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1 Modeling

1.1 Splines

1.1.1 Cubic Hermite Interpolation

Each point is defined by its position h_n and slope h_{m+n} , m being the number of control points. To simplify calculations, it is assumed that $t_0 = 0$ and $t_1 = 1$.

The goal is to convert from a monomial basis

$$\begin{aligned}\phi_0(t) &= 1 \\ \phi_1(t) &= t \\ \phi_2(t) &= t^2 \\ \phi_3(t) &= t^3\end{aligned}$$

to a hermite basis

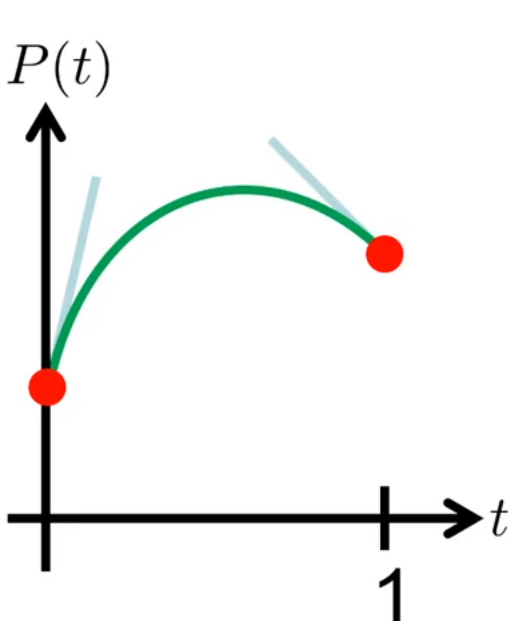
$$\begin{aligned}H_0(t) &= 2t^3 - 3t^2 + 1 \\ H_1(t) &= -2t^3 + 3t^2 \\ H_2(t) &= t^3 - 2t^2 + t \\ H_3(t) &= t^3 - t^2\end{aligned}$$

so that instead of having to manipulate polynomial coefficients

$$f(t) = a\phi_3(t) + b\phi_2(t) + c\phi_1(t) + d\phi_0(t)$$

an easier point slope method can be used:

$$f(t) = h_0H_0(t) + h_1H_1(t) + h_2H_2(t) + h_3H_3(t)$$



$$\begin{aligned}P(t) &= at^3 + bt^2 + ct + d \\ P'(t) &= 3at^2 + 2bt + c\end{aligned}$$

$$\begin{aligned}h_0 &= P(0) = d \\ h_1 &= P(1) = a + b + c + d \\ h_2 &= P'(0) = c \\ h_3 &= P'(1) = 3a + 2b + c\end{aligned}$$

Unknowns in this equation are a , b , c , and d , so a matrix can be used to solve the systems of equations:

$$\begin{pmatrix} 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \\ d \end{pmatrix} = \begin{pmatrix} h_0 \\ h_1 \\ h_2 \\ h_3 \end{pmatrix}$$

a , b , c , and d can be obtained from h values by inverting the matrix:

$$\begin{pmatrix} 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix}^{-1} \begin{pmatrix} h_0 \\ h_1 \\ h_2 \\ h_3 \end{pmatrix} = \begin{pmatrix} a \\ b \\ c \\ d \end{pmatrix}$$

From these solved h values, $P(t)$ can now be converted to a form that is easier for a user to manipulate, in terms of h values:

$$\begin{aligned}
 P(t) &= at^3 + bt^2 + ct + d \\
 &= (2h_0 - 2h_1 + h_2 + h_3)t^3 \\
 &\quad + (-3h_0 + 3h_1 - 2h_2 - h_3)t^2 \\
 &\quad + h_2t + h_0 \\
 &= h_0(2t^3 - 3t^2 + 1) + h_1(-2t^3 + 3t^2) + \\
 &\quad h_2(t^3 - 2t^2 + t) + h_3(t^3 - t^2)
 \end{aligned}$$

Each equation in $P(t) = h_0(2t^3 - 3t^2 + 1) + h_1(-2t^3 + 3t^2) + h_2(t^3 - 2t^2 + t) + h_3(t^3 - t^2)$ that is multiplied by an h value is called a cubic hermite.

1.1.2 More than 1D

A parametric curve described by $\vec{\gamma}(t) = (\gamma_0(t), \gamma_1(t))$ can be converted into hermite basis like this:

$$\underbrace{\vec{\gamma}'(t)}_{\text{Tangent}} = \underbrace{(\gamma_0'(t), \gamma_1'(t))}_{\text{Slopes}}$$

where cubic hermite interpolation can be done for both dimensions.

1.1.3 Cubic blossom

The cubic blossom of a function $\vec{f}(t)$ is $\vec{F}(t_1, t_2, t_3)$.

