Practice Session for week 2

In below table, CPU temperature measurements are shown. Calculate true error and relative true error.

Samples	Temperature (C) Truth Observed			
			True Error	Relative True Error
1	40,58	40,32	40,58-40,32=0,26	0,26/40,58 = %0,6407
2	40,65	40,77	40,65-40,77=-0,12	-0,12/40,65=-%0,2952
3	41,34	41,36	41,34-41,36=-0,02	-0,02/41,34=-%0,0484
4	41,91	42,20	41,91-42,20=-0,29	-0,29/41,91=-%0,6920

Derivative of a function f at a point x is defined by the equation;

$$f'(x) = \lim_{h \to 0} \frac{f(x+h) - f(x)}{h}$$

$$f'(x) \approx \frac{f(x+h) - f(x)}{h}$$

for small h values.

• If
$$f(x) = x^3 + 3x^2$$
 and $h = 0.6$

Find approximate value of f(3)

Find true value of f'(3)

Find true error, relative true error.

$$f'(x) \approx \frac{f(x+h)-f(x)}{h}$$
 $f'(x) \approx \frac{f(3+0.6)-f(3)}{0.6} \approx \frac{f(3.6)-f(3)}{0.6}$

$$f'(x) \approx \frac{(46,656+38,88)-(54)}{0,6} \approx \frac{31,536}{0,6} \approx 52,56$$
 is approximate value.

$$f'(x) = 3x^2 + 6x$$
 is derivative function of function $f(x)$

$$f'(3) = 27 + 18 = 45$$
 is true value.

True error = true value – approximate value = 45-52,56=-7,56Relative true error = true error / true value = -7,56/45=-0,168(%-16,8)

• Compute approximate error and relative approximate error for f'(3) using h = 0.6 and h = 0.4

Approximate error = present approximation – previous approximation

$$f'(x) \approx \frac{f(x+h)-f(x)}{h}$$
 $f'(x) \approx \frac{f(3+0,4)-f(3)}{0,4} \approx \frac{f(3,4)-f(3)}{0,4}$

$$f'(x) \approx \frac{(39,304+34,68)-(54)}{0,4} \approx \frac{73,984-54}{0,4} \approx \frac{19,984}{0,4} \approx 49,96$$
 is present approximation.

Approximate error = 49,96-52,56-2,66-2

Relative approximation error = approximation error / present approximation

$$=\frac{-2,56}{49,96}=-0,051241(\%5,1241)$$

Error measurement on floating point format.

Floating decimal point scientific form: sign x mantisa x 10^{exponent}

$$\sigma x m x 10^e$$
 $-2,5678 x 10^2$ $\sigma = -1$ $m = 2,5678$ $e = 2$

 $[(1)_2 < m < (10)_2]$ 1 is not stored as it is always given to be 1.

Let we have a floating point format ten bit word;

	sn	se	e	e	e	e	m	m	m	m	
--	----	----	---	---	---	---	---	---	---	---	--

- * 1 bit for sign of number
- * 1 bit for sign of exponent
- * 4 bit for exponent
- * 4 bit for mantissa

Display $(123,9631)_{10}$ floating point number for the given format and calculate the true error while representation if it occurs.

$$(123)_{10} = (1111011)_2 \qquad (0*2^7) + (1*2^6) + (1*2^5) + (1*2^4) + (1*2^3) + (0*2^2) + (1*2^1) + (1*2^0) + (0.9631)_{10} = (?)_2$$

	Result	Integer part	Fractional part
0,9631*2	1,9262	1	0,9262
0,9262*2	1,8524	1	0,8524
0,8524*2	1,7048	1	0,7048
0,7048*2	1,4096	1	0,4096

0,4096*2	0,8192	0	0,8192
0,8192*2	1,6384	1	0,6384
0,6384*2	1,2768	1	0,2768
0,2768*2	0,5536	0	0,5536
0,5536*2	1,1072	1	0,1072

 $1111011, 111101101 = 1, 111011111101101 * 2^6$

sn	se	e	e	e	e	m	m	m	m
0	0	0	1	1	0	1	1	1	0

$$1,1110*2^6 = (1111000)_2 = 120$$

True error = 123,9631 - 120 = 3,9631

8 – bit mantissa example for the same question

sn	se	e	e	e	e	m	m	m	m	m	m	m	m
0	0	0	1	1	0	1	1	1	0	1	1	1	1

 $1,111011111*2^6 = (1111011,11,)_2 = 123,75$

True error = 123,9631 - 123,75 = 0,2131

Practice Session for week 3

Use Naive Gauss Elimination to solve the system of linear equations.

