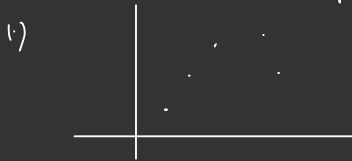


Cubic Splines

$n+1$ piece wise approx



$$1. a_i = f(x_i)$$

solving for d_i in (7)

$$d_i = \frac{1}{3h_i}(c_{i+1} - c_i) \quad (11)$$

now substitute d_i in (5) and solve for b_i

$$b_i = \frac{a_{i+1} - a_i}{h_i} - \frac{2c_i + c_{i+1}}{3}h_i \quad (10)$$

$$\begin{pmatrix} 1 & 0 & 0 & 0 & 0 & \dots & 0 \\ h_0 & 2(h_0 + h_1) & h_1 & 0 & 0 & \dots & 0 \\ 0 & h_1 & 2(h_1 + h_2) & h_2 & 0 & \dots & 0 \\ 0 & 0 & h_2 & 2(h_2 + h_3) & h_3 & \dots & 0 \\ 0 & 0 & 0 & h_3 & 2(h_3 + h_4) & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & 0 & \dots & 1 \end{pmatrix} \begin{pmatrix} c_0 \\ c_1 \\ c_2 \\ c_3 \\ \vdots \\ \vdots \\ c_n \end{pmatrix} = \begin{pmatrix} 0 \\ \frac{3}{h_1}(a_2 - a_1) - \frac{3}{h_0}(a_1 - a_0) \\ \frac{3}{h_2}(a_3 - a_2) - \frac{3}{h_1}(a_2 - a_1) \\ \frac{3}{h_3}(a_4 - a_3) - \frac{3}{h_2}(a_3 - a_2) \\ \vdots \\ \vdots \\ 0 \end{pmatrix}$$

Simple example

Example 1 Construct a natural cubic spline that passes through the points (1, 2), (2, 3), and (3, 5).

Solution This spline consists of two cubics. The first for the interval [1, 2], denoted

$$S_0(x) = a_0 + b_0(x - 1) + c_0(x - 1)^2 + d_0(x - 1)^3,$$

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and the other for [2, 3], denoted

$$S_1(x) = a_1 + b_1(x - 2) + c_1(x - 2)^2 + d_1(x - 2)^3.$$

There are 8 constants to be determined, which requires 8 conditions. Four conditions come from the fact that the splines must agree with the data at the nodes. Hence

$$\begin{aligned} 2 &= f(1) = a_0, & 3 &= f(2) = a_0 + b_0 + c_0 + d_0, & 3 &= f(2) = a_1, & \text{and} \\ 5 &= f(3) = a_1 + b_1 + c_1 + d_1. \end{aligned}$$

Two more come from the fact that $S'_0(2) = S'_1(2)$ and $S''_0(2) = S''_1(2)$. These are

$$S'_0(2) = S'_1(2): \quad b_0 + 2c_0 + 3d_0 = b_1 \quad \text{and} \quad S''_0(2) = S''_1(2): \quad 2c_0 + 6d_0 = 2c_1$$

The final two come from the natural boundary conditions:

$$S''_0(1) = 0: \quad 2c_0 = 0 \quad \text{and} \quad S''_1(3) = 0: \quad 2c_1 + 6d_1 = 0.$$

Solving this system of equations gives the spline

$$S(x) = \begin{cases} 2 + \frac{3}{2}(x - 1) + \frac{1}{4}(x - 1)^3, & \text{for } x \in [1, 2] \\ 3 + \frac{3}{2}(x - 2) + \frac{3}{4}(x - 2)^2 - \frac{1}{4}(x - 2)^3, & \text{for } x \in [2, 3] \end{cases}$$

Example 2 At the beginning of Chapter 3 we gave some Taylor polynomials to approximate the exponential $f(x) = e^x$. Use the data points $(0, 1)$, $(1, e)$, $(2, e^2)$, and $(3, e^3)$ to form a natural spline $S(x)$ that approximates $f(x) = e^x$.

