Part B: Coding In [1]: pip install numpy opency-python scipy Defaulting to user installation because normal site-packages is not writeable Requirement already satisfied: numpy in c:\programdata\anaconda3\lib\site-packages (1.23.5) Requirement already satisfied: opencv-python in c:\users\admin\appdata\roaming\python\python310\site-packages (4.9.0.80) Requirement already satisfied: scipy in c:\programdata\anaconda3\lib\site-packages (1.10.0) Note: you may need to restart the kernel to use updated packages. In [2]: import numpy as np import cv2 import matplotlib.pyplot as plt from scipy.fftpack import dct, idct from skimage.metrics import mean_squared_error import math In [3]: # Example Usage Image from PIL import Image # Load image using PIL image_path = 'nature.png' image = Image.open(image_path) # Convert the image to a numpy array image_array = np.array(image) # Display the shape of the image print("Image shape:", image_array.shape) # Display the image plt.imshow(image_array, cmap='gray') # Use cmap='gray' for grayscale images plt.axis('off') # Hide axes plt.show() Image shape: (225, 225, 3) # Helper function to calculate Compression Ratio def compression_ratio(original_size, compressed_size): return original_size / compressed_size # Helper function to calculate RMSE def rmse(original, reconstructed): return math.sqrt(mean_squared_error(original, reconstructed)) A. Transform Coding (using DCT for forward transform) In [5]: def dct_encode(image): # Apply 2D DCT to the image transformed_image = dct(dct(image.T, norm='ortho').T, norm='ortho') return transformed_image def dct_decode(transformed_image): # Apply 2D Inverse DCT to the image reconstructed image = idct(idct(transformed image.T, norm='ortho').T, norm='ortho') return reconstructed image # Example Usage original_image = cv2.imread('nature.png', cv2.IMREAD_GRAYSCALE) transformed_image = dct_encode(original_image) reconstructed_image = dct_decode(transformed_image) # Calculate Compression Ratio (Example) compressed_size = np.count_nonzero(transformed_image) * transformed_image.itemsize original_size = original_image.size * original_image.itemsize print("Compression Ratio:", compression_ratio(original_size, compressed_size)) # Calculate RMSE print("RMSE:", rmse(original_image, reconstructed_image)) Compression Ratio: 0.125 RMSE: 3.759847506248309e-14 B. Huffman Encoding In [6]: **import** heapq from collections import defaultdict def calculate_frequency(data): freq = defaultdict(int) for pixel in data.flatten(): freq[pixel] += 1 return freq def huffman_encode(image): # Flatten the image and calculate frequency freq = calculate_frequency(image) # Build Huffman Tree heap = [[weight, [symbol, '']] for symbol, weight in freq.items()] heapq.heapify(heap) while len(heap) > 1: lo = heapq.heappop(heap) hi = heapq.heappop(heap) for pair in lo[1:]: pair[1] = '0' + pair[1] for pair in hi[1:]: pair[1] = '1' + pair[1] heapq.heappush(heap, [lo[0] + hi[0]] + lo[1:] + hi[1:]) $huff_dict = dict(sorted(heap[0][1:], key=lambda p: (len(p[-1]), p)))$ # Encode image encoded_image = ''.join(huff_dict[pixel] for pixel in image.flatten()) return encoded_image, huff_dict def huffman_decode(encoded_image, huff_dict, shape): # Decode image using Huffman dictionary reverse_dict = {v: k for k, v in huff_dict.items()} current_code = '' decoded_pixels = [] for bit in encoded_image: current_code += bit if current_code in reverse_dict: decoded_pixels.append(reverse_dict[current_code]) current_code = '' decoded_image = np.array(decoded_pixels).reshape(shape) return decoded_image # Huffman Encoding Example Usage original_image = cv2.imread('nature.png', cv2.IMREAD_GRAYSCALE) encoded_image, huffman_dict = huffman_encode(original_image) reconstructed_image = huffman_decode(encoded_image, huffman_dict, original_image.shape) # Calculate Compression Ratio for Huffman Encoding compressed_size = len(encoded_image) // 8 # Converting bit-length to byte size original_size = original_image.size * original_image.itemsize print("Huffman Compression Ratio:", compression_ratio(original_size, compressed_size)) # Calculate RMSE print("Huffman RMSE:", rmse(original_image, reconstructed_image)) Huffman Compression Ratio: 1.0743389499596792 Huffman RMSE: 0.0 C. LZW Encoding In [7]: def lzw_encode(image): # Convert the image to a list of bytes data = image.flatten().tolist() # Initialize dictionary with single symbols dictionary = {bytes([i]): i for i in range(256)} encoded data = [] w = b""for c in data: wc = w + bytes([c])if wc in dictionary: W = WC else: encoded_data.append(dictionary[w]) dictionary[wc] = len(dictionary) w = bytes([c])encoded_data.append(dictionary[w]) return encoded_data def lzw decode(encoded data): # Reverse the encoding dictionary dictionary = {i: bytes([i]) for i in range(256)} w = bytes([encoded_data.pop(0)]) $decoded_data = [w]$ for k in encoded_data: entry = dictionary.get(k, w + w[:1]) decoded_data.append(entry)

