

# Scientific Computing assignment 6

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**Problem 1)** The following is the code for the python module **DirectDFT.py** with functions **DFT** for computing DFT of a function  $f$  and **iDFT** for computing the inverse DFT of a function  $F$ :

```
#!/usr/bin/env python3
# -*- coding: utf-8 -*-
"""
Created on Fri Dec 14 03:24:57 2018

@author: alfred_mac
"""

import numpy as np
import matplotlib.pyplot as plt
import numpy.linalg as la

def DFT(f):
    N = len(f)
    F = (0+0j)*np.arange(N)
    for i in range(N):
        for j in range(N):
            F[i] += (1/N)*f[j]*np.exp(-2*np.pi*1j*i*j/N)

    return F

def iDFT(F):
    N = len(F)
    f = (0+0j)*np.arange(N)
    for i in range(N):
        for j in range(N):
            f[i] += F[j]*np.exp(2*np.pi*1j*i*j/N)

    return f
```

**Problem 1.a)** The following is the code for comparing the accuracy of the SFTW (code mentioned above) module with the inbuilt Python module for

FFT:

```
#!/usr/bin/env python3
# -*- coding: utf-8 -*-
"""
Created on Fri Dec 14 03:57:33 2018

@author: alfred_mac
"""

import numpy as np
import matplotlib.pyplot as plt
import numpy.linalg as la
import numpy.fft as fft
import DirectDFT as SFTW

#fi = [0,0,1,0,0,0,0]
#fi = [0,0,0,0,0,1,0]
fi = [0,0,0,0,1,0,0,0,0,0]
fo = SFTW.DFT( fi )

f = open("DFT_output_prob1a.txt","a+")
f.write("Input f is:\n")
np.savetxt( f , fi ,fmt='%.2f ' )

fo_act = (1/N)*np.array( fft.fft( fi ) )

f.write("Euclidean norm of the difference between the
estimated result and the actual result as a measure of
accuracy is: ")
f.write( str( sum( abs( fo_act-fo ) ) ) )
f.write("\n")
f.close()
```

The following shows the test results when the above code was tested with different examples. The accuracy of SFTW is mentioned along with it.

Input f is: 0.00 0.00 1.00 0.00 0.00 0.00 0.00 Euclidean norm of the difference between the estimated result and the actual result as a measure of accuracy is: 1.5435090268566047e-15 Input f is: 0.00 0.00 0.00 0.00 0.00 1.00 0.00 Euclidean norm of the difference between the estimated result and the actual result as a measure of accuracy is: 9.066215190086561e-15 Input f is: 0.00 0.00 0.00 1.00 0.00 0.00 0.00 0.00 Euclidean norm of the difference between the estimated result and the actual result as a measure of accuracy is: 4.7146948623413195e-15 Input f is: 0.00 0.00 0.00 0.00 1.00 0.00 0.00 0.00 0.00 0.00 Euclidean norm of the difference between the estimated result and the actual result as a measure of accuracy is: 9.0000000000000002 Input f is: 0.00

0.00 0.00 0.00 1.00 0.00 0.00 0.00 0.00 0.00 Euclidean norm of the difference between the estimated result and the actual result as a measure of accuracy is: 4.734507369567292e-16

**Problem 1.b)** The mathematical derivation for finding the value of  $C$  to equal to  $n$  is given in the handwritten page. Here, the following code verifies that for a simple test case:

```
#!/usr/bin/env python3
# -*- coding: utf-8 -*-
"""
Created on Fri Dec 14 05:34:15 2018

@author: alfred_mac
"""

import numpy as np
import matplotlib.pyplot as plt
import numpy.linalg as la
import numpy.fft as fft
import DirectDFT as SFTW

fi = np.array([0,0,0,0,1,0,0,0,0,0])
fo = SFTW.DFT(fi)

C = sum(fi*np.conj(fi))/sum(fo*np.conj(fo))

print("The value of C for an array of 10 elements is: ",C)
```

The following is the output for the above mentioned code:

**The value of C for an array of 10 elements is: (9.999999999999996+0j)**

**Problem 1.c)** The mathematical derivation for determining the relation between  $\hat{g}_k$  and  $\hat{f}_k$  is given in the handwritten page. Here, the following code verifies that for a simple test case:

```
#!/usr/bin/env python3
# -*- coding: utf-8 -*-
"""
Created on Fri Dec 14 06:05:48 2018

@author: alfred_mac
"""
```

```
import numpy as np
import matplotlib.pyplot as plt
import numpy.linalg as la
```

```

import numpy.fft as fft
import DirectDFT as SFTW

N=10
fi = np.random.random([N,1])
gi = np.zeros([N,1])

gi[1:] = fi[0:-1]
gi[0] = fi[-1]

fo = SFTW.DFT(fi)
go = SFTW.DFT(gi)

go_est = fo*0

for i in range(N):
    go_est[i] = fo[i]*np.exp(-2*np.pi*1j*i/N)

print("go is: ",go)
print("go_est is: ",go_est)
diff = sum(abs(go-go_est))
print("Absolute difference between the estimated and
actual values of DFT of g is:",diff)

