Lossy Compression of Images Overview of various approaches

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Context

- Lossless compression
 - No losses
 - Important when the decompressed version needs to be identical to the original

Lossy compression

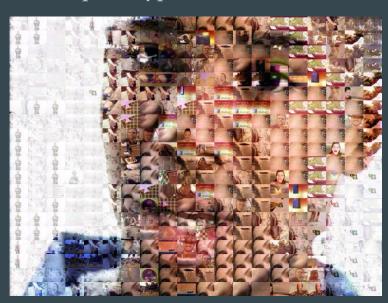
- Loss of some information compared to the original
- Important approach for reducing data size
 - Often better reduction rates than in lossless methods
- No significant degradation is noticed by the end-user
 - Unless if it is constantly compressed/decompressed
 - Generation loss

Context

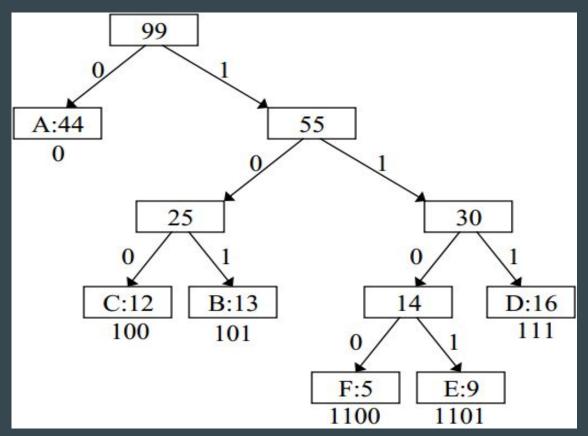
- For this project, we explored means for **lossy** compression of **images**
- Important concepts to take note of
 - Mosaics. Pieces of an image, with certain pixel sizes
 - **Pixel**. Atomic component with an associated color.
 - **RGB**. Popular color system formed by Red, Green and Blue components.
 - Euclidean distance. Distance between 2 points, useful for finding good prototypes
 - **Prototypes.** Mosaics that are the **most representative** of an image.
 - Can be obtained from other images, then applied to the input image

Overview

- 1. Select Image
- 2. Select prototypes
- 3. Re-do image by replacing each mosaic with a prototype
- 4. Build dictionary
- 5. Compress such Dictionary
- 6. Decompress
- 7. See the results



