

## Lagrange Interpolation

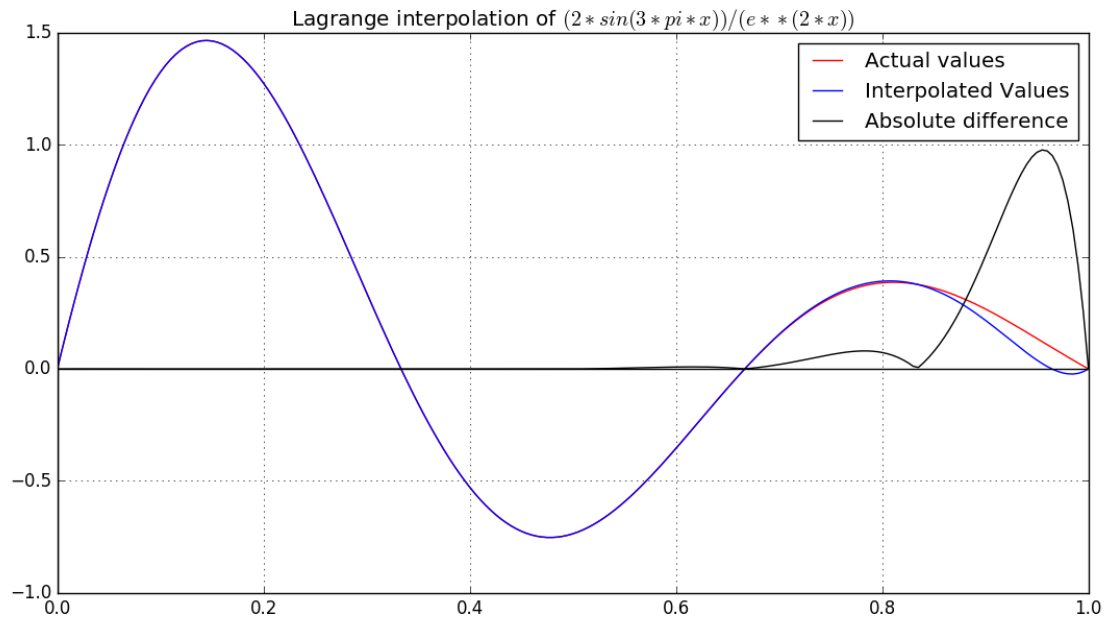


Figure 1. Approximating  $g(x)$  with a Lagrange interpolating polynomial using 11 nodes with different spacing between them. The difference curve, scaled by 10, shows that the error in the approximation increases for successive nodes that are spread farther apart.

The Lagrange interpolant, for the most part, does a good job. We can see that it fails due to its significant error when  $x > 0.8$ . In this portion of the graph,  $g(x)$  remains concave down, but the interpolant changes concavity from concave down to concave up. The difference curve is scaled by a factor of 10, which allows us to see the point at which the error begins. The curve is flat at 0 until  $x$  is close to 0.6 (a hump in the curve can be seen starting from  $x = 0.5$ ). If we look at the differences between the nodes, we can find a clue as to what is going on.

Successive nodes	Difference
$1/12 - 1/24$	.0416
$1/6 - 1/12$	.0833
$1/4 - 1/6$	.0833
$1/3 - 1/4$	.0833
$5/12 - 1/3$	.0833
$1/2 - 5/12$	.0833
$2/3 - 1/2$	.1666
$5/6 - 2/3$	.1666
$1 - 5/6$	.1666

Table 1. Spacing between successive nodes.

