

HTTPS://GITHUB.COM/ANUSHKAWAKANKAR/POISSON-MATTING

POISSON MATTING BY GIRL POWER

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Presentation Outline

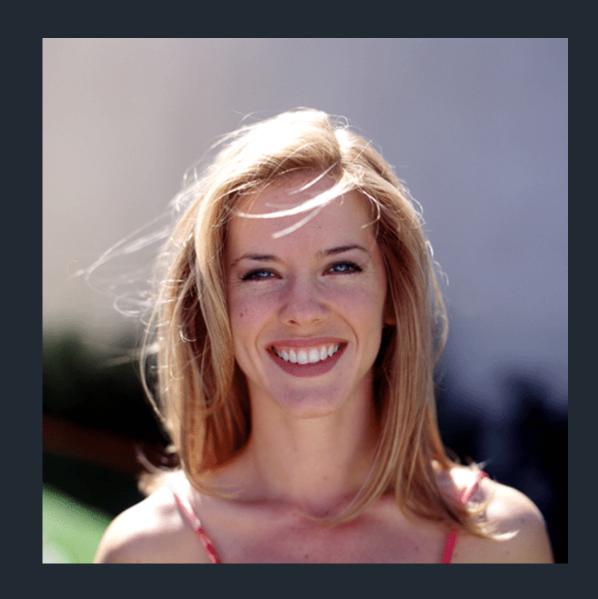
- Introduction to the Problem Statement
- Our approach
- Problems faced
- Final results



How do we change the backgrounds of images?

WE HAVE BEEN GIVEN AN IMAGE AND ITS TRIMAP. NOW WE NEED TO CHANGE THE BACKGROUND OF THE IMAGE. BUT FIRSTLY, WHAT IS A TRIMAP?

THE INPUTS



ORIGINAL



TRIMAP

How to generate an alpha matte?



TRIMAP



ALPHA MATTE



Estimating FG, BG, unknown regions and alpha

- Obtain the grayscale image from the original coloured image
- Inpaint the image to get an approximate fg and bg and then calculate their difference.
- Calculate an estimated alpha matte by giving a weight of 1 to foreground regions, 0 to background regions, and 0.5 to the unknown regions.



Generating the gradient matte

 Using the approx difference, and the second derivative of the intensities of the grayscale image, we calculate alpha gradient by the formula

$$\Delta \alpha = \text{div}(\nabla I)/(F-B)$$

where Δ is the laplacian operator, and div is the divergence operator



Generating the alpha matte

• Next, we generate the alpha matte by solving the laplacian

$$(
abla^2 u)_{ij} = rac{1}{\Delta x^2} (u_{i+1,j} + u_{i-1,j} + u_{i,j+1} + u_{i,j-1} - 4u_{ij}) = g_{ij}$$

To solve this, we use the Gauss-Seidel equation

$$x_i^{(k+1)} = rac{1}{a_{ii}} \left(b_i - \sum_{j=1}^{i-1} a_{ij} x_j^{(k+1)} - \sum_{j=i+1}^n a_{ij} x_j^{(k)}
ight), \quad i = 1, 2, \ldots, n.$$



Now we have an alpha matte



WHY NOT JUST USE CHROMA-KEYING?

We now know which pixels are background and which ones are foreground. Why not just replace the ones in the background with a different image?

THIS IS WHAT HAPPENS...





