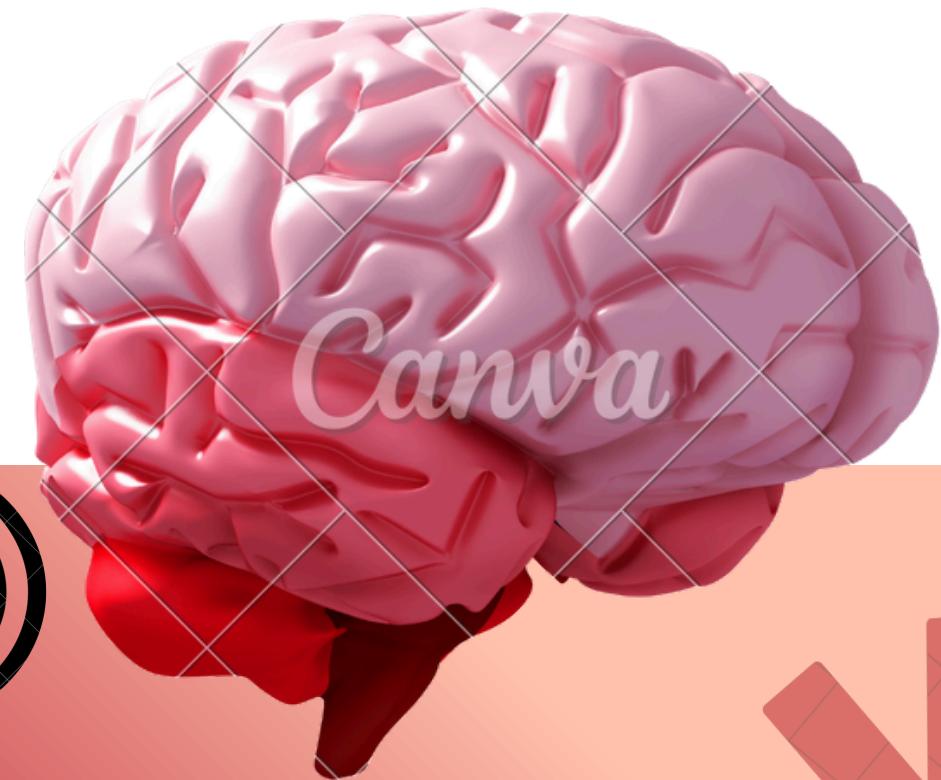


Predicting Parkinson's Disease Severity Using Voice Data

TEAM ELEVEN - 011



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Oxford Parkinson's Disease Detection Dataset

Source

Oxford Parkinson's Disease Detection Dataset
Donated on 6/25/2008

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References

Exploiting Nonlinear Recurrence and Fractal Scaling Properties for Voice Disorder Detection
Little, M.A., McSharry, P.E., Roberts, S.J. et al. Exploiting Nonlinear Recurrence and Fractal Scaling Properties for Voice Disorder Detection. BioMed Eng OnLine 6, 23 (2007).



UCI Machine Learning Repository

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Project Overview

What problem are we solving?

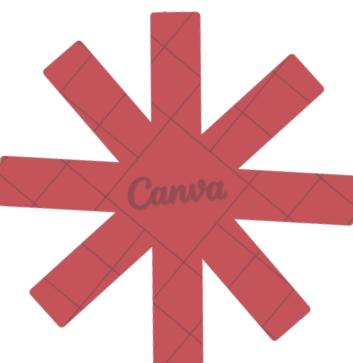
Parkinson's disease affects speech

Goal: **predict disease severity (total_UPDRS) from voice features**

Why Voice Analysis?

