Math 2270-002 Week 11 notes

We will not necessarily finish the material from a given day's notes on that day. We may also add or subtract some material as the week progresses, but these notes represent an in-depth outline of what we plan to cover. These notes cover material in 5.4-5.5, 6.1.

Mon Nov 5

• 5.4-5.5 Brief intro to Matlab, continued discussion of change of variables to understand linear transformations, and introduction to complex eigenvalues and eigenvectors

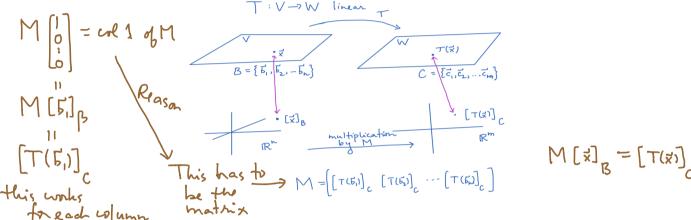
Announcements:

Warm-up Exercise: Find eigenvalues & eigenspace bases for $A = \begin{bmatrix} 4 & 4 \\ -1 & 0 \end{bmatrix}$ ans: $p(\lambda) = (\lambda^{-2})^2$ $E_{\lambda=2} = span \{ \begin{bmatrix} -2 \\ 1 \end{bmatrix} \}$ use labor.

Monday Review and look ahead:

We've been studying eigenvectors and eigenvalues for square matrices A, and the notion of diagonalizability, which we first understood in an algebraic sense.

On Friday we talked about how linear transformations $T: V \to W$ are associated with matrix transformations from $\mathbb{R}^n \to \mathbb{R}^m$, once we choose bases $B = \{\underline{\boldsymbol{b}}_1, \underline{\boldsymbol{b}}_2, \dots \underline{\boldsymbol{b}}_n\}$ for $V, C = \{\underline{\boldsymbol{c}}_1, \underline{\boldsymbol{c}}_2, \dots \underline{\boldsymbol{c}}_m\}$ for W.



This general framework is connected to matrix diagonalizability, and to more general notion of "similar matrices" when we have linear transformations from \mathbb{R}^n to \mathbb{R}^n expressed in one coordinate system, and wish to change to another one. On Friday we understood how the algebraic identity for diagonalizable matrices

$$D = P^{-1} A P$$

is related to this "change of coordinates" framework: For a diagonalizable matrix A, the identity above is also a change of variables identity for understanding the matrix transformation $T(\underline{x}) = A \underline{x}$ first given in standard coordinates; instead with respect to the eigenbasis for \mathbb{R}^n , $B = \{\underline{b}_1, \underline{b}_2, \dots \underline{b}_n\}$ which constitutes the columns of P. In fact, the identity above can be read as

So far, all of our eigendata has been real, but it's possible to have complex eigenvalues and eigenvectors as well, in which case we switch from considering real number scalars to considering complex number scalars and proceed as before. We'll introduce those ideas today and tomorrow, in section 5.5.

On Wednesday, I tentatively plan to begin Chapter 6 on "Orthogonality". This Chapter will lead to amazing applications, but it begins with a review and extension of dot product ideas related to angles between vectors and orthogonality in \mathbb{R}^n , that you learned about in multivariable Calculus for \mathbb{R}^2 and \mathbb{R}^3 . There will be a relatively short homework assignment due next Wednesday, containing preliminary Chapter 6 material.

On Friday we have our second midterm exam, which will cover sections 4.1-4.6, 4.9 (and google page rank), 5.1-5.5. I'll reserve a room for a problem session on Thursday, where we will go over a practice exam.

For your extra credit homework problem, and because you want to familiarize yourself with Matlab for the upcoming chapter as well, I'll demo Matlab.

Here are the two Matlab scripts I've created for this week. You can download these from either our public homework page or CANVAS. Put them into a directory and open them from Matlab. When you run the script with the big arrow at the top of your Matlab window all the commands will be executed. You can modify the script as desired.

The file "matrixcomputations.m" (The .m signifies matlab file.):

% some matlab commands for matrices...if you want more just use google.

% enter a matrix row by row, each row terminates with ";"
A=[1 2 3; 4 5 6] %use spaces or commas to separate row entries
AA=[3 5 9;2 4 6]; %if you put a semi-colon at the end the computation
%isn't echoed in the command window.

%rref computes reduced row echelon form. I named the result % with an equals sign and a name, so %that it would be saved to my workspace after running the script

B=rref(A)

% augmenting a matrix, either by rows or columns: first by row

b=[1,0,0] %row vector ..

