Central issues in modelling

- Construct families of curves, surfaces and volumes that
 - can represent common objects usefully;
 - are easy to interact with;
 interaction includes:
 - manual modelling;
 - fitting to measurements;
 - support geometric computations
 - intersection
 - collision

- Main topics:
 - curves
 - surfaces
 - volumes
 - deformation
- Simple curves
- Simple surfaces
- Continuity and splines
- Bezier surfaces and spline surfaces
- Volume models
- Animation

Parametric vs Implicit

- A parametric curve is given as a function of parameters Examples:
 - circle as $(\cos t, \sin t)$
 - twisted cubic as (t, t*t, t*t*t)
- A parametric surface is given as a function of parameters. Examples:
 - sphere as
 (cos s cos t, sin s cos t, sin t)
- Advantage easy to compute normal, easy to render, easy to put patches together.
- Disadvantage ray tracing is hard

- An implicit curve is given by the vanishing of some functions
 - circle on the plane, x*x
 +y*y-r*r=0
 - twisted cubic in space,
 x*y-z=0, x*z-y*y=0, x*x-y=0
- An implicit surface is given by the vanishing of some functions
 - sphere in space x*x+y*y +z*z-r*r=0
 - plane
 a x+ b y + c z+d=0

Interpolation

- Construct a parametric curve that passes through (interpolates) a set of points.
- Lagrange interpolate:
 - give parameter values associated with each point
 - use Lagrange polynomials
 (one at the relevant point, zero at all others) to construct curve
 - curve is:

 $\sum_{i \in \text{points}} p_i \phi_i^{(l)}(t)$

- degree is (#pts-1)
 - e.g. line through two points
 - quadratic through three.
- Functions phi are known as "blending functions"

Hermite curves

- Hermite interpolate
 - give parameter values and derivatives associated with each point
 - curve passes through given point and the given derivative at that parameter value
 - curve is:

- use Hermite polynomials to construct curve
 - one at some parameter value and zero at others or
 - derivative one at some parameter value, and zero at others

 $\sum_{i \in \text{points}} p_i \phi_i^{(h)}(t) + \sum_{i \in \text{points}} v_i \phi_i^{(hd)}(t)$ $i \in \overline{\text{points}}$ *i* points

Extruded surfaces

- Geometrical model Pasta machine
- Take curve and "extrude" surface along vector
- Many human artifacts have this form rolled steel, etc.



 $(x(s,t), y(s,t), z(s,t)) = (x_c(s), y_c(s), z_c(s)) + t(v_0, v_1, v_2)$

Cones

- From every point on a curve, construct a line segment through a single fixed point in space - the vertex
- Curve can be space or plane curve, but shouldn't pass through the vertex



 $(x(s,t), y(s,t), z(s,t)) = (1-t)(x_c(s), y_c(s), z_c(s)) + t(v_0, v_1, v_2)$

Surfaces of revolution - 1

• Plane curve + axis

(x(s,t), y(s,t), z(s,t)) =

- "spin" plane curve around axis to get surface
- Choice of plane is arbitrary, choice of axis affects surface
- In this case, curve is on x-z plane, axis is z axis.

 $(x_c(s)\cos(t), x_c(s)\sin(t), z_c(s))$



Surfaces of revolution -2

Many artifacts are SOR's, as they're easy to make on a lathe.

Controlling is quite easy concentrate on the cross section.

Axis crossing cross-section leads to ugly geometry.

Ruled surfaces -1

- Popular, because it's easy to build a curved surface out of straight segments eg pavilions, etc.
- Take two space curves, and join corresponding points same s with line segment.
- Even if space curves are lines, the surface is usually curved.

(x(s,t), y(s,t), z(s,t)) = $(1-t)(x_1(s), y_1(s), z_1(s)) +$ $t(x_2(s), y_2(s), z_2(s))$



Normals

- Recall: normal is cross product of tangent in t direction and s direction.
- Cylinder: normal is cross-product of curve tangent and direction vector
- SOR: take curve normal and spin round axis

• Cylinders: small steps along curve, straight segments along t generate polygons; exact normal is known.

• Cone: small steps in s generate straight edges, join with vertex to get triangles, normals known exactly except at vertex.

• SOR: small steps in s generate strips, small steps in t along the strip generate edges; join up to form triangles. Normals known exactly.

• Ruled surface: steps in s generate polygons, join opposite sides to make triangles - otherwise "non planar polygons" result. Normals known exactly.

