



# Vocal analysis as a screening for Parkinson's disease

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# Parkinson's disease (PD)

Neurodegenerative disease

Dopamine cells die, which results in too little dopamine

Host of different symptoms, motoric, cognitive

Voice is affected



# Vocal characteristics Parkinson's patients

Lower pitch

Lower volume

Monotone speaking patterns

More unstable in pitch (jitter)

More unstable in volume (shimmer)



# Statistical goals

Predict PD or not, with:

- Multiple linear regression model (mRLM)
- Logistic regression

H<sub>0</sub>: The predictions based on a regression model are no better than chance.

H<sub>A</sub>: The predictions based on a regression model are better than chance.



# Dataset 1 analysis

32 people

24 PD patients, 8 healthy controls

~20 variables

- Pitch
- Volume
- Jitter
- Shimmer

6 measurements per person



# Dataset 1 analysis

Normalization to Z-scores

Multicollinearity -> Variance Inflation Factor

High VIF scores ( $>10$ ), not illogical

Does not affect predictive power



# Multiple Linear Regression Model (MLRM)

Low F-statistic (0.9937)

High p-value (0.542) -> statistically insignificant

Probably due to dataset



# Dataset 2

252 people

188 PD patients

64 healthy controls

754 variables





# Multiple Linear Regression Model (MLRM)

F-statistic: 3.440

P-value:  $0.000524 < 0.05$

Statistically significant!



# Multiple Linear Regression Model (MLRM)

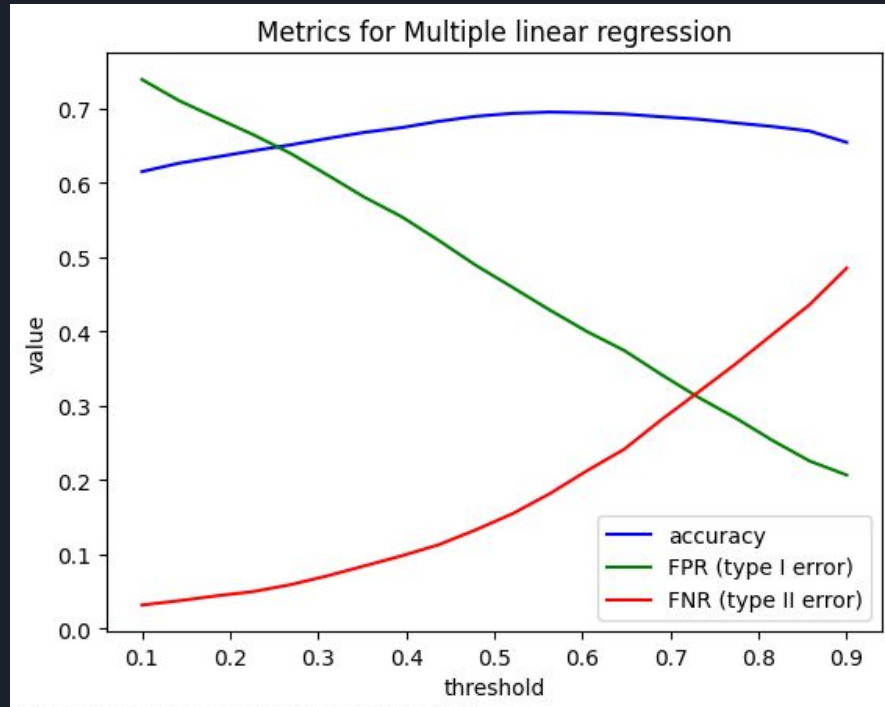
Let's test it!

60 (training) / 40 (test)