Instead, we use this equation

$$I = \alpha F + (1 - \alpha)B$$

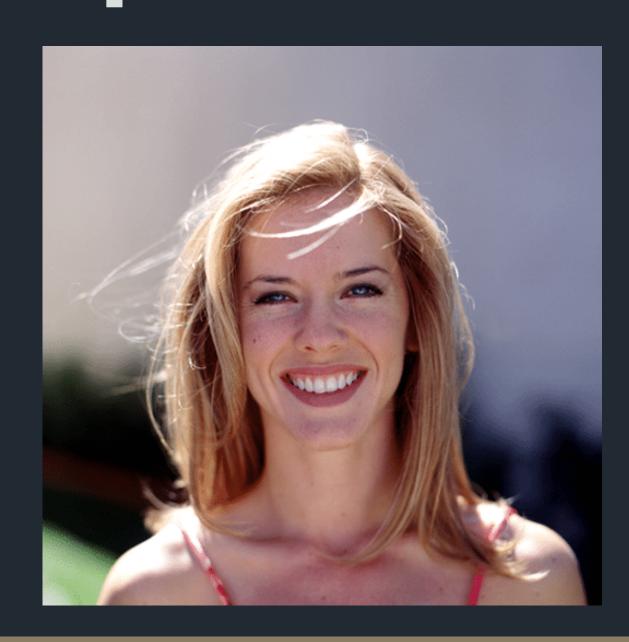
where F is the foreground image, and B is the background image

