

# Content-Aware Image Resizing

Final Project Report

601 Computer Graphics and Image Processing

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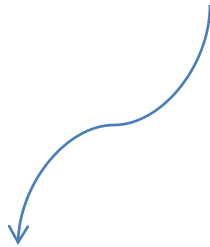
Based on work by S. Avidan and A. Shamir

# Overview

- The problem
- The proposed solution
- Problems with the solution
- Technical details
- Conclusion

# The Problem

- The internet uses dynamic formatting to arrange pictures and text into available space



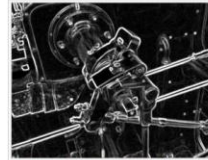
## Sobel operator

From Wikipedia, the free encyclopedia

The **Sobel operator** is used in *image processing*, particularly within *edge detection* algorithms. Technically, it is a *discrete differentiation operator*, computing an approximation of the *gradient* of the image intensity function. At each point in the image, the result of the Sobel operator is either the corresponding gradient vector or the norm of this vector. The Sobel operator is based on convolving the image with a small, separable, and integer valued filter in horizontal and vertical direction and is therefore relatively inexpensive in terms of computations. On the other hand, the gradient approximation which it produces is relatively crude, in particular for high frequency variations in the image.

### Contents [hide]

- 1 Simplified description
- 2 Formulation
- 3 More formally
- 4 Technical details
- 5 Example
- 6 See also
- 7 References
- 8 External links

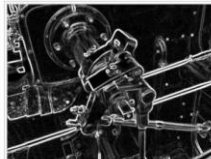


The Sobel operator applied to a colour photograph of a steam engine. (See the original image [here](#).)

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# The Problem

- While text can be rearranged to better fix the screen images are usually only cropped or scaled



scale



crop



# The Problem

Scaling distorts images



Cropping cuts out details

# Some solutions



Original Image and per pixel energy



cropped



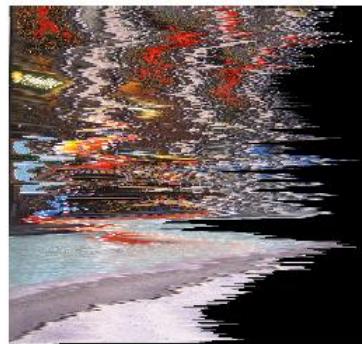
column



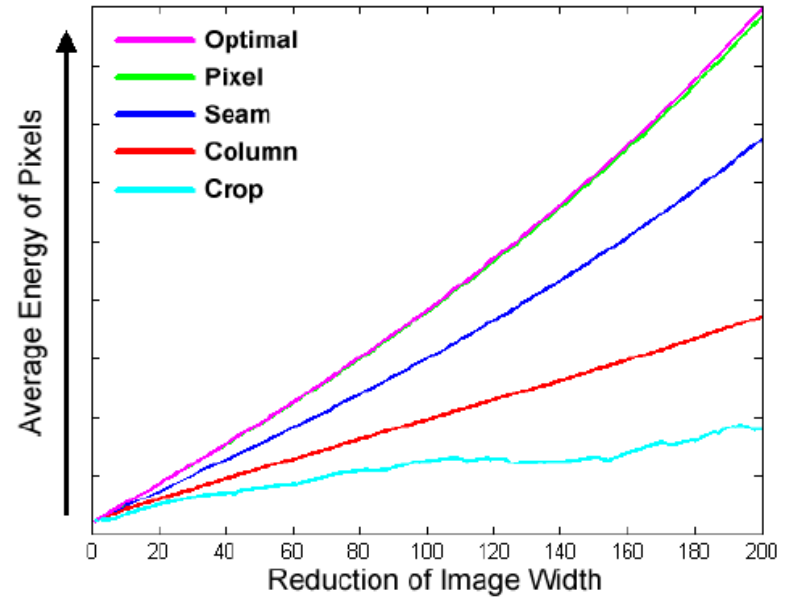
pixel



seam



optimal



# Solution

- Retargetting: Keeping important details while removing visually uninteresting parts



