

Inna Williams

#####

1.

(25 pts) Consider the function $f(x) = 3x^5 + 2x - 1$ and the equation $f(x) = 0$.

#####

(a)

Show (by hand) that there is a unique simple root r in the interval $[0; 1]$.

$$f(x) = 3x^5 + 2x - 1$$

$$f(0) = 0 + 0 - 1 = -1 < 0$$

$$f(1) = 3 + 2 - 1 = 4 > 0$$

On interval $[0,1]$ function changing value from negative to positive.

According intermediate Value Theorem the root will be somewhere in between 0 and 1

```
> bisect:=proc(A,B,TOL,N)
    local i,a,b;a:=A;b:=B;

>
    i:=0;
    while i<=N do
        c:=(a+b)/2:
        if f(c)=0 or (b-a)/2<TOL then
            printf("c=%.8f, f(c)=%.8f\n",c,f(c));
            printf("Number of iterations needed: %d",i);

            return();

            break;
        end if;
        if signum(f(a))*signum(f(c))<0 then
            b:=c;
        else
            a:=c;
        end if;
        i:=i+1;
    end do;

>
    printf("The method failed after %d iterations.\n",N);
    printf("c=%.8f, f(c)=%.8f\n",c,f(c));
    return();
end proc;
```

(b)

Use the bisection procedure (Maple, Matlab, Python) to approximate r to eight

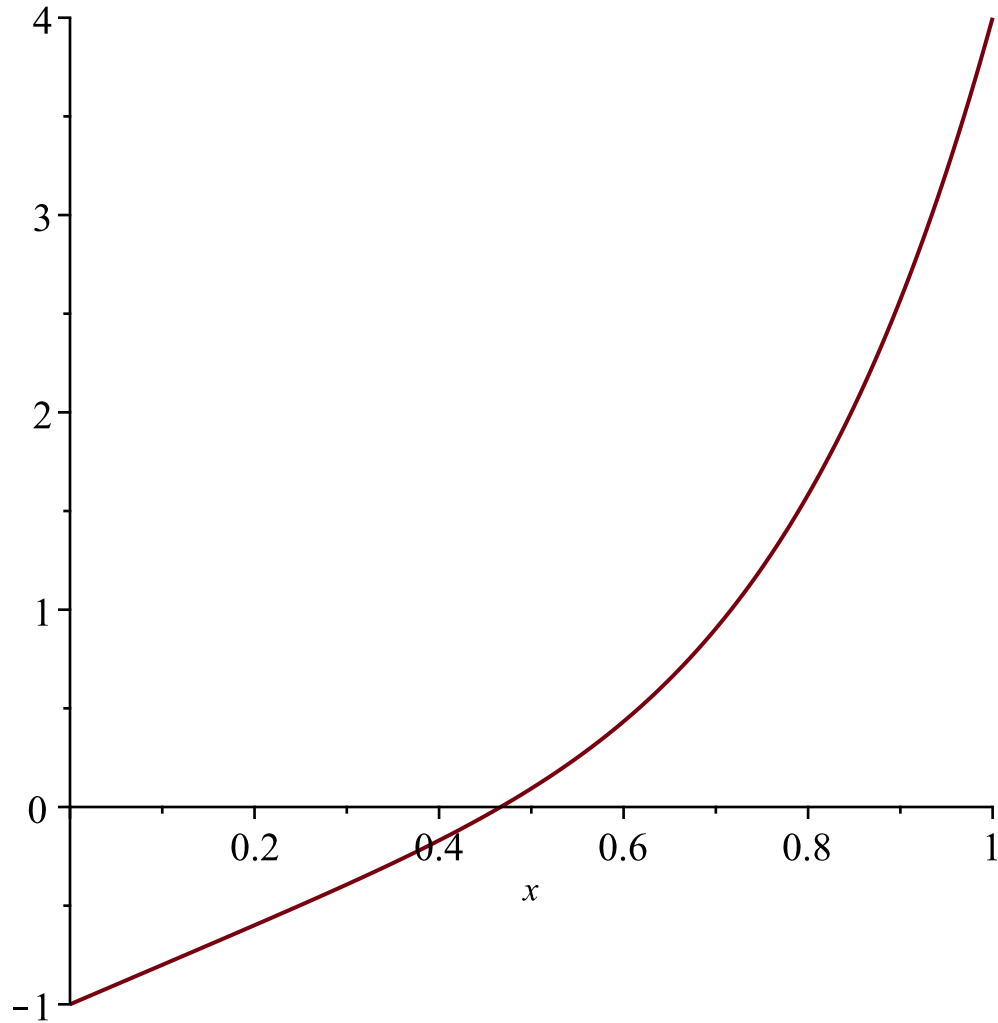
correct decimal places. Report the approximation xc, number of steps needed, and the backward error |f(xc)|

> f := x → 3 · x⁵ + 2 · x − 1;

$$f := x \mapsto 3x^5 + 2x - 1$$

(1)

> plot(f(x), x = 0..1);



> bisect(0, 0.5, 0.5 · 10⁻⁸, 1000);
c = 0.46676574, f(c) = 0.00000000
Number of iterations needed: 26

> BackwardError := abs(f(0.46676574))

$$\text{BackwardError} := 1.6 \cdot 10^{-8}$$

(2)

>

**Answer: Xc = 0.46676574 approximate to 8 decimal places
found in 26 iterations on interval [0.0, 1.0] with Bisection Method**

$$\text{BackwardError} := 1.6 \cdot 10^{-8}$$

```

[>
fixedpoint:=proc(g,X0,TOL,N)
    local i,x0;x0:=X0;
    i:=1;
    while i<=N do
        r:=g(x0);
        printf("Iteration %d: %.8g\n",i,r);
        if abs(r-x0)<TOL then
            printf("Fixed point r=%.8g, g(r)=%.8g\n",r,g(r));
            printf("Number of iterations needed: %d",i);return
        ();
        break;
        end if;
        i:=i+1;x0:=r;
    end do;
    printf("The method failed after %d iterations.\n",N);
    printf("r=%.8g, g(r)=%.8g\n",r,g(r));
    return();
end proc:

```

```

[>
*****

```

(c)

Build a fixed point iteration procedure $g(x) = x$ (not Newton's) to find the root of $f(x)$, starting from $x_0 = 0.5$. Report the approximation x_c , number of steps needed, and the backward error $|f(x_c)|$

$$3 \cdot x^5 + 2 \cdot x - 1 = 0$$

$$2 \cdot x = 1 - 3 \cdot x^5 \Rightarrow x = \frac{(1 - 3 \cdot x^5)}{2} = g(x) \rightarrow FPI$$

```

*****

```

```
> g(x) := x ->  $\frac{(1 - 3 \cdot x^5)}{2}$ 
```

$$g(x) := x \mapsto \frac{1}{2} - \frac{3x^5}{2} \quad (3)$$

```
> x0 := 0.5
```

$$x_0 := 0.5 \quad (4)$$

```
> fixedpoint(g(x),x0,0.5*10-8,50)
```

```

Iteration 1: 0.453125
Iteration 2: 0.47134626
Iteration 3: 0.46510272
Iteration 4: 0.46735357
Iteration 5: 0.46655593
Iteration 6: 0.46684036
Iteration 7: 0.46673916
Iteration 8: 0.46677519
Iteration 9: 0.46676237
Iteration 10: 0.46676693
Iteration 11: 0.46676531
Iteration 12: 0.46676589

```

```

Iteration 13: 0.46676568
Iteration 14: 0.46676575
Iteration 15: 0.46676573
Iteration 16: 0.46676574
Iteration 17: 0.46676573
Fixed point r=0.46676573, g(r)=0.46676573
Number of iterations needed: 17

```

```
> BackwardError := abs(f(0.46676573))
```

BackwardError := 1.10 10⁻⁸

(5)

```
*****
```

**Answer: $X_c = 0.46676573$ approximate to 8 decimal places
found in 17 iterations with initial guess $x_0=0.5$ with FPI Method**

BackwardError := $1.10 \cdot 10^{-8}$

```
*****
```

BackwardError := 1.100000000 10⁻⁸

(6)

```

newton:=proc (f,X0,TOL,N)
    local i,x0;x0:=X0;
    i:=1;
    while i<=N do
        if D(f)(p0)=0 then
            printf("Division by 0. Method failed");break;
        else
            r:=x0-f(x0)/D(f)(x0);
            printf("Iteration %d: %.8g\n",i,r);
            if abs(r-x0)<TOL then
                printf("r=%.8g, f(r)=%.8g\n",r,f(r));
                printf("Number of iterations needed: %d",i);
            return();
            break;
        end if;
        i:=i+1;x0:=r;
    end if;
    end do;
    printf("The method failed after %d iterations.\n",N);
    printf("r=%.8g, f(r)=%.8g\n",r,f(r));
    return();
end proc:

```

```
>
```

```
*****
```

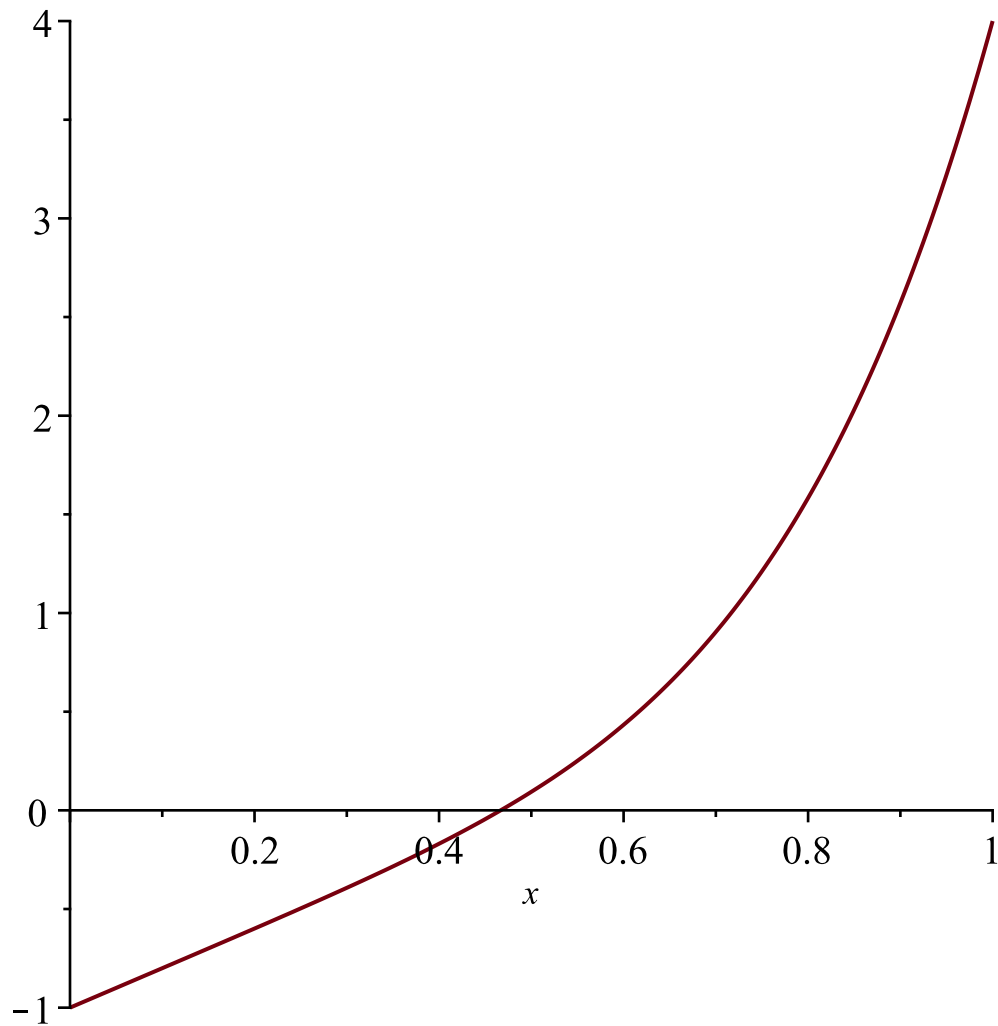
(d)