Voice recordings are non-invasive

Could support remote monitoring of patients



Data Selection and Preparation

Dataset characteristics

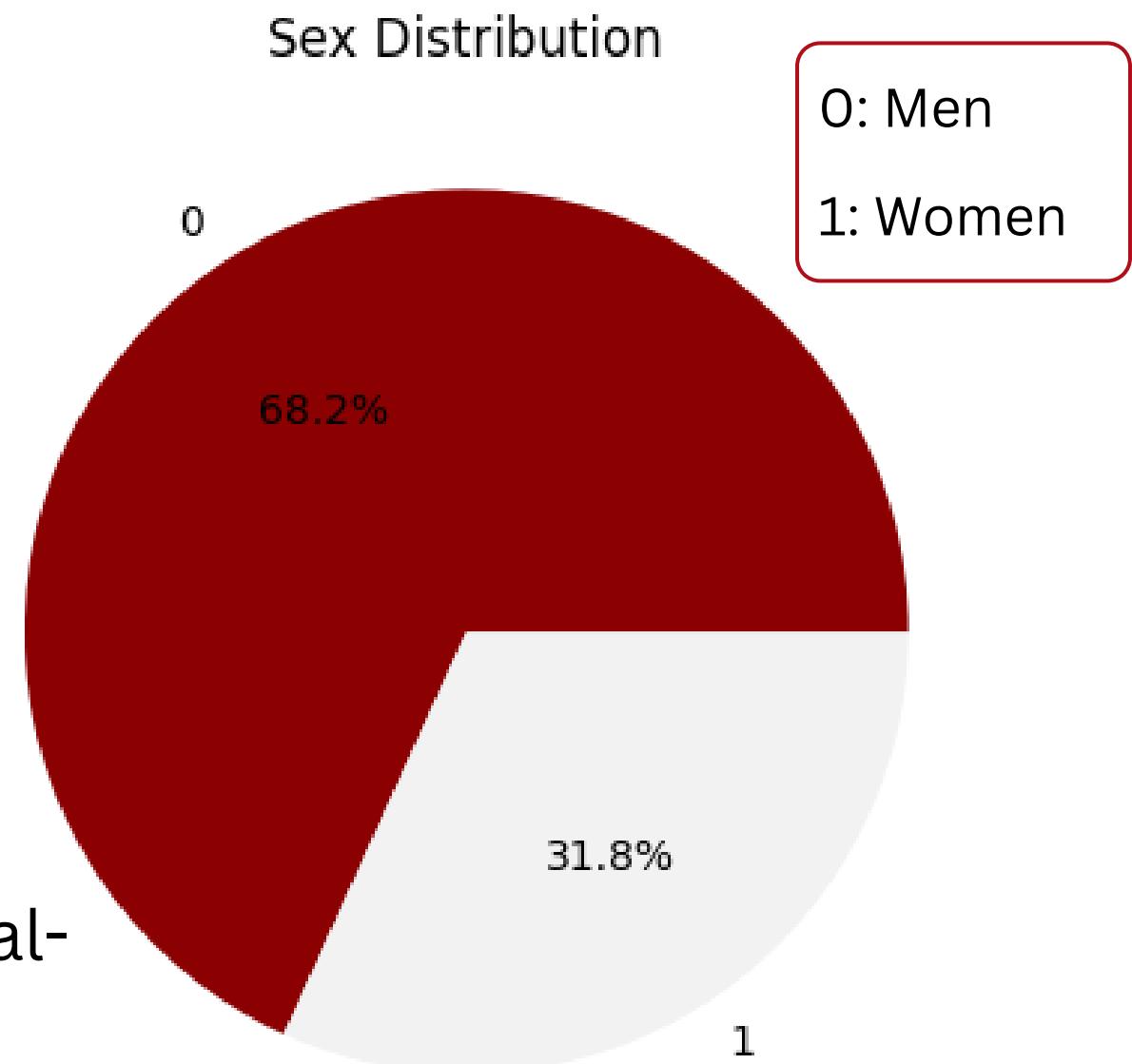
- Oxford Parkinson's Telemonitoring Dataset
- 42 patients with Parkinson's disease
- 5.875 voice recordings, collected remotely
- 140 average recordings per patient
- 6 month trial

Target

- total_UPDRS → overall disease severity score

Features

- Voice measurements such as jitter, shimmer and other signal-based features.



Model Building and Evaluation

MODEL SELECTION

Linear Regression

Used as a baseline model

Helps understand relationship
between voice features and target
(UPDRS score)

WHY THIS MODEL?

