

Math 450 - Homework 6 Solutions

1. Exercise 1.7.1. Find all the invariant distributions of the transition matrix

$$P = \begin{pmatrix} \frac{1}{2} & 0 & 0 & 0 & \frac{1}{2} \\ 0 & \frac{1}{2} & 0 & \frac{1}{2} & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{2} & 0 & 0 & 0 & \frac{1}{2} \end{pmatrix}$$

We need to solve the system $\pi = \pi P$ with the extra normalization condition $\sum_i \pi_i = 1$. These give the 6 equations”

$$\begin{aligned} \pi_1 &= \frac{1}{2}\pi_1 + \frac{1}{2}\pi_5 \\ \pi_2 &= \frac{1}{2}\pi_2 + \frac{1}{4}\pi_4 \\ \pi_3 &= \pi_3 + \frac{1}{4}\pi_4 \\ \pi_4 &= \frac{1}{2}\pi_2 + \frac{1}{4}\pi_4 \\ \pi_5 &= \frac{1}{2}\pi_1 + \frac{1}{4}\pi_4 + \frac{1}{2}\pi_5 \\ 1 &= \pi_1 + \pi_2 + \pi_3 + \pi_4 + \pi_5. \end{aligned}$$

There are 4 independent homogeneous equations and the non-homogeneous normalizing condition. Eliminating the second to last equation and writing the system in standard matrix form gives:

$$\begin{pmatrix} -\frac{1}{2} & 0 & 0 & 0 & \frac{1}{2} \\ 0 & -\frac{1}{2} & 0 & \frac{1}{4} & 0 \\ 0 & 0 & 0 & -\frac{1}{4} & 0 \\ 0 & \frac{1}{2} & 0 & -\frac{3}{4} & 0 \\ 1 & 1 & 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} \pi_1 \\ \pi_2 \\ \pi_3 \\ \pi_4 \\ \pi_5 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}.$$

This gives $\pi_1 = \pi_5$, $\pi_2 = \pi_4 = 0$, and $\pi_1 + \pi_3 + \pi_5 = 1$. Thus invariant measures must have the form $\pi = (a, 0, b, 0, a)$ where a, b are non-negative and $2a + b = 1$, and any such distribution is an invariant probability measure. Note that the transient states 2 and 4 must have probability 0.

2. Exercise 1.7.2. (Ehrenfest model) Gas molecules move about randomly in a box which is divided into two halves symmetrically by a partition. A hole is made in the partition. Suppose there are N molecules in the box. Think of the partitions as two urns containing balls labeled 1 through N . Molecular motion can be modeled by choosing a number between 1 and N at random and moving the corresponding ball from the urn it is presently in to the other. This is a historically important physical model introduced by Ehrenfest in the early days of statistical mechanics to study thermodynamic equilibrium.

- (a) Show that the number of molecules on one side of the partition just after a molecule has passed through the hole evolves as a Markov chain. What are the transition probabilities? (Take as the sets of states the number of molecules in one of the partitions.) Draw a transition diagram of the process.

The set of states of the Markov chain is

$$S = \{0, 1, 2, \dots, N\}$$

representing the number of molecules in one partition of the box. The transition probabilities are as follows:

$$P(X_{n+1} = j | X_n = i) = \begin{cases} i/N & \text{if } j = i - 1 \\ (N - i)/N & \text{if } j = i + 1. \end{cases}$$

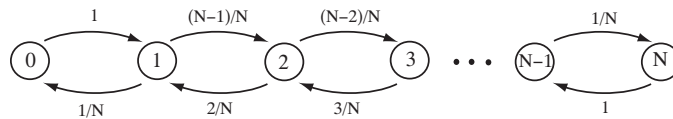


Figure 1: Transition diagram for the Ehrenfest gas model.