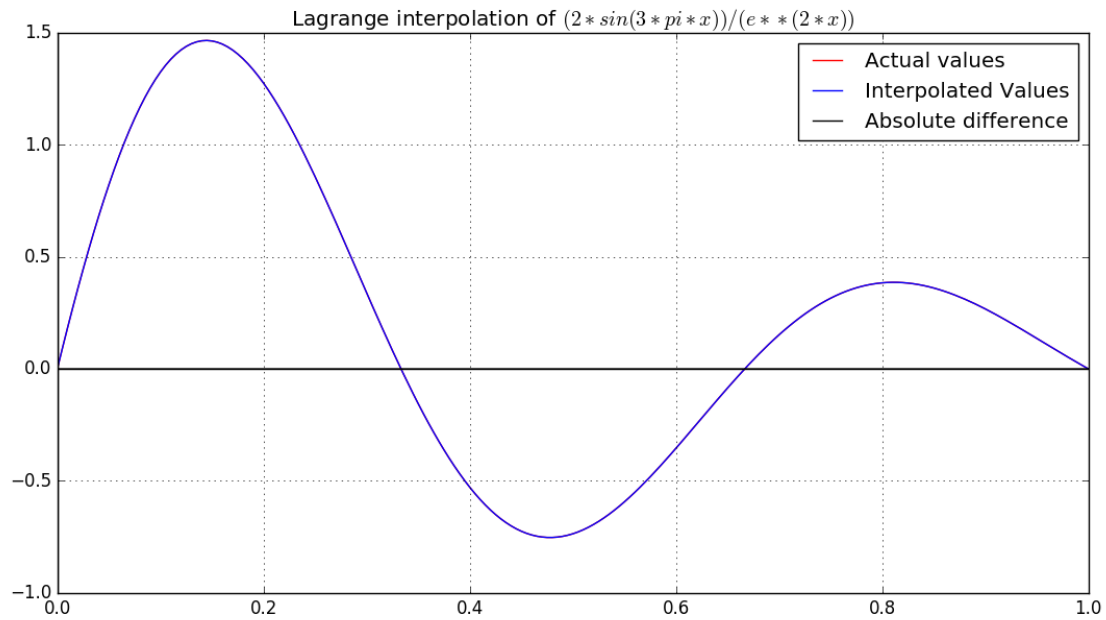


Figure 2. Approximating  $g(x)$  with a Lagrange interpolating polynomial using nodes with equal spacing of 0.0833. The difference curve, scaled by a factor of 10, is not visible. The error that appears in figure 1 has been corrected by using more nodes that are closer together.

