

Classification of Mice on Protein Expression Level

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Introduction

Our project, "Classifying Mice Based on Protein Expression Levels," explores the relationship between protein expression, genetics, behavior, and treatments in mice. Using machine learning, we developed a model to classify mice into eight groups based on 77 proteins.

This research provides insights into how Down syndrome affects learning and memory and evaluates the potential benefits of the drug memantine for trisomic mice.





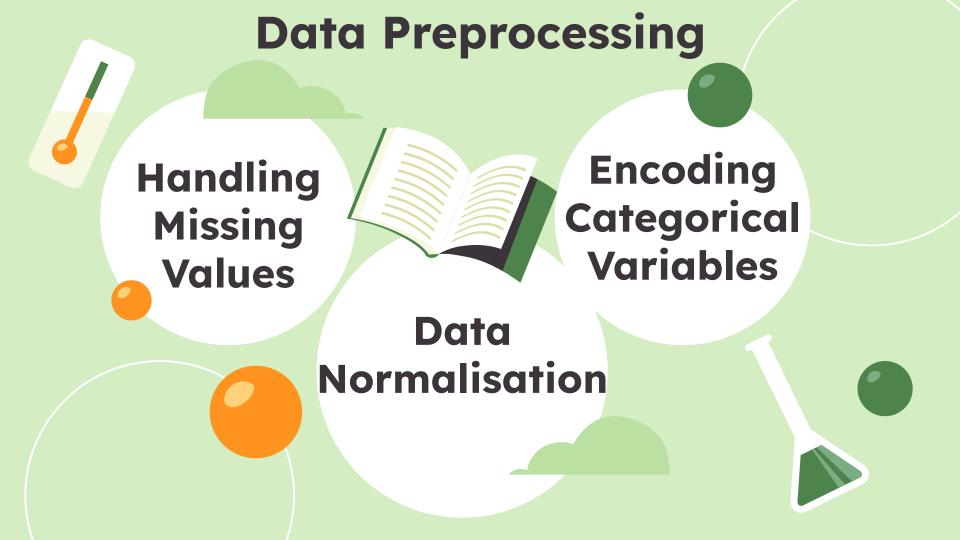


Exploratory Data Analysis

Model Evaluation

Feature Selection

Interpretation & Analysis





Distribution of protein expression levels across different classes.

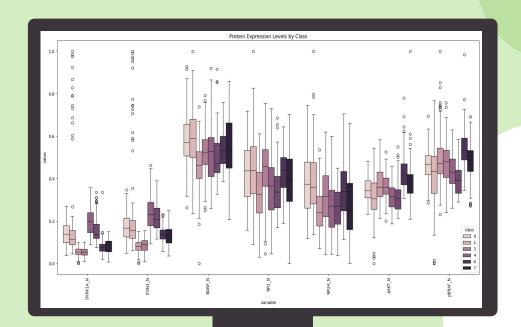
EDA



Examination of the distribution of samples across different classes to check for class imbalance.

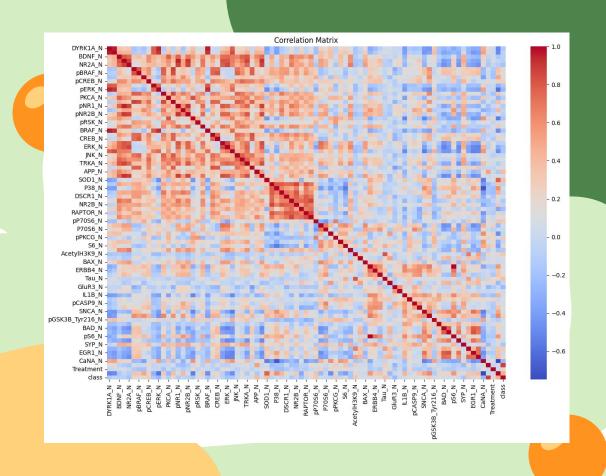


Identification and proportion of missing values in the dataset.

