

Mesh Generation

Per-Olof Persson
persson@berkeley.edu

Department of Mathematics
University of California, Berkeley

Math 228B Numerical Solutions of Differential Equations

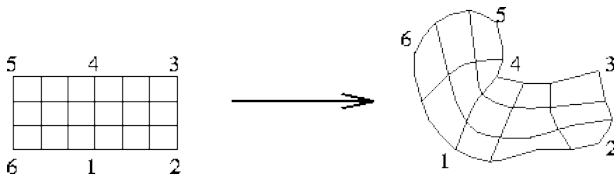
Structured Mesh Generation

Why Structured Meshes?

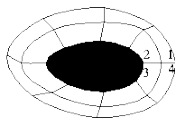
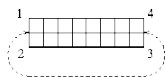
- Lead to very efficient numerical methods
- High quality for sufficiently simple geometries
- Larger grid control when high anisotropy is required
- Multi-block approach allows for realistic geometries

Single-Block Grid Generation

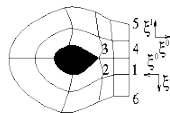
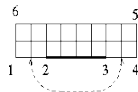
- Construct a one-to-one mapping between a rectangular computational domain and a physical domain
- Ideally, grid size in physical space should be dictated by solver/solution requirements
- Ensure grid quality e.g. smoothness, orthogonality



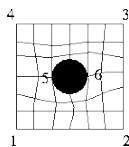
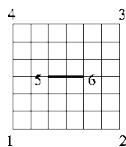
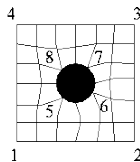
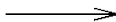
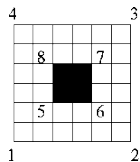
Single-Block Grid Common Topologies



O-Grid



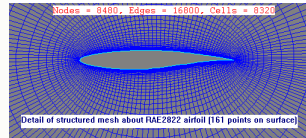
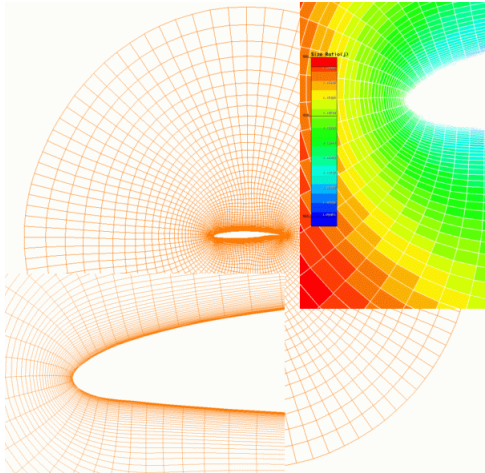
C-Grid



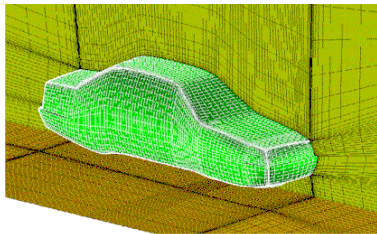
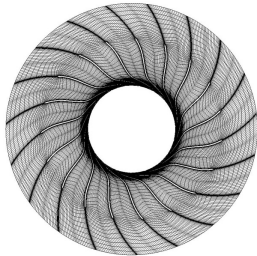
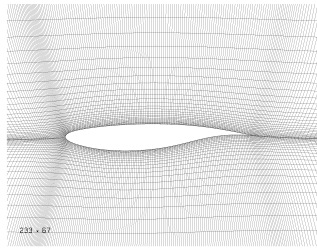
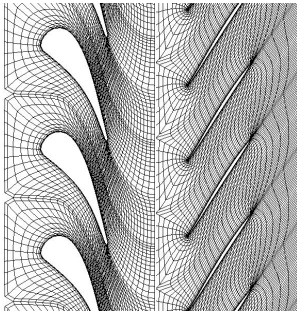
H-Grids