C=[A;b] % add a 3rd row to A, using b

% use semicolons to separate rows

D=rref(C) % the 3 by 3 matrix D is invertible, since D is the identity

E=C^(-1) % inverse matrix...everything is decimals

F= C*E % this is how you multiply matrices! should get identity

% augmenting a matrix by columns

g=[-1;3] % a column vector with two entries

H=[A,g] % augment A with the column g

K=rref(H) %reduce the augmented matrix to find solutions to Ax=g

% eigenvalues and eigenvectors

L=[2,1,0;0,2,0;0,0,3]

lambda=eig(L) % eigenvalues

[M,N]=eig(L) % eigenvalues in first diagonal matrix,

%eigenvectors in second matrix

% note that this matrix is not diagonalizable

The file "PAC_team_rankings.m"

```
% PAC 12 Stochastic Matrix before past weekend
% AZ AZS CAL CU OR OSU STA UCLA USC UT WA WAS
SMa=[0, 0, 1/3, 0, 1/3, 25, 0, 0, 0, 0, 0, 0, 0, 0]
   0, 0, 0, 0, 0, 25, 0, 0 1/3, 0, 0;
                                        %AZS
   0, 0, 0, 0, 0, .25, 0, 0, 0, 0, .5, 0;
                                        %CAL
   0, 1/3, 0, 0, 0, 0, 1/3, 0, 0, 0, 0
                                        %CU
   0, 0, 1/3 0, 0, 0, 0, 0, 0, (0, 5, 0;
                                        %OR
   0, 0, 0, 1/3, 0, 0, 0, 0, \sqrt{0}, 0, 0;
                                        %OSU
   0, 1/3, 0, 0, 1/3, 0, 0, 0, 1/3, 0, 0, 0;
                                         %STA
   1/3, 0, 1/3, 0, 0, 0, 0, 0, 0, 0, 0, 0,
                                         %UCLA
   1/3, 0, 0, 1/3, 0, 0, 0, 0, 0, 0, 0, 1;
                                         %USC
   1/3, 0, 0, 0, 0, 0, .5, 1/3, 1/3, 0, 0, 0;
                                         %UT
  0, 1/3, 0, 1/3, 0, 0, 0, 1/3, 0, 5, 0, 0;
                                         %WA
  0, 0, 0, 0, 1/3,.25,.5, 0, 0, 5, 0, 0; \ \( \text{WAS} \)
```

%create a regular stochastic matrix with google fudge factor: SMb=.15/12*ones(12,12)+.85*SMa

% the columns of a large power of SMb should be essentially identical, % and are the equilibrium solution. The entries can be used to rank % the Pac 12 teams, based on current records. SMb^20

Returning to 5.4-5.5:

$$[T]_{\beta} = P_{\beta \leftarrow E} [T]_{E} P_{E \leftarrow \beta}$$

$$\overrightarrow{\nabla} = C_{1} \overrightarrow{b}_{1} + C_{2} \overrightarrow{b}_{2} + \cdots + C_{n} \overrightarrow{b}_{n} \qquad \overrightarrow{E} = \{\overrightarrow{e}_{1}, \overrightarrow{e}_{2}, ... \overrightarrow{e}_{n}\} \qquad \overrightarrow{E} =$$

$$\begin{bmatrix} \vec{x} \end{bmatrix}_{\beta} = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_k \end{bmatrix}$$

$$M = \begin{bmatrix} T \end{bmatrix}_{\beta} = \begin{bmatrix} c_1(\vec{c_1}) \end{bmatrix}_{\beta} \begin{bmatrix} \tau(\vec{c_2}) \end{bmatrix}_{\beta} \cdot \begin{bmatrix} \tau(\vec{c_2}) \end{bmatrix}_{\beta}$$

Exercise 1 (To review change of basis) Try to pick a better basis to understand the matrix transformation $T(\underline{x}) = A \underline{x}$, even though the matrix A is not diagonalizable. Compute $M = P^{-1}AP$ or compute M directly, to see if it really is a "better" matrix

better matrix.

B =
$$\left\{ \begin{bmatrix} -2 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\} = \left\{ \vec{b}_1, \vec{b}_2 \right\}$$

Check: $A \begin{bmatrix} -2 \\ 1 \end{bmatrix} = \begin{bmatrix} 4 & 4 \\ -1 & 0 \end{bmatrix}$

E = $2 \begin{bmatrix} -2 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 7 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 7 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 7 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 7 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 7 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 7 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 7 \\ 2 \end{bmatrix}, \begin{bmatrix} 7 \\ 2 \end{bmatrix}$

9 = 2

On Thesday

5.5 Complex eigenvalues and eigenvectors.

We'll focus on 2×2 matrices, for simplicity. In this case it will turn out that a matrix with real entries and complex eigenvalues is always similar to a rotation-dilation matrix...

