

# Simplification and Refinement

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# Approximating surfaces with triangle meshes

***Managing the quantity of geometry vs the quality of the representation***

## **Assumptions:**

Surfaces represented by triangle meshes

Accuracy of the approximation is proportional to the number of triangles

Complexity of a surface is proportional to the number of triangles

## **Objective:**

Always produce the simplest mesh that satisfies the accuracy required by the application

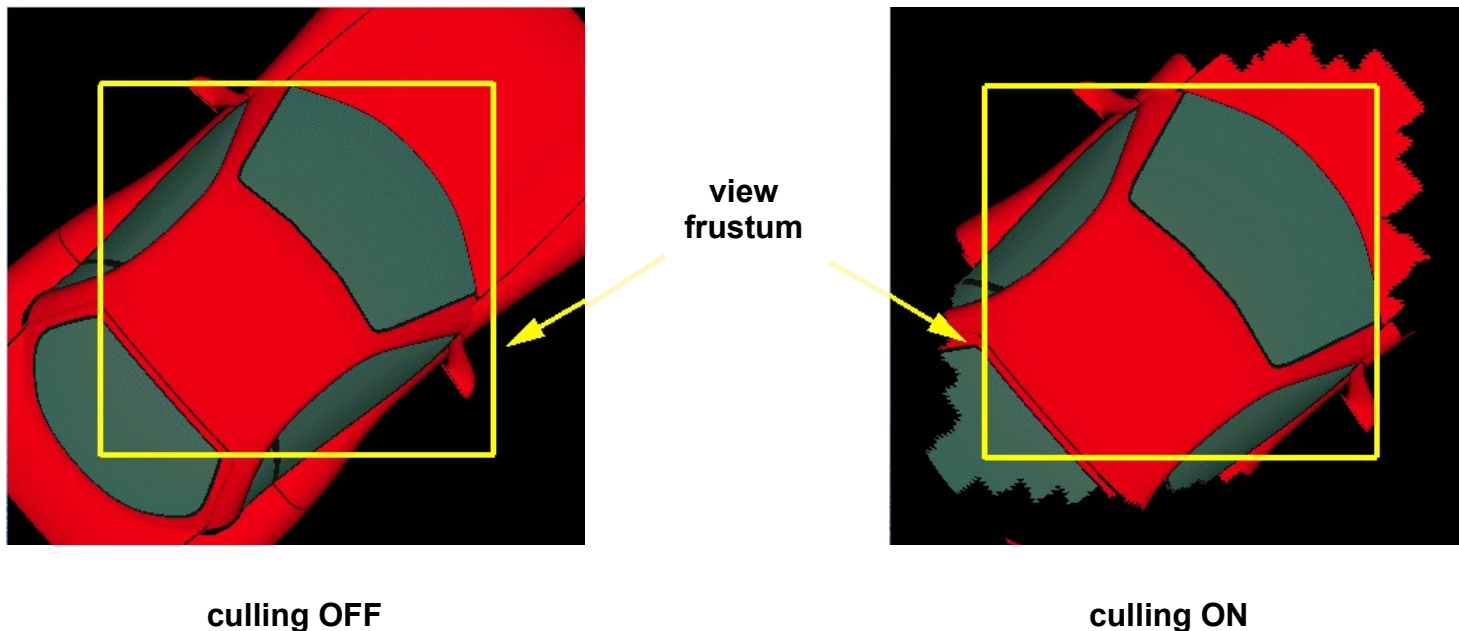
# Reducing rendering cost

Simplifying is not the only way to reduce the needed number of triangles  
At rendering time we can do:

## View Frustum Culling

Drawing only what is in the view frustum

...



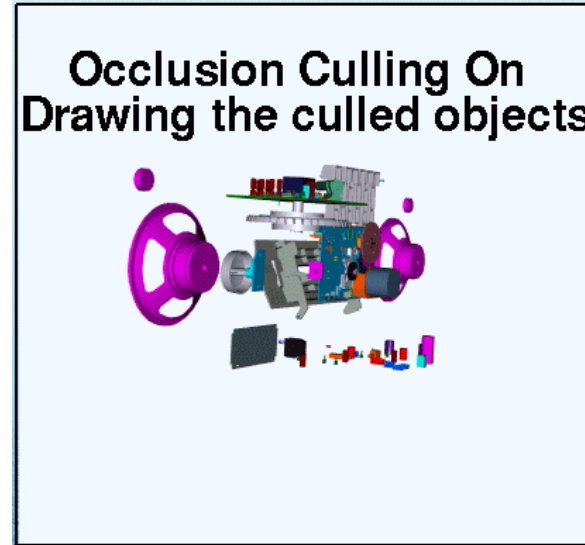
# Reducing Complexity

...

## Occlusion/Visibility culling

Drawing only what it is probably visible

➔ *Improve performances, but not sufficient for very large datasets*



# Simplification

From a computational Point of view, fixed the size of the solution find the **best** representation

The optimal solution in its general form is NP Hard

Heuristics works well  
(even with bounds)

# Metrics

Computing the “difference” between two meshes

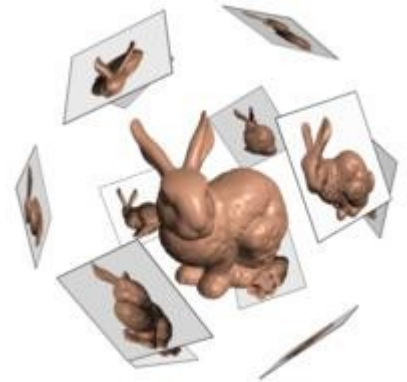
Geometric way

Hausdorff

Perceptual

Working in Image space

Many renderings,  
computing the difference,  
possibly in a human-like way



# Appearance similarity

Difference between two images: (trivial)

$$D(I_1, I_2) = \frac{1}{n^2} \sum_x \sum_y d(I_1(x, y), I_2(x, y))$$

Difference between two objects:

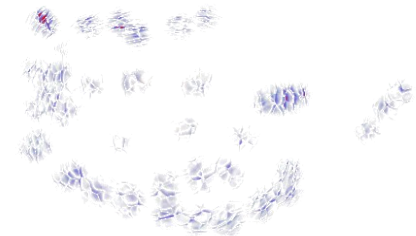
Integrate the above over all possible views



$I_1$



$I_2$



$I_1 - I_2$

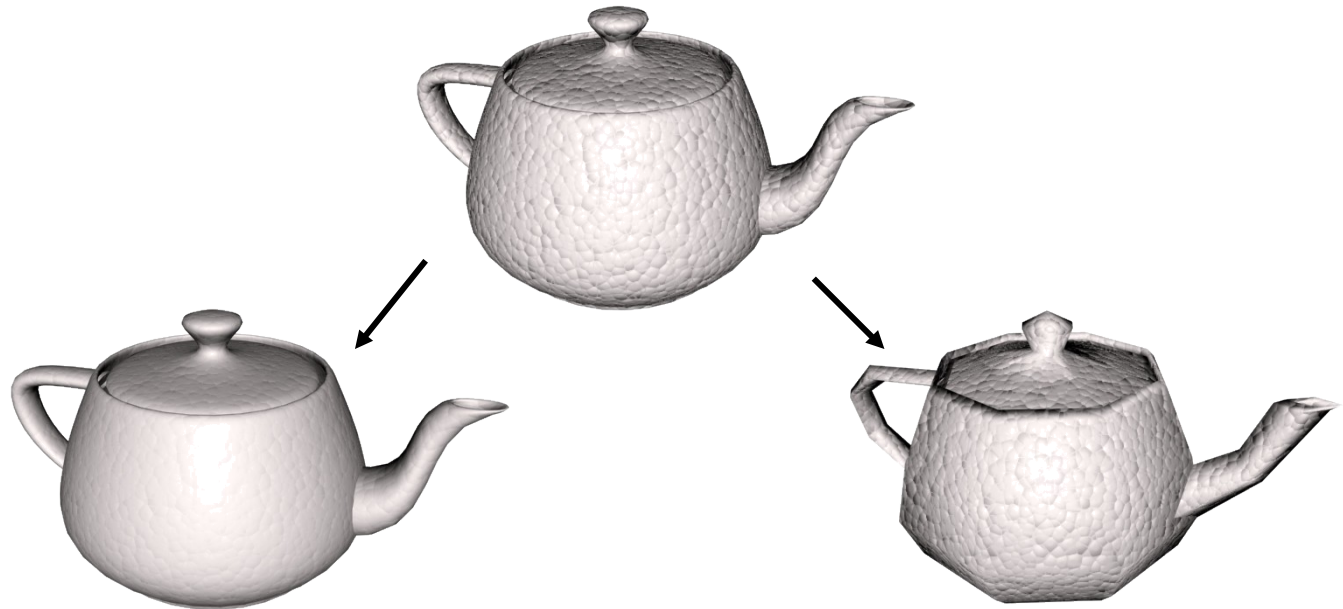
# Approximation error

Quantifies the notion of “similarity”

Two kinds of similarity:

**Geometric similarity** (surface deviation)

**Appearance similarity** (material, normal...)





# Geometric similarity

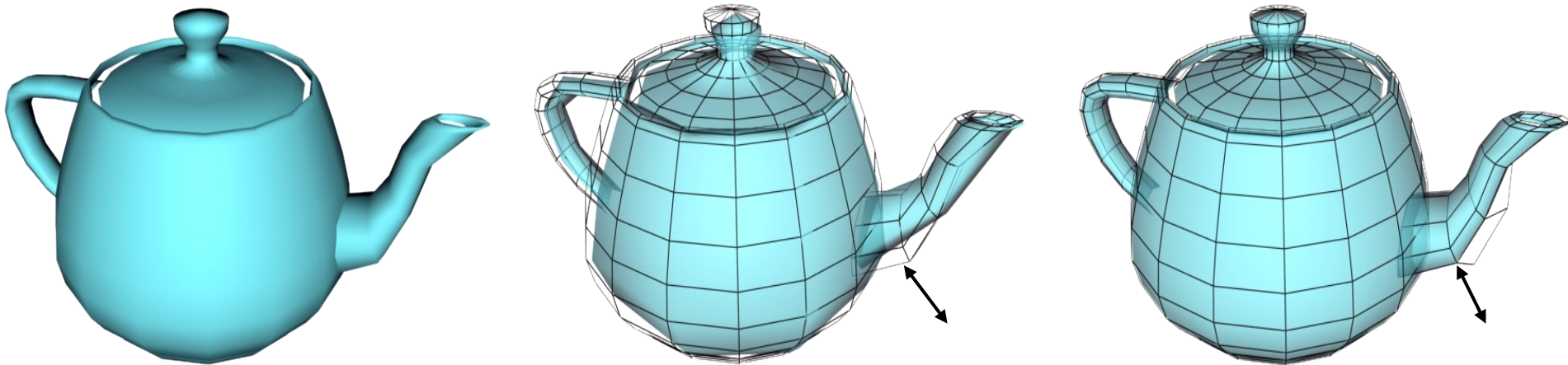
Two main components:

Distance function

Function Norm:

$L_2$ : average deviation

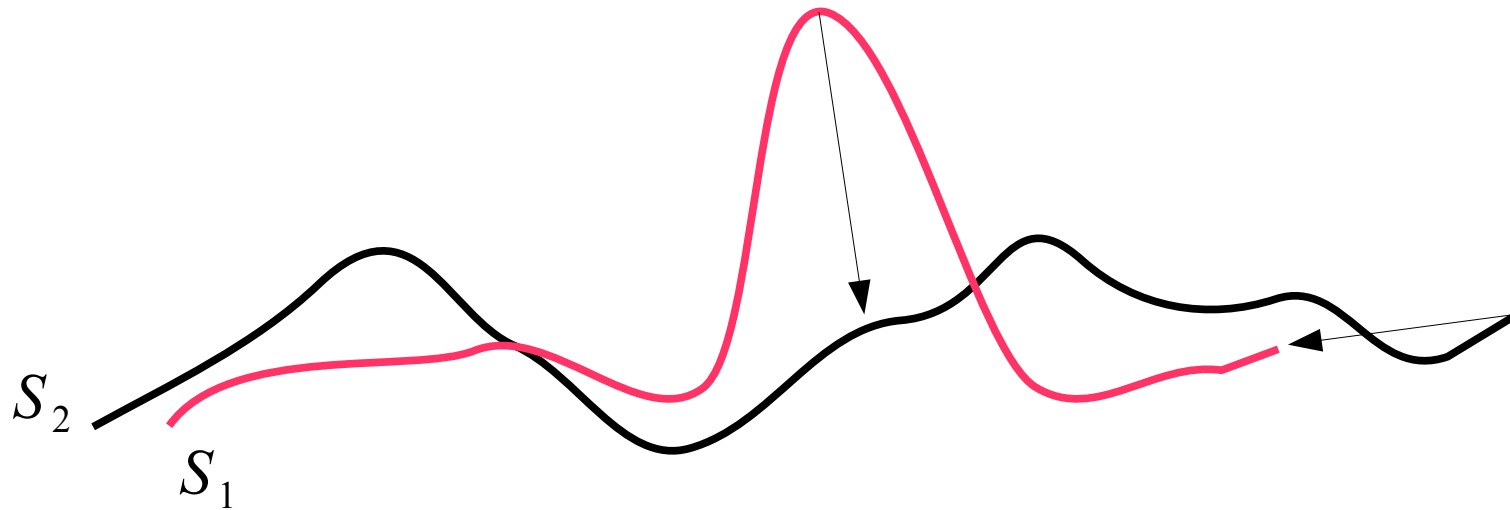
$L_{\text{inf}}$ : maximum deviation - Hausdorff distance



[Image by C. Andujar]

# Hausdorff Distance

$$D_H(S_1, S_2) = \max_{x \in S_1} (\min_{y \in S_2} D(x, y))$$



Symmetric version:

$$D(S_1, S_2) = \max \{ D_H(S_1, S_2), D_H(S_2, S_1) \}$$

# Computing Hausdorff Distance

In an approximate way

Sample a surface

Points uniformly distributed over it

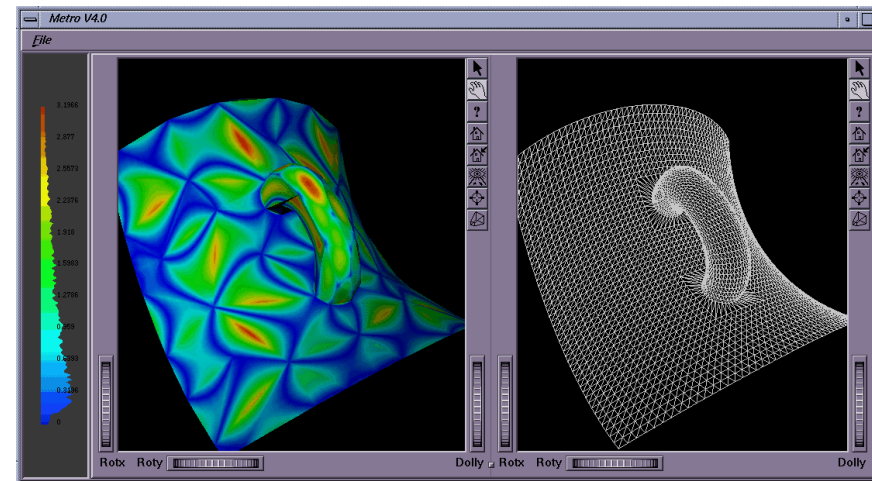
For each point  $x$  compute

$$\min_{y \in S_2} D(x, y)$$

points usually very near to surface,

UG works well, better than trees

Avg are useful too...



# Simplification Algorithms

## Simplification approaches:

**incremental methods** based on local updates

mesh decimation

[Schroeder et al. '92, ... + others]

energy function optimization & progressive meshes

[Hoppe et al. '93, Hoppe '96, Hoppe '97]

**quadric error metrics**

[Garland et al. '97]

**coplanar facets merging**

[Hinker et al. '93, Kalvin et al. '96]

**clustering**

[Rossignac et al. '93, ... + others]

**wavelet-based**

[Eck et al. '95, + others]

**volume-based**

[He+ 95, He+ 96, Andujar+ 02]

# Incremental methods based on *local updates*

**All of the methods such that :**

simplification proceeds as a sequence of small changes of the mesh (in a greedy way)

each update reduces mesh size and  
[~monotonically] decreases the approximation  
precision

**Different approaches:**

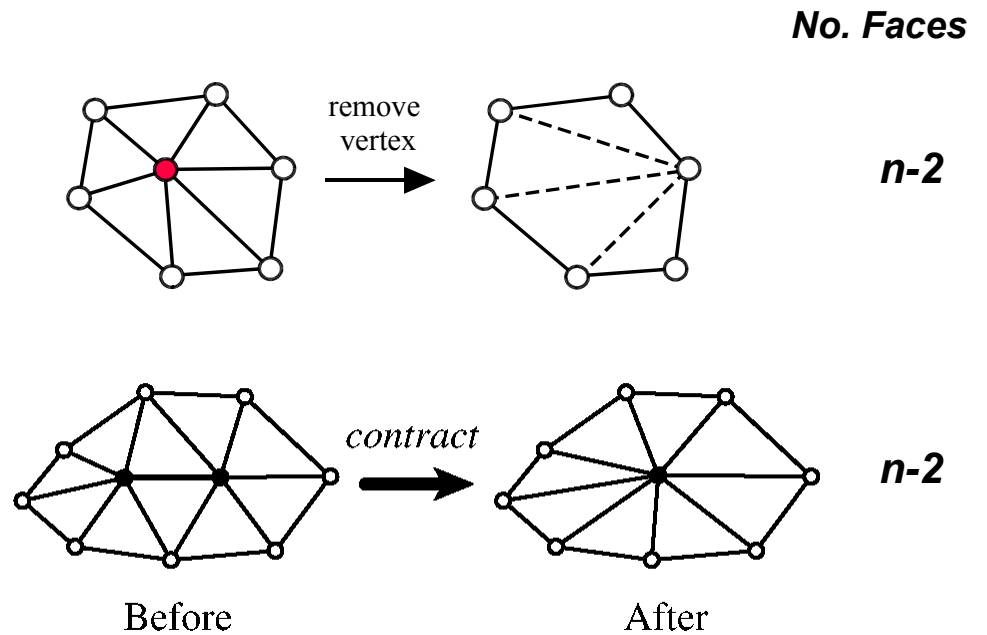
- mesh decimation
- energy function optimization
- progressive meshes
- quadric error metrics

... Incremental methods based on *local updates* ...

## Local update actions:

vertex removal

edge collapse  
preserve location  
new location



# Incremental Simp Method

The common greedy framework:

**While** *mesh size/precision is satisfactory*

**Select** from a priority queue the best element to simplify

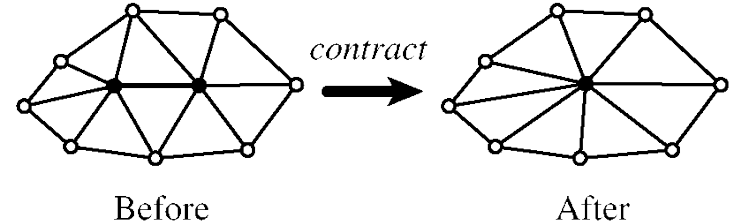
**Perform simplification** modifying the mesh

**Ppdate** priority queue according to the mods;

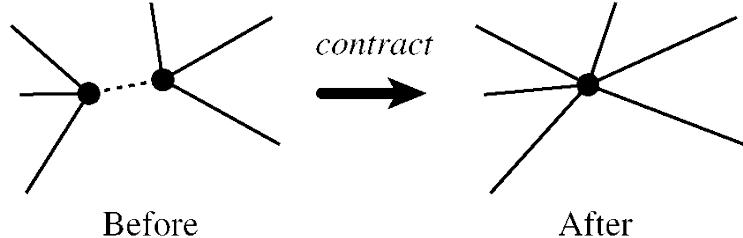
# Simplification using Quadric Error Metrics

[Garland et al. Sig'97]

Based on incremental **edge-collapsing**



**but** can also collapse vertex couples which are **not connected** (topology is not preserved)





# Quadric Error for Surfaces

Let  $\mathbf{n}^T \mathbf{v} + d = 0$  be the equation representing a plane

The squared distance of a point  $\mathbf{x}$  from the plane is

$$D(\mathbf{x}) = \mathbf{x}(\mathbf{n}\mathbf{n}^T)\mathbf{x} + 2d\mathbf{n}^T\mathbf{x} + d^2$$

This distance can be represented as a quadric

$$Q = (A, \mathbf{b}, c) = (\mathbf{n}\mathbf{n}^T, d\mathbf{n}, d^2)$$

$$Q(\mathbf{x}) = \mathbf{x}A\mathbf{x} + 2\mathbf{b}^T\mathbf{x} + c$$

# Quadric

The error is estimated by providing for each vertex  $v$  a quadric  $Q_v$  representing the sum of the all the squared distances from the faces incident in  $v$

The error of collapsing an edge  $e=(v,w)$  can be evaluated as  $Q_w(v)$ .

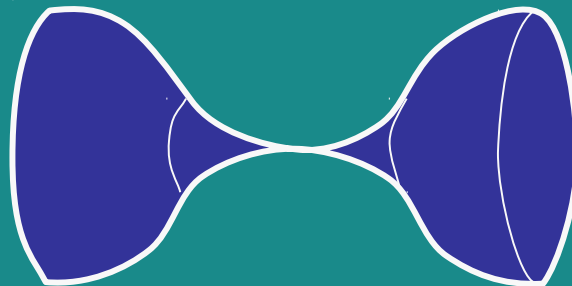
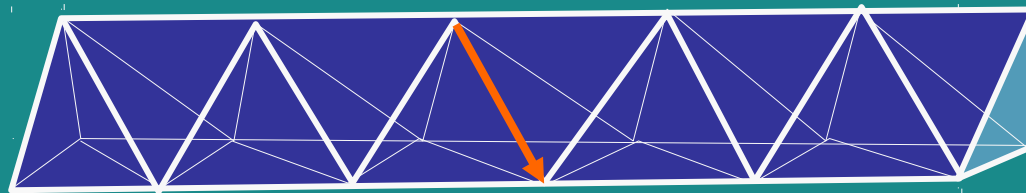
After the collapse the quadric of  $v$  is updated as follow  $Q_v = Q_v + Q_w$

# Topology Preservation

## 2-Manifold

A surface  $\Sigma$  in  $\mathbf{R}^2$  such that any point on  $\Sigma$  has an open neighborhood homeomorphic to an open disc or to half an open disc in  $\mathbf{R}^2$

A edge collapse can create non manifold situations

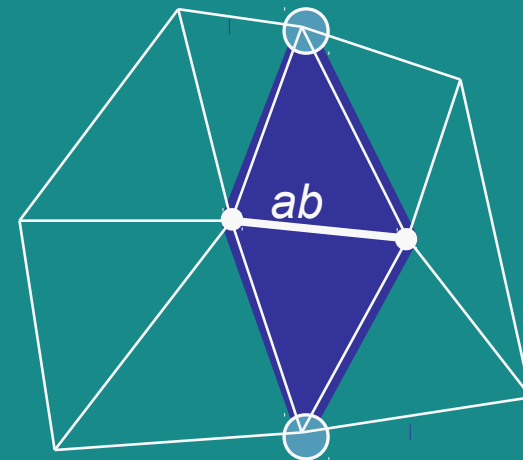
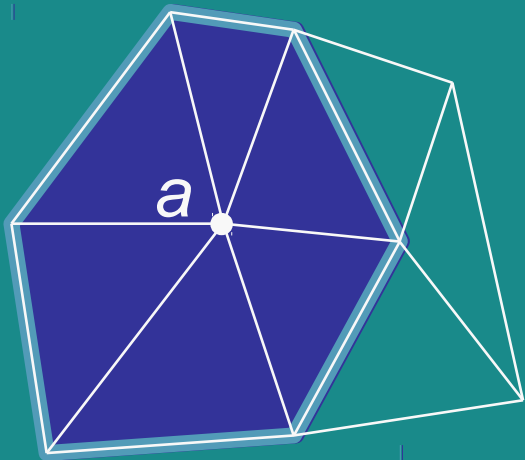


