

# Display-aware Image Editing

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# Very large images are easy to create

- We are interested in 100+ Mpixels.
- large choice of panorama software
- taking plenty of pictures is easy
  - hand-held with a telephoto lens
  - dedicated devices, e.g. GigaPan
  - medical scanners



# Demo

A tour of Hong Kong in 300 megapixels

# Challenges

- A lot of pixels to process
  - 100 Mpixels and above
  - everything is slow
- Scene changes completely with location and zoom level
  - no single adjustment works everywhere

# Our approach

## Display-aware Image Editing

- Only 1 to 4 Mpixels visible at a given time
- Our silver bullet: **Take into account and process only the visible pixels.**



# Related work: Fast image filters

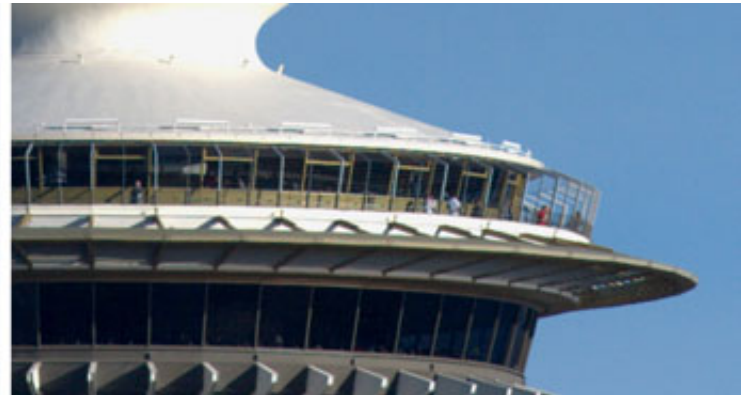
- Bilateral grid [Chen 07, Paris 09], edge-aware wavelets [Fattal 09], guided filtering [He 10], geodesic image editing [Criminisi 11]...
  - designed for a few tens of megapixels
  - slow down on 100 Mpixels and above
    - process every pixel

# Related work: Bounding boxes

- Process only the affected regions [Shantzis 94]
  - global algorithms need touch every pixel
  - ignore zoom level
    - large regions can appear small when fully zoomed out

# Related work: Adaptive tone mapping

- Viewer for HDR panoramas [Kopf 07]
  - adapt viewing parameters to visible image portion
  - does not allow user control