... plus combinations

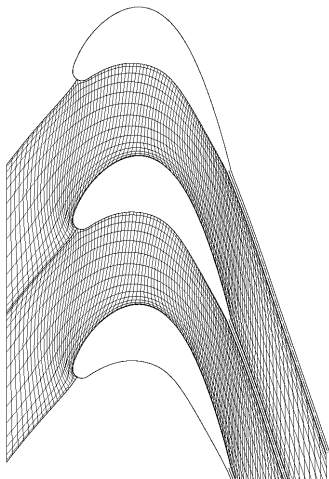
Examples: Single-Block O-Grids



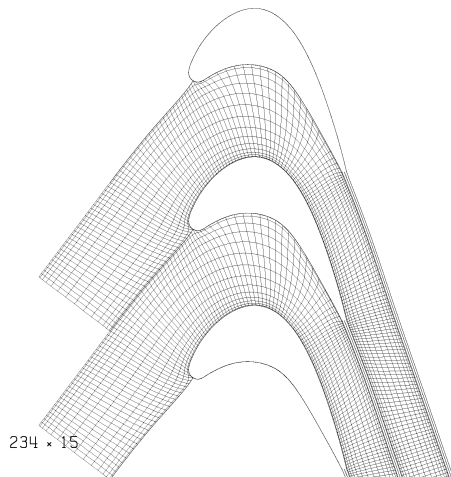
Examples: Single-Block C,H-Grids



Examples: H-Grids



H-Grid

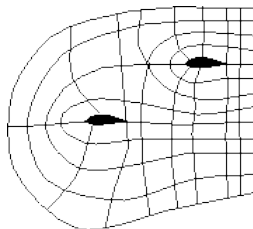


234 × 15

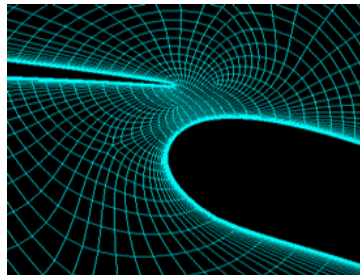
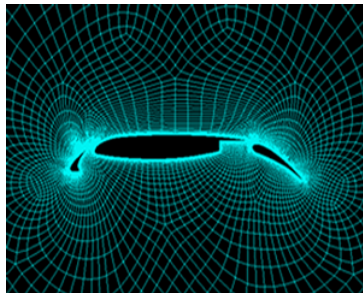
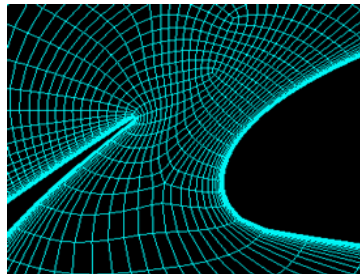
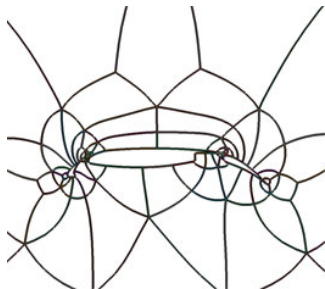
H-Grid/I-Grid

Multi-Block Grid Generation

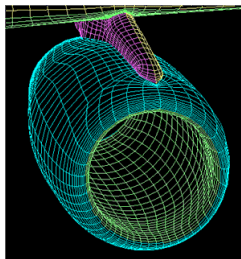
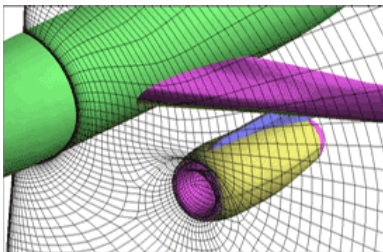
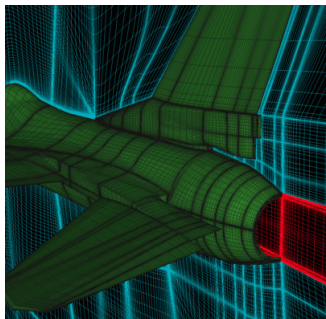
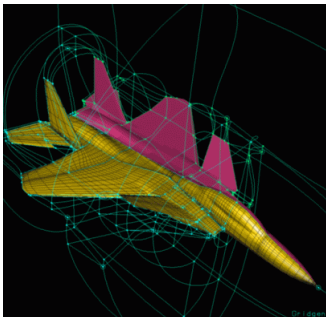
- Subdivide domain into an unstructured assembly of quadrilaterals/hexahedra
- Obtaining block topology automatically is hard
- Obtaining block geometry automatically (e.g. point coordinates) once topology is known is tractable



Examples: Multi-Block Grids

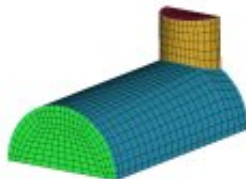
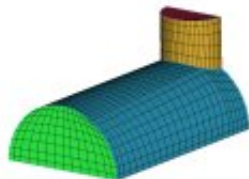
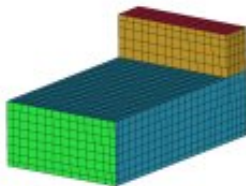
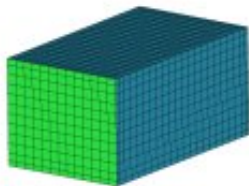


Examples: Multi-Block Grids



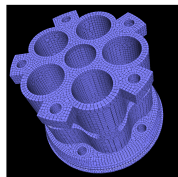
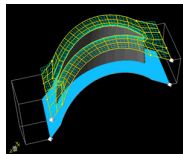
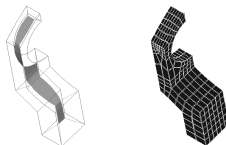
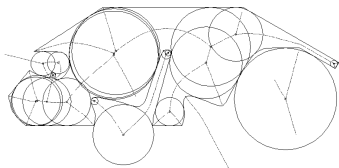
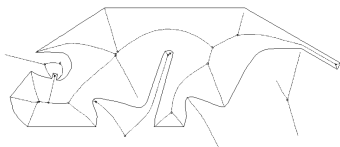
Block Topology Generators

