# ECM1416 Forumlas

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1	$\mathbf{L}$	ineaı	r Algeobra					
1.	1	Vecto	rs					
1	1.1	Vooto	r Norm					
L • .	1.1	vector						
			$  v   = \sqrt{\sum v_i^2}$	(1)				
١.:	1.2	Inner	Product					
			$u \cdot v =  u  v \cos(\theta)$	(2)				
			$x_1, x_2 = x^T, x_3 = \sum_{i=1}^{n} x_i x_i$	(9)				
			$u \cdot v = u^T \cdot v = \sum_{i=1}^n u_i v_i$	(3)				

#### 1.1.3 Rotating Vector

$$v' = \begin{bmatrix} v_x \cos(\theta) - v_y \sin(\theta) \\ v_x \sin(\theta) + v_y \cos(\theta) \end{bmatrix}$$
 (4)

## 1.2 Matracies

# 1.2.1 Matrix Product $(A = (m \times n), B = (n \times p))$

$$A \cdot B = \begin{bmatrix} \sum_{i=1}^{n} a_{1i} b_{i1} & \cdots & \sum_{i=1}^{n} a_{1i} b_{ip} \\ \vdots & \ddots & \vdots \\ \sum_{i=1}^{n} a_{ni} b_{i1} & \cdots & \sum_{i=1}^{n} a_{ni} b_{ip} \end{bmatrix}$$
 (5)

#### 1.2.2 Rotation Matrix

## 1.2.3 Scale Matrix

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \lambda_1 x \\ \lambda_2 y \end{bmatrix} = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$
 (7)

#### 1.2.4 Shear Matrix

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} x \\ \mu x + y \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ \mu & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$
 (9)

#### 1.2.5 Transpose Properties

$$(A+B)^T = A^T + B^T (10)$$

$$(AB)^T = A^T B^T (11)$$

#### 1.2.6 Identity Matrix

$$v = I_n v, \forall v \in \mathbb{R}^n \tag{12}$$

# 1.2.7 Determinant of a $2 \times 2$ Matrix

$$\det\begin{pmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} \end{pmatrix} = ad - bc \tag{13}$$

## 1.2.8 Determinant of a $n \times n$ Matrix

$$\det(A) = \sum_{k=1}^{n} (-1)^{i+k} a_{ik} M_{ik} = \sum_{k=1}^{n} (-1)^{j+k} a_{kj} M_{kj}$$
(14)

#### 1.2.9 Inverse of a $2 \times 2$ Matrix

$$A^{-}1 = \begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{\det(A)} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$
 (15)

## 1.2.10 Inverse of a $n \times n$ Matrix

$$A^{-}1 = \frac{1}{\det(A)}C^{T} = \frac{1}{\det(A)} \begin{bmatrix} C_{11} & C_{21} & \cdots & C_{n1} \\ C_{21} & C_{22} & \cdots & C_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ C_{1n} & C_{2n} & \cdots & C_{nn} \end{bmatrix}$$

$$C_{ij} = (-1)^{i+j} M_{ij}$$

$$(16)$$

#### 1.2.11 Properties of an Inverse

$$(A^{-1})^{-1} = A (17)$$

$$(A^T)^{-1} = (A^{-1})^T (18)$$

$$(AB)^{-1} = B^{-1}A^{-1} (19)$$

$$(kA)^{-1} = k^{-1}A^{-1} (20)$$

#### 1.2.12 Solving a System of Linear Equations

$$Ax = b \tag{21}$$

$$x = A^{-1}b \tag{22}$$

# 2 Differentation

# 2.1 Matrix Calculus

## 2.1.1 Gradient

$$\nabla f = \begin{bmatrix} f_{x_1} \\ f_{x_2} \\ \vdots \\ f_{x_n} \end{bmatrix}$$
 (23)

# 2.1.2 Hessian

$$H_f = \begin{bmatrix} f_{x_1x_1} & f_{x_1x_2} & \dots & f_{x_1x_n} \\ f_{x_2x_1} & f_{x_2x_2} & \dots & f_{x_2x_n} \\ \vdots & \vdots & \ddots & \vdots \\ f_{x_mx_1} & f_{x_mx_2} & \dots & f_{x_mx_n} \end{bmatrix}$$
(24)

## 2.1.3 Jacobian

$$D_{f} = \begin{bmatrix} \nabla f_{1} \\ \nabla f_{2} \\ \vdots \\ \nabla f_{n} \end{bmatrix} = \begin{bmatrix} f_{1x_{1}} & f_{1x_{2}} & \dots & f_{1x_{n}} \\ f_{2x_{1}} & f_{2x_{2}} & \dots & f_{2x_{n}} \\ \vdots & \vdots & \ddots & \vdots \\ f_{nx_{1}} & f_{nx_{2}} & \dots & f_{nx_{n}} \end{bmatrix}$$
(25)

# 2.2 Taylor Expantion

$$f(x) \approx \sum_{i=0}^{n} \frac{f^{i}(a)(x-a)^{i}}{i!}$$

$$(26)$$

# 2.3 Maclorian Expantion (a = 0)

## **2.3.1** $\sin(x)$

$$\sin(x) \approx \sum_{k=0}^{n} \frac{(-1)^k x^{2k+1}}{(2k+1)!}$$
 (27)

# **2.3.2** $\cos(x)$

$$\cos(x) \approx \sum_{k=0}^{n} \frac{(-1)^k x^{2k}}{(2k)!}$$
 (28)

