

```
In[639]:=ResourceFunction["DarkMode"] []
```

HW2

Problem 1

1.

```
In[187]:= ClearAll["Global`*"]  
g[x_] := -16 + 6x +  $\frac{12}{x}$   
gp[x_] = Evaluate[D[g[x], x]]  
gp[2]
```

Out[189]=

$$6 - \frac{12}{x^2}$$

Out[190]=

$$3$$

$$g(x) = -16 + 6x + \frac{12}{x}$$

$$g'(x) = 6 - \frac{12}{x^2}$$

$$g'(a) = g'(2) = 3 > 1$$

So diverges.

2.

```
In[222]:= ClearAll["Global`*"]
g[x_] :=  $\frac{2}{3}x + \frac{1}{x^2}$ 
gp[x_] = Evaluate[D[g[x],x]]
gp[31/3]

gpp[x_] = Evaluate[D[%%,x]]
gpp[31/3]
```

Out[224]=

$$\frac{2}{3} - \frac{2}{x^3}$$

Out[225]=

$$0$$

Out[226]=

$$\frac{6}{x^4}$$

Out[227]=

$$\frac{2}{3^{1/3}}$$

$$g(x) = \frac{2}{3}x + \frac{1}{x^2}$$

$$g'(x) = \frac{2}{3} - \frac{2}{x^3}$$

$$g'(\alpha) = g'(2) = 0$$

$$g''(x) = \frac{6}{x^4}$$

$$g''(\alpha) = g''(2) = \frac{2}{3^{1/3}} \neq 0$$

So 2nd order converges.

3.

```
ClearAll["Global`*"]
g[x_] := 12/(1+x)
gp[x_] = Evaluate[D[g[x],x]]
gp[3]
Abs[%]
```

Out[268]=

$$-\frac{12}{(1+x)^2}$$

Out[269]=

$$-\frac{3}{4}$$

Out[270]=

$$\frac{3}{4}$$

$$g(x) = \frac{12}{1+x}$$

$$g'(x) = -\frac{12}{(1+x)^2}$$

$$g'(\alpha) = g'(2) = -\frac{3}{4}$$

$$\lim_{n \rightarrow \infty} \left| \frac{e_{n+1}}{e_n} \right| = |g'(\alpha)| = \frac{3}{4}$$

So linear converge with rate of $\frac{3}{4}$

Problem 2

```
In[292]:= ClearAll["Global`*"]

g[x_] :=  $\frac{x(x^2+3a)}{3x^2+a}$ 

gp[x_] = Evaluate[D[g[x],x]]
gp[ $\sqrt{a}$ ]

gpp[x_] = Evaluate[D[g[x],x]]
gpp[ $\sqrt{a}$ ]

gppp[x_] = Evaluate[D[gpp[x],x]]
gppp[ $\sqrt{a}$ ]

%/3!
```

$$\text{Out[294]} = -\frac{6x^2(3a+x^2)}{(a+3x^2)^2} + \frac{2x^2}{a+3x^2} + \frac{3a+x^2}{a+3x^2}$$

$$\text{Out[295]} = 0$$

$$\text{Out[296]} = \frac{72x^3(3a+x^2)}{(a+3x^2)^3} - \frac{24x^3}{(a+3x^2)^2} - \frac{18x(3a+x^2)}{(a+3x^2)^2} + \frac{6x}{a+3x^2}$$

$$\text{Out[297]} = 0$$

$$\text{Out[298]} = -\frac{1296x^4(3a+x^2)}{(a+3x^2)^4} + \frac{432x^4}{(a+3x^2)^3} + \frac{432x^2(3a+x^2)}{(a+3x^2)^3} - \frac{144x^2}{(a+3x^2)^2} - \frac{18(3a+x^2)}{(a+3x^2)^2} + \frac{6}{a+3x^2}$$

$$\text{Out[299]} = \frac{3}{2a}$$

$$\text{Out[300]} = \frac{1}{4a}$$

$$g(x) = \frac{x(x^2+3a)}{3x^2+a}$$

$$g'(x) = -\frac{6x^2(3a+x^2)}{(a+3x^2)^2} + \frac{2x^2}{a+3x^2} + \frac{3a+x^2}{a+3x^2}$$