```

The following is the output for the above mentioned code:

```

go is: [0.61932566 + 0.00000000e + 00j - 0.03478393 - 1.28427372e - 03j -
0.07382498 - 5.50837349e - 02j - 0.09483726 - 6.86264145e - 02j - 0.10475228 -
3.88661138e - 02j 0.07684922 + 7.33402803e - 17j - 0.10475228 + 3.88661138e -
02j - 0.09483726 + 6.86264145e - 02j - 0.07382498 + 5.50837349e - 02j -
0.03478393 + 1.28427372e - 03j]
go-est is: [0.61932566 + 0.00000000e + 00j - 0.03478393 - 1.28427372e -
03j - 0.07382498 - 5.50837349e - 02j - 0.09483726 - 6.86264145e - 02j -
0.10475228 - 3.88661138e - 02j 0.07684922 + 8.31102647e - 17j - 0.10475228 +
3.88661138e - 02j - 0.09483726 + 6.86264145e - 02j - 0.07382498 + 5.50837349e -
02j - 0.03478393 + 1.28427372e - 03j] Absolute difference between the estimated
and actual values of DFT of g is: 1.0796099979596523e - 15

```

**Problem 1.d)** The following code verifies both the **DFT** and the **iDFT** function in the **DirectDFT.py** module for a simple test case:

```

#!/usr/bin/env python3
# -*- coding: utf-8 -*-
"""

```

Created on Fri Dec 14 06:49:54 2018

@author: alfred\_mac

```
"""
```

```
import numpy as np
import matplotlib.pyplot as plt
import numpy.linalg as la
import numpy.fft as fft
import DirectDFT as SFTW
```

```
N=10
```

```
fi = (0+0j)*np.arange(N)
fi += np.random.random()
```

```
fo = SFTW.DFT(fi)
```

```
fi1 = SFTW.iDFT(fo)
```

```
print("Input f is: ",fi)
print("iDFT of DFT of f is: ",fi1)
diff = sum((fi-fi1)*np.conj(fi-fi1))
print("Euclidean norm of difference between f and iDFT of
DFT of f is:",diff)
```

The following is the output for the above mentioned code:

```
Input f is: [0.48158163+0.j 0.48158163+0.j 0.48158163+0.j 0.48158163+0.j
0.48158163+0.j 0.48158163+0.j 0.48158163+0.j 0.48158163+0.j 0.48158163+0.j
0.48158163+0.j] iDFT of DFT of f is: [0.48158163+7.80644855e-17j 0.48158163+1.94551855e-
16j 0.48158163-3.16027451e-16j 0.48158163-6.94075799e-17j 0.48158163+1.65920312e-
16j 0.48158163-3.48681348e-16j 0.48158163-2.59278682e-16j 0.48158163+3.44621896e-
16j 0.48158163+1.83879486e-16j 0.48158163+2.63570262e-17j] Euclidean norm
of difference between f and iDFT of DFT of f is: (1.1345371090849671e-30+0j)
```

**Problem 1.e)** The following code verifies how applying DFT four times on a function  $f$  gives back the same function again for a simple test case:

```
#!/usr/bin/env python3
# -*- coding: utf-8 -*-
"""
```

```
Created on Fri Dec 14 07:13:41 2018
```

```
@author: alfred_mac
"""
```

```
import numpy as np
import matplotlib.pyplot as plt
import numpy.linalg as la
import numpy.fft as fft
```

```

import DirectDFT as SFTW

N=10
f0 = (0+0j)*np.arange(N)
f0 += np.random.random()

f1 = SFTW.DFT(f0)
f2 = SFTW.DFT(f1)
f3 = SFTW.DFT(f2)
f4 = SFTW.DFT(f3)

print("Input f is: ",f0)
print("Applying DFT four times on f gives: ",f4)
diff = sum((f0-100*f4)*np.conj(f0-100*f4))
print("Euclidean norm of difference between f and the
result obtained after applying DFT 4 times is:",diff)

```

The following is the output for the above mentioned code:

```

Input f is: [0.44742176+0.j 0.44742176+0.j 0.44742176+0.j 0.44742176+0.j
0.44742176+0.j 0.44742176+0.j 0.44742176+0.j 0.44742176+0.j 0.44742176+0.j
0.44742176+0.j] Applying DFT four times on f gives: [0.00447422+2.35772819e-
18j 0.00447422+2.54352449e-18j 0.00447422-1.49405621e-18j 0.00447422+2.17642095e-
18j 0.00447422-1.24281656e-18j 0.00447422-6.26085601e-18j 0.00447422-1.45134260e-
18j 0.00447422+1.69351020e-18j 0.00447422-9.09180979e-19j 0.00447422+2.58706853e-
18j] Euclidean norm of difference between f and the result obtained after apply-
ing DFT 4 times is: (2.2846566502789448e-30+0j)