Solution We have $n = 3$, $h_0 = h_1 = h_2 = 1$, $a_0 = 1$, $a_1 = e$, $a_2 = e^2$, and $a_3 = e^3$. So the matrix A and the vectors \mathbf{b} and \mathbf{x} given in Theorem 3.11 have the forms

$$A = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 4 & 1 & 0 \\ 0 & 1 & 4 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 0 \\ 3(e^2 - 2e + 1) \\ 3(e^3 - 2e^2 + e) \\ 0 \end{bmatrix}, \quad \text{and} \quad \mathbf{x} = \begin{bmatrix} c_0 \\ c_1 \\ c_2 \\ c_3 \end{bmatrix}.$$

The vector-matrix equation $A\mathbf{x} = \mathbf{b}$ is equivalent to the system of equations

$$\begin{aligned} c_0 &= 0, \\ c_0 + 4c_1 + c_2 &= 3(e^2 - 2e + 1), \\ c_1 + 4c_2 + c_3 &= 3(e^3 - 2e^2 + e), \\ c_3 &= 0. \end{aligned}$$

This system has the solution $c_0 = c_3 = 0$, and to 5 decimal places,

$$c_1 = \frac{1}{5}(-e^3 + 6e^2 - 9e + 4) \approx 0.75685, \quad \text{and} \quad c_2 = \frac{1}{5}(4e^3 - 9e^2 + 6e - 1) \approx 5.83007.$$

Solving for the remaining constants gives

$$\begin{aligned} b_0 &= \frac{1}{h_0}(a_1 - a_0) - \frac{h_0}{3}(c_1 + 2c_0) \\ &= (e - 1) - \frac{1}{15}(-e^3 + 6e^2 - 9e + 4) \approx 1.46600, \\ b_1 &= \frac{1}{h_1}(a_2 - a_1) - \frac{h_1}{3}(c_2 + 2c_1) \\ &= (e^2 - e) - \frac{1}{15}(2e^3 + 3e^2 - 12e + 7) \approx 2.22285, \\ b_2 &= \frac{1}{h_2}(a_3 - a_2) - \frac{h_2}{3}(c_3 + 2c_2) \\ &= (e^3 - e^2) - \frac{1}{15}(8e^3 - 18e^2 + 12e - 2) \approx 8.80977, \\ d_0 &= \frac{1}{3h_0}(c_1 - c_0) = \frac{1}{15}(-e^3 + 6e^2 - 9e + 4) \approx 0.25228, \\ d_1 &= \frac{1}{3h_1}(c_2 - c_1) = \frac{1}{3}(e^3 - 3e^2 + 3e - 1) \approx 1.69107, \end{aligned}$$

and

$$d_2 = \frac{1}{3h_2}(c_3 - c_1) = \frac{1}{15}(-4e^3 + 9e^2 - 6e + 1) \approx -1.94336.$$

The natural cubic spline is described piecewise by

$$S(x) = \begin{cases} 1 + 1.46600x + 0.25228x^3, & \text{for } x \in [0, 1], \\ 2.71828 + 2.22285(x - 1) + 0.75685(x - 1)^2 + 1.69107(x - 1)^3, & \text{for } x \in [1, 2], \\ 7.38906 + 8.80977(x - 2) + 5.83007(x - 2)^2 - 1.94336(x - 2)^3, & \text{for } x \in [2, 3]. \end{cases}$$

Derivation.

n polynomials: \checkmark
 \checkmark
 $S_i(x) = a_i + b_i(x - x_i) + c_i(x - x_i)^2 + d_i(x - x_i)^3$
 $4n$ unknowns

Conditions:
 ① $S_i(x_i) = y_i \quad \forall i = 0, \dots, n$
 ② $S_i(x_{i+1}) = S_{i+1}(x_{i+1}) \quad \forall i = 0, \dots, n-1$
 ③ $S_i'(x_{i+1}) = S_{i+1}'(x_{i+1}) \quad \forall i = 0, \dots, n-1$
 ④ $S_i''(x_{i+1}) = S_{i+1}''(x_{i+1}) \quad \forall i = 0, \dots, n-1$
 ⑤ $S_0''(x_0) = 0; \quad S_{n-1}''(x_n) = 0$