(from ICEM CFD)



Automatic $H \Rightarrow O$ conversion

Block Topology Generators - Medial Axis Transform (MAT)



Single Block Grid Generation - Creating the Mapping

- Conformal Mapping
- Transfinite Interpolation
- Solving PDE's
 - Elliptic
 - Parabolic/Hyperbolic

Conformal Mapping

- Any function $\alpha = f(z)$ such that $\frac{df}{dz} \neq 0$ defines a one-to-one (conformal) mapping between $z = x + iy$ and $\alpha = \xi + i\eta$, or between (x, y) and (ξ, η) .
- The functions $\xi(x, y)$ and $\eta(x, y)$ satisfy the Cauchy-Riemann equations (e.g. $\xi_x = \eta_y$, and $\eta_x = -\xi_y$) and as a consequence, they are harmonic

$$\nabla^2 \xi = 0, \quad \nabla^2 \eta = 0 \quad (\text{smoothness})$$

- Preserve angles (grid orthogonality)
- Preserve ratios
- Lead to high quality grids
- Limited to 2D**

Conformal Mapping Transformations

- Jukowski (maps circle of radius c to segment $[-2c, 2c]$)

$$\alpha = z + \frac{c^2}{z}, \quad \text{or} \quad \frac{\alpha + 2c}{\alpha - 2c} = \left(\frac{z + c}{z - c} \right)^2$$

- Karman-Trefftz

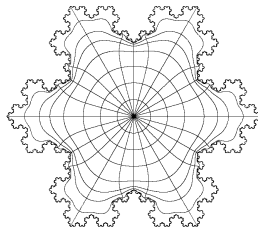
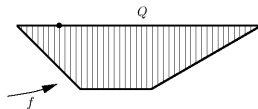
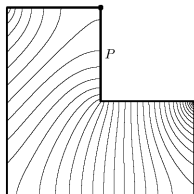
$$\frac{\alpha + 2c}{\alpha - 2c} = \left(\frac{z + c}{z - c} \right)^n$$

- Schwarz-Christoffel (maps polygon into half plane)

$$\frac{d\alpha}{dz} = K \prod_{k=1}^n \left(1 - \frac{z}{z_k} \right)^{\beta_k}$$

Conformal Mapping - Schwarz-Christoffel

Ref. "Schwarz-Christoffel Mapping", *Driscoll and Trefethen*, Cambridge University Press, 2002.



Algebraic Mappings

- Construct a mapping between the boundaries of the unit square (cube) and the boundaries of an "arbitrary" region which is topologically equivalent
- Combine 1D interpolants using Boolean sums to construct mapping - Transfinite Interpolation (TFI)
- Not guaranteed to be one-to-one
- Orthogonality not guaranteed
- Very Fast
- Quite General
- **Grid quality not always assured**

Algebraic Mappings - 1D Interpolants

- General 1D interpolant of $f(x)$ for $x \in (0, 1)$

$$\hat{f}(x) \equiv \Pi_x f = \sum_{i=0}^L \sum_{n=0}^P \alpha_i^n(x) \left. \frac{d^n f}{dx^n} \right|_{x=x_i}$$

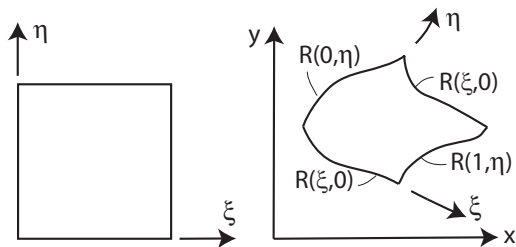
- $\alpha_i^n(x)$ are the blending functions
- Examples
 - Quadratic Lagrange interpolation - $P = 0, L = 2$

$$\Pi_x f = (2x^2 - 3x + 1)f(0) + (4x - 4x^2)f(0.5) + (2x^2 - x)f(1)$$

- Hermite interpolation - $P = 1, L = 2$

$$\begin{aligned} \Pi_x f = & (1 - 3x^2 - 2x^3)f(0) + (x - 2x^2 + x^3)f(1) + \\ & x^2(3 - 2x)^2 f'(0) + x^2(x - 1)f'(1) \end{aligned}$$

Algebraic Mappings - Transfinite Interpolation



- Start from 1D boundary mappings of $\mathbf{R} \equiv (x, y)$, e.g.
 $\mathbf{R}(\xi, 0), \mathbf{R}(\xi, 1), \mathbf{R}(0, \eta), \mathbf{R}(1, \eta)$
- Construct 1D interpolants in the ξ and η directions (e.g. linear)

$$\Pi_x \mathbf{R} = (1 - \xi) \mathbf{R}(0, \eta) + \xi \mathbf{R}(1, \eta)$$

$$\Pi_\eta \mathbf{R} = (1 - \eta) \mathbf{R}(\xi, 0) + \eta \mathbf{R}(\xi, 1)$$