$$x_1 + 2x_2 + x_3 = -8$$

$$2x_1 - 2x_2 - 3x_3 = 0$$

$$-x_1 + x_2 + 2x_3 = 3$$

$$\begin{bmatrix} 1 & 2 & 1 \\ 2 & -2 & -3 \\ -1 & 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -8 \\ 0 \\ 3 \end{bmatrix}$$

Add -2 times row1 to row2

$$\begin{bmatrix} 1 & 2 & 1 \\ 0 & -6 & -5 \\ -1 & 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -8 \\ 16 \\ 3 \end{bmatrix}$$

Add -1 times row1 to row3

$$\begin{bmatrix} 1 & 2 & 1 \\ 0 & -6 & -5 \\ 0 & 3 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -8 \\ 16 \\ -5 \end{bmatrix}$$

Add (1/2) times row2 to row3

$$\begin{bmatrix} 1 & 2 & 1 \\ 0 & -6 & -5 \\ 0 & 0 & 0.5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -8 \\ 16 \\ 3 \end{bmatrix}$$

Now elimination stops.

$$0.5x_3 = 3$$

$$x_3 = 6$$

$$-6x_2 - 5 * 6 = 16$$

$$x_2 = -\frac{23}{3}$$

$$x_1 + 2x_2 + x_3 = -8$$

$$x_1 = \frac{4}{3}$$

Solve same linear system using LU decomposition.

Ax=B

Apply gauss elimination **only matrix A**.

$$A = \begin{bmatrix} 1 & 2 & 1 \\ 2 & -2 & -3 \\ -1 & 1 & 2 \end{bmatrix}$$

 $A = \begin{bmatrix} 1 & 2 & 1 \\ 2 & -2 & -3 \\ -1 & 1 & 2 \end{bmatrix}$ same operation as gauss elimination $U = \begin{bmatrix} 1 & 2 & 1 \\ 0 & -6 & -5 \\ 0 & 0 & 0.5 \end{bmatrix}$

Now find [L] matrix

$$\begin{bmatrix} 1 & 0 & 0 \\ l_{21} & 1 & 0 \\ l_{31} & l_{32} & 1 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 & 0 \\ l_{21} & 1 & 0 \\ l_{31} & l_{32} & 1 \end{bmatrix} \qquad l_{21} = \frac{a_{21}}{a_{11}}, \ l_{31} = \frac{a_{31}}{a_{11}}, \ l_{32} = \frac{a_{32}}{a_{22}}$$

$$L = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -1 & 1 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -1 & -0.5 & 1 \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = \begin{bmatrix} -8 \\ 0 \\ 3 \end{bmatrix}$$

$$Z = \begin{bmatrix} -8 \\ 16 \\ 3 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2 & 1 \\ 0 & -6 & -5 \\ 0 & 0 & 0.5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -8 \\ 16 \\ 3 \end{bmatrix}$$

$$0.5x_3 = 3$$

$$x_3 = 6$$

$$-6x_2 - 5 * 6 = 16$$

$$x_2 = -\frac{23}{3}$$

$$x_1 + 2x_2 + x_3 = -8$$

$$x_1 = \frac{4}{3}$$

Details of LU decomposition.

$$Ux=L^{-1}B$$

$$x=U^{-1}L^{-1}B$$

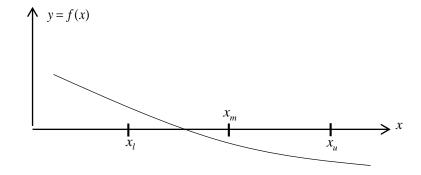
$$x=L^{-1}Z$$

Practice Session for week 4

Example of Bisection Method

A function f(x) is defined as $f(x) = x^2 - 4x - 5$. For y = f(x) = 0 estimate a root of this function using Bisection method. Use [0,48] as initial estimation range points and use absolute relative approximate error notation for error of estimated root at each iteration.