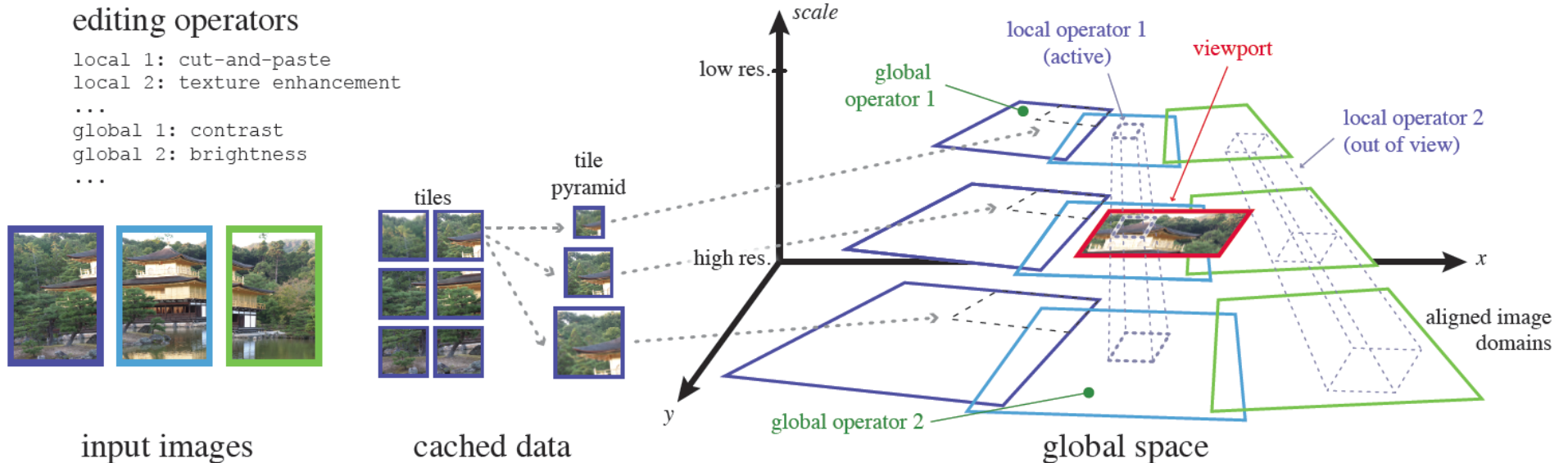
# Talk outline

1. Efficient viewer
2. Display-aware algorithms
3. Display-aware viewing parameters

# Our viewer

- Design choice: displayed image computed on-the-fly each time
  - dependencies straightforward to handle
- Efficient implementation
  - tile-based cache hierarchy
  - GPU-accelerated filters
  - bounding boxes

# Viewer design



We load only the visible data.

We execute only the operators  
that affect the current view.

Extended in [Jeong et al. Vis'10].

# Talk outline

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# Formal discussion

- A filter  $f$  is display-aware if the visible portion of its output can be computed only from the visible portion of the input  $I$ .
- Sufficient to process Laplacian pyramids locally
  - output coefficients depend only on nearby coefficients
  - proof in the paper
  - also a limitation, i.e. constrains algorithm design

# Formal discussion

- A filter  $f$  is display-aware if the visible portion of its output can be computed only from the visible portion of the input  $I$ .
  - $s$  is the operator that display an image on the screen
  - $f$  is display-aware if  $s(f(I)) = f(s(I))$

# Formal discussion

- Displaying an image  $I$  on the screen,  $s(I)$ :
  1. low-pass filter followed by downsampling
    - equivalent to crop in Fourier domain
  2. crop in image space
- To commute with  $s$ , a filter  $f$  should not use data than may be cropped by  $s$ 
  - use only local data in space and frequency
  - **Local processing on Laplacian pyramids**

# Example application: Seamless image compositing



- Standard approach: gradient-domain + Poisson eq.
  - global optimization: does not scale up

# Our approach

- Inspired from [Farbman 09]
  - core operation is computing a smooth interpolant of offset values defined at boundaries
- We use pyramid-based interpolant [Burt 88]
  - classical push-pull operation

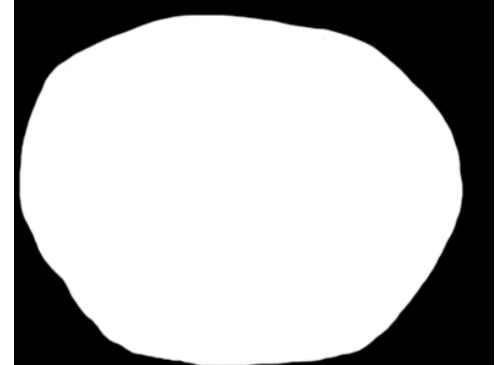
# Input data



foreground

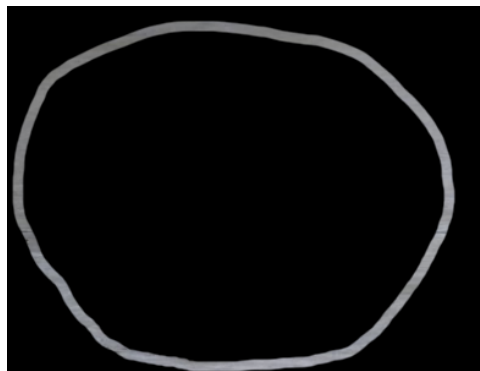


background



mask

# 1- Compute offsets at boundary



foreground

-



background

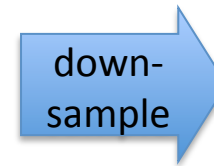
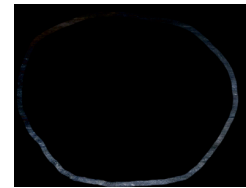
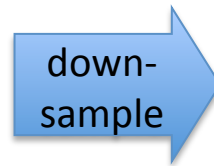
=



offsets

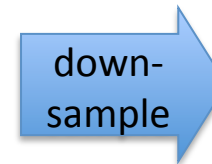
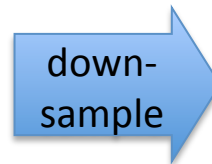
## 2- Build a Laplacian pyramid with the offsets