Predict a continuous value (UPDRS
score) --> fits Linear Regression
Simple and interpretable
Fast training

EVALUATION METRICS

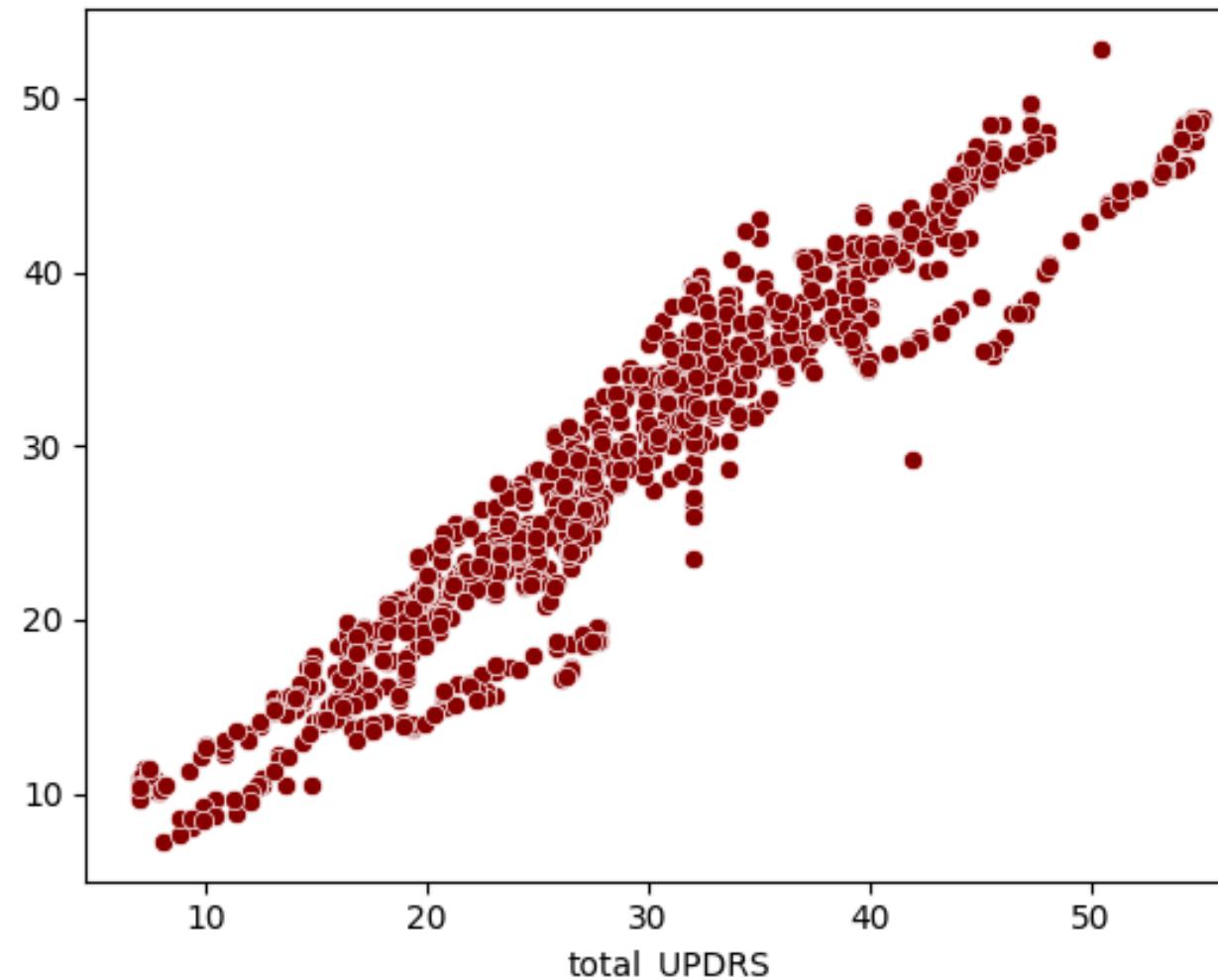
MAE (Mean Absolute Error): Average prediction error in UPDRS points

