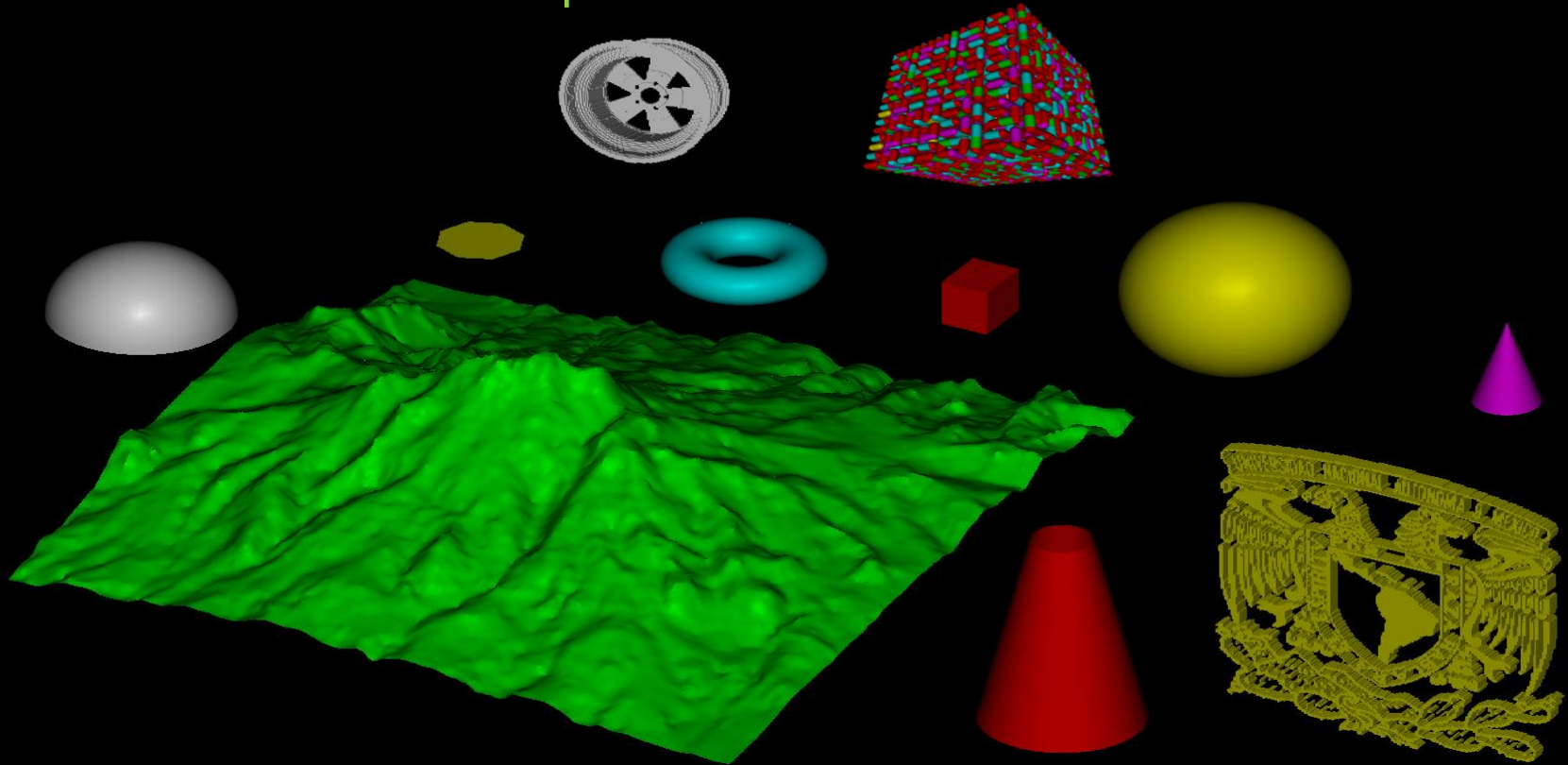


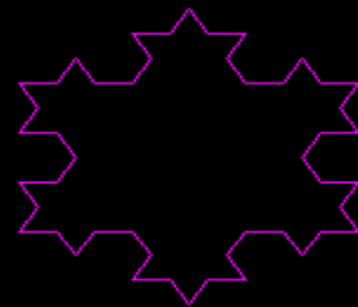
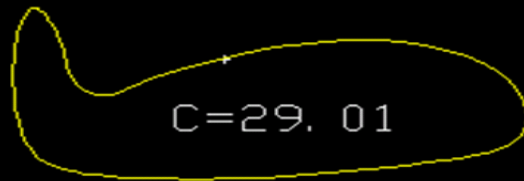
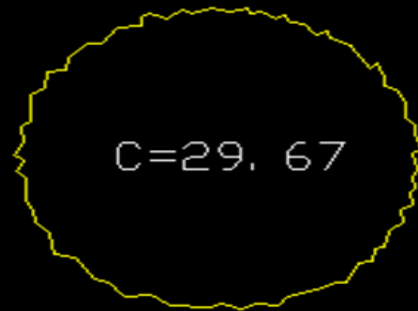
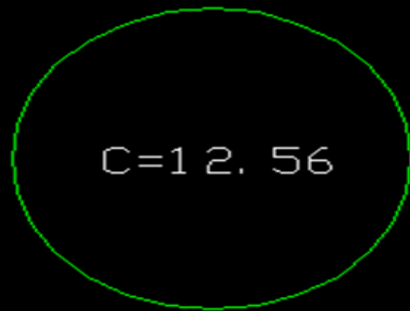
La Compacidad Discreta y sus Aplicaciones

por Ernesto Bribiesca

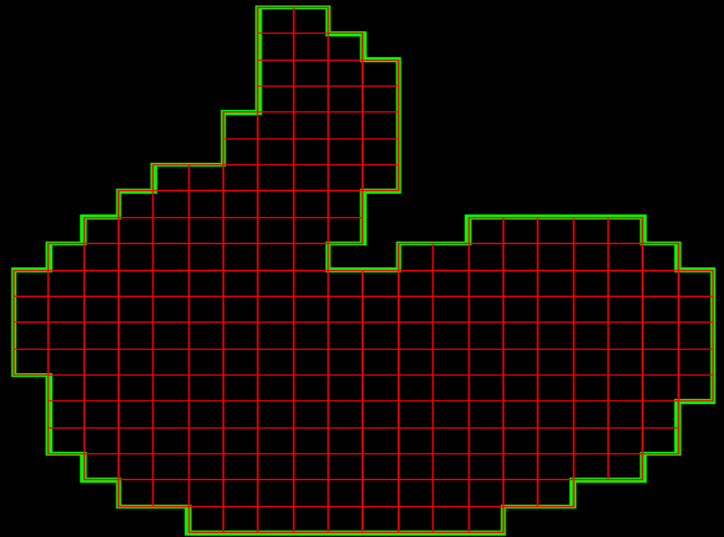
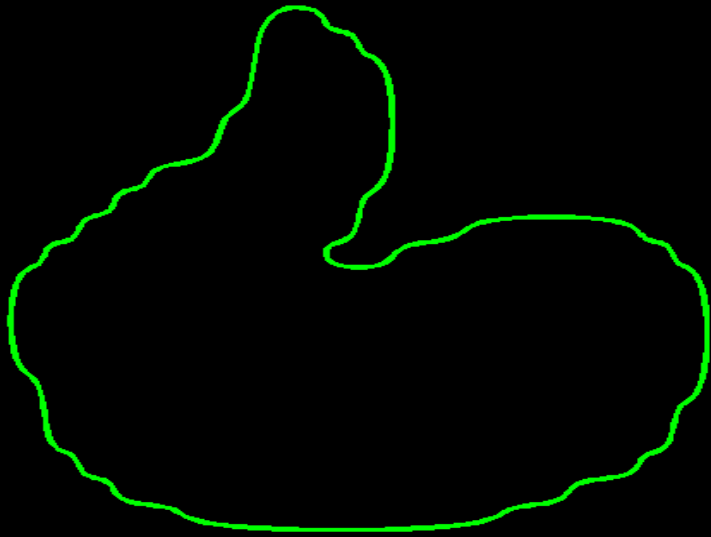


$$C=P^2/A$$

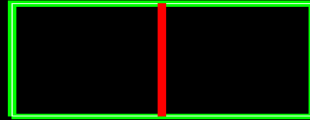
La Compacidad Clásica



Perímetro de Contacto



Perímetro de Contacto



$$P + 2P_c = 4n$$

Donde:

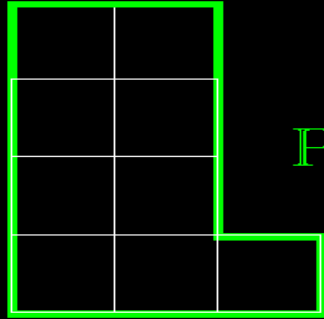
P es el perímetro

P_c es el perímetro de contacto

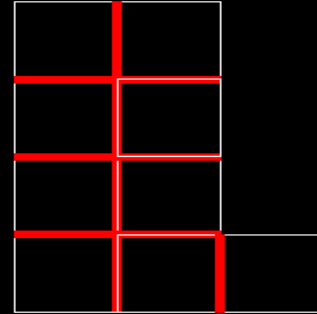
n es el número de píxeles

$$6 + 2(1) = 4(2)$$

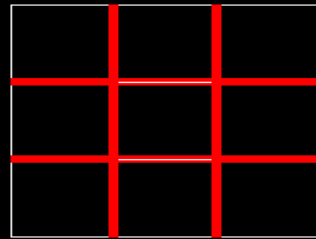
Compacidad Discreta Normalizada con valores del cero al uno