$$\begin{aligned} S_i(x) &= a_i + b_i(x - x_i) + c_i(x - x_i)^2 + d_i(x - x_i)^3 \\ S_{i+1}(x) &= a_{i+1} + b_{i+1}(x - x_{i+1}) + c_{i+1}(x - x_{i+1})^2 + d_{i+1}(x - x_{i+1})^3 \\ S_i'(x) &= b_i + 2c_i(x - x_i) + 3d_i(x - x_i)^2 \\ S_{i+1}'(x) &= b_{i+1} + 2c_{i+1}(x - x_{i+1}) + 3d_{i+1}(x - x_{i+1})^2 \\ S_i''(x) &= 2c_i + 6d_i(x - x_i) \\ S_{i+1}''(x) &= 2c_{i+1} + 6d_{i+1}(x - x_{i+1}) \end{aligned}$$

$h_i = x_{i+1} - x_i$

$$\begin{aligned} ① \quad S_i(x_i) &= y_i \\ a_i &= y_i \\ ② \quad S_{i+1}(x_{i+1}) &= S_i(x_{i+1}) \\ a_{i+1} &= a_i + b_i h_i + c_i h_i^2 + d_i h_i^3 \quad [5] \\ ③ \quad S_{i+1}'(x_{i+1}) &= S_i'(x_{i+1}) \\ b_{i+1} &= b_i + 2c_i h_i + 3d_i h_i^2 \quad [6] \\ ④ \quad S_{i+1}''(x_{i+1}) &= S_i''(x_{i+1}) \\ 2c_{i+1} &= 2c_i + 6d_i h_i \\ c_{i+1} &= c_i + 3d_i h_i \quad [7] \end{aligned}$$

$$\begin{aligned} ⑤ \quad S_0''(x_0) &= 0 & S_{n-1}''(x_n) &= 0 \\ c_0 &= 0 & c_{n-1} &= 0 \end{aligned}$$

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$$P_n(x) = a_n x^n + a_{n-1} x^{n-1} + \cdots + a_1 x + a_0,$$

It is helpful to write the equations as follows:

$$\begin{aligned} a_0 \sum_{i=1}^m x_i^0 + a_1 \sum_{i=1}^m x_i^1 + a_2 \sum_{i=1}^m x_i^2 + \cdots + a_n \sum_{i=1}^m x_i^n &= \sum_{i=1}^m y_i x_i^0, \\ a_0 \sum_{i=1}^m x_i^1 + a_1 \sum_{i=1}^m x_i^2 + a_2 \sum_{i=1}^m x_i^3 + \cdots + a_n \sum_{i=1}^m x_i^{n+1} &= \sum_{i=1}^m y_i x_i^1, \\ &\vdots \\ a_0 \sum_{i=1}^m x_i^n + a_1 \sum_{i=1}^m x_i^{n+1} + a_2 \sum_{i=1}^m x_i^{n+2} + \cdots + a_n \sum_{i=1}^m x_i^{2n} &= \sum_{i=1}^m y_i x_i^n. \end{aligned}$$

Example 2 Fit the data in Table 8.3 with the discrete least squares polynomial of degree at most 2.

Solution For this problem, $n = 2$, $m = 5$, and the three normal equations are

$$\begin{aligned} 5a_0 + 2.5a_1 + 1.875a_2 &= 8.7680, \\ 2.5a_0 + 1.875a_1 + 1.5625a_2 &= 5.4514, \\ 1.875a_0 + 1.5625a_1 + 1.3828a_2 &= 4.4015. \end{aligned}$$

To solve this system using Maple, we first define the equations

$$\begin{aligned} eq1 &:= 5a_0 + 2.5a_1 + 1.875a_2 = 8.7680; \\ eq2 &:= 2.5a_0 + 1.875a_1 + 1.5625a_2 = 5.4514; \\ eq3 &:= 1.875a_0 + 1.5625a_1 + 1.3828a_2 = 4.4015 \end{aligned}$$

Table 8.3

i	x_i	y_i
1	0	1.0000
2	0.25	1.2840
3	0.50	1.6487
4	0.75	2.1170
5	1.00	2.7183

1.5

i	x_i	y_i	$\ln y_i$	x_i^2	$x_i \ln y_i$
1	1.00	5.10	1.629	1.0000	1.629
2	1.25	5.79	1.756	1.5625	2.195
3	1.50	6.53	1.876	2.2500	2.814
4	1.75	7.45	2.008	3.0625	3.514
5	2.00	8.46	2.135	4.0000	4.270
	7.50		9.404	11.875	14.422

If x_i is graphed with $\ln y_i$, the data appear to have a linear relation, so it is reasonable to assume an approximation of the form

$$y = be^{ax}, \quad \text{which implies that} \quad \ln y = \ln b + ax.$$

Extending the table and summing the appropriate columns gives the remaining data in Table 8.5.

