# Math 43900 Problem Solving <br> Fall 2016 <br> Lecture 8 Matrices 

Andrei Jorza

These problems are taken from the textbook, from Engels' Problem solving strategies, from Ravi Vakil's Putnam seminar notes and from Po-Shen Loh's Putnam seminar notes.

## 1 Matrices

## Overview

The way matrices show up in problem solving problems involves the following three main themes:

1. algebraic manipulations of matrices (they can be multiplied and the operation is not commutative),
2. determinants and eigenvalues of matrices,
3. matrices as defining linear maps on vector spaces.

## Basic results

1. You can always add two $m \times n$ matrices.
2. You can always multiply an $m \times n$ matrix and an $n \times p$ matrix to get an $m \times p$ matrix.
3. The trace of a matrix $\operatorname{Tr} A$ is the sum of its diagonal terms. It has the property that $\operatorname{Tr}(A+B)=$ $\operatorname{Tr}(A)+\operatorname{Tr}(B)$ and $\operatorname{Tr}(A B)=\operatorname{Tr}(B A)$ for all matrices $A$ and $B$.
4. The determinant of a matrix $\operatorname{det} A$ is a polynomial expression in the entries of the matrix $A$ and satisfies the following properties:
(a) If in a matrix $A=\left(a_{i j}\right)$ you write $A_{p, q}$ for the $(n-1) \times(n-1)$ where you eliminate the $p$-th row and $q$-th column from $A$ then

$$
\operatorname{det}(A)=a_{11} \operatorname{det} A_{11}-a_{12} \operatorname{det} A_{12}+\cdots+(-1)^{n-1} a_{1, n} \operatorname{det} A_{1, n}
$$

(b) $\operatorname{det}(A B)=\operatorname{det}(A) \operatorname{det}(B)$ for all matrices $A$ and $B$.
(c) If you swap two rows or columns of a matrix $A$ to obtain a matrix $B$ then $\operatorname{det}(B)=-\operatorname{det}(A)$.
(d) If in a matrix $A$ you add a multiple of one row to a different row to get a matrix $B$ then $\operatorname{det}(B)=\operatorname{det}(A)$. The same is true if you add a multiple of a column to a different column.
5. Suppose $A$ is an $n \times n$ matrix. If you can find a nonzero vector (i.e., an $n \times 1$ matrix consisting of a single column) and a scalar $\alpha$ such that $A v=\alpha v$ then $\alpha$ is said to be an eigenvalue of $A$ with eigenvector $v$.
6. If $A$ is an $n \times n$ matrix the characteristic polynomial of $A$ is the monic degree $n$ polynomial

$$
P_{A}(X)=\operatorname{det}\left(X I_{n}-A\right)
$$

(a) A scalar $\alpha$ is an eigenvalue of $A$ if and only if it is a root of $P_{A}(X)$. The roots of $P_{A}(X)$ are the eigenvalues of $A$ and are counted with multiplicity if they are not distinct. E.g., $I_{n}$ has $n$ eigenvalues all equal to 1 .
(b) $P_{A}(X)=X^{n}-(\operatorname{Tr} A) X^{n-1}+\cdots+(-1)^{n} \operatorname{det}(A)$.
(c) Since we know the relation between the coefficients of a polynomial and its roots we deduce that if $\lambda_{1}, \ldots, \lambda_{n}$ are the eigenvalues of $A$ then

$$
\begin{aligned}
\operatorname{Tr}(A) & =\lambda_{1}+\lambda_{2}+\cdots+\lambda_{n} \\
\operatorname{det}(A) & =\lambda_{1} \lambda_{2} \cdots \lambda_{n}
\end{aligned}
$$

(d) If you plug in $A$ into the polynomial $P_{A}(X)$ you always get the 0 matrix: $P_{A}(A)=O$.
(e) If $A$ and $B$ are matrices then $P_{A B}(X)=P_{B A}(X)$ as polynomials.
7. A big result in linear algebra says that for any matrix $A$ you can find an invertible matrix $S$ such that the conjugate $S A S^{-1}$ has a very special shape: the Jordan canonical form. In fact the Jordan canonical form $S A S^{-1}$ has the $n$ eigenvalues on the diagonal but much more is true: $S A S^{-1}$ is block diagonal and each block is of the form

$$
\left(\begin{array}{cccc}
\lambda & 1 & 0 & \ldots \\
0 & \lambda & 1 & \ldots \\
& & \ddots & \ddots \\
0 & \ldots & 0 & \lambda
\end{array}\right)
$$

