

## Exercise Sheet 4

1. Consider the problem

$$\text{minimise } f(x) = (x_1 - 2x_2)^2 + (x_1 - 2)^2$$

Solve the above problem using Algorithm 1 (page 4). Perform two steps (by hand) with  $x^{(0)} = (0, 0)^T$ .

*Hint: In Algorithm 1, in order to compute the step size, consider  $f$  as a function of  $t$  and search for the minimum.*

2. Consider the problem

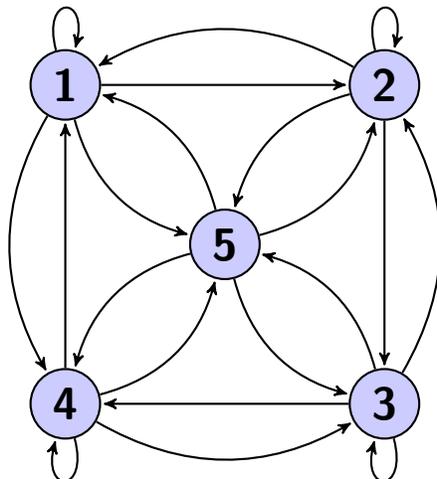
$$\text{minimise } f(x) = \frac{1}{2}(x_1 - 2x_2)^2 + x_1^4$$

Solve the above problem using Algorithm 2 (page 4). Perform two steps (by hand) with  $x^{(0)} = (2, 1)^T$  and  $m = 1/2$ .

3. Determine the classes and specify if they are recurrent or transient for the following transition matrices:

$$P_1 = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 1/2 & 0 & 0 & 1/2 \\ 0 & 1/2 & 0 & 1/2 \\ 0 & 0 & 0 & 1 \end{pmatrix} \quad \text{and} \quad P_2 = \begin{pmatrix} 0 & 1/2 & 1/2 & 0 & 0 \\ 1/2 & 0 & 1/2 & 0 & 0 \\ 1/2 & 1/2 & 0 & 0 & 0 \\ 1/2 & 1/2 & 0 & 0 & 0 \\ 0 & 1/2 & 1/2 & 0 & 0 \end{pmatrix}$$

4. Consider the following random walk with states  $S = \{1, 2, 3, 4, 5\}$ . Assume that at each state the transition probabilities to other adjacent states are all equal and the probability to stay at the same state in the next transition is zero.



- (a) Construct the transition probability matrix  $P$ .
- (b) Show that the Markov chain of the random walk is irreducible and all the states are recurrent.
- (c) Find the steady-state probability distribution.

5. (a) Consider the transition probability matrix

$$P = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}.$$

- i. Show that the Markov chain defined by  $P$  is irreducible but not regular.
- ii. Find the unique stationary distribution. Does this Markov chain converge to the stationary distribution?

(b) Consider the transition probability matrix

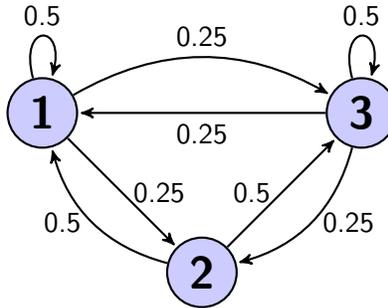
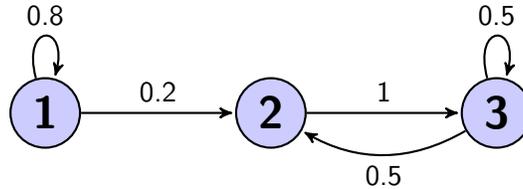
$$P = \begin{pmatrix} 0 & 1 \\ 2/3 & 1/3 \end{pmatrix}.$$

The stationary distribution  $\pi$  is given by

$$\lim_{n \rightarrow \infty} P^n X^{(0)} = \pi,$$

for any initial distribution  $X^{(0)}$ . Compute  $\pi$  using the eigen-decomposition of  $P$ .

6. Consider the following two random walks. Construct the transition probability matrices and find the steady-state probability distributions.



7. Consider a game with five levels, where the 5<sup>th</sup> level is the highest. A player starts at the lowest (1<sup>st</sup> level) and every time he flips a coin. If it turns up head, the player moves up one level. If tails, he moves down to the 1<sup>st</sup> level. When the player reaches the highest level, if it turns up heads he stays there and if tails he moves to the lowest level.
- Find the transition probability matrix.
  - What is the probability that the player will be in the 3<sup>rd</sup> level after his second flipping if he started at the 2<sup>nd</sup> level?
  - What is the probability that the player will be in the 2<sup>nd</sup> level after his third flipping for any starting level?
  - Find the steady-state distribution of the Markov chain (by hand).

ALGORITHM 1: STEEPEST DESCENT WITH EXACT LINE SEARCH

Input: cost function  $f: \mathbb{R}^n \rightarrow \mathbb{R}$ , initial guess  $x^{(0)}$ ,  $k = 0$   
While *convergence criterion not yet satisfied* do  
    compute  $d^{(k)} = -\nabla f(x^{(k)})$   
    compute  $t^{(k)} := \arg \min_{t \geq 0} \{f(x^{(k)} + td^{(k)})\}$   
    define  $x^{(k+1)} := x^{(k)} + t^{(k)}d^{(k)}$   
     $k \leftarrow k + 1$   
end  
define  $x^* := x^{(k)}$

ALGORITHM 2: NEWTON'S METHOD WITH LINE SEARCH (ARMIJO'S RULE)

Input: cost function  $f: \mathbb{R}^n \rightarrow \mathbb{R}$ , initial guess  $x^{(0)}$ ,  
initial step  $t^{(0)} = 1$ , parameter  $0 < m < 1$ ,  $k = 0$   
While *convergence criterion not yet satisfied* do  
    define  $d^{(k)}$  by solving the equation
$$H_f(x^{(k)}) d^{(k)} = -\nabla f(x^{(k)})$$
  
    define  $x^{(k+1)} := x^{(k)} + t^{(k)}d^{(k)}$   
    if  $f(x^{(k+1)}) - f(x^{(k)}) > mt^{(k)}\nabla f(x^{(k)})^T d^{(k)}$  then  
        define  $t^{(k+1)} = t^{(k)}/2$   
    else  $t^{(k+1)} = t^{(k)}$   
     $k \leftarrow k + 1$   
end  
define  $x^* := x^{(k)}$