Math 151A Homework #2

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Problem 0.1.

Let p be some point such that f(p) = 0. Then g has a fixed point at p iff g(p) = p.

(a)

$$f(p) = 0 = p^{4} + 2p^{2} - p - 3$$
$$p + 3 - 2p^{2} = p^{4}$$
$$p = (p + 3 - 2p^{2})^{1/4}$$
$$p = g_{1}(p)$$

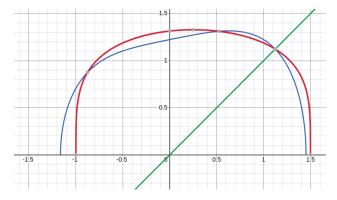
(b)

$$f(p) = 0 = p^{4} + 2p^{2} - p - 3$$
$$p + 3 - p^{4} = 2p^{2}$$
$$p = \left(\frac{x + 3 - x^{4}}{2}\right)^{1/2}$$
$$p = g_{2}(p)$$

Problem 0.2.

```
\begin{array}{l} p_-0 &= 1 \\ p_-1 &= 1.224744871391589 \\ p_-2 &= 0.9936661590774817 \\ p_-3 &= 1.228568645274987 \\ p_-4 &= 0.9875064291508866 \end{array}
```

Here is a graph of g_1 in red and g_2 in blue:



From that graph, we can see that if $p_0 = 1$, then FPI with either g_1 or g_2 will approximate the root $r_1 = 1.124123...$ (and not the root r_2 , which is negative). After 4 iterations, FPI with g_1 approximates r_1 to be 1.107821 and FPI with g_2 approximates that same root as 0.987506. Therefore, I think g_1 is better for approximating the root r_1 .

Problem 0.3.

There are 4 hypotheses needed to show existence and uniqueness of fixed points. If the following are all true:

- $g \in C([a,b])$
- $g(x) \in [a, b]$ for all $x \in [a, b]$
- g'(x) exists when $x \in (a, b)$
- there exists $k \in (0,1)$ such that $|g'(x)| \leq k$ for all $x \in (a,b)$

then g has a unique fixed point on [a, b]. If the first two criteria are true, then g has a fixed point on [a, b], but that alone would not guarantee it is unique.

Let $a=0, b=2\pi, g(x)=\pi+0.5\sin(x/2)$. Then g is continuous on [a,b]. Since the sine of a real number is always in [-1,1], the image g([a,b]) is in $[\pi-0.5,\pi+0.5]$. g is differentiably everywhere, and $g'(x)=0.25\cos(x/2)$. So if k=0.25, then |g'(x)| is always less than or equal to k.

All 4 hypotheses are satisfied, so g has a unique fixed point on $[0, 2\pi]$.

Problem 0.4.

```
>>> from math import \sin, \cos
>>> def f(x): return -x**3 - \cos(x)
...
>>> def f_prime(x): return -3 * x**2 + \sin(x)
...
>>> p = -1
```

```
>>> iteration = 0
>>>  tolerance = 1e-10
>>> \max_{\text{iterations}} = 40
>>> while abs(f(p)) > tolerance and iteration < max_iterations:
          iteration += 1
         p = f(p) / f_prime(p)
         \mathbf{print}(f^*\{iteration = \} \setminus t_{p} = \} \setminus t_{r} = idual = \{abs(f(p))\}^*)
. . .
                    p = -0.880332899571582
                                                 residual = 0.04535115463649053
iteration=1
iteration=2
                    p = -0.8656841631760818
                                                 residual = 0.0006323133963189731
                                                 residual = 1.289199819121123e - 07
iteration=3
                    p = -0.865474075952977
iteration=4
                    p = -0.8654740331016162
                                                 residual = 5.218048215738236e - 15
```

Newton's method converges very quickly. The approximation after two iterations is $p_2 = -0.8656841631760818$ and the residual is $f(p_2) = 0.0006323133963189731$. Newton's method only works because we chose a good value for p_0 though. If p_0 were zero, then $f(p_0)$ would be nonzero and $f'(p_0)$ would be zero, so $p_1 = p_0 - \frac{f(p_0)}{f'(p_0)}$ would be undefined.