Using the normal equations (8.1) and (8.2),

$$a = \frac{(5)(14.422) - (7.5)(9.404)}{(5)(11.875) - (7.5)^2} = 0.5056$$

and

$$\ln b = \frac{(11.875)(9.404) - (14.422)(7.5)}{(5)(11.875) - (7.5)^2} = 1.122.$$

With $\ln b = 1.122$ we have $b = e^{1.122} = 3.071$, and the approximation assumes the form

$$y = 3.071e^{0.5056x}.$$

$$a_0 = \frac{\sum_{i=1}^m x_i^2 \sum_{i=1}^m y_i - \sum_{i=1}^m x_i y_i \sum_{i=1}^m x_i}{m \left(\sum_{i=1}^m x_i^2 \right) - \left(\sum_{i=1}^m x_i \right)^2}$$

$$a_1 = \frac{m \sum_{i=1}^m x_i y_i - \sum_{i=1}^m x_i \sum_{i=1}^m y_i}{m \left(\sum_{i=1}^m x_i^2 \right) - \left(\sum_{i=1}^m x_i \right)^2}.$$

$$p_n(x) = \sum_{j=0}^n a_j x^j \quad \left\{ \begin{array}{l} \text{---} \\ \text{---} \end{array} \right.$$

$$E(a_0, a_1, \dots, a_n) = \sum_{i=1}^m (p_n(x_i) - y_i)^2$$

$$= \sum_{i=1}^m \left(\sum_{j=0}^n a_j x_i^j - y_i \right)^2$$

$$\frac{\partial E}{\partial a_0} = 0, \quad \frac{\partial E}{\partial a_1} = 0 \quad \dots \quad \frac{\partial E}{\partial a_n} = 0 \quad \left\{ \begin{array}{l} \frac{\partial a_j}{\partial a_n} = 1 \\ \frac{\partial a_j}{\partial a_n} = 0 \end{array} \right.$$

$$\frac{\partial E}{\partial a_n} = \sum_{i=1}^m 2 \left(\sum_{j=0}^n a_j x_i^j - y_i \right) \left(\sum_{j=0}^n x_i^j \frac{\partial a_j}{\partial a_n} \right)$$

$$p_n(x) = \sum_{j=0}^n a_j x^j \quad \left\{ \begin{array}{l} \text{---} \\ \text{---} \end{array} \right.$$

$$E(a_0, a_1, \dots, a_n) = \sum_{i=1}^m (p_n(x_i) - y_i)^2$$

$$\frac{\partial E}{\partial a_n} = \sum_{i=1}^m 2 \left(\sum_{j=0}^n a_j x_i^j - y_i \right) \left(\sum_{j=0}^n x_i^j \frac{\partial a_j}{\partial a_n} \right)$$

$$\frac{\partial E}{\partial a_n} = 2 \sum_{i=1}^m \left(\sum_{j=0}^n a_j x_i^j - y_i \right) x_i^k = 0 \quad \dots \quad k=0 \dots n$$

$$\sum_{i=1}^m \sum_{j=0}^n a_j x_i^j x_i^k - \sum_{i=1}^m x_i^k y_i = 0 \quad \forall k=0 \dots n$$

$$n a_0 + \left(\sum_{i=1}^m x_i \right) a_1 + \left(\sum_{i=1}^m x_i^2 \right) a_2 + \dots - \left(\sum_{i=1}^m x_i^n \right) a_n = \sum_{i=1}^m y_i$$

