Abgabe - Übungsblatt [4]

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1. Dezember 2020

Aufgabe 1

Here comes your text ...

Aufgabe 2

```
import numpy as np
from numpy.linalg import eig
from math import sqrt
def ComputeSVD(A:np.matrix):
    """Given a matrix A use the eigen value decomposition
        to\ compute\ a\ SVD
       decomposition. It returns a tuple (U, Sigma, V) of
          np.\ matrix \ objects."""
    k = np. linalg.matrix_rank(A)
    B = A. transpose()@A
    w, V = eig(B)
    #Eigenwerte und Vektoren neu sortieren
    idx = np.argsort(w)
    w = w[idx]
    V = V[:, idx]
    #S berechnen
    S = np.zeros(A.shape)
    for i in range (S. shape [0]):
        for j in range(S.shape[1]):
            i f i==j:
                S[i][j] = sqrt(w[i])
    #U berechnen
    U = np.zeros((k+1,k+1))
    for i in range(k):
        U[:, i] = (1/S[i][i] * A * V[:, i]). flat
    U = np.linalg.qr(U)[0]
```

```
return np. asmatrix (U), np. asmatrix (S), np. asmatrix (V)
def PseudoInverse(A:np.matrix):
    """Given a matrix A use the SVD of A to compute the
        pseudo inverse. It returns the pseudo inverse as a
         np.\ matrix\ object.""
    U, S, V = ComputeSVD(A)
    \#invert S
    S_{inv} = S.H
    for i in range(S_inv.shape[0]):
         S_{inv}[i,i] = 1/S_{inv}[i,i]
    return np.asmatrix (V*S_inv*U.H)
def LinearSolve (A:np.matrix, b:np.ndarray):
    """Given a matrix A and a vector b this function
        solves the linear equations
       A*x=b by solving the least squares problem of
            minimizing | A*x-b | and
        returns the optimal x.""
    x = PseudoInverse(A)*b
    return x
if (__name__ == "__main__"):
    # Try the SVD decomposition
    A = np.matrix([
         [1.0, 1.0, 1.0],
         [1.0, 2.0, 3.0]
         [1.0, 4.0, 9.0],
         [1.0, 8.0, 27.0]
    U, Sigma, V = ComputeSVD(A)
    print(U)
    print(Sigma)
    print(V)
    print ("If _the _following _numbers _ are _ nearly _zero , _SVD_
        seems_to_be_working.")
    \mathbf{print} (np. linalg.norm (U*Sigma*V.H - A))
    \mathbf{print} (np. linalg.norm(U.H*U-np.eye(4))) #
                                                      berprft
        eure Framewroks doch mal auf Fehler...
    \mathbf{print} \, (\, \mathrm{np.\,linalg.norm} \, (\mathrm{V.H*V\!-\!np.\,eye} \, (\, 3\, )\, )\, )
    # Try solving a least squares system
    b = np. matrix([1.0, 2.0, 3.0, 4.0]).T
    x = LinearSolve(A, b)
    print("If _the _following _number_is _nearly _zero, _linear
        _solving_seems_to_be_working.")
    print (np. lin alg. norm (x-np. lin alg. lstsq (A, b, rcond=None
        ) [0])
```

Aufgabe 3

Here comes your text ...

Aufgabe 4

```
import numpy
import numpy as np
from PIL import Image
from matplotlib import cm, pyplot
def Compress(Image, ComponentCount):
             """This function uses a singular value decomposition
                       to compress an image.
                  \param Image An quadratic array providing the image
                             . \quad Entries \quad provide \quad the
                                        brightness \ of \ indidividual \ pixels \ , \ rows
                                                  correspond to scanlines.
                  \param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\param\par
                               to be maintained in the
                                        compressed\ representation .
                  CompressionRatio) such that U*Sigma*V^*
                                           provides an approximation to the original
                                                     image\ when\ Sigma\ is\ a
                                           diagonal matrix with Singular Values on its
                                                     main diagonal.
                                           Compression Ratio\ should\ provide\ the
                                                     quotient of the number of scalars
                                           in Image and the number of scalars in the
                                                     returned representation of
                                           Image."""
def Decompress (U, Singular Values, V):
             """ Given \ a \ compressed \ representation \ of \ an \ image \ as
                      produced by Compress() this
                     function \ reconstructs \ the \ original \ image
                                approximately and returns it."""
if(__name__="__main__"):
           # Define the task
```

```
ImageFileNameList=["Lena", "Stoff", "Stoff2"];
ComponentCountList = [1,4,8,32,64];
# Iterate over all tasks and generate one large plot
PlotIndex = 1;
for ImageFileName in ImageFileNameList:
    ImagePath=ImageFileName+".png";
    img=Image.open(ImagePath);
    # Convert to numpy array
    imgmat = np.array(list(img.getdata(band=0)),
       float)
    # Reshape according to orginal image dimensions
    imgmat.shape = (img.size[1], img.size[0])
    imgmat = np.matrix(imgmat)
    for ComponentCount in ComponentCountList:
        # Define a subplot for this decompressed
            image
        Axes=pyplot.subplot(len(ImageFileNameList),
           len(ComponentCountList), PlotIndex);
        Axes.set_xticks([]);
        Axes.set_yticks([]);
        Axes. set_title (ImageFileName+", _p="+str (
           ComponentCount));
        PlotIndex += 1;
        # Apply compression
        U, Singular Values, V, Compression Ratio=Compress (
           imgmat , ComponentCount ) ;
        # Apply decompression and show the result
        DecompressedImage=Decompress(U, Singular Values
        pyplot.imshow(DecompressedImage,cmap=cm.gray)
        # Compute and print the compression ratio
        print("Compression_ratio_for_p="+str(
           ComponentCount) + " \_ is \_" + str (
           CompressionRatio)+":1.");
    print("");
pyplot.show();
```