

$$1. (a) P(W_h | g_a) = \frac{P(g_a | W_h) \cdot P(W_h)}{P(g_a)} = \frac{0.99 \times 0.85}{0.95} = 88.6\%$$

$$(b) P(g_a | \neg W_h) = \frac{P(\neg W_h | g_a) \cdot P(g_a)}{P(\neg W_h)} = \frac{(1 - P(W_h | g_a)) \cdot P(g_a)}{1 - P(W_h)} = \frac{(1 - 0.886) \cdot 0.95}{1 - 0.95} = 72.3\%$$

$$2. P(g_t) = 0.9 \quad P(b_t) = 0.1 \quad P(b | b_t) = 0.75 \quad P(b | \neg b_t) = 1 - 0.75 = 0.25$$

$$P(b_t | b) = \frac{P(b | b_t) \cdot P(b_t)}{P(b | b_t) \cdot P(b_t) + P(b | \neg b_t) \cdot P(\neg b_t)} = \frac{0.75 \times 0.1}{0.75 \times 0.1 + 0.25 \times 0.9} = 0.25$$

$$P(\neg b_t | b) = \frac{P(b | \neg b_t) \cdot P(\neg b_t)}{P(b | b_t) \cdot P(b_t) + P(b | \neg b_t) \cdot P(\neg b_t)} = \frac{0.25 \times 0.9}{0.75 \times 0.1 + 0.25 \times 0.9} = 0.75$$

75% chance it is a green taxi

$$3. (a) P(H) = \frac{1}{3} \times \frac{3}{3} \times \frac{1}{3} \times \frac{3}{8} = \frac{1}{24} \quad P(NH) = \frac{2}{5} \times \frac{1}{5} \times \frac{3}{3} \times \frac{5}{8} = \frac{9}{200} \quad \text{Not happy}$$

$$(b) P(H) = \frac{1}{4} \times \frac{1}{2} + \frac{3}{4} \times \frac{1}{2} + \frac{1}{4} \times \frac{1}{2} = \frac{5}{8} \quad \text{Happy}$$

$$4. P(C++) = 0.5 \quad P(\text{Java}) = 0.4 \quad P(C++ | ms) = 0.99$$

$$P(\text{Java} | ms) = 0.98 \quad P(ms) = 0.01$$

$$P(ms | C++ \wedge \text{Java}) = \frac{P(C++ \wedge \text{Java} | ms) \cdot P(ms)}{P(C++ \wedge \text{Java})} = \frac{P(C++ | ms) \cdot P(\text{Java} | ms)}{P(C++ \wedge \text{Java})}$$

$$P(ms) = \frac{0.99 \times 0.98 \times 0.01}{0.5 \times 0.4} = 48.5\%$$

5. from the material we know that  $P(Y=1|X) = \frac{1}{1 + \exp(\ln \frac{P(X|Y=0)}{P(X|Y=1)} \frac{P(Y=0)}{P(Y=1)})}$

$$= \frac{1}{1 + \exp(\ln \frac{P(Y=0)}{P(Y=1)} + \sum_i \ln \frac{P(X_i|Y=0)}{P(X_i|Y=1)})}$$

the prior for  $P(Y=1) = \pi$  and  $P(Y=0) = 1 - \pi$

$$P(Y=1|X) = \frac{1}{1 + \exp(\ln \frac{1-\pi}{\pi} + \sum_i \ln \frac{\theta_{i0}^{x_i} (1-\theta_{i0})^{(1-x_i)}}{\theta_{i1}^{x_i} (1-\theta_{i1})^{(1-x_i)}})}$$

$$= \frac{1}{1 + \exp(\ln \frac{1-\pi}{\pi} + \sum_i x_i \ln \frac{\theta_{i0}}{\theta_{i1}} + (1-x_i) \ln \frac{(1-\theta_{i0})}{(1-\theta_{i1})})}$$

$$= \frac{1}{1 + \exp(\ln \frac{1-\pi}{\pi} + (\frac{1-\theta_{i0}}{1-\theta_{i1}}) + \sum_i x_i (\ln \frac{\theta_{i0}}{\theta_{i1}} - \ln \frac{(1-\theta_{i0})}{(1-\theta_{i1})}))}$$

$$w_0 = \ln \frac{1-\pi}{\pi} + \sum_i \ln \frac{(1-\theta_{i0})}{(1-\theta_{i1})} \quad \text{and} \quad w_1 = \ln \frac{\theta_{i0}}{\theta_{i1}} - \ln \frac{(1-\theta_{i0})}{(1-\theta_{i1})}$$

$$P(Y=1|X) = \frac{1}{1 + \exp(\sum_i w_i x_i)}$$

6. based on Tom M. Mitchell's new book

$$P(Y=1|X) = \frac{1}{1 + \exp(\ln \frac{P(Y=0)}{P(Y=1)} + \sum_i \ln \frac{P(X_i|Y=0)}{P(X_i|Y=1)})}$$

$$= \frac{1}{1 + \exp(\ln \frac{1-\pi}{\pi} + \sum_i \ln \frac{P(X_i|Y=0)}{P(X_i|Y=1)})}$$

$$\sum_i \ln \frac{P(X_i|Y=0)}{P(X_i|Y=1)} = \sum_i \ln \left( \frac{h(X_i, \theta_{i0}) \exp(-\theta_{i0}^T X_i + c)}{h(X_i, \theta_{i1}) \exp(-\theta_{i1}^T X_i + c)} \right) = \sum_i \left( \ln \frac{h(X_i, \theta_{i0})}{h(X_i, \theta_{i1})} + (-\theta_{i0}^T X_i + c) - (-\theta_{i1}^T X_i + c) \right) = \sum_i \left( \ln \frac{h(X_i, \theta_{i0})}{h(X_i, \theta_{i1})} + \theta_{i1}^T X_i - \theta_{i0}^T X_i \right)$$

$$P(Y=1|X) = \frac{1}{1 + \exp(\ln \frac{1-\pi}{\pi} + \sum_i \ln \frac{h(X_i, \theta_{i0})}{h(X_i, \theta_{i1})} + \sum_i X_i (\theta_{i1}^T - \theta_{i0}^T))}$$

$$= \frac{1}{1 + \exp(-w_0 - \sum_i w_i x_i)} \quad w_0 = -\ln \frac{1-\pi}{\pi} - \sum_i \ln \frac{h(X_i, \theta_{i0})}{h(X_i, \theta_{i1})}$$

$$w_1 = \theta_{i0}^T - \theta_{i1}^T$$