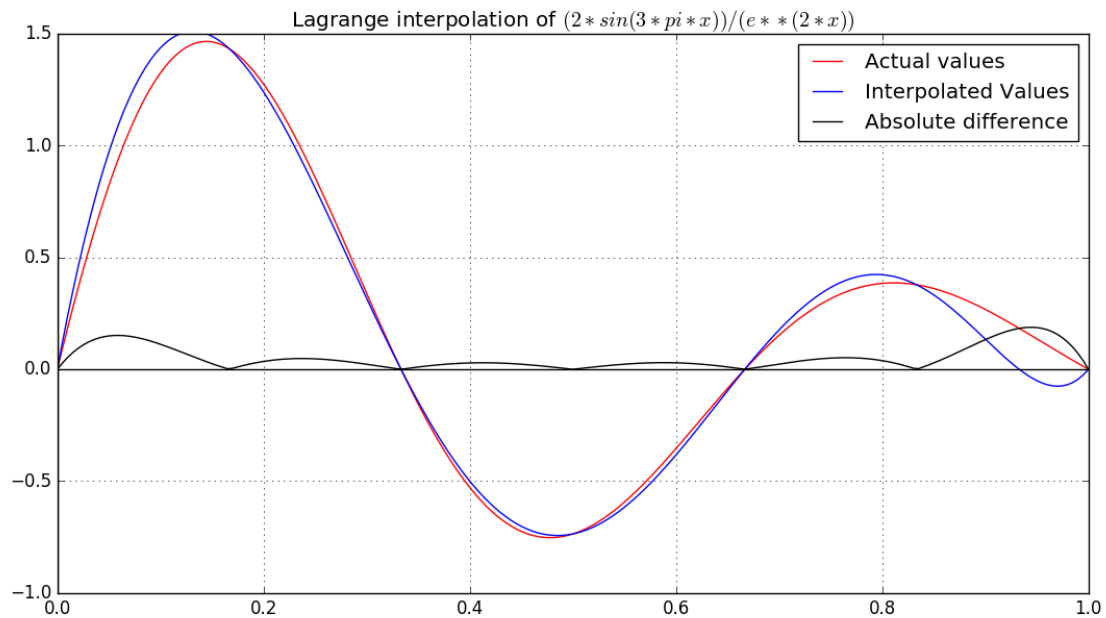


Figure 3. Approximating  $g(x)$  with a Lagrange interpolating polynomial using nodes with equal spacing of 0.1666. This time, the difference curve (which has not been scaled) is not necessary; we can immediately see large differences between the interpolant and the exact curve. Using fewer nodes that are spaced far apart is disastrous for Lagrange interpolation.

The spacing between nodes increases to  $0.1666$  when  $x = 0.5$ , which is the point at which the error curve deviates from  $0$ . From this, we can hypothesize that  $0.1666$  is too large of a distance to separate successive nodes. Figures 2 and 3 agree with this hypothesis. In figure 2, successive nodes are separated by a distance of  $0.0833$ . The Lagrange interpolant approximates  $g(x)$  wonderfully, just as in the first half of figure 1. In figure 3, successive nodes are separated by a distance of  $0.1666$ . We can clearly see that the Lagrange interpolant is not at all a good approximation to  $g(x)$ .  $0.1666$  is simply too much separation to get accurate approximations. This is a great example that shows how sensitive Lagrange interpolation is to the spacing of nodes. Using nodes that are too spaced out will produce abysmal approximations.

## Hermite Interpolation

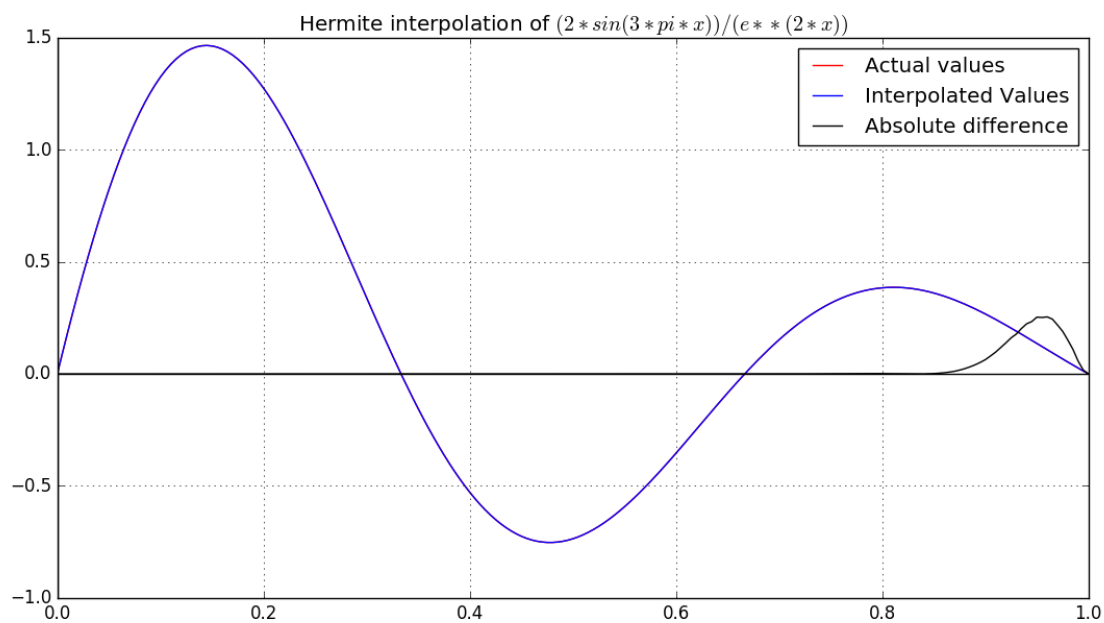


Figure 4. Approximating  $g(x)$  with an Hermite interpolating polynomial using the same nodes as in figure 1. The difference curve, scaled by  $10e7$ , shows that there is almost no error in the approximations.