$$P + 2P_c = 4n$$



$$C_D = \frac{P_c}{P_{c_{\max}}}$$

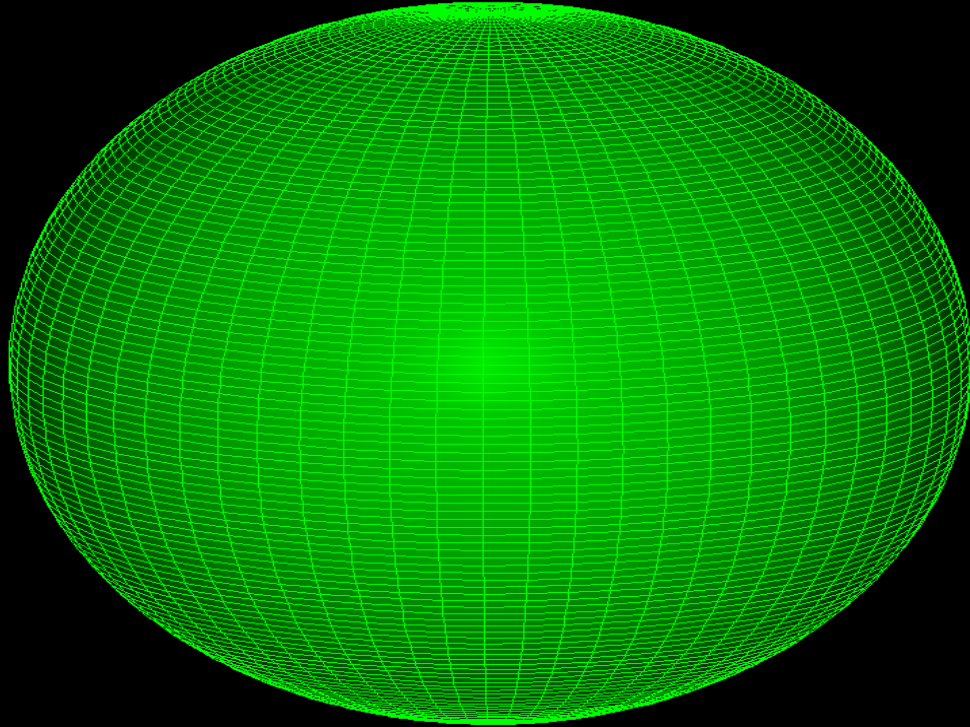


$$P_{c_{\max}} = \frac{4n - 4\sqrt{n}}{2}$$

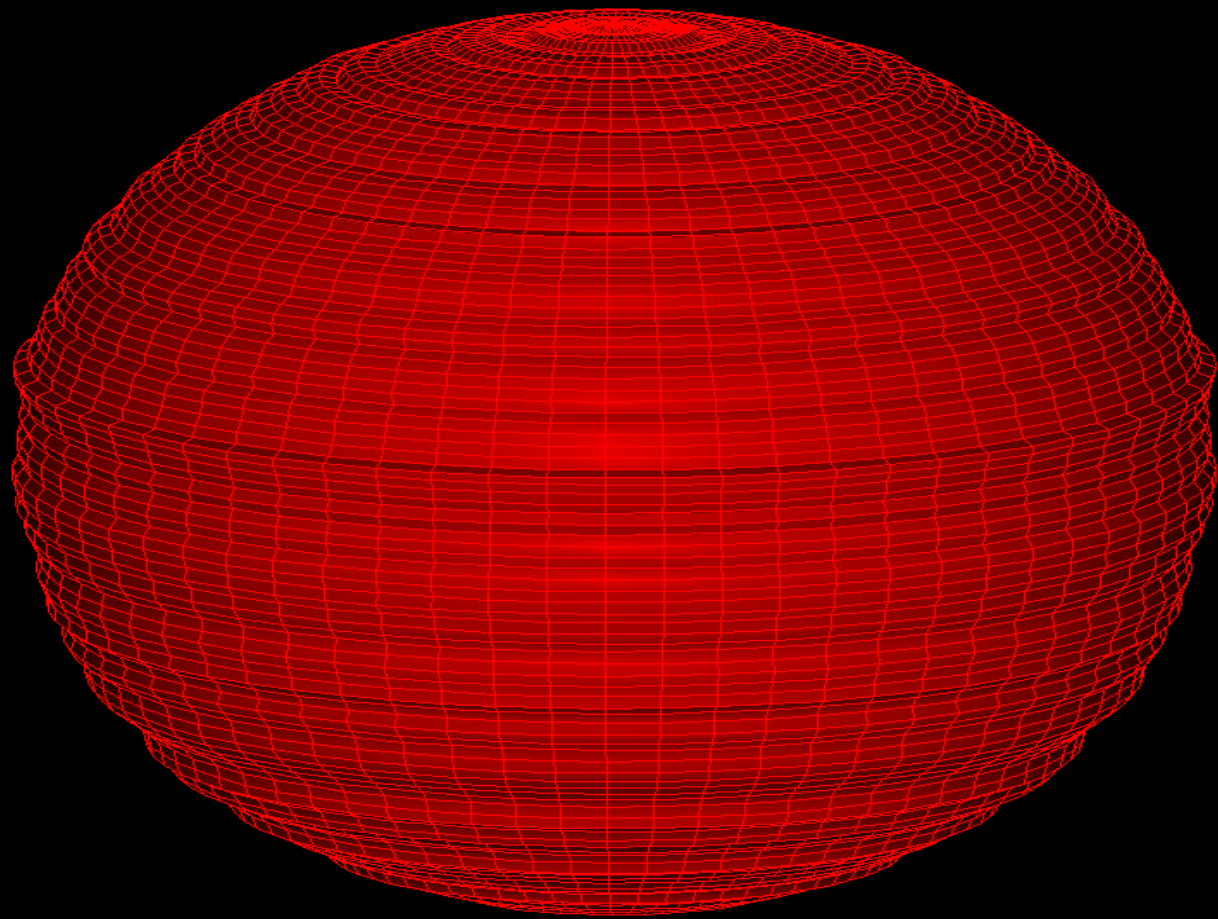
$$P_{c_{\max}} = 2(n - \sqrt{n})$$

$$C_d = \frac{n - P/4}{n - \sqrt{n}}$$

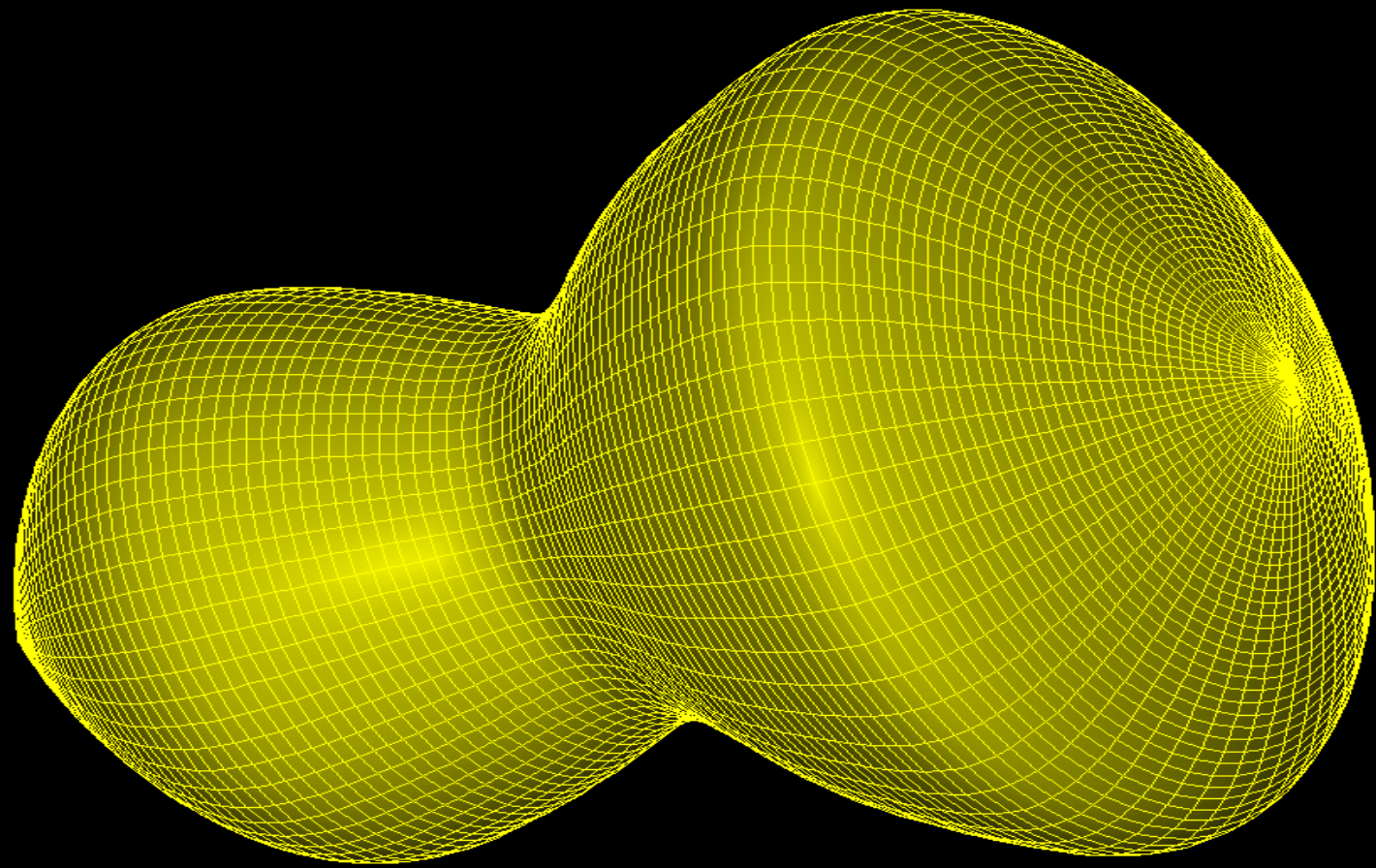
La Compacidad Clásica en tres dimensiones es igual a la superficie al cubo entre el volumen al cuadrado



$$C=113.0973$$



$C=159.8176$



C=123.4025

La Compacidad Discreta

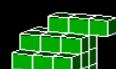
$$C_d = \frac{n - A/6}{n - (\sqrt[3]{n})^2}$$