Definition A matrix of the form $A = \begin{bmatrix} a & -b \\ b & a \end{bmatrix}$ is called a *rotation-dilation* matrix, because for $r = \sqrt{a^2 + b^2}$ we can rewrite A as $A = r \begin{bmatrix} \frac{a}{r} - \frac{b}{r} \\ \frac{b}{r} - \frac{a}{r} \end{bmatrix} = r \begin{bmatrix} \cos(\theta) - \sin(\theta) \\ \sin(\theta) - \cos(\theta) \end{bmatrix}.$

So the transformation $T(\underline{x}) = A\underline{x}$ rotates vectors by an angle θ and then scales them by a factor of r. (So A^2 rotates by an angle 2 θ and scales by r^2 ; A^3 rotates by an angle 3 θ and scales by r^3 , etc.

Exercise 2) Draw the transformation picture for

picture for
$$T\left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}\right) = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

The picture for $T\left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}\right) = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ $= \sqrt{2} \begin{bmatrix} \cos \frac{\pi}{4} & -\sin \frac{\pi}{4} \\ \sin \frac{\pi}{4} & \cos \frac{\pi}{4} \end{bmatrix}$ $A = \sqrt{2} \begin{bmatrix} \sqrt{2} & -\sqrt{2} \\ \sqrt{2} & \sqrt{2} \end{bmatrix}$

and interpret this transformation as a rotation-dilation. Overlay your diagram onto one picture: How about the transformation picture for $T \circ T = T^2$?

て(を)= [-1]

Exercise 3) What are the eigenvalues of a rotation-dilation matrix $A = \begin{bmatrix} a & -b \\ b & a \end{bmatrix}$? How do you think you would go about finding the eigenvectors?

follow warmup.

the same algorithms.

Tues Nov 6

• 5.5 Complex eigendata

Announcements:

On Wed, I won't start Chptrb.

Instead, Appendix on complex *s in text

plue tie in to things like matrix of a linear transformation"

• I'll let you know when I have the review classroom (Th 1:00-2:20)

· I'M say something about midden tomorrow.

Warm-up Exercise: Find the eigenvalues of this matrix, in terms of a, b:

$$A = \begin{bmatrix} a & -b \\ b & a \end{bmatrix}$$

$$\lambda = a \pm bi$$

(Important matrix for 95.5)

$$|A-\lambda I| = |a-\lambda|^{2} + |b|^{2}$$

$$= (\lambda - a)^{2} + |b|^{2}$$

$$= (\lambda - a)^{2} - (ib)^{2} \quad i^{2} = -|a-\lambda|^{2}$$

$$= (\lambda - a)^{2} - (ib)^{2} \quad i^{2} = -|a-\lambda|^{2}$$

$$= (\lambda - a - ib)(\lambda - a + ib)$$

$$(\lambda - a)^{2} + |b|^{2} = 0$$

$$= (\lambda - a - ib)(\lambda - a + ib)$$

$$= [\lambda - (a + ib)][\lambda - (a - ib)]$$

$$(\lambda - a)^{2} = -|b|^{2}$$

$$\lambda = a + ib$$

$$\lambda = a + ib$$

It is possible for a matrix A with real entries to be diagonalizable if one allows complex scalars and vectors, even if it's not diagonalizable with real eigenvalues and eigenvectors. You saw an example of that on a food for thought problem, if you weren't afraid. We'll use a matrix today that we'll use later as well, in section 5.6, to study an interesting discrete dynamical system. This matrix is not a rotation-dilation matrix, but it is *similar* to one, and that fact will help us understand the discrete dynamical system.

Exercise 1) Let

$$B = \begin{bmatrix} .9 & -.4 \\ .1 & .9 \end{bmatrix}$$

Find the (complex) eigenvalues and eigenvectors for B.

Next
$$(B-\lambda I)$$
 to be non-trivial

First $(B-\lambda I) \neq I$
 $\Rightarrow det(B-\lambda I) = 0$.

