# **Transcript for Video Explanation**

## [Video Introduction]

"Hello everyone, welcome to this video where I'll explain the code implementation for our breast cancer diagnosis research project. We've used logistic regression to build a model that can classify breast cancer cases as either benign or malignant based on various cell nucleus characteristics. I'll walk you through the code and explain how it works."

#### [Data Preprocessing]

"First, we load our dataset and check for missing values. We noticed that the 'Unnamed: 32' column contains missing values, so we dropped it. We then visualized the class distribution to understand the balance between benign and malignant cases."

#### [Data Cleaning]

"Next, we performed data cleaning by removing columns associated with 'worst' attributes, columns related to 'perimeter' and 'area,' and columns related to 'concavity' and 'concave points."

### [Model Building]

"We split our dataset into training and testing sets. We defined a logistic regression model using the Statsmodels library and fitted it to our training data. We used a formula that includes the selected features."

## [Model Evaluation]

"Moving on to model evaluation, we predicted the test data and converted the numerical probabilities into categorical labels. We displayed a classification report with precision, recall, and F1-score. We also calculated and displayed the confusion matrix to evaluate true negatives, false positives, false negatives, and true positives."

"We found that our model achieved an accuracy of 96.5%, which is quite promising for breast cancer diagnosis."

### [Conclusion]

"In conclusion, our logistic regression model successfully predicts breast cancer cases as benign or malignant with high accuracy. This code can be a valuable tool for medical professionals in diagnosing breast cancer. Thank you for watching, and don't forget to check out the code on our GitHub repository for a detailed look."

#### [Video Closing]

"Thank you for joining me in this video. If you found it helpful, please like and subscribe for more informative content. You can also access the code and detailed documentation on our GitHub repository. Feel free to leave any questions or comments below, and I'll be happy to assist you. See you in the next video!"