Taylor Series:	$f(x_{i+1}) = f(x_i) + f'(x_i)(x_{i+1} - x_i) + \frac{f''(x_i)}{2!}(x_{i+1} - x_i)^2$
	$+\frac{f'''(x_i)}{3!}(x_{i+1}-x_i)^3+\cdots+\frac{f^{(n+1)}(\xi)}{(n+1)!}(x_{i+1}-x_i)^{n+1}$

<b>Root finding</b> : False Position:		
$x_r = x_u -$	$\frac{f(x_u)(x_l - x_u)}{f(x_l) - f(x_u)}$	

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

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

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

$$x_{i+1} = x_i - \frac{f(x_i)f'(x_i)}{[f'(x_i)]^2 - f(x_i)f''(x_i)}$$

System of linear algebraic equations: [A]  $\{X\} = \{B\}$ . Set [L] [U] = [A] with [L] and [U] as follows:

LU (Doolittle) decomposition:

$$[L] = \begin{bmatrix} 1 & 0 & 0 \\ l_{21} & 1 & 0 \\ l_{31} & l_{32} & 1 \end{bmatrix}; \quad [U] = \begin{bmatrix} u_{11} & u_{12} & u_{13} \\ 0 & u_{22} & u_{23} \\ 0 & 0 & u_{33} \end{bmatrix}$$
Gauss-Seidel: sufficient condition for convergence: 
$$|a_{ii}| > \sum_{j=1; j \neq i}^{n} |a_{ij}|$$

Crout decomposition:

$$\begin{bmatrix} L \end{bmatrix} = \begin{bmatrix} l_{11} & 0 & 0 \\ l_{21} & l_{22} & 0 \\ l_{31} & l_{32} & l_{33} \end{bmatrix}; \quad \begin{bmatrix} U \end{bmatrix} = \begin{bmatrix} 1 & u_{12} & u_{13} \\ 0 & 1 & u_{23} \\ 0 & 0 & 1 \end{bmatrix}$$

$$|a_{ii}| > \sum_{j=1; j \neq i}^{n} |a_{ij}|$$

Polynomial least squares regression analysis:

$$y = a_0 + a_1 x + a_2 x^2 + \dots$$

2x2 matrix (only  $a_0, a_1$ ): linear,

3x3 matrix: quadratic, etc.

**Ouadratic Regression:** 

Error Analysis for linear regression:  
Standard error of the estimate: 
$$s_{y/x} = \sqrt{\frac{S_r}{n-2}}$$
  
 $S_r = \sum_{i=1}^{n} e_i^2 = \sum_{i=1}^{n} (y_i - a_0 - a_1 x_i)^2$ 

$$\begin{bmatrix} n & \sum x_{i} & \sum x_{i}^{2} \\ \sum x_{i} & \sum x_{i}^{2} & \sum x_{i}^{3} \\ \sum x_{i}^{2} & \sum x_{i}^{3} & \sum x_{i}^{4} \end{bmatrix} \begin{Bmatrix} a_{0} \\ a_{1} \\ a_{2} \end{Bmatrix} = \begin{Bmatrix} \sum y_{i} \\ \sum x_{i} y_{i} \\ \sum x_{i}^{2} y_{i} \end{Bmatrix};$$

where 
$$\sum_{i=1}^{n} = \sum_{j=1}^{n}$$

Non linear regression ... (fill in)

Lagrange interpolating polynomials: 
$$f_n(x) = \sum_{i=0}^n L_i(x) f(x_i)$$
; where  $L_i(x) = \prod_{j=0, j \neq i}^n \frac{x - x_j}{x_i - x_j}$ 

$$f_2(x) = \frac{(x - x_1)(x - x_2)}{(x_0 - x_1)(x_0 - x_2)} f(x_0) + \frac{(x - x_0)(x - x_2)}{(x_1 - x_0)(x_1 - x_2)} f(x_1) + \frac{(x - x_0)(x - x_1)}{(x_2 - x_0)(x_2 - x_1)} f(x_2)$$
Newton interpolating polynomials: ....(fill in)