```

**Problem 2)** The following code computes DFT for the given two functions  $g(x)$  and  $h(x)$ :

```

#!/usr/bin/env python3
# -*- coding: utf-8 -*-
"""
Created on Sat Dec 15 02:44:07 2018

@author: alfred_mac
"""

```

```

import numpy as np
import matplotlib.pyplot as plt
import numpy.linalg as la
import numpy.fft as fft

N = 10000
freqs = fft.fftfreq(N, 1.0/6.0)
freqs = fft.fftshift(freqs)

```

```

g = (1/np.sqrt(2*np.pi))*np.exp(-0.5*freqs*freqs)
h = 0.5*np.exp(-abs(freqs))

gk = fft.fft(g)
hk = fft.fft(h)

gk = fft.fftshift(gk)
hk = fft.fftshift(hk)
k = np.arange(N)-N/2
#plt.plot(freqs,g)
#plt.plot(freqs,h)

n=20
plt.plot(k[5000-n:5000+n+1],np.real(gk[5000-n:5000+n+1]))
plt.plot(k[5000-n:5000+n+1],np.real(hk[5000-n:5000+n+1]),'--')

#plt.title('$g(x)=(2\pi)^{-0.5}e^{-0.5x^2}$ and $h(x)=0.5e^{-|x|}$')
plt.title('$G(k)=DFT[g(x)]$ and $H(k)=DFT[h(x)]$')
#plt.legend(['$g(x)$','$h(x)$'],loc='upper right')
plt.legend(['$G(k)$','$H(k)$'],loc='upper right')
#plt.xlabel('$x$')
plt.xlabel('$k$')
#plt.ylabel('$g(x)$ or $h(x)$')
plt.ylabel('$G(k)$ or $H(k)$')
plt.show()

```

The following figures show the functions  $g(x)$  and  $h(x)$  in one plot and its corresponding DFTs -  $\hat{g}_k$  and  $\hat{h}_k$  in another plot.

From the second plot, its clear that the DFT coefficients of  $g(x)$ , i.e.  $\hat{g}_k$  converge faster.

**Problem 3)** The following code interpolates both  $g(x)$  and  $h(x)$  using a trigonometric polynomial DFT for the given two functions  $g(x)$  and  $h(x)$ :

```

#!/usr/bin/env python3
# -*- coding: utf-8 -*-
"""
Created on Sat Dec 15 05:03:40 2018

@author: alfred_mac
"""

```

```

import numpy as np
import matplotlib.pyplot as plt
import numpy.linalg as la
import numpy.fft as fft

```

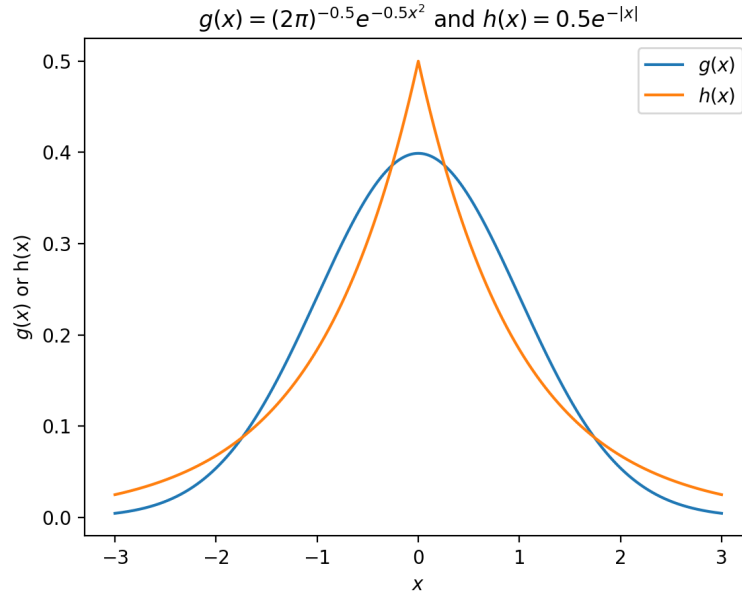


Figure 1: Functions  $g(x)$  and  $h(x)$

$L = 6.0$

```
def interpolate(A, freqs, M):
    n = len(freqs)
    p = np.zeros([M, 1], dtype='complex')
    x = np.arange(M)*L/M - L/2
    for i in range(M):
        p[i] = 0
    # for j in range(n):
    #     p[i] += A[j]*np.exp(2*np.pi*1j*j*x[i]/L)
    for j in range(n):
        if(j<n/2):
            p[i] += A[j]*np.exp(2*np.pi*1j*j*x[i]/L)
        elif(j==n/2):
            p[i] += A[j]*(np.exp(2*np.pi*1j*(n/2)*x[i]/L)+
                np.exp(-2*np.pi*1j*(n/2)*x[i]/L))/2
        else:
            p[i] += A[j]*np.exp(2*np.pi*1j*(j-n)*x[i]/L)

    p = fft.fftshift(p)
```



