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HW #5

Q1: a. Find the best fit quadratic function:

$$f(x) = f(x_1) + \frac{f(x_2) - f(x_1)}{x_2 - x_1}(x - x_1) + \frac{\frac{f(x_3) - f(x_2)}{x_3 - x_2} - \frac{f(x_2) - f(x_1)}{x_2 - x_1}}{x_3 - x_1}(x - x_1)(x - x_2)$$

$$= 0.85 + \frac{0.67 - 0.85}{35 - 30}(x - 30) + \frac{\frac{0.52 - 0.67}{40 - 35} - \frac{0.67 - 0.85}{35 - 30}}{40 - 30}(x - 30)(x - 35)$$

$$= 0.85 + (-0.036)(x - 30) + 0.0006(x - 30)(x - 35)$$

b. Find the best fit exponential function:

$$f(x) = f(x_1) \cdot \left(\frac{f(x_2)}{f(x_1)} \right)^{\frac{x}{x_2 - x_1} - \frac{x_1}{x_2 - x_1}}$$

$$= 0.85 \cdot \left(\frac{0.67}{0.85} \right)^{\frac{x}{5} - 6}$$

c. Find the interpolating polynomial:

$$P_n(x) = f(x) = f(x_0) + f'(x_0)(x - x_0) + \frac{f''(x_0)}{2!}(x - x_0)^2 + \dots + \frac{f^{(n)}(x_0)}{n!}(x - x_0)^n$$

$$= 0.85 + (-0.036)(x - 30) + 0.0006(x - 30)(x - 35) + 0 + 0 + \dots$$

$$= 0.85 - 0.036(x - 30) + 0.0006(x - 30)(x - 35)$$

[Using function
InterpN2]

d. Find the interpolating cubic spline:

[Using function
CubicSpline]

$$P_i = C_0^i + C_1^i(x - x_i) + C_2^i(x - x_i)^2 + C_3^i(x - x_i)^3(x - x_{i+1})$$

$$C_0^i = a = [0.85 \quad 0.67 \quad 0.52 \quad 0.42 \quad 0.34 \quad 0.28 \quad 0.24 \quad 0.21 \quad 0.18]$$

$$C_1^i = b = [-0.037 \quad -0.0341 \quad -0.0247 \quad -0.0173 \quad -0.0142 \quad -0.0098 \quad -0.0066 \quad -0.0035]$$

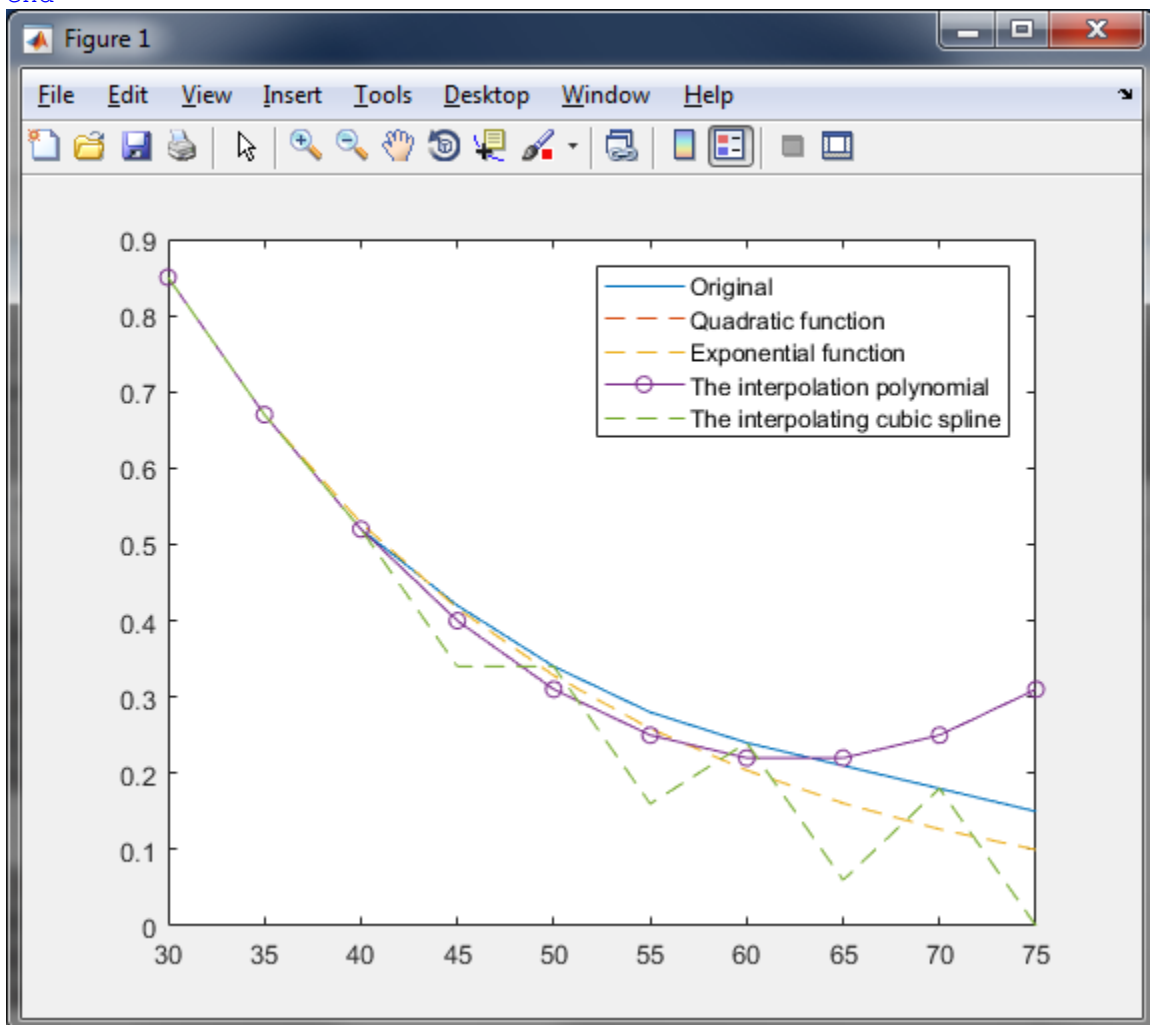
$$C_2^i = c = [0.0002 \quad 0.0014 \quad \dots]$$

