

# Epitomic Image Colorization

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# Outline

## 1 Introduction

- Related Work
- Contribution

## 2 Formulation

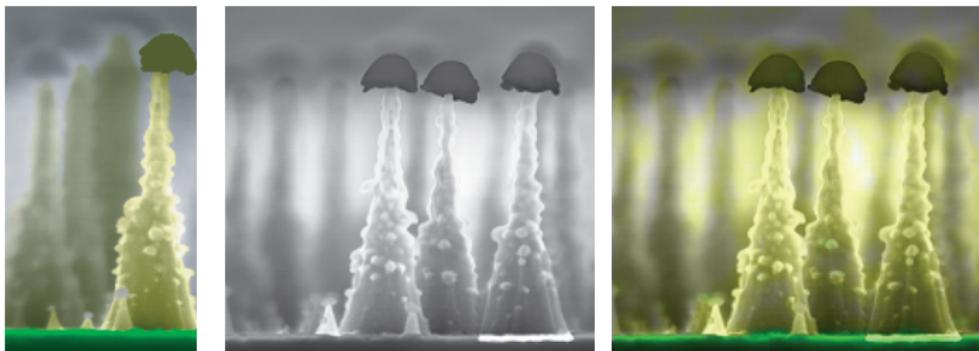
- Epitome
- Robust Patch Dissimilarity Measure via Epitome

## 3 Experimental Results

# Image Colorization

- A process of adding color to grayscale images
  - Increasing the visual appeal of images
  - Information illustration in scientific images
- Manual colorization is time consuming and tedious.
- We focus on automatic image colorization that transfers color from the reference image to the grayscale target image.

# Image Colorization



**Figure 1:** Colorize the Nano Mushroom-like structure by our method.  
From left to right: the reference image, the target image, the colorized target image.

# Related Work

- Transferring Color to Greyscale Images (Welsh et al., 2002)
  - Pixel-level matching by luminance value and neighborhood statistics
  - Suffers from spatial inconsistency
- Image Colorization Using Similar Images (Gupta et al., 2012)
  - A cascade feature matching scheme for matching the target superpixels to the reference superpixels
  - Lacks robust to change in pose or orientation

# Contribution

- We propose a new automatic image colorization method by epitome, called Epitomic Image Colorization
  - Achieve feature matching robust to both noise and the large change in the pose or orientation of the objects
  - Epitome is a generative model which summarizes raw image patches into a condensed representation.
- A new robust patch dissimilarity measure by epitome and the MRF inference.

# Introduction to Epitome

- Epitome (Jovic, Frey, & Kannan, 2003) is a generative model which summarizes raw image patches into a condensed representation similar to Gaussian Mixture Models (GMMs).
- In contrast to tradition GMMs, the Gaussian components of epitome can be overlapping with each other.



Figure 2: Examples of the learned epitome

# Introduction to Epitome

- The epitome  $\mathbf{e}$  is obtained by maximizing the log likelihood function:

$$\mathbf{e} = \arg \max_{\hat{\mathbf{e}}} \log p \left( \{\mathbf{Z}_k\}_{k=1}^Q | \hat{\mathbf{e}} \right), \quad (1)$$

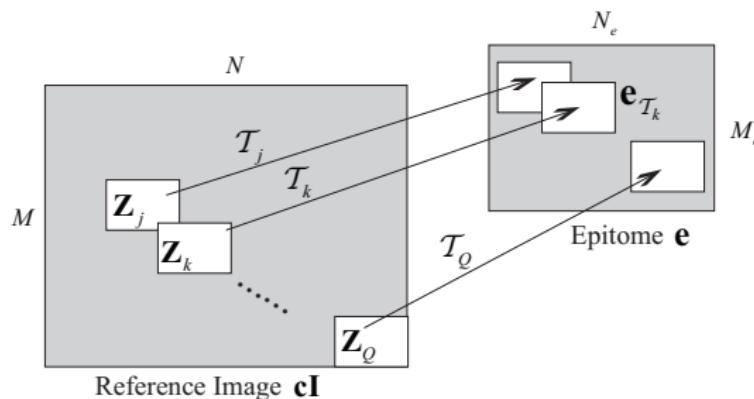


Figure 3: Learn the epitome from the reference image.  $\mathbf{Z}_k$  : patch from the reference image;  $T_k$ : hidden mappings that maps the image patch  $\mathbf{Z}_k$  to the epitome patch.