$C_d = 1$



$C_d = 0.962962$



$C_d = 0.925925$



$C_d = 0.888888$



$C_d = 0.851851$



$C_d = 0.814814$



$C_d = 0.777777$



$C_d = 0.740740$



$C_d = 0.703703$



$C_d = 0.666666$



$C_d = 0.629629$



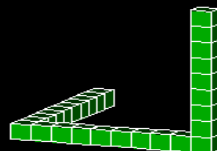
$C_d = 0.592592$



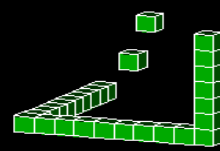
$C_d = 0.555555$



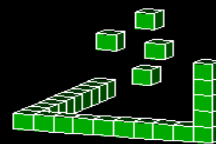
$C_d = 0.518518$



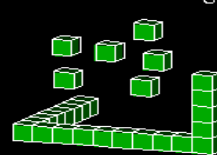
$C_d = 0.481481$



$C_d = 0.444444$



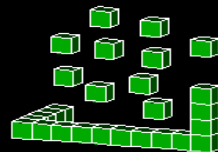
$C_d = 0.407407$



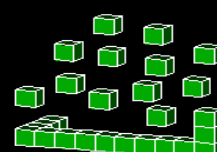
$C_d = 0.370370$



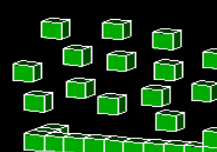
$C_d = 0.333333$



$C_d = 0.296296$



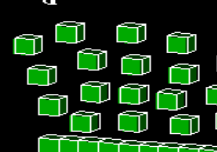
$C_d = 0.259259$



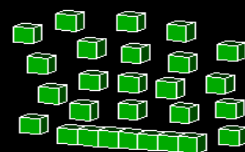
$C_d = 0.222222$



$C_d = 0.185185$



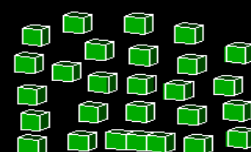
$C_d = 0.148148$



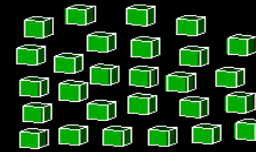
$C_d = 0.111111$



$C_d = 0.074074$

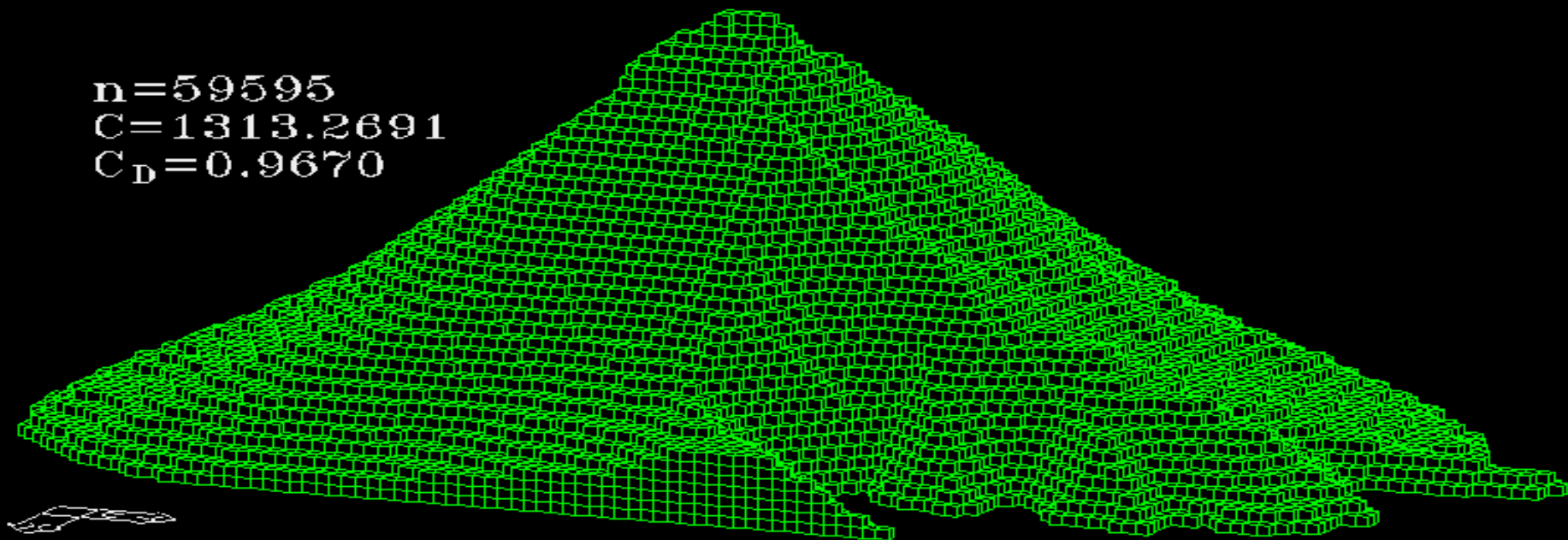


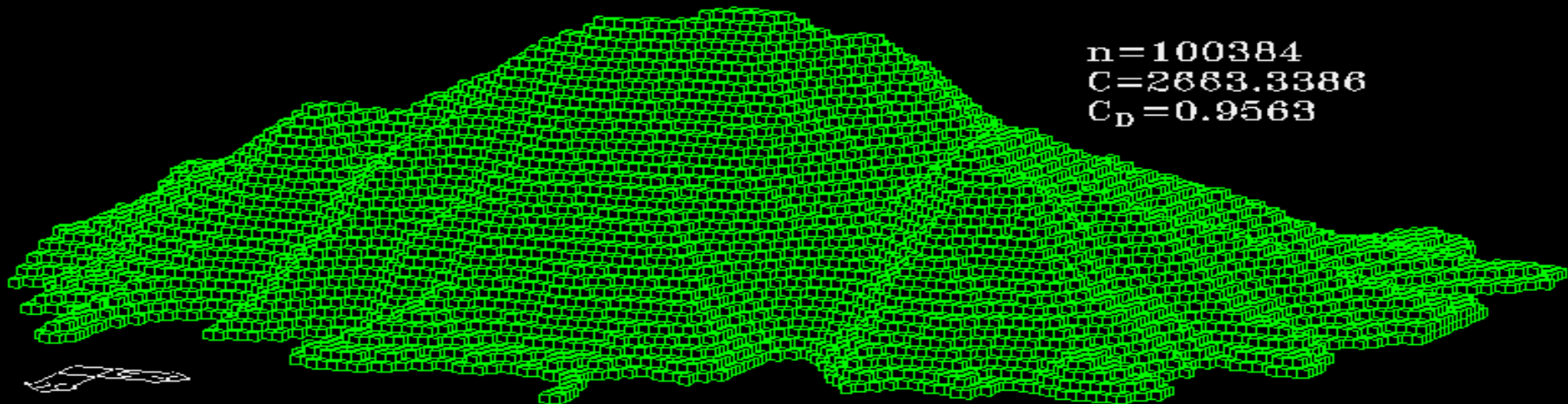
$C_d = 0.037037$

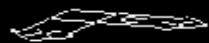
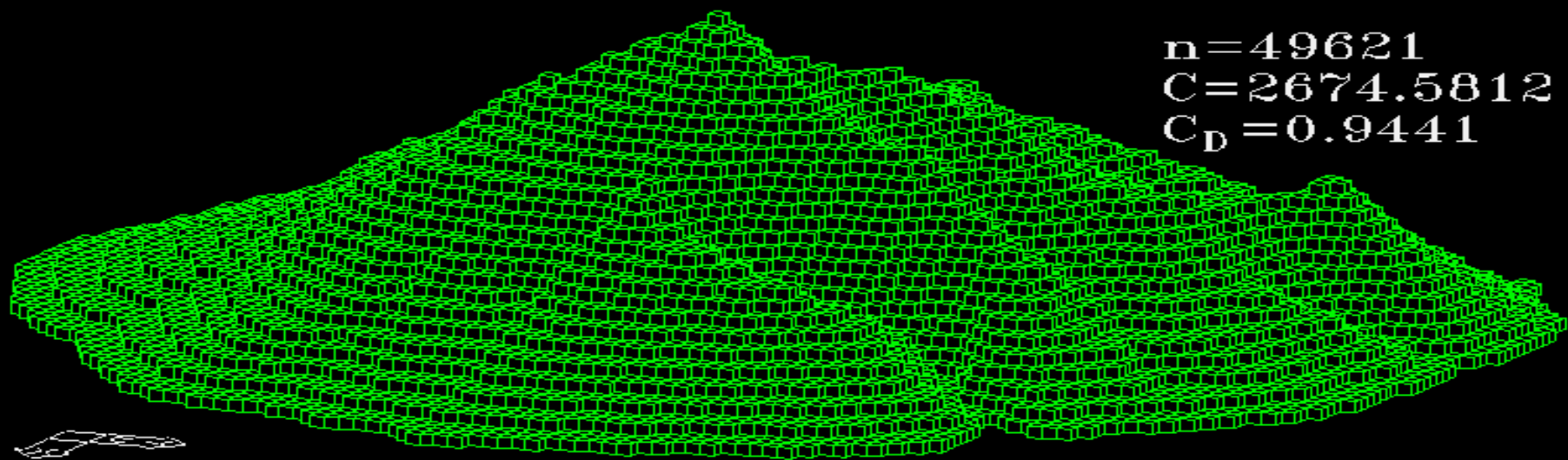


$C_d = 0.0$

$n=59595$
 $C=1313.2691$
 $C_D=0.9670$

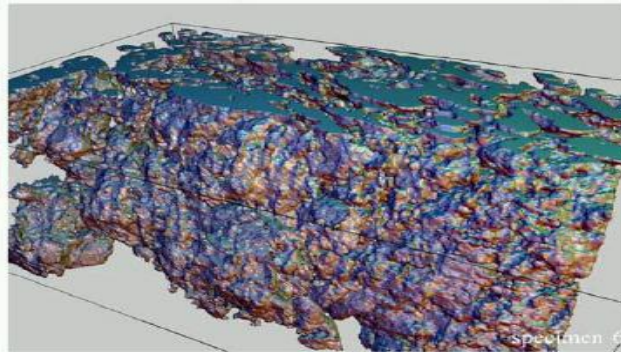
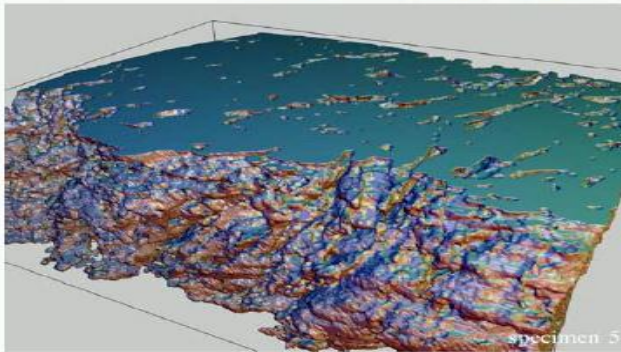
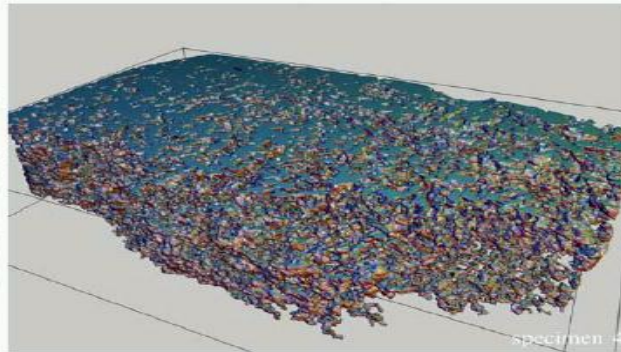
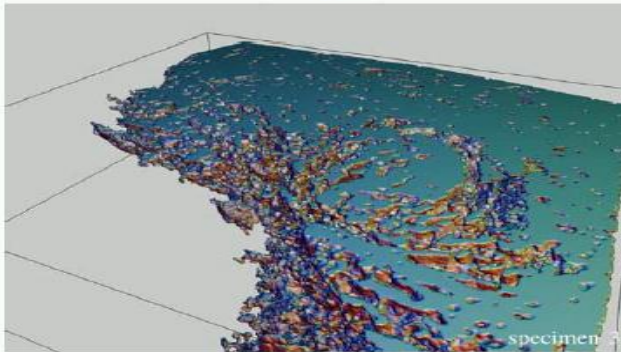
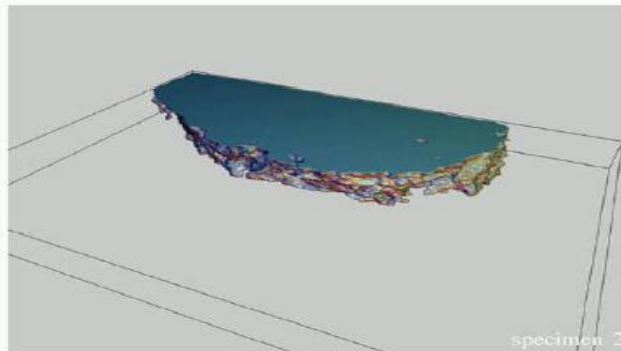
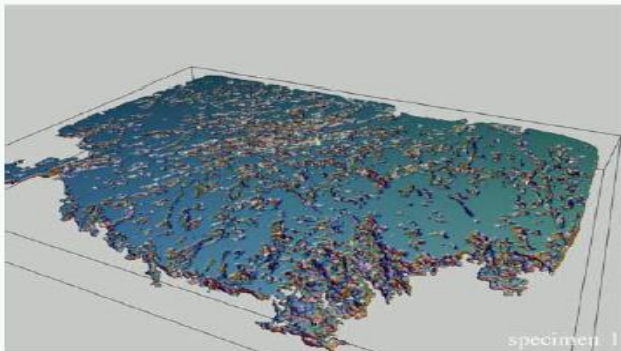


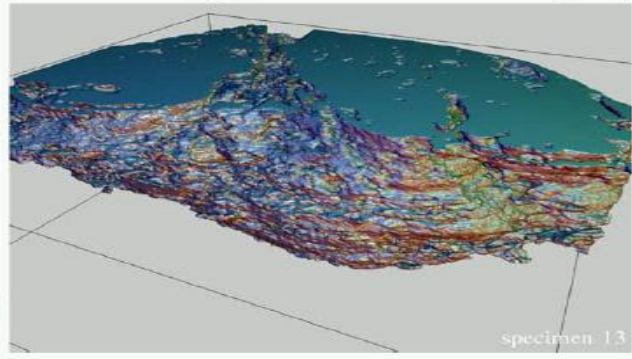
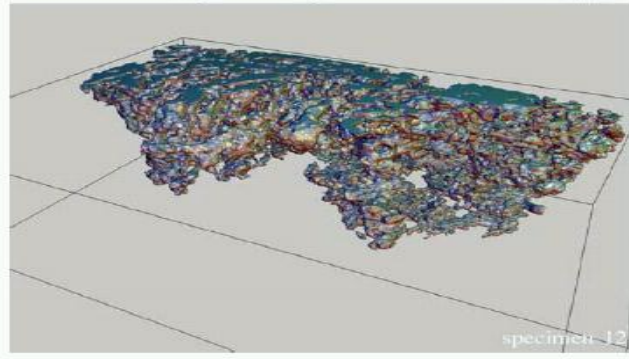
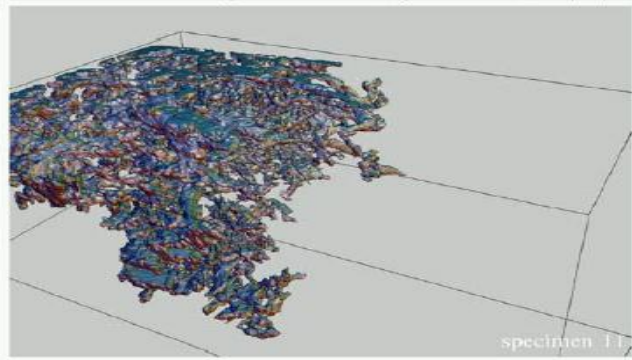
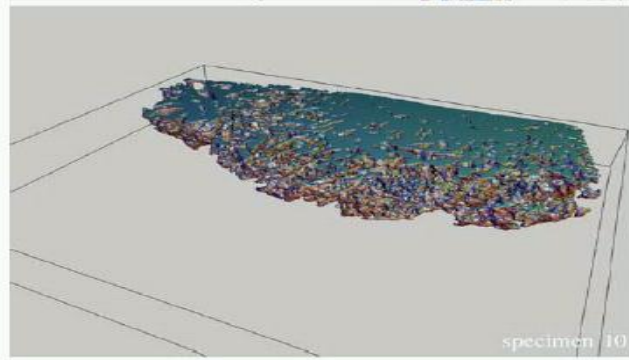
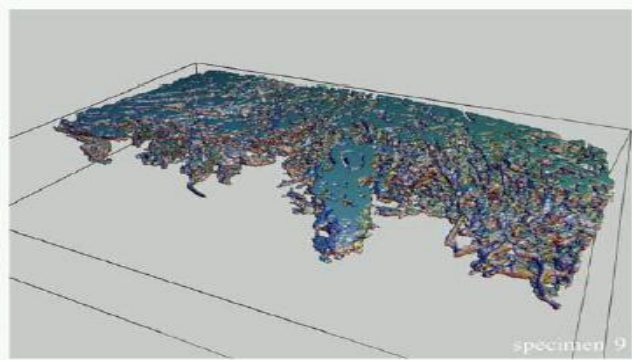
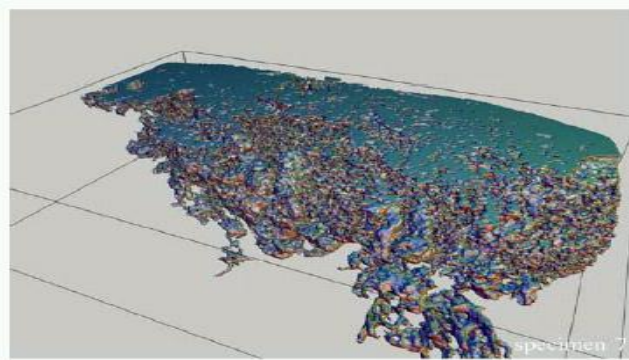




