# A Compact Representation for Topological Decompositions of Non-Manifold Shapes

## David Canino, Leila De Floriani

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February 23, 2013

Manifold shapes (Topological Manifold)

Each neighborhood of every point *p* is homeomorphic to one connected component of a *ball*, centered at *p*.

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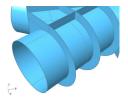
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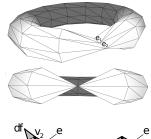
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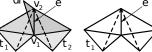
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There is a large amount of research in the literature, see *De Floriani and Hui, 2005* and *Botsch et al., 2010* 



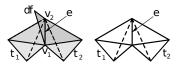


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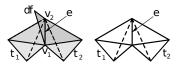
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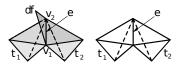
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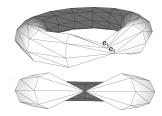
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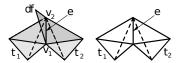
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## Structural Model

connections among meaningful components (global structure)

Main Property of Non-Manifold Shapes

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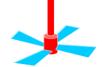
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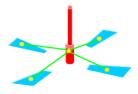
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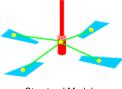
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### Key Idea of our Approach

Expose explicitly and combine *combinatorial* and *structural* information

### Compact Manifold-Connected (MC-) graph

Two-level graph-based representation of the

MC-decomposition, Hui and De Floriani, 2007

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Sequence of top *k*-simplices in  $\Sigma$ , where each simplex is *adjacent* through a (k - 1)-simplex, bounding *at most two* top *k*-simplices

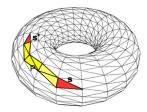
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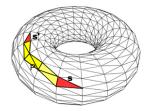
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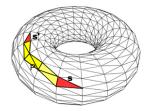
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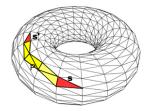
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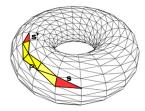
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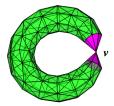
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Superclass of manifolds, they may contain non-manifold singularities

# Manifold-Connected (MC) Decomposition

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Decomposition of a simplicial complex Σ into its *MC-complexes* (*MC-components*)

Image: A matrix

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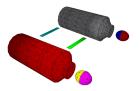
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1 MC-component



6 MC-components



8 MC-components



3 MC-components

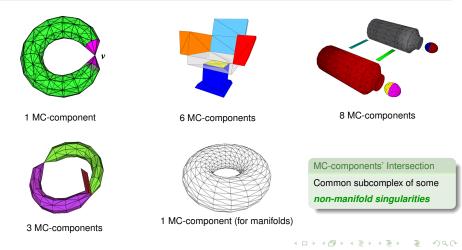


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Lower Level (Combinatorial Aspects)

Describes a non-manifold shape by any topological data structure  $\mathcal{M}_{\Sigma}$  (*unique*):

- the Incidence Simplicial (IS) data structure, De Floriani et al., 2010
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#### The Mangrove TDS Framework (Canino, 2012 - PhD. Thesis)

- Tool for the fast prototyping of topological data structures
- Extensible through dynamic plugins (mangroves)
- Any type of complexes is supported



The Mangrove TDS Library is released as GPL v3 software for the scientific community at http://mangrovetds.sourceforge.net

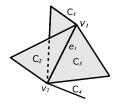
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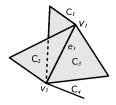
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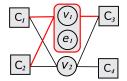
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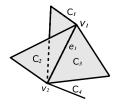
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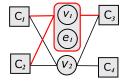


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References are directed toward simplices in  $\mathcal{M}_{\Sigma}$ 

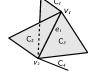
Similar to a spatial index on any non-manifold shape

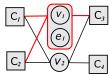


# Storage Cost $S_{\mathcal{C}} = n_{\mathcal{C}} + \sum_{a \in \mathcal{A}_{\Sigma}^{\mathcal{C}}} (k_a + s_a)$

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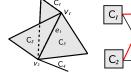


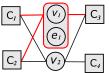
Properties of our Compact MC-graph

- Few hyperarcs
- Minimizes duplications of intersections
- Maximal list of MC-components in hyperarcs

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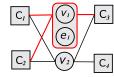
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Pairwise MC-Graph (Boltcheva, Canino, et al., 2011)

 $\mbox{Arcs}\equiv \textit{intersections}$  of only two MC-Components, formed by a

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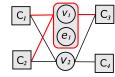


Verbose due to cliques

Less robust wrt to simmetry

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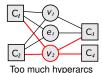
Exploded MC-Graph (Canino and De Floriani, 2011)

A hyper-arc  $\equiv$  one *non-manifold singularity*  $\sigma$ , and connects all the MC-components, bounded by  $\sigma$ 



