IMAGE COMPRESSION PROJECT

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CS663: DIGITAL IMAGE PROCESSING
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Contents

Problem Statement

The problem statement of this project has been taken from the following website:

We have built an image compression engine along the lines of the JPEG algorithm. Along with this, we have implemented PCA (both for coloured and grayscale images) algorithm. We have thoroughly studied a tier-1 conference paper **Edge-Based Image Compression with Homogeneous Diffusion** and implemented the algorithm proposed in the paper.

All the algorithms were tested on a variety of image datasets. The results were compared and analyzed to understand the performance of the algorithms.

Basic Implementation

Here are the steps which were performed as part of the basic implementation:

- Computation of the 2D DCT coefficients of non-overlapping image patches
- Implementation of the quantization step
- Implementation of the Huffman tree
- Writing data to an appropriate file format (.bin) and plotting RRMSE vs BPP

Here is the expression of RMSE:

$$\text{RMSE} = \frac{\sqrt{\frac{1}{N}\sum_{i=1}^{N}\sum_{j=1}^{M}(I_{\text{orig}}(i)(j) - I_{\text{recon}}(i)(j))^2}}{\sqrt{\frac{1}{N}\sum_{i=1}^{N}\sum_{j=1}^{M}(I_{\text{orig}}(i)(j))^2}} \tag{1}$$

where I_{orig} is the original image and I_{recon} is the reconstructed image. BPP stands for the size of the image in bits divided by the number of pixels.

Quality Comparison

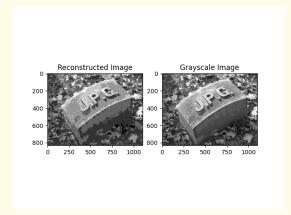Here is the comparison of the reconstucted and the original image for a quality factor of 2:
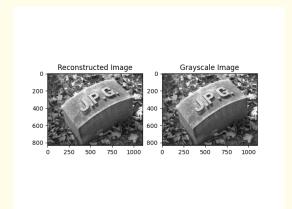

Figure: Original and Reconstructed Image

Quality Comparison

Here is the comparison of the reconstucted and the original image for a quality factor of 10:

Figure: Original and Reconstructed Image

Quality Comparison

Here is the comparison of the reconstucted and the original image for a quality factor of 50:
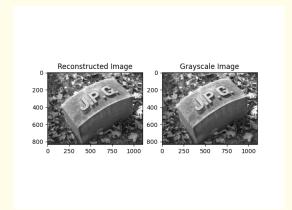
Figure: Original and Reconstructed Image

Quality Comparison

Here is the comparison of the reconstucted and the original image for a quality factor of 80:
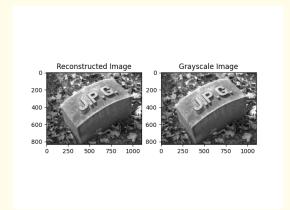
Figure: Original and Reconstructed Image

RMSE vs BPP

For the basic implementation, we have used the dataset from the miscellaneous category of the msrcorid dataset. We picked random 20 images and used 20 quality factors (in the range of 1 to 100) to plot the RMSE vs BPP graph. Here is the graph:
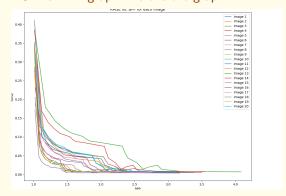

Figure: RMSE vs BPP

Image Compression

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Individual Contributions

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Comparison of Basic vs JPEG

The following plot shows how the basic algorithm performs as compared to the JPEG algorithm.
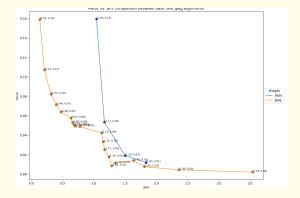

Figure: Comparison of Basic and JPEG

Run Length Encoding

The quantized DCT coefficients are arranged in a zigzag order. This pattern leaves a bunch of consecutive zeros at the end.

In runlength encoding, we replace the consecutive zeros with a pair of numbers: the number of zeros and the value of the next non-zero element. This reduces the size of the data to be stored.