Use Newton's method to approximate r to eight correct decimal places (starting with $x_0 = 0.5$). Report the approximation x_c , number of steps needed, and the backward error $|f(x_c)|$

```
*****
```

```
> f(x);plot(f(x),x=0..1);
```

$3x^5 + 2x - 1$



```
> newton(f, 0.5, 0.5·10-8, 50)
Iteration 1: 0.46808511
Iteration 2: 0.4667677
Iteration 3: 0.46676573
Iteration 4: 0.46676573
r=0.46676573, f(r)=-1e-10
Number of iterations needed: 4
```

```
> BackwardError := abs(f(0.46676573))
```

BackwardError := 1.10 10⁻⁸

(7)

Answer: $X_c = 0.46676573$ approximate to 8 decimal places

found in 4 iterations with initial guess $x_0=0.5$ with Newton Method

BackwardError := 1.10 · 10⁻⁸

BackwardError := 1.100000000 10⁻⁸

(8)

```
Halley:=proc(f,X0,TOL,N)
  local i,x0;x0:=X0;
  i:=1;
  while i<=N do
    if D(f)(p0)=0 then
```

```

        printf("Division by 0. Method failed");break;
    else
        r:=x0-( (2*f(x0)*(D(f))(x0) )/(2*((D(f))(x0))^2-f
(x0)*(D(D(f)))(x0)));
        printf("Iteration %d: %.8g\n",i,r);
        if abs(r-x0)<TOL then
            printf("r=%.8g, f(r)=%.8g\n",r,f(r));
            printf("Number of iterations needed: %d",i);
return();
        break;
    end if;
    i:=i+1;x0:=r;
end if;
end do;
printf("The method failed after %d iterations.\n",N);
printf("r=%.8g, f(r)=%.8g\n",r,f(r));
return();
end proc:

```

>

(e)

Modify the code on Newton's method in order to solve the equation using Halley's method, an iterative process given by

$$X_{k+1} = H(X) = X_k - \frac{2 \cdot f(X_k) \cdot f'(X_k)}{2 \cdot (f'(X_k))^2 - f(X_k) \cdot f''(X_k)}$$

Approximate r to eight correct decimal places (starting with $x_0 = 0.5$). Report the approximation x_c , number of steps needed, and the backward error $|f(x_c)|$

false

(9)

> Halley(f, x0, 0.5·10⁻⁸, 50)

Iteration 1: 0.46672959

Iteration 2: 0.46676573

Iteration 3: 0.46676573

r=0.46676573, f(r)=0

Number of iterations needed: 3

> BackwardError := abs(f(0.46676573))

BackwardError := 1.10 10⁻⁸

(10)

Answer: $X_c = 0.46676573$ approximate to 8 correct decimal places found in 3 iterations with initial guess $x_0=0.5$ with Halley's Method

BackwardError := 1.10 · 10⁻⁸

BackwardError := 1.100000000 10⁻⁸

(11)

#####

2.

(25 pts) The total cost of a federal government program in millions of dollars is shown in the table below. (Notice that the year 2010 data is missing.)

(c)

Perform a least-squares analysis and estimate the 2010 cost and compare to the actual value of \$1,049 (in mil). (Choose an appropriate least squares linear or polynomial model.)

Linear

model : $y = a + b(x - 2005)$

Polynomial 10th degree

model: $y =$

$a + b \cdot (x - 2005) + c \cdot (x - 2005)^2 + d \cdot (x - 2005)^3 + e \cdot (x - 2005)^4 + f \cdot (x - 2005)^5 + g \cdot (x - 2005)^6 + h \cdot (x - 2005)^7 + i \cdot (x - 2005)^8 + j \cdot (x - 2005)^9 + k \cdot (x - 2005)^{10}$

> restart; with(plots) : with(CurveFitting) : with(LinearAlgebra) : actual_cost_2010 := 1049;
actual_cost_2010 := 1049 (12)

> xvalues := ⟨0, 1, 2, 3, 4, 6, 7, 8, 9, 10, 11⟩ : yvalues := ⟨731, 782, 833, 886, 956, 1159, 1267, 1367, 1436, 1505, 1562⟩ :

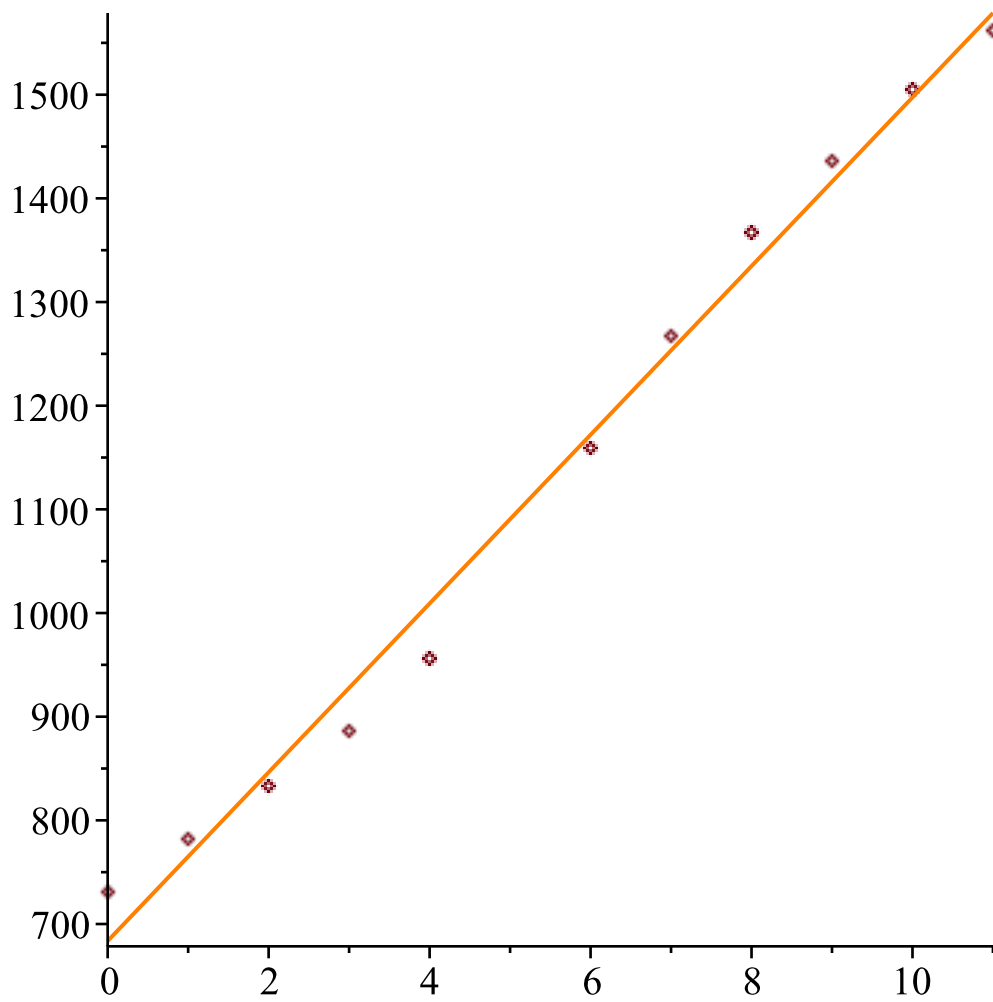
p1 := plot(xvalues, yvalues, style = point) :

> ls_line := CurveFitting[LeastSquares](xvalues, yvalues, x)

ls_line := $\frac{1073259}{1570} + \frac{127771}{1570}x$ (13)

>

> p2_ls_line := plot(ls_line, x = 0 .. 11, color = coral) : display(p1, p2_ls_line)



```

=>
=>
=>
=>

```

```

> interface(rtablesize = 12); yout := map(
    x →  $\frac{1073259}{1570} + \frac{127771}{1570}x$ , xvalues)
    10

```


$$y_{out} := \begin{bmatrix} \frac{1073259}{1570} \\ \frac{120103}{157} \\ \frac{1328801}{1570} \\ \frac{728286}{785} \\ \frac{1584343}{1570} \\ \frac{367977}{314} \\ \frac{983828}{785} \\ \frac{2095427}{1570} \\ \frac{1111599}{785} \\ \frac{2350969}{1570} \\ \frac{247874}{157} \end{bmatrix} \quad (14)$$

```
>
> Calculated_LeastSquare_Line_2010 := evalf(subs(x=2010 - 2005, ls_line))
    Calculated_LeastSquare_Line_2010 := 1090.518471 (15)
```

```
> absolute_error := abs(actual_cost_2010 -
    Calculated_LeastSquare_Line_2010)
    absolute_error := 41.518471 (16)
```

```
> relative_error := 100 * (
    absolute_error / actual_cost_2010 )
    relative_error := 3.957909533 (17)
```

```
> RMSE_ls_line := evalf(
    Norm(yvalues - yout, 2) /
    sqrt(Dimension(xvalues)) )
    RMSE_ls_line := 29.26255267 (18)
```

Answer:

Calculated_LeastSquare_Line_2010 := 1090.518

absolute_error := 41.518

relative_error := 3.958 %

RMSE_ls_line := 29.263

```
>
> unassign('a','b','c','x') : interface(rtablesiz=12);
12 (19)
```

