

## Introduction

Today<sup>1</sup>:

- Classification
- Cumulative density function
- Transformation between variables.
- Hough and Radon transform.

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## Classification I

Consider the sample space set of all the students of IET

$$\Omega_{IET} = \{\omega_i\}_{i=1, \dots, 13}$$

where  $\omega_i$  is the student  $i$ . A camera is filming one student of IET, and from the audio signal  $\mathbf{x}_a$  and visual signal  $\mathbf{x}_v$  (we note  $\mathbf{x} = (\mathbf{x}_a, \mathbf{x}_v)$ ) we observe, we want to infer the identity of this student.

### Definition (Maximum a Posteriori Classifier)

Considering a observed random variable  $\mathbf{x}$

$$\hat{\omega}_{MAP} = \arg \max_{\omega \in \Omega} P(\omega | \mathbf{x}) = \arg \max_{\omega \in \Omega} \frac{P(\mathbf{x} | \omega) P(\omega)}{P(\mathbf{x})}$$

## Classification II

Applied to our problem of IET students recognition, we need to estimate  $P(\omega_i | \mathbf{x})$  for all students  $i$  and select the one that has the maximum posterior probability.

Questions:

- 1 What would you choose to model the marginals  $\{P(\omega_i)\}$ ?
- 2 What would you do to estimate  $P(\mathbf{x} | \omega_i)$ ?

## Classification III

One easy method proposed for classification is the **K nearest neighbour**.

- 1 Draw in the feature space the training samples corresponding to each student.
- 2 When you have a new observation  $\mathbf{x}$  find its  $K$  nearest neighbours.
- 3 Among those  $K$  nearest neighbours, find the class that is the most represented.

## Transformation in between two random variables $Y = g(X)$

### Definition (Cumulative density function)

The **cumulative density function** of a random variable  $X$  with p.d.f.  $f_X(x)$  is defined as

$$F_X(x) = P(X \leq x) = \int_{-\infty}^x f_X(t) dt$$

### Theorem (Transformations in between two random variables)

Consider two random variables  $X$  and  $Y$  such that  $Y = g(X)$  with  $g$  a monotonic and differentiable function ( $g'(x) > 0, \forall x$ ). If  $F_X(x)$  and  $F_Y(y)$  are the respective c.d.f.s. then

$$F_X(g^{-1}(y)) = F_Y(y)$$

## Exercises

- 1 Prove the previous theorem.
- 2 Consider  $X \sim \mathcal{N}(0, 1)$  and  $Y = aX + b$  with  $a > 0$ .
  - 1 What is the probability density function  $f_Y(y)$ ?
  - 2 Compute mean and variance of  $Y$ .Consider the random variable  $Z = X + Y$ . Compute  $f_Z(z)$  w.r.t.  $f_X(x)$  and  $f_Y(y)$ .