# Topology Preservation

Let  $\Sigma$  be a 2 simplicial complex ***without boundary***

$\Sigma'$  is obtained by collapsing the edge  $e = (ab)$

Let  $Lk(\sigma)$  be the set of all the faces of the co-faces of  $\sigma$  disjoint from  $\sigma$



$\Sigma$  and  $\Sigma'$  are homeomorphic ***iff***

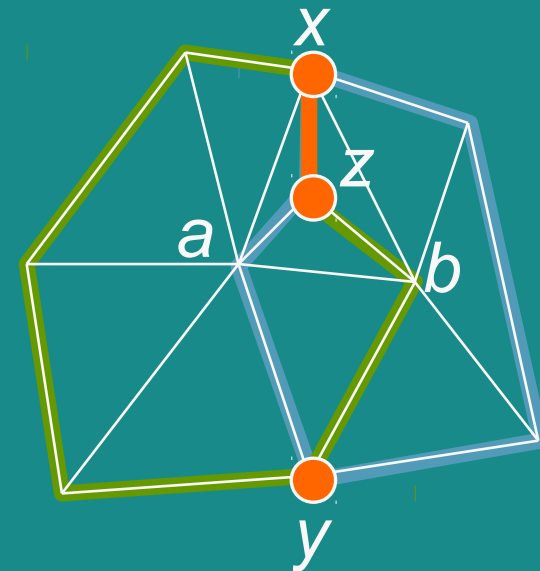
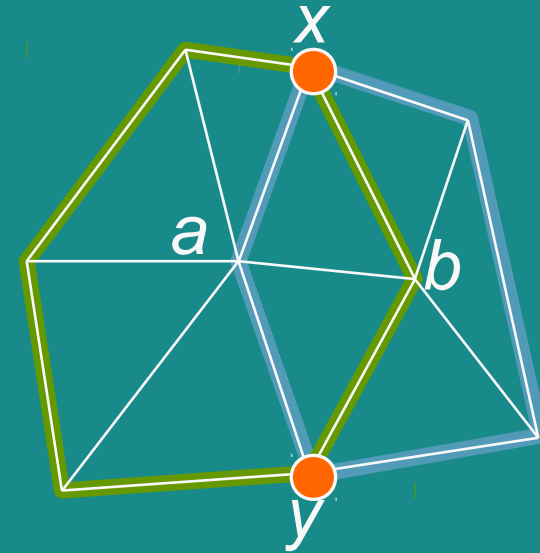
$$Lk(a) \cap Lk(b) = Lk(ab)$$

[Dey 99]

# Topology Preservation

$$Lk(a) \cap Lk(b) = \{x, y\} = Lk(ab)$$

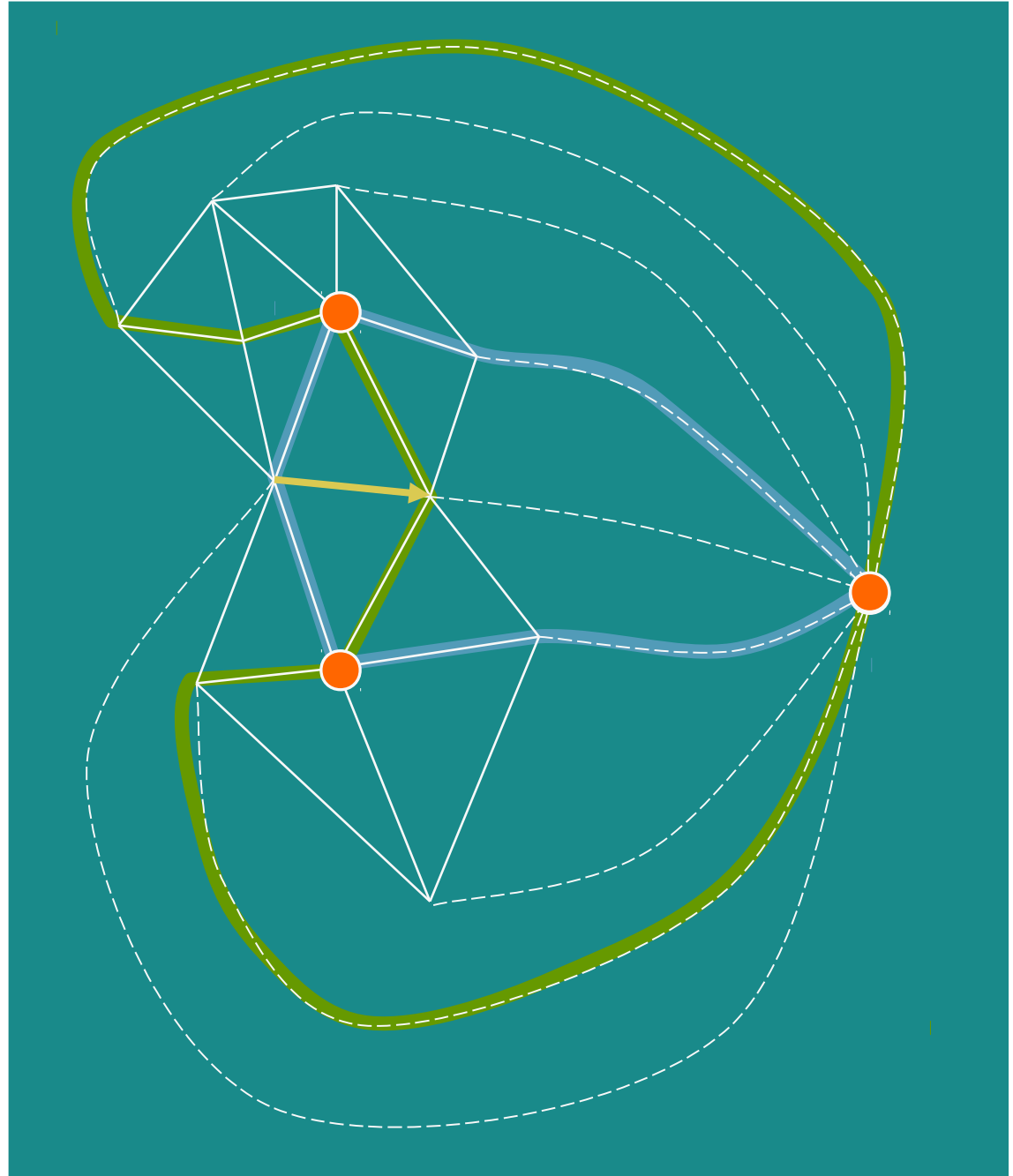
$$Lk(a) \cap Lk(b) = \{x, y, z, zx\} \neq \{y, z\} = Lk(ab)$$



# Topology Preservation

Mesh with boundary can be managed by considering a dummy vertex  $v_d$  and, for each boundary edge  $e_a$  dummy triangle connecting  $e$  with  $v_d$

Think it wrapped on the surface of a sphere

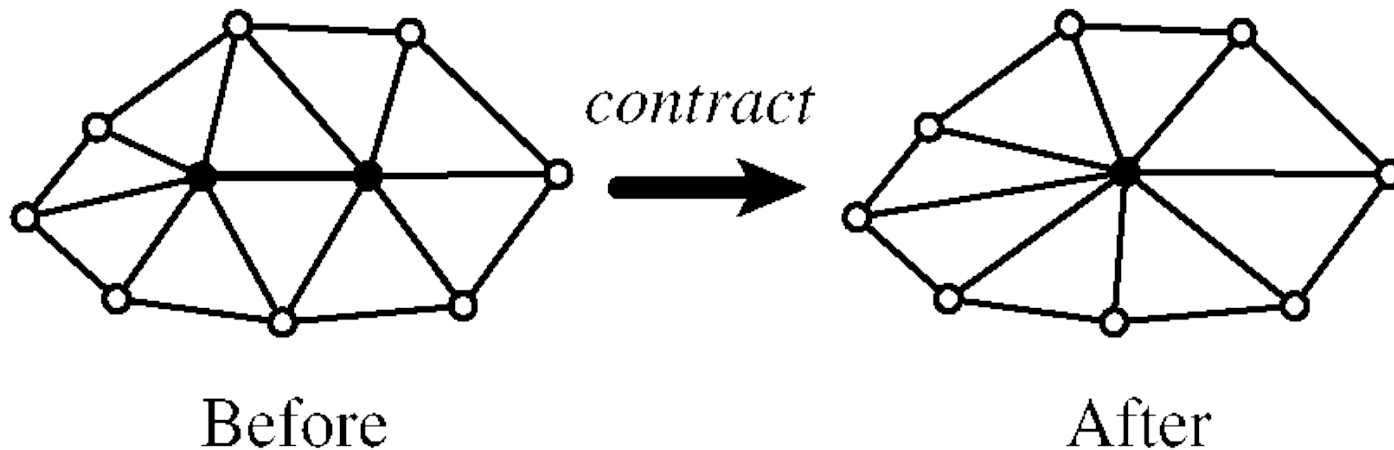


# Lazy heap

Si suppone di avere uno heap con tutte le operazioni

Estraggo da heap e aggiornano la mesh

tali operazioni invalidano/modificano la mesh e quindi le priorità/validità di parte delle azioni già presenti nello Heap



# Lazy Heap

## Due Soluzioni

Link espliciti elementi mesh->heap e  
aggiornamento dello stesso

## Lazy update

Si mettono nello heap tutte le nuove operazioni  
con la nuova priorità

Quando si estrae un'op dall heap si controlla che  
sia sempre valida

Di tanto in tanto garbage collection sullo heap



# Validità collasso

## Dati

Ogni vertice ha una marca temporale:

quando e' stato modificato l'ultima volta

Ogni collasso (coppia di vertici) ha una marca temporale

quando è stato inserito nello heap

## Un collasso è valido se

I due vertici non sono stati cancellati

Il collasso e' stato messo nello heap piu recentemente della data di ultima modifica dei vertici

# Simplification Algorithms

Not-incremental methods:

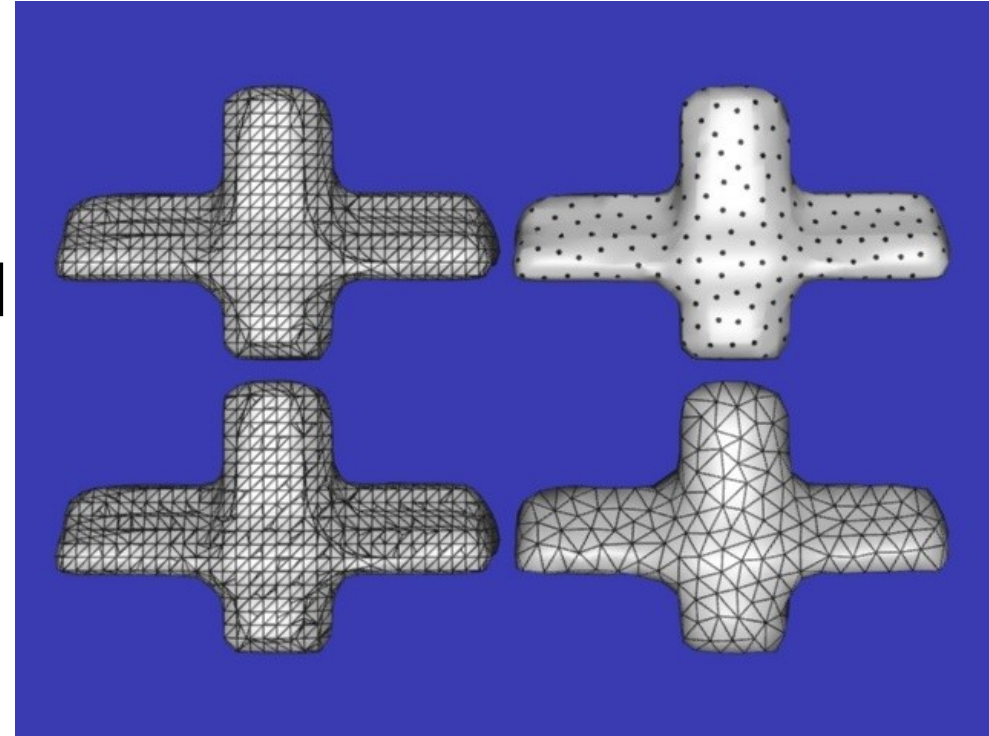
coplanar facets merging  
[Hinker et al. '93, Kalvin et al. '96]

re-tiling  
[Turk '92]

clustering  
[Rossignac et al. '93, ... + others]

wavelet-based  
[Eck et al. '95]

volume-based  
[He+ 95, He+ 96, Andujar+ 02]



# Clustering

## *Vertex Clustering*

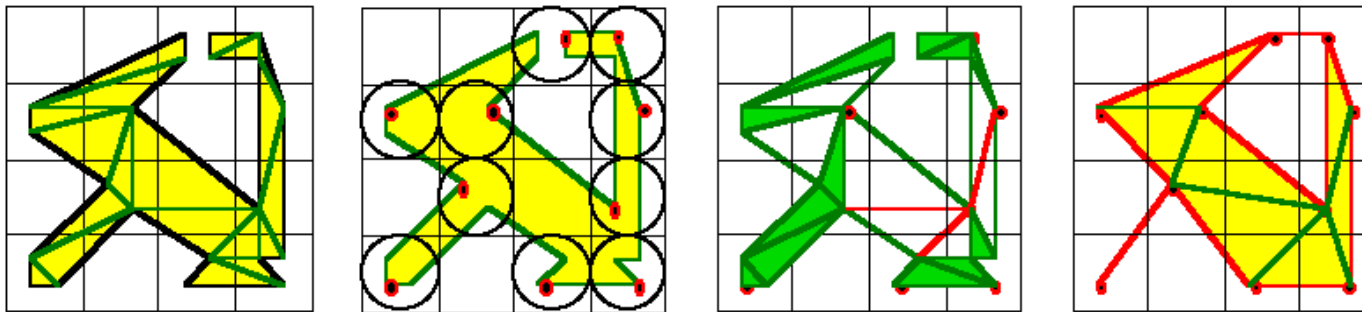
[Rossignac, Borrel '93]

detect and unify **clusters** of nearby vertices (discrete gridding and coordinates truncation)

all faces with two or three vertices in a cluster are removed

does not preserve topology (faces may degenerate to edges, genus may change)

approximation depends on grid resolution



(figure by Rossignac)

# Clustering -- Examples

Simplification of a table lamp,  
Interaction Accelerator

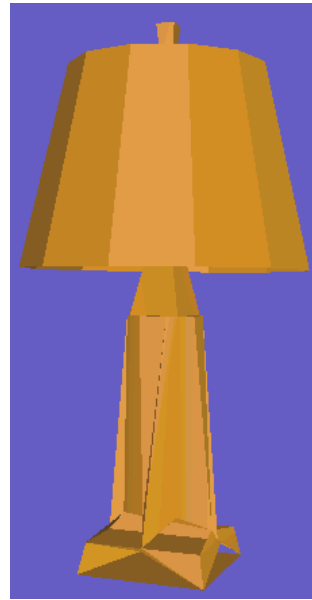
IBM 3D



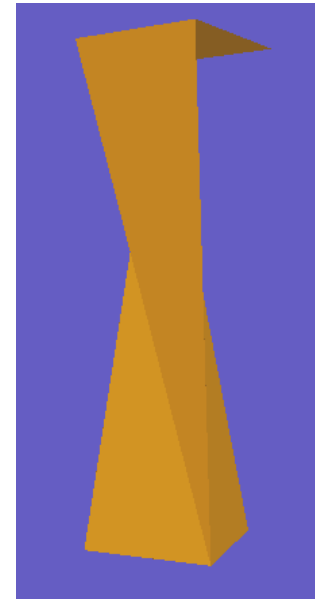
10,108 facets



1,383 facets



474 facets



46 facets

# ... Clustering...

## **Clustering - *Evaluation***

high efficiency

very simple implementation

low quality approximations

does not preserve topology

error is bounded by the grid cell size

REFINEMENT  
or  
Subdivision Surfaces

# Subdivision Surfaces

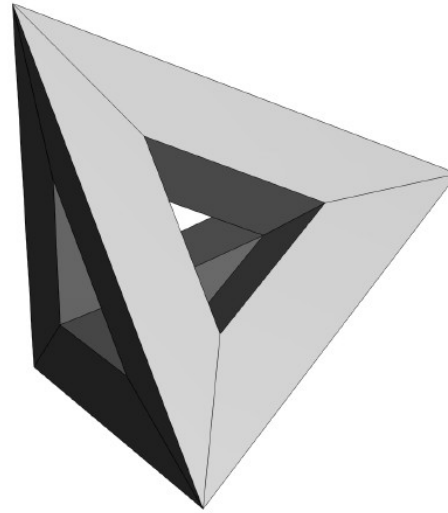
*Subdivision defines a smooth curve or surface as the limit of a sequence of successive refinements.*

Si parte da una mesh poligonale

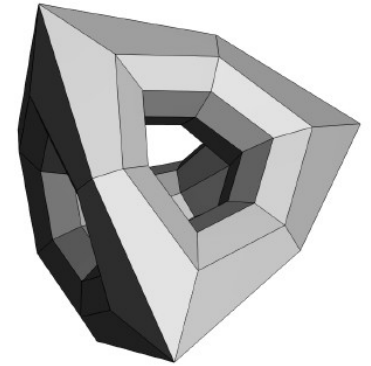
Si suddivide i poligoni che la compongono

Smooth della superficie muovendo i vertici

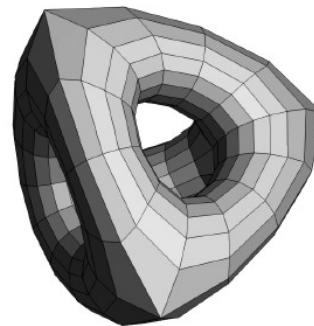
In effetti quello che si vede sono sempre  
approx delle vere subdiv surfaces



(a)



(b)



(c)



(d)



# Esempio

## Geri's Game (1997)

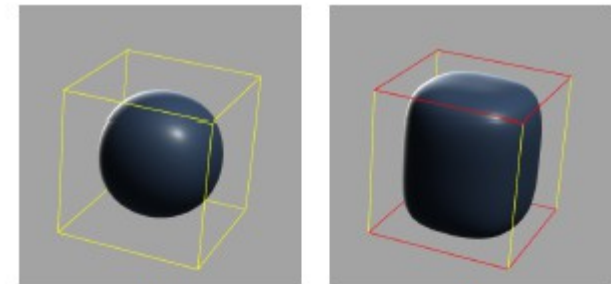
Primo esempio non  
accademico di uso di  
subdivision surfaces



# Smoothness

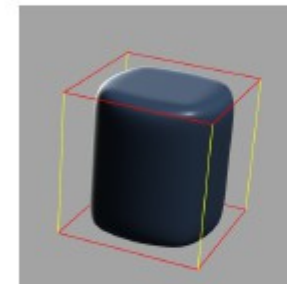
Non solo superfici smooth

Variable sharpness creases

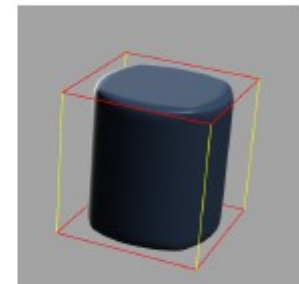


(a)

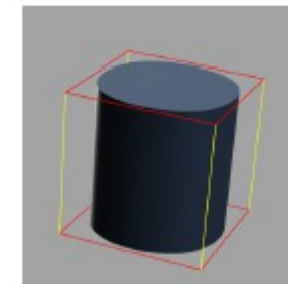
(b)



(c)



(d)



(e)

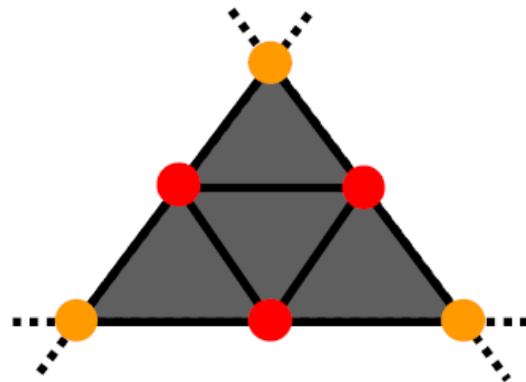
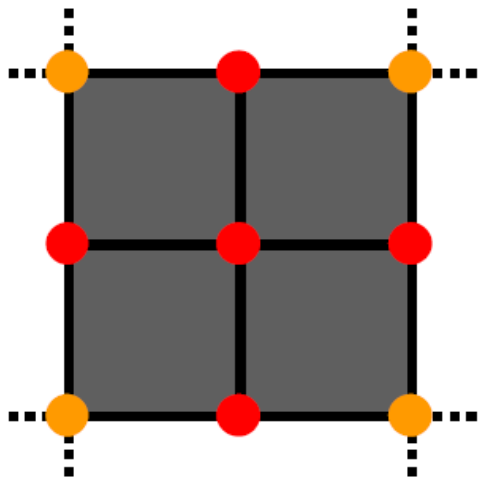
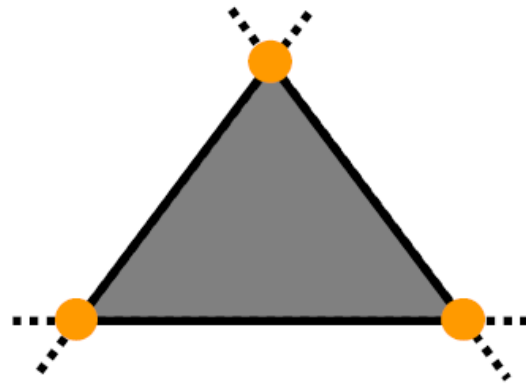
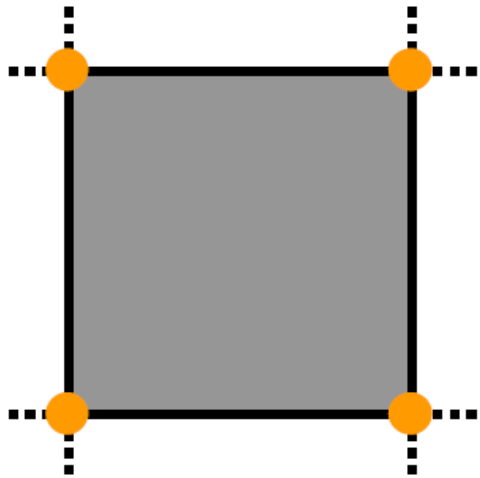
# Subdivision Classification

<b>Primal</b>	Faces are split into sub-faces
<b>Dual</b>	Vertices are split into multiple vertices

