

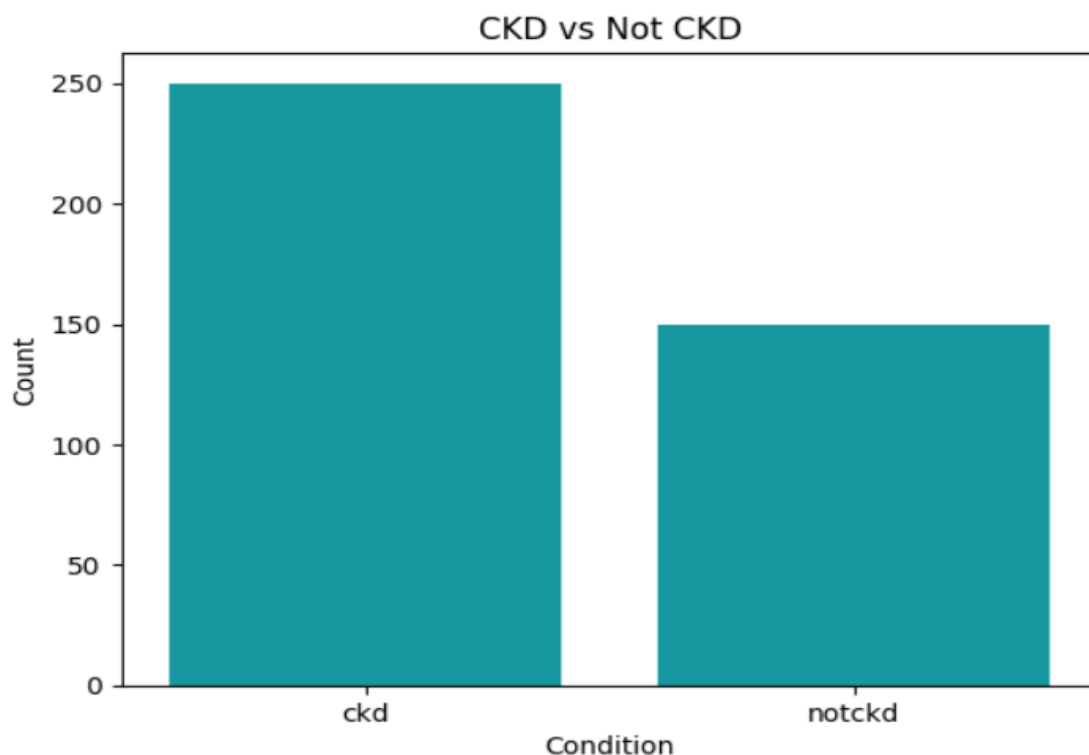
# Chronic Kidney Disease – Visual Analysis

## Exploratory Data Analysis & Model Results

### Section 1 – Data Distribution & Balance

#### Figure 1: CKD vs Non-CKD Bar Chart

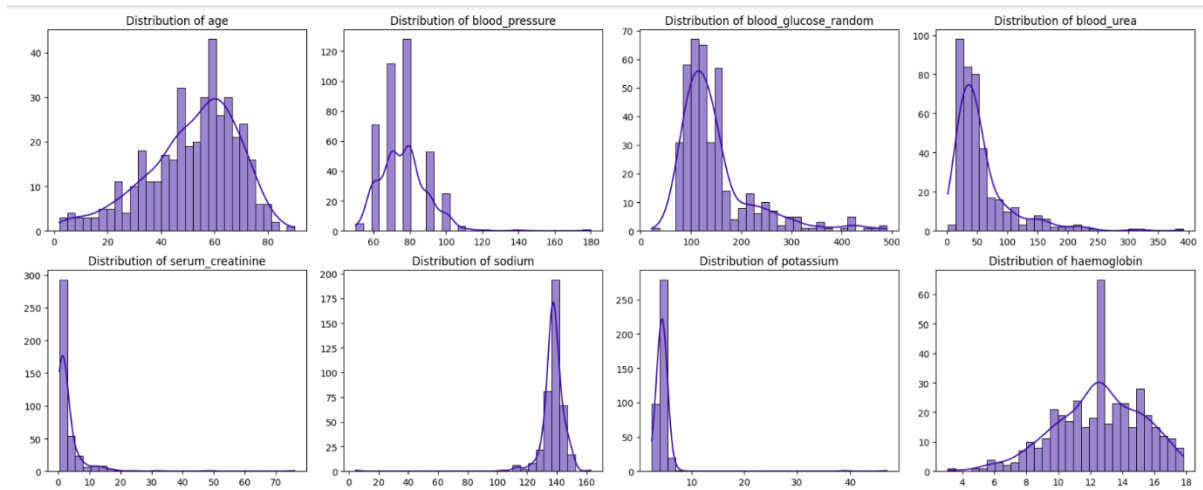
Shows the distribution of patients with CKD compared to those without. Reveals class imbalance, which must be considered during model training.