$$C_3^i = d = [0.0002 \quad 0.0014 \quad \dots]$$

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%Script File for Q1 of HW 5
xvals = [30 35 40 45 50 55 60 65 70 75];
yvals = [0.85 0.67 0.52 0.42 0.34 0.28 0.24 0.21 0.18 0.15];
y1 = 0.85-0.036.*(xvals-30)+0.0006.*(x-30).*(x-35);
y2 = 0.85.*(0.67/0.85).^(x./5-6);
y3 = y1;
for n = 3
    x = [30 35 40 45 50 55 60 65 70 75];
    y = [0.85 0.67 0.52 0.42 0.34 0.28 0.24 0.21 0.18 0.15];
    [a,b,c,d] = CubicSpline(x,y,2,0,0);
    svals = pwCEval(a,b,c,d,x,xvals);
    figure
    plot(xvals,yvals,xvals,y1,'--',xvals,y2,'--',xvals,y3,'o-
',xvals,svals,'--')
    legend('Original','Quadratic function','Exponential function','The
interpolation polynomial','The interpolating cubic spline')
end

```



On the graph, the Quadratic function and the interpolation polynomial are same. The error of all the fixes or interpolation become larger when x (the distance) become larger.

Q2

```
function F = CSInterp(f)
% F = CSInterp(f) [Fast solution]
% f is a column n vector and n = 2m.
% F.a is a column m+1 vector and F.b is a column m-1 vector so that if
% tau = (0:n-1)'*pi/m, then
%
%      f = F.a(1)*cos(0*tau) +...+ F.a(m+1)*cos(m*tau) +
%          F.b(1)*sin(tau)      +...+ F.b(m-1)*sin((m-1)*tau)

n = length(f);
m = n/2;
tau = fft((pi/m)*(0:n-1)');
P = [];
for j=0:m
    if(j==1 || j==m+1)
        P = [P cos(j*tau)]/2;
    else
        P = [P cos(j*tau)];
    end
end
for j=1:m-1, P = [P sin(j*tau)+1]; end
y = P\f;
F = struct('a',y(1:m+1), 'b',y(m+2:n));
```

