

Image Compression via Singular Value Decomposition

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Outline

- 1 Singular Value Decomposition
- 2 Application to Image Compression
- 3 The Code
- 4 Examples

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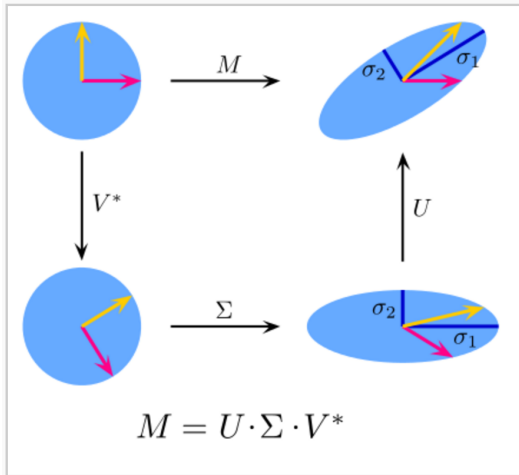
Images are Matrices

- An image can be viewed as a matrix where each entry in the matrix represents a pixel.
 - If we greyscale the image the entries are just numbers!
 - (otherwise they would be something like triples representing RGB values)
- So, we can apply techniques from linear algebra to manipulate images!

Some Linear Algebra

- Given a matrix M we can decompose it as $M = U \Sigma V$ where U , Σ , and V have some nice properties.
- Σ is matrix whose non-zero entries are located on the main diagonal and these values contain information about how the columns of U relate to the matrix M .

Schematic Example



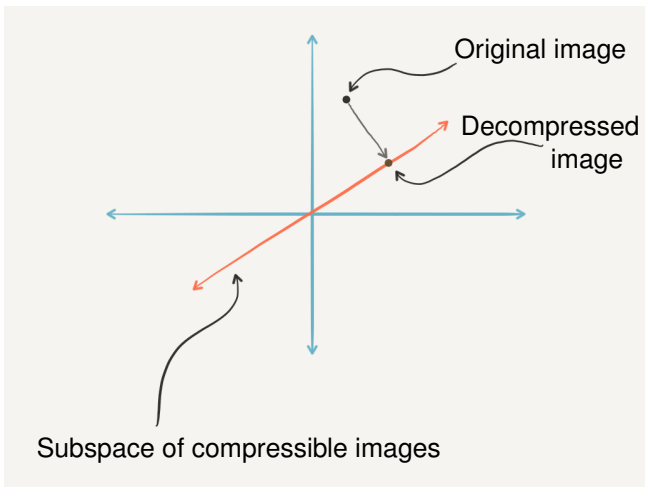
Some Linear Algebra

- The columns of U whose corresponding values in Σ (called singular values) are large can be interpreted as the “most important” in relation to A .
- In a sense we can approximate A by only paying attention to the first N “most important” columns of U .

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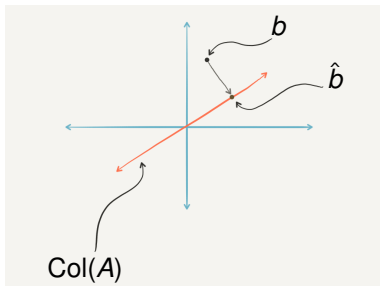
The Big Picture



Where's the Compression?

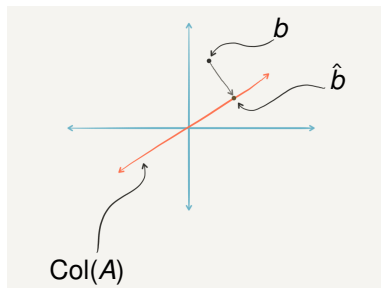
- The original image can be described with two coordinates.
- The compressed can be described with only one coordinate: the distance along the line from the origin!
- So, we have halved the amount of information needed to store the image.

This is Really Least Squares (Some Math)



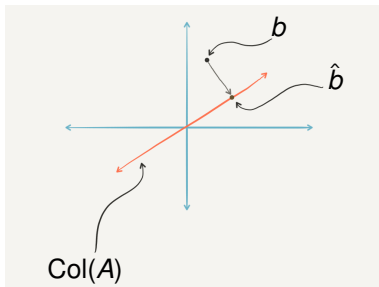
- If A is the matrix of “important” columns of the image matrix M , then the equation $Ax = b$ has a solution if b is in the span of the columns of A .

This is Really Least Squares (Some Math)



- But here, b is not in the column space of A , so we do the best we can: project b into $Col(A)$ and solve the equation $Ax = \hat{b}$ which DOES have a solution.

This is Really Least Squares (Some Math)



- A solution to $Ax = \hat{b}$ is a compressed representation of the image b and we recover the decompressed image \hat{b} by multiplying by A .

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Cards on the Table

- I had trouble working with large images in sklearn due to what seem to be memory issues related to blocking high resolution images.
- The following is a low resolution example generated in python.

Cards on the Table

The following examples were generated using the same algorithm, but implemented in Mathematica.

I Like Dogs



Compressed 90%



Compressed 99%



I Like Dogs



Compressed 90%



Compressed 99%



I Like Dogs



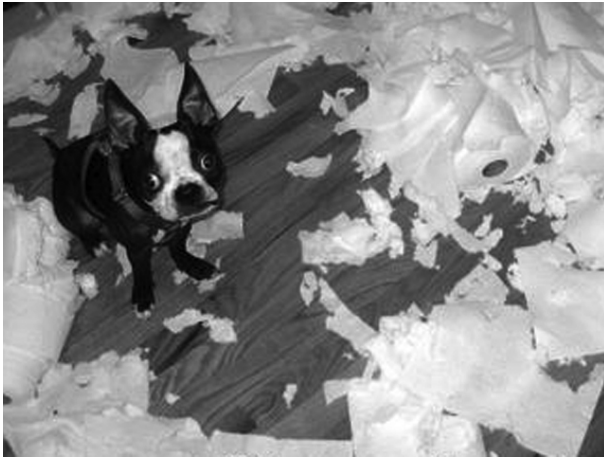
Compressed 90%



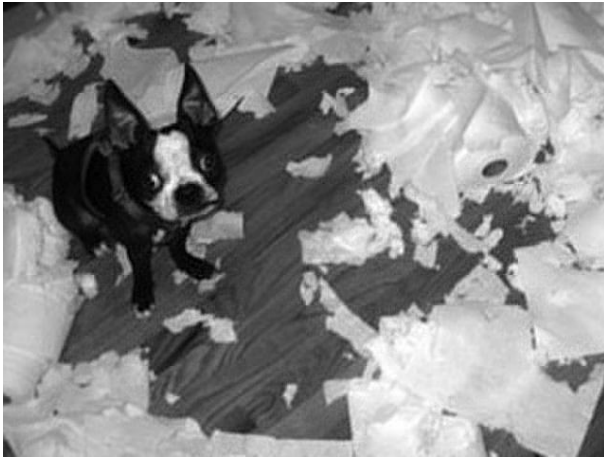
Compressed 99%



I Like Dogs



Compressed 90%



Compressed 99%