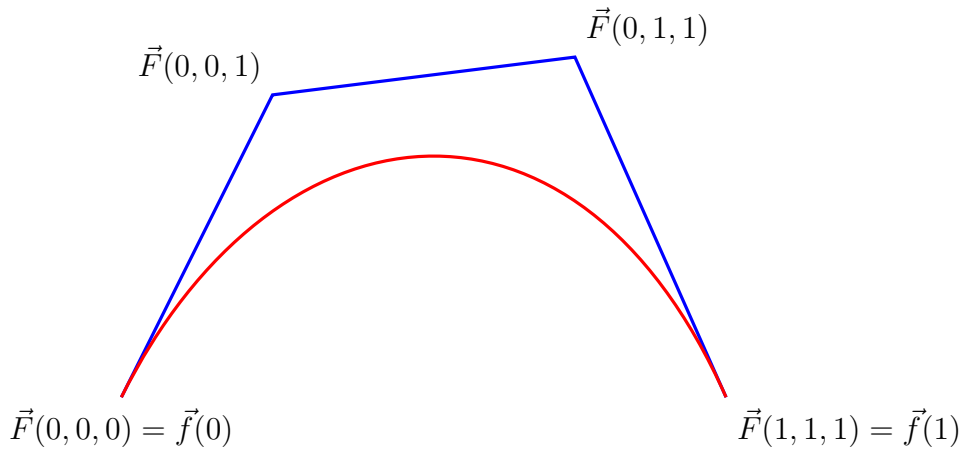
Cubic blossoms have three properties:

1. Symmetric
 - $\vec{F}(t_1, t_2, t_3) = \vec{F}(t_1, t_3, t_2) = \vec{F}(t_3, t_1, t_2) \dots$
2. Affine
 - $\vec{F}(\alpha u + (1 - \alpha)v, t_2, t_3) = \alpha \vec{F}(u, t_2, t_3) + (1 - \alpha) \vec{F}(v, t_2, t_3)$
 - If only one of \vec{F} 's arguments $t_c = \alpha u + (1 - \alpha)v$ is changing between different points on \vec{F} , then any value $\vec{F}(t_c, t_2, t_3)$ in between $\vec{F}(u, t_2, t_3)$ and $\vec{F}(v, t_2, t_3)$ is scaled equivalently with α like t_c is scaled between u and v .
3. Diagonal
 - $\vec{f}(t) = \vec{F}(t, t, t)$

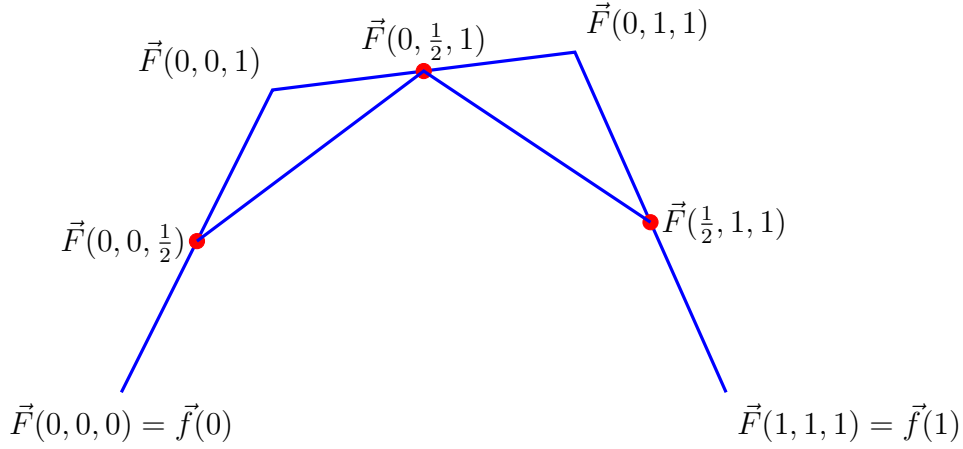
Blossoming examples

- $\vec{f}(t) = t^3 \mapsto \vec{F}(t_1, t_2, t_3) = t_1 t_2 t_3$
- $\vec{f}(t) = t^2 \mapsto \vec{F}(t_1, t_2, t_3) = (t_1 t_2 + t_1 t_3 + t_2 t_3)/3$
- $\vec{f}(t) = t \mapsto \vec{F}(t_1, t_2, t_3) = (t_1 + t_2 + t_3)/3$
- $\vec{f}(t) = 1 \mapsto \vec{F}(t_1, t_2, t_3) = 1$
- $\vec{f}(t) = 3t^3 - t + 1 = 3(t_1 t_2 t_3) - (t_1 + t_2 + t_3)/3 + 1$

