



Colorization by Optimization

Sudharshan Ashwin Renganathan

Aerospace Systems Design Laboratory

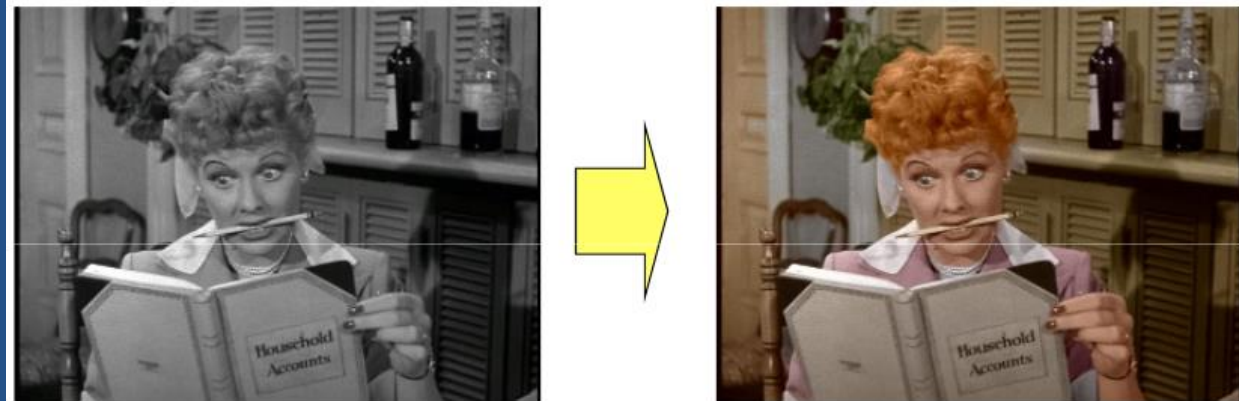
School of Aerospace Engineering

ashwinsr@gatech.edu



Introduction

- Given a grayscale image with some user inputs, we want to generate a *colorized* image



Colorization: a computer-assisted process of adding color to a monochrome image or movie. (Invented by Wilson Markle, 1970)

Motivation

- Typical colorization process involves
 - Delineating *region* boundary for the image
 - Choose a color and apply to region
 - Track regions across frames in case of video



- Limitation – time consuming and labor intensive
- Motivation – How can we automate the process and minimize user input?

Colorization by Optimization

- Works under the simple principle that

“Nearby pixels with similar intensities should have the same color”

- User scribbles image with color
- Code propagates color to regions appropriately
- The idea is completely proprietary to Levin et.al.[1]
- MATLAB source code available in the open internet

[1] <http://www.cs.huji.ac.il/~yweiss/Colorization/>

Approach



Grayscale
Image



Convert to
YIQ format



GS_YIQ

Scribbled
Image



Convert to
YIQ format



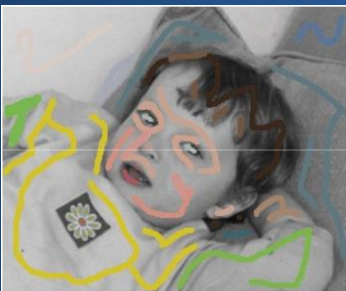
SC_YIQ



Colorize(GS_YIQ, SC_YIQ)



Final Colorized Output



Y – grayscale pixel intensity
(luminance)
I, Q – color palette values
(chrominance)

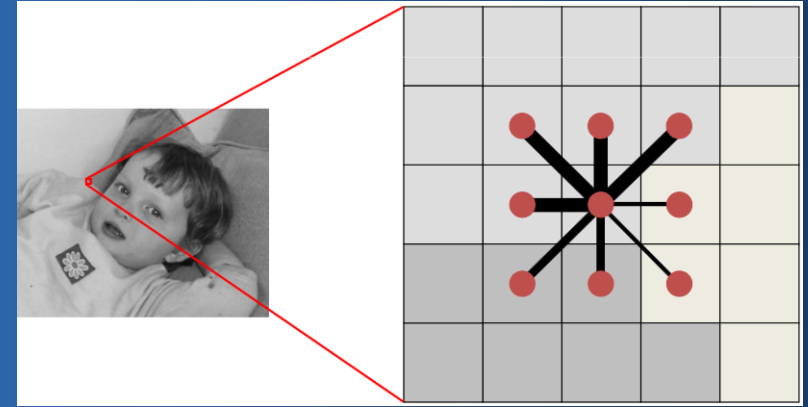


Approach...

