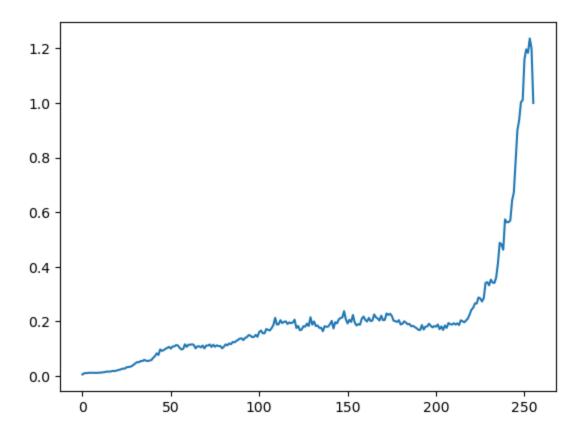
```
In [ ]: import numpy as np
        import matplotlib.pyplot as plt
        import skimage as skimage
        import scipy
        from scipy.sparse import csr array
        from scipy.sparse import lil_array
        from scipy.sparse.linalg import lsqr
        # note we imported the sparse least squares function
In [ ]: # take some images with different exposure times.
        # make sure your camera (phone) does not move between exposures. I used a phone tri
        # images
        Ims = []
        # exposure times
        dts = []
        Ims.append(plt.imread('/Users/yashp/OneDrive/Desktop/UCI Stuff/Junior/Winter/CS 116
        dts.append(0.005)
        Ims.append(plt.imread('/Users/yashp/OneDrive/Desktop/UCI Stuff/Junior/Winter/CS 116
        dts.append(0.02)
        Ims.append(plt.imread('/Users/yashp/OneDrive/Desktop/UCI Stuff/Junior/Winter/CS 116
        dts.append(0.1)
        Ims.append(plt.imread('/Users/yashp/OneDrive/Desktop/UCI Stuff/Junior/Winter/CS 116
        dts.append(1.0)
In [ ]: # resize the images to a reasonable scale
        Imsr = []
        for idx in range(len(Ims)):
            # note this will resize factor of 1/8 for each side and convert to type float
            # "cheap" conversion from color to gray scale by taking the mean of rgb values
            I = np.mean(Ims[idx],axis=2)
            # resize to 1/8 of each dimension
            Ir=skimage.transform.resize(I, (I.shape[0] // 8, I.shape[1] // 8), anti_aliasin
            Imsr.append(Ir)
            print(f"Imsr[{idx}].shape={Imsr[idx].shape} Imsr[{idx}].dtype={Imsr[idx].dtype}
            print(f"values from {np.min(Imsr[idx][:])} to {np.max(Imsr[idx][:])}")
            # convert to uint8 (forcing a value from 0..255) just to have discrete pixel va
            Imsr[idx]=Imsr[idx].astype("uint8")
            print(f"Imsr[{idx}].shape={Imsr[idx].shape} Imsr[{idx}].dtype={Imsr[idx].dtype}
            print(f"values in {np.min(Imsr[idx][:])} .. {np.max(Imsr[idx][:])}")
            print(f"dts[{idx}]={dts[idx]}")
```

```
Imsr[0].shape=(378, 378) Imsr[0].dtype=float64 values from 0.0 to 53.390115578737266
Imsr[0].shape=(378, 378) Imsr[0].dtype=uint8 values in 0 .. 53
dts[0]=0.005
Imsr[1].shape=(378, 378) Imsr[1].dtype=float64 values from 0.0 to 194.26949488627437
Imsr[1].shape=(378, 378) Imsr[1].dtype=uint8 values in 0 .. 194
dts[1]=0.02
Imsr[2].shape=(378, 378) Imsr[2].dtype=float64 values from 12.815263291343918 to 25
0.5942364892112
Imsr[2].shape=(378, 378) Imsr[2].dtype=uint8 values in 12 .. 250
dts[2]=0.1
Imsr[3].shape=(378, 378) Imsr[3].dtype=float64 values from 13.932412441285335 to 25
4.90864716257929
Imsr[3].shape=(378, 378) Imsr[3].dtype=uint8 values in 13 .. 254
dts[3]=1.0
```

```
In [ ]: # simple version of optimization
        # solving for entries in v
        # first 256 entries are for g(0) ... g(255) then next h*w entries are for pixel bri
        # Z_ij is the pixel value at location i in image j
        # dt_j is the exposure time for image j
         # g(z) is the energy x that the sensor receives (exposure) to produce pixel value z
                (the exposure is the irradiance R times exposure time)
        \# x = R*dt
        # for pixel i in image j
        \# x_{ij} = R_{ij}*dt_{j}
        # Z_{ij} = f(x_{ij}) # the pixel value Z_{ij} comes from mapping the exposure through s