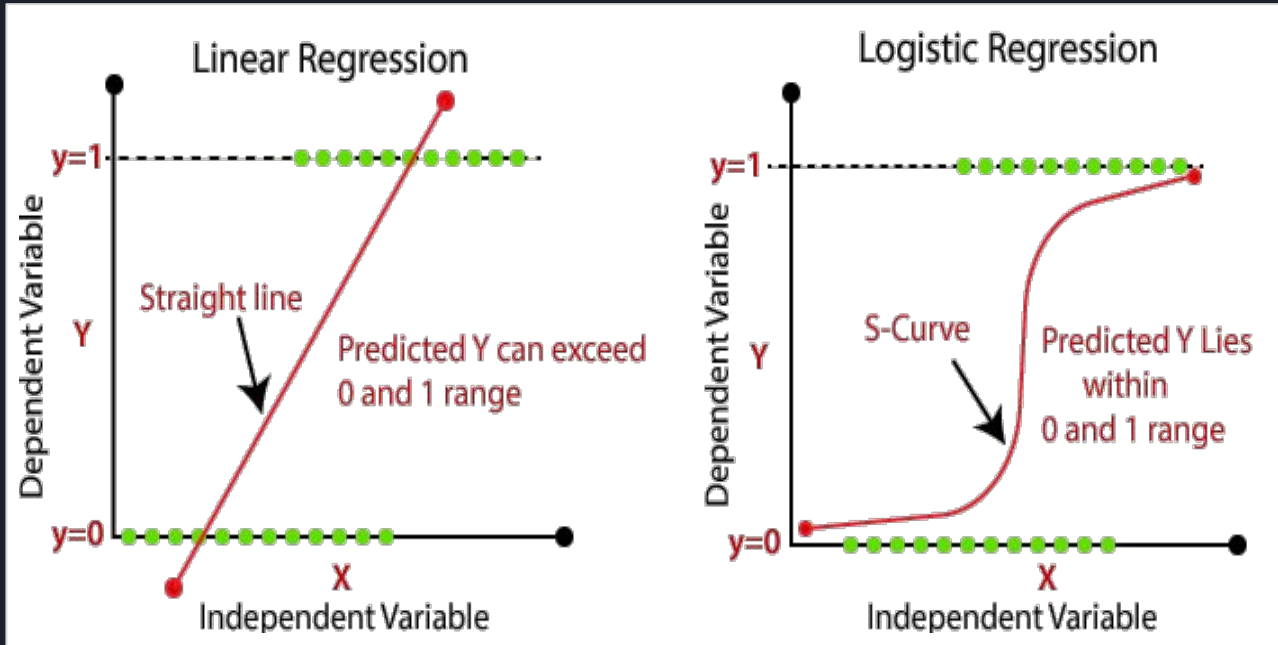
Equal amount of patients and controls in the test set

