Lecture 9

Saliency and Retargeting

Administrative

- A2 is due May 2nd

Administrative

Recitation

- Geometric transformations
- May cover some exam practice questions if time allows as last time

CSE 455 Roadmap

Convolutions
Edges
Descriptors

Motion
Tracking

Camera
3D Geometry

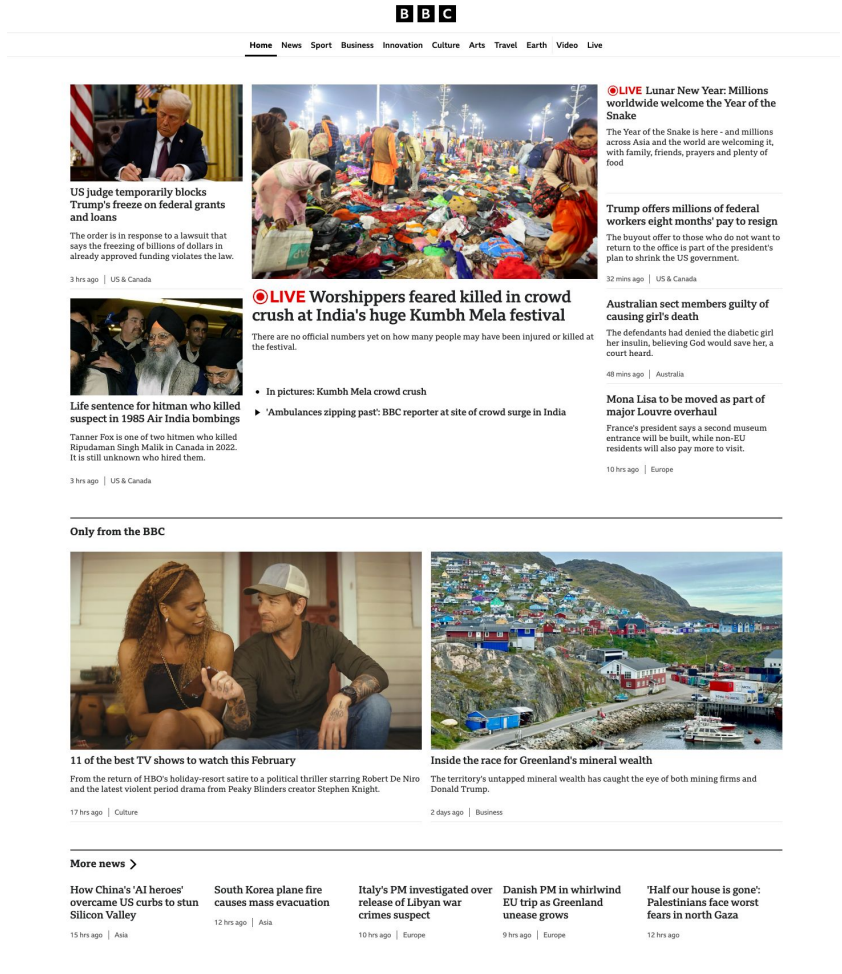
Segmentation
Clustering
Detection

Linear Models
(Conv) Neural networks

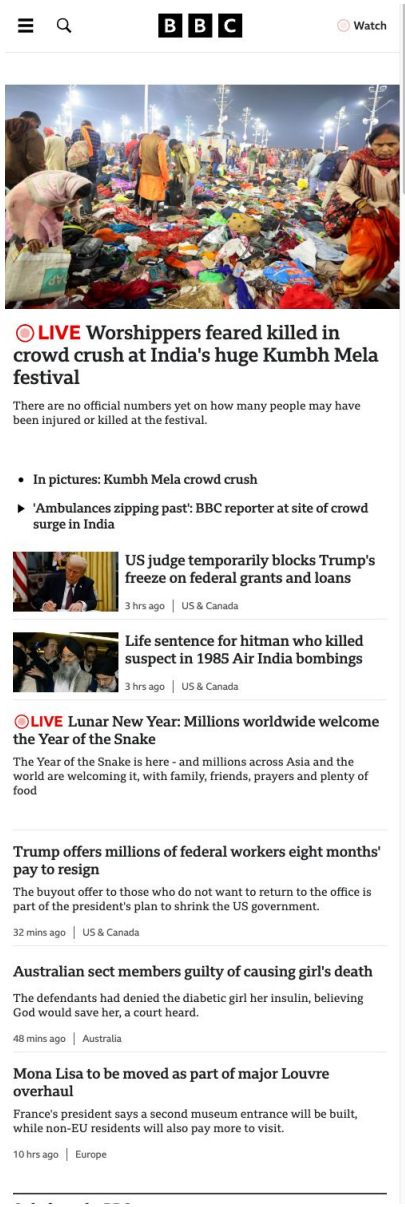
Today’s agenda

- Image retargeting
- Seam carving
- Applications
- Forward algorithm

Content Retargeting

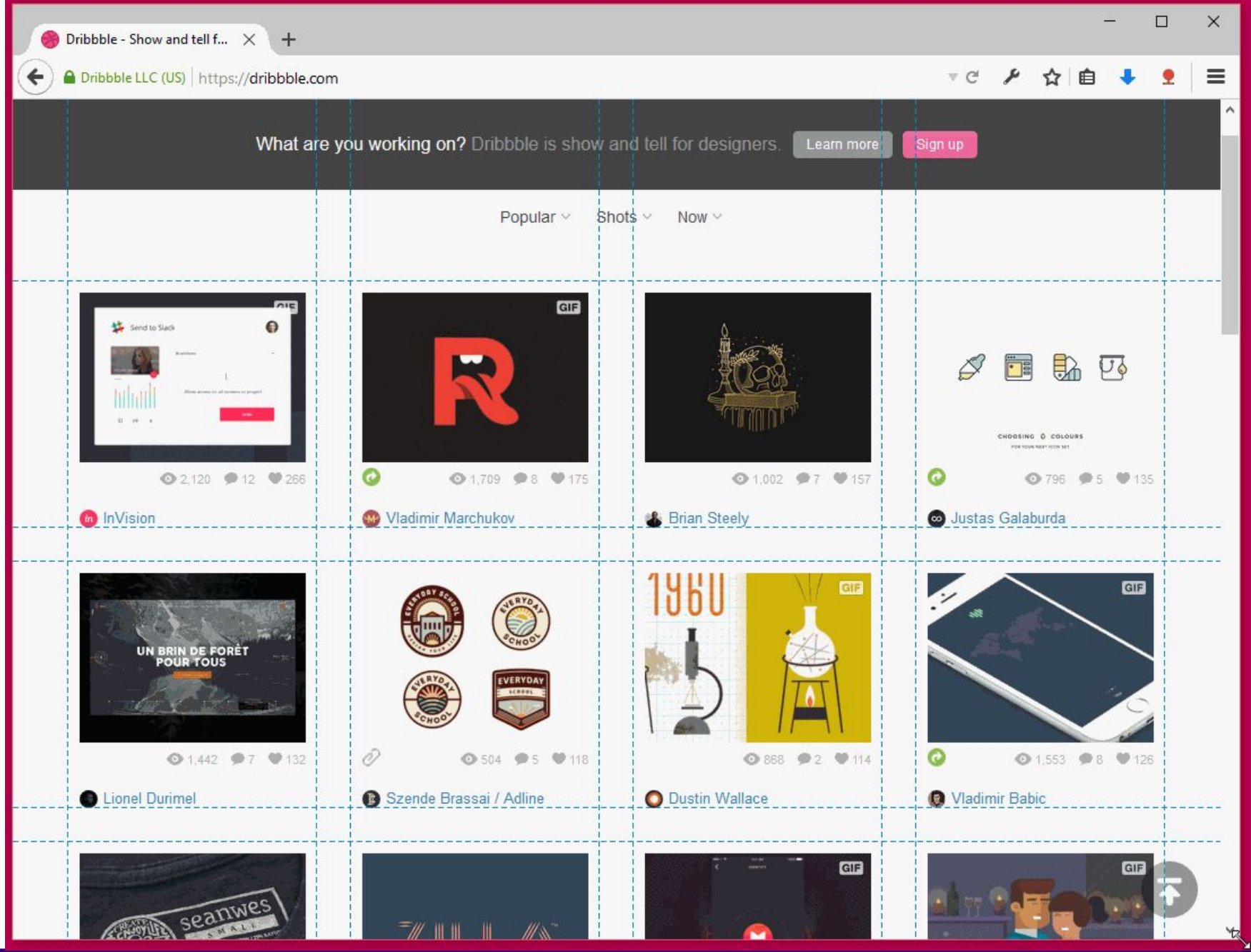


PC



iPhone

Page Layout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Vladimir Marchukov
Brian Steely
Justas Galaburda
UN BRIN DE FORÊT POUR TOUS
Lionel Durimel
Szende Brassai / Adline
Dustin Wallace
Vladimir Babic

Display Devices

Have you ever seen this?


TURKISH
The following film has been modified from its original version. It has been formatted to fit this screen and edited for content.

Simple Media Retargeting Operators

Letterboxing Cropping

Content-aware Retargeting Operators

Content-*aware*

"Important" content

Content-*oblivious*

Content-aware Retargeting

Input

Scale

Crop

Content-aware

“less-Important” content

Image Retargeting (“More” Formally)

Problem statement

- Input image I of size n x m
- Output image I’ of size n’ x m’

Output image should be geometrically and semantically consistent with input image

How can we define consistency?

In large, we would expect retargeting to:

1. Adhere to the geometric constraints (display/aspect ratio)
2. Preserve the important content and structures
3. Limit artifacts

How can we define consistency?

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1. Adhere to the geometric constraints (display/aspect ratio)
2. Preserve the important content and structures
3. Limit artifacts

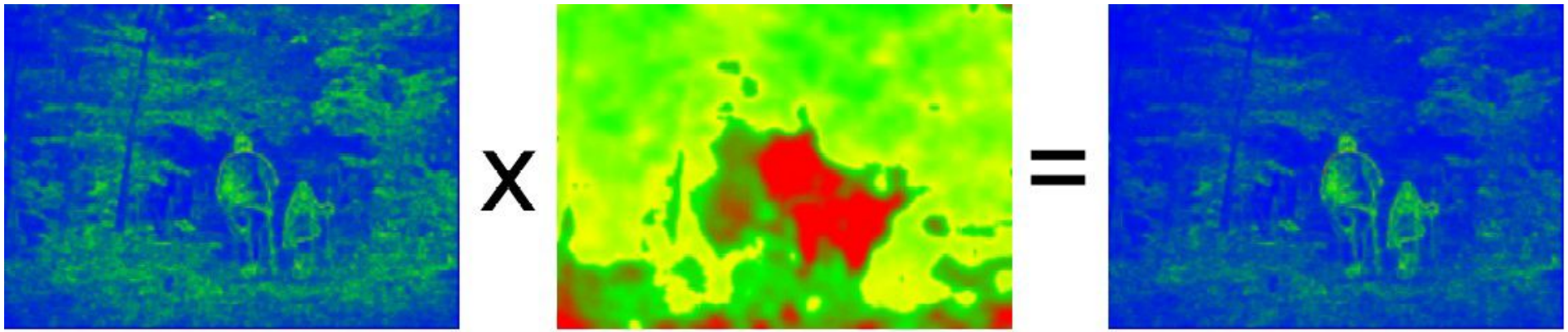
Very ill-posed!

- How do we define what is important?
  - Is there a universal important vs unimportant?
- Would different people find different image regions more or less important?
- What about artistic impression in the original content?

Importance (Saliency) Measures

- A function $\mathcal{S} : n \times m \rightarrow [0, 1]$

- Ideas from human perception



First stage: coarse scan over entire image

Second stage: more focused attention on specific region

Wang et al. A Two-stage approach to saliency detection in images 2008

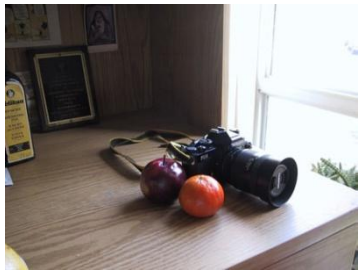
Importance (Saliency) Measures

- A function $\mathcal{S} : n \times m \rightarrow [0, 1]$

- More sophisticated: attention models, eye tracking (gazing studies), face detectors, …

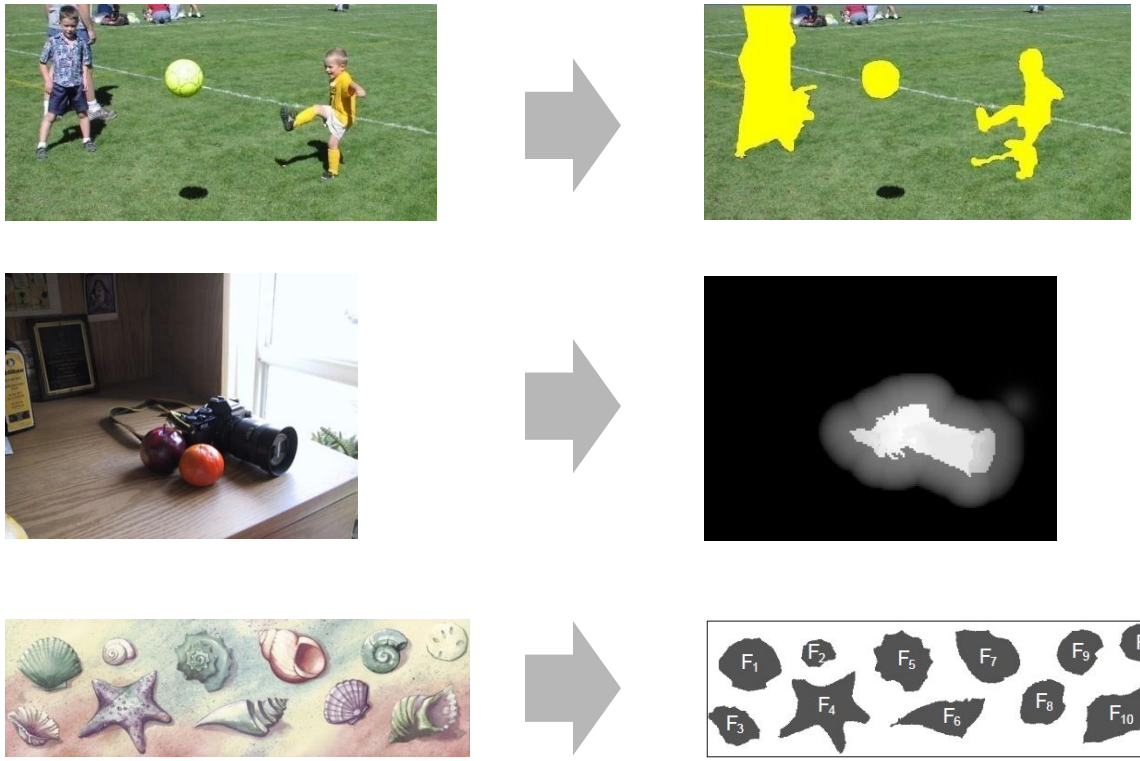
Judd et al. *Learning to predict where people look* ICCV 2009