# Heterogeneous Feature Epitome

- We learn the pixel epitome  $e^{YIQ}$ , the dense SIFT epitome  $e^{SIFT}$  and the LBP epitome  $e^{LBP}$  jointly from the raw pixel, the dense SIFT feature (Lazebnik, Schmid, & Ponce, 2006) and the rotation invariant Local Binary Pattern (LBP) (Ojala, Pietikainen, & Maenpaa, 2002) of the reference image.
- The heterogeneous feature epitome  $\mathbf{e} = (e^{YIQ}, e^{SIFT}, e^{LBP})$

# Robust Patch Dissimilarity Measure via Epitome

- In order to match the target patch to the reference patch for color transfer, we need a robust patch dissimilarity measure.
- We propose a robust dissimilarity measure between the target patch  $\hat{\mathbf{Z}}_i$  and the reference patch  $\mathbf{Z}_j$  with the heterogeneous feature epitome  $\mathbf{e}$  learned from the reference image:

$$\mathcal{D}_{\mathbf{e}}(\hat{\mathbf{Z}}_i, \mathbf{Z}_j) = 1 - p(\hat{\mathcal{T}}_i^* | \mathbf{Z}_j, \mathbf{e}) \quad (2)$$

where  $\hat{\mathcal{T}}_i^*$  is the most probable hidden mapping for  $\hat{\mathbf{Z}}_i$ :

$$\hat{\mathcal{T}}_i^* = \arg \max_{\hat{\mathcal{T}}_i} p(\hat{\mathcal{T}}_i | \hat{\mathbf{Z}}_i, \mathbf{e}) \quad (3)$$

# Robust Patch Dissimilarity Measure via Epitome

- This dissimilarity measure is robust to noise and the large change in the pose or orientation of the objects.

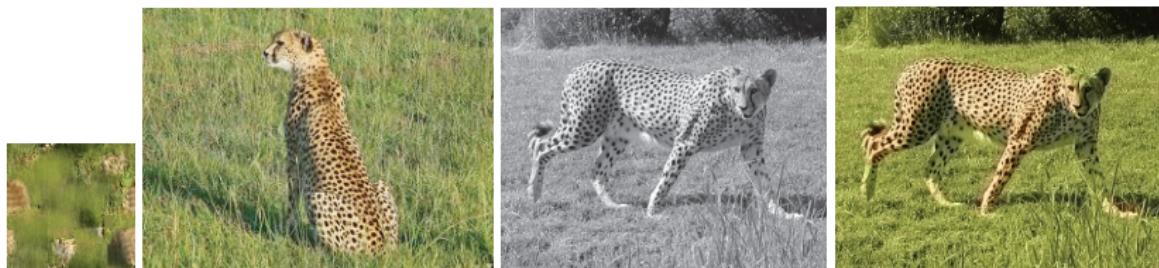


Figure 4: Colorize the cheetah

# Epitomic Image Colorization

- Use the robust patch dissimilarity measure via epitome to find similar reference patches for each target patch
- Transfer color from the similar reference patch to the target patch
- Use MRF inference to obtain a smooth colorization result

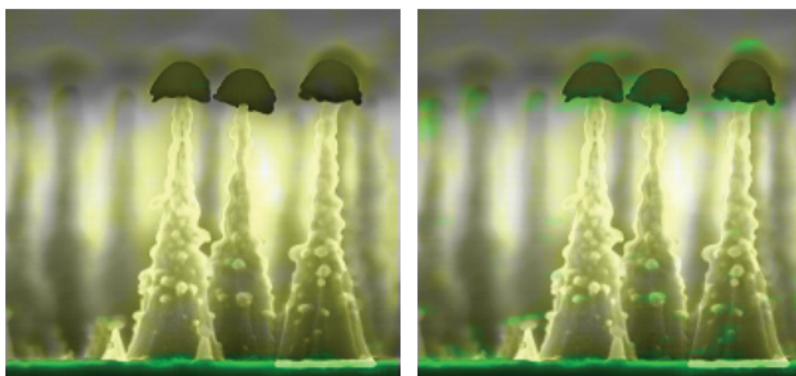


Figure 5: Comparison between colorizing the Nano image with MRF inference (left) or not (right).

