

DIP FINAL PROJECT

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- Team Members: 20171052(CSE) and 20171105(CSD)
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- ID : 25
- TITLE :

Face Illumination Transfer through Edge-preserving Filters

GitHub Link: (https://github.com/saisoorya2000/DIP_Project)

MAIN GOAL OF THE PROJECT

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- Our Main objective is to generate photo-realistic relighting of a frontal face image taken under nearly uniform illumination, with a single reference face image of different illumination. (As shown in below figure)

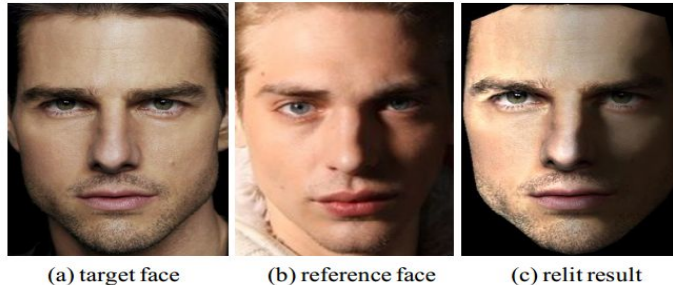


Figure 1. The objective. (a) is the target face image, and (b) is the single reference image. (c) is our relit result which has similar lighting effects to that of (b).

PROBLEM DEFINITION

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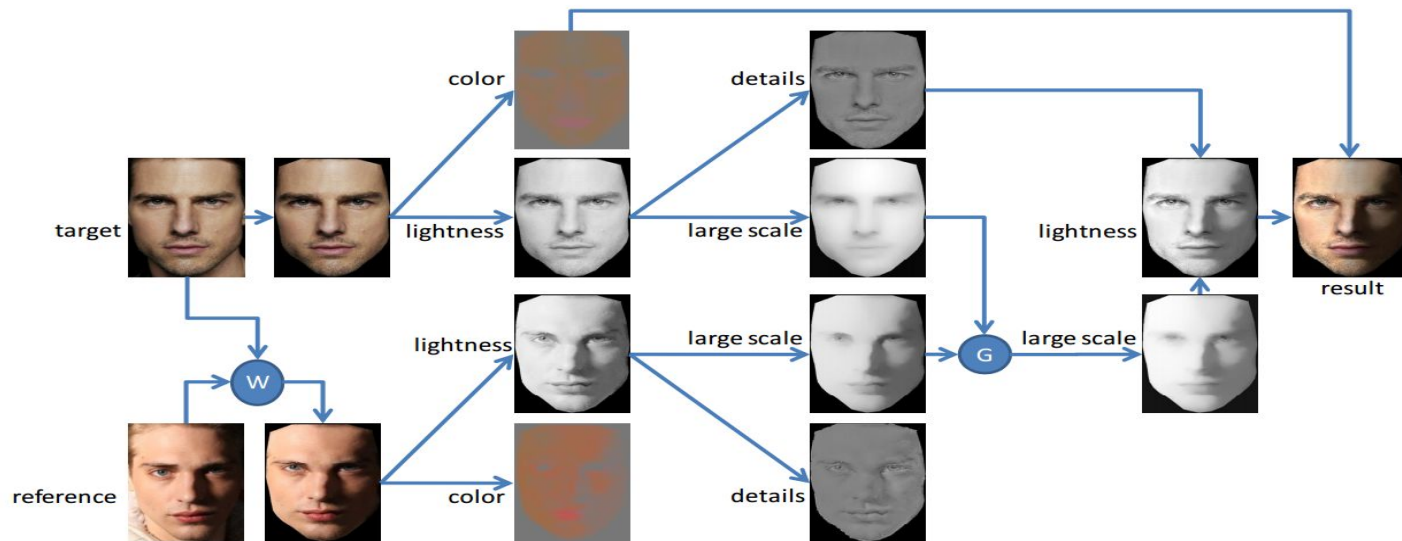
What is the Problem:

Image-based photo-realistic face relighting from a single reference image without any knowledge of the 3D geometry or material information of the target face.

- ❖ We first decompose images into two layers: color and lightness, then decompose the lightness layers of the reference and the target images into **large-scale (illumination variance)** and **detail layers(reflectance variance)**.
- ❖ We then apply **edge-preserving filters** to smooth the large-scale layer of the reference image and preserve the edges of the large-scale layer of the target image.