General Retargeting Framework



General Retargeting Framework

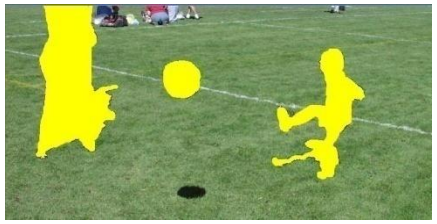
Step 1. Define an energy function
$\mathbf{E(I)}$ (interest, importance, saliency)



General Retargeting Framework

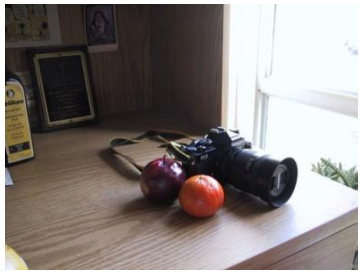
Step 1. Define an energy function $\mathbf{E(I)}$ (interest, importance, saliency)

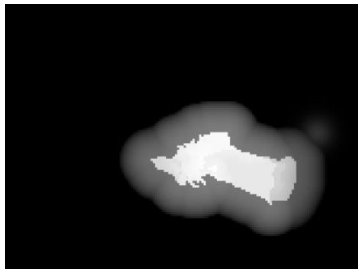
Step 2. Use some operator(s) to change the image $\mathbf{I}$



Recompose

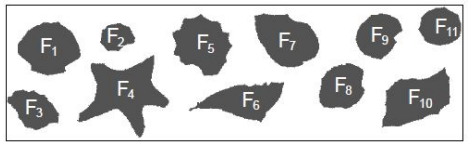
Setlur et al. [2005]





Crop

Santella et al. [2005]



Warp

Gal et al. [2006]

Potential Retargeting Approaches

- Optimal Cropping Window

Potential Retargeting Approaches

- Done manually in the movie industry for many years

Today’s agenda

- Image retargeting
- Seam carving
- Applications
- Forward algorithm

Seam Carving

- Assume input I is size m x n
- Output I is m x n',
  - where n'<n

- Basic Idea: remove unimportant pixels from the image
  - Unimportant = pixels with less "energy"

Seam Carving

- Assume input I is size m x n
- Output I is m x n',
  - where n'<n

- Basic Idea: remove unimportant pixels from the image
  - Unimportant = pixels with less "energy"

$$E(I) = |\frac{\partial I}{\partial x}| + |\frac{\partial I}{\partial y}|$$

$$E(I) = \sqrt{(\frac{\partial I}{\partial x})^2 + (\frac{\partial I}{\partial y})^2}$$

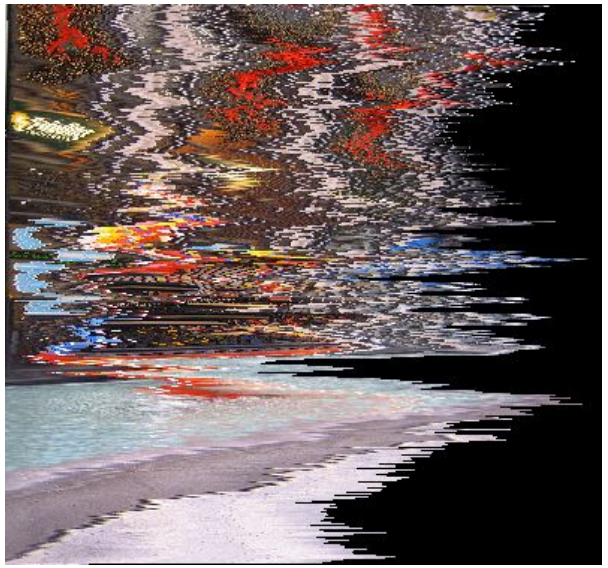
- Intuition for gradient-based energy:
  - Preserve edges – we are sensitive to edges
  - Try remove content from smoother areas
  - Simple enough for producing some nice results

Let’s do an experiment

We calculate the energy for this image.

Q1. Can we just remove the K **pixels** with the lowest energy?

Pixel Removal



Optimal

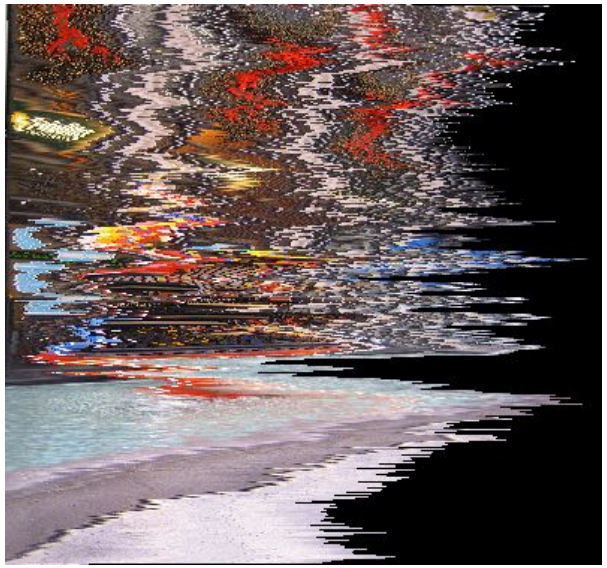
Let’s do an experiment

We calculate the energy for this image.

Q1. Can we just remove the K **pixels** with the lowest energy?

Q2. Can we remove the K **pixels** with the lowest energy **per rows**?

Pixel Removal



Optimal

Least-energy pixels
(per row)

Let’s do an experiment

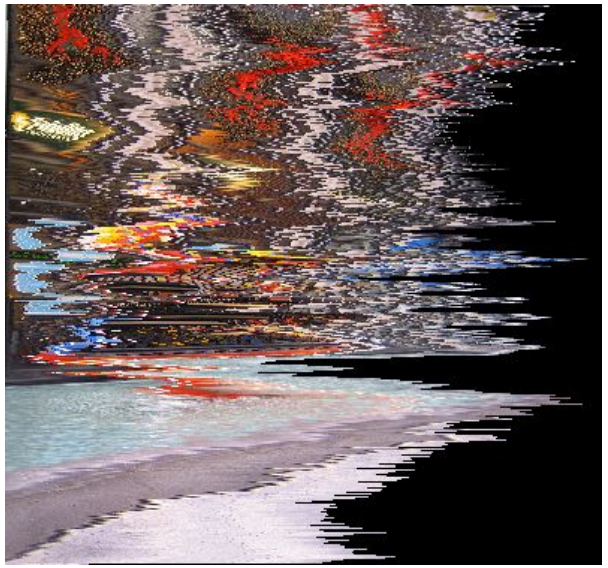
We calculate the energy for this image.

Q1. Can we just remove the K **pixels** with the lowest energy?

Q2. Can we remove the K **pixels** with the lowest energy **per column**?

Q3. Can we remove the K **columns** with the lowest energies?

Pixel Removal



Optimal

Least-energy pixels (per row)

Least-energy columns

Solution: A Seam

- A seam is a connected path of pixels from top to bottom (or left to right). Exactly one in each row (or column)

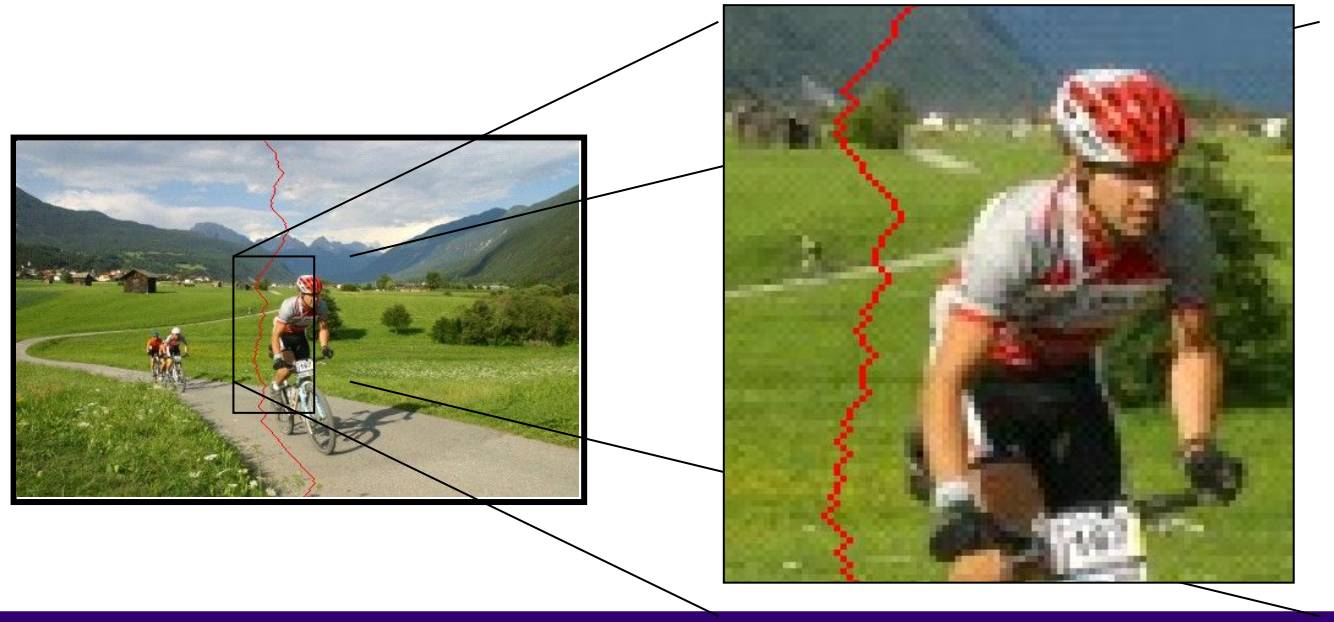
$$s^x = \{s_i^x\}_{i=1}^n$$

$$s^x = \{x(i), i\}_{i=1}^n$$

for every row i

$$\text{s.t. } \forall i, \; |x(i) - x(i-1)| \leq 1$$

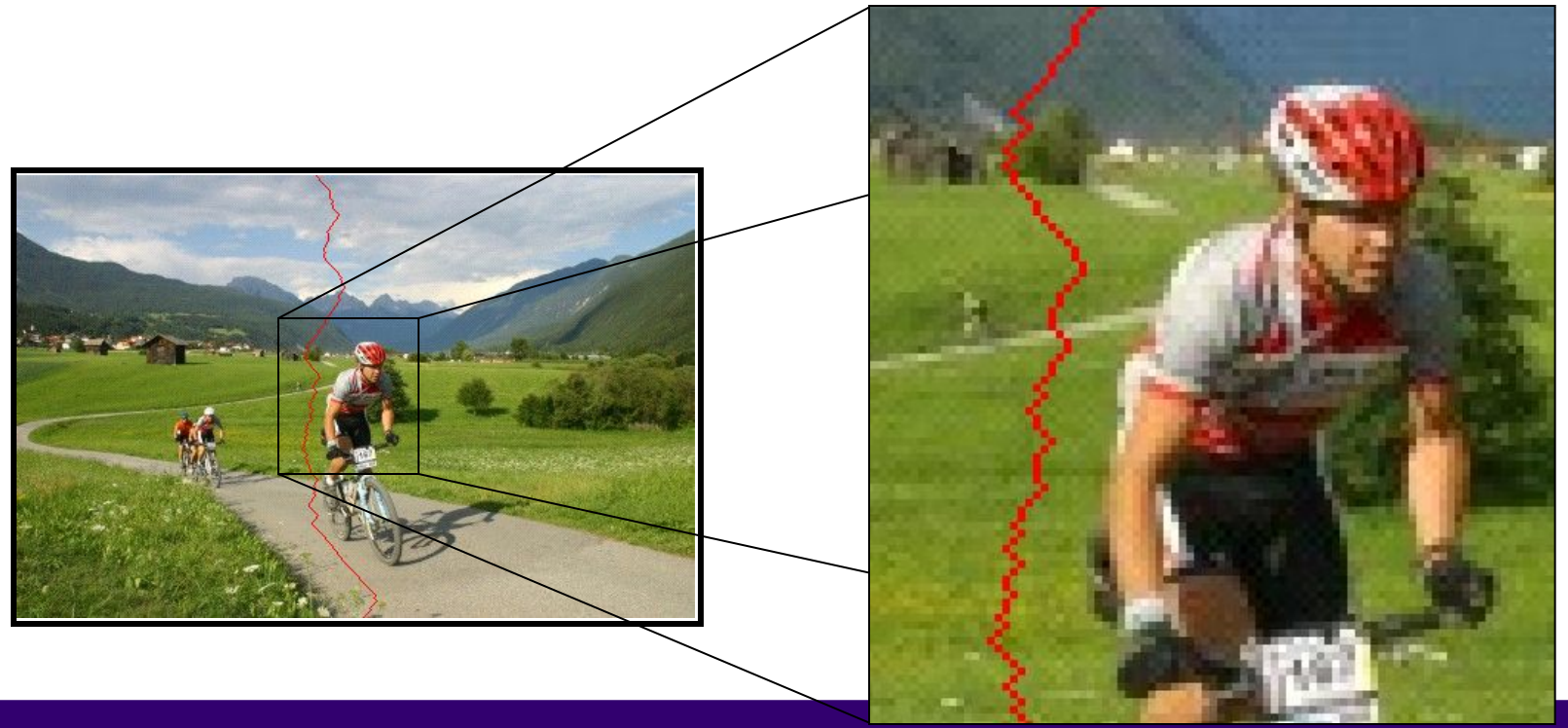
Ensure that seam is “connected”. Columns can only change by a maximum of 1 column



A Seam

- A connected path of pixels from top to bottom (or left to right). Exactly one in each row

$$\mathbf{s^x} = \{s_i^x\}_{i=1}^n = \{(x(i), i)\}_{i=1}^n, \text{ s.t. } \forall i, |x(i) - x(i-1)| \leq 1$$
$$\mathbf{s^y} = \{s_j^y\}_{j=1}^m = \{(j, y(j))\}_{j=1}^m, \text{ s.t. } \forall j |y(j) - y(j-1)| \leq 1$$

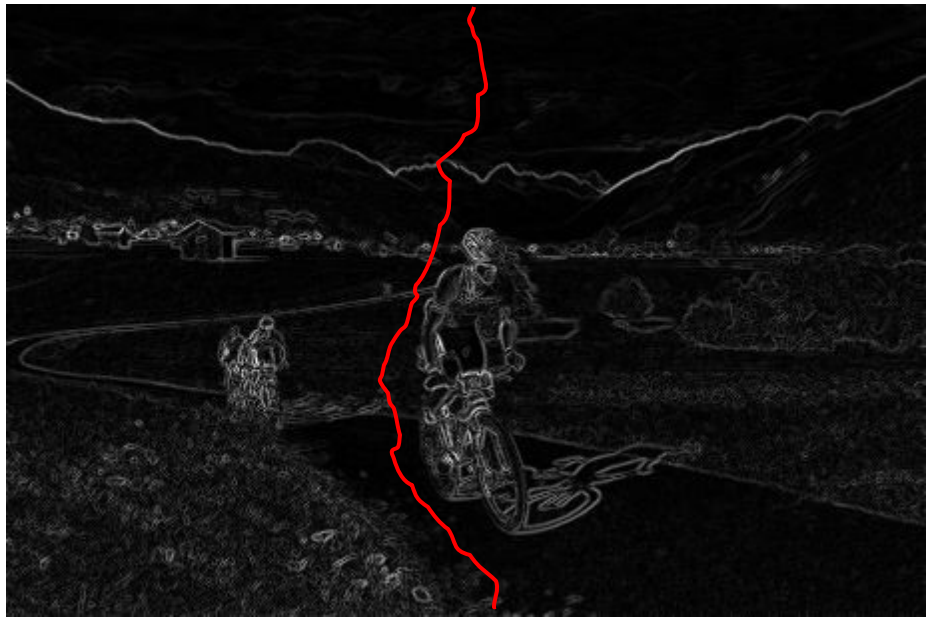


How do we find the optimal Seam?