        \# g() is the inverse of f()
        # so if Z_{ij} = f(x_{ij}) we want g(Z_{ij}) = x_{ij} (=R_{ij}*dt_{j})
        # taking logs of everything:
        \# g(Z_{ij}) = ln(R_{i}) + ln(dt_{j}) \# pixel i, image j with time dt_{j}
         \# g(Z_{ij}) - ln(R_{i}) = ln(dt_{j})
        # making this into a constraint
        # Each pixel i in imag j gives one row of A and one entry in b
         # position: 0..... Z_{ij} .... 255 0 1 2 ... i*w+j ... h*w
        # coefficient: 0 0 ... 1......0 0 ....... -1 .....0
        # entry in b: ln(dt_j)
        # We want to solve for v so that Av=b
        # or the v that minimizes |Av-b|^2
         [h,w]=Imsr[0].shape
        l = len(Imsr)
        # Initialize A as a lil matrix with the appropriate size
        A_{rows} = h * w * l + 1
         \# A_{cols} = 256 + h * w * l + 1
```

```
A cols = 256 + h * w
        A = lil_array((A_rows, A_cols))
        print(A.shape)
        # Initialize b as a numpy array of zeros
        b = np.zeros((A_rows,))
        # setup rows of A and entries in b
        # Loop through each pixel in each image
        # Number of images
        idx = 0
        for j in range(1):
            # Number of rows in each image
            for y in range(h):
                # Number of columns in each image
                for x in range(w):
                    # Calculate the index for the current pixel
                    idx = j * h * w + y * w + x
                    # print(f"idx: {idx}, A_cols: {A_cols}, 256 + idx: {256 + idx}, j * h *
                    # Calculate the pixel value Z_ij from the resized images
                    Z_{ij} = Imsr[j][y, x]
                    # Add a row to A and an entry to b for the current pixel
                    A[idx, Z_{ij}] = 1
                    A[idx, 256 + y*x] = -1
                    b[idx] = np.log(dts[j])
        print("Final index is :", idx)
        # add a constraint that "fixes" g(128) to be say 4 or such
        # This is achieved by adding a row to A and an entry to b
        constraint_idx = A_rows - 1 # Index for the last row (constraint row, empty)
        A[constraint_idx, 128] = 1 # Constraint for fixing g(128)
        A[constraint_idx, 256:] = 0 # Ensure only the g(128) constraint is added to the la
        b[constraint_idx] = 4
       (571537, 143140)
       Final index is : 571535
In [ ]: #solve for least squares solution
        Acsr=csr_array(A)# convert a to a csrarray before calling least squares
        soln = lsqr(Acsr,b,atol=1e-07, btol=1e-07)
        v = soln[0]
In [ ]: plt.plot(np.exp(v[0:256]))
Out[ ]: [<matplotlib.lines.Line2D at 0x1d168178890>]
```



```
In []: # Calculate regularization weights:
    # Hardcode regularization_weight for now
    regularization_weights = []
    def calculate_regularization_weight(image, size_of_image, scaling_factor):
        # Calculate the standard deviation of pixel values in the region
        noise_level = np.std(image)

        # Scale the noise level to determine the regularization weight
        regularization_weight = scaling_factor * noise_level

        return regularization_weight

for image in Ims:
        size_of_image = (377, 377, 377, 377)
        scaling_factor = 0.1
        regularization_weights.append(calculate_regularization_weight(image, size_of_im
        print(regularization_weights)
```

