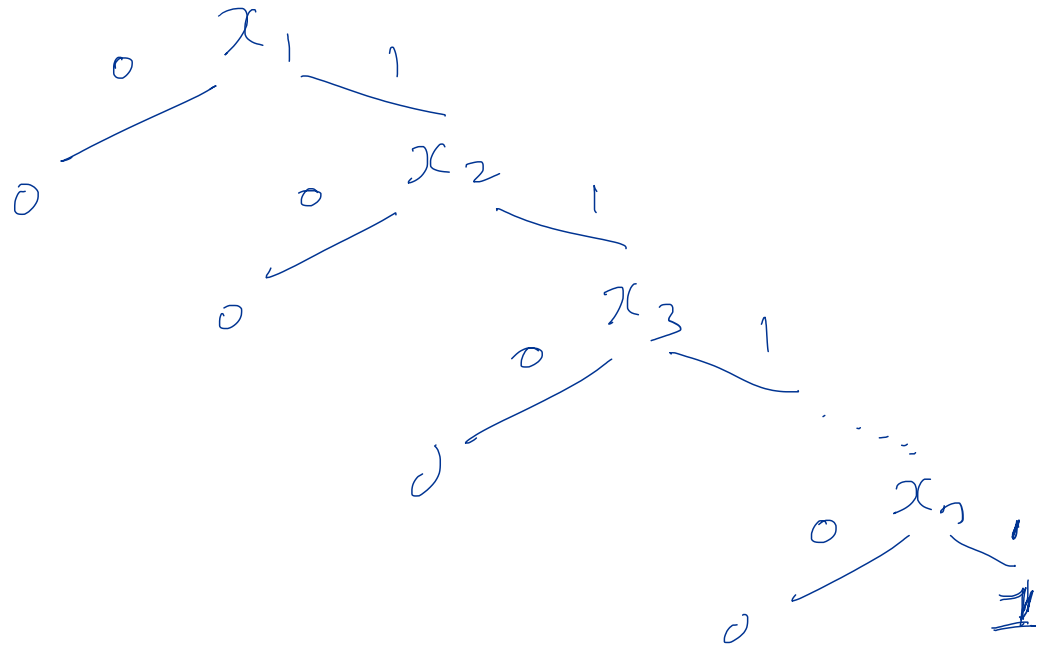


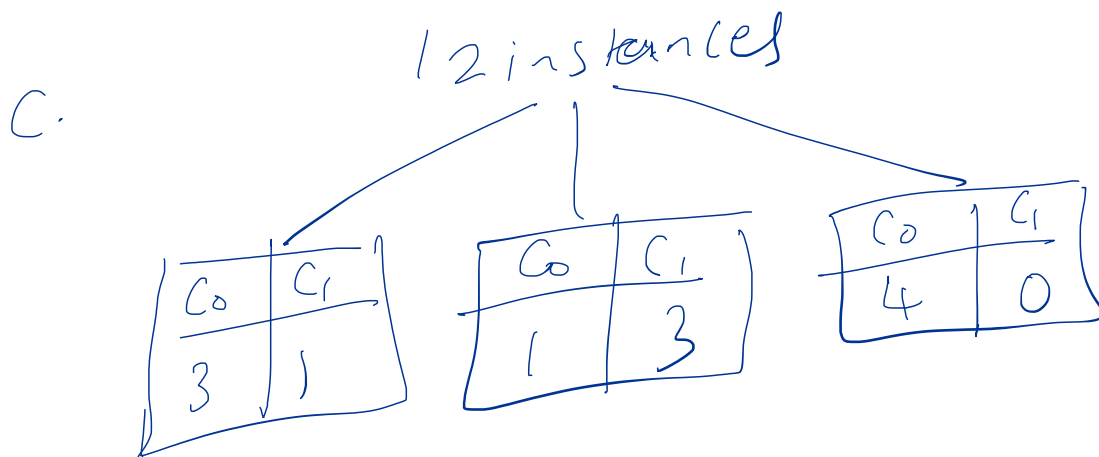
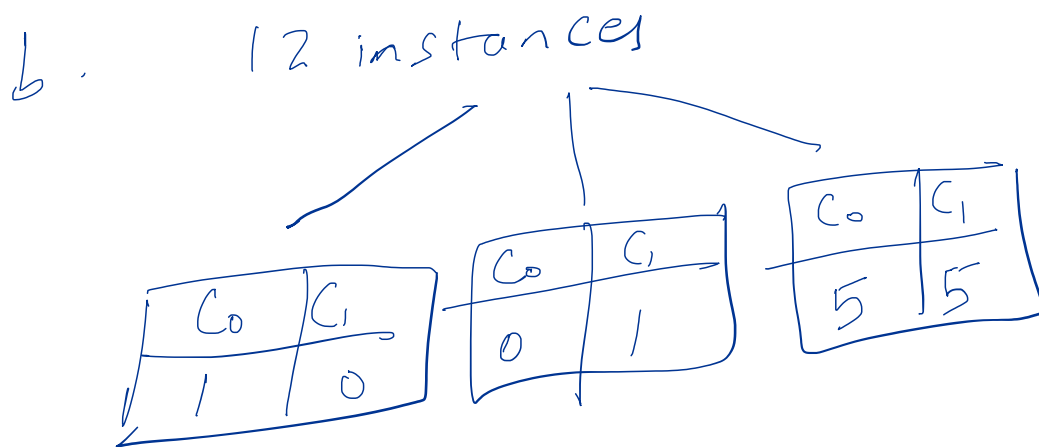
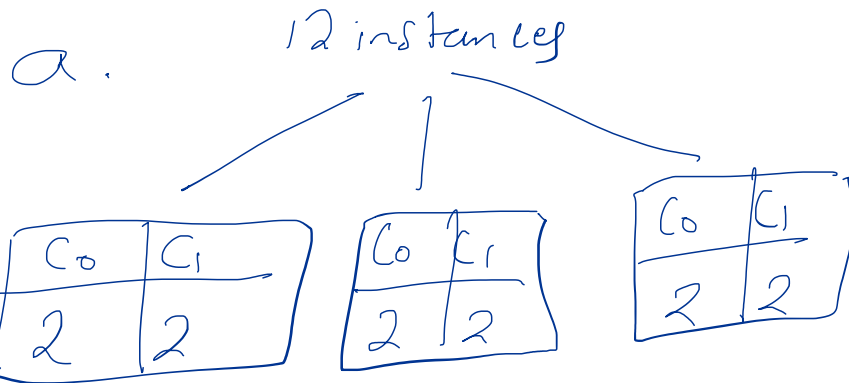
ML Sample Questions

(1) Draw a decision tree for

x_1, x_2, \dots, x_n



② Given 12 instances having Class labels either C_0 or C_1 , which of the following is a better split



C. is a better split since the entropy is nearly 0 after conditioning

② - I have the following dataset from a classification algorithm
Compute precision & recall
w.r.t the "+" class