Q: How many seam do we have for an image? O(???)

$$E(\mathbf{I}) = \left|\frac{\partial}{\partial x}\mathbf{I}\right| + \left|\frac{\partial}{\partial y}\mathbf{I}\right| \Rightarrow s^* = \arg\min_s E(s)$$

The Optimal Seam



Q: How many seam do we have for an image? O(???)

$$E(\mathbf{I}) = \left|\frac{\partial}{\partial x}\mathbf{I}\right| + \left|\frac{\partial}{\partial y}\mathbf{I}\right| \Rightarrow s^* = \arg\min_s E(s)$$

Brute force is not practical, but you must have seen this

An example here, and there are many more

746. Min Cost Climbing Stairs

Easy | Topics | Companies | Hint

You are given an integer array `cost` where `cost[i]` is the cost of `i^th` step on a staircase. Once you pay the cost, you can either climb one or two steps.

You can either start from the step with index `0`, or the step with index `1`.

Return *the minimum cost to reach the top of the floor*.

A: sol[i] = cost[i] + min(sol[i-1], sol[i-2])

Dynamic Programming

Input: Given an energy E(i, j)

5	8	12	3
4	2	3	9
7	3	4	2
5	5	7	8

Energy - E(i, j)

Dynamic Programming

- Create a cost matrix M with the following property:
  - M(i, j) = minimal cost of a seam going through pixel (i, j)
    - starting from j=0

M(i, j)

5	8	12	3
4	2	3	9
7	3	4	2
5	5	7	8

Energy - E(i, j)

Dynamic Programming

M(i, 0) = E(i, 0) of a seam going through pixel (i, j)

5	8	12	3

M(i, j)

5	8	12	3
4	2	3	9
7	3	4	2
5	5	7	8

Energy - E(i, j)

Dynamic Programming

Q. What do you think should be this value?

M(i, j)

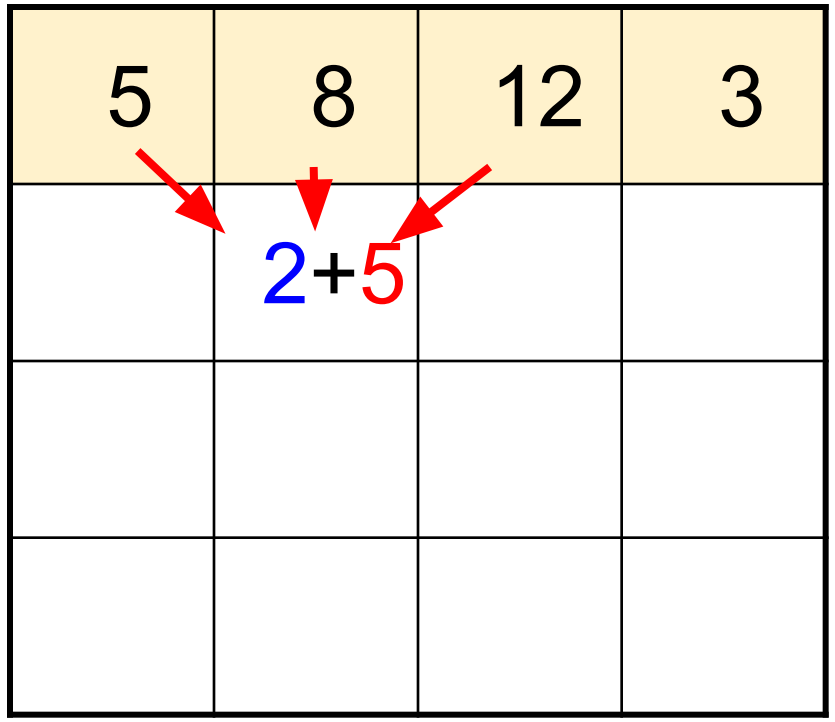
5	8	12	3
	?		

5	8	12	3
4	2	3	9
7	3	4	2
5	5	7	8

Energy - E(i, j)

Dynamic Programming

M(i, j) = total energy of seam going through pixel (i, j) from j=0



M(i, j)

5	8	12	3
4	2	3	9
7	3	4	2
5	5	7	8

Energy - E(i, j)

Dynamic Programming

The recurrence formula

$$\mathbf{M}(i,j) = E(i,j) + \min\left(\mathbf{M}(i-1,j-1), \mathbf{M}(i-1,j), \mathbf{M}(i-1,j+1)\right)$$

5	8	12	3
	2+5		

M(i, j)

5	8	12	3
4	2	3	9
7	3	4	2
5	5	7	8

Energy - E(i, j)

Dynamic Programming

M(i, j)

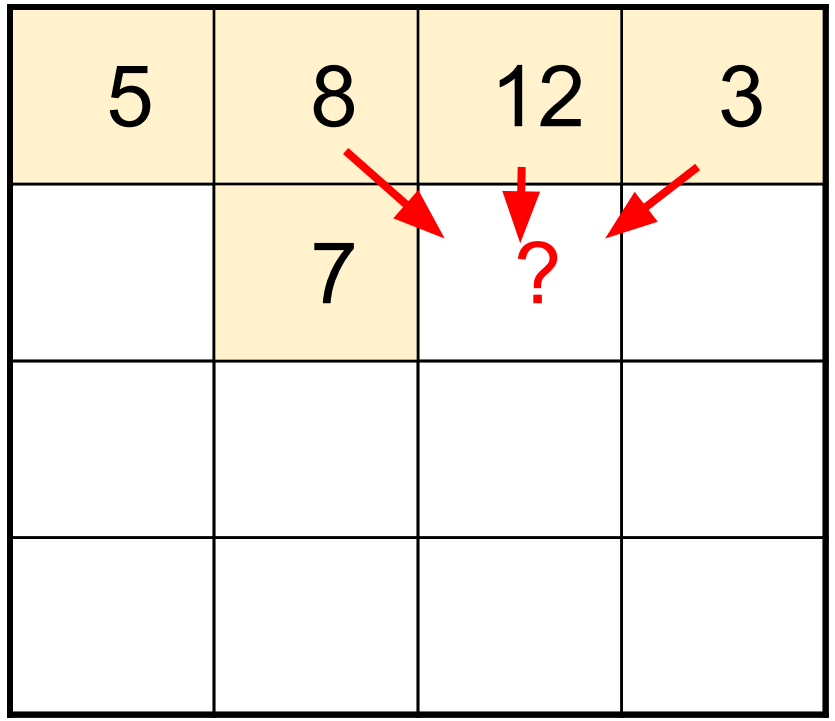
5	8	12	3
	7		

5	8	12	3
4	2	3	9
7	3	4	2
5	5	7	8

Energy - E(i, j)

Dynamic Programming

$$\mathbf{M}(i,j) = E(i,j) + \min\left(\mathbf{M}(i-1,j-1), \mathbf{M}(i-1,j), \mathbf{M}(i-1,j+1)\right)$$



M(i, j)

5	8	12	3
4	2	3	9
7	3	4	2
5	5	7	8

Energy - E(i, j)

Dynamic Programming

$$\mathbf{M}(i,j) = E(i,j) + \min\left(\mathbf{M}(i-1,j-1), \mathbf{M}(i-1,j), \mathbf{M}(i-1,j+1)\right)$$

5	8	12	3
	7	3+3	

M(i, j)

5	8	12	3
4	2	3	9
7	3	4	2
5	5	7	8

Energy - E(i, j)

Dynamic Programming

$$\mathbf{M}(i, j) = E(i, j) + \min\left(\mathbf{M}(i-1, j-1), \mathbf{M}(i-1, j), \mathbf{M}(i-1, j+1)\right)$$

M(i, j)

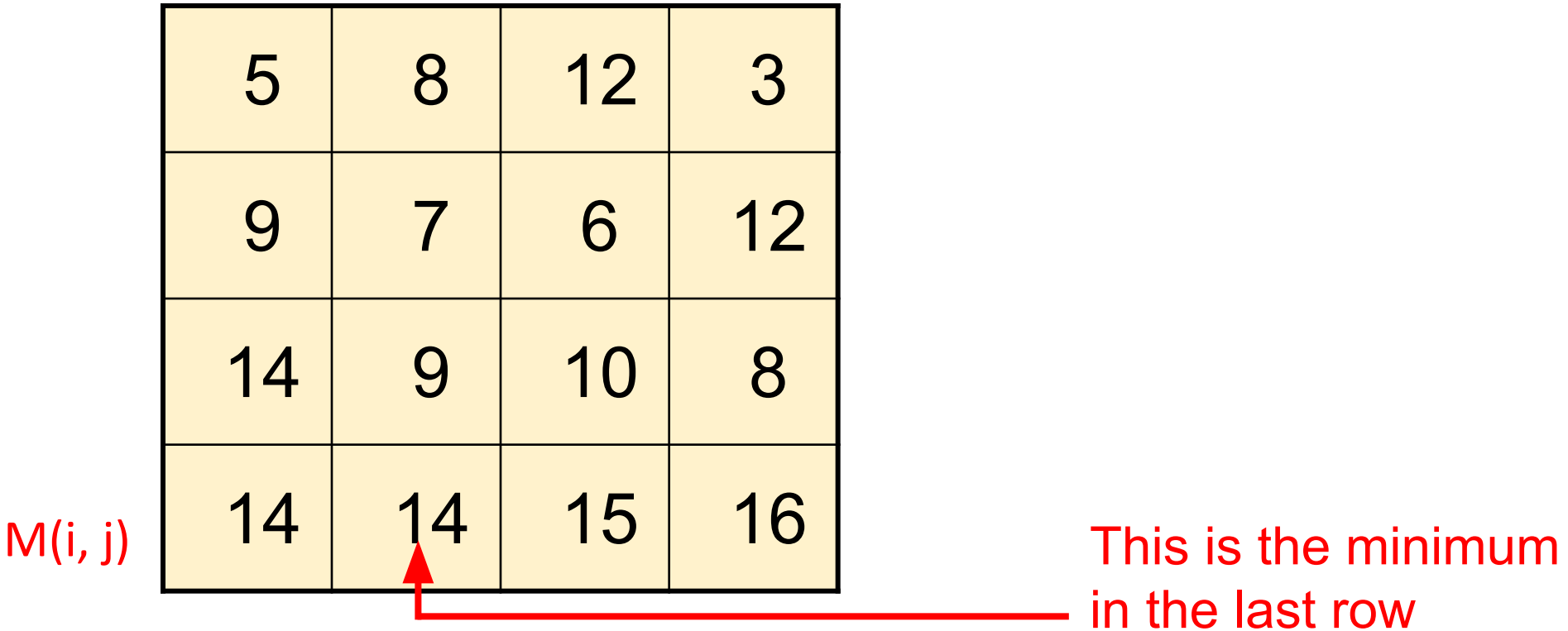
5	8	12	3
9	7	6	12
14	9	10	8
14	14	15	8+8

5	8	12	3
4	2	3	9
7	3	4	2
5	5	7	8

Energy - E(i, j)

Searching for minimum seam

Backtrack: Find the minimum M(i, j=m)



Backtrack

After finding minimum M(i, j) at row j,
find minimum M(i, j-1) but only be looking at neighboring locations: i-1, i, i+1

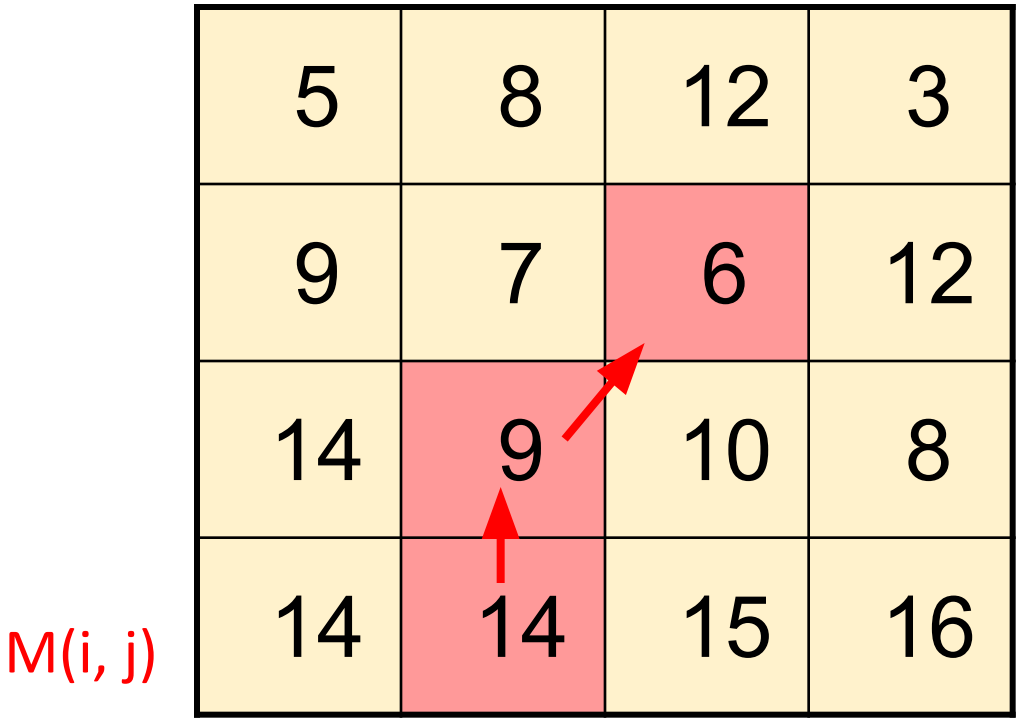
M(i, j)

5	8	12	3
9	7	6	12
14	9	10	8
14	14	15	16

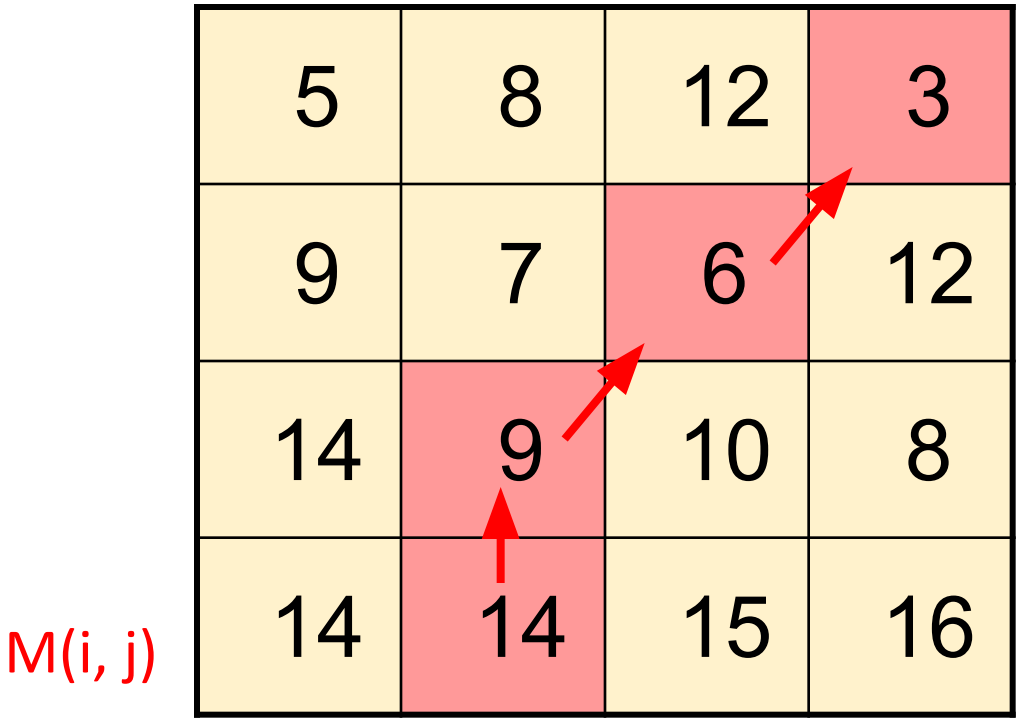
5	8	12	3
4	2	3	9
7	3	4	2
5	5	7	8

Energy - E(i, j)