R² Score: Measures how well the model explains the variability in the data



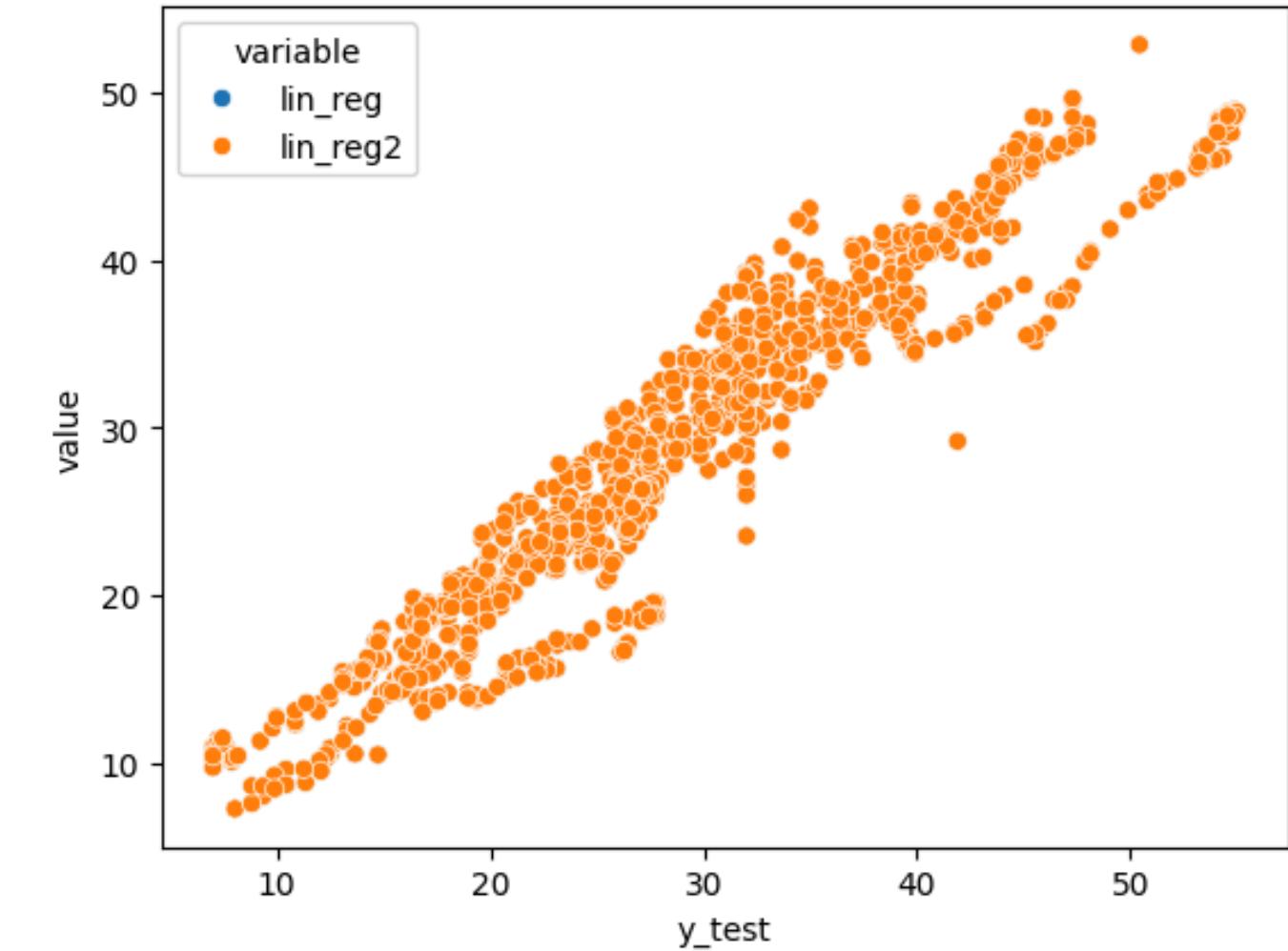
Feature Engineering and Selection

Linear Regression

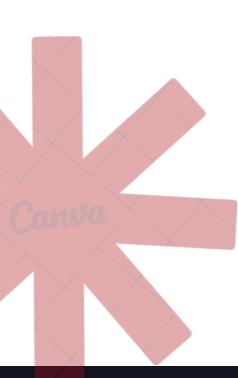


X, y Predictions

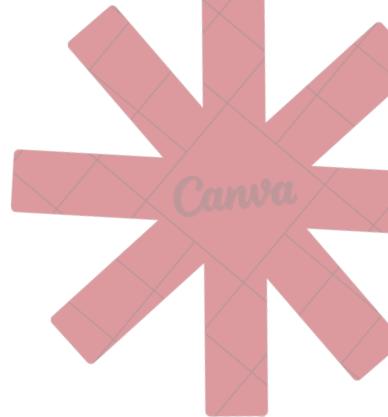
MAE	2.42
MSE	10.69
RMSE	3.27
R2	0.90



Our model predicts that a patient with minimal voice pathology would have a relatively low total UPDRS score of approximately **8.78**.



Hyperparameter Tuning and Model Optimization



Cross-Validation

Model K Fold

R² scores for each fold

0.91381079

0.90167049

0.91176976

0.91351376

0.90381669

Mean R²

0.909

Standard deviation

0.005

Advanced Model: Random Forest

Results:

- MAE ≈ 0.13
- R² ≈ 0.999

Interpretation:

Our model performed extremely well

We achieved:

Very low error

R² score close to 1



Challenges and Learnings

Challenges