RMSE vs BPP

The dataset of images from the miscellaneous category of the msrcorid dataset was used to plot the RMSE vs BPP graph. Here is the graph:
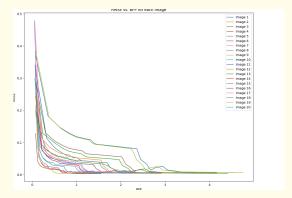

Figure: RMSE vs BPP

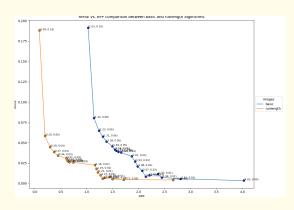Image Compression

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Comparison of Basic and Run Length Encoding

Figure: Comparison of Basic and Run Length Encoding

Comparison of Run Length Encoding and JPEG Algorithm
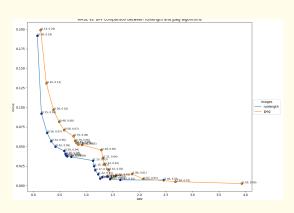
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Figure: Comparison of Basic and Run Length Encoding

Quality Comparison

Here is the comparison of the reconstucted and the original image for a quality factor of 10:

Figure: Original and Reconstructed Image

Quality Comparison

Here is the comparison of the reconstucted and the original image for a quality factor of 80:
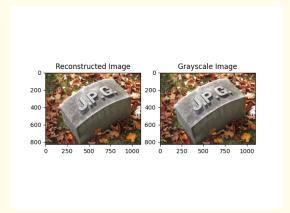
Figure: Original and Reconstructed Image

PCA on a Single Grayscale Image

- Pad the image suitably to make dimensions a multiple of 8.
- Divide the image into patches of size 8×8 and flatten them into column vectors.
- Compute the column correlation and row correlation matrices of the matrix of flattened patches.
- Compute the kronecker product of the two matrices and extract eigenvectors of this matrix with the top k eigenvalues. These eigenvectors are the principal components.
- Now project the original image onto the column space of these eigenvectors. The resulting matrix is the compressed image.
- Reconstruct the image by multiplying the compressed image with the transpose of the eigenvectors.

PCA on a Dataset of Grayscale Images

- Pad the images suitably to make dimensions a multiple of 8.
- Divide each image into patches of size 8×8 and flatten them into column vectors. Create a matrix whose columns are the flattened image patches.
- Compute the column correlation and row correlation matrices of the matrix of flattened patches.
- Compute the kronecker product of the two matrices and extract eigenvectors of this matrix with the top k eigenvalues. These eigenvectors are the principal components.
- Now project each of the original images onto the column space of these eigenvectors. The resulting matrix is the compressed image.
- Reconstruct any image by multiplying the compressed image with the transpose of the eigenvectors.

PCA on a Single Coloured Image

- Pad the image the suitably to make dimensions a multiple of 8.
- Treat each image channel as a grayscale image and run the PCA algorithm on each channel separately.
- Extract the compressed image for each channel separately and merge them to get the compressed coloured image.
- Reconstruct the image by multiplying the compressed image of each channel with the transpose of the eigenvectors of that channel.
- Merge the reconstructed channels to get the reconstructed coloured image.
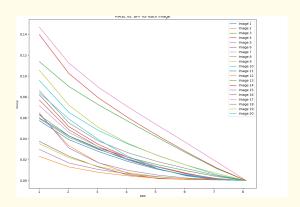
PCA Results

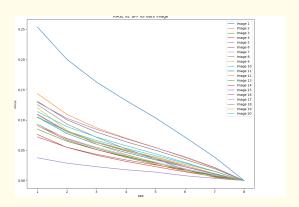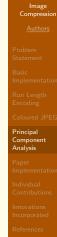Figure: RRMSE vs BPP for PCA on a grayscale image

PCA Results

Figure: RRMSE vs BPP for PCA on a dataset of grayscale images
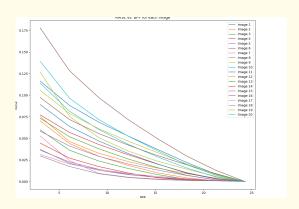
PCA Results

Figure: RRMSE vs BPP for PCA on a coloured image

Edge-Based Image Compression

This paper is based on the fact that edges contain semantically important information in an image. They present a lossy compression method for cartoon-like images.

Encoding Steps

- Firstly, we need to detect edges in the image. The paper proposes Marr-Hildreth edge detector. However, based on our initial experiments, we found that the Canny edge detector works better.

