

# Wine Quality Prediction

Data Analytics Internship

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## PROBLEM STATEMENT

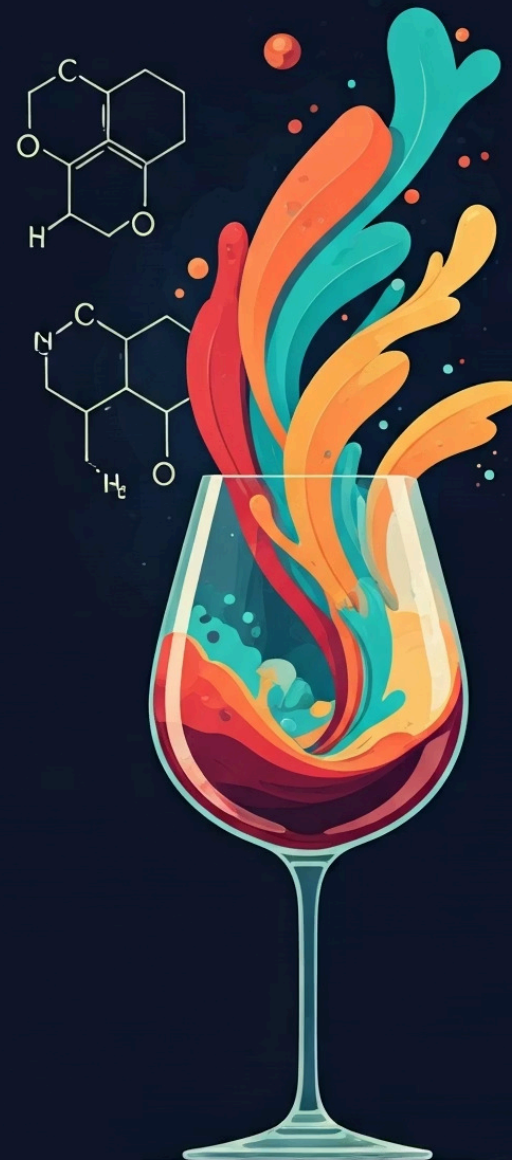
# Predicting Wine Quality

## Goal

Predict wine quality (Good / Bad) using chemical attributes.

## Approach

Compare multiple classifiers and evaluate model performance.



# Understanding Our Data



**1599**

**Total Records**

**11**

**Features**

Chemical attributes

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**Target Variable**

Quality (binary classification)



## DATA EXPLORATION

# Key Data Insights

Our initial exploration revealed important correlations within the dataset.



## No Missing Values

Dataset was clean, requiring minimal preprocessing.



## Alcohol Correlation

Strong positive correlation with wine quality.



## Volatile Acidity

Negatively correlated with wine quality.

## Classifiers Used

We employed a train-test split of 80:20 for model training and evaluation.



### Random Forest

Ensemble learning method for classification.



### Support Vector Classifier

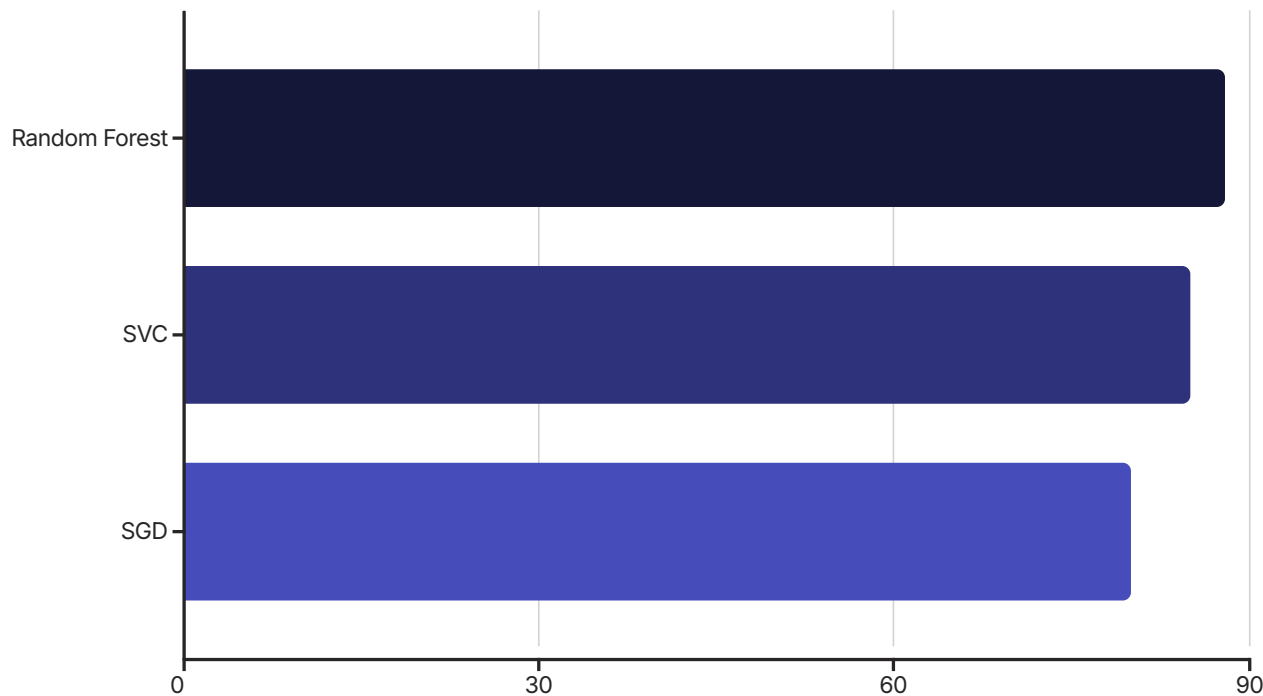
Effective in high-dimensional spaces.



### Stochastic Gradient Descent

Efficient for large-scale learning.

# Comparison of Classifiers



The Random Forest Classifier demonstrated the highest accuracy among the models tested.

## Random Forest

88% Accuracy

## SVC

85% Accuracy

## SGD

80% Accuracy



# Top Influential Chemical Attributes

Certain chemical properties significantly impact wine quality predictions.



## Alcohol

Higher alcohol content generally indicates better quality.



## Volatile Acidity

Lower levels are often associated with higher quality.



## Sulphates

Contribute to wine preservation and quality.



## Density

Related to sugar content and overall body of the wine.



## Conclusion & Key Insights



### Multi-Model Classifier

Developed and compared multiple models.



### Key Factors Identified

Alcohol, volatile acidity, sulphates, and density are crucial.



### Best Accuracy

Achieved 88% accuracy with Random Forest.