

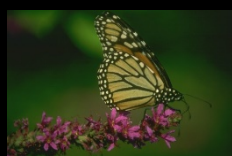
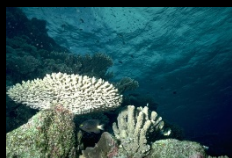
Rolling Guidance Filter

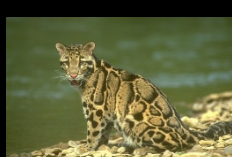
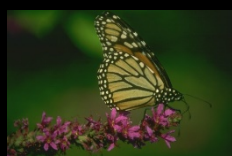
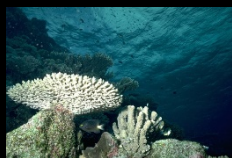
Qi Zhang¹, Xiaoyong Shen¹, Li Xu², Jiaya Jia¹

¹The Chinese University of Hong Kong

²Image & Visual Computing Lab, Lenovo R&T







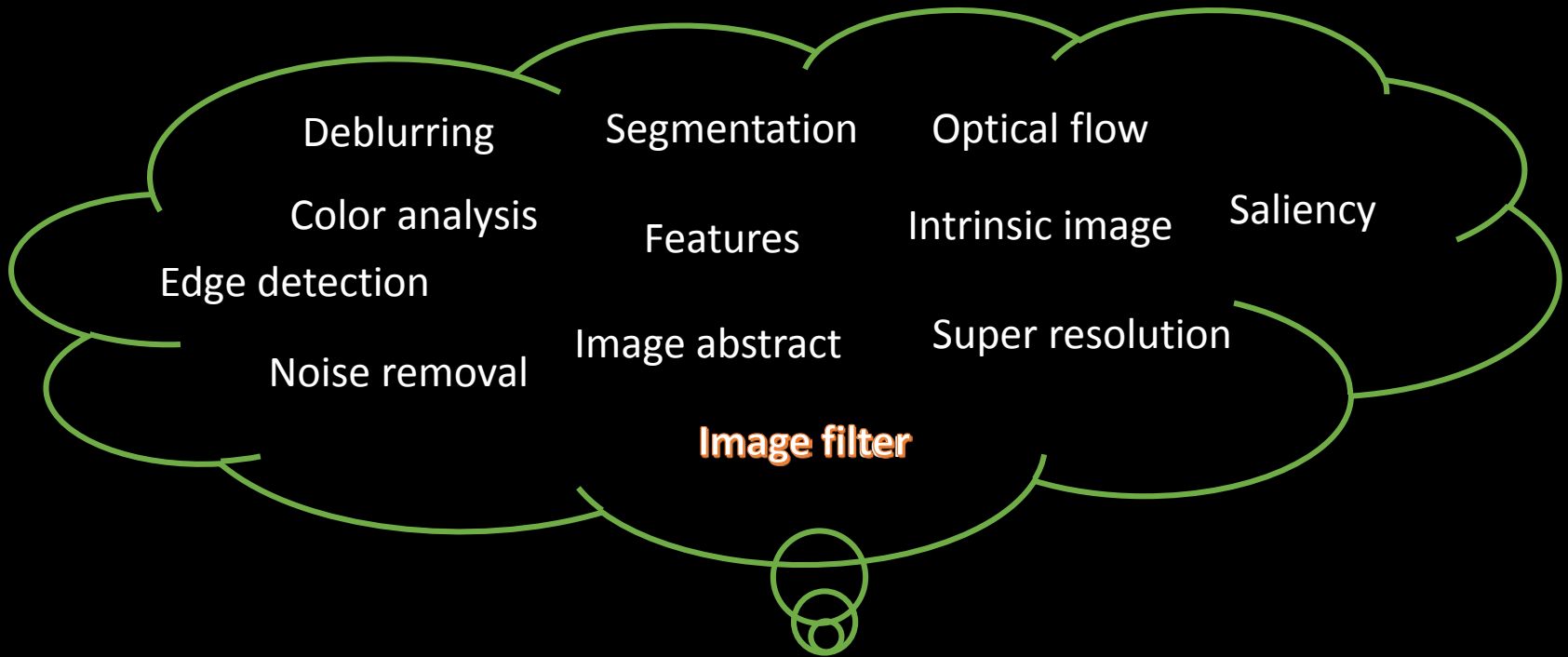
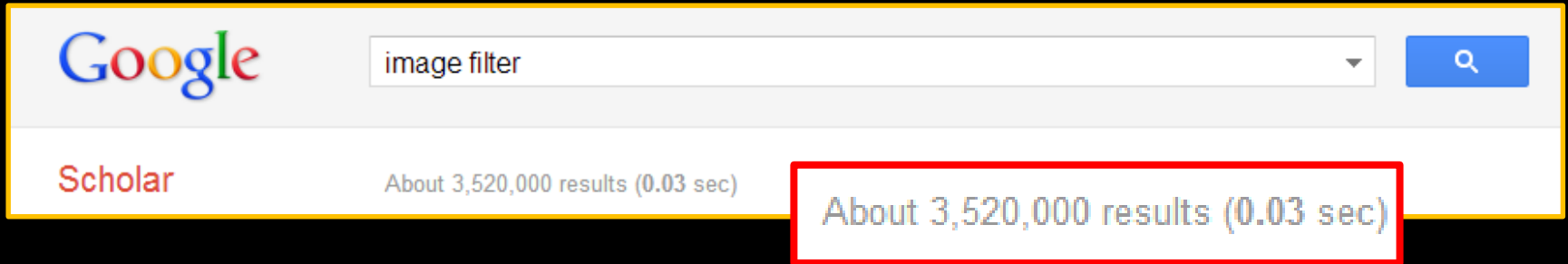


Image Filtering



Anisotropic Diffusion

Percentile Filter

Weighted Least Square Filter

Bilateral Filter

Domain Transform Filter

Local Extrema Filter

Local Mode Filter

Gaussian Filter

Guided Filter

Gabor Filter

Laplacian Filter

Min/Max Filter

Geodesic Filter

Sobel Filter

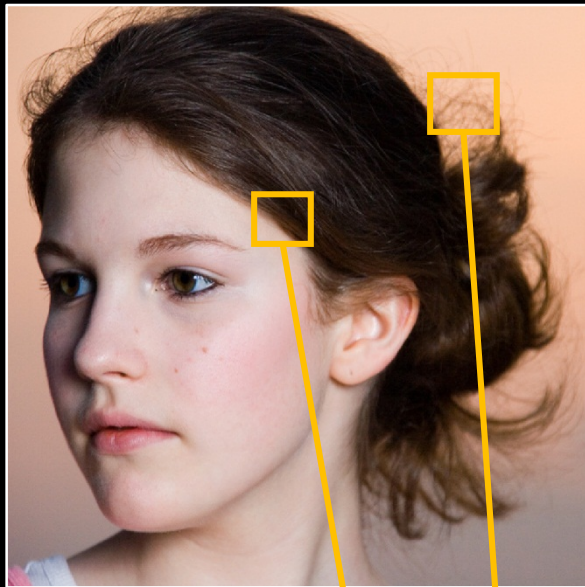
Median Filter

Dominant Filter

Weighted Median Filter

Recursive Bilateral Filter

An Important Steam: Edge Preserving



1998, Bilateral Filter
2008, WLS Filter
2010, Guided Filter
2011, Domain Transform