- Each image has 3 channels Y, U, V
- Y-intensity; UV-Color
- Minimize difference in U,V for each pixel from neighbors
 - Neighbors weighted based on similarity in intensity (Y)
- Constraints – pixels colored by user

$$E(U) = \sum_{i=1}^{npixels} \left[U(i) - \sum_j w_j U(j) \right]^2$$

$$E(V) = \sum_{i=1}^{npixels} \left[V(i) - \sum_j w_j V(j) \right]^2$$



$$w_j = e^{-(Y(i)-Y(j))^2 / \sigma_i^2}$$

- w_j are called 'affinity functions' which are nothing but **weights**
- σ_i^2 is the variance of pixel intensity

Challenges faced

- Use of MATLAB Optimization
 - Number of variables = $O(\text{npixels})$
 - For a 512 x 512 image, this is 262,144 variables
 - MATLAB reports 'insufficient memory' at this scale
 - Limited by number of variables
- Decipher Levin's code to make modular changes
 - Mathematical treatment not published
 - Code is not annotated
 - Hence not easy to understand completely to make desired changes
- Problem was re-scoped
 - Re-formulate the problem on my own.
 - Understand the mathematical treatment.
 - Write my own code to implement formulation
 - Test code on some test images

Goal

- Re-formulate the optimization problem
 - Work out missing steps in the original paper
 - Solve problem analytically and present solution
- Recreate the colorization code
 - Helps understand process better
 - Helps understand the math better
 - Helps make incremental changes easier
- Test code with sample images.

Problem Formulation

$$E(U) = \sum_{i=1}^{npixels} \left[U(i) - \sum_j w_j U(j) \right]^2$$

$$E(U) = \sum_{i=1}^{npixels} U(i)^2 - 2U(i) \sum_j w_j U(j) + \left[\sum_j w_j U(j) \right]^2$$

$$E(U) = \sum_{i=1}^{npixels} U(i) \left(U(i) - 2 \sum_j w_j U(j) \right) + \left[\sum_j w_j U(j) \right]^2$$

- Each cost function can be expressed in matrix form
- A and B are sparse matrices composed only of w_j 's
- Turns out E can be minimized just using Matrix Algebra

$$E(U) = U^T A U + U^T (B B^T) U$$

$$E(U) = U^T (A + B B^T) U$$

$$E(V) = V^T (A + B B^T) V$$

- Input image is of size $m \times n$
- U and V are vectors of size $1 \times mn$
- A and B are sparse square matrices of size $mn \times mn$

Problem Formulation

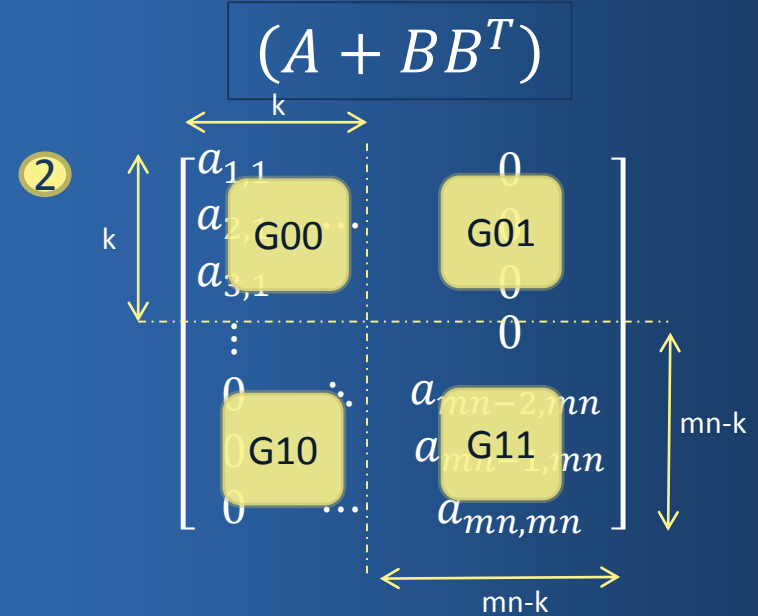
$$E(U) = U^T(A + BB^T)U$$

$$\textcircled{1} \quad E(U) = \begin{bmatrix} U_1 \\ U_2 \\ U_3 \\ \vdots \\ \vdots \\ \vdots \\ U_{mn} \end{bmatrix}^T \begin{bmatrix} a_{1,1} & 0 & & \\ a_{2,1} & \dots & 0 & \\ a_{3,1} & & 0 & \\ \vdots & & 0 & \\ 0 & \ddots & a_{mn-2,mn} & \\ 0 & & a_{mn-1,mn} & \\ 0 & \dots & a_{mn,mn} & \end{bmatrix} \begin{bmatrix} U_1 \\ U_2 \\ U_3 \\ \vdots \\ \vdots \\ \vdots \\ U_{mn} \end{bmatrix}$$

$$\textcircled{3} \quad E = \begin{bmatrix} x \\ q \end{bmatrix}^T \begin{bmatrix} G00 & G01 \\ G10 & G11 \end{bmatrix} \begin{bmatrix} x \\ q \end{bmatrix}$$

