

1.) a.)

Taylor Series of $f(x) = \cos(\frac{x}{3})$ about $x_0 = \pi$:

$$T(x) = \cos(\frac{\pi}{3}) - \frac{1}{3} \cdot \sin(\frac{\pi}{3})(x-\pi) - \frac{1}{2} \cdot \frac{1}{9} \cdot \cos(\frac{\pi}{3})(x-\pi)^2 + \frac{1}{6} \cdot \frac{1}{27} \cdot \sin(\frac{\pi}{3})(x-\pi)^3 + \frac{1}{24} \cdot \frac{1}{81} \cdot \cos(\frac{\pi}{3})(x-\pi)^4 - \frac{1}{120} \cdot \frac{1}{243} \cdot \sin(\frac{\pi}{3})(x-\pi)^5 - \dots$$

$$T(x) = \cos(\frac{\pi}{3}) - \frac{1}{3} \cdot \frac{\sqrt{3}}{2}(x-\pi) - \frac{1}{18} \cdot \frac{1}{2}(x-\pi)^2 + \frac{1}{162} \cdot \frac{\sqrt{3}}{2}(x-\pi)^3 + \frac{1}{1944} \cdot \frac{1}{2}(x-\pi)^4 - \frac{1}{29160} \cdot \frac{\sqrt{3}}{2}(x-\pi)^5 - \dots$$

$$T(x) = \cos(\frac{\pi}{3}) - \frac{\sqrt{3}}{6}(x-\pi) - \frac{1}{36}(x-\pi)^2 + \frac{\sqrt{3}}{324}(x-\pi)^3 + \frac{1}{3888}(x-\pi)^4 - \frac{\sqrt{3}}{58320}(x-\pi)^5 - \dots$$

b.)

$$\text{true} = \cos(\frac{1}{3}) = 0.94495694631$$

$$\text{est} = 1 - \frac{(\frac{1}{3})^2}{2!} + \frac{(\frac{1}{3})^4}{4!} = 1 - \frac{1}{18} + \frac{1}{1944} = 0.94495884773$$

$$\text{absolute error} = |\text{true} - \text{est}| = |0.94495694631 - 0.94495884773| = 0.00000190142$$

$$\text{relative error} = \left| \frac{\text{true} - \text{est}}{\text{true}} \right| = \left| \frac{0.94495694631 - 0.94495884773}{0.94495694631} \right| = 0.00000201217$$

c.)

$$\text{b's absolute error} = 1.901 \times 10^{-6}$$

$$\text{b's relative error} = 2.012 \times 10^{-6}$$

$$\text{if } \hat{x} = 0.999, \hat{f}(\hat{x}) = 0.94506784876$$

$$\text{absolute error} = |0.94495884773 - 0.94506784876| = 0.00010900103$$

$$\text{relative error} = \left| \frac{0.94495884773 - 0.94506784876}{0.94495884773} \right| = 0.00011535002$$

d.)

$$\hat{f}(r) = \cos(\frac{x}{3}) - \frac{1}{3} \sin(\frac{x}{3})(r-x) - \frac{1}{18} \cos(\frac{x}{3})(r-x)^2$$

if $r = x + h$, using only the linear approximation gives:

$$\hat{f}(x+h) = \cos(\frac{x}{3}) - \frac{1}{3} \sin(\frac{x}{3})(x+h-x) = \cos(\frac{x}{3}) - \frac{1}{3} \sin(\frac{x}{3})h$$

$$\text{absolute error: } |f(x+h) - \hat{f}(x+h)| = \left| \cos(\frac{x+h}{3}) - \cos(\frac{x}{3}) - \frac{1}{3} \sin(\frac{x}{3})h \right|$$

$$\text{relative error: } \left| \frac{f(x+h) - \hat{f}(x+h)}{f(x+h)} \right| = \left| \frac{\cos(\frac{x+h}{3}) - \cos(\frac{x}{3}) - \frac{1}{3} \sin(\frac{x}{3})h}{\cos(\frac{x+h}{3})} \right|$$

