- 1. Estimation of density function P(x lwi) using samples patterns
- 2. Estimation of a pastericis probability P (wi/x) directly based on sample patters.

non-parametric methods above.

- 1. parzen window based
- 2. K- Nearest neighbour based,

Density Estimation,

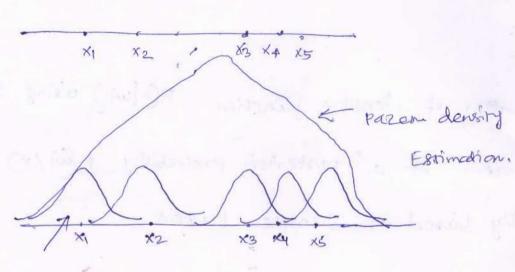
Estimately the PDF P(x7 based an a given set of training Samples D= {24, , -xn}

Histograms are the Simplest way to density
Estimation.

_ no. of training samples

 $p(x) = \frac{n_i}{v_n} = \frac{n_i - i}{v_n}$

This method is not good for many features



frencia for X;

\ \

PNN - Implement the classification cloing taszen windows

KNN - for density estimation.

An estimate for the joint probability is

Por (X, Wi) = Ki & Raptures is scenples

To (X, Wi) = NV. K: of which turns over

Notwee around X

Then I we can estimate P(\vec{ui}) x) by

Por \(\vec{wi}/x\) = \frac{\Prac(x;\vec{wi})}{\Sigma} = \frac{\ki}{\ki}

Linear Discriminant function:

- I we have also a third class of one methods for designing classifiers. They are based on the discontinent bunctions.
- P Discriminant functions (DFs) are a useful tool for representing classifies in a simpler way that could be done with the posterior probabilities.
- * The decision rule given DFs $g_1, ..., g_c$ was classify x to the class w_i If $g_i(x) > g_j(x)$ for all $i \neq j$.
- O with the Baye's classifier, our classifier design prescodure was the following:
 - Assume that class conditional probability (cpdf)

 have a specific parametric form and prior probabilities

 are all known
 - ii) Then using one braining data estimate the optimum Parameter vector for each coalt based on the MLE | Bayesian techniques.

- (2) The trierd class of designing the classifier is based on the Non-parametric from. [methods.
 - nearest neighbour rule?
 - Parzon window

Estimated the posterics

Probabilities directly without

any assumptions on the form of

ccpdfs-

* LDF:

Linear Discriminant functions, is one way of non-Pava metric form.

- i) Instead of estimating parameters for copples, we can parametrisize the DFs and estimate parameters for them [like, weight vectors and bias].
- ii) The estimation will be formulated as a problem of minimizing a criterian functions
- iii) the criterian function can be training errorbue and classification error in classifying the training Samples.