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# Surfaces: Tensor Products

**Milind Sohoni**

<http://www.cse.iitb.ac.in/~sohoni>

## Polynomials in 2 variables

Let  $P^{m,n}[u, v]$  denote the vector space of all polynomials of degree at most  $m$  in  $u$  and  $n$  in  $v$ . Thus, for example,

$$3u^2v - v^3 \in P^{2,3}[u, v] \subset P^{3,3}[u, v]$$

The dimension of  $P^{m,n}[u, v]$  is obviously  $(m + 1)(n + 1)$  and the **Taylor basis** for it is the set:

$$\{u^i v^j \mid 0 \leq i \leq m, 0 \leq j \leq n\}$$

Just as polynomials in one variable served us to parametrize curves, these will serve us to parametrize surfaces.

## Tensor-Product Bases

Actually, if  $B = \{b_0(u), \dots, b_m(u)\}$  is a basis for  $P^m[u]$  and  $C = \{c_0(v), \dots, c_n(v)\}$  is a basis for  $P^n[v]$  then:

$$B \otimes C = \{b_i(u)c_j(v) \mid 0 \leq i \leq m, 0 \leq j \leq n\}$$

is a basis for  $P^{m,n}[u, v]$ .

**Question** : Show that elements of  $B \otimes C$  are linearly independent. Suppose that (as polynomials):

$$\sum_{i=0}^m \sum_{j=0}^n \alpha_{ij} b_i(u) c_j(v) = 0$$

Whence, for every  $u_0$ , we construct the polynomial:

$$p(u_0, v) = \sum_{j=0}^n \left( \sum_{i=0}^m \alpha_{ij} b_i(u_0) \right) c_j(v)$$

We see that  $p(u_0, v) = 0$  for all  $v$ , whence *every* coefficient of  $p(u_0, v)$  must be zero. In other words, for all  $j$  and  $u_0$ ,

$$\sum_{i=0}^m \alpha_{ij} b_i(u_0) = 0$$

Since,  $b_i$ 's are linearly independent, we are forced to conclude that  $\alpha_{ij} = 0$  for all  $i$  and  $j$ .

□

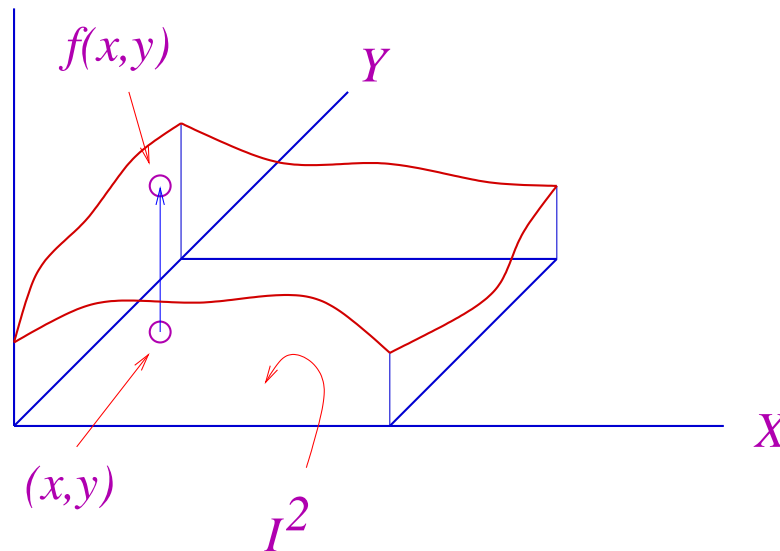
In particular we have the: **Bernstein Basis:**

$$\left\{ \binom{m}{i} u^i (1-u)^{m-i} \binom{n}{j} v^j (1-v)^{n-j} \mid 0 \leq i \leq m, 0 \leq j \leq n \right\}$$

We denote the typical basis element by  $B_i^m(u) B_j^n(v)$ .

## Functions and the Approximation Problem

$I$  with denote the interval  $[0, 1]$  and  $I^2$  the unit square  $[0, 1] \times [0, 1]$ . Let  $f : I^2 \rightarrow \mathbb{R}$  be a function on the unit square.



Is there a polynomial approximation to  $f$ ?

# The Bernstein-Weierstrass Approximation Theorem

Fix  $m$  and  $n$ , and form the data

$$S = \{f_{ij} = f(\frac{i}{m}, \frac{j}{n}) | 0 \leq i \leq m, 0 \leq j \leq n\}$$

We define the **Bernstein Approximation**

$$B^{m,n}(f)(u, v) = \sum_i \sum_j f_{ij} B_i^m(u) B_j^n(v)$$

*Theorem:* Let  $f$  be a function on  $I^2$ , and let  $\epsilon > 0$ . Then there are  $m, n$  such that  $|f(u, v) - B^{m,n}(f)(u, v)| < \epsilon$  for all  $(u, v) \in I^2$ .