Searching for Minimum



Searching for Minimum



The Optimal Seam - dynamic programming

- The recursion relation

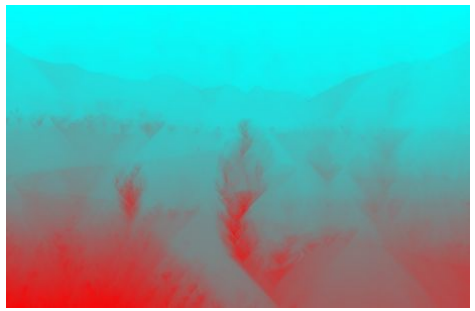
$$\mathbf{M}(i,j) = E(i,j) + \min\left(\mathbf{M}(i-1,j-1), \mathbf{M}(i-1,j), \mathbf{M}(i-1,j+1)\right)$$

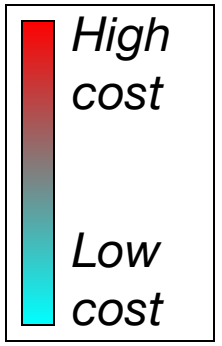
- **Q: What is the time complexity?**

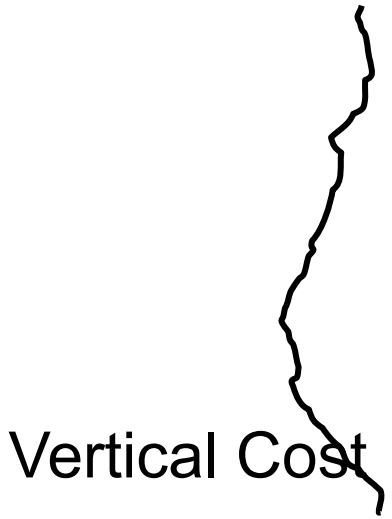
$$O(s \cdot n \cdot m)$$

(s=3 in the original algorithm)

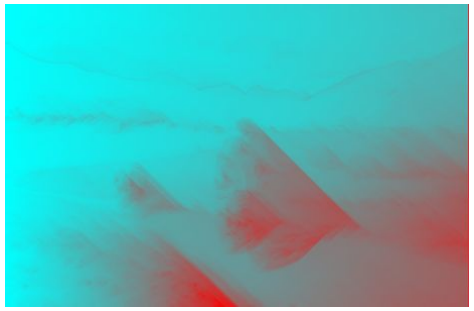
Vertical cost maps

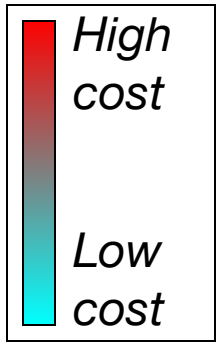






Horizontal cost maps





Horizontal Cost

Seam Carving

The Seam-Carving Algorithm

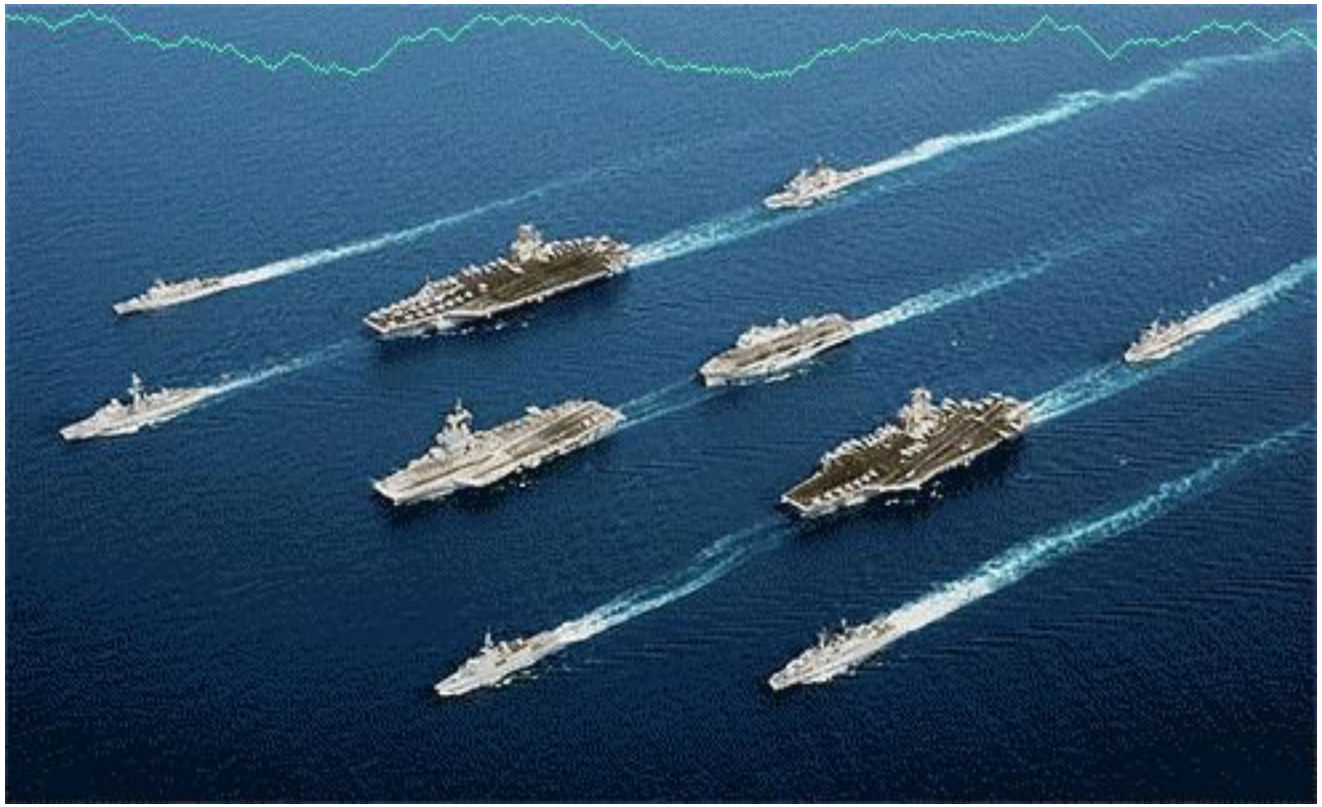
Algorithm: Seam carving
Input: Image I of size m x n
Output: Image I' of size m x n' where n' < n
I' = I
Do d=(n-n') times
    Compute energy map on I'
    Find optimal seam in E
    Remove s from im
Return I'

For vertical resize: transpose the image

Running time: O(dmn) or O(dsmn)

Changing Aspect Ratio

Another example



Example seam carving

Example seam carving

Changing Aspect Ratio

Original

Retargeting

Scaling

Changing Aspect ratio

Cropping

Retarget

Scaling

Changing Aspect Ratio

Original

Retarget

Scaling

Changing Aspect Ratio

Original

Retarget

Scaling

Questions

- Q: Will the result be the same if the image is flipped upside down?

Q. What if we simultaneously want to reduce both width and height?

m x n -> m’ x n’

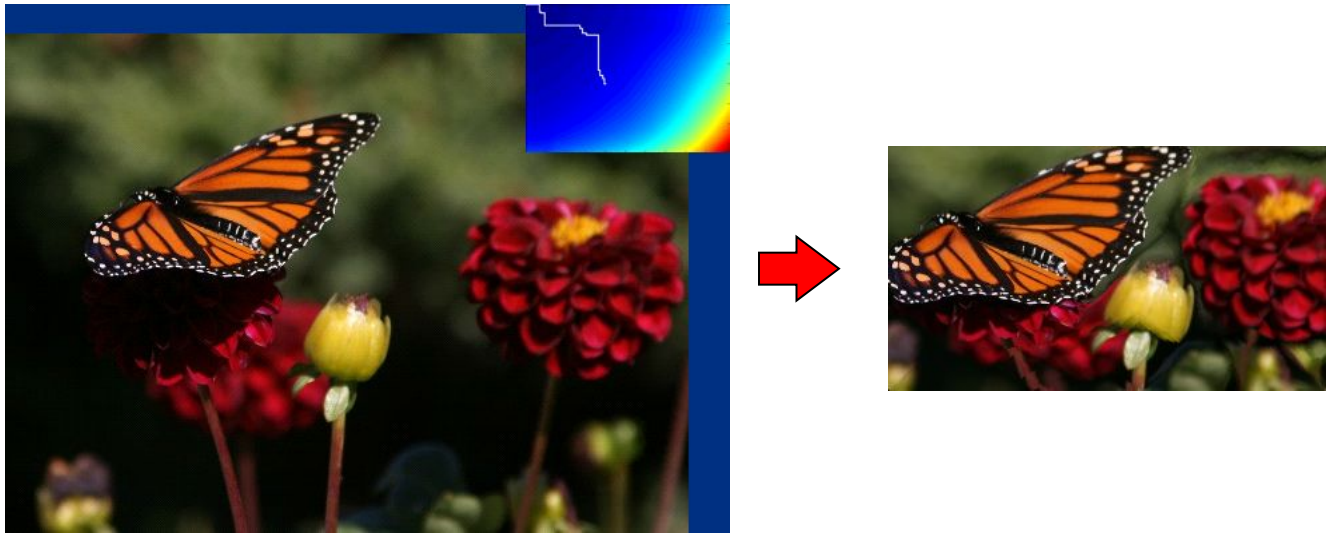
1. Should we remove horizontal seam first?
2. Should we remove vertical seams first?
3. Alternate between the two?
4. Any other ideas?

What if we simultaneously want to reduce both width and height?

m x n -> m’ x n’

1. Should we remove horizontal seam first?
2. Should we remove vertical seams first?
3. Alternate between the two?
4. Any other ideas?

Exercise: the optimal order can be found! Dynamic Prog (again! but this is not easy)



Retargeting in Both Dimensions

- Let T(r,c) denote a new cost matrix of obtaining an image of size (n-r)x(m-c).

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

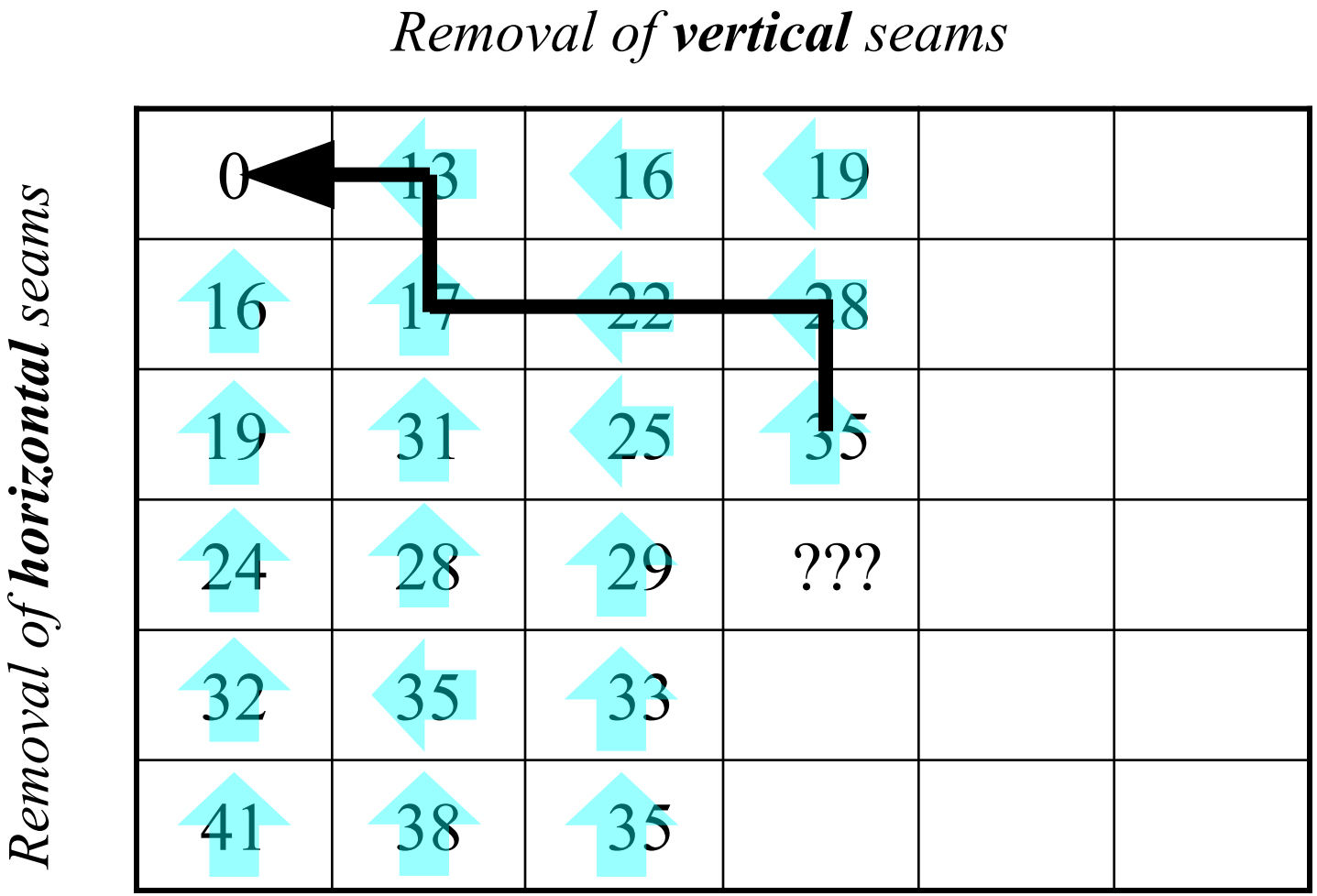
Retargeting in Both Dimensions

- Let T(r,c) denote a new cost matrix of obtaining an image of size (n-r)x(m-c).

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

where $E(\mathbf{s^x}(\mathbf{I_{n-r-1\times m-c}})$ is the cost of removing a horizontal seam from the image $\mathbf{I_{n-r-1\times m-c}}$

Optimal Order Map



Is it optimal…

- … for removing ONE seam?
- … for removing multiple seams?