Gaussian pyramid:

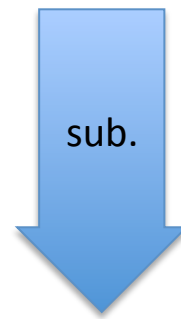
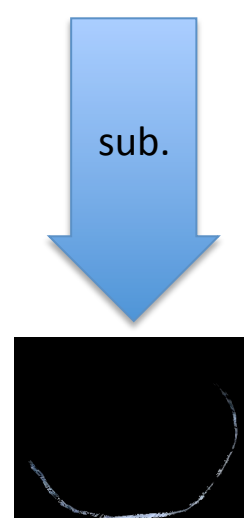


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Gaussian pyramid:

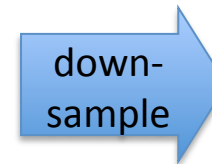
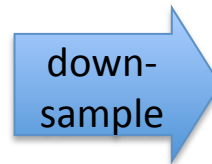


Laplacian pyramid:

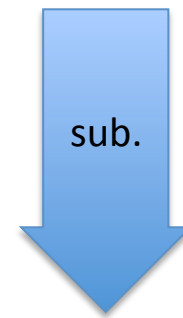
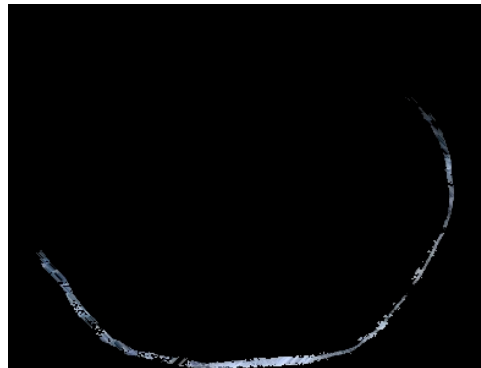


## 2- Build a Laplacian pyramid with the offsets

Gaussian pyramid:

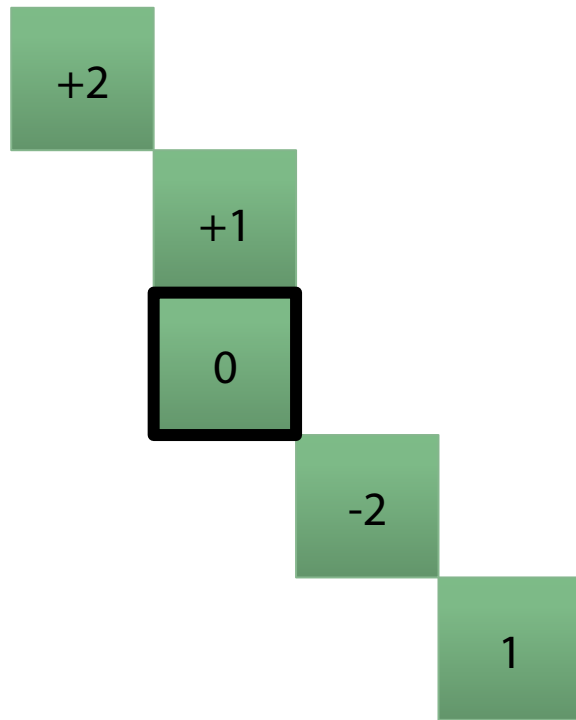


Laplacian pyramid:

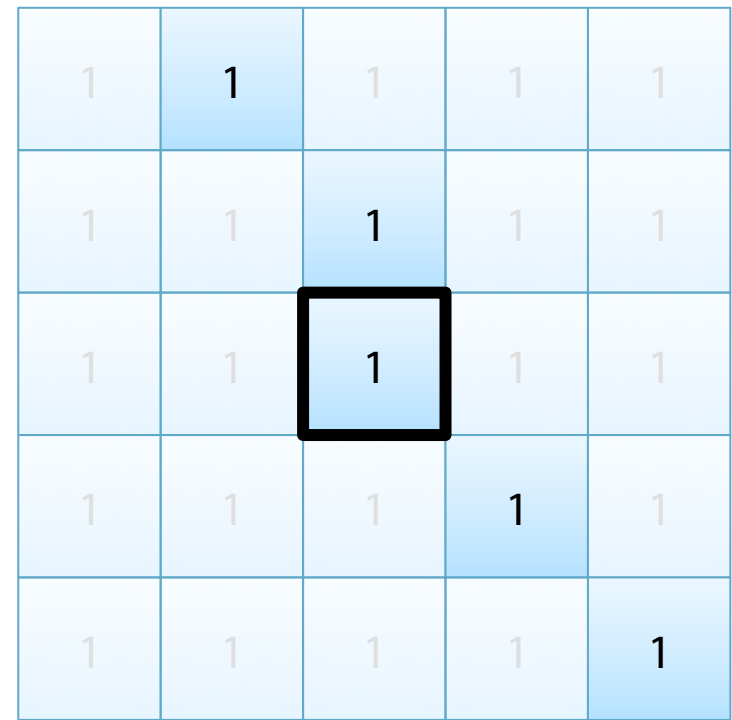


## 2- Build a Laplacian pyramid with the offsets

To downsample, keep only the values where there are data.



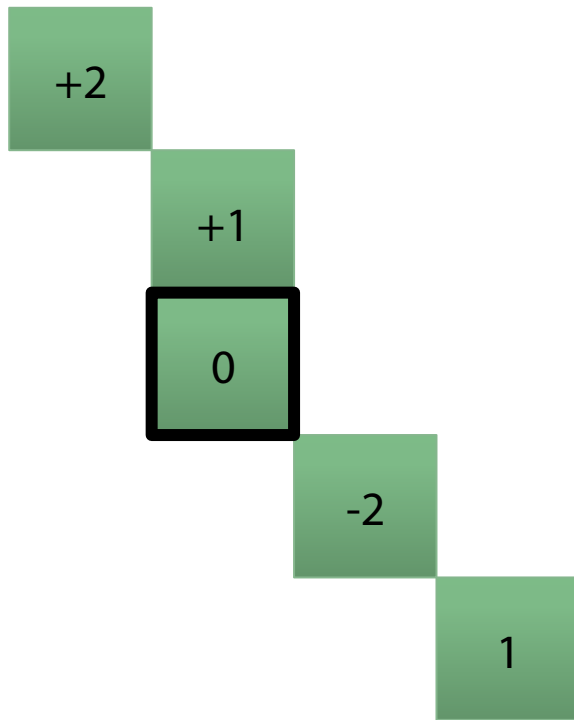
boundary offset values



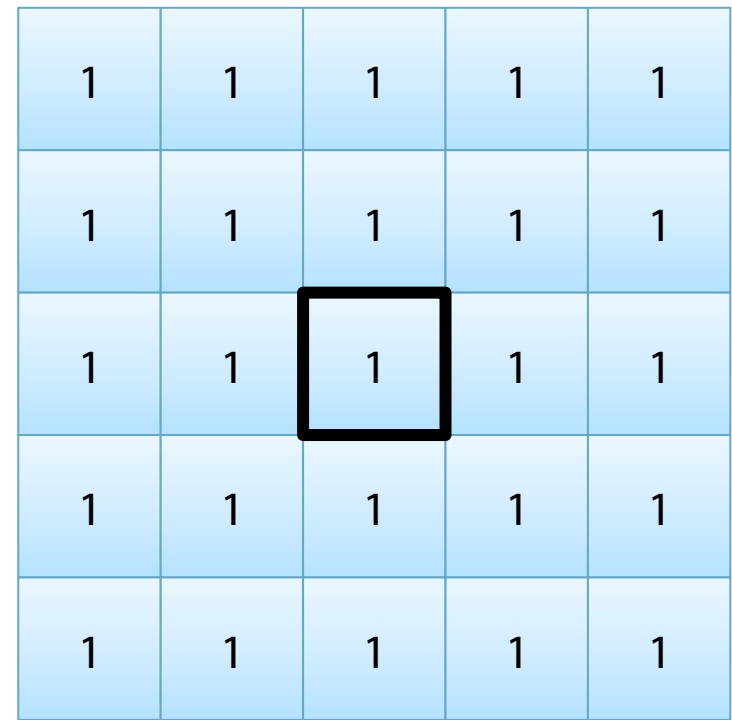
blurring kernel  
(box for illustration)

## 2- Build a Laplacian pyramid with the offsets

To upsample, keep all the values.



