Genetic Biomarkers and Biological Pathways for Autism Spectrum Disorders

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Autism Spectrum Disorders (ASD)

- 1 in every 100 people are affected by ASD (Zeidan et al., 2022).
- Symptoms Include (Matson et al., 2008; Nadeem et al., 2021):
 - Challenges with social interactions
 - Restricted interests
 - Repetitive behavior
 - Gastrointestinal dysfunction
 - Weak immunity
- Early diagnosis can improve later outcomes through reductions in symptom severity (Charman and Baird, 2002).

Background

• Current Diagnostic Methods:

Expert Clinical Observation (Charman and Baird, 2002)

- Challenges in Diagnosis (McCarty and Frye, 2020):
 - Symptoms are considered normal at young age.
 - Limited peer interaction may obscure early symptoms.
 - Overlap with other neuropsychiatric disorders.
 - There is a 33% accuracy when reporting a positive case.

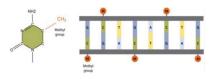
Background

- Compared to similar disorders, ASD has the strongest genetic component (Klin, 2018).
- Several genes have been linked to ASD (PTEN, CHD8, SCN2A, and ADNP)

(Vogt et al., 2015; Yu et al., 2022; Sanders et al., 2018; Arnett et al., 2018).

However, these have not been applied to diagnosis or treatment.

Background



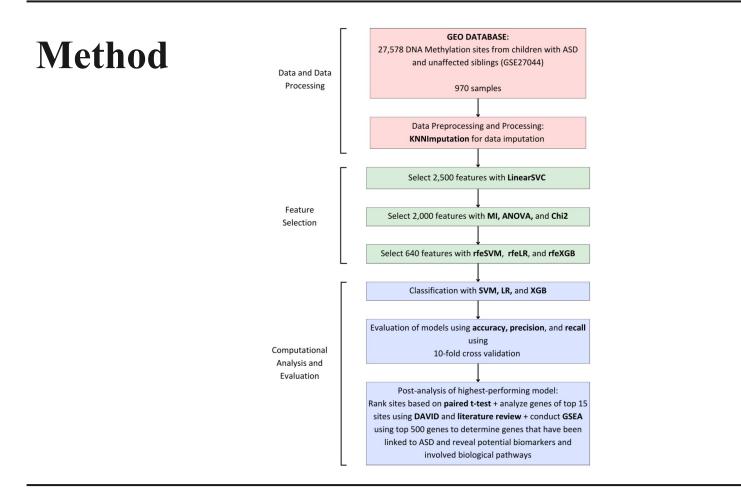
• DNA methylation involves adding a methyl group to a cytosine nucleotide at

the C5 position to form 5-methylcytosine.

- Regulate gene expression without affecting nucleotide sequences.
- Prevent the binding of transcription factors, reducing the expression of a gene.
- Recruit proteins involved in gene expression (Moore et al., 2013).

Research Aim

This research aims to use DNA methylation levels at different sites in children with and without ASD to reveal genetic biomarkers and biological pathways involved in ASD. These can be used for the diagnosis, prediction, and treatment of ASD, leading to improved outcomes for affected individuals.



Method - Dataset

The GSE27044 dataset (Barwick BG et al. 2015) was used for this analysis.

- 27,578 methylation sites from peripheral blood leukocytes
- 488 pediatric-age males with ASD, 482 of their unaffected brothers, and 158 of

indeterminable class. This analysis considered only the 970 samples whose class

is known.

• Different ethnicities but a white majority.

Method - Preprocessing and Feature Selection

 397 missing values (0.00148%) - k-Nearest Neighbor Imputer (kNNImputer) from Scikit-learn was used (Troyanskaya et al., 2001).

- Feature Selection Methods:
 - 2,500 features: LinearSVC (Feng et al., 2019)
 - 2,000 features: ANOVA (Sthle and Wold 1989), Mutual information (Kraskov et al., 2004), Chi2 (H. Zhang et al., 2014).
 - 710 features: **Recursive Feature Elimination** (Guyon et al. 2002).

