

# INF283 | Weekly Exercise 06 | Bayesian inference

## Exercise 1.1

1.

$$P(A) = 1/5$$

$$P(B) = 1/5 \cdot (0 + 0 + 1/7 + 1/11 + 1/20) \approx 0,05675$$

$$P(B|A) = 1/11$$

$$P(A|B) = \frac{P(B|A) \cdot P(A)}{P(B)} = \frac{1/11 \cdot 1/5}{0,05675} \approx 0,32$$

2.

$$P(A) = 1/5$$

$$P(B) = 0,05675$$

$$P(B|A) = 1/20$$

$$P(A|B) = \frac{1/11 \cdot 1/20}{0,05675} \approx 0,176$$

3. Not possible to get an 18 with an 11-sided die, so the probability of getting this die is 0

4. The 20-sided die is the only die that can give a 18, therefore the probability of getting this die is 1.

$$P(\text{Die} = 20 \mid X = 18) = \frac{1/20 \cdot 1/5}{1/100} = 1$$

$$P(B) = 1/5 \cdot 1/20 = \frac{1}{100}$$

## Exercise 1.2

$$\theta \in [0, 1]$$

$$P(H) = \theta \text{ (getting a head)}$$

$$P(T) = 1 - \theta \text{ (getting a tail)}$$

1. Probability mass function

$$P(X=H) = \binom{n}{k} \theta^k (1-\theta)^{n-k}$$

Fair coin:  $\theta = 0,5$

$$= \binom{15}{9} 0,5^9 (1-0,5)^6$$
$$= 0,15274$$

$\theta^n = k \quad n = 15$

b)

b1. Maximum likelihood

$$P(D | \theta) = \theta^9 \cdot (1-\theta)^6$$

$$\text{Solve } P'(D | \theta) = 0$$

$$\hat{\theta} = \frac{9}{15} = 0.6 \quad \left\{ \begin{array}{l} \text{Number of successful events} \\ \text{Total number of sample space} \end{array} \right.$$

b2. Likelihood with  $\theta = 0.6$

$$\begin{aligned} P(D | \theta) &= \theta^{N_h} (1-\theta)^{N_t} = 0.6^9 (1-0.6)^6 \\ &\approx 0.00004 \end{aligned}$$

b3. Likelihood when coin is fair,  $\theta = 0.5$

$$\begin{aligned} P(D | \theta) &= \theta^{N_h} (1-\theta)^{N_t} = 0.5^9 (1-0.5)^6 \\ &\approx 0.00003 \end{aligned}$$

b4. Posterior distribution  $P(\theta | D)$ , uniform prior

$$\theta \sim \text{Beta}(\alpha, \beta)$$

Uniform prior : parameters

$$\alpha = \beta = 1, \quad N_h = 9, \quad N_t = 6$$

Uniform:

$$\alpha = 1$$

$$\beta = 1$$

Posterior :

$$\theta \sim \text{Beta}(N_h + \alpha, N_t + \beta)$$

$$\theta \sim \text{Beta}(9 + 1, 6 + 1) = \text{Beta}(10, 7)$$

b5.

Maximum likelihood

$$P(x=H) = \text{Undefined, cannot divide on 0}$$

$$P(x=H | D=H) = 1$$

$$P(x=T | D=HH) = 0$$

$$P(x=H | D=HHT) = 2/3$$

$$P(x=H | D=HHTH) = 3/4$$

$$P(x=T | D=HHTHH) = 1/5$$

Bayesian inference with uniform priors:

$$P(x=H) = \frac{1+0}{1+0+1+0} = 1/2$$

$$P(x=H | D=H) = \frac{1+1}{1+1+1+0} = 2/3$$

$$P(x=T | D=HH) = \frac{1+0}{1+0+1+2} = 1/4$$

$$P(x=H | D=HHT) = \frac{1+2}{1+2+1+1} = 3/5$$

$$P(x=H | D=HHTH) = \frac{1+3}{1+3+1+1} = 4/6 = 2/3$$

$$P(x=T | D=HHTHH) = \frac{1+1}{1+1+1+4} = 2/7$$

$$N_h = 7 \quad N_+ = 3$$

c)

