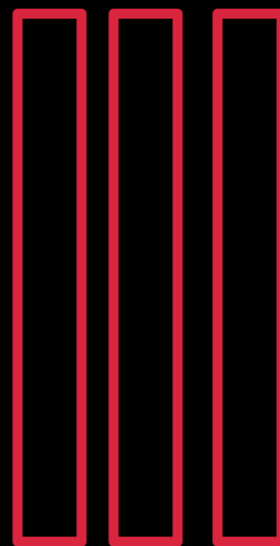
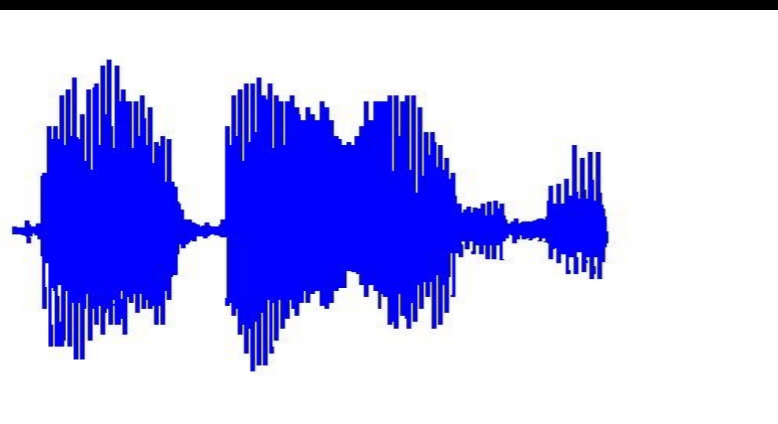


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# Recap ...



MFCC

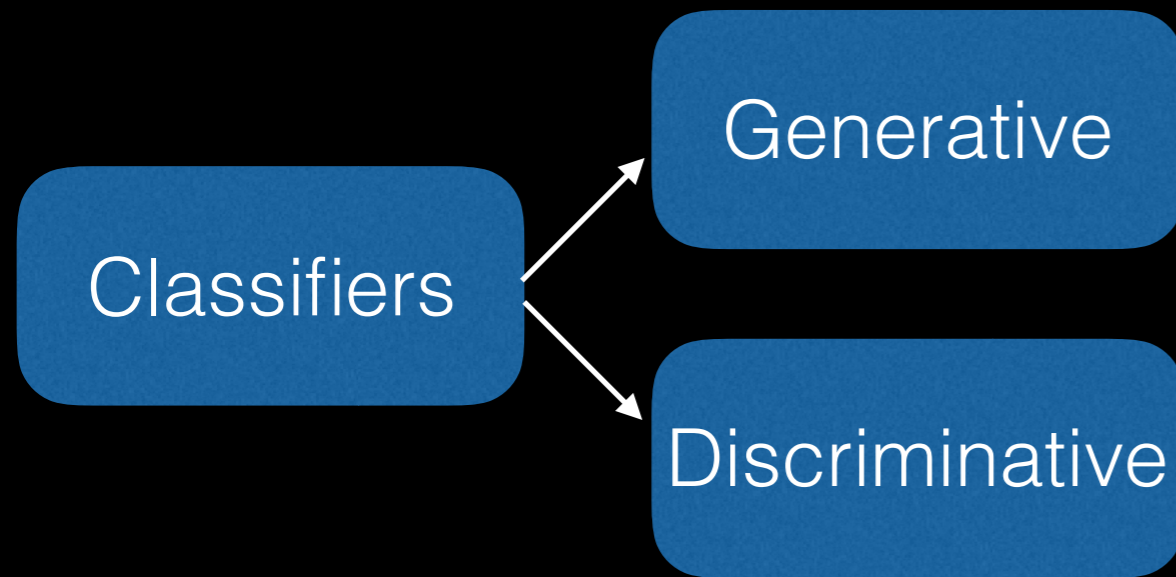


$\mathbf{x}_i, i = 1, \dots, T$

Speech is converted to a sequence of vectors - {Linear prediction parameters, MGCs, MFCC, Sub-band energies etc}

