Problem 1: Calculate the machine epsilon of your computer using the machine epsilon function you are required to write.

To solve this problem, I utilized the program "machine_epsilon.py" that I wrote for estimating my machine's machine epsilon. The program divides a number one by two many times until the difference between the results is undetectable by the machine due to precision limitations of the machine's binary representation of floating-point values. The Figure 1 below shows output of my machine's program.

```
## Cameron Calv ECE T480
   import numpy as np
   def machine_epsilon():
       epsilon = 1
       while epsilon+1 > 1:
           epsilon = epsilon/2
       return epsilon *2
   if __name__ == '__main__':
       print("My calculated Machine Epsilon: ", machine_epsilon())
       print("Numpy's Machine Epsilon: "_np.finfo(float).eps)
       assert(machine_epsilon() == np.finfo(float).eps)
       print("These values are the same!")
   machine_epsilon()
machine_epsilon ×
D:\Python\Python37-32\python.exe "D:/Archives III/School/4 College/
My calculated Machine Epsilon: 2.220446049250313e-16
Numpy's Machine Epsilon: 2.220446049250313e-16
These values are the same!
```

Figure 1: Calculating machine epsilon of my system.

According to the program (which is assert-checked with *numpy*'s *finfo* property) <u>my machine epsilon is</u> **2.220446049250313e-16.**

Within the file *problems_hw1.py* problem 2 is worked out. A snippet of the code and the output is shown below in Figure 2.

```
#Problem 2
          actual = (pi * pi * pi * pi) / (90)
          print("Actual : " + str(actual))
          forward_sum = 0
          for i in range(1, 10000):
              forward_sum = forward_sum + (1/(i*i*i*i))
          print("Forward Approx: "+str(forward_sum))
          print("Relative Accuracy: "+str((forward_sum/actual)*100)+"%")
          backward_sum = 0
          for i in reversed(range(1, 10000)):
              backward_sum = backward_sum + (1/(i*i*i*i))
          print("Backward Approx: "+str(backward_sum))
          print("Relative Accuracy: "+str((backward_sum/actual)*100)+"%\n")
problems_hw1 >
Problem 2: Calculate f(n) = summation(1/(i^4)) for i=1 to n. n = 10000
Actual : 1.082323233711138
Forward Approx: 1.082323233710861
Relative Accuracy: 99.9999999997442%
Backward Approx: 1.0823232337108049
Relative Accuracy: 99.9999999996922%
```

Figure 2: Code and output showing the solution for Problem 2.

According to the output of the code the relative error for both approximations are **almost 99.999%** accurate with the forward approximation being slightly better than the backwards approximation. This is most likely the case because the there is truncation when dealing with the higher *i* numbers as the summation continues. This truncation makes adding backwards and forwards different when done on a limited accuracy machine.

Within the file *problems_hw1.py* problem 3 is worked out. A snippet of the code and the output is shown below in Figure 3.

```
actual = 6.737947*(10**(-3))
          print("Actual : "+str(actual))
          num_terms = 20
          series_approx = 0
          for i in range(num_terms):
              series\_approx = series\_approx + ((-1)**(i))*(((5)**(i))/fact(i))
          print("Series Approx: "+str(series_approx))
          print("Relative Accuracy: "+str((series_approx/actual)*100)+"%")
          inv_series_approx = 0
          for i in range(num_terms):
              inv_series_approx = inv_series_approx + (((5)**(i))/fact(i))
          inv_series_approx = inv_series_approx**(-1)
          print("Inverted Series Approx: "+str(inv_series_approx))
          print("Relative Accuracy: "+str((inv_series_approx/actual)*100)+"%\n")
Problem 3: Estimate e^(-5) using series expansion and inverted series expansion
Actual : 0.006737947
Series Approx: 0.00670634105421557
Relative Accuracy: 99.53092617403446%
Inverted Series Approx: 0.00673794932511709
Relative Accuracy: 100.00003450779727%
```

Figure 3: Code and output for Problem 3.

As shown in the figure, the accuracy of the **first approximation is about 99.53% and the second approximation is about 100.00%.**

Problem 4: The Maclaurin series expansion for $\cos x$ is $\cos x = 1 - x \ 2 \ 2 + x \ 4 \ 4! - x \ 6 \ 6! + x \ 8 \ 8! - ...$ Starting with the simplest version, $\cos x = 1$, add terms one at a time to estimate $\cos(\pi/3)$. After each new term is added, compute the true and approximate percent relative errors. Use your pocket calculator to determine the true value. Add terms until the absolute value of the approximate error estimate falls below an error criterion conforming to two significant figures. Use the fact function you are required to write to calculate the factorials needed.

Within the file *problems_hw1.py* problem 4 is worked out. A snippet of the code and the output is shown below in Figure 4Figure 3.

```
sig_bits = 2
         eps = 10**(-sig_bits)
         approx = 0
         num_terms_count = 0
            num_terms_count = num_terms_count + 1
             approx_sum = 0
             for i in range(num_terms_count):
                approx_sum = approx_sum + ((-1)^{**}(i))^*((pi/3)^{**}(2*i))/(fact(2*i))
            approx = approx_sum
            print(str(num_terms_count)+": Term(s) |\tApprox: "+str(approx)+" |\tError: "+str(abs(approx-actual)))
         print("It took "+str(num_terms_count)+" terms to approximate within "+str(sig_bits)+" significant figures!\n")
Problem 4: Add terms to the approximation of cos(pi/3) until error falls beyond two sigfigs.
Actual : 0.50000000000000001
2: Term(s) | Approx: 0.45168864438392464 | Error: 0.04831135561607547
3: Term(s) | Approx: 0.501796201500181 | Error: 0.001796201500180894
It took 3 terms to approximate within 2 significant figures!
```

Figure 4: Code and snippet of output for Problem 4.

It is shown that it takes **three terms** to get the answer precise enough to be within two significant figures of accuracy.

Problem 5: Consider the function $f(x) = x \cdot 3 - 2x + 4$ on the interval [-2,2] with h = 0.25. Use the forward, backward, and centered finite difference approximations for the first and second derivatives so as to graphically illustrate which approximation is most accurate. Graph all three first derivative finite difference approximation along with the theoretical, and do the same for the second derivative as well. Use the deriv1 and deriv2 functions you are required to make to calculate the derivatives.

Within the file *problems_hw1.py* problem 4 is worked out. The graphs asked for by the problem are shown for in __ and __ respectively.

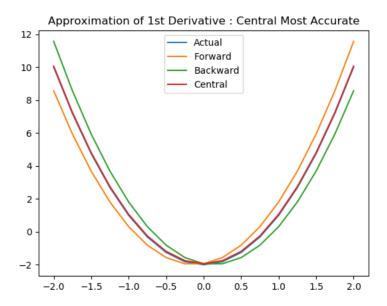


Figure 5: Plot showing comparison of first derivative approximations.

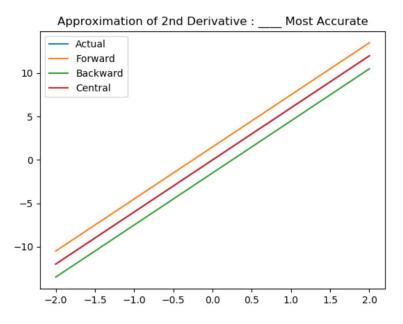


Figure 6: Plot showing comparison of second derivative approximations.