A Proper Choice of Vertices for Triangulation Representation of Digital Images Ivana Kolingerova, Josef Kohout, Michal Rulf, Vaclav Uher, Proc. 2010 International Conference on Computer Vision and Graphics: Part II, pp. 41-48, 2010.

Milestones and Advances in Image Analysis

Stephanie Jennewein

04. December 2012

Motivation

- 2 Triangulation
- A Proper Choice of Vertices

Summary

Triangulation

- representation of digital images
- enables geometric transformations
- very simple
- low compression in comparison with frequency-based methods

Can we save disk space while preserving a good quality?

Strategy

choose the triangulation vertices randomly

What to do:

- assess proper number of vertices (computed from compression rate or given by the user)
- 2 choose set of pixels
- Scompute triangulation with Delaunay triangulation
- decoding

What to do:

- assess proper number of vertices (computed from compression rate or given by the user)
- O choose set of pixels
- Scompute triangulation with Delaunay triangulation
- decoding

2) choose set of pixels

• in general, choose edge points

Edge Point

A strong change in the grey values within a neighbourhood indicates an edge.

source: IPCV 2011-12

- edge detecting operators:
 - Roberts's operator
 - Laplace operator
 - Gaussian operator

Roberts's operator

$$Op_{Roberts}(i, j, f_{i,j}) = |\underbrace{f_{i,j} - f_{i+1,j+1}}_{\left[\begin{array}{c}1 & 0\\0 & -1\end{array}\right]}| + |\underbrace{f_{i+1,j} - f_{i,j+1}}_{\left[\begin{array}{c}1 & 0\\-1 & 0\end{array}\right]}|$$

[i,j] belongs to the set of vertices if

$$Op_{Roberts}(i, j, f_{i,j}) > T$$

- picture f
- pixel i,j
- threshold T

Laplace operator (Laplace4)

$$Op_{Laplace4}(i, j, f_{i,j}) = |\underbrace{f_{i,j-1} + f_{i,j+1} + f_{i-1,j} + f_{i+1,j} - 4f_{i,j}}_{0 \ 1 \ 0 \ 1 \ 0 \ 1 \ 0 \ 1 \ 0 \ 1 \ 0}|_{f_{i,j}}$$

[i,j] belongs to the set of vertices if $Op_{Laplace4}(i, j, f_{i,j}) > T$

- picture f
- pixel i,j
- threshold T

Laplace operator (Laplace8)

 $Op_{Laplace8}(i, j, f_{i,j}) =$



[i,j] belongs to the set of vertices if $Op_{Laplace8}(i, j, f_{i,j}) > T$

- picture f
- pixel i,j
- threshold T

Gaussian operator

$$Op_{GauB}(i, j, f_{i,j}) = \sum_{k=-\nu}^{r} \sum_{l=-\nu}^{r} |f_{i,j} - f_{i+k,j+l} \cdot exp(-\frac{k^2 + l^2}{2\sigma^2})|$$

[i,j] belongs to the set of vertices if $Op_{GauB}(i, j, f_{i,j}) > T$

- picture f
- pixel i,j
- threshold T
- -v and r: influence factor of the point in this area
- σ : vicinity area, width, standart deviation



Figure: The 9-10% pixels with the highest evaluation according to the presented operators, a) Roberts, b) Laplace4, c) Laplace8, d) Gauß

Store coodrinates and the intensity of the chosen pixels

Strategy

choose the pixels randomly

 \Rightarrow don't have to store coordinates <u>Reason:</u> coordinates can be recomputed from the seed of the random generator during decoding

• random point: chosen randomly

• edge point: chosen by an edge operator

What to do:

- assess proper number of vertices (computed from compression rate or given by the user)
- 2 choose set of pixels
- Occupie triangulation with Delaunay triangulation
- decoding

Triangulation - 4) compute triangulation with Delaunay triangulation

• choose triangles in such a way, that the following property is fulfilled:

empty circumcircle criterion:

the circumcircle of any triangle does not contain any of the given vertices in its interior

- goal: maximize the minimum angle of all the angles of the triangles in the triangulation
- ambiguity: two neighbouring triangles have the same circumcircle
- remedy: choose diagonal with lower intensity gradient

What to do:

- assess proper number of vertices (computed from compression rate or given by the user)
- 2 choose set of pixels
- Scompute triangulation with Delaunay triangulation
- decoding

- 5) decoding
 - values of intensities inside triangles are interpolated from the known vertex intensity values
 - coordinates of random points can be reconstructed with the seed of the random generator

Motivation

- 2 Triangulation
- A Proper Choice of Vertices
- Summary

Comparing edge detection operators using edge points and random points

- goal: highest fidelity and at least some compression
- Laplace:
 - best for a low number of edge points
 - and high number of random points
- Roberts:
 - best for a high number of edge points
 - and low number of random points
- Gauß:
 - worst results
 - slowest operator



Figure: The image Fruits: Dependence of MSE on the total number of points of which 8-10% are edge points



Figure: The image Fruits: Dependence of MSE on the total number of points of which 8% are random points

Why not choose only random points?







Figure: 20% of points: only random points (MSE 99.11)

Choosing only edge points



Figure: The image Fruits



Figure: 20% of points: only edge points (MSE 131.94)



Figure: The image Fruits: Only random and only edge points

Proper choice with acceptable quality and some compression:

- for most common images:
 - number random points: 10 15% of the image size
 - number edge points: 5 10% of the image size
- for images with many edges:
 - number random points: 10 15% of the image size
 - number edge points: 15 20% of the image size

24 / 30



Figure: The image Fruits; a) input, b) result - 11% of edge points, 15% of random points, MSE=20.65





Motivation

- 2 Triangulation
- A Proper Choice of Vertices

Summary

- triangulation is simple and good for geometric transformations
- points for vertices of triangles can be chosen randomly, because the coordinates can be reconstructed during decoding
- to keep high quality of the image, one has to find a proper rate of random and edge points
- the Laplace-operator is best since we want to achieve a high number of random points while preserving the quality

References

- Ivana Kolingerova, Josef Kohout, Michal Rulf, Vaclav Uher: A proper choice of vertices for triangulation representation of digital images. Proc. 2010 International Conference on Computer Vision and Graphics: Part II, pp. 41-48, 2010.
- Lecture-Notes: IPCV 2011/12 and 2012/13
- Josef Kohout, On Digital Image Representation by the Delaunay Triangulation.Department of Computer Science and Engineering, University of West Bohemia, Univerzitni 22, 306 14 Plze, Czech Republic besoft@kiv.zcu.cz
- De Berg, M., van Kreveld, M., Overmars, M., Schwarzkopf, O.: Computational geometry. In: Algorithms and applications. Springer, Heidelberg (1997)
- Milan Sonka, Vaclav Hlavac, Roger Boyle: Image Processing, Analysis, and Machine Vision. ITP (1999)
- Galic, I., Weickert, J., Welk, M.: Towards PDE-based Image Compression. In: Para- gios, N., Faugeras, O., Chan, T., Schn orr, C. (eds.) VLSM 2005. LNCS, vol. 3752, pp. 3748. Springer, Heidelberg (2005)