Complex, multifactorial disease → voice explains
only part

Small number of patients (42) → limited variability

Learnings

Simple models are a good baseline

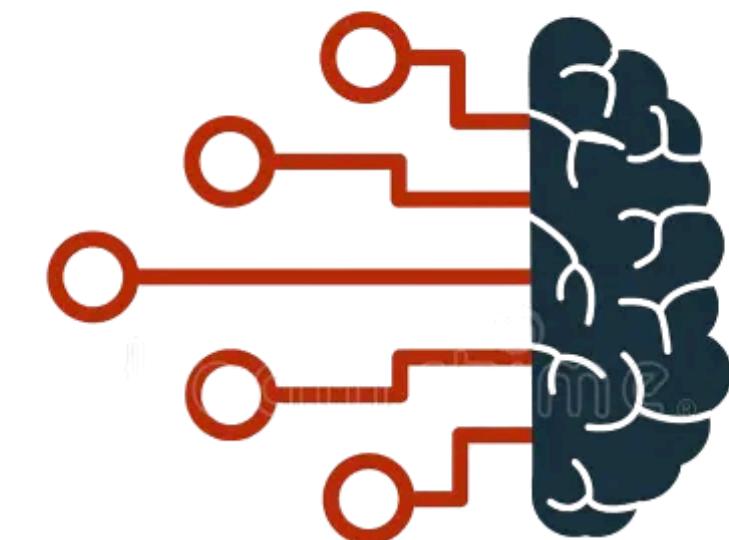
Ensemble models can improve performance



Conclusion

Voice data can help predict Parkinson's severity but is not sufficient alone.

Random Forest improved predictions over Linear Regression, confirming non-linear patterns exist in the data.



Real-World Application and Impact

Practical Applications:

- Accessible pre-screening tool
- Telemedicine / remote diagnosis
- Disease progression monitoring

Potential Impact:

- Early detection enables better treatment
- Cost reduction in healthcare
- Increased accessibility for patients

Ethical Considerations & Limitations:

- Limited dataset size (197 samples)
- Requires clinical validation
- Complements, not replaces, medical diagnosis



