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Document Classification Problem
       We have 2 classes of documents.
(Example: class 1: spam, class 2: important, class 3: All personal k=4
                                                                                                                                                                            has 5000 documents.

11 6000 11

11 10000 11
       Dataset:
                                                                                                                 0 99 1
 hos
                                                                                                           C/9552
  25 K
documents
                                                                                                               1/ 3
                                                                                                                            1 4 11
                                                                                                                                                                           \pi = \frac{5000}{20000} = \frac{5}{29}, \quad \pi_4 = \frac{10}{25}
       Prior probabilities:
                                                                                                                                            \pi_2 = \frac{6}{28} / \pi_3 = \frac{4}{28}
              V = { word 1, word, ..., word | v | 3 has | v | elements.
             P^{(1)} = \begin{pmatrix} P_{11} / \cdots & P_{1N1} \end{pmatrix} \text{ is the model for class } 1.
P^{(2)} = \begin{pmatrix} P_{021} / \cdots & P_{2N1} \end{pmatrix} \text{ is the model for class } 2.
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\frac{P(3)}{P(4)} = \frac{(P_{31}) \cdot \cdots \cdot (P_{31}) \cdot 1}{(P_{41}) \cdot \cdots \cdot (P_{41}) \cdot 1} = 1

\frac{P(4)}{P(4)} = \frac{P(4)}{(P_{41}) \cdot \cdots \cdot (P_{41}) \cdot 1} = 1

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We have 4 models: Model 3 // 3 (//) = P31 - - · P3NI / 2/NI Training process is over

Testing (Classification or Recognition) phase Given a document (word) feature x= [3]
word, word2
word4 word4
word5, word4 Feature vector is obtained using the "bagof words" model. for i=1;4 $d_i = \pi_i P_{i1}^3 P_{i2}^{2} P_{i3}^1 P_{i5}^1 P_{i6}^1$ Answer = pick the i that gives the max probability (likelihood). Class Answer = arg max (Xi)

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