c1. Maximum likelihood

$$P(D|\theta) = \theta^7 \cdot (1-\theta)$$

$$\text{Solve } P'(D|\theta) = 0$$

$$\hat{\theta} = \frac{7}{7+1} = 0,7$$

c2. Likelihood with  $\theta = 0,7$

$$P(D|\theta) = 0,7^7 (1-0,7)^1 \approx 0,00222$$

c3. Likelihood when coin is fair,  $\theta = 0,5$

$$P(D|\theta) = 0,5^7 (1-0,5)^1 \approx 0,00097$$

c4. Posterior distribution  $P(\theta|D)$ , uniform prior

$$\theta \sim \text{Beta}(\alpha, \beta)$$

Uniform prior : parameters

$$\alpha = \beta = 1, \quad N_h = 7, \quad N_+ = 3$$

Posterior :

$$\theta \sim \text{Beta}(N_h + \alpha, N_+ + \beta)$$

$$\theta \sim \text{Beta}(7 + 1, 3 + 1) = \text{Beta}(8, 4)$$

c5.

Maximum likelihood

$$P(x=T) = \text{Undefined, cannot divide on 0}$$

$$P(x=T | D=T) = 1$$

$$P(x=T | D=TT) = 1$$

$$P(x=H | D=TTT) = 0$$

$$P(x=H | D=TTTH) = 1/4$$

$$P(x=H | D=TTTHH) = 2/5$$

Bayesian inference with uniform prior:

$$P(x=T) = \frac{1+0}{1+0+1+0} = 1/2$$

$$P(x=T | D=T) = \frac{1+1}{1+1+1+0} = 2/3$$

$$P(x=T | D=TT) = \frac{1+2}{1+2+1+0} = 3/4$$

$$P(x=H | D=TTT) = \frac{1+0}{1+0+1+3} = 1/5$$

$$P(x=H | D=TTTH) = \frac{1+1}{1+1+1+3} = 2/6 = 1/3$$

$$P(x=H | D=TTTHH) = \frac{1+2}{1+2+1+3} = 3/7$$

## Exercise 1.3

weather: Sunny  
 temperatur: Hot  
 humidity: Normal  
 wind: false

**weather: Sunny**

PlayTennis	Rainy	Overcast	Sunny	Total
Yes	3	4	2	9
No	2	0	3	5
<b>Total</b>	<b>5</b>	<b>4</b>	<b>5</b>	<b>14</b>

**humidity: Normal**

PlayTennis	High	Normal	Total
Yes	4	5	9
No	3	2	5
<b>Total</b>	<b>7</b>	<b>7</b>	<b>14</b>

**temperatur: Hot**

PlayTennis	Hot	Mild	Cool	Total
Yes	3	4	2	9
No	1	2	2	5
<b>Total</b>	<b>4</b>	<b>6</b>	<b>4</b>	<b>14</b>

**wind: False**

PlayTennis	True	False	Total
Yes	3	6	9
No	3	2	5
<b>Total</b>	<b>6</b>	<b>8</b>	<b>14</b>

$$P(\text{PlayTennis} = \text{Yes} \mid \text{weather} = \text{Sunny}, \text{temp} = \text{Hot}, \text{humidity} = \text{Normal}, \text{wind} = \text{False}) \propto$$

$$P(\text{PlayTennis} = \text{Yes}, \text{weather} = \text{Sunny}, \text{temp} = \text{Hot}, \text{humidity} = \text{Normal}, \text{wind} = \text{False})$$

$$= \frac{9}{14} \cdot \frac{2}{9} \cdot \frac{3}{9} \cdot \frac{5}{9} \cdot \frac{6}{9} \approx \underline{0,0176}$$

$$P(\text{PlayTennis} = \text{No}, \text{weather} = \text{Sunny}, \text{temp} = \text{Hot}, \text{humidity} = \text{Normal}, \text{wind} = \text{False})$$

$$= \frac{5}{14} \cdot \frac{3}{5} \cdot \frac{1}{5} \cdot \frac{2}{5} \cdot \frac{2}{5} \approx 0,0069$$

