# Final Project Report

# Techniques for Root Finding, Interpolation & Systems of Linear Equations



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#### **Tools**

We used python 3.6 to develop the project. We also used the following libraries:

- PyQt5: Used in creating the GUI.
- Matplotlib: For plotting all charts and graphs used throughout the application
- Sympy: For parsing input equations and differentiation.

#### Design

We used some simple design patterns in order to facilitate the process of development and keep the project maintainable for more features. We used an MVC architecture where separated the main logic of the methods implemented from the GUI the user interacts with. We also used the Factory design pattern in the root finding part of the program where the user selects whichever method they want to use and the results are generated through the root finder factory.

#### What we added

- Implementation for the modified Newton methods in the lectures:
  - o First modified Newton method which requires multiplicity to be given
  - Second modified Newton method
- A module to solve systems of linear equations (Max: 10 equations) using Gauss-Jordan method

# **Assumptions**

- In interpolation
  - we use all points supplied by the user to interpolate given query points regardless of the polynomial order.
  - Query points have to be within the range of the sample points given, otherwise an error message is shown.

# **Part 1: Root-Finding**

#### **Objective**

Implementing the following six root-finding methods as well as a general algorithm for finding nearly all the roots of a given function.

# 1.1. Implemented Algorithms

#### 1.1.1. Bisection

```
FUNCTION Bisect(x1, xu, es, imax, xr, iter, ea)
  iter = 0
  DO
   xrold = xr
   xr = (x_1 + x_1) / 2
    iter = iter + 1
    IF xr \neq 0 THEN
       ea = ABS((xr - xrold) / xr) * 100
    END IF
    test = f(x1) * f(xr)
    IF test < 0 THEN
       xu = xr
    ELSE IF test > 0 THEN
       x1 = xr
    ELSE
       ea = 0
    END IF
    IF ea < es OR iter ≥ imax EXIT
  END DO
  Bisect = xr
END Bisect
```

#### 1.1.2. False-Position

```
FUNCTION ModFalsePos(xl, xu, es, imax, xr, iter, ea)
 iter = 0
 f = f(x)
 fu = f(xu)
 DO
   xrold = xr
   xr = xu - fu * (x1 - xu) / (f1 - fu)
   fr = f(xr)
   iter = iter + 1
   IF xr <> 0 THEN
     ea = Abs((xr - xrold) / xr) * 100
   END IF
   test = fl * fr
   IF test < 0 THEN
     xu = xr
     fu = f(xu)
      iu = 0
      i7 = i7 + 1
      If il \geq 2 THEN fl = fl / 2
   ELSE IF test > 0 THEN
     x1 = xr
     f = f(x)
      i7 = 0
      iu = iu + 1
     IF iu \ge 2 THEN fu = fu / 2
   ELSE
     ea = 0
   END IF
   IF ea < es OR iter ≥ imax THEN EXIT
 END DO
 ModFalsePos = xr
END ModFalsePos
```

#### 1.1.3. Fixed-Point

```
FUNCTION Fixpt(x0, es, imax, iter, ea)
xr = x0
iter = 0
D0
xrold = xr
xr = g(xrold)
iter = iter + 1
IF xr \neq 0 \text{ THEN}
ea = \left| \frac{xr - xrold}{xr} \right| \cdot 100
END IF
IF ea < es OR iter \ge imax EXIT
END DO
Fixpt = xr
END Fixpt
```

#### 1.4. Newton-Raphson

```
1. Choose \in > 0 (function tolerance |f(x)| < \in)
m > 0 \text{ (Maximum number of iterations)}
x_0 - \text{initial approximation}
k - \text{iteration count}
Compute f(x_0)

2. Do \{q = f'(x_0) \text{ (evaluate derivative at } x_0)
x_1 = x_0 - f_0/q
x_0 = x_1
f_0 = f(x_0)
k = k+1
\}
3. While (|f_0| \ge \in) and (k \le m)

4. x = x_1 the root.
```

