

CSIS3764 DATA SCIENCE

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Machine Learning Naive Bayes

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BAYES THEOREM

- $P(A|B) = \frac{P(B|A)P(A)}{P(B)}$
- A Naive Bayes classifier is a probabilistic machine learning model that's used for classification task
 - Sentiment analysis
 - Spam filtering
 - Recommendation systems



SPAM FILTERING

- 12 Emails
 - 8 Normal
 - 4 Spam
- Words that occur in emails
 - Hello
 - Sally
 - Food
 - Lotto



FREQUENCY OF WORDS

- Normal emails
 - Hello: 8 -> $P(\text{Hello}|\text{Normal}) = 0.47$
 - Sally: 5 -> $P(\text{Sally}|\text{Normal}) = 0.29$
 - Food: 3 -> $P(\text{Food}|\text{Normal}) = 0.18$
 - Price: 1 -> $P(\text{Price}|\text{Normal}) = 0.06$
- Spam emails
 - Hello: 2 -> $P(\text{Hello}|\text{Spam}) = 0.29$
 - Sally: 1 -> $P(\text{Sally}|\text{Spam}) = 0.14$
 - Food: 0 -> $P(\text{Food}|\text{Spam}) = 0$
 - Price: 4 -> $P(\text{Price}|\text{Spam}) = 0.57$



Example: “Hello Sally”

- Probability any message is normal
 - $P(\text{Normal}) = 8/12 = 0.67$
- Probability “Hello” and “Sally” is normal
 - $P(\text{Normal}) \times P(\text{Hello}|\text{Normal}) \times P(\text{Sally}|\text{Normal})$
 $0.67 \times 0.47 \times 0.29 = 0.09$
- Probability any message is spam
 - $P(\text{Spam}) = 4/12 = 0.33$
- Probability “Hello” and “Sally” is spam
 - $P(\text{Spam}) \times P(\text{Hello}|\text{Spam}) \times P(\text{Sally}|\text{Spam})$
 $0.33 \times 0.29 \times 0.14 = 0.01$
- **Message is Normal**



Example: “Food Price Price Price”

- Probability any message is normal
 - $P(\text{Normal}) = 8/12 = 0.67$
- Probability “Food” and “Price⁴” is normal
 - $P(\text{Normal}) \times P(\text{Food}|\text{Normal}) \times P(\text{Price}|\text{Normal})^4$
 $0.67 \times 0.18 \times 0.06^4 = 0.000002$
- Probability any message is spam
 - $P(\text{Spam}) = 4/12 = 0.33$
- Probability “Food” and “Price⁴” is spam
 - $P(\text{Spam}) \times P(\text{Food}|\text{Spam}) \times P(\text{Price}|\text{Spam})^4$
 $0.33 \times 0 \times 0.57 = 0$
- ?



FREQUENCY ADJUSTMENT OF WORDS

- Normal emails
 - Hello: 9 -> $P(\text{Hello}|\text{Normal}) = 0.43$
 - Sally: 6 -> $P(\text{Sally}|\text{Normal}) = 0.29$
 - Food: 4 -> $P(\text{Food}|\text{Normal}) = 0.19$
 - Price: 2 -> $P(\text{Price}|\text{Normal}) = 0.10$
- Spam emails
 - Hello: 3 -> $P(\text{Hello}|\text{Spam}) = 0.27$
 - Sally: 2 -> $P(\text{Sally}|\text{Spam}) = 0.18$
 - Food: 1 -> $P(\text{Food}|\text{Spam}) = 0.09$
 - Price: 5 -> $P(\text{Price}|\text{Spam}) = 0.45$



Example: “Food Price Price Price”

- Probability any message is normal
 - $P(\text{Normal}) = 8/12 = 0.67$
- Probability “Food” and “Price⁴” is normal
 - $P(\text{Normal}) \times P(\text{Food}|\text{Normal}) \times P(\text{Price}|\text{Normal})^4$
 $0.67 \times 0.19 \times 0.10^4 = 0.00001$
- Probability any message is spam
 - $P(\text{Spam}) = 4/12 = 0.33$
- Probability “Food” and “Price⁴” is spam
 - $P(\text{Spam}) \times P(\text{Food}|\text{Spam}) \times P(\text{Price}|\text{Spam})^4$
 $0.33 \times 0.09 \times 0.45 = 0.00122$
- **Message is Spam**