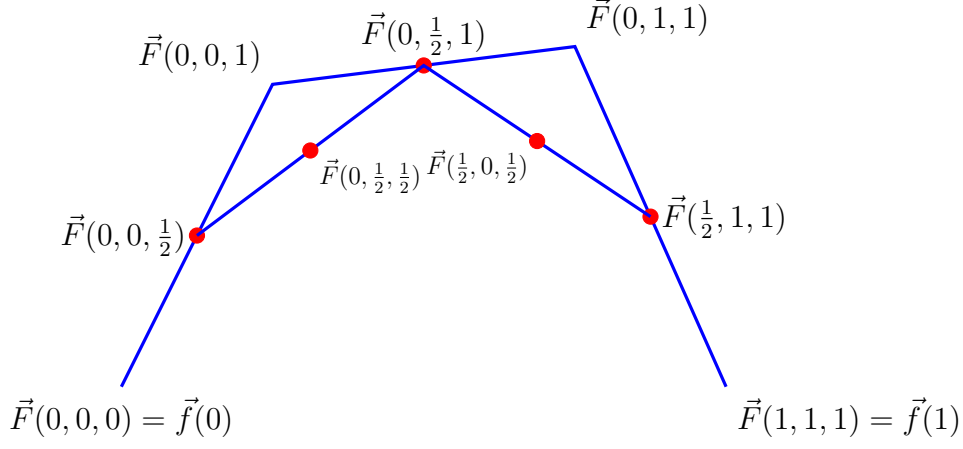
Cubic curves can be blossomed by blossoming each coordinate function separately, which will give a function that maps 3 t variables to two dimensions x and y : $\vec{F}(t_1, t_2, t_3) : \mathbb{R}^3 \mapsto \mathbb{R}^2$. A cubic curve can be obtained from a blossom by specifying four points $\vec{F}(0, 0, 0)$, $\vec{F}(0, 0, 1)$, $\vec{F}(0, 1, 1)$, $\vec{F}(1, 1, 1)$ (which form a cubic control polygon) and subdividing the surface given by the selected points (known as the De Casteljau's Algorithm). Only these four points are required because of the symmetry property of a blossom.



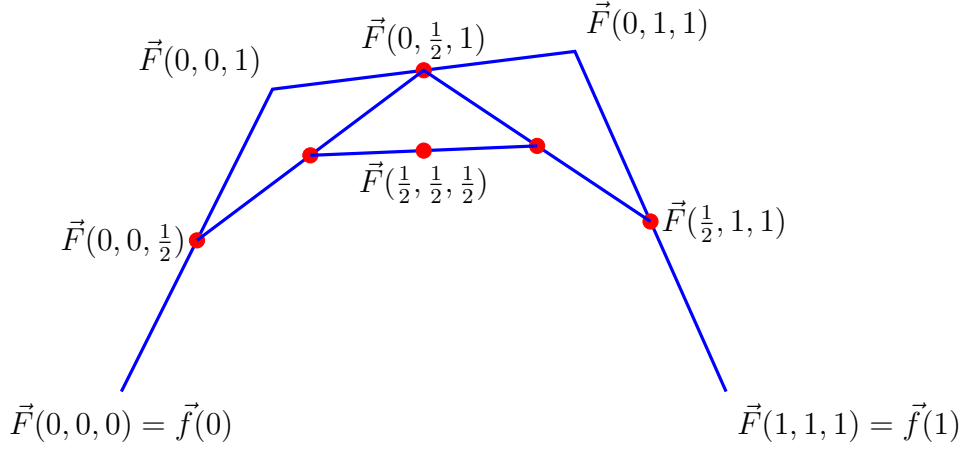
Subdividing the polygon:



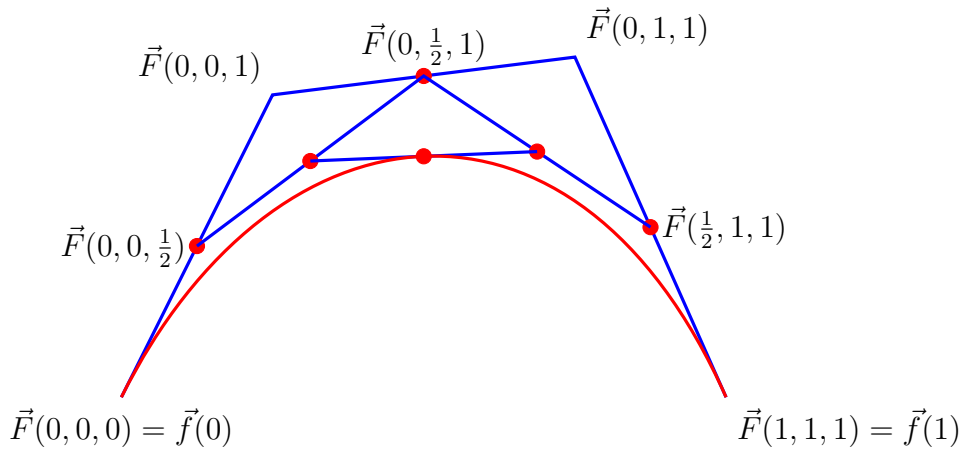
\vec{F} 's arguments at the subdivision points can be determined due to the affine property of a blossom. Further subdivision