Problem 0.5.

```
>>> from math import cos
\Rightarrow  def f(x): return -x**3 - cos(x)
>>> p = [-1, 0]
>>> while len(p) <= 3:
         p.append((p[-2] * f(p[-1]) - p[-1] * f(p[-2])) / (f(p[-1]) - f(p[-2])))
>>>  for i in range(len(p)):
         print (f"p_{i} =_{p[i]} residual={abs(f(p[i]))}")
p_0 = -1 \text{ residual} = 0.45969769413186023
p_1 = 0 residual = 1.0
p_2 = -0.6850733573260451 \text{ residual} = 0.45285023447500383
p_3 = -1.252076488909229 residual=1.649523592018598
After 2 iterations, our approximation (p_3) is pretty bad, but if we kept going, it would eventually converge:
p_4 = -0.8072055385060928
                                     residual = 0.16556011526065983
p_{-5} = -0.8477837694325692
                                     residual = 0.05231270017050571
p_{-6} = -0.8665281869207425
                                     residual = 0.0031747088825289094
p_{-7} = -0.8654557261640932
                                     residual = 5.5076148661736823e - 05
p_{-8} = -0.8654740143806453
                                     residual = 5.632275656974883e - 08
p_{-}9 = -0.8654740331019471
                                     residual = 1.0009770790020411e - 12
p_{-}10 = -0.8654740331016145
                                     residual = 2.220446049250313e - 16
```

Problem 0.6.

Assume $f'(p_0) \neq 0$, and define $g(x) = x - f(x)/f'(p_0)$. Let $p_1 = g(p_0), p_2 = g(p_1)$, and so on. f is continuous on [a, b] and $f'(p_0)$ is assumed to be nonzero, so g is also continuous on [a, b] Suppose p_n converges to p as n goes to infinity. By Taylor's theorem,

$$g(x) = g(x_0) + g'(\xi_x) \cdot (x - x_0)$$

for some ξ_x between x_0 and x. If we let $x = p_{n-1}$ and $x_0 = p_0$, that equation becomes

$$g(p_{n-1}) = g(p) + g'(\xi_n) \cdot (p_{n-1} - p)$$

(for some ξ_n between p_{n-1} and p). That simplifies to

$$p_n = p + g'(\xi_n) \cdot (p_{n-1} - p),$$

which can be rewritten as

$$g'(\xi_n) = \frac{p_n - p}{(p_{n-1} - p)}.$$

Since ξ_n is between p_{n-1} and p_n and p_n converges to p, ξ_n must also converge to p. The function g' is continuous (because $g'(x) = 1 - f'(x)/f'(p_0)$, and f is coninuously differentiable), so that means $g'(\xi_n)$ converges to $g'(p) = 1 - f'(p)/f'(p_0)$. We are given that $f'(p) \neq f'(p_0)$, so

$$\lim_{n \to \infty} \frac{|p_n - p|}{|p_{n-1} - p|} = |g'(p)| > 0.$$

We do not have enough information to show that L < 1, where L := |g'(p)|, but if we assume L < 1, then we have shown p_n converges linearly to p.

Problem 0.7.

Note that for this question, there is some ambiguity about whether to start with p_0 or p_1 for Newton's method. I chose to ignore the given $p_1 = 5$ for Newton's method, and instead start with p_0 .