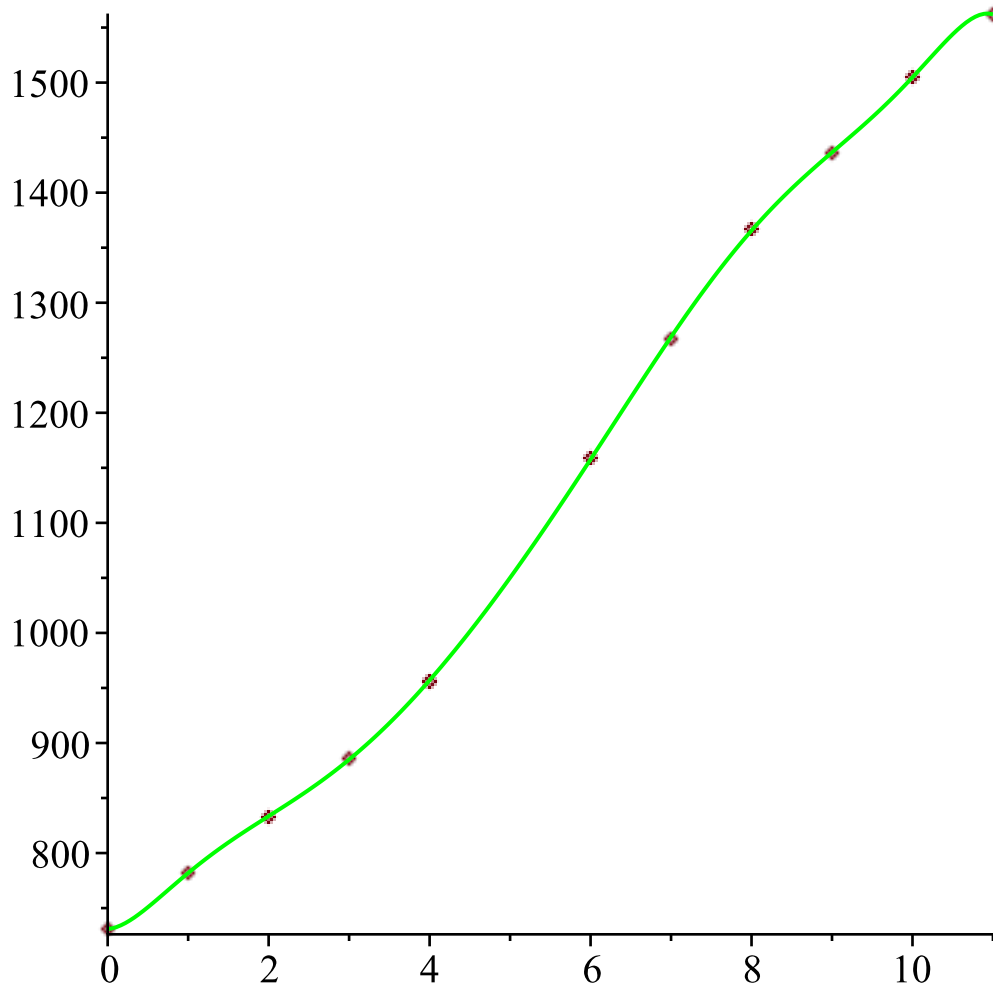
```
>
ls_poly := CurveFitting[LeastSquares](xvalues, yvalues, x, curve = a + b·x + c·x2 + d·x3 + e
·x4 + f·x5 + g·x6 + h·x7 + i·x8 + j·x9 + j·x10)
```

```
ls_poly := 
$$\frac{168887723756}{231031411} + \frac{36482432347}{6930942330} x + \frac{30553404678421}{291099577860} x^2$$


$$- \frac{62620127019091}{698638986864} x^3 + \frac{10259709153157}{268707302640} x^4 - \frac{362768476349}{39808489280} x^5 + \frac{10204101847}{7961697856} x^6$$


$$- \frac{3666157279}{35827640352} x^7 + \frac{40115941}{10537541280} x^8 - \frac{170963}{63225247680} x^{10} - \frac{170963}{63225247680} x^9$$
 (20)
```

```
>
> p3_ls_poly := plot(ls_poly, x=0..11, color=green) :
> display(p1, p3_ls_poly)
```



```
> yout1 := map(x -> 168887723756/231031411 + 36482432347/6930942330 x + 30553404678421/291099577860 x^2
- 62620127019091/698638986864 x^3 + 10259709153157/268707302640 x^4 - 362768476349/39808489280 x^5 + 10204101847/7961697856 x^6
- 3666157279/35827640352 x^7 + 40115941/10537541280 x^8 - 170963/63225247680 x^10 - 170963/63225247680 x^9,
xvalues)
```

$$yout1 := \begin{bmatrix} \frac{168887723756}{231031411} \\ \frac{180633989036}{231031411} \\ \frac{192569316713}{231031411} \\ \frac{204457532491}{231031411} \\ \frac{221098321526}{231031411} \\ \frac{267451409821}{231031411} \\ \frac{293157352687}{231031411} \\ \frac{315495530192}{231031411} \\ \frac{331902617786}{231031411} \\ \frac{26743658705}{17771647} \\ \frac{21227928103}{13590083} \end{bmatrix} \quad (21)$$

```
> Calculated_LeastSquare_Poly_2010 := evalf(subs(x=2010 — 2005, ls_poly))
Calculated_LeastSquare_Poly_2010 := 1050.122300 (22)
```

```
> absolute_error := abs(actual_cost_2010
— Calculated_LeastSquare_Poly_2010)
absolute_error := 1.122300 (23)
```

```
> relative_error := 100*(absolute_error/actual_cost_2010)
relative_error := 0.1069876072 (24)
```

```
> RMSE := evalf( Norm(yvalues — yout1, 2) )
√ Dimension(xvalues)
RMSE := 0.9629895927 (25)
```

Answer:

Calculated Least Square Polynomial 10th degree:

Calculated_LeastSquare_Poly_2010 := 1050.122

absolute_error := 1.122

relative_error := 0.107 %

RMSE := 0.963

(a)

Determine and plot the degree 10 Lagrange polynomial through the data. Use it to estimate the 2010 cost and compare to the actual value of \$1,049 (in mil.).

Lagrange Polinomial:

$$\begin{aligned} f := & x \\ & \rightarrow (731 \cdot ((x-1) \cdot (x-2) \cdot (x-3) \cdot (x-4) \cdot (x-6) \cdot (x-7) \cdot (x-8) \cdot (x-9) \cdot (x-10) \cdot (x-11))) / ((0-1) \cdot (0-2) \cdot (0-3) \cdot (0-4) \cdot (0-6) \cdot (0-7) \cdot (0-8) \cdot (0-9) \cdot (0-10) \cdot (0-11)) \\ & + (782 \cdot ((x-0) \cdot (x-2) \cdot (x-3) \cdot (x-4) \cdot (x-6) \cdot (x-7) \cdot (x-8) \cdot (x-9) \cdot (x-10) \cdot (x-11))) / ((1-0) \cdot (1-2) \cdot (1-3) \cdot (1-4) \cdot (1-6) \cdot (1-7) \cdot (1-8) \cdot (1-9) \cdot (1-10) \cdot (1-11)) \\ & + (833 \cdot ((x-0) \cdot (x-1) \cdot (x-3) \cdot (x-4) \cdot (x-6) \cdot (x-7) \cdot (x-8) \cdot (x-9) \cdot (x-10) \cdot (x-11))) / ((2-0) \cdot (2-1) \cdot (2-3) \cdot (2-4) \cdot (2-6) \cdot (2-7) \cdot (2-8) \cdot (2-9) \cdot (2-10) \cdot (2-11)) \\ & + (886 \cdot ((x-0) \cdot (x-1) \cdot (x-2) \cdot (x-4) \cdot (x-6) \cdot (x-7) \cdot (x-8) \cdot (x-9) \cdot (x-10) \cdot (x-11))) / ((3-0) \cdot (3-1) \cdot (3-2) \cdot (3-4) \cdot (3-6) \cdot (3-7) \cdot (3-8) \cdot (3-9) \cdot (3-10) \cdot (3-11)) \\ & + (956 \cdot ((x-0) \cdot (x-1) \cdot (x-2) \cdot (x-3) \cdot (x-6) \cdot (x-7) \cdot (x-8) \cdot (x-9) \cdot (x-10) \cdot (x-11))) / ((4-0) \cdot (4-1) \cdot (4-2) \cdot (4-3) \cdot (4-6) \cdot (4-7) \cdot (4-8) \cdot (4-9) \cdot (4-10) \cdot (4-11)) \\ & + (1159 \cdot ((x-0) \cdot (x-1) \cdot (x-2) \cdot (x-3) \cdot (x-4) \cdot (x-7) \cdot (x-8) \cdot (x-9) \cdot (x-10) \cdot (x-11))) / ((6-0) \cdot (6-1) \cdot (6-2) \cdot (6-3) \cdot (6-4) \cdot (6-7) \cdot (6-8) \cdot (6-9) \cdot (6-10) \cdot (6-11)) \\ & + (1267 \cdot ((x-1) \cdot (x-2) \cdot (x-3) \cdot (x-4) \cdot (x-6) \cdot (x-7) \cdot (x-8) \cdot (x-9) \cdot (x-10) \cdot (x-11))) / ((7-0) \cdot (7-1) \cdot (7-2) \cdot (7-3) \cdot (7-4) \cdot (7-6) \cdot (7-8) \cdot (7-9) \cdot (7-10) \cdot (7-11)) \\ & + (1367 \cdot ((x-0) \cdot (x-1) \cdot (x-2) \cdot (x-3) \cdot (x-4) \cdot (x-6) \cdot (x-7) \cdot (x-9) \cdot (x-10) \cdot (x-11))) / ((8-0) \cdot (8-1) \cdot (8-2) \cdot (8-3) \cdot (8-4) \cdot (8-6) \cdot (8-7) \cdot (8-9) \cdot (8-10) \cdot (8-11)) \\ & + (1436 \cdot ((x-0) \cdot (x-1) \cdot (x-2) \cdot (x-3) \cdot (x-4) \cdot (x-6) \cdot (x-7) \cdot (x-8) \cdot (x-10) \cdot (x-11))) / ((9-0) \cdot (9-1) \cdot (9-2) \cdot (9-3) \cdot (9-4) \cdot (9-6) \cdot (9-7) \cdot (9-8) \cdot (9-10) \cdot (9-11)) \\ & + (1505 \cdot ((x-0) \cdot (x-1) \cdot (x-2) \cdot (x-3) \cdot (x-4) \cdot (x-6) \cdot (x-7) \cdot (x-8) \cdot (x-10) \cdot (x-11))) / ((10-0) \cdot (10-1) \cdot (10-2) \cdot (10-3) \cdot (10-4) \cdot (10-6) \cdot (10-7) \cdot (10-8) \cdot (10-9) \cdot (10-11)) \\ & + (1562 \cdot ((x-0) \cdot (x-1) \cdot (x-2) \cdot (x-3) \cdot (x-4) \cdot (x-6) \cdot (x-7) \cdot (x-8) \cdot (x-9) \cdot (x-10))) / ((11-0) \cdot (11-1) \cdot (11-2) \cdot (11-3) \cdot (11-4) \cdot (11-6) \cdot (11-7) \cdot (11-8) \cdot (11-9) \cdot (11-10)) \end{aligned}$$