$$g'(\alpha) = g'(\sqrt{a}) = 0$$

$$g''(x) = \frac{72x^3(3a+x^2)}{(a+3x^2)^3} - \frac{24x^3}{(a+3x^2)^2} - \frac{18x(3a+x^2)}{(a+3x^2)^2} + \frac{6x}{a+3x^2}$$

$$g''(\alpha) = g''(\sqrt{a}) = 0$$

$$g'''(x) = -\frac{1296 x^4 (3 a + x^2)}{(a + 3 x^2)^4} + \frac{432 x^4}{(a + 3 x^2)^3} + \frac{432 x^2 (3 a + x^2)}{(a + 3 x^2)^3} - \frac{144 x^2}{(a + 3 x^2)^2} - \frac{18 (3 a + x^2)}{(a + 3 x^2)^2} + \frac{6}{a + 3 x^2}$$

$$g'''(\alpha) = g'''(\sqrt{a}) = \frac{3}{2a} \neq 0$$

So 3rd order converges.

$$x_{n+1} = g(x_n) = g(\sqrt{a}) + (x_n - \sqrt{a})g'(\alpha) + (x_n - \sqrt{a})^2 / 2 * g''(\alpha) + (x_n - \sqrt{a})^3 / 3! * g'''(\alpha)$$

$$g(\sqrt{a}) - x_{n+1} = (\sqrt{a} - x_n)^3 / 3! * g'''(\alpha)$$

$$\lim_{n \rightarrow \infty} \frac{\sqrt{a} - x_{n+1}}{(\sqrt{a} - x_n)^3} = \frac{g'''(\alpha)}{3!} = \frac{1}{4a}$$

Problem 3

$$\mathbf{x}_{n+1} = \mathbf{g}(\mathbf{x}_n)$$

Given $\mathbf{x} = \mathbf{g}(\mathbf{x})$ has a unique solution $\alpha \in D$

By MVT, $\alpha_i - x_{i,n+1} = \partial g_i(\delta_n^{(i)}) / \partial x_1 * (\alpha_i - x_{1,n}) + \dots + \partial g_i(\delta_n^{(i)}) / \partial x_m * (\alpha_i - x_{m,n})$
 where $\delta_n^{(i)}$ denotes a point on the line segment between α and \mathbf{x}_n

$$\alpha - \mathbf{x}_{n+1} = \begin{pmatrix} \partial g_1(\delta_n^{(1)}) / \partial x_1 & \dots & \partial g_1(\delta_n^{(1)}) / \partial x_m \\ \dots & \dots & \dots \\ \partial g_o(\delta_n^{(o)}) / \partial x_1 & \dots & \partial g_o(\delta_n^{(o)}) / \partial x_m \end{pmatrix} (\alpha - \mathbf{x}_n)$$

Let G_n denotes the matrix above where $\alpha - \mathbf{x}_{n+1} = \mathbf{G}_n(\alpha - \mathbf{x}_n)$

$$\|\alpha - \mathbf{x}_{n+1}\|_\infty = \|\mathbf{g}(\alpha) - \mathbf{g}(\mathbf{x}_n)\|_\infty = \|\mathbf{G}_n(\alpha - \mathbf{x}_n)\|_\infty \leq \|\mathbf{G}_n\|_\infty \|\alpha - \mathbf{x}_n\|_\infty \leq \max_{\mathbf{x} \in D} \|\mathbf{G}(\mathbf{x})\|_\infty \|\alpha - \mathbf{x}_n\|_\infty$$

$$\text{i.e. } \|\alpha - \mathbf{x}_{n+1}\|_\infty \leq \lambda \|\alpha - \mathbf{x}_n\|_\infty$$

induction made that $\|\alpha - \mathbf{x}_n\|_\infty \leq \lambda^n \|\alpha - \mathbf{x}_0\|_\infty$

as $\lambda < 1$, $\lim_{n \rightarrow \infty} \|\alpha - \mathbf{x}_n\|_\infty \leq 0$