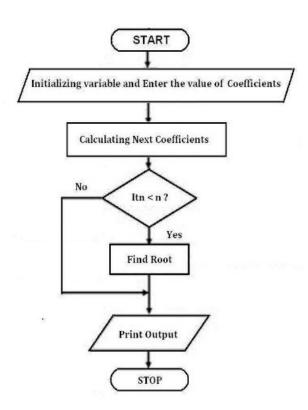
#### 1.1.5. Secant

- 1. Choose  $\in > 0$  (function tolerance  $|f(x)| \le \in$ ) m > 0 (Maximum number of iterations)  $x_0$ ,  $x_1$  (Two initial points near the root)  $f_0 = f(x_0)$   $f_1 = f(x_1)$  k = 1 (iteration count)

  2. Do  $\begin{cases} x_1^5 = x_1^T \left(\frac{x_1^T x_2^0}{x_1^T x_2^T}\right) \\ x_0 = x_1 \\ f_0 = f_1 \\ x_1 = x_2 \\ f_1 = f(x_2) \end{cases}$
- 3. While  $(|f_1| \ge \epsilon)$  and  $(m \le k)$

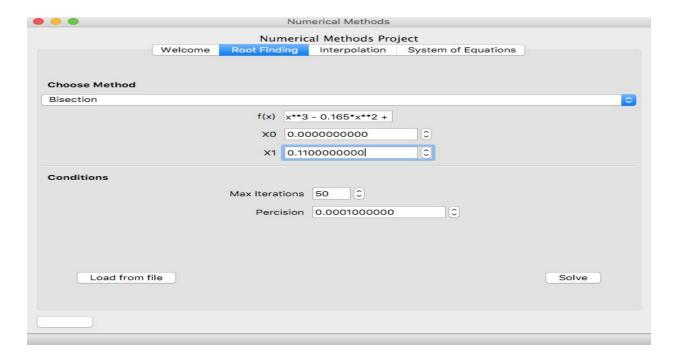
k = k+1

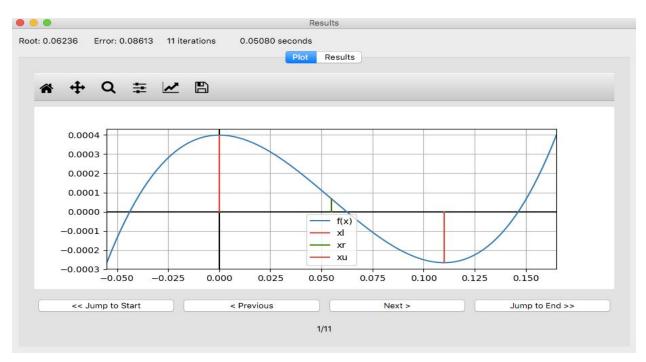
#### 1.1.6. Birge-Vieta

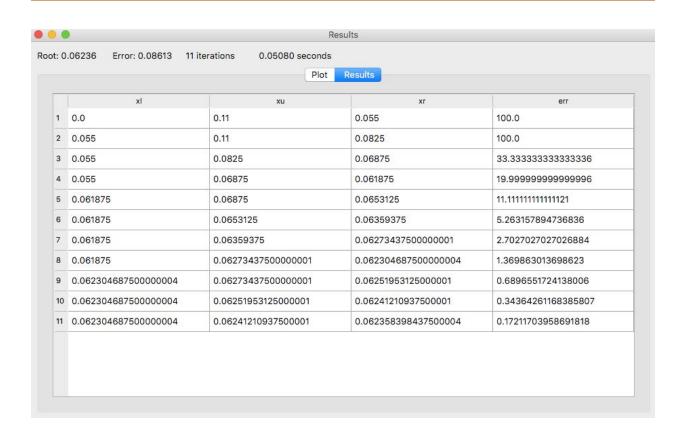


#### Sample runs:

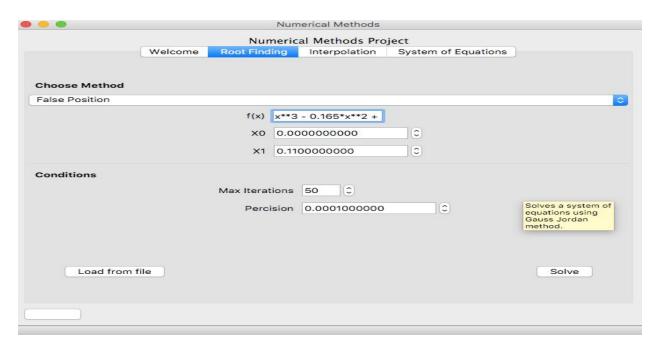
#### 1.1. Bisection

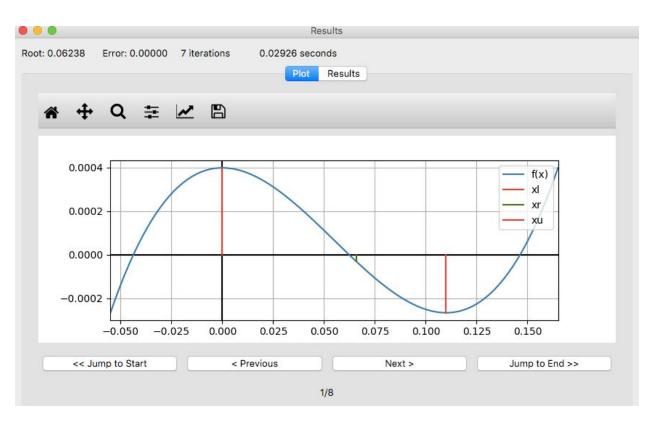






