

Automated Detection of Chondrocytes in Growth Plate Images

Emily Beylerian, Brian de Silva, Ben Gross, Hannah Kastein Mentors: Dr. Maria-Grazia Ascenzi and Hayden Schaeffer

UCLA Applied Mathematics REU 2012

シック・ 川 ・ (川・ (川・ (四)・ (日)・

Automated Detection of Chondrocytes in Growth Plate Images

Growth Plates

Bones

- Longitudinal bone growth via growth plates
- Chondrocytes arranged in vertical columns
- Cell division causes growth



A The local

nan

Our Project

Disorders in the Growth Plate

- Misaligned chondrocytes stunt growth
- Alignment depends on genetic factors

Our Goals

- Develop automated image processing software
- Compare normal and abnormal growth plates



Image Processing

 Want to extract cell locations from matrices of intensity values

Challenges

- Cells beneath plane of focus
- Inconsistencies across image
 - Cell size/shape
 - Appearance of nuclei
 - Stain penetration in background





Pre-Existing Methods

Software

- ImageJ
- CellProfiler
- High-throughput and high-content screening (HT-HCS)

Methods

- Segmentation
- Cartoon-Texture Decomposition
- K-Means Clustering
- Spectral Clustering



Manual detections (blue) overlaid on texture decomposition

Algorithm





Retinex: A Color Contrast Algorithm

- Attempt to imitate and describe human color perception
- Smooths together subtle variations in shading
- Remove cells outside plane of focus







Retinex cont.

Given an initial image, f, finds a reconstructed image, u, such that

$$-\Delta u_{i,j} = F_{i,j}$$

Where

$$\Delta u_{i,j} = u_{i+1,j} + u_{i-1,j} + u_{i,j+1} + u_{i,j-1} - 4u_{i,j}$$

is the discrete Laplacian at mesh point (i,j),

$$F_{i,j} = T(f_{i,j} - f_{i+1,j}) + T(f_{i,j} - f_{i-1,j}) + T(f_{i,j} - f_{i,j+1}) + T(f_{i,j} - f_{i,j-1})$$

and T is a thresholding function such that

$$T(x) = \begin{cases} 0 & \text{if } |x| \le \tau \\ x & \text{if } |x| > \tau \end{cases}$$

- イロト 4 回 ト 4 回 ト 4 回 ト 4 回 ト 4 回 ト 4 回 ト 4

Anisotropic Diffusion

Cartoon





$$\min_{u} \int \Psi \left(\nabla u, \nabla u^{T} \right) dx$$
$$g\left(|\nabla u| \right) = \frac{1}{|\nabla u|^{p}}$$

Related to Perona-Malik Diffusion

不同 トイラト イラト

- Both nonlocal and nonlinear
- Emphasis on preserving edges



Morphological Functions

Convex Hull

- ► Fits polygon to cell outline
- Connects discontinuities linearly
- Need separation between cells



Morphological Functions

Size Thresholding

Shape Thresholding

Isoperimetric Inequality

 $\frac{4\pi A}{L^2}$

- Relationship between shape area and circumference
- Ratio equals one for circle







▲ 臣 ▶ ▲ 臣 ▶ ○ � (♡

Automated Detection of Chondrocytes in Growth Plate Images

Automated Growth Plate Zone Detection

- Classify each object by its Isoperimetric Ratio $\left(\frac{4\pi A}{I^2}\right)$
- Plot object location vs. Ratio and approximate graph with a fourth degree polynomial
- Inflection points at zone boundaries



Results



E 990

Automated Detection of Chondrocytes in Growth Plate Images

Results





Results

Original Growth Plate



Overlay of Final Output





Clustering Error Statistics

Error Analysis

- How well does our algorithm segment the cells in our images?
- Many metrics/statistics for clustering analysis

Metric Requirements

- Can compare an unequal number of clusters
- ► Can handle large variations in cluster sizes
- Must not be computationally complex



Error Statistics

- ► Classes $\{L_j\}$ in our 'ground truth' image $S = \{I_1, ..., I_N\}$
- Clusters $\{C_i\}$ in our segmented image $S' = \{I'_1, \ldots, I'_N\}$

Clustering Purity

- Purity= $\sum_{i} \frac{|C_i|}{N} max_j \frac{|C_i \cap L_j|}{|L_j|}$
- Not robust to clusters which subdivide classes, trivial clusters, or clusters that span multiple classes
- ► Up to 88%



Rand Indicies

- ► Classes $\{L_j\}$ in our 'ground truth' image $S = \{I_1, ..., I_N\}$
- Clusters $\{C_i\}$ in our segmented image $S' = \{I'_1, \ldots, I'_N\}$

Rand Index

►
$$R(S, S') = \frac{1}{\binom{N}{2}} (|A| + |B|)$$

where $A = \{(i, j) | i \neq j, I_i = I_j, I'_i = I'_j\}$ and $B = \{(i, j) | i \neq j, I_i \neq I_j, I'_i \neq I'_j\}$

- Compare clustering of pairs of pixels
- Penalizes false positives and true negatives equally
- ▶ Up to 74%



Conclusion

In summary, we developed a successful algorithm for:

- extraction of chondrocyte location
- zone approximations



Questions?



References

M-G Ascenzi, C. Blanco, I. Drayer, H. Kim, R. Wilson, K.N. Retting, K.M. Lyons, and G. Mohler. 'Effect of Localization, Length, and Orientation of Chondrocytic Primary Cilium on Murine Growth Plate Organization,' *Journal of Theoretical Biology*, 285(1):147-155, September 2011.

M-G Ascenzi, M. Lenox, and C. Farnum. 'Analysis of the Orientation of Primary Cilia in Growth Plate Cartilage: a Mathematical Model Based on Multiphoton Microscopical Images,' *Journal of Structural Biology*, 158(3):293-306, June 2007.

T. Brox, J. Weickert, B. Burgeth, and P. Mrázek. 'Nonlinear Structure Tensors,' *Universität des Saarlandes, Fachrichtung 6.1 - Mathematik*, 113, 2004.

J-M Morel, A.B. Petro, and C. Sbert. 'A PDE Formalization of the Retinex Theory,' *IEEE Transactions on Image Processing*, 2010.

J. Weickert. 'Anisotropic Diffusion in Image Processing,' ECMI, 1998.