Predicting Heart Disease

Implementations of Logistic Regression Models

Motivation

- One of the leading causes of death (World Health Organization, 2021)
- Understanding risk factors to help research



Dataset

- 300.000+ rows
- ~20 variables
- Heart disease as dependent variable



Dataset

Limitations

- Processing:
 - Loss of variability through grouping certain categories as binary
- Biases:
 - USA based dataset
 - Relatively uneven distribution of age among participants

| Health_Good_or_Better - | 1 | -0.045 | 0.25 | -0.22 | -0.032 | -0.12 | -0.18 | -0.23 | -0.21 | 0.017 | -0.12 | 0.053 | -0.11 | -0.16 | -0.13 |
|-------------------------|-------------------------|------------------|------------|-----------------|---------------|----------------|--------------|------------|-------------|------------|------------------|---------------|---------------|------------|-------------------|
| Recent_Checkup - | | 1 | -0.021 | 0.067 | 0.071 | 0.074 | | 0.11 | 0.13 | -0.1 | 0.2 | -0.089 | | | -0.019 |
| Exercise - | 0.25 | -0.021 | 1 | -0.096 | | | -0.085 | -0.14 | -0.12 | | -0.12 | 0.092 | -0.09 | -0.16 | -0.093 |
| Heart_Disease - | -0.22 | 0.067 | -0.096 | | 0.091 | 0.092 | | 0.17 | 0.15 | 0.073 | 0.23 | | | | 0.11 |
| Skin_Cancer - | | 0.071 | | 0.091 | 1 | 0.15 | | | 0.14 | | 0.27 | | | | 0.033 |
| Other_Cancer - | -0.12 | 0.074 | -0.054 | 0.092 | 0.15 | 1 | | 0.066 | 0.13 | | 0.24 | | | | 0.053 |
| Depression - | -0.18 | | -0.085 | | | 0.016 | 1 | 0.053 | 0.12 | -0.14 | -0.1 | -0.091 | | 0.11 | 0.1 |
| Diabetic - | -0.23 | 0.11 | -0.14 | 0.17 | | 0.066 | 0.053 | 1 | 0.14 | | 0.2 | | 0.17 | 0.21 | 0.056 |
| Arthritis - | -0.21 | 0.13 | -0.12 | 0.15 | 0.14 | 0.13 | 0.12 | 0.14 | 1 | -0.1 | 0.37 | -0.098 | 0.074 | 0.14 | 0.12 |
| Sex_Male - | 0.017 | -0.1 | 0.059 | 0.073 | | | -0.14 | -0.0028 | -0.1 | 1 | -0.061 | 0.7 | 0.35 | | 0.073 |
| Age_Normalised - | -0.12 | 0.2 | -0.12 | 0.23 | 0.27 | 0.24 | -0.1 | 0.2 | 0.37 | -0.061 | | -0.12 | -0.065 | | 0.13 |
| Height_norm - | | -0.089 | 0.092 | | 0.0068 | -0.043 | -0.091 | | -0.098 | 0.7 | -0.12 | | 0.47 | | 0.052 |
| Weight_norm - | -0.11 | | -0.09 | | | | | 0.17 | 0.074 | 0.35 | -0.065 | 0.47 | 1 | 0.86 | 0.048 |
| BMI_norm - | -0.16 | | -0.16 | | | | 0.11 | 0.21 | 0.14 | | | | 0.86 | 1 | 0.025 |
| Smoking_History - | -0.13 | | -0.093 | 0.11 | | | 0.1 | | 0.12 | 0.073 | 0.13 | | | 0.025 | 1 |
| | Health_Good_or_Better - | Recent_Checkup - | Exercise - | Heart_Disease - | Skin_Cancer - | Other_Cancer - | Depression - | Diabetic - | Arthritis - | Sex_Male - | Age_Normalised - | Height_norm - | Weight_norm - | BMI_norm - | Smoking_History - |

- 0.8

- 0.6

- 0.4

- 0.2

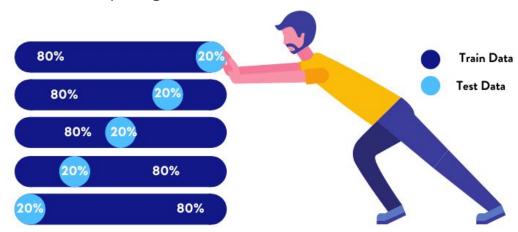
- 0.0

Research Questions and Hypothesis

- **RQ1**: Which features are the most predictive of heart disease when implementing a logistic regression model?
- **RQ2**: Does a logistic regression model based on PCA-generated components predict with higher accuracy than a logistic regression model fit on the 'clean' data?
 - H²-⁰: A logistic regression model using PCA-generated components does not achieve significantly higher accuracy compared to a logistic regression model using the original 'clean' data.
 - RQ2^{sub}_1: What implications do these components have on the interpretation of the predictive capacity of the features?
 - RQ2^{sub_2}: Does the usage of PCA components introduce any bias into the model?
- **RQ3**: Can we extract how many variables and which contribute to a high model accuracy?

Methods

- Logistic Regression
- Principal Component Analysis
- K-Fold Cross-Validation
 - Comparing PCA to Fully Fitted Model
 - Comparing Models With Select Variables



Methods: Coefficient Predictability

- RQ1: Which features are the most predictive of heart disease when implementing a logistic regression model?
 - Logistic regression predicts a binary outcome: heart disease (yes/no).