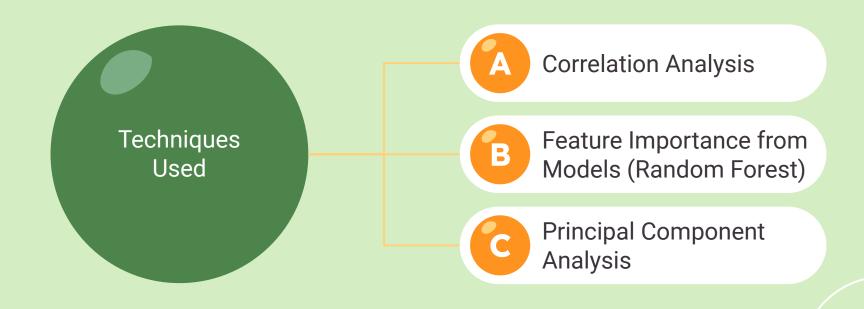


Box Plots

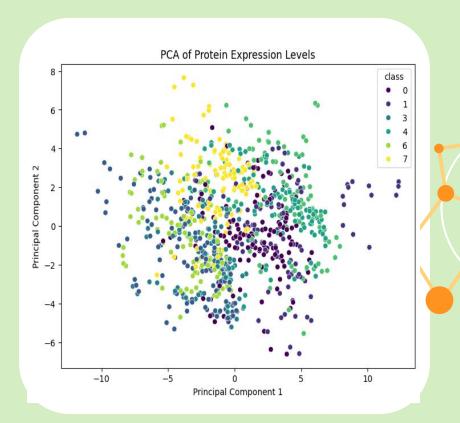
Created to visualise the distribution and variance of protein expression levels across different classes.



Feature Selection



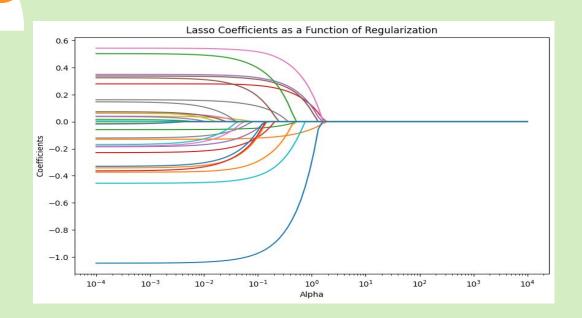




PCA reduces the complexity of the dataset while retaining most of the variance.

Tight, distinct clusters indicate good separability, while overlapping clusters suggest that the classes are not well separated by the chosen features.



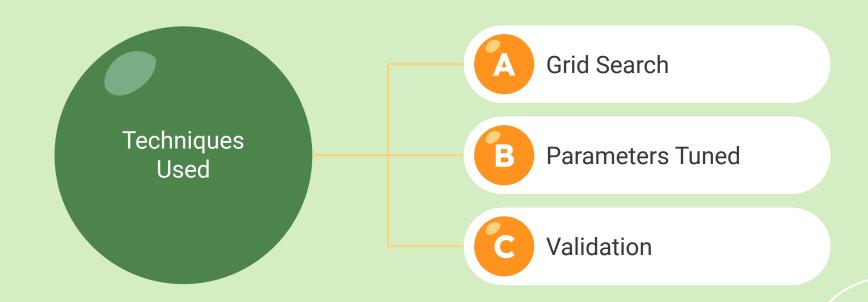


A subset of key proteins was identified as highly discriminative between control and trisomic samples.

Use of Lasso

The Lasso coefficients plot shows how different feature coefficients shrink to zero as the regularisation parameter (alpha) increases, many coefficients shrink to zero, indicating that Lasso regression effectively reduces the number of features by retaining only the most significant ones.

Model Training





Random Forest: An ensemble method that constructs multiple decision trees and outputs the mode of the classes for classification.

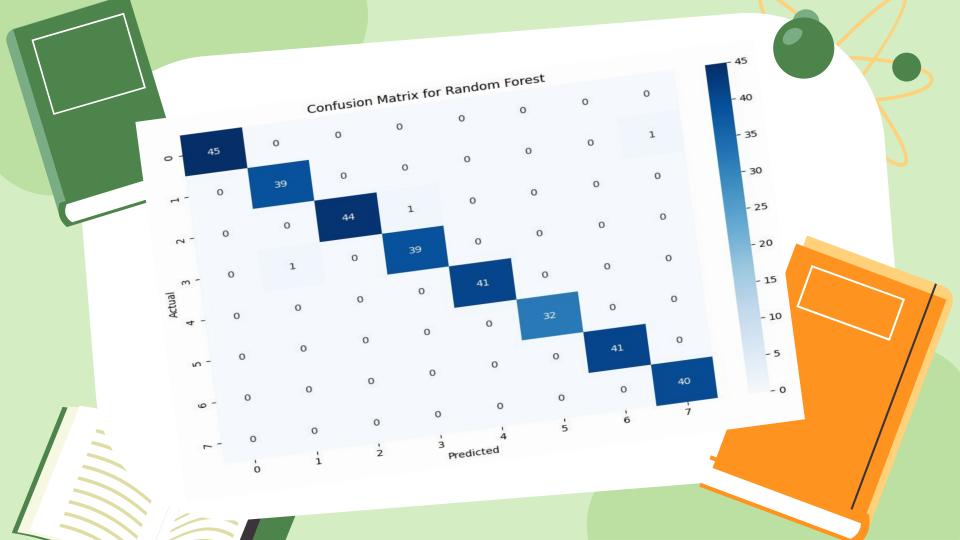
HyperParameter Tuning: An exhaustive search method that tests all possible combinations of predefined hyperparameters to find the best-performing model.