# **2.3.3** $e^x$

$$e^x \approx \sum_{k=0}^n \frac{x^k}{k!} \tag{29}$$

# **2.3.4** $\ln |1+x|$

$$\ln|1+x| \approx \sum_{k=0}^{n} (-1)^k \frac{x^{k+1}}{k+1} \tag{30}$$

# **2.3.5** $c(\frac{1}{1-x})$

$$c(\frac{1}{1-x}) \approx c \sum_{k=0}^{n} x^k \tag{31}$$

# 2.4 Solving ODEs

# 2.4.1 Intergrating Factor

$$\frac{dy}{dx} + P(x)y = Q(x)$$

$$I.F = e^{\int P(x)dx}$$

$$\frac{d}{dx}(e^{\int P(x)dx}y) = Q(x)e^{\int P(x)dx}$$

$$y = \frac{\int Q(x)e^{\int P(x)dx}dx}{e^{\int P(x)dx}}$$
(32)

## 2.4.2 Seperation of Variables

$$\frac{dy}{dx} = g(x)h(y)$$

$$\frac{dy}{dx} \frac{1}{h(y)} = g(x)$$

$$\int \frac{1}{h(y)} dy = \int g(x) dx$$
(33)

# 2.4.3 Eulers Method

$$y_{n+1} = y_n + hf(x_n, y+n) x_{n+1} = x_n + h$$
 (34)

# 3 Probability

# 3.1 Conditional Probability

# 3.1.1 Given That

$$P(A|B) = \frac{P(A \cap B)}{P(B)} \tag{35}$$

# 3.1.2 Independence

$$P(A \cap B) = P(A)P(B) \tag{36}$$

# 3.1.3 Law of Total Probability

$$P(A) = \sum_{n} P(A \cap B_n) = \sum_{n} P(A|B_n)P(B_n) = \sum_{n} P(B_n|A)P(A)$$
 (37)

## 3.1.4 Bayes Theorm

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)} \tag{38}$$

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

$$P(A|X) = \frac{P(X|A)P(A)}{P(X)} = \frac{P(X|A)P(A)}{P(X|A)P(A) + P(X|A^C)P(A^C)}$$
(38)

#### Random Variables 3.2

#### 3.2.1 Descrete Random Variables

$$P(x_i) = P(X = x_i)$$

$$P(X = x_i) \ge 0, \forall i$$

$$\sum P(X = x_i) = 1$$
(40)

## 3.2.2 Continus Random Variables

$$P(a \le X \le b) = \int_{a}^{b} f(x)dx$$

$$f(x) \ge 0, \forall x \in X$$

$$\int_{\Omega} f(x)dx = 1$$

$$f(x) = 0, x \notin X$$

$$(41)$$

$$P(a \le X \le b) = P(a < X \le b)$$

$$= P(a \le X < b)$$

$$= P(a < X < b)$$

$$(42)$$

#### 3.3 **Cumulative Distrubution Function**

#### 3.3.1 General Case

$$F(x) = P(X \le x) \tag{43}$$

#### 3.3.2 X is Discrete

$$F(x) = P(X \le x) = \sum_{\forall x_i \le x} P(x_i)$$
(44)

## 3.3.3 X is Continus

$$F(x) = P(X \le x) = \int_{-\infty}^{x} f(t)dt \tag{45}$$

$$P(a \le X \le b) = F(b) - F(a) \tag{46}$$

# 3.3.4 Properties of CDF

$$0 \le F(x) \le 1 \tag{47}$$

$$x < y \to F(x) \le F(y) \tag{48}$$

$$\lim_{x \to \infty} F(x) = 1 \tag{49}$$

$$\lim_{x \to -\infty} F(x) = 0 \tag{50}$$

# 3.4 Mean, expectation

## 3.4.1 First Moment of X

$$\mu = E[x] = \begin{cases} \sum_{i=1}^{n} x_i P(X = x_i) & \text{Discrete R.V} \\ \int_{-\infty}^{\infty} x f(x) dx & \text{Continus R.V} \end{cases}$$
 (51)

#### 3.4.2 nth Moment of X

$$E[x^n] = \begin{cases} \sum_{i=1}^{N} x_i^n P(X = x_i) & \text{Discrete R.V} \\ \int_{-\infty}^{\infty} x^n f(x) dx & \text{Continus R.V} \end{cases}$$
 (52)

## 3.4.3 Properties of Expectation

$$x \ge 0 \iff E[x] \ge 0 \tag{53}$$

$$E[cx] = cE[x] (54)$$

$$E[x + y] = E[x] + E[y]$$
 (55)

#### 3.4.4 Variance

$$Var(x) = \sigma^2 = E[x^2] - (E[x])^2$$
(56)

## 3.4.5 Standard Deviation

$$\sigma = \sqrt{\operatorname{Var}(x)} \tag{57}$$

$$Var(xC) = c^2 Var(x)$$
 (58)

$$Var(x + y) = Var(x) + Var(y)$$
 if  $x, y$  are indepedent (59)

$$Var(x + y) = Var(x) + Var(y) + 2Cov(x, y)$$
 if  $x, y$  are dependent (60)

(61)

## 3.4.6 Covariance

$$Cov(x,y) = E[xy] - E[x]E[y]$$
(62)

$$Cov(x, x) = E[x^2] - (E[x])^2 = Var(x)$$
 (63)

$$Cov(x, y) = 0$$
  $x, y$  are indepedent (64)

# 4 Markov Chains