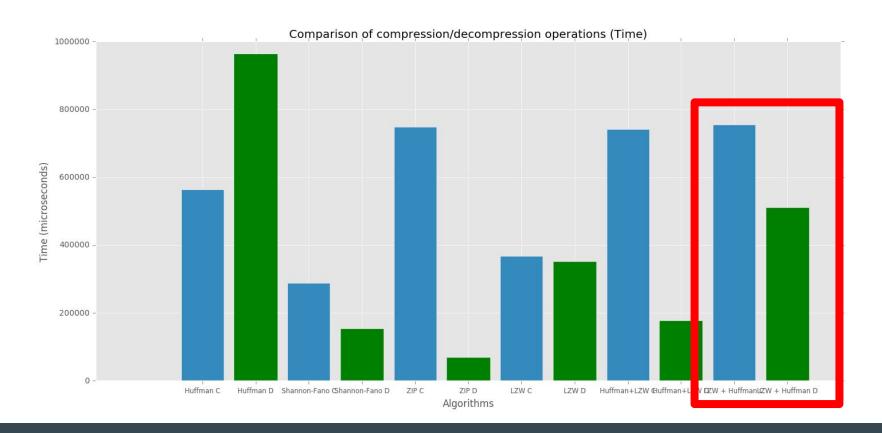
Huffman



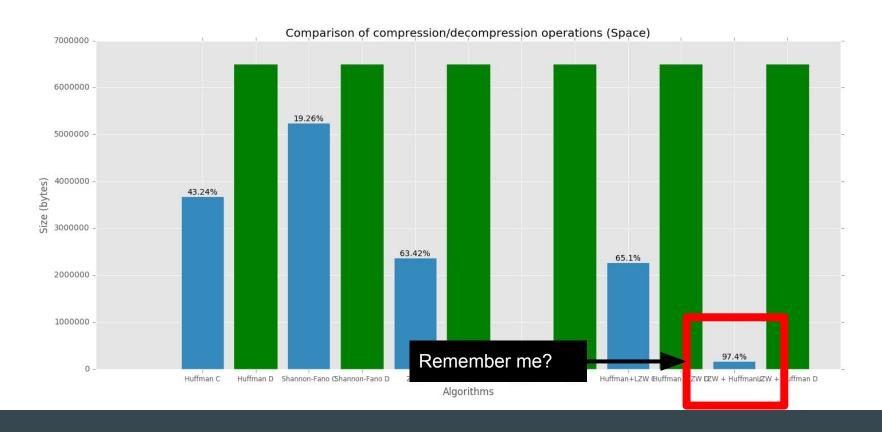
LZW

Input	Current String	Seen this Before?	Encoded Output	New Dictionary Entry/Index
ь	b	yes	nothing	none
ba	ba	no	1	ba / 5
ban	an	no	1,0	an / 6
ban <i>a</i>	na	no	1,0,3	na / 7
bana <i>n</i>	an	yes	no change	none
banan <i>a</i>	an <i>a</i>	no	1,0,3,6	ana / 8
banana_	a_	no	1,0,3,6,0	a_ / 9
banana_b	_b	no	1,0,3,6,0,4	_b / 10
banana_ba	ba	yes	no change	none
banana_ban	ba <i>n</i>	no	1,0,3,6,0,4,5	ban / 11
banana_band	nd	no	1,0,3,6,0,4,5,3	nd / 12
banana_banda	da	no	1,0,3,6,0,4,5,3,2	da / 13
banana_bandan	an	yes	no change	none
banana_bandana	an <i>a</i>	yes	1,0,3,6,0,4,5,3,2,8	none

Remember?



Remember?



LZW Over Huffman

Compression

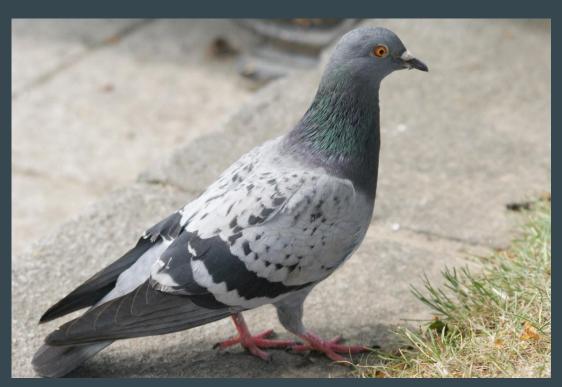
- '[['mosaics/Pigeon-8_xmap_34_2.jpg', 'mosaics/Pigeon-8_xmap_13_8.jpg', ...,'mosaics/Pigeon-8_xmap_11_8.jpg','mosaics/Pigeon-8_xmap_5_22.jpg'], [0, 1, 2, 2, 3, 3, 4, 4, 2, 2, 4, ...,4, 2, 2, 4, 3, 5, 6, 1, 7, 1, 1, 1, 8, 0, 1, 3, 3, 6, 0, 0, 7]]'
- Huffman encode -> 100011001100101010111101110110100001
- LZW encode -> [111,132,142,266,...]

Decompression

- LZW decode -> 100011001100101010111101110110100001
- Huffman decode -> '[['mosaics/Pigeon-8_xmap_34_2.jpg', 'mosaics/Pigeon-8_xmap_13_8.jpg', ...,
 'mosaics/Pigeon-8_xmap_11_8.jpg', 'mosaics/Pigeon-8_xmap_5_22.jpg'], [0, 1, 2, 2, 3, 3, 4, 4, 2, 2, 4, ...,
 4, 2, 2, 4, 3, 5, 6, 1, 7, 1, 1, 1, 8, 0, 1, 3, 3, 6, 0, 0, 7]]'

Seems to be the same, but surely isn't!