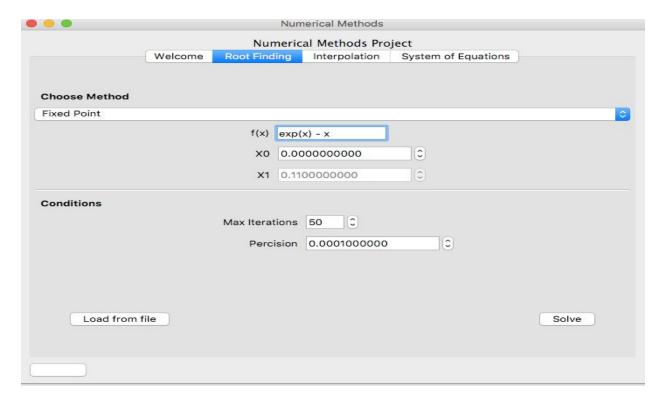
## 1.2. False position

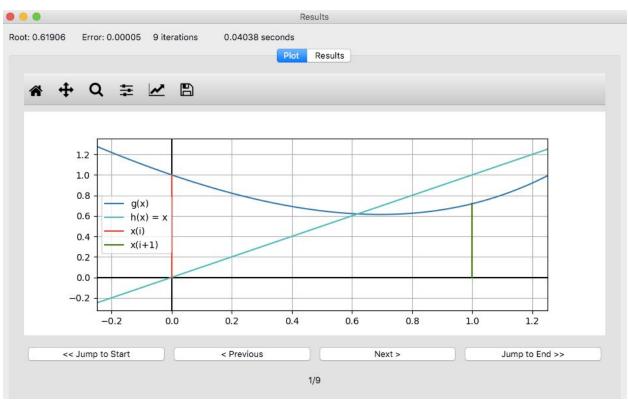


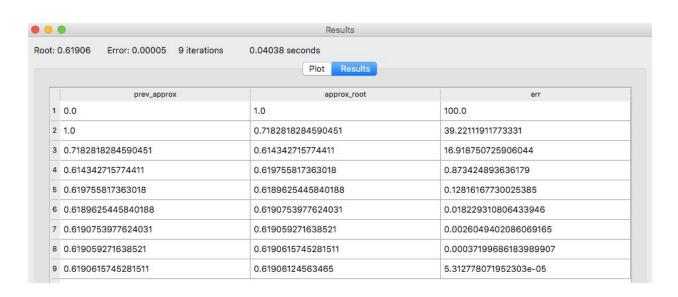


	xl	xu	xr	err
	0.0	0.11	0.0659999999999999	0.0
	0.0	0.0659999999999999	0.0611111111111111111111111111111111111	100.0
	0.061111111111111116	0.0659999999999999	0.06239027683997298	7.99999999999973
	0.061111111111111116	0.06239027683997298	0.06237761907271974	2.050264550264527
	0.061111111111111116	0.06237761907271974	0.062377581624796626	0.02029216158199779
-	0.061111111111111116	0.062377581624796626	0.06237758151407783	6.003426574287142e-05
	0.061111111111111116	0.06237758151407783	0.06237758151375047	1.7749773645547787e-07
	0.06111111111111111	0.06237758151375047	0.06237758151374952	5.248089596522295e-10