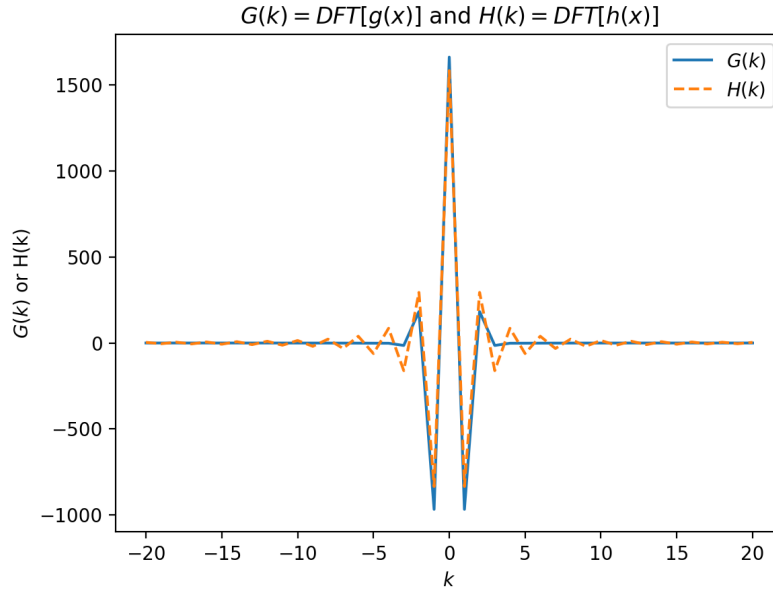


Figure 2: Functions  $\hat{g}_k$  and  $\hat{h}_k$

```

return (p,x)

ng = 1000
nh = 1000
M = 1000
freqsg = fft.fftfreq(ng, 1.0/L)
freqsg = fft.fftshift(freqsg)

freqsh = fft.fftfreq(nh, 1.0/L)
freqsh = fft.fftshift(freqsh)

g = (1/np.sqrt(2*np.pi))*np.exp(-0.5*freqsg*freqsg)
h = 0.5*np.exp(-abs(freqsh))

gk = fft.fft(g)/ng
hk = fft.fft(h)/nh

(Pg,Xg) = interpolate(gk,freqsg,M)
(Ph,Xh) = interpolate(hk,freqsh,M)

#plt.plot(freqsg,g)
#plt.plot(Xg,np.real(Pg),'--')

```

```

plt.plot(freqsh, h)
plt.plot(Xh, np.real(Ph), '--')

#plt.title('$g(x)$ and its trigonometric interpolation $p_{\{g\}}(x)$')
plt.title('$h(x)$ and its trigonometric interpolation $p_{\{h\}}(x)$')
#plt.legend(['$g(x)$', '$p_{\{g\}}(x)$'])
plt.legend(['$h(x)$', '$p_{\{h\}}(x)$'])
plt.xlabel('$x$')
#plt.ylabel('$g(x)$ or $p_{\{g\}}(x)$')
plt.ylabel('$h(x)$ or $p_{\{h\}}(x)$')
plt.show()

```

The following figures show the functions  $g(x)$  and  $h(x)$  in one plot and its corresponding trigonometric polynomial interpolation for  $M = 1000$  and  $n = 1000$  for both  $g(x)$  and  $h(x)$ . Such a low value seems to be enough for interpolation for both the functions.

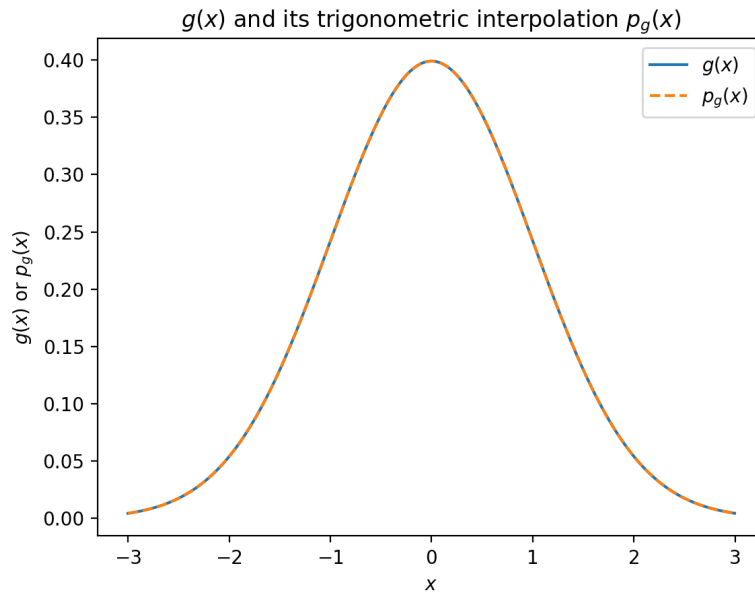


Figure 3: Function  $g(x)$  and its trigonometric polynomial interpolation  $p_g(x)$

