

Using this MRF, we can see that all V_1, \dots, V_n are (conditionally) independent given C . This will allow us to use the known appearance models to get $P(C | V_1, \dots, V_n)$

is $V_i \perp V_j | C$ for all i, j then

$$P(V_1, \dots, V_n | C) = \prod_{i=1}^n P(V_i | C)$$

Using Bayes Rule:

$$P(C | V_1, \dots, V_n) = \frac{P(C, V_1, \dots, V_n)}{P(V_1, \dots, V_n)} = \frac{P(V_1, \dots, V_n | C) P(C)}{P(V_1, \dots, V_n)}$$

Now suppose $C = \{C_i = i, C_j = j, C_{\dots} = \dots\}$
and we sample C and C_i to get $C' = \{C_i = i, C_j = j, C_{\dots} = \dots\}$

$$\text{Then } A(C' | C) = \min \left(1, \frac{P(V_1, \dots, V_n | C') \cdot P(C')}{P(V_1, \dots, V_n)} \right)$$

$$\frac{P(V_1, \dots, V_n | C) \cdot P(C)}{P(V_1, \dots, V_n)}$$

$$= \min \left(1, \frac{P(V_1, \dots, V_n | C') \cdot P(C')}{P(V_1, \dots, V_n | C) \cdot P(C)} \right) \rightarrow P(C) = P(C') = \frac{1}{K_i} \rightarrow \frac{P(C') - 1}{P(C)}$$

$$= \min \left(1, \frac{P(V_1, \dots, V_n | C_i = j, C_j = i, C_{\dots} = \dots)}{P(V_1, \dots, V_n | C_i = i, C_j = j, C_{\dots} = \dots)} \right)$$

$$= \min \left(1, \frac{P(V_i | C_i = j) \cdot P(V_j | C_j = i)}{P(V_i | C_i = i) \cdot P(V_j | C_j = j)} \right)$$