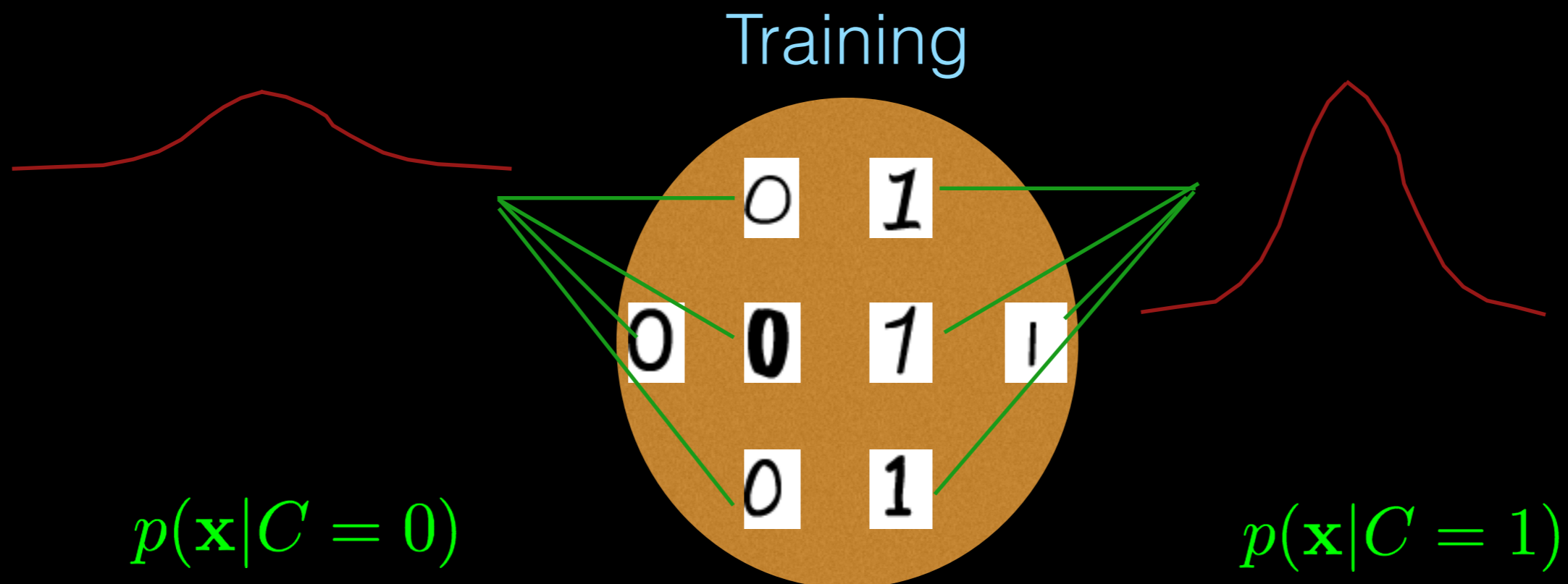
**Recognition** - Determine whether two recordings have the same content or not.