so as $n \rightarrow \infty$, $\mathbf{x}_n \rightarrow \alpha$

the iterates converge to the solution to $\mathbf{x} = \mathbf{g}(\mathbf{x})$

Problem 4

```
In[594]:= ClearAll["Global`*"]
f1[x_, y_] := x^2 + y^2 - 4
f2[x_, y_] := x^2 - y^2 - 1
f1x[x_, y_] = Evaluate[D[f1[x,y],x]]
f1y[x_, y_] = Evaluate[D[f1[x,y],y]]
f2x[x_, y_] = Evaluate[D[f2[x,y],x]]
f2y[x_, y_] = Evaluate[D[f2[x,y],y]]

J[x_,y_] :=  $\begin{pmatrix} f1x[x, y] & f1y[x, y] \\ f2x[x, y] & f2y[x, y] \end{pmatrix}$ 
MatrixForm[J[x,y]]
Jinv[x_,y_] := Inverse[J[x,y]]
MatrixForm[Jinv[x,y]]
```

Out[597]=

$2 x$

Out[598]=

$2 y$

Out[599]=

$2 x$

Out[600]=

$-2 y$

Out[602]//MatrixForm=

$\begin{pmatrix} 2 x & 2 y \\ 2 x & -2 y \end{pmatrix}$

Out[604]//MatrixForm=

$\begin{pmatrix} \frac{1}{4 x} & \frac{1}{4 x} \\ \frac{1}{4 y} & -\frac{1}{4 y} \end{pmatrix}$











```

In 1 1 import numpy as np
2
3 # Define the functions of the system
4 def F(x):
5     return [x[0]**2 + x[1]**2 - 4, x[0]**2 - x[1]**2 - 1]
6
7 # Define the Jacobian matrix of the system
8 def J_inv(x):
9     return [[1/4/x[0], 1/4/x[0]], [1/4/x[1], -1/4/x[1]]]
10
11 # Initial guess
12 x0 = [1.6, 1.2]
13
14 # Newton's method iteration function
15 def newton_method(F, J_inv, x0, tol=1e-10, max_iter=100):
16     x = np.array(x0)
17     for _ in range(max_iter):
18         Fx = np.array(F(x))
19         Jx_inv = np.array(J_inv(x))
20         # Update the guess using the inverse Jacobian
21         x = x - np.dot(Jx_inv, Fx)
22         # Check for convergence
23         if np.linalg.norm(Fx, ord=np.inf) < tol:
24             return x
25         raise ValueError('Newton method did not converge')
26
27 # Perform Newton's method
28 solution = newton_method(F, J_inv, x0)
29
30 solution
31

```

Executed at 2024.04.02 14:38:04 in 75ms

Out 1 ▾

				2 rows ▾			2 rows × 1 columns				
÷	123	0	÷								
0				1.581139							
1				1.224745							

Check Results using Mathematica built-in:

```
In[615]:= FindRoot[{f1[x,y],f2[x,y]},{x,1.6},{y,1.2},WorkingPrecision->10]
```

```
Out[615]= {x -> 1.581138830, y -> 1.224744871}
```

Check Results using true solutions:

```
In[636]:= "x*" = "N[±√2.5]"
           "y*" = "N[±√1.5]"
```

```
Out[636]= x* = (±1.58114)
```

```
Out[637]= y* = (±1.22474)
```

$$J = \begin{pmatrix} 2x & 2y \\ 2x & -2y \end{pmatrix}$$

$$\det(J) = -8xy$$

$$J^{-1} = \frac{1}{\det(J)} \begin{pmatrix} -2y & -2y \\ -2x & 2x \end{pmatrix} = \begin{pmatrix} \frac{1}{4x} & \frac{1}{4x} \\ \frac{1}{4y} & -\frac{1}{4y} \end{pmatrix}$$

using python, $\mathbf{x}_{n+1} = \mathbf{x}_n - \mathbf{J}^{-1}(\mathbf{x}_n) \mathbf{F}(\mathbf{x}_n)$, where $\mathbf{F}(\mathbf{x}) = \begin{pmatrix} f_1(\mathbf{x}) \\ f_2(\mathbf{x}) \end{pmatrix}$,

iterations starts with $\mathbf{x}_0 = \begin{pmatrix} 1.6 \\ 1.2 \end{pmatrix}$,

The output is $\mathbf{x}^* = \begin{pmatrix} x^* \\ y^* \end{pmatrix} = \begin{pmatrix} 1.581139 \\ 1.224745 \end{pmatrix}$, (shown above)

so the ans is (1.581139, 1.224745), which coincides with the given true solutions.