 $(A-\lambda)(A-\lambda)$
 $(A-\lambda)(A-\lambda)$
 $(A-\lambda)^2 = -0.04 = (.2i)^2$
 $A-9 = \pm .2i$
 $A-9 = \pm .2i$
 $A-9 = \pm .2i$
 $A=A+2i$
 $A=A+2i$

General facts we saw illustrated in the example, about complex eigenvalues and eigenvectors: Let A be a matrix with real entries, and let

$$A \underline{\mathbf{v}} = \lambda \underline{\mathbf{v}}$$

Re
$$\lambda = a$$
, Im $\lambda = b$

$$e \underline{v} = \underline{u}, \quad \text{Im } \underline{v} = \underline{w}.$$

with
$$\lambda = a + b i$$
, $\underline{v} = \underline{u} + i \underline{w}$ complex, $(a, b \in \mathbb{R}, \underline{u}, \underline{w} \in \mathbb{R}^n)$. Then we write

$$Re \lambda = a, \text{ Im } \lambda = b$$

$$Re \underline{v} = \underline{u}, \text{ Im } \underline{v} = \underline{w}.$$

$$\text{Must} = \underline{u}$$
So, the equation $A\underline{v} = \lambda \underline{v}$ expands as
$$A\underline{u} + i\underline{w} = (a\underline{u} - b\underline{w}) + i\underline{w}.$$

It will always be true then that the conjugate $\lambda = a - bi$ is also an eigenvalue, and the conjugate vector $\underline{v} = \underline{u} - i \underline{w}$ will be a corresponding eigenvector, because it will satisfy

$$A\vec{u} - i \underline{w} = (a - b i)(\underline{u} - i \underline{w})$$

$$A\vec{u} - i A\vec{w} = (a\vec{u} - b\vec{w}) + i(-b\vec{u} - a\vec{w})$$

Exercise 4 Verify that if the first eigenvector equation holds, then

$$A \underline{\mathbf{u}} = a \underline{\mathbf{u}} - b \underline{\mathbf{w}}$$

$$A \underline{\mathbf{w}} = b \underline{\mathbf{u}} + a \underline{\mathbf{w}}$$

Then check that these equalities automatically make the second conjugate eigenvector equation true as well.

<u>Theorem</u> Let A be a real 2×2 matrix with complex eigenvalues. Then A is similar to a rotation-dilation matrix.

proof: Let a complex eigenvalue and eigenvector be given by $\lambda = a + b i$, $\underline{v} = \underline{u} + i \underline{w}$ complex, ($a, b \in \mathbb{R}, \underline{u}, \underline{w} \in \mathbb{R}^n$) Choose

$$P = [\operatorname{Re} \underline{\boldsymbol{v}} \quad \operatorname{Im} \underline{\boldsymbol{v}}] = [\underline{\boldsymbol{u}} \quad \underline{\boldsymbol{w}}]$$

(One can check that $\{\underline{u}, \underline{w}\}$ is automatically independent.) Then, using the equations of Exercise 4, we mimic what we did for diagonalizable matrices...

And
$$\mathbf{a}$$
 and \mathbf{a} \mathbf{b} \mathbf{a} \mathbf{b} \mathbf{b}

$$P^{-1} A P = \left[\begin{array}{cc} a & b \\ -b & a \end{array} \right].$$

(The matrix on the right is a rotation-dilation matrix ... nobody ever said what the sign of b was. :-))

we'll look this over on Wednesday_

It's a mess, but one can carry out the procedure of the theorem, for the matrix B in exercise 1,

$$B = \begin{bmatrix} .9 & -.4 \\ .1 & .9 \end{bmatrix}$$
using $\lambda = .9 - .2 i$, $\mathbf{y} = \begin{bmatrix} -2 i \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix} + i \begin{bmatrix} -2 \\ 0 \end{bmatrix}$, one gets
$$P = \begin{bmatrix} \operatorname{Re} \mathbf{y} & \operatorname{Im} \mathbf{y} \end{bmatrix} = \begin{bmatrix} 0 & -2 \\ 1 & 0 \end{bmatrix}, \qquad P^{-1} = \frac{1}{2} \begin{bmatrix} 0 & 2 \\ -1 & 0 \end{bmatrix}$$

$$P^{-1}BP = \frac{1}{2} \begin{bmatrix} 0 & 2 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} .9 & -.4 \\ .1 & .9 \end{bmatrix} \begin{bmatrix} 0 & -2 \\ 1 & 0 \end{bmatrix}$$

$$P^{-1}BP = \begin{bmatrix} .9 & -.2 \\ .2 & .9 \end{bmatrix} = \sqrt{.85} \begin{bmatrix} \frac{.9}{\sqrt{.85}} & -\frac{.2}{\sqrt{.85}} \\ \frac{.2}{\sqrt{.85}} & \frac{.9}{\sqrt{.85}} \end{bmatrix}$$

$$P^{-1}BP = r \begin{bmatrix} \cos(\theta) & -\sin(\theta) \end{bmatrix}.$$

$$P^{-1}BP = r \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}.$$

for
$$r = \sqrt{.85} \approx .92$$
, $\theta = \arctan\left(\frac{2}{9}\right) \approx .22$ radians.

Application on next page, to browse through...

