

## **MATH 218 Ch1: Experiments, Models, and Probabilities**

**MATLAB** (software package by Mathworks) will be used to simulate experiments with random outcomes. For this purpose, a *pseudo-random number generator* which produces a sequence of random numbers between 0 and 1 can be utilized, i.e. *rand(m, n)* produces an  $m \times n$  array of uniformly distributed pseudo-random numbers.

For a Matlab simulation we first generate a vector  $R$  of  $N$  random numbers:

```
>> N = 100;  
>> R = rand(1, N);
```

To simulate an experiment that contains an event with probability  $p$ , each random number  $r$  produced by above command will be tested such that if  $r < p$ , event occurs; otherwise it does not occur.

For example, to simulate a coin flipping experiment which yields heads with probability 0.4, tails with probability 0.5, or lands on its edge with probability 0.1, and 100 simulations are needed:

In this case, we generate vector  $X$  as a function of  $R$  to represent 3 possible outcomes:  $X(i)=1$  if flip  $i$  was heads,  $X(i)=2$  if flip  $i$  was tails, and  $X(i)=3$  if flip  $i$  landed on the edge.

```
>> X = (R <= 0.4) + (2*(R > 0.4).*(R <= 0.9)) + (3*(R > 0.9));  
>> [N, Y] = hist(X, 1:3);
```

Plot the number of occurrences of each outcome  $i$  for  $N=100$  trials of this experiment

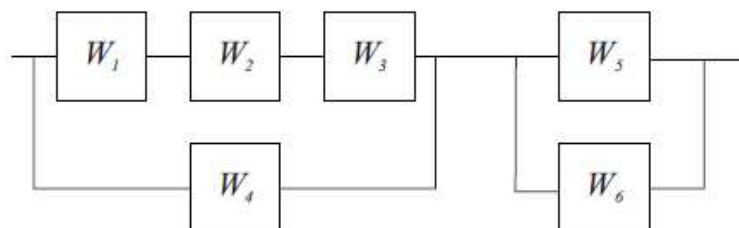
```
>> figure, bar(Y, N)  
>> xlabel('Outcome i'), ylabel('Number of occurrences of outcome i');
```

Next, the following Matlab function can be used to simulate two-coin experiment in Example 1.27 in textbook:

```
>> function [C, H] = twocoin(n); % n: number of trials  
    C = ceil(2*rand(n, 1)); % C(i) = 1 if coin 1, C(i) = 2 if coin 2 is chosen for trial i  
    P = 1 - (C/4); % P(i) = 0.75 if C(i) = 1, otherwise if C(i) = 2  
    H = (rand(n, 1) < P);  
end
```

This function generates vectors  $C$  and  $H$  for  $n$  trials of this experiment.  $C(i)$  indicates which coin (biased =1 or fair =2) is chosen, and  $H(i)$  is the simulated result of a coin flip with heads,  $H(i) = 1$  occurring with probability  $P(i)$ .

Another simulation example is given below about the reliability of a 6 component system which has the following configuration:



To simulate 100 trials of the six-component test (such that each component works with probability  $q$ ), we use the following Matlab function:

```

function N=reliable6(n,q);
% n is the number of 6 component devices
%N is the number of working devices
W=rand(n,6)>q;
D=(W(:,1)&W(:,2)&W(:,3))|W(:,4);
D=D&(W(:,5)|W(:,6));
N=sum(D);

```

The  $n \times 6$  matrix  $W$  is a logical matrix such that  $W(i,j)=1$  if component  $j$  of device  $i$  works properly. Note that  $D(i)=1$  if device  $i$  works. Otherwise,  $D(i)=0$ . Hence,  $N$  is the number of working devices.

The result of 10 repetitions of the 100 trials for  $q=0.2$  is given below:

```

>> for n=1:10, w(n)=reliable6(100,0.2); end
>> w
w =
82 87 87 92 91 85 85 83 90 89
>>

```

While the probability the device works is actually 0.8663.

For **discrete random variables**, the PMFs and CDFs for the families of random variables can be computed as follows:

For finite discrete random variable  $X$  defined by a set of sample values  $S_X = \{s_1, \dots, s_n\}$  with corresponding probabilities  $p_i = P_X(s_i) = P[X = s_i]$ ,  $\mathbf{p} = [p_1 \dots p_n]'$ , the following

Matlab function returns  $y_i = P_X(x_i)$ :

```

function pmf=finitepmf(sx,px,x)
% finite random variable X:
% vector sx of sample space
% elements {sx(1),sx(2), ...}
% vector px of probabilities
% px(i)=P[X=sx(i)]
% Output is the vector
% pmf: pmf(i)=P[X=x(i)]
pmf=zeros(size(x(:)));
for i=1:length(x)
    pmf(i)= sum(px(find(sx==x(i))));
end

```

By applying its definition, the CDF can be calculated simply as

```

function cdf=finitecdf(s,p,x)
% finite random variable X:
% vector sx of sample space
% elements {sx(1),sx(2), ...}
% vector px of probabilities
% px(i)=P[X=sx(i)]
% Output is the vector
% cdf: cdf(i)=P[X=x(i)]
cdf=[];
for i=1:length(x)
    pxi= sum(p(find(s<=x(i))));
    cdf=[cdf; pxi];
end

```

Note that the CDF can also be computed by summing the PMF as

```
>> cdf=cumsum(pmf);
```

The sample values of random variables can be generated based on the calculation of the CDF first. For example, the  $m$  samples of a binomial  $(n, p)$  random variable can be generated as

```
function x=binomialrv(n,p,m)
% m binomial(n,p) samples
r=rand(m,1);
cdf=binomialcdf(n,p,0:n);
x=count(cdf,r);
```