# 03.04 cubic spines

so theres interpolating polynomials for (sort-of) random  $x_i$ , evenly spaced  $x_i$  and chebyshev  $x_i$ . what about other ways to connect the dots? what if the word "continuous" gets taken away?

### code, visual

```
import matplotlib.pyplot as plt
import numpy as np
import scipy as sp

#https://docs.scipy.org/doc/scipy/reference/generated/scipy.interpolate.CubicSpline.html

def main():--

if __name__ == "__main__":--

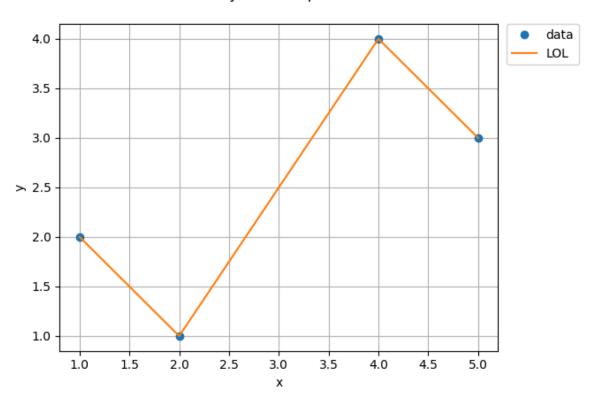
and

if __name__ == "__main__":--

37
```

### $\overline{\mathbf{T}}$

## my second spline



✓ usw

no ones gonna pay money for that but lets write it out seriously:

$$S_1(x) = 2 - (x - 1)$$
 on  $[1, 2]$   
 $S_2(x) = 1 + \frac{3}{2}(x - 2)$  on  $[2, 4]$   
 $S_3(x) = 4 - (x - 4)$  on  $[4, 5]$ .

and if it smoothed out in a cubic-spline kind of way,

$$S_1(x) = 2 - \frac{13}{8}(x-1) + 0 \cdot (x-1)^2 + \frac{5}{8}(x-1)^3$$
 on  $[1,2]$   
 $S_2(x) = 1 + \frac{1}{4}(x-2) + \frac{15}{8}(x-2)^2 - \frac{5}{8}(x-2)^3$  on  $[2,4]$   
 $S_3(x) = 4 + \frac{1}{4}(x-4) - \frac{15}{8}(x-4)^2 + \frac{5}{8}(x-4)^3$  on  $[4,5]$ .

note the smooth transition from one  $S_i$  to the next at the middle base points, or "knots". this is achieved by having neighboring pieces  $S_i$ ,  $S_{i+1}$  have the same zeroth, first and second derivatives at those points.

given n points  $(x_1, y_1), \ldots, (x_n, y_n)$ , there is only one linear spline but there are infinitely many cubic splines. ie, more conditions are needed to decide the spline.

# 1 properties of splines

assume the n given data points  $(x_1, y_1), \ldots, (x_n, y_n)$  are of distinct  $x_i$  and in increasing order. a **cubic spline** S(x) through data points  $(x_1, y_1), \ldots, (x_n, y_n)$  is a set of cubic polynomials.

$$egin{aligned} S_1(x) &= y_1 + b_1(x-x_1) + c_1(x-x_1)^2 + d_1(x-x_1)^3 & ext{on } [x_1,x_2] \ S_2(x) &= y_2 + b_2(x-x_2) + c_2(x-x_2)^2 + d_2(x-x_2)^3 & ext{on } [x_2,x_3] \ &dots \ S_{n-1}(x) &= y_{n-1} + b_{n-1}(x-x_{n-1}) + c_{n-1}(x-x_{n-1})^2 + d_{n-1}(x-x_{n-1})^3 & ext{on } [x_{n-1},x_n] \end{aligned}$$

with the following properties:

property 01 
$$S_i(x_i)=y_i, S_i(x_{i+1})=y_{i+1}$$
 for  $i=1,\ldots,n-1$ .

property 02 
$$S_{i-1}'(x_i) = S_i(x_i)'$$
 for  $i=2,\ldots,n-1$ .

property 03 
$$S_{i-1}''(x_i)=S_i(x_i)''$$
 for  $i=2,\ldots,n-1$ .

property 01 guarantees that spline S(x) interpolates the data points; property 02 forces the slopes neighboring segments to agree where they meet; property 03 does the same wrt curvature.