Recall



$$x_m = \frac{x_l + x_u}{2}$$

$$|\mathbf{E}| = \left| \frac{x_m^{new} - x_m^{old}}{x_m^{new}} \right| *100$$

$$f(x_l) * f(x_m) < 0$$
 the root lies between x_l and x_m

$$f(x_l) * f(x_m) > 0$$
 the root lies between x_m and x_u

$$f(x_l)*f(x_m)=0$$
 x_m is root

$$\begin{aligned} x_l &= 0 & f\left(x_l\right) = f(0) = 0 - 0 - 5 = -5 \\ x_u &= 48 & f\left(x_u\right) = f\left(48\right) = 2304 - 192 - 5 = 2107 & x_m = (0 + 48) / 2 = 24 \\ x_m &= 24 & f\left(x_m\right) = f\left(24\right) = 576 - 96 - 5 = 475 \\ f\left(x_l\right) * f\left(x_m\right) < 0 & the root lies between \ x_l \ and \ x_m \ so \ new \ estimation \ range \left[0,24\right] \\ x_l &= 0 & f\left(x_l\right) = f(0) = 0 - 0 - 5 = -5 \\ x_u &= 24 & f\left(x_u\right) = f\left(24\right) = 576 - 96 - 5 = 475 & x_m = (0 + 24) / 2 = 12 \\ x_m &= 12 & f\left(x_m\right) = f\left(12\right) = 144 - 48 - 5 = 91 \\ f\left(x_l\right) * f\left(x_m\right) < 0 & the \ root lies between \ x_l \ and \ x_m \ so \ new \ estimation \ range \left[0,12\right] \\ x_l &= 0 & f\left(x_l\right) = f\left(0\right) = 0 - 0 - 5 = -5 \\ x_u &= 12 & f\left(x_u\right) = f\left(12\right) = 144 - 48 - 5 = 91 & x_m = (0 + 12) / 2 = 6 \\ x_m &= 6 & f\left(x_m\right) = f\left(6\right) = 36 - 24 - 5 = 7 & x_m = (0 + 6) / 2 = 3 \\ x_l &= 0 & f\left(x_l\right) = f\left(0\right) = 0 - 0 - 5 = -5 \\ x_u &= 6 & f\left(x_u\right) = f\left(6\right) = 36 - 24 - 5 = 7 & x_m = (0 + 6) / 2 = 3 \\ x_m &= 3 & f\left(x_m\right) = f\left(3\right) = 9 - 12 - 5 = -8 \\ f\left(x_l\right) * f\left(x_m\right) > 0 & the \ root lies between \ x_m \ and \ x_u \ so \ new \ estimation \ range \left[3,6\right] \\ x_l &= 3 & f\left(x_l\right) = f\left(3\right) = 9 - 12 - 5 = -8 \\ x_u &= 6 & f\left(x_u\right) = f\left(6\right) = 36 - 24 - 5 = 7 & x_m = (3 + 6) / 2 = 4,5 \\ x_m &= 4,5 & f\left(x_m\right) = f\left(4,5\right) = 20,25 - 18 - 5 = -2,75 \\ f\left(x_l\right) * f\left(x_m\right) > 0 & the \ root lies between \ x_m \ and \ x_u \ so \ new \ estimation \ range \left[4.5,6\right] \end{aligned}$$

Table - Estimation of root for initial range given at each iteration. Iteration will be continued until root is estimated with an acceptable error.

Iteration	x_l	x_u	\mathcal{X}_m	$f(x_l)$	$f(x_u)$	$f(x_m)$	E
1	0	48	24	-5	2107	475	-
2	0	24	12	-5	475	91	%100
3	0	12	6	-5	91	7	%100
4	0	6	3	-5	7	-8	%100
5	3	6	4,5	-8	7	-2,75	%33,33

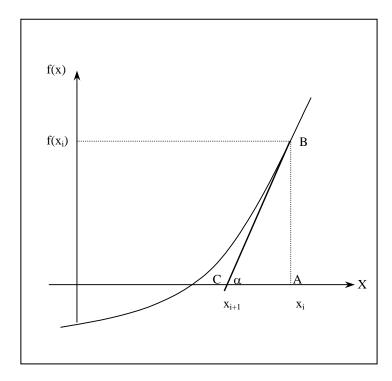
2500 y=f(x)x=0 x=3 2000 x = 4.5x=6 x=12 1500 x=24 x=48 1000 500 -500 --10 10 20 30 40 50

Figure - Bisection lines for given non-linear equation.

Example of Newton Raphson Method

A function f(x) is defined as $f(x) = x^3 - 10x^2 + 100$. For y = f(x) = 0 estimate a root of this function using Newton-Raphson method. Initial guess value of root x_0 is 15 and use absolute relative approximate error notation for error of estimated root.