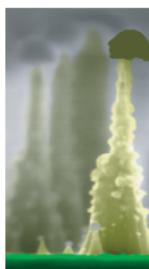
# Parameter Setting

- The area of the heterogeneous feature epitome is no more than  $\frac{1}{4}$  of that of the reference images.
- The patch size is  $9 \times 9$  or  $12 \times 12$ .

# Colorization Results



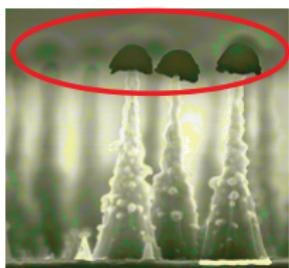
learned epitome



reference image



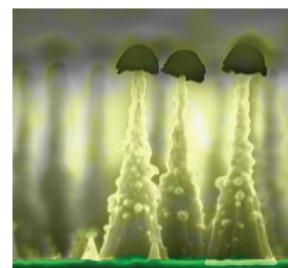
target image



Welsh et al.



Gupta et al.

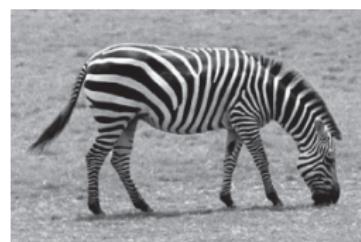


our result.

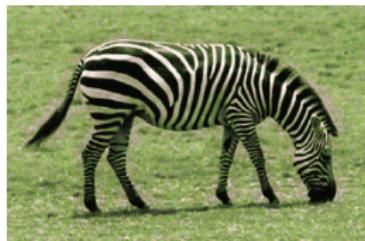
# Colorization Results Cont.



learned epitome reference image



target image



Welsh et al.



Gupta et al.



our result.

# Colorization Results Cont.



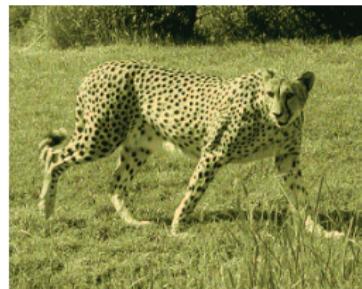
learned epitome



reference image



target image



Welsh et al.



Gupta et al.



our result.

# Colorization Results Cont.



learned epitome



reference image



target image



Welsh et al.



Gupta et al.



our result.

# Colorization Results Cont.



learned epitome



reference image



target image



Welsh et al.

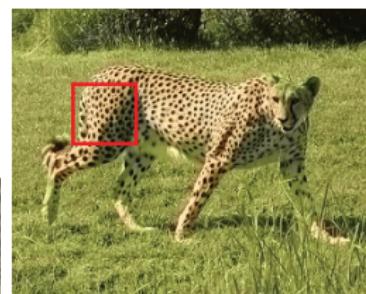
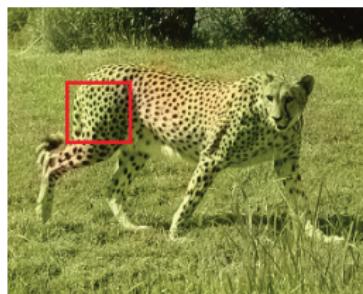
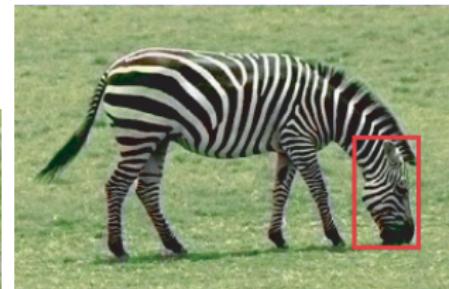
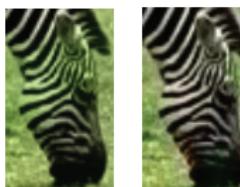
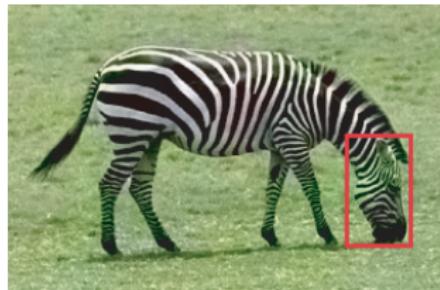


Gupta et al.



our result.

# Colorization Results Cont.



End

Thank you!

# Reference I

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## Reference II

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