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Many duplications of the same

MC-components in hyperarcs

## Experimental Results (with our Mangrove TDS Library)

Digital shapes are freely available from http://indy.disi.unige.it/nmcollection

2D shapes (Storage cost and Properties of MC-graphs)							
Shape	n <sub>C</sub>	a <sub>E</sub>	a <sub>P</sub>	a <sub>C</sub>	SE	S <sub>P</sub>	$S_{C}$
Carter	45	641	79	48	3.8 <i>k</i>	2.6 <i>k</i>	1.2 <i>k</i>
Chandelier	130	616	328	96	2.6 <i>k</i>	2.6 <i>k</i>	1 <i>k</i>
Pinched Pie	120	1.4 <i>k</i>	1.4 <i>k</i>	192	4.8 <i>k</i>	9.6 <i>k</i>	1.9 <i>k</i>
Tower	169	1.4 <i>k</i>	13 <i>k</i>	165	5.9 <i>k</i>	43 <i>k</i>	2.1 <i>k</i>

 $n_C$ : #MC-components  $a_E, a_P, a_C$ : #(hyper)arcs  $S_E, S_P, S_C$ : storage costs

For 2D shapes:

 $\begin{array}{l} a_E \approx 8.9 \times a_C, \, a_P \approx 23 \times a_C \\ S_E \approx 2.8 \times S_C, \, S_P \approx 7.7 \times S_C \end{array}$ 

#### 3D shapes (Storage cost and Properties of MC-graphs)

Shape	n <sub>C</sub>	a <sub>E</sub>	a <sub>P</sub>	a <sub>C</sub>	SE	$S_P$	S <sub>C</sub>
Chime	27	29	47	28	133	210	127
Flasks	8	76	10	6	300	232	98
Teapot	2.9 <i>k</i>	1.2 <i>k</i>	18.1 <i>k</i>	1 <i>k</i>	10.4 <i>k</i>	57.5 <i>k</i>	10.1 <i>k</i>
Wheel	115	136	520	88	675	1.7 <i>k</i>	563

For 3D shapes:

 $\begin{array}{l} a_E \approx 4.1 \times a_C, \, a_P \approx 6.8 \times a_C \\ S_E \approx 1.6 \times S_C, \, S_P \approx 3.2 \times S_C \end{array}$ 

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Our experimental results confirm properties of the Compact MC-graph

# Experimental Results (cont'd)

Comparisons with the Incidence Graph, Edelsbrunner, 1987						
Shape	S <sub>C</sub>	S <sub>IA*</sub>	S <sub>IG</sub>	$S_{C} + S_{IA*}$		
Carter	1.2 <i>k</i>	52 <i>k</i>	95 <i>k</i>	53.2 <i>k</i>		
Chandelier	1 <i>k</i>	120 <i>k</i>	220 <i>k</i>	121 <i>k</i>		
Tower	2.1 <i>k</i>	124 <i>k</i>	221 <i>k</i>	126.1 <i>k</i>		
Flasks	98	29 <i>k</i>	104 <i>k</i>	29.1 <i>k</i>		
Teapot	10.1 <i>k</i>	85 <i>k</i>	220 <i>k</i>	95.1 <i>k</i>		
Sierpinski 3D	458 <i>k</i>	524 <i>k</i>	3.67 <i>M</i>	0.98 <i>M</i>		
Sierpinski 4D	664 <i>k</i>	781 <i>k</i>	11.6 <i>M</i>	1.44 <i>M</i>		
Sierpinski 5D	467 <i>k</i>	559.6 <i>k</i>	7.7M	1 <i>M</i>		

Combined with the IA\* data structure, *Canino et al., 2011* 

> $S_{IA^*}$ : storage cost of the IA\*  $S_{IG}$ : storage cost of the IG

For 2D shapes:  $S_{IG} \approx 1.45 \times S_{IA*}$ For 3D shapes:  $S_{IG} \approx 3.2 \times S_{IA*}$ For 4D shapes:  $S_{IG} \approx 8 \times S_{IA*}$ For 5D shapes:  $S_{IG} \approx 7.7 \times S_{IA*}$ 

# Experimental Results (cont'd)

Comparisons with the Incidence Graph, Edelsbrunner, 1987						
Shape	S <sub>C</sub>	S <sub>IA*</sub>	S <sub>IG</sub>	$S_{C} + S_{IA^*}$		
Carter	1.2 <i>k</i>	52 <i>k</i>	95 <i>k</i>	53.2k		
Chandelier	1 <i>k</i>	120 <i>k</i>	220 <i>k</i>	121 <i>k</i>		
Tower	2.1 <i>k</i>	124 <i>k</i>	221 <i>k</i>	126.1 <i>k</i>		
Flasks	98	29 <i>k</i>	104 <i>k</i>	29.1 <i>k</i>		
Teapot	10.1 <i>k</i>	85 <i>k</i>	220 <i>k</i>	95.1 <i>k</i>		
Sierpinski 3D	458 <i>k</i>	524 <i>k</i>	3.67 <i>M</i>	0.98 <i>M</i>		
Sierpinski 4D	664 <i>k</i>	781 <i>k</i>	11.6 <i>M</i>	1.44 <i>M</i>		
Sierpinski 5D	467 <i>k</i>	559.6 <i>k</i>	7.7 <i>M</i>	1 <i>M</i>		