Recall



$$\tan(\alpha) = \frac{AB}{AC}$$

$$f'(x_i) = \frac{f(x_i)}{x_i - x_{i+1}}$$

$$x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)}$$

Gist of prediction of X_{i+1}

$$|E| = \left| \frac{x_{i+1} - x_i}{x_{i+1}} \right| *100$$

Function :
$$f(x) = x^3 - 10x^2 + 100$$

Derivative of function : $f'(x) = 3x^2 - 20x$

$$x_1 = 15 - \frac{f(15)}{f'(15)} = 15 - \frac{3375 - 2250 + 100}{675 - 300} = 15 - \frac{1225}{375} = 15 - 3, 26 = 11, 74$$

$$x_2 = 11,74 - \frac{f(11,74)}{f'(11,74)} = 11,74 - \frac{1618,09 - 1378,27 + 100}{413,48 - 234,8} = 11,74 - \frac{339,82}{178,68} = 11,74 - 1,89 = 9,85$$

$$x_3 = 9.85 - \frac{f(9.85)}{f'(9.85)} = 9.85 - \frac{955.67 - 970.22 + 100}{291.06 - 197} = 9.85 - \frac{85.45}{94.06} = 9.85 - 1.89 = 8.94$$

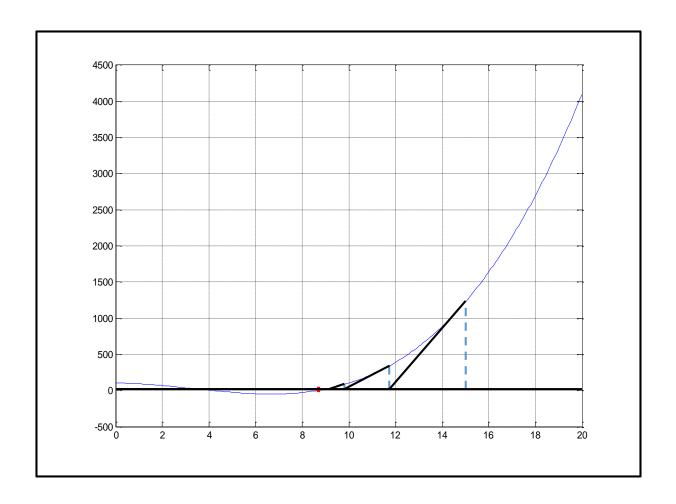
$$x_4 = 8,94 - \frac{f(8,94)}{f'(8,94)} = 8,94 - \frac{714,51 - 799,23 + 100}{239,77 - 178,8} = 8,94 - \frac{15,28}{60,97} = 8,94 - 0,25 = 8,69$$

$$x_5 = 8,69 - \frac{f(8,69)}{f'(8,69)} = 8,69 - \frac{656,23 - 755,16 + 100}{226,54 - 173,8} = 8,69 - \frac{1,07}{52,74} = 8,69 - 0,02 = 8,67$$

Table - Estimation of root for the function described in Q2 at each iteration. Iteration will be continued until root is estimated with an acceptable error.

Iteration	х	f(x)	f'(x)	E
0	$x_0 = 15$	1225	375	-
1	$x_1 = 11,74$	339,82	178,68	%27,76
2	$x_2 = 9,85$	85,45	94,06	%19,18
3	$x_3 = 8,94$	15,28	60,97	%10,17
4	$x_4 = 8,69$	1,07	52,74	%2,87
5	$x_5 = 8,67$	-	-	%0,23

Figure – Graphical analysis of Newton-Raphson method for given non-linear equation.



Practice Session for week 5

Direct Method Example

General formula for an n th order polynomial can be written as $f(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_n x^n$ for (n+1) data points. Data points :

$$(x_0, y_0), (x_1, y_1), \dots (x_n, y_n)$$

Direct Method with linear interpolation

For given data table below model the linear interpolating polynomial and estimate the function value at given point.

X	5	12	6	9
f(x)	28	63	?	?

$$x_1 = 5$$
 $f(5) = 28$ $f(5) = a_0 + a_1 5$ $28 = a_0 + a_1 5$
 $x_2 = 12$ $f(12) = 63$ $f(12) = a_0 + a_1 12$ $63 = a_0 + a_1 12$

$$f(5) = a_0 + a_1 5$$

$$28 = a_0 + a_1 5$$

$$a_0 = 3$$

$$x_2 = 12$$

$$f(12) = 63$$

$$f(12) = a_0 + a_1 12$$

$$63 = a_0 + a_1 12$$

$$a_1 = 5$$

$$f(x) = 3 + 5x$$

$$f(x)=3+5x$$
 $f(6)=3+5\cdot 6=33$

$$f(9) = 3 + 5 \cdot 9 = 48$$

Direct Method with quadratic interpolation

For given data table below model the quadratic interpolating polynomial and estimate the function value at given point.