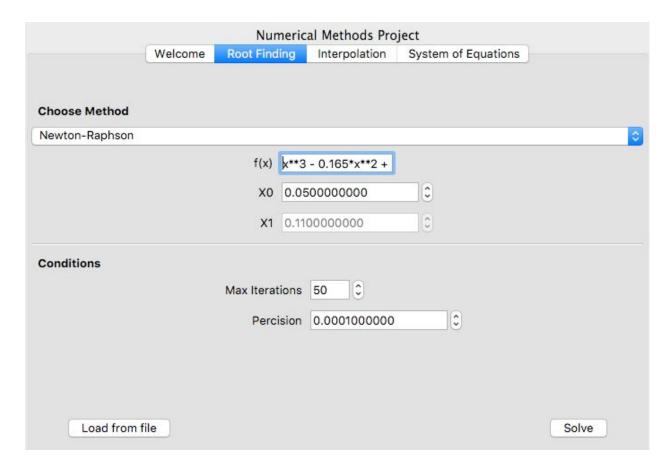
#### 1.3. Fixed point



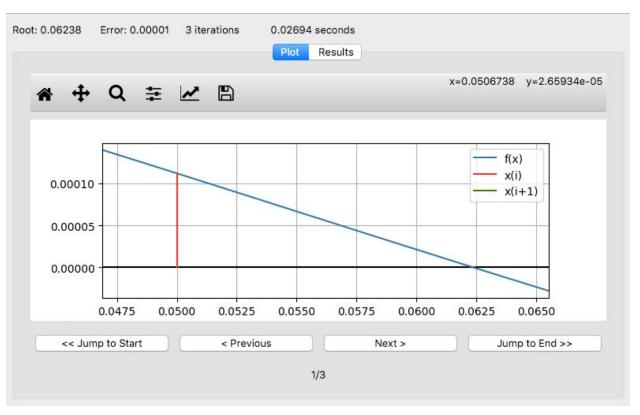




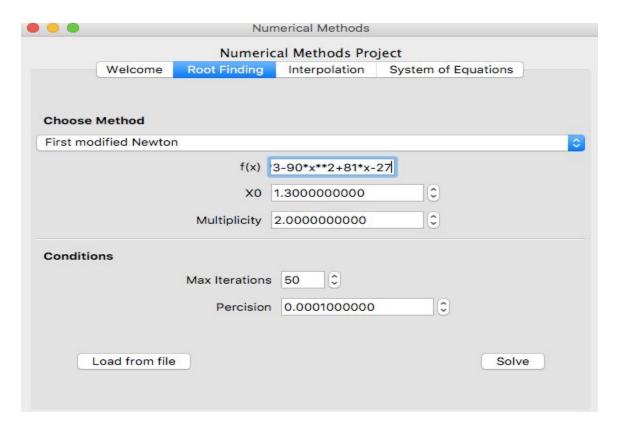
#### 1.4.1. Newton Raphson

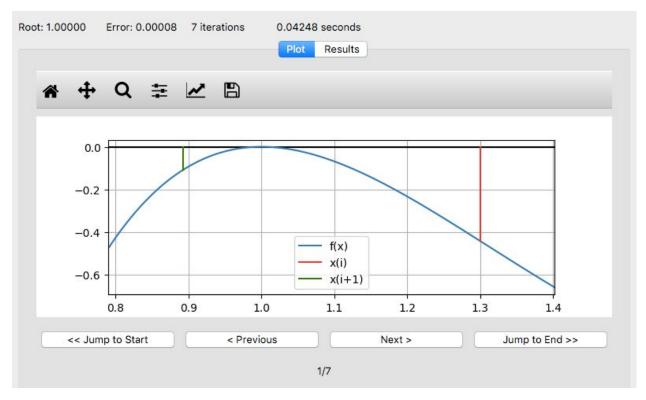


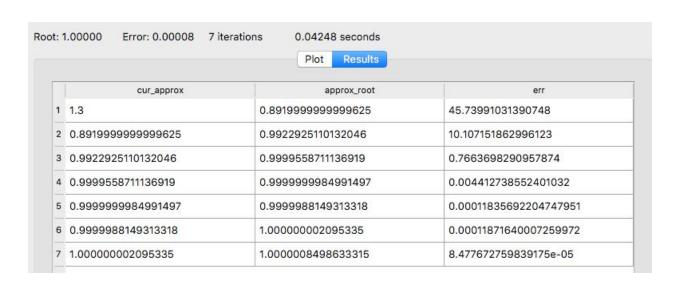
