

HW7

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1 Problem 1

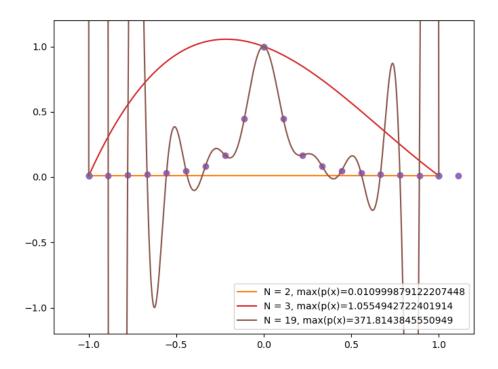
1.a

In order to find the coefficients c of the monomial interpolation, I solved the system $V\mathbf{c} = \mathbf{y}$. To do this, I first created a function that creates the Vandermonde matrix V by filling the i^{th} row with x_i to the power of the j^{th} column. Then, I computed the inverse of this Vandermonde matrix, and multiplied it by \mathbf{y} to find c in the equation $\mathbf{c} = V^{-1}\mathbf{y}$. The resulting vector \mathbf{c} contains the coefficients for the monomial interpolation. See 4 for code conducting these calculations.

1.b

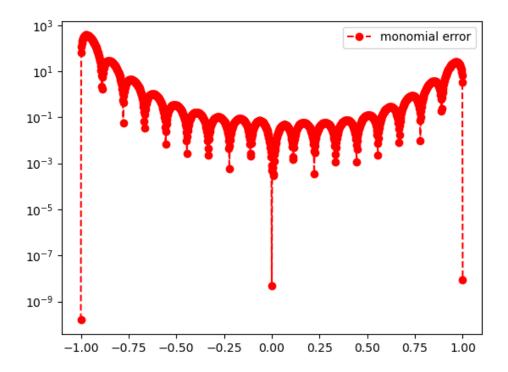
The following plot displays the monomial interpolation of the function $f(x) = \frac{1}{1+(10x)^2}$ for the interpolation points $x_i = -1 + (i-1)\frac{2}{N-1}$. Interpolations are plotted for N = 2, 3, 19. Once N = 19, the maximum value of p(x) > 100. I tested other values of N to discover this fact, but I have removed these values from the plot because it was too cluttered otherwise.





Evidently, as N increases, the polynomial p(x) approximates f(x) more closely for the interpolation points. However, this comes at a tradeoff: the endpoints of polynomials with high N are very oscillatory due to their high degree. Thus, there is large error near the end points. This is exemplified in the plot below of the error for the polynomial estimation using N = 19:

Figure 2: p(x) **Log Error for** N = 19



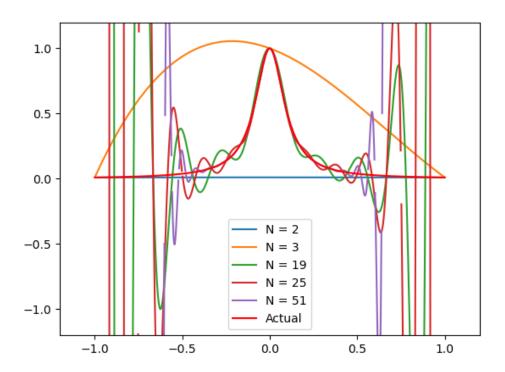
We see that the end points have very large errors, whereas this error decreases for values of x closer to zero. This is the behavior we expected.

2 Problem 2

2.a

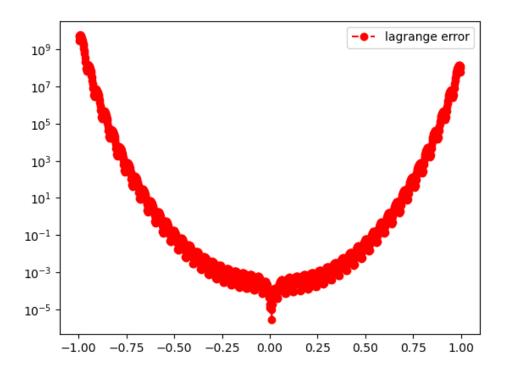
Next we will plot the interpolations of the same function f(x) but using the Barycentric Lagrange method this time. Below are the plots of various p(x) approximations for N = 2, 3, 19, 25, 51.