**Problem 4.a)** The following code verifies the spectral derivative obtained from the interpolated trigonometric polynomial with the actual derivative of the function for the given two functions  $g(x)$  and  $h(x)$ :

```
#!/usr/bin/env python3
```

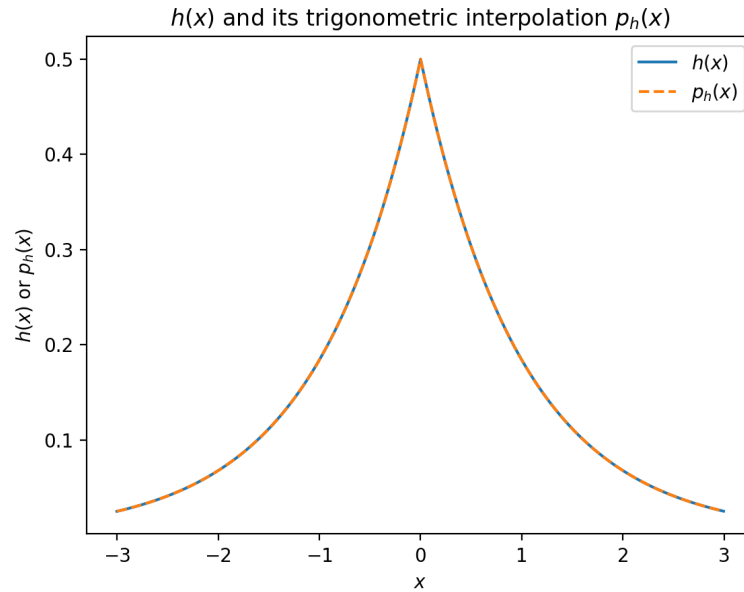


Figure 4: Function  $h(x)$  and its trigonometric polynomial interpolation  $p_h(x)$

```
# -*- coding: utf-8 -*-
"""
Created on Sat Dec 15 07:10:26 2018

@author: alfred_mac
"""

import numpy as np
import matplotlib.pyplot as plt
import numpy.linalg as la
import numpy.fft as fft

L = 6.0

def interpolate(A, freqs, M):
    n = len(freqs)
    p = (0+0j)*np.arange(M) #np.zeros([M,1], dtype='complex')
    x = np.arange(M)*L/M - L/2
    for i in range(M):
        p[i] = 0
    # for j in range(n):
    #     p[i] += (2*np.pi*1j*(j-n/2)/L)*A[j]*np.exp(2*np.pi*1j*j*x[i]/L)
```

```

    for j in range(n):
        if (j<n/2):
            p[i] += (2*np.pi*j/L)*A[j]*
                np.exp(2*np.pi*j*x[i]/L)
        elif (j==n/2):
            p[i] += 0*A[j]*(np.exp(2*np.pi*j*(n/2)*x[i]/L)+
                np.exp(-2*np.pi*j*(n/2)*x[i]/L))/2
        else:
            p[i] += (2*np.pi*j*(j-n)/L)*A[j]*
                np.exp(2*np.pi*j*(j-n)*x[i]/L)

    p = fft.fftshift(p)
    return (p,x)

ng = 10000
nh = 10000
M = 1000
freqsg = fft.fftfreq(ng, 1.0/L)
freqsg = fft.fftshift(freqsg)

freqsh = fft.fftfreq(nh, 1.0/L)
freqsh = fft.fftshift(freqsh)

g = (1/np.sqrt(2*np.pi))*np.exp(-0.5*freqsg*freqsg)
h = 0.5*np.exp(-abs(freqsh))

gk = fft.fft(g)/ng
hk = fft.fft(h)/nh

dgdx = -(freqsg/np.sqrt(2*np.pi))*np.exp(-0.5*freqsg*freqsg)
dhdx = -(freqsh/abs(freqsh))*0.5*np.exp(-abs(freqsh))

#gk = fft.fft(dgdx)/ng
#hk = fft.fft(dhdx)/nh

#(Pg,Xg) = interpolate(gk,freqsg,M)
(Ph,Xh) = interpolate(hk,freqsh,M)

#plt.plot(freqsg,dgdx)
#plt.plot(Xg,np.real(Pg),'--')

#plt.plot(freqsh,dhdx)
#plt.plot(Xh,np.real(Ph),'--')

