

# Eksam assignment: Implementing Akima spline

Jedrzej Jawor

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## 1 Introduction

This report seeks to implement the Akima sub spline and compare it with the cubic spline.

## 2 Theory

Interpolation is a process where one constructs a smooth function to fit a set of discrete data points. This is often done in order to approximate experimental values at points where no data is available.

A spline,  $S(x)$ , is a piecewise polynomial used for interpolation between two points in a discrete data set:

$$S(x) = S_i(x), x \in [x_i, x_{i+1}], \quad (1)$$

for  $i=[1, n-1]$  being the index of the tabulated points.  $S_i$  is a polynomial of order  $k$ .

Splines for a discrete data set are calculated by demanding the continuity of the splines and their derivatives at the tabulated points [1].

Multiple choices of  $k$  are possible, but their efficiency as interpolation function varies as splines are sometimes prone to wiggling, thus reducing their usefulness as interpolating functions.

One of the more common choices is a cubic spline ( $k=3$ ). Cubic splines (cplines) usually have high precision but they can also be prone to wiggling if the interpolated set has a large discontinuity or outliers.

If higher precision near such point is desired, a sub-spline can be used. Sub splines are variations on the spline method that can be used for different purposes.

One sub spline is the Akima sub-spline (aspline). Asplines are especially useful for data sets with discontinuities as they wiggle much less than csplines. However, they come with drawbacks, sacrificing maximal differentiability by treating one of the coefficient as a free variable used for wiggle reduction instead of demanding continuity of the second derivative.

The method and equations for calculating coefficients for csplines and asplines, on which the code used in this \*project is based on, come from [1].

### 3 Results and Discussion

Using the equations for asplines and csplines from [1], an aspline and cspline for a manually tabulated data set is constructed. The discrete points and the splines are shown in figure 3. It can be seen that the akima spline wiggles much less than the cubic spline.

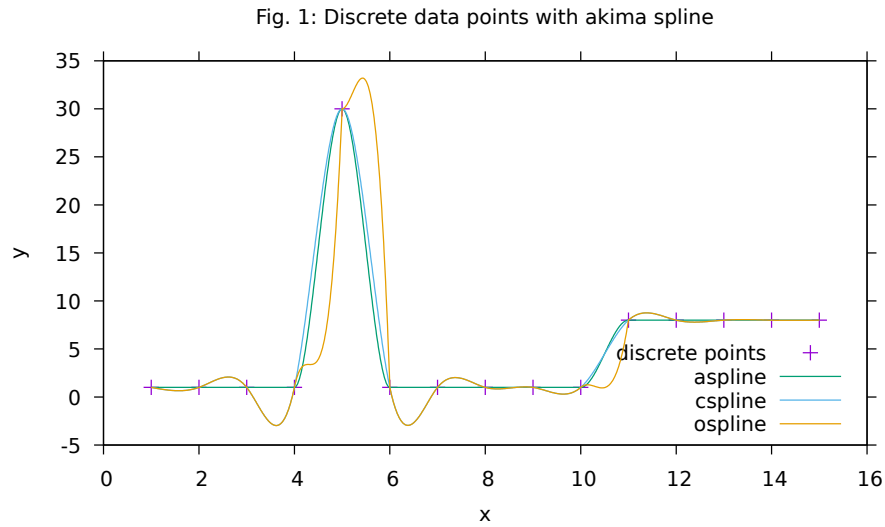


Figure 1: Discrete data points with the akima and cubic splines.

Now the derivatives of the splines are found. For this purpose, a sine function

$$f(x) = \sin(0.5 * x)$$

is tabulated in discrete steps and the a- and c-splines for  $f(x)$  are calculated. The derivative and the second derivative of the a- and csplines are then found. The plots of the two derivatives are shown in figures 3 and ?? together with the analytic derivatives of  $f(x)$ .

The first derivative seems to be alright, with the aspline being only a bit off compared to the cspline. The second derivative of the aspline is not good, consistent with the theory. The cspline on the other hand has a very good second derivative.