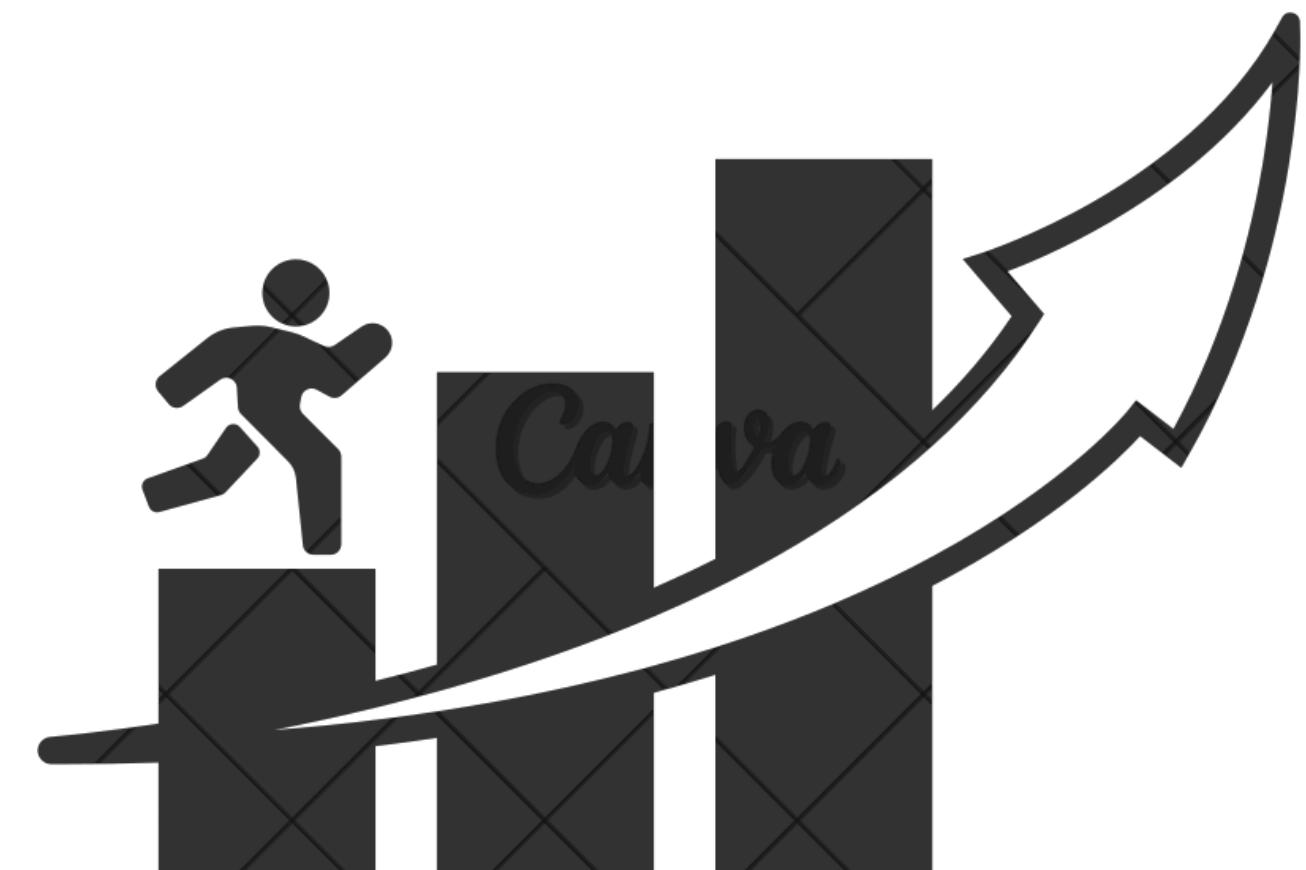
Future Work and Improvements

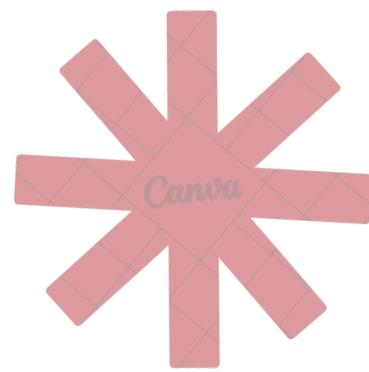
Areas for Expansion:

- Collect larger, more diverse dataset
- Experiment with Deep Learning models (Neural Networks)
- Integrate additional biomarkers (gait analysis, handwriting)

Potential Improvements:

- Mobile application for regular screening
- Longitudinal patient tracking system
- Multi-modal diagnostic approach





THANK YOU!

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