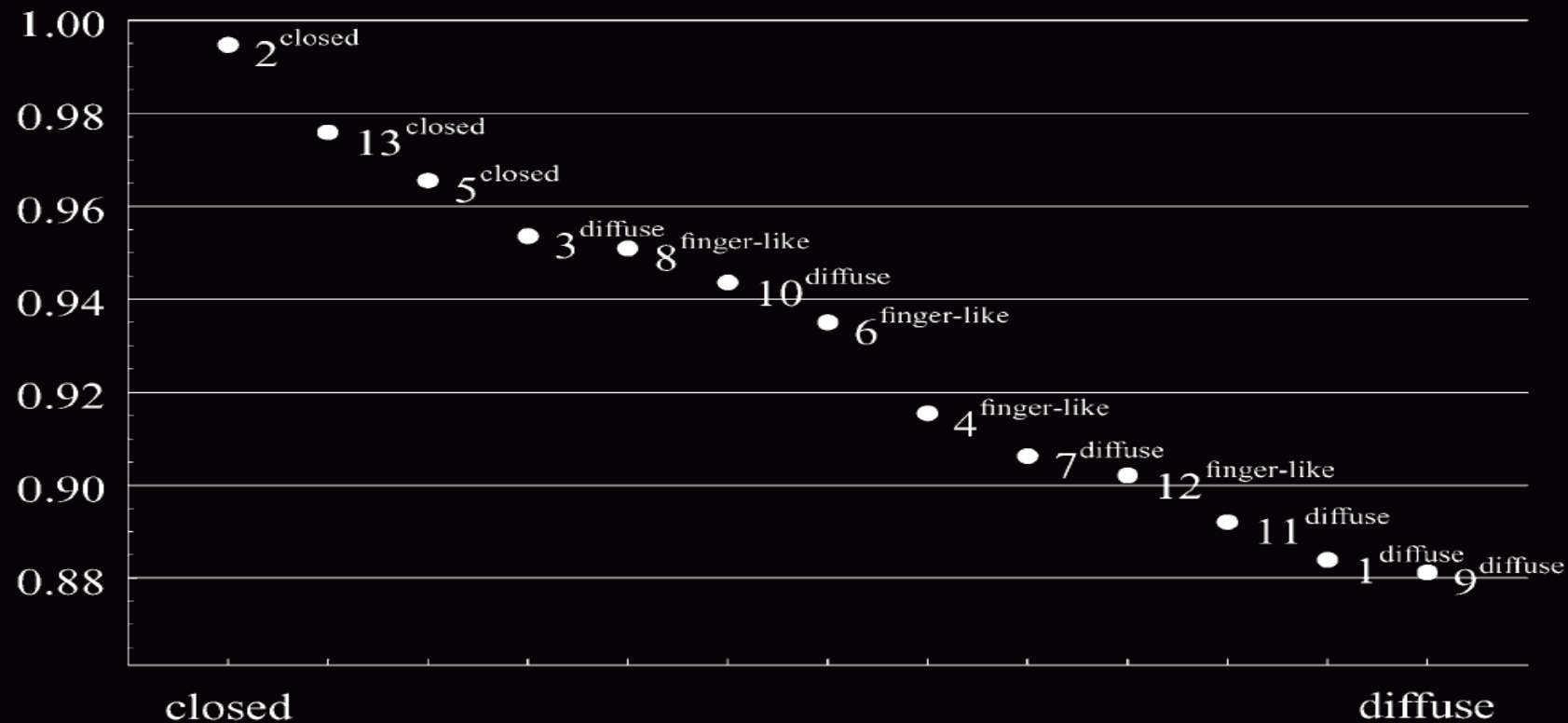
Three-Dimensional Reconstruction and Quantification of Cervical Carcinoma Invasion Fronts From Histological Serial Sections

Ulf-Dietrich Braumann, *Member, IEEE*, Jens-Peer Kuska, *Member, IEEE*, Jens Einkenkel, Lars-Christian Horn,
Markus Löffler, and Michael Höckel





Compactness



Shape Characterization of Extracted and Simulated Tumor Samples using Topological and Geometric Measures

Markus Rohrschneider*, Gerik Scheuermann*, Stefan Hoehme[†], Dirk Drasdo[‡]

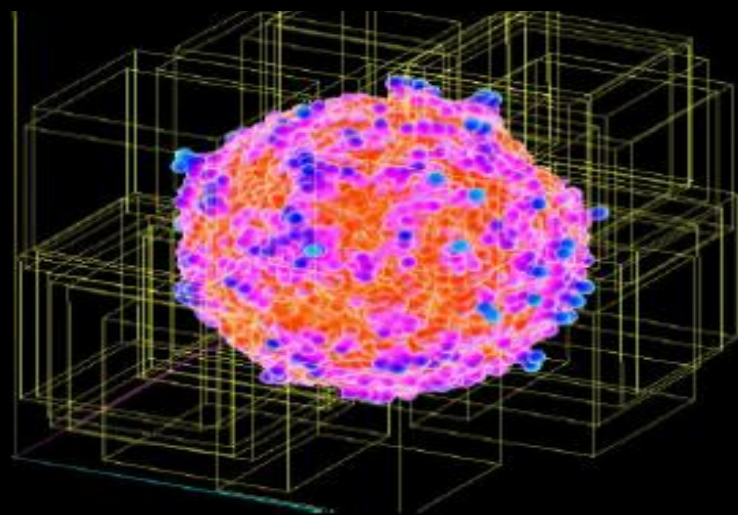


Fig. 2. Boundary of simulated tumor sample id120_mot1. The bounding boxes of 30 random samples are shown.

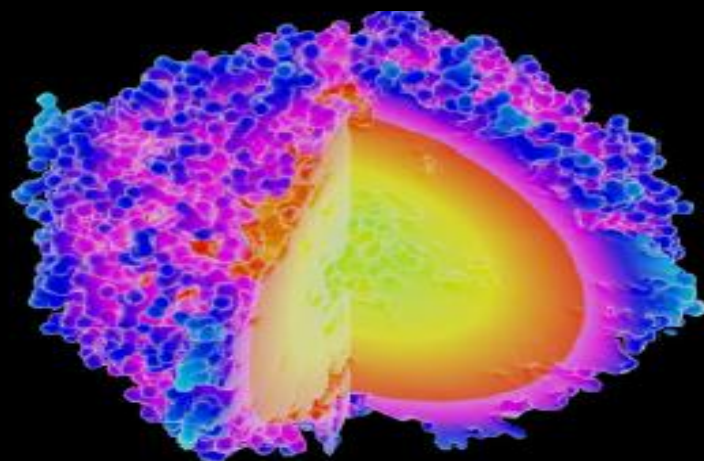


Fig. 3. Boundary of simulated tumor sample id100_mot30. The inner surface due to the necrotic core is shown.

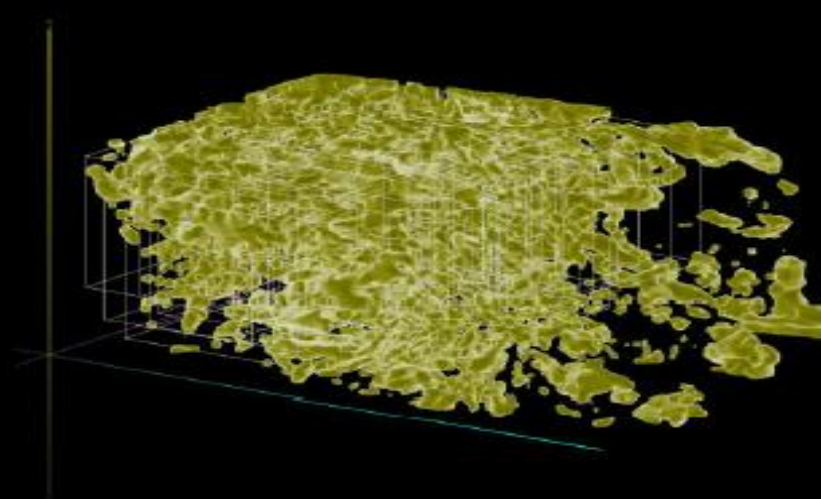


Fig. 6. Boundary of cervix carcinoma specimen 11 (diffuse tumor type). The bounding boxes of 30 random samples containing approx. 13% foreground voxels, are shown.

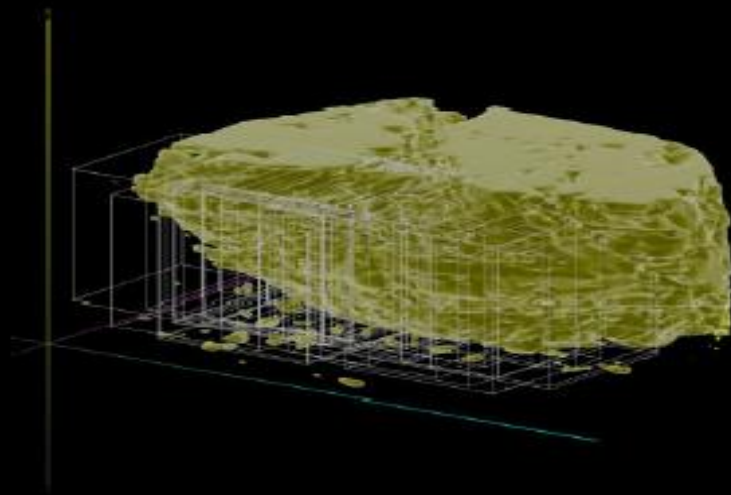
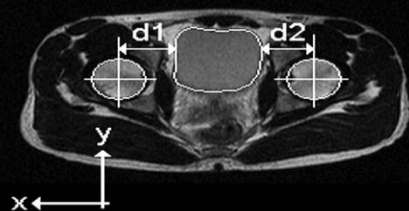


Fig. 7. Boundary of cervix carcinoma specimen 13 (compact tumor type). The bounding boxes of 30 random samples containing approx. 27% foreground voxels, are shown. The object contains many small cavities not visible in this representation.

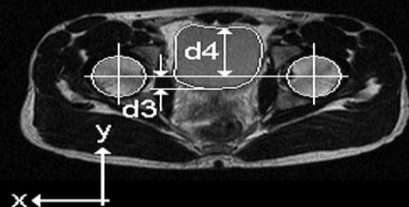
TABLE I
RESULTS OF THE COMPACTNESS COMPUTATION AND TOPOLOGICAL ANALYSIS.

Dataset	Average of 30 Samples						Complete Object					
	Area	Volume	Compactness		β_0	β_1	β_2	Compactness		β_0	β_1	β_2
1. asym_mot1	15618.5	198596	0.85	0.9876	1.00	11.01	0.00	6.67	0.9929	1	176	0
2. asym_mot20	28815.9	192697	5.70	0.9717	1.07	33.03	0.00	60.41	0.9773	2	624	0
3. id095_mot1	27406.1	199477	4.57	0.9750	1.43	40.08	0.00	36.86	0.9809	6	468	0
4. id100_mot1	24598.7	190235	3.64	0.9766	1.33	35.39	0.00	28.55	0.9835	5	475	0
5. id100_mot10	33698.5	189090	9.46	0.9662	1.37	43.27	0.00	74.92	0.9723	5	475	0
6. id100_mot20	34658.0	185317	10.72	0.9647	1.07	40.03	0.00	100.39	0.9688	2	447	0
7. id100_mot30	46132.2	184418	25.53	0.9507	1.00	288.88	1.59	254.65	0.9564	1	2867	17
8. id100_mot1_adh	32872.8	195883	8.19	0.9680	1.03	101.50	0.00	66.16	0.9768	2	1301	0
9. id120_mot1	18581.0	200296	1.41	0.9844	1.00	24.27	0.00	10.76	0.9903	1	316	0
10. cerv11 (diffuse type)	89644.6	134022	354.66	0.8800	110.43	101.11	0.75	3337.46	0.8742	934	618	5
11. cerv13 (compact type)	55138.9	273947	19.75	0.9601	50.37	106.11	36.41	114.21	0.9758	334	1284	755



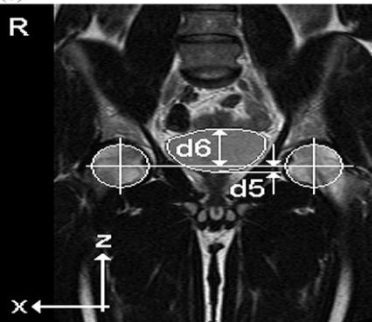
(b)

R



(c)

R



Effect of Bladder Volume, Gender and Body Position on the Shape and Position of the Urinary Bladder

Niels Kristian Kristiansen,^{1,2} Steffen Ringgaard,³ Hans Nygaard^{2,4} and Jens Christian Djurhuus¹

