## 1 Midterm Review!

- 1. There are many equivalent definitions of rank. Give 3. What is the rank of  $3 \times 3$  matrix with a nullspace spanned by  $[1, 2, -2]^T$ ? What are the possible ranks of an  $n \times n$  projection matrix?
- 2. Diagonalization. Let  $A = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 1 & 1 \\ 0 & 1 & 1 \end{bmatrix}$ .
  - (a) Suppose I tell you  $\lambda = 1$  is an eigenvalue for A. Find all other eigenvalues (or show that 1 is the only eigenvalue).
  - (b) Find a basis of eigenvectors for each eigenspace.
  - (c) Diagonalize A.
- 3. Kristina's favourite example to try on T/F questions: Can you diagonalize this matrix:  $A = \begin{bmatrix} 2 & 1 \\ 0 & 2 \end{bmatrix}$ ? Can you invert it? Bonus: write down a T/F question this matrix would be a counterexample for.
- 4. Suppose A is a  $2 \times 2$  matrix with eigenvalue  $\lambda_1 = 2$  corresponding to eigenvector  $\mathbf{v}_1 = [-1, 1]^T$  and  $\lambda_2 = -2$  corresponding to  $\mathbf{v}_2 = [2, -3]^T$ .
  - (a) For  $\mathbf{x}_0 = [0, 1]^T$ , find  $\mathbf{x}_k = A^k \mathbf{x}_0$  in terms of  $\mathbf{v}_1, \mathbf{v}_2, \lambda_1$  and  $\lambda_2$ .
  - (b) Calculate  $A^n$ .
- 5. Prove "complex eigenvalues always come in conjugate pairs" (e.g. if  $\lambda$  in  $\mathbb{C}$  is an eigenvalue of A, then  $\overline{\lambda}$  is too).
- 6. This is not really review, but let's learn the spectral theorem. I'll give a statement of the theorem, read it and then try to answer the last question as best you can without checking back here.

The Spectral Theorem. An  $n \times n$  symmetric matrix has the following properties:

- (a) A has n REAL eigenvalues, counting multiplicities (e.g. no complex eigenvalues, ever. we love symmetric matrices).
- (b) A is orthogonally diagonalizable.

- 7. Suppose I told you a  $3 \times 3$  matrix has exactly 2 distinct real eigenvalues with one eigenvector each (but I don't tell you anything about the complex eigenvlaues). Can you tell me if it's diagonalizable or not?
- 8. Orthogonal  $n \times n$  matrices have lots of nice properties. Write down 3 equivalent definitions of a square matrix being orthogonal. (hint: one defintion has to do with what the matrix does to vectors you multiply it with).
- 9. Let V be the subset of  $\mathbb{R}^4$  satisfying the equations  $\mathbf{x}_1 + 2(\mathbf{x}_2 + \mathbf{x}_3) + \mathbf{x}_4 = 0$ ,  $\mathbf{x}_1 \mathbf{x}_2 + \mathbf{x}_4$  and  $10\mathbf{x}_1 = 0$ . Is this a subspace (hint: is it the nullspace of a matrix)? Find a basis for its orthogonal complement.
- 10. Let  $A = \begin{bmatrix} 1 & -3 & 1 \\ 0 & 2 & 2 \\ 2 & 0 & 4 \\ -3 & 1 & -5 \end{bmatrix}$ . What are the steps to find the projection matrix

P onto A's column space? (hint: is  $A^TA$  invertible?)

- (a) Carry out those steps to find P (can leave P as a product of matrices).
- (b) Find the projection matrix onto A's left nullspace (hint: this shouldn't require much computation).
- 11. Suppose we are trying to fit the points (x,y) = (1,3), (2,0), (-1,1) with a cubic of the form  $y = c_1x + c_2x^3$ .
  - (a) Do we expect to find a solution for  $c_1, c_2$  satisfying all three points?
  - (b) If not, set up the least squares problem that would find us the best possible solution (e.g. what matrix equations do we need to solve?).
  - (c) Will we have a unique solution? (why?) Solve the least squares problem (it's ok to leave the answer as a product of matrices).
- 12. Spectral Theorem T/F:
  - (a) There are symmetric matrices which are not orthogonally diagonalizable (i.e. diagonalizable with  $A = PDP^{-1}$  where P is an orthogonal matrix).
  - (b) If  $A^T = A$  and  $A = PDP^{-1}$  with D diagonal, then all entries of D must be real numbers.
  - (c) The dimension of an eigenspace of a symmetric matrix (i.e. the geometric multiplicity) is sometimes less than the multiplicity of that eigenvalue as a root of the characteristic polynomial (i.e. the algebraic multiplicity).
  - (d) If  $A^T = A$  and if  $\mathbf{u}, v$  satisfy  $A\mathbf{u} = 3\mathbf{u}, A\mathbf{v} = 2\mathbf{v}$ , then  $\mathbf{u} \cdot \mathbf{v} = 0$ .