dictionary[len(dictionary)] = w + entry[:1] return np.array([pixel for byte in decoded_data for pixel in byte]) # LZW Encoding Example Usage original_image = cv2.imread('nature.png', cv2.IMREAD_GRAYSCALE) encoded_data = lzw_encode(original_image) reconstructed_image = lzw_decode(encoded_data).reshape(original_image.shape) # Calculate Compression Ratio for LZW Encoding compressed_size = len(encoded_data) * 4 # Assuming 32-bit integers in encoded data original_size = original_image.size * original_image.itemsize print("LZW Compression Ratio:", compression_ratio(original_size, compressed_size)) # Calculate RMSE print("LZW RMSE:", rmse(original_image, reconstructed_image)) LZW Compression Ratio: 0.4798942099874872 LZW RMSE: 0.0 D. Run-Length Encoding In [8]: def rle_encode(image): pixels = image.flatten() encoded_data = [] last_pixel = pixels[0] count = 1 for pixel in pixels[1:]: if pixel == last_pixel: count += 1 else: encoded_data.append((last_pixel, count)) last_pixel = pixel count = 1 encoded_data.append((last_pixel, count)) return encoded_data def rle_decode(encoded_data, shape): decoded_image = [] for pixel, count in encoded_data: decoded_image.extend([pixel] * count) return np.array(decoded_image).reshape(shape)

RLE Encoding Example Usage original_image = cv2.imread('nature.png', cv2.IMREAD_GRAYSCALE) encoded_data = rle_encode(original_image) reconstructed_image = rle_decode(encoded_data, original_image.shape) # Calculate Compression Ratio for RLE Encoding compressed_size = len(encoded_data) * 2 # Each (pixel, count) pair typically uses 2 bytes original_size = original_image.size * original_image.itemsize print("RLE Compression Ratio:", compression_ratio(original_size, compressed_size))

Calculate RMSE
print("RLE RMSE:", rmse(original_image, reconstructed_image))

RLE Compression Ratio: 0.5954201166729394
RLE RMSE: 0.0

E. Arithmetic Coding

In [9]: import numpy as np
 from PIL import Image
 import math

In [10]: def calculate_frequencies(image):
 frequencies = np.zeros(256) # For 8-bit grayscale image
 for pixel in image.ravel():

frequencies[pixel] += 1

image.ravel()[i] = decoded_pixel

return original_size / encoded_size

def arithmetic_coding_workflow(image_path):
 # Load and convert image to grayscale

Step 4: Calculate compression ratio

Step 1: Calculate frequencies

Step 2: Encode the image

Step 3: Decode the image

Step 5: Calculate RMSE

Arithmetic Encoding Example Usage

print("Arithmetic RMSE:", rmse_value)

Arithmetic Compression Ratio: 405000.0 Arithmetic RMSE: 10.32809329504295

image_path = 'nature.png'

def rmse(original_image, reconstructed_image):

def compression_ratio(original_image, encoded_value):

decoded_pixel = np.searchsorted(cumulative_freq, (encoded_value - low) / range_width)

low = low + range_width * cumulative_freq[decoded_pixel - 1] if decoded_pixel > 0 else low

decoded_pixel = min(decoded_pixel, 255) # Ensure index is within [0, 255]

encoded_size = math.ceil(-math.log2(encoded_value)) # Bits to represent the code

reconstructed_image = arithmetic_decode(encoded_value, original_image.shape, frequencies)

C:\Users\Admin\AppData\Local\Temp\ipykernel_17216\334534403.py:26: RuntimeWarning: invalid value encountered in double_scalars

high = low + range_width * cumulative_freq[decoded_pixel]

original_size = original_image.size * 8 # Each pixel as 8 bits

original_image = np.array(Image.open(image_path).convert('L'))

encoded_value = arithmetic_encode(original_image, frequencies)

comp_ratio = compression_ratio(original_image, encoded_value)

comp_ratio, rmse_value, reconstructed_image = arithmetic_coding_workflow(image_path)

decoded_pixel = np.searchsorted(cumulative_freq, (encoded_value - low) / range_width)

rmse_value = rmse(original_image, reconstructed_image)

return comp_ratio, rmse_value, reconstructed_image

print("Arithmetic Compression Ratio:", comp_ratio)

frequencies = calculate_frequencies(original_image)

return np.sqrt(np.mean((original_image - reconstructed_image) ** 2))

return frequencies / np.sum(frequencies) def arithmetic_encode(image, frequencies): low, high = 0.0, 1.0cumulative_freq = np.cumsum(frequencies) for pixel in image.ravel(): range_width = high - low high = low + range_width * cumulative_freq[pixel] low = low + range_width * cumulative_freq[pixel - 1] if pixel > 0 else low return (low + high) / 2 def arithmetic_decode(encoded_value, shape, frequencies): image = np.zeros(shape, dtype=np.uint8) low, high = 0.0, 1.0cumulative_freq = np.cumsum(frequencies) for i in range(shape[0] * shape[1]): range_width = high - low # Find the decoded pixel value within bounds

return image