Thus the 1-d situation has a complete 2-d analogue.

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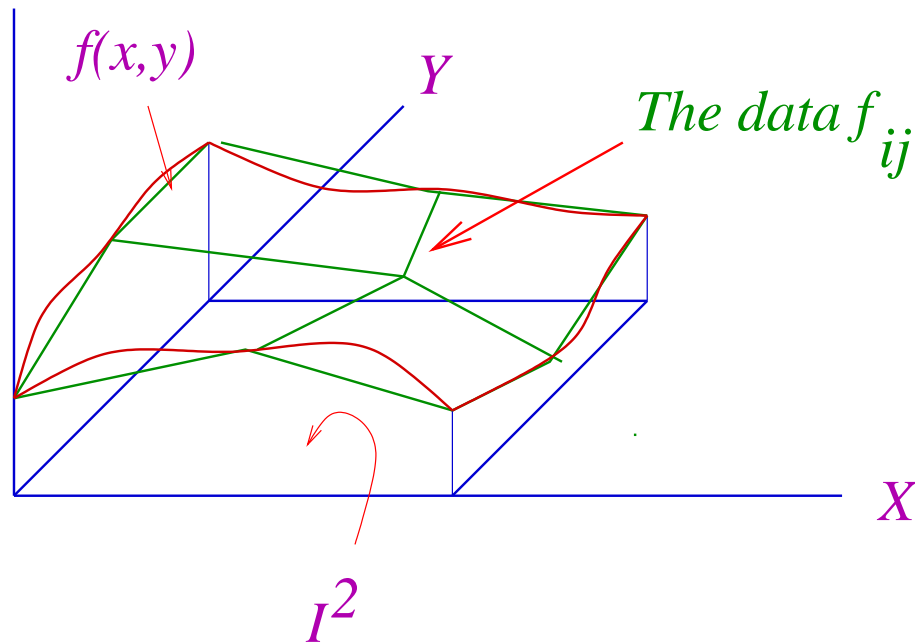
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## The Picture



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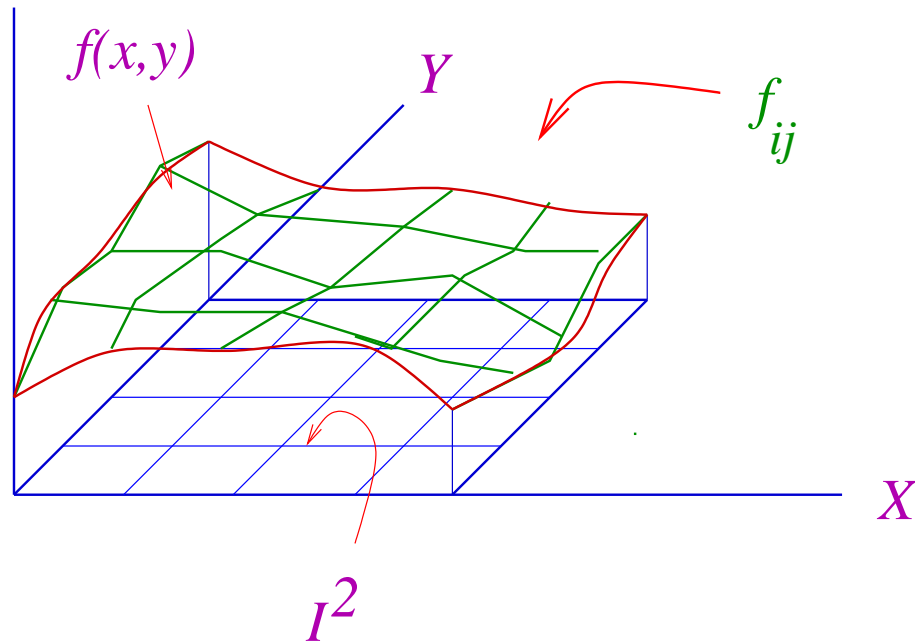
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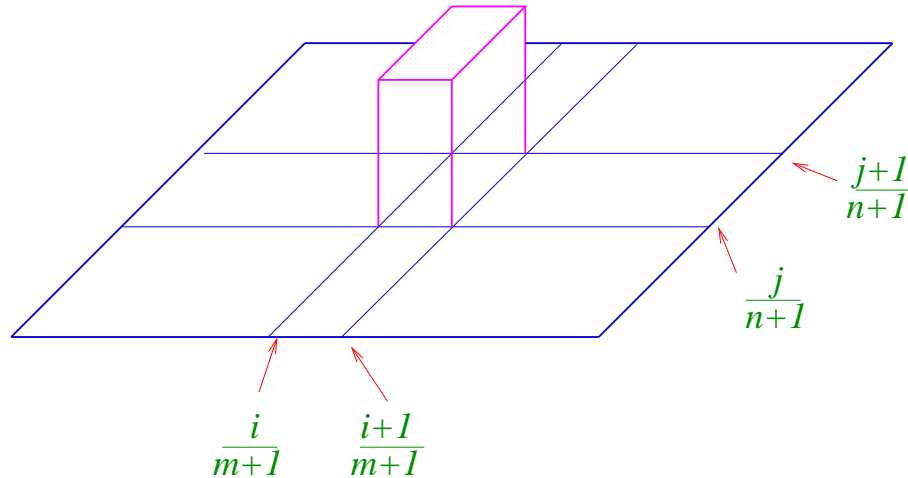
# The Finer Picture