- obtained by iterated linear interpolation
- process is
 known as
 DeCasteljau's
 algorithm

Bezier curves - II

• Blending functions are the Bernstein polynomials

$$c(t) = \sum_{i=0}^{n} p_i B_i^n(t)$$

$$B_i^n(t) = \binom{n}{i} t^i (1-t)^i$$

• e.g. two points

Bezier curves - III

- Bernstein polynomials have several important properties
 - they sum to 1, hence curve lies within convex hull of control points
 - curve interpolates its endpoints
 - curve's tangent at start lies along the vector from p0 to p1
 - tangent at end lies along vector from pn-1 to pn

Bezier curve tricks - I

• "Pull" a curve toward a control point by doubling the control point

Bezier curve tricks-II

- Close the curve by making last point and first point coincident
 - curve has
 continuous
 tangent if first
 segment and last
 segment are
 collinear

Subdivision for Bezier curves

- Use De Casteljau (repeated linear interpolation) to identify points.
- Points as marked in figure give two control polygons, for two Bezier curves, which lie on top of the original.
- Repeated subdivision leads to a polygon that lies very close to the curve
- Limit of subdivision proces is a curve

Fig. 4.5. Decomposition of a Bézier curve into two C^3 continuous curve segments (cf. Fig. 4.4).

de Casteljau Algorithm

- Cascading lerps $\mathbf{p}_{01} = (1-t) \mathbf{p}_0 + t \mathbf{p}_1$ $\mathbf{p}_{12} = (1-t) \mathbf{p}_1 + t \mathbf{p}_2$ $\mathbf{p}_{23} = (1-t) \mathbf{p}_2 + t \mathbf{p}_3$ $\mathbf{p}_{012} = (1-t) \mathbf{p}_{01} + t \mathbf{p}_{12}$ $\mathbf{p}_{123} = (1-t) \mathbf{p}_{12} + t \mathbf{p}_{23}$ $\mathbf{p}_{0123} = (1-t) \mathbf{p}_{012} + t \mathbf{p}_{123}$
- Subdivides curve at **p**₀₁₂₃
 - $\mathbf{p}_0 \mathbf{p}_{01} \mathbf{p}_{012} \mathbf{p}_{0123}$
 - $\mathbf{p}_{0123} \, \mathbf{p}_{123} \, \mathbf{p}_{23} \, \mathbf{p}_{3}$
- Repeated subdivision converges to curve

coordinate free!

Degree Elevation

- Used to add more control over a curve
- Start with

$$\Sigma \mathbf{p}_i \binom{n}{i} t^i (1-t)^{n-i} = \Sigma \mathbf{q}_i \binom{n+1}{i} t^i (1-t)^{n+1-i}$$

- Now figure out the \mathbf{q}_i $(t+(1-t)) \sum \mathbf{p}_i \binom{n}{i} t^i (1-t)^{n-i}$ $= \sum \mathbf{p}_i \binom{n}{i} (t^i (1-t)^{n+1-i} + t^{i+1} (1-t)^{n-i})$
- Compare coefficients

 $\mathbf{q}_{i}\binom{n+1}{i} = \mathbf{p}_{i}\binom{n}{i} + \mathbf{p}_{i-1}\binom{n}{i-1}$ $\mathbf{q}_{i} = (i/(n+1))\mathbf{p}_{i-1} + (n+1-i/(n+1))\mathbf{p}_{i}$

• Repeated elevation converges to curve

Interpolating Splines

- Key idea:
 - high degree
 interpolates
 are badly
 behaved->
 - construct
 curves out of
 low degree
 segments

0

-5

Interpolating Splines - II

- n+1 points;
- write derivatives X'
- X_i is spline for interval between P_i and P_{i+1}

Fig. 3.11. The spline segment X_i .

Interpolating Splines - II

- bolt together a series of Hermite curves with equivalent derivatives.
- But where are the derivative values to come from?
 - Measurements
 - Combination of points
 - Continuity considerations

- Cardinal splines
 - average points
 - t is "tension"
 - specify endpoint tangents
 - or use difference between first two, last two points

$$P_{k} = \left(\frac{1}{2}\right)(1-t)\left(P_{k+1} - P_{k-1}\right)$$
$$P_{k+1} = \left(\frac{1}{2}\right)(1-t)\left(P_{k+2} - P_{k}\right)$$

Tension t=0 gives derivatives as <--• p ${\bf p}_{k+1}$ different values of tension give longer/shorter tangents ${\bf p}_{k+2}$ P_{k-1} t > 0t < 0(Tighter Curve) (Looser Curve)

Interpolating Splines

• Intervals:

$$a = t_0 < t_1 < t_2 < \dots < t_{N-1} < t_N = b.$$

 $\Delta t_i := t_{i+1} - t_i.$

- t values often called "knots"

• Spline form:

$$\begin{aligned} \boldsymbol{X}_{i}(t) &:= \boldsymbol{A}_{i}(t-t_{i})^{3} + \boldsymbol{B}_{i}(t-t_{i})^{2} + \boldsymbol{C}_{i}(t-t_{i}) + \boldsymbol{D}_{i}, \\ & t \in [t_{i}, t_{i+1}], \quad i = 0(1)N-1, \end{aligned}$$