	cur_approx	approx_root	err
1	0.05	0.062422222222221	19.90032039871839
2	0.062422222222221	0.062377576543465846	0.07157328198740867
3	0.062377576543465846	0.06237758151374945	7.96806076917568e-06



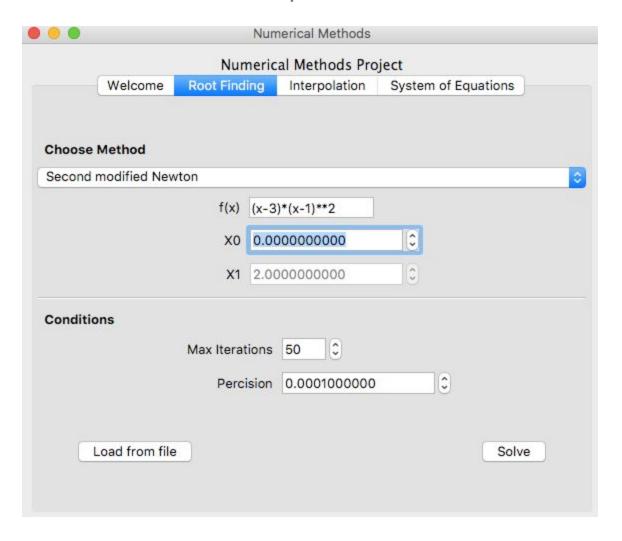
#### 1.4.2. First modified Newton Raphson