Figure 3: Barycentric Lagrange Interpolations for N = 2, 3, 19, 25, 51



The polynomials this time appear to be very strong for the points closer to zero, but wildly erroneous toward the endpoints. This is especially true for higher degree polynomials. Take a look at the log error plot for N = 51:

Figure 4: p(x) **Log Error for** N = 51



The endpoints here have massive errors on the scale of 10e9. However, the values of x closer to zero in the center perform much better, with absolute errors of around 10e5. Again, this is the behavior we expected from the Barycentric Lagrange method.

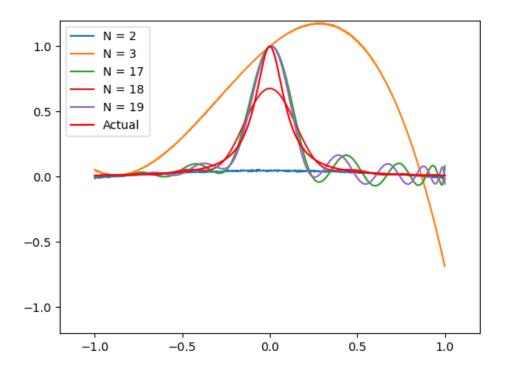
3 Problem 3

Next, we use the interpolation nodes $x_j = cos \frac{(2j-1)\pi}{2N}$. These are known as the Chebyshev points.

The first method of monomial interpolation fails for these points due to the fact that the Vandermond matrix is non-invertible. Therefore, we conclude that the monomial interpolation method is incompatible with the Chebyshev interpolation nodes in this instance.

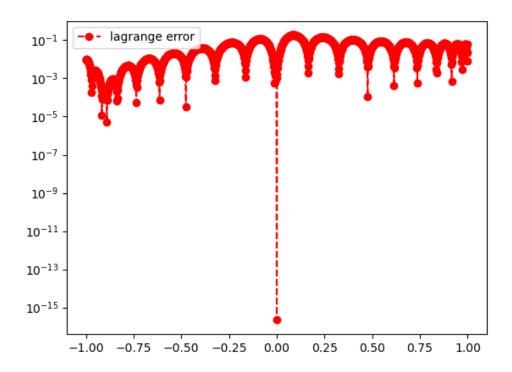
Next, we find that the Barycentric Lagrange method for interpolating the Chebyshev points works. Below is a plot of various p(x) interpolations for N = 2, 3, 17, 18, 19:

Figure 5: Barycentric Lagrange Interpolations for N = 2, 3, 17, 18, 19



Evidently, the behavior near the endpoints is much better now! Although high degree polynomials still oscillate near the endpoints, these are much more controlled, and result in lower errors. Take a look at the log error plot for N = 19, for example:

Figure 6: p(x) **Log Error for** N = 19



A noticible difference here is that the error is more even throughout the interval, as opposed to the "smile" shaped error curves we observed before. This is a known effect of the Chebyshev points. Furthermore, the overall error is lower on average. Therefore, the Chebyshev points do not fail, and in fact appear to work better for Barycentric Lagrange interpolation.