72. Use the method outlined in Exercise 70 to check for which values of the constants a, b, and c the matrix

$$A = \begin{bmatrix} 1 & a & b \\ 0 & 0 & c \\ 0 & 0 & 1 \end{bmatrix}$$
 is diagonalizable.

- 73. Prove the Cayley-Hamilton theorem, $f_A(A) = 0$, for diagonalizable matrices A. See Exercise 7.3.54.
- 74. In both parts of this problem, consider the matrix

$$A = \begin{bmatrix} 1 & 2 \\ 4 & 3 \end{bmatrix},$$

with eigenvalues $\lambda_1 = 5$ and $\lambda_2 = -1$ (see Example 1).

- a. Are the column vectors of the matrices $A \lambda_1 I_2$ and $A \lambda_2 I_2$ eigenvectors of A? Explain. Does this work for other 2×2 matrices? What about diagonalizable $n \times n$ matrices with two distinct eigenvalues, such as projections or reflections? (*Hint:* Exercise 70 is helpful.)
- b. Are the column vectors of

$$A - \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$$

eigenvectors of A? Explain.

"Linear Algebra with Applications" by Offo Bretscher

7.5 Complex Eigenvalues

Imagine that you are diabetic and have to pay close attention to how your body metabolizes glucose. After you eat a heavy meal, the glucose concentration will reach a peak, and then it will slowly return to the fasting level. Certain hormones help regulate the glucose metabolism, especially the hormone insulin. (Compare with Exercise 7.1.52.) Let g(t) be the excess glucose concentration in your blood, usually measured in milligrams of glucose per 100 milliliters of blood. (Excess means that we measure how much the glucose concentration deviates from the fasting level.) A negative value of g(t) indicates that the glucose concentration is below fasting level at time t. Let h(t) be the excess insulin concentration in your blood. Researchers have developed mathematical models for the glucose regulatory system. The following is one such model, in slightly simplified (linearized) form.

$$g(t+1) = ag(t) - bh(t)$$

$$h(t+1) = cg(t) + dh(t)$$

(These formulas apply between meals; obviously, the system is disturbed during and right after a meal.)

In these formulas, a, b, c, and d are positive constants; constants a and d will be less than 1. The term -bh(t) expresses the fact that insulin helps your body absorb glucose, and the term cg(t) represents the fact that the glucose in your blood stimulates the pancreas to secrete insulin.

For your system, the equations might be

$$g(t+1) = 0.9g(t) - 0.4h(t)$$

$$h(t+1) = 0.1g(t) + 0.9h(t),$$

with initial values g(0) = 100 and h(0) = 0, after a heavy meal. Here, time t is measured in hours.

After one hour, the values will be g(1) = 90 and h(1) = 10. Some of the glucose has been absorbed, and the excess glucose has stimulated the pancreas to produce 10 extra units of insulin.

The rounded values of g(t) and h(t) in the following table give you some sense for the evolution of this dynamical system.

t	0	1	2	3	4	5	6	7	8	15	22	29
									-9.3 25.7		1.6 -8.3	

We can "connect the dots" to sketch a rough trajectory, visualizing the long-term behavior. See Figure 1.

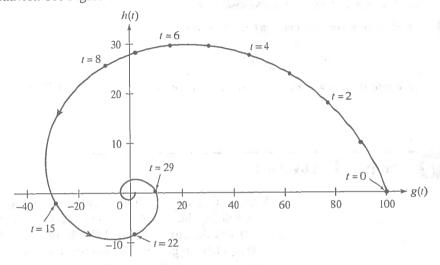


Figure I

We see that after 7 hours the excess glucose is almost gone, but now there are about 30 units of excess insulin in the system. Since this excess insulin helps to reduce glucose further, the glucose concentration will now fall below fasting level, reaching about -30 after 15 hours. (You will feel awfully hungry by now.) Under normal circumstances, you would have taken another meal in the meantime, of course, but let's consider the case of (voluntary or involuntary) fasting.

We leave it to the reader to explain the concentrations after 22 and 29 hours, in terms of how glucose and insulin concentrations influence each other, according to our model. The *spiraling trajectory* indicates an *oscillatory behavior* of the system: Both glucose and insulin levels will swing back and forth around the fasting level, like a damped pendulum. Both concentrations will approach the fasting level (thus the name).

Another way to visualize this oscillatory behavior is to graph the functions g(t) and h(t) against time, using the values from our table. See Figure 2.