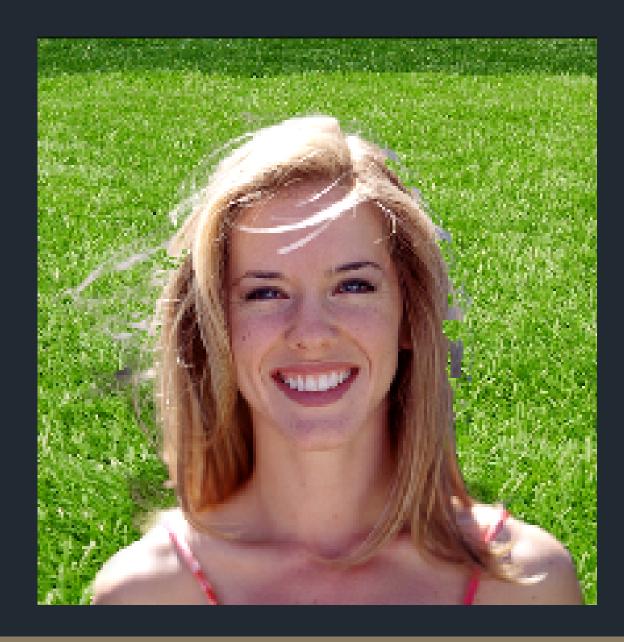
USING THE PREVIOUS EQUATION



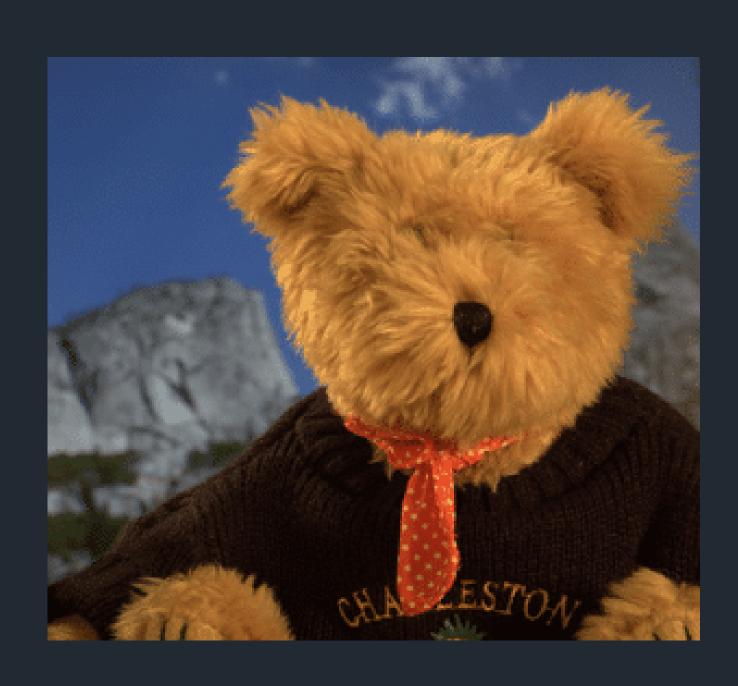


Comparing the input and output



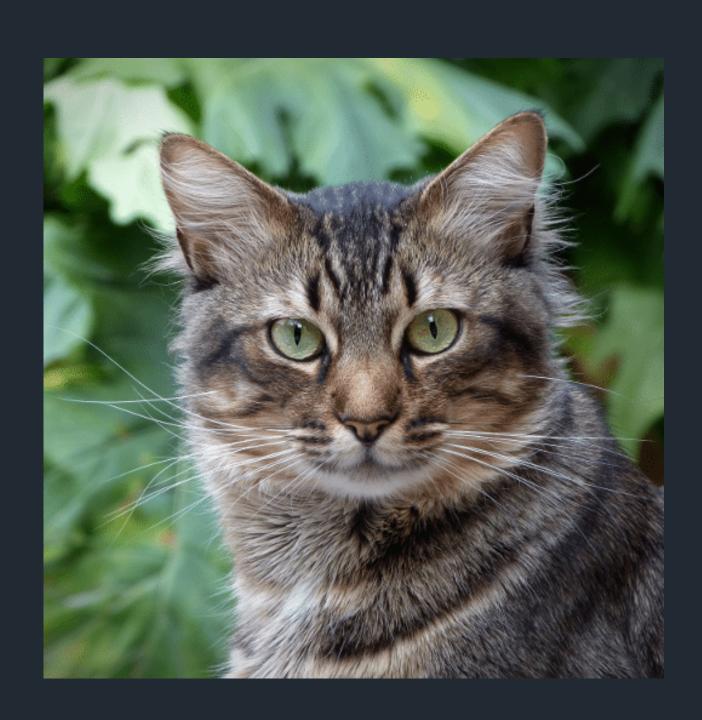


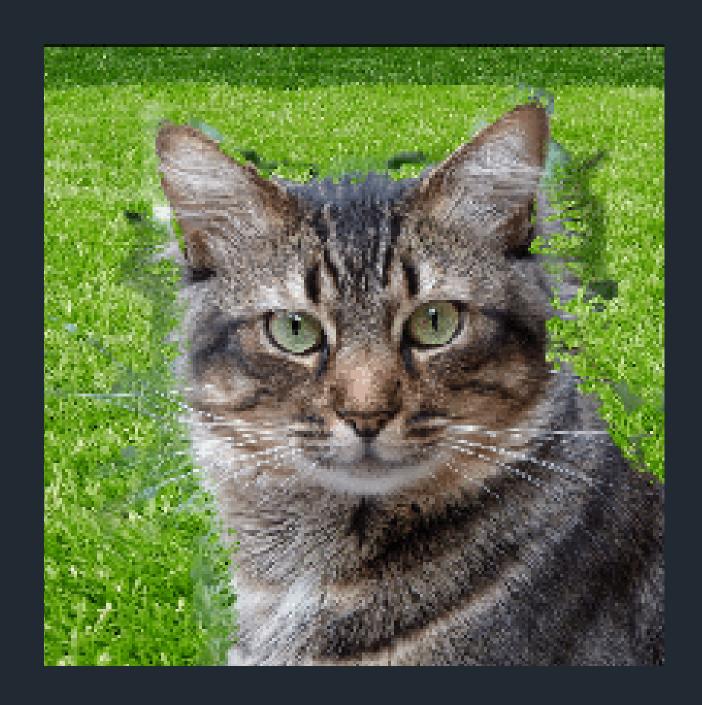




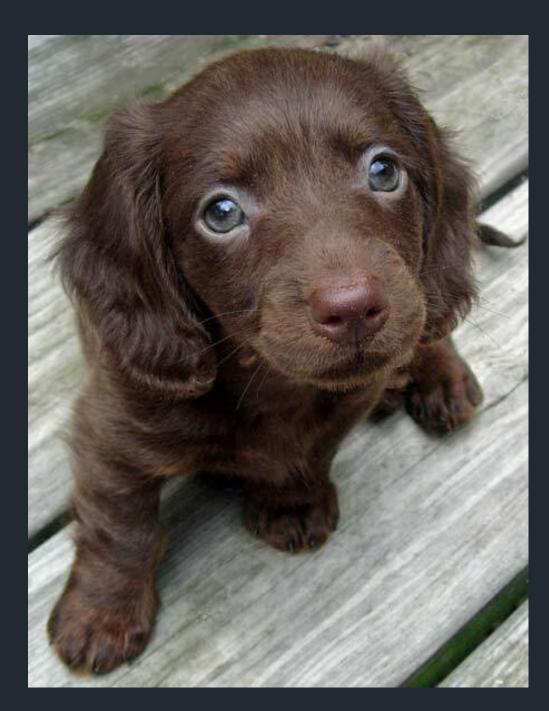


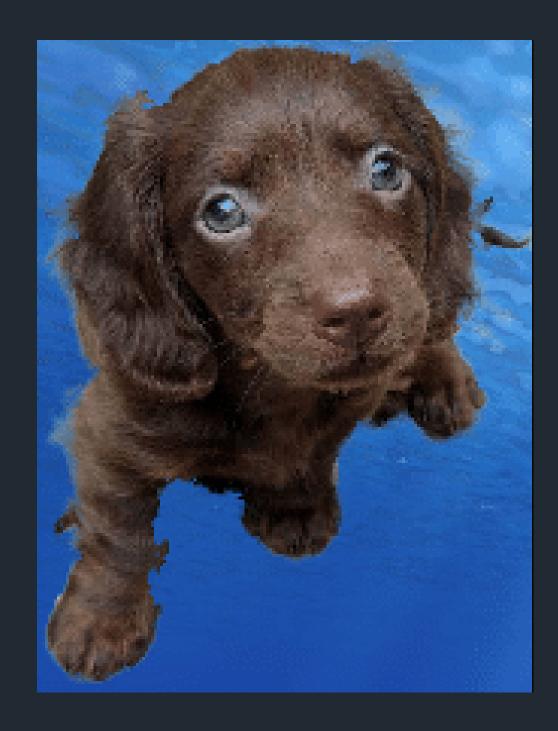