Weak Edge



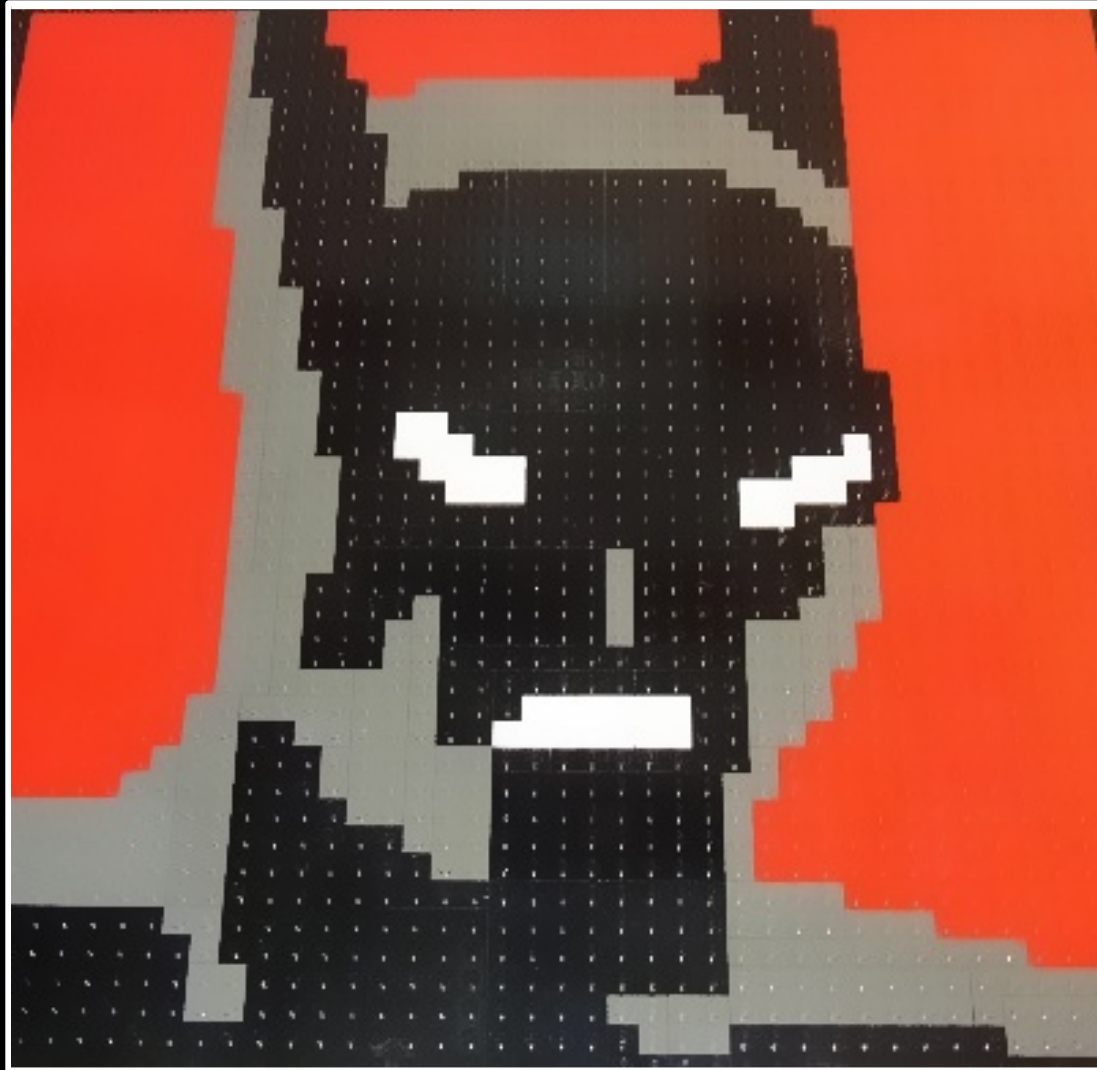
Strong Edge



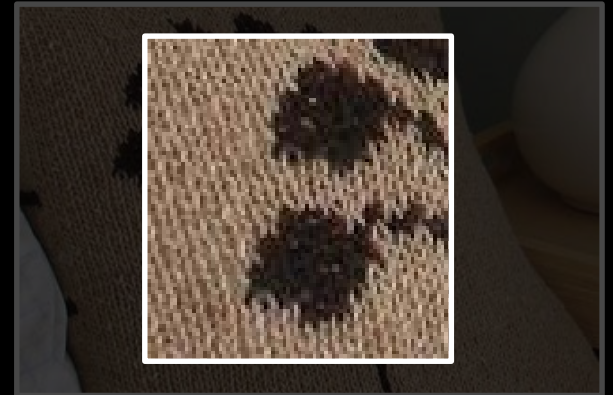
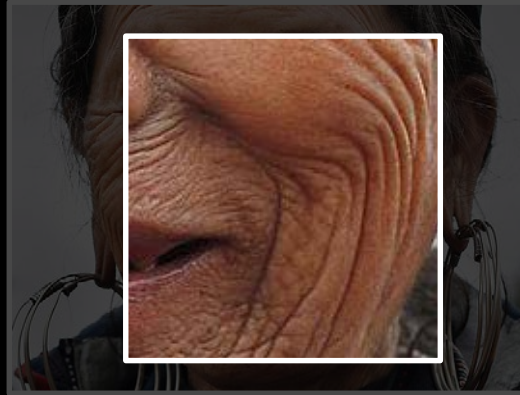
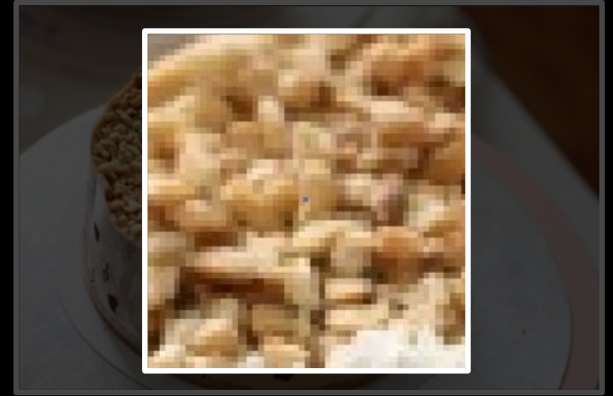
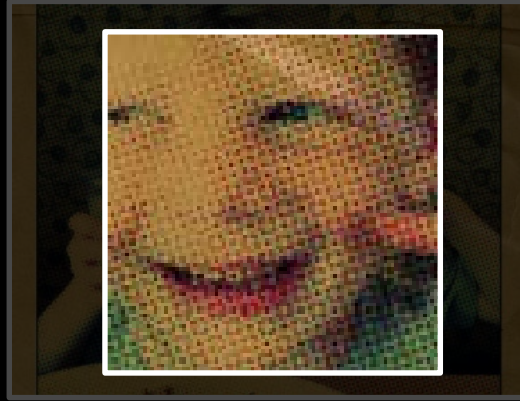
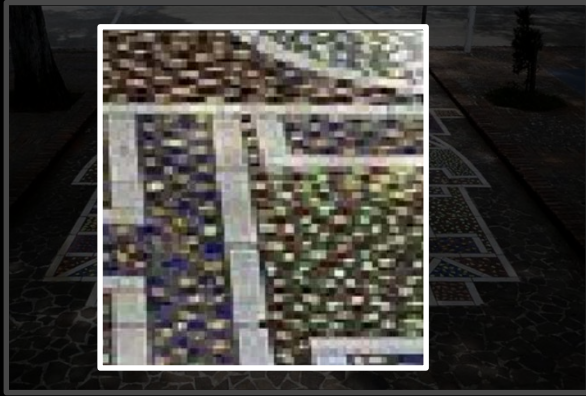
Edge preserving may not work
for pet beatification



Edge preserving also fails the batman



Edge-aware filter (bilateral filter)

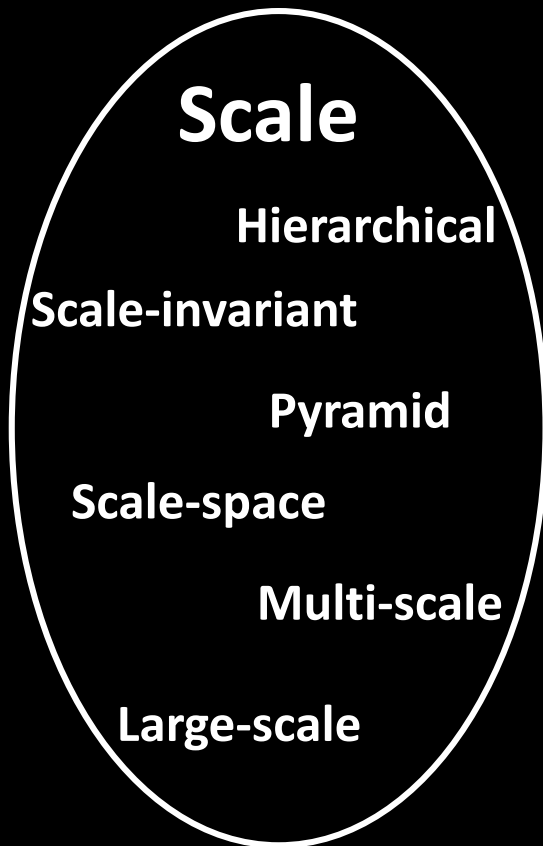


Many **tiny contents** are strong

What better characters them?