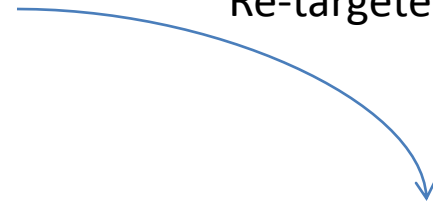
Re-targeting



# Problems with the solution



Re-targeted

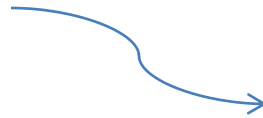




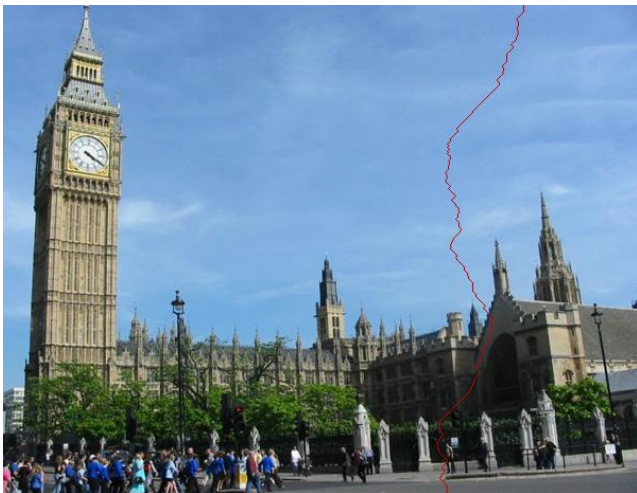
# Technical Details



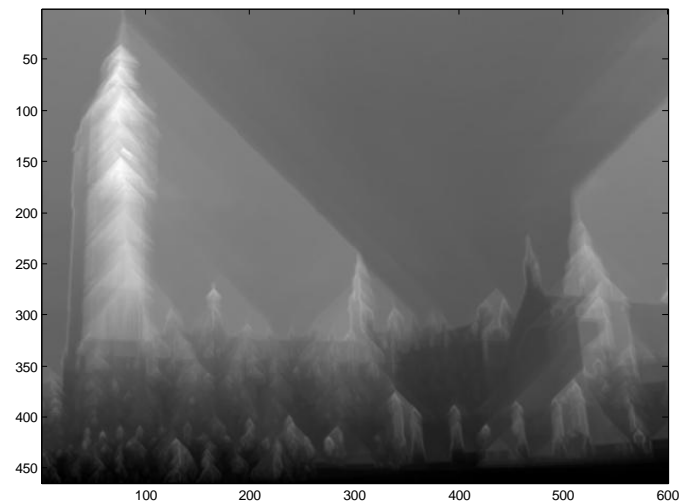
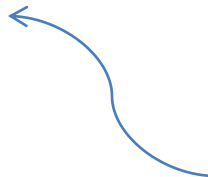
Detect interesting parts



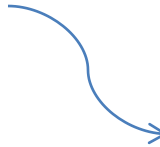
Find minimal path



Remove minimal seam



# Detecting interesting parts



Sobel operator

$$\mathbf{G}_x = \begin{bmatrix} +1 & 0 & -1 \\ +2 & 0 & -2 \\ +1 & 0 & -1 \end{bmatrix} * \mathbf{A} \quad \text{and} \quad \mathbf{G}_y = \begin{bmatrix} +1 & +2 & +1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix} * \mathbf{A}$$

$$\mathbf{G} = \sqrt{\mathbf{G}_x^2 + \mathbf{G}_y^2}$$

# Find Minimal Path

- Assume we have the energy matrix  $E$
- We need to find the minimal path from top to bottom.
- To solve this we need to check all possible paths
- Luckily we have a constraint: the path needs to be 4 connected

# Find Minimal Path - trivial

```
function out = find(E)
    for i = 1:maxx
        temp = findM(i,1)
        if temp < current_minimum
            current_minimum = temp
        end
    end
end
```

```
function out = findM(x,y,E)
    if y > maxy
        out = 0
    else
        out = E(x,y) + min(findM(x-1,y+1,E), findM(x,y+1,E), findM(x+1,y+1,E))
    end
end
```

E =

0.3	0.2	0.4	0.3
0.2	0.3	0.4	0.1
0.1	0.2	0.4	0.3
0.4	0.1	0.3	0.2

Complexity:  $O(\text{maxx} \times 3^{\text{maxy}}) \sim \text{too much}$

# Find Minimal Path - memoization

```
function out = find(E)
    for i = 1:maxx
        temp = findM(i,1)
        if temp < current_minimum
            current_minimum
        end
    end
```

```
function out = findM(x,y,E)
    if isnan(table(x,y))
        table(x,y) = rfindM(x,y,E)
        out = table(x,y)
    else
        out = table(x,y)
    end
```

```
function out = rfindM(x,y,E)
    if y > maxy
        out = 0
    else
        out = E(x,y) + min(findM(x-1,y+1,E), findM(x,y+1,E), find(x+1,y+1,E))
    end
```