```

```

# Plotting error between the sampled derivative (f(xk)) and
# spectral derivative (p(xk))
#plt.plot(Xg[10:990],dgdX[100:(ng-100):int(ng/M)]-
#np.real(Pg[10:990]),'--') # Values at the edges are not
#shown for visual reasons
plt.plot(Xh,dhdX[0:nh:int(nh/M)]-np.real(Ph),'--') # Values
#at the edges are not shown for visual reasons

#plt.title('$g\'(x)$ and its trigonometric interpolation $p_{\{g\}}\'(x)$')
#plt.title('$h\'(x)$ and its trigonometric interpolation $p_{\{h\}}\'(x)$')
#plt.title('Error between $g\'(x)$ and its spectral derivative $p_{\{g\}}\'(x)$')
plt.title('Error between $h\'(x)$ and its spectral derivative $p_{\{h\}}\'(x)$')
#plt.legend(['$g\'(x)$', '$p_{\{g\}}\'(x)$'])
#plt.legend(['$h\'(x)$', '$p_{\{h\}}\'(x)$'])
plt.xlabel('$x$')
#plt.ylabel('$g\'(x)$ or $p_{\{g\}}\'(x)$')
#plt.ylabel('$h\'(x)$ or $p_{\{h\}}\'(x)$')
#plt.ylabel('$g\'(x) - p_{\{g\}}\'(x)$')
plt.ylabel('$h\'(x) - p_{\{h\}}\'(x)$')
plt.show()

```

The following first two figures show the spectral derivative obtained from the interpolated trigonometric polynomial with the actual derivative of the given two functions  $g(x)$  and  $h(x)$ . Also, the next two figures show the error between the actual function and its interpolation for both the given functions:

Here we notice that, unlike for the function  $g'(x)$ , the error between  $h'(x)$  and it's interpolation is too high even in the center region.

**Problem 4.b)** The following code computes the sequence of accuracy measures for different values of  $\Delta x$  for both the functions  $g'(x)$  and  $h'(x)$ :

```

#!/usr/bin/env python3
# -*- coding: utf-8 -*-
"""
Created on Wed Dec 19 09:02:04 2018

@author: alfred_mac
"""

import numpy as np
import matplotlib.pyplot as plt
import numpy.linalg as la
import numpy.fft as fft

L = 6.0

def interpolate(A,freqs,M):

```

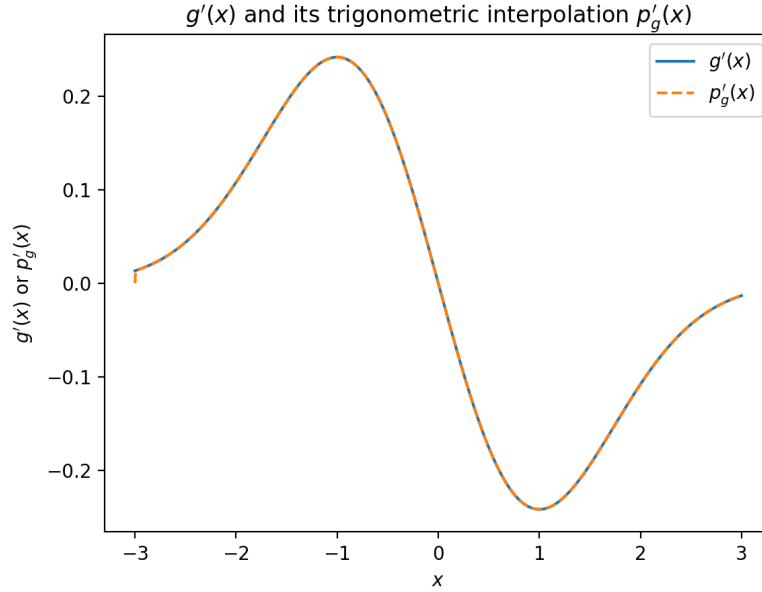


Figure 5: Function  $g'(x)$  and its trigonometric polynomial interpolation  $p_{g'}(x)$

```

n = len(freqs)
p = (0+0j)*np.arange(M) #np.zeros([M,1], dtype='complex')
x = np.arange(M)*L/M - L/2
for i in range(M):
    p[i] = 0
    for j in range(n):
        if(j<n/2):
            p[i] += (2*np.pi*j/L)*A[j]*
                np.exp(2*np.pi*j*x[i]/L)
        elif(j==n/2):
            p[i] += 0*A[j]*(np.exp(2*np.pi*j*(n/2)*x[i]/L)+
                np.exp(-2*np.pi*j*(n/2)*x[i]/L))/2
        else:
            p[i] += (2*np.pi*j*(j-n)/L)*A[j]*
                np.exp(2*np.pi*j*(j-n)*x[i]/L)

p = fft.fftshift(p)
return (p,x)

ng = 10000
nh = 10000
freqsg = fft.fftfreq(ng, 1.0/L)