For h very small, I will approximate $\cos(x+h) = \cos(x)$, so the approximated errors become:

$$\text{absolute error: } \left| \cos\left(\frac{x}{3}\right) - \cos\left(\frac{x}{3}\right) - \frac{1}{3} \sin\left(\frac{x}{3}\right)h \right| = \left| -\frac{1}{3} \sin\left(\frac{x}{3}\right)h \right|$$

$$\text{relative error: } \left| \frac{\cos\left(\frac{x}{3}\right) - \cos\left(\frac{x}{3}\right) - \frac{1}{3} \sin\left(\frac{x}{3}\right)h}{\cos\left(\frac{x+h}{3}\right)} \right| = \left| -\frac{1}{3} \tan\left(\frac{x}{3}\right)h \right|$$

e.)

$$\text{Condition of } f(x), \kappa = \left| \frac{x f'(x)}{f(x)} \right| = \left| \frac{-x \sin\left(\frac{x}{3}\right)}{3 \cos\left(\frac{x}{3}\right)} \right| = -\frac{x}{3} \tan\left(\frac{x}{3}\right)$$

$f(x)$ is ill conditioned for large magnitude x and for x close to $\pm \frac{\pi}{2}, \pm \frac{3\pi}{2}, \pm \frac{5\pi}{2}, \dots$

2.) a.)

$$f(x_0 + h) = f(x_0) + h f'(x_0) + \frac{h^2}{2} f''(x_0) + \frac{h^3}{3!} f'''(x_0) + \dots$$

$$f(x_0 - h) = f(x_0) - h f'(x_0) + \frac{h^2}{2} f''(x_0) - \frac{h^3}{3!} f'''(x_0) + \dots$$

$$f(x_0 + h) - f(x_0 - h) = 2h f'(x_0) + 2 \frac{h^3}{3!} f'''(x_0) + \dots$$

$$\frac{f(x_0+h) - f(x_0-h)}{2h} = f'(x_0) + \frac{h^2}{3!} f'''(x_0) + \dots$$

So an approximation for $f'(x_0)$ is $\frac{f(x_0+h) - f(x_0-h)}{2h}$.

The associated in error in this approximation is $\text{Error} = \frac{h^2}{3!} f'''(x_0) + \frac{h^4}{5!} f^{(5)}(x_0) + \dots$ which is $O(h^2)$