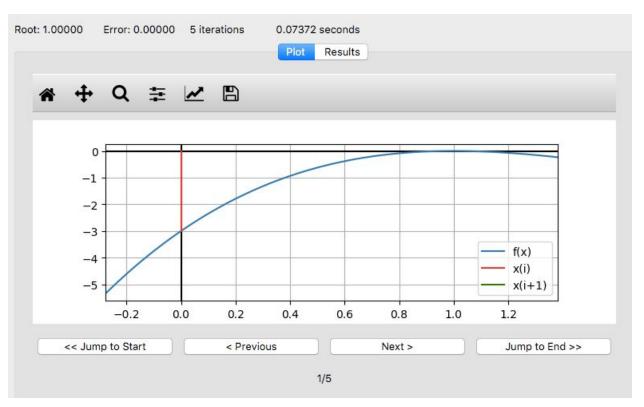




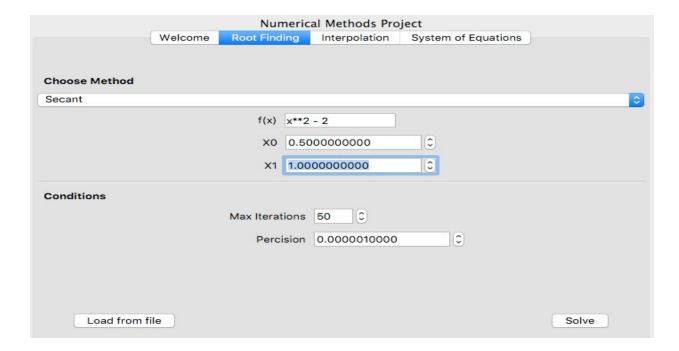
#### 1.4.3. Second modified Newton Raphson

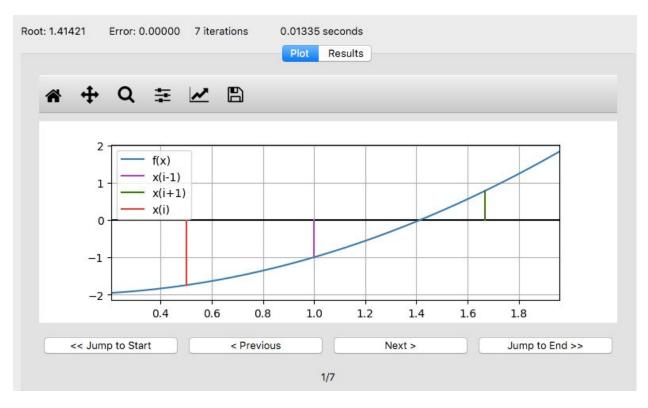


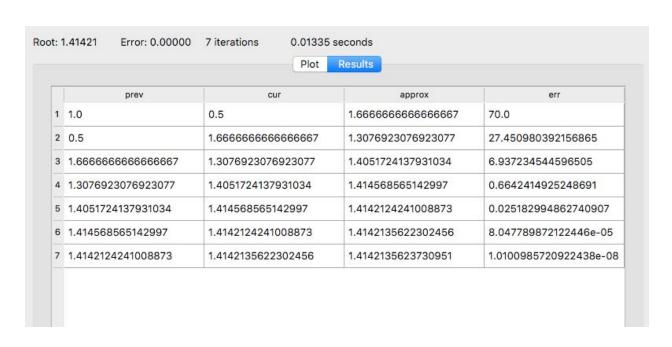
		Plot Results	
	cur_approx	approx_root	err
1	0.0	1.105263157894737	100.0
2	1.105263157894737	1.0030816640986133	10.186757215619698
3	1.0030816640986133	1.0000023814938872	0.3079275271450883
4	1.0000023814938872	1.00000000001418	0.00023814924692244316
5	1.000000000001418	1.0	1.41797684705125e-10