boundary offset values



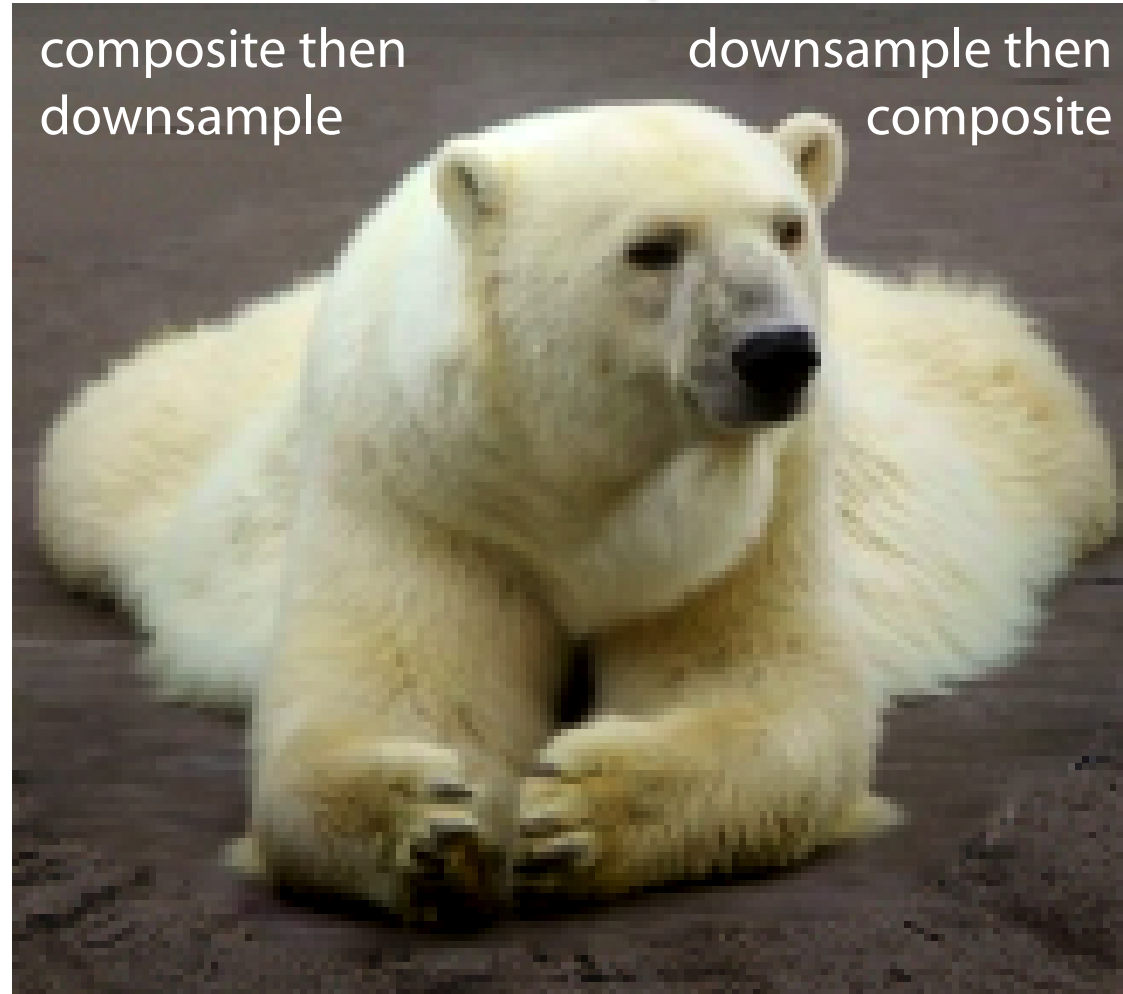
blurring kernel  
(box for illustration)