Original Image



Properties

Name: Pigeon-8.jpg

Size: 194 Kb

Dimensions: 1154*768 pixels

1st Approach

- 1. Take a collection of images
- 2. Choose the n prototypes that are most similar with each mosaic of the original image (matrixEval)
- 3. Build a new image from the prototype selection









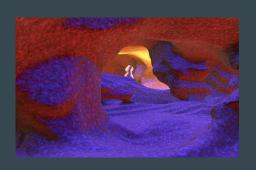












1st Approach



2nd Approach

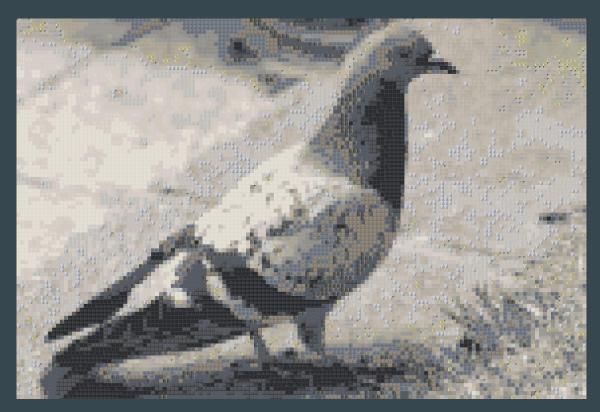
1. One image

- 2. Choose the n prototypes that are most similar with each mosaic of the original image
- 3. Build a new image from the prototype selection

2nd Approach



2nd Approach



Final approach

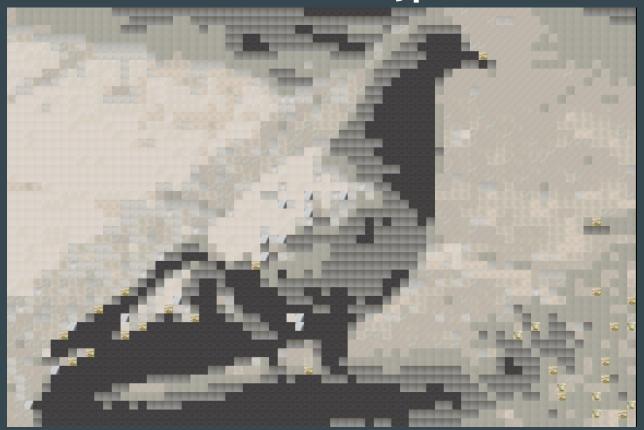
- 1. One Image
- 2. Choose the n prototypes at random
- 3. Train them
- 4. Build a new image from the **trained** prototype selection

































Learning Algorithm

```
y -> most similar prototype
```

x -> mosaic

 α -> alpha [0,1]

i,j -> mosaic location

$$y^{n} = y^{n} + \alpha (x_{ij} - y^{n})$$

Comparison Algorithm

y -> most similar prototype

x -> mosaic

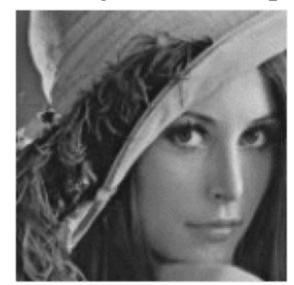
 $\alpha \rightarrow \text{alpha} [0,1]$

p -> number of prototypes

$$d(y^n, x) \le d(y^m, x) \forall m$$

$$d(y, x) = \sqrt{\sum_{i=1}^{p} (y^{i} - x^{i})^{2}}$$

Example of Lossy Compression



Original Lena Image (12KB size)