### Section 2 – Feature Distributions

#### Figure 2: Histograms of Numerical Features

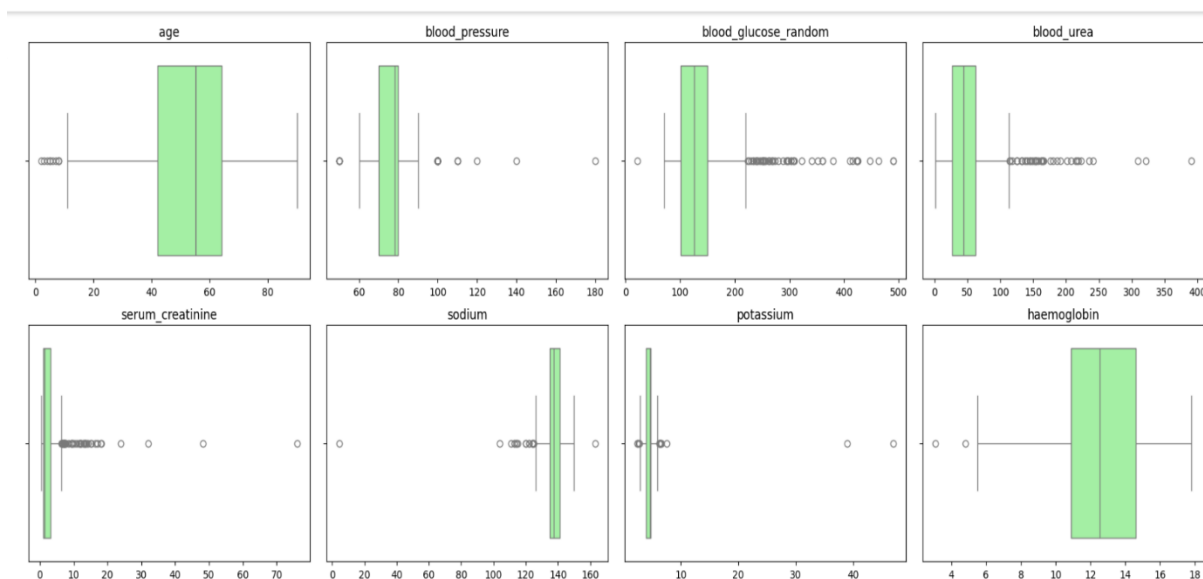
Visualizes spread and skewness of lab values like serum creatinine, haemoglobin, and blood urea. Identifies critical markers for CKD detection.