The Hermite interpolant is extremely good. Out of the four interpolants, this approximates  $g(x)$  the best. We had to scale the difference curve by  $10e7$  to see anything of significance. Like the Lagrange interpolant, the Hermite one has a significant error when  $x > 0.8$ , relative to the rest of the curve. Unlike the Lagrange, this does not seem to be due to spacing between successive nodes. Regardless, Hermite interpolation produces stunningly accurate approximations to  $g(x)$  everywhere on the interval.

## Linear Spline Interpolation

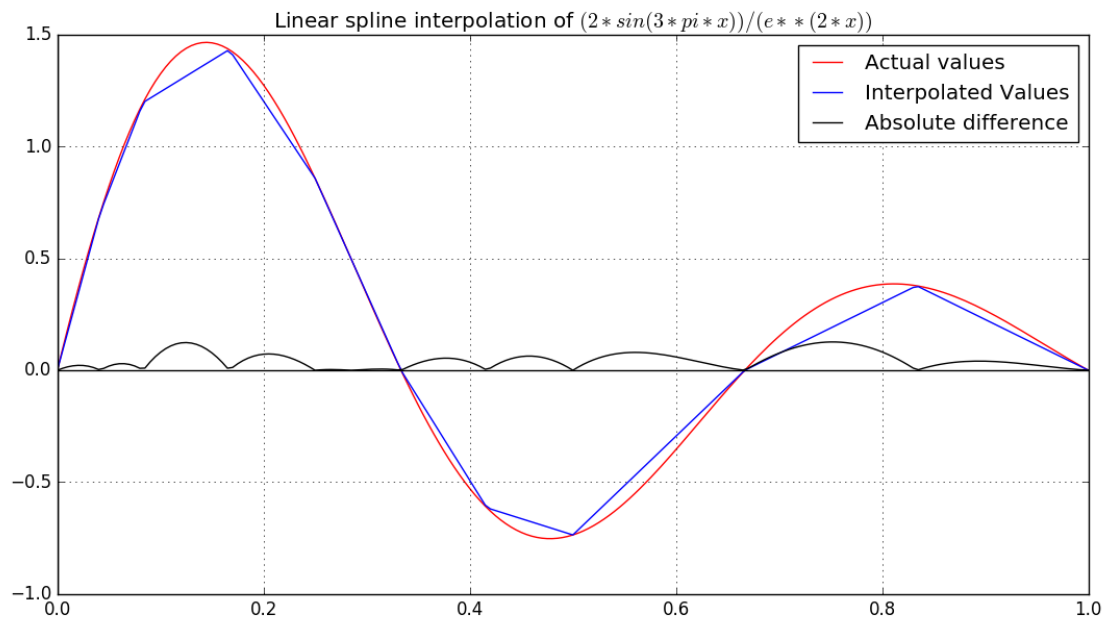


Figure 5. Approximating  $g(x)$  using a linear spline interpolant and the same nodes as in figure 1. Unsurprisingly, the approximations are not good where  $g(x)$  changes slope quickly, but not bad when  $g(x)$ 's slope is close to linear. The difference curve is not scaled.

As expected, the linear spline interpolant was the worst. The difference curve was not scaled. The approximations veer off the true value primarily during intervals where the slope changes quickly.

As figure 5 shows, the difference curve spikes in the intervals that contain local extrema. However, the interpolant does a fine job during the intervals where the slope is close to constant. This shows that linear spline interpolation is not a good choice for curves with slopes that change frequently and rapidly, but it is a good choice for, not surprisingly, linear or close to linear curves.

