# Seam Carving for Content-Aware Image Resizing

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#### **Table of Contents**

O1 Abstract
O4 Algorithm

02 Motivation 05 Results

Problem Definition 06 References

# O1 Abstract

#### **Abstract**

- Seam Carving is a <u>content-aware</u> <u>image resizing</u> method.
- Adjusting unnoticeable pixels to avoid distorting the content of an image.



Original image



Scaling



Seam carving

#### **Abstract**

- The importance of each pixel is determined by its energy value.
- There are several ways to define an energy function, e.g. entropy, L1-norm of gradient.
- <u>Seam</u> is a <u>8-connected path</u> from top to bottom, or left to right
- Enlarging an image by inserting seams.
- Shrinking an image by removing seams.
- Aside from image resizing, seam carving can also be used for <u>content amplification</u> and <u>object removal</u>.

# 02 Motivation

#### **Motivation**

- We found this paper when searching for related articles of image quilting (homework 3).
- Highly practical. Most people resize an image by cropping or scaling; however, it may sacrifice some important contents or lead to serious distortion problem.
- Many applications. It helps us complete some other interesting tasks, like object removal and content enhancement.

# 03 Problem Definition

#### **Problem Definition**

- Do "content-aware" image resizing for both expansion and reduction.
- Apply seam carving on tasks of content amplification and object removal.

# O4 Algorithm

#### **List of Tasks**

- Determine the importance of each pixel.
- Find out a seam to be carved out or inserted.

- Image reduction from one dimension.
- Image enlargement from one dimension.
- Image retargeting from two dimensions.
- Object removal.
- Content amplification.

#### **Energy Function**

- Energy function is used to determine the importance of each pixel.
- A pixel with <u>high energy value</u> is regrarded as an important pixel.
- We have tried  $L_1$ -norm of gradient and entropy.

 $L_1$ -norm of gradient

$$e_1(\mathbf{I}) = \left| \frac{\partial}{\partial x} \mathbf{I} \right| + \left| \frac{\partial}{\partial y} \mathbf{I} \right|$$

entropy

$$E(I(X)) = -\sum_{i=1}^{n} p(x_i) \log_2 p(x_i)$$

#### Find out a Seam - Background

- Example: reduce the width of an image.
- To <u>preserve energy</u> of an image, we would like to remove pixels with low energy values.
- [Remove\_pix] Remove low energy pixels from each row? zig-zag effect
- [Remove\_col] Remove columns with the lowest energy value ? not continuous
- [Crop] Cropping (a sub-window that contains the highest energy)? sacrifice the content

## Find out a Seam - Background

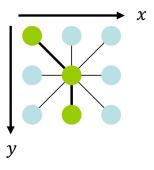


#### Find out a Seam - Definition

- To preserve both <u>energy</u> and <u>visual coherence</u>, we need "seam".
- Seam is a <u>8-connected path</u>.

Vertical seam: 
$$S^{X} = \{(x(i), i)\}_{i=1}^{n}, \text{ s.t. } \forall i, |x(i) - x(i-1)| \le 1$$

Horizontal seam: 
$$s^{Y} = \{j, y(j)\}_{i=1}^{m}$$
, s.t.  $\forall j, |y(j) - y(j-1)| \le 1$ 



• Minimize the cost of a seam by dynammic programming.

Vertical seam: 
$$M(i,j) = e(i,j) + \min(M(i-1,j-1), M(i-1,j), M(i-1,j+1))$$

Horizontal seam: 
$$M(i,j) = e(i,j) + \min(M(i-1,j-1), M(i,j-1), M(i+1,j-1))$$

Image reduction from one dimension.

Reduce the width:  $n \times m \rightarrow n \times m'$ , m > m'

Reduce the height:  $n \times m \rightarrow n' \times m, n > n'$ 

Repeat the following steps until achieving our target.

Step 1: Obtain energy values of all pixels.

Step 2: Find a seam with the lowest cost.

(width: vertical, height: horizontal)

Step 3: Remove the seam from an image.



Image insertion from one dimension.

```
Expand the width: n \times m \to n \times m', m < m'
Expand the height: n \times m \to n' \times m, n < n'
```

There are three methods to insert seams.

```
Method 1: Insert one by one.
```

Method 2: Insert all seams in one time.

Method 3: Insert seams in batches.

• [Insert One by one] Repeat the following steps until achieving our target.

```
Step 1: Find a seam to be <u>removed</u> from the image.

(width: vertical, height: horizontal)
```

Step 2: <u>Insert</u> the seam <u>back</u> to the image.

Step 3: <u>Duplicate</u> the seam by averaging it with its left and right neighbors.

Problem: <u>stretching artifact</u> by choosing the <u>same seam</u>.

"Insert one by one" may create a stretching artifact by choosing the same seam.





• [Insert all seams in one time]

```
Step 1: Find <u>all the seams</u> to be removed.

(Goal: Insert n seams, #seams to be removed: n)
```

Step 2: Insert seams back to the image in backward, and duplicate it afterwards.

Problem: Duplicates all the seams in an image is equivalent to standard scaling.

50% enlargement

• "Insert all seams in one time" may choose all the seams in an image, which is equivalent to standard scaling.





[Insert seams in batches]

Break the process of "Insert all seams in one time" into several steps.





## Image Retargeting from Two Dimensions

Image retargeting from two dimensions.

```
n \times m \rightarrow n' \times m'
```

We propose the new method.

Paper: Horizontal seams + vertical seams → order is the matter. [Not yet]

Ours: Proportional scaling + seam removal.