## Section 3 – Outlier Detection

### Figure 3: Boxplots of Key Numerical Features

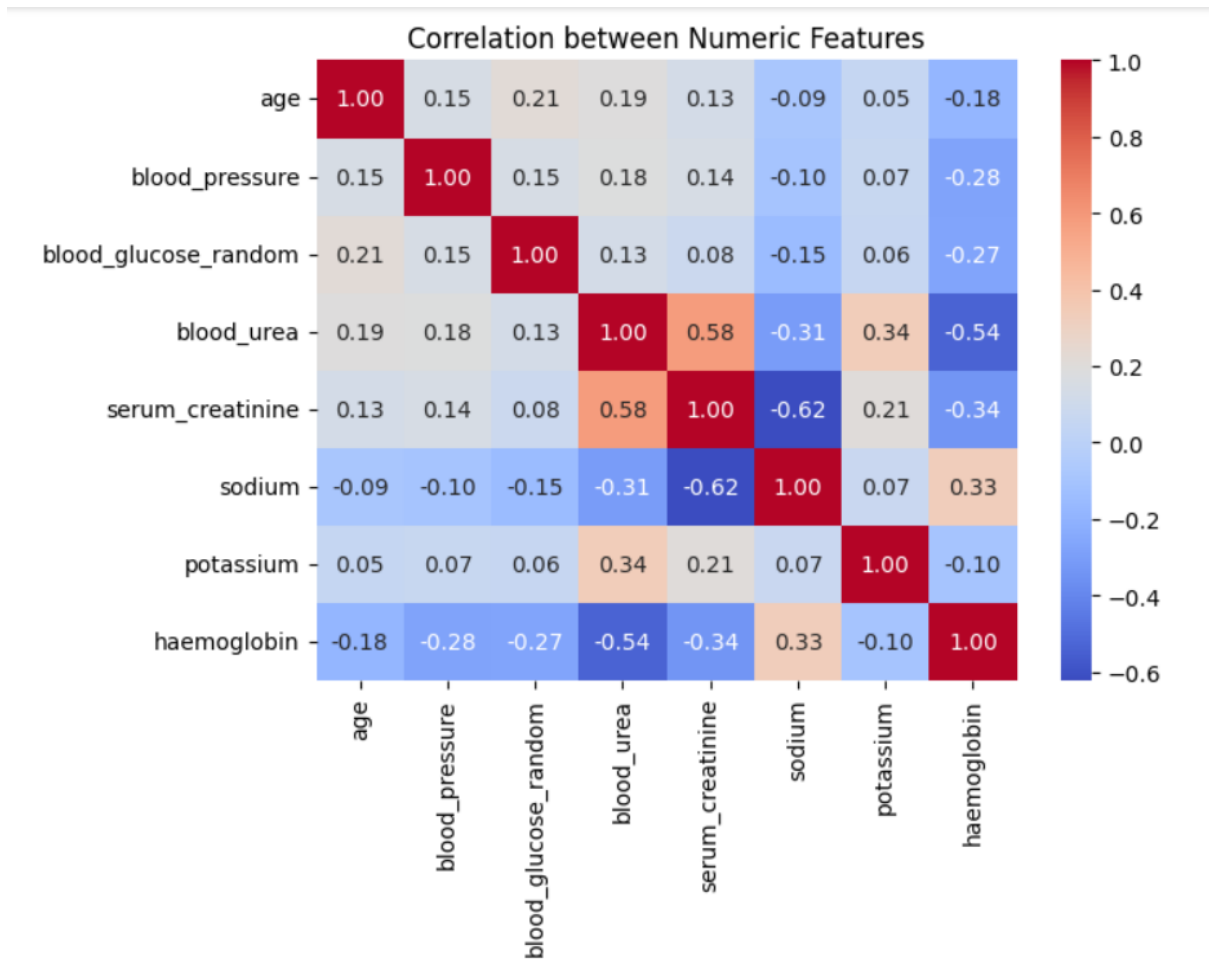
Highlights presence of extreme values in features such as blood urea and serum creatinine. Outlier handling was required to improve model reliability.