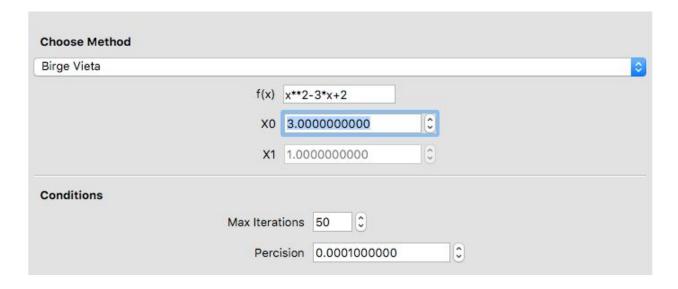
#### 1.5. Secant method

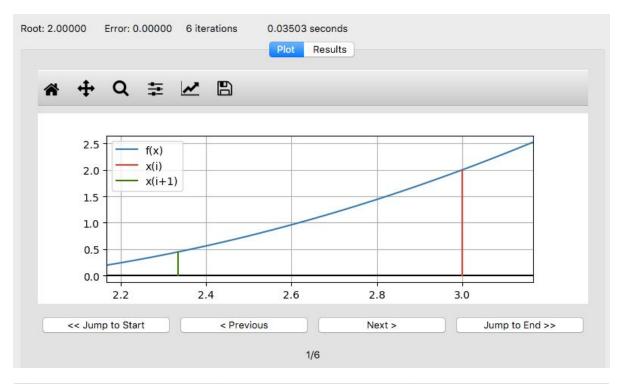






#### 1.6. Birge Vieta





cur_approx	approx_root	err
1 3.0	2.333333333333333	28.571428571428566
2 2.333333333333333	2.06666666666666	12.903225806451657
3 2.0666666666666	2.00392156862745	3.1311154598825954
4 2.00392156862745	2.0000152590218967	0.19531399012744305
5 2.0000152590218967	2.0000000002328306	0.0007629394532138178
6 2.0000000002328306	2.0	1.1641532182693481e-08

#### **Problematic functions**

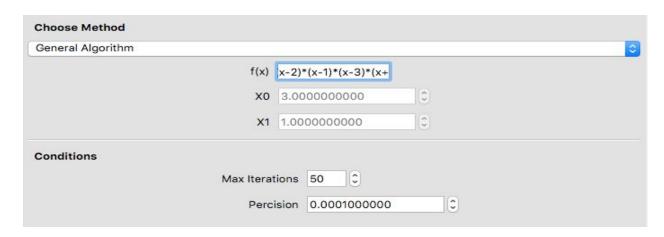
- In Newton's method the function with first derivative equal to zero case divide by the zero problem, we try to move the current approximation using epsilon to escape the the first derivative.
- In Newton's method the existence of deflection point cause the method to jump big interval.
- Open method that not guaranteed to convert set to have maximum iteration bound.

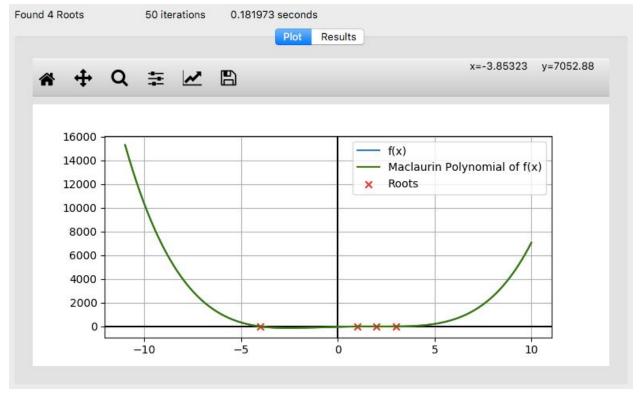
#### 1.2. General Algorithm:

In this part we aimed to evaluate the most number of roots from any given equation. We used a technique of approximating the function given into a tenth degree Maclaurin series and then iteratively using Newton-Raphson method to find up to ten roots for the equation.

#### Sample runs for general algorithm:

$$f(x) = (x - 1)*(x - 2)*(x - 3)*(x + 4)$$





		Roots found	
1	1.0		
2	2.0 3.0 -4.0		
3	3.0		
4	-4.0		

# **Part 2: Interpolation:**

#### **Objective:**

Implementing the 2 interpolation method Lagrange & Newton divided difference.

#### **Specifications:**

- Generate a function passes through all given points.
- Reading points from file or entering them manually.
- Estimate the value for a given queries points.

