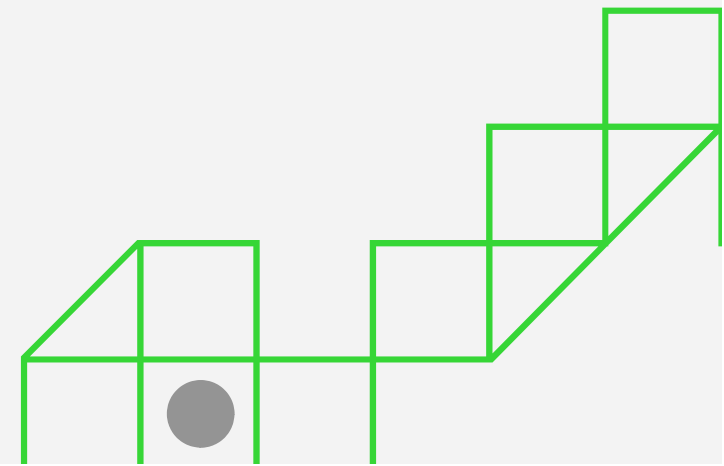
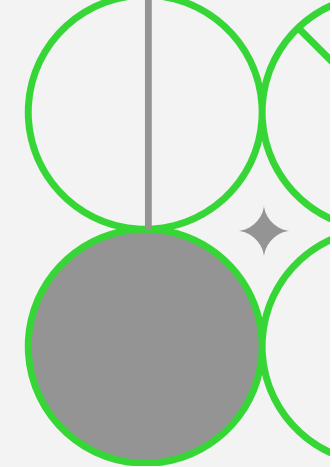




Efficient Iris Classification Using K-Nearest Neighbors Algorithm

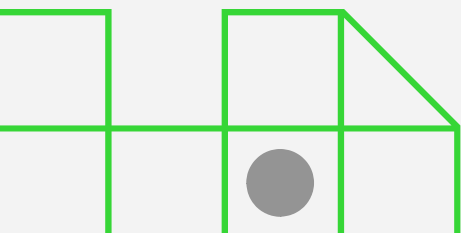




Introduction

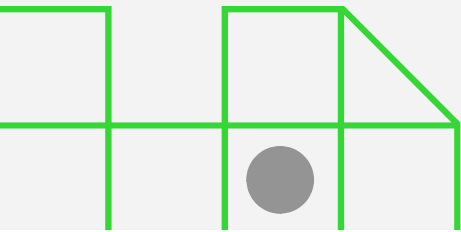
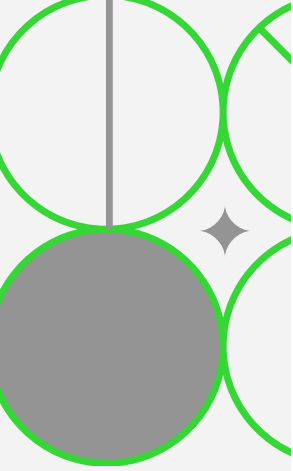
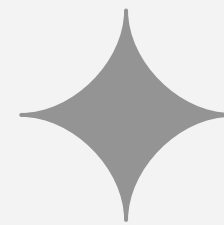
This presentation explores the **efficient** use of the *K-Nearest Neighbors* algorithm for **Iris classification**.

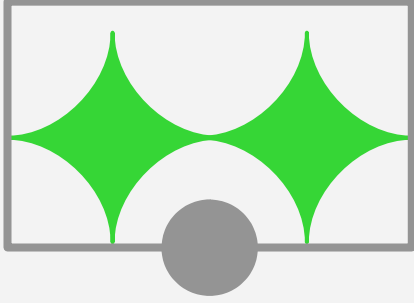
The goal is to achieve accurate and fast classification of Iris flowers based on their features.