Is it optimal…

- … for removing ONE seam?

- … for removing multiple seams?

  - Consider HVV (how many possible orderings?)
  - Cost(V) on HV not necessarily equal Cost(V) on VH
  - But we keep track of only one: min(HV,VH)…

Today's agenda

- Image retargeting
- Seam carving
- Dynamic programming
- More Applications
- Forward algorithm

Image expansion - Repeat the lowest energy seam?

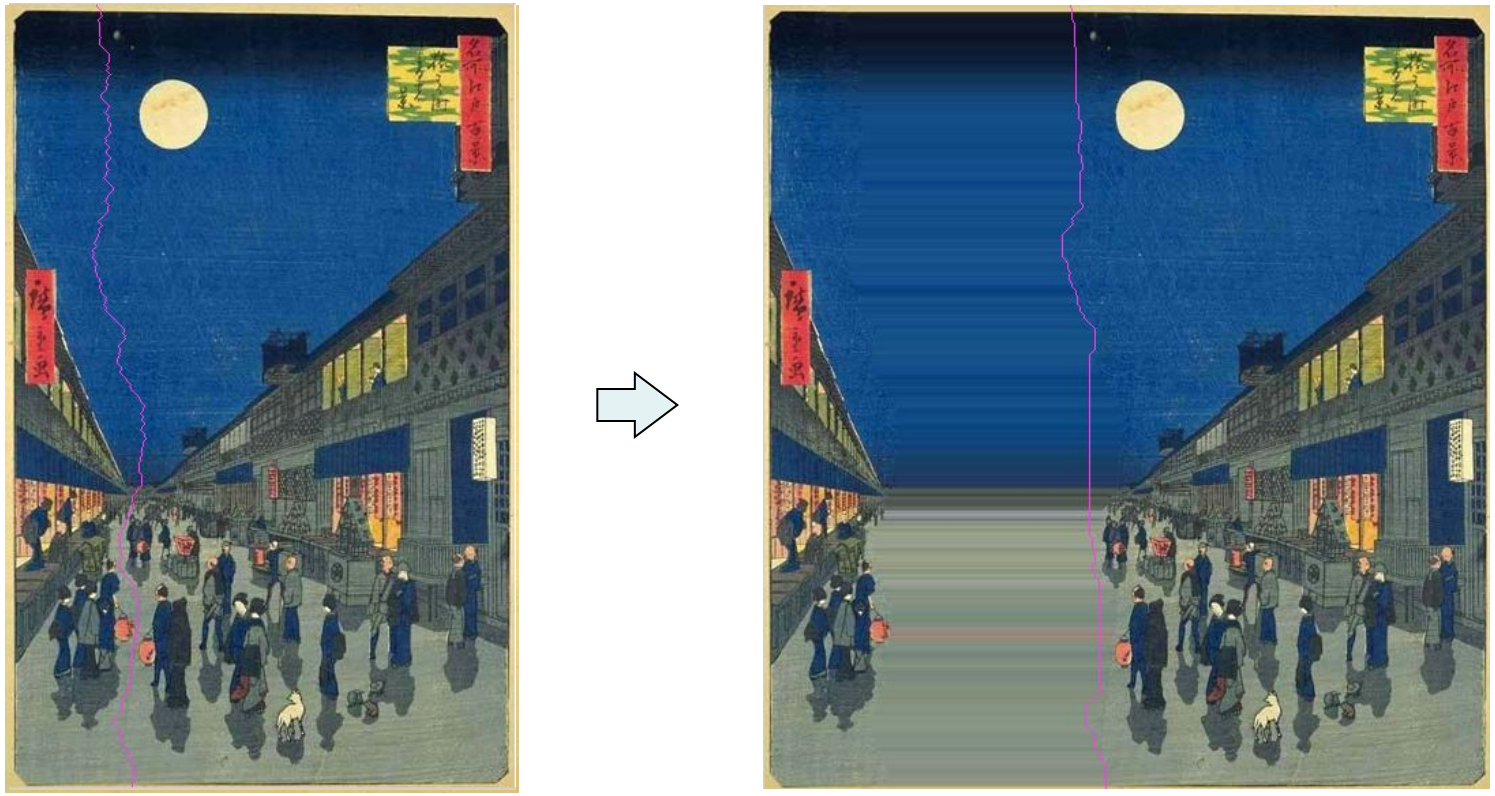
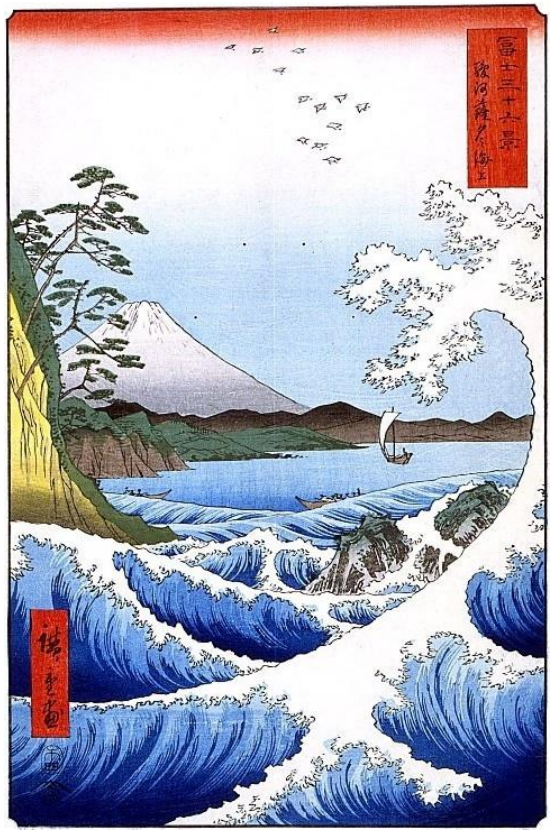


Image Expansion – Repeat the K lowest energy seams

Scaling

Can you tell if this image has been enlarged or reduced?



Combined Insert and Remove

Insert & remove seams

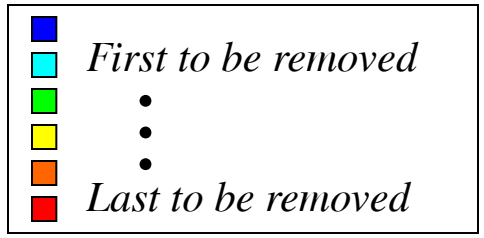
Scaling

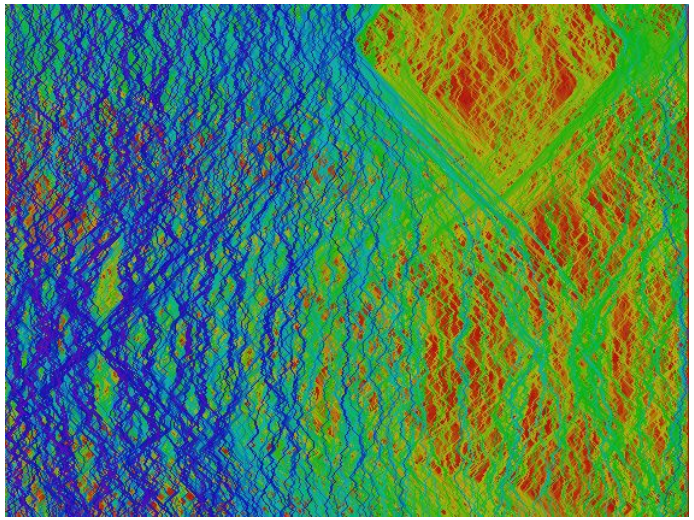
Multi-Size Images

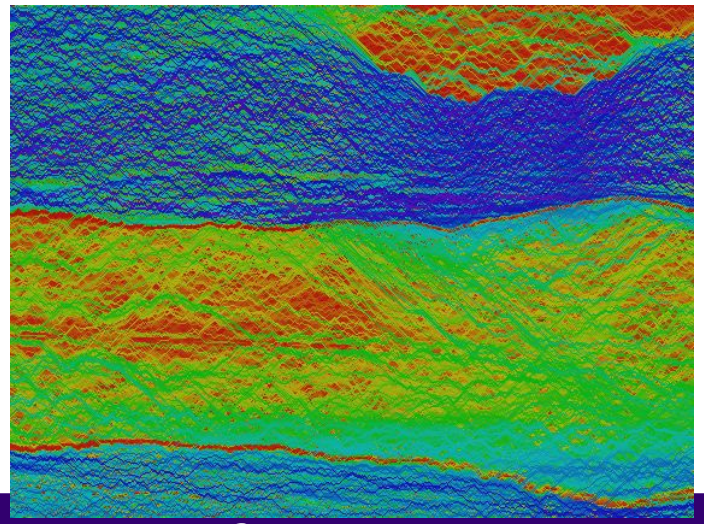
- We can create a new *representation* of an image that will allow adapting it to different sizes!

1. Precompute all seams once
2. Realtime resizing / transmit with content

Multi-Size Images

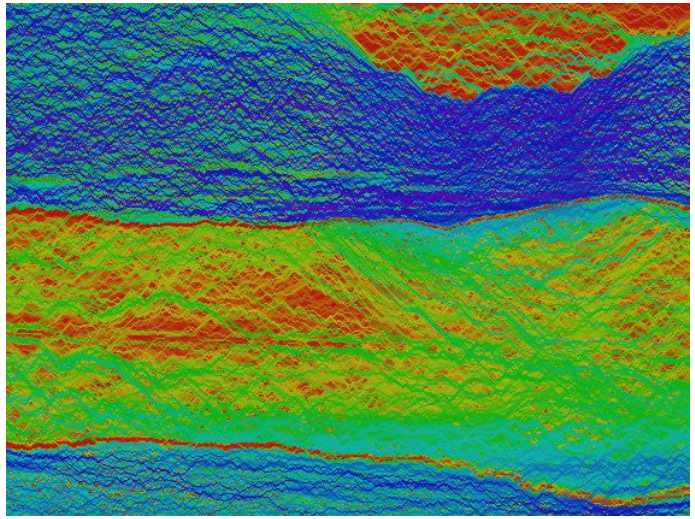




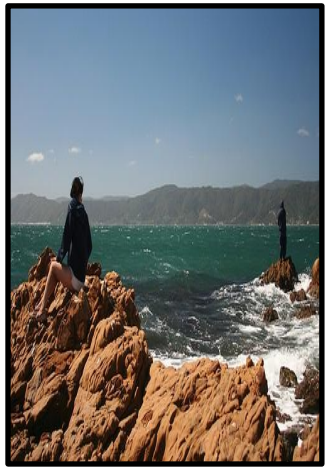


Multi-Size Image Representation

+



Multi-Size Image Representation



Content Enhancement

Q. How to not touch objects when using seam carving?

Replace E(i, j) with user defined energies

Recall our seam equation

$$\mathbf{M}(i,j) = E(i,j) + \min\left(\mathbf{M}(i-1,j-1), \mathbf{M}(i-1,j), \mathbf{M}(i-1,j+1)\right)$$

Set E(i, j) to be infinity is a user wants to keep this pixel
Set E(i, j) to be negative number if a user wants to get rid of it.

Object Removal

Object Removal

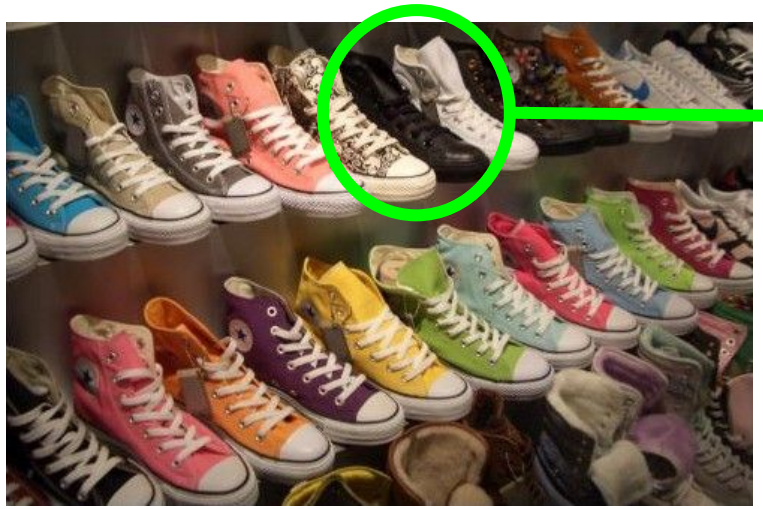
Input | Retargeted | Pigeon Removed | Girl Removed

Let’s delete a shoe from this image

Find the missing Shoe **in** the left image!

Solution: black shoe is gone

Let’s delete another shoe. Find the new missing shoe!



Solution

We can stack these edits. Find the missing show from the bottom right image by deleteing from the upper right

Solution

Use face detector to set energies of faces high

Energy with gradients

Energy with face detectors

With User Constraints

Scaling Retargeting Retargeting

Today’s agenda

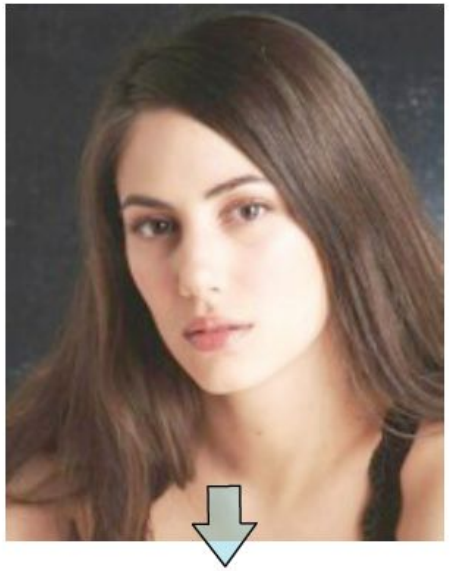
- Image retargeting
- Seam carving
- More Applications
- Forward algorithm

Seam carving creates artifacts breaks edges

Limitations

Content

Structure



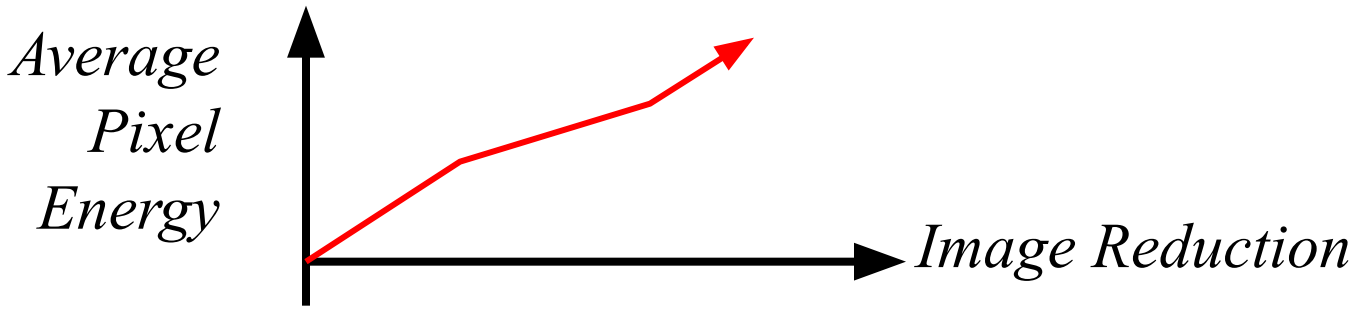
Questions

Q: What happens to the avg pixel energy in the image during seam carving?