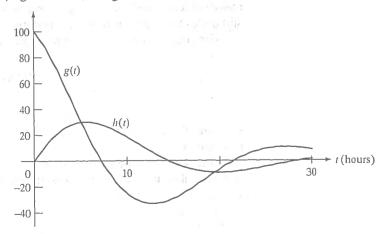


Figure 2

Example: From the algebra yesterday, and after a fair amount of work, For the dynamical system

$$\begin{bmatrix} g_{k+1} \\ h_{k+1} \end{bmatrix} = \begin{bmatrix} .9 & -.4 \\ .1 & .9 \end{bmatrix} \begin{bmatrix} g_k \\ h_k \end{bmatrix}$$

and with $\begin{bmatrix} g_0 \\ h_0 \end{bmatrix} = \begin{bmatrix} 100 \\ 0 \end{bmatrix}$, one can calculate and understand the spiral picture...

$$\begin{bmatrix} g_k \\ h_k \end{bmatrix} = \begin{bmatrix} .9 & -.4 \\ .1 & .9 \end{bmatrix}^k \begin{bmatrix} 100 \\ 0 \end{bmatrix} = -.92^k \begin{bmatrix} 100\cos(k\theta) \\ 50\sin(k\theta) \end{bmatrix}$$
$$\theta \approx .22 \text{ radians}.$$

For
$$r = \sqrt{.85} \approx .92$$
, $\theta = \arctan\left(\frac{2}{9}\right) \approx .22$ radians.

$$B = r P \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix} P^{-1}$$

$$B^{2} = r^{2} P \begin{bmatrix} \cos(2\theta) & -\sin(2\theta) \\ \sin(2\theta) & \cos(2\theta) \end{bmatrix} P^{-1}$$

$$B^{n} = r^{n} P \begin{bmatrix} \cos(n \theta) & -\sin(n \theta) \\ \sin(n \theta) & \cos(n \theta) \end{bmatrix} P^{-1}$$

$$B^{n} \begin{bmatrix} 100 \\ 0 \end{bmatrix} = .92^{n} \begin{bmatrix} 0 & -2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \cos(n\theta) & -\sin(n\theta) \\ \sin(n\theta) & \cos(n\theta) \end{bmatrix} \frac{1}{2} \begin{bmatrix} 0 & 2 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} 100 \\ 0 \end{bmatrix}$$
$$.92^{n} \begin{bmatrix} 0 & -2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \cos(n\theta) & -\sin(n\theta) \\ \sin(n\theta) & \cos(n\theta) \end{bmatrix} \begin{bmatrix} 0 \\ -50 \end{bmatrix}$$
$$= .92^{n} \begin{bmatrix} 0 & -2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 50 \sin(n\theta) \\ -50 \cos(n\theta) \end{bmatrix}$$

$$= .92^{n} \left[\begin{array}{c} 100 \cos(n\theta) \\ 50 \sin(n\theta) \end{array} \right]$$

Wed Nov 7

- Appendix B, the Complex plane $\mathbb C$. Brief review for exam 2.

Substitute Notes Today!!

Announcements: If you made a print out for the Pac 12 problem, you can handit in (otherise you can upload the own file to CANVAS, by 5:00 today)

I'll post practice exam later today

review session Thursday 1-2:20 LCB 222

(small chance this moves. If)

Friday exam 12:50-1:50 so, I'll let you know.

Warm-up Exercise: no warm-up today.

S minules early to S minules lake

Complex number algebra and geometry. Appendix B of text November 7

$$\mathbb{C} := \{ a + b \mid a, b \in \mathbb{R} \text{ and } i^2 := -1 \}$$

Arithmetic: If z = a + b i and w = c + d i then

$$z = w \text{ if and only if } a = c \text{ and } b = d.$$

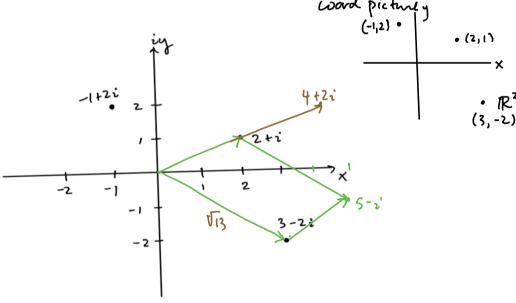
$$z + w := (a + c) + (b + d) i$$

$$(a + bi)(c + di) = z w := (ac - bd) + (ad + bc) i$$

Focusing on just addition and real-number scalar multiplication, \mathbb{C} can be thought of as a *real* vector space of dimension 2. In this case, the natural basis is $\beta = \{1, i\}$. Then the coordinate transfromation is an isomorphism with \mathbb{R}^2 :

$$z = a + b i$$
, $[z]_{\beta} = \begin{bmatrix} a \\ b \end{bmatrix}$.