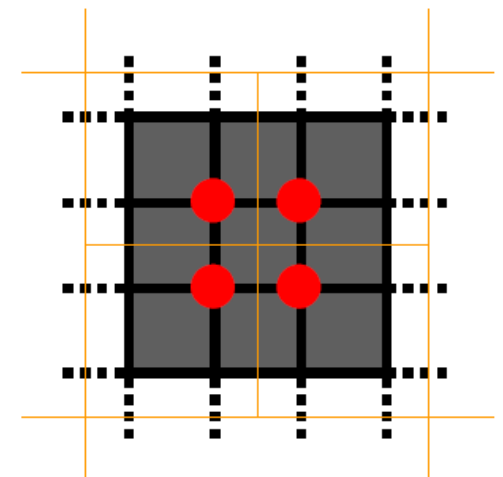
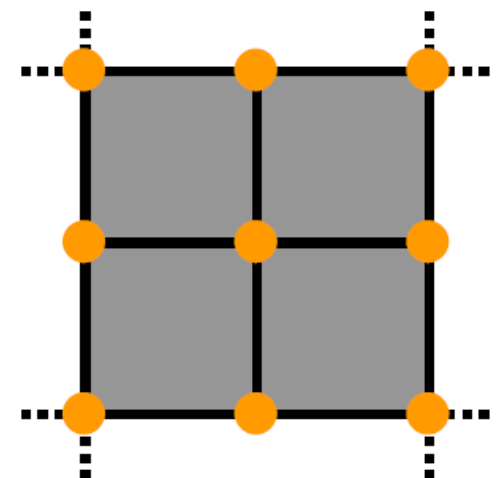
<b>Approximating</b>	Control points not interpolated
<b>Interpolating</b>	Control points are interpolated

# Subdivision Classification

Primal



Dual



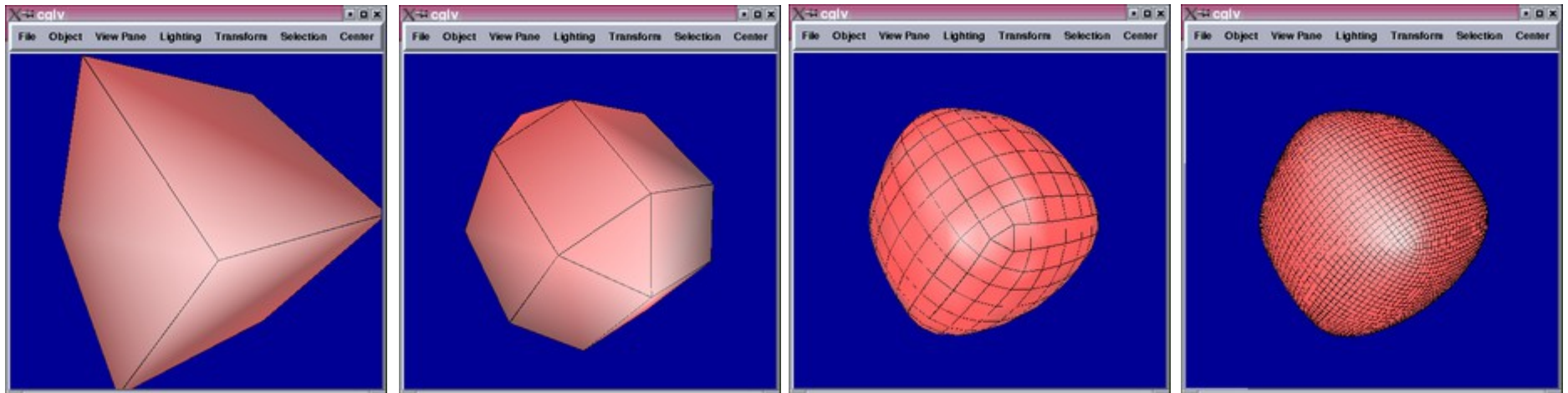
# Subdivision

	<i>Primal</i>		<i>Dual</i>
	Triangles	Rectangles	
Approximating	Loop	Catmull-Clark	Doo-Sabin
Interpolating	Butterfly	Kobbelt	Midedge

# Doo Sabin

Per mesh poligonali

Duale ad ogni vertice corrisponde una nuova faccia



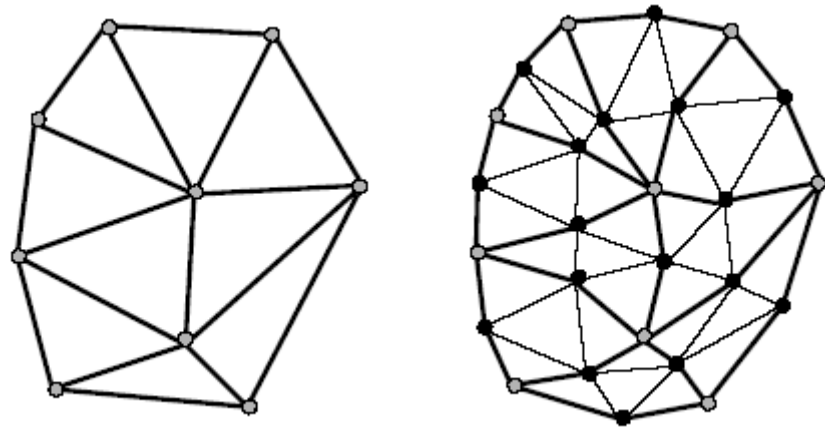


# Loop Refinement Scheme

Funziona per mesh triangolari

Vertex insertion

Ogni edge è diviso in due e i nuovi vertici sono riconnessi per formare nuovi triangoli





# Loop Subdivision

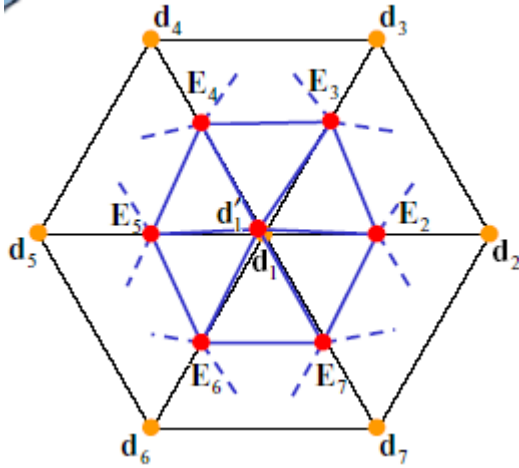
Approssimante (non interpolante)

Continua

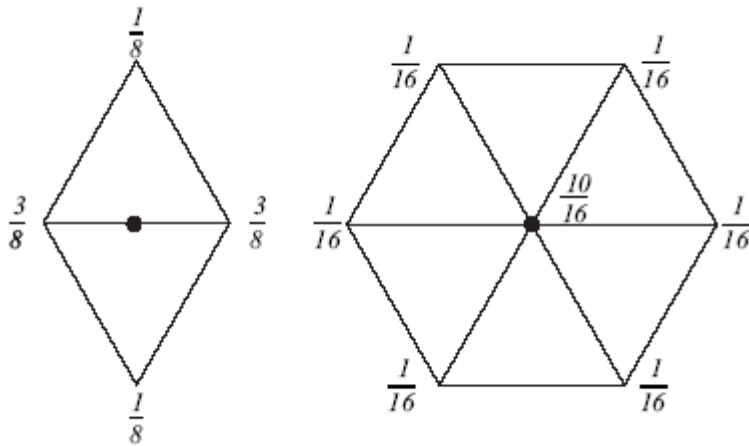
C1 su vertici straordinari (valenza  $\neq 6$ )

C2 elsewhere

# Loop Subdivision



$$E_i = \frac{3}{8}(d_1 + d_i) + \frac{1}{8}(d_{i-1} + d_{i+1})$$



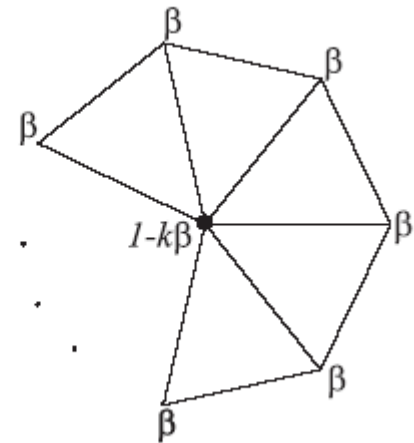
$$\mathbf{d}'_1 = \alpha_n \mathbf{d}_1 + \frac{(1 - \alpha_n)^{n+1}}{n} \sum_{j=2}^{n+1} \mathbf{d}_j$$

$$\alpha_n = \frac{3}{8} + \left( \frac{3}{8} + \frac{1}{4} \cos \frac{2\pi}{n} \right)^2$$

# Loop Subdivision

La scelta di  $\beta$  non è unica

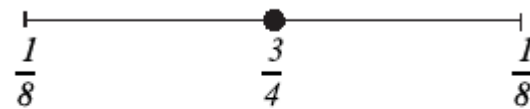
$$\beta = \frac{1}{k} \left( 5/8 - \left( \frac{3}{8} + \frac{1}{4} \cos \frac{2\pi}{k} \right)^2 \right)$$



# Loop Border

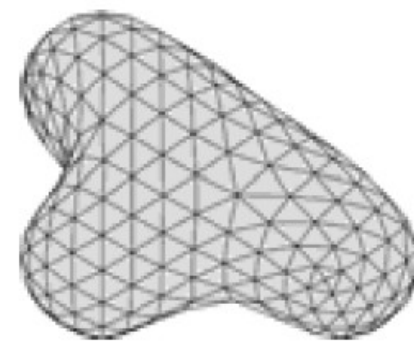
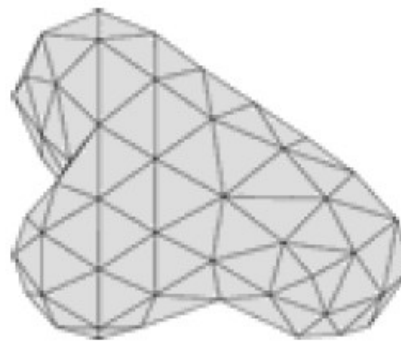
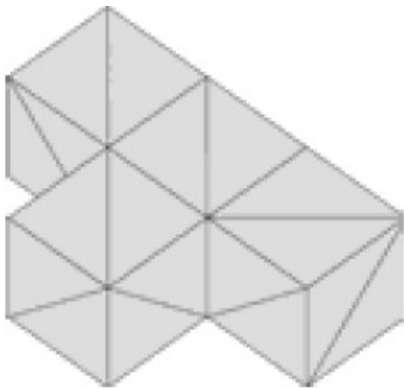


