# Cognitive Algorithms Assignment 1

### Part 1 - Linear Algebra Recap

Due on Tuesday, May 2, 2017 10am via ISIS

## Task 1 (8 points)

- 1. Compute the scalar product of the following vectors  $\begin{pmatrix} 1 \\ -2 \\ 0 \end{pmatrix}$ ,  $\begin{pmatrix} 3 \\ 0 \\ 3 \end{pmatrix}$ .
  - $\square$  3  $\Box$  5  $\Box$  7
- 2. Let  $\mathbf{v}, \mathbf{w} \in \mathbb{R}^n$  be two column vectors. Which of the following statements is always true?
  - $\Box \mathbf{v} \cdot \mathbf{w} = \mathbf{w} \cdot \mathbf{v}$

  - $\nabla \mathbf{v} \cdot \mathbf{w}^T = \mathbf{w} \cdot \mathbf{v}^T$
- 3. The mapping  $f: \mathbb{R}^2 \ni (x,y)^\top \mapsto (x+y,y-x)^\top \in \mathbb{R}^2$  is given by the following matrix:
  - $\Box \left(\begin{array}{cc} 1 & 1 \\ 1 & -1 \end{array}\right)$
- $\square \left(\begin{array}{cc} 0 & 2 \\ -2 & 0 \end{array}\right) \qquad \square \left(\begin{array}{cc} 1 & 1 \\ -1 & 1 \end{array}\right)$
- 4. Which property does matrix multiplication *not* have?
  - $\square$  Associativity: (AB)C = A(BC)
  - $\Box$  Commutativity: AB = BA
  - $\square$  Distributivity: (A+B)C = AC + BC
- 5. Let  $A \in \mathbb{R}^{n \times n}$  be an invertible matrix and  $\mathbf{v}, \mathbf{w} \in \mathbb{R}^n$  two column vectors with  $A \cdot \mathbf{v} = \mathbf{w}$ . Which of the following statements is always true?
  - $\Box A = \mathbf{w} \cdot \mathbf{v}^{-1}$
- $\square \ \mathbf{v} = \mathbf{w} \cdot A^{-1}$
- $\Box \mathbf{v} = A^{-1} \cdot \mathbf{w}$

- 6. The rank of the matrix  $\begin{pmatrix} 4 & 4 & 4 \\ 4 & 4 & 4 \\ 4 & 4 & 4 \end{pmatrix}$  is
  - $\Box$  1

 $\square$  3

 $\Box$  4

- 7. For a square  $n \times n$  matrix A holds
  - $\square$  rank  $A = n \Rightarrow A$  is invertible, but there are invertible A with rank  $A \neq n$
  - $\square$  A is invertible  $\Rightarrow$  rank A = n, but there are A with rank A = n, which are not invertible.
  - $\square$  rank  $A = n \Leftrightarrow A$  is invertible

8. Which of the following matrices is orthogonal:

$$\square \left( \begin{array}{cc} 0 & 1 \\ -1 & 0 \end{array} \right) \qquad \qquad \square \left( \begin{array}{cc} 1 & -1 \\ -1 & 1 \end{array} \right) \qquad \qquad \square \left( \begin{array}{cc} 1 & 1 \\ 1 & 0 \end{array} \right)$$

### Task 2 (3 points)

We consider two functions f and g which transform an input vector  $\mathbf{x} = (x_1, \dots, x_d)^{\top} \in \mathbb{R}^d$  into a scalar:  $f(\mathbf{x}) = \mathbf{u}^{\top} \mathbf{x}$ ,  $\mathbf{u} = (u_1, \dots, u_d)^{\top} \in \mathbb{R}^d$  and  $g(\mathbf{x}) = \mathbf{x}^{\top} \mathbf{x}$ .

1. Compute the partial derivative of f and g with respect to one entry  $x_j$   $(j \in \{1, 2, ..., d\})$ 

(a) 
$$\frac{\partial f(\mathbf{x})}{\partial x_i} =$$

(b) 
$$\frac{\partial g(\mathbf{x})}{\partial x_i} =$$

2. Compute the gradient  $\nabla f(\mathbf{x}) = \left(\frac{\partial f(\mathbf{x})}{\partial x_1}, \dots, \frac{\partial f(\mathbf{x})}{\partial x_d}\right)^{\top}$  for f and g.

(a) 
$$\nabla f(\mathbf{x}) = \left(\frac{\partial f(\mathbf{x})}{\partial x_1}, \dots, \frac{\partial f(\mathbf{x})}{\partial x_d}\right)^{\top} =$$

(b) 
$$\nabla g(\mathbf{x}) = \left(\frac{\partial g(\mathbf{x})}{\partial x_1}, \dots, \frac{\partial g(\mathbf{x})}{\partial x_d}\right)^{\top} =$$

# Task 3 (4 points)

Let  $A \in \mathbb{R}^{n \times n}$  be a symmetric matrix and  $\mathbf{v}, \mathbf{w} \in \mathbb{R}^n$  Eigenvectors of A corresponding to Eigenvalues  $\lambda, \mu \in \mathbb{R}$ , with  $\lambda \neq \mu$ . (Recall the definition of Eigenvectors:  $A\mathbf{v} = \lambda \mathbf{v}$  and  $A\mathbf{w} = \mu \mathbf{w}$ )

Show:  $\mathbf{v}$  and  $\mathbf{w}$  are orthogonal, i.e.  $\mathbf{v}^{\top}\mathbf{w} = 0$ .

Hint:  $\lambda \mathbf{v}^{\top} \mathbf{w} = \dots$