

## Lab 1

1. The rank of  $\mathbf{X}$  is the dimension of the column space, and is also the number of independent columns of  $\mathbf{X}$ . This is '2' - which 2 columns of  $\mathbf{X}$  are independent? (First note that one of them is clearly a linear combination of the other two.) They will form a basis - so write them down and verify that they span  $\text{col}(\mathbf{X})$  and are independent. Finally, relate the rank of  $\mathbf{X}'\mathbf{X}$  to that of  $\mathbf{X}$  - we covered exactly this in class.
2. *The  $n$ -step transition matrix, for a Markov chain with  $s$  states, is given by  $P^{(n)} = P^n$ .* We did this in class for  $n = 2$ . To go from  $n - 1$  to  $n$  (for any  $n$ ) note that  $[\mathbf{P}^{(n)}]_{ij}$  is by definition the probability of going from state  $i$  to state  $j$  in  $n$  steps. So ask where you are after  $n - 1$  steps - you have to be somewhere. As in class:

$$[\mathbf{P}^{(n)}]_{ij} = \sum_{k=1}^s P(\text{go from } i \text{ to } k \text{ in } n-1 \text{ steps, then to } j \text{ in one step})$$

and exactly as in class - review this point now if you need to - you should conclude that this is the  $(i, j)^{\text{th}}$  element of  $\mathbf{P}^{(n-1)}\mathbf{P}$ . So  $\mathbf{P}^{(n)} = \mathbf{P}^{(n-1)}\mathbf{P}$  - how do you finish off now? Fill in the steps of the inductive argument. For part (b) phrase the problem in terms of  $\mathbf{P}^2$ .

3. (a) See the list of axioms in Lecture 3. (b) *Prove: In any vector space the identity element  $\mathbf{0}$  is unique.*

You want to show that if  $\tilde{\mathbf{0}}$  has the same properties as  $\mathbf{0}$ , i.e. if  $\mathbf{x} + \tilde{\mathbf{0}} = \mathbf{x}$  for any  $\mathbf{x}$  (call this statement (\*)), then necessarily  $\tilde{\mathbf{0}} = \mathbf{0}$ . Look at the list of axioms: we have

$$\begin{aligned}\tilde{\mathbf{0}} &= \tilde{\mathbf{0}} + \mathbf{0} \text{ (by \#3)} \\ &= \mathbf{0} + \tilde{\mathbf{0}} \text{ (by \#2)} \\ &= \mathbf{0} \text{ (by statement (*) with } \mathbf{x} = \mathbf{0}\text{)}.\end{aligned}$$

4. (a) Lecture 4. (b) *Prove: If  $W$  is a vector subspace of a vector space  $V$  then  $\dim(W) \leq \dim(V)$ .*

Suppose, for contradiction, that  $\dim(W) = s > r = \dim(V)$ . Then there is a basis  $\mathbf{w}_1, \dots, \mathbf{w}_s$  of  $W$ , and these vectors are also in  $V$ . Look now at Fact 2 from Lecture 3, to get the desired contradiction.

5. Here you will investigate some properties of covariance matrices...

Part (a) starts with expanding the product in the definition of  $\Sigma$  to get

$$\begin{aligned} E[\mathbf{xx}' - \boldsymbol{\mu}\mathbf{x}' - \mathbf{x}\boldsymbol{\mu}' + \boldsymbol{\mu}\boldsymbol{\mu}'] &= E[\mathbf{xx}'] - \boldsymbol{\mu}E[\mathbf{x}'] - E[\mathbf{x}]\boldsymbol{\mu}' + \boldsymbol{\mu}\boldsymbol{\mu}' \\ &= E[\mathbf{xx}'] - \boldsymbol{\mu}\boldsymbol{\mu}' - \boldsymbol{\mu}\boldsymbol{\mu}' + \boldsymbol{\mu}\boldsymbol{\mu}' \\ &= E[\mathbf{xx}'] - \boldsymbol{\mu}\boldsymbol{\mu}'. \end{aligned}$$

What BASIC fact – stressed in the lectures – is used in the first equality of this argument?

Part (b) involves the definition (Lecture 2) of the expected value of a vector or matrix as the vector or matrix of expected values. That independent random variables are uncorrelated starts with

$$\text{COV}[X, Y] = E[(X - \mu_X)(Y - \mu_Y)]$$

and then uses the suggested characterization. For (c) start with the definition:

$$\text{COV}[\mathbf{y}] = E[(\mathbf{y} - \boldsymbol{\mu}_y)(\mathbf{y} - \boldsymbol{\mu}_y)'],$$

where  $\boldsymbol{\mu}_y$  is the mean vector of  $\mathbf{y}$ . Now replace  $\mathbf{y}$  by its expression in terms of  $\mathbf{x}$ , and  $\boldsymbol{\mu}_y$  with its expression in terms of  $\boldsymbol{\mu}_x$ . Then let the linearity properties guide you. After a few steps you will be able to just insert your answer to (a).

6. Suppose we gather data  $X_1, \dots, X_n$  and compute the sample average and variance. Show that ...

Start by writing

$$\begin{aligned} S^2 &= \frac{1}{n-1} (X_1 - \bar{X}, \dots, X_n - \bar{X}) \begin{pmatrix} X_1 - \bar{X} \\ \vdots \\ X_n - \bar{X} \end{pmatrix} \\ &= \frac{1}{n-1} \{(X_1, \dots, X_n) - (\bar{X}, \dots, \bar{X})\} \left\{ \begin{pmatrix} X_1 \\ \vdots \\ X_n \end{pmatrix} - \begin{pmatrix} \bar{X} \\ \vdots \\ \bar{X} \end{pmatrix} \right\}. \end{aligned}$$

Write the second term in braces in terms of  $\mathbf{x}$  and  $\mathbf{1}_n$  (hint:  $\sum X_i = \mathbf{1}'_n \mathbf{x}$ ), and continue.