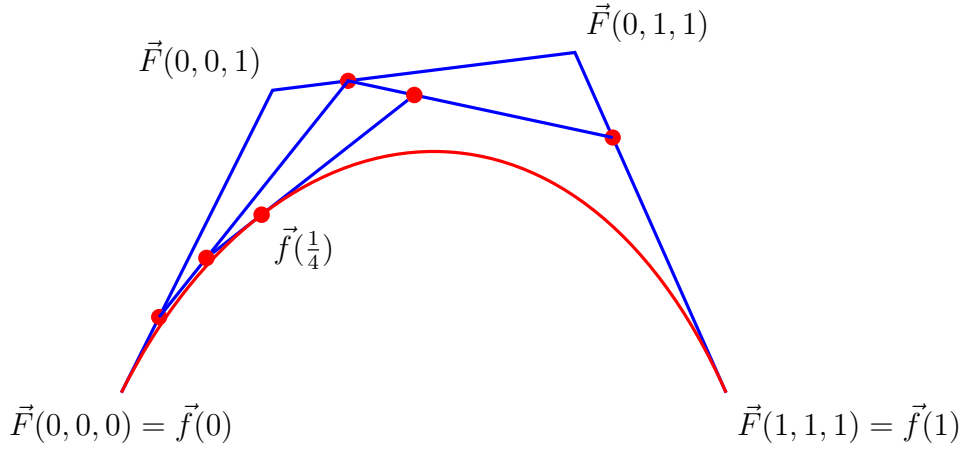
One more subdivision (most recent two subdivision point values omitted because there's no room left)



The final subdivision point can be shown to equal $\vec{f}(\frac{1}{2})$ because of the diagonal property of blossoms, meaning the \vec{f} curve will intersect this third subdivision point:

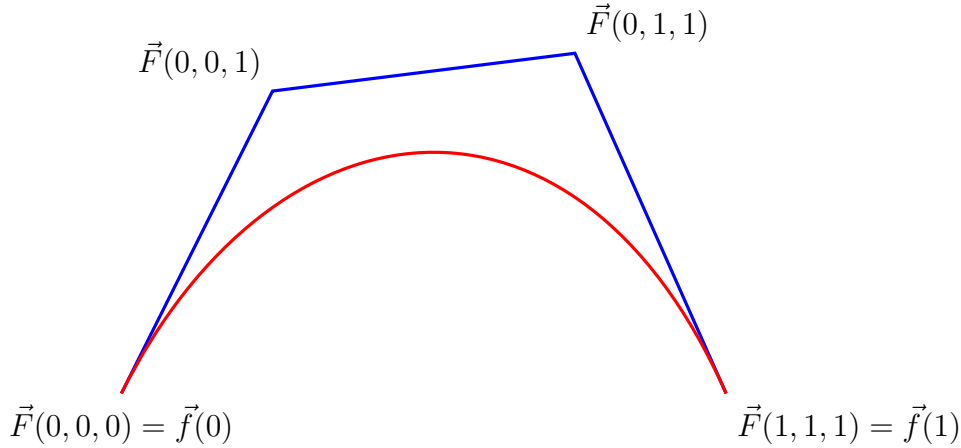


Other values of \vec{f} can be found with subdividing the cubic control polygon on different values. For example, to find $\vec{f}(\frac{1}{4})$, each edge would be divided $\frac{1}{4}$ of the way along it.



1.1.4 Bernstein polynomials

Another basis that can represent cubic curves. Bernstein basis is canonical for Bezier.



The red curve above can be represented as $\vec{f}(t) = \vec{F}(0,0,0)B_0(t) + \vec{F}(0,0,1)B_1(t) + \vec{F}(0,1,1)B_2(t) + \vec{F}(1,1,1)B_3(t)$, knowing

$$\begin{aligned} B_0(t) &= (1-t)^3 = 1 - 3t + 3t^2 - t^3 \\ B_1(t) &= 3t(1-t)^2 = 3t - 6t^2 + 3t^3 \\ B_2(t) &= 3t^2(1-t) = 3t^2 - 3t^3 \\ B_3(t) &= t^3 \end{aligned}$$

Changing basis from monomial to Bernstein:

$$\begin{pmatrix} 1 & -3 & 3 & -1 \\ 0 & 3 & -6 & 3 \\ 0 & 0 & 3 & -3 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ t \\ t^2 \\ t^3 \end{pmatrix} = \begin{pmatrix} B_0(t) \\ B_1(t) \\ B_2(t) \\ B_3(t) \end{pmatrix}$$

And to change from Bernstein to monomial, the inverse of the matrix can be used.

