 v_1 + k_2 v_2 + k_3 v_3 + k_4 v_4$$

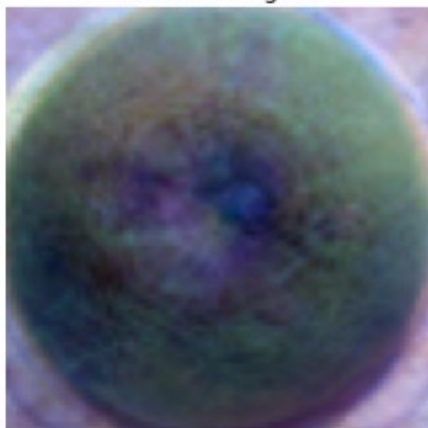
best random image has variance
 λ_1 in that axis.

It is best to have $\sqrt{\lambda_1} < k_1 < \sqrt{\lambda_1}$ to get an image in an acceptable range because just like a gaussian has most of its probability inside the width of $\mu - \sigma$ to $\mu + \sigma$ where σ is the standard deviation similarly $\sqrt{\lambda_1}$ is like standard deviation.



→ Using this algorithm in code q6.py to generate 4 random images.

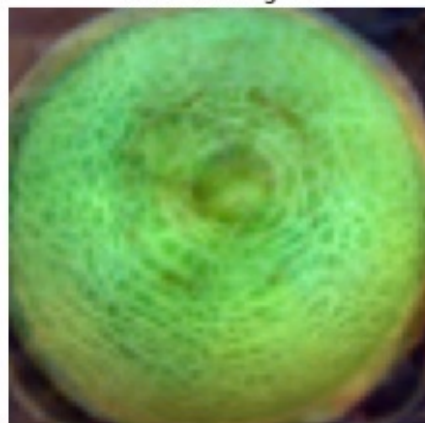
random image:1



random image:2



random image:4



random image:3

