

NAME:

ID:

**Reminders (do not ask what variables stand for):**

$$\mathcal{N}(x|\mu, \sigma^2) = \frac{1}{(2\pi\sigma^2)^{1/2}} \exp \left\{ -\frac{1}{2\sigma^2}(x - \mu)^2 \right\}$$

$$\mathcal{N}(\mathbf{x}|\boldsymbol{\mu}, \boldsymbol{\Sigma}) = \frac{1}{(2\pi)^{D/2}} \frac{1}{|\boldsymbol{\Sigma}|^{1/2}} \exp \left\{ -\frac{1}{2}(\mathbf{x} - \boldsymbol{\mu})^T \boldsymbol{\Sigma}^{-1}(\mathbf{x} - \boldsymbol{\mu}) \right\}$$

$$\text{Sigmoid}(x) = \frac{1}{1+e^{-x}}$$

$$\text{Threshold}(x) = \begin{cases} 1 & \text{if } x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

$$\text{RELU}(x) = \begin{cases} x & \text{if } x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

$$\text{Binary-cross-entropy loss: } -(y \cdot \log \hat{y} + (1-y) \cdot \log (1-\hat{y}))$$

*Entropy of X:*

$$H[X] = -\sum_{x_i} p(X = x_i) \cdot \log_2(p(X = x_i))$$

*Cross-Entropy between p and q:*

$$H(p, q) = -\sum_{x_i} p(X = x_i) \cdot \log_2(q(X = x_i))$$

$$\text{Cross-entropy loss: } -\mathbf{y} \cdot \log(\hat{\mathbf{y}})$$

NAME:

ID:

### 1) 12pt – Basic Math for ML

- a) **3pts** – You have trained a linear regression model  $\hat{f}(x) = a + bx$  that minimizes the squared distance between the target value  $y$  and predicted value  $\hat{f}(x)$ , for a sample  $x$ .

What is the **mean square error** of the estimates  $\hat{f}(x)$  over the given test set  $T = \{(x_i, y_i)\}, i=1..M$ .

Give as a formula, no partial for errors/missing details.

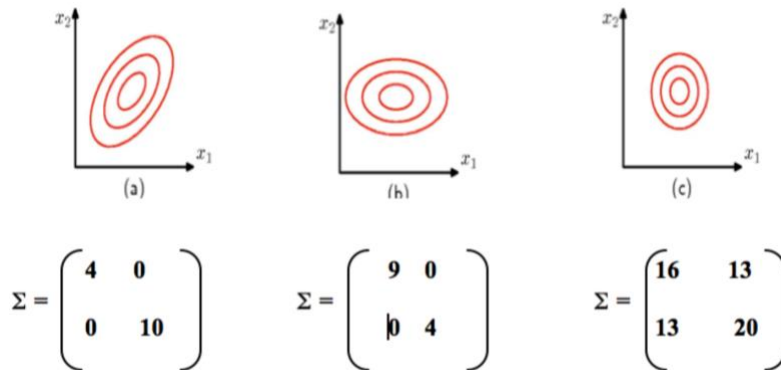
MSE = .....

- b) **3pts** – Apply gradient descent to **minimize** the function  $f(x, y) = x^3 + 2xy + 4$  starting at the point  $p_0 = (x, y) = [1 \ 2]^T$ . Use a learning rate of 0.1 and just go one step.

Gradient =

$p_1 =$

- c) **3pts** – Match given covariance matrices for the following 3 Normal distributions which are shown with their equal density contours. Use arrows to indicate the matchings.



- d) **3pt** – Assume you are given the following conditional probability distributions for left handedness with respect to gender.

$$P(\text{LeftHanded} | \text{Male}) = 0.3.$$

$$P(\text{LeftHanded} | \text{Female}) = 0.2$$

**Compute the probability of picking/encountering a LeftHanded person from the whole population. You must show your work.** Assume:

$$P(\text{Male}) = 0.4$$

$$P(\text{Women}) = 0.6$$

**Answer:** .....

NAME:

ID:

### 5) 10pts – Bayesian Approaches

a) 6pt - Consider a naive Bayes classifier trained on the dataset given below, where **all features are binary**.