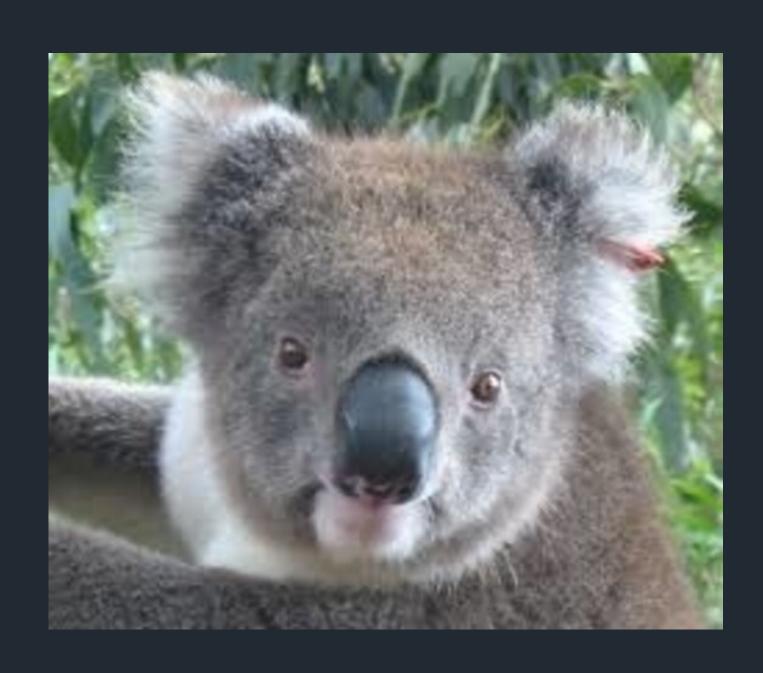


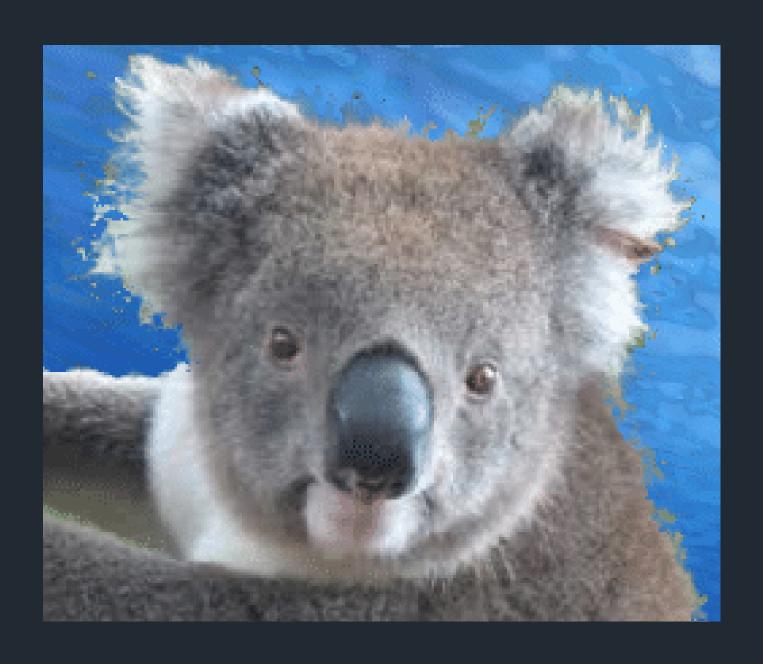












Using Local Poisson Matting

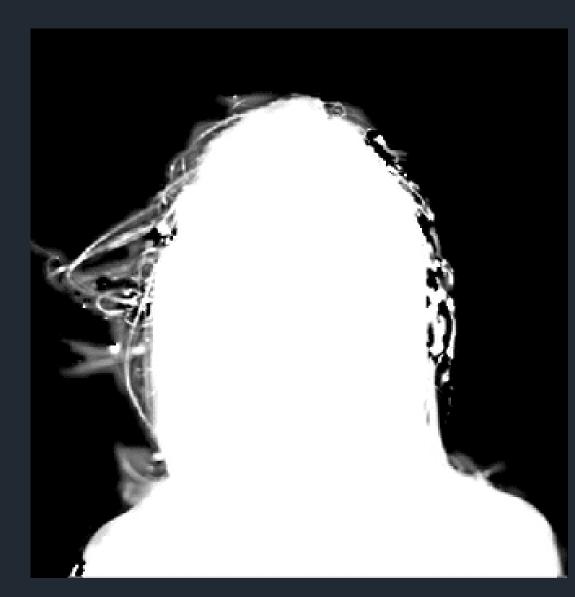
- To refine our results further, we can use Local Poisson matting instead of global matting.
- The user would have to manually select a problem area.