Now the integration of the aspline and cspline is tested. Again using the

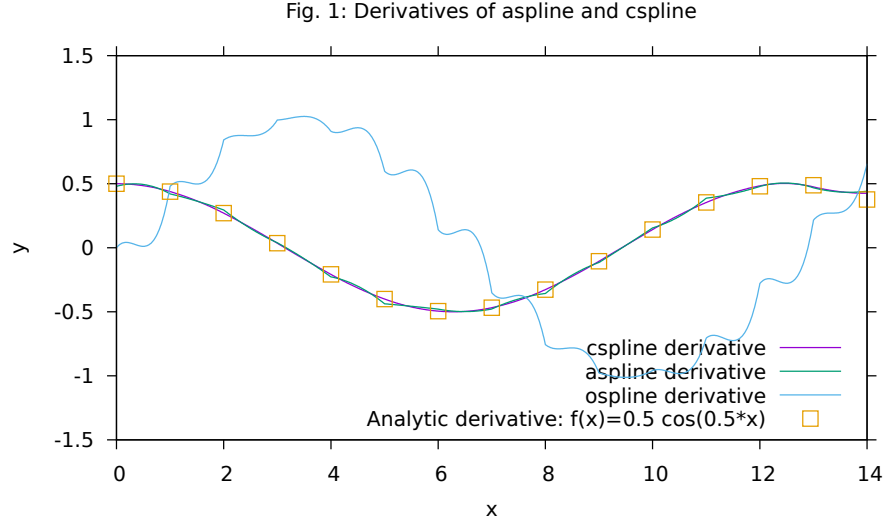


Figure 2: Derivatives of aspline and cspline for  $f(x)$ .

sine function  $f(x)$ . The analytical integral of  $f(x)$  is:

$$\int_0^4 f(x) = 2.83299. \quad (2)$$

The aspline integration yields:

$$\int_0^4 aspline = 2.83194. \quad (3)$$

The cspline integration yields:

$$\int_0^4 cspline = 2.83203. \quad (4)$$

So the integration seems to work quite well for both the cspline and aspline (at least for a nice analytic function like sin).

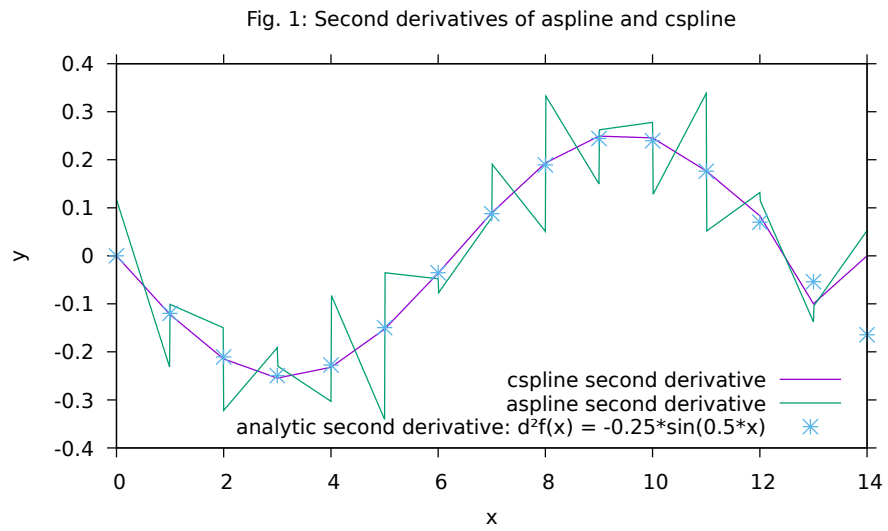


Figure 3: Second derivatives of aspline and cspline for  $f(x)$ .

## References

- [1] Practical programming course book on interpolation.  
*PrakProg: Interpolation.pdf*,  
<http://86.52.112.181/~fedorov/numeric/book/interpolation.pdf>