WORKFLOW

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Face Illumination Transfer

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1)Face Alignment:

- ❖ We take Active Shape Models mark points as control points to warp the reference image according to the target image by using the affine transform.

2)Layer Decomposition:

- ❖ We decouple the image into lightness and color layers.
- ❖ We employ edge-preserving filters such as WLS filter to smooth the lightness layer so as to obtain the large-scale layer and then use division to obtain the detail layer.

3)WLS filter with Adaptive Parameter:

In our task when WLS filter is used it is expected to perform different levels of smoothing on different regions of the image thus we set different smooth levels in different regions .

Face Illumination Transfer : WLS continued..

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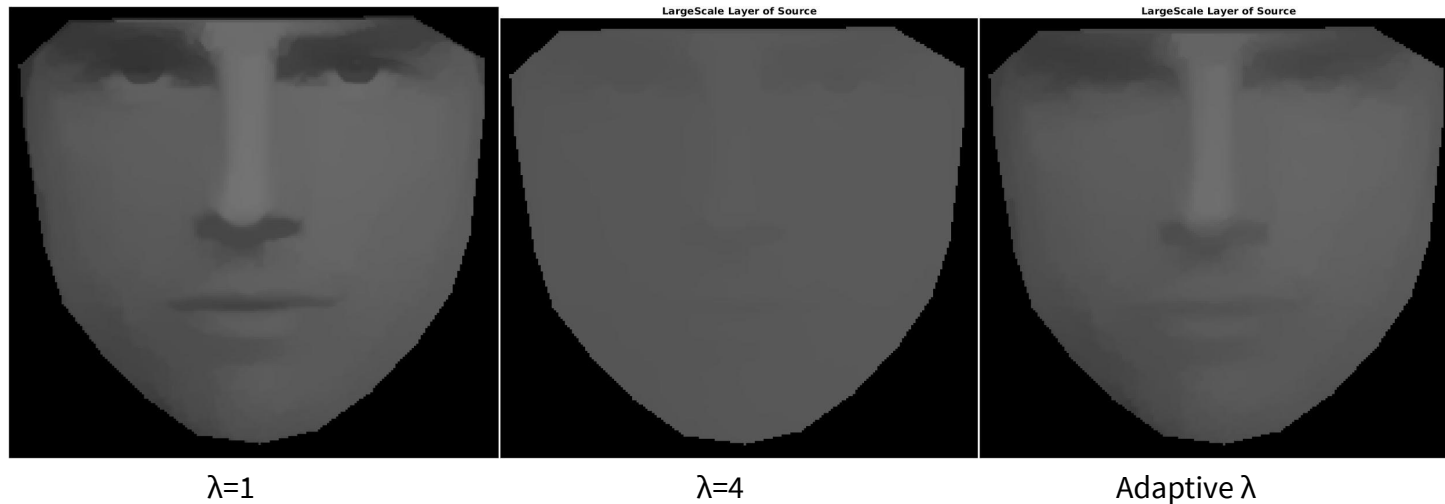
Energy function $\mathbf{E} = |\mathbf{l} - \mathbf{s}|^2 + \mathbf{H}(\nabla \mathbf{s}, \nabla \mathbf{l})$

where, $|\mathbf{l} - \mathbf{s}|^2$ is the data term to keep \mathbf{s} as similar as to \mathbf{l} , and $\mathbf{H}(\nabla \mathbf{s}, \nabla \mathbf{l})$ is the regularization term to make \mathbf{s} as smooth as possible. (' \mathbf{l} ' is lightness layer and ' \mathbf{s} ' is large-scale layer)

- ❖ In $\mathbf{H}(\nabla \mathbf{s}, \nabla \mathbf{l})$, λ is balance factor between the data term and the smoothness term.
- ❖ In the flat region, a small λ is enough to produce a good separation of the large-scale and the detail layers. Most reflectance information can then be retained in the detail layer.
- ❖ However, in the regions such as facial hair and eyebrows, a larger λ is required to perform higher level of smoothing, so that reflectance can be better maintained in the detail layer.

