

## Exercise 2

Junsong Tang

junsong.tang@stat.ubc.ca

## 1 define a Bayesian model

1. The unknown quantity is the value of  $p$ , and the data is the observed value of  $Y_i$ 's.
- 2.

$$X \sim \rho, \rho = (\rho_k)_{k=1}^K$$

$$Y|X \sim \text{Bernoulli}(X)$$

## 2 posterior and point estimates

```

11 prior_probabilities = NULL
12 realizations = NULL
13 for (k in (0 : K)) {
14     prior_probabilities = c(prior_probabilities, (k/K)*(1-k/K))
15     realizations = c(realizations, k/K)
16 }
17 prior_probabilities
18 realizations
19 plot(realizations, prior_probabilities, type="h", main = "prior pmf")
20
21 posterior_probabilities = NULL
22 Z = 0
23 for (k in (0 : K)) {
24     # likelihood = P(X = k/K) * P(Y = (1,1,1) | X = k/K)
25     likelihood = realizations[k+1]**3 * prior_probabilities[k+1]
26     Z = Z + likelihood
27     posterior_probabilities = c(posterior_probabilities, likelihood)
28 }
29 posterior_probabilities = posterior_probabilities / Z
30 posterior_probabilities
31 plot(realizations, posterior_probabilities, type = "h", main = "posterior pmf")
32
33
34

```

3. From the plot, we can see the posterior mode is at  $X = 0.8$

```

41 # posterior mean
42 m = 0
43 for (k in (0 : K)) {
44     x = realizations[k+1]
45     pi_k = posterior_probabilities[k+1]
46     m = m + x * pi_k
47 }
48 m #0.712497759455099
49
50

```

## 3 Bayes action

1. xxx

2. xxx

3. xxx