An Optimization-Based Mesh-Generation Method for Image Representation

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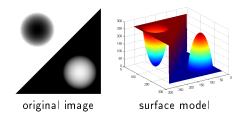
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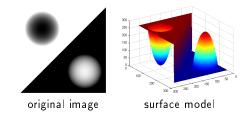
University of Victoria

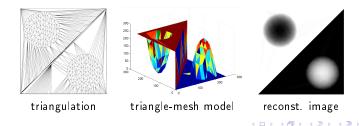
- Introduction to triangle-mesh models of images
- Background
- Proposed mesh-generation method
- Results
- Conclusions

Introduction Triangle Mesh Models of Images

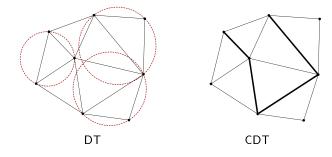


Introduction Triangle Mesh Models of Images





- Two classes of triangulation:
 - Delaunay triangulation (DT)
 - Constrained Delaunay triangulation (CDT)



Mesh-Generation Problem

Find a mesh model of original image ϕ defined in Λ with N number of points that minimizes the measure ϵ of error between ϕ and reconstructed image $\hat{\phi}$

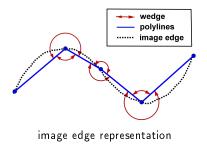
• Mean squared error:
$$\epsilon = \frac{1}{|\Lambda|} \sum_{(x,y) \in \Lambda} \left(\hat{\phi}(x,y) - \phi(x,y) \right)^2$$

- Peak signal-to-noise ratio: $PSNR_{dB} = 20 \log_{10} \left(\frac{2^{\rho} 1}{\sqrt{\epsilon}} \right)$
- ρ is the number of bits/pixel in the image

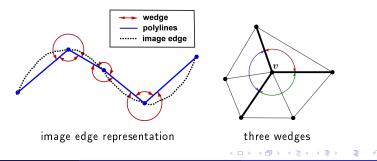
• Sampling density:
$$d = \frac{N}{|\Lambda|}$$

- Explicit representation of discontinuities (ERD)
- Originally proposed by Tu and Adams in 2013
- Based on constrained Delaunay triangulation (CDT)
- Image discontinuities allowed across constrained edges of triangulation
- Wedge-based discontinuity modeling

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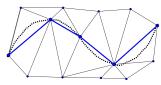
- Initial triangulation.
 - Detect image edges
 - Edges are approximated using polylines
 - Construct a constrained Delaunay triangulation



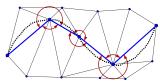
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Initial triangulation.

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- 2 Initial wedge values. Calculate the wedge values
- Point Selection. Select a new sample point to add to the mesh by the error-diffusion technique

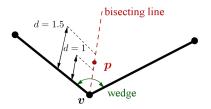
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- 9 Point insertion. Insert the selected point into the triangulation

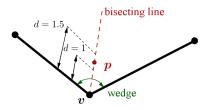
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- 9 Point insertion. Insert the selected point into the triangulation
- Stopping criterion. If the desired number of sample points is not achieved, go to step 3.

- The main focus of our work
- ERDED employs a local line search to calculate wedge values
- Line search is limited to $d \in [1, 1.5]$

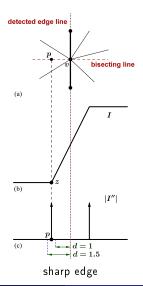


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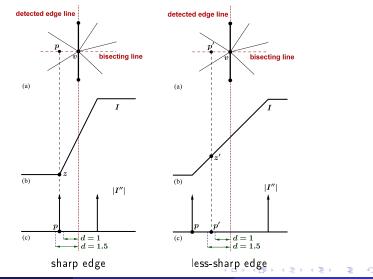
- *p* has the largest maximum magnitude of the second order directional derivative (MMSODD)
- z value at point p is selected as the wedge value associated with wedge

Analysis of Local Line Search



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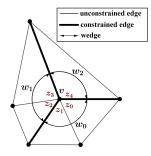
Analysis of Local Line Search



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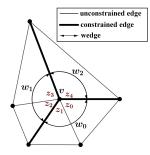
Optimization-Based Algorithm

• Introducing corner z values:



Optimization-Based Algorithm

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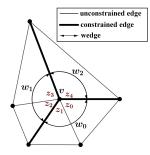


• Optimize corner z values to minimize the squared error inside each face $\rightarrow \min_{\{z_i\}} \sum_{p \in \Omega} \left| \hat{\phi}(p) - \phi(p) \right|^2$

 $\bullet~\Omega$ is the set of grid points in the face of interest

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•
$$w_0 = \frac{z_0 + z_1}{2}, w_1 = \frac{z_2 + z_3}{2}, w_2 = z_4$$

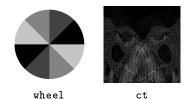


• Photographic, medical, and computer-generated imagery used

peppers



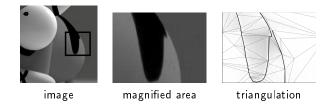




• 18/20 of the test cases by a margin of up to 5.06 dB

| lm ag e | Sampling | PSNR(dB) | |
|-------------|-------------|-----------|-------|
| | Density (%) | Prop osed | ERDED |
| p ep p er s | 0.5 | 22.49 | 22.14 |
| | 1 | 26.47 | 25.97 |
| | 2 | 29.26 | 29.00 |
| | 3 | 30.37 | 30.17 |
| lena | 0.5 | 20.81 | 20.55 |
| | 1 | 26.14 | 25.81 |
| | 2 | 29.38 | 29.28 |
| | 3 | 31.30 | 31.31 |
| ct | 0.125 | 18.72 | 15.63 |
| | 0.25 | 27.42 | 25.97 |
| | 0.5 | 31.62 | 30.06 |
| | 1 | 36.62 | 36.51 |
| bull | 0.125 | 25.22 | 24.68 |
| | 0.25 | 29.47 | 28.86 |
| | 0.5 | 35.35 | 35.15 |
| | 1 | 39.21 | 39.02 |
| wheel | 0.0625 | 35.75 | 30.69 |
| | 0.125 | 36.54 | 34.55 |
| | 0.25 | 37.24 | 35.56 |
| | 0.5 | 37.38 | 37.86 |

Results Subjective Comparison



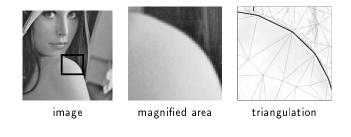


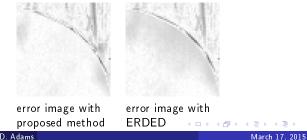
error image with proposed method



error image with ERDED

Results Subjective Comparison Cont'd





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- Modified version of the ERDED mesh-generation method was proposed
- An optimization-based algorithm was employed which
 - Better exploits the image content
 - Reduces the approximation error in the reconstructed image
- The proposed method outperforms the ERDED method both in terms of PSNR and subjective quality
- The improved approximation quality comes at a relatively modest computational cost

THANK YOU

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