**Numerical integration**:  $\int_a^b f(x) dx$ : trapezoidal:  $I = \frac{h}{2} \left( f(x_0) + \sum_{i=1}^{n-1} f(x_i) + f(x_n) \right)$ 

Simpson's 3/8 ....(fill in)

$$I = \frac{h}{3} \left( f(x_0) + 4 \sum_{i=1,3,5...}^{n-1} f(x_i) + 2 \sum_{j=2,4,6...}^{n-2} f(x_j) + f(x_n) \right)$$

Gauss quadrature (Gauss-Legendre polynomials):  $\int_{a}^{b} g(x)dx = \int_{-1}^{1} f(x)dx = c_{0}f(x_{0}) + c_{1}f(x_{1}) + ... + c_{n-1}f(x_{n-1})$ 

Romberg: ....(fill in)

Numerical differentiation using finite divided difference formulas:

Forward: 
$$f'(x_i) = (f(x_{i+1}) - f(x_i))/h + O(h)$$
;  $f'(x_i) = (-f(x_{i+2}) + 4f(x_{i+1}) - 3f(x_i))/(2h) + O(h^2)$   
 $f''(x_i) = (f(x_{i+2}) - 2f(x_{i+1}) + f(x_i))/h^2 + O(h)$ ;  $f''(x_i) = (-f(x_{i+3}) + 4f(x_{i+2}) - 5f(x_{i+1}) + 2f(x_i))/h^2 + O(h^2)$ 

Backward: 
$$f'(x_i) = (f(x_i) - f(x_{i-1}))/h + O(h)$$
;  $f'(x_i) = (3f(x_i) - 4f(x_{i-1}) + f(x_{i-2}))/(2h) + O(h^2)$   
 $f''(x_i) = (f(x_i) - 2f(x_{i-1}) + f(x_{i-2}))/h^2 + O(h)$ ;  $f''(x_i) = (2f(x_i) - 5f(x_{i-1}) + 4f(x_{i-2}) - f(x_{i-3}))/h^2 + O(h^2)$   
Centered:  $f'(x_i) = (f(x_{i+1}) - f(x_{i-1}))/(2h) + O(h^2)$   
 $f''(x_i) = (-f(x_{i+2}) + 8f(x_{i-1}) - 8f(x_{i-1}) + f(x_{i-2}))/(12h) + O(h^4)$   
 $f'''(x_i) = (f(x_{i+1}) - 2f(x_i) + f(x_{i-1}))/h^2 + O(h^2)$   
 $f'''(x_i) = (-f(x_{i+2}) + 16f(x_{i+1}) - 30f(x_i) + 16f(x_{i-1}) - f(x_{i-2}))/(12h^2) + O(h^4)$   
 $f''''(x_i) = (f(x_{i+2}) - 2f(x_{i+1}) + 2f(x_{i-1}) - f(x_{i-2}))/(2h^3) + O(h^2)$   
System of nonlinear equations ... (fill in)

**Solving ODEs** (initial value problems): y' = f(x, y);  $y_0$  given:

**Euler:**  $y_{i+1} = y_i + f(x_i, y_i)h$ 

RK2 methods: **Heun's**: ...(fill in)

**Midpoint**:  $y_{i+1} = y_i + k_2 h$ , where  $k_1 = f(x_i, y_i)$  and  $k_2 = f(x_i + h/2, y_i + k_1 h/2)$ 

Ralston: ...(fill in)

**Classical RK4**:  $y_{i+1} = y_i + (k_1 + 2k_2 + 2k_3 + k_4)h/6$ , where

 $k_1 = f(x_i, y_i);$ 

 $k_2 = f(x_i + h/2, y_i + k_1 h/2);$ 

 $k_3 = f(x_i + h/2, y_i + k_2 h/2)$ 

and  $k_4 = f(x_i + h, y_i + k_3 h)$