property 04a natural spline  $S_1''(x_1)=0, S_{n-1}''(x_n)=0.$ 

n points will have n-1 segments, each represented by a cubic and each cubic will have a coefficient for each of its term. term  $a_0x^0=y_i$ , leaving  $b_i\equiv a_2, c_i\equiv a_i, d_i\equiv a_3$  per equation i.

property 1 at  $x_{i+1} 
ightarrow n-1$  equations:

$$egin{aligned} y_2 &= S_1(x_2) = y_1 + b_1(x_2 - x_1) + c_1(x_2 - x_1)^2 + d_1(x_2 - x_1)^3 \ &dots \ y_n &= S_{n-1}(x_n) = y_{n-1} + b_{n-1}(x_n - x_{n-1}) + c_{n-1}(x_n - x_{n-1})^2 + d_{n-1}(x_n - x_{n-1})^3. \end{aligned}$$

property 02 ightarrow n-2 equations :

$$egin{aligned} 0 &= S_1'(x_2) - S_2'(x_2) = b_1 + 2c_1(x_2 - x_1) + 3d_1(x_2 - x_1)^2 - b_2 \ &\vdots \ 0 &= S_{n-2}'(x_{n-1}) - S_{n-1}'(x_{n-1}) = b_{n-2} + 2c_{n-2}(x_{n-1} - x_{n-2}) + 3d_{n-2}(x_{n-1} - x_{n-2})^2 - b_{n-1}. \end{aligned}$$

property 03 ightarrow n-2 equations :

$$egin{aligned} 0 &= S_1''(x_2) - S_2''(x_2) = 2c_1 + 6d_1(x_2 - x_1) - 2c_2 \ &\vdots \ 0 &= S_{n-2}''(x_{n-1}) - S_{n-1}''(x_{n-1}) = 2c_{n-2} + 6d_{n-2}(x_{n-1} - x_{n-2}) - 2c_{n-1}. \end{aligned}$$

instead of solving  $3 \times (n-1)$  equations all at once, break it down. find  $c_i$  within a subset of those equations, then solve out from there.

consider  $c_n=S_{n-1}''(x_n)\cdot \frac{1}{2}.$  let  $\delta_i=x_{i+1}-x_i, \Delta_i=y_{i-1}-y_i.$  then property 03 equations resolve to

$$d_i = rac{c_{i+1} - c_i}{3\delta_i} \quad i = 1, \ldots, n-1$$

and property 01 equations reduce to

$$egin{aligned} b_i &= rac{\Delta_i}{\delta_i} - c_i \delta_i - d_i \delta_i^2 \ &= rac{\Delta_i}{\delta_i} - c_i \delta_i - rac{\delta_i}{3} (c_{i-1} - c_i) \ &= rac{\Delta_i}{\delta_i} - rac{\delta_i}{3} (2c_i - c_{i+1}) \quad i = 1, \dots, n-1. \end{aligned}$$

ie, both b, d are both in terms of c. so

$$egin{aligned} \delta_1 c_1 + 2(\delta_1 + \delta_2) c_2 + \delta_2 c_3 &= 3\left(rac{\Delta_2}{\delta_2} - rac{\Delta_1}{\delta_1}
ight) \ &dots \ \delta_{n-2} c_{n-2} + 2(\delta_{n-2} + \delta_{n-1}) c_{n-1} + \delta_{n-1} c_n &= 3\left(rac{\Delta_{n-1}}{\delta_{n-1}} - rac{\Delta_{n-2}}{\delta_{n-2}}
ight). \end{aligned}$$

property 04a, natural splines  $\rightarrow$  2 equations:

$$S_1''(x_1) = 0 
ightarrow 2c_1 = 0 \ S_{n-1}''(x_n) = 0 
ightarrow 2c_n = 0.$$

n equations, n unknowns in  $c_i$  ,

then solve for  $b_i$ ,  $d_i$  directly from  $c_i$ . which proves the following theorem.

**theorem 07** for a set of data points  $(x_1, y_1), \ldots, (x_n, y_n)$  with distinct  $x_i$ , there is a unique natural cubic spline fitting the points.