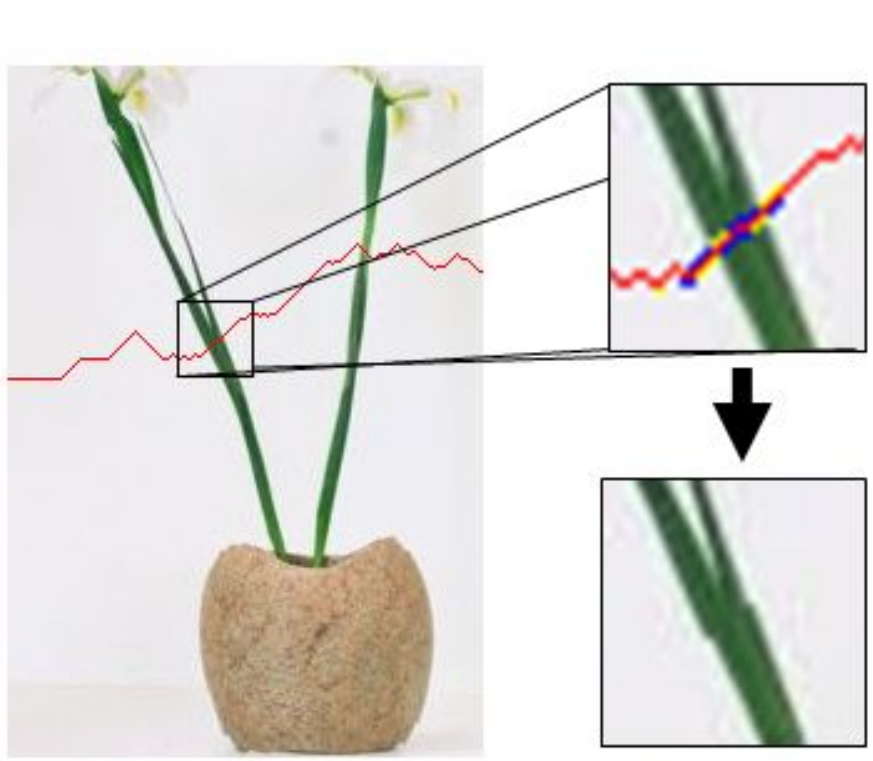
Preserved Energy

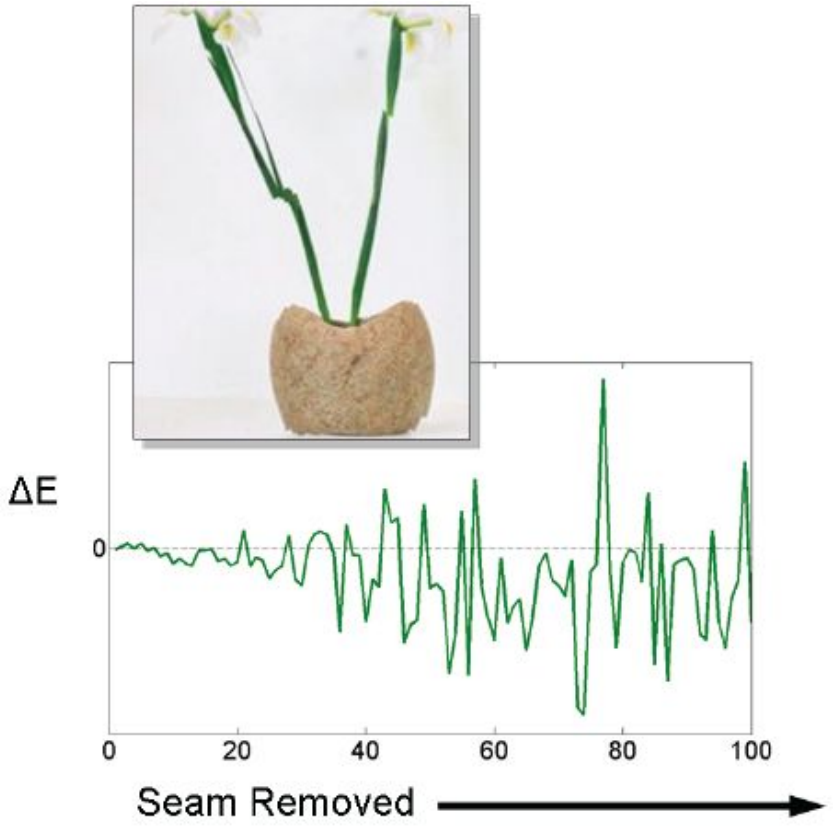
If we measure the average energy of pixels in the image after applying a resizing operator…

…the average should increase!



Inserted Energy

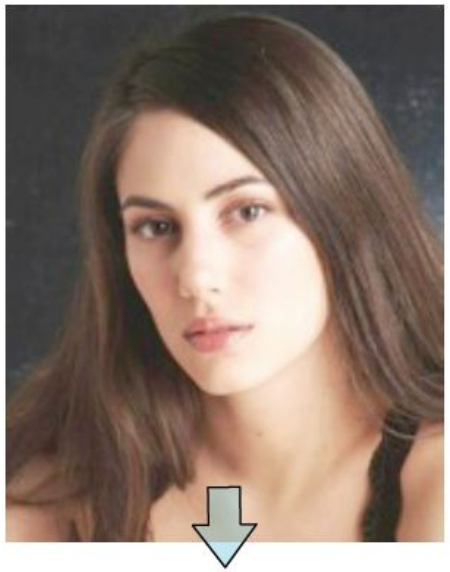




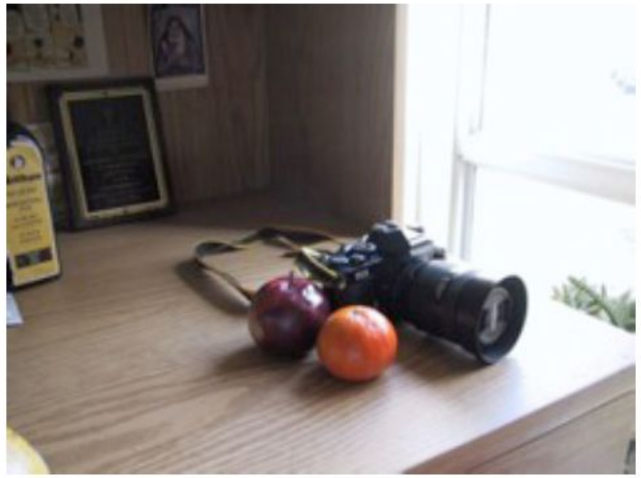
Limitations

Content

Structure



Seam carving creates artifacts breaks edges

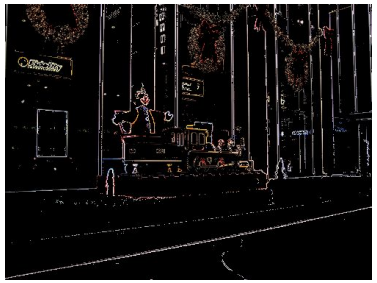


Preserved Energy

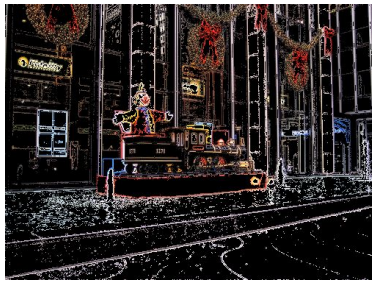
E(i, j)

Energy

Energy increases after every seam removal



15%



30%

40%

75%

Preserved Energy

Average Pixel Energy

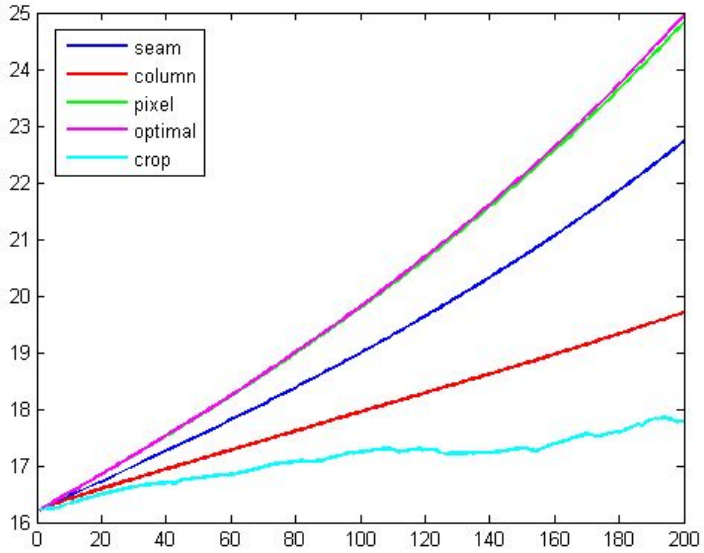


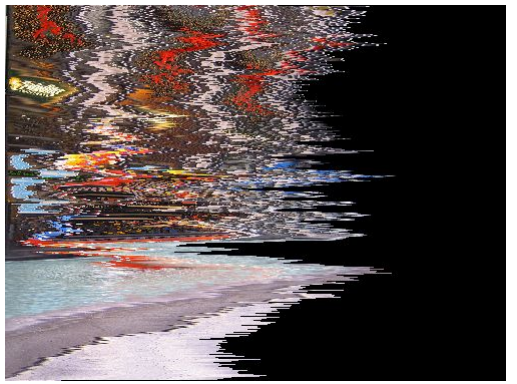
Image Reduction →

crop

column

seam

pixel

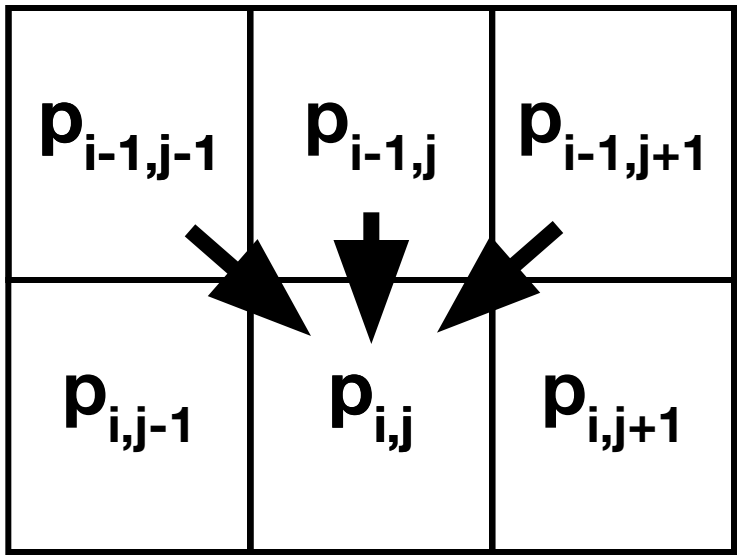


optimal

Minimize Inserted Energy

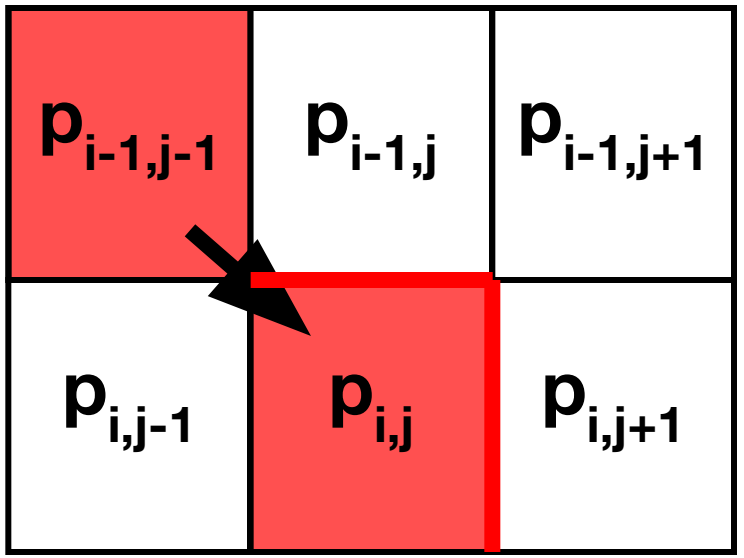
- Instead of removing the seam of least energy, remove the seam that *inserts the least energy* to the image (forward looking) !

Tracking Inserted Energy



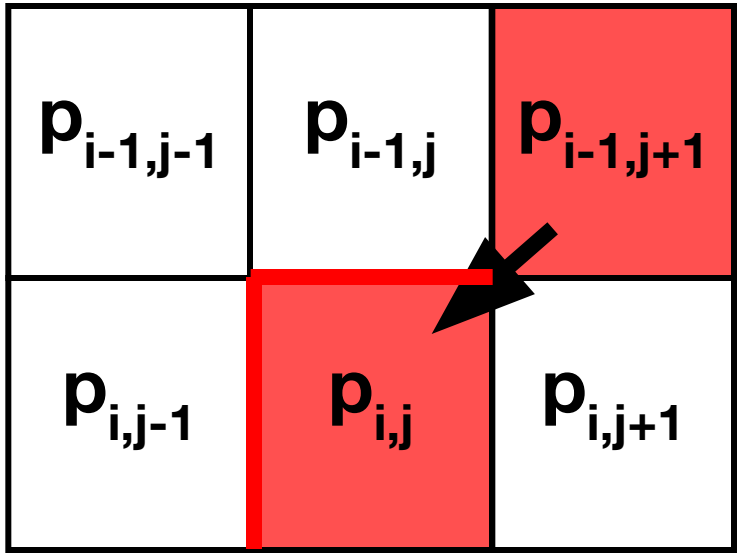
- Three possibilities when removing pixel $P_{i,j}$

Pixel $P_{i,j}$: Left Seam



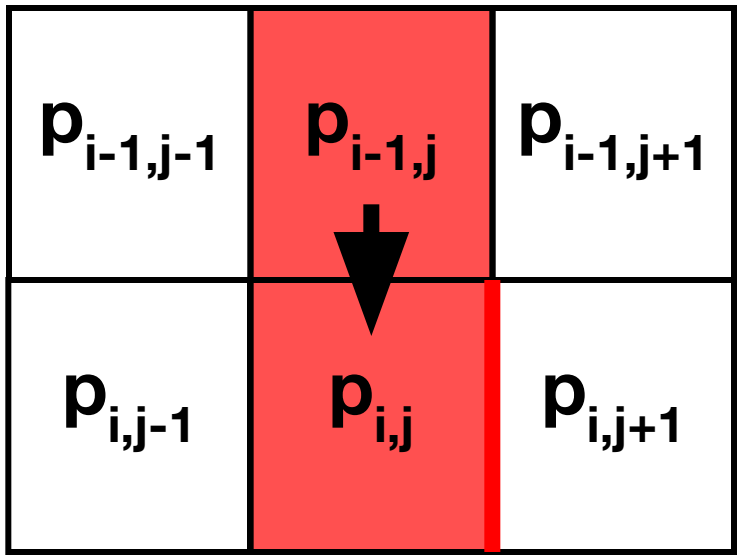
$$C_L(i,j) = |I(i,j+1) - I(i,j-1)| + |I(i-1,j) - I(i,j-1)|$$