### 3- Collapse the pyramid



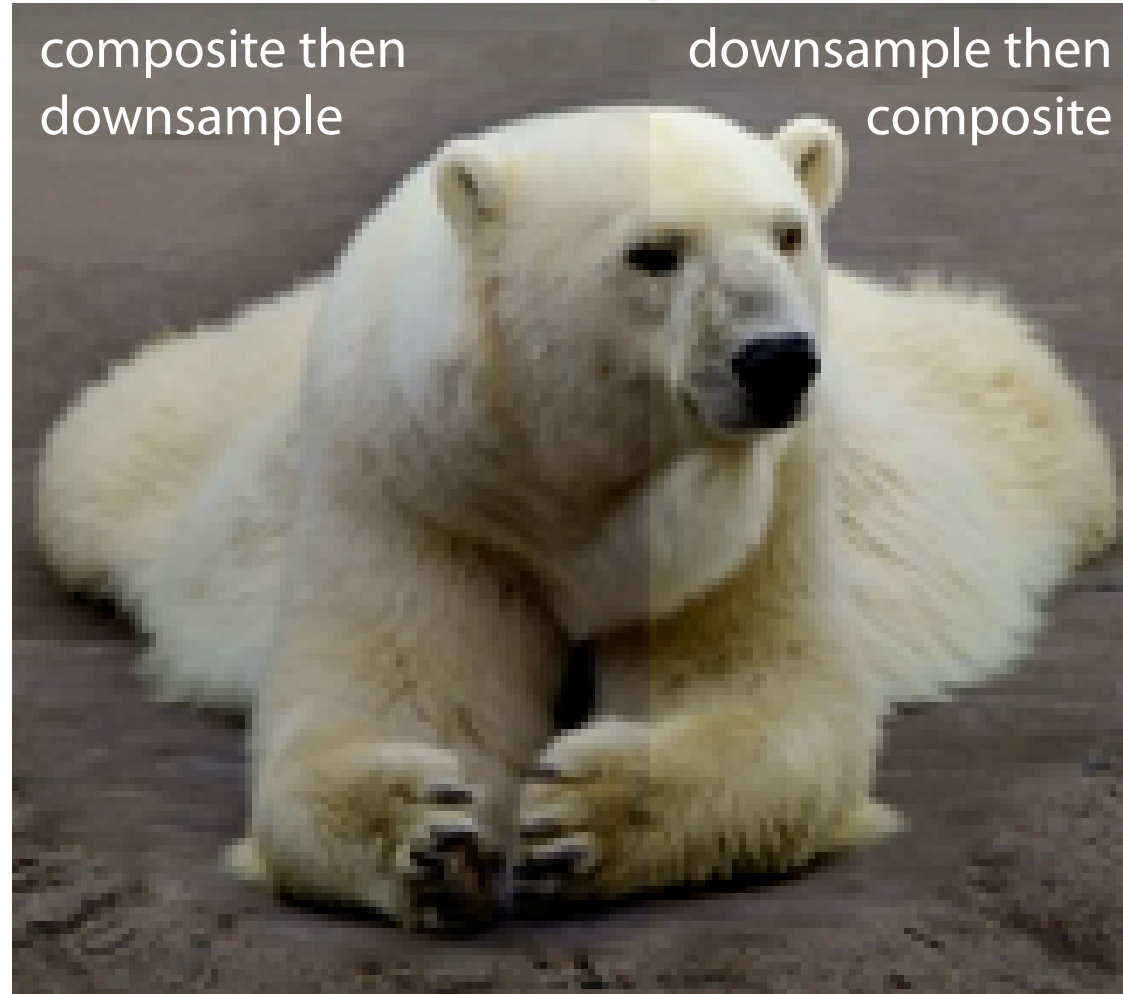
our result

# Is it really display-aware?

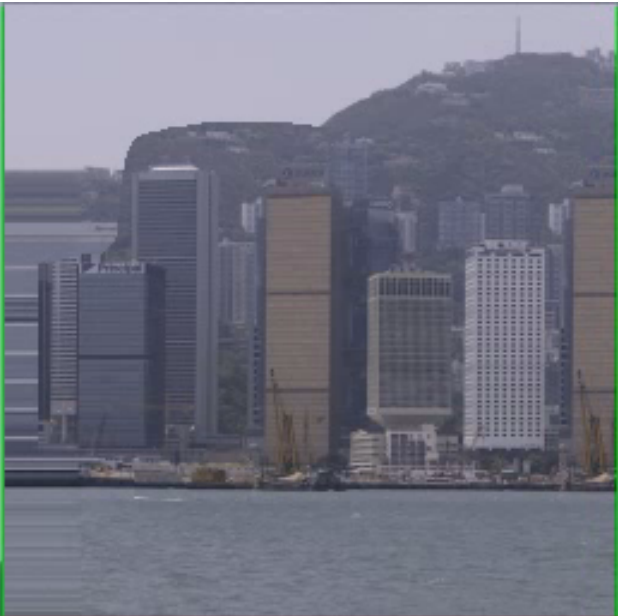


our result

# Is it really display-aware?



Photoshop result (gradient domain)