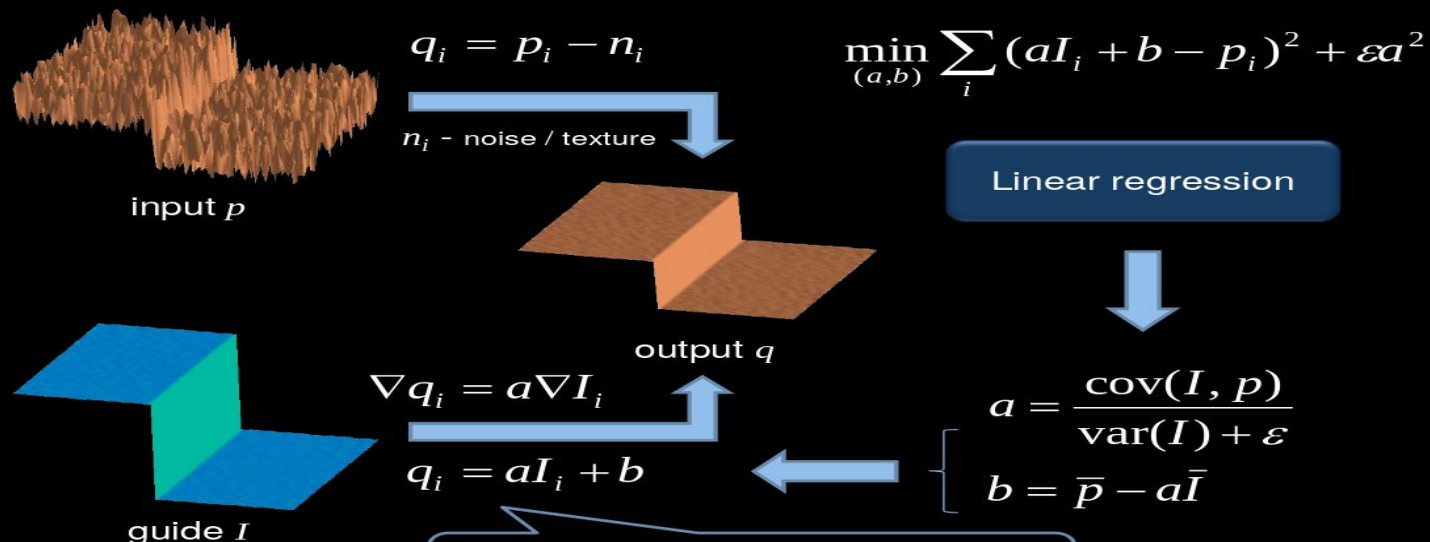
RESULTS OF THE PROJECT (3)

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We can clearly observe that $\lambda=1$ constant over image doesn't smooth well because detail areas require more smoothing whereas $\lambda=4$ over smoothens causing removal of edges. Adaptive filter here uses small λ over flat regions and significantly large λ over facial hair and eyebrows and gives satisfying results.

Guided filter



Face Illumination Transfer : Guided Filter

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4) Guided Filter :

Guided filter has a key assumption that it is a local linear model between guidance I and filtered output q , as q is the linear transform of I in the window w_k centered at the pixel k :

$$q_i = a_k I_i + b_k, \forall i \in w_k$$

,where, a_k and b_k are assumed to be constant in w_k .

The linear coefficients a_k and b_k are determined by minimizing the difference between q and filter input P .

$$\text{Cost-Function } E(a_k, b_k) = \sum ((a_k I_i + b_k - P_i)^2 + \lambda a_k^2).$$

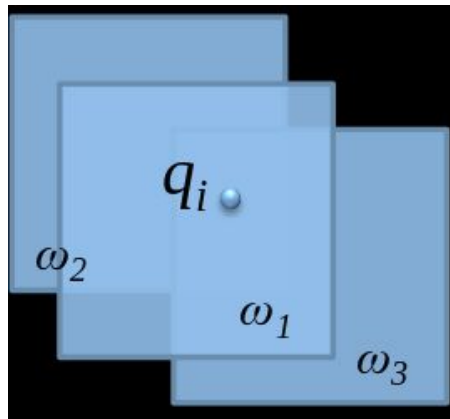
On solving above ridge regression we get a_k and b_k from which we calculate output image.

Face Illumination Transfer : Guided Filter contd..

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Extend to the entire image:

- In all local windows w_k compute the linear coefficients.
- Compute the average over $a_k I_i + b_k$ in all w_k that covers pixel i
- $q_i = \sum_{k| i \in w_k} (a_k I_i + b_k) / |w|$



FINAL STEP

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Filter the reference large-scale layer with the guidance of the target large-scale layer by using guided filter

Desired target lightness is achieved by compositing the filtered reference large-scale layer and the target detail layer.