## The Unit Step

As before it is convenient to associate  $B_i^m(u)B_j^n(v)$  with the 2-dimensional unit step function below. The ‘greville abscissa’ is obviously  $(\frac{i}{m}, \frac{j}{n})$  which occurs within the support of the step.



$$\text{As expected } \int_0^1 \int_0^1 B_i^m(u)B_j^n(v)dudv = \frac{1}{(m+1)(n+1)}.$$

## The Control Polygon

We will now discard the function  $f$ .

Let  $S$  be an  $m \times n$  matrix (in C++ notation, i.e.,  $[0 \dots m][0 \dots n]$ ) with entries in  $\mathbb{R}$  (or  $\mathbb{R}^3$ ).

$S$  is called the **Control Polygon**.

We define  $S(u, v)$  as:

$$S(u, v) = \sum_i \sum_j S[i, j] B_i^m(u) B_j^n(v)$$

$S$  will be called the **tensor-product** surface for the given control polygon.

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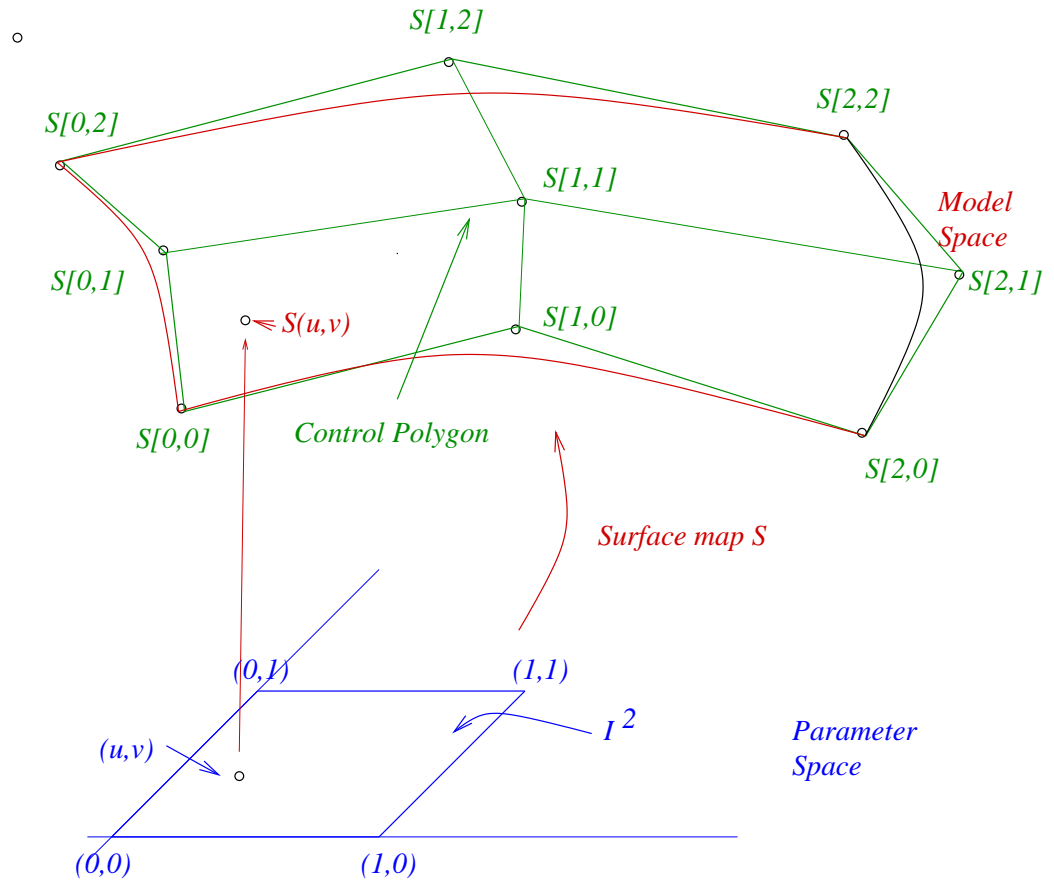
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## An example