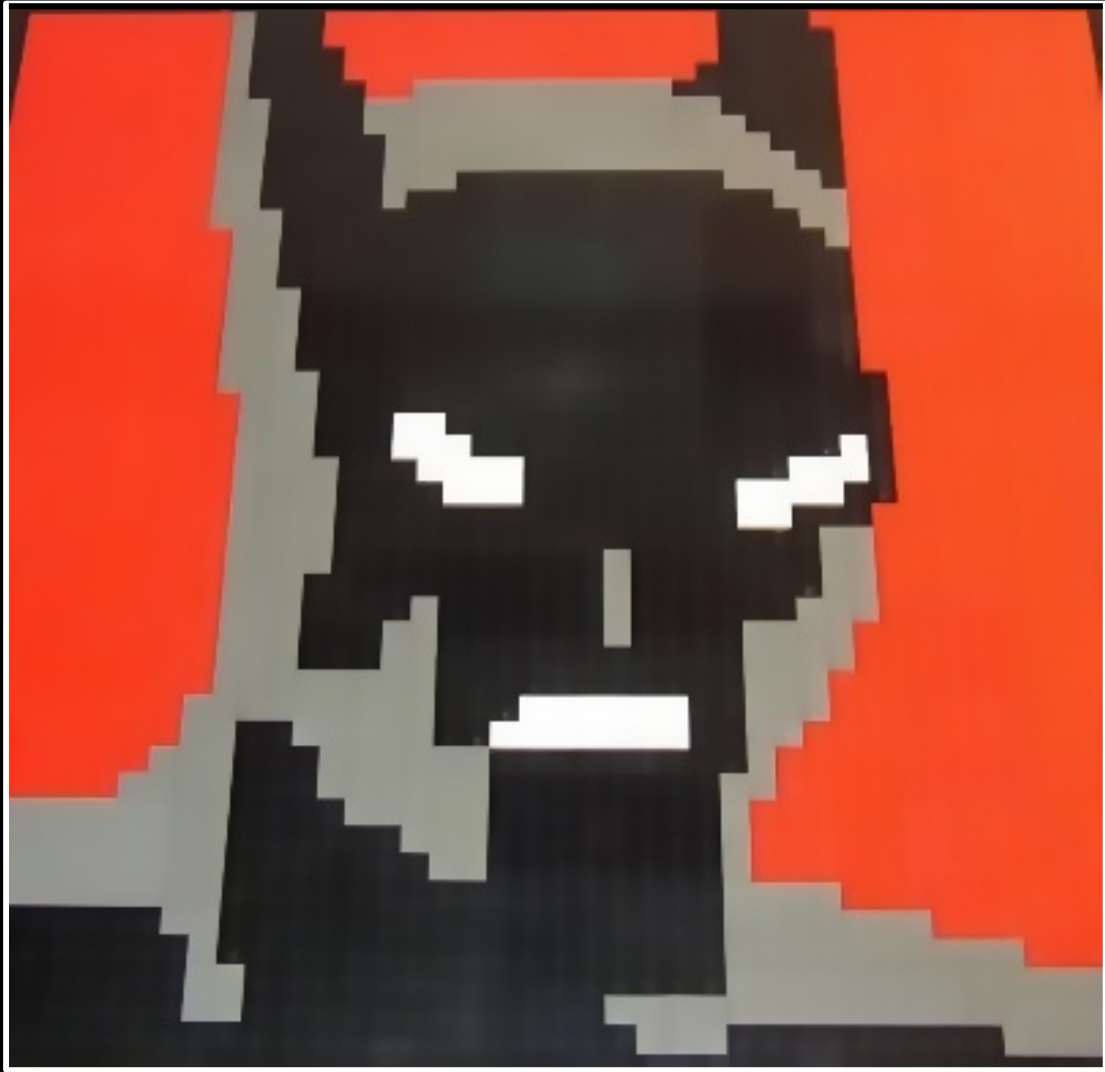
Scale!

Scale in Computer Vision



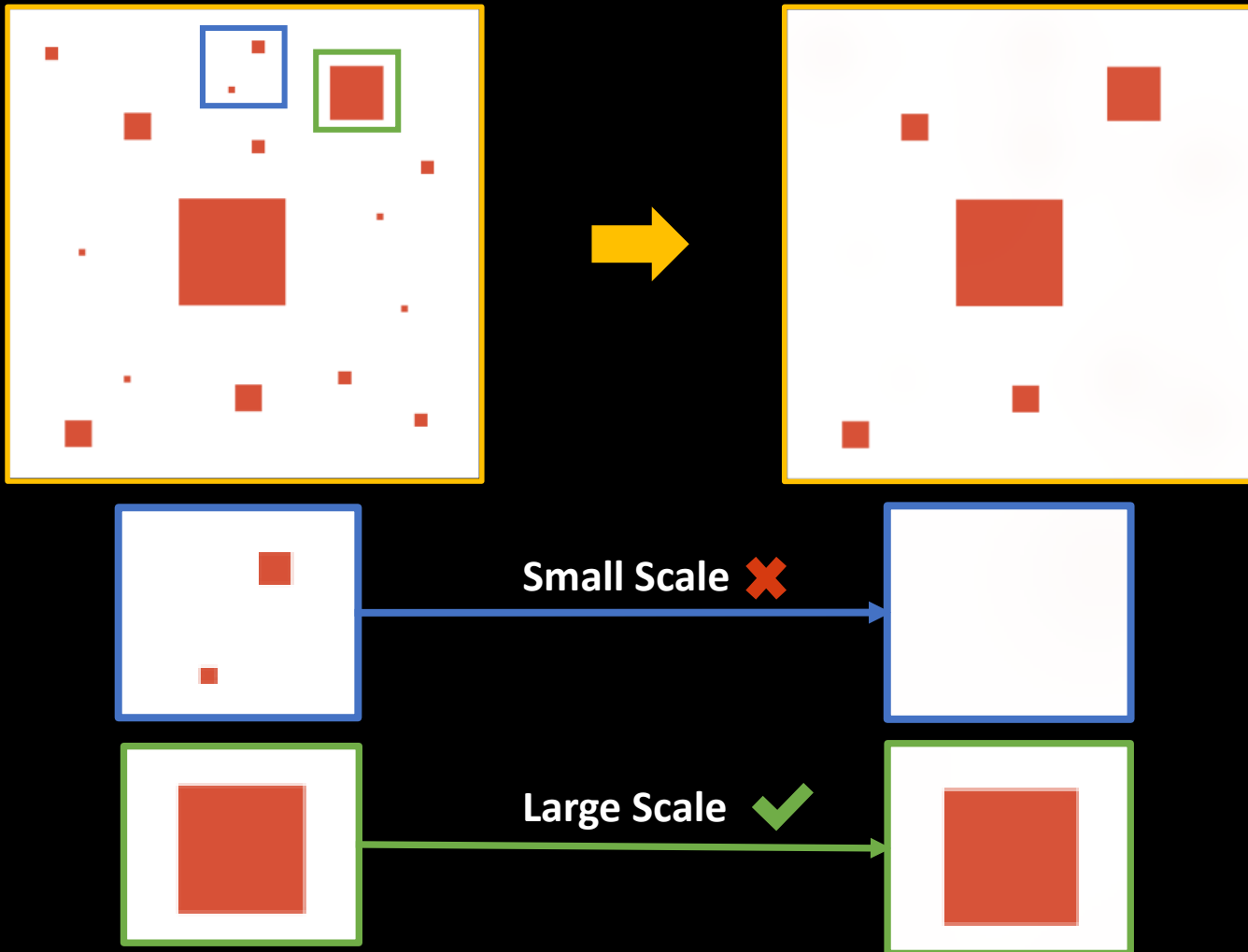
- Segmentation
- Object Detection
- Saliency Detection
- Feature Extraction
- Video Analysis
- Edge Detection
- Optical Flow & Stereo
- Scene Understanding
- Action Recognition
- ...

Scale + Image filter = ?