Algebraic Mappings - Transfinite Interpolation

- Construct two-dimensional interpolant by doing the Boolean sum

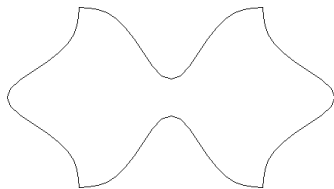
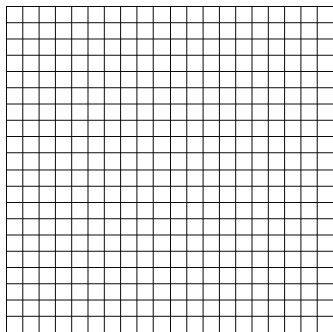
$$\hat{\mathbf{R}}(\xi, \eta) = (\Pi_\xi \oplus \Pi_\eta)\mathbf{R} = (\Pi_\xi + \Pi_\eta - \Pi_\xi \Pi_\eta)\mathbf{R}$$

Expanding

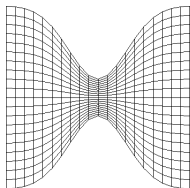
$$\begin{aligned}\hat{\mathbf{R}}(\xi, \eta) = & (1-\xi, \xi) \begin{pmatrix} \mathbf{R}(0, \eta) \\ \mathbf{R}(1, \eta) \end{pmatrix} + (\mathbf{R}(\xi, 0), \mathbf{R}(\xi, 1)) \begin{pmatrix} 1-\eta \\ \eta \end{pmatrix} \\ & - (1-\xi, \xi) \begin{pmatrix} \mathbf{R}(0, 0) & \mathbf{R}(0, 1) \\ \mathbf{R}(1, 0) & \mathbf{R}(1, 1) \end{pmatrix} \begin{pmatrix} 1-\eta \\ \eta \end{pmatrix}\end{aligned}$$

- Extends to **general 1D interpolants** and **any dimension**

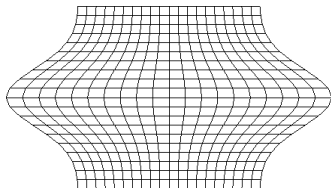
Algebraic Mappings - Example



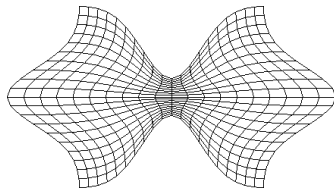
Algebraic Mappings - Example



$$\Pi_\xi \mathbf{R}$$



$$\Pi_\eta \mathbf{R}$$



$$(\Pi_\xi \oplus \Pi_\eta) \mathbf{R}$$

- Use non-regular subdivisions in (ξ, η) (e.g. exponential functions) to obtain desired element sizes in (x, y)
- Use derivative boundary conditions to enforce boundary orthogonality

$$\frac{\partial \mathbf{R}}{\partial \xi} \cdot \frac{\partial \mathbf{R}}{\partial \eta} = 0$$

- Construct mapping by solving a PDE
 - Elliptic Equations (smooth grids)

$$\nabla^2 \xi(x, y) = P(x, y), \quad \nabla^2 \eta(x, y) = Q(x, y)$$

- Hyperbolic equations (orthogonal grids)

$$x_\xi y_\eta - x_\eta y_\xi = J \quad (\text{size control})$$

$$x_\xi x_\eta + y_\xi y_\eta = 0 \quad (\text{orthogonality})$$

- Most widely used approach
- **Grids usually have high quality**

Elliptic Grid Generation

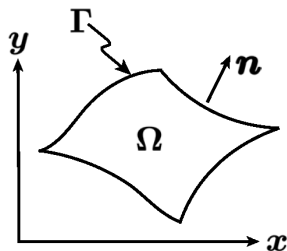
We are interested in solving

$$-\nabla^2 \xi = P \quad \text{in } \Omega$$

$$\xi = g \quad \text{on } \Gamma_D$$

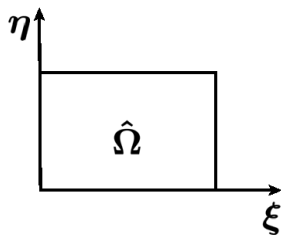
$$\frac{\partial \xi}{\partial n} = h \quad \text{on } \Gamma_N = \Gamma \setminus \Gamma_D$$

where P , g , and h are given.

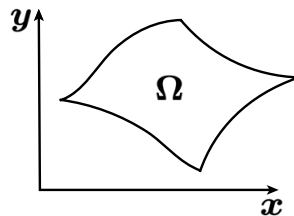


Similarly for $\eta(x, y)$

Elliptic Grid Generation



$$\begin{aligned} x &= x(\xi, \eta) \\ y &= y(\xi, \eta) \\ &\longrightarrow \end{aligned}$$



? ?

$$\nabla^2 \xi = P$$

Can we determine an equivalent problem to be solved on $\hat{\Omega}$?