1.1.5 General spline formulation

$\vec{\gamma}(t) = \text{Geometry} \cdot \text{Spline basis} \cdot \text{Monomial basis}$

- Geometry contains control point coordinates
- Spline basis defines the type of spline (Hermite, Bernstein, etc)
- Monomial basis is a column vector $(1, t, t^2, \dots, t^n)$

Example of a spline represented in the Bernstein basis:

$$P(t) = \begin{pmatrix} x(t) \\ y(t) \end{pmatrix} = \begin{pmatrix} x_1 & x_2 & x_3 & x_4 \\ y_1 & y_2 & y_3 & y_4 \end{pmatrix} \begin{pmatrix} 1 & -3 & 3 & -1 \\ 0 & 3 & -6 & 3 \\ 0 & 0 & 3 & -3 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ t \\ t^2 \\ t^3 \end{pmatrix}$$

1.1.6 Orders of continuity

C^0 = continuous (seam can be sharp)

G^1 = geometric continuity (Tangent vectors align at the seam)

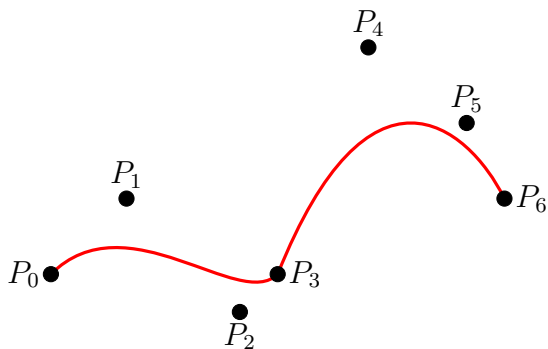
C^1 = parametric continuity (Same velocity at the seam)

C^2 = curvature continuity (Tangents and tangent derivatives are the same)

1.1.7 Cubic B-splines

≥ 4 control points

Chain together splines in fours, popping the back control point off and pushing the next control point on



The full curve is composed of many smaller splines, in which the n th spline is formed by points $P_{n..n+3}$. The benefit of this method is that it guarantees C^1 continuity since every spline shares three control points with the one that comes before and after it.

The end points won't connect with this method, but repeating endpoints will fix it.

The basis functions as well as the basis conversion matrix for cubic b-spline are listed here:

$$B_1(t) = \frac{1}{6}(1-t)^3$$

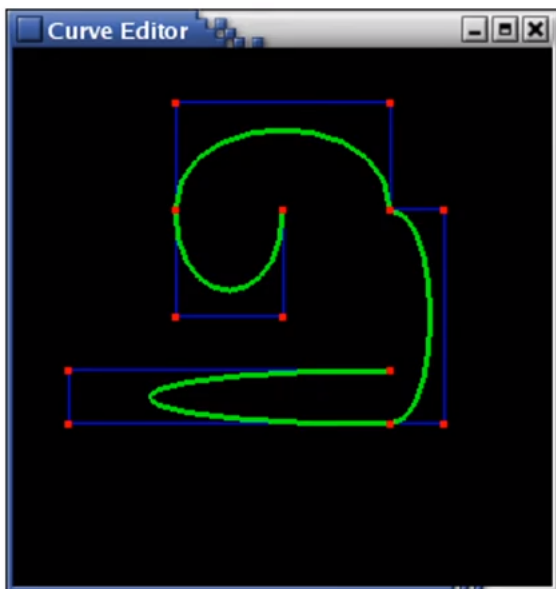
$$B_2(t) = \frac{1}{6}(3t^3 - 6t^2 + 4)$$

$$B_3(t) = \frac{1}{6}(-3t^3 + 3t^2 + 3t + 1)$$

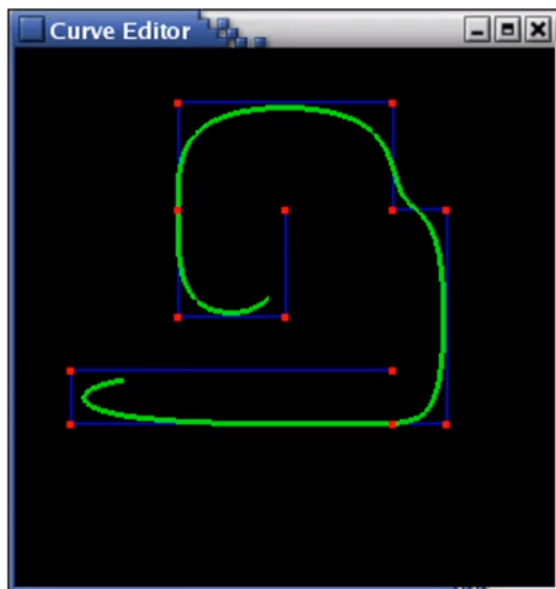
$$B_4(t) = \frac{1}{6}t^3$$

$$B_{B-spline} = \frac{1}{6} \begin{pmatrix} 1 & -3 & 3 & -1 \\ 4 & 0 & -6 & 3 \\ 1 & 3 & 3 & -3 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

1.1.8 Bezier vs B-spline



Bézier



B-Spline

Bezier is derived from Bernstein polynomials, while B-spline uses a different set of basis functions. Additionally, Bezier will try to intersect with control points and only guarantees C^0 continuity, while B-spline does not and guarantees C^1 continuity.

