

## Task Description:

As part of your internship with us, you will have the opportunity to contribute to our organization's knowledge base and provide valuable insights to our audience by writing articles on topics related to Machine Learning. Credit for this article will be given to you. You can showcase this in your resume or interview which showcases your knowledge and deep understanding about this topic. These articles serve as a platform to share knowledge, showcase your understanding of Machine Learning related concepts, and communicate your findings effectively.

## Task Details:

- **Article Length:** Each article should be between 800 and 1,200 words.
- **Originality:** All articles must be original and well-researched. Article content should be Plagiarism Free and should not be AI Generated.
- **Topic Selection:** You have the creative freedom to choose any data analysis or Machine Learning related topic that interests you. You can refer to the list of reference topics below, or you're welcome to propose your own topic.
- **Research:** Conduct thorough research to ensure that your articles are well-informed and up-to date. Cite credible sources when necessary.
- **Quality:** Maintain high-quality writing standards with proper grammar, spelling, and structure. Your articles should be clear, concise, and engaging.
- **Visualization:** Include relevant data visualizations, charts, or graphs to illustrate your points effectively. Ensure that you have the necessary rights to use any visuals.
- **Engagement:** Craft articles that engage our readers and offer value. Use real-world examples, case studies, or practical insights where applicable.

## How to Submit:

Please submit your completed articles via Google form. Link is mentioned in Internship Details file. File name should be in below format:

Batch Name - Your Full Name – Article Name

Eg: If name is Aashish Kharwade and Batch is MIP-ML-04 and Prepared an article on topic 'ML'.  
File name will be:

**Ashish\_Kharwade\_MIP-ML-04\_ML**

Below is the list of suggested topics for reference. You are not bound to select one of these topics. You are free to choose topic of your interest.

1. Decision Trees and Random Forests: Explore the basics of decision trees and the ensemble method, random forests, for improved predictive performance.
2. Gradient Boosting Algorithms: Discuss algorithms like XGBoost, LightGBM, and CatBoost, and their applications in boosting model accuracy.
3. Support Vector Machines (SVM): Understand the fundamentals of SVM and its applications in classification and regression tasks.
4. Principal Component Analysis (PCA): Explore dimensionality reduction techniques and how PCA is used to capture essential features in data.
5. K-Means Clustering: Discuss the basics of unsupervised learning and how K-means clustering can be applied to group data points.
6. Naive Bayes Classifier: Explore probabilistic classifiers and the Naive Bayes algorithm for text classification and spam filtering.
7. K-Nearest Neighbors (KNN): Understand the concept of distance-based classification and regression using the KNN algorithm.
8. Logistic Regression: Dive into the basics of logistic regression for binary and multiclass classification tasks.
9. Neural Network Basics: Explore the foundational concepts of neural networks, including activation functions, layers, and backpropagation.
10. Recurrent Neural Networks (RNN): Understand how RNNs are used for sequence data and time-series prediction.
11. Convolutional Neural Networks (CNN): Discuss the architecture and applications of CNNs for image classification and object detection.
12. Hyperparameter Tuning: Explore techniques for optimizing model hyperparameters to improve performance.
13. Feature Scaling and Normalization: Understand the importance of preparing data through scaling and normalization for better model training.

14. Cross-Validation Techniques: Discuss methodologies like k-fold cross-validation to assess model performance and reduce overfitting.

15. Ensemble Learning Basics: Introduce the concept of ensemble learning and its benefits in combining multiple models for improved predictions.

16. Regularization Techniques: Explore regularization methods such as L1 and L2 regularization to prevent overfitting in machine learning models.

17. Time Complexity in Algorithms: Discuss the concept of time complexity and its significance in understanding algorithm efficiency.

18. Bayesian Networks: Explore probabilistic graphical models and how Bayesian networks represent dependencies between variables.

19. Association Rule Mining: Discuss algorithms like Apriori and FP-growth for discovering patterns in transactional data.

20. Genetic Algorithms: Introduce the concept of genetic algorithms for optimization problems and parameter tuning.