Iris Dataset

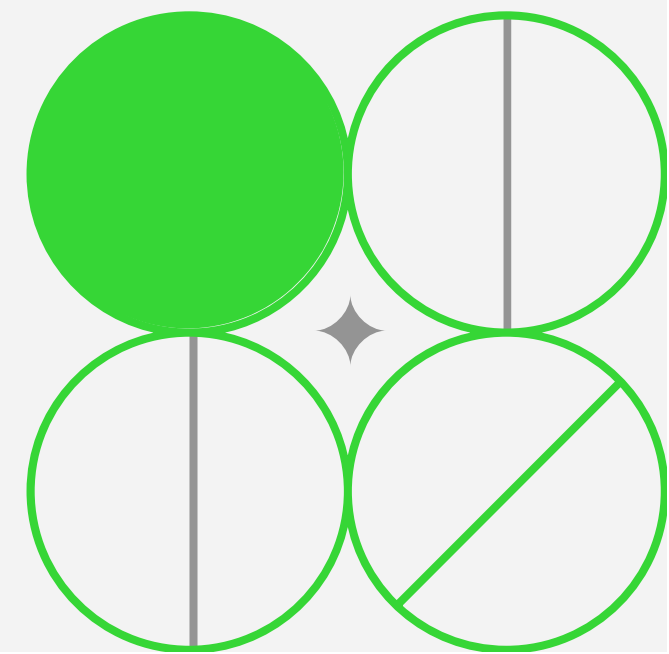
The **Iris dataset** contains measurements of *sepal length*, *sepal width*, *petal length*, and *petal width* for three species of Iris flowers: *setosa*, *versicolor*, and *virginica*. It is a widely used dataset in machine learning.





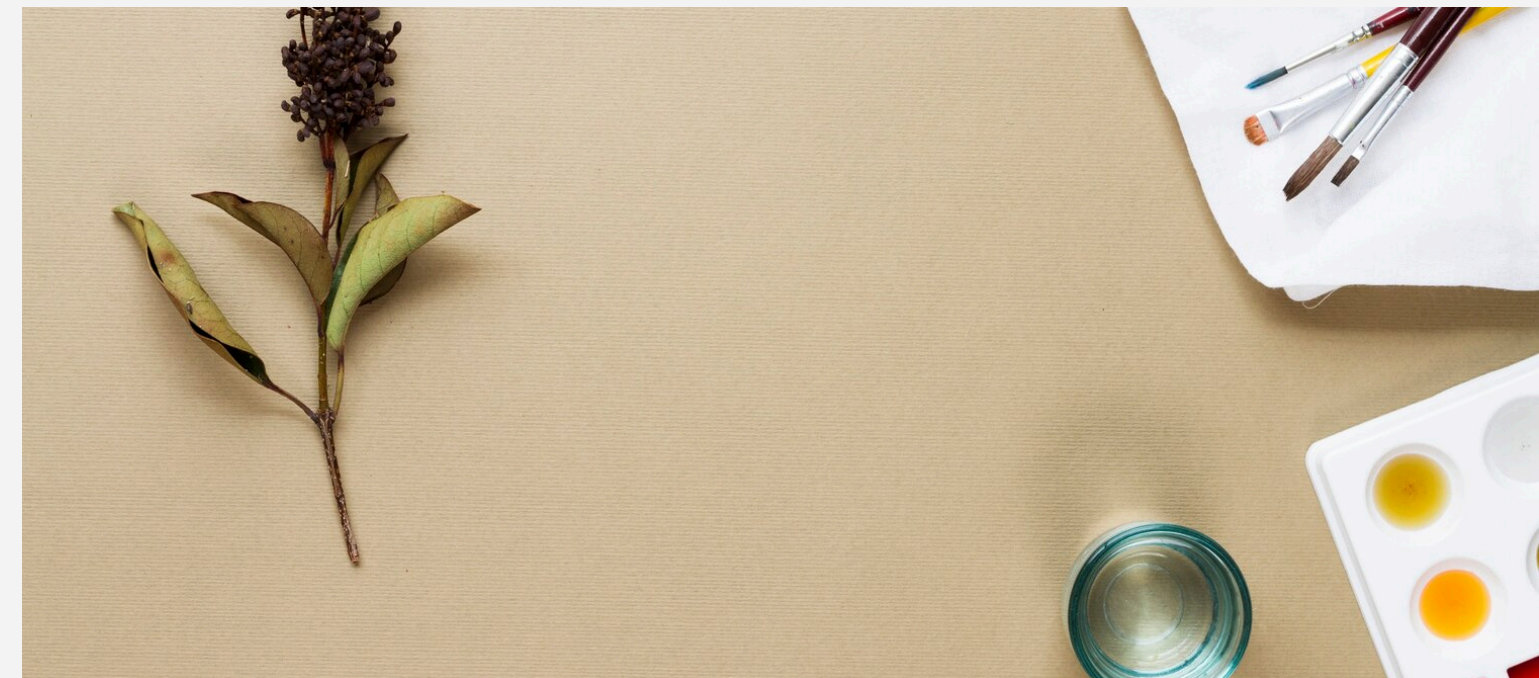
K-Nearest Neighbors Algorithm

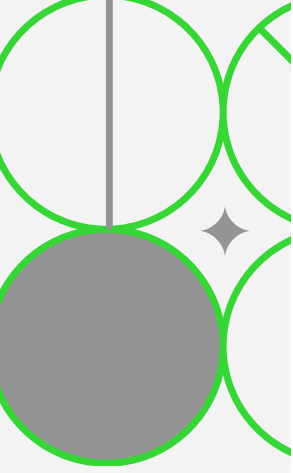
The **K-Nearest Neighbors (K-NN)** algorithm is a simple and effective **classification** algorithm. It classifies a data point based on how its neighbors are classified. It is non-parametric and can handle multi-class classification.