If we identify complex numbers with these coordinates in \mathbb{R}^2 , we get the "complex plane" representation of \mathbb{C} :



Exercise 1a) Illustrate that complex number addition corresponds to vector addition in the complex plane, i.e. in the \mathbb{R}^2 coordinate plane that we have identified with \mathbb{C} as above. Use some of the points labeled above. Also, that real scalar multiplication corresponds to scalar multiplication in the \mathbb{R}^2 coordinate plane.

1b) We define the <u>modulus of z = a + b i to be $|z| = \sqrt{a^2 + b^2}$.</u> Note that this is just the magnitude of the coordinate vector $[a, b]^T$. Compute the modulus of some of the vectors in the diagram above.

Interesting geometry starts happening when you combine the geometry of the complex planes with algebraic operations such as complex multiplication.

Exercise 1 Define the transformation $T: \mathbb{C} \to \mathbb{C}$ by the formula $T(z) := iz \quad \text{i.e. } T(x+iy) := i(x+iy) = -y+ix.$ 1a) Verify that this is a linear transformation of \mathbb{C} .

(1) $T(z+\omega) = i(z+\omega) = iz + i\omega = T(z) + T(\omega)$ (2) T(z) = 11b) Describe T geometrically, in terms of its effect in the (x,y) coordinate plane. Include the matrix for T is a with respect to the basis $\beta = \{1,i\}$. $T(z) = iz = [T(z)] \begin{bmatrix} T(z) \\ z \end{bmatrix} \begin{bmatrix} T(z) \\ z$

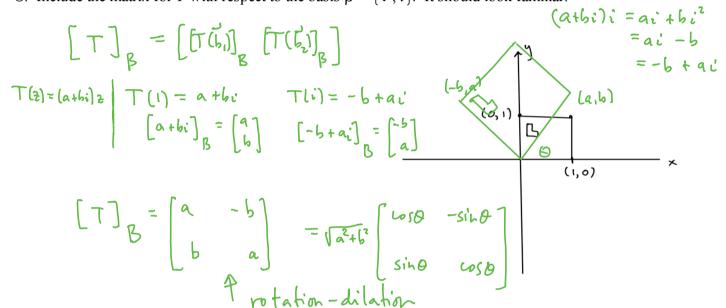
Exercise 2 Let $a, b \in \mathbb{R}$. Define the linear transformation $T : \mathbb{C} \to \mathbb{C}$ by the formula T(z) = (a + b i)z, i.e. $T(x + i y) = (a + b i) \cdot (x + i y)$.

<u>2a</u>)

Verify that this is a linear transformation of C.

<u>2b)</u> Describe *T* geometrically, in terms of its effect in the (x, y) coordinate plane. Include the matrix for *T* with respect to the basis $\beta = \{1, i\}$.

Describe T geometrically, in terms of its effect in the (x, y) coordinate plane which we have identified with \mathbb{C} . Include the matrix for T with respect to the basis $\beta = \{1, i\}$. It should look familiar.



An important algebraic operation for complex numbers is *conjugation*:

Definition: Let z = x + iy. Then the *conjugate* of $z, \overline{z} := x - iy$. Geometrically this is a reflection in the complex plane, across the x-axis. But the major uses of conjugations are algebraic:

Exercise **2** Let z = x + iy, w = u + iv be complex numbers. Then

$$2a) |z|^2 = z \overline{z}.$$

$$\underline{\mathbf{3b}}) \ \overline{z} \ \overline{w} = \overline{z} \ \overline{w} \ .$$

$$(2c)$$
 $z w = 0$ if and only if $z = 0$ or $w = 0$.

3d) If
$$z \neq 0$$
 then $\frac{1}{z}$ exists (i.e. the multiplicative inverse), in fact, $\frac{1}{z} = \frac{\overline{z}}{|z|^2}$.

$$\left(\frac{1}{z}, \frac{\overline{z}}{\overline{z}} = \frac{\overline{z}}{|z|^2}\right)$$

Geometric meaning of complex multiplication:

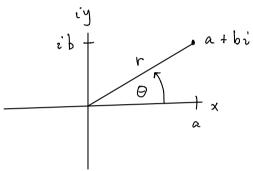
We use the *polar form* of complex numbers, which corresponds to polar coordinates in the \mathbb{R}^2 coordinate plane.

Let
$$z = a + b i$$
.
Let $r = \sqrt{a^2 + b^2} = |z|$.

Then

$$z = r \left(\frac{a}{\sqrt{a^2 + b^2}} + \frac{b}{\sqrt{a^2 + b^2}} i \right)$$
$$z = r \left(\cos \theta + i \sin \theta \right)$$

where θ is the polar coordinate angle.