*Crease and boundary*



*a. Masks for odd vertices*

*b. Masks for even vertices*



# Butterfly

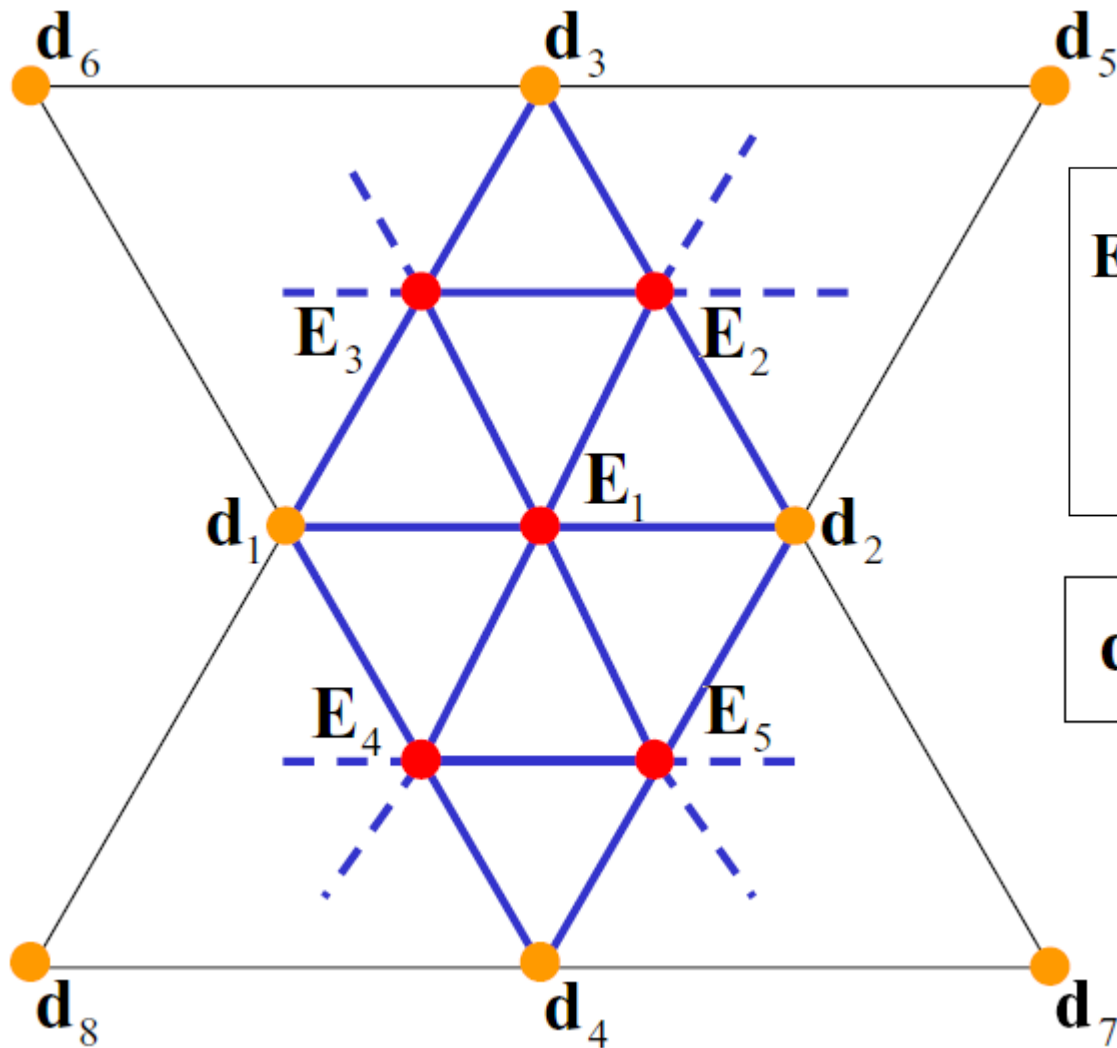
Interpolante (non approssimante)

Continua

C0 su vertici straordinari (valenza  $<4$  e  $>7$ )

C1 elsewhere

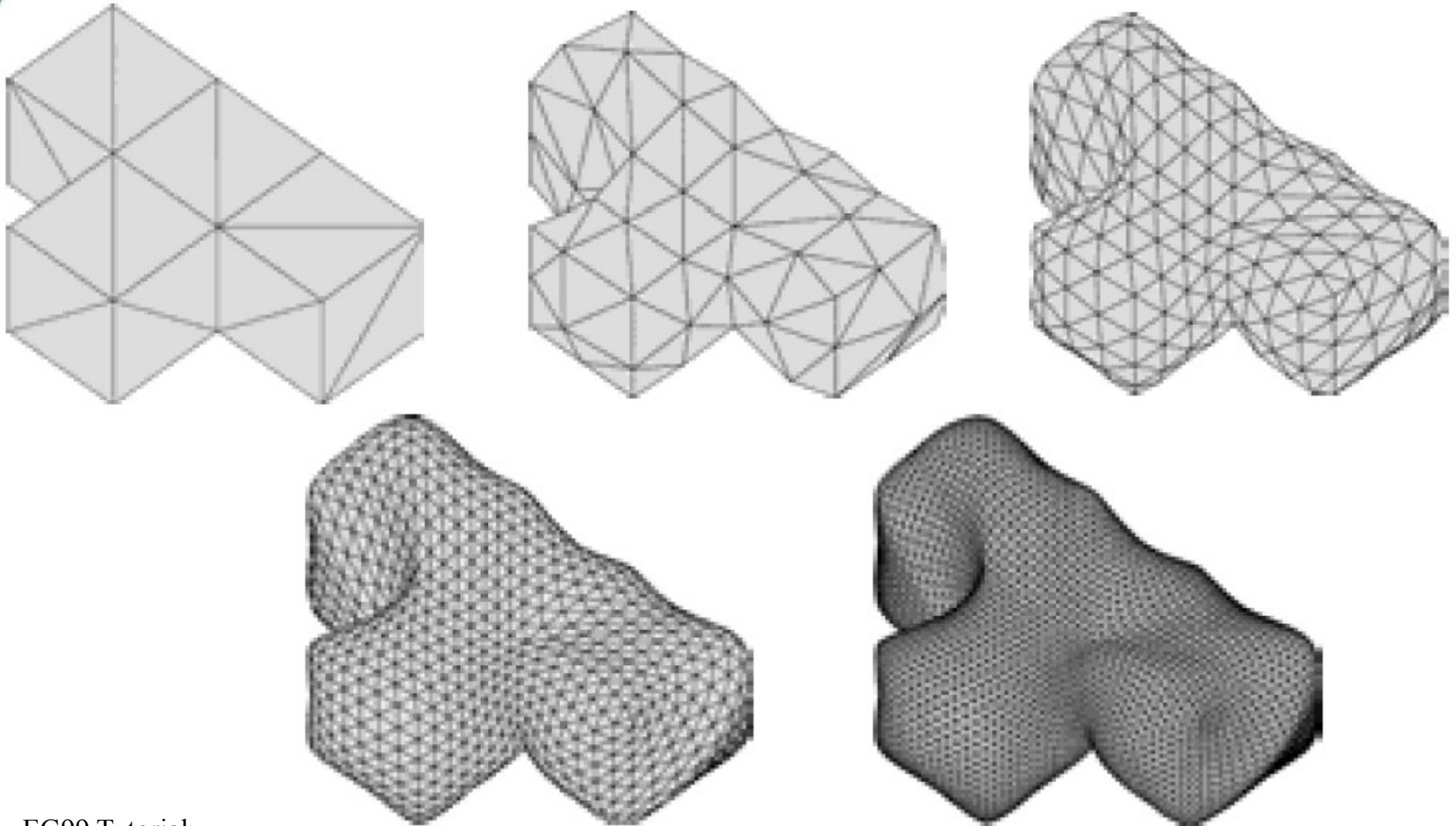
# Butterfly



$$\mathbf{E}_1 = \frac{1}{2}(\mathbf{d}_1 + \mathbf{d}_2) + \omega(\mathbf{d}_3 + \mathbf{d}_4) - \frac{\omega}{2}(\mathbf{d}_5 + \mathbf{d}_6 + \mathbf{d}_7 + \mathbf{d}_8)$$

$$\mathbf{d}'_i = \mathbf{d}_i$$

# Butterfly Subdivision



# Simplification and Refinement

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<http://vcg.isti.cnr.it/~cignoni>



# Approximating surfaces with triangle meshes

## ***Managing the quantity of geometry vs the quality of the representation***

### **Assumptions:**

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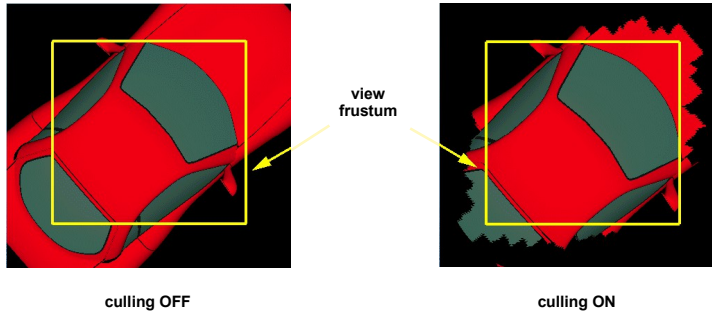
# Reducing rendering cost

Simplifying is not the only way to reduce the needed number of triangles  
At rendering time we can do:

## **View Frustum Culling**

Drawing only what is in the view frustum

...



# Reducing Complexity

...

## Occlusion/Visibility culling

Drawing only what it is probably visible

➔ *Improve performances, but not sufficient for very large datasets*



# Simplification

From a computational Point of view, fixed the size of the solution find the **best** representation

The optimal solution in its general form is NP Hard

Heuristics works well

(even with bounds)

# Metrics

Computing the “difference” between two meshes

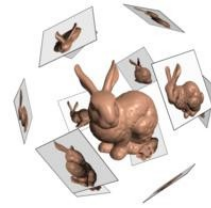
Geometric way

Hausdorff

Perceptual

Working in Image space

Many renderings,  
computing the difference,  
possibly in a human-like way



# Appearance similarity

Difference between two images: (trivial)

$$D(I_1, I_2) = \frac{1}{n^2} \sum_x \sum_y d(I_1(x, y), I_2(x, y))$$

Difference between two objects:

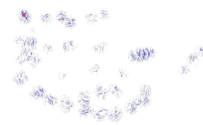
Integrate the above over all possible views



$I_1$



$I_2$



$I_1 - I_2$

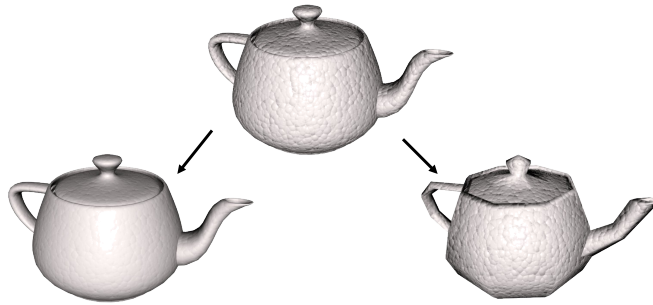
# Approximation error

Quantifies the notion of “similarity”

Two kinds of similarity:

**Geometric similarity** (surface deviation)

**Appearance similarity** (material, normal...)



8

The primary goal of simplification is **to generate an approximation which resembles as much as possible** the original.

# Geometric similarity

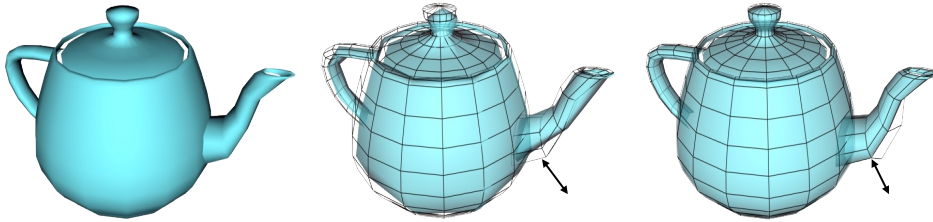
Two main components:

Distance function

Function Norm:

$L_2$ : average deviation

$L_{inf}$ : maximum deviation - Hausdorff distance

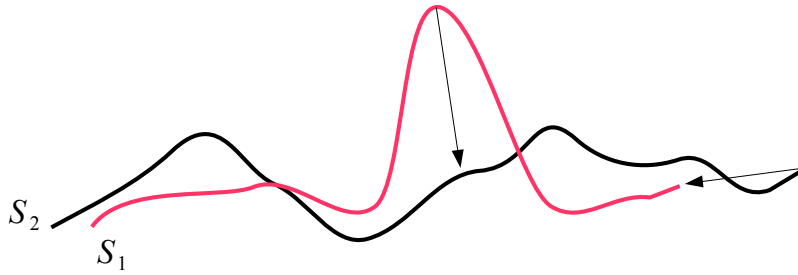


[Image by C. Andujar]



# Hausdorff Distance

$$D_H(S_1, S_2) = \max_{x \in S_1} (\min_{y \in S_2} D(x, y))$$



Symmetric version:

$$D(S_1, S_2) = \max\{D_H(S_1, S_2), D_H(S_2, S_1)\}$$

# Computing Hausdorff Distance

In an approximate way

Sample a surface

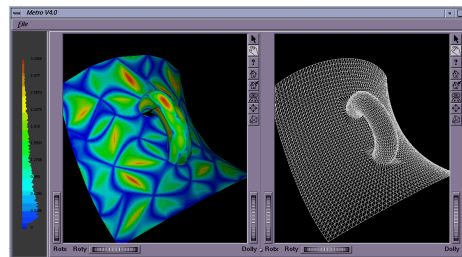
Points uniformly distributed over it

For each point  $x$  compute  $\min_{y \in \mathcal{S}_2} D(x, y)$

points usually very near to surface,

UG works well, better than trees

Avg are useful too...