FEATURE SELECTION

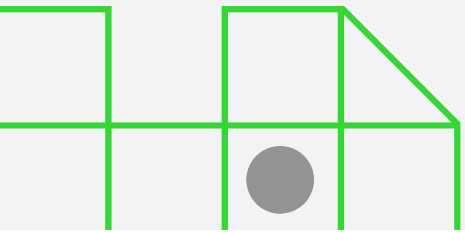
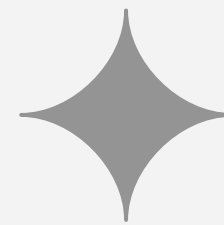
Choosing the right **features** is crucial for the success of the K-NN algorithm. In the context of Iris classification, the sepal and petal measurements serve as the key features for distinguishing between different species of Iris flowers.





Model Training

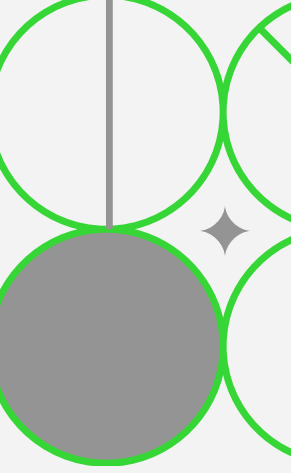
The **training process** involves feeding the K-NN algorithm with the Iris dataset to create a model that can accurately classify new instances of Iris flowers. The algorithm calculates the distance between data points to make predictions.



HYPERPARAMETER TUNING

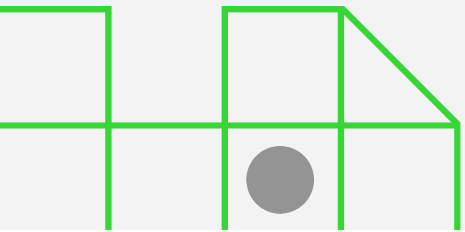
Fine-tuning the **hyperparameters** of the K-NN algorithm, such as the value of K, is essential for optimizing its performance. This process involves finding the optimal balance between bias and variance.





Evaluation Metrics

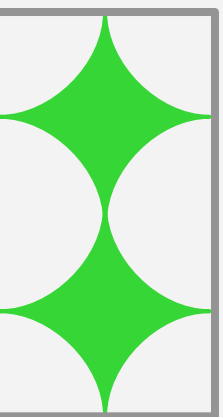
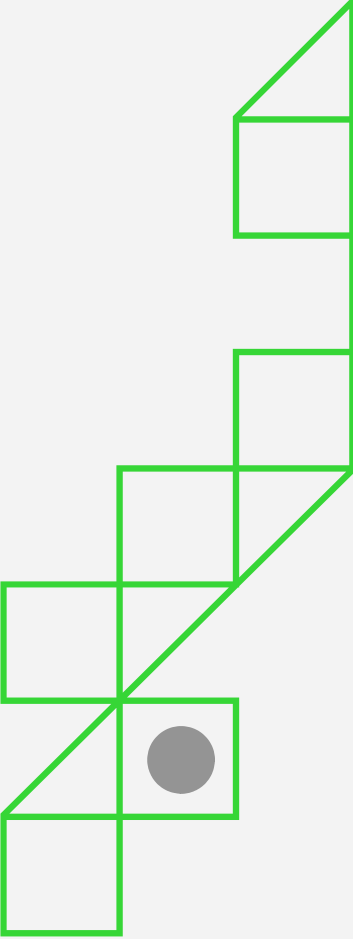
Various **evaluation metrics** such as accuracy, precision, recall, and F1 score are used to assess the performance of the K-NN model for Iris classification. These metrics provide insights into the model's effectiveness.





CONCLUSION

In conclusion, the **K-Nearest Neighbors** algorithm offers an **efficient** approach to **Iris classification**. By leveraging the right features, training the model effectively, and fine-tuning the hyperparameters, accurate classification results can be achieved.



Thanks!