True label	Output label
+	-
-	+
+	+
-	-
+	+

True positives
= 2

False positives
= 1

False negatives
= 1

Precision = $\frac{2}{3}$

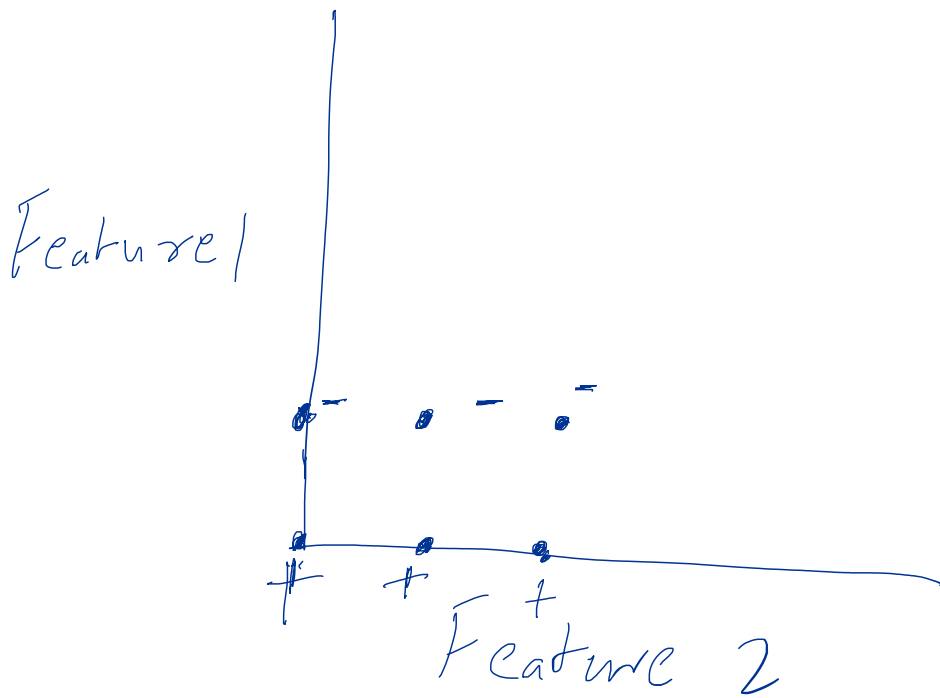
Recall = $\frac{2}{3}$

③ - Suppose a classification algorithm works as follows. If all inputs in my training data is "+" output "+" else output "-" for any training instance. What is its Leave-one-out Cross Validation error if a dataset has 50% +ve & 50% -ve examples

Soln The algorithm gets the output correctly if the label of the instance "left-out" is -ve, else it gets the output wrong. \therefore LOOCV error = 50%

4. Can a perceptron with 2 weights
Classify the following dataset
Correctly.

Feature-1	Feature-2	Class
0	0	+
0	1	+
0	2	+
1	0	-
1	1	-
1	2	-



We cannot find a line passing through the origin that separates "+" from "-".

∴ Perceptron with 2 weights will not get all the labels correctly.

5. Given the following dataset

Car-ahead	Ice	Speed	Brake
No	No	Low	Yes
Yes	Yes	high	Yes
No	Yes	high	No
No	Yes	Low	No
Yes	No	low	No

what will the Naive-Bayes
Algorithm predict for brake given
Car-ahead: Yes, Ice = Yes, Speed = Low

$$P(\text{Car-ahead: Yes} / \text{Brake: Yes}) \\ = 1/2$$

$$P(\text{Car-ahead: Yes} / \text{Brake: No}) \\ = 1/3$$

$$P(\text{Ice: Yes} / \text{Brake: Yes}) \\ = 1/2$$

$$P(\text{Ice: Yes} / \text{Brake: No}) \\ = 2/3$$

$$P(\text{Speed: low} / \text{Brake: Yes}) \\ = 1/2$$

$$P(\text{Speed: low} / \text{Brake: No}) \\ = 2/3$$

$$P(\text{brake} = \text{Yes}) = 2/5$$

$$P(\text{brake} = \text{No}) = 3/5$$

$$\frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{2}{5}$$

<

$$\frac{1}{3} \cdot \frac{2}{3} \cdot \frac{2}{3} \cdot \frac{3}{5}$$

\therefore We choose brake = No

⑦ Given the following training dataset-
what is the precision & recall
w.r.t the "Yes" Class in the
test dataset using the 1-nearest
neighbor Classifier

Feature-value	Class
1	Yes
2	No
4	Yes
5	No
7	Yes
8	No

Training data

Feature-Value	Class
1.1	Yes
2.1	Yes
5.1	No
7.1	Yes
8.1	No

Test dataset

Soln

Value	True-Label (in test data)	Predicted
1.)	Yes	Yes
2.)	Yes	No
5.)	No	No
7.)	Yes	Yes
8.)	No	No

True positives: 2
false positives: 0
False negatives: 1

Precision: 1.0
Recall: $\frac{2}{3}$

⑧. Do you think Naive Bayes is more like Bayesian Learning or like Max-likelihood learning.

Soln

In general, Naive Bayes is neither Bayesian nor Max-likelihood due to the Conditional independence assumption. But if we had to choose it is more like Bayesian learning since we use a prior. Interestingly, we estimate the prior using Max-likelihood learning (instead of assuming a form)

9 In logistic regression, which of the following is true

(a) we assume that the features are independent of each other

(b) we assume that the data instances are independent of each other

Soln (b) is true while (a) is false. (b) is

true because when we formulate the Conditional log-likelihood function we are implicitly assuming this.

⑩. \mathcal{G} VC-dimension a property
of i) training data or
ii) number of features or
iii) Learning algorithm.

Soln

\therefore It is definitely not
a function of training data.

It is inherently the property
of a learning algorithm and
possibly the number of features