

Unit -4 Probabilistic Graphical Model,

1) NAIVE BAYES ALGORITHM:

The Naive Bayes algorithm is a classification technique based on Bayes Theorem. It assumes that all features (attributes) are conditionally independent given the class label.

$$P(c/x) = \frac{P(x/c) P(c)}{P(x)}$$

Where

$c \rightarrow$ class

x - feature ~~class~~ vector

Since $P(x)$ is constant the decision is based on maximizing $P(x/c) P(c)$

Application:

* Spam Detection

* Medical Diagnosis

* Sentiment analysis

2) BAYESIAN BELIEF NETWORK: (BBN)

A Bayesian Belief Network (BBN) also

Known as a Bayesian Network, is a graphical model that represents probabilistic relationships among a set of random variables.

It is represented as a directed Acyclic Graph (DAG).

- * Nodes represent random variables.
- * Edges represents conditional dependencies.
- * Each node has a conditional probability

Table (CPT) that quantifies the effect of parent class

Example :

Consider medical diagnosis network :

Node 1 : Flu

Node 2 : Fever

Node 3 : Cough

Here Fever and Cough depends on Flu.

The network compactly encodes these dependencies.

Advantages :

- * Captures causal relationship.
- * Handles incomplete data.
- * Supports both inference (predict unknowns) and Learning (update probabilities).

Application :

- * Medical diagnosis (diseases & symptoms).
- * Fault detection in engineering system.
- * Decision support system.

3) Hidden Markov Model (HMM).

A Hidden Markov Model (HMM) is a statistical model for systems that evolve over time but where underlying states are hidden.

Instead, we observe outcomes generated, probabilities from the hidden states.

- Components
- a) Hidden states (S)
 - b) Observation (O)
 - c) Transition Probabilities (A)
 - d) Emission probabilities (B)

e) Initial state distribution (π)

Key problems solved by HMM:

* Evaluation: Compute probability of observation sequence given models.

* Decoding: Determine most likely sequence of hidden states (Viterbi algorithm).

* Learning: Estimate parameters.

Application:

* Speech recognition

* Natural Language Processing

* Bio Informatics.

4. BAYESIAN INFERENCE:

Bayesian Inference is a method of statistical inference where probability is used to represent uncertainty about parameters. Unlike frequentist methods, Bayesian methods update beliefs based on new evidence.

$$P(H|D) = \frac{P(D|H) \cdot P(H)}{P(D)}$$

Where

$H \rightarrow$ hypothesis

$D \rightarrow$ Observed data

Advantages :

- * Incorporates prior knowledge
- * Naturally handles uncertainty.
- * Produces full probability distributions.

PROBLEM BASED ON NAIVE BAYES

Problem:

A spam filter uses the Naive Bayes algorithm

Consider the word "offer" appearing in emails.

$$P(\text{spam}) = 0.4, \quad P(\text{Not spam}) = 0.6, \quad P(\text{offer} | \text{spam}) = 0.8$$

$$P(\text{offer} | \text{not spam}) = 0.2$$

If an email contains the word "offer" classify it as spam or not spam.

Solution

$$P(\text{spam} | \text{offer}) = \frac{P(\text{offer} | \text{spam}) P(\text{spam})}{P(\text{offer})}$$

$$P(\text{offer}) = P(\text{offer} | \text{spam}) + P(\text{offer} | \text{Not spam}) P(\text{Not spam}).$$

$$= (0.8)(0.4) + (0.2)(0.6)$$

$$= 0.32 + 0.12$$

$$= 0.44$$

Now

$$P(\text{spam} | \text{offer}) = \frac{0.8 \times 0.4}{0.44} = 0.727$$

$$P(\text{Not spam} | \text{offer}) = \frac{0.2 \times 0.6}{0.44} = 0.273$$

Here the email is classified as "spam".