In [1]: import numpy as np
from PIL import Image

## Load the image

In [2]: hm\_img = Image.open('hidden\_message.png')
hm\_img

Out[2]:



```
In [3]: hm_img_mat = np.array(list(hm_img.getdata()), float)
hm_img_mat.shape
```

Out[3]: (691200, 3)

## **Extract each of the RGB channels**

In [4]: hm\_imgR = hm\_img.getchannel(0) # Red channel in L-mode (Luminance)
hm\_imgR

Out[4]:



```
In [5]: hm_imgG = hm_img.getchannel(1)  # Green channel in L-mode (Luminance)
hm_imgB = hm_img.getchannel(2)  # Blue channel in L-mode (Luminance)
Convert the channels into numpy array to make it easy to manipulate
```

```
In [6]:
         hm_imgR_mat = np.array(hm_imgR)
         print(hm_imgR_mat.shape)
         print('\n', hm_imgR_mat)
         (960, 720)
          [[248 248 246 ... 159 160 164]
           [246 246 249 ... 166 167 162]
           [244 244 247 ... 166 170 170]
           [215 217 208 ... 236 235 236]
           [217 214 214 ... 234 233 234]
           [225 218 213 ... 235 233 231]]
 In [7]: | hm_imgG_mat = np.array(hm_imgG)
         hm_imgG_mat.shape
 Out[7]: (960, 720)
 In [8]: | hm_imgB_mat = np.array(hm_imgB)
         hm_imgB_mat.shape
 Out[8]: (960, 720)
         Create numpy arrays to store pixel values of message image
 In [9]: | dimension_R1, dimension_R2 = hm_imgR_mat.shape
         msg R = np.empty(shape = (dimension R1//4, dimension R2//4))
         msg_R.shape
 Out[9]: (240, 180)
In [10]: | dimension_G1, dimension_G2 = hm_imgG_mat.shape
         msg_G = np.empty(shape = (dimension_R1//4, dimension_R2//4))
         msg_G.shape
Out[10]: (240, 180)
In [11]: | dimension_B1, dimension_B2 = hm_imgB_mat.shape
         msg_B = np.empty(shape = (dimension_R1//4, dimension_R2//4))
         msg_B.shape
Out[11]: (240, 180)
```

Normalize image channel arrays

```
In [12]: def normalize_imgmat(imgmat):
               Min-Max scaling
               Values between [0-1]
             min_value = imgmat.min()
             max_value = imgmat.max()
             normalized_imgmat = (imgmat - min_value) / (max_value - min_value)
              return normalized imgmat
In [13]: hm_imgR_norm = normalize_imgmat(hm_imgR_mat)
          hm imgR norm
Out[13]: array([[0.97254902, 0.97254902, 0.96470588, ..., 0.62352941, 0.62745098,
                  0.64313725],
                 [0.96470588, 0.96470588, 0.97647059, ..., 0.65098039, 0.65490196,
                  0.63529412],
                 [0.95686275, 0.95686275, 0.96862745, ..., 0.65098039, 0.66666667,
                  0.66666667],
                 [0.84313725, 0.85098039, 0.81568627, ..., 0.9254902, 0.92156863,
                  0.9254902 ],
                 [0.85098039, 0.83921569, 0.83921569, ..., 0.91764706, 0.91372549,
                  0.91764706],
                 [0.88235294, 0.85490196, 0.83529412, ..., 0.92156863, 0.91372549,
                  0.90588235]])
In [14]: hm imgG norm = normalize imgmat(hm imgG mat)
         hm_imgB_norm = normalize_imgmat(hm_imgB_mat)
          Compute singular values for each non-overlapping 4 * 4 block in the image
In [15]: def find block sv(arr ch, msg ch, T):
                Find the singular values of each block as \sigma1, \sigma2, \sigma3, \sigma4, sorted in descending
                Encode 1 "bit" of information as the difference between \sigma2 and \sigma3 wrt a threst
               Map the resulting bit value to the corresponding message pixel
```

```
In [15]: def find_block_sv(arr_ch, msg_ch, T):
    Find the singular values of each block as \( \sigma 1, \sigma 2, \sigma 3, \sigma 4, \) sorted in descending Encode 1 "bit" of information as the difference between \( \sigma 2 \) and \( \sigma 3 \) wrt a threst Map the resulting bit value to the corresponding message pixel

'''

n_rows, n_cols = msg_ch.shape
for i in range(n_rows):
    for j in range(n_cols):
        block = arr_ch[4 * i: 4 * (i + 1), 4 * j: 4 * (j + 1)]
        U, sigma, V = np.linalg.svd(block)
        sig_1, sig_2, sig_3, sig_4 = sigma
        msg_pix_val = 0
        if sig_2 - sig_3 > T:
              msg_pix_val = 1
              msg_ch[i][j] = msg_pix_val
        return msg_ch
```

```
In [16]: T = 0.004 # Constant
```

```
msg_uint8 = (np.dstack((msg_R, msg_G, msg_B)) * 255.999) .astype(np.uint8)
    # The reason for using 255.999 instead of 255 is to ensure that rounding errors
    # Multiplying by 255 would round down values less than 1 to 0, which might resul
msg_uint8.shape
```

```
Out[19]: (240, 180, 3)
```

```
In [20]: msg_img = Image.fromarray(msg_uint8)
msg_img
```

Out[20]:

```
Why do mathem-aticians enjoy hiking?
```

## Bonus - A play-on-words response

Mathematicians enjoy hiking because they are "add-venturous" "math-letes" who embark on a "pi-cnic" to find the "X" that marks the "vertex" with the most "sinfully" beautiful "derivative" of the scenery and "acute" "tan-gential" insights "calculated" on the minimum "cos-t" "path", all while avoiding "complex" wildlife, "irrational" obstacles and "undefined" trails!

```
In [20]:
In [20]:
```