> with(plots):with(CurveFitting);with(LinearAlgebra);

with(ArrayTools) : interface(rtablesiz = 15);

[ArrayInterpolation, BSpline, BSplineCurve, Interactive, LeastSquares, Lowess, PolynomialInterpolation, RationalInterpolation, Spline, ThieleInterpolation]

Typesetting:—mparsed(with(LinearAlgebra); with(ArrayTools) : interface(rtablesiz = 15); , (26)
with(LinearAlgebra);

**with(ArrayTools) :
interface(rtablesiz = 15))**



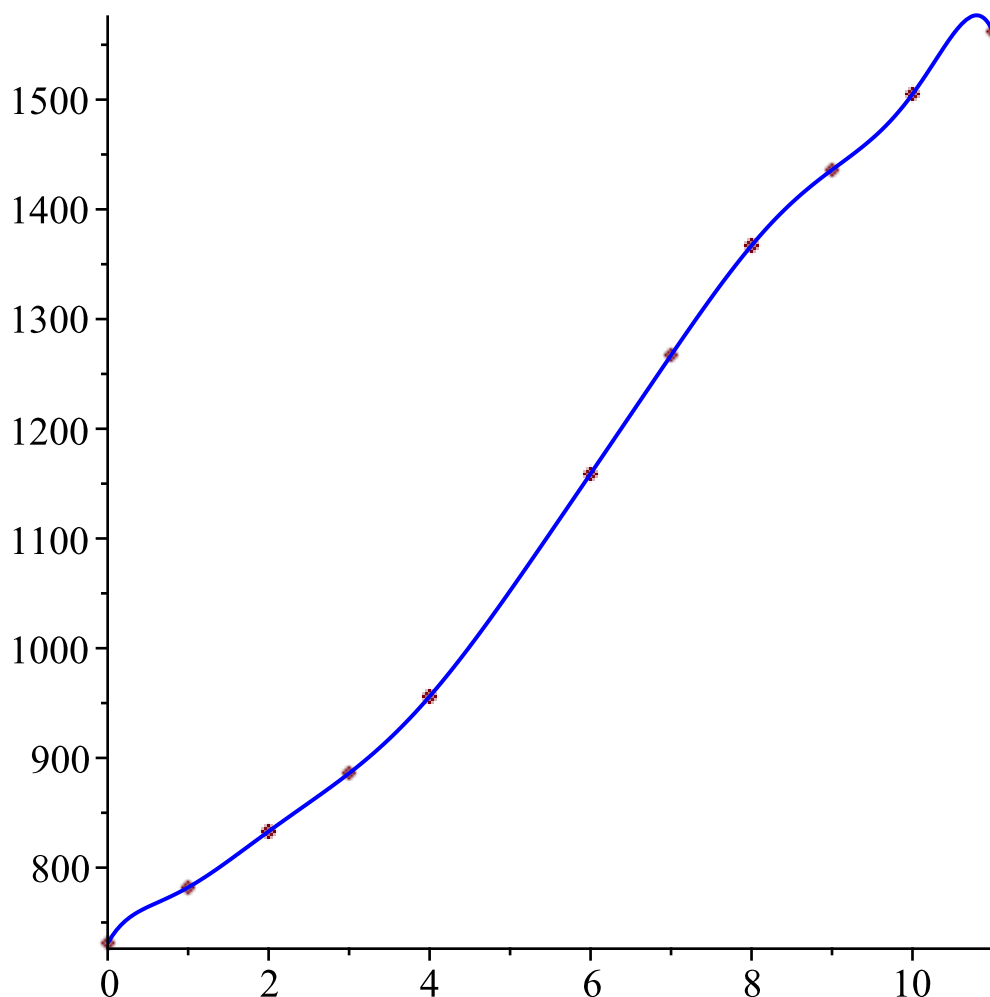
```
> data1 := [[0, 731], [1, 782], [2, 833], [3, 886], [4, 956], [6, 1159], [7, 1267], [8, 1367], [9, 1436], [10, 1505], [11, 1562]]:
```

```
> poly := PolynomialInterpolation(data1, x);
```

$$\text{poly} := -\frac{599}{6652800}x^{10} + \frac{6353}{1330560}x^9 - \frac{299}{2772}x^8 + \frac{300259}{221760}x^7 - \frac{363303}{35200}x^6 + \frac{68993}{1408}x^5 - \frac{19032107}{133056}x^4 + \frac{81659899}{332640}x^3 - \frac{61824641}{277200}x^2 + \frac{152056}{1155}x + 731 \quad (27)$$

```
> p2Lagrange := plot(poly, x=0..11, color=blue) :
```

```
> display(p1, p2Lagrange)
```



```
> yout := map(x → -\frac{599}{6652800}x^{10} + \frac{6353}{1330560}x^9 - \frac{299}{2772}x^8 + \frac{300259}{221760}x^7 - \frac{363303}{35200}x^6 + \frac{68993}{1408}x^5 - \frac{19032107}{133056}x^4 + \frac{81659899}{332640}x^3 - \frac{61824641}{277200}x^2 + \frac{152056}{1155}x + 731, xvalues)
```

$$y_{out} := \begin{bmatrix} 731 \\ 782 \\ 833 \\ 886 \\ 956 \\ 1159 \\ 1267 \\ 1367 \\ 1436 \\ 1505 \\ 1562 \end{bmatrix} \quad (28)$$

$$> \text{calculated_2010_polynomial} := \text{evalf}(\text{subs}(x = 2010 - 2005, \text{poly}))$$

$$\text{calculated_2010_polynomial} := 1052.272727 \quad (29)$$

$$> \text{absolute_error} := \text{abs}(\text{actual_cost_2010} - \text{calculated_2010_polynomial})$$

$$\text{absolute_error} := 3.272727 \quad (30)$$

$$> \text{relative_error} := 100 \cdot \left(\frac{\text{absolute_error}}{\text{actual_cost_2010}} \right)$$

$$\text{relative_error} := 0.3119854147 \quad (31)$$

$$> \text{RMS_lagrange_poly} := \text{evalf} \left(\frac{\text{Norm}(y_{\text{values}} - y_{\text{out}}, 2)}{\sqrt{\text{Dimension}(x_{\text{values}})}} \right)$$

$$\text{RMS_lagrange_poly} := 0. \quad (32)$$

>

Answer: Lagrange 10th Degree Polynomial

calculated_2010_polynomial := 1052.273

absolute_error := 3.273

relative_error := 0.312%

(b)

Determine and plot a clamped cubic spline for the data (using the appropriate slopes at 2005 and 2016). Use it to estimate the 2010 cost and compare to the actual value of \$1,049 (in mil.)

$$> \text{slope_0} := \frac{(\text{data1}[2][2] - \text{data1}[1][2])}{\text{data1}[2][1] - \text{data1}[1][1]}$$

$$\text{slope_0} := 51 \quad (33)$$

$$> \text{slope_1} := \frac{(\text{data1}[11][2] - \text{data1}[10][2])}{\text{data1}[11][1] - \text{data1}[10][1]}$$

$$\text{slope_1} := 57 \quad (34)$$

```

> p1_clamped_spline := plot(data1, style=point, color=black) :
> clamped_spline := evalf(Spline(data1, x, endpoints=[slope_0, slope_1]))
clamped_spline :=

```

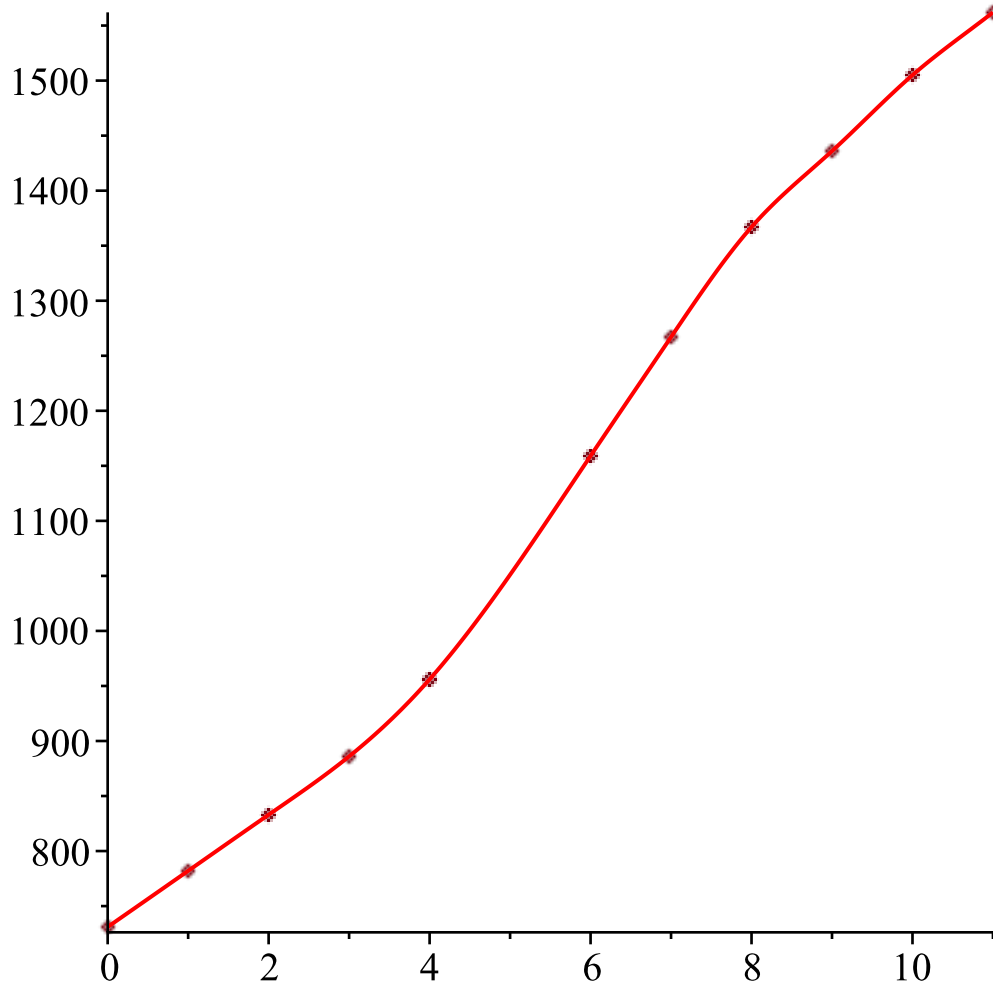
$0.1255477457 x^3 - 0.1255477457 x^2 + 51. x + 731.$	$x < 1.$
$-0.3766432370 x^3 + 1.381025202 x^2 + 49.49342705 x + 731.5021910$	$x < 2.$
$3.381025202 x^3 - 21.16498543 x^2 + 94.58544832 x + 701.4408435$	$x < 3.$
$1.852542428 x^3 - 7.408640466 x^2 + 53.31641342 x + 742.7098784$	$x < 4.$
$-2.778265896 x^3 + 48.16105943 x^2 - 168.9623862 x + 1039.081611$	$x < 6.$
$0.9301165366 x^3 - 18.58982437 x^2 + 231.5429166 x + 238.0710056$	$x < 7.$
$-8.955129268 x^3 + 189.0003375 x^2 - 1221.588217 x + 3628.710317$	$x < 8.$
$11.89040054 x^3 - 311.2923777 x^2 + 2780.753506 x - 7044.200943$	$x < 9.$
$-7.606472873 x^3 + 215.1232043 x^2 - 1956.986733 x + 7169.019772$	$x < 10.$
$6.535490958 x^3 - 209.1357106 x^2 + 2285.602417 x - 6972.944059$	<i>otherwise</i>

```

> p2_clamped_spline := plot(clamped_spline, x=0..11, color=red) :
> display(p1, p2_clamped_spline)