Multiplication!! If

$$z = a + b i = r (\cos \theta + i \sin \theta)$$

 $w = c + d i = \rho (\cos \phi + i \sin \phi)$

Then

$$z w = r (\cos \theta + i \sin \theta) \rho (\cos \phi + i \sin \phi)$$

$$z w = r \rho [(\cos \theta \cos \phi - \sin \theta \sin \phi) + i (\cos \theta \sin \phi + \sin \theta \cos \phi)]$$

$$z w = r \rho [\cos (\theta + \phi) + i \sin (\theta + \phi)].$$

<u>upshot</u>: when you multiply two complex numbers, their moduli are multiplied, and their polar angles are added!

Remark: Using *Euler's formula* that $e^{i\theta} = \cos \theta + i \sin \theta$ the computation above may be expressed as: If

$$z = r e^{i \theta}$$
 and $w = \rho e^{i \phi}$

then

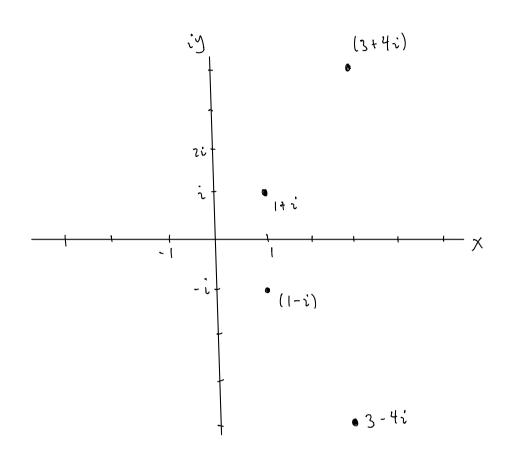
$$z w = r e^{i \theta} \rho e^{i \phi} = r \rho e^{i \theta} e^{i \phi} = r \rho e^{i (\theta + \phi)}$$
.

Exercise 4 Play with complex multiplication algebraically (using the rectangular coordinates of complex numbers) and geometrically (using their polar forms and the previous page).

$$(3-4i)\cdot(3+4i) =$$



$$(1+i)^4 =$$



Topics/concepts list for exam 2

- 4.1 vector spaces and sub vector spaces (subspaces) abstract definitions. realization of subspaces as null spaces or as spans of collections of vectors how to check if a subset is a subspace. examples such as polynomial vector spaces, matrix vector spaces, \mathbb{R}^n , and subspaces of all of these.
- 4.2 *Nul A* and *Col A* for $T(\underline{x}) = A \underline{x}$; *Kernel T* and *Range T* for general linear transformations $T: V \rightarrow W$ how to find *Nul A* and *Col A*, and bases for each.
- 4.3 linearly independent/dependent sets; bases for vector spaces (including subspaces).

how to check whether the vectors in a set span a vector space.

how to check whether a set of vectors is linearly independent.

how to build up bases as growing sets of independent vectors, one vector at a time, until the set spans.

how to cull dependent vectors from a spanning set, until it is an independent set.

4.4 every basis of n vectors for a vector space V yields a coordinate system, via the coordinate isomorphism with \mathbb{R}^n .

answering questions about span and linear independence for sets of vectors in V by using coordinates with respect to a basis.

favorite examples include P_n , $M_{m \times n}$, the polynomial and matrix spaces.

- 4.5 dimension of a vector space. basic facts about dimension, number of vectors required to span, maximum number of independent vectors, dimensions of subspaces.
- 4.6 rank of a matrix. rank + nullity theorem.

connection to reduced row echelon form of the matrix.

how to find Row A, $Nul A^T$.

what Nul A, Row A, Col A, $Nul A^T$ have to do with the geometry of the transformation $T(\underline{x}) = A \underline{x}$.

- 4.9 Markov chains, stochastic and regular stochastic matrices, steady-state vector, google page rank ideas.
- 5.1-5.2 eigenvalues and eigenvectors. Finding eigenvalues via the characteristic equation $det(A \lambda I) = 0$; finding eigenspace bases.
- 5.3 Diagonalizable and non-diagonalizable matrices. Algebraic consequences, e.g. computing large powers of diagonalizable matrices.
- 5.4 Matrices of linear transformations, given domain and codomain bases; change of basis for matrix transformations using "better bases".

Improved understanding of the transformation $T(\underline{x}) = A \underline{x}$ in terms of \mathbb{R}^n basis made out of eigenvectors, as compared to the standard basis.

5.5 Complex eigendata. Finding complex eigenvalues and eigenvectors, especially for 2×2 matrices; rotation-dilation matrices.

computations fluency in the definitions and concepts ability to create examples illustrating definitions and concepts ability to discern whether statements are true or false, based on the material we've covered. use of material from Chapters 1-3 that relates to Chapters 4-5.