# Simplification Algorithms

## Simplification approaches:

**incremental methods** based on local updates

mesh decimation

[Schroeder et al. '92, ... + others]

energy function optimization & progressive meshes

[Hoppe et al. '93, Hoppe '96, Hoppe '97]

**quadric error metrics**

[Garland et al. '97]

**coplanar facets merging**

[Hinker et al. '93, Kalvin et al. '96]

**clustering**

[Rossignac et al. '93, ... + others]

**wavelet-based**

[Eck et al. '95, + others]

**volume-based**

[He+ 95, He+ 96, Andujar+ 02]

# Incremental methods based on *local updates*

**All of the methods such that :**

simplification proceeds as a sequence of small  
changes of the mesh (in a greedy way)

each update reduces mesh size and  
[~monotonically] decreases the approximation  
precision

**Different approaches:**

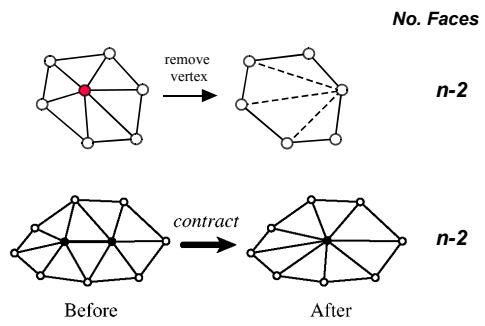
mesh decimation  
energy function optimization  
progressive meshes  
quadric error metrics

... Incremental methods based on **local updates** ...

**Local update actions:**

vertex removal

edge collapse  
preserve location  
new location



# Incremental Simp Method

The common greedy framework:

**While** *mesh size/precision is satisfactory*

**Select** from a priority queue the best element to simplify

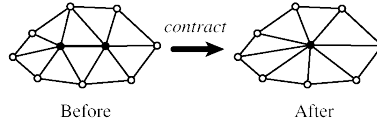
**Perform simplification** modifying the mesh

**Ppdate** priority queue according to the mods;

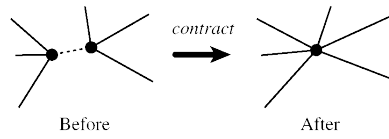
# Simplification using Quadric Error Metrics

[Garland et al. Sig'97]

Based on incremental  
**edge-collapsing**



**but** can also collapse vertex couples  
which are **not connected**  
(topology is not preserved)



## Quadric Error for Surfaces

Let  $\mathbf{n}^T \mathbf{v} + d = 0$  be the equation representing a plane

The squared distance of a point  $\mathbf{x}$  from the plane is

$$D(\mathbf{x}) = \mathbf{x}(\mathbf{nn}^T)\mathbf{x} + 2d\mathbf{n}^T\mathbf{x} + d^2$$

This distance can be represented as a quadric

$$Q = (\mathbf{A}, \mathbf{b}, c) = (\mathbf{nn}^T, d\mathbf{n}, d^2)$$

$$Q(\mathbf{x}) = \mathbf{xAx} + 2\mathbf{b}^T\mathbf{x} + c$$



## Quadric

The error is estimated by providing for each vertex  $v$  a quadric  $Q_v$  representing the sum of all the squared distances from the faces incident in  $v$

The error of collapsing an edge  $e=(v,w)$  can be evaluated as  $Q_w(v)$ .

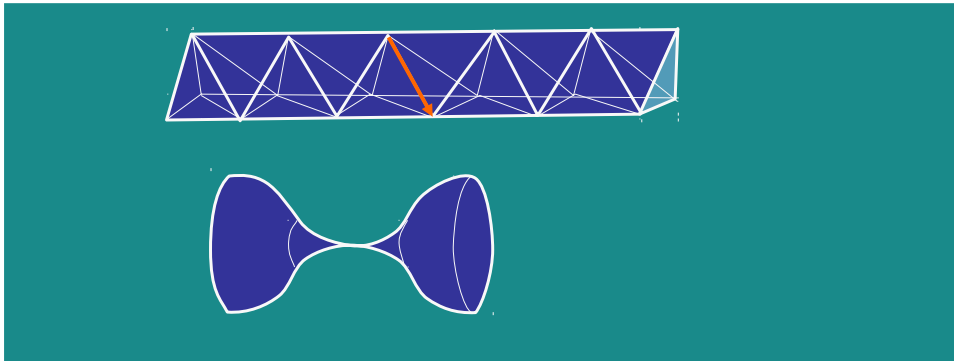
After the collapse the quadric of  $v$  is updated as follow  $Q_v = Q_v + Q_w$

# Topology Preservation

## 2-Manifold

A surface  $\Sigma$  in  $\mathbf{R}^2$  such that any point on  $\Sigma$  has an open neighborhood homeomorphic to an open disc or to half an open disc in  $\mathbf{R}^2$

A edge collapse can create non manifold situations

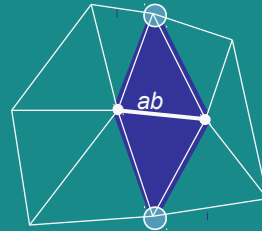
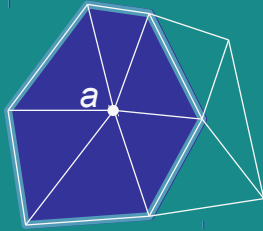


# Topology Preservation

Let  $\Sigma$  be a 2 simplicial complex **without boundary**

$\Sigma'$  is obtained by collapsing the edge  $e = (ab)$

Let  $Lk(\sigma)$  be the set of all the faces of the co-faces of  $\sigma$  disjoint from  $\sigma$



$\Sigma$  and  $\Sigma'$  are homeomorphic **iff**

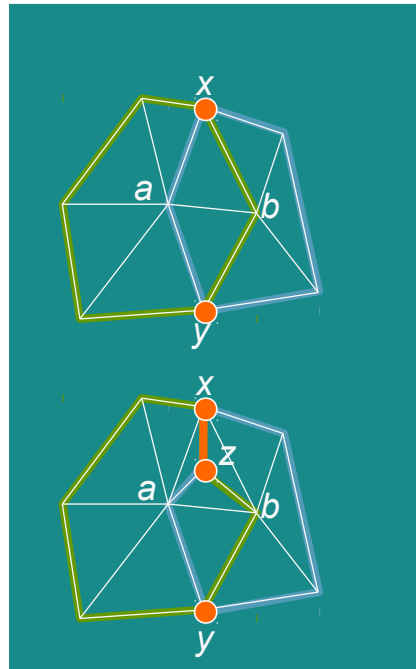
$$Lk(a) \cap Lk(b) = Lk(ab)$$

[Dey 99]

## Topology Preservation

$$Lk(a) \cap Lk(b) = \{x, y\} = Lk(ab)$$

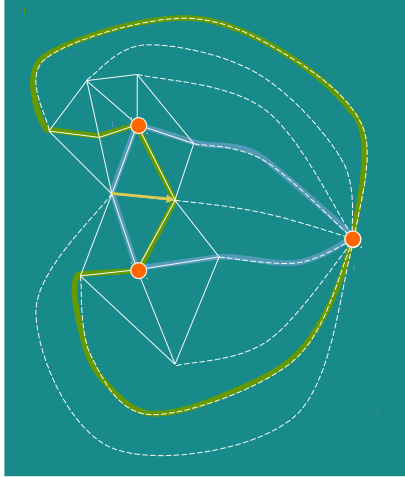
$$Lk(a) \cap Lk(b) = \{x, y, z, zx\} \neq \{y, z\} = Lk(ab)$$



## Topology Preservation

Mesh with boundary can be managed by considering a dummy vertex  $v_d$  and, for each boundary edge  $e$  a dummy triangle connecting  $e$  with  $v_d$

Think it wrapped on the surface of a sphere

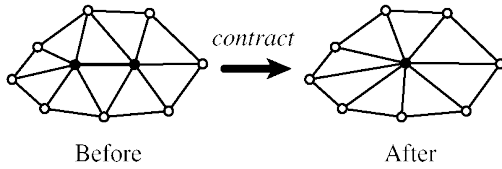


## Lazy heap

Si suppone di avere uno heap con tutte le operazioni

Estraggo da heap e aggiorno la mesh

tali operazioni invalidano/modificano la mesh e quindi le priorità/validità di parte delle azioni già presenti nello Heap



# Lazy Heap

## Due Soluzioni

Link espliciti elementi mesh->heap e  
aggiornamento dello stesso

## Lazy update

Si mettono nello heap tutte le nuove operazioni  
con la nuova priorità

Quando si estrae un'op dall heap si controlla che  
sia sempre valida

Di tanto in tanto garbage collection sullo heap

# Validità collasso

## Dati

Ogni vertice ha una marca temporale:

quando e' stato modificato l'ultima volta

Ogni collasso (coppia di vertici) ha una marca temporale

quando è stato inserito nello heap

## Un collasso è valido se

I due vertici non sono stati cancellati

Il collasso e' stato messo nello heap piu recentemente della data di ultima modifica dei vertici



# Simplification Algorithms

Not-incremental methods:

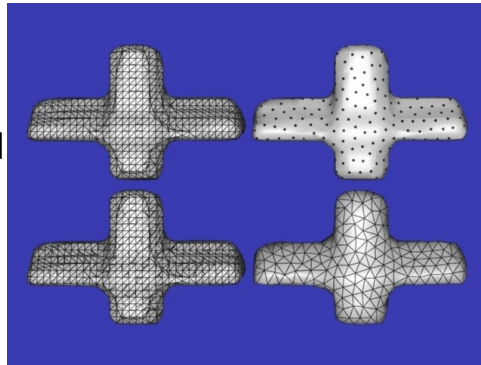
coplanar facets merging  
[Hinker et al. '93, Kalvin et al. '96]

re-tiling  
[Turk '92]

clustering  
[Rossignac et al. '93, ... + others]

wavelet-based  
[Eck et al. '95]

volume-based  
[He+ 95, He+ 96, Andujar+ 02]



# Clustering

## *Vertex Clustering*

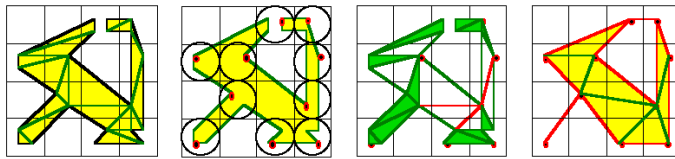
[Rossignac, Borrel '93]

detect and unify **clusters** of nearby vertices (discrete gridding and coordinates truncation)

all faces with two or three vertices in a cluster are removed

does not preserve topology (faces may degenerate to edges, genus may change)

approximation depends on grid resolution



(figure by Rossignac)

# Clustering -- Examples

Simplification of a table lamp,  
Interaction Accelerator

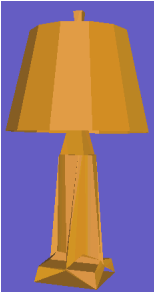
IBM 3D



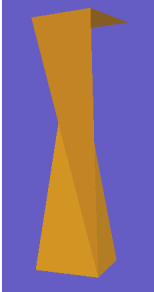
10,108 facets



1,383 facets



474 facets



46 facets

# ... Clustering...

## **Clustering - *Evaluation***

- high efficiency
- very simple implementation
- low quality approximations
- does not preserve topology
- error is bounded by the grid cell size