```

(35)



$\triangleright f := x \rightarrow \text{piecewise}(x < 1, 0.1255477457 x^3 - 0.1255477457 x^2 + 51. x + 731., x < 2,$
 $-0.3766432370 x^3 + 1.381025202 x^2 + 49.49342705 x + 731.5021910, x < 3,$
 $3.381025202 x^3 - 21.16498543 x^2 + 94.58544832 x + 701.4408435, x < 4, 1.852542428 x^3$
 $- 7.408640466 x^2 + 53.31641342 x + 742.7098784, x < 6, -2.778265896 x^3$
 $+ 48.16105943 x^2 - 168.9623862 x + 1039.081611, x < 7, 0.9301165366 x^3$
 $- 18.58982437 x^2 + 231.5429166 x + 238.0710056, x < 8, -8.955129268 x^3$
 $+ 189.0003375 x^2 - 1221.588217 x + 3628.710317, x < 9, 11.89040054 x^3$
 $- 311.2923777 x^2 + 2780.753506 x - 7044.200943, x < 10, -7.606472873 x^3$
 $+ 215.1232043 x^2 - 1956.986733 x + 7169.019772, x \leq 11, 6.535490958 x^3$
 $- 209.1357106 x^2 + 2285.602417 x - 6972.944059)$

$f := x$

(36)

$$\mapsto \left\{ \begin{array}{ll} 0.1255477457 x^3 - 0.1255477457 x^2 + 51. x + 731. & x < 1 \\ -0.3766432370 x^3 + 1.381025202 x^2 + 49.49342705 x + 731.5021910 & x < 2 \\ 3.381025202 x^3 - 21.16498543 x^2 + 94.58544832 x + 701.4408435 & x < 3 \\ 1.852542428 x^3 - 7.408640466 x^2 + 53.31641342 x + 742.7098784 & x < 4 \\ -2.778265896 x^3 + 48.16105943 x^2 - 168.9623862 x + 1039.081611 & x < 6 \\ 0.9301165366 x^3 - 18.58982437 x^2 + 231.5429166 x + 238.0710056 & x < 7 \\ -8.955129268 x^3 + 189.0003375 x^2 - 1221.588217 x + 3628.710317 & x < 8 \\ 11.89040054 x^3 - 311.2923777 x^2 + 2780.753506 x - 7044.200943 & x < 9 \\ -7.606472873 x^3 + 215.1232043 x^2 - 1956.986733 x + 7169.019772 & x < 10 \\ 6.535490958 x^3 - 209.1357106 x^2 + 2285.602417 x - 6972.944059 & x \leq 11 \end{array} \right.$$

> *yout* := map(*f*, *xvalues*)

$$yout := \begin{bmatrix} 731. \\ 782.0000000 \\ 833.0000000 \\ 886.0000001 \\ 955.9999998 \\ 1159.000000 \\ 1266.999997 \\ 1367.000017 \\ 1436.000002 \\ 1505.000011 \\ 1562.000011 \end{bmatrix} \quad (37)$$

> **calculated_clamped_spline_2010** := eval(**clamped_spline**, x = 2010 - 2005)

$$calculated_clamped_spline_2010 := 1051.012929 \quad (38)$$

> **absolute_error** := **abs**(**actual_cost_2010** - **calculated_clamped_spline_2010**)

$$absolute_error := 2.012929 \quad (39)$$

> **relative_error** := 100 · $\left(\frac{\mathbf{absolute_error}}{\mathbf{actual_cost_2010}} \right)$

$$relative_error := 0.1918902765 \quad (40)$$

>

```
> RMSE_clamed_spline := evalf(  $\frac{\text{Norm}(\text{yvalues} - \text{yout}, 2)}{\sqrt{\text{Dimension}(\text{xvalues})}}$  )
RMSE_clamed_spline := 7.032715800 10-6
```

(41)

```
=
>
>
```

Answer for Clamped spline:

calculated_clamped_spline_2010 := 1051.013

absolute_error := 2.0123

relative_error := 0.192 %

RMSE_clamed_spline := 7.033 · 10⁻⁶

(d)

On one plot show the Lagrange polynomial, the clamped cubic spline and the least-squares \014t. Which method gives a better estimate of the 2010 cost?

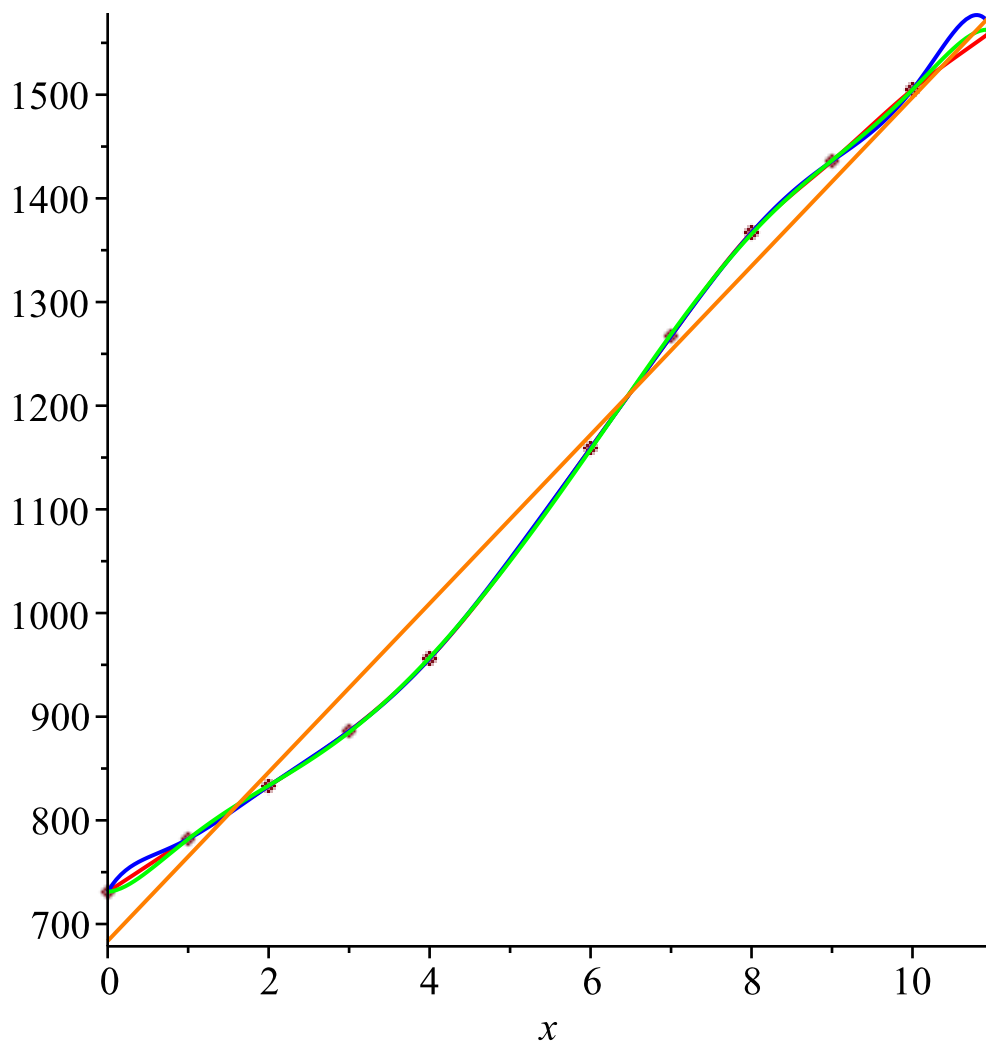
Clamped_spline -> red

Lagrange polynomial->blue

Least square polynomial->green

Least square line-> coral

```
> display(p1, p2_clamped_spline, p2Lagrange, p3_ls_poly, p2_ls_line)
```



On the graph we can see that

Clamped_spline -> red

Lagrange polynomial -> blue

Least square polynomial -> green

almost on the same line

Compare absolute errors:

Least Square Line : absolute_error := 41.52

Least square polynomial: absolute_error := 1.12

Lagrange polynomial: absolute_error := 3.27

Clamped_spline : absolute_error := 2.01

Compare relative errors:

Least Square Line : relative_error := 3.96 %

Least Square polynomial: relative_error := 0.11 %

Lagrange polynomial: relative_error := 0.31 %

Clamped_spline : relative_error := 0.19 %

Compare RMSE:

Least Square Line : RMSE := 29.26

Least Square polynomial: RMSE := 0.96

Lagrange polynomial: RMSE := 0.000000 *gives exact fit through the points*
 Clamped_spline : RMSE := 0.000007

**Answer: Using absolute or relative error for the fit
 we can determine that the better estimate for 2010 cost gives :**

Least Square Polynomial Model :

**with absolute_error := 1.12
 with relative_error := 0.11 %**

(e)

Which method would you choose to estimate the 2019 cost? Explain.

I would use Least Square polynomial to approximate 2019 cost.

There is no reason to think that the relations between the cost can be exactly expressed by a polynomial or spline.

But regression Linear Or Polynomial might do the trick, since the cost might have some slope and a

regression function might be a good approximation, at least locally.

Regression is definitely a better idea than interpolation in this case.

Between Linear or Polynomial I choose Least Square Polynomial regression line because it does have smaller RMSE.

**Least Square Line : RMSE := 29.26
 Least Square polynomial: RMSE := 0.96 < 29.26**

#####

3.