Here is the python program and output I used for this problem. I took out the line that attempted to use the variation of Newton's method from problem 6 with $p_0 = -5$ to find the root, because that did not converge.

```
#!/usr/bin/env python3
from math import cos, sin
\mathbf{def} \ \mathbf{f}(\mathbf{x}) \colon \mathbf{return} \ \mathbf{x} + \cos(\mathbf{x})
\operatorname{def} f_{\operatorname{prime}}(x) : \operatorname{return} 1 - \sin(x)
def newtons_method(f, f_prime, p_0, tolerance, weird_version=False):
     if weird_version:
          print(f"\nVariation_of_Newton's_method_from_problem_6")
     else:
          print(f"\nNewton's_method")
     print (f" {p_0=}\t {tolerance=}\n{'='*53}")
     i = 0
     p = p_0
     while abs(f(p)) > tolerance:
          i += 1
          if weird_version:
               p = f(p) / f_prime(p_0)
          else:
               p = f(p) / f_prime(p)
          print (f"p_{i} = \{i:02\} = \{p: >16.10f\}, = residual = \{abs(f(p)): >16.10f\}")
def secant_method(f, p_0, p_1, tolerance):
     print(f"\nSecant_method")
```

```
print (f" {p_0=}\t {p_1=}\t {tolerance=}\n{'='*53}")
   p = [p_0, p_1]
   while abs(f(p[-1])) > tolerance:
       p.append((p[-2]*f(p[-1])-p[-1]*f(p[-2])) / (f(p[-1])-f(p[-2])))
   for i in range(len(p)):
       print(f"p_{i} = \{i:02\} = \{p[i]: >16.10f\}, = residual = \{abs(f(p[i])): >16.10f\}")
newtons\_method(f, f\_prime, -5, 1e-10)
\operatorname{secant\_method}(f, -5, 5, 1e-10)
newtons\_method(f, f\_prime, -.9, 1e-10)
newtons_method(f, f_prime, -.9, 1e-10, weird_version=True)
\operatorname{secant\_method}(f, -.9, 5, 1e-10)
Newton's_method
p_0 = -5 = tolerance = 1e - 10
p_01 = 109.8205607048, residual = 108.8296839085
p_02 = 1.59607775611, residual = 16.9289898794
p_0 = 03 = 0.6.6151154617, residual = 0.007.5605305903
p_04 = -4.6000997047, residual = -4.7121531554
p_05 = 1743.6197251930, residual = 1743.0281085215
p_006 = 3090.7583906372, residual = 3089.9158322043
p_07 = 3606.1400727092, residual = 3607.0578723786
p_0 = 08 = 1024.2182226354, residual = 1025.2164816044
p_09 = -65.2592470776, erroridual = -66.0148654568
p_10 = 25.3714027324, residual = 24.3997474637
p_111 = -5.6369297160, residual = -4.8385854382
p_12 _=____6.5264723344, __residual _=___7.4970237282
p_1 = 3 = 3.3496519727, residual = 3.3496519727
p_14 = 2.1051947425, extresidual = 2.1051947425
p_15 = -9.3409121107, residual = -10.3373974311
p_16 = 0.1974775656, extresidual = 0.11780421553
p_117 = 1.2681072761, residual = 1.2681072761
p_18 = 0.00 - 0.7718165872, errorightarrow errorightarrow = 0.0551716918
p_19 = 0.0003825038
p_20 = 0.7390851447, residual = 0.00000000193
Secant_method
p_0 = -5 = p_1 = 5 = tolerance = 1e - 10
p_01_=___5.00000000000, __residual_=___5.2836621855
p_02 = 0.02 = 0.2836621855, extresidual = 0.6763747448
p_003 = 1.0593324054, residual = 1.0598780527
p_04 = 0.000 - 0.7046391719, residual = 0.0572061591
p_05 = 0.034943006
p_006 = 0.7391013273, residual = 0.0000271027
p\_07 \, \bot = \!\!\! \_\_\_\_ - 0.7390851257 \, , \, \_\_residual \, \bot = \!\!\! \_\_\_\_\_ 0.0000000125
Newton's method
```

tolerance=1e-10

 $p_0 = -0.9$

```
p_{-}01 =
      -0.7438928778,
                 residual =
                          0.0080548285
p_{-}02 =
      -0.7390902113,
                 residual =
                          0.0000084988
p_{-}03 =
      -0.7390851332,
                 residual =
                          0.0000000000
Variation of Newton's method from problem 6
p_0 = -0.9 ____ tolerance=1e-10
p_01 = 0.01 = 0.0080548285
p_02 = 0.7393761353, residual = 0.0004870559
p_04 = 0.00000018415
p_05 = 0.7390852009, residual = 0.0000001133
p_07 = 0.7390851335, residual = 0.00000000004
```

Secant_method

```
p_0 = -0.9 ____ p_1 = 5 ___ tolerance=1e - 10
```

- (a) With $p_0 = -5$, Newton's method took 21 iterations to reach the desired tolerance. With $p_0 = -5$ and $p_1 = 5$, the secant method took only 7 iterations. With $p_0 = -5$, the variation of Newton's method from problem 6 diverged, which I believe is related to the fact that we couldn't prove L < 1 in problem 6. When L > 1, the method will "diverge linearly" instead of converging linearly.
 - With $p_0 = -0.9$, Newton's method took 3 iterations to reach the desired tolerance. With $p_0 = -0.9$ and $p_1 = 5$, the secant method took 6 iterations. With $p_0 = -0.9$, the variation of Newton's method (from problem 6) took 7 iterations.
- - To obtain these results, I modified the code above to print with more precision otherwise the approximations from each method would be the same.
- (c) See the program output above.

Newton's method converges quadratically, which makes it very good, but only if p_0 is chosen well. When p_0 was 5, it took a while to get close to the answer, but once it did get close, it converged super fast. The variation of Newton's method didn't converge at all, because $f'(p_0)$ was almost exactly zero. On the other hand, when we chose $p_0 = -0.9$, Newton's method converged super fast (only 3 iterations), and the variation of Newton's method converged linearly. The secant method, on the other hand, converged in 7 iterations when p_0 was -5, and in 6 iterations when p_0 was -0.9. The secant method is more consistent because it doesn't depend as much on the initial guess, but it doesn't have the advantage of converging quadratically.