Pixel $P_{i,j}$: Right Seam



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

Pixel $P_{i,j}$: Vertical Seam



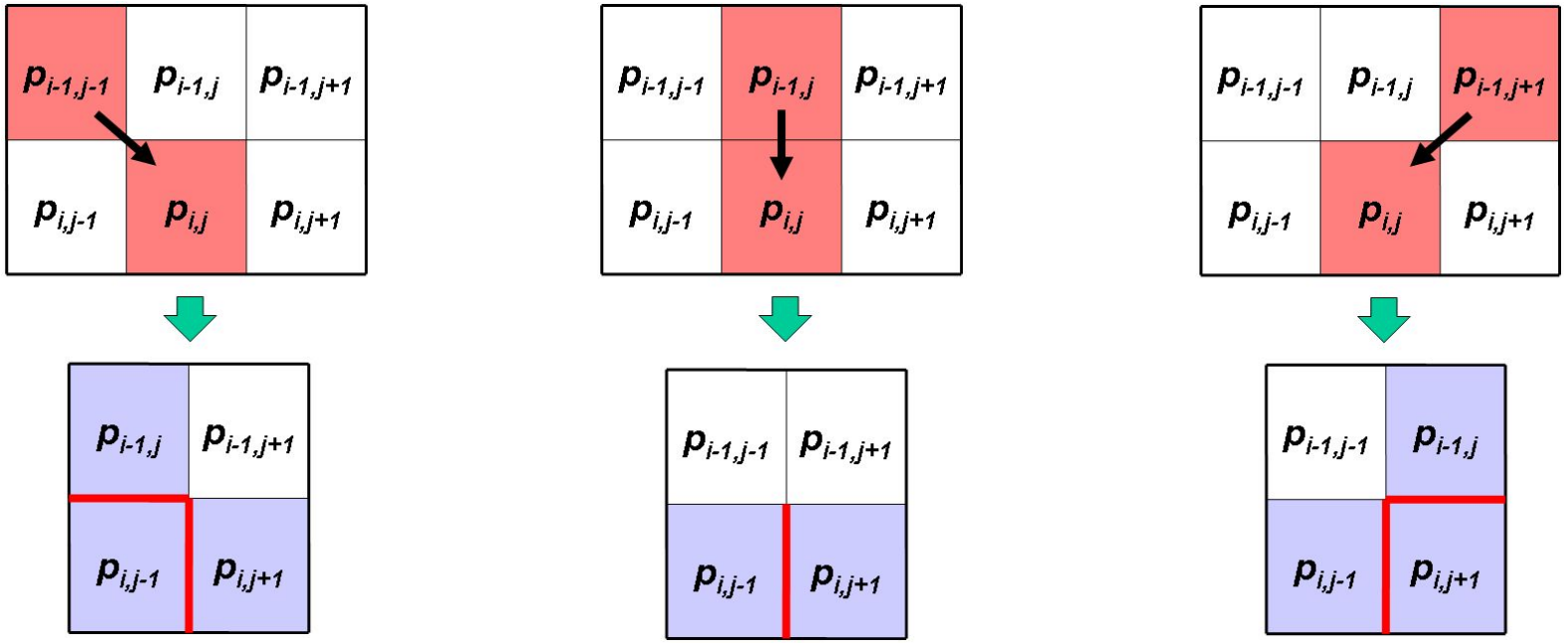
$$C_V(i,j) = |I(i,j+1) - I(i,j-1)|$$

Old Backward Cost Matrix

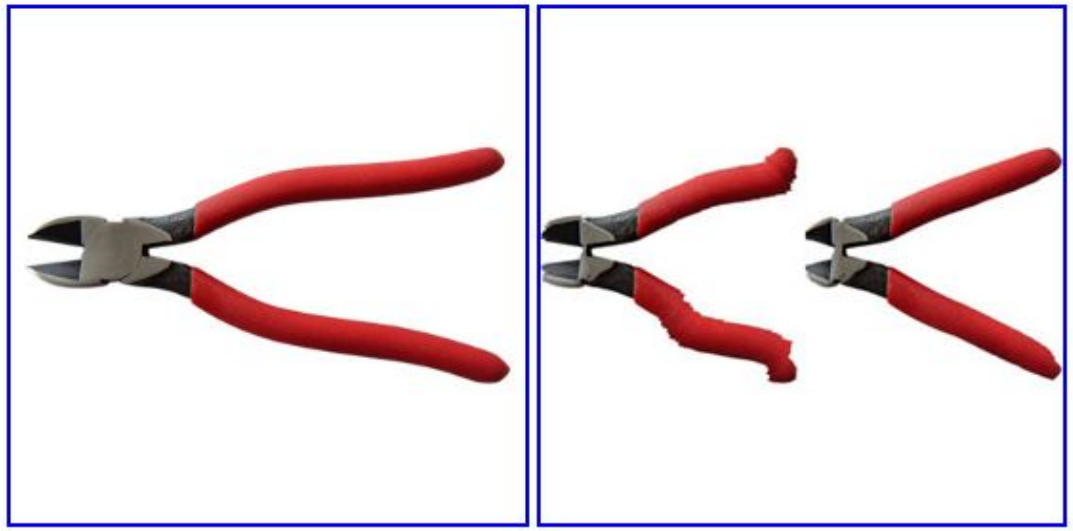
$$M(i,j) = E \quad j) + \min \begin{cases} M(i-1, j-1) \\ M(i-1, j) \\ M(i-1, j+1) \end{cases}$$

New Forward Looking Cost Matrix

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



Results

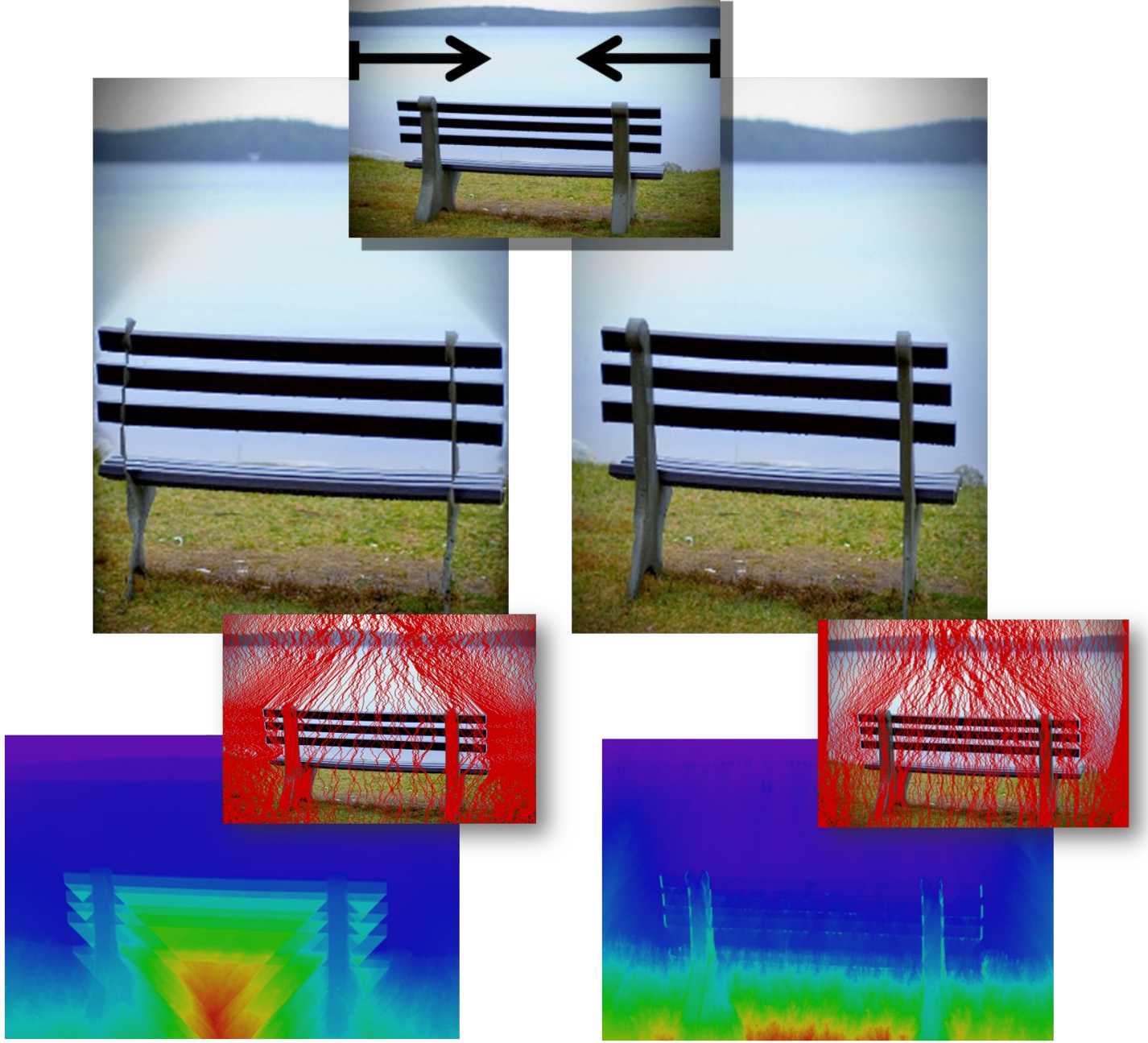


Input Backward Forward

Input

Backward Forward

Results

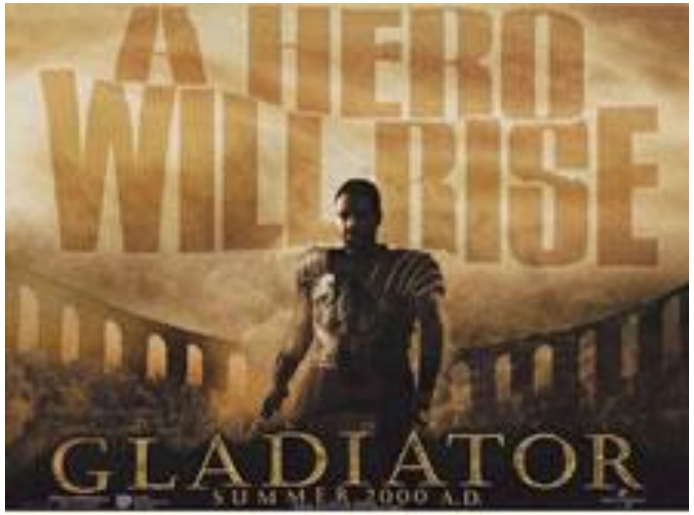


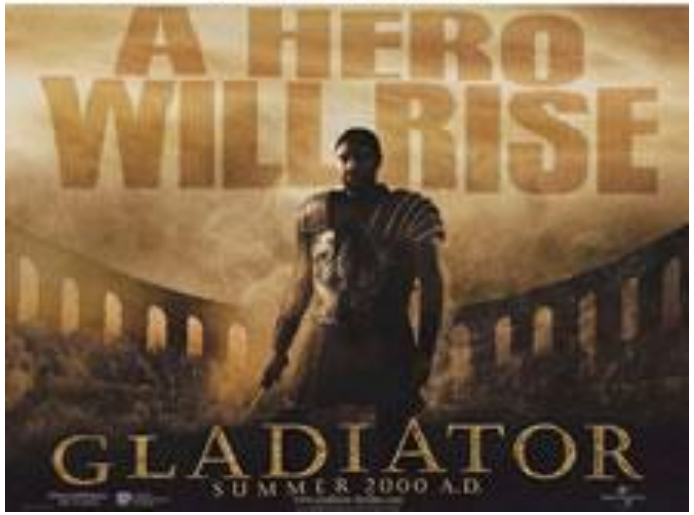
Backward vs. Forward

Backward

Forward

Results


A HERO
WILL RISE
GLADIATOR
SUMMER 2000 A.D.


A HERO
WILL RISE
GLADIATOR
SUMMER 2000 A.D.