with an eigenvalue $\lambda$ on the diagonal and 1-s off diagonal. E.g., for a $2 \times 2$ matrix the possible Jordan canonical forms are

$$
\left(\begin{array}{cc}
\lambda_{1} & 0 \\
0 & \lambda_{2}
\end{array}\right) \text { for } \lambda_{1} \neq \lambda_{2} \text { and }\left(\begin{array}{cc}
\lambda & 0 \\
0 & \lambda
\end{array}\right) \text { or }\left(\begin{array}{cc}
\lambda & 1 \\
0 & \lambda
\end{array}\right)
$$

## 2 Problems

### 2.1 Determinants, traces, characteristic polynomials and eigenvalues

## Easier

1. Show that you can never find two $n \times n$ matrices $A$ and $B$ with real coefficients such that $A B-B A=I_{n}$. [Hint: What is the trace?]
2. Consider an $n \times(n+1)$ matrix $A=\left(a_{i j}\right)$. For a column $k$ write $A_{k}$ for the $n \times n$ matrix you obtain from $A$ by removing the $k$-th column. Show that

$$
a_{11} \operatorname{det} A_{1}-a_{12} \operatorname{det} A_{2}+\cdots+(-1)^{n+1} a_{1, n+1} \operatorname{det} A_{n+1}=0
$$

[Hint: Can you see this expression as the determinant of an $(n+1) \times(n+1)$ matrix?
3. Suppose $P(X)$ is a polynomial and $A$ is an $n \times n$ matrix such that $P(A)=0$. Show that the eigenvalues of $A$ are among the roots of $P(X)$. [Hint: What are the eigenvalues of $P(A)=0$ ? Use Exercise 6.]
4. This is an application of Exercise 15. Suppose $X$ is an antisymmetric matrix, i.e., of the form $X=-X^{t}$. (Think $\left(\begin{array}{cc} & x \\ -x & \end{array}\right)$.) Show that every eigenvalue of $X$ is of the form ai where $i=\sqrt{-1}$ and $a \in \mathbb{R}$. [Hint: If $X v=\alpha v$ compute $\left\langle X^{t} v, v\right\rangle=\langle v, X v\rangle$ in two ways.]
5. Show that $A^{k}=0$ for some $k \geq 0$ if and only if all the eigenvalues of $A$ are 0 in which case $A^{n}=0$ as well.

## Harder

6. (VERY USEFUL) Suppose $A$ is an $n \times n$ matrix and $Q(X)$ is any polynomial. If the eigenvalues of $A$ are $\lambda_{1}, \ldots, \lambda_{n}$ then the eigenvalues of $Q(A)$ (also an $n \times n$ matrix) are $Q\left(\lambda_{1}\right), \ldots, Q\left(\lambda_{n}\right)$.
7. Suppose $A$ is an $n \times n$ real matrix such that $A^{2}=A+I_{n}$. Show that $\operatorname{det}(A)<2^{n}$. In fact show that $\operatorname{det}(A) \leq\left(\frac{1+\sqrt{5}}{2}\right)^{n}$.
8. Suppose $X$ is a real matrix with $X+X^{t}=I_{n}$. Show that $\operatorname{det} X \geq \frac{1}{2^{n}}$.
9. Compute the determinant of the matrix $\left(a_{i j}\right)$ where $a_{i i}=2$ and if $i \neq j$ then $a_{i j}=(-1)^{i-j}$. [Hint: Use row operations to simplify the matrix.]
10. Suppose $A$ is an $n \times n$ matrix whose square $A^{2}$ is either 0 or $I_{n}$. Show that $\operatorname{det}\left(A+I_{n}\right) \geq \operatorname{det}\left(A+I_{n}\right)$.