REFINEMENT  
or  
Subdivision Surfaces

# Subdivision Surfaces

*Subdivision defines a smooth curve or surface as the limit of a sequence of successive refinements.*

Si parte da una mesh poligonale

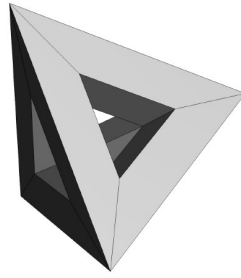
Si suddivide i poligoni che la compongono

Smooth della superficie muovendo i vertici

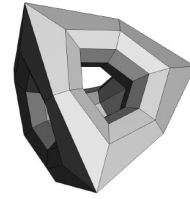
In effetti quello che si vede sono sempre  
approx delle vere subdiv surfaces



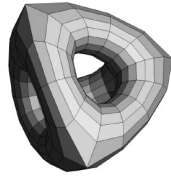
EG99 Tutorial



(a)



(b)



(c)



(d)

# Esempio

## Geri's Game (1997)

Primo esempio non  
accademico di uso di  
subdivision surfaces



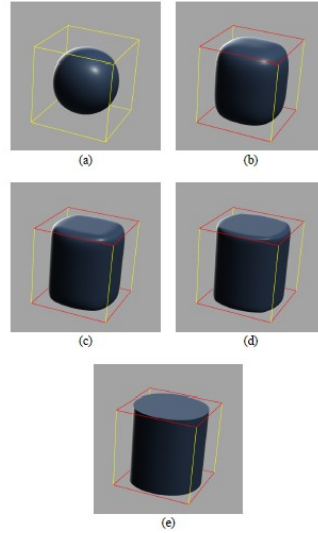
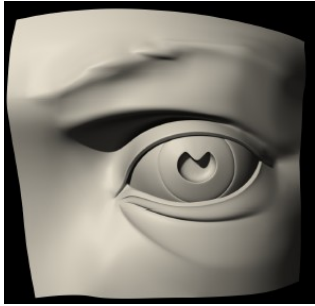
EG99 Tutorial





# Smoothness

Non solo superfici smooth  
Variable sharpness creases

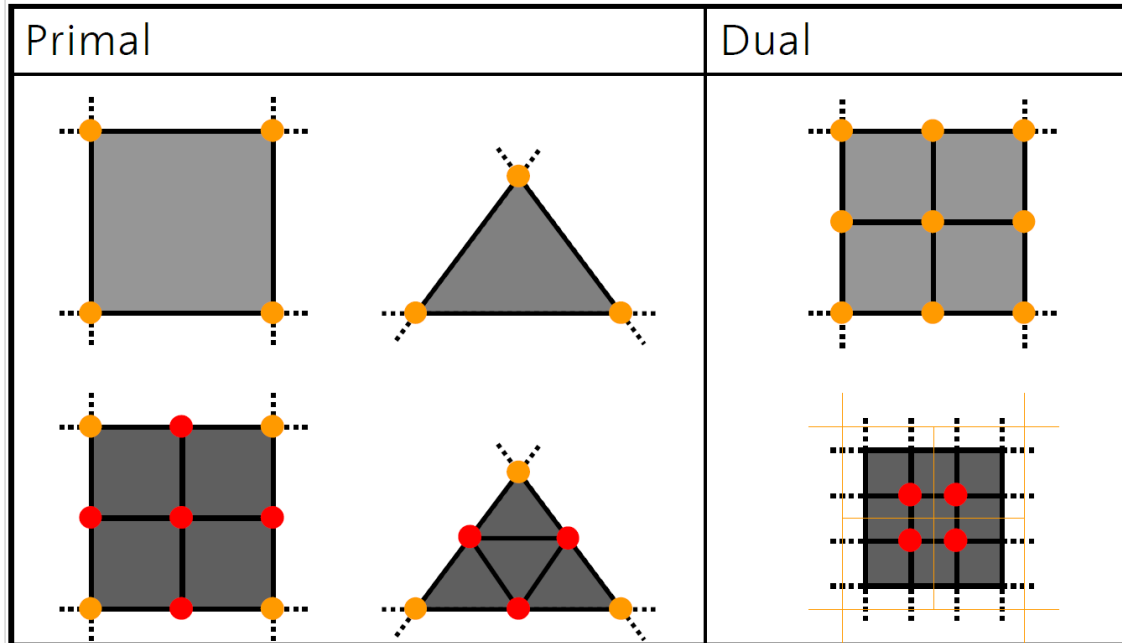


# Subdivision Classification

<b>Primal</b>	Faces are split into sub-faces
<b>Dual</b>	Vertices are split into multiple vertices

<b>Approximating</b>	Control points not interpolated
<b>Interpolating</b>	Control points are interpolated

# Subdivision Classification



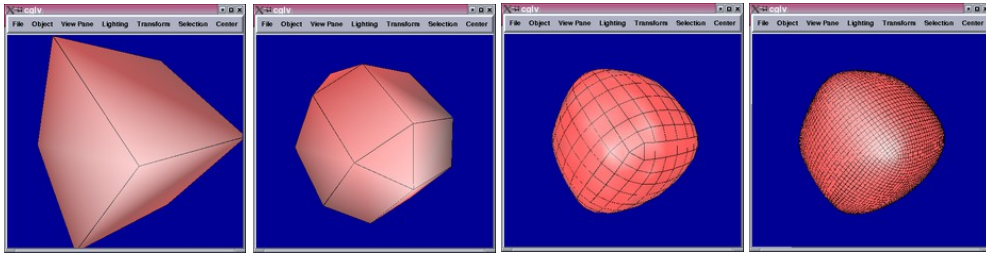
# Subdivision

	<i>Primal</i>		<i>Dual</i>
	Triangles	Rectangles	
<b>Approximating</b>	Loop	Catmull-Clark	Doo-Sabin
<b>Interpolating</b>	Butterfly	Kobbelt	Midedge

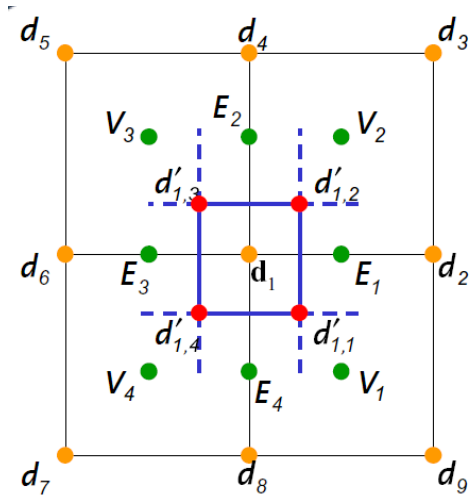
# Doo Sabin

Per mesh poligonali

Duale ad ogni vertice corrisponde una nuova faccia



# Doo Sabin



$$V_2 = \frac{1}{n} \cdot \sum_{j=1}^n d_j$$

$$E_i = \frac{1}{2} (d_1 + d_{2i})$$

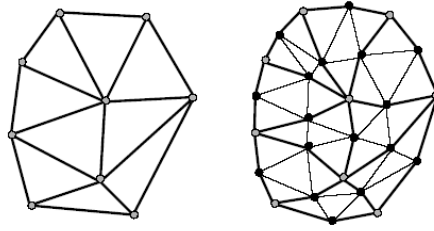
$$d'_{1,j} = \frac{1}{4} (d_1 + E_j + E_{j-1} + V_j)$$

# Loop Refinement Scheme

Funziona per mesh triangolari

Vertex insertion

Ogni edge è diviso in due e i nuovi vertici sono riconnessi per formare nuovi triangoli



# Loop Subdivison

Approssimante (non interpolante)

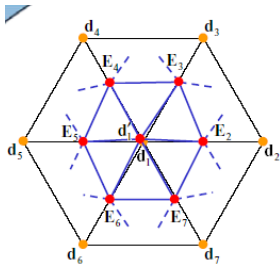
Continua

C1 su vertici straordinari (valenza  $\neq 6$ )

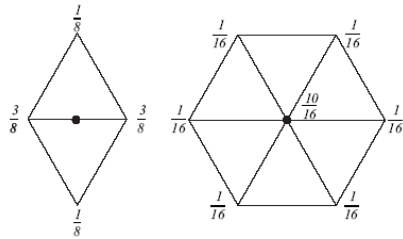
C2 elsewhere



# Loop Subdivision



$$E_i = \frac{3}{8}(d_1 + d_i) + \frac{1}{8}(d_{i-1} + d_{i+1})$$



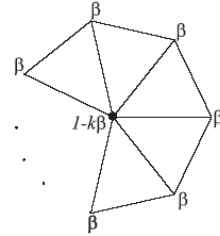
$$\mathbf{d}'_1 = \alpha_n \mathbf{d}_1 + \frac{(1 - \alpha_n)^{n+1}}{n} \sum_{j=2}^{n+1} \mathbf{d}_j$$

$$\alpha_n = \frac{3}{8} + \left( \frac{3}{8} + \frac{1}{4} \cos \frac{2\pi}{n} \right)^2$$

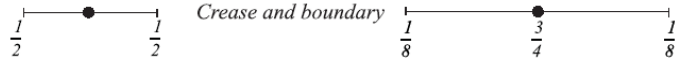
# Loop Subdivision

La scelta di B non è unica

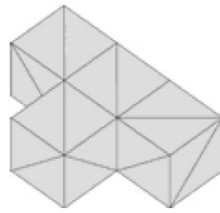
$$\beta = \frac{1}{k} \left( 5/8 - \left( \frac{3}{8} + \frac{1}{4} \cos \frac{2\pi}{k} \right)^2 \right)$$



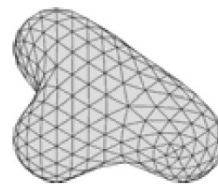
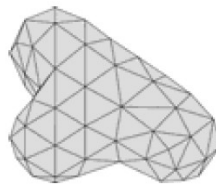
# Loop Border



a. Masks for odd vertices



b. Masks for even vertices



# Butterfly

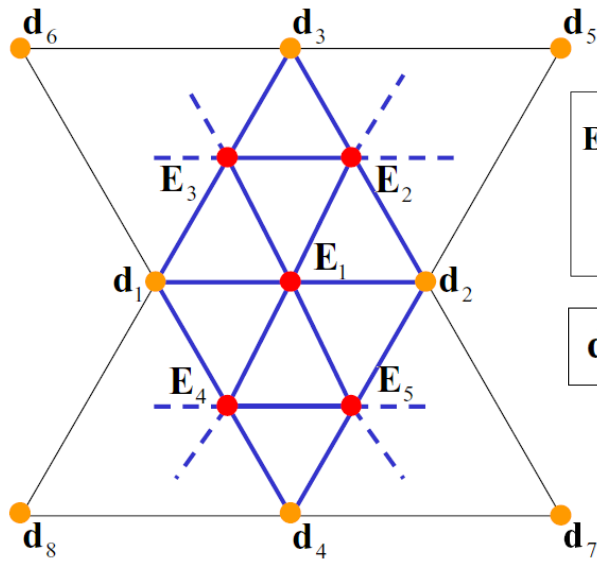
Interpolante (non approssimante)

Continua

C0 su vertici straordinari (valenza  $<4$  e  $>7$ )

C1 elsewhere

# Butterfly



$$\mathbf{E}_1 = \frac{1}{2}(\mathbf{d}_1 + \mathbf{d}_2) + \omega(\mathbf{d}_3 + \mathbf{d}_4) - \frac{\omega}{2}(\mathbf{d}_5 + \mathbf{d}_6 + \mathbf{d}_7 + \mathbf{d}_8)$$

$$\mathbf{d}'_i = \mathbf{d}_i$$

# Butterfly Subdivision