## Some Observations

$$S(u, v) = \sum_i \sum_j S[i, j] B_i^m(u) B_j^n(v)$$

Lets evaluate  $S(0, 0)$ . Since  $B_i^m(0) = 0$  unless  $i = 0$  and  $B_j^n(0) = 0$  unless  $j = 0$ , we have  $S(0, 0) = S[0, 0]$ . Similarly, we have the other 'corner points'. Thus:

$$\begin{aligned} S(0, 0) &= S[0, 0] \\ S(1, 0) &= S[m, 0] \\ S(0, 1) &= S[0, n] \\ S(1, 1) &= S[m, n] \end{aligned}$$

## Boundary Curves

Next, let's look at  $S(u, 0)$ , which is the image of a boundary line of  $I^2$ . Again, since on this curve  $v = 0$ , we have  $B_j^n(0) = 0$  for  $j \neq 0$ . Thus the sum reduces to:

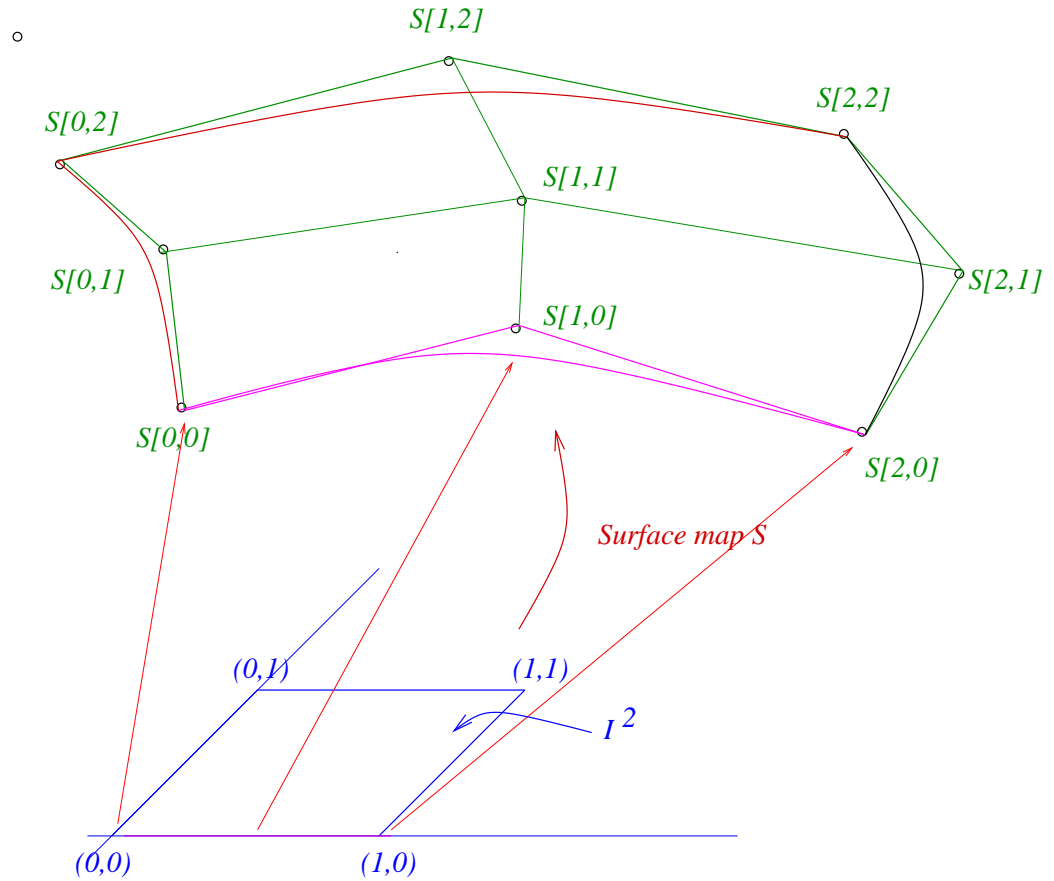
$$S(u, 0) = \sum_{i=0}^m S[i, 0] B_i^m(u)$$

This is clearly the **bezier curve** corresponding to the **first column** of  $S$  as its control points.