4 Appendix

4.1 Code

```
import numpy as np
2 import numpy.linalg as la
  import matplotlib.pyplot as plt
  from numpy.linalg import inv
  from numpy.linalg import norm
  import math
       eval_monomial(xeval, coef, N, Neval):
      yeval = coef[0]*np.ones(Neval+1)
10
11
       print('yeval = ', yeval)
12
13
      for j in range (1, N+1):
14
         for i in range(Neval+1):
```

```
print('yeval[i] = ', yeval[i])
16
            print('a[j] = ', a[j])
17
            print('i = ', i)
            print('xeval[i] = ', xeval[i])
19
           yeval[i] = yeval[i] + coef[j]*xeval[i]**j
20
21
       return yeval
22
  def Vandermonde(xint, N):
25
27
       V = np.zeros((N+1,N+1))
28
       ''' fill the first column'''
29
       for j in range(N+1):
         V[j][0] = 1.0
32
       for i in range (1, N+1):
33
           for j in range(N+1):
              V[j][i] = xint[j]**i
35
36
       return V
37
  def eval_lagrange_bary(xeval, xint, yint, N, f):
41
       \# lj = np.ones(N+1)
42
43
       # for count in range(N+1):
44
            for jj in range(N+1):
45
                 if (jj != count):
                    lj[count] = lj[count] * (xeval - ...
47
          xint[jj])/(xint[count]-xint[jj])
48
       # yeval = 0
       # for jj in range(N+1):
51
         yeval = yeval + yint[jj]*lj[jj]
       phi = 1
       for xi in xint:
           if xi != xeval:
57
               phi *= (xeval - xi)
58
59
       summation = 0
61
       for j in range(N+1):
62
           wj_denom = 1
65
           for i in range(N+1):
66
                if xint[j] != xint[i]:
67
                    wj_denom *= (xint[j] - xint[i])
           wj = 1/wj_denom
71
           summation += (wj/(xeval-xint[j]))*f(xint[j])
72
```

```
74
       yeval = phi*summation
75
       return (yeval)
77
78
  def question1():
81
       # Monomial
82
83
       f = lambda x: 1 / (1 + ((10*x)**2))
85
       N = 10
86
       a = -1
87
       b = 1
       ''' Create interpolation nodes'''
       xint = np.array([-1 + (j-1)*(2/(N-1)) for j in range(1,N+2)])
       '''Create interpolation data'''
93
       yint = f(xint)
94
       ''' Create the Vandermonde matrix'''
96
       V = Vandermonde(xint, N)
       ''' Invert the Vandermonde matrix'''
100
       Vinv = inv(V)
101
        # ----1.a----
102
104
       c = Vinv@yint
105
        # ----1.b----
106
107
       plt.figure(figsize=(8,6))
108
109
       Ns = [2, 3, 19]
110
111
       for N in Ns:
112
           a = -1
113
            b = 1
114
115
            ''' Create interpolation nodes'''
116
            xint = np.array([-1 + (j-1)*(2/(N-1)) for j in range(1,N+2)])
117
            yint = f(xint)
118
119
           ''' Create the Vandermonde matrix'''
120
            V = Vandermonde(xint, N)
121
122
            ''' Invert the Vandermonde matrix'''
123
            Vinv = inv(V)
124
125
           c = Vinv@yint
126
127
           xeval = np.linspace(-1, 1, 1001)
128
            # yeval = f(xeval)
129
            yeval = eval_monomial(xeval,c,N,1000)
130
```

```
131
            ymax = np.max(yeval)
132
133
134
            plt.plot(xint, yint, 'o')
135
            plt.plot(xeval, yeval, label=f'N = \{N\}, max(p(x) = \{ymax\}')
136
137
138
139
        plt.legend()
140
        plt.xlim(-1.2, 1.2)
141
        plt.ylim(-1.2, 1.2)
        plt.savefig("HW7.1.b.png")
143
144
        fex = f(xeval)
145
        plt.figure()
        err_m = abs(yeval-fex)
147
        plt.semilogy(xeval,err_m,'ro--',label='monomial error')
148
        plt.legend()
149
        plt.savefig("HW7.1.c.png")
150
151
        return
152
153
   question1()
154
155
   def question2():
156
157
158
        # Lagrange Barycentric
159
160
        f = lambda x: 1 / (1 + ((10*x)**2))
161
162
       plt.figure()
163
164
       Ns = [2, 3, 19, 25, 51]
166
        a = -1
167
        b = 1
168
        for N in Ns:
170
171
            ''' Create interpolation nodes'''
172
            xint = np.array([-1 + (j-1)*(2/(N-1)) for j in range(1,N+2)])
173
174
            '''Create interpolation data'''
175
            yint = f(xint)
176
177
178