Problem 5

$$\epsilon_i = y_i - (a + bx_i)$$

$$e = \frac{1}{n} \sum_{i=1}^n \epsilon_i^2 = \frac{1}{n} \sum_{i=1}^n [y_i - (a + bx_i)]^2$$

To calculate $\min_{a,b} e$

$$\frac{\partial e}{\partial a} = \frac{\partial e}{\partial b} = 0$$

$$\frac{\partial e}{\partial a} = -\frac{2}{n} \sum_{i=1}^n [y_i - (a + bx_i)] = 0 \text{ (eq1)}$$

$$\frac{\partial e}{\partial b} = -\frac{2}{n} \sum_{i=1}^n x_i [y_i - (a + bx_i)] = 0 \text{ (eq2)}$$

introduce the averages $\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$, $\bar{y} = \frac{1}{n} \sum_{i=1}^n y_i$

from eq1, $-\bar{y} + \hat{a} + \hat{b} \bar{x} = 0$

which is $\hat{a} = \bar{y} - \hat{b} \bar{x}$

from eq2, $\sum_{i=1}^n x_i y_i - \hat{a} \sum_{i=1}^n x_i - \hat{b} \sum_{i=1}^n x_i^2 = 0$

take \hat{a} in, $\sum_{i=1}^n x_i y_i - (\bar{y} - \hat{b} \bar{x}) \sum_{i=1}^n x_i - \hat{b} \sum_{i=1}^n x_i^2 = 0$

$$\sum_{i=1}^n x_i y_i - (\bar{y} - \hat{b} \bar{x}) \sum_{i=1}^n x_i - \hat{b} \sum_{i=1}^n x_i^2 = 0$$

$$\hat{b} = \frac{\sum_{i=1}^n x_i y_i - \bar{y} \sum_{i=1}^n x_i}{\sum_{i=1}^n x_i^2 - \bar{x} \sum_{i=1}^n x_i} = \frac{\sum_{i=1}^n x_i y_i - n \bar{x} \bar{y}}{\sum_{i=1}^n x_i^2 - n \bar{x}^2}$$

$$= \frac{\sum_{i=1}^n x_i y_i - \sum_{i=1}^n x_i \bar{y} - \sum_{i=1}^n \bar{x} y_i + \sum_{i=1}^n \bar{x} \bar{y}}{\sum_{i=1}^n x_i^2 - 2 \sum_{i=1}^n x_i \bar{x} + \sum_{i=1}^n \bar{x}^2} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

so the answer is:

$$\hat{b} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

$$\hat{a} = \bar{y} - \hat{b} \bar{x}$$

Problem 6

$$\text{In[683]:= } \mathbf{A} = \frac{1}{4} \begin{pmatrix} 1 & 3 & 3 \\ -3 & -5 & -3 \\ 3 & 3 & 1 \end{pmatrix}$$

$$\text{Dot} \left[\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} - \mathbf{A}, \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \right]$$

MatrixForm[%]

MatrixForm[%]

Out[683]=

$$\left\{ \left\{ \frac{1}{4}, \frac{3}{4}, \frac{3}{4} \right\}, \left\{ -\frac{3}{4}, -\frac{5}{4}, -\frac{3}{4} \right\}, \left\{ \frac{3}{4}, \frac{3}{4}, \frac{1}{4} \right\} \right\}$$

Out[684]=

$$\left\{ \left\{ -\frac{3}{4} \right\}, \left\{ \frac{15}{4} \right\}, \left\{ -\frac{3}{4} \right\} \right\}$$

Out[685]//MatrixForm=

$$\begin{pmatrix} \frac{1}{4} & \frac{3}{4} & \frac{3}{4} \\ -\frac{3}{4} & -\frac{5}{4} & -\frac{3}{4} \\ \frac{3}{4} & \frac{3}{4} & \frac{1}{4} \end{pmatrix}$$