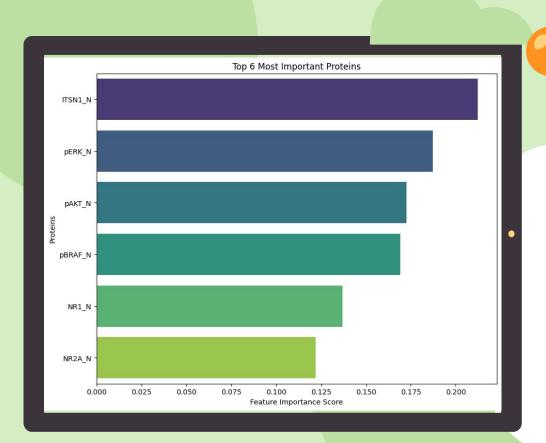
Model Evaluation



Here's a table

Classificatio	n Report for	Reduced	Random Fore	est Model:
	precision	recall	f1-score	support
ø	0.95	1.00	0.98	21
1	0.96	0.87	0.92	31
2	0.97	0.97	0.97	32
3	0.81	0.96	0.88	27
4	0.96	0.88	0.92	25
5	1.00	0.85	0.92	20
6	0.90	0.88	0.89	32
7	0.90	1.00	0.95	28
accuracy			0.93	216
macro avg	0.93	0.93	0.93	216
weighted avg	0.93	0.93	0.93	216

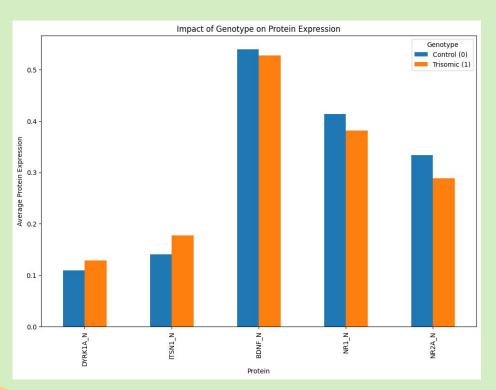








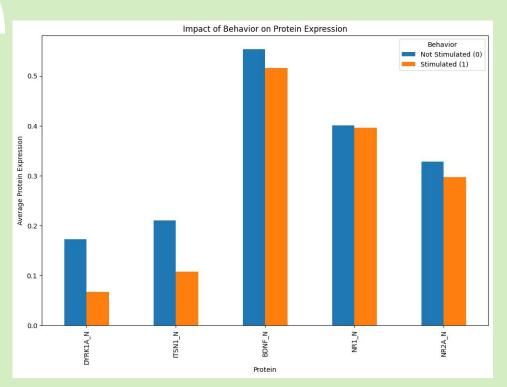
Interpretation and Analysis



Trisomic mice show higher levels of ITSN1 N and DYRK1A N, while control mice have higher BDNF_N, NR1_N, and NR2A_N. These protein differences may be linked to the cognitive impairments in Down syndrome.



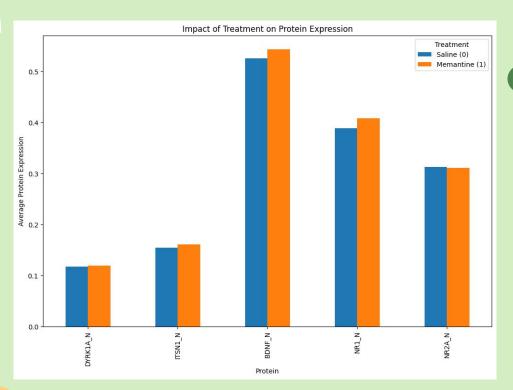
Interpretation and Analysis



Unstimulated mice show higher levels of DYRK1A_N, ITSN1_N, BDNF_N, and NR2A_N compared to stimulated mice. This suggests that stimulation affects these proteins, potentially influencing associative learning mechanisms.



Interpretation and Analysis



Memantine treatment does not drastically alter most protein expression levels compared to saline. However, slight increases in BDNF_N and NR1_N in the memantine group suggest subtle effects on these proteins, possibly related to its therapeutic effects.

Challenges we faced....



High-dimensional data risks overfitting; regularization and cross-validation are essential.

Missing Data

Technical issues may lead to missing protein values, requiring effective handling techniques.

Biological Variability

Natural variations in protein expression among mice can complicate the identification of subtle class differences.

Conclusion

The study identified key proteins, such as DYRK1A_N, ITSN1_N, and BDNF_N, that differentiate between control and trisomic mice. However, the findings are limited by sample size and biological complexity, suggesting the need for further validation with larger datasets and additional experiments. Future research should focus on confirming these results in broader cohorts, exploring underlying biological mechanisms, and integrating other omics data for a more comprehensive understanding of Down syndrome.