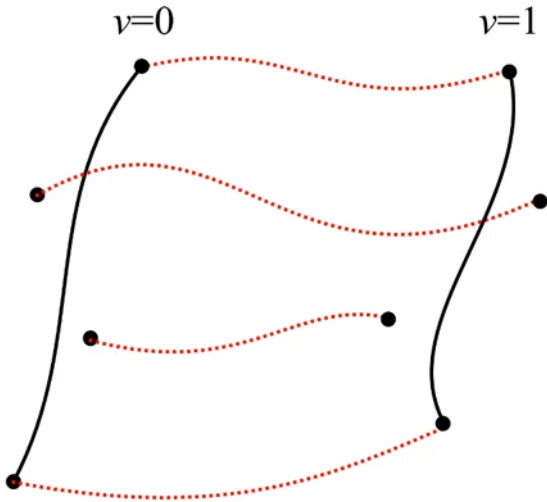
Converting between bezier and b-spline, given G = geometry, B_0 = current basis matrix, $T(t)$ = monomial basis, and B_1 = the basis matrix to convert to:

$$\begin{aligned}\vec{\gamma}(t) &= G \cdot B_0 \cdot T(t) \\ &= G \cdot B_0 \cdot B_1^{-1} \cdot B_1 \cdot T(t) \\ &= (G \cdot B_0 \cdot B_1^{-1}) \cdot B_1 \cdot T(t)\end{aligned}$$

The new geometry matrix is then represented as $G \cdot B \cdot B_1^{-1}$, which shows that to convert between bezier and b-spline, a different set of data is required for the same curve.

1.2 2D Surfaces

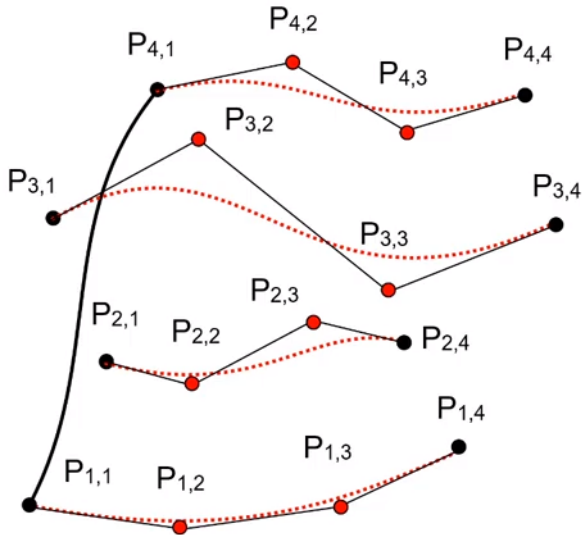
1.2.1 Bicubic Bezier surfaces



Given u changes vertically and v changes horizontally,

$$\begin{aligned}P(u, v) &= (1 - u)^3 P_1(v) \\ &\quad + 3u(1 - u)^2 P_2(v) \\ &\quad + 3u^2(1 - u) P_3(v) \\ &\quad + u^3 P_4(v)\end{aligned}$$

For any constant u for $0 \leq u \leq 1$ there is a bezier curve as a function of v for $0 \leq v \leq 1$ at that u , and vice versa.



$$\begin{aligned}P(u, v) &= \sum_{i=1}^4 B_i(u) P_i(v) \\ P_i(v) &= \sum_{j=1}^4 B_j(v) P_{i,j}\end{aligned}$$

Which can be compactly expressed as:

$$\begin{aligned} P(u, v) &= \sum_{i=1}^4 B_i(u) \left[\sum_{j=1}^4 P_{i,j} B_j(v) \right] \\ &= \sum_{i=1}^4 \sum_{j=1}^4 P_{i,j} B_{i,j}(u, v) \end{aligned}$$

where $B_{i,j}(u, v) = B_i(u)B_j(v)$.

1.2.2 Tangents and Normals for patches

The partial derivatives $\partial P / \partial u$ and $\partial P / \partial v$ are tangent vectors.