Out[686]//MatrixForm=

$$\begin{pmatrix} -\frac{3}{4} \\ \frac{15}{4} \\ -\frac{3}{4} \end{pmatrix}$$

Take $\mathbf{I} - \mathbf{A} = \begin{pmatrix} \frac{1}{4} & \frac{3}{4} & \frac{3}{4} \\ -\frac{3}{4} & -\frac{5}{4} & -\frac{3}{4} \\ \frac{3}{4} & \frac{3}{4} & \frac{1}{4} \end{pmatrix}$ and $\mathbf{x} = \begin{pmatrix} -\frac{3}{4} \\ \frac{15}{4} \\ -\frac{3}{4} \end{pmatrix}$ in,

$$\mathbf{b} = \begin{pmatrix} -\frac{3}{4} \\ \frac{15}{4} \\ -\frac{3}{4} \end{pmatrix}$$

By Neumann series,

$$\mathbf{x} = \mathbf{b} + \mathbf{A}\mathbf{b} + \mathbf{A}^2\mathbf{b} + \dots$$

To implement,

$$\mathbf{x}_{n+1} = \mathbf{A}\mathbf{x}_n + \mathbf{b}$$

```

In 2 1 # Given matrix A and vector x from the problem
2     A = np.array([[1, 3, 3],
3                   [-3, -5, -3],
4                   [3, 3, 1]]) / 4.0
5     x_true = np.array([1, 1, 1])
6
7
8     # Initial guess x0 for the iteration
9     x0 = np.array([1.1, 1.1, 1.1])
10
11    # Since (I - A)x = b, we find b given the true solution x
12    b = np.array([-3/4, 15/4, -3/4])
13
14    # Iteration method to solve  $x_{n+1} = Ax_n + b$ 
15    def iterative_solve(A, b, x0, tol=1e-5, max_iter=1000):
16        x_n = x0
17        for i in range(max_iter):
18            x_n_plus_1 = np.dot(A, x_n) + b
19            # Check for convergence
20            if np.linalg.norm(x_n_plus_1 - x_n, ord=np.inf) < tol:
21                return x_n_plus_1
22            x_n = x_n_plus_1
23        raise ValueError("The method did not converge")
24
25    # Perform the iteration
26    x_approx = iterative_solve(A, b, x0)
27
28    x_approx

```

Executed at 2024.04.03 01:57:09 in 7ms

Out 2

	123	0
0	1.000002	
1	0.999997	
2	1.000002	

3 rows × 1 columns

With Python codes above, the result is (1.000002, 0.999997, 1.000002).

To prove the convergence,

let α denote the true solution, s.t. $\alpha = \mathbf{g}(\alpha) = \mathbf{A}\alpha + \mathbf{b}$

$$\mathbf{x}_{n+1} = \mathbf{A}\mathbf{x}_n + \mathbf{b}$$

$$\mathbf{e}_{n+1} = \alpha - \mathbf{x}_{n+1} = \mathbf{A}\alpha + \mathbf{b} - (\mathbf{A}\mathbf{x}_n + \mathbf{b}) = \mathbf{A}(\alpha - \mathbf{x}_n) = \mathbf{A}\mathbf{e}_n$$

```

In[753]:= {eigenvalues, eigenvectors} = Eigensystem[A];

{eigenvalues, eigenvectors}

V = Transpose[eigenvectors]; (* Mathematica returns eigenvectors as a list of
Dia = DiagonalMatrix[eigenvalues];
VInverse = Inverse[V];

Det[V]

(* Display V, D, and V^-1 *)
{MatrixForm[V], MatrixForm[Dia], MatrixForm[VInverse]}

(* Verify the decomposition *)
MatrixForm[A_reconstructed = V.Dia.VInverse]

```

Out[754]=

$$\left\{ \left\{ -\frac{1}{2}, -\frac{1}{2}, \frac{1}{4} \right\}, \left\{ \{-1, 0, 1\}, \{-1, 1, 0\}, \{1, -1, 1\} \right\} \right\}$$

Out[758]=

$$-1$$

Out[759]=

$$\left\{ \begin{pmatrix} -1 & -1 & 1 \\ 0 & 1 & -1 \\ 1 & 0 & 1 \end{pmatrix}, \begin{pmatrix} -\frac{1}{2} & 0 & 0 \\ 0 & -\frac{1}{2} & 0 \\ 0 & 0 & \frac{1}{4} \end{pmatrix}, \begin{pmatrix} -1 & -1 & 0 \\ 1 & 2 & 1 \\ 1 & 1 & 1 \end{pmatrix} \right\}$$