From the ¹Clinical Institute, Aarhus University Hospital, Skejby Sygehus, Aarhus, Denmark, ²Department of Electronics and Information Technology, University College of Aarhus, Aarhus, Denmark and the ³MR-Center and ⁴Department of Cardiothoracic & Vascular Surgery, Aarhus University Hospital, Skejby Sygehus, Aarhus, Denmark

(Submitted December 15, 2003. Accepted for publication May 19, 2004)

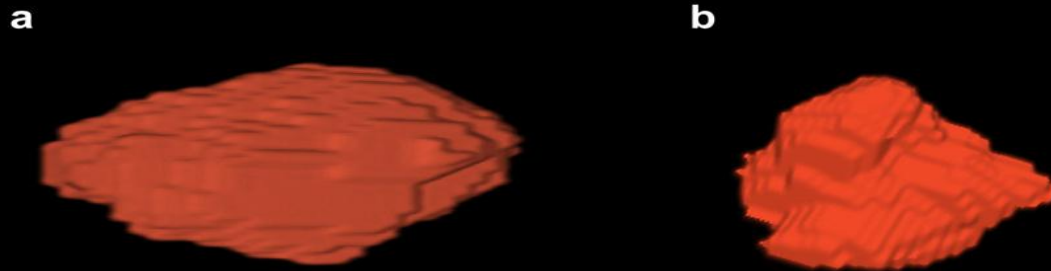
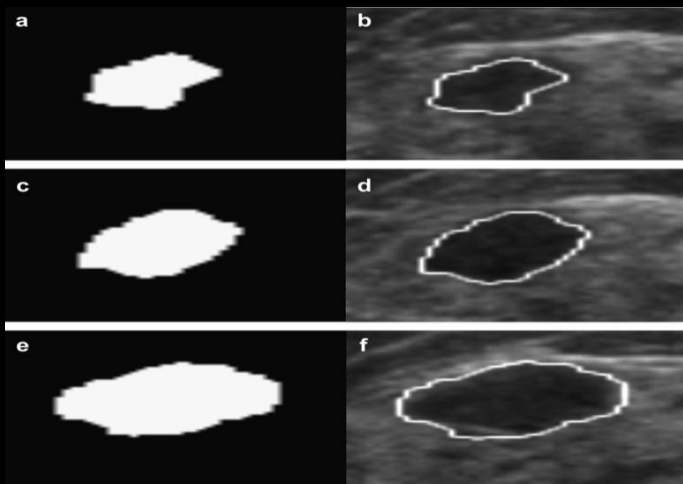
Scand J Urol Nephrol 38: 462–468, 2004

● *Original Contribution*

COMPUTER-AIDED DIAGNOSIS FOR THE CLASSIFICATION OF BREAST MASSES IN AUTOMATED WHOLE BREAST ULTRASOUND IMAGES

WOO KYUNG MOON,^{*} YI-WEI SHEN,[†] CHIUN-SHENG HUANG,[‡] LI-REN CHIANG,[†]
and RUEY-FENG CHANG^{†§}

^{*}Department of Diagnostic Radiology, Seoul National University Hospital, Korea; [†]Department of Computer Science and Information Engineering, National Taiwan University, Taipei, Taiwan; [‡]Department of Surgery, National Taiwan University Hospital, Taipei, Taiwan; and [§]Graduate Institute of Biomedical Electronics and Bioinformatics, National Taiwan University, Taipei, Taiwan



Remote Sensing Letters

Characterization of individual tree crowns using three-dimensional shape signatures derived from LiDAR data

PINLIANG DONG*

Department of Geography, University of North Texas, Denton, TX 76203, USA

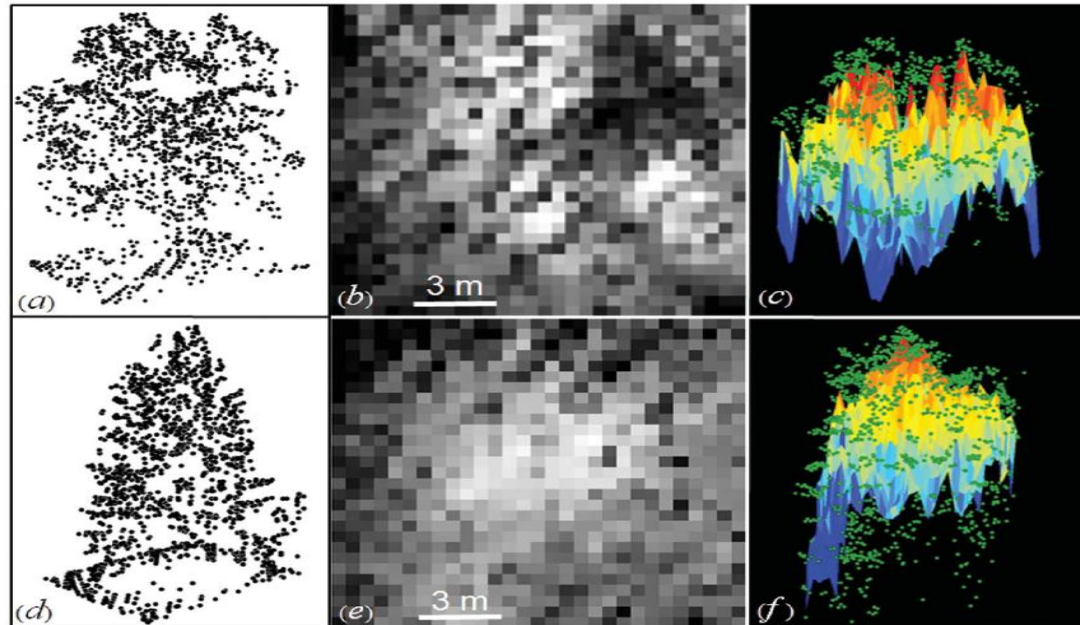


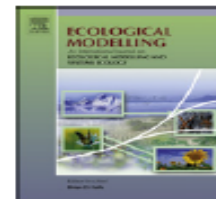
Figure 2. LiDAR data point cloud perspective views, DSM, and combined perspective display of point clouds and DSM. Top row: oak; bottom row: Douglas fir.



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Deriving state-and-transition models from an image series of grassland pattern dynamics

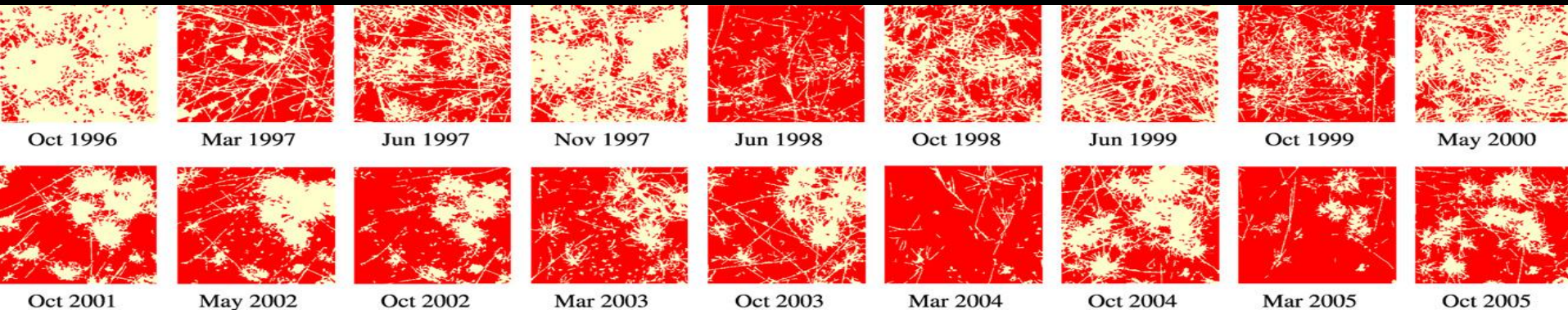
Rohan J. Sadler^{a,b,c,*}, Martin Hazelton^d, Matthias M. Boer^{a,c}, Pauline F. Grierson^{a,c}

^a *Ecosystems Research Group, School of Plant Biology, The University of Western Australia, 35 Stirling Highway, Crawley, 6009, Australia*

^b *School of Mathematics and Statistics, The University of Western Australia, 35 Stirling Highway, Crawley, 6009, Australia*

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A real-time assessment of the ship design complexity

J.-D. Caprace*, P. Rigo

ANAST – University of Liège, Chemin des chevreuils, 1, 4000 Liège, Belgium

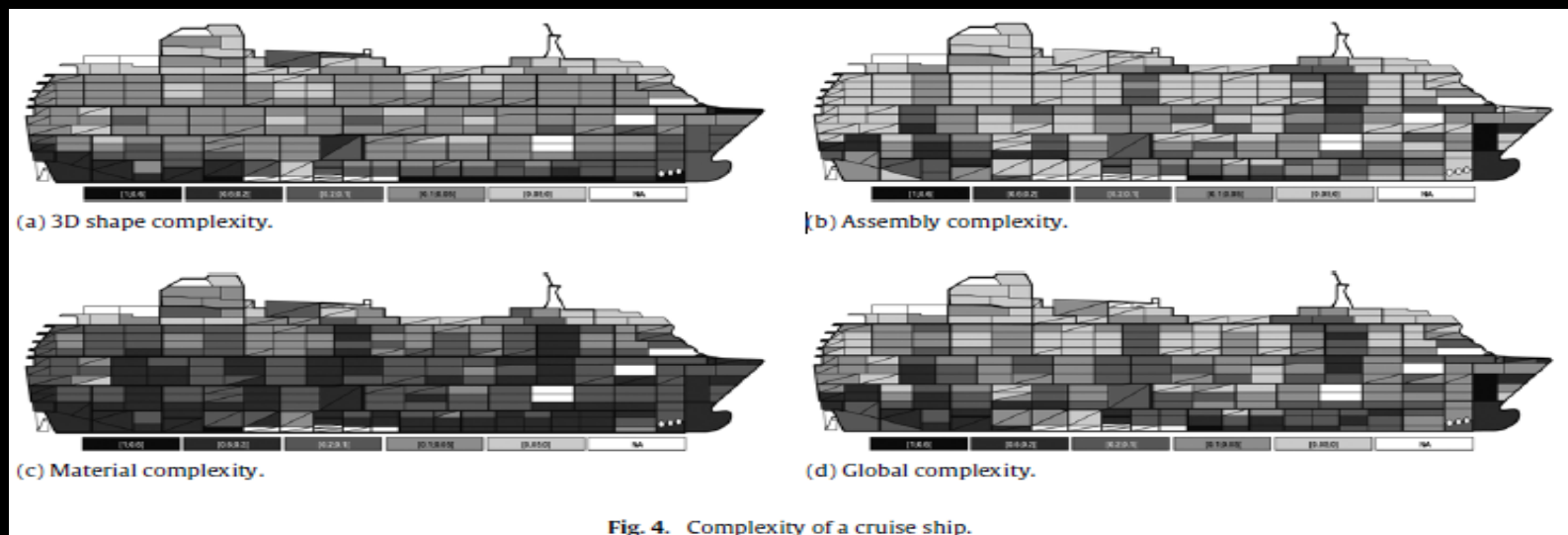


Fig. 4. Complexity of a cruise ship.

Airway segmentation and analysis for the study of mouse models of lung disease using micro-CT

X Artaechevarria¹, D Pérez-Martín¹, M Ceresa¹, G de Biurrun²,
D Blanco², L M Montuenga², B van Ginneken³, C Ortiz-de-Solorzano¹
and A Muñoz-Barrutia¹

7018

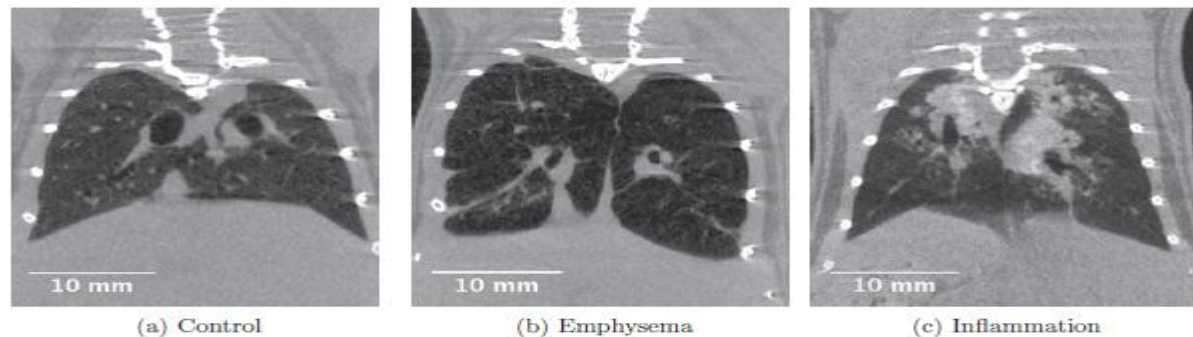
X Artaechevarria *et al*

Figure 2. Sample coronal micro-CT slices of control, emphysematous and inflamed mice lungs. Differences in lung texture and shape are clearly visible.