## Section 4 – Correlations

### Figure 4: Correlation Heatmap

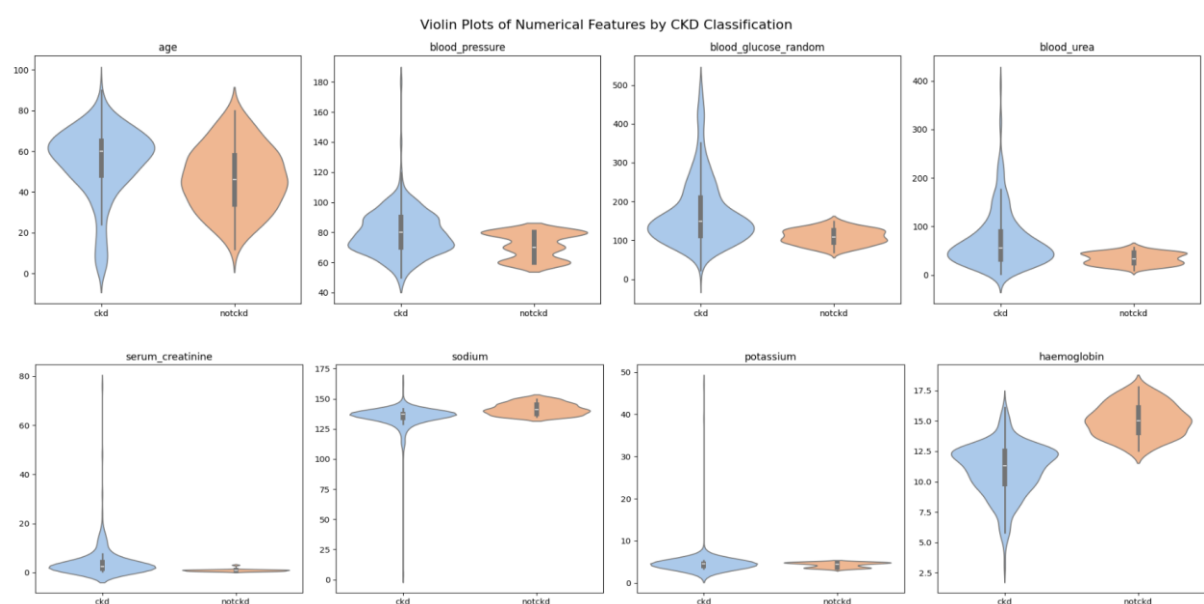
Strong positive correlation of serum creatinine & blood urea with CKD. Haemoglobin shows a negative correlation with CKD cases.



## Section 5 – Bivariate/Multivariate Insights

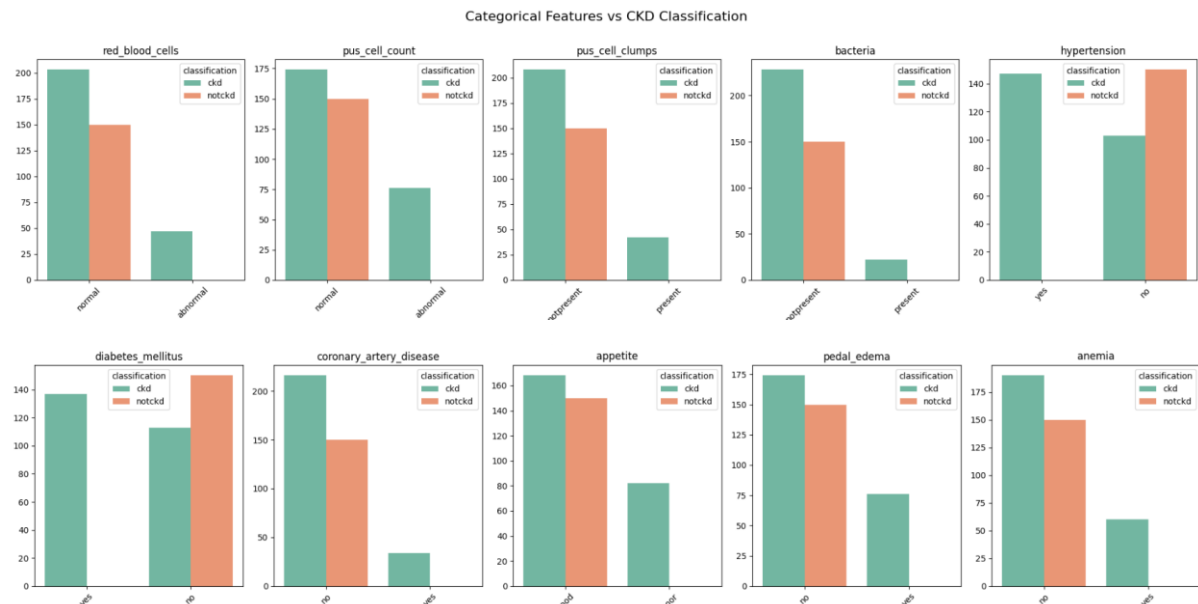
**Figure 5: Violin Plot of Numerical Features**