```

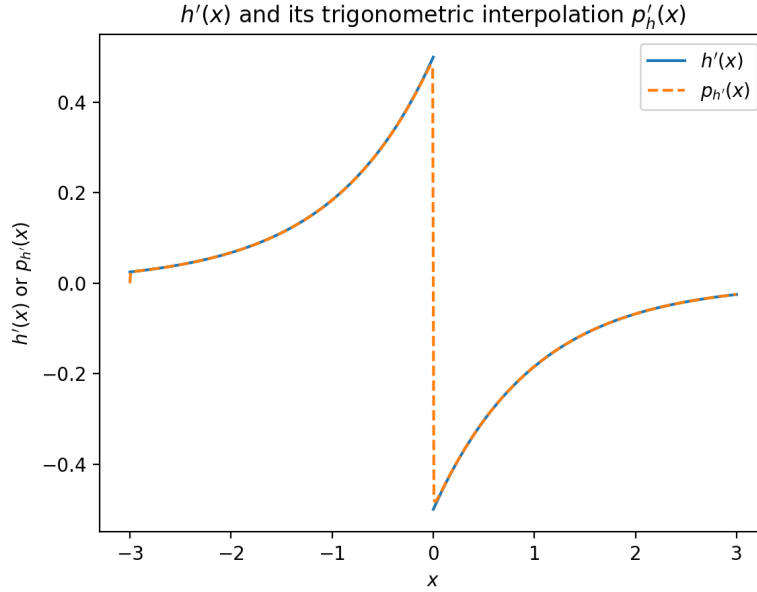


Figure 6: Function  $h'(x)$  and its trigonometric polynomial interpolation  $p_{h'}(x)$

```

freqsg = fft.fftshift(freqsg)

freqsh = fft.fftfreq(nh, 1.0/L)
freqsh = fft.fftshift(freqsh)

g = (1/np.sqrt(2*np.pi))*np.exp(-0.5*freqsg*freqsg)
h = 0.5*np.exp(-abs(freqsh))

gk = fft.fft(g)/ng
hk = fft.fft(h)/nh

dgdxdx = -(freqsg/np.sqrt(2*np.pi))*np.exp(-0.5*freqsg*freqsg)
dhdx = -(freqsh/abs(freqsh))*0.5*np.exp(-abs(freqsh))

M = np.array([1,2,5,10,20,25,40,50,80,100,125,200,250,400,
500,625,1000,2000,5000,10000])
D = np.zeros([len(M),1])

for i in range(len(M)):
    print("Doing for: ",M[i])
#    (Pg,Xg) = interpolate(gk,freqsg,M[i])
    (Ph,Xh) = interpolate(hk,freqsh,M[i])

```

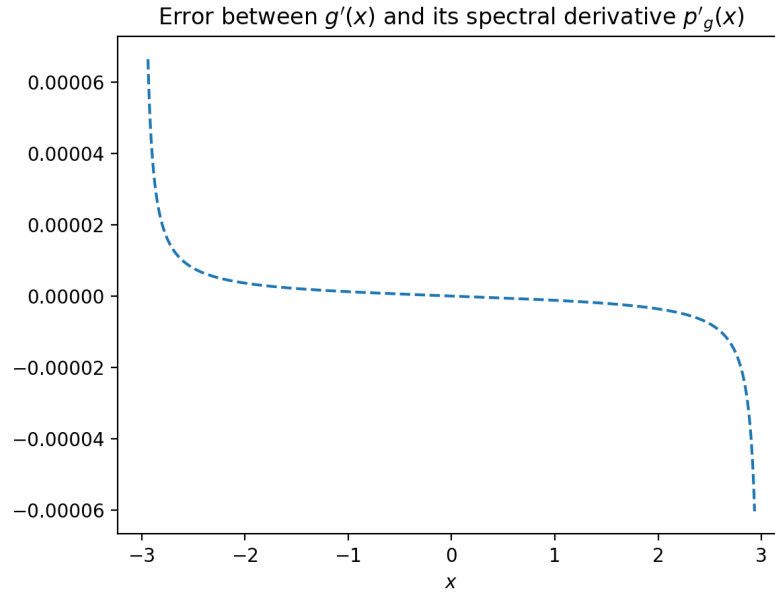


Figure 7: Error between  $g'(x)$  and its trigonometric polynomial interpolation  $p_{g'}(x)$

```
# D[i] = max(abs(dgdx[0:ng:int(ng/M[i])]-np.real(Pg)))
D[i] = max(abs(dhdx[0:nh:int(nh/M[i])]-np.real(Ph)))

plt.semilogx(L/M,D)

#plt.title('Accuracy (or error) vs $\Delta x$ for $g(x)$')
plt.title('Accuracy (or error) vs $\Delta x$ for $h(x)$')
plt.xlabel('$\Delta x$')
plt.ylabel('Accuracy (or error)')
plt.show()
```

The following figures show the variation of the accuracy measure for different values of  $\Delta x$  for both the functions  $g'(x)$  and  $h'(x)$ :