### 2.2 Algebraic operations and linear algebra

## Easier

11. Suppose $\left(x_{n}\right)$ is a sequence defined by the linear recurrence $x_{n+2}=a x_{n+1}+b x_{n}$ for all $n \geq 0$. Show that

$$
\binom{x_{n+2}}{x_{n+1}}=\left(\begin{array}{ll}
a & b \\
1 & 0
\end{array}\right)\binom{x_{n+1}}{x_{n}}
$$

and conclude that for $n \geq 1, x_{n}$ is the first entry of the matrix $\left(\begin{array}{ll}a & b \\ 1 & 0\end{array}\right)^{n-1}\binom{x_{1}}{x_{0}}$.
12. Compute $\left(\begin{array}{cc}\lambda_{1} & 0 \\ 0 & \lambda_{2}\end{array}\right)^{n}$ and $\left(\begin{array}{cc}\lambda & 1 \\ 0 & \lambda\end{array}\right)^{n}$ for all $n$.
13. Suppose $f(x)=a_{0}+a_{1} x+a_{2} x^{2}+\cdots$ is a converging power series. Show that $f\left(S A S^{-1}\right)=S f(A) S^{-1}$.
14. A useful application of Exercise 12. Show that if $f(x)=a_{0}+a_{1} x+a_{2} x^{2}+\cdots$ is an absolutely convergent power series then $f\left(\left(\begin{array}{cc}\lambda_{1} & 0 \\ 0 & \lambda_{2}\end{array}\right)\right)=\left(\begin{array}{cc}f\left(\lambda_{1}\right) & 0 \\ 0 & f\left(\lambda_{2}\right)\end{array}\right)$ and $f\left(\left(\begin{array}{cc}\lambda & 1 \\ 0 & \lambda\end{array}\right)\right)=\left(\begin{array}{cc}f(\lambda) & f^{\prime}(\lambda) \\ 0 & f(\lambda)\end{array}\right)$.
15. If $u$ and $v$ are $n \times 1$ column matrices write $\langle u, v\rangle=u^{t} v$ for the dot product of the two vectors. If $A$ is an $n \times n$ matrix show that $\langle u, A v\rangle=\left\langle A^{t} u, v\right\rangle$. Show that $\langle v, \bar{v}\rangle \geq 0$, where $\bar{v}$ is the complex conjugate of $v$.
16. If $A=\left(a_{i j}\right)$ show that $\operatorname{Tr}\left(A \cdot A^{t}\right)=\sum_{i, j} a_{i j}^{2}$.

## Harder

17. Show that there exists no complex matrix $A$ such that $\sin (A)=\left(\begin{array}{cc}1 & 2016 \\ 0 & 1\end{array}\right)$. (This is a Putnam problem.) [Hint: Use the Exercise 14.]
18. Suppose $A$ and $B$ are $2 \times 2$ complex matrices such that $A B=B A$. Show that you can find two complex numbers $a$ and $b$ such that $B=a A+b$. [Hint: Conjugate $A$ to a Jordan canonical form, then things are much easier.] More generally, if $A$ and $B$ are $n \times n$ matrices such that $A B=B A$ show that $B=P(A)$ where $P$ is a degree at most $n-1$ polynomial.
19. Consider $v_{1}, \ldots, v_{n}$ vectors in $\mathbb{R}^{n}$ and the matrix $A$ whose entry on row $i$ and column $j$ is the dot product $v_{i} \cdot v_{j}$. Let $B$ be the matrix whose columns are the vectors $v_{1}, \ldots, v_{n}$. Show that $\operatorname{det} A=(\operatorname{det} B)^{2}$. [Hint: Show that $A=B B^{t}$.]
20. An application of the previous problem. Suppose $A_{1}, A_{2}, \ldots, A_{n}$ are subsets of some set $\left\{x_{1}, x_{2}, \ldots, x_{n}\right\}$. Show that if $M$ is the $n \times n$ matrix whose entry on row $i$ and column $j$ is the cardinality $\left|A_{i} \cap A_{j}\right|$ then $\operatorname{det} M \geq 0$. [Hint: Apply the previous problem to $v_{i}$ whose entry in position $j$ is 1 if $x_{j}$ is in $A_{i}$ and 0 otherwise.] A cool application: Suppose you choose $a_{1}, \ldots, a_{n}$ divisors of an integer $n$. Show that $\operatorname{det}\left(\operatorname{gcd}\left(a_{i}, a_{j}\right)\right) \geq 0$.
21. Suppose $A$ and $B$ are two $n \times n$ matrices that don't commute and you can find nonzero real numbers $p, q, r$ such that $p A B+q B A=I_{n}$ and $A^{2}=r B^{2}$. Show that $p=q$.
22. Suppose $A$ and $B$ are $n \times n$ real matrices such that $\operatorname{Tr}\left(A \cdot A^{t}+B \cdot B^{t}\right)=\operatorname{Tr}\left(A \cdot B+A^{t} \cdot B^{t}\right)$. Show that $A=B^{t}$. [Hint: Use Exercise 16 and then complete squares.]