Scale-Aware Filtering



Related Work

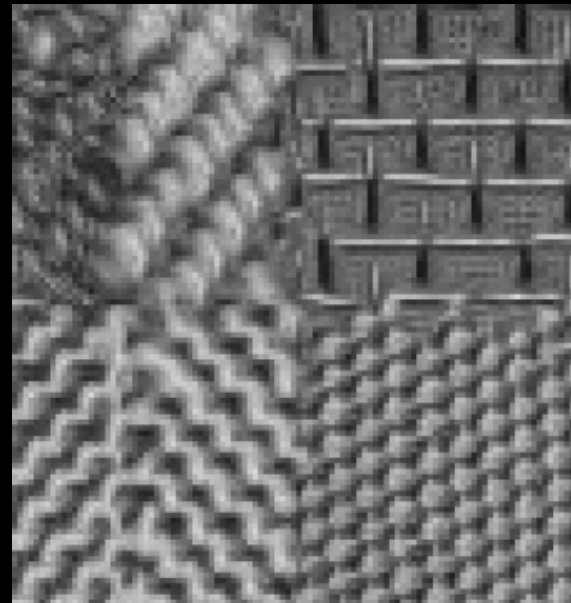
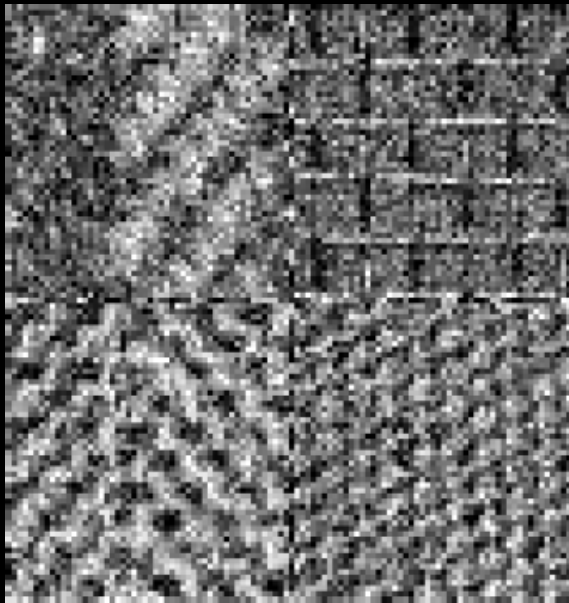
- Texture Smoothing

- Relative Total Variation [Xu et al., 2012]
- Region Covariance [Karacan et al., 2013]
- Weighted Median Filter [Zhang et al., 2014]
- Bilateral Texture Filtering [Cho et al., 2014]



Related Work

- Iterated Filtering Method
 - Iterated Non-local Means (INM) [Brox & Cremers, 2007]



Related Work

- Iterated Filtering Method
 - Iterated Non-local Means (INM) [Brox & Cremers, 2007]
- Comparison
 - Similar **iteration** scheme
 - Different **objective and functionality**
 - Remove texture vs preserve texture
 - Different **usage of iteration**
 - Edge recovery vs energy minimization

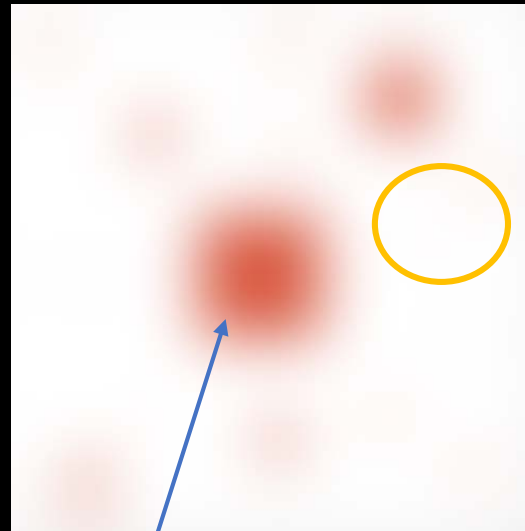
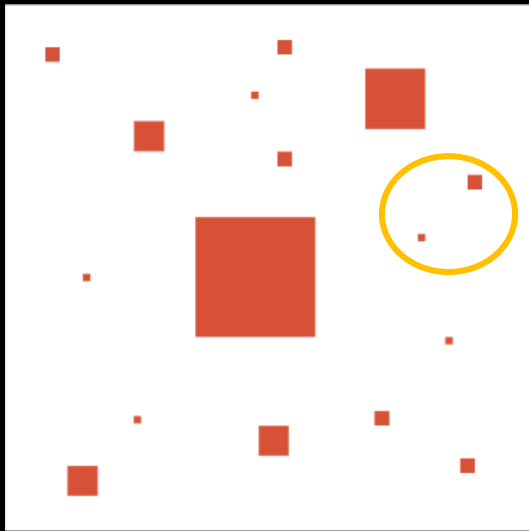
Interesting Fact

Our main algorithm has only **1 line** of code

```
1 Mat rollingGuidanceFilter(Mat im, float scale, int iter){  
2     Mat res = im.mul(0);  
3     while(iter-->0) res = bilateralFilter(im,res,scale,SIGMA_R);  
4     return res;  
5 }
```

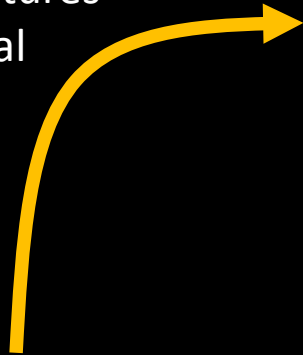
Main Idea

- Scale Space Theory [Lindeberg, 1994]:
 - An object of size t , will be largely smoothed away with Gaussian filter of variance t^2 .

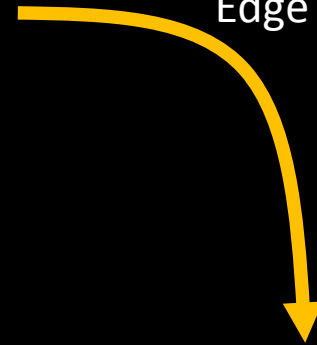


Our Scale-aware Filter

Step 1
Small Structures
Removal



Step 2
Edge Recovery



Step 1: Small Structures Removal

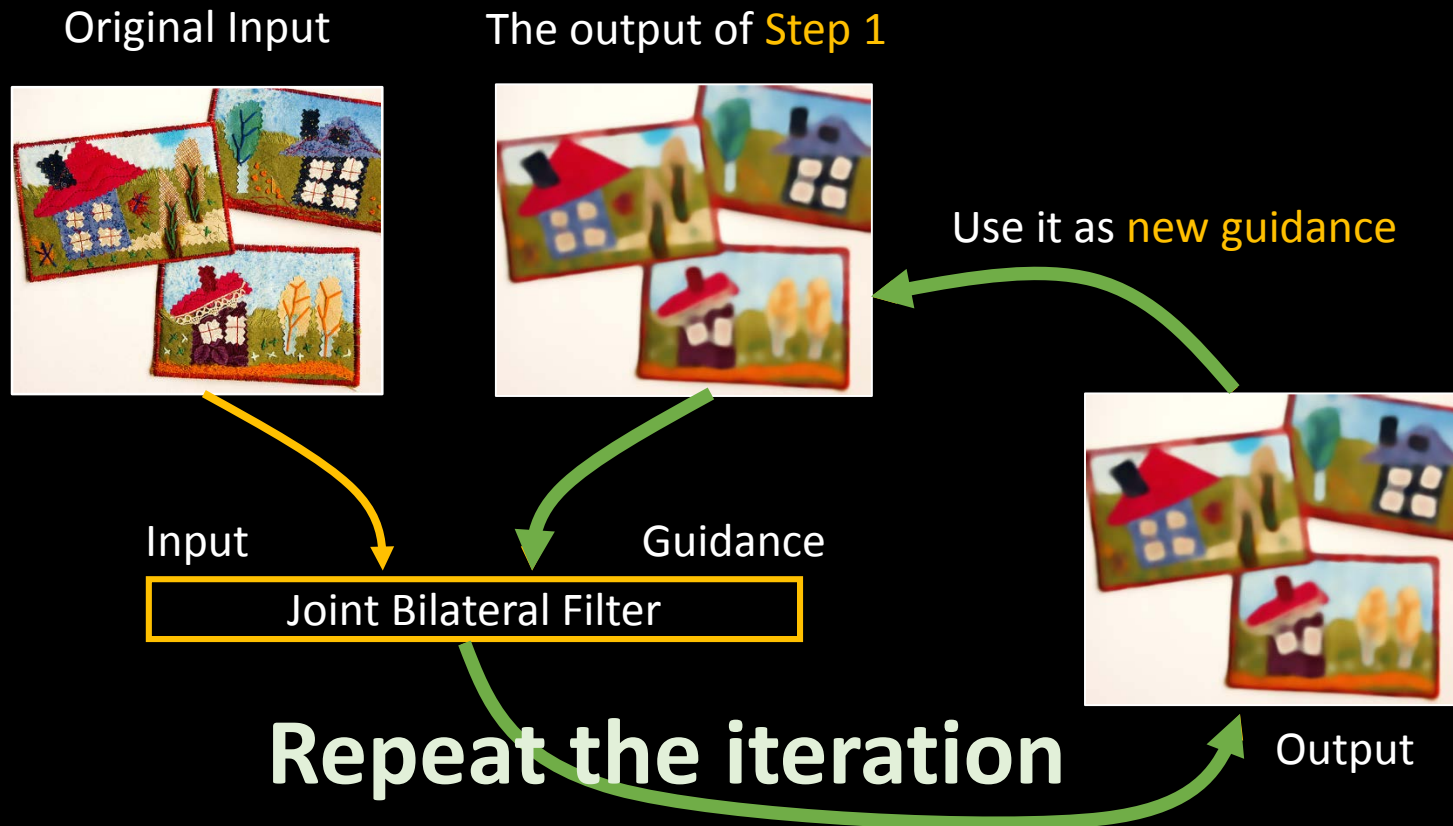
Gaussian Filter

$$J(p) = \frac{\sum_{q \in R(p)} \exp\left(-\frac{|p - q|^2}{2\sigma_s^2}\right) \cdot I(q)}{\sum_{q \in R(p)} \exp\left(-\frac{|p - q|^2}{2\sigma_s^2}\right)}$$