Out[760]//MatrixForm=

$$\begin{pmatrix} \frac{1}{4} & \frac{3}{4} & \frac{3}{4} \\ -\frac{3}{4} & -\frac{5}{4} & -\frac{3}{4} \\ \frac{3}{4} & \frac{3}{4} & \frac{1}{4} \end{pmatrix}$$

To derive $\mathbf{A} = \mathbf{VDV}^{-1}$

$$\det(\mathbf{A} - \lambda \mathbf{I}) = 0 \Rightarrow \lambda_1 = -\frac{1}{2}, \lambda_2 = \frac{1}{4}$$

$$\text{take } \lambda_1 = -\frac{1}{2} \text{ in } (\mathbf{A} - \lambda \mathbf{I})\mathbf{x} = 0 \Rightarrow \mathbf{v}_1 = \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix}, \mathbf{v}_2 = \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}$$

$$\text{take } \lambda_2 = \frac{1}{4} \text{ in } (\mathbf{A} - \lambda \mathbf{I})\mathbf{x} = 0 \Rightarrow \mathbf{v}_3 = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$$

$$\text{s.t. } \mathbf{V} = \begin{pmatrix} -1 & -1 & 1 \\ 0 & 1 & -1 \\ 1 & 0 & 1 \end{pmatrix}$$

$$\det(\mathbf{V}) = -1 \neq 0$$

$$\mathbf{D} = \begin{pmatrix} -\frac{1}{2} & 0 & 0 \\ 0 & -\frac{1}{2} & 0 \\ 0 & 0 & \frac{1}{4} \end{pmatrix}$$

$$\mathbf{V}^{-1} = \frac{1}{\det(\mathbf{V})} \text{adj}(\mathbf{V}) = \begin{pmatrix} -1 & -1 & 0 \\ 1 & 2 & 1 \\ 1 & 1 & 1 \end{pmatrix}$$

$$\mathbf{A} = \mathbf{V}\mathbf{D}\mathbf{V}^{-1} = \begin{pmatrix} -1 & -1 & 1 \\ 0 & 1 & -1 \\ 1 & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} -\frac{1}{2} & 0 & 0 \\ 0 & -\frac{1}{2} & 0 \\ 0 & 0 & \frac{1}{4} \end{pmatrix} \cdot \begin{pmatrix} -1 & -1 & 0 \\ 1 & 2 & 1 \\ 1 & 1 & 1 \end{pmatrix}$$

$$\mathbf{e}_n = \mathbf{A}^n \mathbf{e}_1 = \mathbf{V}\mathbf{D}^n \mathbf{V}^{-1} \mathbf{e}_0$$

$$\|\mathbf{D}\|_{\infty} = \frac{1}{4} < 1$$

$$\lim_{n \rightarrow \infty} \mathbf{e}_n = \lim_{n \rightarrow \infty} \mathbf{V}\mathbf{D}^n \mathbf{V}^{-1} \mathbf{e}_0 = \mathbf{0}$$

i.e. as $n \rightarrow \infty$, $\mathbf{e}_n \rightarrow \mathbf{0}$

$$\mathbf{x}_n \rightarrow \alpha$$

So, convergence to α

Problem 7

$$\mathbf{A} = \begin{pmatrix} 1 & x_0 & x_0^2 & \dots & x_0^n \\ 1 & x_1 & x_1^2 & \dots & x_1^n \\ 1 & x_2 & x_2^2 & \dots & x_2^n \\ \dots & \dots & \dots & \dots & \dots \\ 1 & x_n & x_n^2 & \dots & x_n^n \end{pmatrix}$$

Subtracting a multiple of one column from another column in a matrix \mathbf{A} is equivalent to subtracting the corresponding multiple of one row from another row in the transpose of \mathbf{A} , denoted \mathbf{A}^T . Since the determinant of a matrix is equal to the determinant of its transpose, that is, $\det(\mathbf{A}) = \det(\mathbf{A}^T)$, these operations do not change the determinant's value.