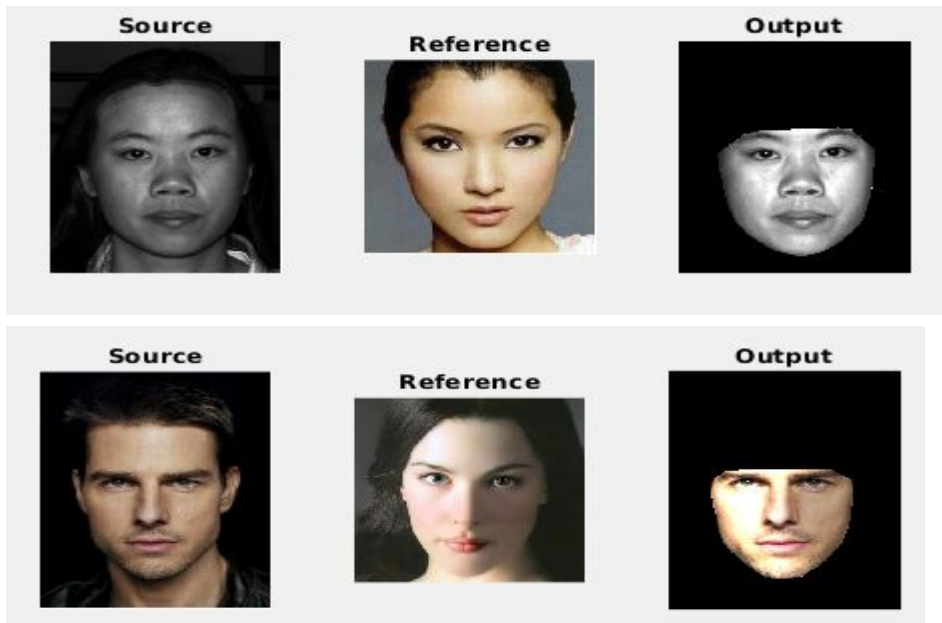
We get the relit result by incorporating the color information of the target face.

Observations

Our method with guided filter stresses that the relit result with the target image in local window can preserve the linear model and maintain better identification characteristics of the target face.

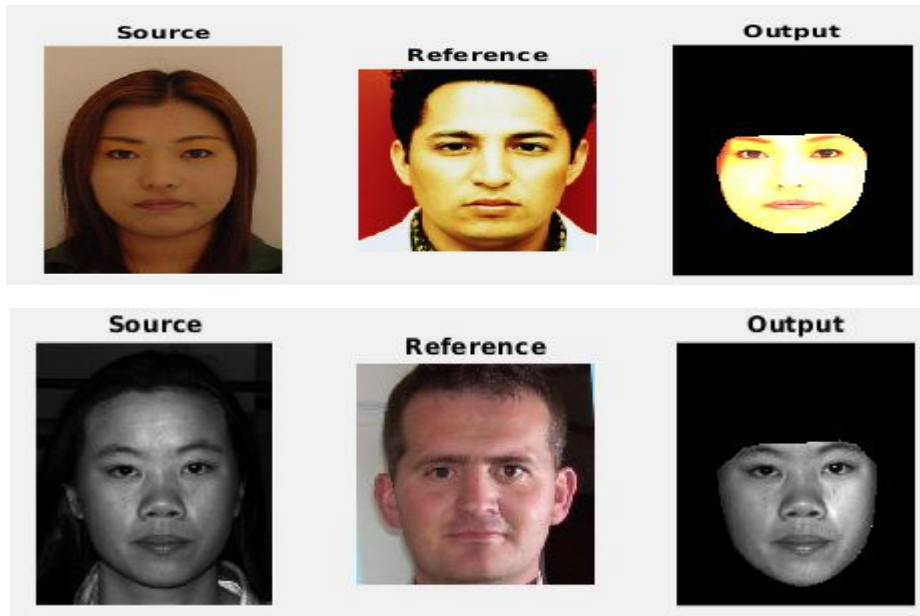
RESULTS

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RESULTS

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DIVISION OF WORK

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Work Item	Major Contributor
Morphing	Soorya
Gradient Generation	Snehith
Adaptive WLS Filter	Snehith
Guided Filter	Soorya + Snehith
Debugging and Misc.	Soorya