#### MTH 100: Lecture 32

# Polynomials Applied to Matrices

Matrix Powers: If A is an nxn matrix [we may take the entries either real or complex], then the product matrix A.A is evell-defined and can be written as  $A^2$ , A.A.A =  $A^3$ , ...

In general,  $A^m = A.A...A$  (m times) for any positive integer m.

For convenience, we define  $A^\circ = I_n$ , the identity matrix.

. If A is invertible and  $A^{-1}$  is its inverse, then for any positive integer m,  $(A^m)^{-1} = (A^{-1})^m$ 

· Remark:

 $A^{i}.A^{j} = A^{i+j}$  and  $(A^{i})^{j} = A^{ij}$  where i, j can be arbitrary integers if A is invertible and non-negative integers if A is not invertible.

Definition. If  $\beta$  is a polynomial given by  $\beta(t) = \alpha_0 + \alpha_1 t + \alpha_2 t^2 + \dots + \alpha_m t^m$  and A is an  $n \times n$  matrix, then  $\beta(A)$  is the matrix given by  $\beta(A) = \alpha_0 I + \alpha_1 A + \alpha_2 A^2 + \dots + \alpha_m A^m$ 

- e Note that this is a new ruse of the symbol of because eve are applying it to matrices and not just scalars.
- If  $\beta$  and  $\gamma$  are two polynomials, then  $\beta \gamma(A) = \beta(A) \gamma(A) = \gamma(A) \beta(A) = (\gamma \beta)(A)$ where  $\beta \gamma$  is the polynomial defined by  $\beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma(t) \quad \text{(usual multiplication of } \beta \gamma(t) = \beta(t) \gamma($

### The Minimal Polynomial of a Matrix

Definition: Given an nxn matrix A, the minimal Bolynomial of A is the (non zero)

monic polynomial of minimal degree such that  $\beta(A) = 0$  (i.e. the Zero matrix)

Monic polynomial means the nonzero coefficient of highest bower of the variable is equal to 1. The monic condition is inserted so as to make the minimal polynomial unique.

Note: Every square matrix must have a minimal folynomial:

Suppose A is an nxn matrix with entries from a field F, then the set ZI, A, A<sup>2</sup>,..., A<sup>n2</sup> ? cannot be linearly independent because this set has n+1 matrices and  $din(F^{n\times n}) = |n^2|$ Let m be the smallest positive integer such that  $\{I, A, A^2, ..., A^m\}$  is linearly dépendent. Then A' is a linear Combination of the preceding matrices. Thus there exist scalars a, a1,..., am-1 Such that  $a_0 I + a_1 A + a_2 A^2 + \cdots + a_{m-1} A^{m-1} + A^m = 0$ 

Note: Reference: Problem 5 of Worksheet 6.

A Famous Result

Theorem (caley - Hamilton Theorem):

Let 9 denote the characteristic folynomial of an nxn matrix A.

Then  $\left[ \Psi(A) = 0 \right]$ 

corollary: The degree of minimal folynomial of any nxn matrix is atmost n.

(Recall that the degree of the characteristic folynomial) of A is n.

Note: We will omit the Broof of Caley-Hamilton Theorem. You may refer to advanced textbooks.

### Remark:

Using Remainder Theorem for Polynomial Division, we can see that if p(x) is the minimal polynomial of A and if q(x) is any other polynomial satisfied by A, then p(x) divides q(x).

## Review of Polynomials:

- We use the notation F[t] to indicate the vector space of polynomials with coefficients from the field F(F could be either IR & \$\psi\$)
- $\lambda \in F$  is called a root of a polynomial  $\beta(t)$  if  $\beta(\lambda) = 0$

· <u>Lemma</u>: Suppose  $P \in F[t]$  is a folynomial of degree m > 1.

Then it is a root of \$ if and only if there exists a folynomial 9 & F[t] with degree (m-1) such that

 $\beta(t) = (t - \lambda) \varphi(t)$ 

e Lemma: Suppose  $\beta \in F[t]$  is a Polynomial of degree m > 0, then  $\beta$  has atmost m distinct roots in F.

Lemma (Division Algorithm or Remainder Theorem): Suppose  $\not=$ ,  $9 \not\in F[t]$  with  $\not= \not= 0$ . Then there exist polynomials r,  $s \in F[t]$  with  $g(t) = \not= (t) s(t) + r(t)$  and either r = 0 or degres d = 0.

Fundamental Theorem of Algebra: Suppose  $\beta \in \Phi[t]$  is a polynomial of degree m 7.1. Then  $\beta$  has a root. Furthermore,  $\beta$  has a factorization of the form  $|\beta(t)| = C(t-\lambda_1)(t-\lambda_2)....(t-\lambda_m)$ 

Lemma: Suppose 9 EF[t]. Then 9(A)=0 if and only if the minimal folynomial of A divides 9. Proof: >: Snppose 9(A)=0 Let p(t) be the minimal folynomial of A. Then \$ \$ 0. Using Remainder Theorem, eve can estite Q(t) = p(t) s(t) + r(t) ..... where either r=0 or degr < deg > Evaluating (1) at A, q(A) = p(A) s(A) + r(A) Since  $\underline{\gamma(A)=0}$ ,  $\varphi(A)=0$  ----  $\underline{2}$ Now, Since & is the minimal folynomial of A, p(A)=0 and hence from 2, r(A)=0 But this is not fossible unless r=0because degr<degp and p is the minimal Bolynomial of A Therefore from (), &(t) = >(t) s(t) and So p divides 9/. E If p(t) is the minimal polynomial of A, then  $\beta(A) = 0$ .

If  $\beta(t)$  divides  $\Upsilon(t)$ , there exists a polynomial S(t) such that  $\Upsilon(t) = \beta(t)S(t)$ .

Evaluating at A,  $\Upsilon(A) = \beta(A)S(A) = 0.S(A) = 0$ .  $\Upsilon(A) = \gamma(A) = 0$ .  $\Upsilon(A) = 0$ .

### Diagonalization of Matrices:

· If A is a diagonal matrix, then its diagonal elements are its eigen values and standard basis vectors are its eigen vectors.

ie if  $A = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \ddots & n \end{bmatrix}$  then  $A e_i = \lambda_i e_i$  where  $e_i = \begin{bmatrix} 0 \\ i \\ 0 \end{bmatrix}$  ith entry

#### Definition:

An nxn matrix A is said to be diagonalizable if A is similar to a diagonal matrix D ie if there exists an invertible matrix P and a diagonal matrix D such that

 $A = PDP^{-1}$ 

Note: If A is diagonalizable then its powers are easy to compute.

Note: If A is diagonalizable, then its eigen values can be found by inspection of D. However, in Bractice, eve have to do things the other way round.

First, we find the eigenvalues from the characteristic equation, then we find P and the diagonal matrix D.