In general, we have:

$$\begin{array}{l} S(u, 0) = \sum_{i=0}^m S[i, 0] B_i^m(u) \\ S(u, 1) = \sum_{i=0}^m S[i, n] B_i^m(u) \\ S(0, v) = \sum_{j=0}^n S[0, j] B_j^n(v) \\ S(1, v) = \sum_{j=0}^n S[m, j] B_j^n(v) \end{array}$$

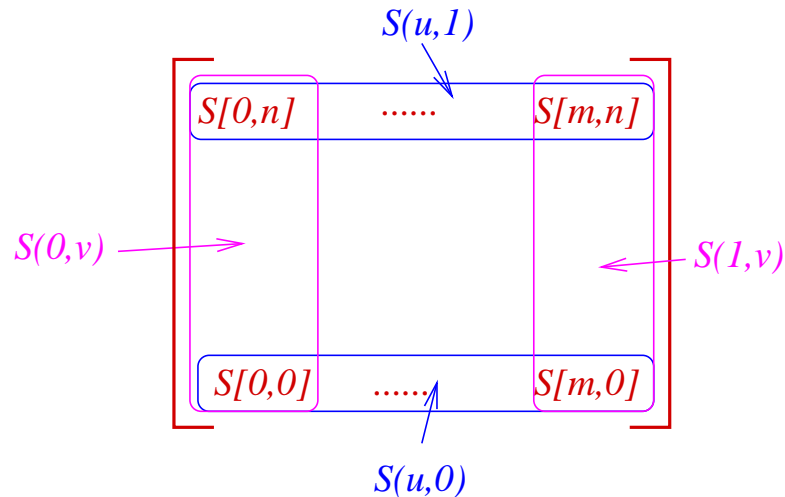
# Pictorially



## And Schematically

In terms of the control matrix, perhaps it is useful to use the *french notation* and number rows and columns from the bottom left corner. Then, we have:

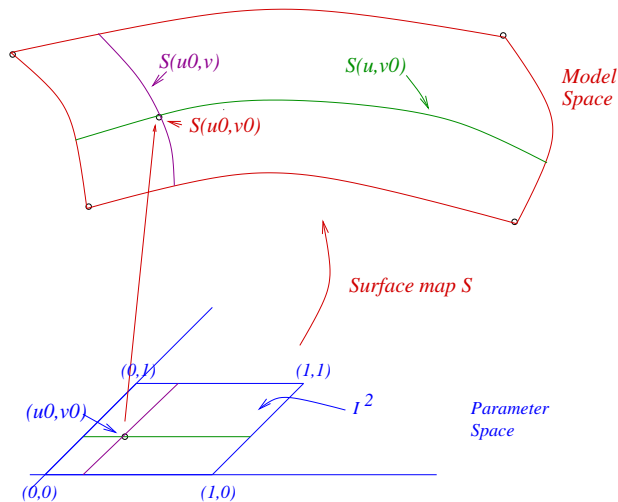
*The Control Matrix S*



## Iso-parametric Lines

But what about general  $S(u_0, v)$  for a fixed  $u_0$  and  $v \in [0, 1]$ ? Or  $S[u, v_0]$  for a fixed  $v_0$  but  $u$  ranging over  $[0, 1]$ ?

These curves (in the model space) are called **iso-parametric** lines. Thus  $S[u_0, v]$  is the iso-parametric line for  $u = u_0$ .





## Iso-parametric Lines contd.

Lets evaluate  $S(u_0, v)$ . Re-arranging the sum  $S(u, v)$ , we see that:

$$S(u_0, v) = \sum_{j=0}^n \left[ \sum_{i=0}^m S[i, j] B_i^m(u_0) \right] B_j^n(v)$$

We call  $\sum_{i=0}^m S[i, j] B_i^m(u_0)$  as  $S[u_0, j]$  and observe that  $S(u_0, v)$  is a **bezier curve with control points**  $[S[u_0, 0], S[u_0, 1], \dots, S[u_0, n]]$ .

Also, note that *each* of these control points  $S[u_0, j]$  is itself moving on a bezier curve parametrized by  $u$ .