Continuity

- Require at endpoints:
 - endpoints equal $X_i(t_i) = X_{i-1}(t_i)$ or $X_i(t_{i+1}) = X_{i+1}(t_{i+1}),$
 - 1'st derivatives $X'_i(t_i) = X'_{i-1}(t_i)$ equal $X''_i(t_i) = X''_{i-1}(t_i)$
 - $= \mathbf{X}_{i-1}^{\prime\prime}(t_i) \quad \text{or} \quad \mathbf{X}_i^{\prime\prime}$ $= \mathbf{X}_{i-1}^{\prime\prime}(t_i) \quad \text{or} \quad \mathbf{X}_i^{\prime\prime}$
- or $X_i(t_{i+1}) = X_{i+1}(t_{i+1}),$ or $X'_i(t_{i+1}) = X'_{i+1}(t_{i+1}),$ or $X''_i(t_{i+1}) = X''_{i+1}(t_{i+1}).$

 2'nd derivatives equal • From endpoint and 1'st derivative:

$$\begin{aligned} X_i(t_i) &= P_i = D_i, & X_i(t_{i+1}) = P_{i+1} = A_i \Delta t_i^3 + B_i \Delta t_i^2 + C_i \Delta t_i + D_i, \\ X'_i(t_i) &= P'_i = C_i, & X'_i(t_{i+1}) = P'_{i+1} = 3A_i \Delta t_i^2 + 2B_i \Delta t_i + C_i, \end{aligned}$$

• So that

$$A_{i} = \frac{1}{(\Delta t_{i})^{3}} [2(P_{i} - P_{i+1}) + \Delta t_{i}(P'_{i} + P'_{i+1})],$$

$$B_{i} = \frac{1}{(\Delta t_{i})^{2}} [3(P_{i+1} - P_{i}) - \Delta t_{i}(2P'_{i} + P'_{i+1})].$$

• Yielding:

$$\begin{split} \boldsymbol{X}_{i}(t) &= \\ \boldsymbol{P}_{i} \left(2\frac{(t-t_{i})^{3}}{(\Delta t_{i})^{3}} - 3\frac{(t-t_{i})^{2}}{(\Delta t_{i})^{2}} + 1 \right) + \boldsymbol{P}_{i+1} \left(-2\frac{(t-t_{i})^{3}}{(\Delta t_{i})^{3}} + 3\frac{(t-t_{i})^{2}}{(\Delta t_{i})^{2}} \right) \\ &+ \boldsymbol{P}_{i}' \left(\frac{(t-t_{i})^{3}}{(\Delta t_{i})^{2}} - 2\frac{(t-t_{i})^{2}}{\Delta t_{i}} + (t-t_{i}) \right) + \boldsymbol{P}_{i+1}' \left(\frac{(t-t_{i})^{3}}{(\Delta t_{i})^{2}} - \frac{(t-t_{i})^{2}}{\Delta t_{i}} \right) \end{split}$$

• Second Derivative:

$$\begin{split} \boldsymbol{X}_{i}''(t) &= 6\boldsymbol{P}_{i}\left(\frac{2(t-t_{i})}{(\Delta t_{i})^{3}} - \frac{1}{(\Delta t_{i})^{2}}\right) + 6\boldsymbol{P}_{i+1}\left(-2\frac{(t-t_{i})}{(\Delta t_{i})^{3}} + \frac{1}{(\Delta t_{i})^{2}}\right) \\ &+ 2\boldsymbol{P}_{i}'\left(3\frac{(t-t_{i})}{(\Delta t_{i})^{2}} - \frac{2}{\Delta t_{i}}\right) + 2\boldsymbol{P}_{i+1}'\left(\frac{3(t-t_{i})}{(\Delta t_{i})^{2}} - \frac{1}{\Delta t_{i}}\right). \end{split}$$

• Want:

x 7'

1 1

$$\boldsymbol{X}_{i-1}^{\prime\prime}(t_i) = \boldsymbol{X}_i^{\prime\prime}(t_i)$$

$$\begin{split} \Delta t_i \boldsymbol{P}_{i-1}' + 2(\Delta t_{i-1} + \Delta t_i) \boldsymbol{P}_i' + \Delta t_{i-1} \boldsymbol{P}_{i+1}' \\ &= 3 \frac{\Delta t_{i-1}}{\Delta t_i} (\boldsymbol{P}_{i+1} - \boldsymbol{P}_i) + 3 \frac{\Delta t_i}{\Delta t_{i-1}} (\boldsymbol{P}_i - \boldsymbol{P}_{i-1}). \end{split}$$