Elliptic Grid Generation

$$\xi = \xi(x, y)$$

$$\eta = \eta(x, y)$$

$$x = x(\xi, \eta)$$

$$y = y(\xi, \eta)$$

$$\begin{pmatrix} d\xi \\ d\eta \end{pmatrix} = \begin{pmatrix} \xi_x & \xi_y \\ \eta_x & \eta_y \end{pmatrix} \begin{pmatrix} dx \\ dy \end{pmatrix} \quad \begin{pmatrix} dx \\ dy \end{pmatrix} = \begin{pmatrix} x_\xi & x_\eta \\ y_\xi & y_\eta \end{pmatrix} \begin{pmatrix} d\xi \\ d\eta \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} \xi_x & \xi_y \\ \eta_x & \eta_y \end{pmatrix} = \begin{pmatrix} x_\xi & x_\eta \\ y_\xi & y_\eta \end{pmatrix}^{-1} = \frac{1}{J} \begin{pmatrix} y_\eta & -x_\eta \\ -y_\xi & x_\xi \end{pmatrix}$$

$$J = x_\xi y_\eta - x_\eta y_\xi$$

Elliptic Grid Generation

$$\begin{aligned}\xi_x &= \frac{y_\eta}{J} & \xi_y &= -\frac{x_\eta}{J} \\ \eta_x &= -\frac{y_\xi}{J} & \eta_y &= \frac{x_\xi}{J}\end{aligned}$$

and

$$\begin{aligned}\xi_{xx} &= \frac{\partial}{\partial x} (\xi_x) = \left(\xi_x \frac{\partial}{\partial \xi} + \eta_x \frac{\partial}{\partial \eta} \right) \left(\frac{y_\eta}{J} \right) \\ &= \frac{1}{J} \left(y_\eta \frac{\partial}{\partial \xi} - y_\xi \frac{\partial}{\partial \eta} \right) \left(\frac{y_\eta}{J} \right) \\ &= \dots \\ \xi_{yy} &= \dots\end{aligned}$$

Elliptic Grid Generation - Thompson's Equations

Finally, $\xi_{xx} + \xi_{yy} = 0$ and $\eta_{xx} + \eta_{yy} = 0$, become

$$\begin{array}{rcl} ax_{\xi\xi} - 2bx_{\xi\eta} + cx_{\eta\eta} & = & 0 \\ ay_{\xi\xi} - 2by_{\xi\eta} + cy_{\eta\eta} & = & 0 \end{array}$$

a, b, c depend on the mapping.

$$a = x_{\eta}^2 + y_{\eta}^2 \quad b = x_{\xi}x_{\eta} + y_{\xi}y_{\eta} \quad c = x_{\xi}^2 + y_{\xi}^2$$

- These equations can be solved using **central finite differences** on a regular grid in the (ξ, η) domain to determine the (x, y) coordinates of each grid point (see notes from 16.920 for details).
- These equations are non-linear and are typically solved using an SOR (Successive Over-Relaxation) Method

Elliptic Grid Generation - Grid Control

Use, $\xi_{xx} + \xi_{yy} = P(x, y)$ and $\eta_{xx} + \eta_{yy} = Q(x, y)$,

$$\begin{array}{rcl} ax_{\xi\xi} - 2bx_{\xi\eta} + cx_{\eta\eta} & = & -J^2(x_{\xi}P + x_{\eta}Q) \\ ay_{\xi\xi} - 2by_{\xi\eta} + cy_{\eta\eta} & = & -J^2(y_{\xi}P + y_{\eta}Q) \end{array}$$

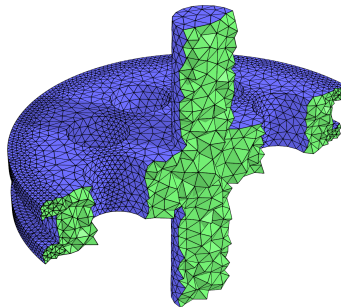
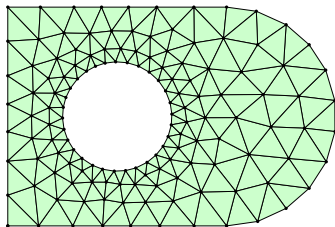
- The functions $P(\xi, \eta)$ and $Q(\xi, \eta)$ can be used to obtain grid control
- Derivative boundary conditions can be used to enforce grid orthogonality at the boundary

Ref. "Numerical Generation of Two-Dimensional Grids by Use of Poisson Equations with Grid Control", *Sorenson and Steger*, in Numerical Grid Generation Techniques, Smith, R.E. (Ed.), NASA-CP-2166, pp. 449-461, 1980