## Complexity:

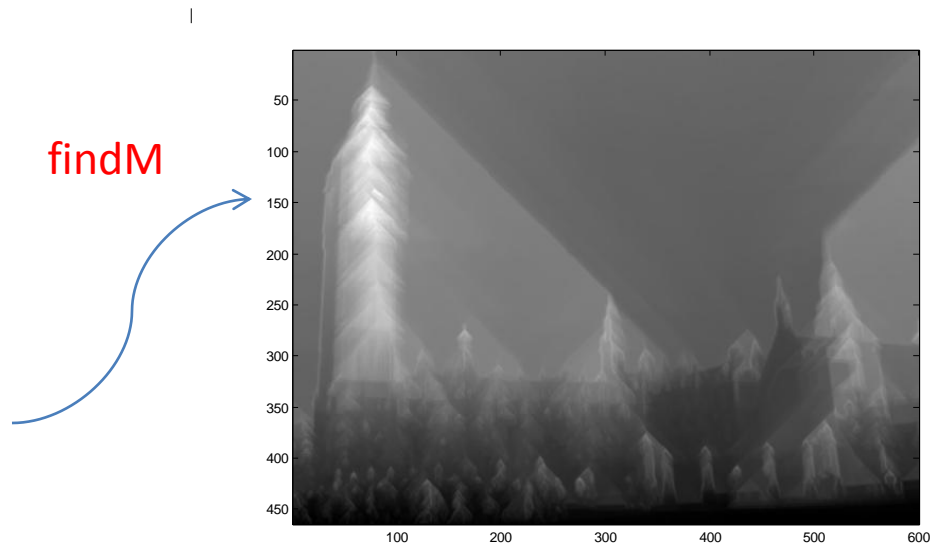
$O(\text{maxx} \times \text{maxy})$  - for making the table

$O(\text{maxx})$  - for the main procedure

STILL A PROBLEM!

# Find Minimal Path – dynamic programming

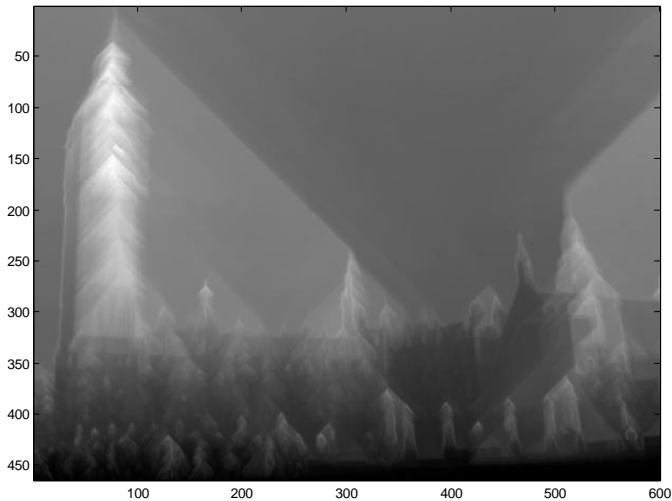
```
function out = findM(E)
    for i = maxy:-1:1
        for j = 1:maxx
            table(x,y) = E(x,y) + min(M(x,y+1),M(x-1,y+1),M(x+1,y+1))
        end
    end
end
```



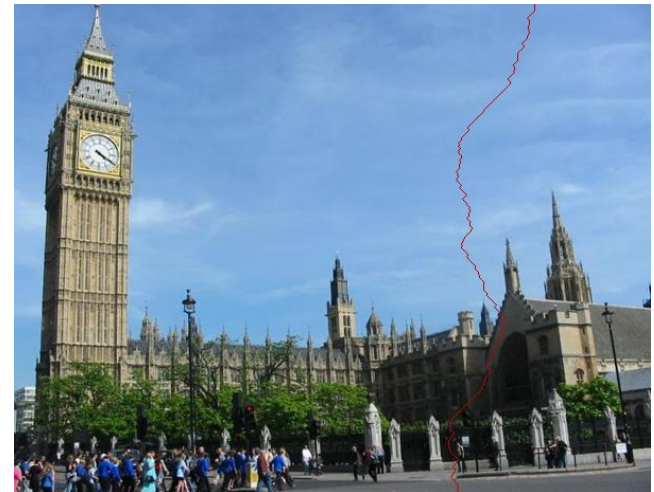
Now we can just do a greedy search

# Find Minimal Path - greedy

```
function seam = findS(table)
[v s] = min(table(:,1))
seam = s
for y = 2:maxy
    [vn sn] = min(table(s-1,y),table(s,y),table(s+1,y))
    s = s + sn - 2
    seam(y) = s
end
```



findS

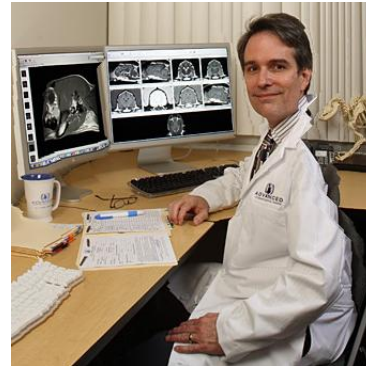


# Improvements

- Improving performance
  - It is possible to code the program in C or Java which will improve the running time by several orders of magnitude.
  - Saving the previous calculations so that the program doesn't compute the same things over and over again
  - Precomputing all possible minimal paths and keeping them in order so that no on-the-fly computing is necessary
- Improving quality
  - Different energy functions that preserve the objects better
    - Entropy, visual saliency, eye gaze movement
  - User corrected seam removal
    - Marking desirable or undesirable objects in frame for removal



# Conclusion



With people and other objects of recognizable shape the algorithm when removing a lot of the image creates undesirable effects



With buildings and similar objects the algorithm performs fairly well, however artifacts such as the interruption of straight lines is obvious