$$= \frac{0,0176}{0,0176 + 0,0069} \approx \underline{\underline{0,718}}$$

## Exercise 1.4

weather: Sunny

temperatur: Hot

humidity: Unknown

wind: Unknown

**weather: Sunny**

PlayTennis	Rainy	Overcast	Sunny	Total
Yes	3	4	2	9
No	2	0	3	5
Total	5	4	5	14

**temperatur: Hot**

PlayTennis	Hot	Mild	Cool	Total
Yes	3	4	2	9
No	1	2	2	5
Total	4	6	4	14

$$P(\text{PlayTennis} = \text{Yes} \mid \text{weather} = \text{Sunny}, \text{temp} = \text{Hot}) \propto$$

$$P(\text{PlayTennis} = \text{Yes}, \text{weather} = \text{Sunny}, \text{temp} = \text{Hot})$$

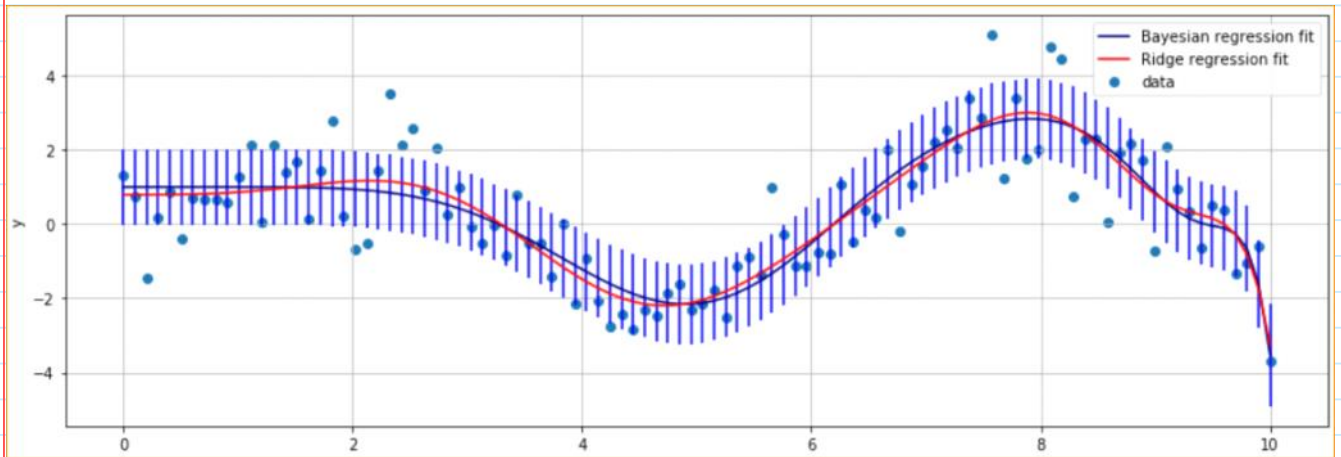
$$= \frac{9}{14} \cdot \frac{2}{9} \cdot \frac{3}{9} \approx 0,047$$

$$P(\text{PlayTennis} = \text{No}, \text{weather} = \text{Sunny}, \text{temp} = \text{Hot})$$

$$= \frac{5}{14} \cdot \frac{3}{5} \cdot \frac{1}{5} \approx 0,043$$

$$= \frac{0,047}{0,047 + 0,043} \approx \underline{\underline{0,52}}$$

## Exercise 1.5



Modified code:

```
plt.figure(figsize=(15, 5))
plt.scatter(X, y, label='data')
plt.plot(X, y_predict_bayesian, color='navy', label='Bayesian regression fit')
plt.plot(X, y_predict_ridge, color='r', label='Ridge regression fit')

for i in range(0, len(y_predict_bayesian_std[0])):
    plt.vlines(X[i], y_predict_bayesian_std[0][i] - y_predict_bayesian_std[1][i],

plt.grid(b=True)
plt.xlabel('X')
plt.ylabel('y')
plt.legend(loc='best')
```