(15 pts) Integrals of functions like $f(x) = \exp(-x^2)$ are hard to evaluate by elementary integration techniques. Use the minimal linear congruential generator with seed $x_0 = 3$

and $n = 10000$ random values to approximate the integral $\int_0^2 \exp(-x^2) dx$

(a)

Type 1 Monte Carlo method: view the integral $\int_0^2 \exp(-x^2) dx$ as twice the av-

erage value of the function $f(x) = \exp(x^2)$ on the interval $[0; 2]$ which you can approximate with the Monte Carlo method

```

LCG := proc (m, a, b, x0, N, scale, shift)
local s, i;
s[0] := x0;
for i to N do s[i] := `mod`(a * s[i-1] + b, m)
end do;
return seq(scale * s[i] - shift, i = 1 .. N)
end proc;

```

$$\begin{aligned}
 & \text{> } n := 10000; \quad x0 := 3; \quad m := 2^{31} - 1; \quad a := 7^5; \quad b := 0; \\
 & \text{> } A := (m, a, b, x0, n, scale, shift) \rightarrow \text{evalf}\left(\frac{LCG(m, a, b, x0, n, scale, shift)}{m}\right) \\
 & \quad \quad \quad A := (m, a, b, x0, n, scale, shift) \mapsto \text{evalf}\left(\frac{LCG(m, a, b, x0, n, scale, shift)}{m}\right) \quad (42)
 \end{aligned}$$

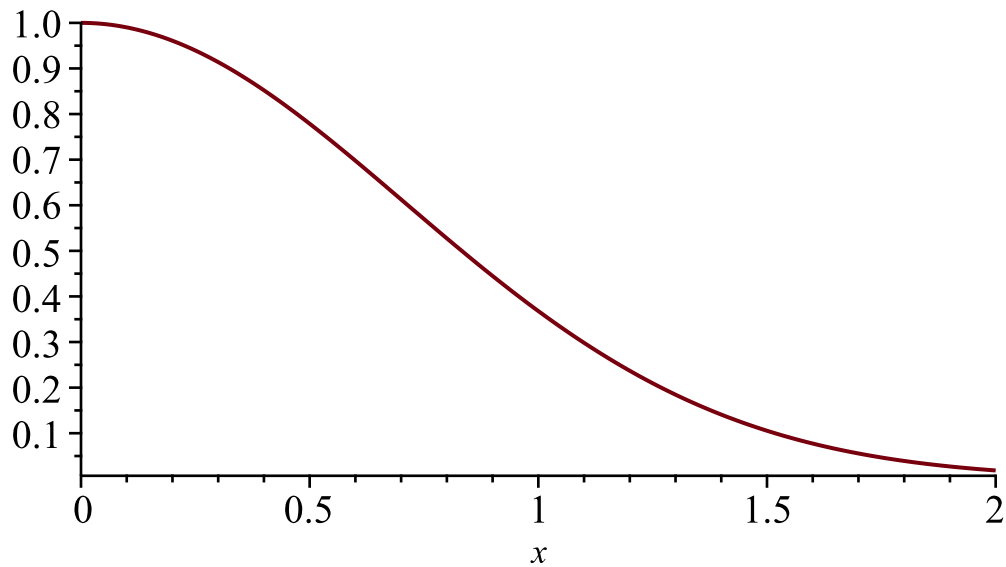
$$\begin{aligned}
 & \text{> } theoretical_value := \text{evalf}\left(\int_0^2 \exp(-x^2) \, dx\right) \\
 & \quad \quad \quad theoretical_value := 0.8820813910 \quad (43)
 \end{aligned}$$

$$\begin{aligned}
 & \text{> } f := x \rightarrow \exp(-x^2) \\
 & \quad \quad \quad f := x \mapsto e^{-x^2} \quad (44)
 \end{aligned}$$

```

> plotf := plot(f(x), x = 0.0 .. 2.0) :
> display(plotf, scaling = constrained)

```



```

> xvalues_mc_type1 := A(m, a, b, x0, n, 2, 0) :
> yvalues_mc_type1 := map(f, [xvalues_mc_type1]) :
> Stat_mc_type1 := 2 * Statistics[Mean](yvalues_mc_type1)
    Stat_mc_type1 := 0.874263292637839

```

(45)

(b) *Type 2 Monte Carlo method : enclose the area under the graph in the rectangular region $[0; 2] \times [0; 1]$, generate random pairs $(x; y)$ in that region and count how many land under the graph of $f(x) = \exp(-x^2)$; then estimate the area under the graph, hence the integral.*

```

> arearatio := proc(l, m)
    local i, k, n;
    i := 1; k := 0; n := nops(l);
    while i ≤ n - 1 do
        if m[i] ≤ f(l[i]) and m[i] ≥ g(l[i])
        then k := k + 1;
        end if;
        i := i + 1;
    end do;

```

```

    return( $\frac{k}{n}$ );
end proc:

```

```
> g := x → 0
```

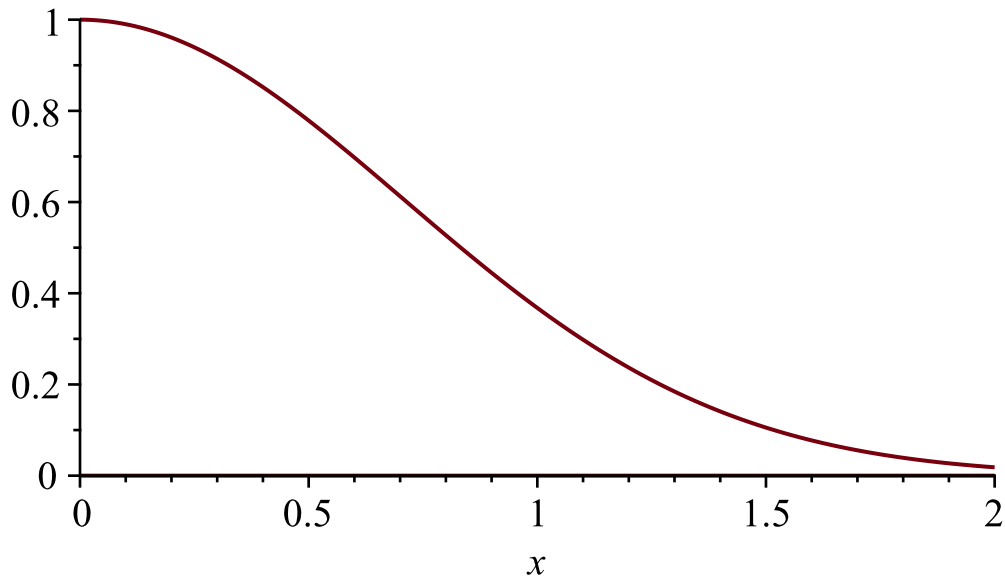
```
g := x ↦ 0
```

(46)

```
>
```

```
> plotg := plot(g(x), x = 0.0 .. 2.0) :
```

```
> display(plotf, plotg, scaling = constrained)
```



```
> xvalues_mc_type2 := A(m, a, b, x0, n, 2, 0) :
```

```
> yvalues_mc_type2 := A(m, a, b, x0, n, 1, 0) :
```

```
> Stat_mc_type2 := 2 · evalf( arearatio([xvalues_mc_type2],
[yvalues_mc_type2]))
```

```
Stat_mc_type2 := 0.8834000000
```

(47)

```
*****
```

```
> absolute_error_type_1 := abs(theoretical_value - Stat_mc_type1)
```

```
absolute_error_type_1 := 0.00781809836216141
```

(48)


```
> absolute_error_type_2 := abs(theoretical_value - Stat_mc_type2)
      absolute_error_type_2 := 0.0013186090
```

(49)

Answer:

theoretical_value := 0.8821 absolute_error_type_1 := 0.007818

a) For Type 1 Monte Carlo area under the graph:

Stat_mc_type1 := 0.8743 absolute_error_type_2 := 0.001319

b) For Type 2 Monte Carlo area under the graph:

Stat_mc_type2 := 0.8834

#####

4

(25 pts) Consider the $n \times n$ matrix with entries $A_{ij} = 1/(i+2j-1)$, set $x = [1; \dots; 1]^T$, and $b = Ax$. You act as the "quality control" person trying to find out what kind of job Maple's LinearSolve command does for this matrix (using double-precision computations).

(a)

Find $\|A\|_\infty$ (by hand) when $n = 5$ and check your work with Maple/Python.

> with(LinearAlgebra) : with(plots) :

> n := [5, 10, 15]

$n := [5, 10, 15]$

(50)

> M := x → Matrix(x, x, (i, j) → (1/(i + 2*j - 1)))

$M := x \mapsto \text{Matrix}(x, x, (i, j) \mapsto \frac{1}{i + 2j - 1})$

(51)

> X := x → Matrix(x, 1, (i, j) → (1))

$X := x \mapsto \text{Matrix}(x, 1, (i, j) \mapsto 1)$

(52)

> A := M(n[1])

$$A := \begin{bmatrix} \frac{1}{2} & \frac{1}{4} & \frac{1}{6} & \frac{1}{8} & \frac{1}{10} \\ \frac{1}{3} & \frac{1}{5} & \frac{1}{7} & \frac{1}{9} & \frac{1}{11} \\ \frac{1}{4} & \frac{1}{6} & \frac{1}{8} & \frac{1}{10} & \frac{1}{12} \\ \frac{1}{5} & \frac{1}{7} & \frac{1}{9} & \frac{1}{11} & \frac{1}{13} \\ \frac{1}{6} & \frac{1}{8} & \frac{1}{10} & \frac{1}{12} & \frac{1}{14} \end{bmatrix}$$

(53)

> AA := Norm(A, infinity)

$AA := \frac{137}{120}$

(54)

> AAdec := evalf(AA)

$AAdec := 1.141666667$

(55)

Answer:

$\|A\|_\infty$ (by hand) = $\frac{137}{120} = 1.1417$ (attached written page)

(b)

Use the `LinearSolve(A,b)` command in Maple (or `linalg.solve` in Python) to compute the solution `xc` in double-precision for `n = 5; 10; 15`.

$$\frac{137}{120} \quad (56)$$

> `with(MTM) :`

> `interface(rtablesize=15);`

$$12 \quad (57)$$

> `n[1]`

$$5 \quad (58)$$

> `A_5 := M(n[1]);`

$$A_5 := \begin{bmatrix} \frac{1}{2} & \frac{1}{4} & \frac{1}{6} & \frac{1}{8} & \frac{1}{10} \\ \frac{1}{3} & \frac{1}{5} & \frac{1}{7} & \frac{1}{9} & \frac{1}{11} \\ \frac{1}{4} & \frac{1}{6} & \frac{1}{8} & \frac{1}{10} & \frac{1}{12} \\ \frac{1}{5} & \frac{1}{7} & \frac{1}{9} & \frac{1}{11} & \frac{1}{13} \\ \frac{1}{6} & \frac{1}{8} & \frac{1}{10} & \frac{1}{12} & \frac{1}{14} \end{bmatrix} \quad (59)$$