b.)

```
In [ ]: import numpy as np
import matplotlib.pyplot as plt

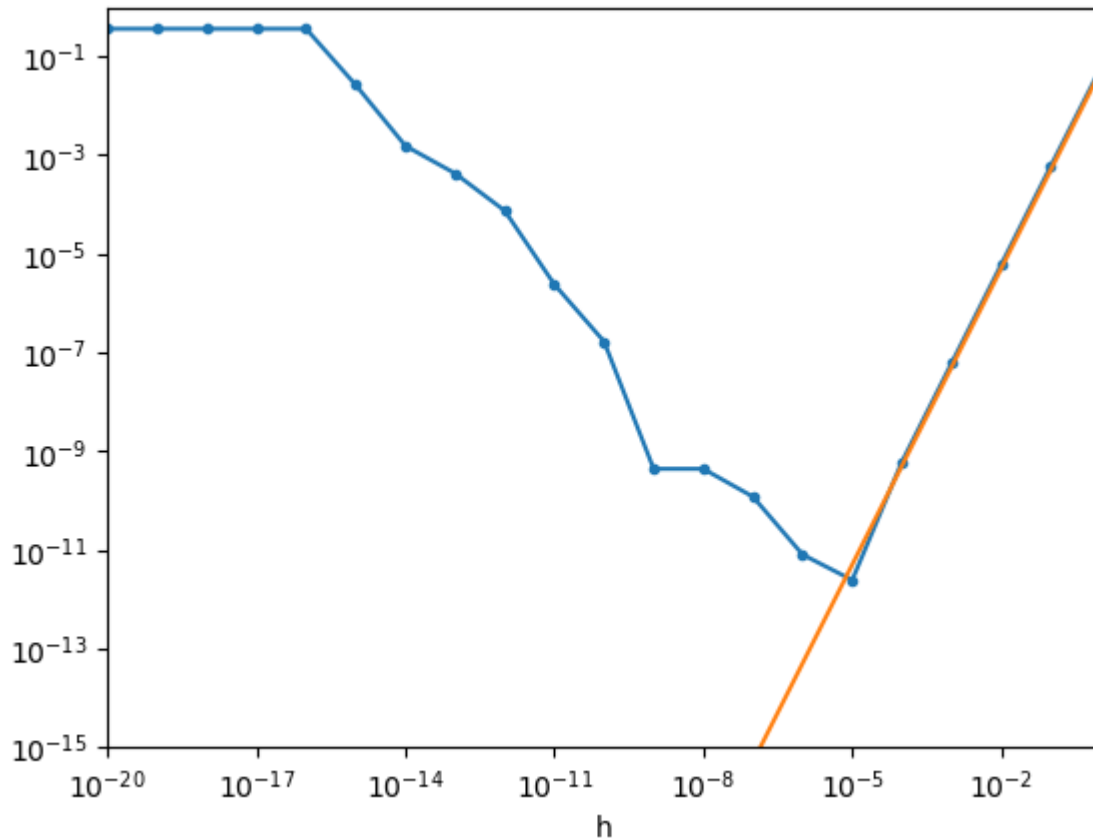
def f(x):
    return np.sin(x)

def derivative(x):
    return np.cos(x)

x0 = 1.2
f0 = f(x0)
fp = derivative(x0)
h = np.logspace(-20, 0, 21)
err = abs(fp - (f(x0+h) - f(x0-h)) / (2*h))

plt.plot(h, err, '-.')
plt.plot(h, 0.05*h**2)
plt.axis([1e-20, 1, 1e-15, 1e0])
plt.xlabel('h')
plt.ylabel('err')
```

```
plt.yscale('log')
plt.xscale('log')
```



The graph of the central difference approximation is steeper against h as the error using this approximation is $O(h^2)$, whereas the error in the forward differences approximation is $O(h)$. This algorithm begins to become inaccurate at h values of roughly 10^{-5} . This is when the cancellation error of $f(x+h) - f(x-h)$ begins to have a significant effect when multiplied by $\frac{1}{2h}$.

3.) a.)