- Now we store the result of edge detection using JBIG(Joint Bi-level Image Experts Group) encoding.

- Then, we encode the contour pixel values. Basically, we store the values from both sides of the edge. The extracted pixel values can be uniformly quantised to 2^q different values, where $q \in \{1, 2, \ldots, 8\}$. Another parameter d, ensures that only every d^{th} value along an edge is stored.

- The quantised and subsampled pixel value signal is then compressed by a PAQ compression method (lossless compression archiver).

- Now, we need to store the encoded data along with all the values of the hyper-parameters.

Decoding Steps

- We split our encoded file into the JBIG data and PAQ data part.

- Both parts are then decoded by using the JBIG and the PAQ method, respectively.

- We reconstruct the quantised colours obtained by the PAQ part. The pixel values between the sampled points are computed by using linear interpolation along each edge.

- The JBIG data provides a bi-level edge image of the original image size. Given the image size and the edge pixel positions, the pixel values are arranged around the edges in the same order in which they have been encoded.

- Now, we have decoded the edge locations and the pixel values surrounding the edges. We use diffusion based inpainting to fill the remaining values.

Results

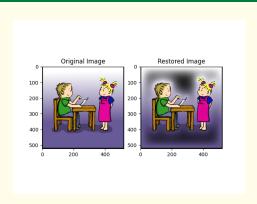Figure: Original vs Reconstructed Image

Results
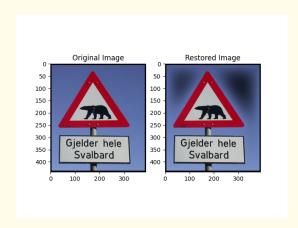
Figure: Original vs Reconstructed Image

To reduce the black patches in regions of the uniform intensity, we propose artificial segmentation of the image, and then handling the small segments normally.

Results

Figure: Original vs Reconstructed Image

Results

Figure: Original vs Reconstructed Image

Results

Figure: Original vs Reconstructed Image

Results

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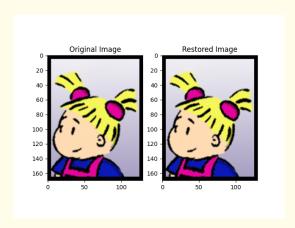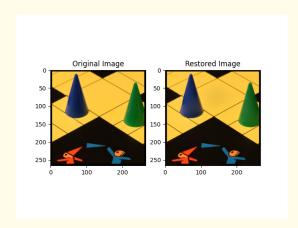Paper Implementation
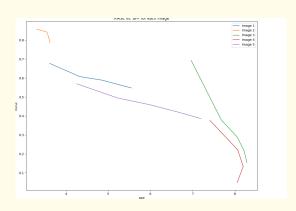
Individual Contributions

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Figure: Original vs Reconstructed Image

Individual Contributions

Saksham Rathi

Basic Implementation
Paper understanding and implementation
Code Modularity
Report Completion

Kavya Gupta

Basic Implementation
Coloured JPEG
Paper understanding and implementation

Shravan Srinivasa Raghavan

PCA
Paper understanding and implementation

Innovations Incorporated

- Runlength encoding and zigzag ordering of DCT coefficients
- PCA for a single grayscale image, a dataset of grayscale images (optimized memory with no redundancies in saving the model), and a single coloured image
- Comparison over different quantization matrices
- JPEG Implementation for a coloured image
- Implementation of the paper "Edge-Based Image Compression with Homogeneous Diffusion"
  - To detect borders of the image, we padded with cells of zero intensity
  - To reduce the black patches in regions of uniform intensity, we are changing the window size (tradeoff between size and accuracy)
  - Varied diffusion time for getting the best output for the target images (tradeoff between time and accuracy)

References

- CS663: Image Compression Slides
- Course Textbook: "Digital Image Processing" by Rafael C. Gonzalez and Richard Woods, 3rd edition
- Using partial differential equations (PDEs) for image compression: M. Mainberger and J.Weickert, "Edge-Based Image Compression with Homogeneous Diffusion", CAIP 2009
- Osman Gokhan Sezer, Onur G. Guleryuz and Yucel Altunbasak, "Approximation and Compression With Sparse Orthonormal Transforms", IEEE Transactions on Image Processing, 2015
- Sample Image Compression Code
- Marr-Hildreth Edge Detector