X	5	12	20	17
f(x)	28	63	80	?

$$x_1 = 5$$
 $f(5) = 2$

$$f(5) = 28$$
 $f(5) = a_0 + a_1 + a_2 + a_2 = 25$

$$28 = a_0 + a_1 + a_2 = 25$$

$$x_2 = 12$$

$$f(12) = 63$$

$$x_2 = 12$$
 $f(12) = 63$ $f(12) = a_0 + a_1 12 + a_2 144$ $63 = a_0 + a_1 12 + a_2 144$

$$63 = a_0 + a_1 12 + a_2 144$$

$$x_2 = 20$$

$$f(20) = 80$$

$$x_2 = 20$$
 $f(20) = 80$ $f(20) = a_0 + a_1 20 + a_2 400$ $80 = a_0 + a_1 20 + a_2 400$

$$80 = a_0 + a_1 20 + a_2 400$$

$$a_0 = -8.5$$

$$a_1 = 8,25$$

$$f(x) = -8,5+8,25x-0,19x^2$$

$$a_2 = -0.19$$

$$f(17) = -8,5+8,25\cdot17-0,19\cdot(17\cdot17) = -8,25+140,25-54,91=77,09$$

Lagrangian Method Example

General form of Lagrangian interpolating polynomial;

$$f_n(x) = \sum_{i=0}^n L_i(x) f(x_i) \qquad \text{for (n+1) data points.} \quad L_i(x) = \prod_{\substack{j=0 \ j \neq i}}^n \frac{x - x_j}{x_i - x_j}$$

Data points : $(x_0, y_0), (x_1, y_1), \dots (x_n, y_n)$

Lagrangian Method with quadratic interpolation

For given data table below model the quadratic interpolating polynomial and estimate the function value at given point.

X	5	9	18	13
f(x)	21	34	49	?

$$f_2(x) = L_0(x)f(x_0) + L_1(x)f(x_1) + L_2(x)f(x_2)$$

$$f_2(x) = L_0(x) \cdot 21 + L_1(x) \cdot 34 + L_2(x) \cdot 49$$

$$L_{0}(x) = \left(\frac{x - x_{1}}{x_{0} - x_{1}}\right) \left(\frac{x - x_{2}}{x_{0} - x_{2}}\right) = \left(\frac{x - 9}{5 - 9}\right) \left(\frac{x - 18}{5 - 18}\right) = \left(\frac{9 - x}{4}\right) \left(\frac{18 - x}{13}\right)$$

$$L_{1}(x) = \left(\frac{x - x_{0}}{x_{1} - x_{0}}\right) \left(\frac{x - x_{2}}{x_{1} - x_{2}}\right) = \left(\frac{x - 5}{9 - 5}\right) \left(\frac{x - 18}{9 - 18}\right) = \left(\frac{x - 5}{4}\right) \left(\frac{18 - x}{9}\right)$$

$$L_{2}(x) = \left(\frac{x - x_{0}}{x_{2} - x_{0}}\right) \left(\frac{x - x_{1}}{x_{2} - x_{1}}\right) = \left(\frac{x - 5}{18 - 5}\right) \left(\frac{x - 9}{18 - 9}\right) = \left(\frac{x - 5}{13}\right) \left(\frac{x - 9}{9}\right)$$

$$f_{2}(x) = \left(\frac{9 - x}{4}\right) \left(\frac{18 - x}{13}\right) \cdot 21 + \left(\frac{x - 5}{4}\right) \left(\frac{18 - x}{9}\right) \cdot 34 + \left(\frac{x - 5}{13}\right) \left(\frac{x - 9}{9}\right) \cdot 49$$

$$f_{2}(13) = (-1) \left(\frac{5}{13}\right) \cdot 21 + (2) \left(\frac{5}{9}\right) \cdot 34 + \left(\frac{8}{13}\right) \left(\frac{4}{9}\right) \cdot 49 = -8,08 + 37,78 + 13,40 = 43,1$$