Normal vector is equal to the cross product of the tangent vectors:

$$\vec{N} = \frac{\partial P}{\partial u} \times \frac{\partial P}{\partial v}$$

\vec{N} is typically not unit, so it needs to be normalized

1.2.3 Matrix notation for patches

$$P(u, v) = \begin{pmatrix} B_1(u) & \cdots & B_4(u) \end{pmatrix} \begin{pmatrix} P_{1,1} & \cdots & P_{1,4} \\ \vdots & \ddots & \vdots \\ P_{4,1} & \cdots & P_{4,4} \end{pmatrix} \begin{pmatrix} B_1(v) \\ \vdots \\ B_4(v) \end{pmatrix}$$

1.2.4 Implicit surfaces

$f(x, y, z) = 0$ is on the surface

$f(x, y, z) < 0$ is inside the surface

$f(x, y, z) > 0$ is outside the surface

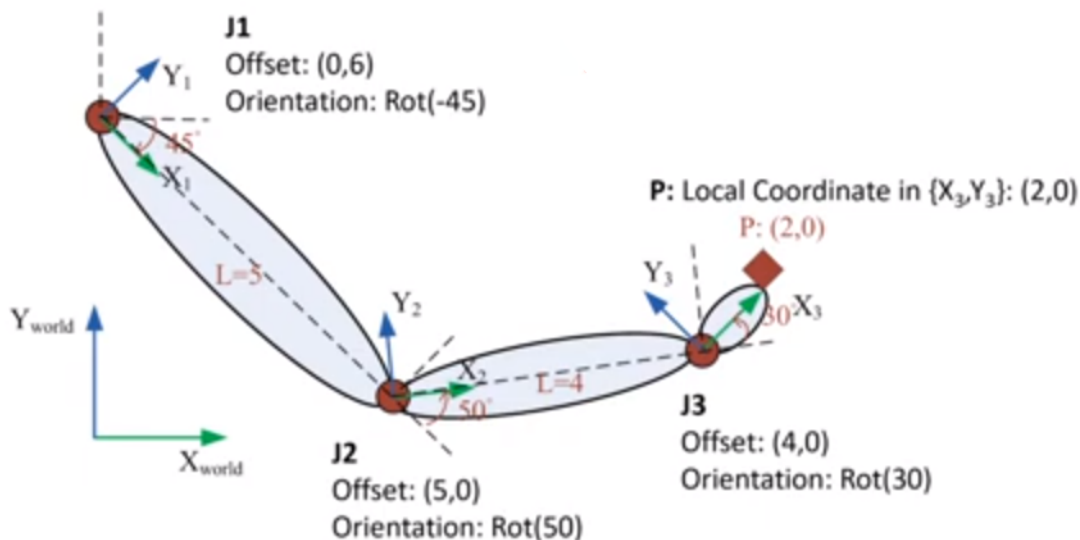
1.3 Hierarchical modeling

1.3.1 Forward kinematics

Specifies base position and joint angles. To compute the final position and orientation of a joint in world coordinates, the motion of all its ancestor joints must be computed as well.

Given values for joint dof (degree of freedom) $\vec{\theta} = [\theta_1, \theta_2, \dots, \theta_n]^T$, compute the end effector in world space $\vec{e} = [e_1, e_2, \dots, e_m]^T$.

$$\vec{e} = F(\vec{\theta})$$



In the diagram above, P in world coordinates can be calculated with $P_{world} = M_{01}M_{12}M_{23}P_{local}$, or $T(0, 6)R(-45)T(5, 0)R(50)T(4, 0)R(30)P_{local}$, T being a transformation matrix and R being a rotation matrix.

1.3.2 Inverse kinematics

Forward kinematics has some issues such as determining what orientation the arm should be in to interact with other objects, which is what inverse kinematics aims to solve.

In forward kinematics, $(\mathbf{e}_1, \mathbf{e}_2) = F(\theta_1, \theta_2, \theta_3)$; however, in inverse kinematics, $(\theta_1, \theta_2, \theta_3) = G(\mathbf{e}_1, \mathbf{e}_2)$.

Inverse kinematics is difficult because sometimes there is no solution / multiple solutions to a problem. Typically it can't be analytically solved and requires numerical methods.

- Jacobian iterative method
- Optimization based methods

Given position of a point in local coordinates \mathbf{v}_s and desired position in world coordinates \mathbf{v}_w , find skeleton parameters \mathbf{p} .

$$\mathbf{v}_w = S(\mathbf{p})\mathbf{v}_s = S(x_h, y_h, z_h, \theta_h, \phi_h, \sigma_h, \theta_t, \phi_t, \sigma_t, \theta_c, \theta_f, \phi_f)\mathbf{v}_s$$

1.3.3 Jacobian

$\mathbf{e} = F(\boldsymbol{\theta})$ where \mathbf{e} is known and $\boldsymbol{\theta}$ is to be solved for.