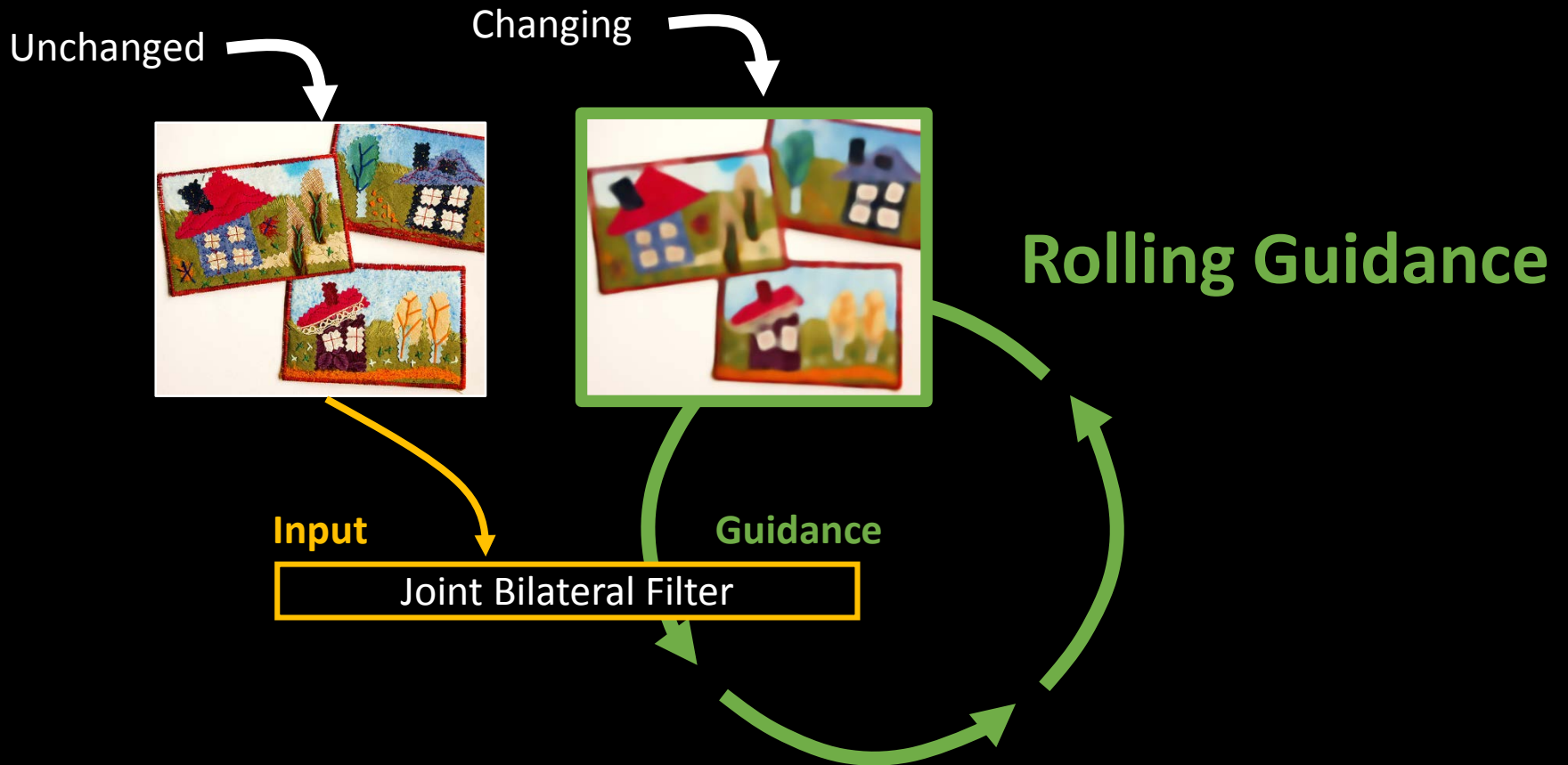


Step 2: Edge Recovery

- A rolling guidance



Rolling Guidance





Guidance for the 1st iteration



Guidance for the 2nd iteration



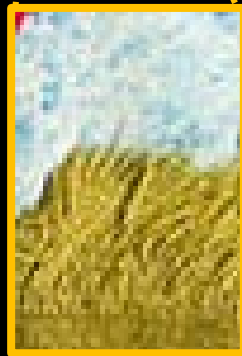
Guidance for the 3rd iteration



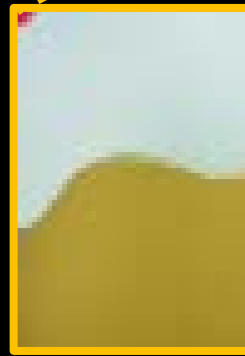
Guidance for the 5th iteration



Input



Output

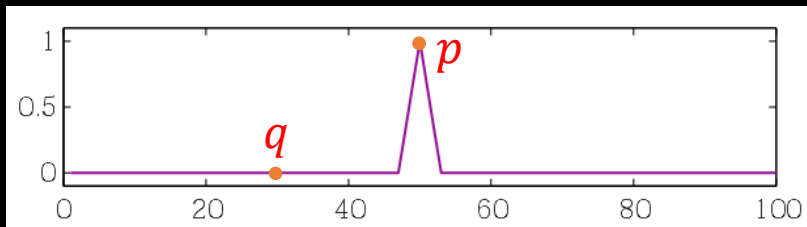


Small structures are **removed**.
Large structure are **NOT** blurred.

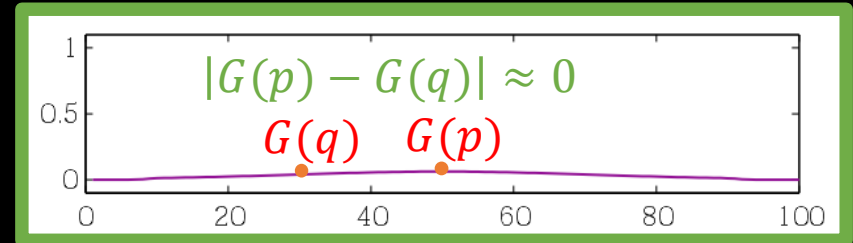
Why does **rolling guidance** work?

Small Structure

Input I



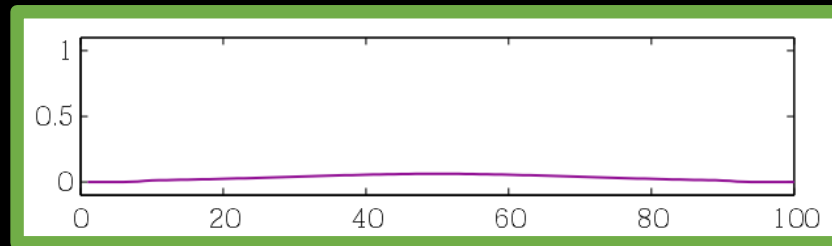
Guidance G (output of step 1)



$$J(p) = \frac{1}{K_p} \sum_{q \in R(p)} \exp\left(-\frac{|p - q|^2}{2\sigma_s^2}\right) \cdot I(q)$$

It becomes a Gaussian filter

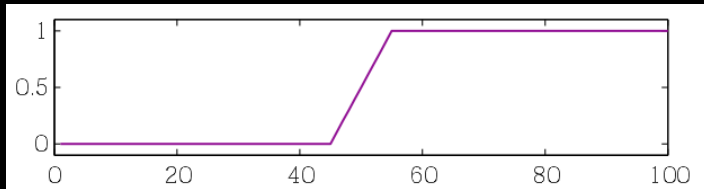
Joint Bilateral Filter



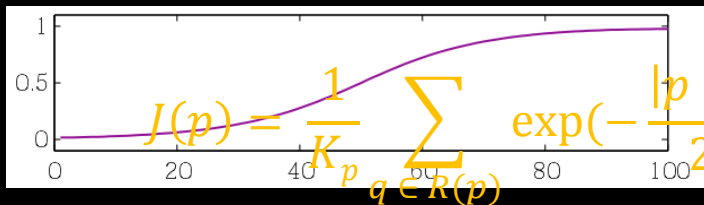
Same

Large Structure

Input Image



Result of Step 1

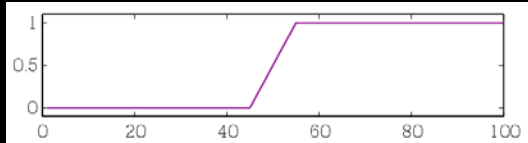


↓

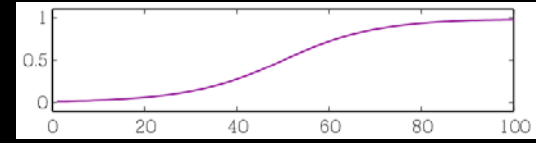
$$\exp\left(-\frac{|G(p) - G(q)|^2}{2\sigma_r^2}\right) \cdot I(q) \neq 1$$

Due to **this range weight**
It generates **sharper** results than Gaussian!