Then, the discrete compactness (C) is computed, as defined by Bribiesca (2008) for a solid volume of n voxels in 3D:

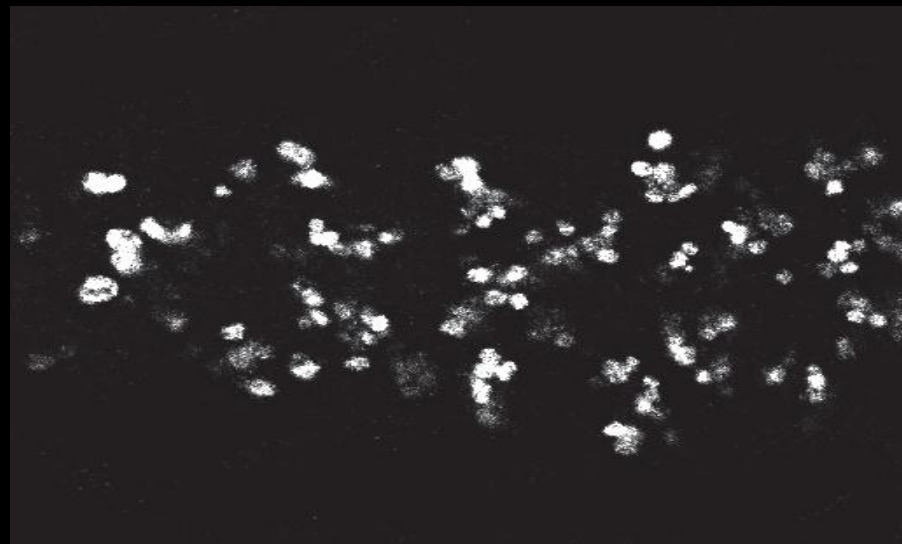
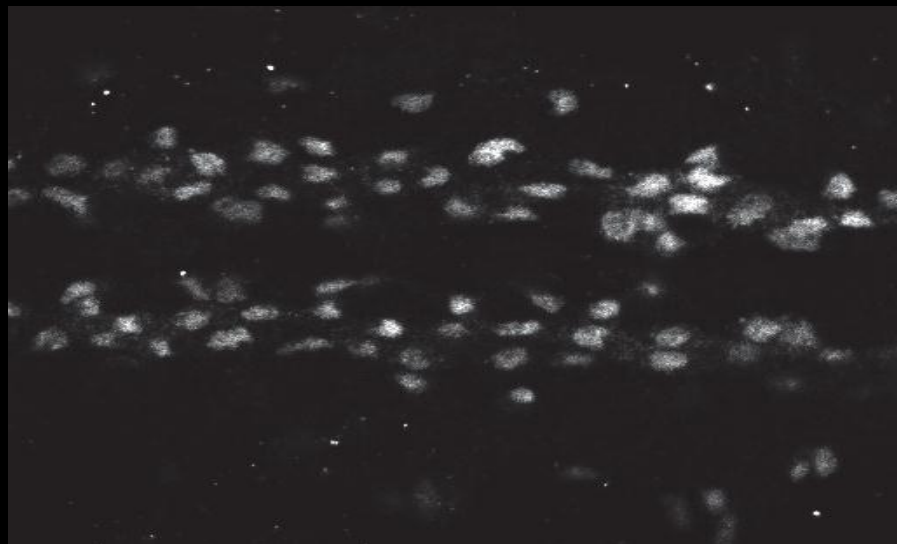
$$C = \frac{n - A/6}{n - (\sqrt[3]{n})^2}, \quad (4)$$

Image Processing Methods for Automatic Cell Counting In Vivo or In Situ Using 3D Confocal Microscopy

Manuel G. Forero¹ and Alicia Hidalgo²

¹Cardiff University,

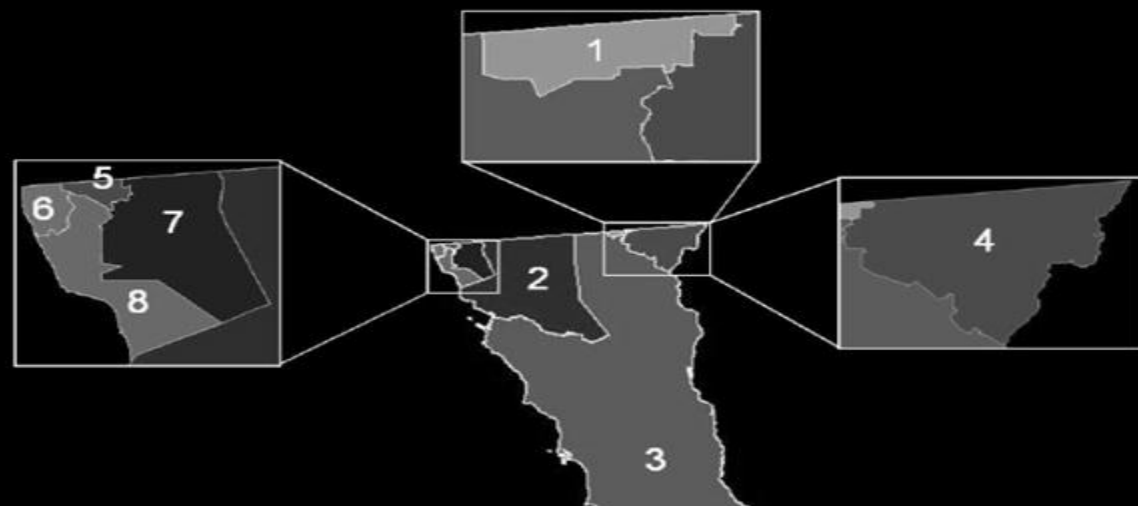
²University of Birmingham,
United Kingdom



*Compacidad en celdas aplicada al diseño
de zonas electorales*

ERIC RINCÓN GARCÍA
MIGUEL ANGEL GUTIÉRREZ ANDRADE ¹

Zonas generadas al aplicar la compacidad por celdas

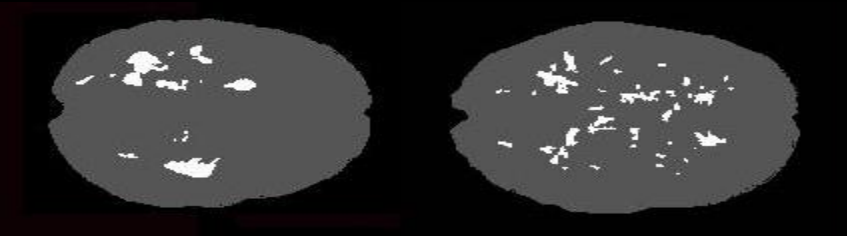


Spatial Dispersion of Lesions as a Surrogate Biomarker for Disability in Multiple Sclerosis

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Ghassan Hamarneh
Medical Image Analysis Lab
Simon Fraser University
Vancouver, BC, Canada, V5A 1S6
hamarneh@sfu.ca



2.2. Lesion dispersion measures

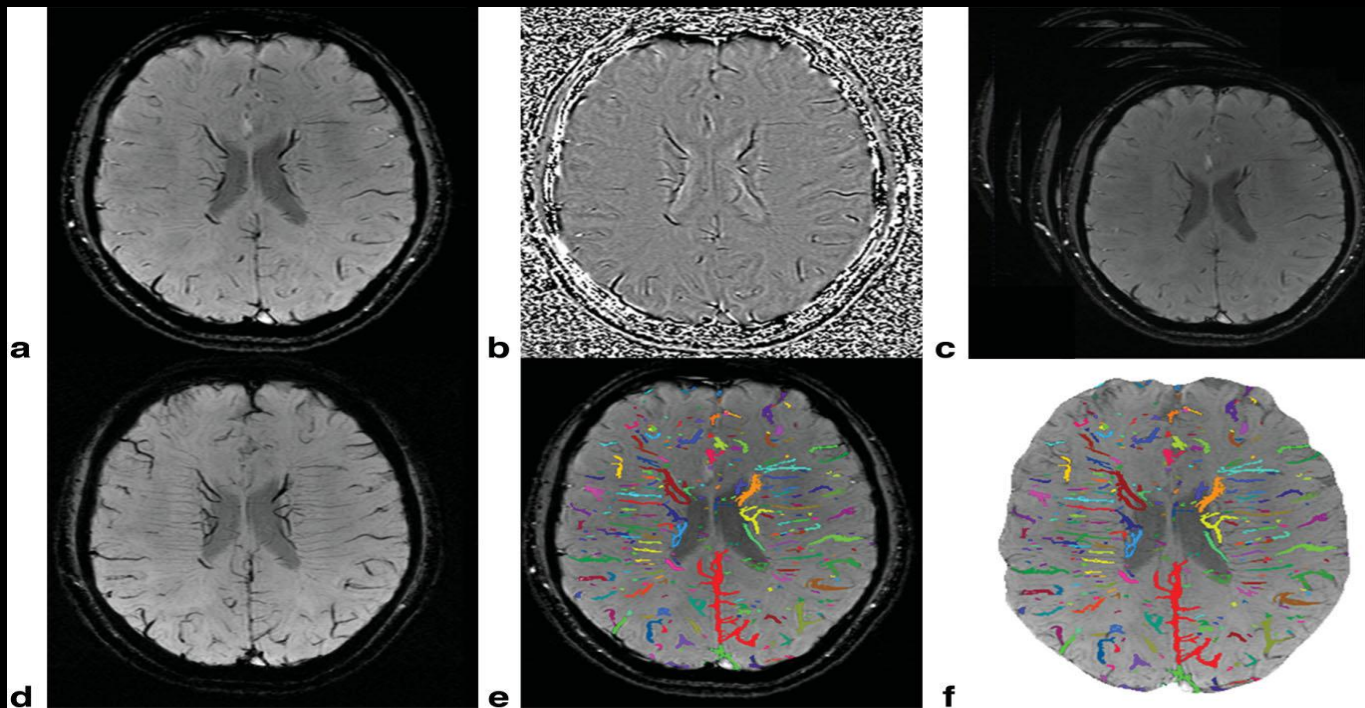
Compactness. Developed by Bribiesca [4] to quantify the connectedness of shapes composed of cubic voxels, compactness is mathematically defined as follows:

$$C = \frac{n - A/6}{n - (\sqrt[3]{n})^2} \quad (1)$$

Clinical Note

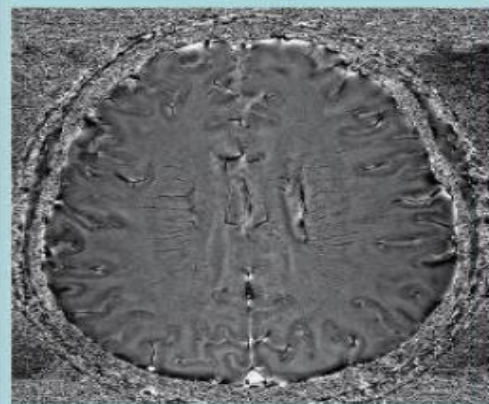
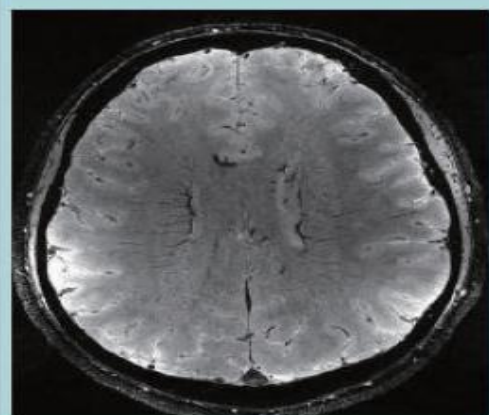
Diminished Visibility of Cerebral Venous Vasculature in Multiple Sclerosis by Susceptibility-Weighted Imaging at 3.0 Tesla

Yulin Ge, MD,^{1*} Vahe M. Zohrabian, MD,¹ Etin-Osa Osa, MD,¹ Jian Xu, PhD,²
Hina Jaggi, RT,¹ Joseph Herbert, MD,³ E. Mark Haacke, PhD,⁴ and
Robert I. Grossman, MD¹