> `x_5 := X(n[1])`

$$x_5 := \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \quad (60)$$

>

> `b_5 := double(A_5 • x_5)`

(61)

$$b_5 := \begin{bmatrix} 1.14166666666667 \\ 0.878210678210678 \\ 0.725000000000000 \\ 0.621800421800422 \\ 0.546428571428571 \end{bmatrix} \quad (61)$$

>

> $Xc_5 := \text{double}(\text{LinearSolve}(\text{Transpose}(A_5).A_5, \text{Transpose}(A_5).b_5))$

$$Xc_5 := \begin{bmatrix} 1.00000011133266 \\ 0.999997123469983 \\ 1.00001487507692 \\ 0.999974724640299 \\ 1.00001343707213 \end{bmatrix} \quad (62)$$

>

> $n[2]$

$$10 \quad (63)$$

> $A_10 := M(n[2]);$

$$A_10 := \begin{bmatrix} \frac{1}{2} & \frac{1}{4} & \frac{1}{6} & \frac{1}{8} & \frac{1}{10} & \frac{1}{12} & \frac{1}{14} & \frac{1}{16} & \frac{1}{18} & \frac{1}{20} \\ \frac{1}{3} & \frac{1}{5} & \frac{1}{7} & \frac{1}{9} & \frac{1}{11} & \frac{1}{13} & \frac{1}{15} & \frac{1}{17} & \frac{1}{19} & \frac{1}{21} \\ \frac{1}{4} & \frac{1}{6} & \frac{1}{8} & \frac{1}{10} & \frac{1}{12} & \frac{1}{14} & \frac{1}{16} & \frac{1}{18} & \frac{1}{20} & \frac{1}{22} \\ \frac{1}{5} & \frac{1}{7} & \frac{1}{9} & \frac{1}{11} & \frac{1}{13} & \frac{1}{15} & \frac{1}{17} & \frac{1}{19} & \frac{1}{21} & \frac{1}{23} \\ \frac{1}{6} & \frac{1}{8} & \frac{1}{10} & \frac{1}{12} & \frac{1}{14} & \frac{1}{16} & \frac{1}{18} & \frac{1}{20} & \frac{1}{22} & \frac{1}{24} \\ \frac{1}{7} & \frac{1}{9} & \frac{1}{11} & \frac{1}{13} & \frac{1}{15} & \frac{1}{17} & \frac{1}{19} & \frac{1}{21} & \frac{1}{23} & \frac{1}{25} \\ \frac{1}{8} & \frac{1}{10} & \frac{1}{12} & \frac{1}{14} & \frac{1}{16} & \frac{1}{18} & \frac{1}{20} & \frac{1}{22} & \frac{1}{24} & \frac{1}{26} \\ \frac{1}{9} & \frac{1}{11} & \frac{1}{13} & \frac{1}{15} & \frac{1}{17} & \frac{1}{19} & \frac{1}{21} & \frac{1}{23} & \frac{1}{25} & \frac{1}{27} \\ \frac{1}{10} & \frac{1}{12} & \frac{1}{14} & \frac{1}{16} & \frac{1}{18} & \frac{1}{20} & \frac{1}{22} & \frac{1}{24} & \frac{1}{26} & \frac{1}{28} \\ \frac{1}{11} & \frac{1}{13} & \frac{1}{15} & \frac{1}{17} & \frac{1}{19} & \frac{1}{21} & \frac{1}{23} & \frac{1}{25} & \frac{1}{27} & \frac{1}{29} \end{bmatrix} \quad (64)$$

```
> x_10 := X(n[2])
```

$$x_{10} := \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \quad (65)$$

```
> b_10 := double(A_10 • x_10)
```

$$b_{10} := \begin{bmatrix} 1.46448412698413 \\ 1.18087457777860 \\ 1.00993867243867 \\ 0.891019505314834 \\ 0.801605339105339 \\ 0.731019505314834 \\ 0.673400210900211 \\ 0.625199399494729 \\ 0.584114496614497 \\ 0.548571047004307 \end{bmatrix} \quad (66)$$

```
> Xc_10 := double(LinearSolve(Transpose(A_10).A_10, Transpose(A_10).b_10))
```

$$Xc_{10} := \begin{bmatrix} 1.00423747408443 \\ 0.659778324466989 \\ 6.25612644809576 \\ -27.5676101516914 \\ 64.4963052615087 \\ -47.1722599531985 \\ -4.64896429150035 \\ -18.4267525991919 \\ 71.7527161688275 \\ -36.4089124180204 \end{bmatrix} \quad (67)$$

```
> n[3]
```

$$A_{15} :=$$
[illegible]

```
> x_15 := X(n[3])
```

$$x_{15} := \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \quad (70)$$

> $b_{15} := \text{double}(A_{15} \cdot x_{15})$

$$b_{15} := \begin{bmatrix} 1.65911449661450 \\ 1.36813069882202 \\ 1.19036449661450 \\ 1.06510039579172 \\ 0.969776261320379 \\ 0.893671824363149 \\ 0.830887372431490 \\ 0.777841708533033 \\ 0.732203161905174 \\ 0.692371623062948 \\ 0.657203161905174 \\ 0.625852776056296 \\ 0.597679352381365 \\ 0.572185513086707 \\ 0.548978053680066 \end{bmatrix} \quad (71)$$

> $Xc_{15} := \text{double}(\text{LinearSolve}(\text{Transpose}(A_{15}).A_{15}, \text{Transpose}(A_{15}).b_{15}))$

$$Xc_{15} := \begin{bmatrix} 0.998731519878267 \\ 1.10886847155932 \\ -0.784935890655999 \\ 10.9580411415985 \\ -19.1138712586410 \\ 2.44522998431929 \\ 41.7914612958163 \\ -42.2779372012657 \\ 17.6877502518298 \\ 21.3585181191549 \\ -75.2593893385501 \\ 60.9398133357623 \\ -17.2268431580972 \\ 38.1198501657362 \\ -25.7767440239996 \end{bmatrix} \quad (72)$$

(c)

Find the infinity norm of the forward error $\|x - xc\|_\infty$, the backward error $\|b - Axc\|_\infty$, and the error magnification factor for $n = 5; 10; 15$.

$$\begin{aligned} &> ForwardError_5_Infinity = Norm(x_5 - Xc_5, infinity) \\ &\quad ForwardError_5_Infinity = 0.0000252753597005384023 \end{aligned} \quad (73)$$

$$\begin{aligned} &> BackwardError_5_infinity := Norm(b_5 - A_5 \cdot Xc_5, infinity) \\ &\quad BackwardError_5_infinity := 1.90433224744879226 \cdot 10^{-11} \end{aligned} \quad (74)$$

$$\begin{aligned} &> ErrorMagnificationFactor_5 := \frac{\frac{Norm(x_5 - Xc_5, infinity)}{Norm(x_5, infinity)}}{\frac{Norm(b_5 - A_5 \cdot Xc_5, infinity)}{Norm(b_5, infinity)}} \\ &\quad ErrorMagnificationFactor_5 := 1.515283676 \cdot 10^6 \end{aligned} \quad (75)$$

$$\begin{aligned} &> ForwardError_10_Infinity = Norm(x_10 - Xc_10, infinity) \\ &\quad ForwardError_10_Infinity = 70.7527161688275186 \end{aligned} \quad (76)$$

$$\begin{aligned} &> BackwardError_10_infinity := Norm(b_10 - A_10 \cdot Xc_10, infinity) \\ &\quad BackwardError_10_infinity := 2.56394261377579369 \cdot 10^{-8} \end{aligned} \quad (77)$$

$$\begin{aligned}
 & \text{ErrorMagnificationFactor}_{10} := \frac{\frac{\text{Norm}(x_{10} - Xc_{10}, \text{infinity})}{\text{Norm}(x_{10}, \text{infinity})}}{\frac{\text{Norm}(b_{10} - A_{10} \cdot Xc_{10}, \text{infinity})}{\text{Norm}(b_{10}, \text{infinity})}} \\
 & \text{ErrorMagnificationFactor}_{10} := 4.041285058 \cdot 10^9
 \end{aligned} \tag{78}$$

$$\begin{aligned}
 & \text{ForwardError}_{15_Infinity} = \text{Norm}(x_{15} - Xc_{15}, \text{infinity}) \\
 & \text{ForwardError}_{15_Infinity} = 76.2593893385501360
 \end{aligned} \tag{79}$$

$$\begin{aligned}
 & \text{BackwardError}_{15_infinity} := \text{Norm}(b_{15} - A_{15} \cdot Xc_{15}, \text{infinity}) \\
 & \text{BackwardError}_{15_infinity} := 6.32045549231463610 \cdot 10^{-9}
 \end{aligned} \tag{80}$$

$$\begin{aligned}
 & \text{ErrorMagnificationFactor}_{15} := \frac{\frac{\text{Norm}(x_{15} - Xc_{15}, \text{infinity})}{\text{Norm}(x_{15}, \text{infinity})}}{\frac{\text{Norm}(b_{15} - A_{15} \cdot Xc_{15}, \text{infinity})}{\text{Norm}(b_{15}, \text{infinity})}} \\
 & \text{ErrorMagnificationFactor}_{15} := 2.001802853 \cdot 10^{10}
 \end{aligned} \tag{81}$$

(d)

Find the condition number of A for n = 5; 10; 15 and compare it with error magnification factor obtained in part (c).

$$\begin{aligned}
 & \text{ConditionNumber}_5 := \text{double}(\text{ConditionNumber}(A_5)) \\
 & \text{ConditionNumber}_5 := 1.97906775000000000 \cdot 10^6
 \end{aligned} \tag{82}$$

$$\begin{aligned}
 & \text{ConditionNumber}_{10} := \text{double}(\text{ConditionNumber}(A_{10})) \\
 & \text{ConditionNumber}_{10} := 1.31334652072236797 \cdot 10^{14}
 \end{aligned} \tag{83}$$

$$\begin{aligned}
 & \text{ConditionNumber}_{15} := \text{double}(\text{ConditionNumber}(A_{15})) \\
 & \text{ConditionNumber}_{15} := 9.28569976203256739 \cdot 10^{21}
 \end{aligned} \tag{84}$$

$$\begin{aligned}
 & \frac{\text{ConditionNumber}_5}{\text{ErrorMagnificationFactor}_5} \\
 & 1.306070791
 \end{aligned} \tag{85}$$

$$\begin{aligned}
 & \frac{\text{ConditionNumber}_{10}}{\text{ErrorMagnificationFactor}_{10}} \\
 & 32498.24009
 \end{aligned} \tag{86}$$

$$\begin{aligned}
 & \frac{\text{ConditionNumber}_{15}}{\text{ErrorMagnificationFactor}_{15}} \\
 & 4.638668462 \cdot 10^{11}
 \end{aligned} \tag{87}$$

Answer:

The condition Number is the maximum possible error magnification factor
As we can see in all this 3 examples EMF < Condition number as expected.

n=5 :

$$\text{ConditionNumber}_5 := 1.9791 \cdot 10^6 \quad \text{ErrorMagnificationFactor}_5 := 1.5153 \cdot 10^6$$

for n=5 EMF and Condition Number have the same order O(1)

n=10 :

$$\text{ConditionNumber}_{10} := 1.3133 \cdot 10^{14} \gg \text{ErrorMagnificationFactor}_{10} := 4.041285058 \cdot 10^9$$

for n=10 Condition Number vs EMF = O(10^5)

n=15 :

$$\text{ConditionNumber}_{15} := 9.2857 \cdot 10^{21} \gg \text{ErrorMagnificationFactor}_{15} := 2.001802853 \cdot 10^{10}$$

for n=15 Condition Number vs EMF = O(10^9)

with $n \rightarrow \infty$ the condition number $\rightarrow \infty$. This matrix is ill conditioned, almost singular, and the computation of its inverse, or solution of a linear system of equations is prone to large numerical errors

(e)

For what value of n does the solution Xc of the linear system start to have no correct significant digits?