Do column operations, $C_i = C_i - x_0 C_{i-1}$ except $i=1$, starting from last column

$$\det(\mathbf{A}) = \det \begin{pmatrix} 1 & 0 & 0 & 0 & \dots & 0 \\ 1 & x_1 - x_0 & x_1(x_1 - x_0) & x_1^2(x_1 - x_0) & \dots & x_1^{n-1}(x_1 - x_0) \\ 1 & x_2 - x_0 & x_2(x_2 - x_0) & x_2^2(x_2 - x_0) & \dots & x_2^{n-1}(x_2 - x_0) \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 1 & x_n - x_0 & x_n(x_n - x_0) & x_n^2(x_n - x_0) & \dots & x_n^{n-1}(x_n - x_0) \end{pmatrix}$$

with co-factor expansion

$$\det(\mathbf{A}) = \det \begin{pmatrix} x_1 - x_0 & x_1(x_1 - x_0) & x_1^2(x_1 - x_0) & \dots & x_1^{n-1}(x_1 - x_0) \\ x_2 - x_0 & x_2(x_2 - x_0) & x_2^2(x_2 - x_0) & \dots & x_2^{n-1}(x_2 - x_0) \\ \dots & \dots & \dots & \dots & \dots \\ x_n - x_0 & x_n(x_n - x_0) & x_n^2(x_n - x_0) & \dots & x_n^{n-1}(x_n - x_0) \end{pmatrix} =$$

$$(x_1 - x_0)(x_2 - x_0) \dots (x_n - x_0) \det \begin{pmatrix} 1 & x_1 & x_1^2 & \dots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \dots & x_2^{n-1} \\ \dots & \dots & \dots & \dots & \dots \\ 1 & x_n & x_n^2 & \dots & x_n^{n-1} \end{pmatrix} = \prod_{1 \leq i \leq n} (x_i - x_0) \det(A')$$

where A' is the Vandermonde matrix in x_1, \dots, x_n . Iterating this process on this smaller Vandermonde matrix,

eventually get $\det(A) = \prod_{1 \leq i \leq n} (x_i - x_0) \prod_{2 \leq i \leq n} (x_i - x_1) \dots \prod_{n \leq i \leq n} (x_i - x_{n-1}) = \prod_{0 \leq j < i \leq n} (x_i - x_j)$

Problem 8

let $t = e^x$

take $x(t) = \ln(t)$ in, $p_n(x(t)) = \sum_{j=0}^n c_j e^{j \ln(t)} = \sum_{j=0}^n c_j t^j = p(t)$

as $t(x) = e^x$ singular increasing on $(-\infty, \infty)$,

and x_0, \dots, x_n are distinct real points,

so t_0, \dots, t_n are also distinct points.

as $p_n(x_i) = y_i$, $i = 0, 1, \dots, n$

$p(t_i) = y_i$, $i = 0, 1, \dots, n$

$$\mathbf{A} = \begin{pmatrix} 1 & t_0 & t_0^2 & \dots & t_0^n \\ 1 & t_1 & t_1^2 & \dots & t_1^n \\ 1 & t_2 & t_2^2 & \dots & t_2^n \\ \dots & \dots & \dots & \dots & \dots \\ 1 & t_n & t_n^2 & \dots & t_n^n \end{pmatrix}, \mathbf{c} = \begin{pmatrix} c_0 \\ c_1 \\ c_2 \\ \dots \\ c_n \end{pmatrix}, \mathbf{y} = \begin{pmatrix} y_0 \\ y_1 \\ y_2 \\ \dots \\ y_n \end{pmatrix}$$

$\mathbf{Ac} = \mathbf{y}$

$$\left(\begin{array}{l} \det(A) = \prod_{0 \leq j < i \leq n} (t_i - t_j) \\ t_i \neq t_j, \text{ so } \det(A) \neq 0 \\ \text{By linear algebra, the solution exists and is unique} \end{array} \right)$$

By THM, given $n+1$ distinct t_i , there is a unique polynomial $p(t)$ of degree $\leq n$ that interpolates y_i at x_i

i.e. there is a unique choice of c_0, \dots, c_n .