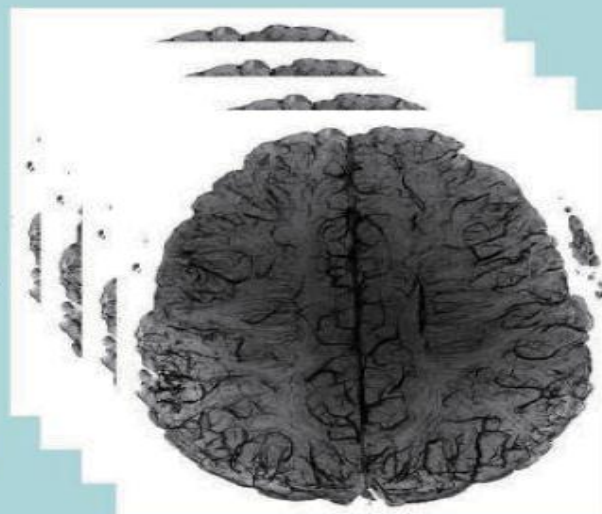


Three-dimensional high resolution venography using susceptibility weighted imaging at 7T

Yulin Ge^{1*}, Samuel Barnes², Samantha Heller¹, Daniel K. Sodickson¹, Lin Tang¹, E. Mark Haacke², Jianping Dai³, Robert I. Grossman¹



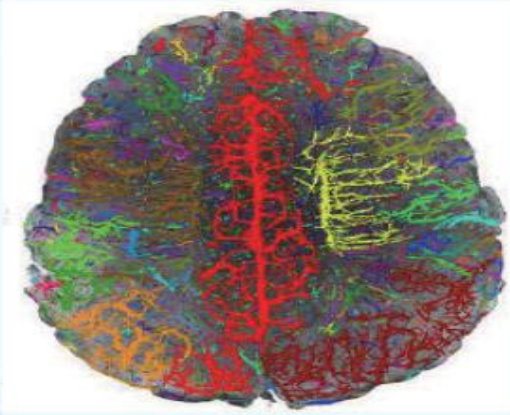
A



C

B

D



E

Select Objects

Object #	Size [# Vx.]	R Anisotro.	Compactn.	Product
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<input checked="" type="checkbox"/> Object 1	21829	0.801237	0.521239	0.038988
<input checked="" type="checkbox"/> Object 2	19095	0.546474	0.505510	0.032400
<input checked="" type="checkbox"/> Object 3	9799	0.579140	0.478457	0.066541
<input checked="" type="checkbox"/> Object 4	8991	0.538915	0.466431	0.068505
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<input checked="" type="checkbox"/> Object 6	4182	0.955450	0.494422	0.076412
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<input checked="" type="checkbox"/> Object 8	2950	0.710544	0.536370	0.050588
<input checked="" type="checkbox"/> Object 9	1726	0.528317	0.521244	0.052524
<input checked="" type="checkbox"/> Object 10	1160	0.623261	0.571984	0.042243
<input checked="" type="checkbox"/> Object 11	941	0.869184	0.658514	0.044648
<input checked="" type="checkbox"/> Object 12	829	0.707759	0.588591	0.077674
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<input checked="" type="checkbox"/> Object 17	559	0.638948	0.486622	0.059642

Select/Unselect All Hide Objects When Clicked

Semiautomated detection of cerebral microbleeds in magnetic resonance images[☆]

Samuel R.S. Bames^{a,b,c,*}, E. Mark Haacke^{a,b,c,d}, Muhammad Ayaz^c, Alexander S. Boikov^{c,f},
Wolff Kirsch^g, Dan Kido^b

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^bDepartment of Radiology, Loma Linda University Medical Center, Loma Linda, CA 92350, USA

^cThe Magnetic Resonance Imaging Institute for Biomedical Research, Detroit, MI 48202, USA

^dDepartment of Radiology, Wayne State University, Detroit, MI 48201, USA

^eHemorrhagic Stroke Research Center, Department of Neurology, Massachusetts General Hospital, Harvard Medical School, Boston, MA 02114, USA

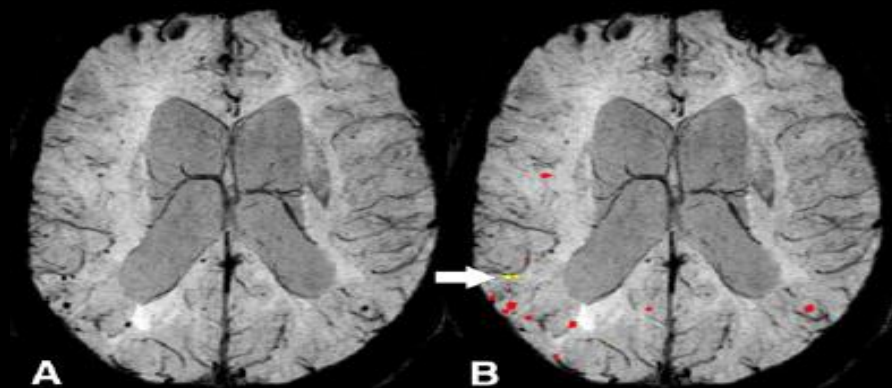
^fWayne State University School of Medicine, Detroit, MI 48201-1998, USA

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Compactness is approximately a ratio of surface area to volume, with spheres having the highest theoretical compactness. Cerebral microbleeds, therefore, have a very high compactness and veins have a very low compactness. Since we are dealing with discrete voxels, compactness is calculated by counting adjacent voxel faces:

$$C_d = \frac{A_C - A_{Cmin}}{A_{Cmax} - A_{Cmin}} \quad (3)$$



A Voxel-Based Measure of Discrete Compactness for Brain Imaging

E. Bribiesca^{1,2}, J. R. Jimenez¹, V. Medina¹, R. Valdes¹, O. Yanez¹

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**PATTERN
RECOGNITION**

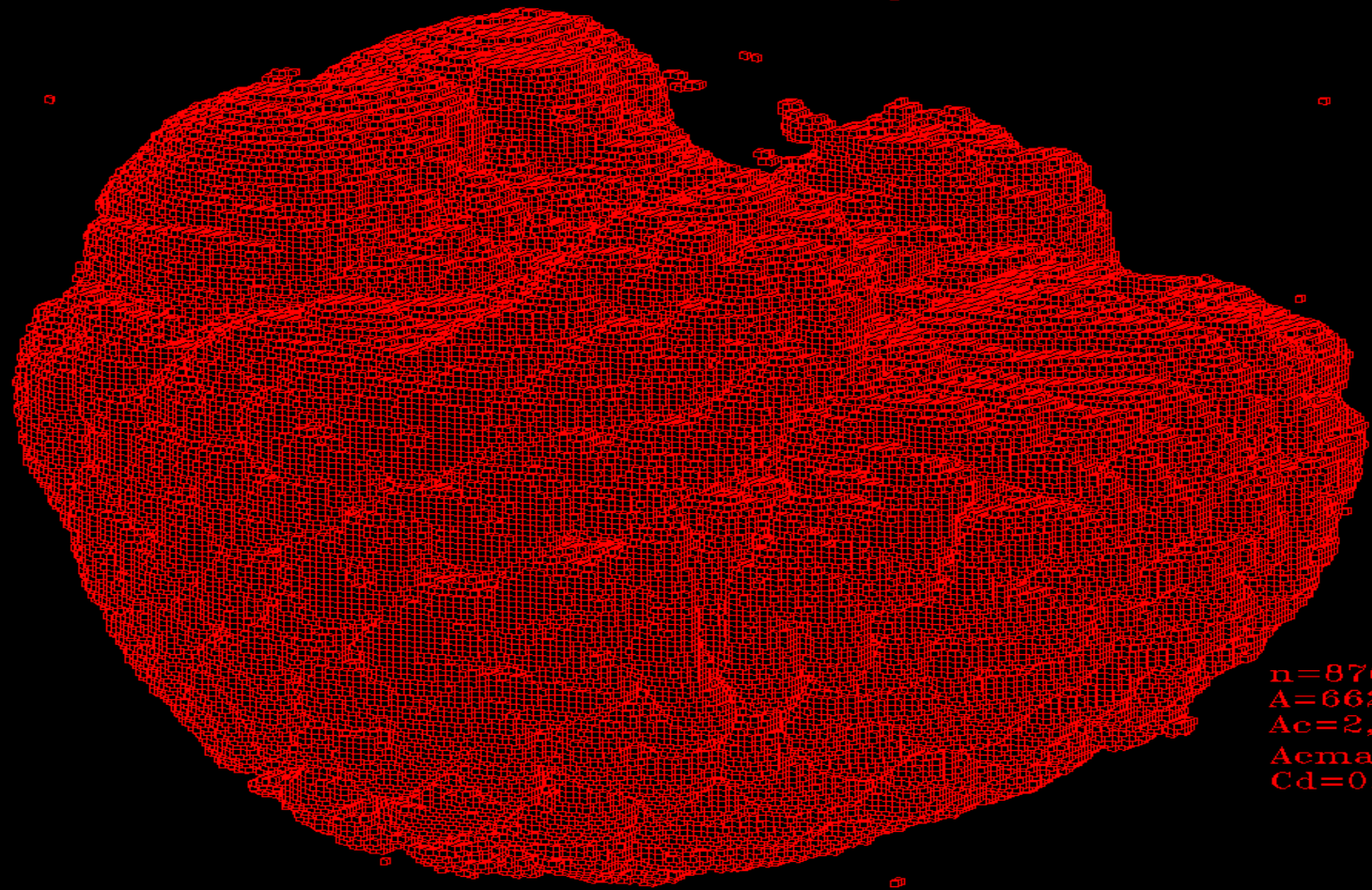
THE JOURNAL OF THE PATTERN RECOGNITION SOCIETY

www.elsevier.com/locate/pr

An easy measure of compactness for 2D and 3D shapes

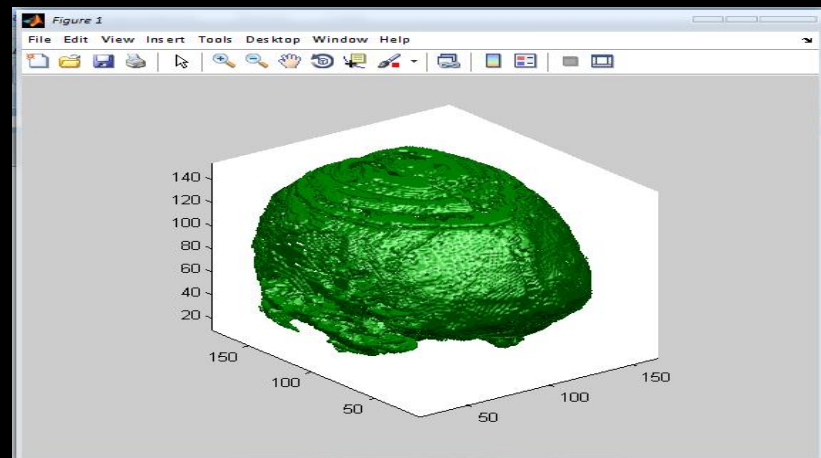
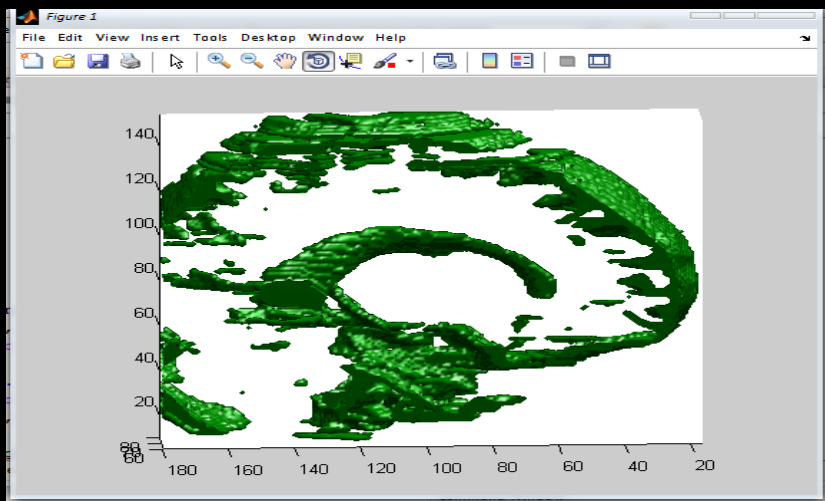
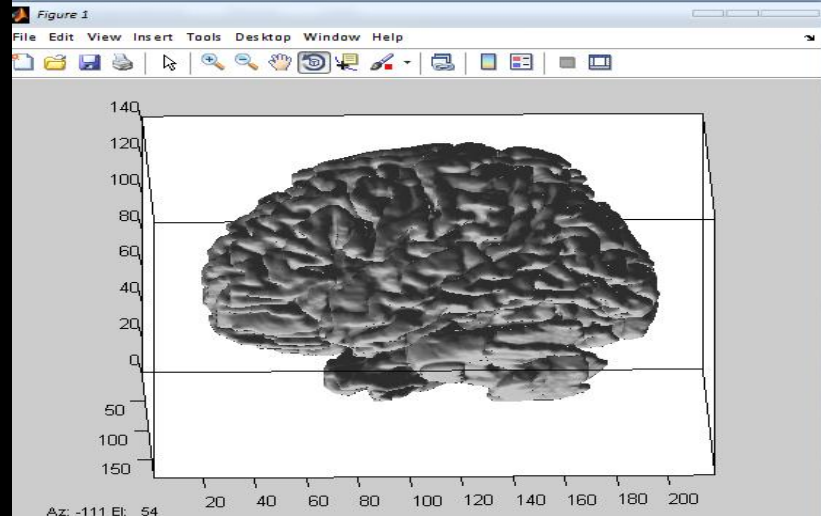
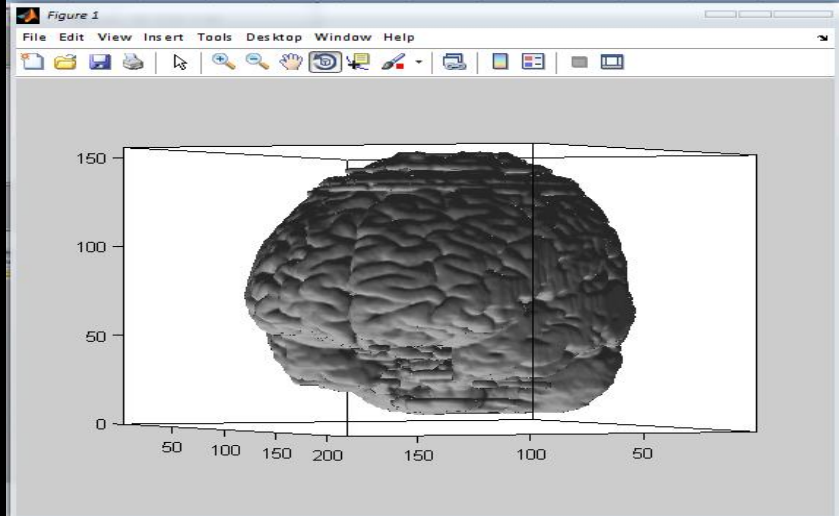
Ernesto Bribiesca*