# Pattern Recognition



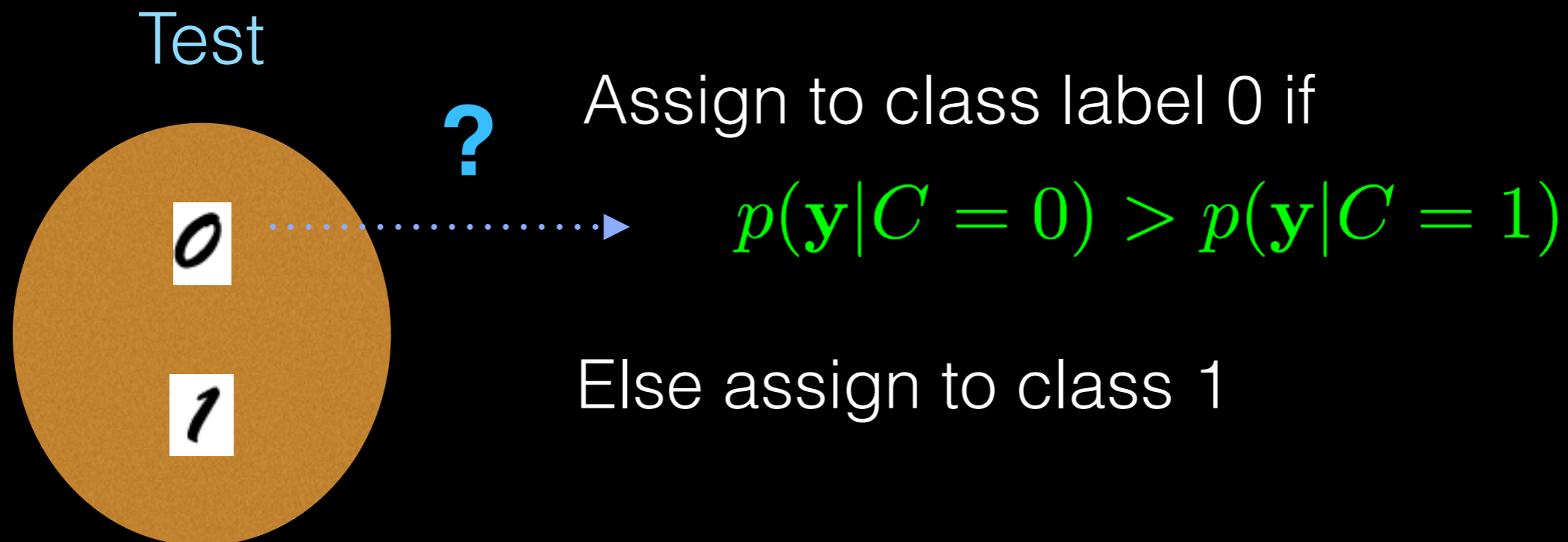
# Generative classifier

- Model the two classes separately using probability distributions -
  - Make each sample  $\mathbf{X}_i$  (28x28) as a vector  $\mathbf{x}_i$  of size 784.
  - Build class dependent probability  $p(\mathbf{x}|C = 0)$  &  $p(\mathbf{x}|C = 1)$



# Generative classifier

- For the test sample
  - Make each sample  $\mathbf{Y}$  (28x28) as a vector  $\mathbf{y}$  of size 784.
  - Compute the probability of generating sample  $\mathbf{y}$  for each class.

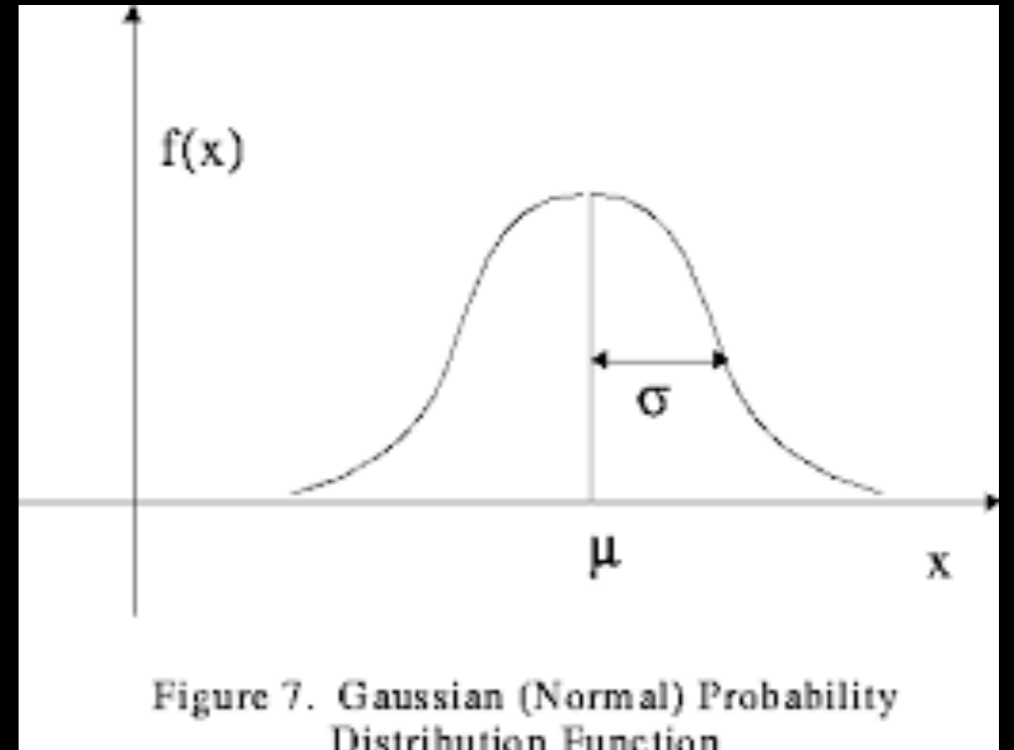


# Gaussian Distribution

$$p(\mathbf{x}|\theta) = \frac{1}{\sqrt{(2\pi)^D |\Sigma|}} \exp\left\{ -\frac{1}{2}(\mathbf{x} - \mu)^* \Sigma^{-1}(\mathbf{x} - \mu) \right\}$$

- Parameters are mean and covariance matrix
- Determine the parameters with Maximum Likelihood Estimation (MLE) process.

$$L(\theta|\mathbf{x}) = p(\mathbf{x}|\theta)$$



(c) Google Images

# Matrix Vector Differentiation Rules

①

## Matrix vector differentiation rules

$$\frac{\partial x^T A x}{\partial x} = (A + A^T) x$$

If  $A$  is symmetric.