Answer: The answer is different for different algorithms. For this solution :

For value of n=5 condition number = $2 \cdot 10^6$

For double precision we expect to lose **6** significant digits, we will be left with **16-6=10** significant digits

For value of n=10 condition number = $1.3 \cdot 10^{14}$

For double precision we expect to lose **14** significant digits, we will be left with **16-14=2** significant digits

> ConditionNumber(double(M(11)))

$$4.804517656 \cdot 10^{15} \quad (88)$$

> ConditionNumber(double(M(12)))

$$1.202196980 \cdot 10^{17} \quad (89)$$

For value of n=11 condition number = $4.8 \cdot 10^{15}$

For double precision we expect to lose **15** significant digits, we will have left with **16-15=1** significant digits

For value of n=12 condition number = $1.8 \cdot 10^{17}$

For double precision we expect to lose **17** significant digits, we will have left with **16-17= -1** significant digits

Start with n=12 the solution Xc of the linear system start to have no correct significant digits

>

#####

5.

(10 pts) Design a fancy letter symbol representing your last name initial, using at least four Bezier curves joined together.

#####

```
> with(plots) :
> bezier:=proc(param)
    local x1, x2, x3, x4, y1, y2, y3, y4, bx, cx, bvy, cy, fx, fy, dx, dy;
    x1:=param[1, 1]; y1:=param[1, 2];
    x2:=param[2, 1]; y2:=param[2, 2];
    x3:=param[3, 1]; y3:=param[3, 2];
    x4:=param[4, 1]; y4:=param[4, 2];
    bx:=3*(x2-x1); bvy:=3*(y2-y1);
    cx:=3*(x3-x2)-bx; cy:=3*(y3-y2)-bvy;
    dx:=x4-x1-bx-cx; dy:=y4-y1-bvy-cy;
    fx:=proc(t) options operator, arrow; x1+bx*t+cx*t^2+dx*t
^3 end proc;
    fy:=proc(t) options operator, arrow; y1+bvy*t+cy*t^2+dy*t
^3 end proc;
    plot([fx(t), fy(t), t=0..1]);
end proc:

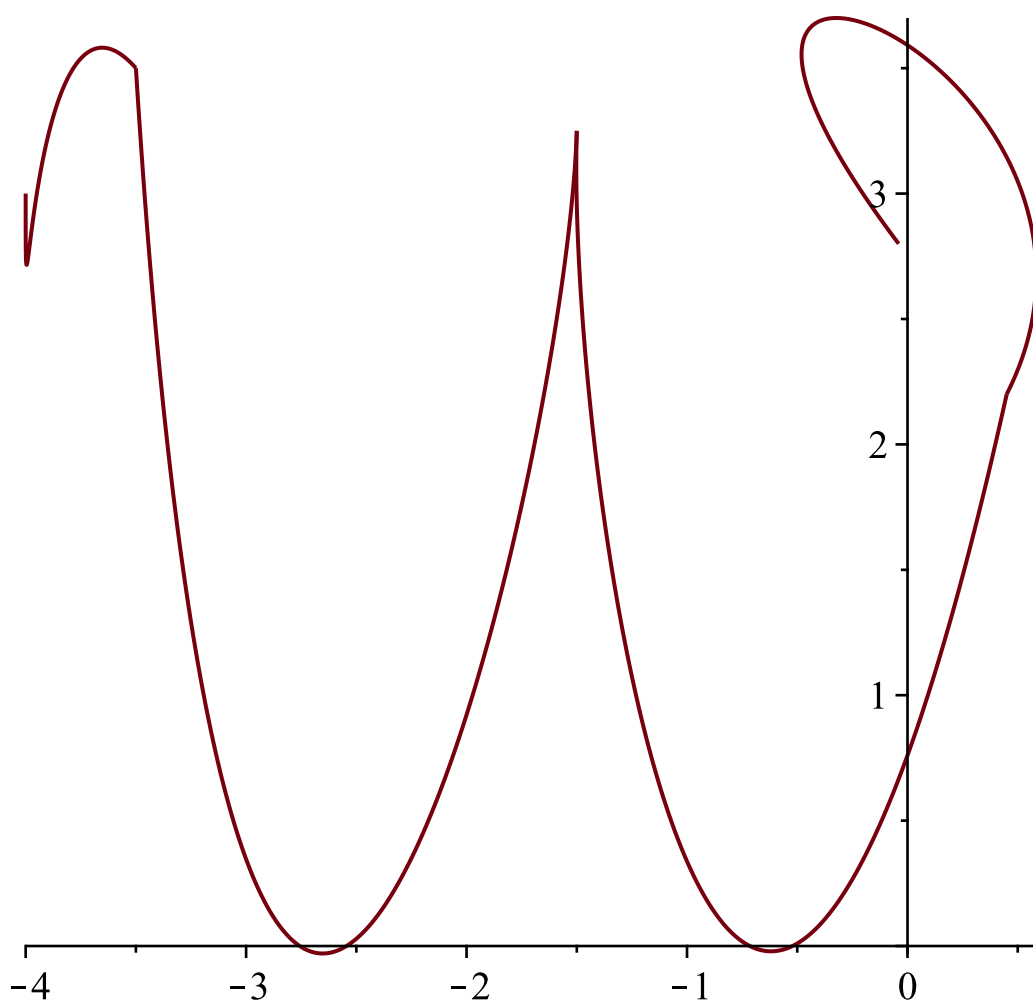
> data1 := [[-4.0, 3.0], [-4.0, 2.0], [-4.0, 4.0], [-3.5, 3.5]];
    data1 := [[-4.0, 3.0], [-4.0, 2.0], [-4.0, 4.0], [-3.5, 3.5]] (90)

> data2 := [[-3.5, 3.5], [-3.0, -3.5], [-1.55, 1.83], [-1.5, 3.25]]
    data2 := [[-3.5, 3.5], [-3.0, -3.5], [-1.55, 1.83], [-1.5, 3.25]] (91)

> data3 := [[-1.5, 3.25], [-1.55, 1.83], [-0.79, -2.7], [0.45, 2.2]];
    data3 := [[-1.5, 3.25], [-1.55, 1.83], [-0.79, -2.7], [0.45, 2.2]] (92)

> data4 := [[0.45, 2.2], [1.2, 3.5], [-1.5, 4.5], [-0.04, 2.80]]
    data4 := [[0.45, 2.2], [1.2, 3.5], [-1.5, 4.5], [-0.04, 2.80]] (93)

> display(bezier(data1), bezier(data2), (bezier(data3), bezier(data4)))
```



Curve 1

data1 := [[-4.0, 3.0], [-4.0, 2.0], [-4.0, 4.0], [-3.5, 3.5]];

$$bx=3*(x2-x1)=3*(-4+4)=0$$

$$cx=3*(x3-x2)=3*(-4+4)=0$$

$$dx = x4-x1-bx-cx= -3.5+4 - 0 - 0= 0.5$$

$$by=3*(y2-y1)=3*(2-3)= -3$$

$$cy=3*(y3-y2)=3*(4-2)= 6$$

$$X1(t) = -4 + 0.5t^3$$

$$Y1(t) = 3 - 3t + 6t^2 - 2.5t^3$$

Curve 2

data2 := [[-3.5, 3.5], [-3.0, -3.5], [-1.55, 1.83], [-1.5, 3.25]]

$$bx=3*(x2-x1)=3*(-3.0+3.5)= 1.5$$

$$cx=3*(x3-x2)=3*(-1.55+3)=4.35$$

$$dx=x4-x1-bx-cx=-1.5+3.5-4.35-1.5=-3.85$$

$$by=3*(y2-y1)=3*(-3.5-3.5)=-21$$

$$cy=3*(y3-y2)=3*(1.83+3.5)=15.99$$

$$X1(t) = -3.5 + 1.5t + 4.35 t^2 - 3.85 t^3$$

$$Y1(t) = 3.5 - 21t + 15.99 t^2 - 4.76 t^3$$

Curve 3

data3 := [[-1.5, 3.25], [-1.55, 1.83], [-0.79, -2.7], [0.45, 2.2]]

$$bx=3*(x2-x1)=3*(-1.55+1.5)=-0.15$$

$$cx=3*(x3-x2)=3*(-0.79+1.55)=2.28$$

$$dx=x4-x1-bx-cx=0.45+1.5-2.28+0.15=-0.18$$

$$by=3*(y2-y1)=3*(1.83-3.25)=-4.26$$

$$cy=3*(y3-y2)=3*(-2.7-1.83)=-13.59$$

$$dy=y4-y1-by-cy=2.2+1.5+4.26+13.59=16.8$$

$$X1(t) = -1.5 - 0.15t + 2.28 t^2 - 0.18 t^3$$

$$Y1(t) = 3.25 - 4.26t - 13.59 t^2 + 16.8 t^3$$

Curve 4

data4 := [[0.45, 2.2], [1.2, 3.5], [-1.5, 4.5], [-0.04, 2.80]]

$$bx=3*(x2-x1)=3*(1.2-0.45)=2.25$$

$$cx=3*(x3-x2)=3*(-1.5-1.2)=-8.1$$

$$dx=x4-x1-bx-cx=-0.04-0.45-2.25+8.1=5.36$$

$$by=3*(y2-y1)=3*(3.5-2.2)=3.9$$

$$cy=3*(y3-y2)=3*(4.5-3.5)=3$$

$$dy=y4-y1-by-cy=2.8-2.2-3.9-3=-6.3$$

$$X1(t) = 0.45 + 2.25t - 8.1 t^2 + 5.36 t^3$$

$$Y1(t) = 2.2 + 3.9t + 3 t^2 - 6.3 t^3$$

=
>
>