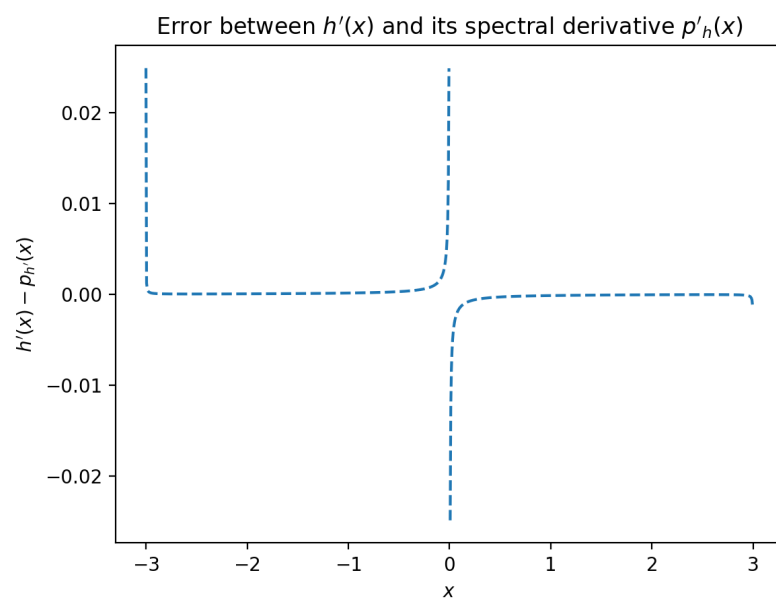


Figure 8: Error between  $h'(x)$  and its trigonometric polynomial interpolation  $p_{h'}(x)$

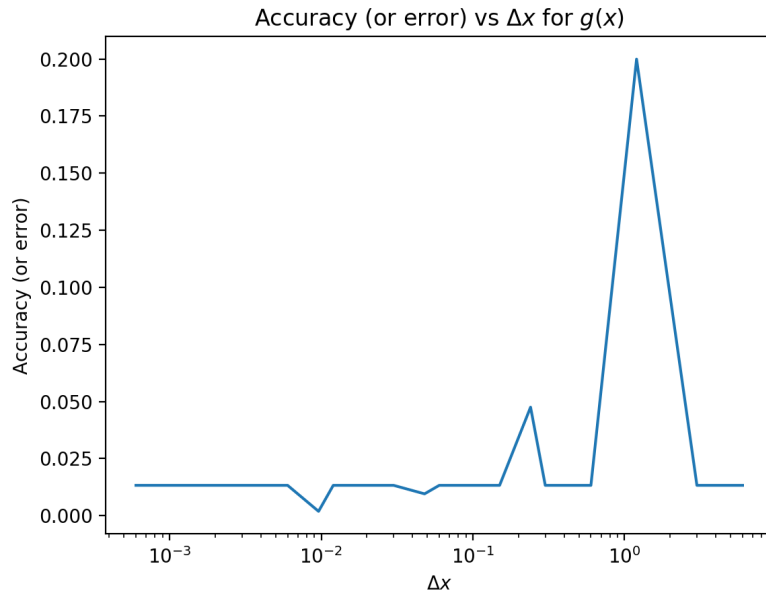


Figure 9: Accuracy vs  $\Delta x$  for the function  $g'(x)$  and  $p'_g(x)$

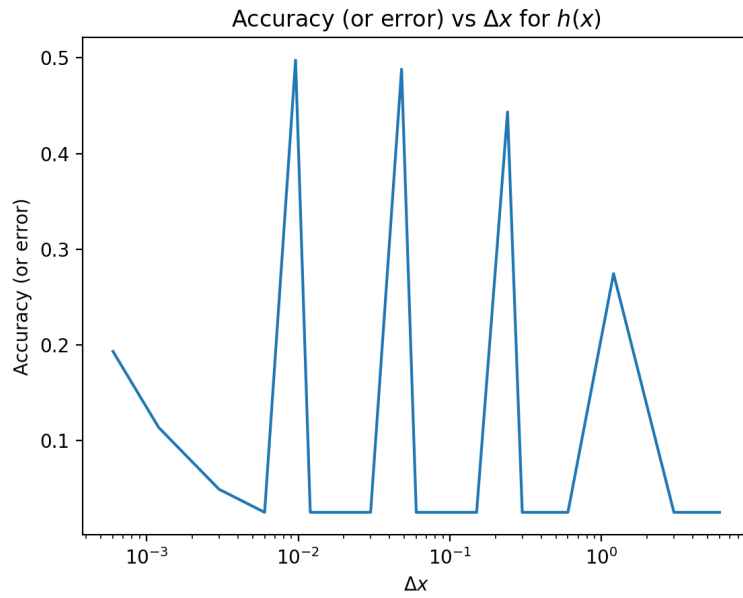


Figure 10: Accuracy vs  $\Delta x$  for the function  $h'(x)$  and  $p'_h(x)$