- (b) Is the chain recurrent? Yes. This is an irreducible finite chain, so all states are recurrent.
- (c) What is the invariant distribution of this chain? We use the notation $p_i = (N - i)/N$ and $q_i = i/N$. Then $\pi = \pi P$ corresponds to the following system of equations:

$$\begin{aligned} \pi_0 &= \pi_1 q_1 \\ \pi_1 &= \pi_0 p_0 + \pi_2 q_2 \\ \pi_2 &= \pi_1 p_1 + \pi_3 q_3 \\ &\dots \\ \pi_i &= \pi_{i-1} p_{i-1} + \pi_{i+1} p_{i+1} \\ &\dots \\ \pi_N &= \pi_{N-1} p_{N-1}. \end{aligned}$$

A simple induction shows that the unique solution (up to a multiplicative constant, which is determined by the equation $\pi_1 + \dots + \pi_N = 1$) is given by:

$$\pi_i = \frac{p_{i-1}p_{i-2} \dots p_0}{q_i q_{i-1} \dots q_1} \pi_0.$$

From this we obtain:

$$\pi_i = \pi_0 \frac{(N-i+1)(N-i+2) \dots N}{i(i-1) \dots 1} = \pi_0 \frac{N!}{i!(N-i)!} = \pi_0 C(N, i).$$

To find π_0 , notice that the sum of the binomial coefficients is 2^N , so from $\pi_0 + \dots + \pi_N = 1$ we get $\pi_0 = 1/2^N$. Therefore,

$$\pi_i = \frac{C(N, i)}{2^N}.$$

- (d) What is the number of steps it takes on average for a partition to become empty given that it was initially empty? In other words, find the expected return time to state 0.

We wish to find $m_0 = E_0[T_0]$. According to theorem 1.7.7, and since the chain is recurrent, this is given by

$$m_0 = \frac{1}{\pi_0} = 2^N.$$

If $N = 100$, and assuming each transition takes about 1/400th of a second, the mean return time to an empty partition is $2^{100}/400$ seconds. This is over 10^{18} centuries.

- (e) Do a computer simulation of this Markov chain for $N = 100$. Start from state 0 (one of the partitions is empty) and follow the chain up to 1000 steps. Draw a graph of the number of molecules in the initially empty partition as a function of the number of steps. On the basis of your answer to the previous item, would you expect to observe during the course of the simulation a return to state 0?

```

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
rand('seed',327)
tic
%number of molecules:
N=100;
%number of time steps:
t=1000;
%States are represented by
%row vectors whose ith entry is zero
%if ith molecule is in second
%compartment, and one if it is in the first.
%We start with and empty first compartment:

```

```

s=zeros(1,N);
%Number of molecules in first compartment:
number=[sum(s)];

for j=1:t
    %choose at random a number between 1 and N:
    i=ceil(N*rand);
    %Let 0 represent the first compartment and 1 the
    %second. Moving molecule i to a different
    %compartment means switching ith entry of s
    %from 0 to 1 or vice-versa.
    s(i)=rem(s(i)+1,2);
    number=[number sum(s)];
end
plot(0:t,number)
grid
toc
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%

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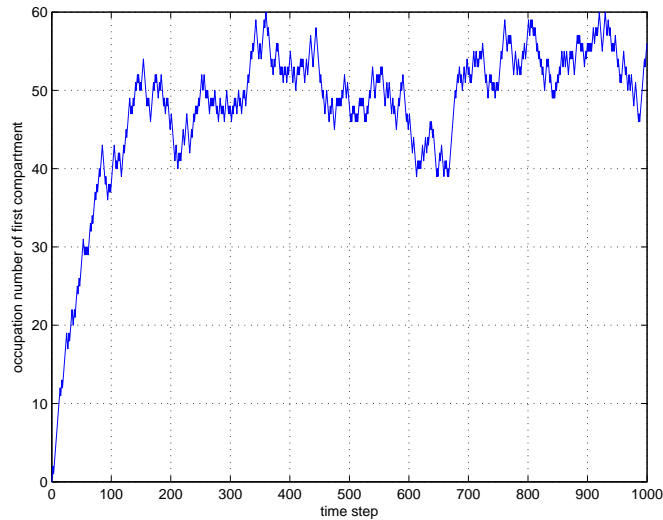


Figure 2: Number of molecules in the first compartment as a function of time. Time is measured in number of steps of the discrete Markov chain.

3. Exercise 1.7.3. A particle moves on the eight vertices of a cube in the following way: at each step the particle is equally likely to move to each of the three adjacent vertices, independently of its past motion. Let i be the

initial vertex occupied by the particle, o the vertex opposite i . Calculate each of the following quantities:

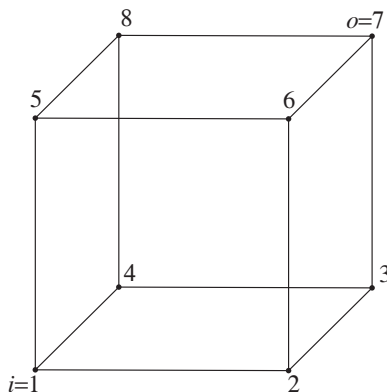


Figure 3: Transition diagram for the random walk on the cube. Each transition has probability $1/3$.

- (a) The expected return time to i is given by $E_i[T_i] = 1/\pi_i$, where π is the stationary probability distribution vector. This is unique since the chain is irreducible and finite. It is easy to show that the constant distribution $\pi_j = 1/8$ (suggested by symmetry) is stationary. Therefore,

$$E_i[T_i] = 8.$$

- (b) The expected number of visits to o until the first return to i is given by theorem 1.7.5 as the number γ_o^i such that the vector $\gamma^i = (\gamma_1^i, \dots, \gamma_8^i)$ satisfies: $\gamma_i^i = 1$, $0 < \gamma_j^i < \infty$, and $\gamma^i P = \gamma^i$. By theorem 1.7.6, this is a constant vector, $\gamma_j^i = 1$ for all j . Therefore,

$$\gamma_o^i = 1.$$

- (c) The expected number of steps until the first visit to o can be obtained using theorem 1.3.5. We write k_j^o for the number of steps until first visit to o , starting at j . We simplify the system using symmetry considerations. Write $u = k_2^o = k_4^o = k_5^o$, $v = k_3^o = k_6^o = k_8^o$, $k = k_1^o$, and $k_7^o = 0$. (We want the value of k .) The system of theorem 1.3.5 now reduces to the following 3 equations:

$$\begin{aligned} k &= 1 + u \\ u &= 1 + \frac{1}{3}k + \frac{2}{3}v \\ v &= 1 + \frac{2}{3}u. \end{aligned}$$

This is easily solved. The value we want is $k = 10$. ($u = 9$ and $v = 7$.)

To find these values by simulation, we use the following program. It produces sample paths of a Markov chain with initial distribution p and transition probabilities matrix P , stopped at the r -th visit to a set A .

```

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
function X=visit(p, P, A, r, n)
%Inputs - p probability distribution of initial state
%        - P transition probability matrix
%        - A set of states where chain is stopped
%        - r number of returns
%        - n maximal time. (If A empty, stop at time n.)
%Output - X sample chain till min{r-th hit time to A, n}.
%
%Note: need function samplefromp.m
q=p;
i=samplefromp(q,1);
X=[i];
m=0;
c=0;
while c<r & m<n-1
    q=P(i,:);
    i=samplefromp(q,1);
    X=[X i];
    c=c+(length(find(A==i))~0);
    m=m+1;
end
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%

```

We now calculate the mean return time, m_T , to $i = 1$, and the mean number m_N of visits to $o = 7$ between visits to i .

```

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
P=zeros(8,8);
P(1,[2 4 5])=1;
P(2,[1 3 6])=1;
P(3,[2 4 7])=1;
P(4,[1 3 8])=1;
P(5,[1 6 8])=1;
P(6,[2 5 7])=1;
P(7,[3 6 8])=1;
P(8,[4 5 7])=1;
P=(1/3)*P;
p=zeros(1,8);
p(1)=1;

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```

A=[1];
t=1000;
rand('seed',371)
T=[]; %return time to i=1
N=[]; %number of visites to o=7
for i=1:t
    X=visit(p, P, A, 1, 100000);
    T=[T length(X)-1];
    N=[N sum(X==7)];
end
m_T=mean(T)
m_N=mean(N)
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
The above gave the value  $m_T = 8.2820$  and  $m_N = 1.0650$ . (Correct
values: 8 and 1, respectively.) For the number of steps till first visit
to  $o = 7$ , starting at  $i = 1$ , the commands below produce the mean
value  $m_K = 9.8920$ . (Correct value: 10.)
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
A=[7];
t=1000;
rand('seed',371)
K=[]; %number of steps till o=7
for i=1:t
    X=visit(p, P, A, 1, 100000);
    K=[K length(X)-1];
end
m_K=mean(K)
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%

```