Perhaps, the matrix notation is more convenient to observe this. We see that:

$$S(u, v) = [B_n^n(v), \dots, B_0^n(v)] \begin{bmatrix} S[0, n] & \dots & S[m, n] \\ \vdots & & \vdots \\ S[0, 0] & \dots & S[m, 0] \end{bmatrix} \begin{bmatrix} B_0^m(u) \\ \vdots \\ B_m^m(u) \end{bmatrix}$$

This may be concisely written as  $S(u, v) = B(v)SB(u)^T$ . Consequently, forming the product as  $S(u, v) = B(v)(SB(u)^T)$ , we see that:

$$S(u_0, v) = [B_n^n(v), \dots, B_0^n(v)] \begin{bmatrix} S[u_0, n] \\ \vdots \\ S[u_0, 0] \end{bmatrix}$$

Also note that  $\sum_i \sum_j B_i^m(u) B_j^n(v) = 1$  and thus  $S(u, v)$  is a **convex combination** of the entries of  $S$ .

## End Tangents and Normals

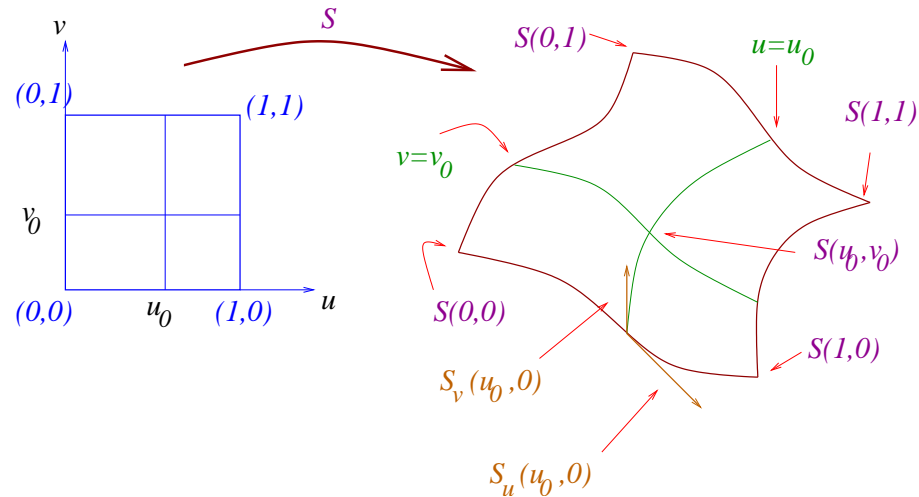
Given a map  $S : I^2 \rightarrow \mathbb{R}^3$  as we have already determined the boundary  $S(0, v)$ ,  $S(u, 0)$ , and so on. Other important data is the first-order data, viz., the tangents.

For convenience, let us consider the boundary point  $S(u_0, 0)$ . At any boundary point, we have **two** tangents to compute.

$$S_u(u_0, 0) = \lim_{u \rightarrow u_0} \frac{S(u, 0) - S(u_0, 0)}{u - u_0}$$
$$S_v(u_0, 0) = \lim_{v \rightarrow u_0} \frac{S(u_0, v) - S(u_0, 0)}{v}$$

These two tangents are shown in the next picture.

## An Example



The quantity  $S_u(u_0, 0)$  is easily computed as the derivative of the boundary  $S(u, 0) = \sum_{i=0}^m S[i, 0]B_i^m(u)$ . We may thus use the curve-tangent law explained earlier to get:

$$S_u(u_0, 0) = m \left[ \sum_{i=0}^{m-1} (S[i+1, 0] - S[i, 0]) B_i^{m-1}(u) \right]$$

$$S_v(u_0, v)$$

This quantity is a bit more delicate, since it is the tangent to the isoparametric curve  $S(u_0, v)$  at  $v = 0$ .

We have seen that:

$$S(u_0, v) = \sum_{j=0}^n S[u_0, j] B_j^n(v)$$

where  $S[u_0, j] = \sum_{i=0}^n S[i, j] B_i^m(u_0)$ .

