## T-61.3050 Machine Learning: Basic Principles Solutions of Homework problems: Exercise Session 3

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## 1 Utility Theory:

Consider the actions  $\alpha_{25}$ ,  $\alpha_{50}$ ,  $\alpha_{75}$  as ordering 25, 50 and 75 pairs of Skis respectively. Let us denote the purchase cost of per pair of Ski when action is  $\alpha_i$ , by  $PC_i$ . Consider the states  $D_{30}$ ,  $D_{40}$ ,  $D_{50}$ ,  $D_{60}$  when the demand of Ski is 30, 40, 50 and 60 pairs respectively. Let us denote the probability of different state by  $p(D_k)$ . It is given that  $p(D_{30}) = .2$ ,  $p(D_{40}) = .4$ ,  $p(D_{50}) = .2$  and  $p(D_{60}) = .2$ . Let us define utility of an action  $\alpha_i$  when state is  $D_k$  by  $U_{i,k}$ 

$$U_{i,k} = \begin{cases} (i \times 75) - (i \times PC_i) & \text{when } k = i \\ (i \times 75) - (i \times PC_i) - (k - i) \times 5 & \text{when } k > i \\ (k \times 75) - (i \times PC_i) + (i - k) \times 25 & \text{when } k < i \end{cases}$$

From utility theory we know, Expected utility of an action  $\alpha_i$  is:,

$$E[\alpha_i] = \sum_k U_{i,k} p(D_k)$$

Table 1: Utility Table

$Actions(\alpha_i) \downarrow$	$D_{30}$	$D_{40}$	$D_{50}$	$D_{60}$	$\mathrm{E}[\alpha_i] \downarrow$
$\alpha_{25}$	600	550	500	450	530
$lpha_{50}$	500	1000	1500	1450	1090
$lpha_{75}$	375	875	1375	1875	1075

$$E[\alpha_{25}] = 600 \times .2 + 550 \times .4 + 500 \times .2 + 450 \times .2 = 530$$

$$E[\alpha_{50}] = 500 \times .2 + 1000 \times .4 + 1500 \times .2 + 1450 \times .2 = 1090$$

$$E[\alpha_{75}] = 375 \times .2 + 875 \times .4 + 1375 \times .2 + 1875 \times .2 = 1075$$

We find that the Expected utility is maximized at action  $\alpha_{50}$ , that is ordering 50 pairs Skis. So the optimum decision is to order 50 pairs of Ski

## 2 Posterior Distribution:

The posterior distribution of  $\theta$  is  $p(\theta|N)=\frac{p(N|\theta)p(\theta)}{p(N)}$ . Now, the likelihood,  $p(N|\theta)=\binom{N}{m}\theta^m(1-\theta)^{N-m}$  which is given. The prior  $p(\theta)$  is Uni(0,1). So, we can write  $p(\theta)=1$ 

So the posterior distribution of  $\theta$  in case of uniform prior is:

$$\begin{array}{lcl} p(\theta|N) & = & \frac{\binom{N}{m}\theta^m(1-\theta)^{N-m}}{p(N)} \\ \\ p(\theta|N) & \propto & \binom{N}{m}\theta^m(1-\theta)^{N-m} \text{ ,As } p(N) \text{ is constant for a given } N \end{array}$$

And the posterior distribution of  $\theta$  in case of prior as beta distribution is:

$$\begin{array}{lll} p(\theta|N) & = & \frac{\binom{N}{m}\theta^m(1-\theta)^{N-m}p(\theta)}{p(N)} \\ p(\theta|N) & = & \frac{\binom{N}{m}\theta^m(1-\theta)^{N-m}Beta(\theta|\alpha,\beta)}{p(N)} \\ p(\theta|N) & = & \frac{\binom{N}{m}\theta^m(1-\theta)^{N-m}\frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)}\theta^{\alpha-1}(1-\theta)^{\beta-1}}{p(N)} \\ p(\theta|N) & = & \frac{\binom{N}{m}\theta^{m+\alpha-1}(1-\theta)^{N-m+\beta-1}\frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)}}{p(N)} \\ p(\theta|N) & \propto & \binom{N}{m}\theta^{m+\alpha-1}(1-\theta)^{N-m+\beta-1} \text{ ,As } p(N) \text{ and } \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \text{ are constant } \end{array}$$

## 3 Bayes Estimation:

Let us assume that the sext Y of five parts has been investigated and X number of parts have been found with defect. The posterior distribution of  $\theta$  is  $p(\theta|Y) = \frac{p(Y|\theta)p(\theta)}{p(Y)}$ . Now, the likelihood,  $p(Y|\theta) = \binom{5}{X}\theta^X(1-\theta)^{5-X}$ . The prior  $p(\theta)$  has Beta(1,9) distribution. So, we can write  $p(\theta) = \frac{\Gamma(1+9)}{\Gamma(1)\Gamma(9)}\theta^{1-1}(1-\theta)^{9-1} = \frac{\Gamma(10)}{\Gamma(1)\Gamma(9)}(1-\theta)^8$ . So the posterior distribution of  $\theta$  in case of prior as Beta(1,9) is:

$$\begin{split} p(\theta|N) &= \frac{\binom{5}{X}\theta^X(1-\theta)^{5-X}\frac{\Gamma(10)}{\Gamma(1)\Gamma(9)}(1-\theta)^8}{p(N)} \\ p(\theta|N) &= \frac{\binom{5}{X}\theta^X(1-\theta)^{5-X+8}\frac{\Gamma(10)}{\Gamma(1)\Gamma(9)}}{p(N)} \\ p(\theta|N) &= \frac{\binom{5}{X}\theta^{(X+1)-1}(1-\theta)^{(14-X)-1}\frac{\Gamma(10)}{\Gamma(1)\Gamma(9)}}{p(N)} \\ p(\theta|N) &= \frac{\binom{5}{X}\theta^{(X+1)-1}(1-\theta)^{(14-X)-1}\frac{\Gamma(X+1+14-X)}{\Gamma(X+1)\Gamma(14-X)}C}{p(N)} \text{ (As } \frac{\Gamma(10)}{\Gamma(1)\Gamma(9)} \text{ is a constant)} \\ p(\theta|N) &= C_1\theta^{(X+1)-1}(1-\theta)^{(14-X)-1}\frac{\Gamma(X+1+14-X)}{\Gamma(X+1)\Gamma(14-X)} \text{ (As } \frac{\Gamma(10)}{\Gamma(1)\Gamma(9)} \text{ is a constant)} \\ p(\theta|N) &= C_1Beta(\theta|X+1,14-X) \end{split}$$

The Expectation of  $\theta$  from a beta function  $Beta(\theta|\alpha,\beta)$  is  $E[\theta] = \frac{\alpha}{\alpha+\beta}$  ( $Ref:http://en.wikipedia.org/wiki/Beta_distribution$ ). In Baye's estimation, we calculate the Expectation of the probability distribution of  $\theta$ . So here we see the Bayes estimation of the posterior probability is  $\frac{X+1}{X+1+14-X}$  ( $C_1=1$ , as Beta is a distribution). So we can write:

The Bayes estimate of the proportion of defective parts is  $\frac{X+1}{15}$