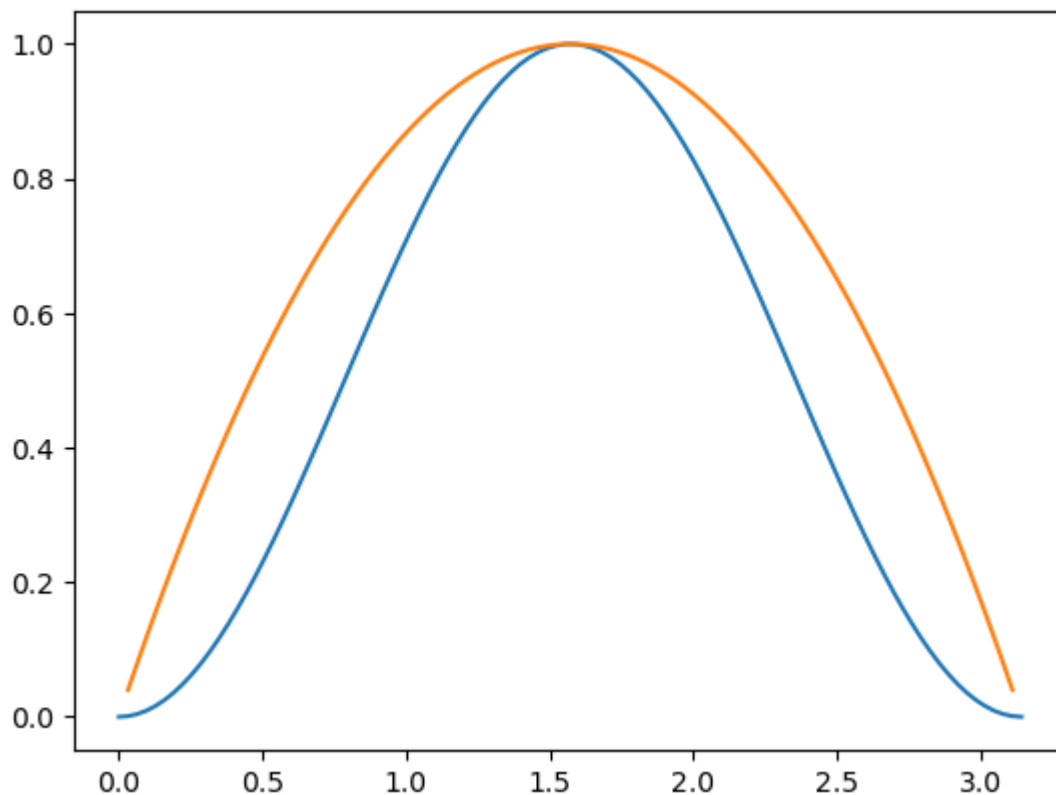
```
In [ ]: def f(x):
         return(np.sin(x)**2)

xs = [0, np.pi/2, np.pi]
ys = f(xs)

w0 = 2/(np.pi**2)
w1 = -4/(np.pi**2)
w2 = 2/(np.pi**2)

def p(x):
    return [x*(x-xs[1])*(x-xs[2])*(ys[0]*w0/x + ys[1]*w1/(x-xs[1]) + ys[2]*w2/(x-xs[2])

xArr = np.linspace(0,np.pi,100)
plt.plot(xArr,f(xArr),'-')
xArr = [x for x in xArr if x not in xs]
plt.plot(xArr,p(xArr),'-')
plt.show()
```