Unstructured Mesh Generation

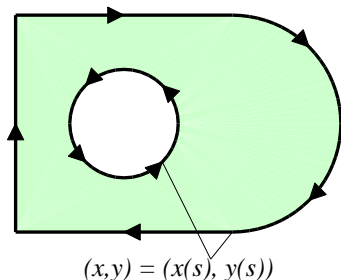
Unstructured Mesh Generation

- Approximate a domain in \mathbb{R}^d by simple geometric shapes
- Determine node points and element connectivity
- Goal: Resolve the domain accurately with well-shaped elements, but use as few elements as possible
- Applications: Numerical solution of PDEs (FEM, FVM, DGM, BEM), interpolation, computer graphics, visualization



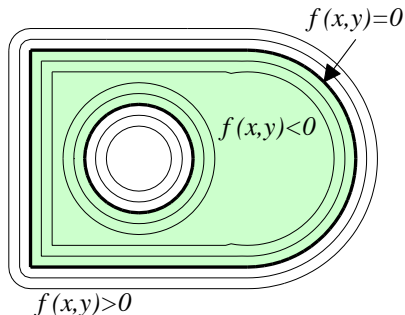
Explicit Geometry

- Parameterized boundaries



Implicit Geometry

- Boundaries from contour



Unstructured Meshing Algorithms

- **Delaunay refinement**

- Refine an initial triangulation by inserting centroid points and updating connectivities
- Efficient and robust, provably good in 2-D

- **Advancing front**

- Propagate a layer of elements from boundaries into domain, stitch together at intersection
- High quality meshes, good for boundary layers, but somewhat unreliable in 3-D

Unstructured Meshing Algorithms

- **Octree mesh**

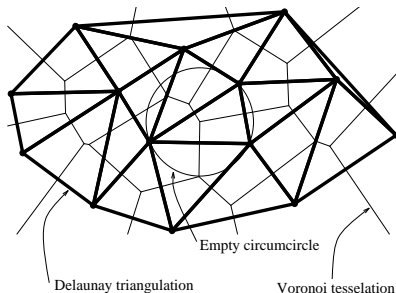
- Create an octree, refine until geometry well resolved, form elements between cell intersections
- Guaranteed quality even in 3-D, but poor element qualities

- **DistMesh**

- Improve initial triangulation by node movements and connectivity updates
- Easy to understand and use, handles implicit geometries, high element qualities, but non-robust and low performance

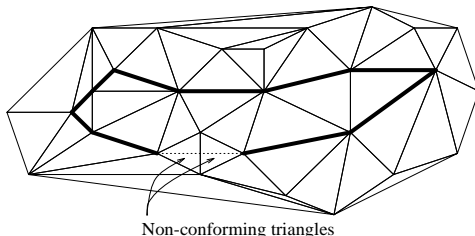
Delaunay Triangulation

- Find non-overlapping triangles that fill the convex hull of a set of points
- Properties:
 - Every edge is shared by at most two triangles
 - The circumcircle of a triangle contains no other input points
 - Maximizes the minimum angle of all the triangles

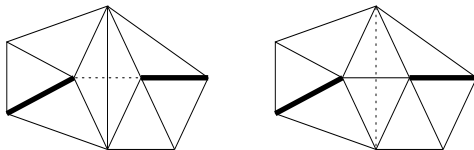


Constrained Delaunay Triangulation

- The Delaunay triangulation might not respect given input edges

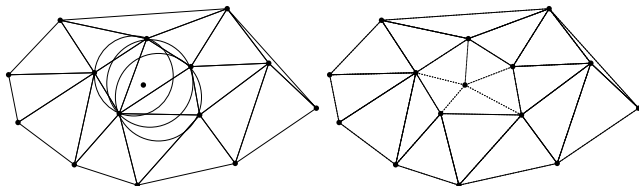


- Use local edge swaps to recover the input edges



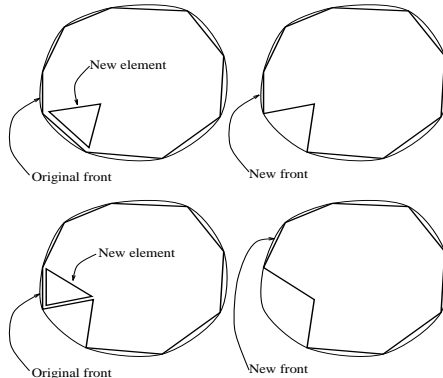
Delaunay Refinement Method

- Algorithm:
 - Form initial triangulation using boundary points and outer box
 - Replace an undesired element (bad or large) by inserting its circumcenter, retriangulate and repeat until mesh is good
- Will converge with high element qualities in 2-D
- Very fast – time almost linear in number of nodes