Newton's Method Example

General form of Newton's interpolating polynomials;

Given
$$(n+1)$$
 data points, $(x_0, y_0), (x_1, y_1), \dots, (x_{n-1}, y_{n-1}), (x_n, y_n)$ as

$$f_n(x) = b_0 + b_1(x - x_0) + \dots + b_n(x - x_0)(x - x_1)\dots(x - x_{n-1})$$

where

$$\begin{aligned} b_0 &= f[x_0] \\ b_1 &= f[x_1, x_0] \\ b_2 &= f[x_2, x_1, x_0] \\ &\vdots \\ b_{n-1} &= f[x_{n-1}, x_{n-2},, x_0] \\ b_n &= f[x_n, x_{n-1},, x_0] \end{aligned}$$

and

$$f[x_n, x_{n-1},, x_0] = \frac{f[x_n, x_{n-1},, x_1] - f[x_{n-1}, x_{n-2},, x_0]}{x_n - x_0}$$

For given data table below model the quadratic interpolating polynomial and estimate the function value at given point.

X	5	9	18	13
f(x)	21	34	49	?

$$f_2(x) = b_0 + b_1(x - x_0) + b_2(x - x_0)(x - x_1)$$

$$b_0 = f[x_0] = f(x_0) = f(5) = 21$$

$$b_1 = f[x_1, x_0] = \frac{f(x_1) - f(x_0)}{x_1 - x_0} = \frac{f(9) - f(5)}{9 - 5} = \frac{34 - 21}{4} = \frac{13}{4} = 3,25$$

$$b_2 = f[x_2, x_1, x_0] = \frac{f[x_2, x_1] - f[x_1, x_0]}{x_2 - x_0} = \frac{\frac{f(x_2) - f(x_1)}{x_2 - x_1} - \frac{f(x_1) - f(x_0)}{x_1 - x_0}}{x_2 - x_0}$$

$$=\frac{\frac{f(18)-f(9)}{18-9}-\frac{f(9)-f(5)}{9-5}}{18-5}=\frac{\frac{49-34}{18-9}-\frac{34-21}{9-5}}{18-5}=\frac{\frac{15}{9}-\frac{13}{4}}{13}=\frac{60-117}{13\cdot 36}=-\frac{57}{468}=-0,12$$
 So,

$$f_2(x) = 21+3,25(x-5)-0,12(x-5)(x-9)$$

 $f_2(13) = 21+3,25(13-5)-0,12(13-5)(13-9)$
 $f_2(13) = 21+26-3,84$

$$f_2(13) = 43,16$$

$$F(t) = F(20) + \frac{F(30) - F(20)}{30 - 20} (t - 20) = \dots$$

Practice Session for week 8

Linear regression with least square method

Minimize error;
$$\sum_{i=1}^{n} (y_i - (ax_i + b))^2$$
 n=number of data points

$$\frac{\partial_{err}}{\partial_a} = -2\sum_{i=1}^n x_i(y_i - ax_i - b) = 0$$

$$\frac{\partial_{err}}{\partial_b} = -2\sum_{i=1}^n (y_i - ax_i - b) = 0$$

Rewrite,

$$a\sum_{i} x_{i}^{2} + b\sum_{i} x_{i} = \sum_{i} (x_{i}y_{i})$$
$$a\sum_{i} x_{i} + (b*n) = \sum_{i} y_{i}$$

Matrix form,

$$\begin{bmatrix} n & \sum x_i \\ \sum x_i & \sum x_i^2 \end{bmatrix} \cdot \begin{bmatrix} b \\ a \end{bmatrix} = \begin{bmatrix} \sum y_i \\ \sum (x_i y_i) \end{bmatrix}$$

X	y=f(x)
0	8,4121
1	7,4882
2	6,4038
3	7,0530
4	6,6072
5	5,3039
6	5,9597
7	5,4933
8	5,7356
9	5,9598

$$n = 10$$

$$\sum x_i = 45$$

$$\sum x_i^2 = 285$$

$$\sum y_i = 64,4166$$
$$\sum (x_i y_i) = 268,1374$$

$$\begin{bmatrix} 10 & 45 \\ 45 & 285 \end{bmatrix} \cdot \begin{bmatrix} b \\ a \end{bmatrix} = \begin{bmatrix} 64,4166 \\ 268,1374 \end{bmatrix}$$

$$b = 7,6275$$
, $a = -0,2635$

$$y = ax + b = -0,2635x + 7,6273$$