

# Artificial Intelligence Laboratory 3:

## Bayesian Network

### Things I experience in this lab

- ✓ • Learn to construct Bayesian Network from data.
- ✓ • Learn to implement Naïve Bayes algorithm for classification tasks.
- ✓ • Learn to use Bayesian Network and Naïve Bayes for inference.

### Task 1

**Read the code in Lab3 task1.ipynb. Learn a Bayesian Network of the smart grid data set. Construct the network and compute the following:**

**Steps to solve this Task:**

- ✓ Create environment for lab 3 project
- ✓ Install *bnlearn* library
- ✓ Import the smart grid file in notebook using pandas
- ✓ Data Encoding/Preprocessing
- ✓ splitting Features and labels
- ✓ From the *bnlearn. structure\_learning( )* library, we **fit** the preprocessed data using
- ✓ Plot the learned structure of Directed Acyclic Graph (DAG)
- ✓ From the *bnlearn. parameter\_learning( )* again model is fit which use the previous model of structure learning
- ✓ From the updated model Conditional Probability Distribution table is plotted
- ✓ performing inference is done using *bnlearn.inference.fit( )*
- ✓ Subtask of task 1 is done using *inference.fit()*, which takes three arguments
  - Updated model
  - Variable
  - evidence

## Task 2

**Implement Naïve Bayes algorithm in Lab3 task2.ipynb, performing classification by making inference:**

**Steps to solve this Task:**

- ✓ Create environment for lab 3 project
- ✓ Install tabulate sklearn
- ✓ Import libraries
- ✓ Convert the list data into pandas DataFrame
- ✓ Data Encoding/Preprocessing
- ✓ splitting Features and labels
- ✓ Model fitting
- ✓ Calculating Accuracy
- ✓ Prediction
- ✓ Part **2d** is done using `model.predict()` method after encoding given data

By using Following Formulas calculation of marginal probability, posterior probability and Likelihood table is created of task **2a,2b,2c**

- ✓ **Bayes Formula**  $P(A | B) = P(B | A) \cdot P(A) / P(B)$
- ✓ **Conditional Probability:**  $P(A | B) = P(A \cap B) / P(B)$
- ✓ **Mean:**  $\mu = \frac{1}{n} \sum_{i=1}^n x_i$
- ✓ **Standard Deviation**  $\sigma = \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - \mu)^2}$
- ✓ **Normal Distribution:**

$$f(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$\mu$  = mean of  $x$

$\sigma$  = standard deviation of  $x$

$\pi \approx 3.14159 \dots$

$e \approx 2.71828 \dots$

- ✓ Task 2e is done by Implementing a Naive Bayes Classifier and performing classification on the Iris dataset
  - Load iris data
  - Split training and testing features and labels
  - Create model of Naive Bayes Classifier using `(from sklearn.naive_bayes import GaussianNB)`
  - Fit the training data in the model
  - Predict the testing data
  - Calculate Performance

## After Doing these Task I observe and learned

- ✓ Learn how to construct Bayesian Network from data.
- ✓ Learn how to inference of the Bayesian
- ✓ Learn to estimate a DAG that captures the dependencies between the variables.
- ✓ Observe the dependencies between the variables using DAG
- ✓ Learn How to do structural learning and parameterised learning practically through bnlearn library
- ✓ Learn to estimate the (conditional) probability distributions of the individual variables.
- ✓ Learn to implement Naïve Bayes using sklearn
- ✓ encoding of data
- ✓ Probability calculation through algorithm also from the formula