Departamento de Ciencias de la Computación, Instituto de Investigaciones en Matemáticas Aplicadas y en Sistemas, Universidad Nacional Autónoma de México, Apdo. Postal 20-726, D.F., 01000, México



n=876,467 voxels
A=662,672
Ac=2,298,065
Acmax=2,601,927
Cd=0.8238

UAM-UNAM



Resultados preliminares

Dra. Verónica Medina, Jorge Luis Pérez

Estructura	Volumen	Área envolvente	Área de contacto	Compacidad discreta	Componentes conexas	Agujeros	Numero de Euler
Materia gris	944,446	685,600	2,490,538	0.880634163	9	19	-10
Materia blanca	678,695	385,866	1,843,152	0.915662579	9	2	7
L.Cefalorraquídeo	326,677	404,404	777,829	0.805371788	57	2	55

Evaluation of the invasion front pattern of squamous cell cervical carcinoma by measuring classical and discrete compactness

Jens Eibenkel^{a,*}, Ulf-Dietrich Braumann^b, Lars-Christian Horn^c, Nadine Pannicke^a,
Jens-Peer Kuska^b, Alexander Schütz^c, Bettina Hentschel^d, Michael Höckel^a

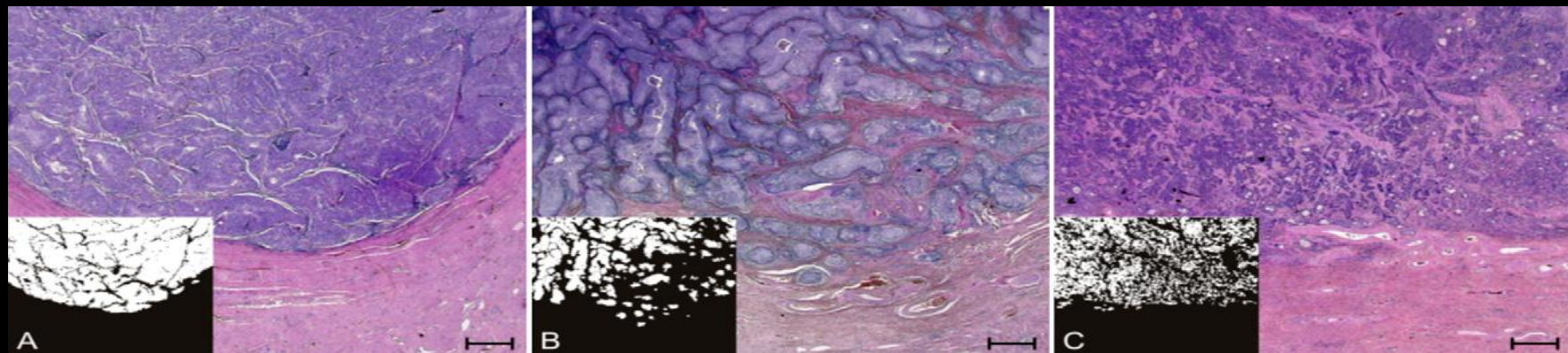
^a Department of Obstetrics and Gynecology, Leipzig University, Germany

^b Interdisciplinary Centre for Bioinformatics & Translational Centre for Regenerative Medicine, Leipzig University, Germany

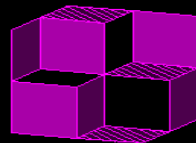
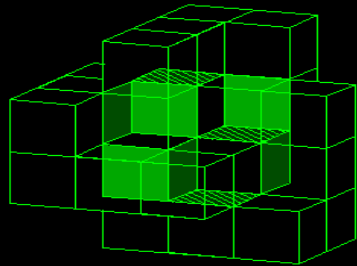
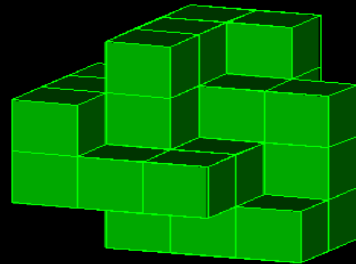
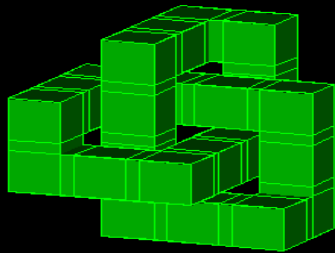
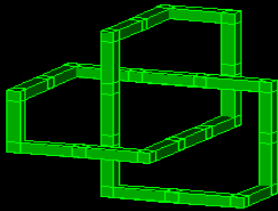
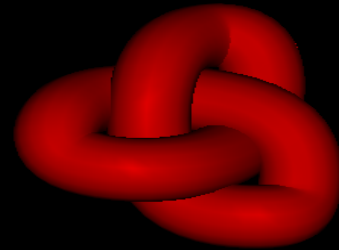
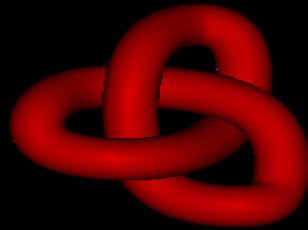
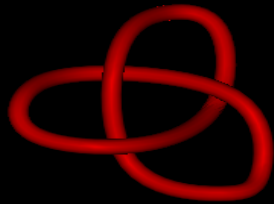
^c Institute of Pathology, Leipzig University, Germany

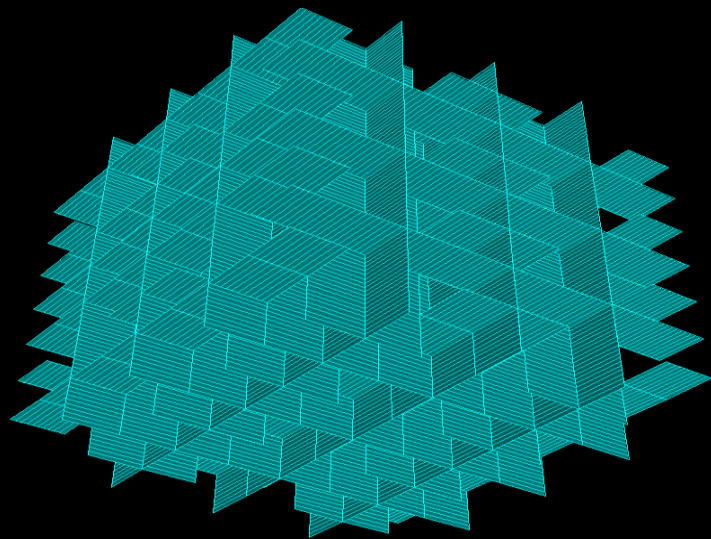
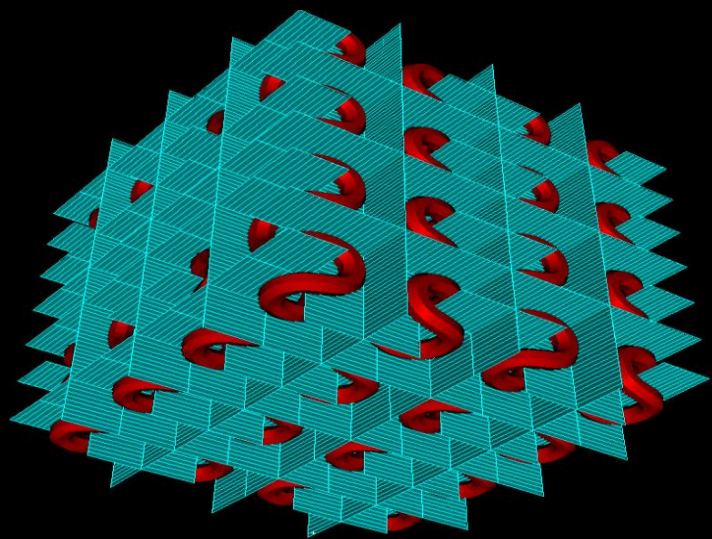
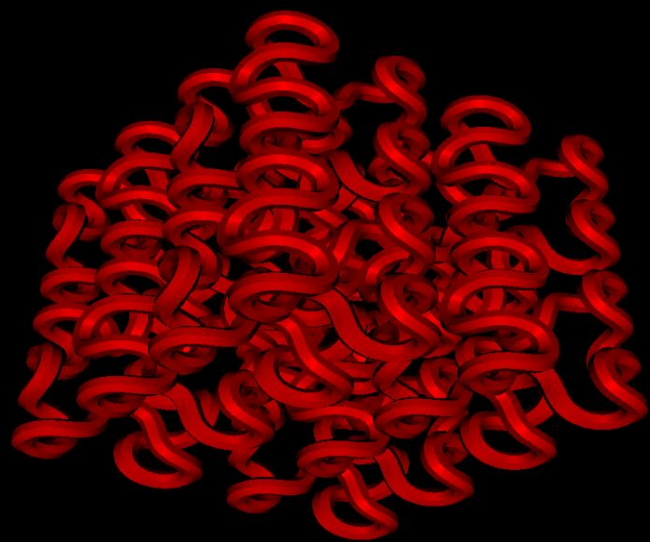
^d Institute of Medical Informatics, Statistics and Epidemiology, Leipzig University, Germany

Received 19 November 2005; received in revised form 20 March 2007; accepted 26 March 2007



Contact Knots

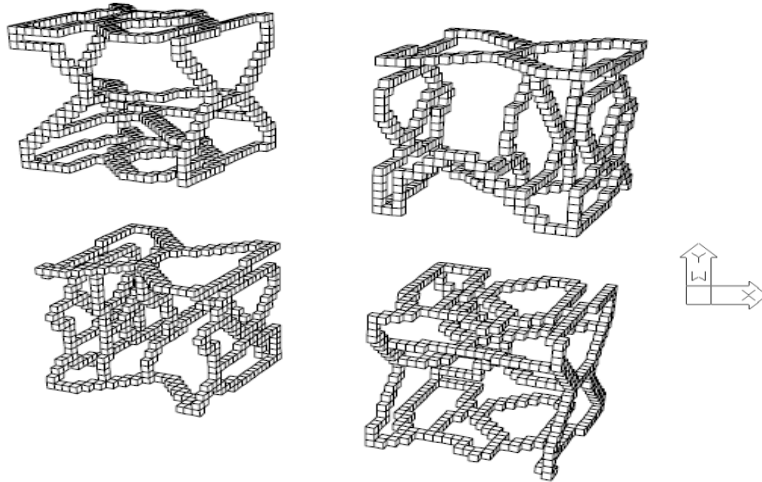




Computation of the Euler number using the contact perimeter

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$$\text{Euler} \\ F + V - E = 2$$

$$\text{Contact Perimeter} \\ H = \frac{2P_c - P}{4} + 1$$