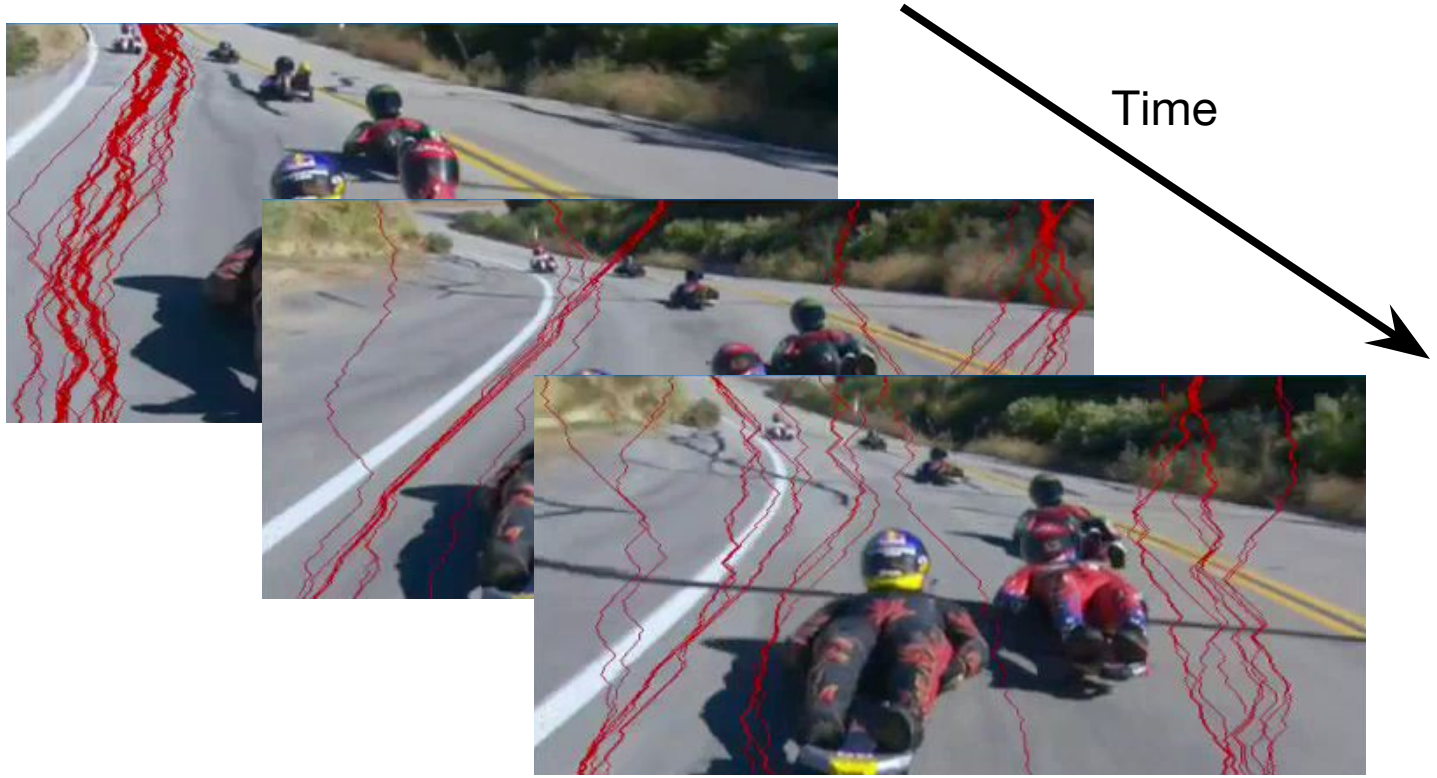
From Images to Videos

In general, video processing is a much (much!) harder problem

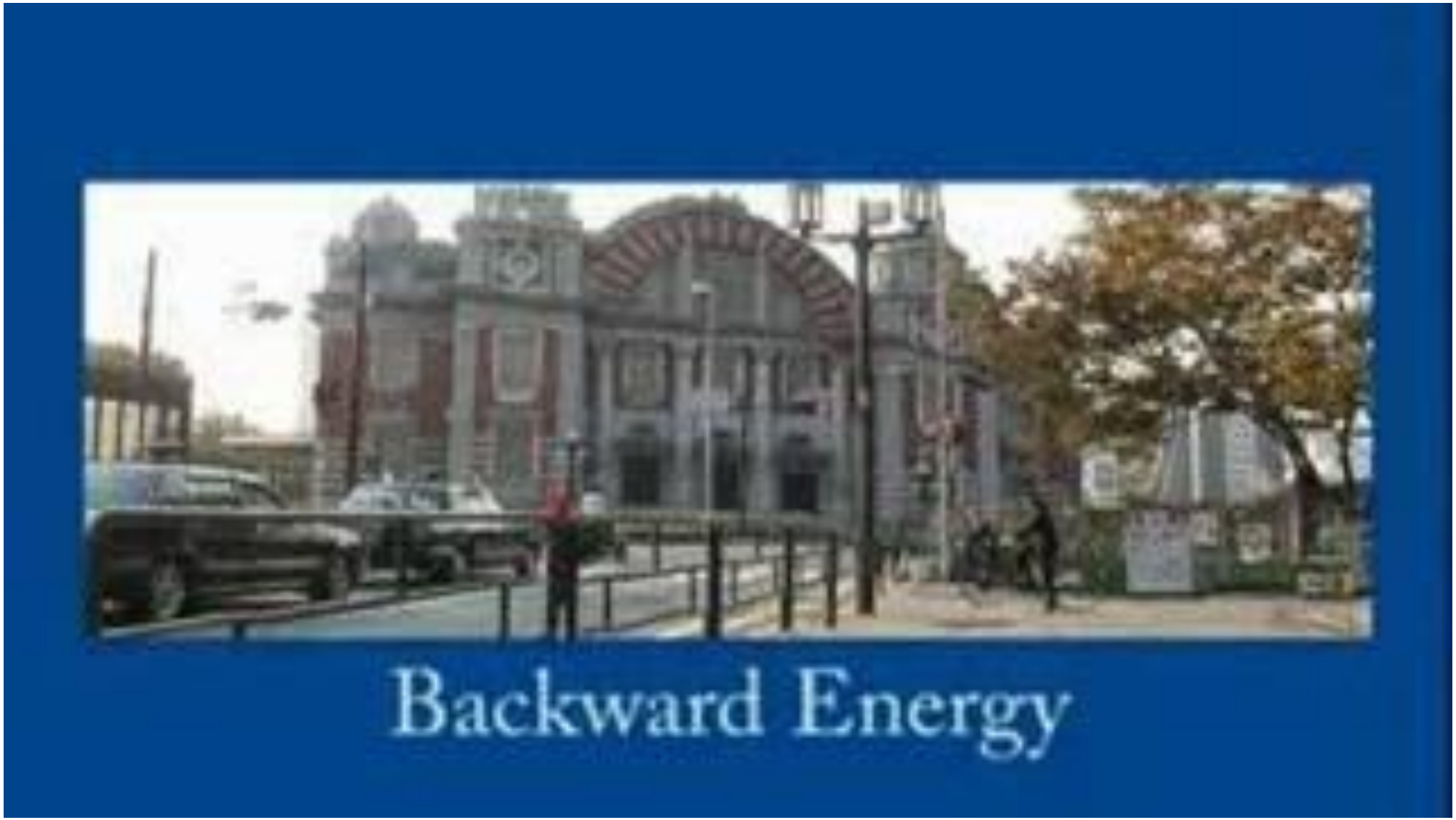
1. Cardinality
   - Suppose 1min of video x 30 fps = 1800 frames
   - Say your algorithm processes an image in 1 minute
     - 1 video would take **30 hours !!**
2. Dimensionality/algorithmic
   - Temporal coherency: human visual system is highly sensitive to motion!

Seam-Carving Video?

- Naive… frame by frame independently



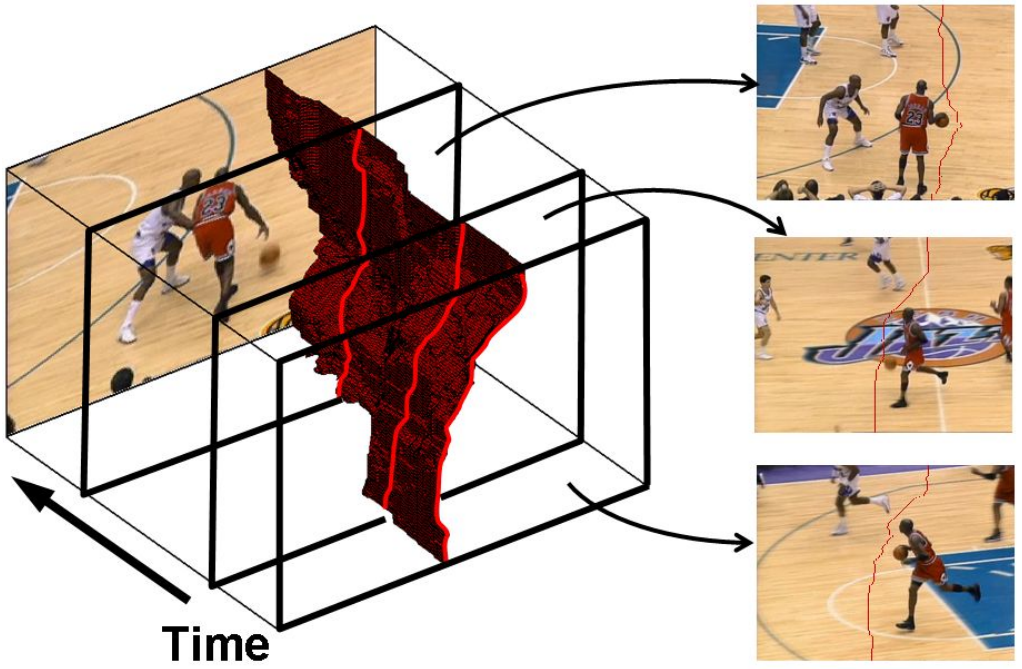
Frame-by-frame Seam-Carving



Let’s check out this video

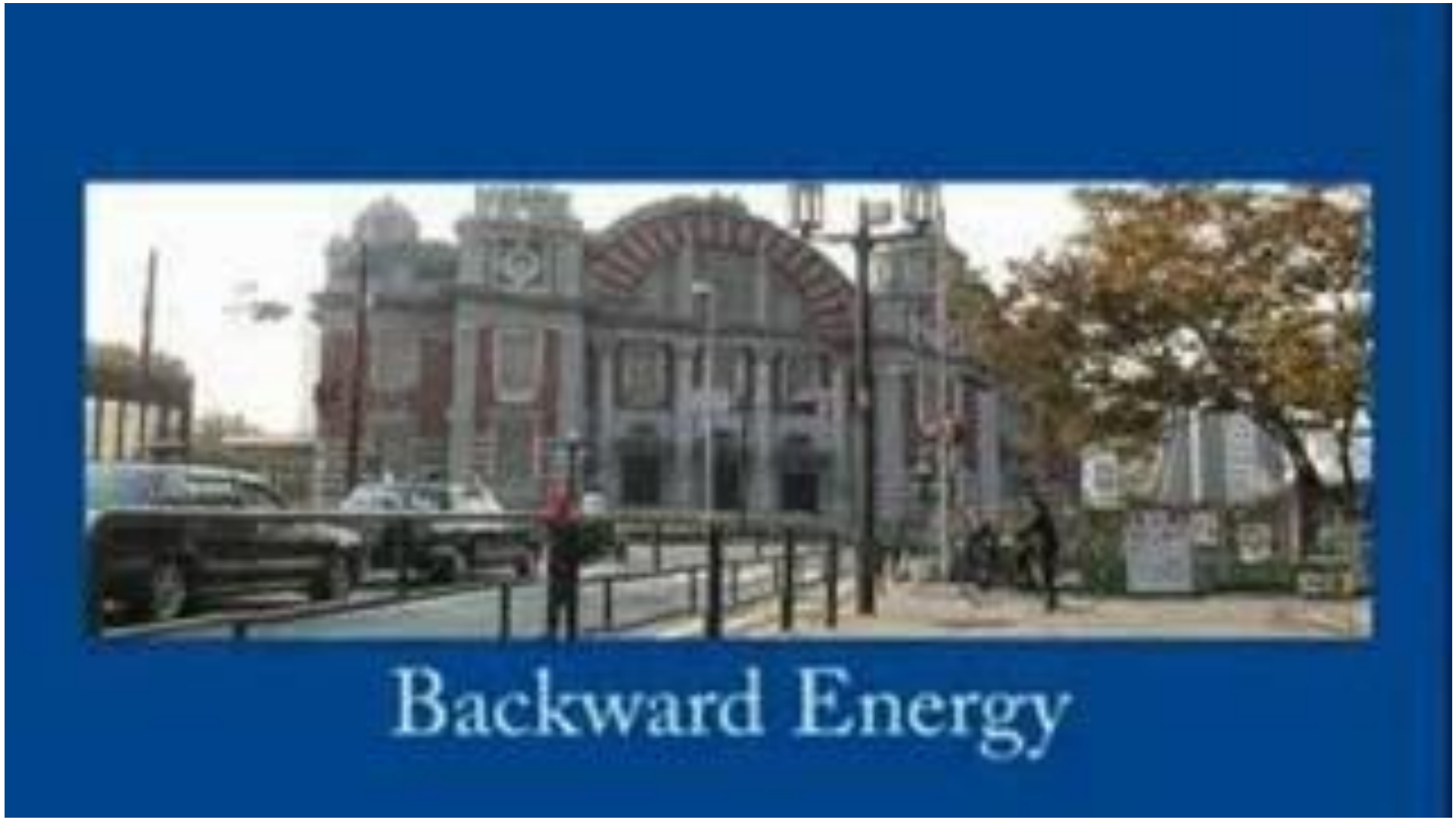
From 2D to 3D

1D paths in images

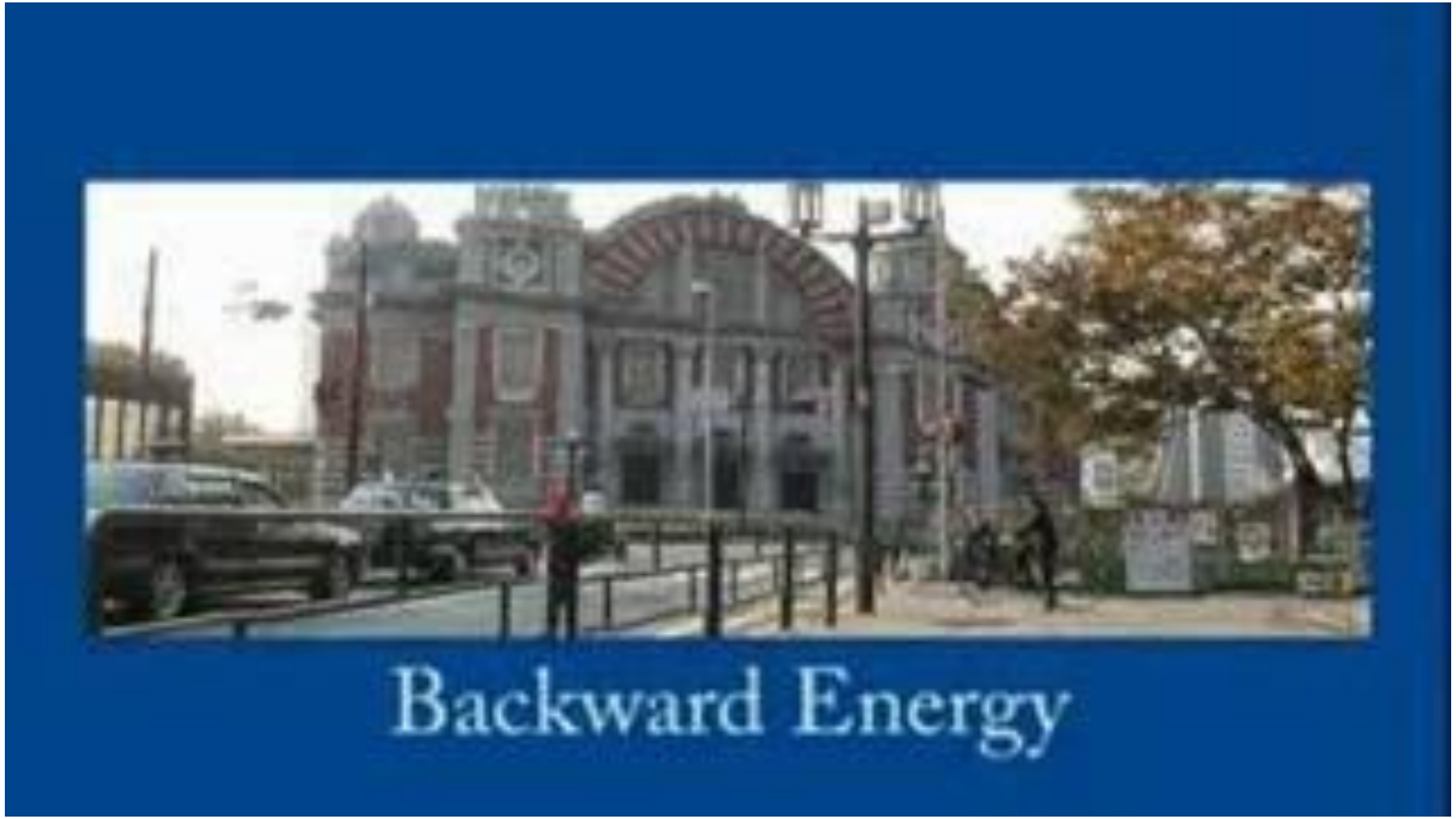


2D manifolds in video cubes

Example video retargeting



Object detection + seam carving



Today’s agenda

- Image retargeting
- Seam carving
- Dynamic programming
- Applications
- Forward algorithm

Next lecture

Motion & Camera

References

- Seam Carving for Content-Aware Image Resizing – Avidan and Shamir 2007
- Content-driven Video Retargeting – Wolf et al. 2007
- Improved Seam Carving for Video Retargeting – Rubinstein et al. 2008
- *Optimized Scale*-and-*Stretch* for Image Resizing – Wang et al. 2008
- Summarizing Visual Data Using Bidirectional Similarity – Simakov et al. 2008
- Multi-operator Media Retargeting – Rubinstein et al. 2009
- Shift-Map Image Editing – Pritch et al. 2009
- Energy-Based Image Deformation – Karni et al. 2009

- Seam carving in Photoshop CS4: http://help.adobe.com/en_US/Photoshop/11.0/WS6F81C45F-2AC0-4685-8FFD-DBA374BF21CD.html

A Local Operator