$$\left(\sum_{i=1}^m x_i \right) a_0 + \left(\sum_{i=1}^m x_i^2 \right) a_1 + \left(\sum_{i=1}^m x_i^3 \right) a_2 + \dots - \left(\sum_{i=1}^m x_i^{n+1} \right) a_n = \sum_{i=1}^m x_i y_i$$

$$\vdots$$

$$\vdots$$

$$\left(\sum_{i=1}^m x_i^n \right) a_0 + \left(\sum_{i=1}^m x_i^{n+1} \right) a_1 + \left(\sum_{i=1}^m x_i^{n+2} \right) a_2 + \dots - \left(\sum_{i=1}^m x_i^{2n} \right) a_n = \sum_{i=1}^m x_i^n y_i$$

$$\begin{pmatrix} m & \sum x_i & \sum x_i^2 & \sum x_i^3 & \dots & \sum x_i^n \\ \sum x_i & \sum x_i^2 & \sum x_i^3 & \sum x_i^4 & \dots & \sum x_i^{n+1} \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ \sum x_i^n & \sum x_i^{n+1} & \sum x_i^{n+2} & \dots & \dots & \sum x_i^{2n} \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{pmatrix} = \begin{pmatrix} \sum y_i \\ \sum x_i y_i \\ \vdots \\ \sum x_i^n y_i \end{pmatrix}$$

$$(y - x\theta)^T (y - x\theta)$$

$$\Rightarrow (y^T - \theta^T x^T) (y - x\theta)$$

$$\frac{1}{2} \frac{\partial}{\partial \theta} (y^T y - y^T x \theta - \theta^T x^T y + \theta^T x^T x \theta) = 0$$

$$\theta = (x^T x)^{-1} x^T y$$

$$\theta_0 + \theta_1 n_1 + \theta_2 n_2 + \dots + \theta_n n_n - y$$

$$y_{pred} = x\theta$$

$$(x\theta - y)$$

$$y_1$$

$$\begin{pmatrix} x_1 & x_2 & \dots & x_n \end{pmatrix} \begin{pmatrix} \theta_0 \\ \theta_1 \\ \vdots \\ \theta_n \end{pmatrix}$$

$$\frac{1}{2} (x^2 - y^2) \mp (x - y)$$

$$T_n(x) = \cos(n \cos^{-1} x)$$

!!

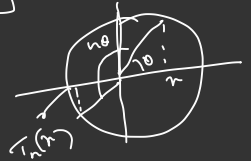
3 cases $\theta = \cos^{-1} x$

$$\cos(n\theta) = 0 \quad n \leq 2$$

$$\Rightarrow n\theta = \frac{\pi}{2} + k\pi$$

$$\theta = \frac{\pi + 2k\pi}{2n}$$

$$T_n \left[\begin{smallmatrix} 1 & 1 \\ 1 & 1 \end{smallmatrix} \right] \rightarrow \left[\begin{smallmatrix} 1 & 1 \\ 1 & 1 \end{smallmatrix} \right]$$



1) $n=0 \rightarrow n=0$ zeros

2) $n=1 \Rightarrow \theta = \frac{\pi}{2} \Rightarrow x=0$

3) $n=2 \Rightarrow \theta = \frac{\pi}{4}, \frac{3\pi}{4} \Rightarrow x = \pm \frac{1}{\sqrt{2}}$

for $n \geq 0$
 \forall exactly n zeros

$$\theta = \frac{\pi}{2n}, \frac{3\pi}{2n}, \frac{5\pi}{2n}, \dots, \frac{(2n-1)\pi}{2n}$$

max & min

$$\cos n\theta = \pm 1$$

$$n\theta = k\pi$$

$$\theta = \frac{k\pi}{n} = 0, \frac{\pi}{2}, \frac{2\pi}{n}, \dots, \pi$$

So $|T_n(x)|$ assume max exactly $n+1$ times.