Lower haemoglobin levels strongly associated with CKD patients.



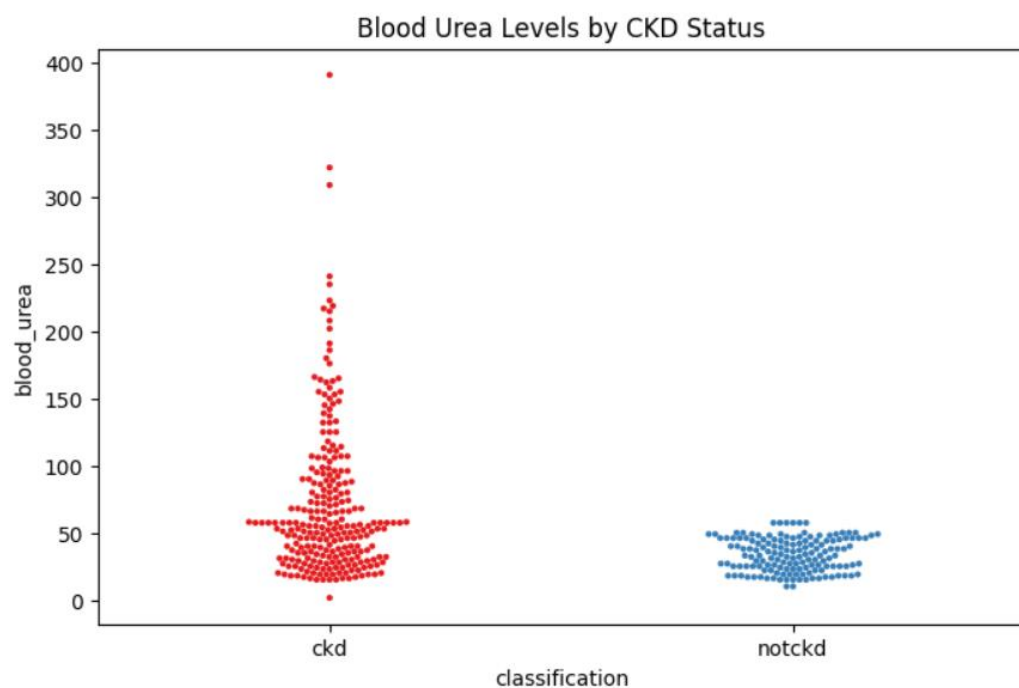
**Figure 6: Barplots for categorical features**

Shows how comorbidities like hypertension and diabetes increase likelihood of CKD.



**Figure 7: Swarmplots**

Swarmplots helped visualize point-wise variations in blood urea and serum creatinine by CKD classification.



## Section 6 – Model Results

### Figure 7: Accuracy Comparison of Different Models

SGBBoost selected as final model with accuracy 98.7%, balancing generalization and performance.



## Section 7 – Summary

The visual analysis confirmed that lab values (serum creatinine, albumin, haemoglobin) and comorbidities (hypertension, diabetes) are key indicators of CKD. These insights guided model feature selection and improved interpretability for medical relevance.