```
function F = CSInterp0(f)
% F = CSInterp(f) [Original solution]
% f is a column n vector and n = 2m.
% F.a is a column m+1 vector and F.b is a column m-1 vector so that if
% tau = (0:n-1)'*pi/m, then
%
%      f = F.a(1)*cos(0*tau) +...+ F.a(m+1)*cos(m*tau) +
%          F.b(1)*sin(tau)      +...+ F.b(m-1)*sin((m-1)*tau)

n = length(f);
m = n/2;
tau = (pi/m)*(0:n-1)';
P = [];
for j=0:m,    P = [P cos(j*tau)]; end
for j=1:m-1, P = [P sin(j*tau)]; end
y = P\f;
F = struct('a',y(1:m+1), 'b',y(m+2:n));
```

```

function Fvals = CSEval(F,T,tvals)
% Fvals = CSEval(F,T,tvals)
% F.a is a length m+1 column vector, F.b is a length m-1 column vector,
% T is a positive scalar, and tvals is a column vector.
% If
%   F(t) = F.a(1) + F.a(2)*cos((2*pi/T)*t) +...+
F.a(m+1)*cos((2*m*pi/T)*t) +
%           F.b(1)*sin((2*pi/T)*t) +...+ F.b(m-
1)*sin((2*m*pi/T)*t)
%
% then Fvals = F(tvals).

Fvals = zeros(length(tvals),1);
tau = (2*pi/T)*tvals;
for j=0:length(F.a)-1, Fvals = Fvals + F.a(j+1)*cos(j*tau); end
for j=1:length(F.b),   Fvals = Fvals + F.b(j)*sin(j*tau); end


% Script File: Pallas
% Plots the trigonometric interpolant of the Gauss Pallas data.

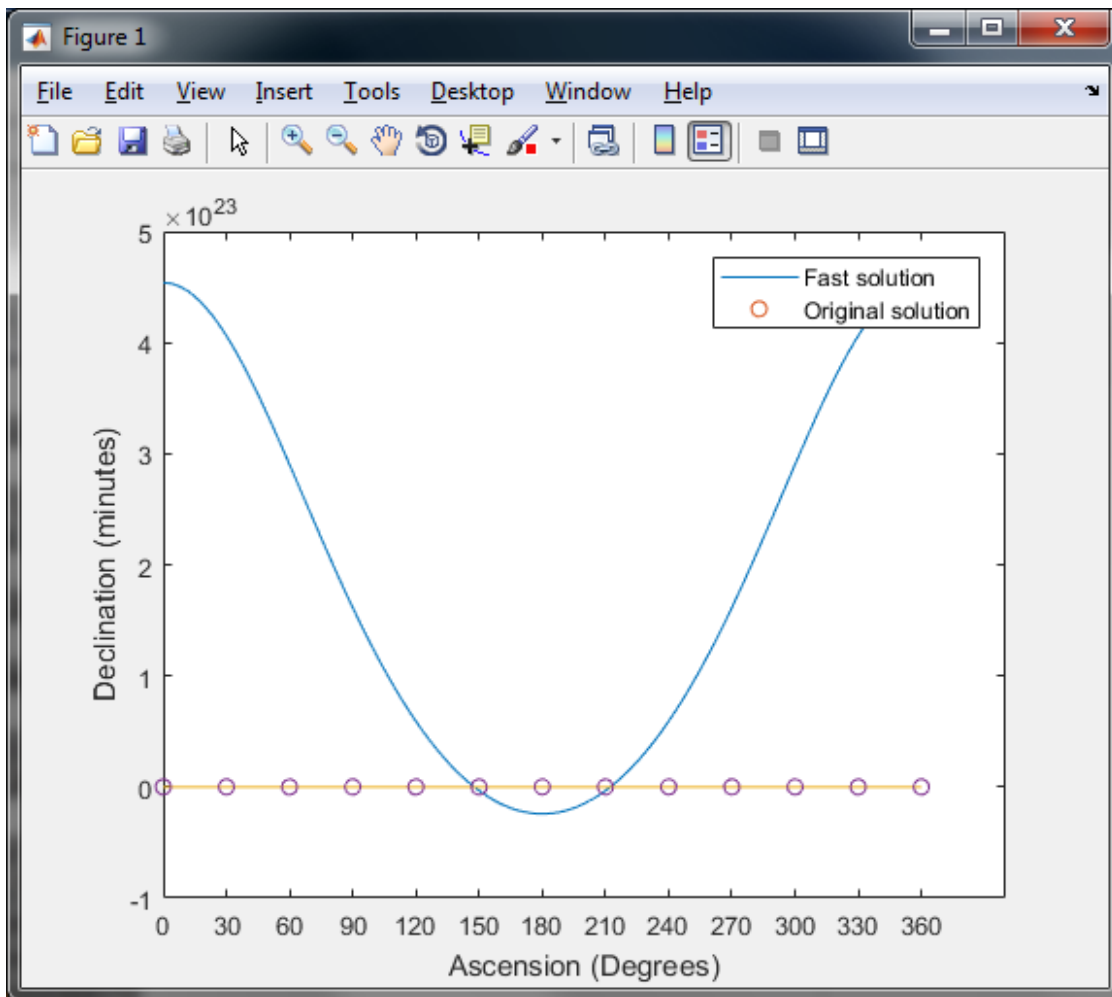
A = linspace(0,360,13)';
D = [ 408 89 -66 10 338 807 1238 1511 1583 1462 1183 804 408]';