$$\frac{\partial |A|}{\partial a_{ij}} = \begin{cases} 2A_{ij} & \text{if } i = j \\ A_{ij} & \text{if } i \neq j \end{cases}$$

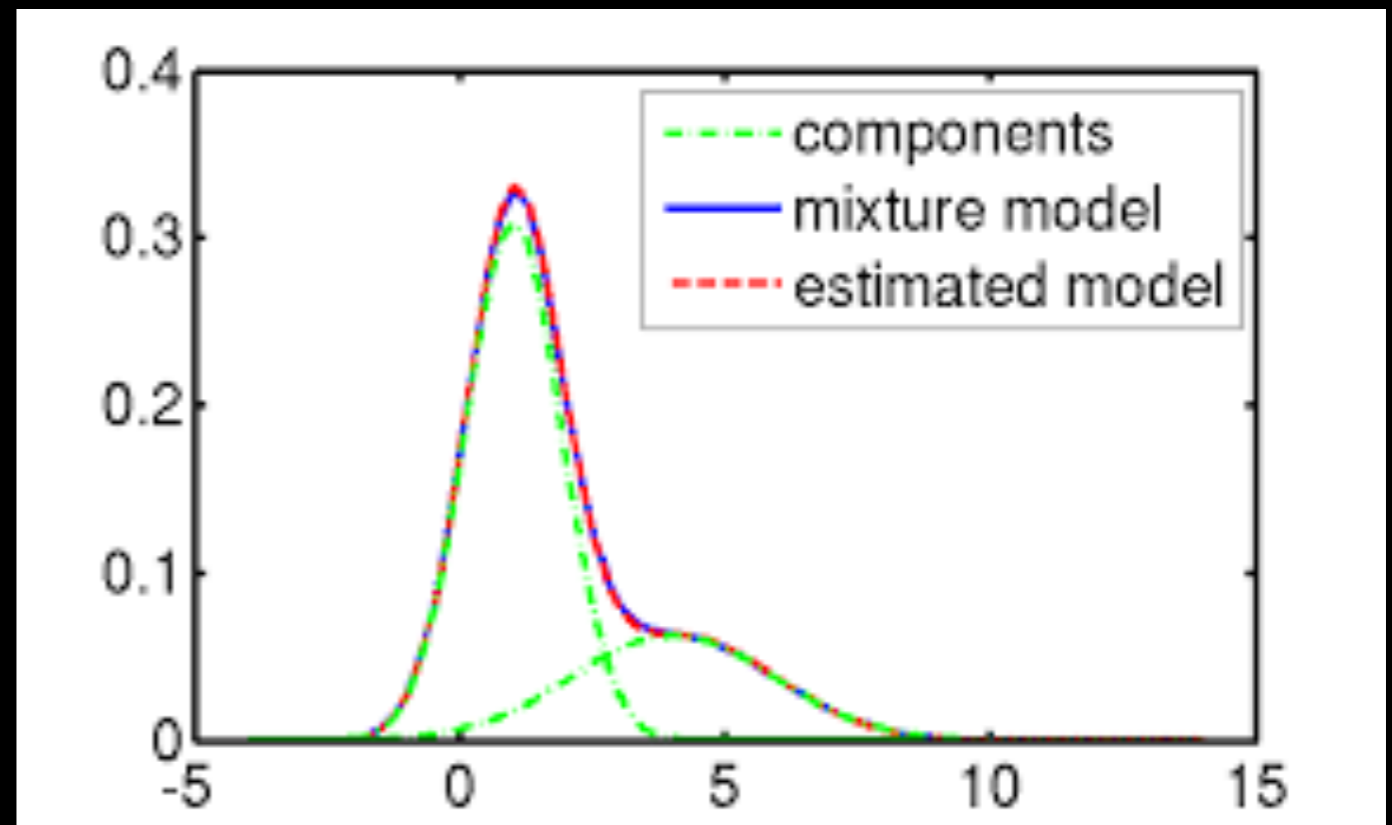
for symmetric  $A$   
 $A_{ij}$  = cofactor of  $A$

$$\frac{\partial \log |A|}{\partial A} = \begin{cases} A_{ij} / |A| & \text{if } i = j \\ 2A_{ij} / |A| & \text{if } i \neq j \end{cases} = 2A^{-1} - \text{diag}(A^{-1})$$

# GMM

$$p(\mathbf{x}|\theta) = \sum_{i=1}^M \alpha_i p_i(\mathbf{x}|\theta_i)$$

- Parameters estimated using EM algorithm.



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