From the task description we get: $P(a \mid r) = 0.3$

$$P(o \mid r) = 0.4$$

 $P(l \mid r) = 0.3$

$$P(a \mid g) = 0.3$$

 $P(o \mid g) = 0.3$

$$P(l \mid q) = 0.4$$

$$P(a \mid b) = 0.4$$

 $P(a \mid b) = 0.5$

$$P(a | b) = 0.5$$

$$P(o \mid b) = 0.5$$

$$P(l \mid b) = 0$$

We can now determine the following probabilities:

$$P(a \cap r) = P(a \mid r) \cdot P(r) = 0.3 \cdot 0.2 = 0.06$$

$$P(o \cap r) = P(o \mid r) \cdot P(r) = 0.4 \cdot 0.2 = 0.08$$

$$P(l \cap r) = P(l \mid r) \cdot P(r) = 0.3 \cdot 0.2 = 0.06$$

$$P(a \cap g) = P(a \mid g) \cdot P(g) = 0.3 \cdot 0.6 = 0.18$$

$$P(o \cap g) = P(o \mid g) \cdot P(g) = 0.3 \cdot 0.6 = 0.18$$

$$P(l \cap g) = P(l \mid g) \cdot P(g) = 0.4 \cdot 0.6 = 0.24$$

$$P(a \cap b) = P(a \mid b) \cdot P(b) = 0.5 \cdot 0.2 = 0.1$$

$$P(o \cap b) = P(o \mid b) \cdot P(b) = 0.5 \cdot 0.2 = 0.1$$

$$P(l \cap b) = P(l \mid b) \cdot P(b) = 0 \cdot 0.2 = 0$$

(a)

$$P(a) = P(a \cap r) + P(a \cap g) + P(a \cap b) = 0.06 + 0.18 + 0.1 = 0.34$$

$$P(o) = P(o \cap r) + P(o \cap g) + P(o \cap b) = 0.08 + 0.18 + 0.1 = 0.36$$

$$P(l) = P(l \cap r) + P(l \cap g) + P(l \cap b) = 0.06 + 0.24 + 0 = 0.3$$

Thus, the probability of selecting an apple is 0.34.

(b)

$$P(r \mid a) = \frac{P(a \cap r)}{P(a)} = \frac{0.06}{0.34} = 0.176$$

$$P(g \mid a) = \frac{P(a \cap g)}{P(a)} = \frac{0.18}{0.34} = 0.529$$

$$P(b \mid a) = \frac{P(a \cap b)}{P(a)} = \frac{0.1}{0.34} = 0.294$$

$$P(r \mid o) = \frac{P(o \cap r)}{P(o)} = \frac{0.08}{0.36} = 0.222$$

$$P(g \mid o) = \frac{P(o \cap g)}{P(o)} = \frac{0.18}{0.36} = 0.5$$

$$P(b \mid o) = \frac{P(o \cap b)}{P(o)} = \frac{0.1}{0.36} = 0.278$$

$$P(r \mid l) = \frac{P(l \cap r)}{P(l)} = \frac{0.06}{0.3} = 0.2$$

$$P(g \mid l) = \frac{P(l \cap g)}{P(l)} = \frac{0.24}{0.3} = 0.8$$

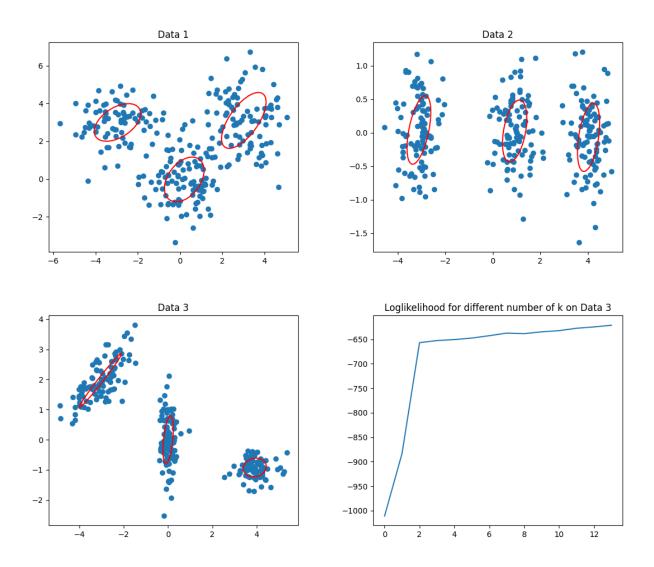
$$P(b \mid l) = \frac{P(l \cap b)}{P(l)} = \frac{0}{0.3} = 0$$

Thus, if we observe that the selected fruit is in fact an orange, the probability that it came from the green box is 0.5.

(a)

(a)

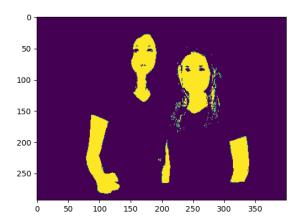
(g)



We can see, that in all three data sets there are 3 clusters (more so in data2 and data3 than in data1). Therefore, K = 3 clusters is a very suitable number of clusters for this application.

K=2 yields very under fitting cluster parameters, even for data1. For K>3 we get clusters that are "overlapping" each other, so we see overfitting.

(h)



Increasing K here improves accuracy on the training data set, thus reducing generalization. For K=9, a model is produced that separates the colors in the test image better than the default of K=3. Increasing K further might lead to poor accuracy on other test data. Since the colors are there dimensional, and we are looking at two classifications (skin and no-skin), intuitively speaking, 3*2 clusters should be a good starting point for finding a suitable number of clusters.

Increasing the threshold *theta* increases the required confidence, that a given pixel color represents skin. This leads to a more conservative classification (so rather non-skin than skin), which is more robust to noise, however, can also be less generalized. We increased *theta* to 3 and got the above result.