# Other Applications

- Image alignment  
(see the paper)



- Edge-aware image editing  
(to appear at Siggraph)



# Talk outline

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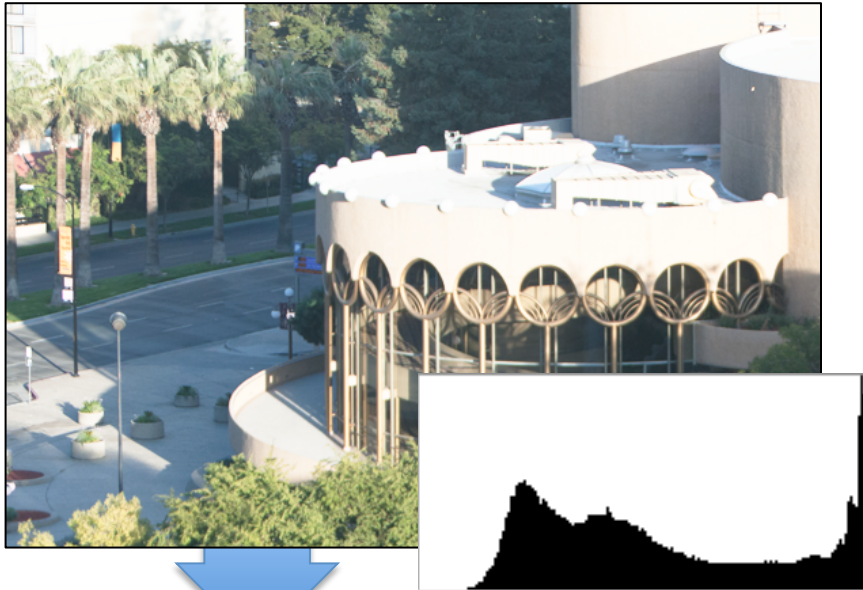
# Objective

- Adapting the viewing parameters (brightness, contrast...) to the current view
- Automatic by default, user can override

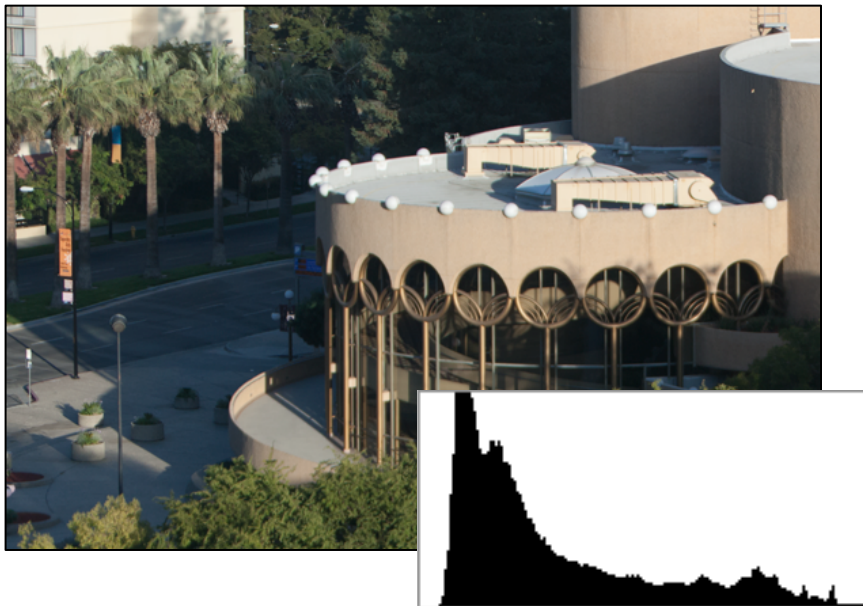
# User study (Mechanical Turks)

- Users adjust photos.
  - instruction: neutral pleasing rendition (postcard)
- We compare before and after.
  - **Image histograms are more similar after adjustment than before.**
  - Confirms the results of [Bae 06]