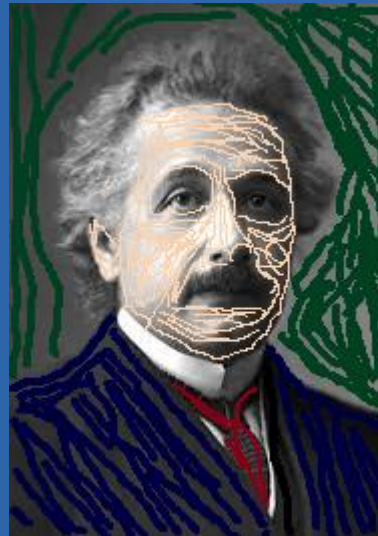
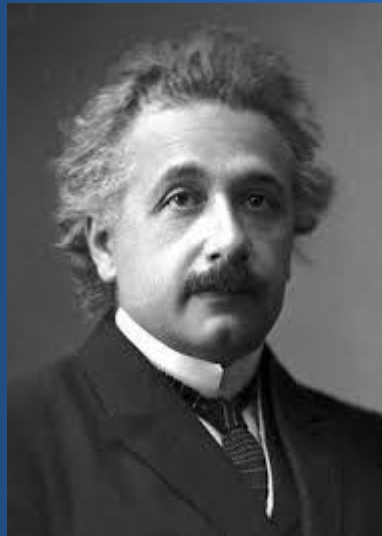
$$\textcircled{4} \quad \frac{\partial E}{\partial x} = x^T(G00 + G00^T) + q(G01 + G10^T) = 0$$

$$\textcircled{5} \quad x = -(G00 + G00^T)^{-1} * q(G01 + G10^T)$$

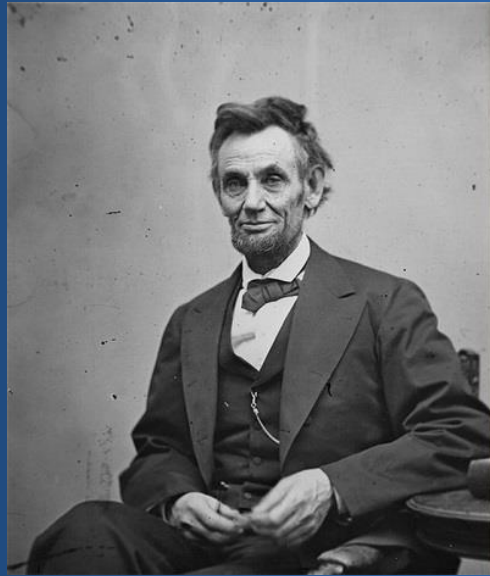


- $G00$ corresponds to un-scribbled pixels
 - Unknowns
- $G11$ corresponds to scribbled pixels
 - Constraints
- $G01$ and $G10$ correspond to both categories
- We split U and V vectors into two vectors
 - x : unknown pixels of length $1 \times k$
 - q : known pixels of length $1 \times mn-k$

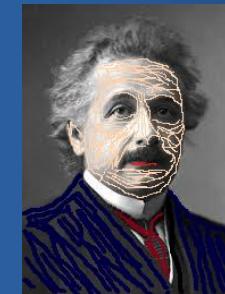
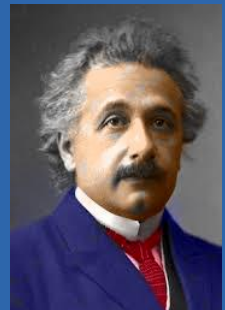
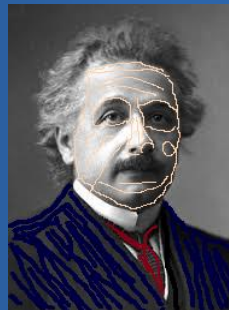
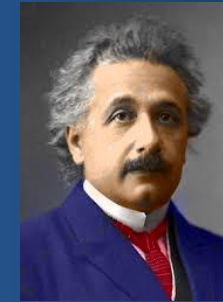
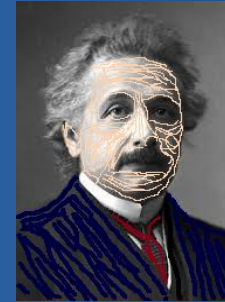
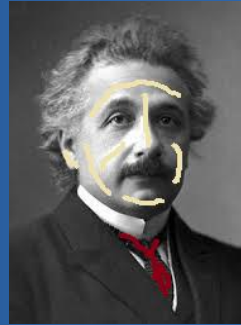
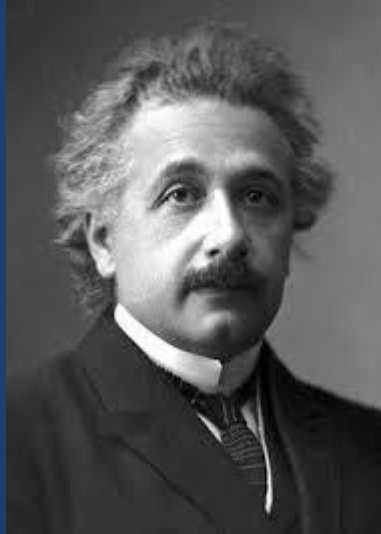
Results



Results



Results (failed cases)



Results (Failed Cases)



Conclusion

- ✓ Colorization by Optimization problem by Levin was understood
 - ✓ Problem was formulated and solution was derived analytically
 - ✓ MATLAB code to implement formulation was developed
- Progressive coloring provides better results
- Coloring close to region boundaries produce better results
- Coloring large and very small regions not very effective
 - More than just a few scribbles required
- Approach not fully automatic
- Some initial goals did not provide expected results but was a learning experience
- Next Steps
 - Video!