## Cubic Spline Interpolation

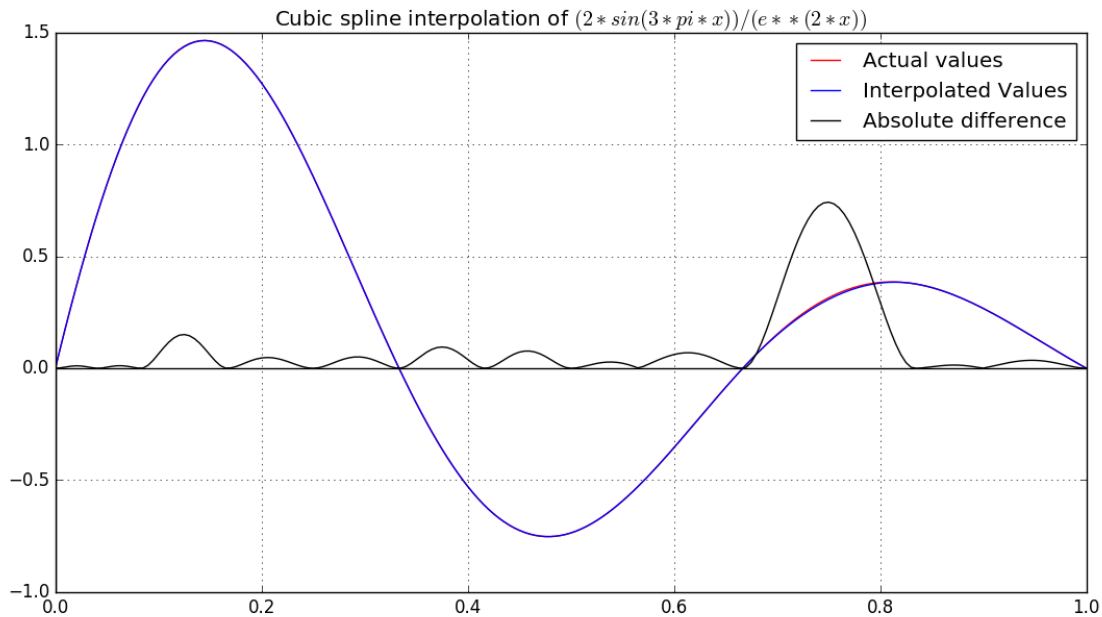


Figure 6. Approximating  $g(x)$  using a cubic spline interpolant and the same nodes as in figure 1. The difference curve has been scaled by 100. It shows some small errors within every interval; however, the errors do not propagate to the other intervals, as was seen in figure 1.

Cubic spline interpolation does a fairly good job at approximating  $g(x)$ . It does a better job than the Lagrange interpolant when  $x > 0.5$ , but is worse when  $x < 0.5$ . The difference curve is scaled by 100, so the errors of the cubic spline interpolant when  $x < 0.5$  are by no means large, however, in the same interval, the Lagrange interpolant's errors are much closer to zero. In the interval from  $x > 0.5$ , the cubic spline interpolant is significantly better than the Lagrange, however, it is significantly worse than the Hermite. Curiously, we can see that the difference curve is zero at regular intervals, except at the large spike when  $x$  is between 0.7 and 0.8. Judging by the curve before 0.7, we would expect it to be zero precisely when it spikes. We would like to attribute this anomaly to the spacing of nodes, but this behavior is not seen in other intervals of equal length. This can actually be seen as a strength of cubic spline interpolation—local errors do not affect global approximations. That is, inaccurate results in one interval do not affect the accuracy in another. This keeps the errors low when  $x > 0.5$ , while in Lagrange interpolation, the errors mostly kept getting worse.

Examining the calculated values and comparing them to the Lagrange Interpolation values shows only minor differences for most of the points where  $x < 0.850$ . It is when we get to values for  $x > 0.850$  that we start to see the difference in  $y$  –values creep up as seen in Table 2.

Whereas differences in the  $y$ -values for values of  $x < 0.855$  remained below  $1E-5$ , by the time we get to  $x > 0.850$  that difference has now grown and reaches a high of  $9.7E-02$ . As described above, these differences between the calculated  $y$ -values and the interpolated values can be attributed to the spacing between nodes.