 Applying the above operations on the region selected results in a better localized matte, and hence a better image.
- The altered equation for local poisson matting is

$$\nabla \alpha = A(\nabla I - D)$$

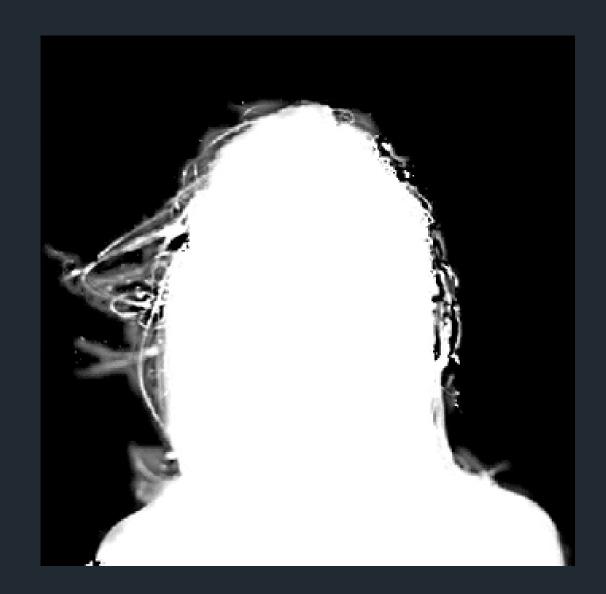
where A = 1/(F - B) and D = $[\alpha \nabla F + (1-\alpha) \nabla B]$.