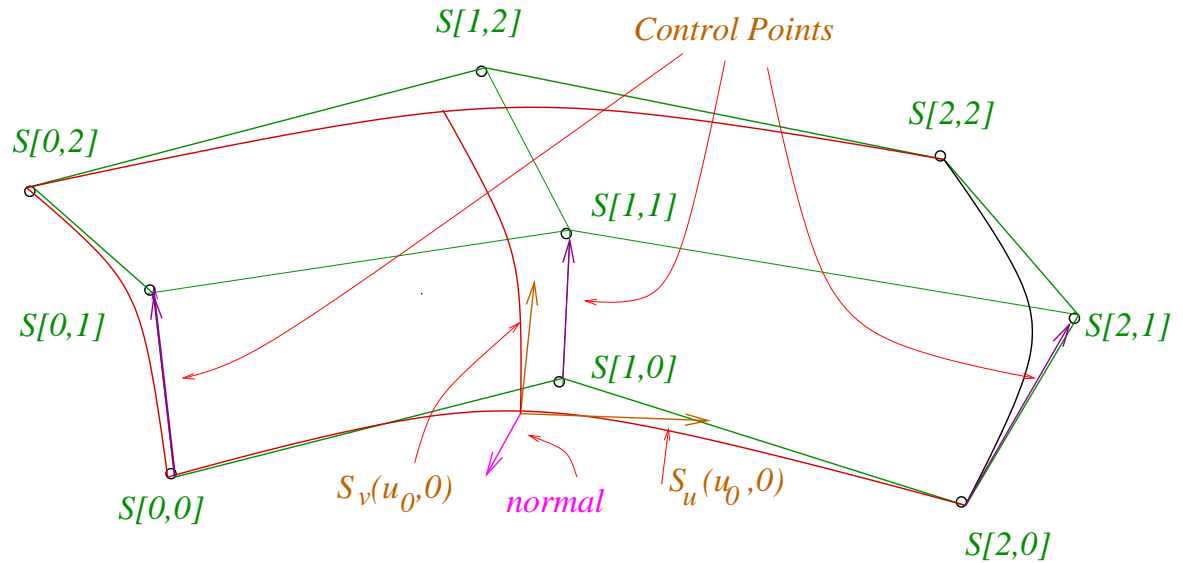
Thus  $S_v(u_0, 0)$ , the end-tangent to this curve, is  $m(S[u_0, 1] - S[u_0, 0])$ .

Back-substituting, we get:

$$\begin{aligned} S_v(u_0, 0) &= m \left[ \sum_{i=0}^n S[i, 1] B_i^m(u_0) - \sum_{i=0}^n S[i, 0] B_i^m(u_0) \right] \\ &= m \left[ \sum_{i=0}^n (S[i, 1] - S[i, 0]) B_i^m(u_0) \right] \end{aligned}$$

Thus  $S_v(u_0, 0)$  is also a bezier with control points  $[S[1, 0] - S[0, 0], S[1, 1] - S[1, 0], \dots, S[m, 1] - S[m, 0]]$ .

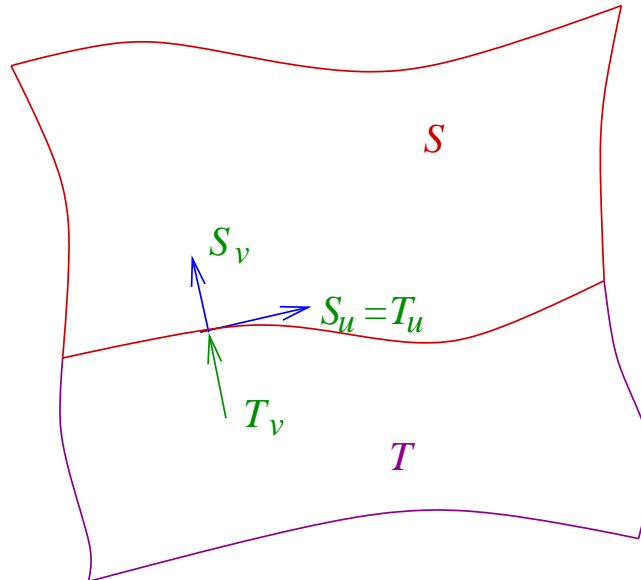
## Pictorially



The **normal** at that point is given by the **cross-product**  $S_v \times S_u$ .

# Splicing

Consider Two surfaces given by control points  $S$  and  $T$ . We would like to have them meet at a common boundary, and **smoothly**. Thus for example, we require  $S(u, 0) = T(u, 1)$  for all  $u \in [0, 1]$ . Furthermore, we require that the normals match too.



## The conditions

The condition  $S(u, 0) = T(u, 1)$  is easily satisfied by having the **bottom** row of  $S$  match the **top** row of  $T$ .

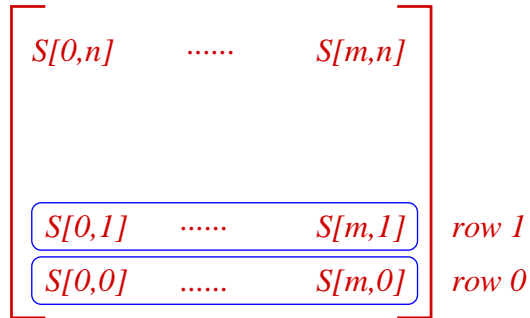
This will also ensure that  $S_u = T_u$  since both are tangents to the same curve.