<u>X Value</u>	<u> Calculated f(x)</u>	<u> Interpolated f(x)</u>	<u> Difference</u>
	0.3542158	0.3435337	
0.855	2	2	0.0106821
	0.3468809	0.3329441	0.0139367
0.86	11	49	62
	0.3388944	0.3213962	0.0174981
0.865	01	47	54
	0.3302875	0.3089224	0.0213650
0.87	39	61	78
	0.3210923	0.2955604	0.0255319
0.875	79	52	28
	0.3113416	0.2813537	0.0299879
0.88	94	06	88
	0.3010688	0.2663522	0.0347166
0.885	82	03	79
	0.2903078	0.2506131	0.0396947
0.89	81	57	24
	0.2790930	0.2342018	0.0448912
0.895	78	3	47
	0.2674592	0.2171924	0.0502667
0.9	19	25	94
	0.2554413	0.1996690	0.0557722
0.905	29	61	68
	0.2430746	0.1817268	0.0613477
0.91	19	35	84
	0.2303944	0.1634729	0.0669214
0.915	03	78	26
	0.2174360	0.1450281	0.0724079
0.92	19	08	11
	0.2042347	0.1265275	0.0777071
0.925	42	91	51
	0.1908257		0.0827027
0.93	06	1.08E-01	04
	0.1772438	0.0899837	0.0872601
0.935	3	11	19
	0.1635237	0.0722985	0.0912251
0.94	35	83	52
0.945	0.1496996	0.0552778	0.0944218

	74	07	67
	0.1358054	0.0391548	0.0966506
0.95	64	64	01
	0.1218744	0.0241886	0.0976857
0.955	11	24	87
	0.1079392		0.0972736
0.96	45	0.0106656	44
		-	
	0.0940320	0.0010976	0.0951297
0.965	59	46	05
		-	
	0.0801842	0.0107519	0.0909361
0.97	48	42	9
		-	
	0.0664264	0.0179127	0.0843392
0.975	45	7	15
		-	
	0.0527884	0.0221573	0.0749458
0.98	72	54	26
		-	
	0.0392992	0.0230215	0.0623208
0.985	86	67	52
		-	
	0.0259869	0.0199966	0.0459835
0.99	26	42	68
		-	
	0.0128784	0.0125256	0.0254041
0.995	73	91	64
1	9.94E-17	9.94E-17	0

Table 2. Calculated values and Lagrange Interpolation values.

Hermite interpolation does a much better job at approximating the y-values than Lagrange interpolation. As we can see by examining the calculated y-values versus the interpolated values in Table 3, the difference between these values is minuscule. For values of  $x < 0.91$ , the difference is as small as  $1E-17$ . It is only once we get to  $x > 0.91$  that the difference climbs to as high as  $1.09E-08$ , which is still a very small difference. The difference for the error can be attributed to the spacing between nodes as previously discussed.

<u>X Value</u>	<u>Calculated f(x)</u>	<u>Interpolated f(x)</u>	<u>Difference</u>
	0.243074	0.243074	9.73E-
0.91	619	628	09
	0.230394	0.230394	1.18E-
0.915	403	415	08
	0.217436	0.217436	1.39E-
0.92	019	033	08
	0.204234	0.204234	1.61E-
0.925	742	758	08
	0.190825	0.190825	1.78E-
0.93	706	724	08
	0.177243	0.177243	2.03E-
0.935	83	85	08
	0.163523	0.163523	2.23E-
0.94	735	757	08
	0.149699	0.149699	2.33E-
0.945	674	698	08
	0.135805	0.135805	2.53E-
0.95	464	49	08
	0.121874	0.121874	2.52E-
0.955	411	436	08
	0.107939	0.107939	2.54E-
0.96	245	27	08
	0.094032	0.094032	2.44E-
0.965	059	084	08
	0.080184	0.080184	2.18E-
0.97	248	27	08
	0.066426	0.066426	1.90E-
0.975	445	464	08
	0.052788	0.052788	1.51E-
0.98	472	487	08
	0.039299	0.039299	1.09E-
0.985	286	297	08
	0.025986	0.025986	5.70E-
0.99	926	932	09
	0.012878	0.012878	1.69E-
0.995	473	474	09
1	9.94E-17	9.94E-17	0

Table 3. Calculated values and Hermite Interpolation values