Sample	Fever	Runny Nose	Throat Pain	Body Ache	Disease
1	High	No	Yes	Yes	Flu
2	High	No	No	Yes	Flu
3	Normal	No	Yes	Yes	Flu
4	Normal	No	Yes	No	Flu
5	High	No	Yes	No	Cold
6	Normal	Yes	Yes	No	Cold
7	Normal	Yes	No	Yes	Cold
8	Normal	Yes	Yes	No	Cold

A new patient comes with the following symptoms:

$x = [\text{Fever}=\text{Normal}, \text{RunnyNose}=\text{No}, \text{ThroatPain}=\text{No}, \text{BodyAche}=\text{Yes}]$

a) 3pts - Give the probabilities needed for estimating  $P(\text{Flu} | X)$ . I.e. write the required probabilities as  $P(A|B)$  etc. and not as numbers.

$$P(\text{Flu} | x) \propto P(\dots) \times P(\dots) \times P(\dots) \times P(\dots)$$

Answer expected:  $P(\text{Flu} | x) \propto P(\text{Fever}=\text{Normal}|\text{Flu}) \times P(\text{RN}=\text{No}|\text{Flu}) \times P(\text{TP}=\text{No}|\text{Flu}) \times P(\text{BA}=\text{Yes}|\text{Flu}) \times P(\text{Flu})$

b) 3pts - Now plugin the numbers calculated from the table (do not use smoothing) as fractions, but leave out the multiplication

$$P(\text{Flu} | x) \propto \dots$$

Answer expected:  $P(\text{Flu} | x) \propto 2/4 \times 4/4 \times \dots$

b) 2pts - Lets assume that you have found that  $P(\text{Flu} | x) = 0.2/P(x)$  and  $P(\text{Cold} | x) = 0.3/P(x)$  in the above problem.

What would be the actual normalized value of  $P(\text{Flu} | x)$ ?

Answer: .....

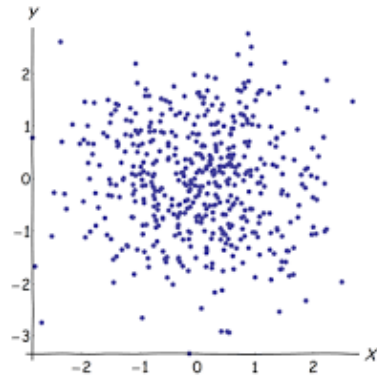
c) 2pt –Select all correct answers (each wrong cancel one correct):

- ☐  $P(\text{Cold} | x) + P(\text{Flu} | x) = 1$
- ☐  $P(x | \text{Cold}) + P(x | \text{Flu}) = 1$
- ☐ Naive Bayes uses a simplifying assumption that attributes are conditionally independent given the class label
- ☐ Naive Bayes uses a simplifying assumption that classes are conditionally independent given the attributes

NAME:

ID:

**6) 8pts – Multivariate Normal Distribution**



**a) 4pts –** Assume you are given the above two-dimensional data, in the form of a scatter plot.

- **1pt** – What can you say about its value (positive?, negative?, 0?, 1?, -1?, ...)? *Be as precise as possible.*
- **1pts** – What is the range of correlation, **in general**? I.e. what numbers can it take? **Answer:** .....
- **1pts** – What is the range of variance, **in general**? I.e. what numbers can it take? **Answer:** .....

...

NAME:

ID:

## 8) 8pts –Classifier Combination

a) 5pts - Circle as True or False. 1pts each correct; -1pts for each incorrect.

- **T / F** - In bagging, **the training set for each base model** may not contain some of the samples in the original training set and may contain a single sample from the original training multiple times.
- **T / F** - Bagging is **not very suitable** for **stable** learning algorithms such as k-NN.
- **T / F** - Ensemble **averaging** is **guaranteed** to have higher accuracy than each of the base classifiers.
- **T / F** - In **random forest**, we select to use a subset of the features for **each tree**.
- **T / F** - **The main advantage** of combining classifiers is reducing **variance**.

b) 3pts – Considering a **majority vote** ensemble on a two class problem, assume that each of the **5 base classifiers** has a probability of error  **$p=0.2$**  on a given input and that their errors are independent.

What can you say about the probability of error for the ensemble?

- Probability that the ensemble makes error < Probability of exactly 3 base classifiers make errors
- Probability that the ensemble makes error > Probability of exactly 3 base classifiers make errors
- Probability that the ensemble makes error = Probability of exactly 3 base classifiers make errors
- Cannot say

*Hint: You can answer this question without using any formulas, with just reasoning*

NAME:

ID:

**10) 15pts – General ML**

**a) 5pts** – A bank loses about 1 million TL a year on fraud cases and wants to detect credit card fraud (fraud or not) using machine learning techniques. Credit card fraud occurs in about 5% of all the bank's credit card transactions.

- i. 2pts - What is the **majority classifier's accuracy** in this classification problem?

**Answer:** .....

- ii. 3pts - What is the **random classifier's accuracy** in this classification problem?

**Answer:** .....