Using Euler's formula:

$$e^{i\theta} = \cos\theta + i\sin\theta$$

$$\cos(\theta + \phi) + i\sin(\theta + \phi) = e^{i(\theta + \phi)}$$

$$= e^{i\theta} e^{i\phi}$$

$$= (\cos\theta + i\sin\theta)(\cos\phi + i\sin\phi)$$

$$= (\cos\theta\cos\phi - \sin\theta\sin\phi) + i(\cos\theta\sin\phi + \sin\theta\cos\phi)$$

De Moivre's formula

$$\begin{aligned}\cos \theta + i \sin \theta &= e^{i\theta} \\ &= (e^{i\theta})^n \\ &= (\cos \theta + i \sin \theta)^n \\ &= \cos^n \theta + n i \cos^{n-1} \theta \cdot \sin \theta - \frac{n(n-1)}{2} \cos^{n-2} \theta \sin^2 \theta + \dots\end{aligned}$$

Setting real parts equal:

$$\begin{aligned}\cos n\theta &= \cos^n \theta - \frac{n(n-1)}{2} \cos^{n-2} \theta \sin^2 \theta + \dots \\ &= \cos^n \theta - \frac{n(n-1)}{2} \cos^{n-2} \theta (1 - \cos^2 \theta) + \dots \\ &= \cos^n \theta - \frac{n(n-1)}{2} \cos^{n-2} \theta + \dots\end{aligned}$$

$\boxed{n = \cos \theta}$

$$T_n(n) = x^n - \frac{n(n-1)}{2} x^{n-2} (1-x^2) + \dots$$

Taylor recursion formula

$$\begin{aligned}T_{n+1}(\theta) &= \cos(n\theta + \theta) \\ &= \cos n\theta \cdot \cos \theta - \sin n\theta \sin \theta\end{aligned}$$

$$\begin{aligned}T_{n-1}(\theta) &= \cos(n\theta - \theta) \\ &= \cos n\theta \cos \theta + \sin n\theta \sin \theta\end{aligned}$$

$$\begin{aligned}T_{n+1} + T_{n-1} &= 2 \cos n\theta \cdot \cos \theta \\ &= 2 \cos \theta T_n(\theta)\end{aligned}$$

$$T_{n+1}(\theta) = 2 \cos \theta T_n(\theta) - T_{n-1}(\theta)$$

$$T_n(x) = \cos(n \cos^{-1} x)$$

$$T_0(x) = 1$$

$$T_1(x) = x$$

$$T_2(x) = 2x^2 - 1$$

$$\begin{aligned}T_n &= 2x(2x^2 - 1) - x \\ &= 4x^3 - 3x\end{aligned}$$

- 1) T_n has n distinct zeros
- 2) $|T_n(x)|$ takes its max at exactly $n+1$ distinct points separ. by zeros
- 3) $T_{\text{even}}(x)$ is even function
- 4) $n > 0$, leading coeff $\rightarrow 2^{n-1}$

Optimally $\hat{T}_n = \frac{T_n(n)}{2^n}$ (minimizing coefficient = 1)

$= 2^n + c_{n-1} 2^{n-1} + \dots$

\hat{T}_n smallest $m = x$
 absolute value on $[1, 1]$

$m \rightarrow -1 \leq n \leq 1$ $|\hat{T}_n(n)| = \frac{1}{2^{n-1}}$

may $|p(n)| \geq \frac{1}{2^{n-1}}$
 $-1 \leq n \leq 1$

~~proof~~ σ

let $m \rightarrow -1 \leq n \leq 1$ $|p(n)| < \frac{1}{2^{n-1}}$

why useful?

$\hat{T}_{n+1}(n) = (n-r_0) \dots (n-r_n)$

$\begin{matrix} k=0 \\ \vdots \\ n \end{matrix}$ $\left\{ \begin{matrix} a_0 \leq \phi_0 \phi_0 & \dots & \end{matrix} \right\}$

$\left[\begin{matrix} \sum \phi_0^2 & \sum \phi_0 \phi_1 & \dots & \sum \phi_0 \phi_n \\ \sum \phi_1 \phi_0 & \sum \phi_1^2 & \dots & \sum \phi_1 \phi_n \\ \vdots & \vdots & \ddots & \vdots \\ \sum \phi_n \phi_0 & \sum \phi_n \phi_1 & \dots & \sum \phi_n^2 \end{matrix} \right] \begin{bmatrix} a_0 \\ \vdots \\ a_n \end{bmatrix} =$

$\left(\begin{matrix} \sum y_i \phi_0 \\ \sum y_i \phi_1 \\ \vdots \end{matrix} \right)$