Input Image

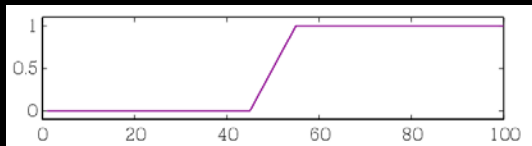


Guidance Image

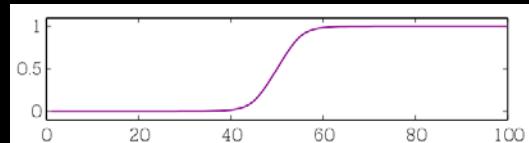


Joint Bilateral Filter

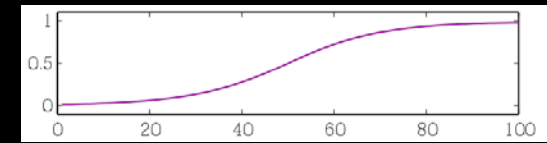
Input



Intermediate iterations



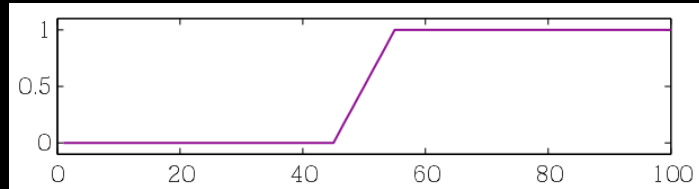
Previous guidance image



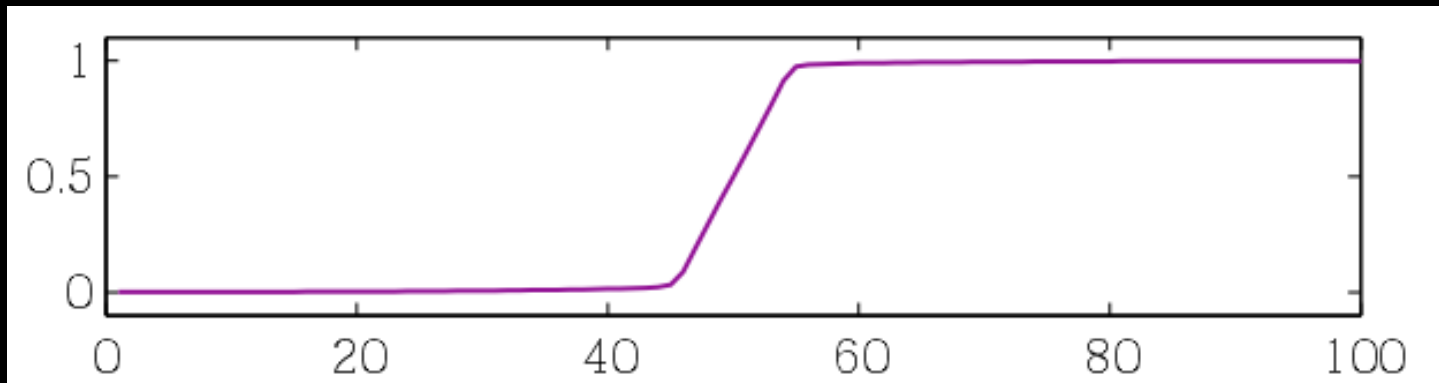
Smoothing

Range weight

Input Image

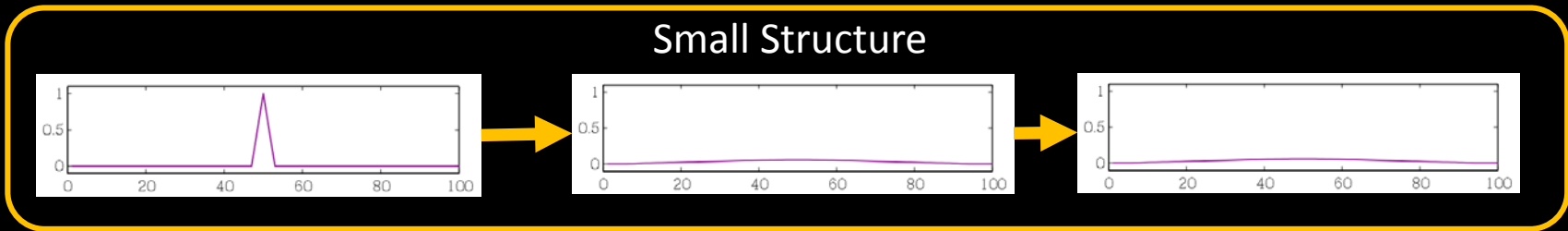
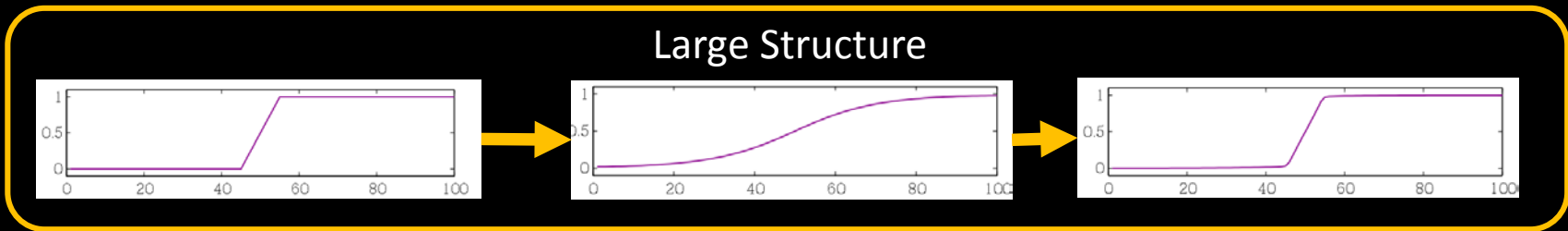