Avals = linspace(0,360,200)';
F = CSInterp(D(1:12));
Fvals = CSEval(F,360,Avals);

F0 = CSInterp0(D(1:12));
Fvals0 = CSEval(F0,360,Avals);

plot(Avals,Fvals,A,D,'o',Avals,Fvals0,A,D,'o')
legend('Fast solution','Original solution')
set(gca,'xTick',linspace(0,360,13))
xlabel('Ascension (Degrees)')
ylabel('Declination (minutes)')

```



Q3

Reference: Gene H. Golub, Charles F. Van Loan, "Matrix Computations",

https://books.google.com/books?id=5U-l8U3P-VUC&pg=PT267&lpg=PT267&dq=least+squares+consider+the+problem+where+p%3D1,2,+or+infinite.+Suppose+b1%3E%3Db2%3E%3Db3&source=bl&ots=7_CwPi_Qes&sig=AdHOn8-CGzckpQGRawzOrx9w7Jo&hl=en&sa=X&ved=0ahUKEwiNztHj29zXAhXCk-AKHZ9UCTgQ6AEILjAB#v=onepage&q&f=false, 5.3 The Full-Rank Least Squares

Problem.

We say the system $Ax = b$ is overdetermined. An overdetermined system has no exact solution since " b " must be an element of $\text{ran}(A)$, a proper subspace of \mathbb{R}^m . This suggests that we strive to minimize $\|Ax - b\|_p$ for some suitable choice of p .

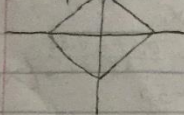
- Minimization in 1-norm and infinity-norm is complicated by the fact that the function $f(x) = \|Ax - b\|_p$ is not differentiable for these value of p .

- The 2-norm is under orthogonal transformation. This means that we can seek an orthogonal Q such that the equivalent problem of minimizing $\|(Q^T A)x - (Q^T b)\|_2$ is "easy" to solve.

Q3. $\min_{x \in \mathbb{R}} \left\| \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} x - \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \right\|_p$,

$p=1$,

$\|e\|_1$



$\min \|Ax - b\|_1$

$\min_{x \in \mathbb{R}} \left\| \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} x - \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \right\|_1$

$\Rightarrow \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} x = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$

$x_1 = b_1, x_2 = b_2, x_3 = b_3$

So, $x_{opt} = b_2$

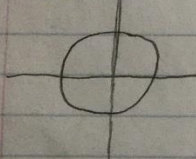
$\min \sum_{i=1}^n y_i$

Subject to $y \leq A \cdot b \leq y$

We know that:
 $b_1 \geq b_2 \geq b_3$

$p=2$,

$\|e\|_2$



$\min \|Ax - b\|_2^2$

$\min \sum_{i=1}^n y_i^2$

Base $A^T \cdot A \cdot x = b \cdot A^T$

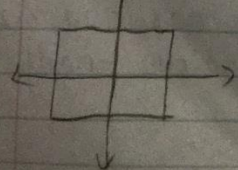
$\begin{bmatrix} 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} x = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \end{bmatrix}$

$3x = b_1 + b_2 + b_3$

$3x_{opt} = (b_1 + b_2 + b_3)$
 $= \frac{(b_1 + b_2 + b_3)}{3}$

$p=\infty$,

$\|e\|_\infty$



$\min \|Ax - b\|_\infty$

$\min y$ or $\max_k |y_k|$

$x_{opt} = \frac{b_1 + b_3}{2}$

$\|e\|_\infty \leq \|e\|_2$

$\|e\|_1 \Leftrightarrow \|e\|_\infty$
 $\frac{b_2}{1} \Leftrightarrow \frac{b_1 + b_3}{2}$