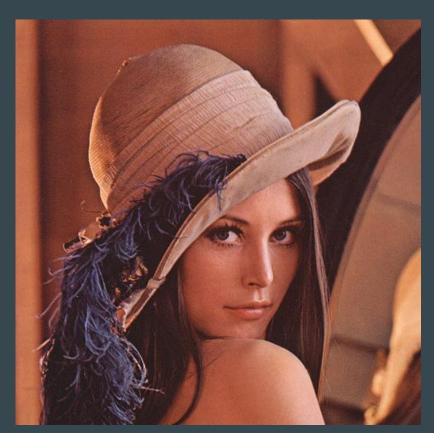


Lena Image, Compressed (85% less information, 1.8KB)



Lena Image, Highly Compressed (96% less information, 0.56KB)

Original Lena image



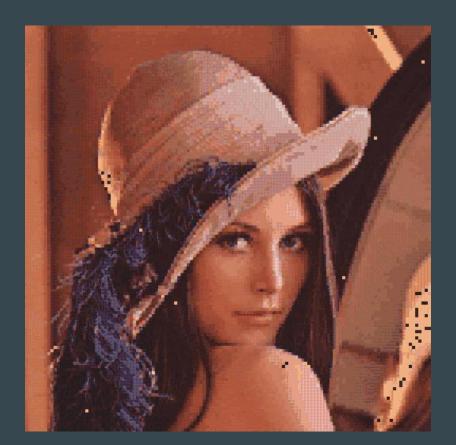
Properties

Name: lena.jpg

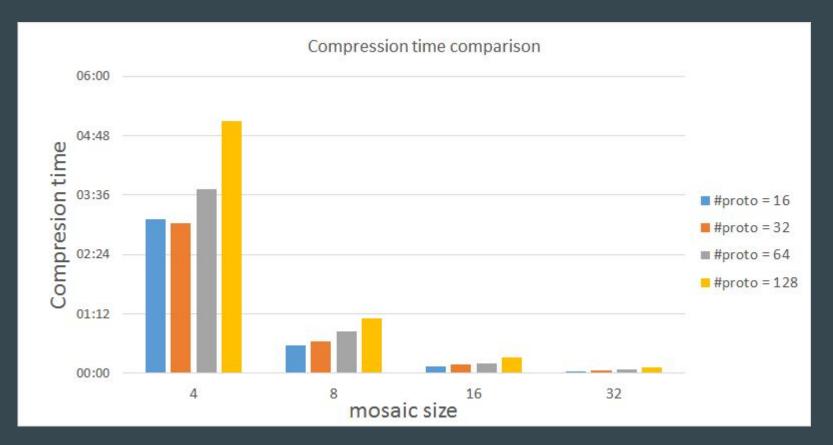
Size: 335 Kb

Dimensions: 512*512 pixels

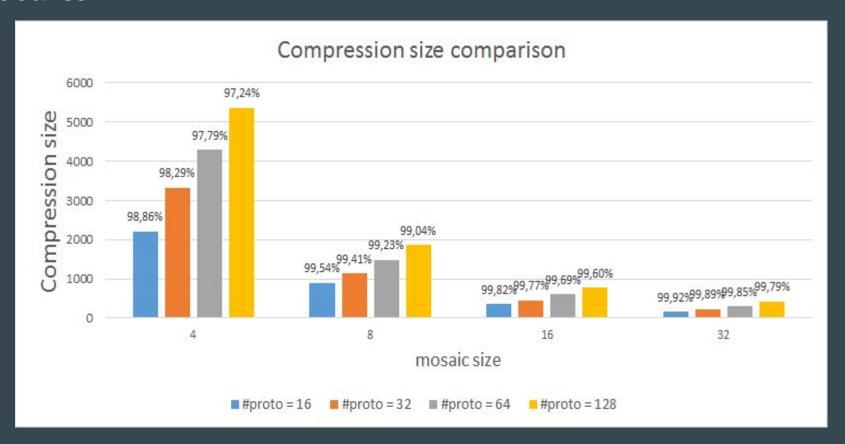
Our version of Lena



Results



Results



References

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Questions?