Missing equations

- Recurrence relations represent d(n-1) equations in d(n+1) unknowns (d is dimension)
- We need to supply the derivative at the start and at the finish (or two equivalent constraints)
- Options:
 - second derivatives vanish at each end (natural spline)
 - give slopes at the boundary
 - vector from first to second, second last to last
 - parabola through first three, last three points
 - third derivative is the same at first, last knot

Parametric vs Geometric Continuity

• Parametric continuity:

- The curve and derivatives up to k are continuous *as a function of parameter value*
 - C^k
- Useful for (for example) animation
- e.g. the interpolating spline from above

- Geometric continuity
 - curve, derivatives up to k'th are the same for equivalent parameter values
 - i.e. there exists a reparametrisation that would achieve parametric continuity
 - D^k
 - Useful, because we often don't require parametric continuity,
 - e.g. take two Hermite curves,
 both parametrised by [0, 1],
 identify endpoints and
 derivatives

More on Geometric Continuity

- Tangent direction is invariant to translation and parametrisation so we can use this to get G1 continuity.
- G2 use curvature
 - property of a curve that is invariant to rotation and translation, and also reparametrisation
 - (1/radius) for best fitting circle
 - the circle whose 2nd derivative is the same as the curve's
 - (equivalent) a circle that intersects the curve in three points arbitrarily close
 - Formula
 - $(x''y' y''x')/(x'^2 + y'^2)^{(3/2)}$
 - dN/ds=kN for N the unit normal

Keep in mind

- Lagrange and Hermite interpolates of the same degree are the same families of curves
 - they just have different control structures
- The interpolating cubic spline is equivalent to a bunch of Hermite cubics, with a different control structure
 - we got the derivatives from the second derivative constraint

- The line of reasoning for interpolating cubic splines works for higher degrees, too
 - but we must either use more derivatives, or supply more information
 - Cubic is the most important case, because cubic splines
 (rather roughly) look like wooden splines
- We chose parameter values for the interpolating curve
 - different choices lead to different curves

Spline blending functions

- "Switches" turn blending functions on and off
- E.g. a piecewise cubic spline obtained by attaching two Hermite curves to one another
 - In principle, there are 8 blending functions (4 points and 4 derivatives)
 - Actually, two points and two derivatives are the same
 - 6 blending functions
 - these are piecewise cubic, easily sketched
 - The properties of the blending functions are what's important
- Now let's consider splines that don't interpolate, by concentrating on the blending functions

B-splines - I

- We obtain a set of blending functions by a recursive definition, with "switches" at the base of the recursion
- Curve:

$$X(t) = \sum_{k=0}^{n} P_k B_{k,d}(t)$$

• where d (called the "order") is:

 $2 \leq d \leq n+1$

B-Spline Blending Functions

- Knots
 - idea: parameter values where curve segments meet, as in Hermite example

$$(t_0, t_1, ..., t_{n+d})$$

where $t_0 \le t_1 \le \dots \le t_{n+d}$

• Blending functions

 $B_{k,1}(t) = \begin{cases} 1 & t_k \le t \le t_{k+1} \\ 0 & \text{otherwise} \end{cases}$ $B_{k,d}(t) = \left(\frac{t - t_k}{t_{k+d-1} - t_k}\right) B_{k,d-1}(t) + \left(\frac{t_{k+d} - t_k}{t_{k+d} - t_{k+1}}\right) B_{k+1,d-1}(t) \end{cases}$

Closed B-Splines

• Periodically extend the control points and the knots

$$P_{n+1} = P_0$$
$$t_{n+1} = t_0$$

• etc

A B-spline curve, with knots at 0,1,... and order 5

Repeated knots

- Definition works for repeated knots (if we are understanding about 0/0)
- Repeated knot reduces continuity. A B-spline blending function has continuity C^{d-2}; if the knot is repeated m times, continuity is now C^{d-m-1}
- e.g. -> quadratic B-spline (i.e. order 3) with a double knot

Fig. 4.22g. A quadratic B-spline with a double knot.

Most useful case

- select the first d and the last d knots to be the same
 - we then get the first and last points lying on the curve
 - also, the curve is tangent to the first and last se

- e.g. cubic case below
- Notice that a control point influences at most d parameter intervals - **local control**

Fig. 4.24b. B-splines for an open B-spline curve with non-uniform knot vector.

Fig. 4.25c. B-spline curve with k = 3, n = 9 and the Bézier curve of degree 9 with the same control polygon.

Bezier curve is the heavy curve

B-Spline properties

- For a B-spline curve of order d
 - if m knots coincide, the curve is C^{d-m-1} at the corresponding point
 - if d-1 points of the control polygon are collinear, then the curve is tangent to the polygon
 - if d points of the control polygon are collinear, then the curve and the polygon have a common segment
 - if d-1 points coincide, then the curve interpolates the common point and the two adjacent sides of the polygon are tangent to the curve
 - each segment of the curve lies in the convex hull of the associated d points