

Model with Categorical Parameter Space

- The frequency (number of claims) is given by die A_1 or A_2 , where a marked face indicates an insurance claim.

$$\Pr\{A_1\} = \Pr\{A_2\} = \frac{1}{2}, \text{ and}$$

$$\Pr\{\text{Claim}|A_1\} = \frac{1}{6}, \quad \Pr\{\text{Claim}|A_2\} = \frac{1}{2}.$$

- The severity (claim amount) is given by spinner B_1 or B_2 .

$$\Pr\{B_1\} = \Pr\{B_2\} = \frac{1}{2}, \text{ and}$$

$$\Pr\{\text{Amount} = 2|B_1\} = \frac{5}{6}, \quad \Pr\{\text{Amount} = 14|B_1\} = \frac{1}{6},$$

$$\Pr\{\text{Amount} = 2|B_2\} = \frac{1}{2}, \quad \Pr\{\text{Amount} = 14|B_2\} = \frac{1}{2}.$$

- Let X_k = total claim costs in year k .

Then $X_k = 0$ if there is no claim,

= 2 if there is a claim, and its amount is 2, and

= 14 if there is a claim, and its amount is 14.

- Note that in this case, $\theta = [A_i, B_j]$ for some $i = 1, 2$ and $j = 1, 2$;

thus, θ is a categorical, as opposed to real-valued, state parameter.

Some Bayesian Calculations

$$\begin{aligned}
 \bullet \quad E[X_1] &= (0)\Pr\{X_1 = 0\} + (2)\Pr\{X_1 = 2\} + (14)\Pr\{X_1 = 14\} \\
 &= (0) \sum_{i=1}^2 \sum_{j=1}^2 \Pr\{X_1 = 0|A_i, B_j\} \Pr\{A_i, B_j\} \\
 &\quad + (2) \sum_{i=1}^2 \sum_{j=1}^2 \Pr\{X_1 = 2|A_i, B_j\} \Pr\{A_i, B_j\} \\
 &\quad + (14) \sum_{i=1}^2 \sum_{j=1}^2 \Pr\{X_1 = 14|A_i, B_j\} \Pr\{A_i, B_j\}, \text{ or}
 \end{aligned}$$

$$E[X_1] = \sum_{i=1}^2 \sum_{j=1}^2 E[X_1|A_i, B_j] \Pr\{A_i, B_j\}.$$

$$\begin{aligned}
 \bullet \quad E[X_2|X_1] &= (0)\Pr\{X_2 = 0|X_1\} + (2)\Pr\{X_2 = 2|X_1\} + (14)\Pr\{X_2 = 14|X_1\} \\
 &= (0) \sum_{i=1}^2 \sum_{j=1}^2 \Pr\{X_2 = 0|X_1, A_i, B_j\} \Pr\{A_i, B_j|X_1\} \\
 &\quad + (2) \sum_{i=1}^2 \sum_{j=1}^2 \Pr\{X_2 = 2|X_1, A_i, B_j\} \Pr\{A_i, B_j|X_1\} \\
 &\quad + (14) \sum_{i=1}^2 \sum_{j=1}^2 \Pr\{X_2 = 14|X_1, A_i, B_j\} \Pr\{A_i, B_j|X_1\} \\
 &= (0) \sum_{i=1}^2 \sum_{j=1}^2 \Pr\{X_2 = 0|A_i, B_j\} \Pr\{A_i, B_j|X_1\} \\
 &\quad + (2) \sum_{i=1}^2 \sum_{j=1}^2 \Pr\{X_2 = 2|A_i, B_j\} \Pr\{A_i, B_j|X_1\} \\
 &\quad + (14) \sum_{i=1}^2 \sum_{j=1}^2 \Pr\{X_2 = 14|A_i, B_j\} \Pr\{A_i, B_j|X_1\},
 \end{aligned}$$

$$\text{where } \Pr\{A_i, B_j|X_1\} = \frac{\Pr\{X_1|A_i, B_j\} \Pr\{A_i, B_j\}}{\sum_{i=1}^2 \sum_{j=1}^2 \Pr\{X_1|A_i, B_j\} \Pr\{A_i, B_j\}} \text{ or}$$

$$\begin{aligned}
 E[X_2|X_1] &= \sum_{i=1}^2 \sum_{j=1}^2 E[X_2|X_1, A_i, B_j] \Pr\{A_i, B_j|X_1\} \\
 &= \sum_{i=1}^2 \sum_{j=1}^2 E[X_2|A_i, B_j] \Pr\{A_i, B_j|X_1\},
 \end{aligned}$$

where $\Pr\{A_i, B_j|X_1\}$ is as above.

Some Bühlmann Calculations

- $C = zX_1 + (1 - z)H,$

where $H = E[X_1]$ and $z = \frac{n}{n + \frac{EVPV}{VHM}}.$

- From above, we know that

$$H = E[X_1] = \sum_{i=1}^2 \sum_{j=1}^2 E[X_1 | A_i, B_j] \Pr\{A_i, B_j\}.$$

- Similarly,

$$VHM = \sum_{i=1}^2 \sum_{j=1}^2 \left(E[X_1 | A_i, B_j] \right)^2 \Pr\{A_i, B_j\} - (E[X_1])^2.$$

- Finally,

$$\begin{aligned} EVPV &= \sum_{i=1}^2 \sum_{j=1}^2 \text{Var}[X_1 | A_i, B_j] \Pr\{A_i, B_j\} \\ &= \sum_{i=1}^2 \sum_{j=1}^2 E[X_1^2 | A_i, B_j] - \left(E[X_1 | A_i, B_j] \right)^2 \Pr\{A_i, B_j\}. \end{aligned}$$