Classification using a threshold

# Multiple Linear Regression Model (MLRM)



# MLRM vs Multiple Logistic regression

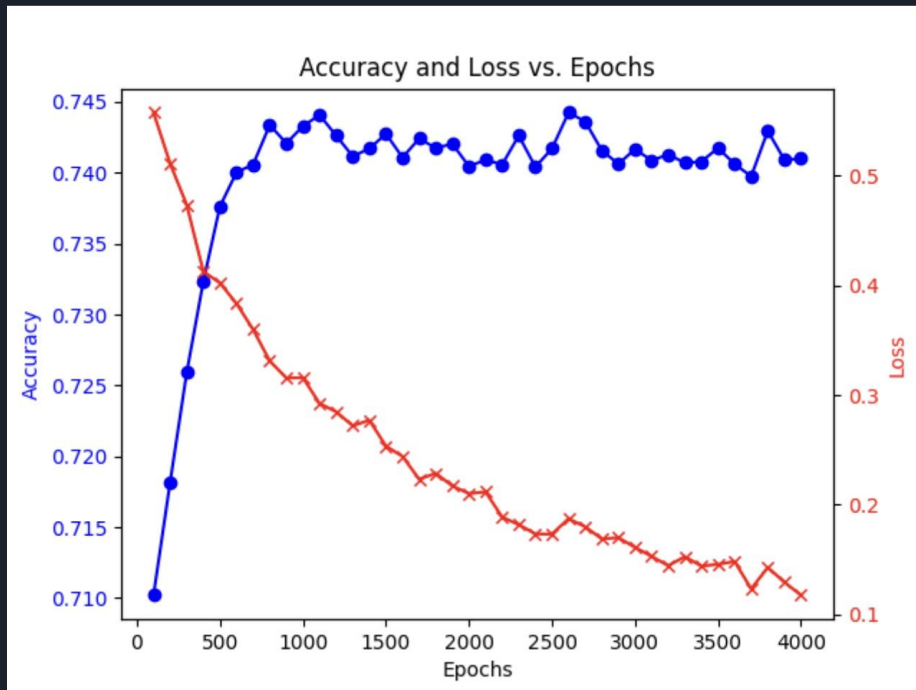


# Initial results: loss function

Demonstrative figure, with lower learning rate

Epochs: times every datapoint is processed

Loss goes down, but prediction accuracy plateaus



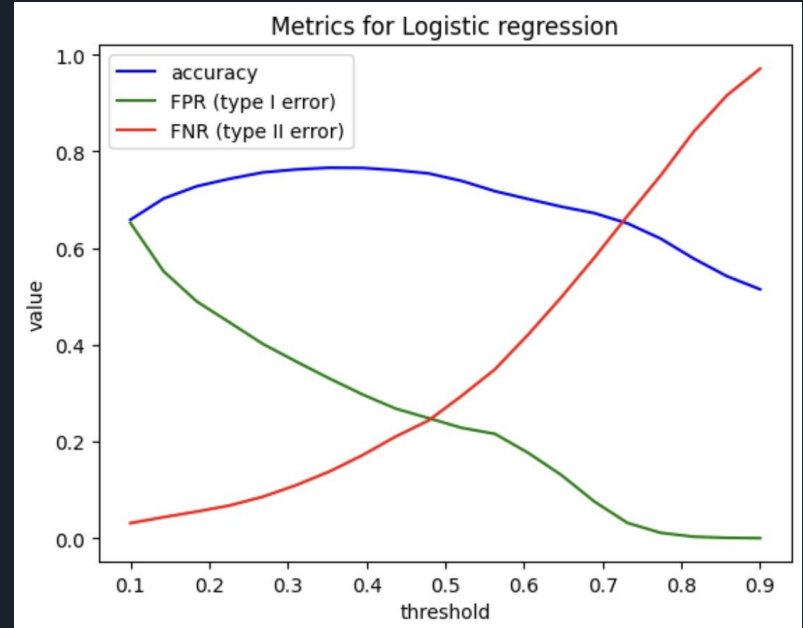
# Multiple Logistic regression

Improvement over MLRM

~75% prediction accuracy

Dependent on threshold

Influences accuracy, but false negatives are also important to avoid (screening)



# Elastic net regularisation

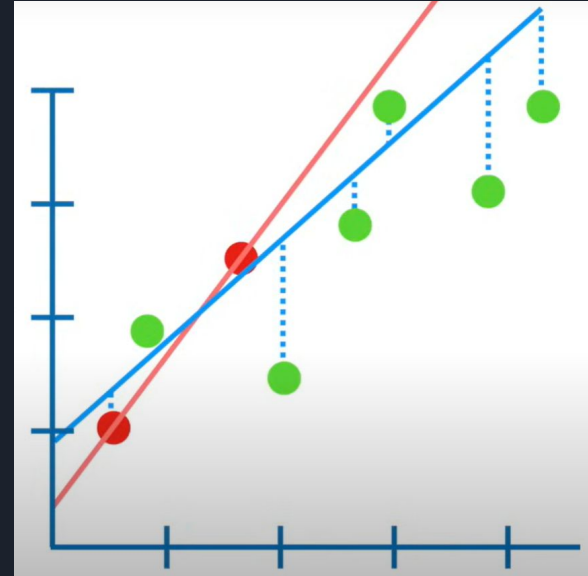
To eliminate variables that do not add to prediction

Lasso (L1): reduce weight, possibly to 0

Ridge (L2): reduce weight with penalty, not to 0

Optimal elastic net parameters:

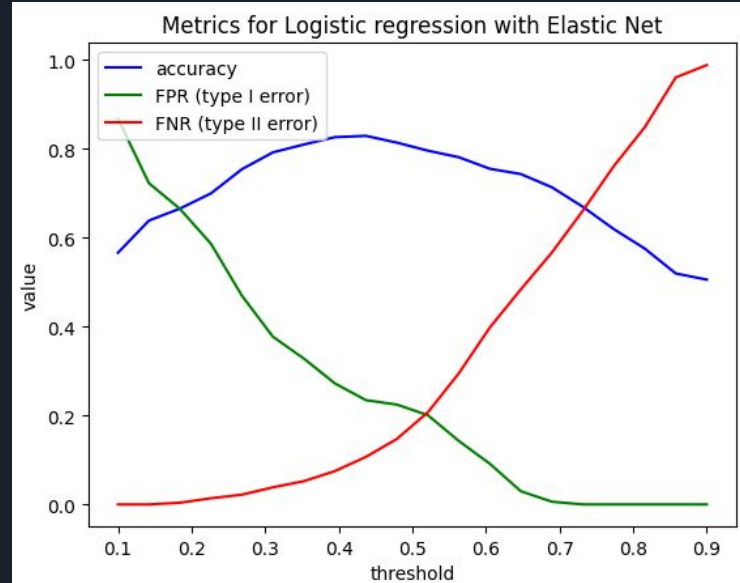
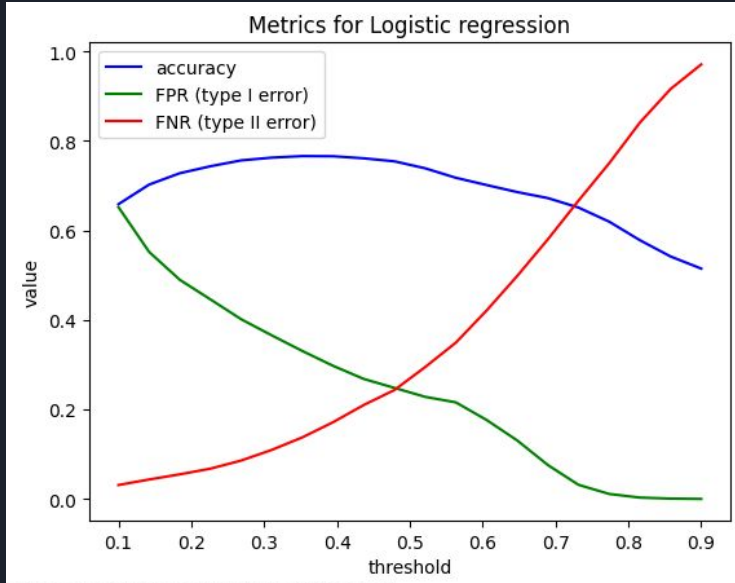
- L1 ratio: 0.1 (mostly ridge)
- alpha: 0.3



# Results after elastic net

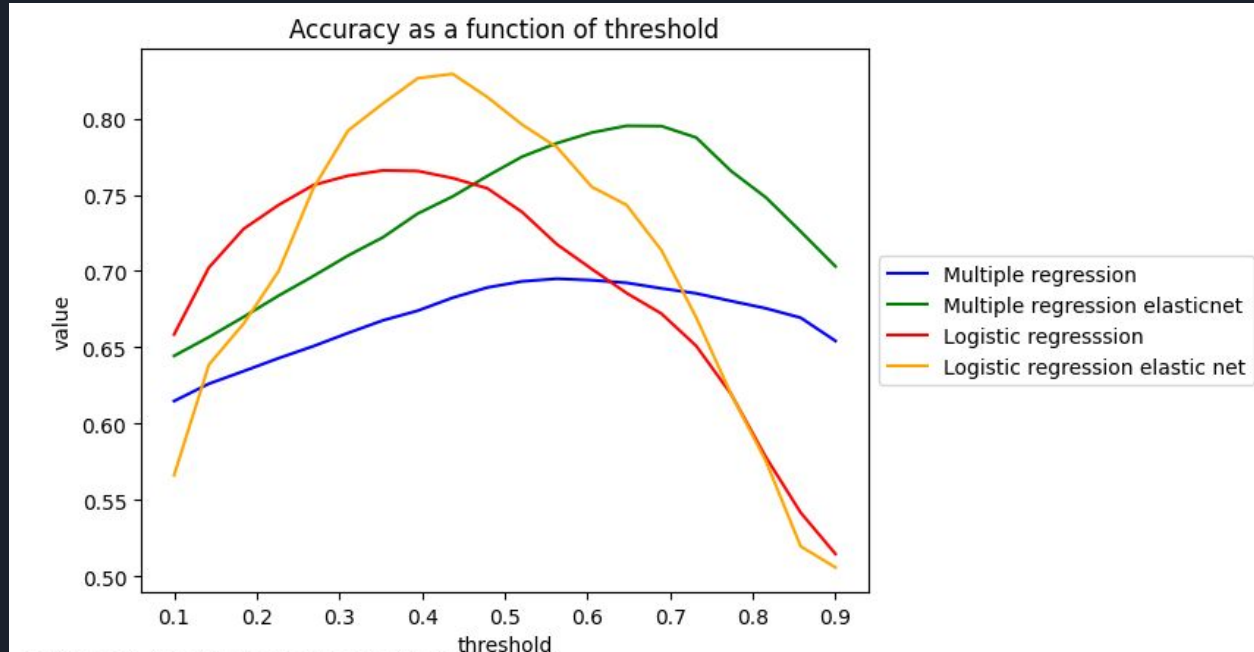
Slightly higher prediction accuracy with 79 variables when threshold is near 0.5