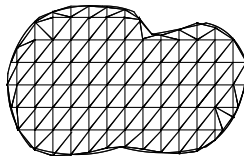
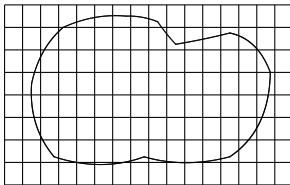
The Advancing Front Method

- Discretise the boundary as initial front
- Add elements into the domain and update the front
- When front is empty the mesh is complete

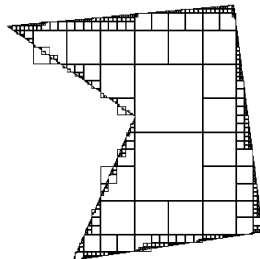


Grid Based and Octree Meshing

- Overlay domain with regular grid, crop and warp edge points to boundary

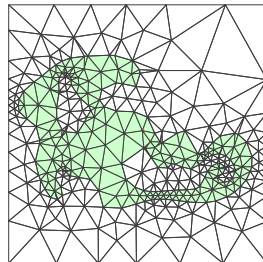
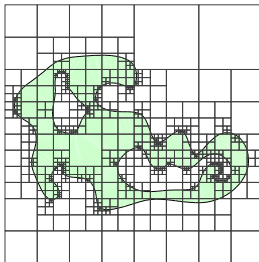
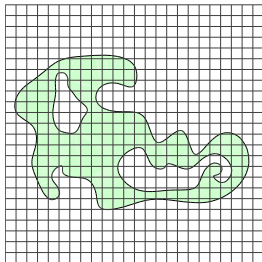


- Octree instead of regular grid gives graded mesh with fewer elements



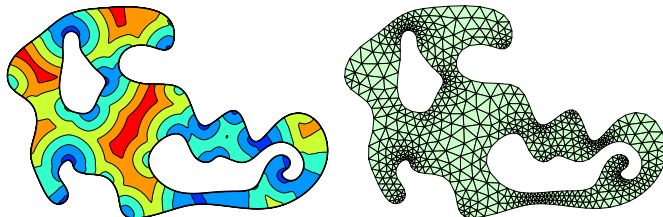
Mesh Size Functions

- Function $h(x)$ specifying desired mesh element size
- Many mesh generators need a priori mesh size functions
 - Physically-based methods such as DistMesh
 - Advancing front and Paving methods
- Discretize mesh size function $h(x)$ on a *background grid*



Mesh Size Functions

- Based on several factors:
 - Curvature of geometry boundary
 - Local feature size of geometry
 - Numerical error estimates (adaptive solvers)
 - Any user-specified size constraints
- Also: $|\nabla h(\mathbf{x})| \leq g$ to limit ratio $G = g + 1$ of neighboring element sizes



Explicit Mesh Size Functions

- A *point-source*

$$h(\mathbf{x}) = h_{\text{pnt}} + g|\mathbf{x} - \mathbf{x}_0|$$

- Any shape, with distance function $\phi(\mathbf{x})$

$$h(\mathbf{x}) = h_{\text{shape}} + g\phi(\mathbf{x})$$

- Combine mesh size functions by min operator:

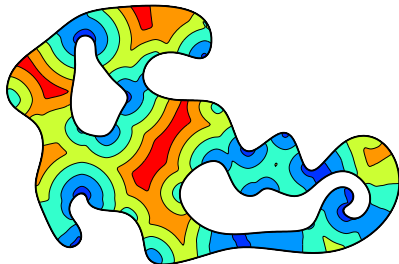
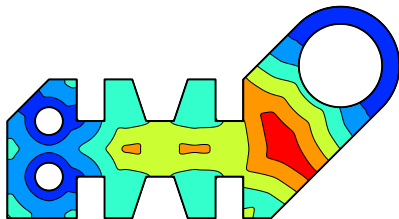
$$h(\mathbf{x}) = \min_i h_i(\mathbf{x})$$

- For more general $h(\mathbf{x})$, solve the *gradient limiting equation* [Persson'05]

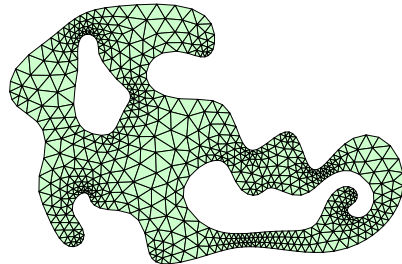
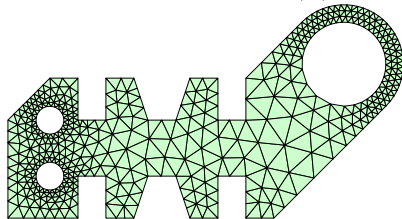
$$\begin{aligned}\frac{\partial h}{\partial t} + |\nabla h| &= \min(|\nabla h|, g), \\ h(t = 0, \mathbf{x}) &= h_0(\mathbf{x}).\end{aligned}$$