## [0.5162206697523292, 4.152722714767378, 7.529810491568016, 8.329813961480818]

```
In []: # Adding some bells and whistles to the optimization setup # # # solving for entries in v # first 256 entries are for g(0) ... g(255) then next h*w entries are for pixel bri # g(x) = \ln(R_i) + \ln(dt_j) # pixel i, image j with time dt_j # g(x) - \ln(R_i) = \ln(dt_j) # # weighted version where weight depends on the pixel value (can care less about 0... # w(R_i)g(x) - g(R_i)\ln(R_i) = w(R_i)\ln(dt_j)
```

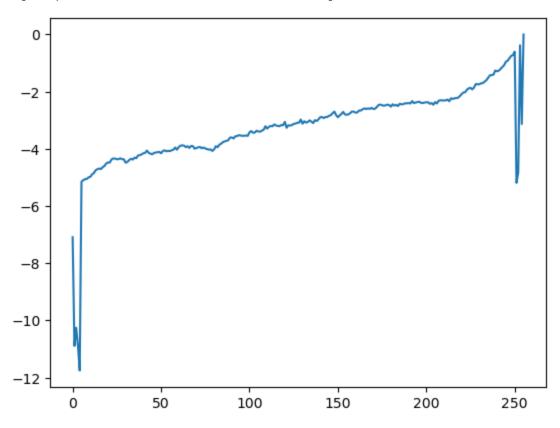
```
# also add regularization so that g(x) tends to be smooth
\# l(g(i+1)-g(i)) - l(g(i)-g(i-1)) = 0
\# lg(i)-2g(i+1)-g(i+2) = 0
# where L is some weight on this constraint
# remember to add a constraint that "fixes" q(128) to be say 4 or such
# Initialize A as a lil_matrix with the appropriate size
A_{rows} = h * w * l + 1
A_{cols} = 256 + h * w
A = lil_array((A_rows, A_cols))
print(A.shape)
# Initialize b as a numpy array of zeros
b = np.zeros((A_rows,))
def weight function(value of pixel):
   if value_of_pixel < 5 or value_of_pixel > 250:
        return 0.1 # Assign a low weight to pixel values less than 5 or greater th
   elif 5 <= value_of_pixel <= 250:</pre>
        return 1.0 # Assign a higher weight to pixel values between 5 and 250
   else:
        return 0.5 # Default weight for other pixel values
# Loop through each pixel in each image
for j in range(1):
   for y in range(h):
        for x in range(w):
            # Calculate the index for the current pixel
            idx = j * h * w + y * w + x
            # print(f"idx: {idx}, A_cols: {A_cols}, 256 + idx: {256 + idx}, j * h *
            # Calculate the pixel value Z_ij from the resized images
            Z_{ij} = Imsr[j][y, x]
            # Add a row to A and an entry to b for the current pixel
            A[idx, Z_ij] = weight_function(Z_ij) # Apply weight based on pixel val
            # Regularization term for g(i) - 2g(i+1) + g(i+2)
            if x < w - 2:
                reg_weight = regularization_weights[j]
                A[idx, 256 + y * w + x] = reg_weight
                A[idx, 256 + y * w + x + 1] = -2 * reg_weight
                # lsqr calculation extremely slow with this for some reason
                \# A[idx, 256 + y * w + x + 2] = reg\_weight
            b[idx] = weight_function(Z_ij) * np.log(dts[j]) # Apply weight to b ba
# Add a constraint that "fixes" g(128) to be 4 or such
constraint_idx = A_rows - 1 # Index for the last row (constraint row)
A[constraint_idx, 128] = 1 # Constraint for fixing g(128)
A[constraint_idx, 256:] = 0 # Ensure only the g(128) constraint is added to the La
b[constraint_idx] = 4 # fixed_value_for_g_128
```

(571537, 143140)

```
In [ ]: Acsr=csr_array(A)
soln = lsqr(Acsr,b,atol=1e-07, btol=1e-07)
v=soln[0]
```

```
In [ ]: plt.plot(v[0:256]) # note plotting g(z) here not exp(g(z))
```

Out[ ]: [<matplotlib.lines.Line2D at 0x1d1092e5d10>]



```
In []: f=plt.figure()
    f.set_size_inches(10, 6)
    ax = f.add_subplot(1,5,1)
    ax.imshow(np.fliplr(np.transpose(np.reshape(v[256:],(h,w)))),cmap="gray")
    ax.axis("off")
    for i in range(len(Imsr)):
        ax = f.add_subplot(1,5,i+2)
        ax.imshow(np.fliplr(np.transpose(np.log(Imsr[i]+1))),cmap="gray")
        ax.axis("off")
```





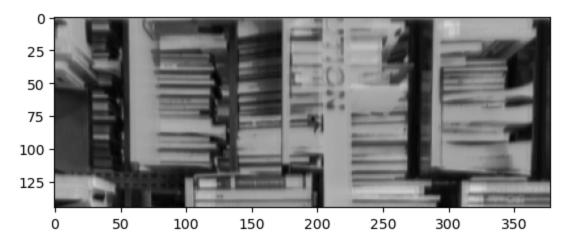






```
In [ ]: da_Im=np.fliplr(np.transpose(np.reshape(soln[0][256:],(h,w))))
    plt.imshow(da_Im[0:145,:],cmap="gray")
```

## Out[]: <matplotlib.image.AxesImage at 0x1d104403950>



```
In []: # show one horixontal line (bottom of image above) in each exposure and the high d
    # note that the hdr image has detail in both the darker and the lighter areas where
    # exposures lack detail in one (e.g. no dark detail in red) or the other (no light
    plt.plot(da_Im[145,:],'k')
    plt.plot(np.flipud(np.log(Imsr[0][:,145])),color='r')
    plt.plot(np.flipud(np.log(Imsr[1][:,145])),color='g')
    plt.plot(np.flipud(np.log(Imsr[2][:,145])),color='b')
    plt.plot(np.flipud(np.log(Imsr[3][:,145])),color='c')
# plotted logs to make it easier to see variations
```

```
C:\Users\yashp\AppData\Local\Temp\ipykernel_17796\2275842349.py:5: RuntimeWarning: d
ivide by zero encountered in log
  plt.plot(np.flipud(np.log(Imsr[0][:,145])),color='r')
C:\Users\yashp\AppData\Local\Temp\ipykernel_17796\2275842349.py:6: RuntimeWarning: d
ivide by zero encountered in log
  plt.plot(np.flipud(np.log(Imsr[1][:,145])),color='g')
```

Out[ ]: [<matplotlib.lines.Line2D at 0x1d104a5ed50>]

