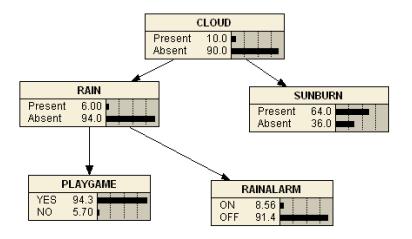
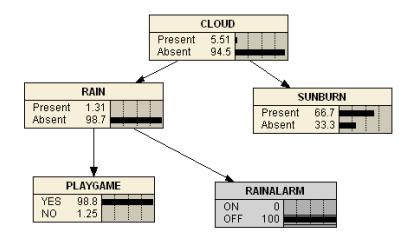
## s4344240 Yangyang Xu

## Prac 10

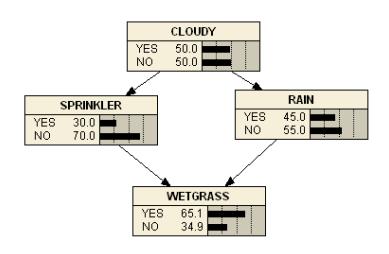


3

## If the rain alarm does NOT go off = GOSE ON



Because the rain alarm is off, thus the rain absent becomes 98.7% is reasonable.



$$P(C \mid W) = \frac{P(W \mid C) \cdot P(C)}{P(W)}$$

P(WIR,8) = 0.95 P(WIR, ~8)=0.9 P(WIR, &)=0.9 P(WIR, ~8)=0.1

P(R,81C) = P(R|C) P(S|C) = 0.8 ×0.1 = 0.08 P(NR,8|C) = P(NR|C) P(S|C) = 0.2 × 0.1 = 0.02 P(R,NS|C) = P(R|C) P(NS|C) = 0.8 × 0.9 = 0.72 P(NR,NS|C) = P(NR|C) P(NS|C) = 0.2 × 0.9 = 0.18

P(R,8|nc) = P(R|nc) P(S|nc) = 0.1 × 0.5-005 P(nR,8|nc) = P(nR|nc) P(S|nc) = 0.9 × 0.5 = 0.45 P(nR,8|nc) = P(nR|nc) P(nS|nc) = 0.1 × 0.5 = 0.05 P(nR,ns|nc) = P(nR|nc) P(ns|nc) = 0.9 × 0.5 = 0.05 P(nR,ns|nc) = P(nR|nc) P(ns|nc) = 0.9 × 0.5 = 0.5

$$\frac{P(w|c)P(c)}{P(w|c)P(c)} = P(w|c)P(c) = P(w|c)P(c)$$

$$P(w|c)P(e) + P(w|nc)P(c)$$

$$P(w|c) = P(w|R,3)P(RS|C) + P(w|nR,3)P(nR,3|C)$$

$$+ P(w|R, nS)P(R, nS|C) + P(w|nR, nS)$$

$$= 0.95 \times 0.08 + 0.9 \times 0.02 + 0.9 \times 0.72 + 0.1 \times 0.18$$

$$= 0.076 + 0.018 + 0.648 + 0.018$$

$$= 0.76$$

$$P(w|nC) = P(w|R,3)P(R, nS|nC) + P(w|nR,3)P(nR,3)$$

$$+ P(w|R, nS)P(R, nS|nC) + P(w|nR,nS)P(nR,3)$$

$$+ P(w|R, nS)P(R, nS|nC) + P(w|nR,nS)P(nR,3)$$

$$= 0.95 \times 0.05 + 0.9 \times 0.45 + 0.9 \times 0.05 + 0.1 \times 0.45$$

$$= 0.95 \times 0.05 + 0.9 \times 0.45 + 0.045 + 0.045$$

$$= 0.96 \times 0.06 + 0.045 + 0.045 + 0.045$$

$$= 0.76 + 0.5425 = 0.76$$
Answey =  $\frac{0.76}{0.76 + 0.5425} = \frac{0.76}{0.76 + 0.5425} = \frac{0.76}{0.7$ 

## Prac 11

Formula used to calculate information of Posterior:

(16.6) 
$$\mu_N = \frac{\sigma^2}{N\sigma_0^2 + \sigma^2}\mu_0 + \frac{N\sigma_0^2}{N\sigma_0^2 + \sigma^2}m$$

$$(16.7) \qquad \frac{1}{\sigma_N^2} \quad = \quad \frac{1}{\sigma_0^2} + \frac{N}{\sigma^2}$$

$$\sigma_0^2 = 1$$

$$\mu 0 = 1$$

N: no. of data = 150

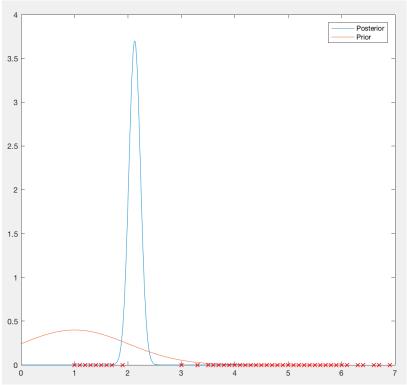
m: mean of data

 $\sigma^2$ : standard deviation of data

```
1 -
       data = meas(:,3);
2
       %% data info
3 -
       my_mean = 1;
4 -
       my_var = 1;
5 -
       data_mean = mean(data);
6 -
       data_std = std(data);
       %% posterior distributions
7
       mu_N=(data_std/(150*1+data_std))*1+(150*1/(150*data_std+data_std))*data_mean;
8 -
       std_N =sqrt(1/(1/1 + size(data,1)/data_std));
9 -
       % Plot Posterior
.0
       x = linspace(0,6,1000);
.1 -
.2 -
       y = normpdf(x,mu_N,std_N);
.3 -
       p1 = plot(x,y);
.4 -
       hold on;
.5
       % Plot Prior
       x2 = linspace(0,6,1000);
.6 -
.7 -
       y2 = normpdf(x2,1,1);
.8 -
       p2 = plot(x2,y2);
.9 -
       hold on;
       %% Plot data
0
       plot(data,0,'rx');
1 -
2 -
       hold off;
       legend([p1 p2], 'Posterior', 'Prior')
```

	_
Name 🛎	Value
<b>⊞</b> data	150x1 double
data_mean	3.7580
data_std     data_std	1.7653
<b>⊞</b> meas	150x4 double
⊞ mu_N	2.1264
<b>⊞</b> my_mean	1
<b>⊞</b> my_var	1
<b>፱</b> p1	1x1 Line
<b>፱</b> p2	1x1 Line
☐ std_N	0.1079
<b>⊞</b> x	1x1000 double
<b>⊞</b> x2	1x1000 double
<b>⊞</b> y	1x1000 double
<u>₩</u> y2	1x1000 double

The std of Posterior is less than the std(s) of prior and data. Thus the graph is skinner.



(Red-Crosses are data)