**Expectation-Maximization for Naïve Bayes** Use **PENCIL** to do this exercise. First Iteration. You have data from a Naïve Bayes model with N = 2 (so there are two binary features,  $A_1$  and  $A_2$ , and one binary class variable C. The structure of the Bayes net is  $A_1 \leftarrow C \rightarrow A_2$ .

The data is as follows: 10 values for  $(A_1, A_2) = (t, t), (t, f), (t, f), (t, f), (t, t), (f, f), (t, t), (t, t), (t, t), (t, t)$ . Fill in the values given in class for the initial guess for the parameters of this model:

$$\begin{array}{c|c} \theta_c = P(C=t) = \boxed{0.6} & \theta_{11} = P(A_1 = t | C = t) = \boxed{0.9} & \theta_{10} = P(A_1 = t | C = f) = \boxed{0.3} \\ \theta_{21} = P(A_2 = t | C = t) = \boxed{0.2} & \theta_{20} = P(A_2 = t | C = f) = \boxed{0.6} \end{array}$$

First, you will caculate  $P(C|A_1, A_2)$  for all possible values of  $A_1, A_2$  and C. Do this by filling in the following table

1	2	3	4	5	6	7	8
$A_1$	$A_2$	C	$P(A_1 C)$	$P(A_2 C)$	P(C)	$P(C, A_1, A_2) = 4 \times 5 \times 6$	$P(C A_1, A_2) = $ normalise 7 over $C$
t	t	t	0.9	0.2	0.6	0.108	0.6
t	t	f	0.3	0.6	0.4	0.072	0.4
t	f	t	0.9	0.8	0.6	0.432	0.9
t	f	f	0.3	0.4	0.4	0.048	0.1
f	t	t	0.1	0.2	0.6	0.012	0.07
f	t	f	0.7	0.6	0.4	0.168	0.93
f	f	t	0.1	0.8	0.6	0.048	0.3
f	f	f	0.7	0.4	0.4	0.112	0.7

Now, "complete" the table of data below with the values in the table you just filled in.

$A_1$	$A_2$	C	$P(C, A_1, A_2)$	$P(C A_1, A_2)$
t	t	t	0.108	0.6
t	$\mathbf{t}$	f	0.072	0.4
t	f	t	0.432	0.9
t	f	f	0.048	0.1
t	f	t	0.432	0.9
t	f	f	0.048	0.1
f	f	t	0.048	0.3
f	f	f	0.112	0.7
t	t	t	0.108	0.6
t	t	f	0.072	0.4
f	$\mathbf{t}$	t	0.012	0.07
f	t	f	0.168	0.93
f	f	t	0.048	0.3
f	f	f	0.112	0.7
t	$\mathbf{t}$	t	0.108	0.6
t	t	f	0.072	0.4
t	t	t	0.108	0.6
t	t	f	0.072	0.4
t	t	t	0.108	0.6
t	t	f	0.072	0.4
put sum	of all rov	vs here:	2.36	10

When the sum of all rows in the second-to-last column stops changing, you can stop iterating. If  $d_j = \{a_{j1}, a_{j2}\}$  is the data for  $j = 1 \dots M$ , then this is :

$$\sum_{j=1}^{M} P(d_j) = \sum_{j=1}^{M} \sum_{c} P(c, a_{j1}, a_{j2})$$

Now, you will re-estimate the parameters using the last column in the table on the last page (the normalised  $P(C|A_1, A_2)$ )

First, compute the sum of all the rows in the table above where C = t, and divide by the sum of all rows to get

$$\theta_C = \frac{\text{sum of all weights where } C = t}{\text{sum of all weights}} = \frac{5.47}{4.53 + 5.47} = \boxed{0.547}$$

Now compute the sum of all the rows in the table above where C = t AND  $A_1 = t$ , to get

$$\theta_{11} = \frac{\text{sum of all weights where } C = t \text{ and } A_1 = t}{\text{sum of all weights where } C = t} = \frac{4.8}{5.47} = \boxed{0.88}$$

Now compute the sum of all the rows in the table above where C = f **AND**  $A_1 = t$ , to get

$$\theta_{10} = \frac{\text{sum of all weights where } C = f \text{ and } A_1 = t}{\text{sum of all weights where } C = f} = \frac{2.2}{4.53} = \boxed{0.49}$$

Now compute the sum of all the rows in the table above where C = t AND  $A_2 = t$ , to get

$$\theta_{21} = \frac{\text{sum of all weights where } C = t \text{ and } A_2 = t}{\text{sum of all weights where } C = t} = \frac{3.07}{5.47} = \boxed{0.56}$$

Now compute the sum of all the rows in the table above where C = f AND  $A_2 = t$ , to get

$$\theta_{20} = \frac{\text{sum of all weights where } C = f \text{ and } A_2 = t}{\text{sum of all weights where } C = f} = \frac{2.93}{4.53} = \boxed{0.65}$$

Finally, copy the values you just calculated back to the first boxes on the last (the model parameters), and start again. Continue doing this until the sum of all the rows stops changing.