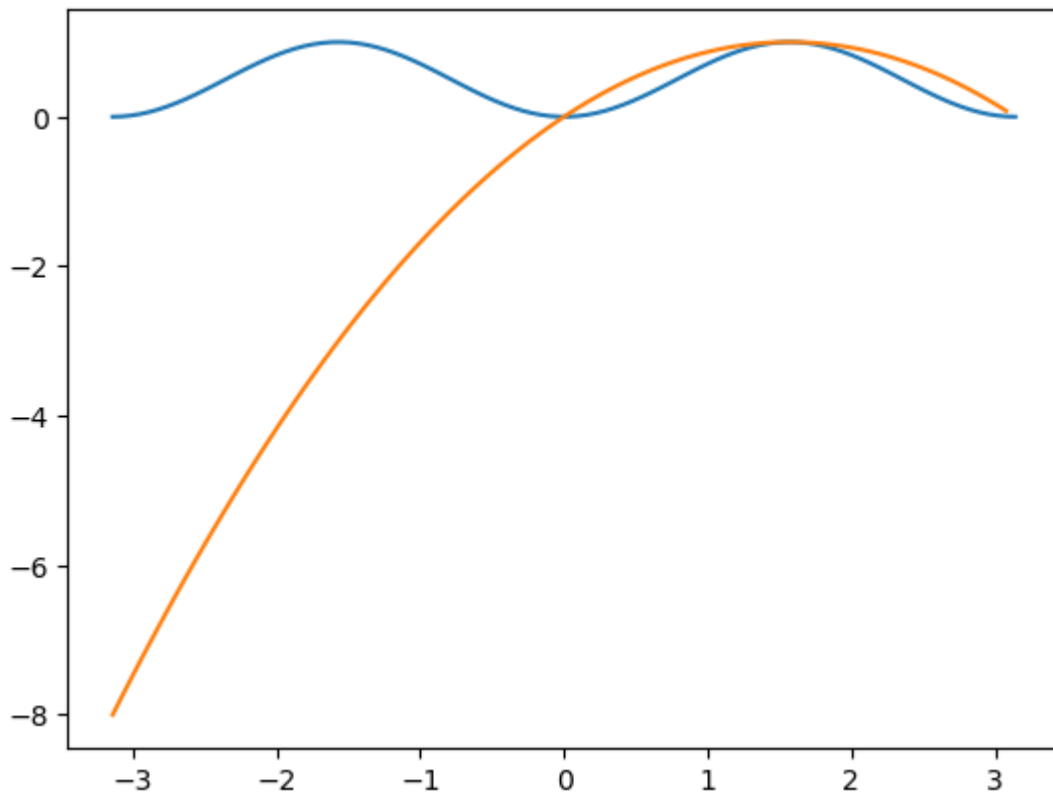


To improve this approximation with the fewest number of extra points, I would add points at $x = -\frac{\pi}{2}$ and $\frac{\pi}{2}$. These two points will significantly help the interpolation hold closely to the curve.

b.)

```
In [ ]: xArrBigger = np.linspace(-np.pi,np.pi,100)
plt.plot(xArrBigger,f(xArrBigger),'-')
xArrBigger = [x for x in xArrBigger if x not in xs]
plt.plot(xArrBigger,p(xArrBigger),'-')
```

```
Out[ ]: [<matplotlib.lines.Line2D at 0x1fd87d6c0d0>]
```



To improve this interpolation, I would add symmetric points, $x = -x$. This should help the curve come back up and down for increasingly negative x s.

In general, to improve an interpolation, add extra points where the interpolation is visually very different from the original function.

4.) a.)

Using Newton Bases,

$$u(x) = (\gamma_0 + \gamma_1 x + \gamma_2 x^2)^{\frac{1}{2}}$$

$$v(x) = u(x)^2 = \gamma_0 + \gamma_1 x + \gamma_2 x^2$$

$$\phi_0 = 1, \phi_1 = (x - x_0), \phi_2 = (x - x_0)(x - x_1)$$

$$v(x) = y_0^2 + \frac{y_1^2 - y_0^2}{x_1 - x_0}(x - x_0) + \frac{\frac{y_2^2 - y_1^2}{x_2 - x_1} - \gamma_1}{x_2 - x_0}(x - x_0)(x - x_1)$$

$$u(x) = \sqrt{v(x)} = \sqrt{y_0^2 + \frac{y_1^2 - y_0^2}{x_1 - x_0}(x - x_0) + \frac{\frac{y_2^2 - y_1^2}{x_2 - x_1} - \gamma_1}{x_2 - x_0}(x - x_0)(x - x_1)}$$

b.)

```
In [ ]: pts = [(0.1,0.5), (1,1), (2,4)]
def u(x):
```

```

gamma1 = (pts[1][1]**2 - pts[0][1]**2/(pts[1][0]-pts[0][0]))
return(np.sqrt(pts[0][1]**2 + gamma1*(x-pts[0][0]) + (((pts[2][1]**2 - pts[1][1]**
2)/(pts[2][0] - pts[1][0]) - gamma1)/(pts[2][0] - pts[0][0]))*(x-pts[0][0])*(x-pts[1]
[0])))

xArr = np.linspace(0,2,100)
plt.plot(xArr, u(xArr), '-')
plt.show()

```

C:\Users\Cameron Holland\AppData\Local\Temp\ipykernel_19380\3740751931.py:5: RuntimeWarning: invalid value encountered in sqrt

```

return(np.sqrt(pts[0][1]**2 + gamma1*(x-pts[0][0]) + (((pts[2][1]**2 - pts[1][1]**
2)/(pts[2][0] - pts[1][0]) - gamma1)/(pts[2][0] - pts[0][0]))*(x-pts[0][0])*(x-pts[1]
[0])))

```