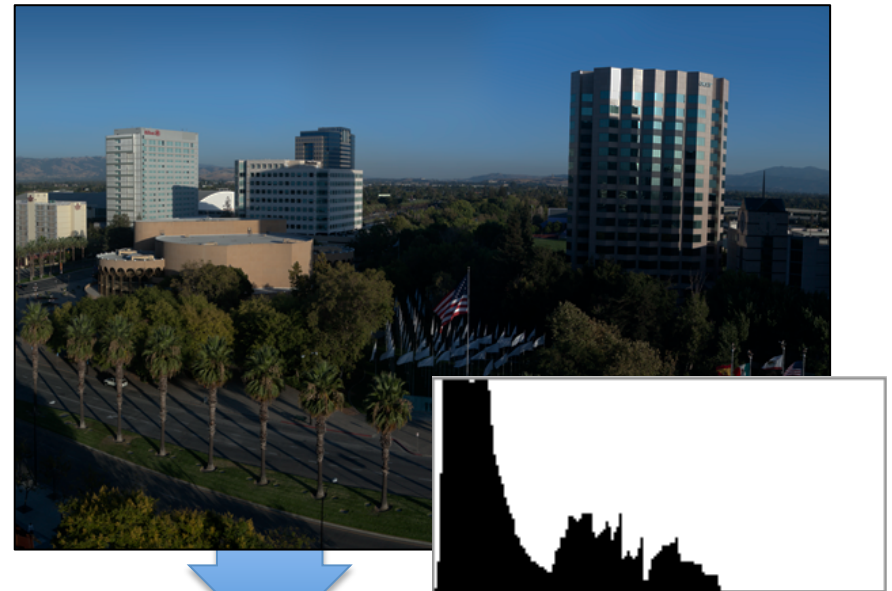
before adjustment



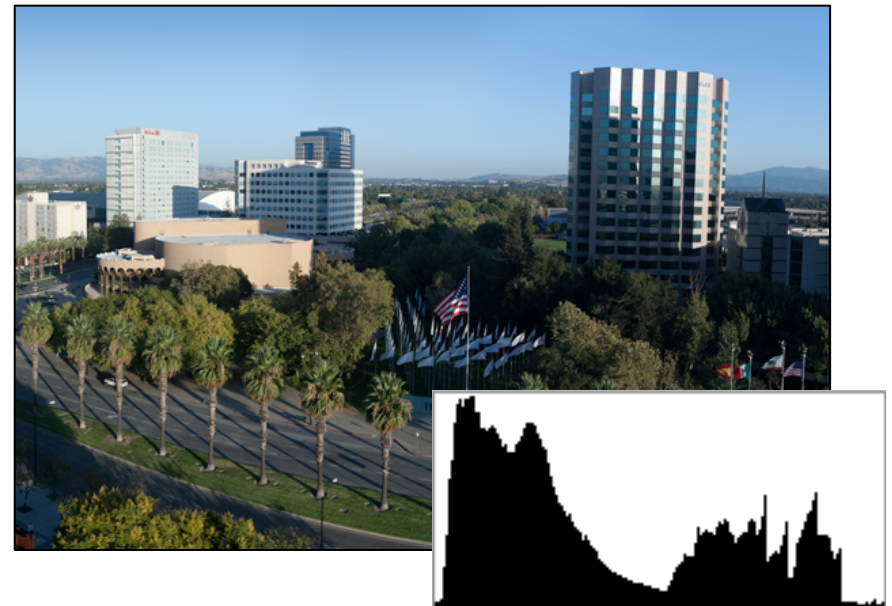
after



before adjustment



after



# Our strategy

Keep histogram as constant as possible.

# Algorithm overview

1. User specifies parameters at a few locations and scales.
2. For unspecified views, find  $k$  nearest neighbors with edits.
  - distance = EMD on unedited image histogram
3. Compute average edited histogram of the  $k$  neighbors
4. Estimate viewing parameters to best match that histogram



Adaptive Viewing

# Conclusion

- We enable the editing of very large images (1+ Gpixel).
- Laplacian pyramids are key.
- Histogram distance characterizes adjustments.
- Well-suited for mobile devices with small screens and limited computational power.



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We thank Min H. Kim for his help producing the supplementary video.

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commitment to your  
privacy, we  
have removed names  
listed for consultation  
from this notice.



HAPPY BIRTHDAY!



# Conclusion

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