

# Fragment-Based Image Completion

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## Outline

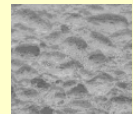
- Goals
- Previous Work
- Fast Approximation
- Confidence Map
- Search
- Compositing Fragments
- Results
- Discussion

## Goals

- Fill holes in images
- Remove unwanted objects from images

## Previous Work

- Texture Synthesis by Non-parametric Sampling – Efros and Leung (ICCV'99)



## Previous Work

- Image Inpainting – Bertalmio and Caselles



## Image Parts

Image

Inverse Matte



## Fast Approximation

- Build an image pyramid (structure which contains images at different scales)
- Down-sample and up-sample image with a kernel at lowest scale until convergence to obtain approximation
- Use this approximation in next highest scale
- At coarser levels, the kernel affects low frequency data, whereas at levels of finer detail, the kernel will approximate higher frequencies

## Fast Approximation

Multiply Image by inverse matte  
and add matte



Downsample and upsample with  
kernel



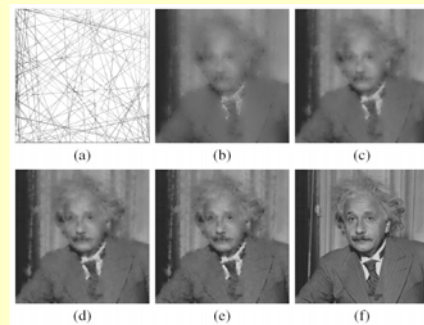
## Fast Approximation



Region after Approximation

Now use this region instead of white space for the approximation step at the next level.

## Fast Approximation



## Fast Approximation



(a)

## Fast Approximation



(b)

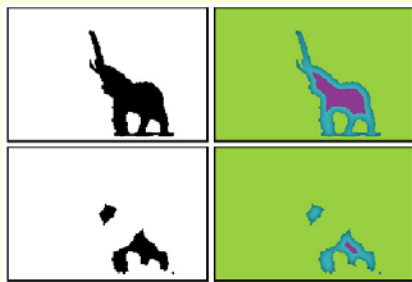
## Confidence Map

- Store a confidence map for the image – each pixel gets a value between 0 and 1, with 1 being the most confident
- This tells us how confident we are in our approximation
- Regions outside the matte have confidence 1

## Confidence Map

- For approximated regions, the confidence map is computed by checking the alpha value of the pixels in the neighborhood
- To select the next sub-region to fill we find the most confident pixel in our approximation and start from there

## Confidence Map



## Search

- For every target fragment we search for a best source match
- This is done across all translations (x,y), 8 orientations  $\theta$ , and 5 scales  $l$ .
- Adds detail to approximated pixels created with kernel, while leaving the known pixels alone

## Search

- How is this done? Consider fragments S and T
- Then we wish to minimize the function:

$$r^* = \arg \min_r \sum_{s=S_r(i), t=T(i), i \in N} (d(s,t)\beta_s\beta_t + (\beta_t - \beta_s)\beta_r).$$

- First term penalizes different values in corresponding pixels with high confidence in both source and target fragments.
- Second term rewards pixels with a higher confidence in source than target, while penalizing pixels with lower confidence in source than target

## Search

Q: But how to figure out how big of a neighborhood to use?

A: Use the contrast criterion of the absolute value of extreme values across channels.

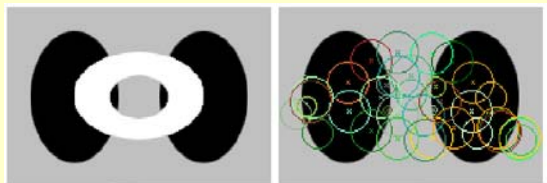
## Search



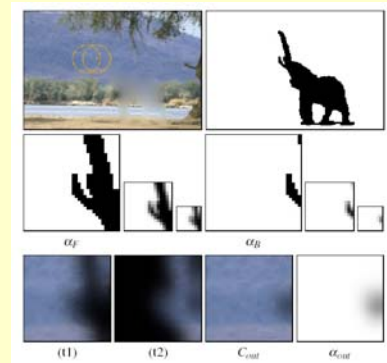
## Search



## Search



## Compositing Fragments



## Results



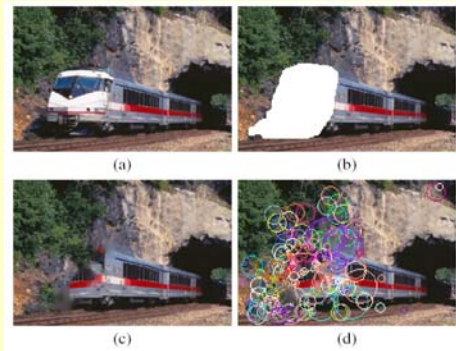
## Results



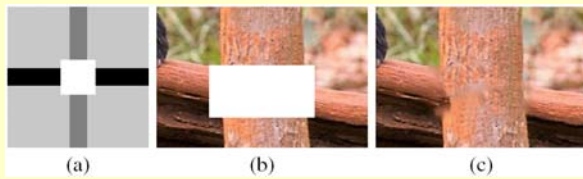
## Results



## Results



## Results



## Results