After Local Poisson Matting



Alpha matte after local matting



Alpha matte after global matting



Boosting Brush

- This is a refining method, which is used when the resulting image is smoother or sharper than expected.
- A boosting brush is applied to increase or decrease A (from the local matting equation) directly
- The modification is as follows -

$$A'_p = [1 + \lambda \exp(-\frac{||p - p_0||^2}{2\sigma^2})] \cdot A_p$$

where p0 is the coordinate of the brush center, σ and λ are user defined parameters to control the size and strength of the effect

High Pass Filtering

- To obtain a clearer alpha matte structure, we can apply High Pass Filtering.
- We use the following formula-

$$D = K * \nabla I$$

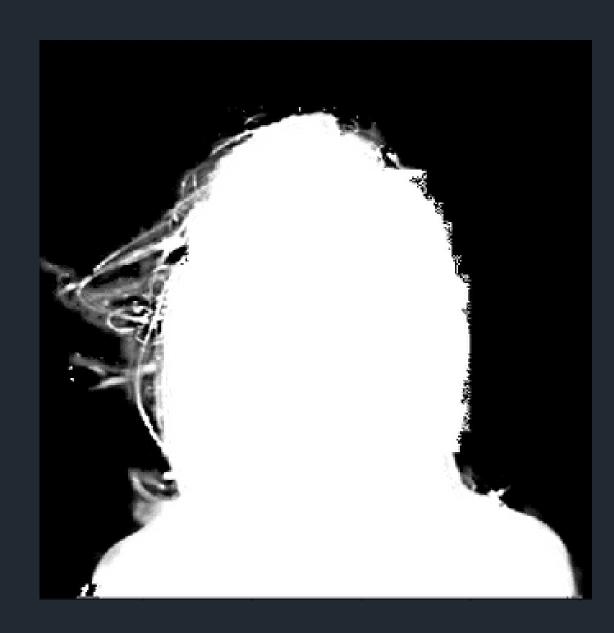
where $K = N(p; p0,\sigma^2)$ is a Gaussian filter centered at p0 and * is the convolution operator.

The new alpha matte is now -

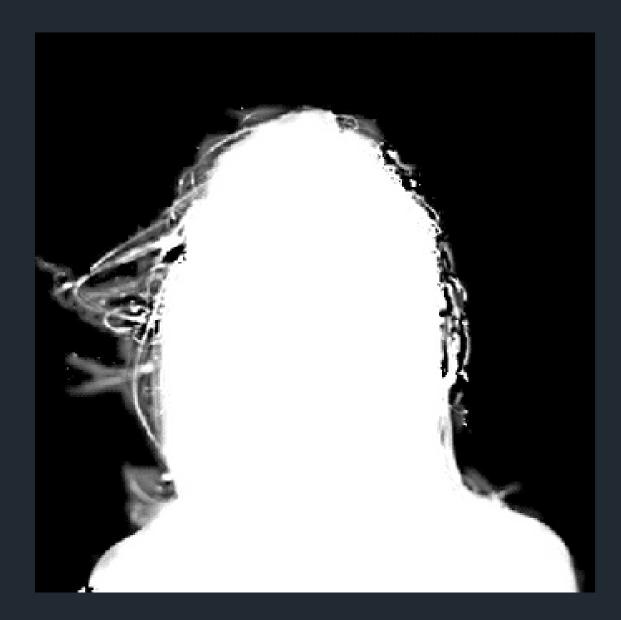
$$\nabla \alpha = A(\nabla I - K * \nabla I)$$



Alpha Matte after refining



After Refining



Before Refining



Other Refining Methods

Some other methods which can be used to refine the images are -

- Diffusion filtering
- Clone brush
- Channel selection

The user can use these methods depending on the results desired and the inputs given.



Distribution of work

ANUSHKA

- Taking inputs and making preliminary masks
- Chroma keying
- Initial local matting operations

SAGRIKA

- Reconstructing
 alpha matte using
 Poisson equations
- Alpha blending
- Local matting by user selection

GOWRI

- Generation of estimate alpha matte
- Optimization of the code
- Trials with refinement



THANK YOU!