

Problem Set 1

Review of Linear Algebra and Probability

Prof. Jim Rehg
CS 7635 Computational Perception
College of Computing
Georgia Institute of Technology

January 7, 2002

Note: For this review problem set only, the policy is slightly unusual. There are four assigned problems which add up to 80 points. To earn the remaining 20 points, you must turn in your solutions by 5 pm on Tuesday Jan 8 (bring them to my office, CCB 253). Otherwise, the solutions are due in class on Monday Jan 14. Please do not collaborate on this problem set.

This policy is for my benefit. I need to see how many of you can do these problems quickly. Note that there is an extra credit problem worth 5 points. If you turn in partial solutions tomorrow and a more complete set on Monday, I will give you the higher of the two grades.

Please show all of your work. You DO NOT need to use Matlab for any of these problems. If you do use Matlab, make sure you understand what is going on and describe it in your write-up. Please don't waste your time type-setting your solutions (let me do that :-). Hand-written answers are fine as long as they are legible and organized. I've tried to leave space on this handout for your solutions.

Vectors and matrices are typeset in bold. A vector \mathbf{v} is a column vector; \mathbf{v}^T is a row vector where "T" denotes the transpose. $P(\cdot)$ denotes a probability density function (PDF), which sums or integrates to one.

1. [20 points]

This question concerns the solutions to systems of linear equations of the form $\mathbf{Ax} = \mathbf{b}$. For each of the $\{\mathbf{A}, \mathbf{b}\}$ pairs below, answer the following questions:

- i. What is the rank of \mathbf{A} ?
- ii. Characterize the solution space (i.e. the space of possible values for \mathbf{x}). Where possible, find
 - The solution vector \mathbf{x} with minimum length.
 - The vector \mathbf{x} satisfying $\arg \min_{\mathbf{x}} \|\mathbf{Ax} - \mathbf{b}\|$
- iii. Where possible, find
 - A vector \mathbf{y} ($\mathbf{y} \neq 0$) such that $\bar{\mathbf{x}} = \mathbf{x} + \alpha\mathbf{y}$ is a valid solution of $\mathbf{A}\bar{\mathbf{x}} = \mathbf{b}$ for all real α .
 - A vector \mathbf{z} ($\mathbf{z} \neq 0$) such that the solution for $\{\mathbf{A}, \mathbf{b} + \alpha\mathbf{z}\}$ (as computed in part ii) remains the same for all real α .

(1a) $\mathbf{A} = \begin{bmatrix} 1 & 3 \\ -4 & 2 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 10 \\ 12 \\ 0 \\ 0 \end{bmatrix} \quad [6 \text{ points}]$

(1b) $\mathbf{A} = \begin{bmatrix} 1 & 2 & 4 & 5 \\ 0 & 0 & 1 & 0 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad [6 \text{ points}]$

(1c) $\mathbf{A} = \begin{bmatrix} 12 & 9 \\ 16 & 12 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 11 \\ 23 \end{bmatrix} \quad [8 \text{ points}]$

2. [20 points]

(a) [4 points]

Write the probability density function for a gaussian random vector \mathbf{x} with mean μ and covariance Σ (i.e. the vector form of the normal density).

(b) [8 points]

Someone has given you a large database of random vectors of length 2, distributed according to a gaussian density with zero mean and unit covariance ($\Sigma = \mathbf{I}$). For a simulation, you need random vectors distributed with the following mean and covariance:

$$\mu = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \quad \Sigma = \begin{bmatrix} 3/8 & -1/8 \\ -1/8 & 3/8 \end{bmatrix}$$

Give an algorithm (with sufficient detail to implement it) for constructing the desired random vectors from the vectors in your database.

(c) [8 points]

Sketch the equiprobability contours of the desired gaussian density in part (b).

3. [20 points]

An eager young professor who is heading back into the lab late at night is trying to guess whether his graduate student is working on her paper or not. He has noticed that whenever this particular student is working, her car is in the parking lot and loud music is playing in the lab. Let W, C, M be three binary random variables corresponding to the events “student Working on paper”, “Car in parking lot”, and “loud Music playing”. Through careful observation, the professor has determined the following conditional probabilities:

$$P(C|W, M) = P(C|W) = \frac{\begin{array}{cc} W = 1 & W = 0 \\ C = 1 & 0.8 & 0.1 \\ C = 0 & 0.2 & 0.9 \end{array}}{=} = P(M|W)$$

In other words, C and M are conditionally independent given W , and e.g. $P(C = 1|W = 1) = P(M = 1|W = 1) = 0.8$. The professor guesses that $P(W = 1) = 0.7$ (she is a good student).

(a) [6 points]

Compute the initial distributions $P(C)$ and $P(M)$, which reflect the state of knowledge before the professor arrives on campus. Hint: First compute the joint probabilities $P(C, W)$ and $P(M, W)$ and then marginalize (sum) over W .

(b) [6 points]

Upon driving into the parking lot, the professor notices his student’s car. Update the distribution over W to reflect this new information (i.e. calculate $P(W|C = 1)$). Hint: Use Bayes Rule.

(c) [6 points]

Calculate the probability that loud music is playing in the lab. Hint: Calculate the updated $P(M, W)$.

(d) [2 points]

Upon reaching his office, the professor remembers that the conference deadline was yesterday! What is $P(M)$?

4. [20 points]

(a) [6 points]

x is a scalar gaussian random variable with mean μ and variance σ_x . Measurements of x are corrupted by an additive gaussian noise source n . The measurement model is $y = x + n$. n has zero mean and variance σ_n . Write the following (Note: These are not trivial to derive, and it's fine if you look them up):

- An expression for $p(x|y)$, the conditional density which encapsulates our knowledge of x as a function of observations y .
- The estimate \hat{X} for a given $y = Y$ that minimizes $E\{\|x - \hat{X}\|^2|y = Y\}$, where $E\{\cdot|y = Y\}$ denotes the conditional expectation with respect to x . This estimate goes under various names, including least squares, minimum mean square, and minimum variance.

(b) [14 points]

Let s be an unknown random variable. A complex random process involving s generates a measurement vector \mathbf{y} of length 3. Through careful experimentation you determine that:

$$E\{\mathbf{y}\mathbf{y}^T\} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 2 & 5 \end{bmatrix} \quad \text{and} \quad E\{s\mathbf{y}\} = \begin{bmatrix} 2 \\ 5 \\ 12 \end{bmatrix}$$

A program left behind by one of your previous graduate students makes it easy to compute estimates of the form $\hat{s} = \mathbf{a}^T \mathbf{y}$ where \mathbf{a} is an arbitrary constant vector. Calculate the value of \mathbf{a} which will yield the minimum mean square estimate of s (i.e. minimizing $E\{\|s - \hat{s}\|^2\}$ where the expectation is over all possible s, \mathbf{y}). Identify the equations you have solved.

5. [Extra Credit: +5 points]

Prove that the exponent in a vector gaussian density (see (2a)) can be written:

$$-\frac{1}{2}\text{tr}\{\boldsymbol{\Sigma}^{-1}(\mathbf{x} - \boldsymbol{\mu})(\mathbf{x} - \boldsymbol{\mu})^T\},$$

where the *trace* $\text{tr}\{\mathbf{A}\}$ of a square matrix \mathbf{A} is the sum of its diagonal elements.