#### natural cubic spline

# given 
$$x = [x(1),...,x(n)]$$
 where  $x(1)<...< x(n),y=[y(1),...,y(n)]$ 

```
for i = 1: n-1 a(i) = y(i) \delta(i) = x(i+1) - x(i) \Delta(i) = y(i+1) - y(i) end solve \ for \ c(1), \dots, c(n) for i = 1: n-1 d(i) = (c(i+1)-c(i))/(3*\delta(i)) b(i) = \Delta(i)/\delta(i) - (2c(i)+c(i+1))*\delta(i)/3 end
```

the natural cubic spline is

$$S_i(x) = a_i + b_i(x - x_i) + c_i(x - x_i)^2 + d_i(x - x_i)^3$$
 on  $[x_i, x_{i+1}]$  for  $i = 1, \dots, n-1$ .

✓ example 14

find natural cubic spline through (0,3),(1,-2),(2-1).

for  $c_i$  where n=3,

$$egin{bmatrix} 1 & 0 & 0 \ 1 & 4 & 1 \ 0 & 0 & 1 \end{bmatrix} egin{bmatrix} c_1 \ c_2 \ c_3 \end{bmatrix} = egin{bmatrix} 0 \ 24 \ 0 \end{bmatrix}$$

✓ code

```
1 # bc im even lazier
2
3 import matplotlib.pyplot as plt
4 import numpy as np
5 import numpy.polynomial as npp
6
7 def main():
8  #input = [[0,3],[1,-2],[2,1]]
9  xs = [0,1,2]
10  ys = [3,-2,1]
```

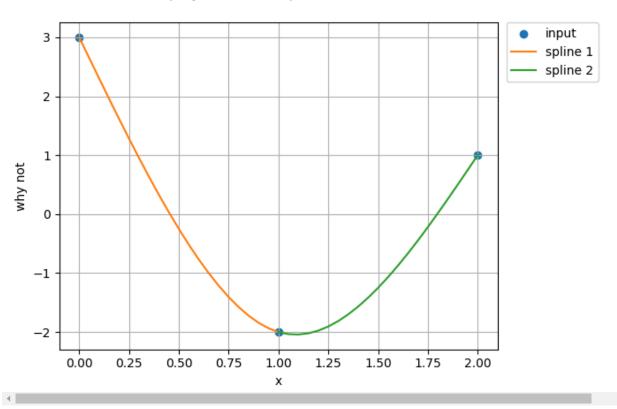
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spline 1 :

$$3.0 - 7.0 \cdot x + 0.0 \cdot x^2 + 2.0 \cdot x^3$$

spline 2 :

### polynomial interpolation error



# 2 endpoint conditions

property 04b curvature-adjusted cubic spline  $S_1^{\prime\prime}(x_1),S_{n-1}^{\prime\prime}(x_n)$  set to arbitrary non-zero values.

property 04c clamped cubic spline  $S_1^\prime(x_1), S_{n-1}^\prime(x_n)$  set to arbitrary non-zero values.

property 04d parabolically terminated cubic spline. first and last points are forced to be at most degree two by specifying  $d_1=0=d_{n-1}$ . equivalently, we can require  $c_1=c_2$ ,  $c_{n-1}=c_n$ , which reduces the matrix equation to a strictly diagonally dominant  $n-2\times n-2$  matrix equation in  $c_2$ ,  $c_{n-1}$ .

property 04e not-a-knot cubic spline  $S_1'''(x_2)=S_2'''(x_2), S_{n-2}'''(x_{n-1})=S_{n-1}'''(x_{n-1})$  set to arbitrary non-zero values. ie, the third derivatives here cause  $S_1=S_2, S_{n-2}=S_{n-1}$ . eg, this means  $x_2$  doesnt need to be a base point bc  $S_2=S_1$ .

#### theorem 08

assume  $n \geq 2$ . then for set of data points  $(x_1, y_1), \ldots, (x_n, y_n)$  and for any one of end conditions given by properties 4a-4c, there is a unique cubic spline satisfying the end conditions and fitting the points. the same is true assuming  $n \geq 3$  for property 4d and  $n \geq 4$  for 4e.

#### ✓ code

```
1 # bc im laziest = hack of previous to include parabolic end conditions
3 import numpy as np
4 import numpy.polynomial as npp
 6 def parabolic_cubic_spline(xs,ys,ni):
 7
8
   xs : x-data
9
    ys : y-data
10
    ni : number of model evaluation points for each spline
1
     import matplotlib.pyplot as plt
2
     import numpy as np
3
     import scipy as sp
4
     #https://docs.scipy.org/doc/scipy/reference/generated/scipy.interpolate.CubicSpline.html
6
7
     def main(): --
35
36
     if __name__ == "__main__":
37
       main()
38
```