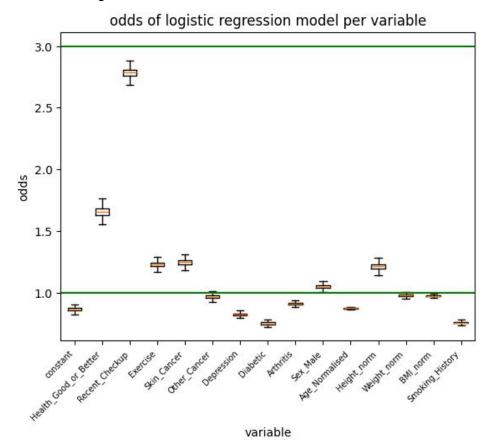
Use training set (75%) for regression model

Test regression model $\rightarrow 0.75$ accuracy

Calculate odds

Results: coefficient predictability

- < 1: low risk
- 1 -3 moderate risk
- > 3: high risk



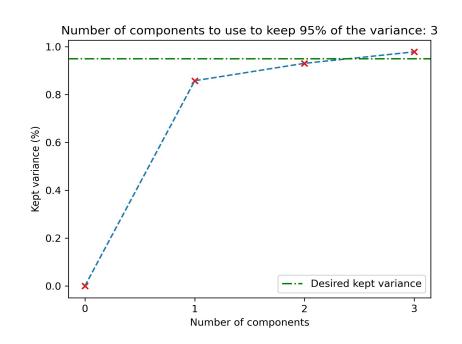
Limitations: coefficient predictability

- Interaction effects
- Multicollinearity
 - Weight, BMI, height
- Interpretation:
 - Recent checkups, exercise

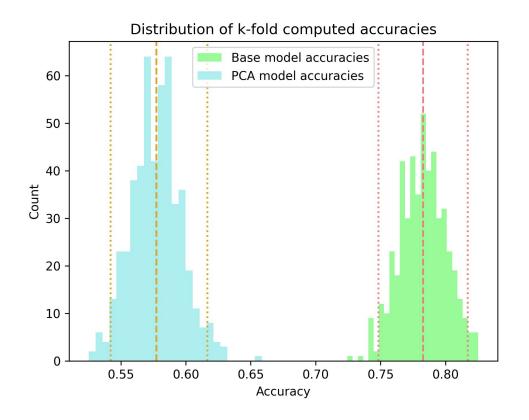
Principal Component Analysis

- 'Summary' of the dataset
- 3 Components for retainment of 95% of the variance

- PC 1: BMI, weight, height
- PC 2: Height, weight
- PC 3: Age



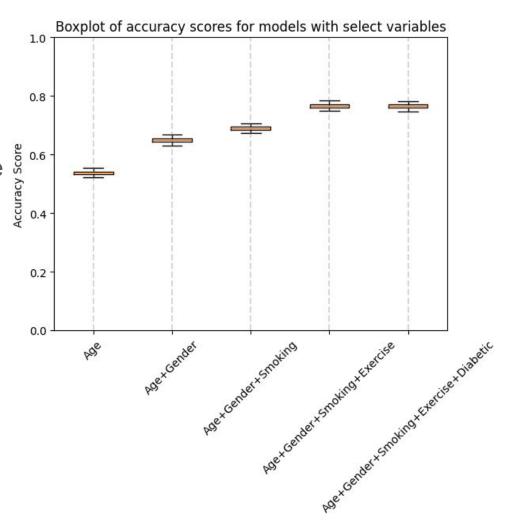
Results: Cross-Validation I



- We fail to reject our null-hypothesis
- Can we answer **RQ2**?
- Product of overfitting?
- AIC suggests this is not the case

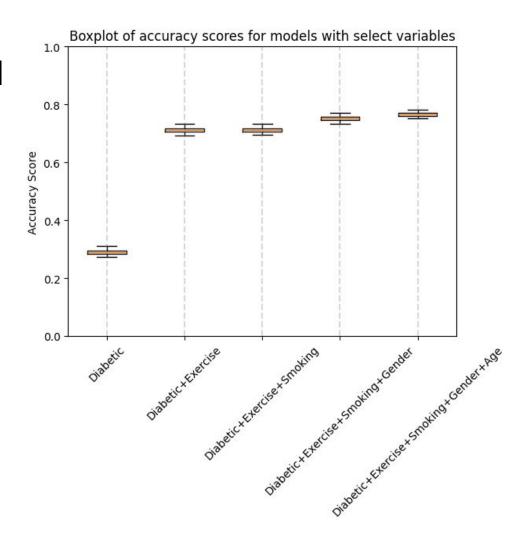
Results: Cross-Validation II

- RQ3: Can we extract how many variables – and which – contribute to a high model accuracy?
- Variable selection based on literature
 - Age
 - Gender
 - Smoking
 - Exercise
 - Diabetes
- Arbitrary order



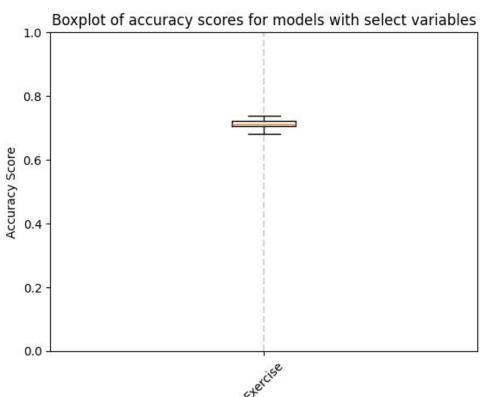
Results: Cross-Validation II

Reversed order of variables



Results: Cross-Validation II

- Is exercise that predictive?
- It seems like it, but we cannot say for certain.



Conclusion and Limitations

- Some insights into variable predictability
- PCA did not yield better results

Conclusion and Limitations

- Potential overfit
- No analysis of potential collinearity
- Correlation ≠ causation

Predicting Heart Disease

Implementations of Logistic Regression Models

Thank you for listening

Any questions?