#### **Algorithms:**

#### Lagrange:

#### **Newton divided differences:**

```
x = symbols('x')
points x = [v['x'] for v in self.points]
points y = [v['y'] for v in self.points]
coff = []
coff.append(points y[0])
divid diff = [v for v in points y]
next_diff = []
for i in range(l_self.points.__len_()):
    for j in range(l_divid_diff.__len__()):
       value = divid_diff[j] - divid diff[j-1]
       value /= (points_x[j+i-1]-points_x[j-1])
       next diff.append(value)
    divid_diff = [v for v in next_diff]
    coff.append(next diff[0])
    next diff.clear()
for i in range (points x. len ()):
    temp = 1
    for j in range(i):
       temp *= (x - points x[j])
    fun += temp*coff[i]
return fun
```

- Implementing the 2 algorithms in python Used built in Library:
  - SymPy for manipulating symbols operations.

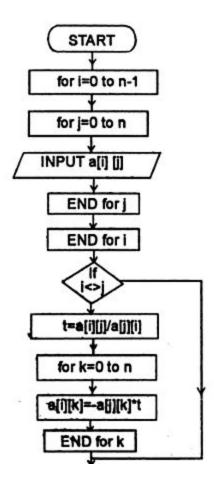
# **Analysis:**

Both Lagrange & Newton divided difference require  $O(n^2)$ 

# Part 3: Systems of linear equations:

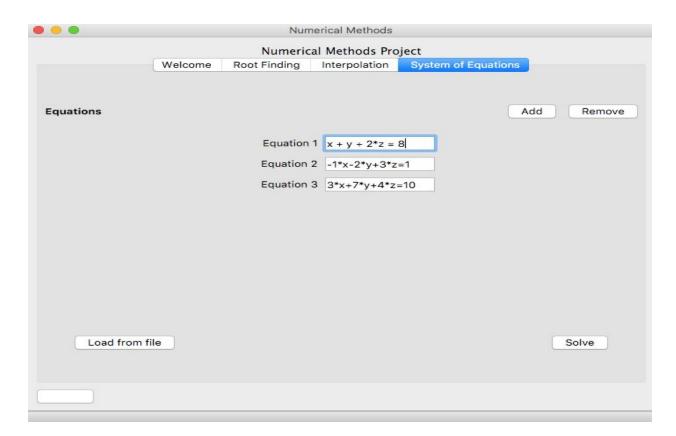
A solver for a system of equations using Gauss-Jordan algorithm

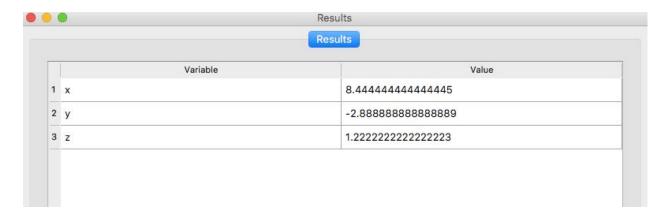
• Gauss-Jordan Algorithm



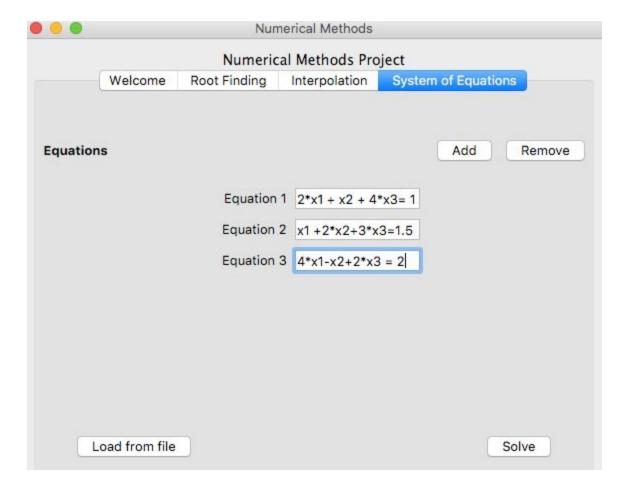
# • Sample runs:

#### Run #1





#### Run #2



		Results	
	Variable	Value	
1 x1		1.0	
2 x2		1.0	
3 x3		-0.5	