The Jacobian is the matrix of partial derivatives of \mathbf{F} : $J = dF/d\boldsymbol{\theta}$

For $\mathbf{F} = [\mathbf{F}_1, \mathbf{F}_2, \dots, \mathbf{F}_m]^T$ and $\boldsymbol{\theta} = [\theta_1, \theta_2, \dots, \theta_n]^T$:

$$J = \begin{bmatrix} \frac{\partial \mathbf{F}_1}{\partial \theta_1} & \dots & \frac{\partial \mathbf{F}_1}{\partial \theta_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial \mathbf{F}_m}{\partial \theta_1} & \dots & \frac{\partial \mathbf{F}_m}{\partial \theta_n} \end{bmatrix}$$

1.4 Skinning

1.4.1 SSD

Skin is made up of vertices, some of which are attached to a single bone, while others are attached to multiple.

Example



1.4.2 Vertex weights

Weight w_{ij} is assigned for each vertex \mathbf{p}_i for each bone \mathbf{b}_j . ("How much will vertex i move with bone j ?")

$w_{ij} = 1$ means \mathbf{p}_i is rigidly attached to bone \mathbf{b}_j . Weights should not be negative, and the sum of weights across all bones for each vertex should equal 1.

The number of bones that can influence a single vertex is typically limited to $N = 4$ bones/vertex.

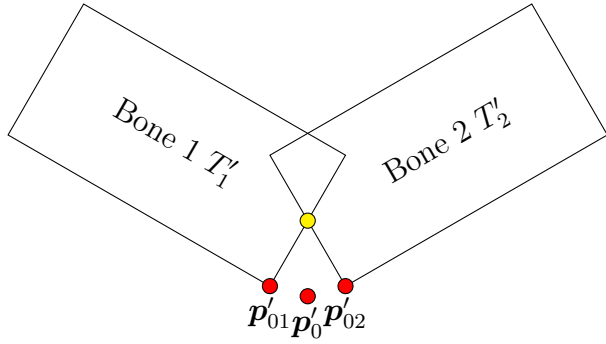
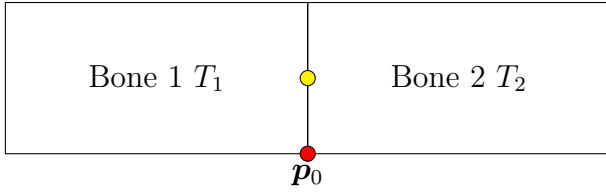
1.4.3 Linear blend skinning

First transform each vertex \mathbf{p}_i with each bone as if it were tied to it rigidly, and then blend the results using weights.

Given \mathbf{p}'_{ij} is vertex i transformed by bone j , T_j is the current transformation of bone j , and \mathbf{p}'_i is the new position of vertex i :

$$\begin{aligned} \mathbf{p}'_{ij} &= T_j \mathbf{p}_i \\ \mathbf{p}'_i &= \sum_j w_{ij} \mathbf{p}'_{ij} \end{aligned}$$

Example



Vertex \mathbf{p}_0 has weights $w_{01} = 0.5$ and $w_{02} = 0.5$. Points \mathbf{p}'_{01} and \mathbf{p}'_{02} are generated by transforming point \mathbf{p}_0 using T'_1 and T'_2 . \mathbf{p}'_0 can be determined using $\mathbf{p}'_0 = w_{01}\mathbf{p}'_{01} + w_{02}\mathbf{p}'_{02} = 0.5\mathbf{p}'_{01} + 0.5\mathbf{p}'_{02}$.

2 Raytracing

2.1 Geometry representation

A ray is represented as $\mathbf{p}(t) = \mathbf{o} + \mathbf{d}t$, \mathbf{o} being the ray origin and \mathbf{d} being the ray direction. All solutions for t under section 2.1 refer to the t parameter for $\mathbf{p}(t)$.

2.1.1 Sphere

Implicit equation: $H(\mathbf{P}) = \mathbf{P} \cdot \mathbf{P} - r^2 = 0$

Intersection detection (\mathbf{c}_o is origin, \mathbf{R}_d is normalized ray direction, \mathbf{c}_s is sphere center, r is sphere radius)

$$\begin{aligned} \mathbf{c}_\Delta &= \mathbf{c}_s - \mathbf{c}_o \\ a &= \mathbf{R}_d \cdot \mathbf{R}_d \\ b &= 2(\mathbf{c}_\Delta \cdot \mathbf{R}_d) \\ c &= \mathbf{c}_\Delta \cdot \mathbf{c}_\Delta - r^2 \\ d &= b^2 - 4ac \\ t &= \frac{b \pm \sqrt{d}}{2a} \end{aligned}$$

If $d < 0$, there is no intersection. If both solutions for t are negative, there is no intersection. If only one solution for t is positive, that is where the intersection occurs. If both solutions for t are positive, the intersection occurs at the closest t .

2.1.2 Triangle

Any point \mathbf{P} on a triangle $(\mathbf{a}, \mathbf{b}, \mathbf{c})$ can be represented as $\mathbf{P}(\alpha, \beta, \gamma) = \alpha\mathbf{a} + \beta\mathbf{b} + \gamma\mathbf{c}$, with $\alpha + \beta + \gamma = 1$.