Lets examine the normal condition next.  $S_u \times S_v \equiv T_u \times T_v$ , is achieved if we force  $S_v$  to be a multiple of  $T_v$ . This is forced by fixing a multiple, say  $\alpha$  and requiring that:

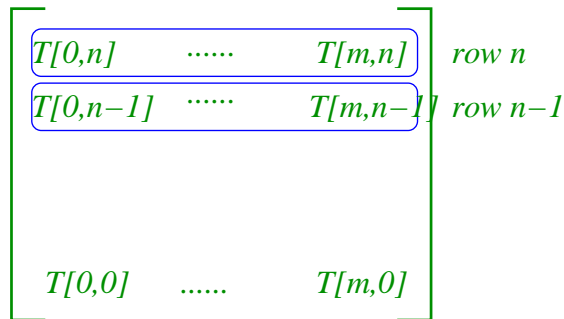
$$S[i, 1] - S[i, 0] = \alpha(T[i, n] - T[i, n - 1])$$



# Schematically

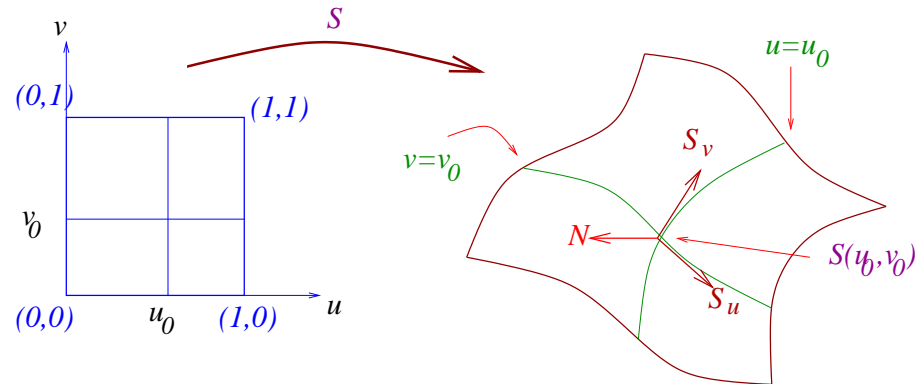


$$\begin{aligned} \text{row } 0 &= \text{row } n \\ \text{row } 0 - \text{row } 1 &= \\ &\text{row } n - \text{row } n-1 \end{aligned}$$



## The Normal System

A surface, if part of a solid, has at *every* point, an **outward normal**. Thus, given a  $(u_0, v_0)$  we are now faced with specifying *uniformly* an outward normal at  $S(u_0, v_0)$ !



Consider the figure above. At the point  $S(u_0, v_0)$ , we have the two tangents  $S_u$  and  $S_v$ . Let  $N = S_u \times S_v$ . Clearly the outward normal at  $S(u_0, v_0)$  must be **either  $N$  or  $-N$** .

## The Sign of the Normal

We claim that if the outward normal at  $S(u_0, v_0)$  is, say,  $-N = -(S_u \times S_v)$ , then it is so at **every**  $u, v$ <sup>a</sup>.

Thus *all* that needs to be stored is a  $sign \in \{+1, -1\}$ . The normal at any point  $S(u, v)$  is given by

$$sign \cdot (S_u \times S_v)$$

**Proof:** Let  $U(u, v)$  be the unit outward normal which exists! Clearly,  $U(u, v)$  is a smooth function on the surface.

Let  $M(u, v) = sign \cdot \frac{S_u \times S_v}{|S_u \times S_v|}$ . We see that (i)  $M(u, v)$  is a smooth function on  $u, v$ , and (ii)  $M(u, v)$  is normal at  $S(u, v)$ .

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<sup>a</sup>provided  $S_u \times S_v$  is never zero

## Continued

Thus at all points  $(u, v)$ , the vectors  $M(u, v)$  and  $U(u, v)$  are **collinear**.  
Now the proof goes in the following 3 steps:

- Since both are unit, we have  $M(u, v)/U(u, v) \in \pm 1$ .
- Since both  $U$  and  $M$  are smooth and unit,  $M(u, v)/U(u, v)$  must be **uniformly** either  $+1$  or  $-1$ .
- But we know that at  $(u_0, v_0)$  it is  $+1$  and thus  $M(u, v) = U(u, v)$ .



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## Things NOT covered

1. Surface Re-construction
2. Subdivision, Evaluation, Degree Elevation
3. Special Surfaces such as Coons-Patch
4. Tangent Planes, Gauss Map and Curvature