Method - Classification and Evaluation

- Classification Methods:
 - Support Vector Machine (Cortes and Vapnik, 1995).
 - Logistic Regression (Montgomery et al., 2021).
 - XGBoost (Chen and Guestrin, 2016).
- Evaluation:
 - 10-fold cross validation (Kohavi, 1995) + accuracy, precision, and recall

Results

Classifier	Feature Selection Method	Accuracy	Precision	Recall
SVM	ANOVA	99.79%	99.80%	99.80%
	MI	99.18%	99.39%	98.98%
	Chi2	99.18%	99.17%	99.16%
LR	ANOVA	90.93%	92.17%	92.21%
	MI	92.37%	92.78%	92.21%
	Chi2	92.47%	93.23%	91.80%
XGB	ANOVA	66.70%	68.00%	65.35%
	MI	65.26%	64.99%	66.77%
	Chi2	67.11%	67.38%	67.48%

Existing Work	Accuracy of Existing Work	Accuracy Achieved
Gopinathan & Geetha (2023)	99.00% (500 CpG sites)	
Feng et al. (2019)	99.70% (678 CpG sites)	99.79% (640 CpG sites)

Comparison with Existing Work

Summary of Performances

Method - Analysis

 Paired t-test to rank features (Hsu & Lachenbruch, 2014; Kim, 2015)- method of comparing paired data to determine the significance of features (to reject the null hypothesis). Using a paired t-test would account for the differences between sibling pairs. An adjusted p-value (Benjamini and Hochberg, 1995) was used to determine significance.

• Gene Set Enrichment Analysis - compares selected genes to existing gene sets to find overlaps (Subramanian et al., 2005)

Results - Some Key Genes

Site	Location	Gene	Function
cg12010995	Intronic	CYP7A1	CYP7A1 codes for an enzyme involved in bile acid synthesis and has been linked to impaired liver regeneration (L. Zhang et al., 2009). While gastrointestinal abnormalities have been associated with ASD (Nadeem et al., 2021), this gene has no clear link in existing literature.
cg23477967	Exonic	TFPI	TFPI plays a role in blood coagulation. However, apart from its primary function, it has been linked to roles in innate immunity, microbial defense, inflammation (Massberg et al., 2010). It has also been linked to venous thrombosis (Schmidt et al., 1999).
cg21660392	Intronic	ABCA8	While the exact function is unknown, the gene is highly expressed in the heart, liver, and muscle. It may play a role in transporting substances across the blood-brain barrier (Albrecht and Viturro 2007).

Results - Some Key Genes

Site	Location	Gene	Function
cg11161873	Exonic	LSMEM1	The exact function of the protein is unknown. However, it has been linked to Parkinson's disease and Alzheimer's disease, suggesting neurological function (Whittle et al. 2024; Bergen et al. 2015).
cg04223956	Intronic	NEK7	NEK7 has been linked to gastric cancer progression (Li et al., 2021) and is involved in protein phosphorylation and regulation of mitotic cell cycle.
cg13059335	Intronic	ADAMDEC1	It is involved in proteolysis, immune response, and negative regulation of cell adhesion. It has been found that it is expressed exclusively in the gastrointestinal tract and has been linked to inflammatory diseases and cancers (Kumagai et al., 2020).

Results - Key Pathways

Gene Set	No. of Overlaps	% of Selected Genes	p-value (†)
Genes differentially expressed in ovarian cancer patients	62	9.69%	4.79 e ⁻¹⁵
Abnormal immune system physiology	60	9.38%	9.17 e ⁻¹⁴
Increased inflammatory response	42	6.56%	6.87 e ⁻¹⁰
Abnormal respiratory system physiology	46	7.19%	1.81 e ⁻⁹
Abnormal reproductive system morphology	40	6.25%	2.95 e ⁻⁹
Abnormal intestine morphology	33	5.16%	3.21 e ⁻⁹
Abnormal digestive system morphology	38	5.94%	4.00 e ⁻⁹
Abnormal lung morphology	40	6.25%	5.17 e ⁻⁹

Conclusions

- The genes CYP7A1, NEK7, TFPI, LSMEM1, ABCA8, and ADAMDEC1 could play a key role in ASD, despite not having been seen previously. These genes correspond to the CpG sites cg12010995, cg04223956, cg23477967, cg11161873, cg21660392, and cg13059335.
- Also of note, genes related to the reproductive and respiratory systems were found to have significant overlap with the selected genes, despite minimal existing research on this link. Additionally, the immune and gastrointestinal systems were revealed to be significantly involved in ASD.

Implications

- Genes hold promise as biomarkers for the diagnosis and prediction of ASD.
- Reveals previously unrecognized links, offering insights into potential symptomatology and therapeutic avenues for ASD and related disorders.
- Reinforces existing evidence of associations between ASD and the gastrointestinal and immune systems, elucidating the physiological effects. However, there is still minimal existing investigation about all these pathways in ASD. Understanding these complex interactions could significantly advance diagnostic approaches.
- Machine learning approach could be made more suited for diverse populations.
- Future research should investigate the roles of KERA, CSN3, C21orf94, ZNF619, MTTP, and PKIA in ASD.

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