Mesh Size Functions – 2-D Examples

Mesh Size Function $h(x)$



Mesh Based on $h(x)$

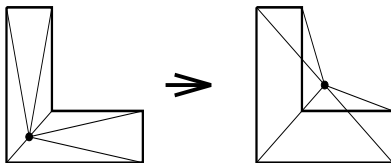


Laplacian Smoothing

- Improve node locations by iteratively moving nodes to average of neighbors:

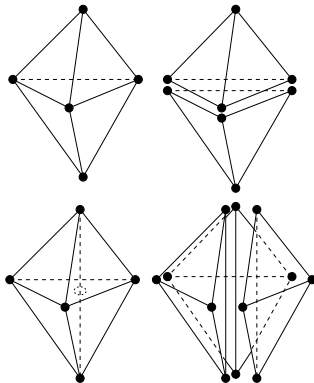
$$\mathbf{x}_i \leftarrow \frac{1}{n_i} \sum_{j=1}^{n_i} \mathbf{x}_j$$

- Usually a good postprocessing step for Delaunay refinement
- However, element quality can get worse and elements might even invert:



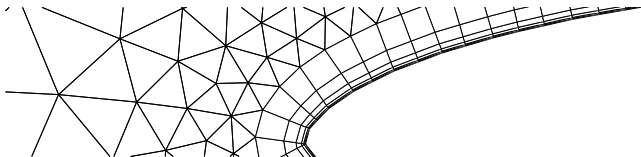
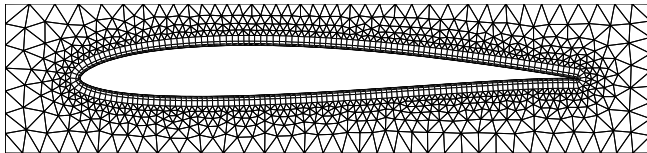
Face and Edge Swapping

- In 3-D there are several swappings between neighboring elements
- Face and edge swapping important postprocessing of Delaunay meshes



Boundary Layer Meshes

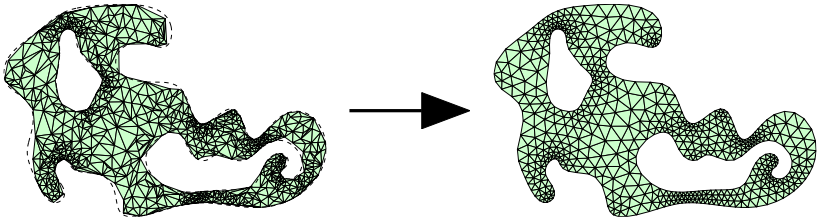
- Unstructured mesh for *offset curve* $\psi(\mathbf{x}) - \delta$
- The structured grid is easily created with the distance function



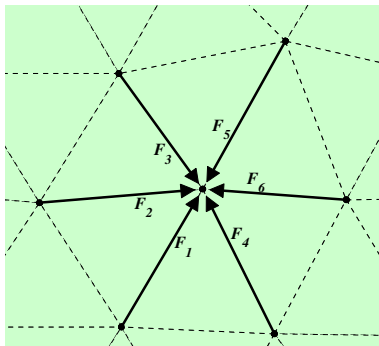
The DistMesh Mesh Generator

The DistMesh Mesh Generator

1. Start with *any* topologically correct initial mesh, for example random node distribution and Delaunay triangulation
2. Move nodes to find force equilibrium in edges
 - Project boundary nodes using implicit function ϕ
 - Update element connectivities



Internal Forces



For each *interior* node:

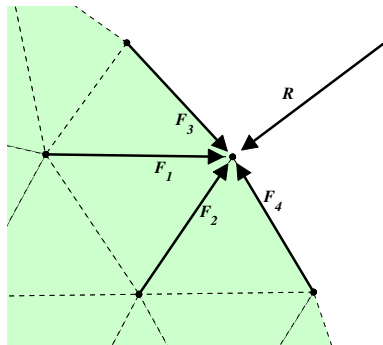
$$\sum_i \mathbf{F}_i = 0$$

Repulsive forces depending on edge length ℓ and equilibrium length ℓ_0 :

$$|\mathbf{F}| = \begin{cases} k(\ell_0 - \ell) & \text{if } \ell < \ell_0, \\ 0 & \text{if } \ell \geq \ell_0. \end{cases}$$

Get expanding mesh by choosing ℓ_0 larger than desired length h

Reactions at Boundaries



For each *boundary* node:

$$\sum_i \mathbf{F}_i + \mathbf{R} = 0$$

Reaction force \mathbf{R} :

- Orthogonal to boundary
- Keeps node along boundary

Node Movement and Connectivity Updates

- Move nodes \mathbf{p} to find force equilibrium:

$$\mathbf{p}_{n+1} = \mathbf{p}_n + \Delta t \mathbf{F}(\mathbf{p}_n)$$

- Project boundary nodes to $\phi(\mathbf{p}) = 0$
- Elements deform, change connectivity based on element quality or in-circle condition (Delaunay)