Guidance Image



3rd Iteration

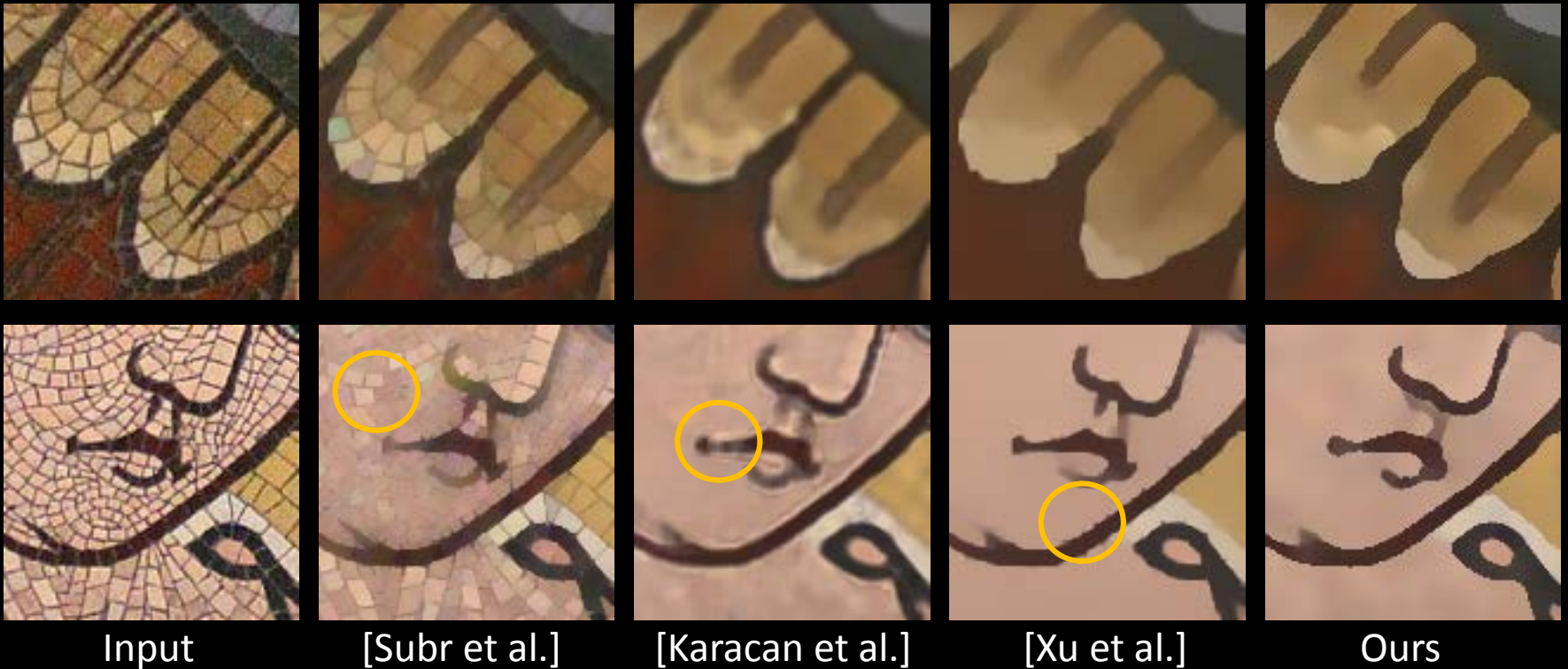
Rolling Guidance Filter



Comparison

Result Comparison

- Result comparison with related work



Performance Comparison

- Performance comparison with related works



Input

Ours

For 4 Megapixel Image

2

seconds

Performance Comparison

- Performance comparison

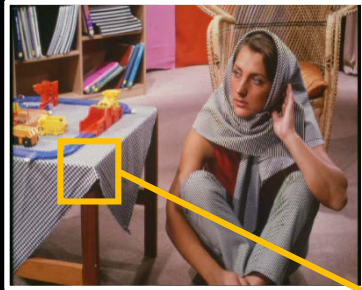
Algorithms	Time (seconds/Megapixel)
Local Extrema [Subr et al., 2009]	95
RTV [Xu et al., 2012]	14
Region Covariance [Karacan et al., 2013]	240
Ours	0.05 (Real-time)

Results & Application

Texture Removal



Texture Removal



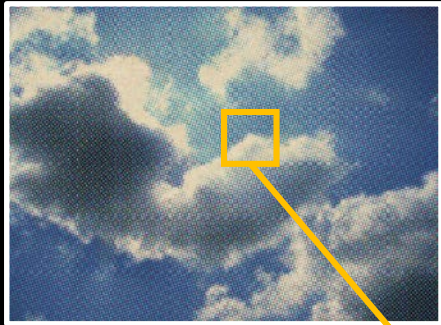
Texture Removal



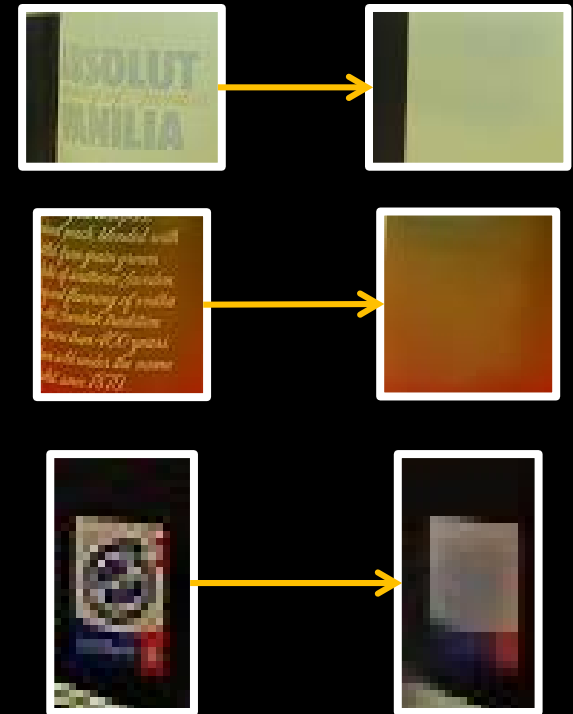
Halftone Image



Halftone Image

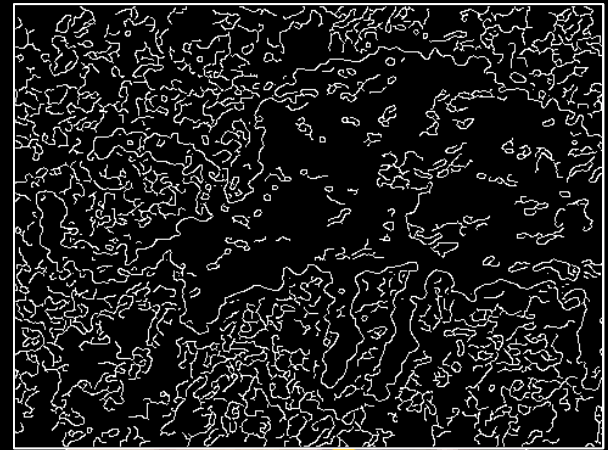
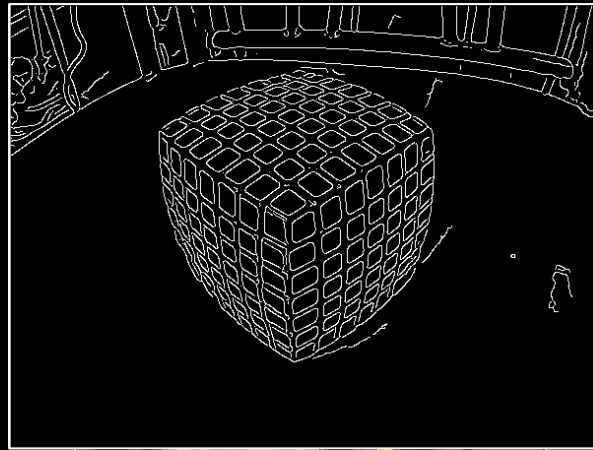
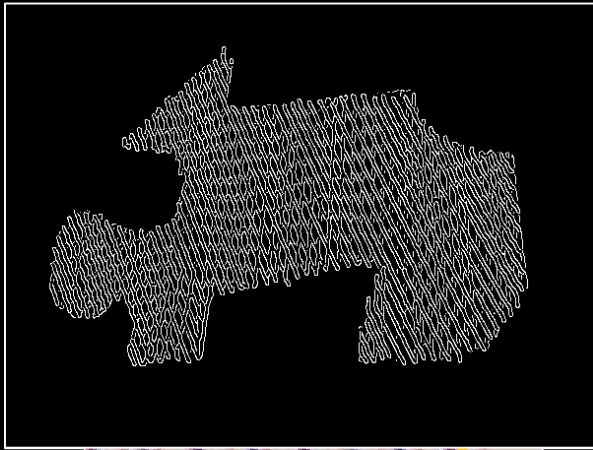


Small Text Removal

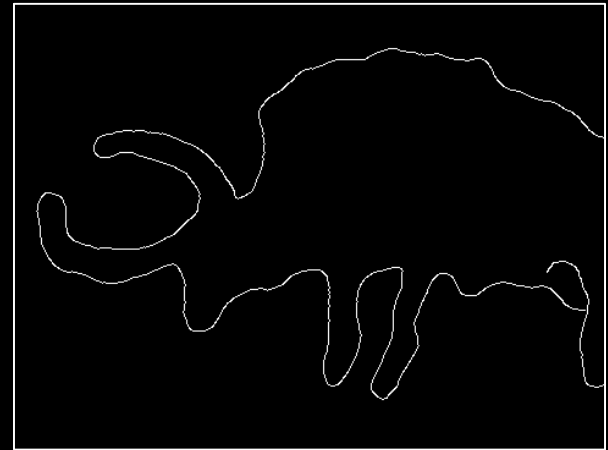
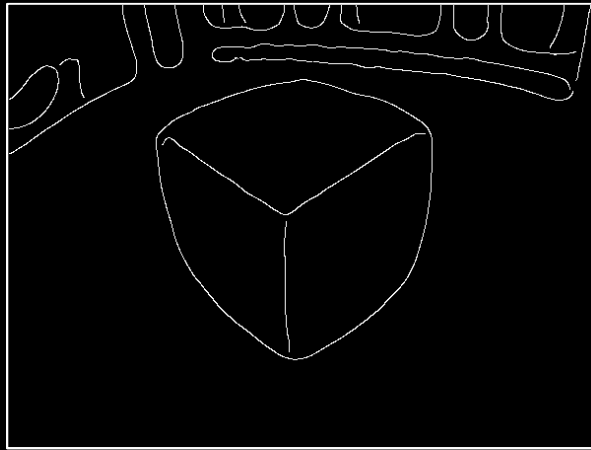
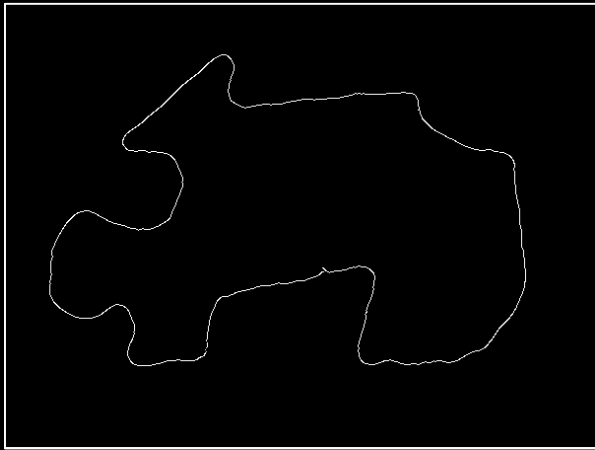


