

- Download the dataset **Breast_Cancer.csv**
 - Using Python, visualize the various characteristics of the dataset.
 - Use Google Colab for implementation.
 - Define your own conditions to learn the characteristics
 - After each visualization write your findings
 - **EDA Questions**
1. What is the distribution of malignant vs benign cases?
 - Use count plots.
 2. Which features differ the most between malignant and benign tumors?
 - Compare boxplots of key features like radius_mean, texture_mean, area_mean.
 3. Is there a strong correlation between radius, perimeter, and area?
 - Heatmap of correlation matrix.
 4. Which features are most useful in distinguishing malignant from benign?
 - Use violin plots for selected features (e.g., concavity_mean, smoothness_mean).
 5. How do "worst" features compare to "mean" features in classification power?
 - Compare distributions of mean vs worst for malignant/benign.
 6. Are there multicollinearity issues among features?
 - Inspect correlation heatmap for redundancy.
 7. Which top 5 features show the largest separation between malignant and benign tumors?
 - Use feature importance pairplots.
 8. What is the average tumor size (radius/area) for malignant vs benign?
 - Use grouped bar plots or descriptive stats.
 9. Are there any outliers in the dataset? How might they affect classification?
 - Boxplots or scatter plots for area and radius.
 11. Which features show the highest variance, and do they contribute to classification?
 - Plot variance of each feature and compare across classes.
 12. How does texture differ between malignant and benign tumors?
 - Use histograms or KDE plots for texture_mean and texture_worst.
 13. Do tumors with higher concavity tend to be malignant?
 - Compare distribution of concavity_mean for both classes.
 14. What is the relationship between smoothness and compactness?
 - Scatter plot colored by diagnosis.
 15. Which class tends to have higher values of fractal dimension?
 - Use violin plots for fractal_dimension_mean.