Combined with the IA\* data structure, *Canino et al., 2011* 

> $S_{IA^*}$ : storage cost of the IA\*  $S_{IG}$ : storage cost of the IG

For 2D shapes:  $S_{IG} \approx 1.45 \times S_{IA*}$ For 3D shapes:  $S_{IG} \approx 3.2 \times S_{IA*}$ For 4D shapes:  $S_{IG} \approx 8 \times S_{IA*}$ For 5D shapes:  $S_{IG} \approx 7.7 \times S_{IA*}$ 

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#### Interesting result (wrt the Incidence Graph)

The *Compact MC-graph*, combined with the *IA*\* data structure, is *more compact* than the incidence graph:

- our contribution is a structural model (topological + structural aspects)
- the IG data structure is a topological data structure (*local connectivity*)

Compact Manifold-Connected (MC-) graph

Two-level graph-based representation of the *MC-decomposition*, *Hui and De Floriani, 2007* 

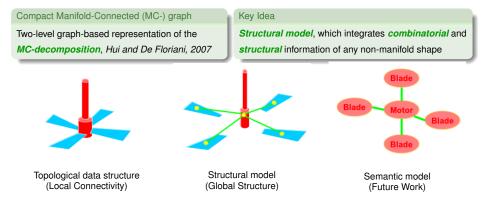
(I)

#### Compact Manifold-Connected (MC-) graph

Two-level graph-based representation of the *MC-decomposition*, *Hui and De Floriani, 2007* 

#### Key Idea

Structural model, which integrates combinatorial and structural information of any non-manifold shape



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Image: A matrix

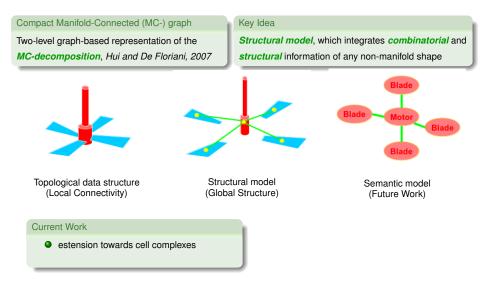


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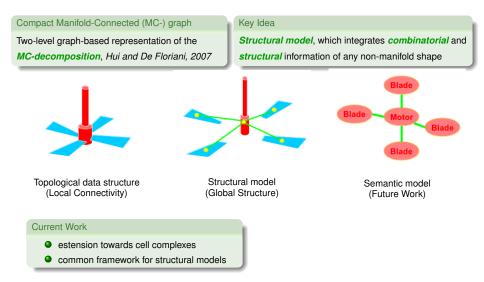
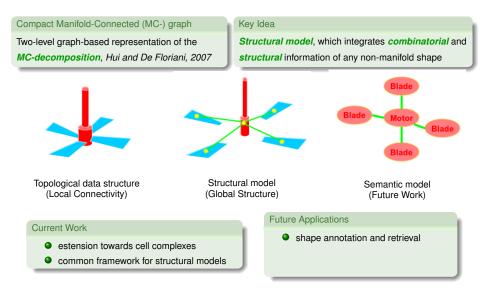
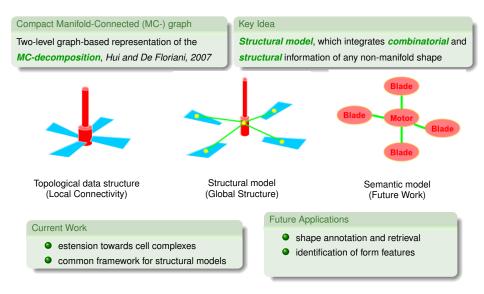
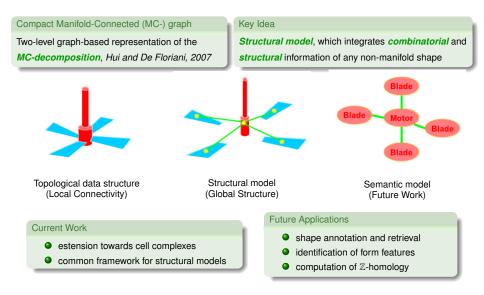


Image: Image:







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These slides are available on http://www.disi.unige.it/person/CaninoD

## Thank you much for your attention!

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