Accuracy is higher around a threshold of 0.4, which also has considerably lower false negatives

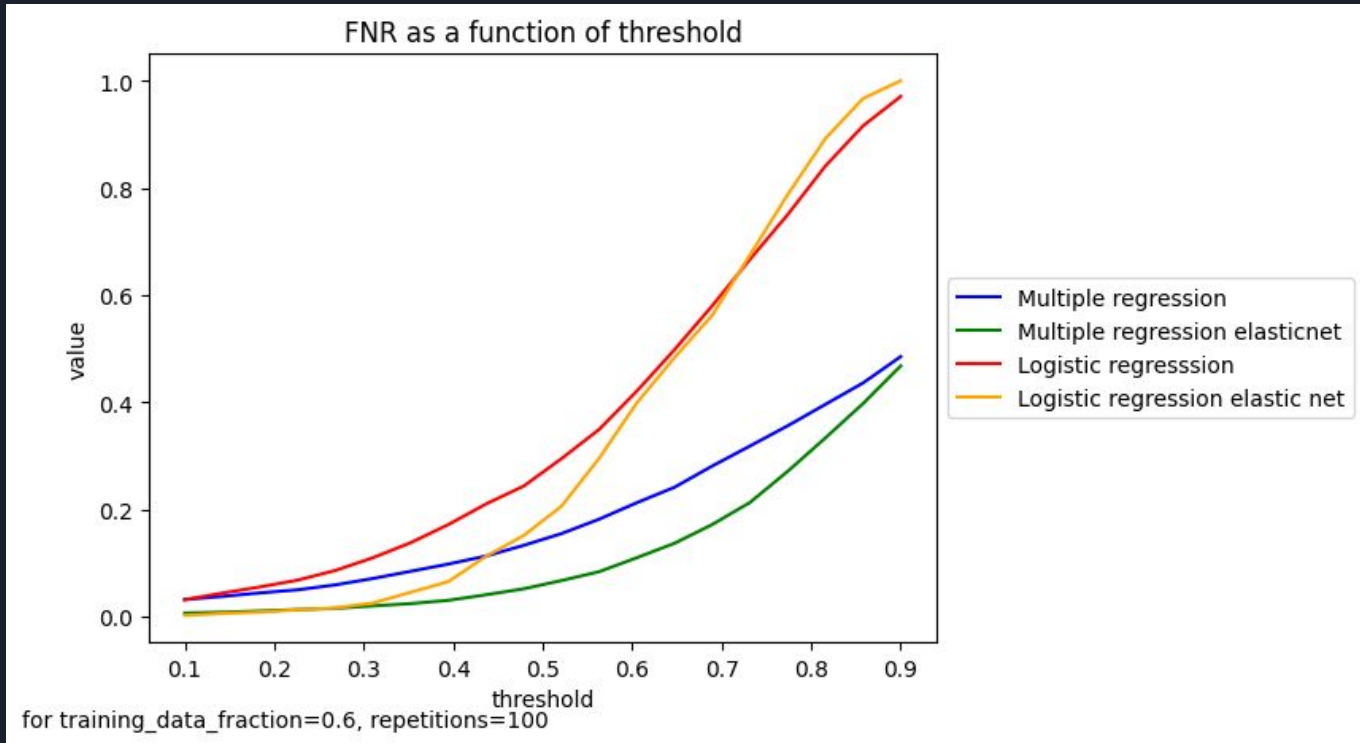




# Which method to choose?



# Which method to choose?





# Limitations to our method

Only ~83% accuracy

Still several false negatives

The current model would be expensive

Real world: much less PD patients



# Conclusion

Prediction is possible

Could be a cheap way to screen for Parkinson's if further researched

Cannot substitute actual diagnostics, but possibly valuable as screening



Questions?