             ''' create points for evaluating the Lagrange interpolating ...
179
                polynomial'''
            Neval = 1000
180
            xeval = np.linspace(a,b,Neval+1)
181
            yeval_l= np.zeros(Neval+1)
182
184
            ''' evaluate lagrange poly '''
            for kk in range(Neval+1):
185
                 yeval_l[kk] = eval_lagrange_bary(xeval[kk], xint, yint, N, f)
186
187
```

```
188
            plt.plot(xeval, yeval_l, label=f'N = {N}')
189
190
        ''' create vector with exact values'''
        Neval = 1000
192
        xeval = np.linspace(a,b,Neval+1)
193
        fex = f(xeval)
194
195
        plt.plot(xeval, fex, 'r-', label='Actual')
196
197
        plt.legend()
198
        plt.xlim(-1.2, 1.2)
        plt.ylim(-1.2, 1.2)
200
        plt.savefig("HW7.2.a.png")
201
202
        plt.figure()
203
        err_l = abs(yeval_l-fex)
204
        plt.semilogy(xeval,err_l,'ro--',label='lagrange error')
205
       plt.legend()
206
        plt.savefig("HW7.2.b.png")
207
208
        return
209
210
   question2()
211
212
   def question3a():
213
214
215
        # Monomial
216
        f = lambda x: 1 / (1 + ((10*x)**2))
217
218
219
220
       plt.figure(figsize=(8,6))
221
222
       Ns = [2, 3, 19]
223
224
        a = -1
225
226
       b = 1
227
        for N in Ns:
228
229
            ''' Create interpolation nodes'''
230
            xint = np.array([math.cos(((2*j -1)*math.pi)/(2*N))) for j in ...
231
                range(1,N+2)])
            yint = f(xint)
232
233
            ''' Create the Vandermonde matrix'''
234
            V = Vandermonde(xint, N)
235
            ''' Invert the Vandermonde matrix'''
237
            Vinv = inv(V)
238
239
            c = Vinv@yint
240
241
242
            xeval = np.linspace(-1, 1, 1001)
            yeval = f(xeval)
243
            yeval = eval_monomial(xeval,c,N,1000)
244
```

```
245
            fex = f(xeval)
246
            ymax = np.max(yeval)
247
248
249
            plt.plot(xint, yint, 'o')
250
            plt.plot(xeval, yeval, label=f'N = \{N\}, max(p(x) = \{ymax\}')
251
252
253
254
        plt.legend()
255
256
        plt.xlim(-1.2, 1.2)
        plt.ylim(-1.2, 1.2)
257
        plt.savefig("HW7.3.a.i.png")
258
259
        plt.figure()
260
        err = abs(yeval-fex)
261
        plt.semilogy(xeval,err,'ro--',label='monomial error')
262
       plt.legend()
263
        plt.savefig("HW7.3.a.ii.png")
264
265
   \# 3a fails due to singular matrix being non-invertivle
266
267
   def question3b():
268
269
        # Lagrange Barycentric
270
271
        f = lambda x: 1 / (1 + ((10*x)**2))
272
273
       plt.figure()
274
275
276
       Ns = [2, 3, 17, 18, 19]
277
        a = -1
278
        b = 1
279
280
        for N in Ns:
281
282
            ''' Create interpolation nodes'''
283
            xint = np.array([math.cos(((2*j -1)*math.pi)/(2*N))) for j in ...
284
                range(1,N+2)])
285
            '''Create interpolation data'''
286
            yint = f(xint)
287
288
289
            ''' create points for evaluating the Lagrange interpolating ...
290
                polynomial'''
            Neval = 1000
291
            xeval = np.linspace(a,b,Neval+1)
            yeval_l= np.zeros(Neval+1)
293
294
            ''' evaluate lagrange poly '''
295
296
            for kk in range(Neval+1):
297
                 yeval_l[kk] = eval_lagrange_bary(xeval[kk], xint, yint, N, f)
298
299
            plt.plot(xeval, yeval_l, label=f'N = {N}')
300
```

```
301
       ''' create vector with exact values'''
302
       Neval = 1000
303
       xeval = np.linspace(a,b,Neval+1)
       fex = f(xeval)
306
       plt.plot(xeval, fex, 'r-', label='Actual')
307
308
       plt.legend()
309
       plt.xlim(-1.2,1.2)
310
       plt.ylim(-1.2,1.2)
311
       plt.savefig("HW7.3.b.i.png")
312
313
314
       plt.figure()
       err = abs(yeval_l-fex)
315
       plt.semilogy(xeval,err,'ro--',label='lagrange error')
316
       plt.legend()
317
       plt.savefig("HW7.3.b.ii.png")
318
319
320 question3b()
```

Note: All code for this assignment is original, written by Jesse Hettleman