## Image Retargeting from Two Dimensions - Author

Optimal order.

Vertical first? Horizontal first? Alternate between the two?

Image reduction from two dimension.

Limitation: 
$$n' < n, m' < m$$

$$\min_{s^{x}, s^{y}, \alpha} \sum_{i=1}^{k} E(\alpha_{i} s_{i}^{x} + (1 - \alpha_{i}) s_{i}^{y})$$

$$T(r, c) = \min \begin{cases} T(r - 1, c) + E\left(s^{x} \left(I_{(n-r-1)\times(m-c)}\right)\right) \\ T(r, c - 1) + E\left(s^{y} \left(I_{(n-r)\times(m-c-1)}\right)\right) \end{cases}$$



#### Image Retargeting from Two Dimensions - Ours

Image retargeting from two dimensions.

Step 1: Set the original size as (h, w) and the target size as (h', w')

Step 2: Calculate the ratio  $r_h = \frac{h'}{h}$  and  $r_w = \frac{w'}{w}$ .

Step 3: Scale the image by the factor of  $r_h$  if  $r_h > r_w$ , and vice versa.

Step 4: Do seam removal to achieve the target size.

Advantages.

Easier ©.

No limitation: Apart from image reduction, we can also do enlargement.

#### **Content Amplification**

- Content amplification.
  - Step 1: Do proportional scaling to enlarge an image.
  - Step 2: Apply seam removal on the image to carve the image back to its original size.





#### **Object Removal**

- Object removal
  - Step 1: Mark the target object to be removed.
  - Step 2: Give the marked pixels <u>low enough energy values</u>, so that they tend to be removed.

    This step should be done in every cycle of seam removal.
  - Step 3: Do <u>seam removal</u> until all marked pixels are gone.

    The orientation is set by user.
  - Step 4: Do seam insertion until the image back to its original size.

# 05 Results



• The dolphin looks good under both energy functions. However, they show different effect!



Original image



 $L_1$ -norm of gradient



Entropy

• Noticeable artifacts: the unnatural variation of light and shadow.



Original image



 $L_1$ -norm of gradient



Entropy

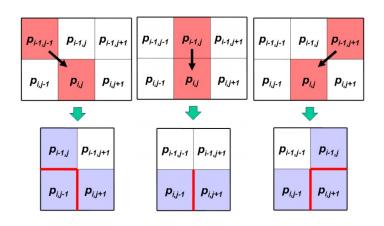
Forward enregy.

$$M(i,j) = \min \begin{cases} M(i-1,j-1) + C_L(i,j) \\ M(i-1,j) + C_U(i,j) \\ M(i-1,j+1) + C_R(i,j) \end{cases}$$

$$C_L = |I(i,j+1) - I(i,j-1)| + |I(i-1,j) - I(i,j-1)|$$

$$C_U = |I(i,j+1) - I(i,j-1)|$$

$$C_R = |I(i,j+1) - I(i,j-1)| + |I(i-1,j) - I(i,j+1)|$$



"Forward energy" works better!



 $L_1$ -norm of gradient



Entropy



Forward energy

The mouse is diminishing.



Original Image



 $L_1$ -norm of gradient



Forward energy

Using a "protective mask" to protect important content of the image.

Give the red region "very high" energy value.





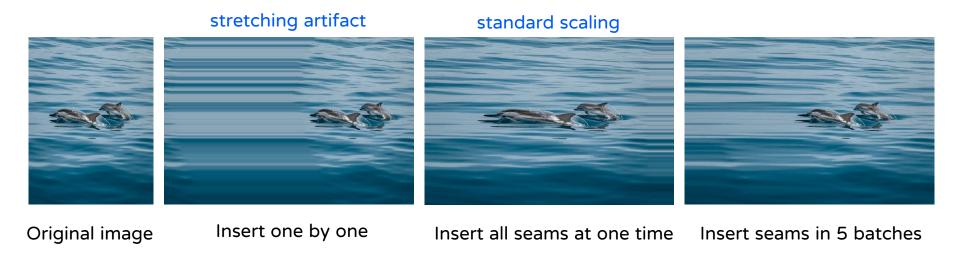


Protective mask

With mask

Without mask

Insert one by one vs. Insert all seams at one time vs. Insert seams in 5 batches.



## **Image Retargeting from Two Dimensions**

Our method supports us to do enlarge or shrink an image from two dimensions!



Original image (360, 640)



Reduction (250, 550)



Expansion (400, 800)

## **Content Amplification**

• They are different in the cloudy sky.



Original Image



 $L_1$ -norm of gradient



Forward energy

## **Object Removal**

Define a "mask".

Give the blue region "very low" energy value.



Original image

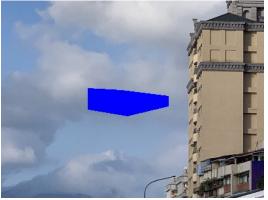


Image with a mask

### **Object Removal**

- Orientation does great impact on the result.
- There are some noticeable artifacts in the image.

Improvement: insert the seams around masks. [Not yet]



Remove horizontally



Remove vertically

# 06 References

#### References

- Avidan, Shai, and Ariel Shamir. "Seam carving for content-aware image resizing." ACM SIGGRAPH 2007 papers. 2007. 10-es.
- Rubinstein, Michael, Ariel Shamir, and Shai Avidan. "Improved seam carving for video retargeting." *ACM transactions on graphics (TOG)* 27.3 (2008): 1-9.

# **Thank You**