Remove Text

Virtual Edge Detection



Virtual Edge Detection



Natural Images

- Usable for
 - Segmentation
 - Saliency
 - Scene understanding
 - Background subtraction
 - Layer separation
 - Outlier removal
 - ...



Determining Scales

Step 1 $J(p) = \frac{1}{K_p} \sum_{q \in R(p)} \exp\left(-\frac{|p - q|^2}{2\sigma_s^2}\right) \cdot I(q)$ Gaussian

Step 2 $J(p) = \frac{1}{K_p} \sum_{q \in R(p)} \exp\left(-\frac{|p - q|^2}{2\sigma_s^2}\right) \cdot \exp\left(-\frac{|G(p) - G(q)|^2}{2\sigma_r^2}\right) \cdot I(q)$ Joint filter

σ_s determine the scale.

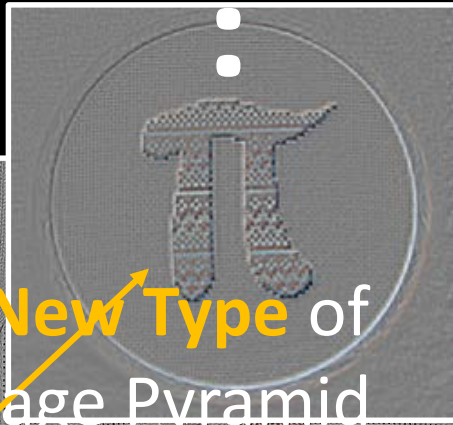
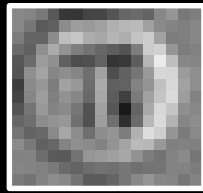
Multi-Scale Filtering



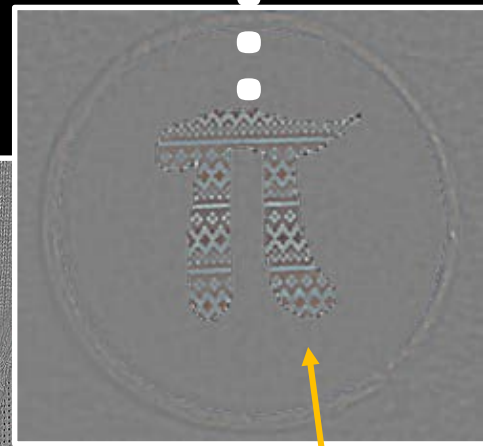
$$\sigma_s = 30$$



Input



A **New Type** of
Image Pyramid



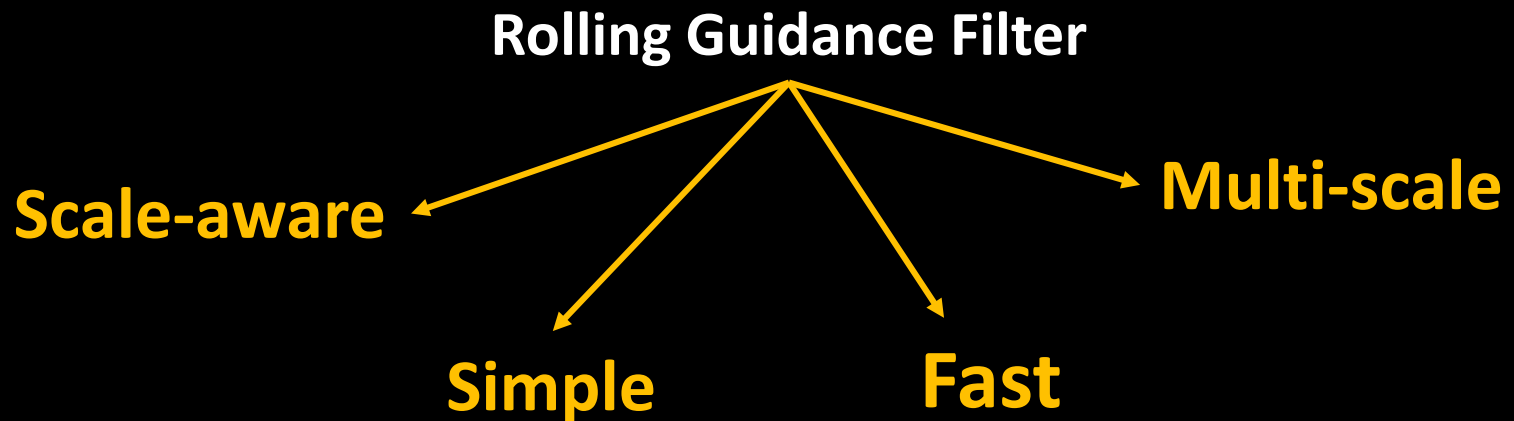
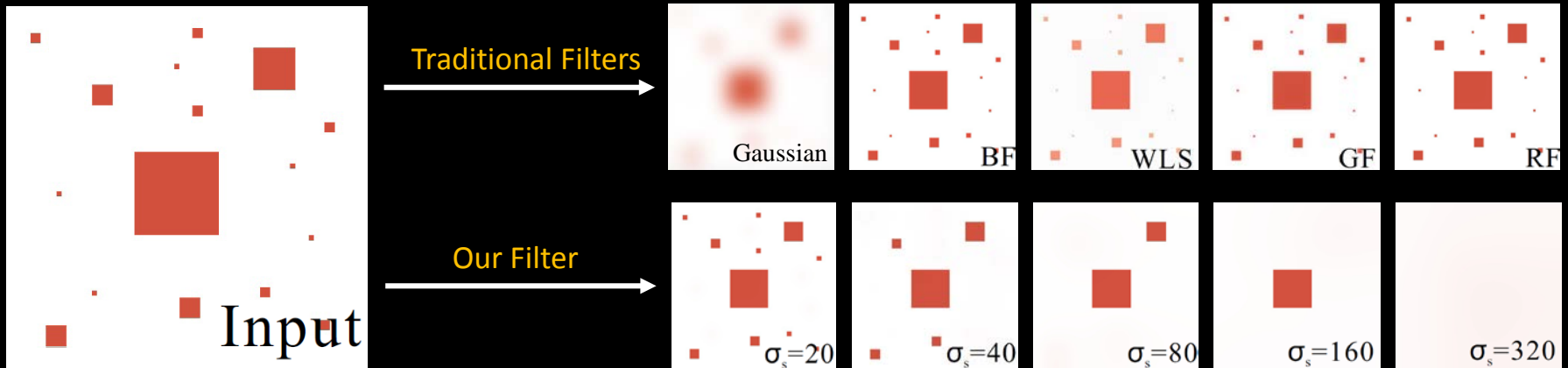
Content of all scales

Content of only current scale

Laplacian Pyramid
(generated with Gaussian)

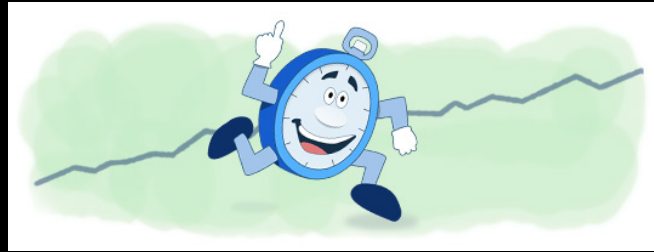
Texture Pyramid
(generated with Rolling Guidance Filter)

Summary



Code Available Online

```
while(iter--) res = bilateralFilter(im,res,scale,SIGMA_R);
```





Thank You

Limitations

- Sharp corners could be rounded
 - It is because sharp corner presents high frequency change.
 - In other words, sharp corners are small-scale structures.